

AN ASSESSMENT OF SAMPLING DETECTABILITY
FOR GLOBAL BIODIVERSITY MONITORING:
RESULTS FROM SAMPLING GRIDS IN
DIFFERENT CLIMATIC REGIONS

Master thesis

by

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Ergebnisse von Sampling GRIDs aus
unterschiedlichen klimatischen Regionen

Abstract

This thesis provides important input for the development of a cost-effective global biodiversity assessment and monitoring system. The study is embedded in a larger project to evaluate possibilities of multiple-species surveys using biodiversity GRIDs. As a pilot study six GRIDs in diverse ecosystem settings are sampled. Sampling methods used for animal species are point transects for birds and trapping webs for arthropods; additionally a line transects add-on protocol is used at some study areas for amphibians, reptiles and butterflies. Within this framework the task is taken over to develop predictive models for sampled animal species with Random Forests. Additionally the data is analyzed to derive abundance estimates with multiple covariate DISTANCE sampling and occupancy estimates through the software PRESENCE.

A total of 5,007 observations from six study areas from all over the world are analyzed in detail. Total sampling time is about 12 weeks. High quality non-random predictive models with a ROC value > 0.5 are gained with Random Forests analysis for 116 described animal narratives. Half of these observations origin from point transect sampling, the other half from trapping web catches. The line transects add-on protocol results in another 3 predictive models. Abundance and occupancy estimates are derived from the data for 46 animal narratives, 23 of those for point transect data, 22 for trapping web data, and 1 for line transect data. Predictive modeling with Random Forests proves to be a very powerful tool. DISTANCE sampling estimates from this study show large confidence interval ranges, but are extremely cost-efficient to gather initial information for multiple species rapidly. PRESENCE estimates are partly unsatisfying because of a large portion of animal narratives with perfect occupancy estimates ($\Psi = 1.0$). It is assumed that this is an effect of small sampling size which will not be problematic for larger amounts of data. This has to be kept in mind when comparing DISTANCE and PRESENCE results. Correlation between DISTANCE and PRESENCE detection probability estimates is negative, while correlation between DISTANCE abundance estimates and PRESENCE occupancy estimates is positive for all but one study area. It is recommended to repeat the comparison when data from more plots is available. On one hand the results, the cost-effectiveness of the study, and possibilities opened by this kind of multiple-species multi-method sampling are promising, on the other hand funding for this visionary approach was not available.

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List of Abbreviations

| | |
|--------------|--|
| 1CR | 1st GRID in Costa Rica |
| 2Ni | 2nd GRID in Nicaragua |
| 3AK | 3rd GRID in Fairbanks, Alaska |
| 4Ru | 4th GRID in Russia |
| 5PG | 5th GRID in Papua New-Guinea |
| 6Ba | 6th GRID in Barrow, Alaska |
| ABMP | Alberta Biodiversity Monitoring Program |
| AIC | Akaike Information Criterion |
| all | all available data (pooled) |
| aur | only data aurally detected |
| Bi | Bird |
| CBD | Convention on Biological Diversity |
| CDS | Conventional DISTANCE Sampling |
| Covariates | one of three Random Forests model definitions |
| DIWPA | Diversitas in Western Pacific and Asia |
| D | Density |
| DS | DISTANCE Sampling |
| DT | DISTANCE Sampling Line Transect |
| GBIF | Global Biodiversity Information Facility |
| GEO | Group on Earth Observations |
| GEOSS | Global Earth Observation System of Systems |
| GIS | Geographic Information Systems |
| GLM | Generalized Linear Models |
| IBOY | International Biodiversity Observation Year |
| Interspecies | one of three Random Forests model definitions |
| IPY | International Polar Year |
| ITIS | Integrated Taxonomic Information System |
| MCDS | Multiple Covariate DISTANCE Sampling |
| MSIM | Multiple Species Inventory and Monitoring Protocol |
| p | Probability of Detection |
| Plot | one of three Random Forests model definitions |
| PR | PRESENCE |
| Psi | Occupancy estimate |
| ran | only data from randomly selected Plots |
| RF | Random Forests |
| ROC | Receiver Operating Characteristic |
| sys | only data from systematically selected Plots |
| TW | Trapping Web |
| vis | only data visually detected |

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1 Introduction

1.1 *Global Biodiversity Crisis and Biodiversity Monitoring*

Biodiversity loss is widely recognized as a crucial survival issue in society, at the latest since most countries of the international community signed the Convention on Biological Diversity at the Earth Summit in Rio de Janeiro in 1992 (Brooks et al. 2002; CBD 2006; McKee et al. 2004). There is vast evidence that the loss of biodiversity is not only an ethical problem, but also substantially financial because important ecosystem services are lost on a global level (Mainka et al. 2005; Millennium Ecosystem Assessment, ongoing). Many countries have recognized these facts and have implemented national biodiversity monitoring strategies to detect changes in biodiversity, usually substituted by monitoring of species richness (Nakashizuka & Stork 2002; Wilson 1992). More and more of these protocols accept a loss of precision for single species by assessing multiple species at the same time, because resources to implement one monitoring system per species are simply not available and multiple-species monitoring on a landscape level is much more resource-efficient (Franklin 1993; Manley et al. 2005; Manley et al. 2004). These systems are also more resilient against sudden changes in the focus of research interest, which may render more specific monitoring systems useless before they are fully implemented (Watson & Novelly 2004). Some regional examples of such monitoring systems are the Multiple Species Inventory and Monitoring Protocol (MSIM) for National Forest System Lands in the United States (Manley & van Horne 2006); the Alberta Biodiversity Monitoring Program (ABMP) in Canada (ABMP 2006; Stadt et al. 2006); Biodiversity Monitoring Switzerland (Küttel 2007); or International Biodiversity Observation Year in Western Pacific and Asia (IBOY-DIWPA, Nakashizuka & Stork 2002).

The use of these systems presents a major step forward to make biodiversity research more relevant, rigorous, compelling and thus more tangible and usable for political planning and implementation processes (Marzluff et al. 2001). But they can have two problems: firstly they are highly specialized for the area within the borders of the country they were developed for. Very often these protocols can only be used in specific environments, for example temperate forests and mountainous areas, but are usually not applicable to ecosystems which do not occur in the country of origin. Political borders are (usually) clear and precise, while changes of biodiversity respectively of nature in general are subtle and gradient. Most of today's threats to biodiversity, for example global climate change, have influences which do not stop

at political and administrative borders, neither do migrating animals nor ecological processes. Secondly, even if the ecosystems in different countries are similar enough, very often the details of data collection and/or processing methods differ too much to compare monitoring results from different countries. As a result they are not allowing for proper generalizations. Green et al. (2005) “*argue that there is a shortage of standardized, regularly repeated measurements of the state of biomes and their biota that could be used to monitor progress toward this goal*”. In the long-term view the intention behind the project is to develop a globally valid biodiversity monitoring system which delivers comparable results at achievable costs in every ecosystem and every country of the world. “*Global conservation assessments require information on the distribution of biodiversity across the planet*” (Ferrier et al. 2004). Achieving this global perspective is obviously a very ambitious goal and might not be completely attained in the very near future. To this date most habitats have not even been assessed once and there is a considerable shortage of biodiversity monitoring on a global scale (Dobson 2005; Green et al. 2005).

Another intention for this project is to work with low-cost methods. The budget available for biodiversity monitoring on a global scale is unfortunately very low. As the method to develop is supposed to be used in many areas of the world which can not or are not willing to afford to invest large amounts of money and resources into the implementation of such a monitoring method, the intention is to work on a “shoestring budget”. “*I’ve become convinced that design for I&M programs must be predicated on the idea that funds are ephemeral and so the core of a monitoring program should be very lean (and relatively inexpensive). Around that core, you can develop add-on protocols and additional sampling that are only implemented when funds are available*” (Morton 2007, pers. comm.). Sampling is therefore primarily conducted for taxonomic groups that are potentially living in almost every terrestrial ecosystem: birds and ground-living insects. One possible add-on protocol for butterflies, amphibians and reptiles is developed for this study; further add-on protocols for other animal groups can be developed at a later stage.

In short, the project idea is to develop a relatively simple low-cost rapid biodiversity assessment and monitoring system, which aims at multiple species and offers multiple ways of analysis. Furthermore this system is supposed to be globally applicable and compatible with current data standards, so that data from this project may contribute to ongoing global biodiversity initiatives (Global Earth Observation System of Systems (GEOSS) 2008; Group

on Earth Observations (GEO) 2008; International Polar Year (IPY) 2008). As a pilot study data was collected with different methods from six diverse regions in the world in form of a biodiversity GRID (as explained in chapter 2.2.1). The study at hand is a partial assessment of some of the most important possibilities to analyze these GRIDs offer for the estimation of animal populations. The results will provide a valuable starting point for more detailed taxonomic studies and provide crucially needed information for setting up sampling schemes with higher accuracy. For that reason data are made fully available to the public and investigators for their own assessment. Full Metadata for the datasets will be uploaded to NBII Clearinghouse website and found online at <http://mercdev3.ornl.gov/nbii/> . In the long-term such data is expected to be easily visualized and connectable to other data sets in public domains (Guralnick et al. 2007).

1.2 Goals of the Study

This thesis supports the overall biodiversity GRID project by analyzing wildlife data from the project at three different analysis levels: prediction, abundance and occupancy. A short overview is given in this introduction; detailed information is available from the methods section.

The first analysis goal, prediction respectively predictive modeling, was in the past in practice limited to Generalized Linear Models (GLMs) by lack of computing power, in spite of ecological data often being non-linear, interactive and multi-dimensional in nature. Recent developments in computer technology and steep price declines of equipment and communication are relaxing these limitations (Bauldock et al. 2001). First studies using machine learning algorithms in ecology are promising and seem to clearly outweigh the traditional GLMs in convenience, speed and accuracy (e.g. Huettmann 1999; Magness et al. 2008; Prasad et al. 2006). Predictive modeling is a tool to achieve global information about biodiversity distribution conveniently (Elith et al. 2006). It has the ability to process all available environmental data to analyze the effect on general biodiversity patterns (Faith 2005). It has also been shown in numerous cases that well-constructed models often show a much better performance and higher consistency in population estimations and habitat modeling than do expert opinions (Pearce et al. 2001; Yamada et al. 2003). Predictive

modeling using data mining can also handle a large variety of data since there are no requirements of parametric assumptions to be met. Additionally these machine learning algorithms have less problems interpreting noisy or sparse information (Elith et al. 2006), partly because interactions between variables are included in analysis (Craig & Huettmann 2008; Magness et al. 2008). The predictions have the advantage that they can be tested for generalizations.

Abundance and population density are probably the most important basic parameters in population dynamics (Krebs 2001). This makes abundance a very valid second analysis goal. Whenever possible it is intended to get true abundance estimates for each species, corrected for imperfect detection of individuals with different methods (Buckland et al. 2001; MacKenzie 2005a). However, with a standardized multiple-species protocol this is not always possible. Especially species with large territories are often difficult to monitor on an eco-regional scale (Manley & van Horne 2006). Therefore as a third point the probability that an area is occupied by a species, known as occupancy or Ψ , is also estimated. Occupancy is the simplest level of interest (Hill et al. 2006), which gives much less information than abundance or density, but still has implications for wildlife management. At the same level of precision it can usually be obtained at lower costs than abundance estimates (Bailey et al. 2007; MacKenzie 2005b). Clearly, occupancy is not the prime goal in this project, but it is accepted as better-than-nothing baseline information. It also allows matching up with studies underway elsewhere worldwide. Results of abundance and occupancy estimates will also be directly compared to each other.

2 Methods

2.1 Study Area

Data from six different study areas from all over the globe was used for this thesis. The data was collected in:

- (1) Costa Rica, lowland tropical rainforest, data collection from 10th to 22nd June 2007 (data collected by Falk Huettmann and Dirk Nemitz)
- (2) Nicaragua, tropical dryforest, data collection from 22nd June to 5th July 2007 (data collected by Falk Huettmann, Dirk Nemitz and Andre Breton)
- (3) Central Alaska (USA), boreal forest, data collection from 14th July to 3rd August 2007 (data collected by Dirk Nemitz and Andre Breton)
- (4) Sakhalin Island (Russia), data collection from 6th to 24th August 2007 (data collected by Falk Huettmann)
- (5) Papua New-Guinea, data collection from 22nd to 28th December 2007 (data collected by Falk Huettmann)
- (6) Northern Alaska (USA), arctic tundra, data collection from 29th June to 4th July 2008 (data collected by Falk Huettmann)



Figure 1: Location of study sites (Google Maps, adjusted)

2.1.1 Study Area 1CR: La Suerte Station, Costa Rica



Figure 2: Location of study area 1CR in Costa Rica (Google Maps, adjusted)

The studied lowland tropical rainforest is located at La Suerte Biological Station at the Río Suerte in north-eastern Costa Rica (Janzen 1983). The station is a teaching and research facility with ca. 20 ha advanced secondary tropical rainforest. It is located about 50 m above sea level, and one of the sites carrying highest biodiversity in the world. According to the owners of the Biological Station it is “*home to thousands of plant and insect species as well as hundreds of species of amphibians, reptiles, birds and mammals*” (Molina 2007). The area receives about 3800 mm annual rainfall on average. It is well known for studies on neotropical primates, especially mantled howling monkeys (*Alouatta palliata*), black-handed spider monkeys (*Ateles geoffroyi*) and white-faced capuchins (*Cebus capucinus*) (Garber & Rehg 1999). According to hand-held GPS measurements the study GRID extends from about 10.26573 to 10.26805 north and from 83.46704 to 83.46919 west.

2.1.2 Study Area 2Ni: Ometepe Island, Nicaragua



Figure 3: Location of study area 2Ni in Nicaragua (Google Maps, adjusted)

The second area where data was collected is a tropical dryforest close to Point San Ramon Village on Ometepe Island in Lake Nicaragua. Ometepe Island encompasses about 276 km² and is the biggest volcanic island in the world that is located in a freshwater lake. The island is dominated by the two volcanoes Concepción and Maderas. Almost the entire flat land is used agriculturally, while secondary tropical dry forest grows on the slopes of the volcanoes. With higher altitude the forest gradually changes into undisturbed virgin tropical cloud forest. Volcanic rocks from former eruptions are scattered all over the island. The island receives about 1,600 mm average annual rainfall and has a medium daily temperature between 26° and 29° Celsius (Steck 1997). The area is especially known for studies on the easily observable mantled howling monkeys (*Alouatta palliata*) population (Garber et al. 1999; Huettmann 1999; Popp et al. 2007). According to hand-held GPS measurements the study GRID extends from about 11.25120 to 11.25388 north and from 85.31858 to 85.32143 west.

2.1.3 Study Area 3AK: Fairbanks, Alaska



Figure 4: Location of study area 3AK in Alaska, USA (Google Maps, adjusted)

The third data collection area is a boreal forest located on the campus of the University of Alaska, Fairbanks, USA. Fairbanks is located in the centre of interior Alaska. The climate is rather continental because of the surrounding mountain ranges, resulting in cold winters as well as warm and dry summers. The temperature ranges from -50° Celsius in January to over 30° Celsius in July. The average annual precipitation is 287 mm (Chapin et al. 2006). Frequent forest fires influence all boreal forests in this eco-region (Kasischke et al. 2006). According to hand-held GPS measurements the study GRID extends from about 64.520482 to 64.521789 north and from 147.512596 to 147.515652 west.

2.1.4 Study Area 4Ru: Verengery Sakhalin Island, Russia



Figure 5: Location of study area 4Ru in Russia (Google Maps, adjusted)

The fourth study area is located in the Russian Far East, on Sakhalin Island in the North Pacific. With about 78,000 km² Sakhalin Island is the largest Russian island. Its climate is rather cold and usually considered to be sub-arctic. Much of the Sea of Okhotsk between the island and the mainland is usually covered by ice during the long winters. The area is regarded as extremely important for conservation of arctic and sub-arctic migratory shorebirds. Despite holding large amounts of oil and gas resources, most of the island's hinterland is relatively undisturbed (Huettmann & Gerasimov 2006). According to hand-held GPS measurements the study GRID extends from about 50.59892 to 50.60234 north and from 143.69052 to 143.69526 east.

2.1.5 Study Area 5PG: Bismarck Range, Papua New-Guinea

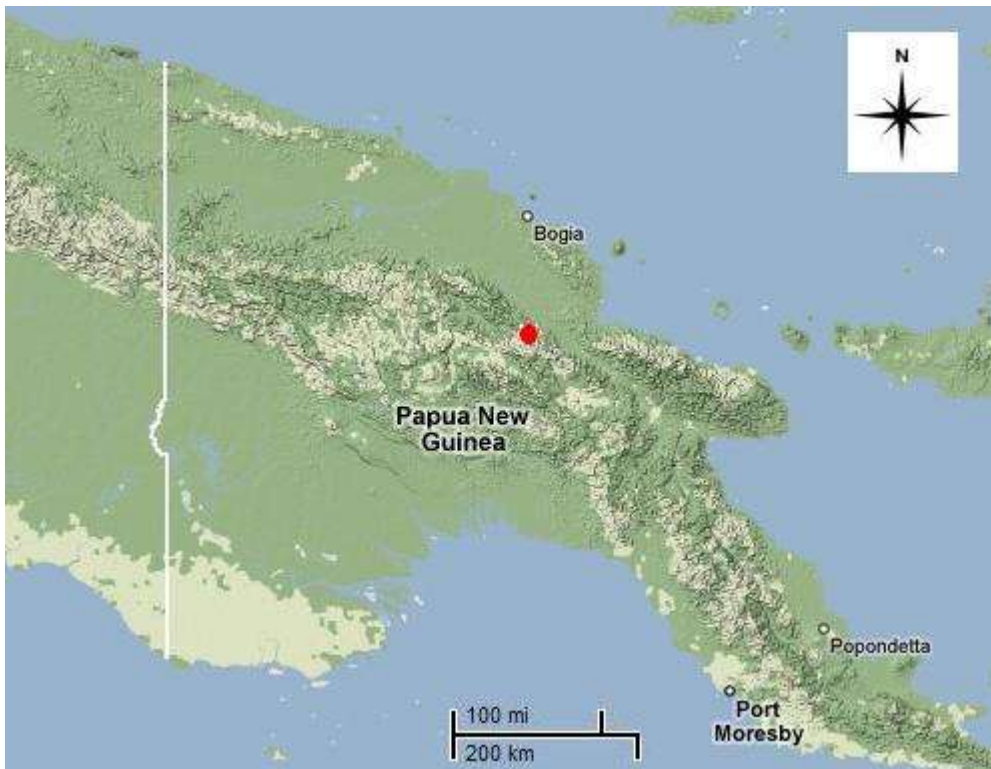


Figure 6: Location of study area 5PG in Papua New-Guinea (Google Maps, adjusted)

The location of the fourth study area is in the Bismarck range in Papua New Guinea. Papua New Guinea is the world's third largest insular state. It is especially known for its enormous scenic, cultural and biological diversity. Papua New Guinea has a high variation in rainfall, altitude, soil, and history of disturbances, resulting in high biodiversity (Miller et al. 1994b). It is estimated that about 5 % of the world's total biodiversity is located in the country, while exact information about species details and taxonomy is very sparse (Miller et al. 1994a). The area covered by the study GRID is located in the Bismarck range at an altitude of ca 850 m, typically being classified as Lowland Humid Forest with an average annual rainfall between 2,500 mm and 3,500 mm (Miller et al. 1994b). The GRID covers prime forest, an adjacent garden and a forest trail.

2.1.6 Study Area 6Ba: Barrow, Alaska



Figure 7: Location of study area 6Ba in Alaska, USA (Google Maps, adjusted)

The sixth study area is Barrow, located in the North of Alaska. It is the northernmost settlement of the United States. The climate is polar, very cold, with less than four months exceeding a mean temperature of 0° Celsius. Because of its dryness the area is classified as desert. Barrow and the surrounding area are extremely important bird habitat (Pitelka 1974). This study area differs from the other five in two main ways. Firstly it contains only one habitat type: arctic tundra without major vegetation and trees. Secondly there was midnight sun during the time the sampling took place, so it was not determinable if the time of sampling relative to the time of sunrise has an effect on the results. According to hand-held GPS measurements the study GRID extends from about 71.24034 to 71.24467 north and from 156.56546 to 156.57717 west.

2.2 Sampling Methods

2.2.1 Biodiversity GRID

For efficiency reasons a systematic sampling approach was chosen (Cochran 1946; Olea 1984). First of all an equally spaced GRID was implemented: 25 points were arranged in five rows and five columns in order to cover a consistent area but also to have a known spatial neighbor relationship among all plots, which is consistent with recommendations given by Ricklefs (2004). The distance between plots was 100 m, resulting in a total GRID size of 500 m x 500 m. While the final GRID system ideally covers the globe systematically without intentional placement, for these initial studies the GRIDs were placed in a way that roughly half to two thirds of the plots fell inside a forested area, the remaining plots at the forest edge or inside the cultural landscape. This survey setup enables other studies on the same data set to make realistic and representative statements about fragmentation effects. The only exception is GRID 6Ba in northern Alaska, where naturally only one habitat type, arctic tundra, occurs. Additionally, five points were randomly placed within the GRID to be able to model the influence of random patterns on the results and their spatial relations (Figure 8).

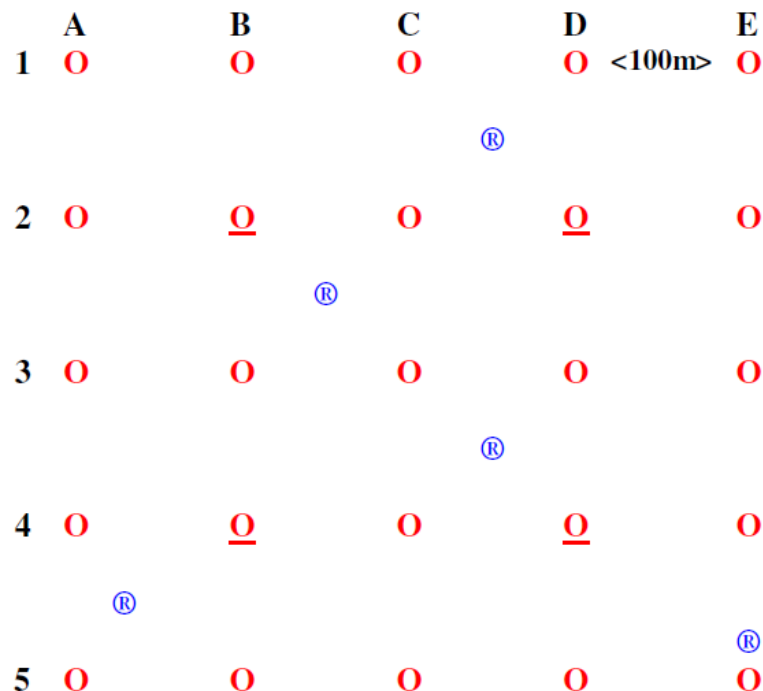


Figure 8: Structure of the biodiversity GRID with 25 systematically selected plots, 5 randomly selected plots, and trapping webs installed at 4 plots (underlined)

The coordinates of each plot were obtained from a regular hand-held GPS receiver and revisited by using the “Go to” function. All plots as well as the path between them were marked with decomposing flagging tape to make recognition in the field easier. A simple schematic map was drawn by hand for each field work participant to ensure that plots are found when the GPS does not receive signals, as was often the case in dense forest settings.

2.2.2 Budget Constraints

The biodiversity GRID is meant as a method for cost-efficient rapid biodiversity assessment that allows for an analysis of spatial relations as well. All methods involved have to work in relatively short time, with low costs and little demand of technological equipment. There is no objection to include more sophisticated methods in add-on protocols, but they are discouraged for the main protocol to keep the inhibition threshold for decision makers low.

Trained taxonomists were not available, as they rarely are for many ecosystems. All notes regarding the observed species were made as precisely as possible, although most of the observers were not trained especially in tropical ornithology or entomology. Data collection followed the motto the more detail the better, but it was not intended to refuse data because of lacking taxonomic details. If the observer did not readily know the correct scientific name of a specimen, a common name or, in lack of knowledge of a common name, a short description was noted. This original field note is referred to as the “narrative name” of an observation respectively of a species. Such process is common when dealing with large numbers of species and in largely unexplored environments, where huge fractions of the biodiversity remains still unknown, or where appropriate taxonomic guide books are missing.

This resulted in good abundance and occupancy estimates, but in less detailed taxonomic data. Such is the characteristic in rapid biodiversity assessments on shoestring budgets, which allow for a first impression and provide detailed information for deeper investigation if desired. This type of rapid assessment additionally serves as a pilot study for further assessments. In the present study the focus lies on spatial global coverage, instead of local detail.

2.2.3 Animal Species Data Collection

In the ideal case, the protocol should result not only in information about the presence or absence of species, but also in an estimate of population size. The DISTANCE sampling approach uses the concept of a detection function based on distance of the observed object from the observer to estimate population density (Buckland et al. 1993; Buckland et al. 2001). It plays a central role in this study and is used in a number of ways.

At each of the 30 plots (25 systematic and 5 random), five minute point transect DISTANCE sampling counts for birds were conducted within 360 degrees (Buckland et al. 2008). A short settle-in period of one minute was granted prior to counting to allow for the snapshot character of DISTANCE sampling, especially meeting the assumption that presence of the observer does not introduce bias by causing responsive movements of animals. Following common practice the point counts took place only in the morning between 5:30 and 10 am. Birds are known to show higher activity at this time, which generally increases detectability and maximizes inventory accuracy. Each bird seen or heard was noted, including an estimate of the radial distance from the observer. Double counts were avoided by the observer's attention and the relatively short counting period.

Observers decided to make two adjustments:

- in study area 4Ru seabird observations were excluded from plot A1;
- in study area 6Ba the survey time was reduced from five to four minutes.

The second method of DISTANCE sampling used was a trapping web (Parmenter & MacMahon 1989). 17 pitfall traps with a diameter of 9 cm each were arranged in a DISTANCE sampling trapping web design to estimate ground-living insects (as described in Buckland et al. 2001, p.216ff). This sampling method is very labor-intensive and could not be implemented at all 30 plots given the short time period available. Thus, four of the plots were systematically selected to capture the general patterns of species and abundances within the GRID: B2, D2, B4 and D4 (underlined in Figure 8) to gather at least some information about ground-living insects. Trapping webs were usually checked every 24 hours; and records were taken every 48 hours. In between check dates the cups were emptied without recording to avoid correlation in time between trapping events, and obtain spatially independent results.

Because of the low number of traps and more available work force it was decided to add a third circle of traps at 3 m from the centre in study areas 4Ru, 5PG and 6Ba. This increased the total number of pitfall traps in these areas to 25.

The third application of DISTANCE sampling was an add-on sampling protocol using DISTANCE sampling line transects, conducted at each of the 30 plots. Transects with a length of 10 m and traversing the plot at its centre were surveyed to estimate numbers of butterflies, amphibians and reptiles.

DISTANCE sampling point counts for birds and trapping webs for ground living insects were repeated three times. These repetitive visits further allow for an analysis with the software PRESENCE, which gives an estimate of general occurrence of a species in the area in a point-based sense. PRESENCE generates a detection function based on multiple visits under the assumption that the population is closed, meaning that no animals leave or enter the area of interest between several visits. Repetitions were not realized for the add-on protocol for DISTANCE sampling line transects.

2.2.4 Vegetation & Environment

Additionally, basic data about the plot environment was collected. If at all possible, the GPS coordinates were noted. A plot picture and a canopy picture were taken with a digital camera to give a general impression of the area and also allow for an analysis of light conditions in other studies on the same data set, e.g. remote sensing investigations (Figure 9). All pictures are available in the raw data file of the digital appendix on the accompanying DVD.



Figure 9: Sample pictures (habitat picture plot D3 in 1CR; canopy picture plot D3 in 3AK)

A short description of the ecosystem was noted as well (for example: pasture, forest interior, forest edge). Height and diameter at breast height were recorded for all trees within 5 m of plot centre. Estimates were noted regarding canopy cover percentage, understory cover percentage, shrub cover percentage (at 1.35 m height), bare soil percentage, duff coverage percentage, leaf browsing percentage, and number of flowers visible. The thickness of epiphytes, hemi-epiphytes, mosses and lichen was noted in categories (none, low, medium, high). Presence/absence of identified plant species or plant families was noted, as well as remarkable animal tracks (e.g. land crab holes, large mammal tracks, etc). Those are referred to as “Covariates 1 to32” in all six study areas, but the actual meaning is different in each. Detailed lists are attached for each study area (page 123). The full protocol is attached in the appendix (page 117 ff). The covariates can have one of four effects:

1. affecting habitat quality (presence/ absence of a species)
2. affecting detectability (detection/ non-detection of a species that is present)
3. affecting both of the above
4. affecting none of the above.

2.3 Analysis Methods

All observations were sorted by plot label and visit number; and marked with an individual observation ID. The data was then cleaned according to the following protocol:

- missing distances were replaced with ‘5000 m’ to be discarded due to data truncation as described by Thomas et al. (2006);
- fields for plots with no observations were emptied, lines with no observations were uniformly marked as ‘none’;
- narrative names assigned during observations were cleaned and summarized to avoid duplicates (“small ant” and “ant, small” are the same narrative, while “tiny ant” is a new one);
- type of identification during observation was standardized to aural/visual;
- habitat type was standardized to 3-5 classes in each region (cp. Table 1);
- sunrise time for each day and region was added, as well as the calculated amount of time between sunrise and observation;
- effective survey effort was calculated for trapping webs;
- comments were worked through and additional information was integrated into the data as far as possible.

Table 1: List of habitat types by study area

| 1CR | 2Ni | 3AK | 4Ru | 5PG | 6Ba |
|--------------|-----------------|-----------------|-----------------|-----------------|---------------|
| Forest edge | Forest edge | Forest edge | Forest trail | Forest edge | Arctic tundra |
| Forest gap | Forest trail | Forest trail | Interior forest | Forest trail | |
| Forest trail | Interior forest | Interior forest | Scrubs | Interior forest | |
| Pasture | Pasture | Pasture | | Pasture | |
| Wetland | Plantation | Wetland | | | |

One table each for observations, plot information and visit information was prepared and imported into MS Access database (Microsoft Office Access 2003 SP3). Additionally, an attempt was made to derive as much taxonomic information about the narrative names noted in the field as possible. All information was cross-checked and taxonomic validity was verified with online information provided by ITIS (Integrated Taxonomic Information System (ITIS), www.itis.gov) and imported into the same Access project database. The full database (DB_MINC.mdb) is available on the digital appendix DVD for this thesis, as well as all raw data sheets (in file “RawData”).

All files for data analysis with other software were derived through individual queries from this central project database. According to the method of plot selection the bird count and DISTANCE transect data was separated into a systematic set (25 plots) and a random set (5 plots). The random set was run as a test set for the systematic set, although small sample sizes were expected to increase standard errors. To increase the number of observations for analysis these two sets were also pooled and additionally analyzed together. Bird count data was further split into aural and visual detections as the form of detection is known to greatly influence detection probability (Marques et al. 2007). If an observation was detected both visually and aurally, then it was allocated to the visual data set. This resulted in a total of nine data sets for each of the six study areas:

- five bird count data sets (systematic, random, pooled, visual detection, aural detection);
- three DISTANCE transect data sets (systematic, random, pooled);
- one trapping web data set.

The exception is the GRID in 6Ba, where all bird detections were visual. In this case there are only three bird count data sets (systematic, random, and pooled).

Usually, each analysis was run using the narrative name given to the specific observation in the field. If further information was expected by summarizing different observation narratives and running the analysis on a higher taxonomic level, especially biological order and biological family, additional analysis targeted those levels. These have been found to be valid surrogates for rapid assessment and monitoring of species diversity (e.g. Negi & Gadgil 2002). Pooling decisions were made on a case-by-case basis (detailed lists are provided under 0 in the appendix).

2.3.1 Random Forests

Each data set was analyzed with Salford Systems Random Forests, version 1.0 (Breiman & Cutler 2005). Random Forests is a machine learning algorithm using sets of classification and regression trees for data mining and to build powerful predictive models (Breiman 2001). The

value of such analysis methods for numerous ecological applications is increasingly recognized.

Three different models were built: the first one used only plot data plus detection/non-detection of narratives at each plot to find patterns in the data and make predictions per plot by given spatial covariates. It is called '*Plot*', targets were all narrative names detected at least at five different plots. Because trapping webs were implemented at four plots only, this model was not applied to trapping web data.

Since there was a maximum of 30 plots at each region and 30 is a low number of samples for machine learning applications, two additional models were run on the complete number of observations, allowing spatial repeats. The first model used only environmental and vegetation covariates collected in the field; it is referred to as '*Covariates*'. The second one additionally used detection/non-detection of other animal species at the same plot as covariates to account for interactions between species; it is referred to as '*Interspecies*'. Table 2 gives an overview about differences between the used models. When aiming at the biological order or biological family level, all narrative names/species belonging to this particular order/family were excluded as covariates. Narratives observed through point or line transect sampling were taken into account when targeting trapping web narratives, because this data was equally available for all four trapping web plots. The only exception was study area 5PG, where the combined number of all point transect and trapping web observations exceeded the software limit for queries, so that only point count observations were used as additional covariates. On the other hand, narratives observed through trapping webs were not taken into account for *Interspecies* models for point and line transect narratives; this data was available for only four of the 30 plots. Targeted was the detection/non-detection of each narrative name with at least five observations. Random Forests settings all remained as 'default' (500 decision trees), only the number of predictors considered for each node was set to the square root of the number of used covariates (rounded up), as indicated in the accompanying software handbook. The best model was selected by highest ROC integral (Fawcett 2006).

All Random Forests import files and project files are available in the accompanying digital appendix DVD (under "ProjectFiles/RandomForests").

Table 2: Random Forests model overview

| Model name | Set of predictors | Response Variable | Spatial Repeats |
|----------------------------|--|--|------------------------|
| <i>Plot</i> | Environmental covariates | Presence/ absence of target (Narrative, Order, Family) | No |
| <i>Covariates</i> | Environmental covariates | Presence/ absence of target (Narrative, Order, Family) | Yes |
| <i>Interspecies</i> | Environmental covariates and presence of other species | Presence/ absence of target (Narrative, Order, Family) | Yes |

2.3.2 DISTANCE Sampling

A full DISTANCE sampling analysis was run for all narrative names with at least 20 observations using DISTANCE 5.0 Release 2 (Thomas et al. 2006). This is considerably lower than the 60-80 observations usually recommended, so inconsistencies resulting from small sample size have to be considered especially for those narratives with less than 60 observations. All of the following model key functions and model series expansion combinations were used (as in Buckland et al. 2001, p. 47):

1. Half-normal/ Cosine
2. Half-normal/ Hermite polynomial
3. Uniform/ Cosine
4. Uniform/ Simple polynomial
5. Hazard-rate/ Cosine
6. Hazard-rate/ Simple polynomial

Additionally, the two Half-normal and two Hazard-rate key function combinations were also analyzed with multiple covariate DISTANCE sampling (MCDS) in combination with each of the ten covariates identified as the most important by Random Forests for point and line transect data (Table 9 to

Table 11 in the results section). For trapping web data only the five most important covariates were used (Table 12 to Table 14 in the results section). All model definitions are listed in detail in the appendix under 7.3. Multiple covariate DISTANCE sampling is especially useful in situations where not enough detections are achieved to stratify the data by habitat and analyze each stratum separately (Aldredge et al. 2007; Marques et al. 2007). Among all models for a narrative name, the best ones for conventional DISTANCE sampling and multiple covariate DISTANCE sampling were selected by visual assessment of model fit and Akaike information criterion (AIC).

All DISTANCE sampling import files and project files are available for investigation in the accompanying digital appendix DVD (under “ProjectFiles/DISTANCE”).

2.3.3 PRESENCE / Occupancy

Occupancy estimations were derived by the software PRESENCE (Hines 2006) for the same narrative names as selected for DISTANCE analysis. For each narrative name one model was run assuming constant detection probability and second assuming different detection probabilities for each visit. Additional runs were conducted adding each of the site and visit specific covariates used in DISTANCE. Observation specific covariates were left out because of the different structure of analysis in PRESENCE, which takes only site and visit specific covariates into account. Categorical and continuous covariates were standardized in MS Excel according to common standards before importing into PRESENCE software (Donovan & Hines 2007). All models are listed in the appendix in chapter 7.4. The best model was selected using AIC.

All PRESENCE import files and project files are available for investigation in the accompanying digital appendix DVD (file “ProjectFiles/PRESENCE”).

3 Results

3.1 General Overview

Statistically there are two totally different estimations of biodiversity at each GRID: a systematic sampling design using 25 plots (respectively four for ground-living insects), and a random sampling design using five plots. They are lumped together for analysis to increase sample size for this initial method evaluation, despite losing the possibility to generalize the results to a larger region. Added together from all 180 plots at six study areas the three data collection routines resulted in a total of 5,007 animal observations (Table 3). 496 different narrative names are registered; Table 4 allocates the narrative names to the study areas in different regions. The add-on protocol for butterflies, amphibians and reptiles yields results in two out of the six study areas. Detailed species lists are given in the appendix (page 146 ff.). These also show the level to which taxonomic identification is possible. The number following some narrative names refers to the title of the picture taken of this particular narrative in the field. This is especially the case for trapping web narratives caught at 2Ni. All of these pictures are available from the digital appendix DVD (under RawData).

Table 3: Number of observations by region

| Study area | Number of bird observations | Number of trapping web observations | Number of line transect observations |
|--------------|-----------------------------|-------------------------------------|--------------------------------------|
| 1CR | 646 | 195 | 18 |
| 2Ni | 361 | 480 | 61 |
| 3AK | 692 | 237 | - |
| 4Ru | 509 | 231 | - |
| 5PG | 440 | 238 | - |
| 6Ba | 419 | 480 | - |
| Total | 3067 | 1861 | 79 |

Table 4: Number of narratives by region

| Study area | Number of bird narrative names | Number of trapping web narrative names | Number of line transect narrative names |
|--------------|--------------------------------|--|---|
| 1CR | 49 | 11 | 5 |
| 2Ni | 33 | 58 | 11 |
| 3AK | 17 | 20 | - |
| 4Ru | 45 | 34 | - |
| 5PG | 86 | 66 | - |
| 6Ba | 22 | 39 | - |
| Total | 252 | 228 | 16 |

Figure 10 shows a comparison of the number of observations in different regions, divided by point transect bird observations; trapping web catches and line transect observations. The GRID in 1CR for example yielded 21.1 % of bird observations, but only 10.5 % of ground insect observations, while having the same survey effort as the GRID in 2Ni with 11.8% of bird observations, but 25.8% of trapping web observations. These figures are raw count data and not corrected for detectability.

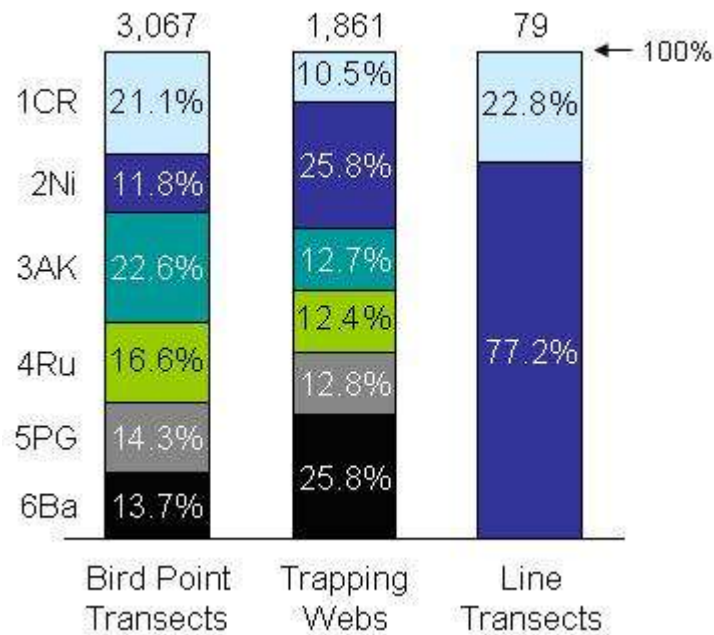


Figure 10: Percent of observations by region and type of survey

Figure 11 divides the number of narrative names into percent by region. This can be seen as a very simple estimate of species richness. For example, the GRID 3AK resulted in 22% of total bird observations (Figure 10), but yielded only 7% of bird species richness (Figure 11).

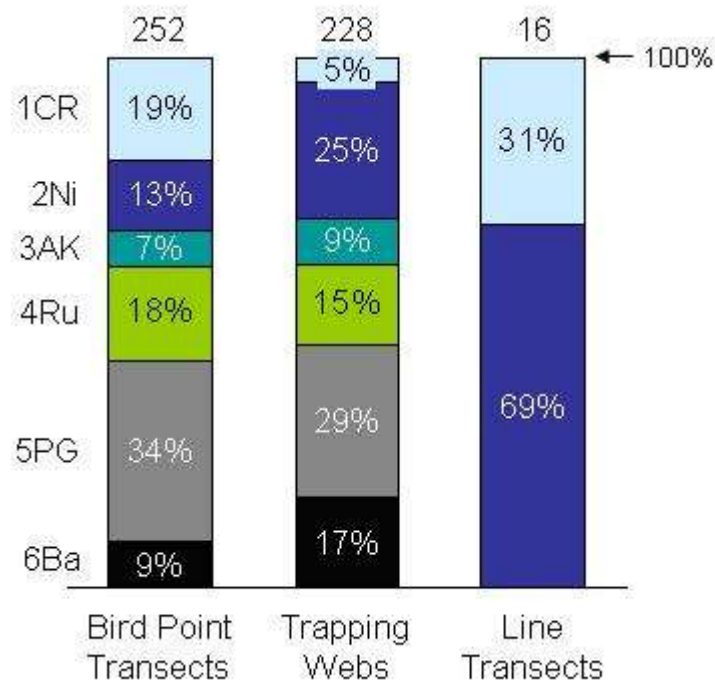


Figure 11: Percent of narrative names by region and type of survey

Despite all efforts to have equal survey effort in all 6 regions this goal was not reached and differences in survey effort have to be kept in mind when interpreting these figures. Table 5 gives an overview over survey effort by sampling method and region. Line transect survey effort was consistent in 1CR and 2Ni, while this sampling method was abandoned in the other regions.

Table 5: Survey effort by region and sampling method

| | Bird point transects | Trapping web: number of traps | Trapping web: area covered | Trapping web: total time |
|------------|----------------------|-------------------------------|----------------------------|--------------------------|
| 1CR | 3x 5 min/ plot | 17/ plot | 4x 19.63 m ² | 311 h |
| 2Ni | 3x 5 min/ plot | 17/ plot | 4x 19.63 m ² | 299 h |
| 3AK | 3x 5 min/ plot | 17/ plot | 4x 19.63 m ² | 296 h |
| 4Ru | 3x 5 min/ plot | 25/ plot | 4x 38.48 m ² | 216 h |
| 5PG | 3x 5 min/ plot | 25/ plot | 4x 38.48 m ² | 192 h |
| 6Ba | 3x 4 min/ plot | 25/ plot | 4x 38.48 m ² | 279 h |

There is no visible trend connecting greater survey effort with higher species richness. For example, the trapping webs in 1CR yielded only 5 % of observed species in spite of having the greatest total survey effort, while the trapping webs in 5PG with the lowest survey effort yielded 29 % of all species catches.

The data from systematic plots is expected to contribute about 80 % of observations, the data from random plots 20 %. Figure 12 shows the distribution of raw count data not corrected for detectability between random and systematic plots for all point transect detections, Figure 13 provides the same information for data collected through line transects. Trapping webs are not shown because they were installed at four systematic plots only. With the exception of bird point transect data at GRID 1CR, the proportion of observations from random plots is generally a little lower than expected. At this point this phenomenon can only be explained with a relatively small sample size because there were no obvious differences between random and systematic plots.



Figure 12: Distribution of point transect observations by plot type (random/systematic)

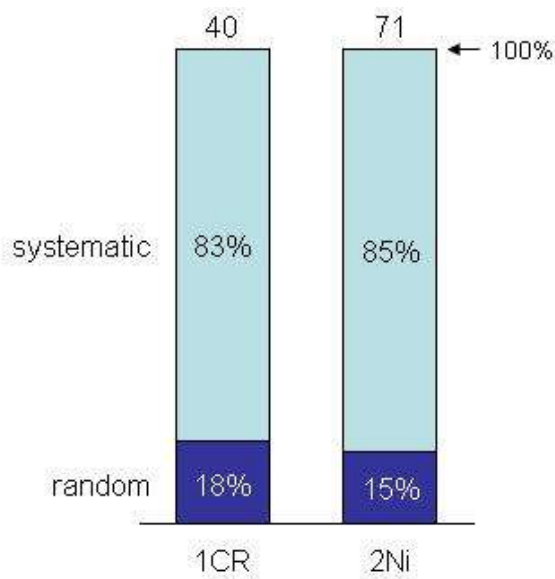


Figure 13: Distribution of line transect observations by plot type (random/systematic)

Figure 12 shows all observations made from all plots. For analysis using plot features the data has to be spatially tied to a plot, meaning observations further away than 50 m are discarded because they possibly are spatially closer to the neighboring plot and its features. Since information of the direction of observations from the observer was not collected, the observations at greater distances than 50 m could not be assigned to one of the neighboring plots. The 50 m border is not relevant for trapping web and line transect data because there were no observations at distances greater than 50 m. Figure 14 shows the distribution for bird point count data, for further analysis only observations within 50 m of the observer is used. The percentage of observed distances greater than 50 m at 3AK is obviously high, while the percentage at 6Ba seems to be low considering that there was no vegetation blocking view in any direction. There is no readily available explanation for these points.

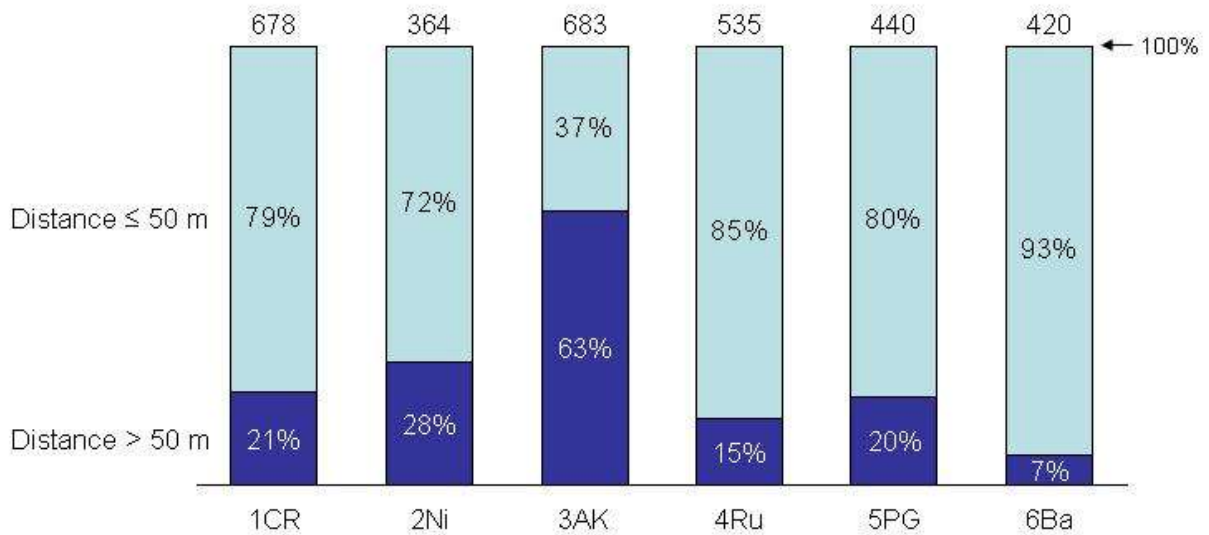


Figure 14: Percent of observations spatially belonging to plot (within 50 m radius)

The last split is between aural and visual detections for point transects (Figure 15). Observations for which information about the form of detection was lacking were disregarded for this analysis. All observations that were detected aurally as well as visually were noted as ‘visual’ and are used only in the assessment of visual detections.

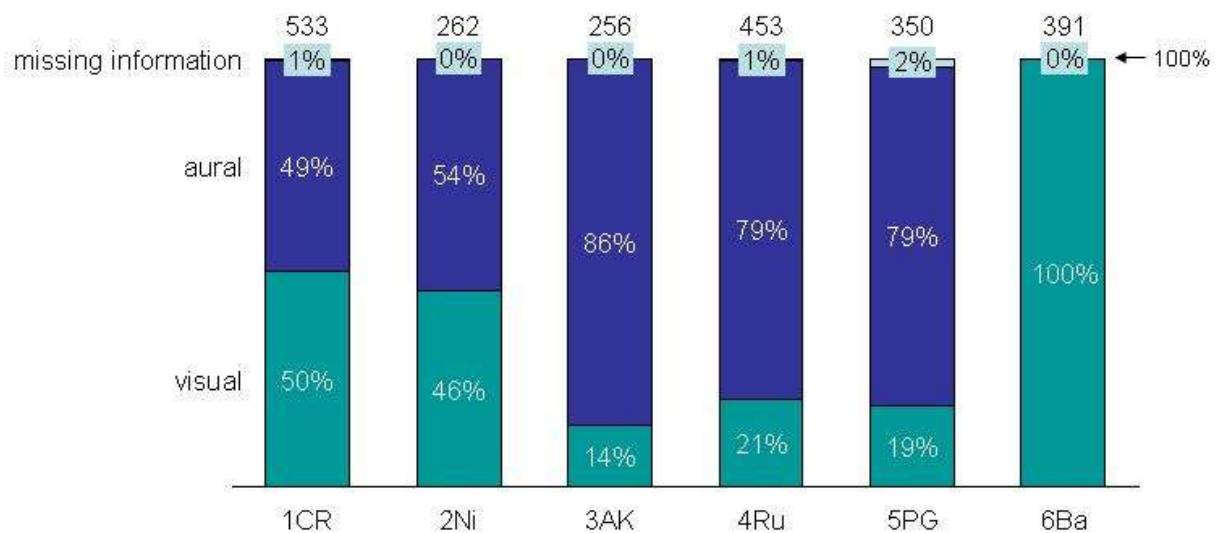


Figure 15: Proportion of aural and visual point transect observations

3.2 Predictive Modeling with Random Forests

Any Random Forests model with a ROC value greater than 0.5 is considered non-random, and therefore a valid predictive model. Complete tables with best ROC values for each data set can be found in the appendix (page 150 ff), while all ROC values are available from the digital appendix DVD. Here only the key results will be displayed. Analysis is run on the pooled data from random and systematic plots and from aural as well as visual detections, unless stated otherwise.

Figure 16 shows the relationship between ROC value and number of observations. An overview about which narrative names are summarized together and analyzed at the biological family and/or biological order level is given in the appendix starting at page 166.

Generally there are many valid models for narrative names with less than 20 observations, but there are also many random models. All models with at least 80 observations, which is recommended as a minimum for DISTANCE sampling analysis, result in valid non-random models in Random Forests. This picture is less clear for analysis at the biological order or family level, in both of these cases there are random models (ROC \leq 0.5) or models with a ROC only slightly higher than 0.5 which build on 200 or more observations. This might be an indicator that pooling in taxonomic classes is not a valid way to receive bigger datasets, especially since differences on the biological level, like habitat requirements, can be huge between two species belonging to the same taxonomic tree.

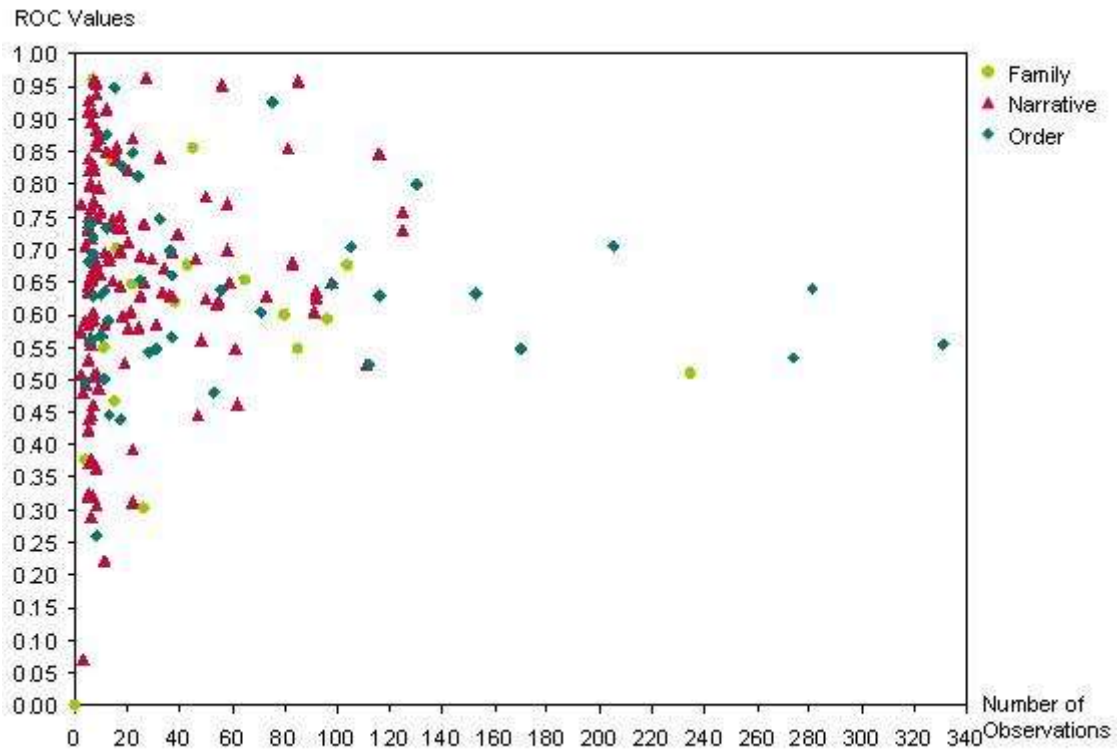


Figure 16: Correlation between number of observations and ROC values of model

3.2.1 ROC Values by Region and Model

Three different models are used for predictive modeling with Random Forests: *Plot*, *Covariates*, and *Interspecies*. *Plot* uses only detection/ non-detection data for each plot, combined with this plot's covariates. *Covariates* uses all detections at a plot combined with its covariates, allowing for spatial repeats. *Interspecies* is basically the same model as *Covariates*, but adds detection/ non-detection of other species as additional covariates to the analysis. Naturally, the *Plot* model was possible for fewer narratives than the other two (cp. chapter 2.3.1).

The following figures from Figure 17 to Figure 28 compare the ROC values from these different models for each narrative name. For each study area there are two figures, one comparing all narratives analyzed with three models (including *Plot*), the second comparing all narratives analyzed with only two models because the number of plots where the narrative was detected was below five and thus not sufficient to run the *Plot* analysis. This affects especially all trapping web data, because it was collected at four plots only. Narrative names

from point and line transects are capitalized to be able to discern them from trapping web narratives.

Of 22 narrative names observed at study area 1CR predictive modeling results in valid models for all but two of them (Dove and Spider), for both of them analysis with *Plot* model is not possible. The *Plot* model failed to result in valid models for three more narrative names (Mealy Parrot, Oropendula, and Parrot). ROC value results from the different models usually are very close together (Figure 17 and Figure 18). The *Plot* model slightly outperforms the other two models in three cases (Parrot large, Toucan, and Woodpecker). *Covariates* proved to be the best model for 11 narrative names and *Interspecies* for 9 narrative names, but both usually yield close results.

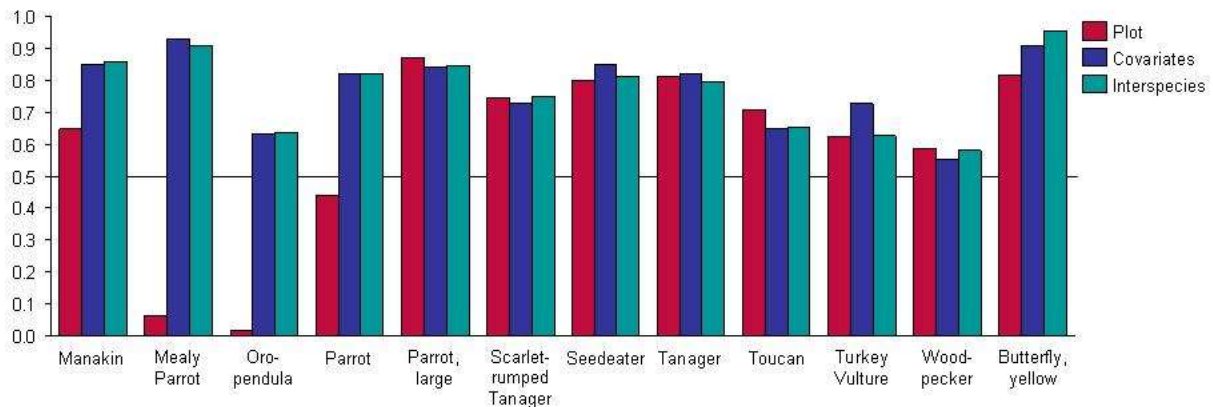


Figure 17: ROC values for narratives at 1CR (analysis with three different models)

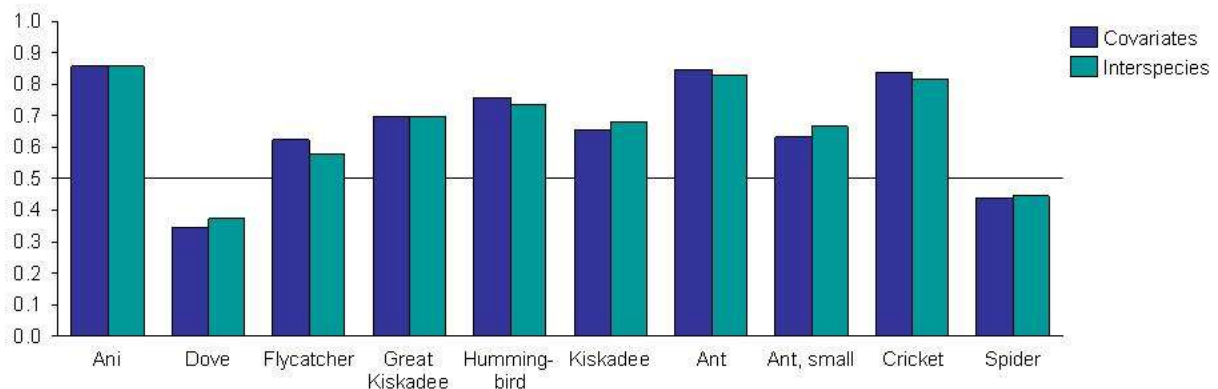


Figure 18: ROC values for narratives at 1CR (analysis with two different models)

30 narrative names are analyzed with Random Forests at study area 2Ni (Figure 19 and Figure 20). Valid models with ROC values > 0.5 are retrieved for 25 of those narratives, the lacking

five were Parakeet, Swallow, Beetle ground, Bristletail, and Caterpillar 877. *Plot* outperforms the other models only for Hawk. Adding species data to the modeling process does not increase ROC values in most cases, *Interspecies* is the best model only in four cases, and only in one of them it is actually better than *Covariates* (Butterfly, yellow), in the other three cases the results are equal. *Covariates* outperforms the other two models in 20 out of all valid models (ROC > 0.5).

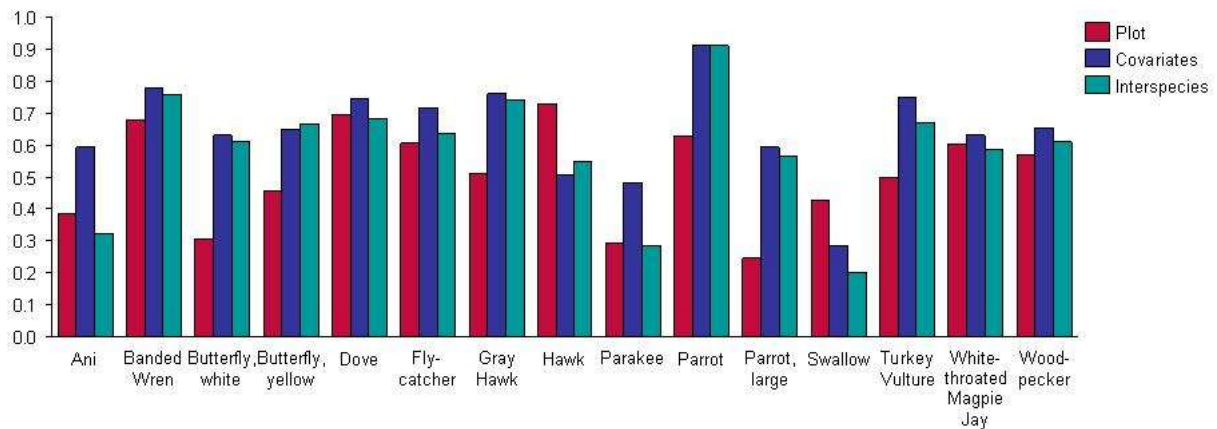


Figure 19: ROC values for narratives at 2Ni (analysis with three different models)

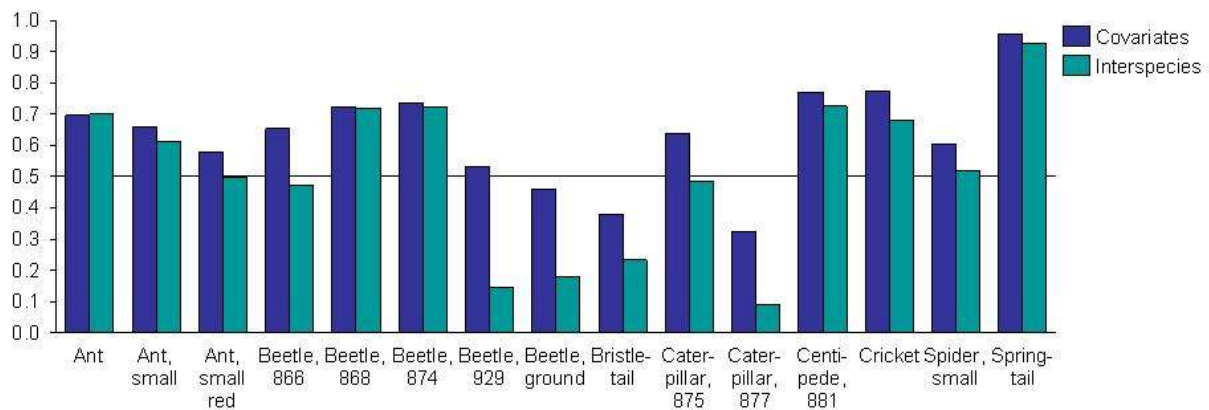


Figure 20: ROC values for narratives at 2Ni (analysis with two different models)

11 Narratives are analyzed from the observations at 3AK, 10 of which result in non-random models (Figure 21 and Figure 22). The only case in which the *Plot* model outperforms the other two is for Squirrel, but there the difference is very clear (ROC 0.86 against ROC 0.56). In all other cases the models *Covariates* and *Interspecies* are again very close, with a maximum ROC value difference of only 0.05 (for Spider, tiny).

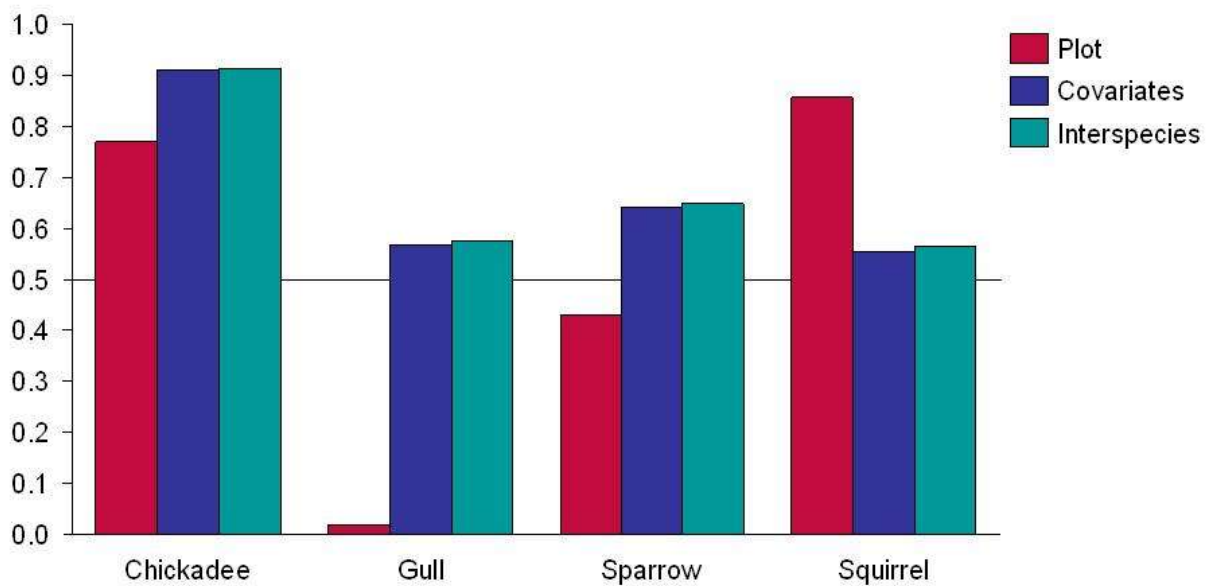


Figure 21: ROC Values for narratives at 3AK (analysis with three different models)

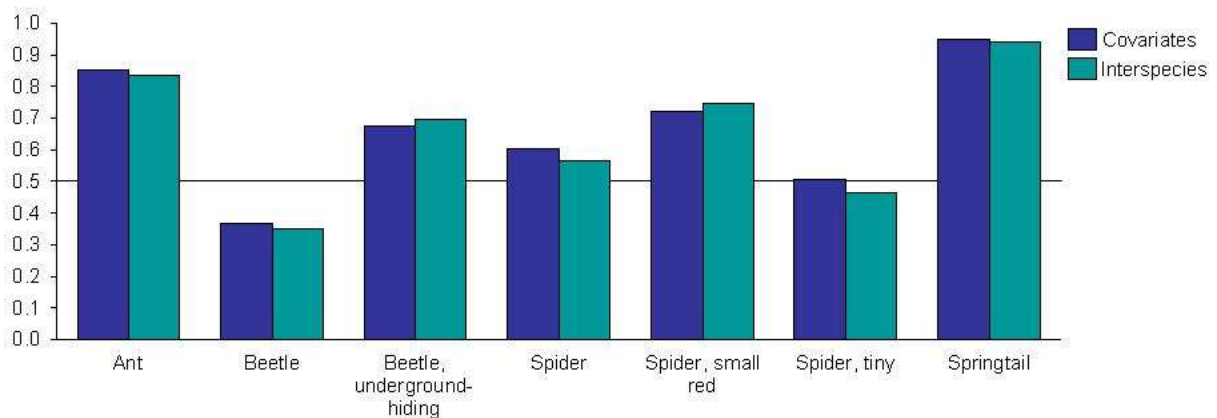


Figure 22: ROC values for narratives at 3AK (analysis with two different models)

23 out of the 27 narratives analyzed from 4Ru result in valid models, no predictive model is gained for Oriental Dove, Oriental Greenfinch, Beetle, and Spider (Figure 23 and Figure 24). The *Plot* model gains the same ROC value as the best other model for Chickadee and Woodpecker, but does not outperform any of the other models. *Covariates* and *Interspecies* results are again very close to each other, in five cases exactly the same ROC values are received. If those are disregarded the *Covariates* model outperforms the other two in 12 cases, the *Interspecies* model in 5 cases.

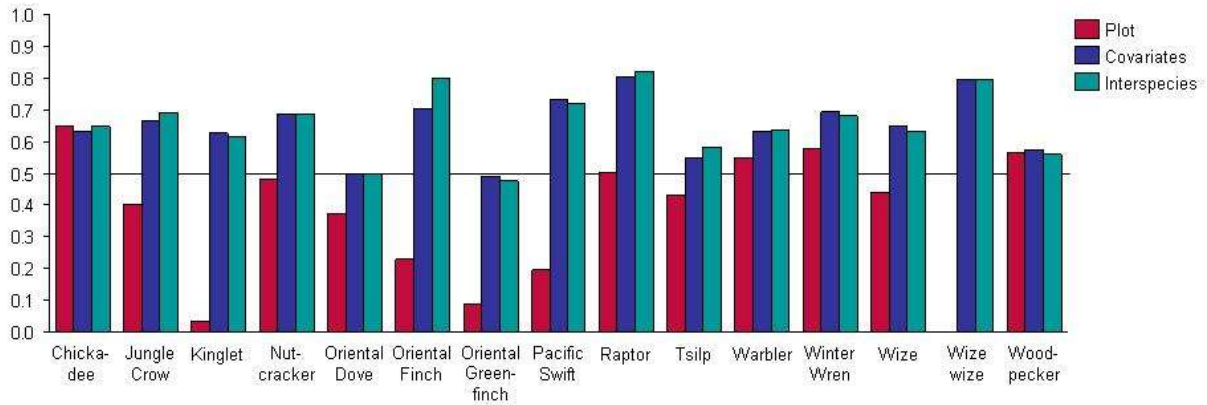


Figure 23: ROC values for narratives at 4Ru (analysis with three different models)

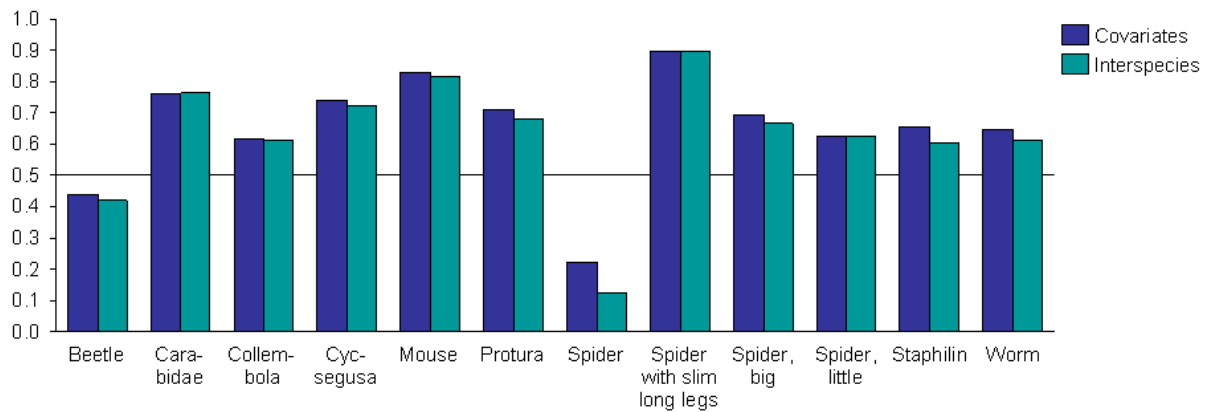


Figure 24: ROC values for narratives at 4Ru (analysis with two different models)

At study area 5PG 31 narratives are analyzed, of which only 23 result in valid models (Figure 25 and Figure 26). The eight narratives not adequately modeled are Parrot, Wize Wize, Woodpecker, tiny black Ant, Balu, Hawk, Hornbill, and Melodious Song. *Plot* outperforms other models for narrative *Wiz Wiz*, *Interspecies* is the best model for predictive modeling of Swallow. In the other 21 cases *Covariates* outperforms the other models.

