

1    **The pace of mitochondrial molecular evolution varies with seasonal migration distance**

2    **Running head: Bird migration and molecular evolution**

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11

12   **Abstract**

13   Animals that engage in long-distance seasonal migration experience strong selective pressures on  
14   their metabolic performance and life history, with potential consequences for molecular  
15   evolution. Species with slow life histories typically show lower rates of synonymous substitution  
16   ( $d_s$ ) than “fast” species. Previous research suggests long-distance seasonal migrants have a  
17   slower life history strategy than short-distance migrants, raising the possibility that rates of  
18   molecular evolution may covary with migration distance. Additionally, long-distance migrants  
19   may face strong selection on metabolically important mitochondrial genes due to their long-  
20   distance flights. Using over 1000 mitochondrial genomes, we assessed the relationship between  
21   migration distance and mitochondrial molecular evolution in 39 boreal-breeding migratory bird  
22   species. We show that migration distance correlates negatively with  $d_s$ , suggesting that the slow  
23   life history associated with long-distance migration is reflected in rates of molecular evolution.

24 Mitochondrial genes in every study species exhibited evidence of purifying selection, but the  
25 strength of selection was greater in short-distance migrants, contrary to our predictions. This  
26 result may indicate effects of selection for cold tolerance on mitochondrial evolution among  
27 species overwintering at high latitudes. Our study demonstrates that the pervasive correlation  
28 between life history and molecular evolutionary rates exists in the context of differential  
29 adaptations to seasonality.

30

31 **Keywords:** Life History, Seasonal Migration, Molecular Evolution,  $d_S$ , Mitochondria

32

33

34 **Introduction**

35 Species' traits are the product of their genome and their environment, but in turn, traits  
36 and the environment also shape the molecular evolution of the genome. For example,  
37 metabolically demanding traits influence molecular evolution of mitochondrial genes (e.g. Shen  
38 et al. 2009; Chong and Mueller 2013; Strohm et al. 2015). More broadly, traits associated with  
39 the slow-fast continuum of life history (Stearns 1983) are correlated with rates of molecular  
40 evolution (Bromham 2020) such that life history evolution is thought to alter the pace of a  
41 lineage's molecular clock (Hwang and Green 2004; Moorjani et al. 2016). Environmental  
42 pressures associated with seasonality can influence life history (Varpe 2017) and metabolic  
43 demands (Weber 2009; Chen et al. 2018), suggesting that variation in adaptation to seasonality  
44 could have molecular evolutionary consequences. However, the linkages between molecular  
45 evolution and differential adaptations to seasonality are rarely explored.

46 In this study, we investigate how patterns of mitochondrial molecular evolution are  
47 related to variation in seasonal migration distance. Migratory animals survive harsh seasonal  
48 conditions on their breeding grounds by temporarily departing until conditions improve (Winger  
49 et al. 2019). Migration distance varies across species, ranging from short-distance movements  
50 within an ecoregion to hemisphere-crossing journeys. Long-distance seasonal migration requires  
51 high metabolic performance (Weber 2009), with potential implications for the dynamics of  
52 selection on the metabolically-important mitochondrial genes (Shen et al. 2009; Strohm et al.  
53 2015). Migration distance has also been recognized as an important axis of life history variation  
54 (the balance between annual survival and reproduction) in birds (Greenberg 1980; Møller 2007;  
55 Bruderer and Salewski 2009; Winger and Pegan 2021). Migration distance may therefore also  
56 influence molecular evolutionary rates through effects on life history (Bromham 2020) that are

57 not directly associated with metabolic demands, but this relationship has not been assessed. Here,  
58 we assess how migration distance correlates with mitochondrial molecular evolution within the  
59 community of migratory birds breeding in the highly seasonal North American boreal region,  
60 and we test hypotheses regarding the roles of life history and metabolic adaptation in mediating a  
61 relationship between molecular evolution and seasonal migration.

62

63 *Metabolic adaptation, life history, and mitochondrial molecular evolution*

64 Reliance on locomotion (migration) for adaptation to seasonality may influence selection  
65 on mitochondrial genes, which play an important role in metabolism. Mitochondria typically  
66 experience purifying selection (i.e. selection that reduces genetic variation) because most  
67 mutations in these genes are deleterious to fitness (Nei et al. 2010; Nabholz et al. 2013; Popadin  
68 et al. 2013). Prior studies have shown that purifying selection tends to be stronger in the  
69 mitochondria of mobile animal species compared to less mobile relatives. This pattern has been  
70 demonstrated in comparisons between flighted and flightless birds (Shen et al. 2009) and insects  
71 (Mitterboeck et al. 2017; Chang et al. 2020), between migratory and nonmigratory fishes  
72 (Strohm et al. 2015), and between amphibians (Chong and Mueller 2013) and mollusks (Sun et  
73 al. 2017) with different locomotory modes. Within flighted birds, species with slow flight and  
74 those that rely on soaring (versus flapping) have been shown to experience relaxed mitochondrial  
75 purifying selection compared to faster-flying species (Shen et al. 2009; De Panis et al. 2021).  
76 Additionally, Montoya et al. (2022) recently demonstrated that flight habit, as represented by  
77 wing morphology, is associated with nonsynonymous mitochondrial evolutionary rate variation  
78 in a large clade of South American birds (Furnariidae). These studies suggest that mitochondrial  
79 genotype plays an especially important role in fitness for organisms that rely on high-energy

80 locomotion, including migratory birds. Metabolic demand may be strongest in long-distance  
81 migrating species if these demands primarily arise from locomotion. However, species that breed  
82 at high latitudes and migrate only short distances for the nonbreeding season may require  
83 alternative metabolic adaptations for dealing with harsh seasonal conditions since their shorter  
84 migrations do not allow them to fully escape cold, resource-depleted winters (Winger et al.  
85 2019). The effect of variation in seasonal migration distance on the strength of mitochondrial  
86 purifying selection is unknown.

87 A second and distinct way in which seasonal migration may influence molecular  
88 evolution is through its relationship with life history and, consequently, molecular evolutionary  
89 rate. The slow-fast continuum of life history is commonly characterized by “life-history traits”  
90 that underly or correlate with differing rates of growth, survival, and reproduction (Read and  
91 Harvey 1989; White et al. 2022). Within major lineages of plants, bacteria, vertebrates, and  
92 invertebrates, species with “slow” life history (i.e., long generation time, low annual fecundity,  
93 large size; Stearns 1983) also exhibit slower molecular substitution rate than “fast” species (i.e.,  
94 those with shorter generation time, higher annual fecundity, and smaller size; Nabholz et al.  
95 2008a; Smith and Donoghue 2008; Thomas et al. 2010; Weller and Wu 2015). Within migratory  
96 birds breeding in the temperate zone, seasonal migration distance covaries with annual fecundity  
97 and survival such that long-distance migrants show “slower” life history (i.e., higher annual  
98 survival, lower annual fecundity) than short-distance migrants (Greenberg 1980; Bruderer and  
99 Salewski 2009; Winger and Pegan 2021). As such, variation in migration distance across species  
100 may affect molecular evolutionary rates because of its association with life history variation.  
101 Specifically, the synonymous substitution rate “ $d_S$ ” often correlates with the slow-fast life history  
102 continuum (Nikolaev et al. 2007, Bromham et al. 2015, Hua et al. 2015; Table 1). Prior studies

103 suggest that life history may influence  $d_s$  through effects on DNA replication rate or selection for  
104 mutation avoidance (reviewed in Bromham 2020), because  $d_s$  is thought to primarily reflect the  
105 underlying mutation rate when synonymous mutations are selectively neutral (Kimura 1983; Nei  
106 et al. 2010; Lanfear et al. 2014). Direct estimates of nuclear germline mutation rates support the  
107 hypothesis that species-level variation in mutation rate correlates with life-history traits  
108 (Bergeron et al. 2023).

109

110 *Predicting the relationship between seasonal migration distance and molecular evolution*

111 Long-distance migratory birds have been shown to exhibit a slower life history than  
112 sympatric breeding short-distance migrants (Winger and Pegan 2021, Fig. 1). Thus, long-  
113 distance migrants travel farther in each migratory trip than short-distance migrants and may also  
114 require more trips per lifetime to achieve the level of lifetime fitness of short-distance migrants  
115 (Møller 2007). Owing to the metabolic demands of migration and the importance of repeated  
116 migration success for fitness in long-distance migrants, the migratory phenotypes of these  
117 species are thought to be under strong variation-reducing natural selection (Conklin et al. 2017).  
118 As such, we hypothesize that long-distance migrants exhibit both lower  $d_s$  (which could reflect  
119 selection against mutation in the mitochondria; Hua et al. 2015) and stronger purifying selection  
120 in their mitochondrial genes than short-distance migrants.

