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1 **Title:** Island biogeography revisited: Museomics reveals affinities of shelf island birds
2 determined by bathymetry and paleo-rivers, not by distance to mainland

3
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35 **ABSTRACT**

36 Island biogeography is one of the most powerful subdisciplines of ecology: its mathematical
37 predictions that island size and distance to mainland determine diversity have withstood the
38 test of time. A key question is whether these predictions follow at a population-genomic
39 level. Using rigorous ancient-DNA protocols, we retrieved ~1000 genomic markers from
40 ~100 historic specimens of two Southeast Asian songbird complexes from across the Sunda
41 Shelf archipelago collected 1893–1957. We show that the genetic affinities of populations on
42 small shelf islands defy the predictions of geographic distance and appear governed by Earth-
43 historic factors including the position of terrestrial barriers (paleo-rivers) and length of
44 persistence of corridors (Quaternary land bridges). Our analyses suggest that classic island-
45 biogeographic predictors may not hold well for population-genomic dynamics on the
46 thousands of shelf islands across the globe, which are exposed to dynamic changes in land
47 distribution during Quaternary climate change.

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69 INTRODUCTION

70 Islands contribute disproportionately towards our understanding of the evolutionary process
71 (MacArthur and Wilson 1967; Whittaker et al. 2017). They provide a window into the
72 workings of the evolutionary forces of isolation, migration, speciation and extinction, as well
73 as the interactions among them (Borregaard et al. 2016; MacArthur and Wilson 1967;
74 Whittaker et al. 2017). MacArthur and Wilson's (1967) seminal work on island biogeography
75 suggested that the two main factors determining the species diversity of a given island
76 depended upon its area and distance from the mainland. However, island size and distance to
77 mainland may not provide a complete picture (Borregaard et al. 2016; Fernández-Palacios et
78 al. 2016; Weigelt et al. 2016; Whittaker et al. 2017), with the role of Quaternary glaciations
79 and concomitant sea-level fluctuations also pertinent to island biogeography (Fernández-
80 Palacios et al. 2016; Weigelt et al. 2016). Cyclical periods of global cooling and warming
81 over the past ~2.6 million years have not only been an engine of diversification at higher
82 latitudes and altitudes, but also affected gene flow and speciation patterns in shelf-island
83 regions (i.e. islands on continental shelves), especially in Southeast Asia (Brown et al. 2013;
84 Chattopadhyay et al. 2017; Cros et al. 2020a; Garg et al. 2018; Garg et al. 2020; Heaney
85 1986; Hewitt 2000; Hewitt 2004; Lim and Sheldon 2011; Moyle et al. 2009; Ng et al. 2017;
86 Rheindt et al. 2020). Many present-day shelf-islands have been connected by land bridges
87 during cooler Quaternary periods with lower sea levels, allowing for possible dispersal of
88 terrestrial organisms and gene flow (Cros et al. 2020a; Garg et al. 2018; Leonard et al. 2015;
89 Lim et al. 2011; Lim et al. 2017; Lim and Sheldon 2011; Ng et al. 2017).

90
91 A key question is understanding the mechanisms by which Quaternary land bridges provide a
92 conduit of gene flow for terrestrial organisms. Such land bridges can be highly species-
93 specific conduits of gene flow, depending upon ecological requirements, and may be
94 semipermeable by allowing more gene flow in one direction than in the other (Cros et al.
95 2020a; Garg et al. 2018). However, a precise quantification of the amount of gene flow that is
96 facilitated by Quaternary land bridges remains problematic, as many organismic groups are
97 capable of active or passive overwater dispersal (Mayr et al. 2001), obviating the need to
98 evoke land bridges as the sole explanatory variable for dispersal events among present-day
99 shelf-islands (Ng et al. 2017).

100

101 In this study, we directly tested the importance of Quaternary land bridges in facilitating gene
102 flow through population-genomic analysis of two terrestrial rainforest songbird species

103 complexes across the Sundaic Region ('Sundaland') in Southeast Asia. Sundaland constitutes
104 the most complex shelf-island archipelago in the world, encompassing the present-day
105 landmasses of Sumatra, Java, Borneo, Peninsular Malaysia (henceforth: the Peninsula), plus
106 numerous satellite islands (Fig. 1), and has undergone the most pronounced changes in land
107 distribution globally across the Quaternary cooling cycles of the past 2.6 million years.
108 During Quaternary cooling cycles, sea levels drop as water is locked up in the form of ice at
109 higher latitudes, leading to the emergence of land bridges connecting Sumatra, Java, Borneo
110 and numerous smaller islands with the mainland, forming Sundaland (Fig. 1). The widely
111 disjunct historic distribution of orangutans and Asian rhinoceroses on disjunct islands,
112 separated by ~500 km of open sea, bears testament to the frequent occurrence of connecting
113 land bridges in the recent past (Mays et al. 2018; Nater et al. 2017). Consequently, Sundaland
114 has become one of the main model regions to test impacts of Quaternary sea level change on
115 evolutionary processes (Cros et al. 2020a; Leonard et al. 2015; Lim et al. 2011; Lim and
116 Sheldon 2011; Lim et al. 2017)

117

118 We leveraged historic museum specimens collected during the 1800s and early 1900s to
119 include samples from remote islands that are difficult to sample in modern times. By
120 determining the genomic affinity of populations on key strategic island groups, we
121 disentangle overwater dispersal from terrestrial dispersal across past land bridges. The study
122 of underwater depth—bathymetry—is important in determining the level of island isolation:
123 for instance, while the small Natuna islands are geographically close to Borneo, bathymetric
124 data indicate that their land connection to Borneo severed far earlier than did their connection
125 to Sumatra and the Peninsula (Fig. 1). A genomic affinity of Natuna populations with those
126 on Sumatra or the Peninsula, despite Natuna's proximity to Borneo, would be a powerful
127 indication that gene flow among these birds is purely governed by the distribution of
128 Quaternary land bridges. Equally, a closer affinity with Borneo would argue for considerable
129 levels of overwater dispersal, especially during periods when intervening water distances
130 were smaller.

131

132 We focus on two widespread and characteristic Sundaic bird species complexes, the black-
133 capped babbler (*Pellorneum capistratum*) and short-tailed babbler (*P. malaccense*), both
134 sedentary rainforest denizens foraging in the understory (Eaton et al. 2016). Both these
135 songbirds are known for their ubiquitous occurrence across a wide range of Sundaic lowland
136 and hill rainforests and their general inability to cross even narrow non-forest gaps, especially

137 open water (e.g. Cros et al. 2020b; Sadanandan and Rheindt 2015; Zakaria et al. 2013). We
138 characterized and compared variation in plumage patterns using a series of museum
139 specimens to assess whether morphological differentiation is congruent with the population-
140 genomic signal. Museums are a treasure trove of important phenotypic and genomic
141 information from remote areas. However, DNA from century-old museum specimens,
142 especially from the tropics, is often heavily degraded and only minuscule amounts can be
143 salvaged (Chattopadhyay et al. 2019; Dabney et al. 2003). Hence, we used a target capture
144 approach to ensure high quality genome-wide data would be obtained from multiple samples.
145 This approach has been successfully utilized previously for similarly degraded samples
146 (Chattopadhyay et al. 2019). At the same time, we implemented multiple cleanup steps and
147 analytical approaches to account for excess DNA damage and increased levels of C to T and
148 G to A substitutions in the ancient DNA of our samples (Chattopadhyay et al. 2019; Dabney
149 et al. 2003).

150

151 We harvested ~1000 genome-wide sequence loci targeting ~1% of the genomes and
152 employed modern approaches of admixture analysis, allowing us to shed light on patterns of
153 isolation and divergence across Sundaland and the role of Quaternary land bridges in
154 dictating dispersal. We tested whether distance to the mainland is the main determinant of
155 population genomic patterns, or whether bathymetric data—reflective of Quaternary land
156 bridges and the course of paleo-rivers—define the population genomic structure of our target
157 species. Based on the first hypothesis, the population genomic affinity of small-island bird
158 populations, e.g. on Natuna, will be similar to that of birds from nearby Borneo (Fig. 1).
159 However, if the history of land connections is more important, then the affinity of these
160 island populations should be with populations from the Peninsula and Sumatra (Fig. 1).

