

1 **Birds as surrogates for mammals and reptiles: are patterns of cross-**  
2 **taxonomic associations stable over time in a human-modified landscape?**

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11

12 **ABSTRACT**

13

14 Cross-taxonomic surrogates can be feasible alternatives to direct measurements of biodiversity in  
15 conservation if validated with robust data and used with explicit goals. However, few studies of cross-  
16 taxonomic surrogates have examined how temporal changes in composition or richness in one taxon can  
17 drive variation in concordant patterns of diversity in another taxon, particularly in a dynamic and heavily  
18 modified landscape. We examined this problem by assessing changes in cross-taxonomic associations  
19 over time between the surrogate (birds) and target vertebrate taxa (mammals, reptiles) that demand high  
20 sampling effort, in a heterogeneous mosaic landscape comprising pine monoculture, eucalypt woodland  
21 remnants and agricultural land. Focussing on four study years (1999, 2001, 2011, 2013) from a dataset  
22 spanning 15 years, we: (1) investigated temporal changes in cross-taxonomic congruency among three  
23 animal taxa, (2) explored how temporal variation in composition and species richness of each taxon might  
24 account for variation in cross-taxonomic congruency, and (3) identified habitat structural variables that  
25 are strongly correlated with species composition of each taxon. We found the strength of cross-taxonomic  
26 congruency varied between taxa in response to both landscape context and over time. Among the three  
27 taxa, overall correlations were weak but were consistently positive and strongest between birds and  
28 mammals, while correlations involving reptiles were usually weak and negative. We also found that  
29 stronger species richness and composition correlations between birds and mammals were not only more  
30 prevalent in woodland remnants in the agricultural matrix, but they also increased in strength over time.  
31 Temporal shifts in species composition differed in rate and extent among the taxa even though these  
32 changes were significant over time, while important habitat structural correlates were seldom shared  
33 across taxa. Our study highlights the role of the landscape matrix and time in shaping animal communities  
34 and the resulting cross-taxonomic associations in the woodland remnants, especially after a major  
35 perturbation event (i.e. plantation establishment). In such dynamic landscapes, differing and taxon-  
36 specific shifts in diversity over time can influence the strength, direction and consistency of cross-

37 taxonomic correlations, therefore posing a ‘temporal’ problem for the use of surrogates like birds in  
38 monitoring and assessments of biodiversity, and conservation management practices.

39

40 **Keywords**

41 Anthropogenic landscapes, conservation, cross-taxonomic surrogates, congruence, indicators of  
42 biodiversity, longitudinal study, matrix, time scales

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66 **1. Introduction**

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68 Land-use change increasingly threatens biodiversity globally by driving habitat loss and degradation (Sala  
69 et al. 2000; Reidsma et al. 2006; Sayer et al. 2013). As a result, there is an urgent need to understand how  
70 diverse groups of biota respond to land-use modification across various scales (e.g., Mattison & Norris  
71 2005, Haines-Young 2009). Such knowledge is integral to informing decisions on how sites should be  
72 conserved and managed (Meir et al. 2004; Vandewalle et al. 2010). However, resource and taxonomic  
73 limitations impose enormous difficulties on sampling large suites of taxonomic groups (Lawton et al.  
74 1998; Schulze et al. 2004; Gardner et al. 2008) to understand broad changes in biodiversity patterns. This  
75 has resulted in multiple surrogate approaches being developed to act as proxies for components of  
76 biodiversity not able to be directly measured (Prendergast & Eversham 1997; Caro 2010; Lindenmayer et  
77 al. 2015), or biota that are costly or logistically difficult to survey within time frames available for  
78 decision-making (Favreau et al. 2006).

79

80 Species-based surrogates of biodiversity are a common type of surrogate (e.g., Caro 2010), and are based  
81 on the hypothesis that the occurrence or diversity of a surrogate or indicator taxon reflects the occurrence  
82 (i.e. co-occurrence) or diversity (i.e. richness, composition) of other sets of target taxa (Rohr et al. 2006;  
83 Rondinini et al. 2006; Gaspar et al. 2010). The best examples of these species surrogates include cross-  
84 taxonomic surrogates (e.g., Kati et al. 2004; Gallardo et al. 2011; Gaspar et al. 2010; Fattorini et al. 2012),  
85 biodiversity indicator species or species groups (e.g., Nally & Fleishman 2002; Roberge & Angelstam  
86 2004; Branton & Richardson 2011), and higher-taxonomic groups (e.g., Báldi 2003; Heino & Soininen  
87 2007).

88

89 Species surrogates of diversity in conservation have several empirical and conceptual shortcomings (e.g.,  
90 Andelman & Fagan 2000; Heink & Kowarik 2010). First, studies of cross-taxonomic relationships have  
91 yielded mixed results in terms of the strength and direction of congruency across different taxa, often

92 varying with the analytical approaches used (Gioria et al. 2011), even when landscape contexts and scales  
93 are broadly similar (Wolters et al. 2006; Lewandowski et al. 2009). At small to intermediate spatial scales  
94 of study, cross-taxonomic congruency of species richness was found to be weak in some studies (e.g.,  
95 Kati et al. 2004; Lovell et al. 2007) but strong in others (e.g., Negi & Gadgil 2002). Such divergent  
96 findings are further exacerbated by the fact that these surrogates are often used to predict occurrence and  
97 diversity of target taxa with different ecological attributes (e.g. dispersal ability, habitat requirements, life  
98 histories) (e.g., Ricketts et al. 1999). Second, many studies testing surrogacy relationships with respect to  
99 a biodiversity target are not clearly defined within a theoretical framework, thus weakening the ecological  
100 basis for using a surrogate (Belovsky *et al.* 2004; Lindenmayer & Likens 2010). Many studies emphasise  
101 the identification of cross-taxonomic surrogate associations, but fail to define the surrogate relationships  
102 clearly, or under a robust framework that incorporate cause-effect relationships and predictive strength  
103 (Barton et al. 2015). Others like Hunter et al. (2016) has pointed out controversies arising from surrogate  
104 concept as a result of differing goals of surrogate application in conservation. Third, many studies of  
105 surrogates are ‘snapshot’ investigations and fail to tackle the problem of how species surrogates perform  
106 over time, or with respect to temporal variability in ecological processes (Anderson 2001; Favreau et al.  
107 2006; Magurran et al. 2010). For any biodiversity surrogate to function as a useful tool for conservation,  
108 it should consistently predict diversity patterns or responses of other species over time (Rodrigues et al.  
109 2000). Understanding of how biodiversity surrogates perform over time (Favreau et al. 2006) is  
110 constrained by the paucity of long-term datasets, with the result that few studies (e.g., Thomson et al.  
111 2007) have examined how long-term shifts in the composition of animal communities associated with  
112 landscape modification may affect cross-taxonomic congruency (see Table 1 for definitions).

113  
114 Biodiversity patterns in general, and individual species in particular, respond to the extent of landscape  
115 modification in different and diverse ways (Fischer & Lindenmayer 2007). Typically, modification of a  
116 landscape leads to changes in habitat spatial configuration and structure (e.g., patch size, matrix quality,  
117 edge effects), which impact animal communities differently, depending on individual species’ ecological

118 needs and their ability to disperse across the wider landscape (Dormann et al. 2007; Driscoll et al. 2013).  
119 Over time, species composition in a biotic community can be affected by dynamic changes in landscape  
120 configuration and vegetation structure or habitat recovery post-disturbance (e.g., Guedo & Lamb 2013).  
121 While it remains unclear how shifts in community composition of one taxon changes relative to other  
122 taxa, a taxonomic group can act as a good surrogate for others if it undergoes turnover (see Table 1 for  
123 definition) in species richness or compositional patterns that are consistent and congruent with other taxa  
124 over space and time. For instance, strong patterns of congruency between turnover of invertebrate and  
125 macroalgal diversity highlight the potential of macroalgae assemblages to act as biodiversity surrogates  
126 for fish and invertebrates (Thomson et al. 2014)

127  
128 In this study, we investigated temporal variation in cross-taxonomic congruency (see Table 1 for  
129 definitions) of diversity between pairs of three taxa, and explored how, (a) temporal shifts in diversity  
130 and, (b) habitat correlates specific to each animal taxon can drive variation in the extent of cross-  
131 taxonomic congruency. We used a large dataset that has been collected over a period of 15 years in a  
132 dynamic, human-modified landscape that has undergone rapid transformation from a woodland-  
133 agriculture mosaic to large tracts of pine monoculture (Lindenmayer et al. 2001; Lindenmayer et al.  
134 2008). We focussed on birds, mammals and reptiles as these taxa are not only frequently used in  
135 conservation assessments (e.g., Westgate et al. 2014), but are also species-rich in our study landscape  
136 (See Table A6, A7, A8 for list of species). In addition, sampling these three taxa demands very different  
137 amounts of effort and resources given the nature of field surveys. For instance, birds can be easily  
138 surveyed and have found to be popular and cost-effective surrogates in inventories of biodiversity (e.g.,  
139 Lawton et al. 1998; Gardner et al. 2008) whereas sampling reptile diversity not only involves a very  
140 different methodology, but also demands specialist knowledge (e.g., McDiarmid et al. 2011). For  
141 mammals, the nocturnal habits and cryptic behaviour of many species (e.g., Suter *et al.* 2000) means  
142 effort-intensive night surveys and baited traps are needed to survey them. Differences in natural history  
143 across taxa, and disparate sampling effort to be invested in different taxonomic groups underscores the

144 need for viable biodiversity surrogates, which could facilitate more optimal use of resources in  
145 inventorying biodiversity.

