

1 Quantifying the recent expansion of native invasive rush species in 2 a UK upland environment

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9 10 11 12 **Abstract**

13 Rushes, such as soft rush (*Juncus effusus* L.), hard rush (*Juncus inflexus* L.) and compact rush
14 (*Juncus conglomeratus* L.) have become problem species within upland grasslands across the
15 UK and the coastal grasslands of western Norway. Indeed, being largely unpalatable to
16 livestock and having a vigorous reproductive ecology means that they can rapidly come to
17 dominate swards. However, rush dominance results in a reduction in grassland biodiversity and
18 farm productivity. Anecdotal evidence from the UK suggests that rush cover within marginal
19 upland grasslands has increased considerably in recent decades. Yet, there is currently no
20 published evidence to support this observation. Here, we use recent and historical Google Earth
21 imagery to measure changes in rush frequency over a 13-year period within four survey years:
22 2005, 2009, 2015 and 2018. During each survey year, we quantified rush presence or absence
23 using a series of quadrats located within 300 upland grassland plots in the West Pennine Moors,
24 UK. Data were analysed in two stages, first, by calculating mean rush frequencies per sample
25 year using all the available plot-year combinations (the full dataset), and second by examining
26 differences in rush frequency using only the plots for which rush frequency data were available
27 in every sample year (the continuous dataset). The full dataset indicated that rush frequency
28 has increased by 82% between 2005 and 2018. Similarly, the continuous dataset suggested that
29 rush frequency has increased by 174% over the same period, with the increases in frequency
30 being statistically significant ($P < 0.05$) between 2005-2018 and 2009-2018. We discuss the
31 potential drivers of rush expansion in the West Pennine Moors, the ecological and agronomic
32 implications of grassland rush infestations, and priorities for future research.

34 **KEYWORDS**

35 Google Imagery, Habitat change, Hill farming, *Juncus* spp., Native invasive species, Upland
36 breeding birds, Upland habitats, Upland land management

37

38 **1. INTRODUCTION**

39 Soft rush (*Juncus effusus* L.), hard rush (*Juncus inflexus* L.) and compact rush (*Juncus*
40 *conglomeratus* L.) (henceforth known as “rushes” in this research paper) are native to the
41 British Isles and occur throughout its many habitats (Preston *et al.*, 2002). Rushes are generally
42 tussock-forming, slowly spreading perennials that have a preference for wet, acidic and
43 nutrient-poor environments (Richards and Clapham, 1941b, c, d; Hill *et al.*, 2004).
44 Nevertheless, they can establish and proliferate under a broad range of environmental
45 conditions (Richards and Clapham, 1941b, c, d; Hill *et al.*, 2004). However, the complete range
46 of conditions under which rushes can survive (i.e. their fundamental niche) remains largely
47 unknown (see, for example, Hamilton *et al.*, 2018).

48 In contrast, we do know about the reproductive ecology of rushes. For example, they
49 can produce between 4500 and 8500 seeds per stem per year (McCarthy, 1971; Kaczmarek-
50 Derda *et al.*, 2014), which, on rush infested ground, equates to approximately 4 to 6.7 million
51 seeds per square metre per season (Moore and Burr, 1948; Ervin and Wetzel, 2001). To produce
52 such large amounts of seed, a single rush plant only uses 0.27% of its annual net biomass
53 production (Ervin and Wetzel, 2001). Depending on species, seeds ripen between July and
54 September and are shed (mainly by the wind during dry conditions) up to the following spring
55 (Richards and Clapham, 1941a, b, c). After shedding, seeds can remain dormant at the soil
56 surface for up to 60 years (Moore and Burr, 1948), and, during this time, they may be dispersed
57 by wind or surface run-off and/or germinate in areas disturbed by cultivation or livestock
58 poaching (Agnew, 1961; McCarthy, 1971; Cairns, 2013). Once established, rushes persist for
59 a long time and usually expand clonally via a shallow system of short rhizomes (Kaczmarek-
60 Derda *et al.*, 2019), which ultimately leads to the formation of dense stands covering entire
61 fields.

62 The vigorous reproductive ecology of rushes may be a contributing factor behind their
63 recent invasion of upland grasslands across the UK and the coastal grasslands of western
64 Norway (Cherrill, 1995; Østrem *et al.*, 2018). Indeed, there is anecdotal evidence from farmers
65 and ecologists in the UK of rush infestations within upland grasslands (Hamilton *et al.*, 2018).
66 Such infestations are problematic because they significantly reduce the agricultural and

67 conservation value of the land (Cairns, 2013; Coyle *et al.*, 2018). However, while there have
68 been several static assessments of grassland rush infestation in the UK (e.g. Hopkins *et al.*,
69 1985; Cherrill, 1995), there are currently no peer-reviewed studies that have attempted to
70 measure changes in grassland rush expansion over time (but, for examples within the grey
71 literature, see: O'Reilly, 2011; Hamilton *et al.*, 2018). The present study aims to address this
72 research gap by providing a direct quantitative assessment of changes in grassland rush
73 frequency between 2005 and 2018 within a large upland area: The West Pennine Moors Site
74 of Special Scientific Interest (SSSI). In addition to presenting our results, we discuss the
75 potential drivers of rush expansion in the West Pennine Moors, the agronomic and ecological
76 implications of grassland rush infestations, and future research priorities.

77

78 **2. MATERIALS AND METHODS**

79 **2.1. Site description and justification**

80 The West Pennine Moors (WPM) Site of Special Scientific Interest (SSSI) is situated in the
81 North West of England (Fig. 1). The site covers an area of approximately 76 km² and an
82 elevation range of 100 to 450 m. It was designated as a SSSI in 2016 due to its extensive mosaic
83 of upland and upland-fringe habitats, which support significant populations of breeding birds,
84 including waders such as curlew (*Numenius arquata* L.), snipe (*Gallinago gallinago* L.) and
85 lapwing (*Vanellus vanellus* L.) (Natural England, 2016). The Centre for Ecology & Hydrology
86 (CEH) Land Cover Map (LCM) data from 2015 (Rowland *et al.*, 2017) indicates that the
87 dominant upland habitats within the SSSI are blanket bog, acid grassland and heather
88 moorland; however, there are also substantial areas of improved grassland and broadleaved
89 woodland (Fig. 1).

90 We chose to measure rush expansion within the WPM SSSI for two reasons. First, the
91 SSSI contains large areas of marginal grassland, i.e., semi-improved and enclosed permanent
92 pasture at or below the moorland line (above this line the land is generally unimproved and
93 unenclosed). These grasslands are vital to hill farmers because they tend to be the most
94 productive areas of their farm (Mansfield, 2008; Nielsen and Sjøgaard, 2000). Also, by
95 providing suitable nesting habitat, marginal grasslands can support large populations of wading
96 bird species (Baines, 1988; Dallimer *et al.*, 2010; Dallimer *et al.*, 2012). Crucially, the value
97 of marginal grasslands to both farmers and birds decreases as rush cover increases: rushes are
98 generally less palatable and digestible to livestock than other grassland species (Grant *et al.*,
99 1984; Nielsen and Sjøgaard, 2000; Tweel and Bohlen, 2008), so increases in rush cover reduce

100 grassland productivity and milk/meat production (Cairns, 2013); likewise, for wading birds,
101 grasslands where rush cover exceeds 30% become suboptimal nesting habitat (RSPB, 2017).
102 The second reason for choosing the WPM SSSI is that there are anecdotal reports from Natural
103 England advisors and farmers of substantial increases in grassland rush cover over the past 20
104 years (K. Rogers, pers. comm., April 15, 2019).

