

1      **Reduced levels of relatedness indicate that**  
2      **great-tailed grackles disperse further at the**  
3      **edge of their range**

4      Lukas, Dieter<sup>1\*</sup>      Blackwell, Aaron D<sup>2</sup>      Edrisi, Maryam<sup>2</sup>  
5      Hardy, Kristin<sup>3,4</sup>      LeGrande, Christa<sup>1</sup>      Marfori, Zara<sup>1</sup>  
6      McCune, Kelsey<sup>5,6</sup>      Sevchik, August<sup>7,8</sup>      Smith, Caroline<sup>2</sup>  
7                                 Logan, Corina J<sup>1,5</sup>



9      **Affiliations:** 1) Max Planck Institute for Evolutionary Anthropology; Leipzig, Germany, 2)  
10     Washington State University, Pullman; USA, 3) University of Rochester; Rochester, USA, 4)  
11     Current affiliation: University of California Davis; Davis, USA, 5) University of California  
12     Santa Barbara; Santa Barbara, USA, 6) Current affiliation: Auburn University; Auburn, USA,  
13     7) Arizona State University; Tempe, USA, 8) Current affiliation: Florida Atlantic University;  
14     Boca Raton, USA, \*Corresponding author: dieter\_lukas@eva.mpg.de

15     **This article is based on the following preregistration:** Logan CJ, McCune KB, Chen  
16     N, Lukas D. 2020. **Implementing a rapid geographic range expansion - the role of**  
17     **behavior and habitat changes**

18     **which has been pre-study peer reviewed and received an In Principle Rec-**  
19     **ommendation by:** Esther Sebastián González (2020) The role of behavior and habi-  
20     tatt availability on species geographic expansion. *Peer Community in Ecology*, 100062.  
21     [10.24072/pci.ecology.100062](https://doi.org/10.24072/pci.ecology.100062). Reviewers: Caroline Nieberding, Tim Parker, and Pizza Ka Yee  
22     Chow



<sup>24</sup> **ABSTRACT**

<sup>25</sup> It is generally thought that behavioral flexibility, the ability to change behavior when circum-  
<sup>26</sup> stances change, plays an important role in the ability of a species to rapidly expand their  
<sup>27</sup> geographic range. To expand into new areas, individuals might specifically show flexibility in  
<sup>28</sup> dispersal behavior, their movement away from their parents to where they themselves repro-  
<sup>29</sup> duce. Great-tailed grackles (*Quiscalus mexicanus*) are a bird species that is rapidly expanding  
<sup>30</sup> its geographic range and are behaviorally flexible. Here, we infer dispersal rates in wild-caught  
<sup>31</sup> grackles from two populations across their range (an older population in the middle of the  
<sup>32</sup> northern expansion front in Arizona nearer the core of their original range versus a young  
<sup>33</sup> population on the northern edge of the expansion front in California) to investigate whether  
<sup>34</sup> grackles show flexibility in their dispersal behavior between these two populations. Based on  
<sup>35</sup> genetic relatedness, we observe no closely related pairs of individuals at the edge, suggesting  
<sup>36</sup> that individuals of both sexes disperse further from their parents and siblings in this population  
<sup>37</sup> than in the population nearer the core. Our analyses also suggest that, in both populations,  
<sup>38</sup> females generally move shorter distances from where they hatched than males. These results  
<sup>39</sup> elucidate that the rapid geographic range expansion of great-tailed grackles is associated with  
<sup>40</sup> individuals, in particular females, differentially expressing dispersal behaviors.

<sup>41</sup> **INTRODUCTION**

<sup>42</sup> Where individuals live, and how this area changes over time, is shaped by the behavior of these  
<sup>43</sup> individuals (Martin, 1998). It has been argued that behavioral flexibility, the ability to change  
<sup>44</sup> behavior when circumstances change (see Mikhalevich et al., 2017 for theoretical background  
<sup>45</sup> on our flexibility definition), plays an important role in the ability of a species to rapidly expand  
<sup>46</sup> their geographic range (Chow et al., 2016; Griffin & Guez, 2014; e.g., Lefebvre et al., 1997; Sol  
<sup>47</sup> et al., 2002, 2005, 2007; Sol & Lefebvre, 2003). One specific behavior where flexibility could  
<sup>48</sup> be crucial at the expanding edge of a population range is dispersal, the movement individuals  
<sup>49</sup> show from the place they were born to the place they reproduce. At the edge, individuals  
<sup>50</sup> might have to travel different distances to reach suitable habitat than nearer the core of their  
<sup>51</sup> range (Simmons & Thomas, 2004) or coordinate to move to a place where they can find a mate  
<sup>52</sup> (Bocedi et al., 2014). Here, we determine whether we can infer flexible changes in dispersal  
<sup>53</sup> behavior in a population at the edge of a rapidly expanding avian species.

<sup>54</sup> There are two flexible changes in dispersal behavior that could be expected at the edge of  
<sup>55</sup> a population's geographic range because they are linked to the expansion rate. First, some  
<sup>56</sup> individuals might disperse longer distances than they would in more established populations  
<sup>57</sup> (Trakhtenbrot et al., 2005). Such increased dispersal distances could facilitate movement across  
<sup>58</sup> uninhabitable environments and bring together a larger number of individuals from different  
<sup>59</sup> areas within the original range. Second, there might be changes in sex-specific dispersal. In  
<sup>60</sup> most animal populations, individuals of one sex tend to disperse less than individuals of the  
<sup>61</sup> opposite sex (Pusey, 1987). This sex bias in dispersal could limit expansion into novel areas,

62 particularly if females do not disperse but remain close to the place they were born (Miller et  
63 al., 2011; Santini et al., 2025). Shorter dispersal distances in females would reduce expansion  
64 rates because most individuals in a novel area would be male vagrants, whereas females and  
65 their offspring would remain in or closer to the range core. Reduced dispersal of females  
66 slows the establishment of breeding populations in the range edge, thereby limiting further  
67 movements beyond the current range edge. Accordingly, an extension of a range might occur  
68 more rapidly in species in which individuals show flexibility in sex specific dispersal biases,  
69 with individuals of both sexes dispersing farther at the edge of the range.

70 In this study, we test whether flexibility in dispersal behavior might have played a role in the  
71 rapid geographic range expansion of great-tailed grackles (*Quiscalus mexicanus*). Great-tailed  
72 grackles show behavioral flexibility in experimental tasks (Logan, 2016) and continue to rapidly  
73 expand their geographic range (Wehtje, 2003), thus offering an opportunity to assess the role  
74 of changes in dispersal behavior during the course of their expansion. Great-tailed grackles  
75 are seasonal breeders, with most young hatched between April to June, and individuals can  
76 begin breeding the year after hatching (Selander & Hauser, 1965). Individuals often forage  
77 in larger geographical areas during the winter before moving to the breeding sites, which is  
78 also when they might disperse. We previously found in the population nearer the core, with a  
79 smaller dataset, that great-tailed grackles show sex biases in dispersal, with females dispersing  
80 shorter distances than males (Sevchik et al., 2022). We also found indications of changes in  
81 the behavior of great-tailed grackles across their range. Compared to the population nearer  
82 to the core, individuals in the population nearer the edge are more persistent, participating in  
83 more of the trials they were offered and often returning to a problem even they did not receive  
84 a reward, and have a higher variance in behavioral flexibility in a reversal learning task (Logan  
85 et al., 2023). We therefore expect flexibility in dispersal behavior between the two populations,  
86 with no sex biases in dispersal at the edge of the population expansion.

