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Quantifying Male-biased Dispersal among Social Groups in the Collared Peccary (Pecari tajacu) Using Analyses Based on mtDNA Variation.

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- 3
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25 Abstract

26 Recent advances in the statistical analysis of microsatellite data permit calculation of sex-specific 27 dispersal rates through sex-and age-specific comparisons of genetic variation. This approach, 28 developed for analysis of data derived from co-dominant autosomal markers, should be 29 applicable to a sex-specific marker such as mitochondrial DNA. To test this premise, we 30 amplified a 449bp control region DNA sequence from the mitochondrial genome of the collared 31 peccary (*Pecari tajacu*), and estimated intra-class correlations among herds sampled from 3 32 Texas populations. Analyses on data partitioned by breeding group showed a clear signal of 33 male-biased dispersal; sex-specific fixation indices associated with genetic variation among 34 social groups within populations yielded values for females ($F_{GP} = 0.91$) which were 35 significantly larger than values for males ($F_{GP} = 0.24$; p = 0.0015). The same general pattern 36 emerged when the analyses were conducted on age classes (albeit nonsignificantly), as well as 37 categories of individuals which were predicted *a posteriori* to be dispersers (adult males) and 38 philopatric (adult females and all immatures). By extending a previously published methodology 39 based on bi-parentally-inherited markers to matrilineally-inherited haploid data, we calculated 40 sex-specific rates of contemporary dispersal among social groups within populations ($m_{\beta} = 0.37$). 41 These results support the idea that mitochondrial DNA haplotype frequency data can be used to 42 estimate sex-specific instantaneous dispersal rates in a social species.

43 Introduction

44 Sex bias in natal dispersal is common; in most mammalian species, males are dispersers 45 while females are philopatric, and the opposite trend is exhibited in birds (Greenwood 1980). 46 Exploring why the sexes differ in their dispersal patterns can shed light on the evolutionary causes of dispersal in general (Goudet et al. 2002), and accurate characterization of dispersal 47 48 behavior is integral to our understanding of the social structure, mating system, and population 49 genetic structure of a species. Yet detection of sex-biased dispersal can be tricky because a 50 dispersal event may occur once in an animal's lifetime, and such events can be difficult to 51 observe directly.

52

53 Measuring dispersal

54 In the last few decades molecular genetics has provided a means of investigating sex-biased 55 dispersal within and among populations (reviewed in Lawson-Handley and Perrin 2007). 56 Several powerful approaches have been developed to detect individual dispersers through 57 assignment tests or to characterize general patterns of dispersal through summary statistics of 58 population genetic structure (F-statistics, relatedness). Most of these approaches utilize 59 autosomal microsatellites as molecular markers, either alone (Goudet et al. 2002; Mossman and 60 Waser 1999; Petit et al. 2001; Waser et al. 2001) or in tandem with a uni-parentally inherited 61 marker such as mitochondrial DNA (mtDNA) or a Y chromosome locus (Escorza-Trevino and 62 Dizon 2000; Girman et al. 1997). The expectation inherent to all these approaches is that greater 63 genetic structure will be evident in the philopatric sex compared to the dispersing sex, thus 64 comparisons of sex-specific F_{ST} estimates should reveal the direction (and suggest the relative 65 strength) of sex-bias in dispersal (Goudet et al. 2002).

66 Because mitochondrial DNA is matrilineally inherited, it is commonly used to infer female-67 biased dispersal rates (Prugnolle and de Meeus 2002). When mtDNA haplotype distribution 68 patterns are examined in isolation, inferences can be made about female dispersal behavior 69 without respect to males, but this approach is qualitative and not widely applied (Hoelzer *et al.* 70 1994). However, it is possible to use mtDNA alone to infer the relative dispersal of both sexes 71 by extending methods developed for autosomal, bi-parentally inherited markers. For instance, 72 the comparisons of sex-specific population differentiation from haplotype frequency data can 73 indicate which sex disperses more (Escorza-Trevino and Dizon 2000; Yang et al. 2003).

74

75 Using sex-specific fixation indices to estimate instantaneous dispersal rates

76 Vitalis (2002) developed a method to quantitatively measure sex bias in instantaneous 77 dispersal rates using data from bi-parentally inherited markers such as microsatellites. This 78 approach allows the inference of sex-specific dispersal rates by comparing sex-specific estimates 79 of genetic differentiation (F_{ST}) measured before and after dispersal. This intuitive method can be 80 further extended to incorporate the hierarchical structure within social species (Fontanillas *et al.*) 81 2004), as it has been recognized that social organization can strongly influence correlations of 82 gene frequencies (Chesser 1991; Chesser and Baker 1996; Slatkin and Voelm 1991; Sugg and 83 Chesser 1994; Vigouroux and Couvet 2000). Herein we develop and use an extension of the 84 Vitalis' (2002) method to estimate instantaneous dispersal rates through analyses of mtDNA 85 haplotype distribution patterns in a social mammal, the collared peccary (*Pecari tajacu*, family 86 Tayassuidae).

