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**Habitat selection in transformed landscapes and the role of forest remnants and shade coffee in the conservation of resident birds**

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## **Habitat selection in transformed landscapes and the role of forest remnants and shade coffee in the conservation of resident birds**

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1    **Abstract**

2

3    1. Biodiversity conservation in transformed landscapes is becoming increasingly important.  
4    However, most assessments of the value of modified habitats rely heavily on species presence  
5    and/or abundance, masking ecological processes such as habitat selection and phenomena like  
6    ecological traps, which may render species persistence uncertain. High species richness has  
7    been documented in tropical agroforestry systems but comparisons with native habitat  
8    remnants generally lack detailed information on species demography and habitat use.

9

10    2. We generated a multi-species, multi-measure framework to evaluate the role of habitat  
11    selection in the adaptation of species to transformed landscapes, and demonstrate that its use  
12    could affect how we value the contribution different land uses make to biodiversity  
13    conservation.

14

15    3. We analyzed seven years of capture-mark-recapture and observation data for twelve species of  
16    resident birds present in native forest remnants and shade coffee plantations in a mega-diverse  
17    region. We assessed whether species behaved adaptively by evaluating the correlation between  
18    measures of habitat preference (occurrence, abundance, fidelity, inter-seasonal variance and  
19    age) and performance (body condition, muscle, primary molt, breeding and juveniles) in forest  
20    and coffee, and generated hypotheses about their role in species persistence.

21

22    4. We documented adaptive habitat selection for seven species, non-ideal selection for four, and  
23    maladaptive selection for one. While many species showed equal-preference and/or equal  
24    performance in many traits, in general we found more evidence for birds preferring and/or

25 performing better in forest than coffee, although relationships between our indicators and  
26 population adaptation need to be studied further before our proposed framework can be  
27 applied to more species and landscapes.

28

29 5. While shade coffee can act as a biodiversity-friendly matrix providing complementary or  
30 supplementary habitat to a wide range of resident bird species, protecting remnants of native  
31 vegetation is still of paramount importance for biodiversity conservation in agricultural  
32 landscapes.

33

34 **Keywords:** agroforestry systems; ecological traps; habitat preference; habitat performance;  
35 habitat selection; novel ecosystems; source and sink; undervalued resources.

36

37 **Selección de hábitat en paisajes transformados y el papel de los remanentes de bosque y**  
38 **cafetales con sombra en la conservación de aves residentes.**

39

## 40 **Resumen**

41

42 1. La conservación de biodiversidad en paisajes transformados es cada vez más importante. Sin  
43 embargo, la mayoría de evaluaciones sobre el valor de los hábitats modificados dependen de la  
44 presencia y/o abundancia de especies, ocultando procesos ecológicos como la selección de hábitat  
45 y fenómenos como las trampas ecológicas, que hacen incierta la persistencia de las especies. Se  
46 han documentado altas riquezas en los sistemas agroforestales tropicales pero las comparaciones  
47 con remanentes de hábitats nativos generalmente no proveen información detallada sobre  
48 demografía y uso de hábitat.

49  
50 2. Generamos un marco metodológico multi-especies y multi-medidas para evaluar el papel de la  
51 selección de hábitat en la adaptación de las especies a los paisajes transformados, y demostramos  
52 que su aplicación podría afectar la forma cómo valoramos la contribución que diferentes usos del  
53 paisaje hacen a la conservación de la biodiversidad.

54  
55 3. Analizamos siete años de datos de captura-marca-recaptura y observaciones para doce especies  
56 de aves residentes presentes en remanentes de bosque nativo y plantaciones de café con sombra  
57 en una región megadiversa. Valoramos si las especies se comportaban de forma adaptativa  
58 evaluando la correlación entre medidas de preferencia (ocurrencia, abundancia, fidelidad,  
59 varianza inter-estacional y edad) y desempeño (condición corporal, músculo, muda primaria,  
60 reproducción y juveniles) en bosques y cafetales, y generamos hipótesis acerca de su papel en la  
61 persistencia de especies.

62  
63 4. Documentamos selección de hábitat adaptativa para siete especies, no-ideal para cuatro y  
64 maladaptativa para una. Aunque muchas especies mostraron preferencias y/o desempeño iguales  
65 en muchos rasgos, en general encontramos más evidencia de aves prefiriendo y/o  
66 desempeñándose mejor en bosques que en cafetales, aunque las relaciones entre nuestros  
67 indicadores y la adaptación poblacional deben ser más estudiadas antes de aplicar el marco que  
68 proponemos a más especies y paisajes.

69  
70 5. Aunque los cafetales con sombra pueden actuar como una matriz amigable con la  
71 biodiversidad, proveyendo hábitat complementario o suplementario a una gran variedad de aves

72 residentes, para asegurar la conservación de biodiversidad en los paisajes agrícolas es  
73 supremamente importante proteger los remanentes de vegetación nativa.

74

75 **Introduction**

76

77 Presently, the majority of the world's species inhabit heterogeneous landscapes combining native  
78 habitat remnants with novel ecosystems varying in their degree of intervention (Johnson et al.  
79 1992; Hobbs et al. 2006). As human population and resource consumption continue to grow,  
80 conservation focus has expanded from the protection of native remnants to include land uses with  
81 intermediate levels of transformation (Daily et al. 2001; Norris 2008). Shade coffee plantations  
82 have been promoted as a biodiversity-friendly production system that can harbor communities  
83 intermediate between those of sun coffee monocultures and remnants of native forest (Perfecto et  
84 al. 1996; Philpott et al. 2008; Jha et al. 2014), so financial tools and social incentives have arisen  
85 to prevent their replacement by homogeneous land uses (Perfecto et al. 2005; Philpott et al.  
86 2007). However, most assessments of the conservation value of tropical agroforestry systems rely  
87 heavily on community-level measures (Hughes et al. 2002; Petit & Petit 2003), and while there is  
88 recent evidence on differences for migrant birds (Bakermans et al. 2009; Bayly et al. 2016),  
89 information on demography and persistence of resident species is still scarce (Komar 2006;  
90 Sánchez-Clavijo et al. 2008; Gleffe et al. 2006; Irizarry et al. 2018).

91

92 When species presence is used to compare the relative conservation value of habitats, the implicit  
93 assumption is that species show neither preference for any of the habitats occupied, nor  
94 differences in performance once they occupy them (Fig. 1e); subsequently the availability of one  
95 habitat could theoretically compensate for the absence of another. While this may sometimes be

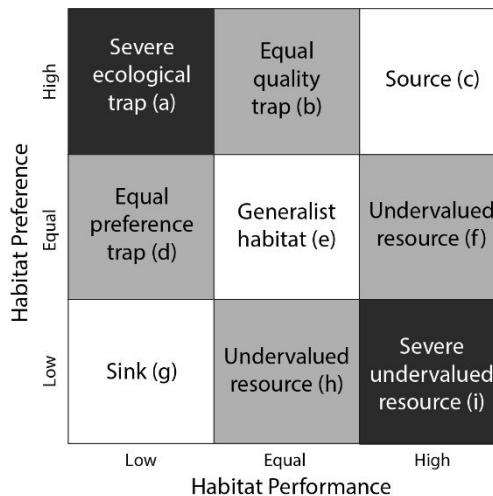
96 the case, assuming this pattern as a rule may lead to inaccurate assessments of population  
97 persistence that cannot be corrected by including abundance (Van Horne 1983; Jones 2001;  
98 Johnson 2007), for example where novel habitats are used in complementary or supplementary  
99 ways by populations that still depend on native remnants for critical resources (Dunning et al.  
100 1992). In contrast, including habitat-specific demography in comparisons allows for the explicit  
101 recognition that species performance varies between habitats and that given adaptive selection  
102 behaviors (Pulliam 1988), individuals prefer to settle in higher-quality habitats (sources – Fig. 1c)  
103 and avoid settling in lower-quality habitats (sinks – Fig. 1g). Under source-sink dynamics, the  
104 best strategy to ensure species persistence in a landscape is to prioritize the conservation of  
105 sources within networks of well-connected sinks (Pulliam & Danielson 1991; Furrer & Pasinelli  
106 2016; Gilroy & Edwards 2017).

107  
108 Measuring habitat preference and performance in a multi-species framework is not  
109 straightforward, especially for species with poorly known natural history (Garshelis 2000; Part et  
110 al. 2007). To further complicate the understanding of species-habitat relationships and their  
111 application to decision-making in conservation, rapid and widespread landscape change may lead  
112 to the uncoupling of cues used for habitat selection from habitat quality, resulting in maladaptive  
113 choices where individuals prefer lower-quality habitats (ecological traps *sensu* Schlaepfer et al.  
114 2002 - Fig. 1a) and avoid higher-quality options (undervalued resources *sensu* Gilroy &  
115 Sutherland 2007 - Fig. 1i). Novel ecosystems may share selection cues with the habitats they  
116 replaced, while offering a different set of resources and hazards that reduce overall fitness  
117 (Schlaepfer et al. 2002; Fletcher et al. 2012; Robertson et al. 2013). As a result, species presence  
118 in novel habitats may mask maladaptive processes, making populations more vulnerable to  
119 further landscape change.

120

121 Source-sink and ecological trap-undervalued resource dichotomies are extreme cases in a  
 122 continuum of possible outcomes from the interaction between habitat availability, selection and  
 123 quality (Battin 2004), and ideally we would evaluate mismatches between preference and  
 124 performance as continuous variables (Kristan 2003). A compromise is to include outcomes  
 125 resulting from equal-preference and equal-performance. Robertson and Hutto (2006)  
 126 distinguished severe ecological traps from equal-preference traps, (Fig. 1d) which while still a  
 127 result of non-ideal selection, are expected to have less severe consequences at the population  
 128 level. Filling in the gaps from these outcomes allows for the recognition of three additional cases  
 129 of mild, non-ideal selection (Fig. 1b, 1f, 1h) that lead to consequences intermediate between  
 130 those of adaptive (Fig. 1c, 1e, 1g) and maladaptive (Fig. 1a, 1i) behaviors.

131



132

133 **Figure 1.** Outcomes when comparing preference and performance for a species using two  
 134 habitats (white: adaptive, grey: non-ideal, black: maladaptive habitat selection).

135

136 We used this extended framework (Fig. 1) to evaluate the correlation between measures of habitat  
137 preference (occurrence, abundance, site fidelity, temporal variance in population size, and  
138 distribution of dominant individuals) and performance (body condition, muscle scores, primary  
139 plumage molts, breeding evidence and frequency of juveniles) for twelve relatively common  
140 resident birds, inhabiting pre-montane forest remnants and shade coffee plantations at mid-  
141 elevations in the Sierra Nevada de Santa Marta, Colombia. Because our focal species are  
142 associated with habitats representing a gradient of tree cover, we expected marked variation in  
143 the outcomes resulting from the interaction of their habitat preference and performance. Our use  
144 of field data to generate hypotheses about the role that native forest and shade coffee play in the  
145 conservation of resident birds serves as a demonstration of a multi-species approach to compare  
146 the contribution of different land-uses to biodiversity conservation in a mega-diverse region.

147

## 148 **Materials and Methods**

149

### 150 *Data Collection*

151

152 Study area - Data was collected in Hacienda La Victoria (Magdalena, Colombia, 11°7'19.84"N,  
153 74°5'34.14"W), chosen for this study because: 1) coffee production in the region started in the  
154 late 1800's and has always taken place under shade trees as commercial polycultures (Moguel &  
155 Toledo 1999) – creating a novel habitat with intermediate levels of transformation that species  
156 have had time to adapt to; 2) the watershed where the farm is situated retains 47% native forest  
157 cover between 600 and 1,700 meters above sea level (Bayly et al. 2012) – meaning there is still  
158 substantial native habitat available; and 3) it is located in a global hotspot for biodiversity and  
159 endemism (Cracraft 1985; Myers et al. 2000), experiencing exponential growth in nature-related

160 tourism (Lara, et al. 2017; Ocampo-Peñuela & Winton 2017) – consequently, assessing the  
161 conservation potential of this productive system is of environmental and economic importance  
162 (see Supporting Information for further details on study area).

163

164 Sampling scheme - Between 2009 and 2015, we set up 17 banding stations, eight in pre-montane  
165 forest remnants and nine within shade coffee plots, at altitudes ranging from 900 to 1,300 m (Fig.  
166 S1). Banding stations consisted of five to ten 12 m mist nets, installed at each site for 5 to 30  
167 days, and operated daily for approximately 6 hours starting at sunrise, following guidelines for  
168 the safe and ethical treatment of animals (Fair et al. 2010). Sampling occurred between mid-  
169 March and mid-November, which corresponds to a gradient of increasing precipitation and  
170 coincides with the onset of breeding for the majority of resident birds, while avoiding sampling  
171 during annual coffee harvests. Effort was concentrated in capturing, marking and recapturing  
172 individuals, but complementary observations were carried out around banding stations. We used  
173 standardized protocols (Ralph et al. 1993; Wolfe et al. 2010) to measure wing chord and body  
174 mass, and assess age class, muscle score, breeding and molting activity of captured individuals.  
175 Variation in sampling effort among sites and seasons was taken into account during data analysis.

176

177 Focal species - 214 resident bird species have been recorded in La Victoria (Bayly & Gómez  
178 2013) but to guarantee adequate sample sizes to assess preference/performance, we required  
179 species that were: frequently detected in both habitats, regularly captured in mist-nets, and  
180 represented a gradient of sensitivity to forest loss. Based on these criteria we selected *Mionectes*  
181 *olivaceus* (Olive-striped Flycatcher), *Mionectes oleagineus* (Ochre-bellied Flycatcher), *Turdus*  
182 *flavipes* (Yellow-legged Thrush), *Turdus albicollis* (White-necked Thrush), *Ramphocelus*

183 *dimidiatus* (Crimson-backed Tanager), *Tangara gyrola* (Bay-headed Tanager; endemic  
184 subspecies), *Saltator maximus* (Buff-throated Saltator), *Saltator striatipectus* (Streaked Saltator),  
185 *Myiothlypis conspicillata* (White-lored Warbler; endemic species), *Basileuterus rufifrons*  
186 (Rufous-capped Warbler), *Myioborus miniatus* (Slate-throated Redstart), and *Euphonia*  
187 *laniirostris* (Thick-billed Euphonia) (Table S1).

188

189 While these species are not representative of the original mountain forest avifauna of the Sierra  
190 Nevada de Santa Marta, they are representative of bird communities in Colombian coffee-  
191 growing regions (Sánchez-Clavijo et al. 2009). Consequently, conclusions from this study do not  
192 extend to highly-sensitive species known to be forest-dependent for their conservation, or open-  
193 area species that exist only in deforested areas, but rather the ones in between that potentially use  
194 shade coffee plantations as suitable habitat.

195

196 *Data Processing*

197

198 We screened the banding database (4,108 captures of 3,214 individuals) to ensure data were  
199 consistent within species and individual capture histories (e.g. that individuals followed logical  
200 age and plumage sequences, or sex and breeding classifications). Inconsistencies were either  
201 resolved based on information from captures of the same individual or changed to unknowns (see  
202 Supporting Information for further details on data processing).

203

204 *Data Analysis*

205

206 We carried out analyses in three levels: using statistical tests to generate separate indicators of  
207 habitat preference and performance; weighting the evidence from level 1 to generate composite  
208 indexes of habitat preference and performance; and contrasting the indexes from level 2 to  
209 evaluate whether habitat selection was acting adaptively in our study system (see Supporting  
210 Information for further details on data analysis).

211

212 Level 1: Separate indicators of habitat preference and performance - We used multiple measures  
213 of preference and performance, under the assumption that more indicators pointing in the same  
214 direction would give us greater confidence in our assessments, but that contradictory patterns  
215 were a warning that more evidence was necessary. We designed five statistical tests to evaluate  
216 which habitat was preferred by each species (hereafter occurrence, abundance, fidelity, inter-  
217 seasonal variance, and age), and five to evaluate in which habitat individuals experienced higher  
218 average performance (body condition, muscle, primary molt, breeding and juveniles - Table 1).

219

220 **Occurrence:** the most commonly used indicator for preference is how often a species is detected  
221 in a habitat, but because higher frequencies can result from either more individuals in a habitat or  
222 a higher tenacity of individuals towards the sites they occupy (Garshelis 2000), occurrence  
223 represents an emergent attribute of different demographic properties that can have contradictory  
224 preference interpretations. Occupancy models take into account differences in detection  
225 probabilities to yield corrected occurrence probabilities in a cost-effective manner (Ruiz-  
226 Gutiérrez et al. 2010; Kéry & Schaub 2012). We used observation data to implement a Bayesian  
227 occupancy model with the effect of habitat (forest/coffee) on the probability of detection ( $p$ ), and  
228 effects of habitat and altitude on the probability of occurrence ( $\psi$ ).

229

230 **Abundance:** in the absence of strong dominance hierarchies and differences in habitat  
231 availability, we would expect more individuals to settle in their preferred habitat over other  
232 available habitats (Pulliam & Danielson 1991), rendering true abundance a good proxy for  
233 preferential choice. Calculating precise estimates of abundance is a field-intensive and data-  
234 hungry procedure, especially for highly-mobile animals with low probabilities of detection (Ruiz-  
235 Gutiérrez et al. 2012). Unlike observation data, capture-mark-recapture data allows for the  
236 recognition of individuals within a species, and is therefore well suited to calculate abundance in  
237 relatively small, intensively-sampled areas. To estimate the total number of individuals present in  
238 our study area across the entire sampling period, we used mist-net data to run capture-recapture  
239 analyses; using closed population models for species with low recapture rates and Jolly-Seber  
240 models for species with higher recapture rates - Royle & Dorazio 2008; Kéry & Schaub 2012).  
241 We ran separate analyses for each species: habitat combination, included the effect of sampling  
242 effort (standardized net hours) on detection probability (**p**) and random variation in apparent  
243 survival (**Φ**) by occasion.

244

245 **Fidelity:** individuals of territorial species are expected to show higher site fidelity when  
246 occupying their preferred habitat relative to less preferred ones (Robertson & Hutto 2006).  
247 Capture-recapture models allow for the estimation of apparent survival (**Φ**), which is a compound  
248 measure of site fidelity and true survival (Kéry & Schaub 2012). We used this parameter as a  
249 surrogate for fidelity because the contribution of mortality to the parameter was expected to be  
250 minimal given the short time between sampling occasions (1-6 months) and the known  
251 longevities of tropical birds (Ruiz-Gutiérrez et al. 2012). For species with high recapture rates, **Φ**

252 per occasion was jointly-estimated in Jolly-Seber models, for species with low recapture rates we  
253 used Cormack-Jolly-Seber models with a similar parameterization (Royle & Dorazio 2008; Kéry  
254 & Schaub 2012). Resulting parameters were then used to calculate mean site fidelity for each  
255 iteration of the model.

256

257 **Inter-seasonal variance:** because species may use different habitats according to life stage or  
258 seasonal variation in resources, measuring abundance during different times of year may yield  
259 different interpretations about habitat preference in places with high seasonality (van Horne  
260 1983; Skagen & Yackel Adams 2010). We assumed a habitat with less inter-seasonal variation in  
261 abundance is experiencing lower individual turnover which can be interpreted as a sign of  
262 preference (Robertson & Hutto 2006). Since Jolly-Seber models allow for the estimation of the  
263 number of individuals present per occasion, we used this output to calculate the coefficient of  
264 variance for each iteration of the model for available species.

265

266 **Age:** abundance may not be a true indicator of habitat preference when individuals follow a  
267 despotic distribution where dominant individuals settle in preferred habitats and displace a larger  
268 number of submissive individuals to less-preferred ones (Fretwell & Lucas 1970). Although  
269 dominance hierarchies are not well described in Neotropical passerines, we assumed that adult  
270 birds would displace immature individuals from their preferred habitat (as shown for migratory  
271 birds in Marra 1993 & Stutchbury 1994). Therefore, a habitat with a significantly higher  
272 proportion of adults could be considered as preferred. To test for segregation according to age,  
273 we calculated the model-averaged coefficient for the effect of habitat on the probability of  
274 capturing an adult over an immature individual, in a model set that included a null model, model

275 with a non-linear effect of day of the year, model with effect of habitat, and model with the  
 276 additive effects of day and habitat.

277

278 **Table 1.** Analyses used to generate indicators of habitat preference (rows 1-5) and performance  
 279 (rows 6-10) for twelve resident birds in native forest and shade coffee (GLM: Generalized linear  
 280 model; AIC: Akaike's information criterion).

<b>Indicator</b>	<b>Input</b>	<b>Analysis</b>	<b>Output</b>	<b>Interpretation</b>
<b>Occurrence</b>	Detection/non-detection matrix (9 sites x 16 occasions)	Hierarchical occupancy model; probability of occurrence ( $\psi$ )	Habitat coefficient (95% credibility intervals)	95% CI did not contain 0
<b>Abundance</b>	Capture/non-capture matrix (individuals x 12 occasions)	Capture-recapture model; total number of individuals (Nsuper)	Posterior distribution for total number of individuals (per habitat)	No overlap in 50% CI
<b>Fidelity</b>	Capture/non-capture matrix (individuals x 12 occasions)	Capture-recapture model; mean apparent survival ( $\phi$ )	Mean and standard deviation of iterations (per habitat)	No overlap in mean +/- SD
<b>Inter-Seasonal Variance</b>	Estimated number of individuals per occasion (Nocc)	Coefficient of variation for (Nocc)	Mean and standard deviation of iterations (per habitat)	No overlap in mean +/- SD
<b>Age</b>	Number of adults vs. immatures	GLM (binomial/logit), AICc model averaging	Habitat coefficient (95% confidence intervals)	95% CI did not contain 0
<b>Body Condition</b>	Vector of body condition index for all captures	GLM (normal/identity), AICc model averaging	Habitat coefficient (95% confidence intervals)	95% CI did not contain 0
<b>Muscle</b>	Number of records with muscle score 3 vs. muscle score 2	GLM (binomial/logit), AICc model averaging	Habitat coefficient (95% confidence intervals)	95% CI did not contain 0
<b>Primary Molt</b>	Number of records with active vs. inactive primary molt	GLM (binomial/logit), AICc model averaging	Habitat coefficient (95% confidence intervals)	95% CI did not contain 0
<b>Breeding</b>	Number of records with active vs. inactive breeding	GLM (binomial/logit), AICc model averaging	Habitat coefficient (95% confidence intervals)	95% CI did not contain 0
<b>Juveniles</b>	Number of juveniles vs. immatures and adults	GLM (binomial/logit), AICc model averaging	Habitat coefficient (95% confidence intervals)	95% CI did not contain 0

281

282 **Body condition:** body mass in birds is expected to reflect overall condition under most  
 283 circumstances, after being corrected by body size (Schulte-Hostedde et al. 2005; Schamber et al.  
 284 2009). We used the residuals of a standard major axis regression between wing chord length and

285 body mass to calculate a scaled index of body condition (Peig & Green 2009) and assumed that  
286 higher condition meant better performance in a given habitat. We calculated the model-averaged  
287 coefficient for the effect of habitat on the index with the same model set we used for age  
288 (changing error structure and link as shown in Table 1). The same statistical procedure using  
289 GLMs and model averaging was used for all habitat performance variables.

290

291 **Muscle:** breast muscle condition is important to species performance; as lower scores indicate  
292 reduced flight capacities from competing energy needs (Lindstrom et al. 2000). Although this  
293 variable is typically used to assess condition in migratory birds, variation in residents could be  
294 related to the amount and/or quality of resources in a habitat. We interpreted higher probabilities  
295 of capturing individuals with a muscle score of 3 (the highest in our classification) as an  
296 indication of better performance in a habitat (healthy immature and adult tropical residents rarely  
297 have scores below 2).

298

299 **Primary molt:** birds undergoing primary feather molts typically associated with annual complete  
300 molts have high energy demands and are compromised in their ability to fly, consequently this  
301 activity is usually undertaken when and where resources are high (Echeverry-Galvis & Córdoba-  
302 Córdoba 2008, Echeverry-Galvis & Hau 2013). We interpret higher probabilities of capturing  
303 individuals undergoing active primary molts as an indication of better performance in a habitat.

304

305 **Breeding:** since breeding-related activities like incubation and food provisioning are energy  
306 demanding, their timing and location are typically constrained by resource availability  
307 (Echeverry-Galvis & Córdoba-Córdoba 2008, Echeverry-Galvis & Hau 2013). We interpret a

308 higher probability of capturing individuals with evidence of active breeding as an indication of  
309 better performance in a habitat.

310

311 **Juveniles:** higher reproductive performance by a species in a habitat will result in higher  
312 proportions of recently-fledged juvenile birds, assuming that their dispersal is limited by  
313 dependence on adults and reduced movement capacities (Cox et al. 2014). We interpret higher  
314 probabilities of capturing juveniles as an indication of better performance in a habitat.

315

316 Level 2: Composite indexes of habitat preference and performance - Because our five tests of  
317 performance used the same statistical method and model set, we generated a quantitative estimate  
318 of differences between habitats for each species by weighting mean effect sizes by their standard  
319 deviation, using a fixed-effects model in a meta-analysis. Species were classified as having  
320 higher performance in forest when the 90% confidence intervals of this estimate were above zero,  
321 as having higher performance in coffee when below zero, and as having equal performance when  
322 they included zero.

323

324 The same method was not available for calculating a composite index of habitat preference  
325 because for two of our indicators the output was a habitat coefficient, while the remaining three  
326 relied on comparisons of habitat-specific parameters. Additionally, we were unable to identify a  
327 clear method for estimating the standard error around measures coming from iterative procedures.  
328 We calculated a preference score by subtracting the number of times evidence showed coffee was  
329 preferred from the number of times forest was preferred (response range -5 to +5). Birds with  
330 positive scores were classified as preferring forest, negative scores as preferring coffee, and zero  
331 as having equal-preference.

332

333 Level 3: Is habitat selection acting adaptively? - Our different approaches to create composite  
334 indexes of habitat preference and performance were not quantitatively comparable. Therefore, we  
335 qualitatively compared each species assigned habitat preference and performance, and used the  
336 framework in Figure 1 to generate hypotheses about the role of shade coffee and native forests.

337

## 338 Results

339

340 We were able to perform 88% of the planned comparisons (52/60 for preference and 54/60 for  
341 performance); omitting those without an adequate sample size for certain species/habitat/response  
342 combinations (which appear as NAs). Outcomes gave us multiple lines of evidence to classify the  
343 twelve species according to preference and performance, and therefore to assess the adaptiveness  
344 of habitat choices (see Supporting Information for further details on sample sizes and data  
345 outputs).

346

347 *Level 1: Separate indicators of habitat preference and performance*

348

349 Of the tests performed for preference, 33 showed statistically unclear (Dushoff et al. 2019)  
350 differences between habitats, 12 favored forest and 7 coffee. Abundance was the most  
351 informative indicator (8 clear differences), followed by fidelity (6), inter-seasonal variance (2),  
352 age (2), and occurrence (1) (Table 2).

353

354 **Table 2.** Indicators of habitat preference for twelve resident birds in native forest and shade  
355 coffee (HC: habitat coefficient (standard error), C: parameter estimate for coffee (standard

356 deviation), F: parameter estimate for forest (standard deviation), Int.: interpretation i.e. preferred  
 357 habitat according to that indicator; coffee was the reference for occurrence while forest was the  
 358 reference for age).

359

Species	Occurrence		Abundance			Fidelity			Inter-seasonal variance			Age	
	HC	Int.	C	F	Int.	C	F	Int.	C	F	Int.	HC	Int.
<i>Mionectes olivaceus</i>	4.72 (8.01)	Equal	1257 (571)	1173 (91)	Equal	0.49 (0.08)	0.66 (0.03)	Forest	NA	NA	-	NA	-
<i>Mionectes oleagineus</i>	-5.88 (6.89)	Equal	253 (65)	674 (83)	Forest	0.68 (0.05)	0.71 (0.03)	Equal	0.51 (0.10)	0.41 (0.06)	Equal	NA	-
<i>Turdus flavipes</i>	4.70 (7.97)	Equal	766 (149)	1239 (153)	Forest	0.63 (0.06)	0.70 (0.04)	Equal	NA	NA	-	-0.40 (0.20)	Coffee
<i>Turdus albicollis</i>	1.44 (8.48)	Equal	NA (41)	482 (41)	Forest	0.50 (0.09)	0.71 (0.03)	Forest	NA	NA	-	1.29 (0.50)	Forest
<i>Ramphocelus dimidiatus</i>	4.55 (8.03)	Equal	613 (165)	329 (64)	Coffee	0.65 (0.04)	0.74 (0.04)	Forest	0.68 (0.12)	0.62 (0.09)	Equal	-0.07 (0.29)	Equal
<i>Tangara gyrola</i>	4.80 (7.93)	Equal	558 (105)	591 (48)	Equal	0.64 (0.04)	0.73 (0.02)	Forest	1.01 (0.13)	0.46 (0.04)	Forest	0.29 (0.20)	Equal
<i>Saltator maximus</i>	4.74 (7.83)	Equal	208 (48)	165 (31)	Equal	0.62 (0.06)	0.68 (0.05)	Equal	NA	NA	-	0.00 (0.31)	Equal
<i>Saltator striatipectus</i>	4.66 (7.96)	Equal	235 (82)	NA (82)	Coffee	0.68 (0.06)	0.54 (0.08)	Coffee	NA	NA	-	-0.87 (0.51)	Equal
<i>Myiothlypis conspicillata</i>	4.64 (7.71)	Equal	317 (97)	604 (100)	Forest	0.65 (0.05)	0.68 (0.04)	Equal	0.70 (0.13)	0.71 (0.08)	Equal	-0.39 (0.26)	Equal
<i>Basileuterus rufifrons</i>	4.71 (7.69)	Equal	481 (94)	319 (115)	Coffee	0.67 (0.04)	0.64 (0.05)	Equal	0.79 (0.10)	0.53 (0.14)	Forest	-0.26 (0.26)	Equal
<i>Myioborus miniatus</i>	14.42 (6.02)	Coffee	231 (72)	273 (187)	Equal	0.64 (0.04)	0.63 (0.06)	Equal	0.80 (0.17)	0.64 (0.16)	Equal	0.07 (0.35)	Equal
<i>Euphonia lanirostris</i>	-2.63 (6.30)	Equal	249 (58)	606 (224)	Forest	0.68 (0.06)	0.53 (0.07)	Coffee	NA	NA	-	-0.27 (0.32)	Equal

360  
 361  
 362 Of the tests for performance, 41 showed no statistically clear differences between habitats, 5  
 363 favored forest and 8 coffee. Breeding was the most informative indicator (5 clear differences),  
 364 followed by body condition (4), muscle (3), primary molt (1), and juveniles (0) (Table 3).

365

366 **Table 3.** Indicators of habitat performance for twelve resident birds in native forest and shade  
 367 coffee (values in the first column correspond to model-averaged habitat coefficients (standard  
 368 errors), second column shows interpretation i.e. in which habitat the species performed better  
 369 according to that indicator; forest was the reference habitat in all analyses).

370

Species	Body Condition	Muscle	Primary Molt	Breeding	Juveniles					
<i>Mionectes olivaceus</i>	0.25 (0.09)	Forest	-0.01 (0.22)	Equal	-0.09 (0.54)	Equal	0.20 (0.45)	Equal	NA	-
<i>Mionectes oleagineus</i>	0.18 (0.11)	Equal	-0.44 (0.24)	Equal	NA	-	-0.50 (0.42)	Equal	NA	-
<i>Turdus flavipes</i>	0.02 (0.08)	Equal	-0.04 (0.17)	Equal	0.41 (0.48)	Equal	-0.92 (0.19)	Coffee	0.80 (0.50)	Equal
<i>Turdus albicollis</i>	-0.30 (0.23)	Equal	0.45 (0.51)	Equal	-1.24 (1.50)	Equal	0.07 (0.64)	Equal	NA	-
<i>Ramphocelus dimidiatus</i>	-0.27 (0.12)	Coffee	0.73 (0.26)	Forest	-0.19 (0.55)	Equal	-0.31 (0.33)	Equal	0.02 (0.83)	Equal
<i>Tangara gyrola</i>	0.11 (0.08)	Equal	-0.40 (0.19)	Coffee	0.07 (0.22)	Equal	-0.41 (0.18)	Coffee	0.99 (0.58)	Equal
<i>Saltator maximus</i>	0.19 (0.14)	Equal	0.35 (0.34)	Equal	0.94 (0.58)	Equal	0.29 (0.34)	Equal	1.05 (1.44)	Equal
<i>Saltator striatipectus</i>	-0.35 (0.22)	Equal	0.23 (0.44)	Equal	-0.44 (0.86)	Equal	-1.47 (0.64)	Coffee	0.78 (0.88)	Equal
<i>Myiothlypis conspicillata</i>	0.40 (0.12)	Forest	0.14 (0.25)	Equal	1.16 (0.88)	Equal	-0.30 (0.27)	Equal	-0.34 (0.85)	Equal
<i>Basileuterus rufifrons</i>	0.04 (0.12)	Equal	0.28 (0.26)	Equal	-1.15 (0.57)	Coffee	-0.97 (0.34)	Coffee	0.80 (0.61)	Equal
<i>Myioborus miniatus</i>	0.31 (0.16)	Equal	1.31 (0.49)	Forest	-1.10 (1.20)	Equal	0.33 (0.40)	Equal	NA	-
<i>Euphonia laniirostris</i>	-0.29 (0.14)	Coffee	-0.19 (0.30)	Equal	NA	-	0.71 (0.33)	Forest	0.20 (1.46)	Equal

371

372 *Level 2: Composite indexes of habitat preference and performance*

373

374 We classified five species as preferring forest (*M. olivaceus*, *M. oleagineus*, *T. albicollis*, *T.*  
 375 *gyrola*, and *M. conspicillata*), two as preferring coffee (*S. striatipectus* and *M. miniatus*) and five  
 376 as having equal-preference (*T. flavipes*, *R. dimidiatus*, *S. maximus*, *B. rufifrons* and *E.*  
 377 *laniirostris*). There were varying degrees of evidence available to make these assignments, with  
 378 preference scores ranging from -2 to 3 (Table 4).

379

380 We classified four species as experiencing higher performance in forest (*M. olivaceus*, *S.*  
 381 *maximus*, *M. conspicillata* and *M. miniatus*), one as experiencing higher performance in coffee  
 382 (*S. striatipectus*) and the remaining seven as experiencing no difference in performance between

383 habitats (*M. oleagineus*, *T. flavipes*, *T. albicollis*, *R. dimidiatus*, *T. gyrola*, *B. rufifrons* and *E.*  
 384 *laniirostris*). There were varying degrees of evidence available to make these assignments, with  
 385 coefficients of the meta-analysis for the effect of habitat ranging from -0.30 to 0.38 (Table 4).

386

387 *Level 3: Is habitat selection acting adaptively?*

388

389 We documented: 1) three species preferring their higher-performance habitat (*M. olivaceus*, and  
 390 *M. conspicillata* for forest, and *S. striatipectus* for coffee); 2) three species preferring forest but  
 391 without consistent evidence for differences in performance (*M. oleagineus*, *T. albicollis* and *T.*  
 392 *gyrola*); 3) four species without consistent differences in either trait (*T. flavipes*, *R. dimidiatus*, *B.*  
 393 *rufifrons* and *E. laniirostris*); 4) *S. maximus* potentially being caught in an equal-preference trap  
 394 (preferring neither habitat despite forest showing higher performance); and 5) *M. miniatus*  
 395 potentially being caught in a severe ecological trap (higher preference for coffee despite higher  
 396 performance in forest) (Table 4).

397

398 **Table 4.** Composite indexes of habitat preference and performance, and hypotheses about the  
 399 role that shade coffee, native forest and habitat selection play at the landscape level for  
 400 populations of twelve resident birds.

401

Species	Preference		Performance		Interpretation		
	Index	Int.	Index	Int.	Coffee	Forest	Selection
<i>Mionectes olivaceus</i>	1	Forest	0.21 (0.08)	Forest	Sink	Source	Adaptive
<i>Mionectes oleagineus</i>	1	Forest	0.04 (0.10)	Equal	Undervalued resource	Equal-quality trap	Non-ideal
<i>Turdus flavipes</i>	0	Equal	-0.08 (0.07)	Equal	Generalist habitat	Generalist habitat	Adaptive
<i>Turdus albicollis</i>	3	Forest	-0.17 (0.20)	Equal	Undervalued resource	Equal-quality trap	Non-ideal
<i>Ramphocelus dimidiatus</i>	0	Equal	-0.12 (0.10)	Equal	Generalist habitat	Generalist habitat	Adaptive
<i>Tangara gyrola</i>	2	Forest	-0.01 (0.06)	Equal	Undervalued resource	Equal-quality trap	Non-ideal
<i>Saltator maximus</i>	0	Equal	0.26 (0.12)	Forest	Equal-preference trap	Undervalued resource	Non-ideal

<i>Saltator striatipectus</i>	-2	Coffee	-0.30 (0.18)	Coffee	Source	Sink	Adaptive
<i>Myiothlypis conspicillata</i>	1	Forest	0.26 (0.10)	Forest	Sink	Source	Adaptive
<i>Basileuterus rufifrons</i>	0	Equal	-0.03 (0.10)	Equal	Generalist habitat	Generalist habitat	Adaptive
<i>Myioborus miniatus</i>	-1	Coffee	0.38 (0.14)	Forest	Severe ecological trap	Severe undervalued resource	Maladaptive
<i>Euphonia laniirostris</i>	0	Equal	-0.14 (0.12)	Equal	Generalist habitat	Generalist habitat	Adaptive

402

403 **Discussion**

404

405 We evaluated the role of native forest and shade coffee in the conservation of resident birds by  
 406 assessing the correlation between measures of habitat preference and performance in a three-level  
 407 analytical approach designed to gauge whether habitat selection was acting adaptively in an  
 408 intermediately-modified landscape.

409

410 At the first level of analysis, where we calculated separate indicators of habitat preference and  
 411 performance, 70% of statistical tests showed no statistically clear differences between habitats,  
 412 partly possibly due to unaccounted variation such as temporal effects, but also reflecting high  
 413 levels of use of both habitats by our focal species. Indicators pointed more commonly towards  
 414 higher preference in forest (12 vs. 7) and higher performance in coffee (8 vs. 5). Evidence  
 415 provided by these indicators was rarely consistent within a species, so no single indicator could  
 416 accurately assess either process for all of them. While using a suite of indicators gave us more  
 417 confidence in our assessments and is recommended when little is known about the behavior of  
 418 the species of interest, further work is needed to validate our chosen variables (see Supporting  
 419 Information for further discussion of these issues). Preference assessment could be improved by  
 420 measuring settlement patterns (Hollander et al. 2011), which in resident species would imply  
 421 being able to follow young during post-natal dispersal (Cox et al. 2014) or conducting choice  
 422 experiments involving the manipulation of selection cues (Robertson & Hutto 2006).

423 Performance assessment could be improved by including parameters more directly related to  
424 survival and reproduction (Johnson 2007), as the relationship between fitness and parameters  
425 such as body condition, muscle and plumage have not been evaluated directly for the focal  
426 species (Labocha & Hayes 2012).

427  
428 At the second level of analysis, the composite index for habitat preference had the disadvantage  
429 of not taking into account uncertainty in estimates or the strength of evidence supporting  
430 preference for each habitat, however, it allowed us to deal with contradictory evidence and gauge  
431 whether unclear differences were pointing to equal-preference or insufficient information.

432 Although effect sizes of habitat at the first level of analysis were generally small, the index of  
433 habitat performance allowed us to weight the strength of evidence when combining the five  
434 indicators (further work could develop an index that allows for non-equal weighing of variables  
435 in relationship to how strongly they correlate with fitness). Half of the resulting assignments  
436 pointed to patterns expected for generalist species (equal-preference and equal-quality), while in  
437 the remaining comparisons evidence favored forest over coffee as the preferred (5 vs. 2) and  
438 better-performance (4 vs. 1) habitat.

439  
440 At the third level of analysis, we found evidence of source-sink dynamics for three species,  
441 although these patterns need to be corroborated with detailed demographic information (Furrer &  
442 Pasinelli 2015; Gilroy & Edwards 2017). Four species showed a pattern consistent with the  
443 equal-preference and equal-quality assumption underlying community-level analyses. Of the four  
444 species displaying evidence of mild non-ideal selection, three appeared to undervalue coffee as a  
445 suitable habitat. If this trend is explored further and proves to be relatively common, detailed

446 studies into the cues used for selection may provide straightforward management actions to raise  
447 the conservation value of coffee and similar novel ecosystems (Gilroy & Sutherland 2007).

448

449 We only found evidence of one species being caught in a severe ecological trap but the strength  
450 of evidence behind preference and performance was weak (statistically clear differences were  
451 only found for occupancy and muscle score). If this pattern were to hold with additional  
452 evidence, actions could be designed to reduce the attractiveness of shade coffee for the species  
453 (Gilroy & Sutherland 2007). Indeed, abundance (Chandler et al. 2013) and demographic  
454 (Mumme 2015) data from Central America suggest a high sensitivity of *Myioborus miniatus* to  
455 landscape structure, making it an interesting species for further study.

456

457 Even though the expansion of coffee plantations in the Sierra Nevada de Santa Marta was rapid  
458 and widespread, it occurred in the late 19<sup>th</sup> century (Carriker 2001), raising the question as to why  
459 species subject to maladaptive selection should persist. Methodologically, tradeoffs between  
460 different aspects of preference and performance may compensate for what appears to be  
461 maladaptive selection (Battin 2004; Chalfoun & Martin 2007), and both density dependence and  
462 the mixing of territorial and floater individuals may interfere with our ability to detect  
463 phenomena that lead to ecological traps (Watkinson 1995; Sherry & Holmes 1996), especially if  
464 population-level phenomena such as source-sink dynamics compensate for differences in  
465 individual-level performance. Ecologically, traps originating from novel components in a  
466 landscape seem less likely to facilitate extinction than those arising from habitat degradation  
467 (Fletcher et al. 2012). We know from simulations of our study system that even if the effect of an  
468 ecological trap is not strong enough to lead to population extinction under current conditions, this  
469 may change rapidly with further landscape change (Sánchez-Clavijo et al. 2016); so that under

470 further forest loss the landscape could lose more species than predicted by models assuming  
471 adaptive habitat selection.

472

473 Our preference/performance results showed no clear pattern of association with individual  
474 species characteristics such as habitat guild, family, size and connectedness of the subspecies  
475 distributions, diet, participation in mixed flocks, body size or previous classifications of  
476 sensitivity to human activity (Table S1). Most of the characteristics hypothesized to make species  
477 vulnerable to traps (speed of evolution, rate of learning, behavioral plasticity, etc. - Battin et al.  
478 2004) are unknown for the majority of tropical species, and even though we assume residents  
479 have better knowledge of their environments than their migrant counterparts, our ability to prove  
480 this is hindered by the lack of studies that go beyond community-level measures for resident  
481 birds.

482

483 *Research and conservation implications*

484

485 For future applications of this framework in the assessment of the value of different land-uses for  
486 biodiversity conservation, we strongly advise using multiple lines of evidence to simultaneously  
487 assess habitat preference and performance, as indicator choice may lead to different  
488 interpretations about the adaptiveness of habitat selection (Robertson & Hutto 2006; Chalfoun &  
489 Martin 2007). The value of this approach will increase as our understanding of the relationship  
490 between the variables of interest and the chosen indicators improves (Noss 1990). A further  
491 refinement will be to extend our framework into a continuous approach to evaluate the  
492 adaptiveness of habitat selection, by evaluating preference and performance in a way that makes  
493 them directly comparable. A quantitative approach would help reduce the subjectivity of

494 interpretations and is more realistic for cases in which the habitats being compared do not hold  
495 independent populations of the species of interest (Kristan 2003; Part et al. 2007).

496  
497 While the ideal way to measure the value of a habitat would be to estimate survival and  
498 reproduction parameters based on long-term demographic and/or movement studies of species  
499 showing a representative range of ecological and behavioral traits (Sekercioglu et al. 2007;  
500 Milder et al. 2014), these are rarely feasible in tropical ecosystems. Our results show that  
501 occupancy and abundance models generated from the wide-scale application of simple sampling  
502 methodologies (Ruiz-Gutiérrez et al. 2016; Irizarry et al. 2018) may neglect important differences  
503 in habitat preference and performance. Capture data allows for the exploration of habitat-species  
504 relationships at a resolution intermediate between these two extremes (Ruiz-Gutiérrez et al. 2012;  
505 Sekercioglu 2012), but to increase the reliability of our inference, we need further research into  
506 the relationships between individual-level data and consequences at the population level. An ideal  
507 approach would be to combine these three scales of analysis.

508  
509 For example, studies comparing “spare” (leaving remnant habitats as intact as possible and  
510 intensifying the use of surrounding lands to provide goods and services) and “share” (promoting  
511 land uses where farming and biodiversity conservation occur simultaneously) approaches to rural  
512 management still assess land use contributions to biodiversity conservation based solely on  
513 species abundance (Fischer et al. 2008; Phalan et al. 2011; von Wehrden et al. 2014; Gilroy &  
514 Edwards 2017). Monitoring approaches using a multi-species, multi-indicator and multi-scale  
515 approach to quantify preference and performance in landscapes of different compositions, could  
516 advance our understanding of biodiversity responses to landscape heterogeneity and support  
517 better decision-making for conservation and sustainable use.

518

519 Even though our site has an unusual combination of large remnants of native forest and shade  
520 coffee plantations with 60% average canopy cover (Guhl 2004; Sánchez-Clavijo et al. 2009), our  
521 findings are similar to those from comparable studies carried out in tropical agricultural  
522 landscapes (Perfecto & Vandermeer 2002; Chandler et al. 2013; Kremen 2015). The mostly  
523 adaptive or neutral selection behaviors found for our focal species suggest that shade coffee can  
524 be a biodiversity-friendly matrix that provides complementary or supplementary habitat to a wide  
525 range of resident birds, even becoming a favorable breeding habitat for some generalist species.  
526 However, higher preference and performance in forest for species with varied life histories,  
527 reinforce the importance of protecting remnants of native vegetation in agricultural landscapes.

528

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530

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**542 Authors' contributions**

543  
544 LMSC and PFQA conceived the ideas, designed methodology and analyzed the data; NJB guided  
545 data collection and provided additional data; LMSC led the writing of the manuscript. All authors  
546 contributed critically to the drafts and gave final approval for publication.

547

**548 Data Availability**

549  
550 Raw data used in this paper is available from SiB Colombia (DOI pending), while curated  
551 datasets for all analyses presented are available from the Dryad Digital Repository (DOI  
552 pending).

553

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753

754 **Supporting Information**

755 Sanchez-Clavijo\_etal2019\_SuppInf\_Part1.pdf

756 Sanchez-Clavijo\_etal2019\_SuppInf\_Part2.pdf

## Habitat selection in transformed landscapes and the role of forest remnants and shade coffee in the conservation of resident birds.

Sánchez-Clavijo, L.M., Bayly, N.J., and Quintana-Ascencio P.F. (2019). *Journal of Animal Ecology* V(i): pp.

**Supporting Information - Part 1:** supplementary details, tables and figures for Materials and Methods, Results and Discussion.

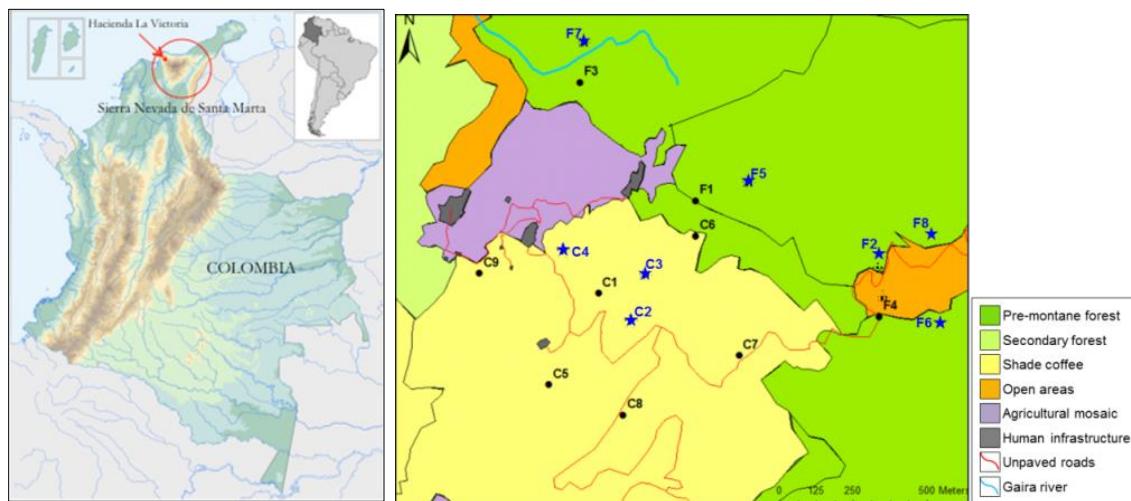
### Table of Contents

<b>1. MATERIALS AND METHODS.....</b>	<b>2</b>
1.1. STUDY AREA & SAMPLING SCHEME .....	2
1.2. FOCAL SPECIES .....	2
1.3. LEVEL 1: SEPARATE INDICATORS OF HABITAT PREFERENCE AND PERFORMANCE .....	3
1.3.1. <i>Occurrence</i> .....	3
1.3.2. <i>Abundance</i> .....	4
1.3.3. <i>Fidelity</i> .....	7
1.3.4. <i>Inter-seasonal variance</i> .....	7
1.3.5. <i>Age</i> .....	7
1.3.6. <i>Body condition</i> .....	9
1.3.7. <i>Muscle</i> .....	11
1.3.8. <i>Primary molt</i> .....	12
1.3.9. <i>Breeding</i> .....	13
1.3.10. <i>Juveniles</i> .....	14
1.4. LEVEL 2: COMPOSITE INDEXES OF HABITAT PREFERENCE AND PERFORMANCE.....	15
1.4.1. <i>Habitat preference</i> .....	15
1.4.2. <i>Habitat performance</i> .....	15
<b>2. RESULTS.....</b>	<b>15</b>
2.1. LEVEL 1: SEPARATE INDICATORS OF HABITAT PREFERENCE AND PERFORMANCE .....	15
2.3.1. <i>Occurrence</i> .....	16
2.3.2. <i>Abundance</i> .....	21
2.3.3. <i>Fidelity</i> .....	40
2.3.4. <i>Inter-seasonal variance</i> .....	47
2.3.5. <i>Output from generalized linear models</i> .....	47
2.2. LEVEL 2: COMPOSITE INDEXES OF HABITAT PREFERENCE AND PERFORMANCE.....	55
2.2.1. <i>Habitat preference</i> .....	55
2.2.2. <i>Habitat performance</i> .....	56
<b>3. DISCUSSION .....</b>	<b>57</b>
3.1. LEVEL 1: SEPARATE INDICATORS OF HABITAT PREFERENCE AND PERFORMANCE .....	57
3.1.1. <i>Habitat preference</i> .....	57
3.1.2. <i>Habitat performance</i> .....	57
3.2. LEVEL 2: COMPOSITE INDEXES OF HABITAT PREFERENCE AND PERFORMANCE.....	58
3.2.1. <i>Habitat preference</i> .....	58
3.2.2. <i>Habitat performance</i> .....	59
<b>4. REFERENCES .....</b>	<b>59</b>

## 1. Materials and Methods

### 1.1. Study area & sampling scheme

Hacienda La Victoria is an 800-hectare agricultural estate devoted to coffee production and forest conservation in the Sierra Nevada de Santa Marta in northern Colombia (Fig. S1A). All coffee was grown in moderate to steep slopes, underneath cultivated shade dominated by *Inga codonantha* and *Albizia carbonaria*, with occasional interspersed trees of other edible and ornamental species, and could be classified as a commercial polyculture (Moguel & Toledo 1999). Canopy height in the plots where sampling took place (Fig. S1B) was generally between 10 and 15m, and canopy percent cover varied greatly around a mean of 60%. Coffee shrub density and height varied according to when plots were last renewed (cut down) or replanted, and groundcover depended on the time of year (high after the dry season, cleared out when the rainy season intensified). Coffee production was the main economic activity of the farm at the time sampling took place, but most plots were not managed intensively, and experienced cycles of temporal local abandonment followed by increases in intensification, and cycles of coffee leaf rust propagation followed by renewal of the coffee shrubs. Plant diversity in remnant forest patches was much higher, with a dominance of species from Lauraceae, Melastomataceae, Araliaceae, Euphorbiaceae, Rubiaceae and Leguminosae. Canopy heights in the forest areas where sampling took place (Fig. S1B) ranged from 15 to 30 m and canopy cover had low variance around a mean of 80%. Most forest sites had a dense understory of palms, ferns, Heliconiaceae and other understory shrubs, including occasional coffee plants either left over from previous plantings or that dispersed naturally from nearby crops. At the time sampling took place forests were only being used for biological research, low-intensity tourism and occasionally for wood extraction.



**Figure S1.** A) Location of Hacienda La Victoria in northern Colombia. B) Schematic map of study site showing the approximate location of banding stations and vegetation cover (F: banding stations in forest, C: banding stations in coffee, black points shows stations where both mist-netting and observations took place, blue stars show stations where only mist-netting took place).

### 1.2. Focal species

According to the literature our focal species represent a gradient of distribution range sizes, affinity towards habitats with different levels of tree cover, diet and social behaviors; however, none are classified as highly sensitive to human activity or as having high conservation or research priorities (Table S1).

**Table S1.** Taxonomical, ecological and behavioral characteristics for 12 focal species of resident birds present in native forest and shade coffee plantations in the Sierra Nevada de Santa Marta, Colombia.

Family	Species <sup>1</sup>	Distribution <sup>2</sup>	Habitat <sup>3</sup>			Diet <sup>3</sup>		Social behavior <sup>3</sup>	Sensitivity to human activity <sup>4</sup>	Conservation priority <sup>4</sup>	Research priority <sup>4</sup>
			Forest	Edge or Woody	Scrub	Fruit	Insects				
Tyrannidae	<i>Mionectes olivaceus galbinus</i>	Restricted	X			X	X	Joins mixed flocks	Medium	Low	Low
Tyrannidae	<i>Mionectes oleagineus parcus</i>	Continuous	X	X		X	X	Joins mixed flocks	Medium	Low	Low
Turdidae	<i>Turdus flavipes venezuelensis</i>	Discontinuous	X	X		X	X	Seasonal migrations	Medium	Low	Medium
Turdidae	<i>Turdus albicollis phaeopygooides</i>	Discontinuous	X			X	X	Follows ant swarms	Medium	Low	Medium
Thraupidae	<i>Ramphocelus dimidiatus dimidiatus</i>	Continuous		X	X	X		Conspecific pairs/groups	Low	Low	Low
Thraupidae	<i>Tangara gyrola toddi</i>	Discontinuous	X	X		X	X	Joins mixed flocks	Medium	Low	Low
Thraupidae	<i>Saltator maximus maximus</i>	Continuous		X	X	X	X	Joins mixed flocks	Low	Low	Low
Thraupidae	<i>Saltator striaticeps perstriatus</i>	Continuous		X	X	X	X	Conspecific pairs/groups	Low	Low	Low
Parulidae	<i>Myiothlypis conspicillata</i>	Restricted	X				X	Conspecific pairs/groups	Medium	Medium	Low
Parulidae	<i>Basileuterus rufifrons mesochrysus</i>	Continuous		X	X		X	Conspecific pairs/groups	Low	Low	Low
Parulidae	<i>Myioborus miniatus sanctaemartiae</i>	Restricted	X	X			X	Joins mixed flocks	Low	Low	Low
Fringillidae	<i>Euphonia laniirostris crassirostris</i>	Continuous		X	X	X		Joins mixed flocks	Low	Low	Low

<sup>1</sup> Subspecies found in the Sierra Nevada de Santa Marta (Ayerbe-Quiñones 2018).<sup>2</sup> *Continuous*: the subspecies found in the Sierra Nevada de Santa Marta has a continuous distribution throughout the rest of northern Colombia and/or the Andes; *Discontinuous*: the subspecies found in the Sierra Nevada de Santa Marta is only found in a few and disjoint populations in northern Colombia; *Restricted*: the subspecies is endemic to the Sierra Nevada de Santa Marta (Ayerbe-Quiñones 2018).<sup>3</sup> Sources: Parker *et al.* 1996; Ayerbe-Quiñones 2018; Schulenberg 2019 (see references for authorship of species accounts).<sup>4</sup> Parker *et al.* 1996

### 1.3. Level 1: Separate indicators of habitat preference and performance

This section contains a detailed account of the data collection, processing, and analysis protocols carried out for the ten chosen indicators of habitat preference and performance.

#### 1.3.1. Occurrence

**Data collection** – estimations of occupancy were based on 16 complementary observation sessions that were carried out in and near our banding stations during 2013 and 2014. Observations took place for two to four hours after sunrise or before sunset and were always carried out by a single observer. Although the initial aim of these sessions was to accumulate resights of color banded birds, observers recorded all detected individuals of the focal species. During some of these sessions, playback was used to increase bird detectability (we looped over a playlist that featured calls and songs from all focal species, plus a

Neotropical owl mix - courtesy of the Cornell Lab of Ornithology). Although the total time spent birdwatching and the use of playback varied among occasions, it was kept constant for each site within occasions and amounted to a mean total of 36 hours per site.

**Data processing** – a detection/non-detection matrix was generated for each species that contained 9 rows (one per banding site) and 16 columns (one per sampling occasion). Additionally, we included the type of habitat (coffee or forest) and the altitude for each site.

