

1 **Bimodal seasonal activity of moths and elevation, weather and land use as**
2 **drivers of their diversity**

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16

17 **ABSTRACT**

18 Moths are an important part of terrestrial insect diversity and contribute substantially to eco-
19 system functioning. Yet, how their activity varies with the season and how different biotic
20 and abiotic factors (elevation, weather, land use) are simultaneously linked to moth commu-
21 nity characteristics are still poorly understood.

22 We analysed a vast moth community dataset from Switzerland collected by a single expert
23 across 50 years containing data of 2.8 Mio moth individuals (1,045 species), covering the
24 whole yearly cycle. Using regression models, we related moth community characteristics (to-
25 tal abundance, species richness, biomass) to season, elevation, weather and land use (land-
26 scape composition).

27 Moth community characteristics showed a clear bimodal seasonal cycle with an activity peak
28 in early spring and one in summer. The different peaks could be clearly linked to moth spe-
29 cies with different overwintering stages, i.e. the spring peak was driven by species overwin-
30 tering as pupae or adults. Along the elevational gradient, we found increases of all moth com-
31 munity characteristics, levelling of at around 2000 m asl. Also, moth activity increased sig-
32 nificantly with increasing temperatures and was higher in landscapes with higher proportions
33 of forests.

34 Based on a moth dataset of unseen extent, we present a well resolved seasonal activity pattern
35 and quantify the role of elevation, landscape composition (forests) and weather (temperature)
36 in driving moth community characteristics. These results will help to better understand varia-
37 tion in moth activity across different temporal and spatial scales and to design targeted con-
38 servation efforts, e.g. in lower elevation sites.

39 INTRODUCTION

40 Moths make up the largest share of the species-rich insect group of lepidopterans, but have
41 been receiving less research interest compared to the popular butterflies (New, 2004). This is
42 despite their important role in ecosystems such as pollinators (Knop et al., 2017; Walton et
43 al., 2020) or food source for bats (Vaughan, 1997) and birds (Evans et al., 2024). Moth com-
44 munities cover species with various ecological strategies, resulting in community changes
45 along biotic and abiotic gradients. Also, moths are susceptible to various global change driv-
46 ers such land use intensification (Mangels et al., 2017), climate change (Hunter et al., 2014)
47 or light pollution (Knop et al., 2017; van Grunsven et al., 2020). The latter is related to their
48 nocturnal activity and their attraction to artificial light sources, which allows to quantify and
49 characterise activity of nocturnal moths using light trap sampling (Jonason et al., 2014). Yet,
50 how moth activity varies over the season and which parameters drive their abundance, spe-
51 cies richness and biomass is still poorly understood. Here, we studied how these moth com-
52 munity characteristics change over the course of the year and how they are related to different
53 biotic and abiotic drivers (elevation, weather, land use).

54 In temperate regions, insects have distinct seasonal cycles, but these cycles have been insuffi-
55 ciently quantified because observational studies, including studies on moths, often have tem-
56 porally limited sampling schemes. Consequently, often a unimodal activity pattern is as-
57 sumed, with one single activity peak of adults during the vegetative period, similar to diel ac-
58 tivity peaks. However, recent evidence on different insect groups indicates that neither diel
59 (Knop et al., 2023; Xu et al., 2021; Zoller et al., 2020) nor seasonal activity (Mellard et al.,
60 2019) show unimodal patterns by default. In the case of moths, it is also highly unlikely that
61 there is only a single activity peak of adults during the vegetative period, given their high
62 species richness and their different overwintering strategies, with species overwintering as

63 egg, larvae, pupae, and adults. However, except for a few studies that analysed the intra-an-
64 nual variability of moth activity over part of the seasonal cycle (Busse et al., 2022; Jonason et
65 al., 2014; Roth et al., 2021) and indicate variability in activity peaks, a comprehensive analy-
66 sis of moth activity peaks over the entire year is missing so far, most likely due to the lack of
67 suitable data. Insufficient understanding of intra-annual variation of insect communities is
68 problematic (Montgomery et al., 2021), as important population changes of species due to
69 global change might be missed, namely of those active early or late in the season, resulting in
70 biased conclusions on susceptibility of moth communities to global change and its conse-
71 quences for the provision of ecosystem services.

72 Besides changes in moth community characteristics across the season, moth communities are
73 driven by various biotic and abiotic factors. These factors might vary spatially between sam-
74 pling sites, but also temporally between days of sampling. For example, the proportion of
75 suitable habitats in a landscape surrounding a sampling site (e.g. forests) is often positively
76 linked to moth abundance and species richness (e.g. Kühne et al., 2022). Also, it has been
77 found that elevation is a driver of moth communities, with highest richness at mid-elevation
78 sites (Beck et al., 2017). Furthermore, weather conditions (temperature, precipitation) are
79 crucial factors linked to moth activity, which vary on a daily basis and result in more individ-
80 uals and species of moths being caught during nights in which temperature is higher (Beck et
81 al., 2010; Jonason et al., 2014; Knop et al., 2018). However, large-scale studies on how moth
82 communities are driven by these abiotic and biotic factors are still rare, and their relative im-
83 portance is so far not well resolved. Better knowledge on the most important drivers of moth
84 community characteristics is however crucial to design specific and efficient conservation
85 measures.

86 In this study, we analyzed a vast dataset collected by a single expert across 50 years and
87 throughout the seasons in Switzerland, spreading across a large elevational gradient. Based

88 on macro-moth abundance, species richness and estimated biomass from this dataset, we ad-
89 dressed the following research questions:

- 90 1. How do moth abundance, richness and biomass change with seasons? How are these
91 patterns influenced by species with different overwintering stages (egg, larva, pupa,
92 adult)?
- 93 2. How do moth abundance, richness and biomass vary with weather conditions (tempera-
94 ture, precipitation)?
- 95 3. How do moth abundance, richness and biomass change along an elevation gradient?
- 96 4. How does landscape composition relate to moth abundance, richness and biomass?

97

98 **METHODS**

99 **Moths dataset**

100 The data were collected with light traps across a 50-year timespan by one of the main experts
101 of the group in Switzerland, Dr. Ladislaus Rezbanyai-Reser (ZOBODAT, 2014–2023), with
102 the purpose to describe the macro-moth communities (including abundance distribution and
103 phenology) at different sites (cf. Rezbanyai-Reser, 2018b). The dataset, which is hosted by
104 *info fauna* (The Swiss Topic Center on Fauna), contains species-level abundance data of
105 macro-moths (Table S1.1 in Supporting Information). The light trap samples allow character-
106 isation of local nocturnal moth communities (Truxa & Fiedler, 2013). All attracted moths
107 were collected and killed with a sampling fluid (mostly chloroform). Light traps were oper-
108 ated in two different ways (Fig. S1.1): Fixed traps, which were installed for a long period
109 (mostly several months) and emptied daily over the whole period in which they were active,
110 and manual traps, which were installed only in single selected nights and which were only ac-
111 tive for some hours (sampling duration of 1–13 hours). For fixed trapping, there was a com-
112 monly used trap model (“type 1”, used in 154 site–year combinations) and a more rarely used
113 trap model (“type 2”, used in 10 site–year combinations). The number of traps installed at a
114 site and the lamps used differed between sampling procedures, sites, and years. The number
115 of traps varied from one trap up to four simultaneously active traps at a single site (Fig. S1.2),
116 but moth data were only available for the pooled samples of all active traps. Three lamp types
117 can be distinguished, which were used differently often: 150–160W mercury mixed-light
118 lamps (150–160W HWL), 80W mercury mixed-light lamps (80W HWL) and 125W mercury
119 vapour lamps (125W HQL) (Fig. S1.3). When several traps were active, several lamp types
120 might have been used, resulting in a joint category 150–160W HWL/125W HQL.

121 The dataset holds data from 171 sites spread across an elevational gradient between 193m asl.
122 and 2454m asl. in Switzerland (Fig. S1.4) and from 49 years (1972–2021; no samples in
123 2019), entailing 663 unique site and year combinations. A total of 37,461 nights were moni-
124 tored, in which 2,814,187 individuals of 1,045 nocturnal macro-moth species (Table S1.1)
125 were caught (day active species that were accidentally collected were excluded from anal-
126 yses). Some sites with manual traps were in proximity and normally ran simultaneously in the
127 same nights (34 site groups including 94 sites, distances between simultaneously active sites
128 ranging from 38m to 6.1km). These sites were still treated as separate sites, but their group-
129 ing was accounted for in the statistical models. For fixed traps, we assumed that stretches of
130 10 or more nights without a single record were due to inactivity of a trap (e.g., due to mis-
131 functioning) and we excluded them from analyses (1614 nights, leaving 35,847 nights). Fixed
132 traps were on average active for 194 nights per year (range: 5–362 nights) with July 24 being
133 the mean sampling day of the year (range: Feb 3 – Aug 28); manual traps in a site were on
134 average active for 8 nights per year (range: 1–71 nights) with July 16 being the mean sam-
135 pling day of the year (range: Feb 3 – Nov 11) (Fig. S1.5). To test how weather conditions
136 during sampling affect moth activity and thus sampling numbers, we determined the tempera-
137 ture and precipitation for each sampling night. We used gridded daily temperature and pre-
138 cipitation data (1.25 degree minute grid; approx. 2.3 km × 1.6 km) provided by MeteoSwiss
139 (<https://www.meteoswiss.admin.ch>) and used the values of the closest grid cell. Mean tem-
140 perature and total precipitation of the two sampling days enclosing the sampling night were
141 used (2 days average/total).