121 To test these hypotheses, we examined the relationship between migration distance and  
122 rates of molecular evolution of the mitochondrial coding genes in a community of small-bodied  
123 migratory songbirds breeding in the boreal forests of North America. The 39 co-distributed  
124 species we studied are ideal for investigating the effects of migration distance on molecular  
125 evolution because they vary greatly in migration distance (e.g., Fig. 1, Table S1), yet they

126 otherwise share similar breeding habitat, population history, and body mass (Winger and Pegan  
127 2021). This system allows us to test hypotheses about migration distance while minimizing  
128 variation in other traits that could influence molecular evolution. We assessed effects of  
129 migration distance on  $d_s$  (synonymous substitution rate) and  $d_N/d_s$  (purifying selection) in a  
130 Bayesian phylogenetic framework (Lartillot and Poujol 2011) with full mitochondrial gene sets  
131 we sequenced for 39 species. Further, we used population genetic datasets from all mitochondrial  
132 genes that we generated for 30 of the species (for a total of 1008 samples used across all  
133 analyses) to assess effects of migration distance on purifying selection at the population level.  
134 Specifically, we assessed  $\pi_N/\pi_S$ , which is a population genetic summary statistic representing the  
135 amount of nonsynonymous versus synonymous polymorphism within a population.

136

#### 137 *Accounting for effects of $N_e$ on substitution rates*

138 Molecular evolution is fundamentally influenced by effective population size ( $N_e$ ), so it is  
139 often difficult to determine whether links between traits and molecular evolutionary rates are  
140 mediated by effects of traits on  $N_e$  versus other hypothesized mechanisms (e.g., Montoya et al.  
141 2022). Therefore, we take advantage of our population-level datasets to directly test for effects of  
142  $N_e$  on molecular evolutionary rates and purifying selection, providing valuable context for the  
143 interpretation of our results. Variation in  $N_e$  can cause variation in substitution rates because the  
144 efficiency of natural selection in purging deleterious mutations is determined by the balance  
145 between strength of selection and strength of drift, which is reflected in  $N_e$  (Ohta 1992).  
146 Specifically, studies on empirical populations have demonstrated that populations with small  $N_e$   
147 typically show weaker purifying selection (i.e., higher  $d_N/d_s$ , e.g., Popadin et al. 2007, Leroy et  
148 al. 2021; and higher  $\pi_N/\pi_S$ , e.g., Chen et al. 2017). Several recent studies found correlations

149 between traits associated with life history and genetic diversity, suggesting that species with  
150 “slow” life histories often have low  $N_e$  (Romiguier et al. 2014; Brüniche-Olsen et al. 2021; De  
151 Kort et al. 2021). There is also evidence that migratory behavior is predictive of population  
152 genetic diversity, a parameter associated with  $N_e$  (García-Berro et al. 2023). It is therefore  
153 important to assess whether molecular rate variation across species can alternatively be explained  
154 by confounding variation in  $N_e$ .

155 Finally, we use estimates of  $N_e$  to test the assumption of neutral evolution at synonymous  
156 sites, which is a fundamental assumption underlying the hypothesis that  $d_S$  reflects mutation rate  
157 (Kimura 1983; Nei et al. 2010; Lanfear et al. 2014). If synonymous substitutions evolve  
158 neutrally, we expect that  $d_S$  should not show a relationship with  $N_e$  because the processes that  
159 lead to a relationship between  $N_e$  and substitution rate involve natural selection.

160

## 161 **Methods**

### 162 *Study system*

163 We focused on 39 species of migratory birds breeding in the North American boreal  
164 forest, representing 11 families (Table S1). These are the same species for which a correlation  
165 between migration distance and the slow-fast life history continuum—*independent* of body  
166 size—has been demonstrated using data on annual fecundity and survivorship (Winger and  
167 Pegan 2021). We focus our analyses on co-distributed populations of the eastern boreal belt of  
168 North America (Omernik 1987, Fig. S1). Some species’ breeding ranges extend into other  
169 ecoregions (e.g., the mountain west or the temperate forests south of the boreal zone), but in  
170 these cases, we only analyze samples from the boreal portion of the range to assess sympatric  
171 populations. The species in the dataset exhibit broad variation in migration distance, with their

172 geographic range centroids shifting between 1048 km and 7600 km between the breeding and  
173 non-breeding periods (Fig. 2, Table S1; Winger and Pegan 2021). These centroid shifts represent  
174 migratory strategies ranging from short-distance movements within the temperate region to the  
175 movement of an entire population across ocean and land barriers from North America to South  
176 America. All species are less than 100 g in mass (range of mean mass across species is 6-87  
177 grams; Table S1) and are broadly similar in habitat use. They are all territorial species with  
178 socially monogamous breeding systems, which suggests that they probably do not vary  
179 substantially in population sex ratio (which can affect  $N_e$ ), although empirical sex ratio data is  
180 not available for these species. Small songbirds are typically capable of breeding in their second  
181 year, and this is true of all species in our study that have been assessed (Billerman et al. 2022).  
182 Additionally, our study species share relatively similar demographic histories, with population  
183 expansions estimated to have mostly occurred during the period of glacial retreat that preceded  
184 the Last Glacial Maximum (~57,000 years before present; Kimmitt et al. 2023).  
185

#### 186 *Life history covariates: Migration distance and mass*

187 Direct measurements of migration distance of individuals are lacking for most of the species in  
188 our system, so we used the distance between the centroid of a species' breeding range and the  
189 centroid of its nonbreeding range to represent the migration distance of the species. Although the  
190 distance between centroids does not represent individual variation in migration distance within a  
191 species, this metric captures broad differences in migratory strategies between species. Our  
192 method for calculating the distance between range centroids is described in detail in Winger and  
193 Pegan (2021). We included mass as a covariate in our analyses because body mass and rates of  
194 molecular evolution are often associated (Figuet et al. 2014; Nabholz et al. 2016), and the

195 relationships between survival and fecundity and migration distance demonstrated by Winger  
196 and Pegan (2021) were recovered after accounting for variation in mass. We obtained mass data  
197 from Dunning (2008) and Billerman et al. (2022).

198

199 *Sampling and DNA sequencing*

200 Our analysis of the relationship between migration distance and  $d_S$  requires one  
201 mitochondrial genome for each species in the study, while analyses of  $N_e$  and  $\pi_N/\pi_S$  require  
202 population-level sampling. For our analysis of  $d_S$ , we obtained whole mitochondrial genomes  
203 from one individual of each of the 39 species in our study by sequencing DNA from tissue  
204 samples associated with a museum specimen, as described below. These specimens were  
205 collected during the breeding season from near the longitudinal center of the boreal forest  
206 (Manitoba, Minnesota, or Michigan; Tables S1,S2). For two species (*Contopus cooperi* and  
207 *Euphagus carolinus*), we used specimen-vouchered tissue samples of individuals salvaged  
208 during migration in Michigan from collision mortalities.

209 For our population-level analyses, we generated a large dataset of 999 additional  
210 mitochondrial genomes for 30 of the 39 species, building on a dataset of 19 species from  
211 Kimmitt et al. (2023). Our larger dataset includes complete coding sequences for 8 to 49  
212 individuals per species (mean 33 individuals per species; Table S1). These individuals were  
213 sampled during the breeding season across a longitudinal transect of the boreal forest from  
214 Alberta to the northeastern United States (Fig. S1, Table S2). Except for 24 blood samples from  
215 New York state, all sequences we used came from frozen or ethanol-preserved tissue samples  
216 associated with museum voucher specimens provided by several museum institutions (Table S2;  
217 *Acknowledgments*).

218 We obtained high-depth mitochondrial genomes captured as a byproduct from low-  
219 coverage whole genome sequencing, as described in detail in Kimmitt et al. (2023). Briefly,  
220 sequencing libraries were prepared using a modified Illumina Nextera library preparation  
221 protocol (Schweizer et al. 2021) and sequenced on HiSeq or NovaSeq machines using services  
222 provided by Novogene and the University of Michigan Advanced Genomics Core. We used  
223 NOVOPlasty v4.3.1 (Dierckxsens et al. 2016) to assemble mitochondrial contigs, specifying a  
224 target genome size of 20-30 kb and using a k-mer of 21. We provided NOVOPlasty with a  
225 conspecific mitochondrial seed sequence (Table S1) for each species. We annotated the contigs  
226 built by NOVOPlasty using Geneious Prime 2020.2.2 (<https://www.geneious.com>) with copies  
227 of mitochondrial genes from GenBank (Table S1). Whenever applicable in the filtering and  
228 analysis steps described below, we used options specifying the vertebrate mitochondrial code.

