Predicting the impact of disease epidemics on wildlife populations is one of the twenty-first century’s main conservation challenges. The long-term demographic responses of wildlife populations to epidemics and the life history and social traits modulating these responses are generally unknown, particularly for K-selected social species. Here we develop a stage-structured matrix population model to provide a long-term projection of demographic responses by a keystone social predator, the spotted hyena, to a virulent epidemic of canine distemper virus (CDV) in the Serengeti ecosystem in 1993/1994 and predict the recovery time for the population following the epidemic. Using two decades of longitudinal data from 625 known hyenas, we demonstrate that although the reduction in population size was moderate, i.e., the population showed high ecological ‘resistance’ to the novel CDV genotype present, recovery was slow. Interestingly, high-ranking females accelerated the population’s recovery, thereby lessening the impact of the epidemic on the population.
Epidemics responsible for a decline in keystone species can alter ecosystem dynamics and diminish biodiversity by increasing the chance of extirpation of host populations, and possibly the extinction of species1–3. Human activities rapidly expand the geographical range and host species spectrum of pathogens, and epidemics caused by exotic pathogens in unexpected hosts are increasing4. Viruses are of particular concern as they evolve rapidly, yielding new strains and adaptations to novel hosts5,6.

We know little about the long-term demographic responses to infectious viral disease epidemics in wildlife species, particularly those with high maternal investment in a low number of offspring during a long lifespan (K-selected species), probably because longitudinal studies on such species are rare5. To characterise these demographic responses, we apply two terms from the field of ecology—ecological resistance: the impact of exogenous disturbance on the state of a system; and recovery: the endogenous process that pulls the disturbed system back towards equilibrium7. In social host species, the relationships between social traits and disease outcomes can be complex8–10. It is therefore unclear how social traits modulate the magnitude of the potential reduction in host population size during an epidemic, or the recovery time to pre-epidemic population size11.

There is a strong need for a robust predictive framework to assess the potential risk that epidemics pose to wildlife populations, to provide projections of the recovery of populations from epidemics and to identify factors modulating population responses, particularly for K-selected and social species. Here we show that stage-structured matrix population model approaches12,13 are highly suitable for these purposes because they allow us to assess how demographic performance (in terms of fecundity, survival and reproductive value) is influenced by infection status, and how disease persistence and dynamics are influenced by host demography and social structure14,15. Such models can make full use of field data as input, i.e., state-specific parameters estimated by capture-mark-recapture (CMR) approaches16, and permit the calculation of the basic reproduction number \( R_0 \) of a pathogen14,15. \( R_0 \) is a measure of the number of individuals infected by introducing a single infected individual into a susceptible population (over the course of its infectious period)17, i.e., a measure of maximum transmissibility, and describes whether an infectious disease will invade or fade out in a population.

Here, we applied a stage-structured matrix population model to a longitudinal dataset of 625 individually known female spotted hyenas \( Crocuta crocuta \) (hereafter hyenas) in the Serengeti National Park, Tanzania, spanning two decades. We investigated the population consequences of a canine distemper virus (CDV) epidemic10,18, caused by a novel genotype better adapted to non-canids such as hyenas and lions \( Panthera leo \) than to canids19. By quantifying and conducting a sensitivity analysis for \( R_0 \) of this CDV strain for the first time in hyenas, we formally demonstrate that it was highly contagious among hyenas during the epidemic and that the most influential parameter to decrease \( R_0 \) was the probability of becoming a socially high-ranking breeder. CDV infection during the epidemic decreased the survival of cubs and subadults, particularly low-ranking ones. The projected hyena population showed a slow recovery, taking at least 16 years to return to pre-epidemic levels. Population growth rate and \( R_0 \) were particularly sensitive to changes in the demographic contribution of high-ranking females, in terms of their probabilities of recruitment, survival and maintenance of high rank. Thus, high-ranking females increased the ecological resistance of the hyena population and helped to accelerate its recovery, thereby favoring the fadeout of the CDV epidemic. To our knowledge, this is the first study in a wildlife host to demonstrate that the interaction of host demographic performance, social status and infection state drive both \( R_0 \) and the population response to a major epidemic.

Results

Demographic, social and infection parameter estimates. Each year, each female was assigned a demographic (cub, subadult, adult breeder or adult non-breeder), a social (high or low social status) and an infection (susceptible, infected or recovered) state. We fitted a multi-event CMR (MECMR) model that accounted for uncertainty on the assignment of infection states, i.e., we included animals with unknown infection states10. We focused on three periods: pre-epidemic (pre-epidem; 1990–1992), epidemic (epidem; 1993–1994), and post-epidemic (post-epidem; 1995–1999). The virulent CDV strain was not detected in any of our study clans after 199710 and it was not detected in any host in the Serengeti ecosystem after 199919.

The annual apparent survival probability (\( \phi \)) varied over time, being highest during pre-epidem and lowest during epidemic for all states, as the effect of periods was additive to the effect of given states (Table 1). The transition probability from the susceptible to the infected state (\( \beta \), the infection probability) also varied across periods; it was highest during epidemic and lowest during pre-epidem. CDV infection requires contact with aerosol droplets and body fluids from virus shedding individuals20. High-ranking subadults, breeders and non-breeders had a higher infection probability than low-ranking ones (Table 1), a consequence of their higher contact rates10,21. As described previously10,20, CDV infection decreased survival only among cubs and subadults but not adults (whether breeders or non-breeders, Table 1). Cubs of low-ranking mothers had a higher infection probability and were less likely to survive CDV infection than those of high-ranking mothers. This is thought to be a consequence of their poorer body condition and lower allocation of resources to immune processes19. High-ranking females were more likely to become breeders (\( \psi \), the probability of assuming a breeder state) than low-ranking females. High-ranking females were less likely than low-ranking ones to maintain their social status (\( r \), the probability of maintaining the current social state). Estimates of all parameters for each period, which were used as input for the subsequent stage-structured matrix population model, are shown in Table 1. Probabilities of detection and assignment of infection states were assumed to be constant throughout the study period.

