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1 **Insect communities under skyglow: diffuse nighttime illuminance induces**
2 **spatiotemporal shifts in movement and predation**

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20
21 **Abstract**

22
23 Artificial light at night (ALAN) is predicted to have far-reaching consequences for natural ecosystems
24 given its influence on organismal physiology and behaviour, species interactions, and community
25 composition. Movement and predation are fundamental ecological processes that are of critical
26 importance to ecosystem functioning. The natural movements and foraging behaviours of nocturnal

27 invertebrates may be particularly sensitive to the presence of ALAN. However, we still lack evidence
28 of how these processes respond to ALAN within a community context. We assembled insect
29 communities to quantify their movement activity and predation rates during simulated moon cycles
30 across a gradient of diffuse nighttime illuminance including the full range of observed skyglow
31 intensities. Using radio frequency identification, we tracked the movements of insects within a
32 fragmented grassland Ecotron experiment. We additionally quantified predation rates using prey
33 dummies. Our results reveal that even low-intensity skyglow causes a temporal shift in movement
34 activity from day to night, and a spatial shift towards open habitats at night. Changes in movement
35 activity are associated with indirect shifts in predation rates. Spatiotemporal shifts in movement and
36 predation have important implications for ecological networks and ecosystem functioning,
37 highlighting the disruptive potential of ALAN for global biodiversity and the provision of ecosystem
38 services.

39

40 **Keywords:** light pollution, ALAN, fragmented landscapes, activity pattern, foraging, nocturnal,
41 crepuscular

42

43 Introduction

44

45 Artificial light at night (ALAN) is a rapidly increasing global phenomenon impacting the physiology
46 and behaviour of organisms [1], their interactions [2,3] and space use [4,5] as well as the
47 composition of species within and across communities [6–8]. It, therefore, has the potential to
48 drastically alter natural ecosystems, and has been proposed as a major driver of insect decline
49 [9,10].

50 To date, studies of the ecological impacts of ALAN have focussed almost exclusively on the responses
51 of animals near to individual bright sources of light (such as streetlights). However, ALAN also affects
52 ecosystems much further from areas of human activity via the phenomenon known as “skyglow”,
53 the diffuse and low-intensity artificial light that is reflected back to Earth by clouds and aerosols in
54 the atmosphere [11–13]. The illuminance of skyglow is often far larger than starlight [14], and can
55 approach the brightness of the full moon [15]. Furthermore, in areas affected by skyglow, overcast
56 nights are no longer dark [16], so the overall range of nighttime illuminance experienced by the

57 ecosystem is reduced by several orders of magnitude compared to natural conditions. For instance,
58 it has been shown that even low levels of artificial light intensities have the capacity to modify
59 foraging efficiency and the strength of interspecific interactions, which lead to corresponding
60 changes in community structure [2], highlighting the need to study the community- and ecosystem-
61 level effects of comparatively low-intensity skyglow [17,18]

62 To better address the effects of ALAN in general and skyglow in particular on biodiversity and
63 ecosystem functioning, we require a holistic understanding of the underlying ecological processes
64 that drive species' distributions and abundances, such as animal space use and biotic interactions.
65 Movement is a key mediator of these processes as it enables animals to explore their environment
66 for food, potential mates and suitable habitats. Unfortunately, the sensitivity of animal movement to
67 the presence of diffuse nighttime illuminance such as skyglow remains unclear.

68 Light pollution has been shown to have widespread effects on movement behaviours across several
69 spatial scales, such as migration [19–21], dispersal [22,23], as well as local movements within and
70 between habitats [24–27]. These effects can be diverse [8]; for instance, at local scales, there is
71 evidence of reduced [28–31] as well as increased movement activity [32] in response to ALAN, which
72 may be induced by the elevated risk or facilitation of predation, respectively. Moreover, animals may
73 shift their activity temporally with or without affecting their overall activity time budget [33–35].
74 Generally, the onset and duration of movement activity among visual predators depend on
75 illumination levels that facilitate successful foraging activity [36]. This implies that ALAN acts on the
76 temporal as well as spatial dimension of movement, with diverse knock-on effects for encounter
77 rates and interactions among hetero- and conspecifics.

78 Predation is a key ecological interaction that determines the structure and functioning of
79 ecosystems. ALAN can drive predation rates (i) by affecting spatiotemporal movement activity (i.e.
80 where and when they move) and local densities, and thus encounters between predator and prey, or
81 (ii) by affecting detection ability [37] (Figure 1). Among animals that rely on visual cues to orient
82 themselves or detect and capture their prey, visual acuity facilitates the processing of spatial
83 information and increases the minimum distances at which potential prey becomes visible. Visual
84 predators, particularly those that possess adaptations which increase visual acuity under the
85 ambient light conditions of their temporal niche [38], are expected to be particularly sensitive to
86 ALAN [2,39]. Habitat structure can modify the effects of ALAN on animal movement and foraging
87 behaviour by altering the trade-off between foraging success and predation risk [25,34,40], for
88 instance, by impeding predators' movements and their visual detection of prey or through the
89 provision of prey refuges [41]. Moreover, ALAN can change species' preferences for certain habitats

90 and thus influence space-use, habitat connectivity [4,5,26], and (co-)occurrence, with profound
91 effects on encounter probabilities. This demonstrates that ALAN has the potential to fundamentally
92 disrupt trophic interactions with implications for food webs, species distributions, biodiversity, and
93 ecosystem functioning (Figure 1).

94 We designed an Ecotron experiment with simulated diel light and moon cycles to elucidate the
95 interactive effects of skyglow and habitat structure on movement and predation. We continuously
96 tracked the movements of individual insects (792 individuals across seven beetle and one bug
97 species) within experimental grassland-patch landscapes using RFID tracking and measured
98 predation rates on artificial caterpillar prey dummies. We quantified how temporal- (Q1) and spatial
99 movement activity (Q2) respond to nighttime illuminance across a gradient from 0.001 lux (starlight)
100 to 30 lux (under a streetlight). Furthermore, we measured the effect of light on predation rates and
101 discuss their dependence on detection probabilities and encounter rates (Q3 & Q4) (Figure 1).

