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1 Identifying the trait syndromes of conservation
2 indicator species: How distinct are British
3 ancient woodland indicator plants from other
4 woodland species?

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15 **Abstract**

16 **Question**

17 Ancient woodland indicator species (AWIs) are plant species which are thought to be
18 restricted to areas of long continuity woodland habitat. In many cases however these
19 species have been identified on the basis of personal, to some extent, subjective
20 experience. Do the species proposed as AWIs according to these lists have traits in

21 common and how distinct is their trait profile from that of other woodland plant
22 species?

23 **Location**

24 United Kingdom

25 **Methods**

26 We applied classification tree analysis to a plant trait database to assess the extent to
27 which proposed AWI species can be clearly separated from other woodland plants
28 based upon their traits. We contrasted AWI species with an objectively defined list of
29 plants that are not considered to be AWIs but that have been commonly recorded in
30 woodlands. We also investigate the effects of phylogeny and region specificity on
31 species' proposed AWI status.

32 **Results**

33 The results provide support for the distinctiveness of plant species thought to be
34 associated with ancient woodland; they were found to be almost exclusively short,
35 perennial species, usually with a high seed weight. Results also indicate that rarer AWIs
36 have a more distinguishable trait profile than more common species. No link was
37 found between phylogeny and AWI status.

38 **Conclusions**

39 AWI species do have a distinguishable trait profile, despite their often partially
40 subjective selection. The results of the classification tree analysis suggest that traits

41 reflecting poor dispersal ability may be partly responsible for confining these species to
42 ancient woodlands. This confirms other studies that emphasise their low ability to
43 colonise secondary woodland sites and hence vulnerability to habitat conversion.

44 **Keywords**

45 Plant traits; classification tree; dispersal ability; phylogeny; rarity.

46 **Nomenclature**

47 Species nomenclature throughout is that of Stace (1997).

48 **Running Head**

49 Traits of ancient woodland indicator species.

50 **1. Introduction**

51 Ancient woodland indicator plants (AWIs) are vascular plant species that are
52 considered to be restricted to areas of long-established woodland habitat. Since they
53 were first proposed as a method of assessing the conservation value of woodland in
54 Lincolnshire by Peterken (1974), lists of plants which are considered AWIs in other
55 regions of Europe and North America have been developed (e.g. Honnay et al. 1998,
56 Motzkin et al. 1999, Verheyen et al. 2003).

57 Areas of ancient woodland, as defined by Peterken (1977), are considered a
58 conservation priority due to their ability to sustain a large number of rare or vulnerable
59 species that are unlikely to colonise isolated younger woodland (Peterken & Game,

60 1984). They may also act as refuges for species dependent on habitat types associated
61 with low farming intensity (Smart et al. 2006). As such, there have been efforts to map
62 remaining ancient woodland habitat (Goldberg et al. 2007) and to protect some of
63 these areas, for example in the UK through notification as Sites of Special Scientific
64 Interest and Priority Habitats under the UK Biodiversity Action Plan (BRIG, 2008). AWI
65 species provide a useful means with which to identify ancient woodland and a simple
66 tool to help assess woodland diversity and gauge the continuity of woodland cover,
67 although they should be used in conjunction with historical land use data (Spencer &
68 Kirby, 1992).

69 Despite the conservation importance of ancient woodland and the use of
70 indicator species in identifying such habitats, concerns remain over the way in which
71 species have been designated as AWIs, often based upon anecdotal evidence of their
72 association with ancient forest (Rolstad et al. 2002). Furthermore, few indicator
73 species are entirely restricted to ancient woodland (Wulf, 2003), meaning that a
74 subjective decision must be taken as to which species occur too frequently outside
75 ancient woodland habitat to be considered AWIs. Too stringent a set of requirements
76 and the resulting list of indicators will be too short to be useful, too loose a definition
77 of an AWI and less specialised plant species may reduce the effectiveness of the
78 indicators chosen (Rose, 1999).

79 Here we test whether lists of species suggested as AWIs for different parts of
80 Britain, often defined at least partly in a subjective way, do have distinctive traits such
81 that they might be considered as a guild of woodland specialists. An objective
82 classification tree method was used to explore differences between species that are

83 currently proposed as AWIs compared to non-AWI species by identifying fundamental
84 life-history traits that can be used to separate species from the two groups.

85 Previous studies have found differences in Ellenberg indicator values between
86 AWI and non-AWI species, with AWIs preferring low light conditions with soils of
87 intermediate nitrogen concentration and wetness (Hermy et al. 1999). However, these
88 Ellenberg values do not represent morphological or behavioural traits and hence offer
89 limited insight into the mechanisms of dispersal, establishment and persistence that
90 define AWI species.

91 The distribution of species associated with ancient woodland habitat has been
92 shown to be limited by dispersal ability and longevity (Wulf, 2003; Hermy & Verheyen,
93 2007). Short species with heavy seeds are thought to have lower ability to colonise
94 new habitat and adapt to land-use change (Verheyen et al. 2003; Hermy & Verheyen,
95 2007). Consequently we hypothesise that dispersal-related traits such as seed terminal
96 velocity and seed weight are likely to prove important factors that can be used to
97 group AWI species together. Due to the shade tolerance of AWI species and their
98 association with low to moderate macro-nutrient availability, specific leaf area (SLA)
99 was also expected to differ between AWIs and non-AWIs. While high SLA has been
100 associated with shade tolerance (Hodgson et al 2011) it is also strongly associated with
101 productive, human modified habitats. High SLA therefore may only be an effective
102 predictor of AWI status after taking into account the presence of other trait states that
103 differentiate species along the productivity and land-use intensity gradient.