142 **Landscape composition**

143 Landscape composition at the study sites was determined based on aerial photo interpretation
144 for the whole of Switzerland at four time steps (1979–1985, 1990–1998, 2004–2009, 2012–

145 2019), which attributed each hectare a categorical land-use and land-cover value (“Are-
146 alstatistik”, Bundesamt für Statistik, 2021). We determined the cover of four broad land-use
147 types (forests, grasslands, croplands, sealed area) (Table S1.2) in the surroundings of the
148 study plots. For forest and grassland cover, we expected positive relations to moth commu-
149 nity characteristics, as they may provide valuable habitats (Archaux et al., 2018). For
150 cropland cover, we expected negative relations to community characteristics as the barely
151 provide habitats and might even be detrimental e.g. due to higher amounts pesticide being
152 present (Archaux et al., 2018; Brühl et al., 2021). Cover of sealed area was used as a proxy of
153 urbanisation, for which we might expect negative effects e.g. due to light pollution (van
154 Grunsven et al., 2020), but could also envision positive relations e.g. due to availability of
155 more diverse food plants (Hall et al., 2017). The cover of these land-use types was deter-
156 mined in a 500m radius around the plots. Starting from the centre point of the hectare of the
157 study site, we included hectares, whose centre point was within the 500m radius of the focal
158 centre point (81ha). The 500m radius was chosen as it covers dispersal distances of many
159 moth species and is a relevant scale for the different land-use types that were involved
160 (Fuentes-Montemayor et al., 2011). Furthermore, it limits spatial autocorrelation among
161 study sites. These land-use covers were attributed to the average year of the aerial photos that
162 were included. For years that fell between the four observation time points, land-use covers
163 were interpolated linearly from the land-use covers in the two adjacent time points. For sam-
164 pling years prior to the first year with land-use data available (~1982) or after the last year
165 with land-use data available (~2015), proportions of the closest year were used, following a
166 conservative scenario of no change.

167 **Statistical analyses**

168 All statistical analyses were performed in R version 4.2.0 (R Core Team, 2022). All code
169 used for analyses is available from the online repository located at
170 <https://github.com/nefffl/moths-CH-Reser>.

171 For each sampling night, we determined three community characteristics: total abundance
172 (sum of individuals across all species), species richness (number of species), and total bio-
173 mass. Total biomass was estimated from species-level dry mass, which was determined from
174 forewing length through allometric equations (Kinsella et al., 2020). Species-level wingspan
175 data, which are more readily available than forewing-length data, were assembled from Jonko
176 (2002–2022) ($n = 981$) and completed with data from other sources (Fibiger, 1990; Potocký
177 et al., 2018; Ronkay et al., 2001) ($n = 12$). Based on data for a subset of study species (Cook
178 et al., 2022), a linear relation between forewing length and wingspan was determined and
179 used to estimate forewing length for all study species. These data were finally used to esti-
180 mate species dry mass. For species where forewing length could not be retrieved ($n = 13$), dry
181 mass was estimated from congeneric species.

182 The three community characteristics were used as response variables in the regression mod-
183 els. We used a zero-inflated negative binomial response distribution for abundance and rich-
184 ness (log link) and a hurdle gamma distribution for biomass (log link), which yielded high
185 agreements between posterior predictive and empirical data distributions (Fig. S1.6). All
186 models included a smoothing term for the day of the year to quantify seasonal patterns, a
187 smoothing term for the elevation of the study site (mean elevation of the study site measured
188 at a hectare) as well as linear terms for weather parameters for the sampling night (tempera-
189 ture and precipitation). To check how the seasonal patterns are influenced by species with
190 different overwintering strategies, we divided the dataset by overwintering stage (egg, larva,

191 pupa, adult) and ran the same models for these data subsets. Overwintering stages were gath-
192 ered from several trait collections (Cook et al., 2022; Mangels et al., 2017; Potocký et al.,
193 2018) and completed based on additional moth trait databases (Jonko, 2002–2022; Ziegler,
194 2005–2022). To analyse the relation of abundance, richness, and biomass to landscape com-
195 position, we included cover of the different land-use types (forest, grassland, cropland, sealed
196 area) as predictor variables.

197 Furthermore, all models had a set of fixed and random factors to account for the sampling de-
198 sign. Fixed effects were trap type (fixed type 1, fixed type 2, or manual), lamp type (four
199 nominal factor levels), number of traps (ordinal factor with four levels) and a two-level nomi-
200 nal factor denoting whether the site was sampled in the previous night to account for persist-
201 ing attraction of light traps. Additionally, there was a smoothing term accounting for the sam-
202 pling duration, which was only included for manual traps and if data on sampling duration
203 was available (2800 out of 4024 manual sampling nights). We chose a smoothing term over a
204 linear effect as we expected moth activity to be changing over the course of a night (e.g. Ma
205 & Ma, 2013). The random structure was composed of the site ID ($n = 171$), the site ID and
206 year combination ($n = 663$), a factor accounting for the sampling night, where proximate sites
207 operated in the same night are grouped together ($n = 34,390$), and a spatio-temporal grouping
208 factor of study sites to account for among-region and year variability ($n = 249$). The latter
209 was defined such that all sites in a year that were within 20km of each other were grouped. In
210 all models, continuous predictor variables were standardised to mean 0 and standard devia-
211 tion 1 prior to analyses. Sum-to-zero contrasts were used for nominal factor variables.

212 The basic structure of all models was built through ‘brms’ (Bürkner et al., 2022) and then
213 manually adapted to meet the specific requirements (covariate for sampling duration only in-
214 cluded if data was available) by directly changing the underlying Stan code of the model. The
215 final models were run through ‘rstan’ (Guo et al., 2022) (4 Markov chain Monte Carlo chains

216 with 2000 iterations each, including 1000 warm-up iterations). Priors followed standard set-
217 tings of the ‘brms’ package representing weakly informative priors (details available in the
218 online repository for the code). Model results were evaluated based on posterior distributions
219 of model predictions.

220 We used the mean and 95% symmetric credible intervals (CIs) based on 2.5% and 97.5%
221 quantiles to summarise posterior distributions. To check whether the four chains mixed well,
222 we calculated Rhat statistics for estimates of intercepts, fixed effect slopes and spline coeffi-
223 cients (smoothing terms) through the package ‘rstan’ (Guo et al., 2022). The standard thresh-
224 old of 1.1 was met by all values of all final models.

225 **RESULTS**

226 **Seasonal patterns in moth community characteristics**

227 All three community characteristics showed a clear seasonal pattern with two distinct peaks
228 (Fig. 1). The first peak, which was clearly lower (richness) or slightly lower (abundance, bio-
229 mass) than the second peak, was at the end of March (March 24 for all community character-
230 istics). The second peak was mid-July (July 12 for abundance and biomass, July 8 for rich-
231 ness) and for abundance and biomass was followed by a plateau peaking again mid-August
232 (August 10 for abundance, August 12 for biomass). The single peaks could be reconstructed
233 for subsets of the dataset defined by species overwintering stages (Fig. 1). At the same time
234 as the first peak, there was a peak of species overwintering as pupa and of the few species
235 overwintering as adults. At the time of the second peak, species overwintering as larva had a
236 coinciding peak. Also, all community characteristics of species overwintering as pupa had a
237 second peak at this time, which however was for abundance and biomass clearly lower than
238 the first peak. Species overwintering in the egg stage, which tend to be rarer than those over-
239 wintering as larva or pupa, had their peak late in the year (between October 22 and November
240 3). Around the same time, the decline of overall community characteristics was slightly miti-
241 gated.

242 **Sampling night specifics and moth community characteristics**

243 Several sampling night specifics were related to the three investigated community character-
244 istics (abundance, richness, biomass) (Fig. 1, Fig. S1.7, Table S1.3). With respect to weather
245 conditions, there was a clear and strong increase of all community characteristics with in-
246 creasing temperatures of the sampling night (Fig. 1, Table S1.3). An increase of sampling
247 night temperature by 50% of its range (corresponding to a rise of 17.2°C) was related to an
248 increase of abundance by a factor 9.68 (95%-CI: 8.98–10.47), of richness by a factor 6.44

249 (95%-CI: 6.16–6.74), and of biomass by a factor 5.14 (95%-CI: 4.77–5.55) (Table 1). For an
250 increase of temperature by 5°C, this corresponds to the factors 1.94 (95%-CI: 1.89–1.98);
251 abundance), 1.72 (95%-CI: 1.70–1.74); richness), and 1.61 (95%-CI: 1.57–1.65); biomass).
252 There was also evidence for slight increases in community characteristics with increasing
253 precipitation in the sampling night (factor 1.12 [95%-CI: 1.02–1.24] for abundance, 1.03
254 [95%-CI: 0.97–1.1] for richness, and 1.19 [95%-CI: 1.07–1.31] for biomass with an increase
255 of 50% of the range, corresponding to 140mm) (Table 1).

256 For other night- and sampling-specific covariates, which were included to correct for data
257 specificities, we found particularly strong relations of community characteristics to trap type,
258 sampling history (i.e. whether there was sampling in the previous night), and to sampling du-
259 ration (Fig. S1.7, Table S1.3). While there was no clear difference between the two fixed trap
260 types, samples from manual traps were estimated to have a factor 6.54 (95%-CI: 4.68–9.3)
261 higher abundance compared to fixed traps of type 1, a factor 3.14 (95%-CI: 2.51–3.97) higher
262 richness, and a factor 5.07 (95%-CI: 3.56–7.21) higher biomass. If there was no sampling
263 taking place at a location in the previous night, samples were estimated to have a factor 0.81
264 (95%-CI: 0.71–0.92) lower abundance, a factor 0.88 (95%-CI: 0.81–0.96) lower richness, and
265 a factor 0.77 (95%-CI: 0.68–0.87) lower biomass. With increasing sampling duration, there
266 were increases in all community characteristics (Fig. S1.7). This increase was not linear, but
267 there was a first peak with a following plateau reached after approx. 7 hours of sampling
268 (flattening of the curve already after approx. 5 hours).

269 **Elevation and landscape composition**

270 All three community metrics increased with elevation, with the increase being close-to linear
271 (Fig. 1, Table S1.3). An increase of elevation by 50% of its range (corresponding to a rise of
272 1130m) was related to an increase of abundance by a factor 3.33 (95%-CI: 2.49–4.44), of

273 richness by a factor 1.38 (95%-CI: 1.13–1.67), and of biomass by a factor 3.54 (95%-CI:
274 2.68–4.67) (Table 1). For all three community characteristics and particularly for richness, a
275 peak with a following plateau or decrease was evident after reaching an elevation of around
276 2000 m asl (Fig. 1).

277 For the effects of the landscape variables (Fig. 2, Table S1.3), there was strong evidence for
278 community characteristics to be higher if cover of forests in the surrounding was high . If for-
279 est cover within a radius of 500m around the plot was higher by 50% of its range (corre-
280 sponding to an approx. proportion of 0.5) , abundance was higher by a factor 1.93 (95%-CI:
281 1.45–2.53), richness by a factor 1.49 (95%-CI: 1.22–1.81), and biomass by a factor 1.72
282 (95%-CI: 1.30–2.29) (Table 1). There was also a tendency for all community characteristics
283 to be higher with increasing proportion of sealed area in the plot surrounding. If sealed area
284 was by 50% of its range higher (corresponding to a proportion of 0.16), abundance was esti-
285 mated to be higher by a factor 1.26 (95%-CI: 0.97–1.64), richness by a factor 1.16 (95%-CI:
286 0.98–1.39), and biomass by a factor 1.26 (95%-CI: 0.98–1.61) (Table 1).

