

1 **Running head**

2 Bayesian estimates of African lion mortality

3 **Title**

4 Bayesian estimates of male and female African lion mortality for future use in pop-  
5 ulation management

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## 22 **Summary**

- 23 1. The global population size of African lions is plummeting, and many small  
24 fragmented populations face local extinction. Extinction risks are amplified  
25 through the common practice of trophy hunting for males, which makes setting  
26 sustainable hunting quotas a vital task.
- 27 2. Various demographic models evaluate consequences of hunting on lion popu-  
28 lation growth. However, none of these models use unbiased estimates of male  
29 age-specific mortality because such estimates do not exist. Until now, estimat-  
30 ing mortality from re-sighting records of marked males has been impossible due  
31 to the uncertain fates of disappeared individuals: dispersal or death.
- 32 3. We develop a new method and infer mortality for male and female lions from  
33 two populations that are typical with respect to their experienced levels of  
34 human impact.
- 35 4. We found that mortality of both sexes differed between the populations and  
36 that males had higher mortality across all ages in both populations. We discuss  
37 the role that different drivers of lion mortality may play in explaining these  
38 differences, and whether their effects need to be included in lion demographic  
39 models.
- 40 5. *Synthesis and applications.* Our mortality estimates can be used to improve  
41 lion population management and, in addition, the mortality model itself has  
42 potential applications in demographically-informed approaches to the conser-  
43 vation of species with sex-biased dispersal.

## 44 **Key words**

45 age-specific mortality, Siler model, dispersal, Serengeti, Hwange, sex differences in  
46 mortality, sex differences in life history, African lions, lion population management,

48 **Introduction**

49 Estimates of mortality for wild animal populations are important to test ecological  
50 and evolutionary theory, and to project future population size and structure for  
51 population management measures (Griffith *et al.* xxxx). Such measures are needed,  
52 for example, for many populations of African lions *Panthera leo* that are facing local  
53 extinction (Packer *et al.* 2011, Riggio *et al.* 2012, Packer *et al.* 2013). Populations  
54 are further decimated through the common practice of shooting males for trophies  
55 (Packer *et al.* 2011). Trophy hunting is an important yet controversial conservation  
56 tool and setting shooting quota at sustainable levels a vital task (Loveridge *et al.*  
57 2007, Lindsey *et al.* 2012). Consequently, multiple demographic models evaluate  
58 consequences of male offtake on lion population growth (Whitman *et al.* 2004, 2007,  
59 Becker *et al.* 2013), but none of these demographic models use unbiased estimates  
60 of male age-specific mortality, because such estimates do not exist. Male mortality  
61 estimates inferred using the conventional Cormack-Jolly-Seber models (Cormack  
62 1964, Jolly 1965, Seber 1965) are biased, because these models cannot account for  
63 male dispersal. More specifically, they are unable to derive mortality information  
64 from records of males that disappeared from monitored populations around the age  
65 of maturity with uncertain fates (i.e. the males have died or dispersed out of the  
66 study area).

67 The most common answer to male data deficiency in population ecology is to  
68 ignore males altogether and to model only the dynamics of the female population  
69 (see for notable exceptions Schindler *et al.* 2015, Childs *et al.* xxxx). This can be  
70 a legitimate approach depending on the study question and species. However, for  
71 lions it is generally recognized that male mortality may affect population dynam-  
72 ics via the mechanistic link of infanticide (Whitman *et al.* 2004, Caro *et al.* 2009).

73 Frequent deaths of adult males disturb the social structure of prides and male coalitions and result in a higher frequency of male takeovers with subsequent deaths of young infants, premature evictions of juveniles with low survival probability, and a risk of injuries for females trying to protect dependent young (Elliot *et al.* 2014). Furthermore, for many applications of demographic models of lions, males are the sex of interest, because the models aim to address issues of sustainable shooting quota in populations hunted for trophies (Whitman *et al.* 2007). Existing studies of lion population dynamics therefore use various approaches to overcome the lack of male age-specific mortality estimates. Some use population summary statistics without further differentiating the age- or sex-structure of the population (Packer *et al.* 2011), others structure their model in a way that makes a distinction of uncertain male records into deaths and dispersal superfluous (Whitman *et al.* 2004, 2007), or use female mortality estimates for both sexes (Becker *et al.* 2013). The development of a model that can provide estimates for both sexes is therefore a significant addition to the toolbox available for species management.

88 Among dioecious organisms, sex differences in life history are the norm. In mammal species that exhibit a polygynous or polygynandrous mating system in which males compete for access to receptive females by physical combat, males are commonly larger than females, mature later, reproduce for fewer years, and die younger than females (Promislow 1992, Andersson 1994, Clutton-Brock & Isvaran 2007). Lions are a prominent example of this type of mating system, and a shorter male than female life span in this species is well known (Packer *et al.* 1988). Yet the degree to which mortality of males and females differ at different ages, and whether sex differences are stable across populations, is poorly understood. Without this knowledge, the management-relevant consequences of sex differences in mortality cannot be unveiled.

99 Here, we develop a mortality analysis method that can incorporate uncertain male records for species with male natal dispersal. In this model, dispersal state

101 (i.e. whether a male with an uncertain fate dispersed or died) is imputed as a latent  
102 state jointly with the coefficients of a parametric mortality model in a Bayesian hier-  
103 archical framework. The parameters of the mortality model have a distinct biological  
104 interpretation (see Data section in Materials and methods), and the mortality model  
105 thus decomposes mortality into age-dependent and age-independent mortality (Siler  
106 1979). Age-independent mortality represents mortality due to external sources that  
107 kill regardless of age (Pletcher 1999). The mortality decomposition therefore allows  
108 comparison of mortality between the sexes or populations not only in terms of dif-  
109 ferent levels but also in terms of differences in underlying processes. Furthermore,  
110 we extend the basic framework of the model to also account for possible secondary  
111 dispersal, where immigrants to the study area out-migrate again after a period of  
112 residency.

113 Using the model, we estimate age-specific mortality rates for males and females  
114 for two populations of African lions that varied with respect to environmental fac-  
115 tors, densities, and human impact. The first population was hunted for trophies  
116 at its boundary, was subjected to killings in accidents and human-wildlife conflicts,  
117 and lived in a food-scarce environment at low densities (the “disturbed” population).  
118 The second population was hardly impacted by humans and lived in a nutrient-rich  
119 environment at high densities (the “undisturbed” population). We compare age-  
120 specific mortality between these two populations with particular focus on a possible  
121 signature of human impact. We expected to find higher levels of mortality, and par-  
122 ticularly of age-independent mortality, in the disturbed population compared to the  
123 undisturbed one. We also expected to confirm previously observed sex differences in  
124 mortality (Packer *et al.* 1988), and to observe an amplification of this difference in  
125 the population impacted by humans, since males were the primary target of trophy  
126 hunting and had a higher risk of being killed by farmers in retaliation for raided  
127 livestock.

