

1 No respect for apex carnivores: distribution and activity patterns of honey badgers in the Serengeti

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9

10 **Abstract**

11 Honey badgers are cryptic carnivores that occur at low densities and range across large areas. The  
12 processes behind site-level honey badger abundance and detection rates are poorly understood, and there  
13 are conflicting results about their avoidance of larger carnivores from different regions. We used data  
14 from 224 camera traps set up in the Serengeti National Park, Tanzania to evaluate patterns in detection  
15 rates, spatial distribution, and activity patterns of honey badgers. Our top models showed that the relative  
16 abundance of larger carnivores (e.g., African lions, *Panthera leo*, and spotted hyenas, *Crocuta crocuta*)  
17 was important, but surprisingly was positively related to honey badger distribution. These results suggest  
18 that honey badgers were not avoiding larger carnivores, but were instead potentially seeking out similar  
19 habitats and niches. We also found no temporal avoidance of larger carnivores. Honey badgers exhibited  
20 seasonal variation in activity patterns, being active at all times during the wet season with peaks during  
21 crepuscular hours, but having a strong nocturnal peak during the dry season. Our detection rates of honey  
22 badgers at individual camera traps were low (3,402 trap nights/detection), but our study shows that with  
23 adequate effort camera traps can be used successfully as a research tool for this elusive mustelid.

24

25 Keywords: abundance, activity patterns, distribution, honey badger, interspecific interactions,

26 *Mellivora capensis*

27

28 The geographical patterns of a species' abundance affect range limits, gene flow and population dynamics  
29 (Brown 1984, Sagarin et al. 2006). The geographical range and abundance of species, however, can be  
30 highly variable: a species can be widespread but have low abundance throughout the range or only be  
31 found in a small geographic range but have high abundance (Rabinowitz et al. 1986). The former likely  
32 best describes honey badgers (*Mellivora capensis*), which are large, solitary mustelids with extensive  
33 home ranges (Johnson et al. 2000, Begg et al. 2005). Honey badgers exist at low densities across the  
34 majority of their range, which includes most of sub-Saharan Africa and portions of the Arabian and  
35 Indian peninsulas (Do Linh San et al. 2016). Despite their extensive distribution, honey badgers have not  
36 been well studied (Begg et al. 2003, Proulx et al. 2016), and there is a need to better understand their  
37 distribution, abundance, habitats and ecology to help define local and overall population status (Do Linh  
38 San et al. 2016) and develop proper conservation programs (Proulx et al. 2016).

39         Detection rates and abundance estimates for carnivores can be affected by many factors.  
40 Detection rates are affected by survey intensity (Rovero and Marshall 2009) and how well surveys  
41 account for the habitat use of the species (Meek et al. 2014), and may also be influenced by species  
42 behavior and activity patterns. For example, seasonal patterns in distribution and activity are often based  
43 on breeding behavior (Vogt et al. 2014, Allen et al. 2015) or prey activity (Sinclair 1979, Durant et al.  
44 2010) and modulated by the light regime (Heurich et al. 2014). The distribution of smaller carnivores can  
45 also be affected by the presence, abundance and activity of larger carnivores (Durant 1998, Wang et al.  
46 2015, Newsome et al. 2017). Honey badgers are sometimes killed by large carnivores (Begg 2001), but  
47 previous studies have given conflicting results regarding honey badger reactions to larger carnivores (e.g.,  
48 Ramesh et al. 2017, Rich et al. 2017), making it unclear whether potential intraguild predation results in  
49 spatial and/or temporal avoidance.

50         The main difficulties in obtaining accurate estimates of honey badger abundance are their large  
51 home ranges, cryptic behavior, and low densities. Mean sizes of home ranges in Kalahari semi-desert  
52 were estimated as 541 km<sup>2</sup> for adult males, 151 km<sup>2</sup> for young males, and 126 km<sup>2</sup> for adult females  
53 (Begg et al. 2005). These relatively large home ranges compared to body size were attributed to low prey

54 availability and a long period of cub dependence (12–16 months). Honey badgers have been documented  
55 in a variety of habitats from rain forests (Bahaa-el-din et al. 2013, Greengrass 2013) to woodlands (Bird  
56 and Mateke 2013) and deserts (Begg et al. 2003), but preferences and selection of specific habitats, as  
57 well as effects of habitat characteristics on local abundance are largely unknown. Honey badgers are often  
58 nocturnal (Bird and Mateke 2013), but activity in South Africa appears to shift to diurnal during the cool  
59 season (Begg et al. 2016).