229 Our initial dataset across all species contained mitochondrial sequences from 1229 total  
230 individuals. To ensure data quality, we used BLAST (<https://blast.ncbi.nlm.nih.gov/Blast.cgi>) to  
231 check species identity and we removed samples with evidence of species misidentification,  
232 chimerism, or introgression from related species (14 samples removed). We aligned and  
233 translated sequences with the R package DECIPHER v2.18.1 (Wright 2016), and we visually  
234 inspected each alignment, ensuring that sequences contained no premature stop codons or other  
235 alignment issues. We used DECIPHER to remove partial stop codons and the untranslated C in  
236 the ND3 sequence of woodpecker (Picidae) species (Mindell et al. 1998). As our population  
237 analyses require complete data matrices, we excluded individuals with incomplete datasets (those  
238 with assemblies that were missing genes and/or with ambiguous base calls; 202 samples  
239 removed). We removed five individuals during population structure analysis, described below.  
240 This data filtering resulted in 1008 complete mitochondrial coding sequences: 999 individuals

241 across 30 species used in the population genomic analyses plus one sequence for each of the 9  
242 additional species we used only in the interspecific Coevol analyses. We concatenated the 13  
243 mitochondrial coding sequences for analysis. The full list of samples, including those removed  
244 from the analyses, can be found in Table S2.

245

246 *Estimating  $\theta$  as a proxy for  $N_e$*

247 We used  $\theta$  as a proxy for effective population size ( $N_e$ ).  $N_e$  can be calculated based on  $\theta$   
248 and mutation rate (Watterson 1975, Nabholz et al. 2008b; Table 1), but accurate estimates of  
249 mitochondrial mutation rate are lacking for most non-model organisms. Accordingly, many  
250 empirical studies interested in  $N_e$  focus on genetic diversity, which is thought to reflect the  
251 harmonic mean of  $N_e$  over time and which does not require mutation rate information to calculate  
252 (e.g., Ellegren and Galtier 2016; Hague and Routman 2016). We hereafter use the genetic  
253 diversity parameter  $\theta$  as a proxy for  $N_e$ . We used LAMARC v2.1.10 (Kuhner 2006) to estimate  $\theta$   
254 for each species. LAMARC estimates  $\theta$  in a maximum likelihood framework using information  
255 about the intervals between coalescence events from sampled genealogies, which the program  
256 generates from population sequence data (Felsenstein 1992; Kuhner et al. 1995; Kuhner 2006).  
257 We imported our population-level full mitochondrial coding sequence data into LAMARC after  
258 converting our concatenated fasta files into the phylip format for each species. We used the  
259 program's likelihood-based method in 10 initial chains (samples = 500, discard = 1000, interval  
260 = 20) and 2 final chains (samples = 10,000, discard = 1000, interval = 20). We used the F84  
261 model of molecular evolution, and we provided a separate transition/transversion ratio for each  
262 species using values we calculated from population mitochondrial coding sequence datasets  
263 using the R package 'spider' (Table S1; Brown et al. 2012). All other input parameters were left

264 at their default values. We examined the output for each species to check for chain convergence,  
265 and we ran two replicate chains for each species to make sure they produced consistent results.  
266 For five species (*Leiothlypis ruficapilla*, *Setophaga castanea*, *Setophaga coronata*, *Setophaga*  
267 *fusca*, and *Vireo olivaceus*), we repeated LAMARC for 25 initial chains instead of 10 to improve  
268 convergence and used the values from these longer runs.

269 Estimation of  $\theta$  can be biased by purifying selection, and the magnitude of this bias may  
270 vary across species due to differences in purifying selection and sample size (Subramanian  
271 2016). To evaluate whether these biases influence our results, we compared  $\theta$  to  $\pi_S$ , or nucleotide  
272 diversity at synonymous polymorphisms, which is not biased by purifying selection assuming  
273 that synonymous sites are evolving neutrally. We estimated  $\pi_S$  from each species using the  
274 python package egglib v3.1.0 (De Mita and Siol 2012) and calculated Pearson's correlation  
275 coefficient between  $\theta$  and  $\pi_S$ . We also repeated Coevol models (described below) with each  
276 proxy of  $N_e$  to assess whether the choice of proxy influences our results.

277

#### 278 *Population Structure*

279 Our population-level analyses (estimation of  $\theta$  and  $\pi_N/\pi_S$ ) assume no geographic  
280 population genetic structure within the samples used. To check this assumption, we calculated  
281 mitochondrial genetic distance between all individuals within each species using “nei.dist()”  
282 from the R package poppr v2.9.3 (Kamvar et al. 2014) and created a neighbor-joining tree with  
283 “nj()” from the R package ape v5.6-2 (Paradis and Schleip 2019). We identified and removed 4  
284 individuals from *Regulus satrapa* and one individual from *Oporornis agilis*, all from Alberta in  
285 the far western part of our sampling area, that were genetically distinct from all other samples in

286 their respective species. Otherwise, there was little evidence of geographic genetic structure in  
287 the mitochondrial genome in these species.

288

289 *Estimating  $d_S$  and  $d_N/d_S$  and their correlations with traits associated with life history*

290 We used Coevol v1.6 (Lartillot and Poujol 2011) to evaluate associations between  
291 migration distance and molecular evolutionary rates using a single representative of each species.  
292 Coevol uses a Bayesian phylogenetic framework to estimate  $d_S$  and  $d_N/d_S$  and to simultaneously  
293 measure the relationship between these traits and covariates of interest (migration distance, mass,  
294 and  $\theta$ ). We included mass to account for the expected relationship between mass and molecular  
295 rates (Nabholz et al. 2016). Models with mass also provide a useful point of comparison,  
296 allowing us to ask whether migration distance correlates with  $d_S$  and  $d_N/d_S$  to the same extent as  
297 (or more or less than) this well-studied life-history trait. Similarly, including  $\theta$  in the models  
298 allows us to assess whether variation in  $N_e$  accounts for differences in molecular evolutionary  
299 rates.

300 We provided Coevol with one complete concatenated mitochondrial coding sequence  
301 from each species and a phylogenetic tree (Fig. 2) we generated with data from birdtree.org (Jetz  
302 et al. 2012) as described in Pegan and Winger (2020). In brief, we sampled 2000 trees  
303 comprising all North American bird species from the Jetz et al. dataset, and we used the python  
304 package “DendroPy” (Sukumaran and Holder 2010) to generate a consensus tree. We then  
305 trimmed this tree to include only the 39 species used in this study. Importantly, Coevol uses the  
306 phylogenetic tree for topological information but estimates relative branching times from the  
307 sequence data (Lartillot and Poujol 2021). Coevol also does not require prior information about  
308 mutation rates. We investigated the potential effects of phylogenetic tree topology on our results

309 by sampling 10 random marginal trees from the original tree dataset (trimmed to include only  
310 relevant species) and re-running Coevol on each tree, which we found to produce consistent  
311 results (Table S3).

312 We created two data subsets for Coevol models: one subset contained all species in the  
313 study and included mass and migration distance as covariates. The other subset included the 30  
314 species for which we had population-level data available; for these, we included  $\theta$  as a covariate  
315 in addition to mass and migration distance. We also repeated these analyses using  $\pi_S$  as a proxy  
316 for  $N_e$  instead of  $\theta$ . For each data subset, we ran Coevol four times: two repeated analyses with  
317 the option “dnds” (estimating  $d_S$ ; models 1 and 2, Table 2) and two with “dsom” (estimating  
318  $d_N/d_S$ ; models 3 and 4, Table 2). We let each analysis run for approximately 20000 steps and  
319 examined the resulting trace files to ensure convergence and evaluate estimated sample sizes  
320 (ESS). All models converged, and all parameters had ESS > 300. We removed the first 500 steps  
321 of each analysis and thinned the posterior sample to retain every 10<sup>th</sup> step to reduce  
322 autocorrelation. Replicate analyses produced highly similar estimates, and the values we report  
323 here represent the mean value of estimates made by each replicate. We present full Coevol model  
324 output in Tables S4-S6.