102

103 Methods

104

105 *General setup and experimental design*

106 We conducted our experiment at the iDiv Ecotron experimental facility, which is an indoor
107 mesocosm facility consisting of independent, experimental chambers called “EcoUnits” [42]. The
108 Ecotron is located in Bad Lauchstädt, Saxony-Anhalt, Germany, at the Experimental Research Station
109 of the Helmholtz Centre for Environmental Research (UFZ, 51.3917° N, 11.8762° E). Multiple
110 environmental conditions in the EcoUnits can be fully controlled (e.g., nutrient supply and irrigation).
111 Each EcoUnit has internal dimensions of 1.46 m × 1.46 m × 1.50 m (L × W × H, aboveground) and
112 1.24 m × 1.24 m × 0.80 m (L × W × H, belowground) with the soil surface area measuring 1.54 m²
113 [42]. We conducted the experiment in 12 EcoUnits from July to October 2020.

114 To assess the interactive effects of diffuse nighttime illuminance and landscape structure on animal
115 movement patterns, we established a patch-grassland system which consisted of four meadow
116 patches within each of the corners of an EcoUnit, separated by an area of bare ground (Figure 2a, b).
117 The EcoUnits were filled with 1.23 m³ of unsterilised and homogenised soil from the vicinity of the
118 iDiv Ecotron, and plant communities of 16 plant species were sown on February 4th 2020 (Table S1).
119 We allowed for a settlement phase of roughly 5 months before starting our measurements.

120 *Light treatment*

121 Across the 12 EcoUnits, we simulated diel light and moon cycles and added a treatment of diffuse
122 nighttime illuminance including the full range of observed skyglow intensities [44,45,16].

123 *Daylight*

124 The daytime lighting (manufactured by Roschwege GmbH, Germany) within all EcoUnits was set to
125 the same daylight settings. Photoperiods were adjusted every four weeks to approximate local
126 sunrise and sunset times throughout the duration of the experimental period. Daylight was gradually
127 (i.e. linearly) brightened or dimmed over the course of two hours before sunrise and sunset,
128 respectively. The maximum brightness of the daytime lighting was approximately 35,000 lux, which
129 corresponds roughly to a sunny day in Germany, and a light spectrum that approximates sunlight.

130 *Moonlight*

131 At night, moonlight within each EcoUnit was simulated by a single sunlike LED (SunLike3030 by Seoul
132 Semiconductor Co. Ltd., Korea) with a light spectrum that approximates sunlight (Figure S1). We
133 simulated moonlight because complete darkness is not a meaningful control [46,47], and organisms
134 have adapted to moon cycles over the course of evolutionary history. Moonlight intensities for clear-
135 sky conditions were modelled for the real time and location of the experiment using an astronomical
136 model of solar and lunar illuminance. The illuminance model calculates direct and diffuse
137 illuminance and was based on the model of Janiczek and DeYoung [48], with several enhancements
138 to increase accuracy. Illuminance of the moonlight LED was adjusted automatically every minute
139 using a Python script running on a raspberry pi, and could be adjusted to 57 illuminance levels
140 spanning from 0 lux (off) to the maximal modelled moonlight brightness of approximately 0.274 lux.

141 *Nighttime illuminance*

142 We established a skyglow treatment with a gradient of diffuse nighttime illuminance that spanned
143 from 0.0014 lux (slightly brighter than starlight) to 30 lux on a log₁₀ scale. The very upper end of our
144 gradient is brighter than the skyglow that is observed in nature today but might cover future
145 scenarios of diffuse nighttime illuminance. The levels of nighttime illuminance at both ends of the
146 gradient were replicated once. We used LED lights (type 2835 by HuiYuan Opto-Electronic Co. Ltd.,
147 China) with a typical blue light peak within their spectrum (Figure S2). This resulted in illuminance of
148 0.0014 (no pollution control), 0.0087, 0.028, 0.081, 0.1, 0.3, 0.94, 3.03, 9.88, and 30.31 lux (in the

149 absence of moonlight). Note that half of the gradient lies below the maximum brightness of the full
150 moon during the experimental phase (0.274 lux). We chose to cover these low light intensities for
151 two major reasons: (1) they represent light levels organisms naturally experience and have adapted
152 to for millions of years, and (2) reflect typical intensities of far-reaching skyglow [14]. To avoid
153 stunning animals by sudden changes in brightness, the treatment lights were always switched on
154 and off at sunrise and sunset, respectively (when daylight was at 50%). To avoid point sources of
155 light and simulate diffuse nighttime illuminance such as skyglow, the light was scattered using
156 diffusion foil.

157 All units were covered with black theatre curtains to block light from outside (see Fig. 2c). The
158 illuminance in the units was calibrated via a sky brightness measurement approach using a fisheye-
159 lens camera [16].

160 *Study animals*

161 We collected the insects for the experimental communities in the surrounding area of Leipzig,
162 Saxony, Germany (51.3213° N, 12.3964° E and 51.2799° N, 12.4119° E) from June to August 2020
163 using pitfall traps. Our species selection (Table S2) depended on seasonal densities and occurrences,
164 and consisted of seven species of carabid ground beetle (Coleoptera: Carabidae) which are primarily
165 crepuscular or nocturnal (Table S2, [49]): *Abax parallelus* (Duftschmid), *Calathus fuscipes* (Goeze),
166 *Carabus granulatus* (Linnaeus), *Carabus nemoralis* (Müller), *Harpalus rufipes* (De Geer), *Nebria*
167 *brevicollis* (Fabricius), and *Pterostichus melanarius* (Illiger) (Table S2). All species were housed in
168 separate containers which were bedded with moistened soil and foliage, and fed ad libitum with
169 beetle jelly from a commercial supplier (The Pet Factory, Germany) prior to the experiment. In total,
170 our Ecotron communities constituted a total of 792 RFID-tagged individuals from seven species
171 across two orders with body masses ranging from 47 mg (*Calathus fuscipes*) to 707 mg (*Carabus*
172 *nemoralis*, Table S2). We distributed them equally across the EcoUnits at densities that reflect a
173 natural abundance-mass relationship (Table S2).

