

Article (refereed) - postprint

Kimberley, Adam; Blackburn, G. Alan; Whyatt, J. Duncan; Smart, Simon M.
2016. **How well is current plant trait composition predicted by modern
and historical forest spatial configuration?** *Ecography*, 39 (1). 67-76.
[10.1111/ecog.01607](https://doi.org/10.1111/ecog.01607)

© 2015 The Authors. *Ecography* © 2015 Nordic Society Oikos

This version available <http://nora.nerc.ac.uk/510539/>

NERC has developed NORA to enable users to access research outputs wholly or partially funded by NERC. Copyright and other rights for material on this site are retained by the rights owners. Users should read the terms and conditions of use of this material at <http://nora.nerc.ac.uk/policies.html#access>

This document is the author's final manuscript version of the journal article, incorporating any revisions agreed during the peer review process. There may be differences between this and the publisher's version. You are advised to consult the publisher's version if you wish to cite from this article.

The definitive version is available at <http://onlinelibrary.wiley.com/>

Contact CEH NORA team at
noraceh@ceh.ac.uk

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

1

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.

9

10

11

12

13

14

15

16

17

18

19

20

21

22

23

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 species whose extinction debt is still
40 to be paid, and faster colonising immigrant species. These results indicate that without
41 management action, rare and threatened species of plant are likely to be lost in the future as
42 a result of changes in forest spatial configuration that have already taken place. The lag seen
43 here for rare specialist plants suggests however that there may still be scope to protect such
44 species before they are lost from forest patches.

45

46

47 **Introduction**

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

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

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

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

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

110 In summary the following hypotheses were tested:

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

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

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

118 **Methods**

119 **Survey data**

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

129 **Species and plant trait data**

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

136 Comparative Plant Ecology database (Grime *et al.* 1995), the LEDA traitbase (Kleyer *et al.*
137 2008), The British Flora (Stace 1997) and PLANTATT (Hill *et al.* 2004). Species rarity was
138 obtained from PLANTATT as the number of occurrences in British 10 km squares in the
139 period 1987-1999.

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

154 **Spatial data**

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

159 Map 2007 (Morton et al. 2011) whilst historical patch area data were digitised from First
160 Series OS maps. These modern and historical area data were then natural log transformed to
161 reduce the skew in their distribution.

162 **Local abiotic conditions**

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

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

185 **Modelling approach**

186 In order to determine the extent to which modern mean community trait values are better
187 predicted by modern or historical patch area data, the spatial data from the two time
188 periods were combined into two new variables; one describing the mean patch area and the
189 other the change in the patch area between the historical and modern period. The amount
190 of change observed in patch area across forest patches is shown in Appendix 1 (Fig. A1).

191 These variables were then used as explanatory variables in models of present day mean
192 values of life history traits and species richness data within forest habitat. Since spatial data
193 was replicated over time but only modern plant species data were available, this modelling
194 approach allowed the effect of modern and historical forest spatial structure to be assessed
195 in a single model for each response variable.

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

206 mean patch area is observed but the change in patch area regression coefficient is close to
207 zero, the results indicate an intermediate community which is equally well explained by both
208 modern and historical spatial data, suggesting an intermediate amount of lag. Since high,
209 low and intermediate values for this metric all indicate important results, testing for a
210 significant difference from zero is not appropriate for the change in patch area term.
211 Confidence intervals are therefore not shown around results for this measure (Figures 3 and
212 4).