**Data analysis** – to determine whether habitat had an effect on the probability of occurrence we used the following two models:

**Observation model for replicated detection/non-detection observations**

$$\text{logit}(p) = \alpha_1 + \beta_1 * \text{habitat}(i)$$

**Ecological model for true occurrence**

$$\text{logit}(\psi_i) = \alpha_2 + \beta_2 * \text{habitat}(i) + \beta_3 * \text{altitude} + \beta_4 * \text{altitude}^2$$

The effect of habitat on the probability of observation (**p**) was included because having a lower and more open canopy means that visibility in shade coffee is usually much higher than in forest remnants, and we wanted to differentiate this from habitat effects on true occurrence, which was our main interest. We did not include the effect of season, year, effort (in number of hours) and method (with or without playback) on **p** because our species were common enough that while these variations probably increased the number of individuals detected per species, we would not expect a clear effect on whether or not they were detected. We included a quadratic effect of altitude on the probability of occurrence (**ψ**) because in tropical mountains it is very common for bird species to have narrow altitudinal ranges (Kattan & Franco 2004; Gómez *et al.* 2015).

We followed a Bayesian implementation of occupancy models using the code available at <http://www.vogelwarte.ch/de/projekte/publikationen/bpa/code-for-running-bpa-using-jags.html> (Kéry & Schaub 2012). Analyses were run using R (R Core Team 2016), JAGS (Plummer 2016) and the “jagsui” package (Kellner 2016). After checking model outputs for convergence, we considered the effect of habitat on **ψ** as statistically clear (Dushoff *et al.* 2019) when parameter estimates for **β2** did not include 0 in their 95% credibility intervals (because of the way we set up the models, positive values of this coefficient indicated higher expected occupancy in coffee, and negative values higher expected occupancy in forest). We also plotted occupancy probability as a function of altitude and habitat to help us interpret results.

### 1.3.2. Abundance

**Data collection** – because all captured individuals were banded with either a uniquely-coded metal ring or a unique combination of color bands, instead of assuming abundances from species counts, we were able to use capture-mark-recapture analyses. Even though we kept the places where we set up mist nets consistent among sampling seasons, there was a lot of variation in effort (Table S2), and therefore in sample sizes according to species, habitat and method, which had to be accounted for during data analysis (Table S3).

**Table S2.** Capture effort in standardized mist-net hours (Mist-netting sites - C: banding stations in shade coffee, F: banding stations in native forest; Sampling events - DW: “dry-wet” seasons (March-May), MW: “mid-wet” seasons (June-August), PW: “peak-wet” (September-November), followed by year of sampling; differences in effort stem from different lengths of sampling occasions, numbers of mist nets set up per site, and/or duration of sampling per day).

		Sampling events												
		PW09	DW10	PW10	DW11	MW11	PW11	DW12	DW13	MW13	DW14	MW14	DW15	
Mist-netting sites	C1	152	1,245	415	1,062	277	-	663	832	174	126	168	109	
	C2	-	-	-	755	-	-	484	-	-	-	-	-	
	C3	-	-	-	-	248	-	-	-	-	-	-	-	
	C4	-	-	-	-	-	-	156	-	-	-	-	-	
	C5	-	-	-	-	-	-	-	599	141	133	179	-	
	C6	-	-	-	-	-	-	-	113	143	126	164	-	
	C7	-	-	-	-	-	-	-	142	164	135	179	-	
	C8	-	-	-	-	-	-	-	129	172	126	175	-	
	C9	-	-	-	-	-	-	-	137	148	117	210	-	
	F1	1,144	2,072	928	2,846	256	1,008	2,448	1,266	147	548	167	1,007	
	F2	-	-	-	1,183	-	-	878	245	-	82	-	-	
	F3	-	-	-	-	262	657	775	625	147	373	178	619	
	F4	-	-	-	-	-	-	1,001	591	170	392	168	-	
	F5	-	-	-	-	-	-	-	434	-	372	-	448	
	F6	-	-	-	-	-	-	-	610	-	570	-	834	
	F7	-	-	-	-	-	-	-	342	-	366	-	479	
	F8	-	-	-	-	-	-	-	-	-	-	-	443	

**Table S3.** Summary of banding data available for capture-mark-recapture analyses for 12 species of resident birds present in native forest and shade coffee plantations in the Sierra Nevada de Santa Marta, Colombia.

Species	Habitat	Captures	Individuals(I)	Recaptures(R)	Ratio R/I
<i>Mionectes olivaceus</i>	Coffee	124	117	7	0.06
	Forest	577	462	115	0.25
<i>Mionectes oleagineus</i>	Coffee	95	66	29	0.44
	Forest	318	235	83	0.35
<i>Turdus flavipes</i>	Coffee	223	195	28	0.14
	Forest	392	339	53	0.16
<i>Turdus albicollis</i>	Coffee	19	19	0	0.00
	Forest	330	244	86	0.35
<i>Ramphocelus dimidiatus</i>	Coffee	126	101	25	0.25
	Forest	119	88	31	0.35
<i>Tangara gyrola</i>	Coffee	182	138	44	0.32
	Forest	398	267	131	0.49
<i>Saltator maximus</i>	Coffee	91	74	17	0.23
	Forest	92	71	21	0.30
<i>Saltator striatipectus</i>	Coffee	70	60	10	0.17
	Forest	34	33	1	0.03
<i>Myiothlypis conspicillata</i>	Coffee	86	64	22	0.34
	Forest	247	188	59	0.31
<i>Basileuterus rufifrons</i>	Coffee	174	131	43	0.33
	Forest	83	61	22	0.36
<i>Myioborus miniatus</i>	Coffee	90	62	28	0.45
	Forest	46	30	16	0.53
<i>Euphonia laniirostris</i>	Coffee	95	80	15	0.19
	Forest	97	89	8	0.09

**Data processing** – to improve precision in habitat-specific demographic estimates we pooled together data from all sites into two habitat classes, and data from each mist-net day into twelve primary sampling events. For each species we produced two data matrices, one for coffee and one for forest, that contained one row for each individual that was identified during the whole sampling time, and one column for each primary sampling event, and simply displayed whether the individual was captured (1) or not (0) during that occasion in that habitat.

**Data analysis** – because of differences in recapture rates among species and habitats, we had to use two different types of models to estimate total abundance over the whole sampling time. For species that had very low recapture rates in either one or the two habitats (*M. olivaceus*, *T. flavipes*, *T. albicollis*, *S. maximus*, *S. striatipectus* and *E. laniirostris*) we used a closed population capture-recapture model (CP), built within a Bayesian framework and taking advantage of parameter-expanded data augmentation (Kéry & Schaub 2012). Data augmentation was introduced to capture–recapture models by Royle *et al.* (2007) and consists of “adding an arbitrary number of zeroes to the dataset and (then) analyzing a reparametrized version of the original model. Essentially, it converts the closed-population model into an occupancy model and turns the problem of estimating abundance N into that of estimating occupancy( $\psi$ )” (Kéry & Schaub 2012). Following the guidelines of the latter authors, to diagnose if data augmentation was enough, we looked at the posterior distributions of N and they were truncated, we repeated the analysis with a larger number of added zeroes.

For species with relatively high recapture rates for both habitats (*M. oleagineus*, *R. dimidiatus*, *T. gyrota*, *M. conspicillata*, *B. rufifrons* and *M. miniatus*) we used a Jolly-Seber (JS) population model parameterized as a multistate model (Royle & Dorazio 2008; Kéry & Schaub 2012), which again was built within a Bayesian framework with data augmentation, and which allowed for the simultaneous estimation of population size and apparent survival. While some model assumptions were probably not met, we assumed that violations would lead to similar inaccuracies in both habitats, and since we were interested in comparing estimated abundance between coffee and forest, rather than on calculating species densities with precision, we chose the models that could best be parameterized given our data.

We modified code available at <http://www.vogelwarte.ch/de/projekte/publikationen/bpa/code-for-running-bpa-using-jags.html> for both models by introducing a quadratic effect of sampling effort (scaled number of standardized mist net hours per habitat, per occasion) on detection probabilities (**p**). This allowed us to account for the fact that effort was uneven between occasions, and we assumed a quadratic relationship because more mist-netting days are expected to yield more captured individuals up to the point where further effort becomes less effective due to individuals learning to avoid the nets. Additionally, for JS models, apparent survival ( $\Phi$ ) was allowed to vary randomly by occasion.

Estimates of the number of individuals alive per occasion (**N**) and over the whole sampling time (**Nsuper**) were calculated as derived population parameters. These analyses were run using R (R Core Team 2016), JAGS (Plummer 2016) and the “jagsui” package (Kellner 2016) and involved running from 3,000 to 300,000 iterations to achieve adequate convergence of all model parameters. After checking model outputs for convergence, we considered species estimations of overall abundance (**Nsuper**) as different between habitats when there was no overlap in their 50% credibility intervals, and plotted their posterior distributions together to visually assess overlap (using package “ggplot2” by Wickham 2016). While parameters were not estimated with the same method for all species, we made sure to always keep the model we used, the number of augmented individuals and the simulation conditions constant for both habitats within each species to make fair comparisons.

Inference from Bayesian population models was considered reliable when Rhat (the potential scale reduction factor) was less than 1.1 and n.eff (measure of effective sample size) was higher or equal than 30. We ran all models with 300 effective iterations, but because certain species: habitat combinations required more iterations to find a reasonable solution, this value was increased by an order of magnitude for each new run until chains showed adequate convergence. Where this was never reached, tables feature an NA in the results (for example abundance in coffee for *T. albicollis* and in forest for *S. striatipectus* for which there were not enough recaptures to adequately run a CRC model).

### 1.3.3. Fidelity

**Data collection and processing** – we used the same datasets that we used to estimate abundance.

**Data analysis** – for species with low recapture rates in either one or the two habitats (*M. olivaceus*, *T. flavipes*, *T. albicollis*, *S. maximus*, *S. striatipectus* and *E. laniirostris*) we used a Cormack-Jolly-Seber (CJS) model to calculate apparent survival ( $\Phi$ ) per occasion (Royle & Dorazio 2008; Kéry & Schaub 2012). Again, we introduced a quadratic effect of sampling effort on detection probabilities ( $p$ ) to modify the available code for this model. These type of models are widely used for the joint estimation of recapture and survival probabilities in animal populations (Kéry & Schaub 2012), but do not allow for the simultaneous estimation of abundance. For species with high recapture rates in both habitats (*M. oleagineus*, *R. dimidiatus*, *T. gyrola*, *M. conspicillata*, *B. rufifrons* and *M. miniatus*) we used  $\Phi$  estimates from the Jolly-Seber (JS) population models. After checking for convergence in estimated parameters, we used model outputs to calculate mean  $\Phi$  between occasions for each iteration, and then derived the mean and standard deviation for all iterations to assess differences in site fidelity between habitats. Differences were considered statistically clear when there was no overlap in habitat-specific mean phi +/- standard deviation. We plotted posterior distributions for both habitats together to assess overlap visually as well.

### 1.3.4. Inter-seasonal variance

**Data collection and processing** – we used the same datasets that we used to estimate abundance.

**Data analysis** – this analysis could only be carried out for the six species for which we found stable solutions for the Jolly-Serber model (*M. oleagineus*, *R. dimidiatus*, *T. gyrola*, *M. conspicillata*, *B. rufifrons* and *M. miniatus*), as it allows for the derived estimation of population size during n-1 of the sampling occasions (Kéry & Schaub 2012). After running the models and checking for parameter convergence, we calculated the coefficient of variance between abundance per occasion for each iteration, and then derived the mean and standard deviation for all iterations to get an estimate of seasonal variance in abundance for each species and habitat combination. We considered differences as statistically clear when there was no overlap of the mean +/- the standard deviation. Since models already accounted for differences in effort, we assumed differences accounted for actual variations in abundance per habitat.

### 1.3.5. Age

**Data collection** – all age information came from the processing of individuals captured via mist-netting. Classification according to age was done following guidelines by Pyle *et al.* 1987, Wolfe *et al.* 2010, and SELVA's unpublished ageing guide for species in the region. The main characteristics used to classify individuals were plumage color, shape, quality, and contrast between flight feathers, aided by secondary features such as body feather density, beak commissure, eye color and cranium, depending on the species. Since SELVA researchers have been working at this site for a long time, all participants of this study received training to appropriately age individuals of our focal species.

The following five categories were used:

- 0: unknown – age could not be determined for the individual at that particular moment  
 1: fledgling – very young birds that had not completed their first molt  
 2: juvenile – young birds that already completed their first molt  
 3: immature – birds that have gone through an incomplete pre-formative molt  
 4: adult – all birds that have gone through at least one complete basic molt

**Data processing** – individual capture histories were checked one by one to ensure that age information was consistent (for example that an individual classified as an adult in one season was not classified as an immature in the following season), and to complete missing information that could be gathered from other captures (for example an individual classified as unknown that had already been captured two years before could be determined *post hoc* to be an adult). We excluded *M. olivaceus* and *M. oleagineus* from this analysis because they do not go through an incomplete pre-formative molt and because age determination criteria were not clear throughout our study period. To focus on dominant vs. subordinate individuals, we restricted the dataset from the ten remaining species to all captured individuals that were classified as immature or adult. We recoded this data matrix to show a value of 1 when the individual was an adult and a value of 0 when it was an immature (Table S4 shows sample sizes available for this analysis).

**Table S4.** Sample size (n) represented by the number of captures for each species according to age category and habitat.

Species	Age category	Coffee (n)	Forest (n)
<i>Turdus flavipes</i>	3 (immature)	52	123
	4 (adult)	205	265
<i>Turdus albicollis</i>	3 (immature)	13	154
	4 (adult)	7	267
<i>Ramphocelus dimidiatus</i>	3 (immature)	31	32
	4 (adult)	106	102
<i>Tangara gyrola</i>	3 (immature)	63	117
	4 (adult)	149	328
<i>Saltator maximus</i>	3 (immature)	34	31
	4 (adult)	64	59
<i>Saltator striatipectus</i>	3 (immature)	46	22
	4 (adult)	33	8
<i>Myiothlypis conspicillata</i>	3 (immature)	28	126
	4 (adult)	60	184
<i>Basileuterus rufifrons</i>	3 (immature)	85	55
	4 (adult)	117	47
<i>Myioborus miniatus</i>	3 (immature)	39	19
	4 (adult)	72	38
<i>Euphonia laniirostris</i>	3 (immature)	34	37
	4 (adult)	61	54

**Data analysis** – since age was transformed into a binary response, to test for differences in the distribution of individuals between habitats according to age, we used the following set of generalized linear models (with binomial errors and logit links):

- AGE ~ 1 (null model)  
 AGE ~ DAY+DAY<sup>2</sup> (considers only the quadratic effect of day of year)  
 AGE ~ HABITAT (considers only the effect of habitat)  
 AGE ~ DAY+DAY<sup>2</sup>+HABITAT (considers day of year and habitat)

Day of year was included as a quadratic variable because most bird species at our study site breed only after the start of the rainy season, and therefore we expected a strong seasonal response in most of our variables. After running the models, we used R (R Core Team 2016) and package “AICcmmodavg” to

perform AICc model averaging (Mazerolle 2016), and to calculate the model-averaged estimate and unconditional standard error for the habitat coefficient. When 95% confidence intervals for this estimate did not include zero, we concluded that there was higher probability of capturing an adult over an immature individual in one vs. the other habitat (positive coefficients indicated higher probability in forest, while negative coefficients indicated higher probability in coffee). We also plotted the probability of capturing an adult over an immature individual as a function of day of year and habitat to help us interpret results.

Even though the effort which led to the number of records available for age and all following variables was uneven between habitats, seasons and years; we believe that samples sizes were generally large enough to overcome these effects, and for species: habitat combinations where sample sizes were smaller, this would be reflected in wider confidence intervals and/or standard errors which were taken into account at the next level of analysis.

### 1.3.6. Body condition

**Data collection** – body mass and wing chord length of all captured individuals were measured using portable electronic balances and standard wing rules, respectively, and following standardized banding station protocols that allow for minimization of measuring error. Repeated measures from individuals were treated as independent because we assume they vary within individuals based on temporal noise and individuals were never processed twice in the same day. Therefore, we assumed their body mass at the time of weighing was a result of the habitat they were captured in, and assumed because sampling took place at the same time period in both habitats, that temporal noise would be the same in coffee and forest.

**Data processing** – body mass and wing chord data was checked for outliers following Tukey's method (Tukey 1949), and records defined as extreme values according to [ $< Q25 - k(IQR)$  or  $> Q75 + k(IQR)$ , with  $k=3$ ] were removed. We chose Peig & Green's scaled mass index as our index of body condition because it aims to correct mass with a body size indicator (in this case wing chord length), recognizes that standard major axis regression is more appropriate than ordinary least squares regression in cases where both variables are measured with error, and is supposedly capable of accounting for the effects of growth and size dimorphism (Green 2001; Peig & Green 2009). To calculate this index, we ran a standard major axis regression between log-transformed wing chord and body mass, and then used the following formula on the residuals (Figure S2 shows the distribution of data for the body condition index per species):

$$\hat{M}_i = M_i \left[ \frac{L_0}{L_i} \right]^{b_{SMA}}$$

where,

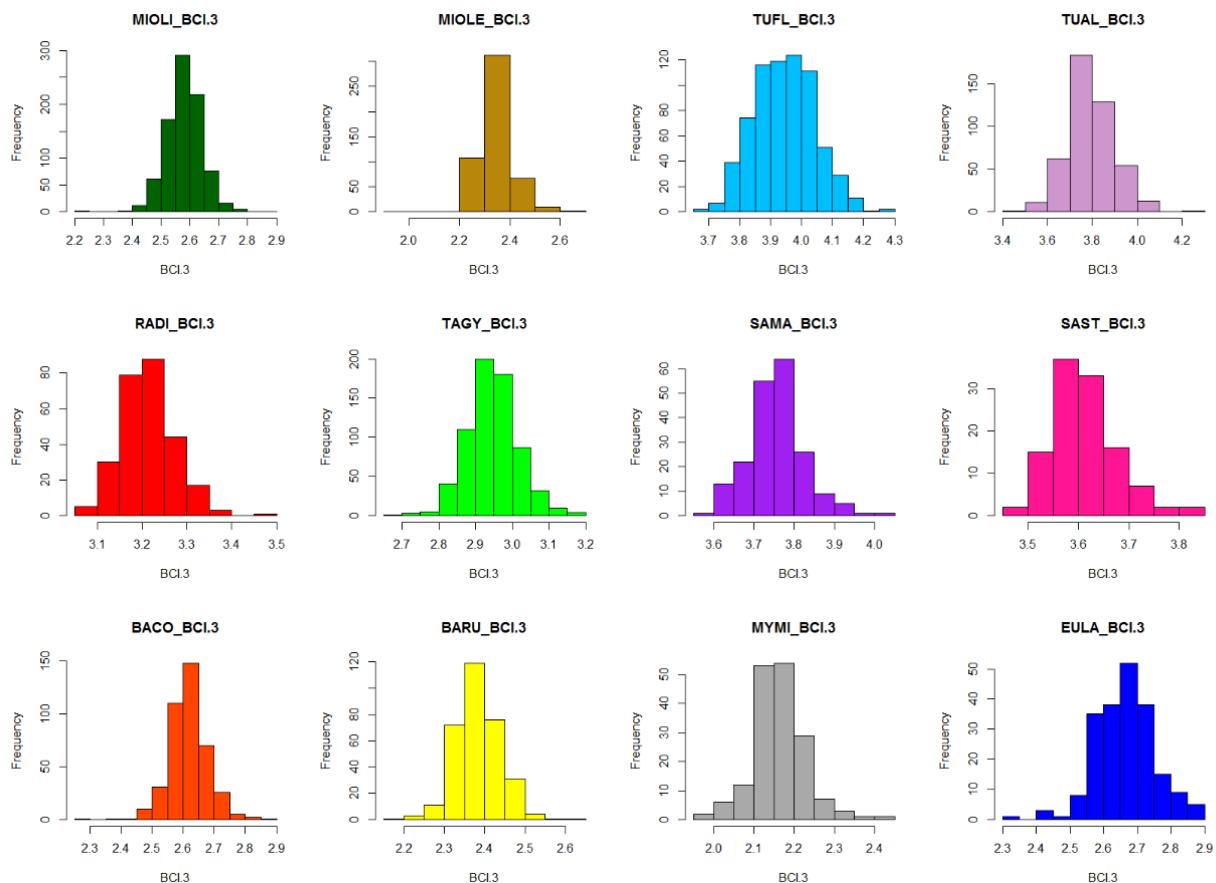
$\hat{M}_i$ : the predicted body mass for individual  $i$  when linear body measure is standardized to  $L_0$

$M_i$ : body mass measurement of individual  $i$

$L_0$ : arithmetic mean value for the study population

$L_i$ : linear body measurement of individual  $i$

$b_{SMA}$ : scaling exponent estimated by the SMA regression of  $M$  on  $L$



**Figure S2.** Histograms for body condition index per species (from left to right starting at the top: *Mionectes olivaceus*, *Mionectes oleagineus*, *Turdus flavipes*, *Turdus albicollis*, *Ramphocelus dimidiatus*, *Tangara gyrola*, *Saltator maximus*, *Saltator striatipectus*, *Myiothlypis conspicillata*, *Basileuterus rufifrons*, *Myioborus miniatus* and *Euphonia laniirostris*).

**Data analysis** – since the body condition index (BCI) was a normally-distributed continuous response, to test for differences in the distribution of individuals between habitats according to age, we used the following set of generalized linear models (with normal errors and identity links) and then used the same process described for age to calculate the model-averaged estimate and unconditional standard error for the habitat coefficient:

BCI ~ 1 (null model)  
 BCI ~ DAY+DAY<sup>2</sup> (considers only the quadratic effect of day of year)  
 BCI ~ HABITAT (considers only the effect of habitat)  
 BCI ~ DAY+DAY<sup>2</sup>+HABITAT (considers day of year and habitat)

When 95% confidence intervals for this estimate did not include zero, we concluded that there was a clear statistical difference in mean body condition between habitats (positive coefficients indicated higher condition in forest, while negative coefficients indicated higher condition in coffee). We also plotted BCI as a function of day of year and habitat to help us interpret results.

### 1.3.7. Muscle

**Data collection** – muscle scores for captured individuals were determined following guidelines by Pyle *et al.* 1987 and ranged from 0 (a highly-wasted breast muscle that looks concave from the breast plate) to 3 (a healthy muscle that forms a convex curve from the breast plate). For adult and immature tropical residents, it usually varies only between scores of 2 and 3, which all station researchers were trained to differentiate.

**Data processing** – we restricted the dataset from all species to all captured individuals that were classified as having muscle scores of 2 or 3, and then recoded this data matrix to show a value of 1 when the score was 3 and a value of 0 when it was 2 (Table S5 shows sample sizes available for this analysis).

**Table S5.** Sample size (n) represented by the number of captures for each species according to muscle score and habitat.

Species	Muscle score	Coffee (n)	Forest (n)
<i>Mionectes olivaceus</i>	2	37	240
	3	95	456
<i>Mionectes oleagineus</i>	2	35	166
	3	64	212
<i>Turdus flavipes</i>	2	152	250
	3	107	155
<i>Turdus albicollis</i>	2	12	256
	3	7	172
<i>Ramphocelus dimidiatus</i>	2	82	59
	3	50	73
<i>Tangara gyrola</i>	2	55	160
	3	154	299
<i>Saltator maximus</i>	2	74	66
	3	23	27
<i>Saltator striatipectus</i>	2	54	19
	3	27	12
<i>Myiothlypis conspicillata</i>	2	35	105
	3	63	208
<i>Basileuterus rufifrons</i>	2	77	33
	3	133	72
<i>Myioborus miniatus</i>	2	55	16
	3	56	41
<i>Euphonia laniirostris</i>	2	55	56
	3	49	44

**Data analysis** – since muscle score was transformed into a binary response, to test for prevalence of higher muscle scores between habitats, we used the following set of generalized linear models (with binomial errors and logit links) and then used the same process described for age to calculate the model-averaged estimate and unconditional standard error for the habitat coefficient:

MUSCLE ~ 1 (null model)

MUSCLE ~ DAY+DAY<sup>2</sup> (considers only the quadratic effect of day of year)

MUSCLE ~ HABITAT (considers only the effect of habitat)

MUSCLE ~ DAY+DAY<sup>2</sup>+HABITAT (considers day of year and habitat)

When 95% confidence intervals for this estimate did not include zero, we concluded that there was a clear statistical difference in muscle scores between habitats (positive coefficients indicated higher condition in forest, while negative coefficients indicated higher condition in coffee). We also plotted the probability of capturing an individual with score 3 over one with score 2 as a function as a function of day of year and habitat to help us interpret results. Because muscle score could be related to age and therefore mix signals of habitat preference and quality, for all species that showed clear statistical

differences we used a chi-square test to check for independence between these two binary variables. Both tests were only statistically clear for *M. miniatus*, so we re-ran the muscle analysis using data for adult individuals only.

### 1.3.8. Primary molt

**Data collection** – primary plumage for all captured individuals was examined to determine the number of plumage generations, in order to assist with age classification. Because primary plumage molting is so easy to detect, we recorded whether individuals were undergoing any of the three following processes: post-juvenile molt (code 2 - when they go from juveniles to immatures), pre-adult molt (code 4 - when they go from juveniles to immatures), and their annual adult primary plumage molt (code 6).

**Data processing** – we restricted the dataset from all species to all captured individuals that were classified as immature or adults to focus on individuals that were assumed to be settled in a habitat, then recoded the data matrix to show a value of 0 when there was no evidence of complete primary plumage molt and a value of 1 when individuals were classified with plumage codes 4 and 6 (Table S6 shows sample sizes available for this analysis, and as can be noted, they could not be performed for *M. oleagineus* or *E. laniirostris*).

**Table S6.** Sample size (n) represented by the number of captures for each species according to primary plumage molt and habitat.

Species	Primary Molt	Coffee (n)	Forest (n)
<i>Mionectes olivaceus</i>	ACTIVE	8	35
	STATIC	68	318
<i>Mionectes oleagineus</i>	ACTIVE	0	5
	STATIC	30	80
<i>Turdus flavipes</i>	ACTIVE	9	53
	STATIC	254	374
<i>Turdus albicollis</i>	ACTIVE	1	33
	STATIC	19	406
<i>Ramphocelus dimidiatus</i>	ACTIVE	15	14
	STATIC	125	123
<i>Tangara gyrola</i>	ACTIVE	51	94
	STATIC	165	373
<i>Saltator maximus</i>	ACTIVE	13	17
	STATIC	90	74
<i>Saltator striatipectus</i>	ACTIVE	10	3
	STATIC	72	29
<i>Myiothlypis conspicillata</i>	ACTIVE	2	10
	STATIC	88	305
<i>Basileuterus rufifrons</i>	ACTIVE	23	6
	STATIC	190	102
<i>Myioborus miniatus</i>	ACTIVE	11	1
	STATIC	102	56
<i>Euphonia laniirostris</i>	ACTIVE	2	2
	STATIC	94	91

**Data analysis** – since primary plumage molt classification was transformed into a binary response, to test for prevalence of individuals undergoing complete primary plumage (PP) molts between habitats, we used the following set of generalized linear models (with binomial errors and logit links) and then used the same process described for age to calculate the model-averaged estimate and unconditional standard error for the habitat coefficient:

PPMolt ~ 1 (null model)  
 PPMolt ~ DAY+DAY<sup>2</sup> (considers only the quadratic effect of day of year)  
 PPMolt ~ HABITAT (considers only the effect of habitat)  
 PPMolt ~ DAY+DAY<sup>2</sup>+HABITAT (considers day of year and habitat)

When 95% confidence intervals for this estimate did not include zero, we concluded that there was a clear statistical difference in the prevalence of primary molt between habitats (positive coefficients indicated higher condition in forest, while negative coefficients indicated higher condition in coffee). We also plotted the probability of capturing an individual with active molt over one with inactive molt as a function as a function of day of year and habitat to help us interpret results.

### 1.3.9. Breeding

**Data collection** – when birds were processed we recorded the presence and state of their cloacal protuberances (ranging from 0 to 3) and their incubation patches (ranging from 0 to 5) by visually assessing them according to standardized criteria. Individuals were classified as male if their cloacal protuberance had a score of either 2 or 3, and as female if their incubation patch score had a score between 2 and 4, but both of these characteristic are also an indicator that the individual could be actively involved in breeding activities (copulation for males and egg incubating for females).

**Data processing** – we removed individuals that were not processed for evidence of breeding and then recoded our data matrix to show a value of 0 when there was no evidence of active breeding (cloacal protuberance 0-1 or incubation patch <2 or >4) and a value of 1 when individuals showed evidence of active breeding (cloacal protuberance 2-3 or incubation patch 2-4) (Table S7 shows sample sizes available for this analysis).

**Table S7.** Sample size (n) represented by the number of captures for each species according to evidence of active breeding and habitat.

Species	Breeding Evidence	Coffee (n)	Forest (n)
<i>Mionectes olivaceus</i>	NO	130	645
	YES	8	36
<i>Mionectes oleagineus</i>	NO	89	345
	YES	12	31
<i>Turdus flavipes</i>	NO	86	214
	YES	177	184
<i>Turdus albicollis</i>	NO	16	252
	YES	4	165
<i>Ramphocelus dimidiatus</i>	NO	104	99
	YES	30	30
<i>Tangara gyrola</i>	NO	82	263
	YES	131	192
<i>Saltator maximus</i>	NO	78	69
	YES	24	29
<i>Saltator striatipectus</i>	NO	56	29
	YES	21	4
<i>Myiothlypis conspicillata</i>	NO	64	208
	YES	33	100
<i>Basileuterus rufifrons</i>	NO	141	76
	YES	76	34
<i>Myioborus miniatus</i>	NO	51	17
	YES	61	39
<i>Euphonia laniirostris</i>	NO	78	58
	YES	24	41

**Data analysis** – since evidence of active breeding was transformed into a binary response, to test for prevalence of individuals undergoing active breeding between habitats, we used the following set of generalized linear models (with binomial errors and logit links) and then used the same process described for age to calculate the model-averaged estimate and unconditional standard error for the habitat coefficient:

BREEDING ~ 1 (null model)
BREEDING ~ DAY+DAY <sup>2</sup> (considers only the quadratic effect of day of year)
BREEDING ~ HABITAT (considers only the effect of habitat)
BREEDING ~ DAY+DAY <sup>2</sup> +HABITAT (considers day of year and habitat)

When 95% confidence intervals for this estimate did not include zero, we concluded that there was a clear statistical difference in the prevalence of breeding individuals between habitats (positive coefficients indicated higher condition in forest, while negative coefficients indicated higher condition in coffee). We also plotted the probability of capturing an individual with active breeding over one with inactive breeding as a function as a function of day of year and habitat to help us interpret results.

### 1.3.10. Juveniles

**Data collection and processing** – for details on data collection see the section on age. As with analyses of adults to immatures, *M. oleagineus* and *M. olivaceus* were not taken into account because aging criteria was not consistent among banders and seasons. We restricted the dataset to individuals that could be classified by age, and then recoded this data matrix to show a value of 1 when the individual was a juvenile and a value of 0 when it was an immature or adult (Table S8 shows sample sizes available for this analysis, and as can be noted, they could not be performed for *M. miniatus* or *T. albicollis*).

**Table S8.** Sample size (n) represented by the number of juveniles captured for each species according to habitat.

Species	Age category	Coffee (n)	Forest (n)
<i>Turdus flavipes</i>	2 (juvenile)	6	39
<i>Turdus albicollis</i>	2 (juvenile)	0	18
<i>Ramphocelus dimidiatus</i>	2 (juvenile)	3	3
<i>Tangara gyrola</i>	2 (juvenile)	4	22
<i>Saltator maximus</i>	2 (juvenile)	5	2
<i>Saltator striatipectus</i>	2 (juvenile)	3	3
<i>Myiothlypis conspicillata</i>	2 (juvenile)	2	5
<i>Basileuterus rufifrons</i>	2 (juvenile)	11	7
<i>Myioborus miniatus</i>	2 (juvenile)	2	0
<i>Euphonia laniirostris</i>	2 (juvenile)	1	2

**Data analysis** – since juveniles captured were transformed into a binary response, to test for prevalence of juveniles between habitats, we used the following set of generalized linear models (with binomial errors and logit links) and then used the same process described for age to calculate the model-averaged estimate and unconditional standard error for the habitat coefficient:

JUVENILES ~ 1 (null model)
JUVENILES ~ DAY+DAY <sup>2</sup> (considers only the quadratic effect of day of year)
JUVENILES ~ HABITAT (considers only the effect of habitat)
JUVENILES ~ DAY+DAY <sup>2</sup> +HABITAT (considers day of year and habitat)

All 95% confidence intervals for this estimate included zero, so we concluded that there were no statistically clear differences in the prevalence of juveniles between habitats for the eight species we tested, probably because sample sizes were too small. We also plotted the probability of capturing a juvenile over an immature or adult as a function of day of year and habitat.

#### *1.4. Level 2: Composite indexes of habitat preference and performance*

##### 1.4.1. Habitat preference

Because all analyses of preference were different, to get a quantitative value of the overall preference for a particular habitat, we subtracted the number of times evidence showed coffee was preferred, from the number of times forest was preferred.

##### 1.4.2. Habitat performance

Because all analyses of performance used the same predictors, the coefficients of the effect of habitat are directly interpretable and comparable (with larger, positive values indicating higher outcome probability in forest, larger, negative values indicating higher outcome probability in coffee, and values close to zero indicating no evidence of habitat differences). To get a quantitative estimate of the overall differences between habitats, we calculated the mean of effects sizes weighted by their standard deviation, in a manner akin to a meta-analysis using the inverse-variance method (rma function with fixed effects in package “metafor”), which creates linear effects models in which the variances of the error terms are known (Viechtbauer 2010). Confidence intervals were narrowed from 95% to 90% to decrease the probability of making a type II error (false negative) since there were many unmeasured variables affecting our response.

## **2. Results**

#### *2.1. Level 1: Separate indicators of habitat preference and performance*

This section contains numerical results and plots for all analyses carried out for the ten chosen indicators of habitat preference and performance. In plot panels species always appear in taxonomic order, in the same position and labelled according to the following species codes (first 2 letters of their genera, followed by first 2-3 letters of the species epithet):

MIOL – *Mionectes olivaceus*  
 MIOLE – *Mionectes oleagineus*  
 TUFL – *Turdus flavipes*  
 TUAL – *Turdus albicollis*  
 RADI – *Ramphocelus dimidiatus*  
 TAGY – *Tangara gyrola*  
 SAMA – *Saltator maximus*  
 SAST – *Saltator striatipectus*  
 MYCO – *Myiothlypis conspicillata*  
 BARU – *Basileuterus rufifrons*  
 MYMI – *Myioborus miniatus*  
 EULA – *Euphonia laniirostris*

Space was left blank when a variable was not tested for a species, plots with a black outline mean the response was clearly higher in forest, and plots with a red outline mean the response was clearly higher in coffee.

### 2.3.1. Occurrence

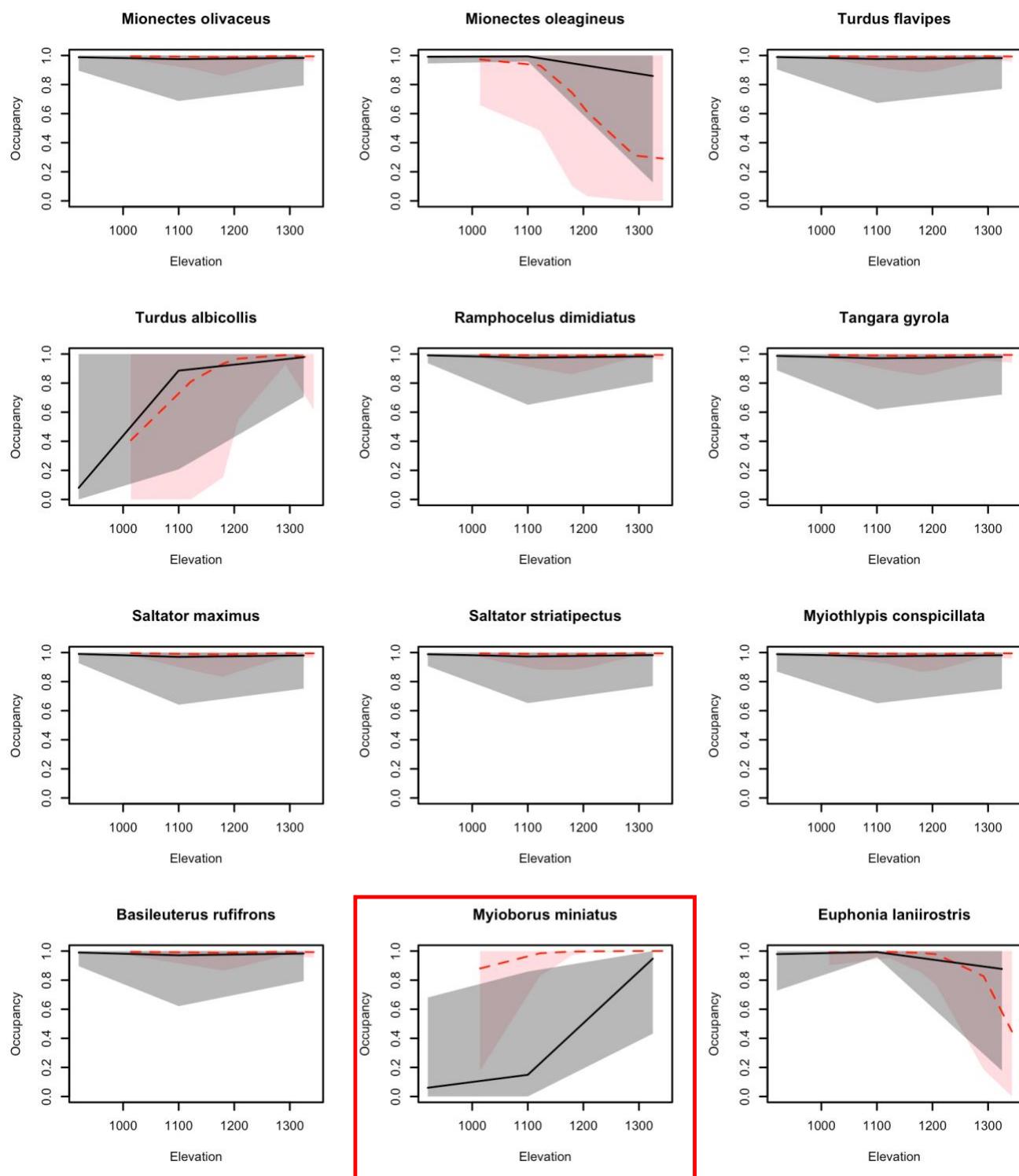
**Table S9.** Numerical outputs for occupancy modelling (**beta1.psi** is the coefficient for the effect of habitat on the probability of occurrence  $\psi$ ). Rhat is the potential scale reduction factor (at convergence, Rhat=1); successful convergence based on Rhat values (all < 1.1), n.eff is a crude measure of effective sample size, overlap0 checks if 0 falls in the parameter's 95% credible interval. f is the proportion of the posterior with the same sign as the mean; i.e., our confidence that the parameter is positive or negative. For this analysis, coffee was the reference habitat.

Coefficient	Mean	SD	2.50%	50%	97.50%	Overlap0	f	Rhat	n.eff
<i>Mionectes olivaceus</i>									
alpha.psi	10.309	7.023	-1.983	9.737	24.797	TRUE	0.947	1.001	1847
<b>beta1.psi</b>	<b>4.722</b>	<b>8.012</b>	<b>-9.874</b>	<b>4.137</b>	<b>22.068</b>	<b>TRUE</b>	<b>0.715</b>	<b>1.000</b>	<b>3000</b>
beta2.psi	-0.525	7.324	-14.672	-0.702	14.925	TRUE	0.542	1.002	1151
beta4.psi	8.215	7.082	-3.625	7.612	23.702	TRUE	0.888	1.000	3000
psi.site[1]	0.988	0.053	0.859	1.000	1.000	FALSE	1.000	1.000	3000
psi.site[2]	0.990	0.051	0.885	1.000	1.000	FALSE	1.000	1.001	3000
psi.site[3]	0.991	0.050	0.911	1.000	1.000	FALSE	1.000	1.033	1157
psi.site[4]	0.997	0.027	0.970	1.000	1.000	FALSE	1.000	1.028	1984
psi.site[5]	0.994	0.043	0.958	1.000	1.000	FALSE	1.000	1.004	2664
psi.site[6]	0.995	0.044	0.971	1.000	1.000	FALSE	1.000	1.032	2182
psi.site[7]	0.975	0.101	0.687	1.000	1.000	FALSE	1.000	1.007	2876
psi.site[8]	0.988	0.075	0.895	1.000	1.000	FALSE	1.000	1.038	1548
psi.site[9]	0.983	0.084	0.794	1.000	1.000	FALSE	1.000	1.007	1670
alpha.p	-1.639	0.381	-2.469	-1.625	-0.952	FALSE	1.000	1.000	3000
beta1.p	-0.333	0.489	-1.287	-0.353	0.669	TRUE	0.760	1.002	1948
occ.fs	9.000	0	9.000	9.000	9.000	FALSE	1.000	NaN	1
deviance	117.496	1.892	115.635	116.918	122.450	FALSE	1.000	1.001	1827
<i>Mionectes oleagineus</i>									
alpha.psi	10.39	5.659	0.426	9.997	22.47	FALSE	0.981	1.002	1234
<b>beta1.psi</b>	<b>-5.876</b>	<b>6.888</b>	<b>-17.453</b>	<b>-6.491</b>	<b>11.976</b>	<b>TRUE</b>	<b>0.864</b>	<b>1.000</b>	<b>3000</b>
beta2.psi	-6.437	5.355	-17.581	-6.126	5.224	TRUE	0.927	1.000	3000
beta4.psi	2.517	5.306	-6.157	1.826	14.882	TRUE	0.672	1.000	3000
psi.site[1]	0.743	0.281	0.101	0.855	1.000	FALSE	1.000	1.000	3000
psi.site[2]	0.611	0.339	0.032	0.65	1.000	FALSE	1.000	1.000	3000
psi.site[3]	0.931	0.136	0.484	0.992	1.000	FALSE	1.000	1.002	3000
psi.site[4]	0.312	0.370	0	0.121	1.000	FALSE	1.000	1.000	3000
psi.site[5]	0.291	0.386	0	0.057	1.000	FALSE	1.000	1.000	3000
psi.site[6]	0.973	0.098	0.658	1.000	1.000	FALSE	1.000	1.000	3000
psi.site[7]	0.994	0.047	0.959	1.000	1.000	FALSE	1.000	1.018	3000
psi.site[8]	0.991	0.066	0.945	1.000	1.000	FALSE	1.000	1.015	3000
psi.site[9]	0.859	0.239	0.127	0.985	1.000	FALSE	1.000	1.001	3000
alpha.p	-0.525	0.305	-1.113	-0.519	0.063	TRUE	0.961	1.000	3000
beta1.p	-1.095	0.555	-2.210	-1.090	-0.049	FALSE	0.980	1.001	3000
occ.fs	6.791	1.080	6.000	6.000	9.000	FALSE	1.000	1.001	2179
deviance	119.015	5.882	112.735	117.181	130.963	FALSE	1.000	1.000	2756
<i>Turdus flavipes</i>									
alpha.psi	10.397	7.109	-1.328	9.792	25.564	TRUE	0.951	1.000	3000
<b>beta1.psi</b>	<b>4.696</b>	<b>7.968</b>	<b>-9.976</b>	<b>4.270</b>	<b>22.020</b>	<b>TRUE</b>	<b>0.718</b>	<b>1.000</b>	<b>3000</b>
beta2.psi	-0.646	7.280	-15.160	-0.730	14.140	TRUE	0.546	1.001	1591
beta4.psi	8.257	7.098	-3.896	7.522	23.769	TRUE	0.896	1.001	3000
psi.site[1]	0.990	0.047	0.884	1.000	1.000	FALSE	1.000	1.008	2268
psi.site[2]	0.991	0.044	0.895	1.000	1.000	FALSE	1.000	1.021	2533
psi.site[3]	0.992	0.046	0.909	1.000	1.000	FALSE	1.000	1.013	2971
psi.site[4]	0.996	0.032	0.974	1.000	1.000	FALSE	1.000	1.006	3000
psi.site[5]	0.993	0.053	0.957	1.000	1.000	FALSE	1.000	1.035	3000
psi.site[6]	0.995	0.04	0.976	1.000	1.000	FALSE	1.000	1.076	1050
psi.site[7]	0.976	0.095	0.673	1.000	1.000	FALSE	1.000	1.021	730
psi.site[8]	0.989	0.074	0.905	1.000	1.000	FALSE	1.000	1.025	2359
psi.site[9]	0.982	0.091	0.770	1.000	1.000	FALSE	1.000	1.029	876
alpha.p	-0.252	0.297	-0.852	-0.250	0.336	TRUE	0.805	1.003	643
beta1.p	0.295	0.364	-0.418	0.288	1.032	TRUE	0.788	1.001	1501
occ.fs	9.000	0	9.000	9.000	9.000	FALSE	1.000	NaN	1
deviance	200.873	2.035	198.896	200.261	206.384	FALSE	1.000	1.000	3000
<i>Turdus albicollis</i>									
alpha.psi	10.951	5.447	2.098	10.331	22.925	FALSE	0.994	1.000	3000
<b>beta1.psi</b>	<b>1.441</b>	<b>8.477</b>	<b>-13.674</b>	<b>0.888</b>	<b>19.433</b>	<b>TRUE</b>	<b>0.543</b>	<b>1.001</b>	<b>1172</b>
beta2.psi	9.041	6.179	-1.977	8.623	22.211	TRUE	0.946	1.000	3000

beta4.psi	-3.733	5.694	-15.019	-3.648	7.692	TRUE	0.778	1.001	1224
psi.site[1]	0.934	0.195	0.152	1.000	1.000	FALSE	1.000	1.004	3000
psi.site[2]	0.968	0.121	0.546	1.000	1.000	FALSE	1.000	1.008	2026
psi.site[3]	0.811	0.348	0.001	1.000	1.000	FALSE	1.000	1.000	3000
psi.site[4]	0.993	0.046	0.927	1.000	1.000	FALSE	1.000	1.007	3000
psi.site[5]	0.975	0.106	0.618	1.000	1.000	FALSE	1.000	1.014	1533
psi.site[6]	0.407	0.455	0	0.070	1.000	FALSE	1.000	1.000	3000
psi.site[7]	0.886	0.213	0.207	0.993	1.000	FALSE	1.000	1.000	3000
psi.site[8]	0.080	0.235	0	0	1.000	FALSE	1.000	1.001	3000
psi.site[9]	0.978	0.096	0.705	1.000	1.000	FALSE	1.000	1.000	3000
alpha.p	-1.015	0.417	-1.881	-0.995	-0.243	FALSE	0.994	1.000	3000
beta1.p	-2.888	0.910	-4.847	-2.815	-1.231	FALSE	0.999	1.000	2815
occ.fs	7.147	0.965	5.000	7.000	9.000	FALSE	1.000	1.000	3000
deviance	59.146	2.868	55.376	58.377	67.486	FALSE	1.000	1.001	3000
<i>Ramphocelus dimidiatus</i>									
alpha.psi	10.337	7.017	-1.574	9.814	25.804	TRUE	0.949	1.002	892
<b>beta1.psi</b>	<b>4.549</b>	<b>8.034</b>	<b>-10.545</b>	<b>4.092</b>	<b>21.516</b>	<b>TRUE</b>	<b>0.718</b>	<b>1.002</b>	<b>1316</b>
beta2.psi	-0.937	7.161	-15.282	-0.758	12.911	TRUE	0.545	1.000	3000
beta4.psi	8.011	6.879	-3.559	7.481	22.420	TRUE	0.888	1.002	1026
psi.site[1]	0.989	0.053	0.860	1.000	1.000	FALSE	1.000	1.005	3000
psi.site[2]	0.990	0.049	0.888	1.000	1.000	FALSE	1.000	1.001	3000
psi.site[3]	0.991	0.048	0.898	1.000	1.000	FALSE	1.000	1.012	3000
psi.site[4]	0.996	0.029	0.976	1.000	1.000	FALSE	1.000	1.021	3000
psi.site[5]	0.994	0.046	0.959	1.000	1.000	FALSE	1.000	1.008	2622
psi.site[6]	0.995	0.039	0.978	1.000	1.000	FALSE	1.000	1.054	1403
psi.site[7]	0.975	0.098	0.651	1.000	1.000	FALSE	1.000	1.024	755
psi.site[8]	0.991	0.062	0.937	1.000	1.000	FALSE	1.000	1.013	1922
psi.site[9]	0.984	0.086	0.809	1.000	1.000	FALSE	1.000	1.005	3000
alpha.p	0.172	0.290	-0.391	0.166	0.735	TRUE	0.723	1.000	3000
beta1.p	0.477	0.354	-0.215	0.476	1.168	TRUE	0.908	1.000	3000
occ.fs	9.000	0	9.000	9.000	9.000	FALSE	1.000	NaN	1
deviance	191.733	1.921	189.812	191.158	197.028	FALSE	1.000	1.002	2467
<i>Tangara gyrola</i>									
alpha.psi	10.129	7.048	-1.612	9.477	25.387	TRUE	0.943	1.000	2819
<b>beta1.psi</b>	<b>4.804</b>	<b>7.928</b>	<b>-10.813</b>	<b>4.448</b>	<b>21.679</b>	<b>TRUE</b>	<b>0.730</b>	<b>1.000</b>	<b>3000</b>
beta2.psi	-0.466	7.412	-15.270	-0.614	13.941	TRUE	0.532	1.001	1198
beta4.psi	8.335	7.043	-3.781	7.777	23.950	TRUE	0.896	1.000	3000
psi.site[1]	0.988	0.056	0.854	1.000	1.000	FALSE	1.000	1.012	2896
psi.site[2]	0.989	0.052	0.874	1.000	1.000	FALSE	1.000	1.004	3000
psi.site[3]	0.989	0.059	0.889	1.000	1.000	FALSE	1.000	1.009	3000
psi.site[4]	0.995	0.038	0.957	1.000	1.000	FALSE	1.000	1.013	2733
psi.site[5]	0.994	0.046	0.940	1.000	1.000	FALSE	1.000	1.011	2830
psi.site[6]	0.994	0.047	0.972	1.000	1.000	FALSE	1.000	1.006	3000
psi.site[7]	0.971	0.113	0.619	1.000	1.000	FALSE	1.000	1.000	3000
psi.site[8]	0.987	0.081	0.888	1.000	1.000	FALSE	1.000	1.010	2022
psi.site[9]	0.980	0.090	0.721	1.000	1.000	FALSE	1.000	1.010	2523
alpha.p	2.025	0.452	1.200	2.003	2.988	FALSE	1.000	1.000	3000
beta1.p	-0.666	0.519	-1.754	-0.660	0.307	TRUE	0.902	1.000	3000
occ.fs	9.000	0	9.000	9.000	9.000	FALSE	1.000	NaN	1
deviance	136.455	2.085	134.471	135.804	142.243	FALSE	1.000	1.001	3000
<i>Saltator maximus</i>									
alpha.psi	10.092	6.928	-1.751	9.413	24.977	TRUE	0.941	1.000	3000
<b>beta1.psi</b>	<b>4.736</b>	<b>7.834</b>	<b>-9.497</b>	<b>4.297</b>	<b>21.324</b>	<b>TRUE</b>	<b>0.715</b>	<b>1.000</b>	<b>3000</b>
beta2.psi	-0.454	7.294	-15.222	-0.466	13.811	TRUE	0.524	1.000	3000
beta4.psi	8.224	6.979	-3.911	7.469	23.385	TRUE	0.898	1.001	2427
psi.site[1]	0.987	0.061	0.833	1.000	1.000	FALSE	1.000	1.011	3000
psi.site[2]	0.989	0.055	0.869	1.000	1.000	FALSE	1.000	1.001	3000
psi.site[3]	0.989	0.058	0.879	1.000	1.000	FALSE	1.000	1.027	2715
psi.site[4]	0.996	0.030	0.967	1.000	1.000	FALSE	1.000	1.008	2046
psi.site[5]	0.994	0.047	0.963	1.000	1.000	FALSE	1.000	1.012	3000
psi.site[6]	0.996	0.039	0.980	1.000	1.000	FALSE	1.000	1.029	3000
psi.site[7]	0.970	0.111	0.641	1.000	1.000	FALSE	1.000	1.002	3000
psi.site[8]	0.99	0.072	0.927	1.000	1.000	FALSE	1.000	1.039	1112
psi.site[9]	0.98	0.096	0.752	1.000	1.000	FALSE	1.000	1.011	1892
alpha.p	-0.616	0.297	-1.227	-0.609	-0.049	FALSE	0.983	1.001	1579
beta1.p	0.695	0.363	-0.011	0.681	1.427	TRUE	0.973	1.001	2424
occ.fs	9.000	0	9.000	9.000	9.000	FALSE	1.000	NaN	1
deviance	197.332	2.046	195.369	196.692	202.784	FALSE	1.000	1.000	3000
<i>Saltator striatipectus</i>									

alpha.psi	10.238	7.122	-1.993	9.700	25.489	TRUE	0.943	1.001	1413
<b>beta1.psi</b>	<b>4.658</b>	<b>7.964</b>	<b>-9.938</b>	<b>4.279</b>	<b>21.617</b>	<b>TRUE</b>	<b>0.705</b>	<b>1.000</b>	<b>3000</b>
beta2.psi	-0.348	7.362	-14.890	-0.484	14.790	TRUE	0.531	1.000	3000
beta4.psi	8.129	7.021	-3.286	7.290	24.162	TRUE	0.892	1.000	3000
psi.site[1]	0.988	0.057	0.881	1.000	1.000	FALSE	1.000	1.031	1099
psi.site[2]	0.990	0.050	0.895	1.000	1.000	FALSE	1.000	1.042	1098
psi.site[3]	0.990	0.057	0.881	1.000	1.000	FALSE	1.000	1.018	2133
psi.site[4]	0.996	0.025	0.975	1.000	1.000	FALSE	1.000	1.023	2413
psi.site[5]	0.994	0.048	0.970	1.000	1.000	FALSE	1.000	1.012	3000
psi.site[6]	0.994	0.053	0.966	1.000	1.000	FALSE	1.000	1.014	3000
psi.site[7]	0.973	0.098	0.652	1.000	1.000	FALSE	1.000	1.008	3000
psi.site[8]	0.988	0.075	0.907	1.000	1.000	FALSE	1.000	1.040	1174
psi.site[9]	0.982	0.086	0.770	1.000	1.000	FALSE	1.000	1.008	2320
alpha.p	-0.343	0.297	-0.913	-0.343	0.228	TRUE	0.876	1.000	3000
beta1.p	0.470	0.366	-0.234	0.468	1.182	TRUE	0.897	1.000	3000
occ.fs	9.000	0	9.000	9.000	9.000	FALSE	1.000	NaN	1
deviance	199.976	2.027	197.968	199.428	205.178	FALSE	1.000	1.003	3000
<i>Myiothlypis conspicillata</i>									
alpha.psi	10.174	6.985	-1.601	9.563	25.194	TRUE	0.948	1.000	3000
<b>beta1.psi</b>	<b>4.643</b>	<b>7.713</b>	<b>-9.336</b>	<b>4.225</b>	<b>20.629</b>	<b>TRUE</b>	<b>0.719</b>	<b>1.001</b>	<b>1444</b>
beta2.psi	-0.398	7.256	-14.189	-0.525	14.119	TRUE	0.533	1.000	3000
beta4.psi	7.994	6.889	-3.824	7.421	22.893	TRUE	0.882	1.000	3000
psi.site[1]	0.989	0.050	0.867	1.000	1.000	FALSE	1.000	1.002	3000
psi.site[2]	0.989	0.052	0.877	1.000	1.000	FALSE	1.000	1.001	3000
psi.site[3]	0.992	0.043	0.921	1.000	1.000	FALSE	1.000	1.029	1895
psi.site[4]	0.996	0.031	0.969	1.000	1.000	FALSE	1.000	1.017	3000
psi.site[5]	0.994	0.048	0.957	1.000	1.000	FALSE	1.000	1.027	2602
psi.site[6]	0.995	0.045	0.975	1.000	1.000	FALSE	1.000	1.024	3000
psi.site[7]	0.974	0.104	0.651	1.000	1.000	FALSE	1.000	1.001	3000
psi.site[8]	0.988	0.072	0.869	1.000	1.000	FALSE	1.000	1.028	1344
psi.site[9]	0.981	0.089	0.750	1.000	1.000	FALSE	1.000	1.002	3000
alpha.p	-0.426	0.296	-1.011	-0.429	0.142	TRUE	0.927	1.001	3000
beta1.p	-0.139	0.365	-0.851	-0.140	0.580	TRUE	0.644	1.001	1538
occ.fs	9.000	0	9.000	9.000	9.000	FALSE	1.000	NaN	1
deviance	192.380	1.977	190.447	191.783	197.668	FALSE	1.000	1.001	1932
<i>Basileuterus rufifrons</i>									
alpha.psi	10.115	7.010	-1.775	9.586	25.039	TRUE	0.940	1.001	1245
<b>beta1.psi</b>	<b>4.710</b>	<b>7.687</b>	<b>-9.389</b>	<b>4.269</b>	<b>20.935</b>	<b>TRUE</b>	<b>0.720</b>	<b>1.000</b>	<b>3000</b>
beta2.psi	-0.670	7.331	-15.371	-0.828	13.952	TRUE	0.550	1.000	3000
beta4.psi	8.156	6.853	-3.664	7.712	22.937	TRUE	0.893	1.000	3000
psi.site[1]	0.988	0.054	0.866	1.000	1.000	FALSE	1.000	1.022	3000
psi.site[2]	0.990	0.054	0.888	1.000	1.000	FALSE	1.000	1.015	3000
psi.site[3]	0.991	0.053	0.902	1.000	1.000	FALSE	1.000	1.018	3000
psi.site[4]	0.996	0.027	0.971	1.000	1.000	FALSE	1.000	1.038	2480
psi.site[5]	0.993	0.053	0.955	1.000	1.000	FALSE	1.000	1.017	3000
psi.site[6]	0.995	0.040	0.969	1.000	1.000	FALSE	1.000	1.004	3000
psi.site[7]	0.971	0.110	0.621	1.000	1.000	FALSE	1.000	1.002	3000
psi.site[8]	0.990	0.063	0.896	1.000	1.000	FALSE	1.000	1.000	3000
psi.site[9]	0.983	0.084	0.794	1.000	1.000	FALSE	1.000	1.004	3000
alpha.p	-0.090	0.287	-0.667	-0.082	0.477	TRUE	0.623	1.004	472
beta1.p	1.575	0.381	0.816	1.569	2.344	FALSE	1.000	1.003	813
occ.fs	9.000	0	9.000	9.000	9.000	FALSE	1.000	NaN	1
deviance	161.068	1.961	159.158	160.460	166.152	FALSE	1.000	1.000	3000
<i>Myioborus miniatus</i>									
alpha.psi	-0.308	4.526	-10.303	0.049	7.698	TRUE	0.494	1.000	3000
<b>beta1.psi</b>	<b>14.423</b>	<b>6.203</b>	<b>4.070</b>	<b>13.916</b>	<b>27.918</b>	<b>FALSE</b>	<b>0.999</b>	<b>1.000</b>	<b>3000</b>
beta2.psi	9.473	4.937	2.066	8.649	21.098	FALSE	0.998	1.000	3000
beta4.psi	1.392	4.257	-7.371	1.581	9.377	TRUE	0.648	1.000	2303
psi.site[1]	0.997	0.024	0.971	1.000	1.000	FALSE	1.000	1.024	3000
psi.site[2]	0.998	0.014	0.989	1.000	1.000	FALSE	1.000	1.076	2343
psi.site[3]	0.985	0.064	0.829	1.000	1.000	FALSE	1.000	1.014	1751
psi.site[4]	1.000	0.005	0.999	1.000	1.000	FALSE	1.000	1.075	2373
psi.site[5]	1.000	0.006	1.000	1.000	1.000	FALSE	1.000	1.055	3000
psi.site[6]	0.880	0.217	0.178	0.989	1.000	FALSE	1.000	1.000	3000
psi.site[7]	0.149	0.242	0	0.022	0.859	FALSE	1.000	1.002	2305
psi.site[8]	0.060	0.166	0	0	0.681	FALSE	1.000	1.001	3000
psi.site[9]	0.948	0.148	0.432	1.000	1.000	FALSE	1.000	1.000	3000
alpha.p	1.600	0.688	0.387	1.537	3.143	FALSE	0.996	1.000	3000
beta1.p	-0.994	0.720	-2.607	-0.943	0.290	TRUE	0.930	1.000	3000

occ.fs	7.000	0	7.000	7.000	7.000	FALSE	1.000	NaN	1
deviance	142.343	2.112	140.294	141.683	148.352	FALSE	1.000	1.001	3000
<i>Euphonia laniirostris</i>									
alpha.psi	13.849	6.533	1.848	13.415	27.187	FALSE	0.987	1.001	2269
<b>beta1.psi</b>	<b>-2.631</b>	<b>6.304</b>	<b>-13.398</b>	<b>-3.195</b>	<b>13.787</b>	<b>TRUE</b>	<b>0.743</b>	<b>1.000</b>	<b>3000</b>
beta2.psi	-6.037	5.732	-17.464	-5.867	7.372	TRUE	0.909	1.000	2499
beta4.psi	0.020	6.104	-9.571	-0.945	16.388	TRUE	0.429	1.002	3000
psi.site[1]	0.987	0.049	0.853	1.000	1.000	FALSE	1.000	1.010	3000
psi.site[2]	0.978	0.069	0.764	1.000	1.000	FALSE	1.000	1.002	3000
psi.site[3]	0.995	0.031	0.953	1.000	1.000	FALSE	1.000	1.035	2353
psi.site[4]	0.827	0.234	0.185	0.945	1.000	FALSE	1.000	1.001	2053
psi.site[5]	0.447	0.384	0.001	0.348	1.000	FALSE	1.000	1.000	3000
psi.site[6]	0.991	0.052	0.902	1.000	1.000	FALSE	1.000	1.034	974
psi.site[7]	0.993	0.054	0.958	1.000	1.000	FALSE	1.000	1.022	3000
psi.site[8]	0.979	0.097	0.728	1.000	1.000	FALSE	1.000	1.003	3000
psi.site[9]	0.877	0.222	0.177	0.991	1.000	FALSE	1.000	1.000	3000
alpha.p	-0.900	0.327	-1.547	-0.893	-0.289	FALSE	0.997	1.001	2673
beta1.p	0.053	0.434	-0.813	0.050	0.884	TRUE	0.545	1.000	3000
occ.fs	8.192	0.394	8.000	8.000	9.000	FALSE	1.000	1.000	3000
deviance	161.461	4.710	157.403	159.345	172.170	FALSE	1.000	1.000	3000



**Figure S3.** Occupancy estimates for each species according to elevation (masl) and habitat (coffee in red and forest in black). Lines are included for heuristic purposes and show a fitted regression model of quadratic effects of elevation + habitat (solid lines) together with their 95% confidence intervals (dotted lines).