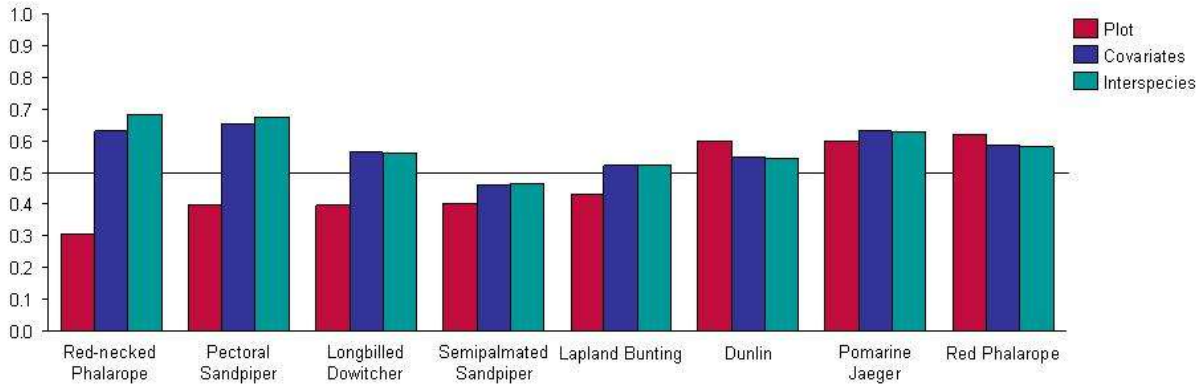


Figure 27: ROC values for narratives at 6Ba (analysis with three different models)

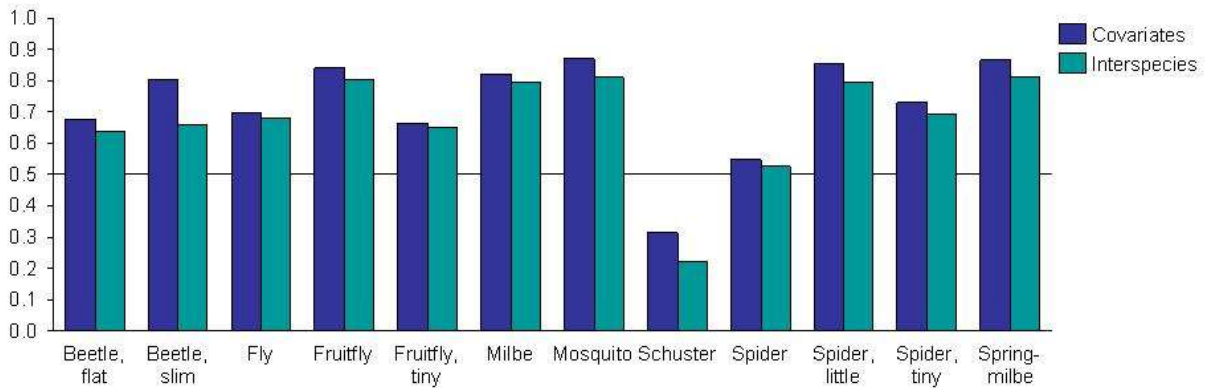


Figure 28: ROC values for narratives at 6Ba (analysis with two different models)

Table 6 shows the total number of models for each study area and the number of valid models with ROC values > 0.5 derived for this particular area. It also summarizes how often each of the models is the best valid model with the highest ROC value. If two models gain the same ROC value, both are regarded as best models, thus adding all best models together results in a number larger than the number of valid models (140 compared to 119). In about 67 % of all cases *Covariates* is the best model to predict a narrative, in 25 % it is *Interspecies*, and in 8 % the *Plot* model.

Table 6: Overview of models with best ROC values by region

| | Total no of models | No of models with ROC>0.5 | Best model: <i>Plot</i> | Best model: <i>Covariates</i> | Best model: <i>Interspecies</i> |
|--------------|--------------------|---------------------------|-------------------------|-------------------------------|---------------------------------|
| 1CR | 22 | 20 | 4 | 11 | 9 |
| 2Ni | 30 | 25 | 1 | 24 | 4 |
| 3AK | 11 | 10 | 1 | 6 | 5 |
| 4Ru | 27 | 23 | 2 | 18 | 11 |
| 5PG | 31 | 23 | 1 | 21 | 1 |
| 6Ba | 20 | 18 | 2 | 14 | 5 |
| Total | 141 | 119 | 11 | 94 | 35 |

3.2.2 Randomly Selected vs. Systematically Selected Plots

For Random Forests analysis all observations from randomly selected and systematically selected plots are added together. Statistically this approach can be further stratified and fine-tuned. The pooling is done to increase sample size and it is based on the assumption that biology, occupancy, abundance, and all other attributes of a population do not differ between random and systematic plots for the GRID area. To check this assumption the data from random and systematic plots is analyzed separately and the best ROC value results compared with the results from the pooled data set. This is also set in relation to the number of observations gained from each of the two plot types. Since 25 plots are systematically selected at each GRID and only five are randomly selected the assumption would be that the random plots yield about 20 % of all observations, the systematic plots about 80 %. Since trapping webs have only been run at systematic plots this analysis is aiming at point and line transect data only.

Figure 29 and Figure 30 show these comparisons for narratives from 1CR and 2Ni. Generally the observations at random plots have a share of between 16 % and 23 % of total observations, coming close to the expected 20 % (Figure 30). Exceptions are Flycatcher at 1CR with only 10% and Banded Wren at 2Ni with 11 %. In effect, Flycatcher is the only narrative for which the data from randomly selected plots does not result in a non-random model, while there is no obvious effect at Banded Wren models. Surprisingly, the small amount of observations at random plots results in better models for Woodpecker and white Butterfly than the systematic or pooled data sets. For all other narratives either the systematic or the pooled data results in better models, with both values usually being close together. In three of these cases the pooled data delivers slightly higher ROC values than the systematic

data, in two cases the ROC values of both are equal, and in the case of Banded Wren the ROC value for systematic data is 0.01 higher than the one for pooled data.

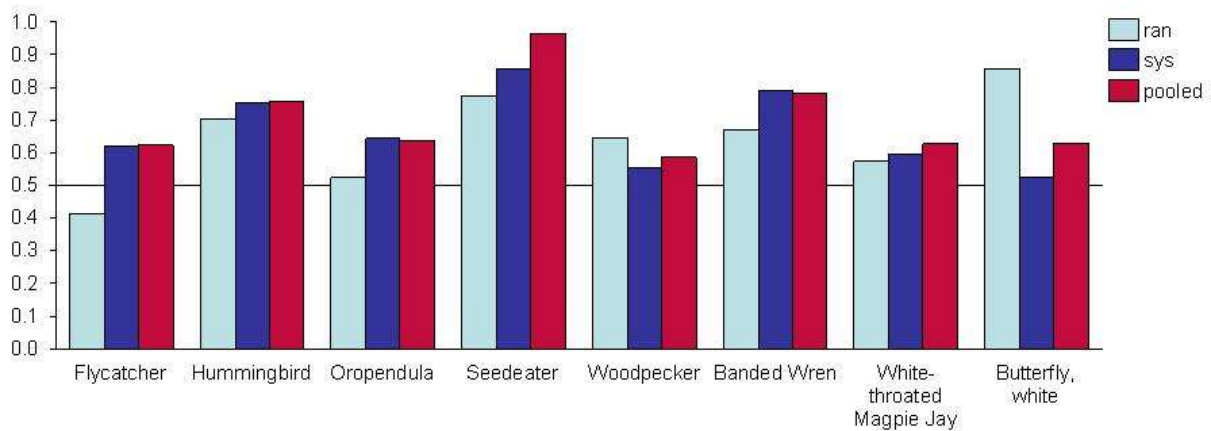


Figure 29: Best ROC values by plot type (1CR & 2Ni)

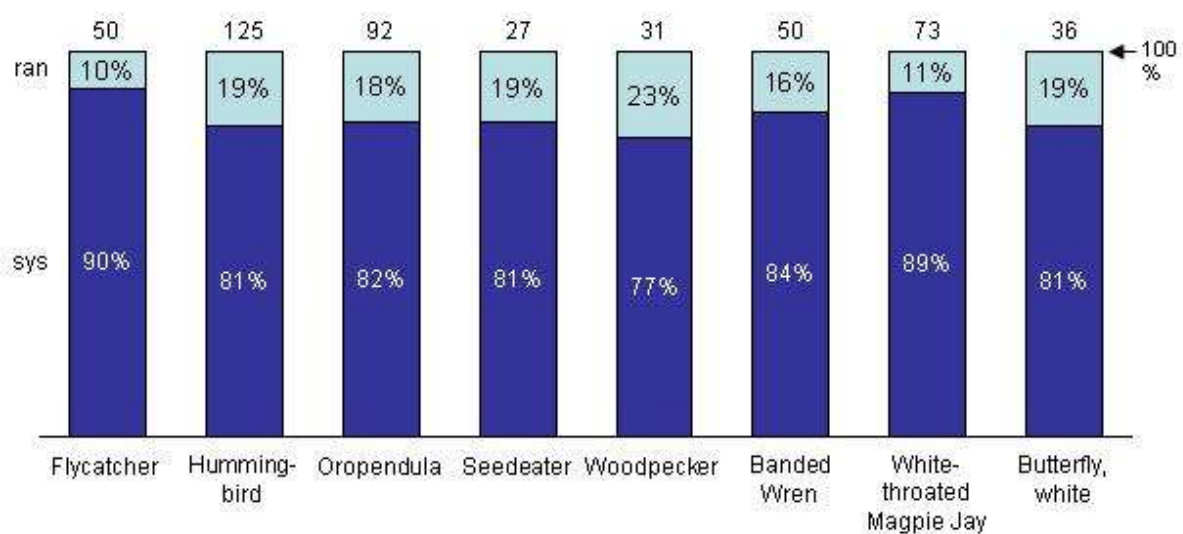


Figure 30: Distribution of observations by plot type (1CR & 2Ni)

ROC values for the different data sets from 3AK and 4Ru are shown in Figure 31, the distribution of observations between plot types in Figure 32. It is remarkable that only two species generate close to 20 % of observations from random plots: Squirrel at 3AK (16 %) and Kinglet at 4Ru (25 %). Proportions of observations from random plots for Sparrow at 3AK and for Chickadee and Wize at 4Ru are all quite low (12 %). Even lower is this proportion for Nutcracker at 4Ru (7 %), while all Warbler and Winter Wren observations at 4Ru stem from systematic plots. Only in the latter two cases an effect on the ROC values

derived through the random data sets is visible because the modeling is impossible without input observations. Nutcracker with only 7 % of observations from random plots even has the highest ROC value from a random data set among all narratives. At 3AK pooling of the data for Squirrel results by far in the best model, while pooled data set and the random data set have the same ROC values for Sparrow, which is only slightly better than the one from systematic data set. At 4Ru the random data set results in the best model for Nutcracker, the systematic data set in the best model for Chickadee, and the pooled data set in the best model for the other four narratives.

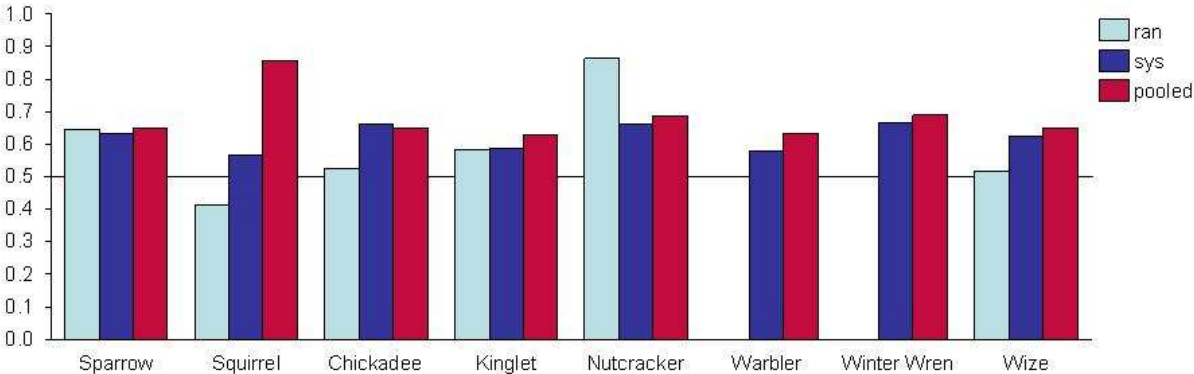


Figure 31: Best ROC values by plot type (3AK & 4Ru)

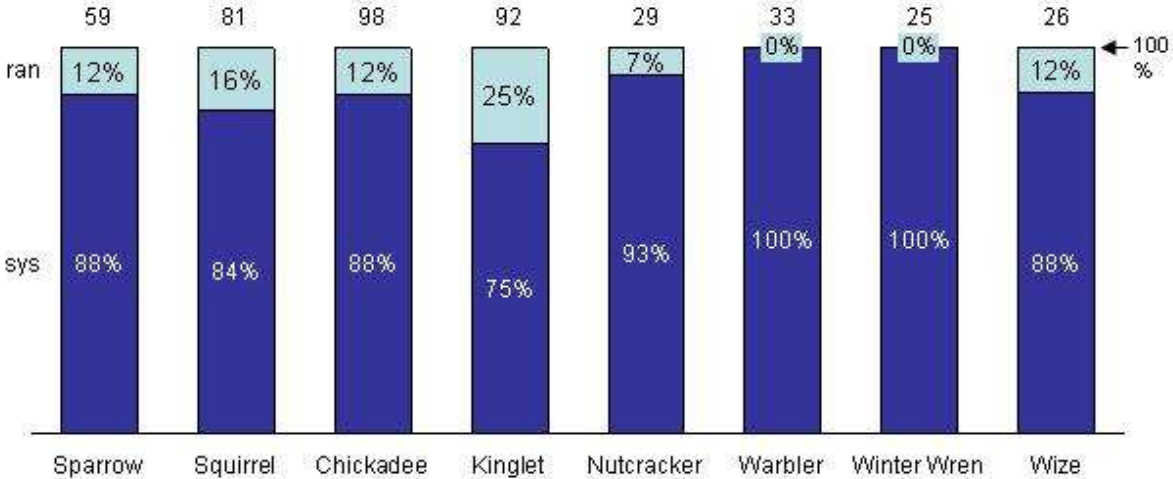


Figure 32: Distribution of observations by plot type (3AK & 4Ru)

At 5PG exactly 20% of observations are made from random plots, while for four of the six narratives from 6Ba the proportion of observations from random plots falls between 15 % and 19 % (Figure 34). Only 11 % of Red Phalarope observations are made from random plots,

while Pectoral Sandpiper observations are only made from systematic plots. As a result, the random data set for Pectoral Sandpiper can not be modeled (Figure 33). The random data set for Red Phalarope results in a poor model with ROC = 0.39, but compared to the other models this does not seem to be a result of the relatively small proportion of observations from random plots. For both Flute and Tsilp from 5PG the highest ROC value is derived for analysis of the pooled data set. At 6Ba the models for Lapland Bunting and Semipalmated Sandpiper based on the random data set have the highest ROC value, for Pomarine Jaeger the one based on the systematic data set, and for the other three narratives based on the pooled data set.

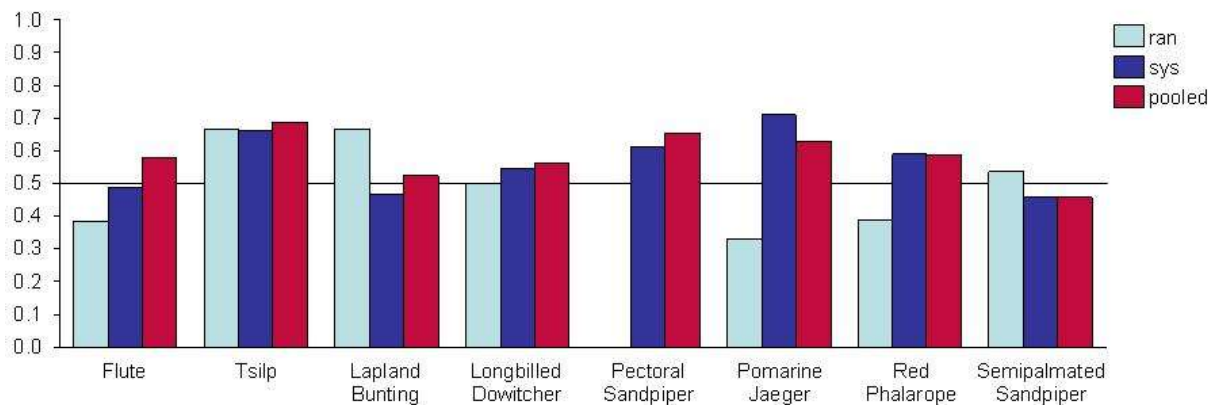


Figure 33: Best ROC values by plot type (5PG & 6Ba)

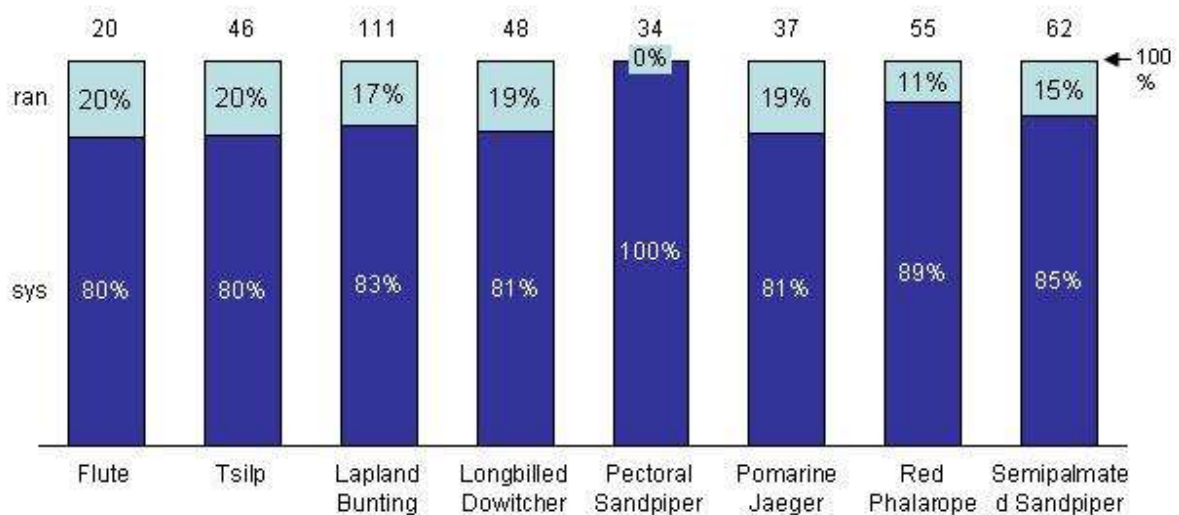


Figure 34: Distribution of observations by plot type (5PG & 6Ba)

Table 7 gives an overview about analysis of which data set resulted in the best model for the narratives analyzed. If two models gain the same ROC value, both are regarded as best models. Thus adding all best models together results in a number larger than the number of valid models (24 compared to 27). Differences between ROC values from systematic and from pooled data sets are relatively small, but usually the pooled data set performs slightly better (resulting in the low number of best models derived through systematic data sets). Sometimes the few observations at random plots analyzed separately result in surprisingly strong models.

Table 7: Best models for data sets from different plot types (random, systematic, pooled)

| | Total no of models | No of models with ROC>0.5 | Best model: random | Best model: systematic | Best model: pooled |
|--------------|---------------------------|-------------------------------------|---------------------------|-------------------------------|---------------------------|
| 1CR | 5 | 5 | 1 | 2 | 4 |
| 2Ni | 3 | 3 | 2 | 0 | 1 |
| 3AK | 2 | 2 | 1 | 0 | 2 |
| 4Ru | 6 | 6 | 1 | 1 | 4 |
| 5PG | 2 | 2 | 0 | 0 | 2 |
| 6Ba | 6 | 6 | 2 | 1 | 3 |
| Total | 24 | 24 | 7 | 4 | 16 |

3.2.3 Aural vs. Visual Bird Detections

It is common knowledge in bird surveys that visual detectability differs from aural detectability (Buckland et al. 2008). Birds can not be seen, but often be heard and identified by their song. To check the effect of pooling these two kinds of detections together the data is analyzed separately and the best ROC value results compared with the best results from the pooled data set. This is also set in relation to the number of observations gained from each of the two kinds of observation. This analysis is only done for the first five study areas. At the 6th study area 6Ba all detections are obtained visually, because the tundra is an open habitat hardly without visual distractions.

Figure 35 compares the ROC values of point transect detections from 1CR, 2Ni and 3AK; while Figure 36 shows the percentages of aural and visual detections. The proportion of visual detections ranges from 89 % for Seedeater in 1CR to only 4 % for Banded Wren in 2Ni and Squirrel in 3AK. The overall effect on ROC values seems to be rather low. For example, although 89 % of Seedeater observations are visual, the model built on the visual data set has almost the same ROC value as the one using the remaining 11 % of aural detections (0.73

compared to 0.72). The 4 % of Banded Wren detections which are visual result in a poorer model than the 96 % of aural detections (0.40 compared to 0.54), but the 4 % of Squirrel detections which are visual actually gain a much better model than the 96 % of aural detections (0.92 compared to 0.48).

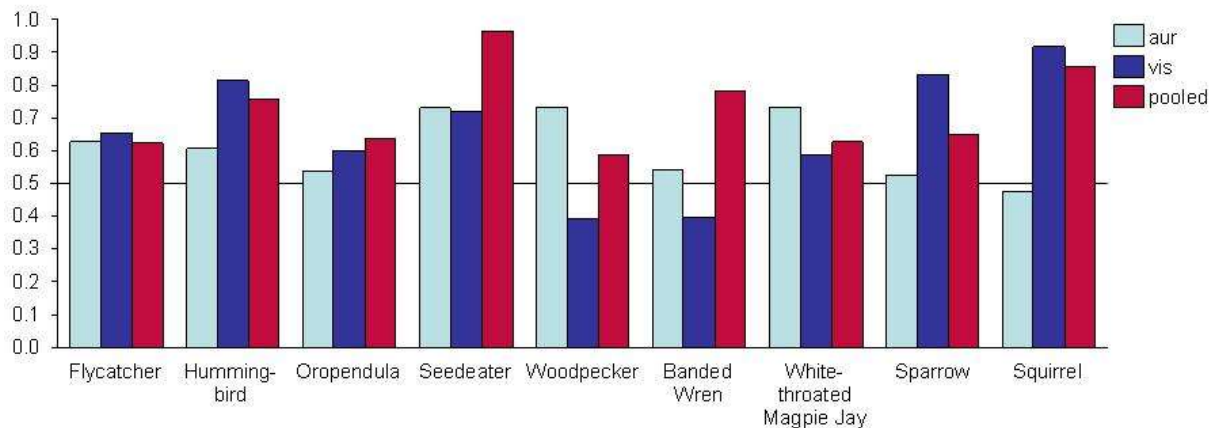


Figure 35: Best ROC values by type of observation (1CR-3AK)

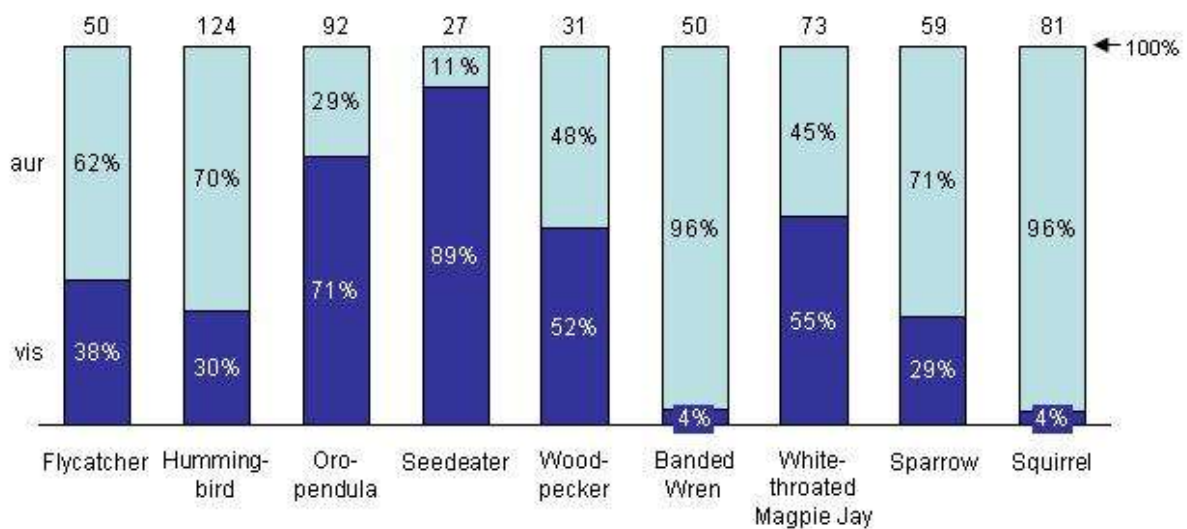


Figure 36: Distribution of observations by type of observation (1CR-3AK)

Figure 38 shows the distribution of detections between aural and visual in 4Ru and 5PG. Clearly most observations in these areas are aural, with percentages between 72 % and 100 %. Chickadee, Kinglet and Nutcracker with between 72 % and 75 % of observations being aural receive good ROC values from these data sets, while ROC values of this data set for narratives with more than 90 % aural detections are relatively poor (Figure 35).

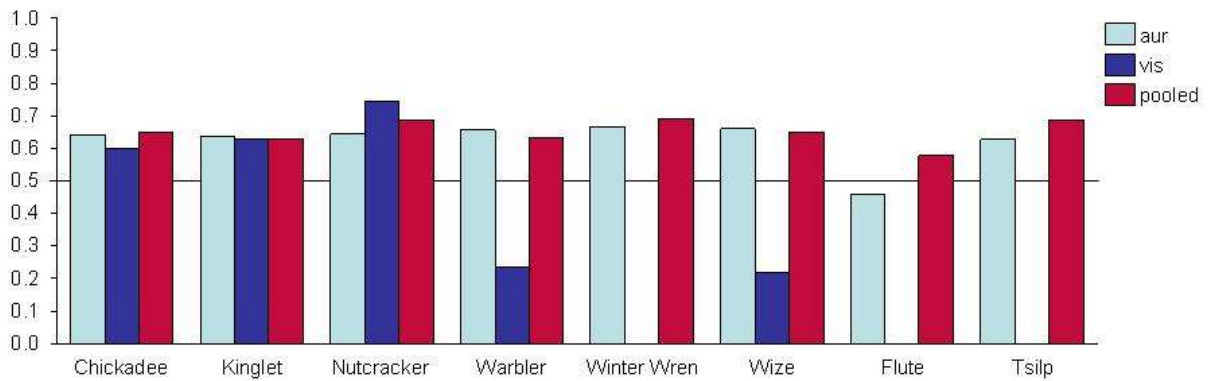


Figure 37: Best ROC values by type of observation (4Ru & 5PG)

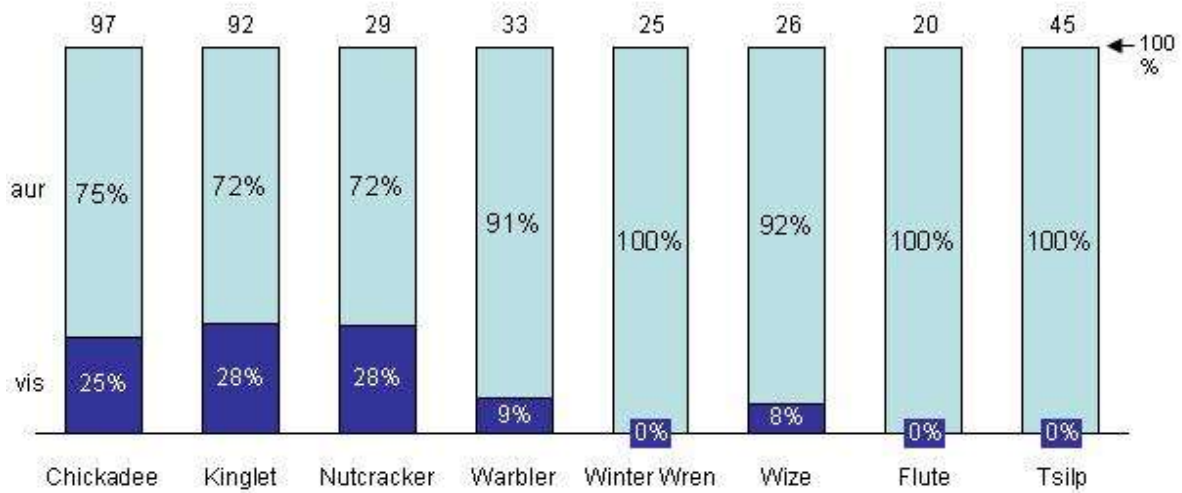


Figure 38: Distribution of observations by type of observation (4Ru & 5PG)

An overview about which data set results in the highest ROC value for a narrative is given in Table 8. The overall distribution is quite even between aural, visual and pooled data sets.

Table 8: Best models for data sets from different types of detection (aural, visual, pooled)

| | Total no of models | No of models with ROC>0.5 | Best model: aural | Best model: visual | Best model: pooled |
|--------------|--------------------|---------------------------|-------------------|--------------------|--------------------|
| 1CR | 5 | 5 | 1 | 2 | 2 |
| 2Ni | 2 | 2 | 1 | 0 | 1 |
| 3AK | 2 | 2 | 0 | 2 | 0 |
| 4Ru | 6 | 6 | 3 | 1 | 2 |
| 5PG | 2 | 2 | 0 | 0 | 2 |
| 6Ba | - | - | - | - | - |
| Total | 17 | 17 | 5 | 5 | 7 |

3.2.4 Biological Family and Order as Analysis Targets

The main unit of interest for any biodiversity assessment is the biological species. But in many cases identification to species level is not possible, or the number of observations is too small for analysis at species level. To make use of this data, predictions at the biological family and biological order level are made in this section, detailed tables showing which narratives are summarized under which family and/or order name can be found in the appendix (page 166 ff). This chapter gives a short overview about analysis trends when moving up the taxonomic tree.

Figure 39 shows the best ROC values for analysis at biological family level, while Figure 40 gives an overview about how many observations the model was built on. A clear trend is not visible, for example 44 observations of Paradisaeidae from 5PG reached a higher ROC than 470 observations of Scolopacidae from 6Ba (0.65 compared to 0.51). Only the model for Tipulidae from 6Ba did not achieve a ROC value > 0.5 . The highest ROC value achieved is 0.86 for Thraupidae from 1CR, building on 90 observations.

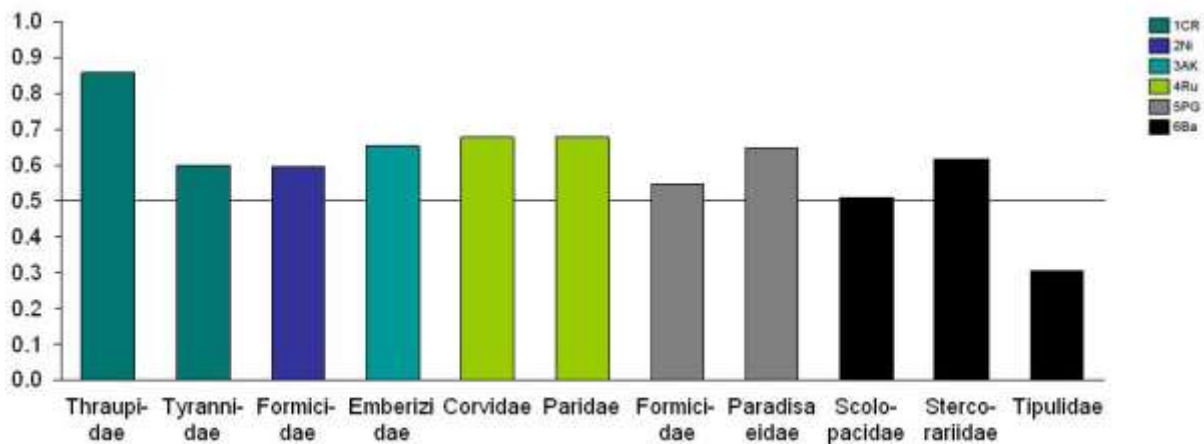


Figure 39: Best ROC values for analysis at biological family level

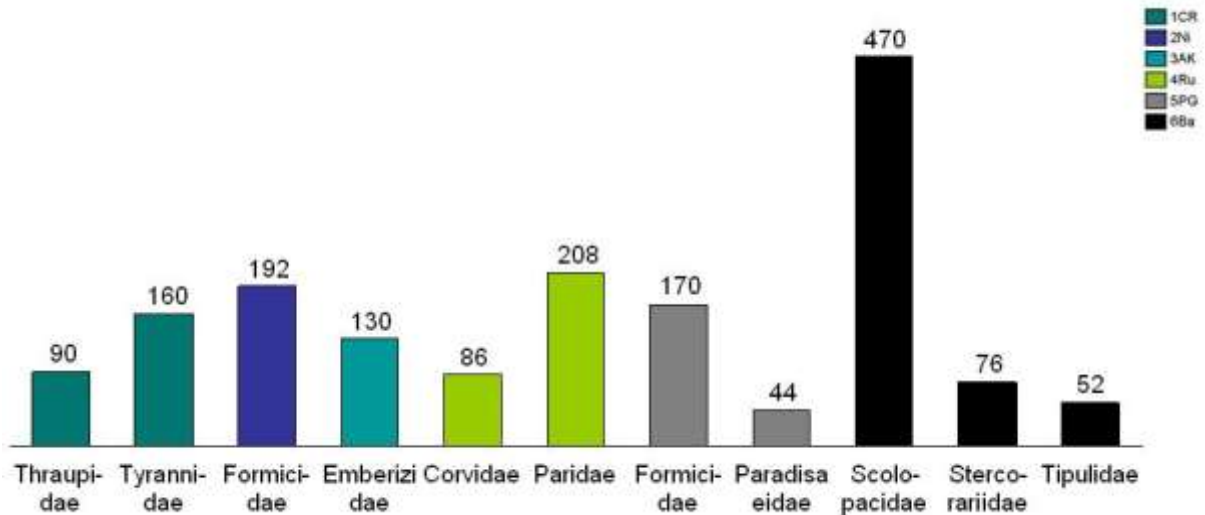


Figure 40: Number of observations pooled by biological family

The second biological level analyzed is the biological order. Results are shown in Figure 41 and Figure 42 for the study areas 1CR, 2Ni and 3AK. Also here larger numbers of detections do not automatically result in higher ROC values. Psittaciformes from 1CR built on the lowest number of observations received a ROC value of 0.85, while Passeriformes from the same study area built on the largest number of observations received a ROC value of 0.64. In this set ROC values > 0.5 indicate valid models for all runs at the biological order level.

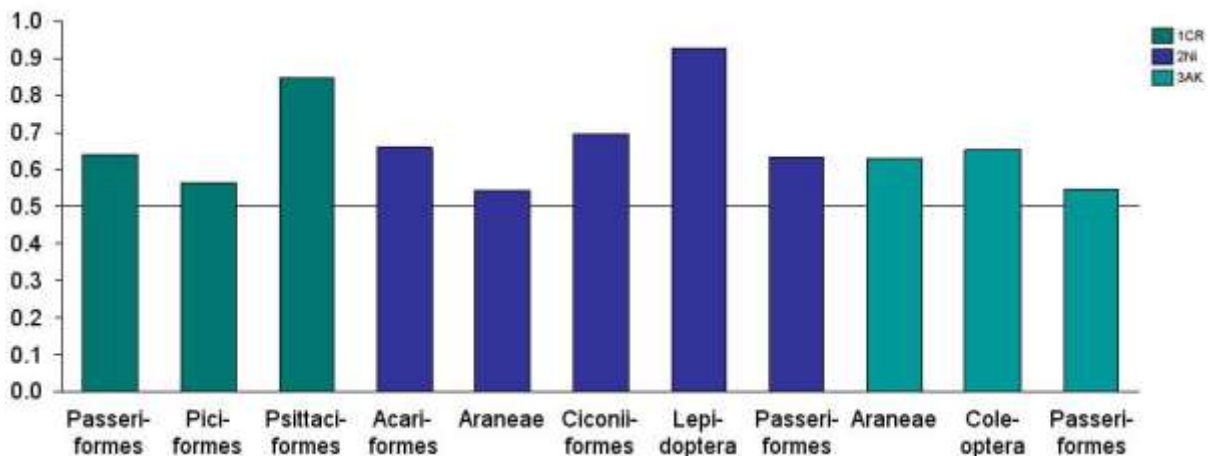


Figure 41: Best ROC values for analysis at biological order level (1CR-3AK)

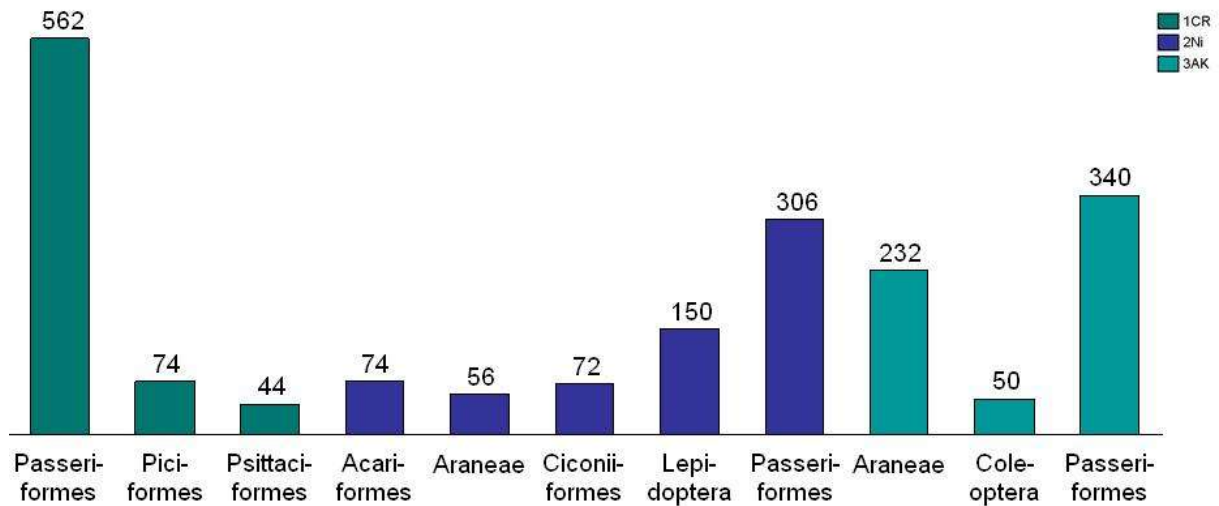


Figure 42: Number of observations pooled by biological order (1CR-3AK)

Figure 43 and Figure 44 illustrate the best ROC values and number of observations for the remaining study areas 4Ru, 5PG and 6Ba. Only one of the orders does not result in higher quality non-random model (ROC of 0.48 for collembola at 5PG). The lowest number of observations compared within this set leads again to the best available model (ROC = 0.81 for Psittaciformes at 5PG), while the highest number of observations resulted in a relatively poor ROC of 0.56 (Passeriformes at 4Ru).

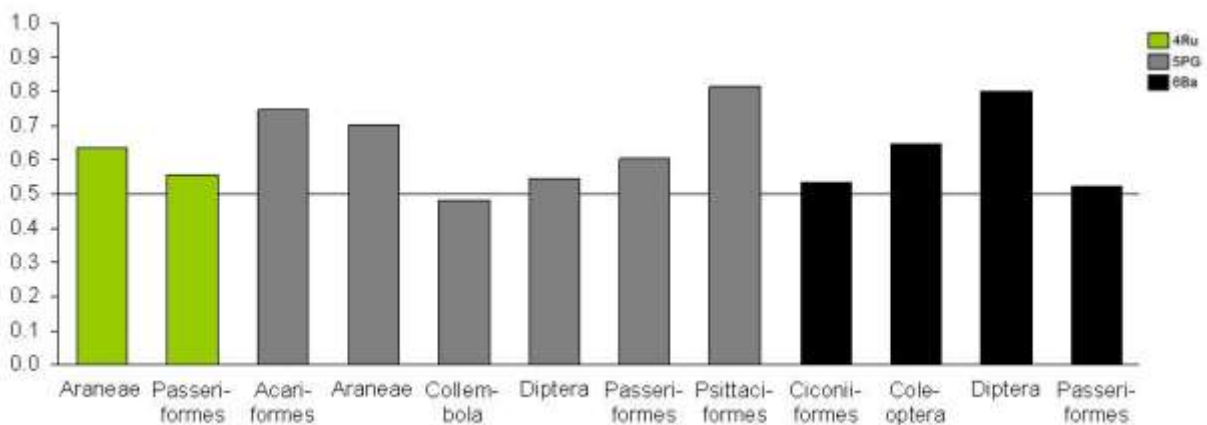


Figure 43: Best ROC values for analysis at biological order level (4Ru-6Ba)

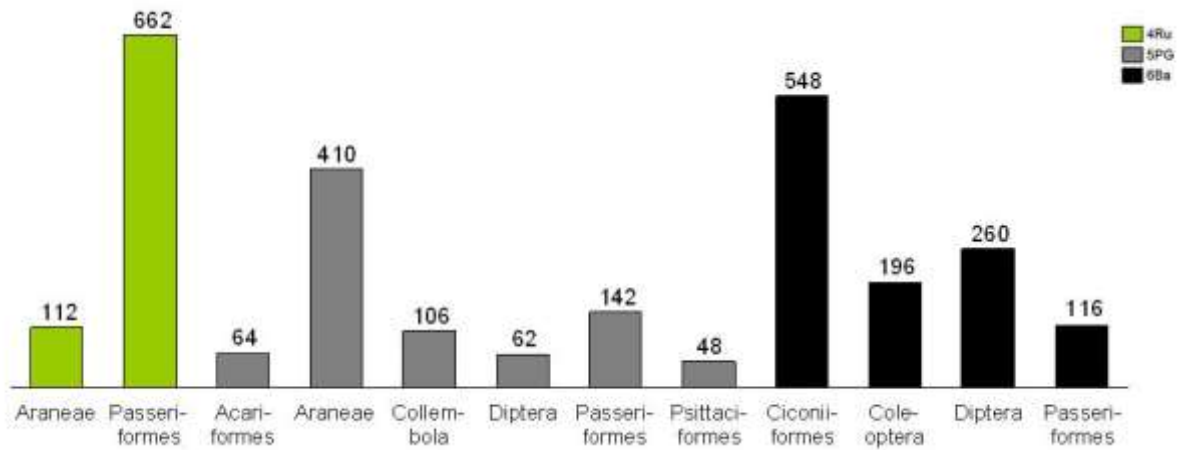


Figure 44: Number of observations pooled by biological order (4Ru-6Ba)

3.2.5 Covariates Identified as Important

Random Forests assigns importance values to each covariate used in a model, the ones identified as most important are used for further analysis. Table 9 to

Table 11 show the ten most important covariates for each narrative in the different study areas for point and line transect data, Table 12 to Table 14 illustrate the five most important covariates for trapping web data. Additionally the model resulting in the best ROC values, the ROC value, and the number of observations are shown for each narrative. Full results can be found in the project files in the digital appendix. Generally speaking all environmental covariates that are spatially tied to a plot, like habitat type, height of highest tree or presence/absence of key plant species, are ‘good’ results, those can easily used for prediction when the spatial data is available. Other covariates that are survey-specific are rather difficult as input variables, because they area unknown prior to sampling (e.g. cluster size, aural or visual identification). Some of them are even indicators that there could have been a problem with survey circumstances, when they should not have an effect but do (e.g. minutes since sunrise, number of visit). A detailed species-based biological discussion of covariate influence is beyond the scope of this thesis.

Table 9: Covariates identified as important for point and line transect observations (1CR-3AK)

| Region | 1CR | 1CR | 1CR | 1CR | 1CR | 2Ni | 2Ni | 2Ni | 3AK | 3AK |
|-----------------|-------------|-------------|--------------|----------------|-------------|--------------|---------------------------|------------------|--------------|--------------|
| Model | Covariates | Covariates | Interspecies | Interspecies | Covariates | Covariates | Covariates | Covariate | Interspecies | Interspecies |
| Target Variable | Flycatcher | Hummingbird | Oropendula | Seedeater | Woodpecker | Banded Wren | White-throated Magpie Jay | Butterfly, white | Sparrow | Squirrel |
| ROC Integral | 0.623 | 0.756 | 0.635 | 0.812 | 0.585 | 0.78 | 0.628 | 0.629 | 0.649 | 0.564 |
| Observations | 50 | 125 | 92 | 27 | 31 | 50 | 73 | 36 | 59 | 81 |
| VarImp01 | Visit No | Duff | Ident | Habitat | Habitat | Ident | Cluster Size | Distance | High. Tree | Duff |
| VarImp02 | Distance | Habitat | Moss/Lichen | Duff | Moss/Lichen | Habitat | Epiphytes | Min_Sunrise | Habitat | Ident |
| VarImp03 | Min_Sunrise | Epiphytes | High. DBH | Moss/Lichen | Canopy | Moss/Lichen | Habitat | Bare Soil | Duff | Min_Sunrise |
| VarImp04 | Habitat | Moss/Lichen | High. Tree | Epiphytes | Epiphytes | Understory | Min_Sunrise | Duff | Moss/Lichen | Cov13 |
| VarImp05 | Shrubs | Canopy | Habitat | Cov05 | Visit No | Duff | Moss/Lichen | High. DBH | Min_Sunrise | Cov14 |
| VarImp06 | Epiphytes | Understory | Epiphytes | Shrubs | Cov04 | Distance | High. DBH | Habitat | Canopy Trees | High. Tree |
| VarImp07 | Plot Type | High. Tree | Distance | Understory | High. DBH | Epiphytes | Ident | Canopy | Canopy | Cov12 |
| VarImp08 | Moss/Lichen | Distance | Canopy Trees | High. Tree | Bare Soil | High. Tree | Distance | Canopy Trees | Squirrel | Cov11 |
| VarImp09 | Ident | Ident | Bare Soil | Canopy | Min_Sunrise | Shrubs | High. Tree | Epiphytes | Cov19 | Moss/Lichen |
| VarImp10 | Cov05 | Cov05 | Flowers | Turkey Vulture | High. Tree | Cluster Size | Shrubs | High. Tree | Cov01 | High. DBH |

Table 10: Covariates identified as important for point and line transect observations (4Ru-5PG)

| Region | 4Ru | 4Ru | 4Ru | 4Ru | 4Ru | 4Ru | 5PG | 5PG |
|-----------------|--------------|--------------|------------|--------------|-------------|------------|-------------|-------------|
| Model | Interspecies | Covariates | Covariates | Interspecies | Covariates | Covariates | Covariates | Covariates |
| Target Variable | Chickadee | Kinglet | Nutcracker | Warbler | Winter Wren | Wize | Flute | Tsilp |
| ROC Integral | 0.647 | 0.625 | 0.685 | 0.634 | 0.691 | 0.65 | 0.578 | 0.686 |
| Observations | 98 | 92 | 29 | 33 | 25 | 26 | 20 | 46 |
| VarImp01 | Moss % | Moss % | Moss % | Moss % | Moss % | Moss % | Habitat | Visit No |
| VarImp02 | High. DBH | Distance | Cov01 | Cov16 | Cov01 | Lichen % | Cov12 | Ident |
| VarImp03 | Lichen % | High. Tree | Cov21 | Lichen % | Habitat | Cov20 | Ident | Min_Sunrise |
| VarImp04 | Cluster Size | Lichen % | Cov12 | High. DBH | Shrubs | Cov18 | Visit No | Habitat |
| VarImp05 | High. Tree | Visit No | Cov05 | Plot Type | Cov21 | Visit No | Cov11 | Canopy |
| VarImp06 | Cov16 | Plot Type | Habitat | Cov31 | Plot Type | Cov30 | Duff | Bare Soil |
| VarImp07 | Understory | Min_Sunrise | Cov20 | Cov15 | Duff | Cov31 | High. Tree | Epiphytes |
| VarImp08 | Wize | Canopy Trees | Understory | Understory | Canopy | Cov28 | Min_Sunrise | Distance |
| VarImp09 | Duff | Understory | Cov15 | Cov20 | Lichen % | Cov16 | Bare Soil | Cov01 |
| VarImp10 | Distance | Shrubs | Cov23 | Canopy | Distance | Understory | Distance | High. Tree |

Table 11: Covariates identified as important for point and line transect observations (6Ba)

| Region | 6Ba | 6Ba | 6Ba | 6Ba | 6Ba | 6Ba |
|-----------------|-----------------|----------------------|--------------------|-----------------|---------------|------------------------|
| Model | Covariates | Covariates | Interspecies | Covariates | Covariates | Interspecies |
| Target Variable | Lapland Bunting | Longbilled Dowitcher | Pectoral Sandpiper | Pomarine Jaeger | Red Phalarope | Semipalmated Sandpiper |
| ROC Integral | 0.523 | 0.561 | 0.671 | 0.629 | 0.585 | 0.463 |
| Observations | 111 | 48 | 34 | 37 | 55 | 62 |
| VarImp01 | Moss % | Grass % | Grass % | Grass % | Moss % | Grass % |
| VarImp02 | Grass % | Moss % | Moss % | Moss % | Grass % | Moss % |
| VarImp03 | Visit No | Lichen % | Diam. Lake | Dist. Lake | Leafs | Diam. Lake |
| VarImp04 | Diam. Lake | Diam. Lake | Leafs | Leafs | Diam. Lake | Dist. Lake |
| VarImp05 | Leafs | Dist. Lake | Dist. Lake | Cov06 | Dist. Lake | Leafs |
| VarImp06 | Dist. Lake | Cov03 | Plot Type | Diam. Lake | Lichen % | Lichen % |
| VarImp07 | Distance | Leafs | Lichen % | Flowers | Cov02 | Visit No |
| VarImp08 | Flowers | Flowers | Cov08 | Lichen % | Flowers | Flowers |
| VarImp09 | Cluster Size | Cov08 | Cov01 | Cluster Size | Visit No | Distance |
| VarImp10 | Cov07 | Cov07 | Cov02 | Cov05 | Cov07 | Cov10 |

Table 12: Covariates identified as important for trapping web catches (1CR-2Ni)

| Region | 1CR | 1CR | 2Ni | 2Ni | 2Ni | 2Ni | 2Ni | 2Ni |
|-----------------|--------------|--------------|----------------|----------------|--------------|----------------|---------------|--------------|
| Model | Interspecies | Interspecies | Interspecies | Covariate | Covariate | Covariate | Covariate | Covariate |
| Target Variable | Ant | Spider | Ant | Ant, small red | Beetle, 868 | Centipede, 881 | Spider, small | Springtail |
| ROC Integral | 0.831 | 0.446 | 0.699 | 0.581 | 0.724 | 0.77 | 0.604 | 0.958 |
| Observations | 116 | 47 | 58 | 24 | 39 | 58 | 21 | 85 |
| VarImp01 | Epiphytes | Min Sunrise | Habitat | Cuplabel | Epiphytes | Status | Visit | Cluster Size |
| VarImp02 | Habitat | Visit | Bug, other red | Visit | Understory | Habitat | Cuplabel | Status |
| VarImp03 | Shrubs | Status | Visit Effort | Habitat | Habitat | Epiphytes | Min Sunrise | Visit |
| VarImp04 | Bare Soil | Moss Lichen | Bug, 870 | Epiphytes | Shrubs | Visit Effort | Cluster Size | Epiphytes |
| VarImp05 | Understory | High DBH | Toad | Visit Effort | Visit Effort | Visit | Epiphytes | Visit Effort |

Table 13: Covariates identified as important for trapping web catches (3AK-5PG)

| Region | 3AK | 3AK | 4Ru | 4Ru | 4Ru | 4Ru | 5PG |
|-----------------|-----------|--------------|------------|-----------|-----------|----------------|-----------------|
| Model | Covariate | Covariate | Covariate | Covariate | Covariate | Covariate | Interspecies |
| Target Variable | Spider | Springtail | Collembola | Cycsegusa | Protura | Spider, little | Ant, tiny black |
| ROC Integral | 0.604 | 0.951 | 0.616 | 0.74 | 0.709 | 0.628 | 0.393 |
| Observations | 91 | 56 | 54 | 26 | 20 | 25 | 22 |
| VarImp01 | Cov18 | Cov13 | Moss % | Cov21 | Cov11 | Cuplabel | Habitat |
| VarImp02 | Cov14 | Cluster Size | Lichen % | Habitat | Lichen % | Visit | Cov01 |
| VarImp03 | Cov11 | Cov12 | Cov01 | Moss % | Cov19 | Cluster Size | Cov06 |
| VarImp04 | Cov08 | Moss Lichen | High. Tree | Cov11 | Moss % | Lichen % | Cov08 |
| VarImp05 | Cov04 | Habitat | Habitat | Lichen % | Habitat | Cov05 | Cov05 |

Table 14: Covariates identified as important for trapping web catches (6Ba)

| Region | 6Ba | 6Ba | 6Ba | 6Ba | 6Ba | 6Ba | 6Ba | 6Ba |
|-----------------|--------------|-----------|-----------|--------------|-----------|-----------|-----------|--------------|
| Model | Covariate | Covariate | Covariate | Covariate | Covariate | Covariate | Covariate | Covariate |
| Target Variable | Beetle, flat | Fly | Fruitfly | Milbe | Mosquito | Schuster | Spider | Spider, tiny |
| ROC Integral | 0.678 | 0.696 | 0.841 | 0.821 | 0.871 | 0.313 | 0.548 | 0.729 |
| Observations | 83 | 37 | 32 | 20 | 22 | 22 | 61 | 125 |
| VarImp01 | Moss % | Status | Status | Lichen % | Status | Visit | Status | Lichen % |
| VarImp02 | Lichen % | Moss % | Moss % | Cov10 | Visit | Status | Lichen % | Cov10 |
| VarImp03 | Grass % | Lichen % | Lichen % | Visit Effort | Cuplabel | Cuplabel | Cov10 | Grass % |
| VarImp04 | Status | Cov10 | Cov10 | Moss % | Grass % | Lichen % | Moss % | Cov01 |
| VarImp05 | Cov02 | Grass % | Grass % | Grass % | Moss % | Cov10 | Cov01 | Cov08 |

3.3 DISTANCE Sampling

46 different models are used for DISTANCE analysis to estimate abundance of each narrative; the different model definitions for each study area are given in the appendix (page 123 ff). Models 1-6 are standard models without covariates; models 7 and higher are covariate DISTANCE models using one covariate each. For each study area and sampling method the actual population densities as well as upper and lower confidence level are shown in this analysis, followed by DISTANCE detection function graphs for each narrative. Missing indicators for confidence levels indicate that they have not been calculated by the software.

3.3.1 DISTANCE Sampling Results: Bird Point Transects

Density estimates by the best available DISTANCE sampling model for narratives at 1CR range from 17 individuals per km² for Flycatcher to 2,908 individuals per km² for Hummingbird, mostly with relatively large confidence intervals (Figure 45). Figure 46, Figure 47 and Figure 48 show the model fit in detection function graphs for these narratives. Only in case of Oropendula the best model is one without covariate use (conventional DISTANCE sampling). In two cases adding plot related covariates results in the best model: Habitat for Flycatcher (model 26) and Duff cover % for Seedeater (model 39). For Hummingbird using the type of identification (aural/ visual) results in the best model fit (model 16), indicating that split of the data in two sets could be beneficial (resulting in much smaller population estimates of 302 or 662 individuals per km², compare chapter 3.3.5:

DISTANCE Sampling Results: Aural vs. Visual Bird Detections). Cluster size as a covariate is found to result in the best model fit for Woodpecker (model 11).

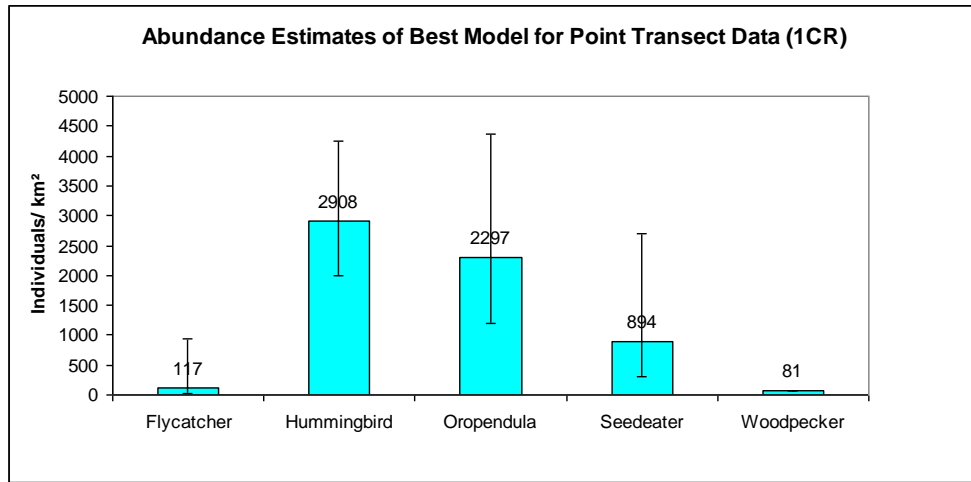


Figure 45: Abundance estimates and confidence intervals of best model for point transect data (1CR)

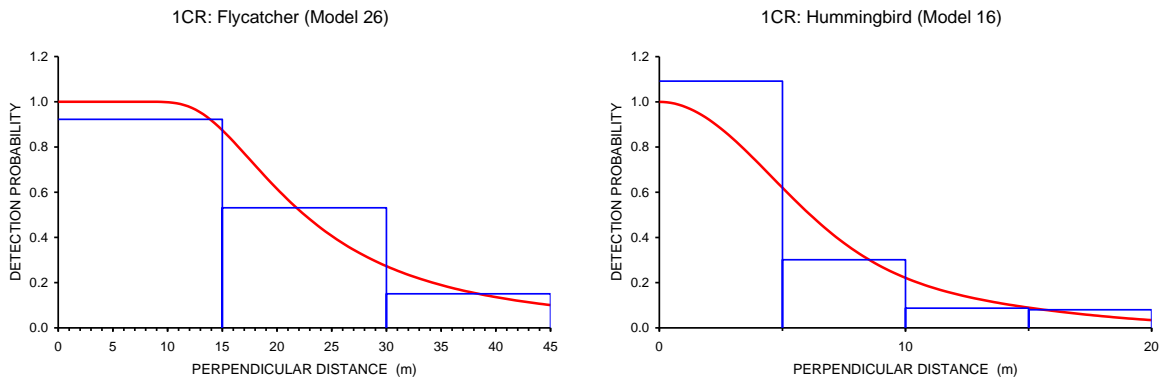


Figure 46: DISTANCE detection functions for Flycatcher and Hummingbird (1CR)

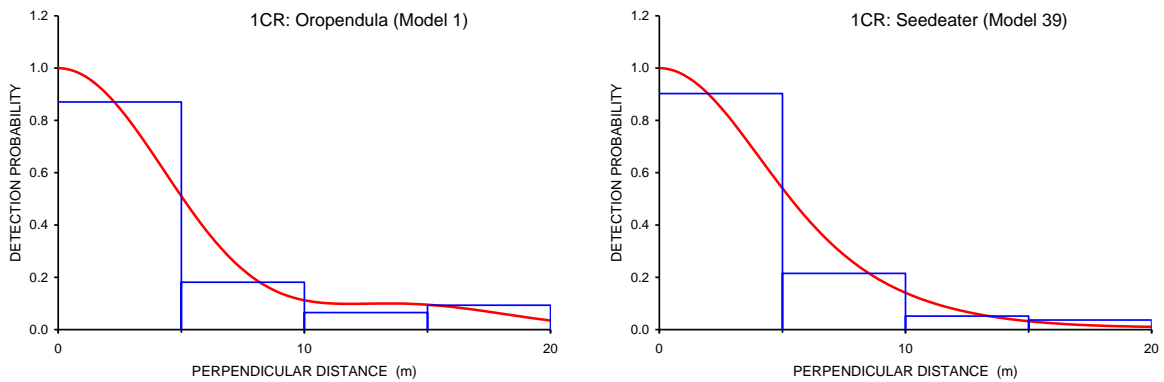


Figure 47: DISTANCE detection functions for Oropendula and Seedeater (1CR)

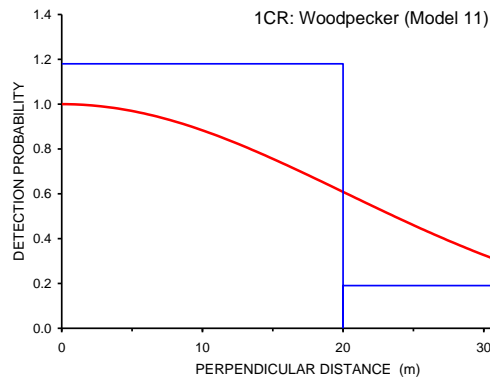


Figure 48: DISTANCE detection function for Woodpecker (1CR)

At study area 2Ni density is estimated for two narratives: White-throated Magpie Jay (152 individuals/ km²) and Banded Wren (160 individuals/ km²). Confidence intervals are relatively high, ranging from 93 to 277 individuals/ km² for Banded Wren and from 88 to 264 individuals/ km² for White-throated Magpie Jay (Figure 49). Both narratives receive best model fit adding Shrub cover % as covariate (Figure 50).

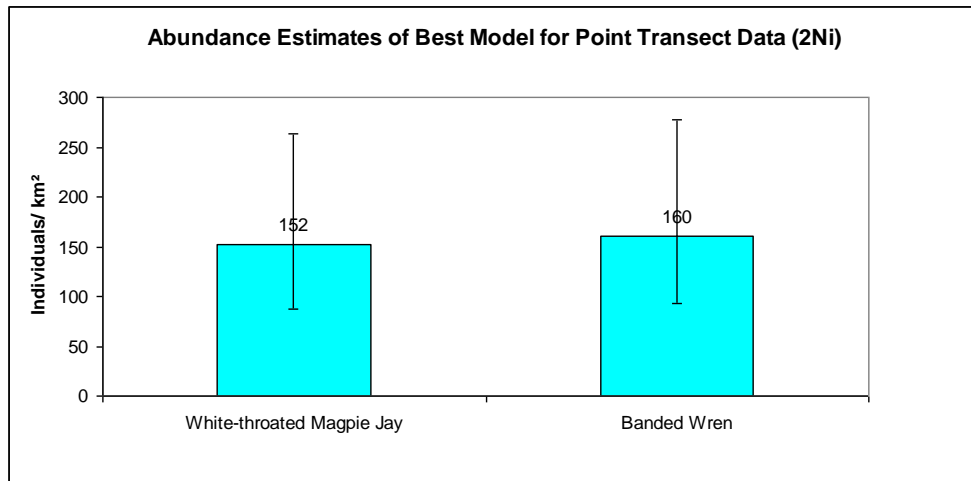


Figure 49: Abundance estimates and confidence intervals of best model for point transect data (2Ni)

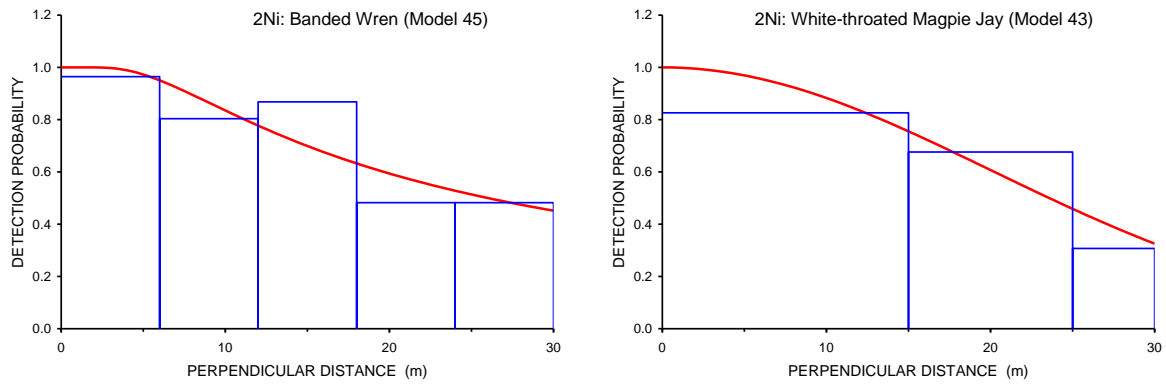


Figure 50: DISTANCE detection functions for Banded Wren and White-throated Magpie Jay (2Ni)

From study area 3AK densities for Sparrow and Squirrel are estimated (Figure 51). The Sparrow population is estimated to have 16 individuals/ km² with confidence interval ranging from 10 to 26 individuals/ km²; the Squirrel population has a very similar estimate of 17 individuals/ km² with confidence interval ranging from 11 to 26 individuals/ km². The best model fit for Sparrow is achieved using minutes since sunrise as covariate (model 12), while the best model fit for Squirrel uses Habitat type as covariate (model 15). Both detection functions are shown in Figure 52. In both cases there were no observations within 5 m of the observer and the number of observations was generally not decreasing smoothly with growing distance.

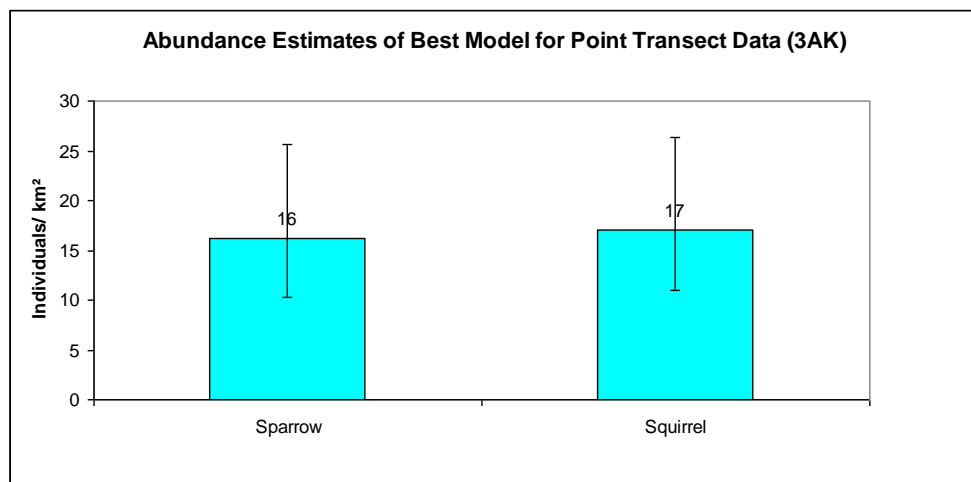


Figure 51: Abundance estimates and confidence intervals of best model for point transect data (3AK)

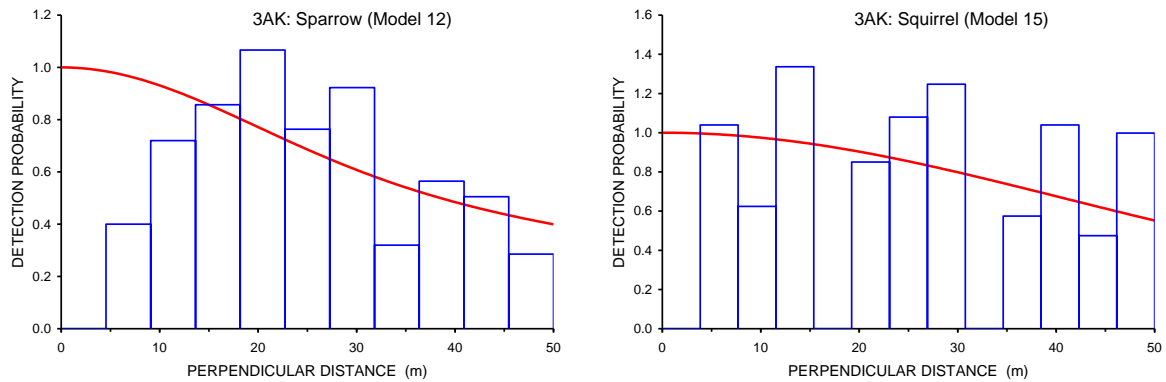


Figure 52: DISTANCE detection functions for Sparrow and Squirrel (3AK)

Density estimates for narratives from study area 4Ru range from 17 individuals/ km² for Winter Wren and Wize to 572 individuals/ km² for Chickadee (Figure 53). Confidence intervals are again relatively large; reaching up to almost 100 % (upper confidence interval for Chickadee is 1077 individuals/ km²). DISTANCE detection functions for the six narratives from 4Ru are shown in Figure 54, Figure 55 and Figure 56. Many of those graphs show problematic trends in the data, like the highest number of observations being at greater distance from the observer (Kinglet and Warbler) or like having no observations within 5 m of the observer (Wize). The latter could possibly be explained by the fact that Wize is an aural identification and most birds closer to the observer will usually be identified aurally as well as visually (it is unknown to which bird species the sound belongs). The model for Chickadee was best with conventional DISTANCE sampling; all other models had a better model fit using plot related covariates. These covariates were Number of flowers (model 43 for Kinglet and model 44 for Nutcracker), Habitat type (model 20 for Warbler), Lichen % (model 67 for Winter Wren), and Covariate 23 (model 111 for Wize). Covariate 23 at study area 4Ru is *Rhodococcum vitis-idaea* (see chapter 7.2: Covariates by Study Area).