161

162 **RESULTS**

163

164 **Plumage analysis**

165 We compared the plumages of museum specimens of *P. capistratum* and *P. malaccense* (see
166 Supplementary Table 1) deposited at the Lee Kong Chian Natural History Museum
167 (Singapore). Within the *P. capistratum* complex, we observed three groups based on
168 plumage. The most striking plumage difference among populations was the color of the
169 supercilium, which divided specimens into (1) a cluster in western Sundaland (subspecies *P.*
170 *c. nigrocapitatum*; Peninsula, Sumatra and Natuna) characterized by a grey supercilium with

171 thin white shaft streaks, (2) a Bornean cluster (subspecies *P. c. capistratoides* at least from
172 Sarawak and *P. c. morrelli* at least from Sabah) with a wholly white supercilium, and (3) a
173 Javan cluster (subspecies *P. c. capistratum*) with an orange anterior and white posterior
174 supercilium (Supplementary Figs. 1A, 1B and 1C). Similarly, lores were white or pale grey in
175 Javan specimens, mid-grey in western Sundaic specimens, and dark grey in Bornean
176 specimens (Supplementary Figs. 1A, 1B and 1D). All western Sundaic specimens had a black
177 moustachial stripe (Supplementary Fig. 1C) that was absent in Javan and Bornean specimens.
178 Additionally, specimens from Java had a paler tail and back than all other populations
179 (Supplementary Fig. 1D), and a uniquely orange lateral suffusion on the white throat.

180

181 In western Sundaland, we observed consistent differences across specimens from the Thai-
182 Malay Peninsula, Sumatra, and Natuna. Specimens from the Peninsula had a brownish black
183 crown while Natuna and Sumatran specimens usually had a more saturated black crown.
184 Similarly, on Borneo, specimens from Sabah had a paler orange belly and breast than
185 specimens from Sarawak. Interestingly, Sarawak specimens' belly and breast coloration
186 resembled that of specimens from western Sundaland. We did not compare the coloration of
187 legs, bills and iris due to post-mortem color change and incomplete label data.

188

189 We observed a high variability in plumage within the *P. malaccense* complex, much of it
190 seemingly independent of location (Supplementary Figs. 1E, 1F and 1G). Nevertheless, a few
191 geographically consistent differences did emerge: Bornean populations (subspecies *P. m.*
192 *poliogene* and *P. m. saturatum*) differed from those in western Sundaland (subspecies *P. m.*
193 *malaccense*) in having a more olive rather than chestnut back (mantle and wings)
194 (Supplementary Figs. 1E, 1F and 1G). Within Borneo, specimens from Sabah (subspecies *P.*
195 *m. poliogene*) exhibited a more rufous tail while specimens from Sarawak showed a less
196 warm-colored brown tail (Supplementary Figs. 1E and 1G). Interestingly, all the other
197 differences between Sarawak and Sabah also differentiated Sarawak populations from
198 western Sundaland. Specimens from Sarawak appeared the most distinct population based on
199 plumage alone: they had a darker crown, darker ear coverts and darker moustachial stripe
200 than all other taxa. They also had more intensely orange flanks and breast whereas other
201 populations had paler, more apricot-colored flanks and breast (Supplementary Fig. 1E).
202 Again, comparisons of the coloration of legs, bills and iris were impossible because of post-
203 mortem color change and a lack of sufficient label information.

204

205 **Sampling, DNA extraction and raw data filtering**

206 We obtained toepads of specimens of both *P. capistratum* (n = 50) and *P. malaccense* (n =
207 46) from the Lee Kong Chian Natural History Museum (Singapore) and Yale Peabody
208 Museum of Natural History (New Haven, Connecticut) (Supplementary Table 1). These
209 specimens were collected over a period of ~70 years from 1893 to 1957. All historical
210 samples were processed in a separate dedicated facility and with fresh gloves, forceps and
211 scalpels for each sample to avoid cross contamination.

212

213 We successfully isolated DNA from ~80% of historic samples (n = 77). Presence of DNA
214 was confirmed using Qubit and an AATI Fragment Analyzer. No detectable DNA was
215 observed in negative controls. All 77 samples along with negative controls were further
216 processed for library preparation and target enrichment. Target enrichment protocols have
217 been shown to be highly effective for ancient DNA samples (Chattopadhyay et al. 2019). We
218 designed target loci (960 loci) for sequence capture protocols that are useful for both
219 population genomic and phylogenomic studies targeting both conserved exons and variable
220 intronic regions (see methods for details). We supplemented historical DNA with fresh
221 samples from Sabah (n = 12) (Supplementary Table 1) following the fieldwork protocols of
222 Cros et al. (2020a). All enriched libraries were sequenced on multiple lanes of HiSeq 4000
223 (150bp paired end runs). The fresh samples from Sabah were processed separately and
224 sequenced in a dedicated lane. We retained ~0.97 billion reads after cleanup steps
225 (Supplementary Table 1). The average number of reads per sample was ~11 million (standard
226 deviation = ~5 million).

227

228 **Data matrix**

229 For DNA sequence-based analysis, we obtained sequence data for 944 out of 960 target loci
230 designed for both species complexes using the HybPiper pipeline. We removed ten historic *P.*
231 *malaccense* samples from downstream processing as their missing data exceeded 85%. After
232 multiple sequence alignments, we performed stringent filtering of alignments using Gblocks
233 and removed 117 loci from the *P. capistratum* dataset and 385 loci from the *P. malaccense*
234 dataset due to high missing data. After removal of loci < 200bp and Z-chromosomal loci, we
235 retained 652 loci for *P. capistratum* and 314 loci for *P. malaccense*. The total sequence
236 matrix length for *P. capistratum* was 454,712 bp (average locus length = 697 bp; minimum =
237 201 bp; maximum = 4,723bp) and 145,877 bp for *P. malaccense* (average locus length = 465
238 bp; minimum = 201 bp; maximum = 2,607 bp).

239

240 For SNP-based analysis, we generated four different datasets for each species complex
241 (dataset I: all SNPs obtained after mapping to *Mixornis gularis* genome; dataset II: only
242 transversions obtained after mapping to *Mixornis gularis* genome; dataset III: single random
243 SNP per target locus; dataset IV: only transversions obtained after mapping to *Parus major*
244 genome). We retrieved between 960 and 208,186 SNPs for *Pellorneum capistratum* and
245 between 958 and 198,711 SNPs for *P. malaccense* across datasets before filtering (Table 1).
246 For dataset IV, after filtering for linkage, deviations from Hardy-Weinberg equilibrium,
247 neutrality and DNA damage, we retained 40,611 transversions for *P. capistratum* and 34,809
248 transversions for *P. malaccense*. As admixture graph analysis can accommodate SNPs
249 located on the Z chromosome, we included these in our analysis of gene flow. In contrast, for
250 analyzing population structure, we removed the outgroup and any resulting monomorphic
251 loci along with SNPs located on the Z chromosome and retained 38,463 transversions for *P.*
252 *capistratum* and 32,735 transversions for *P. malaccense* (Table 1). The overall level of
253 missing data for dataset IV was less than 10% for both species after all cleanup steps. For the
254 other datasets the number of SNPs after cleanup is summarized in Table 1.

255

256 **Phylogenomic reconstruction**

257 We used both concatenation approaches and species tree reconstruction for phylogenomic
258 analysis (see Methods). Sumatran and Natuna populations were embedded within the
259 peninsular population based on the concatenated maximum likelihood trees in both species
260 complexes (Supplementary Figs. 2A and 2C). Our sole *P. malaccense* individual from the
261 Anambas archipelago (Fig. 1) also formed part of the large peninsular clade (Supplementary
262 Fig. 2C). In *P. capistratum*, the Javan population was distinct from both peninsular/Sumatran
263 and Bornean populations, emerging as sister to the latter (Supplementary Figs. 2A and 2B).
264 In the case of *P. malaccense*, the Sabah population formed a clade distinct from Sarawak and
265 basal to all members of the complex (Supplementary Figs. 2C and 2D).