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

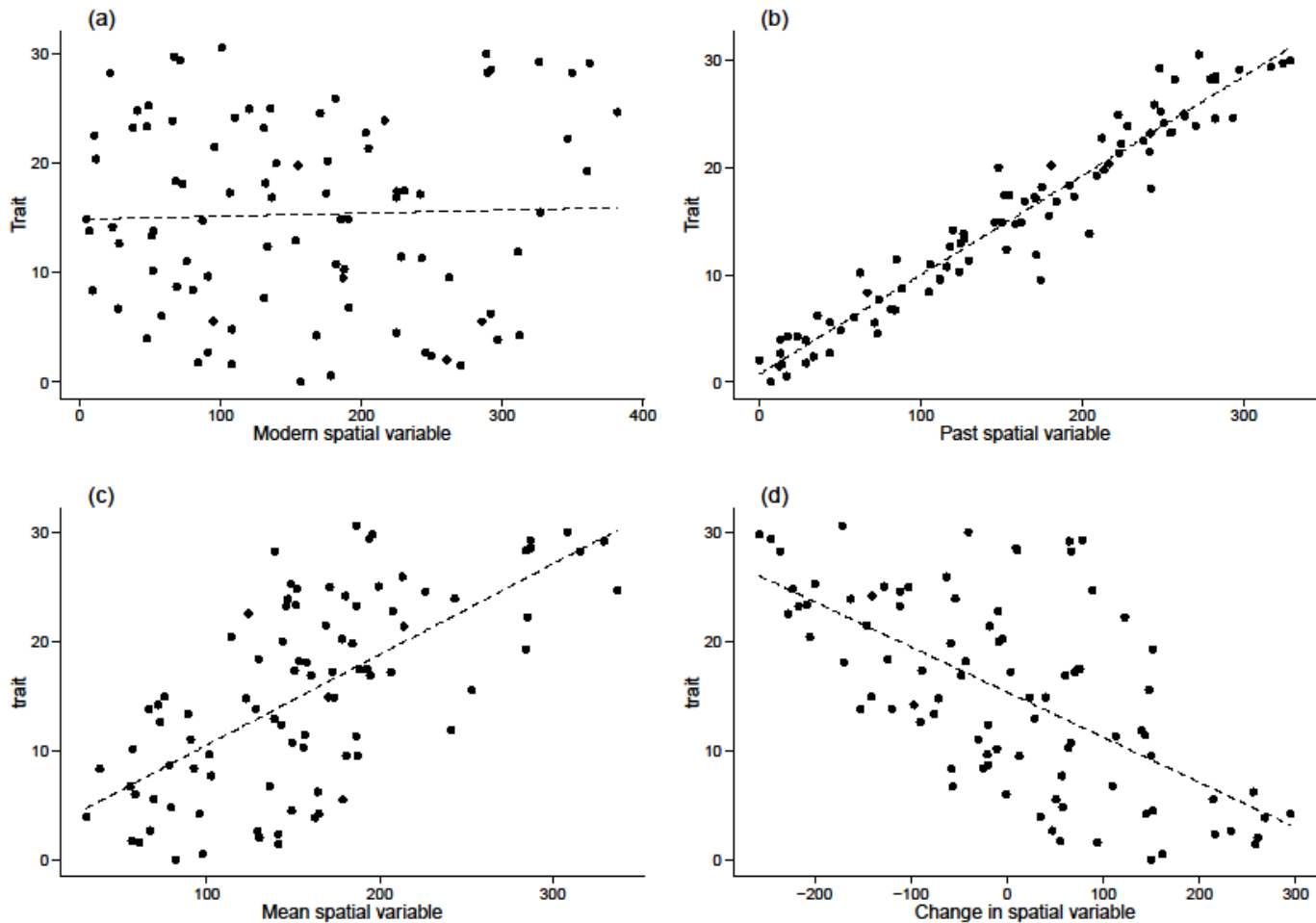
229 historical value avoids collinearity problems where historical patch area is correlated with
230 the amount of change. Thus the two spatial variables used in the analysis were statistically
231 independent.

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.