87 We compare the dispersal behavior of great-tailed grackles between a recently established  
88 edge population (California) and a population that has existed for several generations nearer  
89 the core of their original range (Arizona)(Table 1). We examine whether individuals in the  
90 edge population are more likely to move away from the location where they hatched than  
91 individuals in the population nearer the core, and, in particular, whether we observe less sex  
92 bias in dispersal in the edge population. We infer rates of dispersal by determining whether the  
93 average relatedness (calculated using single nucleotide polymorphisms, SNPs) between pairs of  
94 individuals in each population is lower or higher than what we would expect if individuals move  
95 randomly (Sevchik et al., 2022). Genetic approaches are one of the main ways to determine  
96 patterns of dispersal in birds because actual dispersal events are rare and difficult to study.  
97 In most species, individuals only show limited movement from their place of origin to where  
98 they themselves breed, creating patterns of isolation by distance as pairs of individuals are less  
99 likely to share genetic variants the further away they are (Manel et al., 2003). We measure this  
100 sharing of genetic variants using relatedness (Spong & Creel, 2001), with individuals who show  
101 low relatedness to others in the population being assumed to have dispersed further (Aguillon  
102 et al., 2017). Though we refer to our two sampling sites as two populations, it is important to  
103 note that the distribution of great-tailed grackles appears continuous and connected. Even at

104 the edge, the expansion appears to occur gradually, rather than through the establishment of  
105 separate, distanced populations (Wehtje, 2003). Accordingly, our focus is not on the overall  
106 levels of genetic diversity or relatedness in our sample of individuals, but how genetic variation  
107 at each site is structured according to the sex of individuals and the distances between them  
108 to lead to patterns of relatedness. Due to sampling limitations, our inferences are based on a  
109 single comparison between two populations. Accordingly, we cannot directly infer the causes  
110 of any changes in dispersal behavior. However, we can infer whether there is flexibility in  
111 dispersal behavior in this expanding species if we detect differences in relatedness patterns  
112 between the two populations.

## 113 RESEARCH QUESTION

114 Our research question is listed as it appeared in the preregistration (Logan et al., 2020). We  
115 completed the research and published the results linked to the other research questions in  
116 other articles (Logan et al., 2023, 2025; Summers et al., 2023)

117 **Q2 (dispersal behavior: great-tailed grackles): Are there differences in dispersal behavior  
118 across the great-tailed grackle's geographic range? (Fig. 1, Table 1)**

119 **Prediction 2:** We predict **more dispersal at the edge**: a higher proportion of individuals,  
120 particularly females, which is the sex that appears to disperse less in the population in  
121 the middle of the range expansion (Sevchik et al., 2022), disperse in a more recently estab-  
122 lished population and, accordingly, fewer individuals are closely related to each other. *This  
123 would support the hypothesis* that changes in dispersal behavior are involved in the great-tailed  
124 grackle's geographic range expansion.

125 **Prediction 2 alternative 1:** We predict that the **proportion of individuals dispersing is  
126 not related to when the population established** at a particular site and, accordingly, the  
127 average relatedness is similar across populations. *This supports the hypothesis* that the original  
128 dispersal behavior was already well adapted in this species to facilitate a range expansion.

129 **Table 1.** Population characteristics of the two field sites. The number of generations at a  
130 site is based on a generation length of 5.6 years for this species (BirdLife International, 2018,  
131 note that this species starts breeding at age 1), and on the first year in which this species  
132 was reported or estimated to breed at each location. Woodland, California: Yolo Audubon  
133 Society's newsletter *The Burrowing Owl* (July 2004); Tempe, Arizona: based on 1945 first-  
134 sighting report in nearby Phoenix, Arizona (Wehtje, 2004) to which we added 6 years, which  
135 is the average time between first-sighting and first-breeding - see Table 3 in (Wehtje, 2003).

Site	Range position	Breeding since	Number of years breeding	Average number of generations
Tempe, Arizona	Middle of expansion	1951	69	12.3
Woodland, California	Northern edge	2004	16	2.9

### **136 STATE OF THE DATA AND CHANGES FROM PREREGISTRATION**

137 The preregistration was written (March 2020) prior to collecting any data from the edge  
 138 population (which began in January 2021), therefore we were blind to these data. However,  
 139 we were not blind to some of the data from the Arizona population: some of the relatedness  
 140 data (SNPs used for Hypothesis 2 to quantify relatedness to infer whether individuals disperse  
 141 away from relatives) from the middle population (Arizona) had already been analyzed for other  
 142 purposes (n=57 individuals, see Sevchik et al., 2022). Therefore, we consider it secondary data:  
 143 data that were collected for other investigations. We collected blood samples from 37 more  
 144 grackles in Arizona, and we repeated the analyses for the Arizona population with the complete  
 145 sample. This preregistration was submitted in May 2020 to PCI Ecology for pre-study peer  
 146 review. We received the reviews, and revised and resubmitted in August 2020, and it passed  
 147 pre-study peer review in October 2020.

148 While our ideal plan was to include three field sites, due to restrictions around COVID-19 and  
 149 because we learned about potential risks to the safety of study participants at the initially  
 150 considered third field site, it was not possible for us to accomplish all of our goals within our  
 151 current funding period. We therefore compare only two populations.

## **152 METHODS**

### **153 Sample**

154 **Q2:** Great-tailed grackles were caught in the wild in Tempe, Arizona, and in Woodland and  
 155 Sacramento, California. Adults were identified from their eye color, which changes from brown  
 156 to yellow upon reaching adulthood (Johnson & Peer, 2001). We applied colored leg bands  
 157 in unique combinations for individual identification. Some individuals (~20) were brought  
 158 temporarily into aviaries for behavioral choice tests, and then were released back to the wild  
 159 at their point of capture. We caught grackles with a variety of methods (e.g., walk-in traps,  
 160 mist nets, bow nets), some of which decrease the likelihood of a selection bias for exploratory  
 161 and bold individuals because grackles cannot see the traps (i.e., mist nets).

162 We recorded the latitude and longitude at locations where each grackle was caught. We also  
 163 noted the location where we first resighted the individual after it was released back into the  
 164 wild. We always attempted to resight the individual the day after its release because this  
 165 location is less influenced by the food we put at trapping locations. We used the resight

location for all individuals, but we used the trapping location if we were unable to find them again after release. For Figure 1, we mapped these locations using QGIS (version 3.4.5) and the Google Satellite basemap obtained through QuickMapServices QGIS plugin (Map data ©2020, Google). We used the Categorize point properties feature to distinguish each individual's point on the map based on whether they were adult or juvenile and male or female.

### Sample size rationale

We caught as many great-tailed grackles as we could during the two to three years we spent at each site given that the birds were only brought into the aviaries during the non-breeding season (approximately September through March). We sampled more than the expected 20 grackles per site for the genetic analyses.

### Protocols and open materials

DNA was collected from the grackles, processed, and analyzed for pairwise relatedness using ddRADseq and Stacks as in Sevchik et al. (2022) ([protocol](#)). Our pre-registration only included a brief summary of the methods, we describe them in detail below for full reproducibility.