325 The method implemented in the Coevol software estimates correlation coefficients  
326 between substitution rates and each covariate, as well as partial correlation coefficients (which  
327 hold constant the effects of other covariates in the model). Each correlation or partial correlation  
328 coefficient is accompanied by a posterior probability. In the case of Coevol, posterior  
329 probabilities near 0 indicate strong support for a negative relationship, while posterior  
330 probabilities near 1 indicate strong support for a positive relationship (Lartillot and Poujol 2021).

331

332  $\pi_N/\pi_S$

333  $\pi_N/\pi_S$  is measured by comparing polymorphisms among individuals within a species

334 rather than by comparing between species in a phylogenetic framework (and thus cannot be

335 estimated by Coevol). We estimated  $\pi_N/\pi_S$  from each species with population-level fasta

336 alignments, using the python package egglip v3.1.0 (De Mita and Siol 2012) to create a

337 “CodingDiversity” class with attributes describing nucleotide diversity at codons with

338 synonymous or nonsynonymous polymorphisms. Predictions about the effect of purifying

339 selection on polymorphisms are more complex than predictions about substitution rates because

340 within-population variation can be purged by strong directional selective sweeps in addition to

341 purifying selection (Kryazhimskiy and Plotkin 2008). We predict a negative relationship between

342 migration distance and the  $\pi_N/\pi_S$  ratio, indicating stronger selection (directional or purifying) on

343 mitochondrial function in long-distance migrants. We used linear modeling to test for an effect

344 of migration distance, mass, and  $\theta$  on  $\pi_N/\pi_S$  (Tables S7, S8). Prior to linear modeling, we centered

345 and scaled our predictors using the function “standardize” from the R package “robustHD”

346 (Alfons 2021) with the mean value of each predictor as the center. We used a similar linear

347 modeling approach to test whether  $\theta$  exhibits a relationship with mass or migration distance to

348 ensure that apparent relationships between these traits and molecular rates are not confounded by

349 correlation with  $\theta$ .

350 For each response variable ( $\theta$  and  $\pi_N/\pi_S$ ; Tables S7, S8), we first created a model with all

351 covariates of interest. We then used the function “phylosig()” from the R package phytools v0.7-

352 70 (Revell 2010) to test for phylogenetic signal in the model’s residuals (Revell 2012). For both

353 response variables, the estimate of  $\lambda$  (phylogenetic signal) was low (< 0.2), and the p-value for

354 evidence of phylogenetic signal was > 0.8, so we proceeded with linear modeling rather than

355 using models with phylogenetic covariance matrices. For each response variable, we created a  
356 null (intercept-only) model with no predictors and models with all possible combinations of our  
357 predictors of interest, and we used the function “model.sel()” from the R package MuMIn  
358 v1.43.17 (Bartón 2019) to compare the models’ AICc.

359

## 360 **Results**

361

362 For each model, we report correlation coefficients between traits of interest (migration  
363 distance, mass, or  $\theta$ ) and molecular evolutionary rates ( $d_S$  or  $d_N/d_S$ ) and assess their strength  
364 based on posterior probabilities ( $pp$ ), which are close to 0 in the case of a strong negative  
365 correlation and close to 1 in the case of a strong positive correlation. We also report partial  
366 correlation coefficients and their posterior probabilities, which indicate the relationship between  
367 variables of interest after accounting for the effects of all other covariates.

368 The Pearson correlation coefficient between  $\theta$  and  $\pi_S$  was high (0.77;  $p < 0.0001$ ),  
369 suggesting that these two variables are consistent proxies of  $N_e$ . We found that results of Coevol  
370 models with  $\theta$  as a covariate were consistent with results of models using  $\pi_S$ , so we conclude that  
371 results of analyses with  $\theta$  are not driven by biases in the estimation of  $\theta$ . We hereafter focus on  
372 models using  $\theta$ , and full results of Coevol models using  $\pi_S$  instead of  $\theta$  are presented in Table S6.

373

### 374 *Correlations between migration distance and molecular evolutionary rates ( $d_S$ and $d_N/d_S$ )*

375 Our analyses show that migration distance negatively correlates with  $d_S$  across the 39  
376 species we studied, consistent with our initial predictions (Fig. 2, Fig. S2). For Coevol models  
377 with the full species set, the correlation coefficient relating migration distance to  $d_S$  was -0.39

378 with a posterior probability (*pp*) of 0.018, indicating strong support for a negative relationship.  
379 The partial correlation coefficient (which accounts for mass) between migration distance and  $d_S$   
380 was -0.47 (*pp* = 0.0090).

381 We did not detect evidence of a relationship between migration distance and  $d_N/d_S$   
382 (correlation coefficient = 0.096, *pp* = 0.63). The partial correlation coefficient (accounting for  
383 mass) between migration distance and  $d_N/d_S$  indicated that this relationship was not well  
384 supported (partial correlation coefficient = 0.26, *pp* = 0.82).

385 Results from the Coevol models of the subset of 30 species for which we had estimates of  
386  $\theta$  were consistent with results produced by the full subset (39 species) models, although support  
387 for the correlation between  $d_S$  and migration distance was slightly weaker. In the model  
388 estimating  $d_S$ , migration distance had a correlation coefficient of -0.43 (*pp* = 0.02) and a partial  
389 correlation coefficient of -0.31 (*pp* = 0.11). In the model estimating  $d_N/d_S$ , we did not find  
390 support for a relationship with migration distance, as this variable had a correlation coefficient of  
391 -0.15 (*pp* = 0.32) with  $d_N/d_S$  and a partial correlation coefficient of -0.010 (*pp* = 0.52) with  $d_N/d_S$ .  
392

### 393 *Correlations between mass and molecular evolutionary rates ( $d_S$ and $d_N/d_S$ )*

394 Coevol models with the full species set support the expected negative relationship  
395 between mass and  $d_S$  (correlation coefficient = -0.28, *pp* = 0.065; Fig. 2). This relationship  
396 weakens when effects of migration distance are accounted for (i.e., with partial correlation  
397 coefficient = -0.18, *pp* = 0.20). We did not find a strong correlation between mass and  $d_N/d_S$   
398 (correlation coefficient = -0.25, *pp* = 0.19; partial correlation coefficient = -0.072, *pp* = 0.41). In  
399 models of  $d_S$  with the subset of 30 species that included  $\theta$  as a predictor, mass had a correlation  
400 coefficient of -0.17 (*pp* = 0.21) and a partial correlation coefficient (which controls for the

401 effects of migration distance) of -0.23 ( $pp = 0.15$ ). In models of  $d_N/d_S$  from this subset, mass had  
402 a correlation coefficient of 0.16 ( $pp = 0.7$ ) and a partial correlation coefficient of 0.26 ( $pp =$   
403 0.84).

404

405 *The influence of  $N_e$  on molecular evolutionary rates*

406 In models using the subset of 30 species with population-level data, we did not find  
407 evidence for a correlation between  $\theta$  and  $d_S$  (correlation coefficient = -0.23,  $pp = 0.15$ ; partial  
408 correlation coefficient = -0.12,  $pp = 0.67$ ). This result is consistent with neutral evolution of  
409 synonymous sites among the species we studied. By contrast, we found strong support for the  
410 nearly neutral theory's predicted negative relationship (Ohta 1992; Popadin et al. 2007; Leroy et  
411 al. 2021) between  $\theta$  and  $d_N/d_S$  (correlation coefficient = -0.60,  $pp = 0.025$ ; partial correlation  
412 coefficient = -0.57,  $pp = 0.031$ ; Fig. 3), indicating stronger purifying selection in species with  
413 higher  $N_e$ .

414

415 *Linear modeling of  $\pi_N/\pi_S$*

416 In comparison of AICc, the highest-ranked model of  $\pi_N/\pi_S$  showed a strongly supported  
417 negative relationship between  $\theta$  and  $\pi_N/\pi_S$  (Fig. 4, Table S7, model weight 0.55), as predicted if  
418 purifying selection is stronger in species with higher  $N_e$ . Compared to a model with  $\theta$  alone, a  
419 model with both  $\theta$  and migration distance shows an increase in multiple  $r^2$  from 0.15 to 0.28 and  
420 a decrease in AICc by more than two units, suggesting the inclusion of migration distance  
421 improves model fit. However, contrary to our prediction, migration distance has a weak positive  
422 relationship with  $\pi_N/\pi_S$  (Fig. 4). The estimated coefficient relating  $\theta$  and  $\pi_N/\pi_S$  in the best-fit  
423 model is -0.027 (std error = 0.01) and the estimated effect of migration distance from the best-fit

424 model is 0.022 (std error = 0.01). Model comparison did not support the inclusion of mass as a  
425 predictor of  $\pi_N/\pi_S$  (Table S7).