### 2.3.2. Abundance

**Table S10.** Numerical outputs for abundance modelling (N is the total number of individuals estimated for the whole study period). Rhat is the potential scale reduction factor (at convergence, Rhat=1); successful convergence based on Rhat values (all < 1.1), n.eff is a crude measure of effective sample size, over0 checks if 0 falls in the parameter's 95% credible interval. f is the proportion of the posterior with the same sign as the mean; i.e., our confidence that the parameter is positive or negative.

<b>A) <i>Mionectes olivaceus</i> - Closed Population Model</b>												
<b>COFFEE</b>												
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f	
N	1257	571	548	862	1132	1482	2989	1.10	35	NO	1.00	
p.est[1]	0.0025	0.0012	0.00078	0.001583	0.0022	0.0031	0.0055	1.02	85	NO	1.00	
p.est[2]	0.0172	0.0071	0.0061	0.0121	0.0159	0.0213	0.0333	1.04	63	NO	1.00	
p.est[3]	0.0052	0.0023	0.0018	0.0036	0.0048	0.0065	0.0109	1.03	67	NO	1.00	
p.est[4]	0.0132	0.0057	0.0046	0.0091	0.0123	0.0162	0.0266	1.03	72	NO	1.00	
p.est[5]	0.0068	0.0029	0.0024	0.0048	0.0063	0.0084	0.0138	1.03	64	NO	1.00	
p.est[6]	0.0015	0.0008	0.0004	0.0009	0.0013	0.0019	0.0037	1.02	104	NO	1.00	
p.est[7]	0.0174	0.0072	0.0061	0.0123	0.0161	0.0216	0.0339	1.03	64	NO	1.00	
p.est[8]	0.0110	0.0050	0.0036	0.0074	0.0101	0.0137	0.0232	1.03	79	NO	1.00	
p.est[9]	0.0138	0.0057	0.0049	0.0098	0.0128	0.0171	0.0273	1.04	62	NO	1.00	
p.est[10]	0.0108	0.0045	0.0038	0.0076	0.0101	0.0134	0.0215	1.04	62	NO	1.00	
p.est[11]	0.0157	0.0065	0.0055	0.0112	0.0145	0.0194	0.0308	1.04	63	NO	1.00	
p.est[12]	0.0021	0.0011	0.0007	0.0014	0.0019	0.0027	0.0050	1.02	90	NO	1.00	
mean.p	0.0124	0.0052	0.0044	0.0088	0.0115	0.0154	0.0247	1.04	62	NO	1.00	
beta1	0.8945	0.1586	0.5979	0.7834	0.8923	1.0003	1.2126	1.00	3000	NO	1.00	
beta2	-0.5811	0.1277	-0.8428	-0.6648	-0.5787	-0.4948	-0.3322	1.00	1078	NO	1.00	
psi	0.3057	0.1387	0.1320	0.2102	0.2748	0.3617	0.7198	1.10	34	NO	1.00	
deviance	1369	102	1184	1297	1365	1433	1607	1.07	41	NO	1.00	
<b>FOREST</b>												
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f	
N	1173	91	1015	1110	1170	1231	1367	1.01	419	NO	1.00	
p.est[1]	0.0409	0.0038	0.0341	0.0383	0.0408	0.0434	0.0485	1.00	721	NO	1.00	
p.est[2]	0.0453	0.0045	0.0368	0.0422	0.0451	0.0483	0.0547	1.00	764	NO	1.00	
p.est[3]	0.0396	0.0037	0.0328	0.0371	0.0394	0.0419	0.0470	1.00	758	NO	1.00	
p.est[4]	0.0439	0.0043	0.0357	0.0409	0.0437	0.0466	0.0529	1.00	506	NO	1.00	
p.est[5]	0.0367	0.0039	0.0296	0.0340	0.0365	0.0392	0.0446	1.00	985	NO	1.00	
p.est[6]	0.0437	0.0042	0.0360	0.0408	0.0436	0.0465	0.0522	1.00	735	NO	1.00	
p.est[7]	0.0377	0.0055	0.0277	0.0338	0.0375	0.0412	0.0489	1.00	773	NO	1.00	
p.est[8]	0.0435	0.0043	0.0354	0.0405	0.0433	0.0463	0.0526	1.00	502	NO	1.00	
p.est[9]	0.0363	0.0039	0.0292	0.0335	0.0361	0.0389	0.0443	1.00	1036	NO	1.00	
p.est[10]	0.0465	0.0048	0.0376	0.0431	0.0464	0.0497	0.0565	1.00	737	NO	1.00	
p.est[11]	0.0367	0.0039	0.0296	0.0339	0.0365	0.0392	0.0446	1.00	990	NO	1.00	
p.est[12]	0.0447	0.0044	0.0364	0.0416	0.0446	0.0476	0.0538	1.00	525	NO	1.00	
mean.p	0.0458	0.0047	0.0372	0.0426	0.0456	0.0489	0.0556	1.00	767	NO	1.00	
beta1	0.0905	0.0506	-0.0106	0.0577	0.0902	0.1226	0.1906	1.00	3000	YES	0.96	
beta2	-0.1232	0.0642	-0.2532	-0.1659	-0.1222	-0.0802	0.0012	1.00	3000	YES	0.97	
psi	0.2632	0.0215	0.2238	0.2482	0.2620	0.2763	0.3089	1.00	429	NO	1.00	
deviance	4812	91	4643	4750	4811	4871	4994	1.00	435	NO	1.00	

<b>B) <i>Mionectes oleagineus</i> - Jolly-Seber Model</b>												
<b>COFFEE</b>												
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f	
psi	0.1278	0.0322	0.0801	0.1047	0.1223	0.1449	0.2057	1.01	326	0	1.00	
Ncoffee	253	65	158	206	241	287	411	1.01	312	0	1.00	
p[1]	0.1357	0.0400	0.0706	0.1069	0.1315	0.1597	0.2259	1.01	529	0	1.00	
p[2]	0.2595	0.0873	0.1226	0.1958	0.2480	0.3110	0.4609	1.00	1103	0	1.00	
p[3]	0.1086	0.0332	0.0554	0.0846	0.1047	0.1282	0.1843	1.01	577	0	1.00	
p[4]	0.1972	0.0568	0.1036	0.1564	0.1912	0.2315	0.3247	1.00	612	0	1.00	
p[5]	0.0965	0.0352	0.0431	0.0712	0.0914	0.1162	0.1792	1.00	1063	0	1.00	
p[6]	0.1776	0.0500	0.0944	0.1417	0.1728	0.2082	0.2891	1.01	544	0	1.00	
p[7]	0.3067	0.1153	0.1283	0.2211	0.2908	0.3763	0.5716	1.00	1751	0	1.00	
p[8]	0.1059	0.0324	0.0541	0.0825	0.1021	0.1251	0.1799	1.01	583	0	1.00	
p[9]	0.0988	0.0418	0.0386	0.0689	0.0917	0.1209	0.1998	1.00	1602	0	1.00	
p[10]	0.1283	0.0384	0.0662	0.1006	0.1241	0.1512	0.2152	1.01	541	0	1.00	
p[11]	0.1455	0.0420	0.0764	0.1152	0.1413	0.1709	0.2398	1.01	519	0	1.00	
p[12]	0.1184	0.0360	0.0606	0.0924	0.1143	0.1397	0.2004	1.01	558	0	1.00	
p[13]	0.2720	0.0945	0.1247	0.2027	0.2592	0.3278	0.4902	1.00	1251	0	1.00	
alpha.p	-1.8933	0.3431	-2.5775	-2.1227	-1.8877	-1.6602	-1.2317	1.01	474	0	1.00	
beta1.p	-0.5379	0.2261	-1.0060	-0.6850	-0.5286	-0.3812	-0.1211	1.00	30571	0	1.00	

beta2.p	0.1776	0.1936	-0.1972	0.0464	0.1760	0.3061	0.5638	1.00	22501	1	0.82
<b>phi[1]</b>	<b>0.5001</b>	<b>0.2882</b>	<b>0.0246</b>	<b>0.2507</b>	<b>0.5005</b>	<b>0.7492</b>	<b>0.9752</b>	<b>1.00</b>	<b>3.00E+05</b>	<b>0</b>	<b>1.00</b>
<b>phi[2]</b>	<b>0.6205</b>	<b>0.2437</b>	<b>0.1311</b>	<b>0.4368</b>	<b>0.6472</b>	<b>0.8271</b>	<b>0.9827</b>	<b>1.00</b>	<b>5844</b>	<b>0</b>	<b>1.00</b>
<b>phi[3]</b>	<b>0.7641</b>	<b>0.1692</b>	<b>0.3814</b>	<b>0.6528</b>	<b>0.7944</b>	<b>0.9037</b>	<b>0.9907</b>	<b>1.00</b>	<b>89325</b>	<b>0</b>	<b>1.00</b>
<b>phi[4]</b>	<b>0.5133</b>	<b>0.1799</b>	<b>0.2149</b>	<b>0.3784</b>	<b>0.4942</b>	<b>0.6310</b>	<b>0.9077</b>	<b>1.00</b>	<b>8655</b>	<b>0</b>	<b>1.00</b>
<b>phi[5]</b>	<b>0.8111</b>	<b>0.1410</b>	<b>0.4840</b>	<b>0.7217</b>	<b>0.8393</b>	<b>0.9266</b>	<b>0.9931</b>	<b>1.00</b>	<b>209783</b>	<b>0</b>	<b>1.00</b>
<b>phi[6]</b>	<b>0.7191</b>	<b>0.1705</b>	<b>0.3692</b>	<b>0.5965</b>	<b>0.7324</b>	<b>0.8584</b>	<b>0.9842</b>	<b>1.00</b>	<b>49913</b>	<b>0</b>	<b>1.00</b>
<b>phi[7]</b>	<b>0.8720</b>	<b>0.1085</b>	<b>0.5983</b>	<b>0.8148</b>	<b>0.9003</b>	<b>0.9568</b>	<b>0.9961</b>	<b>1.00</b>	<b>182527</b>	<b>0</b>	<b>1.00</b>
<b>phi[8]</b>	<b>0.8105</b>	<b>0.1454</b>	<b>0.4651</b>	<b>0.7221</b>	<b>0.8419</b>	<b>0.9285</b>	<b>0.9933</b>	<b>1.00</b>	<b>11322</b>	<b>0</b>	<b>1.00</b>
<b>phi[9]</b>	<b>0.7473</b>	<b>0.1649</b>	<b>0.3964</b>	<b>0.6319</b>	<b>0.7668</b>	<b>0.8838</b>	<b>0.9881</b>	<b>1.00</b>	<b>11789</b>	<b>0</b>	<b>1.00</b>
<b>phi[10]</b>	<b>0.8212</b>	<b>0.1389</b>	<b>0.4896</b>	<b>0.7378</b>	<b>0.8518</b>	<b>0.9337</b>	<b>0.9938</b>	<b>1.00</b>	<b>75726</b>	<b>0</b>	<b>1.00</b>
<b>phi[11]</b>	<b>0.4608</b>	<b>0.1911</b>	<b>0.1588</b>	<b>0.3165</b>	<b>0.4347</b>	<b>0.5821</b>	<b>0.8923</b>	<b>1.00</b>	<b>51732</b>	<b>0</b>	<b>1.00</b>
<b>phi[12]</b>	<b>0.4021</b>	<b>0.2357</b>	<b>0.0563</b>	<b>0.2143</b>	<b>0.3624</b>	<b>0.5602</b>	<b>0.9204</b>	<b>1.00</b>	<b>35021</b>	<b>0</b>	<b>1.00</b>
b[1]	0.0895	0.0511	0.0184	0.0516	0.0801	0.1174	0.2136	1.00	6144	0	1.00
b[2]	0.1593	0.0785	0.0264	0.1020	0.1520	0.2093	0.3302	1.00	28650	0	1.00
b[3]	0.1369	0.0811	0.0100	0.0747	0.1288	0.1890	0.3159	1.00	4336	0	1.00
b[4]	0.0377	0.0321	0.0012	0.0131	0.0295	0.0540	0.1192	1.00	24273	0	1.00
b[5]	0.0217	0.0218	0.0005	0.0062	0.0150	0.0299	0.0803	1.00	23909	0	1.00
b[6]	0.0205	0.0206	0.0005	0.0059	0.0142	0.0284	0.0760	1.00	54997	0	1.00
b[7]	0.0784	0.0588	0.0041	0.0340	0.0655	0.1094	0.2254	1.00	11757	0	1.00
b[8]	0.2026	0.0939	0.0315	0.1353	0.1988	0.2652	0.3962	1.00	9946	0	1.00
b[9]	0.0661	0.0569	0.0021	0.0224	0.0513	0.0947	0.2115	1.00	26684	0	1.00
b[10]	0.0956	0.0707	0.0039	0.0392	0.0819	0.1378	0.2627	1.00	22402	0	1.00
b[11]	0.0422	0.0369	0.0013	0.0142	0.0324	0.0601	0.1370	1.00	126937	0	1.00
b[12]	0.0493	0.0424	0.0016	0.0174	0.0383	0.0698	0.1584	1.00	162233	0	1.00
B[1]	23	15	5	12	19	30	61	1.01	831	0	1.00
B[2]	40	21	7	25	37	52	91	1.00	1292	0	1.00
B[3]	36	25	2	18	31	48	97	1.01	774	0	1.00
B[4]	9	8	0	3	7	13	29	1.00	8030	1	1.00
B[5]	5	5	0	1	3	7	19	1.00	6909	1	1.00
B[6]	4	5	0	1	3	6	18	1.00	5662	1	1.00
B[7]	19	15	1	8	16	26	56	1.00	13117	0	1.00
B[8]	53	29	8	33	49	69	121	1.00	2267	0	1.00
B[9]	17	16	0	5	12	23	58	1.00	6564	1	1.00
B[10]	25	21	0	9	20	35	78	1.00	3166	1	1.00
B[11]	10	10	0	3	7	14	38	1.00	5700	1	1.00
B[12]	13	13	0	4	9	17	47	1.00	4221	1	1.00
N[1]	<b>23</b>	<b>15</b>	<b>5</b>	<b>12</b>	<b>19</b>	<b>30</b>	<b>61</b>	<b>1.01</b>	<b>831</b>	<b>0</b>	<b>1.00</b>
N[2]	<b>55</b>	<b>22</b>	<b>25</b>	<b>39</b>	<b>51</b>	<b>66</b>	<b>108</b>	<b>1.01</b>	<b>747</b>	<b>0</b>	<b>1.00</b>
N[3]	<b>77</b>	<b>26</b>	<b>40</b>	<b>59</b>	<b>73</b>	<b>91</b>	<b>144</b>	<b>1.01</b>	<b>351</b>	<b>0</b>	<b>1.00</b>
N[4]	<b>47</b>	<b>16</b>	<b>25</b>	<b>36</b>	<b>44</b>	<b>55</b>	<b>86</b>	<b>1.00</b>	<b>778</b>	<b>0</b>	<b>1.00</b>
N[5]	<b>43</b>	<b>15</b>	<b>22</b>	<b>32</b>	<b>40</b>	<b>50</b>	<b>79</b>	<b>1.00</b>	<b>842</b>	<b>0</b>	<b>1.00</b>
N[6]	<b>35</b>	<b>13</b>	<b>16</b>	<b>26</b>	<b>33</b>	<b>42</b>	<b>68</b>	<b>1.00</b>	<b>891</b>	<b>0</b>	<b>1.00</b>
N[7]	<b>50</b>	<b>19</b>	<b>25</b>	<b>36</b>	<b>46</b>	<b>59</b>	<b>98</b>	<b>1.00</b>	<b>1437</b>	<b>0</b>	<b>1.00</b>
N[8]	<b>94</b>	<b>30</b>	<b>50</b>	<b>73</b>	<b>89</b>	<b>110</b>	<b>166</b>	<b>1.00</b>	<b>830</b>	<b>0</b>	<b>1.00</b>
N[9]	<b>87</b>	<b>31</b>	<b>41</b>	<b>65</b>	<b>82</b>	<b>103</b>	<b>162</b>	<b>1.00</b>	<b>852</b>	<b>0</b>	<b>1.00</b>
N[10]	<b>96</b>	<b>31</b>	<b>51</b>	<b>74</b>	<b>91</b>	<b>112</b>	<b>172</b>	<b>1.01</b>	<b>619</b>	<b>0</b>	<b>1.00</b>
N[11]	<b>53</b>	<b>24</b>	<b>21</b>	<b>36</b>	<b>49</b>	<b>65</b>	<b>113</b>	<b>1.00</b>	<b>951</b>	<b>0</b>	<b>1.00</b>
N[12]	<b>33</b>	<b>20</b>	<b>8</b>	<b>19</b>	<b>28</b>	<b>42</b>	<b>82</b>	<b>1.00</b>	<b>1305</b>	<b>0</b>	<b>1.00</b>
gamma[1]	0.0115	0.0076	0.0021	0.0061	0.0098	0.0150	0.0309	1.01	927	0	1.00
gamma[2]	0.0202	0.0109	0.0035	0.0125	0.0186	0.0261	0.0462	1.00	1331	0	1.00
gamma[3]	0.0182	0.0126	0.0012	0.0091	0.0160	0.0246	0.0497	1.01	785	0	1.00
gamma[4]	0.0049	0.0043	0.0002	0.0017	0.0038	0.0070	0.0160	1.00	7261	0	1.00
gamma[5]	0.0029	0.0030	0.0001	0.0008	0.0020	0.0040	0.0110	1.00	7371	0	1.00
gamma[6]	0.0027	0.0029	0.0001	0.0008	0.0019	0.0038	0.0105	1.00	5224	0	1.00
gamma[7]	0.0103	0.0079	0.0006	0.0045	0.0085	0.0140	0.0303	1.00	10270	0	1.00
gamma[8]	0.0280	0.0156	0.0040	0.0170	0.0257	0.0364	0.0648	1.00	1996	0	1.00
gamma[9]	0.0094	0.0088	0.0003	0.0030	0.0069	0.0130	0.0322	1.00	5656	0	1.00
gamma[10]	0.0139	0.0118	0.0005	0.0052	0.0110	0.0193	0.0440	1.00	2703	0	1.00
gamma[11]	0.0062	0.0061	0.0002	0.0019	0.0045	0.0086	0.0225	1.00	4795	0	1.00
gamma[12]	0.0075	0.0076	0.0002	0.0023	0.0053	0.0102	0.0275	1.00	3441	0	1.00
deviance	530	51	437	495	528	563	637	1.01	400	0	1.00
<b>FOREST</b>											
<b>par</b>	<b>mean</b>	<b>sd</b>	<b>2.50%</b>	<b>25%</b>	<b>50%</b>	<b>75%</b>	<b>97.50%</b>	<b>Rhat</b>	<b>n.eff</b>	<b>over0</b>	<b>f</b>
psi	0.3058	0.0382	0.2377	0.2793	0.3036	0.3294	0.3862	1.02	218	0	1.00
<b>Nforest</b>	<b>674</b>	<b>83</b>	<b>527</b>	<b>616</b>	<b>669</b>	<b>725</b>	<b>848</b>	<b>1.02</b>	<b>204</b>	<b>0</b>	<b>1.00</b>
p[1]	0.1512	0.0278	0.1039	0.1316	0.1488	0.1682	0.2129	1.01	385	0	1.00
p[2]	0.2197	0.0355	0.1569	0.1947	0.2174	0.2421	0.2956	1.01	489	0	1.00
p[3]	0.1593	0.0287	0.1101	0.1390	0.1569	0.1769	0.2227	1.01	382	0	1.00

p[4]	0.2399	0.0391	0.1705	0.2123	0.2374	0.2647	0.3232	1.01	607	0	1.00
p[5]	0.1168	0.0225	0.0785	0.1008	0.1148	0.1306	0.1665	1.00	812	0	1.00
p[6]	0.2862	0.0508	0.1967	0.2502	0.2829	0.3185	0.3947	1.00	1189	0	1.00
p[7]	0.1812	0.0309	0.1275	0.1594	0.1789	0.2003	0.2483	1.01	390	0	1.00
p[8]	0.1286	0.0414	0.0643	0.0985	0.1230	0.1526	0.2246	1.00	7830	0	1.00
p[9]	0.1168	0.0230	0.0778	0.1004	0.1147	0.1309	0.1675	1.00	924	0	1.00
p[10]	0.2933	0.0529	0.2001	0.2558	0.2899	0.3270	0.4062	1.00	1334	0	1.00
p[11]	0.1359	0.0256	0.0927	0.1178	0.1336	0.1514	0.1928	1.01	400	0	1.00
p[12]	0.2870	0.0510	0.1971	0.2508	0.2837	0.3194	0.3959	1.00	1203	0	1.00
p[13]	0.1174	0.0219	0.0801	0.1018	0.1155	0.1308	0.1656	1.01	636	0	1.00
alpha.p	-1.7410	0.2157	-2.1544	-1.8870	-1.7442	-1.5986	-1.3073	1.01	362	0	1.00
beta1.p	-0.5292	0.1383	-0.8060	-0.6213	-0.5267	-0.4341	-0.2643	1.00	16270	0	1.00
beta2.p	0.2324	0.1475	-0.0578	0.1330	0.2327	0.3324	0.5196	1.00	1565	1	0.94
phi[1]	<b>0.4999</b>	<b>0.2885</b>	<b>0.0252</b>	<b>0.2501</b>	<b>0.5000</b>	<b>0.7500</b>	<b>0.9752</b>	<b>1.00</b>	<b>104596</b>	<b>0</b>	<b>1.00</b>
phi[2]	<b>0.7253</b>	<b>0.1510</b>	<b>0.4203</b>	<b>0.6182</b>	<b>0.7315</b>	<b>0.8432</b>	<b>0.9788</b>	<b>1.00</b>	<b>2973</b>	<b>0</b>	<b>1.00</b>
phi[3]	<b>0.7370</b>	<b>0.1530</b>	<b>0.4236</b>	<b>0.6279</b>	<b>0.7470</b>	<b>0.8593</b>	<b>0.9828</b>	<b>1.00</b>	<b>23557</b>	<b>0</b>	<b>1.00</b>
phi[4]	<b>0.5287</b>	<b>0.1286</b>	<b>0.3107</b>	<b>0.4364</b>	<b>0.5169</b>	<b>0.6086</b>	<b>0.8150</b>	<b>1.00</b>	<b>6371</b>	<b>0</b>	<b>1.00</b>
phi[5]	<b>0.8001</b>	<b>0.1411</b>	<b>0.4820</b>	<b>0.7076</b>	<b>0.8234</b>	<b>0.9160</b>	<b>0.9917</b>	<b>1.00</b>	<b>15964</b>	<b>0</b>	<b>1.00</b>
phi[6]	<b>0.5710</b>	<b>0.1417</b>	<b>0.3276</b>	<b>0.4684</b>	<b>0.5586</b>	<b>0.6617</b>	<b>0.8817</b>	<b>1.00</b>	<b>26344</b>	<b>0</b>	<b>1.00</b>
phi[7]	<b>0.9153</b>	<b>0.0742</b>	<b>0.7239</b>	<b>0.8784</b>	<b>0.9357</b>	<b>0.9723</b>	<b>0.9975</b>	<b>1.00</b>	<b>132076</b>	<b>0</b>	<b>1.00</b>
phi[8]	<b>0.5307</b>	<b>0.1566</b>	<b>0.2836</b>	<b>0.4150</b>	<b>0.5085</b>	<b>0.6258</b>	<b>0.8956</b>	<b>1.00</b>	<b>1551</b>	<b>0</b>	<b>1.00</b>
phi[9]	<b>0.7873</b>	<b>0.1386</b>	<b>0.4905</b>	<b>0.6912</b>	<b>0.8041</b>	<b>0.9011</b>	<b>0.9896</b>	<b>1.00</b>	<b>5956</b>	<b>0</b>	<b>1.00</b>
phi[10]	<b>0.6309</b>	<b>0.1548</b>	<b>0.3553</b>	<b>0.5163</b>	<b>0.6214</b>	<b>0.7392</b>	<b>0.9450</b>	<b>1.00</b>	<b>17840</b>	<b>0</b>	<b>1.00</b>
phi[11]	<b>0.7387</b>	<b>0.1414</b>	<b>0.4582</b>	<b>0.6366</b>	<b>0.7432</b>	<b>0.8491</b>	<b>0.9794</b>	<b>1.00</b>	<b>4546</b>	<b>0</b>	<b>1.00</b>
phi[12]	<b>0.8736</b>	<b>0.1032</b>	<b>0.6190</b>	<b>0.8158</b>	<b>0.8991</b>	<b>0.9555</b>	<b>0.9959</b>	<b>1.00</b>	<b>57684</b>	<b>0</b>	<b>1.00</b>
b[1]	0.2204	0.0433	0.1419	0.1900	0.2184	0.2485	0.3110	1.00	4211	0	1.00
b[2]	0.0474	0.0339	0.0021	0.0202	0.0414	0.0682	0.1263	1.00	73647	0	1.00
b[3]	0.0723	0.0383	0.0073	0.0437	0.0699	0.0977	0.1533	1.00	9305	0	1.00
b[4]	0.0151	0.0137	0.0004	0.0048	0.0112	0.0214	0.0509	1.00	30847	0	1.00
b[5]	0.1153	0.0428	0.0379	0.0855	0.1130	0.1425	0.2060	1.00	15438	0	1.00
b[6]	0.0200	0.0176	0.0006	0.0067	0.0152	0.0284	0.0651	1.00	16568	0	1.00
b[7]	0.2978	0.0524	0.1972	0.2621	0.2970	0.3325	0.4024	1.00	3495	0	1.00
b[8]	0.0434	0.0369	0.0014	0.0151	0.0339	0.0619	0.1369	1.00	5301	0	1.00
b[9]	0.0252	0.0215	0.0008	0.0086	0.0196	0.0360	0.0800	1.00	23593	0	1.00
b[10]	0.0273	0.0208	0.0011	0.0108	0.0230	0.0393	0.0774	1.00	10160	0	1.00
b[11]	0.0182	0.0168	0.0005	0.0057	0.0133	0.0257	0.0627	1.00	56425	0	1.00
b[12]	0.0978	0.0360	0.0302	0.0732	0.0965	0.1208	0.1729	1.00	2880	0	1.00
B[1]	150	36	91	125	147	172	231	1.01	426	0	1.00
B[2]	31	23	1	13	27	45	86	1.00	7038	0	1.00
B[3]	48	26	5	29	46	65	105	1.00	1956	0	1.00
B[4]	9	9	0	3	7	13	33	1.00	37844	1	1.00
B[5]	78	30	27	57	75	95	143	1.00	2529	0	1.00
B[6]	13	11	0	4	10	18	42	1.00	186583	1	1.00
B[7]	204	46	126	172	200	231	305	1.01	380	0	1.00
B[8]	29	25	0	10	22	41	91	1.00	30871	1	1.00
B[9]	16	14	0	5	13	23	53	1.00	10463	1	1.00
B[10]	18	13	0	7	15	25	50	1.00	7368	1	1.00
B[11]	12	11	0	3	8	16	41	1.00	11863	1	1.00
B[12]	66	25	21	49	64	81	120	1.00	11864	0	1.00
N[1]	<b>150</b>	<b>36</b>	<b>91</b>	<b>125</b>	<b>147</b>	<b>172</b>	<b>231</b>	<b>1.01</b>	<b>426</b>	<b>0</b>	<b>1.00</b>
N[2]	<b>140</b>	<b>29</b>	<b>91</b>	<b>119</b>	<b>137</b>	<b>157</b>	<b>205</b>	<b>1.00</b>	<b>709</b>	<b>0</b>	<b>1.00</b>
N[3]	<b>152</b>	<b>32</b>	<b>99</b>	<b>129</b>	<b>149</b>	<b>171</b>	<b>223</b>	<b>1.01</b>	<b>470</b>	<b>0</b>	<b>1.00</b>
N[4]	<b>88</b>	<b>20</b>	<b>56</b>	<b>74</b>	<b>86</b>	<b>100</b>	<b>135</b>	<b>1.00</b>	<b>1064</b>	<b>0</b>	<b>1.00</b>
N[5]	<b>149</b>	<b>33</b>	<b>96</b>	<b>125</b>	<b>145</b>	<b>168</b>	<b>223</b>	<b>1.00</b>	<b>831</b>	<b>0</b>	<b>1.00</b>
N[6]	<b>95</b>	<b>21</b>	<b>62</b>	<b>80</b>	<b>93</b>	<b>108</b>	<b>144</b>	<b>1.00</b>	<b>1281</b>	<b>0</b>	<b>1.00</b>
N[7]	<b>292</b>	<b>52</b>	<b>205</b>	<b>256</b>	<b>287</b>	<b>323</b>	<b>407</b>	<b>1.01</b>	<b>332</b>	<b>0</b>	<b>1.00</b>
N[8]	<b>181</b>	<b>50</b>	<b>105</b>	<b>145</b>	<b>173</b>	<b>209</b>	<b>298</b>	<b>1.00</b>	<b>3988</b>	<b>0</b>	<b>1.00</b>
N[9]	<b>156</b>	<b>36</b>	<b>99</b>	<b>130</b>	<b>152</b>	<b>177</b>	<b>240</b>	<b>1.00</b>	<b>2571</b>	<b>0</b>	<b>1.00</b>
N[10]	<b>114</b>	<b>22</b>	<b>78</b>	<b>98</b>	<b>111</b>	<b>127</b>	<b>163</b>	<b>1.00</b>	<b>1133</b>	<b>0</b>	<b>1.00</b>
N[11]	<b>95</b>	<b>23</b>	<b>57</b>	<b>79</b>	<b>93</b>	<b>109</b>	<b>147</b>	<b>1.00</b>	<b>971</b>	<b>0</b>	<b>1.00</b>
N[12]	<b>150</b>	<b>30</b>	<b>102</b>	<b>129</b>	<b>147</b>	<b>167</b>	<b>218</b>	<b>1.00</b>	<b>1446</b>	<b>0</b>	<b>1.00</b>
gamma[1]	0.0677	0.0169	0.0394	0.0556	0.0660	0.0779	0.1056	1.01	477	0	1.00
gamma[2]	0.0155	0.0113	0.0007	0.0065	0.0134	0.0221	0.0423	1.00	5816	0	1.00
gamma[3]	0.0240	0.0131	0.0025	0.0143	0.0228	0.0322	0.0528	1.00	1817	0	1.00
gamma[4]	0.0051	0.0047	0.0001	0.0016	0.0038	0.0072	0.0175	1.00	28841	0	1.00
gamma[5]	0.0394	0.0154	0.0131	0.0287	0.0380	0.0485	0.0737	1.00	2038	0	1.00
gamma[6]	0.0071	0.0062	0.0002	0.0024	0.0054	0.0100	0.0233	1.00	63317	0	1.00
gamma[7]	0.1077	0.0261	0.0639	0.0895	0.1052	0.1232	0.1657	1.01	335	0	1.00
gamma[8]	0.0172	0.0146	0.0006	0.0061	0.0136	0.0244	0.0543	1.00	47734	0	1.00

gamma[9]	0.0103	0.0090	0.0003	0.0035	0.0080	0.0147	0.0335	1.00	6324	0	1.00
gamma[10]	0.0112	0.0085	0.0005	0.0045	0.0094	0.0160	0.0318	1.00	4578	0	1.00
gamma[11]	0.0077	0.0073	0.0002	0.0023	0.0055	0.0107	0.0273	1.00	6628	0	1.00
gamma[12]	0.0413	0.0163	0.0130	0.0301	0.0398	0.0509	0.0778	1.00	4065	0	1.00
deviance	1612	97	1423	1547	1612	1677	1804	1.01	402	0	1.00

<b>C) <i>Turdus flavipes</i> - Closed Population Model</b>											
<b>COFFEE</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
N	<b>766</b>	<b>149</b>	<b>546</b>	<b>662</b>	<b>742</b>	<b>838</b>	<b>1131</b>	<b>1.01</b>	<b>385</b>	<b>NO</b>	<b>1.00</b>
p.est[1]	0.0022	0.0008	0.000968	0.001614	0.0021	0.0027	0.0042	1.00	511	NO	1.00
p.est[2]	0.0482	0.0093	0.0307	0.0419	0.0477	0.0541	0.0679	1.00	389	NO	1.00
p.est[3]	0.0067	0.0018	0.0037	0.0054	0.0065	0.0079	0.0106	1.01	343	NO	1.00
p.est[4]	0.0484	0.0098	0.0306	0.0415	0.0479	0.0544	0.0689	1.00	402	NO	1.00
p.est[5]	0.0100	0.0024	0.0057	0.0083	0.0098	0.0116	0.0151	1.01	306	NO	1.00
p.est[6]	0.0011	0.0005	0.0004	0.0007	0.0010	0.0013	0.0023	1.00	664	NO	1.00
p.est[7]	0.0506	0.0097	0.0323	0.0441	0.0502	0.0569	0.0712	1.00	402	NO	1.00
p.est[8]	0.0415	0.0092	0.0255	0.0351	0.0409	0.0472	0.0613	1.00	395	NO	1.00
p.est[9]	0.0308	0.0061	0.0196	0.0267	0.0306	0.0347	0.0437	1.01	311	NO	1.00
p.est[10]	0.0205	0.0042	0.0127	0.0175	0.0202	0.0232	0.0292	1.01	284	NO	1.00
p.est[11]	0.0390	0.0076	0.0248	0.0339	0.0386	0.0437	0.0551	1.01	344	NO	1.00
p.est[12]	0.0018	0.0007	0.0007	0.0013	0.0017	0.0022	0.0036	1.00	550	NO	1.00
mean.p	0.0258	0.0052	0.0163	0.0223	0.0255	0.0291	0.0367	1.01	295	NO	1.00
beta1	1.5231	0.1712	1.1990	1.4038	1.5160	1.6376	1.8802	1.00	2500	NO	1.00
beta2	-0.7327	0.1118	-0.9585	-0.8076	-0.7290	-0.6549	-0.5153	1.00	1614	NO	1.00
psi	0.1828	0.0364	0.1282	0.1578	0.1776	0.1999	0.2718	1.01	417	NO	1.00
deviance	1931	84	1785	1873	1926	1980	2116	1.01	347	NO	1.00
<b>FOREST</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
N	<b>1239</b>	<b>153</b>	<b>970</b>	<b>1135</b>	<b>1228</b>	<b>1333</b>	<b>1568</b>	<b>1.00</b>	<b>1331</b>	<b>NO</b>	<b>1.00</b>
p.est[1]	0.0206	0.0029	0.0154	0.0187	0.0204	0.0225	0.0268	1.00	3000	NO	1.00
p.est[2]	0.0296	0.0042	0.0220	0.0267	0.0293	0.0322	0.0386	1.00	3000	NO	1.00
p.est[3]	0.0186	0.0027	0.0137	0.0167	0.0184	0.0203	0.0243	1.00	3000	NO	1.00
p.est[4]	0.0381	0.0052	0.0289	0.0345	0.0378	0.0413	0.0495	1.00	3000	NO	1.00
p.est[5]	0.0149	0.0024	0.0106	0.0132	0.0147	0.0164	0.0200	1.00	3000	NO	1.00
p.est[6]	0.0258	0.0036	0.0192	0.0233	0.0255	0.0280	0.0334	1.00	3000	NO	1.00
p.est[7]	0.0331	0.0060	0.0228	0.0290	0.0326	0.0367	0.0465	1.00	1855	NO	1.00
p.est[8]	0.0379	0.0052	0.0287	0.0343	0.0376	0.0411	0.0493	1.00	3000	NO	1.00
p.est[9]	0.0144	0.0024	0.0102	0.0127	0.0142	0.0159	0.0196	1.00	3000	NO	1.00
p.est[10]	0.0345	0.0050	0.0256	0.0312	0.0342	0.0376	0.0454	1.00	3000	NO	1.00
p.est[11]	0.0148	0.0024	0.0106	0.0131	0.0147	0.0164	0.0200	1.00	3000	NO	1.00
p.est[12]	0.0383	0.0052	0.0290	0.0347	0.0379	0.0414	0.0499	1.00	3000	NO	1.00
mean.p	0.0312	0.0045	0.0232	0.0281	0.0309	0.0340	0.0408	1.00	3000	NO	1.00
beta1	0.4604	0.0683	0.3280	0.4146	0.4593	0.5058	0.5983	1.00	3000	NO	1.00
beta2	-0.2480	0.0775	-0.4008	-0.3002	-0.2470	-0.1964	-0.0963	1.00	1782	NO	1.00
psi	0.2857	0.0357	0.2227	0.2614	0.2834	0.3079	0.3601	1.00	1120	NO	1.00
deviance	3570	97	3380	3506	3569	3634	3762	1.00	1524	NO	1.00

<b>D) <i>Turdus albicollis</i> - Closed Population Model</b>											
<b>COFFEE</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
N	<b>2157</b>	<b>1062</b>	<b>192</b>	<b>1406</b>	<b>2191</b>	<b>3019</b>	<b>3927</b>	<b>1.48</b>	<b>8</b>	<b>NO</b>	<b>1.00</b>
p.est[1]	0.0004	0.0009	<0.0001	<0.0001	0.0002	0.0003	0.0020	1.31	30	NO	1.00
p.est[2]	0.0019	0.0037	0.0003	0.0006	0.0009	0.0016	0.0118	1.27	22	NO	1.00
p.est[3]	0.0005	0.0011	<0.0001	0.0001	0.0002	0.0004	0.0031	1.31	24	NO	1.00
p.est[4]	0.0038	0.0075	0.0007	0.0013	0.0018	0.0030	0.0219	1.26	23	NO	1.00
p.est[5]	0.0006	0.0013	0.0001	0.0002	0.0003	0.0005	0.0038	1.31	23	NO	1.00
p.est[6]	0.0003	0.0009	0.0000	0.0000	0.0001	0.0003	0.0017	1.29	36	NO	1.00
p.est[7]	0.0021	0.0040	0.0004	0.0007	0.0010	0.0017	0.0124	1.27	22	NO	1.00
p.est[8]	0.0045	0.0091	0.0007	0.0014	0.0021	0.0037	0.0252	1.25	24	NO	1.00
p.est[9]	0.0012	0.0024	0.0002	0.0004	0.0006	0.0010	0.0074	1.27	22	NO	1.00
p.est[10]	0.0009	0.0018	0.0001	0.0003	0.0004	0.0007	0.0057	1.29	22	NO	1.00
p.est[11]	0.0015	0.0029	0.0002	0.0005	0.0007	0.0012	0.0089	1.27	22	NO	1.00
p.est[12]	0.0003	0.0009	<0.0001	0.0001	0.0001	0.0003	0.0020	1.30	32	NO	1.00
mean.p	0.0011	0.0021	0.0002	0.0003	0.0005	0.0008	0.0066	1.28	22	NO	1.00
beta1	1.1009	0.4594	0.3642	0.7729	1.0527	1.3827	2.1116	1.00	3000	NO	1.00
beta2	-0.1452	0.3292	-0.8286	-0.3509	-0.1408	0.0894	0.4664	1.00	1420	YES	0.66
psi	0.5366	0.2640	0.0465	0.3484	0.5458	0.7515	0.9773	1.48	8	NO	1.00
deviance	294	30	208	285	302	314	326	1.47	9	NO	1.00

<b>FOREST</b>												
<b>par</b>	<b>mean</b>	<b>sd</b>	<b>2.50%</b>	<b>25%</b>	<b>50%</b>	<b>75%</b>	<b>97.50%</b>	<b>Rhat</b>	<b>n.eff</b>	<b>over0</b>	<b>f</b>	
N	<b>482</b>	<b>41</b>	<b>411</b>	<b>452</b>	<b>478</b>	<b>508</b>	<b>575</b>	<b>1.00</b>	<b>1441</b>	<b>NO</b>	<b>1.00</b>	
p.est[1]	0.0390	0.0047	0.0304	0.0358	0.0388	0.0419	0.0490	1.00	827	NO	1.00	
p.est[2]	0.0595	0.0070	0.0467	0.0546	0.0591	0.0641	0.0737	1.00	973	NO	1.00	
p.est[3]	0.0346	0.0044	0.0265	0.0316	0.0343	0.0374	0.0441	1.00	892	NO	1.00	
p.est[4]	0.0891	0.0094	0.0709	0.0828	0.0888	0.0953	0.1075	1.00	1434	NO	1.00	
p.est[5]	0.0270	0.0042	0.0196	0.0241	0.0267	0.0296	0.0360	1.00	1200	NO	1.00	
p.est[6]	0.0504	0.0058	0.0397	0.0463	0.0502	0.0541	0.0623	1.00	854	NO	1.00	
p.est[7]	0.0858	0.0133	0.0617	0.0764	0.0849	0.0942	0.1151	1.00	3000	NO	1.00	
p.est[8]	0.0894	0.0095	0.0712	0.0831	0.0891	0.0956	0.1079	1.00	1488	NO	1.00	
p.est[9]	0.0260	0.0042	0.0187	0.0232	0.0258	0.0286	0.0351	1.00	1264	NO	1.00	
p.est[10]	0.0727	0.0086	0.0569	0.0666	0.0724	0.0782	0.0908	1.00	1169	NO	1.00	
p.est[11]	0.0269	0.0042	0.0195	0.0240	0.0266	0.0295	0.0359	1.00	1207	NO	1.00	
p.est[12]	0.0880	0.0094	0.0701	0.0816	0.0876	0.0942	0.1068	1.00	1353	NO	1.00	
mean.p	0.0636	0.0075	0.0499	0.0583	0.0632	0.0685	0.0791	1.00	1035	NO	1.00	
beta1	0.6012	0.0768	0.4569	0.5479	0.6013	0.6528	0.7486	1.00	3000	NO	1.00	
beta2	-0.2414	0.0829	-0.4053	-0.2958	-0.2405	-0.1864	-0.0787	1.00	3000	NO	1.00	
psi	0.1139	0.0110	0.0945	0.1060	0.1133	0.1208	0.1368	1.00	697	NO	1.00	
deviance	2461	58	2354	2419	2457	2499	2584	1.00	1523	NO	1.00	

**E) Ramphocelus dimidiatus - Jolly-Seber Model****COFFEE**

<b>par</b>	<b>mean</b>	<b>sd</b>	<b>2.50%</b>	<b>25%</b>	<b>50%</b>	<b>75%</b>	<b>97.50%</b>	<b>Rhat</b>	<b>n.eff</b>	<b>over0</b>	<b>f</b>
psi	0.1224	0.0326	0.0704	0.0964	0.1190	0.1455	0.1926	1.04	109	NO	1.00
<b>Ncoffee</b>	<b>613</b>	<b>165</b>	<b>352</b>	<b>481</b>	<b>595</b>	<b>731</b>	<b>972</b>	<b>1.04</b>	<b>106</b>	<b>NO</b>	<b>1.00</b>
p[1]	0.0876	0.0284	0.0454	0.0663	0.0834	0.1046	0.1535	1.04	80	NO	1.00
p[2]	0.2086	0.0725	0.0959	0.1547	0.2001	0.2523	0.3729	1.02	150	NO	1.00
p[3]	0.0618	0.0205	0.0320	0.0464	0.0585	0.0740	0.1099	1.04	73	NO	1.00
p[4]	0.1462	0.0454	0.0750	0.1120	0.1414	0.1738	0.2487	1.03	95	NO	1.00
p[5]	0.0460	0.0186	0.0202	0.0323	0.0426	0.0558	0.0910	1.03	99	NO	1.00
p[6]	0.1273	0.0392	0.0664	0.0976	0.1229	0.1514	0.2160	1.03	86	NO	1.00
p[7]	0.2582	0.0990	0.1057	0.1845	0.2454	0.3184	0.4847	1.01	213	NO	1.00
p[8]	0.0592	0.0197	0.0307	0.0444	0.0560	0.0708	0.1049	1.04	72	NO	1.00
p[9]	0.0453	0.0215	0.0171	0.0299	0.0409	0.0558	0.0984	1.02	134	NO	1.00
p[10]	0.0806	0.0264	0.0416	0.0608	0.0765	0.0964	0.1424	1.04	79	NO	1.00
p[11]	0.0968	0.0308	0.0504	0.0735	0.0925	0.1153	0.1682	1.04	80	NO	1.00
p[12]	0.0712	0.0236	0.0367	0.0535	0.0674	0.0853	0.1268	1.04	77	NO	1.00
p[13]	0.2216	0.0792	0.0989	0.1630	0.2120	0.2691	0.4016	1.01	166	NO	1.00
alpha.p	-2.3936	0.3495	-3.0469	-2.6452	-2.3975	-2.1476	-1.7071	1.03	102	NO	1.00
beta1.p	-0.7366	0.2001	-1.1435	-0.8667	-0.7300	-0.6001	-0.3611	1.00	1156	NO	1.00
beta2.p	0.1709	0.1849	-0.1863	0.0454	0.1694	0.2936	0.5429	1.00	1429	YES	0.83
phi[1]	<b>0.4986</b>	<b>0.2890</b>	<b>0.0256</b>	<b>0.2479</b>	<b>0.4979</b>	<b>0.7506</b>	<b>0.9743</b>	<b>1.00</b>	<b>6661</b>	<b>NO</b>	<b>1.00</b>
phi[2]	<b>0.6622</b>	<b>0.2360</b>	<b>0.1510</b>	<b>0.4965</b>	<b>0.7001</b>	<b>0.8623</b>	<b>0.9868</b>	<b>1.00</b>	<b>588</b>	<b>NO</b>	<b>1.00</b>
phi[3]	<b>0.8577</b>	<b>0.1151</b>	<b>0.5720</b>	<b>0.7937</b>	<b>0.8864</b>	<b>0.9491</b>	<b>0.9951</b>	<b>1.00</b>	<b>2213</b>	<b>NO</b>	<b>1.00</b>
phi[4]	<b>0.8617</b>	<b>0.1119</b>	<b>0.5876</b>	<b>0.7984</b>	<b>0.8885</b>	<b>0.9504</b>	<b>0.9953</b>	<b>1.00</b>	<b>11662</b>	<b>NO</b>	<b>1.00</b>
phi[5]	<b>0.8053</b>	<b>0.1477</b>	<b>0.4544</b>	<b>0.7134</b>	<b>0.8364</b>	<b>0.9254</b>	<b>0.9931</b>	<b>1.00</b>	<b>4809</b>	<b>NO</b>	<b>1.00</b>
phi[6]	<b>0.3599</b>	<b>0.1543</b>	<b>0.1365</b>	<b>0.2482</b>	<b>0.3324</b>	<b>0.4420</b>	<b>0.7473</b>	<b>1.00</b>	<b>12703</b>	<b>NO</b>	<b>1.00</b>
phi[7]	<b>0.8416</b>	<b>0.1320</b>	<b>0.5112</b>	<b>0.7700</b>	<b>0.8755</b>	<b>0.9459</b>	<b>0.9950</b>	<b>1.00</b>	<b>7001</b>	<b>NO</b>	<b>1.00</b>
phi[8]	<b>0.7511</b>	<b>0.1812</b>	<b>0.3417</b>	<b>0.6298</b>	<b>0.7846</b>	<b>0.9027</b>	<b>0.9909</b>	<b>1.00</b>	<b>1165</b>	<b>NO</b>	<b>1.00</b>
phi[9]	<b>0.5916</b>	<b>0.1937</b>	<b>0.2518</b>	<b>0.4431</b>	<b>0.5788</b>	<b>0.7384</b>	<b>0.9613</b>	<b>1.01</b>	<b>340</b>	<b>NO</b>	<b>1.00</b>
phi[10]	<b>0.7745</b>	<b>0.1684</b>	<b>0.3820</b>	<b>0.6691</b>	<b>0.8098</b>	<b>0.9121</b>	<b>0.9913</b>	<b>1.00</b>	<b>8600</b>	<b>NO</b>	<b>1.00</b>
phi[11]	<b>0.4466</b>	<b>0.1980</b>	<b>0.1191</b>	<b>0.2977</b>	<b>0.4260</b>	<b>0.5768</b>	<b>0.8807</b>	<b>1.01</b>	<b>348</b>	<b>NO</b>	<b>1.00</b>
phi[12]	<b>0.1760</b>	<b>0.1616</b>	<b>0.0052</b>	<b>0.0569</b>	<b>0.1297</b>	<b>0.2457</b>	<b>0.6149</b>	<b>1.00</b>	<b>4315</b>	<b>NO</b>	<b>1.00</b>
b[1]	0.0412	0.0298	0.0048	0.0196	0.0344	0.0553	0.1168	1.00	797	NO	1.00
b[2]	0.1378	0.0519	0.0459	0.1013	0.1339	0.1708	0.2479	1.01	207	NO	1.00
b[3]	0.0422	0.0339	0.0015	0.0158	0.0345	0.0607	0.1263	1.00	1536	NO	1.00
b[4]	0.0487	0.0362	0.0020	0.0200	0.0417	0.0699	0.1347	1.00	892	NO	1.00
b[5]	0.0441	0.0409	0.0012	0.0138	0.0325	0.0621	0.1531	1.01	486	NO	1.00
b[6]	0.0129	0.0131	0.0003	0.0036	0.0089	0.0179	0.0478	1.00	1465	NO	1.00
b[7]	0.0574	0.0432	0.0035	0.0265	0.0485	0.0776	0.1663	1.00	2773	NO	1.00
b[8]	0.3780	0.0966	0.1885	0.3129	0.3772	0.4439	0.5679	1.01	332	NO	1.00
b[9]	0.0462	0.0444	0.0012	0.0139	0.0328	0.0649	0.1637	1.01	963	NO	1.00
b[10]	0.0843	0.0646	0.0031	0.0321	0.0706	0.1228	0.2376	1.00	1224	NO	1.00
b[11]	0.0698	0.0499	0.0030	0.0299	0.0610	0.1008	0.1848	1.00	683	NO	1.00
b[12]	0.0375	0.0316	0.0014	0.0136	0.0297	0.0529	0.1182	1.00	4571	NO	1.00
B[1]	25	21	3	11	19	33	79	1.01	509	NO	1.00
B[2]	84	36	28	59	78	102	171	1.01	188	NO	1.00
B[3]	25	22	0	9	20	36	81	1.00	1821	YES	1.00