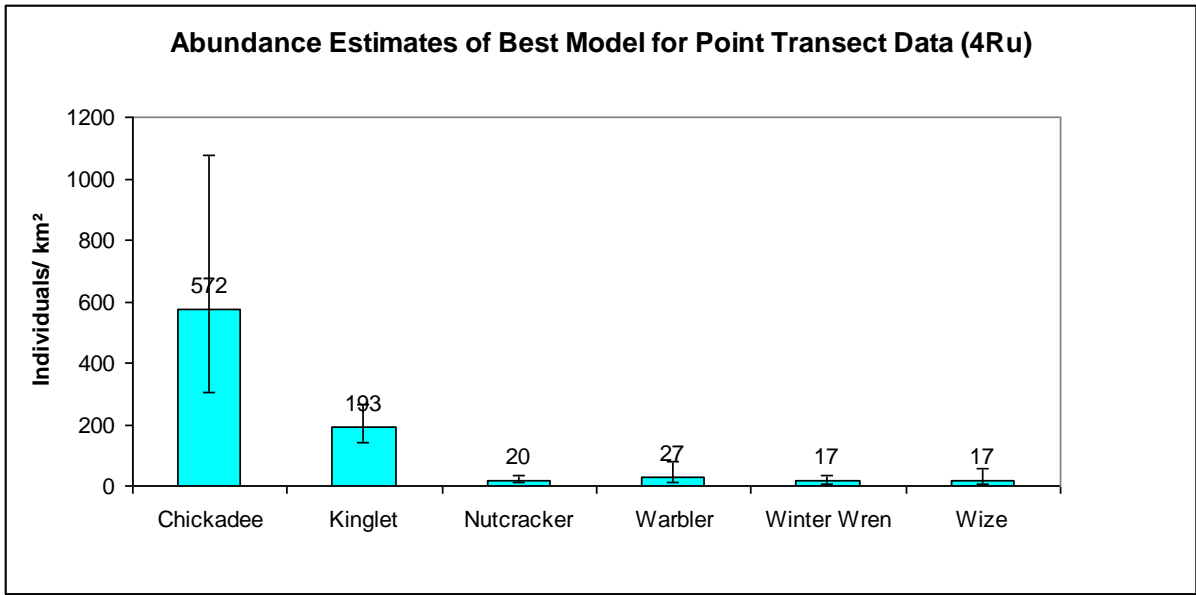


Figure 53: Abundance estimates and confidence intervals of best model for point transect data (4Ru)

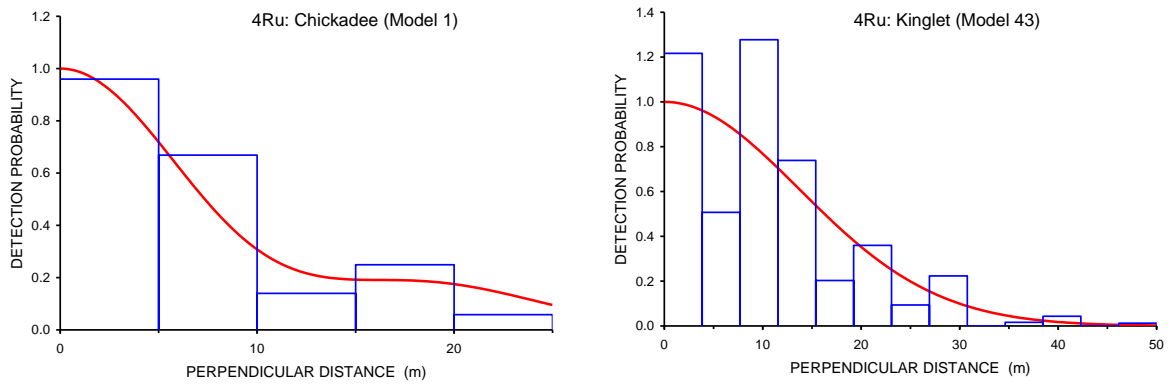


Figure 54: DISTANCE detection functions for Chickadee and Kinglet (4Ru)

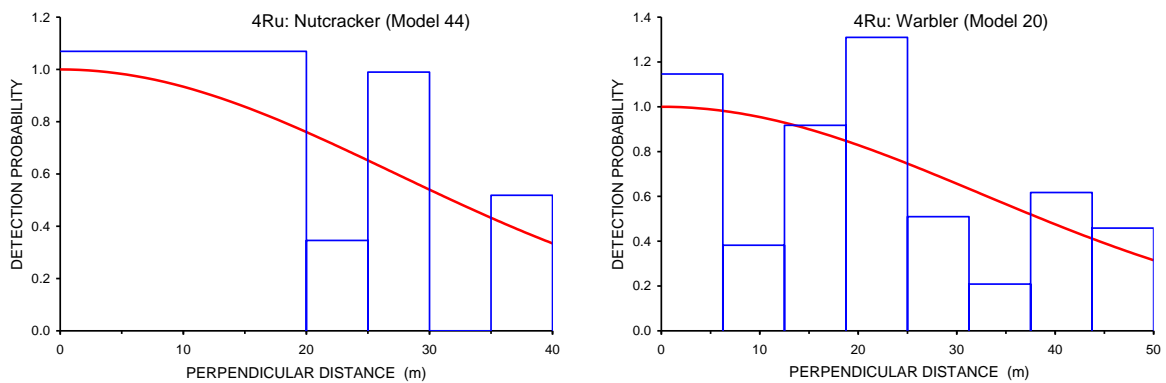


Figure 55: DISTANCE detection functions for Nutcracker and Warbler (4Ru)

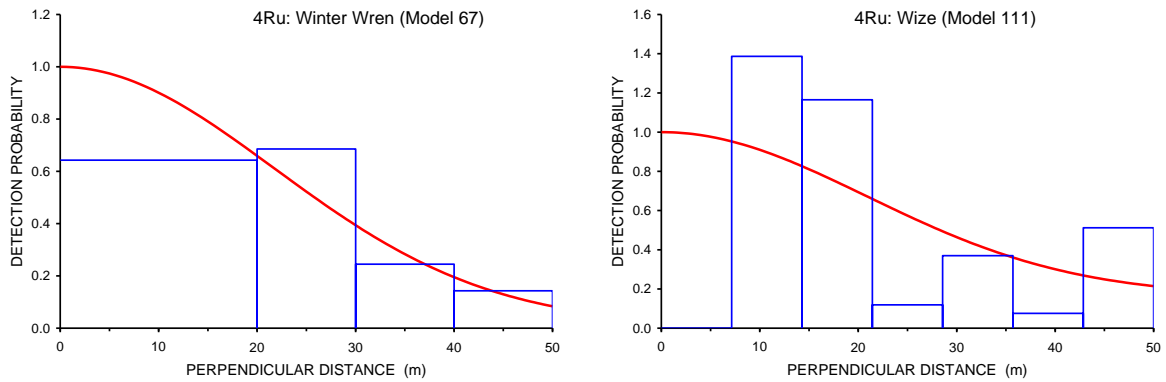


Figure 56: DISTANCE detection functions for Winter Wren and Wize (4Ru)

Abundance estimates are derived for Flute and Tsilp from study area 5PG, both being phonetic descriptions of bird songs (Figure 57). The best estimate for Flute is 17 individuals/ km² (confidence interval from 10 to 27 individuals/ km²), the best one for Tsilp is 67 individuals/ km² (confidence interval not available). The best detection function fit for Flute is achieved without covariates (model 1), while the best one for Tsilp is model 11 using Cluster size as a covariate (Figure 58).

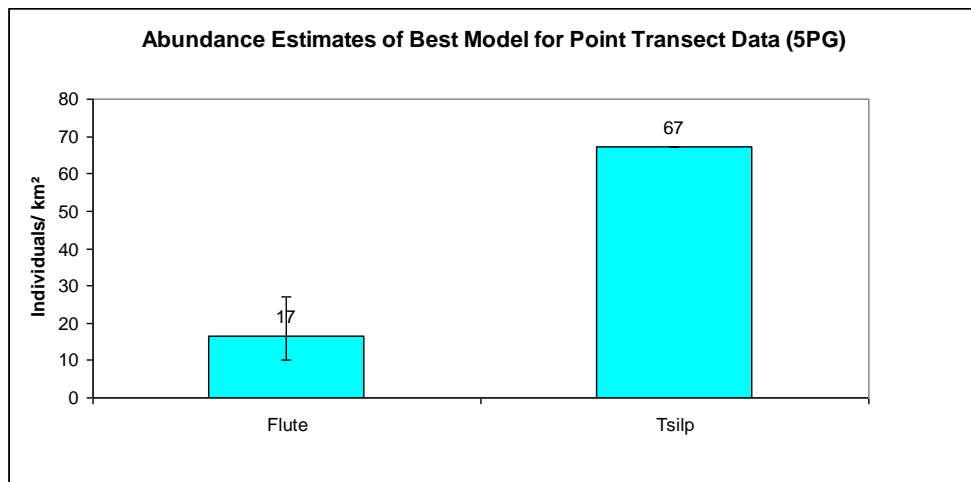


Figure 57: Abundance estimates and confidence intervals of best model for point transect data (5PG)

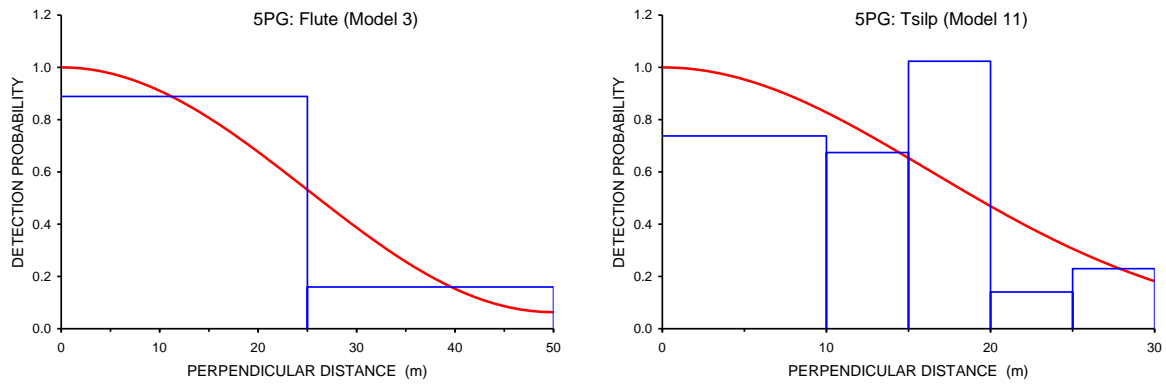


Figure 58: DISTANCE detection functions for Flute and Tsilp (5PG)

Densities between 6 and 70 individuals per km² are estimated for six narratives at study area 6Ba (Figure 59). Confidence intervals are relatively large, ranging up to four times the initial estimate (252 individuals/ km² as upper confidence interval for Pomarine Jaeger). For Lapland Bunting, Longbilled Dowitcher, Pectoral Sandpiper and Semipalmated Sandpiper the Diameter of the nearest lake is the covariate resulting in best model fit with MCDS (models 35-37, with detection functions as shown in Figure 60, Figure 61 and Figure 62). For Pomarine Jaeger model 18 using number of flowers as covariate has the best model fit, for Red Phalarope it is model 52 using detection/ non-detection of coltsfoot as covariate.

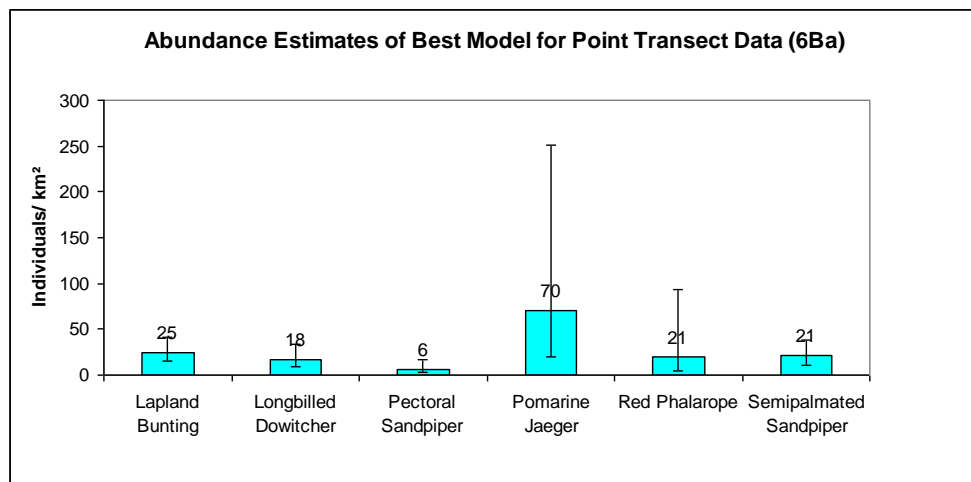


Figure 59: Abundance estimates and confidence intervals of best model for point transect data (6Ba)

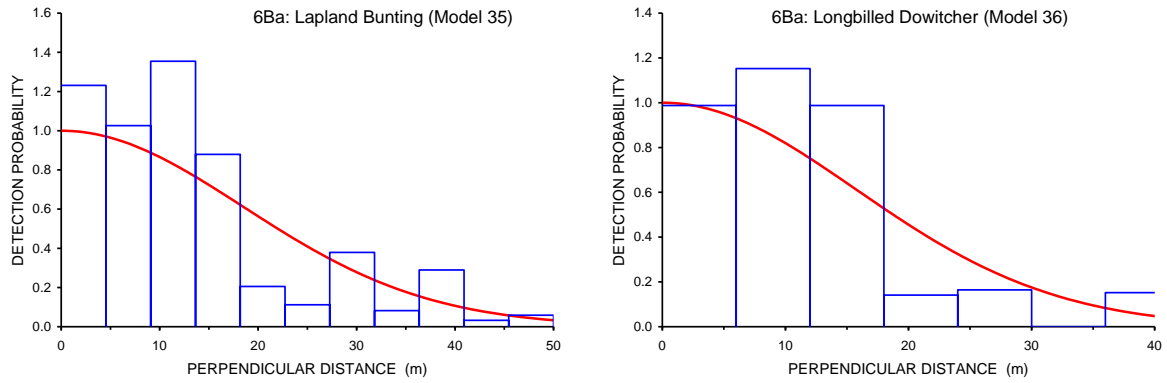


Figure 60: DISTANCE detection functions for Lapland Bunting and Longbilled Dowitcher (6Ba)

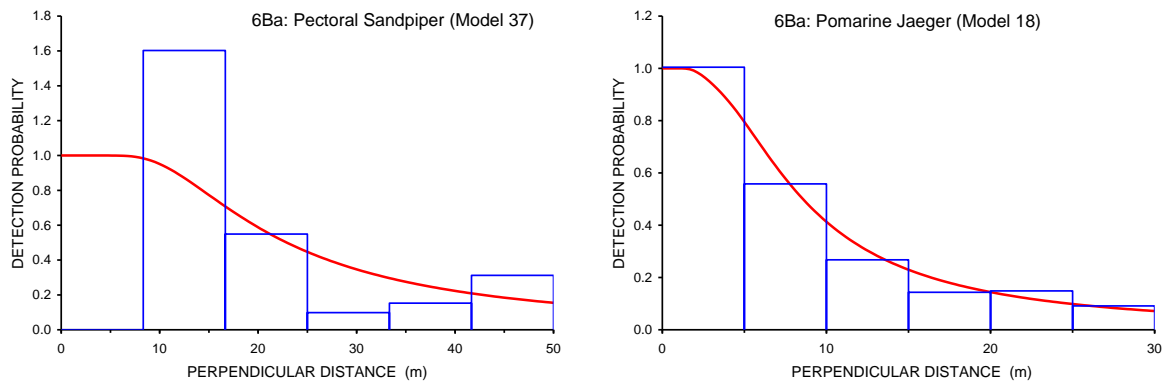


Figure 61: DISTANCE detection functions for Pectoral Sandpiper and Pomarine Jaeger (6Ba)

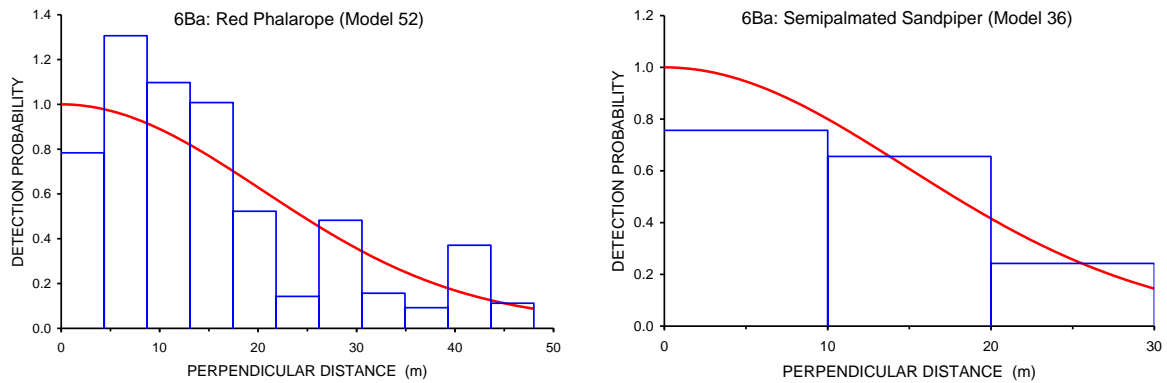


Figure 62: DISTANCE detection functions for Red Phalarope and Semipalmated Sandpiper (6Ba)

Table 15 gives an overview of density estimates and confidence intervals for all narratives. In most cases the relatively large range covered by the confidence interval indicates relatively low precision of the estimates. The last column adds an estimate of the narrative density per

GRID. For each study area the total number of birds per GRID is calculated. Bird totals range from 4 per GRID in 3AK to 1,574 per GRID in 1CR. This calculation disregards all observations which could not be analyzed with DISTANCE because of low sample size.

Table 15: Overview of density estimates and confidence intervals for point transect data

| Study area | Target Narrative | Density (individuals per km ²) | Lower confidence interval | Upper confidence interval | Density (individuals per GRID) |
|------------|---------------------------|--|---------------------------|---------------------------|--------------------------------|
| 1CR | Flycatcher | 117 | 15 | 930 | 29 |
| 1CR | Hummingbird | 2908 | 1992 | 4244 | 727 |
| 1CR | Oropendula | 2297 | 1209 | 4365 | 574 |
| 1CR | Seedeater | 894 | 296 | 2696 | 224 |
| 1CR | Woodpecker | 81 | 0 | 0 | 20 |
| 1CR | Bird Total: | 6297 | - | - | 1574 |
| 2Ni | Banded Wren | 160 | 93 | 277 | 40 |
| 2Ni | White-throated Magpie Jay | 152 | 88 | 264 | 38 |
| 2Ni | Bird Total: | 312 | - | - | 78 |
| 3AK | Sparrow | 16 | 10 | 26 | 4 |
| 3AK | Squirrel | 17 | 11 | 26 | 4 |
| 3AK | Bird Total: | 16 | 10- | 26 | 4 |
| 4Ru | Chickadee | 572 | 304 | 1077 | 143 |
| 4Ru | Kinglet | 193 | 140 | 266 | 48 |
| 4Ru | Nutcracker | 20 | 12 | 33 | 5 |
| 4Ru | Warbler | 27 | 9 | 79 | 7 |
| 4Ru | Winter Wren | 17 | 8 | 34 | 4 |
| 4Ru | Wize | 17 | 5 | 58 | 4 |
| 4Ru | Bird Total: | 846 | - | - | 212 |
| 5PG | Flute | 17 | 10 | 27 | 4 |
| 5PG | Tsilp | 67 | 0 | 0 | 17 |
| 5PG | Bird Total: | 84 | - | - | 21 |
| 6Ba | Lapland Bunting | 25 | 15 | 41 | 6 |
| 6Ba | Longbilled Dowitcher | 18 | 9 | 33 | 5 |
| 6Ba | Pectoral Sandpiper | 6 | 2 | 17 | 2 |
| 6Ba | Pomarine Jaeger | 70 | 19 | 252 | 18 |
| 6Ba | Red Phalarope | 21 | 5 | 93 | 5 |
| 6Ba | Semipalmated Sandpiper | 21 | 11 | 39 | 5 |
| 6Ba | Bird Total: | 161 | - | - | 40 |

3.3.2 DISTANCE Sampling Results: Trapping Web Catches

From study area 1CR abundance estimates for Ant and Spider are calculated (Figure 63). The estimate for ant is 2,741 individuals/ km², while the confidence interval ranges from 656 to 11,448 individuals/ km². The estimate for spider is 702 individuals with a confidence interval

from 403 to 1,225 individuals/ km². Both detection functions are shown in Figure 64. For both narratives model 32 with Shrubs % resulted in best model fit.

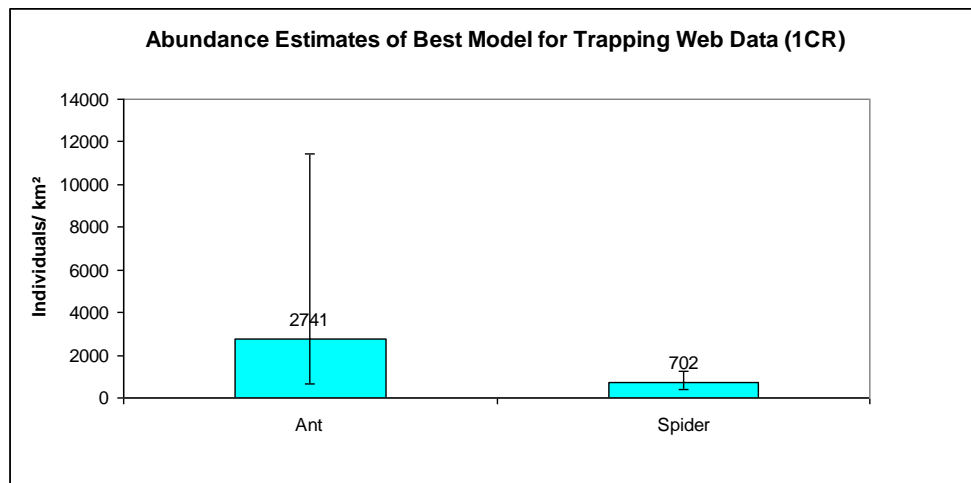


Figure 63: Abundance estimates and confidence intervals of best model for trapping web data (1CR)

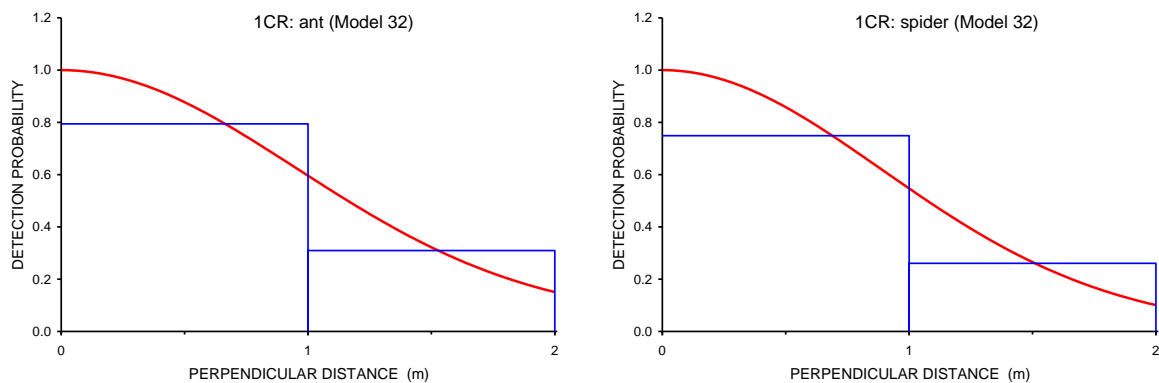


Figure 64: DISTANCE detection functions for Ant and Spider (1CR)

Density estimations at 2Ni range from 271 individuals/ km² for Spider, small to 2,091 individuals/ km² for ant (Figure 65). The estimated density of Springtail was 47,207 individuals/ km², which made the use of a second scale on the right side of the graph necessary. Confidence intervals have a relatively large range, for springtail for example the lower confidence interval is 25,775 individuals/ km² and the upper confidence interval is 86,463 individuals/ km². Five out of the six narratives reach best model fit with the standard models 1-6, without use of covariates (from Figure 66 to Figure 68). Adding a covariate increased model fit only for ant, for which Habitat was used as covariate in MCDS analysis.

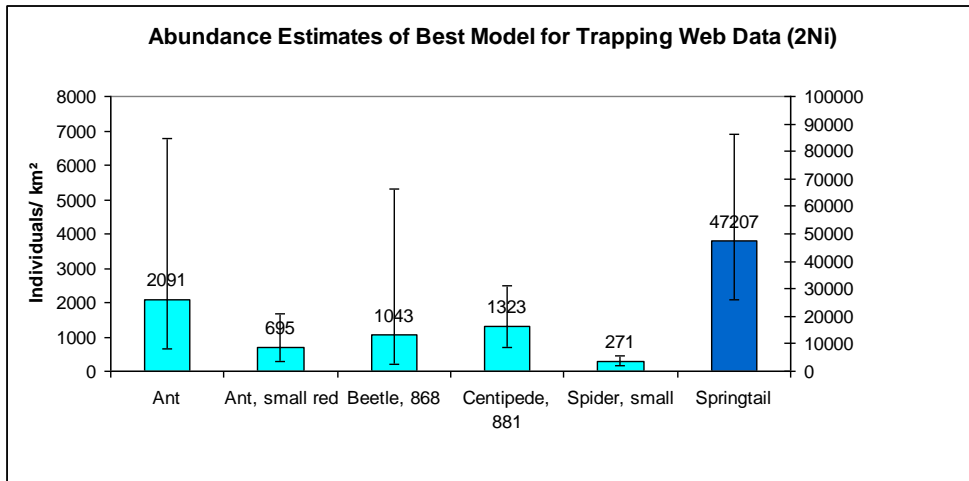


Figure 65: Abundance estimates and confidence intervals of best model for trapping web data (2Ni)

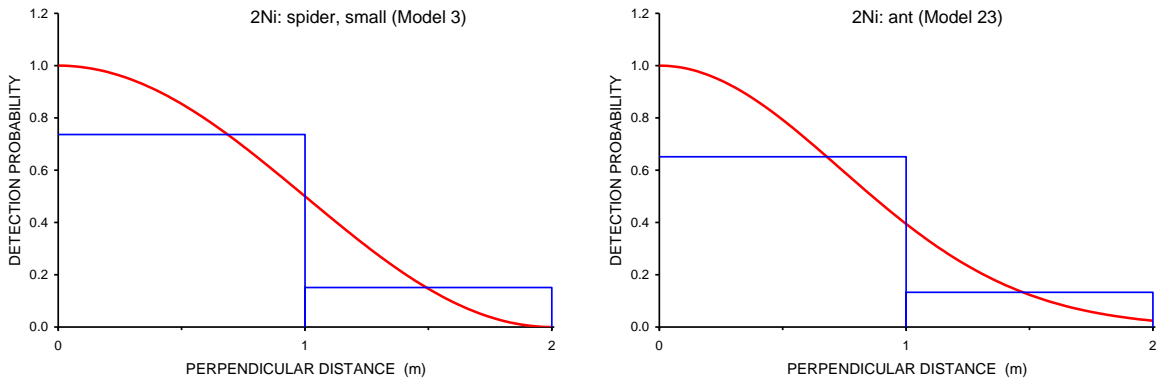


Figure 66: DISTANCE detection functions for Spider, small and Ant (2Ni)

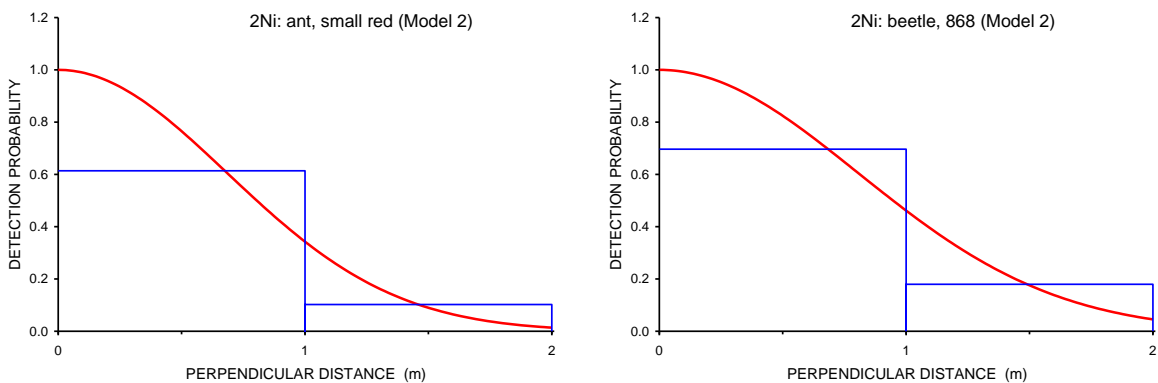


Figure 67: DISTANCE detection functions for Ant, small red and Beetle, 868 (2Ni)

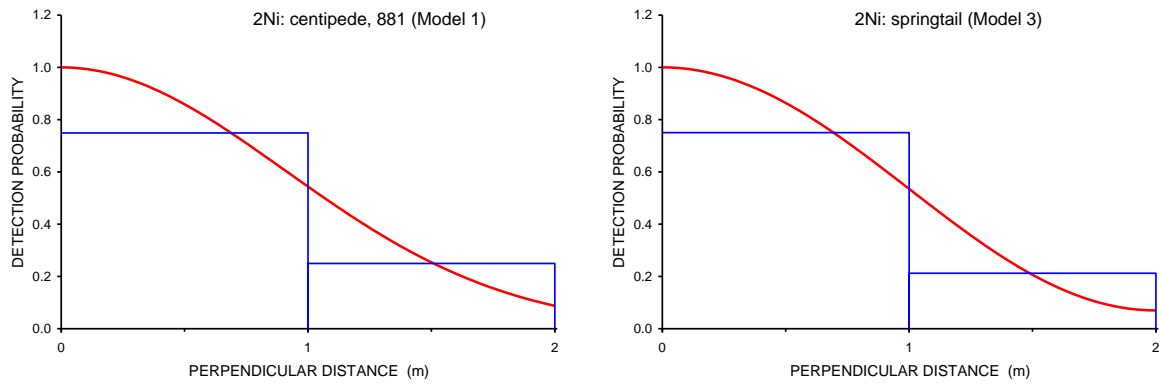


Figure 68: DISTANCE detection functions for Centipede, 881 and Springtail (2Ni)

Sufficient trapping web data for DISTANCE analysis at study area 3AK was collected for Spider and Springtail (Figure 69). Density estimate for Spider is 970 individuals per km² with a confidence interval range from 476 to 1,976 individuals/ km². Density estimate for Springtail is 39,238 individuals/ km² with a confidence interval range from 7,950 to 193,674 individuals/ km². For both narratives the best model fit is achieved with MCDS analysis, best model fits are shown in Figure 70. Model 20 used for Spider has habitat type as covariate, model 71 for springtail uses Covariate 18 for study area 3AK as a covariate (Covariate 18 refers to an unidentified plant species, detailed pictures are available in the digital appendix).

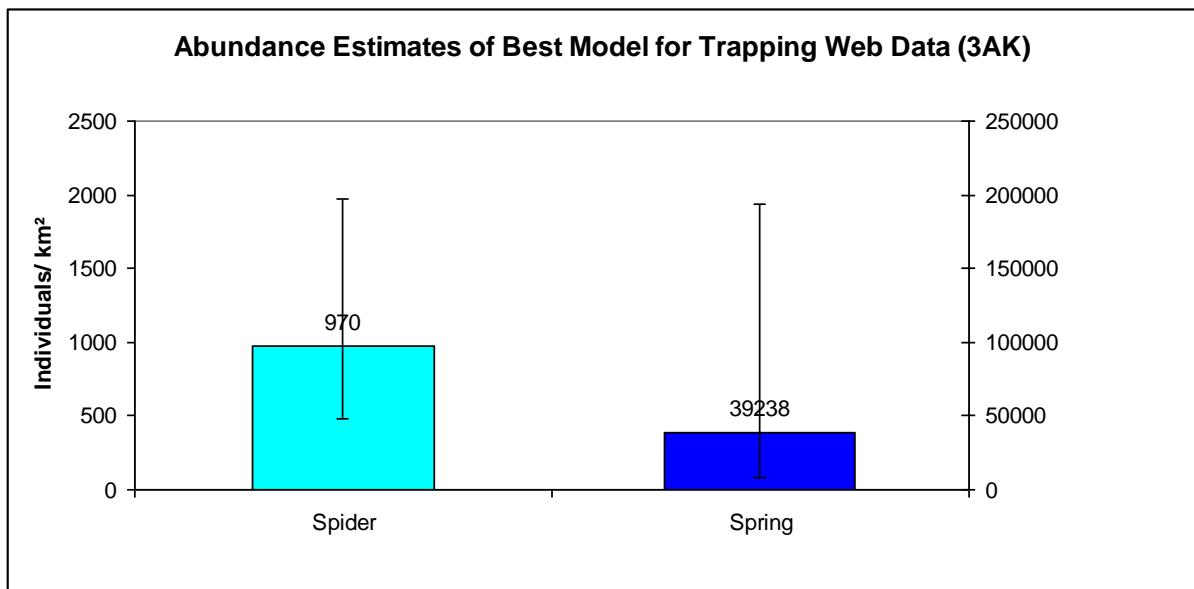


Figure 69: Abundance estimates and confidence intervals of best model for trapping web data (3AK)

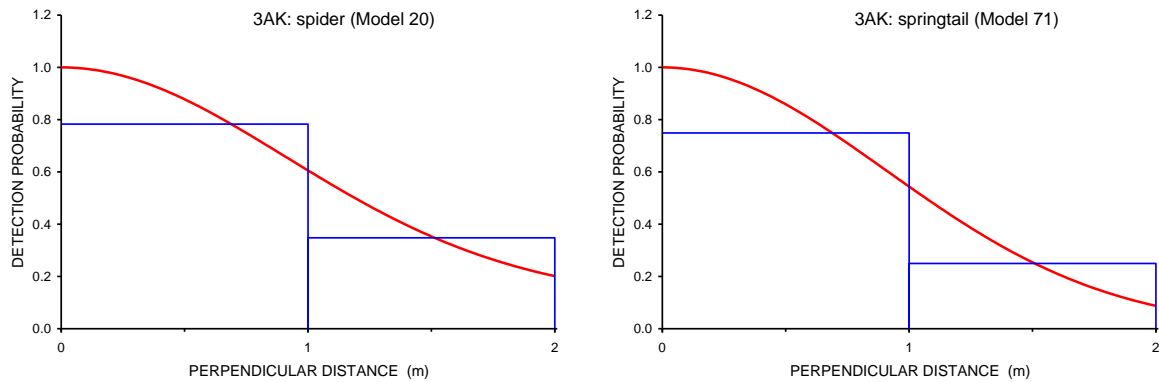


Figure 70: DISTANCE detection functions for Spider and Springtail (3AK)

Abundance at 4Ru is estimated for Cycsegusa, Protura, and Spider, little (Figure 71). Estimates are relatively close together and reach from 163 individuals/ km² for Protura to 281 individuals/ km² for Cycsegusa. Upper confidence interval is up to more than four times the estimate (1,171 individuals/ km² for Cycsegusa). Best model fits for Cycsegusa and Spider little are achieved without use of covariates (model 1 respectively model 2), as shown in Figure 72 and Figure 73. For the analysis of Protura adding *Betula ermanii* as a covariate resulted in best model fit (model 44).

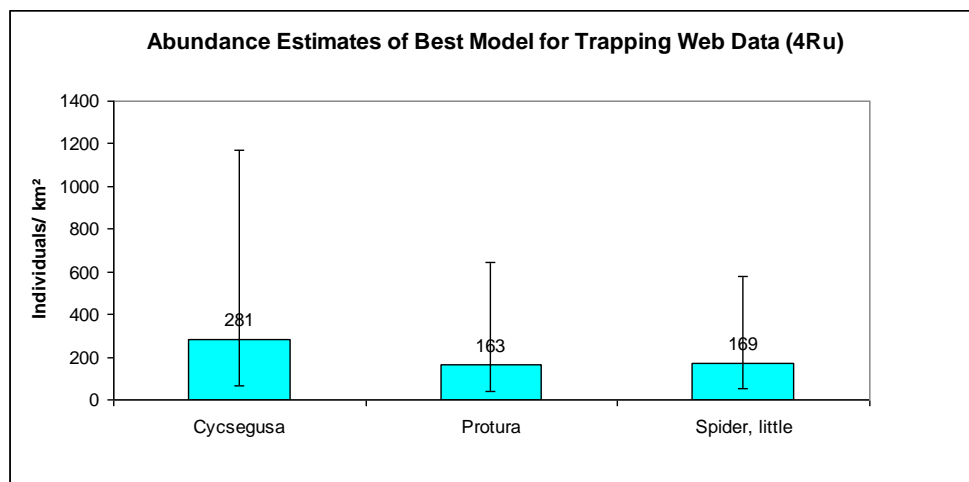


Figure 71: Abundance estimates and confidence intervals of best model for trapping web data (4Ru)

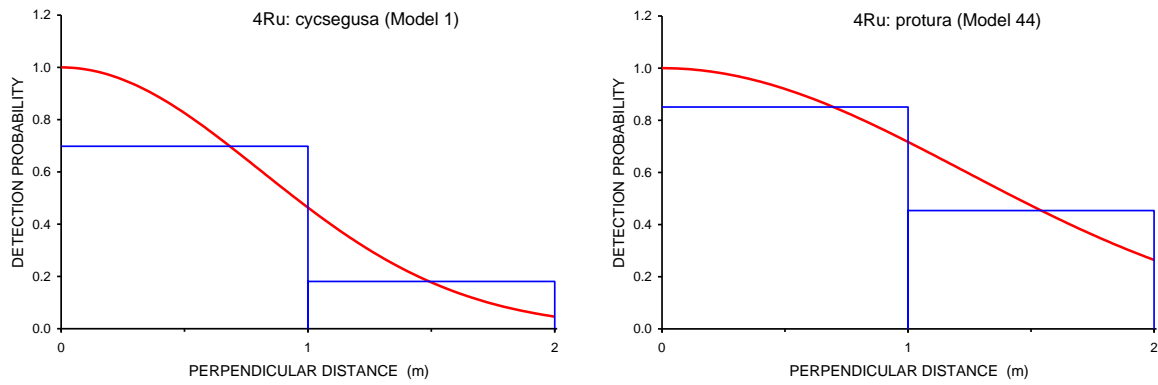


Figure 72: DISTANCE detection functions for Cycsegusa and Protura (4Ru)

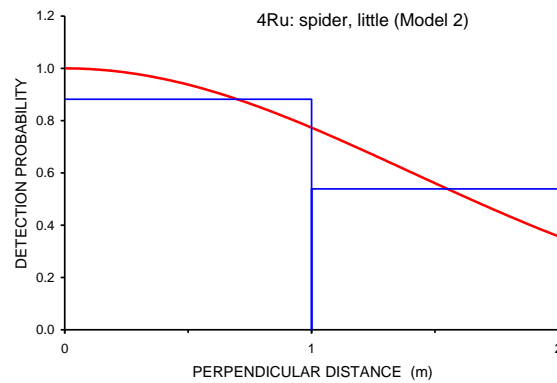


Figure 73: DISTANCE detection function for Spider, little (4Ru)

At 5PG the data allow to calculate density estimates only for tiny black Ant (Figure 74). The actual estimate is 199 individuals/ km²; confidence interval covers a range from 67 to 592 individuals/ km². The best model fit is shown in Figure 75 (model 3, without covariate use).

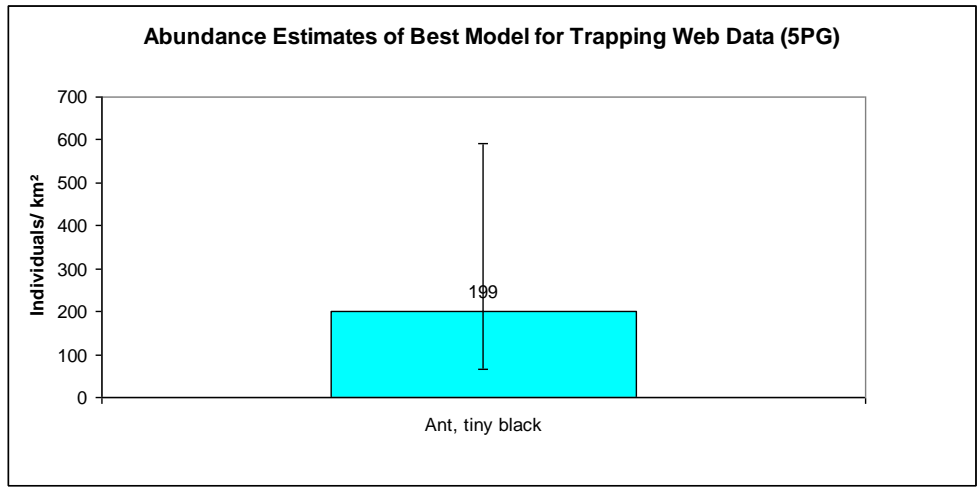


Figure 74: Abundance estimates and confidence intervals of best model for trapping web data (5PG)

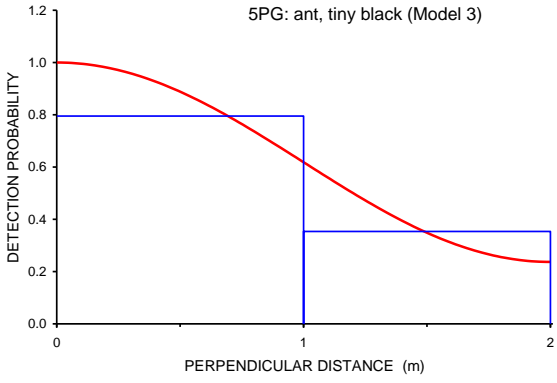


Figure 75: DISTANCE detection functions for Ant, tiny black (5PG)

Figure 76 shows abundance estimates and confidence intervals for eight narratives from 6Ba. Because of the comparably high estimates for flat Beetle and tiny Spider a different scaling is used to display results for these two. DISTANCE detection functions are displayed from Figure 77 to Figure 80. For seven narratives the best model fit is achieved without covariate use, only the model for Spider gained from adding the covariate Cluster size.

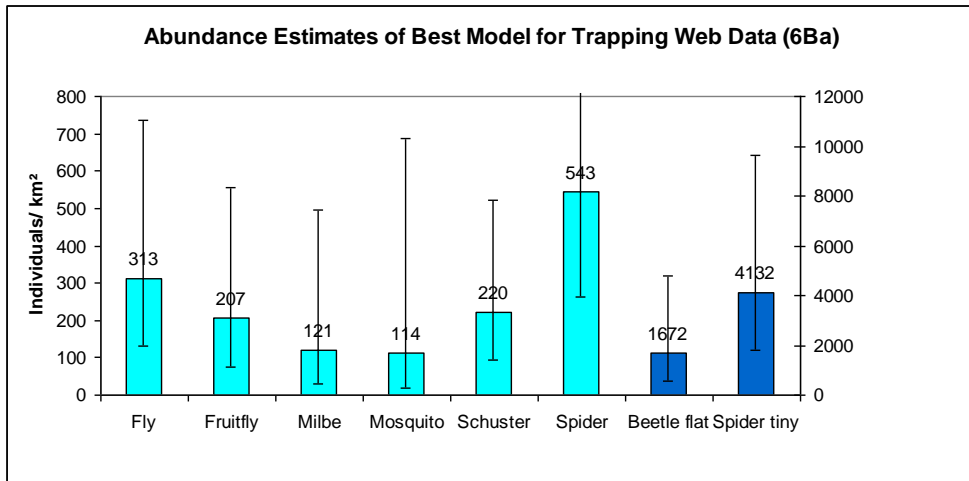


Figure 76: Abundance estimates and confidence intervals of best model for trapping web data (6Ba)

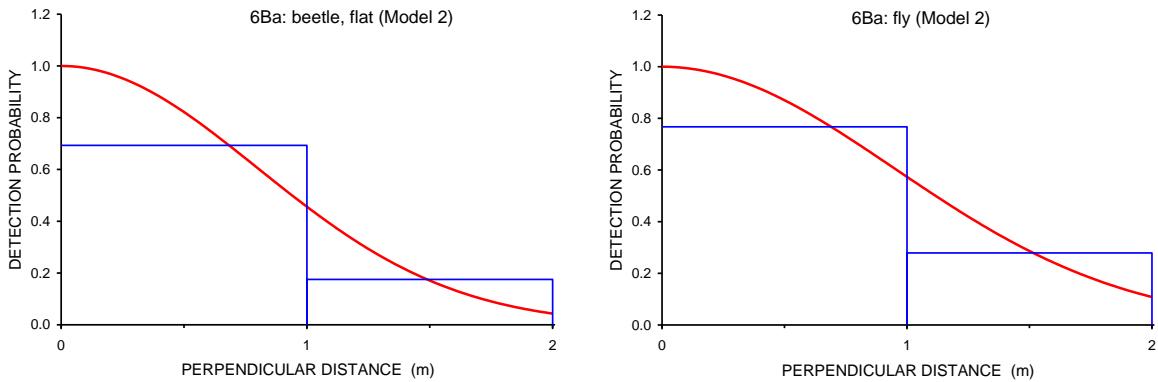


Figure 77: DISTANCE detection functions for Beetle, flat and Fly (6Ba)

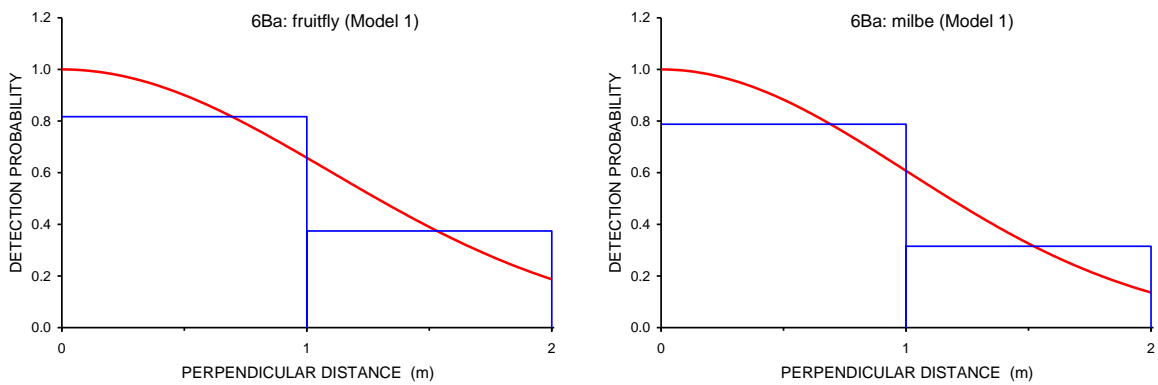


Figure 78: DISTANCE detection functions for Fruitfly and Milbe (6Ba)

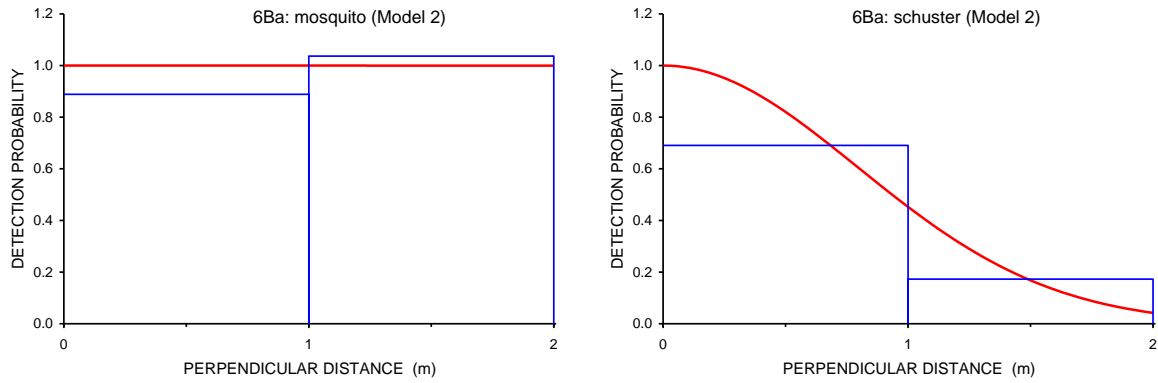


Figure 79: DISTANCE detection functions for Mosquito and Schuster (6Ba)

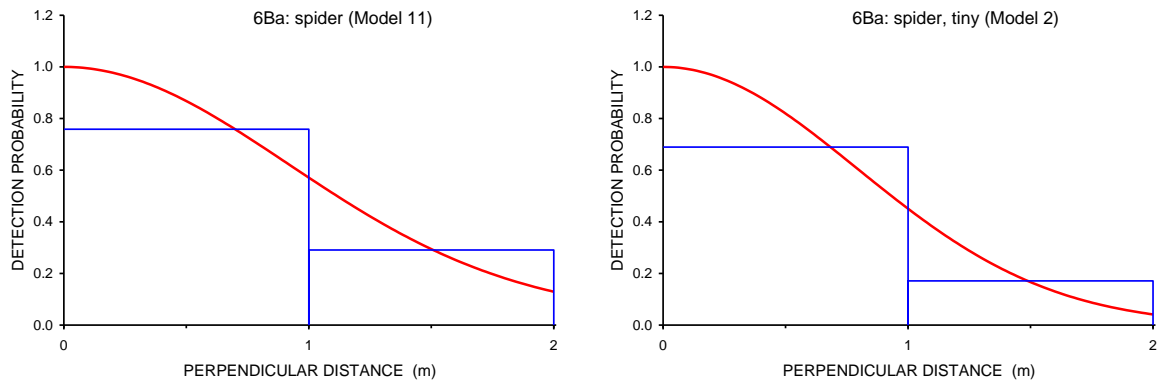


Figure 80: DISTANCE detection functions for Spider and Spider, tiny (6Ba)

Density estimates per km² and per GRID as well as confidence intervals for all narratives are summarized in Table 16. In most cases the range of the confidence interval is relatively large compared to the original estimates. For each study area the total number of arthropods per GRID is calculated. This number can be seen as a simple estimate of arthropod biomass, although it is limited because mean mass per animal is unknown. Arthropod totals range from 50 per GRID in 5PG to 13,158 per GRID in 1CR. This calculation disregards all observations which could not be analyzed with DISTANCE because of low sample size.

Table 16: Overview of density estimates and confidence intervals for trapping web data

| Study area | Target Narrative | Density (individuals per km ²) | Lower confidence interval | Upper confidence interval | Density (individuals per GRID) |
|------------|--------------------------|--|---------------------------|---------------------------|--------------------------------|
| 1CR | Ant | 2741 | 656 | 11448 | 685 |
| 1CR | Spider | 702 | 403 | 1225 | 176 |
| 1CR | Arthropods Total: | 3443 | - | - | 861 |
| 2Ni | Spider, small | 271 | 170 | 432 | 68 |
| 2Ni | Ant | 2091 | 644 | 6787 | 523 |
| 2Ni | Ant, small red | 695 | 289 | 1673 | 174 |
| 2Ni | Beetle, 868 | 1043 | 206 | 5290 | 261 |
| 2Ni | Centipede, 881 | 1323 | 700 | 2501 | 331 |
| 2Ni | Springtail | 47207 | 25775 | 86463 | 11802 |
| 2Ni | Arthropods Total: | 52630 | - | - | 13158 |
| 3AK | Spider | 970 | 476 | 1976 | 243 |
| 3AK | springtail | 39238 | 7950 | 193674 | 9810 |
| 3AK | Arthropods Total: | 40208 | - | - | 10052 |
| 4Ru | Cycsegusa | 281 | 67 | 1171 | 70 |
| 4Ru | Protura | 163 | 41 | 642 | 41 |
| 4Ru | Spider, little | 169 | 50 | 578 | 42 |
| 4Ru | Arthropods Total: | 613 | - | - | 153 |
| 5PG | Ant, tiny black | 199 | 67 | 592 | 50 |
| 5PG | Arthropods Total: | 199 | 67 | 592 | 50 |
| 6Ba | Beetle, flat | 1672 | 584 | 4781 | 418 |
| 6Ba | Fly | 313 | 133 | 738 | 78 |
| 6Ba | Fruitfly | 207 | 77 | 556 | 52 |
| 6Ba | Milbe | 121 | 29 | 497 | 30 |
| 6Ba | Mosquito | 114 | 19 | 686 | 29 |
| 6Ba | Schuster | 220 | 93 | 522 | 55 |
| 6Ba | Spider | 543 | 263 | 1119 | 136 |
| 6Ba | Spider, tiny | 4132 | 1776 | 9614 | 1033 |
| 6Ba | Arthropods Total: | 7322 | - | - | 1831 |

3.3.3 DISTANCE Sampling Results: Line Transect Counts

Enough data to model abundance with DISTANCE was collected for only one narrative through the line transects add-on protocol: white Butterfly at 2Ni. The abundance estimate of 98,778 individuals/ km² was high compared to other insects from trapping web data (Figure 81). Model 4 without covariate use showed the best fit (Figure 82).

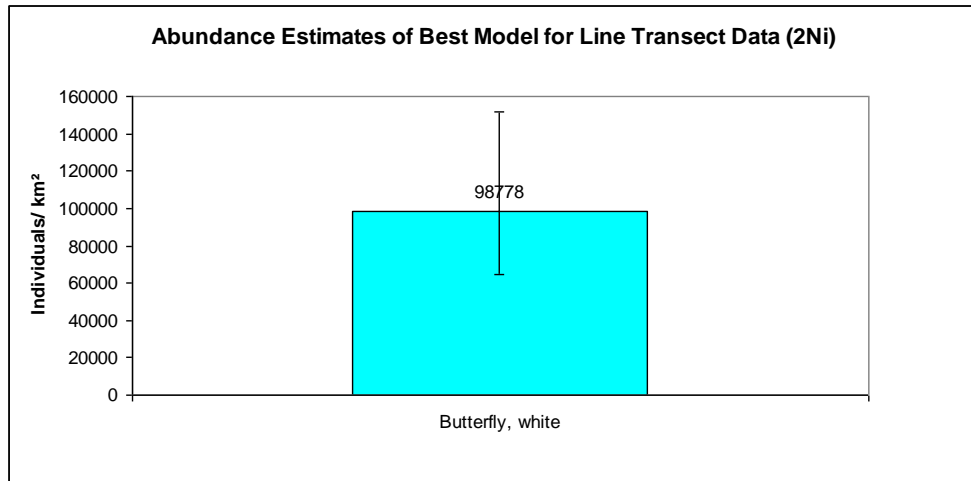


Figure 81: Abundance estimates and confidence intervals of best model for line transect data (2Ni)

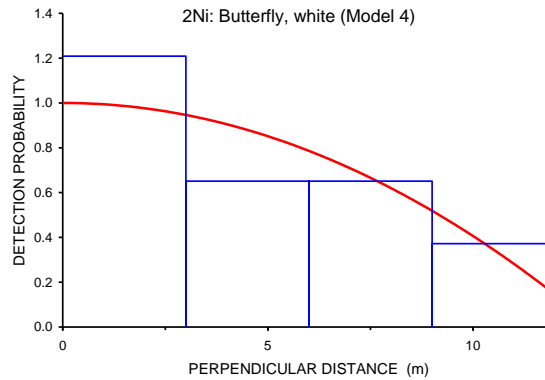


Figure 82: DISTANCE detection function for Butterfly, white (2Ni)

3.3.4 DISTANCE Sampling Results: Randomly vs. Systematically Selected Plots

DISTANCE analysis in above chapters uses pooled data sets under the assumption that true densities are relatively constant between randomly and systematically selected plots. Figure 83, Figure 84 and Figure 85 show comparisons of density estimates for point transect data; for clarity the confidence intervals are not shown in the graphs of this section, but they are available from the digital appendix. Keeping in mind the high confidence intervals shown in the former section some variance between the three data sets (all, ran, sys) is expected. The tendency is that narratives from 1CR and 2Ni have considerably higher estimates for the pooled data set than for the other two. The exception is the Hummingbird estimate, where the

random data set as well as the pooled data set results in extremely high estimates compared to the systematic data set. Estimates from 3AK seem to be relatively balanced.

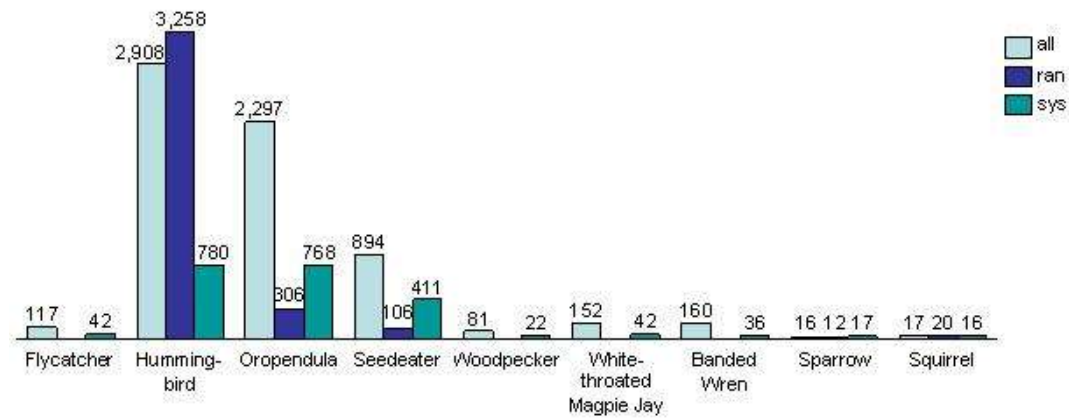


Figure 83: Comparison of abundance estimates for point transect data from random and systematic plots (1CR-3AK)

Abundance estimates for 4Ru and 5PG are also relatively balanced (Figure 84). The most obvious exception is Chickadee at 4Ru, where the estimate based on the random data set is lower compared to the other two.

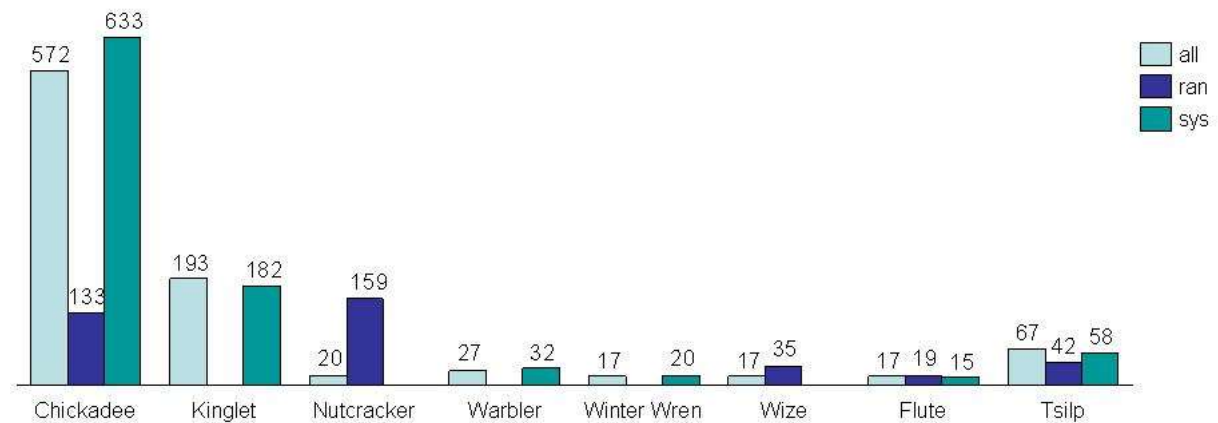


Figure 84: Comparison of abundance estimates for point transect data from random and systematic plots (4Ru-5PG)

Figure 85 shows estimates of best models for narratives from study area 6Ba. All of them seem to be quite balanced; a larger variance in estimates from the random data set is expected because of the considerably lower sample size.

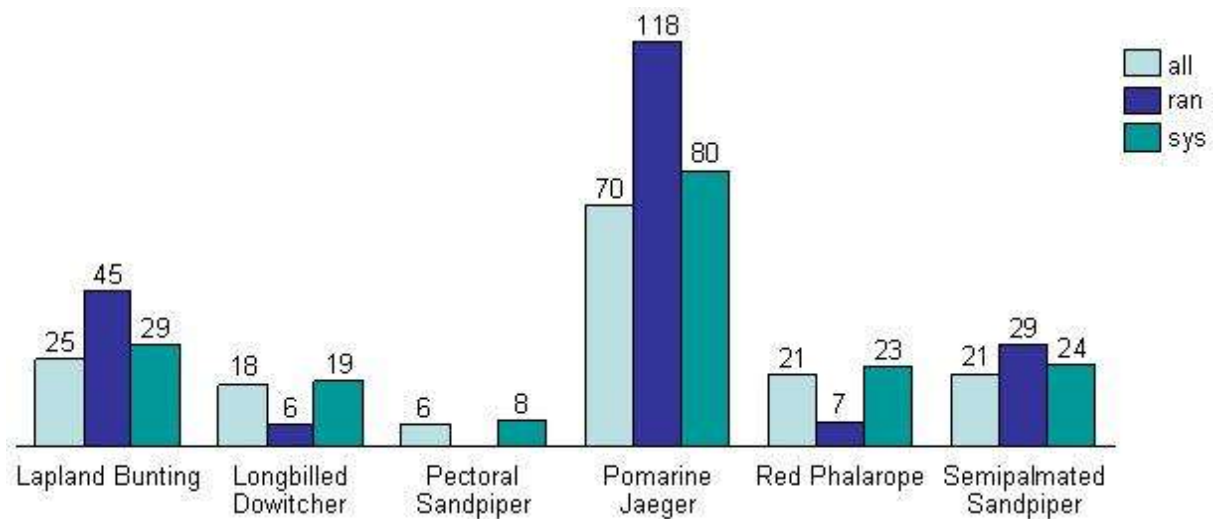


Figure 85: Comparison of abundance estimates for point transect data from random and systematic plots (6Ba)

The only estimate for line transect data shows almost identical values for the systematic and the pooled data set, while the random data set results in a much lower estimate (Figure 86). Trapping web data has not been analyzed this way because it was only collected at systematically selected plots.

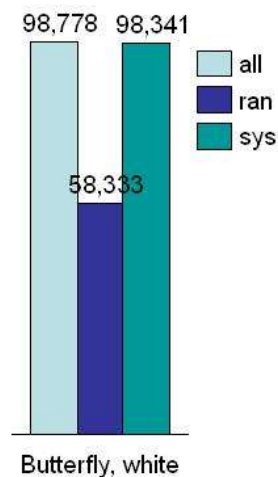


Figure 86: Comparison of abundance estimates for line transect data from random and systematic plots (2Ni)

3.3.5 DISTANCE Sampling Results: Aural vs. Visual Bird Detections

In this section the best model results for pooled data set from point transects at the first five study areas are compared with the ones for aural and visual data sets to check validity of pooling. The results are very different from narrative to narrative, as are the shares of visual and aural detections (compare Figure 36 and Figure 38 above). Figure 87 shows results for the different data sets from 1CR, 2Ni and 3AK. Some of the narratives have relatively close estimates from random and systematic data sets, but much higher values for the pooled estimate (Flycatcher, Hummingbird, White-throated Magpie Jay, Banded Wren). The same applies for Chickadee from 4Ru (Figure 88). This is an indicator that in these cases the pooling may result in too high estimates.

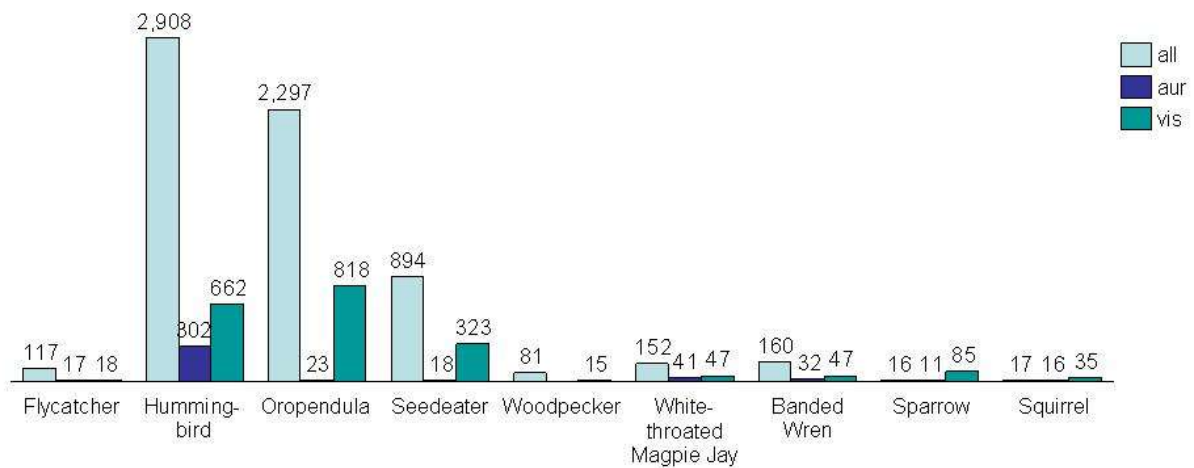


Figure 87: Comparison of abundance estimates for point transect data from aural and visual observations (1CR-3AK)



Figure 88: Comparison of abundance estimates for point transect data from aural and visual observations (4Ru-5PG)

3.3.6 DISTANCE Sampling Results: Biological Family and Order

Figure 89 and Figure 90 show estimates for data pooled by biological order, separately for point count data and trapping web data. In both graphs two different scaling are used. Whenever possible the range of confidence interval is indicated. The larger sample size generated through pooling of data does not result in smaller confidence intervals compared to single narrative analysis. Estimates for the same biological order can be compared between study regions. Density for Ciconniiformes for example is relatively similar at 2Ni and 6Ba, despite the first being close to the equator and the second being at the northernmost point of the American continent. Density for Passeriformes for example is below 100 individuals/ km² at 3AK, 6Ba and 5PG, but reaches 1,000 individuals/ km² and higher at 2Ni, 4Ru and 1CR. Explanations for these differences are not readily available because both groups, the one with low estimates and the one with high estimates, include study areas from arctic as well as tropical zones.

The only order analyzed from line transect data was Lepidoptera from 2Ni with an estimated density of 228,394 individuals/ km² (confidence interval between 161,618 and 322,760 individuals/ km²).

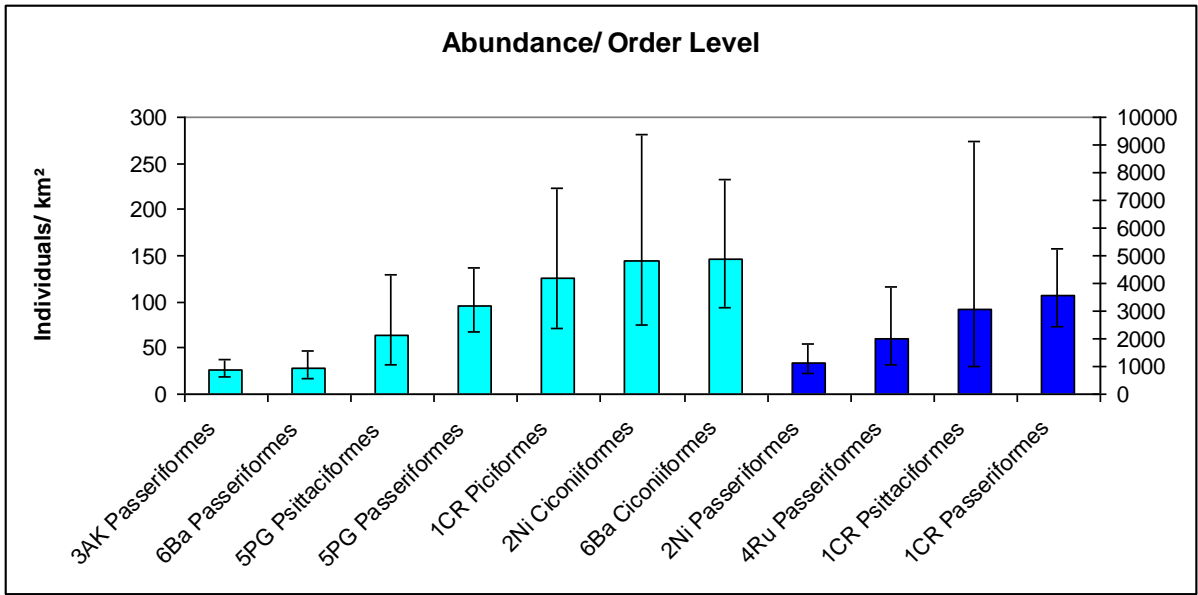


Figure 89: Abundance estimates for point transect data at biological order level

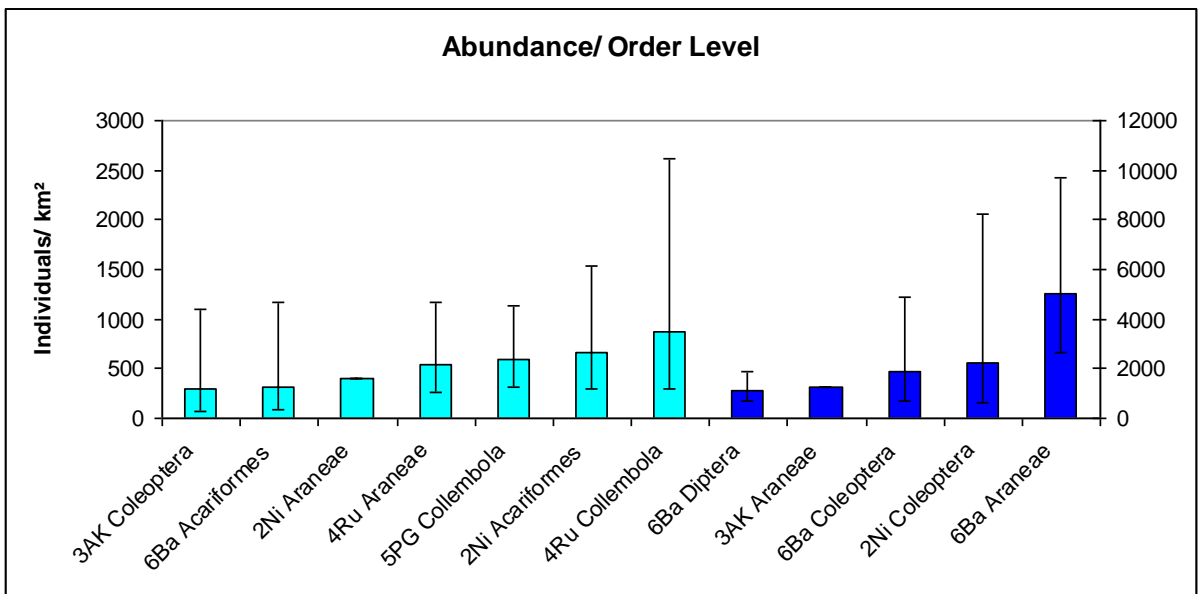


Figure 90: Abundance estimates for trapping web data at biological order level

The same information as above is shown by Figure 91 and Figure 92 for data pooled at the biological family level. Again the confidence interval range is so large that it is safe to assume that pooling does not result in precision gain. Biological family is more specific than order, so aside from Formicidae there are no biological families which are estimated at different study regions. The most interesting point about Formicidae might be that at study area 5PG the abundance estimate of tiny black Ant looks very small (199 individuals/ km²), but the overall ant density is much larger and corrects this first impression (950 individuals/ km²). However,

reasons for the dominance of particular ant species can not be found without a much more detailed survey.

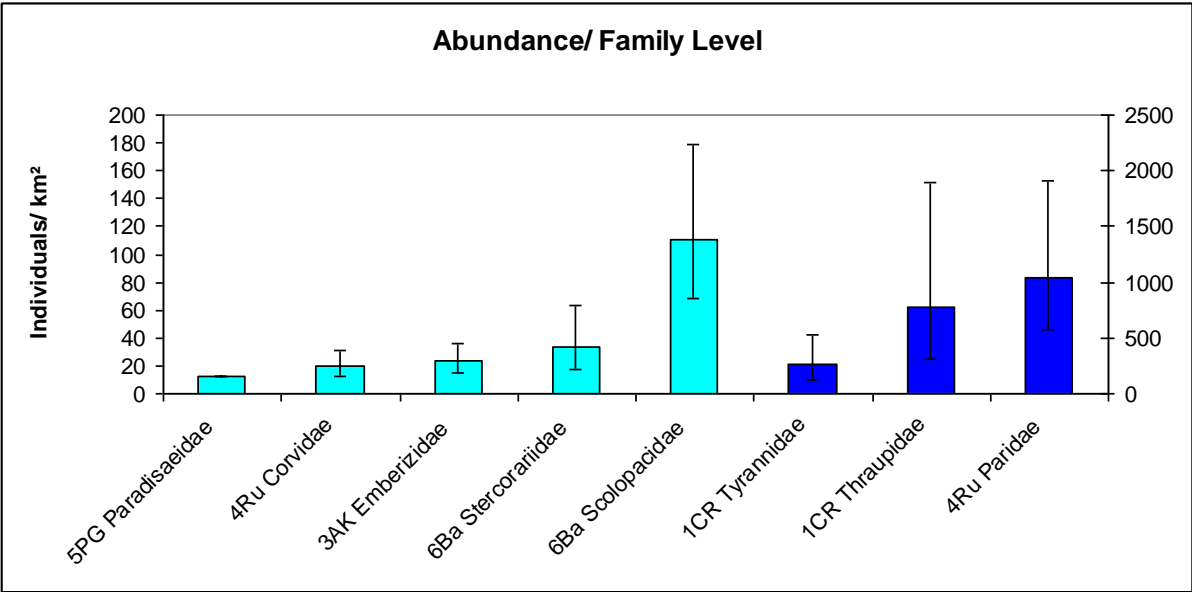


Figure 91: Abundance estimates for point transect data at biological family level

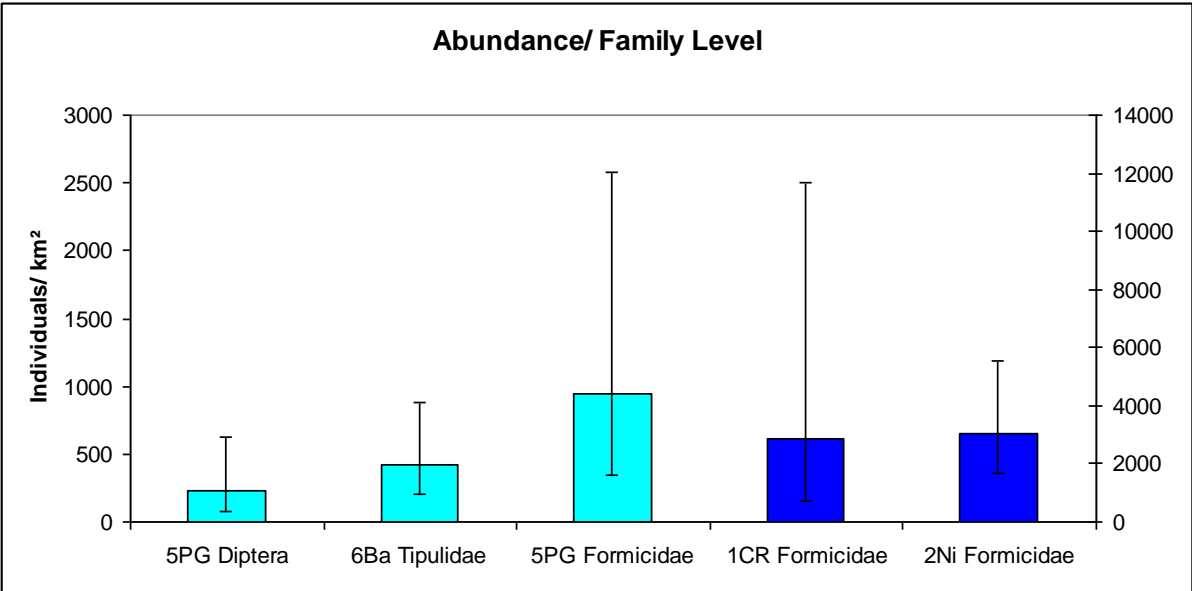


Figure 92: Abundance estimates for trapping web data at biological family level

3.4 PRESENCE / Occupancy

Probability of occupancy (Ψ) at a plot is estimated with up to 12 different models for the same species respectively narrative names for which DISTANCE analysis is run (model overview by study area in appendix, starting at page 124). Estimates and standard errors for a study area are only readily available for models 1 and 2 because of PRESENCE results structure. When covariates are used PRESENCE gives single Ψ values for each of the 30 plots. For brevity it is decided not to display all 30 plot results for each narrative from all six study areas. Calculating Ψ estimates and confidence intervals for each study area considering the covariate values was of a mathematical complexity beyond the limits of this thesis. Detailed results are available for each individual plot from each study area from the PRESENCE project files in the digital appendix. Thus, in this section only results from model 1 (assuming constant probability of detection for all three visits) and model 2 (calculating survey-specific probabilities of detection for each visit) are shown. A table showing best models selected by AIC is given in the appendix (from page 189).