## 128 **Materials and methods**

### 129 **Data**

130 We used life history data from two free-living lion populations that have been mon-  
131 itored for many years. The “disturbed” population lives in the northern range of  
132 Hwange National Park in north-western Zimbabwe. The study area extends to 7000  
133 km<sup>2</sup> and receives 600 mm rainfall seasonally. Vegetation is a mosaic of mixed de-  
134 ciduous woodland and scrubland with limited areas of open or bushed grassland  
135 (Rogers 1993, Loveridge *et al.* 2007). Water is artificially supplied during the dry  
136 season, and the prey assemblage is largely resident. The park borders on hunting  
137 concessions in the north and north-east. Human settlements occur on the north  
138 and east of the park and are mainly used for subsistence agriculture and wildlife ex-  
139 ploitation under the Communal Areas Management Plan for Indigenous Resources  
140 (CAMPFIRE) scheme (Frost & Bond 2008). The park shares a border with wildlife  
141 management areas in Botswana to the west. Life history data were collected be-  
142 tween 1999 and 2013. One female per pride and some resident male nomads and  
143 males of male coalitions wore a radio collar. These prides, resident male nomads,  
144 and male coalitions were located by radio telemetry and censused approximately  
145 once per month. Other males were monitored by opportunistic sightings and pho-  
146 tographs collected from tourists and guides. Field staff identified individual lions  
147 other than the collar-wearing ones from markings such as whisker spot patterns,  
148 scars, and teeth characteristics (Pennycuick & Rudnai 1970, Smuts *et al.* 1978). A  
149 summary of the data used is given in Table 1.

150 The “undisturbed” population occupies a 2000 km<sup>2</sup> study area in the Serengeti  
151 National Park, Tanzania. The area has a south-east to north-west gradient in veg-  
152 etation from short to tall grassland to open woodlands (Packer *et al.* 2005, Mosser  
153 *et al.* 2009). Most rainfall occurs during the wet season, when large herds of migra-  
154 tory herbivores pass through. In response to an increasing abundance of migratory

155 prey, the study population has grown since the start of the study in 1966 (Packer  
156 *et al.* 2005, Packer, unpublished data after 2005). We used life history data collected  
157 between 1966 and 2013. During the early years of the study (1966–1984), observers  
158 gathered data from opportunistic sightings, about 1–3 times per month for most  
159 individuals. Since 1984, tracking the signal of at least one radio-collared female  
160 per pride, observers have sighted each pride 2–6 times per month. The observers  
161 identify individuals from natural markings (Packer *et al.* 1991), and deduce birth  
162 dates of cubs born in the study area from lactation stains on the mothers. A lot  
163 of nomadic males enter the area, most of them migrate through without taking up  
164 residence in the study population. Because of the sparse information on these males,  
165 our analyses excluded all nomadic males that never became residents ( $n = 548$ ). A  
166 summary of data is provided in Table 1.

167 In both studies, trained observers estimate age of individuals with unknown  
168 dates of birth using age indicators such as relative body size, nose colouration, and  
169 eruption and wear of teeth (Smuts *et al.* 1978, Whitman *et al.* 2004). Furthermore,  
170 both data sets contain individuals that died at young ages before sex could be de-  
171 termined (unsexed records). For all data, we identified male records as uncertain  
172 (i.e. the male may potentially have dispersed) if missing males that were born in  
173 the study area (native-borns), and whose deaths were not observed, were older than  
174 1.5 years at disappearance (minimum age at dispersal). Finally, secondary disper-  
175 sal has rarely been observed in the Hwange population (A. Loveridge, unpublished  
176 data). However, the head of the Serengeti study indicated 90 out of 348 immigrants  
177 to possibly have out-migrated again (“potential secondary dispersers”, C. Packer,  
178 unpublished data). This opinion was formed based on the circumstances accompa-  
179 nyng the disappearances. We added these 90 records to the uncertain male records  
180 for which the dispersal state needed to be imputed.

## 181 Mortality analysis

### 182 The parametric mortality and dispersal models

183 We fitted a parametric model for age-specific mortality. With  $X$  being a random  
184 variable for ages at death, the mortality function, or hazard rate of death, for con-  
185 tinuous age  $x$  was

$$\mu(x|\boldsymbol{\theta}) = \lim_{\Delta x \rightarrow 0} \frac{\Pr(x \leq X < x + \Delta x \mid x \leq X, \boldsymbol{\theta})}{\Delta x}, \quad (\text{Eqn. 1})$$

186 where  $\boldsymbol{\theta}$  was a vector of mortality parameters (see Table 2 for a summary of all  
187 random variables, parameters, and indicators). From the mortality rate, the prob-  
188 ability to survive from birth to age  $x$ , or survival function, could be calculated as

189

$$S(x|\boldsymbol{\theta}) = \Pr(X \geq x) = \exp \left[ - \int_0^x \mu(z|\boldsymbol{\theta}) dz \right]. \quad (\text{Eqn. 2a})$$

190 And the probability that death occurred before age  $x$ , or the cumulative density  
191 function (CDF), was

$$F(x|\boldsymbol{\theta}) = \Pr(X < x) = 1 - S(x|\boldsymbol{\theta}), \quad (\text{Eqn. 2b})$$

192 with the probability density function (PDF) for age at death of

$$f(x|\boldsymbol{\theta}) = \frac{d}{dx} F(x|\boldsymbol{\theta}) = S(x|\boldsymbol{\theta}) \mu(x|\boldsymbol{\theta}). \quad (\text{Eqn. 2c})$$

193 To capture the bathtub shape of lion mortality (Packer *et al.* 1998), and to allow  
194 for the estimation of age-independent mortality, we used the Siler model (Siler 1979)  
195 in the form

$$\mu(x|\boldsymbol{\theta}) = e^{a_0 - a_1 x} + c + e^{b_0 + b_1 x}, \quad (\text{Eqn. 3})$$



196 where  $\boldsymbol{\theta}^\top = [a_0, a_1, c, b_0, b_1]$ , with  $a_0, b_0 \in \mathbb{R}$  and  $a_1, c, b_1 > 0$ . The Siler model  
 197 is the sum of three additive mortality hazards (Siler 1979). The first summand  
 198 models the decrease in mortality rates over infant and juvenile ages, with  $a_0$  being  
 199 the initial level and  $a_1$  modelling the rate of decrease. The middle summand is a  
 200 constant hazard  $c$ , also known as a Makeham term (Makeham 1860), that captures  
 201 age-independent mortality. The last summand is the Gompertz law of mortality  
 202 (Gompertz 1825), which captures the exponential increase in mortality rates with  
 203 age from an initial level  $b_0$  with a rate of increase of  $b_1$ .