104 When analysing the explanatory power of multiple traits across many species, it
105 is important to consider the fact that phylogenetic relatedness may result in non-
106 independence between species due to covariance among traits other than those
107 included in the analysis (Felsenstein, 1985). Using phylogeny as an explanatory
108 framework reduces the likelihood of misinterpreting ecological patterns that are
109 driven by common ancestry. AWI species may be largely restricted to certain
110 taxonomic groups. If this is the case, the phylogeny of these species may confound any
111 attempt to separate AWIs from non AWIs based upon specific traits. To investigate the
112 possibility that AWI species can be differentiated as effectively by their ancestral
113 relatedness as by the chosen traits, we performed a second, separate analysis which
114 also attempted to split proposed AWI species from non AWIs, in this case based solely
115 upon their phylogeny.

116 In Britain AWIs can be indicators of ancient woodland across the whole of their
117 range or only considered such in certain regions, despite being distributed much more
118 widely (Kirby, 2006). For example, some species may only be classified as AWIs in
119 relatively more intensively-managed landscapes because ancient woodlands provide
120 the only remaining favourable niche space. The same species may however be more
121 common in semi-natural habitats in less intensively-managed regions, and hence not
122 considered AWIs in these regions because they are evidently not restricted to ancient
123 woods. This wider niche breadth may therefore correspond with a trait profile less
124 readily discriminated from other non-AWI species that occur in the same mid or early
125 successional habitats.

126 We therefore hypothesise that species that are considered AWIs in only a small
127 number of local areas despite being widely distributed across many regions have a less
128 distinctive, more generalist set of traits than those which are AWIs across the whole of
129 their range. This should make them harder to separate from the non-AWI species pool.
130 Conversely, species may only be AWI in a subset of regions because they are rare.
131 Rarer AWIs may have an even more distinctive trait profile if the reason for their rarity
132 is the possession of specialised trait combinations that are associated with restriction
133 to ancient woods.

134 In this paper we test the hypothesis that proposed AWI species can be clearly
135 separated from non-AWI woodland species on the basis of traits linked to poor
136 dispersal and adaptation to low light availability during the peak growing season.
137 Having determined the trait differences between the two groups, we test two
138 hypotheses about the trait profiles of AWI species that are indicators only in certain
139 regions. First, that regional AWIs are less distinguishable from non-AWIs than pan-
140 national AWIs. Second, that those regional AWIs are more distinguishable from non-
141 AWIs but only where they are rare across Britain. Better knowledge of the different
142 sets of traits that are associated with AWI species should provide improved
143 understanding of why their distribution is restricted to ancient woodland and help to
144 develop more effective measures to identify and conserve their habitat in the future.
145 Trait analysis might also suggest other species that might be investigated as possible
146 ancient woodland indicators.

147 **2. Material and methods**

148 **2.1 Classification and regression tree analysis**

149 Classification and regression tree (CART) methods (Breiman et al. 1984) are a set
150 of analytical techniques that can be used to explore and model large sets of data. Their
151 ability to consider interactions between variables and to deal with missing values make
152 them well suited for modelling complex ecological datasets (De'ath & Fabricius, 2000).
153 Here, CART analysis was performed on a database of information on the life history
154 traits of British woodland plant species, using the "rpart" add-on (Therneau, Atkinson
155 & Ripley, 2012) in the statistical software R (R Development Core Team, 2011).

156 CART models are built by applying a series of splits to an input dataset. At each
157 split the data is divided into two groups based upon the value of the explanatory
158 variable (in this case the plant trait) that results in the groups produced being as
159 uniform as possible in terms of the response variable (here species' proposed AWI
160 status). By applying this method to the plant species data a tree model was produced
161 that identifies differences between the traits of the proposed AWI species and other
162 woodland plants (Figure 1). The extent to which the CART model was able to separate
163 the AWIs from non-AWIs at each split also provided a way of assessing the strength of
164 differences between the two groups of species for each trait, as well as the extent to
165 which the proposed AWI species share common characteristics. In order to further
166 investigate the way in which the tree model used the plant traits to group species as
167 either AWI or non-AWI, the final node into which each species was classified was also
168 extracted from the model (see Appendix 1, Table 1).

169 The usual procedure in CART modelling is to fit an overly large (and therefore
170 overfitted) tree model and then prune this back to its optimal level of complexity
171 according to assessment of the cross-validated error (Breiman et al. 1984). Here this
172 was achieved by carrying out 50 sets of tenfold cross-validation and taking an average
173 of the mean cross-validated error of each sized tree, following the method
174 recommended by De'ath & Fabricius (2000). This information was then used to
175 determine the level of tree complexity that provided the lowest mean cross-validated
176 error (here a tree with eight splits). The complexity parameter associated with this size
177 of tree (0.028) was then used in rpart to prune the full tree to its optimal size and
178 produce the classification tree model (Breiman et al. 1984). The control settings used
179 for the fitting function in rpart; the minimum number of observations in a node before
180 attempting a split and the minimum number of observations in a terminal node, were
181 set at 20 and 5 respectively. Changing these settings had little effect on the pruned
182 tree model. Surrogate variables were used where trait data were missing for a
183 particular split, using data for other variables to estimate the missing values (Breiman
184 et al. 1984). If all potential surrogates were missing then species were prevented from
185 continuing through the model rather than being sent in the majority direction (as is the
186 default in rpart). In this case sending observations the way of the majority would have
187 biased the model in favour of non-AWIs, particularly since AWIs had a higher
188 proportion of missing data.

189

190

191 **Table 1.** Summary of input variables used to fit the classification tree model.