266

267 **Population structure**

268 We employed multiple approaches to understand population structure within each babbler
269 species complex and observed similar trends of subdivision across all four SNP datasets
270 generated in this study (Figs. 2 and 3, Supplementary Figs. 3–8). Principal component
271 analysis (PCA) and discriminant analysis of principal components (DAPC) suggested three
272 distinct population groupings in each complex in agreement with phylogenomic results (Figs.

273 2A, 2B, 3A and 3B, Supplementary Fig. 2). In *P. capistratum*, the division entailed (1) a
274 western Sundaic group comprising populations from the Peninsula, Sumatra and Natuna, (2)
275 a Javan group, and (3) a Bornean group (Fig. 2, Supplementary Figs. 3–5). In *P. malaccense*,
276 Bornean populations separated into two deeply divergent groups, one from Sarawak and the
277 other from Sabah (Fig. 3, Supplementary Figs. 6–8) in agreement with previously published
278 studies based on mitochondrial DNA (Lim and Sheldon 2011; Sadanandan and Rheindt
279 2015) and the phylogenomic analysis presented in this study (Supplementary Figs. 2C, and
280 2D). The Sarawak cluster emerged as more closely related to the third group consisting of
281 individuals from the Peninsula, Sumatra, Natuna and Anambas (Fig. 3, Supplementary Figs.
282 6–8).

283

284 The results based on the Bayesian clustering program STRUCTURE were congruent with
285 other analyses for *P. capistratum*, in which the western Sundaic cluster (Peninsula, Sumatra,
286 Natuna) separated from the other clusters (Borneo and Java) at $K=2$ (Fig. 2C), with the latter
287 two separating at $K=3$ (Fig. 2C, Supplementary Figs. 3C, 4C, 5C). STRUCTURE results
288 were less clean for *P. malaccense*, in which only Sabah emerged as clearly distinct for $K=3$
289 (and sometimes $K=2$), whereas the Sarawak cluster did not emerge as visually distinct before
290 $K=4$ (Fig. 3C, Supplementary Figs. 6C, 7C and 8C).

291

292 **Gene flow dynamics**

293

294 The application of D-statistics and admixture graph analysis allowed us to infer the
295 complicated nature of gene flow events and dynamics among populations of all the major
296 Sundaic landmasses investigated (Figs. 4 and 5; Supplementary Table 2). For *P. capistratum*,
297 781 out of 63,725 possible graphs tested by qpbrute exhibited a fit with our data when
298 considering all populations. As this number of graphs was computationally intractable for
299 Bayes factor estimation, we performed subsequent qpbrute analysis in two steps. Initially, we
300 included all populations other than Natuna in our analysis and obtained 41 possible solutions.
301 After Bayes factor estimation, we selected ten graphs as the most likely starting models. For
302 the selected ten graphs, we then included the Natuna population, resulting in 12 unique
303 graphs. Following another round of Bayes factor estimation, five of these admixture graphs
304 displayed a good fit with the data (Fig. 4A). These five admixture graphs had identical
305 topologies and only differed slightly in estimates of admixture proportions for the western
306 Sundaic populations (Fig. 4). They suggested a lack of substantial gene flow between

307 lineages from Java, Borneo and western Sundaland, but pronounced allelic contributions into
308 the three western Sundaic populations from unsampled sources, likely now-extinct
309 populations from the north.

310

311 We obtained a single possible admixture graph for *P. malaccense* out of 9,083 unique graphs
312 tested by qpbrute (Fig. 5). This graph supported post-divergence gene flow from Sabah into
313 western Sundaic populations but not into the adjacent Sarawak population, indicating a
314 potentially strong reproductive barrier between the two Bornean lineages. It also suggested
315 various streams of ancestral allelic contributions into most western Sundaic populations from
316 unsampled, possibly extinct sources (Fig. 5).

317

318 **DISCUSSION**

319

320 *Bathymetric topography predicts genomic affinity of island populations*

321 For both songbird species complexes under study, all phylogenomic and population-genomic
322 approaches unanimously confirmed that Natuna and Anambas island populations are firmly
323 embedded with the peninsular—not Bornean—population cluster (Fig. 2 and 3;
324 Supplementary Figs. 2–8). These conclusions were further corroborated by plumage analysis
325 (Supplementary Fig. 1). These insights defy the fact that Natuna is only ~220 km from the
326 nearest populations on Borneo, less than half (~46%) the distance to the nearest population in
327 Peninsular Malaysia (~480 km) (Fig. 1). Our results unequivocally support the hypothesis
328 that the history of land connections, as dictated by sea level changes and bathymetry, has
329 determined the genetic affinity of shelf-island populations, and that overwater dispersal-
330 related processes have been of much less importance.

331

332 The distance of an island to the nearest mainland has long served as one of the central tenets
333 of classical island biogeography in making inferences about island biota (MacArthur and
334 Wilson 1967). Geographic distance to the mainland has been accepted as the natural criterion
335 explaining why many island biotas are exclusively recruited from one landmass versus the
336 other. Well-documented examples include the American origin of most species inhabiting
337 Bermuda (Sterrer et al. 2004), the European provenance of most species on the Azores
338 (Wallace 1872), and the affinities of British animal populations with those from nearby
339 France, rather than with Scandinavian or Central European populations that share more

340 similarities in climatic regimes they are adapted to (Hewitt 2000; Hewitt 2004; Taberlet et al.
341 1998; Teacher et al. 2009).

342

343 An improvement in our understanding of paleo-climate has led to a realization of the
344 importance of Quaternary sea level change in defining the evolutionary history of terrestrial
345 biota (Cros et al. 2020a; Garg et al. 2018; Lim et al. 2011; Heaney et al. 2005; Ng et al.
346 2017), and has resulted in a new appreciation of bathymetry as a crucial indicator of the
347 extent of land bridges which existed only a few thousand years ago (Garg et al. 2018; Rheindt
348 et al. 2020). Inspection of the bathymetric profile of Sundaland suggests that the land
349 connection between the Peninsula and Natuna/Anambas persisted ~1000 years longer than
350 between Borneo and Natuna/Anambas (Sathiamurthy and Voris 2006) (Fig. 1), supporting
351 the importance of bathymetry and paleo-island distribution in determining the genomic
352 composition of smaller island populations. At the macro-evolutionary level, a re-analysis of
353 species diversity patterns across the planet has shown that Quaternary land connections are an
354 important but overlooked parameter in defining island species diversity (Weigelt et al. 2016)
355 and faunal turnover (Lohman et al. 2011). For instance, one of the steepest and most
356 renowned faunal transition zones runs across Wallace's line (Lohman et al. 2011; Huxley
357 1868; Wallace 1860), separating Sundaland from the Australo-Papuan faunal region.
358 Although only separated by narrow straits, land masses to the east and west of Wallace's line
359 harbor biota of extremely different affinities, attesting to the importance of the deep sea
360 trenches that have precluded the formation of land bridges across this narrow gap (Wallace
361 1860).

362

363 At the population-genetic level, there has so far been a distinct lack of understanding whether
364 the genomic composition of terrestrial populations on present-day shelf islands is largely
365 defined by geographic distance to the nearest large landmass, or by bathymetric topography
366 and consequently by the duration of land connections during the Pleistocene. Genome-wide
367 markers have been instrumental in answering whether shelf island populations are mostly the
368 product of overwater dispersal or of dispersal across historic land connections, which would
369 be difficult to ascertain based on phenotypic data alone (Ng et al. 2017). As Quaternary
370 glacial cycles leave a track record within the genome, patterns of genomic diversity can be
371 used to reconstruct evolutionary history (Cros et al. 2020a; Garg et al. 2018; Lim et al. 2017;
372 Ng et al. 2017; Papadopoulou and Knowles 2017; Rheindt et al. 2020).