204 To model the ages at dispersal, we defined the random variable  $Y$  for age at  
 205 natal dispersal (Table 2). It followed  $Y \sim G_Y(y)$  for ages  $y > 0$ , where  $G_Y(y)$   
 206 was the Gamma distribution function with the parameter vector  $\boldsymbol{\gamma}^\top = [\gamma_1, \gamma_2]$ . The  
 207 probability density function (PDF) of age at natal dispersal was given by

$$g_Y(y|\boldsymbol{\gamma}) = \begin{cases} \frac{\gamma_1^{\gamma_2}}{\Gamma(\gamma_2)}(y - \alpha)^{\gamma_2-1}e^{-\gamma_1(y-\alpha)} & \text{if } y \geq \alpha \\ 0 & \text{if } y < \alpha, \end{cases} \quad (\text{Eqn. 4})$$

208 where  $\alpha$  is the minimum age at natal dispersal ( $\alpha = 1.5$  for both populations)  
 209 and  $\gamma_1, \gamma_2 > 0$ . We further defined the random variable  $Z$  for age at secondary  
 210 dispersal, where the age at secondary dispersal was  $Z \sim G_Z(z)$  for ages  $z > 0$ , with  
 211  $G_Z(z)$  being a second Gamma distribution with the parameter vector  $\boldsymbol{\lambda}^\top = [\lambda_1, \lambda_2]$   
 212 (Table 2). Accordingly, the probability density function (PDF) of age at secondary  
 213 dispersal was given by

$$g_Z(z|\boldsymbol{\lambda}) = \begin{cases} \frac{\lambda_1^{\lambda_2}}{\Gamma(\lambda_2)}(z - \alpha)^{\lambda_2-1}e^{-\lambda_1(z-\alpha)} & \text{if } z \geq \alpha \\ 0 & \text{if } z < \alpha, \end{cases} \quad (\text{Eqn. 5})$$

214 where  $\lambda_1, \lambda_2 > 0$ .

215 **Model variables and functions**

216 In a Bayesian hierarchical framework, the model maximized the posteriors of the  
 217 mortality and dispersal models, while imputing the dispersal state for uncertain  
 218 male records (i.e. dispersed or died) and the sex for unsexed records as latent states  
 219 (see Figure 1 for a flowchart of the model structure). Contributions to the mortality  
 220 and dispersal likelihoods varied according to the sex, dispersal state, and migration  
 221 history of the individual. The likelihood for females was constructed as

$$p(x^F, x^L; \boldsymbol{\theta}) = \begin{cases} \Pr(X = x^L | X > x^F) & \text{if uncensored} \\ \Pr(X > x^L | X > x^F) & \text{if censored,} \end{cases} \quad (\text{Eqn. 6a})$$

222 where  $x^L$  denotes the age at last detection and  $x^F$  is the age at first detection.  
 223 Note that  $x^F = 0$  for individuals born in the study area and  $x^F > 0$  for both  
 224 immigrants and individuals born before monitoring began. The likelihood for native-  
 225 born potential natal dispersers was

$$p(x^F, x^L; \boldsymbol{\theta}, \boldsymbol{\gamma}) = \begin{cases} \Pr(X = x^L, Y > x^L | X > x^F) & \text{if uncensored} \\ \Pr(X > x^L, Y > x^L | X > x^F) & \text{if censored} \\ \Pr(X > x^L, Y = x^L | X > x^F) & \text{if dispersed.} \end{cases} \quad (\text{Eqn. 6b})$$

226 While for immigrants that were potential secondary dispersers the likelihood was

$$p(x^F, x^L; \boldsymbol{\theta}, \boldsymbol{\lambda}) = \begin{cases} \Pr(X = x^L, Z > x^L | Y = x^F, X > x^F) & \text{if uncensored} \\ \Pr(X > x^L, Z > x^L | Y = x^F, X > x^F) & \text{if censored} \\ \Pr(X > x^L, Z = x^L | Y = x^F, X > x^F) & \text{if dispersed.} \end{cases} \quad (\text{Eqn. 6c})$$

227 For the imputation of dispersal state for the uncertain male records, we defined  
 228 a binary random variable  $D$ , which assigned 1 if an individual  $i$  dispersed in its  
 229 last detection age  $x_i^L$ , and 0 if otherwise. We furthermore defined a second binary

230 variable  $S$  ( $s_i = 1$  if female,  $s_i = 2$  if male) for the imputation of sex as another  
 231 latent state for unsexed records (Table 2).

232 Finally, we constructed the full Bayesian model as

$$\begin{aligned}
 p(\mathbf{d}_u, \mathbf{s}_u, \boldsymbol{\theta}, \boldsymbol{\gamma}, \boldsymbol{\lambda} \mid \mathbf{d}_k, \mathbf{s}_k, \mathbf{x}^F, \mathbf{x}^L) &\propto \underbrace{p(\mathbf{d}, \mathbf{s}, \mathbf{x}^F, \mathbf{x}^L \mid \boldsymbol{\theta}, \boldsymbol{\gamma}, \boldsymbol{\lambda})}_{\text{likelihood}} \\
 &\times \underbrace{p(\mathbf{d})p(\mathbf{s})}_{\text{priors for states}} \\
 &\times \underbrace{p(\boldsymbol{\theta})p(\boldsymbol{\gamma})p(\boldsymbol{\lambda})}_{\text{priors for parameters}}, \tag{Eqn. 7}
 \end{aligned}$$

233 where  $\mathbf{d}$  was the indicator vector of dispersal states and  $\mathbf{s}$  was the indicator vector  
 234 for sex. Each of these vectors had two subsets represented by the subscripts  $u$  for  
 235 unknown and  $k$  for known.

## 236 Model fitting and conditional posteriors

237 We fitted the model in Eqn. 7 using a Markov Chain Monte Carlo (MCMC) algorithm  
 238 in four parallel sequences. We randomly drew starting values and set the number  
 239 of iterations to 15000 steps with a burn-in of 5000 initial steps and a thinning  
 240 factor of 20. We used a hierarchical framework that only needed the conditionals  
 241 for posterior simulation by Metropolis-within-Gibbs sampling (Gelfand & Smith  
 242 1990, Clark 2007). This means that, for this particular case, the algorithm divided  
 243 the posterior for the joint distribution of unknowns into five parts: (a) estimation of  
 244 mortality parameters, (b) estimation of natal dispersal parameters, (c) estimation of  
 245 secondary dispersal parameters, (d) imputation of unknown dispersal state, and (c)  
 246 imputation of unknown sexes. We provide details about the conditional posteriors  
 247 and the acceptance probabilities for the different parts in Appendix S1 in Supporting  
 248 Information.

