

Niches for Species

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2 **Niches for Species, a multi-species model to guide woodland management: an**
3 **example based on Scotland's native woodlands**

4

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10

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12

13 **Abstract**

14 Designating and managing areas with the aim of protecting biodiversity requires
15 information on species distributions and habitat associations, but a lack of reliable
16 occurrence records for rare and threatened species precludes robust empirical
17 modelling. Managers of Scotland's native woodlands are obliged to consider 208
18 protected species, which each have their own, narrow niche requirements. To support
19 decision-making, we developed Niches for Species (N4S), a model that uses expert
20 knowledge to predict the potential occurrence of 179 woodland protected species
21 representing a range of taxa: mammals, birds, invertebrates, fungi, bryophytes, lichens
22 and vascular plants. Few existing knowledge-based models have attempted to include
23 so many species. We collated knowledge to define each species' suitable habitat
24 according to a hierarchical habitat classification: woodland type, stand structure and
25 microhabitat. Various spatial environmental datasets were used singly or in
26 combination to classify and map Scotland's native woodlands accordingly, thus

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27 allowing predictive mapping of each species' potential niche. We illustrate how the
28 outputs can inform individual species management, or can be summarised across
29 species and regions to provide an indicator of woodland biodiversity potential for
30 landscape scale decisions. We tested the model for ten species using available
31 occurrence records. Although concordance between predicted and observed
32 distributions was indicated for nine of these species, this relationship was statistically
33 significant in only five cases. We discuss the difficulties in reliably testing predictions
34 when the records available for rare species are typically low in number, patchy and
35 biased, and suggest future model improvements. Finally, we demonstrate how using
36 N4S to synthesise complex, multi-species information into an easily digestible format
37 can help policy makers and practitioners consider large numbers of species and their
38 conservation needs.

39

40 **1. Introduction**

41 Globally, biodiversity is under threat, many species are legally protected but resources
42 for conservation are diminishing (Bottrill et al., 2008; MacDicken et al., 2015;
43 Possingham et al., 2015). Maintaining habitat for species has been part of national and
44 international conservation planning for decades and networks of protected areas exist
45 globally (Orlikowska et al., 2016). However, whilst the IUCN has set a target of
46 designating 10% of terrestrial habitats as protected areas (IUCN, 1993), it is
47 recognised that this percentage of landcover, it's location, spatial configuration, and
48 the actions prescribed within it may not be sufficient to support species, particularly in
49 the face of rapid environmental change (Wiersma et al., 2018; Dinerstein et al., 2017;
50 Rodrigues et al., 2004).

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52 In the context of biodiversity protection in the temperate broadleaved and mixed
53 forest biome, where habitat restoration is a priority, the choice of where to apply
54 conservation effort for most benefit is critical (Dinerstein et al., 2017; Morales-
55 Hidalgo et al., 2015). Such decisions are often directed by international conventions
56 and directives on the environment, which are devolved to a regional level of
57 administration for implementation (JNCC, 2018; EC, 2018). For example, in the UK,
58 Scotland has listed 208 protected woodland species (mammals, birds, invertebrates,
59 fungi, bryophytes, lichens, herptiles and vascular plants which are strongly associated
60 with woodlands) (Scottish Action Coordination Group, 2008). Forestry policy and
61 practice have been designed to deliver habitat enhancement and protection measures
62 for these species (Forestry Commission, 2017), in line with wider conservation effort
63 targeting species which are rare and/or at risk of extinction (Favaro et al., 2014;
64 Winter et al., 2013). However, developing and adhering to these types of guidelines is
65 contingent on knowledge of what habitat features a species requires and how these are
66 distributed. This is complicated by the fact that many of these protected species are
67 cryptic and poorly recorded (Minin and Moilanen, 2014). The challenge is further
68 increased when there is a need to deliver conservation management for multiple
69 protected and data-deficient species simultaneously. This challenge is faced by many
70 land managers and owners.

71

72 To address gaps in species records and poor knowledge on habitat conservation needs,
73 research has focussed on predicting where species are likely to occur using empirical
74 models. These Species Distribution Models (SDMs) relate known species presence-
75 absence or presence-only data with environmental variables to determine species-
76 environment relationships and to predict habitat suitability over large extents (Elith

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77 and Leathwick, 2009; Guisan and Thuiler, 2005; Guisan and Zimmermann, 2000).
78 They have been widely used to characterise and map habitat suitability for single
79 species or taxonomic groups (e.g. Bellamy et al., 2013; Cooper-Bohannon et al.,
80 2016; Johnson and Gillingham, 2008). However, SDMs may fail to accurately predict
81 species habitat suitability when reliable occurrence data are sparse (Stockwell and
82 Peterson, 2002; Wisz et al., 2008; although see Pearson et al., 2007), the full range of
83 environmental variation across a species range is not represented (Austin 2002), the
84 species is not in equilibrium with its environment (Dormann 2007; Soberon and
85 Nakamura, 2009), or the impact of biotic interactions are not considered.

86

87 Whilst spatial data are available on broad woodland types across the UK (Forestry
88 Commission, 2011) and other fine-scale attributes for some UK woodlands (e.g.
89 dominant tree species, woodland structure, deadwood presence; Patterson et al.,
90 2014), species records available via Local Environmental Record Centres or online
91 data portals (e.g. NBN, 2017) typically suffer from sampling bias, low sample sizes
92 and a lack of confirmed absences. This is particularly the case for rare, inconspicuous
93 or cryptic species because of the difficulties in their detection or identification
94 (Phillips et al., 2009; Newbold 2010). In addition, despite advances in data portal
95 accessibility, the complexity and time investment involved in extracting high-
96 resolution records for several hundred species, filtering them for reliability and
97 accuracy, and interpreting the results alongside habitat data, means that this is
98 unlikely to be undertaken by forestry decision makers. Using well recorded and
99 better-known species as surrogates for wider biodiversity has been tested, but studies
100 show surrogates perform less well when used to represent other taxa e.g. birds

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101 representing butterflies (Dorey et al., 2018; Margules and Pressey, 2000; Prendergast
102 et al., 1993).
103
104 Expert-based habitat suitability models (EHSMs) provide a solution as useful
105 alternatives to SDMs when inadequate occurrence records preclude accurate empirical
106 modelling (Fourcade, 2016), or when funds for collecting new substantive datasets are
107 limited (Doswald et al., 2007; Fourcade, 2016; Murray et al., 2009). EHSMs use both
108 expert knowledge and evidence-based reviews from published scientific literature
109 describing a species' habitat requirements and ecology, combined with spatial
110 environmental datasets (e.g. land cover type, topography, aspect) describing the
111 availability of these habitats, to predict the occurrence of species (e.g. Eycott et al.,
112 2012; Ziegler et al., 2015). This approach has been extensively used by conservation
113 agencies in the USA, where many EHSMs have been developed by drawing on the
114 national resource of species specialist knowledge (Crance, 1987; Drew and Collazo,
115 2012; Drew and Perera, 2011). However, EHSMs are usually built for individual
116 species (e.g. Leblond et al., 2014) and validation is nearly always neglected (Iglecia et
117 al., 2012).
118
119 Here we present a multi-species EHSM approach, 'Niches for Species' (N4S), to
120 enable forest policy makers and managers to consider multi-species management
121 within Scottish forests. We use the term 'niche' to describe a set of habitat features
122 that a species is strongly associated with, from which we can estimate species
123 distributions whilst ignoring constraints such as competition. This is analogous to the
124 'potential niche', although we are only considering a narrow set of niche variables
125 (Jackson and Overpeck, 2000). Our aim was to provide a simple-to-interpret spatial

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126 modelling framework for predicting the distribution of suitable habitat for multiple
127 protected species. The main objectives were to develop an approach which could:
128 incorporate all protected species associated with woodland for an entire
129 (administrative) area; provide habitat requirement information for all those species;
130 predict the potential distributions of those species consistently across a range of
131 scales, whilst restricting predictions to climatically suitable areas where possible. Our
132 modelling approach was wider and more ambitious in scope (a greater number of
133 species and a wider range of taxa) than other attempts to inform conservation
134 planning with multi-species models (e.g. Franco et al., 2009; Lentini et al., 2015;
135 Minin and Moilanen, 2014) and as such is a novel application of EHSMs. Although
136 developed for protected woodland species, the framework could be adapted for use
137 with other habitats or suite of species. In addition, we aimed to test the model
138 predictions against species occurrence records, despite our concerns that the low
139 sample size, low resolution and high sampling bias associated with such records could
140 limit agreement with EHSM predictions.

