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1 **Addressing the impact of canine distemper spreading on an isolated**
2 **tiger population in northeast Asia**

3 **Running title: PVA of Amur tiger population**

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26 **Abstract**

27 The continuation of the isolated Amur tiger (*Panthera tigris altaica*) population
28 living on the China-Russia border is facing serious challenges due to factors such as its
29 small size (including 38 individuals) and canine distemper virus (CDV). To assess
30 options to reduce the impact of these negative factors, we use a population viability
31 analysis (PVA) metamodel, which consists of a traditional individual-based
32 demographic model linked to an epidemiological model, to assess options for
33 controlling the impact of negative factors through domestic dog management in
34 protected areas, increasing connectivity to the neighboring large population (including
35 more than 400 individuals) and habitat expansion. Without intervention, under
36 inbreeding depression of 3.14, 6.29 and 12.26 lethal equivalents, our metamodel
37 predicted the extinction within 100 years is 64.4%, 90.6% and 99.8%, respectively. In
38 addition, the simulation results showed that dog management or habitat expansion
39 independently will not ensure tiger population viability for the next 100 years, and
40 connectivity to the neighboring population would only keep the population size from
41 rapidly declining. However, when the above three conservation scenarios are combined,
42 even at the highest level of 12.26 lethal equivalents inbreeding depression, population
43 size will not decline, the probability of extinction will be < 5.8%. Our findings highlight
44 that protecting the Amur tiger necessitates a multifaceted synergistic effort. Our key
45 management recommendations for this population underline the importance of reducing
46 CDV threats and expanding tiger occupancy to its former range in China, but
47 re-establishing habitat connectivity to the neighboring population is an important
48 long-term objective.

49

50 **Keywords:** Amur tiger; Canine distemper virus; Habitat connectivity; Metamodel;
51 Population viability analysis (PVA)

52 **1 INTRODUCTION**

53 The tiger (*Panthera tigris*), a flagship species in Asia, has declined to fewer than
54 5,000 individuals remaining in less than 7% of the surface of its historical range (Joshi
55 *et al.* 2016; Jhala *et al.* 2021). Among the five extant subspecies, the Amur or Siberian
56 tiger (*P. t. altaica*) is the northernmost (Wang *et al.* 2018). It suffered a severe decline in
57 the mid-20th century, and currently has a low level of genetic diversity (Henry *et al.*
58 2009; Dou *et al.* 2016). Fewer than 550 wild Amur tigers exist; they are distributed in
59 two populations separated by urban conglomerates and highways (Miquelle *et al.* 2007;
60 Wang *et al.* 2018). The larger population is spread widely in the Sikhote-Alin mountains,
61 Russia, and the smaller isolated Changbai-Primorye population (< 40 tigers) is restricted
62 to the China-Russia border (Hebblewhite *et al.* 2014; Feng *et al.* 2017). The
63 Changbai-Primorye population represents the main source for recovery of the Amur
64 tiger in much of its former range in northeast China (Miquelle *et al.* 2010; Wang *et al.*
65 2016; Qi *et al.* 2021).

66 Unfortunately, this population is currently facing multiple threats, including
67 insufficient prey resources (Wang *et al.* 2018), fragmentation and degradation of habitat
68 (Hebblewhite *et al.* 2014; Wang *et al.* 2016), poaching (Robinson *et al.* 2015), and
69 inbreeding depression (Henry *et al.* 2009; Ning *et al.* 2021). Furthermore, recent studies
70 have shown that tiger populations are facing an emerging threat from canine distemper
71 virus (CDV). CDV is a nearly globally distributed RNA virus (Deem *et al.* 2000;
72 Adhikari *et al.* 2020) that has caused significant decline in carnivore populations such
73 as Serengeti lions (*P. leo*) (Roelke-Parker *et al.* 1996; Weckworth *et al.* 2020) and
74 Ethiopian wolves (*Canis simensis*) (Gordon *et al.* 2015). As early as 2000, CDV was
75 reported in the Sikhote-Alin Amur tiger population (Quigley *et al.* 2010; Gilbert *et al.*
76 2015), and CDV was a factor responsible for a rapid decline of the Sikhote-Alin
77 population (Miquelle *et al.* 2015). Although not yet detected in the Changbai-Primorye
78 population, studies have shown that CDV has been spreading within sympatric
79 population of Amur leopard (*P. pardus orientalis*) (Sulikhan *et al.* 2018). The

80 consequences of CDV spread to Changbai-Primorye Amur tiger population could be
81 devastating. The multi-host nature of CDV poses a great threat to felid populations that
82 coexist with other hosts. For the Sikhote-Alin and Changbai-Primorye tiger populations,
83 free roaming domestic dogs that are occasionally killed by tigers (Sugimoto *et al.* 2016;
84 Dou *et al.* 2019) constitutes a potential reservoir of infection for tigers (Craft *et al.*
85 2009), who provide 3-6% of the biomass consumed of tigers (Gu *et al.* 2018; Dou *et al.*
86 2019). Few studies have assessed the impact of CDV on a critically small endangered
87 tiger population.

88 Currently, China and Russia are cooperating to conserve this population through
89 joint monitoring (Feng *et al.* 2017). Additionally, in 2016, the Chinese government
90 established a National Park, that expanded three Amur tiger and leopard reserves in
91 China to create a ~15,000 km² Northeast Tiger Leopard National Park. This expansion
92 is linked to the Land of Leopard Reserve, Russia, which has formed an 18,000 km²
93 landscape for large felids. However, human development has effectively blocked
94 continuous genetic exchange between Changbai-Primorye and the Sikhote-Alin
95 population, increasing the risk of inbreeding depression in the Changbai-Primorye
96 population (Henry *et al.* 2009; Sorokin *et al.* 2016).