373

374 *Paleo-rivers: a long-overlooked determinant of population-genetic structure*

375 The role of big rivers in shaping population structure in large areas of tropical rainforest has
376 received much interest. In South America, the Amazon River and its tributaries are known as
377 important barriers between neighboring subspecies and closely related species, many of
378 which have recently been taxonomically upgraded with an improved biological
379 understanding (Burney and Brumfield 2009; Islet et al. 2001; Isler and Maldonado-Coelho
380 2017; Rheindt et al. 2008; Rheindt et al. 2009). Likewise, in Africa, the Congo River is an
381 important divide between young, recently separated species (Prüfer et al. 2012). In Southeast
382 Asia, on the other hand, most rainforest is fragmented archipelagically, and open sea dividing
383 different islands is considered the main shaping force of population structure, with rivers
384 afforded a minor role (Cros et al. 2020a; Lim et al. 2011; Lim et al. 2017; Lohman et al.
385 2011; Mason et al. 2019). At the same time, the present division of Sundaland into a handful
386 of large and numerous smaller islands only represents a snapshot in time, as all larger
387 landmasses in Sundaland have been connected by land bridges for a cumulative ~90% of the
388 last one million years (Cannon et al. 2009; Mason et al. 2019; Sarr et al. 2019). Therefore,
389 rivers may have played a much larger role in shaping population structure here than
390 previously appreciated.

391

392 Our historic museum samples did not encompass a sufficient number of sites to
393 systematically test the divergence effects of large paleo-rivers in Sundaland. However, it did
394 allow us to inspect whether population-genetic divisions are consistent with the river barrier
395 effect specifically in the case of the two small shelf-island archipelagos of Natuna and
396 Anambas. During the global sea-level lows, Natuna has been connected to the Peninsula
397 further west by a hilly watershed with adjacent flat valleys, covered by evergreen tropical
398 rainforest ideal for the babblers under study (Bird et al. 2005; Cannon et al. 2009). On the
399 eastern side, a large paleo-river, the North Sunda River, originating in the Central Sumatran
400 mountains and debouching in the South China Sea, separated Natuna from Bornean land
401 extensions further south (Bird et al. 2005; Voris 2000) (Figs. 4 and 5). Natuna is situated
402 close to the former delta of the North Sunda River, which is the longest exclusively Sundaic
403 river during times of land emergence (Voris 2000). With its length of almost 2,000km
404 forming a vast tropical rainforest watershed, it would have featured a wide lower course and
405 delta, and would have been equivalent in impact to some Amazonian tributaries of similar
406 length, e.g. the Xingu River, that are also known to constitute important population-genetic
407 barriers (da Costa et al. 2016; Isler and Maldonado-Coelho 2017). Paleo-rivers, which have

408 hitherto been afforded little importance in accounting for population subdivisions in
409 Sundaland, offer a compelling explanation for Natuna's deeper genetic rift from Borneo:
410 Natuna's placement just north-west of the delta of the North Sunda Paleo-River, the largest
411 tropical Asian river at the time, would have prevented small and poorly dispersive forest
412 inhabitants from easily crossing over towards Borneo even at times when extensive land
413 connections existed.

414

415 Our data indicate that Sundaic paleo-rivers other than the North Sunda River may also have
416 had an important imprint on population structure. North of present-day Natuna and Anambas,
417 the Siam Paleo-River system constituted an extension of the present-day Chao Phraya, the
418 most dominant river in the Thai plains (Figs. 4 and 5), extending the length of the latter 2–3
419 fold during periods of land emergence (Voris 2000). North of the Siam river, there would
420 have been extensive areas of lowland rainforest that are largely submerged now (Cannon et
421 al. 2009), but have historically survived at the southernmost tip of Vietnam, where a number
422 of Sundaic vertebrates have their northernmost isolated outposts (for examples see Robson
423 2005). *P. malaccense* and *P. capistratum* no longer occur in this Sundaic outpost but may
424 have survived in this area into the present interglacial and gone extinct as a result of the
425 historic destruction of all rainforests here. An ancestral allelic contribution of 10% in the
426 Anambas population of *P. malaccense* from an unsampled, possibly extinct population may
427 well relate to gene flow across the Siam Paleo-River from a diverged northern population that
428 still existed there at the time (Figs. 4 and 5). By the same token, a similar ancestral allelic
429 contribution of 9-16% into the Natuna population of *P. capistratum* may reflect gene flow
430 from a northern, now-extinct stronghold across the Siam Paleo-River into populations that are
431 now stranded on Natuna.

432

433 *Importance of museum collections*

434 Our study demonstrates the timeless importance of historic specimen collections for
435 evolutionary research. In modern times when DNA collection is becoming ever more
436 restrictive, one of the best paths to comprehensive genomic sampling is through historic
437 museum collections. Ancient DNA is prone to degradation and damage through an excess in
438 C to T and G to A substitutions, and the implementation of multiple safeguards is necessary
439 to avoid bias in DNA sequence generation from historic specimens. Our study highlights the
440 utility of target enrichment methods to isolate homologous genomic regions across multiple
441 degraded historic museum samples, and to harvest the DNA signal of hundreds of genomic

442 markers which can capture both phylogenomic and population genomic information. Our
443 study was based on specimens containing highly degraded DNA from the Lee Kong Chian
444 Museum (formerly the Raffles Museum) in Singapore, held at tropical temperatures for many
445 decades before air-conditioning was introduced in the ~1980s. Our successful retrieval of
446 thousands of genome-wide SNPs through the rigorous application of ancient DNA protocols
447 to reduce artifacts due to excess damage demonstrates that even degraded historic museum
448 material can serve an important purpose in molecular research. We hope this approach will be
449 applied to numerous additional organismic groups across the tropics to solve evolutionary
450 problems.

451

452 **METHODS**

453

454 **Plumage analysis**

455 We examined the plumages of 41 museum specimens of *P. capistratum* and 45 museum
456 specimens of *P. malaccense* (Supplementary Table 1) deposited at the Lee Kong Chian Natural
457 History Museum (Singapore). Specifically, we laid out specimen series arranged by geographic
458 area and compared plumage hues for wings, tail, upperparts, underparts and head. We checked
459 label information for the color of beak, legs and iris on live birds.

460

461 **DNA extraction**

462 DNA extractions of historical samples were performed in a separate dedicated ancient DNA
463 facility within a biosafety cabinet. We used DNeasy Blood and Tissue Kits (QIAGEN,
464 Germany) with modifications to extract highly degraded DNA (see Chattopadhyay et al.
465 (2019) for details). We used one or two toepads per sample for DNA extraction. Prior to
466 DNA extraction the toepads were washed two to three times with molecular grade water to
467 remove any PCR inhibitors. We used 360µl of ATL buffer and approximately 100µl of
468 Proteinase K per sample to digest the tissue. The toepads generally required approximately
469 three to five days to completely digest. The volumes of AL buffer and ethanol were adjusted
470 according to the total volume of ATL buffer and Proteinase K. We used Minelute columns
471 (QIAGEN, Germany) for DNA extraction instead of the regular columns provided by the
472 manufacturer as these can help elute single stranded DNA as well as small DNA fragments.
473 As historical samples are highly degraded, these columns were helpful in isolating poor
474 quality DNA. DNA extracted from historical samples was then quantified using an AATI

475 Fragment Analyzer as well as high sensitivity Qubit Assay kits (Invitrogen, USA). For all
476 DNA extractions we carried through a negative control to detect possible contamination. For
477 fresh samples, we extracted DNA from blood using DNeasy Blood and Tissue Kits following
478 the manufacturer's instructions and quantified DNA using high sensitivity Qubit Assay kits.