248

249

250

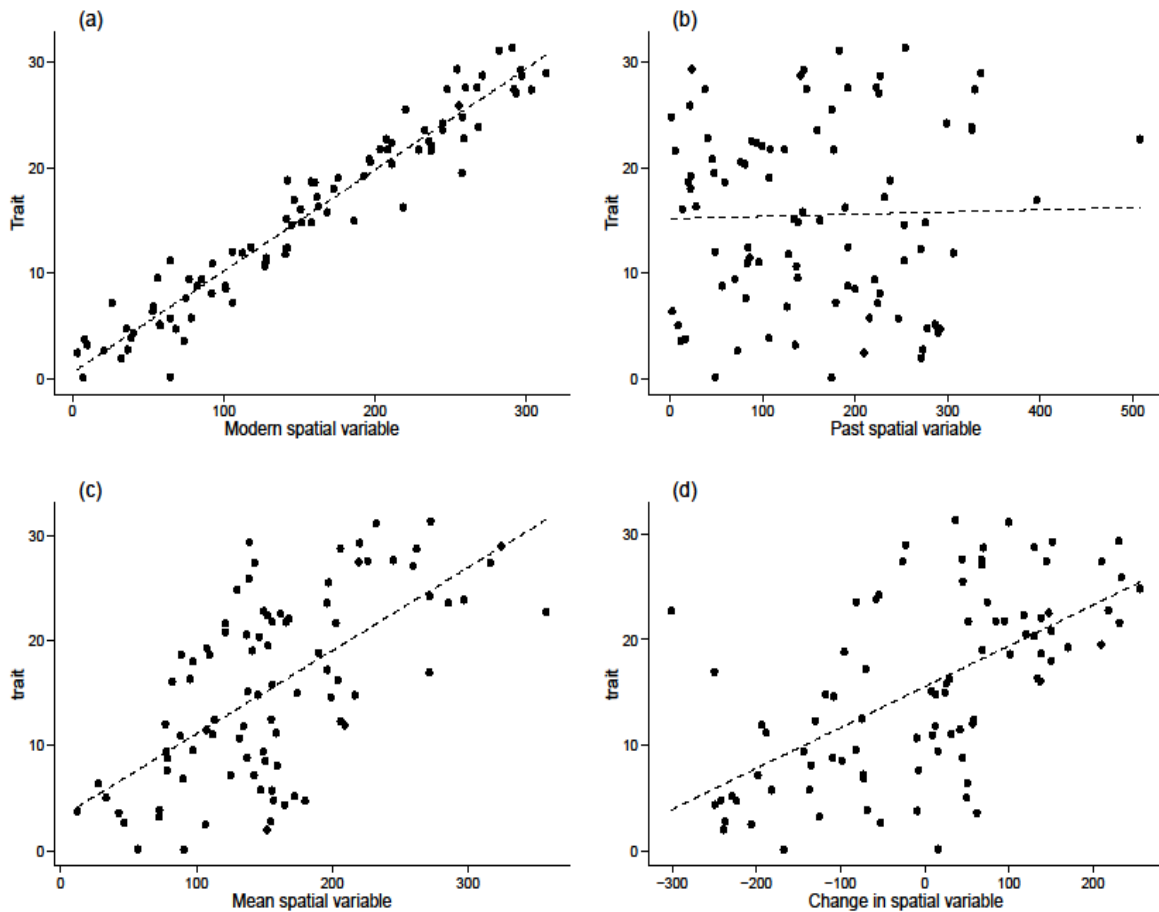


251
 252 **Figure 1. Simulated data showing the relationships between hypothetical mean trait**
 253 **values and (a) a modern spatial variable, (b) a historical spatial variable, (c) mean across**
 254 **modern and historical spatial variables and (d) change between modern and historical**
 255 **spatial 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 ancient woodland indicator richness. These
277 models contained the mean patch area, the change in patch area and the interaction
278 between these two variables, in addition to all local condition variables described above.
279 The interaction term was included in each model to investigate whether patches with
280 varying mean area differ in the extent to which modern spatial data can be used to predict
281 trait composition. A mixed-effects modelling approach was taken, including site (Countryside
282 Survey 1 km square) as a random intercept, using the package lme4 in the statistical
283 software R. This accounted for the spatial autocorrelation introduced by analysing a number
284 of vegetation sampling plots located within the same Countryside Survey sample square.
285 Mean trait values were modelled by linear mixed effects models while generalised linear
286 mixed effects models with a Poisson error distribution were used for species richness and
287 ancient woodland indicator richness models, to account for the count data response. All
288 models were scaled and centred using the R package arm, to produce comparable regression
289 coefficients. These allowed an estimate of the effect sizes of each spatial variable on each
290 plant trait to be made. 95% confidence intervals around these effect sizes were calculated
291 using the bootstrap method in lme4. For linear models response values were also treated in
292 this way to produce standardised effect sizes bounded by ± 1 . For models of count data this
293 was not possible due to the link function used in the generalised linear models. Parameter
294 estimates from the different model types are therefore not directly comparable. The
295 resulting effect sizes and confidence intervals allowed the extent to which present day mean
296 values for different life history traits are better predicted by modern or historical forest
297 spatial configuration to be assessed.