B[4]	29	23	1	11	24	41	88	1.00	30000	NO	1.00
B[5]	27	27	0	8	18	36	102	1.01	481	YES	1.00
B[6]	7	8	0	2	4	10	28	1.00	7584	YES	1.00
B[7]	33	26	2	16	28	44	100	1.01	966	NO	1.00
B[8]	237	92	96	167	224	297	443	1.02	105	NO	1.00
B[9]	28	29	0	7	19	38	105	1.01	341	YES	1.00
B[10]	53	46	1	18	40	74	173	1.01	651	NO	1.00
B[11]	43	35	1	17	35	61	133	1.02	267	NO	1.00
B[12]	23	23	0	7	17	32	83	1.00	750	YES	1.00
N[1]	<b>25</b>	<b>21</b>	<b>3</b>	<b>11</b>	<b>19</b>	<b>33</b>	<b>79</b>	<b>1.01</b>	<b>509</b>	NO	<b>1.00</b>
N[2]	<b>100</b>	<b>37</b>	<b>49</b>	<b>74</b>	<b>93</b>	<b>119</b>	<b>191</b>	<b>1.02</b>	<b>163</b>	NO	<b>1.00</b>
N[3]	<b>111</b>	<b>38</b>	<b>57</b>	<b>84</b>	<b>104</b>	<b>132</b>	<b>204</b>	<b>1.01</b>	<b>174</b>	NO	<b>1.00</b>
N[4]	<b>125</b>	<b>40</b>	<b>69</b>	<b>96</b>	<b>118</b>	<b>147</b>	<b>220</b>	<b>1.01</b>	<b>275</b>	NO	<b>1.00</b>
N[5]	<b>127</b>	<b>47</b>	<b>61</b>	<b>93</b>	<b>118</b>	<b>151</b>	<b>241</b>	<b>1.01</b>	<b>218</b>	NO	<b>1.00</b>
N[6]	<b>51</b>	<b>23</b>	<b>19</b>	<b>34</b>	<b>46</b>	<b>62</b>	<b>109</b>	<b>1.01</b>	<b>325</b>	NO	<b>1.00</b>
N[7]	<b>76</b>	<b>34</b>	<b>34</b>	<b>53</b>	<b>68</b>	<b>91</b>	<b>162</b>	<b>1.01</b>	<b>270</b>	NO	<b>1.00</b>
N[8]	<b>295</b>	<b>96</b>	<b>147</b>	<b>221</b>	<b>282</b>	<b>359</b>	<b>509</b>	<b>1.03</b>	<b>91</b>	NO	<b>1.00</b>
N[9]	<b>199</b>	<b>81</b>	<b>77</b>	<b>139</b>	<b>185</b>	<b>245</b>	<b>390</b>	<b>1.02</b>	<b>146</b>	NO	<b>1.00</b>
N[10]	<b>203</b>	<b>71</b>	<b>96</b>	<b>152</b>	<b>192</b>	<b>244</b>	<b>366</b>	<b>1.02</b>	<b>178</b>	NO	<b>1.00</b>
N[11]	<b>133</b>	<b>55</b>	<b>54</b>	<b>92</b>	<b>123</b>	<b>163</b>	<b>265</b>	<b>1.01</b>	<b>234</b>	NO	<b>1.00</b>
N[12]	<b>44</b>	<b>30</b>	<b>7</b>	<b>23</b>	<b>37</b>	<b>59</b>	<b>123</b>	<b>1.00</b>	<b>549</b>	NO	<b>1.00</b>
gamma[1]	0.0051	0.0041	0.0005	0.0022	0.0040	0.0067	0.0160	1.01	527	NO	1.00
gamma[2]	0.0166	0.0073	0.0055	0.0115	0.0154	0.0204	0.0343	1.01	192	NO	1.00
gamma[3]	0.0053	0.0046	0.0002	0.0019	0.0041	0.0074	0.0169	1.00	1897	NO	1.00
gamma[4]	0.0060	0.0048	0.0003	0.0024	0.0049	0.0084	0.0182	1.00	30000	NO	1.00
gamma[5]	0.0056	0.0057	0.0001	0.0017	0.0039	0.0076	0.0211	1.01	481	NO	1.00
gamma[6]	0.0016	0.0017	0.0000	0.0004	0.0011	0.0022	0.0062	1.00	7406	NO	1.00
gamma[7]	0.0070	0.0055	0.0005	0.0033	0.0058	0.0092	0.0213	1.01	926	NO	1.00
gamma[8]	0.0489	0.0193	0.0194	0.0341	0.0461	0.0613	0.0925	1.02	112	NO	1.00
gamma[9]	0.0062	0.0065	0.0001	0.0018	0.0042	0.0086	0.0238	1.01	347	NO	1.00
gamma[10]	0.0117	0.0102	0.0004	0.0040	0.0089	0.0164	0.0386	1.01	604	NO	1.00
gamma[11]	0.0098	0.0081	0.0004	0.0038	0.0079	0.0137	0.0303	1.02	251	NO	1.00
gamma[12]	0.0054	0.0053	0.0002	0.0017	0.0039	0.0074	0.0195	1.00	731	NO	1.00
deviance	808	74	668	755	805	863	948	1.03	82	NO	1.00
<b>FOREST</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
psi	0.0668	0.0130	0.0441	0.0580	0.0658	0.0746	0.0958	1.08	36	NO	1.00
<b>Nforest</b>	<b>329</b>	<b>64</b>	<b>216</b>	<b>286</b>	<b>323</b>	<b>366</b>	<b>473</b>	<b>1.09</b>	<b>33</b>	NO	<b>1.00</b>
p[1]	0.1387	0.0385	0.0782	0.1115	0.1335	0.1602	0.2293	1.03	77	NO	1.00
p[2]	0.1878	0.0437	0.1160	0.1571	0.1827	0.2129	0.2878	1.03	73	NO	1.00
p[3]	0.1463	0.0394	0.0838	0.1185	0.1412	0.1682	0.2388	1.03	75	NO	1.00
p[4]	0.1983	0.0466	0.1220	0.1655	0.1931	0.2254	0.3050	1.03	81	NO	1.00
p[5]	0.0762	0.0235	0.0401	0.0595	0.0729	0.0892	0.1307	1.02	100	NO	1.00
p[6]	0.2198	0.0577	0.1262	0.1786	0.2132	0.2538	0.3508	1.02	124	NO	1.00
p[7]	0.1638	0.0409	0.0977	0.1352	0.1588	0.1866	0.2590	1.03	71	NO	1.00
p[8]	0.0529	0.0270	0.0180	0.0342	0.0471	0.0650	0.1220	1.01	427	NO	1.00
p[9]	0.0739	0.0232	0.0383	0.0574	0.0706	0.0867	0.1277	1.02	106	NO	1.00
p[10]	0.2228	0.0599	0.1259	0.1801	0.2159	0.2582	0.3591	1.02	133	NO	1.00
p[11]	0.1210	0.0352	0.0667	0.0961	0.1161	0.1407	0.2040	1.03	81	NO	1.00
p[12]	0.2201	0.0579	0.1263	0.1788	0.2134	0.2543	0.3517	1.02	125	NO	1.00
p[13]	0.0820	0.0246	0.0441	0.0644	0.0786	0.0957	0.1391	1.03	91	NO	1.00
alpha.p	-1.8621	0.3166	-2.4668	-2.0751	-1.8702	-1.6567	-1.2123	1.04	69	NO	1.00
beta1.p	-0.5838	0.1782	-0.9406	-0.7033	-0.5811	-0.4637	-0.2373	1.00	11082	NO	1.00
beta2.p	-0.0451	0.2071	-0.4484	-0.1842	-0.0464	0.0927	0.3654	1.00	2157	YES	0.59
phi[1]	<b>0.5009</b>	<b>0.2876</b>	<b>0.0256</b>	<b>0.2528</b>	<b>0.5033</b>	<b>0.7483</b>	<b>0.9756</b>	<b>1.00</b>	<b>30000</b>	NO	<b>1.00</b>
phi[2]	<b>0.6633</b>	<b>0.1959</b>	<b>0.2785</b>	<b>0.5158</b>	<b>0.6706</b>	<b>0.8230</b>	<b>0.9804</b>	<b>1.00</b>	<b>1373</b>	NO	<b>1.00</b>
phi[3]	<b>0.6557</b>	<b>0.1957</b>	<b>0.2699</b>	<b>0.5113</b>	<b>0.6639</b>	<b>0.8140</b>	<b>0.9768</b>	<b>1.00</b>	<b>6201</b>	NO	<b>1.00</b>
phi[4]	<b>0.8147</b>	<b>0.1475</b>	<b>0.4579</b>	<b>0.7275</b>	<b>0.8504</b>	<b>0.9331</b>	<b>0.9940</b>	<b>1.00</b>	<b>30000</b>	NO	<b>1.00</b>
phi[5]	<b>0.6795</b>	<b>0.1952</b>	<b>0.2843</b>	<b>0.5368</b>	<b>0.6953</b>	<b>0.8405</b>	<b>0.9825</b>	<b>1.00</b>	<b>971</b>	NO	<b>1.00</b>
phi[6]	<b>0.7866</b>	<b>0.1522</b>	<b>0.4430</b>	<b>0.6873</b>	<b>0.8122</b>	<b>0.9122</b>	<b>0.9919</b>	<b>1.00</b>	<b>14910</b>	NO	<b>1.00</b>
phi[7]	<b>0.9058</b>	<b>0.0843</b>	<b>0.6903</b>	<b>0.8648</b>	<b>0.9295</b>	<b>0.9700</b>	<b>0.9974</b>	<b>1.00</b>	<b>5070</b>	NO	<b>1.00</b>
phi[8]	<b>0.7348</b>	<b>0.1832</b>	<b>0.3387</b>	<b>0.6064</b>	<b>0.7630</b>	<b>0.8886</b>	<b>0.9892</b>	<b>1.00</b>	<b>1809</b>	NO	<b>1.00</b>
phi[9]	<b>0.4679</b>	<b>0.1768</b>	<b>0.1985</b>	<b>0.3355</b>	<b>0.4398</b>	<b>0.5739</b>	<b>0.8859</b>	<b>1.00</b>	<b>5289</b>	NO	<b>1.00</b>
phi[10]	<b>0.7659</b>	<b>0.1645</b>	<b>0.3994</b>	<b>0.6567</b>	<b>0.7935</b>	<b>0.9019</b>	<b>0.9909</b>	<b>1.00</b>	<b>1203</b>	NO	<b>1.00</b>
phi[11]	<b>0.7880</b>	<b>0.1497</b>	<b>0.4528</b>	<b>0.6901</b>	<b>0.8132</b>	<b>0.9114</b>	<b>0.9917</b>	<b>1.00</b>	<b>9371</b>	NO	<b>1.00</b>
phi[12]	<b>0.8558</b>	<b>0.1184</b>	<b>0.5654</b>	<b>0.7895</b>	<b>0.8854</b>	<b>0.9491</b>	<b>0.9952</b>	<b>1.00</b>	<b>9026</b>	NO	<b>1.00</b>
b[1]	0.1269	0.0443	0.0561	0.0949	0.1214	0.1527	0.2288	1.00	28302	NO	1.00
b[2]	0.0263	0.0227	0.0008	0.0090	0.0204	0.0376	0.0841	1.00	3418	NO	1.00
b[3]	0.0189	0.0184	0.0005	0.0056	0.0134	0.0265	0.0685	1.00	13970	NO	1.00

b[4]	0.0615	0.0387	0.0042	0.0323	0.0565	0.0850	0.1500	1.00	10422	NO	1.00
b[5]	0.0849	0.0515	0.0069	0.0467	0.0781	0.1157	0.2026	1.00	1361	NO	1.00
b[6]	0.0357	0.0320	0.0011	0.0119	0.0269	0.0505	0.1199	1.00	2993	NO	1.00
b[7]	0.3129	0.0803	0.1564	0.2579	0.3128	0.3674	0.4699	1.01	332	NO	1.00
b[8]	0.0489	0.0485	0.0012	0.0140	0.0338	0.0679	0.1791	1.00	30000	NO	1.00
b[9]	0.0322	0.0290	0.0009	0.0104	0.0241	0.0459	0.1075	1.00	2061	NO	1.00
b[10]	0.0351	0.0293	0.0011	0.0125	0.0279	0.0502	0.1095	1.00	1237	NO	1.00
b[11]	0.0502	0.0428	0.0016	0.0174	0.0390	0.0718	0.1599	1.00	4540	NO	1.00
b[12]	0.1664	0.0747	0.0299	0.1138	0.1630	0.2158	0.3219	1.00	4104	NO	1.00
B[1]	42	16	19	30	39	51	81	1.02	148	NO	1.00
B[2]	8	8	0	2	6	11	27	1.01	992	YES	1.00
B[3]	5	6	0	1	4	8	22	1.01	551	YES	1.00
B[4]	20	13	1	10	18	27	51	1.01	332	NO	1.00
B[5]	28	18	2	15	25	37	69	1.00	983	NO	1.00
B[6]	11	11	0	3	8	16	39	1.00	2568	YES	1.00
B[7]	105	32	50	83	103	125	176	1.07	42	NO	1.00
B[8]	16	17	0	4	10	22	62	1.00	908	YES	1.00
B[9]	10	10	0	3	7	14	35	1.01	361	YES	1.00
B[10]	11	10	0	4	8	16	35	1.00	4299	YES	1.00
B[11]	16	15	0	5	12	23	53	1.00	1463	YES	1.00
B[12]	56	30	9	36	52	73	127	1.01	412	NO	1.00
N[1]	<b>42</b>	<b>16</b>	<b>19</b>	<b>30</b>	<b>39</b>	<b>51</b>	<b>81</b>	<b>1.02</b>	<b>148</b>	<b>NO</b>	<b>1.00</b>
N[2]	<b>35</b>	<b>12</b>	<b>17</b>	<b>26</b>	<b>33</b>	<b>42</b>	<b>65</b>	<b>1.02</b>	<b>132</b>	<b>NO</b>	<b>1.00</b>
N[3]	<b>28</b>	<b>11</b>	<b>11</b>	<b>20</b>	<b>26</b>	<b>34</b>	<b>55</b>	<b>1.02</b>	<b>125</b>	<b>NO</b>	<b>1.00</b>
N[4]	<b>43</b>	<b>15</b>	<b>22</b>	<b>33</b>	<b>41</b>	<b>51</b>	<b>78</b>	<b>1.03</b>	<b>96</b>	<b>NO</b>	<b>1.00</b>
N[5]	<b>57</b>	<b>20</b>	<b>28</b>	<b>44</b>	<b>54</b>	<b>68</b>	<b>106</b>	<b>1.02</b>	<b>160</b>	<b>NO</b>	<b>1.00</b>
N[6]	<b>56</b>	<b>18</b>	<b>29</b>	<b>43</b>	<b>53</b>	<b>66</b>	<b>99</b>	<b>1.01</b>	<b>195</b>	<b>NO</b>	<b>1.00</b>
N[7]	<b>157</b>	<b>36</b>	<b>96</b>	<b>132</b>	<b>154</b>	<b>178</b>	<b>239</b>	<b>1.09</b>	<b>34</b>	<b>NO</b>	<b>1.00</b>
N[8]	<b>130</b>	<b>42</b>	<b>59</b>	<b>100</b>	<b>127</b>	<b>157</b>	<b>223</b>	<b>1.03</b>	<b>95</b>	<b>NO</b>	<b>1.00</b>
N[9]	<b>68</b>	<b>25</b>	<b>32</b>	<b>50</b>	<b>64</b>	<b>81</b>	<b>129</b>	<b>1.03</b>	<b>99</b>	<b>NO</b>	<b>1.00</b>
N[10]	<b>62</b>	<b>18</b>	<b>34</b>	<b>49</b>	<b>59</b>	<b>71</b>	<b>103</b>	<b>1.02</b>	<b>128</b>	<b>NO</b>	<b>1.00</b>
N[11]	<b>65</b>	<b>21</b>	<b>32</b>	<b>49</b>	<b>62</b>	<b>77</b>	<b>115</b>	<b>1.02</b>	<b>150</b>	<b>NO</b>	<b>1.00</b>
N[12]	<b>113</b>	<b>33</b>	<b>63</b>	<b>89</b>	<b>107</b>	<b>130</b>	<b>193</b>	<b>1.02</b>	<b>125</b>	<b>NO</b>	<b>1.00</b>
gamma[1]	0.0085	0.0034	0.0034	0.0060	0.0079	0.0104	0.0167	1.02	177	NO	1.00
gamma[2]	0.0018	0.0016	0.0001	0.0006	0.0013	0.0025	0.0059	1.00	1208	NO	1.00
gamma[3]	0.0013	0.0013	0.0000	0.0004	0.0009	0.0018	0.0047	1.01	739	NO	1.00
gamma[4]	0.0041	0.0028	0.0003	0.0021	0.0037	0.0057	0.0107	1.01	318	NO	1.00
gamma[5]	0.0057	0.0037	0.0005	0.0030	0.0051	0.0077	0.0145	1.00	1147	NO	1.00
gamma[6]	0.0024	0.0023	0.0001	0.0008	0.0018	0.0034	0.0083	1.00	4128	NO	1.00
gamma[7]	0.0214	0.0068	0.0097	0.0166	0.0208	0.0255	0.0364	1.07	44	NO	1.00
gamma[8]	0.0034	0.0036	0.0001	0.0010	0.0023	0.0047	0.0133	1.00	838	NO	1.00
gamma[9]	0.0022	0.0021	0.0001	0.0007	0.0017	0.0032	0.0078	1.01	362	NO	1.00
gamma[10]	0.0025	0.0021	0.0001	0.0009	0.0019	0.0035	0.0079	1.00	6516	NO	1.00
gamma[11]	0.0035	0.0031	0.0001	0.0012	0.0027	0.0050	0.0115	1.00	1314	NO	1.00
gamma[12]	0.0119	0.0064	0.0020	0.0075	0.0110	0.0154	0.0272	1.01	359	NO	1.00
deviance	658	54	547	623	659	694	764	1.05	53	NO	1.00

F) Tangara gyrolo - Jolly-Seber Model											
COFFEE											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
psi	0.2654	0.0498	0.1826	0.2288	0.2620	0.2966	0.3712	1.06	50	NO	1.00
<b>Ncoffee</b>	<b>558</b>	<b>105</b>	<b>385</b>	<b>479</b>	<b>551</b>	<b>624</b>	<b>778</b>	<b>1.06</b>	<b>49</b>	<b>NO</b>	<b>1.00</b>
p[1]	0.1489	0.0324	0.0949	0.1259	0.1455	0.1685	0.2217	1.03	81	NO	1.00
p[2]	0.2470	0.0629	0.1409	0.2021	0.2411	0.2858	0.3856	1.01	355	NO	1.00
p[3]	0.1353	0.0300	0.0862	0.1137	0.1318	0.1534	0.2025	1.03	79	NO	1.00
p[4]	0.1951	0.0414	0.1233	0.1660	0.1919	0.2209	0.2852	1.02	128	NO	1.00
p[5]	0.1535	0.0468	0.0812	0.1200	0.1470	0.1799	0.2629	1.01	239	NO	1.00
p[6]	0.1797	0.0373	0.1150	0.1533	0.1767	0.2027	0.2612	1.02	103	NO	1.00
p[7]	0.2885	0.0852	0.1472	0.2265	0.2801	0.3415	0.4765	1.00	853	NO	1.00
p[8]	0.1349	0.0300	0.0859	0.1134	0.1315	0.1530	0.2021	1.03	81	NO	1.00
p[9]	0.1666	0.0602	0.0769	0.1236	0.1571	0.1996	0.3105	1.01	426	NO	1.00
p[10]	0.1443	0.0318	0.0917	0.1216	0.1408	0.1634	0.2156	1.03	80	NO	1.00
p[11]	0.1556	0.0333	0.0995	0.1319	0.1522	0.1758	0.2300	1.03	84	NO	1.00
p[12]	0.1388	0.0308	0.0882	0.1168	0.1353	0.1574	0.2081	1.03	78	NO	1.00
p[13]	0.2578	0.0685	0.1429	0.2086	0.2513	0.3002	0.4088	1.00	446	NO	1.00
alpha.p	-1.7656	0.2544	-2.2552	-1.9380	-1.7706	-1.5961	-1.2555	1.03	75	NO	1.00
beta1.p	-0.3249	0.1707	-0.6698	-0.4380	-0.3215	-0.2080	0.0009	1.00	2352	YES	0.97
beta2.p	0.2266	0.1687	-0.1051	0.1146	0.2260	0.3379	0.5618	1.00	1067	YES	0.91
phi[1]	<b>0.5008</b>	<b>0.2885</b>	<b>0.0261</b>	<b>0.2514</b>	<b>0.5005</b>	<b>0.7510</b>	<b>0.9755</b>	<b>1.00</b>	<b>120000</b>	<b>NO</b>	<b>1.00</b>

	<b>phi[2]</b>	<b>0.6569</b>	<b>0.2384</b>	<b>0.1514</b>	<b>0.4849</b>	<b>0.6950</b>	<b>0.8596</b>	<b>0.9866</b>	<b>1.00</b>	<b>4753</b>	<b>NO</b>	<b>1.00</b>
<b>phi[3]</b>	<b>0.8816</b>	<b>0.0993</b>	<b>0.6319</b>	<b>0.8284</b>	<b>0.9071</b>	<b>0.9596</b>	<b>0.9963</b>	<b>1.00</b>	<b>67412</b>	<b>NO</b>	<b>1.00</b>	
<b>phi[4]</b>	<b>0.8174</b>	<b>0.1305</b>	<b>0.5201</b>	<b>0.7339</b>	<b>0.8400</b>	<b>0.9239</b>	<b>0.9925</b>	<b>1.00</b>	<b>4842</b>	<b>NO</b>	<b>1.00</b>	
<b>phi[5]</b>	<b>0.5704</b>	<b>0.1608</b>	<b>0.3021</b>	<b>0.4511</b>	<b>0.5527</b>	<b>0.6741</b>	<b>0.9229</b>	<b>1.00</b>	<b>120000</b>	<b>NO</b>	<b>1.00</b>	
<b>phi[6]</b>	<b>0.7519</b>	<b>0.1591</b>	<b>0.4180</b>	<b>0.6380</b>	<b>0.7696</b>	<b>0.8836</b>	<b>0.9877</b>	<b>1.00</b>	<b>120000</b>	<b>NO</b>	<b>1.00</b>	
<b>phi[7]</b>	<b>0.9098</b>	<b>0.0798</b>	<b>0.7032</b>	<b>0.8707</b>	<b>0.9320</b>	<b>0.9709</b>	<b>0.9973</b>	<b>1.00</b>	<b>30549</b>	<b>NO</b>	<b>1.00</b>	
<b>phi[8]</b>	<b>0.8861</b>	<b>0.0981</b>	<b>0.6363</b>	<b>0.8358</b>	<b>0.9125</b>	<b>0.9622</b>	<b>0.9966</b>	<b>1.00</b>	<b>16877</b>	<b>NO</b>	<b>1.00</b>	
<b>phi[9]</b>	<b>0.2853</b>	<b>0.1006</b>	<b>0.1363</b>	<b>0.2155</b>	<b>0.2693</b>	<b>0.3370</b>	<b>0.5271</b>	<b>1.00</b>	<b>1163</b>	<b>NO</b>	<b>1.00</b>	
<b>phi[10]</b>	<b>0.7488</b>	<b>0.1748</b>	<b>0.3656</b>	<b>0.6295</b>	<b>0.7762</b>	<b>0.8941</b>	<b>0.9898</b>	<b>1.00</b>	<b>3364</b>	<b>NO</b>	<b>1.00</b>	
<b>phi[11]</b>	<b>0.4866</b>	<b>0.1862</b>	<b>0.1760</b>	<b>0.3480</b>	<b>0.4666</b>	<b>0.6084</b>	<b>0.8946</b>	<b>1.00</b>	<b>3891</b>	<b>NO</b>	<b>1.00</b>	
<b>phi[12]</b>	<b>0.0999</b>	<b>0.1044</b>	<b>0.0024</b>	<b>0.0276</b>	<b>0.0665</b>	<b>0.1367</b>	<b>0.3826</b>	<b>1.00</b>	<b>28456</b>	<b>NO</b>	<b>1.00</b>	
<b>b[1]</b>	0.0254	0.0184	0.0031	0.0120	0.0211	0.0342	0.0723	1.00	3115	NO	1.00	
<b>b[2]</b>	0.1092	0.0392	0.0417	0.0816	0.1058	0.1335	0.1948	1.00	1782	NO	1.00	
<b>b[3]</b>	0.0519	0.0343	0.0026	0.0247	0.0472	0.0738	0.1289	1.00	2276	NO	1.00	
<b>b[4]</b>	0.0466	0.0320	0.0023	0.0213	0.0414	0.0661	0.1209	1.00	1541	NO	1.00	
<b>b[5]</b>	0.0130	0.0122	0.0004	0.0040	0.0095	0.0182	0.0450	1.00	9533	NO	1.00	
<b>b[6]</b>	0.0088	0.0089	0.0002	0.0025	0.0061	0.0121	0.0324	1.00	9330	NO	1.00	
<b>b[7]</b>	0.0314	0.0265	0.0012	0.0121	0.0249	0.0433	0.0996	1.00	634	NO	1.00	
<b>b[8]</b>	0.5730	0.0726	0.4223	0.5253	0.5758	0.6239	0.7061	1.00	636	NO	1.00	
<b>b[9]</b>	0.0300	0.0281	0.0008	0.0093	0.0219	0.0422	0.1031	1.00	2217	NO	1.00	
<b>b[10]</b>	0.0622	0.0434	0.0029	0.0278	0.0551	0.0890	0.1627	1.00	120000	NO	1.00	
<b>b[11]</b>	0.0355	0.0275	0.0013	0.0138	0.0297	0.0508	0.1028	1.00	9576	NO	1.00	
<b>b[12]</b>	0.0132	0.0130	0.0003	0.0038	0.0092	0.0183	0.0480	1.00	120000	NO	1.00	
<b>B[1]</b>	14	11	2	6	11	18	41	1.03	273	NO	1.00	
<b>B[2]</b>	60	21	25	45	58	72	109	1.01	334	NO	1.00	
<b>B[3]</b>	29	20	1	13	25	40	74	1.01	257	NO	1.00	
<b>B[4]</b>	25	17	1	12	22	35	64	1.01	2240	NO	1.00	
<b>B[5]</b>	6	6	0	2	4	9	23	1.00	1822	YES	1.00	
<b>B[6]</b>	4	4	0	1	3	6	16	1.00	2770	YES	1.00	
<b>B[7]</b>	16	13	0	6	13	22	50	1.00	10135	YES	1.00	
<b>B[8]</b>	328	82	189	267	321	382	499	1.04	65	NO	1.00	
<b>B[9]</b>	16	16	0	5	11	22	57	1.00	28633	YES	1.00	
<b>B[10]</b>	35	26	1	15	30	49	96	1.00	932	NO	1.00	
<b>B[11]</b>	19	16	0	7	16	28	59	1.00	1946	YES	1.00	
<b>B[12]</b>	7	7	0	1	4	9	27	1.00	1710	YES	1.00	
<b>N[1]</b>	<b>14</b>	<b>11</b>	<b>2</b>	<b>6</b>	<b>11</b>	<b>18</b>	<b>41</b>	<b>1.03</b>	<b>273</b>	<b>NO</b>	<b>1.00</b>	
<b>N[2]</b>	<b>69</b>	<b>21</b>	<b>37</b>	<b>54</b>	<b>66</b>	<b>81</b>	<b>118</b>	<b>1.02</b>	<b>209</b>	<b>NO</b>	<b>1.00</b>	
<b>N[3]</b>	<b>90</b>	<b>22</b>	<b>55</b>	<b>74</b>	<b>87</b>	<b>102</b>	<b>141</b>	<b>1.04</b>	<b>86</b>	<b>NO</b>	<b>1.00</b>	
<b>N[4]</b>	<b>99</b>	<b>22</b>	<b>65</b>	<b>84</b>	<b>96</b>	<b>111</b>	<b>149</b>	<b>1.04</b>	<b>116</b>	<b>NO</b>	<b>1.00</b>	
<b>N[5]</b>	<b>62</b>	<b>18</b>	<b>35</b>	<b>49</b>	<b>59</b>	<b>72</b>	<b>104</b>	<b>1.01</b>	<b>233</b>	<b>NO</b>	<b>1.00</b>	
<b>N[6]</b>	<b>50</b>	<b>14</b>	<b>29</b>	<b>40</b>	<b>48</b>	<b>57</b>	<b>82</b>	<b>1.01</b>	<b>203</b>	<b>NO</b>	<b>1.00</b>	
<b>N[7]</b>	<b>62</b>	<b>18</b>	<b>37</b>	<b>49</b>	<b>59</b>	<b>71</b>	<b>107</b>	<b>1.01</b>	<b>604</b>	<b>NO</b>	<b>1.00</b>	
<b>N[8]</b>	<b>384</b>	<b>84</b>	<b>243</b>	<b>321</b>	<b>377</b>	<b>438</b>	<b>562</b>	<b>1.04</b>	<b>61</b>	<b>NO</b>	<b>1.00</b>	
<b>N[9]</b>	<b>124</b>	<b>45</b>	<b>58</b>	<b>92</b>	<b>116</b>	<b>147</b>	<b>233</b>	<b>1.00</b>	<b>483</b>	<b>NO</b>	<b>1.00</b>	
<b>N[10]</b>	<b>125</b>	<b>33</b>	<b>73</b>	<b>101</b>	<b>120</b>	<b>143</b>	<b>201</b>	<b>1.02</b>	<b>127</b>	<b>NO</b>	<b>1.00</b>	
<b>N[11]</b>	<b>79</b>	<b>27</b>	<b>38</b>	<b>59</b>	<b>75</b>	<b>94</b>	<b>142</b>	<b>1.02</b>	<b>159</b>	<b>NO</b>	<b>1.00</b>	
<b>N[12]</b>	<b>13</b>	<b>11</b>	<b>1</b>	<b>5</b>	<b>11</b>	<b>18</b>	<b>41</b>	<b>1.00</b>	<b>817</b>	<b>NO</b>	<b>1.00</b>	
<b>FOREST</b>												
<b>par</b>	<b>mean</b>	<b>sd</b>	<b>2.50%</b>	<b>25%</b>	<b>50%</b>	<b>75%</b>	<b>97.50%</b>	<b>Rhat</b>	<b>n.eff</b>	<b>over0</b>	<b>f</b>	
psi	0.2654	0.0230	0.2233	0.2497	0.2641	0.2800	0.3145	1.02	187	NO	1.00	
<b>Nforest</b>	<b>591</b>	<b>48</b>	<b>504</b>	<b>559</b>	<b>588</b>	<b>621</b>	<b>696</b>	<b>1.03</b>	<b>155</b>	<b>NO</b>	<b>1.00</b>	
p[1]	0.2137	0.0280	0.1628	0.1942	0.2124	0.2318	0.2725	1.01	483	NO	1.00	
p[2]	0.2806	0.0299	0.2255	0.2598	0.2795	0.3001	0.3420	1.00	611	NO	1.00	
p[3]	0.2228	0.0280	0.1716	0.2033	0.2215	0.2408	0.2814	1.01	478	NO	1.00	
p[4]	0.2974	0.0330	0.2366	0.2744	0.2960	0.3189	0.3654	1.00	815	NO	1.00	
p[5]	0.1557	0.0223	0.1161	0.1400	0.1542	0.1698	0.2038	1.01	435	NO	1.00	

p[6]	0.3331	0.0445	0.2520	0.3020	0.3312	0.3620	0.4257	1.00	1950	NO	1.00
p[7]	0.2454	0.0277	0.1941	0.2262	0.2444	0.2635	0.3030	1.01	475	NO	1.00
p[8]	0.1421	0.0325	0.0892	0.1193	0.1384	0.1607	0.2166	1.00	1252	NO	1.00
p[9]	0.1541	0.0224	0.1146	0.1383	0.1526	0.1681	0.2024	1.01	446	NO	1.00
p[10]	0.3383	0.0466	0.2532	0.3057	0.3363	0.3686	0.4354	1.00	2251	NO	1.00
p[11]	0.1945	0.0273	0.1455	0.1754	0.1930	0.2120	0.2520	1.01	483	NO	1.00
p[12]	0.3337	0.0447	0.2521	0.3024	0.3318	0.3627	0.4267	1.00	1980	NO	1.00
p[13]	0.1600	0.0227	0.1196	0.1440	0.1586	0.1744	0.2085	1.01	425	NO	1.00
alpha.p	-1.3107	0.1667	-1.6379	-1.4233	-1.3107	-1.1984	-0.9820	1.01	437	NO	1.00
beta1.p	-0.4725	0.1227	-0.7168	-0.5545	-0.4706	-0.3892	-0.2356	1.00	3918	NO	1.00
beta2.p	0.1019	0.1265	-0.1440	0.0162	0.1014	0.1871	0.3514	1.00	9277	YES	0.79
phi[1]	<b>0.4998</b>	<b>0.2892</b>	<b>0.0250</b>	<b>0.2484</b>	<b>0.5002</b>	<b>0.7497</b>	<b>0.9752</b>	<b>1.00</b>	<b>120000</b>	<b>NO</b>	<b>1.00</b>
phi[2]	<b>0.6564</b>	<b>0.1087</b>	<b>0.4576</b>	<b>0.5799</b>	<b>0.6515</b>	<b>0.7275</b>	<b>0.8838</b>	<b>1.00</b>	<b>2946</b>	<b>NO</b>	<b>1.00</b>
phi[3]	<b>0.8376</b>	<b>0.1102</b>	<b>0.5915</b>	<b>0.7649</b>	<b>0.8540</b>	<b>0.9276</b>	<b>0.9924</b>	<b>1.00</b>	<b>15363</b>	<b>NO</b>	<b>1.00</b>
phi[4]	<b>0.5088</b>	<b>0.0974</b>	<b>0.3367</b>	<b>0.4401</b>	<b>0.5026</b>	<b>0.5706</b>	<b>0.7153</b>	<b>1.00</b>	<b>5155</b>	<b>NO</b>	<b>1.00</b>
phi[5]	<b>0.8336</b>	<b>0.1021</b>	<b>0.6144</b>	<b>0.7648</b>	<b>0.8436</b>	<b>0.9147</b>	<b>0.9900</b>	<b>1.00</b>	<b>7316</b>	<b>NO</b>	<b>1.00</b>
phi[6]	<b>0.9025</b>	<b>0.0765</b>	<b>0.7183</b>	<b>0.8579</b>	<b>0.9197</b>	<b>0.9637</b>	<b>0.9966</b>	<b>1.00</b>	<b>7912</b>	<b>NO</b>	<b>1.00</b>
phi[7]	<b>0.9372</b>	<b>0.0541</b>	<b>0.8004</b>	<b>0.9096</b>	<b>0.9518</b>	<b>0.9789</b>	<b>0.9981</b>	<b>1.00</b>	<b>9093</b>	<b>NO</b>	<b>1.00</b>
phi[8]	<b>0.8526</b>	<b>0.1122</b>	<b>0.5883</b>	<b>0.7837</b>	<b>0.8765</b>	<b>0.9435</b>	<b>0.9948</b>	<b>1.00</b>	<b>70236</b>	<b>NO</b>	<b>1.00</b>
phi[9]	<b>0.3087</b>	<b>0.0825</b>	<b>0.1787</b>	<b>0.2502</b>	<b>0.2978</b>	<b>0.3556</b>	<b>0.5000</b>	<b>1.00</b>	<b>120000</b>	<b>NO</b>	<b>1.00</b>
phi[10]	<b>0.7125</b>	<b>0.1591</b>	<b>0.3979</b>	<b>0.5970</b>	<b>0.7178</b>	<b>0.8380</b>	<b>0.9793</b>	<b>1.00</b>	<b>3717</b>	<b>NO</b>	<b>1.00</b>
phi[11]	<b>0.6844</b>	<b>0.1563</b>	<b>0.3853</b>	<b>0.5703</b>	<b>0.6837</b>	<b>0.8020</b>	<b>0.9689</b>	<b>1.00</b>	<b>48465</b>	<b>NO</b>	<b>1.00</b>
phi[12]	<b>0.8447</b>	<b>0.1221</b>	<b>0.5502</b>	<b>0.7731</b>	<b>0.8724</b>	<b>0.9430</b>	<b>0.9947</b>	<b>1.00</b>	<b>120000</b>	<b>NO</b>	<b>1.00</b>
b[1]	0.3243	0.0440	0.2432	0.2935	0.3226	0.3530	0.4150	1.00	7553	NO	1.00
b[2]	0.0144	0.0136	0.0004	0.0044	0.0104	0.0203	0.0504	1.00	97434	NO	1.00
b[3]	0.0947	0.0414	0.0183	0.0654	0.0933	0.1219	0.1799	1.00	20575	NO	1.00
b[4]	0.0606	0.0313	0.0066	0.0375	0.0590	0.0813	0.1262	1.00	3094	NO	1.00
b[5]	0.0388	0.0290	0.0016	0.0157	0.0329	0.0556	0.1086	1.00	30814	NO	1.00
b[6]	0.0910	0.0421	0.0145	0.0610	0.0892	0.1184	0.1790	1.00	6437	NO	1.00
b[7]	0.1332	0.0462	0.0414	0.1019	0.1336	0.1647	0.2230	1.00	9571	NO	1.00
b[8]	0.0620	0.0452	0.0028	0.0264	0.0534	0.0883	0.1700	1.00	2753	NO	1.00
b[9]	0.0097	0.0091	0.0003	0.0030	0.0070	0.0136	0.0337	1.00	14429	NO	1.00
b[10]	0.0128	0.0106	0.0004	0.0046	0.0102	0.0183	0.0395	1.00	120000	NO	1.00
b[11]	0.0112	0.0108	0.0003	0.0034	0.0080	0.0156	0.0401	1.00	32031	NO	1.00
b[12]	0.1474	0.0334	0.0870	0.1242	0.1457	0.1687	0.2183	1.00	3582	NO	1.00
B[1]	194	30	144	174	191	212	263	1.02	343	NO	1.00
B[2]	8	8	0	2	5	11	28	1.00	43392	YES	1.00
B[3]	56	25	11	39	55	72	107	1.00	1923	NO	1.00
B[4]	35	18	4	22	35	47	72	1.00	7669	NO	1.00
B[5]	22	17	0	9	19	32	63	1.00	22237	YES	1.00
B[6]	54	24	9	37	53	69	105	1.00	120000	NO	1.00
B[7]	79	27	25	61	79	98	133	1.00	1458	NO	1.00
B[8]	37	28	1	15	31	52	106	1.00	1212	NO	1.00
B[9]	5	5	0	1	3	7	18	1.00	3446	YES	1.00
B[10]	7	6	0	2	5	10	21	1.00	14048	YES	1.00
B[11]	6	6	0	1	4	8	22	1.00	12367	YES	1.00
B[12]	88	21	53	74	87	101	134	1.00	11651	NO	1.00
N[1]	<b>194</b>	<b>30</b>	<b>144</b>	<b>174</b>	<b>191</b>	<b>212</b>	<b>263</b>	<b>1.02</b>	<b>343</b>	<b>NO</b>	<b>1.00</b>
N[2]	<b>134</b>	<b>22</b>	<b>97</b>	<b>119</b>	<b>133</b>	<b>148</b>	<b>181</b>	<b>1.00</b>	<b>1628</b>	<b>NO</b>	<b>1.00</b>
N[3]	<b>169</b>	<b>26</b>	<b>124</b>	<b>151</b>	<b>167</b>	<b>186</b>	<b>226</b>	<b>1.01</b>	<b>557</b>	<b>NO</b>	<b>1.00</b>
N[4]	<b>121</b>	<b>17</b>	<b>91</b>	<b>109</b>	<b>120</b>	<b>132</b>	<b>159</b>	<b>1.00</b>	<b>1005</b>	<b>NO</b>	<b>1.00</b>
N[5]	<b>124</b>	<b>21</b>	<b>88</b>	<b>109</b>	<b>122</b>	<b>137</b>	<b>169</b>	<b>1.00</b>	<b>965</b>	<b>NO</b>	<b>1.00</b>
N[6]	<b>166</b>	<b>23</b>	<b>127</b>	<b>149</b>	<b>164</b>	<b>180</b>	<b>218</b>	<b>1.00</b>	<b>1983</b>	<b>NO</b>	<b>1.00</b>
N[7]	<b>236</b>	<b>26</b>	<b>189</b>	<b>218</b>	<b>234</b>	<b>252</b>	<b>290</b>	<b>1.01</b>	<b>426</b>	<b>NO</b>	<b>1.00</b>
N[8]	<b>238</b>	<b>40</b>	<b>161</b>	<b>211</b>	<b>237</b>	<b>264</b>	<b>320</b>	<b>1.01</b>	<b>459</b>	<b>NO</b>	<b>1.00</b>
N[9]	<b>77</b>	<b>18</b>	<b>50</b>	<b>64</b>	<b>74</b>	<b>86</b>	<b>118</b>	<b>1.00</b>	<b>804</b>	<b>NO</b>	<b>1.00</b>
N[10]	<b>60</b>	<b>10</b>	<b>43</b>	<b>53</b>	<b>59</b>	<b>66</b>	<b>83</b>	<b>1.00</b>	<b>1640</b>	<b>NO</b>	<b>1.00</b>
N[11]	<b>47</b>	<b>11</b>	<b>28</b>	<b>39</b>	<b>46</b>	<b>53</b>	<b>71</b>	<b>1.00</b>	<b>2322</b>	<b>NO</b>	<b>1.00</b>
N[12]	<b>128</b>	<b>22</b>	<b>92</b>	<b>113</b>	<b>126</b>	<b>142</b>	<b>178</b>	<b>1.00</b>	<b>3085</b>	<b>NO</b>	<b>1.00</b>
gamma[1]	0.0861	0.0144	0.0616	0.0761	0.0848	0.0948	0.1183	1.01	399	NO	1.00
gamma[2]	0.0042	0.0040	0.0001	0.0013	0.0030	0.0059	0.0147	1.00	36972	NO	1.00
gamma[3]	0.0276	0.0124	0.0053	0.0189	0.0270	0.0356	0.0537	1.00	1764	NO	1.00
gamma[4]	0.0180	0.0092	0.0020	0.0113	0.0175	0.0241	0.0375	1.00	9859	NO	1.00
gamma[5]	0.0118	0.0089	0.0005	0.0048	0.0100	0.0169	0.0331	1.00	18602	NO	1.00
gamma[6]	0.0279	0.0129	0.0046	0.0188	0.0273	0.0362	0.0553	1.00	120000	NO	1.00
gamma[7]	0.0423	0.0150	0.0134	0.0321	0.0420	0.0521	0.0724	1.00	1190	NO	1.00
gamma[8]	0.0207	0.0155	0.0009	0.0087	0.0176	0.0293	0.0587	1.00	1112	NO	1.00
gamma[9]	0.0033	0.0031	0.0001	0.0010	0.0024	0.0046	0.0115	1.00	3588	NO	1.00
gamma[10]	0.0043	0.0036	0.0002	0.0016	0.0035	0.0062	0.0133	1.00	8601	NO	1.00

gamma[11]	0.0038	0.0038	0.0001	0.0012	0.0027	0.0053	0.0138	1.00	9978	NO	1.00
gamma[12]	0.0507	0.0129	0.0289	0.0417	0.0495	0.0585	0.0793	1.00	3710	NO	1.00
deviance	1800	81	1644	1744	1799	1855	1961	1.01	332	NO	1.00

<b>G) Saltator maximus - Closed Population Model</b>											
<b>COFFEE</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
N	<b>208</b>	<b>48</b>	<b>139</b>	<b>175</b>	<b>199</b>	<b>232</b>	<b>320</b>	<b>1.00</b>	<b>1398</b>	<b>NO</b>	<b>1.00</b>
p.est[1]	0.0047	0.0025	0.001423	0.002921	0.0042	0.0059	0.0108	1.00	3000	NO	1.00
p.est[2]	0.0718	0.0177	0.0415	0.0591	0.0708	0.0830	0.1092	1.00	3000	NO	1.00
p.est[3]	0.0126	0.0045	0.0057	0.0094	0.0120	0.0152	0.0230	1.00	3000	NO	1.00
p.est[4]	0.0681	0.0186	0.0369	0.0549	0.0663	0.0794	0.1088	1.00	2933	NO	1.00
p.est[5]	0.0179	0.0056	0.0090	0.0139	0.0173	0.0213	0.0307	1.00	3000	NO	1.00
p.est[6]	0.0025	0.0016	0.0005	0.0013	0.0021	0.0032	0.0068	1.00	3000	NO	1.00
p.est[7]	0.0747	0.0184	0.0430	0.0615	0.0735	0.0865	0.1134	1.00	3000	NO	1.00
p.est[8]	0.0585	0.0182	0.0291	0.0455	0.0563	0.0692	0.0999	1.00	3000	NO	1.00
p.est[9]	0.0490	0.0123	0.0279	0.0403	0.0482	0.0569	0.0757	1.00	3000	NO	1.00
p.est[10]	0.0341	0.0090	0.0189	0.0276	0.0334	0.0395	0.0536	1.00	3000	NO	1.00
p.est[11]	0.0601	0.0150	0.0346	0.0494	0.0593	0.0693	0.0918	1.00	3000	NO	1.00
p.est[12]	0.0039	0.0022	0.0011	0.0024	0.0035	0.0050	0.0095	1.00	3000	NO	1.00
mean.p	0.0419	0.0107	0.0234	0.0342	0.0412	0.0487	0.0652	1.00	3000	NO	1.00
beta1	1.3753	0.2509	0.9126	1.2012	1.3687	1.5347	1.8954	1.00	1047	NO	1.00
beta2	-0.6979	0.1718	-1.0547	-0.8068	-0.6952	-0.5843	-0.3626	1.00	2400	NO	1.00
psi	0.0514	0.0122	0.0330	0.0428	0.0495	0.0576	0.0802	1.00	1305	NO	1.00
deviance	721	40	648	693	717	746	806	1.00	1549	NO	1.00
<b>FOREST</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
N	<b>165</b>	<b>31</b>	<b>117</b>	<b>143</b>	<b>160</b>	<b>182</b>	<b>241</b>	<b>1.01</b>	<b>692</b>	<b>NO</b>	<b>1.00</b>
p.est[1]	0.0292	0.0071	0.0169	0.0242	0.0285	0.0338	0.0441	1.00	1409	NO	1.00
p.est[2]	0.0445	0.0107	0.0259	0.0371	0.0435	0.0512	0.0680	1.00	1678	NO	1.00
p.est[3]	0.0262	0.0067	0.0148	0.0214	0.0255	0.0305	0.0407	1.00	1481	NO	1.00
p.est[4]	0.0774	0.0169	0.0473	0.0657	0.0762	0.0881	0.1135	1.00	1420	NO	1.00
p.est[5]	0.0211	0.0065	0.0107	0.0164	0.0203	0.0251	0.0355	1.00	1852	NO	1.00
p.est[6]	0.0374	0.0088	0.0220	0.0313	0.0366	0.0429	0.0564	1.00	1491	NO	1.00
p.est[7]	0.0890	0.0251	0.0472	0.0715	0.0864	0.1038	0.1440	1.00	1873	NO	1.00
p.est[8]	0.0785	0.0171	0.0480	0.0668	0.0772	0.0892	0.1151	1.00	1392	NO	1.00
p.est[9]	0.0205	0.0065	0.0102	0.0158	0.0197	0.0244	0.0350	1.00	1927	NO	1.00
p.est[10]	0.0559	0.0137	0.0326	0.0463	0.0547	0.0645	0.0860	1.00	1879	NO	1.00
p.est[11]	0.0210	0.0065	0.0106	0.0163	0.0203	0.0250	0.0354	1.00	1860	NO	1.00
p.est[12]	0.0746	0.0166	0.0449	0.0631	0.0733	0.0850	0.1104	1.00	1508	NO	1.00
mean.p	0.0478	0.0116	0.0280	0.0398	0.0466	0.0551	0.0733	1.00	1761	NO	1.00
beta1	0.6534	0.1551	0.3659	0.5470	0.6480	0.7554	0.9679	1.00	3000	NO	1.00
beta2	-0.1581	0.1551	-0.4618	-0.2618	-0.1591	-0.0512	0.1475	1.00	3000	YES	0.85
psi	0.0408	0.0083	0.0277	0.0349	0.0399	0.0455	0.0594	1.01	557	NO	1.00
deviance	718	34	656	694	716	740	792	1.00	946	NO	1.00

<b>H) Saltator striatipectus - Closed Population Model</b>											
<b>COFFEE</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
N	<b>235</b>	<b>82</b>	<b>135</b>	<b>178</b>	<b>216</b>	<b>271</b>	<b>430</b>	<b>1.03</b>	<b>148</b>	<b>NO</b>	<b>1.00</b>
p.est[1]	0.0100	0.0045	0.003427	0.006847	0.0093	0.0125	0.0209	1.01	340	NO	1.00
p.est[2]	0.0338	0.0114	0.0153	0.0254	0.0327	0.0413	0.0592	1.01	314	NO	1.00
p.est[3]	0.0135	0.0049	0.0056	0.0099	0.0128	0.0164	0.0245	1.01	280	NO	1.00
p.est[4]	0.0591	0.0196	0.0268	0.0446	0.0571	0.0717	0.1013	1.01	216	NO	1.00
p.est[5]	0.0153	0.0054	0.0066	0.0113	0.0146	0.0185	0.0273	1.01	279	NO	1.00
p.est[6]	0.0086	0.0045	0.0024	0.0054	0.0077	0.0108	0.0199	1.01	434	NO	1.00
p.est[7]	0.0359	0.0121	0.0164	0.0270	0.0347	0.0437	0.0627	1.01	305	NO	1.00
p.est[8]	0.0671	0.0237	0.0289	0.0499	0.0640	0.0813	0.1194	1.01	215	NO	1.00
p.est[9]	0.0245	0.0084	0.0109	0.0183	0.0235	0.0298	0.0428	1.01	322	NO	1.00
p.est[10]	0.0201	0.0069	0.0089	0.0150	0.0193	0.0243	0.0354	1.01	303	NO	1.00
p.est[11]	0.0283	0.0097	0.0127	0.0212	0.0272	0.0345	0.0496	1.01	326	NO	1.00
p.est[12]	0.0096	0.0045	0.0031	0.0064	0.0088	0.0120	0.0204	1.01	361	NO	1.00
mean.p	0.0223	0.0077	0.0099	0.0167	0.0215	0.0271	0.0393	1.01	314	NO	1.00
beta1	0.7404	0.1843	0.4043	0.6102	0.7348	0.8638	1.1095	1.00	3000	NO	1.00
beta2	-0.0364	0.1491	-0.3407	-0.1332	-0.0360	0.0669	0.2448	1.00	1010	YES	0.60
psi	0.0580	0.0203	0.0324	0.0440	0.0537	0.0673	0.1078	1.03	156	NO	1.00
deviance	621	44	547	588	616	649	713	1.02	153	NO	1.00
<b>FOREST</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f

N	<b>2072</b>	<b>1058</b>	<b>270</b>	<b>1054</b>	<b>2317</b>	<b>2971</b>	<b>3679</b>	<b>1.11</b>	<b>23</b>	<b>NO</b>	<b>1.00</b>
p.est[1]	0.0011	0.0014	0.0002	0.0004	0.0006	0.0011	0.0051	1.07	64	NO	1.00
p.est[2]	0.0020	0.0025	0.0004	0.0007	0.0010	0.0021	0.0090	1.09	54	NO	1.00
p.est[3]	0.0010	0.0013	0.0002	0.0003	0.0005	0.0010	0.0045	1.06	68	NO	1.00
p.est[4]	0.0045	0.0056	0.0011	0.0017	0.0023	0.0047	0.0205	1.10	48	NO	1.00
p.est[5]	0.0007	0.0011	0.0001	0.0002	0.0004	0.0007	0.0036	1.06	82	NO	1.00
p.est[6]	0.0015	0.0020	0.0003	0.0006	0.0008	0.0016	0.0070	1.08	57	NO	1.00
p.est[7]	0.0059	0.0080	0.0011	0.0021	0.0031	0.0060	0.0262	1.09	58	NO	1.00
p.est[8]	0.0046	0.0057	0.0011	0.0017	0.0023	0.0048	0.0206	1.10	48	NO	1.00
p.est[9]	0.0007	0.0010	0.0001	0.0002	0.0004	0.0007	0.0036	1.05	85	NO	1.00
p.est[10]	0.0027	0.0034	0.0006	0.0010	0.0014	0.0029	0.0124	1.09	51	NO	1.00
p.est[11]	0.0007	0.0011	0.0001	0.0002	0.0004	0.0007	0.0036	1.06	82	NO	1.00
p.est[12]	0.0042	0.0052	0.0010	0.0016	0.0022	0.0045	0.0190	1.10	48	NO	1.00
mean.p	0.0022	0.0027	0.0005	0.0008	0.0011	0.0023	0.0100	1.09	53	NO	1.00
beta1	0.9027	0.2878	0.3693	0.7030	0.8890	1.0927	1.4943	1.00	2335	NO	1.00
beta2	-0.1946	0.2533	-0.6951	-0.3658	-0.1871	-0.0206	0.2952	1.00	3000	YES	0.77
psi	0.5138	0.2621	0.0674	0.2627	0.5736	0.7348	0.9125	1.11	23	NO	1.00
deviance	487	52	363	456	509	526	541	1.09	29	NO	1.00

**I) Myiothlypis conspicillata - Jolly-Seber Model****COFFEE**

par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
psi	0.1590	0.0474	0.0903	0.1261	0.1519	0.1817	0.2718	1.08	58	0	1.00
<b>Ncoffee</b>	<b>317</b>	<b>97</b>	<b>178</b>	<b>250</b>	<b>302</b>	<b>363</b>	<b>546</b>	<b>1.08</b>	<b>56</b>	<b>0</b>	<b>1.00</b>
p[1]	0.1472	0.0485	0.0697	0.1125	0.1410	0.1752	0.2590	1.03	100	0	1.00
p[2]	0.1530	0.0666	0.0572	0.1045	0.1414	0.1891	0.3141	1.03	134	0	1.00
p[3]	0.1069	0.0370	0.0489	0.0805	0.1019	0.1279	0.1933	1.03	111	0	1.00
p[4]	0.1615	0.0550	0.0744	0.1220	0.1543	0.1935	0.2884	1.04	94	0	1.00
p[5]	0.0431	0.0202	0.0148	0.0287	0.0395	0.0535	0.0924	1.02	189	0	1.00
p[6]	0.1616	0.0528	0.0768	0.1238	0.1551	0.1925	0.2821	1.04	90	0	1.00
p[7]	0.1445	0.0774	0.0424	0.0881	0.1282	0.1835	0.3381	1.02	187	0	1.00
p[8]	0.0998	0.0348	0.0454	0.0750	0.0950	0.1194	0.1810	1.03	112	0	1.00
p[9]	0.0329	0.0184	0.0090	0.0199	0.0290	0.0417	0.0793	1.01	264	0	1.00
p[10]	0.1403	0.0468	0.0659	0.1068	0.1342	0.1671	0.2483	1.03	104	0	1.00
p[11]	0.1536	0.0500	0.0733	0.1179	0.1474	0.1826	0.2681	1.03	96	0	1.00
p[12]	0.1269	0.0432	0.0588	0.0961	0.1212	0.1515	0.2272	1.03	108	0	1.00
p[13]	0.1508	0.0694	0.0533	0.1002	0.1380	0.1878	0.3200	1.02	147	0	1.00
alpha.p	-1.8093	0.3904	-2.5908	-2.0651	-1.8066	-1.5492	-1.0512	1.04	89	0	1.00
beta1.p	-0.4050	0.2501	-0.9056	-0.5712	-0.4023	-0.2349	0.0762	1.00	1983	1	0.95
beta2.p	-0.3643	0.2249	-0.8176	-0.5129	-0.3593	-0.2110	0.0624	1.00	13711	1	0.95
<b>phi[1]</b>	<b>0.5000</b>	<b>0.2886</b>	<b>0.0252</b>	<b>0.2499</b>	<b>0.5002</b>	<b>0.7502</b>	<b>0.9753</b>	<b>1.00</b>	<b>3.00E+05</b>	<b>0</b>	<b>1.00</b>
<b>phi[2]</b>	<b>0.4638</b>	<b>0.2864</b>	<b>0.0205</b>	<b>0.2134</b>	<b>0.4467</b>	<b>0.7045</b>	<b>0.9687</b>	<b>1.00</b>	<b>3.00E+05</b>	<b>0</b>	<b>1.00</b>
<b>phi[3]</b>	<b>0.6271</b>	<b>0.2107</b>	<b>0.2272</b>	<b>0.4662</b>	<b>0.6331</b>	<b>0.7987</b>	<b>0.9767</b>	<b>1.00</b>	<b>8520</b>	<b>0</b>	<b>1.00</b>
<b>phi[4]</b>	<b>0.8130</b>	<b>0.1474</b>	<b>0.4560</b>	<b>0.7265</b>	<b>0.8469</b>	<b>0.9315</b>	<b>0.9938</b>	<b>1.00</b>	<b>57180</b>	<b>0</b>	<b>1.00</b>
<b>phi[5]</b>	<b>0.5466</b>	<b>0.2093</b>	<b>0.1974</b>	<b>0.3823</b>	<b>0.5262</b>	<b>0.7016</b>	<b>0.9585</b>	<b>1.00</b>	<b>23240</b>	<b>0</b>	<b>1.00</b>
<b>phi[6]</b>	<b>0.6541</b>	<b>0.2086</b>	<b>0.2499</b>	<b>0.4946</b>	<b>0.6654</b>	<b>0.8288</b>	<b>0.9821</b>	<b>1.00</b>	<b>41155</b>	<b>0</b>	<b>1.00</b>
<b>phi[7]</b>	<b>0.8472</b>	<b>0.1277</b>	<b>0.5274</b>	<b>0.7780</b>	<b>0.8801</b>	<b>0.9478</b>	<b>0.9953</b>	<b>1.00</b>	<b>150451</b>	<b>0</b>	<b>1.00</b>
<b>phi[8]</b>	<b>0.8466</b>	<b>0.1280</b>	<b>0.5255</b>	<b>0.7772</b>	<b>0.8796</b>	<b>0.9472</b>	<b>0.9953</b>	<b>1.00</b>	<b>99877</b>	<b>0</b>	<b>1.00</b>
<b>phi[9]</b>	<b>0.7809</b>	<b>0.1526</b>	<b>0.4437</b>	<b>0.6778</b>	<b>0.8055</b>	<b>0.9072</b>	<b>0.9909</b>	<b>1.00</b>	<b>31684</b>	<b>0</b>	<b>1.00</b>
<b>phi[10]</b>	<b>0.6661</b>	<b>0.1860</b>	<b>0.3115</b>	<b>0.5251</b>	<b>0.6693</b>	<b>0.8162</b>	<b>0.9786</b>	<b>1.00</b>	<b>6667</b>	<b>0</b>	<b>1.00</b>
<b>phi[11]</b>	<b>0.7755</b>	<b>0.1503</b>	<b>0.4478</b>	<b>0.6733</b>	<b>0.7958</b>	<b>0.8993</b>	<b>0.9898</b>	<b>1.00</b>	<b>22127</b>	<b>0</b>	<b>1.00</b>
<b>phi[12]</b>	<b>0.0972</b>	<b>0.1001</b>	<b>0.0024</b>	<b>0.0273</b>	<b>0.0661</b>	<b>0.1335</b>	<b>0.3691</b>	<b>1.00</b>	<b>68919</b>	<b>0</b>	<b>1.00</b>
b[1]	0.0464	0.0332	0.0055	0.0220	0.0386	0.0624	0.1309	1.00	53231	0	1.00
b[2]	0.1857	0.0768	0.0615	0.1306	0.1763	0.2314	0.3616	1.00	3272	0	1.00
b[3]	0.0432	0.0379	0.0013	0.0143	0.0330	0.0618	0.1406	1.00	64970	0	1.00
b[4]	0.0945	0.0589	0.0064	0.0491	0.0872	0.1314	0.2276	1.00	1962	0	1.00
b[5]	0.0228	0.0242	0.0005	0.0063	0.0153	0.0311	0.0882	1.00	5692	0	1.00
b[6]	0.0186	0.0189	0.0005	0.0053	0.0127	0.0256	0.0695	1.00	38591	0	1.00
b[7]	0.0845	0.0733	0.0033	0.0300	0.0641	0.1179	0.2770	1.00	9949	0	1.00
b[8]	0.3277	0.1257	0.0665	0.2441	0.3331	0.4158	0.5606	1.00	2417	0	1.00
b[9]	0.0703	0.0634	0.0020	0.0227	0.0527	0.0996	0.2364	1.00	9089	0	1.00
b[10]	0.0486	0.0427	0.0015	0.0161	0.0370	0.0694	0.1587	1.00	7938	0	1.00
b[11]	0.0314	0.0298	0.0009	0.0095	0.0226	0.0441	0.1102	1.00	57011	0	1.00
b[12]	0.0261	0.0260	0.0007	0.0075	0.0182	0.0363	0.0960	1.00	112952	0	1.00
B[1]	14	12	2	6	11	19	45	1.01	359	0	1.00
B[2]	60	32	18	38	53	74	142	1.03	152	0	1.00
B[3]	13	13	0	4	9	18	47	1.01	544	1	1.00
B[4]	30	21	2	15	26	40	82	1.02	251	0	1.00
B[5]	7	8	0	1	4	9	29	1.01	589	1	1.00

