

1 **Title:** Fear on the Landscape: How human activity shapes wildlife habitat use in protected areas in
2 Tasmania

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10 **Author contributions**

11 Laura M. Cardona, Jessie C. Buettel and Barry W. Brook conceived the ideas and designed
12 the methodology. Jessie C. Buettel and Barry W. Brook contributed to data analysis and to drafts of the
13 paper. Laura M. Cardona analysed data and led the writing of the manuscript. All authors contributed
14 critically to the drafts and gave final approval for publication.

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21 **Abstract**

22 The growing enthusiasm for outdoor recreation has prompted questions about the effects of different
23 forms of human activity on the habitat use of both predators and prey. Here, we used time-to-event
24 camera trap data from a large-scale survey in Tasmanian protected areas to investigate the influence of
25 motorised (vehicles) and non-motorised (hikers, joggers, and cyclists) recreation on wildlife return times
26 - the time between consecutive detections of the same species – during the warm and cold season.
27 Specifically, we focused on the Tasmanian devil (*Sarcophilus harrisii*), the apex predator in this system,
28 and four common herbivore species: brushtail possums (*Trichosurus vulpecula*), Tasmanian pademelons
29 (*Thylogale billardierii*), Bennett’s wallabies (*Notamacropus rufogriseus*), and bare-nosed wombats
30 (*Vombatus ursinus tasmaniensis*). We found that more motorised and non-motorised events delayed
31 Tasmanian devil return times in both seasons. For herbivores, all species showed longer return times
32 under motorised activity irrespective of season, whereas non-motorised activity had no significant effect
33 on wallabies during the cold season. These results indicate that predator and prey perceive both forms
34 of human activity as a source of risk. This study demonstrates the utility of time-to-event data from
35 camera traps for testing the influences of different disturbance sources on wildlife. These insights are
36 vital for managers striving to balance public access with conservation goals in protected areas.

37

38 **Introduction**

39 Protected areas (PAs) are an important governance mechanism for wildlife conservation (Leader-
40 Williams *et al.*, 1990; Lockwood *et al.*, 2006; Dudley *et al.*, 2010). They aim to act as refuges for species
41 and maintain ecological processes such as predator-prey systems (Arcese *et al.*, 1997). In Tasmania,
42 Australia, the PA network is critically important for biodiversity conservation, covering nearly half of the
43 island’s total land area, which is designated as reserves or national parks (Mendel *et al.*, 2002;

44 Davidson *et al.*, 2021). This extensive network has provided refuge for iconic species that are extinct on
45 the mainland, including the world's largest extant marsupial carnivore, the Tasmanian devil (*Sarcophilus*
46 *harrisii*) (Hollings *et al.*, 2014). Many of these PAs also accommodate public access for recreation through
47 designated road networks (Tasmania Parks and Wildlife Service 2025) and, as such, have
48 historically operated under a dual mandate: to provide opportunities for human activity while ensuring
49 wildlife conservation.

50

51 In recent years, tourism to the island and the popularity of outdoor recreation has increased, and so too
52 have visitation and length of stay to PAs (Balmford *et al.*, 2009; Hardy *et al.*, 2020; Beery *et al.*, 2021;
53 Tasmania Parks and Wildlife Service 2024). Increased recreation has been shown to affect the behaviour
54 and spatiotemporal distribution of wildlife species (Larson *et al.*, 2016). In some systems, species have
55 shifted to more nocturnal activity (Lewis *et al.*, 2021b; Salvatori *et al.*, 2023), delayed their return to
56 trails and roads (Naidoo *et al.*, 2020a; Visscher *et al.*, 2023), and reduced their use of these features with
57 increased human presence (Gump *et al.*, 2023). Other species have been found to use increase their
58 occupancy and habitat use of recreational trails (Lewis *et al.*, 2021b; Gump *et al.*, 2023), or to shift their
59 activity patterns to better align with periods of high human activity (Sytsma *et al.*, 2022).

60

61 Understanding how human activity affects wildlife behaviour and distribution in PAs is important for
62 evaluating their effectiveness in maintaining ecosystem functioning (Reed *et al.*, 2008; Sytsma *et*
63 *al.*, 2022) and devising effective management strategies (Monz *et al.*, 2013). Nevertheless, such
64 assessments are challenging, because wildlife responses to recreation are shaped by multiple
65 factors (Marion *et al.*, 2020). These responses depend not only on the intensity of human activity but
66 also on its type, the spatial and temporal context (e.g. season), and species-specific ecological and life

67 history traits (Tablado *et al.*, 2017). For instance, some mammal species have been shown to be deterred
68 by non-motorised activities (e.g. hikers and bicycles) but attracted to motorised activities (e.g.
69 vehicles) (George *et al.*, 2006b; Procko *et al.*, 2022; Gump *et al.*, 2023). Predator and prey responses to
70 human activity have been found to vary across systems, ranging from mutual avoidance to mutual
71 attraction (Van Scoyoc *et al.*, 2023), and cases where prey are attracted to humans for resources or
72 safety while predators avoid them (Berger 2007; Rogala *et al.*, 2011b).