60 Camera traps are being increasingly used to understand the populations (Chandler and Royle  
61 2013, Parsons et al. 2017) and behaviors of cryptic wildlife (Vogt et al. 2014, Allen et al. 2016). The use  
62 of systematic grid sampling with camera traps allows for continuous surveying across large spatial and  
63 temporal scales. We used this approach in a National Park in Tanzania to understand patterns in detection  
64 rates of honey badgers. Our objectives were to 1) calculate detection rates and determine the viability of  
65 using camera traps for studying honey badgers, 2) determine whether honey badger activity varied  
66 between seasons, and 3) determine the variables that are driving the distribution of detections for honey  
67 badgers, for which we used an *a-priori* modeling framework to test a series of model hypotheses  
68 composed of 10 variables explaining honey badger distribution.

69 Camera trap surveys were conducted in the Serengeti National Park (hereafter referred to as  
70 Serengeti), Tanzania. The Serengeti is characterized by open woodlands in the northern portions, with  
71 treeless grass plains in the south (Grant et al. 2005). Rocky outcrops or ‘kopjes’ support trees and shrubs  
72 and are found throughout the Serengeti, often providing the only cover on the plains (Durant 1998). Mean  
73 annual rainfall ranges from 350 mm in the southeast to 1200 mm in the northwest (Norton-Griffiths et al.  
74 1975) and is concentrated in the wet season (November to May) with very little rain in the dry season  
75 (June to October) and temperatures remaining consistent across seasons (Sinclair 1979, Durant 1998).

76 We used data collected as part of the Snapshot Serengeti Project (see Swanson et al. 2015, 2016)  
77 for monitoring the carnivore community. The project systematically deployed 224 camera traps in a  
78 randomly distributed grid using 5 km<sup>2</sup> intervals, with camera traps placed in strategic locations within 250  
79 m of the center of each grid to maximize wildlife detection. We used the data from July 2010 to May

80 2013 for this study. The date, time, and camera trap site ID was recorded for each photograph, and the  
81 project used citizen scientists to determine the species and behavior (standing, resting, moving or eating)  
82 in the photographs (Swanson et al. 2016). We then proofed each of the photos that were tagged as honey  
83 badgers, and subsequently removed 5 events that had been misidentified as honey badgers. To reduce  
84 pseudoreplication, we considered photos of a species at a camera trap site within 30 min of a previous  
85 photo to be the same event (Wang et al. 2015, Rich et al. 2017). Events with multiple individuals were  
86 treated as a single event with the highest number of individuals documented. We then collected habitat  
87 variables (Appendix A) for each camera trap using available layers in Arc GIS (Environmental Systems  
88 Research Institute, Redlands, CA).

89 We used program R version 3.3.1 (R Core Team 2016) for all statistical analyses. We first  
90 calculated summary statistics for our data and then tested our series of hypotheses and models.

91 We calculated our detection rate as the ‘number of detections per trap night’ for each camera trap,  
92 and reported overall detection rates, and the range and mean for individual camera trap sites. We then  
93 totaled the number of independent events ( $E$ ) for each species, and determined their relative abundance  
94 ( $RA$ ) at each camera trap as:

$$95 \quad RA = (E / TN) \times 100$$

96 where  $TN$  is the total number of trap nights that the camera trap was operational. We used relative  
97 abundance as our measure because they are often an accurate index of abundance (Parsons et al. 2017),  
98 and our limited sample sizes precluded us from using occupancy.

