

1 Ecosystem dynamics in dune heathlands:
2 spatial and temporal effects of
3 environmental drivers on the vegetation

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9

10 **Abstract**

11 Pin-point cover data from 81 Danish dune heathland sites collected over 16 years were analyzed to quantify
12 the effects of key environmental drivers on vegetation dynamics. A spatio-temporal structural equation
13 model within a Bayesian hierarchical framework was used to assess the influence of nitrogen deposition,
14 soil pH, soil C–N ratio, soil type, precipitation, and grazing. The species composition at Danish dune
15 heathlands is changing, and most regression parameters for environmental drivers were significantly
16 different from zero, indicating strong regulatory effects on both large-scale spatial variation and temporal
17 dynamics. Precipitation and soil pH emerged as the most influential factors, underscoring the role of
18 hydrological and edaphic conditions in shaping the vegetation at dune heathlands. The model predicts
19 plant community dynamics and can be applied to forecast the effect of different management scenarios,
20 providing a basis for adaptive site-specific management planning.

21 Keywords: Plant community dynamics of dune heathlands; coastal dune ecosystems; spatial and temporal
22 variation of plant cover; hierarchical Bayesian models; pin-point cover data; structural equation modelling;
23 ecological predictions, adaptive management plans.

24 **Introduction**

25 To effectively manage semi-natural dune heathland ecosystems under changing climate conditions and
26 evolving land-use practices, it is essential to understand and predict how environmental drivers influence
27 vegetation. Developing a robust quantitative understanding of these drivers will allow us to make site-
28 specific predictions of future ecosystem responses to changes such as climate shifts and grazing regimes.
29 This knowledge will also provide critical input for adaptive management plans at the local level (Damgaard
30 2022a; 2025c). To clarify the causal relationships driving the observed large-scale variation and temporal
31 dynamics in dune heathlands, and to quantify the influence of environmental drivers, we developed and
32 fitted a structural equation model of the dune heathland ecosystem using spatial and temporal ecological
33 data from 81 Danish dune heaths collected between 2007 and 2022.

34 The conservation status and resilience of dune heathlands are increasingly compromised by interacting
35 pressures, including nitrogen deposition, successional encroachment of e.g. *Pinus mugo*, and altered land-
36 use practices. Indicators such as mor-layer C/N ratios approaching 25, reduced lichen to moss ratios, and
37 dominance of invasive species signal declining ecological integrity and heightened vulnerability to
38 eutrophication. Maintaining resilience requires management that restores disturbance-driven mosaics,
39 limits nutrient enrichment, and curbs woody and invasive species (Nielsen 2005). Overall, the conservation
40 status of Danish dune heaths is assessed as moderately unfavorable (Fredshavn et al. 2025)

41 The dynamics of semi-natural heathland ecosystems, particularly those dominated by *Calluna vulgaris*,
42 were first investigated by Watt (1947), who provided a comprehensive analysis of the ecological processes
43 underlying spatial vegetation patterns. Subsequent research has expanded upon this foundation (e.g.
44 Gimingham 1978; Gimingham 1988; Gimingham et al. 1981; Løvschal and Damgaard 2022; Usher and
45 Thompson 1993), offering detailed descriptions of heathland vegetation structure across multiple spatial
46 scales and elucidating the role of management interventions in regulating these patterns.

47 The *a priori* selection of the studied environmental drivers was based on current ecological knowledge,
48 where specific environmental drivers, e.g. atmospheric nitrogen deposition, soil type, soil pH and
49 disturbance, have been shown to have an effect on heathland vegetation (e.g. Aerts et al. 1990; Aerts and
50 Heil 1993; Britton et al. 2003; Damgaard et al. 2024; Damgaard et al. 2020; De Graaf et al. 2009; Vogels et
51 al. 2020). However, dune heathland vegetation is likely influenced by additional environmental drivers that
52 could not be incorporated into this study due to data limitations, such as species-specific herbivory, pest
53 dynamics, and historical natural management interventions. Some of these more or less unknown factors
54 can have a geographical regional structure that may be partly explored using latent geographic factors
55 (Ovaskainen et al. 2016), and possible significant effects of such latent geographic factors can generate new
56 testable causal hypotheses. Furthermore, it is expected that soil type and nitrogen deposition may have
57 both direct and indirect effects by affecting soil pH (Damgaard et al. 2014), and such direct or indirect
58 causal pathways are often best modelled using structural equation models (SEM) that are fitted to
59 observed ecological data (Grace et al. 2010). Generally, the effects of changes in environmental variables,
60 e.g. climate, soil physical properties and disturbance regimes, on plant communities are expected to occur
61 with some time-lags (e.g. Svenning and Sandel 2013). Furthermore, the regulating environmental variables
62 are expected to vary considerably among different sites, and it is critical to integrate this large-scale spatial
63 variation into the analysis of the ecosystem dynamics.

64 The objective of this study is to investigate the effects of selected environmental and ecological drivers
65 (nitrogen deposition, soil type, soil pH, soil C-N ratio, precipitation, and grazing) on the vegetation at dune
66 heaths, which is summarized by the multivariate relative cover of six species or aggregated species groups
67 (*Calluna vulgaris*, *Empetrum nigra*, *Avenella flexuosa*, other graminoids, other herbs, and cryptogams).
68 Previously, similar models have been fitted to wet heathlands (Damgaard 2019; 2025b), dry heathlands
69 (Damgaard 2025a), as well as, acid (Damgaard 2022c), and calcareous grassland (Damgaard 2023).

70 **Materials and Methods**

71 **Dune heathlands**

72 Dune heathlands represent a dynamic coastal ecosystem characterized by nutrient-poor, acidic sandy soils
73 and a mosaic of successional stages ranging from grey dunes to decalcified fixed dunes and wet dune
74 slacks. The vegetation is typically composed of dwarf shrubs such as *Calluna vulgaris* and *Empetrum*
75 *nigrum*, interspersed with graminoids (*Deschampsia flexuosa*, *Carex arenaria*) and cryptogams, particularly
76 lichens of the genus *Cladonia* (Nielsen 2005).

77 *Calluna vulgaris* (L.) Hull (heather) is a 10 to 50 cm tall dwarf shrub that is dominant on dune heaths,
78 though not tolerant to heavy grazing, and prefers at least moderately well-drained soils (Gimingham 1960).
79 The developmental phases of heather include pioneer, building, mature and degenerate (Usher and
80 Thompson 1993; Watt 1947), and without management actions such as burning, grazing or cutting, *C.*
81 *vulgaris* plants degenerate when their age exceeds approximately 30 years, thereby leaving room for
82 heathland succession.

83 *Empetrum nigrum* L. (black crowberry) is a low-growing, allelopathic, evergreen dwarf shrub typically
84 forming dense mats on acidic, sandy, and nutrient-poor soils. It is well adapted to wind and salt aerosols. It
85 is often found in late-successional or climax communities, though it can also act as a pioneer species in
86 disturbed habitats. The dominant reproductive strategy is through adventitious rooting and sprouting from
87 the basal parts (Tybirk et al. 2000).

88 *Avenella flexuosa* (L.) Drejer (wavy hair-grass) is a slender, tufted perennial grass typically 20 to 60 cm tall,
89 which is found on acidic, sandy, and nutrient-poor soils. It is widely distributed across European heathlands
90 and often increases in abundance following reduced grazing or elevated nitrogen deposition. It is tolerant

91 of low nutrient availability but can become dominant under disturbance regimes that suppress dwarf
92 shrubs (Aerts and Heil 1993).