We sampled extensively within three populations separated by long distances, with the goalof quantifying local dispersal among breeding groups within populations. We then compared

89 sex- and age-specific estimates of population differentiation based solely on mtDNA haplotype 90 frequencies, using probability based estimates of intra-class correlations of gene frequencies 91 among social groups within populations. We used a resampling approach to test for the 92 significance of the observed age and sex-bias in dispersal. Last, the fixation indices generated by 93 these analyses were used to calculate single-generation sex-specific dispersal rates. Heretofore 94 mtDNA has been used primarily to infer female dispersal patterns, but we demonstrate that this 95 matrilineally-inherited genetic marker can be used to quantify male dispersal rates in the absence 96 of nuclear population genetic data.

97

98 Materials and Methods

99 Study species

100 The collared peccary is a socially complex, pig-like ungulate that forms stable, mixed sex 101 herds of 3 to 30 individuals (Sowls, 1978). These groups associate throughout the year and 102 vigorously defend territories against other social groups (Bissonette 1982; Hellgren et al. 1984; 103 Ellisor and Harwell 1969). Herds are socially cohesive and attempts to immigrate may be met 104 with aggression, although direct observational data on dispersal behavior are still scarce. Male 105 exchange between groups and solitary wandering of both sexes has been observed but natal 106 dispersal has not been adequately described (Day 1985; Ellisor and Harwell 1969; Gabor and 107 Hellgren 2000). Heretofore little population genetic data existed for *P. tajacu* (but see Gongora 108 et al. 2006). Theimer and Keim (1994) utilized mtDNA variation to measure sequence 109 divergence and geographic partitioning in Arizona populations, but their samples were not 110 associated with social groups. There was sufficient heterogeneity in mtDNA haplotype 111 distribution to indicate limited female dispersal across regions (rather than among neighboring

112 herds as is considered here), although it was not clear if the patterns observed were also a

113 signature of founding events (Theimer and Keim 1994).

114

115 Sampling

116 Data were collected from three wild populations of *P. tajacu* in Texas. In the mid-1990s, 102 whole blood samples were collected from the Chaparral Wildlife Management Area 117 118 (CWMA) in south Texas (Gabor and Hellgren 2000). These samples were taken from live-119 trapped animals from 13 social groups, but not all group members were sampled. In 2005, we 120 collected 31 ear snip tissue samples from live-trapped animals from 4 groups in the Welder 121 Wildlife Refuge (WWR) in south Texas. In 2006-2007 we similarly sampled 134 animals from 122 13 groups in Big Bend Ranch State Park (BB) in west Texas, along the Texas-Mexico border. 123 The WWR and BB populations were sampled extensively; every social group at these locations 124 was identified through direct and remote camera observation and trapped in large corrals over 125 several sessions. Groups ranged in size from 2 to 18 animals and mean group size was 8.9. 126 Individuals were uniquely marked with numbered ear tags and the strongest possible effort was 127 made to trap and sample every unmarked individual. All samples include associated data on age 128 class (adult, subadult, juvenile, infant), sex, territory location and social group affiliation. Age 129 class was assigned according to behavior and morphological traits such as pelage, body size and 130 testicular development. Individuals exhibiting immature characteristics such as ginger or spotted 131 pelage, undescended or partially descended testicles, adult-oriented following behavior, or 132 estimated body size of less than 9 kg were classed as infants or juveniles, while individuals 133 which weighed 10-13 kg were classed as "subadults" that were on the cusp of sexual maturity. 134 Whole blood samples were frozen at -20°C, and tissue samples were stored in lysis buffer at

135 room temperature until DNA was extracted for long-term storage at 4°C.

136

137 *Genetic analysis*

138 Blood clot samples (~ 0.5 g) were digested by rotating for 12 hours at 55°C in 750 µL of 139 lysis buffer (100 mM Tris-Cl pH 8, 10 mM EDTA, 1% SDS, ddH₂O), 40 µL of proteinase K (10 140 mg/mL) and 2 μ L of streptokinase (10 U/ μ L). Tissue samples (~ 5 X 5 mm) were digested by 141 rotating for 24 hours at 55°C in 750 µL of lysis buffer and 20 µL of proteinase K (10 mg/mL). 142 Genomic DNA was extracted from blood using a standard phenol-chloroform method, and from 143 tissue samples using either a phenol-chloroform-isopropanol method or ammonium acetate method (Sambrook and Russell 2001). All DNA precipitations were washed twice in 70% 144 145 ethanol, and DNA pellets were resuspended in 250 µL of TLE (10 mM Tris-Cl, 0.1 mM EDTA). A 449 bp region between sites 15,390 and 15,900 of the collared peccary mtDNA D-loop was 146 147 amplified from genomic DNA using porcine primers (Alves et al. 2003). This sequence lies in 148 the hypervariable 5' end of the mitochondrial control region and does not code for any known 149 protein product. PCR volumes were 25 μ L and contained final concentrations of the following 150 reagents: 1.5 mM MgCl₂; 0.5 µM each primer; 0.21 mM dNTPs; 1.25 U *Taq* polymerase (NEB). 151 PCRs were performed in an Eppendorf MasterCycler using the following temperature profile: denaturation for 3 min. at 94° C, followed by 30 cycles of 94° C for 4 s, 55° C for 4 s, and 72° C 152 for 12 s; finishing with a 15 min. extension step at 72° C. PCR products were cleaned using a 153 154 low sodium protocol; 28 µL of a mixture containing 500 ml of absolute ethanol and 20 µL of 3M 155 NaOAc (pH 5.2) was added to each sample, shaken for 15 min, and centrifuged at 2051 g for 35 156 min. This step was followed by 70% ethanol precipitation under centrifugation (twice) and 157 resuspension in 20 μ L ddH₂0.