Population and CDV dynamics. The population growth rate (\( \lambda \)) differed substantially between periods (Fig. 1). \( \lambda \) was highest during pre-epidem and lowest during epidemic. The hyena population increased during pre-epidem but decreased during epidemic and post-epidem (Fig. 1).

We estimated CDV \( R_0 \) in a closed hyena population, based on our three study clans, and hence did not consider the spread of the non-canid strain in other non-canid species, such as the lion, which were infected with this strain during the epidemic19. We modified a previously developed approach for discrete time models14,15. Incorporating population growth rate into the calculation produced a measure of change in the rate of spread of the disease accounting for change in population size, i.e., measuring change in the incidence rate of the disease. We then estimated this modified version of \( R_0 \) during the epidemic period. \( R_0 \) was relatively high, equal to 5.69 ± 0.67, illustrating how CDV infection spread rapidly among hyenas during epidemic.

Sensitivity analyses of \( \lambda \) and \( R_0 \). To determine which vital rates (parameters) contributed most to changes in \( \lambda \) and \( R_0 \), we conducted sensitivity analyses for \( \lambda \) during each period and for \( R_0 \).
Table 1 Input parameter values for the matrix population model

| Process                | Parameter and notation                        | m.l.e ± s.e.m |
|------------------------|------------------------------------------------|---------------|
|                        |                                                | Pre-epidem    | Epidem          | Post-epidem |
| Survival               | H susceptible cubs—φCHS                         | 0.90 ± 0.04   | 0.84 ± 0.07     | 0.86 ± 0.06 |
|                        | L susceptible cubs—φCLs                         | 0.86 ± 0.07   | 0.78 ± 0.10     | 0.81 ± 0.08 |
|                        | H infected cubs—φCHI                            | 0.79 ± 0.06   | 0.67 ± 0.07     | 0.72 ± 0.06 |
|                        | L infected cubs—φCLi                            | 0.62 ± 0.08   | 0.48 ± 0.08     | 0.54 ± 0.08 |
|                        | H susceptible subadults—φSANS                   | 0.97 ± 0.04   | 0.95 ± 0.06     | 0.96 ± 0.05 |
|                        | L susceptible subadults—φSALs                   | 0.90 ± 0.12   | 0.83 ± 0.19     | 0.86 ± 0.16 |
|                        | H and L infected and recovered subadults—φSAR   | 0.69 ± 0.04   | 0.56 ± 0.05     | 0.61 ± 0.03 |
|                        | H and L non-breeders—φNB                        | 0.86 ± 0.02   | 0.78 ± 0.03     | 0.82 ± 0.02 |
|                        | H and L breeders—ψB                             | 0.95 ± 0.01   | 0.91 ± 0.02     | 0.93 ± 0.01 |
|                        | H cubs—μCH                                      | 0.23 ± 0.09   | 0.96 ± 0.02     | 0.67 ± 0.16 |
|                        | L cubs—μCL                                      | 0.45 ± 0.16   | 0.99 ± 0.01     | 0.85 ± 0.10 |
|                        | H subadults, non-breeders and breeders—βH       | 0.10 ± 0.04   | 0.90 ± 0.05     | 0.42 ± 0.16 |
|                        | L subadults, non-breeders and breeders—βL       | 0.02 ± 0.01   | 0.66 ± 0.12     | 0.13 ± 0.08 |
| Infection              | H subadults becoming H breeders—ψSH             | 0.04 ± 0.02   |                |              |
|                        | L subadults becoming L breeders—ψSL             | 0.01 ± 0.01   |                |              |
|                        | H non-breeders becoming H breeders—ψBH          | 0.68 ± 0.02   |                |              |
|                        | L non-breeders becoming L breeders—ψBL          | 0.60 ± 0.03   |                |              |
|                        | H breeders remaining H breeders—ψBH             | 0.49 ± 0.02   |                |              |
|                        | L breeders remaining L breeders—ψBL             | 0.45 ± 0.03   |                |              |
| Social                 | H remaining H—rH                                | 0.94 ± 0.01   |                |              |
|                        | L remaining L—rL                                | 0.97 ± 0.00   |                |              |
| Other                  | Average sex ratio in the population—sex.ratio   | 0.52          |                |              |
|                        | Average litter size in the population—litter.size| 1.53         |                |              |

Demographic states were abbreviated as C, cubs; SA, subadults; B, adult breeders; or NB, adult non-breeders; social states as H, high status or L, low status and infection states as S, susceptible, I, infected or R, recovered. Note that the parameter values for demography and social were calculated across all three periods (pre-epidem, epidemic and post-epidem).

Maximum likelihood estimate and associated standard error (m.l.e ± s.e.m.) of the probabilities of surviving, becoming infected, becoming a breeder, and maintaining the current social state in female spotted hyenas during each period, as estimated via the multi-event capture-mark-recapture (MECMR) model. Sex ratio and litter size were estimated differently.

Figure 1: Population growth rate (λ) of female spotted hyenas. Data shown are for female spotted hyenas in the Serengeti National Park during the three epidemic periods (pre-epidem: 1990–1992; epidemic: 1993–1994; post-epidem: 1995–1999). Empty circles represent outliers, the boxes encompass the first to the third quartiles, inside the box the thick horizontal line shows the median and the whiskers are located at 1.5 × IQR (interquartile range) below the first quartile and at 1.5 × IQR above the third quartile.

During epidemic, for λ, during all periods, the two most influential parameters were the probabilities that high-ranking breeders and non-breeders stayed or became high-ranking breeders (Fig. 2). In all periods, the probability of staying a high-ranking female had an important positive contribution on λ (Fig. 2). For all parameters and during each period, high-ranking females contributed substantially more to λ than low-ranking ones. The survival of non-breeders had a higher contribution than that of breeders because non-breeders were more likely to assume a breeder state than breeders at the next projection interval (Table 1). If breeders and non-breeders were allocated identical probabilities of assuming a breeder state, then the survival of breeders had a higher influence on λ than that of non-breeders.