146  
147 The aim of our study was to evaluate congruence in diversity and species composition measures between  
148 birds, mammals and reptiles over time, and thus uncover evidence for consistent cross-taxonomic  
149 surrogacy (Table 1 for definitions), as the quantification of cross-taxonomic congruency is a critical step  
150 in identifying surrogates (Gioria et al. 2011) . To quantify cross-taxonomic congruency, we used metrics  
151 of correlation between species richness and species composition, as both measures are frequently adopted  
152 in studies of cross-taxon surrogates (e.g., Kati et al. 2004; Sauberer et al. 2004; Gaspar et al. 2010; Cabra-  
153 García et al. 2012) and collectively can offer a comprehensive evaluation of cross-taxonomic congruency  
154 (Su et al. 2004; Gioria et al. 2011). To address our study aims, we posed three questions:

- 155 1. Based on the strength and direction of associations between pairs of taxa, what is the extent of  
156 variation in cross-taxonomic congruence patterns at the species richness and composition levels  
157 over 15 years?

158 Given the limited vagility of reptiles, small spatial requirements (Stow et al. 2014) and the limited effect  
159 posed by habitat fragmentation on lizard communities (e.g., Jellinek et al. 2004) compared to birds or  
160 mammals, we predicted that reptiles were likely to show low congruency in diversity patterns with either  
161 mammals or birds.

162  
163 Cross-taxonomic congruency patterns are often derived from measures of diversity and thus determined  
164 by temporal shifts in the diversity of different taxonomic groups relative to each other. To, (a) explore the  
165 extent of temporal variation in diversity across the taxonomic groups and, (b) determine how different  
166 habitat structural variables in remnant woodlands can influence each taxon in our study, we asked:

- 167 2. In terms of species richness, abundance and composition, what is the extent of temporal change in  
168 three animal taxa over 15 years?

169 3. Are the habitat structure variables that best predict patterns of species composition common to all  
170 three taxa?

171 Based on our findings, we discuss how variation in the predictive strength of surrogates for other aspects  
172 of biodiversity (e.g., other taxonomic groups) can be influenced by taxon-specific temporal shifts and  
173 habitat conditions, as well as implications for the use of cross-taxonomic surrogates in conservation  
174 assessments, inventorying and monitoring.

175

## 176 **2. Materials and methods**

177

### 178 *2.1. Study region*

179 Our study was conducted in the Nanangroe region (34°57'54"S, 148°28'46"E) near Jugiong and Gundagai,  
180 Central New South Wales, Australia. Nanangroe is a dynamic landscape spanning c. 30,000 ha of  
181 agricultural (i.e., grazing) land and exotic tree plantations. Nanangroe was established as a long-term  
182 natural experiment to understand how animal communities respond to differing landscape treatments over  
183 time (Lindenmayer et al. 2008). Much of the original *Eucalyptus*-dominated, box-gum grassy woodland  
184 landscape has been cleared for agriculture in the past two centuries (Yates & Hobbs 1997), leaving what  
185 is best described as a variegated landscape consisting of distinct patches and strips of remnant woodlands  
186 of varying tree densities (McIntyre & Barrett 1992) surrounded by a larger matrix of pastures grazed by  
187 livestock. These woodland remnants are dominated by five *Eucalyptus* species: white box (*E. albens*), red  
188 box (*E. polyanthemos*), yellow box (*E. melliodora*), red stringybark (*E. macrorhyncha*) and Blakely's red  
189 gum (*E. blakelyi*), while the understorey supports a diverse community of native and introduced grasses  
190 and forbs.

191

192 Prior to the commencement of the Nanangroe Natural Experiment in 1999, 52 Eucalypt-woodland  
193 remnants were identified using two landscape contexts and four patch sizes classes (0.5-0.9 ha; 1.0-2.4  
194 ha; 2.5-4.9 ha; 5.0-10 ha). In 1998, the agricultural matrix landscape surrounding these 52 woodland



195 remnants was transformed by the establishment of dense plantations of the exotic Monterey Pine *Pinus*  
196 *radiata* (hereafter these remnants are referred to as “woodland remnants in pine matrix”). In addition,  
197 sampling points in 56 patches of Eucalypt-woodland remnants of broadly similar vegetation classes and  
198 areas were established in surrounding agricultural land (hereafter these remnants are referred to as  
199 “woodland remnants in agricultural matrix”), mostly on farms under private ownership (see Table A1 in  
200 the supplementary material for definitions on landscape contexts). Additionally, 10 sites in cleared and  
201 grazed paddocks and 10 sites in pine plantations were established as “controls”. Inclusive of these two  
202 sets of control sites, there were a total of four landscape contexts examined in our study.

203

## 204 2.2. Animal sampling

205 Permanent transects were marked and established at all 128 study sites prior to the commencement of the  
206 study in 1999. In woodland remnants exceeding one hectare in area, a straight 200m long transect was  
207 established. For a few small remnants less than one hectare in area, a ‘dog-legged’ 200m or 150m transect  
208 was established.

209

210 We sampled bird diversity and abundance at each site using three, five-minute point counts along each  
211 transect, which were conducted between 05:00–10:00hrs during early-middle spring (October-  
212 November). At each point count, observers recorded the numbers of individual species heard or seen  
213 within a 50m radius. Each point was re-sampled by a different observer on another day during the survey  
214 period to minimize bias as a result of weather and variable detection skills by different observers. Bird  
215 surveys were conducted in the years, 1999, 2000, 2001, 2003, 2005, 2007, 2009, 2011, 2012 and 2013.

216

217 To survey reptile abundance and diversity, we conducted standardised, area-constrained searches at two  
218 points along each transect once a year between late winter to early spring (October–November). During  
219 the establishment of the transects, artificial substrates consisting of corrugated metal sheets (c. 1.0m x  
220 1.0m), hardwood timber sleepers (c. 1.0m long, 0.2m thick) and roof tiles (c. 0.3m x 0.3m) were placed at

221 the 0m and 100m points along each transect to simulate microhabitats for small terrestrial reptiles like  
222 snakes, skinks and other lizards. Active searches for reptiles were completed by turning over logs, rocks  
223 and the artificial substrates throughout the sites. Standardised reptile surveys were conducted in the years:  
224 1999, 2000, 2001, 2003, 2006, 2011 and 2013.

225

226 Finally, we surveyed mammal diversity and abundance using standardised, nocturnal spotlighting  
227 searches along each transect, on nights of good weather (i.e. no rain, storms). Mammal spotlight surveys  
228 were conducted in the years: 1999, 2001, 2004, 2005, 2009, 2011 and 2013. Additional details on our  
229 mammal and reptile surveys have been described in Lindenmayer et al. (2001) and (2008).

230

### 231 *2.3. Vegetation sampling*

232 To describe the habitat structure at each study site, we conducted vegetation surveys at all study sites once  
233 every four years. A total of 34 vegetation variables was measured at each site to capture the variation in  
234 vegetation structure from the ground to the canopy. We averaged measures taken from each of three  
235 sampling points to obtain mean values for all habitat structure variables at every site. A full list of the  
236 vegetation variables is available in the supplementary information section (Table A2).

237

### 238 *2.4. Data analysis*

239

#### 240 *2.4.1. Data selection*

241 We used species data from surveys of birds, mammals and reptiles completed in 1999, 2001, 2011 and  
242 2013. Each of these years were selected for our analysis as they included data where all three taxonomic  
243 groups were simultaneously surveyed in the same year and season, and therefore minimized the influence  
244 of temporal effects on our dataset.