73

74 To address this challenge, ground-deployed autonomous technology, such as camera traps (CTs), offers a
75 promising approach. CTs are well-established tools that allow researchers to survey multiple mammal
76 species continuously across various landscapes and over extended periods (Swann *et al.*, 2011;
77 Steenweg *et al.*, 2017; Kays *et al.*, 2020). Recent advances in machine learning approaches and artificial
78 intelligence to automate image identification and anonymise human records, have revolutionised the
79 type of information that can be extracted and analysed from such data (Miller *et al.*, 2017; Tabak *et*
80 *al.*, 2019; Fennell *et al.*, 2023; Vélez *et al.*, 2023). Simultaneously, the proliferation of statistical and
81 computational models for camera data has enabled robust approaches to assess species behaviour,
82 habitat use and interactions while controlling for confounding factors (Sollmann 2018). As a result, CTs
83 applications have expanded to include quantification of human presence and activities facilitating better
84 understanding of human-wildlife interactions (Miller *et al.*, 2017; Cardona *et al.*, 2024).

85

86 To-date however, there has been relatively limited application of CTs to monitor wildlife responses
87 to different types of human activity (Cardona *et al.*, 2024). Furthermore, studies have often pooled
88 wildlife data by seasons, weeks or days and additional research on finer-scale changes (e.g. hourly) is
89 needed (Naidoo *et al.*, 2020a; Blount *et al.*, 2021; Gump *et al.*, 2023; Marion *et al.*, 2024). Time-to-event

90 data obtained from CTs, provides a powerful approach to detecting short term shifts in wildlife temporal
91 and spatial use of habitats that otherwise would not be detected when pooling data into days or weeks.
92 By calculating the time elapsed between a specific event (e.g. species A) and its subsequent detection (A
93 to A) (Niedballa *et al.*, 2019), changes in the time it takes for an animal to reuse an area can be
94 identified and the impact of other wildlife and/or human activity (e.g. species B between A to A) can be
95 explored (Paull *et al.*, 2012; Visscher *et al.*, 2023; Ferry *et al.*, 2024).

96

97 In this study, we used camera-trap time-to-event data to assess short-term changes in wildlife
98 spatiotemporal habitat use in response to different types of human activity in Tasmanian PAs.
99 Specifically, we examined the effects of motorised and non-motorised activities on the time it took the
100 apex predator, the Tasmanian devil, and four herbivore species to return to a previously visited CT site
101 (return time). These herbivores included medium-to-large marsupials species that are also prey of the
102 Tasmanian devil (Pemberton *et al.*, 2008; Andersen *et al.*, 2017): brushtail possum
103 (*Trichosurus vulpecula*, hereafter 'possum'), Bennett's wallaby (*Notamacropus rufogriseus*, hereafter
104 'wallaby'), red-bellied or Tasmanian pademelon (*Thylogale billardierii*, hereafter 'pademelon'), and bare-
105 nosed or 'common' wombat (*Vombatus ursinus tasmaniensis*, hereafter 'wombat'). Despite the common
106 status and wide distribution of these species in Tasmania, information on their responses to recreational
107 activity is relatively limited. Given evidence that wildlife exhibit spatiotemporal avoidance of increased
108 vehicle and pedestrian presence (Frid *et al.*, 2002; Gaynor *et al.*, 2018), we predicted that both
109 motorised and non-motorised human activities would prolong the return times of the studied species.

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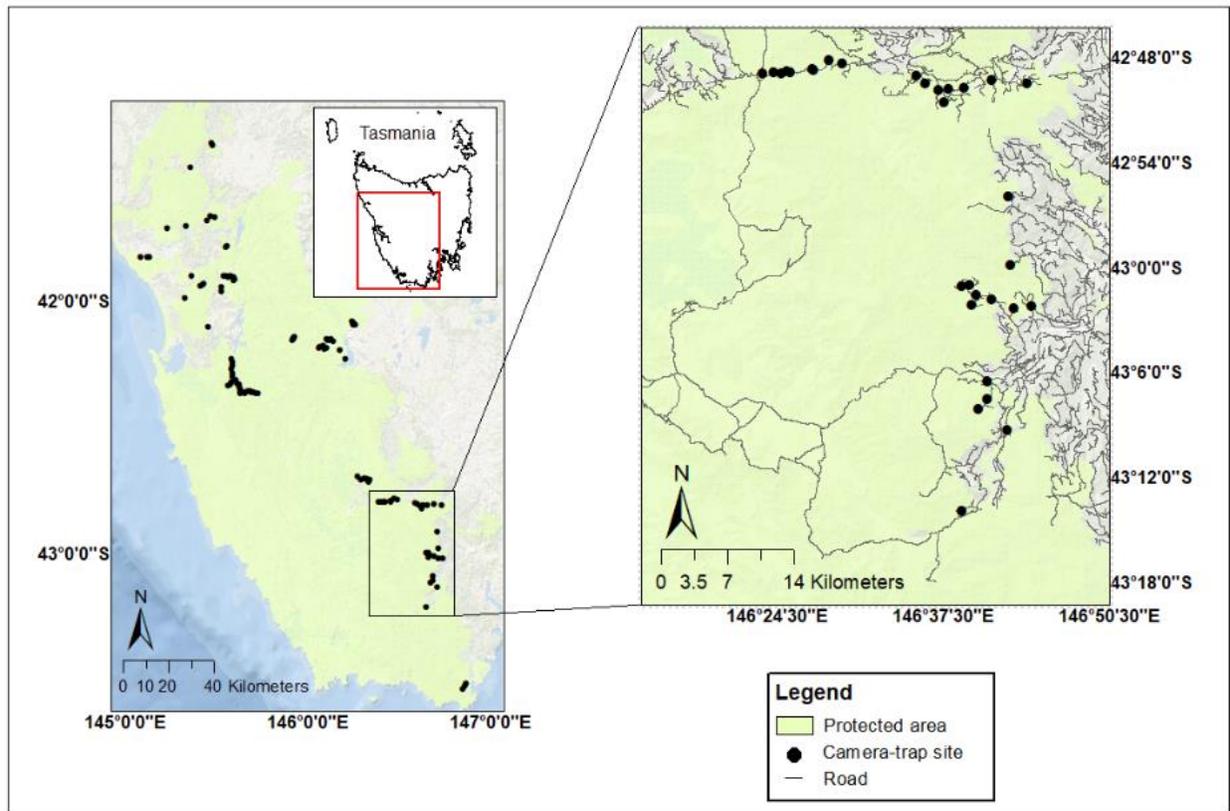
112