287 **DISCUSSION**

288 When relating moth community characteristics to the day of the year (question 1), a clear sea-
289 sonal pattern was evident, with a peak in early spring and one in summer. The distinct pattern
290 could clearly be related to the life cycles of moth groups distinguished by overwintering
291 stage. The spring peak was related to high activity of species overwintering as adults and pu-
292 pupae, the summer peak to species overwintering as pupae or larvae. In fall, egg-overwintering
293 species were peaking, which was also evident from the seasonal pattern of the overall com-
294 munity. While the observed summer peak matches findings from previous studies (Busse et
295 al., 2022; Jonason et al., 2014; Roth et al., 2021), there is at best some indication for the
296 spring peak in these studies because sampling activity only started around that time. Starting
297 sampling campaigns too late within the year result in systematic under-sampling of species
298 groups that overwinter as pupae or adults. As overwintering stage has regularly been related
299 to species' susceptibility to global change drivers such as climate change or land-use intensi-
300 fication (Forsman et al., 2016; Keret et al., 2020; Mangel et al., 2017; Mattila et al., 2006,
301 2008), restricted sampling within the season can lead to biased conclusions. Furthermore, the
302 observed spring peak was rather narrow. If species phenologies are shifting within years due
303 to climate change (Duchenne et al., 2020), while sampling schemes are not adapted to these
304 shifts, wrong conclusions about temporal trends might be drawn (Didham et al., 2020).

305 There was a strong signal of the weather conditions of the sampling night on all moth com-
306 munity characteristics (questions 2). Particularly strong relations were found to the tempera-
307 ture of the sampling night, which is in line with previous findings (Beck et al., 2010; Jonason
308 et al., 2014; Knop et al., 2018) and reflects increases in moth activity with increasing temper-
309 ature. The effect size was estimated to an increase of community characteristics by 50–100%
310 for a temperature increase of only 5°C (while correcting for elevation and season), which is

311 considerable and needs to be accounted for when analyzing datasets covering nights with
312 fluctuating temperatures.

313 Unimodal patterns of species richness along elevational gradients with a mid-elevation peak
314 are common for many species groups (Rahbek, 2005). Large-scale analyses are rare for
315 moths, but mid-elevation peaks seem to be common (Beck et al., 2017). For moths in the
316 Alps, one study based on a spatially and temporally restricted dataset reports an unimodal
317 peak, whose position depends on the season but moves as high as 1800 m asl (Beck et al.,
318 2010). Here, we show an increase of all community characteristics – particularly strong for
319 abundance and biomass – along the studied elevational gradient, which levels off at around
320 2000 m asl (question 3). Because the studied dataset covers few sites above 2000 m asl and
321 none above 2500 m asl, the observed pattern indicates a unimodal pattern in all community
322 characteristics with a peak at around 2000 m asl. Although the peak is comparably high up
323 the elevational gradient, the pattern corresponds well with previous data on moths (Beck et
324 al., 2010) and with data from several other arthropod groups from the Alps (Fontana et al.,
325 2020). The underlying drivers of this pattern might be manifold (Beck et al., 2017; Hodkin-
326 son, 2005; McCoy, 1990). For the studied gradient, land-use related pressures as well as light
327 pollution are largely decreasing with elevation, which might be driving at least some of the
328 observed increase with elevation.

329 With relation to landscape composition (question 4), we found strongest relations of moth
330 community characteristics to the cover of forests. There were more moths present if cover of
331 forests was high, highlighting their importance for sustaining high moth abundance and rich-
332 ness (Fuentes-Montemayor et al., 2012; Kühne et al., 2022). Furthermore, there was indica-
333 tion for a positive effect of a higher cover of sealed area, i.e. of urban areas. As for other pol-
334 linator groups, urban areas may present additional floral resources for moth communities
335 (Hall et al., 2017), which might explain this positive relation. Why the potentially negative

336 effect of higher light pollution (van Grunsven et al., 2020) in these urbanized landscapes was
337 not evident needs further investigation, though. Promoting forest elements across the land-
338 scape mosaic is imminent to support diverse and abundant moth communities.

339 Based on a moth community dataset of unseen extent, we could evaluate the relations of dif-
340 ferent biotic and abiotic factors to moth community characteristics. The bimodal seasonal
341 pattern with two distinct peaks in all community metrics (abundance, richness, biomass) is in
342 this fine resolution new to research and will help to better evaluate and understand results
343 from studies on smaller scales, to plan future sampling campaigns, and to target conservation
344 efforts more effectively. Nightly temperatures are a very important driver of moth numbers,
345 with even small increases having profound effects on all community characteristics. All moth
346 characteristics increased along the investigated elevational gradient, highlighting the im-
347 portance of high-elevation habitats for conserving moth diversity. At the same time, lower
348 numbers of moth individuals and species at lower elevations might be related to the multitude
349 of anthropogenic pressures such as intensive land use and light pollution, which are more
350 prevalent at lower altitudes. Furthermore, there was a positive relation of moth community
351 characteristics to forest cover. Besides reducing anthropogenic pressures such as light pollu-
352 tion, preserving and promoting woodlands, particularly in low-elevation landscapes, might be
353 key to promote diverse and abundant moth communities in a changing world.

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523

524 **DATA ACCESSIBILITY STATEMENT**

525 The raw moths records data is hosted by info fauna and protected by a code of conduct, but
526 might be obtained from info fauna upon request when in accordance with this code of con-
527 duct. Moths records data at coarser spatial resolution and with unique identifier for the sam-
528 pling location and sampling details necessary to reproduce the models will be made available
529 through the GBIF database. The weather data are under restricted access but might be directly
530 obtained from MeteoSwiss (<https://www.meteoswiss.admin.ch>). Other data and codes neces-
531 sary to reproduce the analyses are available from the GitHub repository located at
532 <https://github.com/nefffl/moths-CH-Reser>.

533

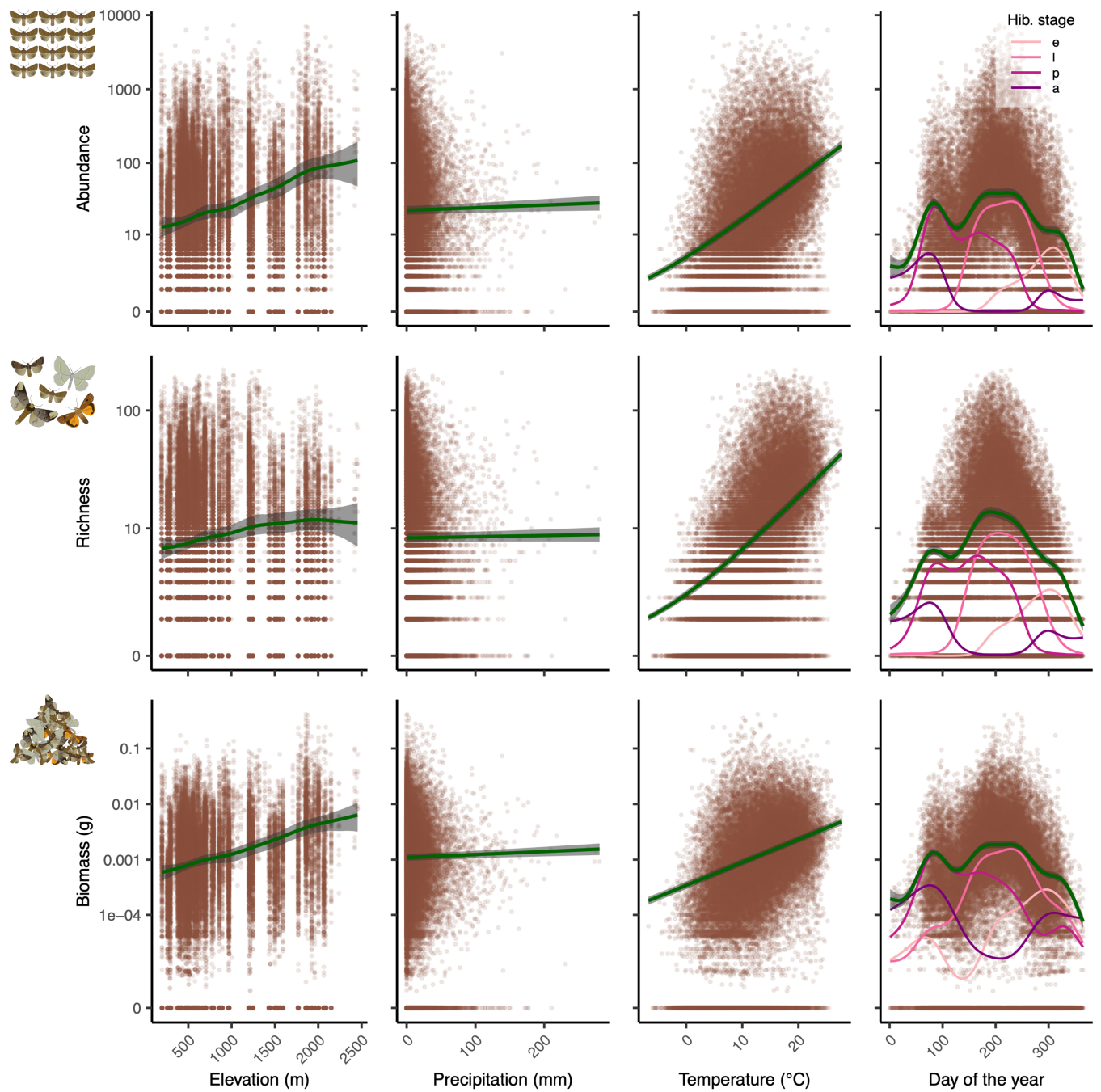
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541 **Table 1** Model predictions on the change in moth abundance, richness, and biomass (g), if a variable is in-
542 creased by 50% of its range of occurring values (indicated in the '50% step' column). Numbers are factors by
543 which the response variable changes. Point estimates (Est.) and 95% credible intervals (CI) are given. Grey
544 shading indicates variables for which the 95%-CI does not include 1.