245

246 2.4.2. Tests for correlations of species richness between different taxa over time (Question 1)

247 We used Spearman's rank correlations to test for cross-taxonomic congruence in species richness patterns  
248 over time between pair-wise combinations of the three taxa for each of four study years and each  
249 landscape context class (including both control sites). The strength of correlation of species richness  
250 between two taxa is often used as a proxy of cross-taxonomic associations (e.g., Hess et al. 2006; Wolters  
251 et al. 2006). Spearman's correlation was chosen over Pearson's correlation as the metric of correlation  
252 strength as species richness was relatively low across sites, particularly for mammals, and is thus likely to  
253 be distributed non-normally. We also calculated correlations between birds, and pooled species richness  
254 of mammals and reptiles combined. Using 1,000 bootstrap replicates, we calculated the 95% confidence  
255 interval for all Spearman's correlations. The strength of the Spearman's correlation coefficient ( $\rho$ ), which  
256 is used as a measure of congruency of species richness between two taxa was interpreted as follows:  
257 correlation values of  $\geq 0.50$  were considered to be strong, between 0.10 to 0.30 to be moderate, and  
258 correlations  $\leq 0.10$  to be weak (see Lamoreux et al. 2006).

259

260 2.4.3. Test for correlations of species composition over time (Question 1)

261 We used partial Mantel tests to investigate the strength of cross-taxonomic congruence in species  
262 composition between pair-wise combinations of animal taxa for each year of four study years. Partial  
263 Mantel tests were used because the data were not independent and Mantel tests are able to address the  
264 problem of partial dependence in dissimilarity matrices (Legendre & Legendre 1998), and have  
265 previously been used to identify correlations between pairs of taxa (e.g., Su et al. 2004; Gioria et al. 2011;  
266 Gaspar et al. 2012). Abundance values for all species were square-root transformed to reduce the potential  
267 over-influence of highly abundant species on among-site dissimilarity values. We quantified species  
268 composition using Bray-Curtis dissimilarity metric between pairs of sites for all landscapes contexts. The

269 advantage of partial Mantel tests over simple Mantel tests is that they can measure the correlation  
270 between two matrices (Paszkowski & Tonn 2000; Su et al. 2004) after considering variation associated  
271 with a matrix of spatial (Euclidean) distances, thus accounting for potential problems of spatial  
272 autocorrelation. Significance of all partial Mantel tests was assessed using a Monte Carlo procedure with  
273 999 permutations. Mantel and Spearman's correlations were implemented using the 'ecodist' package in  
274 R version 1.2.9, while confidence intervals for Spearman's correlations were estimated using 1,000  
275 bootstraps in the 'RVAideMemoire' package (R Development Core Team 2013).

276

#### 277 2.4.4. Test of species composition of two taxa as predictors over time (Question 1)

278 We completed multiple regressions on distance matrices (MRM) (Lichstein 2007) to test if species  
279 composition of two taxa based on dissimilarity matrices can collectively better predict composition of a  
280 target taxa selected *a priori*. Unlike partial Mantel tests which are limited to comparing pairs of taxa, this  
281 approach allows multiple taxa to be used as predictor variables. MRM involves regressing the response  
282 matrix using more than one explanatory matrix, while each matrix contains all combinations of pair-wise  
283 distances between  $n$  number of sample units. We chose not to use bird data as the response variable in any  
284 of our MRM models. This was because birds are usually *the* surrogate taxon in conservation of other  
285 components of biodiversity (e.g., Blair 1999; Sauberer et al. 2004; Larsen et al. 2012) given the relative  
286 ease of collecting bird data compared to data of other taxa. Additionally, we factored geographic distance  
287 into our models as a predictor matrix, since spatial data derived from geographic coordinates are often  
288 available along with species datasets and can be used to reveal ecologically meaningful effects (e.g.,  
289 strong spatial influences on composition may reveal dispersal limitations imposed by space).

290

291 We constructed a set of candidate models using all possible combinations of bird, reptile and mammal  
292 composition and spatial distances as predictor variables, while only mammal or reptile composition was

293 treated as the response. MRM analysis was completed only for species data collected from woodland  
294 remnant sites in pine and agricultural sites as there were too few data for analysis in the control sites due  
295 to low species abundance and richness. As with our partial Mantel tests, we square-root transformed the  
296 animal count data, and used the Bray-Curtis metric to calculate pair-wise species dissimilarity. The  
297 statistical significance of each MRM model was assessed with 999 permutations.

298

#### 299 2.4.5. Analysis of shifts in animal communities over time (Question 2)

300 We plotted site-level, mean species richness and mean abundance for each taxon in both landscape  
301 contexts and control sites to assess temporal changes in species richness and abundance over the four  
302 study years. To visualise changes in community composition between the four landscape contexts over  
303 the four study years, we first performed non-metric multidimensional scaling (NMDS) analysis using the  
304 R function ‘metaMDS’ to ordinate site counts in species space for all three groups and the two main  
305 landscape contexts (woodland remnants in agriculture and pine). For each landscape context, all  
306 ordinations of each taxon were presented together in each plot, but separated by year using coloured  
307 polygons. We then used the multiple response permutation procedure (MRPP) as a non-parametric test for  
308 significant differences in species compositional changes over time. MRPP generates the effect size  
309 statistic  $A$ , which provides a measure of within-group heterogeneity, and a measure of significance  $P$ . The  
310 significance of the effect size  $A$  was assessed using 999 permutations. Bray-Curtis dissimilarity was used  
311 as the measure for species composition in both the NMDS and MRPP analyses.

312

313 To explore how different habitat structural variables influenced each of the three taxa in ordination space,  
314 we fitted vectors for all habitat variables measured in each ordination, to identify those that were  
315 significantly correlated to the two NMDS axes for each taxon. The R function ‘envfit’ available in the  
316 vegan package computes vectors or factor averages of environmental variables fitted to the ordination

317 matrix. The significance of these fitted vectors was then assessed using 999 permutations. Habitat  
318 correlates that were significant at  $P < 0.05$ , and marginally significant  $0.05 < P \leq 0.1$  were retained for  
319 further consideration.

320

321 2.4.6. Evaluating the influence(s) of habitat structural correlates on animal communities (Question 3)

322 We were interested in identifying habitat structural variables consistently associated with species  
323 composition among the three taxa in the Eucalypt-woodland remnants. We constructed a series of  
324 candidate ‘global’ models using multiple regressions on distance matrices for each taxa, and using the full  
325 set of habitat structural variables to explore how the different variables influenced each taxon. Only  
326 habitat variables not strongly correlated with others (Pearson’s  $r < 0.5$ ) were retained in the MRM  
327 analysis after an initial screening of the full set of variables in a correlogram matrix. NMDS and MRPP  
328 analyses were completed using the ‘vegan’ package in R version 2.2-1 (R Development Core Team 2013)  
329 while MRM analysis was carried out using the ‘ecodist’ package in R version 1.2.9 (R Development Core  
330 Team 2013)

331

### 332 **3. Results**

333 *3.1. What is the extent of variation in cross-taxonomic congruence patterns at the richness and*  
334 *composition over 15 years (Question 1)?*

335

336 3.1.1. Change in correlations of species richness over 15 years

337 We found that correlations of species richness varied between different pairs of taxa and across landscape  
338 contexts, but increased in strength and significance over the 15 years (Figure 1, Table A3). In woodland  
339 remnants in the agricultural matrix, species richness was weakly and negatively correlated between

340 reptiles and birds, but none of these correlations were significant (see supplementary material). Mammal  
341 species richness was weakly and negatively correlated with that of birds in 1999, but the correlations  
342 became positive and strengthened over time, with mammal species richness being significantly correlated  
343 with bird species richness in 2011 and 2013 (Spearman's  $\rho = 0.306$  with  $P = 0.022$ ; Spearman's  $\rho = 0.350$   
344 with  $P = 0.01$ ) but not in 1999 and 2001. Additionally, a linear model relating year to correlations of  
345 species richness for bird–mammal congruency was significant (model adjusted  $R$ -square = 0.998,  
346 coefficient estimate = 0.0297,  $P = 0.0007$ ). Species richness correlations between birds and reptiles, and  
347 pooled mammal and reptile richness were weak and insignificant for all years except in 2011 (Spearman's  
348  $\rho = 0.308$  with  $P = 0.022$ ).

349  
350 In woodland remnants in the pine plantation matrix, bird species richness was consistently and positively  
351 correlated with that of mammals, and the strength of these correlations increased with time, with  
352 correlations in 2013 being marginally significant (Spearman's  $\rho = 0.277$  with  $P = 0.065$ ). In addition, bird  
353 species richness was positively correlated with pooled mammal and reptile species richness in later years,  
354 being significantly so in 2013 (Spearman's  $\rho = 0.300$  with  $P = 0.04$ ).