174 *Movement tracking via RFID*

175 We used a Radio Frequency Identification (RFID) tracking system consisting of passive RFID-tags,
176 RFID-readers (transceivers), and a host system (controller) to track the movements of our study
177 animals (see [43] for details). We distributed 36 RFID sensors equally across patch and matrix areas
178 in the EcoUnits (4 sensors in each patch and 20 sensors in the matrix, Fig. 2b). Before adding the
179 study animals to the EcoUnits, we weighed and tagged the individuals with a unique RFID-tag. We

180 kept the insects at 4°C for 15 min before gluing the tag to the elytra of the beetles. We used
181 medium-sized (size: 8.3 x 8.3 x l 0.7 mm, reading range: 25 mm, mass: 35 mg, Murata LXMSAPHA17-
182 176) and small RFID-tags (size: 3.2 x 3.2 x 0.75 mm, reading range: 12 mm, mass: 20 mg, Murata
183 LXMS33HCNK-171), for large- (body mass > 200 mg) and small-bodied (body mass < 200 mg) species,
184 respectively. We recorded the tag-ID together with the identity and body mass of the individual.

185 Movement tracking inside the EcoUnits was performed across two temporal experimental blocks,
186 each corresponding to a period of approximately one lunar cycle (i.e., 28 days: experimental block I:
187 21.07.2020 - 18.08.2020, experimental block II: 15.09.2020 - 13.10.2020). Newly tagged individuals
188 were added a few days prior to the start of the respective experimental block for acclimatisation.
189 During the tracking periods, individuals were identified with a unique timestamp when crossing a
190 sensor and disturbances were minimised by only opening the EcoUnits once for the exchange of prey
191 dummies. Together with the exact position of the RFID sensor in the EcoUnit, this provides unique
192 spatiotemporal information for each tagged individual. We defined detections as distinct and only
193 counted them when they (1) occurred on different sensors or when (2) at least 10 seconds had
194 elapsed (without detection on the same sensor) between two consecutive detections on the same
195 sensor. This prevented the repeated detection of resting or dead animals. We used the number of
196 RFID detections as a measure of the movement activity of the community, which is the product of
197 local densities and individual movement.

198 *Predation rates*

199 Predation rates were estimated across the skyglow gradient by recording bite marks on prey
200 dummies [50–52]. We moulded artificial prey dummies from odourless, non-toxic green plasticine
201 (Noris 8421 by Staedler, Germany) to resemble model caterpillars of a standardised appearance (Fig
202 2d). We mounted 16 prey dummies on pins that were equally spaced within the four habitat patches
203 of each EcoUnit (Figure 2b) for two successive 14-day exposures within each 4-week temporal
204 experimental block. Two independent observers scored the prey dummies by identifying and
205 counting the bite marks left by carabid predators. Although there are limits to the precision of
206 identification [53], we were able to identify and group the parallel marks left by the mandibles of
207 carabid beetles in order to identify the number of successful attacks on individual prey dummies,
208 thereby quantifying predation rates during each 14-day exposure. This approach is likely only able to
209 elucidate predation rates of visual predators, rather than predators that search via olfaction.
210 Notably, visual hunters are the predators that are also likely to be affected by light.

211 *Statistical analysis*

212 We fitted generalised linear mixed effects models (GLMM) using the 'glmmTMB' package [54] in R
213 4.2.2 [55] to investigate the effects of ALAN as diffuse nighttime illuminance on the total movement
214 activity, space use, and predation rates in the insect communities. To test our first hypothesis, that
215 ALAN alters animal movement activity (Q1), we modelled the interactive effects of diffuse nighttime
216 illuminance and diel light cycle (day vs night) on species' movement activity. Movement activity was
217 estimated from the number of detections per day/night and analysed both on the community- and
218 species-level. To assess the effects of ALAN on animal space use during each phase of the diel light
219 cycle (Q2), we modelled the interactive effects of diffuse nighttime illuminance and habitat (patch vs
220 matrix) on species' movement activity (sum of detections per day/night). We used a negative
221 binomial distribution to account for overdispersion in the movement activity data and included the
222 temporal experimental block as a random intercept to account for temporal replication of the 4-
223 week experimental tracking within the same EcoUnit. To test whether the effect of nighttime
224 illuminance on total movement activity of the community is more driven by individual movement or
225 by local densities, we leveraged the individual-level information provided by the RFID tags to correct
226 the sum of detections for differences in local densities. Therefore, we included the number of
227 unique RFID tag detections, aggregated at the corresponding spatial and temporal scale (see Table
228 S3 for detailed information), as an offset term to the GLMM models reported in our supplementary
229 analyses. To evaluate if ALAN mediates the predation rates of visual predators by increasing the
230 detection of prey dummies (Q3), we modelled the effect of diffuse nighttime illuminance on
231 predation rates (bite counts per 14 days). Predation rates during each of the two 4-week
232 experimental blocks were estimated by counting and summing the number of bite marks left by
233 carabid beetles during two successive 14-day exposures. We modelled predation rates using a quasi-
234 Poisson distribution to account for overdispersion, and included the temporal experimental block as
235 a random intercept. To test our final hypothesis, that skyglow affects predation rates via an
236 increased encounter rate with prey (H4), we first aggregated the data on movement activity and
237 predation rates to comparable spatial scales (Fig 2b, Table S3): We modelled the effect of patch-level
238 movement activity (sum of detections per 14 days within each patch, Fig. 2b) on patch-level
239 predation rates (bite counts per 14 days within each patch) using a quasi-Poisson distribution. We
240 included the number of prey dummies recovered from each patch as an offset term, with the
241 temporal experimental block as a random intercept. The patch-level movement activity used in the
242 analyses of predation rates excluded detections of species that are not expected to leave bite marks,
243 i.e., species that are too small to reach mounted prey dummies (body size < 200 mg). Figures were
244 created using the R packages 'ggplot2' [56] and 'ggeffects' [57].

245

246 Results

247

248 We recorded a total of 25,378 RFID-detections across all experimental insect communities. With
249 regard to our first research question (Q1) on how movement activity (measured as the number of
250 detections per day/night, see Methods) is altered by skyglow, we found no significant effect of
251 nighttime illuminance on overall movement activity (Fig. 3a, slope = 0.014, $p = 0.530$, Table S4a).
252 However, with increased nighttime illuminance and despite some variability across species (see
253 Table S5), we found a significant decrease in movement activity at the community level during the
254 day (Fig. 3b, slope = -0.099, $p = 0.008$, Table S6a) and a significant increase in movement activity
255 during the night (Fig. 3b, slope = 0.069, $p = 0.014$, Table S6a). Together, these results imply a
256 temporal shift in activity from day to night without effects on the overall activity time budget.

