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Wetlands in disturbed landscapes support higher avian biodiversity

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20 **Abstract**

21 Avian demographic and community responses to land development have been closely studied.
22 However most of this work has focused on forest dwelling species. The responses of wetland-
23 dependent birds to landscape alteration are less known. Based on prior work in forested
24 environments, I predicted that wetland dependent bird abundance and richness would be lower in
25 agricultural and urban/suburban environments compared with environments dominated by native
26 forest. Contrary to expectations, I found that avian richness and abundance was higher in
27 developed versus undeveloped landscapes. In particular, I observed greater richness and
28 abundance of birds in wetlands within agricultural landscapes. These results suggest that
29 developed landscapes may offer opportunities to meet conservation objectives for wetland birds.

30

31 **Introduction**

32 Avian responses to fragmented forest habitats are well described. Many studies report
33 the effects of clearcut logging and agriculture on forest songbirds, particularly in the context of
34 edge and isolation effects (Donovan et al. 1997; Major et al. 2001; Schmiegelow & Monkkonen
35 2002; Turner 1996). Their findings indicate that forest-dependent birds are generally influenced
36 negatively by altered landscapes. Because similar responses have been documented across a
37 wide variety of taxa and systems (Hansen et al. 2005), conservation frameworks often
38 incorporate assumptions that landscape alteration negatively impacts biotic communities (e.g.
39 Temple & Terry 2007). In contrast to this convention, recent investigations suggest that
40 disturbed landscapes may provide important habitat for species of conservation concern (e.g.
41 Rosenzweig 2003; Balcome et al. 2005; White & Main 2005; Hodgkison et al. 2007; Kareiva et
42 al. 2007; Milne & Bennett 2007; Winfree et al. 2007). These alternative perspectives suggest

43 that avian responses to habitat alteration may vary by species and habitat type. In light of these
44 conflicting views, developing further understanding of the effect of landscape development on
45 biotic communities remains a critical challenge for conservation scientists and practitioners.

46 Freshwater wetlands have undergone considerable anthropogenic losses and alterations
47 (Dahl 1990, 2000). However, few studies have estimated the community response of wetland-
48 dependent birds occupying wetlands situated in fragmented—as compared with undeveloped—
49 landscapes (Cox et al. 2000; but see DeLuca et al. 2004; Guadagnin et al. 2005; Guadagnin &
50 Maltchik 2007). Based on the patterns prevailing among forest dwelling birds, we might expect
51 congruent responses by wetland-dependent birds. That is, that conversion of landscapes adjacent
52 to intact wetlands should have negative effects on abundance and richness of wetland dependent
53 species. However, this prediction remains largely untested.

54 Here, I evaluate the distribution and abundance of wetland-dependent birds across land
55 cover types, ranging from forested to highly developed. I hypothesized that wetland-dependent
56 bird communities varied across these land cover types. Specifically, I predicted that richness and
57 abundance was lowest in wetlands situated within developed landscapes. Contrary to this
58 prediction, I observed higher wetland-dependent bird richness and abundance in developed rather
59 than undeveloped landscapes.

60

61 **Methods**

62 *Natural history and study sites*

63 I conducted surveys of wetland-dependent bird richness and abundance at 16 permanent,
64 open-canopy wetlands located in the state of Connecticut between latitudes 41° 28' 37" N –
65 42° 1' 50" N and longitudes 72° 6' 32" W – 73° 5' 5" W. Wetlands included for observation

66 were selected from a pool of candidate wetlands classified by land cover in a prior study (Skelly
67 et al 2006; D. K. Skelly unpublished data). From among this compliment, I surveyed those
68 wetlands in which private landowners granted access. To achieve sufficient replication across
69 treatments, I included for survey two additional wetlands (S.P.B., unpublished data). Wetlands
70 ranged in size from 245 m² – 47,124 m². For a subset of wetlands, size was determined optically
71 using a coincidence rangefinder (Skelly et al. 2006). For the remaining wetlands, size was
72 determined digitally from the U.S. Fish & Wildlife Service National Wetland Inventory (NWI)
73 using ArcView GIS v. 9.1.

74 Prior to conducting avian surveys, I groundtruthed all wetlands to verify accuracy of
75 previous land cover classifications. Land cover type encompassing wetlands varied, and
76 included native forest, commercial agriculture, and urban/suburban development. Forest
77 wetlands were characterized by a matrix of relatively contiguous deciduous and coniferous
78 forest. Agriculture wetlands were characterized by row crops (berries, corn, and shade-grown
79 tobacco) cultivated at commercially active farms. Impervious surface and turf characterized the
80 areas surrounding urban/suburban wetlands. These wetlands were either situated within
81 residential yards or business park complexes, or located adjacent to community housing
82 developments.