355  
356 3.1.2. Change in correlations of species composition over 15 years

357 We found that partial Mantel correlations between distance matrices of animal groups were often weak  
358 and insignificant (Table 2, Figure 2). For instance, in woodland remnants in the agricultural matrix, bird  
359 and reptile composition was negatively correlated in all study years except in 2001. Bird and mammal  
360 composition were mostly positively correlated over the four study years, although only correlations in  
361 later years – 2011 and 2013 were moderately strong and significant (Mantel  $R = 0.306$  with  $P = 0.002$ ;  
362 Mantel  $R = 0.168$  with  $P = 0.008$ ). None of the correlations between reptile and mammal composition  
363 were strong or significant, and fluctuated between being weakly positive and negative over time.

364

365 In woodland remnants in the pine plantation matrix, bird and reptile composition were positively  
366 correlated only in 1999 (Mantel  $R = 0.1912$  with  $P = 0.035$ ), but negatively correlated in all other years.  
367 Although consistently positive, we found that correlations of bird and mammal composition were weak  
368 and insignificant across all study years except 2001 (Mantel  $R = 0.279$  with  $P = 0.012$ ). None of the  
369 correlations between mammals and reptiles were significant, and were mostly negative. Overall, we found  
370 that while correlations involving reptiles were usually negative and weak (Figure 2), correlations between  
371 mammal and bird species composition were consistently positive, and appeared to have strengthened over  
372 time, at least for woodland remnants in the agricultural matrix (2011 Mantel  $R = 0.306$  with  $P = 0.002$ ).  
373 Such a trend did not apply for woodland remnants in the pine matrix, as correlation strength peaked in  
374 2001, but declined thereafter.

375

### 376 3.1.3. Change in predictive strength of two taxa for a single target animal group over 15 years

377 We found that MRM models incorporating distance matrices of birds, reptiles and spatial distances were  
378 able to predict mammalian composition, albeit weakly for woodland remnants sites in the agricultural  
379 matrix, but relationships declined in predictive strength between 2011 ( $R^2 = 0.182$ ) and 2013 ( $R^2 =$   
380  $0.0565$ ) (Table 3). Of three explanatory variables including reptile composition and spatial distances, bird  
381 species composition explained 85.9% of the variation in mammal species composition in 2011 but only  
382 42.5% in 2013, although bird composition remained significant as a predictor in both years. Reptile  
383 composition and spatial distance were weak and non-significant predictors in all candidate models  
384 explaining mammal composition in 2011 and 2013. For candidate models using birds, mammals and  
385 spatial distance to predict reptile composition, bird and mammal composition never emerged as  
386 significant predictor, being weakly but positively correlated in most years. However, spatial distance  
387 appeared to be a significant and relatively important predictor of reptile composition, explaining 48.3%



388 and 55.9% of the variation of reptile composition in 2011 and 2013 respectively (Table 3).

389

390 All candidate models incorporating species compositional and spatial distances for woodland remnants in  
391 the pine matrix explained very little variation in either reptile or mammal composition. Although bird  
392 species composition explained 31.0% of the variation in mammalian composition in 2013, it was not a  
393 significant predictor in other years, and in fact was negatively correlated in 2011. Neither mammal nor  
394 bird species composition with spatial distances were useful predictors of reptile species composition in  
395 woodland remnants in the pine matrix, although mammal species composition was marginally significant  
396 as a predictor in 2011 and 2013 ( $0.05 < P < 0.1$ ).

397

398 *3.2. What is the extent of temporal changes in three animal communities over 15 years (Question 2)?*

399 *3.2.1. Changes in species richness and abundance of three taxa over 15 years*

400 Across the study landscape, we found that mammal and reptile species richness showed clear increases  
401 over the four study years, while bird species richness increased marginally between 1999 and 2011, but  
402 declined in 2013 (Figure 3, Supplementary Tables A6, A7, A8). Species richness and abundance, and  
403 their change over 15 years in both pine and agricultural control sites were limited especially for mammals  
404 and reptiles, and consistently lower than corresponding woodland sites in either landscape contexts. At a  
405 site level, we found weak and insignificant patterns of change in bird species richness for woodland  
406 remnants in the pine matrix over time (Figure 3) while mean site abundance increased from 57.7 to 68.9  
407 individuals (Mann-Whitney  $U = 693$ ,  $Z = 1.65$ ,  $P > 0.05$ ). By comparison, bird species richness in  
408 woodland remnants in the agricultural matrix increased more rapidly over time, from 12.1 species in 1999  
409 to 15.0 species per site (Mann-Whitney  $U = 971$ ,  $Z = -3.231$ ,  $P < 0.05$ ) in 2013. The trends in reptile  
410 richness and abundance over time for woodland remnants in the pine matrix were less clear compared to  
411 those in the agricultural matrix, but changed somewhat faster, and were significant. For example, mean

412 reptile species richness in woodland remnants in agriculture increased from 0.357 species per site in 1999  
413 to 2.52 species in 2013 (Mann-Whitney  $U = 244$ ,  $Z = 5.771$ ,  $P < 0.001$ ) while mean reptile richness in  
414 woodland remnants in pine increased from 0.54 to 2.33 species over the same period (Mann-Whitney  $U =$   
415  $971$ ,  $Z = -3.231$ ,  $P < 0.001$ ). Unlike birds or reptiles, both mammal species richness and abundance  
416 showed consistent increases in the two landscape contexts over the study period. For example, mean  
417 mammal species richness for woodland remnants in the pine matrix increased from 0.475 to 1.27 species  
418 in 2013 (Mann-Whitney  $U = 517$ ,  $Z = 3.368$ ,  $P < 0.001$ ), while mean abundance increased from 0.775  
419 individuals in 1999 to 2.4 individuals (Mann-Whitney  $U = 531$ ,  $Z = 3.245$ ,  $P < 0.05$ ). Likewise, mean  
420 mammal richness for woodland remnants in the agricultural matrix doubled over the same period, from  
421 0.518 species in 1999 to 1.03 species per site in 2013 (Mann-Whitney  $U = 1076.5$ ,  $Z = -2.6009$ ,  $P < 0.01$ ).

422

### 423 3.2.2. Changes in community composition of three taxa over 15 years

424

425 In woodland remnants in the pine matrix, we found that points representing mammal species composition  
426 in ordination space clustered towards the negative end of NMDS axis 1 in 1999, but became less clustered  
427 in subsequent years, and shifted positively along the axis (Figure 4a). The MRPP results indicated that  
428 mammal assemblages differed over the four years ( $A = 0.0464$ ,  $P < 0.01$ ). Points representing reptiles  
429 were well spread in ordination space in 1999 (Figure 4b), but became increasingly clustered towards the  
430 positive end of NMDS axis 1 in later years, with these changes in species assemblage being significantly  
431 different over time ( $A = 0.0504$ ,  $P < 0.01$ ). Similarly, points representing birds were sparsely clustered in  
432 1999, but subsequently clustered closely towards the negative end of NMDS axis 1 in 2011 and 2013  
433 (Figure 4c). Such a change in the bird assemblage over time was also found to be significant in our MRPP  
434 analysis ( $A = 0.0741$ ,  $P < 0.01$ ) and suggests that bird composition in these woodland remnants were  
435 become increasingly similar over the 15 years.

436

437 For woodland remnants in the agriculture matrix, points representing mammal species were sparsely  
438 clustered in ordination space, and appeared to be even less so in 2013 and 2011 than in 1999 and 2001  
439 (Figure 4d). Although changes in the mammal assemblage over time were significant, they were weaker  
440 than the species compositional changes observed for the other two groups ( $A = 0.0193$ ,  $P < 0.05$ ), and for  
441 all animal groups in woodland remnants in the pine matrix.

442

443 We did not plot the ordinations for reptiles due to the very large scatter of points. However, we noted that  
444 reptile assemblages in woodland remnants in the agricultural matrix changed significantly over 15 years  
445 ( $A = 0.103$   $P < 0.001$ ) (see also Figure A2 in the supplementary material for cluster dendrograms  
446 representing differences in Bray-Curtis dissimilarity across the study years). Likewise, the bird  
447 assemblage in these woodland remnants differed significantly over 15 years ( $A = 0.0222$ ,  $P < 0.001$ ). The  
448 change in bird species composition is shown in the positive shift in clusters of points representing bird  
449 species composition in ordination space along NMDS axis 1 (Figure 4e).