Trait	Variable type	Possible categories and ranges of values	No. missing values	
			AWI (n = 138)	Non-AWI (n = 423)
Maximum height	Continuous	4-5800 centimetres	0	0
Lifespan	Categorical	Perennial/biennial/ annual	0	0
Growth form	Categorical	Woody species/grass/sedge/ forb/fern/other monocotyledon	0	0
Seed weight (weight of 1000 seeds)	Continuous	0.001-12980 grams	45	66
Seed terminal velocity	Continuous	0.110-5.42 metres per second	66	151
Specific leaf area	Continuous	3.64-86.10 millimetres squared per milligram	54	35
Seed bank persistence	Categorical	Transient seeds/seeds persist for a short time/some persistent seeds/large bank of persistent seeds all year round	39	0
Dispersed by wind	Boolean	True/false	43	0
Dispersed by water	Boolean	True/false	43	0
Dispersed by animal vector	Boolean	True/false	43	0
Dispersed by human vector	Boolean	True/false	43	0

192

193 **2.2 Testing for effects of phylogeny**

194 In order to test for relationships between species' phylogeny and their AWI
 195 status a second CART analysis was performed. This involved using molecular
 196 phylogenetic data on the genus, family and order of 1888 British plant species, taken
 197 from PLANTATT (Hill, Preston & Roy, 2004). These phylogenetic factors were used as
 198 explanatory variables in a classification tree model, which attempted to distinguish
 199 AWIs from non AWIs. The methods used to build and prune the tree model were those

200 described in section 2.1. The accuracy with which this model was able to classify these
201 species provided a way of assessing the strength with which AWI status is linked to
202 phylogeny, and therefore whether variation in AWI status can be reliably attributed to
203 species' traits.

204 **2.3 Effects of rarity and regional AWI status**

205 The classification tree analysis grouped proposed AWI species into one of two
206 categories based upon their traits; either identifying them as potential AWIs or as non-
207 AWIs. It was predicted that the probability of an proposed AWI species being identified
208 as an AWI would increase with species' rarity, since rarer AWIs were expected to have
209 a more distinct trait profile. However, species commonness and assignment as AWI
210 only in local regions should reflect a more generalist trait profile therefore associated
211 with a greater chance of being classified as a non-AWI. We used multiple logistic
212 regression in the R package MASS (Venables & Ripley, 2002) to test the hypothesis that
213 the probability of proposed AWIs being correctly classified by the tree model was
214 related to their rarity and the number of regions for which they are AWIs. Species' AWI
215 status in various areas of Britain; Derbyshire, Lincolnshire, Carmarthen, North
216 Yorkshire, Dorset, Worcestershire, Somerset and Angus is documented in Kirby (2006)
217 and a count of the number of these (eight) regions in which each species is considered
218 an AWI was used in the analysis. Species' rarity was determined from PLANTATT (Hill,
219 Preston & Roy, 2004) and measured as number of occurrences in British 10 km squares
220 in the period 1987-1999. The interaction between rarity and number of AWI regions
221 was also included in the model. Due to the degree of intercorrelation between rarity

222 and number of regions a type III likelihood ratio test was carried out to determine the
223 significance of the explanatory variables. This prevented the order in which variables
224 were entered into the model affecting the results. Out of the 138 AWI species used in
225 the CART analysis, 108 were included in the logistic regression, leaving out 29 AWI
226 species unclassified by the tree model due to lack of data and one species for which
227 information on regional AWI status was not available.

228 **2.4 Plant species data**

229 The species used in the classification tree analysis included 138 that had been
230 proposed as ancient woodland indicator plants (AWIs) in at least part of Britain, based
231 on the list collated by Kirby (2006) and 423 other woodland species not considered
232 ancient woodland indicators (non-AWIs) but recorded in quadrats located in woodland
233 as part of the 2007 Countryside Survey of Great Britain (Norton et al. 2012). This
234 approach enabled the use of randomly sampled representative data for woodlands
235 across Britain to define a species pool of non-AWIs that nevertheless occur in
236 woodland habitat. Crucially this reduced the extent to which differences between the
237 traits of AWIs and non-AWIs were obscured by trait differences linked to species
238 preferences for non-woodland habitats. The list of AWIs used was created by
239 combining twelve existing lists of proposed indicators across Britain drawn up by
240 numerous authors, as described in Kirby et al. (2012). Although a number of the
241 species on these lists were proposed as AWIs based upon independent data showing
242 their association with ancient woodland, some have been assessed based only upon
243 the judgement of the expert surveyors. By comparing the traits of these proposed

244 AWIs with those of other woodland species we aim to establish whether these species
245 do have a different set of characteristic traits and thus are a useful conservation tool.

246 Eleven plant traits were used to build the classification tree model (Table 1),
247 representing those life history attributes considered most likely to differ between
248 AWIs and non-AWIs. This included various dispersal related traits; seed weight, seed
249 terminal velocity and maximum recorded species height (Soons et al. 2004, Thomson
250 et al. 2011). A number of categorical variables were included in the model, relating to
251 species' ability to use a number of dispersal vectors. Species could be assigned more
252 than one dispersal vector; for example a species could be considered both wind and
253 water dispersed. Since recent work suggests that dispersal vector variables based upon
254 seed morphology are in fact weak predictors of the actual ability of species to disperse
255 through the landscape (Tackenberg et al. 2003; Eycott et al. 2007) we expected that
256 these variables would not be successful predictors of AWI status of woodland plants.

257 In addition to the dispersal centred traits, data on species' lifespan, seedbank
258 persistence, growth form and specific leaf area (SLA) were also used in the
259 classification model. SLA in particular has been shown to be a key trait in determining
260 plant species' resource use strategy (Westoby, 1998) and is also correlated with a
261 number other traits such as growth rate, leaf lifespan and leaf nitrogen content (Reich
262 et al. 1997). Together these traits therefore represented a number of the competitive
263 and shade tolerant strategies likely to differ between AWIs and non AWI species.