## 249 Mortality and dispersal priors

250 The Siler parameters for the prior for females were  $a_{0p} = -1.4$  ( $\sigma = 0.5$ ),  $a_{1p} = 0.65$   
251 ( $\sigma = 0.25$ ),  $c_p = 0.07$  ( $\sigma = 0.25$ ),  $b_{0p} = -3.8$  ( $\sigma = 0.5$ ), and  $b_{1p} = 0.2$  ( $\sigma = 0.25$ ),  
252 and for males  $a_{0p} = -1.2$  ( $\sigma(a_{0p}) = 0.5$ ),  $a_{1p} = 0.7$  ( $\sigma(a_{1p}) = 0.25$ ),  $c_p = 0.16$   
253 ( $\sigma(c_p) = 0.25$ ),  $b_{0p} = -3.5$  ( $\sigma(b_{0p}) = 0.5$ ), and  $b_{1p} = 0.23$  ( $\sigma(b_{1p}) = 0.25$ ). For  
254 dispersal, the Gamma parameters (shape and scale) for the dispersal priors were set  
255 to  $\gamma_p = \lambda_p = \{8, 2\}$  with  $\sigma(\gamma_p) = \sigma(\lambda_p) \{2, 1\}$ . All priors for parameters were fairly  
256 uninformative and within the bounds given by the life expectancies of medium to  
257 long-lived animals. The priors for sex as a latent state corresponded to the empirical  
258 sex ratios at model start ages (Table 1).

## 259 Model application and posterior analysis

260 To study the differences in mortality between the sexes and between the two popu-  
261 lations, we fitted the model with both sex and population as covariates, and allowed  
262 for an interaction effect between sex and population. Since the negative exponential  
263 part of the Siler model may have problems in capturing the very steep decline in  
264 infant mortality after birth, the mortality at adjacent infant and juvenile ages can  
265 be overestimated by models fitted from birth. To evaluate this issue and with the  
266 goal of providing the best possible estimates, as well as estimates across the entire  
267 life span, we fitted the model from three different starting ages: birth, 0.5 years of  
268 age, and 1 year of age. Since the latter was not affected by the constraints of the  
269 Siler function we used this model for further investigations.

270 We predicted mortality rates for each sex and population using the parameter  
271 estimates of each MCMC iteration after burn-in and thinning. We then used these  
272 predictions to calculate the mean and credible intervals of age-specific mortality  
273 rates. Since we were *a priori* interested in the effects of population and sex on mor-  
274 tality, we decided against taking a model selection approach and instead calculated  
275 Kullback-Leibler (KL) divergences of the mortality parameter posterior densities

276 (Kullback & Leibler 1951, McCulloch 1989, Burnham & Anderson 2001, Barthold  
277 *et al.*, unpublished data). The KL divergence compares two probability density dis-  
278 tributions and can be interpreted to measure the amount of information lost when  
279 using the second probability density distribution to approximate the first one. After  
280 a simple calibration of the KL values (McCulloch 1989), the values range from 0.5  
281 to 1, where a value of 0.5 indicates that the distributions are identical, and 1 that  
282 they do not overlap at all.

## 283 **Mortality measures**

284 We report mortality information as mortality rates, defined as the instantaneous  
285 hazard of death, and also known as the force of mortality (see Equation Eqn.1).  
286 From the continuous age-specific mortality rate, the discrete age-specific probability  
287 of survival  $_{\Delta x}p_x$  can be calculated. It is defined as the probability to survive from  
288 age  $x$  to age  $x + \Delta x$ , with  $\Delta x = 1$  for annual survival. Survival probabilities  
289 are a common mortality measure in capture–recapture studies, where they are also  
290 termed survival rate. To make our mortality estimates widely usable, we describe  
291 this calculation in Appendix S2. We also calculated life expectancy at the model  
292 start age. Finally, we computed the PDFs for age at death for males and females  
293 in the Serengeti and Hwange (see Equation Eqn.2c). All analyses were conducted  
294 using the statistical programming language R (R Core Team 2012).

## 295 **Results**

### 296 **Population differences in mortality**

297 The models converged for all estimated parameters and all starting ages (Figure 2,  
298 see also Figure S1 to S3 for traces). However, the posterior distributions for Hwange  
299 were wider than those for the Serengeti, which was expected due to the smaller  
300 sample size of the Hwange data (Figure 2). This is also reflected in the wider con-

301 fidence bands around the mean estimated mortality rates for Hwange compared to  
302 the Serengeti (Figure 3). A model that allowed all Serengeti immigrants with un-  
303 certain fates at the age of last detection to be potential secondary dispersers did  
304 not converge. We therefore decided to restrict the potential secondary dispersers  
305 as described in the data section. Overall mortality of both sexes was U-shaped in  
306 the Serengeti with high initial cub mortality, low mortality of prime-aged adults,  
307 and an age-dependent increase in mortality during the older ages (Figure 3, left  
308 panels). Hwange lions also showed higher senescent mortality than prime-age mor-  
309 tality, although the confidence bands were wider than for the Serengeti population  
310 (Figure 3, right panels). The main difference in overall mortality between the two  
311 populations was that, in Hwange, we could only detect moderately elevated levels  
312 of cub mortality compared to the mortality of prime-aged adults. Furthermore, this  
313 result only held for the model that estimated mortality from birth (Figure 3, upper  
314 right panel).

315 The KL divergences comparing Serengeti and Hwange females revealed that mor-  
316 tality of females differed between the two populations (Figure 2, lower right panel).  
317 Females in the Serengeti had higher initial cub mortality ( $a_0$ ) and a steeper decline  
318 in mortality over infant and juvenile ages ( $a_1$ ), yet slightly lower levels of prime-  
319 adult mortality, and similar levels of senescent mortality, when compared to females  
320 in Hwange ( $b_0$ , and  $b_1$ ). Due to higher cub mortality, they also had a shorter life  
321 expectancy at birth (or 0.5 years and 1 year of age where applicable) (Figure 3).

322 As with females, the mortality of Serengeti males differed from the mortality  
323 of Hwange males (Figure 2). Initial cub mortality and the increase in senescent  
324 mortality ( $a_0$ ,  $b_1$ ) were lower in Hwange, but age-independent mortality ( $c$ ) was  
325 much higher, resulting in higher mean mortality rates across the prime-adult ages  
326 for Hwange males compared to Serengeti males (Figures 3, 2). The PDFs for age at  
327 dispersal for males are illustrated in Figure S4.

