Selection and local adaptation in capuchin monkeys revealed through fluorescence-activated cell sorting of feces (fecalFACS)

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ABSTRACT

Background: Capuchins have the largest relative brain size of any monkey and a remarkable lifespan of 55 years, despite their small body size. Distributed widely across Central and South America, they are inventive and extractive foragers, known for their sensorimotor intelligence, dietary diversity, and ecological flexibility. Despite decades of research into their ecology and life history, little is known about the genomics of this radiation.

Results: We assemble a *de novo* reference genome for *Cebus imitator*, and provide the first genome annotation of a capuchin monkey. We identified 20,740 and 9,556 for protein-coding and non-coding genes, and recovered 23,402 orthologous groups. Through a comparative genomics approach across a diversity of mammals, we identified genes under positive selection associated with longevity and brain development, which are of particular relevance to capuchin and primate comparative biology. Additionally, we compared populations in distinct habitats, facilitated by our novel method for minimally-biased, whole-genome sequencing from fecal DNA using fluorescence activated cell sorting (FACS). By analyzing 23 capuchin genomes from tropical dry forest and rainforest, we identified population divergence in genes involved in water balance, kidney function, and metabolism, consistent with local adaptation to resource seasonality.

Conclusions: Our comparative study of capuchin genomics provides new insights into the molecular basis of brain evolution and longevity. These data also improve our understanding of processes of local adaptation to diverse and physiologically challenging environments. Additionally, we provide a technological advancement in use of non-invasive genomics to study free-ranging mammals through FACS.

KEYWORDS

Brain size, longevity, intelligence, seasonality, dry forest, non-invasive genomics, flow cytometry, platyrrhine

BACKGROUND

Large brains, long lifespans, extended juvenescence, tool use, and problem solving are hallmark characteristics of the great apes, and are of enduring interest in studies of human evolution [1–6]. Similar suites of traits have arisen in other lineages, including some cetaceans, corvids and, independently, in another radiation of primates, the capuchin monkeys. Like great apes, they have diverse diets, consume and seek out high-energy resources, and engage in complex extractive foraging techniques [7,8] to consume difficult-to-access invertebrates and nuts [8]. Their propensity for tool use and their ecological flexibility may have contributed to their convergence with the great apes [9]offering opportunities for understanding their evolution via comparative methods [10–12].

Capuchins also offer excellent opportunities to study local adaptation to diverse habitats. While little is known about the migration, history, and population dynamics of this species, they presently range from Panama to northern Honduras [13–15], where they occupy a wide diversity of habitats, including rainforests and, in the northern extent of their range, tropical dry forests. Particular challenges of the tropical dry forest are staying hydrated during the seasonally prominent droughts, high temperatures in the absence of foliage, and coping metabolically with periods of fruit dearth (Figure 1). The sensory challenges of food search in dry versus humid biomes are also distinct. For example, odor detection and propagation is affected by temperature and humidity [16,17], and color vision is hypothesized to be adaptive in the search for ripe fruits and young reddish leaves against a background of thick, mature foliage [18], which

98 is absent for long stretches in dry deciduous forests. The behavioral plasticity of capuchins is
99 widely acknowledged as a source of their ability to adapt to these dramatically different habitats
100 [19–22]. However, physiological processes including water balance and metabolic adaptations
101 to low caloric intake, and sensory adaptations to food search, are also anticipated to be targets
102 of natural selection, as seen in other mammals [23–26].
103 Comparative genomics offers a unique opportunity to examine the molecular
104 underpinnings of traits relevant to the evolution of humans and other great apes. Furthermore,
105 the ecological flexibility of white-faced capuchins allows us to assess the influence of divergent
106 and changing habitats on the process of adaptive evolution in these primates. In order to
107 address the genetic underpinnings of capuchin adaptive evolution, we assembled the first
108 reference genome of *Cebus imitator*. Additionally, to better understand the local adaptation of
109 capuchins, we conducted high-coverage re-sequencing (10X - 47X) of 10 individuals from
110 populations inhabiting distinct environments--one in a lowland evergreen rainforest (n=4), and
111 another from a lowland tropical dry forest (n=6). Importantly, to facilitate the population-wide
112 analyses without the need for potentially harmful invasive sampling of wild primates, we
113 sequenced an additional 13 individuals at low coverage using a novel method for minimally-
114 biased, whole-genome sequencing from fecal DNA using fluorescence-activated cell sorting
115 (fecalFACS) that we developed (Figure 2). In our positive selection analyses, we focus on
116 genes that may underlie sensation, cognition, and lifespan due to their relevance to capuchin-
117 specific biology and adaptation. In our population comparison we predict genes related to water
118 balance, metabolism will differ between dry forest and rainforest populations, reflecting local
119 adaptation to different habitats. With respect to sensory systems, due to higher diversity of flora
120 and fauna, including colorful fruits presented against green foliage, and increased humidity,
121 genes underlying color vision (opsin genes) and chemosensation (olfactory and taste receptor
122 genes) are predicted to be more diverse in rainforest habitats [27–29].
123

124 **RESULTS**

125

126 **1. Comparative Genomics**

127

128 ***Genome assembly and gene annotation***

129

130 Our reference genome assembly for *Cebus imitator* is comprised of 7,742 scaffolds
131 (including single contig scaffolds) with an N50 scaffold length of 5.2 Mb and an N50 contig
132 length of 41 kb. The final ungapped assembly length is 2.6 Gb (GenBank accession:
133 GCA_001604975.1). Our estimate of total interspersed repeats using WindowMasker [30]
134 output is 45.8%. The numbers of annotated genes are 20,740 and 9,556 for protein-coding and
135 non-coding genes, respectively (Table S1). Measures of gene representation using the known
136 human RefSeq set of 56,230 transcripts show an average of >94% coverage with a mean
137 identity of 92.5%. Overall, our draft assembly metrics and gene representation are consistent
138 with other non-human primate (NHP) short-read reference assemblies [31].

139

140 ***Positive selection analyses***

141

142 We recovered 23,402 Orthologous Groups (OGs). Capuchins share 18,475 OGs with
143 human, 17,589 OGs with rhesus macaque, 15,582 OGs with mouse, and 14,404 OGs with dog.
144 When we included orthologous genes that are present simultaneously in all 15 species, we
145 recovered 7,519 OGs, which we subsequently used in the natural selection analyses ($d_N/d_S=\omega$).
146 We identified 612 genes under positive selection ($p<0.05$ after FDR correction) in the *Cebus*
147 lineage using the branch model. We also performed a branch-site test using codeml in
148 PAML[32] and identified a second set of 748 genes under positive selection in *Cebus* (Table S2
149 Sheet 2). The results of our enrichment analysis for biological processes using DAVID [33]

