

1 Traits of plant communities in fragmented forests: The  
2 relative influence of habitat spatial configuration and  
3 local abiotic conditions.

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11 **Running headline:** Traits of plant communities in fragmented forests

12 **Summary**

- 13 1. The plant trait composition of forest fragments is thought to be partly determined by  
14 forest spatial properties, although the relative importance of habitat configuration  
15 and local abiotic drivers is poorly understood.
- 16 2. To address this issue, large-scale habitat extent data were combined with detailed  
17 field survey information for temperate broad-leaved deciduous forest patches to  
18 quantify the relative effects of spatial and abiotic filters on plant community mean  
19 trait values.
- 20 3. Local conditions such as shade and soil fertility had the largest effect on mean trait  
21 values, but aspects of habitat configuration also had significant partial effects on a  
22 number of traits.

23 4. Mean trait values within older forest patches were more strongly influenced by  
24 forest spatial configuration than in younger patches.

25 5. *Synthesis*. Results suggest that, in addition to the effects of greater light availability  
26 and competition in small patches and at forest edges, aspects of habitat  
27 configuration such as patch size and isolation are themselves important factors  
28 limiting the occurrence of forest specialist species. Large areas of core forest habitat  
29 contain a greater proportion of rare, poor dispersing species, although these effects  
30 were less visible in more recently established forest. This highlights the importance  
31 of maintaining existing large and old forest patches as a refuge for forest specialist  
32 plants. The results of this comparison of spatial and abiotic variables suggest that  
33 controlling the spatial properties of forest patches is likely to prove an effective way  
34 of managing plant species diversity, provided that sites with appropriate abiotic  
35 conditions are chosen.

36 **Key-words:** dispersal traits, environmental conditions, forest age, forest conservation,  
37 habitat fragmentation, patch area, rarity, seedbank persistence, specific leaf area.

## 38 **Introduction**

39 Forests, particularly those of long continuity, are a conservation priority in many areas due  
40 to their potential to act as a refuge for rare or threatened species (Peterken & Game, 1984;  
41 Wulf, 1997). An assessment of the way in which the spatial configuration of these habitats  
42 affects species with different life history traits is therefore essential to allow accurate  
43 modelling of the impacts of ongoing landscape change on forest specialist plants.

44 Urbanisation and agricultural intensification have dramatically changed landscapes  
45 worldwide, causing the fragmentation and loss of many habitat types (Foley *et al.* 2005). In a

46 fragmented landscape, habitat availability is reduced for target organisms, with favourable  
47 patches generally smaller and less well connected. Consequently the populations of species  
48 which are dependent on this habitat may be smaller and at greater risk of localised  
49 extinction (Fischer & Lindenmayer, 2007). Here, the effects of a number of variables  
50 describing forest configuration, condition and history upon plant community mean trait  
51 values were investigated. The aim was to quantify the partial covariance between mean  
52 trait values and forest spatial configuration given variation in patch age, soil quality and  
53 levels of shade, thus allowing the strength of the effects of both spatial properties and local  
54 abiotic conditions on mean trait values within forest patches to be compared.

55 Although species with particular life history traits exhibit a negative response to habitat loss  
56 and fragmentation, the occurrence of most plant species is dependent upon habitat quality  
57 rather than habitat configuration (Dupré & Ehrlén, 2002). Direct filters such as substrate pH,  
58 soil moisture and macronutrient availability within forests are strongly related to plant  
59 species occurrence (Critchley et al. 2002; Corney et al. 2006; Smart et al. 2010) and as such  
60 may be the biggest drivers of mean trait values. Light availability at ground level is also  
61 important, since a number of forest specialists possess traits such as high specific leaf area  
62 and small stature which make them well adapted to low light conditions (Hermy et al. 1999).  
63 It was therefore hypothesised that mean trait values in this study would be more strongly  
64 affected by local abiotic condition variables than forest spatial configuration.

65 Species that are most vulnerable to the effects of landscape fragmentation and habitat loss  
66 tend to be those that have characteristics that do not favour effective dispersal in space or  
67 time (Henle et al. 2004; Kolb & Diekmann, 2005; Ockinger et al. 2010). Species with fast  
68 falling seeds and no persistent seedbank are generally less able to rescue threatened

69 populations through immigration from nearby patches (Jacquemyn et al. 2003; Ozinga et al.  
70 2009; Schleicher et al. 2011; Jacquemyn et al. 2012). Dispersal method is also important,  
71 with abiotically dispersed species more prone to extinction in fragmented landscapes than  
72 those which are capable of dispersing via animal vectors (Marini et al. 2012). Consequently  
73 species possessing traits such as these occur less frequently in small, isolated habitat  
74 patches, which become dominated by a higher proportion of more persistent, longer lived  
75 and better dispersed species (Kolb and Diekmann 2005; Lindborg, 2007). Factors such as  
76 patch area and the amount of forest habitat in the surrounding landscape should therefore  
77 be important determinants of mean trait values within forest patches, as should distance to  
78 the nearest forest edge, since species dependent upon the interior of forest patches tend to  
79 possess traits linked to higher shade tolerance and lower dispersal capability (Hermy et al.  
80 1999, Pellissier et al. 2013). The presence of lag effects, which result in a lack of coupling  
81 between contemporary habitat structure and species composition (Lindborg & Eriksson,  
82 2001; Purschke et al. 2012) may however reduce the strength of this association.