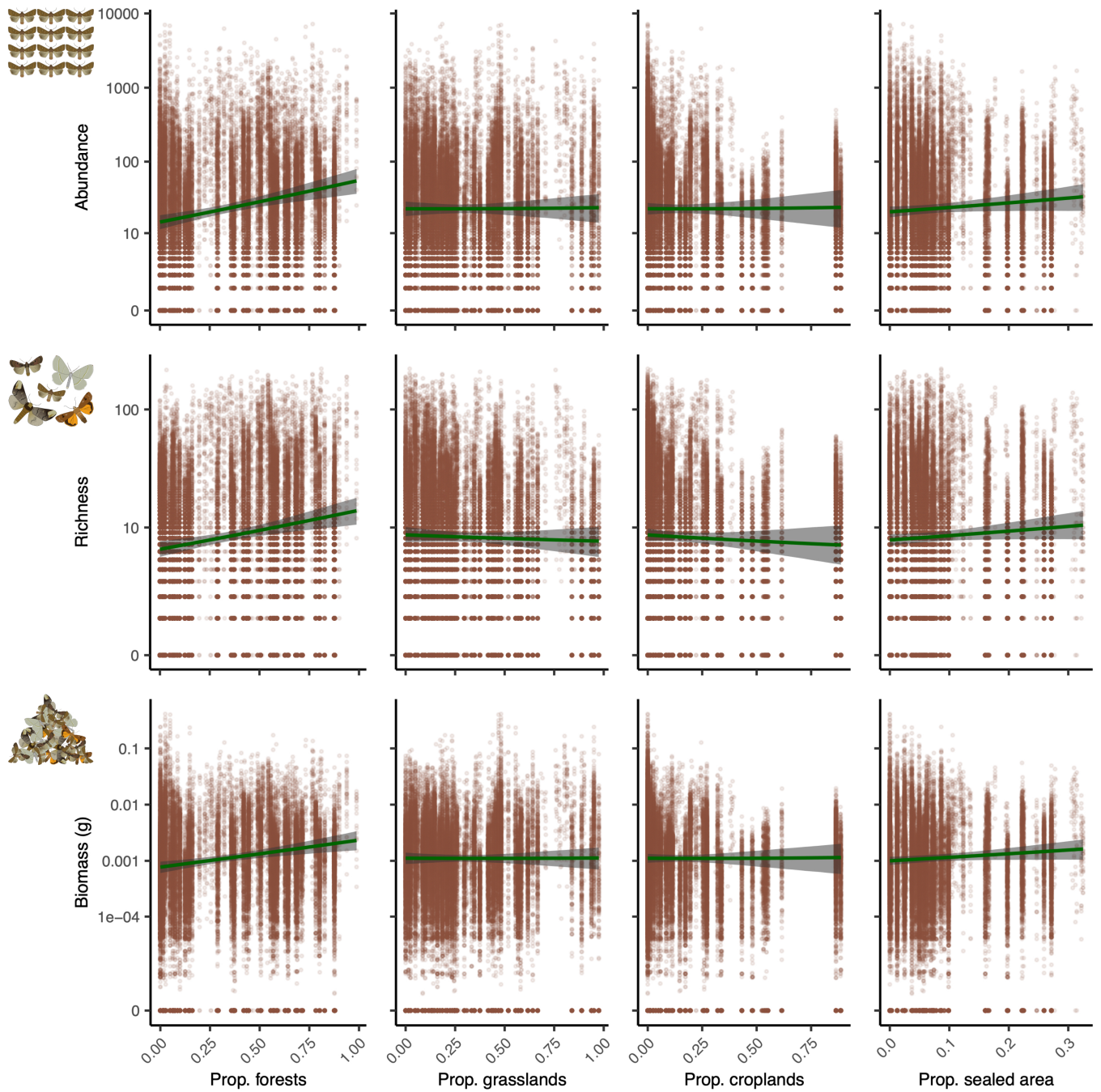
Variable	50% step	Abundance			Richness			Biomass (g)		
		Est.	Lower 95%-CI	Upper 95%-CI	Est.	Lower 95%-CI	Upper 95%-CI	Est.	Lower 95%-CI	Upper 95%-CI
Elevation ¹	1130 m	3.33	2.49	4.44	1.38	1.13	1.67	3.54	2.68	4.67
Precipitation	140 mm	1.12	1.02	1.24	1.03	0.97	1.10	1.19	1.07	1.31
Temperature	17.2 °C	9.68	8.98	10.47	6.44	6.16	6.74	5.14	4.77	5.55
Prop. forests	0.494	1.93	1.45	2.53	1.49	1.22	1.81	1.72	1.3	2.29
Prop. grasslands	0.488	1.01	0.73	1.39	0.93	0.74	1.14	0.99	0.74	1.36
Prop. croplands	0.443	1.00	0.69	1.43	0.89	0.69	1.15	0.99	0.69	1.42
Prop. sealed area	0.162	1.26	0.97	1.64	1.16	0.98	1.39	1.26	0.98	1.61

545 ¹Term included as a smoothing term. Because the prediction is close to linear, change factors still harbour useful
546 information. The values are the means across 100 different 50% ranges.



547

548 **Fig. 1** Conditional effects of different model covariates (elevation, precipitation and temperature of sampling
 549 night, day of the year) on moth abundance, richness, and biomass. The y axis shows (predicted) abundance, rich-
 550 ness, or biomass per sampling night. Point estimates are in green, shaded areas show 95% credible intervals.
 551 Underlying points show data per sampling night ($n = 35,847$). For day of the year, conditional effects from mod-
 552 els run on subgroups defined by hibernation stage (e: egg; l: larva; p: pupa; a: adult) are shown along the condi-
 553 tional effects from the full model. Note that the y axes are on log scale (after adding the minimal non-zero value
 554 to all values). Effects of other covariates are shown in Fig. S1.7. Detailed model results in Table S1.3.



555

556 **Fig. 2** Conditional effects of different landscape variables on moth abundance, richness, and biomass. Land-
 557 scape variables are the covers of different land-use types (forests, grasslands, croplands, sealed area) within a
 558 radius of 500m around the study site. The y axis shows (predicted) abundance, richness, or biomass per sam-
 559 pling night. Point estimates are in green, shaded areas show 95% credible intervals. Underlying points show
 560 data per sampling night ($n = 35,847$). Note that the y axes are on log scale (after adding the minimal non-zero
 561 value to all values). Detailed model results in Table S1.3.

562

SUPPORTING INFORMATION: Appendix S1

Table S1.1 Overview of all moth species recorded with family and superfamily attribution.

