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1 **The lethal 23%: predator demography influences predation risk for threatened prey**

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11 **Running head:** Predator demographics

12

13 **Article Impact Statement**

14 Predator demographics are an important determinant of predation impact and certain

15 predator demographics should be targeted for protection of threatened prey

16 **Key Words**

17 Threatened species, predation, Australia, cat, body size, demography, diet, prey

18

19 **Abstract**

20 Globally, wildlife managers often control predator populations to protect biodiversity,  
21 livestock or other valued resources. Most assume that the predation impact of each  
22 individual predator is the same and that removing any individual predator produces a  
23 benefit to the target species. However, research suggests predation efficacy can vary within  
24 a predator species according to phenotypic characteristics. Understanding these individual  
25 differences may be critical for managing predation impacts on particular categories of prey  
26 including small populations where predation effects are amplified. We used dietary data  
27 from 1748 feral cats euthanased over 30 years during a control program in arid Australia to  
28 determine whether any predator attributes could predict their effect on different prey  
29 weight classes.

30 Feral cats in our study ate a wide range of prey including reptiles, mammals, birds and  
31 invertebrates. Demography (body mass) was a highly significant predictor of diet. Cats  
32 weighing 3 kg fed predominantly on prey < 50g increasing to > 500g when cats attained a  
33 body mass of 6 kg. Of more significance was that diet varied between demographic groups  
34 but also within a single demographic group (adult males) based on body mass, with results  
35 having significant implications for threatened prey. Modelling indicated that for a fixed  
36 predator population size, the predation rate on prey in a >500g weight class increased by  
37 28% when the ratio of large (>4.2 kg) to small adult males varied according to its natural  
38 range (12-80% over the 30 year study). Results suggest that variations in predator  
39 demography can significantly impact predation rates on prey species and should be included  
40 in predator-prey models for small prey populations. On average, large male cats comprised  
41 23% of the population and our findings suggest that targeting the “lethal demographic” and  
42 manipulating predator demography should be prioritised along with lowering predator  
43 density to reduce predation impacts.

44

45 **Introduction**

46 Introduced predators have caused the decline and extinction of hundreds of native species  
47 around the world. They have decimated native mammal populations in Australia (Burbidge  
48 and McKenzie 1989; Woinarski et al. 2012) and native birds in New Zealand, Japan, the  
49 United States and on islands (Arcilla et al. 2015, Young et al. 2013, Nogales et al. 2013).  
50 Furthermore, introduced predators have caused the failure of numerous global  
51 reintroduction programs through high levels of predation (Moseby et al. 2011; Clayton et al.  
52 2014; Armstrong et al. 2006). Introduced predators are thought to exert a greater impact on  
53 prey than native predators due to the absence of shared evolutionary history (Saul &  
54 Jeschke 2015).

55

56 Introduced predators also have an impact on agriculture through predation on livestock  
57 (Greentree et al. 2000; McLeod et al. 2010). Due to the threats posed by introduced  
58 predators, their control is a common management action implemented for the protection of  
59 threatened species and livestock around the world (Courchamp et al. 2003; Burrows et al.  
60 2003; Whitehead *et al.* 2010). The cost of these control programs can be high, for example  
61 Australia spends approximately 16 million dollars a year on the control of foxes for the  
62 protection of wildlife and domestic stock (McLeod 2004). Reducing the abundance of  
63 predators is the goal of most programs based on the assumption that predator abundance is  
64 the most important determinant of predation risk. This assumption is underpinned by the  
65 conceptual model of predator-prey dynamics stemming from the Lotka-Volterra model  
66 (Volterra 1926; Lotka 1932) that explains changes in prey abundance based on predator

67 abundance. A well-known example is the lynx-hare relationship where lynx populations  
68 increase and decrease in synchrony but slightly lagging behind fluctuations in abundance of  
69 snow-shoe hares (Keith *et al.* 1984). Hollings (1966) expanded predator-prey theory by  
70 adding a functional response, arguing that the rate at which predators consume prey is a  
71 function of their attack rate and processing time, which varies at different prey densities.  
72 However, predator-prey models are based on assumptions, e.g. that the predator is  
73 dependent on a single prey species, that the predator has a limitless appetite, and that a  
74 fixed proportion of encounters leads to the death of the prey. These assumptions are rarely  
75 met in the wild, impeding application of these models in conservation (Abrams 2007).  
76  
77 Importantly, most predator management programs and models of predator-prey dynamics  
78 assume that each predator in the population has equivalent hunting efficacy (e.g. the  
79 constant in Lotka's predator consumption rate or Hollings' predator attack rate). In the case  
80 of vertebrates, this assumption is often violated because differences in the genetic make-  
81 up, demography and life-time experience of predators influences individual's hunting  
82 behaviour (McGregor *et al.* 2014; Dickman & Newsome 2015; Moseby *et al.* 2015; Pettorelli  
83 *et al.* 2015). For example, Austin *et al.* (2004) found significant intraspecific differences in  
84 foraging behaviour in seals with likely resultant effects on predation models. In feral cats,  
85 some individuals can learn to hunt vulnerable prey, leading to accelerated killings and  
86 sometimes local extinction (Clout and Craig 1995; Moseby *et al.* 2015; Hardman *et al.* 2016).  
87 Within small populations, such as threatened or reintroduced wildlife, predation effects are  
88 amplified (Saul & Jeschke 2015) and intraspecific differences in predator attack rate or  
89 preferred prey have the potential to drive catastrophic population declines. Thus, assuming

90 that all predators exert similar impacts on prey has the potential to incorrectly inform  
91 management approaches (Pettorelli *et al.* 2015).

92

93 Some researchers have incorporated changes in predation rate into their models when  
94 assessing predator impacts, for example, Nilson *et al.* (2005) incorporated the variation in  
95 predation risk of young vs adult moose in predator-prey models of wolf impacts. However,  
96 those studies that do focus on the variation in prey selection of predators tend to focus on  
97 population changes such as seasonal (Davidson *et al.* 2013) or temporal changes in predator  
98 diet or changes due to habitat structure (Hebblewhite *et al.* 2005), pack size (Packer *et al.*  
99 1990, Loveridge *et al.* 2006) or prey breeding season (Davidson *et al.* 2013). Unfortunately,  
100 the influence of individual predator traits on predation impacts has received less focus (but  
101 see Funston & Mille 2006), despite studies reporting intraspecific differences in predation  
102 impacts related to predator sex (Marlow *et al.* 2015), age (Litvaitis *et al.*, 1986), experience  
103 (Estes *et al.* 2003) and body size (Moseby *et al.* 2015; Kutt 2012). Feline predators are  
104 particularly flexible, exploitative and opportunistic, Cheetahs (*Acinonyx jubatus*) vary in prey  
105 selection based on sex (Cooper *et al.* 2007), Lynx (*Lynx lynx*) on reproductive status (Pierce  
106 *et al.* 2000), Cougars (*Puma concolor*) on age (Ross *et al.* 1997) and domestic cats (*Felis*  
107 *catus*) on sex and body size (Kutt 2011; Moseby *et al.* 2015). Individual felines can also  
108 specialise on certain prey based on individual experience and learn to hunt difficult prey  
109 (Knopff & Boyce 2007; Dickman & Newsome 2015; Moseby *et al.* 2015).

110

111 Feral cats are an introduced species in Australia and predation by feral cats is listed as a Key  
112 Threatening Process for more than 100 fauna species under the Commonwealth  
113 Environment Protection and Biodiversity Conservation Act (1999). Most of these threatened

114 prey weigh between 35g and several kilograms. We used data from 1748 feral cats  
115 euthanased over 30 years during a predator control program in arid Australia to determine  
116 whether any predator attributes could predict their predation rate on fauna. We first  
117 determined whether the composition and diversity of feral cat diet varied amongst  
118 demographic groups and how this related to prey abundance and environmental variables.  
119 Secondly, we modelled the effects of different demographic ratios on the predation rates on  
120 populations of mammals >500g. This prey size is typical of many mammal species known to  
121 be threatened by feral cats and which have been subjected to high rates of decline and  
122 extinction in the study area (Moseby *et al.* 2011; Woinarski *et al.* 2012). We compared the  
123 potential impacts of various demographic combinations and used the results to determine if  
124 predation impacts can vary significantly even when predator and prey density remain  
125 constant.