97 Given the isolation and low genetic diversity of this population and the specific
98 threats of CDV, a population viability analysis (PVA) is urgently needed for the
99 Changbai-Primorye tiger population. Previous PVAs of the Amur tiger assessed the
100 impact of habitat connectivity and fragmentation (Carroll & Miquelle 2006; Tian *et al.*
101 2011), and the influence of CDV on the larger Sikhote-Alin population (Gilbert *et al.*
102 2020). However, when CDV and other wildlife diseases are incorporated in PVA models
103 they are usually treated as a general random effect and operate at the same time scale as
104 the PVA. Epidemic disease such as CDV have dynamics that are often strongly related
105 to the specific host populations and they operate at a rapid temporal scale not captured
106 in a typical PVA analysis using an annual time step (Smith *et al.* 2009; Shoemaker *et al.*

107 2014). Therefore, in this study, we modeled the viability of the Changbai-Primorye tiger
108 population combining a traditional PVA (Lacy & Miller 2020) with a separate CDV
109 epidemiological model (Lacy *et al.* 2020). This metamodel approach is commonly used
110 to simulate the spread of epidemic diseases within wild populations (Bradshaw *et al.*
111 2012).

112 Here, we assess the probability of the persistence of the Changbai-Primorye
113 population for the next 100 years. We (i) use a metamodel modelling approach to
114 determine the impact of CDV infection, lack of connectivity and restricted habitat
115 which threaten the long-term viability of this population and (ii) assess the most
116 effective combinations of management strategies that increase CDV infection control,
117 expand habitat into its former range in China and establish population connectivity with
118 the larger Sikhote-Alin population. Our results provide guidance for conservation
119 policies to enhance the viability of the Changbai-Primorye population.

120 **2 MATERIALS AND METHODS**

121 **2.1 Study area**

122 The smaller Changbai-Primorye tiger population occurs in the Changbai Mountains
123 in the Jilin and Heilongjiang provinces of China and in the Land of Leopards National
124 Park in southwestern Primorye Province (Figure 1). The Changbai Mountains are one of
125 the highest priority Conservation Areas in China. This landscape consists of a large
126 network of habitat patches, which are connected with the Land of Leopards National
127 Park in Russia (Hebblewhite *et al.* 2012). In 2015, based on long-term monitoring, we
128 estimated the Changbai-Primorye population contained at least 38 tigers (13 adult
129 females, 10 adult males, 3 unidentified adults, and 12 sub-adults/cubs) occupying 9,000
130 km² along the China-Russia border (Feng *et al.* 2017). The elevations of the rugged
131 landscape ranges from 5m to 1477 m and has a temperate continental monsoon climate
132 that supports a temperate coniferous broad-leaved mixed forest. Due to long-term
133 deforestation, many low-elevation forests in these areas have been transformed into

134 secondary deciduous forests (Wang *et al.* 2016). The main prey species of Amur tigers
135 in this area are wild boar (*Sus scrofa*), sika deer (*Cervus nippon*), Siberian roe deer
136 (*Capreolus pygargus*), and domesticated species, such as cows and dogs (Kerley *et al.*
137 2015). The area has been subjected to cattle grazing, ginseng planting, frog farming, and
138 edible fern collection for decades, particularly on the Chinese side (Feng *et al.* 2021).

139 **2.2 Modelling overview**

140 We developed a metamodel using two spatially implicit submodels to assess the
141 extinction risk of the tiger population (Figure 2): an age- and sex-structured,
142 individual-based stochastic demographic model based on Vortex software (version
143 10.5.0) (Lacy & Miller 2020) and an individual-based epidemiological model that
144 simulates epizootics such as CDV built with Outbreak software (version 2.11.0) (Lacy
145 *et al.* 2020). The two submodels were connected by the Metamodel Manager (version
146 1.0.6) (Pollak & Lacy 2020), which transfers information back and forth between
147 submodels as “open data” that are updated by each before being passed on. All
148 scenarios were projected for 100 years, with 500 model runs, each one with a specific
149 draw of parameter values from its probability distributions or specified; the results of
150 population size and genetic diversity were averaged across the runs. The population
151 survival probability was calculated as the percentage of simulations ending with
152 population survival.

153 **2.2.1 Tiger demographic model**

154 We used Vortex to build a demographic model for the tiger population with one
155 year as a time step. Cubs and juveniles do not maintain their own territories (Tian *et al.*
156 2011) and male territories overlap with those of 1-3 females, while the overlap between
157 females is relatively low (Hernandez-Blanco *et al.* 2015; Xiao *et al.* 2016). Thus, in this
158 study, the habitat carrying capacity was measured in terms of female tiger home ranges;
159 the initial habitat of 9,000 km² can accommodate approximately 23 females with an
160 average home range of approximately 400 km² (Hernandez-Blanco *et al.* 2015). The

161 gestation period of the Amur tiger is 95-100 days, and the time of caring for the cubs is
162 approximately 18.8 months (Kerley *et al.* 2003). So, we used 2 years both for time that
163 young were dependent on their mother and the interbirth interval. If a female tiger died,
164 all of her currently dependent offspring also died. For the mortality rates of the Amur
165 tiger, we refer to mortality data from 1992 to 2012 in the Sikhote-Alin Mountains,
166 Russia (Robinson *et al.* 2015). Poaching has been a severe threat to wild tiger survival
167 (Kenney *et al.* 1995), but because of the Changbai-Primorye population's strong
168 protection and transboundary cooperation, we used a mortality rate of subadults and
169 adults as the sum of the natural mortality rate (5.1%) and 50% of the Sikhote-Alin
170 population's poaching death rate (10.3%) (Robinson *et al.* 2015). In addition, inbreeding
171 depression has been confirmed to exist in wild tiger populations (Smith & Mcdougal
172 1991), and the Amur tiger population in China has reached a moderate level of
173 inbreeding (Ning *et al.* 2021), but accurate lethal equivalents to measure the severity of
174 inbreeding depression have not been available. So we conservatively set the base line
175 lethal equivalents to 3.14 (Ralls *et al.* 1988) and then examine the impact of higher
176 lethal equivalents rated (6.29 and 12.26) more recently estimated by O'Grady *et al.*
177 (2006). Other model parameters are from previously reported tiger studies (see Table 1,
178 Appendix S1 for the details and main parameter input).