Superfamily	Family	Species
Bombycoidea	Brahmaeidae	Lemonia taraxaci
	Endromidae	Endromis versicolora
	Saturniidae	Aglia tau, Saturnia pavonia, S. pavoniella, S. pyri
	Sphingidae	Acherontia atropos, Agrius convolvuli, Deilephila elpenor, D. porcellus, Hyles euphorbiae, H. gallii, H. livornica, H. vespertilio, Laothoe populi, Mimas tiliae, Proserpinus proserpina, Smerinthus ocellata, Sphinx ligustri, S. pinastri
Cossoidea	Cossidae	Cossus cossus, Phragmataecia castaneae, Zeuzera pyrina
Drepanoidea	Drepanidae	Achlya flavicornis, Cilix glaucata, Cymatophorina diluta, Drepana curvatula, D. falcataria, Falcaria lacertinaria, Habrosyne pyritoides, Ochropacha duplaris, Polyploca ridens, Sabra harpagula, Tethea ocularis, T. or, Tetheella fluctuosa, Thyatira batis, Watsonalla binaria, W. cultraria, W. uncinula
Geometroidea	Geometridae	Abraxas grossulariata, A. sylvata, Acasis appensata, A. viretata, Adactylotis contaminaria, Aethalura punctulata, Agriopis aurantiaria, A. bajaria, A. leucophaearia, A. marginaria, Alcis deversata, A. jubata, A. repandata, Aleucis distinctata, Alsophila aceraria, A. aescularia, Angerona prunaria, Anticlea derivata, Anticollix sparsata, Apeira syringaria, Aplocera plagiata, A. praeformata, Apocheima hispidaria, Arichanna melanaria, Ascotis selenaria, Aspitates gilvaria, Asthena albulata, A. anseraria, Biston betularia, B. strataria, Bupalus piniaria, Cabera exanthemata, C. pusaria, Campaea margaritaria, Campptogramma bilineata, C. scripturata, Carsia sororiata, Cataclysmes riguada, Catarhoe cuculata, C. rubidata, Cepphis advenaria, Charissa ambiguata, C. glaucinaria, C. italo-helvetica, C. obscurata, C. pullata, C. variegata, Chesias legatella, C. rufata, Chiasmia clathrata, Chlorissa cloraria, C. viridata, Chloroclysta miata, C. siterata, Chloroclystis v-ata, Cidaria fulvata, Cleora cinctaria, Cleorodes licheneria, Coenotephria ablutaria, C. salicata, C. tophaceata, Colostygia aptata, C. aqueata, C. kollariaria, C. laetaria, C. olivata, C. pectinataria, C. puengeleri, C. turbata, Colotois pennaria, Comibaena bajularia, Cosmorhoe ocellata, Costaconvexa polygrammata, Crocallis elinguaris, C. tusciaria, Cyclophora albicellaria, C. albipunctata, C. annularia, C. linearia, C. pendularia, C. porata, C. punctaria, C. pupillaria, C. quercimontaria, C. ruficiliaria, C. suppunctaria, Deileptenia ribeata, Dysstroma citrata, D. truncata, Earophila badiata, Ecliptopera capitata, E. silaceata, Ectopis crepuscularia, Electrophaes corylata, Elophos caelibaria, E. operaria, Ematurga atomaria, Emmiltis pygmaearia, Ennomos alniaria, E. erosaria, E. fuscantaria, E. quercaria, E. quercinaria, Entephria caesiata, E. cyanata, E. flavata, E. flavicinctata, E. infidaria, E. nobiliaria, Epilobophora sabinata, Epione repandaria, E. vespertaria, Epirrhoe alternata, E. galiata, E. molluginata, E. rivata, E. tristata, Epirrita autumnata, E. christyi, E. dilutata, Erannis defoliaria, Euchoeca nebulata, Eulithis mellinata, E. populata, E. prunata, E. testata, Euphyia biangulata, E. frustata, E. unangulata, Eupithecia abbreviata, E. abietaria, E. absinthiata, E. actaeata, E. analoga, E. assimilata, E. carpophagata, E. cauchiata, E. centaureata, E. conterminata, E. cretaceata, E. denotata, E. denticulata, E. distinctaria, E. dodoneata, E. egenaria, E. ericeata, E. exigua, E. expallidata, E. extraversaria, E. gemellata, E. graphata, E. gueneata, E. haworthiata, E. icterata, E. immundata, E. impurata, E. indigata, E. innotata, E. insigniata, E. intricata, E. inturbata, E. irriguata, E. lanceata, E. laquaearia, E. lariciata, E. linariata, E. nanata, E. ochridata, E. orphnata, E. pernotata, E. pimpinellata, E. plumbeolata, E. pusillata, E. pyreneata, E. satyrata, E. schiefereri, E. selinata, E. semigraphata, E. silenata, E. sinuosaria, E. subfasciata, E. subumbrata, E. succenturiata, E. tantillaria, E. tenuiata, E. tripunctaria, E. trisignaria, E. undata, E. valerianata, E. venosata, E. veratraria, E. virgaureata, E. vulgata, Eustroma reticulata, Fagivorina arenaria, Gagitodes sagittata, Gandaritis pyraliata, Geometra papilionaria, Gnophos furvata, G. obfuscata, Gymnoscelis ruffifasciata, Heliomata glarearia, Hemistola chrysoprasaria, Hemithea aestivaria, Horisme aemulata, H. calligraphata, H. radiciaria, H. tersata, H. vitalbata, Hydrelia flammeolaria, H. sylvata, Hydria cervinalis, H. montivagata, H. undulata, Hydriomena furcata, H. impluviata, H. ruberata, Hylaea fasciaria, Hypomecis punctinalis, H. roboraria, Idaeia aureolaria, I. aversata, I. biselata, I. calunetaria, I. contiguaria, I. degeneraria, I. deversaria, I. dilutaria, I. dimidiata, I. emarginata, I. flaveolaria, I. fuscovenosa, I. humiliata, I. inquinata, I. laevigata, I. macilentaria, I. moniliata, I. muricata, I. nitidata, I. obsoletaria, I. ochrata, I. politaria, I. rubraria, I. rufaria, I. rusticata, I. seriata, I. sericeata, I. serpentata, I. straminata, I. subsericeata, I. sylvestraria, I. trigeminata, I. typicata, Isturgia arenacearia, Jodis lactaria, J. putata, Lampropteryx suffumata, Larentia clavaria, Ligdia adustata, Lobophora halterata, Lomaspiis marginata, Lomographa bimaculata, L. temerata, Lycia alpina, L. hirtaria, L. zonaria, Macaria alternata, M. artesiaria, M. brunneata, M. liturata, M. notata, M. signaria, M. wauaria, Martania taeniata, Melanthia alaudaria, M. procellata, Menophra abruptaria, M. nyctemeraria, Mesoleuca albicillata, Mesotype didymata, M. parallelolineata, M. verberata, Minoa murinata, Nebula achromaria, N. nebulata, Nothocasis sertata, Odontopera bidentata, Operophtera brumata, O. fagata, Opisthograptis luteolata, Orthonama obstipata, O. vittata, Ourapteryx sambucaria, Pachycnemia hippocastanaria, Paradarisa consonaria, Parectropis similaria, Pareulype berberata, Pasiphila chloerata, P. debiliata, P. rectangulata, Pelurga comitata, Pennithera firmata, Perconia strigillaria, Peribatodes perversaria, P. rhomboidaria, P. secundaria, Perizoma affinitata, P. albulata, P. alchemillata, P. bifaciata, P. blandiata, P. flavofasciata, P. hydrata, P. incultaria, P. juracolaria, P. lugdunaria, P. minorata, P. obsoletata, Petrophora chlorosata, Phaiogramma etruscaria, Phigalia pilosaria, Philereime transversata, P. vetulata, Plagodis dolabraria, P. pulveraria, Pterygia rubiginata, Pseudoterpna pruinata, Pterapherapteryx sexualata, Pungeleria capreolaria, Rheumaptera subastata, Rhodometra saccharia, Rhodostrophia calabra, R. vibicaria, Sciadia zelleraria, Scopula caricaria, S. confinaria, S. decorata, S. floslactata, S. imitaria, S. immorata, S. immutata, S.

		incanata, S. marginepunctata, S. nigropunctata, S. ornata, S. rubiginata, S. submutata, S. subpunctaria, S. ternata, S. umbelaria, S. virgulata, Scotopteryx angularia, S. bipunctaria, S. chenopodiata, S. luridata, S. moeniata, S. mucronata, S. octodurensis, S. vicinaria, Selenia dentaria, S. lunularia, S. tetralunaria, Selidosema brunnearia, S. plumaria, Siona lineata, Spargania luctuata, Stegania cararia, S. trimaculata, Synopsia sociaria, Tephronia sepiaria, Thaleria fimbrialis, Thera britannica, T. cembrae, T. cognata, T. cupressata, T. juniperata, T. obeliscata, T. variata, T. vetustata, Theria primaria, T. rupicaprararia, Thetidia smaragdaria, Timandra comae, Trichopteryx carpinata, T. polycommata, Triphosa dubitata, T. sabaudiata, T. tauteli, Venusia blomeri, V. cambrica, Xanthorhoe biriviata, X. decoloraria, X. designata, X. ferrugata, X. fluctuata, X. incurcata, X. montanata, X. quadrifasiata, X. spadicearia, Yezognophos dilucidaria, Y. serotinaria, Y. vittaria
Hepialoidea	Hepialidae	Gazoryctra ganna, Hepialus humuli, Korscheltellus fusconebulosa, K. lupulina, Phymatopus hecta, Triodia sylvina
Lasiocampoidea	Lasiocampidae	Cosmotriche lobulina, Dendrolimus pini, Eriogaster arbusculae, E. catax, E. lanestris, Euthrix potatoria, Gastro-pacha quercifolia, Lasiocampa quercus, L. trifolii, Macrothylacia rubi, Malacosoma alpicola, M. castrensis, M. neustria, Odonestis pruni, Phylloidesma tremulifolia, Poecilocampa alpina, P. populi, Trichiura crataegi
Noctuoidea	Erebidae	Arctia caja, A. flavia, A. matronula, A. testudinaria, A. villica, Arctornis l-nigrum, Atolmis rubricollis, Autophila dilucida, Callimorpha dominula, Calliteara pudibunda, Calyptra thalictri, Catephia alchymista, Catocala coniuncta, C. dilecta, C. electa, C. elocata, C. fraxini, C. fulminea, C. nupta, C. nymphaea, C. nymphagoga, C. promissa, C. puerpera, C. sponsa, Chelis simplonica, Colobochoyla salicalis, Coscinia cribraria, Cybosis mesomella, Diacrisia purpurata, D. sannio, Diaphora mendica, D. sordida, Dysgonia algira, Eilema caniola, E. cereola, E. complana, E. depressa, E. griseola, E. lutreola, E. lutarella, E. palliatella, E. pseudocomplana, E. pygmaeola, E. sororcula, Epatolmis luctifera, Eublemma ostrina, E. parva, E. polygramma, E. purpurina, Euplagia quadripunctaria, Euproctis chrysorrhoea, Grammodes stolidia, Gynaephora fascalina, Herminia grisealis, H. tarsicrinalis, H. tarsipennalis, H. tenuialis, Hypena crassalis, H. lividalis, H. obesalis, H. obsitalis, H. proboscidalis, H. rostralis, Hyphenodes humidalis, Hyphantria cunea, Idia calvaria, Laspeyria flexula, Leucoma salicis, Lithosia quadra, Lygephila cracca, L. lusoria, L. pastinum, L. viciae, Lymantria dispar, L. monacha, Macrochilo cribrumalis, Metachrostis dardouini, Miltochrista miniata, Minucia lunaris, Nudaria mundana, Ocneria rubea, Ocnogyna parasita, Orgyia antiqua, O. recens, Paidia rica, Paracolax tristalis, Parascotia fuliginaria, Pechipogo strigilata, Pelosia muscerda, P. obtusa, Phragmatobia fuliginosa, Phytometra viridaria, Polypogon gryphalis, P. plumigeralis, P. tentacularia, Rivula sericealis, Schrankia costaestrigalis, S. taenialis, Scoliopteryx libatrix, Setina irrorella, Sphrageidus similis, Spilarctia lutea, Spilosoma lubricipeda, S. urticae, Thumatha senex, Trisateles emortualis, Tyria jacobaeae, Watsonarctia deserta, Zanclognatha lunalis, Z. zelleralis
	Euteliidae	Eutelia adalatrix
	Noctuidae	Abrostola agnorista, A. asclepiadis, A. tripartita, A. triplasia, Acontia trabealis, Acosmetia caliginosa, Acronicta aceris, A.alni, A. auricoma, A. cuspidis, A. euphorbiae, A. leporina, A. megacephala, A. menyanthidis, A. psi, A. rumicis, A. strigosa, A. tridens, Actebia multifida, A. praecox, Actinotia polyodon, Aedia funesta, A. leucomelas, Agrochola helvola, A. laevis, A. litura, A. lota, A. lychnidis, A. macilentata, A. nitida, A. pistacinoides, A. ruticilla, Agrotis bigramma, A. cinerea, A. clavis, A. exclamationis, A. fatidica, A. ipsilon, A. puta, A. segetum, A. simplonia, A. trux, A. vestigialis, Allophyes oxyacanthae, Ammoconia caecimacula, A. senex, Amphipoea fucosa, A. lucens, A. oculea, Amphipyra berbera, A. cinnamomea, A. livida, A. perflua, A. pyramidea, A. tragopoginis, Anaplectoides prasina, Anarta melanopa, A. myrtilli, A. odontites, A. trifolii, Anorthoa munda, Antitype chi, A. suda, Apamea anceps, A. aquila, A. crenata, A. epomidion, A. furva, A. illyria, A. lateritia, A. lithoxylaea, A. maillardi, A. monoglypha, A. platinea, A. remissa, A. rubrirena, A. scolopacina, A. sordens, A. sublustris, A. unanims, A. zeta, Aporophya lutulenta, A. nigra, Apterogenum ypsilon, Archanara dissoluta, A. neurica, Asteroscopus sphinx, Atethmia centrigo, Athetis gluteosa, A. hospes, A. pallustris, Atypha pulmonaris, Auchmis detersa, Autographa aemula, A. bractea, A. gamma, A. jota, A. pulchrina, Axylia putris, Brachionycha nubeculosa, Brachyolomia viminalis, Bryophila domestica, B. ereptricula, B. petricolor, B. raptricula, B. ravula, Bryopsis muralis, Calamia tridens, Calliergis ramosa, Callopietria juvenina, C. latreillei, Calophasia lunula, C. platyptera, Caradrina aspersa, C. clavipalpis, C. flavirena, C. gilva, C. kadenii, C. montana, C. morpheus, C. selini, C. terrea, C. wulschlegeli, Ceramica pisi, Cerapteryx graminis, Cerastis leucographa, C. rubricosa, Charanyca trigrammica, Chersotis alpestris, C. andereggii, C. cuprea, C. fimbriola, C. margaritacea, C. multangula, C. ocellina, C. oreina, C. rectangularis, Chilodes maritima, Chloantha hyperici, Chrysodeixis chalcites, Cirrha gilvago, C. icteritia, C. ocellaris, Clemathada calberlai, Colocasia coryli, Conistra erythrocephala, C. ligula, C. rubiginea, C. rubiginosa, C. torrida, C. vaccinii, Cosmia affinis, C. pyralina, C. trapezina, Craniophora ligustri, Cryphia algae, C. ochsi, C. simulatricula, Crypsedra gemmea, Cucullia absinthii, C. artemisiae, C. asteris, C. campanulae, C. gnaphalii, C. lactucae, C. lucifuga, C. lychnitis, C. prenanthis, C. santonici, C. scrophulariae, C. thapsiphaga, C. umbratica, C. verbasci, Dasyptolia ferdinandi, D. templi, Deltote bankiana, D. deceptoria, D. pygarga, D. uncula, Denticucullus pygmina, Diachrysis chrysitis aggr., D. chryson, D. nadeja, Diarsia brunnea, D. dahlii, D. mendica, D. rubi, Dichagyris candelsequa, D. flammata, D. forcipula, D. musiva, D. nigrescens, D. signifera, D. vallesiaca, Dichonia convergens, Dicycla oo, Diloba caeruleocephala, Dryobotodes eremita, Dypterygia scabriuscula, Egira conspicillaris, Elaphria venustula, Enargia paleacea, Enterpia laudeti, Epilecta linogrisea, Epimecia ustula, Epipsilia grisescens, E. latens, Episema glaucina, Eriopygodes imbecilla, Eucarta amethystina, E. virgo, Euchalcia modestoides, E. variabilis, Eugnorisma depuncta, E. glareosa, Eugraphe sigma, Euplexia lucipara, Eupsilia transversa, Eurois occulta, Euxoa aquilina, E. birivia, E. cos, E. culminicola, E. decora, E. distinguenda, E. eruta, E. nigricans, E. obeliscus, E. recussa, E. vitta, Gortyna flavago, Graphiphora augur, Griposia aprilina, Hada plebeja, Hadenia albimacula, H. bicurris, H. caesia, H. compta, H. confusa, H. filograna, H. irregularis, H. luteocincta, H. magnolii, H. perplexa, H. tephroleuca, Hecatera bicolorata, H. dysodea, Helicoverpa armigera, Heliothis nubigera, H. peltigera, H. viriplaca, Helotropha leucostigma, Hoplodrina ambigua, H. blanda, H. octogenaria, H. respersa, H. superstes, Hydracraia micacea, Hyppa rectilinea, Ipimorpha retusa, I. subtusa, Jodia croceago, Lacanobia aliena, L. contigua, L. oleracea, L. splendens, L. suasa, L. thalassina, L. w-latinum, Lamprotes c-aureum, Lasionhada proxima, Lateroligia ophiogramma, Lenisa geminipuncta, Leucania comma, L. loreyi, L. obsoleta, Lithophane consocia, L.