150 identified genes that underlie brain development, sensation, and lifespan, which were of
151 particular interest given the derived features of capuchin biology (Figure 3).
152
153 *Brain development and longevity*
154
155 Capuchins have the largest brain to body ratio of any monkey, and are known for their
156 sensorimotor intelligence [7] and derived cognitive abilities [8]. Of the 748 genes identified as
157 being under positive selection in the branch-site model, 17 were previously associated with
158 brain development (Table S2, Sheet 6, row 18) and 5 were linked to neurogenesis (Table S2,
159 Sheet 6, row 116). For example, *WDR62*, *BPTF*, *BBS7*, *NUP113*, mutations are directly
160 associated with brain size and related malformations, including microcephaly [34–37]. *MTOR*
161 signaling malfunction is also implicated in developmental brain malformations [38], and *NUP113*
162 is involved in nuclear migration during mammalian brain development [39]. Several genes are
163 linked with brain tumor formation (including *ZNF217*), and others with cognitive ability (e.g.
164 *PHF8 [40]*).
165 We found 27 aging-related genes, as identified in the GenAge database [41,42], under
166 positive selection in capuchins including *PARP1*, *MTOR*, *SREBF1*, *INSR*, *HTT*, *RB1* and *MDM2*
167 (Table S2, Sheet 7). Of note, poly (ADP-ribose) polymerase 1 (PARP1) putatively serves as a
168 determinant of mammalian aging due to its activity in the recovery of cells from DNA damage. In
169 previous studies, gene expression levels of *PARP1* were inversely correlated with mammalian
170 lifespan [43]. Another large body of research has associated the mechanistic target of rapamycin
171 (*MTOR*) with aging and longevity in various organisms [44], making it a prime candidate for
172 therapeutic interventions in aging [45]. *MTOR* acts as a regulator of cell growth and proliferation
173 while also being generally involved in metabolism and other processes. Additional key genes in
174 aging and metabolism include sterol regulatory element binding transcription factor 1 (*SREBF1*),
175 which acts as a regulator of the metabolic benefits of caloric restriction [46,47], and the insulin

176 receptor (*INSR*), a major player in longevity regulation [48]. As for specific age-related diseases,
177 huntingtin (*HTT*) is under selection in mammals; *HTT* is not only involved in Huntington's disease but
178 has also been associated with longevity in mice [49]. Lastly, various cell cycle regulators (e.g., *RB1*,
179 *MDM2*) are also under positive selection in capuchins, and indeed, cell cycle is an enriched term
180 among positively selected genes (Table S2), though these could be related to other life history traits
181 like developmental schedules that correlate with longevity.
182
183 **2. <u>Population Genomics and Local Adaptation</u>**
184
185 ***Population structure, genetic diversity, and demographic history in Costa Rican white-***
186 ***faced capuchins***
187
188 Of the 24 capuchin DNA samples that we sequenced and mapped to our reference
189 genome, 15 were fecal-derived (Table S3). When comparing the high coverage tissue-derived
190 genomes from the Santa Rosa site to those generated from our novel application of
191 fluorescence-activated cell sorting to isolate fecal-sourced cells (fecalFACS), we observed no
192 substantial difference in quality, coverage, heterozygosity, or GC content (Figures 2, S1, S2,
193 Supplemental Text). This includes the first (to our knowledge) high-coverage (12.2 X) whole
194 mammalian genome generated from a fecal sample.
195 The pattern of clustering in our maximum likelihood single nucleotide variant (SNV) tree
196 recapitulates the expected patterns of geographic distance and ecological separation in our
197 samples (Figure 4). Likewise, in the projected PCA all individuals from the seasonal dry forests
198 in the northwest are sharply discriminated from individuals inhabiting the southern rainforests
199 along PC1. These relationships are not perturbed by depth of coverage, or source material
200 (tissue-based vs fecalFACS genomic libraries) (Figure 4). Levels of heterozygosity calculated in
201 overlapping 1 Mb genomic windows (with a step size of 100 kb) were significantly higher in the

202 southern population (W = 1,535,400,000, p-value < 2.2e-16; Figure 5A, Figure S3).
203 Furthermore, the median pairwise heterozygosity for each southern individual (range: 0.00065 -
204 0.00071) was higher than any northern monkey (0.00047 - 0.00057) (W = 0, p-value =
205 0.009524; Table S5). In the northern population, we also identified long runs of homozygosity
206 significantly more often (W = 24, p-value = 0.009524), and more of the longest runs (>= 5 Mb)
207 (W = 1315.5, p-value = 0.03053; Figures 5B, S4, S5). Pairwise sequential Markovian coalescent
208 (PSMC) analysis of demographic history (Figure 5C) reveals that white-faced capuchins had a
209 peak effective population size of ~60,000 effective individuals ~1 mya, which declined to fewer
210 than 20,000 during the middle to late Pleistocene transition. After recovering during the middle
211 Pleistocene, they declined precipitously through the late Pleistocene to fewer than 5,000
212 effective individuals.
213
214 ***Local adaptation to seasonal dry forest biome***
215
216 We predicted that genes related to water balance, metabolism, and sensation would
217 differ between dry forest and rainforest populations, reflecting local adaptation to different
218 habitats. To test this, we searched for associations between genes in windows with high F_{ST} and
219 divergent non-synonymous SNVs with high or moderate effect. Of the 299 genes identified in
220 high F_{ST} windows, 39 had highly differentiated non-synonymous SNVs (Figure S6, Tables S6,
221 S7). Our enrichment analysis identified a single significant GO biological process: regulation of
222 protein localization to cilium (GO: 1903564). Remarkably--and in accordance with our
223 hypothesis--disruptions of cilia proteins are predominantly associated with disorders of the
224 kidney and retina (ciliopathies) [50]. Furthermore, enrichment of genes associated with disease
225 states was linked to kidney function, metabolism, and muscular wasting. These genes are good
226 candidates for adaptive resilience to seasonal water and food shortages and warrant further
227 investigation. We highlight several genes of particular promise (Figure 6, Figure S7).

228 Evidence of adaptation to food and water scarcity

229

230 Multiple candidate genes indicate that dry-forest capuchins could be adapted to
231 seasonal drought-like conditions and food scarcity. *SERPINC1* encodes antithrombin III, which
232 is involved in anticoagulant and anti-inflammatory responses associated with numerous kidney-
233 related disorders including salt-sensitive hypertension, proteinuria, and nephrotic syndrome [51–
234 56]. *SERPINC1* is overexpressed in the renal cortex of Dahl salt-sensitive rats fed a high salt
235 diet [55], suggesting that variants could be facilitating water/salinity balance in the dry-forest
236 capuchins. Additionally, sequence variants of *AXDND1* (as identified in the GeneCards
237 database) are associated with nephrotic syndrome, a kidney disorder that commonly presents
238 with edema and proteinuria. *BCAS3* is expressed in multiple distal nephron cells types [57], and
239 is associated with four pleiotropic kidney functions (concentrations of serum creatinine, blood
240 urea nitrogen, uric acid, and the estimated glomerular filtration rate based on serum creatinine
241 level) [58]. Although we did not identify any population-specific non-synonymous SNVs in
242 *BCAS3*, the genomic windows encompassing the gene rank among the highest regions of F_{ST} in
243 our dataset, and intronic variation in BCAS3 putatively impacts estimated glomerular filtration
244 rate in humans [57].