158	PCR products were then directly sequenced in both directions using Big Dye 3.1
159	chemistry. Sequencing products were purified using the low sodium protocol described above,
160	and then electrophoresed using an AB Prism 3730XL sequencer (Applied Biosystems).
161	Sequence data were aligned and edited with Sequencher 4.5 (Gene Codes). Nuclear copies of
162	mtDNA genes (numts) can greatly confound evolutionary analyses, and we avoided numts using
163	methods described in Triant and DeWoody (2007). For example, a few individuals (<5%),
164	harbored apparently heterozygous sites so we reamplified their DNA and completely
165	resequenced the amplicons in both directions. In every case, this procedure completely resolved
166	the mismatch and suggested the initial discrepancy was probably a result of Taq error.
167	We converted sequences into NEXUS format and imported them into PAUP* 4.0
168	(Swofford 2003) for haplotype assignment. Haplotypes were determined through reconstruction
169	of unrooted phylogenetic trees using a neighbor-joining algorithm. Direct sequencing of a sub-
170	set of the CWMA population revealed that some of the mtDNA haplotypes could be
171	discriminated by restriction digest with the MboI enzyme, but all individuals from WWR and BB
172	were typed by direct sequencing.
173	
174	Statistical analyses
175	Among-populations differentiation
176	MtDNA haplotype frequencies were calculated by hand for all three populations. Genetic
177	differentiation among populations was inferred from F_{ST} estimates (Weir and Cockerham 1984)

178 and exact tests of population differentiation (Raymond and Rousset 1995) using the software

179 package Arlequin Version 3.1 (Excoffier *et al.* 2005). For the latter, *p*-values were estimated

180 from a Markov chain set to 110,000 steps including 10,000 dememorization steps. All analyses

181 were based on pure haplotype frequency data rather than nucleotide differences.

182

183 Within-populations differentiation

184 Because *P. tajacu* populations are subdivided into breeding groups, we incorporated 185 breeding group as a hierarchical level. We calculated identity probabilities by simple counting of 186 identical pairs of genes at different hierarchical levels (Q_1 for pairs of genes within groups, Q_2 187 for pairs of genes sampled among groups within populations, and Q_3 for pairs of genes sampled 188 in different populations). We then estimated the intra-class correlations by taking appropriate 189 ratios of identity probabilities, weighted according to the number of pairs in each sample (see 190 Rousset 2007), following the definitions of F-statistics as functions of identity probabilities 191 between pairs of genes (see Appendix). Since the distances among populations are large in this 192 study (range of 225 km to 945 km between the three sampling sites), we considered the three 193 populations as independent replicates in the analysis, and we restricted our analyses to estimate 194 within-population dispersal. We focused on the level of genetic differentiation among social 195 groups within populations as measured by the parameter F_{GP} . The notation is adapted from 196 Wright (1965). This approach is different from that of Fontanillas et al. (2004) who considered 197 dispersal both among populations and among breeding groups. Although the samples from each 198 site were collected in different years, F_{GP} estimates do not depend upon identity between pairs of 199 genes from different populations and temporally discontinuous sampling is therefore unlikely to 200 undermine the approach. We employed a bootstrapping procedure to calculate confidence limits 201 around estimates of F_{GP} for each class of individuals. Using the statistical software package R 202 (R Development Core Team 2008), we generated 25,000 bootstrap samples, with each sample 203 being produced by random resampling (with replacement) of the 255 nucleotide sites from the

mtDNA haplotypes (254 sites + 1 indel). This allowed us to calculate F_{GP} estimates for each sample and generate a distribution; confidence intervals endpoints were then calculated as the 2.5% and the 97.5% percentiles of this distribution. This procedure is strictly equivalent to that implemented in the software package Arlequin Version 3.1 (Excoffier *et al.* 2005) to generate 95% confidence limits by bootstrapping genetic differentiation values in a locus-by-locus AMOVA (see, e.g., Langergraber et al. 2007).