## 328 **Sex differences in mortality**

329 Mortality also varied between the sexes. In the Serengeti population, mortality of  
330 males was higher than mortality of females across all ages (Figure 3, left panels).  
331 There was an overlapping of confidence bands of male and female cub mortality in  
332 the model fitted from birth (Figure 3, upper left panel). However, this result may  
333 stem from the imputation of sex as a latent state for the many unsexed individu-  
334 als (Table 1), which increased uncertainty. The sex difference in mortality is also  
335 reflected in the KL divergences (Figure 2).

336 In Serengeti, males and females had identical levels of age-independent mortality  
337 ( $c$  parameter). The other parameter estimates showed little or no overlap, with  
338 the parameter distributions governing early-life mortality overlapping the least. In  
339 Hwange, males also had higher mortality than females. However, the difference in  
340 mortality was almost entirely due to different age-independent mortality indicated  
341 by the  $c$  parameter, which was strikingly higher for males than for females. This  
342 indicates a source of age-independent mortality that was sex-selective.

343 Due to the higher male than female mortality in both populations, life expectancy  
344 at birth, or at 0.5 year and 1 year of age for the models fitted from these respec-  
345 tive ages, was lower for males than for females in both populations (Figure 3). For  
346 the model fitted from birth, the mean PDFs for age at death showed that more  
347 male than female deaths occurred up to the age of about six years in both popula-  
348 tions, although because of different reasons: higher age-dependent male than female  
349 mortality in Serengeti and higher age-independent male than female mortality in  
350 Hwange (Figure 2).

## 351 **Discussion**

352 Estimates of age-specific mortality, particularly for males and for populations dis-  
353 turbed by humans, are often a missing piece of information for developing demo-

354 graphic models for population management. Here, we found that, contrary to ex-  
355 pectation, males in the disturbed Hwange population outlived those in the undis-  
356 turbed Serengeti population by approximately one year, and Hwange females out-  
357 lived Serengeti females by approximately two years, despite Hwange’s history of lions  
358 being killed in accidents, as trophies, and as retaliation for raided livestock (Fig-  
359 ure 3) (Loveridge *et al.* 2007). Although adult mortality was, as predicted, higher in  
360 Hwange, the difference was too small to compensate for the much lower mortality of  
361 cubs in Hwange compared to the Serengeti. Despite mortality varying between the  
362 populations, females outlived males in both of them. However, the sex difference was  
363 driven by different mechanisms in each locality: lower age-dependent female than  
364 male mortality in Serengeti and lower age-independent female than male mortality  
365 in Hwange. In the following, we discuss how possible drivers of lion mortality may  
366 explain our findings.

### 367 **The effect of hunting areas bordering on national parks**

368 The decomposition of adult mortality rates into age-dependent and age-independent  
369 mortality revealed that many Hwange lions die of extrinsic causes that kill regardless  
370 of age (high  $c$  parameters). Particularly for males, who had a higher  $c$  parameter  
371 than females, one of these causes is likely trophy hunting that occurs at the park  
372 boundary (Loveridge *et al.* 2007). Trophy hunting around Hwange National Park  
373 recently made international news when the cherished Hwange lion “Cecil” was lured  
374 out of the park and killed (e.g. via The New York Times; Rogers 2015). Hunting  
375 threatens males who reside at the park boundary but also those that migrate from  
376 deeper in the park to the commonly male-depleted edge (Loveridge *et al.* 2007).  
377 The signature that trophy hunting leaves on Hwange mortality in the form of high  
378  $c$  parameters becomes strikingly clear in the comparison of the two populations.  
379 In contrast to what we found for Hwange lions, mortality due to age-independent,  
380 extrinsic causes is low in the Serengeti (low  $c$  parameters) for both sexes. Mortality



381 of adult lions is instead dominated by the increase in senescent mortality with age ( $b_0$   
382 and  $b_1$  parameters). The difference in the composition of adult mortality between the  
383 populations highlights the threat that hunting poses in areas adjacent to protected  
384 areas (Loveridge *et al.* 2007). Given the dominant role of trophy hunting in driving  
385 mortality of Hwange males, an important future application of our framework will be  
386 to refine our mortality estimates by studying how they vary over time with different  
387 hunting intensities.

### 388 **Density dependence as a driver of lion mortality**

389 Our results confirmed that adult mortality is higher in the disturbed Hwange popu-  
390 lation than in the undisturbed Serengeti population; however, the difference in adult  
391 mortality would potentially be larger were it not for density dependence effects (i.e.  
392 the positive correlation between lion population density and their mortality rates).  
393 Since the start of the Serengeti study, the population has been growing due to a long-  
394 term increase in prey availability. However, growth did not occur continuously but  
395 through step-wise increases in mean population sizes, which remained stable across  
396 multiple years (Packer *et al.* 2005). This pattern of population growth is created by  
397 the dynamics of between- and within-group competition. Only when an exception-  
398 ally large cohort of cubs recruits to a pride can a large enough fraction of the pride  
399 split off to successfully compete with other prides for space. In years where these  
400 conditions are not met, mortality may be density-dependent due to within-group  
401 competition (Packer *et al.* 2001, Mosser *et al.* 2009). Density-dependent mortality  
402 has been indirectly observed for lions in Kruger National Park, where lion density  
403 was positively associated with prey biomass density. As prey biomass increased,  
404 lion mortality first declined then increased, indicating that mortality increases at  
405 greater lion densities (Ferreira & Funston 2010). Over the past 15 years the Serengeti  
406 population size has remained stable (C. Packer, unpublished data), suggesting that  
407 the population may have reached carrying capacity. Therefore, density dependence,

408 alongside epizootic diseases, may cause mortality to be relatively high throughout  
409 the study period despite the observed long-term population growth and absence of  
410 human impact (Packer *et al.* 1999, 2005).

411 In comparison, the Hwange population has, because of conservation measures,  
412 increased by 46 % since 2000, with a 200 % increase in the number of adult males  
413 since 2004 (A. Loveridge, unpublished data), when the population sex ratio was  
414 heavily skewed towards females due to trophy hunting (Loveridge *et al.* 2007). Since  
415 2004, the trophy hunting regime has changed markedly with smaller quotas now in  
416 place. However, other sources of anthropogenic mortality including poaching and  
417 conflict mortality remain unchanged. Anthropogenic mortality certainly increases  
418 the Hwange mortality rates, yet anthropogenic mortality may also keep the popu-  
419 lation at low densities and thus in a perpetual state of density-independent growth  
420 with the associated low age-dependent mortality. Therefore, the levels of adult  
421 mortality in the Serengeti and Hwange populations may be driven by two different  
422 mechanisms: density dependence in the Serengeti and anthropogenic mortality in  
423 Hwange.