113 **Methods.**

114 **Existing field sites and camera deployment**

115 We used time-to-event data derived from a pre-existing camera trap (CT) network established in 2018
116 for the purpose of wildlife monitoring project across Tasmania, Australia. The study involved 119 camera
117 sites (model: Cuddeback Xchange 1279) located in protected areas (PAs) across the southeastern, central
118 highlands, northwestern and western regions (Figure 1). These PAs comprised National Parks,
119 Conservation areas, and other reserves (see Table S1 for the list of protected areas), and covered a
120 variety of habitat types, from temperate forests such as Eucalypt woodlands and tall open forests, to
121 grasslands, herb lands, sedgeland, bushlands, rainforests, vine thickets, low closed forests, and tall
122 closed shrublands. A wide range of non-consumptive human activities take place within these areas
123 during both day and night, such as camping, hiking, vehicle driving, dog walking, fishing, mountain
124 biking, four-wheel driving, all-terrain vehicle (ATV) driving and boating. These activities are regulated
125 through existing management policies and educational programs (Tasmanian government 2021). Native
126 forest extraction and wildlife hunting is prohibited, and activities such as four-wheel-driving, walking
127 dogs and hiking are restricted to designated areas and roads. Over the past decade, public interest in
128 outdoor recreation and nature-based activities in Australia has increased (Hardy *et al.*, 2020; Ecotourism
129 Australia 2023), resulting in higher visitation to many of these PAs (Tasmania Parks and Wildlife Service
130 2024).

131



133

134 **Figure 1.** Map of Tasmania showing the camera trap sites located along unsealed roads used in this study
 135 (black points). The green shaded area indicates the boundary of the protected areas (obtained from
 136 the Tasmanian Reserve Estate spatial layer). The map on the right shows an example of the camera
 137 deployments at a finer spatial scale.

138

139 Cameras were placed along unsealed roads (gravel or dirt roads). We acknowledge that road- based
 140 designs limit results to the habitat use of wildlife and humans to roads rather than the whole study area.
 141 Nevertheless, restricting cameras to roads eliminated confounding effects associated with edge-effects,
 142 such as shifts in community composition and variation in detection probabilities, that arise when
 143 sampling designs include both on- and off-road locations (Fuentes-Montemayor *et al.*, 2009; Bötsch *et*

144 *al.*, 2018b; Greco *et al.*, 2025). Study designs that exploit roads are also more cost effective (Geyle *et*
145 *al.*, 2020), efficient for detecting large mammals and carnivores (Mann *et al.*, 2015; Tanwar *et al.*, 2021),
146 and allow researchers to directly and simultaneously monitor human activity and wildlife (Wolf *et*
147 *al.*, 2012; Miller *et al.*, 2017; Oberosler *et al.*, 2017; Salvatori *et al.*, 2023; Procko *et*
148 *al.*, 2024). Accordingly, this approach was functional to our research aims.

149

150 Cameras were at least 100 m apart (typically 250 m to 5 km), not baited and fixed to stakes or trees at an
151 average height of 30 cm off the ground, adjusted to target medium to large mammal species. They were
152 equipped with a passive motion sensor and either an infra-red or white flash, depending on the
153 perceived risk of theft in the area. Infra-red cameras were programmed with 30-second delay for both
154 day and night, while white flash units operated with 30-second delay during the day and one minute at
155 night. Cameras were re-visited and serviced every 4-6 months to retrieve the images, replace batteries,
156 remove vegetation obstructing the lens, and check for any theft occurrences. Camera-trap imagery for
157 this study was collected over a four-year period, from July 2018 to June 2022. A total of 92 cameras were
158 deployed in 2018 (and remained in place throughout the study), with an additional 23 cameras added in
159 2019, followed by one in 2020, and three in 2021. Before the end of the four-year period a total of 18
160 CTs were stolen. In total, the 119 camera traps operated for 93,622 days (average of 787 days, or 2.16
161 years, of cumulative operation per camera site).