83 Human activity has long lasting effects on abiotic conditions and therefore on patterns of  
84 species richness and composition within secondary forests (Dupouey et al. 2002; Vellend et  
85 al, 2007). As such, mean trait values within more recently established forest patches are  
86 likely to be more strongly determined by these historical environmental effects than by  
87 forest configuration, particularly where sited on former agricultural land with conditions  
88 which are unfavourable to many forest specialist species (Dupouey et al. 2002). Conversely,  
89 older patches are likely to contain a greater proportion of ancient woodland indicator  
90 species, characterised by poor competitive and colonising ability (Verheyen et al. 2003;  
91 Kimberley et al. 2013); traits which are likely to make them more vulnerable to habitat loss

92 and fragmentation. These species should be more frequent in large patches or far from the  
93 edges of forests, but mainly in the long-established habitat in which they almost exclusively  
94 exist due to time lags caused by the slow colonisation of younger forests by ancient  
95 woodland specialists (Jacquemyn et al. 2003). For these reasons, trait values in older forests  
96 should be more dependent on habitat configuration than those in younger forests.

97 In summary the following hypotheses were tested;

- 98 1. Variation in abiotic conditions exerts a stronger selective filter on mean trait values  
99 than forest spatial configuration and age.
- 100 2. Relationships between forest spatial configuration, patch age and mean plant trait  
101 values in British forest patches are still detectable having accounted for variation in  
102 abiotic conditions.
- 103 3. The spatial properties of older forest patches have a stronger effect on mean trait  
104 values than those of younger patches.

## 105 **Materials and methods**

### 106 **Plant trait data**

107 Plant species occurrence data were collected in 406 randomly stratified sampling plots  
108 located in broadleaved deciduous forest habitat within 1 km<sup>2</sup> regions across Great Britain as  
109 part of the 2007 Countryside Survey (Norton et al. 2012). Mean values for a number of life  
110 history traits within each of these plots were then obtained by averaging available trait data  
111 across species present. The mean trait values obtained were then used as response  
112 variables in the subsequent modelling. To allow the amount of shade present to be included  
113 as an explanatory variable without introducing circularity to the analysis, trees and shrubs

114 were excluded from this process. Mean trait values were left un-weighted by species'  
115 abundance. This approach places each species, whether subordinate or dominant, on an  
116 equal footing and avoids confounding the results by introducing the influence of variation in  
117 cover as a result of local competitive sorting. Plant trait information was obtained from the  
118 Electronic Comparative Plant Ecology database (Grime et al. 1995), the LEDA traitbase  
119 (Kleyer et al. 2008), Stace (1997) and PLANTATT (Hill, Preston & Roy, 2004). Species' rarity  
120 was obtained from PLANTATT (Hill, Preston & Roy, 2004) as the number of occurrences in  
121 British 10 km squares in the period 1987-1999.

122 In total, 445 species were present across the vegetation sampling plots. The difficulties in  
123 obtaining trait data for so many species meant that data were not available for all traits for  
124 all species. The five traits tested, along with the percentage of species with missing values  
125 were; log natural seed weight (23.4%), seed terminal velocity, (35.5%), specific leaf area  
126 (10.3%), seedbank persistence (39.6%) and rarity (0.9%). Following the Bayesian approach of  
127 Thompson & McCarthy (2008), missing trait values for species were drawn randomly from a  
128 posterior statistical distribution of trait values which was created based upon the  
129 distribution of known values for other species within the same genus and family, using a  
130 hierarchical model written in WinBUGS (Lunn et al. 2000). This approach is superior to  
131 simply imputing mean values since missing values were estimated taking into account all  
132 available information for related species.

### 133 **Local conditions**

134 The approach taken was to include measurement of influential abiotic conditions in the  
135 analysis but to treat them as "nuisance" covariates whose effects would be removed prior  
136 to estimating the magnitude of the effect of forest spatial configuration on mean trait

137 values. By including both the spatial characteristics of forest patches and data on local  
138 conditions in the same analysis it was possible to evaluate the relative importance of these  
139 different sets of variables in determining the mean trait values within forest habitat.

140 Two types of Countryside Survey vegetation sampling plot were employed in the analysis,  
141 linear plots (10 m<sup>2</sup> in area), located parallel to forest streamsides and forest tracks, and area  
142 plots (200 m<sup>2</sup> in area), located within the wider areal extent of each patch but not sampling  
143 a linear feature. Shade was estimated on a three point scale for all vegetation plots and  
144 plots designated unshaded, partially shaded or fully shaded by field surveyors. Within each  
145 of the area plots (n = 87) soil pH, volumetric soil moisture content and carbon to nitrogen  
146 ratio were measured based on a 15 cm topsoil sample taken at the same time as the flora  
147 was recorded in each plot. In the linear plots (n = 319) directly measured soil data were not  
148 available. Values within these plots were estimated using published equations derived from  
149 a national calibration of observed values of the three soil variables against the mean  
150 Ellenberg values of plants in 1033 plots from a stratified, random sample of the range of  
151 British vegetation types (Smart et al. 2010). The mean Ellenberg values used in these  
152 equations to generate soil variables were derived only from the trees and shrubs which  
153 were excluded from the calculation of mean trait values, thus avoiding the problem of  
154 circularity when the estimated soil variables were used to model mean trait values. In order  
155 to account for differences in response between the area and linear plots, plot type was  
156 included as a categorical explanatory variable. Climate and residual geographic variation  
157 across Britain were accounted for by the inclusion of the northing of each sample plot as a  
158 continuous explanatory variable (Corney et al. 2006).