245 The association of *BCAS3* and serum creatinine is of particular interest, given field
246 observations of capuchins from SSR, whose urinary creatinine levels are associated with
247 decreased muscle mass during periods of seasonally low fruit availability [59]. Creatinine is a
248 byproduct of the metabolism of creatine phosphate in skeletal muscle, which is normally filtered
249 by the kidneys, and can be used as a clinical biomarker of kidney function, chronic kidney
250 disease [58], and as a monitor of muscle mass [59,60]. Multiple congenital muscular
251 dystrophies--including mild forms that present with muscular wasting--and abnormal circulating
252 creatine kinase concentration (HP:0040081) are associated with candidate genes *ITGA7, ISPD*
253 (*CRPPA*), and SYNE2 [61–64]. Furthermore, transgenic overexpression of *ITGA7* has been

254 shown to reduce muscular pathologies caused by mutations in *LAMA2* [65], which falls in a high
255 F_{ST} window. The relationship among seasonal resource availability, kidney function, and muscle
256 mass is further supported by a potential adaptive role in capuchin sugar metabolism and
257 frugivory. In particular, *GLIS3* is one of two genes known to be associated with type 1 diabetes,
258 type 2 diabetes, and neonatal diabetes [66], and appears to be diverging between populations.
259 Additionally, while insulin receptor substrate (*IRS4*) did not fall in a high F_{ST} window, we
260 observed a non-synonymous SNP of medium effect fixed between populations. Given the
261 appearance of both diabetes and kidney disorders in our gene sets, we conducted an *a*
262 *posteriori* search of our high F_{ST} gene set for overrepresentation of genes associated with
263 diabetic nephropathy (EFO_0000401) in the GWAS catalogue (Table S6). Seven genes were
264 present in both our gene set of 299 high F_{ST} genes and the diabetic nephropathy set of 117.
265 Given the 16,553 annotated genes with HGNC Gene IDs, seven overlapping genes would occur
266 with $p = 0.00046$ when permuted 100,000 times. We take this as promising evidence that these
267 genes have been under selection in the northwestern population.
268
269 <u>Evidence of adaptation in sensory systems</u>
270
271 Given the ecological differences between the northern dry and southern wet forests, we
272 predicted that the evergreen, humid environment of the lowland rainforest would favor enhanced
273 diversity of both color vision (opsin genes) and olfaction (olfactory receptor genes) would result
274 in population specific variation in chemosensory and visual genes. We did not find support for
275 this; in both populations, we observed similar polymorphism at each of the three medium/long
276 wave cone opsin tuning sites (180, A/S; 277, F/Y; and 285, T/A) (Table S8). None of these
277 codons is a novel variant [67,68], providing no support for the hypothesis of differences in the
278 perception of photopic (cone-driven) vision between the two biomes. However, we did observe
279 some evidence for population specific variation associated with the photoreceptive layers of the

280 retina. First, creatine kinase plays an important role in providing energy to the retinal pigment
281 epithelium [69]. Secondly, we identified a fixed non-synonymous SNV in *CCDC66*, which falls in
282 a high F_{ST} region and is heavily expressed in photoreceptive layers of the retina.
283 Electroretinography of *CCDC66* -/- mice reveals a significant reduction in scotopic (rod-driven)
284 photoreceptor response [70], indicating a potential effect on vision in low-light conditions.
285 Curiously, *CCDC66* -/- mice, also display neurodegeneration of the olfactory bulb, and have
286 reduced odor discrimination performance of lemon smells [71]. Turning to olfaction, we identified
287 614 olfactory receptor (OR) genes and pseudogenes in the capuchin reference genome (408
288 intact, 45 truncated, and 161 pseudogenized (Table S9). To test for population differences in the
289 OR gene repertoire, we assembled each olfactory gene/pseudogene independently in each
290 individual. The proportion of total functional ORs was stable across individuals and populations,
291 with trivial fluctuations (North $\bar{x}$ = 411, s = 1.6; South $\bar{x}$ = 408.5, s = 1.3), possibly driven by a
292 small difference in OR family 5/8/9 (Table S10). We also identified 8 vomeronasal and 28 taste
293 receptor and taste receptor-like genes (Table S11), two of which (*TAS1R* and *TAS2R4*) have
294 non-synonymous SNVs with fixed variants in the north (Table S12). The functional significance
295 of these variants is unknown, but may be revealed via cellular expression systems in future
296 research [72].
297
298 **DISCUSSION**
299
300 **Comparative genomics of white-faced capuchins**
301
302 Among primates, capuchin monkeys are known for their relatively large brains, cognitive
303 capacity, and sensorimotor intelligence [7,8,73,74]. Accordingly, it is perhaps unsurprising to
304 see positive selection in the *Cebus* lineage and evidence of shifts in gene function linked to
305 brain function and development relative to other primates. In particular, positive selection in

306 *WDR62*, *BPTF*, *BBS7*, *NUP133*, and *MTOR*, and *PHF8* supports our hypothesis that the
307 capuchin lineage has undergone adaptation linked to brain development. The association of
308 several of these genes with size related brain malformations [34–37], such as microcephaly,
309 suggests that they could be influencing the large relative brain size of capuchins. Furthermore,
310 the evidence of selection in PHF8, which is associated with human cognitive capacity, aligns
311 with the link between brain size and intelligence that has been observed in other primates [75].
312 While we highlight here the putative functional roles of these genes, which are based on clinical
313 studies and comparative genomics, we acknowledge that further examination of their function in
314 the context of capuchin biology is warranted.

315 In the context of longevity, it is noteworthy that we observed genes under selection
316 associated with DNA damage response, metabolism, cell cycle and insulin signaling [76]. Of
317 particular interest are: *PARP1*, *MTOR*, *SREBF1*, *INSR1*, and *HTT*. Damage to the DNA is
318 thought to be a major contributor to aging [77]. Previous studies have also shown that genes
319 involved in DNA damage responses exhibit longevity-specific selection patterns in mammals
320 [41]. It is therefore intriguing that *PARP1*, a gene suggested to be a determinant of mammalian
321 aging [43], is under selection in capuchins. Other genomes of long-lived mammals also revealed
322 genes related to DNA repair and DNA damage responses under selection [78,79]. In the context
323 of longevity, it is also noteworthy that we observed genes under selection associated with
324 metabolism, cell cycle and insulin signaling. Other genome sequencing efforts of long-lived
325 mammals also revealed changes in such pathways [79,80]. Intriguingly, short-lived species also
326 exhibit genes under selection related to insulin receptors, raising the possibility that the same
327 pathways associated with aging in model organisms are involved in the evolution of both short-
328 and long lifespans [81], an idea supported by our results. Of course, because aging-related
329 genes often play multiple roles, for example in growth and development, it is impossible to tell
330 for sure whether selection in these genes is related to aging or to other life-history traits, like
331 growth rates and developmental times, that in turn correlate with longevity [82]. Therefore,

332 although we should be cautious about the biological significance of our findings, it is tempting to
333 speculate that, like in other species, changes to specific aging-related genes or pathways, could
334 contribute to the longevity of capuchins.
335
336 **Population genomics and local adaptation with fecalFACS**
337
338 Through a novel use of flow cytometry/FACS, we have developed a new method for the
339 isolation of epithelial cells from mammalian feces for population genomics. We generated the
340 first high-coverage, minimally biased mammalian genome solely from feces. Additionally, we
341 have demonstrated that fecalFACS can be used to generate low coverage SNP datasets that
342 are suitable for population assignment and clustering. FecalFACS is cost-effective and
343 minimizes the biases that commonly occur in traditional bait-and-capture approaches to the
344 enrichment of endogenous DNA from feces. Furthermore, fecalFACS does not require costly
345 impractical preservation of biomaterial in liquid nitrogen; rather, we rely on room-temperature
346 stable storage in RNAlater. FecalFACS offers great benefits to the field of mammalian
347 conservation and population genomics.
348 White-faced capuchins are the most northerly distributed member of the Cebinae, having
349 dispersed over the isthmus of Panama in a presumed speciation event with *C. capucinus* in
350 South America ~1.6 mya [13–15]. After expanding during the early late Pleistocene, white-faced
351 capuchins appear to have undergone a dramatic reduction in effective population size. This
352 pattern predates the movement of humans into Central America, and could reflect a series of
353 population collapses and expansions caused by glacial shifts and fluctuating forest cover
354 availability during the Pleistocene. At a finer scale, we observed a clear demarcation between
355 the northern dry- and southern wet-forest populations in Costa Rica. Higher levels of
356 heterozygosity in the south and lower levels in the northwest are in accordance with the
357 hypothesis that capuchins dispersed northwards across Costa Rica. White-faced capuchins in