83

84 *Avian surveys*

85 I defined wetland-dependent birds as waterfowl, waders, and grebes, belonging
86 respectively to the orders *Anseriformes*, *Ciconiiformes* or *Charadriiformes*, and
87 *Podicipediformes*. I surveyed intensively wetland-dependent bird richness and abundance by
88 conducting five point count surveys at each of the 16 total wetlands according to standard

89 protocols (Bibby 1992). A 10-min observation period comprised each of the first and second
90 surveys conducted at all wetlands. A 1-min observation period comprised each of the remaining
91 three surveys. These three additional surveys were included in the design because of the
92 advantages associated with increasing counts (Tozer et al. 2006; Verner & Ritter 1986). All five
93 surveys per wetland were conducted between 11 July – 06 August 2006, a time period that
94 coincides with post-breeding and pre-migration. To include potential seasonal variation in
95 richness and abundance associated with the breeding period, I conducted five additional point
96 count surveys (identical in design to those of 2006) between 22 May – 08 July 2007. In total, I
97 conducted 736 person-minutes of observation (two 10-min surveys + three 1-min surveys X 16
98 wetlands X 2 years = 736) across 160 point count surveys. For all surveys, the order in which
99 wetlands were visited was haphazard. To minimize the risk of temporal induced bias, I
100 conducted surveys in blocks. Within each block, I surveyed all wetlands within a period of
101 several days, the exact period depending on weather conditions. I conducted 10-min point count
102 surveys between one half hour before sunrise and 1030 hr, a period of increased bird activity
103 (Verner & Ritter 1986). I conducted 1-min point count surveys throughout the day because
104 counts during all hours may yield higher richness (Verner & Ritter 1986), and evening counts are
105 known to yield equivalent sampling efficiency for wetland-dependent birds (Krzys et al. 2002).
106 Each survey was conducted from a single location at the wetland edge. Surveys were not
107 conducted during high wind or rain events. Prior to entering each wetland, I observed open
108 water from a vantage point and recorded all birds detected visually and aurally (Naugle et al.
109 2000). I recorded all birds detected during the approach and egress into and out of each wetland.
110 All adults and juveniles detected within ca. 100 m of the wetland were recorded. Immediately
111 following the 10-min point count, I walked the wetland perimeter to increase the probability of

112 detecting secretive species (Naugle et al. 2000). At two of the largest wetlands (one forested,
113 one urban/suburban), I chose not to walk through a small portion (< 27%) of the perimeter due to
114 a combination of shear size, deep water, and thick vegetation. This likely had little influence on
115 the results as no new individuals were detected during any perimeter walks.

116

117 *Land cover assessment*

118 I conducted all geographic analyses in ArcView GIS v. 9.1. For each wetland, I
119 delineated a buffer zone based on a distance of 200 m from the wetland perimeter. I chose this
120 distance because visual inspection of the land cover map indicated that larger buffer zones
121 diluted the land adjacency pattern of interest. Wetland perimeters for 14 wetlands were based on
122 NWI maps. For two wetlands that were not identified on the NWI maps, I traced the perimeter
123 of the wetland as depicted in a 2004 aerial photograph (Univ. of CT Center for Land Use
124 Education and Research [UConn CLEAR] 2007). I overlaid buffers onto a land-cover map
125 generated through supervised classification of 30 m spatial resolution Landsat imagery acquired
126 in 2002 (UConn CLEAR 2007). Within the land cover map, I resampled the data to increase the
127 usable resolution such that each 30 x 30 m pixel was subdivided into four identical pixels. The
128 original land-cover dataset contained eleven classes. From these, I extracted and summed the
129 proportion of land-cover classes within each buffer by defining three types of land cover: forest,
130 agriculture, and urban/suburban (King et al. 2005).

131 I defined urban/suburban land as the sum of *developed, turf & grass, and barren*
132 categories, which described respectively high-density commercial, industrial, and residential
133 areas; cultivated lawns; and non-vegetated non-agricultural areas (e.g. mines, quarries).
134 Percentage agriculture was defined by one class, *other grasses & agriculture*, which included

135 agricultural crop and pasture fields. I defined forest as the summation of two classes: *coniferous*
136 *forest* and *deciduous forest*. Prior to summing proportions, I subtracted from each buffer any
137 pixels belonging to a wetland classification.

