

How well is current plant trait composition predicted by modern and historical forest spatial configuration?

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2 Adam Kimberley, G. Alan Blackburn, J. Duncan Whyatt & Simon M. Smart

3 Kimberley, A. (Corresponding author, adakim@ceh.ac.uk) & Smart, S, (ssma@ceh.ac.uk): Centre for

4 Ecology and Hydrology, Lancaster Environment Centre, Library Avenue, Bailrigg, Lancaster, United

5 Kingdom, LA1 4AP.

6 Kimberley, A., Blackburn, G.A. (alan.blackburn@lancaster.ac.uk) & Whyatt, J.D.

7 (d.whyatt@lancaster.ac.uk): Lancaster Environment Centre, Library Avenue, Lancaster University,

8 Bailrigg, Lancaster, United Kingdom, LA1 4YQ.

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24 **Abstract**

25 There is increasing evidence to suggest that a delayed response of many forest species to
26 habitat loss and fragmentation leads to the development of extinction debts and
27 immigration credits in affected forest habitat. These time lags result in plant communities
28 which are not well predicted by present day landscape structure, reducing the accuracy of
29 biodiversity assessments and predictions for future change. Here, species richness data and
30 mean values for five life history characteristics within deciduous broadleaved forest habitat
31 across Great Britain were used to quantify the degree to which aspects of present day forest
32 plant composition are best explained by modern or historical forest patch area. Ancient
33 forest specialist richness, mean rarity and mean seed terminal velocity were not well
34 predicted by modern patch area, implying the existence of a degree of lag in British forest
35 patches. Mean seedbank persistence values were more closely related to modern patch area
36 than historical, particularly in larger patches. The variation in response for different mean
37 trait values suggests that species respond to landscape change at different rates depending
38 upon their combinations of different trait states. Current forest understorey communities
39 are therefore likely to consist of a mixture of declining extinction debt species and colonising
40 immigrant species. These results indicate that without management action, rare and
41 threatened species of plant are likely to be lost in the future as a result of changes in forest
42 spatial configuration that have already taken place. The lag seen here for rare specialist
43 plants suggests however that there may still be scope to protect such species before they
44 are lost from forest patches.

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48 **1. Introduction**

49 The spatial configuration of forest habitat is an important determinant of the richness and
50 composition of forest understorey plant communities (Jacquemyn et al. 2003, Lindborg
51 2007, Kimberley et al. 2014). Large, well connected patches support greater numbers of rare
52 species and species which possess low dispersal and competitive ability (Kolb and Diekmann
53 2005). This is particularly the case where such forests are of long continuity (Kimberley *et al.*
54 2014). Species with fast falling seeds and which are unable to persist within the seedbank
55 tend to be lost from forest habitat following landscape fragmentation and habitat loss, partly
56 because they are less able to rescue threatened populations through immigration or through
57 regeneration from the seedbank (Ozinga et al. 2009, Jacquemyn et al. 2012, Lindborg et al,
58 2012).

59 Recent evidence suggests that the response of forest communities to landscape change is
60 not immediate, with many species taking years to be lost from fragmented habitat or to
61 colonise expanding forest areas. This results in the formation of “extinction debts” and
62 “immigration credits” (Lindborg and Eriksson 2004, Metzger et al. 2009), where species
63 assemblages remain more strongly correlated with historical landscape structure than
64 modern habitat configurations (Kuussaari et al. 2009, Jackson and Sax 2010, Purschke et al.
65 2012). The consequent lack of coupling between biodiversity estimates and present day
66 landscape configuration is likely to reduce the ability of present day forest configuration to
67 explain and predict future patterns of plant species occurrence (Jackson and Sax 2010). This
68 has important implications for forest conservation and management strategies which
69 depend on accurate estimates of current biodiversity.

70 Although the impact of forest area, configuration and history has been investigated in
71 previous studies (Dupré & Ehrlén, 2002, Lindborg et al, 2012), relatively little work has
72 directly focussed on quantifying the extent of lag effects in forest habitat and determining
73 whether they differ between plant traits in a predictable manner. Here, we combine a
74 national scale dataset of plant species occurrence in forest patches with past and present
75 forest extent data. We then used these data to investigate the degree to which current plant
76 community composition is explained by historical rather than modern forest patch area.

77 Extinction debts are associated with species with low rates of population turnover such as
78 those with long life spans or the ability to persist within the seedbank. Such species may
79 remain as remnant populations for some time following unfavourable landscape change,
80 even when their eventual local extinction is likely (Eriksson 1996, Lindborg 2007, Vellend et
81 al. 2011). Forest habitat which has reduced in size may therefore still retain a
82 disproportionate number of the rare, forest specialist species that survived in previously
83 larger forest patches (Vellend et al, 2006, Kimberley et al, 2014). Conversely, immigration
84 credits result from the slow colonisation of new forest area by poorly dispersing species
85 (Verheyen et al. 2003, Jackson and Sax 2010). Forest patches which have been recently
86 established or which have seen an increase in the amount of forest habitat may therefore
87 still be dominated by better dispersing species; those with low seed weight and seed
88 terminal velocity or seeds which persist within the seed bank, in the absence of forest
89 specialist plants (Kimberley et al. 2014). Over time as the immigration credit is paid, many of
90 these forest specialists are likely to arrive, although the rate at which this occurs depends
91 upon proximity to source populations and the permeability of the intervening habitat matrix
92 (Peterken 2000, Brunet et al. 2011).

93 Where extinction debts and immigration credits exist in forest patches, the proportion of
94 species with linked traits such as high seed weight and terminal velocity and high seedbank
95 persistence are likely to lag behind landscape change. Combinations of life history
96 characteristics such as high seed terminal velocity and high specific leaf area are also known
97 to differentiate slow-dispersing, shade tolerant specialists largely restricted to long-
98 continuity, ancient woodland from forest plants that are more readily dispersed and more
99 typical of secondary forest (Kimberley et al. 2013). Such species are also more likely to be
100 rare. Thus ancient forest species tend to be stress tolerant and poor colonisers of new
101 habitat (Hermy et al. 1999) and therefore may be more prone to lag behind changes in forest
102 configuration. Lag effects in forest plants are often long lasting and have been observed
103 more than a century after forest fragmentation (Vellend et al. 2011). We therefore
104 hypothesised that present day forest community mean values for these traits would be
105 better explained by historical rather than modern forest patch area in patches which have
106 undergone area change. In addition to the trait-based approach, the relationships between
107 both total species richness and ancient woodland specialist richness (based on the list of
108 ancient woodland indicators in Kirby (2006)) and modern forest spatial configuration were
109 also analysed in order to determine whether species-based patterns could be discerned
110 alongside trait-based relationships with historical change in landscape structure.