424 Low density is also the most likely explanation for the observed low cub mortality  
425 in Hwange. While estimates of cub mortality in both populations fall within the  
426 previously reported ranges for other populations (Van Orsdol *et al.* 1985, Becker  
427 *et al.* 2013), the difference is remarkable and causes a considerable difference in life  
428 expectancy at birth between the two populations. In support of this interpretation,  
429 cub survival has previously been shown to decrease in the presence of juveniles and  
430 sub-adults in the pride, which in turn depends on female density (Packer *et al.*  
431 2001). However, alternative explanations for the difference in cub mortality are also  
432 possible. For example, a larger number of early-occurring deaths may go unrecorded  
433 in Hwange because cubs may be older at first sighting in the densely vegetated  
434 Hwange landscape when compared to the more open Serengeti. Nevertheless, our  
435 findings, both for cub and adult mortality, suggest that density dependence may be

436 an important driver of lion mortality. If this is the case, then density dependence  
437 needs to be included in demographic models that aim to determine sustainable  
438 trophy hunting quota by projecting the population's development. The impact of  
439 density dependence on lion mortality may be quantified in future work by using our  
440 model to estimate how mortality varies with density.

## 441 **Co-variation of hunting mortality with cub and female mor-** 442 **tality**

443 Hunting of adult males can decrease population size by more than just the trophy  
444 head count if hunting mortality of adult males co-varies with male mortality by  
445 other causes, or with mortality of females and cubs. With rising adult male mortal-  
446 ity cub mortality can increase because of an increase in the rate of pride takeover  
447 and infanticide (Whitman *et al.* 2004). Similarly, juvenile mortality can increase  
448 because of the eviction of juveniles that are too young to survive on their own  
449 (Elliot *et al.* 2014). It is also plausible that adult females defending their young,  
450 and resident males fighting intruders, could be fatally injured. Based on these hy-  
451 potheses, we expected to find higher cub mortality in the population with higher  
452 adult male mortality. However, we found the opposite, which could be explained by  
453 other factors including density dependence. Longitudinal analyses within the popu-  
454 lations may reveal the expected relationship between cub and adult male mortality.  
455 Furthermore, the fact that both Hwange males and females have elevated levels of  
456 age-independent mortality suggests possible co-variation between male and female  
457 adult mortality. By carefully increasing the time-resolution of our mortality esti-  
458 mates, our framework could in future help unveil these co-variations among age-and  
459 sex-specific mortality, which are necessary to project lion population development  
460 under changed male mortality rates.

## 461 **Conclusion**

462 We have shown how the combination of a Bayesian hierarchical framework with a  
463 parametric mortality model can provide mortality estimates for both sexes in species  
464 with sex-biased dispersal. We provided mortality estimates for both sexes from two  
465 lion populations that experience varying environments, densities, and exposures  
466 to human impact. In the undisturbed population, natural mortality governs the  
467 mortality trajectory of lions, while in the disturbed population trophy hunting and  
468 other anthropogenic mortality left a clear signature on mortality. Because of the  
469 detected variation in mortality between the two populations, we pinpoint the study  
470 of lion mortality drivers as an important research area. Our framework for estimating  
471 lion mortality can be used to test the role of these potential drivers, including density  
472 dependence, trophy hunting, and co-variation of adult male mortality with cub and  
473 female mortality. The framework can therefore be employed to refine demographic  
474 models built to make population management recommendations for lions, but can  
475 also be applied to other species where dispersal behaviour of one or both sexes has  
476 hindered the estimation of mortality.

## 477 **Acknowledgements**

478 We are grateful to the Director of Zimbabwe Parks and Wildlife Management Au-  
479 thority for permission to undertake this research and the assistance of ZPWMA and  
480 Hwange Lion Project field staff for field data collection and facilitation of research.  
481 This work was funded and supported by Darwin Initiative for Biodiversity Grants  
482 162-09-015 and EIDPO002, Mitsubishi Fund for Europe and Africa, R.G. Franken-  
483 berg Foundation, Boesak and Kruger Foundation, Rufford Maurice Laing Founda-  
484 tion, SATIB Trust, Eppley Foundation, Riv and Joan Winant and Recanati-Kaplan  
485 Foundation. DWM thanks the Recanati-Kaplan Foundation and the Robertson  
486 Foundation. JAB acknowledges funding from the International Max Planck Re-

487 search Network on Aging (MaxNetAging). JAB thanks Owen Jones, Jacques Deere,  
488 Emily Simmonds, Ben Sheldon, Jason Matthiopoulos, Tim Coulson, and the re-  
489 viewers, Dave Koons and Johnny Wilson, and editors, Marc Cadotte and Jessica  
490 Metcalf, for helpful comments on the manuscript.

## 491 **Data Accessibility**

492 The data and R code used in this study have been deposited in the Dryad data  
493 repository: <http://dx.doi.org/10.5061/dryad.2382q> (Barthold *et al.* 2016).

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617 fects of trophy selection and environmental disturbance on a simulated population  
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## 619 **Supporting Information**

620 **Appendix S1** Detailed description of the conditional posteriors and acceptance  
621 probabilities.

622 **Appendix S2** Method to convert continuous mortality rate into discrete probabili-  
623 ties of survival.

624 **Table S1** Estimated Siler model coefficients for the models fitted from birth, 0.5  
625 years, and 1 year of age, respectively.

626 **Figure S1, S2, S3** Traces of mortality and dispersal parameter estimation (models  
627 fitted from ages 0, 0.5, and 1, respectively).

628 **Figure S4** Probability density functions for age at dispersal (model fitted from 1  
629 year of age).

## Tables

Table 1: Sample sizes for males (M), females (F), and individuals that died before sex could be determined (U)

Sample	<i>Serengeti</i>			<i>Hwange</i>			<i>Sex ratio</i>
	M	F	U	M	F	U	F:M
Birth <sup>†</sup>	1466(316)*	1507	875	174(32)	244	140	0.51**
0.5	988(315)	1095	62	168(32)	235	60	0.53
1	763(315)	905	4	157(32)	225	32	0.55

<sup>†</sup>“Birth” indicates the sample that includes all individuals. “0.5” and “1” indicate the samples that include individuals that survived to at least 0.5 years and 1 year, respectively.

\*For males sample sizes refer to the number of native-born individuals, followed by the number of immigrants in brackets.

\*\*Female to male sex ratio among all native-borns (pooled data, excluding immigrants) assuming a sex ratio of 0.5 among individuals that died before sex could be determined.