159 **Spatial woodland variables & Patch age**

160 To determine the configuration of forest patches around vegetation samples, the geo-  
161 referenced Countryside Survey plot data was overlain with forest extent data obtained from  
162 Land Cover Map (LCM) 2007 using ArcMap 10.0 software (ESRI, 2011). LCM 2007 is a  
163 satellite-derived dataset containing information on the spatial extent of various habitat  
164 types across Great Britain (Morton et al. 2011). Use of this data set enabled the spatial  
165 configuration of broadleaved forest patches in the wider landscape around vegetation plots  
166 to be assessed.

167 Figure 1 illustrates the spatial variables obtained from the various data sources for an  
168 example Countryside Survey vegetation plot. These were “patch area” (the area of the  
169 forest patch containing the vegetation sampling plot), “distance to edge” (the Euclidean  
170 distance between each vegetation plot and the nearest point of forest edge) and “buffer  
171 forest” (the percentage of forest habitat within a 1 km buffer area around the vegetation  
172 plot). To reflect the fact that the majority of plant species have maximum dispersal  
173 distances of less than 1 km (Thompson et al. 2011), only forest habitat within 1 km of  
174 vegetation sampling plots was considered. Forest area further than this was therefore  
175 assumed to be too far away from vegetation plots to have a significant impact on trait  
176 values and therefore not included in this statistic, even where contiguous with patches  
177 within the 1 km area. Patch area and distance to edge were both natural log transformed  
178 prior to inclusion within the modelling, to reduce the positive skew in their distributions.

179 Finally, the age of forest patches was estimated using First Edition Ordnance Survey maps  
180 (County Series) dated from 1849 to 1899. Presence or absence of woodland patches on  
181 these historical maps was used to divide present day woodland patches into either younger



182 woodland (established after 1899, n = 255) or older woodland patches (established before  
183 1899, n = 151).

#### 184 **Statistical modelling**

185 Mean trait values within forest patches were modelled using the various local conditions  
186 and forest spatial properties described earlier. For each trait a full linear mixed effects  
187 model, containing all of the spatial and abiotic explanatory variables, was built. Interactions  
188 between forest age and patch area, buffer woodland and distance to edge were also  
189 included and all models were fitted using the Countryside Survey 1 km square identifier as a  
190 random intercept. This accounted for spatial autocorrelation caused by the presence of  
191 multiple plots within the same 1 km sampling region.

192 The R package MuMIn (Barton, 2012) was then used to fit all subsets nested within the full  
193 model described above. Models were standardised following the procedure of Grueber et  
194 al. (2011), in order to provide effect sizes on a comparable scale. To avoid bias resulting  
195 from the low ratio of observations to parameters, models were compared using an adjusted  
196 Akaike information criterion (AICc) statistic, as recommended by Burnham and Anderson  
197 (2002). AICc is a measure of model performance which compares the maximum likelihood  
198 estimate of models, adjusted for increasing model complexity. The model with the lowest  
199 AICc value is considered the best performing model (of the set tested). All models with AICc  
200 values within 4 of the lowest value were then selected as a “confidence set”, thus including  
201 possible models possessing a considerable level of empirical support (Burnham and  
202 Anderson, 2002). These confidence sets were then used to derive relative importance values  
203 and model averaged effect sizes for each explanatory variable. Relative importance  
204 represents the probability of a variable being present in the best performing model for a

205 particular trait, and was calculated in MuMIn using the relative Akaike weights of models  
206 within the confidence set (Burnham and Anderson, 2002). Model averaged effect sizes were  
207 calculated for each explanatory variable by averaging the parameter estimates across each  
208 model in which a given variable occurred. 95% confidence intervals were also generated for  
209 these effect sizes and a significant effect of a variable is indicated where the confidence  
210 intervals do not overlap with zero (Grueber et al. 2011). The resulting statistics provide a  
211 way of assessing which spatial, age and local variables affect each trait, and the magnitude  
212 of these effects.

213 Recent work on the same plant species pool showed that their traits helped discriminate  
214 ancient woodland specialists from other woodland species yet segregation of species into  
215 either group was not explained at all by phylogeny (Kimberley et al. 2013). It is therefore  
216 unlikely that ancestral relatedness is responsible for artefactual correlations between traits  
217 and the explanatory variables used to quantify forest age and patch geometry. For this  
218 reason phylogeny was not included in any analyses.

## 219 **Results**

### 220 **Selection probability**

221 The relative importance values shown in Table 1 indicate the probability of each explanatory  
222 variable being selected in the best performing model (of the set tested) for each plant trait.  
223 Where a selection probability > 0.50 the variable in question is more likely to be included in  
224 the best performing model than not, and is therefore considered an important predictor.  
225 Abiotic predictors had a selection probability > 0.50 in a higher proportion of cases (20 out  
226 of 30) than the spatial/age predictors (11 out of 35) although both sets of variables were

227 important predictors across the range of traits. This suggests that a strong local filtering  
228 effect is operating upon mean plant trait values but that forest spatial configuration is still  
229 an important driving factor.

### 230 **Model averaged effect sizes**

#### 231 **Seed weight**

232 Amount of shade present had the strongest effect on mean seed weight values in  
233 vegetation sampling plots, with significantly heavier seeds found in plots which were fully  
234 shaded compared to plots which were fully lit (Fig 2a and Table S1 in Supporting  
235 information). Increasing northing and C: N ratio were both found to lead to a significant  
236 reduction in mean seed weight, albeit with an effect size of lesser magnitude. Despite  
237 distance to nearest edge having a high probability of inclusion in the best performing model  
238 it was only found to have a weak effect on mean seed weight values.