257 Furthermore, we observed a spatial shift in movement activity of the insect community in response
258 to nighttime illuminance (Q2). During daytime, the movement activity within the matrix decreased
259 with nighttime illuminance (Fig. 4a, slope = -0.183, $p = 0.001$, Table S7a), while the movement
260 activity increased in the habitat patches (Fig. 4a, slope = 0.114, $p = 0.048$, Table S7a). This suggests a
261 shift in space use towards denser habitats during daytime. In contrast, we found that nighttime
262 movement activity increased within the matrix in response to the effect of nighttime illuminance
263 (Fig. 4b, slope = 0.072, $p = 0.028$, Table S8a), while no significant effect was observed within habitat
264 patches (Fig. 4b, slope = 0.018, $p = 0.653$, Table S7). This suggests that the increased nocturnal
265 movement activity of the insect community as well as the corresponding decreased movement
266 activity during day predominantly took place within the matrix.

267 To elucidate whether the effects of nighttime illuminance on movement activity are driven by
268 changes in individual movement or local densities, we performed all analyses (Q1 & Q2) with the
269 local densities as an offset. All results remained virtually identical (see Figures S3 & S4 and Tables
270 S4b, S6b, S7b & S8b). Furthermore, the number of detected individuals per EcoUnit during the
271 second half of each experimental block was not significantly affected by the diffuse nighttime
272 illumination (see Table S9), suggesting that abundances were unaffected by the light treatment.
273 Together, this indicates that nighttime illuminance drives the community-level movement activity
274 mainly through changes in individual movement.

275 Contrary to our expectations, there was no significant effect of nighttime illuminance on the overall
276 predation rate (estimated by the number of bite marks on individual prey dummies) by the
277 experimental insect community (Fig. 5a, slope = 0.029, $p = 0.127$, Table S10) (Q3). This reflects the
278 neutral effect of nighttime illuminance on the overall movement activity (Fig. 3a). In contrast to
279 movement activity (Fig. 3b), bite marks could not be associated with night or day, as they were only
280 collected every two weeks. However, we did find a strong correlation between patch-level predation
281 rate and movement activity (Fig. 5b, slope = 0.131, $p = 0.002$, Table S11), i.e., we counted
282 significantly more bite marks on prey dummies from patches that reported a higher movement
283 activity. Together with our results showing no effect of nighttime illuminance on local densities, this
284 supports our expectation that higher movement activity, which enables more frequent encounters
285 between predators and artificial caterpillar prey dummies, is the primary driver of predation rates
286 (Q4).

287

288 Discussion

289

290 We experimentally exposed artificial grassland communities to a gradient of diffuse nighttime
291 illuminance, and demonstrated that nighttime illuminance elicits spatiotemporal shifts in movement
292 and predation of insects. We found shifts in community-level movement activity from daytime to
293 nighttime (Q1) as well as shifts in habitat use from vegetated habitat patches to open habitat at
294 night (Q2). While we did not detect an overall response of predation rates to nighttime illuminance,
295 we deduce spatiotemporal shifts in predation rates via their strong correlation with patch-level
296 movement activity (Q3 & Q4).

297 We tracked the movements of individual animals within experimental insect communities using an
298 RFID sensor array in order to investigate their response to diffuse nighttime illuminance such as
299 skyglow (Q1). The lightweight, passive RFID tags [58] are well-suited to the tracking of small animals
300 such as insects [59]. Moreover, this approach enabled us to track the movement of the insects in
301 darkness as well as complex physical habitats, which is limited with other methods such as image-
302 based tracking [60]. Despite finding no effect of nighttime illuminance on the overall activity time
303 budget (Figure 3a), we did detect a temporal shift in movement activity from day to night (Figure
304 3b). Our community is composed primarily of crepuscular and nocturnal species that are likely to be
305 able to extend their temporal niche into the night when artificial light maintains their ability to see
306 and thus forage. As nocturnal foraging probably evolved to reduce competition and predation

307 pressure, crepuscular species may benefit from opportunities that reduce their interactions with
308 diurnal species, which explains the simultaneous reduction in movement activity during the day.
309 Furthermore, by keeping the total time budget constant, these species avoid an overall increase in
310 their total energy expenditure.

311 In addition to this temporal shift, we observed a change in the insect communities' space use in
312 response to skyglow (Q2), marked by a concomitant increase in nocturnal- and decrease in diurnal
313 movement activity within the bare-soil matrix. The dense vegetation within the habitat patches in
314 our experimental landscapes (Figure 2a) reduces light intrusion and visibility, in contrast to the open
315 ground of the interstitial matrix area. This can have important implications for animal movement
316 and foraging behaviour, for instance, by facilitating foraging or increasing predation risk [25,40,61].
317 The observed increase in movement activity in the matrix at night (Figure 4b) fits our interpretation
318 that crepuscular species shift their activity towards nocturnality due to increased foraging and
319 exploration opportunities. During daytime, not only does the overall movement activity decrease,
320 but there is an additional shift in activity from the matrix to the habitat patches in response to
321 increasing nighttime illuminance (Figure 4a). A shift towards nocturnal exploration activity could
322 result in a preference for habitat patches that provide protection from potential predators during
323 the day.

324 Predation rates can be driven by detection success as well as by the probability of encounters
325 between predators and their prey. Higher detection ability facilitated by increased visibility under
326 ALAN should generally lead to higher predation rates. However, as we did not find a significant effect
327 of nighttime illuminance on the total number of attacks (bite counts per 14 days) on prey dummies
328 (Q3, Figure 5a), we can deduce that there is likely also no significant effect on predation rates via
329 detection probability within our experimental grassland communities. In addition to the predators'
330 movement activity, local densities, and detection success, predation rates could also be influenced
331 by the predators' decision to forage or the behaviour of the prey. In contrast to studies that employ
332 immobile prey dummies, future studies that simultaneously track the movement of prey could
333 elucidate the role of prey responses to ALAN in determining the outcomes of predator-prey
334 interactions (e.g. [61]). Moreover, our use of prey dummies does not allow us to differentiate
335 daytime and nighttime predation rates; nevertheless, we found a strong spatiotemporal association
336 between patch-level predation rate and movement activity (Q4). Therefore, based on the absence of
337 an effect of nighttime illuminance on movement activity (Figure 3a), we expect a corresponding
338 absence of an effect on predation rates when measured across day and night, which is supported by
339 our analysis (Figure 5a). This highlights that nighttime illuminance drives predation through the rate

340 of encounters between predators and their prey rather than via the predators' visual detection of
341 prey. Together with the observed temporal and spatial shift in movement activity (Figure 3b and
342 Figure 4b, respectively), our results suggest that diffuse nighttime illuminance leads to a congruent
343 spatiotemporal shift in predation rates.