450

451 *3.3. Are the habitat variables that drive shifts in species composition over time shared among the three*  
452 *taxa (Question 3)?*

453

454 3.3.1. Significant habitat structure variables correlated with each taxonomic group

455

456 We found that the habitat variables strongly correlated with species communities differed among taxa and  
457 between woodland remnants in the two key landscape contexts (Table 4, see also Table A5), and few  
458 variables were shared. Bird species composition was correlated with more habitat structure variables than

459 either reptiles or mammals for woodland remnants in both landscape contexts. ‘Blackberry cover’ was a  
460 recurrent explanatory variable for birds and reptiles, correlating strongly with at least one NMDS axis for  
461 each group. In woodland remnants surrounded by pine, ‘crown structure’ ( $R^2 = 0.163, P < 0.05$ ) and  
462 ‘basal count’ ( $R^2 = 0.272, P < 0.01$ ) were moderately and significantly correlated with both NMDS axes  
463 for birds, while ‘blackberry cover’ and ‘woodland strata’ appeared only weakly correlated. While  
464 ‘blackberry cover’ ( $R^2 = 0.292, P < 0.01$ ), ‘dominant cover’ ( $R^2 = 0.226, P < 0.05$ ) and ‘shrub stem count’  
465 ( $R^2 = 0.154, P < 0.05$ ) appeared to be important correlates for reptile species composition in woodland  
466 remnants surrounded by pine, we found that no habitat variables were strongly and significantly  
467 correlated to either NMDS axes for mammals.

468

469 Bird species composition in the woodland remnants in the agricultural matrix was significantly correlated  
470 with eight habitat structure variables, with two of these variables shared with mammals (‘exposed rock’,  
471 ‘blackberry cover’). While reptiles were found not to be significantly correlated with any habitat variables  
472 in the landscape contexts, we found that mammal species composition in woodland remnants in the  
473 agricultural matrix was strongly correlated with four variables, with ‘blackberry cover’ ( $R^2 = 1.00, P <$   
474  $0.05$ ) again being very strongly and positively correlated with NMDS axis 1, and negatively with NMDS  
475 axis 2.

476

## 477 **4. Discussion**

### 478 *4.1. Overview*

479 We assessed the strength and direction of cross-taxonomic correlations in species richness and  
480 composition between three vertebrate taxa that feature frequently in conservation assessments (e.g.,  
481 Lawton et al. 1998; Schulze et al. 2004; Westgate et al. 2014). We then compared these associations over  
482 each of four study years spread over 15 years to assess whether cross-taxonomic congruency was

483 consistent over time, a requisite of a good biodiversity surrogate. Below, we discuss our key findings and  
484 outline some of the implications of our research for the use of cross-taxonomic surrogates in dynamic  
485 landscapes undergoing rapid transformation.

486

#### 487 4.2. Variation in cross-taxonomic congruency patterns over time

488 Whether based on a species richness or a species composition approach, we found that the strength of  
489 congruency between pairs of animal taxa varied with the taxon examined, landscape context, and over  
490 time (Question 1). Between pairs of taxa, we found that correlations from both approaches ranged from  
491 being very weak to moderate, and that correlations in either species richness or composition tended to be  
492 positive and stronger between birds and mammals than between either group and reptiles. We also found  
493 that species richness and composition correlations increased in strength over time for woodland remnants  
494 in both agricultural and pine matrixes. The prevalence of stronger and significant associations between  
495 taxa in woodland remnants in the agricultural matrix in the later years compared to woodland remnants in  
496 the pine matrix underscores the role played by the matrix in shaping animal communities in remnant  
497 woodland patches (e.g., Ricketts 2001), possibly by influencing the dispersal of different species (e.g.,  
498 Dormann et al. 2007; Driscoll et al. 2013). For instance, the stronger cross-taxonomic associations may  
499 arise from greater dispersal into, and out of these woodland patches by species in all three taxa through  
500 the comparatively more open agricultural matrix. Additionally, the effects of the pine plantation matrix on  
501 animal communities in the Eucalypt-woodland patches embedded within may be further accentuated by  
502 the limited food resources available (e.g. flowering plants, arthropods) and a different set of microclimatic  
503 conditions resulting from the dense pine cover.

504

#### 505 4.3. Change in species richness and composition of animal communities over time

506 We found that species richness and abundances of mammals and reptiles showed larger shifts than birds  
507 over time for woodland remnants in both pine and agricultural matrixes. We also found that the extent of  
508 temporal change in species composition differed with taxa, and between the two major landscape contexts  
509 (Question 2). Our findings of taxa-specific shifts in diversity and abundance here mirror the variation in  
510 congruency across taxa described earlier, and add yet another line of evidence to the influences exerted by  
511 the landscape matrix on shaping the animal communities occurring within these habitat patches. It is  
512 likely that woodland remnants in the pine matrix showed lower cross-taxonomic congruence a decade  
513 after the initial disturbance period (when pine monoculture was established) because the dense pine  
514 plantation matrix may have acted as a barrier to the dispersing reptiles and mammals (Mortelliti et al.  
515 2014), thus influencing species richness and composition of both taxa over time. Our finding here  
516 highlights the problem posed by differential turnover in species diversity across taxonomic groups to  
517 cross-taxonomic surrogacy because it compromises the temporal consistency required if these surrogates  
518 are to be used in conservation monitoring and biodiversity assessments.

519

#### 520 4.4. Differing habitat structure variables correlated with animal taxa

521 Our analyses of the influence of habitat structure variables on animal taxa indicated that the explanatory  
522 variables that fit best with the NMDS axes were different for each group (Table 3) at the landscape scale,  
523 although one variable was frequently shared (blackberry index). When an MRM approach was used to  
524 evaluate the relative influence of habitat structure correlates, we again found that there were few or no  
525 shared correlates between any two taxa. Other studies of cross-taxonomic surrogates have also reported  
526 such differences of explanatory variables across taxa (e.g., Dauber et al. 2003). Blackberry (*Rubus*  
527 *fruticosus* sp. agg.) is widely recognized as one of the most invasive plant species across Australia  
528 (Dehaan et al. 2013) and has increasingly spread across our study sites. Blackberry forms dense patches in  
529 woodland remnants along creeks in our study sites and is likely to have modified habitats and

530 microclimatic conditions for many terrestrial species. This may account for its strong correlation with the  
531 composition of all three animal taxa as revealed in our ordination analysis (Figure 4).

532

533 We hypothesize that the broadly differing set of correlated habitat structure variables identified in our  
534 analyses is the outcome of divergent habitat requirements of birds, reptiles and mammals at the landscape  
535 scale. Bird species composition was predicted by more habitat variables than reptiles or mammals in both  
536 landscape contexts. This pattern is likely due to the fact that while the majority of reptile species (e.g.  
537 skinks) and mammal species are more affected by habitat structural variables on the ground, bird species  
538 composition are more strongly affected by a larger set of habitat variables associated with trees (e.g. stand  
539 height, number of trees, number of strata), due to the arboreal behaviour of many species (Barton et al.  
540 2014). The differential associations of each taxa with specific sets of habitat attributes and their changing  
541 relationship over time, may explain the weak cross-taxonomic congruency observed in our study, and has  
542 also been highlighted by other studies of cross-taxonomic associations (e.g., Dauber et al. 2003; Azeria et  
543 al. 2009; Heino et al. 2009).

544

#### 545 *4.5. Implications for the use of cross-taxonomic surrogates in conservation*

546 Our findings have several key implications for the use of some vertebrate taxa, particularly birds, as  
547 surrogates or broad indicators for the diversity of other taxa in conservation. First, variation in species  
548 richness and composition over time and among the taxa studied suggests that species richness and  
549 compositional approaches to quantifying surrogates of species diversity should be applied cautiously. Our  
550 finding that stronger cross-taxonomic associations in composition and species richness occurred in  
551 woodland remnants in the agricultural matrix alludes to the role played by the landscape matrix in shaping  
552 animal communities, either by limiting or promoting species dispersal (e.g., Driscoll *et al.* 2013).  
553 Differences in dispersal ability and spatial requirements may have influenced cross-taxonomic

554 associations at the landscape scale, and may explain why both birds and mammals were better correlated  
555 with each other, but were often weakly and negatively correlated with reptiles, which are not only  
556 predominantly terrestrial but less vagile, and thus have smaller spatial requirements (e.g. Stow et al.  
557 2014).

558

559 Second, our finding of stronger and more positive associations between bird and mammal diversity in  
560 both landscape contexts over time suggests that animal communities can become increasingly similar and  
561 more stable, possibly in response to changes in vegetation structure as woodland remnants regenerate and  
562 mature in the years following initial disturbance (e.g., change of the landscape matrix when pines were  
563 planted). Strengthening of these cross-taxonomic relationships may be also paralleled by increases in  
564 mean species richness at the site level for both taxa (Figure 3). Communities in heavily modified  
565 landscapes are likely to show lower community stability and higher temporal turnover in species  
566 composition. However these communities can become more stable with time post-disturbance (Leibold  
567 2009) and with increased overall species diversity (van Ruijven & Berendse 2007). We hypothesise that  
568 increased community stability and higher diversity at the landscape scale may have a role in driving  
569 stronger cross-taxonomic congruency at the species richness and composition levels observed in our study  
570 in 2011 and 2013, and suggest that cross-taxonomic surrogates may not be very useful for assessing  
571 biodiversity in landscapes that have recently been subject to heavy anthropogenic disturbance.