93 **Sampling design**

94 Hierarchical time-series data from 81 dune heathland sites (Fig. 1) that had been monitored at least three
95 times in the period from 2007 to 2022 were used in the analysis. Fifty-four of the 81 sites are NATURA 2000
96 habitat sites and are protected under the Habitat Directive (EU 1992). All sites included several plots
97 classified as dune heathland (EU habitat type: 2140) according to the habitat classification system used for
98 the European Habitat Directive EU (EU 2013). The area of the sites ranged from 0.36 ha to 2697 ha, with a
99 median area of 15.5 ha. A total of 612 unique plots were used in the analysis. The sampling was performed
100 in the summer, and all plots at a site were sampled on the same day. Otherwise, sampling intensity was
101 irregular among sites and years, but all sites were monitored within a six-year period. Typically, ten plots
102 were sampled from each site each time. Including resampling over the years, a total of 1530 plots were
103 used in the analyses. The plots were resampled with GPS-certainty (< 10 meters).

104 The data are a subset of the data collected in the Danish habitat surveillance program NOVANA (Nielsen et
105 al. 2012; Nygaard et al. 2024).

106 **Variables and measurements**

107 **Plant cover data**

108 The plant cover, which is the relative projected area covered by a species, was measured for all higher
109 plants by the pin-point method using a square frame (50 cm X 50 cm) of 16 grid points that were equally
110 spaced by 10 cm (Nielsen et al. 2012). At each grid point, a thin pin was inserted into the vegetation and
111 the plant species that were touched by the pin were recorded and used as an estimate of cover (Damgaard
112 and Irvine 2019; Levy and Madden 1933; Lindquist 1931). Since the pin-point cover data after 2007 were

113 recorded for each pin separately, the species cover data are readily aggregated into cover data for classes
114 of species at a higher taxonomic or functional level. At each grid point, the pin may hit different plant
115 species from the same species class and, in those cases, the hits are only counted as a single hit of the
116 species class at the grid point.

117 In this study, the species were classified into six groups: *C. vulgaris*, *E. nigra*, and *A. flexuosa*, any other
118 graminoids, any other herbs or cryptogams. The assumed distribution of pin-point cover data for single
119 species and the joint distribution for multiple species are outlined in the electronic supplement (Appendix
120 A).

121 **Nitrogen deposition**

122 Nitrogen deposition at each plot was calculated for each year using a spatial atmospheric deposition model
123 in the period from 2005 to 2014 (Ellermann et al. 2012). The mean site nitrogen deposition ranged from
124 5.83 kg N ha⁻¹ year to 18.69 kg N ha⁻¹ year⁻¹, with a mean deposition of 10.89 kg N ha⁻¹ year⁻¹.
125 Anthropogenic nitrogen deposition has reached a maximum in Denmark and is currently decreasing
126 (Ellermann et al. 2018).

127 **Soil pH**

128 Soil pH was measured in randomly selected plots from the uppermost 5 cm of the soil (four samples were
129 amassed into a single sample). The soils were passed through a 2mm sieve to remove gravel and coarse
130 plant material, and pH_{KCl} was measured on a 1 M KCl-soil paste (1:1). The soil sampling intensity was
131 irregular among sites and years, but typically between one and four plots were sampled from each site at
132 each time point. When a plot was resampled, the pH at the plot was calculated as the mean of the samples.
133 In total, 595 independent soil pH values were used in the analysis. The measured soil pH ranged from 2.9 to
134 6.7, with a mean soil pH of 3.77.

135 **C-N ratio in the soil**

136 Soil C-N ratio was measured in randomly selected plots from the uppermost 5 cm of the soil (four samples
137 were amassed into a single sample). Total C in each sample was determined by dry combustion and N by
138 the Kjeldahl method. The soil sampling intensity was irregular among sites and years, but typically between
139 one and four plots were sampled from each site at each time point. When a plot was resampled, the C-N
140 ratio at the plot was calculated as the mean of the samples. Measurements outside of the domain [10, 60]
141 were assessed to be unrealistic, and measurements outside the domain were set to either 10 or 60. The
142 mean site C-N ratio was 22.9. Moreover, the average soil C–N ratio on heathlands has been observed to
143 decrease in the period from 2004 to 2014 (Strandberg et al. 2018).

144 Since soil C-N ratio is expected to be influenced by the vegetation, e.g. by the amount of slowly
145 decomposing dwarf shrub litter, soil C-N ratio may be thought of more as an environmental covariable than
146 an environmental driver.

147 **Soil type**

148 The texture of the topsoil for each site was obtained from a raster based map of Danish soils (Greve et al.
149 2007). The categorical classification of the soil (JB-nr.) was made on an ordinal scale with decreasing
150 particle size, 1: coarse sandy soil, 2: fine sand soil, 3: coarse loamy soil, 4: fine loamy soil. There were some
151 records with other soil types, but because of possible classification errors they were treated as missing
152 values. The mean soil type was 1.49.

153 **Precipitation**

154 Site-specific precipitation was measured by the average annual precipitation in the period 2001 to 2010,
155 with a spatial resolution of 10 km (DMI 2014). The annual precipitation ranged from 587 mm to 903 mm,
156 with a mean precipitation of 798 mm.

157 **Grazing**

158 Land-use was summarized by possible signs of grazing, e.g. the presence of livestock or short vegetation
159 within fences was recorded by the observer at each plot for each sampling year since 2007 as a binary
160 variable (sign of grazing = 1, no sign of grazing = 0), i.e. if grazing was 0.5, then this probability may arise by
161 a number of ways, e.g. if half the plots at the site showed signs of being grazed each year or all plots were
162 grazed every second year. The mean grazing variable ranged from 0 to 1 among sites, but most sites had no
163 grazing and the mean grazing intensity at the site level was 0.11. Unfortunately, the grazing variable does
164 not include information on which animals were used for grazing, stocking densities or grazing duration, and
165 is therefore a quite imprecise variable that must be interpreted together with general knowledge on the
166 typically used grazing regime of dune heathlands.

167 **Geographic regions**

168 The 81 dune heathland sites were grouped into four arbitrary geographic regions without using prior
169 information (Fig. 1). These regions were used to investigate possible latent geographic factors.

170 **Spatio-temporal modelling**

171 To further understand the observed changes in species composition at dune heathlands(Damgaard 2025d),
172 and the causes for the moderately unfavorable conservation status of Danish dune heaths (Fredshavn et al.
173 2025; Nygaard et al. 2020), the observed changes in pin-point cover data were fitted to site variation in
174 selected abiotic and land-use environmental variables in a spatio-temporal structural equation model
175 (SEM) (Fig. 2).

176 It was decided to fit the SEM within a Bayesian hierarchical model structure using latent variables to model
177 the effect of measurement and sampling uncertainties (Fig. 2). This use of a hierarchical model structure is
178 important, since it has been demonstrated that ignoring measurement and sampling uncertainties may
179 lead to model and prediction bias (Damgaard 2020; Damgaard and Weiner 2021). Furthermore, it is an

180 advantage when making ecological predictions to separate measurement and sampling uncertainties from
181 process uncertainty. The hierarchical SEM approach and the motivation for using it are explained further in
182 (Damgaard 2025e). The mathematical and statistical details of the spatio-temporal modelling are explained
183 in the electronic supplement (Appendix B)

184 The procedures for estimating the most important single species cover and change in species cover for all
185 sampled dune heathland plots since the beginning of the monitoring program in 2004 are explained in the
186 electronic supplement (Appendix C), and the used code and additional tests of fitting properties etc. may
187 be found in Appendix D.

188 **Results**

189 The estimated cover and change in cover of the most common species in Danish dune heathland plots are
190 shown in Table 1. The most abundant species were *E. nigrum* and *C. vulgaris*, followed by *A. flexuosa* and
191 *Carex arenaria*. Significant declines were observed for *E. nigrum*, *Erica tetralix*, *Salix repens*, and *Carex*
192 *nigra*, with the latter showing the strongest negative trend on the logit scale. In contrast, *Cladonia sp.*
193 exhibited significant positive trends, while *C. vulgaris* and graminoids remained relatively stable.