264 The trait information was obtained from the Electronic Comparative Plant
265 Ecology database (Grime et al. 1995), the LEDA traitbase (Kleyer et al. 2008) and other

266 reference materials including Stace (1997) and PLANTATT (Hill et al. 2004). Where
267 species' dispersal vectors were not available they were inferred from relevant
268 literature and by inspection of plant parts in the illustrations of the British Flora (Ross-
269 Craig, 1948-74).

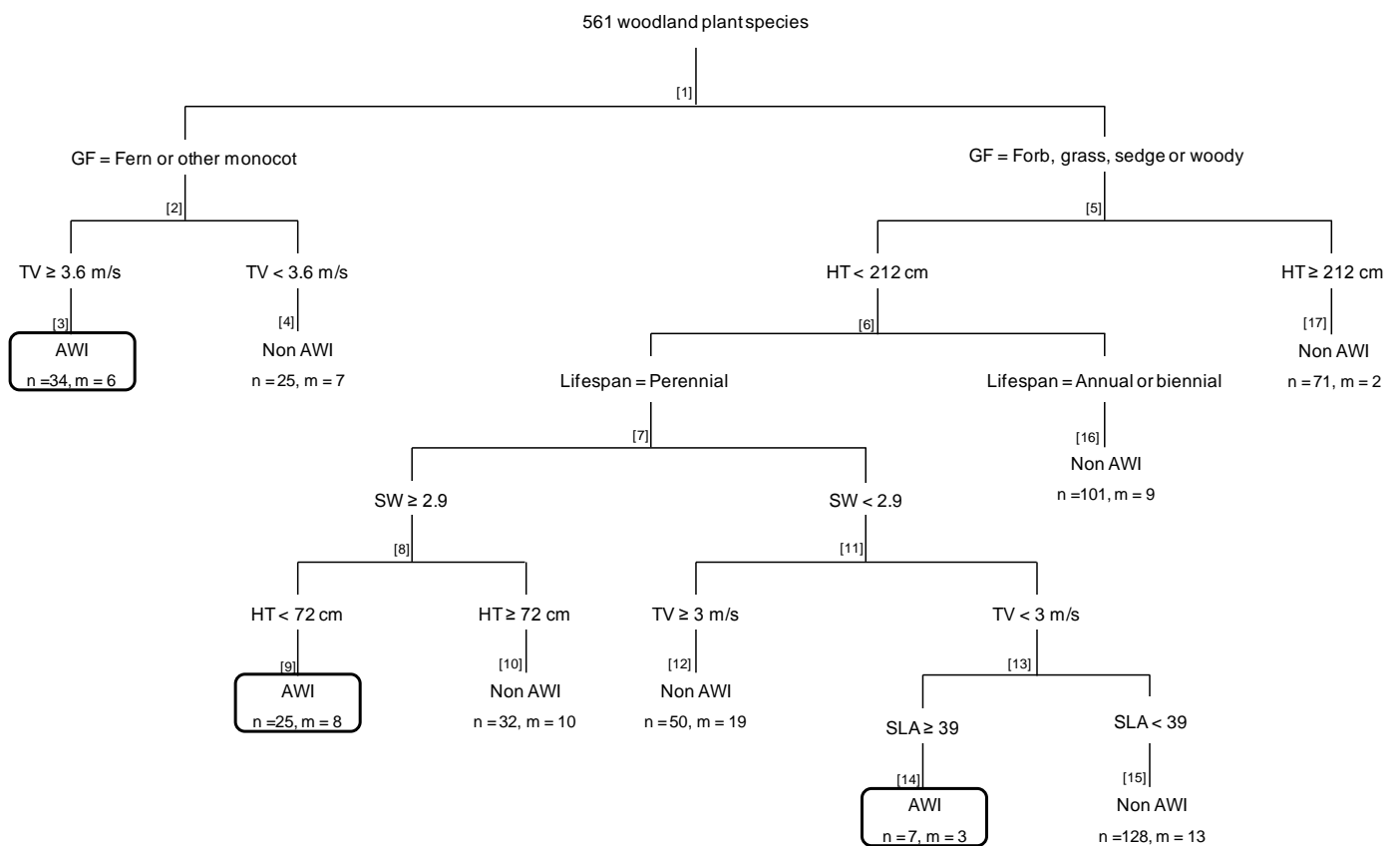
270 Although efforts were made to minimise gaps in the database through obtaining
271 information from as many sources as possible, the difficulty in obtaining trait data for
272 all species meant that a number of missing values were still present in the database
273 (Table 1). One advantage of CART techniques is their ability to handle missing values
274 without entirely removing incomplete records from the model; however rates of
275 misclassification may be higher for traits with a large number of missing values such as
276 seed terminal velocity due to the lower amount of information present.

277 **3. Results**

278 **3.1 Trait analysis**

279 The final classification tree model (Figure 1) retained six of the plant trait
280 variables tested; seed weight, seed terminal velocity, maximum species height,
281 lifespan, growth form and specific leaf area. None of the four dispersal vector variables
282 nor seedbank persistence were used by the tree model to discriminate between AWI
283 species and non-AWIs, although the effect of these traits may be represented by some
284 of the other variables, for example through the continuous variables describing seed
285 characteristics.

286 The tree model firstly separated ferns and other monocots (59 species, largely
 287 geophytes with underground storage organs) from other growth forms. The AWI status
 288 of the former group was best reflected by their seed terminal velocity; those with fast
 289 falling seeds were classified as AWIs, those with slow falling seeds as non-AWIs (Node
 290 2, Figure 1). At this node only 7 proposed AWIs were classed as non AWI species.