105

106 **2.2. Detecting rush (*Juncus* spp.) using Google Earth imagery**

107 Rush tussocks are visible on colour aerial imagery, but only within habitats where the
108 surrounding vegetation is much shorter and of a different colour or tone. The marginal
109 grasslands within the WPM SSSI meet these criteria. For example, Figures 2 and 3 demonstrate
110 that, compared to other upland habitats, there is a considerable height and colour differential
111 between rush tussocks and the surrounding vegetation (mainly Poaceae spp.) within these
112 grasslands, and these differences mean that rush tussocks are clearly visible on the
113 corresponding aerial imagery. Thus, rush frequency within marginal grasslands can be
114 quantified using aerial imagery and, if historical aerial imagery is available, one can measure
115 changes in rush frequency over time. Google Earth (Google Inc) provides historical aerial
116 imagery of the WPM SSSI for 2005, 2009, 2015 and 2018. However, images from 2009 and
117 2015 only provide partial coverage of the SSSI. Using the available Google Earth imagery data,
118 we aimed to quantify changes in rush frequency within the marginal grasslands of the WPM
119 SSSI during four time periods: 2005, 2009, 2015 and 2018.

120 We decided to use aerial imagery instead of field surveys because there is a lack of
121 historical field data on rush expansion within the marginal grasslands of the WPM SSSI.
122 Furthermore, while field surveys are likely to be more accurate, rush expansion can be
123 measured more efficiently using aerial imagery, which means that larger areas of grassland can
124 be surveyed. Furthermore, the use of aerial imagery is much more convenient for sampling
125 more remote or inaccessible areas and you do not require prior permission from landowners.

126

127 **2.3. GIS selection of marginal grassland parcels**

128 We used CEH LCM 2015 vector data (Rowland *et al.*, 2017) to select marginal grassland
129 parcels that lay within or intersected the WPM SSSI boundary. Since the CEH LCM 2015 does
130 not have a ‘Marginal grassland’ land cover category (Rowland *et al.*, 2017) we adopted the
131 ‘Improved grassland’ land cover category as a surrogate because Google Earth aerial imagery
132 revealed this to be the best proxy for marginal grassland within the WPM SSSI. According to
133 the CEH LCM 2015, ‘Improved grassland’ is “characterised by vegetation dominated by a few

134 fast-growing grasses such as *Lolium* spp., and also white clover (*Trifolium repens*), on fertile,
135 neutral soils. Improved Grasslands are typically either managed as pasture or mown regularly
136 for silage production” (NERC, 2017).

137 In total, 340 improved grassland parcels lay within or intersected the WPM SSSI
138 boundary. However, 40 grassland parcels were excluded from our survey because Google Earth
139 imagery revealed that non-grassland habitats constituted $\geq 25\%$ of their extent. We used the
140 remaining 300 grassland parcels as discrete sampling units in which we measured temporal
141 changes in rush frequency (see Supplemental File 1). These parcels varied in size from 0.5 to
142 18.8 ha (mean parcel area of 2.8 ± 0.1 ha) and occurred at elevations ranging from 140 to 341
143 m (mean parcel elevation of 253.5 ± 2.4 m)

144

145 **2.4. Retrieval and processing of Google Earth imagery**

146 We downloaded Google Earth images from 2005, 2009, 2015 and 2018 that corresponded to
147 the 300 marginal grassland sample parcels we intended to survey. Google Earth images were
148 available for every sample parcel in 2005 and 2018 but only for a selection of parcels in 2009
149 and 2015. Furthermore, even when an image was available for a given survey year, there were
150 specific instances when it could not be used for a given sample parcel. For example, if the
151 sample parcel had been mown, was shaded, covered in bare earth (e.g. temporary ground
152 disturbance, such as ploughing) or there was low contrast between rush tussocks and the
153 surrounding vegetation. Consequently, we used a different number of grassland sample parcels
154 during each survey year (Table 1). Further information on image availability and usage is
155 provided in the Supplementary Information (Files 2 & 3).

156 A total of 205 high-resolution Google Earth images were downloaded (Table 2). All
157 images were selected from an eye altitude of 1 km while all Google Earth layers were switched
158 off. Also, before a Google Earth image was captured, the compass and tilt were reset, and the
159 ‘Atmosphere’, ‘Sun’ and ‘Water surface’ options from the ‘View’ menu were also deselected.
160 After an image was downloaded, it was imported into ArcGIS and then georeferenced. Google
161 Earth images are orthorectified, but the original images are captured using different camera
162 angles (Google Inc). Therefore, to enhance subsequent alignment, the images were
163 planimetrically corrected. We began by georeferencing 2018 images to the Ordnance Survey
164 Open Carto base map layer within ESRI ArcGIS 10.4 using four control points per image (e.g.
165 building corners, road intersections, field boundary intersections). We then aligned 2005, 2009
166 and 2015 images to the georeferenced 2018 images using between 4 and 35 control points per
167 image, i.e., we stopped adding control points once a reasonable alignment had been achieved.

168 Root Mean Square (RMS) error is a measure of the difference between known locations and
169 locations that have been georeferenced, i.e., it is a measure of georeferencing accuracy.
170 Therefore, care was taken to ensure that the RMS error of each georeferenced image was <1
171 (Table 2). Additional information about the aerial images used in this study is contained within
172 Supplemental File 2 (image date, the number of georeferenced points used and the RMS error
173 per image).

174

175 **2.5. Sampling strategy**

176 We used a stratified random sampling approach whereby we recorded rush frequency per
177 grassland parcel within ten randomly placed 2 x 2 m quadrats sited in a 20 x 20 m randomly
178 located sample area. The same random quadrats were used during each survey year (2005,
179 2009, 2015 and 2018). To begin with, a negative 20 m buffer was applied to each of the 300
180 grassland parcels. This was done to ensure that the randomly located sampling plots did not
181 extend outside the grassland parcel boundary. We then created a single randomly located 20 x
182 20 m sampling plot within each of the 300 marginal grassland parcels using the ‘Create
183 Random Points’ and ‘Buffer’ tools within ArcGIS. After this, we used the same process as
184 above to create ten random 2 x 2 m quadrats within each 20 x 20 m sample plot. During this
185 process, we set the ‘Minimum Allowed Distance’ to 1.5 m to ensure that the quadrats did not
186 overlap. Finally, we recorded whether rush tussocks were present or absent within each of the
187 ten quadrats for each available plot and survey year combination (see Supplemental File 3 for
188 raw frequency data). Figure 4 provides an illustrative example of how rush frequency was
189 recorded across survey years.

190

191 **2.6. Accuracy and limitations of the method**

192 We validated the accuracy of our rush detection method by ground-truthing 45 (15%) of the 20
193 x 20 m sample plots. Validation plots were selected using a convenience sample, i.e., plots
194 were selected based on their proximity to roads and public footpaths. The first stage of the
195 validation process involved visiting all 45 of the 20 x 20 m validation plots and recording
196 whether rush tussocks were present or absent. A shapefile containing all 45 of the 20 x 20 m
197 validation plots was loaded into Google Maps (Google Inc) so that they could be accurately
198 located using a tablet in the field. It is important to note that we recorded rush as absent if
199 individual stems of young rush plants were present, but rush tussocks were absent. We did this
200 because individual rush stems are not visible on aerial imagery, but rush tussocks are.

201 Consequently, our approach is likely to underestimate rush frequency. Ground truthing took
202 place on the 20th of September 2019.