572

573 Third, our findings suggest that high rates of taxa-specific turnover and among-group differences in  
574 habitat correlates, can affect the degree of congruency in diversity patterns between different taxa. For  
575 example, birds showed significant shifts in species composition over time, but with little increase in  
576 richness or abundance. By contrast, the reptile communities showed significant temporal turnover, and  
577 increases in overall diversity and abundance (Figure 3). Differing rates of temporal turnover shown by



578 change in Bray-Curtis dissimilarity over the study period may account for the large variation in the  
579 Mantel correlations over time. While many studies have explored cross-taxonomic congruency using  
580 large sets of species data (e.g., Schulze et al. 2004; Grenyer et al. 2006; Stoch et al. 2009), we note that  
581 few have examined congruency patterns in relation to temporal changes in species richness, abundance  
582 and composition. This temporal problem continues to persist because most surrogate studies are based on  
583 short-termed datasets (Favreau et al. 2006). Therefore, we suggest that strong congruencies observed  
584 between two taxonomic groups at one point in time may be ephemeral, especially in highly disturbed  
585 landscapes undergoing change. Our results thus offer some support to the predictions by Prendergast &  
586 Eversham (1997) that differential responses to the environment (in this case, habitat structure variables),  
587 may be responsible for driving weakly congruent patterns of diversity. From a conservation standpoint,  
588 the use of one or few taxa as cross-taxonomic surrogates, especially birds, is likely to be problematic  
589 since it could inherently fail to represent diversity patterns of other taxa (e.g., Dauber et al. 2003) and  
590 their responses to changing habitat structure (e.g., Barton et al. 2014).

591

#### 592 *4.6. Ecological basis of surrogacy relationships and scope for future research*

593 Snapshot-type studies of cross-taxonomic surrogates are ubiquitous in the literature but lack a temporal  
594 dimension, thus failing to take into consideration ecological processes that take time to manifest (e.g.  
595 Bond 2001, Favreau et al. 2006). Since many ecological patterns and processes are highly dynamic in  
596 time and space (Morgan et al. 1994), short-term studies will inherently fail to capture the temporal  
597 variability of communities and their effects on cross-taxonomic comparisons. Moreover, many such  
598 surrogate studies are also conducted at scales too large for surrogacy patterns to be meaningful for  
599 conservation (Westgate et al. 2014), often at a continental to global scale. However, Grenyer et al. (2006)  
600 and others (e.g. Weibull et al. 2003) have noted that congruency between taxa tends to be highly scale  
601 dependent; levels of congruency may be particularly low if these patterns are measured at the fine spatial

602 resolutions relevant to conservation. There is thus a need for more studies of cross-taxonomic surrogacy  
603 at these fine spatial scales which these surrogates are to be applied.

604

605 Our findings of stronger associations at the species richness and composition level between mammals and  
606 birds, both which are known to be better dispersers and have larger spatial requirements than reptiles,  
607 underscores the role of dispersal and spatial scale in shaping animal communities (e.g., Howeth &  
608 Leibold 2010). These ecological factors needs to be considered when identifying species surrogates for  
609 conservation application in dynamic landscapes. Our findings also raise problems for the efficacy of using  
610 biodiversity surrogates in dynamic, human-modified landscapes because cross-taxonomic congruency  
611 changes over time with temporal shifts in diversity (e.g., Wolters et al. 2006).

612

613 Finally, an immediate goal for ecologists studying indicators of biodiversity should be to identify clearer  
614 links between different taxonomic groups and in relation to underlying ecological processes, to ensure  
615 that taxa used as surrogates are grounded within a more robust, science-driven framework that considers  
616 causal links that allows for validation across spatial and temporal contexts (e.g., Lindenmayer & Likens  
617 2011; Barton et al. 2015). Identifying shared responses and relationships to landscape and habitat  
618 structure variables between species, and between different taxa could be a first step in understanding these  
619 associations in a mechanistic manner. This, in turn, needs to be coupled with a better understanding of  
620 how temporal processes may alter these relationships, although doing so will demand greater investments  
621 into collecting long-term data.

622

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943 **Figure and table annotations**

944 **Figures**

945 **Figure 1.** Plots showing variation in Spearman's  $\rho$  (congruency of species richness) over the four study  
946 years. Error bars represent 95% confidence intervals after 1,000 bootstraps. Diamond-shaped points  
947 represent woodland remnants in pine (treatment) while circle-shaped points represent woodland remnants  
948 in the agricultural matrix. Only the relationship between year and congruency of species richness for bird-  
949 mammal congruency was found to be significant (model adjusted  $R$ -square = 0.998, coefficient estimate =  
950 0.0297,  $P = 0.0007$ )

951

952 **Figure 2.** Plots showing variation in partial Mantel  $R$  (congruency of species composition) over the study  
953 period. Error bars represent the 95% confidence intervals after 999 permutations. Diamond shaped points  
954 represent woodland remnants in the pine matrix (treatment) while circle-shaped points represent  
955 woodland remnants in the agricultural matrix.

956

957 **Figure 3.** Scatterplots showing shifts in mean site species richness and abundance for birds, reptiles and  
958 mammals over the study years spanning 1999 and 2013. (Legend: shaded diamond-shaped points  
959 represent woodland remnants in pine (treatment) while shaded circle-shaped points represent remnants in  
960 agriculture; unshaded diamond- and circle-shaped points represent the control sites in the respective pine  
961 and agricultural matrix)

962

963 **Figure 4.** NMDS ordination plots for (a) mammal, (b) reptile and (c) bird communities in woodland  
964 remnants surrounded in the pine (treatment) matrix, and (d) mammals and (e) birds in woodland remnants  
965 in the agricultural matrix. The ordination plot for reptiles in agricultural woodland remnants is not shown

966 due to its wide scatter of point clusters. Number of dimensions and stress values for all NMDS  
967 ordinations are shown on each plot. (Legend: black diamond – 1999, purple triangle – 2001, blue circle –  
968 2011, red square – 2013)

969

970 **Table 1.** Glossary of selected important terms in the concept of cross-taxonomic surrogacy and their  
971 definitions.

972

973 **Table 2.** Results of partial Mantel correlations of species composition for three taxa over the study period.  
974 Results for pine control (PIN) sites are not presented as there was only adequate species data for one site.

975

976 **Table 3.** Multiple regression in matrix (MRM) models and summary statistics for predictor variables.  
977 Predictor variables included bird, mammal and reptile composition, and geographic space. See Table A4  
978 for model attributes for years 1999 and 2001.

979

980 **Table 4.** Significant habitat structure correlates of bird, reptile and mammals in two different landscape  
981 contexts, identified with non-metric multidimensional scaling.

