

Foraging constraints reverse the scaling of activity time in carnivores

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Abstract36
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The proportion of time an animal spends actively foraging in a day determines its long-term fitness. Here, we derive a general mathematical model for the scaling of this activity time with body size in consumers. We show that this scaling can change from positive (increasing with size) to negative (decreasing with size) if detectability and availability of preferred prey sizes is a limiting factor. These predictions are supported by a global dataset on 73 terrestrial carnivore species from 8 families spanning >3 orders of magnitude in size. Carnivores weighing $\sim 5\text{kg}$ experience high foraging costs because their diets include significant proportions of relatively small (invertebrate) prey, and therefore show an increase in activity time with size. This shifts to a negative scaling in larger carnivores as they shift to foraging on less-costly vertebrate prey. Our model can be generalized to other classes of terrestrial and aquatic consumers, and offers a general framework for mechanistically linking body size to population fitness and vulnerability in consumers.

48 Introduction

49 The activity budget of an animal, that is, the time that it allocates to different behaviors on
50 a daily basis, strongly affects its fitness by determining interaction rates with its resources,
51 predators and competitors. The activity budget can also inform conservation efforts by helping
52 predict the spatial and temporal distribution of resources necessary for an animal population to
53 remain viable under habitat loss or climate change^{1,2,3}. For example, accurate models of long-
54 term activity and geographical ranges of mammalian carnivores are necessary for determining
55 the appropriate size of protected areas of endangered mammals^{1,4}.

56 However, although empirical patterns of activity budgets in the field are now widely recorded
57 due to improved tracking technologies, there is currently no theoretical framework for predicting
58 them, or generalizing our knowledge of these patterns across species, habitats, or foraging
59 strategies. In particular, although animals can vary widely in how they manage their time,
60 there is great potential for developing predictive models for activity budgets by including general
61 constraints due to biomechanical (e.g., locomotion and searching) and energetic (e.g., basal or
62 resting metabolic rate) limitations^{5,6,7}.

63 Activity time on a daily basis should be strongly related to the minimum energetic require-
64 ments of an animal — more the energy requirement per unit time, more the time spent actively
65 foraging for resources. An animal’s energetic requirements can be estimated from its metabolic
66 rate, which scales positively with body mass (m). In the case of vertebrates, this scaling ranges
67 between $m^{0.65}$ (for field metabolic rate) and $m^{0.9}$ (for active metabolic rate)^{8,9,10,11}. Therefore,
68 metabolic rate per unit body mass (mass-specific metabolic rate) in vertebrates scales nega-
69 tively with body size with an exponent ranging between -0.35 and -0.1 . That is, individuals
70 from small-sized species have higher maintenance costs per unit body mass and necessarily need
71 to consume energy at a faster rate than those from larger ones. In other words, all else being
72 equal, smaller vertebrates need to forage for longer periods than larger ones. This leads to a
73 simple prediction: the amount of time a vertebrate spends active in its daily cycle also scales
74 with body mass within the range $m^{-0.35} - m^{-0.1}$.

75 However, this prediction of a negative scaling of activity time with body size hinges on the
76 key assumption that energy intake rate scales identically to the rate of energy use (metabolic
77 rate). In reality however, intake rates in the field are typically limited by resource availability
78 (i.e., prey abundance), and ability of the consumer to search, detect, attack and handle prey.
79 This can result in deviations of the scaling of intake rate from that of metabolic rate^{12,13,15}.
80 Because of these constraints on field intake rates, animals of different sizes need to optimize
81 their activity budget by choosing the right resource sizes (e.g., many small vs. few large prey)
82 and foraging strategy (e.g., active-capture vs. sit-and-wait) to meet their energetic needs^{16,13,17}.
83 Thus these limitations on foraging, and therefore intake rates, may ultimately lead to deviations
84 from the expectation of a universal decrease in activity time with body size.

85 Here we derive a general mathematical model for the size-scaling of the activity budget of
86 consumers under field conditions, incorporating key metabolic and biomechanical constraints
87 on foraging costs. We then develop a specification of the model appropriate for terrestrial
88 mammalian carnivores, and test its predictions by compiling a global dataset of high-resolution
89 activity budgets. In the Discussion we show how our model can be specified or extended to
90 other classes of consumers.

91 Results

92 Our model links consumer and resource body size to the minimum proportion of time (T_p) that
93 an individual consumer must spend foraging on a daily basis to maintain its energy balance (see

94 Methods, and Supplementary Information for detailed derivation):

$$95 \quad T_p = \frac{B_0 m^\beta}{B_0 m^\beta - A_0 m^\alpha + I_0 m^\iota} \quad (1)$$

96 Here, m is an individual consumer's adult average body mass (its size), I_0 , B_0 , and A_0 are
97 the standard (i.e., for a 1 kg consumer) intake, resting metabolic, and active metabolic rates
98 respectively, while the scaling exponents ι , β and α respectively quantify the size-dependence
99 of these three fundamental rates. This model makes a key prediction (see Methods and Sup-
100 plementary Information for derivation): if l is the exponent of the scaling of energy loss (either
101 while resting or actively foraging), if

$$102 \quad \iota < l, \quad (2)$$

103 the scaling relationship of T_p with body size changes from negative (T_p decreasing with size) to
104 positive (T_p increasing with size). The critical value l for the intake rate scaling exponent (at
105 which the scaling relationship reverses) is expected to lie between approximately 0.70 if resting
106 (through the exponent β) dominates energy expenditure, and 0.80 if active foraging (through
107 the exponent α) dominates. This result about the reversal of activity time scaling remains
108 robust to considerable variation in scaling exponents due to uncertainty in their estimation
109 as well as biological variation such as differences in the scaling of basal, field, and maximum
110 metabolic rates (Supplementary Information).