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143 **2. Material and methods**

144 **2.1 The Niches for Species framework**

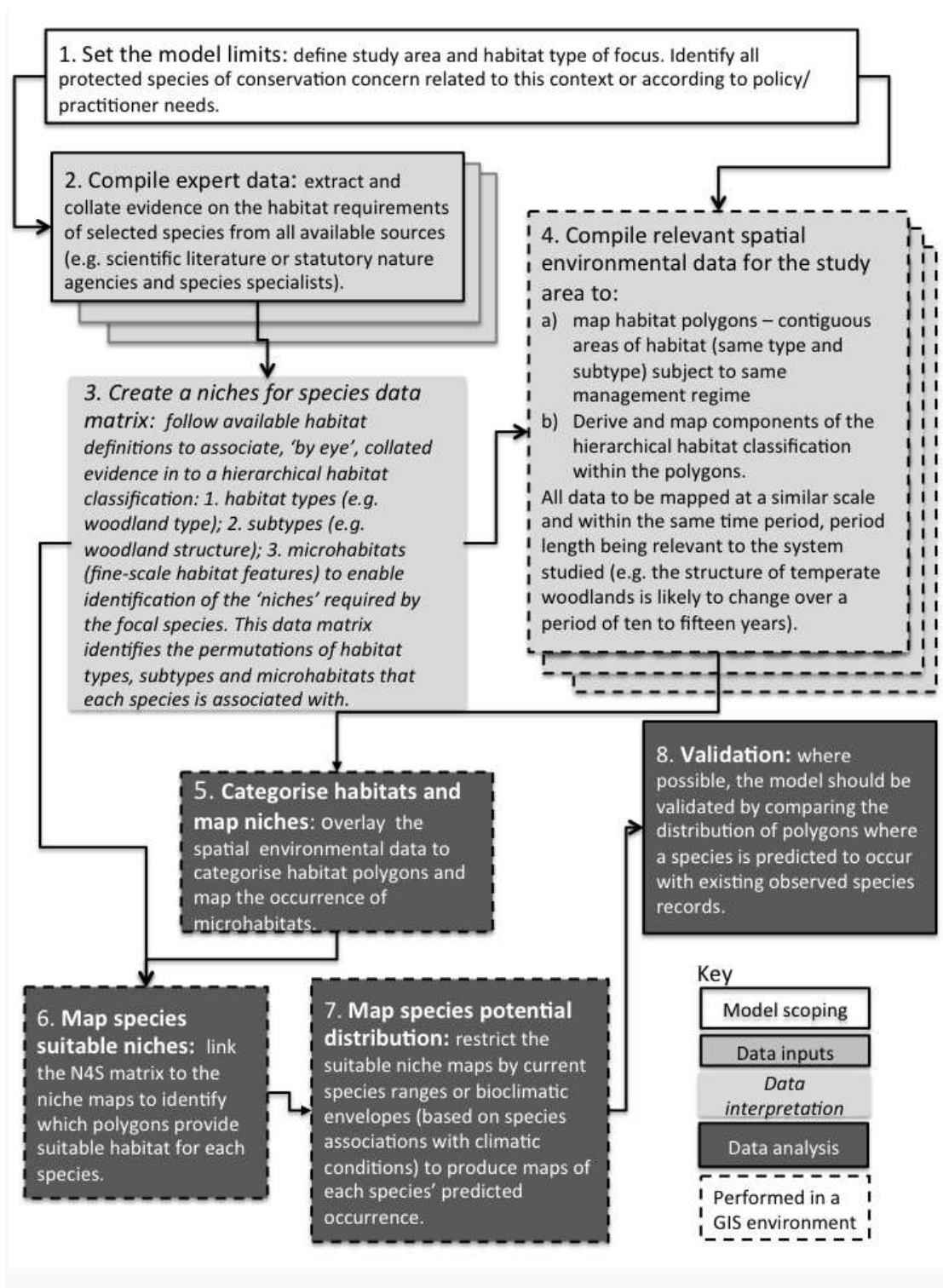
145 There are eight stages to the modelling framework (Figure 1). Stage 3 is unique to the
146 N4S methodology; the development of a hierarchical habitat classification provided a
147 structured system for categorizing species' niches. The incorporation of microhabitat
148 information is rarely implemented in these types of landscape-scale, spatial
149 approaches, despite their strong association with biodiversity (Michel & Winter,
150 2009). By nesting the levels, we take account of context dependency in species-

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151 microhabitat associations i.e. species microhabitats may only be important in certain
152 types and structures of the habitat. Stage 8 (validation) is rarely performed in EHSM
153 development. Details on how we have implemented these stages for woodland
154 protected species in Scotland are given in Section 2.2., along with the list of attributes
155 used and their sources (Tables 1 to 3). Output maps from Stage 7 can display single
156 species predictions or aggregate information by polygon to show predicted species
157 richness, for example.

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Figure 1: A schematic flow chart illustrating the steps involved in Niches for Species (N4S) expert-based habitat suitability modelling framework to map the distribution of niches and species potential occurrence.

170
171

172 **2.2 Our woodland application**

173 We applied the N4S model to map the potential distribution of woodland protected
174 species in Scottish native woodlands.

175

176 **2.2.1 Expert knowledge on species-habitat requirements**

177 We reviewed the available data documenting the habitat requirements for 208
178 protected species, considered to occur in Scotland and use woodland as their primary
179 habitat (Scottish Action Coordination Group, 2008). These represented a wide range
180 of taxonomic groups: lichens, bryophytes and liverworts; invertebrates; fungi; birds;
181 vascular plants; mammals; reptiles and amphibians. Evidence sources were classified
182 in to four categories:

183 **Evidence type 1**- information from habitat association analyses supplied directly to
184 the authors by species experts in the statutory nature agencies (Scottish Natural
185 Heritage, Natural England, Natural Resources Wales), and nature non-government
186 organisations (NGOs) (Butterfly Conservation, Plantlife Scotland, British Trust for
187 Ornithology). These sources were used particularly where peer reviewed information
188 was lacking on habitat associations under British conditions.

189 **Evidence type 2** – books and peer reviewed scientific articles detailing protected
190 species requirements; these were sourced by searching online journals and journal
191 directories. Example search strings and references used are shown in Table S1 in
192 Appendix A (online supplementary material).

193 **Evidence type 3** - information obtained from publications produced by nature
194 agencies and nature NGOs and from websites likely to be subject to peer-review e.g.
195 for Lepidoptera we used Butterfly Conservation (Butterfly Conservation, 2017)

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196 **Evidence type 4** – web sites where the review process was unconfirmed, and which
197 might include anecdotal evidence.

198 For most taxa, roughly half of the sources of evidence were peer-reviewed websites
199 and grey literature (evidence type 3), and the remainder were drawn evenly from the
200 other three sources of evidence (Table 1). Differences in the use of evidence source
201 by taxon is indicated when the percentage of data fields supported is considered
202 (Table 1). Here there is a reliance on specialist and less available knowledge (type 1
203 and type 1) for the more cryptic species (e.g. lichens, fungi), compared to more
204 widely accessible reports and information notes provided by nature conservation
205 NGO's and nature agencies (type 3), for the better-known taxa (e.g. birds, vascular
206 plants). Overall, only a low proportion of data fields were supported by type 4 sources
207 of evidence, where data accuracy is uncertain, as it may not have been confirmed or
208 checked by species experts (Table 1).

209

210 We collated the information systematically for each species, recording associations
211 with woodland type or tree species, and microhabitat requirements. Microhabitats
212 represent features of the habitat that may be present at a particular location for a
213 minimum of 5 to 10 years and offer particular microclimates and conditions which
214 may be used by some species only at certain times of the year. Details on species
215 requirements throughout the lifecycle, including differences at early and mature life
216 stages, where appropriate (e.g. for invertebrate species) were also collected. All
217 information included was referenced. A sample of the resulting database is given in
218 Table S2, Appendix A (online supplementary material).