126

## 127 **Methods**

### 128 *Study area*

129 This study was conducted in the vicinity of the Arid Recovery Reserve in the Roxby Downs  
130 region of northern South Australia (30 29'S, 136 53'E). The area is comprised of arid  
131 shrubland supporting longitudinal sand dunes overlying clay interdunal swales. The climate  
132 is arid with erratic annual rainfall averaging 166 mm ([www.BOM.gov.au](http://www.BOM.gov.au)). The area supports  
133 multiple land uses including cattle grazing, mining and conservation. Feral cats are  
134 widespread throughout the region (Read and Bowen 2001) and live wild and independent of  
135 human contact (Moseby *et al.* 2009a).

### 136 *Cat diet samples*

137 Between January 1990 and January 2020, feral cats were removed from a 960,000 ha area in  
138 the Roxby Downs region through shooting, cage trapping and padded leghold trapping. Cats  
139 were removed principally as part of a local control program to reduce the pressure around  
140 the Arid Recovery Reserve (Moseby *et al.* 2011), but also as part of feral control on pastoral  
141 properties and around a mine.

142

143 For this analysis, we focused on feral cats, defined as those collected at least 3km from  
144 refuse dumps, residential or industrial areas. Cats were weighed, sexed and then dissected  
145 to examine stomach contents and to determine reproductive status of females. The lightest  
146 pregnant female cat was 2.3 kg, which was adopted as the minimum weight of both sexes at  
147 sexual maturity. This was slightly lower than the defined adult body mass of 2.5 kg recorded  
148 in other studies (Jones & Coman 1982; Brothers *et. al.* 1985). The coat colour of cats was  
149 recorded as either tabby, ginger, or black. Body condition was not included due to  
150 inconsistencies with field scoring methods. Stomach contents were examined for prey items  
151 and where possible each item was identified to species level. Invertebrate species were  
152 identified to Order. A comprehensive 20 year inventory of prey species from the local area  
153 was used as a guide (Read 1998; Read *et al.* 2000; Read & Cunningham 2010; Moseby *et al.*  
154 2009b).

155

### 156 *Cat and Rabbit density*

157 The density of feral cats and one of their key prey items, the European rabbit (*Orytolagus*  
158 *cuniculus*) was estimated using two 20km spotlight transects conducted in the study area  
159 every two months from April 1989 until 2013 and then less frequently for the duration of  
160 the study (see Read and Bowen 2001 for methods). Results were averaged across quarters



161 and if no transects were conducted within a quarter, then the previous quarter's estimate  
162 was used.

### 163 *Variables influencing Diet*

164 We divided prey into six prey groups; invertebrates, birds, herpetofauna (reptiles and frogs),  
165 rabbits, small mammals (<20g) and rodents 20 – 60g (comprising the spinifex hopping  
166 mouse (*Notomys alexis*) and nationally threatened plains mouse (*Pseudomys australis*)).  
167 Carrion was not considered as it was recorded in <2% of samples. We used a count of each  
168 prey group per stomach, with the exception of rabbits and birds which were considered as  
169 binary variables (present/absent) as they were almost exclusively recorded as single prey  
170 items.

171 First, we explored data for simple relationships between cat weight and prey size. The  
172 average live weight of each prey species was derived from trapping and shooting studies at  
173 the study site (Read 1998; Read 1999; Read & Cunningham 2010; Moseby *et al.* 2009b). We  
174 then averaged the live weight of all the prey items found within each cat stomach, then ran  
175 a Generalised Linear Model (R, v3.5.1, [www.r-project.org](http://www.r-project.org)) with a negative binomial  
176 distribution against cat weight, and compared this to a null model using an information  
177 theory framework ie we considered a model with an Akaike weight > 0.5 and delta AICc > 4  
178 having relatively better support compared to the null models (Burnham and Anderson  
179 1998). r

180 Next, for each prey group, we compared their abundance in each cat's stomach with a range  
181 of individual cat characteristics and extrinsic variables. For prey categories using count data  
182 we used two-stage hurdle generalised linear models (Potts and Elith 2006; Zuur *et al.* 2009),  
183 as data were over-dispersed and contained many zeros. These two-stage models ran a

184 binary logit model in tandem with a truncated negative binomial model fitted only to counts  
185 greater than 0. The former models the probability of eating a prey species, the latter models  
186 the probability of how many are eaten. For rabbits and birds (binary variables), we applied  
187 generalised linear models with a binomial distribution.

188 The explanatory variables used in each prey group model of cat diet included cat weight  
189 (both a linear and parabolic relationship peaking at 3 kg used in separate models), sex,  
190 pelage (ginger, tabby, black), rainfall recorded in the last 12 months, either maximum  
191 nightly temperature (for birds and herpetofauna only, as this would affect activity) or  
192 maximum temperature averaged over each month (as more representative of climatic  
193 conditions), and method of sampling (shot or trapped). Rainfall and temperature data were  
194 sourced from the Bureau of Meteorology ([www.BOM.gov.au](http://www.BOM.gov.au)) from the closest registered  
195 recording location (Olympic Dam Aerodrome, within 80 km of each sample). Rabbit density  
196 was also included in the models and abundance of medium sized rodents (20-60g) was  
197 included as a categorical variable on a scale of 0-5 (0=absent, 5=highest abundance). This  
198 index was used instead of trapping rates as sampling was not consistent between habitats  
199 and time periods. Capture rates of rodents in the weight range 20-60g fluctuated from 0 to  
200 10%, and were allocated scores ranging from 1-5 (Read 1994; Moseby & Read 2001; Moseby  
201 *et al.* 2009; Arid Recovery unpublished data).

202 As prey consumption of certain groups could be affected by dietary preference, we also  
203 added a variable 'preferred food' into the model. This variable quantified the number of  
204 prey items present in each cat's stomach that were from a more preferred prey group. We  
205 ranked prey groups based on preferences found in existing literature (Paltridge *et al.* 1997,  
206 2002; Pavey *et al.* 2008; Spencer *et al.* 2014; Read *et al.* 2019; Kutt 2012), then for each prey

207 group we added the number of all of the more preferred prey items in each cat stomach as  
208 a variable. We ranked preferred prey highest to lowest; rodents 20 – 60g; rabbits, small  
209 mammals (<20g), birds, herpetofauna, then invertebrates.

210

211 For the prey models using bird and rabbit prey groups, we included an interaction term  
212 between cat weight and average monthly temperature, as birds and rabbits are most likely  
213 to have young from Winter to Spring (Read *et al.* 2000; Bowen & Read 1998) which could be  
214 targeted by smaller cats.

215 For each of the six prey groups, we ran independent analyses to determine the best model  
216 for predicting their presence or count in cat stomachs using information theory, based on  
217 Akaike Information Criterion (AIC) weights. We used AIC instead of AICc, as our sample size  
218 divided by candidate variables was always greater than 200 (Burham & Anderson 1998). As  
219 we considered all variables possible important predictors of cat diet, we ran 12 models with  
220 different combinations of biologically plausible variables, where each variable was present  
221 within 4-6 models. A global and null model was also added to each model set. The sampling  
222 method (trapped vs shot) was included in all models bar the null, as trapped cats were less  
223 likely to have eaten in the previous 12 hours than shot cats. We then looked at model  
224 averaged coefficients, and added another model containing only variables with > 73%  
225 relative importance, suggesting an AIC delta difference of 2 or less (Richards 2005). For each  
226 analysis, we considered the model with the most support to be the model with a delta of 0 if  
227 >2 difference.

228 Once we had selected the models with the most support, we investigated model fit and  
229 predictive power using quartile-residual plots for all models, rootograms for zero-inflated

230 hurdle models (Kleiber & Zeileis 2016) and receiver operator curves for the Area Under  
231 Curves (AUC) for binomial models.

### 232 *The influence of demography on kill rates of different prey groups*

233 We tested whether the proportion of large males in the population would change predation  
234 pressure on the rabbit prey category at a population level, when all other variables were  
235 taken into consideration. The topprey group models were used to simulate the stomach  
236 contents of 100 random cats each day for 365 days given a range of realistic demographic  
237 and environmental variables. We used an input of temperatures based on long term daily  
238 averages obtained from the study area and we simulated an entire year. We used a  
239 demographic ratio of male:female:subadult of 2:2:1 based on the long term average from  
240 shot cats in the study area and yearly rainfall was taken from a normal distribution from the  
241 site. The index of rodents 20-60g was a uniformly drawn random number between 1 and 5.  
242 As some prey group outcomes are influenced by others, we included any of our significant  
243 prey preference hierarchies in the models. We assumed each stomach content would  
244 represent 10 hours of hunting; the time taken for 95% of small and large food items to  
245 transit the stomach of owned cats (Chandler *et al.* 1997).