	<p>furcifera, L. leautieri, L. ornitopus, L. semibrunnea, L. socia, Litologia literosa, Luperina dumerilii, L. testacea, Luteohadena luteago, Lycophotia porphyrea, Macdunnoughia confusa, Mamestra brassicae, Melanchra persicariae, Mesapamea secalis aggr., Mesogona acetosellae, M. oxalina, Mesologia furuncula, Mniotype adusta, M. satura, M. solieri, Moma alpium, Mormo maura, Mythimna albipuncta, M. anderreggii, M. congrua, M. conigera, M. ferrago, M. impura, M. l-album, M. pallens, M. pudorina, M. riparia, M. sicula, M. straminea, M. turca, M. unipuncta, M. vitellina, Naenia typica, Noctua comes, N. fimbriata, N. interjecta, N. interposita, N. janthe/janthina aggr., N. orbona, N. pronuba, N. tirrenica, Nonagria typhae, Ochropleura plecta, Oligia dubia, O. latruncula, O. strigilis, O. versicolor, Opigena polygona, Orbona fragariae, Orthosia cerasi, O. cruda, O. gothica, O. gracilis, O. incerta, O. miniosa, O. opima, O. populeti, Pachetra sagittigera, Panchrysis deaurata, P. v-argenteum, Panemeria tenebrata, Panolis flammea, Panthea coenobita, Papestra biren, Paradiarsia punicea, Parastichtis suspecta, Peridroma saucia, Phlogophora meticulosa, P. scita, Photedes captiuncula, P. fluxa, P. minima, P. morrisii, Phragmatiphila nexa, Phyllophila oblitterata, Plusia festucae, Polia bombycina, P. hepatica, P. nebulosa, P. serratilinea, Polychrysis moneta, Polymixis rufocincta, P. xanthomista, Polyphaenis sericata, Protolampra sobrina, Pseudeustrotia candidula, Pyrrhia umbra, Rhizedra lutosa, Rhyacia helvetina, R. lucipeta, R. simulans, Rusina ferruginea, Sedina buettneri, Sideridis kitti, S. lampra, S. reticulata, S. rivularis, S. turbida, Spaelotis ravida, S. senna, Spodoptera exigua, Standfussiana lucernea, S. wiskotti, Staurophora celsia, Sunira circellaris, Syngrapha ain, S. hochenwarthi, S. interrogationis, Thalpophila matura, Tholera cespitis, T. decimalis, Tiliacea aurago, T. citrigo, Trachea atriplicis, Trichoplusia ni, Trigonophora flammea, Tyta luctuosa, Xanthia togata, Xestia alpicola, X. ashworthii, X. baja, X. c-nigrum, X. castanea, X. collina, X. ditrapezium, X. lorezi, X. ochreago, X. rhaetica, X. sexstrigata, X. speciosa, X. stigmatica, X. triangulum, X. xanthographa, Xylena exsoleta, X. solidaginis, X. vetusta, Xylocampa areola</p>
Nolidae	<p>Bena bicolorana, Earias biplaga, E. clorana, E. vernana, Meganola albula, M. strigula, M. togatulalis, Nola aerugula, N. confusalis, N. cucullatella, N. subchlamydula, Nycteola asiatica, N. degenerana, N. revayana, N. sicilana, Pseudoips prasinana</p>
Notodontidae	<p>Cerura erminea, C. vinula, Clostera anachoreta, C. anastomosis, C. curtula, C. pigra, Drymonia dodonaea, D. oblitterata, D. querna, D. ruficornis, D. velitaris, Furcula bicuspis, F. bifida, F. furcula, Gluphisia crenata, Harpyia milhauseri, Leucodonta bicoloria, Notodonta dromedarius, N. tritophus, N. ziczac, Odontotia carmelita, Peridea anceps, Phalera bucephala, Pheosia gnoma, P. tremula, Pterostoma palpina, Ptilodon capucina, P. cucullina, Ptilophora plumigera, Spatalia argentina, Stauropus fagi, Thaumetopoea pityocampa, T. processionea</p>

Table S1.2 Attribution of land-use and land-cover categories distinguished in the “Arealstatistik” to broad land-use types used in this study. Numbers show codes used in the “Arealstatistik”.

Land-use type	Arealstatistik categories
Forests	Forests (without agricultural usage) (NOLU300)
Grasslands	Natural meadows (NOLU222), Pastures (NOLU223), Alpine meadows (NOLU241), Alpine and Jura pastures (NOLU242), Alpine sheep pastures (NOLU243)
Croplands	Orchards (NOLU201), Vineyards (NOLU202), Horticulture (NOLU203), Arable land (NOLU221)
Sealed area	Paved areas (NOLC11), Buildings (NOLC12), Greenhouses (NOLC13)

Table S1.3 Detailed model results for the models analysing the relation of elevation, landscape composition, weather, season and other covariates to moth abundance, richness, and biomass. Fixed effects of linear effects and splines and standard deviations of random factors and splines are given. For factors, sum-to-zero contrasts were used. Point estimates and 95% credible intervals (95%-CI) are given. If 95%-CI do not include zero, numbers are bold.