138 I used the summed proportion data to assign each wetland a land cover classification
139 based on the most common land cover type. Urban/suburban wetlands were located within
140 buffers consisting of 51-90 % urban/suburban cover. Agriculture wetlands were located within
141 buffers consisting of 42-86 % agriculture. Forest wetlands were located within buffers
142 containing 78-97 % forest cover. For each wetland, subdominant land-cover types never
143 exceeded 34 %. I could not describe a complete 200 m buffer for one wetland located near the
144 Connecticut-Massachusetts border. While the wetland basin was entirely within Connecticut,
145 approximately one fifth of the buffer area was located in Massachusetts. To avoid problems
146 associated with using an independently created land cover database, I based the buffer
147 classification for this wetland on the Connecticut portion of the buffer. Inspection of aerial
148 photographs suggested there was little distinction in cover distribution between states.

149

150 *Statistical analyses*

151 *Mixed-model approach*

152 I analyzed the effect of land cover classification—forest, agriculture, and
153 urban/suburban—on species richness and abundance using two approaches: (1) I used a
154 categorical predictor approach to analyze the effect of the three land cover classes on richness
155 and abundance and (2) I used a continuous predictor approach to assess the effect of proportional
156 land cover data on richness and abundance. In both categorical and continuous analyses, I used
157 mixed-effects models to analyze repeated observations at the year level, and to account for the

158 combination of fixed effects (specifically, class and proportion land cover) and random effects
159 (specifically, wetland; Thogmartin et al. 2004). This approach avoided problems with limited
160 power associated with traditional approaches relying on random effects modeling of repeated
161 observations (e.g. ANOVA; R. Baayen et al. unpublished data).

162

163 *Avian richness and abundance*

164 ANOVA and mixed model analyses of richness and abundance were conducted using R
165 v. 2.5.0 (R Development Core Team 2007); goodness of fit tests were conducted using S-Plus v.
166 7.0 (Insightful Corp., Seattle, WA). Prior to analyzing the effect of land cover on richness and
167 abundance, I first used ANOVA to test for an effect of land-cover on wetland size. I log-
168 transformed wetland area values because they were non-normal and thus violated the
169 assumptions of parametric analyses. Wetland size was distributed randomly among classes
170 (ANOVA: $F = 1.44$, $df = 2,13$, $p = 0.27$), and did not vary significantly with respect to land
171 cover proportion data (Linear Regression of cumulated land cover: $df = 1,14$, $F = 3.24$, $p = 0.09$;
172 Multiple Linear Regression of partitioned land cover: $df = 2,13$, $F = 3.23$, $p = 0.07$). I therefore
173 did not include wetland size as a covariate in subsequent analyses.

174 I defined bird richness for each wetland as the total number of wetland-dependent bird
175 species detected across all observations. I defined abundance as the maximum number of
176 individuals detected among each species across all observations (Betts et al. 2005). Within the
177 mixed model, I analyzed the effect of the three land cover classes (i.e. forest, agriculture, and
178 urban/suburban) on richness and abundance. I used Kolmogorov-Smirnov (K-S) tests to assess
179 the goodness of fit (GOF) of the data to multiple distributions. Because the distribution of avian
180 richness data was not significantly different from a Poisson distribution (K-S GOF: $p = 0.11$), I

181 chose a Poisson family to characterize the error term in the mixed model. Poisson distributions
182 often characterize count data (Thogmartin et al. 2004). Because abundance data were over-
183 dispersed, I log transformed the original counts to achieve normality (K-S GOF: $p = 0.75$).

184 I used a similar mixed model analysis to determine the effect of land cover proportion
185 data on richness and abundance. To address issues of multicollinearity associated with
186 proportion data (King et al. 2005), I removed from analysis the single land cover class—percent
187 forest—present at all sites (Newbold & Eadie 2004). To estimate the influence of anthropogenic
188 landscape development on wetland birds, I performed a mixed model analysis in which the
189 single predictor variable was the cumulative proportion of the landscape containing
190 urban/suburban and agriculture land cover. In cases where either richness or abundance of birds
191 was associated with human development, I performed two subsequent mixed model analyses
192 using the continuous predictor approach. Each of these subsequent models included two
193 explanatory variables: proportion urban/suburban and proportion agriculture. The first of these
194 models lacked an interaction term for the land cover types, while the second model contained the
195 interaction term. In each case, Akaike information criterion (AIC) values indicated that model fit
196 did not improve notably (i.e. AIC values did not decrease by greater than 2) with the addition of
197 the interaction term. I therefore chose to analyze for significance the more parsimonious models
198 lacking the interaction term.