99 To understand the activity patterns of honey badgers we first changed the time of each event to  
100 radians and then fit the data to a circular kernel density and used its distribution to estimate the activity  
101 level using the *overlap* package (Ridout and Linkie 2009). To determine whether honey badger activity  
102 varied by season we used the *overlapEst* function to test whether activity patterns varied between the wet  
103 and dry season, where we considered  $\Delta_1 > 0.80$  to be strong overlap and 0.50-0.79 to be medium overlap  
104 (Lynam et al. 2013). We then used the *overlapEst* function to test whether honey badgers varied in  
105 activity patterns from larger carnivores (spotted hyena, *Crocuta crocuta*, and leopard, *Panthera pardus*).

106 We did not test for African lions because their activity data was biased in favor of mid-day activity,  
107 because several camera traps were located at location regularly used as a daybed (A. Swanson, Personal  
108 Comm.), and therefore we believe that the records do not reflect true circadian activity for this species.

109 To understand our detection patterns for honey badgers we tested among the 14 *a-priori* models  
110 (Table 1). We fitted our models using Generalized Linear Models (GLMs) with a binomial logit link  
111 indicating whether a honey badger was detected or not at a camera trap. We compared our models using  
112 AIC weight ( $AIC_w$ ) (Burnham and Anderson 2002), and considered each of the models until a cumulative  
113  $AIC_w = 0.90$  to be our top models.

114 We had a total of 98,644 trap nights at 224 camera traps. We detected 29 honey badger events at  
115 23 individual camera traps, among which 6 events included 2 honey badgers. In 86% of events the honey  
116 badgers were moving, but we also documented standing (10%) and eating (4%).

117 Our overall detection rate among all camera traps was 3,402 trap nights/detection. Our detection  
118 rates for the 23 individual camera traps with detections ranged from 94 to 720 trap nights/detection, with  
119 a mean rate for individual camera traps of 404.4 ( $\pm 40.6$  SE) trap nights/detection. October had the  
120 highest number of detections ( $n = 5$ ), followed by 3 detections in each of February, March, and June. No  
121 detections were recorded in July or August.

122 We found that honey badger activity was generally similar among seasons with a medium overlap  
123 ( $\Delta_1 = 0.69$ ). The main difference being that during the wet season ( $n = 15$ ) honey badgers were active  
124 during all times, with peaks during crepuscular hours, but during the dry season ( $n = 14$ ) they showed a  
125 strong nocturnal peak (Figure 1). Honey badgers had strong temporal overlap with both spotted hyenas  
126 ( $\Delta_1 = 0.85$ ) and leopards ( $\Delta_1 = 0.82$ ) (Figure 1).

127 Our best model 'Lion Abundance' had 2 times more explanatory value ( $AIC_w = 0.57$ ) as the next  
128 best model 'Apex Carnivore Avoidance' ( $AIC_w = 0.27$ ). These two were our top models (Table 2),  
129 however, the variables in the models did not always affect honey badger abundance in the way we  
130 predicted. In 'Lion Abundance', African lions had a positive relationship ( $\beta_{lion} = 0.009$ ), and in 'Apex

131 Carnivore Avoidance', both African lions and spotted hyenas had positive relationships ( $\beta_{\text{lion}} = 0.009$ ,  
132  $\beta_{\text{hyena}} = 0.001$ ).

133 The small number of studies focused on honey badgers in the published literature may be directly  
134 related to the difficulty of finding and detecting this elusive mustelid. In our study, we documented 29  
135 honey badger events across 98,644 trap nights (an average of 3,402 trap nights needed for each detection).  
136 Some of the camera traps had notably higher success, but even our most frequently visited camera trap  
137 still needed on average 94 nights per detection. These detection rates appear to be much lower than those  
138 found with nocturnal vehicle surveys (e.g., 0.1 honey badgers/km; Waser 1980), or those of other  
139 carnivores (e.g., Ramesh et al. 2017, Rich et al. 2017). The low detection rate limited our ability to detect  
140 differences in behaviors and detections by month or time of day, and also precluded us from using  
141 occupancy modeling for our analyses.