479

480 **Library preparation**

481 We prepared whole genome libraries using NEB II Ultrakits (New England BioLabs, USA).
482 For ancient samples we included an additional step of DNA repair using the FFPE DNA
483 repair kit (New England BioLabs, USA) prior to library preparation. Historical samples are
484 prone to DNA damage and this step reduced the levels of damage observed in later steps
485 (Chattopadhyay et al. 2019). We carried through a dedicated negative control for each batch
486 of library preparation. For fresh samples we used a bioruptor pico (Diagenode, Belgium) to
487 fragment the DNA, performing 13 cycles of 30s ON and 30s OFF to obtain DNA fragments
488 of ~250 bp. Samples were then subjected to library preparation using the NEB Ultra II kit.
489 All samples were dual indexed using 8bp barcodes.

490

491 **Design of target loci**

492 Target enrichment protocols have been shown to be highly effective for ancient DNA
493 samples (Chattopadhyay et al. 2019). We designed target loci for sequence capture protocols
494 that are useful for both population genomic and phylogenomic studies targeting both
495 conserved exons and variable intronic regions. We used EvolMarkers (Lim et al. 2012) to
496 identify conserved single copy coding sequences in the striped tit-babbler genome (*Mixornis*
497 *gularis*, QVAJ00000000.1), collared flycatcher genome (*Ficedula albicollis*,
498 GCA_000247815.1) and zebra finch genome (*Taeniopygia guttata*, GCF_003957565.1). The
499 striped tit-babbler genome is the phylogenetically closest genome available for both target
500 species. To identify conserved exons, EvolMarkers performs a BLAST search, for which we
501 set a minimum of 55% identity and e-value of less than 10E-15. Only exons longer than
502 500bp were used for downstream analysis. We identified a total of 1,161 exons. Then we
503 isolated 500 bp upstream and downstream of these conserved exons from the striped tit-
504 babbler genome to recover variable intronic regions using bedtools 2.28.0 (Quinlan and Hall
505 2010). We further checked for overlapping targets and merged all overlapping loci in
506 bedtools. We then removed any loci with a GC content less than 40% or more than 60%. Loci
507 which contained repeat elements were identified using RepeatMasker 4.0.7 (Smit et al. 2015)
508 and removed. We finally retained 960 loci (1.99Mb), which were used by MYcroarray (USA)

509 to design RNA baits. We used 73,928 100bp baits with 4X tiling density for in-solution target
510 enrichment.

511

512 **Target enrichment**

513 We performed in-solution hybrid capture for whole genome libraries to enrich target loci. A
514 modified version of the myBaits (USA) protocol was used for hybridization (myBaits manual
515 version 3). For ancient samples, we diluted the baits and used them at 50% strength, carrying
516 out independent hybridization reactions at 60 °C for 40 hours. For fresh samples, we pooled
517 three uniquely barcoded samples at equimolar concentrations and carried out hybridization at
518 65 °C for 20 hours. We used a lower temperature and longer duration for the hybridization of
519 ancient samples as suggested by the MYBaits manual. Following hybridization, the samples
520 were cleaned as suggested in the MYBaits manual and we performed PCR for the enriched
521 libraries using IS5 and IS6 primers (Kircher et al. 2012). The final libraries were cleaned
522 using Ampure beads and pooled at equimolar concentrations. We sequenced the enriched
523 libraries on multiple lanes of HiSeq 4000 (150bp paired end runs). Fresh and ancient samples
524 were run on separate lanes. All negative controls were also sequenced to rule out
525 contamination.

526

527 **Data filtering and cleanup**

528 We obtained 1.4 billion 150bp paired-end reads. Reads with a PHREAD score below 10 were
529 removed by the service provider. We ran FASTQC 0.11.7 (Andrews 2010) to check for
530 adapter contamination. We used cutadapter 1.12 (Martin 2011) to trim adapters from the
531 reads and performed another quality check in FastQC. Subsequently, we used Trimmomatic
532 0.38 (Bolger et al. 2014) to remove any remaining adapter sequences. Trimmomatic also
533 removes any low-quality reads and reads less than 36bp in length. Finally, we performed a
534 third FastQC run to check for sequence quality and confirm complete adapter removal. Next,
535 we removed PCR duplicates using the dedupe program within bmap 36.84 (Bushnell 2014).
536 Furthermore, for the ancient samples we used mapDamage 2.0 (Jónsson et al. 2013) to
537 remove preservation-related post-mortem substitutions. We observed high rates of transitions
538 (G to A and C to T) at the ends of reads. Hence, we first rescaled our bam files using
539 mapDamage. Then the rescaled bam files were converted to fastq reads and we trimmed 10bp
540 from both the 5' and 3' ends in the historic samples using Seqtk 1.2-r94
541 (<https://github.com/lh3/seqtk>) to reduce bias due to DNA degradation. The cleaned reads
542 were used for downstream processing.

543

544 Data matrix generation

545 We used two different approaches to generate data matrices for each of the two babbler
546 species complexes. In the first approach, we generated sequence data matrices for target loci.
547 In the second approach, we generated genome-wide SNPs.

548

549 A. Sequence data generation

550 To generate a sequence data matrix of target loci, we used the HybPiper 1.2 pipeline
551 (Johnson et al. 2016), which is specifically designed for sequence capture protocols. We
552 generated sequence data with an outgroup individual for phylogenomic analysis. The
553 resultant sequences were then aligned using MAFFT v7.130b (Kato and Standley 2016). We
554 used auto settings within MAFFT to identify the best approach for sequence alignment. The
555 alignments were further cleaned using strict settings (default settings) within Gblocks 0.91b
556 (Castresana 2000) to remove poorly aligned regions. We further removed loci with a
557 sequence length less than 200bp and those located on the Z chromosome for phylogenomic
558 analysis.

559

560 We obtained sequence data for 944 out of 960 target loci designed for both species
561 complexes using the HybPiper pipeline. We removed ten historic *P. malaccense* samples
562 from downstream processing as their missing data exceeded 85%. After multiple sequence
563 alignments, we performed stringent filtering of alignments using Gblocks and removed 117
564 loci from the *P. capistratum* dataset and 385 loci from the *P. malaccense* dataset due to high
565 missing data. After removal of loci < 200bp and Z-chromosomal loci, we retained 652 loci
566 for *P. capistratum* and 314 loci for *P. malaccense*. The total sequence matrix length for *P.*
567 *capistratum* was 454,712 bp (average locus length = 697 bp; minimum = 201 bp; maximum =
568 4,723bp) and 145,877 bp for *P. malaccense* (average locus length = 465 bp; minimum = 201
569 bp; maximum = 2,607 bp).

570

571 B. SNP data generation

572 We generated four different SNP sets for each of the two babbler species complexes. For the
573 first three SNP sets, we mapped the clean reads to the striped tit-babbler genome using
574 BWA-MEM 0.7.17-r1188 (Li and Durbin 2009). The mapped reads were sorted and
575 converted to a bam format using SAMTOOLS 1.9 (Li et al. 2009; Li 2011). We used the
576 GATK SNP caller within ANGSD 0.923-3-ga8ed56f (Korneliussen et al. 2014) to identify

577 SNPs for each species complex separately. For both species complexes, we used a p-value
578 cut-off of $1E-6$. Further, only SNPs with a PHREAD score ≥ 30 (99.9% accuracy) were
579 retained. We allowed for no missing data during SNP calling. The SNPs were further filtered
580 using VCFtools 0.1.16 (Danecek et al. 2011), and any locus with a read depth < 10 per sample
581 was removed. We then removed linked loci using PLINK 1.9 (Purcell et al. 2007) by
582 applying the indep-pairwise algorithm with a sliding window size of 50 SNPs, a step size of
583 10 and an r^2 correlation coefficient cut-off of 0.9. We also removed any loci not in Hardy-
584 Weinberg equilibrium using PLINK while correcting the p-value for multiple comparisons.
585 We further tested for selection in BAYESCAN 2.1 (Foll and Gaggiotti 2008) using default
586 settings and removed any locus under positive selection. The striped tit-babbler genome was
587 also blasted to the chromosomal assembly of the great tit genome (*Parus major*,
588 GCA_001522545.3) using blastn to identify scaffolds which map to the Z chromosome. The
589 great tit genome is the phylogenetically closest genome available that has been assembled to
590 the chromosome level. SNPs which mapped to the Z chromosome of the great tit genome
591 were removed from downstream processing. The cleaned, unlinked, neutral autosomal SNPs
592 constituted dataset I. C to T and G to A substitutions are the most common post-mortem
593 substitutions observed in historic samples and hence can bias results (Chattopadhyay et al.
594 2019). To account for DNA damage due to historic sampling, we therefore generated a
595 second dataset in which only transversions were included in analyses (dataset II). We also
596 generated a third dataset which included only a single random SNP from each target locus.
597 We did not perform any test for linkage for this dataset (dataset III).