358 SSR are near the northernmost limits of their range, which extends as far north as Honduras,
359 and we predict they may represent some of the least genetically diverse members of their
360 species. Given the limitations of the available sampling sites, it is possible that the appearance
361 of an ecological divide is actually evidence of isolation by distance; however, given that the
362 single individual from Cañas clusters closely with the individuals from SSR, despite a
363 geographic distance of more than 100 km, we suggest that isolation by distance does not
364 completely explain the population differentiation.

365 We found evidence that the recent northern expansion of *Cebus imitator* has undergone
366 local adaptation to the extreme seasonality of rainfall and food availability. The effects of
367 seasonality have been linked to biological consequences for Costa Rican white-faced capuchins
368 in other contexts. Seasonal fluctuations in food abundance and rainfall are associated with
369 compositional changes in the gut microbiome of capuchins at SSR [83], which differs markedly
370 from that observed in capuchins inhabiting nonseasonal forests [84]. Previous capuchin
371 research in the dry forest also demonstrates seasonal negative energy balance and periods of
372 pronounced muscle loss through catabolic processes [59]. These observations fit with the notion
373 that animals living in seasonal environments, or pursuing seasonal migrations, are more likely to
374 have weight fluctuations through binge-subsist cycles that map onto food abundance [85].
375 Accordingly, we observed population-specific variation in genes implicated in water/salt balance,
376 kidney function, muscle wasting, and metabolism (e.g. *SERPINC1*, *BCAS3*, *ITGA7*, *ISPD, and*
377 *GLIS3*). In light of this, we contend that seasonal drought and food shortages would create an
378 environment favoring efficient catabolism when needed and adaptations for maintaining water
379 balance. Given that selection operates on both gene function and regulation, we suspect the
380 observed variation is affecting gene expression or enzymatic efficiency, which offers a
381 promising avenue for future research.

382 Additionally, we found evidence of 408 OR genes, 28 TASR, and 7 VR genes that are
383 putatively functional. These numbers are similar to, or slightly higher than, the number of

384 chemosensory genes identified in other anthropoid primates [86–90]. The VR gene repertoire of
385 capuchins highlights the persistent role of the vomeronasal organ that is used in social
386 communication of other mammals, but that has been nearly lost from all African and Asian
387 monkeys [89]. Like most other primates in the Americas, capuchins possess an intriguing color
388 vision system characterized by extensive intraspecific genotypic and phenotypic variation [32–35].
389 Contrary to our prediction, dry-forest and rainforest capuchins have similar numbers of color
390 vision, taste and olfactory receptors. This indicates that there are not greater numbers of
391 functional sensory genes in areas of higher vegetative biodiversity or humidity. The tuning of the
392 chemosensory receptors may vary between habitat types and may be elucidated by future work.
393 While we did observe population-specific variation in CCDC66, which is expressed in
394 photoreceptive layers of the retina and may affect odor discrimination, the ecological
395 significance of this result, if any, is unclear at present but may warrant future attention.
396

397 **CONCLUSION**

398

399 We provide the first annotated reference assembly for a capuchin monkey. We observed
400 evidence of selection in *C. imitator* on genes involved with brain function and cognitive capacity.
401 These results are in accordance with the remarkably long life span, large brain, and high degree
402 of sensorimotor intelligence that has been observed in capuchins. These genes are good
403 candidates for further investigation of traits which have evolved in parallel in apes and other
404 mammals. Through a novel use of flow cytometry/FACS, we developed a new method for the
405 isolation of epithelial cells from mammalian feces for population genomics. FecalFACS allowed
406 us to generate both the first high-coverage, minimally biased mammalian genome solely from
407 feces, as well as low coverage SNV datasets for population level analyses. In our population
408 level analysis of wet- and dry-forest capuchins, we observed both evidence of population
409 structure between and local adaptation to these different habitats. In particular, we identified

410 selection in genes related to food and water scarcity, as well as muscular wasting, all of which
411 have been observed during seasonal extremes in the dry forest population.
412
413 **METHODS**
414
415 *Study populations and sample collection*
416 Central American white-faced capuchins *(Cebus imitator)*, a member of the gracile
417 radiation of capuchins (genus *Cebus*) [91], were recently recognized as a species, distinct from
418 *C. capucinus* in South America [14]. *Cebus imitator* occupies a wide diversity of habitats,
419 spanning lowland rainforests and cloud forests in Panama and throughout southern, eastern,
420 and central Costa Rica, and tropical dry forests in northwestern Costa Rica and Nicaragua. The
421 annual precipitation and elevation of rainforest versus dry forest biomes in their current range
422 vary dramatically, leading to considerable variation in the resident flora and fauna [92,93]. We
423 sampled individual Costa Rican capuchin monkeys from populations inhabiting two distinct
424 habitats. 1) lowland rainforest around Quepos, Puntarenas Province; and 2) tropical dry forest at
425 two sites in Guanacaste Province. In total, we collected samples from 23 capuchins, a list of
426 which is provided in table S3.
427 We sampled capuchins inhabiting a lowland tropical rainforest biome by collaborating
428 with *Kids Saving the Rainforest* (KSTR) in Quepos, Costa Rica. We acquired blood samples
429 from 4 wild capuchins from nearby populations who were undergoing treatment at the facility
430 (although we were unable to collect paired fecal samples). For one of these individuals, an adult
431 male white-faced capuchin that was mortally wounded by a vehicle in Costa Rica, we
432 additionally sampled tissues from several organs. DNA derived from the kidney was used for the
433 reference genome assembly.
434 We collected 21 samples from 19 individuals in the northern tropical dry forest. 16 fecal
435 samples and 4 tissue samples were from free-ranging white-faced capuchin monkeys (*Cebus*