111 As such, equation (1) and the prediction of a critical value of intake rate scaling (equa-
112 tion (2)) provides an simple, intuitive model for determining the necessary intake rate scaling
113 to maintain a negative scaling of activity time with increasing body size. Furthermore, for a
114 given scaling of active and resting metabolic rates equation (1) can be used to estimate the the
115 body mass threshold below which the scaling exponent of intake rate must increase to maintain
116 activity time below a biologically feasible limit (for example, assuming that the daily activity
117 time proportion cannot exceed, say, 0.5).

118 Next, to obtain a mechanistic basis for the constraints on intake rate scaling and determine
119 where the body mass threshold for a qualitative change in the daily activity proportion may lie
120 under field conditions, we show that ι can be broken down into the contributions of different
121 constraints on foraging, and therefore intake rate:

$$122 \quad \iota = p_v + p_x + (p_k + 1)p_d \quad (3)$$

123 Here, p_v is the size-scaling exponent of body velocity, p_x the exponent of prey abundance, p_k is
124 the exponent of resource size relative to consumer size, i.e., the resource-consumer (e.g., prey-
125 predator) size ratio, and p_d is the exponent of detection distance. Of these, p_x , p_k , and p_d are
126 most important because body velocity (through its scaling exponent p_v) contributes to both
127 energy gain and loss while actively foraging. These are ubiquitous constraints imposed by field
128 conditions on the intake rate and therefore activity time: the prey-predator size-ratio scaling
129 determines the effect of availability of or preference for resources of different sizes relative to
130 predator, reaction distance scaling determines the effect of detectability of these resources, and
131 abundance scaling determines the effect of prey rarity which translates into higher foraging costs
132 by decreasing the number of resource encounters per unit time spent foraging. These foraging
133 constraints can be interpreted partly as inherent, biomechanical constraints of consumers, and
134 partly as properties of the local ecological conditions, which may constrain availability of prey
135 of different sizes.

136 An example of the effect of each of these three foraging constraints on the predicted scaling
137 of intake rate and therefore, activity time, is shown in Fig. 1. To generate these predictions, we
138 use size-scaling relationships for energy loss rates (B and A) and components of intake rate (I)
139 appropriate for terrestrial mammalian carnivores, as these have been extensively studied^{16,19}.
140 This yields the prediction that terrestrial carnivores are expected to have a scaling of activity

