Original Contribution

Social Network Predicts Exposure to Respiratory Infection in a Wild Chimpanzee Group

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Abstract: Respiratory pathogens are expected to spread through social contacts, but outbreaks often occur quickly and unpredictably, making it challenging to simultaneously record social contact and disease incidence data, especially in wildlife. Thus, the role of social contacts in the spread of infectious disease is often treated as an assumption in disease simulation studies, and few studies have empirically demonstrated how pathogens spread through social networks. In July–August 2015, an outbreak of respiratory disease was observed in a wild chimpanzee community in Kibale National Park, Uganda, during an ongoing behavioral study of male chimpanzees, offering a rare opportunity to evaluate how social behavior affects individual exposure to socially transmissible diseases. From May to August 2015, we identified adult and adolescent male chimpanzees displaying coughs and rhinorrhea and recorded 5-m proximity data on males (N = 40). Using the network k-test, we found significant relationships between male network connectivity and the distribution of cases within the network, supporting the importance of short-distance contacts for the spread of the respiratory outbreak. Additionally, chimpanzees central to the network were more likely to display clinical signs than those with fewer connections. Although our analyses were limited to male chimpanzees, these findings underscore the value of social connectivity data in predicting disease outcomes and elucidate a potential evolutionary cost of being social.

Keywords: Pathogens, Sociality, Contact network, Primates, Cough, Dominance

INTRODUCTION AND PURPOSE

Group life is important for many species as it may reduce the risk of predation and improve foraging efficiency, but one major cost of sociality is the increased opportunity for direct pathogen transmission (Alexander 1974; Krause and...
Contact networks have emerged as a valuable way to examine how social interactions influence pathogen transmission, especially for species with heterogeneous contact rates (Craft and Caillaud 2011; Rushmore 2014; Craft 2015; Reynolds et al. 2015; Croft et al. 2016; Romano et al. 2016; Webber et al. 2016; Springer et al. 2017). Several recent examples have demonstrated strong correlations between social networks and transmission networks. For example, social networks have been associated with tuberculosis infection in meerkats (Drewe 2010), Salmonella enterica in sleepy lizards (Bull et al. 2012), chronic wasting disease in mule deer (Salazar et al. 2016), skin disease in dolphins (Powell 2019), and Escherichia coli in opossums (Blyton et al. 2014), giraffe (VanderWaal et al. 2014), and captive rhesus macaques (Balasubramaniam et al. 2019). Theoretical models have generalized these empirical findings for wildlife by showing that in heterogeneous contact networks, individuals systematically vary in their likelihood to acquire and transmit pathogens to a large number of susceptible individuals (Rushmore et al. 2017; White et al. 2017).

Infectious diseases have emerged as one of the main threats to endangered wildlife populations (Daszak et al. 2000; Smith et al. 2009), and group-living taxa, such as primates, are particularly susceptible (Wolfe et al. 1998; Wallis and Rick Lee 1999; Chapman et al. 2005; Nunn and Gillespie 2016). Some of the best-documented cases of wildlife disease outbreaks concern great apes (Leendertz et al. 2004, 2006; Kaur et al. 2008; Körndgen et al. 2008; Palacios et al. 2011; Grützmacher et al. 2016; Emery Thompson et al. 2018). For example, Ebola devastated wild gorilla populations in the early 2000s (Bermejo et al. 2006; Caillaud et al. 2006). More recently, multiple respiratory disease outbreaks in wild chimpanzees have been linked to zoonotic viruses (Kaur et al. 2008; Körndgen et al. 2008; Williams et al. 2008; Scully et al. 2018; Negrey et al. 2019). Understanding pathogen spread and factors or interventions that influence individual infection risk has therefore become a conservation priority (Nunn et al. 2008; Carne et al. 2013; Rushmore 2014).