194 The selected environmental variables covaried at the 81 dune heathland sites (Fig. S1), which again is
195 expected to lead to covariance among parameter estimates and affect the fitting properties of the model
196 negatively. Nevertheless, plots of the mean latent vs. expected logit-transformed cover variables
197 demonstrated a relatively good fit of the large-scale spatial variation in cover (Fig. 3A; between 46% and
198 56% of the variation is explained, Table S2), and the model fitted the temporal process of the change in
199 cover very well (Fig. 3B; > 97% of the variation is explained, Table S2). Furthermore, the Dunn–Smyth
200 residuals of the marginal observed cover data of the six species classes were approximately normally
201 distributed (Fig. S2).

202 To prevent possible prediction bias, the different sources of uncertainty, i.e. measurement and sampling
203 uncertainty when measuring plant cover, nitrogen deposition, soil pH, soil C-N ratio and soil type, as well as
204 the structural uncertainties due to the modelled soil pH, large-scale (among sites) spatial variation and the
205 temporal processes, were modelled explicitly. The most important source of measurement uncertainty was
206 the plant cover measurement due to the significant small-scale spatial aggregation of plant species, which
207 was modelled by the parameter δ in the Dirichlet - multinomial mixture distribution. The median estimated
208 value of δ was 0.188 with a relatively narrow credible interval (Table S1). Generally, the estimates of
209 structural uncertainties were relatively low, although the large-scale spatial variation was relatively high
210 (Table S1, Fig. 3A).

211 Most of the regression parameters that measure the large-scale spatial and temporal effect of the abiotic
212 variables on the vegetation were significantly different from zero (Table S1), suggesting that the modelled
213 environmental and land-use factors have a regulating effect on both the large-scale spatial variation in
214 cover of the six species classes as well as plant community dynamics in dune heathlands. All explanatory
215 variables had significant effects on the large-scale spatial variation of five or all six species classes (Fig. 4A,
216 Table S1) and, there was significant geographic variation among the four assigned Danish regions, where
217 the large-scale cover of other herbs and cryptogams showed a qualitatively different geographic
218 distribution than the other species (Fig. 4A, Table S1), but without showing a clear pattern (results not
219 shown).

220 The selected environmental drivers, all had significant effects on the temporal variation of all six species
221 classes, except for nitrogen deposition on other graminoids (Fig. 4B, Table S1). Notably, relatively high
222 nitrogen deposition was associated with an observed decrease in the cover of *C. vulgaris* and *A. flexuosa*,
223 and an increase in the cover of *E. nigrum*. Relatively high soil pH was associated with an observed increase
224 in *A. flexuosa* and a decrease in cryptogams. Relatively high C/N ratio in the soil was associated with an
225 observed increase in the cover of *C. vulgaris*, *E. nigrum*, and *A. flexuosa*, and a decrease in the cover of

cryptogams. Grazing was associated with an increase in the cover of *C. vulgaris*. Sandy soil was associated with an observed increase in the cover of *C. vulgaris* and cryptogams. Finally, relatively high precipitation was associated with an increase in the cover of *E. nigrum* and a decrease in the cover of cryptogams (Fig. 4B).

In this study of dune heathlands, relatively high nitrogen deposition was associated with low soil pH (γ_N , Table S1), and soil pH was found to be significantly lower on more clayey soils compared to sandy soils (γ_S , Table S1).

Discussion

Plant community dynamics and environmental drivers

The species composition of Danish dune heathlands has changed since 2004. Several dwarf shrubs and wetland-associated species, such as *Carex nigra* and *Salix repens*, showed a decline in cover, while *Cladonia* species increased in cover. *C. vulgaris* and graminoids remained relatively stable throughout the monitoring period.

Most of the regression parameters that measure the effect of the environmental drivers on the change in plant species cover were significantly different from zero. These results suggest that the selected modelled environmental variables have regulating effects on the observed large-scale spatial variation as well as plant community dynamics in dune heathlands.

Overall, precipitation and soil pH emerged as the most important drivers of dune heathland vegetation composition, highlighting the importance of hydrological and edaphic conditions in shaping species distributions. Precipitation showed the strongest overall effect, positively associated with *E. nigrum* while

246 reducing cryptogams. Soil pH was another key driver, promoting *A. flexuosa* but negatively affecting other
 247 herbaceous species, indicating a shift toward acid-tolerant graminoids under lower pH conditions.

248 The annual precipitation in the future climate in Denmark is predicted to increase, but with decreasing
 249 summer precipitation and longer summer drought periods (DMI 2017). Generally, it is uncertain how this
 250 combination of more extreme weather will influence the future dune heathland vegetation (Olmeda et al.
 251 2020). The study demonstrated significant acidifying effects of nitrogen deposition and soil pH was found to
 252 be significantly higher on sandy soils compared to less sandy soils. The soil acidification effects of nitrogen
 253 deposition are due either to nitrate leaching or removal of base cations from the system by nature
 254 management (Williams and Anderson 1999), and the detrimental effects of this acidification, e.g. reduced
 255 biodiversity of higher plants, have clearly been demonstrated at heathlands in the Netherlands (Bobbink et
 256 al. 2022; Vogels et al. 2020), which generally have received higher nitrogen deposition than Danish
 257 heathlands.

258 The C/N ratio exhibited moderate positive effects on *C. vulgaris*, *E. nigrum*, and graminoids, whereas
 259 cryptogams responded slightly negatively. Grazing intensity produced mixed responses: graminoids and
 260 other herbs showed slight positive associations, while cryptogams and *C. vulgaris* declined. Soil type effects
 261 were generally small, where clayey soils tended to favor graminoids and herbs. Nitrogen deposition had
 262 only minor positive coefficients across most groups, suggesting limited direct influence compared to other
 263 environmental gradients.

264 **Spatial variation**

265 The fine-scale spatial aggregation of the three focal species, along with the three aggregated classes of
 266 other plant species, was modeled using the parameter δ within the Dirichlet-multinomial mixture
 267 distribution framework (see Appendix A). The estimated degree of spatial aggregation at this scale
 268 significantly increased the uncertainty in the expected cover measurements compared to a scenario where

269 plant species are randomly distributed. If this overdispersion in the pin-point cover data is not accounted
270 for in the statistical model, the signal-to-noise ratio will be substantially inflated, which is likely to result in
271 misleading conclusions (Damgaard 2013).

272 All species classes exhibited significant regional variation across the four designated geographic regions
273 (Fig. 1). This finding suggests that part of the large-scale spatial residual variation, which is not explained by
274 the modeled environmental drivers, may be attributed to yet unidentified factors that differ among these
275 regions. Future research should aim to uncover the historical causes underlying this spatial variation in
276 species abundance, for instance, by compiling and analyzing detailed records of site-specific nature
277 management practices and hydrological conditions.

278 **Uncertainties and application of the model**

279 In general, structural equation modeling (SEM) does not provide definitive proof of hypothesized causal
280 relationships. However, it does allow for testing whether specific causal pathways are supported by the
281 data. Demonstrating true causality requires system manipulation to observe whether the responses
282 predicted by the SEM actually occur (Granger 1969; Pearl 2009). Additionally, there is often an unknown
283 time lag between changes in environmental variables and their effects on vegetation cover (e.g. Svenning
284 and Sandel 2013), making it uncertain which temporal window of environmental change should be used in
285 the model. Nevertheless, due to the typically high degree of temporal spatial autocorrelation, where sites
286 with relatively high values of an environmental variable at one time tend to maintain those values over
287 subsequent periods, the model can still reliably estimate the relative effects of environmental variables
288 across sites and years.