3.4.1 PRESENCE Results: Occupancy Estimates

Occupancy estimates for point transect data is shown in Figure 93 and Figure 94. Both models give relatively constant results for all narratives, model 1 with constant p sometimes having slightly higher estimates. Differences between confidence intervals ranges are also small.

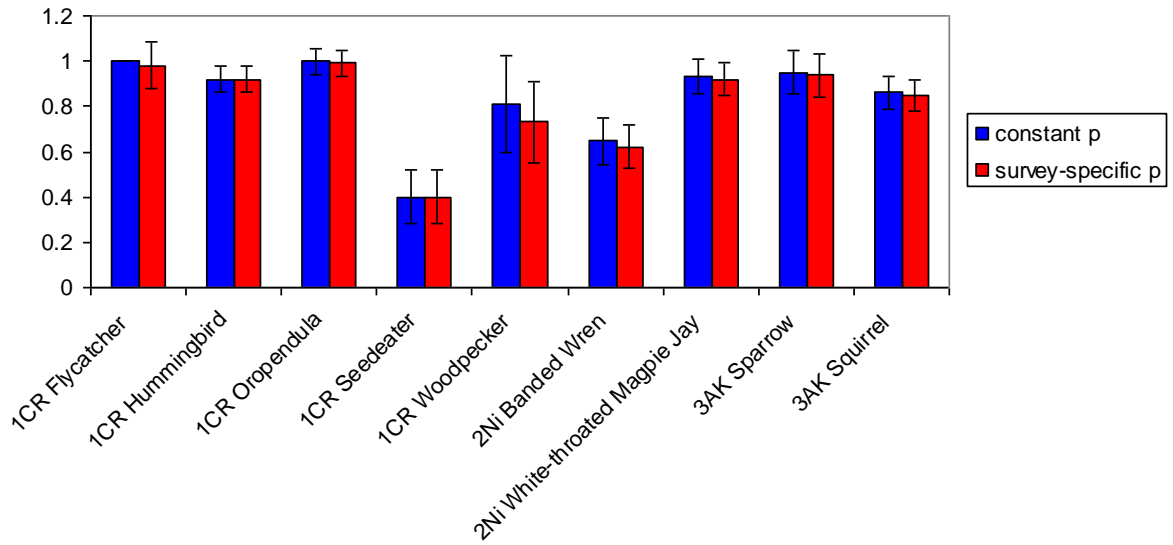


Figure 93: Occupancy estimates and confidence intervals of two models for point transect data (1CR-3AK)

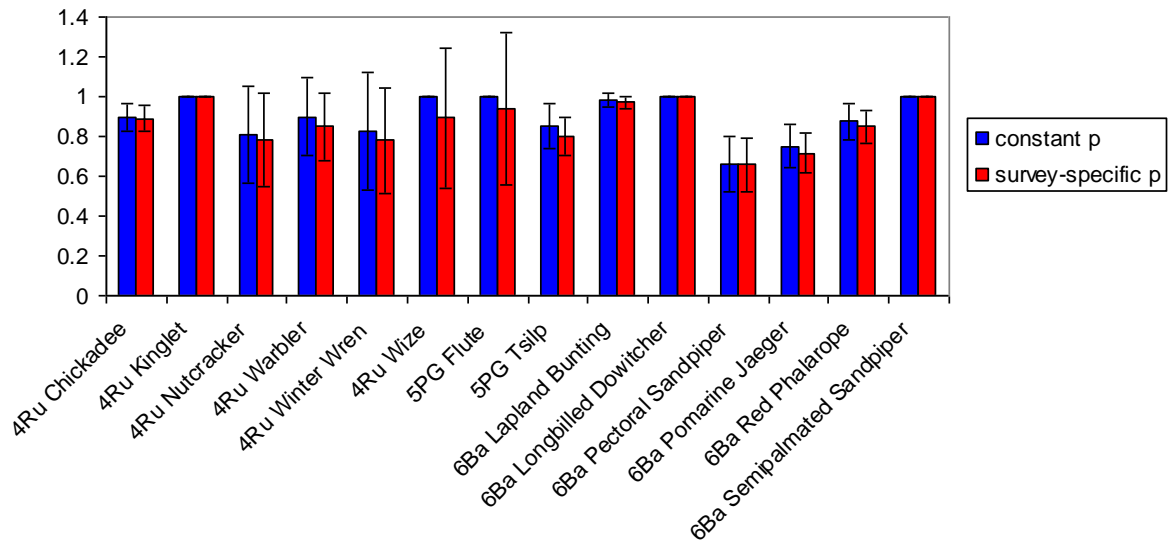


Figure 94: Occupancy estimates and confidence intervals of two models for point transect data (4Ru-6Ba)

Figure 95 and Figure 96 display occupancy estimates for model 1 and model 2 for trapping web data. All but three narratives reach occupancy estimates of 1.0 without confidence interval, meaning that the animals the narratives refer to occupy the plots in the study area with certainty. The three narratives with Psi estimates < 1.0 are Springtail from 3AK, Cycsegusa from 4Ru, and Milbe from 6Ba. There is no immediate explanation why these three narratives differ from the others. The large number of trapping web narratives with

perfect occupancy estimates of 1.0 is probably the result of two phenomena: small number of trapping webs, and small size of each web. The small number of trapping webs leads to relatively little variation in habitat types. Combined with the small size of each web it also leads to low numbers of catches for common and rare species, so that these can not be analyzed properly. On the other hand, abundant species are caught at each plot and reach the necessary number of observations, but also naturally reach very high to perfect occupancy estimates.

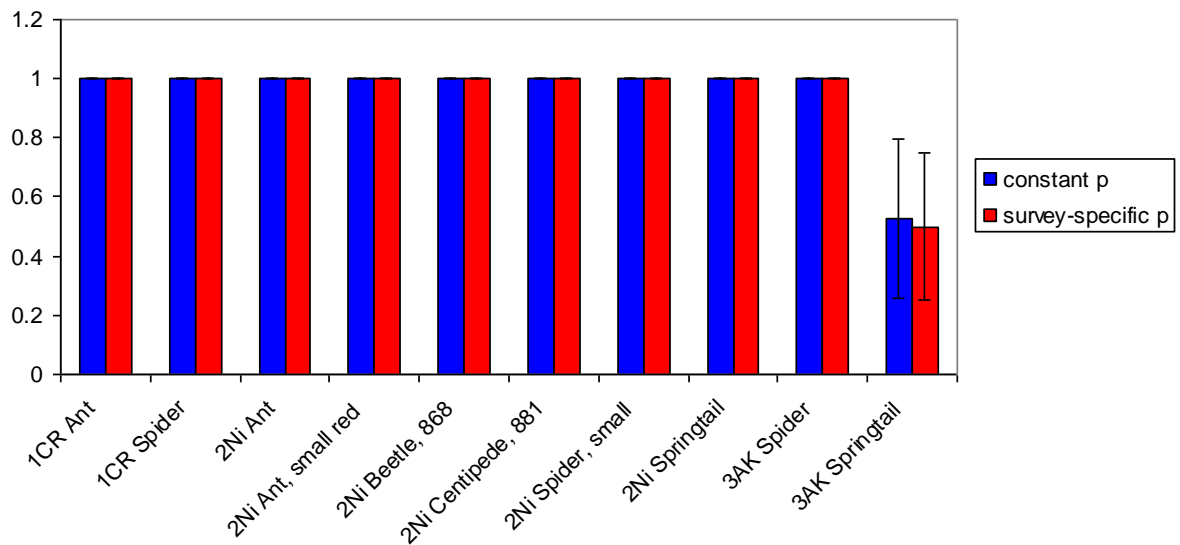


Figure 95: Occupancy estimates and confidence intervals of two models for trapping web data (1CR-3AK)

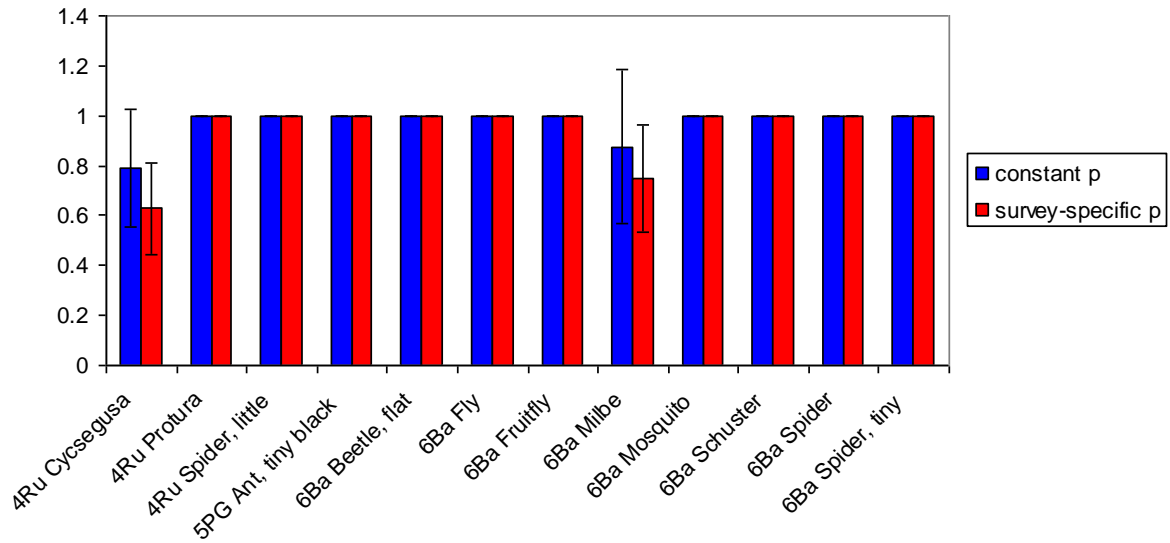


Figure 96: Occupancy estimates and confidence intervals of two models for trapping web data (4Ru-6Ba)

3.4.2 PRESENCE Results: Randomly vs. Systematically Selected Plots

The comparison of analysis for pooled, random and systematic data sets for point transect data shows that differences in occupancy estimates are much smaller than for density estimates in former chapters (Figure 97 and Figure 98). The main exceptions are Seedeater at 1CR and Pomarine Jaeger at 6Ba, which both have comparably high occupancy estimates for the random data set.

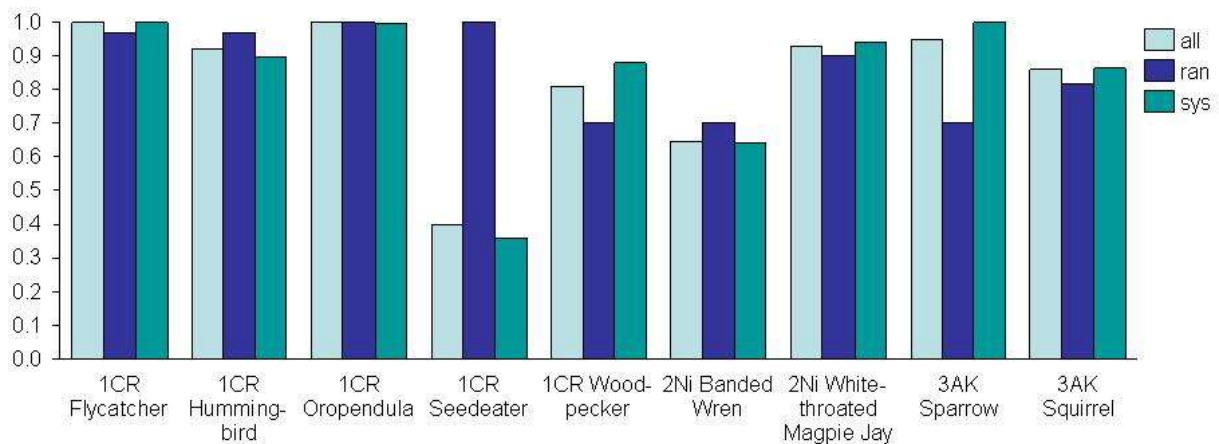


Figure 97: Comparison of occupancy estimates for point transect data from random and systematic plots (1CR-3AK)

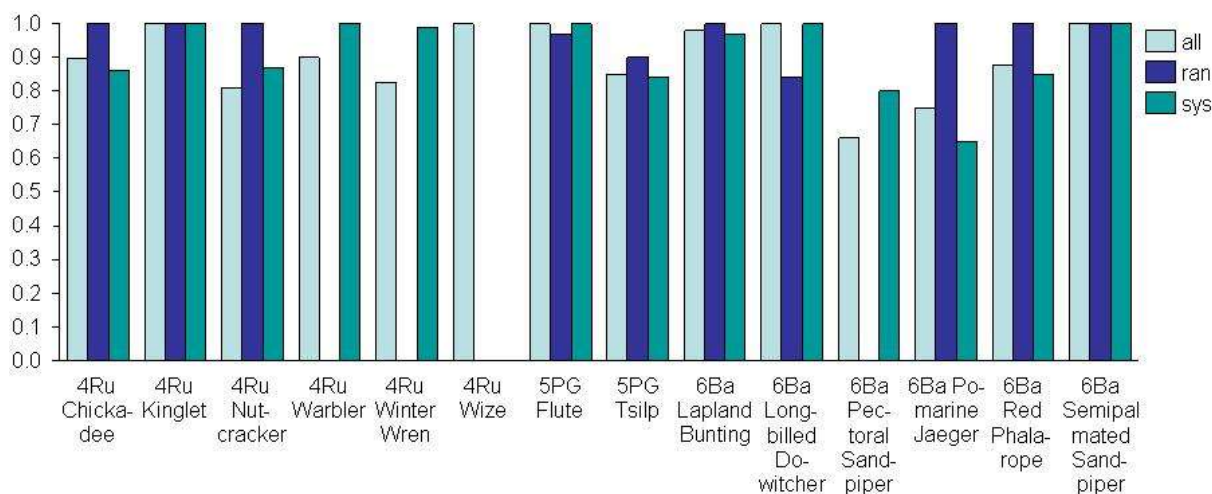


Figure 98: Comparison of occupancy estimates for point transect data from random and systematic plots (4Ru-6Ba)

3.4.3 PRESENCE Results: Aural vs. Visual Bird Detections

Figure 99 and Figure 100 compare occupancy estimates for point transect data from pooled data set with data sets including only aural and only visual detections. Some narratives have relatively close estimates for all three data sets (e.g. Hummingbird at 1CR or Warbler at 4Ru), indicating that there was no problem in pooling two kinds of detection together. Especially in study areas 4Ru and 5PG some narratives are detected only aurally, resulting in missing estimates for visual data set and equal or very close estimates for the aural and the pooled data set. In many cases one kind of detection results in a considerably higher estimate, for example Seedeater at 1CR or Banded Wren at 2Ni. This strongly suggests that at least for those narratives where this is the case the pooling may negatively affect estimates and analysis should be separated by type of detection.

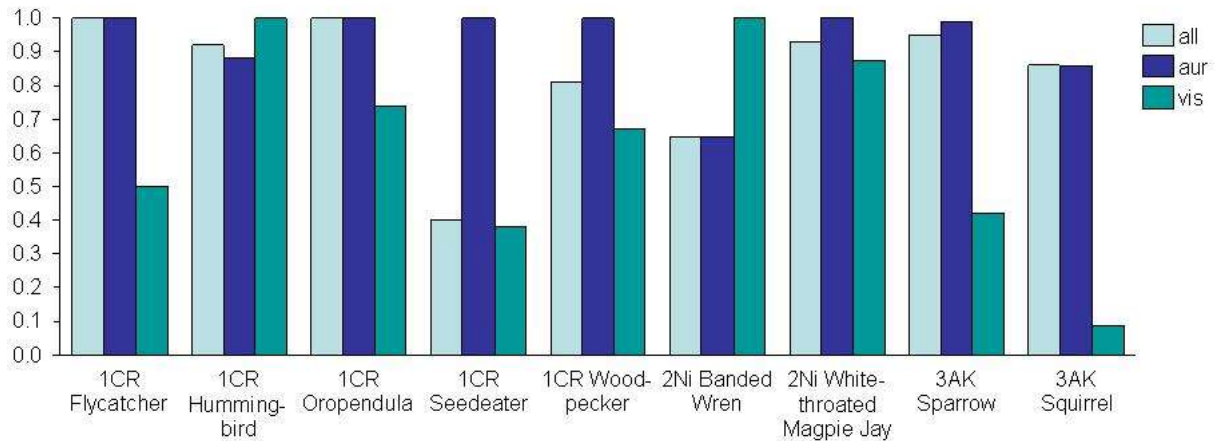


Figure 99: Comparison of occupancy estimates for point transect data from aural and visual detections (1CR-3AK)

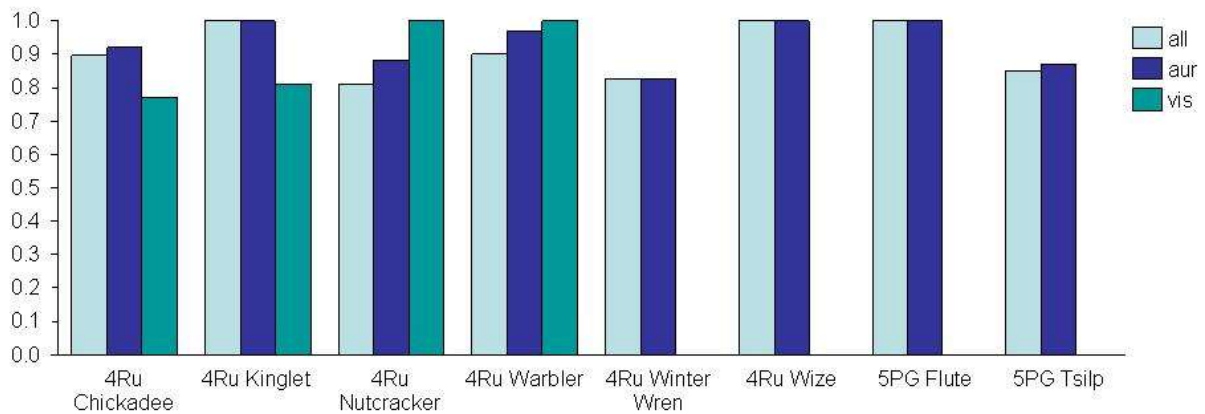


Figure 100: Comparison of occupancy estimates for point transect data from aural and visual detections (4Ru-5PG)

3.4.4 PRESENCE Results: Biological Family and Order

At the biological order level the results and confidence intervals calculated through the two PRESENCE models are relatively close together (Figure 101 and Figure 102). The only exception is Passeriformes from 4Ru, for which the constant detection probability (model 1) results in considerably lower occupancy estimate. Trapping web estimates are all at 1.0 and show no differences for different biological orders, which is difficult to analyze (see also chapter 3.4.1). The same tendency can be seen for point transect data, where only two of eleven biological orders show occupancy estimates lower than 1.0.

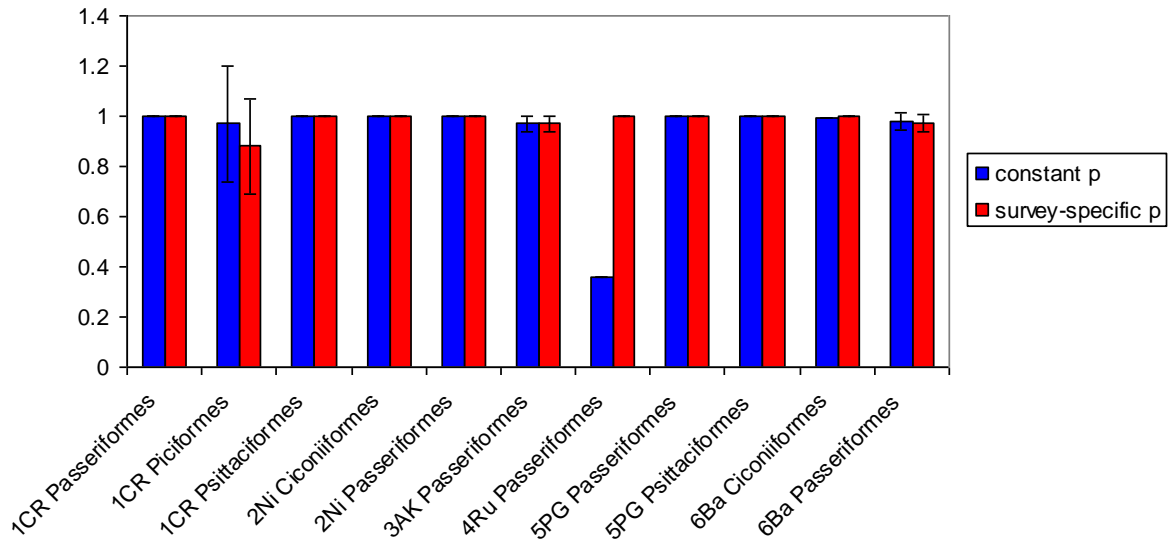


Figure 101: Occupancy estimates and confidence intervals of two models for point transect data at biological order level

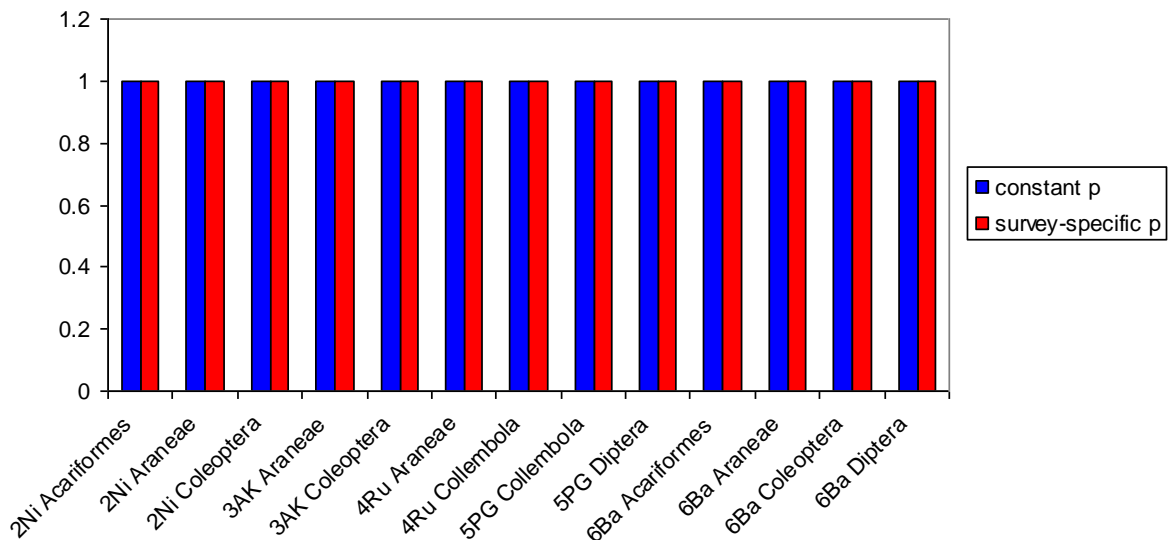


Figure 102: Occupancy estimates and confidence intervals of two models for trapping web data at biological order level

Results at biological family level for point transect data are more diverse (Figure 103), while for trapping web data all estimates show certain occupancy ($\Psi = 1.0$, Figure 104). Results of both models are close together, with the exception being Paradisaeidae from 5PG. Here model 1 results in an occupancy estimate of 1.0 and model 2 comes close to this estimate, but shows an exceptionally large confidence interval. Explanation for this exception can not be offered.

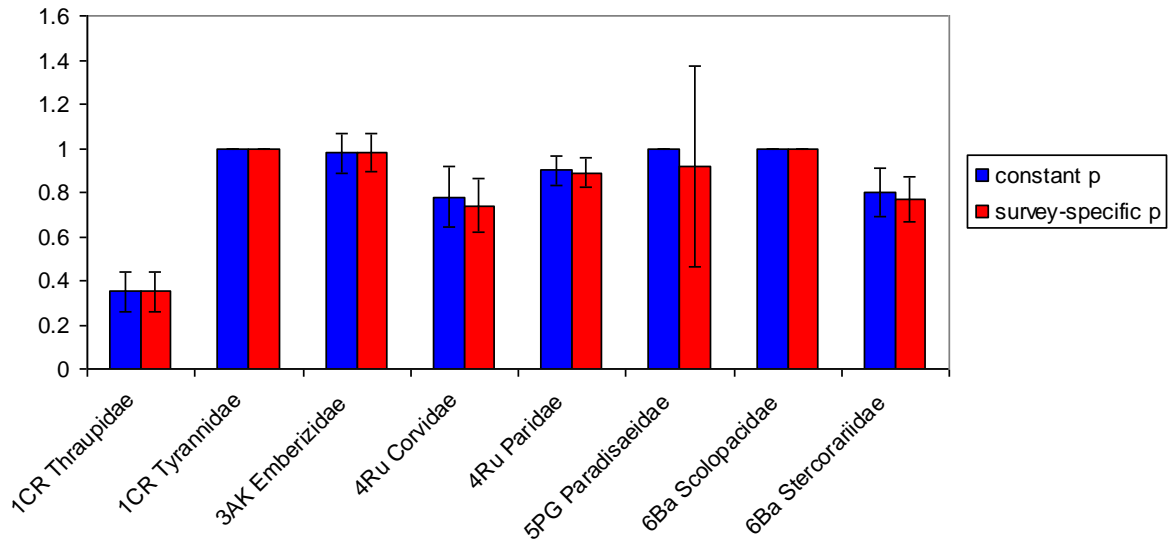


Figure 103: Occupancy estimates and confidence intervals of two models for point transect data at biological family level

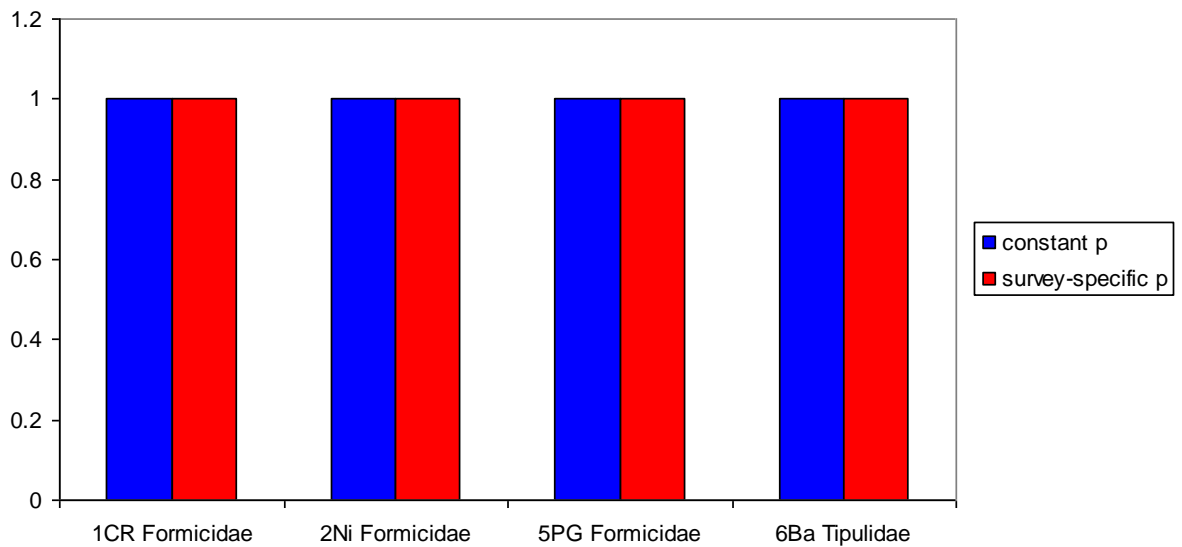


Figure 104: Occupancy estimates and confidence intervals of two models for trapping web data at biological family level

3.5 Comparing DISTANCE and PRESENCE Results

Abundance as estimated by DISTANCE is a direct estimate of population size. Occupancy estimates as derived by PRESENCE are expected to correlate with population size and are seen as indicator for population trends by many wildlife biologists. Following this assumption the estimates of DISTANCE and PRESENCE can be expected to correlate. In this chapter this assumption is analyzed in two parts, separately for point transect and trapping web results:

1. correlation between p estimated by DISTANCE and p estimated by PRESENCE
2. correlation between d estimated by DISTANCE and Ψ estimated by PRESENCE.

For this analysis only data from systematic plots is used to avoid any differences between the analysis methods and their ability to handle the combined data, although real differences in the data from the two types of plot would not be expected.

3.5.1 Comparing Point Transect Results

The correlation graph for DISTANCE detection probabilities and PRESENCE detection probabilities from point transect data shows a slightly negative correlation (Figure 105). There is no immediate explanation for this phenomenon, and more detailed analysis did not bring much different results. Both detection probabilities seem to be not directly comparable.

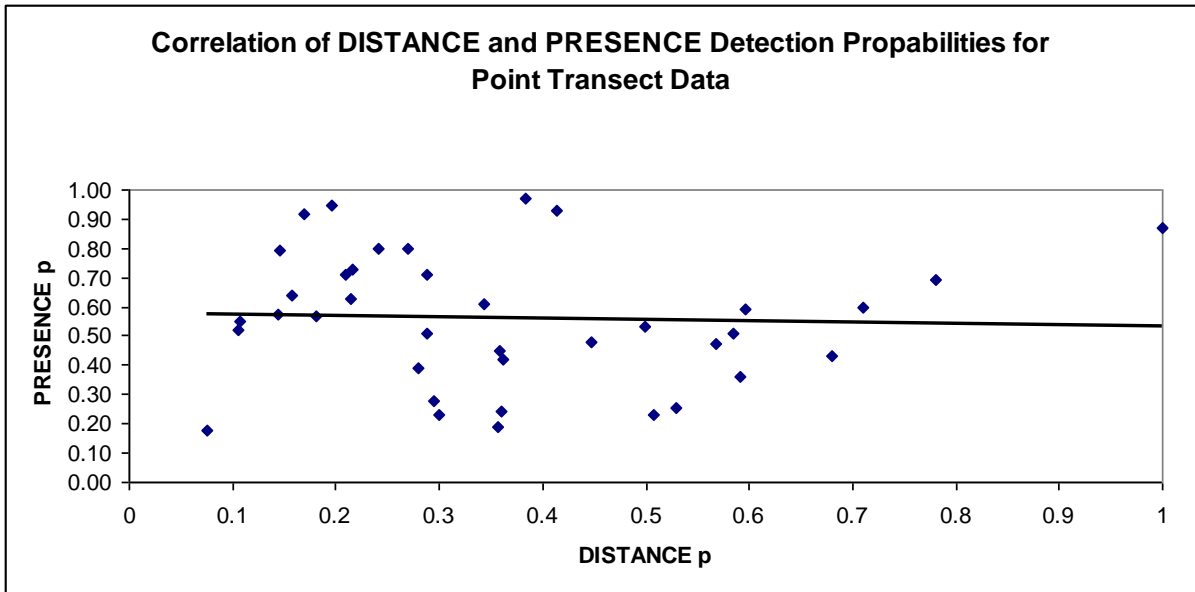


Figure 105: Correlation of DISTANCE and PRESENCE detection probabilities for point transect data (all study sites)

Abundance and occupancy estimates correlate weakly positive (Figure 106). For brevity reasons only those estimates are analyzed in more detail.

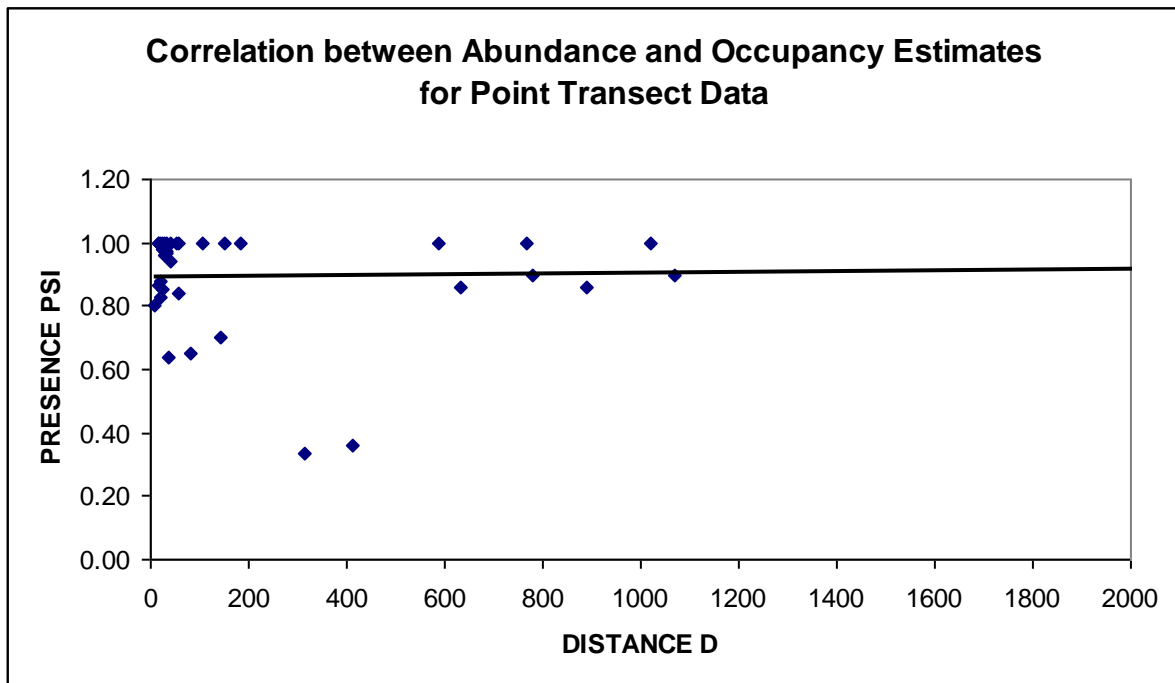


Figure 106: Correlation between abundance and occupancy estimates for point transect data (all study sites)

Correlation for point transects data differs substantially between study areas (Figure 107 to Figure 109). For most study areas relatively low positive correlation can be observed (1CR, 2Ni, 4Ru, and 5PG). The graph for study area 3AK shows a rather steep positive correlation. There is no immediately available explanation why this study site differs so clearly from the other four. The results for study area 6Ba are sticking out even more; there a negative correlation between the two estimates is observed (Figure 109).

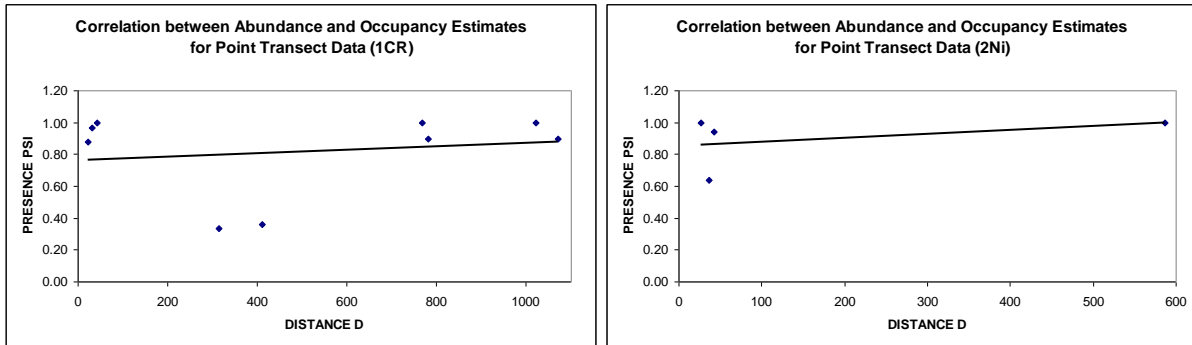


Figure 107: Correlation between abundance and occupancy estimates for point transect data (study sites 1CR and 2Ni)

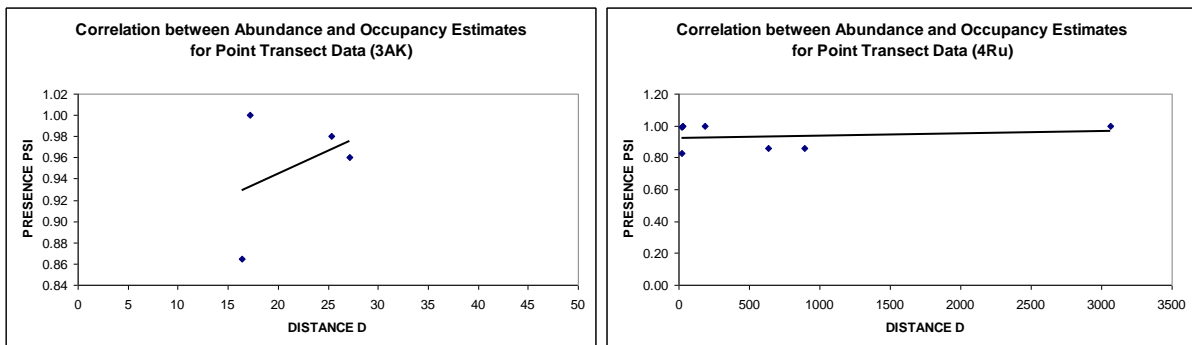


Figure 108: Correlation between abundance and occupancy estimates for point transect data (study sites 3AK and 4Ru)

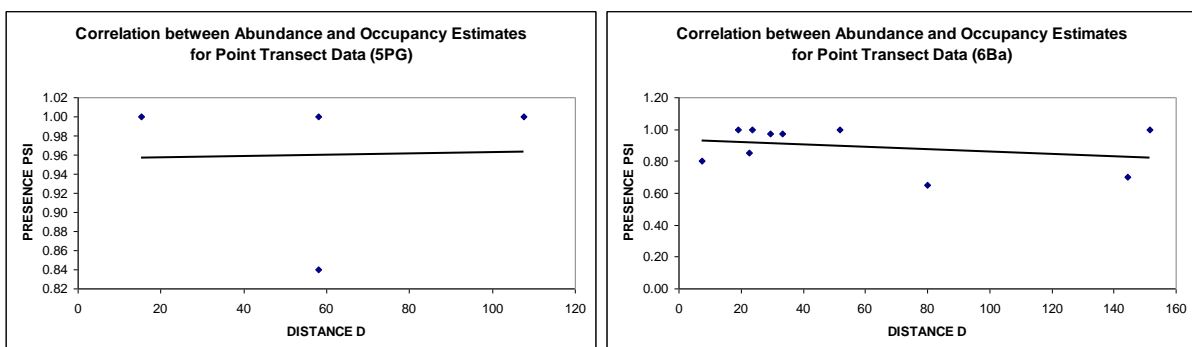


Figure 109: Correlation between abundance and occupancy estimates for point transect data (study sites 5PG and 6Ba)

3.5.2 Comparing Trapping Web Results

Detection probability as estimated by PRESENCE compared to the one estimated by DISTANCE are negatively correlated also for trapping web data (Figure 110). This underlines the impression from the former chapter that both estimates are possibly not directly comparable, despite having the same label.

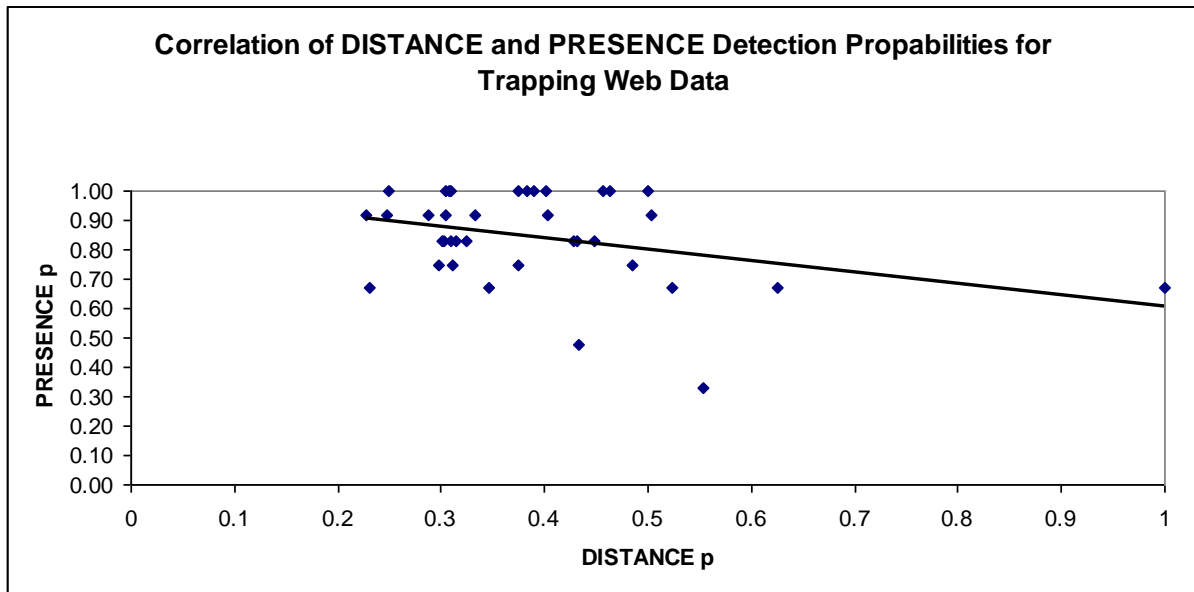


Figure 110: Correlation of DISTANCE and PRESENCE detection probabilities for trapping web data

Figure 111 shows the trend for correlations between occupancy and density estimates: the constantly high occupancy estimates for trapping web data result in a simple horizontal line with no variation for higher density estimates. The same trend can be seen in the separated figures for study areas 1CR and 2Ni (Figure 112), figures for study areas 3AK and 5PG are not displayed because they basically have the same outlook. The large numbers of narratives with perfect occupancy result probably from small number and size of trapping webs (as formerly discussed in chapter 3.4.1).

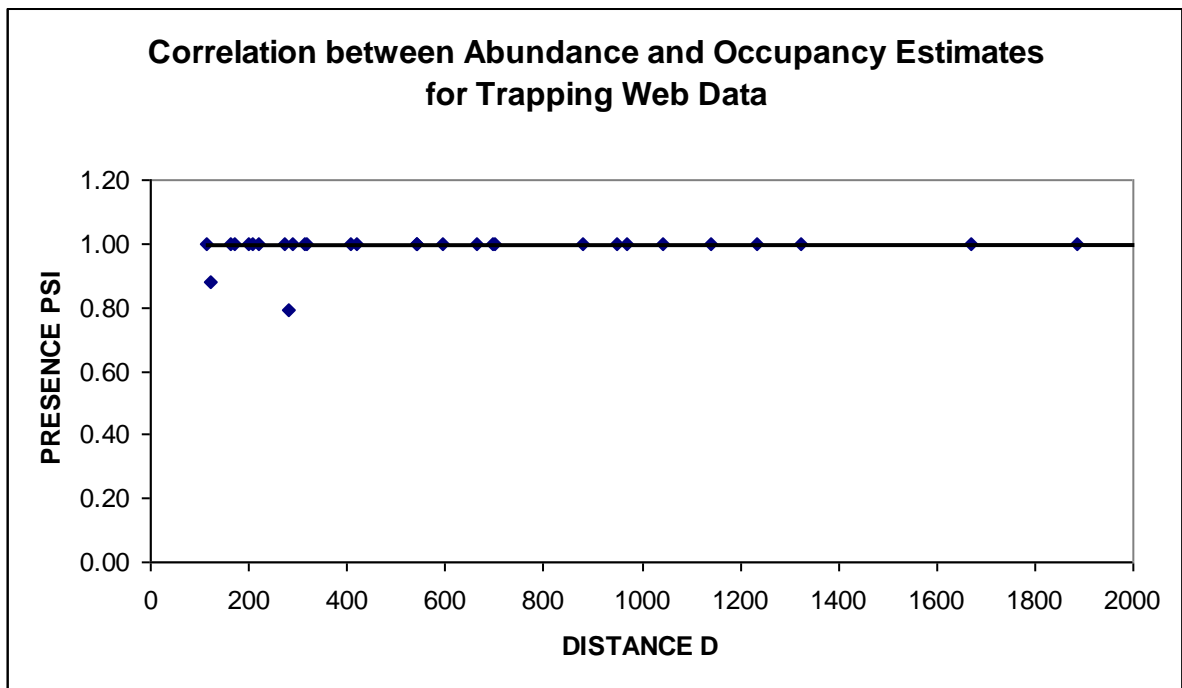


Figure 111: Correlation between abundance and occupancy estimates for trapping web data (all Study Sites)

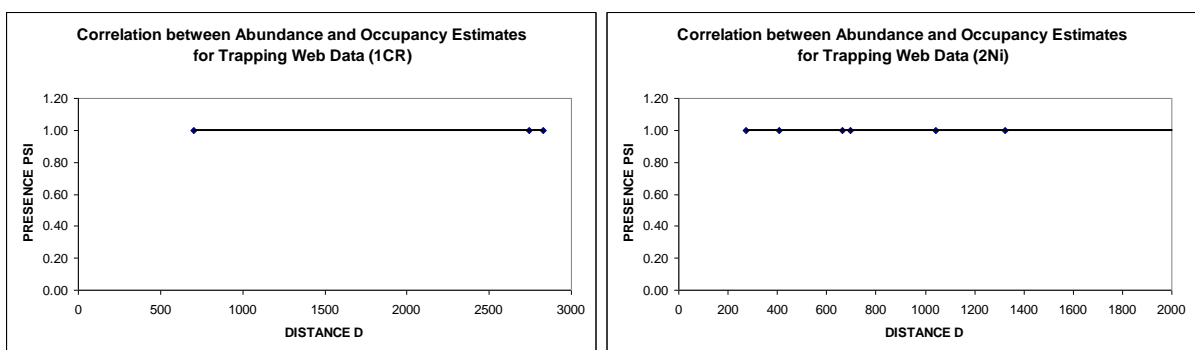


Figure 112: Correlation between abundance and occupancy estimates for trapping web data (study sites 1CR and 2Ni)

At study sites 4Ru and 6Ba a slightly positive correlation can be observed (Figure 113). A closer look reveals that this is in both cases the result of one data point with low DISTANCE density estimate being off the 1.0-occupancy line, while all other data points are exactly on this horizontal line (as for the other study areas). It is also at least questionable if a valid correlation can be built on only four data points.

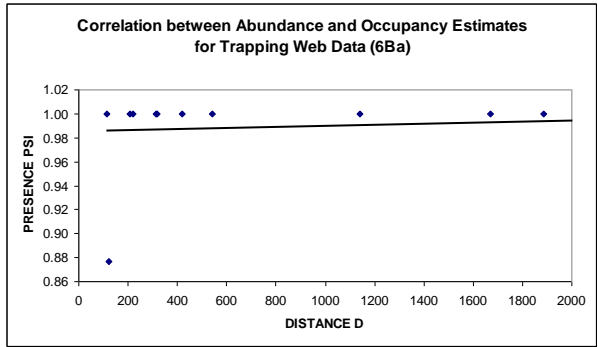
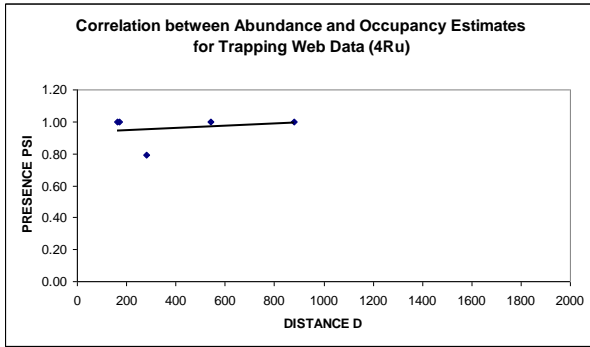


Figure 113: Correlation between abundance and occupancy estimates for trapping web data (study sites 4Ru and 6Ba)

4 Discussion

4.1 Discussion of Results

The amount of information gathered through each GRID was enormous given the relatively short sampling time of 10-14 days. Table 17 displays the number of narratives for which valid predictive models were retained through Random Forests analysis (ROC value > 0.5); as well as the number of narratives for which abundance estimates through DISTANCE sampling and occupancy estimates through PRESENCE analysis were gained. The line transects add-on protocol added another three narratives to Random Forests analysis for 1CR and 2Ni and one narrative to DISTANCE/ PRESENCE analysis for 2Ni (not included in Table 17). In short, 116 valid predictive models were gained through only 12 weeks of sampling! 45 of those also delivered abundance and occupancy estimates. In Random Forests analysis each of the three constructed models (*Plot*, *Covariates*, *Interspecies*) proved to be the best model for some of the narratives analyzed, thus each of them was ultimately useful. A consideration built on this observation is the construction of more model definitions, especially if additional spatially assigned covariate data becomes available from other sources. The quality of the data was in some cases not very good; especially confidence interval ranges for DISTANCE sampling were relatively large. It is assumed that larger data sets will enable stratification of analysis by habitat and solve this problem. Larger amounts of data would also open the possibility to split the data in different sets and analyze those separately if this is assumed to be beneficial for analysis precision, as is probably the case for aural and visual detection of some narratives (chapter 3.3.5). PRESENCE analysis showed a high tendency for perfect probability of occupancy ($\Psi = 1.0$), especially for trapping web results. It is expected that this problem can be solved with larger data sets, especially larger numbers of plots respectively more GRIDs per study area. The effective PRESENCE sampling size for each narrative in each study area was 30 per visit, because this is the number of plots and detection/ non-detection is entering the analysis only once per plot and species (similar to the *Plot* model in Random Forests). DISTANCE analysis on the other hand benefited from spatial repeats, because all distances of all narrative detections at each plot were used for the modeling of the detection function (similar to the *Covariates* and *Interspecies* models in Random Forests).

Table 17: Total number of narratives analyzed by study area, sampling method and analysis method

| | Random Forests | | | DISTANCE / PRESENCE | | |
|---------------|----------------|-----------------|---------------|---------------------|-----------------|---------------|
| | Total | Point Transects | Trapping Webs | Total | Point Transects | Trapping Webs |
| 1CR | 19 | 16 | 3 | 7 | 5 | 2 |
| 2Ni | 23 | 11 | 12 | 8 | 2 | 6 |
| 3AK | 10 | 4 | 6 | 4 | 2 | 2 |
| 4Ru | 23 | 13 | 10 | 9 | 6 | 3 |
| 5PG | 23 | 7 | 16 | 3 | 2 | 1 |
| 6Ba | 18 | 7 | 11 | 14 | 6 | 8 |
| Total: | 116 | 58 | 58 | 45 | 23 | 22 |

The comparison of ROC values for data from systematic and random plots showed relatively small differences (chapter 3.2.2). In most cases the ROC values for data from systematic plots were very close to the pooled ones, which is to be expected as survey effort for this data had a share of roughly 80 % at each study site (25 plots compared to 5 plots). In a few cases the ROC values of data from random plots were surprisingly high. In these cases the most likely explanation is that the input data was very clean, allowing for very strong models. Random Forests has no tendency to overfit data with larger numbers of covariates for small sample sizes (Breiman 2001). In DISTANCE and PRESENCE analysis the differences between the estimates from random and systematic plots were generally low, results usually were very close. Few exceptions stand out, where the difference between estimates from random and systematic plots was rather large (e.g. Hummingbird at 1CR and Chickadee at 4Ru).

The comparison of results from visual and aural data sets implied that for a detailed analysis the individual species' biology has to be taken into account. In many cases the analysis results were similar, but there were also many narratives for which the proportion of observations by one of the two means of detection (aural/ visual) as well as the gained results differed considerably. In some cases the larger number of observations was of one type of detection, while the ROC value derived through Random Forests analysis was larger using the other type of detection. There were also a number of narratives for which the pooled data resulted in considerably higher DISTANCE abundance estimates than the separated data sets, which seems to be implausible (e.g. Oropendula at 1CR and Chickadee at 4Ru). A multitude of biological reasons can explain such differences, for example gender dimorphism in singing behavior (affecting aural detectability), gender dimorphism in coloration of plumage (affecting visual detectability), differences in general behavior/ secrecy by age or gender,

different effect of environmental covariates on (especially visual) detectability etc. It is highly recommended to separate between aurally and visually collected data, although this was not possible continuously in the study at hand because of relatively low sample size.

There was a considerable difference between analysis at the narrative level and analysis at higher taxonomic levels. Figure 16 for example shows that 80 and more observations for a narrative generally resulted in valid Random Forests models with ROC > 0.5 (page 37). For higher taxonomic levels there were some with more than 200 observations which did not or only barely result in valid models, and generally a high number of observations did not automatically imply a high ROC value (chapter 3.2.4). This indicates that pooling by taxonomic classes like family or order is not a recommended way to receive bigger datasets. Differences on the biological level, like habitat requirements, can be huge between two species belonging to the same taxonomic class, especially when looking at such diverse ones as Passeriformes. The benefits gained through the analysis of higher taxonomic classes were rather low; the range of DISTANCE confidence intervals was not reduced. In PRESENCE analysis the pooling caused similar problems for point transect data as did the small number of plots for trapping web data: low variance between plots, because some bird belonging to the order Passeriformes was detected at almost every plot, and therefore universally high occupancy estimates (Psi = 1.0 for all analyses at the biological order level). The assumption that higher taxonomic categories can be used as valid surrogates for rapid assessment and monitoring of species diversity gains no support from this study.

Generally both PRESENCE and DISTANCE analysis are supposed to estimate trends in animal populations. Joseph et al. (2006) for example directly compare the two methods and give a recommendation for which types of species which type of sampling can be used. They come to the conclusion that “*Abundance surveys were best if the species was expected to be recorded more than 16 times/year; otherwise, presence-absence surveys were best*” (Joseph et al. 2006). Support in the study at hand for interchangeability of these two methods is ambiguous. Probability of detection as estimated by DISTANCE correlated slightly negative with probability of detection as estimated by PRESENCE. The two might be mathematically different in fundamental ways while only sharing the same label, but this hypothesis was untested. Detailed comparison of these two different p estimates and their methodological differences is beyond the scope of this thesis. The main results, DISTANCE density estimates and PRESENCE occupancy estimates, showed in all but one study area positive correlations

for point transects (chapter 3.5.1). Correlations for trapping web data were basically not analyzable because of the large number of perfect or near perfect occupancy estimates ($\Psi = 1.0$, cp. chapter 3.5.2). It is also imaginable that the strict GRID design, which is not tailored to gain high precision results for any of the two methods, does in fact work in favor for one of the two.

4.2 Discussion of the GRID Approach

This section discusses problems with the biodiversity GRID in general and the six study sites specifically. Despite the global relevance and scope of the project it was not funded by relevant funding bodies. This resulted in very few sampling sites which in addition had been selected opportunistically: GRIDs were installed in areas of ongoing other research. The coverage and diversity was still extremely high so that problems of this way of selecting are not expected, but with better funding a more careful design could have been implemented. Probably more important would be a higher number of study sites, since effectively the sample size for the whole globe is six study areas. Another very important area definitely needing attention is the development of a similar approach for aquatic or partly aquatic ecosystems, which have been ignored completely in this work despite their importance for biodiversity.

One reason for the lack of funding could be the visionary approach taken. The intention to have a globally applicable multiple-species monitoring and rapid biodiversity assessment scheme is contrary to the recommendation of many scientists to aim for maximum precision by designing each survey individually for each species of interest (Bailey et al. 2007; MacKenzie & Royle 2005; Pollock et al. 2002). The argument against this point of view is that it is ultimately more cost-effective and useful to aim for several dozens of species estimates with a precision of plus-minus 80 % (or similar) than to aim for only 1 species estimate with a precision of plus-minus 5 %. It has also been shown that the assumption that information of single species can serve as surrogate information for biodiversity in general is often not valid (cp. also van Jaarsveld et al. 1998). Manley et al. (2004) make a point that history of ecological research is rather dubious in some cases, which can also result in favoring multiple-species surveys: *“Any effort that relies solely on a small set of indicator*

species will be subject to skepticism given the history of misuse, overuse, and poor performance of the indicator concept”.

In addition, the large data set produced through this type of survey resulted in a treasure for data mining approaches and pilot study data for more detailed study of species of special interest, for which pilot studies would have to be conducted anyway. A promising idea that to the author’s knowledge has not been tested would be to use predictive modeling to identify study sites for adaptive sampling of species of special interest, optimizing precision per study effort (as described by Pollard & Buckland 2004). This could prove especially useful for the monitoring of endemic species with small regional distributions, which might be sampled inadequately by a global biodiversity GRID system, depending on plot density. Distances of 100 m between plots seem to be ideal, but can probably not be achieved on a global scale. Assuming a land area of about 130 million km² (without Antarctica) this would result in roughly 13 billion plots. Increasing the distance to 500 m would still add up to 2.6 billion plots; while 5 km distances as have been used by Magness et al. (2008) in Alaska would result in 260 million plots. This sounds huge at first, but political will built on economic and social considerations clearly decided to protect biodiversity, while so far the necessary actions to do so are lacking. Information is essential for conservation and protection, while “*the extent of global data gathering underway is inadequate to meet the challenge set out at the WSSD in Johannesburg*” (Green et al. 2005). To act accordingly is certainly costly but so is the cost of restoration, with the latter one often being even higher than combined costs of monitoring and conservation (Dobson 2005). The decision ultimately boils down to one question: how valuable is reliable knowledge? (MacKenzie 2005b).

4.3 Discussion of Sampling Methods

The overall biodiversity GRID approach has some promising aspects, especially global coverage and avoidance of bias common in many population studies (e.g. roadside bias, Kadmon et al. 2004). However, some aspects of the sampling methods can be discussed, one being differences in taxonomic knowledge and identification skills between different observers. This was already observable in this relatively small pilot study and will probably grow to a major challenge for a truly global GRID system. The low number of identified

species in study area 3AK for example could partly be an effect of less ornithological experience of the observer compared to the observers at the other study sites, qualifying especially the simple species richness estimates from chapter 3.1 (Figure 10 and Figure 11). Generally speaking many biological aspects can hinder identification when observers are lacking specific experience, for example species' gender and age dimorphism in appearance and behavior (the latter one also affecting detectability). On the other hand it is well known also for more traditional survey approaches that "*misidentifications at the species level are common*" (Guralnick et al. 2007). It can be argued that lower precision in taxonomic identification is a minor issue compared to the analysis methods and results offered by the GRID, much as lower estimate precision discussed above. In addition it has been found that a feedback system to integrate observers experience with the sampling methods in different ecosystems is essential. Observers sometimes decided to make immediate adjustments in the field, like exclusion of seabirds from plot A1 in study area 4Ru. A communication system has to be implemented to ensure that this information will be taken into account when analyzing the data sets. It also has to be checked whether it would make sense to include the adjustment in the general survey protocol.

A major issue with the data as collected for this pilot study is the large number of observations with subjective descriptions. An observation noted as "tiny ant" may be a "small ant" for the next observer, thus real monitoring of trends in time by visiting the same plot several times might prove difficult, especially when different observers are surveying. There is no immediate solution for this problem, because even if time and financial resources allowed for an extensive training period prior to sampling, for many regions and ecosystems qualified trainers and literature would still not be available, especially not for more than one taxonomic group. However, when considering that no other relevant data exist, such approaches will help to further fine-tune sampling efforts in the future. And at least for aural detections of bird species automated identification methods are under development (Brandes 2008). These add other technological and financial challenges, but those are expected to be smaller than those from providing adequate training for a large number of observers. It is far easier and more reliable to teach a bird song to a computer and multiply digitally than to train human observers one by one. Thus it would still be costly technology; but it is also expected to be a cost-effective method. Similarly automated approaches for identification of insect species and visual bird detections would be extremely helpful for further development of the biodiversity GRID approach, but are to the author's knowledge not (yet) in sight.

Another important question is the importance of time of the year when the survey is carried out. Buckland et al. (2008) recommend the breeding season for bird surveys. The opportunistically selected study sites in this project were not ideal to meet this criterion. Especially at study area 3AK most aural bird detections were by single call, no bird song melodies were detected, indicating that surveys took place in the last phase of or maybe even after the breeding phase. To conduct studies exactly at the best time of the year will provide a considerable planning challenge in a global project.

Three other issues have been observed which mainly affect trapping webs. The first is survey effort. Despite all endeavor to keep survey effort constant between all study plots and visits, this goal was already not achievable for only 24 sampled plots at 6 study areas. It can be assumed that the differences are not very important because all trapping events took place over night (at least 12 hours trapping time) and differences in trapping time were after all relatively marginal. The second issue with trapping webs is the availability of weather protection and/or a trapping fluid. For budget reasons in this study all cups were set dry and unprotected from rain and predators. Both points do not seem to cause any immediate problems, but a more sophisticated approach would be to protect the cups with small roofs and use a trapping fluid, which would also avoid predatory arthropods to eliminate each others while in trap. However, this could also exclude insects that fly into the trap. These issues require more study. The last point is that some species traits that are regarded to have important influences on precision (e.g. home range size and movement rates, Lukacs et al. 2005) can not be taken into consideration for study design when using multiple-species trapping webs. As stated before and shown in this study the gain in number of animals surveyed in combination with sophisticated modeling approaches outweighs this lack of precision.

The tested line transects add-on protocol produced observations only in the first two study areas, and enough for analysis only for butterfly species. Snakes were detected, but only when walking between the plots and not enough to run a valid analysis. For many species it is probably just the small survey effort of 300 m per GRID that is not sufficient to collect enough data. Thus, add-on protocols using line transects probably have to use longer lines, which might result in problems to assign them to spatial covariate data. Occupancy estimates

for this sampling method would have been especially interesting, but the necessary three repeats were not realizable.

Violation of assumptions necessary for DISTANCE and PRESENCE analysis has not been tested explicitly in this study. Repeats of sampling at each plot have been done within few days, so that extensive movement of animals is unlikely, but not impossible. Movement of animals in and out of the sampled area would be problematic for PRESENCE analysis, which assumes a constant population. However, results did not imply that this is a problem in the study at hand. Results of models with constant detection probability were generally very close to those with survey-specific detection probability (chapter 3.4.1). DISTANCE assumptions were a bit trickier. Recent research suggests that the assumption of perfect detectability close to the observer, at a distance of 0 m ($g(0) = 1$), should be vigorously tested in each study (Bächler & Liechti 2007). This is simply impossible given the number of study sites and different ecosystems, the short time and the budget constraints. This assumption might have been violated for four of the 45 point transect DISTANCE estimates, where detection functions clearly showed no observations in the first segment from the observer (chapter 3.3.1). Another assumption crucial for DISTANCE analysis is high precision of distance estimates. In this study all distances for point and line transect observations were estimated by the observers without technical distance measurement tools, because technical devices were not available and would have failed in some of the environments anyway, especially for aural detections. Besides, the GRID system with distances of 100 m between plots and additional markers at 50 m between plots proved to be extremely helpful for the observers to validate their estimates.

The last minor issue worth mentioning in this section is a possible effect the preparation of a GRID may have on animal behavior. Especially in dense lowland tropical rainforest as encountered in Costa Rica there is a necessity to cut trails if one wants to do intensive repetitive sampling in the area over several days or even weeks. On the one hand it would be interesting to know if these trail works actually affect the gathered observations, but on the other hand there is little one can do to investigate this issue. To count and survey animals scientists have to walk through the forest, and to do this in a lowland tropical rainforest in Costa Rica some trails have to be cut prior to sampling.

4.4 Discussion of Analysis Methods

Most of the encountered analysis problems have been mentioned in the results section already. In this section a summary of general limits of the study are given. The biggest problem for analysis is the relatively small data set. On the one hand more than 5,000 observations are available for analysis, but on the other hand these were collected in 6 study areas with three different survey methods. Strictly speaking the splitting in most cases should have gone even further to separate data collected at randomly selected plots from data collected at systematically selected plots, and aural detections from visual detections at the same time. But all analysis methods used usually profit from larger amounts of data. The decision here was to report detailed results for the pooled data sets and additionally display trends for split data sets separately. The trends indicated in many cases that a general split would have been beneficial. Additionally, the split was always only at one level, random data was separated from systematic data, but not split further in aural detections from random data and visual detections from random data, simply because the data sets became too small for analysis this way. Another option that is certainly promising and could not be used because of lacking data is stratification by habitat. In this study habitat was used in different ways of analysis only as a covariate, simply because the data set was too small to stratify by habitat. In short: the more data the better the algorithms work and the more precisely data sets can be split by important features.

To analyze the relationship between species and covariates spatially the observations have to be assigned to one particular plot, which can be difficult in open habitats where neighboring plots can be observed from a plot. The solution used for this study was to truncate all data at 50 m, half of the distance between two neighboring plots. Mathematically this way of handling cuts out some of the data: the greatest possible distance between two GRID plots is ca. 141 m measured diagonally, resulting in some data within 70 m spatially belonging to a plot. However, without indication to which direction the observations were made from the plot a clearer analysis was not possible and some data that could be used is omitted for clarity. Maybe the use of a plot form other than circle would be beneficial, but then again all other plot forms are considerably more effort to use in the field, especially combined with point transect sampling.

Another issue with the covariates is an extremely low variation at study site 6Ba. Most of the covariates gathered in the field can be expected to correlate with presence/ absence and abundance of trees (highest tree, canopy cover, duff cover, plant species presence/ absence etc.). This worked well in environments where the landscape was diverse, including pasture, wetland, and different kinds of forests. It proved to be more problematic in a relatively homogenous ecosystem like arctic tundra. Other estimates which would not be visible in dense forests, like diameter of or distance to the next lake, were added to the protocol in this case. All in all the predictive modeling still worked well at this study site, but the recommendation resulting for global monitoring is to gather additional data with higher small-scale variations at each plot, especially pedological data available in all terrestrial ecosystems. Another idea is to use data from other surveys or other publicly available geo-referenced data, for example distance to roads from available maps or slope gradient and aspect from digital elevation models. The possibilities this approach opens are amazing. The only thing to be kept in mind is the GRID spacing: a satellite picture resolution with pixel sizes of 2 km x 2 km will not provide adequate covariates for modeling of a GRID with much smaller spacing (e.g. 100 m, as in this study).

Technically a problem was encountered with the MS Access database, which reached its limits already in this relatively small study. The use of point transect data for the *Interspecies* model aiming at the prediction of trapping web narratives was not possible. The combined number of trapping web and point transect data simply exceeded the limitations in the number of columns a query in MS Access can have. For biodiversity GRID studies on a larger scale a more sophisticated database application is therefore highly recommended.

5 Conclusions

This study demonstrates three of the analysis methods which carefully designed biodiversity GRIDs offer to ecological research. The available analysis options are by a magnitude more. Especially autocorrelation issues between plots and questions regarding fragmentation and change over time come to mind. Adding that all sort of spatial data, especially also those resulting from remote sensing, can be connected to the GRID data and that continuous efforts exist to make research data publicly available, the possibilities to conduct relevant analysis are enormous.

The results shown so far are more than promising. Three of the most sophisticated current methods in ecology are already involved: Random Forests as a powerful data mining tool to construct predictive models, DISTANCE sampling for the estimation of population abundance, and Occupancy estimation with PRESENCE to gain information for species with low sample sizes. Results lack in precision for each single species, but are promising regarding first snapshot assessments of multiple-species. Such an approach is urgently needed to improve cost-effectiveness of ecological research, while at the same time more precise study designs have their place in evaluation of known risk species for which more detailed population estimates are necessary. Challenges have been faced by the current study, but those are to be expected when working on a global scale. A number of recommendations could be given to improve the involved methods. The most pressing next step is to sample more study sites and build a stronger database. It is unlikely that the biodiversity GRID approach will be accepted and implemented by many country governments within a short time frame, so another urgent point of development is the connection with other data sources. Additionally the project would gain from development of a meta-software with the ability to batch several other software solutions, and from a closer investigation of the comparability of DISTANCE and PRESENCE results as well as detection probabilities estimated by those two programs.

Biodiversity GRIDs are an important step into the direction to fill holes in global biodiversity information for conservation and management in a cost-efficient way. The challenge to make this approach work is to move political decision makers to act according to their declarations of intent. The protection of biodiversity is not a selfless act of charity...