219

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221 Table 1: Number of sources of evidence by evidence type (and the percentage of data
222 field entries supported) used in identifying habitat requirements, by taxon.

Taxon (number species)	Collated expert knowledge covering individual species (type 1)	Peer-reviewed papers and books (type 2)	Websites (known quality review process) and nature agency reports (type 3)	Websites (unknown review process) and anecdotal evidence (type 4)
Lower plants (Lichens , Liverworts and Bryophytes)(69)	3 (82%)	6 (2%)	5 (16%)	0
Invertebrates (52)	8 (2%)	7 (53%)	36 (28%)	13 (17%)
Fungi (21)	1 (6%)	1 (58%)	9 (36%)	0
Birds (16)	1 (33%)	4 (11%)	7 (56%)	0
Vascular plants (10)	1 (20%)	2(21%)	6 (45%)	5 (14%)
Mammals (8)	2 (43%)	3 (9%)	8 (48%)	0
Herptiles (Amphibians and Reptiles) (3)	2 (70%)	0	2 (30%)	0

223

224 For 179 of the 208 protected woodland species (69 lower plants (lichens, bryophytes
225 and liverworts); 52 invertebrates; 21 fungi; ten vascular plants; 16 birds; three
226 herptiles (amphibians and reptiles) and eight mammals), there was sufficient
227 information on habitat requirements for their inclusion in the N4S model. These
228 species were allocated to woodland niches.

229

230 **2.2.2 Habitat classification - Niches for Species (N4S) matrix**

231 We constructed a hierarchical woodland classification which captured the habitat
232 requirements for all species based on the collated expert information. Where possible,
233 the classification used established descriptors of woodland habitat already familiar to
234 forestry decision-makers e.g. woodland type and structure class (Figure 2):

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235 **i. Habitat type:** At the highest level of the habitat classification is woodland type.
 236 Seven native woodland types are recognised and described (Maddock 2008) (Figure
 237 2).
 238 **ii. Habitat subtype:** At the second level of the classification hierarchy is structure
 239 type. Any woodland type may have stands (representing a portion of the woodland
 240 with the same structure, size and age, and considered a single management unit)
 241 according to six structure types – these include five stand development stages and a
 242 sixth *permanently open* type (Table 2).

243 Table 2: Summary of structure types used in the classification of niches providing
 244 habitat for 179 protected woodland species in Scotland in the Niches for Species
 245 model. The structure types are based on the Native Woodland Survey Scotland
 246 (NWSS) survey criteria (NWSS, 2013; Patterson et al., 2014)

Structure type	Description
Permanently Open	Open habitats: grassland, water or areas where there are constraints to planting trees e.g. rocks, geology, roads.
Temporary open	Area that has been thinned, clear felled, coppiced in last 4 years.
Regeneration and Scrub	Woodland without an overstorey - tree seedlings (< 1m tall), saplings (trees > 1m tall and with girth of up to 7cm diameter at breast height (1.5m)) and shrubs.
Pole stage	Trees and shrubs fill the area and compete, ground flora is shaded out and no other plants colonise. Some canopy trees and understorey shrubs die due to competition. Trees and shrubs not yet bearing seed/fruit (immature). Trees have a diameter at breast height of > 7cm and < 20-30cm and are usually above 5m height.
Mature	Trees producing seed/berries. Crown/canopy usually spreading and at its maximum development. Canopy die-back (up to 10%) from competition for light and/or wind/snow damage.
Veteran ancient	Characterised by the presence of individual trees which have a large girth and show least three signs of old growth and decay. e.g. major trunk cavities/progressive hollowing, fungal fruiting bodies (e.g. from heart rotting species), high aesthetic interest (e.g. pollard or old coppice stool).

247 Sources of expert knowledge often documented which of the woodland types, and
 248 which of the stand structure types, a species was associated with. However, where the
 249 expert review did not provide this information, we used the canopy or understorey
 250 tree species, or the ground flora the species was associated with to guide its allocation

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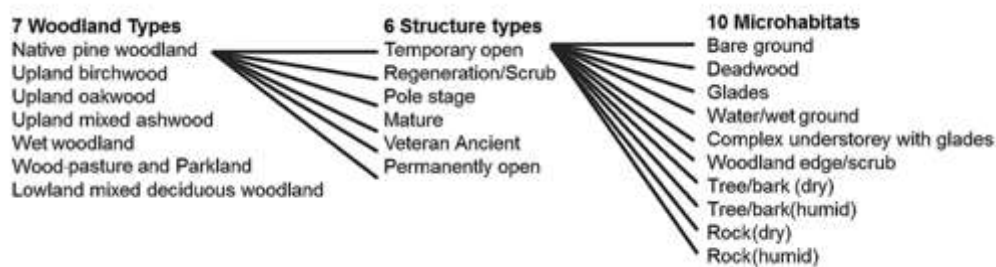
251 to the woodland type following the National Vegetation Classification (Rodwell,
252 1991). Where stand structure was not specified in the expert knowledge review for a
253 species, we used information on species' detailed resource and microclimate
254 preferences to inform the structure class within which a species was associated, such
255 as: the use of old growth tree features; the requirement for openness or shade; a
256 reliance on tree seeds; a preference for foliage density at different heights in the
257 canopy.

258 **iii. Microhabitat:** From the Stage 2 review describing species resource needs we
259 identified ten microhabitats (Figure 2) within each woodland type and structure class
260 that covered various fine-scale requirements of every protected species. These
261 microhabitat types nested within each structure type (Figure 2).

262

263 Having defined each unique woodland type-structure-microhabitat combination as a
264 niche, each species was associated with one or several of these to reflect the range of
265 woodland niches it is associated with according to the review evidence; these
266 associations formed a N4S matrix.

267



268

269 Figure 2: Hierarchical representation of the breakdown of a species resource requirement niche
270 to illustrate the Niches for Species system of habitat classification into niche components.

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272

273

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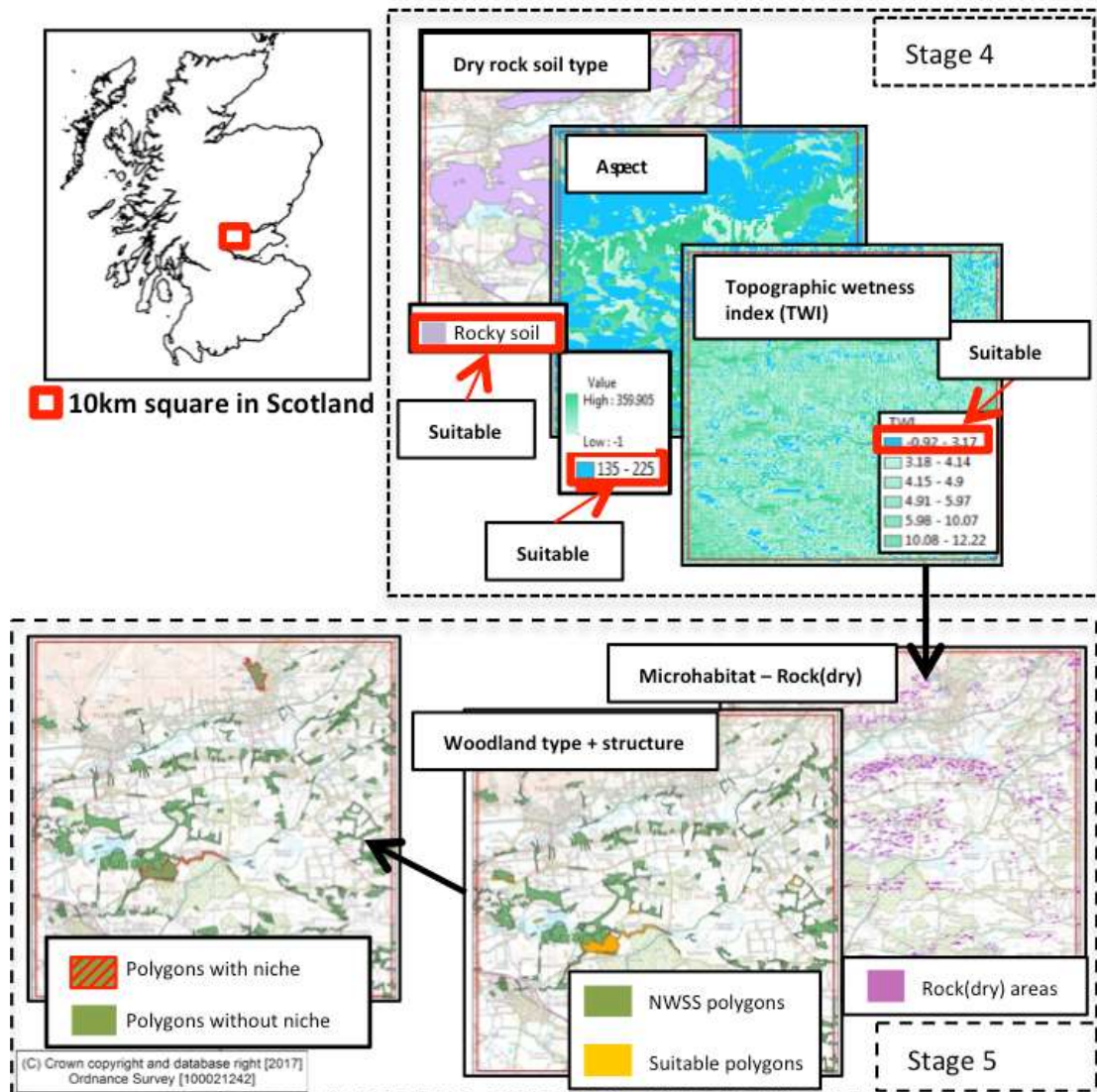
275 **2.2.3 Mapping woodland polygons and niche distributions**