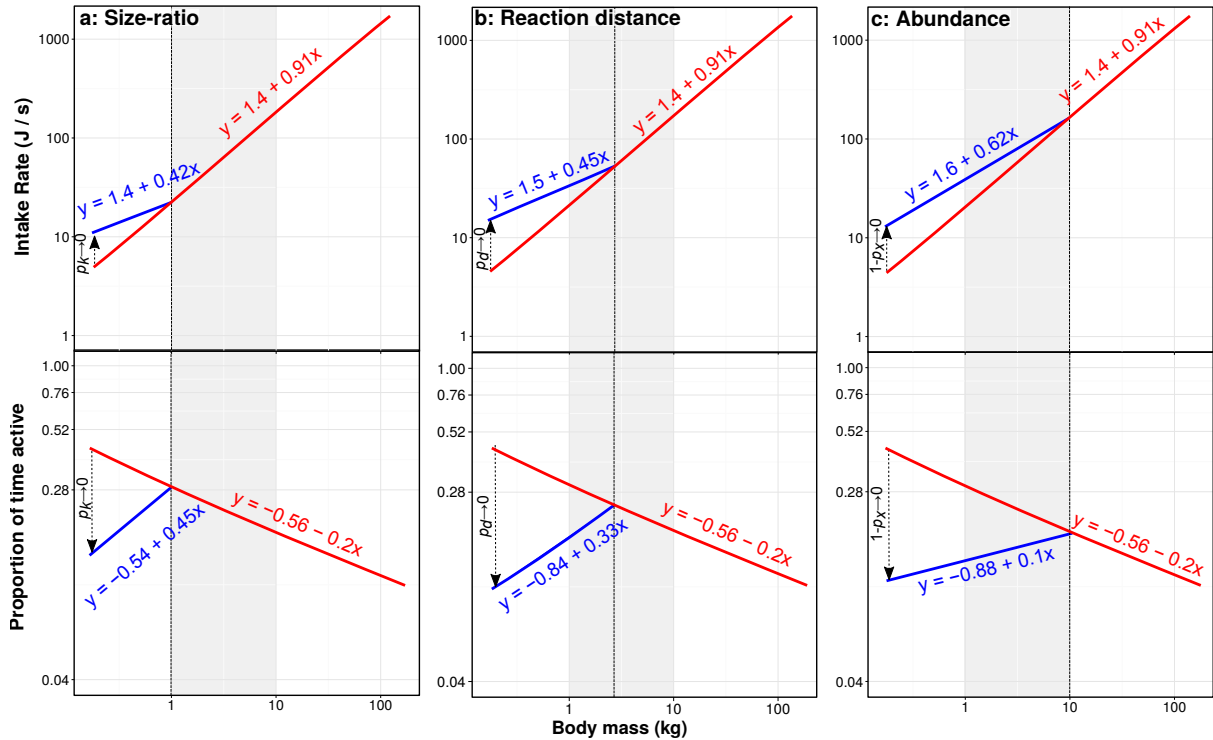


Figure 1: **The size-scaling model of activity time in terrestrial carnivores.** The three panels show effect on scaling of the activity time (lower figures), due to a weakening of intake rate scaling ι (upper figures) through three constraints of field conditions: **a.** Scaling of prey-predator size-ratio p_k , which captures the constraint of availability of suitable (relative) prey size with increasing predator size; **b.** Scaling of reaction distance p_d , which captures the constraint on ability of predators to detect prey of a certain size relative to themselves; **c.** Scaling of prey biomass abundance p_x , which captures the constraint of availability of sufficient prey individuals with increasing predator size. In all plots, blue lines represent the deviation of scaling of intake rate or activity budget from the “normal” scaling (red lines) — when none of these constraints exist. Note that all the intersection points of pairs of scaling (red and blue) lines lie between 1–10 kg predator weight range (highlighted).

141 time with an exponent of ≈ -0.2 (Fig. 1), which changes to a positive scaling exponent if
 142 foraging is subject to one or all of three biomechanical constraints: either size, reaction distance,
 143 or abundance of preferred prey do not keep up with increasing carnivore size. The precise value
 144 of the positive scaling exponent depends upon which of these constraints applies and to what
 145 degree. Underlying this is the scaling of intake rate, predicted to be ≈ 0.9 , which changes to
 146 an exponent between 0.4 – 0.65 depending on the type and strength of the same biomechanical
 147 constraints. The intersection between the two scaling relationships occurs between $\approx 1 - 10$ kg
 148 carnivore size.

149 Thus our model provides a nuanced, mechanistic explanation for the costs of foraging in
 150 small carnivores, and links these costs to the scaling of activity time in small vs. large terrestrial
 151 carnivores. Specification of the general model (equation (1)) for other types of consumers will
 152 likely yield different scaling predictions for intake rate (equation (3)) and activity time scaling
 153 (equation (2)).

154 Next, to test our theoretical predictions, we compiled independent datasets on intake rates
 155 and activity budgets on 73 species of terrestrial carnivores spanning most of the extant size of
 156 the Order (Supplementary Information). From these data, we first calculated the proportion of
 157 time spent active (T_p) in a day across 38 carnivore species. We restricted the activity budget

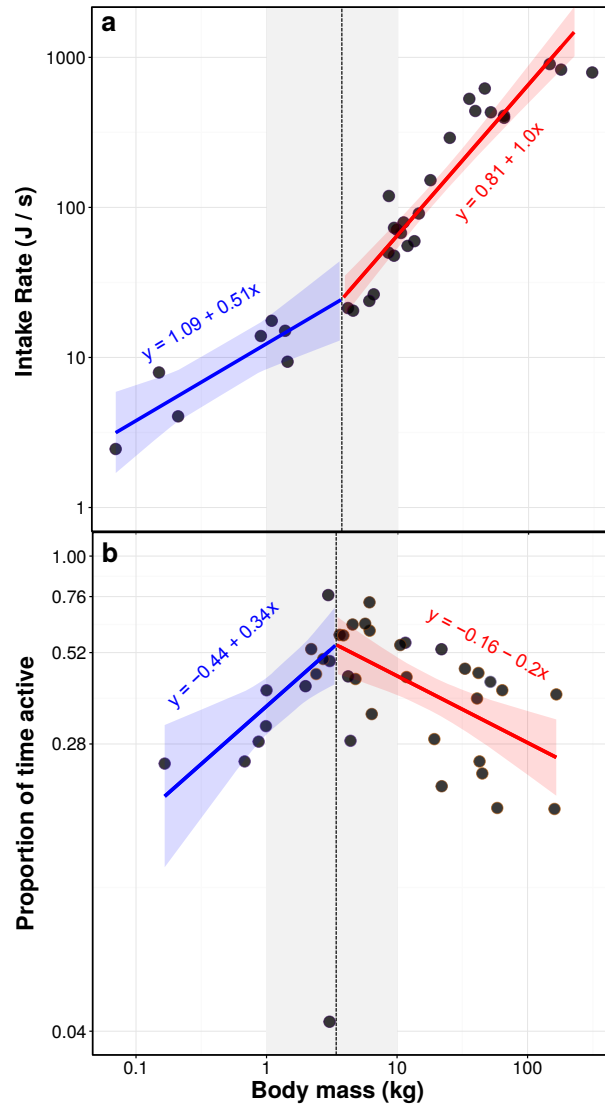


Figure 2: **Observed size-scaling of Energy Intake Rate and Activity Proportions among Carnivores.** In both panels, the blue and red lines represent the two segments of the breakpoint regression (with 95% prediction bounds), the vertical line is their intersection (the breakpoint), and the vertical shaded area the 1–10 kg interval predicted by the theory (Fig. 1). **a:** The scaling of energy intake rate ($n = 32$ species). Breakpoint is at 3.74 kg, and the two scaling exponents (i.e., the slopes) are significantly different (Davies' test, $p = 0.014$). **b:** Activity budget scaling ($n = 38$ species). Breakpoint is at 3.42 kg, and the two scaling exponents are significantly different (Davies' test, $p = 0.0006$). The outlier represents activity data of one Egyptian mongoose (*Herpestes ichneumon*) collected over the course of 1 month in Spain (Supplementary Information). Also see Fig. S5 & S10 for species- and family-level contributions to these results.

