

2007

Behavioral Indicators of Predator Space Use: Studying Species Interactions Through the Behavior of Predators

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This is an Accepted Manuscript of an article published by Taylor & Francis in *Israel Journal of Ecology and Evolution* in 2007, available online: <http://www.tandfonline.com/10.1560/IJEE.53.3.389>

Recommended Citation

Schmidt, Kenneth A. and Schaubert, Eric M. "Behavioral Indicators of Predator Space Use: Studying Species Interactions Through the Behavior of Predators." *Israel Journal of Ecology and Evolution* 53, No. 3-4 (Jan 2007): 389-406. doi:10.1560/IJEE.53.3.389.

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1 Running Head: Behavioral indicators and predator-free space
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7 **Behavioral indicators of predator space use:**
8 **Studying species interactions through the behavior of predators**

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ABSTRACT

Predation has major impacts on survival and reproductive success for many species. To quantify these effects, ecologists often choose to intensively study prey populations to measure predation rates and/or estimate predator abundance. But in some cases, predation rates are less strongly related to predator abundance per se than to spatial and temporal patterns of predator space use; thus quantifying the latter may provide meaningful surrogates of predation rates that scale up to larger areas. This is particularly true when safety for prey, especially sessile and vulnerable prey, is strongly linked to predator-free space. Our own research programs have used two general types of behavioral indicators to quantify space use by predators: giving-up densities, as a surrogate for patch quitting harvest rates, and activity density. We discuss two general mechanisms by which predator-free (or predator-poor) space is created and link these mechanisms to behavioral indicators that can be easily collected in the field. We then summarize our past work on predation on passerine nests and moth pupae to demonstrate how using behavioral indicators of space use can reveal much about the impact of a predator on its prey. We demonstrate that behavioral indicators can be used as: (1) leading indicators for predation rates, (2) surrogates for information otherwise difficult to obtain, (3) integrative measures of the strength of species interactions, and may (4) reveal the outcomes of ecological interactions, such as prey persistence.

Keywords: behavioral indicators, foraging theory, GUDs, incidental predation, predator space use

58

INTRODUCTION

59 Predation has major impacts on survival and reproductive success for many species. To quantify
60 these effects, ecologists often choose to intensively study predator and prey populations to
61 measure predation rates and/or estimate predator abundance. But in some cases, predation rates
62 are less strongly related to predator abundance per se than to spatial and temporal patterns of
63 predators' space use. The consequence is that predator density can be a lagging or misleading
64 indicator of ecological relationships when the factors that govern predator space use, e.g., the
65 predator's motivational state or patch use decisions based on the local levels of risk and reward,
66 do not scale with or lag behind conspecific population density (Brown et al. 1999, Schmidt et al.
67 2001a, Pusenius and Schmidt 2002). Under these circumstances, developing informative
68 behavioral metrics (i.e., behavioral indicators) that do scale with predation rates is imperative
69 and of obvious practical value. Two metrics or behavioral indicators of predation rates we
70 consider below are quantifying the amount of space occupied by predators or the intensity at
71 which they exploit space.

72 Consider nest predation as a specific example of these points. Nest predation is often the
73 leading cause of reproductive failure in birds (Martin 1992) and other nesting organisms (e.g.,
74 turtles, Chaloupka and Limpus 1991), and thus is an important determinant of population vital
75 rates. A predator's attack on a songbird nest is the outcome of numerous behavioral decisions
76 made by the predator and prey. Perhaps none are as important as the decisions of where the
77 predator should forage and for how long. To see this, first consider that although parents can
78 actively select nest sites to reduce predation risk, nests themselves cannot move once established
79 and so prey escape tactics are relatively passive. Second, outside colonial species that rely on
80 predator satiation as a means to reduce the magnitude of predation, nests are cryptic as well as

81 ephemeral, dispersed and occurring at low densities, and in some cases outnumbered by their
82 predators. The latter can occur because most nest predators are foraging generalists that rely on
83 a broad prey base and so can out number some classes of prey items. In New York State, for
84 instance, the 2.25-ha trapping grids managed by our colleague, Richard Ostfeld, can hold > 200
85 individual white-footed mice and eastern chipmunks in some years, whereas breeding bird
86 densities are < 10 pairs/ha and much lower when we consider only ground- or low-shrub nesting
87 species.

88 Based in part on the ratio of predator/prey abundances, the generalist feeding habitats of
89 most nest predators, and the low intake (nests often comprise $\leq 2\%$ of predator diet), that many
90 predation events are incidental in nature (Vickery et al. 1992, Yanes and Saurez 1996); that is,
91 predation occurs as the result of encountering relatively rare and hence “unexpected” prey items
92 while searching for primary prey or simply occupying high quality resource patches. Hence,
93 rates of incidental predation will correlate with patterns of predator space use and exploitation,
94 and developing models or methods to track the latter will benefit from a consideration of the
95 predator’s patch use behavior. We develop these explicit relationships below.

96 While we have conceptualized our ideas largely based on our empirical studies of
97 breeding songbirds (KAS) and pupae of gypsy moths (EMS), our framework should be
98 amenable to prey that are encountered largely incidentally by generalist predators, including prey
99 organisms that are relatively uncommon or ephemeral, sessile or otherwise are largely inactive
100 during at least one life history stage. For example, this can also include the vulnerable young of
101 cheetahs (Durrant 1998) and deer (Pojar and Bowden 2004). After presenting our conceptual
102 background and supporting empirical studies we expand our framework to consider incidental or
103 non-incidental prey that benefit from alternative mechanisms that ‘anchor’ predators in space in turn
104 creating spatial heterogeneity in predator activity that prey may find refuge in. Through developing

105 a framework and summarizing our past work we argue that behavioral indicators of predator
106 space use can reveal much about the predator's impact on its prey and hence the habitat quality
107 and reproductive success of the prey. We will demonstrate that behavioral indicators, and the
108 framework from which they emerge, may be particularly valuable as: (1) leading as opposed to
109 lagging indicators of the magnitude of predation, (2) surrogates for information otherwise
110 difficult to obtain, (3) integrative measures of the strength of species interactions, and lastly may
111 (4) reveal novel insights into ecological interactions.