B[6]	5	6	0	1	3	7	22	1.01	588	1	1.00
B[7]	27	26	1	9	19	36	96	1.01	622	0	1.00
B[8]	107	54	21	70	100	135	238	1.03	156	0	1.00
B[9]	22	24	0	6	16	31	84	1.01	606	1	1.00
B[10]	15	15	0	4	11	21	54	1.01	943	1	1.00
B[11]	9	11	0	2	6	13	38	1.01	635	1	1.00
B[12]	8	10	0	2	5	11	33	1.01	678	1	1.00
N[1]	14	12	2	6	11	19	45	1.01	359	0	1.00
N[2]	67	33	26	44	59	81	151	1.03	126	0	1.00
N[3]	53	24	20	36	49	65	112	1.04	115	0	1.00
N[4]	73	28	36	54	68	86	143	1.05	83	0	1.00
N[5]	46	23	15	29	41	57	105	1.03	142	0	1.00
N[6]	33	17	12	22	30	41	75	1.04	130	0	1.00
N[7]	56	32	19	33	47	69	138	1.02	198	0	1.00
N[8]	155	55	78	116	145	180	294	1.06	81	0	1.00
N[9]	143	56	66	105	133	169	282	1.05	96	0	1.00
N[10]	107	40	53	79	100	126	205	1.05	104	0	1.00
N[11]	92	36	43	67	85	109	181	1.04	108	0	1.00
N[12]	16	14	1	6	12	21	53	1.01	437	0	1.00
gamma[1]	0.0074	0.0060	0.0008	0.0032	0.0058	0.0098	0.0232	1.01	400	0	1.00
gamma[2]	0.0298	0.0161	0.0086	0.0187	0.0264	0.0371	0.0709	1.03	160	0	1.00
gamma[3]	0.0071	0.0068	0.0002	0.0022	0.0052	0.0099	0.0249	1.01	533	0	1.00
gamma[4]	0.0156	0.0112	0.0010	0.0076	0.0134	0.0210	0.0432	1.02	249	0	1.00
gamma[5]	0.0039	0.0046	0.0001	0.0010	0.0024	0.0051	0.0161	1.01	596	0	1.00
gamma[6]	0.0032	0.0035	0.0001	0.0008	0.0020	0.0042	0.0127	1.01	591	0	1.00
gamma[7]	0.0144	0.0142	0.0005	0.0048	0.0102	0.0195	0.0520	1.01	580	0	1.00
gamma[8]	0.0565	0.0293	0.0112	0.0368	0.0523	0.0710	0.1279	1.04	135	0	1.00
gamma[9]	0.0131	0.0137	0.0003	0.0038	0.0091	0.0178	0.0490	1.02	465	0	1.00
gamma[10]	0.0090	0.0088	0.0003	0.0028	0.0065	0.0124	0.0322	1.01	621	0	1.00
gamma[11]	0.0060	0.0067	0.0001	0.0017	0.0040	0.0081	0.0235	1.02	473	0	1.00
gamma[12]	0.0051	0.0061	0.0001	0.0013	0.0032	0.0067	0.0212	1.02	495	0	1.00
deviance	531	54	429	494	529	565	644	1.05	66	0	1.00
<b>FOREST</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
psi	0.2805	0.0465	0.2025	0.2477	0.2768	0.3081	0.3853	1.03	138	0	1.00
Nforest	<b>604</b>	<b>100</b>	<b>438</b>	<b>533</b>	<b>596</b>	<b>663</b>	<b>831</b>	<b>1.03</b>	<b>131</b>	<b>0</b>	<b>1.00</b>
p[1]	0.1146	0.0281	0.0684	0.0945	0.1114	0.1314	0.1781	1.01	312	0	1.00
p[2]	0.1955	0.0376	0.1304	0.1688	0.1923	0.2189	0.2770	1.01	274	0	1.00
p[3]	0.1220	0.0291	0.0740	0.1013	0.1188	0.1394	0.1872	1.01	299	0	1.00
p[4]	0.2254	0.0423	0.1519	0.1954	0.2221	0.2519	0.3173	1.01	308	0	1.00
p[5]	0.1140	0.0269	0.0702	0.0948	0.1109	0.1299	0.1752	1.01	558	0	1.00
p[6]	0.3016	0.0579	0.2006	0.2607	0.2973	0.3379	0.4266	1.01	516	0	1.00
p[7]	0.1453	0.0316	0.0921	0.1228	0.1422	0.1647	0.2150	1.01	272	0	1.00
p[8]	0.1936	0.0609	0.1014	0.1491	0.1844	0.2285	0.3369	1.00	9664	0	1.00
p[9]	0.1172	0.0277	0.0721	0.0974	0.1140	0.1336	0.1802	1.01	611	0	1.00
p[10]	0.3140	0.0608	0.2077	0.2710	0.3095	0.3522	0.4451	1.01	570	0	1.00
p[11]	0.1033	0.0264	0.0607	0.0845	0.1002	0.1189	0.1632	1.01	343	0	1.00
p[12]	0.3029	0.0582	0.2013	0.2618	0.2986	0.3394	0.4286	1.01	522	0	1.00
p[13]	0.1079	0.0256	0.0662	0.0896	0.1049	0.1231	0.1661	1.01	475	0	1.00
alpha.p	-0.2037	0.2755	-2.6109	-2.2596	-2.0764	-1.8884	-1.5292	1.01	247	0	1.00
beta1.p	-0.5866	0.1513	-0.8906	-0.6864	-0.5841	-0.4842	-0.2957	1.00	172331	0	1.00
beta2.p	0.5437	0.1736	0.2161	0.4246	0.5393	0.6582	0.8958	1.00	1507	0	1.00
phi[1]	<b>0.4999</b>	<b>0.2886</b>	<b>0.0249</b>	<b>0.2507</b>	<b>0.4993</b>	<b>0.7490</b>	<b>0.9751</b>	<b>1.00</b>	<b>300000</b>	<b>0</b>	<b>1.00</b>
phi[2]	<b>0.6489</b>	<b>0.2036</b>	<b>0.2420</b>	<b>0.4996</b>	<b>0.6602</b>	<b>0.8134</b>	<b>0.9782</b>	<b>1.00</b>	<b>3118</b>	<b>0</b>	<b>1.00</b>
phi[3]	<b>0.6308</b>	<b>0.1902</b>	<b>0.2725</b>	<b>0.4891</b>	<b>0.6292</b>	<b>0.7775</b>	<b>0.9698</b>	<b>1.00</b>	<b>55939</b>	<b>0</b>	<b>1.00</b>
phi[4]	<b>0.6886</b>	<b>0.1905</b>	<b>0.2991</b>	<b>0.5503</b>	<b>0.7042</b>	<b>0.8450</b>	<b>0.9833</b>	<b>1.00</b>	<b>12738</b>	<b>0</b>	<b>1.00</b>
phi[5]	<b>0.4095</b>	<b>0.1315</b>	<b>0.2024</b>	<b>0.3161</b>	<b>0.3919</b>	<b>0.4837</b>	<b>0.7206</b>	<b>1.00</b>	<b>47735</b>	<b>0</b>	<b>1.00</b>
phi[6]	<b>0.7992</b>	<b>0.1486</b>	<b>0.4552</b>	<b>0.7049</b>	<b>0.8282</b>	<b>0.9212</b>	<b>0.9926</b>	<b>1.00</b>	<b>48997</b>	<b>0</b>	<b>1.00</b>
phi[7]	<b>0.9056</b>	<b>0.0835</b>	<b>0.6901</b>	<b>0.8644</b>	<b>0.9287</b>	<b>0.9696</b>	<b>0.9973</b>	<b>1.00</b>	<b>57761</b>	<b>0</b>	<b>1.00</b>
phi[8]	<b>0.6290</b>	<b>0.1768</b>	<b>0.3171</b>	<b>0.4935</b>	<b>0.6187</b>	<b>0.7612</b>	<b>0.9654</b>	<b>1.00</b>	<b>1347</b>	<b>0</b>	<b>1.00</b>
phi[9]	<b>0.3972</b>	<b>0.1172</b>	<b>0.2144</b>	<b>0.3126</b>	<b>0.3812</b>	<b>0.4645</b>	<b>0.6700</b>	<b>1.00</b>	<b>2548</b>	<b>0</b>	<b>1.00</b>
phi[10]	<b>0.8154</b>	<b>0.1334</b>	<b>0.5105</b>	<b>0.7298</b>	<b>0.8397</b>	<b>0.9246</b>	<b>0.9928</b>	<b>1.00</b>	<b>43610</b>	<b>0</b>	<b>1.00</b>
phi[11]	<b>0.7965</b>	<b>0.1362</b>	<b>0.4974</b>	<b>0.7043</b>	<b>0.8152</b>	<b>0.9085</b>	<b>0.9906</b>	<b>1.00</b>	<b>7026</b>	<b>0</b>	<b>1.00</b>
phi[12]	<b>0.7637</b>	<b>0.1549</b>	<b>0.4315</b>	<b>0.6561</b>	<b>0.7822</b>	<b>0.8914</b>	<b>0.9888</b>	<b>1.00</b>	<b>21312</b>	<b>0</b>	<b>1.00</b>
b[1]	0.1243	0.0399	0.0577	0.0957	0.1204	0.1485	0.2127	1.00	1487	0	1.00
b[2]	0.0502	0.0323	0.0028	0.0247	0.0463	0.0709	0.1223	1.00	56043	0	1.00
b[3]	0.0398	0.0308	0.0016	0.0159	0.0331	0.0568	0.1155	1.00	17333	0	1.00
b[4]	0.1117	0.0417	0.0293	0.0838	0.1116	0.1393	0.1952	1.00	10937	0	1.00
b[5]	0.0122	0.0115	0.0003	0.0037	0.0088	0.0171	0.0423	1.00	86042	0	1.00

b[6]	0.0203	0.0165	0.0008	0.0076	0.0165	0.0289	0.0614	1.00	6551	0	1.00
b[7]	0.4396	0.0620	0.3160	0.3980	0.4403	0.4820	0.5590	1.00	3748	0	1.00
b[8]	0.0496	0.0464	0.0013	0.0152	0.0360	0.0702	0.1721	1.00	3791	0	1.00
b[9]	0.0178	0.0159	0.0005	0.0058	0.0135	0.0254	0.0587	1.00	6659	0	1.00
b[10]	0.0280	0.0216	0.0010	0.0108	0.0233	0.0404	0.0801	1.00	7305	0	1.00
b[11]	0.0371	0.0290	0.0014	0.0144	0.0308	0.0534	0.1080	1.00	27713	0	1.00
b[12]	0.0693	0.0356	0.0075	0.0433	0.0674	0.0926	0.1446	1.00	2593	0	1.00
B[1]	76	29	33	55	71	91	147	1.02	259	0	1.00
B[2]	30	20	1	14	27	42	76	1.00	1540	0	1.00
B[3]	24	19	0	9	19	33	72	1.00	1627	1	1.00
B[4]	67	26	18	50	66	83	123	1.00	1128	0	1.00
B[5]	6	7	0	2	5	9	24	1.00	6079	1	1.00
B[6]	11	9	0	4	9	16	35	1.00	58401	1	1.00
B[7]	271	64	167	225	264	309	415	1.02	193	0	1.00
B[8]	29	28	0	9	21	41	104	1.00	23055	1	1.00
B[9]	10	9	0	3	7	14	33	1.00	67720	1	1.00
B[10]	16	12	0	6	13	23	46	1.00	49643	1	1.00
B[11]	22	18	0	8	18	32	66	1.00	2170	1	1.00
B[12]	41	21	4	26	40	54	88	1.00	26384	0	1.00
N[1]	<b>76</b>	<b>29</b>	<b>33</b>	<b>55</b>	<b>71</b>	<b>91</b>	<b>147</b>	<b>1.02</b>	<b>259</b>	<b>0</b>	<b>1.00</b>
N[2]	<b>78</b>	<b>22</b>	<b>44</b>	<b>62</b>	<b>75</b>	<b>90</b>	<b>129</b>	<b>1.01</b>	<b>304</b>	<b>0</b>	<b>1.00</b>
N[3]	<b>73</b>	<b>26</b>	<b>34</b>	<b>54</b>	<b>69</b>	<b>87</b>	<b>134</b>	<b>1.01</b>	<b>425</b>	<b>0</b>	<b>1.00</b>
N[4]	<b>117</b>	<b>27</b>	<b>75</b>	<b>98</b>	<b>114</b>	<b>133</b>	<b>179</b>	<b>1.01</b>	<b>293</b>	<b>0</b>	<b>1.00</b>
N[5]	<b>53</b>	<b>17</b>	<b>28</b>	<b>41</b>	<b>51</b>	<b>63</b>	<b>94</b>	<b>1.01</b>	<b>601</b>	<b>0</b>	<b>1.00</b>
N[6]	<b>54</b>	<b>14</b>	<b>33</b>	<b>44</b>	<b>52</b>	<b>61</b>	<b>86</b>	<b>1.01</b>	<b>641</b>	<b>0</b>	<b>1.00</b>
N[7]	<b>321</b>	<b>69</b>	<b>209</b>	<b>271</b>	<b>314</b>	<b>361</b>	<b>475</b>	<b>1.02</b>	<b>180</b>	<b>0</b>	<b>1.00</b>
N[8]	<b>227</b>	<b>65</b>	<b>125</b>	<b>179</b>	<b>219</b>	<b>266</b>	<b>373</b>	<b>1.00</b>	<b>3185</b>	<b>0</b>	<b>1.00</b>
N[9]	<b>96</b>	<b>26</b>	<b>56</b>	<b>78</b>	<b>93</b>	<b>110</b>	<b>155</b>	<b>1.00</b>	<b>815</b>	<b>0</b>	<b>1.00</b>
N[10]	<b>93</b>	<b>19</b>	<b>63</b>	<b>80</b>	<b>91</b>	<b>104</b>	<b>137</b>	<b>1.01</b>	<b>577</b>	<b>0</b>	<b>1.00</b>
N[11]	<b>97</b>	<b>25</b>	<b>56</b>	<b>79</b>	<b>94</b>	<b>112</b>	<b>155</b>	<b>1.01</b>	<b>480</b>	<b>0</b>	<b>1.00</b>
N[12]	<b>115</b>	<b>26</b>	<b>74</b>	<b>97</b>	<b>112</b>	<b>131</b>	<b>174</b>	<b>1.01</b>	<b>572</b>	<b>0</b>	<b>1.00</b>
gamma[1]	0.0352	0.0139	0.0145	0.0253	0.0330	0.0426	0.0686	1.02	282	0	1.00
gamma[2]	0.0146	0.0098	0.0008	0.0070	0.0131	0.0204	0.0375	1.00	1512	0	1.00
gamma[3]	0.0118	0.0096	0.0005	0.0045	0.0095	0.0167	0.0361	1.00	1501	0	1.00
gamma[4]	0.0331	0.0132	0.0089	0.0242	0.0322	0.0410	0.0619	1.00	982	0	1.00
gamma[5]	0.0037	0.0036	0.0001	0.0011	0.0027	0.0052	0.0131	1.00	5010	0	1.00
gamma[6]	0.0062	0.0051	0.0002	0.0023	0.0050	0.0088	0.0189	1.00	31207	0	1.00
gamma[7]	0.1382	0.0353	0.0817	0.1130	0.1342	0.1588	0.2184	1.03	176	0	1.00
gamma[8]	0.0176	0.0165	0.0005	0.0055	0.0129	0.0246	0.0611	1.00	22990	0	1.00
gamma[9]	0.0065	0.0059	0.0002	0.0021	0.0049	0.0092	0.0219	1.00	10618	0	1.00
gamma[10]	0.0102	0.0079	0.0004	0.0040	0.0085	0.0145	0.0294	1.00	8318	0	1.00
gamma[11]	0.0141	0.0118	0.0005	0.0052	0.0113	0.0200	0.0436	1.00	1245	0	1.00
gamma[12]	0.0260	0.0140	0.0030	0.0162	0.0247	0.0342	0.0575	1.00	4038	0	1.00
deviance	1253	90	1078	1192	1253	1314	1430	1.01	251	0	1.00

<b>J) Basileuterus rufifrons - Jolly-Seber Model</b>											
<b>COFFEE</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
psi	0.2306	0.0450	0.1597	0.1974	0.2250	0.2583	0.3311	1.04	160	0	1.00
<b>Ncoffee</b>	<b>481</b>	<b>94</b>	<b>335</b>	<b>411</b>	<b>469</b>	<b>539</b>	<b>692</b>	<b>1.04</b>	<b>153</b>	<b>0</b>	<b>1.00</b>
p[1]	0.1444	0.0338	0.0880	0.1200	0.1411	0.1652	0.2187	1.01	598	0	1.00
p[2]	0.3269	0.0817	0.1865	0.2681	0.3202	0.3786	0.5038	1.00	1299	0	1.00
p[3]	0.1102	0.0272	0.0657	0.0905	0.1074	0.1268	0.1707	1.01	606	0	1.00
p[4]	0.2328	0.0518	0.1435	0.1958	0.2286	0.2654	0.3450	1.01	716	0	1.00
p[5]	0.1004	0.0284	0.0548	0.0800	0.0970	0.1170	0.1652	1.00	821	0	1.00
p[6]	0.2038	0.0447	0.1269	0.1719	0.2002	0.2319	0.3008	1.01	632	0	1.00
p[7]	0.3977	0.1068	0.2113	0.3201	0.3905	0.4678	0.6246	1.00	2051	0	1.00
p[8]	0.1072	0.0264	0.0639	0.0881	0.1045	0.1234	0.1663	1.01	602	0	1.00
p[9]	0.1045	0.0336	0.0520	0.0804	0.1000	0.1236	0.1824	1.00	1117	0	1.00
p[10]	0.1345	0.0321	0.0813	0.1113	0.1313	0.1543	0.2054	1.01	605	0	1.00
p[11]	0.1577	0.0360	0.0970	0.1318	0.1545	0.1801	0.2366	1.01	592	0	1.00
p[12]	0.1219	0.0297	0.0729	0.1003	0.1188	0.1400	0.1880	1.01	611	0	1.00
p[13]	0.3458	0.0884	0.1936	0.2820	0.3388	0.4022	0.5367	1.00	1472	0	1.00
alpha.p	-1.8059	0.2733	-2.3388	-1.9926	-1.8059	-1.6199	-1.2731	1.01	407	0	1.00
beta1.p	-0.6829	0.1919	-1.0711	-0.8094	-0.6788	-0.5521	-0.3188	1.00	12325	0	1.00
beta2.p	0.2672	0.1545	-0.0345	0.1634	0.2666	0.3700	0.5729	1.00	10109	1	0.96
<b>phi[1]</b>	<b>0.4989</b>	<b>0.2884</b>	<b>0.0247</b>	<b>0.2491</b>	<b>0.4993</b>	<b>0.7483</b>	<b>0.9745</b>	<b>1.00</b>	<b>3.00E+05</b>	<b>0</b>	<b>1.00</b>
<b>phi[2]</b>	<b>0.7948</b>	<b>0.1656</b>	<b>0.3896</b>	<b>0.6989</b>	<b>0.8354</b>	<b>0.9278</b>	<b>0.9934</b>	<b>1.00</b>	<b>50482</b>	<b>0</b>	<b>1.00</b>
<b>phi[3]</b>	<b>0.5692</b>	<b>0.1762</b>	<b>0.2612</b>	<b>0.4382</b>	<b>0.5564</b>	<b>0.6912</b>	<b>0.9332</b>	<b>1.00</b>	<b>8868</b>	<b>0</b>	<b>1.00</b>

<b>phi[4]</b>	<b>0.8234</b>	<b>0.1360</b>	<b>0.5006</b>	<b>0.7404</b>	<b>0.8530</b>	<b>0.9339</b>	<b>0.9938</b>	<b>1.00</b>	<b>30121</b>	<b>0</b>	<b>1.00</b>
<b>phi[5]</b>	<b>0.7920</b>	<b>0.1475</b>	<b>0.4613</b>	<b>0.6945</b>	<b>0.8170</b>	<b>0.9139</b>	<b>0.9917</b>	<b>1.00</b>	<b>34096</b>	<b>0</b>	<b>1.00</b>
<b>phi[6]</b>	<b>0.4905</b>	<b>0.1564</b>	<b>0.2377</b>	<b>0.3758</b>	<b>0.4714</b>	<b>0.5858</b>	<b>0.8507</b>	<b>1.00</b>	<b>14127</b>	<b>0</b>	<b>1.00</b>
<b>phi[7]</b>	<b>0.8912</b>	<b>0.0938</b>	<b>0.6525</b>	<b>0.8427</b>	<b>0.9166</b>	<b>0.9639</b>	<b>0.9967</b>	<b>1.00</b>	<b>280830</b>	<b>0</b>	<b>1.00</b>
<b>phi[8]</b>	<b>0.7895</b>	<b>0.1545</b>	<b>0.4324</b>	<b>0.6911</b>	<b>0.8191</b>	<b>0.9160</b>	<b>0.9918</b>	<b>1.00</b>	<b>46451</b>	<b>0</b>	<b>1.00</b>
<b>phi[9]</b>	<b>0.5498</b>	<b>0.1472</b>	<b>0.3076</b>	<b>0.4428</b>	<b>0.5326</b>	<b>0.6398</b>	<b>0.8869</b>	<b>1.00</b>	<b>11309</b>	<b>0</b>	<b>1.00</b>
<b>phi[10]</b>	<b>0.8042</b>	<b>0.1443</b>	<b>0.4724</b>	<b>0.7117</b>	<b>0.8322</b>	<b>0.9228</b>	<b>0.9926</b>	<b>1.00</b>	<b>28493</b>	<b>0</b>	<b>1.00</b>
<b>phi[11]</b>	<b>0.7255</b>	<b>0.1626</b>	<b>0.3905</b>	<b>0.6101</b>	<b>0.7368</b>	<b>0.8561</b>	<b>0.9830</b>	<b>1.00</b>	<b>85965</b>	<b>0</b>	<b>1.00</b>
<b>phi[12]</b>	<b>0.0883</b>	<b>0.0837</b>	<b>0.0025</b>	<b>0.0280</b>	<b>0.0642</b>	<b>0.1227</b>	<b>0.3106</b>	<b>1.00</b>	<b>102345</b>	<b>0</b>	<b>1.00</b>
b[1]	0.0592	0.0298	0.0159	0.0373	0.0544	0.0759	0.1302	1.00	75030	0	1.00
b[2]	0.1122	0.0481	0.0268	0.0785	0.1088	0.1421	0.2161	1.00	7070	0	1.00
b[3]	0.0353	0.0278	0.0013	0.0135	0.0291	0.0507	0.1032	1.00	94653	0	1.00
b[4]	0.0570	0.0338	0.0043	0.0311	0.0536	0.0787	0.1318	1.00	6490	0	1.00
b[5]	0.0501	0.0360	0.0025	0.0222	0.0433	0.0705	0.1365	1.00	4225	0	1.00
b[6]	0.0085	0.0085	0.0002	0.0024	0.0059	0.0118	0.0314	1.00	17352	0	1.00
b[7]	0.0246	0.0210	0.0009	0.0091	0.0194	0.0343	0.0781	1.00	10304	0	1.00
b[8]	0.4953	0.0773	0.3410	0.4434	0.4963	0.5485	0.6431	1.00	2013	0	1.00
b[9]	0.0329	0.0303	0.0009	0.0103	0.0242	0.0465	0.1117	1.00	21063	0	1.00
b[10]	0.0409	0.0351	0.0013	0.0138	0.0316	0.0588	0.1305	1.00	17531	0	1.00
b[11]	0.0561	0.0416	0.0023	0.0230	0.0481	0.0806	0.1545	1.00	15224	0	1.00
b[12]	0.0280	0.0231	0.0011	0.0106	0.0222	0.0392	0.0870	1.00	221606	0	1.00
B[1]	28	15	8	17	25	36	66	1.00	1318	0	1.00
B[2]	53	23	14	38	51	66	105	1.00	2489	0	1.00
B[3]	16	14	0	6	13	23	50	1.00	3538	1	1.00
B[4]	27	16	2	15	25	36	62	1.00	5772	0	1.00
B[5]	23	18	1	10	20	33	67	1.00	11597	0	1.00
B[6]	3	4	0	1	2	4	13	1.00	16469	1	1.00
B[7]	11	9	0	4	9	15	34	1.00	56817	1	1.00
B[8]	245	68	140	196	235	286	403	1.03	190	0	1.00
B[9]	15	15	0	4	11	22	54	1.00	5845	1	1.00
B[10]	19	18	0	6	14	27	65	1.00	2232	1	1.00
B[11]	27	21	1	10	22	38	80	1.00	1344	0	1.00
B[12]	13	12	0	4	10	18	45	1.00	3407	1	1.00
N[1]	28	15	8	17	25	36	66	1.00	1318	0	1.00
N[2]	76	22	45	61	73	88	128	1.01	810	0	1.00
N[3]	59	20	28	45	57	71	106	1.01	615	0	1.00
N[4]	75	19	46	62	73	86	119	1.01	585	0	1.00
N[5]	83	23	48	67	80	96	138	1.01	869	0	1.00
N[6]	42	13	24	33	40	49	73	1.01	968	0	1.00
N[7]	49	14	30	39	46	56	84	1.00	1356	0	1.00
N[8]	285	71	175	233	274	328	448	1.03	179	0	1.00
N[9]	169	50	93	133	162	196	288	1.01	436	0	1.00
N[10]	153	40	91	125	148	175	245	1.02	367	0	1.00
N[11]	138	39	77	110	132	160	227	1.01	351	0	1.00
N[12]	24	16	4	13	21	32	65	1.00	1486	0	1.00
gamma[1]	0.0137	0.0075	0.0034	0.0082	0.0122	0.0176	0.0321	1.00	1490	0	1.00
gamma[2]	0.0257	0.0113	0.0065	0.0179	0.0245	0.0321	0.0515	1.00	2601	0	1.00
gamma[3]	0.0084	0.0069	0.0003	0.0031	0.0068	0.0120	0.0255	1.00	3534	0	1.00
gamma[4]	0.0135	0.0082	0.0010	0.0074	0.0126	0.0184	0.0321	1.00	5271	0	1.00
gamma[5]	0.0122	0.0091	0.0006	0.0053	0.0103	0.0169	0.0345	1.00	10982	0	1.00
gamma[6]	0.0021	0.0021	0.0001	0.0006	0.0014	0.0029	0.0077	1.00	14576	0	1.00
gamma[7]	0.0059	0.0050	0.0002	0.0023	0.0047	0.0082	0.0185	1.00	40532	0	1.00
gamma[8]	0.1252	0.0361	0.0692	0.0989	0.1198	0.1467	0.2088	1.03	186	0	1.00
gamma[9]	0.0094	0.0089	0.0003	0.0029	0.0068	0.0132	0.0330	1.00	3340	0	1.00
gamma[10]	0.0119	0.0109	0.0004	0.0038	0.0089	0.0168	0.0402	1.00	1389	0	1.00
gamma[11]	0.0166	0.0134	0.0007	0.0064	0.0136	0.0233	0.0502	1.01	876	0	1.00
gamma[12]	0.0086	0.0080	0.0003	0.0030	0.0064	0.0117	0.0294	1.01	1663	0	1.00
deviance	936	72	801	885	934	983	1083	1.02	320	0	1.00
<b>FOREST</b>											
<b>par</b>	<b>mean</b>	<b>sd</b>	<b>2.50%</b>	<b>25%</b>	<b>50%</b>	<b>75%</b>	<b>97.50%</b>	<b>Rhat</b>	<b>n.eff</b>	<b>over0</b>	<b>f</b>
psi	0.1598	0.0560	0.0803	0.1195	0.1490	0.1884	0.2999	1.01	872	0	1.00
<b>Nforest</b>	<b>319</b>	<b>115</b>	<b>157</b>	<b>236</b>	<b>296</b>	<b>377</b>	<b>607</b>	<b>1.01</b>	<b>861</b>	<b>0</b>	<b>1.00</b>
p[1]	0.0935	0.0443	0.0333	0.0617	0.0849	0.1157	0.2031	1.00	3562	0	1.00
p[2]	0.1451	0.0618	0.0569	0.1005	0.1342	0.1779	0.2952	1.00	3389	0	1.00
p[3]	0.0986	0.0462	0.0355	0.0655	0.0897	0.1219	0.2128	1.00	3470	0	1.00
p[4]	0.1632	0.0687	0.0646	0.1135	0.1512	0.2000	0.3298	1.00	3561	0	1.00
p[5]	0.0851	0.0373	0.0333	0.0582	0.0783	0.1043	0.1770	1.00	6123	0	1.00
p[6]	0.2092	0.0887	0.0813	0.1448	0.1940	0.2574	0.4238	1.00	4214	0	1.00
p[7]	0.1140	0.0514	0.0426	0.0770	0.1044	0.1405	0.2401	1.00	3327	0	1.00

p[8]	0.1290	0.0813	0.0336	0.0727	0.1087	0.1624	0.3432	1.00	14223	0	1.00
p[9]	0.0867	0.0383	0.0337	0.0591	0.0796	0.1062	0.1814	1.00	6535	0	1.00
p[10]	0.2168	0.0923	0.0835	0.1495	0.2009	0.2671	0.4397	1.00	4334	0	1.00
p[11]	0.0848	0.0402	0.0303	0.0560	0.0770	0.1049	0.1847	1.00	3828	0	1.00
p[12]	0.2100	0.0891	0.0815	0.1453	0.1947	0.2584	0.4255	1.00	4226	0	1.00
p[13]	0.0823	0.0359	0.0322	0.0564	0.0760	0.1010	0.1705	1.00	5376	0	1.00
alpha.p	-2.3756	0.5098	-3.3668	-2.7213	-2.3778	-2.0341	-1.3668	1.00	3519	0	1.00
beta1.p	-0.5045	0.2358	-0.9915	-0.6555	-0.4972	-0.3449	-0.0606	1.00	27479	0	0.99
beta2.p	0.3930	0.2708	-0.1225	0.2106	0.3869	0.5688	0.9441	1.00	28023	1	0.93
phi[1]	<b>0.5002</b>	<b>0.2887</b>	<b>0.0251</b>	<b>0.2501</b>	<b>0.5001</b>	<b>0.7505</b>	<b>0.9750</b>	<b>1.00</b>	<b>211101</b>	<b>0</b>	<b>1.00</b>
phi[2]	<b>0.4982</b>	<b>0.2342</b>	<b>0.1163</b>	<b>0.3104</b>	<b>0.4754</b>	<b>0.6745</b>	<b>0.9546</b>	<b>1.00</b>	<b>6470</b>	<b>0</b>	<b>1.00</b>
phi[3]	<b>0.5012</b>	<b>0.2359</b>	<b>0.1112</b>	<b>0.3122</b>	<b>0.4807</b>	<b>0.6802</b>	<b>0.9562</b>	<b>1.00</b>	<b>23463</b>	<b>0</b>	<b>1.00</b>
phi[4]	<b>0.6593</b>	<b>0.2226</b>	<b>0.2012</b>	<b>0.4960</b>	<b>0.6851</b>	<b>0.8471</b>	<b>0.9848</b>	<b>1.00</b>	<b>62273</b>	<b>0</b>	<b>1.00</b>
phi[5]	<b>0.6936</b>	<b>0.2127</b>	<b>0.2310</b>	<b>0.5464</b>	<b>0.7278</b>	<b>0.8714</b>	<b>0.9875</b>	<b>1.00</b>	<b>102014</b>	<b>0</b>	<b>1.00</b>
phi[6]	<b>0.7073</b>	<b>0.1933</b>	<b>0.2967</b>	<b>0.5704</b>	<b>0.7317</b>	<b>0.8681</b>	<b>0.9867</b>	<b>1.00</b>	<b>18877</b>	<b>0</b>	<b>1.00</b>
phi[7]	<b>0.7694</b>	<b>0.1674</b>	<b>0.3882</b>	<b>0.6606</b>	<b>0.8006</b>	<b>0.9069</b>	<b>0.9912</b>	<b>1.00</b>	<b>21578</b>	<b>0</b>	<b>1.00</b>
phi[8]	<b>0.5914</b>	<b>0.2272</b>	<b>0.1795</b>	<b>0.4108</b>	<b>0.5914</b>	<b>0.7791</b>	<b>0.9756</b>	<b>1.00</b>	<b>44190</b>	<b>0</b>	<b>1.00</b>
phi[9]	<b>0.6314</b>	<b>0.2103</b>	<b>0.2294</b>	<b>0.4706</b>	<b>0.6384</b>	<b>0.8032</b>	<b>0.9777</b>	<b>1.00</b>	<b>53904</b>	<b>0</b>	<b>1.00</b>
phi[10]	<b>0.4959</b>	<b>0.2154</b>	<b>0.1496</b>	<b>0.3270</b>	<b>0.4699</b>	<b>0.6490</b>	<b>0.9411</b>	<b>1.00</b>	<b>30028</b>	<b>0</b>	<b>1.00</b>
phi[11]	<b>0.7846</b>	<b>0.1593</b>	<b>0.4158</b>	<b>0.6834</b>	<b>0.8163</b>	<b>0.9151</b>	<b>0.9919</b>	<b>1.00</b>	<b>227946</b>	<b>0</b>	<b>1.00</b>
phi[12]	<b>0.7438</b>	<b>0.1811</b>	<b>0.3391</b>	<b>0.6226</b>	<b>0.7739</b>	<b>0.8942</b>	<b>0.9896</b>	<b>1.00</b>	<b>14334</b>	<b>0</b>	<b>1.00</b>
b[1]	0.1702	0.0768	0.0567	0.1139	0.1576	0.2134	0.3549	1.00	3161	0	1.00
b[2]	0.0361	0.0325	0.0011	0.0120	0.0273	0.0507	0.1206	1.00	112732	0	1.00
b[3]	0.0262	0.0263	0.0007	0.0076	0.0181	0.0361	0.0971	1.00	151614	0	1.00
b[4]	0.0417	0.0350	0.0016	0.0154	0.0330	0.0586	0.1312	1.00	80103	0	1.00
b[5]	0.1499	0.0753	0.0233	0.0952	0.1432	0.1967	0.3164	1.00	6042	0	1.00
b[6]	0.0784	0.0614	0.0029	0.0303	0.0648	0.1125	0.2293	1.00	13488	0	1.00
b[7]	0.1350	0.0818	0.0102	0.0720	0.1255	0.1868	0.3182	1.00	10811	0	1.00
b[8]	0.1132	0.0993	0.0035	0.0379	0.0857	0.1608	0.3735	1.00	12697	0	1.00
b[9]	0.0851	0.0624	0.0038	0.0364	0.0733	0.1206	0.2354	1.00	13651	0	1.00
b[10]	0.0430	0.0359	0.0014	0.0154	0.0342	0.0615	0.1333	1.00	24019	0	1.00
b[11]	0.0429	0.0378	0.0013	0.0142	0.0325	0.0614	0.1398	1.00	98142	0	1.00
b[12]	0.0782	0.0550	0.0041	0.0358	0.0683	0.1093	0.2103	1.00	24966	0	1.00
B[1]	57	37	15	31	47	72	161	1.01	1201	0	1.00
B[2]	11	13	0	3	7	15	44	1.01	4348	1	1.00
B[3]	8	10	0	2	5	10	34	1.00	5036	1	1.00
B[4]	13	13	0	4	9	17	47	1.00	6129	1	1.00
B[5]	48	31	7	27	42	63	127	1.00	1758	0	1.00
B[6]	24	22	0	9	19	34	81	1.00	3471	1	1.00
B[7]	44	32	3	21	37	58	126	1.00	2381	0	1.00
B[8]	36	37	1	11	25	48	138	1.00	6821	0	1.00
B[9]	27	23	1	10	21	37	86	1.00	7312	0	1.00
B[10]	13	12	0	4	10	18	45	1.00	12937	1	1.00
B[11]	13	14	0	4	9	18	50	1.00	5862	1	1.00
B[12]	25	21	1	10	20	33	80	1.00	5824	0	1.00
N[1]	<b>57</b>	<b>37</b>	<b>15</b>	<b>31</b>	<b>47</b>	<b>72</b>	<b>161</b>	<b>1.01</b>	<b>1201</b>	<b>0</b>	<b>1.00</b>
N[2]	<b>36</b>	<b>22</b>	<b>11</b>	<b>21</b>	<b>31</b>	<b>45</b>	<b>94</b>	<b>1.01</b>	<b>1673</b>	<b>0</b>	<b>1.00</b>
N[3]	<b>25</b>	<b>17</b>	<b>5</b>	<b>13</b>	<b>20</b>	<b>32</b>	<b>70</b>	<b>1.01</b>	<b>2286</b>	<b>0</b>	<b>1.00</b>
N[4]	<b>29</b>	<b>18</b>	<b>8</b>	<b>16</b>	<b>24</b>	<b>36</b>	<b>75</b>	<b>1.00</b>	<b>2569</b>	<b>0</b>	<b>1.00</b>
N[5]	<b>69</b>	<b>34</b>	<b>24</b>	<b>44</b>	<b>61</b>	<b>85</b>	<b>157</b>	<b>1.01</b>	<b>1326</b>	<b>0</b>	<b>1.00</b>
N[6]	<b>72</b>	<b>33</b>	<b>29</b>	<b>49</b>	<b>65</b>	<b>88</b>	<b>155</b>	<b>1.01</b>	<b>1408</b>	<b>0</b>	<b>1.00</b>
N[7]	<b>100</b>	<b>44</b>	<b>41</b>	<b>68</b>	<b>90</b>	<b>121</b>	<b>211</b>	<b>1.01</b>	<b>1324</b>	<b>0</b>	<b>1.00</b>
N[8]	<b>93</b>	<b>51</b>	<b>28</b>	<b>57</b>	<b>82</b>	<b>116</b>	<b>223</b>	<b>1.00</b>	<b>2418</b>	<b>0</b>	<b>1.00</b>
N[9]	<b>84</b>	<b>41</b>	<b>31</b>	<b>55</b>	<b>75</b>	<b>103</b>	<b>189</b>	<b>1.00</b>	<b>2715</b>	<b>0</b>	<b>1.00</b>
N[10]	<b>51</b>	<b>24</b>	<b>20</b>	<b>34</b>	<b>46</b>	<b>62</b>	<b>114</b>	<b>1.00</b>	<b>2367</b>	<b>0</b>	<b>1.00</b>
N[11]	<b>54</b>	<b>27</b>	<b>20</b>	<b>35</b>	<b>48</b>	<b>66</b>	<b>121</b>	<b>1.00</b>	<b>2006</b>	<b>0</b>	<b>1.00</b>
N[12]	<b>65</b>	<b>32</b>	<b>25</b>	<b>43</b>	<b>58</b>	<b>79</b>	<b>146</b>	<b>1.00</b>	<b>2137</b>	<b>0</b>	<b>1.00</b>
gamma[1]	0.0281	0.0185	0.0068	0.0152	0.0231	0.0356	0.0790	1.01	1229	0	1.00
gamma[2]	0.0061	0.0066	0.0002	0.0018	0.0041	0.0081	0.0232	1.01	4147	0	1.00
gamma[3]	0.0044	0.0052	0.0001	0.0011	0.0028	0.0058	0.0187	1.00	4945	0	1.00
gamma[4]	0.0070	0.0068	0.0002	0.0023	0.0050	0.0094	0.0250	1.00	5442	0	1.00
gamma[5]	0.0251	0.0163	0.0035	0.0137	0.0217	0.0327	0.0669	1.00	1686	0	1.00
gamma[6]	0.0133	0.0120	0.0005	0.0047	0.0102	0.0183	0.0445	1.00	3100	0	1.00
gamma[7]	0.0236	0.0181	0.0017	0.0110	0.0196	0.0314	0.0699	1.01	1875	0	1.00
gamma[8]	0.0199	0.0202	0.0006	0.0063	0.0138	0.0264	0.0758	1.00	4923	0	1.00
gamma[9]	0.0154	0.0134	0.0006	0.0060	0.0121	0.0209	0.0504	1.00	4545	0	1.00
gamma[10]	0.0078	0.0075	0.0003	0.0026	0.0058	0.0106	0.0274	1.00	8518	0	1.00
gamma[11]	0.0082	0.0087	0.0002	0.0024	0.0056	0.0110	0.0312	1.00	3462	0	1.00
gamma[12]	0.0148	0.0132	0.0007	0.0060	0.0115	0.0196	0.0492	1.00	3320	0	1.00

deviance	492	65	367	447	490	535	623	1.00	2299	0	1.00
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<b>K) <i>Myioborus miniatus</i> - Jolly-Seber Model</b>											
<b>COFFEE</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
psi	0.1175	0.0353	0.0695	0.0931	0.1105	0.1344	0.2097	1.03	192	0	1.00
<b>Ncoffee</b>	<b>231</b>	<b>72</b>	<b>136</b>	<b>182</b>	<b>216</b>	<b>265</b>	<b>420</b>	<b>1.03</b>	<b>185</b>	<b>0</b>	<b>1.00</b>
p[1]	0.1936	0.0590	0.0954	0.1512	0.1876	0.2294	0.3256	1.01	465	0	1.00
p[2]	0.4357	0.1233	0.2184	0.3460	0.4281	0.5178	0.6940	1.00	2782	0	1.00
p[3]	0.1326	0.0440	0.0615	0.1009	0.1274	0.1586	0.2331	1.01	401	0	1.00
p[4]	0.3219	0.0861	0.1720	0.2603	0.3157	0.3768	0.5070	1.00	876	0	1.00
p[5]	0.0936	0.0404	0.0346	0.0644	0.0869	0.1154	0.1902	1.00	685	0	1.00
p[6]	0.2825	0.0763	0.1503	0.2281	0.2767	0.3310	0.4477	1.01	661	0	1.00
p[7]	0.5091	0.1497	0.2356	0.3992	0.5046	0.6153	0.8045	1.00	7143	0	1.00
p[8]	0.1262	0.0421	0.0583	0.0959	0.1212	0.1511	0.2223	1.01	395	0	1.00
p[9]	0.0917	0.0472	0.0279	0.0579	0.0824	0.1154	0.2082	1.00	1064	0	1.00
p[10]	0.1774	0.0557	0.0858	0.1373	0.1713	0.2108	0.3027	1.01	449	0	1.00
p[11]	0.2148	0.0631	0.1084	0.1696	0.2090	0.2536	0.3544	1.01	491	0	1.00
p[12]	0.1552	0.0504	0.0733	0.1190	0.1495	0.1851	0.2699	1.01	426	0	1.00
p[13]	0.4562	0.1307	0.2241	0.3608	0.4489	0.5446	0.7266	1.00	3566	0	1.00
alpha.p	-1.4710	0.3851	-2.2500	-1.7251	-1.4655	-1.2118	-0.7281	1.01	387	0	1.00
beta1.p	-0.8889	0.2708	-1.4494	-1.0638	-0.8788	-0.7032	-0.3841	1.00	2480	0	1.00
beta2.p	0.1941	0.2400	-0.2766	0.0337	0.1929	0.3532	0.6700	1.00	5549	1	0.79
<b>phi[1]</b>	<b>0.5001</b>	<b>0.2879</b>	<b>0.0249</b>	<b>0.2510</b>	<b>0.5002</b>	<b>0.7486</b>	<b>0.9745</b>	<b>1.00</b>	<b>3.00E+05</b>	<b>0</b>	<b>1.00</b>
<b>phi[2]</b>	<b>0.7879</b>	<b>0.1683</b>	<b>0.3794</b>	<b>0.6885</b>	<b>0.8277</b>	<b>0.9238</b>	<b>0.9931</b>	<b>1.00</b>	<b>9981</b>	<b>0</b>	<b>1.00</b>
<b>phi[3]</b>	<b>0.7429</b>	<b>0.1680</b>	<b>0.3800</b>	<b>0.6271</b>	<b>0.7636</b>	<b>0.8808</b>	<b>0.9874</b>	<b>1.00</b>	<b>8410</b>	<b>0</b>	<b>1.00</b>
<b>phi[4]</b>	<b>0.8227</b>	<b>0.1351</b>	<b>0.5036</b>	<b>0.7401</b>	<b>0.8515</b>	<b>0.9322</b>	<b>0.9936</b>	<b>1.00</b>	<b>76966</b>	<b>0</b>	<b>1.00</b>
<b>phi[5]</b>	<b>0.7216</b>	<b>0.1823</b>	<b>0.3398</b>	<b>0.5906</b>	<b>0.7428</b>	<b>0.8741</b>	<b>0.9873</b>	<b>1.00</b>	<b>57155</b>	<b>0</b>	<b>1.00</b>
<b>phi[6]</b>	<b>0.4287</b>	<b>0.1801</b>	<b>0.1546</b>	<b>0.2941</b>	<b>0.4004</b>	<b>0.5361</b>	<b>0.8544</b>	<b>1.00</b>	<b>50671</b>	<b>0</b>	<b>1.00</b>
<b>phi[7]</b>	<b>0.8186</b>	<b>0.1447</b>	<b>0.4669</b>	<b>0.7338</b>	<b>0.8525</b>	<b>0.9354</b>	<b>0.9940</b>	<b>1.00</b>	<b>32881</b>	<b>0</b>	<b>1.00</b>
<b>phi[8]</b>	<b>0.8455</b>	<b>0.1287</b>	<b>0.5233</b>	<b>0.7751</b>	<b>0.8784</b>	<b>0.9473</b>	<b>0.9951</b>	<b>1.00</b>	<b>35792</b>	<b>0</b>	<b>1.00</b>
<b>phi[9]</b>	<b>0.4221</b>	<b>0.1823</b>	<b>0.1509</b>	<b>0.2870</b>	<b>0.3901</b>	<b>0.5267</b>	<b>0.8636</b>	<b>1.00</b>	<b>9288</b>	<b>0</b>	<b>1.00</b>
<b>phi[10]</b>	<b>0.7067</b>	<b>0.1981</b>	<b>0.2882</b>	<b>0.5647</b>	<b>0.7341</b>	<b>0.8736</b>	<b>0.9878</b>	<b>1.00</b>	<b>26583</b>	<b>0</b>	<b>1.00</b>
<b>phi[11]</b>	<b>0.3972</b>	<b>0.2001</b>	<b>0.0929</b>	<b>0.2451</b>	<b>0.3666</b>	<b>0.5204</b>	<b>0.8640</b>	<b>1.00</b>	<b>10821</b>	<b>0</b>	<b>1.00</b>
<b>phi[12]</b>	<b>0.3976</b>	<b>0.2409</b>	<b>0.0506</b>	<b>0.2025</b>	<b>0.3560</b>	<b>0.5622</b>	<b>0.9242</b>	<b>1.00</b>	<b>106636</b>	<b>0</b>	<b>1.00</b>
b[1]	0.0839	0.0401	0.0234	0.0543	0.0780	0.1072	0.1776	1.00	9107	0	1.00
b[2]	0.0724	0.0480	0.0047	0.0359	0.0648	0.1002	0.1856	1.00	1667	0	1.00
b[3]	0.0611	0.0413	0.0034	0.0290	0.0546	0.0857	0.1577	1.00	14017	0	1.00
b[4]	0.0511	0.0377	0.0022	0.0215	0.0438	0.0727	0.1418	1.00	4307	0	1.00
b[5]	0.0357	0.0336	0.0010	0.0112	0.0260	0.0498	0.1256	1.00	9692	0	1.00
b[6]	0.0139	0.0140	0.0003	0.0040	0.0096	0.0192	0.0517	1.00	12410	0	1.00
b[7]	0.0399	0.0329	0.0020	0.0166	0.0321	0.0542	0.1242	1.00	2028	0	1.00
b[8]	0.4647	0.1010	0.2639	0.3967	0.4661	0.5347	0.6570	1.00	748	0	1.00
b[9]	0.0540	0.0493	0.0016	0.0174	0.0403	0.0762	0.1820	1.00	30951	0	1.00
b[10]	0.0495	0.0429	0.0015	0.0168	0.0383	0.0708	0.1583	1.00	14740	0	1.00
b[11]	0.0449	0.0354	0.0018	0.0177	0.0371	0.0637	0.1327	1.00	18032	0	1.00
b[12]	0.0288	0.0282	0.0007	0.0084	0.0202	0.0401	0.1044	1.00	103022	0	1.00
B[1]	19	10	6	12	17	24	44	1.01	823	0	1.00
B[2]	16	10	1	8	14	21	39	1.00	47620	0	1.00
B[3]	14	10	0	6	12	19	37	1.00	2005	1	1.00
B[4]	11	8	0	5	9	15	30	1.00	5430	1	1.00
B[5]	8	8	0	2	5	10	29	1.00	2052	1	1.00
B[6]	2	3	0	0	1	3	10	1.00	8714	1	1.00
B[7]	8	6	0	4	7	11	24	1.00	91344	1	1.00
B[8]	115	51	48	80	104	138	248	1.03	203	0	1.00
B[9]	12	13	0	3	8	17	46	1.01	1561	1	1.00
B[10]	11	11	0	3	8	15	40	1.00	2982	1	1.00
B[11]	10	9	0	4	8	14	33	1.00	2365	1	1.00
B[12]	6	8	0	1	4	8	27	1.00	2701	1	1.00
<b>N[1]</b>	<b>19</b>	<b>10</b>	<b>6</b>	<b>12</b>	<b>17</b>	<b>24</b>	<b>44</b>	<b>1.01</b>	<b>823</b>	<b>0</b>	<b>1.00</b>
<b>N[2]</b>	<b>31</b>	<b>10</b>	<b>17</b>	<b>24</b>	<b>29</b>	<b>36</b>	<b>56</b>	<b>1.00</b>	<b>1782</b>	<b>0</b>	<b>1.00</b>
<b>N[3]</b>	<b>37</b>	<b>12</b>	<b>19</b>	<b>28</b>	<b>35</b>	<b>43</b>	<b>66</b>	<b>1.01</b>	<b>734</b>	<b>0</b>	<b>1.00</b>
<b>N[4]</b>	<b>42</b>	<b>12</b>	<b>25</b>	<b>34</b>	<b>40</b>	<b>47</b>	<b>70</b>	<b>1.01</b>	<b>609</b>	<b>0</b>	<b>1.00</b>
<b>N[5]</b>	<b>38</b>	<b>14</b>	<b>17</b>	<b>28</b>	<b>35</b>	<b>45</b>	<b>72</b>	<b>1.01</b>	<b>692</b>	<b>0</b>	<b>1.00</b>
<b>N[6]</b>	<b>17</b>	<b>7</b>	<b>8</b>	<b>12</b>	<b>16</b>	<b>21</b>	<b>35</b>	<b>1.00</b>	<b>1113</b>	<b>0</b>	<b>1.00</b>
<b>N[7]</b>	<b>23</b>	<b>8</b>	<b>13</b>	<b>17</b>	<b>20</b>	<b>26</b>	<b>44</b>	<b>1.00</b>	<b>3639</b>	<b>0</b>	<b>1.00</b>
<b>N[8]</b>	<b>135</b>	<b>52</b>	<b>67</b>	<b>99</b>	<b>123</b>	<b>158</b>	<b>270</b>	<b>1.03</b>	<b>195</b>	<b>0</b>	<b>1.00</b>
<b>N[9]</b>	<b>67</b>	<b>34</b>	<b>25</b>	<b>44</b>	<b>59</b>	<b>82</b>	<b>153</b>	<b>1.01</b>	<b>476</b>	<b>0</b>	<b>1.00</b>
<b>N[10]</b>	<b>56</b>	<b>23</b>	<b>26</b>	<b>40</b>	<b>51</b>	<b>66</b>	<b>113</b>	<b>1.01</b>	<b>415</b>	<b>0</b>	<b>1.00</b>