210

211 Class-specific analyzes

212 Dispersal is a trait that can be partitioned into pre- and post-dispersal conditions, 213 therefore our first analysis partitioned the data by age. We performed independent analyses on 214 data partitioned into two age sets, respectively for adults and immatures (the latter including both 215 juveniles and infants). Subadults were classed as immatures and then as adults in sequential 216 analyses. Each age-specific data set was composed of individuals assigned to their respective 217 populations and social groups, and intra-class correlations (F_{GP}) were calculated among social 218 groups within populations from identity probabilities of pairs of genes (see above). Only those 219 social groups containing a representative individual from each treatment were included in the 220 analysis (e.g. in the independent analyses on adult and immature data sets, a social group must 221 have contained at least 1 adult and 1 immature to be included). We then duplicated the analysis 222 with the data partitioned by sex rather than age. From these results, we were able to distinguish a 223 putative class of dispersing individuals, from a putative class of non-dispersers. We therefore 224 performed a posteriori, independent analyses on data sets of putative dispersers and non-225 dispersers.

226

227 We used a resampling scheme after Goudet *et al.* (2002) to test whether the estimated fixation indices among social groups within replicate populations (F_{GP}) for specific classes (age, 228 229 sex, or putative dispersal class) departed significantly from the null hypothesis that dispersal is 230 independent from the class of individuals. Resampling tests were all performed with the 231 statistical software package R (R Development Core Team 2008). For each class, we generated 232 25,000 randomized datasets, by re-assigning the age (or sex, or dispersal class) of each haplotype 233 randomly within each breeding group. By doing so, we kept the number of individuals from 234 each class constant within each breeding group. We calculated the probabilities of identities 235 between pairs of genes for each resampled dataset, and obtained the distribution of class-specific $F_{\rm GP}$ estimates under the null hypothesis that dispersal behavior or capability is independent from 236 237 age, sex, or dispersal class. We then calculated p-values as the proportion of times where F_{GP} from the randomized datasets was larger than or equal to the observed F_{GP} on the original 238 239 dataset.

240

241 Estimating dispersal

To calculate a sex-specific dispersal rate within a single generation, we adapted Vitalis' (2002) approach and extended it to mtDNA data. In Vitalis (2002), the ratio of the sex-specific differentiation evaluated after juvenile dispersal (\hat{F}_{GP}^{XX}) divided by the differentiation evaluated before dispersal (\hat{F}_{GP}^{*}) gives the sex-specific dispersal rate. Appendix 1 shows that this relationship also applies to uni-parentally inherited markers, and:

248
$$\hat{m}_{X} \approx 1 - \sqrt{\frac{\hat{F}_{GP}^{XX}}{\hat{F}_{GP}^{*}}} \quad \text{for all } X \in \{\mathcal{O}, \mathcal{Q}\}$$
(1)

249

250	gives the sex-specific dispersal rate. Here we use this simple model to compare fixation indices
251	before and after dispersal at the within-population level, focusing on dispersal of individuals
252	among breeding groups. This equation assumes that the number of breeding groups, n , is large
253	(infinite); by considering an infinitely large n , we slightly overestimate dispersal rate m_x (e.g.,
254	10% relative bias with $n = 10$). We estimated instantaneous sex-specific dispersal rates for <i>P</i> .
255	tajacu by applying equation 1, using fixation indices estimates for adult males, adult females and
256	all immatures of both sexes (Table 2). Confidence intervals for dispersal rates were obtained by
257	means of a bootstrap procedure, similar to that used for F_{GP} (see above), modified as follows.
258	For each bootstrap sample, F_{GP} estimates were calculated for adult males (resp. adult females)
259	and all immatures, and male- (resp. female-) specific migration rates were calculated using
260	equation (1). Confidence intervals for sex-specific dispersal rates were then derived from the
261	0.025 and 0.975 percentiles of the bootstrap distribution.

262

263 **Results**

264 *mtDNA haplotype distribution patterns*

A total of 18 nucleotide sites were variable (17 substitutions and a single indel) over 449 bp. We recovered 6 mtDNA haplotypes from 267 individual collared peccaries among the 3 sites sampled (Table 1). Haplotype A was observed in all sampling sites, but haplotype B was unique to the CWMA, and haplotype C was found in both the WWR and the CWMA. The BB population was almost fixed for haplotype E (96%). Haplotypes F and G were only found in the CWMA, and were represented by single individuals (both males).