The three most influential parameters to decrease R0 during epidemic were increases in the probabilities that high-ranking breeders remained and high-ranking non-breeders became high-ranking breeders, and an increase in the probability of staying high-ranking across all demographic and infection states (Fig. 3). The most influential parameters to boost R0 were increases in the probabilities of becoming infected; the probability of staying low-ranking; and the survival of susceptible cubs and subadults (Fig. 3). As the infection probability was very high during epidemic (Table 1), a high survival of susceptible cubs and subadults indicated a high infection risk. For each parameter except the survival of susceptible cubs, high-ranking females contributed substantially more than low-ranking ones to either an increase or decrease in R0.

Stable stage distribution (asymptotic values). The proportions of female cubs, subadults, breeders and non-breeders varied little across periods (Supplementary Fig. 1). On average (±s.d.), breeders (0.35 ± 0.04) and non-breeders (0.34 ± 0.04) were more common than cubs (0.18 ± 0.03) or subadults (0.12 ± 0.02).

The proportions of low and high social states were similar and varied little across periods (Supplementary Fig. 2), with slightly more low-ranking females (0.54 ± 0.04) than high-ranking ones (0.46 ± 0.03). The higher proportion of low-ranking females was a consequence of a higher probability of transition from a high to a low social state than from a low to a high social state (see Table 1).

The proportions of susceptible, infected and recovered females substantially varied across periods (Fig. 4 and Supplementary...
Asymptotic, predicted values as expected from a stable stage distribution were as follows: during pre-epidem, the proportion of susceptible females was high (0.60 ± 0.02) and the proportions of infected and recovered were low (0.08 ± 0.01 and 0.31 ± 0.03, respectively). During epidem, the proportion of susceptible individuals was very low (0.01 ± 0.001), whereas the relative proportions of infected and recovered individuals was high (0.17 ± 0.002 and 0.82 ± 0.07, respectively). During post-epidem, there was a notable increase in the proportion of susceptible and a decrease in the proportion of recovered individuals (susceptible: 0.15 ± 0.01, infected: 0.16 ± 0.001, recovered: 0.70 ± 0.06).

Transient dynamics. Figure 4 shows how the proportions of females in each infection state changed over the first 10 projected years and then reached a stable stage distribution. During epidem and post-epidem, the pool of susceptibles was rapidly depeleted, owing to the very high probabilities of infection (Table 1).

Reproductive values. As is typical for the reproductive value, representing the expected current and future reproductive output, it initially increased with age, that is, from cubs to subadults and also from subadults to breeders and non-breeders. Across study periods, the reproductive value of subadults was 2.2 times higher than that of cubs and the reproductive value of adults (pooled breeders and non-breeders) was 6.2 and 2.8 times higher than that of cubs and subadults, respectively (Supplementary Fig. 4). Among adults, the reproductive value of breeders was higher than that of non-breeders across periods (averaged values across periods: breeders: 17.4 ± 0.11, non-breeders 14.1 ± 0.16). Reproductive values varied little across periods (Supplementary Fig. 4).
High-ranking females, irrespective of their demographic state or infection state, had a reproductive value (31.6 ± 1.69) almost twice as large as that of low-ranking females (16.3 ± 1.22) across all periods, demographic states and infection states (Supplementary Fig. 5).

Across all periods, reproductive values of females were similar between different infection states (susceptible: 17.1 ± 1.46, infected: 16.5 ± 1.44, recovered: 14.3 ± 1.54, Supplementary Fig. 6). The lower value for recovered individuals, as compared to susceptible and infected ones, resulted from the fact that the recovered state did not include any cub (because cubs were never diagnosed as recovered in our original data set), whereas both the susceptible and the infected states included cubs.

Ecological resistance and recovery. We projected changes in population size using the values of $\lambda$ obtained for each of the three periods onto the period from 2000 and 2010, when the virulent strain was absent from the ecosystem (Fig. 5, pink curve —see Supplementary Fig. 7 and Supplementary Results for an alternative model). In addition, we predicted changes in population size between 2010 and 2020, based on the parameters estimated for the period 2000–2010. Ecological resistance, the opposite of the magnitude of the reduction (16%) in population size, was 84% and recovery time, the time needed for the population to regain its pre-epidemic size, was 16 years.

We fitted a MECMR model without any contribution of social status to all processes to show how population size would vary between 1990 and 2020 if all parameter values would be similar for high-ranking and low-ranking females (Fig. 5, blue curve). By doing so, we highlight the role of high-ranking females in increasing ecological resistance and accelerating population recovery after the epidemic. For this model, the drop in...
population size was 23%, ecological resistance was therefore 77% and full recovery was not reached by 2020 (Fig. 5).

Discussion
The predicted impact of the CDV epidemic on female hyenas was severe. The projected hyena population was reduced by 16% and needed more than a decade to return to its pre-epidemic size. CDV is a global multi-host pathogen, infecting and causing mortality in a wide range of hosts, mostly carnivores23. In Africa, CDV is considered to be an exotic pathogen imported from the USA to South Africa by human activity in the previous century24.

As expected from other morbilliviruses, such as measles or phocine distemper virus, $R_0$ in hyenas was elevated during the epidemic, demonstrating the capacity of the novel, virulent strain first observed during this epidemic to invade its new host population in 1993. Here we estimated a CDV $R_0$ in a closed hyena population, not considering contacts between hyenas and lions or other carnivores present in the Serengeti ecosystem. We consider it unlikely that our estimate of CDV $R_0$ in hyenas was substantially affected by other carnivores since CDV transmission occurs primarily via physical contacts, the frequency of contacts among hyenas (which are highly social) is substantially higher than between hyenas and other carnivores, and infectiousness was mostly among juveniles stationed at communal dens, who have limited contact to other non-canid species shedding CDV20,25.