289 The statistical modeling uncertainty was decomposed into measurement uncertainty and uncertainties
290 associated with the modeled spatial and temporal processes. The primary source of measurement
291 uncertainty stemmed from plant cover estimates, largely due to pronounced small-scale spatial aggregation

292 of plant species (see above). However, uncertainties in nitrogen deposition, soil pH, soil C–N ratio, and soil
293 type were also quantified and incorporated into the model. Overall, the structural uncertainty linked to
294 temporal processes was relatively minor, whereas the large-scale spatial processes introduced a greater
295 degree of structural uncertainty.

296 One key advantage of partitioning different sources of uncertainty within the structural equation model
297 (SEM) is its utility for predictive purposes (Damgaard 2022b; Damgaard 2025c). Given the excellent fit of
298 the temporal model, it is suggested that the included environmental variables are sufficient for predicting
299 the average dynamics of plant communities in dune heathlands. This optimistic conclusion is somewhat
300 unexpected, considering the absence of several potentially influential regulatory factors such as historical
301 nature management interventions. In the absence of a well-established causal framework, caution is
302 warranted when applying the fitted model to generate site-specific ecological predictions for adaptive
303 management planning. The environmental variables used in this study may be correlated with unknown
304 causal drivers, rare but impactful contingent events, or relevant factors for which data are unavailable.
305 Nevertheless, the modeling results offer valuable insights for site managers regarding the relative
306 importance of various environmental drivers and management scenarios (Fig. 5). For instance, because the
307 direct and indirect effects of nitrogen deposition are not readily observable, their impact is difficult to
308 assess without a statistical model and is often underestimated by site managers.

309

310 Tables

311 Table 1: Maximum likelihood estimates of mean site cover for the most common species in Danish dune
 312 heathland sites. The cover is summarized by the maximum likelihood estimates of mean site cover (μ) and
 313 the beta distribution shape parameter (ν) (Appendix C). The estimated trends in cover in the period 2004 -
 314 2021 were calculated using a Bayesian framework using three linear models (Appendix C). The 2.5%, 50%,
 315 and 97.5% percentiles of the marginal posterior probability distribution of average yearly change (on a logit
 316 scale) in the model that was best supported by the data are reported (significant changes are shown in
 317 bold).

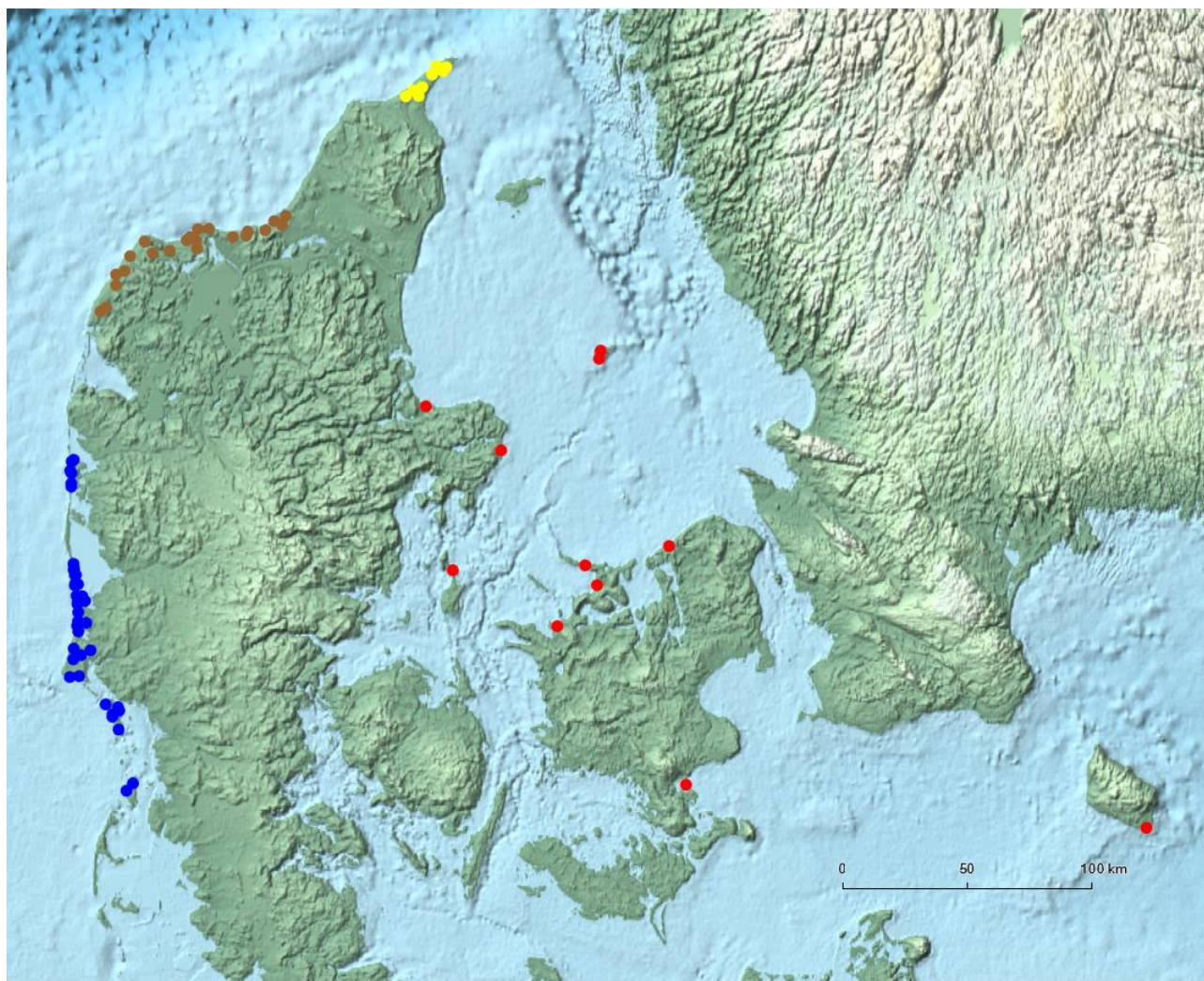
Species	Cover		Trends in cover			Model
	μ	ν	2.50%	50%	97.50%	
<i>Empetrum nigrum</i>	0.297	1.17	-0.064	-0.043	-0.023	3
<i>Calluna vulgaris</i>	0.278	1.22	-0.043	-0.012	0.015	3
<i>Avenella flexuosa</i>	0.168	1.29	-0.009	0.011	0.03	3
<i>Carex arenaria</i>	0.148	2.59	-0.03	-0.013	0.003	3
<i>Vaccinium uliginosum</i>	0.063	1.17	-0.035	-0.009	0.018	3
<i>Cladonia sp.</i>	0.056	2.01	0.003	0.037	0.065	2
<i>Molinia caerulea</i>	0.054	0.58	-0.06	-0.025	0.011	3
<i>Ammophila arenaria</i>	0.047	1.62	-0.009	0.009	0.028	1
<i>Erica tetralix</i>	0.045	1.93	-0.072	-0.045	-0.018	3
<i>Myrica gale</i>	0.03	0.71	-0.12	-0.032	0.038	3
<i>Salix repens</i>	0.026	2.29	-0.139	-0.081	-0.033	3
<i>Carex nigra</i>	0.022	2.31	-0.252	-0.155	-0.081	3