203 During the second stage of the validation process, the most recent Google Earth images
204 used during our survey (2018) were inspected to determine whether rush was present or absent
205 within each of the 45 plots visited in the field. Unfortunately, due to the lack of site-specific
206 field data, we could not validate rush presence within the plots during earlier study years (2005,
207 2009, 2015). The field and 2018 aerial image data were then compared, and this indicated there
208 was 100% agreement between the two datasets (see Supplemental File 4 for raw validation
209 data). Despite the complete agreement between aerial imagery and field data, the 2 x 2 m
210 quadrat polygons used during our survey are only likely to have sampled the same approximate
211 (rather than exact) area within each grassland parcel between sample years. This is because
212 Google Earth imagery is orthorectified, but the source images are captured using different
213 camera angles, which means perfect alignment between survey years is impossible.
214 Nevertheless, the RMS error of georeferenced images was extremely low during each survey
215 year (Table 2). Furthermore, during the georeferencing process, care was taken to ensure that
216 the field boundaries of the sample grassland parcels were aligned between survey years.
217 Finally, it is also worth noting that other types of tall vegetation (e.g. thistles or nettles) may
218 look similar to rushes on aerial imagery. However, such vegetation was rare within validation
219 plots. In short, while our approach is not perfect, we believe that we have minimised error
220 sufficiently to be confident that our approach is an accurate and valid technique for measuring
221 rush frequency within marginal grasslands.

222

223 **2.7. Data analysis**

224 All statistical tests were performed in R v.3.6.0 (R Core Team, 2019). Plot within study year
225 served as a replicate during data analysis. For every plot-year combination (i.e. replicate), we
226 summed the number of quadrats containing rush, which gave a rush frequency score of between
227 0 and 10. We subsequently examined temporal changes in rush frequency in two stages.

228

229 *2.7.1. Stage one: measuring rush frequency using the complete dataset*

230 Initially, we used descriptive statistics to explore changes in mean rush frequency across all
231 survey years using all the sample plots for which frequency data were available: 294 sample
232 plots in 2005, 106 sample plots in 2009, 189 sample plots in 2015 and 283 sample plots in
233 2018. We also calculated and graphed the proportion of plots per study year in which rush
234 frequency was: 0 (absent), 1-3, 4-6, 7-9 or 10 (dominant).

235

236 2.7.2. Stage two: measuring rush frequency using only continuous data

237 During the second stage of analysis, we only used those plots for which continuous rush
238 frequency data were available, i.e., the plots that had frequency data available for 2005, 2009,
239 2015 and 2018 (91 of the 300 plots examined). Using these data, we tested for changes in rush
240 frequency over time (2005, 2009, 2015 and 2018) using a Friedman's test. We used Friedman's
241 test instead of a repeated-measures ANOVA because the data failed to meet several parametric
242 assumptions, namely, normality and the homogeneity of variances. Friedman's test was
243 followed up by post hoc comparisons between individual survey years using Wilcoxon signed-
244 rank tests in which pairwise significance values were adjusted using the Bonferroni correction
245 method.

246 Using the continuous frequency data, we then calculated and graphed three additional
247 parameters. First, we calculated the average percent change in rush frequency per plot between
248 2005-2009, 2009-2015 and 2015-2018. Second, each of the 91 plots was assigned to one of
249 three categories depending on whether rush frequency remained stable, increased or decreased
250 between 2005 and 2018: 'No change' (=), 'Positive' (+) or 'Negative' (-). Finally, we
251 calculated the number of plots per study year in which rush frequency was: 0 (absent), 1-3, 4-
252 6, 7-9 or 10 (dominant).

253

254 3. RESULTS

255 3.1. Examining rush frequency using the complete dataset

256 The complete dataset suggests that rush frequency has increased by 81.7% over the whole study
257 period between 2005 and 2018 (Fig. 5a). In line with these increases, rush absence decreased,
258 and rush dominance increased within sample plots between 2005 and 2018 (Fig. 5b). For
259 example, rush was absent in 57.3% of the plots during 2005 but only absent in 35.3% of plots
260 in 2018 (Fig. 5b). Conversely, rush was dominant in only 6.8% of plots in 2005, but 16.3% of
261 plots in 2018 (Fig. 5b).

262

263 3.2. Examining rush frequency using only continuous data

264 For the 91 plots for which we had continuous data, we recorded an increase in rush frequency
265 during each consecutive study year (Fig. 6a). Overall, mean rush frequency increased by
266 174.2% between 2005 and 2018. The Friedman test results indicated that the differences in
267 rush frequency across all study years were significant ($d.f. = 3, \chi^2 = 48.5, p < 0.001$).

268 Furthermore, post hoc Wilcoxon signed-rank test comparisons suggested that there were
269 significant differences in rush frequency between 2005-2018 ($p = 0.003$) and 2009-2018 ($p =$
270 0.023) (Fig. 6a). Conversely, changes in rush frequency between 2005-2009, 2005-2015 and
271 2009-2015 and 2015-2018 were not significant.

272 The largest percentage increases in rush frequency within the WPM SSSI occurred
273 between 2009-2015 and 2015-2018, with mean percentage increases in rush frequency per plot
274 of $51.9 \pm 17.2\%$ and $53.8 \pm 15.7\%$ recorded during these periods respectively (Fig. 6b). Overall,
275 between 2005-2018 rush frequency remained unchanged within 45 plots (49.5% of plots),
276 increased within 39 plots (42.9% of plots) and decreased within seven plots (7.7% of plots)
277 (Fig. 6c). Finally, during each consecutive study year (2005, 2009, 2015 & 2018) the number
278 of plots in which rush was absent decreased and the number of plots in which rush was
279 dominant increased (Fig. 6d).

280

281 **4. DISCUSSION**

282 Our results provide quantitative evidence of rush expansion within the marginal upland
283 grasslands of the WPM SSSI between 2005 and 2018. Both datasets suggest that rush frequency
284 has increased by 81.7% (all data) to 174.2% (continuous data) during the study period.
285 Moreover, the continuous dataset indicates that between 2005-2018 rush frequency increased
286 within 42.9% of plots, but only decreased within 7.7% of plots. The continuous data also shows
287 that the largest increases in rush frequency occurred more recently between 2009-2015 (51.9%)
288 and 2015-2018 (53.8%), with only moderate increases recorded between 2005-2009 (22.3%).
289 These findings corroborate the results reported in the grey literature, which suggest that there
290 have been significant increases in rush cover or frequency over time within the upland hay
291 meadows of northern England (O'Reilly, 2011; Hamilton *et al.*, 2018). However, our study
292 differs in that: we measured rush expansion within marginal semi-improved upland grasslands
293 (as opposed to upland hay meadows); we used a much greater number of sample fields and
294 quadrats; we measured changes in rush frequency across a greater number of time periods (we
295 used four time periods, whereas studies in the grey literature used two); and, more importantly,
296 we used a consistent survey method across each time period.

297 Despite recording large and significant increases in rush cover, by 2018, there were still
298 between 35.3% (all data) to 53.9% (continuous data) of plots in which rushes were absent.
299 Furthermore, the continuous data also shows that within 42 of the 91 plots examined (46.2%
300 of continuous data plots) rushes were absent throughout the entire duration of the study (i.e.

301 during 2005, 2009, 2015 and 2018). Given that rush frequency did not increase within every
302 grassland parcel and that the greatest increases in rush frequency happened after 2009, recent
303 changes in field-level management appear to be the most likely cause of rush expansion within
304 the WPM SSSI. Nevertheless, the drivers behind the recent expansion of rushes within upland
305 grasslands are currently unknown.