426

427  *$N_e$  does not confound patterns of rate correlations*

428 We used linear modeling to test whether migration distance or mass show a relationship  
429 with  $\theta$ , our proxy of  $N_e$ . We did not find strong evidence that mass or migration distance are  
430 correlated with  $\theta$  among the 30 species we studied. The null model for  $\theta$  (an intercept-only  
431 model with no predictors) showed the lowest AICc, suggesting that the addition of mass and  
432 migration distance as predictors did not improve model fit (Table S8, model weight 0.45).  
433 However, the model with migration distance as a predictor was within 2 AICc units of the null  
434 model and showed a model weight of 0.30, indicating considerable model uncertainty. The  
435 estimated effect of migration distance on  $\theta$  was positive but had a negligible effect size in the  
436 second-best model (estimate = 0.0017, std error = 0.0013 model multiple  $r^2 = 0.054$ ).

437

438 **Discussion**

439

440 *Seasonal migration distance correlates with mitochondrial  $d_S$*

441 We examined the relationship between life history and patterns of mitochondrial  
442 sequence evolution within North American boreal birds. These species occupy a region where  
443 strong seasonality demands specialized adaptations that carry life history tradeoffs (Varpe 2017;  
444 Winger and Pegan 2021). Our results implicate the life-history axis of seasonal migration  
445 distance as a novel correlate of mitochondrial synonymous substitution rate ( $d_S$ ). Previous work  
446 demonstrates that, even after accounting for body size, long-distance migrants in this system

447 have slower life history strategies than short-distance migrants, showing higher annual adult  
448 survival and lower fecundity (Winger and Pegan 2021). Here, we find that the slow life history  
449 of long-distance migrants is accompanied by a slower rate of neutral molecular evolution in the  
450 mitochondria of these species compared with that of shorter-migrating species in the region.  
451 Indeed, among the 39 species we studied, the correlation between migration distance and  $d_S$  is  
452 stronger than the correlation between mass and  $d_S$ , which is notable given that the relationship  
453 between mass and substitution rate has been documented in previous work (Nabholz et al. 2016).  
454 As such, we suggest that the association between migration distance and the slow-fast life history  
455 continuum extends to effects on  $d_S$ .

456

457 *What evolutionary processes link migration distance with mitochondrial  $d_S$ ?*

458 Substitution rates are fundamentally influenced by mutation rate, which provides new  
459 molecular variants with potential to become substitutions, and by natural selection, which  
460 influences whether variants are fixed as substitutions or lost. The correlation between migration  
461 distance and  $d_S$  therefore reflects one or both processes.  $d_S$  is often treated as a proxy for  
462 mutation rate alone based on the assumption that natural selection does not operate on  
463 synonymous sites (Nei et al. 2010), but in some cases, synonymous sites are known to evolve  
464 non-neutrally (Chamary et al. 2006; Künstner et al. 2011; Wei et al. 2014; Wynn and Christensen  
465 2015). If synonymous sites are not evolving neutrally, nearly neutral theory suggests that the  
466 relationship between  $d_S$  and migration distance could be explained by larger  $N_e$  in long-distance  
467 migrants (Ohta 1992). We tested the key assumption that synonymous sites evolve neutrally by  
468 assessing the relationship between  $d_S$  and our proxy for  $N_e$  ( $\theta$ ) (Table S4). We found no  
469 correlation, suggesting that synonymous sites are indeed evolving neutrally in our system. We

470 also found no correlation between  $\theta$  and migration distance (Table S8). Together, these results  
471 suggest that variation in  $d_S$  among species with different migration distances is not well  
472 explained by variation in natural selection or effective population size. Rather, we suggest that  
473 the negative relationship between migration distance and  $d_S$  may reflect a negative relationship  
474 between migration distance and mutation rate.

475

476 *Why might long-distance migrants have a lower mitochondrial mutation rate?*

477 We predicted that migration distance would correlate with  $d_S$  because of its relationship  
478 with the slow-fast continuum of life history in these species independent of body size (Winger  
479 and Pegan 2021). In turn, a species' position on the slow-fast life history continuum is  
480 hypothesized to affect mutation rate (Bromham 2020). There are several potential mechanisms to  
481 explain the link between life history and mutation rate, and the relative importance of each is not  
482 clear (Bromham 2020). The "copy error effect" hypothesis suggests that the explanation is  
483 related to generation time, assuming that "fast" species with short generation times and young  
484 age at first reproduction experience higher rates of germline replication (and thus replication-  
485 induced mutation) than species with "slow" life histories (Li et al. 1996; Thomas et al. 2010;  
486 Lehtonen and Lanfear 2014).

487 However, recent studies comparing cell division rates with directly-measured mutation  
488 rates suggest that replication-induced copy errors may not be the only driver of differences in  
489 mutation rate between lineages (Wu et al. 2020; Wang et al. 2022). The "mutation avoidance"  
490 hypothesis offers another non-exclusive explanation for lower  $d_S$  in organisms with slow life  
491 history based on higher mutation costs in longer-lived species (Bromham 2020). Under this  
492 hypothesis, organisms with slow life history are predicted to have adaptations that reduce the

493 introduction of mutations from DNA damage or DNA replication and repair processes (Galtier et  
494 al. 2009; Tian et al. 2019; Zhang et al. 2021; Cagan et al. 2022). Long-distance migrants may be  
495 especially sensitive to the costs of mitochondrial mutation, which may cause mitochondrial  
496 senescence (Galtier et al. 2009; Hua et al. 2015), because of the high physical performance  
497 demanded by their migratory behavior across their entire lifespans (Møller 2007; Conklin et al.  
498 2017). Further research is necessary to understand what processes contribute to the apparent  
499 reduction of mutation rate in species at the slow end of the slow-fast continuum of life history.

500 Another possible link between migration distance and mutation rate is oxidative damage  
501 from metabolism, which is recognized as a potential source of mutation rate variation (Martin  
502 and Palumbi 1993, Gillooly et al. 2005, Berv and Field 2018; but see Lanfear et al. 2007, Galtier  
503 et al. 2009). Thus, a potential explanation for our results—lower mitochondrial  $ds$  in long-  
504 distance migrants—is that long-distance migrants incur less metabolically-induced DNA damage  
505 than short-distance migrants. This explanation is initially surprising in light of studies showing  
506 that migratory birds experience oxidative damage from endurance flight (Jenni-Eiermann et al.  
507 2014; Skrip and McWilliams 2016). However, we suggest that three plausible and non-exclusive  
508 scenarios could lead to lower metabolically-induced DNA damage in long-distance compared to  
509 short-distance migrants. First, long-distance migrants may have better adaptations for flight  
510 efficiency (Weber 2009; Elowe et al. 2023), reducing the oxidative damage they experience per  
511 mile traveled. Second, the mutation avoidance hypothesis predicts that long-distance migrants  
512 may have more efficient DNA repair mechanisms than short-distance migrants, which could  
513 reduce metabolically-induced mutation rates even when long-distance flight does induce high  
514 oxidative stress. Last, short-distance migrants in our boreal study system may experience greater  
515 oxidative damage arising from their increased need for winter cold tolerance than long-distance

516 migrants that winter in the tropics. The mitochondria also play an important role in the metabolic  
517 challenge of maintaining homeostasis during cold weather and resource shortages (Bicudo et al.  
518 2001; Chen et al. 2018). Short-distance boreal migrants likely face more of these kinds of  
519 challenges than long-distance migrants during migration and winter (Winger and Pegan 2021).  
520 Despite the view that long-distance migration is an extreme performance challenge, its  
521 alternative—spending the winter within the temperate zone—is also an extreme metabolic  
522 challenge for small-bodied homoeothermic endotherms that do not hibernate (Dawson and  
523 Yacoe 1983; Winger et al. 2019). Further investigation of the comparative metabolic challenges  
524 faced by short versus long-distance boreal migrants is needed to clarify whether and how  
525 migration distance influences metabolically-induced mutation in the mitochondria.