199

200 *Species composition*

201 To estimate the relationship between species composition and land cover, I used
202 CANOCO software (Version 4.5) to conduct a redundancy analysis (RDA) of species
203 composition against proportion agriculture and proportion urban/suburban. Composition data

204 were based on square root-transformed maximum abundance of all species detected at each site.
205 The RDA was based on the matrix of species correlation coefficients, and was conducted using a
206 split-plot design, such that each wetland site contained two abundance observations—one from
207 each year. The permutation analysis for significance—based on 9999 permutations—was freely
208 interchangeable at the site level and modeled using a time series approach at the observation
209 level (ter Braak & Smilauer 1998). This approach ensured that multiple observations at each site
210 shared an error term, thus yielding a mixed-model. Following a significant result from this test, I
211 reanalyzed the data based on the subset of all species found at multiple sites. This exclusion of
212 singleton species—those species recorded at a single site—enabled inference into the robustness
213 of the model results. Since both analyses were significant, I report here only the former analysis
214 containing all recorded species.

215

216 **Results**

217 *Avian richness and abundance*

218 I detected 15 wetland-dependent bird species during a total of 160 observations. Species
219 detected most commonly (those present across 9-12 % of all observations) include Mallard (*Anas*
220 *platyrhynchos*), Green Heron (*Butorides virescens*), Canada Goose (*Branta canadensis*), and
221 Great Blue Heron (*Ardea herodias*). Species detected less commonly (those present across 6-7
222 % of all observations) include Wood Duck (*Aix sponsa*) and Killdeer (*Charadrius vociferus*).
223 The species detected least commonly (those present < 3 % of all observations) include American
224 Black Duck (*Anas rubripes*), American Bittern (*Botaurus lentiginosus*), Red-necked Grebe
225 (*Podiceps grisegena*), Spotted Sandpiper (*Actitis macularius*), Solitary Sandpiper (*Tringa*
226 *solitaria*), Great Egret (*Ardea alba*), Willet (*Tringa semipalmata*), Common Merganser (*Mergus*

227 *merganser*), and Hooded Merganser (*Lophodytes cucullatus*). Of these, *B. lentiginosus* and *A.*
228 *alba* and are considered, respectively, endangered and threatened in the state of CT (CT DEP,
229 2007). In all, eight species were singletons.

230 Across sixteen wetlands, land cover classification was associated with avian richness
231 (mixed-model: $P = 0.032$), but not avian abundance (mixed-model: $P = 0.087$; Fig. 1). A
232 treatment level contrast revealed that agriculture wetlands had greater richness than forest
233 wetlands ($P = 0.007$). On average, agriculture wetlands contained 4.5 times the number of
234 species detected in forest wetlands. No other contrasts revealed differences in richness across
235 class. The proportion of a landscape that was developed was positively related to both avian
236 richness (mixed model: $P = 0.022$, AIC = 43.4) and abundance (mixed-model: $P = 0.019$, AIC =
237 45.5; Table 1). The model incorporating the terms for proportion urban/suburban and proportion
238 agriculture was significant for both richness (mixed model: $P = 0.023$, AIC = 43.1) and
239 abundance (mixed-model: $P = 0.036$, AIC = 46.4; Table 2).

240

241 *Species composition*

242 Results from a redundancy analysis (RDA) indicated a relationship between community
243 composition and proportion agriculture ($P = 0.027$), but not with proportion urban/suburban ($P =$
244 0.635). Neither was there evidence for an interaction between agriculture and urban/suburban
245 land covers on community composition ($P = 0.315$). The total variance associated with the
246 community data fitted to land cover data was 18.1 %, as indicated by the sum of the canonical
247 eigenvalues. A triplot of the ordination (Fig. 3) shows arrows pointing in the direction of
248 steepest increase of their respective values within the RDA. Species arrow length represents the
249 multiple correlation of each species with the ordination axes (ter Braak & Smilauer 1998). The

250 majority of species associate strongly with agriculture, while a subset associate most strongly
251 with the interaction of agriculture and urban/suburban landscapes. Two species (*L. cucullatus*
252 and *T. solitaria*) show weak or negative correlation with both the ordination axes and the
253 environmental variables.