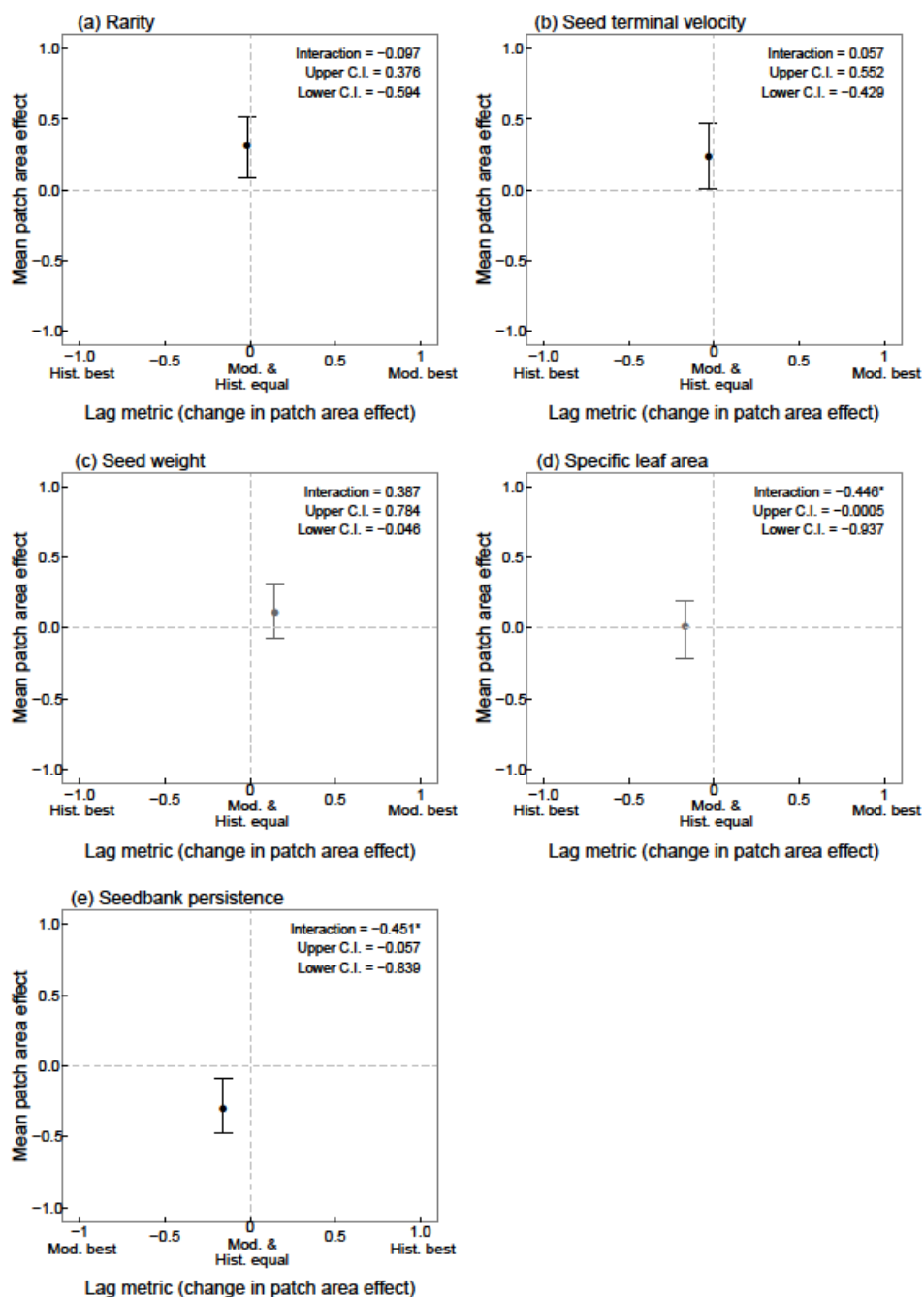
298 A number of significant effects of the abiotic variables, northing and plot type were
299 detected, discussion of which is beyond the scope of this article. Here we focus on partial
300 spatial relationships with trait composition having accounted for variation explained by local
301 environmental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