#### 239 **Seed terminal velocity**

240 Seed terminal velocity was significantly affected by a number of the local condition  
241 variables, with amount of shade again having the strongest effect (Fig 2b, Table S2). More  
242 shaded plots were found to contain sets of species with faster falling seeds, as were plots  
243 with a low soil moisture content and a high soil pH value. Although the local condition  
244 variables had the strongest effect on mean seed terminal velocity values, the amount of  
245 buffer forest also had a significant effect on this trait, with species with faster falling seeds  
246 found in patches with more forest habitat in the 1 km buffer area. Furthermore, the effect  
247 size observed for this spatial variable was similar in magnitude to the effects of soil moisture

248 and soil pH. This suggests that the spatial structure of forest habitat is influencing plant  
249 species composition with a similar degree of strength to the local conditions.

250 A significant effect was also found for the interaction between forest patch age and distance  
251 to nearest patch edge. This suggests that the influence of core forest habitat depends upon  
252 the age of the patch in question. As Figure 3 suggests, the relationship between mean seed  
253 terminal velocity of plots and increasing distance to the edge is slightly stronger in older  
254 habitat than in younger.

### 255 **Specific leaf area**

256 Although none of the spatial variables tested were found to influence mean specific leaf  
257 area (SLA) values within plots (Figure 2c and Table S3), a strong relationship was identified  
258 between the levels of soil carbon present and mean SLA. Where soil C: N was high, lower  
259 SLA values were observed within plots. Increasing shade also had a significant, albeit smaller  
260 effect on this trait, with heavily shaded plots containing species with a higher mean SLA  
261 than more open plots. Mean SLA values were also higher in more northerly plots and in  
262 plots with lower soil moisture.

### 263 **Seedbank persistence**

264 The amount of shade present was found to have the strongest effect on mean seedbank  
265 persistence values, with fully-shaded plots containing species with a less persistent mean  
266 seedbank than non-shaded plots (Figure 2d). Weak but still significant relationships were  
267 also found between increasing soil pH and increasing Northing and higher mean seedbank  
268 persistence.

269 The interaction between age and patch area had a weak but non-significant (at the 95%  
270 confidence level) effect on mean seedbank persistence values within vegetation sampling  
271 plots (Figure 2d and Table S4) suggesting that the relationship between patch area and  
272 mean seedbank persistence may be stronger in older forests than younger forests. This is  
273 supported by Figure 3, where mean seedbank persistence decreases with increasing patch  
274 area in old forest patches but shows little response in younger forest patches.

## 275 **Rarity**

276 Forest patches with high levels of shade and soil C: N contained a greater proportion of rare  
277 species. Conversely, plots in the north of Britain were found to have, on average, species  
278 which are more common (Figure 2e and Table S5). Again, the spatial variables did not have a  
279 significant effect on rarity considering all forests together, but patch area was found to have  
280 a stronger effect on mean rarity in older forests, shown by the significant effect of the  
281 interaction between patch area and patch age in Figure 2e. Figure 3 suggests that, in older  
282 forest patches, as the area of forest patches increases, the average rarity of species present  
283 increases.

## 284 **Discussion**

### 285 **Abiotic conditions**

286 As expected, abiotic conditions within forest patches were found to be key determinants of  
287 plant species composition. Principal amongst these was the amount of shade in vegetation  
288 plots, which had the strongest effect on four of the five traits tested. These patterns likely  
289 reflect the different strategies needed to survive in relatively open woodlands compared

290 with more dense forest habitat. For example, greater light availability has been shown to  
291 favour species possessing traits associated with a high relative growth rate, such as lower  
292 seed mass (Reich et al. 1998). Such patterns were observed in this study; well lit forests  
293 contained species with significantly lower mean seed weights and mean seed terminal  
294 velocities compared to plots which were fully shaded. Species found in shadier patches had  
295 a less persistent seedbank on average, possibly since soil turnover is an unpredictable and  
296 rare event in forest environments. Persistent banks of buried seeds are a less common  
297 regenerative strategy in these conditions than, for example, non-flowering ramets or  
298 cohorts of persistent juveniles (Grime, 2001).

299 The effect of increasing shade on mean SLA values supports previous work showing that,  
300 under low light conditions, shade tolerant species possess higher SLA (Hodgson et al. 2011).  
301 In temperate broadleaf forests such as those studied here, thinner leaves, and hence higher  
302 SLA, promote greater light capture for least expenditure on structural tissues which can  
303 then afford to be shed every autumn. This is in contrast to tropical forest trees where the  
304 longer growing season favours year round photosynthesis and growth but at a cost of  
305 greater investment in structural tissue, resulting in lower SLA (Baltzer & Thomas, 2010).

306 Increasing soil C: N ratio had the strongest effect on mean SLA values; on more productive  
307 soils (those with a low C: N ratio) mean SLA values were higher. Under these conditions high  
308 macronutrient availability can fund growth strategies that divert resources into rapidly  
309 accumulating plant biomass comprising leaves of low longevity and higher tissue nitrogen  
310 content (Ordoñez et al. 2009). This is consistent with the link between soil fertility and leaf  
311 mass per unit area revealed by the Leaf Economics Spectrum (Wright et al. 2004). Thus in  
312 temperate forests high SLA is not a reliable indicator of shade-tolerance associated with

313 ancient forest because in some areas high SLA can also indicate the presence of nutrient-  
314 demanding generalist herbs (Hodgson et al. 2011).