179 **2.2.2 Epidemiological modelling**

180 We developed an individual-based epidemiological model with one day as a time
181 step using Outbreak, to simulate CDV transmission in the tiger population in the
182 low-risk outbreak year. The model incorporates complex processes such as
183 interindividual transmission, pathogen environmental transmission, incubation period,
184 infection period and infection outcome. In the model, there are two main ways for tigers
185 to become infected with CDV: (i) contact with environmental disease sources, such as
186 predation on small and medium-sized wild predators and dogs carrying the virus or (ii)
187 intraspecific transmission through social contact. According to Gilbert *et al.* (2014), the

188 probability of CDV infection of Amur tigers by feeding on infected domestic dogs or
189 small carnivores is 1.4%, and the average daily infection probability is 0.003836%
190 (Appendix S2: Table S1). We set the probability of infection in the interaction process
191 as 1.4% of the cumulative prevalence rate of CDV within one year (see Table 2,
192 Appendix S2 for the details and main parameter input). Furthermore, to simulate CDV's
193 cyclical high-risk prevalence (Roscoe 1993), we applied 5 years of cyclically high
194 infection risk to the model (Gilbert *et al.* 2014). In the high risk outbreak year, we set an
195 additional mortality rate of 11.48% in the "catastrophes" module of Vortex (Table 1).

196 **2.3 Sensitivity analysis**

197 We developed a set of scenarios to assess how uncertainty about parameter values
198 affect metamodel outcomes. We tested 9 key parameters in the sensitivity analysis. For
199 lethal equivalents (LEs), in addition to a base line rate of 3.14, we examined values of
200 6.29 and 12.26 estimated through meta-analyses of multiple wild species (O'Grady *et al.*
201 2006; Kenney *et al.* 2014). To evaluate the impact of temporal high-risk cycles of CDV
202 infection (CCI) we used cycles of 3, 7 and no high infection cycle. For poaching
203 intensity (PI) we tested poaching mortality rates of 0 and 10.3% as reported by
204 Robinson *et al.* (2015) (see Appendix S1 for the details). For the other 6 parameters,
205 mortality rate after CDV infection (MCI), successfully breeding female proportion
206 (BFP), infant-cub (0-1 year old) mortality rate (IMR), adult (>3 years old) female
207 mortality rate (FMR), adult (>3 years old) male mortality rate (MMR), and carrying
208 capacity (K), we increased or decreased the baseline value by 40% (He *et al.* 2020). A
209 total of 19 scenarios were created for this analysis. We varied each key parameter while
210 holding all other parameters to baseline values. The outputs of each sensitive scenario
211 were compared to the baseline scenario: mean stochastic population growth rate (R),
212 mean population number at the 100th year (N), population genetic diversity at the 100th
213 year (GD), probability of extinction (PE), and mean time of extinction (TE). The
214 sensitivity index (Pulliam *et al.* 1992) of the mean stochastic population growth rate for

215 each simulation parameter was calculated as follows:

$$216 \quad S_R = | (\Delta R / R) / (\Delta P / P) |$$

217 $\Delta R/R$ represents the variable ratio of the mean stochastic population growth rate over
218 100 years; $\Delta P/P$ represents the variable ratio of the parameters. The sensitivity index is
219 proportional to the influence of the parameters on the output of the model.

220 **2.4 Tests of alternative management actions**

221 Based on the recent threats faced by the Amur tiger, we developed three scenarios
222 to assess how different management actions and their combinations affect the
223 probability of extinction of the Changbai-Primorye population (Appendix S4: Table S1
224 and Appendix S3).

225 *Scenario A: CDV control*

226 For Scenario A, we reduced the risk of CDV infection in Amur tigers by managing
227 free-ranging domestic dogs (e.g., vaccination, hereafter referred to as “dog control”).
228 The probability of the tiger being infected with CDV due to predation on dogs was
229 reduced to 0, the probability of CDV transmission between individuals in low-risk years
230 was 0.99%, the daily average pathogenic environmental infection probability was
231 0.002712%, and the mortality rate of CDV infection in high-risk years was 83.8%.
232 These parameters were derived from published studies on the influence of CDV on
233 Amur tigers (Gilbert *et al.* 2014, 2015).

234 *Scenario B: habitat connectivity*

235 For Scenario B, we constructed a larger population (the initial population contains
236 454 individuals and the habitat can hold 502 adult female tigers) in the model, based on
237 the population characteristics of the neighboring larger Sikhote-Alin population. Then
238 we established habitat connectivity between large and small populations by exchanging
239 individuals between populations. According to the research of Henry *et al.* (2009),
240 before the complete separation of the two populations, the probability of individuals in
241 the large population dispersing to the small population is 0.22%, and the probability of

242 individuals in the small population dispersing to the large population is 1.3%. In the
243 model we use this pair of dispersion rates.

244 *Scenario C: habitat expansion*

245 Scenario C was divided into two options. For the first option scenario C45, tiger
246 habitat increased to 18,000 km² within 20 years, which could accommodate 45 female
247 tigers, based on the plan of Wang *et al.* (2018) that designated protected habitat
248 becomes occupied. For the second option scenario C100, tiger habitat increased to
249 40,000 km² in 30 years and can accommodate 100 female tigers, based on Hebblewhite
250 *et al.*'s (2012) estimate of potential recoverable core habitat in China (see Appendix S3).

251 **2.5 Inbreeding depression**

252 To avoid underestimating the degree of inbreeding depression in wildlife
253 populations, in addition to the inbreeding depression test for 3.14LEs, we also examined
254 the impact of 6.29 LEs based on Kenney *et al.* (2014) and 12.26 LEs based on O'Grady
255 *et al.* (2006), to assess population trends under different cases of inbreeding depression.
256 We set the absence of any management alternatives as the baseline for each inbreeding
257 depression scenario. In the simulation of 3.14 LEs, the inbreeding only affected the
258 first-year survival. In the simulation of 6.29 LEs, 3.94 LEs were used to impact
259 fecundity and 2.35 LEs to reduce first-year survival. In the simulation of 12.26 LEs, we
260 added 5.97 LEs for altering survival from age 1 to sexual maturity based on 6.29 LEs.

261 **3 RESULTS**

262 **3.1 Baseline scenario**

263 The baseline scenario (no management actions) predicted that under inbreeding
264 depression of 3.14 LEs, the small population would decrease at a mean rate of 0.018 per
265 year, and would have a 64.4% extinction probability within 100 years. The mean time of
266 extinction was 66 years, the mean population size was 7 individuals and the population
267 genetic diversity would decline to 57.0% of its original level (Table 3).