6 References

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7 Appendix

7.1 Data: Biodiversity GRID Fieldsheets

| | |
|--------------------|----------------|
| Site ID: | Crew: |
| Date: | Latitude: |
| Start time: | Longitude: |
| End time: | Elevation: |
| Trapping Web? Y/N: | Picture taken? |

| | |
|-----------------------|--------|
| | Notes: |
| Canopy cover (%): | |
| Open soil (%): | |
| Open water (%): | |
| Groundcover veg. (%): | |
| Land type: | |
| | |

Open soil: note main cause for open soil (e.g. cutline, trail, cattle)

Open water: note main type of water body (e.g. pond, lake, river, puddle)

Land type: note main habitat type (e.g. forest, river, lake, agriculture, rangeland)

| | |
|-------------------|--------|
| Temperature (°C): | Notes: |
| Rainfall? Y/N | |

Weather: note extraordinary weather conditions (e.g. strong wind, storm, hail)

Traces sheet:

| Species | Type of sign |
|----------------|---------------------|
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Arthropod Trapping Web:

| Species | Cup ID | Count |
|----------------|---------------|--------------|
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7.2 Covariates by Study Area

| Covariate | 1CR | 2Ni | 3AK | 4Ru | 5PG | 6Ba |
|-------------|----------------|-----|------------|----------------------------|--------------|------------------|
| Covariate01 | Melastomatacea | | Spruce | Picea_jesoensis | Fern | CottonGrass |
| Covariate02 | Costaceae | | Birch | Alnus_hirsuta | TreeFern | Coltsfoot |
| Covariate03 | Marantacea | | Nothofagus | Betula_ermanii | Heliconia | WhiteKelchFlower |
| Covariate04 | Heliconea | | Equisetum | Abies_sachalinensis | Impatiens | WhiteTowerPlant |
| Covariate05 | Palm | | Salicaceae | Larix_cajanderi | Grass | Sphagnum |
| Covariate06 | Piperaceae | | Plant 01 | Picea_sachalinensis | SquashFlower | Willow |
| Covariate07 | Mimosae | | Plant 02 | Pinus_pumila | Pandanas | LemmingTrails |
| Covariate08 | Fern | | Plant 03 | Salix_caprea | PapayaTree | HareFeces |
| Covariate09 | Diefenbachia | | Plant 04 | Abies_sachalinensis | Bamboo | FoxFeces |
| Covariate10 | Cycadaceae | | Plant 05 | Sorbaria_sorbifolia | BananaTree | ShorebirdFeces |
| Covariate11 | WalkingPalm | | Plant 06 | Maianthemum_dilatatum | Lianas | |
| Covariate12 | Crabholes | | Plant 07 | Calamagrostis_lansgdorfii | Orchids | |
| Covariate13 | | | Plant 08 | Daris_hexaphylla | FarmSpecies | |
| Covariate14 | | | Plant 09 | Spirea_betulifolia | PigTracks | |
| Covariate15 | | | Plant 10 | Equisetum | | |
| Covariate16 | | | Plant 11 | Lycopodium | | |
| Covariate17 | | | Plant 12 | Chamaepericlymenum_canadse | | |
| Covariate18 | | | Plant 13 | Lilium | | |
| Covariate19 | | | Plant 14 | Vaccinium_ovalifolium | | |
| Covariate20 | | | | Dryopteris | | |
| Covariate21 | | | | Ledum | | |
| Covariate22 | | | | Oxyria_digyna | | |
| Covariate23 | | | | Rhodococcum_vitis-idaea | | |
| Covariate24 | | | | Veratrum | | |
| Covariate25 | | | | Rubus_sachalinensis | | |
| Covariate26 | | | | Carex | | |
| Covariate27 | | | | Chamerion | | |
| Covariate28 | | | | UsneaLichen | | |
| Covariate29 | | | | AnimalBurrows | | |
| Covariate30 | | | | BearTrail | | |
| Covariate31 | | | | ScaleLichen | | |

7.3 DISTANCE Sampling Model Definitions

| Type | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extension |
|------|------------|----------|--------------------|------------------------|-----------------|----------------|
| Bi | 1CR | 1 | Half-normal | Cosine | none | none |
| Bi | 1CR | 2 | Half-normal | Hermite polynomial | none | none |
| Bi | 1CR | 3 | Uniform | Cosine | none | none |
| Bi | 1CR | 4 | Uniform | Simple polynomial | none | none |
| Bi | 1CR | 5 | Hazard-rate | Cosine | none | none |
| Bi | 1CR | 6 | Hazard-rate | Simple polynomial | none | none |
| Bi | 1CR | 7 | Half-normal | Cosine | VISIT | 1 |
| Bi | 1CR | 8 | Half-normal | Hermite polynomial | VISIT | 1 |
| Bi | 1CR | 9 | Hazard-rate | Cosine | VISIT | 1 |
| Bi | 1CR | 10 | Hazard-rate | Simple polynomial | VISIT | 1 |
| Bi | 1CR | 11 | Half-normal | Cosine | CLUSTER_SIZE | 2 |
| Bi | 1CR | 12 | Half-normal | Hermite polynomial | CLUSTER_SIZE | 2 |
| Bi | 1CR | 13 | Hazard-rate | Cosine | CLUSTER_SIZE | 2 |
| Bi | 1CR | 14 | Hazard-rate | Simple polynomial | CLUSTER_SIZE | 2 |
| Bi | 1CR | 15 | Half-normal | Cosine | IDENT | 3 |
| Bi | 1CR | 16 | Half-normal | Hermite polynomial | IDENT | 3 |
| Bi | 1CR | 17 | Hazard-rate | Cosine | IDENT | 3 |
| Bi | 1CR | 18 | Hazard-rate | Simple polynomial | IDENT | 3 |
| Bi | 1CR | 19 | Half-normal | Cosine | MINSINCEDAWN | 4 |
| Bi | 1CR | 20 | Half-normal | Hermite polynomial | MINSINCEDAWN | 4 |
| Bi | 1CR | 21 | Hazard-rate | Cosine | MINSINCEDAWN | 4 |
| Bi | 1CR | 22 | Hazard-rate | Simple polynomial | MINSINCEDAWN | 4 |
| Bi | 1CR | 23 | Half-normal | Cosine | HABITAT | 5 |
| Bi | 1CR | 24 | Half-normal | Hermite polynomial | HABITAT | 5 |
| Bi | 1CR | 25 | Hazard-rate | Cosine | HABITAT | 5 |
| Bi | 1CR | 26 | Hazard-rate | Simple polynomial | HABITAT | 5 |
| Bi | 1CR | 27 | Half-normal | Cosine | EPIPHYTESCAT | 6 |
| Bi | 1CR | 28 | Half-normal | Hermite polynomial | EPIPHYTESCAT | 6 |
| Bi | 1CR | 29 | Hazard-rate | Cosine | EPIPHYTESCAT | 6 |
| Bi | 1CR | 30 | Hazard-rate | Simple polynomial | EPIPHYTESCAT | 6 |
| Bi | 1CR | 31 | Half-normal | Cosine | MOSSLICHENCAT | 7 |
| Bi | 1CR | 32 | Half-normal | Hermite polynomial | MOSSLICHENCAT | 7 |
| Bi | 1CR | 33 | Hazard-rate | Cosine | MOSSLICHENCAT | 7 |
| Bi | 1CR | 34 | Hazard-rate | Simple polynomial | MOSSLICHENCAT | 7 |
| Bi | 1CR | 35 | Half-normal | Cosine | BARESOILPERC | 8 |
| Bi | 1CR | 36 | Half-normal | Hermite polynomial | BARESOILPERC | 8 |
| Bi | 1CR | 37 | Hazard-rate | Cosine | BARESOILPERC | 8 |
| Bi | 1CR | 38 | Hazard-rate | Simple polynomial | BARESOILPERC | 8 |
| Bi | 1CR | 39 | Half-normal | Cosine | DUFFCOVERPERC | 9 |
| Bi | 1CR | 40 | Half-normal | Hermite polynomial | DUFFCOVERPERC | 9 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|--------------------|----------------|
| Bi | 1CR | 41 | Hazard-rate | Cosine | DUFFCOVERPERC | 9 |
| Bi | 1CR | 42 | Hazard-rate | Simple polynomial | DUFFCOVERPERC | 9 |
| Bi | 1CR | 43 | Half-normal | Cosine | SHRUBSPERC135CM | 10 |
| Bi | 1CR | 44 | Half-normal | Hermite polynomial | SHRUBSPERC135CM | 10 |
| Bi | 1CR | 45 | Hazard-rate | Cosine | SHRUBSPERC135CM | 10 |
| Bi | 1CR | 46 | Hazard-rate | Simple polynomial | SHRUBSPERC135CM | 10 |
| Bi | 1CR | 47 | Half-normal | Cosine | CANOPYPERC | 11 |
| Bi | 1CR | 48 | Half-normal | Hermite polynomial | CANOPYPERC | 11 |
| Bi | 1CR | 49 | Hazard-rate | Cosine | CANOPYPERC | 11 |
| Bi | 1CR | 50 | Hazard-rate | Simple polynomial | CANOPYPERC | 11 |
| Bi | 1CR | 51 | Half-normal | Cosine | UNDERSTORYCOVER PE | 12 |
| Bi | 1CR | 52 | Half-normal | Hermite polynomial | UNDERSTORYCOVER PE | 12 |
| Bi | 1CR | 53 | Hazard-rate | Cosine | UNDERSTORYCOVER PE | 12 |
| Bi | 1CR | 54 | Hazard-rate | Simple polynomial | UNDERSTORYCOVER PE | 12 |
| Bi | 1CR | 55 | Half-normal | Cosine | LEAFBROWSINGPER C | 13 |
| Bi | 1CR | 56 | Half-normal | Hermite polynomial | LEAFBROWSINGPER C | 13 |
| Bi | 1CR | 57 | Hazard-rate | Cosine | LEAFBROWSINGPER C | 13 |
| Bi | 1CR | 58 | Hazard-rate | Simple polynomial | LEAFBROWSINGPER C | 13 |
| Bi | 1CR | 59 | Half-normal | Cosine | FLOWERSNO | 14 |
| Bi | 1CR | 60 | Half-normal | Hermite polynomial | FLOWERSNO | 14 |
| Bi | 1CR | 61 | Hazard-rate | Cosine | FLOWERSNO | 14 |
| Bi | 1CR | 62 | Hazard-rate | Simple polynomial | FLOWERSNO | 14 |
| Bi | 1CR | 63 | Half-normal | Cosine | CANOPYTREESNO | 15 |
| Bi | 1CR | 64 | Half-normal | Hermite polynomial | CANOPYTREESNO | 15 |
| Bi | 1CR | 65 | Hazard-rate | Cosine | CANOPYTREESNO | 15 |
| Bi | 1CR | 66 | Hazard-rate | Simple polynomial | CANOPYTREESNO | 15 |
| Bi | 1CR | 67 | Half-normal | Cosine | HIGHESTTREEM | 16 |
| Bi | 1CR | 68 | Half-normal | Hermite polynomial | HIGHESTTREEM | 16 |
| Bi | 1CR | 69 | Hazard-rate | Cosine | HIGHESTTREEM | 16 |
| Bi | 1CR | 70 | Hazard-rate | Simple polynomial | HIGHESTTREEM | 16 |
| Bi | 1CR | 71 | Half-normal | Cosine | HIGHESTDBHCM | 17 |
| Bi | 1CR | 72 | Half-normal | Hermite polynomial | HIGHESTDBHCM | 17 |
| Bi | 1CR | 73 | Hazard-rate | Cosine | HIGHESTDBHCM | 17 |
| Bi | 1CR | 74 | Hazard-rate | Simple polynomial | HIGHESTDBHCM | 17 |
| Bi | 1CR | 75 | Half-normal | Cosine | PLOT_TYPE | 18 |
| Bi | 1CR | 76 | Half-normal | Hermite polynomial | PLOT_TYPE | 18 |
| Bi | 1CR | 77 | Hazard-rate | Cosine | PLOT_TYPE | 18 |
| Bi | 1CR | 78 | Hazard-rate | Simple polynomial | PLOT_TYPE | 18 |
| Bi | 1CR | 79 | Half-normal | Cosine | COVARIATE04 | 28 |
| Bi | 1CR | 80 | Half-normal | Hermite | COVARIATE04 | 28 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDS covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|-----------------|----------------|
| | | | | polynomial | | |
| Bi | 1CR | 81 | Hazard-rate | Cosine | COVARIATE04 | 28 |
| Bi | 1CR | 82 | Hazard-rate | Simple polynomial | COVARIATE04 | 28 |
| Bi | 1CR | 83 | Half-normal | Cosine | COVARIATE05 | 29 |
| Bi | 1CR | 84 | Half-normal | Hermite polynomial | COVARIATE05 | 29 |
| Bi | 1CR | 85 | Hazard-rate | Cosine | COVARIATE05 | 29 |
| Bi | 1CR | 86 | Hazard-rate | Simple polynomial | COVARIATE05 | 29 |
| Bi | 1CR | 87 | Half-normal | Cosine | COVARIATE12 | 35 |
| Bi | 1CR | 88 | Half-normal | Hermite polynomial | COVARIATE12 | 35 |
| Bi | 1CR | 89 | Hazard-rate | Cosine | COVARIATE12 | 35 |
| Bi | 1CR | 90 | Hazard-rate | Simple polynomial | COVARIATE12 | 35 |
| Bi | 1CR | 91 | Half-normal | Cosine | MANAKIN | 36 |
| Bi | 1CR | 92 | Half-normal | Hermite polynomial | MANAKIN | 36 |
| Bi | 1CR | 93 | Hazard-rate | Cosine | MANAKIN | 36 |
| Bi | 1CR | 94 | Hazard-rate | Simple polynomial | MANAKIN | 36 |
| Bi | 1CR | 95 | Half-normal | Cosine | TURKEY_VULTURE | 37 |
| Bi | 1CR | 96 | Half-normal | Hermite polynomial | TURKEY_VULTURE | 37 |
| Bi | 1CR | 97 | Hazard-rate | Cosine | TURKEY_VULTURE | 37 |
| Bi | 1CR | 98 | Hazard-rate | Simple polynomial | TURKEY_VULTURE | 37 |
| Bi | 2Ni | 1 | Half-normal | Cosine | none | none |
| Bi | 2Ni | 2 | Half-normal | Hermite polynomial | none | none |
| Bi | 2Ni | 3 | Uniform | Cosine | none | none |
| Bi | 2Ni | 4 | Uniform | Simple polynomial | none | none |
| Bi | 2Ni | 5 | Hazard-rate | Cosine | none | none |
| Bi | 2Ni | 6 | Hazard-rate | Simple polynomial | none | none |
| Bi | 2Ni | 7 | Half-normal | Cosine | VISIT | 1 |
| Bi | 2Ni | 8 | Half-normal | Hermite polynomial | VISIT | 1 |
| Bi | 2Ni | 9 | Hazard-rate | Cosine | VISIT | 1 |
| Bi | 2Ni | 10 | Hazard-rate | Simple polynomial | VISIT | 1 |
| Bi | 2Ni | 11 | Half-normal | Cosine | CLUSTER_SIZE | 2 |
| Bi | 2Ni | 12 | Half-normal | Hermite polynomial | CLUSTER_SIZE | 2 |
| Bi | 2Ni | 13 | Hazard-rate | Cosine | CLUSTER_SIZE | 2 |
| Bi | 2Ni | 14 | Hazard-rate | Simple polynomial | CLUSTER_SIZE | 2 |
| Bi | 2Ni | 15 | Half-normal | Cosine | IDENT | 3 |
| Bi | 2Ni | 16 | Half-normal | Hermite polynomial | IDENT | 3 |
| Bi | 2Ni | 17 | Hazard-rate | Cosine | IDENT | 3 |
| Bi | 2Ni | 18 | Hazard-rate | Simple polynomial | IDENT | 3 |
| Bi | 2Ni | 19 | Half-normal | Cosine | MINSINCEDAWN | 4 |
| Bi | 2Ni | 20 | Half-normal | Hermite polynomial | MINSINCEDAWN | 4 |
| Bi | 2Ni | 21 | Hazard-rate | Cosine | MINSINCEDAWN | 4 |
| Bi | 2Ni | 22 | Hazard-rate | Simple polynomial | MINSINCEDAWN | 4 |
| Bi | 2Ni | 23 | Half-normal | Cosine | HABITAT | 5 |
| Bi | 2Ni | 24 | Half-normal | Hermite polynomial | HABITAT | 5 |
| Bi | 2Ni | 25 | Hazard-rate | Cosine | HABITAT | 5 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|--------------------|----------------|
| Bi | 2Ni | 26 | Hazard-rate | Simple polynomial | HABITAT | 5 |
| Bi | 2Ni | 27 | Half-normal | Cosine | EPIPHYTESCAT | 6 |
| Bi | 2Ni | 28 | Half-normal | Hermite polynomial | EPIPHYTESCAT | 6 |
| Bi | 2Ni | 29 | Hazard-rate | Cosine | EPIPHYTESCAT | 6 |
| Bi | 2Ni | 30 | Hazard-rate | Simple polynomial | EPIPHYTESCAT | 6 |
| Bi | 2Ni | 31 | Half-normal | Cosine | MOSSLICHENCAT | 7 |
| Bi | 2Ni | 32 | Half-normal | Hermite polynomial | MOSSLICHENCAT | 7 |
| Bi | 2Ni | 33 | Hazard-rate | Cosine | MOSSLICHENCAT | 7 |
| Bi | 2Ni | 34 | Hazard-rate | Simple polynomial | MOSSLICHENCAT | 7 |
| Bi | 2Ni | 35 | Half-normal | Cosine | BARESOILPERC | 8 |
| Bi | 2Ni | 36 | Half-normal | Hermite polynomial | BARESOILPERC | 8 |
| Bi | 2Ni | 37 | Hazard-rate | Cosine | BARESOILPERC | 8 |
| Bi | 2Ni | 38 | Hazard-rate | Simple polynomial | BARESOILPERC | 8 |
| Bi | 2Ni | 39 | Half-normal | Cosine | DUFFCOVERPERC | 9 |
| Bi | 2Ni | 40 | Half-normal | Hermite polynomial | DUFFCOVERPERC | 9 |
| Bi | 2Ni | 41 | Hazard-rate | Cosine | DUFFCOVERPERC | 9 |
| Bi | 2Ni | 42 | Hazard-rate | Simple polynomial | DUFFCOVERPERC | 9 |
| Bi | 2Ni | 43 | Half-normal | Cosine | SHRUBSPERC135CM | 10 |
| Bi | 2Ni | 44 | Half-normal | Hermite polynomial | SHRUBSPERC135CM | 10 |
| Bi | 2Ni | 45 | Hazard-rate | Cosine | SHRUBSPERC135CM | 10 |
| Bi | 2Ni | 46 | Hazard-rate | Simple polynomial | SHRUBSPERC135CM | 10 |
| Bi | 2Ni | 47 | Half-normal | Cosine | CANOPYPERC | 11 |
| Bi | 2Ni | 48 | Half-normal | Hermite polynomial | CANOPYPERC | 11 |
| Bi | 2Ni | 49 | Hazard-rate | Cosine | CANOPYPERC | 11 |
| Bi | 2Ni | 50 | Hazard-rate | Simple polynomial | CANOPYPERC | 11 |
| Bi | 2Ni | 51 | Half-normal | Cosine | UNDERSTORYCOVER PE | 12 |
| Bi | 2Ni | 52 | Half-normal | Hermite polynomial | UNDERSTORYCOVER PE | 12 |
| Bi | 2Ni | 53 | Hazard-rate | Cosine | UNDERSTORYCOVER PE | 12 |
| Bi | 2Ni | 54 | Hazard-rate | Simple polynomial | UNDERSTORYCOVER PE | 12 |
| Bi | 2Ni | 55 | Half-normal | Cosine | HIGHESTTREEM | 16 |
| Bi | 2Ni | 56 | Half-normal | Hermite polynomial | HIGHESTTREEM | 16 |
| Bi | 2Ni | 57 | Hazard-rate | Cosine | HIGHESTTREEM | 16 |
| Bi | 2Ni | 58 | Hazard-rate | Simple polynomial | HIGHESTTREEM | 16 |
| Bi | 2Ni | 59 | Half-normal | Cosine | HIGHESTDBHCM | 17 |
| Bi | 2Ni | 60 | Half-normal | Hermite polynomial | HIGHESTDBHCM | 17 |
| Bi | 2Ni | 61 | Hazard-rate | Cosine | HIGHESTDBHCM | 17 |
| Bi | 2Ni | 62 | Hazard-rate | Simple polynomial | HIGHESTDBHCM | 17 |
| Bi | 2Ni | 63 | Half-normal | Cosine | PLOT_TYPE | 18 |
| Bi | 2Ni | 64 | Half-normal | Hermite polynomial | PLOT_TYPE | 18 |
| Bi | 2Ni | 65 | Hazard-rate | Cosine | PLOT_TYPE | 18 |
| Bi | 2Ni | 66 | Hazard-rate | Simple polynomial | PLOT_TYPE | 18 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDS covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|-----------------|----------------|
| Bi | 2Ni | 67 | Half-normal | Cosine | TURKEY_VULTURE | 37 |
| Bi | 2Ni | 68 | Half-normal | Hermite polynomial | TURKEY_VULTURE | 37 |
| Bi | 2Ni | 69 | Hazard-rate | Cosine | TURKEY_VULTURE | 37 |
| Bi | 2Ni | 70 | Hazard-rate | Simple polynomial | TURKEY_VULTURE | 37 |
| Bi | 3AK | 1 | Half-normal | Cosine | none | none |
| Bi | 3AK | 2 | Half-normal | Hermite polynomial | none | none |
| Bi | 3AK | 3 | Uniform | Cosine | none | none |
| Bi | 3AK | 4 | Uniform | Simple polynomial | none | none |
| Bi | 3AK | 5 | Hazard-rate | Cosine | none | none |
| Bi | 3AK | 6 | Hazard-rate | Simple polynomial | none | none |
| Bi | 3AK | 7 | Half-normal | Cosine | IDENT | 3 |
| Bi | 3AK | 8 | Half-normal | Hermite polynomial | IDENT | 3 |
| Bi | 3AK | 9 | Hazard-rate | Cosine | IDENT | 3 |
| Bi | 3AK | 10 | Hazard-rate | Simple polynomial | IDENT | 3 |
| Bi | 3AK | 11 | Half-normal | Cosine | MINSINCEDAWN | 4 |
| Bi | 3AK | 12 | Half-normal | Hermite polynomial | MINSINCEDAWN | 4 |
| Bi | 3AK | 13 | Hazard-rate | Cosine | MINSINCEDAWN | 4 |
| Bi | 3AK | 14 | Hazard-rate | Simple polynomial | MINSINCEDAWN | 4 |
| Bi | 3AK | 15 | Half-normal | Cosine | HABITAT | 5 |
| Bi | 3AK | 16 | Half-normal | Hermite polynomial | HABITAT | 5 |
| Bi | 3AK | 17 | Hazard-rate | Cosine | HABITAT | 5 |
| Bi | 3AK | 18 | Hazard-rate | Simple polynomial | HABITAT | 5 |
| Bi | 3AK | 19 | Half-normal | Cosine | MOSSLICHENCAT | 7 |
| Bi | 3AK | 20 | Half-normal | Hermite polynomial | MOSSLICHENCAT | 7 |
| Bi | 3AK | 21 | Hazard-rate | Cosine | MOSSLICHENCAT | 7 |
| Bi | 3AK | 22 | Hazard-rate | Simple polynomial | MOSSLICHENCAT | 7 |
| Bi | 3AK | 23 | Half-normal | Cosine | DUFFCOVERPERC | 9 |
| Bi | 3AK | 24 | Half-normal | Hermite polynomial | DUFFCOVERPERC | 9 |
| Bi | 3AK | 25 | Hazard-rate | Cosine | DUFFCOVERPERC | 9 |
| Bi | 3AK | 26 | Hazard-rate | Simple polynomial | DUFFCOVERPERC | 9 |
| Bi | 3AK | 27 | Half-normal | Cosine | CANOPYPERC | 11 |
| Bi | 3AK | 28 | Half-normal | Hermite polynomial | CANOPYPERC | 11 |
| Bi | 3AK | 29 | Hazard-rate | Cosine | CANOPYPERC | 11 |
| Bi | 3AK | 30 | Hazard-rate | Simple polynomial | CANOPYPERC | 11 |
| Bi | 3AK | 31 | Half-normal | Cosine | CANOPYTREESNO | 15 |
| Bi | 3AK | 32 | Half-normal | Hermite polynomial | CANOPYTREESNO | 15 |
| Bi | 3AK | 33 | Hazard-rate | Cosine | CANOPYTREESNO | 15 |
| Bi | 3AK | 34 | Hazard-rate | Simple polynomial | CANOPYTREESNO | 15 |
| Bi | 3AK | 35 | Half-normal | Cosine | HIGHESTTREEM | 16 |
| Bi | 3AK | 36 | Half-normal | Hermite polynomial | HIGHESTTREEM | 16 |
| Bi | 3AK | 37 | Hazard-rate | Cosine | HIGHESTTREEM | 16 |
| Bi | 3AK | 38 | Hazard-rate | Simple polynomial | HIGHESTTREEM | 16 |
| Bi | 3AK | 39 | Half-normal | Cosine | HIGHESTDBHCM | 17 |
| Bi | 3AK | 40 | Half-normal | Hermite | HIGHESTDBHCM | 17 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDS covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|-----------------|----------------|
| | | | | polynomial | | |
| Bi | 3AK | 41 | Hazard-rate | Cosine | HIGHESTDBHCM | 17 |
| Bi | 3AK | 42 | Hazard-rate | Simple polynomial | HIGHESTDBHCM | 17 |
| Bi | 3AK | 43 | Half-normal | Cosine | COVARIATE01 | 25 |
| Bi | 3AK | 44 | Half-normal | Hermite polynomial | COVARIATE01 | 25 |
| Bi | 3AK | 45 | Hazard-rate | Cosine | COVARIATE01 | 25 |
| Bi | 3AK | 46 | Hazard-rate | Simple polynomial | COVARIATE01 | 25 |
| Bi | 3AK | 47 | Half-normal | Cosine | COVARIATE07 | 31 |
| Bi | 3AK | 48 | Half-normal | Hermite polynomial | COVARIATE07 | 31 |
| Bi | 3AK | 49 | Hazard-rate | Cosine | COVARIATE07 | 31 |
| Bi | 3AK | 50 | Hazard-rate | Simple polynomial | COVARIATE07 | 31 |
| Bi | 3AK | 51 | Half-normal | Cosine | COVARIATE11 | 34 |
| Bi | 3AK | 52 | Half-normal | Hermite polynomial | COVARIATE11 | 34 |
| Bi | 3AK | 53 | Hazard-rate | Cosine | COVARIATE11 | 34 |
| Bi | 3AK | 54 | Hazard-rate | Simple polynomial | COVARIATE11 | 34 |
| Bi | 3AK | 55 | Half-normal | Cosine | COVARIATE12 | 35 |
| Bi | 3AK | 56 | Half-normal | Hermite polynomial | COVARIATE12 | 35 |
| Bi | 3AK | 57 | Hazard-rate | Cosine | COVARIATE12 | 35 |
| Bi | 3AK | 58 | Hazard-rate | Simple polynomial | COVARIATE12 | 35 |
| Bi | 3AK | 59 | Half-normal | Cosine | COVARIATE13 | 38 |
| Bi | 3AK | 60 | Half-normal | Hermite polynomial | COVARIATE13 | 38 |
| Bi | 3AK | 61 | Hazard-rate | Cosine | COVARIATE13 | 38 |
| Bi | 3AK | 62 | Hazard-rate | Simple polynomial | COVARIATE13 | 38 |
| Bi | 3AK | 63 | Half-normal | Cosine | COVARIATE14 | 39 |
| Bi | 3AK | 64 | Half-normal | Hermite polynomial | COVARIATE14 | 39 |
| Bi | 3AK | 65 | Hazard-rate | Cosine | COVARIATE14 | 39 |
| Bi | 3AK | 66 | Hazard-rate | Simple polynomial | COVARIATE14 | 39 |
| Bi | 3AK | 67 | Half-normal | Cosine | COVARIATE19 | 43 |
| Bi | 3AK | 68 | Half-normal | Hermite polynomial | COVARIATE19 | 43 |
| Bi | 3AK | 69 | Hazard-rate | Cosine | COVARIATE19 | 43 |
| Bi | 3AK | 70 | Hazard-rate | Simple polynomial | COVARIATE19 | 43 |
| Bi | 3AK | 71 | Half-normal | Cosine | SQUIRREL | 45 |
| Bi | 3AK | 72 | Half-normal | Hermite polynomial | SQUIRREL | 45 |
| Bi | 3AK | 73 | Hazard-rate | Cosine | SQUIRREL | 45 |
| Bi | 3AK | 74 | Hazard-rate | Simple polynomial | SQUIRREL | 45 |
| Bi | 4Ru | 1 | Half-normal | Cosine | none | none |
| Bi | 4Ru | 2 | Half-normal | Hermite polynomial | none | none |
| Bi | 4Ru | 3 | Uniform | Cosine | none | none |
| Bi | 4Ru | 4 | Uniform | Simple polynomial | none | none |
| Bi | 4Ru | 5 | Hazard-rate | Cosine | none | none |
| Bi | 4Ru | 6 | Hazard-rate | Simple polynomial | none | none |
| Bi | 4Ru | 7 | Half-normal | Cosine | VISIT | 1 |
| Bi | 4Ru | 8 | Half-normal | Hermite polynomial | VISIT | 1 |
| Bi | 4Ru | 9 | Hazard-rate | Cosine | VISIT | 1 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|--------------------|----------------|
| Bi | 4Ru | 10 | Hazard-rate | Simple polynomial | VISIT | 1 |
| Bi | 4Ru | 11 | Half-normal | Cosine | CLUSTER_SIZE | 2 |
| Bi | 4Ru | 12 | Half-normal | Hermite polynomial | CLUSTER_SIZE | 2 |
| Bi | 4Ru | 13 | Hazard-rate | Cosine | CLUSTER_SIZE | 2 |
| Bi | 4Ru | 14 | Hazard-rate | Simple polynomial | CLUSTER_SIZE | 2 |
| Bi | 4Ru | 15 | Half-normal | Cosine | MINSINCEDAWN | 4 |
| Bi | 4Ru | 16 | Half-normal | Hermite polynomial | MINSINCEDAWN | 4 |
| Bi | 4Ru | 17 | Hazard-rate | Cosine | MINSINCEDAWN | 4 |
| Bi | 4Ru | 18 | Hazard-rate | Simple polynomial | MINSINCEDAWN | 4 |
| Bi | 4Ru | 19 | Half-normal | Cosine | HABITAT | 5 |
| Bi | 4Ru | 20 | Half-normal | Hermite polynomial | HABITAT | 5 |
| Bi | 4Ru | 21 | Hazard-rate | Cosine | HABITAT | 5 |
| Bi | 4Ru | 22 | Hazard-rate | Simple polynomial | HABITAT | 5 |
| Bi | 4Ru | 23 | Half-normal | Cosine | DUFFCOVERPERC | 9 |
| Bi | 4Ru | 24 | Half-normal | Hermite polynomial | DUFFCOVERPERC | 9 |
| Bi | 4Ru | 25 | Hazard-rate | Cosine | DUFFCOVERPERC | 9 |
| Bi | 4Ru | 26 | Hazard-rate | Simple polynomial | DUFFCOVERPERC | 9 |
| Bi | 4Ru | 27 | Half-normal | Cosine | SHRUBSPERC135CM | 10 |
| Bi | 4Ru | 28 | Half-normal | Hermite polynomial | SHRUBSPERC135CM | 10 |
| Bi | 4Ru | 29 | Hazard-rate | Cosine | SHRUBSPERC135CM | 10 |
| Bi | 4Ru | 30 | Hazard-rate | Simple polynomial | SHRUBSPERC135CM | 10 |
| Bi | 4Ru | 31 | Half-normal | Cosine | CANOPYPERC | 11 |
| Bi | 4Ru | 32 | Half-normal | Hermite polynomial | CANOPYPERC | 11 |
| Bi | 4Ru | 33 | Hazard-rate | Cosine | CANOPYPERC | 11 |
| Bi | 4Ru | 34 | Hazard-rate | Simple polynomial | CANOPYPERC | 11 |
| Bi | 4Ru | 35 | Half-normal | Cosine | UNDERSTORYCOVER PE | 12 |
| Bi | 4Ru | 36 | Half-normal | Hermite polynomial | UNDERSTORYCOVER PE | 12 |
| Bi | 4Ru | 37 | Hazard-rate | Cosine | UNDERSTORYCOVER PE | 12 |
| Bi | 4Ru | 38 | Hazard-rate | Simple polynomial | UNDERSTORYCOVER PE | 12 |
| Bi | 4Ru | 39 | Half-normal | Cosine | LEAFBROWSINGPERC | 13 |
| Bi | 4Ru | 40 | Half-normal | Hermite polynomial | LEAFBROWSINGPERC | 13 |
| Bi | 4Ru | 41 | Hazard-rate | Cosine | LEAFBROWSINGPERC | 13 |
| Bi | 4Ru | 42 | Hazard-rate | Simple polynomial | LEAFBROWSINGPERC | 13 |
| Bi | 4Ru | 43 | Half-normal | Cosine | FLOWERSNO | 14 |
| Bi | 4Ru | 44 | Half-normal | Hermite polynomial | FLOWERSNO | 14 |
| Bi | 4Ru | 45 | Hazard-rate | Cosine | FLOWERSNO | 14 |
| Bi | 4Ru | 46 | Hazard-rate | Simple polynomial | FLOWERSNO | 14 |
| Bi | 4Ru | 47 | Half-normal | Cosine | CANOPYTREESNO | 15 |
| Bi | 4Ru | 48 | Half-normal | Hermite polynomial | CANOPYTREESNO | 15 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|-----------------|----------------|
| Bi | 4Ru | 49 | Hazard-rate | Cosine | CANOPYTREESNO | 15 |
| Bi | 4Ru | 50 | Hazard-rate | Simple polynomial | CANOPYTREESNO | 15 |
| Bi | 4Ru | 51 | Half-normal | Cosine | HIGHESTTREEM | 16 |
| Bi | 4Ru | 52 | Half-normal | Hermite polynomial | HIGHESTTREEM | 16 |
| Bi | 4Ru | 53 | Hazard-rate | Cosine | HIGHESTTREEM | 16 |
| Bi | 4Ru | 54 | Hazard-rate | Simple polynomial | HIGHESTTREEM | 16 |
| Bi | 4Ru | 55 | Half-normal | Cosine | HIGHESTDBHCM | 17 |
| Bi | 4Ru | 56 | Half-normal | Hermite polynomial | HIGHESTDBHCM | 17 |
| Bi | 4Ru | 57 | Hazard-rate | Cosine | HIGHESTDBHCM | 17 |
| Bi | 4Ru | 58 | Hazard-rate | Simple polynomial | HIGHESTDBHCM | 17 |
| Bi | 4Ru | 59 | Half-normal | Cosine | PLOT_TYPE | 18 |
| Bi | 4Ru | 60 | Half-normal | Hermite polynomial | PLOT_TYPE | 18 |
| Bi | 4Ru | 61 | Hazard-rate | Cosine | PLOT_TYPE | 18 |
| Bi | 4Ru | 62 | Hazard-rate | Simple polynomial | PLOT_TYPE | 18 |
| Bi | 4Ru | 63 | Half-normal | Cosine | MOSSPERC | 19 |
| Bi | 4Ru | 64 | Half-normal | Hermite polynomial | MOSSPERC | 19 |
| Bi | 4Ru | 65 | Hazard-rate | Cosine | MOSSPERC | 19 |
| Bi | 4Ru | 66 | Hazard-rate | Simple polynomial | MOSSPERC | 19 |
| Bi | 4Ru | 67 | Half-normal | Cosine | LICHENPERC | 20 |
| Bi | 4Ru | 68 | Half-normal | Hermite polynomial | LICHENPERC | 20 |
| Bi | 4Ru | 69 | Hazard-rate | Cosine | LICHENPERC | 20 |
| Bi | 4Ru | 70 | Hazard-rate | Simple polynomial | LICHENPERC | 20 |
| Bi | 4Ru | 71 | Half-normal | Cosine | COVARIATE01 | 25 |
| Bi | 4Ru | 72 | Half-normal | Hermite polynomial | COVARIATE01 | 25 |
| Bi | 4Ru | 73 | Hazard-rate | Cosine | COVARIATE01 | 25 |
| Bi | 4Ru | 74 | Hazard-rate | Simple polynomial | COVARIATE01 | 25 |
| Bi | 4Ru | 75 | Half-normal | Cosine | COVARIATE04 | 28 |
| Bi | 4Ru | 76 | Half-normal | Hermite polynomial | COVARIATE04 | 28 |
| Bi | 4Ru | 77 | Hazard-rate | Cosine | COVARIATE04 | 28 |
| Bi | 4Ru | 78 | Hazard-rate | Simple polynomial | COVARIATE04 | 28 |
| Bi | 4Ru | 79 | Half-normal | Cosine | COVARIATE05 | 29 |
| Bi | 4Ru | 80 | Half-normal | Hermite polynomial | COVARIATE05 | 29 |
| Bi | 4Ru | 81 | Hazard-rate | Cosine | COVARIATE05 | 29 |
| Bi | 4Ru | 82 | Hazard-rate | Simple polynomial | COVARIATE05 | 29 |
| Bi | 4Ru | 83 | Half-normal | Cosine | COVARIATE08 | 32 |
| Bi | 4Ru | 84 | Half-normal | Hermite polynomial | COVARIATE08 | 32 |
| Bi | 4Ru | 85 | Hazard-rate | Cosine | COVARIATE08 | 32 |
| Bi | 4Ru | 86 | Hazard-rate | Simple polynomial | COVARIATE08 | 32 |
| Bi | 4Ru | 87 | Half-normal | Cosine | COVARIATE12 | 35 |
| Bi | 4Ru | 88 | Half-normal | Hermite polynomial | COVARIATE12 | 35 |
| Bi | 4Ru | 89 | Hazard-rate | Cosine | COVARIATE12 | 35 |
| Bi | 4Ru | 90 | Hazard-rate | Simple polynomial | COVARIATE12 | 35 |
| Bi | 4Ru | 91 | Half-normal | Cosine | COVARIATE15 | 40 |
| Bi | 4Ru | 92 | Half-normal | Hermite | COVARIATE15 | 40 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDS covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|-----------------|----------------|
| | | | | polynomial | | |
| Bi | 4Ru | 93 | Hazard-rate | Cosine | COVARIATE15 | 40 |
| Bi | 4Ru | 94 | Hazard-rate | Simple polynomial | COVARIATE15 | 40 |
| Bi | 4Ru | 95 | Half-normal | Cosine | COVARIATE16 | 41 |
| Bi | 4Ru | 96 | Half-normal | Hermite polynomial | COVARIATE16 | 41 |
| Bi | 4Ru | 97 | Hazard-rate | Cosine | COVARIATE16 | 41 |
| Bi | 4Ru | 98 | Hazard-rate | Simple polynomial | COVARIATE16 | 41 |
| Bi | 4Ru | 99 | Half-normal | Cosine | COVARIATE18 | 42 |
| Bi | 4Ru | 100 | Half-normal | Hermite polynomial | COVARIATE18 | 42 |
| Bi | 4Ru | 101 | Hazard-rate | Cosine | COVARIATE18 | 42 |
| Bi | 4Ru | 102 | Hazard-rate | Simple polynomial | COVARIATE18 | 42 |
| Bi | 4Ru | 103 | Half-normal | Cosine | COVARIATE20 | 44 |
| Bi | 4Ru | 104 | Half-normal | Hermite polynomial | COVARIATE20 | 44 |
| Bi | 4Ru | 105 | Hazard-rate | Cosine | COVARIATE20 | 44 |
| Bi | 4Ru | 106 | Hazard-rate | Simple polynomial | COVARIATE20 | 44 |
| Bi | 4Ru | 107 | Half-normal | Cosine | COVARIATE21 | 46 |
| Bi | 4Ru | 108 | Half-normal | Hermite polynomial | COVARIATE21 | 46 |
| Bi | 4Ru | 109 | Hazard-rate | Cosine | COVARIATE21 | 46 |
| Bi | 4Ru | 110 | Hazard-rate | Simple polynomial | COVARIATE21 | 46 |
| Bi | 4Ru | 111 | Half-normal | Cosine | COVARIATE23 | 47 |
| Bi | 4Ru | 112 | Half-normal | Hermite polynomial | COVARIATE23 | 47 |
| Bi | 4Ru | 113 | Hazard-rate | Cosine | COVARIATE23 | 47 |
| Bi | 4Ru | 114 | Hazard-rate | Simple polynomial | COVARIATE23 | 47 |
| Bi | 4Ru | 115 | Half-normal | Cosine | COVARIATE28 | 48 |
| Bi | 4Ru | 116 | Half-normal | Hermite polynomial | COVARIATE28 | 48 |
| Bi | 4Ru | 117 | Hazard-rate | Cosine | COVARIATE28 | 48 |
| Bi | 4Ru | 118 | Hazard-rate | Simple polynomial | COVARIATE28 | 48 |
| Bi | 4Ru | 119 | Half-normal | Cosine | COVARIATE30 | 49 |
| Bi | 4Ru | 120 | Half-normal | Hermite polynomial | COVARIATE30 | 49 |
| Bi | 4Ru | 121 | Hazard-rate | Cosine | COVARIATE30 | 49 |
| Bi | 4Ru | 122 | Hazard-rate | Simple polynomial | COVARIATE30 | 49 |
| Bi | 4Ru | 123 | Half-normal | Cosine | COVARIATE31 | 50 |
| Bi | 4Ru | 124 | Half-normal | Hermite polynomial | COVARIATE31 | 50 |
| Bi | 4Ru | 125 | Hazard-rate | Cosine | COVARIATE31 | 50 |
| Bi | 4Ru | 126 | Hazard-rate | Simple polynomial | COVARIATE31 | 50 |
| Bi | 4Ru | 127 | Half-normal | Cosine | WIZE | 51 |
| Bi | 4Ru | 128 | Half-normal | Hermite polynomial | WIZE | 51 |
| Bi | 4Ru | 129 | Hazard-rate | Cosine | WIZE | 51 |
| Bi | 4Ru | 130 | Hazard-rate | Simple polynomial | WIZE | 51 |
| Bi | 5PG | 1 | Half-normal | Cosine | none | none |
| Bi | 5PG | 2 | Half-normal | Hermite polynomial | none | none |
| Bi | 5PG | 3 | Uniform | Cosine | none | none |
| Bi | 5PG | 4 | Uniform | Simple polynomial | none | none |
| Bi | 5PG | 5 | Hazard-rate | Cosine | none | none |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDS covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|-----------------|----------------|
| Bi | 5PG | 6 | Hazard-rate | Simple polynomial | none | none |
| Bi | 5PG | 7 | Half-normal | Cosine | VISIT | 1 |
| Bi | 5PG | 8 | Half-normal | Hermite polynomial | VISIT | 1 |
| Bi | 5PG | 9 | Hazard-rate | Cosine | VISIT | 1 |
| Bi | 5PG | 10 | Hazard-rate | Simple polynomial | VISIT | 1 |
| Bi | 5PG | 11 | Half-normal | Cosine | CLUSTER_SIZE | 2 |
| Bi | 5PG | 12 | Half-normal | Hermite polynomial | CLUSTER_SIZE | 2 |
| Bi | 5PG | 13 | Hazard-rate | Cosine | CLUSTER_SIZE | 2 |
| Bi | 5PG | 14 | Hazard-rate | Simple polynomial | CLUSTER_SIZE | 2 |
| Bi | 5PG | 15 | Half-normal | Cosine | IDENT | 3 |
| Bi | 5PG | 16 | Half-normal | Hermite polynomial | IDENT | 3 |
| Bi | 5PG | 17 | Hazard-rate | Cosine | IDENT | 3 |
| Bi | 5PG | 18 | Hazard-rate | Simple polynomial | IDENT | 3 |
| Bi | 5PG | 19 | Half-normal | Cosine | MINSINCEDAWN | 4 |
| Bi | 5PG | 20 | Half-normal | Hermite polynomial | MINSINCEDAWN | 4 |
| Bi | 5PG | 21 | Hazard-rate | Cosine | MINSINCEDAWN | 4 |
| Bi | 5PG | 22 | Hazard-rate | Simple polynomial | MINSINCEDAWN | 4 |
| Bi | 5PG | 23 | Half-normal | Cosine | HABITAT | 5 |
| Bi | 5PG | 24 | Half-normal | Hermite polynomial | HABITAT | 5 |
| Bi | 5PG | 25 | Hazard-rate | Cosine | HABITAT | 5 |
| Bi | 5PG | 26 | Hazard-rate | Simple polynomial | HABITAT | 5 |
| Bi | 5PG | 27 | Half-normal | Cosine | EPIPHYTESCAT | 6 |
| Bi | 5PG | 28 | Half-normal | Hermite polynomial | EPIPHYTESCAT | 6 |
| Bi | 5PG | 29 | Hazard-rate | Cosine | EPIPHYTESCAT | 6 |
| Bi | 5PG | 30 | Hazard-rate | Simple polynomial | EPIPHYTESCAT | 6 |
| Bi | 5PG | 31 | Half-normal | Cosine | BARESOILPERC | 8 |
| Bi | 5PG | 32 | Half-normal | Hermite polynomial | BARESOILPERC | 8 |
| Bi | 5PG | 33 | Hazard-rate | Cosine | BARESOILPERC | 8 |
| Bi | 5PG | 34 | Hazard-rate | Simple polynomial | BARESOILPERC | 8 |
| Bi | 5PG | 35 | Half-normal | Cosine | DUFFCOVERPERC | 9 |
| Bi | 5PG | 36 | Half-normal | Hermite polynomial | DUFFCOVERPERC | 9 |
| Bi | 5PG | 37 | Hazard-rate | Cosine | DUFFCOVERPERC | 9 |
| Bi | 5PG | 38 | Hazard-rate | Simple polynomial | DUFFCOVERPERC | 9 |
| Bi | 5PG | 39 | Half-normal | Cosine | CANOPYPERC | 11 |
| Bi | 5PG | 40 | Half-normal | Hermite polynomial | CANOPYPERC | 11 |
| Bi | 5PG | 41 | Hazard-rate | Cosine | CANOPYPERC | 11 |
| Bi | 5PG | 42 | Hazard-rate | Simple polynomial | CANOPYPERC | 11 |
| Bi | 5PG | 43 | Half-normal | Cosine | HIGHESTTREEM | 16 |
| Bi | 5PG | 44 | Half-normal | Hermite polynomial | HIGHESTTREEM | 16 |
| Bi | 5PG | 45 | Hazard-rate | Cosine | HIGHESTTREEM | 16 |
| Bi | 5PG | 46 | Hazard-rate | Simple polynomial | HIGHESTTREEM | 16 |
| Bi | 5PG | 47 | Half-normal | Cosine | HIGHESTDBHCM | 17 |
| Bi | 5PG | 48 | Half-normal | Hermite polynomial | HIGHESTDBHCM | 17 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|-----------------|----------------|
| Bi | 5PG | 49 | Hazard-rate | Cosine | HIGHESTDBHCM | 17 |
| Bi | 5PG | 50 | Hazard-rate | Simple polynomial | HIGHESTDBHCM | 17 |
| Bi | 5PG | 51 | Half-normal | Cosine | COVARIATE01 | 25 |
| Bi | 5PG | 52 | Half-normal | Hermite polynomial | COVARIATE01 | 25 |
| Bi | 5PG | 53 | Hazard-rate | Cosine | COVARIATE01 | 25 |
| Bi | 5PG | 54 | Hazard-rate | Simple polynomial | COVARIATE01 | 25 |
| Bi | 5PG | 55 | Half-normal | Cosine | COVARIATE06 | 30 |
| Bi | 5PG | 56 | Half-normal | Hermite polynomial | COVARIATE06 | 30 |
| Bi | 5PG | 57 | Hazard-rate | Cosine | COVARIATE06 | 30 |
| Bi | 5PG | 58 | Hazard-rate | Simple polynomial | COVARIATE06 | 30 |
| Bi | 5PG | 59 | Half-normal | Cosine | COVARIATE11 | 34 |
| Bi | 5PG | 60 | Half-normal | Hermite polynomial | COVARIATE11 | 34 |
| Bi | 5PG | 61 | Hazard-rate | Cosine | COVARIATE11 | 34 |
| Bi | 5PG | 62 | Hazard-rate | Simple polynomial | COVARIATE11 | 34 |
| Bi | 5PG | 63 | Half-normal | Cosine | COVARIATE12 | 35 |
| Bi | 5PG | 64 | Half-normal | Hermite polynomial | COVARIATE12 | 35 |
| Bi | 5PG | 65 | Hazard-rate | Cosine | COVARIATE12 | 35 |
| Bi | 5PG | 66 | Hazard-rate | Simple polynomial | COVARIATE12 | 35 |
| Bi | 6Ba | 1 | Half-normal | Cosine | none | none |
| Bi | 6Ba | 2 | Half-normal | Hermite polynomial | none | none |
| Bi | 6Ba | 3 | Uniform | Cosine | none | none |
| Bi | 6Ba | 4 | Uniform | Simple polynomial | none | none |
| Bi | 6Ba | 5 | Hazard-rate | Cosine | none | none |
| Bi | 6Ba | 6 | Hazard-rate | Simple polynomial | none | none |
| Bi | 6Ba | 7 | Half-normal | Cosine | VISIT | 1 |
| Bi | 6Ba | 8 | Half-normal | Hermite polynomial | VISIT | 1 |
| Bi | 6Ba | 9 | Hazard-rate | Cosine | VISIT | 1 |
| Bi | 6Ba | 10 | Hazard-rate | Simple polynomial | VISIT | 1 |
| Bi | 6Ba | 11 | Half-normal | Cosine | CLUSTER_SIZE | 2 |
| Bi | 6Ba | 12 | Half-normal | Hermite polynomial | CLUSTER_SIZE | 2 |
| Bi | 6Ba | 13 | Hazard-rate | Cosine | CLUSTER_SIZE | 2 |
| Bi | 6Ba | 14 | Hazard-rate | Simple polynomial | CLUSTER_SIZE | 2 |
| Bi | 6Ba | 15 | Half-normal | Cosine | FLOWERSNO | 14 |
| Bi | 6Ba | 16 | Half-normal | Hermite polynomial | FLOWERSNO | 14 |
| Bi | 6Ba | 17 | Hazard-rate | Cosine | FLOWERSNO | 14 |
| Bi | 6Ba | 18 | Hazard-rate | Simple polynomial | FLOWERSNO | 14 |
| Bi | 6Ba | 19 | Half-normal | Cosine | PLOT_TYPE | 18 |
| Bi | 6Ba | 20 | Half-normal | Hermite polynomial | PLOT_TYPE | 18 |
| Bi | 6Ba | 21 | Hazard-rate | Cosine | PLOT_TYPE | 18 |
| Bi | 6Ba | 22 | Hazard-rate | Simple polynomial | PLOT_TYPE | 18 |
| Bi | 6Ba | 23 | Half-normal | Cosine | MOSSPERC | 19 |
| Bi | 6Ba | 24 | Half-normal | Hermite polynomial | MOSSPERC | 19 |
| Bi | 6Ba | 25 | Hazard-rate | Cosine | MOSSPERC | 19 |
| Bi | 6Ba | 26 | Hazard-rate | Simple polynomial | MOSSPERC | 19 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDS covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|-----------------|----------------|
| Bi | 6Ba | 27 | Half-normal | Cosine | LICHENPERC | 20 |
| Bi | 6Ba | 28 | Half-normal | Hermite polynomial | LICHENPERC | 20 |
| Bi | 6Ba | 29 | Hazard-rate | Cosine | LICHENPERC | 20 |
| Bi | 6Ba | 30 | Hazard-rate | Simple polynomial | LICHENPERC | 20 |
| Bi | 6Ba | 31 | Half-normal | Cosine | LEAFS | 21 |
| Bi | 6Ba | 32 | Half-normal | Hermite polynomial | LEAFS | 21 |
| Bi | 6Ba | 33 | Hazard-rate | Cosine | LEAFS | 21 |
| Bi | 6Ba | 34 | Hazard-rate | Simple polynomial | LEAFS | 21 |
| Bi | 6Ba | 35 | Half-normal | Cosine | DIAMNEXTLAKE | 22 |
| Bi | 6Ba | 36 | Half-normal | Hermite polynomial | DIAMNEXTLAKE | 22 |
| Bi | 6Ba | 37 | Hazard-rate | Cosine | DIAMNEXTLAKE | 22 |
| Bi | 6Ba | 38 | Hazard-rate | Simple polynomial | DIAMNEXTLAKE | 22 |
| Bi | 6Ba | 39 | Half-normal | Cosine | DISTNEXTLAKE | 23 |
| Bi | 6Ba | 40 | Half-normal | Hermite polynomial | DISTNEXTLAKE | 23 |
| Bi | 6Ba | 41 | Hazard-rate | Cosine | DISTNEXTLAKE | 23 |
| Bi | 6Ba | 42 | Hazard-rate | Simple polynomial | DISTNEXTLAKE | 23 |
| Bi | 6Ba | 43 | Half-normal | Cosine | GRASSPERC | 24 |
| Bi | 6Ba | 44 | Half-normal | Hermite polynomial | GRASSPERC | 24 |
| Bi | 6Ba | 45 | Hazard-rate | Cosine | GRASSPERC | 24 |
| Bi | 6Ba | 46 | Hazard-rate | Simple polynomial | GRASSPERC | 24 |
| Bi | 6Ba | 47 | Half-normal | Cosine | COVARIATE01 | 25 |
| Bi | 6Ba | 48 | Half-normal | Hermite polynomial | COVARIATE01 | 25 |
| Bi | 6Ba | 49 | Hazard-rate | Cosine | COVARIATE01 | 25 |
| Bi | 6Ba | 50 | Hazard-rate | Simple polynomial | COVARIATE01 | 25 |
| Bi | 6Ba | 51 | Half-normal | Cosine | COVARIATE02 | 26 |
| Bi | 6Ba | 52 | Half-normal | Hermite polynomial | COVARIATE02 | 26 |
| Bi | 6Ba | 53 | Hazard-rate | Cosine | COVARIATE02 | 26 |
| Bi | 6Ba | 54 | Hazard-rate | Simple polynomial | COVARIATE02 | 26 |
| Bi | 6Ba | 55 | Half-normal | Cosine | COVARIATE03 | 27 |
| Bi | 6Ba | 56 | Half-normal | Hermite polynomial | COVARIATE03 | 27 |
| Bi | 6Ba | 57 | Hazard-rate | Cosine | COVARIATE03 | 27 |
| Bi | 6Ba | 58 | Hazard-rate | Simple polynomial | COVARIATE03 | 27 |
| Bi | 6Ba | 59 | Half-normal | Cosine | COVARIATE05 | 29 |
| Bi | 6Ba | 60 | Half-normal | Hermite polynomial | COVARIATE05 | 29 |
| Bi | 6Ba | 61 | Hazard-rate | Cosine | COVARIATE05 | 29 |
| Bi | 6Ba | 62 | Hazard-rate | Simple polynomial | COVARIATE05 | 29 |
| Bi | 6Ba | 63 | Half-normal | Cosine | COVARIATE06 | 30 |
| Bi | 6Ba | 64 | Half-normal | Hermite polynomial | COVARIATE06 | 30 |
| Bi | 6Ba | 65 | Hazard-rate | Cosine | COVARIATE06 | 30 |
| Bi | 6Ba | 66 | Hazard-rate | Simple polynomial | COVARIATE06 | 30 |
| Bi | 6Ba | 67 | Half-normal | Cosine | COVARIATE07 | 31 |
| Bi | 6Ba | 68 | Half-normal | Hermite polynomial | COVARIATE07 | 31 |
| Bi | 6Ba | 69 | Hazard-rate | Cosine | COVARIATE07 | 31 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDS covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|-----------------|----------------|
| Bi | 6Ba | 70 | Hazard-rate | Simple polynomial | COVARIATE07 | 31 |
| Bi | 6Ba | 71 | Half-normal | Cosine | COVARIATE08 | 32 |
| Bi | 6Ba | 72 | Half-normal | Hermite polynomial | COVARIATE08 | 32 |
| Bi | 6Ba | 73 | Hazard-rate | Cosine | COVARIATE08 | 32 |
| Bi | 6Ba | 74 | Hazard-rate | Simple polynomial | COVARIATE08 | 32 |
| Bi | 6Ba | 75 | Half-normal | Cosine | COVARIATE10 | 33 |
| Bi | 6Ba | 76 | Half-normal | Hermite polynomial | COVARIATE10 | 33 |
| Bi | 6Ba | 77 | Hazard-rate | Cosine | COVARIATE10 | 33 |
| Bi | 6Ba | 78 | Hazard-rate | Simple polynomial | COVARIATE10 | 33 |
| Bi | 6Ba | 79 | Half-normal | Cosine | POMARINE_JAEGER | 52 |
| Bi | 6Ba | 80 | Half-normal | Hermite polynomial | POMARINE_JAEGER | 52 |
| Bi | 6Ba | 81 | Hazard-rate | Cosine | POMARINE_JAEGER | 52 |
| Bi | 6Ba | 82 | Hazard-rate | Simple polynomial | POMARINE_JAEGER | 52 |
| TW | 1CR | 1 | Half-normal | Cosine | none | none |
| TW | 1CR | 2 | Half-normal | Hermite polynomial | none | none |
| TW | 1CR | 3 | Uniform | Cosine | none | none |
| TW | 1CR | 4 | Uniform | Simple polynomial | none | none |
| TW | 1CR | 5 | Hazard-rate | Cosine | none | none |
| TW | 1CR | 6 | Hazard-rate | Simple polynomial | none | none |
| TW | 1CR | 7 | Half-normal | Cosine | VISIT | 1 |
| TW | 1CR | 8 | Half-normal | Hermite polynomial | VISIT | 1 |
| TW | 1CR | 9 | Hazard-rate | Cosine | VISIT | 1 |
| TW | 1CR | 10 | Hazard-rate | Simple polynomial | VISIT | 1 |
| TW | 1CR | 11 | Half-normal | Cosine | STATUS | 4 |
| TW | 1CR | 12 | Half-normal | Hermite polynomial | STATUS | 4 |
| TW | 1CR | 13 | Hazard-rate | Cosine | STATUS | 4 |
| TW | 1CR | 14 | Hazard-rate | Simple polynomial | STATUS | 4 |
| TW | 1CR | 15 | Half-normal | Cosine | HABITAT | 5 |
| TW | 1CR | 16 | Half-normal | Hermite polynomial | HABITAT | 5 |
| TW | 1CR | 17 | Hazard-rate | Cosine | HABITAT | 5 |
| TW | 1CR | 18 | Hazard-rate | Simple polynomial | HABITAT | 5 |
| TW | 1CR | 19 | Half-normal | Cosine | EPIPHYTESCAT | 6 |
| TW | 1CR | 20 | Half-normal | Hermite polynomial | EPIPHYTESCAT | 6 |
| TW | 1CR | 21 | Hazard-rate | Cosine | EPIPHYTESCAT | 6 |
| TW | 1CR | 22 | Hazard-rate | Simple polynomial | EPIPHYTESCAT | 6 |
| TW | 1CR | 23 | Half-normal | Cosine | MOSSLICHENCAT | 7 |
| TW | 1CR | 24 | Half-normal | Hermite polynomial | MOSSLICHENCAT | 7 |
| TW | 1CR | 25 | Hazard-rate | Cosine | MOSSLICHENCAT | 7 |
| TW | 1CR | 26 | Hazard-rate | Simple polynomial | MOSSLICHENCAT | 7 |
| TW | 1CR | 27 | Half-normal | Cosine | BARESOILPERC | 8 |
| TW | 1CR | 28 | Half-normal | Hermite polynomial | BARESOILPERC | 8 |
| TW | 1CR | 29 | Hazard-rate | Cosine | BARESOILPERC | 8 |
| TW | 1CR | 30 | Hazard-rate | Simple polynomial | BARESOILPERC | 8 |
| TW | 1CR | 31 | Half-normal | Cosine | SHRUBSPERC135CM | 9 |

| Typ e | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extention |
|-------|------------|----------|--------------------|------------------------|--------------------|----------------|
| TW | 1CR | 32 | Half-normal | Hermite polynomial | SHRUBSPERC135CM | 9 |
| TW | 1CR | 33 | Hazard-rate | Cosine | SHRUBSPERC135CM | 9 |
| TW | 1CR | 34 | Hazard-rate | Simple polynomial | SHRUBSPERC135CM | 9 |
| TW | 1CR | 35 | Half-normal | Cosine | UNDERSTORYCOVER PE | 11 |
| TW | 1CR | 36 | Half-normal | Hermite polynomial | UNDERSTORYCOVER PE | 11 |
| TW | 1CR | 37 | Hazard-rate | Cosine | UNDERSTORYCOVER PE | 11 |
| TW | 1CR | 38 | Hazard-rate | Simple polynomial | UNDERSTORYCOVER PE | 11 |
| TW | 1CR | 39 | Half-normal | Cosine | HIGHESTDBHCM | 12 |
| TW | 1CR | 40 | Half-normal | Hermite polynomial | HIGHESTDBHCM | 12 |
| TW | 1CR | 41 | Hazard-rate | Cosine | HIGHESTDBHCM | 12 |
| TW | 1CR | 42 | Hazard-rate | Simple polynomial | HIGHESTDBHCM | 12 |
| TW | 1CR | 43 | Half-normal | Cosine | MINSINCEDAWN | 15 |
| TW | 1CR | 44 | Half-normal | Hermite polynomial | MINSINCEDAWN | 15 |
| TW | 1CR | 45 | Hazard-rate | Cosine | MINSINCEDAWN | 15 |
| TW | 1CR | 46 | Hazard-rate | Simple polynomial | MINSINCEDAWN | 15 |
| TW | 2Ni | 1 | Half-normal | Cosine | none | none |
| TW | 2Ni | 2 | Half-normal | Hermite polynomial | none | none |
| TW | 2Ni | 3 | Uniform | Cosine | none | none |
| TW | 2Ni | 4 | Uniform | Simple polynomial | none | none |
| TW | 2Ni | 5 | Hazard-rate | Cosine | none | none |
| TW | 2Ni | 6 | Hazard-rate | Simple polynomial | none | none |
| TW | 2Ni | 7 | Half-normal | Cosine | VISIT | 1 |
| TW | 2Ni | 8 | Half-normal | Hermite polynomial | VISIT | 1 |
| TW | 2Ni | 9 | Hazard-rate | Cosine | VISIT | 1 |
| TW | 2Ni | 10 | Hazard-rate | Simple polynomial | VISIT | 1 |
| TW | 2Ni | 11 | Half-normal | Cosine | CLUSTER_SIZE | 2 |
| TW | 2Ni | 12 | Half-normal | Hermite polynomial | CLUSTER_SIZE | 2 |
| TW | 2Ni | 13 | Hazard-rate | Cosine | CLUSTER_SIZE | 2 |
| TW | 2Ni | 14 | Hazard-rate | Simple polynomial | CLUSTER_SIZE | 2 |
| TW | 2Ni | 15 | Half-normal | Cosine | CUPLABEL | 3 |
| TW | 2Ni | 16 | Half-normal | Hermite polynomial | CUPLABEL | 3 |
| TW | 2Ni | 17 | Hazard-rate | Cosine | CUPLABEL | 3 |
| TW | 2Ni | 18 | Hazard-rate | Simple polynomial | CUPLABEL | 3 |
| TW | 2Ni | 19 | Half-normal | Cosine | STATUS | 4 |
| TW | 2Ni | 20 | Half-normal | Hermite polynomial | STATUS | 4 |
| TW | 2Ni | 21 | Hazard-rate | Cosine | STATUS | 4 |
| TW | 2Ni | 22 | Hazard-rate | Simple polynomial | STATUS | 4 |
| TW | 2Ni | 23 | Half-normal | Cosine | HABITAT | 5 |
| TW | 2Ni | 24 | Half-normal | Hermite polynomial | HABITAT | 5 |
| TW | 2Ni | 25 | Hazard-rate | Cosine | HABITAT | 5 |
| TW | 2Ni | 26 | Hazard-rate | Simple polynomial | HABITAT | 5 |

| Type | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extension |
|------|------------|----------|--------------------|------------------------|--------------------|----------------|
| TW | 2Ni | 27 | Half-normal | Cosine | EIPHYTESCAT | 6 |
| TW | 2Ni | 28 | Half-normal | Hermite polynomial | EIPHYTESCAT | 6 |
| TW | 2Ni | 29 | Hazard-rate | Cosine | EIPHYTESCAT | 6 |
| TW | 2Ni | 30 | Hazard-rate | Simple polynomial | EIPHYTESCAT | 6 |
| TW | 2Ni | 31 | Half-normal | Cosine | SHRUBSPERC135CM | 9 |
| TW | 2Ni | 32 | Half-normal | Hermite polynomial | SHRUBSPERC135CM | 9 |
| TW | 2Ni | 33 | Hazard-rate | Cosine | SHRUBSPERC135CM | 9 |
| TW | 2Ni | 34 | Hazard-rate | Simple polynomial | SHRUBSPERC135CM | 9 |
| TW | 2Ni | 35 | Half-normal | Cosine | CANOPYPERC | 10 |
| TW | 2Ni | 36 | Half-normal | Hermite polynomial | CANOPYPERC | 10 |
| TW | 2Ni | 37 | Hazard-rate | Cosine | CANOPYPERC | 10 |
| TW | 2Ni | 38 | Hazard-rate | Simple polynomial | CANOPYPERC | 10 |
| TW | 2Ni | 39 | Half-normal | Cosine | UNDERSTORYCOVER PE | 11 |
| TW | 2Ni | 40 | Half-normal | Hermite polynomial | UNDERSTORYCOVER PE | 11 |
| TW | 2Ni | 41 | Hazard-rate | Cosine | UNDERSTORYCOVER PE | 11 |
| TW | 2Ni | 42 | Hazard-rate | Simple polynomial | UNDERSTORYCOVER PE | 11 |
| TW | 2Ni | 43 | Half-normal | Cosine | HIGHESTDBHCM | 12 |
| TW | 2Ni | 44 | Half-normal | Hermite polynomial | HIGHESTDBHCM | 12 |
| TW | 2Ni | 45 | Hazard-rate | Cosine | HIGHESTDBHCM | 12 |
| TW | 2Ni | 46 | Hazard-rate | Simple polynomial | HIGHESTDBHCM | 12 |
| TW | 2Ni | 47 | Half-normal | Cosine | HIGHESTTREEM | 13 |
| TW | 2Ni | 48 | Half-normal | Hermite polynomial | HIGHESTTREEM | 13 |
| TW | 2Ni | 49 | Hazard-rate | Cosine | HIGHESTTREEM | 13 |
| TW | 2Ni | 50 | Hazard-rate | Simple polynomial | HIGHESTTREEM | 13 |
| TW | 2Ni | 51 | Half-normal | Cosine | CANOPYTREESNO | 14 |
| TW | 2Ni | 52 | Half-normal | Hermite polynomial | CANOPYTREESNO | 14 |
| TW | 2Ni | 53 | Hazard-rate | Cosine | CANOPYTREESNO | 14 |
| TW | 2Ni | 54 | Hazard-rate | Simple polynomial | CANOPYTREESNO | 14 |
| TW | 2Ni | 55 | Half-normal | Cosine | MINSINCEDAWN | 15 |
| TW | 2Ni | 56 | Half-normal | Hermite polynomial | MINSINCEDAWN | 15 |
| TW | 2Ni | 57 | Hazard-rate | Cosine | MINSINCEDAWN | 15 |
| TW | 2Ni | 58 | Hazard-rate | Simple polynomial | MINSINCEDAWN | 15 |
| TW | 2Ni | 59 | Half-normal | Cosine | VISIT_EFFORT | 16 |
| TW | 2Ni | 60 | Half-normal | Hermite polynomial | VISIT_EFFORT | 16 |
| TW | 2Ni | 61 | Hazard-rate | Cosine | VISIT_EFFORT | 16 |
| TW | 2Ni | 62 | Hazard-rate | Simple polynomial | VISIT_EFFORT | 16 |
| TW | 2Ni | 63 | Half-normal | Cosine | BUG__870 | 38 |
| TW | 2Ni | 64 | Half-normal | Hermite polynomial | BUG__870 | 38 |
| TW | 2Ni | 65 | Hazard-rate | Cosine | BUG__870 | 38 |
| TW | 2Ni | 66 | Hazard-rate | Simple polynomial | BUG__870 | 38 |
| TW | 2Ni | 67 | Half-normal | Cosine | BUG__OTHER_RED | 39 |

| Type | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extension |
|------|------------|----------|--------------------|------------------------|------------------|----------------|
| TW | 2Ni | 68 | Half-normal | Hermite polynomial | BUG_OTHER_RED | 39 |
| TW | 2Ni | 69 | Hazard-rate | Cosine | BUG_OTHER_RED | 39 |
| TW | 2Ni | 70 | Hazard-rate | Simple polynomial | BUG_OTHER_RED | 39 |
| TW | 2Ni | 71 | Half-normal | Cosine | INSECT_869 | 40 |
| TW | 2Ni | 72 | Half-normal | Hermite polynomial | INSECT_869 | 40 |
| TW | 2Ni | 73 | Hazard-rate | Cosine | INSECT_869 | 40 |
| TW | 2Ni | 74 | Hazard-rate | Simple polynomial | INSECT_869 | 40 |
| TW | 2Ni | 75 | Half-normal | Cosine | SPIDER_SMALL_RED | 41 |
| TW | 2Ni | 76 | Half-normal | Hermite polynomial | SPIDER_SMALL_RED | 41 |
| TW | 2Ni | 77 | Hazard-rate | Cosine | SPIDER_SMALL_RED | 41 |
| TW | 2Ni | 78 | Hazard-rate | Simple polynomial | SPIDER_SMALL_RED | 41 |
| TW | 2Ni | 79 | Half-normal | Cosine | TOAD | 42 |
| TW | 2Ni | 80 | Half-normal | Hermite polynomial | TOAD | 42 |
| TW | 2Ni | 81 | Hazard-rate | Cosine | TOAD | 42 |
| TW | 2Ni | 82 | Hazard-rate | Simple polynomial | TOAD | 42 |
| TW | 3AK | 1 | Half-normal | Cosine | none | none |
| TW | 3AK | 2 | Half-normal | Hermite polynomial | none | none |
| TW | 3AK | 3 | Uniform | Cosine | none | none |
| TW | 3AK | 4 | Uniform | Simple polynomial | none | none |
| TW | 3AK | 5 | Hazard-rate | Cosine | none | none |
| TW | 3AK | 6 | Hazard-rate | Simple polynomial | none | none |
| TW | 3AK | 7 | Half-normal | Cosine | VISIT | 1 |
| TW | 3AK | 8 | Half-normal | Hermite polynomial | VISIT | 1 |
| TW | 3AK | 9 | Hazard-rate | Cosine | VISIT | 1 |
| TW | 3AK | 10 | Hazard-rate | Simple polynomial | VISIT | 1 |
| TW | 3AK | 11 | Half-normal | Cosine | CLUSTER_SIZE | 2 |
| TW | 3AK | 12 | Half-normal | Hermite polynomial | CLUSTER_SIZE | 2 |
| TW | 3AK | 13 | Hazard-rate | Cosine | CLUSTER_SIZE | 2 |
| TW | 3AK | 14 | Hazard-rate | Simple polynomial | CLUSTER_SIZE | 2 |
| TW | 3AK | 15 | Half-normal | Cosine | CUPLABEL | 3 |
| TW | 3AK | 16 | Half-normal | Hermite polynomial | CUPLABEL | 3 |
| TW | 3AK | 17 | Hazard-rate | Cosine | CUPLABEL | 3 |
| TW | 3AK | 18 | Hazard-rate | Simple polynomial | CUPLABEL | 3 |
| TW | 3AK | 19 | Half-normal | Cosine | HABITAT | 5 |
| TW | 3AK | 20 | Half-normal | Hermite polynomial | HABITAT | 5 |
| TW | 3AK | 21 | Hazard-rate | Cosine | HABITAT | 5 |
| TW | 3AK | 22 | Hazard-rate | Simple polynomial | HABITAT | 5 |
| TW | 3AK | 23 | Half-normal | Cosine | MOSSLICHENCAT | 7 |
| TW | 3AK | 24 | Half-normal | Hermite polynomial | MOSSLICHENCAT | 7 |
| TW | 3AK | 25 | Hazard-rate | Cosine | MOSSLICHENCAT | 7 |
| TW | 3AK | 26 | Hazard-rate | Simple polynomial | MOSSLICHENCAT | 7 |
| TW | 3AK | 27 | Half-normal | Cosine | VISIT_EFFORT | 16 |
| TW | 3AK | 28 | Half-normal | Hermite polynomial | VISIT_EFFORT | 16 |