315 Other multivariate studies have assessed the effects of spatial and abiotic factors on  
316 community composition using species occurrence data, thus only accounting for their  
317 overall effect on various different traits (Foster et al. 1998; Vellend et al. 2007). In this study  
318 mean trait values were analysed separately, allowing the differences in the way traits  
319 respond to important variables to be detected. Care must be taken when interpreting these  
320 results however, due to correlations between pairs of traits. For example, part of the  
321 observed effect of shade on seedbank persistence may be due to the close relationship  
322 between this trait and seed mass (Westoby et al. 2002) which is also linked to light  
323 availability.

#### 324 **Importance of habitat configuration**

325 Previous studies have related changes in the composition of forest vegetation with  
326 alterations in environmental conditions and levels of disturbance following land use changes  
327 (Foster et al. 1998). Our findings confirm the links between prevailing abiotic conditions  
328 within forest patches and mean community trait values, but also indicate that forest habitat  
329 configuration has an important effect. Rare species with fast falling seeds and no persistent  
330 seedbank responded to both the area of forest patches and the amount of surrounding  
331 forest habitat, even when abiotic factors were accounted for. This suggests that such  
332 species are not restricted to large, old forest fragments solely due to the increased  
333 disturbance and competition at the edges of small or young patches, but also because  
334 aspects of landscape context such as patch size and isolation are acting as important filters  
335 on the occurrence of these species.

336 Conservation strategies often centre on protecting and increasing areas of existing forest  
337 habitat (e.g. Forestry Commission, 2011); particularly in land sparing scenarios where large  
338 habitat reserves, separate from an agricultural matrix, are the main focus of effects to  
339 conserve biodiversity (Phalan et al. 2011). The results of this study provide some support for  
340 such measures, since positive relationships were found between the presence of large core  
341 forest areas and the occurrence of rare, poorly dispersing species, even after accounting for  
342 abiotic environmental conditions. Increasing the size of forest patches should therefore help  
343 to promote the occurrence of many forest specialist species. The effects of patch area and  
344 buffer forest however were often secondary to those of abiotic factors such as shade and  
345 soil C: N ratio. This suggests that the maximum benefit to these species will be obtained by  
346 focussing conservation and restoration efforts on areas where the soil and shade conditions  
347 are most favourable. Attempts to increase characteristic forest biodiversity on unfavourable  
348 sites may lead to any positive effects of extra available tree cover being negated by the  
349 stronger effects of soil fertility and light availability. In addition to the effects of forest  
350 configuration observed here, other aspects of landscape structure may also be important  
351 determinants of mean trait values. The heterogeneity of the matrix landscape around forest  
352 patches for example is likely to affect the ability of poorly dispersing species to colonise  
353 habitat patches (Matlack & Monde, 2004), while large amounts of nearby woody linear  
354 features may act as a refuge for forest specialist plants, increasing resilience to patch area  
355 and isolation (Petit et al. 2004). In addition to these other important possible covariates, the  
356 high levels of noise relative to signal found in large-scale randomised survey samples such as  
357 Countryside Survey (Smart et al. 2012), may explain the small effect sizes seen here for most  
358 variables.



359 As hypothesised, plant community mean trait values within younger forest patches were  
360 not strongly affected by forest spatial structure, possibly due to the absence of the  
361 inefficient dispersers which are most affected by habitat structure and typify older forests  
362 (Verheyen et al. 2003; Schleicher et al. 2011). Large areas of young forest habitat may be  
363 missing many of these species, despite providing suitable habitat. This immigration credit  
364 (Jackson & Sax, 2010) may result in a future change in trait composition towards one that  
365 more strongly reflects patch spatial characteristics, as the recently established forest is  
366 gradually colonised by poorer dispersers. Any newly created habitat however is likely to take  
367 time to realise benefits to biodiversity, with studies suggesting that secondary woodlands  
368 take around 70 years to develop a similar level of species diversity to ancient forests (Flinn &  
369 Vellend, 2005). Achieving this rate of community assembly also critically depends upon  
370 adjacency to existing ancient forest (Brunet et al. 2011).

371 Ancient forest habitat is generally thought to be of higher conservation value due to its  
372 ability to sustain a large number of rare species that are considered less capable of  
373 colonising isolated younger forest (Peterken & Game, 1984). Our results suggest that this is,  
374 on average, only the case for large older patches. Smaller forests, even where they are of  
375 long continuity, are less able to support these rare species (Figure 3). Although ruderal  
376 species possessing lighter seeds and more persistent seedbanks are not characteristic of the  
377 flora of long continuity forest habitat, they were still found to dominate the flora of older  
378 yet smaller forest patches. Moreover, species with no persistent seedbank were more  
379 frequent only in forest patches which were both large and old (Figure 2d, Figure 3). Higher  
380 mean values for seed terminal velocity were also observed at greater distances to forest  
381 edge in older forests (Figure 2b, Figure 3), suggesting a clear distinction between core and

382 periphery species. In order to obtain the conservation benefits of old growth forest, such  
383 habitat must also be large in size and contain a high proportion of core habitat. Priority  
384 should therefore be given to measures that maintain and increase the area of old growth  
385 forest habitat where the aim is to conserve rare, poorly dispersing ancient forest specialist  
386 species.