268 **3.2 Sensitivity analysis**

269 The results showed that the three parameters with the highest sensitivity indices
270 were BFP, MCI and IMR (Table 3, Figure S1 in Appendix S4). The effect of different
271 parameter variations on the simulation were as follows: for LEs of 6.29 and 12.26, the
272 extinction probability at 100 years was 90.6% and 99.8%, respectively and mean
273 population size declined to 1 and 0, respectively (Table 3). For CCI, the longer the
274 interval between high infection rates, the slower the population declines. However,
275 when the infectious cycle increased to 7 years, the population extinction probability was
276 still high (43%), but when cycle of CDV high infection risk was removed, the
277 probability of extinction dropped to 0, and the population slowly grew from an initial
278 size of 38 to 58 individuals after 100 years. For PI, when the mortality caused by
279 poaching was set at 0% and 10.3%, the extinction probability was 1.8% and 99.8%,
280 respectively. When poaching was fully eradicated, the population grew slowly, reaching
281 a size of 56 after 100 years. The population exhibited a growing trend when MCI, IMR,
282 and FMR were reduced by 40% and BFP was increased by 40%. For MMR and K, their
283 increase or decrease had little influence on the population (Table 3).

284 **3.2 Tests of alternative management actions**

285 Just reducing domestic dogs (scenario A), even in the mildest inbreeding depression
286 category (3.14 LEs), resulted in a negative growth rate (-0.006), the population
287 extinction probability was 40.6%, and the population size declined to 16 within the 100
288 years. In the other two more severe inbreeding depression cases, the population is at a
289 high risk of extinction, with extinction probability above 75% (Figure 3 and Table S2 in
290 Appendix S4).

291 Increasing habitat connectivity (scenario B), with inbreeding depression at 3.14 and
292 6.29 LEs, resulted in a mean stochastic growth rate was 0.013 and 0.008, the mean
293 population genetic diversity all declined to 89%, the population size reached 43 and 37,
294 and the extinction probability within 100 years dropped to 1.4% and 4.2%, respectively.
295 However, in the case of 12.26 LEs, the mean stochastic growth rate was 0 and the

296 extinction probability was 14.4% (Figure 3 and Table S2 in Appendix S4).

297 Expanding the tiger's range (scenario C) alone was inadequate. For both schemes
298 (C45, C100), the simulation results were essentially the same as the baseline under the
299 different inbreeding depression, except for a slight difference in population size in the
300 100th year. The population remained at high risk of extinction (Figure 3 and Table S2 in
301 Appendix S4).

302 In summary, as the value of LEs increases, the benefits of implementing just one of
303 the scenarios was largely ineffective. Habitat connectivity provided the biggest benefit.
304 Implementing both habitat expansion and reducing the number of dogs in the forest
305 have smaller benefits, but increasing habitat is the least beneficial (Figure 3). We chose
306 the expansion scheme of 45 females (C45) as the representative of the habitat expansion
307 scenario and simulated four combinations of the three scenarios overlapping each other.
308 We found that the benefits of the four combinations were much higher than those of
309 single management action (Appendix S4: Table S2, Table S3), regardless of the LEs
310 (Figure 4). In the combination of scenario A and C at three levels of LEs, the population
311 size reached 44, 13 and 0 within the 100th year, the genetic diversity was 71.3%, 66.7%,
312 0.00%, the mean stochastic growth rate was -0.002, -0.018, and -0.049, and the
313 population extinction probability was 26.4%, 57.8%, and 99.8%, respectively; in the
314 combination of scenarios B and C with the 3 possible levels of LEs, the population size
315 was 73, 56, and 11, and the genetic diversity was 90.5%, 89.7%, and 88.4%,
316 respectively, the mean stochastic growth rate was 0.008, 0.003, and -0.016, and the
317 extinction probability was 2.2%, 3.4%, and 44.2% within 100 years, respectively; in the
318 combination of scenario A and B of three levels LE, the population size was 59, 56, 32,
319 the genetic diversity in the 100th year was 93.3%, 93.3%, and 90.0%, and the mean
320 stochastic growth rate was 0.030, 0.024, and 0.004, respectively. And there was no risk
321 of population extinction within 100 years in the 3.14 and 6.29 LEs but the extinction
322 probability with 12.26 LEs was 5.6%; when the three actions were combined, the

323 population size was 110, 97, 41, genetic diversity in the 100th year was 93.8%, 93.8%,
324 91.1%, and the mean stochastic growth rate was 0.021, 0.016, and 0.005, respectively.
325 There was no risk of extinction in the 3.14 and 6.29 LEs cases, but the extinction
326 probability with 12.26 LEs was 5.6% (Appendix S4: Table S3).

327 **4 DISCUSSIONS**

328 By incorporating a metamodel, we were better able to capture the potential
329 dynamics of the rapid spread of CDV within the Changbai-Primorye tiger population
330 and thus obtain a more realistic prediction of the impact of CDV on tigers. Compared
331 with previous PVA tiger studies (Carroll & Miquelle 2006; Tian *et al.* 2011), the
332 contributions of our study are as follows: (i) the metamodel allowed us to incorporate
333 the detailed, rapid dynamics of CDV into a tiger PVA, (ii) we incorporated recent
334 estimates of inbreeding depression that may be more realistic than those used in
335 previous models, and (iii) under these different levels of inbreeding depression and a
336 realistic model of CDV, we were able to explore management options for recovery of
337 the Changbai-Primorye tiger population. Through the simulation of different
338 management measures and their combinations, our results clearly show the importance
339 of habitat connectivity.

340 **4.1 Insights from sensitivity assessment**

341 Our results highlighted that with an outbreak of CDV, this population may require a
342 combination of management strategies to insure its persistence. The extinction
343 probability was greater than 60% under all realistic baseline estimations of lethal
344 equivalents without any intervention. Sensitivity analysis examined the impact of
345 variation in key demographic parameters on the model output results. Variation in the
346 proportion of breeding females was the most sensitive parameter in our sensitivity
347 analysis but based on observations in Nepal and Thailand (James L.D. Smith pers.
348 comm.) there is actually little variation in proportion of females that get pregnant. In
349 contrast, variation in female and infant-juvenile mortality does occur in wild

350 populations and can have a strong negative impact of recruitment. In tigers and other
351 large felids, female survival is a major determinant of population viability (Kenney et al.
352 2014). This finding is consistent with PVA studies of other mammals such as the fennec
353 fox (*Vulpes zerda*), the Asian elephant (*Elephas maximus*) (He et al. 2020; Franklin et al.
354 2021). Therefore, it is critical to ensure female tigers have good breeding habitat and
355 low human-caused mortality to help increase their survival rates (Miquelle et al. 2015).
356 Sensitivity analysis also highlighted that poaching intensity, cycles of CDV high
357 infection risk and mortality rate after CDV infection all had critical impacts on the
358 probability of extinction. Without the threat of poaching and CDV, the current
359 population would most likely remain viable for the next 100 years.