246 We reran the model simulation 100 times using different ratios of adult males that were  
247 either small (the first and second quartile of adult male's cat body mass in the population,  
248 2.3-4.2 kg) or large (third and fourth quartile 4.3 kg-6 kg). Female weight ratios were not  
249 included due to the small proportion of females weighing over 4.2kg. In every model, we  
250 assumed the same feral cat population size and that rabbit density was moderate and stable  
251 (40 km<sup>-2</sup>, the average from 2009-2019). We then ran a generalised linear model where each

252 datapoint was one of the 100 simulations of the relationship between large:small male  
253 ratio, and predicted annual rabbit kill.

254 To investigate whether the percentage of large adult male cats in a population was related  
255 to density, for each year we summed cat density and portion of large adult males in the  
256 sampled population and ran Generalised Linear Model (R, v3.5.1, [www.r-project.org](http://www.r-project.org))  
257 comparing large males against estimated cat density, then compared this to a null model  
258 using an information theory framework ie we considered a model with an Akaike weight >  
259 0.5 and delta AICc > 4 as having relatively stronger support compared to the null models  
260 (Burnham and Anderson 1998).

## 261 **Results**

262 There were 3025 recorded cats euthanased in the study region between January 1990 and  
263 January 2020. Of those, 1748 were caught away from human habitation and had a full suite  
264 of demographic and stomach data. For these cats, 1288 were shot and 460 were trapped in  
265 leghold (>95%) or cage traps. Adult female cats (average weight=3.33, SE=0.02) were  
266 typically smaller than adult male cats (average weight=4.19, SE=0.03, Fig 1). The  
267 demographic ratio averaged across all years between male:female:subadult was 2.6 : 2.4 : 1,  
268 but ranged from 15 : 13 : 1 to 1.1 : 0.8 : 1. For adult male cats the percentage of large cats in  
269 the population each year varied from 12% to 80% (average= 45%). Demography varied  
270 independently of density, as a model of cat density against the percentage of large adult  
271 male cats had less support than a null model (AICc of model compared to null: loglik 26.7 vs  
272 26.6, AICc -46.5 vs -48.6, delta 2.3 vs 0, weight =0.2 vs 0.8).

## 273 *Diet vs demography*

274 There were 75 different vertebrate species recorded in the diet, and 12 invertebrate  
275 families. When the size of cats was compared with their stomach contents, subadult cats  
276 less than 2.3 kg predominantly ate food items less than 10g in weight such as invertebrates,  
277 small mammals and reptiles, but as cats increased in size, so too did their intake of larger  
278 food sources (Fig. 2). Once cats attained a body mass of more than 4 kg their average weight  
279 of dietary items increased markedly (Fig.2). Larger cats over 5.5 kg predominantly ate prey  
280 greater than 500g, especially rabbits. The average weight of all prey species in a cat's  
281 stomach increased on a log scale with cat weight (Fig 3, AIC of model compared to null:  
282 loglik -3412 vs 3513, delta 0 vs 200, weight =1 vs 0). Dietary diversity within stomachs of  
283 each cat size declined with cat weight (AIC of model compared to null: loglik -2790 vs 5584,  
284 delta 0 vs 9, weight =0.99 vs 0.1), with stomachs from larger cats containing fewer dietary  
285 categories than smaller cats (-0.05/kg,  $z=3.3$ ,  $P<0.001$ ).

286 When models were run comparing stomach contents with individual cat attributes and  
287 environmental variables, methodological and environmental variables were prominent in  
288 nearly all top ranking models (Table 1). For each prey category, the most parsimonious  
289 model was selected with a substantial margin over other candidate models (Akaike weight >  
290 0.95, and delta > 2 of next strongest model, see Suppl Material 1). Most variables within  
291 each model had a high relative importance (>0.95) and were significant, except for certain  
292 interaction terms. All hurdle models had stable rootograms and QQ plots (see Suppl  
293 Material 2), all binomial models had strong predictive power (rabbit AUC = 0.79, birds =  
294 0.62).

295 Cats that were trapped were around 50% less likely to have prey present for all groups than  
296 shot cats. In hotter temperatures, cats ate fewer rabbit and birds, yet more small mammals,

297 herpetofauna and invertebrates (Table 1). We found no difference in diet between cats of  
298 different pelage colouration. Interestingly, for a given weight class, female cats were slightly  
299 more likely to eat rabbits than males. If cats had already eaten more preferred prey, they  
300 were less likely to eat rabbit or small mammals, however, no relationship was found for  
301 birds and herpetofauna, and a positive association with invertebrates (Table 1).

302 Cat weight was of high relative importance in the highest ranking models for all prey groups  
303 except birds. Rabbits became more prevalent in stomachs of cats of larger sizes (Fig. 4), with  
304 an interaction with average monthly maximum temperature (Table 1). Smaller cats were  
305 more likely to eat rabbit during cooler months (Fig. 5) whereas larger cats ate rabbits  
306 regardless of temperature. Invertebrates and herpetofauna were predominantly eaten by  
307 smaller cats, whilst rodents 20-60g displayed an approximately parabolic relationship (Fig.  
308 4).

### 309 **Using models to predict rabbit kill rates under different demographics**

310 There was a strong significant relationship between the ratio of large to small adult male  
311 cats and rabbit consumption compared to the null model (loglik = -895.7 vs -906.7, delta = 0  
312 vs 20, weight = 1 vs 0, coef=34, se=7,  $R^2 = 0.19$  F = 24.3 1 on 98 DF,  $P < 0.001^{***}$ ). The  
313 greater the ratio of large to small male cats in the population the higher the predation  
314 impact would be on rabbits (Fig.6). For a random population of 100 cats, every 10%  
315 increase in the proportion of the male population over 4.2g would result in an additional  
316 340 extra rabbits per year being killed. The actual proportion of large to small adult cats  
317 varied from 13 to 80% over the study period.

318

319

320 **Discussion**

321

322 The significant influence of cat body mass on predation of different prey groups suggests  
323 that individuals within a population can disproportionately affect prey populations. This  
324 result has implications for managing the undesired impacts of predators including high  
325 predation rates of introduced predators on threatened species. The recorded ratios of large  
326 to small adult male cats over our 30 year study ranged from 13% to 80% demonstrating that  
327 demographic ratios vary significantly in the wild even within demographic groups. We found  
328 that for a predator and prey population of a fixed size, the predation rate on prey >500g  
329 increased by 28% when comparing the lowest and highest ratios recorded during our study.  
330 Our results support other studies that have found differences in predation rates based on  
331 body size both within a species (see Pettorelli *et al.* 2015 for review, Kutt 2012) and  
332 between species (Carbone *et al.* 1999; Vezina 1985). However, our results contrast with the  
333 results of a more limited study by Yip *et al.* (2014) who found no relationship between the  
334 types of prey consumed and cat size.

335 Results suggest that targeted control of individual predators within certain demographic  
336 groups may significantly improve management outcomes depending on the size of prey  
337 being protected. When modelling the impacts of predators on threatened or small prey  
338 populations, the proportion of predators in each adult size category may need to be  
339 explicitly included in predator prey models. Unfortunately, most predator prey models use  
340 population means and consider predators to be identical individuals (e.g. Fryxell *et al.* 2007;  
341 Nilsen *et al.* 2005; Post *et al.* 1999; Stenseth *et al.* 1997; Vucetich *et al.* 2005, 2011). Some



342 researchers have included sex or age class in predator prey models to improve their  
343 predictive power e.g. Nilsen *et al.* (2007) but in general, individual traits are rarely  
344 considered (Pettorelli *et al.* 2011). This can have significant implications for developing  
345 appropriate management strategies for threatened prey populations where intraspecific  
346 differences in hunting behaviour are likely to have amplified effects.

347

348 Our study is significant as, whilst previous studies have shown differences in predation rates  
349 between sub-adults and adults within a population, we have shown that even changes  
350 within a single demographic group (adult male cats) are likely to affect predation rates.

351 Although this has implications for the control of introduced predators for the protection of  
352 threatened species, our results have broader implications including for practitioners  
353 protecting native prey or livestock from native or exotic predators. Understanding how  
354 predation rates change with demography can assist managers with developing more  
355 targeted control methods, improving cost effectiveness and efficacy. In our study, prey  
356 weighing over 500g were more likely to be eaten by cats over 4.2 kg in body mass. The  
357 proportion of cats in this size category averaged 23% of the population each year but ranged  
358 from 7% to 45% over the 30 year study period. This lethal 23% is supported by other field  
359 studies that have shown male (Marlow *et al.* 2015), large 4 kg males (Moseby *et al.* 2015) or  
360 individual cats (Hardman *et al.* 2016) can have catastrophic impacts on threatened species  
361 populations. Targeting this 'lethal demographic' would increase the efficacy of predator  
362 control for protection of threatened species that have a body mass of more than 500 g and  
363 may explain why some control programs that record a decline in predator density are  
364 unable to demonstrate a resultant increase in threatened species abundance (Walsh *et al.*  
365 2012).