111 In summary the following hypotheses were tested:

- 112 1. Plant community traits are better predicted by historical patch area than by modern
113 patch area within forest patches greater than 100 years old.

114 2. Traits associated with restriction to ancient forest habitat such as seed terminal
115 velocity and seedbank persistence are likely to be those most strongly linked to
116 historical forest patch area.

117 3. Richness of species restricted to ancient forest will be more closely related to
118 historical forest patch area than overall species richness.

119 **2. Methods**

120 **2.1. Survey data**

121 Digitised First Edition Ordnance Survey County Series (OS) maps (dated between 1849 and
122 1899) and data from the Countryside Survey, a national ecological surveillance programme
123 for Great Britain (Norton *et al.* 2012), were used to identify 82 patches of British
124 broadleaved forest which were established prior to 1899 and that were still recorded as
125 forest in 2007. Forest understory plant species occurrence data were then obtained for 151
126 vegetation sampling plots within these patches, assessed as part of Countryside Survey
127 2007. Two types of vegetation sampling plot were employed in the analysis; linear plots (10
128 m² in area), located parallel to forest streamsides and forest tracks, and area plots (200 m² in
129 area), located within the wider areal extent of each patch but not sampling a linear feature.

130 **2.2. Species and plant trait data**

131 Plant community mean trait values for a number of life history characteristics were
132 calculated for each plot by averaging the individual traits of all species present. These mean
133 values were then used as response variables in subsequent modelling. Mean trait values
134 were left un-weighted by species abundance. This allowed both subordinate and dominant
135 species to be considered equally, thus avoiding the confounding effect of variation in cover
136 due to local competitive sorting. Plant trait information was obtained from the Electronic
137 Comparative Plant Ecology database (Grime *et al.* 1995), the LEDA traitbase (Kleyer *et al.*

138 2008), The British Flora (Stace 1997) and PLANTATT (Hill et al. 2004). Species rarity was
139 obtained from PLANTATT as the number of occurrences in British 10 km squares in the
140 period 1987-1999.

141 Excluding trees and shrubs, 250 species occurred across the vegetation plots. Since trait data
142 were not available for all traits for all species, an approach was taken to minimise this
143 problem by estimating the missing values using a Bayesian hierarchical model written in
144 WinBUGs (Lunn et al. 2000), following the approach of Thompson and McCarthy (2008) as
145 applied in Kimberley et al. (2014). Imputing missing values in this manner is preferable to
146 removing them entirely, since estimated values take into account both between and within
147 family similarity among those species with known trait values. The five traits tested, along
148 with the percentage of species with missing values were; log natural seed weight (17.6%),
149 seed terminal velocity (29.6%), specific leaf area (5.2%), seedbank persistence (24.8%) and
150 rarity (0.4%). Seedbank persistence was assessed on a four point scale (1 = Transient seed, 2
151 = Persistent until next growing season, 3 = Small concentrations of persistent seeds, 4 =
152 Large year round bank of persistent seeds). In addition to the mean trait values, counts of
153 both overall plant species richness and ancient woodland indicator (AWI) species richness
154 were also obtained, using the list of AWIs in Kirby (2006).

155 **2.3. Spatial data**

156 Patch area data for forest patches around each Countryside Survey vegetation plot were
157 derived for two periods; modern (2007) and historical (pre 1899), by overlaying forest extent
158 data onto the geo-referenced Countryside Survey plot data using GIS techniques (ESRI,
159 2011). Modern forest patch area data were extracted from the satellite derived Land Cover
160 Map 2007 (Morton et al. 2011) whilst historical patch area data were digitised from First

161 Series OS maps. These modern and historical area data were then natural log transformed to
162 reduce the skew in their distribution.

163 **2.4. Local abiotic conditions**

164 Local conditions within forests are also important determinants of community composition
165 (Dupré and Ehrlén 2002, Kimberley et al. 2014). In order to obtain a more realistic estimate
166 of the effects of modern and historical forest configuration on mean community trait values
167 we included a number of abiotic variables measured at the same locations as the plant
168 species composition. Shade was estimated on a three point scale for all vegetation plots and
169 plots designated unshaded, partially shaded or fully shaded by field surveyors. Within each
170 of the area plots ($n = 46$) soil pH and carbon to nitrogen ratio were measured based on a 15
171 cm topsoil sample taken at the same time as the flora was recorded in each plot. In the
172 linear plots ($n = 105$) directly measured soil data were not available. Values within these
173 plots were estimated using published equations derived from a national calibration of
174 observed values of the three soil variables against the mean Ellenberg values of plants in
175 1033 plots from a stratified, random sample of the range of British vegetation types (Smart
176 et al. 2010). The mean Ellenberg values used in these equations to generate soil variables
177 were derived only from the trees and shrubs which were excluded from the calculation of
178 mean trait values for the herbaceous understorey (the dependent variables in the present
179 study). This may result in a less accurate estimate of soil conditions present in vegetation
180 plots due to the lower sample size of woody species present, however the problem of
181 circularity when the estimated soil variables were used to model mean trait values is
182 avoided through this method. In order to account for differences in response between the
183 area and linear plots, plot type was included as a categorical explanatory variable. Climate

184 and residual geographic variation across Britain were accounted for by the inclusion of the
185 northing of each sample plot as a continuous explanatory variable (Corney et al. 2006).

186 **2.5. Modelling approach**

187 In order to determine the extent to which modern mean community trait values are better
188 predicted by modern or historical patch area data, the spatial data from the two time
189 periods were combined into two new variables; one describing the mean patch area and the
190 other the change in the patch area between the historical and modern period. The amount
191 of change observed in patch area across forest patches is shown in Appendix 1 (Fig. A1).
192 These variables were then used as explanatory variables in models of present day mean
193 values of life history traits and species richness data within forest habitat. Since spatial data
194 was replicated over time but only modern plant species data were available, this modelling
195 approach allowed the effect of modern and historical forest spatial structure to be assessed
196 in a single model for each response variable.

197 Results from the models can be interpreted as follows: the relationship between trait and
198 mean patch area indicates whether the trait in question is significantly affected by forest
199 patch area. In cases where a significant effect exists, the parameter estimate for the change
200 in patch area versus modern trait relationship can then be used to indicate whether the trait
201 is better modelled using the modern or historical spatial data. Where the relationship
202 between mean patch area and trait is positive, a value for the change in area parameter of
203 greater than zero will indicate a community that is better predicted by the modern spatial
204 data. If the change in area parameter is negative, the results indicate present day trait data
205 are more strongly correlated with historical patch area (this is reversed where the
206 relationship between mean patch area and trait is negative). Where a significant effect of

207 mean patch area is observed but the change in patch area regression coefficient is close to
208 zero, the results indicate an intermediate community which is equally well explained by both
209 modern and historical spatial data, suggesting an intermediate amount of lag. Since high,
210 low and intermediate values for this metric all indicate important results, testing for a
211 significant difference from zero is not appropriate for the change in patch area term.
212 Confidence intervals are therefore not shown around this measure.