272 We overlaid mtDNA haplotype distribution onto the social group territory distribution for 273 all populations. At the local level, haplotype distribution did not exhibit geographic structuring 274 in the CWMA or the WWR; all haplotypes present at each sampling site were found distributed 275 throughout that site. In the BB population, haplotype A was found only in the eastern portion of 276 the sampling site. At the regional level across Texas, we observed significant population differentiation. Pairwise F_{ST} estimates ranged from 0.31 to 0.86 between populations and 277 278 pairwise exact tests of population differentiation were highly significant (p = 0.001), indicating 279 that these populations are significantly divergent from one another. 280

281 Patterns of genetic variation revealed by F-statistics as functions of identity probabilities

282 Because dispersal status is often dependent upon age, we tested for an age bias in 283 dispersal. To that end, we pooled infants and juveniles (categorized hereafter as "immatures") in 284 one class, and adults in another class. It was not clear if individuals categorized as subadults 285 were sufficiently developed to be considered as adults, therefore we performed a preliminary 286 analysis on adult-only and immature-only data sets partitioned into social groups, which revealed 287 a decrease in F_{GP} when subadults were included in the adult class (not shown). This result 288 indicates that individual genetic variation in the subadult class is apportioned among rather than 289 within social groups, and therefore subadults were classed as adults in all subsequent analyses. 290 We estimated fixation indices among social groups for each sex, with individuals partitioned into 291 known breeding groups (Table 2). It is clear that F_{GP} for adults (0.30 [0.03, 0.38]) is much 292 smaller than that for immatures (0.60 [0.60, 1.00]), as would be expected if the adult class 293 included dispersed individuals. To test for significance of these quantitative differences, we used 294 a randomization approach, and generated randomized data sets by assigning an age randomly to

295 each mtDNA haplotype. Under the null hypothesis that dispersal is not age-biased, we expect 296 the observed F_{GP} of adults and immatures not to depart significantly from the null distribution. 297 For adults, there was a large proportion of randomized data sets with a differentiation among 298 groups within populations (F_{GP}) larger than the observed, although this proportion did not 299 achieve significance (p = 0.79; Fig. 1A). In contrast, for immatures of both sexes, there was only 300 a small proportion of randomized data sets giving a F_{GP} larger than the observed, although the 301 test was not significant (p = 0.20; Fig. 1B). In general terms, these results clearly indicate a 302 greater amount of dispersal among social groups for adults when compared to immatures. To test for a signal of sex-biased dispersal, intra-class correlations (F_{GP}) were estimated 303 304 for each sex with individuals partitioned into known breeding groups (Table 2). It can be seen 305 from these results that F_{GP} among social groups is much smaller for males (0.23 [0.09, 0.36]) 306 than it is for females (0.90 [0.87, 1.00]), which indicates that even when pre-dispersal age 307 individuals are included in the male class the sex difference is still apparent. To test the 308 significance of the sex difference, we used a randomization approach identical to the one 309 described for age bias, and generated randomized data sets by assigning a sex randomly to each 310 mtDNA haplotype. For males, there was a very large proportion of randomized data sets with a 311 larger F_{GP} than the observed (p = 0.98; Fig. 1C). In contrast, for females, there was only a very 312 small proportion of randomized data sets giving a F_{GP} larger than the observed, and the test was 313 therefore highly significant (p < 0.001; Fig. 1D). These results suggest that dispersal is strongly 314 male-biased in *P. tajacu*. 315 The inferred dispersal pattern of *P. tajacu* being of adult male dispersal, we conducted a

The inferred dispersal pattern of *P. tajacu* being of adult male dispersal, we conducted a further analysis, *a posteriori*, on data partitioned by putative dispersal condition: the data were partitioned into "dispersers" (adult males) and "philopatrics" (immature males and all females)

and separate analyses performed on individuals assigned to breeding groups. As expected, F_{GP} for the philopatric class was much larger (0.76 [0.73, 0.99]) than was seen for adult males (0.24 [0.07, 0.36]). For adult males, there was a very large proportion of randomized data sets with a F_{GP} larger than the observed (p = 0.99; Fig. 1E). In contrast, for putative non-dispersers, the test was highly significant, with very few datasets giving a F_{GP} larger than the observed (p < 0.001; Fig. 1F).

324

325 Dispersal rate estimates

326 The instantaneous sex-specific dispersal rate among social groups within populations was 327 estimated using equation (1). We used the F_{GP} estimates among social groups within populations 328 (Table 2) for adult males (dispersers) ($F_{GP} = 0.24$), for adult females ($F_{GP} = 0.91$), and for predispersal individuals of both sexes (also categorized as "immature"; $F_{GP} = 0.60$). This yielded a 329 330 male-specific dispersal rate estimate (m_{σ}) of 0.37 [0.32, 0.65]. Equation (1) only makes sense if 331 there is a significant difference between F_{GP} measured after dispersal and before dispersal. Since 332 the confidence limits of F_{GP} for adult females ([0.88; 1.00]) and immatures ([0.60; 1.00]) largely overlap, we were unable to calculate a female-specific dispersal rate from equation (1). 333

334

335 Discussion

We have demonstrated that maternally-inherited genes can be used to describe the contemporary dispersal patterns of males (and the overall dispersal patterns of females) within an analytical framework based on intra-class genetic correlations. This was accomplished through comparisons of age- and sex-specific intra-class correlations partitioned hierarchically within populations. A second aim was to show that instantaneous sex-specific dispersal rates can be

341 calculated from sex-specific estimates of differentiation using single-locus haplotypic data.