366

367 Studies have recorded feral cats preying on species as large as 220-280g (Dickman 1996; Yip  
368 et al. 2014), up to their own body weight (Paltridge *et al.* 1997) and consuming mammals  
369 weighing 1.6 kg (Marlow *et al.* 2015), 2 kg (Moseby *et al.* 2015; Hardman *et al.* 2016) and  
370 even 4 kg (Read *et al.* 2019). Our finding suggests that the preferred prey weight of 40g or  
371 less determined by Pearre and Maass (1998) for *Felis catus* globally is too simplistic and that  
372 feral cats may be a significant threat to species up to several kilograms in body mass.

373 The interacting effects of cat size, prey availability, prey preferences, rainfall and season on  
374 cat diet supports other studies that suggests predation impacts are complex and depend on  
375 a range of intrinsic and extrinsic factors (Pettorelli *et al.* 2011). Some studies have found  
376 that larger predator species consume a higher diversity of prey than smaller ones (Radloff  
377 and du Toit 2004; Schoener 1969). This relationship has also been demonstrated within the  
378 same species (Yip *et al.* 2014) which, in sexually dimorphic species, is thought to be a  
379 response to the higher protein requirements of larger males (Nagy 1987). However, we  
380 found large cats to have lower diversity diets. Part of this reduction in diversity may be an  
381 artefact caused by the local extinction of many mammals >100g from the study area  
382 (Moseby *et al.* 2011) limiting the availability of larger prey. However, reptiles weighing up  
383 to 450 g were included in the diet of cats and were present in the study area (Read 1994).  
384 Single prey species are thought to only form the bulk of the diet of individual cats that have  
385 developed specialist hunting methods (Dickman 2009 but see Dickman and Newsome 2015)  
386 or where prey are very abundant and easy to hunt (Fitzgerald and Turner 2000; Yip *et al.*  
387 2014; Denny and Dickman 2010). However, our results suggest that specialisation may occur  
388 when cats attain a body size where they can improve their hunting efficiency by selecting

389 and targeting larger prey, rendering these cats particularly destructive in wildlife protection  
390 programs.

391

392 Sex is another phenotypic characteristic that has been shown to affect individual predation  
393 rates in weasels (*Mustela nivalis*) (Sundell, 2003), Cheetahs (*Acinonyx jubatus*) (Cooper *et al.*  
394 2007) and wolf spiders (*Pardosa vancouveri*) (Hardman & Turnbull 1974). However, sex only  
395 weakly influenced rabbit consumption when body mass was held constant. Breeding female  
396 cats may be targeting young rabbits which emerge most often in Winter and Spring (Bowen  
397 & Read 1998). Juvenile rabbits may be easier for smaller non-breeding female cats and/or  
398 younger inexperienced cats to hunt. This is supported by the highly significant interaction  
399 between weight and temperature for rabbit consumption with smaller cats feeding on  
400 rabbits in the cooler months and larger cats feeding on rabbits regardless of temperature.

401 Although male cats generally attain a larger body mass than females it appears to be size  
402 rather than sex per se that drives differences in prey consumption. Studies that have found  
403 male cats to specialise on threatened prey species and cause significant predation impacts  
404 (Marlow *et al.* 2015) are likely recording a size rather than a sex effect (Moseby *et al.* 2015).

405 However, the influence of body size could not be effectively separated from age. Heavier  
406 cats are likely to be older but the absence of a reliable ageing mechanism for feral cats  
407 makes it difficult to separate out the two demographic effects. Age is an important  
408 determinant of predation rate in other feline species but usually these studies compare  
409 subadult with adult animals (Litvaitis *et al.* 1986; Knopff *et al.* 2010). We considered  
410 subadults separately from adult cats in an attempt to separate age from size effects and,  
411 like other felid studies, found young cats more likely to consume invertebrate and smaller

412 prey items than adult cats. However, whether this result is simply due to smaller body size  
413 or is in fact related to age effects such as lack of experience or shyness is unknown.

414

415 The presence of 20-60g rodents at a site may provide some protection for other mammals.  
416 Small mammal and rabbit presence in cat stomachs were lower if 20-60g rodents were  
417 present, suggesting that when rodent abundance is high, cats will preferentially eat rodents.  
418 This protection may be most pronounced during wet years as higher rainfall in the previous  
419 12 months increased the incidence of 20-60g rodents in the diet probably due to rainfall-  
420 induced resource pulses leading to localised rodent population booms (Brandle & Moseby  
421 1999; Letnic et al. 2005). Alternatively, this protection may be unsustainable: when rodent  
422 numbers crash during droughts, cats may prey switch to other fauna (Courchamp et al 2000;  
423 Read and Bowen 2001).

424 Our findings have significant implications for conservation of small populations of  
425 threatened species where predation is a major threat. Controlling the “lethal demographic”  
426 is likely to be just as important as reducing the density of introduced predators per se.  
427 Predator control methods that remove large cats over 4 kg are likely to yield significantly  
428 greater conservation benefits to prey weighing more than 500g than removing random cats  
429 from the population. When reintroducing or protecting threatened species of this size,  
430 practitioners could target cats that prey on these species through the use of toxic implants  
431 in prey species (Read *et al.* 2015), audio lures that include female mating calls and olfactory  
432 lures that are attractive to male cats. Control methods that are likely to target small cats,  
433 young cats or inexperienced hunters, such as cage traps and control methods that use food  
434 based lures (Short *et al.* 2002), should be used sparingly.

435

436 Our results reveal that accounting for individual variability in predation impacts may help  
437 resolve the discrepancy between patterns predicted in theoretical versus wild populations  
438 and provide more accurate models for predicting predation impacts on threatened prey  
439 populations. We urge global conservation managers and researchers to measure individual  
440 differences in predation rates of threatened prey by other predator species and incorporate  
441 this knowledge into prioritising on ground management actions that target predators with  
442 'lethal demographics'.

443

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445

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452

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755 Table 1. Details of variables within models with the strongest support for each prey group  
 756 predicted within cat stomachs. ‘Type’ refers to type of model. If modelling a count, two-  
 757 stage hurdle models were used with a zero and count, while binary models used a binomial  
 758 ‘GLMuhi’. The variable pelage was also modelled, but not present in any top ranked model.

Prey category		Variable details	AIC relative importance	Model averaged coefficient	SE	Z	P
Rodents 20-60g	Zero	Intercept		-3.46	0.21	16.33	<0.001
		Sample method- trapped	0.95	-0.37	0.15	-2.45	0.014
		Cat weight (parabolic)	1	-0.61	0.11	-5.41	<0.001
		Rodent abundance index (0-5)	1	0.87	0.05	16.05	<0.001
	count	Intercept		-0.73	0.34	-2.14	0.032
		Abundance index (0-5)	1	0.23	0.07	3.42	<0.001
		Sample method- trapped	0.95	-0.45	0.18	-2.53	0.011
Rabbits	binomial	Intercept		1.35	0.92	1.48	0.14
		Sample method- trapped	1	-0.64	0.14	-4.45	<0.001
		Preferred food	1	-0.56	0.12	-4.78	<0.001
		Cat weight (linear)	0.95	-0.06	0.23	-0.26	0.792
		Average month temp. (°C)	1	-0.11	0.03	-3.74	<0.001
		Rabbit density	1	0.45	0.21	2.2	0.028
		Rainfall in last 12 months	0.97	-0.37	0.1	-3.74	<0.001
		Sex- Male	0.95	-0.23	0.13	-1.73	0.084
		Weight (linear) × av. month temp. (°C)	0.97	0.02	0.01	2.57	0.01
Small mammals	Zero	Intercept		-1.68	0.4	-4.16	<0.001