254

255 **Discussion**

256 Conservation biologists have typically held that landscape conversion leads to declines in
257 biodiversity (Pimm & Askins 1995; Sala et al. 2000). This pattern is well established for forest
258 dwelling birds. Many studies also report the negative impacts of land development on wetland
259 wildlife (e.g. Findlay & Houlihan 1997; Lehtinen et al. 1999), further supporting the view that
260 anthropogenic impacts erode biodiversity. Based on results from previous forest bird studies, I
261 expected that bird richness and abundance would decrease in wetlands surrounded by natural
262 habitat that had been converted by anthropogenic land use. In contrast to this prediction, I found
263 that wetland-dependent bird richness and abundance increased in the context of landscapes
264 converted for human use.

265 Both sets of continuous models indicated that wetland-dependent bird richness and
266 abundance increased with the proportion of the landscape that was developed. However, the
267 model containing two terms for land development (proportion urban/suburban and proportion
268 agriculture) was more parsimonious (i.e. lower AIC value) than the model treating development
269 as a single term—comprising the sum of proportion urban/suburban plus proportion agriculture
270 (Tables 1 and 2). Furthermore, in the two-termed model, only agriculture was associated
271 significantly with richness and abundance. RDA results indicated a similar trend, whereby
272 community structure was strongly associated with agriculture. While the majority of species are

273 shown to aggregate along the steepest line of agricultural increase, a small cluster straddles the
274 parameter space representing the interaction of agriculture and urban/suburban landscapes.
275 Strikingly, only two species (*T. solitaria* and *L. cucullatus*) showed weak association with the
276 terms for landscape development.

277 Taken together, these results suggest that different forms of landscape conversion do not
278 have equivalent influences on wetland-dependent birds. Specifically, while both predominant
279 land type and proportion of land type adjacent to wetlands predicted abundance and richness,
280 agricultural landscapes appear to have the strongest influence on this outcome. In addition to the
281 strong association with agriculture, it is important to note that richness in urban/suburban
282 wetlands did not differ significantly from that of forest wetlands, and that abundance showed a
283 weak positive association with urban/suburban landscapes. Moreover, the two threatened
284 species—*B. lentiginosus* and *A. alba*—were detected only in a suburban/urban and an agriculture
285 wetland respectively.

286 That species richness and abundance were shown here to increase with increasing
287 anthropogenic development prompts questions concerning the mechanisms by which developed
288 landscapes might positively influence wetland-dependent bird communities. As one possibility,
289 fertilizer runoff in agriculture wetlands might bolster community richness via bottom-up trophic
290 effects. Alternatively, decreased density of wetlands in developed landscapes (Gibbs 2000) may
291 function to concentrate biota in remaining habitat (White & Main 2005).

292 Insight into these patterns will require further experimental studies. However, partial
293 insight may be gained from previous observational studies that have documented wetland-
294 dependent bird distributions associated with increasing agricultural food resources (Fasola &
295 Ruiz 1996; Czech & Parsons 2002). In particular, Czech & Parsons (2002) note that agricultural

296 areas—especially those dedicated to the cultivation of rice and sorghum—provide important
297 habitat for waterfowl. Similarly, postharvest waste corn remaining on fields is a primary energy
298 source for North American waterfowl, and is believed to augment the success of management
299 programs (Krapu et al. 2004). In practice, the U.S. Fish & Wildlife Service promotes the
300 planting of grain crops adjacent to wetlands as part of a strategy for waterfowl management on
301 agency refuges (Cross & Vohs 1988). The fostering of wetland-dependent wildlife via
302 agricultural land cover typically associated with declines in biodiversity marks an important
303 departure from the views traditionally embraced in conservation science.

304 Increasing rates of land conversion for human settlement and agricultural use have been
305 cited as two of the most important threats to global biodiversity (Ricketts & Imhoff 2003). This
306 is certainly true for a wide diversity of taxa. However, an emerging body of literature reports the
307 positive influence of disturbed landscapes on certain communities of native wildlife species
308 (Balcome et al. 2005; Hodgkison et al. 2007; Winfree et al. 2007), thus underscoring the role of
309 context dependency in describing species responses to landscape conversion. Some conservation
310 goals may be met in disturbed habitats while others are not (Rosenzweig 2003). For instance,
311 many wetland bird species may persist in disturbed wetlands while most wetland amphibian
312 species do not (Gibbs et al. 2005). Parsing contrasting influences of landscape conversion
313 presents a critical challenge for conservation scientists and decisionmakers intent on estimating
314 tradeoffs in ecosystem attributes in “domesticated landscapes” (Kareiva et al. 2007). Estimating
315 the structure of such tradeoffs will require additional studies to address explicitly the nature of
316 community level responses to multiple land cover types across broad spatial and temporal scales
317 (Miller & Hobbs 2002). Insights gained from such an approach may reveal new opportunities

318 for conservation and facilitate a cooperative framework for meeting the frequently conflicting
319 objectives of conservation and land development.