306

307 **4.1. Factors controlling rush expansion within upland grasslands**

308 *4.1.1. Field-level factors*

309 One possible field-level factor driving the recent increase in rushes within upland grasslands is
310 inadequate drainage. The gradual decline in the number of farmworkers combined with the low
311 profitability of upland farming means that farmers do not have the time, labour or money to
312 maintain existing drains or install a new drainage system. Given the preference of rushes
313 (especially *J. effusus*) for damp conditions (Richards and Clapham, 1941b, c, d; Hill *et al.*,
314 2004), the recent decline in operational and efficient field drainage systems may have
315 facilitated rush expansion. Surprisingly, Hamilton *et al.* (2018) found no evidence of a
316 relationship between drainage and temporal changes in rush cover within the upland hay
317 meadow sites they studied, but this could have been due to difficulties in relocating quadrat
318 samples between repeat surveys and/or the assessment of hay meadow vegetation at the quadrat
319 rather than field scale (e.g. two to three repeat quadrats per hay meadow).

320 Drainage capacity may have been further reduced in recent times by the increasing use
321 of heavier farm machinery. For example, Hamilton *et al.* (2018) found that none of the upland
322 hay meadow sites they studied had modern field drains, with many fields being described by
323 farmers as having ‘old’ or ‘Victorian’ drainage systems (44.2% of farmers asked). Such old
324 drainage systems are likely to have collapsed under the weight of heavier modern machinery
325 and, because farmers are unable to repair or replace them, the soil in these fields will have
326 become much wetter and thereby more favourable to rushes. The use of heavy farm machinery
327 may have also caused soil compaction (Keller *et al.*, 2019), which, in turn, may have facilitated
328 rush expansion via increased soil surface wetness due to the creation of an impenetrable pan of
329 soil preventing surface water from percolating down to the sub-soil and any existing field
330 drains (Chyba *et al.*, 2014; Chyba *et al.*, 2017).

331 During the headage era (1980-2005) hill farmers were paid a subsidy based on the
332 number of sheep within their flock (Thomson, 2011). This policy led to the overstocking of
333 sheep and may well have led to increased soil compaction and surface wetness (and thereby

334 rush expansion) within marginal grasslands (Wathern *et al.*, 1985; Fuller and Gough, 1999;
335 Sutherland, 2002). For example, sheep grazing can increase soil bulk density and reduce soil
336 infiltration capacity within upland grasslands (Marshall *et al.*, 2014). Overstocking of sheep
337 may also lead to poaching, especially on undrained fields with wet soils (Bilotta *et al.*, 2007).
338 The creation of bare ground via poaching would facilitate the spread of rushes by providing
339 the germination niches required by overwintering seeds lying dormant at the soil surface
340 (Agnew, 1961; McCarthy, 1971; Cairns, 2013). Poaching induced rush germination may even
341 occur at low stocking densities in rush dominated grasslands because, due to the low
342 palatability of rushes (Grant *et al.*, 1984; Nielsen and Sjøgaard, 2000; Tweel and Bohlen,
343 2008), sheep may concentrate their feeding activity within the small patches of grass that
344 remain. Thus, what should be a low stocking density in a rush-free grassland, becomes a high
345 stocking density that causes localised poaching on the few remaining areas of productive
346 grassland.

347 Sheep numbers within the British uplands have declined substantially since the
348 outbreak of Foot and Mouth Disease in 2001 and the end of headage in 2005 (SAC, 2008;
349 Thomson, 2011). Nevertheless, stocking densities may still be high enough to cause localised
350 soil compaction and surface ponding in upland grasslands (e.g. Marshall *et al.*, 2014). Thus,
351 current stocking levels may still be promoting rush expansion, especially in rush dominated
352 fields where grazing is restricted to small areas of palatable grass.

353 Another possible field-level factor that has encouraged rush expansion is a reduction in
354 management intensity. Many of the upland grassland agri-environment schemes available to
355 farmers restrict the application of inorganic fertilisers or livestock manures and lime (RPA,
356 2019a; RPA, 2019b). Before the widespread adoption of such schemes, farmers would
357 regularly fertilise their fields and increase the pH by liming, with both actions making the
358 conditions more favourable to grasses and less favourable to rushes (Hill *et al.*, 2004; Cairns,
359 2013). Consequently, rushes may have been held back due to farmers making the grasses more
360 competitive (Cairns, 2013).

361 The cessation of traditional farming practices may have also created a series of field-
362 level factors that may have contributed to the spread of rushes within upland grasslands. For
363 example, upland farmers used to keep a much wider range of livestock than just sheep,
364 including native cattle and pony breeds (Fuller and Gough, 1999) that, unlike sheep, find rush
365 more palatable (Grant *et al.*, 1984; O'Reilly, 2012; Coyle *et al.*, 2018). Native cattle and ponies
366 may have been present in enough numbers to control rush expansion. Farmers also used to
367 mow, bale and remove grassland cuttings every year, which could have reduced rush seed fall

368 and germination. Furthermore, the practice of burning rushes within marginal grasslands (i.e.
369 swaling) has disappeared in upland areas across the UK. This practice would have had a
370 negative effect on rush abundance via reductions in biomass and seed load (Ghantous and
371 Sandker, 2015) and would have also increased the competitiveness of grass (in relation to
372 rushes) via increases in soil nutrients and pH (e.g. Niering and Dreyer, 1989; Dudley and
373 Lajtha, 1993; Brockway *et al.*, 2002).

374 To truly understand if and what field-level factors are contributing to rush expansion,
375 we need to combine our satellite imagery approach with historical management data.
376 Unfortunately, accurate historical data was not available for the grassland parcels used in this
377 study, but such data is likely to be available in other areas across the UK.

378

379 *4.1.2. Climatic factors*

380 North West England and North Wales (the climatic region in which this study took place) were
381 3% wetter between 2005 and 2018 than they were between 1981-2010 and 7% wetter than they
382 were between 1961-1990 (Met Office, 2020b). Furthermore, recent increases in wetness during
383 winter and summer have been even greater within the study region (Met Office, 2020b). For
384 example, winters between 2005-2018 were 5% wetter than winters between 1981-2010 and
385 14% wetter than winters between 1961-1990 (Met Office, 2020b). Likewise, summers between
386 2005-2018 were 13% wetter than summers between 1981-2010 and 14% wetter than summers
387 between 1961-1990 (Met Office, 2020b). By facilitating more favourable conditions for rushes
388 (i.e. wetter and warmer), the recent increases in wetness may have compounded field-level
389 drivers of rush expansion, such as inadequate drainage, soil compaction and poaching.

390 Alongside the observed increases in precipitation, there has been a recent reduction in
391 the number of days of air frost across the study region. For example, between 2005–2018, there
392 have been 6% fewer days of air frost compared to the 1981–2010 average (Met Office, 2020a).
393 Similarly, compared to the 1961–1990 average, there have been 16% fewer days of air frost
394 between 2009-2018 (Met Office, 2020a). Several studies suggest that rush regrowth after
395 cutting (or grazing) is reduced when plants are exposed to freezing temperatures (Folkestad *et*
396 *al.*, 2010; Østrem *et al.*, 2018). Thus, combined with the cessation of traditional management
397 (e.g. swaling, use of a wider range of native grazers or the cutting and removing grassland
398 arisings), the recent reductions in the number of air frost days may have also contributed to
399 grassland rush expansion.

400

401 **4.2. Implications of rush expansion within upland grasslands**

402 The expansion of rushes within upland grasslands has several negative consequences. First and
403 foremost, as rushes increase, palatable and productive grasses tend to be outcompeted.
404 Consequently, rush infestations reduce farm productivity. For example, Cairns (2013) states
405 that a “15% rush infestation in a productive grass sward, could reduce output by 1.25t
406 DM/ha/annum. If the field is cut for big bale silage on upland in-bye fields, the value of this
407 lost production could be as high as £192/ha (£78/acre)”. As Hamilton *et al.* (2018) note, such
408 large losses are extremely significant on livestock farms in marginal upland areas within
409 England where the average farm income is between £130/ha and £141/ha (Rural Business
410 Research, 2018 data from North West and North East England, respectively). Secondly, rush
411 infestations lead to declines in plant and bird biodiversity. For instance, as more grassland area
412 is taken up by rushes, there is less space for other grassland species. Also, while snipe and
413 curlew may nest in rush-dominated fields, redshank (*Tringa tetanus* L.) and lapwing prefer to
414 nest in fields with a mixture of scattered rush tussocks (no more than 30% cover) and grassland
415 patches in which to feed (RSPB, 2017; Coyle *et al.*, 2018).