598
599 For the final dataset (dataset IV) we mapped clean reads to the great tit genome and carried
600 out SNP calling in ANGSD as mentioned above. We included an outgroup for this SNP set
601 and allowed for 20% missing data. We further pruned the dataset for linkage disequilibrium
602 and deviations from neutrality and Hardy-Weinberg equilibrium using the same approaches
603 as mentioned above. We only retained transversions for dataset IV. This latter dataset had the
604 advantage of chromosomal information required for admixture graph analyses (see section on
605 ABBA-BABA). For the population structure analysis using dataset IV, we further pruned the
606 loci located on the Z chromosome.

607

608 In the end, we retrieved between 960 and 208,186 SNPs for *P. capistratum* and retained 883
609 to 54,906 SNPs after filtering (Table 1). For *P. malaccense* between 958 and 198,711 SNPs

610 were obtained across datasets before filtering and we retained 895 to 33,708 SNPs for further
611 analysis (Table 1).

612

613 **Phylogenomic analysis**

614 We used both concatenation approaches and species tree reconstruction for phylogenomic
615 analysis. For concatenation-based analyses, we used RAxML 8.2 (Stamatakis 2014) to
616 reconstruct a maximum likelihood tree using the GTR+GAMMA model of substitution. We
617 ran the AMAS pipeline (Borowiec 2016) to concatenate loci. For species tree reconstruction,
618 we used the phyluce pipeline 1.6.6 (Faircloth 2016) to process the sequence data. We
619 generated individual maximum likelihood gene trees using RAxML within the phyluce
620 pipeline for downstream use in MP-EST 1.6 (Liu and Edwards 2010) to estimate the species
621 tree. We followed Liu et al. (2017) for species tree estimation and generated 100 bootstrap
622 trees per locus. We further generated 100 bootstrap files containing one bootstrap replicate
623 per locus. These files were then used for MP-EST species tree estimation. The 100 bootstrap
624 species trees were further used to estimate nodal support and to generate a majority-rule
625 consensus tree using Phylip v3.69 (Felsenstein 2005). Both the concatenated tree and species
626 tree were viewed in FigTree v1.4.2 (<http://tree.bio.ed.ac.uk/software/figtree/>).

627

628 **Population genomic analysis**

629 We employed multiple approaches to understand population subdivision within each babbler
630 species complex. First, we performed PCA using the SNPrelate package in R (Zheng et al.
631 2012). PCA is a multi-variate approach to identify structure within datasets, but is at the same
632 time independent of population genetic assumptions. For dataset IV we removed the loci with
633 missing data for PCA analysis. Second, we performed DAPC in the adegenet 2.1.1 R package
634 (Jombart 2008). One major feature that sets DAPC apart from PCA is that it tends to
635 maximize between-group differences while placing less emphasis on within-group
636 variability⁷³. Finally, we adopted a Bayesian clustering approach using STRUCTURE
637 (Pritchard 2000) to identify subdivision within each species complex. We used the structure
638 threader program for STRUCTURE analysis and performed ten runs each for K=1 to 10. For
639 each K, we performed 100,000 burn-ins and 500,000 Markov chain Monte Carlo steps.
640 Structure runs were performed without any *a priori* assumptions of population assignment.
641 We ran the pophelper R package (Francis 2017) to visualize STRUCTURE results and
642 compared across multiple K values to assess genomic assignments of individuals. We used R
643 version 3.5.3 (R Core Team 2019) for all analyses.

644

645 **ABBA-BABA tests and admixture graph analysis**

646 We used the four-taxon ABBA-BABA test to understand gene flow among various island
 647 populations of babblers (Durand et al. 2011; Green et al. 2010). The ABBA-BABA test is a
 648 powerful method to differentiate between secondary admixture and incomplete lineage
 649 sorting (Durand et al. 2011; Green et al. 2010). We performed ABBA-BABA tests using
 650 ANGSD considering only transversions to account for DNA damage. Only sites with a
 651 mapping quality and PHRED score ≥ 30 were considered. To test for significance, we
 652 performed jackknifing of 20kb blocks. Only test scenarios congruent with the phylogenomic
 653 analysis were considered. Significant secondary admixture was inferred when Z scores
 654 exceeded -3 or +3. For ABBA-BABA analyses, we used the bam files obtained by mapping
 655 clean reads to the chromosomal assembly of the great tit genome.

656

657 In addition to ABBA-BABA tests, we performed admixture graph analysis with qpbrute
 658 (Leathlobhair et al. 2018; Liu et al. 2019), which implements a heuristic algorithm to
 659 iteratively fit complex admixture models using qpGraph (part of ADMIXTOOLS 5.1
 660 (Patterson et al. 2012)) as well as estimate Bayes factors to determine the best admixture
 661 graph: qpGraph reconstructs relationships among populations using a phylogenetic approach
 662 allowing for admixture events. For a given topology, f_2 , f_3 and f_4 statistics are estimated for
 663 all taxa and the observed and expected allele frequencies are calculated for the observed data
 664 and model. Within qpbrute, for a given outgroup, taxa are added iteratively. If a node cannot
 665 initially be included and outliers are observed for the f_4 statistic, then all possible admixture
 666 events are attempted. If a node ends up being excluded, its subgraph is discarded, but if a
 667 node is successfully included, remaining nodes are added iteratively to the subgraph. All
 668 possible combinations of taxa are tested within qpbrute to ensure a complete coverage of
 669 graph space. We used dataset IV along with an outgroup and also included SNPs mapping to
 670 the Z chromosome for this analysis.

671

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981

982 **ETHICS STATEMENT**

983 This study complied with all ethical regulations. Protocols were approved by the National
984 University of Singapore Institutional Animal Care and Use Committee (IACUC, Protocol
985 Number: L2017–00459). Permits for sampling in Sabah were approved by the Danum Valley
986 Management Committee, the Sabah Forestry Department, and the Sabah Biodiversity
987 Council (permit numbers: JKM/MBS.1000-2/2 JLD.3 (118); JHL 100.7/27 and export
988 samples [permit number: JKM/MBS.1000-2/3 JLD.2 (65)].

989

990 **AUTHOR CONTRIBUTIONS**

991 KMG and FER designed the research. KMG performed laboratory work with input from BC.
992 KMG and BC designed the target enrichment loci with input from FER. EC performed
993 toepad sampling and plumage analysis. EC conducted fieldwork with the help of SB, ST and
994 DPE. KMG and FER wrote the paper with input from all co-authors.

995

996 **COMPETING INTERESTS**

997 The authors declare no competing interests.

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999 **DATA AVAILABILITY**

1000 The data underlying this article are available in the article and in its online supplementary
1001 material. Raw data generated in this study have been submitted to the NCBI SRA database
1002 (BioProject ID: PRJNA701111).