213 Both present and past spatial data would be expected to predict plant composition equally
214 well where the plant composition is in an intermediate state, having moved away from the
215 historic forest configuration following landscape change but not yet well predicted by
216 current spatial data. However modern and historical patch area would also be expected to
217 be equivalent in their ability to predict modern trait values where only small amounts of
218 spatial change has occurred. In order to prevent any lag effects being obscured by a lack of
219 change between time periods it was therefore important to ensure that the dataset was not
220 dominated by patches which were stable in area between historical and modern data
221 sources. To reduce this problem 40 plots, randomly selected from those present in patches
222 which had undergone less than a 10% change in patch area, were removed from the dataset
223 prior to the analysis. This provided a set of patches with an approximately even distribution
224 of amount of change which could be used in subsequent modelling (Supplementary material
225 Appendix 1, Fig. A1).

226 The analysis allowed the identification of traits which are similarly well predicted by both
227 modern and historical patch area as well as permitting the amount of change between time
228 periods to be taken into account in the analysis. Use of the mean patch area rather than the
229 historical value avoids collinearity problems where historical patch area is correlated with

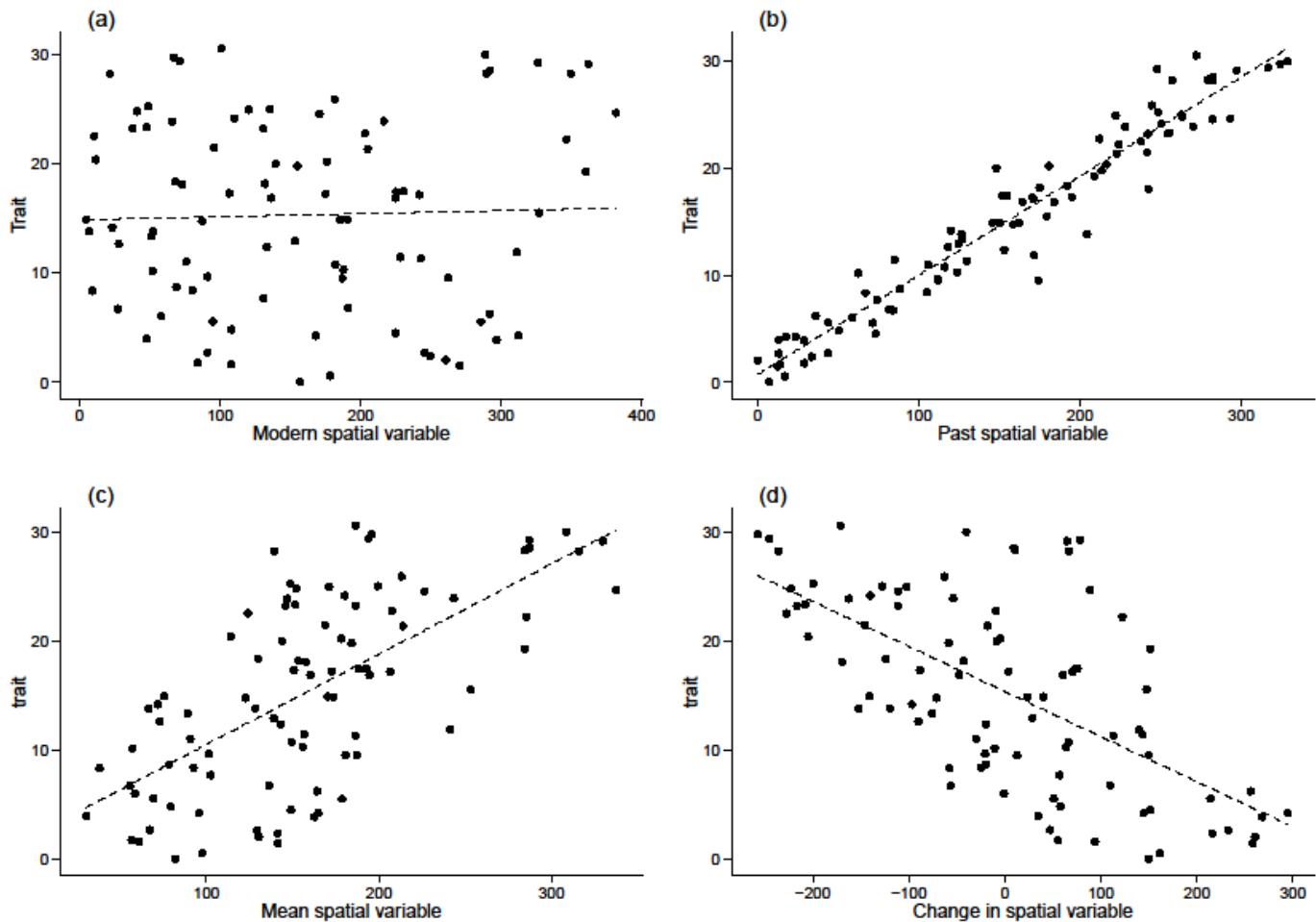
230 the amount of change. Thus the two spatial variables used in the analysis were statistically
231 independent.

232 The approach can be demonstrated using simulated examples. An artificial dataset was
233 created with information on modern trait composition, modern patch area and historical
234 patch area, where all patches had undergone a randomly allocated amount of change (either
235 positive or negative). The data were constructed such that modern values for a hypothetical
236 life history trait were strongly correlated with historical patch area but had no relationship
237 with a modern patch area (Figure 1a, b). Figure 1 shows the result of fitting the mean patch
238 area (Figure 1c) and change in patch area (Figure 1d) terms against the trait values. The trait
239 values which were associated with spatial variable values in the historical data have not
240 changed despite these patches having undergone change. Thus the patch area has changed
241 – high becoming low and low becoming high – but the trait values have not (Fig 1a). In such
242 a situation a relationship between trait and mean spatial variable is observed (Figure 1c),
243 and necessarily results in a strong negative correlation between change in the spatial
244 variable and the modern trait variable (Fig 1d), from which the stronger relationship
245 between trait and historical patch area can be inferred. If the historical patch area versus
246 trait relationship had been negative then this effect would have resulted in a positive slope
247 in Fig 1d.