344 Organisms show diverse and context-dependent responses to ALAN [8,62], which was also reflected
345 by some variability in how the species in our experimental community responded. This might
346 translate to distinct community-level responses depending on, for example, the species composition
347 of the community and the ecosystem it is integrated within (e.g. [63]). Here, we could show that in
348 grasslands, diffuse nighttime illuminance such as skyglow can influence movement and by extension
349 prey encounters of seven predominantly crepuscular species. Our focus on diffuse nighttime
350 illuminance and the range of our experimental gradient covers most of the real-world light
351 conditions from natural starlight to cities [14], suggesting that our results are relevant for light
352 pollution experienced by invertebrates in open habitats such as grasslands and agricultural fields
353 throughout the world. We were able to show that even low levels of nighttime illuminance can
354 cause substantial changes in animal movement and consequently predation rates. For instance,
355 more than 50% of the observed change in movement activity (Fig. 3b) across the whole gradient of
356 nighttime illuminance occurred at illuminance levels that were below that of an average full moon
357 (approx. 0.3 lux). This strong response to low illuminance levels is to be expected among organisms
358 that have adapted to respond to subtle changes in illuminance such as the moon cycle [61,64] and to
359 life under starlight [65].

360 Our evidence for skyglow affecting fundamental ecological processes such as the movement of
361 invertebrates and predation suggests cascading and far-reaching repercussions for landscape
362 connectivity, biodiversity and ecosystem functioning. For instance, the shift of predator activity to
363 open habitats, as shown here, could increase the predation risk of dispersing nocturnal prey and
364 diminish landscape connectivity. This is particularly relevant for animals which rely on behavioural
365 shifts towards nocturnal activity to buffer against thermal and water stress [66–68]. Spatiotemporal
366 shifts in predation rates can also have strong implications for species interactions, either by rewiring
367 food webs or by modifying the strengths of interactions. This can fundamentally change the
368 structure of food webs and their stability [69]: A temporal shift in the activity of crepuscular species
369 into the night as found in our study could lead to new interactions with nocturnal species, and in
370 turn, cause interactions with diurnal species to be lost or weakened. Furthermore, changes to
371 encounter probabilities and, consequently, interaction strengths can alter energy fluxes in food
372 webs. For instance, an increase in predation rates at night could lead to higher energy fluxes, with

373 knock-on effects on the stability of nocturnal as well as diurnal food webs [70–72]. Such
374 spatiotemporal changes in movement activity have also been shown to affect other types of
375 ecological networks such as plant-pollinator and host-parasitoid networks [2,7,73]. This suggests
376 that skyglow, as a recent and intensifying anthropogenic disturbance [13], has far-reaching
377 consequences, and the potential to fundamentally disrupt natural communities and the services
378 they provide.

379

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389

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581

582 Figure captions

583

584 **Figure 1.** Concept illustrating how ALAN may cascade from physiological and behavioural processes
585 (A), to interactions and functions (B), and ultimately to community and ecosystem responses (C).
586 Orange circles and black arrows indicate our research questions: How does ALAN affect temporal-
587 (Q1) and spatial movement activity (Q2) in a patchy habitat? Does ALAN predominantly affect
588 predation rates through detection probabilities (Q3), or are predation rates rather driven by the
589 effects of spatiotemporal movement activity on encounter rates between predator and prey (Q4)?

590

591 **Figure 2.** Experimental design (a) Interior view of the grassland habitat patches established in an
592 EcoUnit, (b) schematic representation of the patch design highlighting the distribution of RFID
593 sensors and prey dummies across the EcoUnit. (c) EcoUnits covered with black theatre curtains to
594 prevent cross-contamination with light. (d) Pictures of a beetle with medium-sized RFID tag (taken
595 from an experiment using the same tracking approach and setting, [43]) and an artificial caterpillar
596 prey dummy with bite marks.

597

598 **Figure 3.** Movement activity (sum of detections per day/night) in response to nighttime illuminance.
599 EcoUnit-level daily movement activity (per 24 hours) (a) and daytime (light blue) and nighttime
600 movement activity (dark blue) (b). Dashed lines represent non-significant relationships ($p > 0.05$).
601 Shaded regions represent 95% confidence intervals.

602

603 **Figure 4.** Movement activity (sum of detections per day/night) in habitat patches (green) and matrix
604 (orange) in response to nighttime illuminance during the day (a) and during night (b). Dashed lines
605 represent non-significant relationships ($p > 0.05$). Shaded regions represent 95% confidence
606 intervals.

607

608 **Figure 5.** (a) EcoUnit-level predation rate (bite counts per 14 days) in response to nighttime
609 illuminance. (b) Patch-level predation rate (bite counts per 14 days per patch) in response to patch-
610 level movement activity (sum of detections per 14 days per patch, Fig. 2). Dashed lines represent
611 non-significant relationships ($p > 0.05$). Shaded regions represent 95% confidence intervals.

612

613 Data accessibility

614 The underlying data are available in an open access repository at

615 <https://doi.org/10.5281/zenodo.8017438> [74]

616

Supplementary information

Insect communities under skyglow: diffuse nighttime illuminance induces spatiotemporal shifts in movement and predation

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Tables S1-S11

Figures S1-S4

Table S1. Sown plant species in each EcoUnit. For every species 1000 seeds m⁻² were sown.

Species	Sown (mg/1000 seeds)
<i>Bromus hordeaceus</i>	3443.0
<i>Campanula patula</i>	65.0
<i>Cynosurus cristatus</i>	556.7
<i>Festuca pratensis</i>	2467.3
<i>Festuca rubra</i>	869.3
<i>Lathyrus pratensis</i>	10889.5
<i>Leucanthemum vulgare</i>	1004.0
<i>Lotus corniculatus</i>	1244.3
<i>Luzula campestris</i>	681.3
<i>Medicago lupulina</i>	2079.2
<i>Plantago media</i>	384.9
<i>Prunella vulgaris</i>	796.3
<i>Ranunculus repens</i>	1804.8
<i>Trifolium repens</i>	692.1
<i>Veronica chamaedrys</i>	205.4
<i>Vicia sepium</i>	12995.5

Table S2. Body masses and abundances of animals used for RFID movement tracking.