Adults allocated to the infected state were likely not infectious themselves, i.e. not actively shedding the virus10,19,20. Nikolin et al.19 provide a more detailed discussion of the likely involvement of various carnivore species from an ecosystem-wide perspective.

An interesting finding was that reproductive values varied little with infection states (Supplementary Fig. 6). This is probably because subclinical CDV infection in adult females, if present, did not reduce their survival or chance of becoming a breeder (see Table 1 and ref. 10), and possibly also because of the rapid transition to the recovered state. Had infection negatively impacted female survival or fecundity, we would have expected larger differences in female reproductive values between the susceptible, infected and recovered states. Interestingly, CDV $R_0$ decreased as the probability for high-ranking females to maintain their social status increased, probably because high-ranking females were more likely to produce a litter than low-ranking females, and their cubs were less likely to become infected and to die of CDV infection (Table 1). We detected an elevated demographic contribution by high-ranking females to population growth (Fig. 2). The importance of high social status for population growth is further illustrated by the predicted persistent population reduction and lack of full recovery when we modelled the female population as homogeneous in terms of social states (Fig. 5, blue curve).

In contrast to the lion population, which was reduced by 30% during the same epidemic18, hyenas were relatively ecologically resistant with a potential reduction of only 16%. This difference could be explained by the fact that in hyenas CDV infection only decreased juvenile survival (Table 1), whereas in lions mortality occurred in all age classes18. As the hyena population growth rate was highly sensitive to adult survival (Fig. 2), the ecological resistance of hyenas would have been far lower if infection had decreased adult survival.

Although survival of adult hyenas was not affected by CDV, our projections show that hyenas needed potentially more time to recover from this epidemic (~16 years, Fig. 5) than lions (~3 years)26. The most likely reasons for such a difference between the two species are different life histories and maternal investment patterns. Lionesses have no dominance hierarchy, produce...
litters of 1–6 cubs, nurse communally and share food. In contrast, female hyenas have a linear dominance hierarchy and compete intensively for food, produce smaller litters of 1–3 cubs, and do not nurse communally. Instead, hyenas provide exceptionally rich milk and high maternal input in terms of individual lactation effort which can last for 12–20 months and intense within-litter competition can result in a reduction in litter size.

During the epidemic, hyena cub mortality was also spread across several cohorts born between 1993 and 1994, increasing the time required for the population to compensate increased cub mortality. Our results thus suggest that for K-selected species with a slow development of young, high resilience may depend on good ecological resistance to the initial epidemic rather than a fast recovery if pathogen infection is virulent to juveniles but benign to adults. In addition, in social species with dominance hierarchies, it is likely that high-ranking animals will substantially contribute to population recovery after an outbreak.

Epidemics of emerging infectious diseases pose a threat to many K-selected species. For example, Ebola outbreaks recently caused a severe decline in a western lowland gorilla population; this population has not yet recovered. Substantial advances in our understanding of infectious diseases and ability to forecast their outcome has been considerably improved by the development of continuous time models. However, the theoretical development of these models has outpaced their fusion with data, partly owing to the practical difficulty of estimating their parameters.

To accurately predict and improve the diagnosis of the impact of future epidemics on vulnerable wildlife populations, it is thus essential that we develop appropriate data-driven models. Because discrete time modelling approaches can easily incorporate (typically uncertain) field data and because they allow the interactions between host demography, sociality and disease dynamics to be assessed, we expect their use to increase in disease ecology. By monitoring changes in the stable state proportions of individuals in terms of susceptible, infected and recovered (Supplementary Fig. 3) through time, discrete time modelling approaches allow to predict when a population is at risk following the emergence of a virulent strain. This would be expected when a higher proportion of susceptible individuals than recovered ones occurs. These types of models can also be useful to compare the outcomes of management interventions by predicting their impacts on changes in population size and assessing which vital rates have the greatest impact on population viability. One interesting development of our deterministic model, which assumes constant environmental effects and constant survival rates during each epidemic period, would be to include the effects of stochastic forces (environment or demography) or density-dependent or frequency-dependent processes on the dynamics of disease transmission.

The states of the matrix population model. We only modelled the female section of the hyena population, as typical for demography models, which often focus on demographic sex. The four demographic states for the stage-structured matrix population model were as follows: cub (C), subadult (SA), breeder (B) and non-breeder (NB) (Table 2). Cubs were younger than 1 year, subadults aged between 1 and 2 years. Cubs entirely depended on maternal milk for at least the first 6 months. Subadults at least partially, as weaning occurs at 12 months of age. Breeders gave birth to a litter during a given year, as documented by a freshly ruptured clitoris caused by parturition and/or subsequent lactation, whereas non-breeders did not. As breeders were mostly lactating females, considering this state allowed us to account for the elevated energetic cost of lactation and the possible delayed and indirect costs of losing their offspring during the outbreak.

The two social states were low (L) or high social status (H) (Table 2). These social states were based on the positions of females in strictly linear adult female dominance hierarchies in study clans. To construct these hierarchies we recorded, for each clan, submissive behavioural interactions during dyadic interactions between adult females. Hierarchies were updated after recruitment or loss of adult females and after periods of social instability. To compare the social status of each female (across clans and years), females were assigned standardised ranks. Standardised ranks were evenly distributed between the highest (+1) and lowest (−1) standardised rank. Adult (i.e. breeder and non-breeder) females were classified as holding a high status if their standardised rank was between 0.1 and +1 or a low status if their standardised rank was between −1 and 0.1. Every year, we assigned half of the females with a standardised rank above the median to the high social status H and half of the female with a standardised rank below the median standardised rank to low social status L. We assigned the social state of an offspring to its mother to non-adult females and subadults and its adoptive mother to adopted ones (as offspring typically acquire a rank immediately below that of the female that reared them). High-ranking females have preferential access to food resources in clan territories and commute less frequently to areas outside the territory than low-status females. As a result, they produce their first offspring at an earlier age and produce more milk, which results in their cubs growing faster and surviving better to adulthood than the cubs of low-ranking females. We found previously that the cubs of high status females have a higher probability of surviving infection with CDV (Table 1) than those of low-ranking females.