Table 2: Random variables, observed variables, and indicators

<b><i>Modelled random variables</i></b>	
$X$	Random variable for age at death, where $x$ is any age element
$Y$	Random variable for age at natal dispersal with elements $y$
$Z$	Random variable for age at secondary dispersal with elements $z$
$D$	Binary random variable for disperser or non-disperser
$S$	Binary random variable for sex
<b><i>Observed variables and indicators</i></b>	
$\mathbf{x}^F$	Vector of ages at first detection ( $x_i^F = t_i^F - b_i$ )
$\mathbf{x}^L$	Vector of ages at last detection ( $x_i^L = t_i^L - b_i$ )
$\mathbf{m}$	Indicator vector for immigrants ( $m_i = 1$ if immigrant)
$\boldsymbol{\omega}$	Indicator vector for potential natal dispersers ( $\omega_i = 1$ if $m_i = 0$ , $s_i = 0$ , and uncertain fate at $x_i^L \geq \alpha$ , and $\omega_i = 0$ otherwise)
$\mathbf{v}$	Indicator vector for potential secondary dispersers ( $v_i = 1$ if $m_i = 1$ , $s_i = 0$ , uncertain fate at $x_i^L$ , and expert indicated potential secondary dispersal, and $v_i = 0$ otherwise)
<b><i>Updated indicators</i></b>	
$\mathbf{d}$	Indicator vector for dispersers ( $d_i = 1$ if disperser and $d_i = 0$ otherwise)
$\mathbf{s}$	Indicator vector for sex ( $s_i = 1$ if female and $s_i = 0$ otherwise)
<b><i>Parameters</i></b>	
$\boldsymbol{\theta}$	Vector of mortality parameters
$\boldsymbol{\gamma}$	Vector of natal dispersal parameters
$\boldsymbol{\lambda}$	Vector of secondary dispersal parameters
<b><i>Functions</i></b>	
<i>Mortality</i>	
$\mu(x   \boldsymbol{\theta})$	Mortality (Siler model)
$S(x   \boldsymbol{\theta})$	Survival
$F(x   \boldsymbol{\theta})$	Cumulative density function (CDF) for age at death ( $F(x) = 1 - S(x)$ )
$f(x   \boldsymbol{\theta})$	Probability density function (PDF) for age at death
<i>Dispersal</i>	
$g_Y(y   \boldsymbol{\gamma})$	PDF for age at natal dispersal (gamma distribution)
$G_Y(y   \boldsymbol{\gamma})$	CDF for age at natal dispersal
$g_Z(z   \boldsymbol{\lambda})$	PDF for age at secondary dispersal (gamma distribution)
$G_Z(z   \boldsymbol{\lambda})$	CDF for age at secondary dispersal

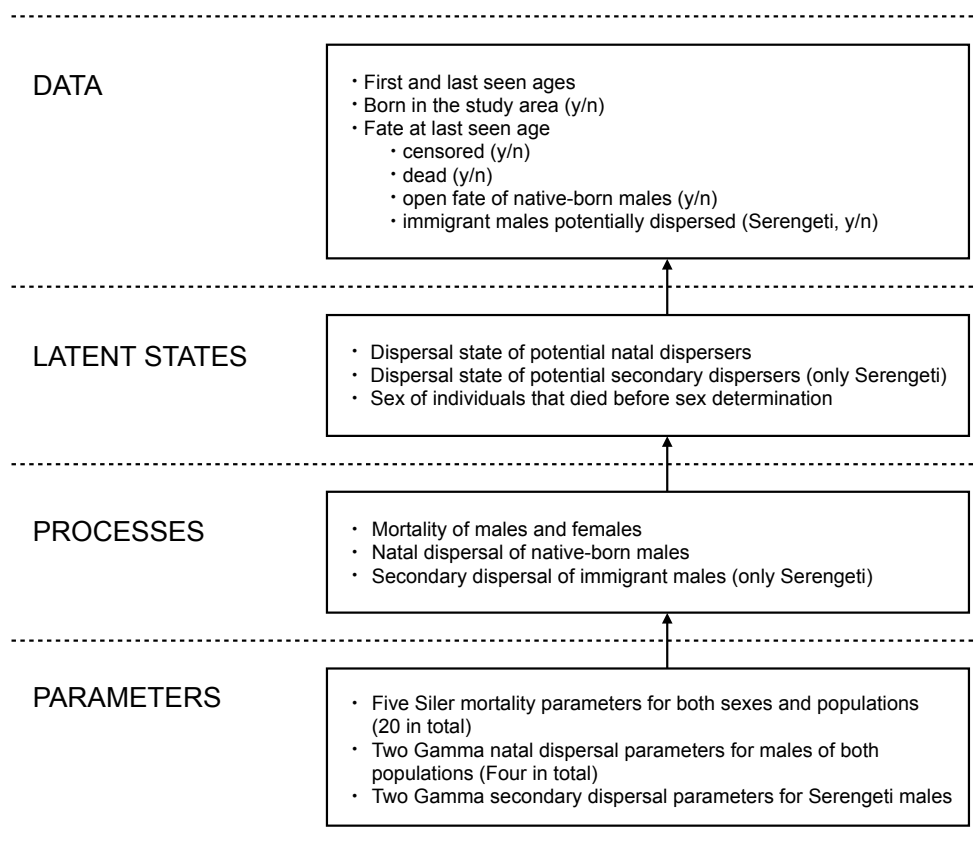


Figure 1: Structure of the Bayesian hierarchical model to infer age-specific mortality of the Serengeti and Hwange lions.