303 **Results**

304 **Trait data**

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

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

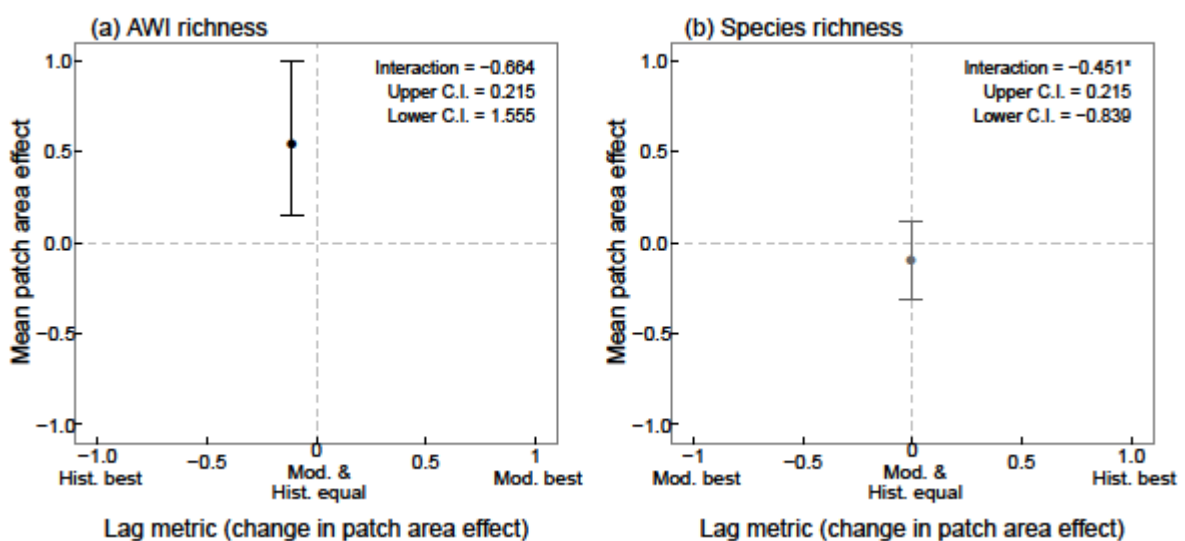


325

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

334 **Species data**

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



340

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

350 **Discussion**

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

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

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

382 Although existing time lags are likely to lead to ongoing change in forest community
383 composition, if the amount of change in forest extent between time periods is small the
384 degree of future change in plant composition is also likely to be limited, even where this
385 change takes some time to occur. It is therefore also important to consider the amount of
386 change which occurred between time periods when interpreting these results. It is likely that

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

398 The variation in the degree to which modern or historical forest patch area best explains
399 mean trait values suggests that different species are responsible for each individual trait
400 relationship. For a species to persist but be bound for extinction it requires both strong
401 ability to persist and weak dispersal capability. Any lag observed in patches which have lost
402 area may be due to forest specialist species which have a particular combination of
403 established phase traits (slow, shade-tolerant vegetative growth) and regenerative traits
404 (poor dispersal) and therefore have the potential to persist for some time after landscape
405 change (Saar et al, 2012, Kimberley et al, 2013). Forest specialist species without this trait
406 combination are likely to be lost relatively quickly from fragmented patches while species
407 with these characteristics remain until they are either out-competed by more ruderal
408 immigrants or otherwise suffer mortality from disturbance, herbivory or disease (Grime,
409 2001, Jackson & Sax 2010). On the other hand immigrant species must be both rapidly

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

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

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

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

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

456 the inter-correlation between pairs of traits must be taken into account. For example, part of
457 the observed effect of patch area on seedbank persistence may be due to the close
458 relationship between this trait and seed mass (Westoby et al. 2002). Correlations between
459 mean trait values are shown in Appendix 3 (Appendix 3, Fig. A5).

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

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

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

501 **Acknowledgements**

502 This research was funded through a NERC algorithm studentship to A.K., project code
503 NEC03454. Ordnance Survey County Series maps (1st Edition: 1849-1899) were obtained
504 from the EDINA Historic Digimap Service, <http://edina.ac.uk/digimap>, Landmark Information
505 Group, UK.

506 **References**

507 Bossuyt, B. et al. 2002. Seed bank and vegetation composition of forest stands of varying age
508 in central Belgium: consequences for regeneration of ancient forest vegetation. *Plant*
509 *Ecol.* 162: 33-48.

510 Brunet, J. 2007. Plant colonization in heterogeneous landscapes: an 80-year perspective on
511 restoration of broadleaved forest vegetation. *J. Appl. Ecol.* 44: 563–572.

512 Brunet, J. et al. 2011. Understory succession in post-agricultural oak forests: Habitat
513 fragmentation affects forest specialists and generalists differently. *For. Ecol. Manage.*
514 262: 1863–1871.

515 Chabrierie, O. et al. 2013. Maturation of forest edges is constrained by neighbouring
516 agricultural land management. *J. Veg. Sci.* 24: 58–69.

517 Corney, P.M. et al. 2006. Relationships between the species composition of forest field-layer
518 vegetation and environmental drivers, assessed using a national scale survey. *J. Ecol.* 94;
519 383–401.

520 Dupré, C. and Ehrlén, J. 2002. Habitat configuration, species traits and plant distributions. *J.*
521 *Ecol.* 90; 796–805.