The three infection states for female hyenas were susceptible (S), infected (I) and recovered (R) (Table 2). The outcomes of diagnostic procedures were employed to assign these states: (1) RT-PCR screening for the presence or absence of CDV RNA in samples, (2) CDV antibody titres in serum and (3) the observation of clinical signs associated with CDV infection in hyenas, and the secondary infections it causes in this species (for more details on these procedures, including on antibodies used and sample sizes, see the Supplementary Methods and Supplementary Table 1). CDV infection could occur only once in life because hyenas develop life-long immunity to this virus if they survive the infection. As a result, they produce their first offspring at an earlier age and produce more milk, which results in their cubs growing faster and surviving better to adulthood than the cubs of low-ranking females. We found previously that the cubs of high status females have a higher probability of surviving infection with CDV (Table 1) than those of low-ranking females.

Our MECMR model included individuals with an unknown infection state, that is, we modelled uncertainty on the infection state (Supplementary Methods). As we focused on uncertainty related to infection states, we did not include a small fraction of individuals with unknown or unclear demographic and social states. Including such individuals would imply to account also for uncertainty on the demographic and social states and this would result in over-parameterised MECMR models. Excluding this small fraction of individuals is unlikely to have biased the parameters estimated in this study because MECMR models account for potential heterogeneity in detection probability and sample sizes remained large even after exclusion of this fraction. All females were highly detectable, regardless of their status, as we show below and in the Supplementary Results. To fulfil a main assumption of CMR analyses, we assumed that we focused exclusively on hyenas detected during our routine observations at clan communal dens, i.e. did not include hyenas that were detected opportunistically away from clan communal den areas. As we were interested in the key impact and the long-term perspective (20 years) of the impact of the epidemic population dynamics rather than a fine-scaled, short-term one, the duration of the infected state in our model exceeded the actual duration of the infectious period with CDV. For the same reason, we did not include an exposed state, i.e., we did not model SEIR dynamics.
CDV prevalence and the proportion of infected female hyenas in each demographic state are shown in Supplementary Fig. 8 and Supplementary Fig. 9, respectively. Sample sizes for each combination of a given demographic, social and infection state are provided in the Supplementary Table 2.

**Structure of the meta-matrix.** The overall stage-structured matrix population model, termed the meta-matrix $M$, was of the form:

$$M = \begin{pmatrix} C & Z & Z & F & Z \\ SA & Z & Z & F & Z \\ B & Z & B_{SA} & B_{B} & B_{NB} \\ NB & Z & NB_{SA} & NB_{B} & NB_{NB} \end{pmatrix}$$

(Eq. (1))

Each entry in $M$ in Eq. (1) was a 6x6 submatrix that accounted for survival, social and infection processes also corresponding to the transition from a starting demographic state (4 columns corresponding to the demographic state: cub (C), subadult (SA), breeder (B), non-breeder (NB) on the top of each column in Eq. (1)) to the following demographic state (4 rows corresponding to C, SA, B, NB, on the left side of the matrix). Thus, $M$ presented in Eq. (1) is a block-matrix notation of total dimension 24x24. By combining the demographic, social and infection states, the female hyenas, i.e. the transitions between the 4 demographic states (shown as grey circles and with C cub, SA subadult, B breeder, NB non-breeder), conditional on survival and considering potential changes in social and infection states. The matrices were named according to the following demographic state, with the starting demographic state as index. $F$ was the fertility matrix, accounting for cub production by breeders states. $Social$ a 6x6 submatrix accounting for transitions among social states and $DemoB_{SA}$ a 6x6 submatrix accounting for the demographic transitions, i.e. the probability of becoming a breeder (recruitment) for subadults.

The other submatrices $SA_{C}$, $NB_{SA}$, $B_{B}$, $NB_{B}$ and $NB_{SA}$ in $M$ (Eq. (1), Fig. 6) were built using the same logic, and were as follows:

$$SA_{C} = Survival_{C} \times Social \times DemoB_{SA}$$

(Eq. (4))

$$NB_{SA} = Survival_{SA} \times Infection_{C} \times Social \times DemoNB_{SA}$$

(Eq. (5))

$$B_{B} = Survival_{B} \times Infection_{C} \times Social \times DemoB_{B}$$

(Eq. (6))

$$NB_{B} = Survival_{B} \times Infection_{C} \times Social \times DemoNB_{B}$$

(Eq. (7))

$$NB_{SA} = Survival_{SA} \times Infection_{C} \times Social \times DemoNB_{SA}$$

(Eq. (8))

For the 6x6 submatrix accounting for fertility $F$ (Eq. (1), Fig. 6), we assumed that all births and deaths occurred simultaneously at the end of the projection interval, that is, we modelled the population with a birth pulse reproduction and a pre-breeding census. Heterogeneity in the timing of births was handled in the statistical analysis of our data. The fertility submatrix was:

$$F = Survival_{C} \times Infection_{C} \times Social \times fecundity$$

(Eq. (10))

In Eq. (10), $Survival_{C}$ was a 6x6 submatrix accounting for cub survival, $Infection_{C}$ a 6x6 submatrix accounting for transitions between infection states for cubs, $Infection_{SA}$ a 6x6 submatrix accounting for transitions between infection states for newborns and $fecundity$, the elementary fertility submatrix.