291 **Figure 1.** Classification tree model showing how different plant trait variables
 292 contribute to species' AWI status. Split abbreviations; GF = growth form, TV = seed
 293 terminal velocity, SLA = specific leaf area, SW = seed weight, HT = maximum height.
 294 Node labels are given in square brackets and can be cross-referenced to the species

295 lists in the appendix (Appendix 1, Table 1). n = number of species within each terminal
296 node, m = number of species misclassified at each terminal node.

297 In other growth forms (forbs, grasses, sedges and woody species) tall species
298 were not considered to be AWIs. Only two proposed AWI species had a maximum
299 height of greater than or equal to 212 cm, causing them to be classified as non-AWI
300 species according to the tree model (Figure 1). Among those plants shorter than 212
301 cm, most annual and biennial species were classified as non-AWI species, with 9
302 proposed AWIs terminating in this node, out of 101 species in total. Of the remaining
303 species (perennial forbs, grasses, sedges and woody species shorter than 212 cm),
304 species with light, slow falling seeds were not classified as AWIs unless they had an
305 extremely large SLA. Species with heavy seeds were classified as AWIs if shorter than
306 72 cm but not if taller than 72cm.

307 88 species were not classified due to missing values; 29 AWI species and 59
308 non-AWIs. The traits that most clearly distinguished the two groups were height and
309 lifespan; these two splits identifying 161 non-AWI species, while only including 11
310 proposed AWI species. The least certain group, node number 12 on Figure 1,
311 contained species with relatively light, fast falling seeds. This group contained almost
312 equal numbers of both proposed AWIs and non-AWIs.

313 **3.2 Phylogeny and AWI status**

314 When the genus, family and order of plant species were used to predict their
315 AWI status, the resulting classification tree did not retain any of the three explanatory
316 variables; an optimal tree model was returned which contained no splits. Including the

317 phylogenetic variables in this model only resulted in the cross-validated error of the
318 tree increasing. This provides strong evidence that phylogeny is not an effective
319 predictor of species AWI status.

320 **3.3 Regional AWIs**

321 Results of the logistic regression found no significant relationship between the
322 number of regions for which a species was considered an AWI and its probability of
323 misclassification (Chi squared = 0.0506, $p = 0.82200$). The interaction between rarity
324 and number of regions was also non-significant (Chi squared = 1.0808, $p = 0.29853$).
325 Rarity on its own however did have a significant effect, with rarer AWI species more
326 likely to be correctly classified by the tree model (Chi squared = 4.4219, $p = 0.03548$).

327

328 **4. Discussion**

329 The results of the CART analysis largely support the hypothesis that dispersal-
330 related traits are useful in discriminating AWIs from other plant species found in
331 woodlands. Maximum species height, seed weight and seed terminal velocity all
332 emerged as key correlates with AWI status. Phylogeny was found to have no influence
333 on species' AWI status, with none of genus, family or order being able to predict
334 species AWI status successfully. This indicates that AWIs are not confined to a
335 particular group of related species, rather being spread across a wider range of taxa.
336 Since none of the phylogenetic variables were capable of discriminating successfully
337 between AWI species and non AWIs, it is unlikely that the discriminating power of the

338 traits analysed here is confounded by the common ancestry of these species. Hence
339 these traits seem to be those which best explain the restriction of many proposed AWI
340 species to ancient woodlands.

341 Small stature, found in almost all AWI species, is associated with a number of
342 strategies for tolerating low light throughout much of the growing season (Westoby,
343 1998). Vernal species are constrained to complete seasonal leaf production and
344 flowering in the narrow window between unfavourable spring temperatures and
345 canopy leafing after which carbon fixation and biomass production is strongly light-
346 limited (Augsburger et al. 2005). Survival for these species may therefore centre on
347 tolerating or avoiding shade rather than growing woody biomass. Where light (or
348 another resource) is less limiting, taller species, identified almost exclusively as non-
349 AWIs, may have the competitive advantage.

350 AWI plants tend to be perennial species with heavy seeds; traits which other
351 studies have linked to poor colonising ability (Verheyen et al. 2003). Low dispersal
352 ability is thought under some conditions to reduce the ability of species to form viable
353 metapopulations, leading to higher vulnerability to habitat loss and fragmentation and
354 slower response to changes in landscape structure (Fischer & Lindenmayer, 2007). The
355 delayed response to landscape change shown by many perennial forest plants can lead
356 to an extinction debt forming in disturbed areas, with a number of existing species
357 destined for eventual extinction under the modified conditions (Eriksson, 1996;
358 Kuussaari et al., 2009). Many AWIs in fragmented habitat patches may therefore exist

359 as part of such remnant populations and consequently be at risk of future extinction
360 from such habitat.

361 As predicted, the dispersal vector variables were not useful in discriminating
362 between AWIs and other woodland plants. This is likely due to the poor ability of such
363 categorical variables based upon seed morphology to reflect observed dispersal rates
364 of plant species (Tackenberg, 2003).

365 In the classification tree model, traits such as growth form, lifespan and height
366 provided an effective initial separation between proposed AWIs and non-AWI species,
367 suggesting that the two groups tend to have distinct values for these characteristics.
368 Higher misclassification rates at nodes lower down in the tree model may occur
369 because important discriminating information has not been included, either because
370 the values for included traits are missing or because key traits have not been included.
371 However it may also mean that what is important in determining AWI status is the
372 interaction between the plant traits and their landscape context. For example if all that
373 is asked of an AWI is that it occurs much less in secondary woodland than in ancient
374 woodland this could still be consistent with a species occurring in a range of low-
375 productivity mid-successional habitats (e.g. Motzkin et al 1999). Species that are less
376 likely to occur in secondary woodland but can occur in other non-woodland habitats of
377 long continuity include those in node 4, such as the fern *Oreopteris limbosperma* and
378 the horsetails *Equisetum sylvaticum* and *E. telmateia*. These species are predicted by
379 the tree model to be non-AWIs since they have low seed terminal velocity (Figure 1;
380 Appendix 1, Table 1) and are widespread in Britain, occurring on linear features such as