| Type | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extension |
|------|------------|----------|--------------------|------------------------|-----------------|----------------|
| TW | 3AK | 29 | Hazard-rate | Cosine | VISIT_EFFORT | 16 |
| TW | 3AK | 30 | Hazard-rate | Simple polynomial | VISIT_EFFORT | 16 |
| TW | 3AK | 31 | Half-normal | Cosine | COVARIATE04 | 20 |
| TW | 3AK | 32 | Half-normal | Hermite polynomial | COVARIATE04 | 20 |
| TW | 3AK | 33 | Hazard-rate | Cosine | COVARIATE04 | 20 |
| TW | 3AK | 34 | Hazard-rate | Simple polynomial | COVARIATE04 | 20 |
| TW | 3AK | 35 | Half-normal | Cosine | COVARIATE07 | 21 |
| TW | 3AK | 36 | Half-normal | Hermite polynomial | COVARIATE07 | 21 |
| TW | 3AK | 37 | Hazard-rate | Cosine | COVARIATE07 | 21 |
| TW | 3AK | 38 | Hazard-rate | Simple polynomial | COVARIATE07 | 21 |
| TW | 3AK | 39 | Half-normal | Cosine | COVARIATE08 | 22 |
| TW | 3AK | 40 | Half-normal | Hermite polynomial | COVARIATE08 | 22 |
| TW | 3AK | 41 | Hazard-rate | Cosine | COVARIATE08 | 22 |
| TW | 3AK | 42 | Hazard-rate | Simple polynomial | COVARIATE08 | 22 |
| TW | 3AK | 43 | Half-normal | Cosine | COVARIATE09 | 23 |
| TW | 3AK | 44 | Half-normal | Hermite polynomial | COVARIATE09 | 23 |
| TW | 3AK | 45 | Hazard-rate | Cosine | COVARIATE09 | 23 |
| TW | 3AK | 46 | Hazard-rate | Simple polynomial | COVARIATE09 | 23 |
| TW | 3AK | 47 | Half-normal | Cosine | COVARIATE10 | 24 |
| TW | 3AK | 48 | Half-normal | Hermite polynomial | COVARIATE10 | 24 |
| TW | 3AK | 49 | Hazard-rate | Cosine | COVARIATE10 | 24 |
| TW | 3AK | 50 | Hazard-rate | Simple polynomial | COVARIATE10 | 24 |
| TW | 3AK | 51 | Half-normal | Cosine | COVARIATE11 | 25 |
| TW | 3AK | 52 | Half-normal | Hermite polynomial | COVARIATE11 | 25 |
| TW | 3AK | 53 | Hazard-rate | Cosine | COVARIATE11 | 25 |
| TW | 3AK | 54 | Hazard-rate | Simple polynomial | COVARIATE11 | 25 |
| TW | 3AK | 55 | Half-normal | Cosine | COVARIATE12 | 26 |
| TW | 3AK | 56 | Half-normal | Hermite polynomial | COVARIATE12 | 26 |
| TW | 3AK | 57 | Hazard-rate | Cosine | COVARIATE12 | 26 |
| TW | 3AK | 58 | Hazard-rate | Simple polynomial | COVARIATE12 | 26 |
| TW | 3AK | 59 | Half-normal | Cosine | COVARIATE13 | 27 |
| TW | 3AK | 60 | Half-normal | Hermite polynomial | COVARIATE13 | 27 |
| TW | 3AK | 61 | Hazard-rate | Cosine | COVARIATE13 | 27 |
| TW | 3AK | 62 | Hazard-rate | Simple polynomial | COVARIATE13 | 27 |
| TW | 3AK | 63 | Half-normal | Cosine | COVARIATE14 | 28 |
| TW | 3AK | 64 | Half-normal | Hermite polynomial | COVARIATE14 | 28 |
| TW | 3AK | 65 | Hazard-rate | Cosine | COVARIATE14 | 28 |
| TW | 3AK | 66 | Hazard-rate | Simple polynomial | COVARIATE14 | 28 |
| TW | 3AK | 67 | Half-normal | Cosine | COVARIATE15 | 29 |
| TW | 3AK | 68 | Half-normal | Hermite polynomial | COVARIATE15 | 29 |
| TW | 3AK | 69 | Hazard-rate | Cosine | COVARIATE15 | 29 |
| TW | 3AK | 70 | Hazard-rate | Simple polynomial | COVARIATE15 | 29 |
| TW | 3AK | 71 | Half-normal | Cosine | COVARIATE18 | 30 |
| TW | 3AK | 72 | Half-normal | Hermite | COVARIATE18 | 30 |

| Type | Study area | Model no | Model key function | Model series expansion | MCDS covariates | Name extension |
|------|------------|----------|--------------------|------------------------|------------------|----------------|
| | | | | polynomial | | |
| TW | 3AK | 73 | Hazard-rate | Cosine | COVARIATE18 | 30 |
| TW | 3AK | 74 | Hazard-rate | Simple polynomial | COVARIATE18 | 30 |
| TW | 3AK | 75 | Half-normal | Cosine | ANT__SMALL | 43 |
| TW | 3AK | 76 | Half-normal | Hermite polynomial | ANT__SMALL | 43 |
| TW | 3AK | 77 | Hazard-rate | Cosine | ANT__SMALL | 43 |
| TW | 3AK | 78 | Hazard-rate | Simple polynomial | ANT__SMALL | 43 |
| TW | 3AK | 79 | Half-normal | Cosine | BOREAL_CHICKADEE | 44 |
| TW | 3AK | 80 | Half-normal | Hermite polynomial | BOREAL_CHICKADEE | 44 |
| TW | 3AK | 81 | Hazard-rate | Cosine | BOREAL_CHICKADEE | 44 |
| TW | 3AK | 82 | Hazard-rate | Simple polynomial | BOREAL_CHICKADEE | 44 |
| TW | 4Ru | 1 | Half-normal | Cosine | none | none |
| TW | 4Ru | 2 | Half-normal | Hermite polynomial | none | none |
| TW | 4Ru | 3 | Uniform | Cosine | none | none |
| TW | 4Ru | 4 | Uniform | Simple polynomial | none | none |
| TW | 4Ru | 5 | Hazard-rate | Cosine | none | none |
| TW | 4Ru | 6 | Hazard-rate | Simple polynomial | none | none |
| TW | 4Ru | 7 | Half-normal | Cosine | VISIT | 1 |
| TW | 4Ru | 8 | Half-normal | Hermite polynomial | VISIT | 1 |
| TW | 4Ru | 9 | Hazard-rate | Cosine | VISIT | 1 |
| TW | 4Ru | 10 | Hazard-rate | Simple polynomial | VISIT | 1 |
| TW | 4Ru | 11 | Half-normal | Cosine | CLUSTER_SIZE | 2 |
| TW | 4Ru | 12 | Half-normal | Hermite polynomial | CLUSTER_SIZE | 2 |
| TW | 4Ru | 13 | Hazard-rate | Cosine | CLUSTER_SIZE | 2 |
| TW | 4Ru | 14 | Hazard-rate | Simple polynomial | CLUSTER_SIZE | 2 |
| TW | 4Ru | 15 | Half-normal | Cosine | CUPLABEL | 3 |
| TW | 4Ru | 16 | Half-normal | Hermite polynomial | CUPLABEL | 3 |
| TW | 4Ru | 17 | Hazard-rate | Cosine | CUPLABEL | 3 |
| TW | 4Ru | 18 | Hazard-rate | Simple polynomial | CUPLABEL | 3 |
| TW | 4Ru | 19 | Half-normal | Cosine | HABITAT | 5 |
| TW | 4Ru | 20 | Half-normal | Hermite polynomial | HABITAT | 5 |
| TW | 4Ru | 21 | Hazard-rate | Cosine | HABITAT | 5 |
| TW | 4Ru | 22 | Hazard-rate | Simple polynomial | HABITAT | 5 |
| TW | 4Ru | 23 | Half-normal | Cosine | HIGHESTTREEM | 13 |
| TW | 4Ru | 24 | Half-normal | Hermite polynomial | HIGHESTTREEM | 13 |
| TW | 4Ru | 25 | Hazard-rate | Cosine | HIGHESTTREEM | 13 |
| TW | 4Ru | 26 | Hazard-rate | Simple polynomial | HIGHESTTREEM | 13 |
| TW | 4Ru | 27 | Half-normal | Cosine | MINSINCEDAWN | 15 |
| TW | 4Ru | 28 | Half-normal | Hermite polynomial | MINSINCEDAWN | 15 |
| TW | 4Ru | 29 | Hazard-rate | Cosine | MINSINCEDAWN | 15 |
| TW | 4Ru | 30 | Hazard-rate | Simple polynomial | MINSINCEDAWN | 15 |
| TW | 4Ru | 31 | Half-normal | Cosine | LICHENPERC | 18 |
| TW | 4Ru | 32 | Half-normal | Hermite polynomial | LICHENPERC | 18 |
| TW | 4Ru | 33 | Hazard-rate | Cosine | LICHENPERC | 18 |

| Type | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extension |
|------|------------|----------|--------------------|------------------------|-----------------|----------------|
| TW | 4Ru | 34 | Hazard-rate | Simple polynomial | LICHENPERC | 18 |
| TW | 4Ru | 35 | Half-normal | Cosine | MOSSPERC | 19 |
| TW | 4Ru | 36 | Half-normal | Hermite polynomial | MOSSPERC | 19 |
| TW | 4Ru | 37 | Hazard-rate | Cosine | MOSSPERC | 19 |
| TW | 4Ru | 38 | Hazard-rate | Simple polynomial | MOSSPERC | 19 |
| TW | 4Ru | 39 | Half-normal | Cosine | COVARIATE01 | 31 |
| TW | 4Ru | 40 | Half-normal | Hermite polynomial | COVARIATE01 | 31 |
| TW | 4Ru | 41 | Hazard-rate | Cosine | COVARIATE01 | 31 |
| TW | 4Ru | 42 | Hazard-rate | Simple polynomial | COVARIATE01 | 31 |
| TW | 4Ru | 43 | Half-normal | Cosine | COVARIATE03 | 32 |
| TW | 4Ru | 44 | Half-normal | Hermite polynomial | COVARIATE03 | 32 |
| TW | 4Ru | 45 | Hazard-rate | Cosine | COVARIATE03 | 32 |
| TW | 4Ru | 46 | Hazard-rate | Simple polynomial | COVARIATE03 | 32 |
| TW | 4Ru | 47 | Half-normal | Cosine | COVARIATE05 | 33 |
| TW | 4Ru | 48 | Half-normal | Hermite polynomial | COVARIATE05 | 33 |
| TW | 4Ru | 49 | Hazard-rate | Cosine | COVARIATE05 | 33 |
| TW | 4Ru | 50 | Hazard-rate | Simple polynomial | COVARIATE05 | 33 |
| TW | 4Ru | 51 | Half-normal | Cosine | COVARIATE11 | 25 |
| TW | 4Ru | 52 | Half-normal | Hermite polynomial | COVARIATE11 | 25 |
| TW | 4Ru | 53 | Hazard-rate | Cosine | COVARIATE11 | 25 |
| TW | 4Ru | 54 | Hazard-rate | Simple polynomial | COVARIATE11 | 25 |
| TW | 4Ru | 55 | Half-normal | Cosine | COVARIATE19 | 34 |
| TW | 4Ru | 56 | Half-normal | Hermite polynomial | COVARIATE19 | 34 |
| TW | 4Ru | 57 | Hazard-rate | Cosine | COVARIATE19 | 34 |
| TW | 4Ru | 58 | Hazard-rate | Simple polynomial | COVARIATE19 | 34 |
| TW | 4Ru | 59 | Half-normal | Cosine | COVARIATE21 | 35 |
| TW | 4Ru | 60 | Half-normal | Hermite polynomial | COVARIATE21 | 35 |
| TW | 4Ru | 61 | Hazard-rate | Cosine | COVARIATE21 | 35 |
| TW | 4Ru | 62 | Hazard-rate | Simple polynomial | COVARIATE21 | 35 |
| TW | 5PG | 1 | Half-normal | Cosine | none | none |
| TW | 5PG | 2 | Half-normal | Hermite polynomial | none | none |
| TW | 5PG | 3 | Uniform | Cosine | none | none |
| TW | 5PG | 4 | Uniform | Simple polynomial | none | none |
| TW | 5PG | 5 | Hazard-rate | Cosine | none | none |
| TW | 5PG | 6 | Hazard-rate | Simple polynomial | none | none |
| TW | 5PG | 7 | Half-normal | Cosine | VISIT | 1 |
| TW | 5PG | 8 | Half-normal | Hermite polynomial | VISIT | 1 |
| TW | 5PG | 9 | Hazard-rate | Cosine | VISIT | 1 |
| TW | 5PG | 10 | Hazard-rate | Simple polynomial | VISIT | 1 |
| TW | 5PG | 11 | Half-normal | Cosine | CUPLABEL | 3 |
| TW | 5PG | 12 | Half-normal | Hermite polynomial | CUPLABEL | 3 |
| TW | 5PG | 13 | Hazard-rate | Cosine | CUPLABEL | 3 |
| TW | 5PG | 14 | Hazard-rate | Simple polynomial | CUPLABEL | 3 |
| TW | 5PG | 15 | Half-normal | Cosine | HABITAT | 5 |

| Type | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extension |
|------|------------|----------|--------------------|------------------------|-----------------|----------------|
| TW | 5PG | 16 | Half-normal | Hermite polynomial | HABITAT | 5 |
| TW | 5PG | 17 | Hazard-rate | Cosine | HABITAT | 5 |
| TW | 5PG | 18 | Hazard-rate | Simple polynomial | HABITAT | 5 |
| TW | 5PG | 19 | Half-normal | Cosine | EPIPHYTESCAT | 6 |
| TW | 5PG | 20 | Half-normal | Hermite polynomial | EPIPHYTESCAT | 6 |
| TW | 5PG | 21 | Hazard-rate | Cosine | EPIPHYTESCAT | 6 |
| TW | 5PG | 22 | Hazard-rate | Simple polynomial | EPIPHYTESCAT | 6 |
| TW | 5PG | 23 | Half-normal | Cosine | COVARIATE01 | 31 |
| TW | 5PG | 24 | Half-normal | Hermite polynomial | COVARIATE01 | 31 |
| TW | 5PG | 25 | Hazard-rate | Cosine | COVARIATE01 | 31 |
| TW | 5PG | 26 | Hazard-rate | Simple polynomial | COVARIATE01 | 31 |
| TW | 5PG | 27 | Half-normal | Cosine | COVARIATE05 | 33 |
| TW | 5PG | 28 | Half-normal | Hermite polynomial | COVARIATE05 | 33 |
| TW | 5PG | 29 | Hazard-rate | Cosine | COVARIATE05 | 33 |
| TW | 5PG | 30 | Hazard-rate | Simple polynomial | COVARIATE05 | 33 |
| TW | 5PG | 31 | Half-normal | Cosine | COVARIATE06 | 37 |
| TW | 5PG | 32 | Half-normal | Hermite polynomial | COVARIATE06 | 37 |
| TW | 5PG | 33 | Hazard-rate | Cosine | COVARIATE06 | 37 |
| TW | 5PG | 34 | Hazard-rate | Simple polynomial | COVARIATE06 | 37 |
| TW | 5PG | 35 | Half-normal | Cosine | COVARIATE08 | 22 |
| TW | 5PG | 36 | Half-normal | Hermite polynomial | COVARIATE08 | 22 |
| TW | 5PG | 37 | Hazard-rate | Cosine | COVARIATE08 | 22 |
| TW | 5PG | 38 | Hazard-rate | Simple polynomial | COVARIATE08 | 22 |
| TW | 5PG | 39 | Half-normal | Cosine | COVARIATE11 | 25 |
| TW | 5PG | 40 | Half-normal | Hermite polynomial | COVARIATE11 | 25 |
| TW | 5PG | 41 | Hazard-rate | Cosine | COVARIATE11 | 25 |
| TW | 5PG | 42 | Hazard-rate | Simple polynomial | COVARIATE11 | 25 |
| TW | 5PG | 43 | Half-normal | Cosine | COVARIATE12 | 26 |
| TW | 5PG | 44 | Half-normal | Hermite polynomial | COVARIATE12 | 26 |
| TW | 5PG | 45 | Hazard-rate | Cosine | COVARIATE12 | 26 |
| TW | 5PG | 46 | Hazard-rate | Simple polynomial | COVARIATE12 | 26 |
| TW | 6Ba | 1 | Half-normal | Cosine | none | none |
| TW | 6Ba | 2 | Half-normal | Hermite polynomial | none | none |
| TW | 6Ba | 3 | Uniform | Cosine | none | none |
| TW | 6Ba | 4 | Uniform | Simple polynomial | none | none |
| TW | 6Ba | 5 | Hazard-rate | Cosine | none | none |
| TW | 6Ba | 6 | Hazard-rate | Simple polynomial | none | none |
| TW | 6Ba | 7 | Half-normal | Cosine | VISIT | 1 |
| TW | 6Ba | 8 | Half-normal | Hermite polynomial | VISIT | 1 |
| TW | 6Ba | 9 | Hazard-rate | Cosine | VISIT | 1 |
| TW | 6Ba | 10 | Hazard-rate | Simple polynomial | VISIT | 1 |
| TW | 6Ba | 11 | Half-normal | Cosine | CLUSTER_SIZE | 2 |
| TW | 6Ba | 12 | Half-normal | Hermite polynomial | CLUSTER_SIZE | 2 |

| Type | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extension |
|------|------------|----------|--------------------|------------------------|-----------------|----------------|
| TW | 6Ba | 13 | Hazard-rate | Cosine | CLUSTER_SIZE | 2 |
| TW | 6Ba | 14 | Hazard-rate | Simple polynomial | CLUSTER_SIZE | 2 |
| TW | 6Ba | 15 | Half-normal | Cosine | CUPLABEL | 3 |
| TW | 6Ba | 16 | Half-normal | Hermite polynomial | CUPLABEL | 3 |
| TW | 6Ba | 17 | Hazard-rate | Cosine | CUPLABEL | 3 |
| TW | 6Ba | 18 | Hazard-rate | Simple polynomial | CUPLABEL | 3 |
| TW | 6Ba | 19 | Half-normal | Cosine | STATUS | 4 |
| TW | 6Ba | 20 | Half-normal | Hermite polynomial | STATUS | 4 |
| TW | 6Ba | 21 | Hazard-rate | Cosine | STATUS | 4 |
| TW | 6Ba | 22 | Hazard-rate | Simple polynomial | STATUS | 4 |
| TW | 6Ba | 23 | Half-normal | Cosine | BARESOILPERC | 8 |
| TW | 6Ba | 24 | Half-normal | Hermite polynomial | BARESOILPERC | 8 |
| TW | 6Ba | 25 | Hazard-rate | Cosine | BARESOILPERC | 8 |
| TW | 6Ba | 26 | Hazard-rate | Simple polynomial | BARESOILPERC | 8 |
| TW | 6Ba | 27 | Half-normal | Cosine | VISIT_EFFORT | 16 |
| TW | 6Ba | 28 | Half-normal | Hermite polynomial | VISIT_EFFORT | 16 |
| TW | 6Ba | 29 | Hazard-rate | Cosine | VISIT_EFFORT | 16 |
| TW | 6Ba | 30 | Hazard-rate | Simple polynomial | VISIT_EFFORT | 16 |
| TW | 6Ba | 31 | Half-normal | Cosine | GRASSPERC | 17 |
| TW | 6Ba | 32 | Half-normal | Hermite polynomial | GRASSPERC | 17 |
| TW | 6Ba | 33 | Hazard-rate | Cosine | GRASSPERC | 17 |
| TW | 6Ba | 34 | Hazard-rate | Simple polynomial | GRASSPERC | 17 |
| TW | 6Ba | 35 | Half-normal | Cosine | LICHENPERC | 18 |
| TW | 6Ba | 36 | Half-normal | Hermite polynomial | LICHENPERC | 18 |
| TW | 6Ba | 37 | Hazard-rate | Cosine | LICHENPERC | 18 |
| TW | 6Ba | 38 | Hazard-rate | Simple polynomial | LICHENPERC | 18 |
| TW | 6Ba | 39 | Half-normal | Cosine | MOSSPERC | 19 |
| TW | 6Ba | 40 | Half-normal | Hermite polynomial | MOSSPERC | 19 |
| TW | 6Ba | 41 | Hazard-rate | Cosine | MOSSPERC | 19 |
| TW | 6Ba | 42 | Hazard-rate | Simple polynomial | MOSSPERC | 19 |
| TW | 6Ba | 43 | Half-normal | Cosine | COVARIATE01 | 31 |
| TW | 6Ba | 44 | Half-normal | Hermite polynomial | COVARIATE01 | 31 |
| TW | 6Ba | 45 | Hazard-rate | Cosine | COVARIATE01 | 31 |
| TW | 6Ba | 46 | Hazard-rate | Simple polynomial | COVARIATE01 | 31 |
| TW | 6Ba | 47 | Half-normal | Cosine | COVARIATE02 | 36 |
| TW | 6Ba | 48 | Half-normal | Hermite polynomial | COVARIATE02 | 36 |
| TW | 6Ba | 49 | Hazard-rate | Cosine | COVARIATE02 | 36 |
| TW | 6Ba | 50 | Hazard-rate | Simple polynomial | COVARIATE02 | 36 |
| TW | 6Ba | 51 | Half-normal | Cosine | COVARIATE08 | 22 |
| TW | 6Ba | 52 | Half-normal | Hermite polynomial | COVARIATE08 | 22 |
| TW | 6Ba | 53 | Hazard-rate | Cosine | COVARIATE08 | 22 |
| TW | 6Ba | 54 | Hazard-rate | Simple polynomial | COVARIATE08 | 22 |
| TW | 6Ba | 55 | Half-normal | Cosine | COVARIATE10 | 24 |
| TW | 6Ba | 56 | Half-normal | Hermite | COVARIATE10 | 24 |

| Type | Study area | Model no | Model key function | Model series expansion | MCDs covariates | Name extension |
|------|------------|----------|--------------------|------------------------|-----------------|----------------|
| | | | | polynomial | | |
| TW | 6Ba | 57 | Hazard-rate | Cosine | COVARIATE10 | 24 |
| TW | 6Ba | 58 | Hazard-rate | Simple polynomial | COVARIATE10 | 24 |

7.4 PRESENCE Model Definitions

| Study area | Type | Model |
|------------|------|----------------------------|
| 1CR | Bi | 1 group, Constant P |
| 1CR | Bi | 1 group, Survey-specific P |
| 1CR | Bi | BareSoil |
| 1CR | Bi | CanopyPerc |
| 1CR | Bi | CanopyTrees |
| 1CR | Bi | Cov04 |
| 1CR | Bi | Cov05 |
| 1CR | Bi | Cov12 |
| 1CR | Bi | DuffCover |
| 1CR | Bi | Epiphytes |
| 1CR | Bi | Flowers |
| 1CR | Bi | Habitat |
| 1CR | Bi | HighestDBH |
| 1CR | Bi | HighestTree |
| 1CR | Bi | LeafBrowsing |
| 1CR | Bi | Manakin |
| 1CR | Bi | Min |
| 1CR | Bi | MossLichen |
| 1CR | Bi | Shrubs |
| 1CR | Bi | TurkeyVulture |
| 1CR | Bi | Understory |
| 2Ni | Bi | 1 group, Constant P |
| 2Ni | Bi | 1 group, Survey-specific P |
| 2Ni | Bi | BareSoil |
| 2Ni | Bi | CanopyPerc |
| 2Ni | Bi | DuffCover |
| 2Ni | Bi | Epiphytes |
| 2Ni | Bi | Habitat |
| 2Ni | Bi | HighestDBH |
| 2Ni | Bi | HighestTree |
| 2Ni | Bi | Min |
| 2Ni | Bi | MossLichen |
| 2Ni | Bi | Shrubs |
| 2Ni | Bi | TurkeyVulture |
| 2Ni | Bi | Understory |
| 3AK | Bi | 1 group, Constant P |
| 3AK | Bi | 1 group, Survey-specific P |
| 3AK | Bi | CanopyPerc |
| 3AK | Bi | CanopyTrees |
| 3AK | Bi | Cov07 |
| 3AK | Bi | Cov1 |
| 3AK | Bi | Cov11 |
| 3AK | Bi | Cov12 |
| 3AK | Bi | Cov13 |
| 3AK | Bi | Cov14 |
| 3AK | Bi | Cov19 |
| 3AK | Bi | DuffCover |
| 3AK | Bi | Habitat |
| 3AK | Bi | HighestDBH |

| Study area | Type | Model |
|-------------------|-------------|----------------------------|
| 3AK | Bi | HighestTree |
| 3AK | Bi | Min |
| 3AK | Bi | Model |
| 3AK | Bi | MossLichen |
| 3AK | Bi | Squirrel |
| 4Ru | Bi | 1 group, Constant P |
| 4Ru | Bi | 1 group, Survey-specific P |
| 4Ru | Bi | CanopyPerc |
| 4Ru | Bi | CanopyTrees |
| 4Ru | Bi | Cov01 |
| 4Ru | Bi | Cov04 |
| 4Ru | Bi | Cov05 |
| 4Ru | Bi | Cov08 |
| 4Ru | Bi | Cov12 |
| 4Ru | Bi | Cov15 |
| 4Ru | Bi | Cov16 |
| 4Ru | Bi | Cov18 |
| 4Ru | Bi | Cov19 |
| 4Ru | Bi | Cov20 |
| 4Ru | Bi | Cov21 |
| 4Ru | Bi | Cov23 |
| 4Ru | Bi | Cov28 |
| 4Ru | Bi | Cov30 |
| 4Ru | Bi | Cov31 |
| 4Ru | Bi | DuffCover |
| 4Ru | Bi | Flowers |
| 4Ru | Bi | Habitat |
| 4Ru | Bi | HighDBH |
| 4Ru | Bi | HighTree |
| 4Ru | Bi | LichenPerc |
| 4Ru | Bi | Min |
| 4Ru | Bi | MossPerc |
| 4Ru | Bi | Shrubs |
| 4Ru | Bi | Understory |
| 4Ru | Bi | Wize |
| 5PG | Bi | 1 group, Constant P |
| 5PG | Bi | 1 group, Survey-specific P |
| 5PG | Bi | BareSoil |
| 5PG | Bi | CanopyPerc |
| 5PG | Bi | Cov01 |
| 5PG | Bi | Cov06 |
| 5PG | Bi | Cov11 |
| 5PG | Bi | Cov12 |
| 5PG | Bi | DuffCover |
| 5PG | Bi | Epiphytes |
| 5PG | Bi | Habitat |
| 5PG | Bi | HighestDBH |
| 5PG | Bi | HighestTree |
| 5PG | Bi | Min |
| 6Ba | Bi | 1 group, Constant P |
| 6Ba | Bi | 1 group, Survey-specific P |
| 6Ba | Bi | BareSoil |
| 6Ba | Bi | Cov01 |

| Study area | Type | Model |
|-------------------|-------------|----------------------------|
| 6Ba | Bi | Cov02 |
| 6Ba | Bi | Cov03 |
| 6Ba | Bi | Cov05 |
| 6Ba | Bi | Cov06 |
| 6Ba | Bi | Cov07 |
| 6Ba | Bi | Cov08 |
| 6Ba | Bi | Cov10 |
| 6Ba | Bi | DiamLake |
| 6Ba | Bi | DistLake |
| 6Ba | Bi | Flowers |
| 6Ba | Bi | GrassPerc |
| 6Ba | Bi | Leafs |
| 6Ba | Bi | LichenPerc |
| 6Ba | Bi | MossPerc |
| 6Ba | Bi | PomarineJaeger |
| 6Ba | Bi | SurveyEffort |
| 1CR | TW | 1 group, Constant P |
| 1CR | TW | 1 group, Survey-specific P |
| 1CR | TW | BareSoil |
| 1CR | TW | Epiphytes |
| 1CR | TW | Habitat |
| 1CR | TW | HighestDBH |
| 1CR | TW | Min |
| 1CR | TW | MossLichen |
| 1CR | TW | Shrubs |
| 1CR | TW | Understory |
| 2Ni | TW | 1 group, Constant P |
| 2Ni | TW | 1 group, Survey-specific P |
| 2Ni | TW | Bug870 |
| 2Ni | TW | BugOtherRed |
| 2Ni | TW | CanopyPerc |
| 2Ni | TW | CanopyTrees |
| 2Ni | TW | Epiphytes |
| 2Ni | TW | Habitat |
| 2Ni | TW | HighestDBH |
| 2Ni | TW | HighestTree |
| 2Ni | TW | Insect869 |
| 2Ni | TW | Min |
| 2Ni | TW | Shrubs |
| 2Ni | TW | SpiderSmallRed |
| 2Ni | TW | Toad |
| 2Ni | TW | Understory |
| 2Ni | TW | VisitEffort |
| 3AK | TW | 1 group, Constant P |
| 3AK | TW | 1 group, Survey-specific P |
| 3AK | TW | AntSmall |
| 3AK | TW | BorealChickadee |
| 3AK | TW | Cov04 |
| 3AK | TW | Cov07 |
| 3AK | TW | Cov08 |
| 3AK | TW | Cov09 |
| 3AK | TW | Cov10 |
| 3AK | TW | Cov11 |

| Study area | Type | Model |
|-------------------|-------------|----------------------------|
| 3AK | TW | Cov12 |
| 3AK | TW | Cov13 |
| 3AK | TW | Cov14 |
| 3AK | TW | Cov15 |
| 3AK | TW | Cov18 |
| 3AK | TW | Habitat |
| 3AK | TW | Min |
| 3AK | TW | MossLichen |
| 3AK | TW | SurveyEffort |
| 4Ru | TW | 1 group, Constant P |
| 4Ru | TW | 1 group, Survey-specific P |
| 4Ru | TW | Cov01 |
| 4Ru | TW | Cov03 |
| 4Ru | TW | Cov05 |
| 4Ru | TW | Cov11 |
| 4Ru | TW | Cov19 |
| 4Ru | TW | Cov21 |
| 4Ru | TW | Habitat |
| 4Ru | TW | HighestTree |
| 4Ru | TW | LichenPerc |
| 4Ru | TW | Min |
| 4Ru | TW | MossPerc |
| 5PG | TW | 1 group, Constant P |
| 5PG | TW | 1 group, Survey-specific P |
| 5PG | TW | Cov01 |
| 5PG | TW | Cov05 |
| 5PG | TW | Cov06 |
| 5PG | TW | Cov08 |
| 5PG | TW | Cov11 |
| 5PG | TW | Cov12 |
| 5PG | TW | Epiphytes |
| 5PG | TW | Habitat |
| 5PG | TW | Min |
| 6Ba | TW | 1 group, Constant P |
| 6Ba | TW | 1 group, Survey-specific P |
| 6Ba | TW | BareSoil |
| 6Ba | TW | Cov01 |
| 6Ba | TW | Cov02 |
| 6Ba | TW | Cov08 |
| 6Ba | TW | Cov10 |
| 6Ba | TW | GrassPerc |
| 6Ba | TW | LichenPerc |
| 6Ba | TW | MossPerc |
| 6Ba | TW | SurveyEffort |

7.5 Detailed Species Lists (Valid ITIS Taxonomy)

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|---------------------------|----------------|----------------|----------------|----------------|
| Bi | 1CR | 9 | Ani | Cuculiformes | Cuculidae | Crotophaga | not identified |
| Bi | 1CR | 8 | Bird | not identified | not identified | not identified | not identified |
| Bi | 1CR | 1 | Bird of Prey | not identified | not identified | not identified | not identified |
| Bi | 1CR | 1 | Bird, big | not identified | not identified | not identified | not identified |
| Bi | 1CR | 3 | Crake | Gruiformes | Rallidae | not identified | not identified |
| Bi | 1CR | 15 | Dove | Columbiformes | Columbidae | not identified | not identified |
| Bi | 1CR | 2 | Falcon | Ciconiiformes | Falconidae | Falco | not identified |
| Bi | 1CR | 61 | Flycatcher | Passeriformes | Tyrannidae | not identified | not identified |
| Bi | 1CR | 1 | Golden-bellied Flycatcher | Passeriformes | Tyrannidae | Myiodynastes | hemichrysus |
| Bi | 1CR | 3 | Golden-hooded Tanager | Passeriformes | Thraupidae | Tangara | larvata |
| Bi | 1CR | 2 | Gray-necked Woodpecker | Piciformes | Picidae | not identified | not identified |
| Bi | 1CR | 28 | Great Kiskadee | Passeriformes | Tyrannidae | Pitangus | sulphuratus |
| Bi | 1CR | 2 | Groove-billed Ani | Cuculiformes | Cuculidae | Crotophaga | sulcirostris |
| Bi | 1CR | 128 | Hummingbird | Apodiformes | Trochilidae | not identified | not identified |
| Bi | 1CR | 10 | Kiskadee | Passeriformes | Tyrannidae | Pitangus | not identified |
| Bi | 1CR | 1 | Lattice-tailed Trogon | Trogoniformes | Trogonidae | Trogon | clathratus |
| Bi | 1CR | 1 | Laughing Falcon | Ciconiiformes | Falconidae | Herpetotheres | cachinnans |
| Bi | 1CR | 1 | Lesser Kiskadee | Passeriformes | Tyrannidae | Pitangus | lictor |
| Bi | 1CR | 1 | Little | not identified | not identified | not identified | not identified |
| Bi | 1CR | 16 | Manakin | Passeriformes | Pipridae | not identified | not identified |
| Bi | 1CR | 5 | Mealy Parrot | Psittaciformes | Psittacidae | Amazona | farinosa |
| Bi | 1CR | 2 | Motmot | Coraciiformes | Momotidae | not identified | not identified |
| Bi | 1CR | 125 | Oropendula | Passeriformes | Icteridae | Psarocolius | not identified |
| Bi | 1CR | 2 | Pale-vented Thrush | Passeriformes | Turdidae | Turdus | obsoletus |
| Bi | 1CR | 20 | Parrot | Psittaciformes | Psittacidae | not identified | not identified |
| Bi | 1CR | 14 | Parrot, large | Psittaciformes | Psittacidae | not identified | not identified |
| Bi | 1CR | 1 | Parrot, little | Psittaciformes | Psittacidae | not identified | not identified |
| Bi | 1CR | 1 | Rainbird | not identified | not identified | not identified | not identified |
| Bi | 1CR | 1 | Raptor, small | not identified | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|---------------------------|----------------|----------------|----------------|----------------|
| Bi | 1CR | 3 | Saltatron | not identified | not identified | not identified | not identified |
| Bi | 1CR | 12 | Scarlet-rumped Tanager | Passeriformes | Thraupidae | Ramphocelus | passerinii |
| Bi | 1CR | 30 | Seedeater | Passeriformes | Thraupidae | not identified | not identified |
| Bi | 1CR | 3 | Smooth-billed Ani | Cuculiformes | Cuculidae | Crotophaga | ani |
| Bi | 1CR | 47 | Songbird | Passeriformes | not identified | not identified | not identified |
| Bi | 1CR | 1 | Songbird, brown | Passeriformes | not identified | not identified | not identified |
| Bi | 1CR | 1 | Songbird, little | Passeriformes | not identified | not identified | not identified |
| Bi | 1CR | 1 | Squirrel Cuckoo | Cuculiformes | Cuculidae | Piaya | cayana |
| Bi | 1CR | 1 | Steep-forehead Flycatcher | Passeriformes | Tyrannidae | not identified | not identified |
| Bi | 1CR | 2 | Swallow | Passeriformes | Hirundinidae | not identified | not identified |
| Bi | 1CR | 1 | Swift | Apodiformes | Apodidae | not identified | not identified |
| Bi | 1CR | 10 | Tanager | Passeriformes | Thraupidae | not identified | not identified |
| Bi | 1CR | 1 | Thrush | Passeriformes | not identified | not identified | not identified |
| Bi | 1CR | 1 | Tick Bird | not identified | not identified | not identified | not identified |
| Bi | 1CR | 7 | Toucan | Piciformes | Ramphastidae | not identified | not identified |
| Bi | 1CR | 3 | Treecreper | Passeriformes | Certhiidae | not identified | not identified |
| Bi | 1CR | 12 | Turkey Vulture | Ciconiiformes | Ciconiidae | Cathartes | aura |
| Bi | 1CR | 3 | Vulture | Ciconiiformes | Ciconiidae | not identified | not identified |
| Bi | 1CR | 38 | Woodpecker | Piciformes | Picidae | not identified | not identified |
| Bi | 1CR | 4 | Yellow-bellied Flycatcher | Passeriformes | Tyrannidae | Empidonax | flaviventris |
| DT | 1CR | 1 | Butterfly, blue-black | Lepidoptera | not identified | not identified | not identified |
| DT | 1CR | 3 | Butterfly, small yellow | Lepidoptera | not identified | not identified | not identified |
| DT | 1CR | 3 | Butterfly, white | Lepidoptera | not identified | not identified | not identified |
| DT | 1CR | 8 | Butterfly, yellow | Lepidoptera | not identified | not identified | not identified |
| DT | 1CR | 3 | Frog, red Dendrobatus | Anura | Dendrobatidae | Dendrobates | pumilio |
| TW | 1CR | 116 | ant | Hymenoptera | Formicidae | not identified | not identified |
| TW | 1CR | 7 | ant, small | Hymenoptera | Formicidae | not identified | not identified |
| TW | 1CR | 1 | ant, winged | Hymenoptera | Formicidae | not identified | not identified |
| TW | 1CR | 1 | beetle, ground | Coleoptera | not identified | not identified | not identified |
| TW | 1CR | 2 | beetle, long & slim | Coleoptera | not identified | not identified | not identified |
| TW | 1CR | 1 | bug | Hemiptera | not identified | not identified | not identified |
| TW | 1CR | 16 | cricket | Orthoptera | not identified | not identified | not identified |
| TW | 1CR | 1 | moth | Lepidoptera | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|--------------------------|----------------|----------------|----------------|----------------|
| TW | 1CR | 1 | salamander | Caudata | not identified | not identified | not identified |
| TW | 1CR | 47 | spider | Araneae | not identified | not identified | not identified |
| TW | 1CR | 2 | wasp | Hymenoptera | not identified | not identified | not identified |
| Bi | 2Ni | 5 | Ani | Cuculiformes | Cuculidae | Crotophaga | not identified |
| Bi | 2Ni | 65 | Banded Wren | Passeriformes | Troglodytidae | Thryothorus | pleurostictus |
| Bi | 2Ni | 1 | Black-headed Trogon | Trogoniformes | Trogonidae | Trogon | melanocephalus |
| Bi | 2Ni | 4 | Brown-crested Flycatcher | Passeriformes | Tyrannidae | Myiarchus | tyrannulus |
| Bi | 2Ni | 4 | Cattle Egret | Ciconiiformes | Ardeidae | Bubulcus | ibis |
| Bi | 2Ni | 11 | Dove | Columbiformes | Columbidae | not identified | not identified |
| Bi | 2Ni | 11 | Flycatcher | Passeriformes | Tyrannidae | not identified | not identified |
| Bi | 2Ni | 17 | Gray Hawk | Ciconiiformes | Accipitridae | Buteo | nitidus |
| Bi | 2Ni | 3 | Great Kiskadee | Passeriformes | Tyrannidae | Pitangus | sulphuratus |
| Bi | 2Ni | 2 | Groove-billed Ani | Cuculiformes | Cuculidae | Crotophaga | sulcirostris |
| Bi | 2Ni | 7 | Hawk | Ciconiiformes | Accipitridae | not identified | not identified |
| Bi | 2Ni | 3 | Hoffmann's Woodpecker | Piciformes | Picidae | Melanerpes | hoffmannii |
| Bi | 2Ni | 4 | Hummingbird | Apodiformes | Trochilidae | not identified | not identified |
| Bi | 2Ni | 2 | Jay | Passeriformes | Corvidae | not identified | not identified |
| Bi | 2Ni | 1 | Magnificent Frigatebird | Ciconiiformes | Fregatidae | Fregata | magnificens |
| Bi | 2Ni | 1 | Masked Tityra | Passeriformes | Cotingidae | Tityra | semifasciata |
| Bi | 2Ni | 5 | Parakeet | Psittaciformes | Psittacidae | Aratinga | not identified |
| Bi | 2Ni | 14 | Parrot | Psittaciformes | Psittacidae | not identified | not identified |
| Bi | 2Ni | 9 | Parrot, large | Psittaciformes | Psittacidae | not identified | not identified |
| Bi | 2Ni | 1 | Pauraque | Strigiformes | Caprimulgidae | Nyctidromus | albicollis |
| Bi | 2Ni | 3 | Red-billed Pigeon | Columbiformes | Columbidae | Patagioenas | flavirostris |
| Bi | 2Ni | 1 | Seedeater | Passeriformes | Thraupidae | not identified | not identified |
| Bi | 2Ni | 13 | Songbird | Passeriformes | not identified | not identified | not identified |
| Bi | 2Ni | 6 | Swallow | Passeriformes | Hirundinidae | not identified | not identified |
| Bi | 2Ni | 1 | Swift | Apodiformes | Apodidae | not identified | not identified |
| Bi | 2Ni | 1 | Tanager | Passeriformes | Thraupidae | not identified | not identified |
| Bi | 2Ni | 41 | Turkey Vulture | Ciconiiformes | Ciconiidae | Cathartes | aura |
| Bi | 2Ni | 5 | unknown | not identified | not identified | not identified | not identified |
| Bi | 2Ni | 3 | Vaux's Swift | Apodiformes | Apodidae | Chaetura | vauxi |
| Bi | 2Ni | 1 | Vulture | Ciconiiformes | Ciconiidae | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|---------------------------|----------------|----------------|----------------|----------------|
| Bi | 2Ni | 99 | White-throated Magpie Jay | Passeriformes | Corvidae | Calocitta | formosa |
| Bi | 2Ni | 16 | Woodpecker | Piciformes | Picidae | not identified | not identified |
| Bi | 2Ni | 1 | Yellow-naped Parrot | Psittaciformes | Psittacidae | Amazona | auropalliata |
| DT | 2Ni | 2 | Butterfly, black-red | Lepidoptera | not identified | not identified | not identified |
| DT | 2Ni | 3 | Butterfly, black-yellow | Lepidoptera | not identified | not identified | not identified |
| DT | 2Ni | 3 | Butterfly, grey | Lepidoptera | not identified | not identified | not identified |
| DT | 2Ni | 1 | Butterfly, large yellow | Lepidoptera | not identified | not identified | not identified |
| DT | 2Ni | 2 | Butterfly, orange | Lepidoptera | not identified | not identified | not identified |
| DT | 2Ni | 2 | Butterfly, orange-white | Lepidoptera | not identified | not identified | not identified |
| DT | 2Ni | 1 | Butterfly, small black | Lepidoptera | not identified | not identified | not identified |
| DT | 2Ni | 1 | Butterfly, small white | Lepidoptera | not identified | not identified | not identified |
| DT | 2Ni | 1 | butterfly, swallowtail | Lepidoptera | Papilionidae | Papilio | not identified |
| DT | 2Ni | 36 | Butterfly, white | Lepidoptera | not identified | not identified | not identified |
| DT | 2Ni | 9 | Butterfly, yellow | Lepidoptera | not identified | not identified | not identified |
| TW | 2Ni | 58 | ant | Hymenoptera | Formicidae | not identified | not identified |
| TW | 2Ni | 4 | ant, red | Hymenoptera | Formicidae | not identified | not identified |
| TW | 2Ni | 9 | ant, small | Hymenoptera | Formicidae | not identified | not identified |
| TW | 2Ni | 1 | ant, small black | Hymenoptera | Formicidae | not identified | not identified |
| TW | 2Ni | 24 | ant, small red | Hymenoptera | Formicidae | not identified | not identified |
| TW | 2Ni | 1 | beetle, 1002-1004 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 6 | beetle, 866 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 39 | beetle, 868 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | beetle, 872 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 2 | beetle, 873 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 14 | beetle, 874 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | beetle, 884 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 4 | beetle, 891 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 4 | beetle, 893 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 5 | beetle, 929 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | beetle, 933 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | beetle, 934 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | beetle, 937 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | beetle, 939-941 | Coleoptera | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|----------------------|----------------|----------------|----------------|----------------|
| TW | 2Ni | 1 | beetle, 957 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | beetle, 999 | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 7 | beetle, ground | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 14 | beetle, other | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | beetle, small | Coleoptera | not identified | not identified | not identified |
| TW | 2Ni | 6 | bristletail | not identified | not identified | not identified | not identified |
| TW | 2Ni | 1 | bug, 1001 | Hemiptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | bug, 870 | Hemiptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | bug, 926 | Hemiptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | bug, 928 | Hemiptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | bug, 946 | Hemiptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | bug, 961 | Hemiptera | not identified | not identified | not identified |
| TW | 2Ni | 3 | bug, other | Hemiptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | bug, other red | Hemiptera | not identified | not identified | not identified |
| TW | 2Ni | 6 | caterpillar, 875 | Lepidoptera | not identified | not identified | not identified |
| TW | 2Ni | 5 | caterpillar, 877 | Lepidoptera | not identified | not identified | not identified |
| TW | 2Ni | 2 | caterpillar, 942-943 | Lepidoptera | not identified | not identified | not identified |
| TW | 2Ni | 58 | centipede, 881 | not identified | not identified | not identified | not identified |
| TW | 2Ni | 4 | centipede, 882 | not identified | not identified | not identified | not identified |
| TW | 2Ni | 3 | centipede, 944 | not identified | not identified | not identified | not identified |
| TW | 2Ni | 7 | cricket | Orthoptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | earth worm | Haptotaxida | not identified | not identified | not identified |
| TW | 2Ni | 2 | insect, 869 | not identified | not identified | not identified | not identified |
| TW | 2Ni | 7 | insect, other | not identified | not identified | not identified | not identified |
| TW | 2Ni | 36 | mite, red | Acariformes | Acariformes | not identified | not identified |
| TW | 2Ni | 1 | mite, red 925 | Acariformes | not identified | not identified | not identified |
| TW | 2Ni | 1 | moth | Lepidoptera | not identified | not identified | not identified |
| TW | 2Ni | 1 | scorpion, 938 | Scorpiones | not identified | not identified | not identified |
| TW | 2Ni | 1 | snail, 1006 | not identified | not identified | not identified | not identified |
| TW | 2Ni | 3 | spider | Araneae | not identified | not identified | not identified |
| TW | 2Ni | 1 | spider, black 892 | Araneae | not identified | not identified | not identified |
| TW | 2Ni | 1 | spider, red | Araneae | not identified | not identified | not identified |
| TW | 2Ni | 21 | spider, small | Araneae | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|----------------------------|----------------|----------------|----------------|----------------|
| TW | 2Ni | 2 | spider, small red | Araneae | not identified | not identified | not identified |
| TW | 2Ni | 85 | springtail | Collembola | not identified | not identified | not identified |
| TW | 2Ni | 1 | toad | Anura | Bufo | Bufo | not identified |
| TW | 2Ni | 12 | woodlouse | Isopoda | not identified | not identified | not identified |
| TW | 2Ni | 1 | woodlouse, 954-955 | Isopoda | not identified | not identified | not identified |
| TW | 2Ni | 1 | worm | not identified | not identified | not identified | not identified |
| Bi | 3AK | 3 | American Robin | Passeriformes | Turdidae | Turdus | migratorius |
| Bi | 3AK | 3 | Boreal Chickadee | Passeriformes | Paridae | Poecile | hudsonica |
| Bi | 3AK | 7 | Chickadee | Passeriformes | Paridae | not identified | not identified |
| Bi | 3AK | 7 | Corvidae | Passeriformes | Corvidae | not identified | not identified |
| Bi | 3AK | 2 | Dark-eyed Junco | Passeriformes | Emberizidae | Junco | hyemalis |
| Bi | 3AK | 14 | Gray Jay | Passeriformes | Corvidae | Perisoreus | canadensis |
| Bi | 3AK | 39 | Gull | Ciconiiformes | Laridae | not identified | not identified |
| Bi | 3AK | 3 | Junco | Passeriformes | Emberizidae | Junco | not identified |
| Bi | 3AK | 1 | Northern Flicker | Piciformes | Picidae | Colaptes | auratus |
| Bi | 3AK | 4 | Sandhill Crane | Gruiformes | Gruidae | Grus | canadensis |
| Bi | 3AK | 302 | Songbird | Passeriformes | not identified | not identified | not identified |
| Bi | 3AK | 99 | Sparrow | Passeriformes | Emberizidae | not identified | not identified |
| Bi | 3AK | 200 | squirrel | Rodentia | Sciuridae | Sciurus | not identified |
| Bi | 3AK | 1 | Tit | not identified | not identified | not identified | not identified |
| Bi | 3AK | 1 | White-crowned Sparrow | Passeriformes | Emberizidae | Zonotrichia | leucophrys |
| Bi | 3AK | 3 | Woodpecker | Piciformes | Picidae | not identified | not identified |
| Bi | 3AK | 3 | Yellow-rumped Warbler | Passeriformes | Parulidae | Dendroica | coronata |
| TW | 3AK | 12 | ant | Hymenoptera | Formicidae | not identified | not identified |
| TW | 3AK | 2 | ant, small | Hymenoptera | Formicidae | not identified | not identified |
| TW | 3AK | 1 | bee | Hymenoptera | not identified | not identified | not identified |
| TW | 3AK | 8 | beetle | Coleoptera | not identified | not identified | not identified |
| TW | 3AK | 17 | beetle, underground-hiding | Coleoptera | not identified | not identified | not identified |
| TW | 3AK | 1 | bug | Hemiptera | not identified | not identified | not identified |
| TW | 3AK | 1 | caterpillar, black-hairy | Lepidoptera | not identified | not identified | not identified |
| TW | 3AK | 2 | fly | Diptera | not identified | not identified | not identified |
| TW | 3AK | 1 | grasshopper | Orthoptera | Acrididae | not identified | not identified |
| TW | 3AK | 4 | green insect | not identified | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|---------------------|----------------|----------------|----------------|----------------|
| TW | 3AK | 2 | mite, red | Acariformes | not identified | not identified | not identified |
| TW | 3AK | 3 | mouse | Rodentia | Muridae | not identified | not identified |
| TW | 3AK | 7 | other insect | not identified | not identified | not identified | not identified |
| TW | 3AK | 91 | spider | Araneae | not identified | not identified | not identified |
| TW | 3AK | 1 | spider, small | Araneae | not identified | not identified | not identified |
| TW | 3AK | 3 | spider, small black | Araneae | not identified | not identified | not identified |
| TW | 3AK | 14 | spider, small red | Araneae | not identified | not identified | not identified |
| TW | 3AK | 7 | spider, tiny | Araneae | not identified | not identified | not identified |
| TW | 3AK | 56 | springtail | Collembola | not identified | not identified | not identified |
| TW | 3AK | 4 | woodlouse | Isopoda | not identified | not identified | not identified |
| Bi | 4Ru | 4 | Bird | not identified | not identified | not identified | not identified |
| Bi | 4Ru | 1 | Blue Flank juv | not identified | not identified | not identified | not identified |
| Bi | 4Ru | 2 | bluetail | Passeriformes | Muscicapidae | Tarsiger | not identified |
| Bi | 4Ru | 105 | Chickadee | Passeriformes | Paridae | not identified | not identified |
| Bi | 4Ru | 2 | Crow | Passeriformes | Corvidae | not identified | not identified |
| Bi | 4Ru | 2 | Dove | Columbiformes | Columbidae | not identified | not identified |
| Bi | 4Ru | 3 | Emberiza | Passeriformes | Emberizidae | Emberiza | not identified |
| Bi | 4Ru | 1 | Falcon | Ciconiiformes | Falconidae | Falco | not identified |
| Bi | 4Ru | 1 | Finch | Passeriformes | Fringillidae | not identified | not identified |
| Bi | 4Ru | 2 | Flycatcher | Passeriformes | Muscicapidae | not identified | not identified |
| Bi | 4Ru | 3 | Grasshopper Warbler | Passeriformes | Sylviidae | Locustella | naevia |
| Bi | 4Ru | 1 | Gull | Ciconiiformes | Laridae | not identified | not identified |
| Bi | 4Ru | 4 | Hazelgrouse | Galliformes | Phasianidae | Tetrastes | bonasia |
| Bi | 4Ru | 32 | Jungle Crow | Passeriformes | Corvidae | Corvus | levaillantii |
| Bi | 4Ru | 1 | Juv passerine | Passeriformes | not identified | not identified | not identified |
| Bi | 4Ru | 94 | Kinglet | Passeriformes | Regulidae | Regulus | not identified |
| Bi | 4Ru | 1 | Kohlmeise | Passeriformes | Paridae | Parus | major |
| Bi | 4Ru | 1 | longtailed tit | Passeriformes | Paridae | not identified | not identified |
| Bi | 4Ru | 3 | Merganser | Anseriformes | Anatidae | Mergus | merganser |
| Bi | 4Ru | 42 | Nutcracker | Passeriformes | Corvidae | Nucifraga | not identified |
| Bi | 4Ru | 3 | nuthatch | Passeriformes | Sittidae | Sitta | not identified |
| Bi | 4Ru | 6 | Oriental Dove | Columbiformes | Columbidae | Streptopelia | orientalis |
| Bi | 4Ru | 8 | Oriental Finch | Passeriformes | Fringillidae | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|------------------------|----------------|----------------|----------------|----------------|
| Bi | 4Ru | 10 | Oriental Greenfinch | Passeriformes | Fringillidae | Carduelis | sinica |
| Bi | 4Ru | 2 | Oriental Pigeon | Columbiformes | Columbidae | not identified | not identified |
| Bi | 4Ru | 18 | Pacific Swift | Apodiformes | Apodidae | Apus | pacificus |
| Bi | 4Ru | 4 | passerine | Passeriformes | not identified | not identified | not identified |
| Bi | 4Ru | 2 | Rain Call Bird | not identified | not identified | not identified | not identified |
| Bi | 4Ru | 5 | Raptor | not identified | not identified | not identified | not identified |
| Bi | 4Ru | 1 | Raven | Passeriformes | Corvidae | Corvus | not identified |
| Bi | 4Ru | 1 | stellers sea eagle juv | Ciconiiformes | Accipitridae | Haliaeetus | pelagicus |
| Bi | 4Ru | 2 | Tannenmeise | Passeriformes | Paridae | Periparus | ater |
| Bi | 4Ru | 2 | Teseewee | not identified | not identified | not identified | not identified |
| Bi | 4Ru | 1 | Thriller | not identified | not identified | not identified | not identified |
| Bi | 4Ru | 1 | Thrush | Passeriformes | Turdidae | not identified | not identified |
| Bi | 4Ru | 1 | Tistiwee | not identified | not identified | not identified | not identified |
| Bi | 4Ru | 1 | Titi titititi | not identified | not identified | not identified | not identified |
| Bi | 4Ru | 11 | Tsilp | not identified | not identified | not identified | not identified |
| Bi | 4Ru | 2 | Wagtail | Passeriformes | Motacillidae | Motacilla | not identified |
| Bi | 4Ru | 38 | Warbler | Passeriformes | not identified | not identified | not identified |
| Bi | 4Ru | 2 | Weidenmeise | Passeriformes | Paridae | Poecile | montana |
| Bi | 4Ru | 37 | Winter Wren | Passeriformes | Troglodytidae | Troglodytes | troglodytes |
| Bi | 4Ru | 30 | wize | not identified | not identified | not identified | not identified |
| Bi | 4Ru | 6 | wize wize | not identified | not identified | not identified | not identified |
| Bi | 4Ru | 10 | Woodpecker | Piciformes | Picidae | not identified | not identified |
| TW | 4Ru | 1 | aimbia | not identified | not identified | not identified | not identified |
| TW | 4Ru | 1 | aimbia, little | not identified | not identified | not identified | not identified |
| TW | 4Ru | 5 | Beetle | Coleoptera | not identified | not identified | not identified |
| TW | 4Ru | 3 | bibienka | not identified | not identified | not identified | not identified |
| TW | 4Ru | 6 | Carabidae | Coleoptera | Carabidae | not identified | not identified |
| TW | 4Ru | 1 | caterpillar | Lepidoptera | not identified | not identified | not identified |
| TW | 4Ru | 6 | Cenocosiets | Opiliones | not identified | not identified | not identified |
| TW | 4Ru | 3 | Cestianka | Lithobiomorpha | Lithobiidae | Lithobius | not identified |
| TW | 4Ru | 1 | changa | not identified | not identified | not identified | not identified |
| TW | 4Ru | 1 | changa (2) | not identified | not identified | not identified | not identified |
| TW | 4Ru | 1 | cinacost | not identified | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|----------------------------|----------------|----------------|----------------|----------------|
| TW | 4Ru | 54 | Collembola | Collembola | not identified | not identified | not identified |
| TW | 4Ru | 1 | collisea | not identified | not identified | not identified | not identified |
| TW | 4Ru | 4 | costianka | Lithobiomorpha | Lithobiidae | Lithobius | not identified |
| TW | 4Ru | 26 | cycsegusa | not identified | not identified | not identified | not identified |
| TW | 4Ru | 1 | expoxata | not identified | not identified | not identified | not identified |
| TW | 4Ru | 3 | fly little | Diptera | not identified | not identified | not identified |
| TW | 4Ru | 1 | fly, small special | Diptera | not identified | not identified | not identified |
| TW | 4Ru | 1 | insects | not identified | not identified | not identified | not identified |
| TW | 4Ru | 7 | mouse | Rodentia | Muridae | not identified | not identified |
| TW | 4Ru | 1 | nayesdink | not identified | not identified | not identified | not identified |
| TW | 4Ru | 1 | nayesdink, little | not identified | not identified | not identified | not identified |
| TW | 4Ru | 20 | Protura | Protura | not identified | not identified | not identified |
| TW | 4Ru | 1 | Sinocoset | Opiliones | not identified | not identified | not identified |
| TW | 4Ru | 11 | Spider | Araneae | not identified | not identified | not identified |
| TW | 4Ru | 6 | spider with slim long legs | Araneae | not identified | not identified | not identified |
| TW | 4Ru | 11 | spider, big | Araneae | not identified | not identified | not identified |
| TW | 4Ru | 25 | spider, little | Araneae | not identified | not identified | not identified |
| TW | 4Ru | 2 | spider, midsize | Araneae | not identified | not identified | not identified |
| TW | 4Ru | 1 | spider, palekolane | Araneae | not identified | not identified | not identified |
| TW | 4Ru | 6 | Staphilin | Staphylinidae | not identified | not identified | not identified |
| TW | 4Ru | 1 | tick | Ixodida | not identified | not identified | not identified |
| TW | 4Ru | 1 | ucene | not identified | not identified | not identified | not identified |
| TW | 4Ru | 17 | worm | not identified | not identified | not identified | not identified |
| Bi | 5PG | 6 | Balu | not identified | not identified | not identified | not identified |
| Bi | 5PG | 60 | bird | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Bird Chreak | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | bird fly over | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | bird, medium | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Birds of Prey | not identified | not identified | not identified | not identified |
| Bi | 5PG | 3 | Black Hawk | Ciconiiformes | Accipitridae | not identified | not identified |
| Bi | 5PG | 1 | Broken Flute | not identified | not identified | not identified | not identified |
| Bi | 5PG | 6 | call | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Canopy Bird | not identified | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|----------------------------|----------------|----------------|----------------|----------------|
| Bi | 5PG | 1 | Check bird | not identified | not identified | not identified | not identified |
| Bi | 5PG | 2 | Chickchickchickachick | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Chilk Twitz | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Chilp | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Ching | not identified | not identified | not identified | not identified |
| Bi | 5PG | 3 | Chirp | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | chirp loud | not identified | not identified | not identified | not identified |
| Bi | 5PG | 2 | Chitter | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Clink | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Cockatoo | Psittaciformes | Psittacidae | Cacatua | not identified |
| Bi | 5PG | 5 | Craw, Bird of Paradise | Passeriformes | Paradisaeidae | not identified | not identified |
| Bi | 5PG | 26 | dove | Columbiformes | Columbidae | not identified | not identified |
| Bi | 5PG | 1 | dove, psurr deep | Columbiformes | Columbidae | not identified | not identified |
| Bi | 5PG | 1 | falcon | Ciconiiformes | Falconidae | Falco | not identified |
| Bi | 5PG | 1 | feep | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Fiep | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Fitz | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | flowerpiercer | not identified | not identified | not identified | not identified |
| Bi | 5PG | 23 | Flute | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Flute melodious | not identified | not identified | not identified | not identified |
| Bi | 5PG | 2 | Flute song | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Flycatcher | Passeriformes | Monarchidae | not identified | not identified |
| Bi | 5PG | 1 | Flycatcher tschirrp | Passeriformes | Monarchidae | not identified | not identified |
| Bi | 5PG | 1 | Flycatcher, similar willie | Passeriformes | Monarchidae | not identified | not identified |
| Bi | 5PG | 10 | Fowl | Galliformes | not identified | not identified | not identified |
| Bi | 5PG | 2 | fruit pecker | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | fruit pecker white cheek | not identified | not identified | not identified | not identified |
| Bi | 5PG | 5 | gleaner | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | gleaner, white cheek | not identified | not identified | not identified | not identified |
| Bi | 5PG | 7 | Hawk | Ciconiiformes | Accipitridae | not identified | not identified |
| Bi | 5PG | 9 | Hornbill | Bucerotiformes | Bucerotidae | not identified | not identified |
| Bi | 5PG | 1 | Jackah call | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Jackljakl | not identified | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|-------------------------------------|----------------|----------------|----------------|----------------|
| Bi | 5PG | 2 | Kau Kau, Bird of Paradise | Passeriformes | Paradisaeidae | not identified | not identified |
| Bi | 5PG | 1 | loud call | not identified | not identified | not identified | not identified |
| Bi | 5PG | 8 | melodious song | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Melodious Song, like sylvia warbler | not identified | not identified | not identified | not identified |
| Bi | 5PG | 2 | Palm Cockatoo | Psittaciformes | Psittacidae | Probosciger | aterrimus |
| Bi | 5PG | 9 | Parakeet | Psittaciformes | Psittacidae | Aratinga | not identified |
| Bi | 5PG | 6 | Parrot | Psittaciformes | Psittacidae | not identified | not identified |
| Bi | 5PG | 2 | Parrot, little | Psittaciformes | Psittacidae | not identified | not identified |
| Bi | 5PG | 2 | pewee like North American Pewee | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Piwi | not identified | not identified | not identified | not identified |
| Bi | 5PG | 2 | pschorr | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Psitt | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Queek | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Quit | not identified | not identified | not identified | not identified |
| Bi | 5PG | 30 | Rezina, rezina | Passeriformes | Paradisaeidae | not identified | not identified |
| Bi | 5PG | 1 | schrill | not identified | not identified | not identified | not identified |
| Bi | 5PG | 3 | song | not identified | not identified | not identified | not identified |
| Bi | 5PG | 34 | Songbird | Passeriformes | not identified | not identified | not identified |
| Bi | 5PG | 1 | Songbird little | Passeriformes | not identified | not identified | not identified |
| Bi | 5PG | 4 | songbird tshirp | Passeriformes | not identified | not identified | not identified |
| Bi | 5PG | 1 | Songbird tsilp | Passeriformes | not identified | not identified | not identified |
| Bi | 5PG | 6 | Swallow | Passeriformes | Hirundinidae | not identified | not identified |
| Bi | 5PG | 7 | swirl | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | sylvia song | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Thrush | Passeriformes | Turdidae | not identified | not identified |
| Bi | 5PG | 1 | Trach trach | not identified | not identified | not identified | not identified |
| Bi | 5PG | 2 | tschick | not identified | not identified | not identified | not identified |
| Bi | 5PG | 2 | tschirp | not identified | not identified | not identified | not identified |
| Bi | 5PG | 2 | tsi tsi | not identified | not identified | not identified | not identified |
| Bi | 5PG | 50 | tsilp | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | Tsilp tsilp | not identified | not identified | not identified | not identified |
| Bi | 5PG | 2 | Tsirp | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | wae wae wae | not identified | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|--------------------------|----------------|----------------|----------------|----------------|
| Bi | 5PG | 1 | wake wake | not identified | not identified | not identified | not identified |
| Bi | 5PG | 1 | warning call | not identified | not identified | not identified | not identified |
| Bi | 5PG | 28 | White Cockatoo | Psittaciformes | Psittacidae | Cacatua | alba |
| Bi | 5PG | 1 | Wieeh | not identified | not identified | not identified | not identified |
| Bi | 5PG | 3 | Willie | not identified | not identified | not identified | not identified |
| Bi | 5PG | 2 | witz | not identified | not identified | not identified | not identified |
| Bi | 5PG | 5 | wiz wiz | not identified | not identified | not identified | not identified |
| Bi | 5PG | 6 | Wize wize | not identified | not identified | not identified | not identified |
| Bi | 5PG | 5 | woodpecker | Piciformes | Picidae | not identified | not identified |
| Bi | 5PG | 1 | wren | Passeriformes | Troglodytidae | not identified | not identified |
| TW | 5PG | 1 | ant, big | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 3 | ant, big black | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 8 | ant, big yellow | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 19 | ant, black | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 1 | ant, little | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 6 | ant, little black | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 2 | ant, little red | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 1 | ant, medium black | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 2 | ant, red | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 3 | ant, tiny | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 22 | ant, tiny black | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 12 | ant, tiny red | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 1 | ant, tiny yellow | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 5 | ant, yellow | Hymenoptera | Formicidae | not identified | not identified |
| TW | 5PG | 1 | bug | Hemiptera | not identified | not identified | not identified |
| TW | 5PG | 1 | bug, coackroach type | Dictyoptera | not identified | not identified | not identified |
| TW | 5PG | 1 | bug, little | Hemiptera | not identified | not identified | not identified |
| TW | 5PG | 1 | bug, medium | Hemiptera | not identified | not identified | not identified |
| TW | 5PG | 1 | bug, tiny | Hemiptera | not identified | not identified | not identified |
| TW | 5PG | 1 | caterpillar | Lepidoptera | not identified | not identified | not identified |
| TW | 5PG | 45 | Collembola | Collembola | not identified | not identified | not identified |
| TW | 5PG | 2 | collembola long antennae | Collembola | not identified | not identified | not identified |
| TW | 5PG | 3 | collembola, big yellow | Collembola | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|--------------------------------------|----------------|----------------|----------------|----------------|
| TW | 5PG | 1 | collembola, black-yellow | Collembola | not identified | not identified | not identified |
| TW | 5PG | 2 | collembola, yellow | Collembola | not identified | not identified | not identified |
| TW | 5PG | 3 | earth grille, mid size long antennae | Orthoptera | not identified | not identified | not identified |
| TW | 5PG | 1 | eintagsfliege, 4 wing | Ephemeroptera | not identified | not identified | not identified |
| TW | 5PG | 1 | floh | Siphonaptera | not identified | not identified | not identified |
| TW | 5PG | 6 | fly | Diptera | not identified | not identified | not identified |
| TW | 5PG | 1 | fly with legs and antennae | Diptera | not identified | not identified | not identified |
| TW | 5PG | 1 | fly, tiny | Diptera | not identified | not identified | not identified |
| TW | 5PG | 7 | fruitfly | Diptera | not identified | not identified | not identified |
| TW | 5PG | 1 | fruitfly black | Diptera | not identified | not identified | not identified |
| TW | 5PG | 1 | fruitfly grey | Diptera | not identified | not identified | not identified |
| TW | 5PG | 2 | fruitfly, blue | Diptera | not identified | not identified | not identified |
| TW | 5PG | 1 | fruitfly, pink | Diptera | not identified | not identified | not identified |
| TW | 5PG | 9 | fruitfly, white | Diptera | not identified | not identified | not identified |
| TW | 5PG | 1 | Grille small | Orthoptera | not identified | not identified | not identified |
| TW | 5PG | 1 | grille, mid size long antennae | Orthoptera | not identified | not identified | not identified |
| TW | 5PG | 1 | insect | not identified | not identified | not identified | not identified |
| TW | 5PG | 1 | kaefer middle | Coleoptera | not identified | not identified | not identified |
| TW | 5PG | 1 | laufkaefer | Coleoptera | Carabidae | not identified | not identified |
| TW | 5PG | 3 | marienkaeferlarve | Coleoptera | Coccinellidae | not identified | not identified |
| TW | 5PG | 3 | milbe, red | Acariformes | not identified | not identified | not identified |
| TW | 5PG | 2 | milbe, spring | Acariformes | not identified | not identified | not identified |
| TW | 5PG | 2 | millipede, big | not identified | not identified | not identified | not identified |
| TW | 5PG | 1 | millipede, small | not identified | not identified | not identified | not identified |
| TW | 5PG | 1 | miskaefer medium | Coleoptera | Geotrupidae | not identified | not identified |
| TW | 5PG | 1 | mosquito | Diptera | Culicidae | not identified | not identified |
| TW | 5PG | 1 | mosquito, jumping | Diptera | Culicidae | not identified | not identified |
| TW | 5PG | 1 | rainworm | Haplotaxida | Lumbricidae | not identified | not identified |
| TW | 5PG | 2 | rainworm little | Haplotaxida | Lumbricidae | not identified | not identified |
| TW | 5PG | 1 | schnellkaefer | Coleoptera | Elateridae | not identified | not identified |
| TW | 5PG | 4 | spider, little | Araneae | not identified | not identified | not identified |
| TW | 5PG | 3 | spider, little black | Araneae | not identified | not identified | not identified |
| TW | 5PG | 3 | spider, little long legs | Araneae | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|---------------------------|----------------|----------------|----------------|----------------|
| TW | 5PG | 1 | spider, medium | Araneae | not identified | not identified | not identified |
| TW | 5PG | 1 | spider, tiny black | Araneae | not identified | not identified | not identified |
| TW | 5PG | 2 | spring floh, blue | not identified | not identified | not identified | not identified |
| TW | 5PG | 8 | springfloh | not identified | not identified | not identified | not identified |
| TW | 5PG | 5 | springfloh mit antennae | not identified | not identified | not identified | not identified |
| TW | 5PG | 1 | Springfloh yellow | not identified | not identified | not identified | not identified |
| TW | 5PG | 5 | springfloh, long antennae | not identified | not identified | not identified | not identified |
| TW | 5PG | 1 | tausenfuesser medium | not identified | not identified | not identified | not identified |
| TW | 5PG | 1 | weevil medium | Coleoptera | Curculionidae | not identified | not identified |
| TW | 5PG | 1 | wurm | not identified | not identified | not identified | not identified |
| Bi | 6Ba | 1 | Brant | Anseriformes | Anatidae | Branta | bernicla |
| Bi | 6Ba | 1 | Dowitcher | Ciconiiformes | Scolopacidae | Limnodromus | not identified |
| Bi | 6Ba | 18 | Dunlin | Ciconiiformes | Scolopacidae | Calidris | alpina |
| Bi | 6Ba | 2 | Eider Duck | Anseriformes | Anatidae | Somateria | mollissima |
| Bi | 6Ba | 1 | Glaucous Gull | Ciconiiformes | Laridae | Larus | hyperboreus |
| Bi | 6Ba | 112 | Lapland Bunting | Passeriformes | Emberizidae | Calcarius | lapponicus |
| Bi | 6Ba | 1 | Lemming | Rodentia | Muridae | Lemmus | not identified |
| Bi | 6Ba | 48 | Longbilled Dowitcher | Ciconiiformes | Scolopacidae | Limnodromus | scolopaceus |
| Bi | 6Ba | 1 | Longtailed Duck | Anseriformes | Anatidae | Clangula | hyemalis |
| Bi | 6Ba | 1 | Loon | Ciconiiformes | Gaviidae | Gavia | immer |
| Bi | 6Ba | 1 | Pacific Loon | Ciconiiformes | Gaviidae | Gavia | pacifica |
| Bi | 6Ba | 1 | Parasitic Jaeger | Ciconiiformes | Stercorariidae | Stercorarius | parasiticus |
| Bi | 6Ba | 37 | Pectoral Sandpiper | Ciconiiformes | Scolopacidae | Calidris | melanotos |
| Bi | 6Ba | 1 | Phalarope | Ciconiiformes | Scolopacidae | Phalaropus | not identified |
| Bi | 6Ba | 49 | Pomarine Jaeger | Ciconiiformes | Stercorariidae | Stercorarius | pomarinus |
| Bi | 6Ba | 59 | Red Phalarope | Ciconiiformes | Scolopacidae | Phalaropus | fulicarius |
| Bi | 6Ba | 13 | Red-necked Phalarope | Ciconiiformes | Scolopacidae | Phalaropus | lobatus |
| Bi | 6Ba | 65 | Semipalmated Sandpiper | Ciconiiformes | Scolopacidae | Calidris | pusilla |
| Bi | 6Ba | 1 | Snow Bunting | Passeriformes | Emberizidae | Plectrophenax | nivalis |
| Bi | 6Ba | 2 | Spectacled Eider | Anseriformes | Anatidae | Somateria | fischeri |
| Bi | 6Ba | 1 | Swans | Anseriformes | Anatidae | Cygnus | not identified |
| Bi | 6Ba | 3 | Western Sandpiper | Ciconiiformes | Scolopacidae | Calidris | mauri |
| TW | 6Ba | 4 | beetle | Coleoptera | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|----------------------|----------------|----------------|----------------|----------------|
| TW | 6Ba | 83 | beetle, flat | Coleoptera | not identified | not identified | not identified |
| TW | 6Ba | 1 | beetle, gold-green | Coleoptera | not identified | not identified | not identified |
| TW | 6Ba | 1 | beetle, green | Coleoptera | not identified | not identified | not identified |
| TW | 6Ba | 1 | beetle, little | Coleoptera | not identified | not identified | not identified |
| TW | 6Ba | 1 | beetle, little green | Coleoptera | not identified | not identified | not identified |
| TW | 6Ba | 6 | beetle, slim | Coleoptera | not identified | not identified | not identified |
| TW | 6Ba | 1 | caterpillar worm | Lepidoptera | not identified | not identified | not identified |
| TW | 6Ba | 1 | caterpillar, big | Lepidoptera | not identified | not identified | not identified |
| TW | 6Ba | 1 | caterpillar, hairy | Lepidoptera | not identified | not identified | not identified |
| TW | 6Ba | 1 | caterpillar, small | Lepidoptera | not identified | not identified | not identified |
| TW | 6Ba | 37 | fly | Diptera | not identified | not identified | not identified |
| TW | 6Ba | 2 | fly, little | Diptera | not identified | not identified | not identified |
| TW | 6Ba | 32 | Fruitfly | Diptera | not identified | not identified | not identified |
| TW | 6Ba | 3 | fruitfly, little | Diptera | not identified | not identified | not identified |
| TW | 6Ba | 8 | fruitfly, tiny | Diptera | not identified | not identified | not identified |
| TW | 6Ba | 3 | larvae | not identified | not identified | not identified | not identified |
| TW | 6Ba | 1 | larvae big | not identified | not identified | not identified | not identified |
| TW | 6Ba | 1 | larvae with legs | not identified | not identified | not identified | not identified |
| TW | 6Ba | 1 | larvae, long | not identified | not identified | not identified | not identified |
| TW | 6Ba | 1 | Marienkaeferlarve | Coleoptera | Coccinellidae | not identified | not identified |
| TW | 6Ba | 1 | Microworm | not identified | not identified | not identified | not identified |
| TW | 6Ba | 20 | Milbe | Acariformes | not identified | not identified | not identified |
| TW | 6Ba | 2 | Milbe, micro | Acariformes | not identified | not identified | not identified |
| TW | 6Ba | 1 | Milbe, tiny | Acariformes | not identified | not identified | not identified |
| TW | 6Ba | 22 | mosquito | Diptera | Culicidae | not identified | not identified |
| TW | 6Ba | 1 | Rueckenschwimmkaefer | Hemiptera | Notonectidae | not identified | not identified |
| TW | 6Ba | 22 | Schuster | Diptera | Tipulidae | not identified | not identified |
| TW | 6Ba | 2 | schuster, big | Diptera | Tipulidae | not identified | not identified |
| TW | 6Ba | 1 | Schuster, large | Diptera | Tipulidae | not identified | not identified |
| TW | 6Ba | 1 | Schuster, no wings | Diptera | Tipulidae | not identified | not identified |
| TW | 6Ba | 61 | spider | Araneae | not identified | not identified | not identified |
| TW | 6Ba | 1 | spider (underwater) | Araneae | not identified | not identified | not identified |
| TW | 6Ba | 3 | spider, big | Araneae | not identified | not identified | not identified |

| Type | Study area | No of Obs. | Narrative | Order | Family | Genus | Species |
|------|------------|------------|-------------------|-------------|----------------|----------------|----------------|
| TW | 6Ba | 15 | spider, little | Araneae | not identified | not identified | not identified |
| TW | 6Ba | 125 | spider, tiny | Araneae | not identified | not identified | not identified |
| TW | 6Ba | 8 | Springmilbe | Acariformes | not identified | not identified | not identified |
| TW | 6Ba | 1 | Springmilbe, tiny | Acariformes | not identified | not identified | not identified |
| TW | 6Ba | 3 | springschwanz | Collembola | not identified | not identified | not identified |