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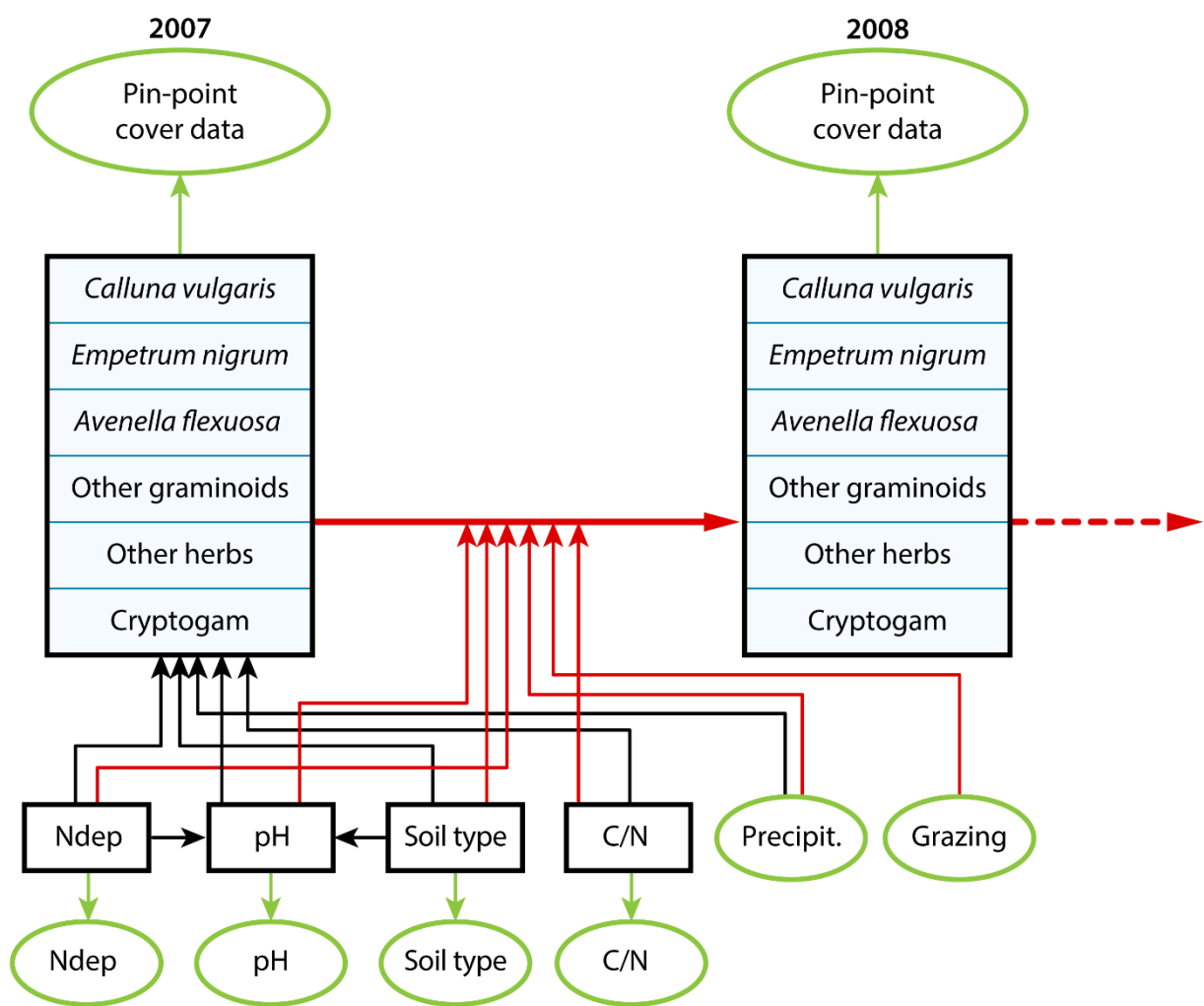
320 **Figures**

321 Fig. 1. Map of the selected 81 Danish dune heathland sites. The different colors represent a classification of
322 the different sites into four geographical regions.



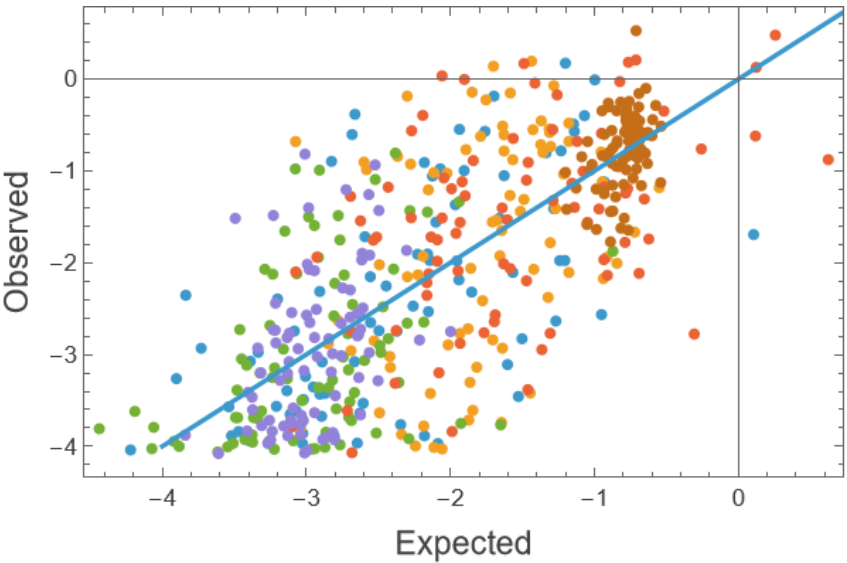
323

324 Fig. 2. Outline of hierarchical SEM. The spatial variation in vegetation cover in 2007 is modelled by nitrogen
 325 deposition (Ndep), soil pH (pH), soil C-N ratio (C/N), soil type and precipitation (Precipit.). The yearly change
 326 in vegetation cover from 2007 to 2022 (only a single yearly change is shown in the figure) is modelled by all
 327 the former variables as well as grazing. The black boxes are latent variables and the green ovals are data.
 328 The black arrows denote large-scale spatial processes, and the red arrows denote temporal processes
 329 (Appendix B).



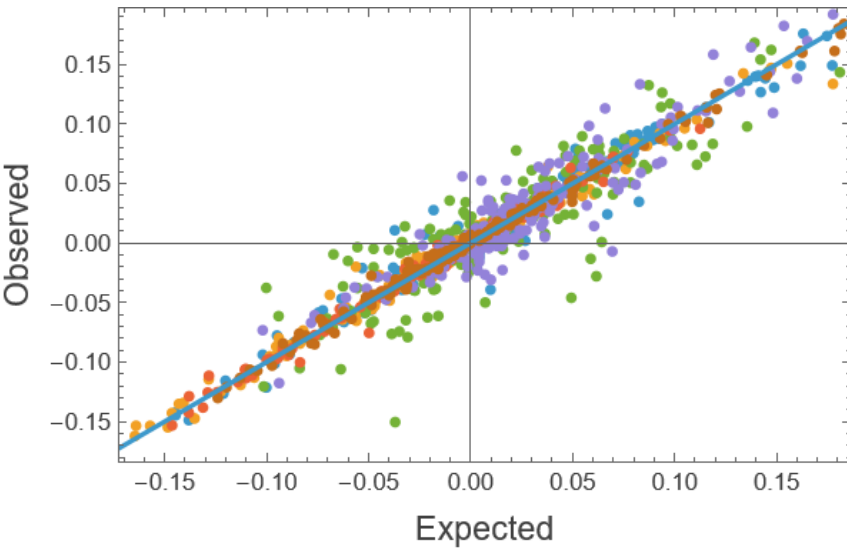
331 Fig 3. Plots of observed vs. expected logit-transformed cover of the large-scale spatial process (A) and the
 332 temporal (B). The proportions of the variance explained for each species may be found in Table S3. Blue:
 333 *Calluna vulgaris*, yellow: *Empetrum nigra*, green: *Avenella flexuosa*, red: other graminoids, purple: other
 334 herbs, blue: cryptogams.