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**Table 2 Notation and definition of demographic, social and infection states in the stage-structured matrix population model**

| The 4 demographic states | The 2 social states | The 3 infection states |
|--------------------------|---------------------|------------------------|
| Cub (C): 0 – 365 days    | Low social status (L): below median standardised social rank (low-ranking) | Susceptible (S): never infected with CDV |
| Subadult (SA): 366 – 720 days | High social status (H): equal or above median standardised social rank (high-ranking) | Infected (I): infected with CDV |
| Adult breeder (B): have given birth to a litter (that year) | | Recovered (R): immune to CDV |
| Adult non-breeder (NB): have not given birth to a litter (that year) | | |

These states were assigned to 625 female hyenas each year during 1990 and 2010. We considered uncertainty in the assignment of infection states. For further details on the methods used to assign these states see and Supplementary Methods.
The structure of such block-matrix model in M (Eq. (1)) was similar to the structure of the MECDM model used to estimate demographic, social and infection parameters, reflecting the transitions from one state to another on a yearly basis\(^{10}\).

**Description of all submatrices.** Each entry in the submatrices described below corresponded to the probability of transition (or survival), or the fecundity, from a starting combination of social and infection states to a following combination of social and infection states. As indicated above, the 6 rows and 6 columns in these submatrices corresponded to: Low social status—Susceptible, Low social status—Infected, Low social status—Recovered, High social status—Susceptible, High social status—Infected, High social status—Recovered.

We built six diagonal demography submatrices; three to account for the transition to the breeder state (Demo\(_{BA}\), Demo\(_{NB}\), Demo\(_{NBBA}\) (Fig. 7a) and three others to account for the transition to the non-breeder state (Demo\(_{BA}\), Demo\(_{NB}\), Demo\(_{NBBA}\)) (Fig. 7b), both accessed from the subadult, breeder or non-breeder state. The parameter \(\psi\) in these submatrices denoted the transition probability to the breeder state and \(1 - \psi\) the transition probability to the non-breeder state. As indicated above and in Table 2, all surviving cubs became subadults by definition, explaining why there is no demography submatrix in Eq. (4). These matrices were as follows:

\[
\text{Demo}_{BA} = \begin{bmatrix}
L - S & \psi_{SA} & 0 & 0 & 0 & 0 \\
L - I & 0 & \psi_{SA} & 0 & 0 & 0 \\
L - R & 0 & 0 & \psi_{SA} & 0 & 0 \\
H - S & 0 & 0 & 0 & \psi_{SAH} & 0 \\
H - I & 0 & 0 & 0 & 0 & \psi_{SAH} \\
H - R & 0 & 0 & 0 & 0 & 0
\end{bmatrix}
\]

(11)

\[
\text{Demo}_{BA} = \begin{bmatrix}
L - S & \psi_{BL} & 0 & 0 & 0 & 0 \\
L - I & 0 & \psi_{BL} & 0 & 0 & 0 \\
L - R & 0 & 0 & \psi_{BL} & 0 & 0 \\
H - S & 0 & 0 & 0 & \psi_{BH} & 0 \\
H - I & 0 & 0 & 0 & 0 & \psi_{BH} \\
H - R & 0 & 0 & 0 & 0 & 0
\end{bmatrix}
\]

(12)

\[
\text{Demo}_{NB} = \begin{bmatrix}
L - S & 1 - \psi_{NBL} & 0 & 0 & 0 & 0 \\
L - I & 0 & 1 - \psi_{NBL} & 0 & 0 & 0 \\
L - R & 0 & 0 & 1 - \psi_{NBL} & 0 & 0 \\
H - S & 0 & 0 & 0 & 1 - \psi_{NBH} & 0 \\
H - I & 0 & 0 & 0 & 0 & \psi_{NBH} \\
H - R & 0 & 0 & 0 & 0 & 0
\end{bmatrix}
\]

(13)

\[
\text{Demo}_{BA} = \begin{bmatrix}
L - S & 1 - \psi_{BL} & 0 & 0 & 0 & 0 \\
L - I & 0 & 1 - \psi_{BL} & 0 & 0 & 0 \\
L - R & 0 & 0 & 1 - \psi_{BL} & 0 & 0 \\
H - S & 0 & 0 & 0 & 1 - \psi_{BH} & 0 \\
H - I & 0 & 0 & 0 & 0 & \psi_{BH} \\
H - R & 0 & 0 & 0 & 0 & 0
\end{bmatrix}
\]

(14)

\[
\text{Demo}_{NB} = \begin{bmatrix}
L - S & 1 - \psi_{NBL} & 0 & 0 & 0 & 0 \\
L - I & 0 & 1 - \psi_{NBL} & 0 & 0 & 0 \\
L - R & 0 & 0 & 1 - \psi_{NBL} & 0 & 0 \\
H - S & 0 & 0 & 0 & 1 - \psi_{NBH} & 0 \\
H - I & 0 & 0 & 0 & 0 & \psi_{NBH} \\
H - R & 0 & 0 & 0 & 0 & 0
\end{bmatrix}
\]

(15)

\[
\text{Demo}_{BA} = \begin{bmatrix}
L - S & 1 - \psi_{BL} & 0 & 0 & 0 & 0 \\
L - I & 0 & 1 - \psi_{BL} & 0 & 0 & 0 \\
L - R & 0 & 0 & 1 - \psi_{BL} & 0 & 0 \\
H - S & 0 & 0 & 0 & 1 - \psi_{BH} & 0 \\
H - I & 0 & 0 & 0 & 0 & \psi_{BH} \\
H - R & 0 & 0 & 0 & 0 & 0
\end{bmatrix}
\]

(16)

\[
\text{Demo}_{BA} = \begin{bmatrix}
L - S & r_L & 0 & 0 & 0 & 0 \\
L - I & 0 & r_L & 0 & 0 & 0 \\
L - R & 0 & 0 & r_L & 0 & 0 \\
H - S & 1 - r_L & 0 & 0 & \psi_{NBH} & 0 \\
H - I & 0 & 1 - r_L & 0 & 0 & \psi_{NBH} \\
H - R & 0 & 0 & 1 - r_L & 0 & 0
\end{bmatrix}
\]

(17)

We built a social submatrix to account for the fact that subadults, breeders and non-breeders could either stay within their social state or change it (Fig. 8a). The parameters \(r_L\) and \(r_H\) in this submatrix represented the probabilities of staying in a low or high social state, respectively, and \(1 - r_L\) and \(1 - r_H\) the probabilities of becoming high or low ranking, respectively. Cubs had the same social state as their female hyenas corresponded to the submatrix Social (Eq. (17)). The two social states are indicated as grey circles (L: low ranking; H: high-ranking). The arrows show the transition probabilities, with \(r_L\) and \(r_H\) the probabilities of staying in a low or high social state, respectively, and with \(1 - r_L\) and \(1 - r_H\) the probabilities of becoming high or low-ranking, respectively.