We previously generated genotypes for 57 individuals from Arizona in 2018 (Sevchik et al., 2022). For the current analyses, we added genotypes for 37 individuals from Arizona and 35 individuals from California. In brief, we collected 150uL of blood from individual birds by brachial or medial metatarsal venipuncture. Samples were centrifuged at 15x gravity for 10 minutes directly after collection to separate the serum from the cellular fraction. The serum layer was removed and 600uL of lysis buffer was added to the remaining packed cells. Tubes containing packed cells and lysis buffer were stored at room temperature for up to 5 years before DNA extraction. The time gap between sampling and extraction could have reduced the quality of the samples, as we observed that some samples had clotted. This potentially explains the reduced SNP recovery rate for one of the samples we had to exclude (see below). DNA was extracted from the samples using the DNeasy Blood and Tissue kit (Qiagen). Extracted DNA samples were shipped with ice packs to the Cornell Lab of Ornithology for ddRAD sequencing in August 2023. The sequencing to generate single-nucleotide polymorphism (SNP; where at a given position in the genome two different bases, alleles, can occur) genotypes was performed at the Cornell University Lab of Ornithology. Fragments were digested with a combination of two restriction enzymes (SbfI-HF and MspI), cleaned, size-selected, amplified using a low-cycle PCR process, and pooled together for sequencing on an Illumina NextSeq500.

We performed the SNP processing and selection as in Thrasher et al. (2018), processing the samples from the two populations separately. For Arizona, we combined the genotypes of the individuals sequenced in 2018 with the genotypes sequenced in 2023 prior to the processing so that all individuals have the same set of alleles to compare for the relatedness estimation. Occurrences of rare alleles are likely to differ among the two separate populations, therefore combining the data from the two populations could potentially lead to ascertainment biases,

where alleles that occur in the population with the larger sample but not in the population with the smaller sample are included as informative whereas alleles that occur only in the population with the smaller sample are excluded. This would bias the relatedness estimation in the smaller population because differences among individuals in this population would be lost. The re-processing means that, for those individuals already included in Sevchik et al. (2022), the genotypes, and the resulting pairwise relatedness estimates, are slightly different compared to those previously estimated. For both populations, loci were considered only if they were present in 95% of the samples ( $r$ ) and had a minimum minor allele frequency of 0.05 (min maf).

To prepare, check, and describe the genotype data (expected heterozygosity, probability of identity), we used functions in the R packages ‘adegenet’ (Jombart, 2008), ‘pegas’ (Paradis, 2010), and ‘popgenutils’ (Tourvas, 2020).

For each population, we calculated the pairwise relatedness among all dyads of adult individuals using the estimator by Queller & Goodnight (1989), which was more robust for our inferences in a subset of the Arizona data (Sevchik et al., 2022), as implemented in the package ‘related’ (Pew et al., 2015) in R. For the relatedness calculation, we only used the genotypes of individuals in the respective population to derive the allele frequencies that set the baseline chance of sharing alleles. That means that, overall, in both populations average relatedness will be close to zero. Individuals who share fewer alleles than expected have a negative relatedness value, while individuals who share more alleles than expected have a positive value. Our focus is not on comparing the overall levels of genetic diversity in the two populations, but whether there is structure in the sharing of alleles that lead to patterns of relatedness among individuals of the same sex. We identified as potential kin those pairs of individuals whose estimated relatedness was equal to or larger than 0.25 (closer relatives, at the level of half-siblings) or larger than 0.125 (distant relatives, at the level of cousins)(Goudet et al., 2018).

We recorded the longitude and latitude of the first locations where individuals were observed after they had been caught and released, or for those individuals who were not resighted, the location where they were trapped. To calculate the geographic distance among pairs of individuals based on these locations, we used the function ‘distm’ in the package ‘geosphere’ (Hijmans, 2022) with the Vincenty ellipsoid great circle function.

## Open data

All data for analyses are available at Edmond (Lukas & Logan, 2024). Raw genotype files are available in the Sequence Read Archive of the National Center for Biotechnology Information (NCBI, accession number: PRJNA658480).

## Blinding during analysis

<sup>239</sup> Blinding is usually not involved in the final analyses because the experimenters collect the data  
<sup>240</sup> (and therefore have seen some form of it) and run the analyses. However, when processing the  
<sup>241</sup> genetic data, the experimenters and the people who conducted the lab work were blind to the  
<sup>242</sup> relatedness amongst the birds.

## <sup>243</sup> ANALYSIS

<sup>244</sup> We did not exclude any data except for instances where missing data made analyses not  
<sup>245</sup> reliable. Samples with a low DNA quantity and quality produce data for only a small number  
<sup>246</sup> of SNP loci. Relatedness estimates are only reliable if they are based on several hundred SNP  
<sup>247</sup> loci (Foroughirad et al., 2019; Wang, 2016), because small numbers of loci can lead to high  
<sup>248</sup> variances in the estimates. Analyses were conducted in R [current version 4.4.3; R Core Team  
<sup>249</sup> (2023)] and Stan (version 2.18, Carpenter et al., 2017). We used functions in the package  
<sup>250</sup> ‘rethinking’ (McElreath, 2020) to construct and summarize the linear models. Following the  
<sup>251</sup> social convention of this approach, we report the 89% compatibility intervals (89% CI) of the  
<sup>252</sup> posterior sample.

<sup>253</sup> Our response variable is the average relatedness between all pairs of individuals within one  
<sup>254</sup> sex. As in Sevchik et al. (2022), we analysed this in two ways: first, as a continuous variable  
<sup>255</sup> ranging between -1 and +1, reflecting average relatedness as whether individuals share more  
<sup>256</sup> or less alleles than expected by chance; and second, as a categorical variable coded as yes/no,  
<sup>257</sup> reflecting whether the average relatedness among a pair of individuals is more or less than the  
<sup>258</sup> threshold that kin are expected to have ( $r$  0.125 and  $r$  0.25). We had planned to include as  
<sup>259</sup> explanatory variables the site diameter, the site sample size, and the number of generations  
<sup>260</sup> at a site. However, because we were able to only obtain samples from two populations, we  
<sup>261</sup> did not include these variables in the models because it would be impossible to say which of  
<sup>262</sup> the factors might explain the site differences (see also our Discussion). We did however use  
<sup>263</sup> the site diameter data to ensure that the two populations were comparable. Permutations  
<sup>264</sup> (i.e., randomly assigning site ID to individuals) and general linear models estimating average  
<sup>265</sup> relatedness of each individual to all others at that site (averagerelatedness  $\sim \alpha[\text{site}]$ ) were used  
<sup>266</sup> to determine whether individuals at one site are more closely related to each other than the  
<sup>267</sup> individuals at another site.