276 To map woodland polygons, we used the Native Woodland Survey of Scotland
277 (NWSS) (Patterson et al., 2014). This spatial dataset provided information on native
278 woodlands across Scotland according to their type (Biodiversity Action Plan Priority
279 Woodland types: Maddock, 2008), structure and other features. The data were
280 gathered from all of Scotland's native woodlands during 2006-2013 by trained
281 surveyors according to a standard protocol (NWSS, 2013). Attributes are provided at
282 the scale of the woodland polygon, which is defined as a discrete area ≥ 0.5 ha and
283 having a minimum width of 20 m, and in which structural elements occupying a
284 minimum of 5% of the woodland area have been mapped. Therefore, a polygon can
285 be considered analogous to a stand, and there are approximately 95,800 NWSS
286 polygons mapped across Scotland, ranging in size from 0.5 ha to 800 ha with a mean
287 size of c.4 ha (Figure 3).

288 The NWSS data provided information that allowed us to classify most woodland
289 polygons into the two higher-level niche component categories, woodland type and
290 woodland structure (Table 3). To identify 'wood-pasture and parkland' woodland
291 type, which is not a NWSS woodland category, we overlaid Scotland's

292

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293

294 Figure 3: Graphical representation of the Niches for Species model development for
 295 Stage 4- deriving niche components (this example for microhabitat type rock (dry))
 296 from environmental spatial data, and Stage 5- categorising habitats and mapping
 297 niches by combining microhabitat presence with NWSS polygon information (in this
 298 example 'suitable' NWSS polygons are of habitat type upland oak woodland and
 299 subtype (structure) is mature).

300

301 Country Parks dataset (Scottish Natural Heritage) and updated the woodland type of
 302 any polygons with a centroid overlapping a park. The 'permanent open' or 'temporary
 303 open' woodland structures were identified as NWSS open habitat or clear fell
 304 polygons. These open polygons lacked woodland type information, so they were
 305 assigned the same woodland type as the adjacent woodland polygon with the shared
 306 longest border, calculated using a Geographic Information Software (GIS) (Esri,

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307 2013). We made this assumption in the absence of historical NWSS data that might
308 provide evidence of earlier woodland type.

309

310 To map the distribution of the 10 microhabitats, we reviewed the relevance of NWSS
311 data attributes alongside various other spatial environmental datasets (singly or in
312 combination) available for Scotland (Table 3). Data layers were extracted from non-
313 NWSS data by selecting polygons (vector data) or cells (raster layers) using a GIS,
314 that met specified attributes. For example, areas that were likely to have wet sites
315 were identified as those falling within 25 m of linear water features or wetland habitat
316 features identified from vector landcover maps, or as flat cells ($\leq 0.5^\circ$ slope) with high
317 topographic wetness index values (Sørensen et al., 2006) using a 25 m digital
318 elevation model (Table 3). Sources of all data layers used and whether vector or raster
319 are provided in Table 3.

320

321 **2.2.4 Mapping niche occurrence in polygons using spatial environmental data**

322 Once the NWSS woodland polygons had been classified by type and structure,
323 microhabitat presence-absence was predicted by overlaying the NWSS polygons with
324 the various microhabitat input data layers in a GIS. A rule-set was established for
325 mapping the presence of microhabitats that depended on particular combinations of
326 microhabitat input layers. The simplest microhabitat to map was *deadwood*, as NWSS
327 surveyors estimated deadwood volume on a single site visit per woodland polygon
328 during the seven year-long field survey (Table 3) (NWSS, 2013). The remaining nine
329 microhabitats were more complex to map, requiring more than a single data source
330 (for details of data sources used to map the microhabitats see Table 3). For example,
331 identifying the microhabitat *rock (dry)* used several spatial layers (Figure 3) combined

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332 using a logical rule-set to integrate information on land cover, soil and topographic
333 position (e.g. slope and aspect) of the polygon. The rule-sets were automated in
334 ArcGIS Model Builder (v10.2) (Esri, 2013). The ‘zonal statistics’ tool was used to
335 identify polygons overlapping input raster cells, and ‘select by location’ used to
336 identify polygons intersected by input vector layers. Any amount of overlap between
337 a NWSS polygon and a microhabitat input layer resulted in recording the microhabitat
338 ‘presence/absence’ in the polygon (although microhabitat ‘absence’ unused), and the
339 area or amount of microhabitat cover within a polygon was not considered.
340

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341 Table 3: Rule-set for combining spatial environmental data (type- vector=V, raster=R and sources of
 342 data shown in brackets) to describe potential niches present in the native woodlands of Scotland.
 343

Niche Component	GIS rule
Woodland type¹	
Upland mixed ashwood	Dominant NWSS woodland type for polygon (NWSS)
Upland birchwood	
Upland oakwood	
Lowland mixed deciduous	
Native pine	
Wet woodland	
Wood-pasture and parkland	Any NWSS polygon with centroids overlapping the Scotland's Country Parks dataset (NWSS; Scotland's Country Parks)
Structure type	
Permanently open	NWSS polygons recorded as 'open land' habitat type, which were ≥ 1 ha and shared an edge with a wooded NWSS polygon ² (NWSS)
Temporary open	NWSS polygons recorded as 'clear fell' dominant habitat type which were ≥ 1 ha and shared an edge with a wooded NWSS polygon ² (NWSS)
Regeneration or Scrub	'Native woodland' or 'Nearly-native woodland' NWSS polygons with dominant structure recorded as 'Visible regeneration', 'Established regeneration' or 'Shrub' or 'Scrub' (NWSS)
Pole	'Native woodland' or 'Nearly-native woodland' NWSS polygons with dominant structure recorded as 'Pole Immature' or 'Pole immature' (NWSS)
Mature	'Native woodland' or 'Nearly-native woodland' NWSS polygons with dominant structure recorded as 'Mature' (NWSS)
Veteran ancient	'Native woodland' or 'Nearly-native woodland' NWSS polygons with dominant structure recorded as 'Veteran' (NWSS)
Microhabitat	
Deadwood	NWSS polygons where deadwood was recorded by surveyor (NWSS)
Water/wet ground	NWSS polygons where (a) NVC ³ types associated with wet woodland habitats were recorded or, (b) they were intersected by: (i) inland water or wetland habitat polygons (OSMM or LCM inland water features) or, (ii) DEM cells with low slope ($\leq 0.5^\circ$) and within the top seven deciles of topographic wetness index values (NWSS; OSMM; LCM; DEM)
Woodland edge/scrub	NWSS polygons where (a) scrub was recorded by the surveyor (NWSS) or, (b) that have 'hard edges' i.e. aren't completely surrounded by other woodland polygons (NWSS; NFI)
Tree/bark (dry)	NWSS polygons with hard woodland edges (see woodland edge / scrub description) that overlap DEM cells with a southerly aspect ($135 - 225^\circ$) and are within the bottom decile of topographic wetness index values (NWSS; DEM)
Tree/bark (humid)	NWSS polygons that (a) overlap DEM cells with a northerly aspect ($>315^\circ$ or $\leq 45^\circ$) or, (b) overlap DEM cells with low slope ($\leq 0.5^\circ$) and are (c) within the top seven deciles of topographic wetness index values (TWI) or, (d) within 25 m of inland water or wetland habitats (OSOR or LCM inland water