342

343 Dispersal in Pecari tajacu

344 In our study we quantitatively demonstrated that dispersal in collared peccaries is 345 strongly biased toward males, and that approximately one-third of males dispersed from their 346 natal groups in this single generation. This is a minimum estimate, as some individuals die 347 before or during dispersal, and the lack of mtDNA variation undoubtedly prevented our detection 348 of dispersal between some groups. Moreover, the pronounced local genetic structure indicates 349 that males preferentially disperse over short distances, perhaps into neighboring herds; this is 350 congruent with trapping data (Gabor and Hellgren 2000). The results from the age-based 351 analysis indicate that dispersal in this species is usually accomplished by subadults (18-24 352 months). At this age, they have not reached their full body mass and may be forced out by larger, 353 resident males.

354

355 Measuring dispersal biases

356 Our approach allowed us to organize data into age classes, sex classes, social groups, and 357 populations and then test hypotheses about the dispersal rate of each class. For example, by 358 performing separate analyses on sex-specific datasets, we were able to both detect a sex-bias in 359 dispersal and also determine which sex contributed to the pattern. Because the method relies on contrasts of sex-specific estimates of population differentiation, rather than absolutes, the power 360 361 to detect differences among hierarchies is limited only by the intensity of the bias (Vitalis 2002). 362 In this study there was sufficient contrast between pre- and post-dispersal age classes in males to 363 provide a direct estimate of the instantaneous dispersal rate.

364 The method presented here should be applicable to any species in which there is a bias in dispersal, whether that bias is conditional on sex, age, or some other phenotype, so long as trait 365 variation can be readily distinguished and assigned to different hierarchical levels. This 366 367 approach does not impose spatial distance (or a distance proxy) onto the analysis, as is seen in 368 other approaches such as spatial autocorrelation (Smouse and Peakall 1999). Such approaches 369 force investigators to make assumptions about how distance interacts with social organization 370 when it may be inappropriate or irrelevant (e.g., when sampling a highly mobile species, or at a 371 scale where an individual is equally likely to disperse to any location under consideration). Our 372 approach removes metric distance and location from the equation, and shifts the focus onto how the genetic variation is distributed across space irrespective of distance, which is especially 373 374 useful for addressing questions of how sociality influences dispersal.

375

376 Measuring sex-biased dispersal with uni-parentally inherited markers

377 The approach discussed herein relies upon contrasts: we compared the genetic structure 378 of the pre-dispersal class to the sex-specific genetic structure of the post-dispersal class to 379 estimate instantaneous dispersal within a single generation (Lawson-Handley and Perrin 2007). 380 When autosomal markers are used, the expectation is that genetic structure will be more apparent 381 in the pre-dispersal class compared to the post-dispersal class as a whole, and even more 382 apparent in the non-dispersing sex (whichever sex it may be). When a uni-parentally inherited marker is used the expectation is similar, but not identical, to what is seen for bi-parentally 383 384 inherited markers. For instance, under a system of male-biased dispersal mtDNA haplotypes are 385 carried within males into breeding groups, but males do not contribute mtDNA to the subsequent 386 generation and thus the contrast between pre-dispersal individuals and adult males is substantial.

387 However, under a system of female-biased dispersal haplotypes would be re-distributed within 388 and among populations each generation. Thus a contrast between genetic differentiation for pre-389 and post-dispersal individuals would be difficult to detect. As a result, this approach is most 390 useful for deriving instantaneous sex-specific dispersal rates with mtDNA data under a system of 391 male-biased dispersal, or a double uniparental system of mitochondrial DNA inheritance (e.g. 392 *Mytilus* mussels). Here we use mtDNA haplotypes as a tag, but any physical or genetic tag that 393 could be identified in males and females before and after dispersal may play the same role as 394 mtDNA markers in this context. 395 We have demonstrated that mtDNA can be used in isolation to estimate sex-specific dispersal 396 in the current generation,. The main caveat is that mtDNA is, in effect, a single genetic marker 397 that might be biased by selection (Bazin et al. 2005). Yet, because we based our analyzes upon

398 differences of variation in male and female within a single generation, it is difficult to imagine a

399 pattern of selection that would undermine the approach.