436 *imitator*) at in the Sector Santa Rosa (SSR), part of the Área de Conservación Guanacaste in
437 northwestern Costa Rica, which is a 163,000 hectare tropical dry forest nature reserve (Figure
438 1). Behavioral research of free-ranging white-faced capuchins has been ongoing at SSR since
439 the 1980's which allows for the reliable identification of known individuals from facial features
440 and bodily scars [94]. The 16 fresh fecal samples were collected from 14 white-faced capuchin
441 monkeys immediately following defecation (Table S3). We placed 1 mL of feces into conical 15
442 mL tubes pre-filled with 5 mL of RNAlater. RNAlater preserved fecal samples were sent to the
443 University of Calgary, where they were stored at room temperature for up to three years. To
444 evaluate other preservation methods, we also collected two additional capuchin monkey fecal
445 samples (SSR-FL and a section of SSR-ML), which we stored in 1X PBS buffer and then frozen
446 in liquid nitrogen with a betaine cryopreservative [95]. Given the logistical challenges of carrying
447 liquid nitrogen to remote field sites, we prioritized evaluation of samples stored in RNAlater. We
448 also collected tissue and blood samples opportunistically. During the course of our study, 4
449 individual capuchin monkeys died of natural causes at SSR, from whom we were able to collect
450 tissue samples, which were stored in RNAlater. Additionally, we collected a blood sample from 1
451 northern dry-forest individual housed at KSTR that originated near the town of Cañas, which is
452 ~100 km southwest of SSR.
453
454 *Genome-wide sequencing, genome assembly and gene annotation*
455 We assembled a reference genome for *Cebus imitator* from DNA extracted from the
456 kidney of a male Costa Rican individual (KSTR64) using a short read approach (Illumina HiSeq
457 2500). Based on a genome estimate of 3 Gb, the total sequencing depth generated was 81X,
458 including 50X of overlapping read-pairs (200 bp insert), 26X and 5X of 3 and 8 kbs insert read
459 pairs, respectively. The combined sequence reads were filtered and assembled using default
460 parameter settings with ALLPATHS-LG [96]. To improve the quality of gene annotation, we
461 isolated total RNA from the whole blood of an adult male white-faced capuchin (ID: CNS-HE)

462 permanently residing at the KSTR wildlife rehabilitation center. The blood was immediately
463 stored in a PAXgene blood RNA tube (Qiagen), and frozen at ultralow temperatures for
464 subseqeunt use. To extract total RNA, we used the PAXgene Blood RNA kit following the
465 manufacturer's recommended protocols. A RiboZero library construction protocol was followed
466 according to the manufacturer's specifications and sequenced on an Illumina HiSeq 2000
467 instrument creating 150 bp paired-end reads. We assembled the FASTQ sequence files into
468 transcripts with Trinity, and submitted the assembled transcriptome to the National Center for
469 Biotechnology Information (NCBI) to assist in gene annotation. The capuchin genome assembly
470 was annotated with the NCBI pipeline previously described here:
471 (http://www.ncbi.nlm.nih.gov/books/NBK169439/).
472
473 *Phylogenetic arrangement and data treatment*
474 The phylogenetic arrangement in this study included 14 species as outgroups to *C.*
475 *imitator*: three Platyrrhini (*Callithrix jacchus*, *Aotus nancymaae*, *Saimiri* boliviensis), six
476 Catarrhini (*Macaca mulatta*, *Rhinopithecus roxellana*, *Nomascus leucogenys*, *Pan troglodytes*,
477 *Homo sapiens*, *Pongo abelii*), one Strepsirrhini (*Microcebus murinus*), one rodent (*Mus*
478 *musculus*), and three Laurasiatheria (*Canis lupus familiaris*, *Bos taurus,* and *Sus scrofa*).
479 Genomic cds were downloaded from Ensembl and NCBI (Table S13). The sequences per
480 genome were clustered using CD-HITest version 4.6 [97] with a sequence identity threshold of
481 90% and an alignment coverage control of 80%. To remove low quality sequences and keep the
482 longest transcript per gene, we used TransDecoder.LongOrfs and TransDecoder.Predict
483 (https://transdecoder.github.io) with default criteria.
484
485 *Orthology identification*
486 The orthology assessment was performed with OMA stand-alone v. 2.3.1 [98]. The OMA
487 algorithm makes strict pairwise "all-against-all" sequence comparisons and identifies the

488 orthologous pairs (genes related by speciation events) based on evolutionary distances. These
489 orthologous genes were clustered into Orthologous Groups (OGs). All OGs included one
490 ortholog sequence from capuchin and at least one outgroup. The tree topology was obtained
491 from TimeTree (http://www.timetree.org/). We identified 7,519 OGs present among the 15
492 species. Each orthogroup shared by all species was translated into amino acids using the
493 function pxtlate -s in phyx [99]. Amino acid sequences were aligned using the L-INS-i algorithm
494 from MAFFT v.7 [100]. We generated codon alignments using pxaa2cdn in phyx. To avoid false
495 positives in low quality regions, the codon alignments were cleaned with the codon.clean.msa
496 algorithm in rphast [101], using human as a reference sequence. We used conservative
497 methodologies of homology and data cleaning to obtain a smaller number of orthologous genes
498 that avoided false positives with high confidence.
499
500 *Positive natural selection analysis through codon-based models of evolution and enrichment*
501 *tests*
502 To evaluate signals consistent with positive selection in the *C. imitator* genome, we
503 explored variation in the ratio of non-synonymous and synonymous substitutions ($d_N/d_S=\omega$) in
504 the ancestor of *Cebus*. We used branch and branch-site substitution models with a maximum
505 likelihood approach in PAML v4.9 [32], which we implemented through the python framework
506 ETE-toolkit with the ete-evol function [102]. We compared the null model where the omega (ω)
507 value in the branch marked as foreground was set with 1, with the model where the ω value was
508 estimated from the data [103]. Likelihood ratio tests (LRT) were used to test for significance
509 between the models and probability values were adjusted with a false discovery rate correction
510 for multiple testing with a q-value < 0.05 for the two positive selection models (branch and
511 branch-site).
512 We performed functional annotation analysis using DAVID bioinformatics resources -
513 DAVID 6.7 [33], to ascertain which ontology processes the genes with signals of positive

514 selection were involved. We focussed the enrichment analysis on two functional categories:
515 Biological Processes (BP) and Genetic Association Database (GAD). GAD is a database from
516 complex diseases and disorders in humans. Finally, the genes with positive section signal were
517 intersected with the GenAge database (build 19,307 genes) [42].
518
519 *FACS*
520 Before isolating cells by fluorescence-activated cell sorting, fecal samples were prepared
521 using a series of washes and filtration steps. Fecal samples were vortexed for 30 s and
522 centrifuged for 30 s at 2,500 g. Then the supernatant was passed through a 70 μm filter into a
523 50 mL tube and washed with DPBS. After transferring the resultant filtrate to a 15 mL tube, it
524 was centrifuged at 1,500 RPM for 5 minutes to pellet the cells. Then we twice washed the cells
525 with 13 mL of DPBS. We added 500 μL of DPBS to the pellet and re-filtered through a 35 μm
526 filter into a 5 mL FACS tube. We prepared a negative control (to control for auto-fluorescence)
527 with 500 μL of DPBS and one drop of the cell solution. To the remaining solution, we added 1
528 μL of AE1/AE3 Anti-Pan Cytokeratin Alexa Fluor® 488 antibody or TOTO-3 DNA stain, which
529 we allowed to incubate at 4°C for at least 30 minutes.
530 We isolated cells using a BD FACSAria™ Fusion (BD Biosciences) flow cytometer at the
531 University of Calgary Flow Cytometry Core. To sterilize the cytometer's fluidics before
532 processing each sample, we ran a 3% bleach solution through the system for four minutes at
533 maximum pressure. We assessed background fluorescence and cellular integrity by processing
534 the negative control sample prior to all prepared fecal samples. For each sample we first gated
535 our target population by forward and side scatter characteristics that were likely to minimize
536 bacteria and cellular debris (Figure S8). Secondary and tertiary gates were implemented to
537 remove cellular agglomerations. Finally, we selected cells with antibody or DNA fluorescence
538 greater than background levels. In cases when staining was not effective, we sorted solely on
539 the first three gates. Cells were pelleted and frozen at -20°C.