158 data compilation to biotelemetry and GPS tracking studies because other methods (e.g., camera-
159 trapping, direct observation) are likely to under sample the proportion of time active in smaller
160 animals due do their poorer temporal and spatial resolution. We find that there is an overall
161 tent-shaped relationship between T_p and carnivore body size across the entire size range (Fig. 2).
162 Larger carnivore species generally become less active (T_p decreases) as body size increases, while
163 smaller carnivores become more active (T_p increases) as body size increases.

Table 1: **Parameter values for the piecewise regression model fitted to the Activity Time, Intake Rate and Size Ratio data vs. Body Mass data.** All variables were log₁₀-transformed. The ΔAICc is the difference in the small-sample size Akaike Information Criterion value for the piecewise ordinary least squares (OLS) vs. a single-line OLS model fitted to the data. The intercept and slope ($\pm 95\%$ Confidence Intervals) pairs are for the fitted OLS models below and above the breakpoint (cf. Fig. 2 & 3). Further details on the model fitting and selection results are in the Supplementary Information.

	Activity proportion	Intake Rate	Size Ratio
ΔAICc	-13.9	-6.54	-10.6
r^2	0.43	0.94	0.55
Breakpoint	0.53 ± 0.14	0.57 ± 0.34	0.68 ± 0.20
Intercepts	-0.44, -0.16	1.09, 0.81	-1.24, -2.74
Slopes	0.34 ± 0.11 , -0.20 ± 0.05	0.51 ± 0.15 , 1.00 ± 0.08	0.21 ± 0.44 , 2.41 ± 0.36

164 This overall pattern is best explained by a piecewise regression (Table 1, Supplementary
165 Table S2). The breakpoint — the body weight where the scaling relationship reverses — is
166 estimated to be at 3.42 kg (95% confidence intervals (CIs): 1.79 – 6.52), within the 1 – 10 kg
167 range predicted by our model (cf. Fig. 1 & 2). This breakpoint is around the average weight
168 of a Gray Fox (*Urocyon cinereoargenteus*). The slopes of the piecewise model’s left and right
169 segments were also significantly different (0.34 ± 0.11 , and -0.2 ± 0.18 respectively, $p < 0.0001$)
170 (Table 1). Thus, the empirically-observed exponent for the activity time for large carnivores is
171 statistically indistinguishable from the value predicted by our model. The value of the exponent
172 for small carnivores (0.34 ± 0.11) is also within the range predicted by the model (Fig. 1).

173 Next, we used the dataset on intake rates (32 species) to test the mechanistic link between
174 the scaling of intake rate and activity time predicted by our theory (Fig. 1; equations (2) & (3)).
175 Figure 2 shows strong support for our prediction: a change in activity time scaling is coupled
176 with a significant weakening of the intake rate scaling in smaller carnivores (Fig. 1). The
177 qualitative change in the intake rate scaling takes place at ~ 3.74 kg (95% CIs: 0.86 – 8.34), —
178 statistically indistinguishable from the breakpoint for the scaling of the activity budget (Fig. 2).
179 The slopes of the piecewise model’s two segments, to the left and to the right of the breakpoint,
180 differed significantly (0.51 ± 0.29 and 1 ± 0.15 respectively, $p = 0.013$) (Table 1). Furthermore,
181 in the upper panel of Fig. 2, the slope to the left of the breakpoint is below the 0.7–0.8 value
182 predicted by our theory (the exponent α ; equation (2)). Thus overall, we find strong support
183 for the predicted mechanistic link between intake rate and activity time.

184 We did not detect a significant phylogenetic signal in the activity budget or intake rate
185 datasets (Supplementary Information). The results also remain qualitatively unchanged after
186 fitting a linear mixed effect model to the data with study and species identity as random
187 effects (Supplementary Information), nor does accounting for seasonal resource availability or
188 reanalyzing the data for only the three most data-rich Families (Supplementary Information).

189 Thus our results indicate strong constraints on intake (foraging) rates, and therefore on
190 activity times in small terrestrial carnivores (below a body size of ~ 5 kg). These constraints
191 arise in small carnivores because some or all of the following: prey biomass abundance increases
192 too weakly with increasing predator body mass, prey body sizes increase too weakly (that is,

193 larger species among small carnivores feed on prey that are sub-optimally small), or reaction
194 distance does not increase or increases too weakly (Fig. 1).

195 To gain further insights into which of these constraints might dominate the observed patterns
196 in activity budget scaling, we compiled a third independent dataset on prey-predator body sizes
197 relationships for terrestrial carnivores (Supplementary Information). We tested whether smaller
198 carnivores fed on qualitatively different relative prey sizes than larger carnivores. The results
199 (Fig. 3) show that there is indeed a significant shift in the relative size of preferred prey between
200 small and large carnivores — prey size scales strongly and positively with body size in large
201 carnivores, but not in small carnivores. A breakpoint regression indicates that the shift lies
202 within the 1–10 kg size range, at about 4.8 kg (95% CIs: 2.34 – 7.25) a value that is somewhat
203 larger but statistically indistinguishable from the breakpoints for the activity budget and intake
204 rate scaling relationships (Fig. 2). Thus, the fact that smaller terrestrial carnivores tend to feed
205 on prey items of relatively constant size that are much smaller than themselves (Fig. 3) at least
206 partly explains the reversal of scaling of the activity budget.