142 Our top models suggest that honey badger distribution is not negatively affected by interspecific  
143 interactions with larger carnivores, but instead honey badgers seek out similar habitats and niches as  
144 larger carnivores. In our top two models the interactions with African lions and spotted hyenas were  
145 important, while our third best model also included leopards, and presence of all three species coincided  
146 with higher detection of honey badgers at camera trap sites. Honey badgers also exhibited strong temporal  
147 overlap with spotted hyenas and leopards suggesting there was also no temporal avoidance of larger  
148 carnivores. It could be that our low sample size limited our ability to determine what drives site-level  
149 detection of honey badgers, but the data suggest that honey badgers are not avoiding larger carnivores in  
150 Serengeti, as was suggested by Ramesh et al. (2017) for badgers and leopards in South Africa. Instead,  
151 our results support the study by Rich et al. (2017) and Balme et al. (2017), who reported a lack of  
152 repulsion of carnivore guild species. Although it has been documented that large carnivores occasionally  
153 kill honey badgers (Begg 2001), honey badgers are notorious for their aggressive behavior and high  
154 resource holding potential even when faced with much larger carnivores (Estes 1992). Their aggressive  
155 threat display is often successful in preventing predation even against largest of carnivores, such as lions,  
156 spotted hyenas, and leopards (Begg 2001). Our results suggest that honey badgers may exhibit less

157 avoidance of large carnivores in space and time than other, less aggressive mesocarnivores, which are  
158 often constrained by apex carnivores (Durant 1998, Hayward and Slotow 2009, Newsome et al. 2017).

159         The benefit of systematic grid sampling with camera traps is that it allows for continuous  
160 surveying across large spatial and temporal scales (Swanson et al. 2015, Rich et al. 2017) while not  
161 affecting the activity or behavior of wildlife (Vogt et al. 2014, Allen et al. 2016). We found that honey  
162 badgers exhibited seasonal variation in their activity, which is in accord with studies from other  
163 ecosystems (Begg et al. 2016, Bird and Mateke 2013). During the wet season honey badgers were active  
164 throughout the 24-hour cycle, but during the dry season honey badgers showed a strong nocturnal peak.  
165 Honey badgers avoid the hottest times of the day by staying in burrows in other systems (Begg et al.  
166 2003, Bird and Mateke 2013), but temperatures in the Serengeti stay similar throughout the year,  
167 suggesting that the circadian cycles are not dictated solely by temperatures. This could, however, be an  
168 artifact of our low sample sizes, or the placement of cameras as honey badgers may be seeking out shaded  
169 areas during day similar to African lions.

170         Adjustments to the systematic camera trap design used may allow for higher detection rates in the  
171 future. First, it is important to understand that these camera traps were placed in areas that would allow  
172 for detections of all wildlife species (Swanson et al. 2015), and not just honey badgers. A camera trap  
173 study targeting honey badgers could increase success by choosing microhabitat variables most frequently  
174 used by honey badgers, such as heterogeneous aspects of the landscape, or placing camera traps near  
175 known tracks, marking sites, food sources or dens. Adjustments like these may be especially important  
176 for species like honey badger that are characterized by very low detection rates. Nevertheless, due to the  
177 large home ranges of honey badgers and low population densities, detection rates are always likely to be  
178 low. The camera traps were placed in the center of 5 km<sup>2</sup> grids, which should be sufficiently large enough  
179 for determining population sizes of honey badgers in a spatially explicit capture recapture modeling  
180 framework (Chandler and Royle 2013). Other important topics for future research include increasing  
181 detection rates, understanding honey badger selection of microhabitat features, and determining what may  
182 affect the distribution of honey badgers. While research of honey badger is difficult and presents many



183 challenges, we encourage future research to understand their ecology and to develop effective  
184 conservation and management efforts, where needed.

185

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189 earlier drafts that greatly improved the manuscript.

190

### 191 **Appendix A. Supplementary data**

192 Supplementary data associated with this article can be found, in the online version, at <http://...>

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194

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289

290 Table 1. Our *a-priori* models to explain honey badger distribution at camera sites in Serengeti National  
 291 Park. We provide the name of the model, the variables included and the hypothesis behind the model  
 292 together with references used to develop it.