Given the impact that directly transmitted diseases have on primate populations, the relationship between social networks and pathogen spread has recently become a focus of research for wild primates. For example, Kuehl et al. (2008) concluded that social contacts increased pathogen transmission in young western chimpanzees after determining that increased infant play behavior corresponded to peaks in infant mortality. Another study tracked individually marked lice (Lemurpediculus verruculosus) in a mouse lemur population and provided direct evidence of between-individual transfers during social contacts (Zohdy et al. 2012). Brown spider monkeys that have more physical contact with conspecifics exhibit a higher gastrointestinal parasite species richness than less connected individuals (Rimbach et al. 2015). Similarly, mangabeys that occupy more central positions in proximity networks are more likely to be reinfected by helminths after chemotherapeutic treatment (Friant et al. 2016). Finally, being members of the same social group is a strong predictor of intestinal bacterial flora composition in sifaka (Springer et al. 2016; Perofsky et al. 2017). In addition to social connectivity, dominance status is associated with gastrointestinal helminthic parasite infection in wild chimpanzees (Muehlenbein and Watts 2010), nematode infection in Japanese macaques (MacIntosh et al. 2012), and influences immune function in macaques (Tung et al. 2012; Snyder-Mackler et al. 2016), although this pattern is inconsistent across species (Habig and Archie 2015). Indeed, the influence of dominance status and social network position on infection may vary depending on the specific disease or parasite in question (MacIntosh et al. 2012).

Social connectivity and social status appear to influence exposure to infection in wild primates, but the role of sociality in shaping respiratory outbreaks—a major conservation threat for great apes—remains poorly documented (Vogel 2003; Leendertz et al. 2006; Rushmore et al. 2013; Rushmore 2014; Romano et al. 2016; Emery Thompson et al. 2018; Scully et al. 2018; Negrey et al. 2019). Chimpanzees are an ideal taxon for exploring links between sociality and infection due to their fission–fusion social dynamics. The members of a chimpanzee community rarely are all together at the same time; instead, individuals form temporary subgroups that last a few hours to a few days (Nishida 1968; Goodall 1986). Within-community variation in sociality creates heterogeneity and potential sub-structuring, increasing the chance of detecting whether pathogen transmission tracks the underlying contact network. In addition, adult male chimpanzees form a linear dominance hierarchy, which allows assessment of the interconnection between dominance rank, social contacts, and infection status (e.g., Rushmore et al. 2013).

In July and August 2015, the Ngogo chimpanzees of Kibale National Park, Uganda, experienced an outbreak of non-lethal respiratory disease, which we report here for the first time. The outbreak occurred during an ongoing study...
of male chimpanzee behavior, providing a rare opportunity to examine the role of social networks in predicting disease outcomes for an acute respiratory pathogen in a wildlife population. Here, we test if social contacts influenced the distribution of clinical signs among male chimpanzees, using data collected on 40 adolescent and adult male chimpanzees from the Ngogo community. We also examined the importance of a male’s position in the social network relative to his dominance status and age in predicting the likelihood of exhibiting clinical signs. We tested the hypotheses that (1) contact heterogeneity among dyads affects disease dynamics, and (2) males more central in the network are more likely to exhibit clinical signs, controlling for dominance status and age. We also tested (3) the effect of dominance on centrality and (4) the effect of age on centrality.

**METHODS**

**Study Site and Subjects**

We conducted observations of chimpanzees at Ngogo, Kibale National Park, Uganda (00° 29′ 53″ N, 30° 25′ 30″ E). Located in the center of the Park, the Ngogo chimpanzee community territory is surrounded on all sides by other chimpanzee communities and the only contact with humans is from researchers and research staff. The habitat consists mostly of mature rainforest interspersed between areas of regenerating forest, grasslands, and swamps (Struhsaker 1997). During the data collection period (May 1 to August 29, 2015), the Ngogo chimpanzee community consisted of 193 individuals, including 53 adult and adolescent males (32 adult males, 21 adolescent males), 78 adult and adolescent females (63 adult females, 15 adolescent females), 15 juveniles (10 males, 5 females), and 46 infants. Ages of chimpanzees are known to the nearest year for subjects under 20 years and estimated based on physical appearance and known genetic relationships for adults over 20 years (Wood et al. 2017).

For the purposes of this study, we focused on 32 adult male chimpanzees (16–53 years) and eight middle-to-late adolescents (12–15 years). Although including all age and sex classes would offer the most complete network for analyzing transmission dynamics (De Nys et al. 2013; Rushmore et al. 2013; Emery Thompson et al. 2018), our focus on males was partly due to methodological constraints: outbreaks are unpredictable, and during the time of the outbreak, we had the greatest precision in social and health data on mature male chimpanzees, as they were the subjects of concurrent studies by two of the authors. Although limiting to a subset of males is not ideal, there was no a priori reason to believe this would skew results. Mature male chimpanzees, especially in East Africa, are more gregarious than adult females (Wrangham and Smuts 1980; Goodall 1986; Gilby and Wrangham 2008; Carne et al. 2013; Langergraber et al. 2013; Machanda et al. 2013) and occupy a larger range within the community territory than adult females (Thompson et al. 2007; Kahlenberg et al. 2008). Thus, the male network is expected to reflect the overall network. This is similar to other studies that have limited networks to the more social sex (MacIntosh et al. 2012; Duboscq et al. 2016; Balasubramaniam et al. 2018). The focus on males is also relevant as male chimpanzees have more clearly defined dominance hierarchies than do females (Wakefield 2008; Foerster et al. 2016), and we aimed to examine the influence of dominance status on clinical signs.