162

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166 **Data analysis**

167 **Determining independent events of wildlife and humans**

168 To address temporal autocorrelation, we first identified independent events (hereafter ‘events’) in the
169 camera-trap imagery of humans and wildlife. Independent events for wildlife species – possums,
170 wallabies, pademelons, wombats and Tasmanian devils – were defined as detections of the same species
171 occurring at least 30 minutes apart at each camera-trap site. This threshold, based on prior studies in
172 similar mammal communities (Cunningham *et al.*, 2019), minimises the likelihood of repeat detections
173 of the same individual or group. Applying this criterion resulted in 10,257 events for pademelons, 8,034
174 for possums, 8,943 for wombats, 3,126 for wallabies, and 9,230 for Tasmanian devils.

175

176 Camera trap imagery was processed using the open-source object detection
177 software MegaDetector v5 (Beery *et al.*, 2019a), using a confidence threshold of 50% (Brook *et*
178 *al.*, 2025). This automated approach identifies vehicles and people from remote camera data (Fennell *et*
179 *al.*, 2022), and enables classification without viewing images, thereby preserving individual privacy. We
180 categorised vehicles as ‘motorised’ forms of activity (vehicles other than bicycles), and people as ‘non-
181 motorised’ forms of activity (people on foot and on bicycles) (Reilly *et al.*, 2017; Procko *et al.*, 2022;
182 Gump *et al.*, 2023). To define independent events we applied a one minute threshold, as longer intervals
183 risked conflating distinct events given the rapid passage of vehicles and pedestrians (Procko *et*
184 *al.*, 2022). This resulted in 21,972 events of motorised activity, and 2,875 of non-motorised activity.

185

186

187

188 **Model design**

189 All analyses were undertaken in Program R v4.3 ([R Development Core Team 2010](#)).

190 We used a statistical time-to-event analysis based on a mixed-effects Cox proportional hazards
191 model (Cox 1972) from the R package 'coxme' (Therneau 2015). This model was used to examine 'return
192 times' of possums, pademelons, wallabies, wombats and Tasmanian devils, defined as the time
193 difference (in hours) between the end of one event of a species and the beginning of the next
194 consecutive event of the same species (Figure 2a) (Visscher *et al.*, 2023). Return times were calculated
195 for each camera operational period (hereafter 'period'), which was defined as the duration during which
196 a camera remained continuously active. A period ended when a camera either malfunctioned (e.g.,
197 stopped operating due to battery depletion), was stolen, or reached the end of the field season. Because
198 cameras could malfunction between servicing visits, some camera sites had multiple periods. Unlike
199 return times of marked individuals commonly measured using GPS telemetry (Anderson *et al.*, 2008;
200 English *et al.*, 2014; Martin *et al.*, 2015), our approach focused on revisit times of an animal to fixed
201 camera-trap sites, thus capturing data at the population level.

202

203 To meet the proportional hazards assumption of the Cox model, and focus on an ecologically
204 relevant timeframe, return times were left- and right-censored (Rulli *et al.*, 2018; Visscher *et al.*, 2023).
205 We left-censored return times at five hours, as shorter intervals likely reflect immediate local foraging
206 rather than meaningful habitat revisits from a risk-avoidance perspective (See Figure S1 for the
207 distribution of return times for all four species). Right-censoring was set at 24 hours to restrict analyses
208 to single diel cycles, as longer thresholds risk introducing non-proportional effects due to repeated daily
209 patterns in disturbance sources (e.g. human and predator activity patterns) and animal behaviour that
210 could confound covariate estimates (Fisher *et al.*, 1999). To account for variation across sites such as

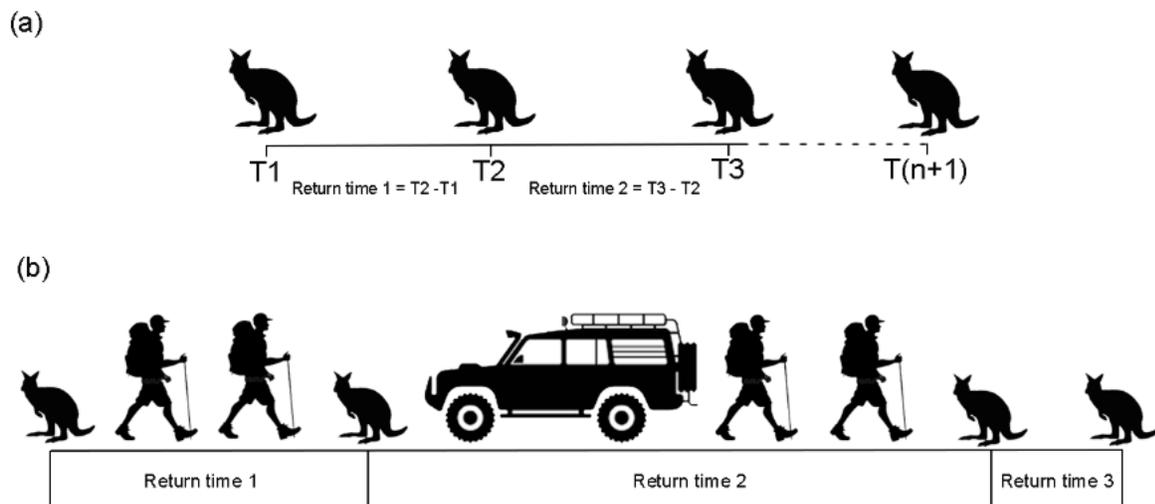
211 potential differences in animal densities, and to address the non-independence of observations within
212 camera-trap locations, site was included as a random intercept.