Species	Experimental block I (21.07. - 18.08.2020)		Experimental block II (15.09. - 13.10.2020)	
	Mean body mass [mg]	Abundance per unit	Mean body mass [mg]	Abundance per unit
<i>Abax parallelus</i>	302	2	269	5
<i>Calathus fuscipes</i>	67	6	75	6
<i>Carabus granulatus</i>	NA	NA	273	1
<i>Carabus nemoralis</i>	538	5	550	3
<i>Harpalus rufipes</i>	112	2	128	6
<i>Nebria brevicollis</i>	69	6	NA	NA
<i>Pterostichus melanarius</i>	151	4	NA	NA

Table S3. Overview of the response and predictor variables used to answer our four research questions (Q1-Q4) including units as well as spatial and temporal aggregation level. Aggregation on patch level means aggregation across the four individual patches in an EcoUnit contrary to aggregation on individual patch-level as in Q4.

Question	Response	Unit - response	Aggregation level - response		Predictor	Unit - predictor	Aggregation level - predictor	
			Spatial	Temporal			Spatial	Temporal
Q1	activity	number of detections	EcoUnit	day/night	ALAN	lux	NA	NA
Q2	activity	number of detections	matrix/ patch	day/night	ALAN	lux	NA	NA
Q3	predation	number of bites	none	14 days	ALAN	lux	NA	NA
Q4	predation	number of bites	individual patch	14 days	activity	number of detections	individual patch	14 days

Table S4a. Generalised linear mixed effects model results of the influence of nighttime illuminance on EcoUnit-level daily movement activity (sum of detections per day/night) (Q1). Estimates are given on ln-scale, intercept is given at 1 lux.

Predictor	Estimate	Std. Error	p-value
Intercept	2.743	0.347	< 0.001
Slope: log10 nighttime illuminance	0.014	0.022	0.530

Table S4b. Generalised linear mixed effects model results of the influence of nighttime illuminance on EcoUnit-level individual daily movement (sum of detections per day/night, corrected for individual densities) (Q1). Estimates are given on ln-scale, intercept is given at 1 lux.

Predictor	Estimate	Std. Error	p-value
Intercept	2.793	0.253	< 0.001
Slope: log10 nighttime illuminance	0.025	0.022	0.247

Table S5. Estimates for species-level responses of daytime and nighttime movement activity to nighttime illuminance. Insufficient data corresponds to species with fewer than 10 data points. Estimates are given on ln-scale, intercept is given at 1 lux.

		Intercept			Slope: log ₁₀ nighttime illuminance		
		Estimate	Std. Error	p-value	Estimate	Std. Error	p-value
<i>Abax parallelus</i>	Day	2.833	0.152	<0.001	-0.036	0.045	0.414
	Night	2.890	0.144	<0.001	0.065	0.034	0.053
<i>Calathus fuscipes</i>	Day	0.907	0.188	<0.001	-0.075	0.055	0.169
	Night	1.247	0.140	<0.001	-0.115	0.041	0.005
<i>Carabus nemoralis</i>	Day	2.969	0.187	<0.001	0.270	0.121	0.024
	Night	3.158	0.119	<0.001	0.293	0.059	<0.001
<i>Carabus granulatus</i>	Day	3.624	0.122	<0.001	-0.483	0.098	<0.001
	Night	4.016	0.107	<0.001	-0.120	0.081	0.143
<i>Nebria brevicollis</i>	Day	0.404	0.523	0.440	insufficient data		
	Night	0.770	0.118	<0.001	0.044	0.088	0.612
<i>Pterostichus melanarius</i>	Day	0.522	0.196	0.007	0.238	0.150	0.113
	Night	1.006	0.125	<0.001	-0.314	0.089	<0.001
<i>Harpalus rufipes</i>	Day	-0.366	1.118	0.743	insufficient data		
	Night	1.004	0.147	<0.001	-0.071	0.092	0.442

Table S6a. Generalised linear mixed effects model results of the interaction between nighttime illuminance and diel light cycle (day vs night) on movement activity (number of sensor crossing events per day/night) (Q1). Estimates are given on ln-scale, intercept is given at 1 lux.

Predictor	Estimate	Std. Error	p-value
Intercept [day]	2.527	0.362	< 0.001
Intercept [night]	2.840	0.360	< 0.001
Slope: log10 nighttime illuminance [day]	- 0.099	0.038	0.008
Slope: log10 nighttime illuminance [night]	0.069	0.028	0.014

Table S6b. Generalised linear mixed effects model results of the interaction between nighttime illuminance and diel light cycle (day vs night) on individual movement (number of sensor crossing events per day/night, corrected for individual densities) (Q1). Estimates are given on ln-scale, intercept is given at 1 lux.

Predictor	Estimate	Std. Error	p-value
Intercept [day]	2.736	0.264	< 0.001
Intercept [night]	2.809	0.260	< 0.001
Slope: log10 nighttime illuminance [day]	- 0.103	0.038	0.007
Slope: log10 nighttime illuminance [night]	0.090	0.026	< 0.001

Table S7a. Generalised linear mixed effects model results of the interaction between nighttime illuminance and habitat (patch vs matrix) on daytime movement activity (sum of detections per day/night) (Q2). Estimates are given on ln-scale, intercept is given at 1 lux.

Predictor	Estimate	Std. Error	p-value
Intercept [matrix]	2.649	0.336	< 0.001
Intercept [patch]	2.157	0.341	< 0.001
Slope: log10 nighttime illuminance [matrix]	-0.183	0.054	< 0.001
Slope: log10 nighttime illuminance [patch]	0.114	0.058	0.048

Table S7b. Generalised linear mixed effects model results of the interaction between nighttime illuminance and habitat (patch vs matrix) on individual daytime movement (sum of detections per day/night, corrected for individual densities) (Q2). Estimates are given on ln-scale, intercept is given at 1 lux.

Predictor	Estimate	Std. Error	p-value
Intercept [matrix]	2.774	0.245	< 0.001
Intercept [patch]	2.510	0.254	< 0.001
Slope: log10 nighttime illuminance [matrix]	-0.202	0.050	< 0.001
Slope: log10 nighttime illuminance [patch]	0.127	0.058	0.030

Table S8a. Generalised linear mixed effects model results of the interaction between nighttime illuminance and habitat (patch vs matrix) on nighttime movement activity (sum of detections per day/night) (Q2). Estimates are given on ln-scale, intercept is given at 1 lux.