207 Discussion

208 We have developed a mechanistic model to predict the body size-scaling of the time consumers
209 need to be active (i.e., foraging) for maintaining energetic balance. By specifying this model
210 for terrestrial carnivores and by analyzing a global empirical dataset, we find that somewhat
211 counter-intuitively, small and large terrestrial carnivores have an opposite scaling of the pro-
212 portion of time spent active (T_p) in a day. Specifically, small carnivores below a 1–10kg size
213 range show an increase in daily activity time with body size, with only larger carnivores showing
214 the decrease in activity time with size expected from their lower mass-specific energy needs⁵.
215 This reversal of scaling occurs because small carnivores face additional constraints while for-
216 aging, which limits their intake rate and negates the advantage of decreasing mass-specific
217 metabolic rates with increasing size. These results also provide an explicit, (bio)mechanistic,
218 and empirically-validated theoretical model for the cost of small-prey eating hypothesized by
219 Carbone et al^{19,16}, who predicted a similar upper threshold size (14.5 kg) for small prey eating.

220 The three constraints — prey-predator size ratio, reaction distance, and resource abundance
221 — are general in that all of them are likely to be experienced by predators under field conditions.
222 That is, unlike in larger carnivores, where bigger species feed on proportionally bigger prey, in
223 small carnivores preferred prey size changes little as body size increases, as can be seen in Fig. 3.
224 Thus, in small carnivores, though foraging on small prey is initially relatively easy due to low
225 hunting costs, as predator size increases, prey become increasingly difficult to detect, attack
226 and handle.

227 Size-ratio and reaction distance are tightly interlinked, because smaller size-ratios (prey
228 much smaller than predator) also decrease reaction distance^{12,13}. Current models and data
229 suggest that reaction distances scale positively with body size when considering visual con-
230 straints, so that larger species have bigger reaction distances^{23,12,13}. However, field conditions
231 impose multiple constraints on how far an animal can see, including vantage point, line of sight,
232 prey conspicuousness and maneuverability, all of which likely raise additional challenges for
233 small predators hunting for much smaller prey. Indeed, it has been suggested that natural habi-
234 tats show fractal (i.e., self-similar) visual structure²⁴ which implies that, compared to large-prey
235 eaters, small-prey eaters hunt for better-hidden prey in effectively more complex landscapes.

236 Thus, although data are currently lacking on the scaling of reaction distance in terrestrial
237 carnivores, the weak scaling of both intake rates and prey sizes seen in small carnivores indicates
238 that species up to about 10kg face somewhat insurmountable challenges in the way of feeding
239 on optimally-sized prey. This is likely compounded by the increasing costs of rapid maneuver-
240 ing necessary for small-prey hunting^{21,22}, and greater prevalence of nocturnal foraging (which
241 presumably also limits prey detectability) in small carnivores. Carnivore species above the 1–

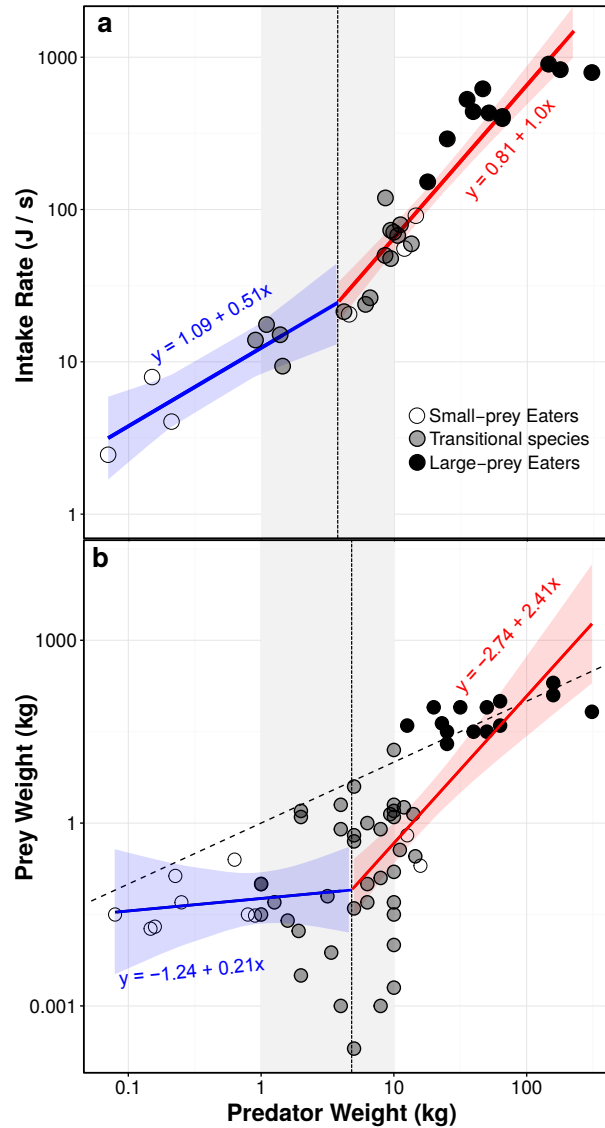


Figure 3: **Scaling of intake rate and prey vs. predator size (size-ratio) among carnivores.** In both panels, the blue and red lines represent the two segments of the breakpoint regression (with 95% prediction bounds), the vertical line is their intersection (the breakpoint), and the vertical shaded area the 1–10kg interval predicted by the theory (Fig. 1). **a:** Same as Fig. 2a, but with the data classified by relative prey-size based feeding strategies instead of taxonomy. **b:** scaling of prey with predator size ($n = 63$ species). The breakpoint estimated here is 4.8 kg: above this, size of preferred prey scales positively with predator size, whereas below this, prey size remains roughly constant even though consumer body mass increases. The two slopes are significantly different (Davies' test, $p = 0.0025$). The diagonal dashed line represents the predator-prey sizes where the size-ratio equals 1; values below it indicate prey are smaller than the predator. The classes of prey-size based feeding strategies shown were defined on the basis of the range of prey types taken, as explained in Supplementary Information Section 2.4. Note that in both plots, the transitional species lie largely within the 1–10kg (shaded) range.