540

541 *DNA Extraction and Shotgun Sequencing*

542 We extracted fecal DNA (fDNA) with the QIAGEN DNA Micro kit, following the "small
543 volumes of blood" protocol. To improve DNA yield, we increased the lysis time to three hours,
544 and incubated 50 µL of 56°C elution buffer on the spin column membrane for 10 minutes. DNA
545 concentration was measured with a Qubit fluorometer. Additionally, to calculate endogenous
546 DNA enrichment, we extracted DNA directly from five fecal samples prior to their having
547 undergone FACS. We extracted DNA from the nine tissue and blood samples using the
548 QIAGEN Gentra Puregene Tissue kit and DNeasy blood and tissue kit, respectively.
549 For the fecal samples, DNA was fragmented to 350 bp with a Covaris sonicator. We built whole
550 genome sequencing libraries with the NEB Next Ultra 2 kit using 10-11 PCR cycles. Fecal
551 genomic libraries were sequenced on an Illumina NextSeq (2x150 PE) at the University of
552 Calgary genome sequencing core and an Illumina HighSeq 4000 at the McDonnell Genome
553 Institute at Washington University in St. Louis (MGI). Using ½ of one HiSeq 4000 lane, we
554 achieved an average coverage of 12.2X across the *Cebus imitator* 1.0 genome (sample SSR-
555 ML). Other fecal samples were sequenced to average depths of 0.1-4.4X (Table S4). High-
556 coverage (10.3-47.6X), whole genome shotgun libraries were prepared for the blood and tissue
557 DNA samples and sequenced on an Illumina X Ten system at MGI. For population analyses
558 within capuchins, we mapped genomic data from all 23 individuals sequenced (Table S3) to the
559 reference genome.

560

561 *Mapping and SNV Generation*

562 Reads were trimmed of sequencing adaptors with Trimmomatic [104]. Subsequently, we
563 mapped the *Cebus* reads to the *Cebus imitator* 1.0 reference genome (GCF_001604975.1) with
564 BWA mem [105] and removed duplicates with Picard Tools
565 (http://broadinstitute.github.io/picard/) and SAMtools [106]. We called SNVs for each sample

566 independently using the *Cebus* genome and the GATK UnifiedGenotyper pipeline (*-out_mode*
567 *EMIT_ALL_SITES*) [107]. Genomic VCFs were then combined using GATK's CombineVariants
568 restricting to positions with a depth of coverage between 3 and 100, mapping quality above 30,
569 no reads with mapping quality zero, and variant PHRED scores above 30. Sequencing reads
570 from one of the high coverage fecal samples (SSR-FL) bore a strong signature of human
571 contamination (16%), and were thus excluded from SNV generation. We included reads from
572 nine tissue/blood samples and one frozen fecal sample with high coverage (SSR-ML). In total,
573 we identified 4,184,363 SNVs for downstream analyses.
574 To remove potential human contamination from sequenced libraries, we mapped
575 trimmed reads to the *Cebus imitator* 1.0 and human (hg38) genomes simultaneously with
576 BBsplit [108]. Using default BBsplit parameters, we binned separately reads that mapped
577 unambiguously to either genome. Ambiguously mapping reads (i.e. those mapping equally well
578 to both genomes) were assigned to both genomic bins, and unmapped reads were assigned to
579 a third bin. We calculated the amount of human genomic contamination as the percentage of
580 total reads unambiguously mapping to the human genome (Table S4). After removing
581 contaminant reads, all libraries with at least 0.5X genomic coverage were used for population
582 structure analysis.
583 In order to test the effect of fecalFACS on mapping rates, we selected five samples at
584 random (SSR-CH, SSR-NM, SSR-LE, SSR-PR, SSR-SN) to compare pre- and post-FACS
585 mapping rates. To test for an increase in mapping percentage, we ran a one-sample paired
586 Wilcoxon signed-rank test on the percentages of reads that mapped exclusively to the *Cebus*
587 genome before and after FACS. Additionally, we ran Pearson's product moment correlations to
588 test for an effect of the number of cells (log10 transformed) on rates of mapping, read
589 duplication, and nanograms of input DNA. The above tests were all performed in R.
590
591 *High coverage fecal genome comparison*

592 We made several comparisons between our high-coverage feces-derived genome and
593 the blood/tissue-derived genomes using window-based approaches. For each test, the feces-
594 derived genome should fall within the range of variation for members of its population of origin
595 (SSR). Deviations from this, for example all fecal genomes clustering together, would indicate
596 biases in our DNA isolation methods. To assess this, we constructed 10 KB windows with a 4KB
597 slide along the largest scaffold (21,314,911 bp) in the *C. imitator* reference genome. From these
598 windows, we constructed plots of coverage density and the distribution of window coverage
599 along the scaffold. Secondly, we assessed the level of heterozygosity in 1 MB / 200 KB sliding
600 windows throughout the ten largest scaffolds. For each high-coverage genome, we plotted the
601 density distribution of window heterozygosity. We measured genome-wide GC content with the
602 Picard Tools CollectGcBiasMetrics function. The percentage of GC content was assessed
603 against the distribution of normalized coverage and the number of reads in 100 bp windows per
604 the number reads aligned to the windows.
605
606 *Population Structure*
607 Given the large degree of difference in coverage among our samples, (less than 1X to
608 greater than 50X), we performed pseudodiploid allele calling on all samples. For each library, at
609 each position in the SNV set, we selected a single, random read from the sequenced library.
610 From that read, we called the variant information at the respective SNV site for the given library.
611 In so doing, we generated a VCF with a representative degree of variation and error for all
612 samples.
613 To assess population structure and infer splits between northern and southern groups of
614 Costa Rican white-faced capuchins, we constructed principal components plots with
615 EIGENSTRAT [109] and built population trees with TreeMix [110]. Because we ascertained
616 variants predominantly with libraries that were of tissue/blood origin, we built principal
617 components solely with SNVs from these libraries and projected the remaining fecal libraries

618 onto the principal components. For our maximum likelihood trees, we used two outgroups
619 (*Saimiri sciureus*, and *Cebus albifrons*), with *S. sciureus* serving as the root of the tree. Given
620 the geographic distance and anthropogenic deforestation between northern and southern
621 populations, we assumed no migration. To account for linkage disequilibrium, we grouped SNVs
622 into windows of 1,000 SNVs. Population size history was inferred from the highest coverage
623 non-reference individual, SSR-RM08 with PSMC [111], using default parameters
624 (https://github.com/lh3/psmc).
625
626 *Local adaptation, F_{ST}, heterozygosity, relatedness*
627 For all analyses of local adaptation and heterozygosity between populations, we
628 excluded individuals from our low coverage dataset. We tested for the degree of relatedness
629 among all high and low coverage individuals using READ [112] and identified two of the high-
630 coverage individuals from SSR, SSR-ML and SSR-CR, as potential first degree relatives (Figure
631 S9, Table S14). For all statistical analyses of high-coverage samples, we removed SSR-ML,
632 because SSR-CR was sequenced to higher average depth.
633 For each individual, we calculated heterozygosity in 1 Mb / 200 Kb sliding windows
634 across the genome for all scaffolds at least 1 Mb in length. Windows were generated with
635 BedTools windowMaker [113] and heterozygosity was calculated as the per-site average within
636 each window. Based upon a visual inspection of the average heterozygosity values across the
637 genome (Figure 5), we classified a window as part of a run of homozygosity if the window's
638 average heterozygosity fell below 0.0002. Descriptive statistics and two-sided Wilcoxon tests
639 were calculated in R.
640 For each high-coverage sample, we calculated the Hudson's F_{ST} ratio of averages [114]
641 in 20 kb windows with a slide of 4 Kb across the genome. Among the genes present in each
642 window in the top 0.5% and top 0.1% of F_{ST} values, we searched for SNPs with high or
643 moderate effects using SnpEff and identified those SNPs with high F_{ST} values (> 0.75) using