416 Rush dominated fields, particularly bordering heather moorland, could also be a
417 significant, but currently unidentified, wildfire risk, especially given that we know rushes are
418 combustible (e.g. as highlighted by the historical practice of swaling, but also see Ghantous
419 and Sandker, 2015). Furthermore, fields in which rush cover exceeds 50% will have a
420 significant amount of biomass that is likely to become very dry (and thereby more combustible)
421 during summer. To date, the wildfire risk posed by moorland edge rush infestation has not been
422 investigated. If rush infestations *do* pose a significant wildfire risk, we would need to reduce
423 rush cover at and just below the moorland line. Such a task would be difficult, given that we
424 still do not know the most effective way to control rush infestations within grassland habitats
425 (O'Reilly, 2012; Coyle *et al.*, 2018).

426

427 **4.3. Research priorities**

428 Our protocol for measuring rush frequency is subjective and restricted to grassland habitats
429 where there is a clear height, colour or tone differential between rush tussocks and the
430 surrounding vegetation. Therefore, an obvious next step would be to develop a more objective
431 and automated protocol for quantifying rush abundance across multiple habitats. One approach
432 would be to use Light Detection and Ranging (LiDAR) data to differentiate rush tussocks from
433 the surrounding grassland vegetation in the same way tree canopies can be differentiated from
434 the understory vegetation and the forest floor (e.g. Latifi *et al.*, 2015; Hamraz *et al.*, 2017).
435 Rush tussocks are generally less than one metre wide (see Supplemental File 4), which means

436 that LiDAR with a spatial resolution of 1 metre or less would be the most appropriate for
437 mapping soft rush. However, in other habitats (e.g. acid grassland, heather moorland or blanket
438 bog) where there is less of a height differential between rushes and the surrounding vegetation,
439 LiDAR may have to be replaced by or supplemented with spectral band analysis using satellite
440 images, such as SENTINEL-2 or LANDSAT 8 (Davidson *et al.*, 2016; Erinjery *et al.*, 2018;
441 Forkuor *et al.*, 2018). Notwithstanding the points above, the development and implementation
442 of an automated protocol for measuring rush abundance in upland grasslands across the UK
443 are currently hampered by the limited coverage of high-resolution LiDAR data (spatial
444 resolutions of $\leq 1\text{m}$).

445 Four further research gaps need to be addressed. Firstly, we need to replicate our
446 satellite imagery approach across different areas of the UK and further validate the method by
447 using both contemporary and historical field data. Secondly, we need to determine the drivers
448 behind the recent expansion in rushes within upland grasslands across the UK. This could be
449 achieved by mapping changes in rush frequency over time and exploring how different
450 management and environmental factors have influenced these changes. Potential drivers of
451 rush expansion to explore are historical changes in management (e.g. changes in drainage
452 efficiency, reduction in stocking levels and restricted fertiliser inputs), changes in climate (e.g.
453 changes in rainfall and temperature) and environmental factors (e.g. slope, aspect and
454 proximity to standing water). Climatic and topographical data for the UK are freely available
455 online (e.g. Met Office and Ordnance Survey), and historical management data could be
456 obtained by interview or questionnaire.

457 Thirdly, we need to establish the most effective rush control techniques to give land
458 managers the tools to reduce rush dominance. The effectiveness of several rush control
459 techniques have been explored within several studies (see Coyle *et al.*, 2018; O'Reilly, 2012
460 and references therein), but not in any depth or within an experimental framework that
461 compares the efficacy of different control methods across different farms with varying
462 environmental and management contexts (i.e. in a way that provides practical knowledge to
463 farmers and land managers).

464 Finally, we need to quantify the fundamental niche of soft rush, hard rush and compact
465 rush. Knowledge of the environmental tolerances of these invasive rush species will enable us
466 to better understand the drivers behind the recent expansion in rushes within upland grasslands
467 and allow us to reduce rushes where they have become dominant.

468

469 **5. CONCLUSIONS**

470 This is the first peer-reviewed study to document the recent increases in rush abundance within
471 upland grasslands. Our data suggest that the frequency of rushes within the marginal grasslands
472 of the West Pennine Moors SSSI has increased by 81.7% to 174.2% between 2005-2018. It is
473 not clear why such increases may have occurred. However, they may be due to changes in
474 field-level management, which have been further compounded by recent increases in rainfall
475 and reductions in the number of air frost days. Future research into rush ecology, expansion
476 and management is urgently required to determine the broader extent of the problem in England
477 and to combat the negative consequences of grassland rush infestations on the upland farm
478 economy and grassland biodiversity.

479

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483 with a rush management workshop which we held on the 1st of August 2019 in the Forest of
484 Bowland as part of this project. We would also like to thank Ian Cairns (Agrifood Technical
485 Services) for attending our rush management workshop and giving a very informative talk and
486 field demonstration to farmers about grassland rush control. Finally, we would like to thank the
487 anonymous reviewers for their helpful suggestions to improve the manuscript.

488

489 **COMPETING INTERESTS**

490 M. Ashby has provided independent ecological advice and evidence synthesis services to the
491 Moorland Association since April 2019 and the Game & Wildlife Conservation Trust since
492 October 2019.

493

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649

650 **SUPPORTING INFORMATION**

651 Supplemental File 1 – Grassland Parcel and Sample Plot Data

652 Supplemental File 2 – Aerial Imagery Data

653 Supplemental File 3 – Raw Rush Frequency Data

654 Supplemental File 4 – Method Validation Data and Rush Tussock Dimensions Field Data

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TABLES

Table 1. The number of grassland parcels used for each survey year. The ‘All years’ category refers to sample parcels for which data were available across all four survey years (i.e. continuous data).

Survey year	Number of parcels used
2005	293
2009	106
2015	189
2018	283
All years	91

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Table 2. Descriptive statistics for the georeferenced Google Earth images used for each survey year. RMS error minimised using a 1st order polynomial (Affine) transformation. For further information about the Google Earth images used see Supplemental File 2.