242 10kg size range escape these constraints by feeding on relatively larger prey^{19,20} (Fig. 3), which
243 results in a switch to a steeper scaling of intake rate and correspondingly, a negative scaling of
244 activity time (Fig. 2).

245 Handling time, which includes the time spent in pursuing and capturing prey¹⁵ following
246 detection and reaction, would be subject to similar constraints. Handling rates are unimodal
247 with respect to prey-predator size ratios, that is, they decline at both extremes of prey-predator
248 body size ratios^{16,25,20}. Therefore, the main advantage of feeding on small prey — a relatively
249 short handling time^{16,20} — is increasingly negated for larger small-prey eaters as their prey
250 become sub-optimally smaller relative to themselves. This would compound with the constraints
251 of reaction distance. Future work should aim to directly test this hypothesized weaker scaling
252 of reaction distance among small carnivores, as well as the variation in handling time scaling
253 across different size-classes of carnivores.

254 A weak scaling of prey abundance can accentuate the constraints faced by small carnivores
255 (Fig. 1). Resource biomass abundance, which scales positively with resource body mass and
256 therefore also predator body mass, (the exponent $1 - p_x$ in equation (7), Supplementary In-
257 formation equation S13), can partly offset the higher cost of foraging for small prey in small
258 carnivores. If the numerical abundance of the prey (exponent p_x) of small-prey eaters itself
259 scaled more strongly, such that larger prey items were rarer, small-prey eating would become
260 more costly with increasing predator size, leading to an even stronger positive scaling of the
261 proportion of time active with size. Whether, in general, this is true within the size range of prey
262 species relevant to small carnivores is currently unknown. Therefore, future work should also
263 aim to quantify the scaling of abundance of target prey for different size-classes of carnivores.

264 The upshot of these constraints on foraging is that in order to maintain energetic balance,
265 small-prey eaters have to start preferentially taking larger prey beyond a certain body size
266 range¹⁶, or evolve morphological and behavioral specializations to feed on small prey (e.g.,
267 the Aardwolf; *Proteles cristata*²⁶). In this regard, it is notable that we do find an abrupt
268 diversification of prey sizes taken (including the Aardwolf example) within the 1–10kg size
269 range (the shaded area in Fig. 3, consisting mostly of “transitional” species). This may explain
270 why small carnivores (Viverrids, as well as many Canids and Felids) in the 1–10 kg body weight
271 range increase the diversity of size range of their prey — becoming more generalized to offset
272 the increased costs of foraging. A signature of these foraging constraints on small carnivore
273 foraging may also be seen in long-term home range size and usage. Understanding animal home
274 range sizes and usage is important for design of protected areas^{1,4}, and is a promising avenue
275 for future work based on the findings of this study. Also, because the same biomechanical
276 constraints highlighted here for small-prey eaters could apply to large-prey eating terrestrial
277 carnivores if sufficiently large prey are unavailable (or go extinct), our model may also provide
278 a mechanistic explanation for body size limits to large carnivores, and why gigantic forms
279 in many extant carnivore families have appeared and become extinct time and again in the
280 paleontological record^{16,27}.

281 Our results thus shed light on the behavioral adaptations involved in offsetting the higher
282 energetic requirements of increasing size, and reveal ecological challenges faced by small carni-
283 vores. Small carnivores may be particularly susceptible to habitat degradation if this leads to
284 an increase in foraging activity and therefore adds to an already sub-optimal activity budget.
285 As human-induced environmental changes become ever more common and severe, these species
286 may be among the first and more seriously affected — calling for further studies to inform
287 adequate conservation policies. From this perspective, the modelling framework we propose
288 here can be used to develop a better understanding or *a priori* predictions for daily activity
289 times within individual species across their geographical range. In particular, by appropriately
290 parameterizing the resource (prey) abundance scaling constant x_0 , the model can be used to
291 predict how different populations of the same species respond to spatial or temporal variability
292 in resource availability and quality over its geographical range. For example, the percentage

293 increase in daily activity time due to a decline in resource abundance over time or space for a
294 species can be predicted using the model, and then tested using field data. Similarly, by appro-
295 priately parameterizing the size-ratio scaling constant k_0 and exponent p_k , the model can be
296 used to predict the effect of variation in availability of appropriate or preferred prey on activity
297 budgets of a species across its range. This would provide key insights into threats to species'
298 energy budgets and therefore, ultimately population sizes and sizes of protected areas necessary
299 for maintaining a “healthy” activity budget in a target species.

300 Also, to develop such accurate, species-specific predictions, more realistic parameterizations
301 of the cost of locomotion will be needed. For example, the equation for scaling of the cost of
302 locomotion we use¹⁸ (equation (6)) likely underestimates the actual metabolic costs experienced
303 by carnivores in the field. Therefore, we would expect a higher intercept of the scaling of activity
304 time than predicted here (Fig. 1). The increasingly cheaper techniques available for field
305 measures of carnivore energetics hold great promise for more accurate predictions for specific
306 species or groups of species (e.g., mustelids vs. canids and felids; see SI section 2.5.6) using this
307 modelling framework.