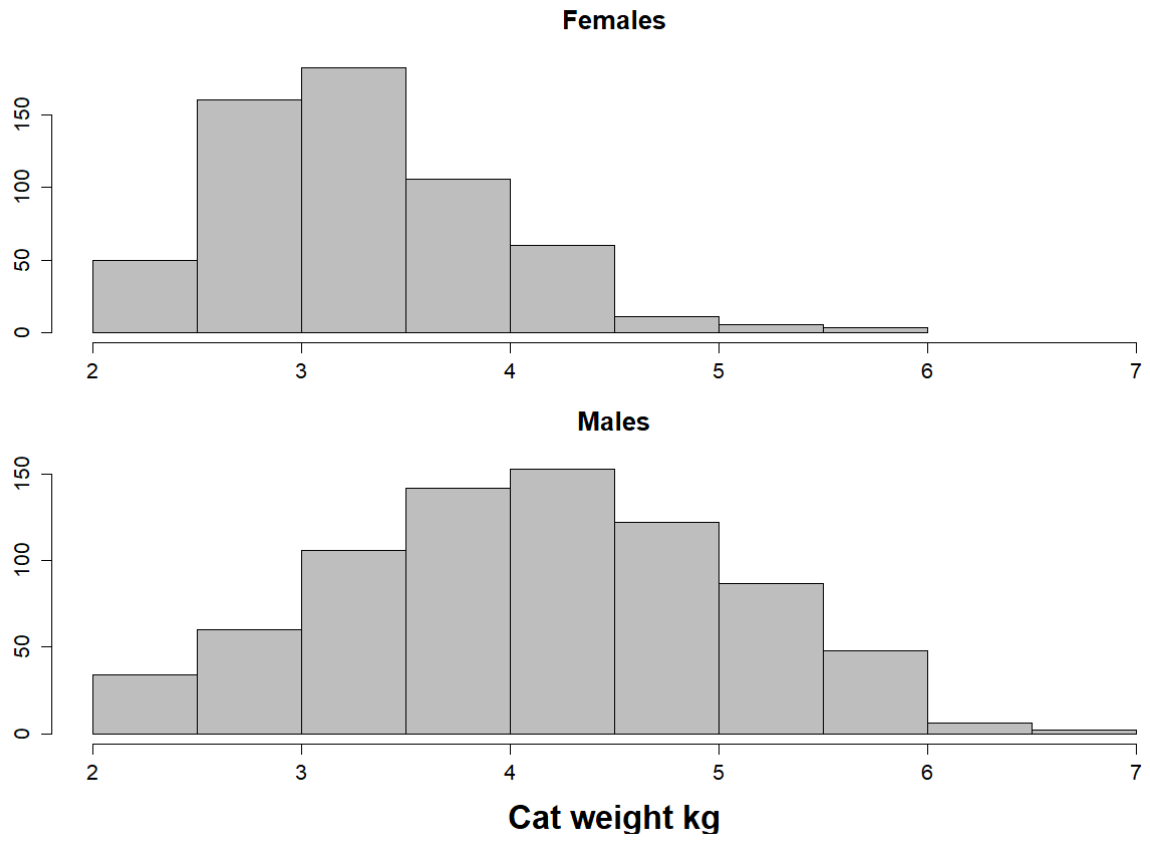
		Sample method- trapped	1	-0.87	0.14	-6.18	<0.001
		Rain in last 12 months	1	0.47	0.09	5.48	<0.001
		Cat weight (linear)	1	-0.15	0.05	-3.23	0.001
		Preferred food	0.97	-0.15	0.06	-2.58	0.01
		Average month temp. (°C)	0.95	0.02	0.01	2.63	0.009
	Count	Intercept	0.95	-2.66	2.44	-1.09	0.277
		Rain in last 12 months	0.95	0.29	0.13	2.23	0.026
Birds	Binomial	Intercept		-0.64	0.37	-1.72	0.085
		Sample method -trapped	1	-0.9	0.24	-3.69	<0.001
		Average monthly temp. (°C)	1	-0.05	0.01	-4.09	<0.001
Herpetofauna	Zero	Intercept		-1.7	0.32	-5.37	<0.001
		Sample method-trapped	1	-1.43	0.17	-8.36	<0.001
		Maximum daily temperature	1	0.04	0.01	5.58	<0.001
		Cat weight (linear)	1	-0.12	0.05	-2.51	0.012
	Count	Intercept		-8.8	52.57	-0.17	0.867
		Sample method- trapped	1	-1.03	0.35	-2.98	0.003
Invertebrates	Zero	Intercept	1	-2.81	0.49	-5.78	<0.001
		Sample method- trapped	1	-0.51	0.16	-3.19	0.001
		Cat weight (linear)	1	-0.36	0.06	-6.43	<0.001
		Rain in last 12 months	1	-0.22	0.1	-2.11	0.035
		Preferred food	1	0.14	0.03	5.05	<0.001
		Average month temp. (°C)	1	0.09	0.01	8.52	<0.001
	Count	Intercept		-8.72	53.01	-0.16	0.869
		Sample method- trapped	1	-1	0.28	-3.55	<0.001
		Cat weight (linear)	1	-0.44	0.1	-4.38	<0.001
		Preferred food	1	0.11	0.06	1.97	0.049

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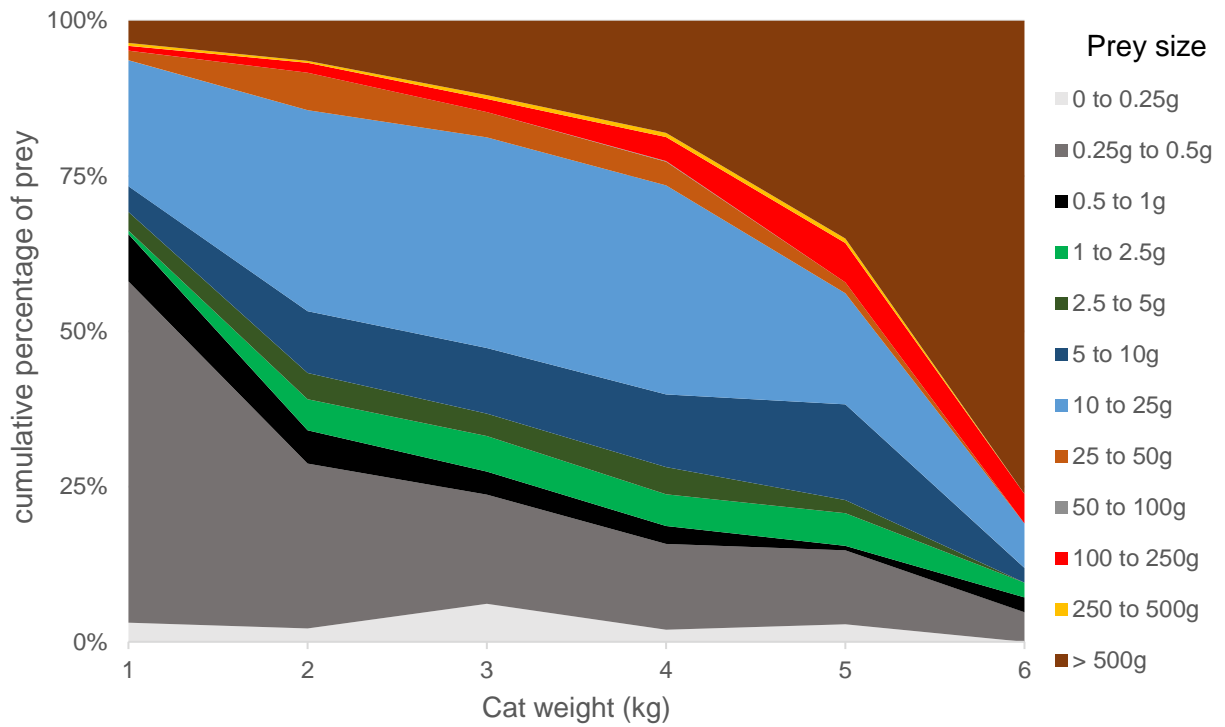
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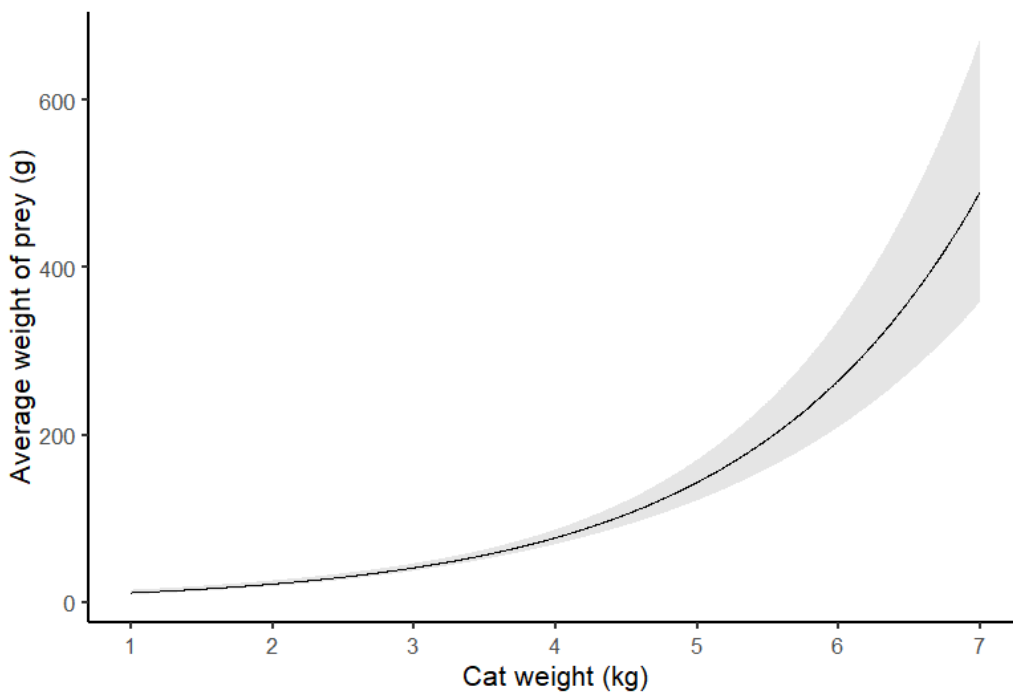
764 Fig 1. Histogram of the spread of weights of adult female and male cats sampled during the  
765 study.