112

113

CONCEPTUAL FRAMEWORK

114 *Quitting harvest rates, space use, and incidental predation* – The branch of optimal foraging
115 theory known as patch-use theory has developed around the question of how much time to
116 allocate to a resource patch. The answer is generally given in the form of a patch giving-up rule.
117 In the case of prescient foragers with random search the appropriate rule is to forage until the
118 rate of resource harvest (a declining function of time spent foraging in the patch) equals the
119 accrual rate of foraging costs (Charnov 1976, Brown 1988). Brown (1988, 1994) recognized
120 three general foraging costs: predation (P), missed opportunity (MOC; the value of time should it
121 be spent on alternative fitness-enhancing activities) and metabolic (C) costs. This giving-up rule
122 is referred to as a fixed quitting harvest rate (QHR) rule because it results in patches that have
123 equal harvest rates after a forager ceases (i.e., quits) to exploit the patch irrespective of initial
124 patch quality. It can be written as: $QHR = P + MOC + C$. When the initial harvest rate of a
125 patch is less than $P + MOC + C$ the patch is unprofitable (i.e., the optimal time spent foraging is
126 zero). A forager that can assess and respond to spatial heterogeneity in resource abundance
127 should exploit only those resource patches that it considers profitable (Stephens and Krebs 1986,

128 Schmidt and Brown 1996, Schmidt et al. 2001). Unprofitable and therefore unused space (i.e.,
129 predator-free space) can provide refuge that may be especially important for incidental prey;
130 prey too sparse for predators to base patch-use decisions on, yet they are attacked whenever
131 generalist predators encounter them while foraging for more common foods (Vickery et al.
132 1991).

133 The giving-up rule also indicates two general routes to patch rejection (see Fig. 1)
134 corresponding to "top-down" and "bottom-up" forces, respectively: predators reject patches
135 associated with a high risk of predation from apex predators (e.g., nest-robbing chipmunks avoid
136 patches associated with greater risk of predation from weasels) and predators reject patches
137 lacking sufficient resources (nest-robbing chipmunks avoid patches with few red maple seeds).
138 These routes are non-mutually exclusive in the sense that patch rejection always occurs because
139 there are insufficient resources to offset the high cost (including the risk of predation; Brown and
140 Kotler 2004) of foraging in the patch. Nonetheless, it is also valuable to separate these two
141 components because the abundance and distribution of resources and apex predators (maple
142 seeds and weasels, respectively, in our chipmunk example) are largely independently determined
143 (but see Sih 1998 for a summary of models that suggest apex predator abundance should match
144 the abundance of resources for its prey).

145 More formally, Schmidt et al. (2001a) constructed a model explicitly linking generalist
146 predator foraging behavior, space use, and predation rates on incidental prey. In the model,
147 predators assess patch quality and only forage in those resource patches considered profitable
148 (i.e., initial patch harvest rate exceed the predator's quitting harvest rate, QHR); unprofitable
149 space is unused (Schmidt 2004). By varying the QHR, its relationship to space use and
150 incidental predation rates could be determined (also see Schmidt et al. 2001 for explicit

151 equations). If the QHR is low (i.e., resources are extremely valuable) most patches are profitable
152 and nearly all available space is used. If the QHR is high (i.e., resources are relatively cheap or
153 predation risk is very high) many patches are unprofitable and space use declines. Therefore,
154 changing the QHR drives an inverse but non-linear change in the amount of space predators
155 forage within (Schmidt et al. 2001, Pusenius and Schmidt 2002). In one analysis of the model
156 that included uncommon incidental prey dispersed (independent of primary prey) among the
157 patches, increasing the QHR reduced incidental prey consumption. Overall consumption closely
158 mirrored predator space use when primary prey were distributed in a Poisson fashion (Fig. 3),
159 but this result may not follow for other distributions of prey or if primary and incidental prey are
160 themselves inversely correlated.

161 Space need not be treated dichotomously as predator-free or predator-occupied and this
162 fact can further strengthen the relationship between the QHR and space use. The QHR is both a
163 qualitative (predator-free/occupied) and quantitative (time spent searching for primary prey)
164 indicator of space use/exploitation. Assuming random search within a patch and holding initial
165 primary prey density fixed, search time (t) is a decreasing function of QHR or the final density of
166 primary prey remaining in a patch (Schmidt and Brown 1996). Predator-free space is equivalent
167 to patches where $t=0$.

168

169 *Spatial anchoring and predator space use* – Within our framework of incidental predation
170 predator activity is “anchored” to high reward/low risk sites creating predator-free or predator-
171 poor space at other sites. Spatial heterogeneity in predation risk may also arise from alternative
172 mechanisms of spatial anchoring of the predator that may be important for both incidental and
173 non-incidental prey. This includes interactions with conspecifics (e.g., territoriality) and the

174 need to return to a central home location (e.g., nest or den). For example, Sergio et al. (2003)
175 observed that black kite (*Milvus migrans*) territories were aggregated within interstitial areas
176 outside the sphere of influence of its intra-guild predator, the eagle owl (*Bubo bubo*). All black
177 kite nests within 1 km of eagle owl nests were unsuccessful, and kite nests beyond 1 km showed
178 a linear increase in productivity with greater distance from owl nests. Likewise, Mech (1977)
179 noted that wolves form stable territories that are separated by buffer zones up to 2 km wide, into
180 which wolves rarely venture to avoid antagonistic interactions with neighboring packs. Nearly
181 all surviving white-tailed deer in a declining population were within these buffer zones (Taylor
182 and Perkins 1991). In this vein, Schaubert (2000) modeled predator-free space as gaps between
183 individual home ranges under different assumptions of territory spacing. Uniform territory
184 spacing creates a linear decrease in predator-free space with increasing predator abundance until
185 all space is occupied, whereas random spacing results in an exponential decrease, as has been
186 observed for white-footed mice (Schmidt et al. 2006). With random territory spacing, some
187 predator-free space persists even at high predator abundance, which was critical for gypsy moth
188 persistence in Schaubert's (2000) analysis. Through these mechanisms of territoriality and
189 central place foraging, predator space may not correlate well with predator abundance, leaving
190 the latter a poor indicator of predation risk to prey. Such cases call for measuring space use
191 itself.

192 We end this section by noting that colleagues have suggested to us that true predator-free
193 space does not exist – eventually predators will move through all available space. This remains
194 to be demonstrated but it does have a ring of truth; predator-free space can best be considered as
195 a transient phenomenon. But despite its transience, predator-free space can be critical when prey
196 vulnerability is short-lived. Examples include gypsy moth pupae (vulnerable for ~ 13 days) and
197 songbird nests (vulnerable for ~ 2-4 weeks) that comprise our own studies, but also includes

198 cheetahs, which are vulnerable to intraguild predation largely as young cubs, and deer, which are
199 vulnerable to predation from coyotes largely as young fawns (Pojar and Bowden 2004, Vreeland et
200 al. 2004). Survival of these prey requires their locations remain predator-free for only days or weeks
201 to have large repercussions. When prey are mobile, as when young cubs or fawns are moved, then
202 even if individual pockets of predator-free space wink on and off, the mean amount of predator-free
203 space is likely to be fairly stable in comparison.