387

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539 Table 1. Probabilities of spatial and local abiotic explanatory variables being included in the  
540 best performing model of the model set tested for five life history traits. Variables with a  
541 selection probability of greater than 0.5 are shown in bold

542

543	<b>Explanatory variable</b>	<b>Seed weight</b>	<b>Seed terminal velocity</b>	<b>Specific leaf area</b>
544		<b>Seedbank Persistence</b>	<b>Rarity</b>	<b>Important responses</b>
545	<b>Spatial/age variables</b>			

546	Distance to edge	<b>0.53</b>	<b>0.77</b>	0.40	0.34	<b>0.52</b>	3
547	Patch area	0.18	0.31	0.19	<b>0.91</b>	<b>0.82</b>	2
548	Buffer forest	0.21	<b>1.00</b>	0.19	0.44	0.28	1
549	Age	0.35	<b>0.78</b>	<b>0.58</b>	<b>0.65</b>	<b>0.58</b>	4
550	Age x Distance to edge		0.11	<b>0.74</b>	0.05	0.17	0.13 1
551	Age x Patch area		0.14	0.04	0.01	0.47	0.43 0
552	Age x Buffer forest		0.01	0.14	0.00	0.06	0.05 0
553	<b>Abiotic variables</b>						
554	Shade	<b>1.00</b>	<b>1.00</b>	<b>0.94</b>	<b>1.00</b>	<b>1.00</b>	5