335 A: large-scale spatial process



336

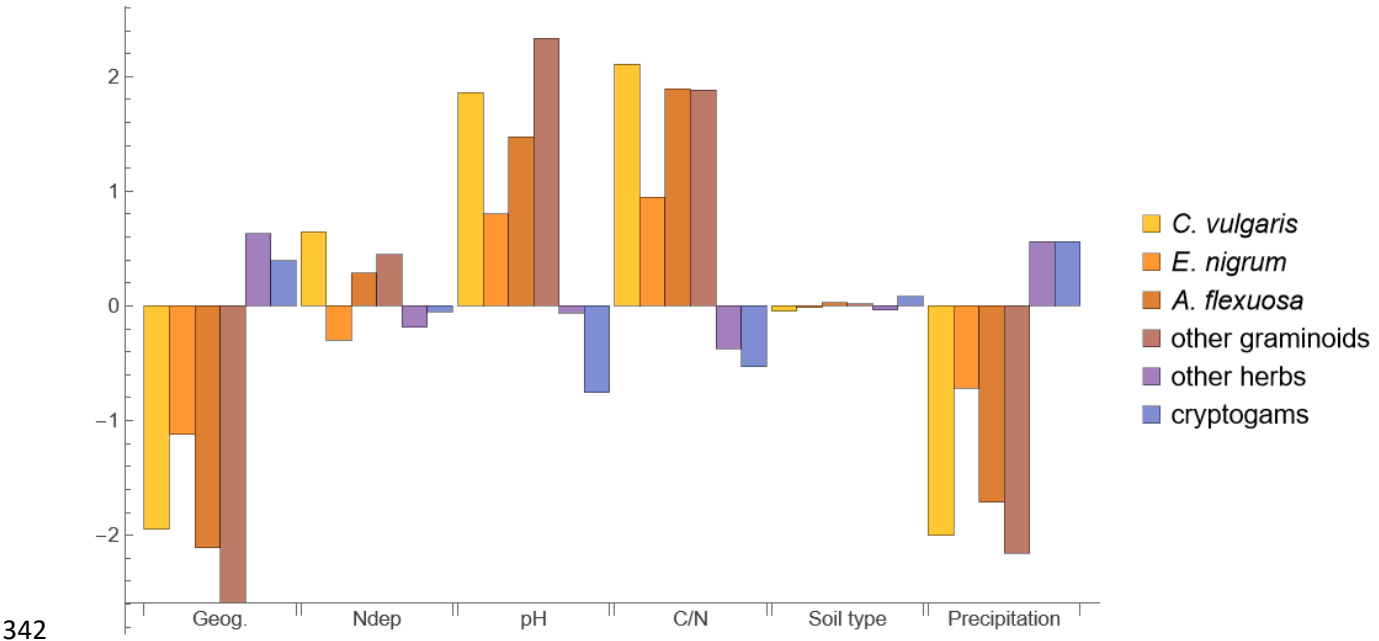
337 B: temporal process



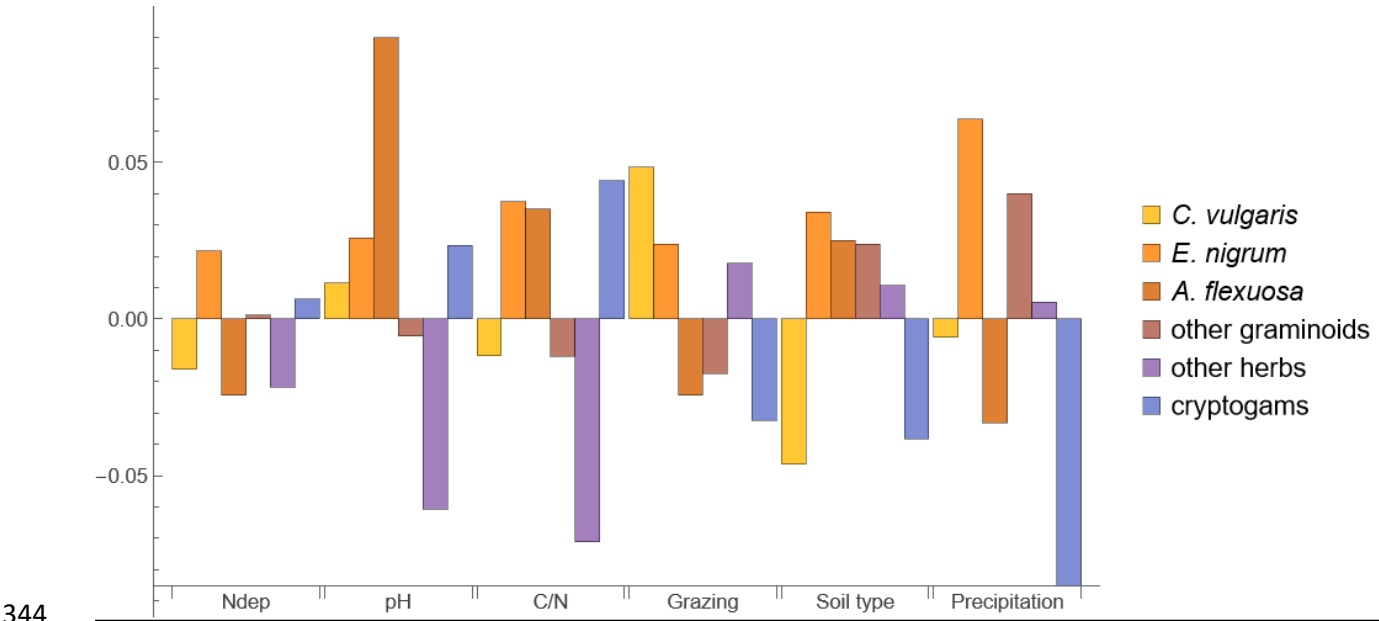
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339 Fig 4. Standardized regression coefficients of the SEM for the large-scale spatial process (A) and the
 340 temporal (B).