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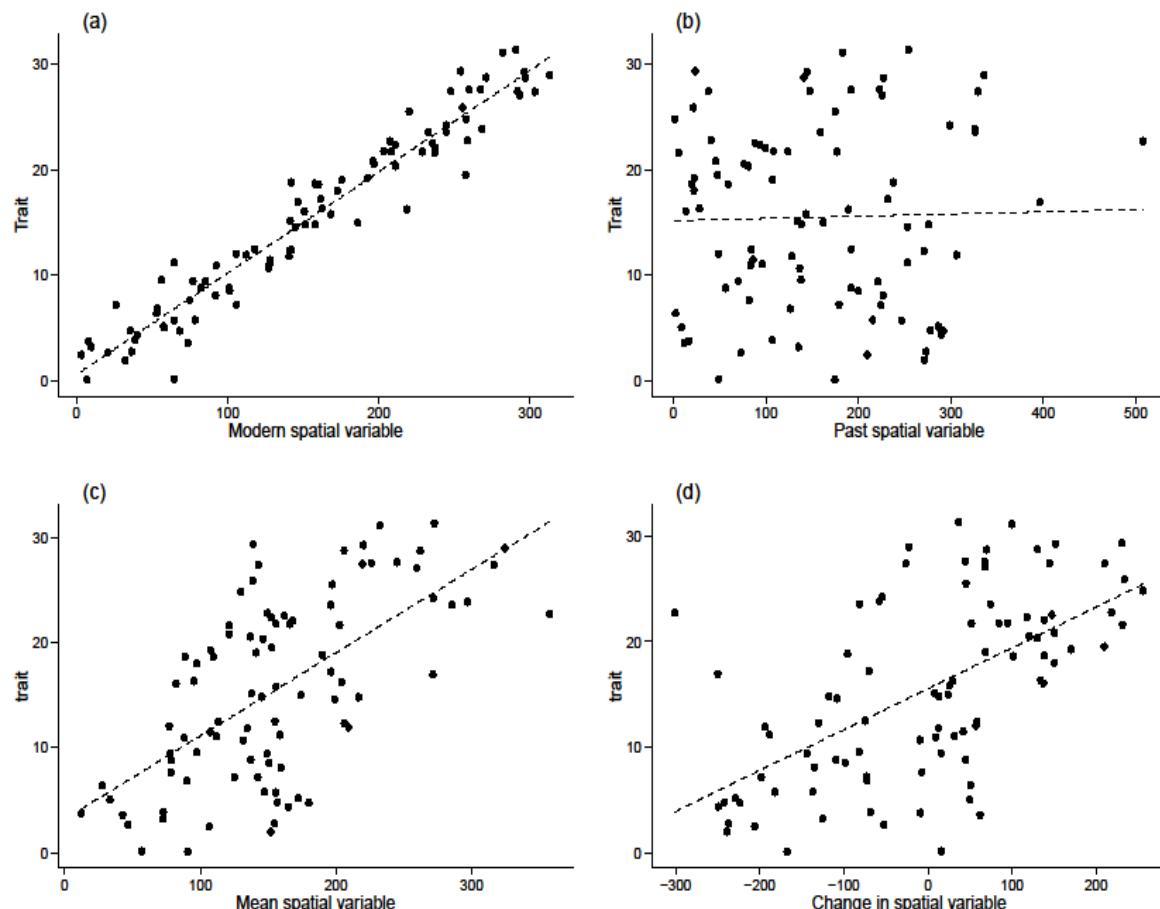


251
252 **Figure 1. Simulated data showing the relationships between hypothetical mean trait**
253 **values and (a) a modern spatial variable, (b) a historical spatial variable, (c) mean across**
254 **modern and historical spatial variables and (d) change between modern and historical**
255 **spatial variables, where trait data is best explained by historical spatial conditions. Dashed**
256 **lines show linear models between trait and each individual explanatory variable.**

257 A further simulation shows the pattern recovered by the analysis where the same strong
258 positive spatial-trait relationship occurs but in this case with modern patch area. A second
259 dataset was created; this time such that modern values for the hypothetical life history trait
260 were strongly correlated with modern patch area but had no relationship with historical
261 patch area (Figure 2a, b). The same modelling approach of fitting mean and change in patch
262 area against trait was then applied. This again results in a relationship between trait and
263 mean patch area (Figure 2c); however in this case the relationship between trait and

264 modern patch area is revealed by the positive relationship between trait and change in
265 patch area (Figure 2d).

266



267

268 **Figure 2. Simulated data showing the relationships between hypothetical mean trait**
269 **values and (a) a modern spatial variable, (b) a historical spatial variable, (c) mean across**
270 **modern and historical spatial variables and (d) change between modern and historical**
271 **spatial variables, where trait data is best explained by modern spatial conditions. Dashed**
272 **lines show linear models between trait and each individual explanatory variable.**

273

274 The modelling approach demonstrated in the simulated examples was applied to the real
275 data for the 111 vegetation sampling plots used. A single model was fitted for each mean
276 plant trait, along with species richness and AWI richness. These models contained the mean
277 patch area, the change in patch area and the interaction between these two variables, in
278 addition to all local condition variables described above. The interaction term was included
279 in each model to investigate whether patches with varying mean area differ in the extent to
280 which modern spatial data can be used to predict trait composition. A mixed-effects
281 modelling approach was taken, including site (Countryside Survey 1 km square) as a random
282 intercept, using the package lme4 in the statistical software R. This accounted for the spatial
283 autocorrelation introduced by analysing a number of vegetation sampling plots located
284 within the same Countryside Survey sample square. Mean trait values were modelled by
285 linear mixed effects models while generalised linear mixed effects models with a Poisson
286 error distribution were used for species richness and AWI richness models, to account for
287 the count data response. All models were scaled and centred using the R package arm, to
288 produce comparable regression coefficients. These allowed an estimate of the effect sizes of
289 each spatial variable on each plant trait to be made. 95% confidence intervals around these
290 effect sizes were calculated using the bootstrap method in lme4. For linear models response
291 values were also treated in this way to produce standardised effect sizes bounded by ± 1 . For
292 models of count data this was not possible due to the link function used in the generalised
293 linear models. Parameter estimates from the different model types are therefore not directly
294 comparable. The resulting effect sizes and confidence intervals allowed the extent to which
295 present day mean values for different life history traits are better predicted by modern or
296 historical forest spatial configuration to be assessed.