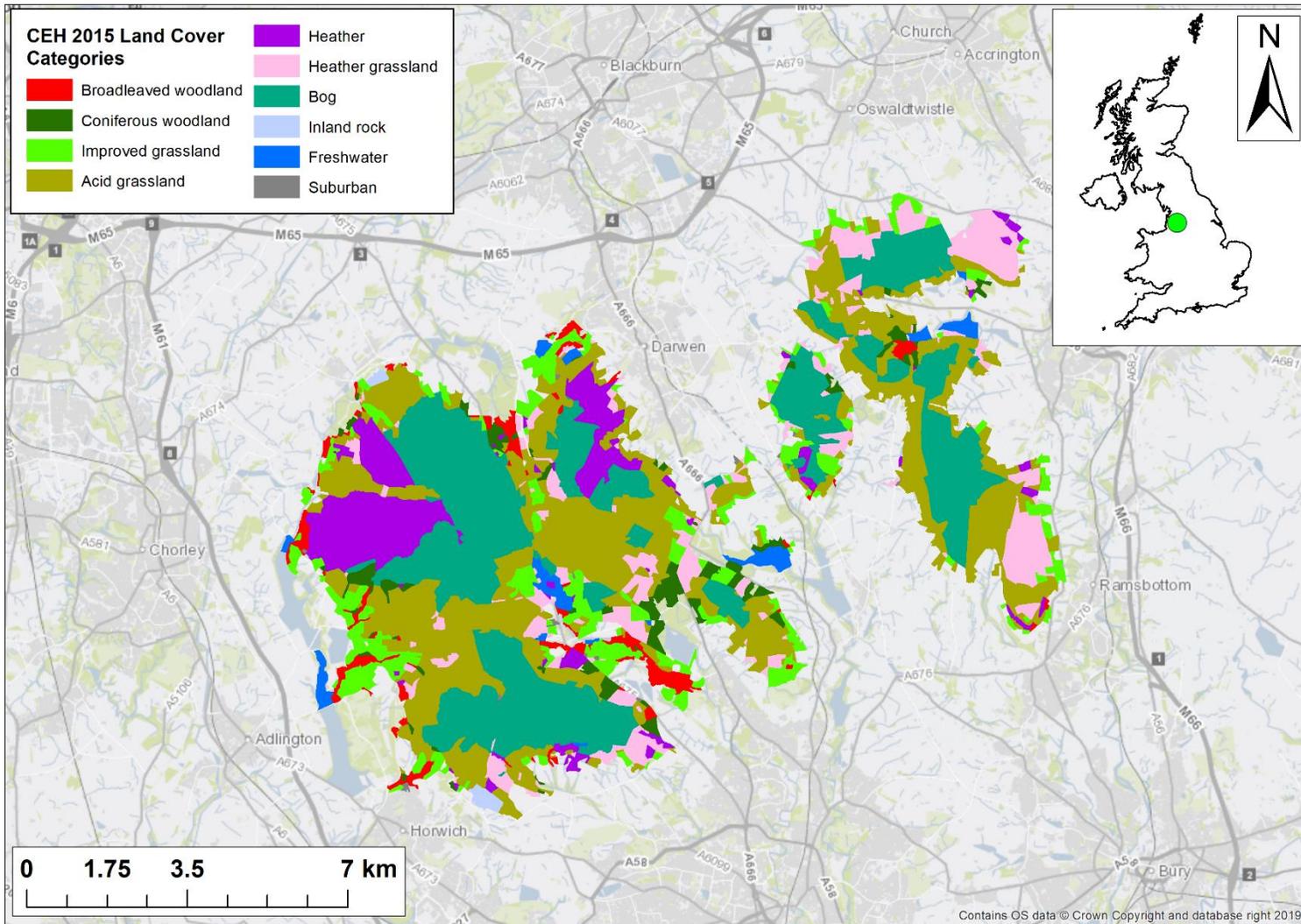
Survey year	No of images	Georeference points		RMS error	
		Mean \pm SEM	Min-Max	Mean \pm SEM	Min-Max
2005	70	10.4 \pm 0.7	4-30	0.4 \pm 0.0	0.02-0.56
2009	19	8.0 \pm 0.9	4-17	0.3 \pm 0.0	0.07-0.51
2015	46	9.1 \pm 0.8	4-35	0.3 \pm 0.0	0.03-0.75
2018	70	4.0 \pm 0.0	4-4	0.2 \pm 0.0	0.02-0.36

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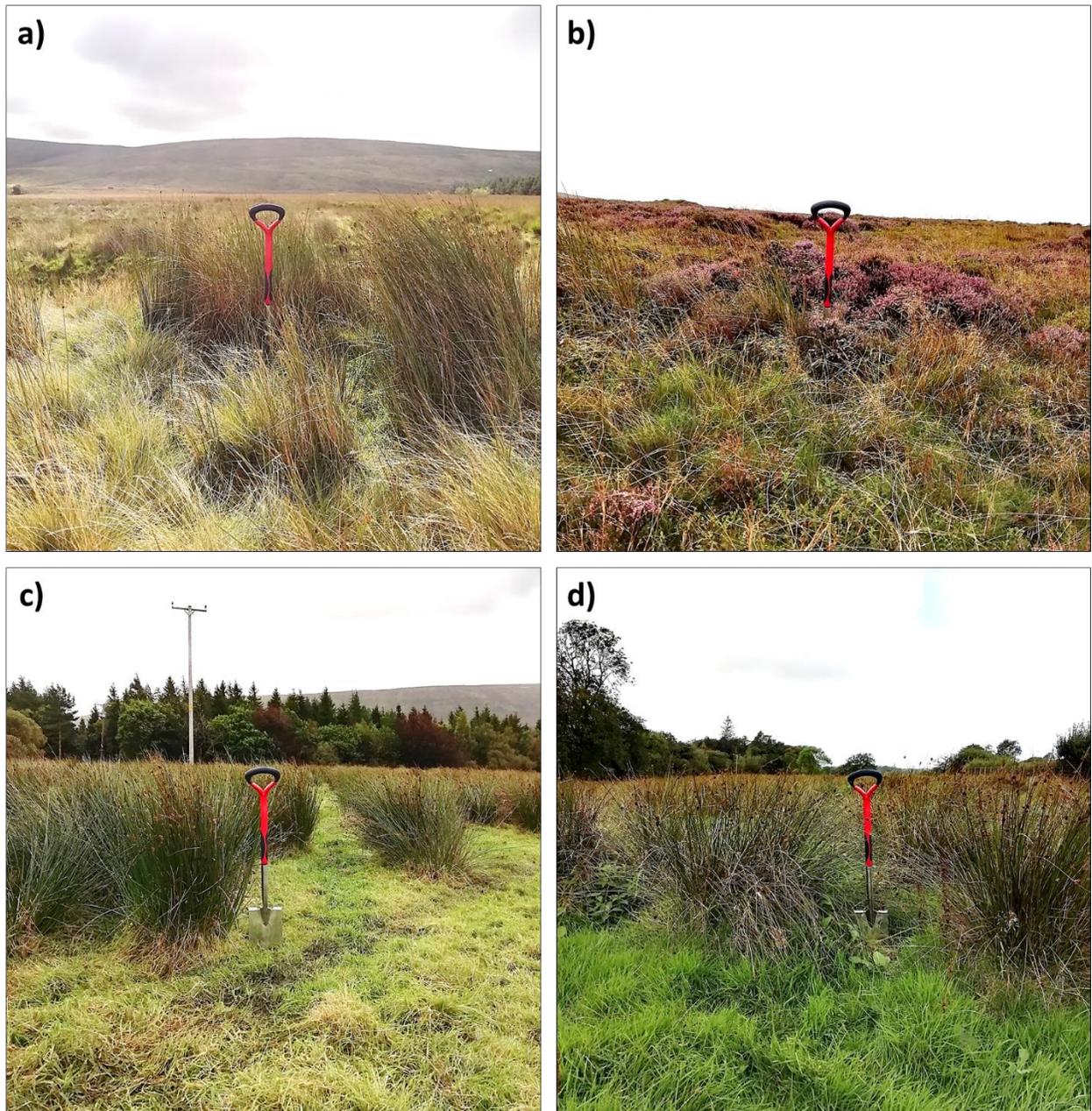
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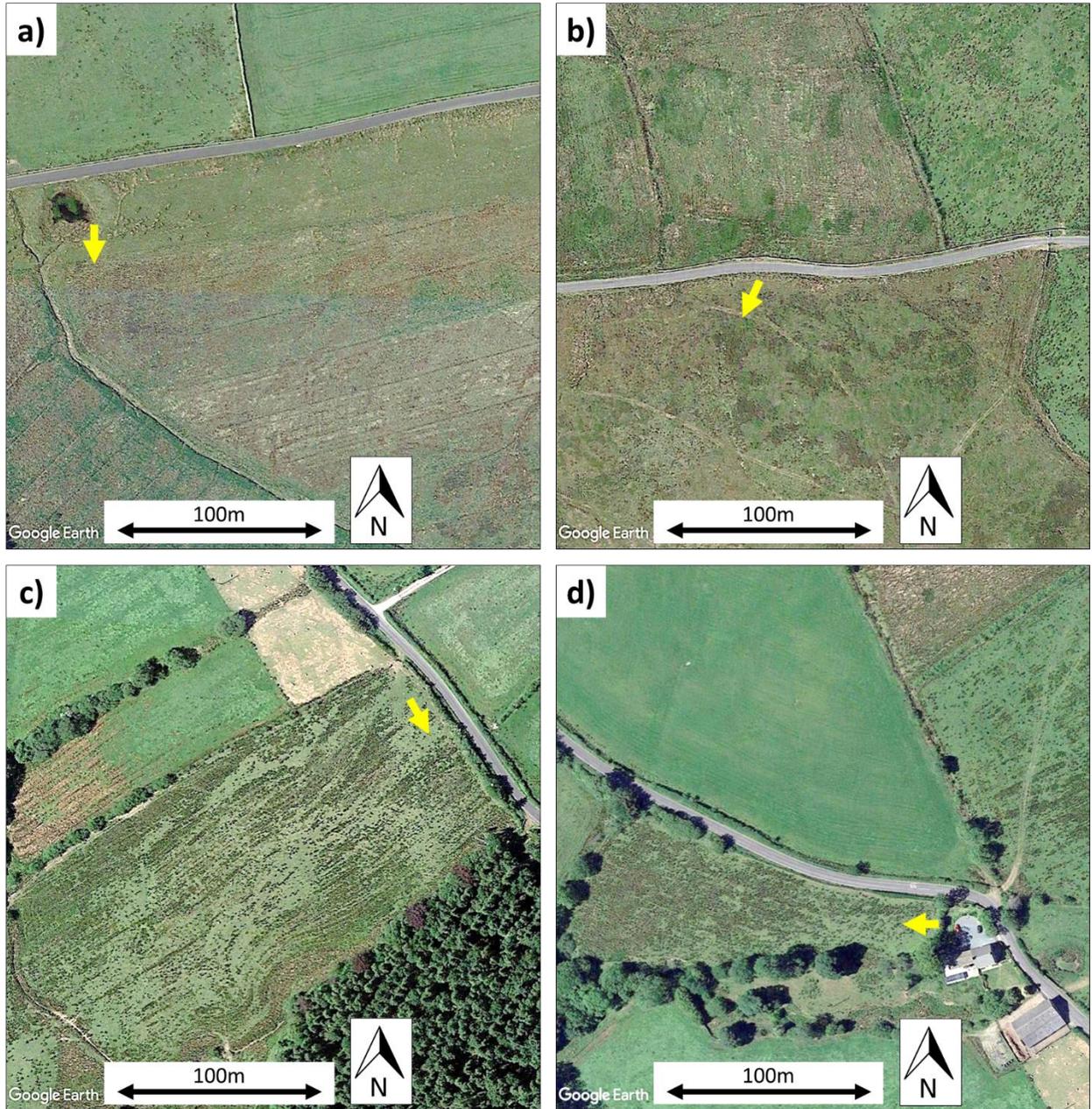
688 **Figure 1.** CEH land cover categories present within the West Pennine Moors SSSI (Rowland *et al.*, 2017). Inset: Location of the West Pennine
 689 Moors SSSI (green circle) in the UK. The base map used is the Ordnance Survey Open Background map accessed through ArcGIS 10.4.