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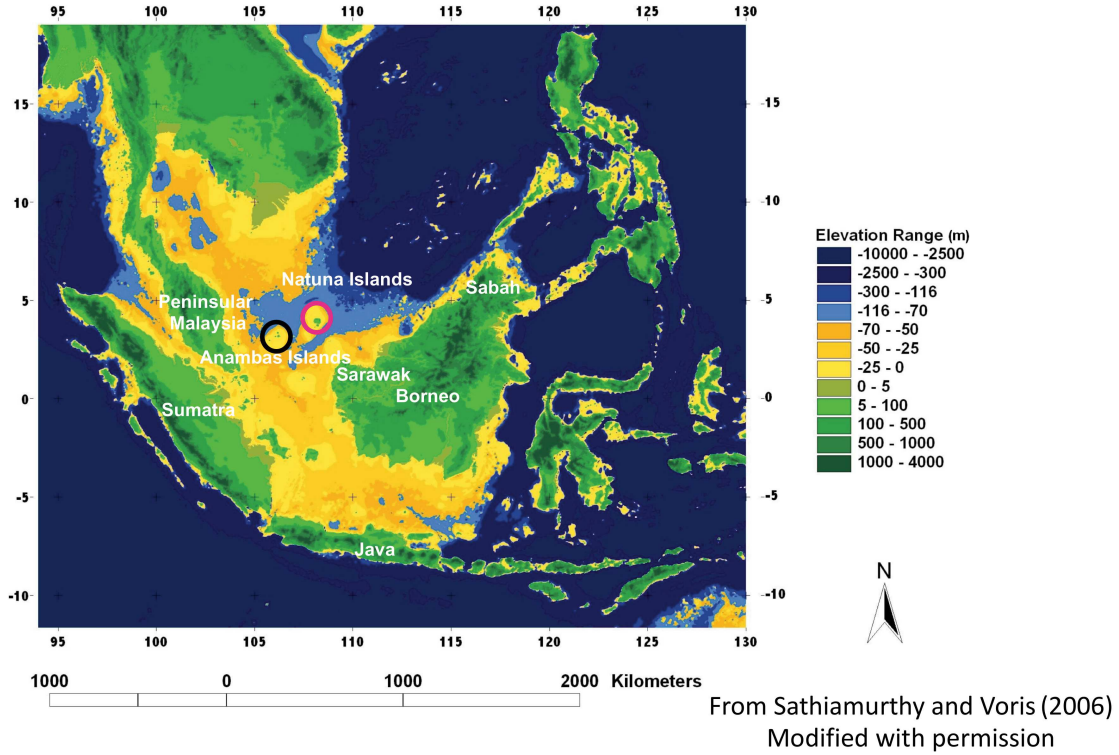
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1012 FIGURES

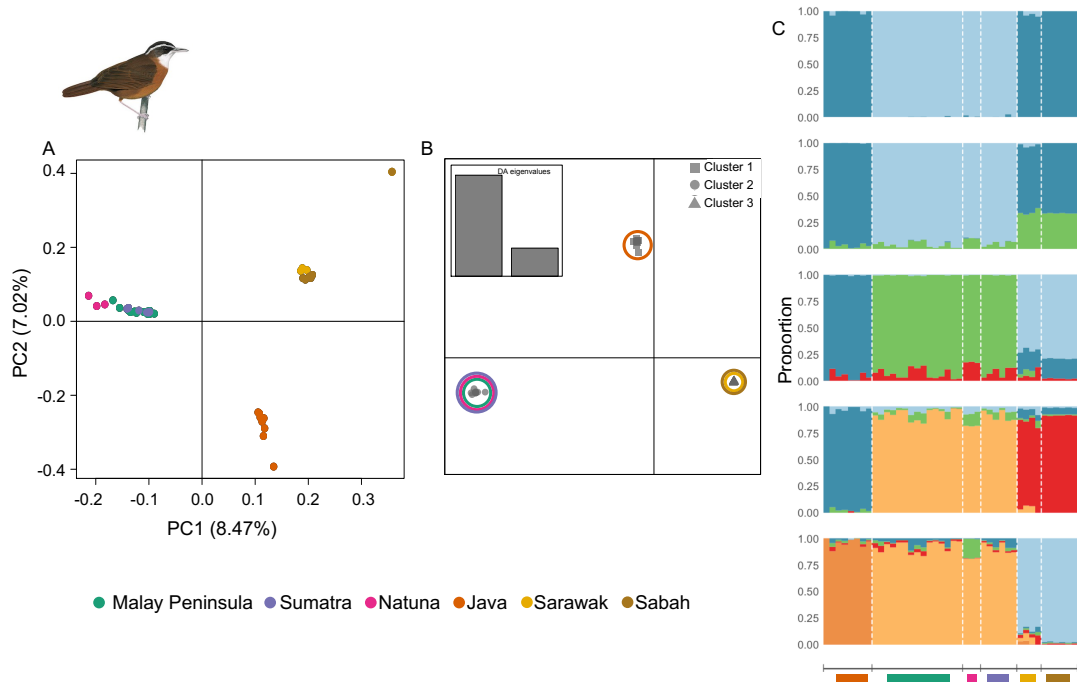
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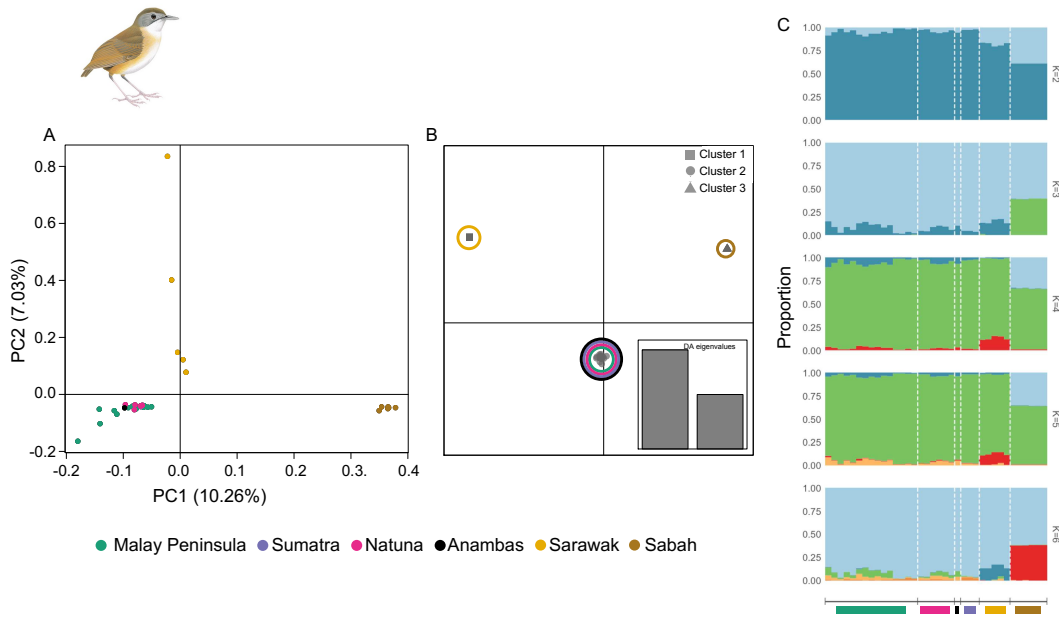
1015 Fig. 1: Map showing Sundaland and sea depths across the Sunda Shelf. Note that land would
1016 have extended approximately to the -116m isobath (border between light-blue and mid-blue)
1017 at the peak of glacial episodes.