204

205 **EMPIRICAL STUDIES**

206 *Behavioral Indicators: monitoring quitting harvest rates and space use* – Quitting harvest rates
207 can be notoriously difficult to monitor in the field. Fortunately, counting the number of resource
208 items remaining within a patch after a foraging bout, known as a giving-up density or GUD,
209 often provides a useful surrogate of quitting harvest rates that is easily measured in the field
210 (Brown 1988). Food patches should yield diminishing returns such that current amount of food
211 in a patch correlates with the instantaneous harvest rate. These conditions are often easily
212 fulfilled through the use of artificial food patches (Brown 1988) created by mixing a small
213 quantity (to prevent satiation) of food (e.g., seeds) within a patch substrate (e.g., sand). Giving-
214 up densities have been widely applied to quantifying individual foraging costs (see recent review
215 by Brown and Kotler (2004), but as discussed above, GUDs collected across larger spatial scales
216 can be used as surrogates for space use (Schmidt et al. 2001).

217 The technique of giving-up densities is an important tool in the ecologist's toolbox for
218 measuring the landscape of foraging costs perceived by predators and the risk experienced by
219 prey. However, not all prey are distributed within idealized food patches that the GUD
220 technique often requires. Furthermore, the food patches themselves lure animals so that GUDs
221 do not necessarily indicate specific sites predators would not use in the absence of a baited patch.

222 There are techniques to get around some of these concerns; for example by correlating GUDs to
223 the physical landscape one may learn that a predator around foraging along habitat edges. An
224 alternative to GUDs that we discuss here is directly quantifying predator activity. Predator
225 activity density, i.e., the aggregate use of a site per unit area and time, is a more proximal
226 measure of the landscape of risk experienced by prey. Activity density reflects the total *intensity*
227 *of use* in a specific area by a number of individuals, rather than quantifying the *number* of
228 individuals. The concept of activity density is similar to the idea of population density, but
229 applies at the spatial scale of individual space use decisions. For instance, one raccoon (*Procyon*
230 *lotor*) that traverses a given patch of space 6 times in a week should represent greater risk (i.e.,
231 higher likelihood of encounter) to prey than 3 raccoons that each traverse the same patch once a
232 week. We have used two different indicators of small mammal activity density, (1) traps with
233 zero captures and (2) track plates.

234

235 *Case studies* – Schmidt and colleagues (2001, Schmidt and Ostfeld 2003) tested the explicit links
236 between QHRs (measured by GUDs), space use (quantified as the proportion single oat grains
237 removed by mice off a lattice grid arrangement) in the white-footed mouse (nest predator), and
238 predation on passerine nests (prey). They used artificial nests to increase sample sizes and have
239 more control of the distribution of nests (Schmidt et al. 2001, Schmidt and Ostfeld 2003). These
240 studies demonstrated the hypothesized relationships between (1) GUDs (collected at random
241 locations within the same 2.25 ha plots where artificial nests were deployed) and space use in
242 mice (Fig 4a) and (2) mouse space use and predation of nests by mice (Fig 4b). In contrast, plot-
243 level mouse density failed to predict space use or predation rates on nests. Noteworthy in this
244 regard, Schmidt and Ostfeld (2003) observed 7-fold variation in the amount of space occupied

245 (based on oat grain removal) among the six study plots that in turn varied less than 2-fold in
246 mouse density. Clearly in this example mouse abundance did not scale with their spatial activity,
247 but it was the latter that determined nest predation. Interestingly, annual variation in mouse
248 abundance does predict annual variation in nest success for the ground-nesting veery (*Catharus*
249 *fuscescens*) suggesting that intra-annual spatial variability may be greater than inter-annual
250 variability (Schmidt et al. 2006). Unfortunately, we do not have a temporal sequence of spatial
251 activity or GUDs for comparison to mouse density. However, mouse GUDs were by far the
252 highest recorded at the site in 2000 when mouse populations had collapsed and nest predation
253 rates were their lowest.

254

255 *Spatiotemporal shifts in space use or predation rates* – Whereas the above example focused on
256 spatial variability in QHRs and space use, understanding temporal changes is also important and
257 can be addressed using similar techniques. Schmidt and Whelan (1999) documented a reprieve
258 from nest predation for forest songbirds when a mulberry (*Morus spp.*) mast drew raccoons out
259 of the forest within a suburban arboretum. We were first alerted to this through a drop in nest
260 predation and shift from density-dependent to density independent predation mid-way through an
261 artificial nest study. Artificial nests were replenished daily in the experiment, so intact nests
262 remaining after a night of foraging by raccoons represented a giving-up nest density much the
263 same as seeds remaining in a tray of sand after a bout of foraging. We subsequently
264 demonstrated a concomitant drop in nest predation on shrub-nesting birds (American robin,
265 *Turdus migratorius*, and wood thrush, *Hylocichla mustelina*). Importantly, artificial nest GUDs
266 informed us of current temporal changes in predation rates on real nests. Similarly, Schmidt
267 (2006) showed the density of active veery nests drops as much as 20% or more (due to

268 predation) between the full and new lunar phases when white-footed mice demonstrated a 3-fold
269 increase in the amount of space occupied (based again on oat grain removal), typical of nocturnal
270 rodents (Kotler et al. 1991, Mandelik et al. 2003, Abramsky et al. 2004). These observations
271 suggest that GUDs can be used as an indicator of temporal shifts in space use/activity and
272 predation rates as alternative resources rapidly renew (or pulse) in the environment, or through
273 rapid temporal variation in predation risk from apex predators. The two studies described above
274 relied on ‘natural’ GUDs (or standing crop of nests after a night of foraging by raccoons or mice)
275 instead of using artificial food patches. In these cases the experimenter has little control of
276 resource renewal rates that can cloud interpretation of the data (see Morris and Mukherjee, this
277 volume). But systems such as these should be amenable to the use of artificial food patches for
278 monitoring and understanding the causes and consequences of temporal shifts in predator
279 behavior and predation rates. Raccoon food patches were developed after Schmidt and Whelan’s
280 observations (*Berris, unpublished*). Likewise, Schmidt (2006) did attempt to collect mouse
281 GUDs, but artifacts due to playbacks of owl vocalizations interfered with isolating a moon effect
282 (*KAS, unpublished*).