341 A: large-scale spatial process

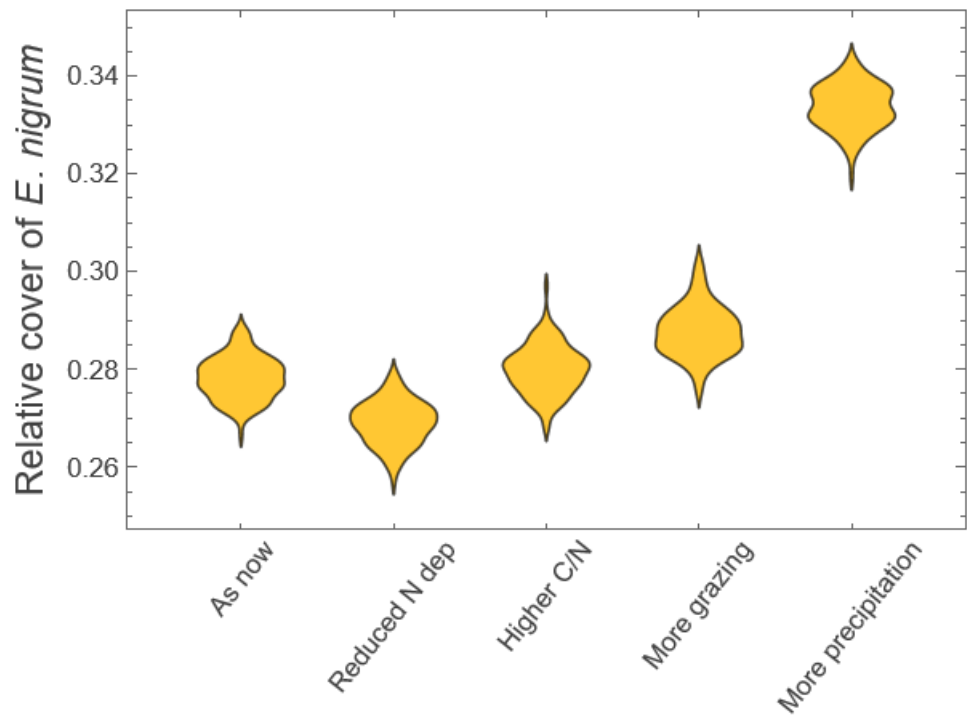


343 B: temporal process

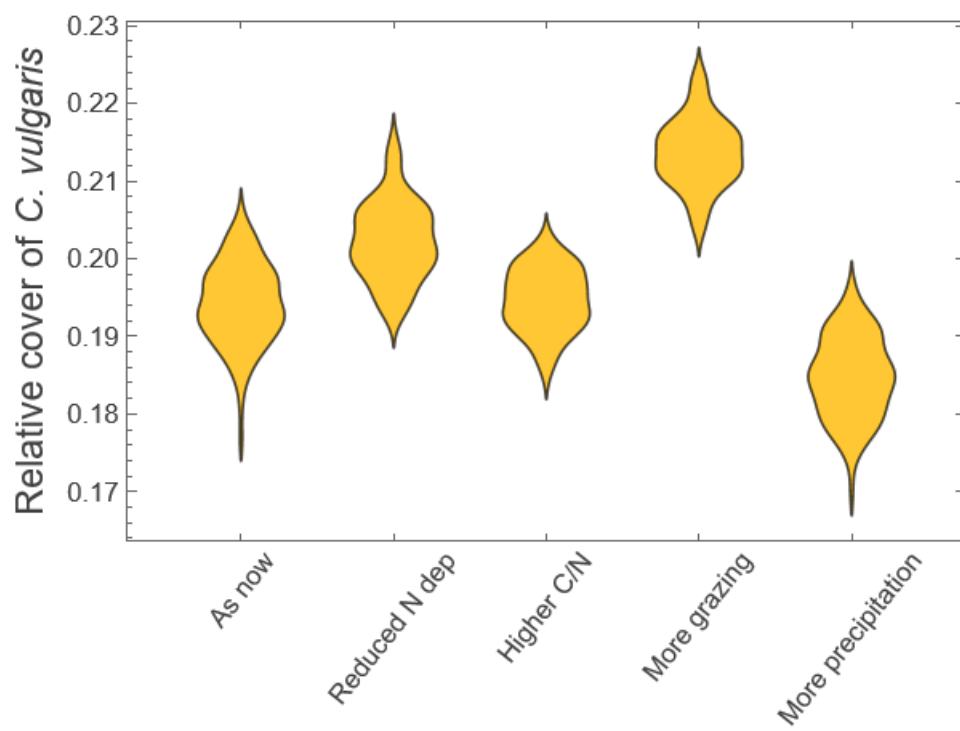


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346 Fig. 5. Predicted distribution of the cover of *E. nigrum* and *C. vulgaris* for a specific dune heathland site
 347 (Husby klit in West Jutland) after five years under four different scenarios (Damgaard 2022a; 2025c). The
 348 initial cover of *E. nigrum* and *C. vulgaris* was 0.30 and 0.18, respectively. The scenarios were 1: As now, 2:
 349 reduced N deposition, from 12.2 kg N ha⁻¹ year⁻¹ to 8 kg N ha⁻¹ year⁻¹ (note that it will require intensive
 350 mangement actions to reduce plant available N in the soil to the level expected at equilibrium under the
 351 reduced N deposition scenario), 3: Higher C/N ratio in soil, from 19.7 to 40, 4: more grazing 0 to 0.5
 352 (average of a binary variable where 0 is no grazing) 5: more precipitation, from 827 mm year⁻¹ to 1100 mm
 353 year⁻¹. The other environmental variables were pH in soil: 3.66, soil texture (jb nr.): 1 (coarse sandy soil).



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357 **References**

- 358 Aerts R, Berendse F, Caluwe Hd, Schmitz M. 1990. Competition in heathland along an experimental gradient
359 of nutrient availability. *Oikos*. 57:310-318.
- 360 Aerts R, Heil GW. 1993. *Heathlands. Patterns and processes in a changing environment.*: Kluwer Academic
361 Publishers.
- 362 Bobbink R, Loran C, Tomassen H. 2022. Review and revision of empirical critical loads of nitrogen for
363 europe. German Environment Agency.
- 364 Britton A, Marrs R, Pakeman R, Carey P. 2003. The influence of soil-type, drought and nitrogen addition on
365 interactions between *calluna vulgaris* and *deschampsia flexuosa*: Implications for heathland
366 regeneration. *Plant Ecol*. 166(1):93-105.
- 367 Damgaard C. 2013. Hierarchical and spatially aggregated plant cover data. *Ecol Inform*. 18:35-39.
- 368 Damgaard C. 2015. Modelling pin-point cover data of complementary vegetation classes. *Ecol Inform*.
369 30:179-184.
- 370 Damgaard C. 2019. Spatio-temporal structural equation modeling in a hierarchical bayesian framework:
371 What controls wet heathland vegetation? *Ecosystems*. 22:152-164.
- 372 Damgaard C. 2020. Measurement uncertainty in ecological and environmental models. *Trends Ecol Evol*.
373 35:871-873.
- 374 Damgaard C. 2022a. Adaptive management plans rooted in quantitative ecological predictions of
375 ecosystem processes: Putting monitoring data to practical use. *Environmental Conservation*. 49:27-
376 32.
- 377 Damgaard C. 2022b. Processes and predictions in plant ecological models: Logic and causality. *EcoEvoRxiv*.
- 378 Damgaard C. 2022c. Spatio-temporal modelling of the effect of selected environmental and land-use
379 factors on acid grassland vegetation. *Journal of Plant Ecology*. 15(2):253-264.
- 380 Damgaard C. 2023. Spatio-temporal modelling of the effect of environmental and land-use factors on
381 species-rich calcareous grasslands. *Basic and Applied Ecology*. 72:22-29.
- 382 Damgaard C. 2025a. Ecosystem dynamics in dry heathlands: Spatial and temporal effects of environmental
383 drivers on the vegetation.
- 384 Damgaard C. 2025b. Ecosystem dynamics in wet heathlands: Spatial and temporal effects of environmental
385 drivers on the vegetation. *Rangeland Ecology & Management*. 100:47-55.
- 386 Damgaard C. 2025c. Local ecological predictions as input to adaptive management of natural plant
387 communities. *Biol Conserv*. 302:110951.
- 388 Damgaard C. 2025d. Observed vegetation changes in danish dry heathlands since 2004. *Flora*. 327:152728.
- 389 Damgaard C. 2025e. Processes and predictions in ecological models: Logic and causality. *Journal of*
390 *Forecasting*. 44(5):1658-1665.
- 391 Damgaard C, Bak JL, Strandberg M, Hansen RR. 2024. The resilience of heathland ecosystems: A working
392 hypothesis. *Acta Oecologica*. 125:104037.
- 393 Damgaard C, Hansen RR, Hui FKC. 2020. Model-based ordination of pin-point cover data: Effect of
394 management on dry heathland. *Ecol Inform*. 60:101155.
- 395 Damgaard C, Irvine KM. 2019. Using the beta distribution to analyze plant cover data. *J Ecol*. 107:2747–
396 2759.
- 397 Damgaard C, Strandberg MT, Kristiansen SM, Nielsen KE, Bak JL. 2014. Is *erica tetralix* abundance on wet
398 heathlands controlled by nitrogen deposition or soil acidification? *Environmental Pollution*. 184:1-
399 8.
- 400 Damgaard C, Weiner J. 2021. The need for alternative plant species interaction models. *Journal of Plant*
401 *Ecology*. 14:771-780.
- 402 De Graaf MCC, Bobbink R, Smits NAC, Van Diggelen R, Roelofs JGM. 2009. Biodiversity, vegetation gradients
403 and key biogeochemical processes in the heathland landscape. *Biol Conserv*. 142(10):2191-2201.