360 The sensitivity analysis showed that changes in environmental carrying capacity did
361 not reduce the probability of extinction. This could be because the Changbai-Primorye
362 population has been kept below the environmental capacity for long periods of time due
363 to multiple threats (Robinson et al. 2015; Wang et al. 2018; Ning et al. 2021), and the
364 influence of inbreeding depression did not decline with habitat expansion. Even if with
365 a large enough habitat, it is difficult for the population to spread to any additional
366 habitat due to the high extinction risk within the population.

367 Given the difficulty in estimating rates of inbreeding depression in the wild and
368 recent debate as to the extent to which purging reduces inbreeding depression
369 (Armstrong et al. 2021), further research is needed to estimate current levels of
370 inbreeding depression in tigers and other species. Alternatively, we may have
371 over-estimated mortality of adult females in our model and that is restricting recovery.
372 Kenney et al. (2014) estimated a much higher annual rate of adult female survival than
373 we used in our model. Future modeling of this population will benefit from efforts to
374 improve estimation of this parameter through our current intensive camera trapping.
375 Simultaneously, it is critical to monitor CDV in both tigers and domestic species.

376 **4.2 Management interventions**

377 With the newly established the Northeast Tiger and Leopard National Park in China,
378 tigers are now expanding their range into northeast China. However, the distribution
379 further from the Russian border is still spotty and the number of tigers is low (Wang *et*
380 *al.* 2016; Qi *et al.* 2021). Most of the habitat expansion in China is by young males.
381 Female tigers and their young are confined to the China-Russia border and breeding
382 habitat is still estimated at < 6,000 km² (Wang *et al.* 2018; Qi *et al.* 2021). At this
383 juncture, urgent actions are needed.

384 Our study revealed that maintaining connectivity among the two populations was
385 essential and a prerequisite for preventing drastic reduction of population. The
386 simulation of different management scenarios showed that habitat connectivity was the
387 only measure that effectively increased growth rates in the tiger population. Moreover,
388 the scenarios combining habitat connectivity performed relatively well in population
389 projections. In contrast, dog control, habitat expansion and this combination resulted in
390 the population reaching its maximum size in less than 40 years and then gradually
391 declining. It is thus clear that gene exchange among populations plays a key role in the
392 recovery and long-term population persistence.

393 A true barrier now exists due to continuing development along the Razdolnaya
394 River basin, which prevents movements of tigers between the Changbai landscape and
395 the southern Sikhote-Alin landscape; the populations show clear genetic differentiation
396 (Sorokin *et al.* 2016). Furthermore, a recent study reported poor health status for the
397 Changbai-Primorye population as well as 50% of individual relationships were cousins
398 or half-sibs (Ning *et al.* 2021). Miquelle *et al.* (2015) used least-cost distance analyses
399 to identify a single potential corridor to retain connectivity for the two subpopulations,
400 but no evidence demonstrates that this corridor is actually used by tigers. When habitat
401 connectivity is difficult to achieve, a more immediate, and practical strategy, is
402 translocation of individuals from the Sikhote-Alin population. A similar movement of
403 Texas panthers into the Florida panther population (*P. c. coryi*) resulted in a dramatic

404 population recovery (Johnson *et al.* 2010; Hostetler *et al.* 2013). Similarly, the
405 reintroduction of tigers in Panna, India, resulted in rapid population recovery (Sankar *et*
406 *al.* 2010).

407 CDV infection is regarded as the main threat to large felids conservation not only in
408 Asia but also worldwide (Adhikari *et al.* 2020). Domestic dogs are a proven source of
409 CDV for wild animals and transmit the CDV to the tiger population through direct
410 predation by tigers or indirectly through interaction with other wildlife, which can
411 increase the mortality risk of tigers (Gilbert *et al.* 2014; Dou *et al.* 2019). However,
412 currently, domestic dogs have widely invaded into the Amur tiger's habitat (Gilbert *et al.*
413 2020). According to our camera-trapping data and field surveys, domestic dogs are still
414 entering the parkland and committed wildlife attacks in 2021; on the Chinese side,
415 unvaccinated domestic dogs had a high level of positive antibodies for CDV due to
416 exposure to the disease (37% in 202 domestic dogs serum tests in 2018, unpublished
417 data). Although dog management practices were not the most effective when compared
418 to habitat connectivity, and domestic dogs may not be the only source of CDV infection
419 (Gilbert *et al.* 2020), dog management was the easiest and safest method to achieve
420 among all conservation actions, and when combined with other measures, it can
421 effectively improve population trajectories by increasing the population recovery speed.
422 Given CDV is preventable, strict vaccination of domestic dogs with the commercial
423 attenuated vaccine is needed. To ensure the long-term success of tiger conservation, we
424 suggest that the local government implement policies aimed at gradually controlling
425 dogs, if not completely prohibiting all dogs in the tiger's core range.

426 Although the establishment of national park has expanded the habitat of Amur
427 tigers, through the monitoring of camera trap data in the past few years, the core range
428 of Amur tigers still stays at the Sino-Russian border and does not spread to China on a
429 large scale. As the results in our simulations show, the expansion of habitat did not
430 significantly increase the population size, especially in the case of severe inbreeding

431 depression. This may be due to the fact that the Amur tiger population face many threats,
432 such as low initial population size, inbreeding depression and CDV, which cause the
433 population size to remain below the carrying capacity for a long time. Even with
434 sufficiently large habitats, it is difficult for tiger population to spread due to the high
435 risk of extinction. The same result was obtained when assessing the extinction risk of
436 tigers in central India (Thatte *et al.* 2018). Although our model did not indicate that
437 habitat expansion alone would reduce the probability of extinction, habitat expansion
438 has several benefits that should not be overlooked. Habitat expansion would increase
439 structurally and spatial diversity of the landscape, which in turn might decrease the
440 spread of CDV. Also shifting from people intensive logging activities to conservation
441 management may reduce the impact of humans and their dogs serving as a vector for
442 CDV. In India, a topographically and vegetatively diverse landscape also reduced the
443 risk of extinction (Thatte *et al.* 2018).