381 road verges, streamsides and hedge banks, especially in the more oceanic west and
382 north. They are not however typical of the productive, disturbed conditions that often
383 persist as abiotic legacy effects within secondary woodland (Gilliam, 2007). These
384 species may therefore still be valid AWIs where their relative abundance in ancient
385 rather than secondary woodland is more important than their absolute restriction to
386 woodland. Other species where this applies include *Geranium sylvaticum* and *Stachys*
387 *officinalis*, both of which are considered AWIs, but also occur outside the woodland
388 environment in unimproved hay meadows, and *Cardamine amara*, *Conopodium majus*,
389 *Hypericum tetrapterum* and *Wahlenbergia hederacea* which occur widely in non-
390 woodland habitats but where they do occur in woodland this is more likely to be of
391 long continuity than secondary.

392 A number of widespread species (for example at node 9, *Cruciata laevipes*,
393 *Ranunculus ficaria*, *Symphytum tuberosum* and *Viola hirta*) associated with linear
394 features and were predicted to be AWI based on their trait sets. The management of
395 such features often involves infrequent pulse disturbance such as cutting that sets
396 back succession creating disturbance regimes and abiotic conditions that resemble
397 those of woodland gaps. Short perennial herbs with limited seed dispersal in space or
398 time are also characteristic of long-established meadows and pastures (Hodgson &
399 Grime, 1990) and hence such species might be classed as having AWI type traits.
400 Examples include *Cirsium acaule* and *Sanguisorba minor* (node 9; Appendix 1, Table 1)
401 both short perennials of grazed calcareous grassland and best considered as outliers
402 within the woodland species pool analysed. Adding in further traits related to shade

403 tolerance, along with traits that could discriminate grazing tolerance might have
404 allowed better separation of these species (Pakeman, 2004).

405 Preferences of some AWIs for non-woodland habitats may also mean that
406 species are only considered indicators in regions where the non-woodland habitat in
407 which they are found elsewhere in Britain is absent. The situation is however
408 complicated for species such as *Hyacinthoides non-scripta* where the range of habitats
409 they can occupy changes geographically as a function of temperature and not
410 necessarily habitat availability (Blackman & Rutter, 1954). Moving toward the western
411 fringes of the British Isles, mean minimum winter temperatures increase and this frost-
412 sensitive species becomes increasingly common in mid-successional habitats.

413 Node 15 comprised a large, well-differentiated group of perennial herbs with
414 light, slow falling seeds; likely to be more widely dispersed than the typical AWI
415 (Appendix 1, Table 1). Most were predicted to be non-AWI but a subset of proposed
416 AWIs were predicted to be non-AWI, including *Carex acutiformis*, *C.remota*, *Fragaria*
417 *vesca* and *Scrophularia nodosa*. All are either grazing intolerant or not favoured by
418 high productivity and so likely to find woodland a favourable refuge. Their wide
419 distribution may however make them less reliable as AWIs.

420 Rarity was found to have a significant effect on whether or not a proposed AWI
421 species was considered to possess AWI-like traits by the tree model. The rarity of these
422 species may be due to highly specialised sets of traits, such as preference for high
423 levels of shade and infrequent disturbance, which confine them to a narrow range of
424 conditions. These species are likely to be more dependent on ancient woodland

425 habitat and therefore more distinct from other woodland plants with a more general
426 set of traits and consequently looser association with old growth forest.

427 Other characteristics may differentiate between AWIs and other woodland
428 plants but for which trait data were not available. For example the amount of nuclear
429 DNA that a species possesses is associated with a number of plant traits such as shade
430 tolerance, phenology and generation time (Bennet, 1987) and as such might prove
431 effective in distinguishing AWIs from other woodland plants. Growth rate may also be
432 important, since plants with shade tolerant strategies have lower rates of growth
433 (Coley, 1988) thus typical AWI species may have slower growth than non-AWI plants.
434 Inclusion of relative growth rate in the classification tree model may have been able to
435 improve the rate of successful classification but we would expect the discriminatory
436 power associated with this trait to have been captured by specific leaf area given the
437 strong correlation between the two.

438

439 **5. Conclusions**

440 Clear trait-based patterns emerged from the CART modelling, suggesting that a
441 distinct trait profile is associated with AWI species: despite many lists being at least
442 partly based on subjective assessments they do appear to be a distinct guild of plants.
443 In summary an AWI species is most likely to be a short perennial with heavy, fast falling
444 seeds; often poorly dispersing species, not favoured by intensive disturbance regimes
445 and high productivity. Such a step constitutes a useful generalisation that subsumes

446 taxonomic identity and should aid further understanding of the mechanisms that
447 confine these species to older woodlands. This knowledge may help better
448 parameterise models of landscape connectivity for resilience mapping (e.g. Vos et al.
449 2008).

450 The functional distinctiveness of AWI species provides some support for the use
451 of such species as a group to identify areas of conservation importance. However we
452 also found trait-based similarities between many AWI species and non-AWIs that are
453 found in rarer, less frequently disturbed semi-natural habitats. Some of these might
454 merit further investigation to see if they might also be AWI where they occur in
455 woodland.

456 The strength of the association between these AWIs and ancient woodland
457 habitat depends on landscape context. This should be considered when using the
458 presence or absence of such indicator species to assess the conservation importance
459 of woodland habitat. Rarer AWI species were more clearly discriminated from non-
460 AWI woodland species on the basis of their traits and as such these species may be
461 most reliable as indicators of ancient woodland.