Parameter	Type	Variable	Abundance			Richness			Biomass			
			Estimate	Lower 95%-CI	Upper 95%-CI	Estimate	Lower 95%-CI	Upper 95%-CI	Estimate	Lower 95%-CI	Upper 95%-CI	
Fixed effect	Fixed	Proportion forests	0.377	0.215	0.532	0.228	0.115	0.341	0.310	0.150	0.476	
		Proportion grasslands	0.00396	-0.160	0.169	-0.0365	-0.154	0.0692	-0.00262	-0.157	0.159	
		Proportion croplands	0.000358	-0.135	0.133	-0.0435	-0.135	0.0512	-0.00280	-0.136	0.130	
		Proportion sealed area	0.0971	-0.0149	0.206	0.0627	-0.00646	0.137	0.0970	-0.00918	0.199	
		Precipitation	0.0138	0.00202	0.0254	0.00405	-0.00306	0.0115	0.0206	0.00817	0.0325	
		Temperature	0.727	0.703	0.752	0.597	0.582	0.611	0.524	0.500	0.549	
		Trap type (contr. sum 1)	-0.426	-0.737	-0.112	-0.325	-0.543	-0.106	-0.320	-0.637	-0.00668	
		Trap type (contr. sum 2)	-1.03	-1.62	-0.454	-0.494	-0.918	-0.0831	-0.982	-1.56	-0.398	
		Lamp type (contr. sum 1)	-0.300	-0.619	0.0186	-0.174	-0.375	0.0277	-0.281	-0.588	0.0435	
		Lamp type (contr. sum 2)	-0.0653	-0.721	0.594	-0.125	-0.500	0.269	-0.207	-0.859	0.489	
		Lamp type (contr. sum 3)	-0.378	-0.689	-0.0532	-0.189	-0.389	0.0214	-0.377	-0.688	-0.0541	
		Nr. of traps (linear)	0.323	-0.677	1.30	0.245	-0.391	0.920	0.263	-0.691	1.18	
		Nr. of traps (quadratic)	-0.408	-1.16	0.305	-0.280	-0.762	0.183	-0.513	-1.22	0.196	
		Nr. of traps (cubic)	-0.337	-0.701	0.0257	-0.176	-0.408	0.0706	-0.391	-0.767	-0.0303	
		Sampl. prev. night (contr. sum 1)	-0.106	-0.169	-0.0421	-0.0635	-0.103	-0.0222	-0.131	-0.190	-0.0699	
Spline	Spline	Elevation	1.97	-3.10	5.20	0.246	-3.06	2.44	3.19	-0.838	6.09	
		Day of year	-23.3	-28.0	-18.7	-8.57	-12.3	-4.76	-18.7	-23.2	-14.4	
		Hrs. active	6.63	1.85	11.2	3.97	0.878	7.49	5.54	0.843	10.2	
SD	Random	Spat-temp. cluster	0.267	0.196	0.334	0.133	0.0907	0.173	0.270	0.204	0.337	
		Sampling night x Location	0.795	0.774	0.816	0.481	0.472	0.489	0.599	0.579	0.619	
		Site ID	0.521	0.453	0.601	0.369	0.321	0.421	0.501	0.434	0.575	
		Site ID x Year	0.234	0.196	0.276	0.141	0.120	0.164	0.228	0.190	0.271	
	Spline	Spline	Elevation	1.60	0.0461	5.87	0.789	0.0489	2.47	1.06	0.0347	3.63
			Day of year	12.6	8.07	20.5	5.88	3.79	9.30	11.0	7.11	17.3
			Hrs. active	1.87	0.938	3.49	1.21	0.528	2.46	1.91	0.890	3.84

Nr. of traps: Number of traps
 Sampl. prev. night: Sampling in previous night
 Spat-temp. cluster: Spatio-temporal cluster



Fig. S1.1 Pictures of the three trap types that were used for sampling. **(a)** fixed trap ("type 1"), **(b)** manual trap, **(c)** fixed trap ("type 2"). Pictures from Rezbanyai-Reser (2014, 2018a).

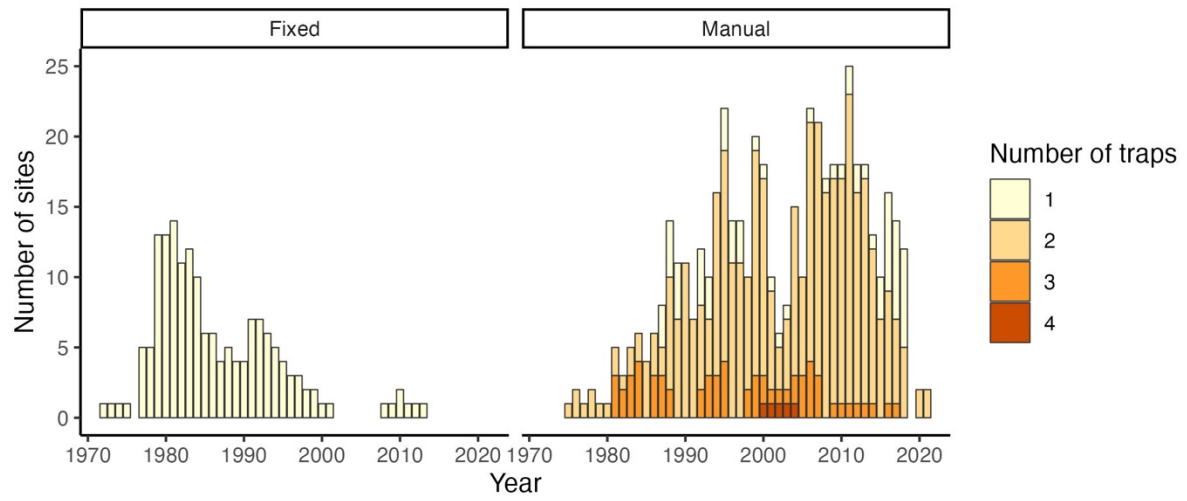


Fig. S1.2 Number of sites per year with respect to the number of traps that were active. Panels on the left show fixed trap samples, panels on the right show manual trap samples.

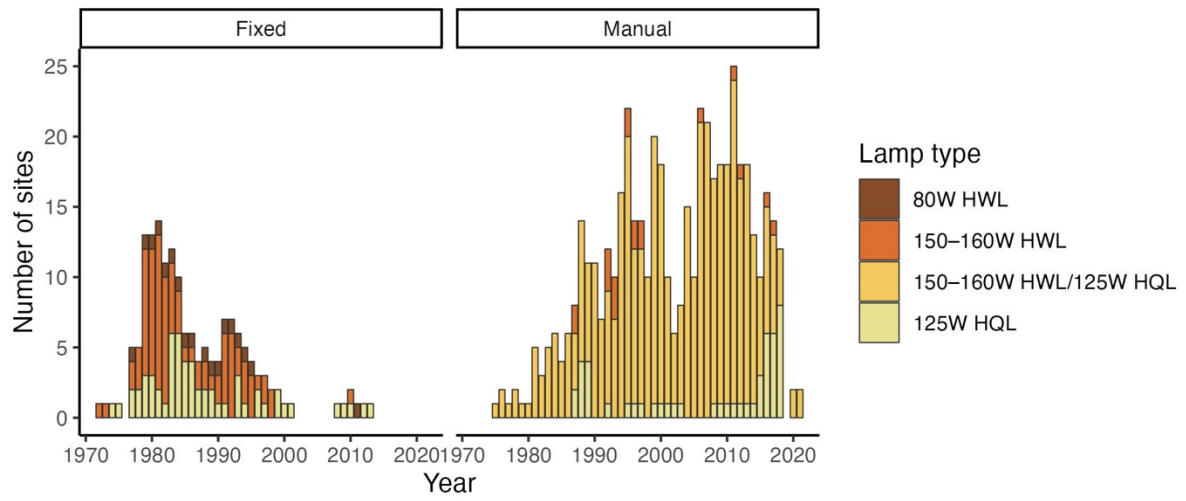


Fig. S1.3 Number of sites per year with respect to the lamp type that was used. Panels on the left show fixed trap samples, panels on the right show manual trap samples.

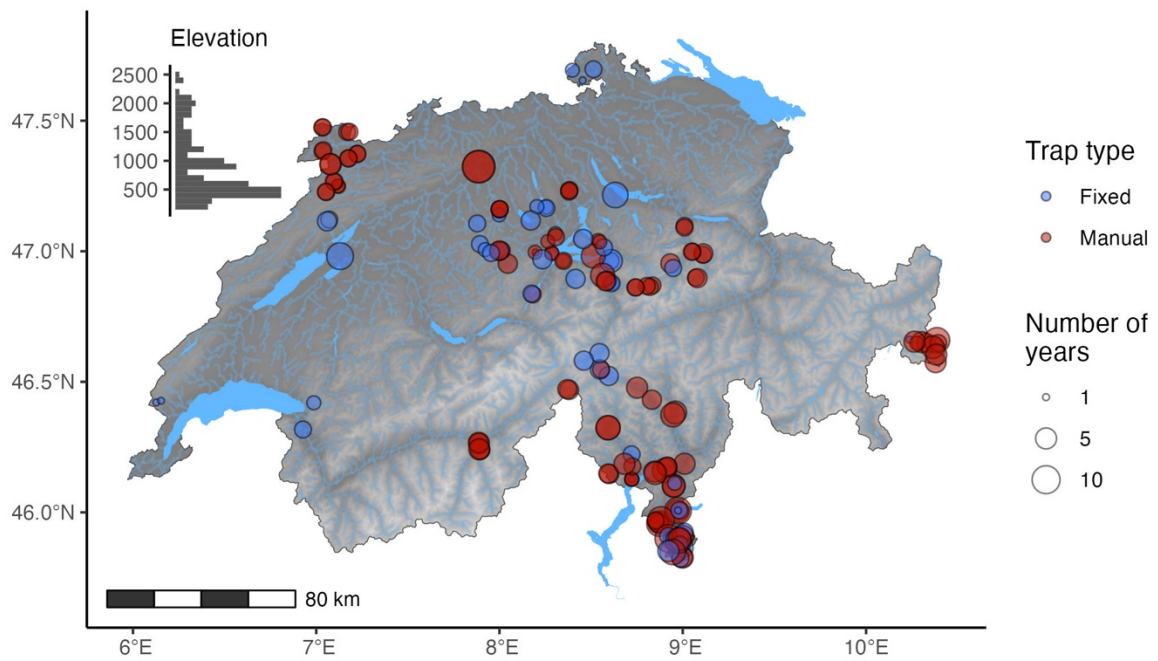


Fig. S1.4 Spatial distribution of sampling sites in Switzerland. Points show sampling sites, colour indicates trap-type and point size indicates the number of years in which the site was sampled. Shading of the map shows elevation (darker shading for lower elevation). The top left inlet shows the distribution of study sites across the elevation gradient (meter above sea level).