213

214 **Model covariates**

215 We modelled wildlife return times using two continuous predictor variables: the number of motorised
216 and the number of non-motorised events recorded during each return interval (Figure 2b). Representing
217 these predictors as continuous counts, rather than binary presence-absence indicators, allowed us to
218 quantify how each additional event – regardless of whether it occurred during the day or night –
219 influenced wildlife return times. All three predictors were included in the modelling as they all had a
220 Pearson's correlation coefficient $r < 0.7$. We did not undertake model selection as we were interested in
221 effect size rather than predictive modelling (as per the recommendation of Yates *et al.*, 2023).

222



223

224 **Figure 2.** A timeline illustration of (a) the procedure to calculate return times (RTs) using time-to-event-
225 data from a single camera trap. T1 is the time (in hours) elapsed from the moment the camera
226 started operating to the moment it captured the first event of a mammal species. T2, T3 and next

227 T_i values are the times for subsequent independent events (using a threshold of 30 minutes) of that
228 same species. T_{n+1} is the time of the last independent event of the species before the camera trap
229 stopped operating due to either battery depletion, theft, removal due to high risk of theft, or the end of
230 the study period. RTs are calculated as $RT_i = T_{(i+1)} - T_i$, for $i = 1, 2, \dots, n$. (b) scenarios where zero, one, or
231 more than one independent event involving motorised or non-motorised human activity (using a
232 threshold of one minute) occurred during the return time interval of a mammal species.

233

234 **Hazard modelling framework**

235 Cox proportional hazard models were examined for each wildlife species during two distinct periods of
236 the year, referred to as 'season'. This approach allowed us to minimise the influence of seasonal changes
237 in wildlife and human activity patterns. The 'warm' season encompassed summer and parts of late
238 spring and early autumn, defined as November to March (Southern Hemisphere). This period is
239 characterised by higher average maximum temperatures and greater visitor numbers to
240 Tasmania (Cook *et al.*, 2000; Hardy *et al.*, 2020). In contrast, the 'cold' season spanned April to October
241 and was associated with lower average maximum temperatures. Return times were grouped according
242 to these two seasonal divisions.

243 For each model, we estimated hazard ratios (HRs) and 95% confidence intervals (CIs) for all predictor
244 variables. The HR represents the multiplicative effect of a predictor on the hazard of a species return
245 time. A HR less than one suggests that the predictor variable reduces the hazard, leading to longer
246 return times. Predictor variables with 95% CI that did not overlap with one were considered to have
247 evidence of an effect on wildlife return times.