444 In summary, our findings highlight the critical need for integrated conservation
445 strategies and actions that link wildlife populations, landscape planning, CDV research,
446 and international cooperation while also meeting human development goals. Our
447 modeling of the potential impact of CDV highlights the critical need for integrated
448 conservation strategies that link and expand wildlife populations, institute broad
449 landscape planning, and include continuous surveys of potential CDV outbreaks.
450 Fortunately, transborder cooperation in monitoring and management is already in place.

451 **4.3 Methodological considerations**

452 Metamodelling requires estimating a number of parameters to build each
453 submodel with an acceptable degree of realism, and hence there is a degree of
454 uncertainty associated with their results (Shoemaker *et al.* 2014). However, metamodels
455 can assist managers and policymakers in making decisions for the conservation of
456 endangered populations in an uncertain environment (Reed *et al.* 2002; Lawson *et al.*
457 2021). Therefore, this analysis provides the relative benefits of various possible actions

458 rather than an absolute and accurate prediction of future population trends. Inbreeding
459 could have an impact on the survival and reproduction of adult tigers, especially in the
460 face of changing environmental conditions (Coltman *et al.* 1999; Fox & Reed 2011),
461 and population connectivity may slightly mitigate inbreeding depression in populations.
462 However, the above two considerations were not included in the model because of the
463 difficulty of accurately estimating the relevant parameters and their limited influence on
464 the overall population trends. In addition, since prey resources are also an important
465 factor affecting tiger population continuation and management and policy decisions
466 operate on short-term projections, it is necessary to construct more complex and
467 shorter-term projections that include prey species in the future.

468

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630 **Table 1.** Parameters input for the baseline scenario of the tiger population model in Vortex.

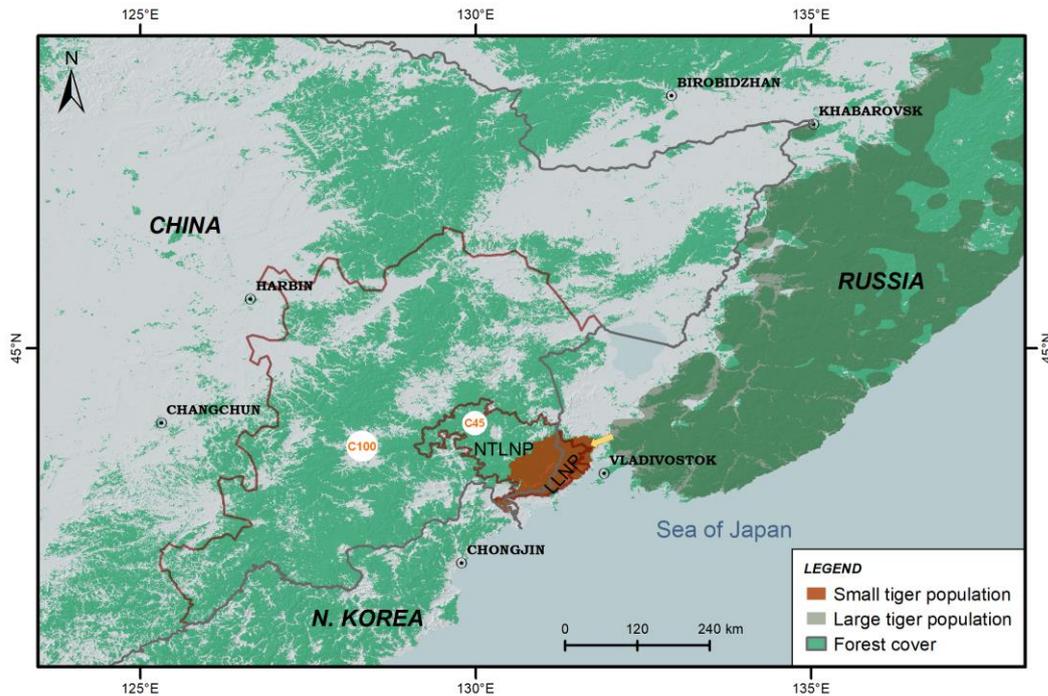
Parameter	Baseline value
Lethal equivalents	3.14
Percent due to recessive lethal alleles (%)	50
Breeding system	Polygynous
Age of first offspring (female/male)	3/4
Maximum age of reproduction	15
Maximum lifespan	15
Maximum number of broods per year	1
Maximum number of progenies per brood	6
Share of males at birth (%)	50
Cubs depend on the mother (years)	2
Successfully breeding female proportion (%)	70
Breeding male proportion (%)	70
Litter size (%)	
1 offspring	10
2 offspring	38
3 offspring	38
4 offspring	10
5 offspring	3
6 offspring	1
Mortality of infant-cubs (%)	40 (0-1)
Mortality of juveniles (%)	20 (1-2)
Mortality of subadults (%)	18.368 (2-3)
Mortality of adult females (%)	6.273 (>3)
Mortality of adult males (%)	15.539 (>3)
Initial population size	38
Number of infant-cubs (female/male)	2/2
Number of juveniles (female/male)	2/2
Number of subadults (female/male)	2/2
Number of adults (female/male)	14/12
Carrying capacity (K)	23 (adult females)
Catastrophes (high risk of CDV)	
Frequency (%)	20
Reproduction	1
Survival (%)	88.52

631 **Table 2.** Parameters input for the baseline scenario of the canine distemper epidemiological
632 model in Outbreak.