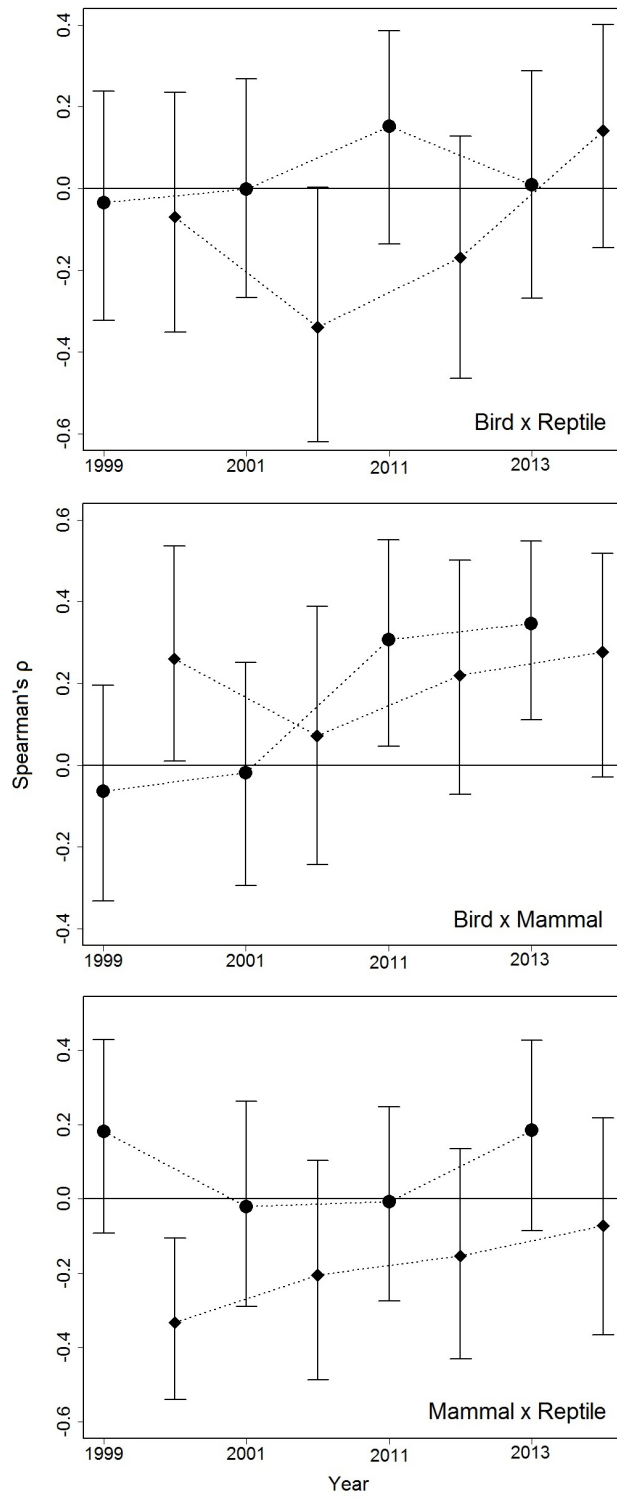
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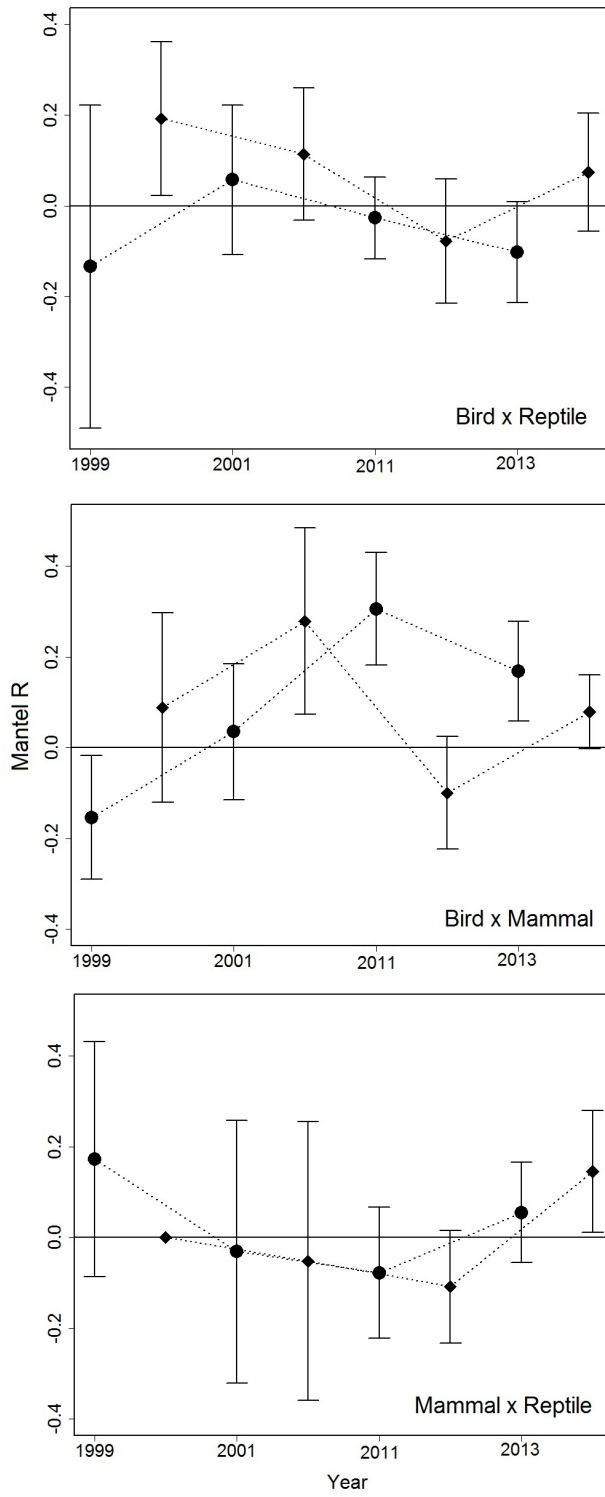
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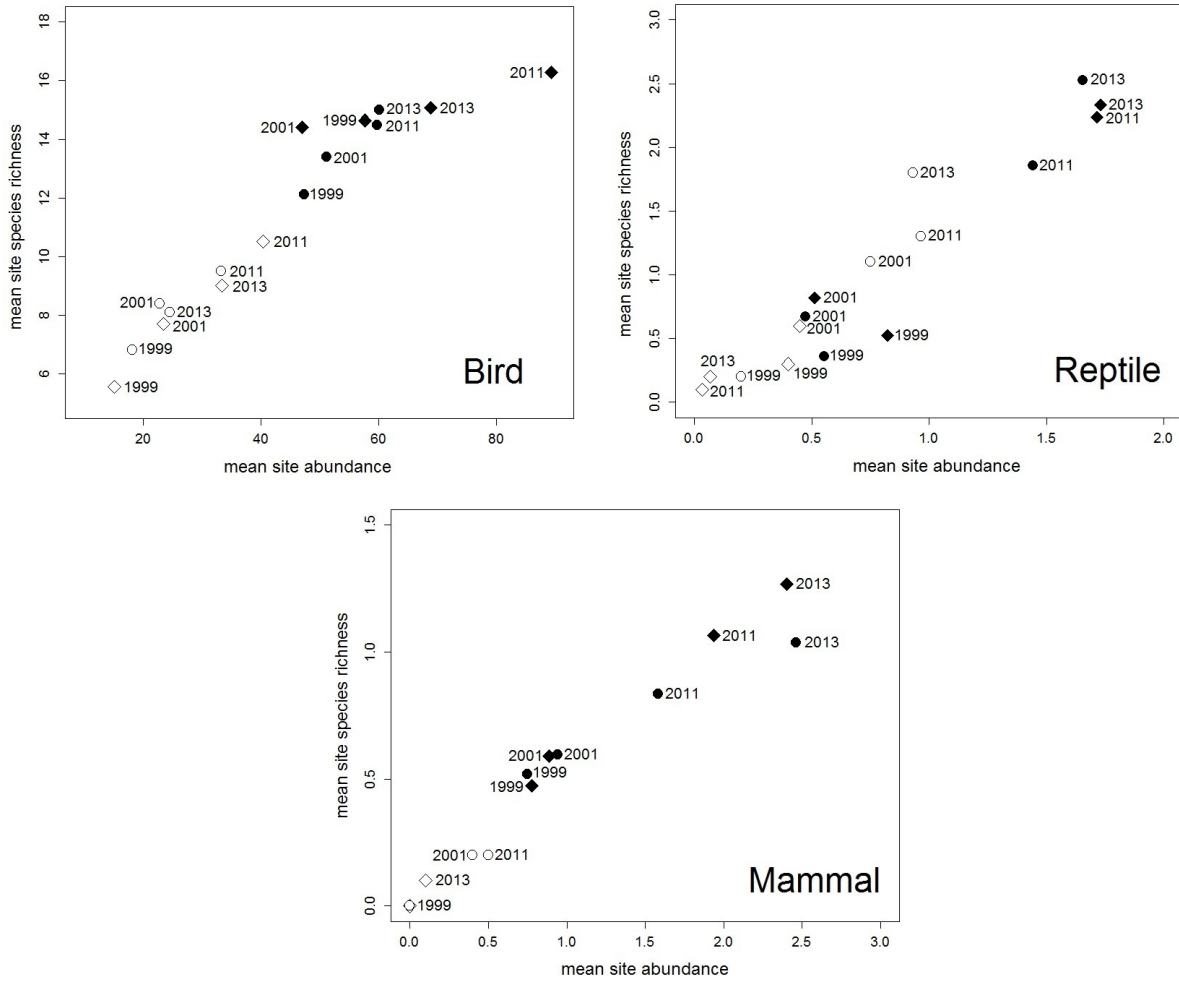
989 **Figure 2**