526

527 *Purifying selection is not stronger in long-distance migrants*

528 Whereas evolutionary rate at synonymous sites ( $d_S$ ) may primarily reflect mutation rate,  
529 evolution at nonsynonymous sites is expected to strongly reflect natural selection because  
530 nonsynonymous mutations alter the amino acid sequence of a gene's protein product. We found  
531 that the ratio of nonsynonymous to synonymous substitutions ( $d_N/d_S$ ) among our species is  
532 universally much less than 1 (Fig. 3), indicating that the mitochondrial genes we studied are  
533 under purifying selection in all species in the system. We similarly found low ratios of  
534 nonsynonymous to synonymous polymorphisms within each population ( $\pi_N/\pi_S$ ; Fig. 4), which is  
535 also consistent with purifying selection. Moreover, both  $d_N/d_S$  and the  $\pi_N/\pi_S$  ratio are strongly  
536 correlated with  $\theta$ , our proxy for  $N_e$  (Fig. 3, 4), as expected under nearly neutral theory (Ohta  
537 1992). A nuance of our results is that  $d_N/d_S$  reflects the accumulation of substitutions across the  
538 entire history of a lineage, whereas population parameters such as  $\theta$  and  $\pi_N/\pi_S$  may be more

539 strongly influenced by recent demographic processes. However, that we and others (e.g.,  
540 Popadin et al. 2007; Leroy et al. 2021) find empirical evidence for the relationship between  $\theta$   
541 and  $d_N/d_S$  predicted by nearly neutral theory, despite this potential mismatch in evolutionary  
542 timescales, suggests that similar demographic processes may shape empirical estimates of  
543 genetic diversity and molecular evolutionary rates.

544 Our results are consistent with the general finding that mitochondrial genes tend to  
545 experience strong purifying selection (Nabholz et al. 2013; Popadin et al. 2013). However, we  
546 did not find evidence supporting our prediction that long-distance migrants would show stronger  
547 purifying selection (i.e., lower  $d_N/d_S$  and  $\pi_N/\pi_S$ ) than short-distance migrants. This finding may  
548 reflect the reality that all species in our system face generally strong mitochondrial purifying  
549 selection, such that the endurance flights of long-distance migrants do not incur much stronger  
550 selection than the level that exists among all the species we studied. Our results also imply that  
551 short-distance migrants in the boreal region do not experience *relaxed* purifying selection on  
552 mitochondrial genes compared to long-distance migrants. As noted above, short-distance boreal  
553 migrants contend with metabolic challenges associated with cold winter temperatures which may  
554 also exert selection on the mitochondria (Chen et al. 2018), as well as the metabolic demands of  
555 flight.

556

#### 557 *Migration distance and the costs of mitochondrial mutations*

558 In this study, we based our predictions on several complementary hypotheses about the  
559 costs of mutation in species with slow life history and high demand for physiological  
560 performance, such as long-distance migrants. From the perspective of molecular evolution, the  
561 mutation avoidance hypothesis (Bromham 2020) and studies on the relationship between lifespan

562 and mutation rate (Nabholz et al. 2008a; Galtier et al. 2009; Tian et al. 2019; Zhang et al. 2021)  
563 predict that phenotype-altering genetic variation is harmful enough to induce selection for  
564 mutation avoidance in organisms with slow life history. From the perspective of population  
565 biology, the hypothesis proposed by Conklin et al. (2017) predicts that “slow” species with high  
566 performance demands experience a strong selective filter on phenotypic performance in early  
567 life, reducing phenotypic variation in these populations. While Conklin et al. (2017) frame their  
568 hypothesis around reduction of phenotypic variation, a similar prediction about reduction of  
569 genetic variation emerges from a series of studies showing that mitochondrial purifying selection  
570 is stronger in species with higher locomotory metabolic demands (Shen et al. 2009; Chong and  
571 Mueller 2013; Strohm et al. 2015; Mitterboeck et al. 2017; Sun et al. 2017; Chang et al. 2020; De  
572 Panis et al. 2021). Together, these hypotheses led us to predict that the costs of mitochondrial  
573 mutation in long-distance migrants, which have slow life histories, would cause them to exhibit  
574 slower mitochondrial mutation rates and stronger mitochondrial purifying selection than short-  
575 distance migrants.

576 Our predictions were only partially supported. The negative relationship we found  
577 between migration distance and  $d_S$  is consistent with lower mitochondrial mutation rate in long-  
578 distance migrants, but we did not find evidence that these species experience stronger  
579 mitochondrial purifying selection than do short-distance migrants. To reconcile these findings  
580 and advance our understanding of how long-distance migration influences molecular  
581 evolutionary dynamics, further research is needed on the relative metabolic demands of long-  
582 distance flight versus cold tolerance and on the consequences of mitochondrial genetic variation  
583 for migratory phenotype. Additionally, studying molecular rates across the nuclear genome will

584 help clarify which dynamics we report here are related to selection on the mitochondrial genome  
585 and which reflect more general interactions between life history and molecular evolution.

586

587 *Conclusions: seasonal adaptation provides novel context for studying the links between life*  
588 *history and molecular evolutionary rates*

589       Adaptation to seasonality entails life history tradeoffs (Varpe 2017). Organisms balance  
590 these tradeoffs in different ways, creating variation in life history strategy within communities  
591 that inhabit seasonal environments (e.g., Winger and Pegan 2021). Our study demonstrates that  
592 life history variation related to seasonality can influence molecular evolutionary rates, which has  
593 implications for the accurate reconstruction of evolutionary history (Berv and Field 2018; Shafir  
594 et al. 2020; Ritchie et al. 2022). More broadly, communities adapted to seasonal habitats provide  
595 an important context to investigate potential drivers of the relationship between life history and  
596 molecular evolution. Co-distributed species show varying adaptations to seasonality—e.g., cold  
597 tolerance, migration, hibernation—and they express these strategies to different degrees (Auteri  
598 2022). Cold adaptations can influence biological processes hypothesized to be relevant for  
599 germline replication rate or mutation rate (e.g., Wang et al. 2022), even among species that show  
600 little variation in commonly-studied life history proxies such as body mass. Comparative studies  
601 using seasonal communities can therefore allow us to draw new insights into how life history  
602 tradeoffs affect mutation rate, one of the most fundamental processes in evolution.

603

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607

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609

610 **Data accessibility statement:** Sequence data are available on GenBank. Accession numbers  
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646