248

249 **Results**

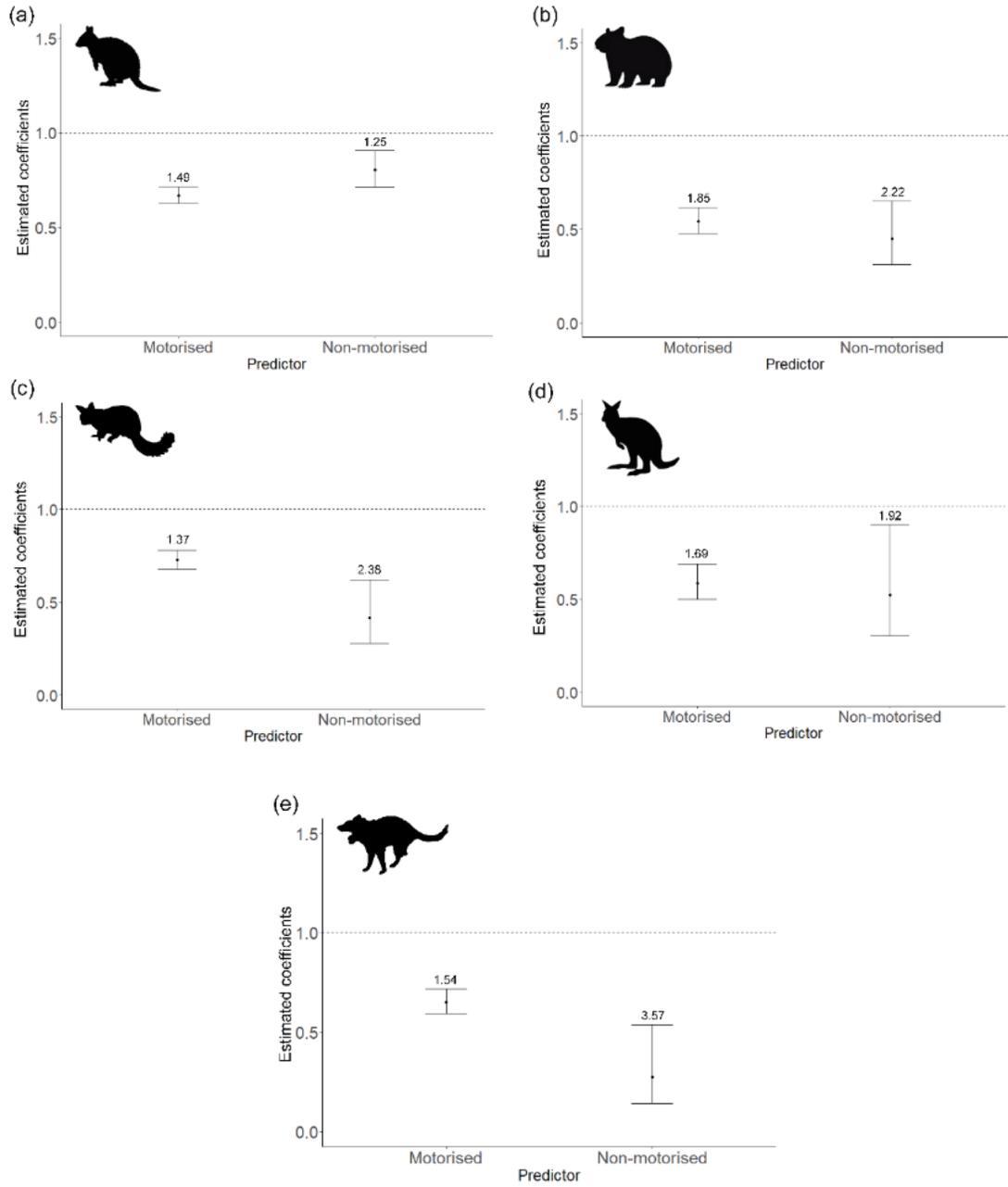
250 **Warm season**

251 A total of 4,707 return times were recorded for pademelons, 4,253 for wombats, 3,310 for possums,
252 1,733 for wallabies, and 3,471 for Tasmanian devils. During the warm season roads averaged
253 approximately 1.83 motorised events per day (range = 1.00–5.59) and 1.49 non-motorised events per
254 day (range = 1.11 – 2.40).

255

256 We found that both the number of motorised and non-motorised events had a negative effect on return
257 times for wildlife (see Figure 3 and Table S2 for model outputs). For Tasmanian devils (predator)
258 each additional motorised event delayed return times by approximately 35% (hazard ratio, hereafter HR:
259 0.65; 95% confidence interval, hereafter CI: 0.59-0.72), while each additional non-motorised event
260 delayed return times by approximately 72% (HR: 0.28; 95% CI: 0.14-0.54). Among herbivores (prey), the
261 longest delay associated with motorised events was for wombats (HR = 0.54; 95% CI:0.48-0.62), and with
262 non-motorised events for possums (HR:0.42, 95% CI:0.28-0.62).

263



264

265 **Figure 3.** Warm season hazard ratio (HR) estimates and 95% confidence intervals (CIs; vertical lines) of
 266 for predictors included in the Cox model for (a) Tasmanian pademelon, (b) Bare-nosed wombat, (c)
 267 Brushtail possum (d) Bennett's wallaby, and (e) Tasmanian devil. Predictors with HR was less than one
 268 and 95% CI not overlapping with one were interpreted as having a significant negative effect in the
 269 hazard, and thus, delaying return times for that species. A HR equal to one indicates no effect. The

270 numbers above each point on the plot represent the return time multipliers calculated as 1/HR.

271 Predictors include: Motorised = motorised forms of activities (e.g. vehicles and ATVs), and non-motorised
272 = non-motorised forms of activity (e.g. walkers and bicycles).

273

274 **Cold season**

275 A total of 3,582 return times were recorded for pademelons, 3,198 for wombats, 2,745 for possums, 834
276 for wallabies, and 4,132 for Tasmanian devils. During the cold season roads averaged approximately 1.54
277 motorised events per day (range = 1.00 –4.28), and 1.47 non-motorised events per day (range = 1.07 –
278 2.44).