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992 **Figure 3**



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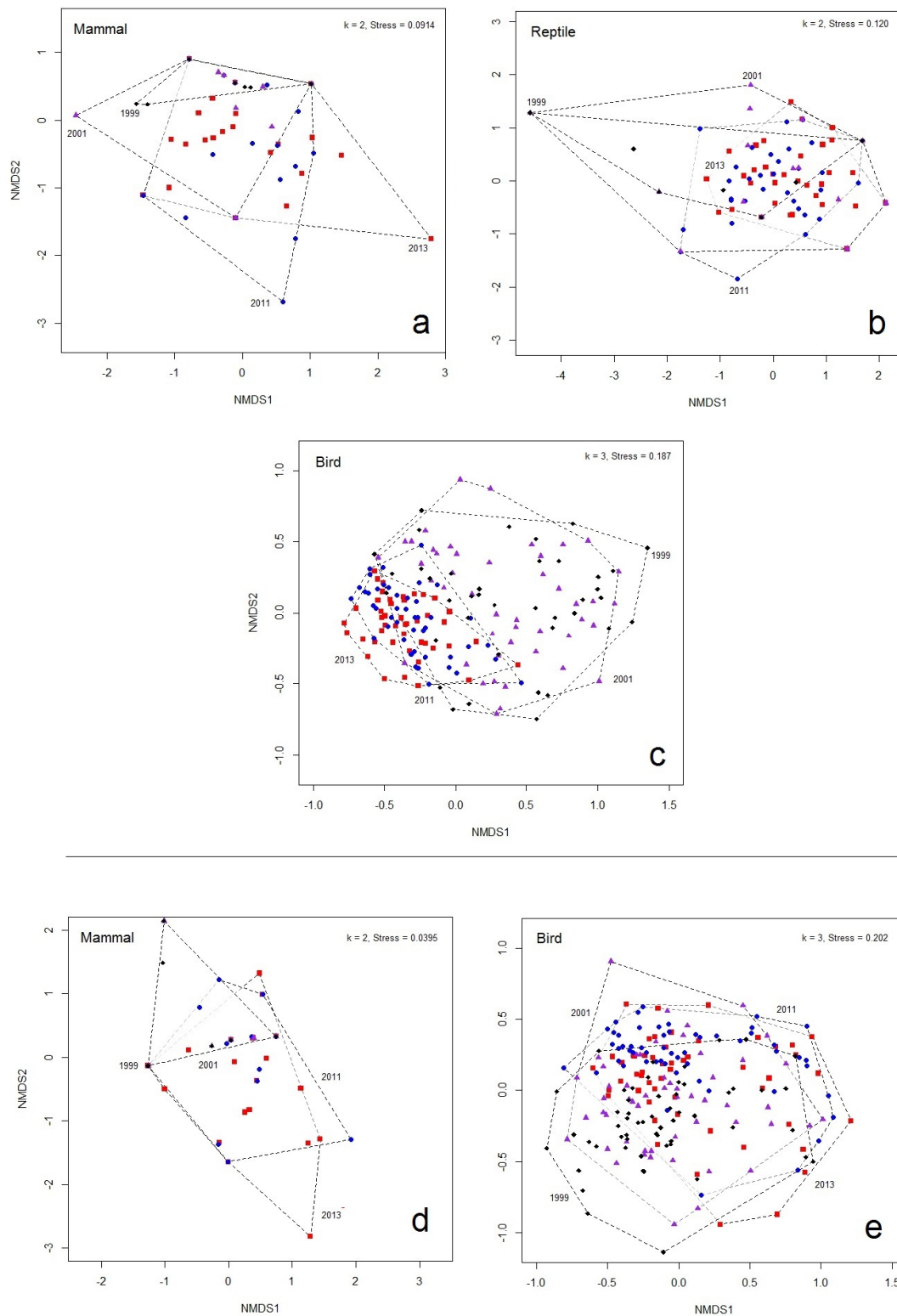
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1002 **Table 1**

<b>Term</b>	<b>Definition</b>
Cross-taxonomic surrogacy	The hypothesis that changes in the diversity or composition in a defined taxon (the surrogate) reflects a similar and commensurate change in another taxon (the target).
Congruence	The degree of concordance between measures of two defined taxonomic units (e.g. Fattorini et al. 2012; Westgate et al. 2014). Often measured by the level of correlation between diversity metrics of the defined taxonomic groups (e.g. Su et al. 2004), and is an important requisite in identifying cross-taxonomic surrogates (Gioria et al. 2011)
Indicator species	A species that can be used as a surrogate or proxy measure for the distribution and occurrence of other species, species groups (Ricketts et al 1999) and environmental conditions.
Species-based surrogate	A surrogate approach based on data of individual species, defined groups of species or measures of species diversity.
Species richness	The total number of species in a defined biotic community; also a commonly used metric in measures of biodiversity.
Species composition	A metric of a biodiversity that considers the identity and relative abundance of species in a defined biotic community.
Species temporal turnover	Change in species composition in a biotic community over time.

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1005 **Table 2**

<b>Taxa correlated</b>	<b>Mantel <i>R</i></b>	<b><i>n</i></b>	<b>Mantel <i>R</i></b>	<b><i>n</i></b>	<b>Mantel <i>R</i></b>	<b><i>n</i></b>	<b>Mantel <i>R</i></b>	<b><i>n</i></b>
<i>Agriculture control (AGR)</i>		1999		2001		2011		2013
Bird vs Reptile	-	-	-0.403	4	0.424*	7	0.0531	9
Bird vs Mammal	-	-	-	-	-	-	-	-
Reptile vs Mammal	-	-	-	-	-	-	-	-
<i>Woodland remnants in agricultural matrix</i>								
Bird vs Reptile	-0.134	17	0.0576	20	-0.0271	45	-0.102	50
Bird vs Mammal	-0.154	7	0.0350	11	0.306**	25	0.168*	50
Reptile vs Mammal	0.173	25	-0.0314	25	-0.0779	32	0.0548	33
<i>Woodland remnants in pine matrix</i>								
Bird vs Reptile	0.191*	17	0.114	21	-0.0780	41	0.0739	42
Bird vs Mammal	0.0880	13	0.279**	9	-0.0996	28	0.0791	29
Reptile vs Mammal	-	-	-0.0521	17	-0.109	32	0.145	32
Significance $P < 0.001$ **, $P \leq 0.05$ *, $0.05 < P \leq 0.1$ •								

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1016 **Table 3**

Predictor variable	2013		2011	
	Coefficient	P	Coefficient	P
<b>Woodland remnants in pine matrix</b>				
<i>Mammal ~ Bird + Reptile + Space</i>		$R^2 = 0.0309$		$R^2 = 0.0254$
Bird	0.312	0.157	-0.385	0.0911
Reptile	0.150	0.0650	-0.144	0.0731
Space	-0.0500	0.915	0.475	0.184
<i>Reptile ~ Bird + Mammal + Space</i>		$R^2 = 0.0309$		$R^2 = 0.0285$
Bird	0.312	0.167	-0.321	0.285
Mammal	0.150	0.0771	-0.100	0.098
Space	-0.0500	0.924	0.555	0.165
<b>Woodland remnants in agricultural matrix</b>				
<i>Mammal ~ Bird + Reptile + Space</i>		$R^2 = 0.0565^*$		$R^2 = 0.182^*$
Bird	0.425	0.00400*	0.859	0.00100*
Reptile	0.0979	0.258	-0.0870	0.376
Space	0.195	0.429	-0.0420	0.881
<i>Reptile ~ Bird + Mammal + Space</i>		$R^2 = 0.0287$		$R^2 = 0.0176$
Bird	-0.225	0.131	0.00942	0.955
Mammal	0.0582	0.280	-0.0620	0.343
Space	0.559	0.0130*	0.483	0.0450*
Significance $P < 0.001$ **, $P \leq 0.05$ *, $0.05 < P \leq 0.1$ .				

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1022 **Table 4**

Variable	NMDS1	NMDS2	R <sup>2</sup>	Variable	NMDS1	NMDS2	R <sup>2</sup>
<b>Woodland remnants in pine matrix</b>				<b>Woodland remnants in agricultural matrix</b>			
<i>Bird species composition</i>				<i>Bird species composition</i>			
% crown affected	0.581	-0.814	0.163*	Blackberry	0.463	-0.886	0.104·
Basal count	0.907	-0.420	0.272**	Dead trees	0.777	0.629	0.265**
Blackberry	-0.997	-0.0793	0.149·	Exposed rock	0.900	-0.436	0.162**
Logs 10-20cm	0.961	-0.276	0.114·	Ground cover	0.377	0.926	0.137*
Number of strata	-0.762	-0.647	0.129·	Number of strata	0.801	0.599	0.190**
<i>Reptile species composition</i>				<i>Reptile species composition</i>			
Blackberry	0.989	-0.149	0.292**	Number of trees	0.543	0.840	0.214**
Dominant cover	-0.487	0.874	0.226*	Shrub cover	0.941	0.339	0.236**
Number of strata	0.999	0.0545	0.145·	Stand height	-0.566	0.825	0.119*
Stem count 11-20cm	0.846	-0.534	0.154*	Subdominant cover	0.522	0.853	0.119*
<i>Mammal species composition</i>				<i>Mammal species composition</i>			
Logs >50cm	0.336	0.942	0.181·	Foliage depth	0.486	-0.874	0.116·
				Blackberry	0.957	-0.290	1.000*
				Exposed rock	0.00146	1.000	0.241*
				Foliage depth	-0.00110	1.000	0.190*
				Stand height	-0.00122	1.000	0.168*
Significance $P < 0.001$ **, $P \leq 0.05$ *, $0.05 < P \leq 0.1$ ·							

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