**Behavioral Data**

Behavioral and proximity data were collected from May 1 to August 29, 2015 as part of long-term studies on adolescent and adult male chimpanzees. Chimpanzees were observed continuously for 60-min focal-follow episodes, and the identity of chimpanzees within 5 meters of the focal subject was recorded instantaneously at 10-min intervals. To assess dominance status, we used data on pant grunts, a formal signal of submission directed up the hierarchy and given by low-ranking chimpanzees to higher-ranking individuals (Bygott 1979; de Waal 1982). Males were given ordinal ranks (i.e., the alpha male had a rank of 1, the beta had a rank of 2, etc.), which were then transformed by subtracting the rank from the total number of males in the hierarchy and dividing this by the total number of males minus one. After transformation, the alpha male had a rank of 1 and the lowest-ranking individual had a rank of 0 (Muller et al. 2006).

Because chimpanzees live in fission–fusion societies, not all males were available for observation every day. We balanced the number of focal follows by rotating through subjects on an opportunistic basis, prioritizing males who had been observed less often than others (irrespective of clinical signs). On occasion, chimpanzees were observed twice in a single day, although in such cases focal follows were separated by at least five hours. We conducted a total
of 659 h of focal observations representing an average of 16.5 ± 6.2 h per study subject.

Health Data

We conducted health assessments as part of routine observations of all chimpanzees, similar to those used at other chimpanzee sites (Krief et al. 2005; Lonsdorf et al. 2018). Individuals observed to cough or exhibit rhinorrhea (i.e., nasal drainage) more than once on a given day across the study period (regardless of focal status) were considered to have clinical signs of respiratory infection. If observers saw an individual for at least 3 h on a given day with no clinical signs, they were considered healthy for that day. Agreement among field observers was high (agreement in 13/14 instances when all three observers monitored the same individuals). Males that were never observed to exhibit clinical signs for the duration of the outbreak were considered healthy.

Social Network and Centrality Metrics

We created an adjacency matrix of pairwise association indices based on proximity data. To calculate an association index for a given pair, we summed the number of scans the pair was observed in proximity, regardless of which individual was the focal subject, and then we divided this by the total number of scans either individual was observed as a focal subject. We built a weighted network, in which nodes represented chimpanzee individuals and edges were weighted according to pairwise association indices.

We used the weighted network to calculate strength centrality, which is the summed edge weight for a given node (Freeman et al. 1991; Newman 2010). We also calculated two other centrality metrics: eigenvector centrality (function of connectedness of a nodes’ associates) and flow-betweenness centrality (frequency with which a node is along the shortest path between pairs of nodes). Given that strength was strongly correlated with eigenvector (Spearman’s rank correlation, $r_s = 0.71$, $P < 0.001$) and flow-betweenness centrality ($r_s = 0.69$, $P < 0.001$), our statistical analyses focused on strength centrality. We calculated centrality metrics and built social networks in R (R Core Team 2018) using the sna (Butts 2016) and igraph packages (Csardi and Nepusz 2006).

Statistical Analyses

We assessed the relevance of the social network on clinical signs using two methods. First, we used the path-based k-test to assess whether the topology of chimpanzees with clinical signs in the observed network likely resulted from transmission along network edges (VanderWaal et al. 2016). The path-based k-test assesses if the mean weighted distance between discernibly infected individuals in a weighted network is shorter than expected by chance (VanderWaal et al. 2016). The path-based k-test uses a permutation-based procedure of “node-label swapping”—cases of infection are randomly re-assigned within the network—to create a null distribution. We report p values based on the distribution of the simulated statistics over 1000 randomizations. The k-test is useful for modeling pathogen transmission as it takes into account clustering within the network and is robust to missing nodes and missing cases (VanderWaal et al. 2016). The k-test can be used on social or spatial data, although we limited our analysis to social data and do not assess the relative contribution of shared space use on social structure or pathogen transmission.