**Fig. 8 Transitions between social and infection states for female hyenas.** a Transitions between social states in subadult, breeder and non-breeder female hyenas corresponded to the submatrix Social (Eq. (17)). The two social states are indicated as grey circles (L: low ranking; H: high-ranking). The arrows show the transition probabilities, with \(r_L\) and \(r_H\) the probabilities of staying in a low or high social state, respectively, and with \(1 - r_L\) and \(1 - r_H\) the probabilities of becoming high or low-ranking, respectively. b Transitions between infection states in subadult, breeder and non-breeder female hyenas, corresponding to the submatrix Infection (Eq. (20)). The three infection states are indicated as grey circles (S susceptible, I infected, R recovered). The arrows show the transition probability between those states, with \(\beta\) the probability of transition from a susceptible state to an infected state (the infection probability). The infection probability varied with the social state (see Table 1) — this is not shown, for simplicity.

We built three infection submatrices: one for newborns (Infection\(_{NB}\), Eq. (18)), one for cubs (Infection\(_{C}\), Eq. (19)) and one for subadults, breeders and non-
breeders (Infection, Eq. 20, Fig. 8b). As mentioned above, we considered that infected individuals became recovered at the following interval, conditional on their survival. We also considered that recovered individuals acquired life-long immunity to CDV, which implied that they remained recovered until their death or disappearance (as in19). For the infection submatrix for newborns Infection_0, part of the submatrix F (in Eqs. (1) and (10), Fig. 6), we considered that mothers only produced susceptible newborns, irrespective of their own infection status. Indeed, as we assumed that adults were not infectious and hence did not excrete the virus actively as young individuals,10,19,20, we considered the transmission of CDV to be strictly horizontal. These key features of the model generated susceptible individuals in the model population at each projection interval (in addition to those already present in the initial population vector, details below). The 6 × 6 matrix for newborns, Infection_0, was as follows:

\[
\text{Infection}_0 = \begin{bmatrix}
L - S & 1 & 1 & 0 & 0 & 0 & 0 \\
L - I & 0 & 0 & 0 & 0 & 0 & 0 \\
L - R & 0 & 0 & 0 & 0 & 0 & 0 \\
H - S & 0 & 0 & 0 & 1 & 1 & 1 \\
H - I & 0 & 0 & 0 & 0 & 0 & 0 \\
H - R & 0 & 0 & 0 & 0 & 0 & 0 \\
\end{bmatrix}
\]

In Eq. (18), the value 1 represented the production of susceptible newborns by mothers of low social status and high social status (by definition all breeders produced a litter thus the probability of female offspring production was equal to 1).

For cubs, the only possible infection transitions were from a susceptible to a susceptible state or from a susceptible to an infected state. The submatrix for cubs, Infection_c (Eq. 19) was thus:

\[
\text{Infection}_c = \begin{bmatrix}
L - S & 1 - \beta_{CL} & 0 & 0 & 0 & 0 & 0 \\
L - I & \beta_{CL} & 0 & 0 & 0 & 0 & 0 \\
L - R & 0 & 1 & 0 & 0 & 0 & 0 \\
H - S & 0 & 0 & 0 & 1 - \beta_{CH} & 0 & 0 \\
H - I & 0 & 0 & 0 & \beta_{CH} & 0 & 0 \\
H - R & 0 & 0 & 0 & 0 & 1 & 0 \\
\end{bmatrix}
\]

In Eq. (19), \(\beta_{CL}\) and \(\beta_{CH}\) were the infection probabilities for low-ranking and high-ranking cubs, respectively. The values 1 corresponded to infected cubs becoming recovered at the next interval (conditional on their survival). As there was no cub in the recovered state in the data set19, there was no transition from recovered cub to recovered cub.

The submatrix Infection accounted for possible transitions between susceptible, infected and recovered states for subadults, breeders and non-breeders (Fig. 8b). \(\beta_{L}\) and \(\beta_{H}\) were the infection probabilities for low-ranking and high-ranking subadults, breeders and non-breeders, respectively. The probabilities of transition from a susceptible to an infected state (and hence 1 – \(\beta_{L}\) and 1 – \(\beta_{H}\)) the probabilities of transition from a susceptible to a susceptible state for low-ranking and high-ranking subadults, breeders and non-breeders, respectively. The values 1 corresponded to infected individuals becoming or staying recovered at the next interval (conditional on their survival) (Fig. 8b). This matrix was as follows:

\[
\text{Infection} = \begin{bmatrix}
L - S & 1 - \beta_{L} & 0 & 0 & 0 & 0 & 0 \\
L - I & \beta_{L} & 0 & 0 & 0 & 0 & 0 \\
L - R & 0 & 1 & 0 & 0 & 0 & 0 \\
H - S & 0 & 0 & 0 & 1 - \beta_{H} & 0 & 0 \\
H - I & 0 & 0 & \beta_{H} & 0 & 0 & 0 \\
H - R & 0 & 0 & 0 & 0 & 1 & 1 \\
\end{bmatrix}
\]

As cubs were either susceptible or infected and never diagnosed as recovered in our original data set19, in the last step of the model development we deleted the two columns and two rows corresponding to recovered cubs. We thus ended up with a 22 × 22 meta-matrix \(M\) (Eq. (1)).