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467 **7. References**

468 Augspurger, C.K., Cheeseman, J.M. & Salk, C.F. Light gains and physiological capacity of
469 understorey woody plants during phonological avoidance of canopy shade. *Functional*
470 *Ecology* 19: 537-546.

471 Bennet, M.D. 1987. Variation in genomic form in plants and its ecological implications.
472 *New Phytologist* 106: 177-200.

473 BRIG (Biodiversity Reporting and Information Group) (ed. A. Maddock). 2008. *UK*
474 *Biodiversity Action Plan Priority Habitat Descriptions*. UK Biodiversity Partnership.
475 http://jncc.defra.gov.uk/PDF/UKBAP_PriorityHabitatDesc-Rev2011.pdf (PDF, 1.3Mb).
476 Accessed 13/12/2012.

477 Blackman, G.E. & Rutter, A.J. 1954. *Endymion nonscriptus* (L.) Garcke. *Journal of*
478 *Ecology* 42: 629-638.

479 Breiman, L., Friedman, J., Olshen, R. & Stone, C. 1984. *Classification and regression*
480 *trees*. Wadsworth International Group, Belmont, California.

481 Coley, P.D. 1988. Effects of plant growth rate and leaf lifetime on the amount and type
482 of anti-herbivore defense. *Oecologia* 74: 531-536.

483 De'ath, G. & Fabricius, K.E. 2000. Classification and regression trees: a powerful yet
484 simple technique for ecological data analysis. *Ecology* 81: 3178-3192.

485 Eriksson, O. 1996. Regional dynamics of plants: a review of evidence for remnant,
486 source-sink and metapopulations. *Oikos* 77: 248-258.

487 Eycott, A.E., Watkinson, A.R., Hemami, M.R. & Dolman, P.M. 2007. The dispersal of
488 vascular plants in a forest mosaic by a guild of mammalian herbivores. *Oecologia* 154:
489 107-118.

490 Felsenstein, J. 1985. Phylogenetics and the comparative method. *American Naturalist*
491 125: 1-15.

492 Fischer, J. & Lindenmayer, D.B. 2007. Landscape modification and habitat
493 fragmentation: a synthesis. *Global Ecology and Biogeography* 16: 265-280.

494 Gilliam, F.S. 2007. The ecological significance of the herbaceous layer in temperate
495 forest ecosystems. *BioScience* 57: 845-858.

496 Goldberg, E., Kirby, K., Hall, J. & Latham, J. 2007. The ancient woodland concept as a
497 practical conservation tool in Great Britain. *Journal for Nature Conservation* 15: 109-
498 119.

499 Grime, J.P., Hodgson, J.G., Hunt, R. & Thompson, K. 1995. *The Electronic Comparative*
500 *Plant Ecology*. London. Chapman & Hall.

501 Hermy, M., Honnay, O., Firbank, L., Grashof-Bokdam, C., Lawesson, J.E. 1999. An
502 ecological comparison between ancient and other forest plant species of Europe, and
503 the implications for forest conservation. *Biological Conservation* 91: 9-22.

504 Hermy, M. & Verheyen, K. 2007. Legacies of the past in the present-day forest
505 biodiversity: a review of past land-use effects on forest plant species composition and
506 diversity. *Ecological Research* 22: 361-371.

507 Hill, M.O., Preston, C.D. & Roy, D.B. 2004. *PLANTATT - attributes of British and Irish*
508 *Plants: status, size, life history, geography and habitats*. Centre for Ecology and
509 Hydrology, Huntingdon.

510 Hodgson, J.G. & Grime, J.P. 1990. The role of dispersal mechanisms, regenerative
511 strategies and seedbanks in the vegetation dynamics of the British landscape. In:
512 “*Species Dispersal in Agricultural Habitats*” Eds; Bunce, R.G.H., Howard, D.C. Institute
513 of Terrestrial Ecology. Belhaven Press, London, 65-81.

514 Hodgson, J.G., Montserrat-Marti, G., Charles, M., Jones, G., Wilson, P., Shipley, B.,
515 Sarafi, M., Cerabolini, B.E.L., Cornelissen, J.H.C., Band, S.R., Bogard, A., Castro-Diez, P.,
516 Guerrero-Campo, J., Palmer, C., Perez-Rontome, M.C., Carter, G., Hynd, A., Romo-Diez,
517 A., de Torres Espuny, L. & Royo Pla, F. 2011. Is leaf dry matter content a better
518 predictor of soil fertility than specific leaf area? *Annals of Botany* 108: 1337 - 1345

519 Honnay, O., Degroote, B. & Hermy, M. 1998. Ancient forest plant species in western
520 Belgium: A species list and possible ecological mechanisms. *Belgian Journal of Botany*
521 130: 139-145

522 Kirby, K. 2006. Ancient Woodland Indicator (AWI) plants, in Rose, F. (ed) *The wildflower*
523 *key*. Penguin Group, London.

524 Kirby, K.J., Pyatt, D.G. & Rodwell, J. 2012. Characterization of the woodland flora and
525 woodland communities in Britain using Ellenberg Values and Functional Analysis. In
526 *Working and Walking in the Footsteps of Ghosts: volume 1 the wooded landscape’*,
527 edited by I D Rotherham, M Jones, C Handley, Wildtrack Publishing, Sheffield pp66-86.

528

529 Kleyer M., Bekker R.M., Knevel, I.C., Bakker, J.P., Thompson, K.Sonnenschein, M.,
530 Poschlod, P., Van Groenendael, J.M., Klimeš, L., Klimešová, J., Klotz, S., Rusch, G.M.,
531 Hermy, M., Adriaens, D., Boedeltje, G., Bossuyt, B., Dannemann, A., Endels, P.,
532 Götzenberger, L., Hodgson, J.G., Jackel, A-K., Kühn, I., Kunzmann, D., Ozinga, W.A.,
533 Römermann, C., Stadler, M., Schlegelmilch, J., Steendam, H.J., Tackenberg, O.,
534 Wilmann, B., Cornelissen, J.H.C., Eriksson, O., Garnier, E. & Peco, B. 2008. The LEDA
535 traitbase: a database of life-history traits of the NW European flora. *Journal of Ecology*
536 96: 1266-1274.