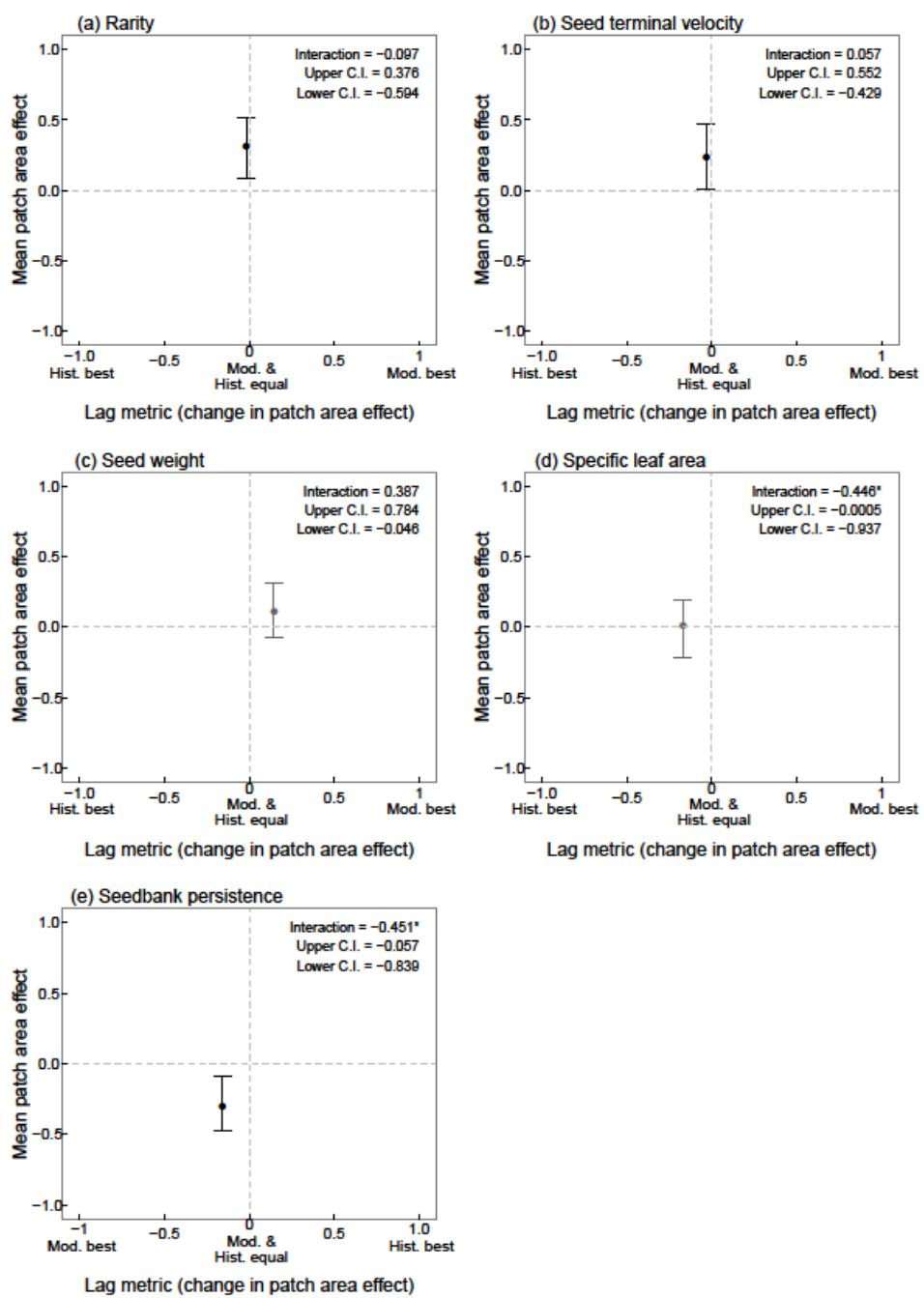
297 A number of significant effects of the abiotic variables, northing and plot type were
298 detected, discussion of which is beyond the scope of this article. Here we focus on partial
299 spatial relationships with trait composition having accounted for variation explained by local
300 environmental conditions. Full modelling results are however shown in Appendix 2
301 (Appendix 2, Table A2).

302 **3. Results**

303 **3.1. Trait data**

304 Mean patch area was a significant predictor for three of the five community mean response
305 variables tested; seedbank persistence, seed terminal velocity and species rarity (Figure 3).
306 Rarer species with faster falling seeds and less persistent seedbanks were found in patches
307 with a high average area across the two time periods, suggesting that forest configuration
308 has an important effect on the occurrence of species with these traits. The lag metric was
309 close to zero for both seed terminal velocity and rarity (change in area term, Figure 3a,b),
310 suggesting that both modern and historical patch area explain these traits equally well,
311 despite the gradient of change in patch area present across the sampled woodlands. This
312 must therefore mean that communities have not remained static and hence stayed
313 correlated with historic patch configuration, but neither have they completely readjusted to
314 the modern patch configuration. The lag metric for seedbank persistence however was less
315 than zero (Figure 3e). Given the negative relationship between mean patch area and this
316 trait this indicates that mean seedbank persistence values were better predicted by the
317 modern patch area than the historical.

318 The interaction between mean patch area and change in patch area had a significant
 319 negative effect on mean seed bank persistence values (Figure 3e). As mean patch area
 320 increases, the negative relationship between trait and change in area becomes stronger. This
 321 suggests that mean seedbank persistence was better predicted by modern patch area in
 322 forest patches with a larger mean area across the two time periods than in patches with a
 323 smaller mean area.