**Asymptotic analyses of the matrix population model.** We used R v. 3.5.0. (R Core Team 2017)21. We used the function ‘is.matrix’/’irreducible’ in R’s package popbio v1.3-921 to verify that the meta-matrix was irreducible. Matrix models are termed irreducible when their associated life cycles contain the transition rates to facilitate pathways from all states to all other states. Irreducible matrices are ergodic: the stable asymptotic growth rate is independent from the initial stage structure in the population projection. Both conditions should ideally be met for further analyses21,22.

When \(M(t) = M\) is considered constant (i.e., no time dependence, no stochasticity), the behaviour of such a deterministic model is well-known22. The modelled population is then characterised by an asymptotic growth rate and a stable asymptotic distribution over the 22 states, which are the dominant eigenvalue and the corresponding right eigenvector, respectively17,21. We used the function ‘ecolfit’ and the ‘productive value’ in R’s package popbio 2.4.425 to determine the stable stage distribution and the reproductive values, based on the definitions provided by22.
We used the functions ‘pop.projection’ and ‘stage.vector.plot’ from R’s package popbio 2.4.44 to plot the short-term population dynamics and the convergence to the stable stage distribution (Fig. 4), using the initial state probabilities estimated from the MECMR model (see Supplementary Results) as initial stage vector. The first function calculates the population growth rate and stable stage distribution by repeated projections of the Eq. (2).

We included a modification in the approach developed by44 to estimate the basic reproduction number $R_0$ of this CDV during the epidemic period, considering only hyena-to-hyena transmission (i.e. no interspecific interactions as discussed). Our modified version of $R_0$ accounted for the population’s growth rate and by doing this, we replaced the assumption of a stationary host population by accounting for observed population dynamics and thus also for changes in the proportion of infected individuals. As for equivalent time-continuous matrix models, $R_0$ is the dominant eigenvalue of the next generation matrix, itself being the product of the fundamental matrix by the reproductive matrix. The entries in the reproductive matrix in an epidemiological context correspond to the rate at which new infections are produced by infected individuals. In our case, there is only one infected state, thus only one type of infection is produced by only one type of (infected) individuals. We detail our modified approach below.

The starting point is the decomposition of the change in the number of infected in transmission and transitions:

$$A = T + R$$  \hspace{1cm} (26)

where $A$ is the projection matrix, $T$ is the transition matrix and $R$ is the reproductive matrix. Under asymptotic conditions, the changes in one time step are within a population that grows at rate $\lambda$. To represent the change in the number of infected as a change in proportion relative to the overall population, one can discount for population growth rate at $\lambda$ and write the contribution matrix $C$ to the “next generation of infected” (i.e. our modification of $R_0$) as:

$$C = \lambda^{-1}(R - \lambda^{-1}I)^{-1} = R(\lambda - T)^{-1}$$  \hspace{1cm} (27)

which reduces to:

$$C = \lambda^{-1}R(I - \lambda^{-1}T)^{-1} = R(\lambda - T)^{-1}$$  \hspace{1cm} (28)

Sensitivity analyses of $\lambda$, $R_0$ and stochasticity. To determine which parameters contributed most to $A$ and $R_0$ and predict the results of future changes in parameter estimates, we performed sensitivity analyses (Figs. 2 and 3). When elements of a population are composed of several classes, the classical first order sensitivity analysis is not recommended, as it does not allow one to disentangle the effects of demographic, social and infection parameters. We therefore conducted lower-level sensitivity analyses for $\lambda$ and $R_0$. For $\lambda$ we applied the function ‘vital-sens’ from the R package popbio 2.4.44. For $R_0$, we used a numerical approach in which we varied each parameter by 0.1% while maintaining the others constant, and calculated $R_0$ at each iteration as previously described.

To account for parameter uncertainty (Figs. 1, 2, 3 and 5), we used Monte Carlo iterations. At each successive period (pre-epidemic, epidemic, post-epidemic, plus the epidemic periods (pre-epidemic, post-epidemic). For survival and transition probabilities for the period 2000–2010, initial state, detection and assignment Probabilities, see Supplementary Results.

The sex-ratio at birth is balanced and was estimated in a previous study to be equal to 0.5246. Here we used this estimate as input value for the parameter sex. ratio (Table 1, fecundity matrix, Eq. (21)). This parameter estimate had no standard error associated with it. We set an arbitrary standard error value equal to 0.03 for the sensitivity analyses.

In a previous study, we estimated average litter size based on observations conducted in our three main study clans between 1987 and 200647 and the monitoring of the fate of 1124 spotted hyena cubs from 735 litters, comprising 351 (47.8%) singleton, 379 (51.6%) twin and five (0.7%) triplet litters. Thus, for the average litter size (parameter litter.size, Table 1, fecundity matrix, Eq. (21)) we used the value 1.53 (1124/735). This parameter estimate had no standard error associated with it. We set an arbitrary standard error value equal to 0.03 for the sensitivity analyses.

Code availability. The R programming code to replicate all our analyses and figures is available from Github: https://github.com/LucileMarescot/SIR-Matrix-model. It is divided in three main sections: (1) code 1_Model construction loads the input values for the three epidemic periods and builds the matrix model, (2) code 2_Model analysis-asymptotic presents the asymptotic analysis of the matrix model and (3) code 3_Model analysis-stochastic presents the stochastic analysis of the matrix model. Each section is fully annotated so that each step can be replicated.

Data availability. The original data from Marescot et al. (2018) supplied the parameter estimates used in this study and are archived in figshare48: https://doi.org/10.6084/m9.figshare.5840970. The parameter estimates used in this study are provided in Table 1 and in Supplementary Results.

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