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Niche Component	GIS rule
	features)(NWSS; DEM; TWI; OSOR; LCM)
Complex understorey with glades	NWSS polygons with 10 - 70% canopy cover and (a) regeneration (established or visible; $\geq 10\%$ cover) and shrub structures ($\geq 10\%$ cover) or, (b) ≥ 6 canopy structure types recorded by the surveyor (NWSS)
Glade	NWSS polygons with 10 - 70% canopy cover (NWSS)
Rock (dry)	NWSS polygons intersected by soil polygons with 'rocky' properties and DEM cells with a southerly aspect ($135 - 225^\circ$) and within the bottom decile of topographic wetness index values (NWSS; Scottish soils; DEM; TWI)
Rock (humid)	NWSS polygons intersected by rocky soil polygons and (a) overlap DEM cells with a northerly aspect ($>315^\circ$ or $\leq 45^\circ$) or, (b) overlap DEM cells with low slope ($\leq 0.5^\circ$) and are (c) within the top seven deciles of topographic wetness index values or, (d) within 25 m of inland water or wetland habitats (NWSS; Scottish soils; DEM; TWI; OSOR; LCM).
Bare ground	NWSS polygons? intersected by a footpath or forest track feature (footpaths)

344 Data sources: NWSS = Native Woodland Survey Scotland (V) (Patterson et al., 2014); Scotland's
 345 Country Parks = Scottish Natural Heritage (V); OSMM = Ordnance Survey Master Map (V)(Ordnance
 346 Survey, 2016); LCM = Centre for Ecology and Hydrology Land Cover Map 2007 vector map
 347 (V)(Morton et al., 2011); DEM= 25 m resolution digital elevation model (R)(EU-DEM, 2016); NFI =
 348 Forestry Commission's National Forest Inventory map (V)(Forestry Commission, 2011); TWI =
 349 topographic wetness index (R)(Sørensen et al., 2006; EU-DEM, 2016); OSOR = Ordnance Survey
 350 Open Rivers (V); Scottish soils = a 'mash-up' of two different scale maps at 1:10,000 and 1:250,000
 351 (V)(Lilly et al., 2010); Footpaths = Forestry Commission Scotland forest paths, tracks, rides, and
 352 boundaries (V)(FC Scotland, 2016).

353 ¹see Maddock (2008) for definitions

354 ²Assigned the woodland type of the wooded polygon (those classified as Native woodland' or 'Nearly-native
 355 woodland') with which they shared the longest border length with.

356 ³ National Vegetation Classification (NVC) see Rodwell (1991).

357

358

359 2.2.5 Mapping species habitat suitability

360 Using Model Builder and Python scripts in ArcGIS, we implemented a rule-set to link
 361 the NWSS niche map with the N4S matrix. A NWSS polygon was predicted to be
 362 suitable when the combination of woodland type-woodland structure and microhabitat
 363 presence matched a species' habitat requirements. Binary fields were added to the
 364 spatial database to indicate a polygon's predicted suitability (0 or 1) for each species.

365

366 2.2.6 Mapping species potential distribution

367 As many of the protected species have restricted ranges across Scotland, we limited
 368 predicted species occurrence by classifying any NWSS polygons outside of modelled
 369 current bioclimatic envelopes as unsuitable. Bioclimatic envelopes were available for

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370 51 species (23 species of invertebrate, 17 lower plants, 1 vascular plant (Ellis et al.,
371 2014; Pearce-Higgins et al., 2015)) (Table S3, Appendix A – online supplementary
372 material). In the absence of these data we mapped population ranges from 10 km
373 resolution NBN Gateway species records (NBN, 2017) for all survey years using the
374 Minimum Convex Polygons (MCPs) (Rurik and Macdonald, 2003) in ArcGIS. MCPs
375 were generated for 90 species representing all taxa. For 38 species there was
376 insufficient data (fewer than three 10 km squares adjacent to one another) (Table S3
377 Appendix A – online supplementary material).

378

379 **2.3 Validation of model**

380 **2.3.1 Validation species occurrence data**

381 We selected ten species to use in a validation exercise. The validation compared the
382 potential distribution predicted by N4S with existing species occurrence records. The
383 validation species were selected to represent a range of woodland types, taxonomic
384 groups, and traits (wide to narrow niche breadth; vagile to sessile; easy to observe to
385 cryptic). We used only data recorded at a 100 m resolution or finer (≥ 6 figure grid
386 references) to ensure we could accurately attribute records to polygons (Dymytrova et
387 al., 2016). Records were used from a sixteen-year period (2000 to 2016), in line with
388 the NWSS data (surveyed 2006 – 2013). To gain insights into how well the N4S
389 model predicted areas without the potential to support protected species, we
390 incorporated pseudo-absence records into the analysis as adequate absence records
391 were not available. Pseudo-absence records were created following the “surveyed
392 absence” or “target group” strategy which uses location records of species from the
393 same taxonomic group, where it is assumed that the focal species was not recorded as
394 it was absent (Gomez-Rodriguez et al., 2012; Hanberry et al., 2012; Phillips et al.,

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395 2009). The choice of only 10 validation species was largely influenced by the
396 availability of species records for which we could obtain some pseudoabsence data.

397

398 Ultimately, choice of validation species was constrained by data availability. For two
399 bird species - *Muscicapa striata*, *Turdus philomelos* - data at the required resolution
400 were available only from surveys of one woodland type (native pine woodland)
401 limiting testing of model predictions to between woodland type and structure with and
402 without microhabitat. N4S model predictions were fully tested for the remaining eight
403 validation species: three lower plants- *Collema fasciculare*, *Pseudocyphellaria*
404 *norvegica*, *Gomphillus calyciodes*; one vascular plant- *Linnaea borealis*; and four
405 invertebrates- *Cupido minimus*, *Carterocephalus palaemon*, *Boloria euphrosyne*,
406 *Osmia uncinata*.

407

408 **2.3.2 Validation data analysis**

409 Duplicate species records (same date and location) were removed. The proportion of
410 field records falling within polygons predicted to be suitable or unsuitable for each of
411 the validation species were calculated for presence and pseudo-absence records. We
412 applied a cumulative binomial probability test (R Core Team 2012) to estimate
413 whether the number of presence records lying within suitable polygons of the N4S
414 model was greater than could have been predicted by chance alone, according to the
415 area of suitable woodland habitat available within the species' range.

416

417 We also tested the degree of agreement between the N4S model predictions and the
418 information from the presence/pseudo-absence datasets by constructing confusion
419 matrices using SAS version 9.3 (SAS, 2011) (Table S1, Appendix B - online

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420 supplementary material) and generating Cohen's Kappa statistic (k), where $k=1$
421 indicates perfect agreement, $k=0$ agreement by chance alone and $k<0$ disagreement
422 (Cunningham, 2009). A system of subdivision of k has been suggested, for which we
423 tested the six categories: "No agreement" ($k<0$); "Slight agreement" ($k\geq 0$ and <0.2);
424 "Fair agreement" ($k\geq 0.2$ and <0.4); "Moderate agreement" ($k\geq 0.4$ and <0.6);
425 "Substantial agreement" ($k\geq 0.6$ and <0.8); "Almost perfect agreement" ($k\geq 0.8$ and
426 <1.0) (Landis and Koch, 1977). The deviation of k values from zero was tested
427 statistically ($H_0: k = 0$; one-sided probability reported as testing agreement i.e. $k>0$).
428 All tests were performed for each species and at three levels of the habitat
429 classification hierarchy i.e. where occurrence of the target species was predicted from
430 the presence of 1) suitable woodland type only, 2) woodland type + structure type or
431 3) woodland type + structure type + microhabitat type.