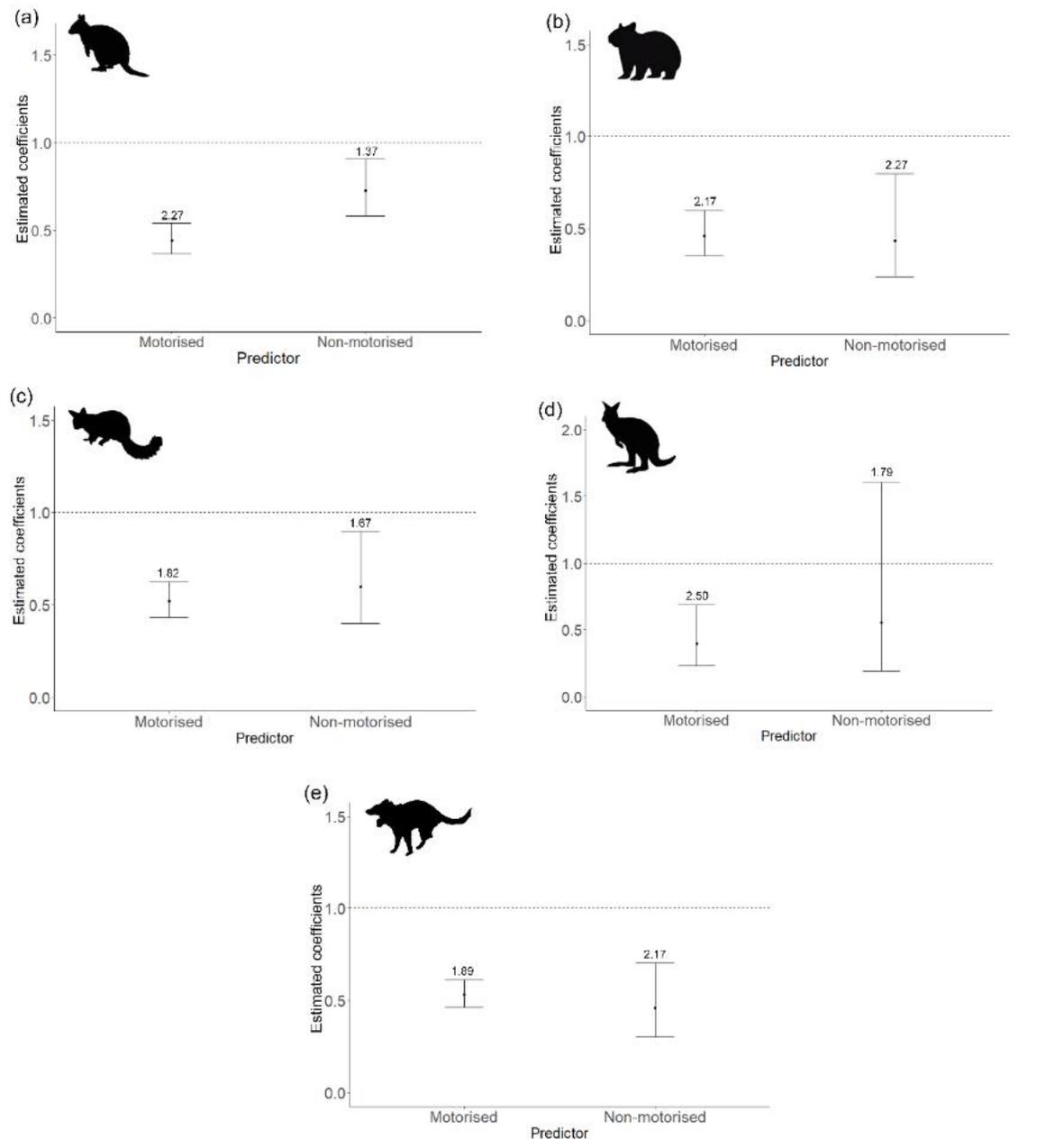
279

280 During the cold season, we found that an increase in motorised events had a negative effect on return
281 times for all wildlife species (see Figure 4 and Table S2 for model outputs). For instance,
282 an additional motorised event delayed return times for Tasmanian devils by approximately 47% and for
283 wallabies by approximately 60% (Figure 4c and e). However, during this season, an increase in non-
284 motorised events significantly delayed return times only for Tasmanian devils (HR: 0.46; 95% CI: 0.30-
285 0.70) and three herbivore species: pademelons (HR: 0.73; 95% CI: 0.58-0.91) wombats (HR: 0.44; 95% CI:
286 0.34-0.80), and possums (HR: 0.60; 95% CI: 0.44-0.63). This suggests species-specific responses to non-
287 motorised recreation that vary by season, likely driven by lower human visitation levels in Tasmanian
288 protected areas (PAs) during the cold season and behavioural traits associated with individual species.

289

290

291



292

293 **Figure 4.** Cold season hazard ratio (HR) estimates and 95% confidence intervals (CIs; vertical lines) of for

294 predictors included in the Cox model for (a) Tasmanian pademelon, (b) Bare-nosed wombat, (c) Brushtail

295 possum (d) Bennett's wallaby, and (e) Tasmanian devil. Predictors with HR was less than one and 95% CI

296 not overlapping with one were interpreted as having a significant negative effect in the hazard, and thus,

297 delaying return times for that species. A HR equal to one indicates no effect. The numbers above each

298 point on the plot represent the return time multipliers calculated as $1/HR$. Predictors include: Motorised

299 = motorised forms of activities (e.g. vehicles and ATVs), and non-motorised = non-motorised forms of
300 activity (e.g. walkers and bicycles).

301

302 **Discussion**

303 In this study, we used a time-to-event analysis to assess how an increase in the number of events
304 involving motorised and non-motorised recreation influence return times of Tasmanian devils and four
305 prey species, in protected areas (PAs) of Tasmania. The results showed that predator and prey delayed
306 their return times following more motorised and non-motorised events, indicating that they generally
307 avoided human presence along roads.

308

309 In line with our predictions, we found that both predator and prey species avoided motorised and non-
310 motorised activities along roads. This pattern contrasts with the human shield hypothesis which predicts
311 that predators avoid human activity while prey do not, using human presence as a shield against direct
312 competitors or predators (Berger 2007; Muhly *et al.*, 2011a). Therefore, our findings align with the
313 'landscape of fear' framework, which suggests that non-consumptive recreational activities (e.g. hiking
314 and four-wheel-driving) shape wildlife behaviour and habitat use through perceived risk (Frid *et*
315 *al.*, 2002; Gaynor *et al.*, 2019). By mutually avoiding humans, both predator and prey may be forced into
316 greater spatial and temporal overlap, leading to increase predation and its non-consumptive effects (e.g.,
317 risk effects) (Van Scoyoc *et al.*, 2023). However, changes in return times may represent adaptive
318 strategies with wildlife moving into adjacent habitat to cope with a disturbance. In such cases, human-
319 altered predator-prey overlap may be temporary and unlikely to have lasting consequences for
320 ecological communities (Van Scoyoc *et al.*, 2023). Addressing this uncertainty requires further research
321 that integrates behavioural, physiological and demographic responses.