- 522 Eriksson, O. 1996 Regional dynamics of plants: A review of evidence for remnant, source-
523 sink and metapopulations. *Oikos* 77:248-258.
- 524 ESRI. 2011. ArcGIS Desktop: Release 10. Environmental Systems Research Institute,
525 Redlands, CA.
- 526 Ewers, R. M. et al. 2013. Using landscape history to predict biodiversity patterns in
527 fragmented landscapes. - *Ecol. Lett.* 16: 1221–33.
- 528 Fahrig, L. 2003. Effects of Habitat Fragmentation on Biodiversity. *Annu. Rev. Ecol. Evol. Syst.*
529 34: 487-515.
- 530 Grime, J.P. et al. 1995. *The Electronic Comparative Plant Ecology*. Chapman & Hall, London.
- 531 Grime, J. P. 2001. *Plant strategies, vegetation processes, and ecosystem properties*. Second
532 edn. John Wiley and Sons, Chichester, UK.
- 533 Helm, A. et al. 2006. Slow response of plant species richness to habitat loss and
534 fragmentation. - *Ecol. Lett.* 9: 72–7.
- 535 Hermy, M. et al. 1999. An ecological comparison between ancient and other forest plant
536 species of Europe, and the implications for forest conservation. *Biol. Cons.* 91: 9-22.
- 537 Hill, M. O. et al. 2004. *PLANTATT - attributes of British and Irish plants: status, size, life*
538 *history, geography and habitats*. Centre for Ecology and Hydrology, Huntingdon.
- 539 Jackson, S. T. and Sax, D. F. 2010. Balancing biodiversity in a changing environment:
540 extinction debt, immigration credit and species turnover. *Trends Ecol. Evol.* 25: 153–60.

541 Jacquemyn, H. et al. 2003. Influence of environmental and spatial variables on regional
542 distribution of forest plant species in a fragmented and changing landscape. *Ecography*.
543 26: 768–776.

544 Jacquemyn, H. et al. 2012. Evolutionary changes in plant reproductive traits following habitat
545 fragmentation and their consequences for population fitness. *J. Ecol.* 100: 76–87.

546 Kimberley, A. et al. 2013. Identifying the trait syndromes of conservation indicator species:
547 how distinct are British ancient woodland indicator plants from other woodland
548 species? - *Appl. Veg. Sci.* 16: 667–675.

549 Kimberley, A. et al. 2014. Traits of plant communities in fragmented forests: The relative
550 influence of habitat spatial configuration and local abiotic conditions. *J. Ecol.* 102: 632–
551 640.

552 Kirby, K. 2006. Ancient Woodland Indicator (AWI) plants, in Rose, F. (ed) *The wildflower key*.
553 Penguin Group, London, pp. 558–561.

554 Kleyer, M. et al. 2008. The LEDA Traitbase: a database of life-history traits of the Northwest
555 European flora. *J. Ecol.* 96: 1266–1274.

556 Kolb, A. and Diekmann, M. 2005. Effects of Life-History Traits on Responses of Plant Species
557 to Forest Fragmentation. *Conserv. Biol.* 19: 929–938.

558 Kolk, J. and Naaf, T. 2015. Herb layer extinction debt in highly fragmented temperate forests
559 – Completely paid after 160 years? *Biol. Cons.* 182: 164-172.

560 Kuussaari, M. et al. 2009. Extinction debt: a challenge for biodiversity conservation. *Trends*
561 *Ecol. Evol.* 24: 564–71.

562 Lindborg, R. 2007. Evaluating the distribution of plant life-history traits in relation to current
563 and historical landscape configurations. *J. Ecol.* 95: 555–564.

564 Lindborg, R. et al. 2012. Effect of habitat area and isolation on plant trait distribution in
565 European forests and grasslands. *Ecography* 34:1–8.

566 Lindborg, R. and Eriksson, O. 2004. Historical Landscape Connectivity Affects Present Plant
567 Species Diversity. *Ecology* 85: 1840–1845.

568 Lunn, D.J. et al. 2000. WinBUGS - a Bayesian modelling framework: concepts, structure, and
569 extensibility. *Stat. Comput.* 10: 325–337.

570 McCollin, D. et al. 2000. Hedgerows as habitat for woodland plants. *J. Environ. Manage.* 60:
571 77–90.

572 Metzger, J. P. et al. 2009. Time-lag in biological responses to landscape changes in a highly
573 dynamic Atlantic forest region. *Biol. Conserv.* 142: 1166–1177.

574 Morton, R.D. et al. 2011. Final Report for LCM2007 - the new UK land cover map.
575 Countryside Survey Technical Report No 11/07. NERC/Centre for Ecology & Hydrology
576 pp. 112.

577 Naaf, T. and Kolk, J. 2015. Colonization credit of post-agricultural forest patches in NE
578 Germany remains 130–230 years after reforestation, *Biol. Cons.* 182: 155–163.

579 Norton, L. R. et al. 2012. Measuring stock and change in the GB countryside for policy--key
580 findings and developments from the Countryside Survey 2007 field survey. *J. Environ.*
581 *Manage.* 113: 117–27.