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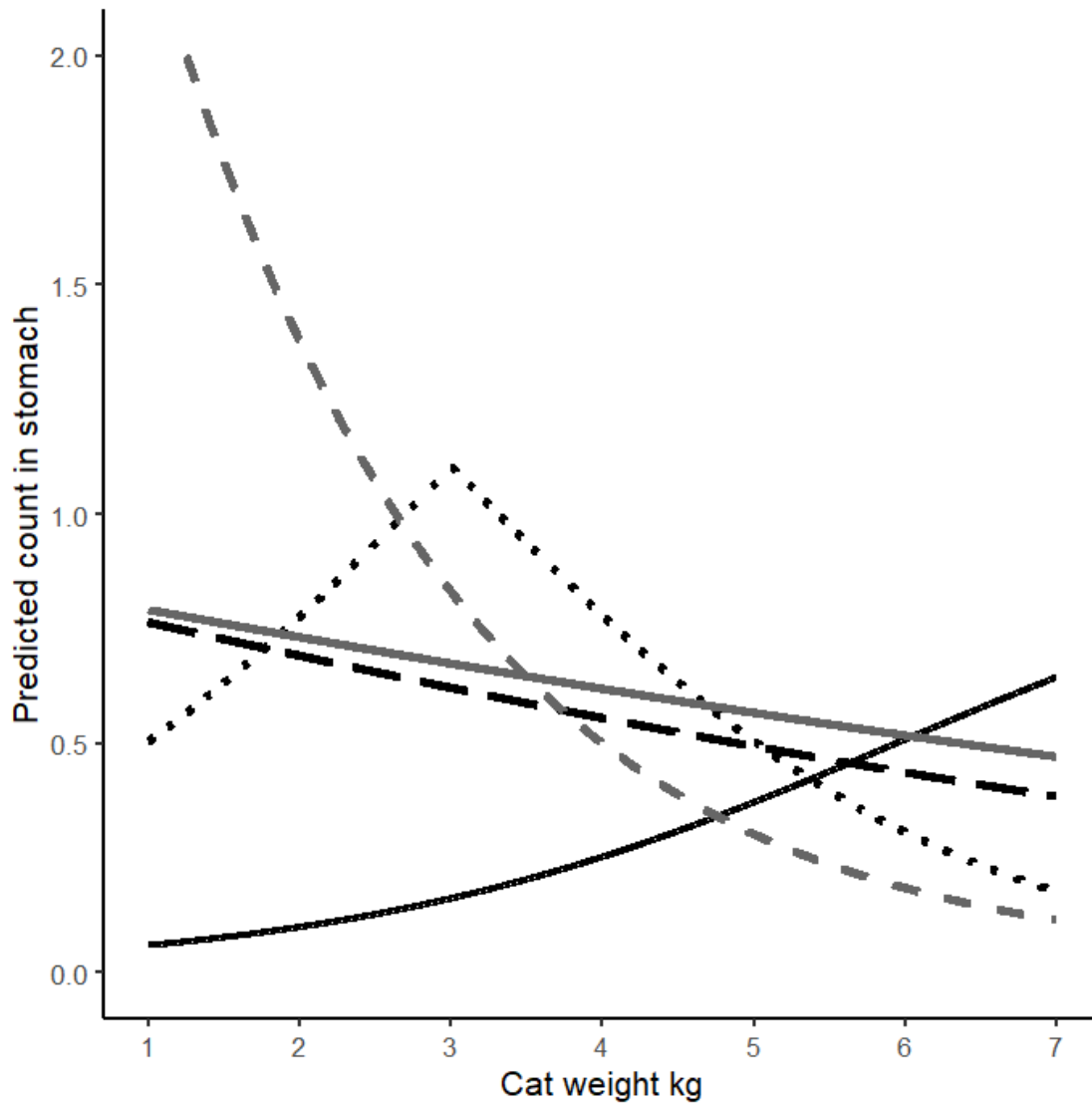
767

768 Figure 2. Distribution of prey size found inside cat stomachs for cats in each cat weight class.



769

770 Figure 3: Model predictions of the relationship between cat weight and average weight of all  
 771 prey species found in each stomach, including 95% confidence interval (grey shading).



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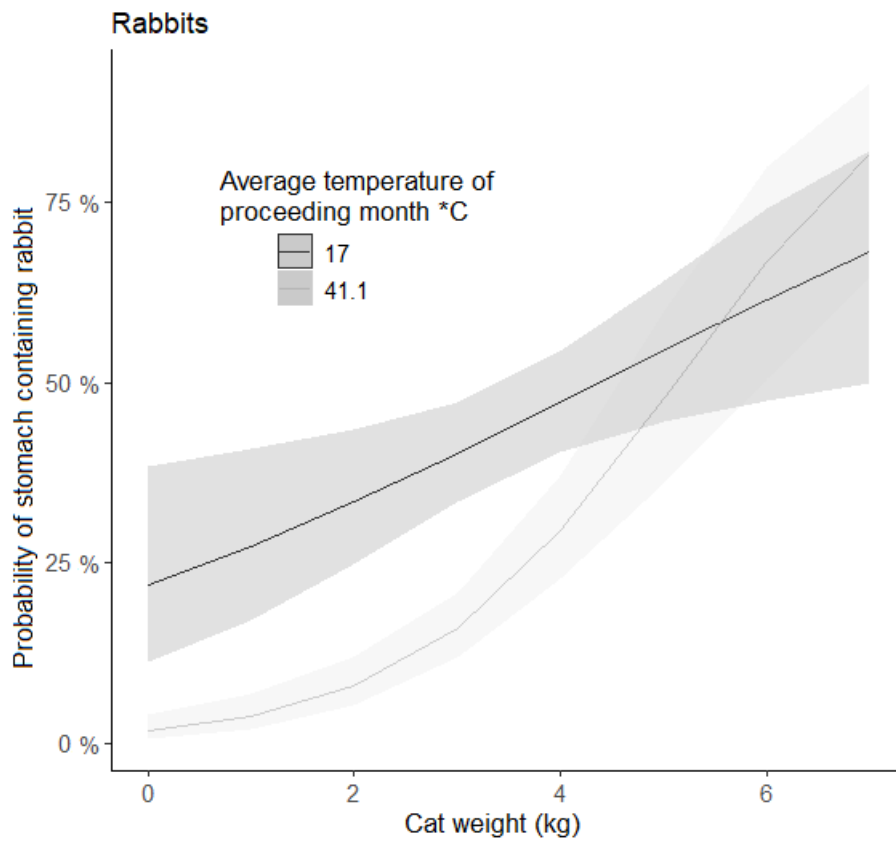
773 Figure 4: Predicted relationships between cat weight and prey categories found in cat  
 774 stomachs from models with the strongest support for the five prey categories; rodents 20-  
 775 60g (dots), rabbits (solid black), small mammals <20g (black dash), herpetofauna (solid grey)  
 776 and invertebrates (grey dash). All other variables were set as constant.