644 VCFtools. We searched for functional enrichment of our population gene set using ToppFun in
645 the ToppGene Suite [115]. In an effort to identify candidate genes for further investigation, we
646 set low thresholds for significance (FDR p-value < 0.1 and the minimum number of genes per
647 category to 1).
648
649 Chemosensory genes
650 The chemosensory behaviors of capuchins have been well-studied [31], and taste and
651 olfaction are suspected to play an important role in their foraging ecology. 614 orthologous olfactory
652 receptor genes were identified in the *Cebus* reference genome using previously described
653 methods [116]. Briefly, putative OR sequences were identified by conducting TBLASTN
654 searches against the capuchin reference assembly using functional human OR protein
655 sequences as queries with an e-value threshold of 1e-20. For each reference-derived OR gene,
656 we added 500 bp of flanking sequence to both the 5' and 3' ends in a bed file. For each
657 individual, we extracted the OR gene region from the gVCF and generated a consensus
658 sequence defaulting to the reference allele at variable site using bcftools [117]. The number of
659 intact, truncated, and pseudogenized OR genes in each individual were identified using the
660 ORA pipeline [118]. We considered an OR gene to be putatively functional if its amino acid
661 sequence was at least 300 amino acids in length. ORA was further used to classify each OR
662 into the appropriate class and subfamily with an e-value cutoff of 1e-10 to identify the functional
663 OR subgenome for each individual [118,119]. Taste and vomeronasal receptor genes were
664 identified in the NCBI genome annotation, and variable positions were located by scanning the
665 VCF with VCFtools [120] and bash. The positions of *Cebus* opsin tuning sites have been
666 identified previously [68]. With the high-coverage dataset, we identified the allele(s) present at
667 each locus in the VCF. For each low-coverage fecal-derived genome, we located the position of
668 the tuning site in the bam file using SAMtools [117] tview and manually called the variant when
669 possible.

670

671 **DECLARATIONS**

672

673 **Ethics approval and consent to participate**

674

675 This research adhered to the laws of Costa Rica, the United States, and Canada and
676 complied with protocols approved by the Área de Conservación Guanacaste and by the Canada
677 Research Council for Animal Care through the University of Calgary's Life and Environmental
678 Care Committee (ACC protocol AC15-0161). Samples were collected with permission from the
679 Área de Conservación Guanacaste (ACG-PI-033-2016) and CONAGEBIO (R-025-2014-OT-
680 CONAGEBIO; R-002-2020-OT-CONAGEBIO). Samples were exported from Costa Rica under
681 permits from CITES and Area de Conservacion Guanacaste (2016-CR2392/SJ #S 2477, 2016-
682 CR2393/SJ #S 2477, DGVS-030-2016-ACG-PI-002-2016; 012706) and imported with
683 permission from the Canadian Food and Inspection Agency (A-2016-03992-4).

684

685 **Consent for Publication**

686

687 Not applicable

688

689 **Availability of data and materials**

690

691 The reference genome is available at NCBI through BioProjects PRJNA298580 and
692 PRJNA328123. RNAseq reads used in genome annotation can be accessed through
693 PRJNA319062. The sequencing reads used in the local adaptation will be released by NCBI
694 upon publication (PRJNA610850), and are available immediately to reviewers and editors upon
695 request from the corresponding author.

Competing Interests

The authors declare that they have no competing interests.

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Author contributions

ADM and JDO conceived of the project

JDO, ADM, YN, WCW, RK, JdC, JPM, and TMB contributed to the study design

JDO, DTM, MdM, MJM, LFK, and MCJ performed computational analyses

JDO, RK, and JT conducted flow cytometry

JDO, ADM, CF, JH, and EL conducted molecular lab work

ADM, GHP, ADF, JAH, and SK contributed samples

ADM, RK, and TMB provided computational and laboratory resources

JDO, ADM, MJM, JPM, and DTM wrote the manuscript with commentary from all authors

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FIGURES

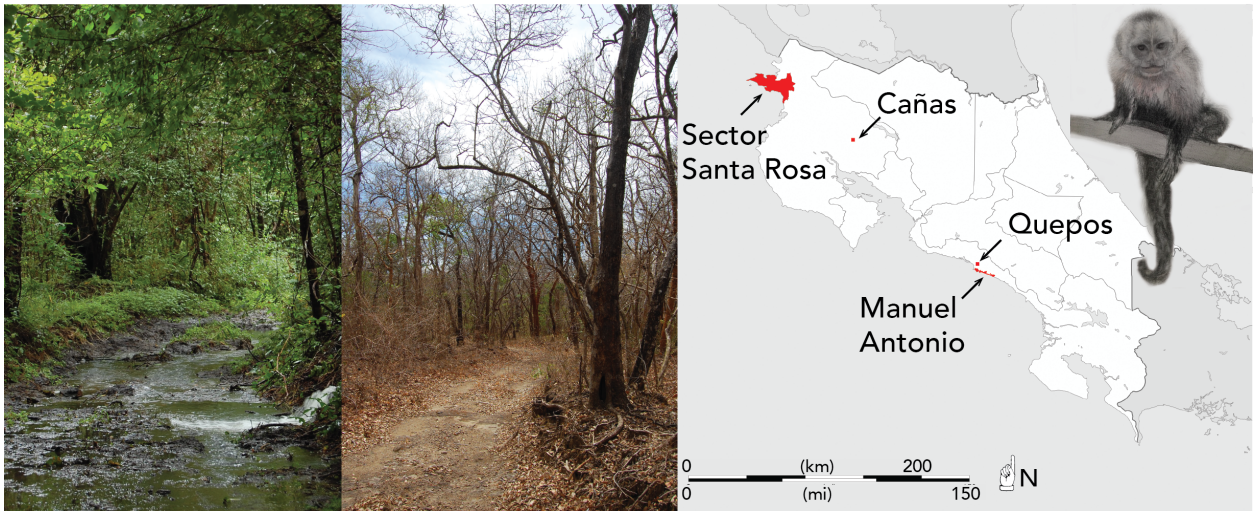

Figure 1: Sector Santa Rosa (SSR) during wet (left) and dry (middle) seasons. Right: Map of sampling locations in Costa Rica. The two northern sites, SSR and Cañas, have tropical dry forest biomes, whereas the two southern sites, Quepos and Manuel Antonio are tropical wet forests. Photos - Amanda Melin; Drawing of white faced capuchin monkey - Alejandra Tejada-Martinez; Map: Eric Gaba – Wikimedia Commons user: Sting