308 Another source of variation in daily activity time is likely to be the seasonal changes in energy
309 requirements for breeding (e.g., searching for mates and defending territories) and overwintering
310 (e.g., storing fat for hibernation) in many species. Therefore, though our result about the
311 dual scaling of activity time remains qualitatively robust across resource-rich and resource-poor
312 seasons (SI section 2.5.6 & Fig. S9), elaborating our model to include seasonal energy loss
313 terms will allow a more accurate predictions of activity time. This will have to be coupled
314 with tracking datasets at sufficient resolution to allow a proper investigation of the effects of
315 seasonal bursts or declines in daily energy expenditure on activity patterns. For this, the general
316 bias towards relatively larger carnivore species evident in the published literature on activity
317 patterns (and therefore also in our compiled data; see Fig. 2) needs to be addressed first. The
318 results of this study emphasizes the need for some correction of focus of tracking studies from
319 bigger, charismatic carnivores to smaller, more elusive species.

320 From a more theoretical perspective, our model framework could be adapted and extended
321 to explore the role of biomechanical constraints in the field on activity budgets for a wide
322 range of organisms, opening up research avenues for understanding links between behavioral
323 and population processes. Specifically, using appropriate scaling models for reaction distance
324 and resource abundance, the model can be extended to herbivores, invertebrate predators, or
325 consumers that forage in three spatial dimensions ($3D$; such as in pelagic environments^{13,14,17}.
326 Extending the model to ectothermic consumers (e.g., all invertebrate predators) would require
327 appropriate models for temperature dependence of metabolic rates and body velocity^{28,15}. For
328 example, to generate predictions for aquatic predators, the equations and parameter values
329 for the energy costs of inertial aquatic locomotion (equation (6)) and basal metabolic rate
330 (equation (5)) for aquatic organisms could be used, coupled with a change from a $2D$ to $3D$
331 intake rate model¹³. Although herbivores may be less constrained by reaction distance, they
332 may still be subject to size-ratio or resource abundance scaling constraints on intake rate²⁹.
333 As recent studies on other classes of mammals both substantiate³⁰ and contradict³¹ the dual
334 scaling relationship of activity budget with body size found in the present work, our results
335 prompt further, in-depth investigation of the effects these constraints have on non-carnivorous
336 mammals.

337 In conclusion, the proportion of time that animals need to dedicate to foraging depends
338 upon the biomechanical constraints they face in the field. We find strong evidence that small
339 terrestrial carnivores face such constraints, likely arising from a combination of sub-optimal
340 prey-predator size ratios, weak scaling of reaction (effective detection) distance and possibly,
341 prey abundance. These constraints change the energetic advantage of increasing body size to a
342 disadvantage, and the scaling of proportion of activity time reverses in small vs. large carnivores.
343 By quantifying the mechanistic links between field conditions and activity budgets, our model

344 offers insights into the constraints on animal fitness in the field, and what foraging strategies
 345 may be preferred in different biotic and abiotic contexts. Further work along these lines would
 346 provide field scientists, conservationists, and theoreticians with a powerful tool with which to
 347 explore how species adjust to both environmental and physiological changes, expanding on our
 348 knowledge of the ecology, evolution, and conservation of as of yet poorly understood consumer
 349 species and consumer-resource interactions.

350 Methods

351 **Model development.** We start by deriving a model for the minimum proportion of time (T_p)
 352 that an individual consumer must spend foraging on a daily basis to maintain its energy balance
 353 (see Supplementary Information for derivation):

$$354 \quad T_p = \frac{B}{I + B - A} \quad (4)$$

355 where I as the individual's energy intake rate, B its energy loss rate when resting and A its en-
 356 ergy loss rate when active. All rates are in $J \cdot s^{-1}$. We then impose biomechanical and metabolic
 357 constraints on the three components of equation (4) using metabolic scaling theory^{5,8,13,15} ap-
 358 propriate for terrestrial carnivores (Supplementary Information for further details). Specifically,
 359 for energy loss rate while resting, we use the scaling of basal metabolic rate,

$$360 \quad B = B_0 m^\beta \quad (5)$$

361 Where m is the body mass (in kg) of the predator. For energy loss during movement, we use
 362 Taylor et al.'s model¹⁸ (see Supplementary Information for more details):

$$363 \quad A = A_{0,1} v_0 m^{a_1 + p_v} + A_{0,2} m^{a_2} \quad (6)$$

364 Where v_0 , $A_{0,1}$, and $A_{0,2}$ are constants (Supplementary Table S1). The first term of the sum
 365 ($A_{0,1} v_0 m^{a_1 + p_v}$) quantifies the increase in energy consumption during movement as a function of
 366 body size (exponent a_1) and body velocity (exponent p_v) (incremental cost), while the second
 367 term ($A_{0,2} m^{a_2}$) quantifies the energy needed to initiate the movement (zero speed cost). For
 368 energy intake rate we use the consumption-rate model derived for individuals foraging in $2D$
 369 (two euclidean dimensions)^{13,15},

$$370 \quad I = I_0 m^\iota \quad (7)$$

371 where

$$372 \quad I_0 = 2v_0 d_0 x_0 k_0^{1-p_x+p_d} \quad (8)$$

373 and

$$374 \quad \iota = p_v + p_x + (p_k + 1)p_d \quad (9)$$

375 Here, p_v , p_d and $1 - p_x$ are the scaling exponents of the predator's velocity (same as in equa-
 376 tion (6)), reaction distance and prey biomass abundance respectively. Reaction distance is the
 377 minimum distance at which the consumer can detect the prey and react to it¹³. The exponent
 378 p_k is for the scaling of prey-predator body size-ratio ($\frac{m_R}{m}$). I_0 is the product of the scaling con-
 379 stants of velocity (v_0), reaction distance (d_0), size-ratio ($k_0^{p_d}$), and prey biomass abundance (x_0).
 380 Substituting equations eqs. (5) to (7) into equation (4) followed by some simplifications and
 381 approximations (detailed in Supplementary Information) gives the biomechanically-constrained
 382 activity budget model (equation (1)). Analysis of this model to determine the inflection point,
 383 which satisfies the condition $\frac{d \log(T_p)}{dm} = 0$, yields the result shown in equation (2).