Predictor	Estimate	Std. Error	p-value
Intercept [matrix]	3.089	0.278	< 0.001
Intercept [patch]	2.178	0.282	< 0.001
Slope: log10 nighttime illuminance [matrix]	0.072	0.033	0.028
Slope: log10 nighttime illuminance [patch]	0.018	0.041	0.653

Table S8b. Generalised linear mixed effects model results of the interaction between nighttime illuminance and habitat (patch vs matrix) on individual nighttime movement (sum of detections per day/night, corrected for individual densities) (Q2). Estimates are given on ln-scale, intercept is given at 1 lux.

Predictor	Estimate	Std. Error	p-value
Intercept [matrix]	3.012	0.199	< 0.001
Intercept [patch]	2.253	0.206	< 0.001
Slope: log10 nighttime illuminance [matrix]	0.087	0.032	0.008
Slope: log10 nighttime illuminance [patch]	0.072	0.041	0.077

Table S9. Linear model results of the effect of nighttime illuminance [lux] on the number of detected individuals for the second half of each temporal experimental block. Estimates are given on ln-scale, intercept is given at 1 lux.

Predictor	Estimate	Std. Error	p-value
Intercept	6.157	1.926	0.048
Slope: log10 nighttime illuminance	-0.278	0.213	0.198

Table S10. Generalised linear mixed effects model results of the influence of nighttime illuminance on EcoUnit-level predation rate (bite counts per 14 days) (Q3).

Predictor	Estimate	Std. Error	p-value
Intercept	1.421	0.037	< 0.001
Slope: log ₁₀ nighttime illuminance	0.029	0.019	0.127

Table S11. Generalised linear mixed effects model results of the influence of individual-patch-level movement activity (sum of detections) on individual-patch-level predation rate (bite counts per 14 days per individual patch) (Q4).

Predictor	Estimate	Std. Error	p-value
Intercept	1.308	0.048	< 0.001
Slope: log10 movement activity	0.131	0.042	0.002

Figure S1. Spectrum of the LED used to simulate moonlight in the EcoUnits.

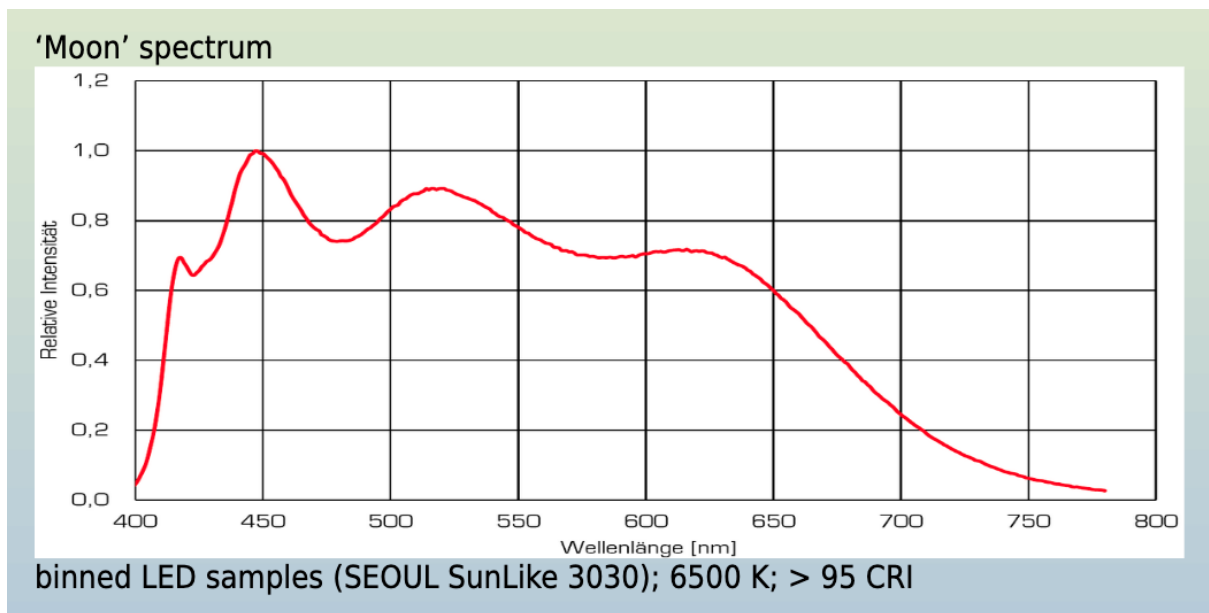
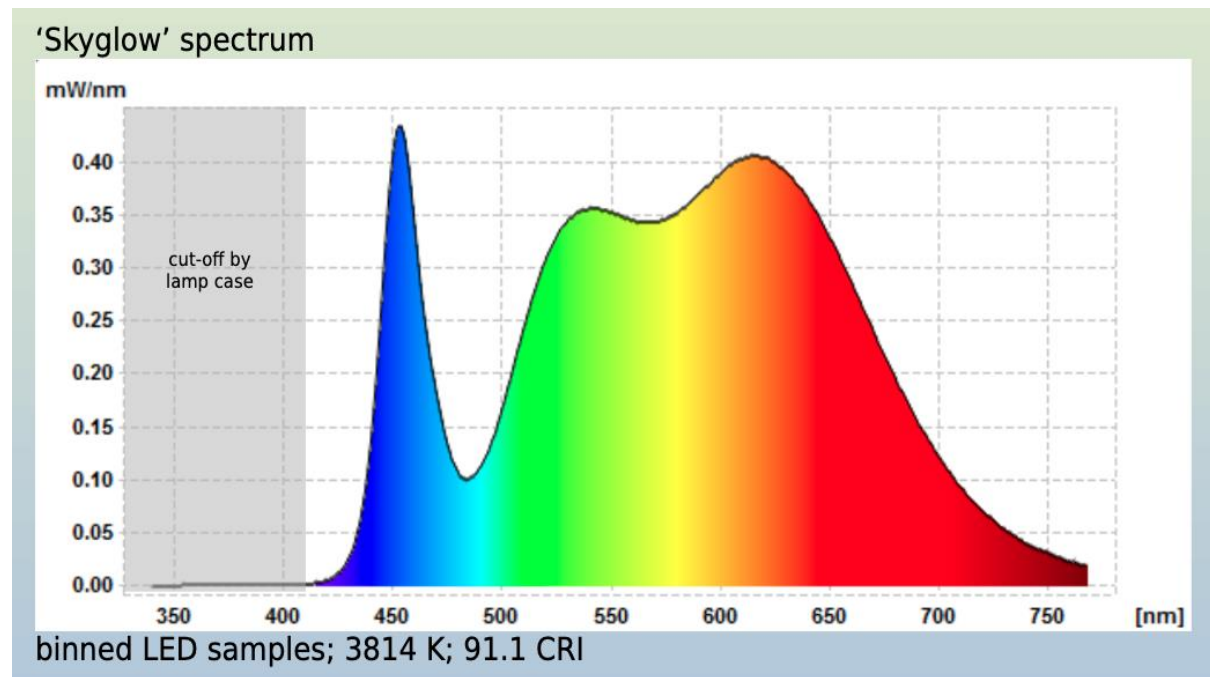