691

692 **Figure 2.** The upper photos show the homogeneous height and colour contrast found between rushes and
 693 the surrounding vegetation within (a) Acid Grassland and (b) Heather Moorland. The lower photos show
 694 the heterogeneous height and colour contrast found between rushes and the surrounding vegetation within
 695 the Marginal Grasslands (c & d). The large height and colour contrast between rushes and the surrounding
 696 vegetation within Marginal Grassland parcels mean that it is clearly visible on Google Earth imagery (see
 697 Fig.3). The spade pictured is approximately 1 m tall. All photos were taken on the 11th of September 2019.

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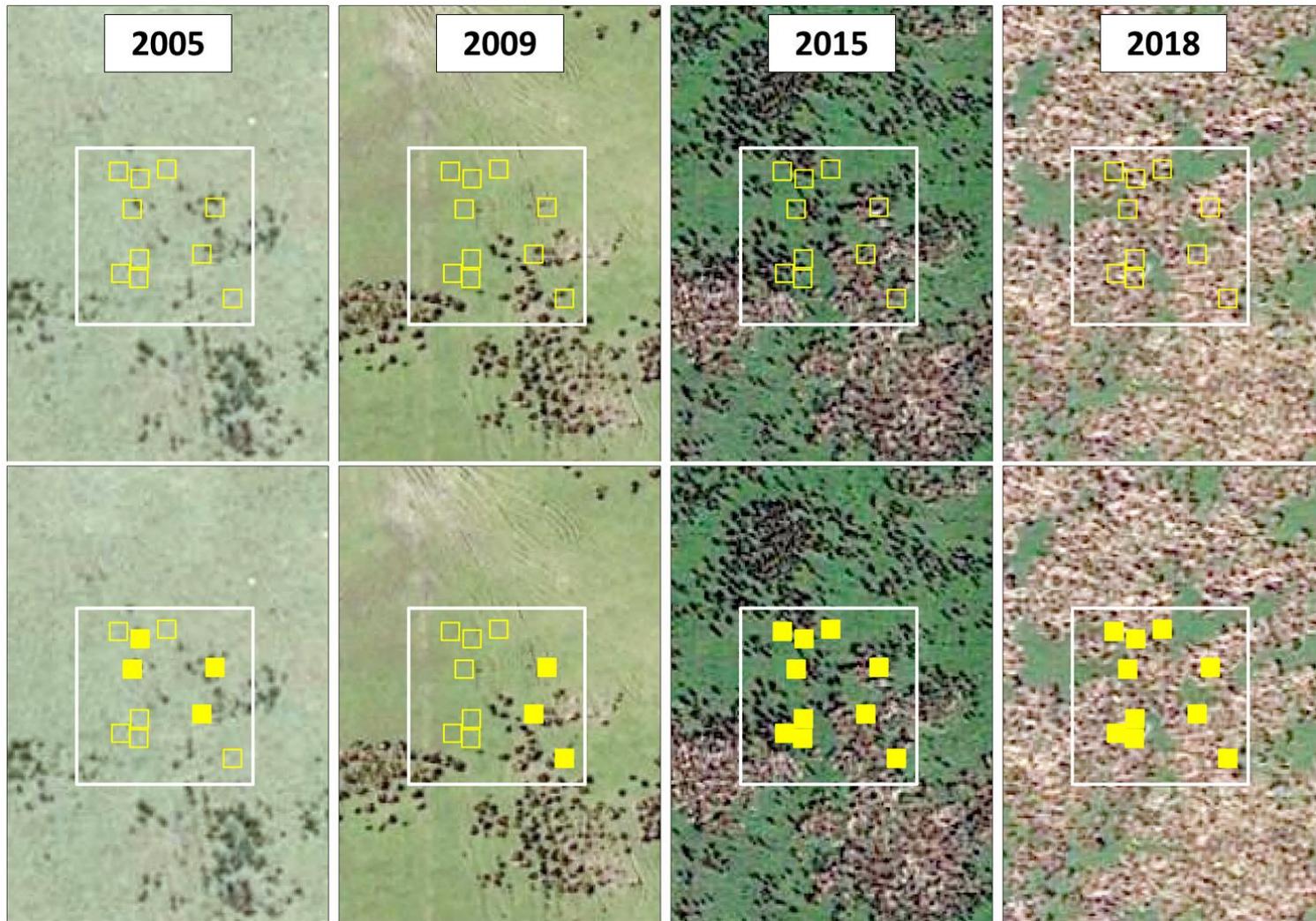
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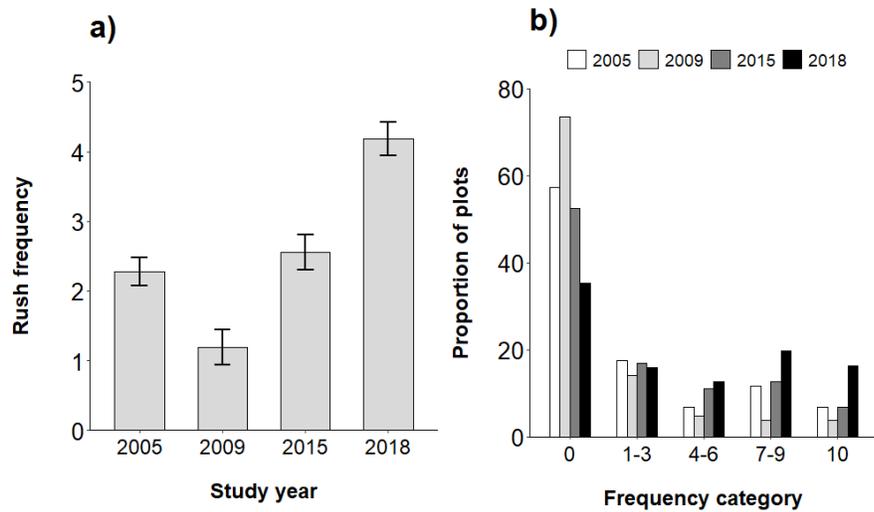
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Figure 3. Modified Google Earth images corresponding to photographs a, b, c and d in Fig. 2. The yellow arrow denotes the location and direction of the corresponding photograph. Note how rushes cannot be seen clearly within (a) Acid Grassland parcels and areas of (b) Heather Moorland, but they can be seen clearly within Marginal Grassland (c & d).