7.6 Random Forests Models with Highest ROC Values

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------|-----------|
| 1CR | Bi | all | <i>Covariates</i> | Ani | 0.856 |
| 1CR | Bi | aur | <i>Covariates</i> | Ani | n/a |
| 1CR | Bi | ran | <i>Covariates</i> | Ani | 0.927 |
| 1CR | Bi | sys | <i>Covariates</i> | Ani | 0.751 |
| 1CR | Bi | vis | <i>Covariates</i> | Ani | 0.714 |
| 1CR | Bi | all | <i>Interspecies</i> | Ani | 0.858 |
| 1CR | Bi | aur | <i>Interspecies</i> | Ani | n/a |
| 1CR | Bi | ran | <i>Interspecies</i> | Ani | 0.927 |
| 1CR | Bi | sys | <i>Interspecies</i> | Ani | 0.728 |
| 1CR | Bi | vis | <i>Interspecies</i> | Ani | 0.714 |
| 1CR | Bi | all | <i>Covariates</i> | Dove | 0.344 |
| 1CR | Bi | aur | <i>Covariates</i> | Dove | 0.65 |
| 1CR | Bi | ran | <i>Covariates</i> | Dove | 0.032 |
| 1CR | Bi | sys | <i>Covariates</i> | Dove | 0.496 |
| 1CR | Bi | vis | <i>Covariates</i> | Dove | 0.042 |
| 1CR | Bi | all | <i>Interspecies</i> | Dove | 0.374 |
| 1CR | Bi | aur | <i>Interspecies</i> | Dove | 0.672 |
| 1CR | Bi | ran | <i>Interspecies</i> | Dove | 0.032 |
| 1CR | Bi | sys | <i>Interspecies</i> | Dove | 0.496 |
| 1CR | Bi | vis | <i>Interspecies</i> | Dove | 0.042 |
| 1CR | Bi | all | <i>Covariates</i> | Flycatcher | 0.623 |
| 1CR | Bi | aur | <i>Covariates</i> | Flycatcher | 0.626 |
| 1CR | Bi | ran | <i>Covariates</i> | Flycatcher | 0.414 |
| 1CR | Bi | sys | <i>Covariates</i> | Flycatcher | 0.620 |
| 1CR | Bi | vis | <i>Covariates</i> | Flycatcher | 0.654 |
| 1CR | Bi | all | <i>Interspecies</i> | Flycatcher | 0.578 |
| 1CR | Bi | aur | <i>Interspecies</i> | Flycatcher | 0.581 |
| 1CR | Bi | ran | <i>Interspecies</i> | Flycatcher | 0.327 |
| 1CR | Bi | sys | <i>Interspecies</i> | Flycatcher | 0.58 |
| 1CR | Bi | vis | <i>Interspecies</i> | Flycatcher | 0.649 |
| 1CR | Bi | all | <i>Covariates</i> | Great Kiskadee | 0.698 |
| 1CR | Bi | aur | <i>Covariates</i> | Great Kiskadee | 0.697 |
| 1CR | Bi | ran | <i>Covariates</i> | Great Kiskadee | 0.753 |
| 1CR | Bi | sys | <i>Covariates</i> | Great Kiskadee | 0.668 |
| 1CR | Bi | vis | <i>Covariates</i> | Great Kiskadee | 0.643 |
| 1CR | Bi | all | <i>Interspecies</i> | Great Kiskadee | 0.698 |
| 1CR | Bi | aur | <i>Interspecies</i> | Great Kiskadee | 0.732 |
| 1CR | Bi | ran | <i>Interspecies</i> | Great Kiskadee | 0.724 |
| 1CR | Bi | sys | <i>Interspecies</i> | Great Kiskadee | 0.662 |
| 1CR | Bi | vis | <i>Interspecies</i> | Great Kiskadee | 0.659 |
| 1CR | Bi | all | <i>Covariates</i> | Hummingbird | 0.756 |
| 1CR | Bi | aur | <i>Covariates</i> | Hummingbird | 0.608 |
| 1CR | Bi | ran | <i>Covariates</i> | Hummingbird | 0.704 |
| 1CR | Bi | sys | <i>Covariates</i> | Hummingbird | 0.752 |
| 1CR | Bi | vis | <i>Covariates</i> | Hummingbird | 0.813 |
| 1CR | Bi | all | <i>Interspecies</i> | Hummingbird | 0.738 |
| 1CR | Bi | aur | <i>Interspecies</i> | Hummingbird | 0.584 |
| 1CR | Bi | ran | <i>Interspecies</i> | Hummingbird | 0.692 |
| 1CR | Bi | sys | <i>Interspecies</i> | Hummingbird | 0.734 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------|-----------|
| 1CR | Bi | vis | <i>Interspecies</i> | Hummingbird | 0.797 |
| 1CR | Bi | all | <i>Covariates</i> | Kiskadee | 0.655 |
| 1CR | Bi | aur | <i>Covariates</i> | Kiskadee | 0.648 |
| 1CR | Bi | ran | <i>Covariates</i> | Kiskadee | n/a |
| 1CR | Bi | sys | <i>Covariates</i> | Kiskadee | 0.666 |
| 1CR | Bi | vis | <i>Covariates</i> | Kiskadee | 0.046 |
| 1CR | Bi | all | <i>Interspecies</i> | Kiskadee | 0.679 |
| 1CR | Bi | aur | <i>Interspecies</i> | Kiskadee | 0.645 |
| 1CR | Bi | ran | <i>Interspecies</i> | Kiskadee | n/a |
| 1CR | Bi | sys | <i>Interspecies</i> | Kiskadee | 0.655 |
| 1CR | Bi | vis | <i>Interspecies</i> | Kiskadee | 0.046 |
| 1CR | Bi | all | <i>Covariates</i> | Manakin | 0.851 |
| 1CR | Bi | aur | <i>Covariates</i> | Manakin | 0.741 |
| 1CR | Bi | ran | <i>Covariates</i> | Manakin | 0.753 |
| 1CR | Bi | sys | <i>Covariates</i> | Manakin | 0.815 |
| 1CR | Bi | vis | <i>Covariates</i> | Manakin | n/a |
| 1CR | Bi | all | <i>Interspecies</i> | Manakin | 0.857 |
| 1CR | Bi | aur | <i>Interspecies</i> | Manakin | 0.741 |
| 1CR | Bi | ran | <i>Interspecies</i> | Manakin | 0.709 |
| 1CR | Bi | sys | <i>Interspecies</i> | Manakin | 0.825 |
| 1CR | Bi | vis | <i>Interspecies</i> | Manakin | n/a |
| 1CR | Bi | all | <i>Plot</i> | Manakin | 0.646 |
| 1CR | Bi | all | <i>Covariates</i> | Mealy Parrot | 0.930 |
| 1CR | Bi | aur | <i>Covariates</i> | Mealy Parrot | n/a |
| 1CR | Bi | ran | <i>Covariates</i> | Mealy Parrot | n/a |
| 1CR | Bi | sys | <i>Covariates</i> | Mealy Parrot | 0.883 |
| 1CR | Bi | vis | <i>Covariates</i> | Mealy Parrot | 0.87 |
| 1CR | Bi | all | <i>Interspecies</i> | Mealy Parrot | 0.908 |
| 1CR | Bi | aur | <i>Interspecies</i> | Mealy Parrot | n/a |
| 1CR | Bi | ran | <i>Interspecies</i> | Mealy Parrot | n/a |
| 1CR | Bi | sys | <i>Interspecies</i> | Mealy Parrot | 0.883 |
| 1CR | Bi | vis | <i>Interspecies</i> | Mealy Parrot | 0.892 |
| 1CR | Bi | all | <i>Plot</i> | Mealy Parrot | 0.063 |
| 1CR | Bi | all | <i>Covariates</i> | Oropendula | 0.631 |
| 1CR | Bi | aur | <i>Covariates</i> | Oropendula | 0.535 |
| 1CR | Bi | ran | <i>Covariates</i> | Oropendula | 0.524 |
| 1CR | Bi | sys | <i>Covariates</i> | Oropendula | 0.641 |
| 1CR | Bi | vis | <i>Covariates</i> | Oropendula | 0.6 |
| 1CR | Bi | all | <i>Interspecies</i> | Oropendula | 0.635 |
| 1CR | Bi | aur | <i>Interspecies</i> | Oropendula | 0.524 |
| 1CR | Bi | ran | <i>Interspecies</i> | Oropendula | 0.505 |
| 1CR | Bi | sys | <i>Interspecies</i> | Oropendula | 0.642 |
| 1CR | Bi | vis | <i>Interspecies</i> | Oropendula | 0.576 |
| 1CR | Bi | all | <i>Plot</i> | Oropendula | 0.018 |
| 1CR | Bi | all | <i>Covariates</i> | Parrot | 0.820 |
| 1CR | Bi | aur | <i>Covariates</i> | Parrot | 0.897 |
| 1CR | Bi | ran | <i>Covariates</i> | Parrot | 0.037 |
| 1CR | Bi | sys | <i>Covariates</i> | Parrot | 0.951 |
| 1CR | Bi | vis | <i>Covariates</i> | Parrot | 0.753 |
| 1CR | Bi | all | <i>Interspecies</i> | Parrot | 0.818 |
| 1CR | Bi | aur | <i>Interspecies</i> | Parrot | 0.897 |
| 1CR | Bi | ran | <i>Interspecies</i> | Parrot | 0.032 |
| 1CR | Bi | sys | <i>Interspecies</i> | Parrot | 0.906 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------------|-----------|
| 1CR | Bi | vis | <i>Interspecies</i> | Parrot | 0.789 |
| 1CR | Bi | all | <i>Plot</i> | Parrot | 0.438 |
| 1CR | Bi | all | <i>Covariates</i> | Parrot, large | 0.870 |
| 1CR | Bi | aur | <i>Covariates</i> | Parrot, large | 0.804 |
| 1CR | Bi | ran | <i>Covariates</i> | Parrot, large | n/a |
| 1CR | Bi | sys | <i>Covariates</i> | Parrot, large | 0.841 |
| 1CR | Bi | vis | <i>Covariates</i> | Parrot, large | 0.916 |
| 1CR | Bi | all | <i>Interspecies</i> | Parrot, large | 0.847 |
| 1CR | Bi | aur | <i>Interspecies</i> | Parrot, large | 0.804 |
| 1CR | Bi | ran | <i>Interspecies</i> | Parrot, large | n/a |
| 1CR | Bi | sys | <i>Interspecies</i> | Parrot, large | 0.846 |
| 1CR | Bi | vis | <i>Interspecies</i> | Parrot, large | 0.918 |
| 1CR | Bi | all | <i>Plot</i> | Parrot, large | 0.476 |
| 1CR | Bi | all | <i>Covariates</i> | Scarlet-rumped Tanager | 0.747 |
| 1CR | Bi | aur | <i>Covariates</i> | Scarlet-rumped Tanager | n/a |
| 1CR | Bi | ran | <i>Covariates</i> | Scarlet-rumped Tanager | 0.463 |
| 1CR | Bi | sys | <i>Covariates</i> | Scarlet-rumped Tanager | 0.729 |
| 1CR | Bi | vis | <i>Covariates</i> | Scarlet-rumped Tanager | 0.481 |
| 1CR | Bi | all | <i>Interspecies</i> | Scarlet-rumped Tanager | 0.748 |
| 1CR | Bi | aur | <i>Interspecies</i> | Scarlet-rumped Tanager | n/a |
| 1CR | Bi | ran | <i>Interspecies</i> | Scarlet-rumped Tanager | 0.276 |
| 1CR | Bi | sys | <i>Interspecies</i> | Scarlet-rumped Tanager | 0.729 |
| 1CR | Bi | vis | <i>Interspecies</i> | Scarlet-rumped Tanager | 0.481 |
| 1CR | Bi | all | <i>Plot</i> | Scarlet-rumped Tanager | 0.938 |
| 1CR | Bi | all | <i>Covariates</i> | Seedeater | 0.798 |
| 1CR | Bi | aur | <i>Covariates</i> | Seedeater | 0.731 |
| 1CR | Bi | ran | <i>Covariates</i> | Seedeater | 0.774 |
| 1CR | Bi | sys | <i>Covariates</i> | Seedeater | 0.851 |
| 1CR | Bi | vis | <i>Covariates</i> | Seedeater | 0.721 |
| 1CR | Bi | all | <i>Interspecies</i> | Seedeater | 0.812 |
| 1CR | Bi | aur | <i>Interspecies</i> | Seedeater | 0.636 |
| 1CR | Bi | ran | <i>Interspecies</i> | Seedeater | 0.752 |
| 1CR | Bi | sys | <i>Interspecies</i> | Seedeater | 0.858 |
| 1CR | Bi | vis | <i>Interspecies</i> | Seedeater | 0.711 |
| 1CR | Bi | all | <i>Plot</i> | Seedeater | 0.962 |
| 1CR | Bi | all | <i>Covariates</i> | Tanager | 0.811 |
| 1CR | Bi | aur | <i>Covariates</i> | Tanager | n/a |
| 1CR | Bi | ran | <i>Covariates</i> | Tanager | 0.853 |
| 1CR | Bi | sys | <i>Covariates</i> | Tanager | 0.820 |
| 1CR | Bi | vis | <i>Covariates</i> | Tanager | 0.631 |
| 1CR | Bi | all | <i>Interspecies</i> | Tanager | 0.797 |
| 1CR | Bi | aur | <i>Interspecies</i> | Tanager | n/a |
| 1CR | Bi | ran | <i>Interspecies</i> | Tanager | 0.853 |
| 1CR | Bi | sys | <i>Interspecies</i> | Tanager | 0.795 |
| 1CR | Bi | vis | <i>Interspecies</i> | Tanager | 0.599 |
| 1CR | Bi | all | <i>Plot</i> | Tanager | 0.958 |
| 1CR | Bi | all | <i>Covariates</i> | Toucan | 0.706 |
| 1CR | Bi | aur | <i>Covariates</i> | Toucan | 0.045 |
| 1CR | Bi | ran | <i>Covariates</i> | Toucan | n/a |
| 1CR | Bi | sys | <i>Covariates</i> | Toucan | 0.649 |
| 1CR | Bi | vis | <i>Covariates</i> | Toucan | 0.801 |
| 1CR | Bi | all | <i>Interspecies</i> | Toucan | 0.652 |
| 1CR | Bi | aur | <i>Interspecies</i> | Toucan | 0.045 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------|-----------|
| 1CR | Bi | ran | <i>Interspecies</i> | Toucan | n/a |
| 1CR | Bi | sys | <i>Interspecies</i> | Toucan | 0.595 |
| 1CR | Bi | vis | <i>Interspecies</i> | Toucan | 0.809 |
| 1CR | Bi | all | <i>Plot</i> | Toucan | 0.567 |
| 1CR | Bi | all | <i>Covariates</i> | Turkey Vulture | 0.622 |
| 1CR | Bi | aur | <i>Covariates</i> | Turkey Vulture | n/a |
| 1CR | Bi | ran | <i>Covariates</i> | Turkey Vulture | n/a |
| 1CR | Bi | sys | <i>Covariates</i> | Turkey Vulture | 0.726 |
| 1CR | Bi | vis | <i>Covariates</i> | Turkey Vulture | 0.32 |
| 1CR | Bi | all | <i>Interspecies</i> | Turkey Vulture | 0.624 |
| 1CR | Bi | aur | <i>Interspecies</i> | Turkey Vulture | n/a |
| 1CR | Bi | ran | <i>Interspecies</i> | Turkey Vulture | n/a |
| 1CR | Bi | sys | <i>Interspecies</i> | Turkey Vulture | 0.672 |
| 1CR | Bi | vis | <i>Interspecies</i> | Turkey Vulture | 0.182 |
| 1CR | Bi | all | <i>Plot</i> | Turkey Vulture | 0.768 |
| 1CR | Bi | all | <i>Covariates</i> | Woodpecker | 0.585 |
| 1CR | Bi | aur | <i>Covariates</i> | Woodpecker | 0.73 |
| 1CR | Bi | ran | <i>Covariates</i> | Woodpecker | 0.648 |
| 1CR | Bi | sys | <i>Covariates</i> | Woodpecker | 0.553 |
| 1CR | Bi | vis | <i>Covariates</i> | Woodpecker | 0.392 |
| 1CR | Bi | all | <i>Interspecies</i> | Woodpecker | 0.577 |
| 1CR | Bi | aur | <i>Interspecies</i> | Woodpecker | 0.732 |
| 1CR | Bi | ran | <i>Interspecies</i> | Woodpecker | 0.581 |
| 1CR | Bi | sys | <i>Interspecies</i> | Woodpecker | 0.535 |
| 1CR | Bi | vis | <i>Interspecies</i> | Woodpecker | 0.366 |
| 1CR | Bi | all | <i>Plot</i> | Woodpecker | 0.377 |
| 2Ni | Bi | all | <i>Covariates</i> | Ani | 0.591 |
| 2Ni | Bi | all | <i>Interspecies</i> | Ani | 0.321 |
| 2Ni | Bi | all | <i>Plot</i> | Ani | 0.385 |
| 2Ni | Bi | aur | <i>Covariates</i> | Ani | 0.076 |
| 2Ni | Bi | aur | <i>Interspecies</i> | Ani | 0.038 |
| 2Ni | Bi | ran | <i>Covariates</i> | Ani | 0.036 |
| 2Ni | Bi | ran | <i>Interspecies</i> | Ani | 0.036 |
| 2Ni | Bi | sys | <i>Covariates</i> | Ani | 0.695 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Ani | 0.498 |
| 2Ni | Bi | vis | <i>Covariates</i> | Ani | 0.336 |
| 2Ni | Bi | vis | <i>Interspecies</i> | Ani | 0.211 |
| 2Ni | Bi | all | <i>Covariates</i> | Banded Wren | 0.780 |
| 2Ni | Bi | all | <i>Interspecies</i> | Banded Wren | 0.757 |
| 2Ni | Bi | all | <i>Plot</i> | Banded Wren | 0.674 |
| 2Ni | Bi | aur | <i>Covariates</i> | Banded Wren | 0.541 |
| 2Ni | Bi | aur | <i>Interspecies</i> | Banded Wren | 0.54 |
| 2Ni | Bi | ran | <i>Covariates</i> | Banded Wren | 0.67 |
| 2Ni | Bi | ran | <i>Interspecies</i> | Banded Wren | 0.604 |
| 2Ni | Bi | sys | <i>Covariates</i> | Banded Wren | 0.790 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Banded Wren | 0.756 |
| 2Ni | Bi | vis | <i>Covariates</i> | Banded Wren | 0.398 |
| 2Ni | Bi | vis | <i>Interspecies</i> | Banded Wren | 0.272 |
| 2Ni | Bi | all | <i>Covariates</i> | Dove | 0.747 |
| 2Ni | Bi | all | <i>Interspecies</i> | Dove | 0.679 |
| 2Ni | Bi | all | <i>Plot</i> | Dove | 0.693 |
| 2Ni | Bi | aur | <i>Covariates</i> | Dove | 0.508 |
| 2Ni | Bi | aur | <i>Interspecies</i> | Dove | 0.505 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------|-----------|
| 2Ni | Bi | ran | <i>Covariates</i> | Dove | n/a |
| 2Ni | Bi | ran | <i>Interspecies</i> | Dove | n/a |
| 2Ni | Bi | sys | <i>Covariates</i> | Dove | 0.695 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Dove | 0.623 |
| 2Ni | Bi | vis | <i>Covariates</i> | Dove | 0.833 |
| 2Ni | Bi | vis | <i>Interspecies</i> | Dove | 0.834 |
| 2Ni | Bi | all | <i>Covariates</i> | Flycatcher | 0.717 |
| 2Ni | Bi | all | <i>Interspecies</i> | Flycatcher | 0.637 |
| 2Ni | Bi | all | <i>Plot</i> | Flycatcher | 0.604 |
| 2Ni | Bi | aur | <i>Covariates</i> | Flycatcher | 0.642 |
| 2Ni | Bi | aur | <i>Interspecies</i> | Flycatcher | 0.787 |
| 2Ni | Bi | ran | <i>Covariates</i> | Flycatcher | 0.024 |
| 2Ni | Bi | ran | <i>Interspecies</i> | Flycatcher | 0.152 |
| 2Ni | Bi | sys | <i>Covariates</i> | Flycatcher | 0.928 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Flycatcher | 0.843 |
| 2Ni | Bi | vis | <i>Covariates</i> | Flycatcher | n/a |
| 2Ni | Bi | vis | <i>Interspecies</i> | Flycatcher | n/a |
| 2Ni | Bi | all | <i>Covariates</i> | Gray Hawk | 0.759 |
| 2Ni | Bi | all | <i>Interspecies</i> | Gray Hawk | 0.738 |
| 2Ni | Bi | all | <i>Plot</i> | Gray Hawk | 0.510 |
| 2Ni | Bi | aur | <i>Covariates</i> | Gray Hawk | 0.805 |
| 2Ni | Bi | aur | <i>Interspecies</i> | Gray Hawk | 0.748 |
| 2Ni | Bi | ran | <i>Covariates</i> | Gray Hawk | 0.232 |
| 2Ni | Bi | ran | <i>Interspecies</i> | Gray Hawk | 0.152 |
| 2Ni | Bi | sys | <i>Covariates</i> | Gray Hawk | 0.767 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Gray Hawk | 0.793 |
| 2Ni | Bi | vis | <i>Covariates</i> | Gray Hawk | 0.682 |
| 2Ni | Bi | vis | <i>Interspecies</i> | Gray Hawk | 0.712 |
| 2Ni | Bi | all | <i>Covariates</i> | Hawk | 0.506 |
| 2Ni | Bi | all | <i>Interspecies</i> | Hawk | 0.546 |
| 2Ni | Bi | all | <i>Plot</i> | Hawk | 0.728 |
| 2Ni | Bi | aur | <i>Covariates</i> | Hawk | 0.63 |
| 2Ni | Bi | aur | <i>Interspecies</i> | Hawk | 0.628 |
| 2Ni | Bi | ran | <i>Covariates</i> | Hawk | 0.048 |
| 2Ni | Bi | ran | <i>Interspecies</i> | Hawk | 0.036 |
| 2Ni | Bi | sys | <i>Covariates</i> | Hawk | 0.580 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Hawk | 0.57 |
| 2Ni | Bi | vis | <i>Covariates</i> | Hawk | 0.044 |
| 2Ni | Bi | vis | <i>Interspecies</i> | Hawk | 0.039 |
| 2Ni | Bi | all | <i>Covariates</i> | Parakeet | 0.480 |
| 2Ni | Bi | all | <i>Interspecies</i> | Parakeet | 0.283 |
| 2Ni | Bi | all | <i>Plot</i> | Parakeet | 0.290 |
| 2Ni | Bi | aur | <i>Covariates</i> | Parakeet | n/a |
| 2Ni | Bi | aur | <i>Interspecies</i> | Parakeet | n/a |
| 2Ni | Bi | ran | <i>Covariates</i> | Parakeet | 0.119 |
| 2Ni | Bi | ran | <i>Interspecies</i> | Parakeet | 0.036 |
| 2Ni | Bi | sys | <i>Covariates</i> | Parakeet | 0.176 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Parakeet | 0.042 |
| 2Ni | Bi | vis | <i>Covariates</i> | Parakeet | 0.274 |
| 2Ni | Bi | vis | <i>Interspecies</i> | Parakeet | 0.229 |
| 2Ni | Bi | all | <i>Covariates</i> | Parrot | 0.910 |
| 2Ni | Bi | all | <i>Interspecies</i> | Parrot | 0.91 |
| 2Ni | Bi | all | <i>Plot</i> | Parrot | 0.624 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|---------------------------|-----------|
| 2Ni | Bi | aur | <i>Covariates</i> | Parrot | 0.042 |
| 2Ni | Bi | aur | <i>Interspecies</i> | Parrot | 0.038 |
| 2Ni | Bi | ran | <i>Covariates</i> | Parrot | 0.762 |
| 2Ni | Bi | ran | <i>Interspecies</i> | Parrot | 0.811 |
| 2Ni | Bi | sys | <i>Covariates</i> | Parrot | 0.867 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Parrot | 0.869 |
| 2Ni | Bi | vis | <i>Covariates</i> | Parrot | 0.897 |
| 2Ni | Bi | vis | <i>Interspecies</i> | Parrot | 0.898 |
| 2Ni | Bi | all | <i>Covariates</i> | Parrot, large | 0.592 |
| 2Ni | Bi | all | <i>Interspecies</i> | Parrot, large | 0.562 |
| 2Ni | Bi | all | <i>Plot</i> | Parrot, large | 0.243 |
| 2Ni | Bi | aur | <i>Covariates</i> | Parrot, large | 0.042 |
| 2Ni | Bi | aur | <i>Interspecies</i> | Parrot, large | 0.038 |
| 2Ni | Bi | ran | <i>Covariates</i> | Parrot, large | 0.036 |
| 2Ni | Bi | ran | <i>Interspecies</i> | Parrot, large | 0.036 |
| 2Ni | Bi | sys | <i>Covariates</i> | Parrot, large | 0.624 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Parrot, large | 0.516 |
| 2Ni | Bi | vis | <i>Covariates</i> | Parrot, large | 0.378 |
| 2Ni | Bi | vis | <i>Interspecies</i> | Parrot, large | 0.332 |
| 2Ni | Bi | all | <i>Covariates</i> | Swallow | 0.283 |
| 2Ni | Bi | all | <i>Interspecies</i> | Swallow | 0.201 |
| 2Ni | Bi | all | <i>Plot</i> | Swallow | 0.424 |
| 2Ni | Bi | aur | <i>Covariates</i> | Swallow | 0.038 |
| 2Ni | Bi | aur | <i>Interspecies</i> | Swallow | 0.038 |
| 2Ni | Bi | ran | <i>Covariates</i> | Swallow | 0.036 |
| 2Ni | Bi | ran | <i>Interspecies</i> | Swallow | 0.036 |
| 2Ni | Bi | sys | <i>Covariates</i> | Swallow | 0.475 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Swallow | 0.42 |
| 2Ni | Bi | vis | <i>Covariates</i> | Swallow | 0.357 |
| 2Ni | Bi | vis | <i>Interspecies</i> | Swallow | 0.183 |
| 2Ni | Bi | all | <i>Covariates</i> | Turkey Vulture | 0.750 |
| 2Ni | Bi | all | <i>Interspecies</i> | Turkey Vulture | 0.67 |
| 2Ni | Bi | all | <i>Plot</i> | Turkey Vulture | 0.498 |
| 2Ni | Bi | aur | <i>Covariates</i> | Turkey Vulture | n/a |
| 2Ni | Bi | aur | <i>Interspecies</i> | Turkey Vulture | n/a |
| 2Ni | Bi | ran | <i>Covariates</i> | Turkey Vulture | 0.679 |
| 2Ni | Bi | ran | <i>Interspecies</i> | Turkey Vulture | 0.679 |
| 2Ni | Bi | sys | <i>Covariates</i> | Turkey Vulture | 0.640 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Turkey Vulture | 0.544 |
| 2Ni | Bi | vis | <i>Covariates</i> | Turkey Vulture | 0.576 |
| 2Ni | Bi | vis | <i>Interspecies</i> | Turkey Vulture | 0.533 |
| 2Ni | Bi | all | <i>Covariates</i> | White-throated Magpie Jay | 0.628 |
| 2Ni | Bi | all | <i>Interspecies</i> | White-throated Magpie Jay | 0.586 |
| 2Ni | Bi | all | <i>Plot</i> | White-throated Magpie Jay | 0.601 |
| 2Ni | Bi | aur | <i>Covariates</i> | White-throated Magpie Jay | 0.734 |
| 2Ni | Bi | aur | <i>Interspecies</i> | White-throated Magpie Jay | 0.705 |
| 2Ni | Bi | ran | <i>Covariates</i> | White-throated Magpie Jay | 0.573 |
| 2Ni | Bi | ran | <i>Interspecies</i> | White-throated Magpie Jay | 0.514 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|---------------------------|-----------|
| | | | | Jay | |
| 2Ni | Bi | sys | <i>Covariates</i> | White-throated Magpie Jay | 0.598 |
| 2Ni | Bi | sys | <i>Interspecies</i> | White-throated Magpie Jay | 0.574 |
| 2Ni | Bi | vis | <i>Covariates</i> | White-throated Magpie Jay | 0.587 |
| 2Ni | Bi | vis | <i>Interspecies</i> | White-throated Magpie Jay | 0.565 |
| 2Ni | Bi | all | <i>Covariates</i> | Woodpecker | 0.651 |
| 2Ni | Bi | all | <i>Interspecies</i> | Woodpecker | 0.61 |
| 2Ni | Bi | all | <i>Plot</i> | Woodpecker | 0.569 |
| 2Ni | Bi | aur | <i>Covariates</i> | Woodpecker | 0.59 |
| 2Ni | Bi | aur | <i>Interspecies</i> | Woodpecker | 0.511 |
| 2Ni | Bi | ran | <i>Covariates</i> | Woodpecker | 0.104 |
| 2Ni | Bi | ran | <i>Interspecies</i> | Woodpecker | 0.146 |
| 2Ni | Bi | sys | <i>Covariates</i> | Woodpecker | 0.667 |
| 2Ni | Bi | sys | <i>Interspecies</i> | Woodpecker | 0.602 |
| 2Ni | Bi | vis | <i>Covariates</i> | Woodpecker | 0.502 |
| 2Ni | Bi | vis | <i>Interspecies</i> | Woodpecker | 0.479 |
| 3AK | Bi | all | <i>Covariates</i> | Chickadee | 0.909 |
| 3AK | Bi | all | <i>Interspecies</i> | Chickadee | 0.912 |
| 3AK | Bi | all | <i>Plot</i> | Chickadee | 0.769 |
| 3AK | Bi | aur | <i>Covariates</i> | Chickadee | 0.933 |
| 3AK | Bi | aur | <i>Interspecies</i> | Chickadee | 0.933 |
| 3AK | Bi | ran | <i>Covariates</i> | Chickadee | n/a |
| 3AK | Bi | ran | <i>Interspecies</i> | Chickadee | n/a |
| 3AK | Bi | sys | <i>Covariates</i> | Chickadee | 0.911 |
| 3AK | Bi | sys | <i>Interspecies</i> | Chickadee | 0.888 |
| 3AK | Bi | vis | <i>Covariates</i> | Chickadee | n/a |
| 3AK | Bi | vis | <i>Interspecies</i> | Chickadee | n/a |
| 3AK | Bi | all | <i>Covariates</i> | Gull | 0.570 |
| 3AK | Bi | all | <i>Interspecies</i> | Gull | 0.574 |
| 3AK | Bi | all | <i>Plot</i> | Gull | 0.019 |
| 3AK | Bi | aur | <i>Covariates</i> | Gull | n/a |
| 3AK | Bi | aur | <i>Interspecies</i> | Gull | n/a |
| 3AK | Bi | ran | <i>Covariates</i> | Gull | n/a |
| 3AK | Bi | ran | <i>Interspecies</i> | Gull | n/a |
| 3AK | Bi | sys | <i>Covariates</i> | Gull | 0.531 |
| 3AK | Bi | sys | <i>Interspecies</i> | Gull | 0.327 |
| 3AK | Bi | vis | <i>Covariates</i> | Gull | 0.279 |
| 3AK | Bi | vis | <i>Interspecies</i> | Gull | 0.214 |
| 3AK | Bi | all | <i>Covariates</i> | Sparrow | 0.642 |
| 3AK | Bi | all | <i>Interspecies</i> | Sparrow | 0.649 |
| 3AK | Bi | all | <i>Plot</i> | Sparrow | 0.430 |
| 3AK | Bi | aur | <i>Covariates</i> | Sparrow | 0.528 |
| 3AK | Bi | aur | <i>Interspecies</i> | Sparrow | 0.509 |
| 3AK | Bi | ran | <i>Covariates</i> | Sparrow | 0.623 |
| 3AK | Bi | ran | <i>Interspecies</i> | Sparrow | 0.647 |
| 3AK | Bi | sys | <i>Covariates</i> | Sparrow | 0.632 |
| 3AK | Bi | sys | <i>Interspecies</i> | Sparrow | 0.624 |
| 3AK | Bi | vis | <i>Covariates</i> | Sparrow | 0.832 |
| 3AK | Bi | vis | <i>Interspecies</i> | Sparrow | 0.822 |
| 3AK | Bi | all | <i>Covariates</i> | squirrel | 0.555 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------|-----------|
| 3AK | Bi | all | <i>Interspecies</i> | squirrel | 0.564 |
| 3AK | Bi | all | <i>Plot</i> | Squirrel | 0.855 |
| 3AK | Bi | aur | <i>Covariates</i> | squirrel | 0.467 |
| 3AK | Bi | aur | <i>Interspecies</i> | squirrel | 0.478 |
| 3AK | Bi | ran | <i>Covariates</i> | squirrel | 0.401 |
| 3AK | Bi | ran | <i>Interspecies</i> | squirrel | 0.412 |
| 3AK | Bi | sys | <i>Covariates</i> | squirrel | 0.566 |
| 3AK | Bi | sys | <i>Interspecies</i> | squirrel | 0.563 |
| 3AK | Bi | vis | <i>Covariates</i> | squirrel | 0.917 |
| 3AK | Bi | vis | <i>Interspecies</i> | squirrel | 0.917 |
| 4Ru | Bi | all | <i>Covariates</i> | Chickadee | 0.629 |
| 4Ru | Bi | all | <i>Interspecies</i> | Chickadee | 0.647 |
| 4Ru | Bi | all | <i>Plot</i> | Chickadee | 0.649 |
| 4Ru | Bi | aur | <i>Covariates</i> | Chickadee | 0.623 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Chickadee | 0.64 |
| 4Ru | Bi | ran | <i>Covariates</i> | Chickadee | 0.528 |
| 4Ru | Bi | ran | <i>Interspecies</i> | Chickadee | 0.528 |
| 4Ru | Bi | sys | <i>Covariates</i> | Chickadee | 0.658 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Chickadee | 0.66 |
| 4Ru | Bi | vis | <i>Covariates</i> | Chickadee | 0.587 |
| 4Ru | Bi | vis | <i>Interspecies</i> | Chickadee | 0.601 |
| 4Ru | Bi | all | <i>Covariates</i> | Jungle Crow | 0.665 |
| 4Ru | Bi | all | <i>Interspecies</i> | Jungle Crow | 0.689 |
| 4Ru | Bi | all | <i>Plot</i> | Jungle Crow | 0.402 |
| 4Ru | Bi | aur | <i>Covariates</i> | Jungle Crow | 0.799 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Jungle Crow | 0.765 |
| 4Ru | Bi | ran | <i>Covariates</i> | Jungle Crow | 0.862 |
| 4Ru | Bi | ran | <i>Interspecies</i> | Jungle Crow | 0.862 |
| 4Ru | Bi | sys | <i>Covariates</i> | Jungle Crow | 0.574 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Jungle Crow | 0.634 |
| 4Ru | Bi | vis | <i>Covariates</i> | Jungle Crow | 0.119 |
| 4Ru | Bi | vis | <i>Interspecies</i> | Jungle Crow | 0.034 |
| 4Ru | Bi | all | <i>Covariates</i> | Kinglet | 0.625 |
| 4Ru | Bi | all | <i>Interspecies</i> | Kinglet | 0.613 |
| 4Ru | Bi | all | <i>Plot</i> | Kinglet | 0.034 |
| 4Ru | Bi | aur | <i>Covariates</i> | Kinglet | 0.638 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Kinglet | 0.637 |
| 4Ru | Bi | ran | <i>Covariates</i> | Kinglet | 0.583 |
| 4Ru | Bi | ran | <i>Interspecies</i> | Kinglet | 0.552 |
| 4Ru | Bi | sys | <i>Covariates</i> | Kinglet | 0.576 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Kinglet | 0.588 |
| 4Ru | Bi | vis | <i>Covariates</i> | Kinglet | 0.626 |
| 4Ru | Bi | vis | <i>Interspecies</i> | Kinglet | 0.629 |
| 4Ru | Bi | all | <i>Covariates</i> | Nutcracker | 0.685 |
| 4Ru | Bi | all | <i>Interspecies</i> | Nutcracker | 0.685 |
| 4Ru | Bi | all | <i>Plot</i> | Nutcracker | 0.478 |
| 4Ru | Bi | aur | <i>Covariates</i> | Nutcracker | 0.622 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Nutcracker | 0.643 |
| 4Ru | Bi | ran | <i>Covariates</i> | Nutcracker | 0.862 |
| 4Ru | Bi | ran | <i>Interspecies</i> | Nutcracker | 0.862 |
| 4Ru | Bi | sys | <i>Covariates</i> | Nutcracker | 0.644 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Nutcracker | 0.659 |
| 4Ru | Bi | vis | <i>Covariates</i> | Nutcracker | 0.742 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|---------------------|-----------|
| 4Ru | Bi | vis | <i>Interspecies</i> | Nutcracker | 0.663 |
| 4Ru | Bi | all | <i>Covariates</i> | Oriental Dove | 0.494 |
| 4Ru | Bi | all | <i>Interspecies</i> | Oriental Dove | 0.494 |
| 4Ru | Bi | all | <i>Plot</i> | Oriental Dove | 0.370 |
| 4Ru | Bi | aur | <i>Covariates</i> | Oriental Dove | 0.5 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Oriental Dove | 0.262 |
| 4Ru | Bi | ran | <i>Covariates</i> | Oriental Dove | n/a |
| 4Ru | Bi | ran | <i>Interspecies</i> | Oriental Dove | n/a |
| 4Ru | Bi | sys | <i>Covariates</i> | Oriental Dove | 0.538 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Oriental Dove | 0.538 |
| 4Ru | Bi | vis | <i>Covariates</i> | Oriental Dove | 0.039 |
| 4Ru | Bi | vis | <i>Interspecies</i> | Oriental Dove | 0.051 |
| 4Ru | Bi | all | <i>Covariates</i> | Oriental Finch | 0.701 |
| 4Ru | Bi | all | <i>Interspecies</i> | Oriental Finch | 0.801 |
| 4Ru | Bi | all | <i>Plot</i> | Oriental Finch | 0.226 |
| 4Ru | Bi | aur | <i>Covariates</i> | Oriental Finch | 0.291 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Oriental Finch | 0.191 |
| 4Ru | Bi | ran | <i>Covariates</i> | Oriental Finch | 0.862 |
| 4Ru | Bi | ran | <i>Interspecies</i> | Oriental Finch | 0.862 |
| 4Ru | Bi | sys | <i>Covariates</i> | Oriental Finch | 0.756 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Oriental Finch | 0.572 |
| 4Ru | Bi | vis | <i>Covariates</i> | Oriental Finch | 0.14 |
| 4Ru | Bi | vis | <i>Interspecies</i> | Oriental Finch | 0.201 |
| 4Ru | Bi | all | <i>Covariates</i> | Oriental Greenfinch | 0.487 |
| 4Ru | Bi | all | <i>Interspecies</i> | Oriental Greenfinch | 0.475 |
| 4Ru | Bi | all | <i>Plot</i> | Oriental Greenfinch | 0.087 |
| 4Ru | Bi | aur | <i>Covariates</i> | Oriental Greenfinch | 0.417 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Oriental Greenfinch | 0.427 |
| 4Ru | Bi | ran | <i>Covariates</i> | Oriental Greenfinch | n/a |
| 4Ru | Bi | ran | <i>Interspecies</i> | Oriental Greenfinch | n/a |
| 4Ru | Bi | sys | <i>Covariates</i> | Oriental Greenfinch | 0.473 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Oriental Greenfinch | 0.493 |
| 4Ru | Bi | vis | <i>Covariates</i> | Oriental Greenfinch | n/a |
| 4Ru | Bi | vis | <i>Interspecies</i> | Oriental Greenfinch | n/a |
| 4Ru | Bi | all | <i>Covariates</i> | Pacific Swift | 0.734 |
| 4Ru | Bi | all | <i>Interspecies</i> | Pacific Swift | 0.72 |
| 4Ru | Bi | all | <i>Plot</i> | Pacific Swift | 0.194 |
| 4Ru | Bi | aur | <i>Covariates</i> | Pacific Swift | 0.78 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Pacific Swift | 0.756 |
| 4Ru | Bi | ran | <i>Covariates</i> | Pacific Swift | 0.633 |
| 4Ru | Bi | ran | <i>Interspecies</i> | Pacific Swift | 0.607 |
| 4Ru | Bi | sys | <i>Covariates</i> | Pacific Swift | 0.714 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Pacific Swift | 0.696 |
| 4Ru | Bi | vis | <i>Covariates</i> | Pacific Swift | 0.577 |
| 4Ru | Bi | vis | <i>Interspecies</i> | Pacific Swift | 0.58 |
| 4Ru | Bi | all | <i>Covariates</i> | Raptor | 0.802 |
| 4Ru | Bi | all | <i>Interspecies</i> | Raptor | 0.821 |
| 4Ru | Bi | all | <i>Plot</i> | Raptor | 0.500 |
| 4Ru | Bi | aur | <i>Covariates</i> | Raptor | 0.771 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Raptor | 0.771 |
| 4Ru | Bi | ran | <i>Covariates</i> | Raptor | n/a |
| 4Ru | Bi | ran | <i>Interspecies</i> | Raptor | n/a |
| 4Ru | Bi | sys | <i>Covariates</i> | Raptor | 0.746 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------|-----------|
| 4Ru | Bi | sys | <i>Interspecies</i> | Raptor | 0.768 |
| 4Ru | Bi | vis | <i>Covariates</i> | Raptor | n/a |
| 4Ru | Bi | vis | <i>Interspecies</i> | Raptor | n/a |
| 4Ru | Bi | all | <i>Covariates</i> | Tsilp | 0.547 |
| 4Ru | Bi | all | <i>Interspecies</i> | Tsilp | 0.583 |
| 4Ru | Bi | all | <i>Plot</i> | Tsilp | 0.431 |
| 4Ru | Bi | aur | <i>Covariates</i> | Tsilp | 0.598 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Tsilp | 0.521 |
| 4Ru | Bi | ran | <i>Covariates</i> | Tsilp | n/a |
| 4Ru | Bi | ran | <i>Interspecies</i> | Tsilp | n/a |
| 4Ru | Bi | sys | <i>Covariates</i> | Tsilp | 0.490 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Tsilp | 0.501 |
| 4Ru | Bi | vis | <i>Covariates</i> | Tsilp | n/a |
| 4Ru | Bi | vis | <i>Interspecies</i> | Tsilp | n/a |
| 4Ru | Bi | all | <i>Covariates</i> | Warbler | 0.631 |
| 4Ru | Bi | all | <i>Interspecies</i> | Warbler | 0.634 |
| 4Ru | Bi | all | <i>Plot</i> | Warbler | 0.548 |
| 4Ru | Bi | aur | <i>Covariates</i> | Warbler | 0.658 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Warbler | 0.651 |
| 4Ru | Bi | ran | <i>Covariates</i> | Warbler | n/a |
| 4Ru | Bi | ran | <i>Interspecies</i> | Warbler | n/a |
| 4Ru | Bi | sys | <i>Covariates</i> | Warbler | 0.576 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Warbler | 0.569 |
| 4Ru | Bi | vis | <i>Covariates</i> | Warbler | 0.238 |
| 4Ru | Bi | vis | <i>Interspecies</i> | Warbler | 0.115 |
| 4Ru | Bi | all | <i>Covariates</i> | Winter Wren | 0.691 |
| 4Ru | Bi | all | <i>Interspecies</i> | Winter Wren | 0.68 |
| 4Ru | Bi | all | <i>Plot</i> | Winter Wren | 0.576 |
| 4Ru | Bi | aur | <i>Covariates</i> | Winter Wren | 0.665 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Winter Wren | 0.667 |
| 4Ru | Bi | ran | <i>Covariates</i> | Winter Wren | n/a |
| 4Ru | Bi | ran | <i>Interspecies</i> | Winter Wren | n/a |
| 4Ru | Bi | sys | <i>Covariates</i> | Winter Wren | 0.667 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Winter Wren | 0.666 |
| 4Ru | Bi | vis | <i>Covariates</i> | Winter Wren | n/a |
| 4Ru | Bi | vis | <i>Interspecies</i> | Winter Wren | n/a |
| 4Ru | Bi | all | <i>Covariates</i> | wize | 0.650 |
| 4Ru | Bi | all | <i>Interspecies</i> | wize | 0.633 |
| 4Ru | Bi | all | <i>Plot</i> | wize | 0.438 |
| 4Ru | Bi | aur | <i>Covariates</i> | wize | 0.659 |
| 4Ru | Bi | aur | <i>Interspecies</i> | wize | 0.653 |
| 4Ru | Bi | ran | <i>Covariates</i> | wize | 0.515 |
| 4Ru | Bi | ran | <i>Interspecies</i> | wize | 0.515 |
| 4Ru | Bi | sys | <i>Covariates</i> | wize | 0.622 |
| 4Ru | Bi | sys | <i>Interspecies</i> | wize | 0.614 |
| 4Ru | Bi | vis | <i>Covariates</i> | wize | 0.216 |
| 4Ru | Bi | vis | <i>Interspecies</i> | wize | 0.205 |
| 4Ru | Bi | all | <i>Covariates</i> | wize wize | 0.797 |
| 4Ru | Bi | all | <i>Interspecies</i> | wize wize | 0.797 |
| 4Ru | Bi | all | <i>Plot</i> | wize wize | 0.000 |
| 4Ru | Bi | aur | <i>Covariates</i> | wize wize | 0.803 |
| 4Ru | Bi | aur | <i>Interspecies</i> | wize wize | 0.803 |
| 4Ru | Bi | ran | <i>Covariates</i> | wize wize | n/a |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------------|-----------|
| 4Ru | Bi | ran | <i>Interspecies</i> | wize wize | n/a |
| 4Ru | Bi | sys | <i>Covariates</i> | wize wize | 0.799 |
| 4Ru | Bi | sys | <i>Interspecies</i> | wize wize | 0.781 |
| 4Ru | Bi | vis | <i>Covariates</i> | wize wize | n/a |
| 4Ru | Bi | vis | <i>Interspecies</i> | wize wize | n/a |
| 4Ru | Bi | all | <i>Covariates</i> | Woodpecker | 0.572 |
| 4Ru | Bi | all | <i>Interspecies</i> | Woodpecker | 0.559 |
| 4Ru | Bi | all | <i>Plot</i> | Woodpecker | 0.566 |
| 4Ru | Bi | aur | <i>Covariates</i> | Woodpecker | 0.531 |
| 4Ru | Bi | aur | <i>Interspecies</i> | Woodpecker | 0.519 |
| 4Ru | Bi | ran | <i>Covariates</i> | Woodpecker | 0.034 |
| 4Ru | Bi | ran | <i>Interspecies</i> | Woodpecker | 0.034 |
| 4Ru | Bi | sys | <i>Covariates</i> | Woodpecker | 0.519 |
| 4Ru | Bi | sys | <i>Interspecies</i> | Woodpecker | 0.52 |
| 4Ru | Bi | vis | <i>Covariates</i> | Woodpecker | n/a |
| 4Ru | Bi | vis | <i>Interspecies</i> | Woodpecker | n/a |
| 5PG | Bi | all | <i>Covariates</i> | Balu | 0.070 |
| 5PG | Bi | all | <i>Interspecies</i> | Balu | 0.04 |
| 5PG | Bi | aur | <i>Covariates</i> | Balu | 0.073 |
| 5PG | Bi | aur | <i>Interspecies</i> | Balu | 0.042 |
| 5PG | Bi | ran | <i>Covariates</i> | Balu | n/a |
| 5PG | Bi | ran | <i>Interspecies</i> | Balu | n/a |
| 5PG | Bi | sys | <i>Covariates</i> | Balu | 0.072 |
| 5PG | Bi | sys | <i>Interspecies</i> | Balu | 0.042 |
| 5PG | Bi | vis | <i>Covariates</i> | Balu | n/a |
| 5PG | Bi | vis | <i>Interspecies</i> | Balu | n/a |
| 5PG | Bi | all | <i>Covariates</i> | call | 0.583 |
| 5PG | Bi | all | <i>Interspecies</i> | call | 0.446 |
| 5PG | Bi | aur | <i>Covariates</i> | call | 0.473 |
| 5PG | Bi | aur | <i>Interspecies</i> | call | 0.393 |
| 5PG | Bi | ran | <i>Covariates</i> | call | n/a |
| 5PG | Bi | ran | <i>Interspecies</i> | call | n/a |
| 5PG | Bi | sys | <i>Covariates</i> | call | 0.496 |
| 5PG | Bi | sys | <i>Interspecies</i> | call | 0.272 |
| 5PG | Bi | vis | <i>Covariates</i> | call | n/a |
| 5PG | Bi | vis | <i>Interspecies</i> | call | n/a |
| 5PG | Bi | all | <i>Covariates</i> | Craw, Bird of Paradise | 0.507 |
| 5PG | Bi | all | <i>Interspecies</i> | Craw, Bird of Paradise | 0.318 |
| 5PG | Bi | aur | <i>Covariates</i> | Craw, Bird of Paradise | 0.367 |
| 5PG | Bi | aur | <i>Interspecies</i> | Craw, Bird of Paradise | 0.227 |
| 5PG | Bi | ran | <i>Covariates</i> | Craw, Bird of Paradise | n/a |
| 5PG | Bi | ran | <i>Interspecies</i> | Craw, Bird of Paradise | n/a |
| 5PG | Bi | sys | <i>Covariates</i> | Craw, Bird of Paradise | 0.621 |
| 5PG | Bi | sys | <i>Interspecies</i> | Craw, Bird of Paradise | 0.465 |
| 5PG | Bi | vis | <i>Covariates</i> | Craw, Bird of Paradise | n/a |
| 5PG | Bi | vis | <i>Interspecies</i> | Craw, Bird of Paradise | n/a |
| 5PG | Bi | all | <i>Covariates</i> | Flute | 0.578 |
| 5PG | Bi | all | <i>Interspecies</i> | Flute | 0.519 |
| 5PG | Bi | aur | <i>Covariates</i> | Flute | 0.451 |
| 5PG | Bi | aur | <i>Interspecies</i> | Flute | 0.458 |
| 5PG | Bi | ran | <i>Covariates</i> | Flute | 0.383 |
| 5PG | Bi | ran | <i>Interspecies</i> | Flute | 0.288 |
| 5PG | Bi | sys | <i>Covariates</i> | Flute | 0.488 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------|-----------|
| 5PG | Bi | sys | <i>Interspecies</i> | Flute | 0.499 |
| 5PG | Bi | vis | <i>Covariates</i> | Flute | n/a |
| 5PG | Bi | vis | <i>Interspecies</i> | Flute | n/a |
| 5PG | Bi | all | <i>Covariates</i> | Fowl | 0.654 |
| 5PG | Bi | all | <i>Interspecies</i> | Fowl | 0.6 |
| 5PG | Bi | aur | <i>Covariates</i> | Fowl | 0.607 |
| 5PG | Bi | aur | <i>Interspecies</i> | Fowl | 0.722 |
| 5PG | Bi | ran | <i>Covariates</i> | Fowl | 0.031 |
| 5PG | Bi | ran | <i>Interspecies</i> | Fowl | 0.031 |
| 5PG | Bi | sys | <i>Covariates</i> | Fowl | 0.691 |
| 5PG | Bi | sys | <i>Interspecies</i> | Fowl | 0.691 |
| 5PG | Bi | vis | <i>Covariates</i> | Fowl | 0.038 |
| 5PG | Bi | vis | <i>Interspecies</i> | Fowl | 0.038 |
| 5PG | Bi | all | <i>Covariates</i> | gleaner | 0.750 |
| 5PG | Bi | all | <i>Interspecies</i> | gleaner | 0.683 |
| 5PG | Bi | aur | <i>Covariates</i> | gleaner | n/a |
| 5PG | Bi | aur | <i>Interspecies</i> | gleaner | n/a |
| 5PG | Bi | ran | <i>Covariates</i> | gleaner | 0.031 |
| 5PG | Bi | ran | <i>Interspecies</i> | gleaner | 0.031 |
| 5PG | Bi | sys | <i>Covariates</i> | gleaner | 0.815 |
| 5PG | Bi | sys | <i>Interspecies</i> | gleaner | 0.731 |
| 5PG | Bi | vis | <i>Covariates</i> | gleaner | 0.174 |
| 5PG | Bi | vis | <i>Interspecies</i> | gleaner | 0.149 |
| 5PG | Bi | all | <i>Covariates</i> | Hawk | 0.322 |
| 5PG | Bi | all | <i>Interspecies</i> | Hawk | 0.268 |
| 5PG | Bi | aur | <i>Covariates</i> | Hawk | 0.571 |
| 5PG | Bi | aur | <i>Interspecies</i> | Hawk | 0.589 |
| 5PG | Bi | ran | <i>Covariates</i> | Hawk | 0.031 |
| 5PG | Bi | ran | <i>Interspecies</i> | Hawk | 0.031 |
| 5PG | Bi | sys | <i>Covariates</i> | Hawk | 0.350 |
| 5PG | Bi | sys | <i>Interspecies</i> | Hawk | 0.275 |
| 5PG | Bi | vis | <i>Covariates</i> | Hawk | 0.301 |
| 5PG | Bi | vis | <i>Interspecies</i> | Hawk | 0.117 |
| 5PG | Bi | all | <i>Covariates</i> | Hornbill | 0.291 |
| 5PG | Bi | all | <i>Interspecies</i> | Hornbill | 0.24 |
| 5PG | Bi | aur | <i>Covariates</i> | Hornbill | 0.109 |
| 5PG | Bi | aur | <i>Interspecies</i> | Hornbill | 0.04 |
| 5PG | Bi | ran | <i>Covariates</i> | Hornbill | 0.041 |
| 5PG | Bi | ran | <i>Interspecies</i> | Hornbill | 0.031 |
| 5PG | Bi | sys | <i>Covariates</i> | Hornbill | 0.390 |
| 5PG | Bi | sys | <i>Interspecies</i> | Hornbill | 0.294 |
| 5PG | Bi | vis | <i>Covariates</i> | Hornbill | 0.746 |
| 5PG | Bi | vis | <i>Interspecies</i> | Hornbill | 0.859 |
| 5PG | Bi | all | <i>Covariates</i> | melodious song | 0.308 |
| 5PG | Bi | all | <i>Interspecies</i> | melodious song | 0.193 |
| 5PG | Bi | aur | <i>Covariates</i> | melodious song | 0.315 |
| 5PG | Bi | aur | <i>Interspecies</i> | melodious song | 0.25 |
| 5PG | Bi | ran | <i>Covariates</i> | melodious song | 0.417 |
| 5PG | Bi | ran | <i>Interspecies</i> | melodious song | 0.156 |
| 5PG | Bi | sys | <i>Covariates</i> | melodious song | 0.250 |
| 5PG | Bi | sys | <i>Interspecies</i> | melodious song | 0.174 |
| 5PG | Bi | vis | <i>Covariates</i> | melodious song | n/a |
| 5PG | Bi | vis | <i>Interspecies</i> | melodious song | n/a |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------|-----------|
| 5PG | Bi | all | <i>Covariates</i> | Parakeet | 0.884 |
| 5PG | Bi | all | <i>Interspecies</i> | Parakeet | 0.842 |
| 5PG | Bi | all | <i>Plot</i> | Parakeet | 0.537 |
| 5PG | Bi | aur | <i>Covariates</i> | Parakeet | n/a |
| 5PG | Bi | aur | <i>Interspecies</i> | Parakeet | n/a |
| 5PG | Bi | ran | <i>Covariates</i> | Parakeet | 0.457 |
| 5PG | Bi | ran | <i>Interspecies</i> | Parakeet | 0.33 |
| 5PG | Bi | sys | <i>Covariates</i> | Parakeet | 0.932 |
| 5PG | Bi | sys | <i>Interspecies</i> | Parakeet | 0.823 |
| 5PG | Bi | vis | <i>Covariates</i> | Parakeet | 0.512 |
| 5PG | Bi | vis | <i>Interspecies</i> | Parakeet | 0.341 |
| 5PG | Bi | all | <i>Covariates</i> | Parrot | 0.320 |
| 5PG | Bi | all | <i>Interspecies</i> | Parrot | 0.257 |
| 5PG | Bi | all | <i>Plot</i> | Parrot | 0.000 |
| 5PG | Bi | aur | <i>Covariates</i> | Parrot | 0.044 |
| 5PG | Bi | aur | <i>Interspecies</i> | Parrot | 0.044 |
| 5PG | Bi | ran | <i>Covariates</i> | Parrot | 0.021 |
| 5PG | Bi | ran | <i>Interspecies</i> | Parrot | 0.021 |
| 5PG | Bi | sys | <i>Covariates</i> | Parrot | 0.072 |
| 5PG | Bi | sys | <i>Interspecies</i> | Parrot | 0.042 |
| 5PG | Bi | vis | <i>Covariates</i> | Parrot | 0.053 |
| 5PG | Bi | vis | <i>Interspecies</i> | Parrot | 0.024 |
| 5PG | Bi | all | <i>Covariates</i> | Rezina, rezina | 0.733 |
| 5PG | Bi | all | <i>Interspecies</i> | Rezina, rezina | 0.676 |
| 5PG | Bi | all | <i>Plot</i> | Rezina, rezina | 0.208 |
| 5PG | Bi | aur | <i>Covariates</i> | Rezina, rezina | 0.753 |
| 5PG | Bi | aur | <i>Interspecies</i> | Rezina, rezina | 0.754 |
| 5PG | Bi | ran | <i>Covariates</i> | Rezina, rezina | 0.687 |
| 5PG | Bi | ran | <i>Interspecies</i> | Rezina, rezina | 0.7 |
| 5PG | Bi | sys | <i>Covariates</i> | Rezina, rezina | 0.751 |
| 5PG | Bi | sys | <i>Interspecies</i> | Rezina, rezina | 0.701 |
| 5PG | Bi | vis | <i>Covariates</i> | Rezina, rezina | 0.843 |
| 5PG | Bi | vis | <i>Interspecies</i> | Rezina, rezina | 0.8 |
| 5PG | Bi | all | <i>Covariates</i> | Swallow | 0.795 |
| 5PG | Bi | all | <i>Interspecies</i> | Swallow | 0.839 |
| 5PG | Bi | all | <i>Plot</i> | Swallow | 0.031 |
| 5PG | Bi | aur | <i>Covariates</i> | Swallow | 0.045 |
| 5PG | Bi | aur | <i>Interspecies</i> | Swallow | 0.045 |
| 5PG | Bi | ran | <i>Covariates</i> | Swallow | 0.865 |
| 5PG | Bi | ran | <i>Interspecies</i> | Swallow | 0.865 |
| 5PG | Bi | sys | <i>Covariates</i> | Swallow | 0.804 |
| 5PG | Bi | sys | <i>Interspecies</i> | Swallow | 0.841 |
| 5PG | Bi | vis | <i>Covariates</i> | Swallow | 0.496 |
| 5PG | Bi | vis | <i>Interspecies</i> | Swallow | 0.496 |
| 5PG | Bi | all | <i>Covariates</i> | swirrl | 0.554 |
| 5PG | Bi | all | <i>Interspecies</i> | swirrl | 0.535 |
| 5PG | Bi | all | <i>Plot</i> | swirrl | 0.056 |
| 5PG | Bi | aur | <i>Covariates</i> | swirrl | 0.556 |
| 5PG | Bi | aur | <i>Interspecies</i> | swirrl | 0.502 |
| 5PG | Bi | ran | <i>Covariates</i> | swirrl | n/a |
| 5PG | Bi | ran | <i>Interspecies</i> | swirrl | n/a |
| 5PG | Bi | sys | <i>Covariates</i> | swirrl | 0.613 |
| 5PG | Bi | sys | <i>Interspecies</i> | swirrl | 0.539 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------|-----------|
| 5PG | Bi | vis | <i>Covariates</i> | swirrl | n/a |
| 5PG | Bi | vis | <i>Interspecies</i> | swirrl | n/a |
| 5PG | Bi | all | <i>Covariates</i> | tsilp | 0.686 |
| 5PG | Bi | all | <i>Interspecies</i> | tsilp | 0.654 |
| 5PG | Bi | all | <i>Plot</i> | tsilp | 0.196 |
| 5PG | Bi | aur | <i>Covariates</i> | tsilp | 0.627 |
| 5PG | Bi | aur | <i>Interspecies</i> | tsilp | 0.627 |
| 5PG | Bi | ran | <i>Covariates</i> | tsilp | 0.667 |
| 5PG | Bi | ran | <i>Interspecies</i> | tsilp | 0.599 |
| 5PG | Bi | sys | <i>Covariates</i> | tsilp | 0.661 |
| 5PG | Bi | sys | <i>Interspecies</i> | tsilp | 0.624 |
| 5PG | Bi | vis | <i>Covariates</i> | tsilp | n/a |
| 5PG | Bi | vis | <i>Interspecies</i> | tsilp | n/a |
| 5PG | Bi | all | <i>Covariates</i> | White Cockatoo | 0.825 |
| 5PG | Bi | all | <i>Interspecies</i> | White Cockatoo | 0.777 |
| 5PG | Bi | all | <i>Plot</i> | White Cockatoo | 0.466 |
| 5PG | Bi | aur | <i>Covariates</i> | White Cockatoo | 0.674 |
| 5PG | Bi | aur | <i>Interspecies</i> | White Cockatoo | 0.516 |
| 5PG | Bi | ran | <i>Covariates</i> | White Cockatoo | n/a |
| 5PG | Bi | ran | <i>Interspecies</i> | White Cockatoo | n/a |
| 5PG | Bi | sys | <i>Covariates</i> | White Cockatoo | 0.830 |
| 5PG | Bi | sys | <i>Interspecies</i> | White Cockatoo | 0.75 |
| 5PG | Bi | vis | <i>Covariates</i> | White Cockatoo | 0.721 |
| 5PG | Bi | vis | <i>Interspecies</i> | White Cockatoo | 0.543 |
| 5PG | Bi | all | <i>Covariates</i> | wiz wiz | 0.591 |
| 5PG | Bi | all | <i>Interspecies</i> | wiz wiz | 0.433 |
| 5PG | Bi | all | <i>Plot</i> | wiz wiz | 0.635 |
| 5PG | Bi | aur | <i>Covariates</i> | wiz wiz | 0.66 |
| 5PG | Bi | aur | <i>Interspecies</i> | wiz wiz | 0.574 |
| 5PG | Bi | ran | <i>Covariates</i> | wiz wiz | n/a |
| 5PG | Bi | ran | <i>Interspecies</i> | wiz wiz | n/a |
| 5PG | Bi | sys | <i>Covariates</i> | wiz wiz | 0.557 |
| 5PG | Bi | sys | <i>Interspecies</i> | wiz wiz | 0.616 |
| 5PG | Bi | vis | <i>Covariates</i> | wiz wiz | n/a |
| 5PG | Bi | vis | <i>Interspecies</i> | wiz wiz | n/a |
| 5PG | Bi | all | <i>Covariates</i> | Wize wize | 0.274 |
| 5PG | Bi | all | <i>Interspecies</i> | Wize wize | 0.241 |
| 5PG | Bi | all | <i>Plot</i> | Wize wize | 0.448 |
| 5PG | Bi | aur | <i>Covariates</i> | Wize wize | 0.428 |
| 5PG | Bi | aur | <i>Interspecies</i> | Wize wize | 0.433 |
| 5PG | Bi | ran | <i>Covariates</i> | Wize wize | n/a |
| 5PG | Bi | ran | <i>Interspecies</i> | Wize wize | n/a |
| 5PG | Bi | sys | <i>Covariates</i> | Wize wize | 0.228 |
| 5PG | Bi | sys | <i>Interspecies</i> | Wize wize | 0.097 |
| 5PG | Bi | vis | <i>Covariates</i> | Wize wize | 0.038 |
| 5PG | Bi | vis | <i>Interspecies</i> | Wize wize | 0.038 |
| 5PG | Bi | all | <i>Covariates</i> | woodpecker | 0.372 |
| 5PG | Bi | all | <i>Interspecies</i> | woodpecker | 0.295 |
| 5PG | Bi | all | <i>Plot</i> | woodpecker | 0.308 |
| 5PG | Bi | aur | <i>Covariates</i> | woodpecker | 0.134 |
| 5PG | Bi | aur | <i>Interspecies</i> | woodpecker | 0.073 |
| 5PG | Bi | ran | <i>Covariates</i> | woodpecker | 0.031 |
| 5PG | Bi | ran | <i>Interspecies</i> | woodpecker | 0.031 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|----------------------|-----------|
| 5PG | Bi | sys | <i>Covariates</i> | woodpecker | 0.370 |
| 5PG | Bi | sys | <i>Interspecies</i> | woodpecker | 0.248 |
| 5PG | Bi | vis | <i>Covariates</i> | woodpecker | 0.242 |
| 5PG | Bi | vis | <i>Interspecies</i> | woodpecker | 0.156 |
| 6Ba | Bi | all | <i>Covariates</i> | Dunlin | 0.547 |
| 6Ba | Bi | all | <i>Interspecies</i> | Dunlin | 0.541 |
| 6Ba | Bi | all | <i>Plot</i> | Dunlin | 0.597 |
| 6Ba | Bi | ran | <i>Covariates</i> | Dunlin | 0.256 |
| 6Ba | Bi | ran | <i>Interspecies</i> | Dunlin | 0.139 |
| 6Ba | Bi | sys | <i>Covariates</i> | Dunlin | 0.602 |
| 6Ba | Bi | sys | <i>Interspecies</i> | Dunlin | 0.602 |
| 6Ba | Bi | all | <i>Covariates</i> | Lapland Bunting | 0.523 |
| 6Ba | Bi | all | <i>Interspecies</i> | Lapland Bunting | 0.521 |
| 6Ba | Bi | all | <i>Plot</i> | Lapland Bunting | 0.428 |
| 6Ba | Bi | ran | <i>Covariates</i> | Lapland Bunting | 0.666 |
| 6Ba | Bi | ran | <i>Interspecies</i> | Lapland Bunting | 0.627 |
| 6Ba | Bi | sys | <i>Covariates</i> | Lapland Bunting | 0.468 |
| 6Ba | Bi | sys | <i>Interspecies</i> | Lapland Bunting | 0.46 |
| 6Ba | Bi | all | <i>Covariates</i> | Longbilled Dowitcher | 0.561 |
| 6Ba | Bi | all | <i>Interspecies</i> | Longbilled Dowitcher | 0.56 |
| 6Ba | Bi | all | <i>Plot</i> | Longbilled Dowitcher | 0.394 |
| 6Ba | Bi | ran | <i>Covariates</i> | Longbilled Dowitcher | 0.5 |
| 6Ba | Bi | ran | <i>Interspecies</i> | Longbilled Dowitcher | 0.484 |
| 6Ba | Bi | sys | <i>Covariates</i> | Longbilled Dowitcher | 0.545 |
| 6Ba | Bi | sys | <i>Interspecies</i> | Longbilled Dowitcher | 0.546 |
| 6Ba | Bi | all | <i>Covariates</i> | Pectoral Sandpiper | 0.653 |
| 6Ba | Bi | all | <i>Interspecies</i> | Pectoral Sandpiper | 0.671 |
| 6Ba | Bi | all | <i>Plot</i> | Pectoral Sandpiper | 0.393 |
| 6Ba | Bi | ran | <i>Covariates</i> | Pectoral Sandpiper | n/a |
| 6Ba | Bi | ran | <i>Interspecies</i> | Pectoral Sandpiper | n/a |
| 6Ba | Bi | sys | <i>Covariates</i> | Pectoral Sandpiper | 0.613 |
| 6Ba | Bi | sys | <i>Interspecies</i> | Pectoral Sandpiper | 0.615 |
| 6Ba | Bi | all | <i>Covariates</i> | Pomarine Jaeger | 0.629 |
| 6Ba | Bi | all | <i>Interspecies</i> | Pomarine Jaeger | 0.626 |
| 6Ba | Bi | all | <i>Plot</i> | Pomarine Jaeger | 0.597 |
| 6Ba | Bi | ran | <i>Covariates</i> | Pomarine Jaeger | 0.331 |
| 6Ba | Bi | ran | <i>Interspecies</i> | Pomarine Jaeger | 0.307 |
| 6Ba | Bi | sys | <i>Covariates</i> | Pomarine Jaeger | 0.709 |
| 6Ba | Bi | sys | <i>Interspecies</i> | Pomarine Jaeger | 0.716 |
| 6Ba | Bi | all | <i>Covariates</i> | Red Phalarope | 0.585 |
| 6Ba | Bi | all | <i>Interspecies</i> | Red Phalarope | 0.577 |
| 6Ba | Bi | all | <i>Plot</i> | Red Phalarope | 0.620 |
| 6Ba | Bi | ran | <i>Covariates</i> | Red Phalarope | 0.387 |
| 6Ba | Bi | ran | <i>Interspecies</i> | Red Phalarope | 0.371 |
| 6Ba | Bi | sys | <i>Covariates</i> | Red Phalarope | 0.589 |
| 6Ba | Bi | sys | <i>Interspecies</i> | Red Phalarope | 0.594 |
| 6Ba | Bi | all | <i>Covariates</i> | Red-necked Phalarope | 0.630 |
| 6Ba | Bi | all | <i>Interspecies</i> | Red-necked Phalarope | 0.682 |
| 6Ba | Bi | all | <i>Plot</i> | Red-necked Phalarope | 0.304 |
| 6Ba | Bi | ran | <i>Covariates</i> | Red-necked Phalarope | 0.66 |
| 6Ba | Bi | ran | <i>Interspecies</i> | Red-necked Phalarope | 0.634 |
| 6Ba | Bi | sys | <i>Covariates</i> | Red-necked Phalarope | 0.576 |
| 6Ba | Bi | sys | <i>Interspecies</i> | Red-necked Phalarope | 0.59 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|------------------------|-----------|
| 6Ba | Bi | all | <i>Covariates</i> | Semipalmated Sandpiper | 0.458 |
| 6Ba | Bi | all | <i>Interspecies</i> | Semipalmated Sandpiper | 0.463 |
| 6Ba | Bi | all | <i>Plot</i> | Semipalmated Sandpiper | 0.403 |
| 6Ba | Bi | ran | <i>Covariates</i> | Semipalmated Sandpiper | 0.536 |
| 6Ba | Bi | sys | <i>Covariates</i> | Semipalmated Sandpiper | 0.461 |
| 6Ba | Bi | sys | <i>Interspecies</i> | Semipalmated Sandpiper | 0.462 |
| 1CR | DT | all | <i>Covariates</i> | Butterfly, yellow | 0.906 |
| 1CR | DT | ran | <i>Covariates</i> | Butterfly, yellow | 0.4 |
| 1CR | DT | sys | <i>Covariates</i> | Butterfly, yellow | 0.969 |
| 1CR | DT | all | <i>Interspecies</i> | Butterfly, yellow | 0.953 |
| 1CR | DT | ran | <i>Interspecies</i> | Butterfly, yellow | 0.6 |
| 1CR | DT | sys | <i>Interspecies</i> | Butterfly, yellow | 0.969 |
| 1CR | DT | all | <i>Plot</i> | Butterfly, yellow | 0.815 |
| 2Ni | DT | all | <i>Covariates</i> | Butterfly, white | 0.629 |
| 2Ni | DT | all | <i>Interspecies</i> | Butterfly, white | 0.609 |
| 2Ni | DT | all | <i>Plot</i> | Butterfly, white | 0.305 |
| 2Ni | DT | ran | <i>Covariates</i> | Butterfly, white | 0.857 |
| 2Ni | DT | ran | <i>Interspecies</i> | Butterfly, white | 0.821 |
| 2Ni | DT | sys | <i>Covariates</i> | Butterfly, white | 0.528 |
| 2Ni | DT | sys | <i>Interspecies</i> | Butterfly, white | 0.487 |
| 2Ni | DT | all | <i>Covariates</i> | Butterfly, yellow | 0.65 |
| 2Ni | DT | all | <i>Interspecies</i> | Butterfly, yellow | 0.665 |
| 2Ni | DT | all | <i>Plot</i> | Butterfly, yellow | 0.456 |
| 2Ni | DT | ran | <i>Covariates</i> | Butterfly, yellow | n/a |
| 2Ni | DT | ran | <i>Interspecies</i> | Butterfly, yellow | n/a |
| 2Ni | DT | sys | <i>Covariates</i> | Butterfly, yellow | 0.598 |
| 2Ni | DT | sys | <i>Interspecies</i> | Butterfly, yellow | 0.574 |
| 1CR | TW | all | <i>Covariates</i> | ant | 0.847 |
| 1CR | TW | all | <i>Interspecies</i> | ant | 0.831 |
| 1CR | TW | all | <i>Covariates</i> | ant, small | 0.634 |
| 1CR | TW | all | <i>Interspecies</i> | ant, small | 0.667 |
| 1CR | TW | all | <i>Covariates</i> | cricket | 0.838 |
| 1CR | TW | all | <i>Interspecies</i> | cricket | 0.816 |
| 1CR | TW | all | <i>Covariates</i> | spider | 0.44 |
| 1CR | TW | all | <i>Interspecies</i> | spider | 0.446 |
| 2Ni | TW | all | <i>Covariates</i> | ant | 0.696 |
| 2Ni | TW | all | <i>Interspecies</i> | ant | 0.699 |
| 2Ni | TW | all | <i>Covariates</i> | ant, small | 0.661 |
| 2Ni | TW | all | <i>Interspecies</i> | ant, small | 0.612 |
| 2Ni | TW | all | <i>Covariates</i> | ant, small red | 0.581 |
| 2Ni | TW | all | <i>Interspecies</i> | ant, small red | 0.496 |
| 2Ni | TW | all | <i>Covariates</i> | beetle, 866 | 0.652 |
| 2Ni | TW | all | <i>Interspecies</i> | beetle, 866 | 0.473 |
| 2Ni | TW | all | <i>Covariates</i> | beetle, 868 | 0.724 |
| 2Ni | TW | all | <i>Interspecies</i> | beetle, 868 | 0.72 |
| 2Ni | TW | all | <i>Covariates</i> | beetle, 874 | 0.736 |
| 2Ni | TW | all | <i>Interspecies</i> | beetle, 874 | 0.722 |
| 2Ni | TW | all | <i>Covariates</i> | beetle, 929 | 0.53 |
| 2Ni | TW | all | <i>Interspecies</i> | beetle, 929 | 0.142 |
| 2Ni | TW | all | <i>Covariates</i> | beetle, ground | 0.461 |
| 2Ni | TW | all | <i>Interspecies</i> | beetle, ground | 0.181 |
| 2Ni | TW | all | <i>Covariates</i> | bristletail | 0.379 |
| 2Ni | TW | all | <i>Interspecies</i> | bristletail | 0.232 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|----------------------------|-----------|
| 2Ni | TW | all | <i>Covariates</i> | caterpillar, 875 | 0.637 |
| 2Ni | TW | all | <i>Interspecies</i> | caterpillar, 875 | 0.487 |
| 2Ni | TW | all | <i>Covariates</i> | caterpillar, 877 | 0.324 |
| 2Ni | TW | all | <i>Interspecies</i> | caterpillar, 877 | 0.089 |
| 2Ni | TW | all | <i>Covariates</i> | centipede, 881 | 0.77 |
| 2Ni | TW | all | <i>Interspecies</i> | centipede, 881 | 0.728 |
| 2Ni | TW | all | <i>Covariates</i> | cricket | 0.773 |
| 2Ni | TW | all | <i>Interspecies</i> | cricket | 0.68 |
| 2Ni | TW | all | <i>Covariates</i> | spider, small | 0.604 |
| 2Ni | TW | all | <i>Interspecies</i> | spider, small | 0.519 |
| 2Ni | TW | all | <i>Covariates</i> | springtail | 0.958 |
| 2Ni | TW | all | <i>Interspecies</i> | springtail | 0.925 |
| 3AK | TW | all | <i>Covariates</i> | ant | 0.85 |
| 3AK | TW | all | <i>Interspecies</i> | ant | 0.833 |
| 3AK | TW | all | <i>Covariates</i> | beetle | 0.365 |
| 3AK | TW | all | <i>Interspecies</i> | beetle | 0.349 |
| 3AK | TW | all | <i>Covariates</i> | beetle, underground-hiding | 0.676 |
| 3AK | TW | all | <i>Interspecies</i> | beetle, underground-hiding | 0.696 |
| 3AK | TW | all | <i>Covariates</i> | spider | 0.604 |
| 3AK | TW | all | <i>Interspecies</i> | spider | 0.567 |
| 3AK | TW | all | <i>Covariates</i> | spider, small red | 0.724 |
| 3AK | TW | all | <i>Interspecies</i> | spider, small red | 0.746 |
| 3AK | TW | all | <i>Covariates</i> | spider, tiny | 0.506 |
| 3AK | TW | all | <i>Interspecies</i> | spider, tiny | 0.463 |
| 3AK | TW | all | <i>Covariates</i> | springtail | 0.951 |
| 3AK | TW | all | <i>Interspecies</i> | springtail | 0.939 |
| 4Ru | TW | all | <i>Covariates</i> | Beetle | 0.441 |
| 4Ru | TW | all | <i>Interspecies</i> | Beetle | 0.421 |
| 4Ru | TW | all | <i>Covariates</i> | Carabidae | 0.76 |
| 4Ru | TW | all | <i>Interspecies</i> | Carabidae | 0.763 |
| 4Ru | TW | all | <i>Covariates</i> | Collembola | 0.616 |
| 4Ru | TW | all | <i>Interspecies</i> | Collembola | 0.614 |
| 4Ru | TW | all | <i>Covariates</i> | cycsegusa | 0.74 |
| 4Ru | TW | all | <i>Interspecies</i> | cycsegusa | 0.723 |
| 4Ru | TW | all | <i>Covariates</i> | mouse | 0.83 |
| 4Ru | TW | all | <i>Interspecies</i> | mouse | 0.817 |
| 4Ru | TW | all | <i>Covariates</i> | Protura | 0.709 |
| 4Ru | TW | all | <i>Interspecies</i> | Protura | 0.681 |
| 4Ru | TW | all | <i>Covariates</i> | Spider | 0.223 |
| 4Ru | TW | all | <i>Interspecies</i> | Spider | 0.124 |
| 4Ru | TW | all | <i>Covariates</i> | spider with slim long legs | 0.897 |
| 4Ru | TW | all | <i>Interspecies</i> | spider with slim long legs | 0.897 |
| 4Ru | TW | all | <i>Covariates</i> | spider, big | 0.694 |
| 4Ru | TW | all | <i>Interspecies</i> | spider, big | 0.666 |
| 4Ru | TW | all | <i>Covariates</i> | spider, little | 0.628 |
| 4Ru | TW | all | <i>Interspecies</i> | spider, little | 0.627 |
| 4Ru | TW | all | <i>Covariates</i> | Staphilin | 0.658 |
| 4Ru | TW | all | <i>Interspecies</i> | Staphilin | 0.604 |
| 4Ru | TW | all | <i>Covariates</i> | worm | 0.645 |
| 4Ru | TW | all | <i>Interspecies</i> | worm | 0.613 |
| 5PG | TW | all | <i>Covariates</i> | ant, big yellow | 0.955 |