Figure S2. Spectrum of the LED used to simulate skyglow in the EcoUnits.



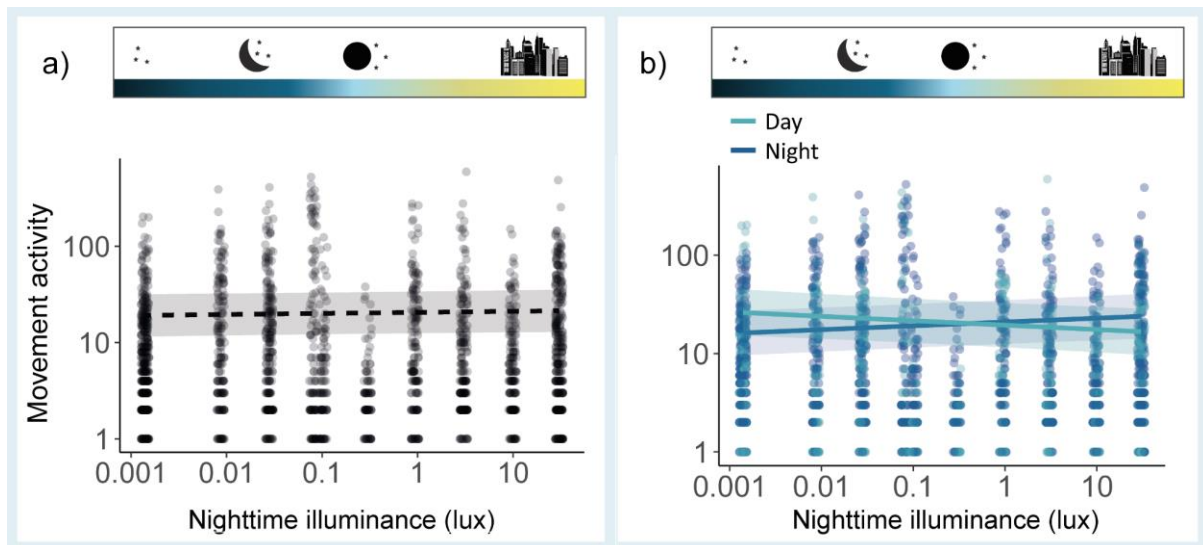


Figure S3. Individual movement (sum of detections per day/night, corrected for individual densities) in response to nighttime illuminance. EcoUnit-level daily movement (per 24 hours) **(a)** and daytime (light blue) and nighttime movement (dark blue) **(b)**. Dashed lines represent non-significant relationships ($p > 0.05$). Shaded regions represent 95% confidence intervals.

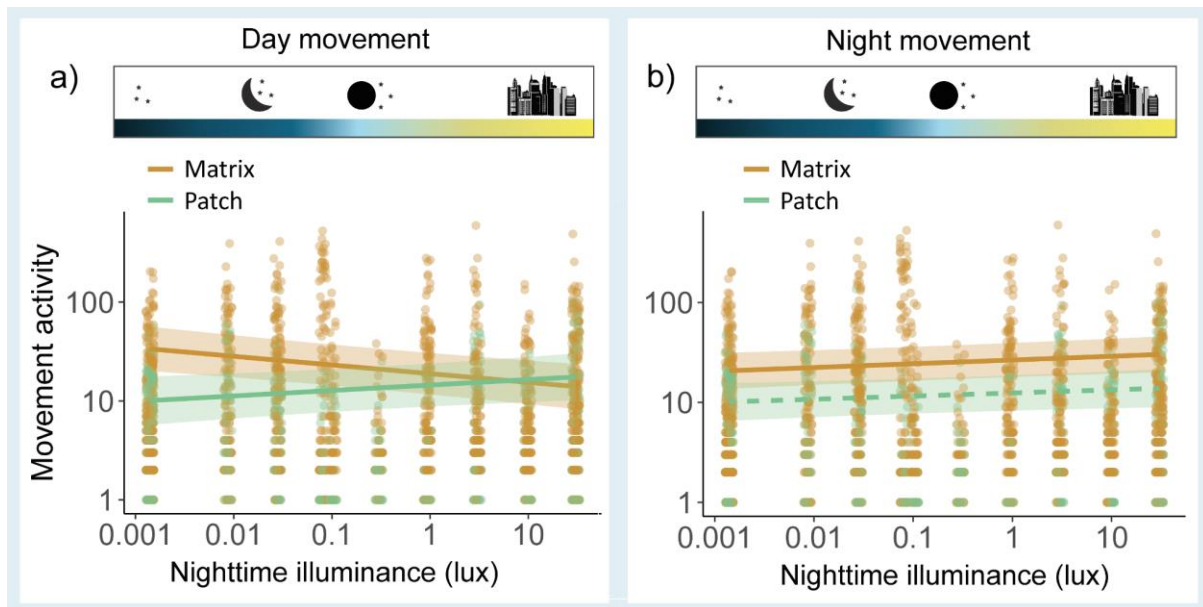


Figure S4. Individual movement (sum of detections per day/night, corrected for individual densities) in habitat patches (green) and matrix (orange) in response to nighttime illuminance during the day (a) and during night (b). Dashed lines represent non-significant relationships ($p > 0.05$). Shaded regions represent 95% confidence intervals.