432

433 **2.4 Choice of Niches for Species model outputs**

434 The N4S model output (map of protected woodland species potential occurrence
435 based on the availability of niches) can be viewed at a variety of scales. We selected
436 three scales considered appropriate for different policy or practice queries: 1) a
437 national-scale overview of species richness which may be applicable to supporting
438 strategic forest policy decisions, 2) a landscape-scale assessment of species richness
439 which may support tactical decision making in forest planning, and 3) an individual
440 species map with associated habitat data which we envisaged might be used in
441 practice for operational decisions guiding management interventions.

442

443

444

445 **3. Results**

446

447 **3.1 Spatial environmental data used to map niche occurrence in polygons**

448 Most of the niche components were derived directly from the NWSS data (Table 3).

449 For the remainder, information was derived from other available spatial datasets and

450 their reliability was limited by their relevance, accuracy and precision (Table 3; Table

451 S4, Appendix A - online supplementary material). For example, there were no fine

452 resolution spatial data available to describe the microhabitat *bare ground*. Therefore,

453 we assumed this microhabitat would be found along footpaths and tracks, and used

454 spatial data on these features to map the likely occurrence of this microhabitat.

455

456 **3.2 Validation**

457 The strength of the agreement (i.e. higher Kappa value) varied among species (Table

458 4). There was some agreement ($Kappa > 0$) between model predictions and the

459 occurrence for nine of the ten validation species (No agreement found for *T.*

460 *philomelos*), but this was 'Slight' for seven of the remaining nine species' (Landis and

461 Koch 1977). Higher Kappa values ($Kappa = 0.296$ - 'Fair agreement' to $Kappa =$

462 0.807 - 'Perfect agreement') occurred for the species *O. uncinata* and for *L. borealis*.

463 Results from the probability tests (Kappa and binomial) were largely consistent. For

464 five of the ten validation species associations between distribution records and

465 predicted availability of suitable polygons was better than would be expected if

466 species occurrence had been allocated at random to the woodland polygons. For two

467 species the results approached statistical significance, for one level of the habitat

468 classification hierarchy (e.g. woodland type + structure) tested. Judged on the

469 frequency of agreement (between actual and predicted occurrence of species when the

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470 N4S model was run at the three levels of niche hierarchy complexity) the N4S model
471 appeared to perform equally well at the intermediate (woodland type + structure) and
472 most detailed (woodland type + structure + microhabitat) hierarchy levels (Table 4).
473 However, the agreements with the highest levels of significance ($p < 0.05$) for the
474 binomial test and Kappa value occurred when the model included microhabitat (Table
475 4). This suggests that where agreements are found these are stronger when niche
476 identification included microhabitat features.
477

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478 Table 4 Summary of correspondence between the habitat availability for ten validation species predicted using Niches for Species (N4S) model and records of species
 479 occurrence and pseudo-absence at three levels of niche hierarchy (1 = woodland type only; 2= woodland type + stand structure; 3= woodland type + stand structure +
 480 microhabitat). Kappa (k) subdivisions: "No agreement" ($k < 0$); "Slight agreement" ($k \geq 0$ and < 0.2); "Fair agreement" ($k \geq 0.2$ and < 0.4); "Moderate agreement" ($k \geq 0.4$ and < 0.6);
 481 "Substantial agreement" ($k \geq 0.6$ and < 0.8); "Almost perfect agreement" ($k \geq 0.8$ and < 1.0) (Landis and Koch, 1977). One-sided probability reported as testing for where k is
 482 positive; $H_0: k = 0$. "Binomial" refers to a binomial probability test; H_0 : the number of validation species records found within suitable woodland polygons is no better than
 483 random within the sampled woodland polygons. Sampled polygons being those containing a pseudo-absence record a validation species record or both. Probability test level
 484 of significance (for both Kappa and binomial tests): * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$, na = not applicable for one-sided test, p value reported where non-significant.

Validation species	Niche hierarchy					
	1		2		3	
	Kappa value (p =)	Binomial	Kappa value (p =)	Binomial	Kappa value (p =)	Binomial
<i>Collema fasciculare</i>	Slight agreement 0.105 (p=0.067)	p=0.098	Slight agreement 0.095 (p=0.103)	p=0.147	Slight agreement 0.022 (p=0.386)	p=0.483
<i>Pseudocyphellaria norvegica</i>	No agreement -0.107 (na)	p>0.999	Slight agreement 0.005 (p=0.455)	p=0.444	Slight agreement 0.014 (p=0.358)	p=0.253
<i>Gomphillus calyciodes</i>	Slight agreement 0.008 (p=0.419)	p=0.518	Slight agreement 0.126(p=0.053)	p=0.078	Slight agreement 0.108(p=0.081)	p=0.118
<i>Linnaea borealis</i>	Almost perfect agreement 0.807 (***)	**	Slight agreement 0.128 (***)	**	Slight agreement 0.065 (***)	***
<i>Cupido minimus</i>	No agreement -0.0013 (na)	p=0.863	Slight agreement 0.042 (***)	**	Slight agreement 0.045 (***)	***
<i>Carterocephalus palaemon</i>	No agreement -0.018 (na)	p>0.999	Slight agreement 0.022 (p=0.075)	p =0.056	Slight agreement 0.004 (p=0.381)	p=0.262
<i>Boloria euphrosyne</i>	Slight agreement 0.013 (***)	**	No agreement -0.025 (na)	P=0.999	No agreement -0.016 (na)	P=0.999
<i>Osmia ucinata</i>	Slight agreement 0.006 (p=0.139)	p=0.231	Fair agreement 0.296 (***)	***	Fair agreement 0.223(***)	***
<i>Muscicapa striata</i>	NA	NA	Slight agreement 0.041 (*)	p=0.076	No agreement -0.016 (na)	p=0. 641
<i>Turdus philomelos</i>	NA	NA	No agreement -0.024 (na)	P=0.999	Insufficient values	p=0.999

485
486

487 **3.3 Example N4S model outputs**

488

489 **3.3.1 National species richness map**

490 The national scale map (Figure 4) highlights the extent of native woodlands covered by
491 the NWSS dataset (included in the N4S model), and shows the potential occurrence of
492 protected woodland species within these woodlands. Woodlands with high species
493 richness (>20 to 30 protected woodland species per woodland polygon) are reasonably
494 well spread throughout Scotland although the native woodlands of the River Dee valley
495 and the River Spey valley in north-eastern Scotland stand out as being areas of
496 particularly high species richness.

497

498 **3.3.2 Landscape scale species richness output**

499 The 10 km x 10 km area of upland Scotland selected to illustrate the N4S model
500 landscape scale output (Figure 5) depicts a highly wooded landscape area, where nearly
501 half of the area (4,377 ha) comprises native woodlands. A few polygons have the niche
502 potential for a high number of protected woodland species (up to 31) and most have the
503 potential to support ten or more species. However, several polygons have low species
504 richness (0 to 10 protected species per polygon).

505

506 **3.3.3 Individual species-niche output**

507 More detailed information can be extracted from the N4S model (Figure 6). For example,
508 the lower plant *Dumortiera hirsuta* was one of the species predicted to occur in the
509 sample landscape we have used to illustrate the finest scale output from the N4S model.
510 The model output gives the locations of the polygons *D. hirsuta* is predicted to occur
511 within (Figure 6). These comprise woodland types *upland oakwood* and *upland mixed*