### <sup>268</sup> Comparison of average relatedness between the two populations

<sup>269</sup> We compared the overall levels of average relatedness, as well as the average relatedness among  
<sup>270</sup> the females and among the males, between the population in Arizona and the population in  
<sup>271</sup> California using a linear model:

$$pairwiserelatedness_d \sim Normal(\mu_d, \theta)$$

$$\mu_d = \alpha_{pop[d]} + \beta_{sex[d], pop[d]}$$

$$\alpha_{pop[d]} \sim Normal(0, 1)$$

$$\begin{bmatrix} \beta_{d,1} \\ \beta_{d,2} \end{bmatrix} \sim MVNormal \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, S_{sex} \right)$$

$$S = \begin{pmatrix} \sigma_{sex=1} & 0 \\ 0 & \sigma_{sex=2} \end{pmatrix} R \begin{pmatrix} \sigma_{sex=1} & 0 \\ 0 & \sigma_{sex=2} \end{pmatrix}$$

$$R \sim LKJcorr(4)$$

$$\sigma_{sex} \sim Exponential(1)$$

$$\theta \sim Exponential(1)$$

where the  $pairwiserelatedness_d$ , the relatedness among all pairs of relatives in the two populations  $pop$  of either  $sex$ , is assumed to be distributed according to a normal distribution with mean  $\mu$  and variance  $\theta$ . We assumed that relatedness overall might be higher in one population than the other, and therefore included an interaction between population and sex, such that the intercepts are defined by a two dimensional Gaussian distribution ( $MVNormal$ ) with means of 0, because we separately include the population means as  $\alpha_{pop}$ , and covariance matrices  $S$  reflecting the two  $sexes$ . The covariance matrix,  $S$ , is factored into separate standard deviations,  $\sigma_{sex}$ , and a correlation matrix,  $R$ . The prior for the correlation matrix is set to come from the Lewandowski-Kurowicka-Joe ( $LKJcorr$ ) distribution, and is set to be weakly informative and skeptical of extreme correlations near -1 or 1.

## Comparison of degree of kinship between the two populations

We compared the number of individuals classified as either close or distant relatives in the two populations using a binomial model:

$$kin_d \sim Binomial(1, p_d)$$

$$logit(p_d) = \alpha_{pop[d]} + \beta_{sex[d], pop[d]}$$

$$\alpha_{pop[d]} \sim Normal(0, 1)$$

$$\begin{bmatrix} \beta_{d,1} \\ \beta_{d,2} \end{bmatrix} \sim MVNormal \left( \begin{bmatrix} 0 \\ 0 \end{bmatrix}, S_{sex} \right)$$

$$S = \begin{pmatrix} \sigma_{sex=1} & 0 \\ 0 & \sigma_{sex=2} \end{pmatrix} R \begin{pmatrix} \sigma_{sex=1} & 0 \\ 0 & \sigma_{sex=2} \end{pmatrix}$$

$$R \sim LKJcorr(4)$$

$$\sigma_{sex} \sim Exponential(1)$$

285

286 where the  $kin_d$  reflects whether the relatedness of a given pair of individuals is or is not larger  
287 than the threshold for either close or distant relatives. All remaining terms as above.

288 Second, we compared the presence of kin in the two populations using permutations. Average  
289 relatedness declines the more individuals are included in the calculation (Lukas et al., 2005).  
290 Permutations are a way to account for this by assessing whether any observed differences  
291 remain when comparing the same number of individuals. We randomly took 10,000 draws  
292 of the same number of individuals we had in the California population, which was a smaller  
293 sample, from the genotypes we had in the Arizona population (e.g., randomly drawing 13 of  
294 the female genotypes in Arizona and calculating the number of kin observed in this sample,  
295 before repeating the random draw another 9999 times, each time calculating the number of kin  
296 observed in the sample). We then compared the observed number of kin in California to the  
297 numbers obtained in the 10,000 random samples to assess whether the kinship composition in  
298 California is similar or different to that observed in Arizona

### 299 **Sex biases in dispersal in the two populations**

300 To determine whether, in either or both populations, individuals of one sex are more likely to  
301 disperse farther than individuals of the opposite sex, we first compared the average relatedness  
302 among females to the average relatedness among males in the same population. We performed  
303 10,000 random draws, drawing the same number of individuals from the whole population  
304 as there are females or males in that population, to assess whether the relatedness among  
305 individuals of one sex is different than that observed in a random sample of individuals of the  
306 same size from that population. Next, we determined the geographic distances among those  
307 pairs of individuals identified as potential close or distant kin. We again performed 10,000  
308 draws, drawing the same number as there are kin of that sex from all the females or males in  
309 that population to assess the expected distance among such a sample of same-sex individuals.  
310 If the distances among the 10,000 draws are generally larger than those observed among kin,  
311 then we infer that kin of that sex remain closer together than what would be expected by  
312 chance. Finally, we performed assessments of spatial autocorrelation to link the pairwise  
313 relatedness among individuals of each sex to the geographic distances of their locations.

314 To test whether males and females show different patterns of genetic isolation by geographic  
315 distance, we followed analyses as in Aguillon et al. (2017). In each population, for males and  
316 females separately, we assessed the strength of the association between the the matrices of  
317 average relatedness and of geographic distance using Mantel correlograms with the function  
318 ‘mantel.correlog’ in the ‘vegan’ package (Oksanen et al., 2013) in R. For each of the four  
319 associations (two sexes in two populations), we performed 10,000 permutations to assess the  
320 strength of the association. The approach involves partitioning the geographic locations into  
321 a series of discrete distance classes. We used two methods to create the distance classes.  
322 First, we attempted to have about equal numbers of pairs of individuals within each distance  
323 class, creating nine distance classes of (0-100m, 100-200m, 200-300m, 300-400m, 400-500m,  
324 500-750m, 750-1000m, 1 000-1250m, and 1250-2000m). With the second method, we only

325 created two distance classes to increase the sample size in each class, splitting the distance  
326 according to the limit at which most close kin were detected (0-400m and 400-1400m). For each  
327 distance class, a normalized Mantel statistic is calculated using permutations of values within  
328 that distance class. The permutation statistics, plotted against distance classes, produce  
329 a multivariate correlogram. A negative correlation between genetic relatedness and spatial  
330 distance indicates that the more closely related individuals are found closer to each other,  
331 indicating that these individuals likely disperse shorter distances than those individuals where  
332 a positive correlation is found.

## 333 RESULTS

### 334 Summary statistics

#### 335 California SNP data

336 We retained 493 SNPs. Data was missing for 3.3% of all alleles (individuals missing information  
337 for either one or both of their chromosomes for that particular position). None of the SNPs  
338 showed a particular underrepresentation of information. The missingness was due to the  
339 incomplete genotype of one individual (C116RY, adult male), who had missing data at 459 of  
340 the 493 SNPs (93%), whereas all other individuals had data missing at four or fewer SNPs.  
341 We excluded this individual from the further analyses, because relatedness calculations based  
342 on so few SNPs were, as expected, highly stochastic and led to extreme deviations (see code  
343 chunk ‘kin composition’ in the Rmd file for illustration). For the remaining individuals, all  
344 SNPs had two alleles and the observed heterozygosity (individuals carrying one copy each  
345 of the two bases) was 0.29, identical to the heterozygosity expected in a population with the  
346 same allele frequencies and random mating. The probability of identity for siblings, the chance  
347 that two siblings will show the same genotypes given the allele frequencies across these 493  
348 SNP loci and random mating among individuals, is less than  $10^{-64}$ . This indicates that any  
349 relatedness we detect among individuals is likely to reflect biological relatedness, rather than  
350 resulting from limited sampling making individuals more similar.