283

284 *Variation in the predator community* – Coexisting consumer species often differ considerably in
285 their giving-up densities; in fact this observation has become a tool in the search for mechanisms
286 of coexistence between competitors (Brown 1989, Kotler and Brown 1999). However, the
287 relevance of interspecific variation in GUDs for species interactions outside of competition has
288 received far less attention. Our model suggests interspecific variation, through effects on space
289 use and predation, can be far-reaching. For instance, Schmidt (*unpublished*) collected GUDs
290 from both white-footed mice and eastern chipmunks (*Tamias striatus*) foraging in the same food

291 patches that were replenished between nighttime exploitation by mice and daytime exploitation
292 by chipmunks. Mice (20g) had much lower GUDs than the larger bodied, diurnally-active
293 chipmunk (100g). It is unknown whether body mass, metabolic demands or other life history
294 characteristics might lead to a predictable hierarchy of GUDs among species. Regardless, these
295 species differences, based on our modeling approach, predict that mice should exploit a greater
296 amount of space than chipmunks, all else equal (see Fig. 1). If true, two additional predictions
297 follow: (1) mice should cause higher mortality rates on nests than chipmunks (i.e., mice use more
298 space) and (2) given the greater area of space used, mice will show compensatory predation on
299 nests following chipmunk removal, but not vice versa. Both of these predications have been
300 upheld (see Schmidt et al. 2001b and Fig. 5). Unfortunately, Schmidt and Ostfeld (2003) were
301 unable to quantify space use in chipmunks, so the prediction of interspecific differences in space
302 use commensurate with differences GUDs has not been directly tested.

303 The assumption of all 'else equal' certainly doesn't hold in this example, and there are
304 numerous alternative explanations for species differences in nest predation rates. For instance,
305 the artificial nests used by Schmidt et al. (2001b) may not have mimicked real nests in important
306 ways, i.e., chipmunks may use begging calls as cues to locate nests (Haskell 1999). Still, the
307 conclusion we hope the reader takes away is that QHRs/GUDs are an empirical tool as well as
308 the center piece of a larger set of conceptual ideas (e.g., mechanisms of coexistence, foraging
309 ecology, incidental predation) that can be applied to individual species as well as communities.
310 As used here, QHR theory and GUD methodology may provide a useful framework toward
311 understanding the distribution of species interaction strengths within communities, or predicting
312 the consequences, such as additive or non-additive predation resulting from the loss or
313 introduction of predator species.

314
315 *Zero-capture data* – Our colleague, Richard Ostfeld, has maintained six 2.25 ha small mammal
316 trapping grids at IES since 1995. We have used small mammal abundance data from these grids
317 to document temporal trends in nest predation rates on ground-nesting birds, such as the veery.
318 In early studies (Schmidt et al. 2001b) we used crude measures of mouse abundance: minimum
319 number known alive (MNA). These studies were later followed by more sophisticated estimates
320 of plot-level abundance through the program MARK (Schmidt and Ostfeld in press). At the
321 same time, we began using a very crude metric of local activity density - simply quantifying the
322 proportion of trap stations where no mice were captured over the course of a capture season. Not
323 only do these metrics vary in their sophistication, but the field and analytical effort needed to
324 produce each estimate varies considerably. In the end, correlating temporal and spatial
325 variability in mouse abundance/activity to nest predation on two thrush species required only the
326 simplest metric, zero-capture trap data (Schmidt et al. 2006).

327 The proportion of traps with zero mouse captures predicted nest predation in veeries both
328 annually and locally within a 30 × 30m area (nine trap stations centered on each nest; what we
329 called a nest neighborhood; Fig. 6). Zero-capture data also demonstrated an exponential decline
330 in unused space as predicted from the random territory placement model of Schaubert (2000)
331 discussed above. However, the decline reached a plateau, not predicted in the model signifying
332 that even at extreme mouse densities a significant proportion (~10%) of space, on average,
333 remained unoccupied (Schmidt et al. 2006). This method of quantifying predator-free space by
334 enumerating information only from trap stations that record no predator activity (i.e., captures)
335 thus provides a useful surrogate for monitoring spatiotemporal heterogeneity in nest predation
336 rates. In some ways, this should not be surprising. If prey are capable of appropriately

337 responding to predation risk then knowledge of where predators are not may be a more
338 meaningful way to track predation rates since the researcher is attempting to assess predation risk
339 in ways that may be similar to what prey use. In our example, this works even though prey (i.e.,
340 stationary nests) have to 'live' with their choice for 4 weeks. However, if prey were not
341 incidental we suspect predators would quickly respond to the distribution of their prey.

342

343 *Track plates* – When terrestrial predators are identifiable from their tracks, unbaited track plates
344 offer an efficient method to quantify predator activity density at relevant spatial scales without
345 substantially altering predator behavior or their pattern of space use as a result of luring the
346 predator (e.g., traps or cameras baited with food or scent lures; Connors et al. 2005). Connors et
347 al. (2005) designed track plates to use with small mammals by coating acetate transparency
348 sheets with a mixture of graphite powder, mineral oil (for water resistance), and alcohol (carrier).
349 These proved highly effective and convenient for recording tracks and had little evident impact
350 on visitation by white-footed mice (Connors et al. 2005).

351 We used the activity data obtained from track plates to measure the magnitude, spatial
352 scale, and temporal persistence (over periods of months to years) of hot spots and cold spots in
353 mouse activity density (Connors 2005, Schaubert et al. in revision). These characteristics appear
354 to be highly relevant to the persistence of gypsy moths in eastern oak forests, where predatory
355 mice are typically abundant and pupal predation rates are high (Campbell et al. 1975, Elkinton et
356 al. 1996, Jones et al. 1998). The vulnerable pupal stage is sedentary and adult female moths are
357 flightless, so egg masses are most likely to be laid in sites of low predation risk (i.e., spatial
358 selection). We found that track activity at a tree was a strong indicator of predation rate on
359 gypsy moth pupae deployed there (Fig. 7).

360 If hot and cold spots of predation persist between years and have spatial scales larger than
361 typical dispersal distances of larvae, then larvae can inherit the level of risk their mothers
362 experienced. This combination of spatial selection and inheritance can reduce overall predation
363 on moth populations below the threshold necessary to cause negative population growth
364 (Schauber et al. 2007). However, estimating predation from either naturally occurring or
365 experimentally deployed pupae would likely yield biased results and actually alter the behavior
366 of the foraging mice (Schauber and Jones 2006). Thus, using track plates to measure predator
367 activity density as an indicator of predation risk was the only feasible way to test our hypotheses
368 by quantifying the persistence and scale of relative refugia.