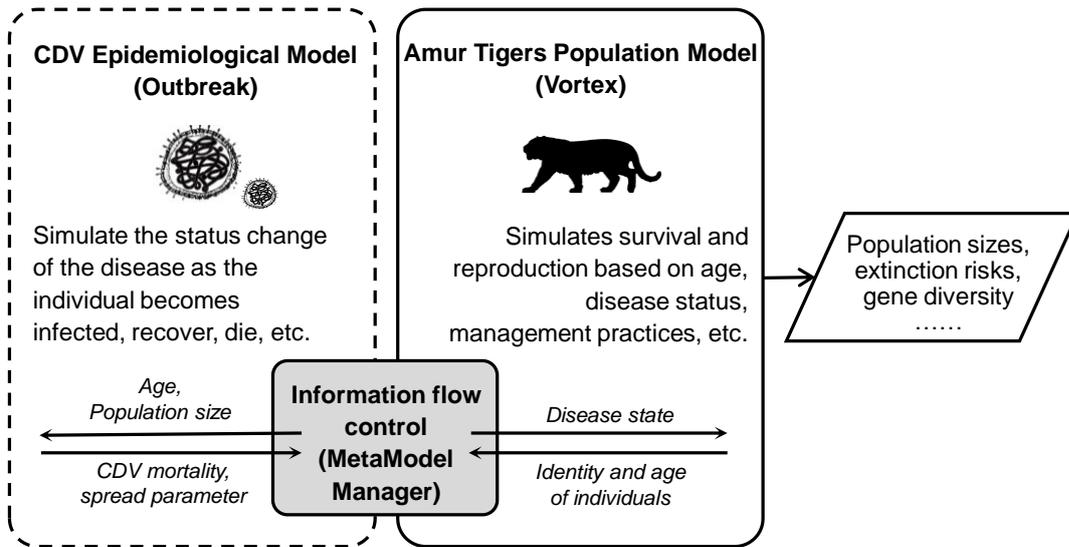
633	Parameter	Baseline value
	Probability that an individual never becomes susceptible (%)	0
	Earliest age of susceptibility (days)	15-30
	Transmission probability from an infectious mother to a newborn	1
	Time for maternally derived immunity to protect an offspring (days)	150-180
	Number of individuals that an individual encounters per day	0.03
	Transmission probability among individuals during an encounter (%)	1.40
	Daily transmission probability from the environment (%)	0.003836
	Duration of the incubation period (days)	2-7
	Duration of the infectious period (days)	30-60
	Probability of recovering and acquiring permanent immunity (%)	60
	Mortality rate after CDV infection (%)	40

634 **Table 3.** Baseline candidate model results. R(SD): mean (standard deviation) of stochastic
635 population growth rate across all years and iterations; N (SD): mean (standard deviation) of
636 number of tigers at year 100th; GD (SD): initial genetic diversity (heterozygosity) remaining
637 in population at year 100th; PE: probability of extinction, defined as only 1 sex remains at
638 year 100th; TE: mean time of extinction (in years) for all iterations that went extinct within
639 100 years; SR: The sensitivity index of the mean stochastic population growth rate for each
640 simulation parameter.

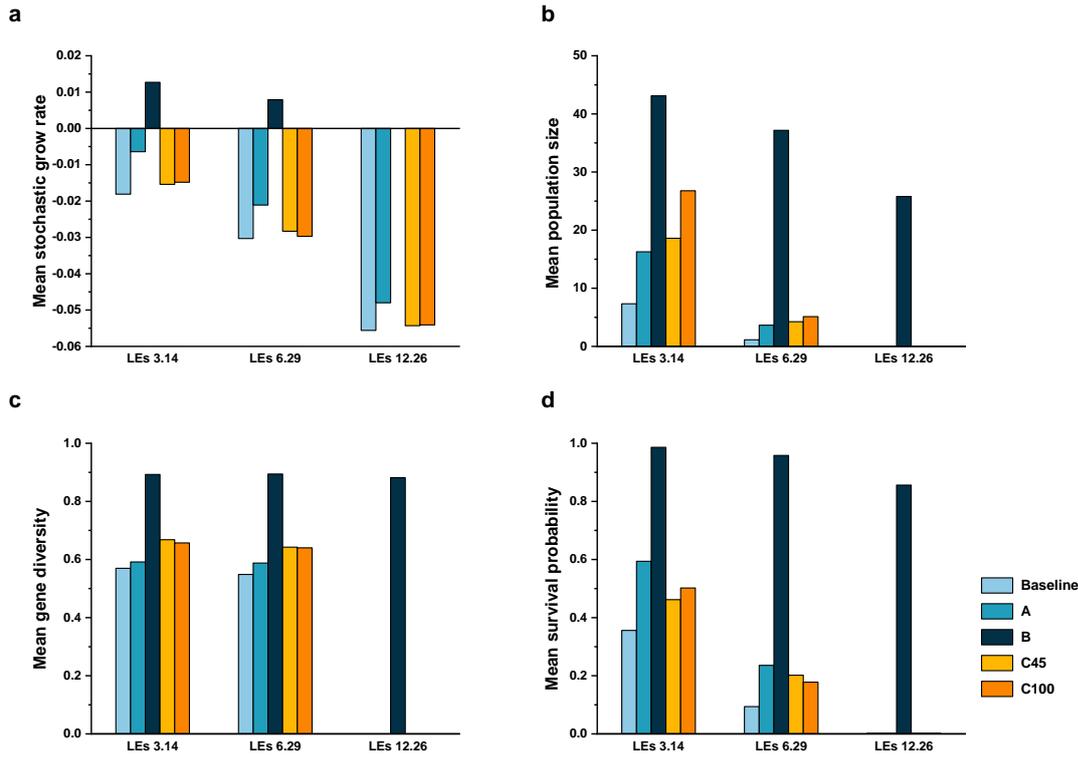
Scenario	R (SD)	N (SD)	GD (SD)	PE	TE	S _R
Baseline	-0.018 (0.186)	7(13)	0.570 (0.1700)	0.644	66	
Lethal equivalents (LEs)						0.691
6.29	-0.030 (0.192)	1 (4)	0.549 (0.189)	0.906	63	
12.26	-0.056 (0.191)	0 (0)	0 (0)	0.998	40	
The cycles of CDV high infection risk (CCI)						1.989
NA	0.057 (0.105)	58 (9)	0.738 (0.093)	0.000	-	
3	-0.037 (0.214)	1 (5)	0.493 (0.176)	0.928	55	
7	-0.008 (0.171)	14 (18)	0.585 (0.178)	0.430	70	
Poached intensity (PI)						2.798
0%	0.037 (0.125)	56 (17)	0.758 (0.092)	0.018	62	
100%	-0.064 (0.244)	0 (0)	0 (0)	0.998	36	
Mortality rate after CDV infection (MCI)						3.791
+40%	-0.042 (0.226)	0 (3)	0.454 (0.238)	0.966	51	
-40%	0.013 (0.138)	33 (20)	0.664 (0.147)	0.128	74	
Successfully breeding female proportion (BFP)						5.228
+40%	0.017 (0.151)	39 (23)	0.674 (0.150)	0.116	66	
-40%	-0.059 (0.218)	0 (0)	0 (0)	1.000	40	
Infant-cub (0-1 year old) mortality rate (IMR)						3.743
+40%	-0.043 (0.204)	0 (1)	0.482 (0.284)	0.982	52	
-40%	0.011 (0.163)	33 (24)	0.659 (0.140)	0.202	71	
Adult (>3 years old) female mortality rate (FMR)						3.342
+40%	-0.040 (0.211)	1 (3)	0.412 (0.253)	0.954	53	
-40%	0.009 (0.157)	30 (20)	0.648 (0.150)	0.168	70	
Adult (>3 years old) male mortality rate (MMR)						0.470
+40%	-0.021 (0.194)	5 (10)	0.497 (0.222)	0.732	63	
-40%	-0.014 (0.173)	11(15)	0.614 (0.183)	0.536	68	
Carrying capacity (K)						0.275
+40%	-0.015 (0.176)	14 (21)	0.624 (0.166)	0.526	65	
-40%	-0.019 (0.209)	2 (6)	0.428 (0.223)	0.844	60	