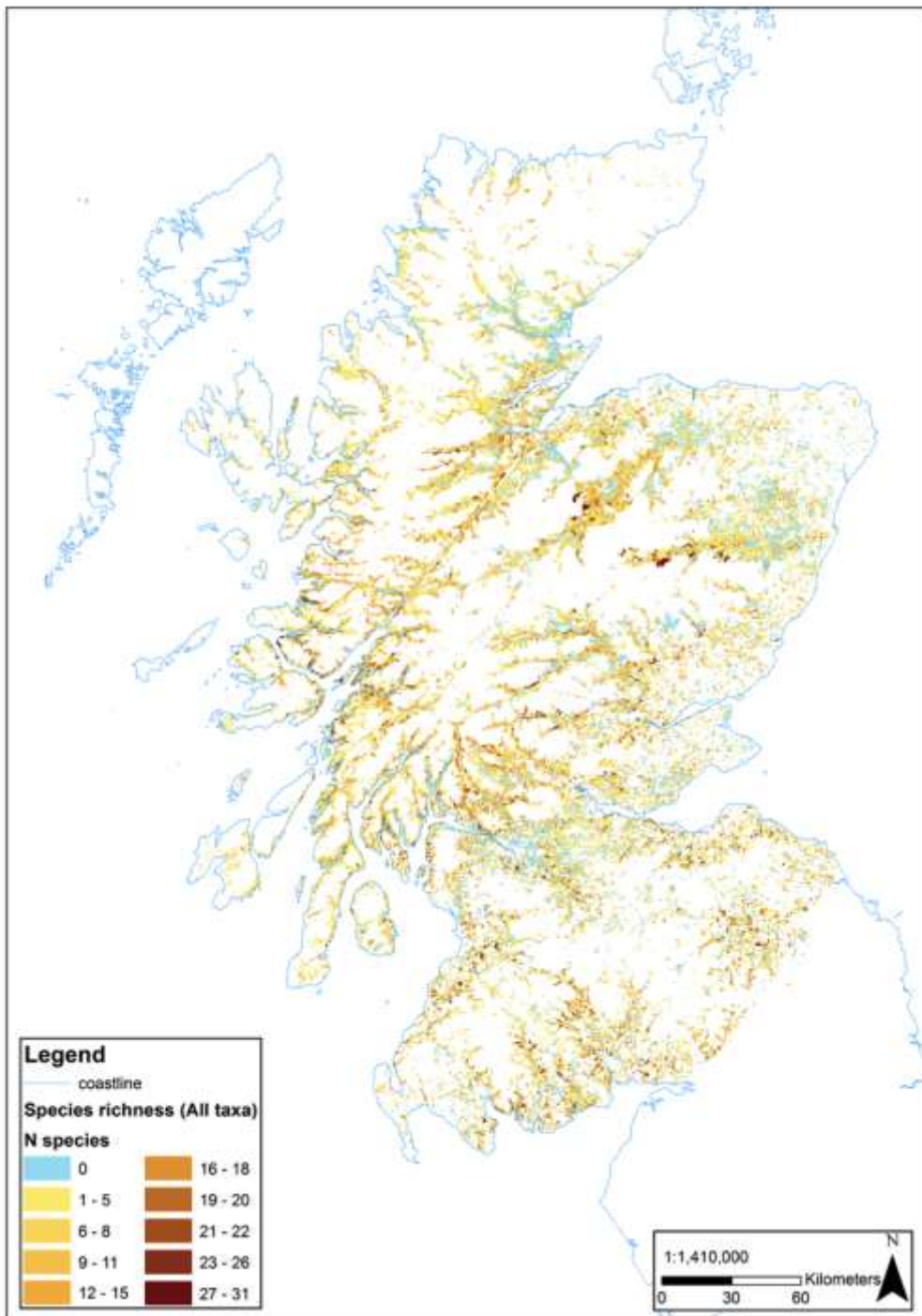
Niches for Species

512 *ashwood*, all with a *mature* stand structural stage. *Dumortiera hirsuta* is most likely to be

513 associated with the *water/wet ground* and *rock(humid)* and *bare ground* microhabitats

514 where available within these polygons.

515

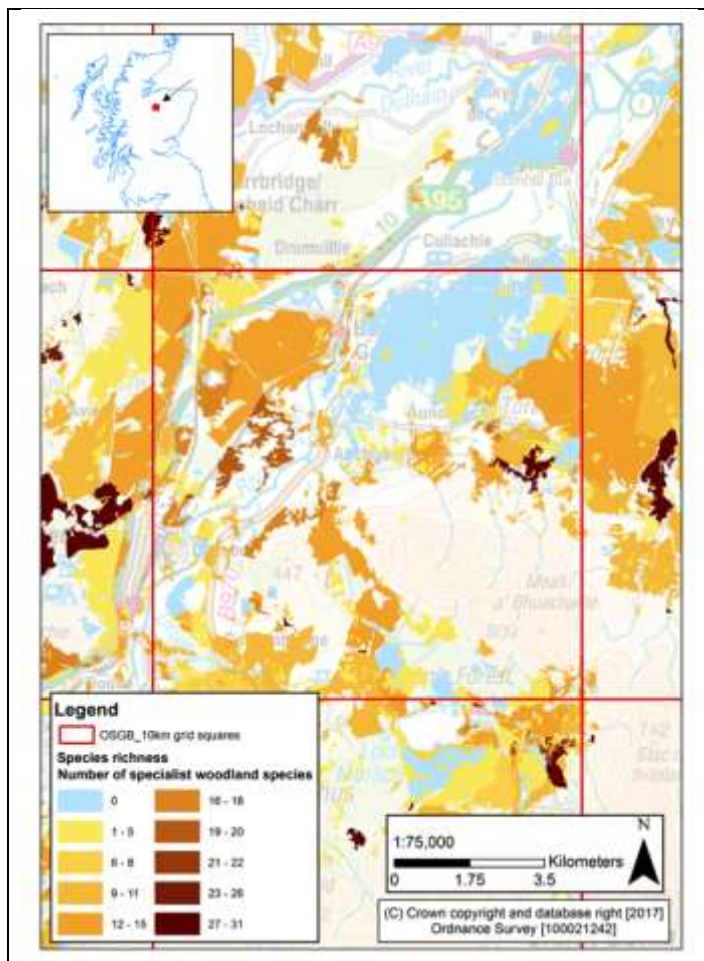


516

517 Figure 4: Species richness of native woodlands in Scotland based on the predicted potential distribution of
518 all 179 protected woodland species.

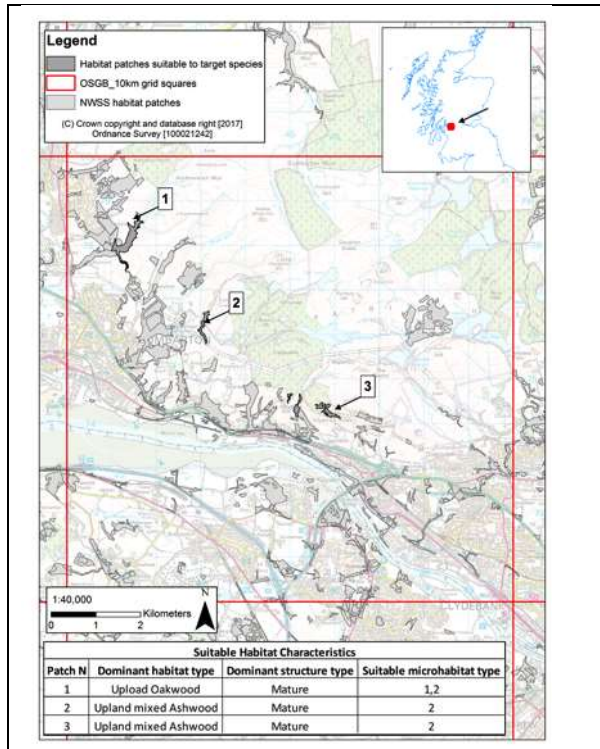
Niches for Species

519
520
521
522



523 Figure 5: Sample output from the Niches for Species model showing predicted distribution of protected
524 woodland species richness by native woodland polygon in a 10 km x 10 km (Ordnance Survey Great
525 Britain) area of a typical upland landscape in Scotland.
526

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527 Figure 6: Sample output from the Niches for Species model showing predicted location of *Dumortiera*
 528 *hirsuta*, a protected woodland species in native woodland polygons in a 10 km x 10 km area (Ordnance
 529 Survey Great Britain). Niche information associated with *D. hirsuta* is included; a niche is defined by the
 530 combination of *woodland type*, *structure type* and *microhabitat* (1 = *water/wet ground*, 2 = *rock(humid)*).
 531 (NWSS- Native Woodland Survey Scotland).