384 **Model parameterization.** We provide a detailed account of the model parameterization
 385 in the Supplementary Information, along with values of all scaling parameters. In short, we used

386 published scaling relationships for all variables in equation (1). We parameterized equation (5)
387 by reanalyzing carnivores’ data from Kolokotronis et al.⁸, after dropping aquatic and omnivore
388 species. For equation (6), we used the values in the original paper by Taylor et al.¹⁸ re-expressed
389 in $\text{J} \cdot \text{s}^{-1}$ from the original $\text{J} \cdot (\text{kg} \cdot \text{s})^{-1}$. The scaling equation for intake rate I involves three
390 different relationships (equation (7)): we used the reaction distance equation from¹³, the size-
391 ratio scaling relationship from⁵ and the prey biomass abundance models of^{16,13}. For the velocity
392 term v , which appears in both the scaling of A and I , we used the relationship by⁴³ as cited
393 in⁵. We used an energy content value of $1\text{kg wet mass} = 7 \cdot 10^6 \text{ J}$ in all conversions⁵.

394 **Sensitivity analyses.** We tested our model for sensitivity to both variation in its mathe-
395 matical structure and in the values of the parameters used. Results from the structural sensi-
396 tivity analysis are shown in Supplementary Fig. S1 & S2, and indicate that our carnivore model
397 specification is robust to simplification of its mathematical structure. We also used equation (1)
398 to determine the contribution of each of the three constraints (prey abundance, prey size and
399 prey reaction distance) alone to the size-scaling of intake rate and activity budget (Fig. 1). As
400 can be seen in Fig. 1, each of these constraints can by itself result in a qualitative (negative to
401 positive) shift in the scaling of activity budget, with the shift occurring within a size range of
402 about 1–10 kg predator body weight. Finally, we sampled each of our seven scaling exponents
403 10,000 times independently from a Gaussian distribution to test for robustness to variation in
404 the parameterization of all scaling relationships. Each time, we re-calculated I and T_p . Sup-
405 plementary Fig. S3 shows the results of this analysis, highlighting that our main results are
406 qualitatively robust to uncertainty in the values of our scaling parameterizations.

407 **Data collection.** We collected data on the daily activity budget of terrestrial carnivore
408 species from both published literature and existing databases. We focused on data collected
409 via high-resolution radio-tracking techniques (VHF, GPS and Accelerometers) to build a con-
410 sistent dataset that would allow for direct comparison between different species and because
411 of substantial variability in the accuracy of different techniques (Supplementary Information).
412 We used a set of keywords defined *a priori* and selected only studies reporting full 24 hours
413 activity cycles based on 1 or more complete years of sampling. Using open-source software³⁴, we
414 digitized graphs and tables, and then converted all data collected to SI units of time (s). We did
415 not include marine (e.g., Pinnipeds) and omnivore (e.g., Ursids) species in our dataset. We used
416 an existing dataset for intake rate data¹⁶. Similarly, when not available in the original sources,
417 we used an existing dataset for average body weight³⁵. We obtained size-ratio data (i.e, mass
418 of predators and of their preferred prey) from the published literature^{19,16,36}. We classified diet
419 of carnivores species based on the percentages of different food categories present in their diet,
420 and then classified them as either “large-prey eaters” or “small-prey eaters” (Supplementary
421 Information).

422 **Data analyses.** We conducted all analyses in R (v. 3.3.0³⁷) with significance levels set as
423 $\alpha = 0.05$. Our dataset showed substantial pseudo-replication: we accounted for this by taking
424 the geometric mean of repeated measures, which allowed us to obtain a single average value of
425 activity over 24 hours for every species in our dataset (Supplementary Information). A special
426 case of pseudo-replication is represented by phylogenetic relatedness³⁸. To account for this, we
427 tested both our activity times and intake rate datasets for phylogenetic signal using a recently
428 published tree for carnivores³⁹. Using R package “geiger”⁴⁰, we fitted 3 models to each dataset:
429 a maximum likelihood model, a brownian motion model with $\lambda = 0$ and brownian motion
430 model with $\lambda = 0$. We used an information theory approach to establish the better model
431 and found no evidence of phylogenetic signal in either dataset (Supplementary Information).
432 To quantify the relationship between the activity times and body weight in our dataset, we
433 fitted 3 different regression models to the \log_{10} -transformed variables: an ordinary least squares
434 (OLS), a second degree (quadratic) polynomial and a piecewise regression using R package
435 “segmented”⁴¹ (Supplementary Information). To test for differences in the slopes of the two
436 segments of the piecewise regression, we used the Davies’ test⁴¹. Analyses of the intake rate

437 and size-ratio data, as well as those on the effects of seasonality, followed similar procedures
438 (Supplementary Information). We repeated these analyses on the un-transformed data, using a
439 linear mixed model with Study and Species as random effect fitted using R package “nlme”⁴².

440 **Code and Data Availability.** The computer code and data used in the present analy-
441 ses are available from Figshare public repositories (identifiers 10.6084/m9.figshare.5466295 and
442 10.6084/m9.figshare.5464150 respectively).

443 **Competing Interests Statement.** The authors declare no competing interests.

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