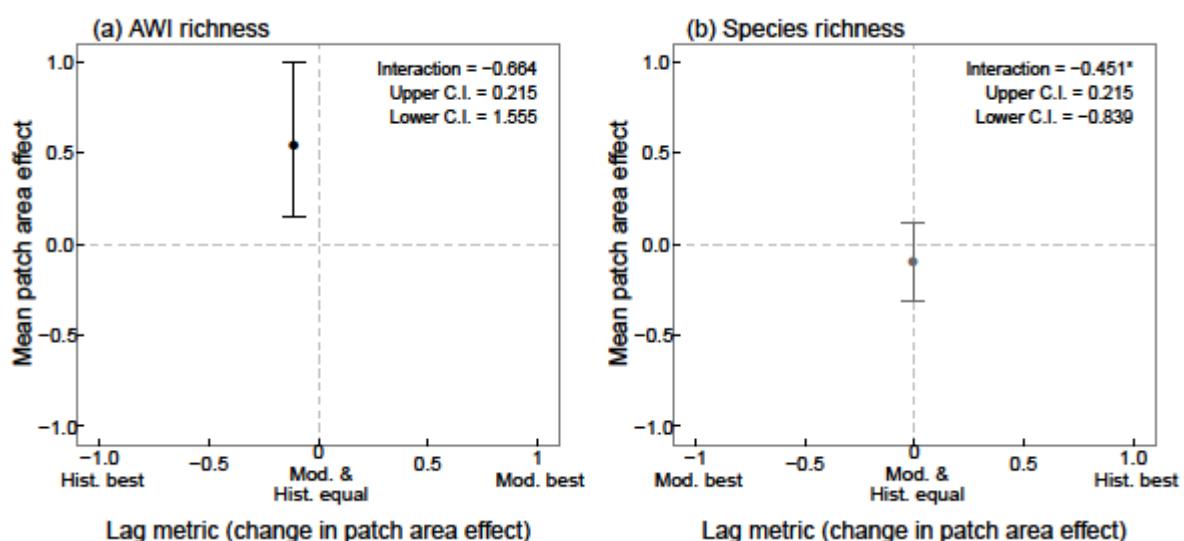


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325 **Figure 3. Standardised effect sizes quantifying the influence of patch area in models of five**
326 **mean trait values in forest vegetation sampling plots. Error bars represent 95% confidence**
327 **intervals. Where displayed confidence intervals do not overlap 0 a significant effect of**
328 **patch area is indicated. The position of the point on the x axis shows the extent to which**
329 **present day trait values are best predicted by historical or modern patch area. Text in the**
330 **top right of each panel shows the parameter estimate and upper and lower confidence**
331 **intervals for interaction terms. Parameter estimates for local abiotic variables (also**
332 **included in models) are not shown here.**

333 **3.2 Species data**

334 Mean patch area had a significant effect on AWI species richness but no effect on overall
335 species richness (Figure 4). This suggests that ancient forest specialists are more sensitive to
336 patch area than other forest plants. Change in patch area had a weak negative effect on AWI
337 species richness, indicating that the number of ancient forest specialists is slightly better
338 predicted by historical patch area than modern.



339

340 **Figure 4. Standardised parameter estimates quantifying the influence of patch area in**
341 **models of overall species richness and ancient woodland indicator (AWI) richness in forest**
342 **vegetation sampling plots. Error bars represent 95% confidence intervals. Where displayed**
343 **confidence intervals do not overlap 0 a significant effect of patch area is indicated. The**
344 **position of the point on the x axis shows the extent to which present day trait values are**
345 **best predicted by historical or modern patch area. Text in the top right of each panel**
346 **shows the parameter estimate and upper and lower confidence intervals for interaction**
347 **terms. Parameter estimates for local abiotic variables (also included in models) are not**
348 **shown here.**

349 **4. Discussion**

350 The important effects of forest spatial configuration on understorey plant composition
351 within forest patches were confirmed by the relationships identified here between mean
352 patch area and three of the five mean community values tested here. The strength with
353 which different traits could be predicted by modern rather than historical forest patch area
354 varied, indicating that while some species may be quickly lost from fragmented habitat,
355 many are likely to persist for some time following landscape change. Such variation in
356 response to changes in habitat fragmentation has important consequences for conservation
357 planning because it suggests that there may be a window of time in which to introduce
358 measures to help vulnerable species (Wearn et al. 2012).

359 The analytical approach taken here allowed intermediate situations to be identified, where a
360 mean trait value is affected by patch area but the trait is equally well predicted by both
361 modern and historical forest extent. Results suggest that this is the current case for both
362 rarity and seed terminal velocity, implying the existence of weak time lags for these

363 characteristics. This supports previous studies which have found that plant communities take
364 time to respond following landscape change (Lindborg and Eriksson 2004, Metzger et al.
365 2009, Saar et al. 2012). Rare species and those with heavy, fast falling seeds are likely to be
366 less able to disperse effectively and rescue threatened populations through immigration
367 (Kolb & Diekmann, 2005). Many such species are therefore unlikely to be able to persist
368 long-term following the loss of forest patch area. Since many rare, forest specialist plants are
369 perennial species however (Kimberley et al, 2013), they may survive in remnant populations
370 for some time following landscape change (Eriksson, 1996). The slow loss of species with
371 these characteristics may explain why mean seed terminal velocity and rarity were equally
372 well predicted by modern and historical patch area. This is further evidenced by the fact that
373 AWI richness within forest patches was more closely related to historical patch area than
374 modern. Hence there is likely to be a disproportionate drop in the occurrence of these
375 vulnerable plant species in the future as existing extinction debts are paid in patches which
376 have decreased in area. In many cases these species are also likely to be slow to colonise
377 forest patches which have increased in size, particularly in isolated patches (Brunet, 2011).
378 Hence maintaining large areas of older forest is important to avoid the loss of populations of
379 rare or poorly dispersing ancient woodland specialist plants (Kimberley *et al.* 2013).
380 Although existing time lags are likely to lead to ongoing change in forest community
381 composition, if the amount of change in forest extent between time periods is small the
382 degree of future change in plant composition is also likely to be limited, even where this
383 change takes some time to occur. It is therefore also important to consider the amount of
384 change which occurred between time periods when interpreting these results. It is likely that
385 a large alteration in patch size is needed to produce a significant, long lasting time lag. Here

386 only a weak lag was identified for mean rarity and seed terminal velocity, possibly due to a
387 small amount of change between historical and modern patch area for many patches.
388 Further application of this method to forests which have undergone more substantial or very
389 recent changes in area may reveal whether this is indeed the case. If so, the greatest benefit
390 of increasing forest patch area may be seen in patches which have recently undergone a
391 large reduction in area. The time lag identified here for rarity and seed terminal velocity may
392 also be weak due to the difference in species richness and composition between area and
393 linear plots used in this analysis. If linear plots contain a higher proportion of ruderal species
394 with characteristics consistent with a more rapid response to landscape change,
395 communities are likely to be closer to those predicted by modern forest patch area.

396 The variation in the degree to which modern or historical forest patch area best explains
397 mean trait values suggests that different species are responsible for each individual trait
398 relationship. For a species to persist but be bound for extinction it requires both strong
399 ability to persist and weak dispersal capability. Any lag observed in patches which have lost
400 area may be due to forest specialist species which have a particular combination of
401 established phase traits (slow, shade-tolerant vegetative growth) and regenerative traits
402 (poor dispersal) and therefore have the potential to persist for some time after landscape
403 change (Kimberley et al, 2013, Saar et al, 2012). Forest specialist species without this trait
404 combination are likely to be lost relatively quickly from fragmented patches while species
405 with these characteristics remain until they are either out-competed by more ruderal
406 immigrants or otherwise suffer mortality from disturbance, herbivory or disease (Grime,
407 2001, Jackson & Sax 2010). On the other hand immigrant species must be both rapidly
408 dispersed and shade-tolerant slow growers to truly survive in undisturbed forest

409 understorey. For example ruderal species with high investment in many small seeds with low
410 terminal velocity, high relative growth rates and high seedbank persistence can respond
411 more rapidly to landscape change, quickly colonising new forest edges, new small areas of
412 secondary woodland including previously larger patches which have lost forest area
413 (Tabarelli et al. 1999).