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				Sex		Age class	8	
Population	Haplotype	n	Freq.	F	М	Ι	J	А
WWR	С	24	0.77	10	14	1	6	17
	А	7	0.23	3	4	0	3	4
CWMA	В	43	0.43	24	19	1	10	32
	А	38	0.38	21	17	1	9	28
	С	19	0.19	9	10	3	2	14
	G	1	0.01	0	1	0	0	1
	F	1	0.01	0	1	0	0	1
BB	Е	129	0.96	61	68	13	35	81
	А	5	0.04	3	2	0	0	5

Table 1 Distribution of mtDNA haplotypes in three wild populations of *P. tajacu* in Texas, across sex and age classes.

n: sample size; F: females, M: males; I: infants; J: juveniles; A: adults (including subadults)

501 **Table 2.** Intra-class correlations for pairs of genes among social groups within replicate populations estimated by means of identity

- 502 probabilities. F_{GP} estimates used to calculate sex-specific migration rates from equation (1) are given in bold. Confidence intervals are
- 503 noted in brackets, and represent the 2.5% and the 97.5% quantiles produced by bootstrapping nucleotide sites over 25,000 samples. P-
- 504 values give the results of significance tests for differentiation among groups. They were calculated as the proportion of times where
- 505 F_{GP} from randomized datasets was larger than or equal to the observed F_{GP} on the original dataset. Randomized datasets were

506	obtained by by	permuting haplotypes	at random among groups	within populations	(25,000 permutations).
				1 1	

Category	$F_{\rm GP}$ estimate	<i>p</i> -value
All data	0.50 [0.46, 0.59]	<i>p</i> < 0.001
Adults	0.30 [0.03, 0.38]	<i>p</i> < 0.001
Immatures	0.60 [0.60, 1.00]	<i>p</i> < 0.001
Females	0.90 [0.87, 1.00]	p < 0.001
Males	0.23 [0.09, 0.36]	p < 0.001
Adult females	0.91 [0.88, 1.00]	p < 0.001
Adult males (dispersers)	0.24 [0.07, 0.36]	p = 0.013
Philopatrics	0.76 [0.73, 0.99]	<i>p</i> < 0.001

507 Figure Legends

508

509

510	Figure 1. Re-sampled data null distributions for each class of individuals. Observed F_{GP} for
511	each analysis represented by hatched vertical line. Significance tested over 25,000 permutations.
512	Histogram class heights are represented as black dots, and the smoothed density was obtained
513	using the Average Shifted Histogram (ASH) algorithm (Scott 1992) with smoothing parameter m
514	= 20.

515 Appendix

516 Sex-specific differentiation before and after (instantaneous) dispersal

517 F-statistics are defined as intra-class correlations for the probability of identity in state 518 (IIS correlations) (Cockerham and Weir 1987; Rousset 1996). Yet, the infinite allele model 519 (IAM) provides the value of the probability of identity by descent (IBD probabilities) and in the 520 low mutation rate limit, for two given classes, IIS and IBD correlations converge to the same 521 value (Slatkin 1991; Rousset 1996). Thus, the properties of *F*-statistics can be deduced from the 522 properties of intra-class correlations for IBD probabilities (Rousset 1996). 523 In the following, we consider mitochondrial DNA (mtDNA) markers, i.e. haploid markers, transmitted by females only. Let Q_1^{XY} , Q_2^{XY} and Q_3^{XY} be the IBD probability of two 524 525 mtDNA gene copies sampled from two individuals of sex X and Y after dispersal among individuals within the same breeding group, among breeding groups within the same 526 527 populations, and among distinct populations, respectively. These individuals may be two males, two females, or one male and one female. Let Q_1^* , Q_2^* and Q_3^* be the corresponding IBD 528 529 probabilities for gene copies sampled before dispersal. For pairs of genes sampled before dispersal there is no need to consider distinct coefficients for different pairs of individuals of the 530 531 same or opposite sex (Vitalis 2002).

Let us consider an infinite island model of population structure (Wright 1951) where each population is isolated and further subdivided into *n* breeding groups, and where dispersal among breeding groups is achieved by juveniles, before reproduction. Here, because we consider that populations are independent, we restrict our analyses to estimate within-population dispersal only. We therefore focus on the IBD probabilities among individuals within social groups (Q_1^{XY} and Q_1^*), and among social groups within populations (Q_2^{XY} and Q_2^*), and further consider the IBD

probabilities between pairs of genes from distinct populations (Q_3^{XY} and Q_3^*) to be nil in the 538 539 model. A migrant individual in a breeding group is equally likely to come from any of the (n-1) other breeding groups. Generations do not overlap. Let m_x denote the probability that 540 541 an individual of sex X has immigrated. Each generation, after migration, the frequency of pairs 542 of individuals taken at random in one breeding group that come from a single group before migration is $a_{XY} = (1 - m_x)(1 - m_y) + m_x \cdot m_y/(n-1)$, for pairs of individuals of sex X and Y with 543 $X \in \{\sigma, Q\}$ and $Y \in \{\sigma, Q\}$. Conversely, the frequency of pairs of individuals taken at random 544 from two breeding group after migration that originate from the same group before migration is 545 $b_{xy} = (1 - a_{xy})/(n - 1)$ (see, e.g., Rousset 1996). We assume that the mutation rate is low (i.e., μ 546 547 $\ll m_X$), so that virtually no mutation arises over a single generation.