320

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328

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447 Table 1. Results of the cumulated mixed-model analyses of avian richness and log-transformed
 448 abundance across proportion of the landscape that was developed. Developed is defined as the
 449 cumulated proportion of agriculture plus urban/suburban land cover.

Land cover development term	Estimate	Std. Error	df/Chi df	Test Stat.	AIC	<i>p</i>
Avian species richness				<i>Z</i>		
developed	1.679	0.727	3/1	2.308	43.42	0.022
Avian species abundance				<i>Chisq.</i>		
developed	0.771	0.319	3/1	5.518	48.73	0.019

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464 Table 2. Results of the mixed-model analyses of avian richness and log-transformed abundance
 465 across proportion of the landscape that was developed. Models contain partitioned land cover
 466 terms.

Land cover development term	Estimate	Std. Error	df/Chi df	Test Stat.	AIC	<i>p</i>
				Chisq.		
Avian species richness			4/2	9.507	41.17	0.009
				Z		
agriculture	1.796	0.671		3.465		0.001
urban/suburban	0.577	0.682		1.621		0.105
				Chisq.		
Avian species abundance			4/2	8.564	48.18	0.036
				T		
agriculture	1.364	0.380		3.085		0.004
urban/suburban	0.763	0.321		1.775		0.086

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476 **FIGURE LEGENDS**

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478 Figure 1. Mean (± 1 SE) avian species richness (open bars) and avian abundance (solid bars)
479 classed by dominant land cover. Mean richness is calculated as the average of the maximum
480 richness recorded in 2006 and 2007. Mean abundance is calculated as the average of the
481 maximum abundances recorded in 2006 and 2007. All abundance values are log-transformed.

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483 Figure 2. a) Log-transformed avian abundance—and b) log-transformed avian species richness—
484 against the proportion of land that was developed. Richness values plus 0.5 are log-transformed
485 only for graphical presentation. Diamonds indicate log-transformed maximum avian abundance
486 and richness for each set of observations in 2006 and 2007. Land development portrayed here
487 represents the cumulated proportion of agriculture and suburban/urban land cover within a 200 m
488 buffer surrounding each wetland. The solid line indicates the fitted values from the fixed-effect
489 component of the mixed models.

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491 Figure 3. Ordination diagram based on redundancy analysis of avian species composition against
492 proportion agriculture, proportion urban/suburban, and their interaction. An open circle (O)
493 denotes an urban/suburban wetland (labeled: Urb); A closed circle (\bullet) denotes a forest wetland
494 (labeled: For); An X denotes an agricultural wetland (labeled: Ag). Dotted arrows represent
495 species response, labeled with species common names. Bold arrows represent the two land cover
496 terms in the model (proportion agriculture; proportion urban suburban) along with a term for
497 their interaction (agriculture*urban/suburban). The angle between arrows indicates correlation,

498 with positive correlations represented by angles less than 90 deg and negative correlations at
499 angles greater than 90 deg.