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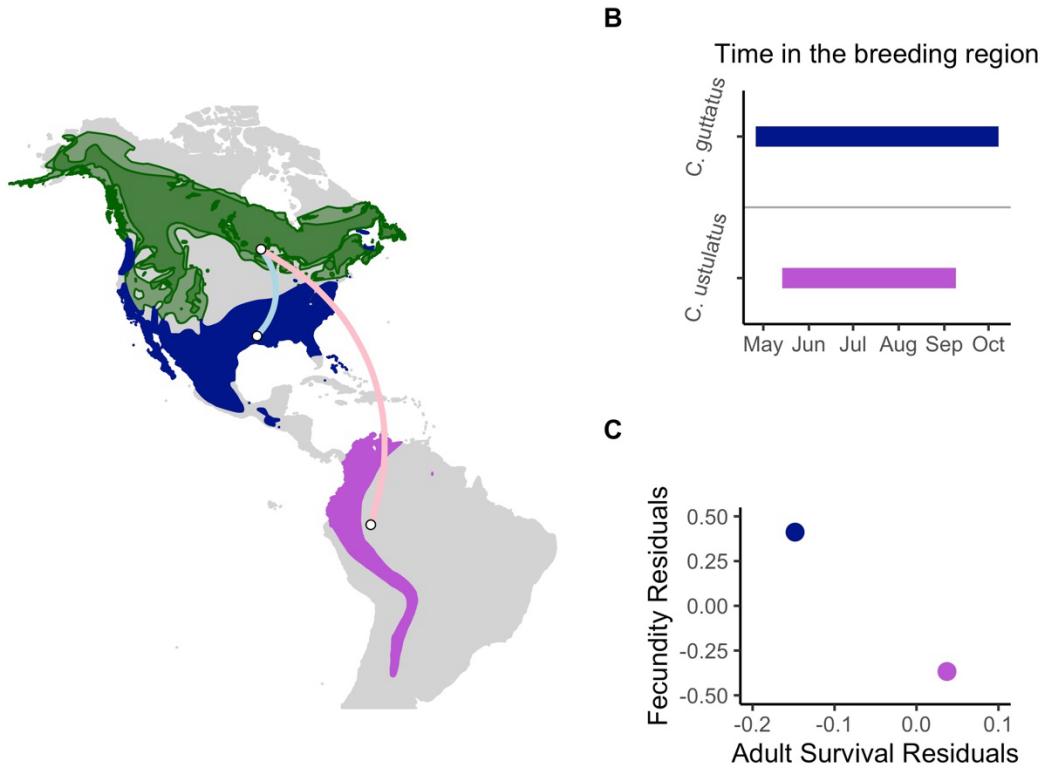
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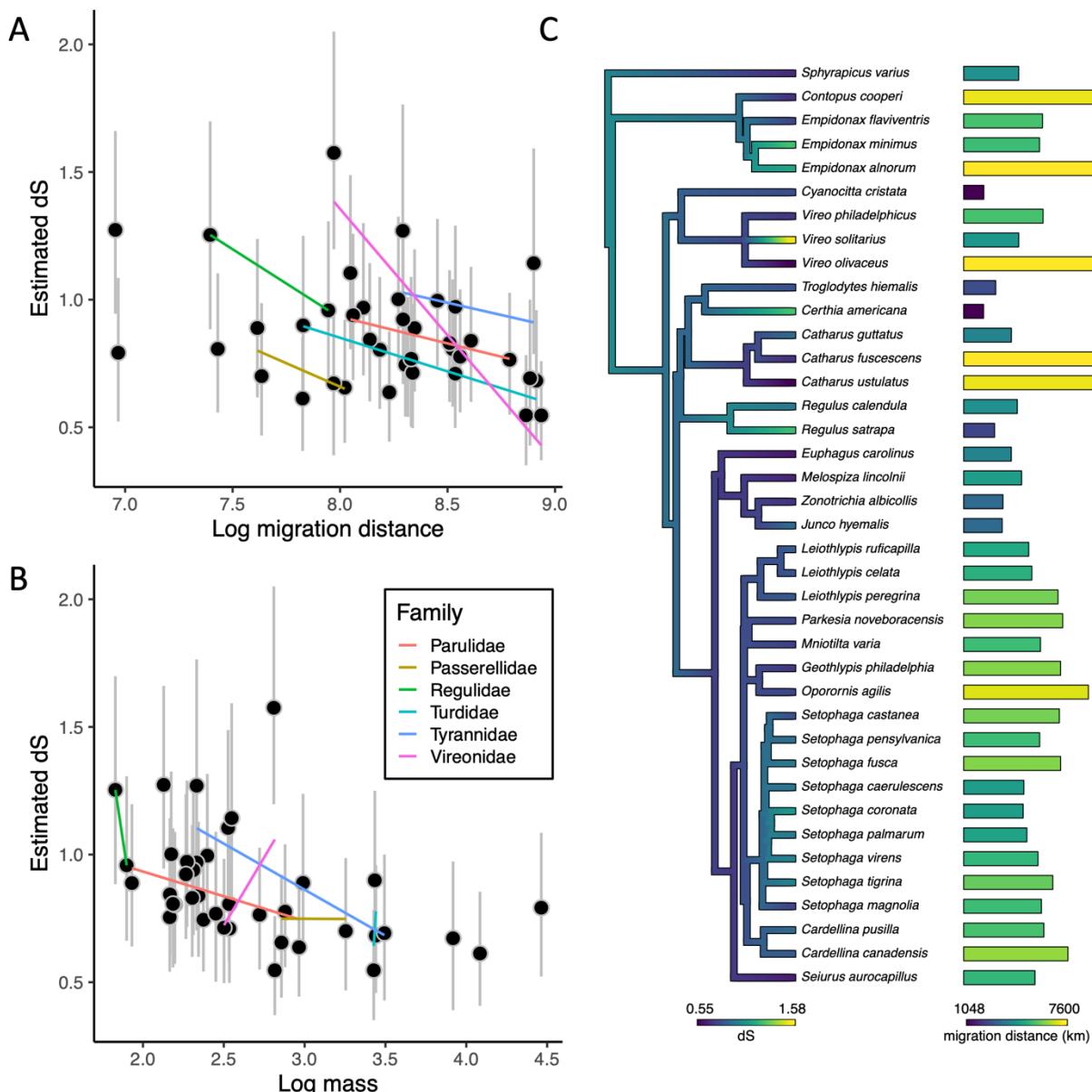
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**Figure 1.** An example contrast between a shorter-distance migrant *Catharus guttatus* and a closely related longer-distance migrant *Catharus ustulatus swainsoni* illustrates the relationship between migration distance and life history in our study system. Both species have broadly overlapping breeding ranges (green), but *C. guttatus* (dark blue nonbreeding range) migrates a shorter distance (blue migratory route) than *C. u. swainsoni* (purple nonbreeding range, pink migratory route) (panel A). Accordingly, *C. guttatus* spends more time in its breeding range than *C. u. swainsoni* (panel B). With more time in the breeding range and the possibility of raising a second brood, the short-distance migrant has higher fecundity but lower adult survival—i.e., faster life history—than the long-distance migrant (panel C, showing model residuals from mass-corrected analysis of fecundity and survival). The short-distance migrant spends the winter in colder, more resource-depleted regions than the long-distance migrant.

914 Figure and data adapted from Winger and Pegan (2021). Our sampling for this study occurred  
915 only within the eastern boreal belt (Fig. S1).

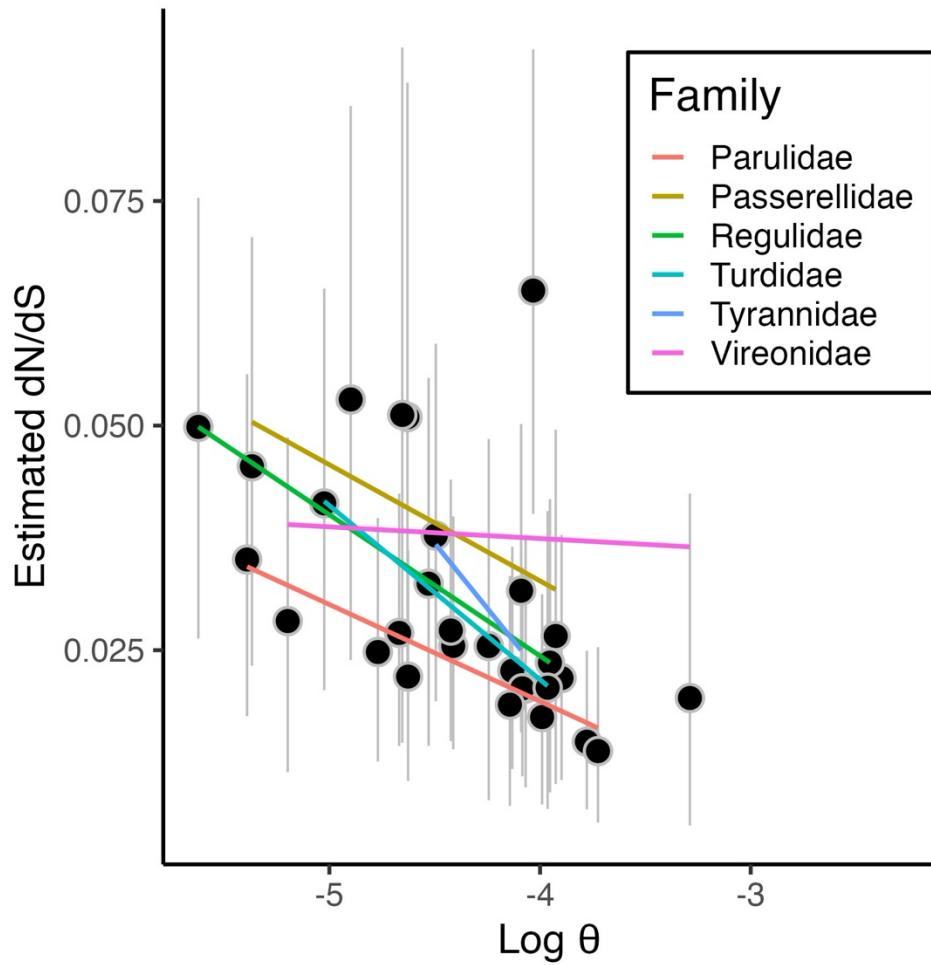
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920 **Figure 2.**  $d_S$  versus traits associated with life history (A, B) and a phylogenetic tree showing  $d_S$   
 921 and migration distance for each species (C). In panels A and B, posterior mean tip estimates of  $d_S$   
 922 (black dots) from Coevol are shown compared to migration distance (A), and mass (B) from  
 923 models using our full species set. Gray vertical bars indicate 95% credible intervals for each

924 estimate. These analyses reveal that both migration distance and mass have a negative  
925 relationship with  $d_S$ . Plotted lines use linear models to visualize the relationship between  
926 estimated tip  $d_S$  and a given covariate within each family of birds (when represented in our  
927 dataset by two or more species), demonstrating a consistently negative relationship between  $d_S$   
928 and migration distance within and among major clades in our system. In panel C, the  
929 phylogenetic tree was created in phytools (Revell 2012) and is colored based on posterior mean  
930 tip and node estimates of  $d_S$  from Coevol.