369 Overall mouse activity decreased greatly between 2004 and 2005, concomitant with a
370 crash in mouse abundance following acorn failure (Connors 2005, Schaubert et al. *in revision*).
371 However, only by monitoring the spatial and temporal dynamics of mouse activity were we to
372 fully understand the implications of this crash. Temporal autocorrelation of mouse activity
373 between months was high during years of relatively high mouse density (2003 and 2004),
374 providing stable refugia that prey could exploit, whereas this temporal autocorrelation was
375 considerably lower during the crash year (Connors 2005, Schaubert et al. *in revision*). We
376 hypothesize that the increased stability of refugia during periods of high predator density is
377 expected to buffer prey populations from the drastic mast-driven fluctuations that white-footed
378 mouse populations exhibit (Wolff 1996, Ostfeld et al. 1996, McCracken et al. 1999). Such
379 density dependence in space-use stability may arise via greater social stability during periods of
380 high density when summer reproduction is suppressed (Falls et al. 2007), or perhaps just from
381 demographic stochasticity in mouse populations at small spatial scales when overall density is

382 low. These mechanisms are ripe for future empirical investigation, and highlight the
383 insufficiency of predator population density as a sole indicator of risk to prey.

384

385 **Concluding remarks** – Our conceptual approach centers on an animal's quitting harvest rate for
386 two reasons. First, mechanistically both space use and time devoted to foraging any particular
387 patch are related to the QHR and patterns of adaptive foraging as developed above. Second, the
388 QHR (or GUD) is a phenomenological outcome of a predator's assessment of its foraging costs
389 and benefits (Brown 1988, Brown and Kotler 2004). The QHR thus presents a window into the
390 world of a predator by reflecting its ecological relationships with its environment, such as its
391 energetic state (influenced by its density and the availability of primary prey), the instantaneous
392 risk of predation from apex predators, and current valuation of food and safety. QHRs/GUDs
393 thereby incorporate the influence of multiple trophic levels (Fig. 1) and may provide a more
394 informative and causal link to space use and rates of predation (i.e., strength of species
395 interactions) than predator density would by itself. GUDs can also reveal the operation of
396 density dependence (see Morris and Mukherjee 2007). For instance, changes in population
397 density of the predator may manifest as a foraging cost (mainly the marginal value of energy
398 increases with density, as resources are depleted; Schmidt et al. 2004) that can be observed
399 through changes in QHRs and GUDs (Bowers et al. 1993, Davidson and Morris 2001, Morris
400 and Mukherjee 2007). As with any ecological model, the set of linkages depicted in Fig. 3
401 represent a subset of those that actually exist, and some classes of predators may not reasonably
402 fit within the current conceptualization. Nonetheless, we believe the model generalizes a broad
403 and important group of interactions characterizing predator-prey interactions that fall within the
404 purview of incidental predation. It also provides a cornerstone to conceptualizing behavioral

405 mechanisms that anchor predators in space, such as territoriality and central place foraging, that
406 are likely to generalize to both incidental and non-incidental prey (e.g., wolf-deer interactions;
407 Mech 1977).

408 Although monitoring GUDs reveals insights into the costs and benefits of foraging, the
409 use of artificial food patches to collect GUDs may not reveal predation risk for prey at any
410 specific point, whereas quantifying activity density may. For instance, collecting GUDs in a
411 forested landscape to monitor nest predation levels might tell us that predator space use and
412 predation risk (on nests) in the environment is low but GUDs alone cannot reveal which
413 individual songbird nests (or other prey) are most at risk because overall levels of predation risk
414 and missed opportunity both manifest at large spatial scales. For example, abundant food
415 elsewhere in an animal's territory will influence GUDs collected at a point source (Brown and
416 Alkon 1990, Schmidt et al. 2001a). Likewise, GUDs collected at a relatively risk-free (from the
417 predator's perspective) point in space but in an otherwise high risk environment would fail to
418 accurately reveal the mean likelihood incidental predation outside local site where GUDs were
419 collected. Alternatively, demographic inertia could cause the spatial and temporal distributions
420 of activity of predators to imperfectly reflect the distribution of habitat quality (Schauber et al in
421 revision). Quantifying activity density concomitantly with GUDs can enable researchers to
422 disentangle and compare the landscape of risk experienced by prey (determined by activity
423 density of predators) with the landscape of foraging costs experienced by the predator.

424 It was our goal to promote the use of behavioral indicators as an effective tool to increase
425 our understanding of predator-prey interactions, not as a substitute for measuring abundances
426 and other population-level characteristics. A detailed field study of breeding birds will
427 accurately measure nest predation rates, take video images of nest predators, and collect data on

428 avian demography and possible information on site and individual variation in quality to monitor
429 processes such as density-dependence in reproductive success. Yet many studies or basic
430 monitoring schemes have more modest goals. Even if money is no objective, detailed field
431 studies of breeding populations cannot estimate nest predation in advance of the breeding season
432 - within season estimates will be too late to implement management programs intent on boosting
433 reproductive productivity. In contrast, GUDs and activity density can be collected prior to the
434 breeding season to monitor predator behavior or as surrogates for nest predation rates. Lastly,
435 GUDs and activity density can greatly increase the cost-effectiveness of conservation efforts,
436 particularly in poor nations that cannot mount an effective field effort to monitor avian
437 reproductive ecology and nest predator behavior. Outside of monitoring actual predation events
438 (artificial nests or video images from real nests) the predator has largely been absent from nest
439 predation studies, although some excellent examples exist (e.g., Vigallon and Marzluff 2005).
440 GUDs and activity density are complementary and cost-effective indicators of the "worlds" that
441 predators and their prey inhabit.

442

443

ACKNOWLEDGEMENTS

444 Funding for our studies of predator-free space in nesting songbird communities and gypsy moth
445 populations has been provided through grants from the National Science Foundation to KAS and
446 RS Ostfeld (DEB 0089588) and to CG Jones, BJ Goodwin, RS Ostfeld and EMS (DEB
447 0212215). BP Kotler and CJ Whelan provided helpful comments on a previous draft of this
448 manuscript. We thank Marius van der Merwe and Joel S. Brown for graciously sharing their
449 unpublished research on the landscape of fear in African ground squirrels. Last of all, we thank
450 BP Kotler, JS Brown, and DW Morris for their invitation to contribute to this special feature on

451 behavioral indicators, but most of all of their vision and leadership for bridging behavioral
 452 ecology and conservation.

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FIGURE LEGENDS

Figure 1. A simplified framework for viewing the operation of patch-use based interactions on prey mortality through the predator. Predator-free or predator-poor space arises through bottom-up and top-down factors that influence the predator's QHR. The QHR is an integrative measure that is influenced by, among other factors not shown, the predator's density, and the abundances of primary food and predation risk from apex predators. When possible, we indicate the likely relationship(s) (0, +, -) between two factors when the first is increased, e.g., increasing the QHR creates more behavioral refugia and reduces foraging time.