555	C:N ratio	<b>0.75</b>	0.17	<b>1.00</b>	0.18	<b>1.00</b>	3
556	Soil moisture	<b>0.58</b>	<b>1.00</b>	<b>1.00</b>	0.19	0.47	3
557	Soil pH0.20	<b>1.00</b>	<b>0.50</b>	<b>1.00</b>	<b>0.72</b>		4
558	Plot type	0.16	<b>1.00</b>	<b>1.00</b>	0.10	0.21	2
559	Northing	<b>1.00</b>	0.49	0.22	<b>0.81</b>	<b>1.00</b>	3
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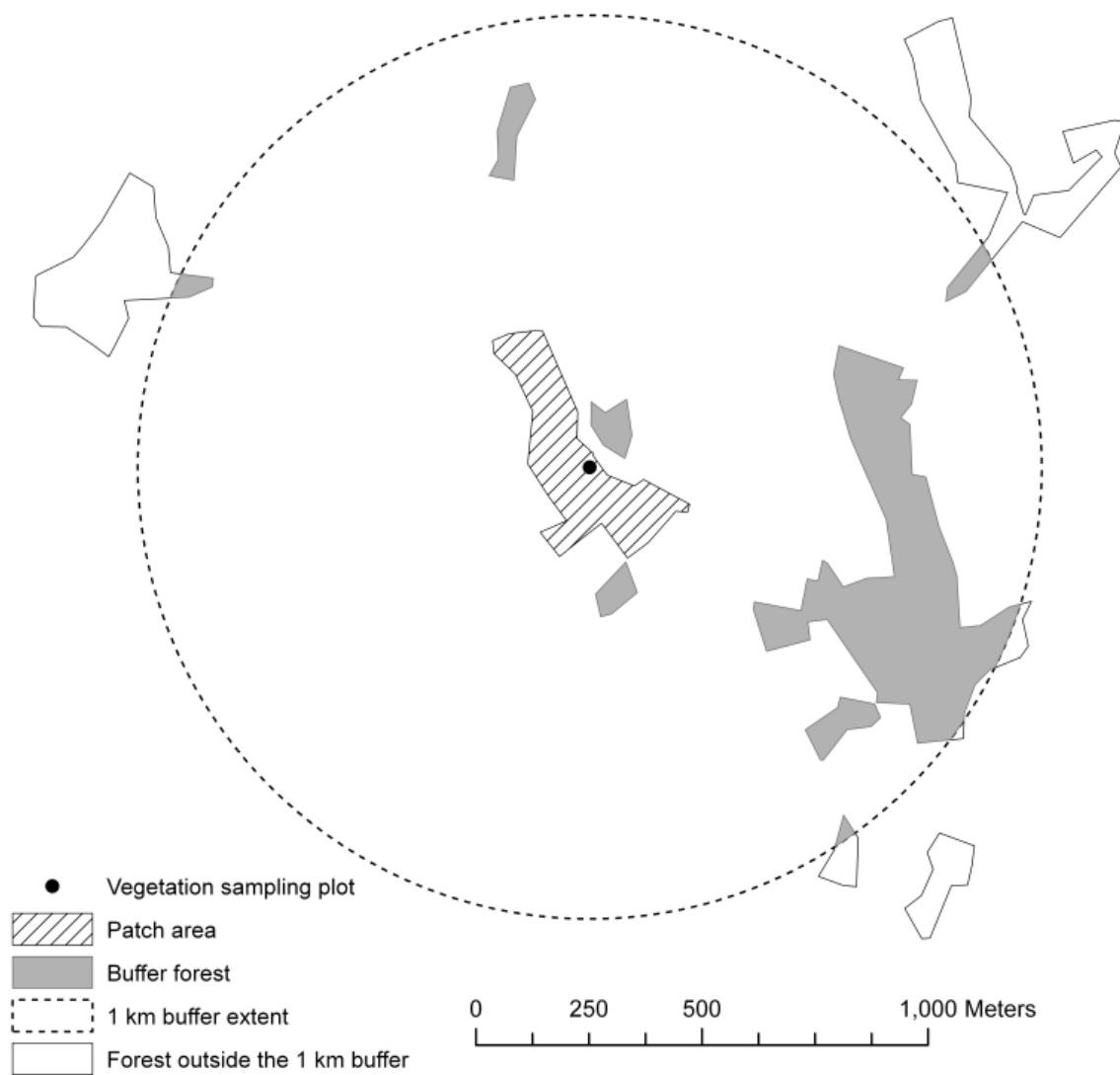
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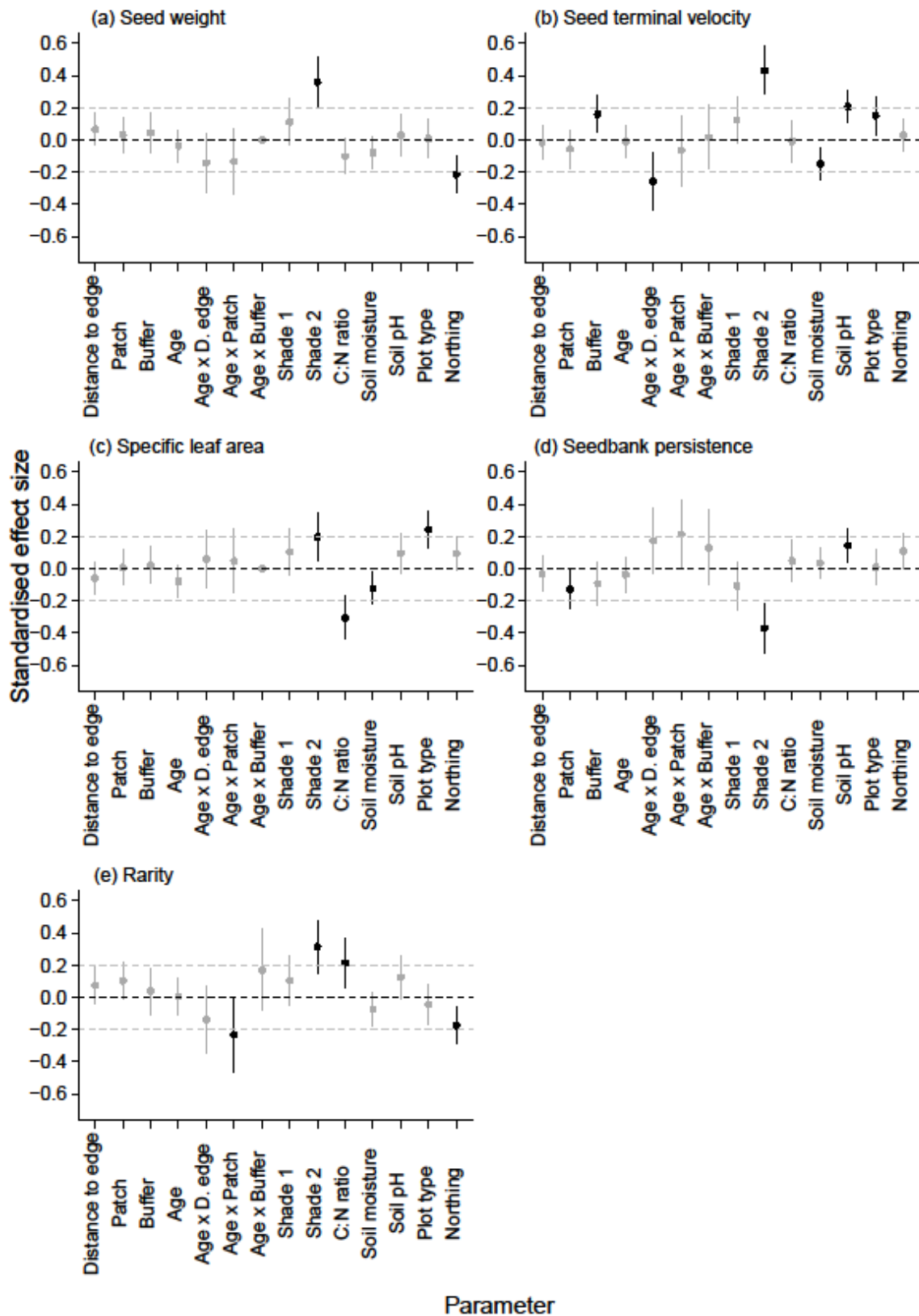
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573 **Fig 1.** Map showing an example Countryside Survey vegetation sampling plot and  
 574 surrounding Land Cover Map forest data. These data were used to calculate the various  
 575 spatial metrics for the patch in which the plot occurs. The hatched area of forest habitat  
 576 shows “Patch area” while the grey shaded area represents “Buffer forest”. Forest habitat  
 577 outside the 1km buffer area was not considered within the Buffer forest variable, even  
 578 where contiguous with patches inside the buffer.