293

Name	Variables	Hypothesis
Habitat Preference	HABT*GRHT	Distribution of honey badgers will be dictated by their preferred habitat (Durant et al. 2000, Vanderhaar and Hwang 2003).
Water Availability	RIVR	The distribution of honey badgers will be dictated by the proximity to water and riparian habitats (Durant et al. 2000).
Habitat and Water Availability	HABT * GRHT + RIVR	The distribution of honey badgers will be dictated by a combination of available habitat and water (Durant et al. 2000, Vanderhaar and Hwang 2003).
Cover	COVR	The availability of cover will dictate the distribution of honey badgers (Durant et al. 2000).
Kopje	KOPJ	The availability of kopjes for cover, possible den sites, and habitat heterogeneity will dictate the distribution of honey badgers (Durant et al. 2000).
Cover and Kopje	COVR + KOPJ	The distribution of honey badgers will be dictated by areas that provide cover (Durant et al. 2000).
Human Abundance	HUAB	The distribution of honey badgers will be dictated by trying to avoid areas frequently used by humans (Proulx et al. 2016).
Leopard Abundance	LPAB	The distribution of honey badgers will be dictated by trying to avoid areas frequently used by leopards (Begg 2001, Do Linh San et al. 2016, Ramesh et al. 2017).
Leopard Abundance and Cover	LPAB + COVR	The distribution of honey badgers will be dictated by avoiding leopards and cover to escape (Ramesh et al. 2017).
Lion Abundance		The distribution of honey badgers will be dictated by trying to avoid areas used by African lions (Begg 2001, Do Linh San et al. 2016).
Cascading Carnivore Abundance	LPAB * LNAB	Honey badgers will be more frequently found where lions are abundant due to their deterring the leopards (Ramesh et al. 2017)
Apex Carnivore Abundance	LNAB + HYAB	The distribution of honey badgers will be dictated by trying to avoid areas frequently used by group-living apex carnivores (Begg 2001, Do Linh San et al. 2016).
Trapping Effort	TRAP	The number of camera trap nights will have greater influence on the distribution of honey badgers than biological factors (Wegge et al. 2004).

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295

296 Table 2. The results of our *a-priori* model selection for honey badger distribution in Serengeti National  
 297 Park, with individual models ranked based on their AIC<sub>w</sub>.

298

Name	Variables	AIC	ΔAIC	AIC <sub>w</sub>	Cum AIC <sub>w</sub>
Lion Abundance	LNAB	140.09	0.00	0.57	0.57
Apex Carnivore Avoidance	LNAB + HYAB	141.58	1.49	0.27	0.84
Cascading Predation	LPAB * LNAB	143.64	3.55	0.10	0.94
Cover and Kopje	COVR + KOPJ	146.82	6.73	0.02	0.96
Cover	COVR	146.94	6.85	0.02	0.98
Leopard Abundance and Cover	LPAB + COVR	148.89	8.80	0.01	0.99
Kopje	KOPJ	149.89	9.80	0.00	0.99
Human Avoidance	HUAB	150.18	10.09	0.00	0.99
Trapping Effort	TRAP	151.88	11.79	0.00	1.00
Water Availability	RIVR	152.13	12.04	0.00	1.00
Leopard Abundance	LPAB	152.53	12.44	0.00	1.00
Habitat and Water Availability	HABT * COVR + RIVR	152.62	12.53	0.00	1.00
Habitat Preference	HABT * GRHT	154.94	14.85	0.00	1.00

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301 Figure 1. The temporal activity and overlap (gray-shaded area), estimated with kernel densities, of honey  
302 badgers and large carnivores in the Serengeti. (A) represents honey badger activity during different  
303 seasons, with wet season represented as a solid line and dry season represented as a dotted line. (B&C)  
304 represent overlap of honey badger activity with spotted hyenas (B) and leopards (C), where honey badger  
305 activity is represented as solid lines and the large carnivore activity as a dotted line.

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