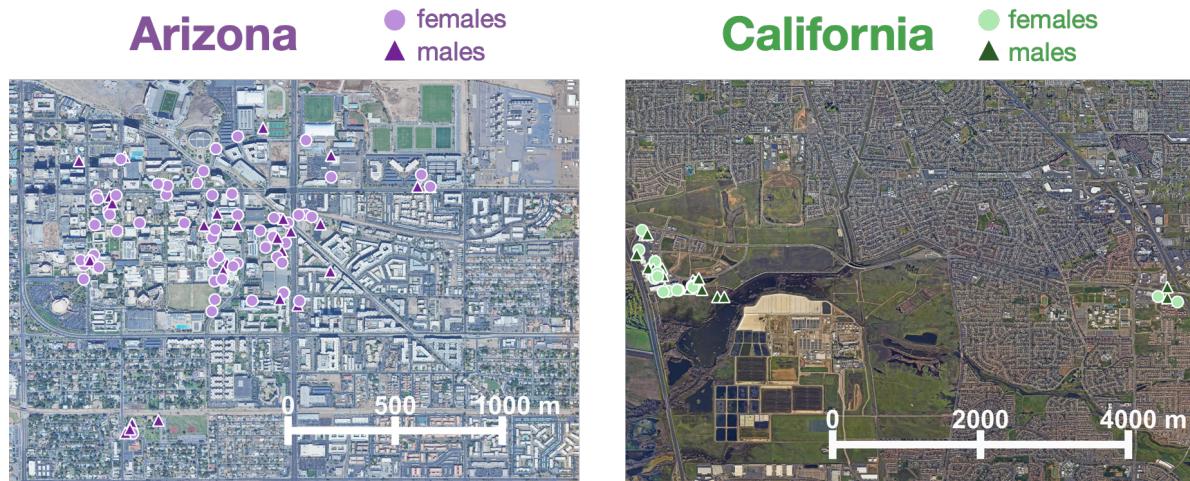
#### 351 Arizona SNP data

352 We retained 462 SNPs. Data was missing for 3.0% of all alleles (individuals missing information  
353 for either one or both of their chromosomes for that particular position). None of the SNPs  
354 showed a particular underrepresentation of information. There were three individuals whose  
355 genotypes were less complete (A072KB, adult female, missing data at 191 (41%) of SNPs;  
356 A088YR, adult male, missing data at 174 (38%) of SNPs; A059NB, adult female, at 148 (32%)  
357 of SNPs), whereas all other individuals had data missing at less than 10% of SNPs. Here,  
358 we did not exclude any individuals because the number of SNPs with information was still  
359 sufficiently high for all genotypes to reduce the noise in the relatedness estimation. All SNPs  
360 had two alleles and the observed heterozygosity (individuals carrying one copy each of the two  
361 bases) was 0.29, similar to the heterozygosity expected in a population with the same allele  
362 frequencies and random mating. The probability of identity for siblings, the chance that two

363 siblings will show the same genotypes given the allele frequencies across these 462 SNP loci  
364 and random mating among individuals, is less than  $10^{-60}$ .

365 *Sample*

366 In total, we included genotype information for 52 females and 27 males in Arizona, and 13  
367 females and 15 males in California. In Arizona, all birds were found within a maximum of  
368 1,991m from each other (median 669m). In California, birds were found at multiple locations  
369 (Figure 1). Twelve females and twelve males were found at a location in Sacramento that  
370 spanned about the same range as the population in Arizona (maximum geographic distance  
371 1,592m, median 474m). Three birds (one female, two males) were resighted at a separate  
372 location in Sacramento ~7,000m away from the main location. In addition, one male was  
373 trapped and resighted at a location ~33,000m away in Woodland. Therefore, the maximum  
374 and average geographic distances between the locations of individuals are much higher for  
375 the California sample than the Arizona sample. For the set of analyses that include pairwise  
376 geographic distances among individuals, we performed the analyses only with the birds found  
377 at the single location in Sacramento in order to keep the California population comparable to  
378 the Arizona population (i.e., we excluded these four birds).



379  
380 Figure 1: Maps showing where individuals were located in Arizona (left, purple) and in Cal-  
381 ifornia (right, green). In both populations, most individuals were sampled within 1,500m of  
382 each other, with females (lighter color circles) and males (darker color triangles) being found  
383 throughout the populations. For the analyses incorporating location data in California, we  
384 only used the birds from the cluster on the left and excluded the individuals to the far right  
385 and the individuals used in a separate location (Woodland). Excluding these individuals makes  
386 the distribution more comparable between the two populations (see also Figure 2). We wanted  
387 to show here that individuals in the population in California appear to roam across larger areas  
388 - individuals on the far right were sighted together with the individuals on the cluster to the  
389 left.

390 **Difference in dispersal behavior between the two populations**

391 *Comparison of average relatedness in the two populations*

392 Overall, the average relatedness among individuals in the two populations is slightly negative,  
393 which is more pronounced in California (average relatedness: Arizona -0.013, California -  
394 0.037). This slight skew toward negative relatedness values suggests that both populations,  
395 but particularly the population in California, might contain individuals who have immigrated  
396 into these populations and are therefore sharing fewer alleles than would be expected by chance.  
397 In Arizona, males (-0.009) have slightly higher average relatedness than females (-0.013). In  
398 California, females (-0.024) have higher average relatedness than males (-0.048).

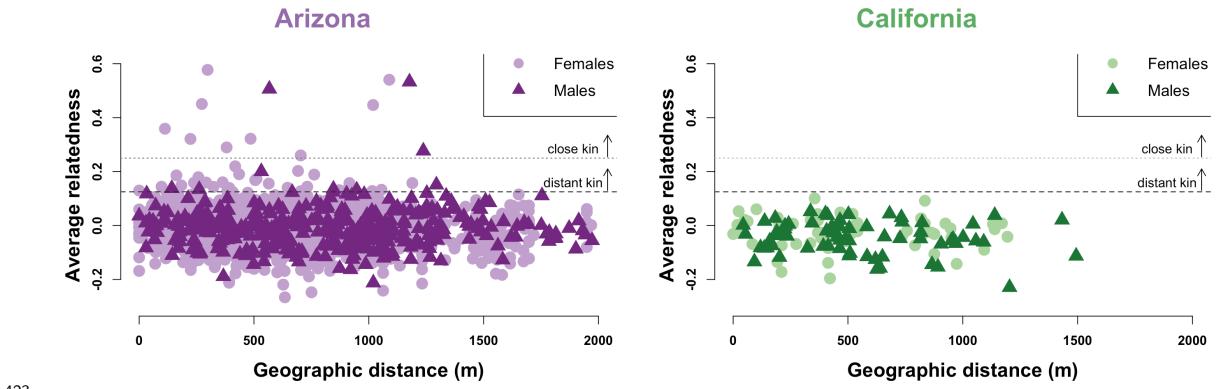
399 The model comparing levels of pairwise relatedness between the two populations indicates  
400 that the values in California are consistently lower than the values in Arizona because their  
401 confidence intervals do not cross zero (median of contrast for females -0.003, 89%CI of contrast  
402 -0.006 to -0.001, for males median -0.009, 89% CI -0.011 to -0.007).