322

323 Our results showed that motorised activity was significantly associated with longer return times across
324 all species and seasons, whereas non-motorised activity did not significantly influence wallaby return
325 times during the cold season. This lack of response suggests that wallabies have higher tolerance
326 thresholds to non-motorised recreation when overall human activity is lower, consistent with other
327 findings that contextual factors mediate wildlife responses to disturbance (Tablado *et al.*, 2017; Smith *et*
328 *al.*, 2021; Sytsma *et al.*, 2022). Nevertheless, further research involving non-observational approaches
329 (e.g., experimental) is needed to fully characterise the conditions underlying non-response of wallabies
330 to non-motorised activity during this season. Furthermore, the apparent reduced effect of non-
331 motorised recreation in the cold season can be explained by that fact that during this season longer
332 intervals between animal detections inherently contain a smaller number of recreational events
333 (Dymit *et al.* 2025). In addition, grouping all forms of non-motorised activities (e.g., hiking and mountain
334 biking) into a single category may have masked distinct effects of specific activities. For instance, some
335 mammals have been found to perceive mountain biking more similarly to vehicles than to
336 hikers (Naidoo *et al.*, 2020a). Future research should prioritise examining wildlife responses to these
337 activities separately. Another future direction will be identifying road use thresholds; point at which
338 motorised and non-motorised activity begins to cause significant impacts on wildlife (Marion *et*
339 *al.*, 2002; Dertien *et al.*, 2021).

340

341 Within the context of managing PAs, our findings have important implications. If managers aim to
342 mitigate the most harmful disturbance first, our results suggest prioritising strategies targeting
343 motorised users, which showed to be a consistent source of perceived risk across species. This
344 consistency likely reflects the higher speeds, louder noise emissions (Barber *et al.*, 2010; Tablado *et*

345 *al.*, 2017), and the added threat of collisions (Hobday 2010; Nguyen *et al.*, 2019) associated with
346 vehicles. Practical measures could include enforcing clearly marked speed limits (Miller *et*
347 *al.*, 2020c), and implementing seasonal restrictions during peak periods (e.g. the warm season in
348 Tasmania) (Schummer *et al.*, 2003; Miller *et al.*, 2020c). However, seasonal restrictions, should balance
349 human and ecological benefits and be justified with population level evidence (Miller *et al.*, 2020;
350 Schummer *et al.*, 2003). At the same time, given the significant effects of non-motorised activities on
351 wildlife, and the projected rise in visitation to many Tasmania PAs (Tasmania Parks and Wildlife Service
352 2024; Tourism Tasmania corporate 2024) managers should incorporate ongoing monitoring of human-
353 use trends (Rice *et al.*, 2020). This will allow them track changes in visitation and recreational behaviour
354 over time and adjust conservation strategies accordingly.

355

356 While our study characterises the effects of motorised and non-motorised recreation on wildlife, there
357 are limitations to the inferences we can make. First, we focused on estimating wildlife habitat use to
358 human activities on roads, and thus it was not possible to determine the spatial magnitude to which
359 wildlife may be displaced from roads by anthropogenic activity. Furthermore, our model does not
360 incorporate differences in detection rates and behavioural dynamics between species and types of
361 recreational activity. This limitation has been shown to produce false indications of avoidance in studies
362 using time-based approaches such as avoidance-attraction ratio (AAR) methods, which compare time
363 intervals between species detections (Parsons *et al.* 2016; Dymit *et al.*, 2025). Although our time-to-
364 event approach differs from AAR methods by modelling time to the next detection and including the
365 type of recreational activity as a covariate, we recommend further research using GPS or accelerometer
366 data to account for the directionality of species movements (Dymit *et al.*, 2025; Flowers, 2019). Satellite
367 tracking collars on focal species would also allow to determine the spatial magnitude to which wildlife
368 may be displaced at finer behavioural scales (Suraci *et al.*, 2020), while also enabling assessments of

369 whether responses of wildlife vary between sex and age classes. Second, the presence of a third species
370 between return times can introduce constraints in the observed patterns (Niedballa *et al.*, 2019;
371 Van Scoyoc *et al.*, 2023). While one possible approach to addressing this would have been to exclude
372 time intervals where a third species was present, this would have come at the cost of reduced sample
373 size and thus, statistical power (Niedballa *et al.*, 2019). Finally, our framework did not account for
374 variation in visitation levels across PAs or across years because such data was unavailable for many areas,
375 and to maintain a robust sample size. A future study with sufficient sample sizes across PAs with
376 contrasting human visitation levels, or across years within a single PA where visitation is monitored,
377 could help disentangle the influence of variations on overall human visitation levels on wildlife return
378 times.

379

380 **Conclusion**

381 In this study we illustrate how time-to-event data on wildlife and humans, collected from a large-scale
382 camera trap survey, can help researchers and park managers test the immediate spatiotemporal
383 responses of wildlife (e.g., return times) to recreational activities. Our findings provide evidence that
384 both predator and prey perceive human activities in protected areas as a source of risk, offering
385 important insights into the extent of recreational impacts on wildlife and their interactions. Given the
386 complexity of wildlife responses to human activities, standardised methods such as camera traps can aid
387 in the endeavour of accounting for diverse factors influencing these interactions.

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