582 Ozinga, W. et al. 2009. Dispersal failure contributes to plant losses in NW Europe. *Ecol. Lett.*
583 12: 66–74.

584 Peterken, G. F. 2000. Rebuilding Networks of Forest Habitats in Lowland England. *Landsc.*
585 *Res.* 25: 291–303.

586 Purschke, O. et al. 2012. Linking landscape history and dispersal traits in grassland plant
587 communities. *Oecologia* 168: 773–83.

588 Saar, L. et al. 2012. Which plant traits predict species loss in calcareous grasslands with
589 extinction debt? - *Divers. Distrib.* 18: 808–817.

590 Smart, S. M. et al. 2001. An assessment of the potential of British hedges to act as corridors
591 and refuges for Ancient Woodland Indicator plants. In Barr, C., Petit, S. (eds) *Hedgerows*
592 *of the world: their ecological functions in different landscapes. Proceedings of the 10th*
593 *Annual Conference of the International Association for Landscape Ecology*, pp 137–146.

594 Smart, S.M. et al. 2010. Empirical realised niche models for British higher and lower plants –
595 development and preliminary testing. *J. Veg. Sci.* 21: 643–656.

596 Smart, S. M. et al. 2014. Quantifying the impact of an extreme climate event on species
597 diversity in fragmented temperate forests: the effect of the October 1987 storm on
598 British broadleaved woodlands. *J. Ecol.* 102: 1273–1287.

599 Stace, C. 1997. *New Flora of the British Isles.* Cambridge University Press.

600 Tabarelli, M. et al. 1999. Effects of habitat fragmentation on plant guild structure in the
601 montane Atlantic forest of southeastern Brazil. *Biol. Conserv.* 91: 119–127

602 Thompson, K. & McCarthy, M.A. 2008. Traits of British alien and native urban plants. *J. Ecol.*
603 96: 853–859.

604 Vellend, M. et al. 2006. Extinction Debt of Forest Plants Persists for More than a Century
605 following Habitat Fragmentation. *Ecology*. 87: 542-548

606 Verheyen, K. et al. 2003. Herbaceous plant community structure of ancient and recent
607 forests in two contrasting forest types. *Basic Appl. Ecol.* 4: 537–546.

608 Wearn, O. R. et al. 2012. Extinction debt and windows of conservation opportunity in the
609 Brazilian Amazon. *Science*. 337: 228–32.

610 Westoby, M. et al. 2002. Plant ecological strategies: some leading dimensions of variation
611 between species. *Annu. Rev. Ecol. Syst.*, 33: 125–159.

612 Willi, J. C. et al. 2005. The Modification of Ancient Woodland Ground Flora at Arable Edges.
613 *Biodivers. Conserv.* 14: 3215–3233.

614

615

616

617

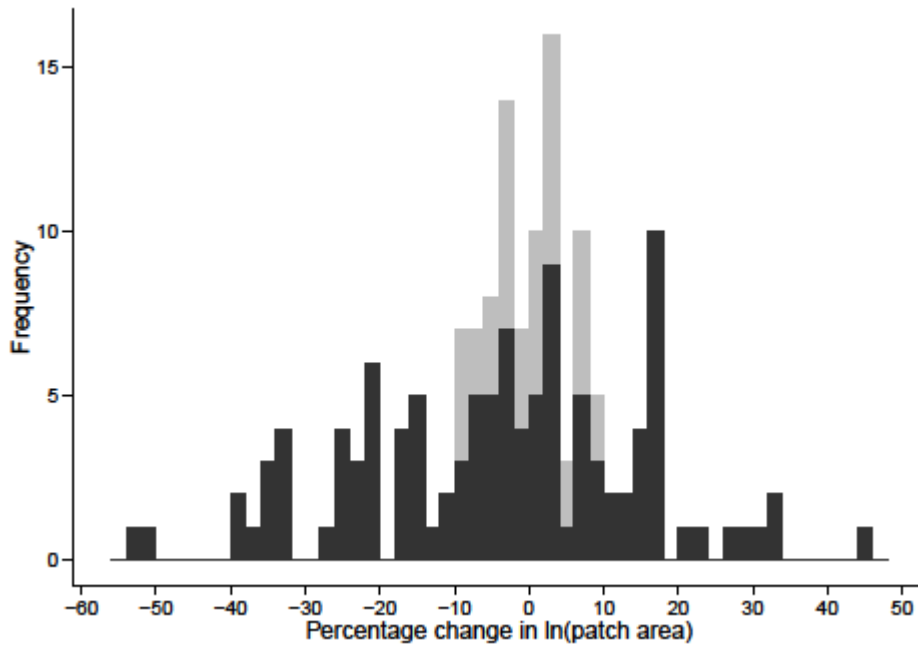
618

619

620

621 **Supplementary Materials**

622 **Appendix 1: Histograms showing the amount of change in each spatial variable for forests**
623 **patches.**