403 *Comparison of the degree of likely kin in the two populations*

404 Overall, in both populations we identified very low numbers of dyads that are potentially kin  
405 (Figure 2). In California, none of the dyads are estimated to be related with  $r \geq 0.25$ , and only  
406 one opposite sex dyad is estimated to be related with  $r \geq 0.125$  (out of 105 male-male dyads, 78  
407 female-female dyads, and 195 opposite sex dyads). In Arizona, 3 male-male dyads (0.9% of the  
408 351 male-male dyads), 12 opposite sex dyads (0.9% of the 1404 dyads), and 9 female-female  
409 dyads (0.7% of 1326 dyads) are estimated to be related with  $r \geq 0.25$ . With the lower threshold  
410 of  $r \geq 0.125$ , 9 (2.5%) of all male-male dyads, 39 (2.9%) of female-female dyads, and 32 (2.3%)  
411 of opposite sex dyads in Arizona are classified as related.

412 A binomial model indicates that the probability that any dyad would be kin at  $r \geq 0.25$  is higher  
413 among individuals in Arizona than in California (median posterior estimate of difference in  
414 probabilities for close kin  $r \geq 0.25$ : 0.6%, 89% CI 0.4 to 1.0%, for more distant kin  $r \geq 0.125$  median  
415 difference 2.2%, 89% CI 1.6 to 2.9%). The differences in probability hold for both female-  
416 female dyads ( $r \geq 0.25$ : 0.7, 89% CI 0.3 to 1.0%;  $r \geq 0.125$ : 2.5%, 89% CI 1.7 to 3.3%) and for  
417 male-male dyads ( $r \geq 0.25$ : 0.6%, 89% CI 0.2 to 1.1%;  $r \geq 0.125$ : 2.3%, 89% CI 1.2 to 3.2%).

418 The permutations support that the absence of same-sex individuals related at  $r \geq 0.125$  in the  
419 California population is not simply due to the smaller sample of individuals. There are no  
420 relatives in only 12% of permutations drawing 13 individuals from among the 52 females in  
421 Arizona, and there are no relatives in only 2% of permutations drawing 15 individuals from  
422 among the 27 males in Arizona.



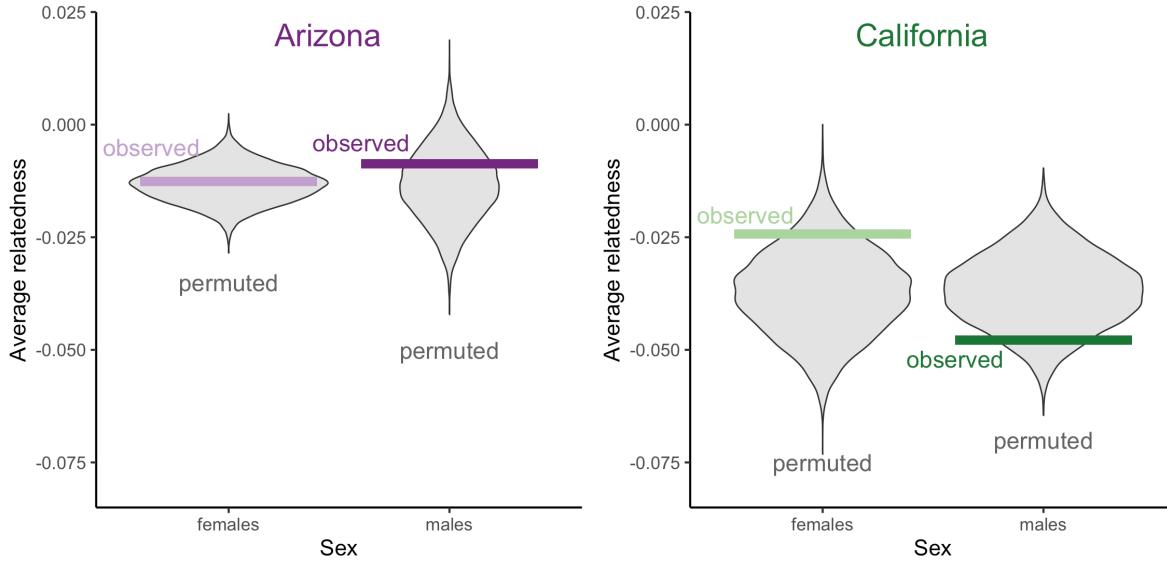
423

424 Figure 2: The relatedness and geographic distance observed among grackles in Arizona (left, purple)  
 425 and in California (right, green). Average relatedness is higher in Arizona than in  
 426 California. In Arizona, there are several pairs of females (lighter circles) and males (darker  
 427 triangles) who are related at levels higher than 0.25 (dotted line, close kin of half-sibling or  
 428 closer) and 0.125 (dashed line, distant kin of cousin or closer), while there are no close or  
 429 distant same-sex kin in California. In terms of potential sex-biases in dispersal, in Arizona,  
 430 we observe more females than males related at levels of distant kin and of close kin. Closely  
 431 related females tend to be found at shorter geographic distances than the average female pair,  
 432 while closely related males are not found at short geographic distances.

### 433 Sex biases in dispersal in the two populations

#### 434 Average relatedness within the sexes

435 Average relatedness among both the females and the males in Arizona is not different from what  
 436 would be expected by chance. Randomly drawing the same number of individuals from the  
 437 full sample gives an average relatedness that is lower than that observed among the females in  
 438 45% of permutations and for males in 29% of permutations. In contrast, the observed average  
 439 relatedness among females in California is slightly higher than what would be expected by  
 440 chance, with 90% of the permutations drawing the same number of individuals from the  
 441 overall population as there are females leading to lower average relatedness than that observed  
 442 among the females (Figure 3). In contrast, the observed average relatedness among males is  
 443 slightly less than what would be expected by chance given the relatedness among individuals  
 444 in this population, with 91% of permutations giving higher levels of average relatedness than  
 445 that observed among the males (Figure 3).

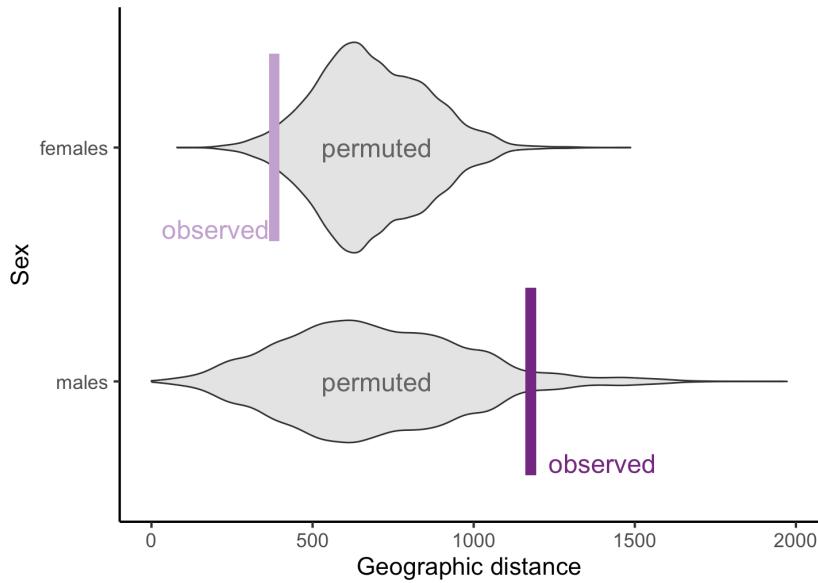


446

447 Figure 3: Observed levels of average relatedness among females and males in the two popu-  
 448 lations compared to expected levels of relatedness based on permutations. Overall, average  
 449 relatedness is lower in California (right panel, green lines) than in Arizona (left panel, purple  
 450 lines). In Arizona, the observed average relatedness among females (light purple line, left)  
 451 and males (dark purple line, right) are similar and fall within the expected levels of average  
 452 relatedness based on permutations (grey shaded areas). In California, the average relatedness  
 453 observed among females (light green line, left) is higher than the average relatedness observed  
 454 in most permutations, while the average relatedness observed among males (dark green line,  
 455 right) is lower than the relatedness observed in most permutations. 10,000 permutations were  
 456 performed for each sex in each population, drawing a subset of individuals from the sample  
 457 of individuals in that population and calculating their average relatedness. The width of the  
 458 violin plots (grey shaded areas) reflects the number of permutations during which a particular  
 459 level of average relatedness was observed.