414 What we see integrated into the mean trait values is likely to be the trait-controlled sum of
415 the dynamics of fast-responding species more rapidly dispersed in time (through persistent
416 seedbanks) and space (through light, slower falling seeds) arriving at different rates from
417 surrounding habitats, coexisting with extinction debt species that are better fitted to
418 historical spatial configurations and hence are likely to decline further. These two processes
419 may occur at different rates however, with extinction debts in forest understorey plants
420 being paid sooner (after around 160 years) (Kolk & Naaf, 2015) than immigration credits
421 (which can remain for much longer) (Naaf & Kolk, 2015). If extinction debts in forest patches
422 in this analysis which have lost area have largely been paid, this may partly explain why only
423 weak lags were identified here for mean seed terminal velocity and rarity.

424 Mean seedbank persistence values lag less behind changes in patch area than mean seed
425 terminal velocity and rarity, particularly in large forest patches. High seedbank persistence
426 allows species to regenerate vulnerable or locally extinct populations from the soil
427 seedbank. The absence of such persistent species in larger forest patches (Kimberley et al,
428 2014) may result in a community which is faster to respond to changing patch area because
429 more species present in the vegetation possess no persistent seedbank. Such species are
430 likely to be quickly lost when habitat area is reduced. The species present above-ground are
431 also often poorly correlated with the species present in the seedbank (Bossuyt et al. 2002).

432 Many species present in forest seedbanks may therefore be rapidly growing species and
433 widely dispersed which are absent from the above-ground vegetation but likely to appear
434 and thrive following disturbance to the soil or canopy (Bossuyt et al. 2002). When forest
435 patches lose area or are newly disturbed they may swiftly gain these ruderal species from
436 the existing seedbank, reducing the lag for this trait (Smart et al 2014). In smaller patches
437 this effect may be weaker due to a higher original proportion of species with a persistent
438 seedbank (Kimberley et al, 2014). This suggests that large patches are likely to be quickest to
439 pay their extinction debts when they are reduced in size and further confirms the fact that
440 species which are particularly dependent on large, core areas of habitat may be first to
441 become extinct following the loss and fragmentation of forest habitat. The creation of small
442 patches of new forest is therefore likely to be of less benefit than extending existing forest
443 habitat (Peterken 2000).

444 One limitation of analysing the data in this way is that there is no way of knowing when
445 changes in spatial properties between the two time periods have occurred. Interpretation of
446 the results must therefore be done with care, since modern forest configuration would be
447 expected to have a stronger effect than historical if most of the spatial change was longer
448 ago. The large number of data points from across a wide geographic area used here however
449 ensured that a realistic assessment of current patterns in British forests could be made.
450 Furthermore, because the same forest habitats were analysed for all traits tested,
451 comparisons of the relative strength with which modern forest configuration affects
452 different mean trait values are still valid. Mean trait values were analysed separately to allow
453 differences in the response of traits to important variables to be detected. As such however,
454 the inter-correlation between pairs of traits must be taken into account. For example, part of

455 the observed effect of patch area on seedbank persistence may be due to the close
456 relationship between this trait and seed mass (Westoby et al. 2002). Correlations between
457 mean trait values are shown in Appendix 3 (Appendix 3, Fig. A5).

458 Although only forest patch area was tested here, this variable is often correlated with a
459 number of other forest configuration variables such as the amount of forest present in the
460 landscape or the amount of core forest habitat (Fahrig, 2003). In reality, time lags in forest
461 habitat are likely to depend on interactions between the size of patches, the amount of
462 nearby forest (particularly that of long continuity) and the amount of edge habitat present.
463 For example, newly created forest patches within a short distance of ancient forest habitat
464 have been shown to accumulate forest specialist species more quickly (Brunet *et al.* 2011),
465 while young forest patches which are highly isolated from ancient forest habitat mostly
466 accumulate species adapted for effective dispersal which tend not to be ancient woodland
467 specialists (Brunet 2007). Hedges and other semi-natural habitat types also have some
468 ability to act as a refuge for forest specialist species (McCollin et al. 2000, Smart et al. 2001),
469 potentially enabling such species to persist for longer, and therefore exhibit a stronger lag
470 effect, in landscapes where such features are common. The landscape context of changing
471 forest habitat is therefore also likely to be an important determinant of the extent to which
472 time lags develop. High intensity agriculture in neighbouring land use has been shown to
473 reduce the ability of forest specialist species to exist near forest edge habitat (Chabrerie et
474 al. 2013). Where forest patches are surrounded by intensive agricultural land, forest edge is
475 likely to be quickly colonised and dominated by weedy generalist species with higher
476 seedbank persistence (Willi et al. 2005). Where forest edge is buffered by less intensive land
477 uses however, stronger lags may be occurring as forest specialist species take longer to be

478 out-competed by immigrants. Hence, some forest specialist species may still be able to
479 persist even in small patches or at forest edges, so long as they are already established
480 before fragmentation and that the forest patch is appropriately surrounded by non-intensive
481 land. Hence buffering forest habitat with less intensive habitat types and linear refuges may
482 allow many vulnerable forest species to persist following landscape change, but this issue
483 requires further research.

484 In future, as existing immigration credits and extinction debts are paid, forest species
485 composition is likely to shift towards present day patterns of habitat configuration, with
486 fragmented forest likely to lose shade tolerant, poor dispersers and gain populations of
487 immigrant species. Likewise forest patches which are increasing in size will begin to recruit
488 suitable populations of forest plants and lose species more fitted to smaller patches with a
489 high edge to area ratio. The fact that mean rarity and seed terminal velocity were equally
490 strongly affected by modern and historical forest configuration in long established British
491 forest patches highlights the importance of accounting for historical forest spatial
492 configuration when modelling patterns of plant species occurrence (Ewers *et al.* 2013).
493 Failure to do so risks both underestimating the strength with which forest configuration
494 affects species and failing to identify species which are at risk of local extinction (Helm *et al.*
495 2006). However extinction debts in particular do present an opportunity to initiate measures
496 to prevent the loss of threatened species (Kuussaari *et al.* 2009) and the time lag identified
497 here for rare species and inefficient dispersers suggests that many vulnerable species could
498 benefit from well targeted management action.

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503 Group, UK.