Figure 2. Creation of nest predator-free space and consequences for incidental prey, i.e., passerine nests. There are two ways to interpret the figure (i) bottom-up and (ii) top-down creation. In (i) local primary prey density decreases from blue to red contour lines, whereas risk of attack from apex predators is homogeneous throughout the environment. The predator has a fixed quitting harvest rate (QHR) that increases from panel (a) → (c). Predator-free space is shown by removing contour lines in panels (b) and (c); i.e., predator-free space occurs where local resource abundance < QHR. A higher QHR creates more predator-free space and fewer nests are vulnerable to predation.

Alternatively, in (ii) contour lines indicate increasing (blue to red) QHR due to local risk from apex predators, whereas the abundance of primary prey is homogeneous throughout the environment. Primary prey abundance

602 In either case, In nature both scenarios are likely to apply simultaneously; the difference is
 603 whether one conceives of spatial heterogeneity created primarily through the distribution or
 604 reward (i) or risk (ii). In all cases, predator-free space occurs where local primary prey densities
 605 falls below the predator's local QHR. Increasing the QHR changes the distribution and
 606 decreases the abundance of predator-free space, which is especially important for incidental prey
 607 such as songbird nests (represented by the clutch of eggs). Maybe we should just stick to
 608 illustrating bottom-up creation since I am finding it hard to illustrate both in simple terms that
 609 will avoid confusion. Ignore the highlighted section if that the case.

610

611 Figure 3. (A) Proportion of space used and (B) prey consumption as a function of a predator's
 612 giving-up density (GUD) in a simulated landscape produced in MATLAB. Primary (n=1000)
 613 and incidental (n=50) prey were randomly and independently assigned to one of 100 patches
 614 (this produces a Poisson distribution of prey). Only patches with initial primary prey quality >
 615 GUD were foraged, and of these final prey density = GUD. Random search time for a given
 616 foraged patch (i) was calculated as: $t_i = (1/a_p) \log(N_i/GUD)$, where a_p is the encounter rate on
 617 primary prey and N_i and GUD are, respectively, the initial and final number of primary prey in
 618 patch (i). The probability, I_p , an individual incidental prey item was consumed from a foraged
 619 patch was calculated as: $I_p = [1 - e^{-a_i t_i}]$. Actual incidental prey consumption was determined by
 620 comparing I_p to a random number (uniform distribution). Figure shows the mean results from
 621 1000 simulations at each specified GUD. $a_p = 0.6$, $a_i = 0.6$.

622

623 Figure 4. Relationships between (A) space use (quantified as the proportion oat grains removed
 624 by mice off a lattice grid arrangement) and mouse giving-up density (log-transformed) and (B)

625 the percentage of nests depredated by mice and space use. Each data point represents one of six
626 2.25 ha plots from which the data were collected. See Schmidt and Ostfeld (2003) for more
627 details of the experimental methods. Figures reprinted with permission from Schmidt and
628 Ostfeld (2003).

629

630 Figure 5. Artificial nest daily mortality rates (DMRs = number of nest depredated/nest exposure
631 days; after Mayfield 1975) as a function of rodent abundance. (A) Per capita nest DMR (log-
632 transformed) specific to mice (solid circles) or chipmunks (open circles) as determined by incisor
633 marks on clay eggs. Each data point represents one of six 2.25 ha plots from which the data
634 were collected (3 plots omitted for mice due to an experimental mouse removal). (B) Overall
635 (all predators included) nest DMRs (\pm SE) on control (n=3) or removal (n=3) plots; mice were
636 removed in 1997, chipmunks in 1998). See Schmidt et al. (2001a) for more details of the
637 experimental methods. Figures reprinted with permission from Schmidt et al. (2001a).

638

639 Figure 6. Veery nest success (proportion of nests fledged) as a function of zero-capture data:
640 proportion of the 9 traps stations comprising a trap neighborhood that recording zero mouse
641 captures during spring-summer trapping sessions; see Schmidt et al. 2006 for more details).

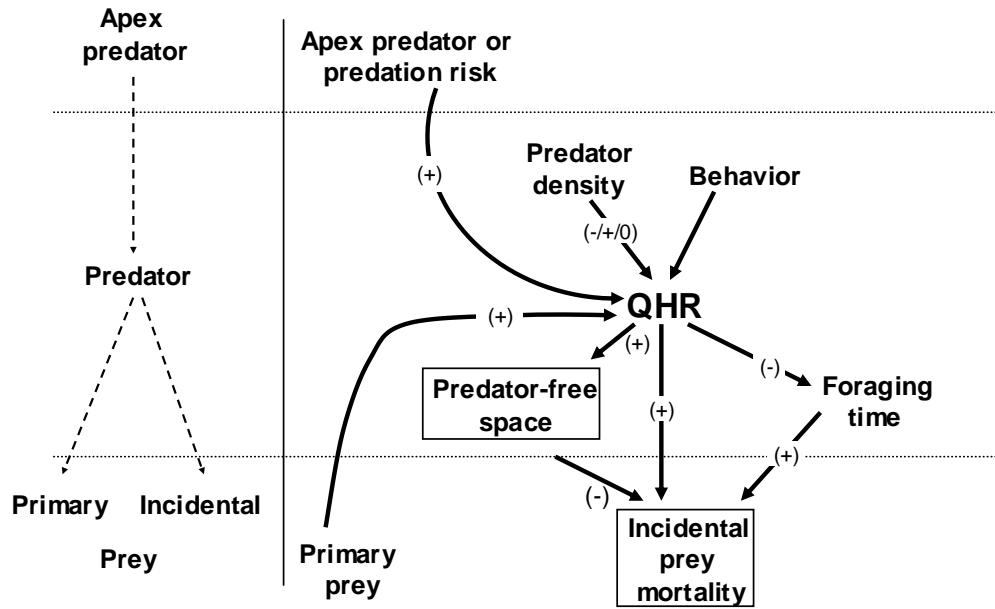
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643 Figure 7. Mean (+ SE) daily predation of gypsy moth pupae deployed on (1.5 m height) or at the
644 base of (ground) trees, as a function of the percentage of track plates recording mouse tracks at
645 those trees over a 10-12 day period at the Institute of Ecosystem Studies, Millbrook, New York,
646 2002-2005.