532

533 4. Discussion

534 We provide a framework to link expert species knowledge with spatial environmental

535 datasets to predict simultaneously, for multiple taxa, the availability of habitat for

536 protected species. In applying our N4S model to protected woodland species in

537 Scotland, we found that the type and accessibility of expert knowledge on habitat

538 requirements varied between taxa, but that there was sufficient information to include

539 179 of the 208 species. Relevant spatial environmental data were also available to

540 classify native woodlands into type and structure, and to map the distribution of most

541 microhabitats with confidence. Species records did not consistently accord with the

542 predictions of species occurrence by the model: good agreement was shown for five out

543 of ten of the validation species, based on the niche hierarchy giving best results. By

544 mapping protected species potential occurrence, the quality of habitat for

545 supporting biodiversity can be visualized in a simple form by spatial outputs of
546 protected species richness by woodland polygons; interpreted from the same input
547 data either at the whole administrative region, landscape or forest level.

548

549 **4.1 Adequacy of data and model strengths**

550 The relatively simple species-habitat association evidence in the N4S model has been
551 drawn from information provided by species experts, and, although of good quality,
552 much of the information was not published and therefore needed to be sought directly
553 from the experts. Based on the percentage of data field entries for different taxa
554 supported by each of the four evidence types, it appears information for cryptic species is
555 less accessible (mostly via expert knowledge- evidence type 1 and peer-reviewed
556 journals- evidence type 2) than for the better-known species, as expected. The literature
557 on biodiversity indicators suggests there is a sound basis to making links between habitat
558 features and species occurrence (Regnery et al., 2013; Gao et al., 2015) and the inclusion
559 of fine scale habitat features (e.g. structure type and microhabitats in the N4S model) can
560 be important for certain species (Harvey and Platenberg, 2009; Dymytrova et al., 2016;
561 Horak, 2017).

562

563 We have high confidence in the quality of the spatial data as 65% of the 23 different
564 habitat categories and microhabitats used in the N4S model (7 woodland types, 6
565 structure types and 10 microhabitat types) were derived directly from existing attributes
566 in the input datasets. A third of these attributes relied on information derived from
567 various other spatial data. However, we had low confidence in predicting just one
568 attribute - the *bare ground* microhabitat. Although beyond the scope of this study, we
569 recommend validating the N4S model using a targeted survey of polygons in which an

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570 assessment of the predicted niche occurrence has been verified. This would increase our
571 confidence in how well spatial data combine to describe features on the ground. We have
572 relied on the detailed woodland survey NWSS data to locate many of the niches and such
573 data may not be universally available. Nevertheless, the approach illustrated, of
574 classifying habitat niches and describing these using spatial data would allow the use of
575 alternative or replacement spatial datasets. We recommend sourcing and integrating
576 alternative spatial data to ensure that habitat layers remain current. For example, we
577 could integrate a forest structure layer interpreted from aerial photography or LiDAR data
578 (where this is available) to update the woodland structure information within the
579 polygons (McInerney et al., 2011).

580

581 The N4S model does not take account of interactions among species and assumes that if
582 the correct habitat is available there will be no constraints on potential species use. This
583 assumption, like SDMs in general, may lead to over prediction of species occurrence
584 (Phillips et al., 2006). Although N4S does not account for the landscape surrounding a
585 woodland patch, broader scale influences that affect species distribution are factored in to
586 the N4S model by constraining predictions by bio-climatic or distribution envelopes.
587 Inclusion of envelopes has been shown to improve model performance in SDMs based on
588 species records (Lobo et al., 2011; Zarnetske et al., 2007), primarily because it enables
589 some environmental data to be incorporated.

590

591 **4.2 Model validation**

592 Consistent with our expectations, validation showed limited correspondence between
593 predicted potential species locations (woodland polygons) and recent species presence
594 records (agreements were significant for half of the validation species). Including detailed

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595 information about species' resource requirement (microhabitat) in our expert-based
596 habitat suitability model did appear to improve the model performance in the validation
597 tests for the subset of species where agreement was found between predicted and
598 recorded species occurrence. It is possible that this is due to weak relationships between
599 some but not all taxa represented by the validation species and microhabitats (Goa et al.,
600 2015; Regnery et al., 2013) and could also result from poor spatial definition of
601 microhabitats from the data sets we have used. However, we anticipated that poor model
602 performance could also result from the lack of availability of high-resolution species
603 presence records available for validation. Although the resource of species records for
604 Britain is large, surveys are not always carried out systematically (instead favoured
605 locations are targeted for survey), it is uncommon for all areas to be surveyed regularly,
606 and only rarely is species absence data collected (NBN, 2017). In studies when data
607 meeting these survey criteria are deployed, good agreement has been found between the
608 empirical data and the expert-based classifications of habitat choice (Leblond et al., 2014;
609 Reif et al., 2010). The lack of availability of good quality species records has been argued
610 (e.g. Phillips et al., 2006) as a reason to develop predictive models of distribution based
611 on knowledge, as in N4S, rather than records.

612

613 **4.3 Application**

614 Niches for Species is now being applied in several ways with model uncertainty
615 described according to the scale of application. For forest policy makers, the model
616 provides an analysis of the whole of the native woodland resource in Scotland (both
617 within and outside protected areas) and indicates where there are species 'hotspots' or
618 habitats where particular sets of rare or threatened species are likely to occur. As N4S
619 considers all the protected species of interest for the region, it performs as well, or better

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620 than the current alternative national analysis method conducted for the UK using coarser
621 (2km resolution) data, and the better recorded species e.g. birds as surrogates (Franco et
622 al., 2009). Furthermore, the N4S model has the advantage of providing information on
623 the habitats associated with areas that may be prioritised due to potential protected
624 species occurrences: Franco et al. (2009) recognised that the lack of such information was
625 a shortcoming in their SDM. For forestry decision making, visualising the configuration
626 of potential protected species occurrence at the landscape-scale can help planners
627 consider how to minimise forest operations impacts on species rich areas (Forestry
628 Commission 2017; UKWAS, 2008). When used in a scenario analysis, N4S can provide
629 planners with estimates of how potential species lists and overall species richness may
630 vary with choice of woodland type and location, as a result of decisions to meet
631 woodland expansion targets (Sing et al., 2013). At this scale of application, uncertainties
632 regarding the accuracy of the model should be checked by applying local experience to
633 compare habitat types, and likely diversity of niches with the location of species rich
634 areas indicated by the model. At a finer scale, knowledge of potential occurrence of a
635 protected species within a woodland polygon may alert the need for an expert survey to
636 confirm species presence. This is consistent with the recommended application of many
637 SDMs (Buechling and Tobalske, 2011; Dymytrova et al., 2016; Lentini et al., 2015).

638

639 Forestry practitioners and policy makers are tasked with applying management in ways
640 that will meet international and national obligations for conserving biodiversity in the
641 most efficient manner (CBD, 2010; Forestry Commission, 2017). Obligations are
642 articulated through law and policy devolved to a country/regional level. In all cases
643 information is needed on where the most threatened species occur within the landscape,
644 and how species presence may change in response to habitat management at a variety of

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645 scales (Barrows et al., 2005; Egoh et al., 2014;). Our challenge was to produce a model
646 which encompassed all the protected species Scottish forestry decision makers are legally
647 obliged to consider. Our approach incorporates the available wealth of ecological
648 knowledge on species using high resolution spatial data, and avoids the need to rely
649 solely on species records or surrogate species. The N4S model provides forest decision
650 makers with information on the occurrence of niches for nearly all the protected species
651 associated with woodlands in Scotland. For many species, actual locations may not be
652 known due to their rarity and/or their cryptic nature; and additionally, there may be
653 uncertainty about habitat features to which their location is related. The output map
654 format with associated attribute table listing the woodland type, structure and predicted
655 presence and absence of each microhabitat and protected species, helps to improve
656 knowledge of species needs and location of potential niches.

657

658 Niches for Species can help forest practitioners guide conservation management, but we
659 acknowledge some weaknesses, which may limit its application, and suggest
660 improvements. The model may lack high levels of accuracy that would otherwise be
661 valuable to forest policy makers and practitioners. However, high levels of accuracy are
662 not always needed by decision makers, and more timely action may ultimately be more
663 cost effective than delayed action (Cook et al., 2013). This is particularly so at a time of
664 austerity and a decline in priority afforded to biodiversity policy. We recommend this
665 expert-based EHSM approach as a method to integrate complex information relating to
666 multiple and often data-deficient species in a format which allows land policy makers and
667 managers to consider equally, large numbers of species and their conservation needs.

668

669

670

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681

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