N[11]	31	15	12	21	28	38	68	1.01	646	0	1.00
N[12]	18	12	3	9	15	23	49	1.01	1038	0	1.00
gamma[1]	0.0097	0.0053	0.0025	0.0059	0.0088	0.0124	0.0226	1.01	1023	0	1.00
gamma[2]	0.0081	0.0052	0.0006	0.0042	0.0073	0.0110	0.0205	1.00	85508	0	1.00
gamma[3]	0.0072	0.0052	0.0004	0.0033	0.0062	0.0099	0.0198	1.00	2339	0	1.00
gamma[4]	0.0059	0.0044	0.0003	0.0025	0.0050	0.0082	0.0165	1.00	5716	0	1.00
gamma[5]	0.0043	0.0043	0.0001	0.0013	0.0030	0.0058	0.0158	1.00	2185	0	1.00
gamma[6]	0.0016	0.0017	0.0000	0.0005	0.0011	0.0023	0.0061	1.00	11223	0	1.00
gamma[7]	0.0045	0.0036	0.0003	0.0020	0.0037	0.0060	0.0136	1.00	160940	0	1.00
gamma[8]	0.0585	0.0267	0.0232	0.0402	0.0529	0.0706	0.1272	1.03	205	0	1.00
gamma[9]	0.0071	0.0075	0.0002	0.0021	0.0049	0.0096	0.0265	1.01	1310	0	1.00
gamma[10]	0.0066	0.0065	0.0002	0.0020	0.0047	0.0090	0.0233	1.01	2224	0	1.00
gamma[11]	0.0059	0.0053	0.0002	0.0022	0.0046	0.0081	0.0196	1.00	1867	0	1.00
gamma[12]	0.0040	0.0046	0.0001	0.0010	0.0025	0.0053	0.0162	1.00	2114	0	1.00
deviance	432	50	343	398	429	464	539	1.01	355	0	1.00
<b>FOREST</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
psi	0.1176	0.0532	0.0504	0.0800	0.1043	0.1424	0.2597	1.07	57	0	1.00
<b>Nforest</b>	<b>227</b>	<b>107</b>	<b>93</b>	<b>152</b>	<b>200</b>	<b>278</b>	<b>516</b>	<b>1.07</b>	<b>56</b>	<b>0</b>	<b>1.00</b>
p[1]	0.0833	0.0441	0.0248	0.0515	0.0745	0.1053	0.1934	1.06	67	0	1.00
p[2]	0.1040	0.0498	0.0339	0.0675	0.0952	0.1309	0.2253	1.06	63	0	1.00
p[3]	0.0852	0.0443	0.0257	0.0531	0.0765	0.1076	0.1956	1.06	65	0	1.00
p[4]	0.1117	0.0540	0.0360	0.0720	0.1021	0.1409	0.2430	1.05	69	0	1.00
p[5]	0.0861	0.0424	0.0272	0.0550	0.0783	0.1087	0.1897	1.05	74	0	1.00
p[6]	0.1320	0.0692	0.0385	0.0812	0.1183	0.1680	0.3029	1.04	95	0	1.00
p[7]	0.0912	0.0454	0.0288	0.0582	0.0828	0.1148	0.2030	1.06	62	0	1.00
p[8]	0.1202	0.0752	0.0290	0.0668	0.1021	0.1536	0.3150	1.02	214	0	1.00
p[9]	0.0874	0.0431	0.0276	0.0559	0.0795	0.1104	0.1923	1.05	76	0	1.00
p[10]	0.1354	0.0721	0.0386	0.0825	0.1208	0.1727	0.3144	1.04	101	0	1.00
p[11]	0.0805	0.0433	0.0235	0.0492	0.0717	0.1019	0.1889	1.05	70	0	1.00
p[12]	0.1323	0.0695	0.0385	0.0813	0.1186	0.1685	0.3042	1.04	96	0	1.00
p[13]	0.0836	0.0416	0.0261	0.0531	0.0758	0.1055	0.1853	1.05	71	0	1.00
alpha.p	-2.5302	0.5726	-3.6730	-2.9132	-2.5194	-2.1397	-1.4280	1.05	66	0	1.00
beta1.p	-0.2243	0.2590	-0.7470	-0.3947	-0.2196	-0.0488	0.2690	1.00	165550	1	0.81
beta2.p	0.2527	0.2940	-0.3076	0.0533	0.2466	0.4450	0.8482	1.00	2393	1	0.81
phi[1]	<b>0.4996</b>	<b>0.2887</b>	<b>0.0248</b>	<b>0.2490</b>	<b>0.5001</b>	<b>0.7493</b>	<b>0.9747</b>	<b>1.00</b>	<b>1.38E+05</b>	<b>0</b>	<b>1.00</b>
phi[2]	<b>0.4339</b>	<b>0.2808</b>	<b>0.0187</b>	<b>0.1908</b>	<b>0.4015</b>	<b>0.6605</b>	<b>0.9610</b>	<b>1.00</b>	<b>4423</b>	<b>0</b>	<b>1.00</b>
phi[3]	<b>0.3745</b>	<b>0.2688</b>	<b>0.0136</b>	<b>0.1455</b>	<b>0.3226</b>	<b>0.5700</b>	<b>0.9401</b>	<b>1.00</b>	<b>72975</b>	<b>0</b>	<b>1.00</b>
phi[4]	<b>0.4206</b>	<b>0.2783</b>	<b>0.0168</b>	<b>0.1789</b>	<b>0.3859</b>	<b>0.6410</b>	<b>0.9564</b>	<b>1.00</b>	<b>43457</b>	<b>0</b>	<b>1.00</b>
phi[5]	<b>0.7591</b>	<b>0.1726</b>	<b>0.3721</b>	<b>0.6445</b>	<b>0.7900</b>	<b>0.9019</b>	<b>0.9906</b>	<b>1.00</b>	<b>5604</b>	<b>0</b>	<b>1.00</b>
phi[6]	<b>0.8286</b>	<b>0.1386</b>	<b>0.4887</b>	<b>0.7496</b>	<b>0.8623</b>	<b>0.9396</b>	<b>0.9944</b>	<b>1.00</b>	<b>16435</b>	<b>0</b>	<b>1.00</b>
phi[7]	<b>0.8867</b>	<b>0.1007</b>	<b>0.6258</b>	<b>0.8375</b>	<b>0.9150</b>	<b>0.9637</b>	<b>0.9967</b>	<b>1.00</b>	<b>186968</b>	<b>0</b>	<b>1.00</b>
phi[8]	<b>0.6558</b>	<b>0.2158</b>	<b>0.2309</b>	<b>0.4909</b>	<b>0.6732</b>	<b>0.8387</b>	<b>0.9840</b>	<b>1.00</b>	<b>2871</b>	<b>0</b>	<b>1.00</b>
phi[9]	<b>0.4662</b>	<b>0.2095</b>	<b>0.1429</b>	<b>0.3032</b>	<b>0.4352</b>	<b>0.6054</b>	<b>0.9255</b>	<b>1.00</b>	<b>7466</b>	<b>0</b>	<b>1.00</b>
phi[10]	<b>0.7285</b>	<b>0.1914</b>	<b>0.3073</b>	<b>0.5974</b>	<b>0.7606</b>	<b>0.8886</b>	<b>0.9895</b>	<b>1.00</b>	<b>12940</b>	<b>0</b>	<b>1.00</b>
phi[11]	<b>0.6943</b>	<b>0.2021</b>	<b>0.2684</b>	<b>0.5500</b>	<b>0.7193</b>	<b>0.8637</b>	<b>0.9868</b>	<b>1.00</b>	<b>49999</b>	<b>0</b>	<b>1.00</b>
phi[12]	<b>0.6878</b>	<b>0.2113</b>	<b>0.2397</b>	<b>0.5371</b>	<b>0.7174</b>	<b>0.8655</b>	<b>0.9867</b>	<b>1.00</b>	<b>3196</b>	<b>0</b>	<b>1.00</b>
b[1]	0.1086	0.0715	0.0139	0.0548	0.0938	0.1472	0.2842	1.00	850	0	1.00
b[2]	0.1017	0.0707	0.0063	0.0480	0.0886	0.1408	0.2735	1.00	4653	0	1.00
b[3]	0.0928	0.0731	0.0040	0.0368	0.0759	0.1313	0.2754	1.00	6877	0	1.00
b[4]	0.1498	0.0799	0.0180	0.0920	0.1420	0.1978	0.3290	1.00	2875	0	1.00
b[5]	0.0276	0.0284	0.0007	0.0077	0.0187	0.0380	0.1045	1.00	26591	0	1.00
b[6]	0.0404	0.0424	0.0009	0.0109	0.0269	0.0554	0.1565	1.00	51459	0	1.00
b[7]	0.2784	0.1152	0.0585	0.1974	0.2765	0.3573	0.5076	1.00	8889	0	1.00
b[8]	0.0667	0.0681	0.0016	0.0184	0.0450	0.0921	0.2529	1.00	3360	0	1.00
b[9]	0.0280	0.0279	0.0007	0.0081	0.0195	0.0387	0.1029	1.00	34510	0	1.00
b[10]	0.0344	0.0334	0.0009	0.0104	0.0245	0.0478	0.1232	1.00	7582	0	1.00
b[11]	0.0316	0.0321	0.0008	0.0089	0.0216	0.0436	0.1190	1.00	34905	0	1.00
b[12]	0.0401	0.0410	0.0010	0.0111	0.0271	0.0553	0.1515	1.00	26690	0	1.00
B[1]	26	24	2	10	19	33	90	1.05	119	0	1.00
B[2]	24	22	1	9	18	31	85	1.03	181	0	1.00
B[3]	22	23	0	7	15	29	83	1.02	222	1	1.00
B[4]	34	25	4	18	28	44	99	1.02	191	0	1.00
B[5]	6	8	0	1	3	7	26	1.01	384	1	1.00
B[6]	9	12	0	2	5	11	40	1.02	387	1	1.00
B[7]	65	40	12	37	56	82	168	1.04	107	0	1.00
B[8]	15	18	0	3	9	19	65	1.01	552	1	1.00
B[9]	6	8	0	1	3	7	26	1.02	333	1	1.00
B[10]	7	9	0	1	4	9	32	1.01	491	1	1.00

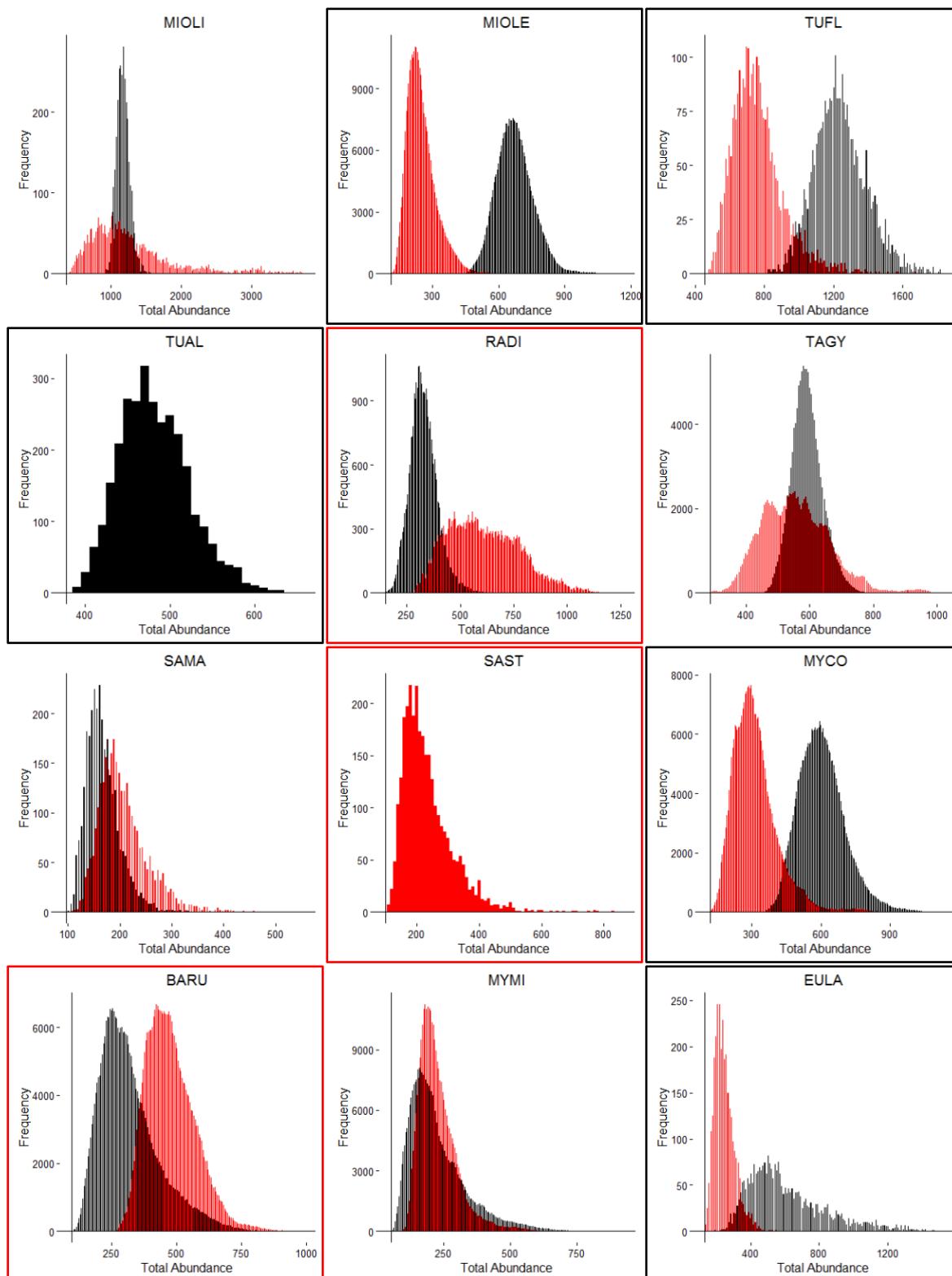
B[11]	7	9	0	1	4	9	32	1.01	485	1	1.00
B[12]	9	13	0	2	5	11	43	1.01	556	1	1.00
N[1]	26	24	2	10	19	33	90	1.05	119	0	1.00
N[2]	34	25	6	17	27	43	103	1.04	116	0	1.00
N[3]	33	28	4	15	26	43	108	1.03	138	0	1.00
N[4]	48	29	15	28	40	59	124	1.04	108	0	1.00
N[5]	41	24	13	24	35	51	106	1.04	111	0	1.00
N[6]	43	27	13	25	36	53	113	1.04	118	0	1.00
N[7]	104	49	43	69	91	126	233	1.05	69	0	1.00
N[8]	81	45	26	50	70	100	197	1.03	117	0	1.00
N[9]	41	26	12	23	34	50	109	1.03	125	0	1.00
N[10]	36	22	12	21	30	44	94	1.03	127	0	1.00
N[11]	31	21	9	18	26	39	87	1.03	141	0	1.00
N[12]	30	22	7	16	24	38	89	1.03	167	0	1.00
gamma[1]	0.0131	0.0120	0.0012	0.0052	0.0098	0.0170	0.0455	1.04	124	0	1.00
gamma[2]	0.0123	0.0113	0.0006	0.0047	0.0092	0.0163	0.0434	1.03	186	0	1.00
gamma[3]	0.0115	0.0119	0.0004	0.0037	0.0080	0.0152	0.0438	1.02	222	0	1.00
gamma[4]	0.0181	0.0133	0.0020	0.0091	0.0149	0.0233	0.0532	1.03	185	0	1.00
gamma[5]	0.0034	0.0042	0.0001	0.0008	0.0021	0.0045	0.0149	1.01	391	0	1.00
gamma[6]	0.0051	0.0066	0.0001	0.0012	0.0030	0.0065	0.0228	1.02	368	0	1.00
gamma[7]	0.0346	0.0221	0.0064	0.0195	0.0294	0.0441	0.0927	1.04	103	0	1.00
gamma[8]	0.0085	0.0104	0.0002	0.0021	0.0052	0.0110	0.0375	1.01	471	0	1.00
gamma[9]	0.0037	0.0048	0.0001	0.0009	0.0023	0.0048	0.0163	1.03	300	0	1.00
gamma[10]	0.0046	0.0057	0.0001	0.0012	0.0028	0.0059	0.0197	1.02	414	0	1.00
gamma[11]	0.0044	0.0058	0.0001	0.0010	0.0025	0.0055	0.0199	1.02	423	0	1.00
gamma[12]	0.0057	0.0079	0.0001	0.0012	0.0032	0.0071	0.0264	1.02	470	0	1.00
deviance	306	42	231	276	304	334	393	1.06	56	0	1.00

**L) Euphonia laniirostris - Closed Population Model****COFFEE**

par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
N	249	58	164	207	240	282	392	1.00	1022	NO	1.00
p.est[1]	0.0001	0.0002	<0.0001	<0.0001	0.0001	0.0002	0.0007	1.00	1023	NO	1.00
p.est[2]	0.0695	0.0177	0.0389	0.0566	0.0678	0.0805	0.1085	1.00	1118	NO	1.00
p.est[3]	0.0011	0.0009	0.0001	0.0005	0.0008	0.0014	0.0033	1.00	894	NO	1.00
p.est[4]	0.0950	0.0238	0.0549	0.0783	0.0931	0.1099	0.1483	1.00	2907	NO	1.00
p.est[5]	0.0023	0.0015	0.0005	0.0012	0.0020	0.0029	0.0061	1.00	1016	NO	1.00
p.est[6]	<0.0001	0.0001	<0.0001	<0.0001	<0.0001	<0.0001	0.0002	1.00	1	NO	1.00
p.est[7]	0.0789	0.0201	0.0443	0.0646	0.0771	0.0913	0.1231	1.00	1136	NO	1.00
p.est[8]	0.0764	0.0214	0.0415	0.0612	0.0735	0.0895	0.1247	1.00	1632	NO	1.00
p.est[9]	0.0245	0.0072	0.0129	0.0191	0.0237	0.0290	0.0406	1.00	1848	NO	1.00
p.est[10]	0.0100	0.0039	0.0042	0.0071	0.0096	0.0123	0.0191	1.00	1808	NO	1.00
p.est[11]	0.0417	0.0109	0.0234	0.0338	0.0406	0.0483	0.0652	1.00	1328	NO	1.00
p.est[12]	0.0001	0.0001	<0.0001	<0.0001	0.0001	0.0001	0.0005	1.00	1109	NO	1.00
mean.p	0.0165	0.0055	0.0080	0.0125	0.0159	0.0198	0.0286	1.00	2048	NO	1.00
beta1	3.3987	0.6257	2.2767	2.9606	3.3572	3.7926	4.7046	1.00	945	NO	1.00
beta2	-1.4503	0.3297	-2.1337	-1.6693	-1.4279	-1.2202	-0.8623	1.00	599	NO	1.00
psi	0.0613	0.0147	0.0389	0.0506	0.0591	0.0695	0.0962	1.01	949	NO	1.00
deviance	712	44	635	681	710	743	807	1.00	1324	NO	1.00

**FOREST**

par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
N	606	224	309	444	553	721	1176	1.06	52	NO	1.00
p.est[1]	0.0060	0.0024	0.0024	0.0042	0.0057	0.0074	0.0119	1.02	107	NO	1.00
p.est[2]	0.0162	0.0060	0.0069	0.0117	0.0154	0.0198	0.0302	1.03	83	NO	1.00
p.est[3]	0.0045	0.0019	0.0017	0.0031	0.0042	0.0056	0.0093	1.02	124	NO	1.00
p.est[4]	0.0303	0.0108	0.0131	0.0222	0.0291	0.0370	0.0552	1.03	74	NO	1.00
p.est[5]	0.0025	0.0012	0.0008	0.0016	0.0022	0.0031	0.0057	1.01	189	NO	1.00
p.est[6]	0.0111	0.0042	0.0047	0.0079	0.0106	0.0135	0.0207	1.02	87	NO	1.00
p.est[7]	0.0192	0.0084	0.0073	0.0130	0.0178	0.0239	0.0399	1.02	112	NO	1.00
p.est[8]	0.0298	0.0107	0.0129	0.0218	0.0286	0.0365	0.0544	1.03	75	NO	1.00
p.est[9]	0.0023	0.0011	0.0007	0.0014	0.0020	0.0028	0.0053	1.01	202	NO	1.00
p.est[10]	0.0245	0.0091	0.0106	0.0178	0.0234	0.0299	0.0451	1.03	81	NO	1.00
p.est[11]	0.0025	0.0012	0.0008	0.0016	0.0022	0.0030	0.0056	1.01	190	NO	1.00
p.est[12]	0.0311	0.0111	0.0135	0.0228	0.0299	0.0378	0.0562	1.03	75	NO	1.00
mean.p	0.0187	0.0069	0.0080	0.0134	0.0178	0.0228	0.0345	1.03	82	NO	1.00
beta1	1.2315	0.1907	0.8766	1.0990	1.2240	1.3562	1.6264	1.00	2046	NO	1.00
beta2	-0.7123	0.1733	-1.0638	-0.8252	-0.7131	-0.5898	-0.3894	1.00	3000	NO	1.00
psi	0.1484	0.0551	0.0749	0.1087	0.1353	0.1766	0.2881	1.06	52	NO	1.00
deviance	961	69	841	912	955	1007	1103	1.04	54	NO	1.00



**Figure S4.** Posterior distributions for the estimates of total abundance for the twelve focal species in coffee (red bars) and forest (black bars); less overlap means higher probability of differences between habitats. Estimation method and simulations conditions varied among species but not between habitats within a species. There was not enough data to make estimations for *T. albicollis* in coffee and *S. striatipectus* in forest.

### 2.3.3. Fidelity

**Table S11.** Numerical outputs for Cormack-Jolly-Seber models of species with low recapture probabilities (for species with high recapture probabilities see Table S10b, e, f, i, j and k) (**phi** is the parameter for apparent survival). Rhat is the potential scale reduction factor (at convergence, Rhat=1); successful convergence based on Rhat values (all < 1.1), n.eff is a crude measure of effective sample size, over0 checks if 0 falls in the parameter's 95% credible interval. f is the proportion of the posterior with the same sign as the mean; i.e., our confidence that the parameter is positive or negative.

A) <i>Mionectes olivaceus</i> - Cormack-Jolly-Seber											
COFFEE											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
phi.occ[1]	0.4936	0.2929	0.0262	0.2304	0.4882	0.7509	0.9753	1.00	3000	NO	1.00
phi.occ[2]	0.4151	0.2815	0.0180	0.1667	0.3820	0.6398	0.9544	1.00	1317	NO	1.00
phi.occ[3]	0.2298	0.2250	0.0053	0.0613	0.1507	0.3213	0.8587	1.01	286	NO	1.00
phi.occ[4]	0.4063	0.2539	0.0475	0.1955	0.3637	0.5896	0.9455	1.00	778	NO	1.00
phi.occ[5]	0.4676	0.2536	0.0795	0.2536	0.4445	0.6675	0.9549	1.01	145	NO	1.00
phi.occ[6]	0.6452	0.2329	0.1710	0.4627	0.6728	0.8426	0.9849	1.00	3000	NO	1.00
phi.occ[7]	0.6891	0.2204	0.2001	0.5348	0.7276	0.8751	0.9862	1.00	3000	NO	1.00
phi.occ[8]	0.5544	0.2559	0.0901	0.3471	0.5618	0.7677	0.9737	1.00	3000	NO	1.00
phi.occ[9]	0.4727	0.2632	0.0672	0.2538	0.4374	0.6877	0.9659	1.00	445	NO	1.00
phi.occ[10]	0.5684	0.2509	0.1091	0.3646	0.5763	0.7791	0.9785	1.00	545	NO	1.00
phi.occ[11]	0.3984	0.2849	0.0130	0.1408	0.3510	0.6255	0.9575	1.00	619	NO	1.00
p.est[1]	0.3281	0.2502	0.0286	0.1286	0.2544	0.4704	0.9182	1.00	453	NO	1.00
p.est[2]	0.0452	0.0336	0.0070	0.0217	0.0367	0.0589	0.1345	1.00	960	NO	1.00
p.est[3]	0.2192	0.1641	0.0376	0.1012	0.1707	0.2847	0.6745	1.00	454	NO	1.00
p.est[4]	0.0268	0.0404	0.0002	0.0043	0.0141	0.0338	0.1267	1.01	2524	NO	1.00
p.est[5]	0.1797	0.1321	0.0352	0.0880	0.1437	0.2287	0.5512	1.00	481	NO	1.00
p.est[6]	0.3890	0.2934	0.0198	0.1366	0.3138	0.6073	0.9749	1.00	497	NO	1.00
p.est[7]	0.0408	0.0311	0.0053	0.0188	0.0332	0.0534	0.1224	1.00	1063	NO	1.00
p.est[8]	0.0300	0.0561	0.0000	0.0025	0.0110	0.0341	0.1742	1.00	3000	NO	1.00
p.est[9]	0.0811	0.0592	0.0160	0.0420	0.0669	0.1038	0.2192	1.01	623	NO	1.00
p.est[10]	0.1145	0.0829	0.0245	0.0592	0.0931	0.1454	0.3238	1.01	548	NO	1.00
p.est[11]	0.0625	0.0455	0.0116	0.0316	0.0513	0.0794	0.1820	1.01	722	NO	1.00
mean.p	0.0953	0.0694	0.0196	0.0490	0.0781	0.1217	0.2689	1.01	584	NO	1.00
beta1	-1.4148	0.8803	-3.4641	-1.9425	-1.3183	-0.8060	0.0335	1.00	945	YES	0.97
beta2	0.0002	0.7476	-1.5673	-0.4663	0.0331	0.4918	1.4661	1.00	3000	YES	0.52
deviance	63	5	54	60	63	67	74	1.00	438	NO	1.00
FOREST											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
phi.occ[1]	0.5090	0.1459	0.2564	0.4054	0.4998	0.6044	0.8233	1.00	643	NO	1.00
phi.occ[2]	0.8553	0.1112	0.5833	0.7902	0.8794	0.9432	0.9940	1.00	2029	NO	1.00
phi.occ[3]	0.4833	0.1143	0.2891	0.4019	0.4722	0.5534	0.7300	1.01	429	NO	1.00
phi.occ[4]	0.6326	0.1460	0.3786	0.5235	0.6209	0.7319	0.9326	1.00	3000	NO	1.00
phi.occ[5]	0.8565	0.1078	0.5998	0.7919	0.8749	0.9444	0.9953	1.01	431	NO	1.00
phi.occ[6]	0.8037	0.1226	0.5511	0.7160	0.8154	0.9020	0.9900	1.02	103	NO	1.00
phi.occ[7]	0.3817	0.1316	0.1955	0.2912	0.3590	0.4430	0.7265	1.02	147	NO	1.00
phi.occ[8]	0.7010	0.1579	0.3980	0.5865	0.7055	0.8228	0.9782	1.00	1854	NO	1.00
phi.occ[9]	0.7963	0.1299	0.5251	0.7065	0.8124	0.9015	0.9873	1.00	3000	NO	1.00
phi.occ[10]	0.5761	0.1677	0.2906	0.4523	0.5615	0.6864	0.9387	1.00	2081	NO	1.00
phi.occ[11]	0.6271	0.1802	0.2965	0.4906	0.6176	0.7588	0.9616	1.01	331	NO	1.00
p.est[1]	0.1774	0.0265	0.1305	0.1584	0.1761	0.1943	0.2358	1.00	3000	NO	1.00
p.est[2]	0.1308	0.0273	0.0856	0.1111	0.1278	0.1477	0.1910	1.00	1902	NO	1.00
p.est[3]	0.1963	0.0285	0.1450	0.1760	0.1947	0.2146	0.2570	1.00	2847	NO	1.00
p.est[4]	0.1448	0.0298	0.0927	0.1237	0.1428	0.1637	0.2106	1.00	3000	NO	1.00
p.est[5]	0.2446	0.0406	0.1737	0.2154	0.2418	0.2702	0.3317	1.00	605	NO	1.00
p.est[6]	0.1455	0.0266	0.1003	0.1258	0.1430	0.1624	0.2038	1.00	3000	NO	1.00
p.est[7]	0.2377	0.0766	0.1125	0.1809	0.2300	0.2870	0.4042	1.00	448	NO	1.00
p.est[8]	0.1490	0.0311	0.0942	0.1266	0.1469	0.1691	0.2170	1.00	3000	NO	1.00
p.est[9]	0.2525	0.0433	0.1763	0.2211	0.2493	0.2801	0.3452	1.00	550	NO	1.00
p.est[10]	0.1208	0.0273	0.0769	0.1013	0.1178	0.1377	0.1818	1.00	1962	NO	1.00
p.est[11]	0.2454	0.0409	0.1739	0.2161	0.2427	0.2713	0.3330	1.00	598	NO	1.00
mean.p	0.1264	0.0275	0.0811	0.1065	0.1235	0.1434	0.1883	1.00	1770	NO	1.00
beta1	-0.3178	0.1568	-0.6319	-0.4244	-0.3128	-0.2085	-0.0213	1.00	2935	NO	0.98
beta2	0.4348	0.1975	0.0393	0.3035	0.4363	0.5681	0.8120	1.01	328	NO	0.98
deviance	788	13	762	779	788	797	813	1.00	1812	NO	1.00

C) <i>Turdus flavipes</i> - Cormack-Jolly-Seber												
COFFEE												
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f	
phi.occ[1]	0.5041	0.2910	0.0250	0.2499	0.5128	0.7564	0.9755	1.00	3000	NO	1.00	
phi.occ[2]	0.5517	0.2270	0.1619	0.3729	0.5391	0.7293	0.9668	1.00	2324	NO	1.00	
phi.occ[3]	0.5905	0.2303	0.1745	0.4076	0.5855	0.7883	0.9746	1.00	3000	NO	1.00	
phi.occ[4]	0.5774	0.2041	0.2164	0.4165	0.5638	0.7320	0.9602	1.00	1594	NO	1.00	
phi.occ[5]	0.7104	0.1836	0.3243	0.5767	0.7276	0.8606	0.9857	1.00	3000	NO	1.00	
phi.occ[6]	0.8160	0.1454	0.4576	0.7319	0.8515	0.9318	0.9944	1.00	1754	NO	1.00	
phi.occ[7]	0.8449	0.1211	0.5495	0.7717	0.8712	0.9411	0.9951	1.00	3000	NO	1.00	
phi.occ[8]	0.6862	0.1903	0.3074	0.5418	0.6941	0.8464	0.9863	1.00	711	NO	1.00	
phi.occ[9]	0.7059	0.1875	0.3179	0.5716	0.7247	0.8607	0.9858	1.00	715	NO	1.00	
phi.occ[10]	0.4947	0.2047	0.1669	0.3391	0.4678	0.6448	0.9291	1.00	1300	NO	1.00	
phi.occ[11]	0.4125	0.2473	0.0574	0.2122	0.3631	0.5761	0.9484	1.01	473	NO	1.00	
p.est[1]	0.3417	0.1245	0.1368	0.2499	0.3267	0.4231	0.6082	1.00	457	NO	1.00	
p.est[2]	0.0655	0.0231	0.0306	0.0488	0.0621	0.0787	0.1210	1.00	2282	NO	1.00	
p.est[3]	0.2069	0.0663	0.0982	0.1582	0.2004	0.2467	0.3539	1.00	974	NO	1.00	
p.est[4]	0.0585	0.0280	0.0198	0.0385	0.0533	0.0729	0.1272	1.00	2593	NO	1.00	
p.est[5]	0.1688	0.0528	0.0833	0.1302	0.1632	0.2007	0.2894	1.00	1772	NO	1.00	
p.est[6]	0.4437	0.1663	0.1590	0.3187	0.4325	0.5615	0.7759	1.00	414	NO	1.00	
p.est[7]	0.0628	0.0221	0.0293	0.0468	0.0598	0.0752	0.1161	1.00	2048	NO	1.00	
p.est[8]	0.0640	0.0372	0.0164	0.0375	0.0555	0.0823	0.1557	1.00	3000	NO	1.00	
p.est[9]	0.0892	0.0306	0.0427	0.0666	0.0848	0.1070	0.1623	1.00	3000	NO	1.00	
p.est[10]	0.1139	0.0371	0.0563	0.0862	0.1090	0.1363	0.2013	1.00	3000	NO	1.00	
p.est[11]	0.0766	0.0268	0.0360	0.0569	0.0725	0.0916	0.1416	1.00	3000	NO	1.00	
mean.p	0.0994	0.0333	0.0477	0.0748	0.0946	0.1191	0.1802	1.00	3000	NO	1.00	
beta1	-1.0018	0.3352	-1.6546	-1.2263	-0.9927	-0.7733	-0.3671	1.00	456	NO	1.00	
beta2	0.3897	0.2824	-0.1528	0.1946	0.3861	0.5738	0.9483	1.00	825	YES	0.92	
deviance	225	6	213	221	225	229	238	1.00	1198	NO	1.00	
FOREST												
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f	
phi.occ[1]	0.3321	0.1753	0.0878	0.2027	0.2991	0.4374	0.7754	1.02	140	NO	1.00	
phi.occ[2]	0.7024	0.1764	0.3472	0.5701	0.7137	0.8467	0.9807	1.00	3000	NO	1.00	
phi.occ[3]	0.7520	0.1645	0.3915	0.6343	0.7716	0.8871	0.9878	1.00	1046	NO	1.00	
phi.occ[4]	0.7374	0.1599	0.4073	0.6246	0.7498	0.8681	0.9865	1.00	2539	NO	1.00	
phi.occ[5]	0.7262	0.1667	0.4007	0.5992	0.7386	0.8653	0.9859	1.00	3000	NO	1.00	
phi.occ[6]	0.8480	0.1175	0.5647	0.7784	0.8735	0.9423	0.9946	1.00	822	NO	1.00	
phi.occ[7]	0.6595	0.1761	0.3445	0.5271	0.6577	0.7963	0.9720	1.00	700	NO	1.00	
phi.occ[8]	0.6826	0.1701	0.3625	0.5530	0.6834	0.8176	0.9760	1.00	526	NO	1.00	
phi.occ[9]	0.6185	0.1809	0.2927	0.4786	0.6133	0.7542	0.9584	1.01	351	NO	1.00	
phi.occ[10]	0.8254	0.1287	0.5324	0.7426	0.8497	0.9309	0.9935	1.01	635	NO	1.00	
phi.occ[11]	0.8607	0.1134	0.5848	0.7991	0.8898	0.9506	0.9948	1.00	667	NO	1.00	
p.est[1]	0.0857	0.0201	0.0521	0.0714	0.0838	0.0976	0.1299	1.00	905	NO	1.00	
p.est[2]	0.0546	0.0170	0.0276	0.0423	0.0525	0.0648	0.0926	1.00	3000	NO	1.00	
p.est[3]	0.0999	0.0229	0.0615	0.0834	0.0983	0.1136	0.1518	1.00	707	NO	1.00	
p.est[4]	0.0630	0.0181	0.0345	0.0503	0.0611	0.0732	0.1048	1.00	1566	NO	1.00	
p.est[5]	0.1405	0.0362	0.0822	0.1151	0.1364	0.1616	0.2203	1.00	621	NO	1.00	
p.est[6]	0.0638	0.0177	0.0351	0.0510	0.0621	0.0745	0.1028	1.00	2072	NO	1.00	
p.est[7]	0.1338	0.0556	0.0566	0.0944	0.1226	0.1623	0.2707	1.01	667	NO	1.00	
p.est[8]	0.0657	0.0188	0.0361	0.0525	0.0636	0.0762	0.1095	1.00	1372	NO	1.00	
p.est[9]	0.1477	0.0392	0.0851	0.1200	0.1430	0.1706	0.2348	1.00	628	NO	1.00	
p.est[10]	0.0486	0.0162	0.0234	0.0367	0.0465	0.0582	0.0848	1.00	3000	NO	1.00	
p.est[11]	0.1412	0.0365	0.0826	0.1157	0.1371	0.1625	0.2217	1.00	621	NO	1.00	
mean.p	0.0519	0.0167	0.0254	0.0398	0.0499	0.0619	0.0894	1.00	3000	NO	1.00	
beta1	-0.4338	0.1989	-0.8375	-0.5653	-0.4320	-0.3041	-0.0404	1.00	1149	NO	0.98	
beta2	0.5918	0.2416	0.1285	0.4284	0.5856	0.7457	1.0882	1.00	2118	NO	1.00	
deviance	439	8	425	434	439	444	454	1.00	529	NO	1.00	

D) <i>Turdus albicollis</i> - Cormack-Jolly-Seber												
COFFEE												
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f	
phi.occ[1]	0.4979	0.2875	0.0235	0.2513	0.4954	0.7436	0.9719	1.00	2870	NO	1.00	
phi.occ[2]	0.4922	0.2900	0.0220	0.2424	0.4892	0.7372	0.9753	1.00	3000	NO	1.00	
phi.occ[3]	0.4972	0.2867	0.0248	0.2527	0.4986	0.7421	0.9718	1.00	1771	NO	1.00	
phi.occ[4]	0.5030	0.2856	0.0307	0.2579	0.5080	0.7500	0.9752	1.00	3000	NO	1.00	
phi.occ[5]	0.5036	0.2865	0.0252	0.2547	0.5170	0.7468	0.9710	1.00	3000	NO	1.00	
phi.occ[6]	0.4983	0.2888	0.0231	0.2468	0.5095	0.7416	0.9753	1.00	1221	NO	1.00	
phi.occ[7]	0.5036	0.2918	0.0263	0.2454	0.5059	0.7561	0.9768	1.00	3000	NO	1.00	
phi.occ[8]	0.4967	0.2917	0.0230	0.2393	0.5012	0.7432	0.9750	1.00	1613	NO	1.00	

	<b>phi.occ[9]</b>	<b>0.4831</b>	<b>0.2913</b>	<b>0.0193</b>	<b>0.2284</b>	<b>0.4767</b>	<b>0.7377</b>	<b>0.9710</b>	<b>1.00</b>	<b>1328</b>	<b>NO</b>	<b>1.00</b>
phi.occ[10]	<b>0.4941</b>	<b>0.2900</b>	<b>0.0250</b>	<b>0.2477</b>	<b>0.4886</b>	<b>0.7411</b>	<b>0.9801</b>	<b>1.00</b>	<b>1112</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[11]	<b>0.4962</b>	<b>0.2903</b>	<b>0.0233</b>	<b>0.2444</b>	<b>0.4866</b>	<b>0.7441</b>	<b>0.9765</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>	
p.est[1]	0.0193	0.1226	<0.0001	<0.0001	<0.0001	<0.0001	0.1599	1.02	2036	NO	1.00	
p.est[2]	0.0017	0.0239	<0.0001	<0.0001	<0.0001	<0.0001	0.0049	1.09	3000	NO	1.00	
p.est[3]	0.0080	0.0709	<0.0001	<0.0001	<0.0001	<0.0001	0.0275	1.05	1512	NO	1.00	
p.est[4]	0.0046	0.0570	<0.0001	<0.0001	<0.0001	<0.0001	0.0065	1.09	1394	NO	1.00	
p.est[5]	0.0060	0.0579	<0.0001	<0.0001	<0.0001	<0.0001	0.0193	1.08	1158	NO	1.00	
p.est[6]	0.0306	0.1581	<0.0001	<0.0001	<0.0001	<0.0001	0.7019	1.00	3000	NO	1.00	
p.est[7]	0.0016	0.0228	<0.0001	<0.0001	<0.0001	<0.0001	0.0047	1.12	3000	NO	1.00	
p.est[8]	0.0112	0.0859	<0.0001	<0.0001	<0.0001	<0.0001	0.0572	1.02	2608	NO	1.00	
p.est[9]	0.0033	0.0362	<0.0001	<0.0001	<0.0001	<0.0001	0.0096	1.06	2042	NO	1.00	
p.est[10]	0.0039	0.0411	<0.0001	<0.0001	<0.0001	<0.0001	0.0205	1.09	1305	NO	1.00	
p.est[11]	0.0026	0.0305	<0.0001	<0.0001	<0.0001	<0.0001	0.0071	1.02	3000	NO	1.00	
mean.p	0.0036	0.0388	<0.0001	<0.0001	<0.0001	<0.0001	0.0154	1.08	1573	NO	1.00	
beta1	-0.6842	5.6036	-9.5923	-5.5339	-1.0143	4.0625	9.3889	1.00	3000	YES	0.55	
beta2	-1.5233	5.3965	-9.6026	-6.1098	-2.1459	2.6795	8.8916	1.00	3000	YES	0.62	
deviance	0	0	0	0	0	0	1	1.02	3000	YES	1.00	
<b>FOREST</b>												
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f	
phi.occ[1]	<b>0.2449</b>	<b>0.0957</b>	<b>0.0945</b>	<b>0.1762</b>	<b>0.2313</b>	<b>0.3016</b>	<b>0.4566</b>	<b>1.01</b>	<b>1110</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[2]	<b>0.8327</b>	<b>0.1258</b>	<b>0.5377</b>	<b>0.7542</b>	<b>0.8578</b>	<b>0.9356</b>	<b>0.9945</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[3]	<b>0.8165</b>	<b>0.1313</b>	<b>0.5122</b>	<b>0.7355</b>	<b>0.8374</b>	<b>0.9227</b>	<b>0.9925</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[4]	<b>0.7827</b>	<b>0.1380</b>	<b>0.4851</b>	<b>0.6884</b>	<b>0.7979</b>	<b>0.8932</b>	<b>0.9875</b>	<b>1.00</b>	<b>2597</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[5]	<b>0.6711</b>	<b>0.1455</b>	<b>0.3989</b>	<b>0.5629</b>	<b>0.6685</b>	<b>0.7800</b>	<b>0.9557</b>	<b>1.00</b>	<b>872</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[6]	<b>0.8383</b>	<b>0.1109</b>	<b>0.5985</b>	<b>0.7651</b>	<b>0.8551</b>	<b>0.9306</b>	<b>0.9930</b>	<b>1.00</b>	<b>1867</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[7]	<b>0.5583</b>	<b>0.1442</b>	<b>0.3219</b>	<b>0.4495</b>	<b>0.5402</b>	<b>0.6501</b>	<b>0.8780</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[8]	<b>0.6986</b>	<b>0.1554</b>	<b>0.4071</b>	<b>0.5789</b>	<b>0.6981</b>	<b>0.8207</b>	<b>0.9770</b>	<b>1.00</b>	<b>515</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[9]	<b>0.6734</b>	<b>0.1661</b>	<b>0.3755</b>	<b>0.5440</b>	<b>0.6653</b>	<b>0.8061</b>	<b>0.9734</b>	<b>1.00</b>	<b>498</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[10]	<b>0.8491</b>	<b>0.1101</b>	<b>0.5968</b>	<b>0.7767</b>	<b>0.8673</b>	<b>0.9414</b>	<b>0.9941</b>	<b>1.01</b>	<b>443</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[11]	<b>0.8455</b>	<b>0.1135</b>	<b>0.5868</b>	<b>0.7738</b>	<b>0.8654</b>	<b>0.9386</b>	<b>0.9940</b>	<b>1.00</b>	<b>1175</b>	<b>NO</b>	<b>1.00</b>	
p.est[1]	0.2370	0.0393	0.1654	0.2096	0.2343	0.2618	0.3192	1.00	677	NO	1.00	
p.est[2]	0.1266	0.0302	0.0750	0.1044	0.1245	0.1453	0.1928	1.00	2100	NO	1.00	
p.est[3]	0.2842	0.0448	0.2020	0.2539	0.2819	0.3125	0.3800	1.00	610	NO	1.00	
p.est[4]	0.1142	0.0259	0.0680	0.0958	0.1124	0.1299	0.1711	1.00	1226	NO	1.00	
p.est[5]	0.4042	0.0646	0.2810	0.3599	0.4015	0.4458	0.5373	1.00	719	NO	1.00	
p.est[6]	0.1603	0.0331	0.1016	0.1370	0.1582	0.1808	0.2321	1.00	1238	NO	1.00	
p.est[7]	0.2295	0.0739	0.1098	0.1754	0.2205	0.2754	0.3967	1.00	738	NO	1.00	
p.est[8]	0.1185	0.0269	0.0708	0.0995	0.1168	0.1348	0.1776	1.00	1090	NO	1.00	
p.est[9]	0.4232	0.0682	0.2949	0.3764	0.4205	0.4681	0.5614	1.00	754	NO	1.00	
p.est[10]	0.1009	0.0264	0.0569	0.0817	0.0988	0.1166	0.1592	1.00	3000	NO	1.00	
p.est[11]	0.4063	0.0650	0.2822	0.3618	0.4035	0.4482	0.5398	1.00	723	NO	1.00	
mean.p	0.1163	0.0290	0.0674	0.0949	0.1143	0.1339	0.1813	1.00	2598	NO	1.00	
beta1	-0.8054	0.1889	-1.1849	-0.9307	-0.7992	-0.6787	-0.4520	1.00	1408	NO	1.00	
beta2	0.7372	0.2174	0.3197	0.5905	0.7338	0.8742	1.1767	1.00	3000	NO	1.00	
deviance	512	13	486	503	513	522	538	1.00	1876	NO	1.00	

<b>G) Saltator maximus - Cormack-Jolly-Seber</b>												
<b>COFFEE</b>												
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f	
phi.occ[1]	<b>0.3742</b>	<b>0.2702</b>	<b>0.0125</b>	<b>0.1407</b>	<b>0.3276</b>	<b>0.5715</b>	<b>0.9382</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[2]	<b>0.5323</b>	<b>0.2151</b>	<b>0.1679</b>	<b>0.3675</b>	<b>0.5156</b>	<b>0.6923</b>	<b>0.9460</b>	<b>1.00</b>	<b>1182</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[3]	<b>0.7287</b>	<b>0.1963</b>	<b>0.2976</b>	<b>0.5902</b>	<b>0.7687</b>	<b>0.8904</b>	<b>0.9901</b>	<b>1.00</b>	<b>2234</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[4]	<b>0.6702</b>	<b>0.1980</b>	<b>0.2860</b>	<b>0.5205</b>	<b>0.6845</b>	<b>0.8316</b>	<b>0.9815</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[5]	<b>0.6283</b>	<b>0.2036</b>	<b>0.2410</b>	<b>0.4746</b>	<b>0.6308</b>	<b>0.7897</b>	<b>0.9754</b>	<b>1.00</b>	<b>2621</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[6]	<b>0.7683</b>	<b>0.1732</b>	<b>0.3661</b>	<b>0.6570</b>	<b>0.8031</b>	<b>0.9088</b>	<b>0.9919</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[7]	<b>0.6823</b>	<b>0.2024</b>	<b>0.2696</b>	<b>0.5298</b>	<b>0.7100</b>	<b>0.8519</b>	<b>0.9852</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[8]	<b>0.6608</b>	<b>0.2098</b>	<b>0.2425</b>	<b>0.5058</b>	<b>0.6796</b>	<b>0.8354</b>	<b>0.9854</b>	<b>1.00</b>	<b>1354</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[9]	<b>0.6176</b>	<b>0.2236</b>	<b>0.1913</b>	<b>0.4464</b>	<b>0.6176</b>	<b>0.8037</b>	<b>0.9801</b>	<b>1.00</b>	<b>2292</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[10]	<b>0.6307</b>	<b>0.2175</b>	<b>0.2071</b>	<b>0.4666</b>	<b>0.6408</b>	<b>0.8098</b>	<b>0.9786</b>	<b>1.00</b>	<b>2874</b>	<b>NO</b>	<b>1.00</b>	
phi.occ[11]	<b>0.5382</b>	<b>0.2568</b>	<b>0.0821</b>	<b>0.3320</b>	<b>0.5395</b>	<b>0.7503</b>	<b>0.9737</b>	<b>1.00</b>	<b>700</b>	<b>NO</b>	<b>1.00</b>	
p.est[1]	0.4502	0.1671	0.1692	0.3268	0.4349	0.5584	0.8018	1.00	3000	NO	1.00	
p.est[2]	0.0957	0.0421	0.0349	0.0655	0.0879	0.1204	0.1964	1.00	1228	NO	1.00	
p.est[3]	0.2956	0.1052	0.1338	0.2185	0.2796	0.3587	0.5334	1.00	3000	NO	1.00	
p.est[4]	0.0813	0.0492	0.0166	0.0457	0.0704	0.1064	0.2054	1.00	3000	NO	1.00	
p.est[5]	0.2456	0.0874	0.1093	0.1815	0.2327	0.2985	0.4474	1.00	3000	NO	1.00	
p.est[6]	0.5474	0.2008	0.1752	0.3947	0.5501	0.6989	0.9132	1.00	3000	NO	1.00	
p.est[7]	0.0913	0.0403	0.0329	0.0621	0.0840	0.1145	0.1896	1.00	1296	NO	1.00	
p.est[8]	0.0895	0.0654	0.0124	0.0424	0.0722	0.1204	0.2592	1.00	3000	NO	1.00	

p.est[9]	0.1328	0.0554	0.0502	0.0917	0.1229	0.1649	0.2629	1.00	1271	NO	1.00
p.est[10]	0.1691	0.0655	0.0685	0.1214	0.1581	0.2084	0.3188	1.00	1773	NO	1.00
p.est[11]	0.1134	0.0489	0.0418	0.0771	0.1049	0.1418	0.2306	1.00	1175	NO	1.00
mean.p	0.1480	0.0598	0.0584	0.1040	0.1375	0.1832	0.2879	1.00	1423	NO	1.00
beta1	-1.0776	0.4426	-2.0003	-1.3600	-1.0527	-0.7850	-0.2669	1.00	3000	NO	1.00
beta2	0.3722	0.3934	-0.3694	0.1100	0.3675	0.6274	1.1763	1.00	1377	YES	0.83
deviance	124	6	112	120	124	128	136	1.00	3000	NO	1.00
<b>FOREST</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
phi.occ[1]	<b>0.7106</b>	<b>0.1969</b>	<b>0.2893</b>	<b>0.5739</b>	<b>0.7371</b>	<b>0.8744</b>	<b>0.9864</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[2]	<b>0.6758</b>	<b>0.1960</b>	<b>0.2935</b>	<b>0.5290</b>	<b>0.6902</b>	<b>0.8377</b>	<b>0.9818</b>	<b>1.00</b>	<b>839</b>	<b>NO</b>	<b>1.00</b>
phi.occ[3]	<b>0.6304</b>	<b>0.2122</b>	<b>0.2245</b>	<b>0.4633</b>	<b>0.6427</b>	<b>0.8027</b>	<b>0.9761</b>	<b>1.00</b>	<b>813</b>	<b>NO</b>	<b>1.00</b>
phi.occ[4]	<b>0.5596</b>	<b>0.2132</b>	<b>0.1902</b>	<b>0.3880</b>	<b>0.5453</b>	<b>0.7249</b>	<b>0.9677</b>	<b>1.01</b>	<b>240</b>	<b>NO</b>	<b>1.00</b>
phi.occ[5]	<b>0.7373</b>	<b>0.1811</b>	<b>0.3410</b>	<b>0.6132</b>	<b>0.7648</b>	<b>0.8867</b>	<b>0.9891</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[6]	<b>0.8031</b>	<b>0.1485</b>	<b>0.4562</b>	<b>0.7106</b>	<b>0.8346</b>	<b>0.9237</b>	<b>0.9925</b>	<b>1.00</b>	<b>964</b>	<b>NO</b>	<b>1.00</b>
phi.occ[7]	<b>0.7131</b>	<b>0.1842</b>	<b>0.3385</b>	<b>0.5761</b>	<b>0.7316</b>	<b>0.8675</b>	<b>0.9892</b>	<b>1.00</b>	<b>1761</b>	<b>NO</b>	<b>1.00</b>
phi.occ[8]	<b>0.5516</b>	<b>0.2171</b>	<b>0.1764</b>	<b>0.3774</b>	<b>0.5394</b>	<b>0.7193</b>	<b>0.9572</b>	<b>1.00</b>	<b>530</b>	<b>NO</b>	<b>1.00</b>
phi.occ[9]	<b>0.6739</b>	<b>0.2102</b>	<b>0.2368</b>	<b>0.5172</b>	<b>0.6991</b>	<b>0.8521</b>	<b>0.9809</b>	<b>1.00</b>	<b>1921</b>	<b>NO</b>	<b>1.00</b>
phi.occ[10]	<b>0.7328</b>	<b>0.1907</b>	<b>0.3193</b>	<b>0.6037</b>	<b>0.7653</b>	<b>0.8913</b>	<b>0.9907</b>	<b>1.00</b>	<b>1706</b>	<b>NO</b>	<b>1.00</b>
phi.occ[11]	<b>0.7352</b>	<b>0.1927</b>	<b>0.3010</b>	<b>0.6096</b>	<b>0.7683</b>	<b>0.8945</b>	<b>0.9924</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
p.est[1]	0.2018	0.0625	0.0995	0.1555	0.1957	0.2411	0.3426	1.00	664	NO	1.00
p.est[2]	0.1446	0.0615	0.0531	0.0996	0.1352	0.1796	0.2863	1.00	416	NO	1.00
p.est[3]	0.2255	0.0691	0.1135	0.1743	0.2180	0.2666	0.3784	1.00	1068	NO	1.00
p.est[4]	0.1385	0.0559	0.0529	0.0987	0.1299	0.1710	0.2701	1.00	729	NO	1.00
p.est[5]	0.2863	0.0999	0.1267	0.2133	0.2741	0.3470	0.5084	1.00	3000	NO	1.00
p.est[6]	0.1627	0.0603	0.0679	0.1188	0.1547	0.1991	0.3010	1.00	437	NO	1.00
p.est[7]	0.2168	0.1135	0.0563	0.1344	0.1960	0.2787	0.4953	1.00	2354	NO	1.00
p.est[8]	0.1414	0.0567	0.0545	0.1009	0.1326	0.1741	0.2769	1.00	831	NO	1.00
p.est[9]	0.2962	0.1059	0.1261	0.2195	0.2829	0.3616	0.5321	1.00	3000	NO	1.00
p.est[10]	0.1297	0.0610	0.0414	0.0854	0.1187	0.1646	0.2739	1.01	418	NO	1.00
p.est[11]	0.2874	0.1005	0.1267	0.2140	0.2748	0.3486	0.5123	1.00	3000	NO	1.00
mean.p	0.1388	0.0617	0.0472	0.0935	0.1286	0.1733	0.2837	1.01	414	NO	1.00
beta1	-0.4522	0.3442	-1.1371	-0.6824	-0.4400	-0.2247	0.1955	1.00	3000	YES	0.91
beta2	0.4347	0.3934	-0.3202	0.1624	0.4269	0.7050	1.2159	1.00	731	YES	0.87
deviance	149	7	136	144	149	153	162	1.00	2081	NO	1.00

<b>COFFEE</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
phi.occ[1]	<b>0.6852</b>	<b>0.2173</b>	<b>0.2167</b>	<b>0.5360</b>	<b>0.7188</b>	<b>0.8701</b>	<b>0.9868</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[2]	<b>0.5660</b>	<b>0.2388</b>	<b>0.1271</b>	<b>0.3795</b>	<b>0.5665</b>	<b>0.7553</b>	<b>0.9773</b>	<b>1.01</b>	<b>188</b>	<b>NO</b>	<b>1.00</b>
phi.occ[3]	<b>0.7274</b>	<b>0.1991</b>	<b>0.2756</b>	<b>0.5970</b>	<b>0.7662</b>	<b>0.8932</b>	<b>0.9885</b>	<b>1.00</b>	<b>989</b>	<b>NO</b>	<b>1.00</b>
phi.occ[4]	<b>0.6235</b>	<b>0.2345</b>	<b>0.1730</b>	<b>0.4385</b>	<b>0.6427</b>	<b>0.8236</b>	<b>0.9815</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[5]	<b>0.6373</b>	<b>0.2320</b>	<b>0.1733</b>	<b>0.4630</b>	<b>0.6562</b>	<b>0.8356</b>	<b>0.9838</b>	<b>1.01</b>	<b>215</b>	<b>NO</b>	<b>1.00</b>
phi.occ[6]	<b>0.6894</b>	<b>0.2185</b>	<b>0.2168</b>	<b>0.5365</b>	<b>0.7274</b>	<b>0.8738</b>	<b>0.9864</b>	<b>1.01</b>	<b>257</b>	<b>NO</b>	<b>1.00</b>
phi.occ[7]	<b>0.7383</b>	<b>0.1824</b>	<b>0.3324</b>	<b>0.6177</b>	<b>0.7671</b>	<b>0.8891</b>	<b>0.9883</b>	<b>1.00</b>	<b>1420</b>	<b>NO</b>	<b>1.00</b>
phi.occ[8]	<b>0.7232</b>	<b>0.1888</b>	<b>0.3171</b>	<b>0.5841</b>	<b>0.7465</b>	<b>0.8832</b>	<b>0.9913</b>	<b>1.00</b>	<b>429</b>	<b>NO</b>	<b>1.00</b>
phi.occ[9]	<b>0.7301</b>	<b>0.1853</b>	<b>0.3314</b>	<b>0.5977</b>	<b>0.7603</b>	<b>0.8860</b>	<b>0.9890</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[10]	<b>0.6817</b>	<b>0.2133</b>	<b>0.2384</b>	<b>0.5251</b>	<b>0.7098</b>	<b>0.8620</b>	<b>0.9872</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[11]	<b>0.7032</b>	<b>0.2138</b>	<b>0.2345</b>	<b>0.5507</b>	<b>0.7402</b>	<b>0.8859</b>	<b>0.9869</b>	<b>1.00</b>	<b>1567</b>	<b>NO</b>	<b>1.00</b>
p.est[1]	0.1259	0.0915	0.0142	0.0584	0.1039	0.1705	0.3602	1.00	447	NO	1.00
p.est[2]	0.0846	0.0434	0.0252	0.0532	0.0755	0.1080	0.1894	1.01	425	NO	1.00
p.est[3]	0.1269	0.0633	0.0355	0.0805	0.1162	0.1615	0.2792	1.01	374	NO	1.00
p.est[4]	0.0371	0.0327	0.0025	0.0139	0.0282	0.0501	0.1232	1.01	614	NO	1.00
p.est[5]	0.1271	0.0585	0.0420	0.0849	0.1176	0.1596	0.2655	1.00	402	NO	1.00
p.est[6]	0.1277	0.1184	0.0062	0.0427	0.0938	0.1723	0.4552	1.00	546	NO	1.00
p.est[7]	0.0787	0.0414	0.0225	0.0487	0.0703	0.1007	0.1782	1.01	405	NO	1.00
p.est[8]	0.0314	0.0342	0.0009	0.0084	0.0204	0.0424	0.1239	1.01	773	NO	1.00
p.est[9]	0.1123	0.0526	0.0378	0.0741	0.1021	0.1411	0.2403	1.00	551	NO	1.00
p.est[10]	0.1224	0.0551	0.0434	0.0831	0.1120	0.1519	0.2566	1.00	528	NO	1.00
p.est[11]	0.1014	0.0491	0.0322	0.0655	0.0920	0.1274	0.2223	1.00	504	NO	1.00
mean.p	0.1177	0.0540	0.0399	0.0786	0.1081	0.1466	0.2508	1.00	554	NO	1.00
beta1	-0.3925	0.5471	-1.4358	-0.7656	-0.3975	-0.0483	0.7472	1.00	1939	YES	0.78
beta2	-0.4437	0.5142	-1.5620	-0.7724	-0.4030	-0.0997	0.4964	1.00	1555	YES	0.81
deviance	97	5	88	93	96	100	107	1.00	3000	NO	1.00
<b>FOREST</b>											
par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
phi.occ[1]	<b>0.5073</b>	<b>0.2909</b>	<b>0.0300</b>	<b>0.2530</b>	<b>0.5146</b>	<b>0.7615</b>	<b>0.9730</b>	<b>1.00</b>	<b>2557</b>	<b>0</b>	<b>1.00</b>
phi.occ[2]	<b>0.4933</b>	<b>0.2850</b>	<b>0.0304</b>	<b>0.2442</b>	<b>0.4932</b>	<b>0.7370</b>	<b>0.9713</b>	<b>1.00</b>	<b>3000</b>	<b>0</b>	<b>1.00</b>

phi.occ[3]	<b>0.4909</b>	<b>0.2889</b>	<b>0.0237</b>	<b>0.2388</b>	<b>0.4853</b>	<b>0.7423</b>	<b>0.9725</b>	<b>1.00</b>	<b>3000</b>	<b>0</b>	<b>1.00</b>
phi.occ[4]	<b>0.4839</b>	<b>0.2878</b>	<b>0.0212</b>	<b>0.2394</b>	<b>0.4744</b>	<b>0.7267</b>	<b>0.9777</b>	<b>1.00</b>	<b>3000</b>	<b>0</b>	<b>1.00</b>
phi.occ[5]	<b>0.4669</b>	<b>0.2859</b>	<b>0.0239</b>	<b>0.2226</b>	<b>0.4471</b>	<b>0.7094</b>	<b>0.9692</b>	<b>1.00</b>	<b>737</b>	<b>0</b>	<b>1.00</b>
phi.occ[6]	<b>0.4882</b>	<b>0.2900</b>	<b>0.0224</b>	<b>0.2412</b>	<b>0.4790</b>	<b>0.7360</b>	<b>0.9741</b>	<b>1.00</b>	<b>3000</b>	<b>0</b>	<b>1.00</b>
phi.occ[7]	<b>0.4788</b>	<b>0.2939</b>	<b>0.0257</b>	<b>0.2199</b>	<b>0.4590</b>	<b>0.7366</b>	<b>0.9748</b>	<b>1.00</b>	<b>386</b>	<b>0</b>	<b>1.00</b>
phi.occ[8]	<b>0.6318</b>	<b>0.2485</b>	<b>0.1254</b>	<b>0.4444</b>	<b>0.6694</b>	<b>0.8506</b>	<b>0.9864</b>	<b>1.00</b>	<b>784</b>	<b>0</b>	<b>1.00</b>
phi.occ[9]	<b>0.6120</b>	<b>0.2511</b>	<b>0.1174</b>	<b>0.4128</b>	<b>0.6364</b>	<b>0.8315</b>	<b>0.9838</b>	<b>1.00</b>	<b>2653</b>	<b>0</b>	<b>1.00</b>
phi.occ[10]	<b>0.6579</b>	<b>0.2387</b>	<b>0.1360</b>	<b>0.4965</b>	<b>0.6985</b>	<b>0.8550</b>	<b>0.9855</b>	<b>1.00</b>	<b>2860</b>	<b>0</b>	<b>1.00</b>
phi.occ[11]	<b>0.6410</b>	<b>0.2435</b>	<b>0.1377</b>	<b>0.4526</b>	<b>0.6760</b>	<b>0.8461</b>	<b>0.9835</b>	<b>1.00</b>	<b>3000</b>	<b>0</b>	<b>1.00</b>
p.est[1]	0.0359	0.0894	<0.0001	0.0002	0.0029	0.0260	0.3101	1.00	1737	0	1.00
p.est[2]	0.0168	0.0667	<0.0001	<0.0001	0.0001	0.0036	0.1733	1.00	3000	0	1.00
p.est[3]	0.0383	0.0805	<0.0001	0.0010	0.0074	0.0374	0.2706	1.00	1506	0	1.00
p.est[4]	0.0010	0.0092	<0.0001	<0.0001	<0.0001	<0.0001	0.0079	1.06	3000	0	1.00
p.est[5]	0.0863	0.1038	0.0015	0.0186	0.0489	0.1150	0.3791	1.00	3000	0	1.00
p.est[6]	0.0277	0.0890	<0.0001	<0.0001	0.0003	0.0100	0.2883	1.00	3000	0	1.00
p.est[7]	0.0332	0.1492	<0.0001	<0.0001	<0.0001	0.0002	0.6120	1.04	381	0	1.00
p.est[8]	0.0014	0.0134	<0.0001	<0.0001	<0.0001	<0.0001	0.0086	1.04	2552	0	1.00
p.est[9]	0.1121	0.1294	0.0019	0.0239	0.0650	0.1518	0.4760	1.00	2736	0	1.00
p.est[10]	0.0048	0.0292	<0.0001	<0.0001	<0.0001	0.0005	0.0382	1.03	3000	0	1.00
p.est[11]	0.0887	0.1060	0.0015	0.0194	0.0506	0.1174	0.3891	1.00	3000	0	1.00
mean.p	0.0121	0.0546	<0.0001	<0.0001	<0.0001	0.0021	0.1241	1.00	3000	0	1.00
beta1	-5.9333	2.5793	-9.8063	-8.1456	-6.1405	-4.0003	-0.7178	1.00	2105	0	0.99
beta2	1.2717	4.4057	-7.3890	-1.8189	1.2987	4.6546	9.1068	1.00	428	1	0.61
deviance	13	2	9	11	12	14	19	1.00	3000	0	1.00