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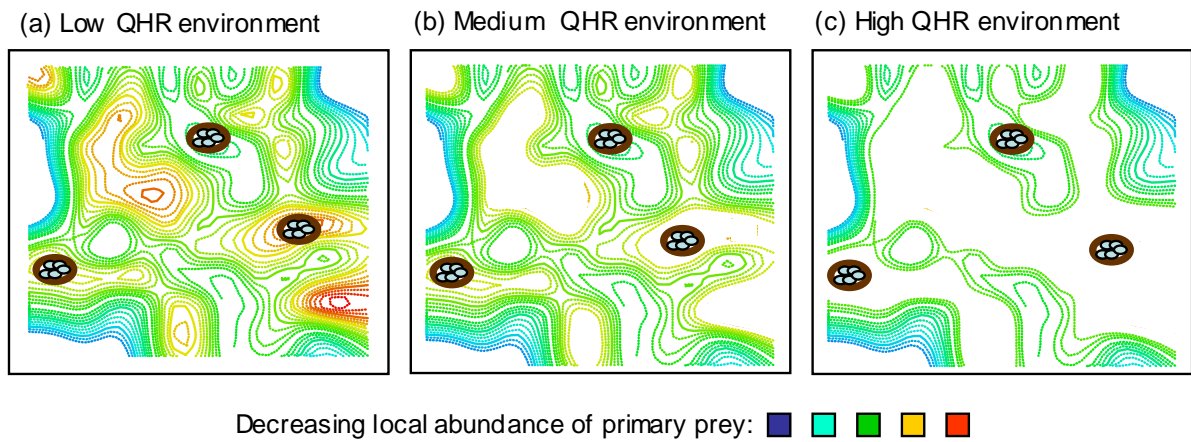
649 **FIG. 1**
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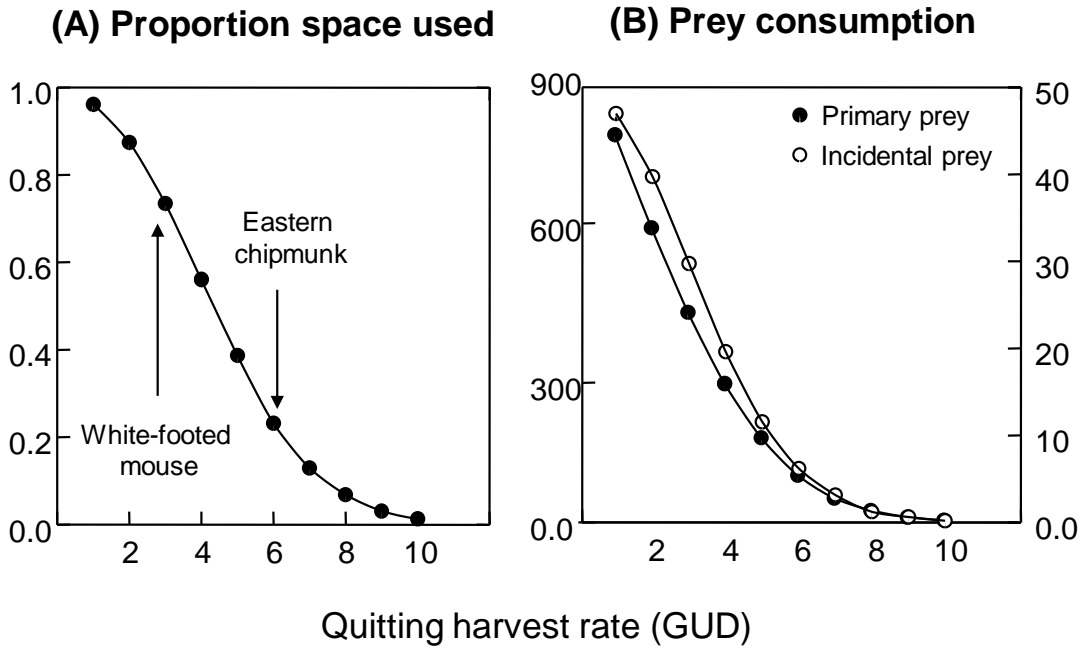
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659 **FIG. 2**
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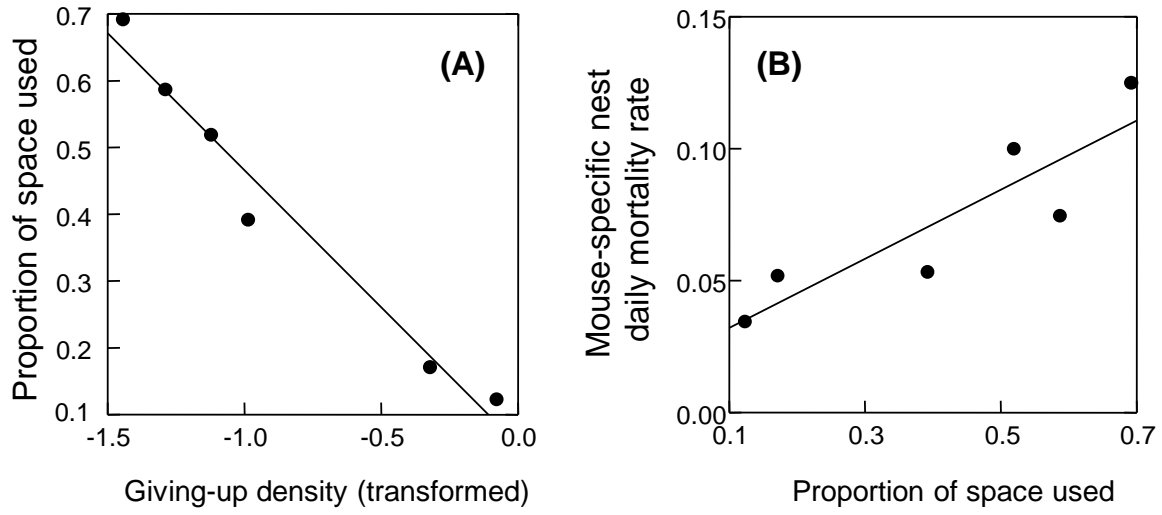
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665 **FIG. 3**
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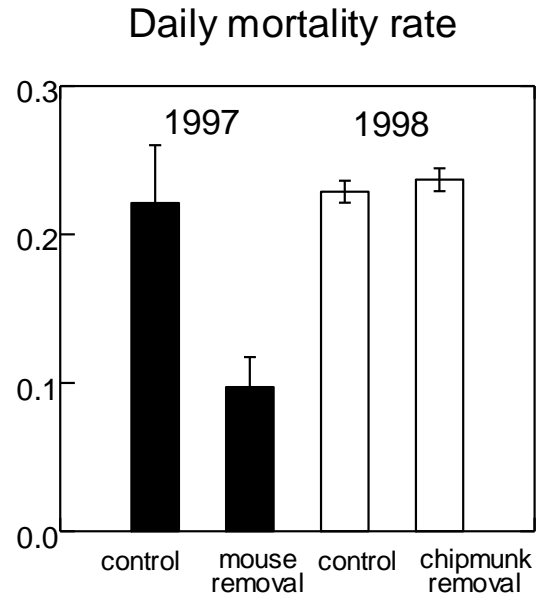