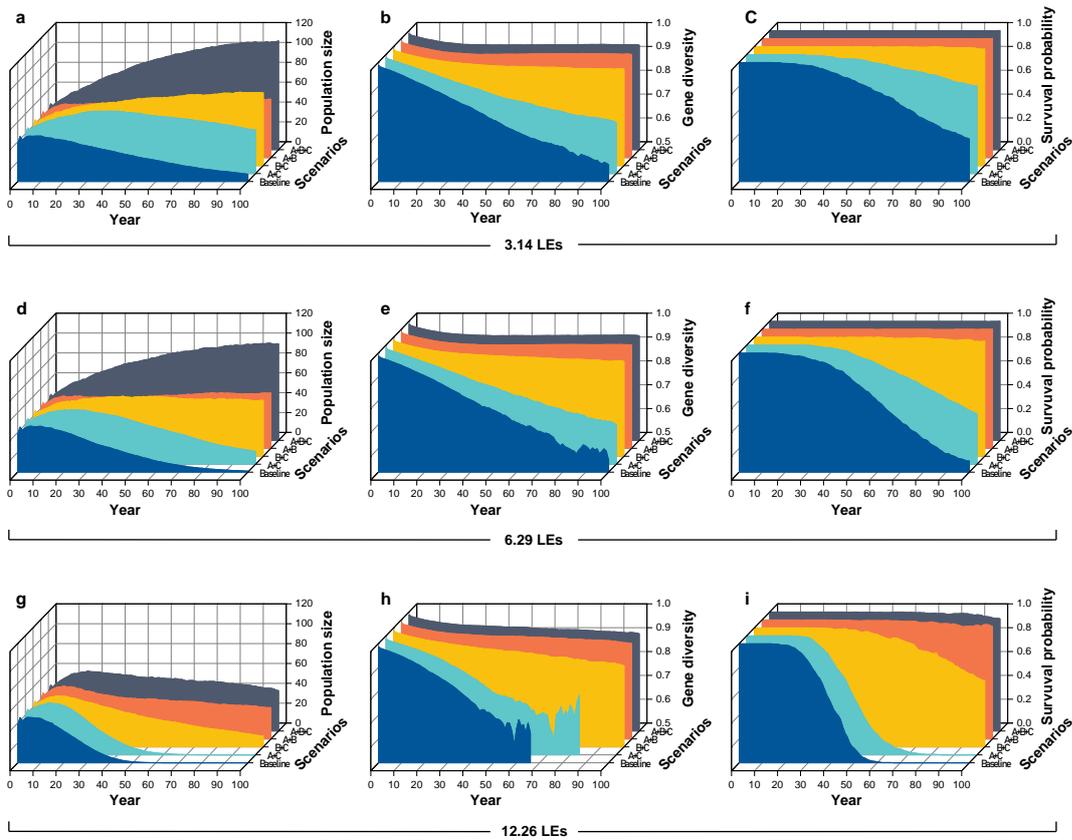
642 **Figure 1.** Current range of the Amur tiger in Northeast China and Far East Russia. Small
 643 tiger population is areas with confirmed tiger breeding activity within the last 10 years. C100
 644 and C45 indicate two habitat expansion schemes (see method for details). NTLNP and LLNP
 645 is the Northeast Tiger and Leopard National Park of China and the Land of Leopard National
 646 Park of Russia, respectively. Yellow arrow represents potential important ecological corridors
 647 for tigers between small and large populations.



649 **Figure 2.** Schematic flowchart of the Amur tiger-CDV metamodel structure used for this
 650 analysis. The PVA program acts as the system model (solid outline) to simulate individual
 651 survival and reproduction based on individual and population state variables (shown in italics)
 652 passed from other models. Epidemiological model (dashed outlines) simulates individual
 653 transitions in disease status. The central information flow control program passes state
 654 variables between the system and modifier models at appropriate time steps. Rounded
 655 rectangles represent software components, and parallelograms identify model output
 656 variables from Vortex.
 657



658 **Figure 3.** Estimates of stochastic population growth (a), population size (b), genetic diversity
659 (c) and survival probability (d) for the small Amur tiger population under different
660 management alternatives and lethal equivalents (LEs) over 100 years based on the average of
661 500 model runs. Scenario A: dog control in the Amur tiger habitat; Scenario B: establish
662 habitat connectivity between large and small populations; Scenario C45: habitat expanded to
663 accommodate 45 females; Scenario C100: habitat expanded to accommodate 100 females.
664



666 **Figure 4.** Changes in population size, genetic diversity, and survival probability of small
 667 Amur tiger population under different combinations of management actions and lethal
 668 equivalents (LEs) over 100 years based on the average of 500 model runs. Scenario A: dog
 669 control in the Amur tiger habitat; Scenario B: establish habitat connectivity between large and
 670 small populations; Scenario C: habitat expanded to accommodate 45 females.