| Study area | Type | Data | Model | Target narrative | ROC value |
|------------|------|------|---------------------|---------------------------|-----------|
| 5PG | TW | all | <i>Interspecies</i> | ant, big yellow | 0.927 |
| 5PG | TW | all | <i>Covariates</i> | ant, black | 0.498 |
| 5PG | TW | all | <i>Interspecies</i> | ant, black | 0.525 |
| 5PG | TW | all | <i>Covariates</i> | ant, little black | 0.644 |
| 5PG | TW | all | <i>Interspecies</i> | ant, little black | 0.6 |
| 5PG | TW | all | <i>Covariates</i> | ant, tiny black | 0.39 |
| 5PG | TW | all | <i>Interspecies</i> | ant, tiny black | 0.393 |
| 5PG | TW | all | <i>Covariates</i> | ant, tiny red | 0.915 |
| 5PG | TW | all | <i>Interspecies</i> | ant, tiny red | 0.855 |
| 5PG | TW | all | <i>Covariates</i> | ant, yellow | 0.734 |
| 5PG | TW | all | <i>Interspecies</i> | ant, yellow | 0.658 |
| 5PG | TW | all | <i>Covariates</i> | fly | 0.565 |
| 5PG | TW | all | <i>Interspecies</i> | fly | 0.527 |
| 5PG | TW | all | <i>Covariates</i> | fruitfly | 0.603 |
| 5PG | TW | all | <i>Interspecies</i> | fruitfly | 0.561 |
| 5PG | TW | all | <i>Covariates</i> | fruitfly, white | 0.795 |
| 5PG | TW | all | <i>Interspecies</i> | fruitfly, white | 0.783 |
| 5PG | TW | all | <i>Covariates</i> | springfloh | 0.51 |
| 5PG | TW | all | <i>Interspecies</i> | springfloh | 0.385 |
| 5PG | TW | all | <i>Covariates</i> | springfloh mit antennae | 0.84 |
| 5PG | TW | all | <i>Interspecies</i> | springfloh mit antennae | 0.818 |
| 5PG | TW | all | <i>Covariates</i> | springfloh, long antennae | 0.561 |
| 5PG | TW | all | <i>Interspecies</i> | springfloh, long antennae | 0.475 |
| 6Ba | TW | all | <i>Covariates</i> | beetle, flat | 0.678 |
| 6Ba | TW | all | <i>Interspecies</i> | beetle, flat | 0.639 |
| 6Ba | TW | all | <i>Covariates</i> | beetle, slim | 0.802 |
| 6Ba | TW | all | <i>Interspecies</i> | beetle, slim | 0.661 |
| 6Ba | TW | all | <i>Covariates</i> | fly | 0.696 |
| 6Ba | TW | all | <i>Interspecies</i> | fly | 0.679 |
| 6Ba | TW | all | <i>Covariates</i> | Fruitfly | 0.841 |
| 6Ba | TW | all | <i>Interspecies</i> | Fruitfly | 0.802 |
| 6Ba | TW | all | <i>Covariates</i> | fruitfly, tiny | 0.664 |
| 6Ba | TW | all | <i>Interspecies</i> | fruitfly, tiny | 0.651 |
| 6Ba | TW | all | <i>Covariates</i> | Milbe | 0.821 |
| 6Ba | TW | all | <i>Interspecies</i> | Milbe | 0.797 |
| 6Ba | TW | all | <i>Covariates</i> | mosquito | 0.871 |
| 6Ba | TW | all | <i>Interspecies</i> | mosquito | 0.810 |
| 6Ba | TW | all | <i>Covariates</i> | Schuster | 0.313 |
| 6Ba | TW | all | <i>Interspecies</i> | Schuster | 0.223 |
| 6Ba | TW | all | <i>Covariates</i> | spider | 0.548 |
| 6Ba | TW | all | <i>Interspecies</i> | spider | 0.527 |
| 6Ba | TW | all | <i>Covariates</i> | spider, little | 0.852 |
| 6Ba | TW | all | <i>Interspecies</i> | spider, little | 0.798 |
| 6Ba | TW | all | <i>Covariates</i> | spider, tiny | 0.729 |
| 6Ba | TW | all | <i>Interspecies</i> | spider, tiny | 0.694 |
| 6Ba | TW | all | <i>Covariates</i> | Springmilbe | 0.867 |
| 6Ba | TW | all | <i>Interspecies</i> | Springmilbe | 0.811 |

7.7 Allocation of Narrative Names to Biological Order/Family

| Study area | Level | Target | Pooled narratives |
|------------|-------|----------------|---------------------------|
| 1CR | Order | Passeriformes | Flycatcher |
| 1CR | Order | Passeriformes | Golden-bellied Flycatcher |
| 1CR | Order | Passeriformes | Golden-hooded Tanager |
| 1CR | Order | Passeriformes | Great Kiskadee |
| 1CR | Order | Passeriformes | Kiskadee |
| 1CR | Order | Passeriformes | Lesser Kiskadee |
| 1CR | Order | Passeriformes | Manakin |
| 1CR | Order | Passeriformes | Oropendula |
| 1CR | Order | Passeriformes | Pale-vented Thrush |
| 1CR | Order | Passeriformes | Scarlet-rumped Tanager |
| 1CR | Order | Passeriformes | Seedeater |
| 1CR | Order | Passeriformes | Songbird |
| 1CR | Order | Passeriformes | Songbird, brown |
| 1CR | Order | Passeriformes | Songbird, little |
| 1CR | Order | Passeriformes | Steep-forehead Flycatcher |
| 1CR | Order | Passeriformes | Swallow |
| 1CR | Order | Passeriformes | Tanager |
| 1CR | Order | Passeriformes | Thrush |
| 1CR | Order | Passeriformes | Treecreper |
| 1CR | Order | Passeriformes | Yellow-bellied Flycatcher |
| 1CR | Order | Piciformes | Gray-necked Woodpecker |
| 1CR | Order | Piciformes | Toucan |
| 1CR | Order | Piciformes | Woodpecker |
| 1CR | Order | Psittaciformes | Mealy Parrot |
| 1CR | Order | Psittaciformes | Parrot |
| 1CR | Order | Psittaciformes | Parrot, large |
| 1CR | Order | Psittaciformes | Parrot, little |
| 2Ni | Order | Acariformes | mite, red |
| 2Ni | Order | Acariformes | mite, red 925 |
| 2Ni | Order | Araneae | spider |
| 2Ni | Order | Araneae | spider, black 892 |
| 2Ni | Order | Araneae | spider, red |
| 2Ni | Order | Araneae | spider, small |
| 2Ni | Order | Araneae | spider, small red |
| 2Ni | Order | Ciconiiformes | Cattle Egret |
| 2Ni | Order | Ciconiiformes | Gray Hawk |
| 2Ni | Order | Ciconiiformes | Hawk |
| 2Ni | Order | Ciconiiformes | Magnificent Frigatebird |
| 2Ni | Order | Ciconiiformes | Turkey Vulture |
| 2Ni | Order | Ciconiiformes | Vulture |
| 2Ni | Order | Lepidoptera | Butterfly, black-red |
| 2Ni | Order | Lepidoptera | Butterfly, black-yellow |
| 2Ni | Order | Lepidoptera | Butterfly, grey |
| 2Ni | Order | Lepidoptera | Butterfly, large yellow |
| 2Ni | Order | Lepidoptera | Butterfly, orange |
| 2Ni | Order | Lepidoptera | Butterfly, orange-white |
| 2Ni | Order | Lepidoptera | Butterfly, small black |
| 2Ni | Order | Lepidoptera | Butterfly, small white |
| 2Ni | Order | Lepidoptera | butterfly, swallowtail |
| 2Ni | Order | Lepidoptera | Butterfly, white |

| Study area | Level | Target | Pooled narratives |
|-------------------|--------------|---------------|----------------------------|
| 2Ni | Order | Lepidoptera | Butterfly, yellow |
| 2Ni | Order | Lepidoptera | caterpillar, 875 |
| 2Ni | Order | Lepidoptera | caterpillar, 877 |
| 2Ni | Order | Lepidoptera | caterpillar, 942-943 |
| 2Ni | Order | Lepidoptera | moth |
| 2Ni | Order | Passeriformes | Banded Wren |
| 2Ni | Order | Passeriformes | Brown-crested Flycatcher |
| 2Ni | Order | Passeriformes | Flycatcher |
| 2Ni | Order | Passeriformes | Great Kiskadee |
| 2Ni | Order | Passeriformes | Jay |
| 2Ni | Order | Passeriformes | Masked Tityra |
| 2Ni | Order | Passeriformes | Seedeater |
| 2Ni | Order | Passeriformes | Songbird |
| 2Ni | Order | Passeriformes | Swallow |
| 2Ni | Order | Passeriformes | Tanager |
| 2Ni | Order | Passeriformes | White-throated Magpie Jay |
| 3AK | Order | Araneae | spider |
| 3AK | Order | Araneae | spider, small |
| 3AK | Order | Araneae | spider, small black |
| 3AK | Order | Araneae | spider, small red |
| 3AK | Order | Araneae | spider, tiny |
| 3AK | Order | Coleoptera | beetle |
| 3AK | Order | Coleoptera | beetle, underground-hiding |
| 3AK | Order | Passeriformes | American Robin |
| 3AK | Order | Passeriformes | Boreal Chickadee |
| 3AK | Order | Passeriformes | Chickadee |
| 3AK | Order | Passeriformes | Corvidae |
| 3AK | Order | Passeriformes | Dark-eyed Junco |
| 3AK | Order | Passeriformes | Gray Jay |
| 3AK | Order | Passeriformes | Junco |
| 3AK | Order | Passeriformes | Songbird |
| 3AK | Order | Passeriformes | Sparrow |
| 3AK | Order | Passeriformes | White-crowned Sparrow |
| 3AK | Order | Passeriformes | Yellow-rumped Warbler |
| 4Ru | Order | Araneae | Spider |
| 4Ru | Order | Araneae | spider with slim long legs |
| 4Ru | Order | Araneae | spider, big |
| 4Ru | Order | Araneae | spider, little |
| 4Ru | Order | Araneae | spider, midsize |
| 4Ru | Order | Araneae | spider, palekolane |
| 4Ru | Order | Passeriformes | bluetail |
| 4Ru | Order | Passeriformes | Chickadee |
| 4Ru | Order | Passeriformes | Crow |
| 4Ru | Order | Passeriformes | Emberiza |
| 4Ru | Order | Passeriformes | Finch |
| 4Ru | Order | Passeriformes | Flycatcher |
| 4Ru | Order | Passeriformes | Grasshopper Warbler |
| 4Ru | Order | Passeriformes | Jungle Crow |
| 4Ru | Order | Passeriformes | Juv passerine |
| 4Ru | Order | Passeriformes | Kinglet |
| 4Ru | Order | Passeriformes | Kohlmeise |
| 4Ru | Order | Passeriformes | longtailed tit |
| 4Ru | Order | Passeriformes | Nutcracker |
| 4Ru | Order | Passeriformes | nuthatch |

| Study area | Level | Target | Pooled narratives |
|------------|-------|----------------|----------------------------|
| 4Ru | Order | Passeriformes | Oriental Finch |
| 4Ru | Order | Passeriformes | Oriental Greenfinch |
| 4Ru | Order | Passeriformes | passerine |
| 4Ru | Order | Passeriformes | Raven |
| 4Ru | Order | Passeriformes | Tannenmeise |
| 4Ru | Order | Passeriformes | Thrush |
| 4Ru | Order | Passeriformes | Wagtail |
| 4Ru | Order | Passeriformes | Warbler |
| 4Ru | Order | Passeriformes | Weidenmeise |
| 4Ru | Order | Passeriformes | Winter Wren |
| 5PG | Order | Acariformes | milbe, red |
| 5PG | Order | Acariformes | milbe, spring |
| 5PG | Order | Araneae | spider, little |
| 5PG | Order | Araneae | spider, little black |
| 5PG | Order | Araneae | spider, little long legs |
| 5PG | Order | Araneae | spider, medium |
| 5PG | Order | Araneae | spider, tiny black |
| 5PG | Order | Collembola | Collembola |
| 5PG | Order | Collembola | collembola long antennae |
| 5PG | Order | Collembola | collembola, big yellow |
| 5PG | Order | Collembola | collembola, black-yellow |
| 5PG | Order | Collembola | collembola, yellow |
| 5PG | Order | Diptera | fly |
| 5PG | Order | Diptera | fly with legs and antennae |
| 5PG | Order | Diptera | fly, tiny |
| 5PG | Order | Diptera | fruitfly |
| 5PG | Order | Diptera | fruitfly black |
| 5PG | Order | Diptera | fruitfly grey |
| 5PG | Order | Diptera | fruitfly, blue |
| 5PG | Order | Diptera | fruitfly, pink |
| 5PG | Order | Diptera | fruitfly, white |
| 5PG | Order | Diptera | mosquito |
| 5PG | Order | Diptera | mosquito, jumping |
| 5PG | Order | Passeriformes | Craw, Bird of Paradise |
| 5PG | Order | Passeriformes | Flycatcher |
| 5PG | Order | Passeriformes | Flycatcher tschirrp |
| 5PG | Order | Passeriformes | Flycatcher, similar willie |
| 5PG | Order | Passeriformes | Kau Kau, Bird of Paradise |
| 5PG | Order | Passeriformes | Rezina, rezina |
| 5PG | Order | Passeriformes | Songbird |
| 5PG | Order | Passeriformes | Songbird little |
| 5PG | Order | Passeriformes | songbird tshirp |
| 5PG | Order | Passeriformes | Songbird tsilp |
| 5PG | Order | Passeriformes | Swallow |
| 5PG | Order | Passeriformes | Thrush |
| 5PG | Order | Passeriformes | wren |
| 5PG | Order | Psittaciformes | Cockatoo |
| 5PG | Order | Psittaciformes | Palm Cockatoo |
| 5PG | Order | Psittaciformes | Parakeet |
| 5PG | Order | Psittaciformes | Parrot |
| 5PG | Order | Psittaciformes | Parrot, little |
| 5PG | Order | Psittaciformes | White Cockatoo |
| 6Ba | Order | Ciconiiformes | Dowitcher |
| 6Ba | Order | Ciconiiformes | Dunlin |

| Study area | Level | Target | Pooled narratives |
|-------------------|--------------|---------------|---------------------------|
| 6Ba | Order | Ciconiiformes | Glaucous Gull |
| 6Ba | Order | Ciconiiformes | Longbilled Dowitcher |
| 6Ba | Order | Ciconiiformes | Loon |
| 6Ba | Order | Ciconiiformes | Pacific Loon |
| 6Ba | Order | Ciconiiformes | Parasitic Jaeger |
| 6Ba | Order | Ciconiiformes | Pectoral Sandpiper |
| 6Ba | Order | Ciconiiformes | Phalarope |
| 6Ba | Order | Ciconiiformes | Pomarine Jaeger |
| 6Ba | Order | Ciconiiformes | Red Phalarope |
| 6Ba | Order | Ciconiiformes | Red-necked Phalarope |
| 6Ba | Order | Ciconiiformes | Semipalmated Sandpiper |
| 6Ba | Order | Ciconiiformes | Western Sandpiper |
| 6Ba | Order | Coleoptera | beetle |
| 6Ba | Order | Coleoptera | beetle, flat |
| 6Ba | Order | Coleoptera | beetle, gold-green |
| 6Ba | Order | Coleoptera | beetle, green |
| 6Ba | Order | Coleoptera | beetle, little |
| 6Ba | Order | Coleoptera | beetle, little green |
| 6Ba | Order | Coleoptera | beetle, slim |
| 6Ba | Order | Coleoptera | Marienkaeferlarve |
| 6Ba | Order | Diptera | fly |
| 6Ba | Order | Diptera | fly, little |
| 6Ba | Order | Diptera | Fruitfly |
| 6Ba | Order | Diptera | fruitfly, little |
| 6Ba | Order | Diptera | fruitfly, tiny |
| 6Ba | Order | Diptera | mosquito |
| 6Ba | Order | Diptera | Schuster |
| 6Ba | Order | Diptera | schuster, big |
| 6Ba | Order | Diptera | Schuster, large |
| 6Ba | Order | Diptera | Schuster, no wings |
| 6Ba | Order | Passeriformes | Lapland Bunting |
| 6Ba | Order | Passeriformes | Snow Bunting |
| 1CR | Family | Thraupidae | Golden-hooded Tanager |
| 1CR | Family | Thraupidae | Scarlet-rumped Tanager |
| 1CR | Family | Thraupidae | Seedeater |
| 1CR | Family | Thraupidae | Tanager |
| 1CR | Family | Tyrannidae | Flycatcher |
| 1CR | Family | Tyrannidae | Golden-bellied Flycatcher |
| 1CR | Family | Tyrannidae | Great Kiskadee |
| 1CR | Family | Tyrannidae | Kiskadee |
| 1CR | Family | Tyrannidae | Lesser Kiskadee |
| 1CR | Family | Tyrannidae | Steep-forehead Flycatcher |
| 1CR | Family | Tyrannidae | Yellow-bellied Flycatcher |
| 2Ni | Family | Formicidae | ant |
| 2Ni | Family | Formicidae | ant, red |
| 2Ni | Family | Formicidae | ant, small |
| 2Ni | Family | Formicidae | ant, small black |
| 2Ni | Family | Formicidae | ant, small red |
| 3AK | Family | Emberizidae | Dark-eyed Junco |
| 3AK | Family | Emberizidae | Junco |
| 3AK | Family | Emberizidae | Sparrow |
| 3AK | Family | Emberizidae | White-crowned Sparrow |
| 4Ru | Family | Corvidae | Crow |
| 4Ru | Family | Corvidae | Jungle Crow |

| Study area | Level | Target | Pooled narratives |
|-------------------|--------------|----------------|---------------------------|
| 4Ru | Family | Corvidae | Nutcracker |
| 4Ru | Family | Corvidae | Raven |
| 4Ru | Family | Paridae | Chickadee |
| 4Ru | Family | Paridae | Kohlmeise |
| 4Ru | Family | Paridae | longtailed tit |
| 4Ru | Family | Paridae | Tannenmeise |
| 4Ru | Family | Paridae | Weidenmeise |
| 5PG | Family | Formicidae | ant, big |
| 5PG | Family | Formicidae | ant, big black |
| 5PG | Family | Formicidae | ant, big yellow |
| 5PG | Family | Formicidae | ant, black |
| 5PG | Family | Formicidae | ant, little |
| 5PG | Family | Formicidae | ant, little black |
| 5PG | Family | Formicidae | ant, little red |
| 5PG | Family | Formicidae | ant, medium black |
| 5PG | Family | Formicidae | ant, red |
| 5PG | Family | Formicidae | ant, tiny |
| 5PG | Family | Formicidae | ant, tiny black |
| 5PG | Family | Formicidae | ant, tiny red |
| 5PG | Family | Formicidae | ant, tiny yellow |
| 5PG | Family | Formicidae | ant, yellow |
| 5PG | Family | Paradisaeidae | Craw, Bird of Paradise |
| 5PG | Family | Paradisaeidae | Kau Kau, Bird of Paradise |
| 5PG | Family | Paradisaeidae | Rezina, rezina |
| 6Ba | Family | Scolopacidae | Dowitcher |
| 6Ba | Family | Scolopacidae | Dunlin |
| 6Ba | Family | Scolopacidae | Longbilled Dowitcher |
| 6Ba | Family | Scolopacidae | Pectoral Sandpiper |
| 6Ba | Family | Scolopacidae | Phalarope |
| 6Ba | Family | Scolopacidae | Red Phalarope |
| 6Ba | Family | Scolopacidae | Red-necked Phalarope |
| 6Ba | Family | Scolopacidae | Semipalmated Sandpiper |
| 6Ba | Family | Scolopacidae | Western Sandpiper |
| 6Ba | Family | Stercorariidae | Parasitic Jaeger |
| 6Ba | Family | Stercorariidae | Pomarine Jaeger |
| 6Ba | Family | Tipulidae | Schuster |
| 6Ba | Family | Tipulidae | schuster, big |
| 6Ba | Family | Tipulidae | Schuster, large |
| 6Ba | Family | Tipulidae | Schuster, no wings |

7.8 Best Models (DISTANCE Sampling)

| Study area | Target narrative | Type | Data | Model definition | ESW/EDR | D | D LCL | D UCL | D CV | P | P LCL | P UCL |
|------------|------------------|------|------|------------------|---------|--------|--------|--------|------|-----|-------|-------|
| 1CR | Flycatcher | Bi | all | 26 | 27.1 | 117.3 | 14.8 | 929.7 | 1.3 | 0.4 | 32.0 | 0.0 |
| 1CR | Flycatcher | Bi | aur | 26 | 30.6 | 17.4 | 6.8 | 44.3 | 0.5 | 0.5 | 18.0 | 0.2 |
| 1CR | Flycatcher | Bi | ran | 26 | | | | | | | | |
| 1CR | Flycatcher | Bi | sys | 26 | 26.9 | 41.8 | 24.5 | 71.2 | 0.3 | 0.4 | 28.0 | 0.2 |
| 1CR | Flycatcher | Bi | vis | 26 | 26.7 | 17.9 | 0.8 | 385.2 | 2.5 | 0.4 | 12.0 | 0.0 |
| 1CR | Hummingbird | Bi | all | 16 | 8.7 | 2908.0 | 1992.3 | 4244.5 | 0.2 | 0.2 | 94.0 | 0.1 |
| 1CR | Hummingbird | Bi | aur | 16 | 12.8 | 302.3 | 184.0 | 496.6 | 0.3 | 0.4 | 61.0 | 0.3 |
| 1CR | Hummingbird | Bi | ran | 16 | 5.5 | 3257.9 | 1207.0 | 8793.8 | 0.5 | 0.1 | 19.0 | 0.0 |
| 1CR | Hummingbird | Bi | sys | 16 | 9.3 | 780.4 | 511.5 | 1190.6 | 0.2 | 0.2 | 73.0 | 0.2 |
| 1CR | Hummingbird | Bi | vis | 16 | 6.1 | 661.9 | 409.4 | 1070.1 | 0.2 | 0.1 | 33.0 | 0.1 |
| 1CR | Oropendula | Bi | all | 1 | 7.7 | 2296.9 | 1208.7 | 4364.7 | 0.3 | 0.1 | 42.0 | 0.1 |
| 1CR | Oropendula | Bi | aur | 1 | 18.4 | 23.0 | 7.0 | 76.3 | 0.6 | 0.8 | 10.0 | 0.3 |
| 1CR | Oropendula | Bi | ran | 1 | 13.4 | 306.2 | 61.3 | 1528.7 | 0.9 | 0.4 | 5.0 | 0.1 |
| 1CR | Oropendula | Bi | sys | 1 | 7.6 | 768.3 | 389.0 | 1517.6 | 0.4 | 0.1 | 36.0 | 0.1 |
| 1CR | Oropendula | Bi | vis | 1 | 6.8 | 817.9 | 422.0 | 1585.3 | 0.3 | 0.1 | 31.0 | 0.1 |
| 1CR | Seed eater | Bi | all | 39 | 7.2 | 893.8 | 296.3 | 2696.3 | 0.6 | 0.1 | 14.0 | 0.1 |
| 1CR | Seed eater | Bi | aur | 39 | | 17.7 | 3.2 | 97.1 | 1.0 | | | |
| 1CR | Seed eater | Bi | ran | 39 | | 106.1 | 10.5 | 1070.6 | 1.0 | | | |
| 1CR | Seed eater | Bi | sys | 39 | 6.5 | 410.5 | 132.2 | 1275.0 | 0.6 | 0.1 | 13.0 | 0.1 |
| 1CR | Seed eater | Bi | vis | 39 | 7.0 | 322.7 | 102.0 | 1021.3 | 0.6 | 0.1 | 13.0 | 0.1 |
| 1CR | Woodpecker | Bi | all | 11 | 25.1 | 80.9 | 0.0 | 0.0 | 0.0 | 0.5 | 22.0 | 0.4 |
| 1CR | Woodpecker | Bi | aur | 11 | | | | | | | | |
| 1CR | Woodpecker | Bi | ran | 11 | | | | | | | | |
| 1CR | Woodpecker | Bi | sys | 11 | 25.4 | 22.3 | 0.0 | 0.0 | 0.0 | 0.5 | 15.0 | 0.4 |
| 1CR | Woodpecker | Bi | vis | 11 | 23.5 | 14.6 | 0.0 | 0.0 | 0.0 | 0.4 | 9.0 | 0.3 |
| 2Ni | BandedWren | Bi | aur | 45 | 26.8 | 31.6 | 18.3 | 54.7 | 0.3 | 0.8 | 27.0 | 0.6 |
| 2Ni | BandedWren | Bi | ran | 45 | | | | | | | | |
| 2Ni | BandedWren | Bi | sys | 45 | 25.3 | 36.1 | 19.3 | 67.5 | 0.3 | 0.7 | 23.0 | 0.5 |
| 2Ni | BandedWren | Bi | vis | 45 | | 47.2 | 8.4 | 264.7 | 0.9 | | | |

| Study area | Target narrative | Type | Data | Model definition | ESW/EDR | D | D LCL | D UCL | D CV | P | P LCL | P UCL |
|------------|---------------------------|------|------|------------------|---------|-------|-------|---------|-------|-----|-------|-------|
| 2Ni | White-throated Magpie Jay | Bi | all | 43 | 23.3 | 152.4 | 88.0 | 263.8 | 0.3 | 0.6 | 30.0 | 0.5 |
| 2Ni | White-throated Magpie Jay | Bi | aur | 43 | 23.5 | 40.9 | 0.1 | 20472.3 | 100.0 | 0.6 | 28.0 | 0.0 |
| 2Ni | White-throated Magpie Jay | Bi | ran | 43 | | | | | | | | |
| 2Ni | White-throated Magpie Jay | Bi | sys | 43 | 23.2 | 42.3 | 23.2 | 77.3 | 0.3 | 0.6 | 24.0 | 0.5 |
| 2Ni | White-throated Magpie Jay | Bi | vis | 43 | | 47.2 | 8.4 | 264.7 | 0.9 | | | |
| 3AK | Sparrow | Bi | all | 12 | 38.2 | 16.2 | 10.3 | 25.7 | 0.2 | 0.6 | 57.0 | 0.5 |
| 3AK | Sparrow | Bi | aur | 12 | 38.7 | 11.2 | 6.9 | 18.4 | 0.3 | 0.6 | 40.0 | 0.4 |
| 3AK | Sparrow | Bi | ran | 12 | 37.9 | 11.5 | 1.7 | 78.0 | 1.0 | 0.6 | 5.0 | 0.1 |
| 3AK | Sparrow | Bi | sys | 12 | 38.2 | 17.2 | 10.6 | 28.0 | 0.2 | 0.6 | 50.0 | 0.4 |
| 3AK | Sparrow | Bi | vis | 12 | 15.6 | 85.4 | 6.1 | 1202.5 | 1.9 | 0.1 | 15.0 | 0.0 |
| 3AK | Squirrel | Bi | all | 15 | 43.2 | 17.0 | 11.0 | 26.4 | 0.2 | 0.7 | 76.0 | 0.6 |
| 3AK | Squirrel | Bi | aur | 15 | 44.8 | 15.6 | 10.4 | 23.3 | 0.2 | 0.8 | 73.0 | 0.6 |
| 3AK | Squirrel | Bi | ran | 15 | 38.8 | 20.3 | 7.0 | 58.9 | 0.5 | 0.6 | 11.0 | 0.3 |
| 3AK | Squirrel | Bi | sys | 15 | 44.2 | 16.4 | 10.4 | 26.1 | 0.2 | 0.8 | 63.0 | 0.6 |
| 3AK | Squirrel | Bi | vis | 15 | 10.0 | 35.4 | 4.0 | 310.0 | 1.1 | 1.0 | 2.0 | 0.0 |
| 4Ru | Chickadee | Bi | all | 1 | 12.2 | 572.2 | 304.0 | 1077.1 | 0.3 | 0.2 | 66.0 | 0.1 |
| 4Ru | Chickadee | Bi | aur | 1 | 17.1 | 214.6 | 125.5 | 366.9 | 0.3 | 0.5 | 50.0 | 0.3 |
| 4Ru | Chickadee | Bi | ran | 1 | 23.5 | 132.7 | 48.3 | 364.4 | 0.5 | 0.9 | 11.0 | 0.4 |
| 4Ru | Chickadee | Bi | sys | 1 | 11.4 | 632.8 | 323.9 | 1236.1 | 0.3 | 0.2 | 54.0 | 0.1 |
| 4Ru | Chickadee | Bi | vis | 1 | 9.7 | 237.0 | 90.4 | 621.1 | 0.5 | 0.2 | 14.0 | 0.1 |
| 4Ru | Kinglet | Bi | all | 43 | 19.7 | 193.3 | 140.2 | 266.4 | 0.2 | 0.2 | 84.0 | 0.1 |
| 4Ru | Kinglet | Bi | aur | 43 | 19.1 | 131.7 | 90.4 | 191.8 | 0.2 | 0.1 | 61.0 | 0.1 |
| 4Ru | Kinglet | Bi | ran | 43 | | | | | | | | |
| 4Ru | Kinglet | Bi | sys | 43 | 19.8 | 182.4 | 128.5 | 258.9 | 0.2 | 0.2 | 61.0 | 0.1 |
| 4Ru | Kinglet | Bi | vis | 43 | | | | | | | | |
| 4Ru | Nutcracker | Bi | aur | 44 | 31.0 | 12.5 | 6.6 | 23.8 | 0.3 | 0.6 | 15.0 | 0.4 |
| 4Ru | Nutcracker | Bi | ran | 44 | | 159.2 | 16.5 | 1532.0 | 1.1 | | | |
| 4Ru | Nutcracker | Bi | sys | 44 | | | | | | | | |
| 4Ru | Nutcracker | Bi | vis | 44 | 14.5 | 26.2 | 1.2 | 550.5 | 2.0 | 0.1 | 6.0 | 0.0 |
| 4Ru | Warbler | Bi | all | 20 | 38.4 | 26.8 | 9.0 | 79.2 | 0.6 | 0.6 | 30.0 | 0.2 |
| 4Ru | Warbler | Bi | aur | 20 | 43.3 | 11.3 | 7.0 | 18.4 | 0.2 | 0.8 | 27.0 | 0.6 |
| 4Ru | Warbler | Bi | ran | 20 | | | | | | | | |
| 4Ru | Warbler | Bi | sys | 20 | 38.4 | 32.1 | 10.9 | 94.3 | 0.6 | 0.6 | 30.0 | 0.2 |

| Study area | Target narrative | Type | Data | Model definition | ESW/EDR | D | D LCL | D UCL | D CV | P | P LCL | P UCL |
|------------|----------------------|------|------|------------------|---------|-------|-------|------------|-------|-----|-------|-------|
| 4Ru | Warbler | Bi | vis | 20 | 4.9 | 93.6 | 0.0 | 281037.8 | 1.5 | 0.0 | 1.0 | 0.0 |
| 4Ru | Winter Wren | Bi | all | 67 | 30.0 | 16.5 | 8.1 | 33.5 | 0.4 | 0.4 | 19.0 | 0.2 |
| 4Ru | Winter Wren | Bi | aur | 67 | 30.0 | 16.5 | 8.1 | 33.5 | 0.4 | 0.4 | 19.0 | 0.2 |
| 4Ru | Winter Wren | Bi | ran | 67 | | | | | | | | |
| 4Ru | Winter Wren | Bi | sys | 67 | 30.0 | 19.8 | 9.9 | 39.7 | 0.4 | 0.4 | 19.0 | 0.2 |
| 4Ru | Winter Wren | Bi | vis | 67 | | | | | | | | |
| 4Ru | Wize | Bi | all | 111 | 33.2 | 17.4 | 5.3 | 57.7 | 0.6 | 0.4 | 24.0 | 0.1 |
| 4Ru | Wize | Bi | aur | 111 | 32.3 | 17.1 | 3.6 | 81.0 | 0.9 | 0.4 | 22.0 | 0.1 |
| 4Ru | Wize | Bi | ran | 111 | 19.1 | 35.0 | 6.4 | 190.5 | 0.7 | 0.9 | 1.0 | 0.0 |
| 4Ru | Wize | Bi | sys | 111 | | | | | | | | |
| 4Ru | Wize | Bi | vis | 111 | | 32.5 | 7.7 | 137.2 | 0.8 | | | |
| 5PG | Flute | Bi | all | 3 | 29.2 | 16.6 | 10.2 | 26.9 | 0.2 | 0.3 | 19.0 | 0.3 |
| 5PG | Flute | Bi | aur | 3 | 29.2 | 17.1 | 10.6 | 27.7 | 0.2 | 0.3 | 19.0 | 0.3 |
| 5PG | Flute | Bi | ran | 3 | 30.0 | 18.9 | 5.5 | 64.9 | 0.5 | 1.0 | 4.0 | 1.0 |
| 5PG | Flute | Bi | sys | 3 | 29.9 | 15.2 | 8.6 | 27.0 | 0.3 | 0.4 | 15.0 | 0.3 |
| 5PG | Flute | Bi | vis | 3 | | | | | | | | |
| 5PG | Tsilp | Bi | all | 11 | 20.8 | 67.1 | 0.0 | 0.0 | 0.0 | 0.5 | 39.0 | 0.0 |
| 5PG | Tsilp | Bi | aur | 11 | 23.0 | 55.6 | 0.0 | 0.0 | 0.0 | 0.6 | 38.0 | 0.5 |
| 5PG | Tsilp | Bi | ran | 11 | 24.5 | 42.5 | 0.0 | 0.0 | 0.0 | 0.7 | 4.0 | 0.3 |
| 5PG | Tsilp | Bi | sys | 11 | 22.6 | 58.1 | 0.0 | 0.0 | 0.0 | 0.6 | 33.0 | 0.5 |
| 5PG | Tsilp | Bi | vis | 11 | | | | | | | | |
| 6Ba | Lapland Bunting | Bi | all | 35 | 26.2 | 25.0 | 15.3 | 40.8 | 0.2 | 0.3 | 52.0 | 0.2 |
| 6Ba | Lapland Bunting | Bi | ran | 35 | | 45.3 | 4.5 | 457.5 | 1.0 | | | |
| 6Ba | Lapland Bunting | Bi | sys | 35 | 26.0 | 29.5 | 18.1 | 48.0 | 0.2 | 0.3 | 50.0 | 0.2 |
| 6Ba | Longbilled Dowitcher | Bi | all | 36 | 22.3 | 17.6 | 9.2 | 33.4 | 0.3 | 0.3 | 26.0 | 0.2 |
| 6Ba | Longbilled Dowitcher | Bi | ran | 36 | 30.0 | 5.9 | 0.0 | ##### # | 100.0 | 1.0 | 1.0 | 0.0 |
| 6Ba | Longbilled Dowitcher | Bi | sys | 36 | 22.2 | 19.0 | 9.7 | 37.4 | 0.3 | 0.3 | 23.0 | 0.2 |
| 6Ba | Pectoral Sandpiper | Bi | all | 37 | 30.1 | 6.3 | 2.4 | 16.8 | 0.5 | 0.4 | 16.0 | 0.2 |
| 6Ba | Pectoral Sandpiper | Bi | ran | 37 | | | | | | | | |
| 6Ba | Pectoral Sandpiper | Bi | sys | 37 | 30.1 | 7.6 | 2.9 | 20.0 | 0.5 | 0.4 | 16.0 | 0.2 |
| 6Ba | Pomarine Jaeger | Bi | all | 18 | 13.6 | 69.9 | 19.4 | 251.7 | 0.7 | 0.2 | 19.0 | 0.1 |
| 6Ba | Pomarine Jaeger | Bi | ran | 18 | | 117.9 | 38.6 | 359.9 | 0.5 | | | |

| Study area | Target narrative | Type | Data | Model definition | ESW/EDR | D | D LCL | D UCL | D CV | P | P LCL | P UCL |
|------------|------------------------|------|------|------------------|---------|---------|---------|----------|------|-----|-------|-------|
| 6Ba | Pomarine Jaeger | Bi | sys | 18 | 12.8 | 80.0 | 25.4 | 252.0 | 0.6 | 0.2 | 16.0 | 0.1 |
| 6Ba | Red Phalarope | Bi | all | 52 | 28.7 | 20.6 | 4.6 | 93.0 | 0.9 | 0.4 | 53.0 | 0.1 |
| 6Ba | Red Phalarope | Bi | ran | 52 | 40.0 | 6.6 | 1.7 | 26.5 | 0.7 | 1.0 | 5.0 | 0.3 |
| 6Ba | Red Phalarope | Bi | sys | 52 | 28.1 | 22.8 | 4.8 | 108.7 | 0.9 | 0.3 | 47.0 | 0.1 |
| 6Ba | Semipalmated Sandpiper | Bi | all | 36 | 19.8 | 20.8 | 11.1 | 38.9 | 0.3 | 0.4 | 24.0 | 0.3 |
| 6Ba | Semipalmated Sandpiper | Bi | ran | 36 | | 29.5 | 2.9 | 297.4 | 1.0 | | | |
| 6Ba | Semipalmated Sandpiper | Bi | sys | 36 | 20.0 | 23.6 | 12.5 | 44.4 | 0.3 | 0.4 | 23.0 | 0.3 |
| 2Ni | Butterfly, white | DT | all | 4 | 8.6 | 98777.7 | 64350.3 | 151624.0 | 0.2 | 0.6 | 30.0 | 0.5 |
| 2Ni | Butterfly, white | DT | ran | 4 | 15.0 | 58333.3 | 9184.3 | 370500.8 | 0.6 | 1.0 | 7.0 | 1.0 |
| 2Ni | Butterfly, white | DT | sys | 4 | 8.5 | 98341.1 | 58020.3 | 166682.8 | 0.3 | 0.6 | 23.0 | 0.4 |
| 1CR | Ant | TW | all | 32 | 1.3 | 2741.2 | 656.4 | 11448.0 | 0.5 | 0.4 | 113.0 | 0.4 |
| 1CR | Spider | TW | all | 32 | 1.2 | 702.5 | 402.9 | 1224.7 | 0.2 | 0.4 | 45.0 | 0.3 |
| 2Ni | Ant | TW | all | 23 | 1.0 | 2090.7 | 644.1 | 6786.6 | 0.5 | 0.2 | 55.0 | 0.2 |
| 2Ni | Ant, small red | TW | all | 2 | 1.0 | 695.1 | 288.8 | 1672.8 | 0.4 | 0.2 | 23.0 | 0.1 |
| 2Ni | Beetle, 868 | TW | all | 2 | 1.1 | 1042.8 | 205.6 | 5290.1 | 0.7 | 0.3 | 38.0 | 0.2 |
| 2Ni | Centipede, 881 | TW | all | 1 | 1.2 | 1322.9 | 699.7 | 2501.3 | 0.3 | 0.4 | 57.0 | 0.2 |
| 2Ni | Spider, small | TW | all | 3 | 1.1 | 271.0 | 170.1 | 431.9 | 0.2 | 0.3 | 20.0 | 0.2 |
| 2Ni | Springtail | TW | all | 3 | 1.2 | 47207.5 | 25774.5 | 86463.1 | 0.3 | 0.3 | 84.0 | 0.3 |
| 3AK | Spider | TW | all | 20 | 1.4 | 969.6 | 475.8 | 1975.9 | 0.3 | 0.5 | 89.0 | 0.4 |
| 3AK | Spider | TW | all | 71 | | | | | | | | |
| 4Ru | Cycsegusa | TW | all | 1 | 1.1 | 280.9 | 67.4 | 1170.5 | 0.6 | 0.3 | 15.0 | 0.1 |
| 4Ru | Protura | TW | all | 44 | 1.5 | 162.5 | 41.2 | 641.6 | 0.7 | 0.6 | 12.0 | 0.2 |
| 4Ru | Spider, little | TW | all | 2 | 1.6 | 169.2 | 49.5 | 578.1 | 0.6 | 0.6 | 16.0 | 0.2 |
| 5PG | Ant, tiny black | TW | all | 3 | 1.4 | 199.3 | 67.2 | 591.6 | 0.5 | 0.5 | 13.0 | 0.2 |
| 6Ba | Beetle, flat | TW | all | 2 | 1.1 | 1671.5 | 584.4 | 4780.6 | 0.4 | 0.3 | 50.0 | 0.2 |
| 6Ba | Fly | TW | all | 2 | 1.3 | 313.1 | 132.9 | 737.5 | 0.4 | 0.4 | 22.0 | 0.2 |
| 6Ba | Fruitfly | TW | all | 1 | 1.4 | 206.8 | 77.0 | 555.6 | 0.5 | 0.5 | 18.0 | 0.2 |
| 6Ba | Milbe | TW | all | 1 | 1.3 | 120.7 | 29.3 | 497.3 | 0.7 | 0.4 | 10.0 | 0.2 |
| 6Ba | Mosquito | TW | all | 2 | 2.0 | 114.1 | 19.0 | 685.7 | 1.0 | 1.0 | 8.0 | 0.2 |
| 6Ba | Schuster | TW | all | 2 | 1.1 | 220.3 | 92.9 | 522.3 | 0.4 | 0.3 | 13.0 | 0.1 |
| 6Ba | Spider | TW | all | 1 | 1.3 | 542.8 | 263.2 | 1119.5 | 0.3 | 0.4 | 43.0 | 0.2 |
| 6Ba | Spider, tiny | TW | all | 2 | 1.1 | 4131.6 | 1775.6 | 9613.9 | 0.3 | 0.3 | 88.0 | 0.2 |

7.9 Best Models (PRESENCE)

| Study area | Type | Target narrative | Data | Model | AIC | Likelihood | No. Par. | (-2*Log Like) |
|------------|------|------------------|------|----------------------------|--------|------------|----------|---------------|
| 1CR | Bi | Flycatcher | all | 1 group, Survey-specific P | 100.45 | 1 | 4 | 92.45 |
| 1CR | Bi | Hummingbird | all | Flowers | 106.51 | 1 | 3 | 100.51 |
| 1CR | Bi | Oropendula | all | TurkeyVulture | 124.15 | 1 | 3 | 118.15 |
| 1CR | Bi | Seedeater | all | Habitat | 58.01 | 1 | 7 | 44.01 |
| 1CR | Bi | Woodpecker | all | HighestTree | 96.77 | 1 | 3 | 90.77 |
| 1CR | Bi | Passeriformes | all | 1 group, Survey-specific P | 46.26 | 1 | 4 | 38.26 |
| 1CR | Bi | Piciformes | all | Min | 102.8 | 1 | 3 | 96.80 |
| 1CR | Bi | Psittaciformes | all | Min | 74.21 | 1 | 3 | 68.21 |
| 1CR | Bi | Thraupidae | all | Habitat | 54.89 | 1 | 7 | 40.89 |
| 1CR | Bi | Tyrannidae | all | 1 group, Survey-specific P | 116.26 | 1 | 4 | 108.26 |
| 1CR | Bi | Flycatcher | aur | 1 group, Survey-specific P | 94.31 | 1 | 4 | 86.31 |
| 1CR | Bi | Hummingbird | aur | DuffCover | 113.28 | 1 | 3 | 107.28 |
| 1CR | Bi | Oropendula | aur | HighestDBH | 104.62 | 1 | 3 | 98.62 |
| 1CR | Bi | Passeriformes | aur | LeafBrowsing | 114.08 | 1 | 3 | 108.08 |
| 1CR | Bi | Piciformes | aur | HighestDBH | 61.92 | 1 | 3 | 55.92 |
| 1CR | Bi | Psittaciformes | aur | Cov05 | 37.39 | 1 | 3 | 31.39 |
| 1CR | Bi | Seedeater | aur | HighestDBH | 26.11 | 1 | 3 | 20.11 |
| 1CR | Bi | Thraupidae | aur | HighestDBH | 26.11 | 1 | 3 | 20.11 |
| 1CR | Bi | Tyrannidae | aur | 1 group, Survey-specific P | 116.13 | 1 | 4 | 108.13 |
| 1CR | Bi | Woodpecker | aur | HighestDBH | 58.94 | 1 | 3 | 52.94 |
| 1CR | Bi | Flycatcher | ran | Shrubs | 18.37 | 1 | 3 | 12.37 |
| 1CR | Bi | Hummingbird | ran | Shrubs | 18.37 | 1 | 3 | 12.37 |
| 1CR | Bi | Oropendula | ran | 1 group, Constant P | 21.4 | 1 | 2 | 17.40 |
| 1CR | Bi | Passeriformes | ran | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 1CR | Bi | Piciformes | ran | DuffCover | 22.64 | 1 | 3 | 16.64 |
| 1CR | Bi | Psittaciformes | ran | Shrubs | 15.53 | 1 | 3 | 9.53 |
| 1CR | Bi | Seedeater | ran | BareSoil | 13.64 | 1 | 3 | 7.64 |
| 1CR | Bi | Thraupidae | ran | BareSoil | 13.64 | 1 | 3 | 7.64 |
| 1CR | Bi | Tyrannidae | ran | DuffCover | 22.64 | 1 | 3 | 16.64 |
| 1CR | Bi | Woodpecker | ran | Shrubs | 18.37 | 1 | 3 | 12.37 |
| 1CR | Bi | Flycatcher | sys | 1 group, Survey-specific P | 85.66 | 1 | 4 | 77.66 |
| 1CR | Bi | Hummingbird | sys | Flowers | 89.08 | 1 | 3 | 83.08 |
| 1CR | Bi | Oropendula | sys | TurkeyVulture | 105.71 | 1 | 3 | 99.71 |
| 1CR | Bi | Passeriformes | sys | 1 group, Survey-specific P | 43.92 | 1 | 4 | 35.92 |
| 1CR | Bi | Piciformes | sys | Min | 84.74 | 1 | 3 | 78.74 |
| 1CR | Bi | Psittaciformes | sys | Min | 62.72 | 1 | 3 | 56.72 |
| 1CR | Bi | Seedeater | sys | Min | 48.93 | 1 | 3 | 42.93 |
| 1CR | Bi | Thraupidae | sys | Min | 46.98 | 1 | 3 | 40.98 |
| 1CR | Bi | Tyrannidae | sys | 1 group, Survey-specific P | 93.93 | 1 | 4 | 85.93 |
| 1CR | Bi | Woodpecker | sys | Min | 79.47 | 1 | 3 | 73.47 |
| 1CR | Bi | Flycatcher | vis | MossLichen | 73.85 | 1 | 5 | 63.85 |
| 1CR | Bi | Hummingbird | vis | Flowers | 114.27 | 1 | 3 | 108.27 |
| 1CR | Bi | Oropendula | vis | LeafBrowsing | 115.39 | 1 | 3 | 109.39 |
| 1CR | Bi | Passeriformes | vis | CanopyTrees | 112.32 | 1 | 3 | 106.32 |
| 1CR | Bi | Piciformes | vis | Min | 88.66 | 1 | 3 | 82.66 |
| 1CR | Bi | Psittaciformes | vis | Cov12 | 58.81 | 1 | 3 | 52.81 |
| 1CR | Bi | Seedeater | vis | DuffCover | 43.39 | 1 | 3 | 37.39 |
| 1CR | Bi | Thraupidae | vis | DuffCover | 38.82 | 1 | 3 | 32.82 |
| 1CR | Bi | Tyrannidae | vis | MossLichen | 84.09 | 1 | 5 | 74.09 |

| Study area | Type | Target narrative | Data | Model | AIC | Likelihood | No. Par. | (-2*Log Like) |
|------------|------|---------------------------|------|----------------------------|--------|------------|----------|---------------|
| 1CR | Bi | Woodpecker | vis | HighestDBH | 72.66 | 1 | 3 | 66.66 |
| 1CR | TW | Ant | all | 1 group, Survey-specific P | 13.55 | 1 | 4 | 5.55 |
| 1CR | TW | Formicidae | all | 1 group, Survey-specific P | 13.55 | 1 | 4 | 5.55 |
| 1CR | TW | Spider | all | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 2Ni | Bi | Banded Wren | all | Habitat | 104.83 | 1 | 7 | 90.83 |
| 2Ni | Bi | Ciconiiformes | all | DuffCover | 102.14 | 0.831 | 3 | 96.14 |
| 2Ni | Bi | Passeriformes | all | 1 group, Survey-specific P | 93.1 | 1 | 4 | 85.10 |
| 2Ni | Bi | White-throated Magpie Jay | all | 1 group, Survey-specific P | 125.05 | 1 | 4 | 117.05 |
| 2Ni | Bi | Banded Wren | aur | Habitat | 104.83 | 1 | 7 | 90.83 |
| 2Ni | Bi | Ciconiiformes | aur | 1 group, Survey-specific P | 57.53 | 1 | 4 | 49.53 |
| 2Ni | Bi | Passeriformes | aur | 1 group, Survey-specific P | 117.99 | 1 | 4 | 109.99 |
| 2Ni | Bi | White-throated Magpie Jay | aur | 1 group, Survey-specific P | 106.75 | 1 | 4 | 98.75 |
| 2Ni | Bi | Banded Wren | ran | Shrubs | 18.37 | 1 | 3 | 12.37 |
| 2Ni | Bi | Ciconiiformes | ran | 1 group, Survey-specific P | 22 | 1 | 4 | 14.00 |
| 2Ni | Bi | Passeriformes | ran | 1 group, Constant P | 21.4 | 1 | 2 | 17.40 |
| 2Ni | Bi | White-throated Magpie Jay | ran | Shrubs | 22.3 | 1 | 3 | 16.30 |
| 2Ni | Bi | Banded Wren | sys | Habitat | 87.77 | 1 | 7 | 73.77 |
| 2Ni | Bi | Ciconiiformes | sys | DuffCover | 82.28 | 1 | 3 | 76.28 |
| 2Ni | Bi | Passeriformes | sys | 1 group, Survey-specific P | 77.57 | 1 | 4 | 69.57 |
| 2Ni | Bi | White-throated Magpie Jay | sys | Min | 103.8 | 1 | 3 | 97.80 |
| 2Ni | Bi | Banded Wren | vis | Understory | 16.81 | 1 | 3 | 10.81 |
| 2Ni | Bi | Ciconiiformes | vis | DuffCover | 83.05 | 0.045 | 3 | 77.05 |
| 2Ni | Bi | Passeriformes | vis | HighestTree | 125.38 | 1 | 3 | 119.38 |
| 2Ni | Bi | White-throated Magpie Jay | vis | HighestTree | 115.4 | 1 | 3 | 109.40 |
| 2Ni | TW | Acariformes | all | 1 group, Constant P | 10.88 | 1 | 2 | 6.88 |
| 2Ni | TW | Ant | all | 1 group, Constant P | 10.88 | 1 | 2 | 6.88 |
| 2Ni | TW | Ant, small red | all | 1 group, Survey-specific P | 19.09 | 1 | 4 | 11.09 |
| 2Ni | TW | Araneae | all | 1 group, Constant P | 14.81 | 1 | 2 | 10.81 |
| 2Ni | TW | Beetle, 868 | all | 1 group, Survey-specific P | 13.55 | 1 | 4 | 5.55 |
| 2Ni | TW | Centipede, 881 | all | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 2Ni | TW | Coleoptera | all | 1 group, Constant P | 10.88 | 1 | 2 | 6.88 |
| 2Ni | TW | Formicidae | all | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 2Ni | TW | Spider, small | all | 1 group, Constant P | 17.5 | 1 | 2 | 13.50 |
| 2Ni | TW | Springtail | all | 1 group, Survey-specific P | 8 | 1 | 4 | 0.00 |
| 3AK | Bi | Emberizidae | all | 1 group, Constant P | 128.73 | 1 | 2 | 124.73 |
| 3AK | Bi | Passeriformes | all | Squirrel | 80.79 | 1 | 3 | 74.79 |
| 3AK | Bi | Sparrow | all | Min | 126.6 | 1 | 3 | 120.60 |
| 3AK | Bi | Squirrel | all | DuffCover | 104.87 | 0.007 | 3 | 98.87 |
| 3AK | Bi | Emberizidae | ran | Cov13 | 22.64 | 1 | 3 | 16.64 |
| 3AK | Bi | Passeriformes | ran | 1 group, Survey-specific P | 14.73 | 1 | 4 | 6.73 |
| 3AK | Bi | Sparrow | ran | DuffCover | 18.37 | 1 | 3 | 12.37 |
| 3AK | Bi | Squirrel | ran | 1 group, Survey-specific P | 17.5 | 1 | 4 | 9.50 |
| 3AK | TW | Araneae | all | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 3AK | TW | Coleoptera | all | 1 group, Constant P | 19.28 | 1 | 2 | 15.28 |
| 3AK | TW | Spider | all | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 3AK | TW | Springtail | all | 1 group, Survey-specific P | 13.55 | 1 | 4 | 5.55 |
| 4Ru | Bi | Chickadee | all | Cov05 | 117.41 | 1 | 3 | 111.41 |

| Study area | Type | Target narrative | Data | Model | AIC | Likelihood | No. Par. | (-2*Log Like) |
|------------|------|------------------------|------|----------------------------|--------|------------|----------|---------------|
| 4Ru | Bi | Corvidae | all | Cov21 | 111.38 | 1 | 3 | 105.38 |
| 4Ru | Bi | Kinglet | all | HighDBH | 116.48 | 1 | 3 | 110.48 |
| 4Ru | Bi | Nutcracker | all | Cov21 | 94.34 | 1 | 3 | 88.34 |
| 4Ru | Bi | Paridae | all | Cov05 | 117.41 | 1 | 3 | 111.41 |
| 4Ru | Bi | Passeriformes | all | DuffCover | 44.62 | 1 | 3 | 38.62 |
| 4Ru | Bi | Warbler | all | 1 group, Survey-specific P | 110.77 | 1 | 4 | 102.77 |
| 4Ru | Bi | Winter Wren | all | Cov01 | 88.36 | 1 | 3 | 82.36 |
| 4Ru | Bi | Wize | all | 1 group, Survey-specific P | 86.15 | 1E-04 | 4 | 78.15 |
| 4Ru | TW | Araneae | all | 1 group, Constant P | 10.88 | 1 | 2 | 6.88 |
| 4Ru | TW | Collembola | all | 1 group, Constant P | 14.81 | 1 | 2 | 10.81 |
| 4Ru | TW | Cycsegusa | all | HighestTree | 17.46 | 1 | 3 | 11.46 |
| 4Ru | TW | Protura | all | Min | 18.37 | 1 | 3 | 12.37 |
| 4Ru | TW | Spider, little | all | 1 group, Survey-specific P | 17 | 1 | 4 | 9.00 |
| 5PG | Bi | Flute | all | CanopyPerc | 86.98 | 1 | 3 | 80.98 |
| 5PG | Bi | Paradisaeidae | all | 1 group, Survey-specific P | 74.43 | 1 | 4 | 66.43 |
| 5PG | Bi | Passeriformes | all | 1 group, Survey-specific P | 124.3 | 1 | 4 | 116.30 |
| 5PG | Bi | Psittaciformes | all | CanopyPerc | 96.87 | 1 | 3 | 90.87 |
| 5PG | Bi | Tsilp | all | 1 group, Survey-specific P | 114.31 | 1 | 4 | 106.31 |
| 5PG | TW | Ant, tiny black | all | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 5PG | TW | Collembola | all | 1 group, Constant P | 14.81 | 1 | 2 | 10.81 |
| 5PG | TW | Diptera | all | 1 group, Constant P | 17.5 | 1 | 2 | 13.50 |
| 5PG | TW | Formicidae | all | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 6Ba | Bi | Ciconiiformes | all | 1 group, Survey-specific P | 42.2 | 1 | 4 | 34.20 |
| 6Ba | Bi | Lapland Bunting | all | Cov03 | 101.1 | 1 | 3 | 95.10 |
| 6Ba | Bi | Longbilled Dowitcher | all | Cov03 | 117.68 | 1 | 3 | 111.68 |
| 6Ba | Bi | Passeriformes | all | Cov03 | 101.1 | 1 | 3 | 95.10 |
| 6Ba | Bi | Pectoral Sandpiper | all | DiamLake | 99.84 | 1 | 3 | 93.84 |
| 6Ba | Bi | Pomarine Jaeger | all | 1 group, Survey-specific P | 115.45 | 1 | 4 | 107.45 |
| 6Ba | Bi | Red Phalarope | all | GrassPerc | 122.82 | 1 | 3 | 116.82 |
| 6Ba | Bi | schuster | all | 1 group, Constant P | 14.81 | 1 | 2 | 10.81 |
| 6Ba | Bi | Scolopacidae | all | 1 group, Constant P | 57.99 | 1 | 2 | 53.99 |
| 6Ba | Bi | Semipalmated Sandpiper | all | 1 group, Constant P | 128.05 | 1 | 2 | 124.05 |
| 6Ba | Bi | Stercorariidae | all | 1 group, Survey-specific P | 119.02 | 1 | 4 | 111.02 |
| 6Ba | Bi | Ciconiiformes | ran | 1 group, Constant P | 19.01 | 1 | 2 | 15.01 |
| 6Ba | Bi | Lapland Bunting | ran | 1 group, Survey-specific P | 18.01 | 1 | 4 | 10.01 |
| 6Ba | Bi | Longbilled Dowitcher | ran | Cov06 | 21.28 | 1 | 3 | 15.28 |
| 6Ba | Bi | Passeriformes | ran | 1 group, Survey-specific P | 18.01 | 1 | 4 | 10.01 |
| 6Ba | Bi | Pectoral Sandpiper | ran | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 6Ba | Bi | Pomarine Jaeger | ran | 1 group, Survey-specific P | 14.73 | 1 | 4 | 6.73 |
| 6Ba | Bi | Red Phalarope | ran | 1 group, Survey-specific P | 17.5 | 1 | 4 | 9.50 |
| 6Ba | Bi | Scolopacidae | ran | 1 group, Constant P | 19.01 | 1 | 2 | 15.01 |
| 6Ba | Bi | Semipalmated Sandpiper | ran | 1 group, Constant P | 23.1 | 1 | 2 | 19.10 |
| 6Ba | Bi | Stercorariidae | ran | 1 group, Survey-specific P | 14.73 | 1 | 4 | 6.73 |
| 6Ba | Bi | Ciconiiformes | sys | 1 group, Survey-specific P | 21.94 | 1 | 4 | 13.94 |
| 6Ba | Bi | Lapland Bunting | sys | Cov03 | 80.72 | 1 | 3 | 74.72 |

| Study area | Type | Target narrative | Data | Model | AIC | Likelihood | No. Par. | (-2*Log Like) |
|------------|------|------------------------|------|----------------------------|--------|------------|----------|---------------|
| 6Ba | Bi | Longbilled Dowitcher | sys | Flowers | 96.19 | 1 | 3 | 90.19 |
| 6Ba | Bi | Passeriformes | sys | Cov03 | 80.72 | 1 | 3 | 74.72 |
| 6Ba | Bi | Pectoral Sandpiper | sys | 1 group, Constant P | 98.16 | 1 | 2 | 94.16 |
| 6Ba | Bi | Pomarine Jaeger | sys | Flowers | 90.99 | 0.937 | 3 | 84.99 |
| 6Ba | Bi | Red Phalarope | sys | GrassPerc | 100.56 | 1 | 3 | 94.56 |
| 6Ba | Bi | Scolopacidae | sys | 1 group, Constant P | 40.74 | 1 | 2 | 36.74 |
| 6Ba | Bi | Semipalmated Sandpiper | sys | 1 group, Constant P | 107.85 | 1 | 2 | 103.85 |
| 6Ba | Bi | Stercorariidae | sys | Flowers | 96 | 0.705 | 3 | 90.00 |
| 6Ba | TW | Acariformes | all | 1 group, Constant P | 17.5 | 1 | 2 | 13.50 |
| 6Ba | TW | Araneae | all | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 6Ba | TW | beetle, flat | all | 1 group, Constant P | 10.88 | 1 | 2 | 6.88 |
| 6Ba | TW | Coleoptera | all | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 6Ba | TW | Diptera | all | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 6Ba | TW | fly | all | 1 group, Constant P | 4 | 1 | 2 | 0.00 |
| 6Ba | TW | fruitfly | all | 1 group, Constant P | 17.5 | 1 | 2 | 13.50 |
| 6Ba | TW | Milbe | all | 1 group, Survey-specific P | 16.32 | 1 | 4 | 8.32 |
| 6Ba | TW | mosquito | all | 1 group, Survey-specific P | 17 | 1 | 4 | 9.00 |
| 6Ba | TW | spider | all | 1 group, Constant P | 10.88 | 1 | 2 | 6.88 |
| 6Ba | TW | spider, tiny | all | 1 group, Survey-specific P | 13.55 | 1 | 4 | 5.55 |
| 6Ba | TW | Tipulidae | all | 1 group, Constant P | 10.88 | 1 | 2 | 6.88 |

8 Declaration

Hiermit versichere ich gemäß § 9 Abs. 5 der Prüfungsordnung für den integrierten binationalen Master-Studiengang Internationaler Naturschutz (engl.: International Nature Conservation) vom 16.08.2006, dass ich die vorliegende Arbeit selbstständig verfasst und keine anderen als die angegebenen Hilfsmittel verwendet habe. Diese Arbeit wurde nicht in der gleichen oder einer ähnlichen Form bereits einem anderen Prüfungsausschuss vorgelegt und wurde bisher noch nicht veröffentlicht.

Hereby I affirm – according to § 9 section 5 of the examination regulations for the integrated bi-national Master programme International Nature Conservation (deutsch: Internationaler Naturschutz) from 16.08.2006 – that I have penned the present thesis autonomously and that I did not use any other resources than those specified above. This work was not submitted previously in same or similar form to another examination committee and was not yet published.

Göttingen, 05 December 2008 _____

Ort/Place, Datum/Date



Name/Name