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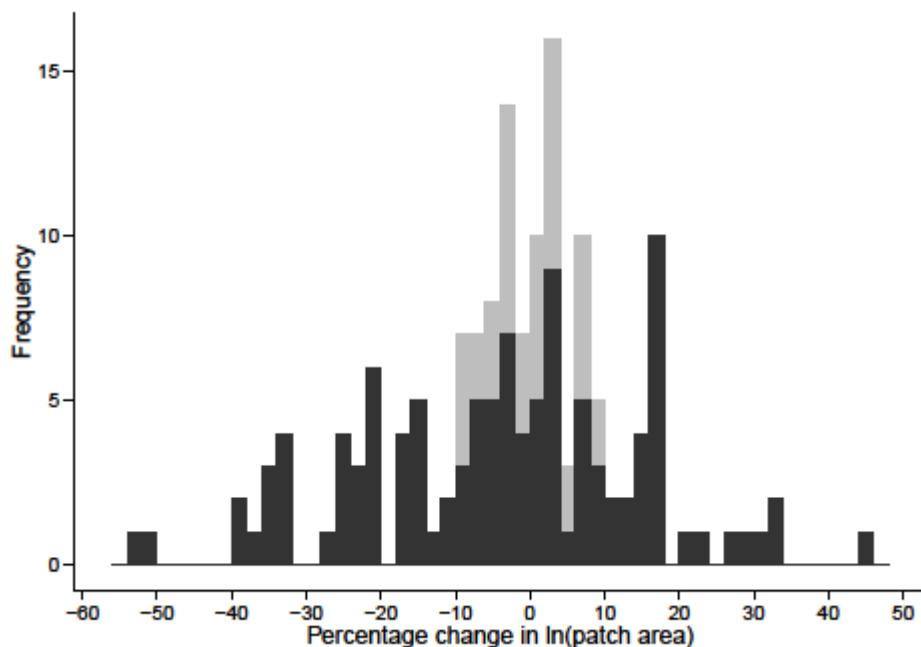
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632 **Supplementary Materials**

633 **Appendix 1: Histograms showing the amount of change in each spatial variable for forests**

634 **patches.**



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636 **Figure A1: Histograms showing the amount of change observed for three aspects of forest
 637 spatial configuration between 1899 and 2007 in forest patches over 100 years in age across
 638 Great Britain, around 151 vegetation sampling plots. Grey area shows the data removed
 639 prior to modelling.**

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644 **Appendix 2: Effect sizes and 95% confidence intervals for explanatory variables in models
 645 of mean trait values and species richness.**

646 Table A2: Effect sizes and 95% confidence intervals for explanatory variables in patch area
 647 models for different traits

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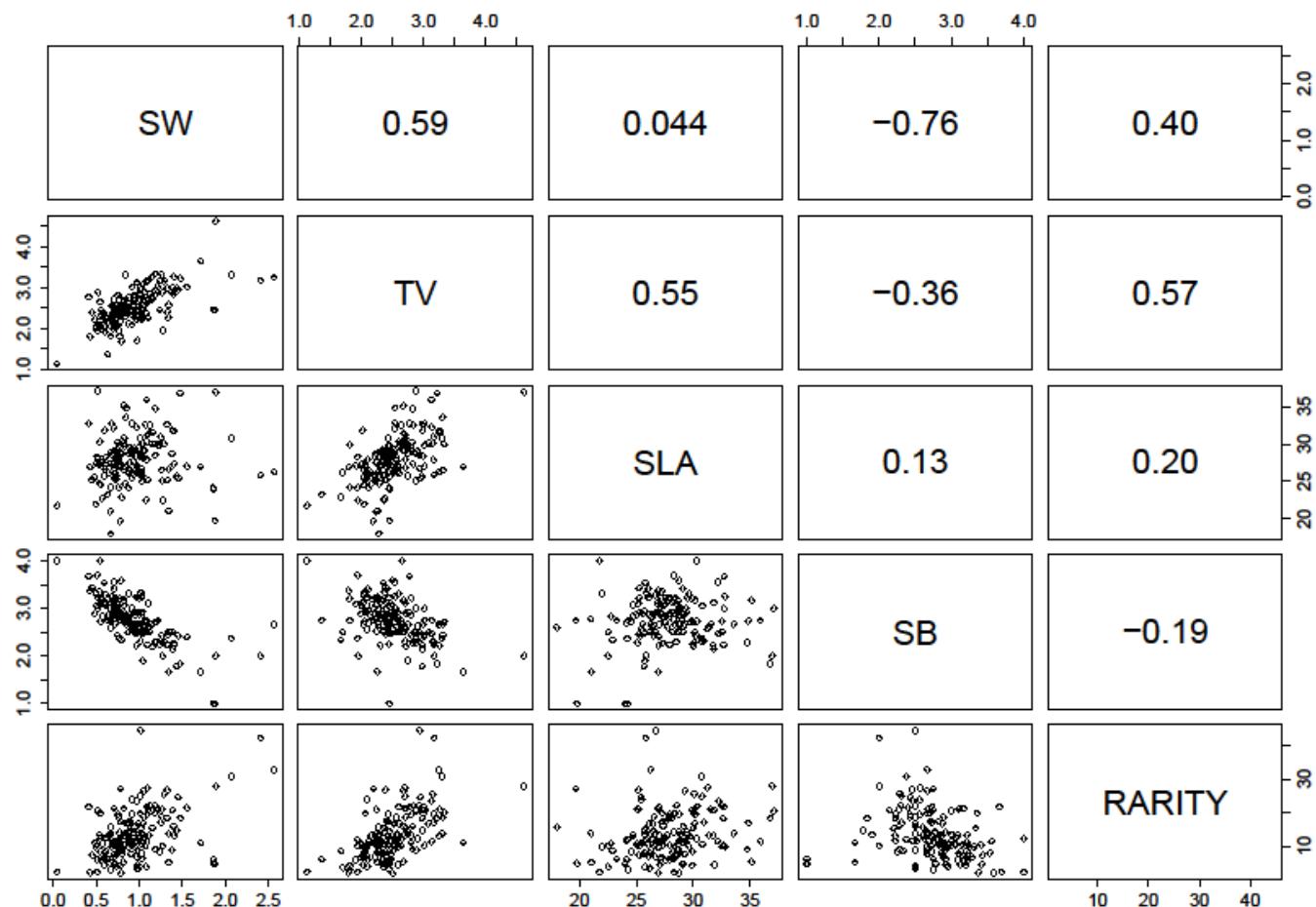
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656 **Appendix 3: Pairs plot displaying correlations between mean trait values within 151**
657 **vegetation sampling plots.**



658 **Figure A5: Pairs plot displaying correlations between mean trait values within vegetation**

659 **sampling plots.**

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