We also examined the extent to which the presence of clinical signs was predicted by individual connectivity within the social network, social rank, and age (Rushmore et al. 2013; Habig and Archie 2015). We conducted a logistic regression with a binomial error distribution using the “glm” function in R (R Core Team 2018). The presence/absence of clinical signs was the response variable, and strength centrality, ordinal rank, and age were the predictor variables. To account for the non-independence of data inherent to social network metrics, we assessed variable importance by creating p values using node-level permutation with 30,000 iterations (Rushmore et al. 2013). Specifically, we calculated the regression using observed data. We then permuted the response variable data and saved estimates after each regression, calculating p values based on the permuted values. We assessed collinearity of predictor variables using the “vif” function in package car (Fox and Weisberg 2019); all variance inflation factors were less than 2, indicating little-to-no collinearity. In two additional analyses, we conducted node-level permutation-based linear regressions to assess (1) the effect of dominance rank and (2) age on strength centrality.

Individual differences in the number of observations can increase both centrality and the likelihood of detecting sick individuals. We controlled for variation in sampling.
effort by creating dyadic association indices that accounted for time spent observing each chimpanzee in a given dyad. Further, variation in observation hours across individuals did not affect the traits of interest: There was no relationship between number of observations and (a) centrality in the social network ($\beta = -0.001 \pm 0.001, P = 0.139$) or (b) whether a chimpanzee exhibited clinical signs ($\beta = 0.007 \pm 0.009, P = 0.425$). These analyses suggest to us that we collected sufficient data on each individual such that additional observations did not greatly increase our ability to determine (a) position in the social network and (b) which animals had clinical signs. Nevertheless, there may be patterns in the data that could generate a spurious outcome in a more complex model, which we did not account for here.

RESULTS

Clinical signs of respiratory infection were observed between July 20 and August 12, 2015 (Fig. 1) with high morbidity and zero mortality. Among adolescent and adult male chimpanzees, 32 out of 53 (60%) individuals exhibited clinical signs, including cough and/or rhinorrhea, representing 30 of the 40 males (75%) for which we had proximity data. We also observed 11 females with clinical signs, which was likely an underestimate, as females were not a focus of the study and were excluded from the analyses. Although it was difficult to estimate a minimum or average duration of clinical signs, a subset of males ($N = 20$) observed with clinical signs over time had a maximum duration of 11 days between onset and recovery.

Figure 1. Timing of clinical signs in individuals based on social neighborhoods. Three social clusters emerged from the observed contact network, corresponding to three qualitative subgroups acknowledged by researchers: west, central, and east. Each shaded dot represents the first time an individual was observed with clinical signs.

Figure 2. Social network predicts clinical signs of infection. Individual chimpanzees (nodes) are colored according to their presentation (red) or absence (green) of clinical signs. This contact network was created using the Fruchterman-Reingold layout algorithm in R’s igraph package. We overlay polygons based on the three social neighborhoods (west = green, central = red, and east = blue).

The k-test indicated that the male chimpanzee social network was epidemiologically relevant for the observed respiratory outbreak. Male chimpanzees with clinical signs were closer to one another in the network than expected by
chance (mean weighted path length to nearest infections: 15.04, \( P = 0.014 \)) (Fig. 3).

A permutation-based multiple regression model revealed that a male chimpanzee's position in the social network, but not his rank or age, was related to the probability of exhibiting clinical signs. Individuals with high-strength centrality were more likely to exhibit clinical signs than less central male chimpanzees (\( \beta_{\text{strength}} = 9.47, \ P = 0.006 \); Fig. 4). There were minimal effects of rank (\( \beta_{\text{rank}} = -1.14, \ P = 0.304 \)) and age (\( \beta_{\text{age}} = -0.031, \ P = 0.224 \)) on the probability of exhibiting clinical signs. Although age and rank did not predict whether individuals exhibited clinical signs, these traits were positively related to network centrality. In two additional permutation-based regression models, high-ranking males exhibited greater strength centrality than low-ranking males (\( \beta_{\text{rank}} = 0.46, \ P < 0.001 \), Supplementary Figure 1), and older males exhibited greater strength centrality than younger males (\( \beta_{\text{age}} = 0.009, \ P < 0.001 \)).