460 *Distances among kin of the different sexes*

461 In Arizona, the only population where our relatedness calculations indicated likely kin, females  
 462 related at  $r = 0.25$  are found a median of 391m from each other, while males related at this level  
 463 are found 1,177m from each other (Figure 2). Similarly, females related at  $r = 0.125$  are found a  
 464 median of 435m from each other, while males related at this level are found 846m from each  
 465 other. These differences in distance are not due to females generally being found closer to  
 466 each other, because in only 2% of permutations drawing from the same number of female and  
 467 male dyads as those that are related at the respective levels, are the differences in distance as  
 468 large or larger than those observed (Figure 4). We cannot perform this analysis for California  
 469 because there are no kin among either sex (Figure 2).



470

471 Figure 4: In Arizona, females related at  $r = 0.25$  (light purple line, top) are found closer to each  
 472 other than the average set of females (grey shaded area, top), while males related at  $r = 0.25$   
 473 (dark purple line, bottom) are found at larger distances from each other than the average set  
 474 of males (grey shaded area, bottom). We cannot perform this analysis for California, because  
 475 none of the same-sex dyads are related at  $r = 0.25$ . 10,000 permutations were performed for each  
 476 sex, drawing a subset of individuals from the total sample matching the number of close kin  
 477 and calculating their average geographic distance. The width of the violin plot (grey shaded  
 478 area) reflects the number of permutations during which a particular average distance was  
 479 observed.

480 *Spatial autocorrelation between geographic distance and relatedness in the two sexes*

481 The spatial autocorrelation analyses indicate that, in both Arizona and California, female  
 482 relatives likely stay close to each other while male relatives move away from each other (Figure  
 483 2). For Arizona, more closely related females are found at shorter distances from each other  
 484 (negative correlation between relatedness and geographic distance based on a Mantel test when  
 485 females are close,  $-0.08$ ,  $p=0.02$ ; positive correlation when females are far,  $0.06$ ,  $p=0.03$ ). In  
 486 contrast, at short distances males are not related to each other ( $0.05$ ,  $p=0.21$ ), but relatives are  
 487 found at larger distances ( $-0.11$ ,  $p=0.02$ ). The same pattern is found for females in California,  
 488 though with the smaller sample size, the effects are not significant (females close  $-0.15$ ,  $p=0.12$ ,  
 489 distant  $0.15$ ,  $p=0.12$ ), whereas for males there is no obvious pattern (close  $0.04$ ,  $p=0.38$ ; distant  
 490  $0.04$ ,  $p=0.43$ ). These results are similar when using the larger number of distance classes, with  
 491 correlations switching from negative to positive for females as distance increases, and from  
 492 positive to negative for males.

493 **DISCUSSION**

494 Our results provide support for our prediction that natal dispersal is higher in great-tailed  
495 grackle populations that are closer to the edge of the expansion range. We find that the average  
496 levels of relatedness, as well as the number of pairs of same-sex individuals that are closely  
497 related are lower in the population in California than in the population in Arizona. Grackles  
498 have been breeding since 2004 in California and since 1951 in Arizona. Our analyses suggest  
499 that the observed differences between the two populations in the levels of relatedness are  
500 unlikely to be simply due to the larger sample of individuals included in the Arizona population.  
501 While the results support our main prediction, further assessment of the hypothesis that  
502 individuals in edge populations behave differently than those nearer the core of the range is  
503 required, because our inferences rely on only a single comparison between two populations that  
504 might also differ in other aspects besides the age at which they were established. Independent  
505 of the potential aspects involved, our findings do reveal flexibility in dispersal behavior in this  
506 species that is rapidly expanding its range. We also find that, in both populations, females  
507 are more likely to remain closer to same-sex relatives than males, suggesting that females  
508 disperse shorter distances than males. These findings, with our larger sample from this article,  
509 confirm our previous inferences for the population in Arizona (Sevchik et al., 2022), that the  
510 sex biases in dispersal in great-tailed grackles are the opposite to that observed in most other  
511 bird species.

512 In the population closer to the edge of the range in California, our relatedness analyses indicate  
513 that no pair of same-sex individuals is related at the level of cousins ( $r = 0.125$ ) or higher. Our  
514 inference is based on a relatively small sample of 13 females and 15 males, which is nevertheless  
515 larger than the minimum sample size set in our preregistration. In addition, all analyses  
516 suggest that the low relatedness, and in particular the absence of related same sex dyads,  
517 is unexpected given the levels of relatedness we observe among the individuals in Arizona.  
518 While the permutation analysis suggests that there might be a chance to observe no female  
519 relatives in such a sample, this approach is limited because it does not fully take into account  
520 the potential contingencies in the observed data (for example, if a mother is present with two  
521 daughters, these dyads are not independent). We therefore performed an unregistered post-  
522 hoc analysis using a social network approach that accounts for such potential interdependence  
523 using functions of the package ‘STRAND’ in R (Ross et al., 2024). We coded whether a given  
524 pair of individuals in either population was likely kin or not ( $r = 0.125$ ) and determined whether  
525 the likelihood that individuals are in a kin dyad is different between the two populations.  
526 These models also indicate that the likelihood that individuals in California are closely related  
527 is substantially lower than that of individuals in Arizona (for all individuals: 89%CI estimate  
528 of difference in likelihood 0.94 to 6.81; for females 89% CI -2.36 to 4.79; for males 89% CI -0.04  
529 to 6.16). Our results suggest that beyond the radius that we sampled, California individuals  
530 of both sexes disperse further from where they hatched than individuals in Arizona.

531 Previous theoretical and empirical studies predict such increased dispersal at the edge of a  
532 population expansion. Multiple processes could contribute to the higher dispersal at the edge

533 of the population expansion. The higher frequency of dispersers at the edge could result from  
534 simple sorting processes, whereby highly dispersive individuals are over-represented in edge  
535 populations because they are more likely to end up in these novel areas (Shine et al., 2011;  
536 Travis & Dytham, 2002). Alternatively, or in addition, the conditions at the edge could shift  
537 the trade-off of the costs and benefits towards dispersing in the edge population (Chuang &  
538 Peterson, 2016; Simmons & Thomas, 2004). Such trade-offs linked to expansion have been  
539 observed in relation to dispersal of aggressive individuals in bluebirds (Duckworth & Badyaev,  
540 2007) and morphological adaptation for speed in cane toads (Clarke et al., 2019).