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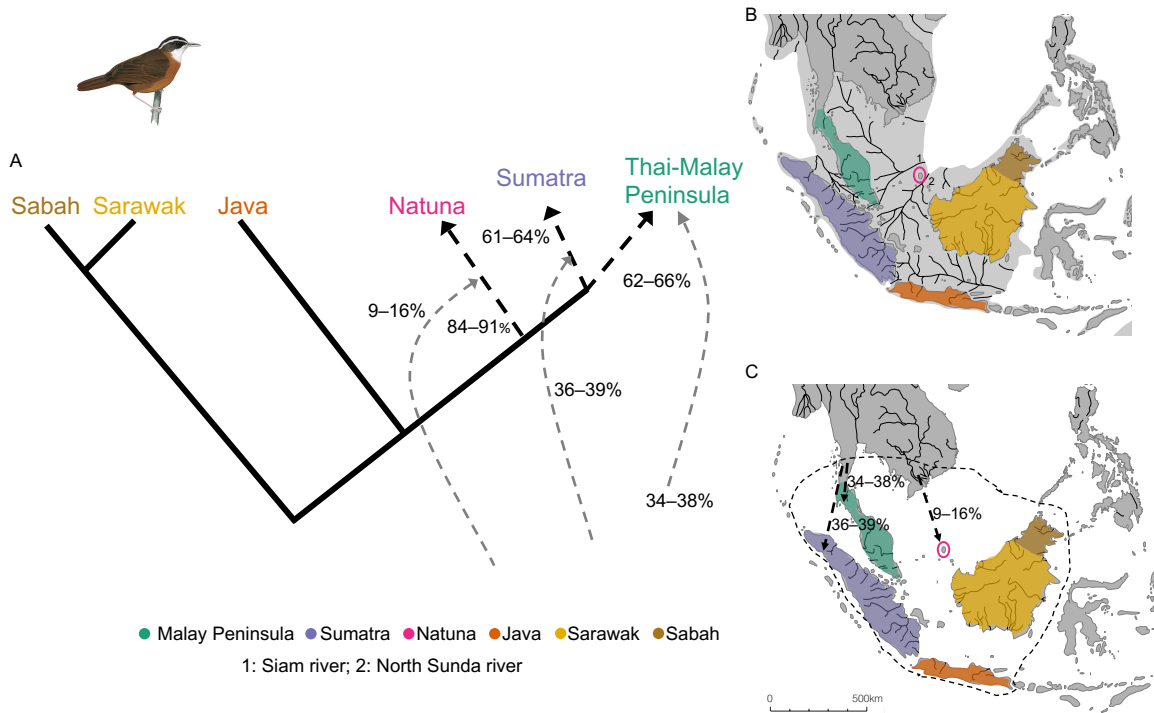
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Fig. 2: Population subdivision observed in *Pellorneum capistratum* based on transversions isolated after mapping to the *Parus major* genome. Three population clusters observed based on (A) principal component (PC) analysis (number of transversions = 10,848), (B) discriminant analysis (DA) of principal components, in which clusters are color coded based on population affinity (number of transversions = 38,463), (C) genetic assignment of *P. capistratum* individuals using STRUCTURE for K=2 to K=6 (number of transversions = 38,463).



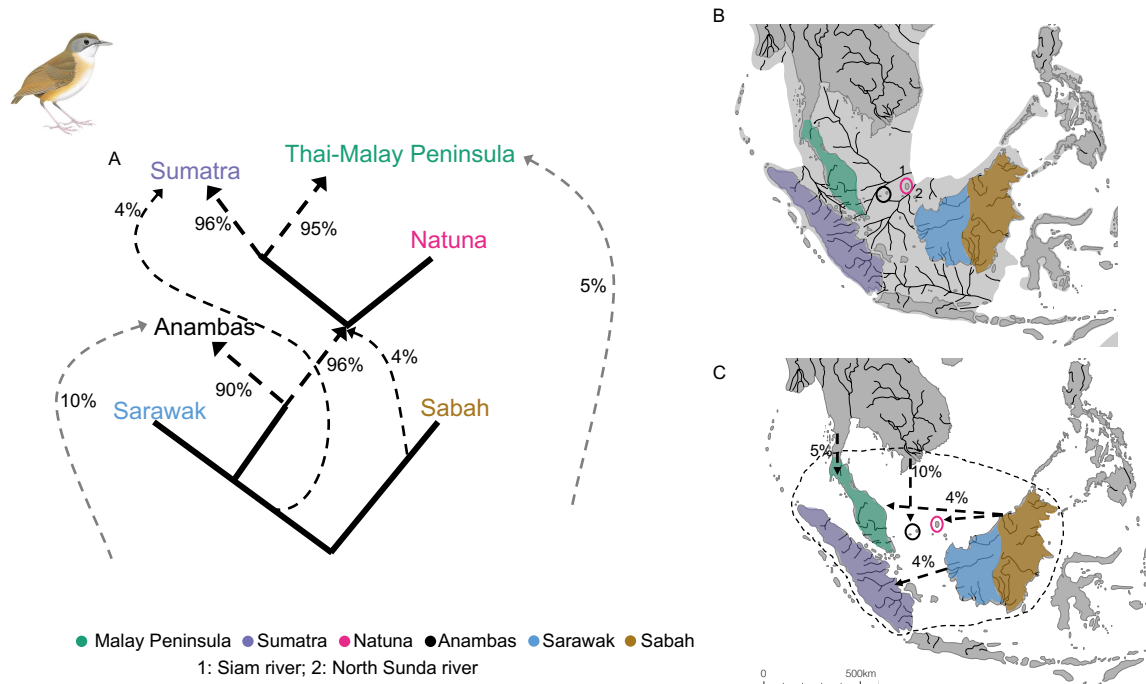
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Fig. 3: Population subdivision observed in *Pellorneum malaccense* based on transversions isolated after mapping to the *Parus major* genome. Three population clusters observed based on (A) principal component (PC) analysis (number of transversions = 6,091), (B) discriminant analysis (DA) of principal components, in which clusters are color coded based on population affinity (number of transversions = 32,735), (C) genetic assignment of *P. malaccense* individuals using STRUCTURE for K=2 to K=6 (number of transversions = 32,735).



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Fig. 4: Admixture graph of gene flow dynamics observed in *Pellorneum capistratum*. (A) Schematic of genetic ancestry and admixture proportions among populations of *P. capistratum* based on a combination of five best-fit admixture graphs, all of which exhibited an identical topology and only differed slightly in estimates of the extent of admixture from an unknown source for the western Sundaic population; (B) map during sea level recession of ~120m at glacial maximum around 20,000 years before present, showing major paleo-rivers and present-day taxon distributions; (C) current map of South East Asia with taxon distributions, major rivers, and arrows indicating potential gene flow from unsampled or extinct populations (see panel A). The black stippled line indicates potential habitat during sea level lows. Panels B and C modified with permission from Voris (2000).



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Fig. 5: Admixture graph of gene flow dynamics observed in *Pellorneum malaccense*. (A) Schematic of genetic ancestry and admixture proportions among populations of *P. malaccense* based on single best-fit admixture graph; (B) map during sea level recession of ~120m at glacial maximum around 20,000 years before present, showing major paleo-rivers and present-day taxon distributions; (C) current map of South East Asia with taxon distributions, major rivers, and arrows indicating potential gene flow among populations or from unsampled or extinct populations (see panel A). The black stippled line indicates potential habitat during sea level lows. Panels B and C modified with permission from Voris (2000).

1059 **TABLES**

1060

1061 Table 1: Number of SNPs identified and filtered for different datasets. Abbreviations: HWE – Hardy-Weinberg equilibrium; NA – not
 1062 applicable.

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Dataset	Number of unfiltered SNPs	Number of SNPs removed due to linkage	Number of SNPs removed that were not in HWE	Number of non-neutral SNPs removed	Number of SNPs removed mapping to Z chromosome	Number of transitions removed	Number of SNPs retained	Number of SNPs with no missing data
<i>Pelloroneum capistratum</i>								
Dataset I: All SNPs obtained after mapping to <i>Mixornis gularis</i> genome	82,468	24,025	1,904	82	1,246	NA	54,906	54,906
Dataset II: Only transversions obtained after mapping to	82,468	24,025	1,904	82	1,246	37,690	17,216	17,216

<i>Mixornis gularis</i> genome								
Dataset III: single random SNP per target locus	960	NA	30	12	35	NA	883	883
Dataset IV: Only transversions obtained after mapping to <i>Parus major</i> genome	208,186	67,451	3,357	1,130	1,470	96,315	38,463	10,848
<i>Pellorneum malaccense</i>								
Dataset I: All SNPs obtained after mapping to <i>Mixornis gularis</i> genome	57,300	21,595	440	240	1,317	NA	33,708	33,708
Dataset II: Only transversions obtained after mapping to	57,300	21,595	440	240	1,317	22,558	11,150	11,150

<i>Mixornis gularis</i> genome								
Dataset III: single random SNP per target locus	958	NA	17	2	44	NA	895	895
Dataset IV: Only transversions obtained after mapping to <i>Parus major</i> genome	198,711	83,460	749	1,490	1,491	78,786	32,735	6,091

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