The genes sampled among individuals after dispersal in a breeding group at any generation come from the same breeding group before dispersal with probability a_{XY} and from different breeding group with probability $(1 - a_{XY})$. These mtDNA gene copies are then IBD with probabilities Q_1^* and Q_2^* , respectively. Thus,

552
$$Q_1^{XY} = a_{XY}Q_1^* + (1 - a_{XY})Q_2^*$$
 (A.1)

553 Similarly, for genes sampled before dispersal,

554
$$Q_2^{XY} = b_{XY}Q_1^* + (1 - b_{XY})Q_2^*$$
 (A.2)

555 We can rewrite equations 1 and 2 as:

556
$$Q_1^{XY} = a_{XY} \left(Q_1^* - Q_2^* \right) + Q_2^*,$$
 (A.3)

557 and:

558
$$Q_2^{XY} = b_{XY} \left(Q_1^* - Q_2^* \right) + Q_2^*.$$
 (A.4)

559 Subtracting equations A.3 and A.4, we obtain:

560
$$\left(Q_1^{XY} - Q_2^{XY}\right) = d_{XY}\left(Q_1^* - Q_2^*\right),$$
 (A.5)

561 where $d_{xy} = (a_{xy} - b_{xy})$. Rearranging equation A.4 gives:

562
$$(1-Q_2^{XY}) = (1-Q_2^*) - b_{XY}(Q_1^* - Q_2^*).$$
 (A.6)

563 Taking the ratio of equation A.6 over A.5 gives:

564
$$\frac{1-Q_2^{XY}}{Q_1^{XY}-Q_2^{XY}} = \frac{1}{d_{XY}} \frac{1-Q_2^*}{Q_1^*-Q_2^*} - \frac{b_{XY}}{d_{XY}}.$$
 (A.7)

Since we consider independent populations, the relevant parameter to infer dispersal among breeding groups is F_{GP} . From the definition of sex-specific *F*-statistics as appropriate ratios of sex-specific IBD probabilities (Vitalis 2002), we get:

568
$$F_{GP}^* = \frac{Q_1^* - Q_2^*}{1 - Q_2^*}$$
(A.8)

569 before dispersal, and:

570
$$F_{GP}^{XY} = \frac{Q_1^{XY} - Q_2^{XY}}{1 - Q_2^{XY}}$$
(A.9)

571 after dispersal. We get from equation A.7:

572
$$\frac{1}{F_{GP}^{XY}} = \frac{1}{d_{XY}} \frac{1}{F_{GP}^*} - \frac{b_{XY}}{d_{XY}}, \qquad (A.10)$$

573 i.e., multiplying both sides by $(d_{XY} \cdot F_{GP}^{XY})$ and rearranging:

574
$$\frac{F_{GP}^{XY}}{F_{GP}^{*}} = d_{XY} + b_{XY}F_{GP}^{XY} \text{ for all } (X, Y) \in \{\mathcal{O}, \mathcal{Q}\}$$
(A.11)

which is equation 15 in Vitalis (2002) except that here it holds for mtDNA markers, and it does not assume equilibrium. This relation is valid for any generation, for samples taken before and after dispersal. The last term in the right-hand side of equation A.11 is negligible compared to d_{XY} . Thus, F_{GP}^{XX} differs from F_{GP}^* by a factor $d_{XX} = (1 - m_X [n/(n-1)])^2 \approx (1 - m_X)^2$, for large *n*.

579 Therefore, taking the ratio of sex-specific F_{GP}^{XX} evaluated after juvenile dispersal, over F_{GP}^{*} 580 evaluated before dispersal gives the sex-specific migration rate:

581
$$m_X \approx 1 - \sqrt{\frac{F_{GP}^{XX}}{F_{GP}^*}}$$
 for all $X \in \{\mathcal{O}, \mathcal{Q}\}$ (A.12)

Following Rousset (2007), we estimated the intra-class correlations by taking appropriate ratios of identity probabilities, weighted according to the number of pairs in each sample. From the definitions of *F*-statistics as functions of identity probabilities between pairs of genes (equations A.8 and A.9), we get:

586

587
$$\hat{F}_{GP}^* = \frac{\hat{Q}_1^* - \hat{Q}_2^*}{1 - \hat{Q}_2^*}$$
, and (A.13)

588
$$\hat{F}_{GP}^{XY} = \frac{\hat{Q}_1^{XY} - \hat{Q}_2^{XY}}{1 - \hat{Q}_2^{XY}}, \qquad (A.14)$$

where \hat{Q}_i^* and \hat{Q}_i^{XY} denote the estimates of identity probabilities between pairs of genes, calculated by simple counting of identical pairs of genes at the *i*th hierarchical level. An estimate of the sex-specific migration rate therefore reads:

592
$$\hat{m}_X \approx 1 - \sqrt{\frac{\hat{F}_{GP}^{XX}}{\hat{F}_{GP}^*}}$$
 for all $X \in \{ \sigma, Q \}$ (A.15)