537 Kuussaari, M., Bommarco, R., Heikkinen, K., Helm, A., Krauss, J., Lindborg, R., Ockinger,
538 E., Partel, M., Pino, J., Roda, F., Stefanescu, C., Teder, T., Zobel, M., & Steffan-
539 Dewenter, I. 2009. Extinction debt: a challenge for biodiversity conservation. *Trends in*
540 *Ecology & Evolution* 24: 564-571.

541 Motzkin, G., Wilson, P., Foster, D.R. & Allen, A. 1999. Vegetation Patterns in
542 Heterogeneous Landscapes: The Importance of History and Environment. *Journal of*
543 *Vegetation Science* 10: 903-920.

544 Norton, L.R., Maskell, L.C., Smart, S.S., Dunbar, M.J., Emmett, B.E., Carey, P.D.,
545 Williams, P., Crowe, A., Chandler, K., Scott, W.A. & Wood, C.M. 2012. Measuring stock
546 and change in the GB countryside for policy – key findings and developments from the
547 Countryside Survey 2007 field survey. *Journal of Environmental Management* 113: 117-
548 127.

549 Pakeman, R.J. 2004. Consistency of plant species and trait responses to grazing along a
550 productivity gradient: a multi-site analysis. *Journal of Ecology* 92: 893-905.

551 Peterken, G.F. 1974. A method for assessing woodland flora for conservation using
552 indicator species. *Biological Conservation* 6: 239-245.

553 Peterken, G.F. 1977. Habitat conservation priorities in British and European
554 woodlands. *Biological Conservation* 11: 223-236.

555 Peterken, G.F. & Game, M., 1984. Historical factors affecting the number and
556 distribution of vascular plant species in the woodlands of central Lincolnshire. *Journal*
557 *of Ecology* 72: 155-182.

558 R Development Core Team, 2011. *R: A language and environment for statistical*
559 *computing*. R Foundation for Statistical Computing, Vienna, Austria. ISBN 3-900051-07-
560 0, URL <http://www.R-project.org/>

561 Reich, P.B., Walters, M.B. & Ellsworth, D.S. 1997. From tropics to tundra: global
562 convergence in plant functioning. *Proceedings of the National Academy of Sciences* 94:
563 13730-13734.

564 Rolstad, J. Gjerde, I., Gundersen, V.S. & Sætersal, M. 2002. Use of indicator species to
565 assess forest continuity: a critique. *Conservation Biology* 16: 253-257.

566 Rose, F. 1999. Indicators of ancient woodland - the use of vascular plants in evaluating
567 ancient woods for nature conservation. *British Wildlife* 10: 241-251

568 Ross-Craig, S. 1948-74. *Drawings of British Plants, Vols 1-8*. Bell, London.

569 Smart, S.M., Marrs, R.H., Le Duc, M.G., Thompson, K., Bunce, R.G.H., Firbank, L.G. &
570 Rossall, M.J. 2006. Spatial relationships between intensive land cover and residual

571 plant species diversity in temperate farmed landscapes. *Journal of Applied Ecology* 43:
572 1128-1137.

573 Soons, M.B., Heil, G.W., Nathan, R. & Katul, G.G. 2004. Determinants of long-distance
574 seed dispersal by wind in grasslands. *Ecology* 85: 3056–3068.

575 Spencer, J.W., Kirby, K.J. 1992. An inventory of ancient woodland for England and
576 Wales. *Biological Conservation* 62: 77-94

577 Stace, C. 1997. *New Flora of The British Isles*. Cambridge: Cambridge Univ. Press

578 Tackenberg, O., Poschlod, P., Bonn, S. 2003. Assessment of Wind Dispersal Potential in
579 Plant Species. *Ecological Monographs* 73: 191-205.

580 Therneau, T.M. & Atkinson, B. (R port by Ripley, B.). 2012. *rpart: Recursive Partitioning*.
581 R package version 3.1-51. <http://CRAN.R-project.org/package=rpart>

582 Thomson, F.J., Moles, A.T., Auld, T.D. & Kingsford, R.T. 2011. Seed dispersal distance is
583 more strongly correlated with plant height than with seed mass. *Journal of Ecology*
584 99: 1299–1307.

585 Venables, W.N. & Ripley, B.D. 2002. *Modern Applied Statistics with S. Fourth Edition*.
586 Springer, New York.

587 Verheyen, K., Honnay, O., Motzkin, G., Hermy, M. & Foster, D.R. 2003. Response of
588 forest plant species to land-use change: a life-history approach. *Journal of Ecology* 91:
589 563-577.

590 Vos, C.C., Berry, P., Opdam, P., Baveco, H., Nijhof, B., O’Hanley, J., Bell, C. & Kuipers, H.
591 2008. Adapting landscapes to climate change: examples of climate proof ecosystem
592 networks and priority adaptation zones. *Journal of Applied Ecology* 45: 1722-1731.

593 Westoby, M. 1998. A leaf-height-seed (LHS) plant ecology strategy scheme. *Plant and*
594 *Soil* 199: 213-227

595 Wulf, M. 2003. Preference of plant species for woodlands with differing habitat
596 continuities. *Flora - Morphology, Distribution, Functional Ecology of Plants* 198: 444-
597 460

598 **List of appendices**

599 Appendix 1: Table of the 561 plant species used in the CART analysis.