Male chimpanzees at Ngogo were clustered into three subgroups (Fig. 2) (based on the leading non-negative eigenvector of the modularity matrix, Csardi and Nepusz 2006; Newman 2006), which represented east, central, and west “neighborhoods.” There was a perfect agreement among the authors on which individual belonged to which neighborhood, and perfect agreement between the quantitative and qualitative measures of neighborhood membership. The first chimpanzees to exhibit clinical signs were all part of the central neighborhood, and clinical signs were observed in these chimpanzees from July 20 to July 30, 2015. Additional males close to the central neighborhood that are considered part of an eastern neighborhood exhibited clinical signs from August 1 to August 6, 2015.

**Figure 3.** Graphical results of the path-based network k-test. The blue shaded area represents the null distribution of the mean distance between infected cases when the infected nodes were randomized within the network. The red vertical line represents the observed mean distance to nearest infected case.

**Figure 4.** Network centrality strength predicts clinical signs of infection. This plot shows a fitted logistic regression of the effect of individual node strength centrality on the individual’s probability of exhibiting clinical signs. Each data point represents a single chimpanzee.
Males from the western neighborhood exhibited clinical signs from July 30 to August 12, 2015 (Fig. 1).

**Discussion**

The contact network, based on proximity relationships between adolescent and adult male chimpanzees, predicted which male chimpanzees exhibited respiratory infection clinical signs. Male chimpanzees that occupied central network positions were more likely to exhibit clinical signs than those less connected, whereas dominance rank and age had minimal effects on the likelihood of exhibiting clinical signs. Additional analyses revealed that high-ranking and older males were more connected in the network than low-ranking and younger males, respectively, indicating a potential indirect connection with the likelihood of infection.

The findings presented here add to a growing body of evidence that an individual’s social network is an important predictor of infectious disease risk. It is worth noting that this occurred during a particularly high degree of substructuring, with males from the western neighborhood and central/east neighborhoods avoiding each other for over one month. It may have been due to this high substructuring that the social network was epidemiologically relevant (Sah et al. 2017). In addition, a qualitative assessment of our data indicates a temporal component to the transmission, with the disease spreading first within neighborhoods, and only later transferred to other neighborhoods. It is possible that this clustering was related to the disease itself (Farine 2018), with chimpanzees traveling over shorter distances (e.g., vervet monkeys, Chapman et al. 2016) due to sickness behavior (e.g., lethargy in wild mice, Lopes et al. 2016) or behavioral avoidance (e.g., “self quarantine” behavior in ants, Heinze and Walter 2010). Given our understanding of long-term social dynamics in Ngogo chimpanzees, we consider it more likely that factors unrelated to disease caused the clustering. The western subgroup has since become isolated from central and eastern subgroup (Ngogo Chimpanzee Project, unpublished data), and the respiratory outbreak seems to have occurred soon after the western subgroup first exhibited signs of separating, which occurred for other social or ecological reasons.

Social relationships mediate immune health on multiple scales (Hock and Fefferman 2012; Ezenwa et al. 2016). Strong relationships may buffer individuals against infection (Cohen et al. 1997; Hostinar 2015) and decrease inflammation (Kim et al. 2016), as observed in humans, or reduce parasites due to grooming, as demonstrated in macaques (Duboscq et al. 2016). Other aspects of sociality may incur costs and benefits. Our study provides indirect evidence that social status mediates exposure to infectious disease via social networks, as previously hypothesized (Rushmore et al. 2013). By achieving high rank and accruing its reproductive benefits (Boesch et al. 2006; Wroblewski et al. 2009; Langergraber et al. 2013), individuals may be exposed to infectious pathogens at higher rates due to greater sociality. Across primates, high dominance rank may increase resistance to and/or tolerance of infectious pathogens, despite potential greater exposure to pathogenic agents due to central positions in the network. High-ranking macaques exhibit more robust immune responses and lower parasite loads than low-ranking conspecifics (MacIntosh et al. 2012; Georgiev et al. 2015; Snyder-Mackler et al. 2016), while low-ranking baboons (Papio cynocephalus) exhibit slower wound healing rates (Archie et al. 2012). This trade-off may explain the results of a recent longitudinal study of respiratory infection in wild chimpanzees in which old, high-ranking males and young, low-ranking males exhibited clinical signs of infection more frequently than other age-rank classes (Emery Thompson et al. 2018). Integrating data on social behavior, immune activity, and pathogen burdens will elucidate the complex relationships affecting disease exposure and susceptibility in social animals.