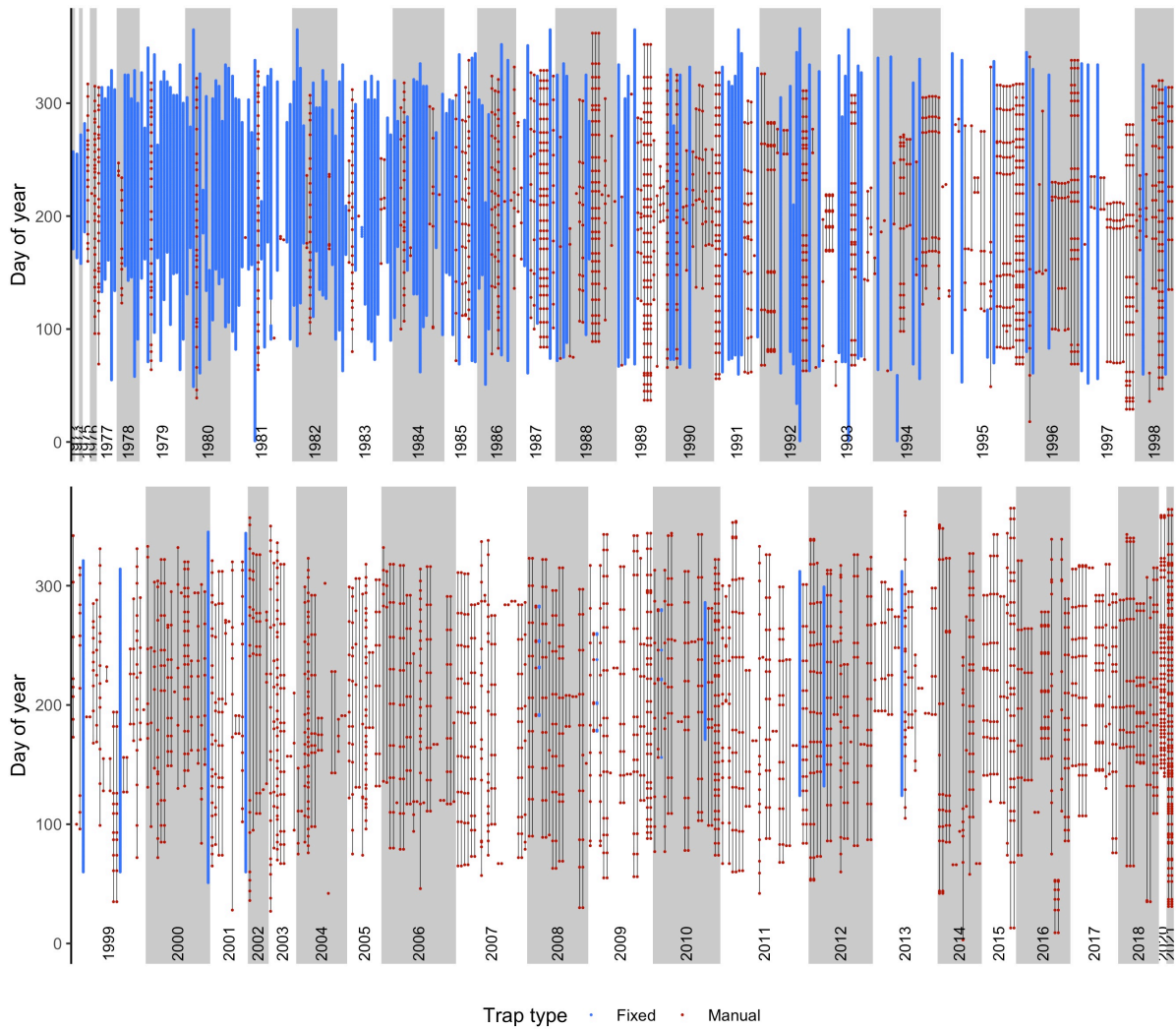


Fig. S1.5 Overview of single sampling nights, arranged chronologically across the whole study period. Each vertical line connects samples from one site and one year. They build blocks of one year, indicates by the shading of the background (starting top left with 1972, ending bottom right with 2021). The y axis denotes the day of the year (between 1 and 366) of the sampling night. Single sampling night are shown as points (which might overlay for very dense sampling periods), colors differentiate the fixed samplings (blue) and manual samplings (red).

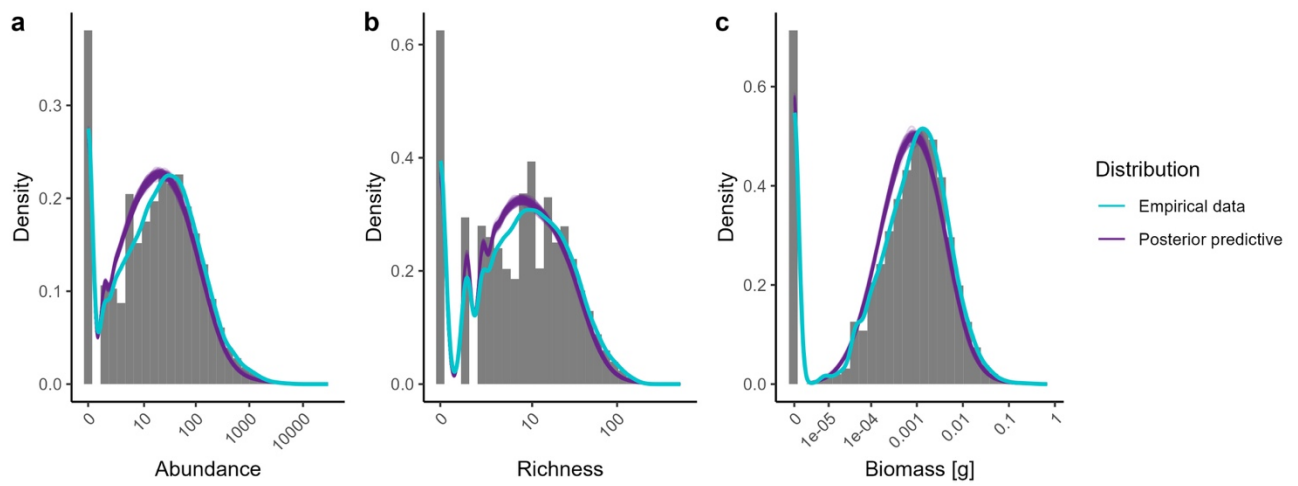


Fig. S1.6 Comparison of empirical data distribution and 100 draws of the posterior predictive distribution from the (a) abundance, (b) richness and (c) biomass models. Lines show kernel density estimates for empirical data (turquoise) and posterior predictive (purple) distributions, bars show histograms of the empirical data distributions. For abundance and richness, a zero-inflated negative binomial distribution was used, whereas for biomass, a hurdle gamma distribution was used. Note that the x axes are on log scale (after adding the minimal non-zero value to all values).

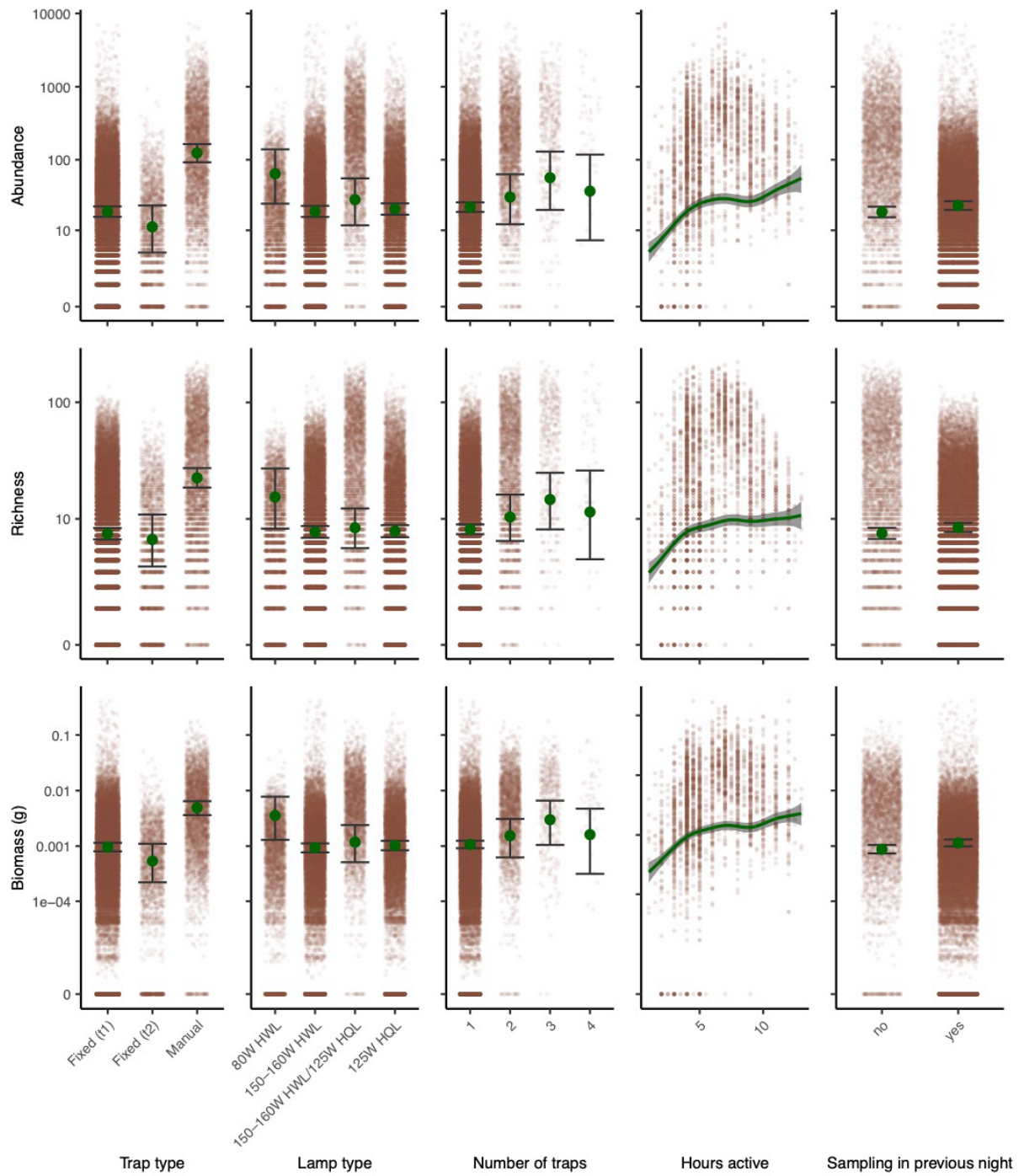


Fig. S1.7 Conditional effects of different model covariates in the abundance, richness, and biomass model. The y axis shows predicted abundance, richness, or biomass. Point estimates are in green, shaded areas or error bars show 95% uncertainty intervals. Underlying points show data per sampling night ($n = 35,847$). Detailed model results in Table S1.3.