579 **Fig 2.** Model averaged effect sizes of 14 explanatory variables on mean trait values in forest  
580 plots. Points show the average effect size taken from multimodel inference analysis, while  
581 the error bars indicate 95 percent confidence intervals. Where the confidence intervals do  
582 not overlap zero (black points), a significant effect is indicated. The further a point is from

583 zero, the stronger the effect. Dashed horizontal lines at +0.2 and -0.2 delimit small from  
584 medium sized effects according to Cohen (1988). Shade 1 shows the difference between  
585 unshaded and partially shaded plots, Shade 2 the difference between unshaded and fully  
586 shaded.

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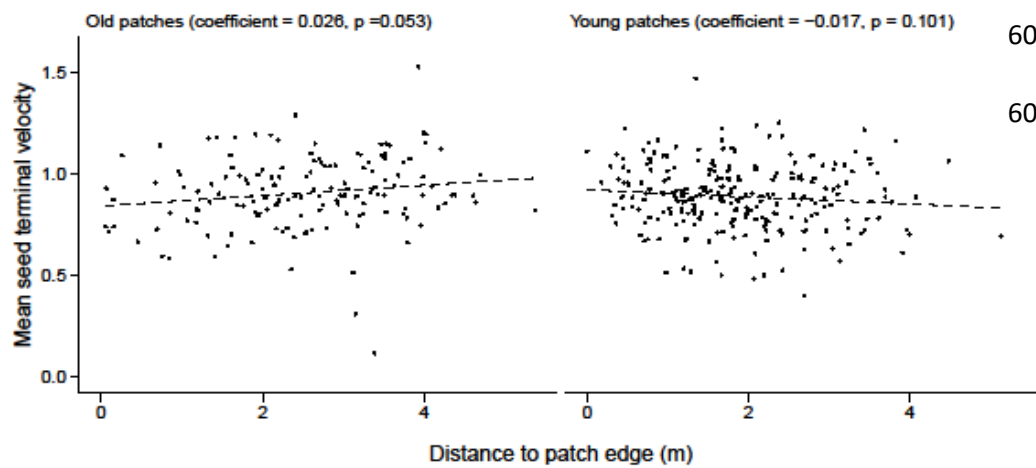
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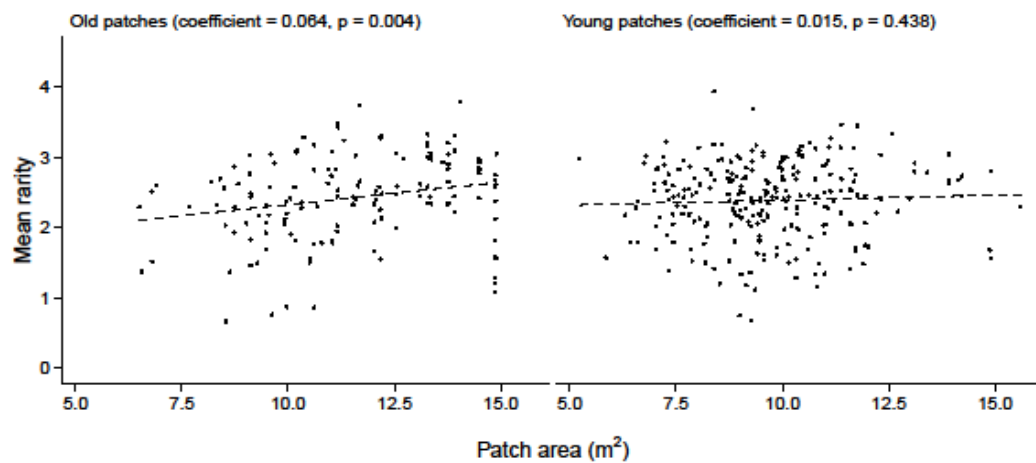
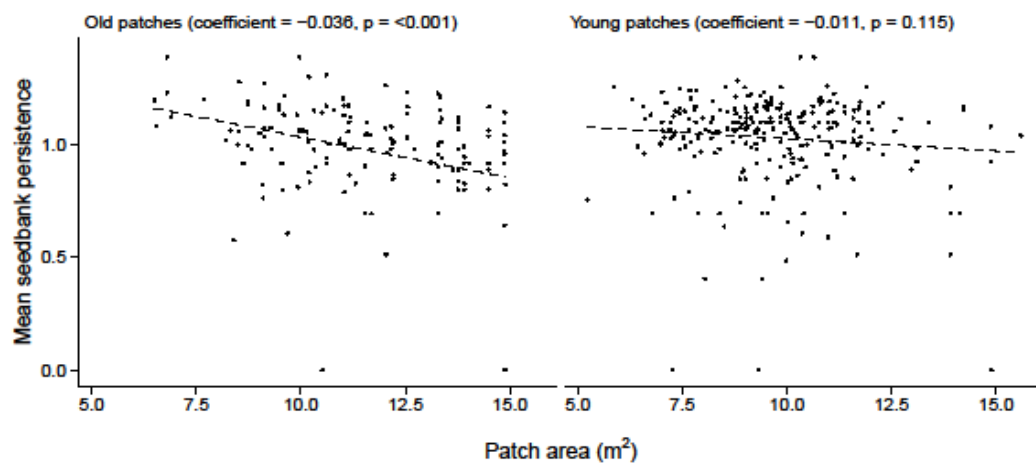
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606 Fig

607 3.



608 Relationships between spatial variables and mean trait values in older forest and young  
609 forest patches. Patch area and distance to patch edge were both log transformed. Dashed  
610 lines represent a linear model of trait versus spatial predictor. Regression co-efficients and *P*  
611 values for these models are also displayed.

612 **Supporting information**

613 **Appendix S1: Model averaged effect sizes and selection probability values of explanatory**  
614 **variables.**

615 Table S1: Model averaged effect sizes and selection probability values for seed weight

616 Table S2: Model averaged effect sizes and selection probability values for seed terminal  
617 velocity

618 Table S3: Model averaged effect sizes and selection probability values for specific leaf area

619 Table S4: Model averaged effect sizes and selection probability values for seedbank  
620 persistence

621 Table S5: Model averaged effect sizes and selection probability values for species rarity

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