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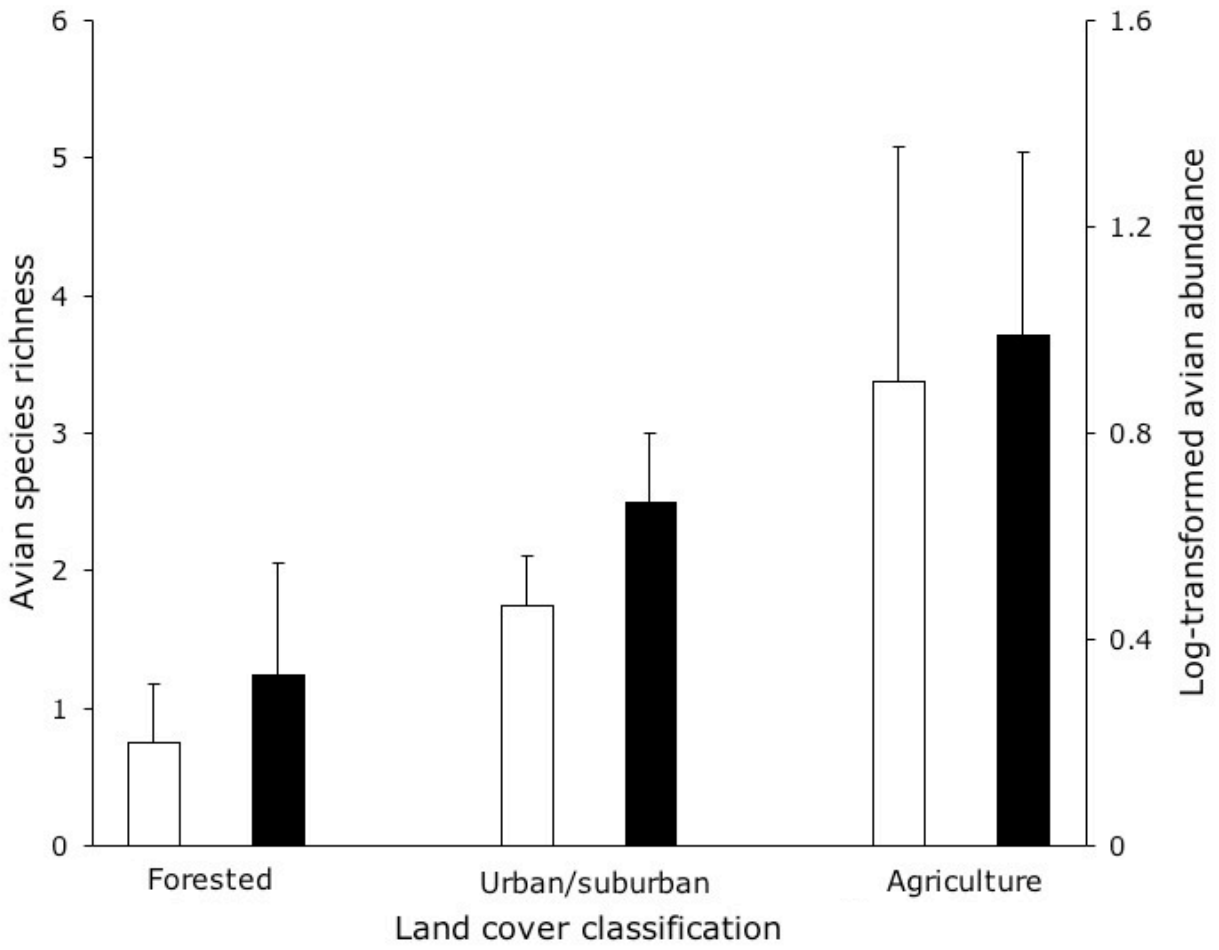
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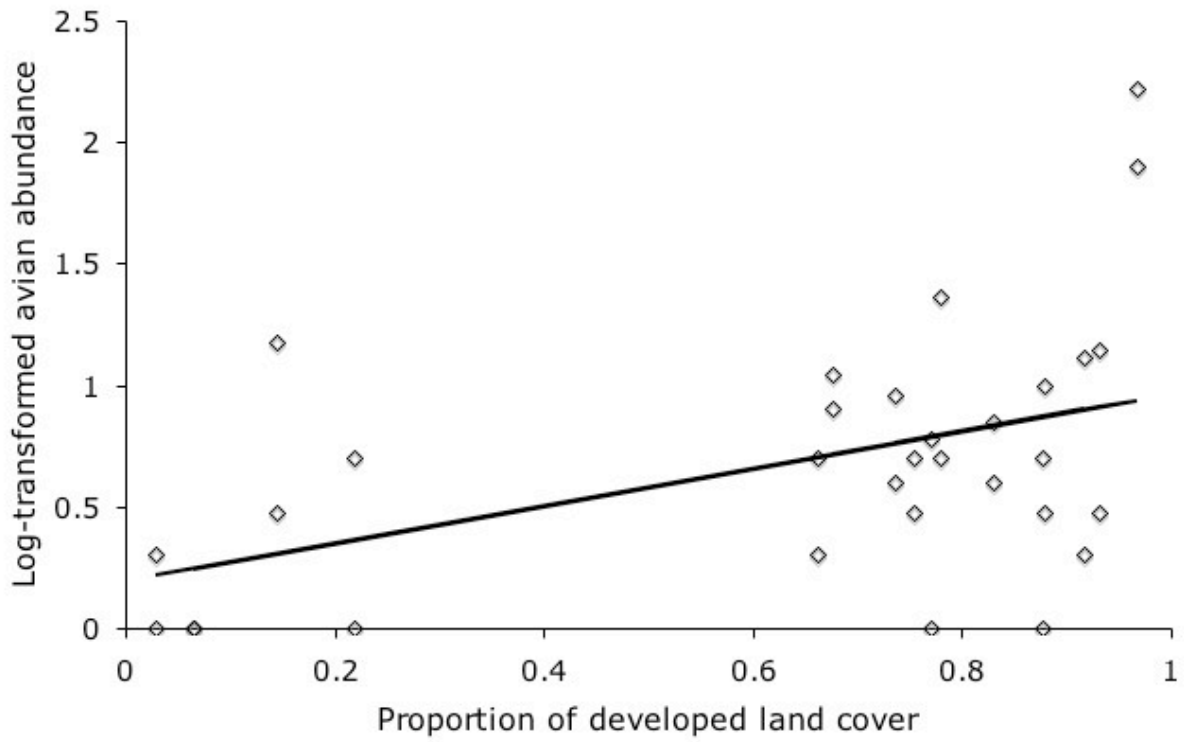