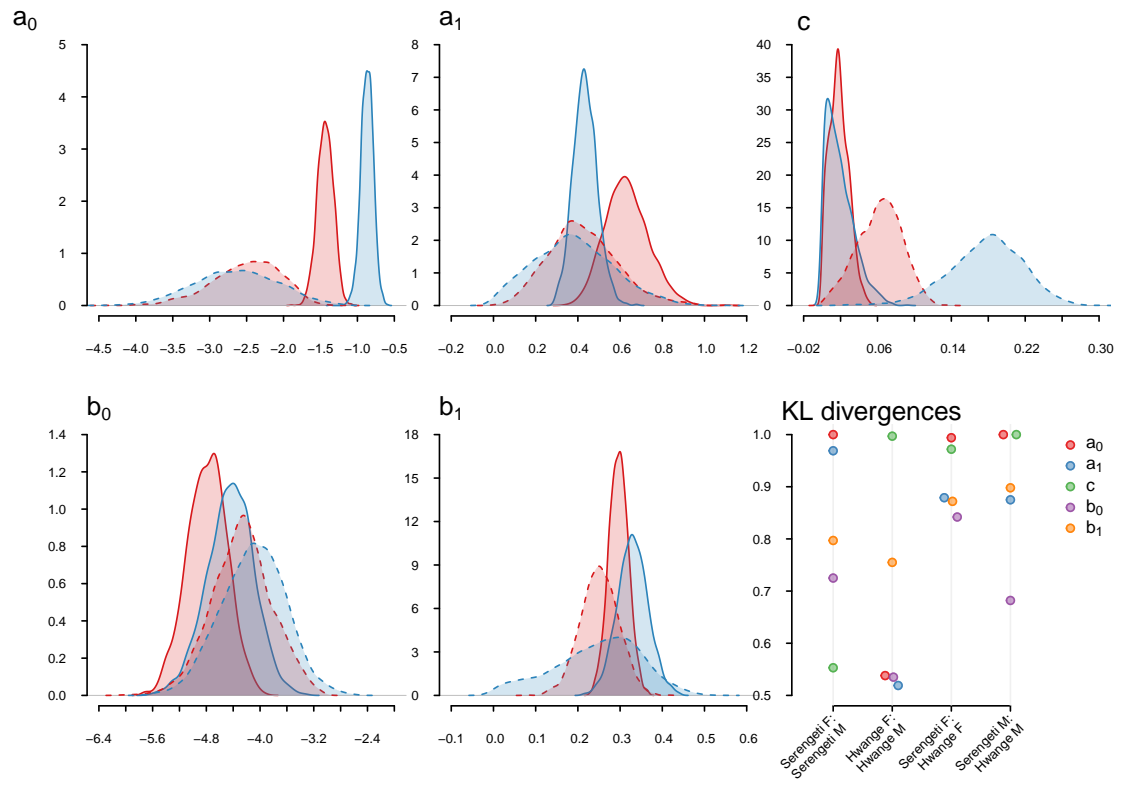


Figure 2: Posterior distributions of Siler parameter estimates for female (pink) and male (blue) African lions of the Serengeti (solid lines) and Hwange (dashed lines) populations. The Siler parameters and their biological interpretation are: initial level ( $a_0$ ) and rate of exponential decrease with age of infant mortality ( $a_1$ ), age-independent mortality ( $c$ ), and initial level ( $b_0$ ) and rate of exponential increase of mortality with age ( $b_1$ ). Also shown are Kullback-Leibler (KL) divergences comparing parameter posteriors between females (F) and males (M) within populations, and within sexes between the populations. Note that some KL divergence estimates are jittered in x-axis direction to improve visibility. The analysis was conditioned on survival to the first year of life.

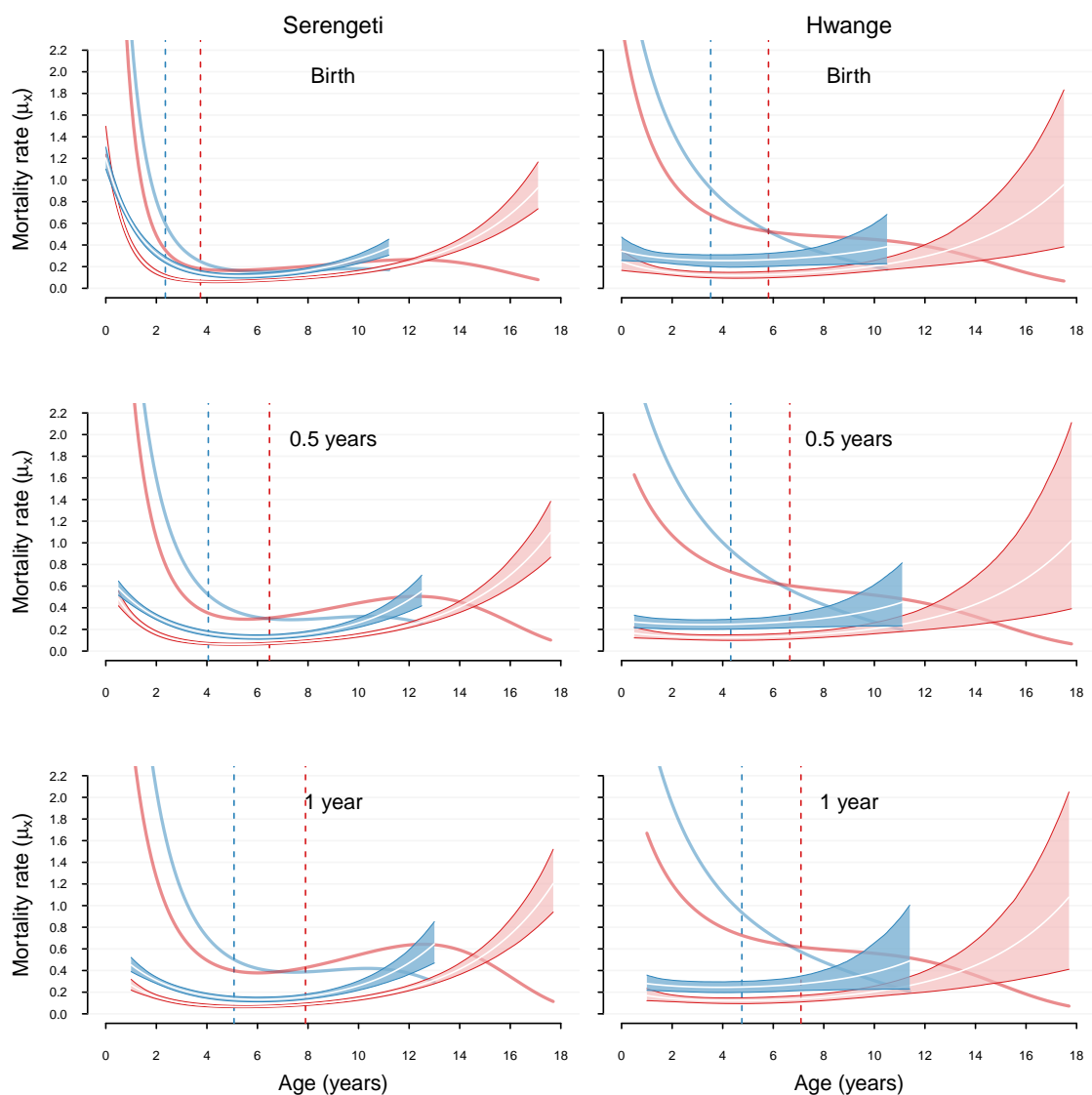


Figure 3: Age-specific mortality estimates for male (blue) and female African lions (pink) of the Serengeti population (left panels) and the Hwange population (right panels). Polygons represent 95 % credible intervals of age-specific mortality rates with white lines indicating the mean. Solid lines indicate the probability density function (PDF) for age at death, scaled so that the areas under the curves are equal and multiplied with a scaling factor of 100 to improve visibility. The dashed lines indicate life expectancy at birth. Mortality rates and PDFs are plotted until the ages when 95 % of a synthetic same-sex cohort would be dead. The first row of panels shows results of the model fitted from birth. The second and third row show results of models fitted to individuals that died or disappeared at ages older than 0.5 and 1 year, respectively.