624

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

629

630

631

632

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

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

637

638

639

640

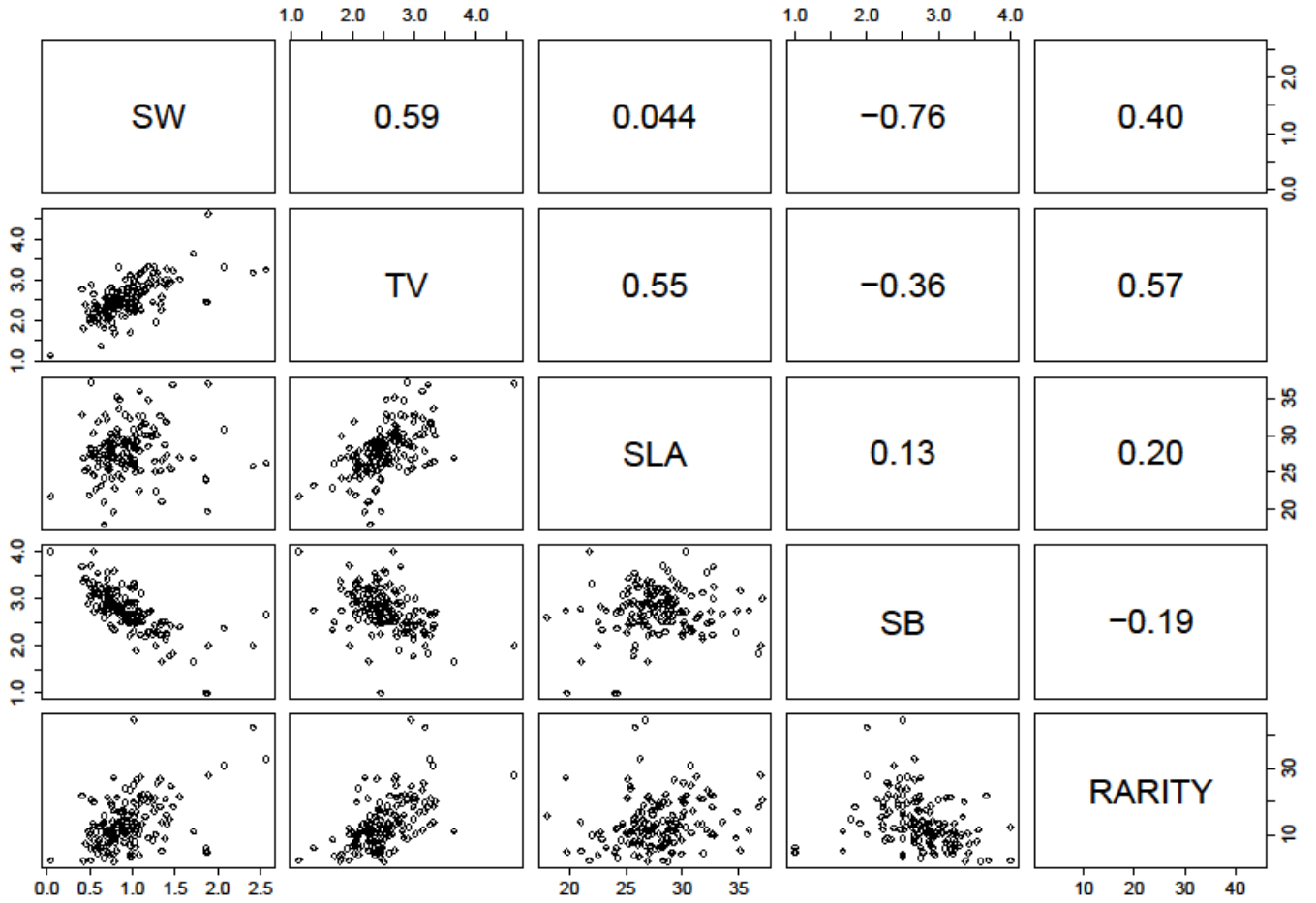
641

642

643

644

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



647 **Figure A3: Pairs plot displaying correlations between mean trait values within vegetation**
 648 **sampling plots.**

649