541 In the population in Arizona, we observed a small number of pairs of individuals related at  
542 the level of cousins or higher. However, while only 3% of all dyads are related at r 0.125, 60%  
543 of females (31 out of the 52) and 56% of males (15 out of the 27) have at least one same-sex  
544 relative in the population. The kin composition we observed among great-tailed grackles in  
545 Arizona is similar to what has been reported for ravens, where 2.2% of dyads were classified  
546 as close kin and 20% of individuals had a close kin in their foraging group (the study used  
547 a different approach to estimate relatedness so the category is between our cut-offs of 0.25  
548 and 0.125; Parker et al. (1994)). The raven study also suggested that kinship, besides parent-  
549 offspring relations, did not play a major role in structuring social interactions. Both ravens and  
550 grackles form foraging groups, where individuals are generally resighted at the same location  
551 with the same set of others. However, groups are not closed and cohesive, unlike the stable  
552 groups found in cooperatively breeding birds or several social mammals, where levels of kinship  
553 are generally higher than what we observed here and kinship plays an important role in social  
554 relationships (Pereira et al., 2023).

555 In both populations, we find indications of a sex bias in dispersal, with females apparently  
556 dispersing shorter distances than males. Despite the absence of close relatives in California,  
557 the analyses linking relatedness to geographic distance also supports a similar bias in this  
558 population. This confirms our previous conclusion with a smaller sample in Arizona (Sevchik  
559 et al., 2022). We find more male relatives in Arizona than in our earlier study that used a subset  
560 of these Arizona individuals (Sevchik et al., 2022). This indicates that, while males disperse  
561 more than females, they apparently do not move much further than the distances involved in  
562 our sampling areas (2,000m). With our approach, we cannot track individual movements. Sex  
563 biases in dispersal could either arise because, on average, all males move larger distances than  
564 all females. Alternatively, differences could arise because a higher frequency of males compared  
565 to females disperse, even though, when they disperse, both males and females move similar  
566 distances (Sutherland et al., 2000). The male bias in dispersal also matches with observational  
567 reports of which individuals are first observed at the edge of the range expansion. An earlier  
568 study found that, of the first sightings of a great-tailed grackle in a new location, where the  
569 sex of the individual was reported, the pioneer individual was a male in 65% of instances  
570 (Dinsmore & Dinsmore, 1993).

571 The male sex bias in dispersal, and the reduced sex bias in the edge population where females  
572 also appear to disperse more, might interact with the ongoing range expansion of the great-  
573 tailed grackles. In most sexually reproducing species, the distribution and movement of females

574 determines the range limits (Miller & Compagnoni, 2022). Particularly in species where single  
575 males mate with multiple females, as in the great-tailed grackles, we would expect that mod-  
576 erate levels of female-biased dispersal would increase the range expansion speed because this  
577 would lead to the sex ratio of multiple females per mating male in the new populations (Miller  
578 et al., 2011). Accordingly, the adaptability of dispersal behavior in grackles, with both sexes  
579 showing more dispersal at the edge than nearer the core, might contribute to their ability to  
580 expand into new areas.

581 These results elucidate that the rapid geographic range expansion of great-tailed grackles is  
582 associated with individuals differentially expressing dispersal behaviors. Additional studies  
583 are needed to determine the robustness and potential mechanisms involved in finding different  
584 dispersal behavior in an edge population. It is not clear whether these differences reflect the  
585 particular conditions of edge populations or other ecological conditions that could also influence  
586 dispersal behavior. The differences in dispersal behavior could also be linked to differences  
587 in other behavior, rather than directly reflecting a response to whether the individuals are at  
588 the edge or nearer the core of the distribution. Our previous comparison of several behaviors  
589 indicated higher persistence and interindividual differences in behavioral flexibility in the edge  
590 population, but no differences in average exploration, innovativeness, or behavioral flexibility  
591 (Logan et al., 2023). We hope that our findings will stimulate additional studies into the traits  
592 that characterize individuals and populations at the edge of population range expansions.

593 **ETHICS**

594 This research is carried out in accordance with permits from the:

- 595 1) US Fish and Wildlife Service (scientific collecting permit number MB76700A-0,1,2)
- 596 2) US Geological Survey Bird Banding Laboratory (federal bird banding permit number  
597 23872)
- 598 3) Arizona Game and Fish Department (scientific collecting license number SP594338 [2017],  
599 SP606267 [2018], SP639866 [2019], and SP402153 [2020])
- 600 4) Institutional Animal Care and Use Committee at Arizona State University (protocol  
601 number 17-1594R)
- 602 5) California Department of Fish and Wildlife (scientific collecting permit [specific use]  
603 number S-192100001-19210-001)
- 604 6) Institutional Animal Care and Use Committee at the University of California Santa  
605 Barbara (protocol number 958)
- 606 7) RegionalSan access permit (number AP 2021-01)

607 **AUTHOR CONTRIBUTIONS**

608 **Dieter Lukas:** Hypothesis development, data analysis and interpretation, write up, revising/  
609 editing

610 **Aaron Blackwell:** data collection, data interpretation, revising/editing

611 **Maryam Edrisi:** data collection, revising/editing

612 **Kristin Hardy:** data collection, revising/editing

613 **Christa LeGrande:** data collection, revising/editing

614 **Zara Marfori:** data collection, revising/editing

615 **Kelsey McCune:** data collection, data interpretation, revising/editing

616 **August Sevchik:** data collection, data interpretation, revising/editing

617 **Caroline Smith:** data collection, revising/editing

618 **Corina Logan:** Hypothesis development, data collection, data interpretation, revising/  
619 editing, materials/funding.

620 **FUNDING**

621 This research was funded by the Department of Human Behavior, Ecology and Culture at the  
622 Max Planck Institute for Evolutionary Anthropology.

623 **CONFLICT OF INTEREST DISCLOSURE**

624 We, the authors, declare that we have no financial conflicts of interest with the content of this  
625 article. CJ Logan and D Lukas are Recommenders at PCI Ecology, and CJ Logan was on the  
626 Managing Board at PCI Ecology (2018-2022).

627 **ACKNOWLEDGEMENTS**

628 We thank: our PCI Ecology recommender, Esther Sebastián González, and reviewers, Caroline  
629 Nieberding, Tim Parker, and Pizza Ka Yee Chow, for their helpful feedback on the preregistration;  
630 Luisa Bergeron, Melissa Folsom, Zoe Johnson-Ulrich, Maggie MacPherson, and Carol  
631 Rowney for sample collection and processing; Bronwyn Butcher and the Cornell Lab of Or-  
632 nithology for running ddRADseq and teaching our students in these methods; Nancy Chen  
633 for connecting us with students who wanted to learn ddRADseq; Xuwen Geng and Xin Yi  
634 He for support running ddRADseq; Woodland-Davis Clean Water Agency, RegionalSan, and  
635 Conaway Ranch for hosting the research on their land; Kristine Johnson for technical advice  
636 on great-tailed grackles; Julia Cissewski and Sophie Kaube for tirelessly solving problems in-  
637 volving financial transactions and contracts; Richard McElreath for project support; and Ken  
638 Kosik for being a UCSB sponsor of the Cooperation Agreement with the Max Planck Institute  
639 for Evolutionary Anthropology.

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