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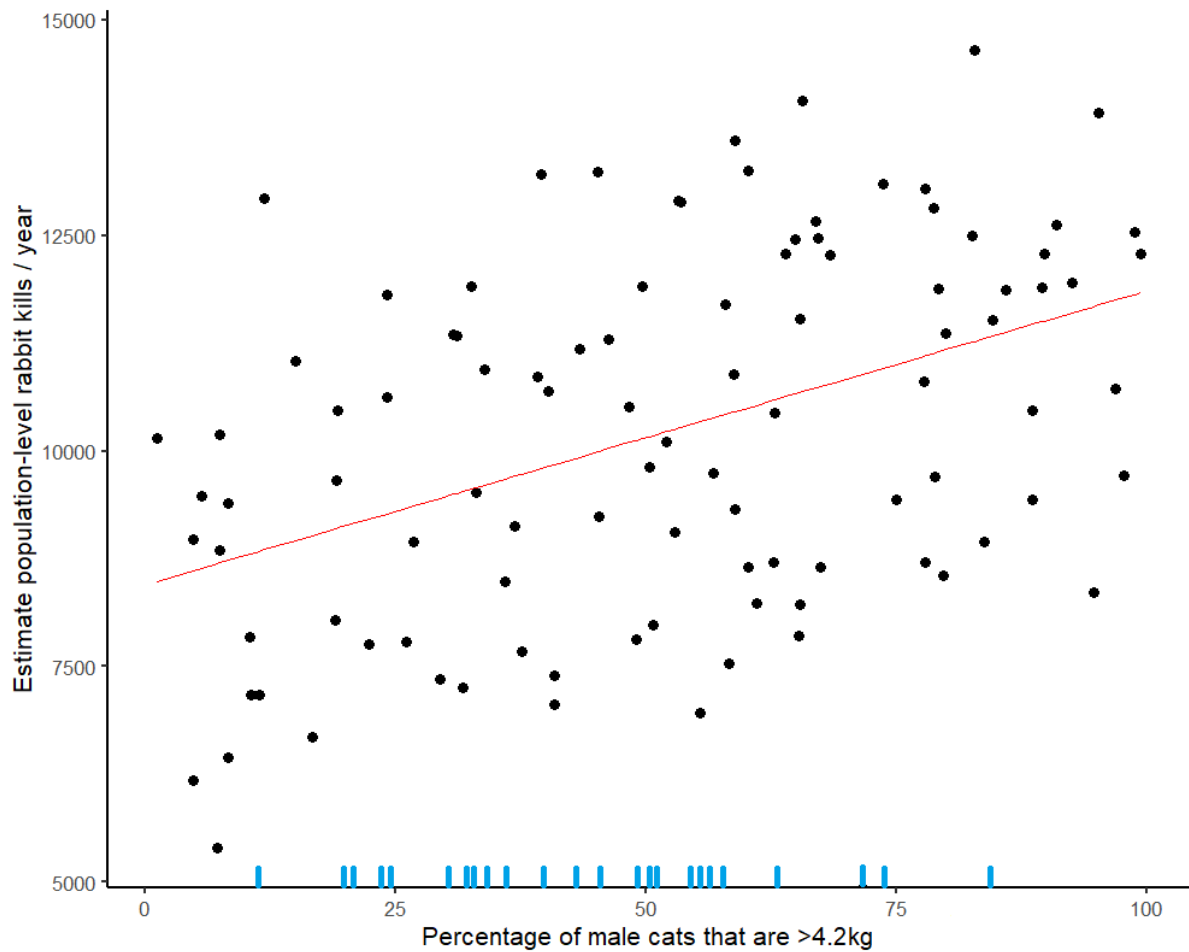


781

782 Figure 5: The probability of a cat stomach containing rabbit in relation to cat weight during

783 the hottest months (pale grey) and coldest months (dark grey) at the study site.





784

785 Figure 6. Estimated annual kill rate of rabbits in 100 simulated populations of feral cats,  
 786 comparing scenarios with varying percentage of the male feral cats >4.2 kg. Constants in all  
 787 scenarios is cat population size (100 cats), a 2:2:1 ratio of male:female:subadult and rabbit  
 788 density of 40km<sup>-2</sup>. The variables of rainfall, temperature and number of rodents 20-60g  
 789 were drawn from a random sample according to real patterns around the study area. Blue  
 790 tick marks indicate natural spread of the percentage of males >4.2 kg sampled from the  
 791 study area in the years between 1990 and 2019.

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795 Supplementary Material 1. Results from model selection tables comparing variables  
 796 affecting prey groups in cat stomachs, using AIC values (see Table 1). Only the top model,  
 797 second best model, global model and null model are presented for brevity.

<b>Prey</b>						
<b>Group</b>	<b>Model details</b>	<b>df</b>	<b>LogLik</b>	<b>AIC</b>	<b>Delta</b>	<b>Weight</b>
<b>Rodents 20-60g</b>						
	Top model	8	-1067	2150.1	0	0.78
	Second model					
	(global)	12	-1064.5	2153.2	3.1	0.16
	Null	3	-1264.5	2534.9	384.8	0
<b>Rabbits</b>						
	Top model	9	-903.4	1825	0	0.81
	Second model					
	(global)	11	-903.1	1828.3	3.3	0.15
	Null	1	-1046.3	2094.6	269.6	0
<b>Small mammals</b>						
	Top model	9	-1633.4	3284.8	0	0.75
	Second model					
	(global)	12	-1631.5	3287.3	2.4	0.22
	Null	3	-1685.9	3377.9	93.1	0
<b>Birds</b>						
	Top model	3	-500.3	1006.6	0	0.46
	Second model	7	-497.6	1009.2	2.5	0.13

global model	10	-496.1	1012.4	5.7	0.03
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Null	1	-516.4	1034.9	28.2	0
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### Herpetofauna

Top model	7	-1504.8	3023.7	0	0.83
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Second model	6	-1508	3028	4.3	0.1
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global model	10	-1507.3	3034.7	11	0
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Null	3	-1575.7	3157.4	133.8	0
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### Insects