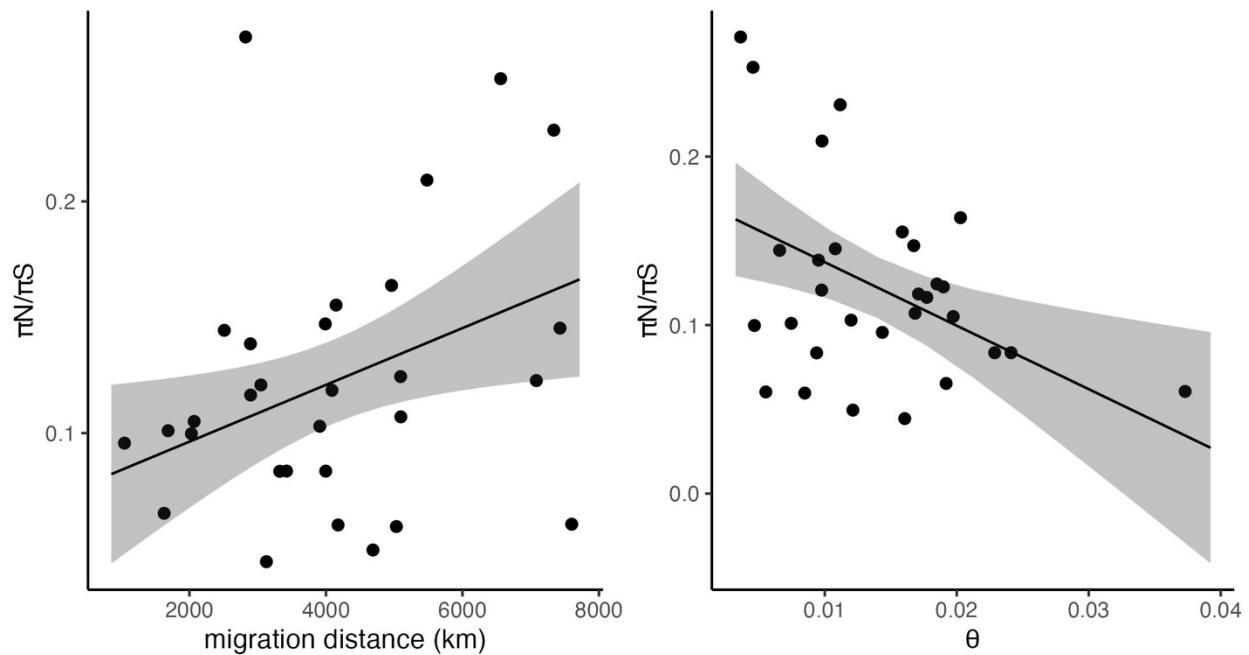
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934 **Figure 3.**  $d_N/d_S$  versus  $\theta$ . Posterior mean tip estimates (black dots) of  $d_N/d_S$  are shown compared  
935 to  $\theta$  from a Coevol model including species for which we could estimate  $\theta$ . Gray vertical bars  
936 indicate 95% credible intervals for each estimate. As in Fig. 2, plotted lines use linear models to  
937 visualize the relationship between mean tip  $d_N/d_S$  and  $\theta$  within each family of birds (when  
938 represented in our dataset by two or more species), demonstrating a consistently negative  
939 relationship between  $\theta$  and  $d_N/d_S$  within and among major clades in our system.

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942 **Figure 4.** The relationship between  $\pi_N/\pi_S$  and migration distance (left) and  $\theta$  (right).  $\pi_N/\pi_S$  is  
 943 strongly influenced by  $\theta$ , as expected if purifying selection removes more nonsynonymous  
 944 variation in species with larger  $N_e$ .  $\pi_N/\pi_S$  increases with migration distance, after accounting for  
 945 effects of  $\theta$ . Regression lines and 95% confidence intervals show the marginal effect of each  
 946 variable as calculated by “`ggpredict()`” from the R package `ggeffects` v0.16.0 (Lüdecke 2018)  
 947 using the best-fit model, which included both predictors.

948

949 **Table 1.** Definitions of abbreviations for molecular substitution rates and population genetic  
 950 parameters and predictions for their relationships with migration distance.  
 951

Concept	Abbr.	Description and assumptions	Predictions (this study)
Synonymous substitution rate	$ds$	Assuming synonymous sites evolve neutrally, $ds$ primarily reflects $\mu$ (Nei et al. 2010; Lanfear et al. 2014)	Negative relationship between migration distance and $ds$
Nonsynonymous substitution rate	$d_N$	Assuming nonsynonymous sites are generally deleterious, $d_N$ is influenced by both $\mu$ and $N_e$ (reviewed in Nei 2005)	NA
$d_N/ds$ ratio	$d_N/ds$	Assuming nonsynonymous mutations are generally deleterious, $d_N/ds$ reflects strength of purifying selection on $d_N$ while accounting for variation in $\mu$ . Low $d_N/ds$ = strong purifying selection. (Nei 2005; Kryazhimskiy and Plotkin 2008)	Negative relationship between $\theta$ and $d_N/ds$ , reflecting the influence of $N_e$ on $d_N/ds$ . Negative relationship between migration distance and $d_N/ds$ , indicating positive relationship between migration distance and purifying selection strength.
Mutation rate	$\mu$	May be influenced by life history; reviewed in (Bromham 2020)	NA, $\mu$ not measurable in our data
Effective population size	$N_e$	Defined as the ideal population size experiencing the same level of genetic drift as observed in the data (Waples 2022). Estimated in mitochondrial data as $\theta / \mu$ . (Watterson 1975; Nabholz et al. 2008a)	NA, see $\theta$
Theta	$\theta$	Population genetic parameter representing genetic variation. Assuming low variation in $\mu$ , variation in $\theta$ primarily reflects variation in $N_e$	Negative relation between $\theta$ and $d_N/ds$ and between $\theta$ and $\pi_N/\pi_S$
Synonymous nucleotide diversity	$\pi_S$	Population genetic parameter representing population-level nucleotide diversity at synonymous sites.	NA
Nonsynonymous nucleotide diversity	$\pi_N$	Population genetic parameter representing population-level nucleotide diversity at synonymous sites.	NA
$\pi_N/\pi_S$ ratio	$\pi_N/\pi_S$	Reduction of $\pi_N$ compared to $\pi_S$ is expected to reflect natural selection, but the relationship is more complex than with $d_N/ds$	Negative relationship between migration distance and $\pi_N/\pi_S$ , indicating positive relationship between migration distance and selection. Negative relationship between $\theta$ and $\pi_N/\pi_S$ , indicating purifying selection on nonsynonymous polymorphisms.

952  
 953

954 **Table 2.** A summary of analyses. Models 1 and 2 use Coevol test our hypothesis that  
 955 synonymous substitution rate ( $d_S$ ) is influenced by migration distance, with mass and  $\theta$  as  
 956 additional covariates. Models 3 and 4 use the same approach with Coevol to estimate  
 957 correlations between traits of interest and  $d_N/d_S$ . Models including  $\theta$  use only 30 species because  
 958 we did not have population-level data available to estimate  $\theta$  for all 39 species. Coevol does not  
 959 analyze molecular evolutionary parameters based on population-level data, so we used linear  
 960 modeling to test whether traits of interest influence  $\pi_N/\pi_S$  (model 5). Finally, we also used linear  
 961 modeling to test for potential confounding relationships between  $\theta$  and life history-associated  
 962 traits of interest (mass and migration distance; model 6).

		<b>Data subset</b>	<b>Method</b>
1	$d_S \sim \text{migration distance} + \text{mass}$	full (39 species)	Coevol
2	$d_S \sim \text{migration distance} + \text{mass} + \theta$	theta (30 species)	Coevol
3	$d_N/d_S \sim \text{migration distance} + \text{mass}$	full (39 species)	Coevol
4	$d_N/d_S \sim \text{migration distance} + \text{mass} + \theta$	theta (30 species)	Coevol
5	$\pi_N/\pi_S \sim \text{migration distance} + \text{mass} + \theta$	theta (30 species)	linear modeling
6	$\theta \sim \text{migration distance} + \text{mass}$	theta (30 species)	linear modeling

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964