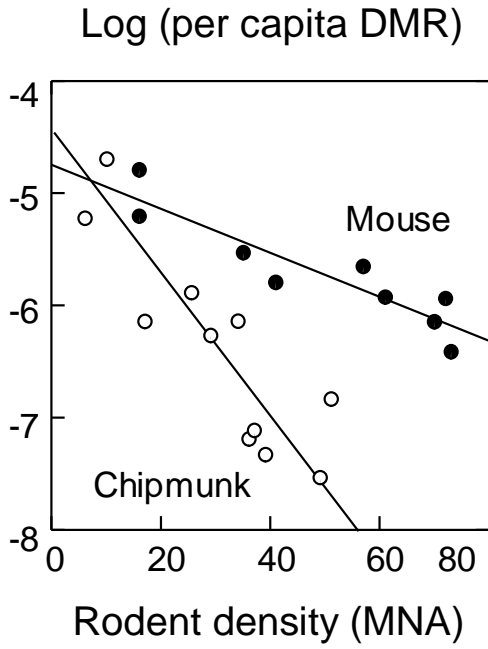
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670 **FIG. 4**
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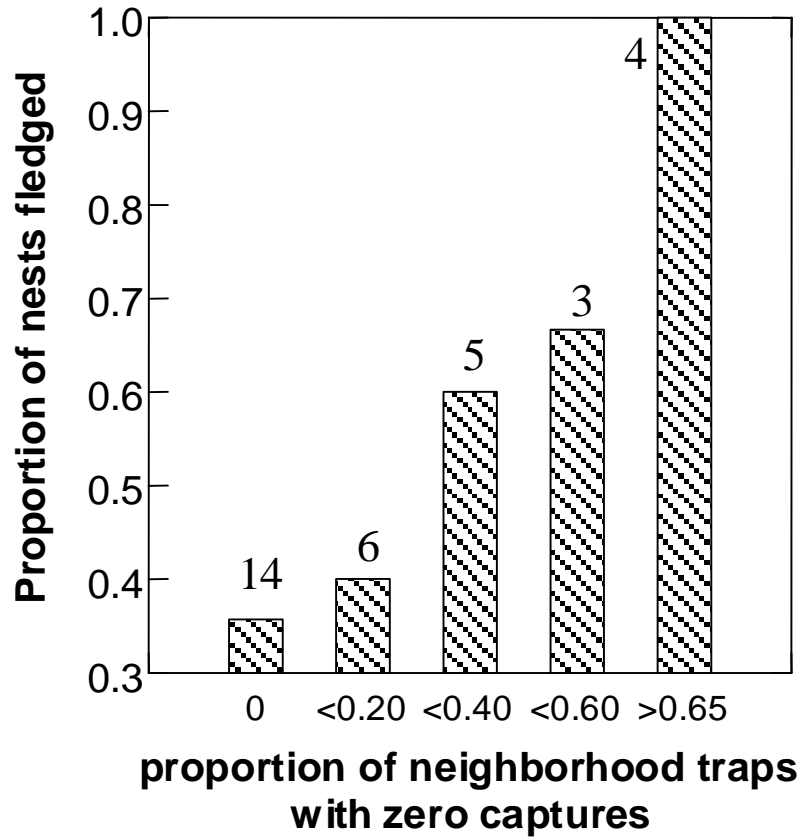
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677 **FIG. 5**
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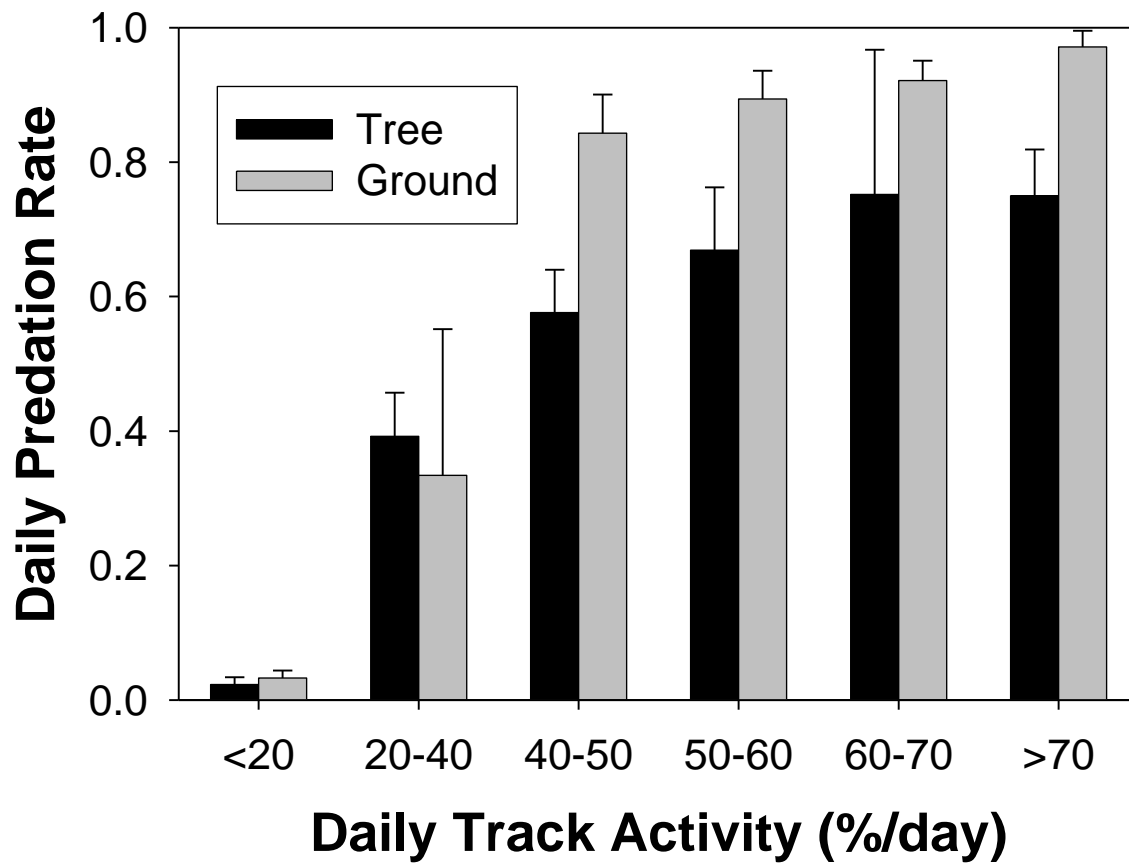
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693 **FIG. 6**
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707 Fig. 7

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