704

705 **Figure 4.** An illustrative example of recording rush frequency within the ten quadrats (yellow squares) in the sample plots (white squares) across
 706 each sample year. Along the bottom row, quadrats are filled if rush is present and unfilled if rush is absent. Quadrats along the top row are left
 707 unfilled for comparison. We recorded rush as present if any part of a rush tussock (no matter how small) was within the quadrat boundary.



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710 **Figure 5.** Results from the analysis of the complete dataset: a) mean rush frequency per year (error bars are
 711 standard errors of the mean); and, b) the proportion of plots per year in which rush frequency was: 0
 712 (absent), 1-3, 4-6, 7-9 or 10 (dominant). Rush frequency was measured within ten quadrats per sample plot
 713 per year.

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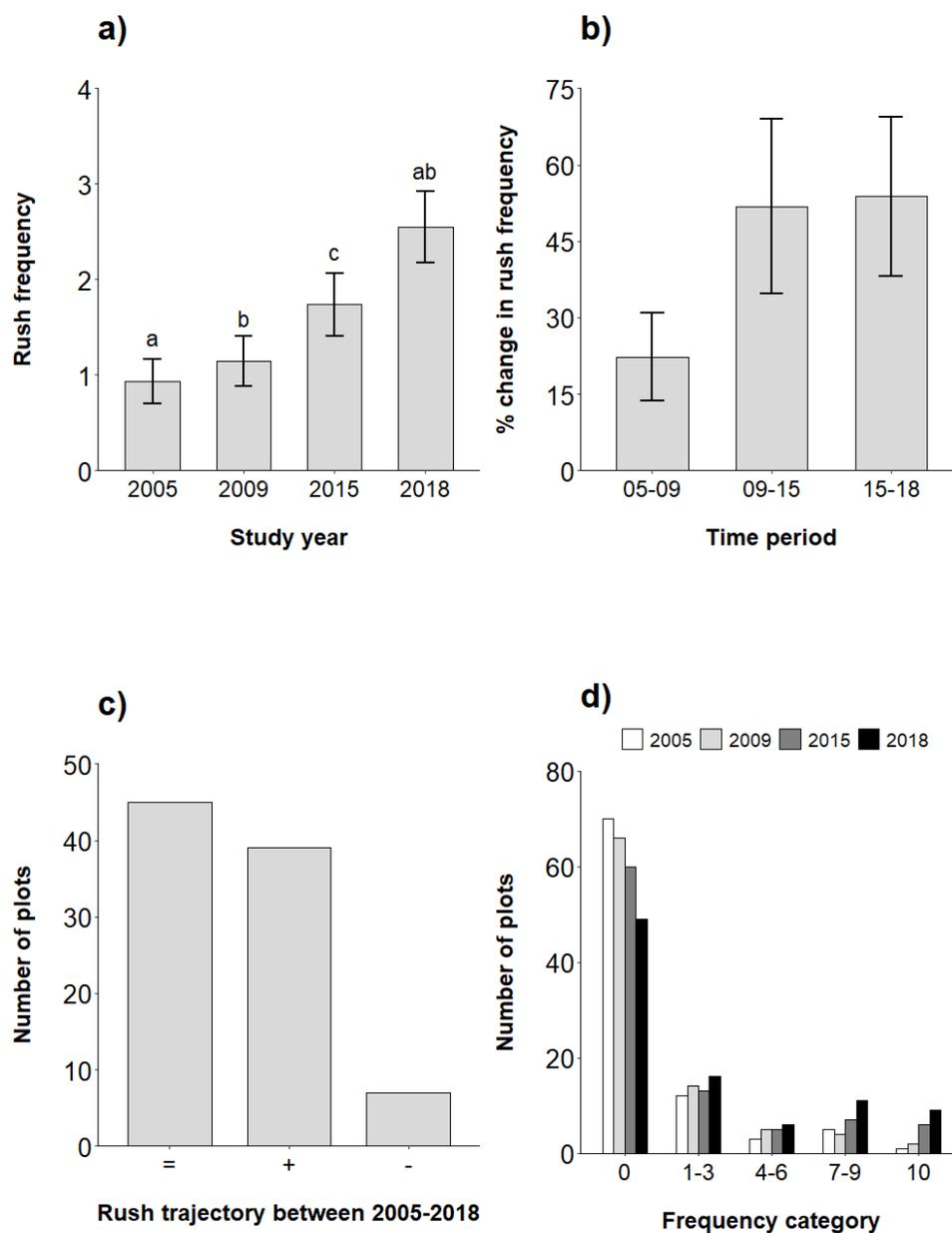
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 730 **Figure 6.** Results from the analysis of the continuous dataset: a) mean rush frequency per year with bars
 731 marked with different letters being significantly different ($p < 0.05$) according to post hoc comparisons
 732 between individual survey years using Wilcoxon signed-rank tests adjusted using the Bonferroni correction
 733 method; b) the mean percentage change in rush frequency per plot between 2005-2009, 2009-2015 and
 734 2015-2018; c) the number of continuous data plots in which rush frequencies displayed no change (=), were
 735 positive (+) or were negative (-) between 2005 and 2018; and, d) the proportion of plots per year in which

- 736 rush frequency was: 0, 1-3, 4-6, 7-9 or 10. For figures a) and b) error bars are standard errors of the mean.
- 737 Rush frequency was measured within ten quadrats per sample plot per year.