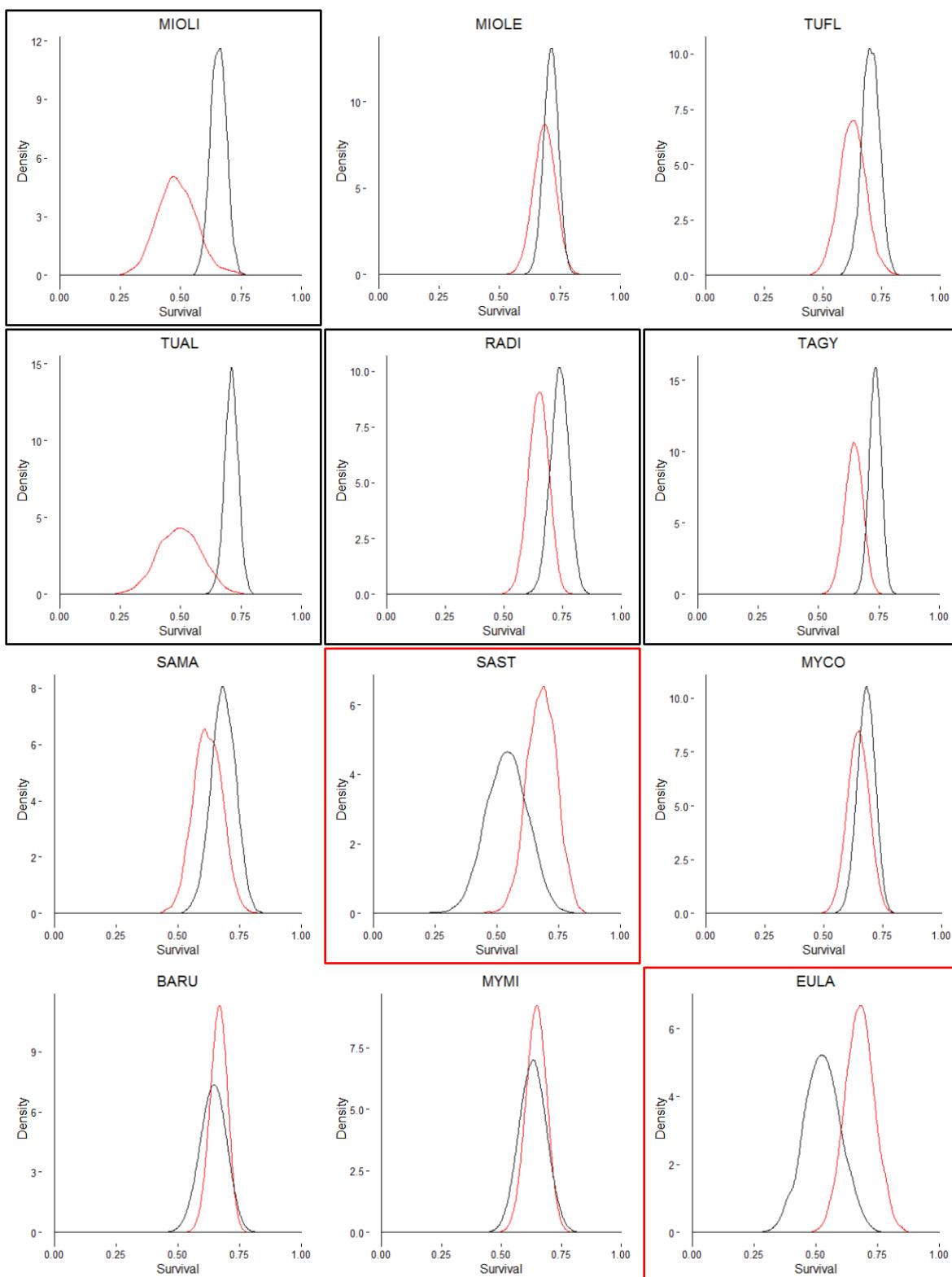
**L) Euphonia laniirostris - Cormack-Jolly-Seber****COFFEE**

par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
phi.occ[1]	<b>0.5037</b>	<b>0.2900</b>	<b>0.0248</b>	<b>0.2550</b>	<b>0.5078</b>	<b>0.7539</b>	<b>0.9788</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[2]	<b>0.6954</b>	<b>0.1941</b>	<b>0.2967</b>	<b>0.5590</b>	<b>0.7156</b>	<b>0.8574</b>	<b>0.9853</b>	<b>1.00</b>	<b>1113</b>	<b>NO</b>	<b>1.00</b>
phi.occ[3]	<b>0.7298</b>	<b>0.1881</b>	<b>0.3158</b>	<b>0.5956</b>	<b>0.7577</b>	<b>0.8882</b>	<b>0.9899</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[4]	<b>0.6530</b>	<b>0.2011</b>	<b>0.2649</b>	<b>0.5011</b>	<b>0.6627</b>	<b>0.8176</b>	<b>0.9780</b>	<b>1.01</b>	<b>327</b>	<b>NO</b>	<b>1.00</b>
phi.occ[5]	<b>0.7333</b>	<b>0.1812</b>	<b>0.3403</b>	<b>0.6106</b>	<b>0.7615</b>	<b>0.8849</b>	<b>0.9866</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[6]	<b>0.7777</b>	<b>0.1680</b>	<b>0.3949</b>	<b>0.6692</b>	<b>0.8109</b>	<b>0.9160</b>	<b>0.9931</b>	<b>1.00</b>	<b>2895</b>	<b>NO</b>	<b>1.00</b>
phi.occ[7]	<b>0.7965</b>	<b>0.1593</b>	<b>0.4098</b>	<b>0.7097</b>	<b>0.8341</b>	<b>0.9227</b>	<b>0.9926</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[8]	<b>0.7828</b>	<b>0.1684</b>	<b>0.3905</b>	<b>0.6801</b>	<b>0.8168</b>	<b>0.9218</b>	<b>0.9925</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[9]	<b>0.6495</b>	<b>0.2369</b>	<b>0.1756</b>	<b>0.4743</b>	<b>0.6807</b>	<b>0.8558</b>	<b>0.9848</b>	<b>1.01</b>	<b>353</b>	<b>NO</b>	<b>1.00</b>
phi.occ[10]	<b>0.6959</b>	<b>0.2160</b>	<b>0.2206</b>	<b>0.5494</b>	<b>0.7372</b>	<b>0.8760</b>	<b>0.9901</b>	<b>1.00</b>	<b>677</b>	<b>NO</b>	<b>1.00</b>
phi.occ[11]	<b>0.4493</b>	<b>0.2827</b>	<b>0.0179</b>	<b>0.2057</b>	<b>0.4352</b>	<b>0.6815</b>	<b>0.9605</b>	<b>1.00</b>	<b>541</b>	<b>NO</b>	<b>1.00</b>
p.est[1]	0.3457	0.1525	0.1135	0.2306	0.3198	0.4375	0.6956	1.00	2113	NO	1.00
p.est[2]	0.0298	0.0174	0.0076	0.0173	0.0260	0.0377	0.0728	1.00	789	NO	1.00
p.est[3]	0.1305	0.0628	0.0431	0.0849	0.1191	0.1645	0.2826	1.00	1392	NO	1.00
p.est[4]	0.0765	0.0352	0.0268	0.0518	0.0702	0.0944	0.1649	1.00	3000	NO	1.00
p.est[5]	0.0895	0.0444	0.0273	0.0576	0.0813	0.1119	0.1958	1.00	1095	NO	1.00
p.est[6]	0.5424	0.1950	0.1801	0.3951	0.5373	0.6919	0.9058	1.00	3000	NO	1.00
p.est[7]	0.0303	0.0172	0.0079	0.0180	0.0268	0.0385	0.0741	1.00	818	NO	1.00
p.est[8]	0.1232	0.0632	0.0359	0.0785	0.1117	0.1543	0.2834	1.00	3000	NO	1.00
p.est[9]	0.0353	0.0212	0.0084	0.0204	0.0308	0.0454	0.0884	1.00	742	NO	1.00
p.est[10]	0.0475	0.0267	0.0121	0.0283	0.0417	0.0609	0.1144	1.00	791	NO	1.00
p.est[11]	0.0311	0.0189	0.0076	0.0177	0.0269	0.0397	0.0793	1.00	745	NO	1.00
mean.p	0.0398	0.0233	0.0098	0.0231	0.0348	0.0510	0.0988	1.00	755	NO	1.00
beta1	-1.1805	0.4142	-2.0937	-1.4255	-1.1595	-0.8959	-0.4355	1.00	3000	NO	1.00
beta2	1.1439	0.3997	0.3747	0.8740	1.1324	1.4087	1.9820	1.00	770	NO	1.00
deviance	119	6	109	115	119	123	131	1.00	2304	NO	1.00

**FOREST**

par	mean	sd	2.50%	25%	50%	75%	97.50%	Rhat	n.eff	over0	f
phi.occ[1]	<b>0.4276</b>	<b>0.2841</b>	<b>0.0165</b>	<b>0.1802</b>	<b>0.3837</b>	<b>0.6616</b>	<b>0.9577</b>	<b>1.00</b>	<b>2733</b>	<b>NO</b>	<b>1.00</b>
phi.occ[2]	<b>0.6578</b>	<b>0.2182</b>	<b>0.2095</b>	<b>0.4954</b>	<b>0.6802</b>	<b>0.8359</b>	<b>0.9853</b>	<b>1.00</b>	<b>1617</b>	<b>NO</b>	<b>1.00</b>
phi.occ[3]	<b>0.7074</b>	<b>0.2020</b>	<b>0.2657</b>	<b>0.5678</b>	<b>0.7393</b>	<b>0.8750</b>	<b>0.9880</b>	<b>1.00</b>	<b>1005</b>	<b>NO</b>	<b>1.00</b>
phi.occ[4]	<b>0.4280</b>	<b>0.2559</b>	<b>0.0489</b>	<b>0.2193</b>	<b>0.3841</b>	<b>0.6228</b>	<b>0.9429</b>	<b>1.00</b>	<b>588</b>	<b>NO</b>	<b>1.00</b>
phi.occ[5]	<b>0.5046</b>	<b>0.2625</b>	<b>0.0699</b>	<b>0.2871</b>	<b>0.4877</b>	<b>0.7233</b>	<b>0.9635</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[6]	<b>0.5908</b>	<b>0.2527</b>	<b>0.1101</b>	<b>0.3930</b>	<b>0.6046</b>	<b>0.8059</b>	<b>0.9788</b>	<b>1.00</b>	<b>2077</b>	<b>NO</b>	<b>1.00</b>
phi.occ[7]	<b>0.3924</b>	<b>0.2672</b>	<b>0.0359</b>	<b>0.1682</b>	<b>0.3309</b>	<b>0.6002</b>	<b>0.9405</b>	<b>1.01</b>	<b>260</b>	<b>NO</b>	<b>1.00</b>
phi.occ[8]	<b>0.3869</b>	<b>0.2510</b>	<b>0.0459</b>	<b>0.1796</b>	<b>0.3328</b>	<b>0.5567</b>	<b>0.9446</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
phi.occ[9]	<b>0.4445</b>	<b>0.2564</b>	<b>0.0532</b>	<b>0.2342</b>	<b>0.4119</b>	<b>0.6407</b>	<b>0.9424</b>	<b>1.00</b>	<b>1452</b>	<b>NO</b>	<b>1.00</b>
phi.occ[10]	<b>0.6112</b>	<b>0.2342</b>	<b>0.1550</b>	<b>0.4307</b>	<b>0.6236</b>	<b>0.8090</b>	<b>0.9808</b>	<b>1.00</b>	<b>566</b>	<b>NO</b>	<b>1.00</b>
phi.occ[11]	<b>0.6481</b>	<b>0.2289</b>	<b>0.1784</b>	<b>0.4748</b>	<b>0.6739</b>	<b>0.8408</b>	<b>0.9832</b>	<b>1.00</b>	<b>3000</b>	<b>NO</b>	<b>1.00</b>
p.est[1]	0.1198	0.0710	0.0300	0.0691	0.1034	0.1532	0.3002	1.00	912	NO	1.00
p.est[2]	0.0578	0.0473	0.0063	0.0244	0.0456	0.0766	0.1860	1.01	299	NO	1.00

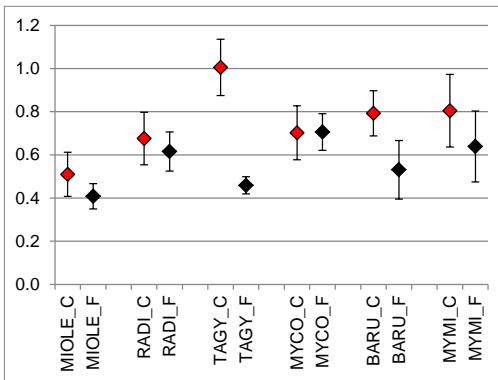
p.est[3]	0.1562	0.0889	0.0416	0.0928	0.1345	0.2008	0.3800	1.00	2998	NO	1.00
p.est[4]	0.0787	0.0835	0.0086	0.0295	0.0545	0.0961	0.3115	1.01	743	NO	1.00
p.est[5]	0.2736	0.1587	0.0613	0.1580	0.2370	0.3607	0.6728	1.00	2912	NO	1.00
p.est[6]	0.0738	0.0534	0.0113	0.0357	0.0610	0.0968	0.2101	1.01	348	NO	1.00
p.est[7]	0.2763	0.2834	0.0072	0.0642	0.1570	0.4070	0.9747	1.00	658	NO	1.00
p.est[8]	0.0866	0.0952	0.0089	0.0309	0.0575	0.1020	0.3716	1.01	798	NO	1.00
p.est[9]	0.2943	0.1705	0.0637	0.1686	0.2557	0.3898	0.7143	1.00	2151	NO	1.00
p.est[10]	0.0473	0.0413	0.0042	0.0186	0.0354	0.0634	0.1570	1.01	290	NO	1.00
p.est[11]	0.2758	0.1600	0.0615	0.1592	0.2385	0.3642	0.6784	1.00	2808	NO	1.00
mean.p	0.0533	0.0452	0.0052	0.0218	0.0408	0.0708	0.1745	1.01	292	NO	1.00
beta1	-0.8394	0.6094	-2.1336	-1.2199	-0.8059	-0.4442	0.2851	1.00	457	YES	0.93
beta2	1.1050	0.8830	-0.4276	0.5399	1.0109	1.5798	3.0750	1.01	361	YES	0.92
deviance	70	5	62	67	70	73	80	1.00	3000	NO	1.00



**Figure S5.** Posterior distributions for the estimates of apparent survival for the twelve focal species in coffee (red lines) and forest (black lines); less overlap means higher probability of differences between habitats. Estimation method and simulations conditions varied among species but not between habitats within a species.

### 2.3.4. Inter-seasonal variance

Inter-seasonal variance was calculated from the outputs of Jolly-Seber models (**N[i]** in Table S10b, e, f, i, j and k) for six species with high recapture rates.



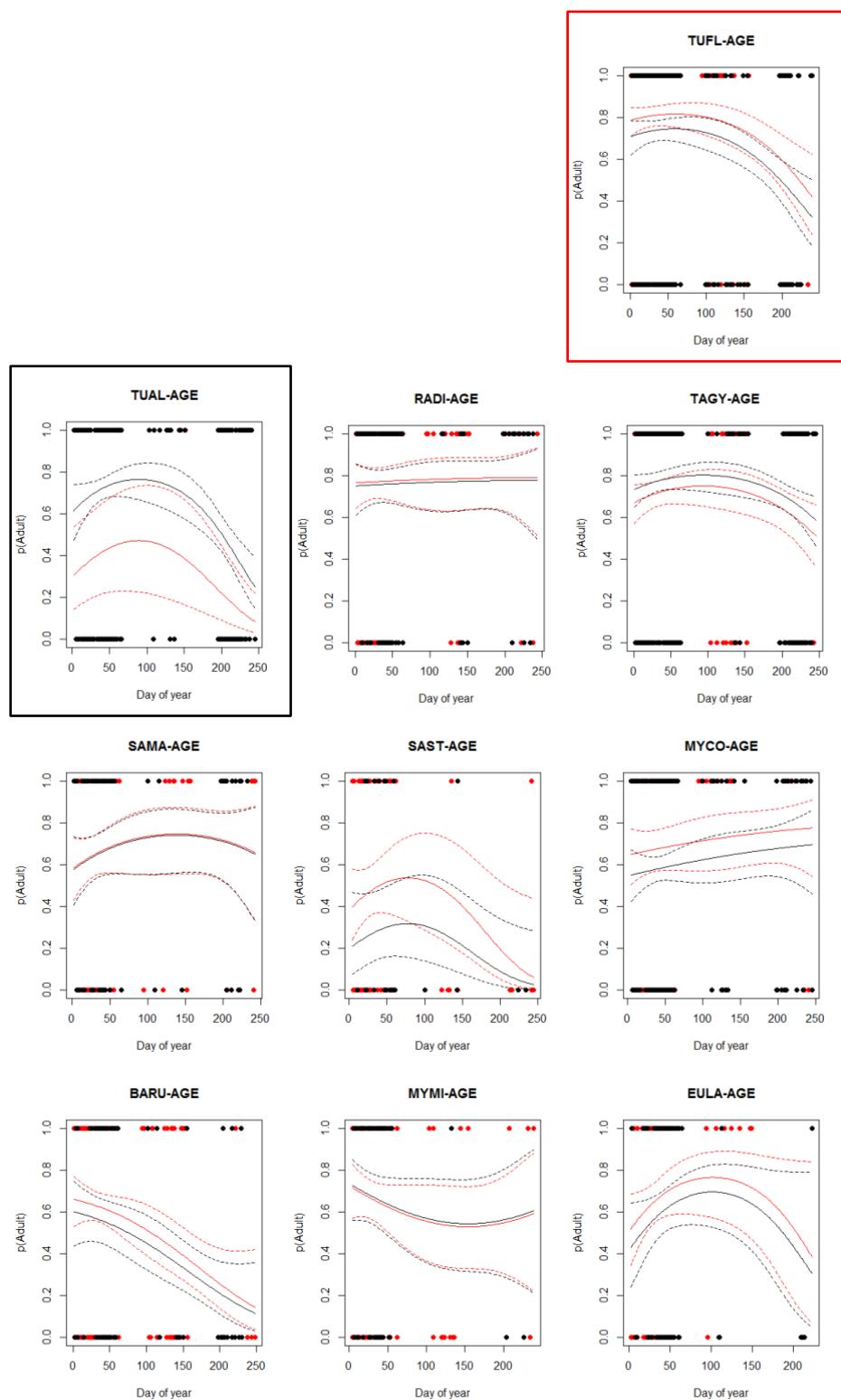
**Figure S6.** Mean and standard deviation for the coefficient of variation among estimates of abundance per occasion (calculated from output of the Jolly-Serber model).

### 2.3.5. Output from generalized linear models

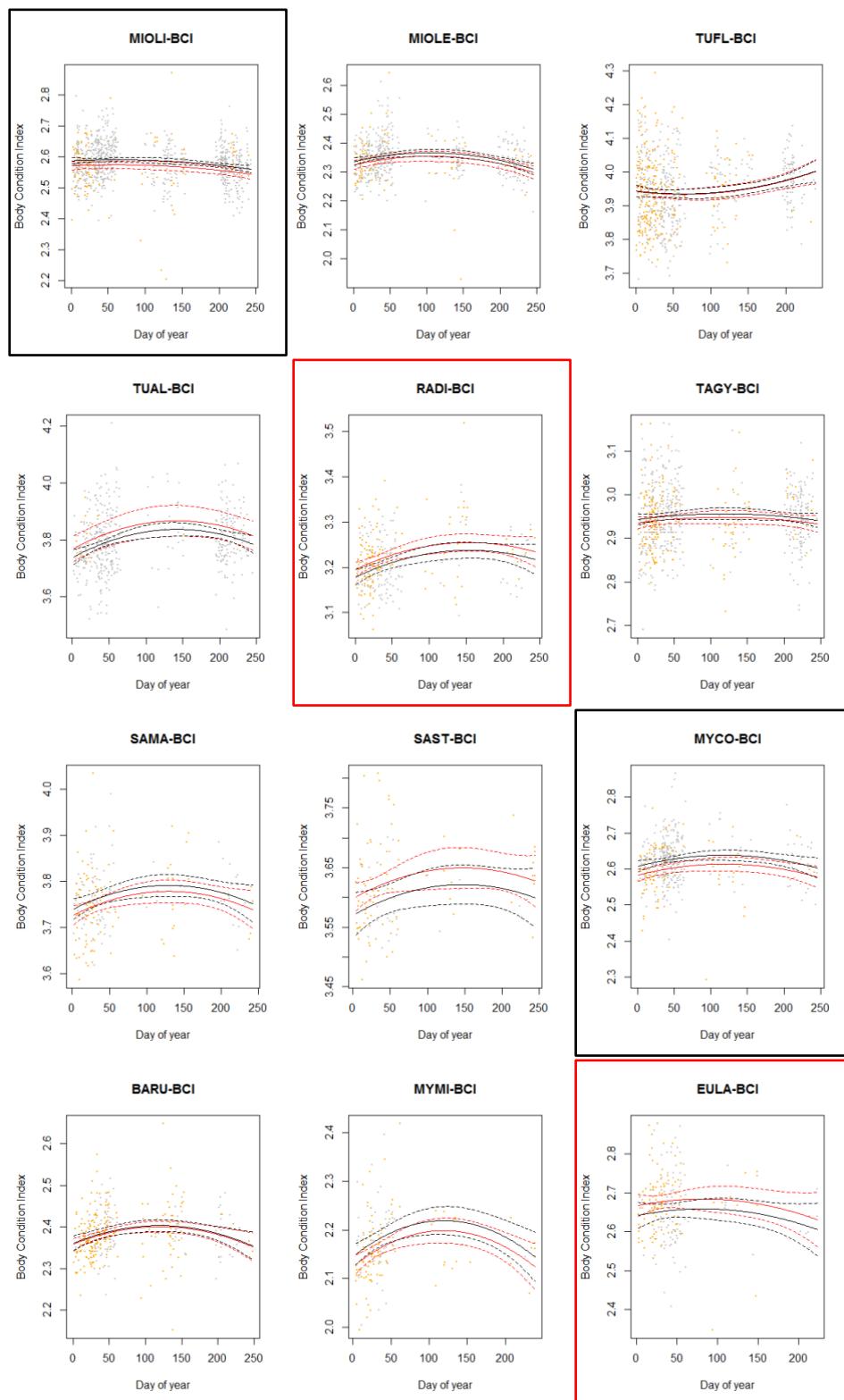
**Table S12.** Model averaging results for generalized linear models of habitat preference and performance variables (b0 to b4 stand for coefficient estimates, values in bold were significantly different from 0). For all these analyses, forest was the reference habitat.

Response	Species	b0	Est.	Err.	b1	Est.	Err.	b2	Est.	Err.	b3	Est.	Err.
Age	<i>Turdus flavipes</i>	Intercept	<b>1.22</b>	0.18	DAYS	0.39	0.36	DAYS^2	<b>-0.75</b>	0.35	HABITAT	<b>-0.4</b>	0.2
Age	<i>Turdus albicollis</i>	Intercept	-0.61	0.57	DAYS	1.36	0.7	DAYS^2	<b>-1.88</b>	0.69	HABITAT	<b>1.29</b>	0.5
Age	<i>Ramphocelus dimidiatus</i>	Intercept	<b>1.2</b>	0.16	DAYS	0.06	0.59	DAYS^2	-0.02	0.59	HABITAT	-0.07	0.29
Age	<i>Tangara gyrola</i>	Intercept	<b>0.89</b>	0.16	DAYS	0.7	0.5	DAYS^2	-0.89	0.49	HABITAT	0.29	0.2
Age	<i>Saltator maximus</i>	Intercept	<b>0.64</b>	0.17	DAYS	0.7	0.67	DAYS^2	-0.58	0.67	HABITAT	0	0.31
Age	<i>Saltator striatipectus</i>	Intercept	-0.43	0.25	DAYS	0.9	0.94	DAYS^2	-1.55	1.07	HABITAT	-0.87	0.51
Age	<i>Basileuterus rufifrons</i>	Intercept	0.18	0.14	DAYS	-0.24	0.49	DAYS^2	-0.35	0.51	HABITAT	-0.26	0.26
Age	<i>Myiothlypis conspicillata</i>	Intercept	<b>0.63</b>	0.24	DAYS	0.17	0.43	DAYS^2	-0.03	0.43	HABITAT	-0.39	0.26
Age	<i>Myioborus miniatus</i>	Intercept	<b>0.63</b>	0.17	DAYS	-0.56	0.56	DAYS^2	0.4	0.56	HABITAT	0.07	0.35
Age	<i>Euphonia laniirostris</i>	Intercept	<b>0.53</b>	0.19	DAYS	0.78	0.47	DAYS^2	-0.78	0.46	HABITAT	-0.27	0.32
Body condition	<i>Mionectes olivaceus</i>	Intercept	-0.2	0.1	DAYS	0.15	0.18	DAYS^2	-0.3	0.18	HABITAT	<b>0.25</b>	0.09
Body condition	<i>Mionectes oleagineus</i>	Intercept	-0.08	0.1	DAYS	<b>0.69</b>	0.21	DAYS^2	<b>-0.78</b>	0.21	HABITAT	0.18	0.11
Body condition	<i>Turdus flavipes</i>	Intercept	0	0.05	DAYS	-0.18	0.16	DAYS^2	<b>0.32</b>	0.16	HABITAT	0.02	0.08
Body condition	<i>Turdus albicollis</i>	Intercept	0.13	0.21	DAYS	<b>1.07</b>	0.3	DAYS^2	<b>-0.93</b>	0.3	HABITAT	-0.3	0.23
Body condition	<i>Ramphocelus dimidiatus</i>	Intercept	0.11	0.09	DAYS	<b>0.78</b>	0.24	DAYS^2	<b>-0.57</b>	0.24	HABITAT	<b>-0.27</b>	0.12
Body condition	<i>Tangara gyrola</i>	Intercept	-0.03	0.07	DAYS	0.27	0.22	DAYS^2	-0.27	0.22	HABITAT	0.11	0.08
Body condition	<i>Saltator maximus</i>	Intercept	-0.04	0.1	DAYS	<b>0.78</b>	0.29	DAYS^2	<b>-0.69</b>	0.29	HABITAT	0.19	0.14
Body condition	<i>Saltator striatipectus</i>	Intercept	0.05	0.11	DAYS	0.62	0.41	DAYS^2	-0.51	0.41	HABITAT	-0.35	0.22
Body condition	<i>Basileuterus rufifrons</i>	Intercept	0	0.06	DAYS	<b>0.77</b>	0.22	DAYS^2	<b>-0.7</b>	0.22	HABITAT	0.04	0.12
Body condition	<i>Myiothlypis conspicillata</i>	Intercept	<b>-0.31</b>	0.1	DAYS	<b>0.47</b>	0.2	DAYS^2	<b>-0.47</b>	0.19	HABITAT	<b>0.4</b>	0.12
Body condition	<i>Myioborus miniatus</i>	Intercept	-0.07	0.1	DAYS	<b>1</b>	0.27	DAYS^2	<b>-0.95</b>	0.27	HABITAT	0.31	0.16
Body condition	<i>Euphonia laniirostris</i>	Intercept	<b>0.11</b>	0.11	DAYS	0.18	0.24	DAYS^2	-0.25	0.23	HABITAT	<b>-0.29</b>	0.14
Breeding	<i>Mionectes olivaceus</i>	Intercept	<b>-5.55</b>	0.85	DAYS	<b>11.45</b>	2.28	DAYS^2	<b>-14.61</b>	3.16	HABITAT	0.2	0.45
Breeding	<i>Mionectes oleagineus</i>	Intercept	<b>-3.81</b>	0.61	DAYS	<b>9.79</b>	1.96	DAYS^2	<b>-11.73</b>	2.59	HABITAT	-0.5	0.42
Breeding	<i>Turdus flavipes</i>	Intercept	0.17	0.17	DAYS	<b>3.98</b>	0.55	DAYS^2	<b>-6.21</b>	0.85	HABITAT	<b>-0.92</b>	0.19
Breeding	<i>Turdus albicollis</i>	Intercept	<b>-2.96</b>	0.59	DAYS	<b>9.79</b>	1.29	DAYS^2	<b>-14.88</b>	2.04	HABITAT	0.07	0.64
Breeding	<i>Ramphocelus dimidiatus</i>	Intercept	<b>-1.74</b>	0.27	DAYS	<b>3.71</b>	1.01	DAYS^2	<b>-5.54</b>	1.49	HABITAT	-0.31	0.33
Breeding	<i>Tangara gyrola</i>	Intercept	-0.11	0.19	DAYS	<b>1.45</b>	0.47	DAYS^2	<b>-2.85</b>	0.54	HABITAT	<b>-0.41</b>	0.18
Breeding	<i>Saltator maximus</i>	Intercept	<b>-1.49</b>	0.29	DAYS	<b>3.13</b>	1.09	DAYS^2	<b>-4.53</b>	1.57	HABITAT	0.29	0.34
Breeding	<i>Saltator striatipectus</i>	Intercept	<b>-1.14</b>	0.32	DAYS	<b>3.17</b>	1.28	DAYS^2	<b>-3.71</b>	1.78	HABITAT	<b>-1.47</b>	0.64
Breeding	<i>Basileuterus rufifrons</i>	Intercept	<b>-2.08</b>	0.3	DAYS	<b>10.15</b>	1.38	DAYS^2	<b>-14.36</b>	1.92	HABITAT	<b>-0.97</b>	0.34
Breeding	<i>Myiothlypis conspicillata</i>	Intercept	<b>-0.86</b>	0.21	DAYS	<b>3.26</b>	0.61	DAYS^2	<b>-3.98</b>	0.83	HABITAT	-0.3	0.27
Breeding	<i>Myioborus miniatus</i>	Intercept	-0.15	0.24	DAYS	<b>5.12</b>	1.05	DAYS^2	<b>-8.15</b>	1.62	HABITAT	0.33	0.4
Breeding	<i>Euphonia laniirostris</i>	Intercept	<b>-1.07</b>	0.27	DAYS	<b>1.3</b>	0.52	DAYS^2	<b>-1.12</b>	0.51	HABITAT	<b>0.71</b>	0.33

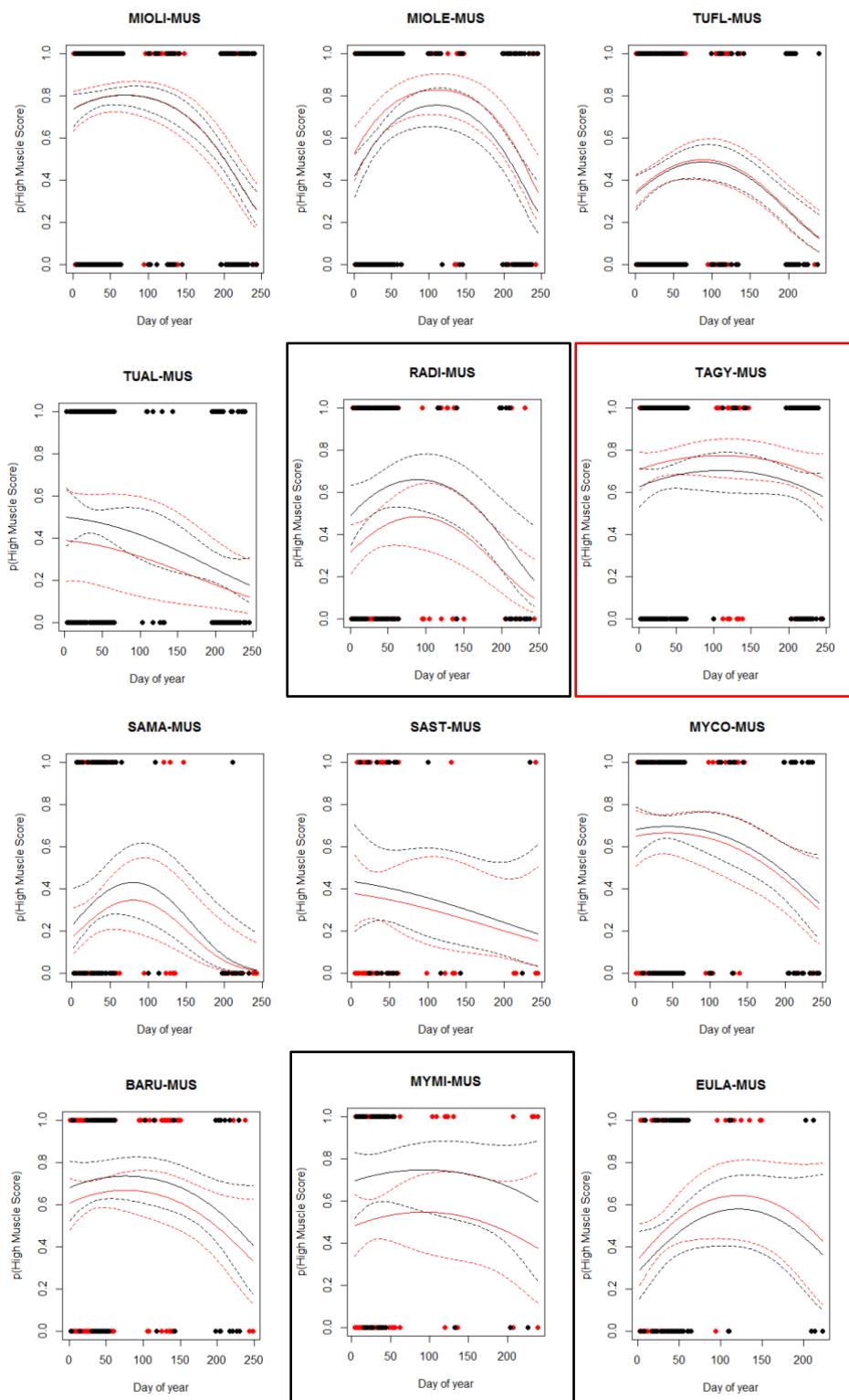
Juveniles	<i>Turdus flavipes</i>	Intercept	<b>-4.32</b>	0.6	DAYs	<b>4.56</b>	0.95	DAYs^2	<b>-3.01</b>	0.78	HABITAT	0.8	0.5
Juveniles	<i>Ramphocelus dimidiatus</i>	Intercept	<b>-3.81</b>	0.46	DAYs	-0.03	1.66	DAYs^2	0.02	1.66	HABITAT	0.02	0.83
Juveniles	<i>Tangara gyrola</i>	Intercept	<b>-4.04</b>	0.61	DAYs	<b>3.58</b>	1.15	DAYs^2	<b>-3.02</b>	1.09	HABITAT	0.99	0.58
Juveniles	<i>Saltator maximus</i>	Intercept	-15.93	15.31	DAYs	27.22	27.57	DAYs^2	-22.98	23.44	HABITAT	1.05	1.44
Juveniles	<i>Saltator striatipectus</i>	Intercept	<b>-3.26</b>	0.61	DAYs	<b>3.57</b>	1.69	DAYs^2	-3.09	1.66	HABITAT	0.78	0.88
Juveniles	<i>Basileuterus rufifrons</i>	Intercept	<b>-5.9</b>	1.28	DAYs	<b>10.08</b>	2.75	DAYs^2	<b>-8.25</b>	2.42	HABITAT	0.8	0.61
Juveniles	<i>Myiothlypis conspicillata</i>	Intercept	<b>-3.99</b>	0.52	DAYs	1.81	1.51	DAYs^2	-1.64	1.54	HABITAT	-0.34	0.85
Juveniles	<i>Euphonia laniirostris</i>	Intercept	<b>-4.49</b>	0.83	DAYs	-0.46	1.82	DAYs^2	1.06	1.52	HABITAT	0.2	1.46
Muscle	<i>Mionectes olivaceus</i>	Intercept	<b>0.76</b>	0.13	DAYs	<b>0.94</b>	0.45	DAYs^2	<b>-1.67</b>	0.45	HABITAT	-0.01	0.22
Muscle	<i>Mionectes oleagineus</i>	Intercept	<b>0.57</b>	0.25	DAYs	<b>1.98</b>	0.51	DAYs^2	<b>-2.18</b>	0.51	HABITAT	-0.44	0.24
Muscle	<i>Turdus flavipes</i>	Intercept	<b>-0.44</b>	0.1	DAYs	<b>0.94</b>	0.33	DAYs^2	<b>-1.18</b>	0.34	HABITAT	-0.04	0.17
Muscle	<i>Turdus albicollis</i>	Intercept	-0.59	0.37	DAYs	-0.01	0.68	DAYs^2	-0.49	0.69	HABITAT	0.45	0.51
Muscle	<i>Ramphocelus dimidiatus</i>	Intercept	<b>-0.5</b>	0.2	DAYs	1	0.53	DAYs^2	<b>-1.3</b>	0.55	HABITAT	<b>0.73</b>	0.26
Muscle	<i>Tangara gyrola</i>	Intercept	<b>0.96</b>	0.19	DAYs	0.53	0.51	DAYs^2	-0.61	0.51	HABITAT	<b>-0.4</b>	0.19
Muscle	<i>Saltator maximus</i>	Intercept	<b>-1.24</b>	0.24	DAYs	1.63	0.83	DAYs^2	<b>-2.34</b>	0.99	HABITAT	0.35	0.34
Muscle	<i>Saltator striatipectus</i>	Intercept	<b>-0.65</b>	0.21	DAYs	-0.04	0.85	DAYs^2	-0.29	0.88	HABITAT	0.23	0.44
Muscle	<i>Basileuterus rufifrons</i>	Intercept	<b>0.59</b>	0.14	DAYs	0.45	0.45	DAYs^2	-0.64	0.45	HABITAT	0.28	0.26
Muscle	<i>Myiothlypis conspicillata</i>	Intercept	<b>0.64</b>	0.15	DAYs	0.2	0.42	DAYs^2	-0.51	0.42	HABITAT	0.14	0.25
Muscle	<i>Myioborus miniatus</i>	Intercept	0.04	0.2	DAYs	0.32	0.57	DAYs^2	-0.4	0.58	HABITAT	<b>0.92</b>	0.35
Muscle	<i>Euphonia laniirostris</i>	Intercept	-0.15	0.17	DAYs	0.81	0.47	DAYs^2	-0.71	0.47	HABITAT	-0.19	0.3
Primary molt	<i>Mionectes olivaceus</i>	Intercept	<b>-6.56</b>	1.68	DAYs	<b>15.95</b>	4.13	DAYs^2	<b>-12.02</b>	2.96	HABITAT	-0.09	0.54
Primary molt	<i>Turdus flavipes</i>	Intercept	<b>-4.51</b>	0.56	DAYs	<b>5.12</b>	1.01	DAYs^2	<b>-3.39</b>	0.80	HABITAT	0.41	0.48
Primary molt	<i>Turdus albicollis</i>	Intercept	<b>-4.16</b>	1.19	DAYs	<b>9.75</b>	2.03	DAYs^2	<b>-7.94</b>	1.75	HABITAT	-1.24	1.50
Primary molt	<i>Ramphocelus dimidiatus</i>	Intercept	<b>-4.97</b>	1.01	DAYs	<b>7.00</b>	1.63	DAYs^2	<b>-4.52</b>	1.18	HABITAT	-0.19	0.55
Primary molt	<i>Tangara gyrola</i>	Intercept	<b>-2.10</b>	0.23	DAYs	<b>5.16</b>	0.7	DAYs^2	<b>-6.67</b>	0.94	HABITAT	0.07	0.22
Primary molt	<i>Saltator maximus</i>	Intercept	<b>-3.09</b>	0.58	DAYs	<b>4.12</b>	1.12	DAYs^2	<b>-3.17</b>	1.04	HABITAT	0.94	0.58
Primary molt	<i>Saltator striatipectus</i>	Intercept	<b>-3.33</b>	0.66	DAYs	<b>6.20</b>	1.67	DAYs^2	<b>-4.95</b>	1.47	HABITAT	-0.44	0.86
Primary molt	<i>Basileuterus rufifrons</i>	Intercept	<b>-2.82</b>	0.38	DAYs	<b>4.10</b>	0.92	DAYs^2	<b>-3.09</b>	0.88	HABITAT	<b>-1.15</b>	0.57
Primary molt	<i>Myiothlypis conspicillata</i>	Intercept	<b>-5.39</b>	1.06	DAYs	<b>6.20</b>	1.49	DAYs^2	<b>-5.04</b>	1.33	HABITAT	1.16	0.88
Primary molt	<i>Myioborus miniatus</i>	Intercept	<b>-4.27</b>	0.82	DAYs	<b>6.52</b>	1.56	DAYs^2	<b>-5.44</b>	1.58	HABITAT	-1.10	1.20



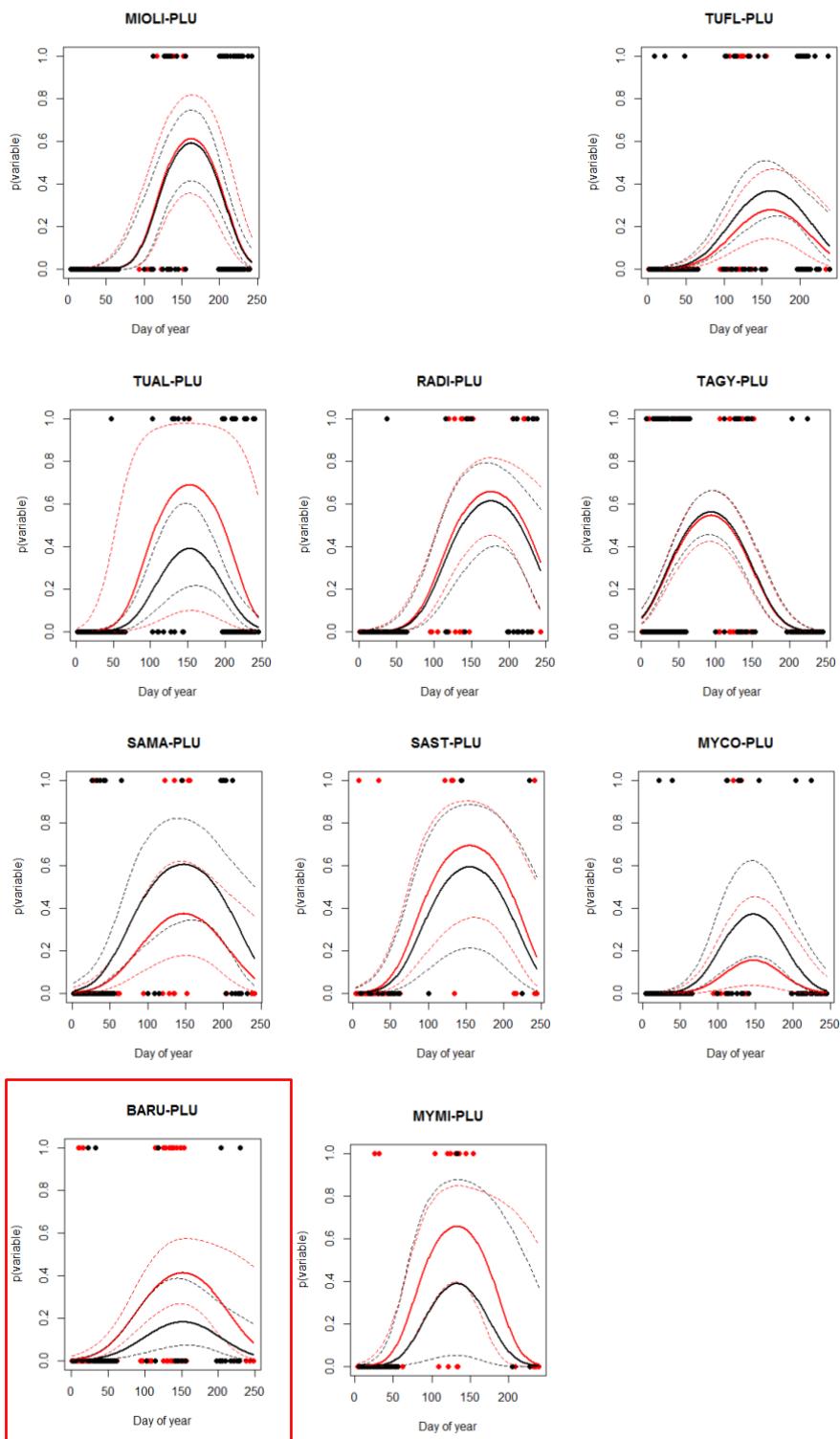
**Figure S7.** Effect of day of year (day 0: earliest date of sampling) and habitat (coffee in red and forest in black) in estimates for the probability of capturing an adult individual (over an immature one). Predicted model (solid line) and 95% confidence intervals (dotted lines) for the generalized linear model of DAY + DAY<sup>2</sup> + HABITAT with binomial error family and logit link; points show observed data.



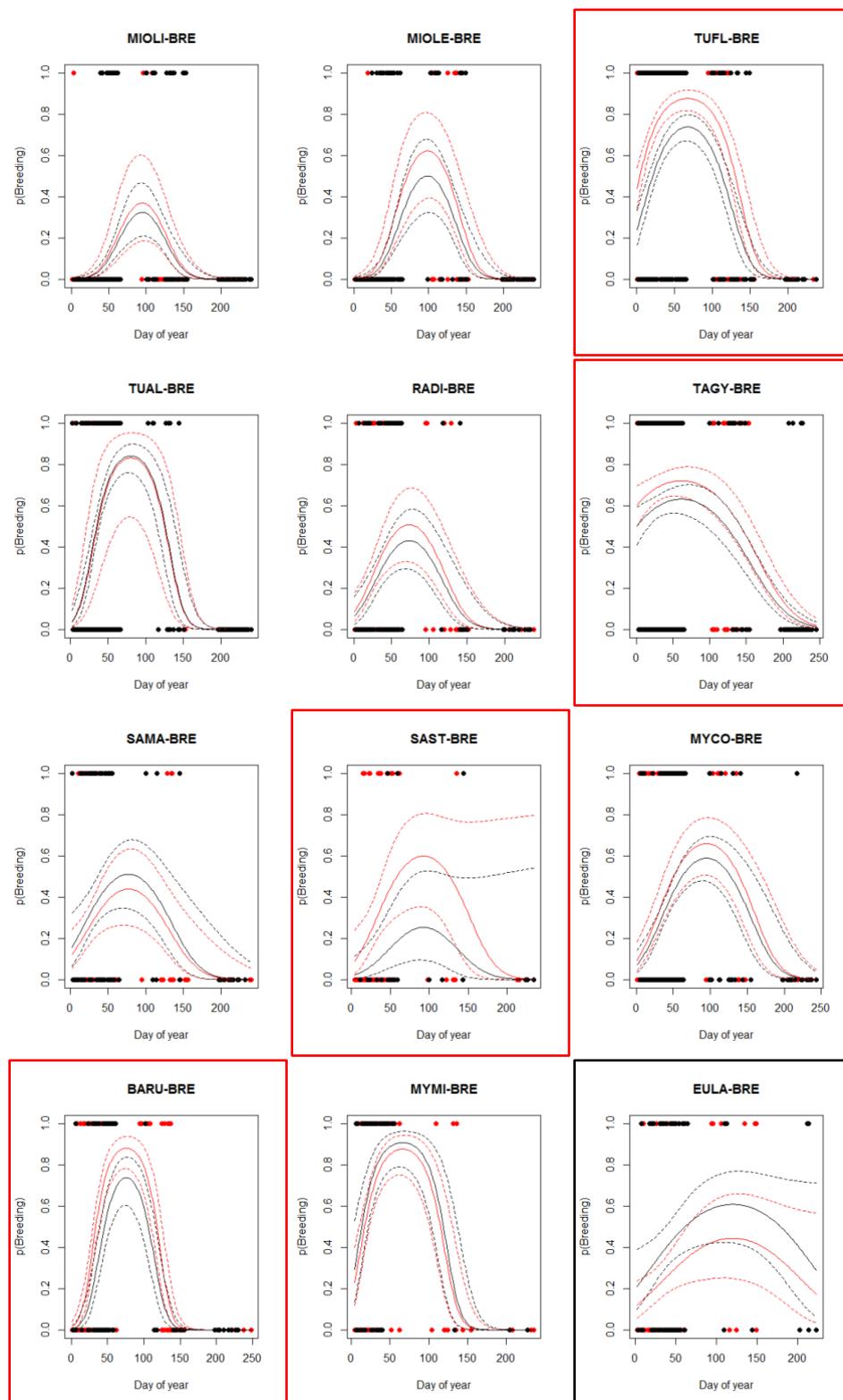
**Figure S8.** Effect of day of year (day 0: earliest date of sampling) and habitat (coffee in red and forest in black) on body condition index. Predicted model (solid line) and 95% confidence intervals (dotted lines) for the generalized linear model of  $\text{DAY} + \text{DAY}^2 + \text{HABITAT}$  with normal error family and identity link; orange points show observed data for coffee and grey points observed data for forest.



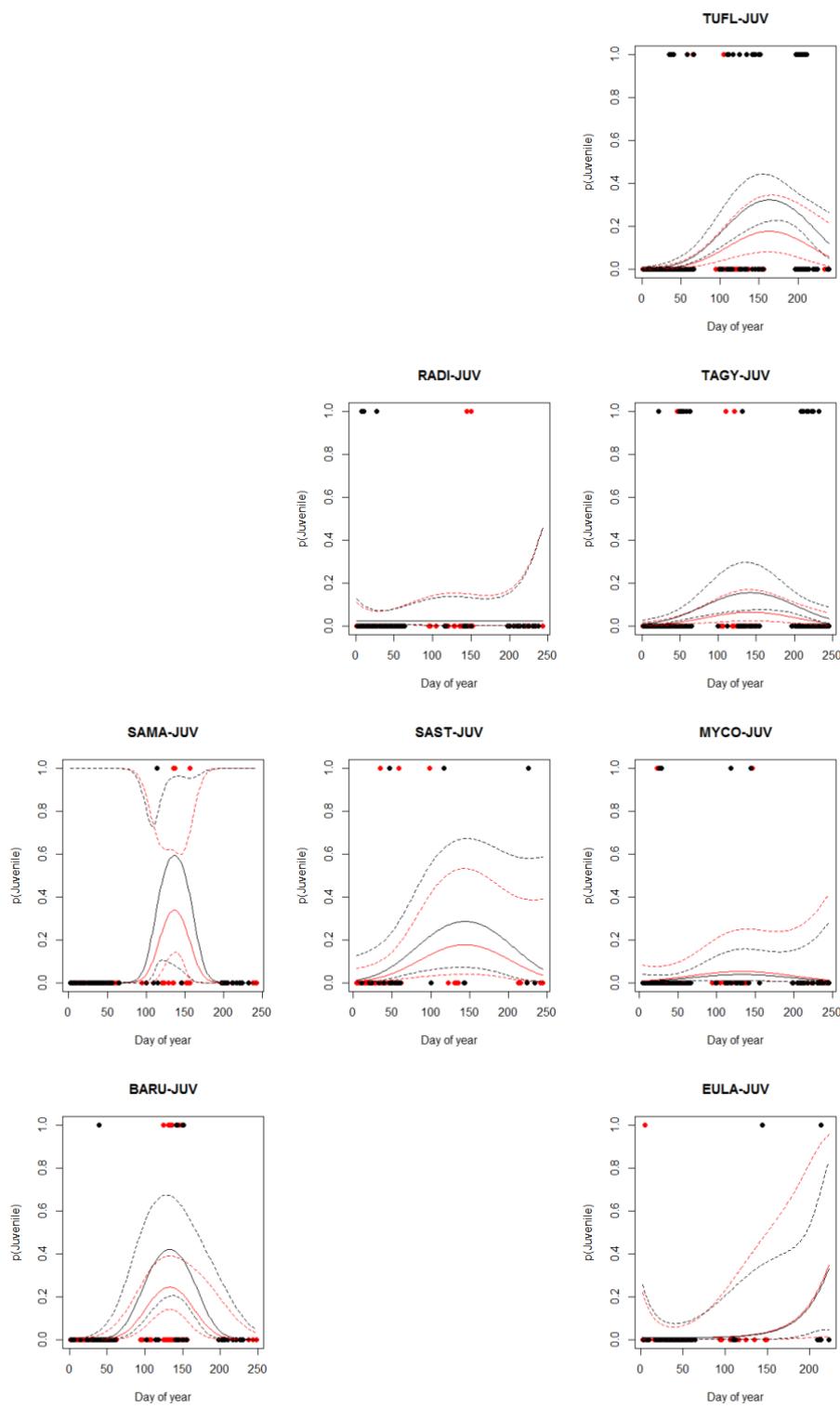
**Figure S9.** Effect of day of year (day 0: earliest date of sampling) and habitat (coffee in red and forest in black) in estimates for the probability of capturing an individual with high muscle score (over one with medium). Predicted model (solid line) and 95% confidence intervals (dotted lines) for the generalized linear model of DAY + DAY<sup>2</sup> + HABITAT with binomial error family and logit link; points show observed data.



**Figure S10.** Effect of day of year (day 0: earliest date of sampling) and habitat (coffee in red and forest in black) in estimates for the probability of capturing an individual with active primary plumage molt (over one without). Predicted model (solid line) and 95% confidence intervals (dotted lines) for the generalized linear model of DAY + DAY<sup>2</sup> + HABITAT with binomial error family and logit link; points show observed data.



**Figure S11.** Effect of day of year (day 0: earliest date of sampling) and habitat (coffee in red and forest in black) in estimates for the probability of capturing an individual in active breeding (over an inactive one). Predicted model (solid line) and 95% confidence intervals (dotted lines) for the generalized linear model of  $\text{DAY} + \text{DAY}^2 + \text{HABITAT}$  with binomial error family and logit link; points show observed data.



**Figure S12.** Effect of day of year (day 0: earliest date of sampling) and habitat (coffee in red and forest in black) in estimates for the probability of capturing a juvenile individual (over an adult or immature). Predicted model (solid line) and 95% confidence intervals (dotted lines) for the generalized linear model of DAY + DAY<sup>2</sup> + HABITAT with binomial error family and logit link; points show observed data.