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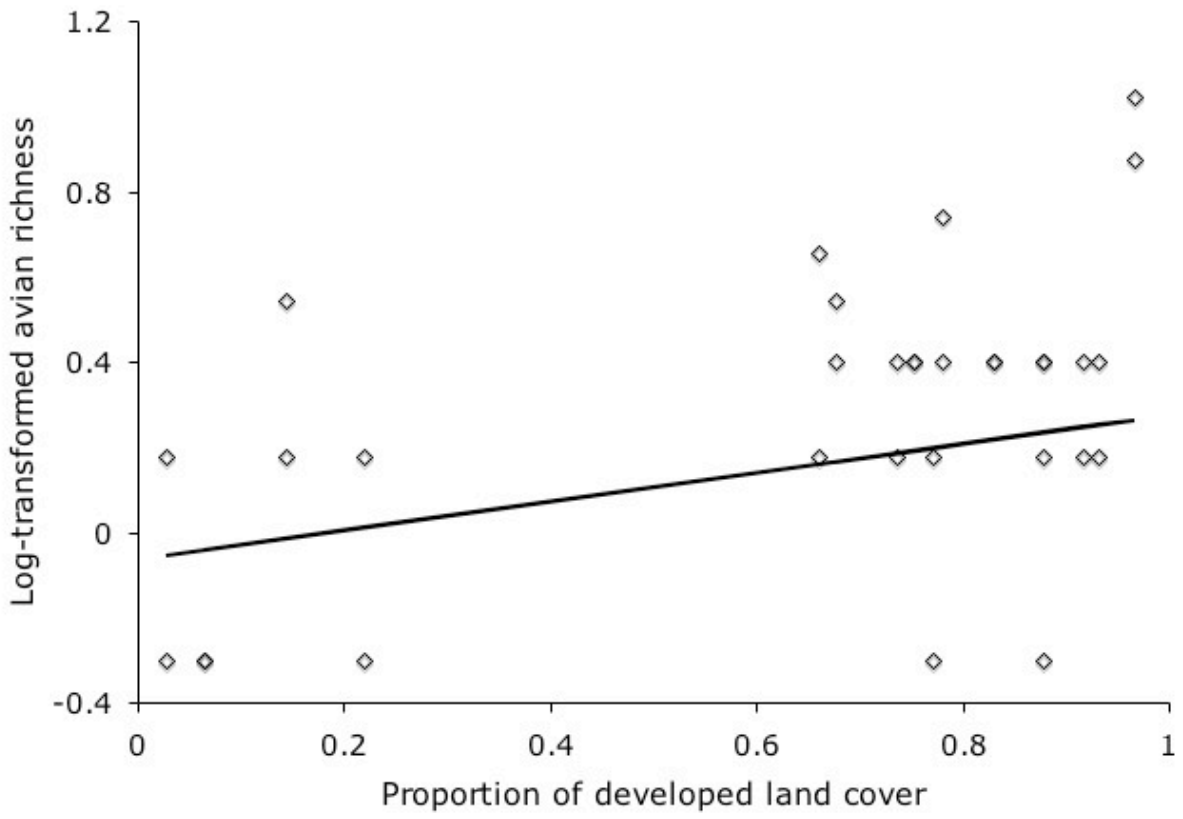
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