404 DMI. 2014. Average annual precipitation in the period 2001 to 2010 with a spatial resolution of 10 km.
 405 Copenhagen: Danmarks Meteorologiske Institut.
 406 Fremtidens klima i danmark. 2017. København: Danmarks Meteorologiske Institut; [accessed].
 407 <https://www.dmi.dk/klima/fremtidens-klima/danmark/>.
 408 Ellermann T, Andersen HV, Bossi B, Christensen J, Løfstrøm P, Monies C, Grundahl L, Geels C. 2012.
 409 Atmosfærisk deposition 2011 – novana. Aarhus: Nationalt Center for Miljø og Energi.
 410 Ellermann T, Nygaard J, Christensen JH, Løfstrøm P, Geels C, Nielsen IE, Poulsen MB, Monies C,
 411 Gyldenkerne S, Brandt J et al. 2018. Nitrogen deposition on danish nature. Atmosphere. 9(11).
 412 EU. 1992. Council directive 92/43/EEC of 21 May 1992 on the conservation of natural habitats and of wild
 413 fauna and flora. In: Commission E, editor.
 414 EU. 2013. Interpretation manual of European Union habitats. Bruxelles: European Commission, DG
 415 Environment, Nature and Biodiversity.
 416 Fredshavn J, Nygaard B, Ejrnæs R, Johansson LS, Dahl K, Christensen JPA, Kjær C, Elmeros M, Mortensen
 417 RM, Møller JD et al. 2025. Bevaringsstatus for naturtyper og arter – 2025. Habitatdirektivets artikel
 418 17-rapportering. Aarhus Universitet, DCE – Nationalt Center for Miljø og Energi.
 419 Gimingham C. 1978. Calluna and its associated species: Some aspects of co-existence in communities. Plant
 420 Ecol. 36(3):179-186.
 421 Gimingham CH. 1960. Biological flora of the British Isles. No. 74. Calluna vulgaris (L.) Hull. J Ecol. 48:455-483.
 422 Gimingham CH. 1988. A reappraisal of cyclical processes in Calluna heath. Vegetatio. 77(1/3):61-64.
 423 Gimingham CH, Hobbs RJ, Mallik AU. 1981. Community dynamics in relation to management of heathland
 424 vegetation in Scotland. Vegetatio. 46(1):149-155.
 425 Grace JB, Anderson TM, Olff H, Scheiner SM. 2010. On the specification of structural equation models for
 426 ecological systems. Ecological Monographs. 80:67–87.
 427 Granger CWJ. 1969. Investigating causal relations by econometric models and cross-spectral methods.
 428 Econometrica. 37(3):424-438.
 429 Greve MH, Greve MB, Bøcher PK, Balstrøm T, Breuning-Madsen H, Krogh L. 2007. Generating a Danish
 430 raster-based topsoil property map combining choropleth maps and point information. Danish
 431 Journal of Geography. 107:1-12.
 432 Levy EB, Madden EA. 1933. The point method of pasture analyses. New Zealand Journal of Agriculture.
 433 46:267-279.
 434 Lindquist B. 1931. Den skandinaviska bokskogens biologi. Svenska Skogsvårdsföreningens Tidskrift. 3:179-
 435 485.
 436 Løvschal M, Damgaard CF. 2022. Mapping the ecological resilience of Atlantic postglacial heathlands. J Appl
 437 Ecol. n/a(n/a).
 438 Newton AC, Stewart GB, Myers G, Diaz A, Lake S, Bullock JM, Pullin AS. 2009. Impacts of grazing on lowland
 439 heathland in north-west Europe. Biol Conserv. 142:935-947.
 440 Nielsen KE. 2005. Testing favourable conservation status for dune heathlands, Denmark. National
 441 Environmental Research Institute, Denmark.
 442 Nielsen KE, Bak JL, Bruus M, Damgaard C, Ejrnæs R, Fredshavn JR, Nygaard B, Skov F, Strandberg B,
 443 Strandberg M. 2012. Naturdata.Dk - Danish monitoring program of vegetation and chemical plant
 444 and soil data from non-forested terrestrial habitat types. Biodiversity & Ecology. 4:375.
 445 Kontrol- og overvågning af terrestriske habitatnaturtyper 2004 – 2022. Novana. 2024. Aarhus Universitet: DCE –
 446 Nationalt Center for Miljø og Energi; [accessed]. www.novana.au.dk.
 447 Nygaard B, Damgaard C, Bladt J, Ejrnæs R. 2020. Fagligt grundlag for vurdering af bevaringsstatus for
 448 terrestriske naturtyper. Artikel 17-rapporteringen 2019. Aarhus Universitet, DCE – Nationalt Center
 449 for Miljø og Energi.
 450 Olmeda C, ŠeffEROVÁ V, Underwood E, Millan L, Gil T, Naumann S. 2020. EU action plan to maintain and
 451 restore to favourable conservation status the habitat type 4030 European dry heaths. European
 452 Commission.

453 Ovaskainen O, Roy DB, Fox R, Anderson BJ. 2016. Uncovering hidden spatial structure in species
 454 communities with spatially explicit joint species distribution models. *Methods in Ecology and*
 455 *Evolution*. 7(4):428-436.

456 Pearl J. 2009. *Causality. Models reasoning, and inferences*. 2, editor. Cambridge: Cambridge University
 457 Press.

458 Strandberg M, Nielsen KE, Damgaard C. 2018. Habitat monitoring reveals decreasing morlayer c:N ratios in
 459 danish heathlands. *Ecological Indicators*. 89:538-542.

460 Svenning J-C, Sandel B. 2013. Disequilibrium vegetation dynamics under future climate change. *Am J Bot*.
 461 100:1–21.

462 Tybirk K, Nilsson M-C, Michelsen A, Kristensen HL, Shevtsova A, Tune Strandberg M, Johansson M, Nielsen
 463 KE, Riis-Nielsen T, Strandberg B et al. 2000. Nordic *empetrum* dominated ecosystems:
 464 Function and susceptibility to environmental changes. *AMBIO: A Journal of the Human*
 465 *Environment*. 29(2):90-97, 98.

466 Usher MB, Thompson DBA. 1993. Variation in the upland heathlands of great britain: Conservation
 467 importance. *Biol Conserv*. 66(1):69-81.

468 Vogels JJ, Weijters MJ, Bobbink R, Bijlsma R-J, Lamers LPM, Verberk WCEP, Siepel H. 2020. Barriers to
 469 restoration: Soil acidity and phosphorus limitation constrain recovery of heathland plant
 470 communities after sod cutting. *Applied Vegetation Science*. 23(1):94-106.

471 Watt AS. 1947. Pattern and process in the plant community. *J Ecol*. 35:1-22.

472 Williams BL, Anderson HA. 1999. The role of plant and soil processes in determining the fate of atmospheric
 473 nitrogen. In: Langan SJ, editor. *The impact of nitrogen deposition on natural and semi-natural*
 474 *ecosystems*. Dordrecht: Kluwer.

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477 **Electronic supplements**

478 *The following electronic supplements may be downloaded from <https://osf.io/ckhqy/overview>*

479 Table S1 - Marginal distribution of parameters

480 Table S2 – Proportion of variance explained

481 Fig. S1 – Pairwise scatter plot of variables

482 Fig. S2 - Dunn–Smyth residuals

483 Appendix A - Distribution of pin-point plant cover data

484 Appendix B - Spatio-temporal modelling

485 Appendix C - Estimating species cover and change in species cover

486 Appendix D – Mathematica notebook (a free reader may be downloaded from

487 <https://www.wolfram.com/player/>)

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