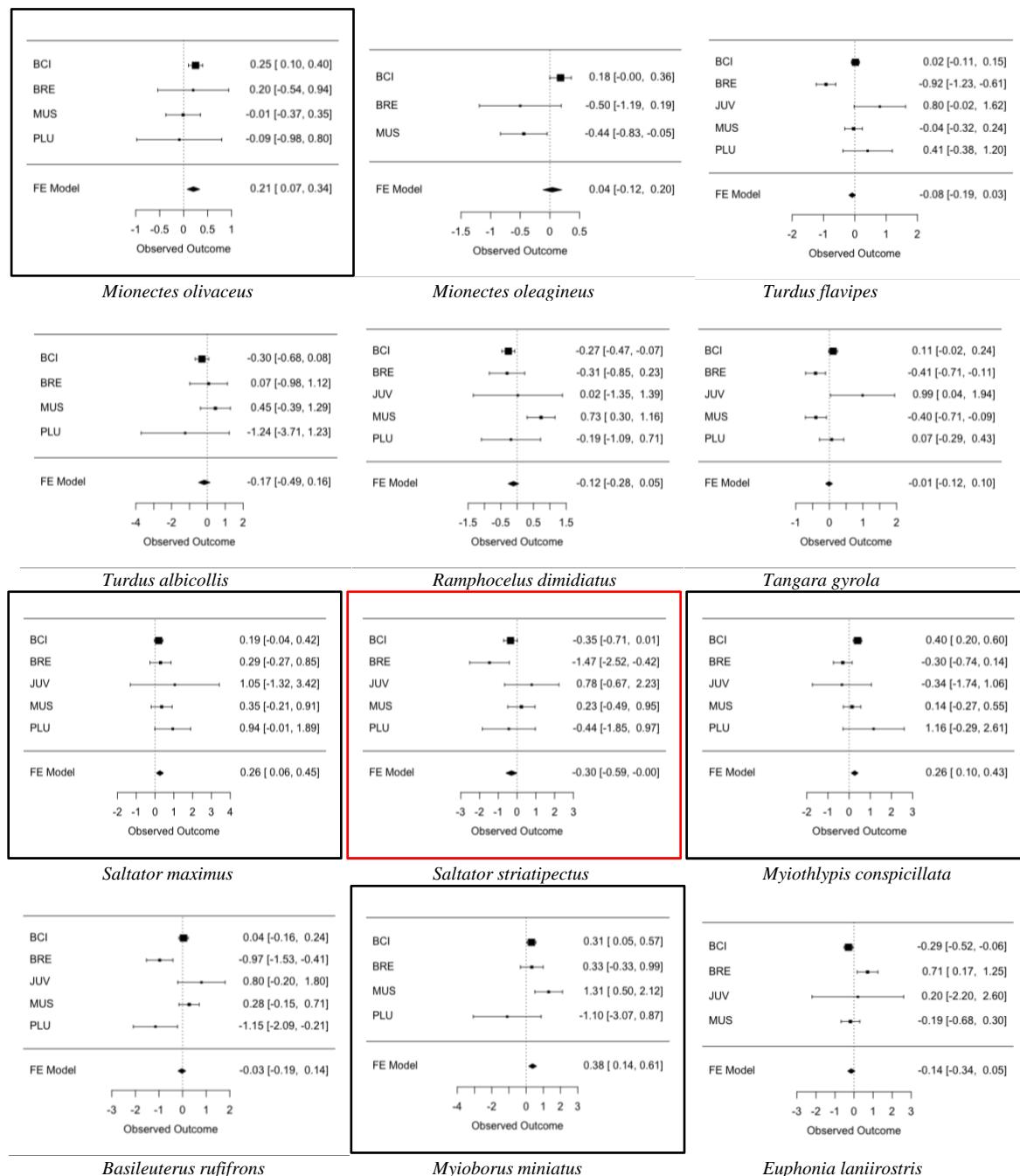
## 2.2. Level 2: Composite indexes of habitat preference and performance

### 2.2.1. Habitat preference

**Table S13.** Results for index of habitat preference.

Species	No. indicators showing Forest (F) was preferred	No. indicators showing Coffee (C) was preferred	F-C
<i>Mionectes olivaceus</i>	1	0	+1
<i>Mionectes oleagineus</i>	1	0	+1
<i>Turdus flavipes</i>	1	1	0
<i>Turdus albicollis</i>	3	0	+3
<i>Ramphocelus dimidiatus</i>	1	1	0
<i>Tangara gyrola</i>	2	0	+2
<i>Saltator maximus</i>	0	0	0
<i>Saltator striatipectus</i>	0	2	-2
<i>Myiothlypis conspicillata</i>	1	0	+1
<i>Basileuterus rufifrons</i>	1	1	0
<i>Myioborus miniatus</i>	0	1	-1
<i>Euphonia laniirostris</i>	1	1	0

## 2.2.2. Habitat performance



**Figure S13.** Summary of evidence from meta-analysis of habitat performance with five separate indicators (black frame: quality clearly higher in forest, red frame: quality clearly higher in coffee, no frame: no clear differences; variables: BCI – body condition, MUS – muscle, PLU – primary molt, BRE – breeding, JUV – juveniles; squares: mid-point represents the mean effect estimate for each study, area represents weight given to it in the model, and lines represent 95% confidence intervals (negative values indicate higher performance in coffee, positive values higher performance in forest); diamond: overall effect of habitat on individual fitness with 95% confidence intervals (fixed effects model of the means weighted by the inverse of the variance)).

### 3. Discussion

#### 3.1. Level 1: Separate indicators of habitat preference and performance

##### 3.1.1. Habitat preference

Despite abundance being the most informative indicator for preference, six species showed contrary evidence from other variables. For example, higher abundance in forest was contradicted in *T. flavipes* by a higher proportion of adults in coffee, and in *E. laniirostris* by a higher site fidelity. Other species showed no differences in abundance between habitats and yet for *M. olivaceus* (fidelity), *T. gyrola* (fidelity and variance) and *M. miniatus* (occupancy) there was evidence of preference for one habitat based on interpretation of other indicators. These results emphasize that abundance alone may lead to completely different conclusions about species-habitat relationships than using a suite of indicators of preference (Robertson & Hutto 2006). Occurrence was not a reliable variable in our study, probably because we did not sample enough sites to properly capture heterogeneity for our species. Nonetheless, being the only proposed indicator estimable via visual observations (Ruiz-Gutiérrez *et al.* 2010), further study into its potential as a proxy for preference is warranted. Differences in site fidelity and temporal variance in population size were not correlated with the proportion of individuals that were captured in both habitats within or between seasons (Table S13), giving support to our interpretations despite the fact that parameters were estimated at the population and not the individual level. Only the thrushes showed habitat segregation according to age, which may arise from the strong dominance hierarchies documented in this family (Brown *et al.* 2000; Smith 2001), and from differences in habitat use by age documented elsewhere (Cohen *et al.* 2004; Cohen & Lindell 2005; Verhulst *et al.* 2014).

**Table S14.** Percent of captured individuals which were recaptured in both habitats within and between primary sampling events.

Species	Within	Between
<i>Mionectes olivaceus</i>	2%	24%
<i>Mionectes oleagineus</i>	7%	30%
<i>Turdus flavipes</i>	4%	57%
<i>Turdus albicollis</i>	1%	4%
<i>Ramphocelus dimidiatus</i>	2%	32%
<i>Tangara gyrola</i>	1%	25%
<i>Saltator maximus</i>	1%	24%
<i>Saltator striatipectus</i>	1%	8%
<i>Myiothlypis conspicillata</i>	1%	7%
<i>Basileuterus rufifrons</i>	0%	2%
<i>Myioborus miniatus</i>	1%	4%
<i>Euphonia laniirostris</i>	8%	46%

##### 3.1.2. Habitat performance

Considering our sample sizes, we were surprised to find so many statistically unclear results for performance variables. Other than true equal-preference or quality, statistically unclear comparisons may have resulted from aggregating variables measured on different days into occasions, and at different sites into habitat types. Given the difficulty of inferring the degree to which an individual's condition is the result of occupying the habitat where it is captured (Sherry & Holmes 1996), for future analyses it may be advisable to remove floaters by ranking individuals according to how many times they were captured at a site per occasion. As with preference, the chosen variables sometimes complemented (both breeding and primary molt signaled better performance in coffee for *B. rufifrons*, while breeding and muscle signaled better performance of *T. gyrola* in forest), and sometimes contradicted (opposing patterns of body condition and muscle for *R. dimidiatus*, and of body condition and breeding for *E. laniirostris*) each other. Opposing evidence may reflect tradeoffs between different aspects of fitness or may result from

confounding variables not taken into account in our analyses (Battin 2004), and reiterates the importance of including a suite of indicators to evaluate performance. The relative frequency of juveniles in the population performed poorly in our analyses, indicating that increased effort to capture juveniles was likely needed. For four species, further research into the higher proportions of individuals showing evidence of breeding in coffee might provide insight into the complementary and supplementary use of both habitats (Dunning *et al.* 1992).

### 3.2. Level 2: Composite indexes of habitat preference and performance

#### 3.2.1. Habitat preference

Habitat selection results from a variety of individual, social, behavioral, and environmental cues interacting with each other at different scales (Jones 2001; Fletcher Jr & Sieving 2010; Schmidt, Dall & Van Gils 2010; DeCesare *et al.* 2014) and since the indicators of preference used in this study reflected different components of life history, we did not expect them to necessarily provide consistent results within species. Preference classifications did seem to coincide with species habitat descriptions in the literature (Table S1), with the three species classified as being present only in forest (*M. olivaceus*, *T. albicollis* and *M. conspicillata*) showing the expected pattern of higher preference towards forest, the four species common in forested and woody areas (*M. oleagineus*, *T. flavipes*, *T. gyrola* and *M. miniatus*) showing all three options of preference, and the five species common in woody and scrubby vegetation (*R. dimidiatus*, *S. maximus*, *S. striatipectus*, *B. rufifrons* and *E. laniirostris*) showing either higher preference for coffee (*S. striatipectus*) or equal preferences (all other species). This pattern supports the idea that despite structural similarities between forest remnants and shade coffee plantations, the latter is still perceived by forest species as a less preferred habitat, but that as species criteria for selection shift towards more open habitats, differences in preference between forest and coffee are either not there or harder to detect.

However, it is noticeable that preferring coffee over forest was not a common choice, even among the species less associated with forest. Mechanistic explanations behind this pattern could be related to shade coffee in our study site not sharing enough characteristics used as selection cues with the original habitat that species evolved to select, for example having a simpler understory structure or a lower diversity of species offering food resources. Rainfall seasonality has a strong effect on many of the variables we estimated and this introduced further complexity. Forests are typically less affected by changes in precipitation than open habitats, and this leads to seasonal shifts in foraging niche, increases in site fidelity or changes in use related to the formation of mixed-species flocks (Dietsch 2003; Schmidt 2004; Jedlicka *et al.* 2006; Colorado 2011).

A logical next step will be to determine which habitat and social characteristics are being used as selection cues (Schlaepfer, Runge & Sherman 2002; Battin 2004; Robertson & Hutto 2006; Gilroy & Sutherland 2007). Previous studies in shade coffee point to the role that keystone plant species such as the genus *Inga* (which dominated shade at our study site) have in providing seasonal resources, to the importance of epiphytes for foraging and breeding, and to the presence of scattered trees in the landscape to attract habitat-sensitive species (Johnson 2000; Sekercioglu *et al.* 2007; Cruz-Angón, Sillett & Greenberg 2008). Nonetheless, to get a complete picture of habitat and landscape effects on species distribution and persistence, we need long-term demographic studies that allow the distinction of the effects that such variables can have on preference, use and/or performance.

### 3.2.2. Habitat performance

Ideally, habitat quality assessments should take into account survival and reproductive success, as well as individual condition, and be able to discriminate the effects of density-dependence from other causes of variation (Pulliam & Danielson 1991; Vickery, Hunter & Wells 1992; Donovan & Thompson 2001; Burel & Baudry 2005; Johnson *et al.* 2006). Although we did not have enough information to estimate these vital rates, the variables we chose for performance are all factors that could ultimately affect survival and reproduction. Furthermore, they reflect performance at shorter intervals than population-wide estimates, which can be more appropriate to answer our questions given the larger than expected home ranges we found for our study species (evidenced by very low recapture rates and by many individuals being captured in different sites and/or habitats within a sampling occasion – Table S14), and that even within the same habitat patch, wanderer and sedentary birds can experience different mortality rates (Rappole, Ramos & Winker 1989).

Differences in performance were harder to find, but two of the three forest species showed higher performance in forest (*M. olivaceus* and *M. conspicillata*), while one showed no consistent differences in performance (*T. albicollis*). Following the expected pattern for the woody area species, three showed no differences in performance and one (*M. miniatus*) showed higher performance in forest. Lastly, of the scrubby area species three showed no differences, only one had evidence of better performance in coffee (*S. striatipectus*) and surprisingly one had evidence of better performance in forest (*S. maximus*). Our findings seem to contradict previous studies in tropical agroecosystems which found that resident birds do not commute regularly to forest (Sekercioglu *et al.* 2007) but coincide with evidence for better habitat quality in forest than plantations for wintering Neotropical migrants (Bakermans *et al.* 2009; Chandler & King 2011; Bayly *et al.* 2016).

Other than methodological issues, better performance in forest for species with a range of habitat associations once again points to characteristics in the coffee plantations in our study site that may be making it a lower quality habitat than existing forest remnants. Caution should be exercised with habitat quality classifications, since the “best-quality habitat” may vary across seasons, years and individual life stages (Lindell & Smith 2003; Cohen & Lindell 2004). We need more information to understand if this apparent forest dependence is related to issues of habitat supplementation and complementation (Dunning, Danielson & Pulliam 1992), seasonality in resources and predation (Wunderle Jr & Latta 1998; Dietsch 2003; Shochat *et al.* 2005), changes in community composition brought about by the presence of migrants (Greenberg, Reitsma & Angon 1996; Jedlicka *et al.* 2006), or to varying levels of landscape heterogeneity (Tscharntke *et al.* 2012), to name just a few possible mechanisms.

## 4. References

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## Habitat selection in transformed landscapes and the role of forest remnants and shade coffee in the conservation of resident birds.

Sánchez-Clavijo, L.M., Bayly, N.J., and Quintana-Ascencio P.F. (2019). *Journal of Animal Ecology* **V**(i): pp.

**Supporting Information - Part 2:** R code used for data analysis, following the order and structure of the manuscript.

### 1. Level 1: Separate indicators of habitat preference and performance

#### 1.1. Habitat preference

##### 1.1.1. Occurrence

```
### Occupancy model with effect of habitat on p and quadratic altitude & habitat on psi
### Based on code available at
### http://www.vogelwarte.ch/de/projekte/publikationen/bpa/code-for-running-bpa-using-jags.html
### Kéry & Schaub 2012 Bayesian population analysis using WinBUGS: a hierarchical perspective.

## Part 1. Get ready

setwd("")
data<-read.table("",header=TRUE)
y<-as.matrix(data[,5:20]) #Select columns with detected (1) / not detected (0) data
habitat<-data$hab
alt<-as.vector(scale(data$altitude))
library(jagsUI)

## Part 2. Set up and run Bayesian model in JAGS

cat("
  model {

    # Priors
    alpha.psi ~ dnorm(0,0.01)
    beta1.psi ~ dnorm(0,0.01)
    beta2.psi ~ dnorm(0,0.01)
    beta4.psi ~ dnorm(0,0.01)
    alpha.p ~ dnorm(0,0.01)
    beta1.p ~ dnorm(0,0.01)

    # Likelihood

    # Ecological model for true occurrence
    for (i in 1:R) {
      z[i] ~ dbern(psi[i])
      psi[i] <- 1 / (1 + exp(-lpsi.lim[i]))
      lpsi.lim[i] <- min(999, max(-999, lpsi[i]))
      lpsi[i] <- alpha.psi + beta1.psi*habitat[i] + beta2.psi*alt[i] + beta4.psi*pow(alt[i],2)

    # Observation model for replicated detection/non-detection observations
    for (j in 1:T){
      y[i,j] ~ dbern(mu.p[i,j])
      mu.p[i,j] <- z[i] * p[i,j]
      p[i,j] <- 1 / (1+exp(-lp.lim[i,j]))
      lp.lim[i,j] <- min(999, max(-999, lp[i,j]))
      lp[i,j] <- alpha.p + beta1.p*habitat[i]
    } #
  } #

  # Derived quantities
  occ.fs <- sum(z[]) # Number of occupied sites
}
```

```

for (s in 1:9){
  odds1[s] <- alpha.psi + beta1.psi*habitat[s] + beta2.psi*alt[s] + beta4.psi*pow(alt[s],2)
  psi.site[s] <- exp(odds1[s]) / (1+exp(odds1[s]))
} #site

for (i in 1:R){
  for (j in 1:T){
    odds2[i,j] <- alpha.p + beta1.p*habitat[i]
    pij[i,j] <- exp(odds2[i,j]) / (1+exp(odds2[i,j]))
  } #occasion
} #site

}",file="name.txt")

data<-list(y=y,R=nrow(y),T=ncol(y),habitat=habitat,alt=alt)
zst<-apply(y,1,max,na.rm=TRUE)
inits<-function(){list(z=zst,alpha.p=rnorm(1,-3,3),alpha.psi=rnorm(1,-3,3))}
params<-c("alpha.psi","beta1.psi","beta2.psi","beta4.psi","psi.site","alpha.p","beta1.p","occ.fs")
ni<-120000
nb<-20000
nt<-100
nc<-3

m=jags(data,inits,params,model="name.txt",n.chains=nc,n.thin=nt,n.iter=ni,n.burnin=nb);m

## Note: output for psi for all species was manually put together into a single file

## Part 3. Build graphs

d<-read.table("occupancy.txt",header=T)
d<-data[data$SPP=="",] # Select species to make graph
attach(d)

plot(ALT,psi_m,type="n",ylab="Occupancy",xlab="Elevation",main="Turdus flavipes",ylim=c(0.95,1.0))
points(ALT[HABITAT=="COFFEE"],psi_m[HABITAT=="COFFEE"],col="red",pch=16)
points(ALT[HABITAT=="FOREST"],psi_m[HABITAT=="FOREST"],col="black",pch=16)
ALT2<-ALT^2
m<-lm(psi_m~ALT+ALT2+HABITAT)
x<-seq(min(ALT),max(ALT),1)
one<-rep("COFFEE",length(x))
two<-rep("FOREST",length(x))
y1<-predict(m,list(ALT=x,ALT2=x^2,HABITAT=one),interval=c("confidence"),level=0.95,type="response")
y2<-predict(m,list(ALT=x,ALT2=x^2,HABITAT=two),interval=c("confidence"),level=0.95,type="response")
lines(x,y1[,1],col="red")
lines(x,y1[,2],col="red",lty=2)
lines(x,y1[,3],col="red",lty=2)
lines(x,y2[,1],col="black")
lines(x,y2[,2],col="black",lty=2)
lines(x,y2[,3],col="black",lty=2)

```

### 1.1.2. Abundance

```

### Closed population model with effect of quadratic effort on p
### Based on code available at
### http://www.vogelwarte.ch/de/projekte/publikationen/bpa/code-for-running-bpa-using-jags.html
### Kéry & Schaub 2012 Bayesian population analysis using WinBUGS: a hierarchical perspective.
### Forest version (code for coffee is exactly the same)
### Used for species with low recapture rates

## Part 1. Get ready

setwd("")"
CH<-read.table("",header=F)
CH<-as.matrix(CH)
nz<-2000 # Set number of augmented individuals
CH.aug<-rbind(CH,matrix(0,ncol=dim(CH)[2],nrow=nz)) # Bind real data with augmented individuals

```

```

eff <- c(-0.6733507,-0.1116703,-0.8040866,1.0728217,-1.0522429,-0.3574055,1.7222646,1.1242687,-1.0855321,0.2690376,-1.0558744,0.9517698) #
Vector of effort per occasion (scaled mist net hours)
library(jagsUI)

## Part 2. Set up and run Bayesian model in JAGS

cat("
  model {

    # Priors and constraints
    psi ~ dunif(0,1)
    for (t in 1:n.occ){
      logit(p[t]) <- mp + beta1*x[t] + beta2*pow(x[t],2)
      p.est[t] <- 1 / (1+exp(-mp-beta1*x[t]-beta2*pow(x[t],2)))
    } #t
    mp ~ dnorm(0,0.001)
    mean.p <- 1 / (1+exp(-mp))
    beta1 ~ dnorm(0,0.001)(-10,10)
    beta2 ~ dnorm(0,0.001)(-10,10)

    # Likelihood
    for (i in 1:M){
      z[i] ~ dbern(psi)
      for (t in 1:n.occ){
        y[i,t] ~ dbern(mu[i,t])
        mu[i,t] <- z[i] * p[t]
      } #t
    } #i

    # Derived quantities
    N <- sum(z[])
  }
)",file="name.txt")

data <- list(y=CH.aug,M=dim(CH.aug)[1],n.occ=dim(CH.aug)[2],x=eff)
inits <- function(){list(z=rep(1,dim(CH.aug)[1]),mp=rnorm(1),beta1=runif(1,-5,5),beta2=runif(1,-5,5))}
params <- c("N","p.est","mean.p","beta1","beta2","psi")
ni <- 12000
nb <- 2000
nt <- 10
nc <- 3

m=jags(data=inits,inits=inits,parameters=params,model="name.txt",n.chains=nc,n.iter=ni,n.burnin=nb,n.thin=nt,parallel=TRUE)

## Part 3. Collect output

write.table(m$summary,"C:/summ.txt",sep="\t") #Saves summary statistics for all iterations
write.table(m$sims.list,"C:/sims.txt",sep="\t") #Saves complete results for each iteration
pdf("C:/plots.pdf") #Next three lines save diagnostic plots in pdf files
plot(m)
dev.off()

## Part 4. Build graphs

library(ggplot2)
C_sim<-read.table("",header=T) #Simulation file for coffee
F_sim<-read.table("",header=T) #Simulation file for forest
Ncof<-C_sim$N
Nfor<-F_sim$N
Nsup<-c(Ncof,Nfor)
type<-c(rep(1,length(Ncof)),rep(2,length(Nfor)))
p.graph<-cbind(type,Nsup)
p.graph<-as.data.frame(p.graph)
p.graph$type<-factor(p.graph$type)
levels(p.graph$type)<-c("Coffee","Forest")
ggplot(p.graph,aes(p.graph$Nsup,fill=p.graph$type))+
  geom_histogram(position="dodge",binwidth=20)+
  scale_fill_manual(values=c("red","black"))+
  xlab("Total Abundance")+

```

```

ylab("Frequency")+
ggtitle("SPECIES")+
xlim(min(Nsup),max(Nsup))+
theme_classic()+
geom_vline(xintercept=min(Nsup))+
geom_hline(yintercept=0)+
theme(axis.text=element_text(size=10),axis.title=element_text(size=12))+  

theme(legend.position="none")

```

```

### Jolly-Seber model with effect of quadratic effort on p and random variation of phi per occasion
### Parameterized as a multistate model based on code available at
### http://www.vogelwarte.ch/de/projekte/publikationen/bpa/code-for-running-bpa-using-jags.html
### Kéry & Schaub 2012 Bayesian population analysis using WinBUGS: a hierarchical perspective.
### Forest version (code for coffee is exactly the same)
### Used for species with high recapture rates

### Parameters
# phi: survival probability
# gamma: removal entry probability
# p1: capture probability in coffee
# p2: capture probability in forest

### States (S)
# 1 not yet entered
# 2 alive
# 3 dead

### Observations (O)
# 1 captured
# 2 not captured

## Part 1. Get ready

setwd("")  

nz<-2000 # Set number of augmented individuals  

CH<-read.table("",header=F)  

V0<-rep(0,dim(CH)[1])  

CH.du<-cbind(V0,CH)  

mat.n<-matrix(0,ncol=dim(CH.du)[2],nrow=nz) # Bind real data with augmented individuals  

colnames(mat.n)<-colnames(CH.du)  

CH.msc<-rbind(CH.du,mat.n)  

CH.ms[CH.ms==0]<-2  

eff_forest <- c(1144,2072,928,4029,518,1666,5102,4114,463,2701,512,3829) # Vector of mist net hours  

e_for <- as.vector(scale(eff_forest))  

eff <- c(0,e_for)  

library(jagsUI)

## Part 2. Set up and run Bayesian model in JAGS

cat("model {  

# Priors and constraints  

for (t in 1:(n.occasions-1)){  

  gamma[t] ~ dunif(0,1)  

  phi[t] ~ dunif(0,1)  

} #  

alpha.p ~ dnorm(0,0.001)  

beta1.p ~ dnorm(0,0.001)  

beta2.p ~ dnorm(0,0.001)  

# Define state-transition and observation matrices  

for (j in 1:M){  

  # Define probabilities of state S(t+1) given S(t)  

  for (t in 1:(n.occasions-1)){  

    ps[1,j,t,1] <- 1-gamma[t]  

    ps[1,j,t,2] <- gamma[t]  

    ps[1,j,t,3] <- 0
  }
}

```

```

ps[2,j,t,1] <- 0
ps[2,j,t,2] <- phi[t]
ps[2,j,t,3] <- 1-phi[t]
ps[3,j,t,1] <- 0
ps[3,j,t,2] <- 0
ps[3,j,t,3] <- 1

# Define probabilities of O(t) given S(t)
po[1,j,t,1] <- 0
po[1,j,t,2] <- 1
po[2,j,t,1] <- p[t]
po[2,j,t,2] <- 1-p[t]
po[3,j,t,1] <- 0
po[3,j,t,2] <- 1

} #t
} #i

# Likelihood

for (t in 1:n.occasions){
lin.p[t] <- alpha.p + beta1.p * eff[t] + beta2.p * pow(eff[t],2)
p[t] <- exp(lin.p[t]) / (1+exp(lin.p[t]))
} #t

for (i in 1:M){

# Define latent state at first occasion
z[i,1] <- 1

for (t in 2:n.occasions){

# State process: draw S(t) given S(t-1)
z[i,t] ~ dcat(ps[z[i,t-1], i, t-1,])

# Observation process: draw O(t) given S(t)
y[i,t] ~ dcat(po[z[i,t], i, t-1,])

} #t
} #i

# Calculate derived population parameters

for (t in 1:(n.occasions-1)){
qgamma[t] <- 1-gamma[t]
} #t

cprob[1] <- gamma[1]

for (t in 2:(n.occasions-1)){
cprob[t] <- gamma[t] * prod(qgamma[1:(t-1)])
} #t

psi <- sum(cprob[])

for (t in 1:(n.occasions-1)){
b[t] <- cprob[t] / psi
} #t

for (i in 1:M){
  for (t in 2:n.occasions){
    al[i,t-1] <- equals(z[i,t], 2)
  } #
  for (t in 1:(n.occasions-1)){
    d[i,t] <- equals(z[i,t]-al[i,t],0)
  } #
  alive[i] <- sum(al[i,])
} #i
}

```

```

for (t in 1:(n.occasions-1)){
  N[t] <- sum(al[,t])
  B[t] <- sum(d[,t])
} #t

for (i in 1:M){
  w[i] <- 1-equals(alive[i],0)
} #i

Nforest <- sum(w[])

} #model
", file="name.txt")

data <- list(y=CH.ms,n.occasions=dim(CH.ms)[2],M=dim(CH.ms)[1],eff=eff)

js.multistate.init <- function(ch,nz){
  ch[ch==2] <- NA
  state <- ch
  for (i in 1:nrow(ch)){
    n1 <- min(which(ch[i,]==1))
    n2 <- max(which(ch[i,]==1))
    state[i,n1:n2] <- 2
  }
  state[state==0] <- NA
  get.first <- function(x) min(which(!is.na(x)))
  get.last <- function(x) max(which(!is.na(x)))
  f <- apply(state, 1, get.first)
  l <- apply(state, 1, get.last)
  for (i in 1:nrow(ch)){
    state[i,1:f[i]] <- 1
    if(l[i]!=ncol(ch)) state[i, (l[i]+1):ncol(ch)] <- 3
    state[i, f[i]] <- 2
  }
  m.z <- matrix(1, ncol = ncol(ch), nrow = nz)
  colnames(m.z) <- colnames(state)
  state <- rbind(state,m.z)
  state <- as.matrix(state)
  state[,1] <- NA
  return(state)
}

inits <- function(){list(z = js.multistate.init(CH.du,nz),
  alpha.p=runif(1,-1,1),beta1.p=runif(1,-1,1),beta2.p=runif(1,-1,1),
  phi=runif(12,0,1))}

params <- c("Nforest","p","alpha.p","beta1.p","beta2.p","phi","psi","b","B","N","gamma")

ni <- 12000
nb <- 2000
nt <- 10
nc <- 3

Jm=jags(data=data,inits=inits,parameters=params,model="name.txt",n.chains=nc,n.iter=ni,n.burnin=nb,n.thin=nt,parallel=TRUE)

## Part 3. Collect output

write.table(m$summary," C:/summ.txt",sep="\t") #Saves summary statistics for all iterations
write.table(m$sims.list," C:/summ.txt",sep="\t") #Saves complete results for each iteration
pdf("C:/plots.pdf") #Next three lines save diagnostic plots in pdf files
plot(m)
dev.off()

## Part 4. Build graphs

library(ggplot2)
C_sim<-read.table("",header=T) #Simulation file for coffee
F_sim<-read.table("",header=T) #Simulation file for forest
Ncof<-C_sim$Ncoffee

```

```

Nfor<-F_sim$Nforest
Nsup<-c(Ncof,Nfor)
type<-c(rep(1,length(Ncof)),rep(2,length(Nfor)))
p.graph<-cbind(type,Nsup)
p.graph<-as.data.frame(p.graph)
p.graph$type<-factor(p.graph$type)
levels(p.graph$type)<-c("Coffee","Forest")
ggplot(p.graph,aes(p.graph$Nsup,fill=p.graph$type))+
  geom_histogram(position="dodge",binwidth=5)+
  scale_fill_manual(values=c("red","black"))+
  xlab("Total Abundance")+
  ylab("Frequency")+
  ggtitle("MYCO")+
  xlim(min(Nsup),max(Nsup))+
  theme_classic()+
  geom_vline(xintercept=min(Nsup))+
  geom_hline(yintercept=0)+
  theme(axis.text=element_text(size=10),axis.title=element_text(size=12))+
  theme(legend.position="none")

```

### 1.1.3. Fidelity

```

### Cormack-Jolly-Seber model with effect of quadratic effort on p and fixed phi
### Based on code available at
### http://www.vogelwarte.ch/de/projekte/publikationen/bpa/code-for-running-bpa-using-jags.html
### Kéry & Schaub 2012 Bayesian population analysis using WinBUGS: a hierarchical perspective.
### Forest version (code for coffee is exactly the same)
### Used for species with low recapture rates

## Part 1. Get ready

setwd("")
CH<-read.table("",header=F)
CH<-as.matrix(CH)
get.first<-function(x)min(which(x!=0))
f<-apply(CH,1,get.first)
eff <- c(-0.6733507,-0.1116703,-0.8040866,1.0728217,-1.0522429,-0.3574055,1.7222646,1.1242687,-1.0855321,0.2690376,-1.0558744) # Vector of
effort per occasion (scaled mist net hours)
library(jagsUI)

## Part 2. Set up and run Bayesian model in JAGS

cat("model {

  # Priors and constraints
  for (i in 1:nind){
    for (t in f[i]:(n.occasions-1)){
      phi[i,t] <- phi.occ[t]
      logit(p[i,t]) <- mu + beta1*x[t] + beta2*pow(x[t],2)
    }
  }
  for (t in 1:(n.occasions-1)){
    phi.occ[t] ~ dunif(0,1)
    p.est[t] <- 1 / (1+exp(-mu-beta1*x[t]-beta2*pow(x[t],2)))
  }
  mu ~ dnorm(0,0.001)
  mean.p <- 1 / (1+exp(-mu))
  beta1 ~ dnorm(0,0.001)(-10,10)
  beta2 ~ dnorm(0,0.001)(-10,10)

  # Likelihood
  for (i in 1:nind){

    # Define latent state at first capture
    z[i,f[i]] <-1
    for (t in (f[i]+1):n.occasions){

      # State process

```

```

z[i,t] ~dbern(mu1[i,t])
mu1[i,t]<-phi[i,t-1] *z[i,t-1]

# Observation process
y[i,t] ~dbern(mu2[i,t])
mu2[i,t]<-p[i,t-1]*z[i,t]
} #
} #i

}",file="name.txt")

known.state.cjs <- function(ch){
  state <- ch
  for (i in 1:dim(ch)[1]){
    n1 <- min(which(ch[i,]==1))
    n2 <- max(which(ch[i,]==1))
    state[i,n1:n2] <- 1
    state[i,n1] <- NA
  }
  state[state==0] <- NA
  return(state)
}

data<-list(y=CH,f=f,nind=dim(CH)[1],n.occasions=dim(CH)[2],z=known.state.cjs(CH),x=eff)

cjs.init.z <- function(ch,f){
  for (i in 1:dim(ch)[1]){
    if (sum(ch[i,])==1) next
    n2 <- max(which(ch[i,]==1))
    ch[i,f[i]:n2] <- NA
  }
  for (i in 1:dim(ch)[1]){
    ch[i,1:f[i]] <- NA
  }
  return(ch)
}

inits <- function(){list(z=cjs.init.z(CH,f),mu=rnorm(1),beta1=runif(-5,5),beta2=runif(1,-5,5),phi.occ=runif(11,0,1))}
params<-c("phi.occ","p.est","mean.p","beta1","beta2")
ni <- 12
nb <- 2
nt <- 1
nc <- 3

m=jags(data=data,inits=inits,parameters=params,model="name.txt",n.chains=nc,n.iter=ni,n.burnin=nb,n.thin=nt,parallel=TRUE)

## Part 3. Collect output

write.table(m$summary,"C:/summ.txt",sep="\t") #Saves summary statistics for all iterations
write.table(m$sims.list,"C:/sims.txt",sep="\t") #Saves complete results for each iteration
pdf("C:/plots.pdf") #Next three lines save diagnostic plots in pdf files
plot(m)
dev.off()

## Part 4. Calculate summary statistics for phi and build graphs

sim<-read.table("",header=T) #Simulation file for species: habitat of interest
n.iter<-dim(sim)[1]
phi_m<-as.vector(rep(0,n.iter))
phi_v<-as.vector(rep(0,n.iter))
phi_sd<-as.vector(rep(0,n.iter))
phi_cv<-as.vector(rep(0,n.iter))
for(i in 1:n.iter){
  phi_m[i]<-mean(as.numeric(sim[i,20:30])) #20:30 are the columns in which phi [1-11] were recorded
  phi_v[i]<-var(as.numeric(sim[i,20:30]))
  phi_sd[i]<-sd(as.numeric(sim[i,20:30]))
  phi_cv[i]<-phi_sd[i]/phi_m[i]
}
phi.F<-matrix(0,2,4)

```

```

phi.F[1,1]<-round(mean(phi_m),4)
phi.F[1,2]<-round(mean(phi_v),4)
phi.F[1,3]<-round(mean(phi_sd),4)
phi.F[1,4]<-round(mean(phi_cv),4)
phi.F[2,1]<-round(var(phi_m),4)
phi.F[2,2]<-round(var(phi_v),4)
phi.F[2,3]<-round(var(phi_sd),4)
phi.F[2,4]<-round(var(phi_cv),4)

library(ggplot2)
C_sim<-read.table("",header=T) #Simulation file for coffee
F_sim<-read.table("",header=T) #Simulation file for forest

Cphi<-as.vector(rep(0,n.iter))
for(i in 1:n.iter){
  Cphi[i]<-mean(as.numeric(C_sim[i,20:30]))
}

Fphi<-as.vector(rep(0,n.iter))
for(i in 1:n.iter){
  Fphi[i]<-mean(as.numeric(F_sim[i,20:30]))
}

phi2<-c(Cphi,Fphi)
type<-c(rep(1,length(Cphi)),rep(2,length(Fphi)))
p.graph<-cbind(type,phi2)
p.graph<-as.data.frame(p.graph)
p.graph$type<-factor(p.graph$type)
levels(p.graph$type)<-c("Coffee","Forest")
ggplot(p.graph,aes(p.graph$phi2,color=p.graph$type))+ 
  geom_density()+
  scale_color_manual(values=c("red","black"))+
  xlab("Survival")+
  ylab("Density")+
  ggtitle("MYCO")+
  xlim(0,1)+
  theme_classic()+
  geom_vline(xintercept=0)+
  theme(axis.text=element_text(size=10),axis.title=element_text(size=12))+
  theme(legend.position="none")

```

#### 1.1.4. Inter-seasonal variance

```

### Calculation for inter-seasonal variance per iteration uses output for Jolly-Seber model as input

F_sim<-read.table("",header=T) #Simulation file for forest
n.iter<-dim(F_sim)[1]
F_m<-as.vector(rep(0,n.iter))
F_sd<-as.vector(rep(0,n.iter))
F_CV<-as.vector(rep(0,n.iter))
for(i in 1:n.iter){
  F_m[i]<-mean(as.numeric(F_sim[i,55:66]))
  F_sd[i]<-sd(as.numeric(F_sim[i,55:66]))
  F_CV[i]<-F_sd[i]/F_m[i]
}

mean(F_CV)
sd(F_CV)

C_sim<-read.table("",header=T) #Simulation file for coffee
n.iter<-dim(C_sim)[1]
C_m<-as.vector(rep(0,n.iter))
C_sd<-as.vector(rep(0,n.iter))
C_CV<-as.vector(rep(0,n.iter))
for(i in 1:n.iter){
  C_m[i]<-mean(as.numeric(C_sim[i,55:66]))
  C_sd[i]<-sd(as.numeric(C_sim[i,55:66]))
  C_CV[i]<-C_sd[i]/C_m[i]
}

```

```

}

mean(C_CV)
sd(C_CV)

```

### 1.1.5. Age

```

### Generalized linear models for mist-net data - AGE

## Part 1. Get ready (and modify and define response)

library(AICcmodavg)
setwd("")
data<-read.table("",header=T)
d<-data[data$SP=="SSPP",] #Select species
d1<-d[d$AGE=="3" | d$AGE=="4",] #Subset to only immature and adult individuals
d1$AGE<-as.character(d1$AGE)
d1$AGE[d1$AGE=="3"] <- 0 # FAIL defined as capturing an immature individual
d1$AGE[d1$AGE=="4"] <- 1 # SUCCESS defined as capturing an adult individual
d1$AGE<-as.numeric(d1$AGE)
d1$RESP<-d1$AGE
d1$HAB<-as.factor(d1$HAB)
d1$DAY2<-d1$DAY^2
d1$DAYs<-scale(d1$DAY)
d1$DAY2s<-scale(d1$DAY2)
attach(d1)

## Part 2. Define models and run model selection

M0<-glm(RESP~1,family=binomial)
Mt<-glm(RESP~DAYs+DAY2s,family=binomial)
Mh<-glm(RESP~HAB,family=binomial)
Mth<-glm(RESP~DAYs+DAY2s+HAB,family=binomial)
summary(M0)
summary(Mt)
summary(Mh)
summary(Mth)
Cand.mods <- list(M0,Mt,Mh,Mth)
modnames<-c("M0","Mt","Mh","Mth")
aictab(Cand.mods,modnames,sort=TRUE)
modavg(Cand.mods,parm="((Intercept)",modnames)
modavg(Cand.mods,parm="DAYs",modnames)
modavg(Cand.mods,parm="DAY2s",modnames)
modavg(Cand.mods,parm="HABFOREST",modnames)

## Part 3. Build graphs

plot(DAY,RESP,type="n",ylab="p(variable)",xlab="Day of year",main="SSPP-RES",ylim=c(0,1))
points(DAY[HAB=="COFFEE"],RESP[HAB=="COFFEE"],col="red",pch=16)
points(DAY[HAB=="FOREST"],RESP[HAB=="FOREST"],col="black",pch=16)
m<-glm(RESP~DAY+DAY2+HAB,family=binomial)
plotdatc<-data.frame(DAY=min(DAY):max(DAY))
plotdatc$DAY2<-plotdatc$DAY^2
plotdatc$HAB<-"COFFEE"
preddatc<-predict(m,newdata=plotdatc,se.fit=TRUE)
with(preddatc,lines(min(DAY):max(DAY),exp(fit)/(1+exp(fit)),col="red",lwd=2))
with(preddatc,lines(min(DAY):max(DAY),exp(fit+1.96*se.fit)/(1+exp(fit+1.96*se.fit)),lty=2,col="red"))
with(preddatc,lines(min(DAY):max(DAY),exp(fit-1.96*se.fit)/(1+exp(fit-1.96*se.fit)),lty=2,col="red"))
plotdatf<-data.frame(DAY=min(DAY):max(DAY))
plotdatf$DAY2<-plotdatf$DAY^2
plotdatf$HAB<-"FOREST"
preddatf<-predict(m,newdata=plotdatf,se.fit=TRUE)
with(preddatf,lines(min(DAY):max(DAY),exp(fit)/(1+exp(fit)),col="black",lwd=2))
with(preddatf,lines(min(DAY):max(DAY),exp(fit+1.96*se.fit)/(1+exp(fit+1.96*se.fit)),lty=2))
with(preddatf,lines(min(DAY):max(DAY),exp(fit-1.96*se.fit)/(1+exp(fit-1.96*se.fit)),lty=2))

```

## 1.2. Habitat performance

### 1.2.1. Body condition

```
### Generalized linear models for mist-net data - BCI

## Part 1. Calculate Body Condition Index for each capture event following guidelines from
## Peig, J. & Green, A.J. (2009) New perspectives for estimating body condition from mass/length data: ## the scaled mass index as an alternative
method. Oikos 118:1883-1891.

library(smatr)
setwd("")
data<-read.table("",header=T) #data table with all measurements
d1<-data[is.na(data$WC) & !is.na(data$BM),] #WC stands for wing chord, BM for body mass
attach(d1)

d2<-d1[d1$SP=="SSPP",] #select species
k<-3 #recommended value to exclude outliers that are probably mistakes but not extreme cases
QLS_WC<-quantile(d2$WC)
LQ_WC<-QLS_WC[[2]]
UQ_WC<-QLS_WC[[4]]
IQR_WC<-UQ_WC - LQ_WC
d3<-d2[d2$WC<(UQ_WC+(k*IQR_WC)) & d2$WC>(LQ_WC-(k*IQR_WC)),]
QLS_BM<-quantile(d3$BM)
LQ_BM<-QLS_BM[[2]]
UQ_BM<-QLS_BM[[4]]
IQR_BM<-UQ_BM - LQ_BM
d4<-d3[d3$BM<(UQ_BM+(k*IQR_BM)) & d3$BM>(LQ_BM-(k*IQR_BM)),] #These lines remove outliers from data

Mi.2<-log(d4$BM) #log of BM
Li.2<-log(d4$WC) #log of WC
Lo.2<-mean(Li.2)
sma_model.2<-sma(Mi.2~Li.2,method="SMA") #Standard major axis regression

M1.2<-sma_model.2$coeff[[1]]
M2.2<-as.matrix(M1.2)
bSMA.2<-M2.2[2,1]
BCI.2<-Mi.2*((Lo.2/Li.2)^bSMA.2) #Apply formula suggested by Peig & Green 2009
t2<-cbind(d4,BCI.2) #Create a data vector with BCI values for each capture that had both WC and BM
write.table(t2,"",sep="\t")

## Part 2. Define generalized linear models and run model selection

library("bbmle")
library(AICmodavg)
data<-read.table("t2.txt",header=T) #Output from previous step
attach(data)
DAY2<-DAY^2
DAYs<-scale(DAY)
DAY2s<-scale(DAY2)
BCIs<-scale(BCI.2)

M0<-lm(BCIs~1)
Mt<-lm(BCIs~DAYs+DAY2s)
Mh<-lm(BCIs~HAB)
Mth<-lm(BCIs~DAYs+DAY2s+HAB)
summary(M0)
summary(Mt)
summary(Mh)
summary(Mth)
Cand.mods <- list(M0,Mt,Mh,Mth)
modnames<-c("M0","Mt","Mh","Mth")
aictab(Cand.mods,modnames,sort=TRUE)
modavg(Cand.mods,parm="(Intercept)",modnames)
modavg(Cand.mods,parm="DAYs",modnames)
modavg(Cand.mods,parm="DAY2s",modnames)
modavg(Cand.mods,parm="HABFOREST",modnames)
```

```
## Part 3. Build graphs

DAY2<-DAY^2
m<-lm(BCI.2~DAY+DAY2+HAB)
plot(DAY,BCI.2,type="n",ylab="Body Condition Index",xlab="Day of year",main="TUFL-BCI",ylim=c(3.9,4.1))
x<-seq(min(DAY),max(DAY),1)
one<-rep("COFFEE",length(x))
two<-rep("FOREST",length(x))
y1<-predict(m,list(DAY=x,DAY2=x^2,HAB=one),interval=c("confidence"),level=0.95,type="response")
y2<-predict(m,list(DAY=x,DAY2=x^2,HAB=two),interval=c("confidence"),level=0.95,type="response")
lines(x,y1[,1],col="red",lwd=2)
lines(x,y1[,2],col="red",lty=2)
lines(x,y1[,3],col="red",lty=2)
lines(x,y2[,1],col="black",lwd=2)
lines(x,y2[,2],col="black",lty=2)
lines(x,y2[,3],col="black",lty=2)
```

## 1.2.2. Muscle

```
### Generalized linear models for mist-net data - MUSCLE

## Part 1. Get ready (and modify and define response)

library(AICmodavg)
setwd("")
data<-read.table("",header=T)
d<-data[data$SP=="SSPP",] #Select species
d6<-d[d$MUS=="2" | d$MUS=="3",] #Subset to individuals with muscle score 2 and 3
d6<-d6[!is.na(d6$MUS),]
d6$MUS<-as.character(d6$MUS)
d6$MUS[d6$MUS=="2"] <- 0 # FAIL defined as individuals with medium muscle
d6$MUS[d6$MUS=="3"] <- 1 # SUCCESS defined as individuals with high muscle
d6$MUS<-as.numeric(d6$MUS)
d6$RESP<-d6$MUS
d6$HAB<-as.factor(d6$HAB)
d6$DAY2<-d6$DAY^2
d6$DAYs<-scale(d6$DAY)
d6$DAY2s<-scale(d6$DAY2)
attach(d6)

## Part 2. Define models and run model selection
##### Same code as for AGE

## Part 3. Build graphs
##### Same code as for AGE
```

## 1.2.3. Primary molt

```
### Generalized linear models for mist-net data – PP MOLT

## Part 1. Get ready (and modify and define response)

library(AICmodavg)
setwd("")
data<-read.table("",header=T)
d<-data[data$SP=="SSPP",] #Select species
d4<-d[d$AGE=="3" | d$AGE=="4",] #Subset to only immature and adult individuals
d4<-d4[!is.na(d4$PLU),]
d4$PLU<-as.character(d4$PLU)
d4$PLU[d4$PLU=="2"] <- 0 # FAIL defined as individual not IN FULL PPMolting
d4$PLU[d4$PLU=="3"] <- 0 # FAIL defined as individual not IN FULL PPMolting
d4$PLU[d4$PLU=="5"] <- 0 # FAIL defined as individual not IN FULL PPMolting
d4$PLU[d4$PLU=="4"] <- 1 # SUCCESS defined as individual is IN FULL PPMolting
d4$PLU[d4$PLU=="6"] <- 1 # SUCCESS defined as individual is IN FULL PPMolting
d4$PLU<-as.numeric(d4$PLU)
```

```

d4$RESP<-d4$PLU
d4$HAB<-as.factor(d4$HAB)
d4$DAY2<-d4$DAY^2
d4$DAYs<-scale(d4$DAY)
d4$DAY2s<-scale(d4$DAY2)
attach(d4)

## Part 2. Define models and run model selection
##### Same code as for AGE

## Part 3. Build graphs
##### Same code as for AGE

```

### 1.2.4. Breeding

```

### Generalized linear models for mist-net data – BREEDING

## Part 1. Get ready (and modify and define response)

library(AICmodavg)
setwd("")
data<-read.table("",header=T)
d<-data[data$SP=="SSPP",] #Select species
d3<-d[lis.na(d$BRE),]
d3$BRE<-as.character(d3$BRE)
d3$BRE[d3$BRE=="NO"] <- 0 # FAIL defined as individual not breeding
d3$BRE[d3$BRE=="YES"] <- 1 # SUCCESS defined as individual is breeding
d3$BRE<-as.numeric(d3$BRE)
d3$RESP<-d3$BRE
d3$HAB<-as.factor(d3$HAB)
d3$DAY2<-d3$DAY^2
d3$DAYs<-scale(d3$DAY)
d3$DAY2s<-scale(d3$DAY2)
attach(d3)

## Part 2. Define models and run model selection
##### Same code as for AGE

## Part 3. Build graphs
##### Same code as for AGE

```

### 1.2.5. Juveniles

```

### Generalized linear models for mist-net data – JUVENILES

## Part 1. Get ready (and modify and define response)

library(AICmodavg)
setwd("")
data<-read.table("",header=T)
d<-data[data$SP=="SSPP",] #Select species
d5<-d[d$AGE!="0",] #Remove individuals with age unknown
d5$AGE<-as.character(d5$AGE)
d5$AGE[d5$AGE=="4"] <- 0 # FAIL defined as AHY individual
d5$AGE[d5$AGE=="3"] <- 0 # FAIL defined as AHY individual
d5$AGE[d5$AGE=="2"] <- 1 # SUCCESS defined as JUVENILE individual
d5$AGE<-as.numeric(d5$AGE)
d5$RESP<-d5$AGE
d5$HAB<-as.factor(d5$HAB)
d5$DAY2<-d5$DAY^2
d5$DAYs<-scale(d5$DAY)
d5$DAY2s<-scale(d5$DAY2)
attach(d5)

## Part 2. Define models and run model selection
##### Same code as for AGE

```

```
## Part 3. Build graphs
##### Same code as for AGE
```

## 2. Level 2: Composite indexes of habitat preference and performance

### 2.1. Habitat preference

No R code was used for this part of the analysis.

### 2.2. Habitat performance

```
### Meta-analysis of habitat performance

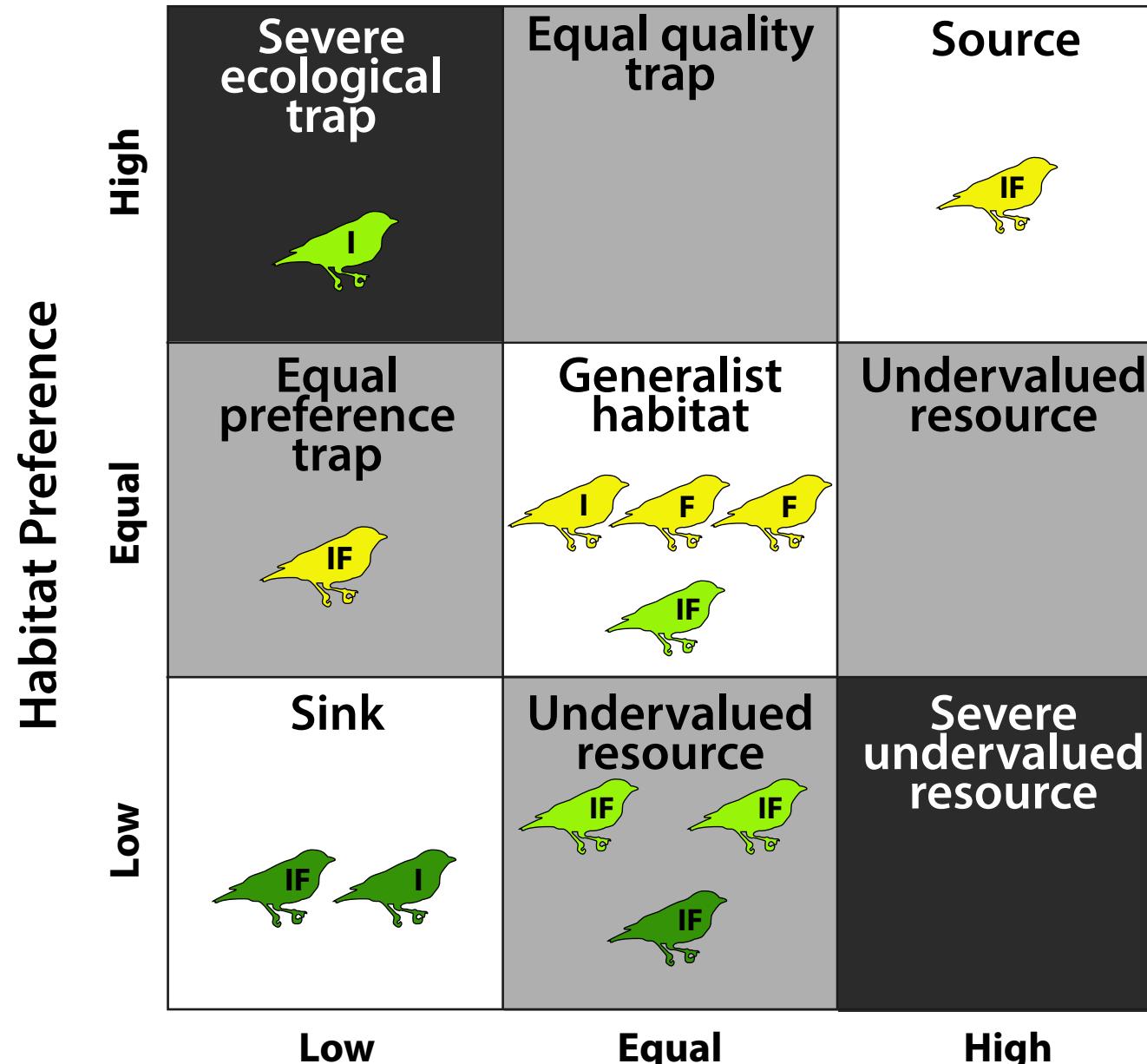
setwd("")
library("metafor")
data<-read.table("",header=T) #Read file with the mean and standard error for each species: variable
d<-data[data$SPP=="SSPP"] #Select species
d<-d[!is.na(d$MEAN),] #Remove NAs (analysis not carried out for a species)
x<-rma(yi=MEAN,sei=STE,data=d,method="FE",level=90);x
forest(x,slab=c("BCI","BRE","JUV","MUS","PLU"))
```

## 3. Level 3: Is habitat selection acting adaptively?

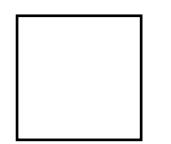
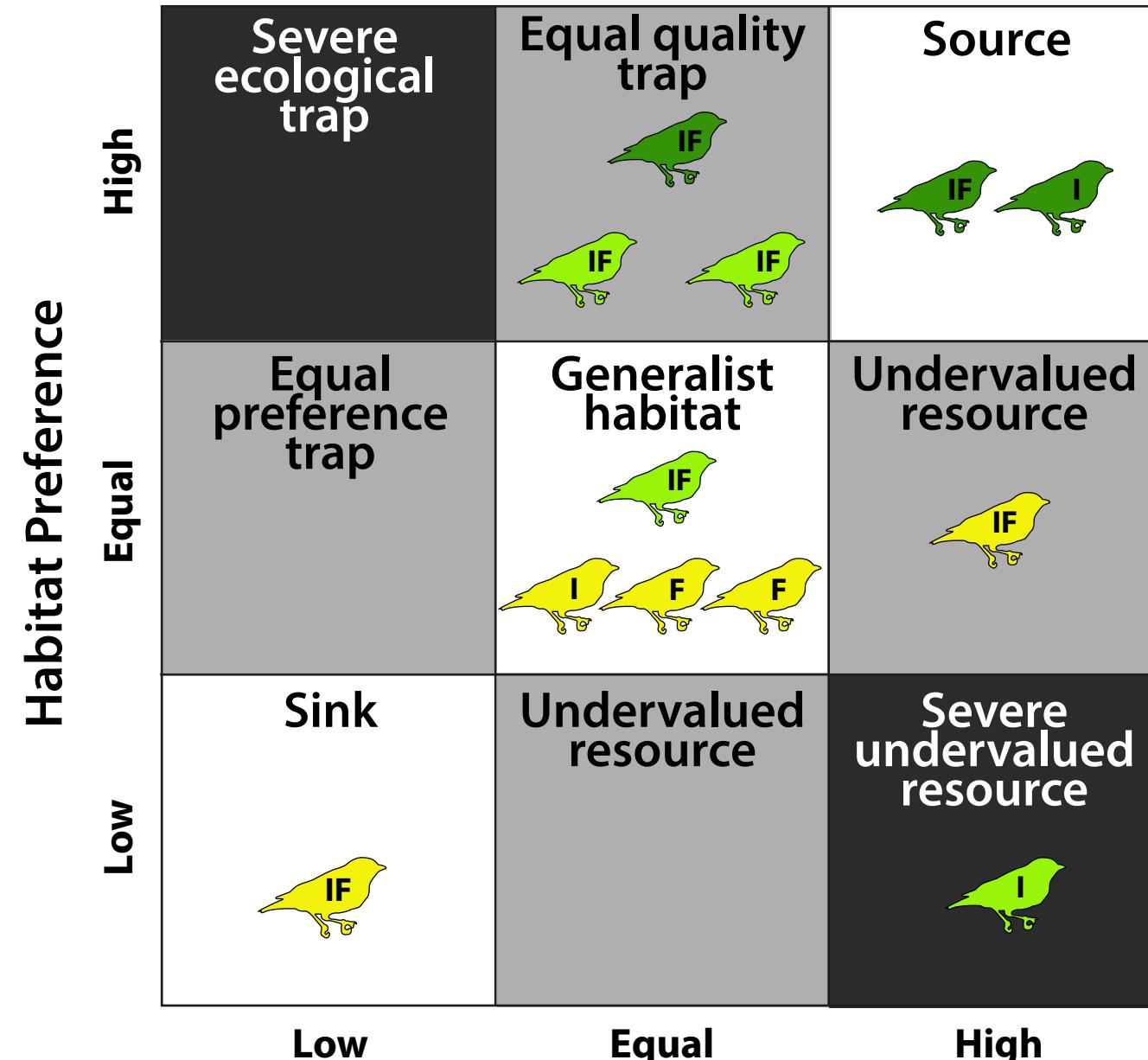
No R code was used for this part of the analysis.



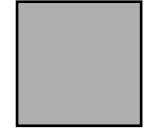
# Shade Coffee



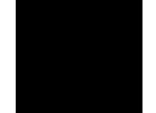
# Forest Remnants



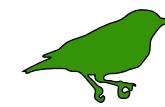
Adaptive outcomes



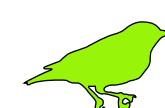
Non-ideal outcomes



Maladaptive outcomes



Forest species



Forest and woody-area species



Woody area and scrub species



Insectivores



Frugivores



Insectivores/Frugivores