Top model	11	-1459.8	2941.8	0	0.62
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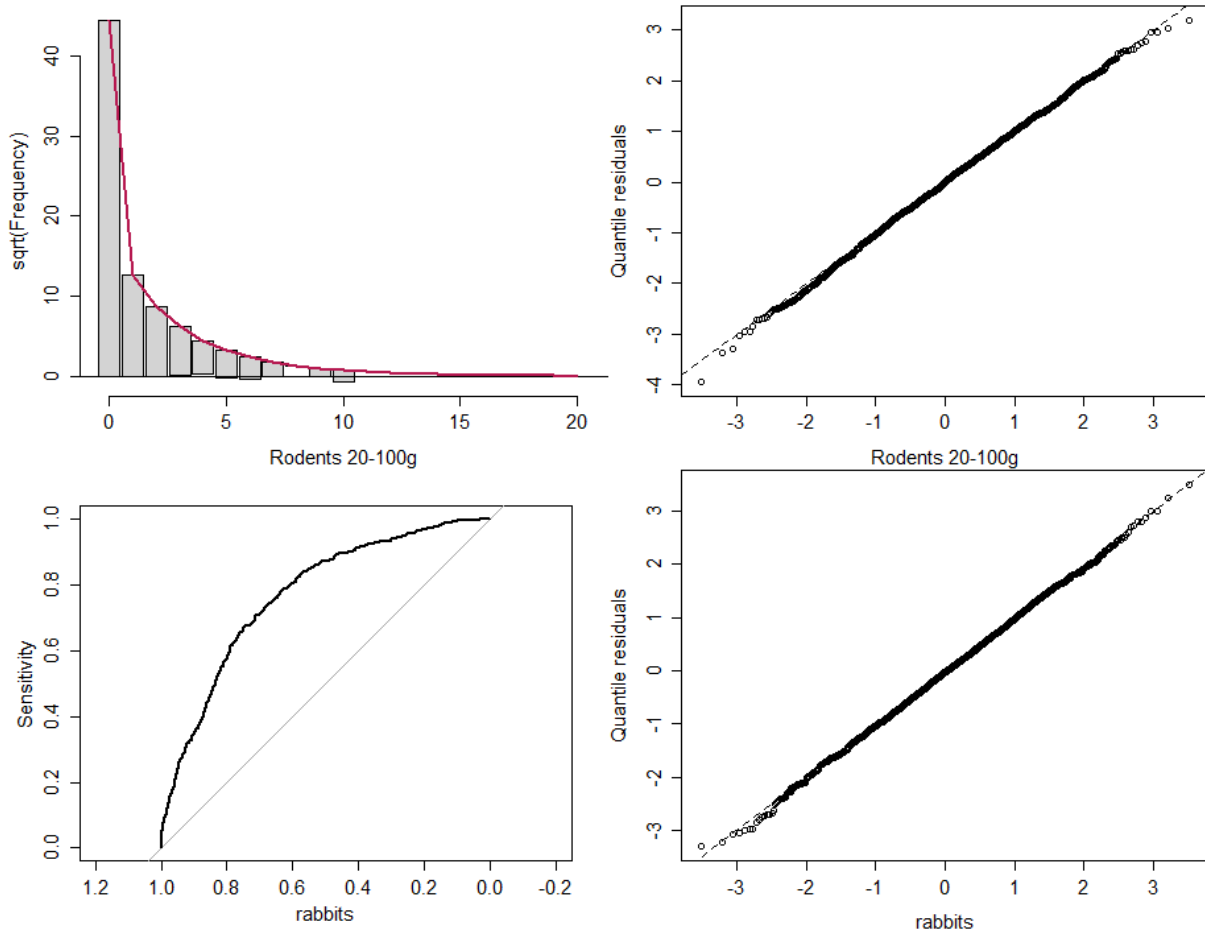
Second model					
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(global)	12	-1459.3	2942.8	1	0.38
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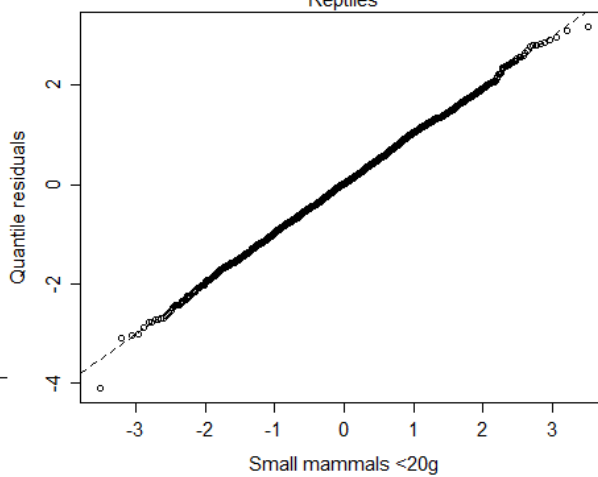
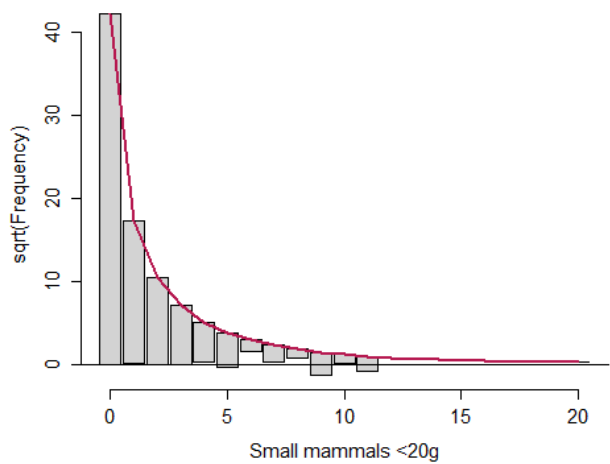
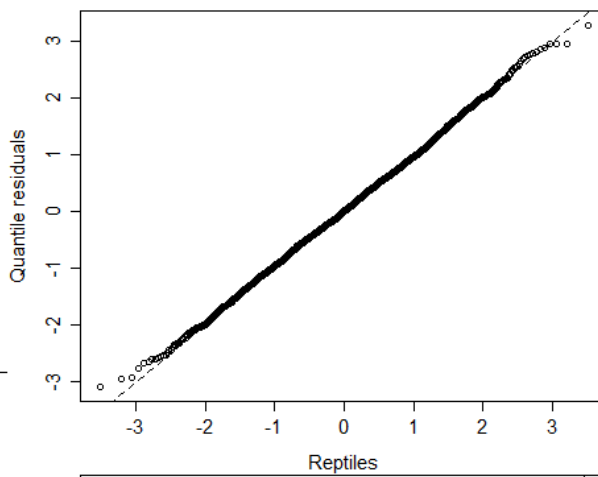
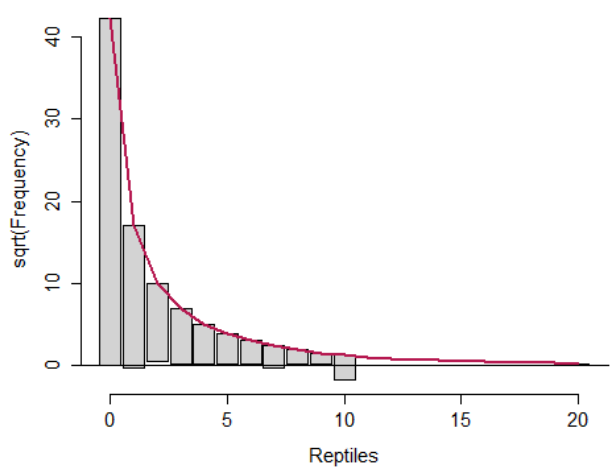
Null	3	-1581.4	3168.9	227.1	0
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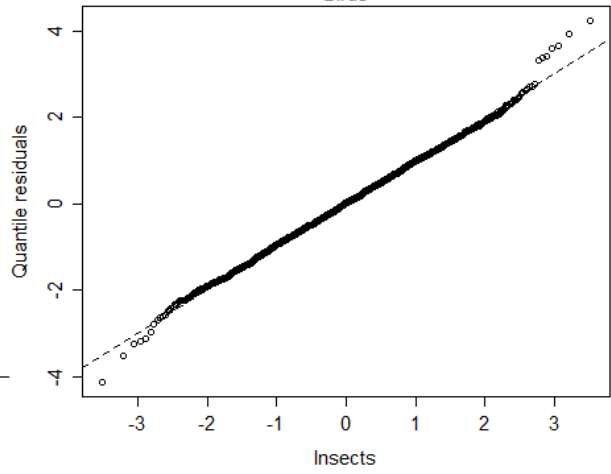
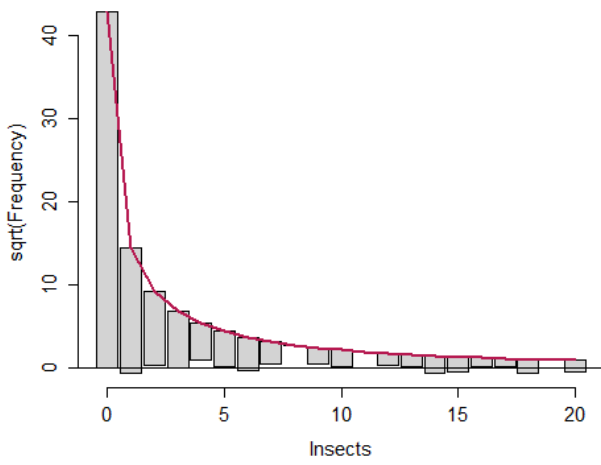
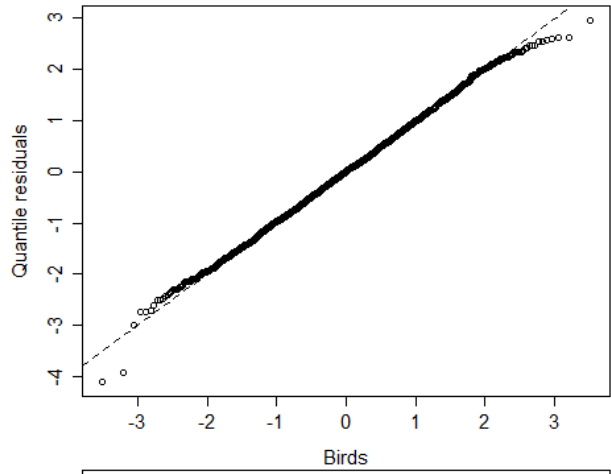
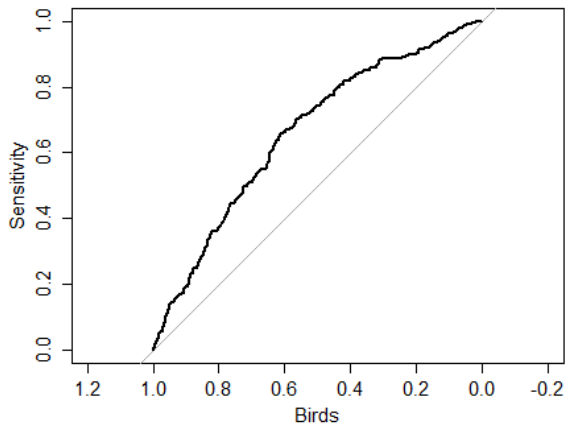
799 Supplementary Material 2. Model performance diagnostics for the top models of the  
800 (Supplementary Material 1) likelihood of certain prey groups being recorded in feral cat stomachs.  
801 For two-stage hurdle models of count data, we present rootograms and QQ residual plots (rodents  
802 20-60g, reptiles, small mammals <20g, insects). For binary models of presence-absence (Rabbits and  
803 birds), we present Area Under Curves (AUC) plots and QQ residual plots



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