Although we demonstrated links between social network position and exposure to infectious disease, there are several limitations to our study. We based our study on clinical signs of respiratory infection and did not have direct evidence of pathogen prevalence or distribution within the population. Indeed, it is possible that different individuals were infected with different pathogens, although we consider this unlikely given the temporal overlap of clinical signs. In addition, we did not have the precision to calculate the onset and recovery of disease within individuals to examine temporal trends in more detail or to derive important epidemiological parameters like the basic reproductive number (R₀) or recovery rate.

A further limitation is that we focused only on adolescent and adult males. Accounting for all age and sex classes would offer a more complete network for analyzing transmission dynamics, and females and dependent offspring may be especially relevant for pathogen transmission among chimpanzees (De Nys et al. 2013; Rushmore et al. 2013; Ezenwa et al. 2016).
Had we found no effect of the social network, we would not have been able to rule out the possibility that the pathogen spread via connections among females, infants, and juveniles. The fact that we still observed more connected males to be more likely to exhibit clinical signs indicates that even the limited male network was epidemiologically relevant. It remains possible, however, that our results are false positives, and accounting for females would reveal that clinical signs were distributed randomly within the network. Given that male chimpanzees tend to associate with a greater number of partners and traverse a wider range within the territory compared with females (Wrangham and Smuts 1980; Murray et al. 2007; Gilby and Wrangham 2008; Kahlenberg et al. 2008; Carne et al. 2013; Langergraber et al. 2013; Machanda et al. 2013), we consider this possibility unlikely. Although challenging to fully capture during acute respiratory infections, future studies should ideally detail complete networks, including individuals of all age and sex classes, networks at multiple social and temporal scales (Chen et al. 2014; Pinter-Wollman et al. 2014; Farine 2017, 2018), and other epidemiologically relevant traits including environmental exposure, interspecific transmission, and extra-group conspecifics (Wallis and Lee 1999; Nunn and Dokey 2006; Bonnell et al. 2010; Gogarten et al. 2019; Johnson et al. 2019). Future studies should also assess the relative contribution of social and spatial data when assessing transmission networks (Albery et al. 2020). Space use within the territory may play an important role in the spread of pathogens, and our study was limited by only using social data to construct networks.

The dynamics between sociality and disease are especially important for great apes, as disease outbreaks represent a major threat to wild populations (Vogel 2003; Leendertz et al. 2006; Nunn and Gillespie 2016). Of particular interest is the potential risk posed by human observers (Gilardi et al. 2015). A major source of mortality in gorillas and chimpanzees is from respiratory infections, and there is mounting evidence that many of these infections are from human sources (Kaur et al. 2008; Köndgen et al. 2008; Palacios et al. 2011; Grützmacher et al. 2016; Dunay et al. 2018; Negrey et al. 2019). Future studies should examine the underlying cause of disease outbreaks. Doing so requires standardizing methods to monitor health (Krief et al. 2005; Löhrich et al. 2018; Lonsdorf et al. 2018) and record social networks (White et al. 2017). All the while, human observers must maintain strict health protocols, such as wearing facemasks, sanitizing hands regularly, and avoiding entering the forest when contagious (Gilardi et al. 2015). The combination of health monitoring, noninvasive sampling for infection, and social behavior data collection will continue to provide key insights into disease dynamics while at the same time protecting wild ape populations (Leendertz et al. 2006). Such precautions are especially important for conservationists and field biologists in light of the current coronavirus pandemic (Lappan et al. 2020; Trivedy 2020).

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Authors’ Contributions

A.A.S. conceived of the study, collected the data, performed the analysis, and wrote the paper. J.R. and D.C. designed and performed the analysis and wrote the paper, contributing equally to the manuscript, along with J.D.N. who conceived of the study, collected the data, contributed to the data analysis, and wrote the paper. J.C.M. collected the data and contributed to writing the paper. D.M.L. designed and performed earlier versions of the analysis and contributed to writing the paper.

Compliance with Ethical Standards

Ethical Approval All applicable institutional and/or national guidelines for the care and use of animals were followed. The study was exempt from review by the University Committee on Use and Care of Animals at the University of Michigan. The Uganda Wildlife Authority (UWA/TDO/33/02) and the Uganda National Council for Science and Technology (NS 401) approved the study.
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