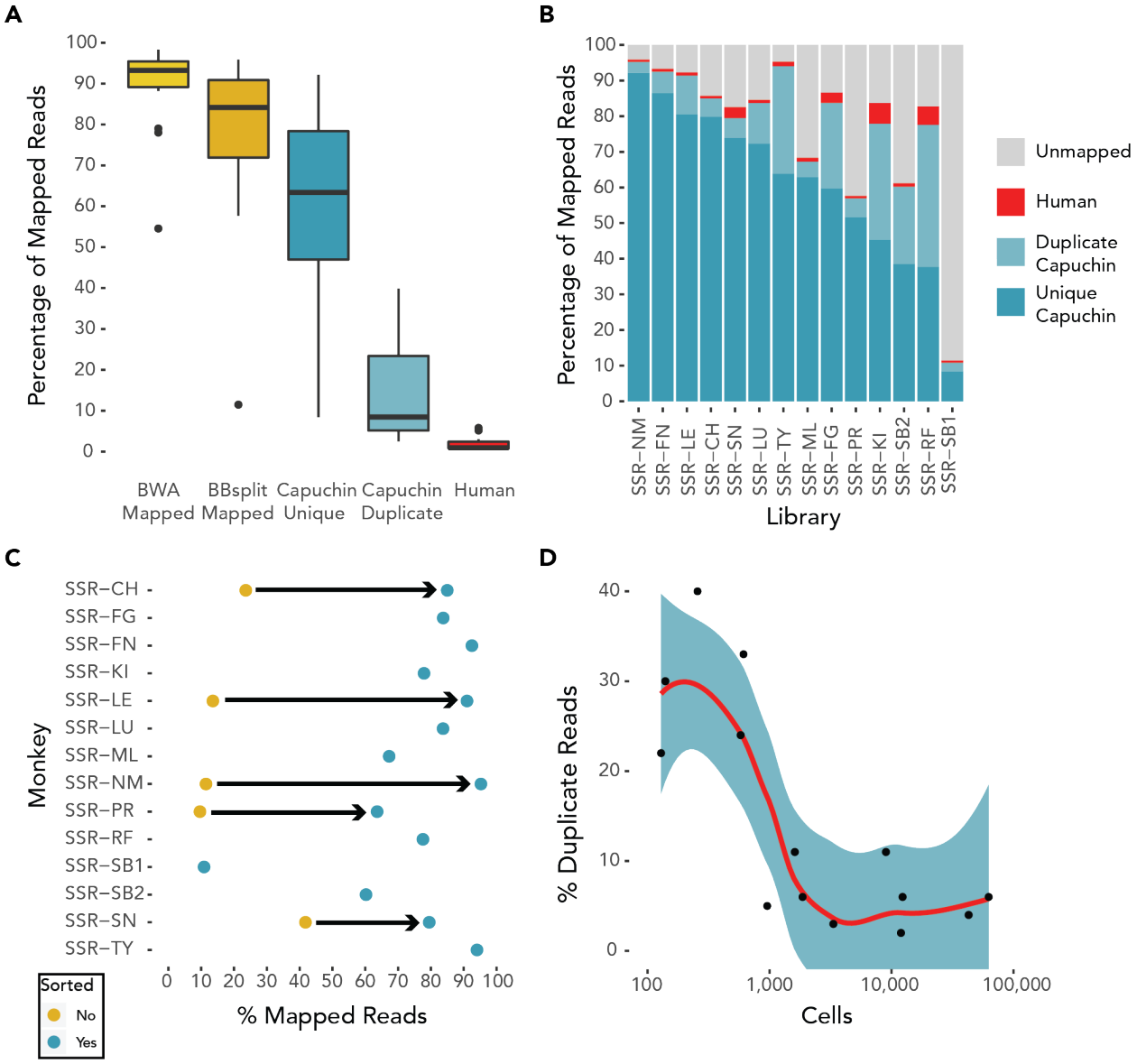

Figure 2: Mapping percentages of sequencing reads from RNAlater preserved fDNA libraries prepared with FACS for A) all samples [Box-plot elements: center line, median; box limits, upper and lower quartiles; whiskers, 1.5x interquartile range; points, outliers], and B) individual libraries. C) Increase in mapping rate for RNAlater preserved samples. D) Relationship between mapped read duplication and number of cells with LOESS smoothing. The duplicate rate decreases sharply once a threshold of about 1,000 cells is reached.

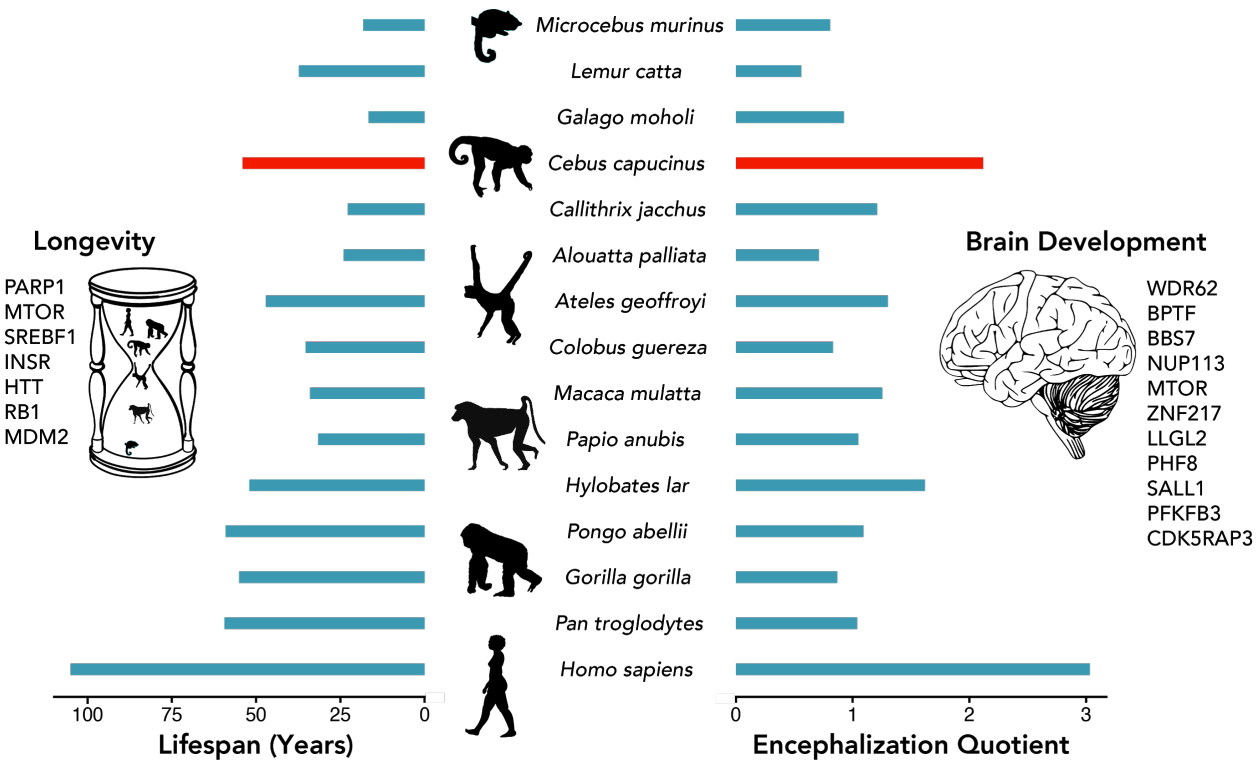

Figure 3: Genes under positive selection in white-faced capuchin monkeys that are associated with longevity and brain development. Values from *Cebus capucinus* are used in place of *Cebus imitator*, given the recent taxonomic split.

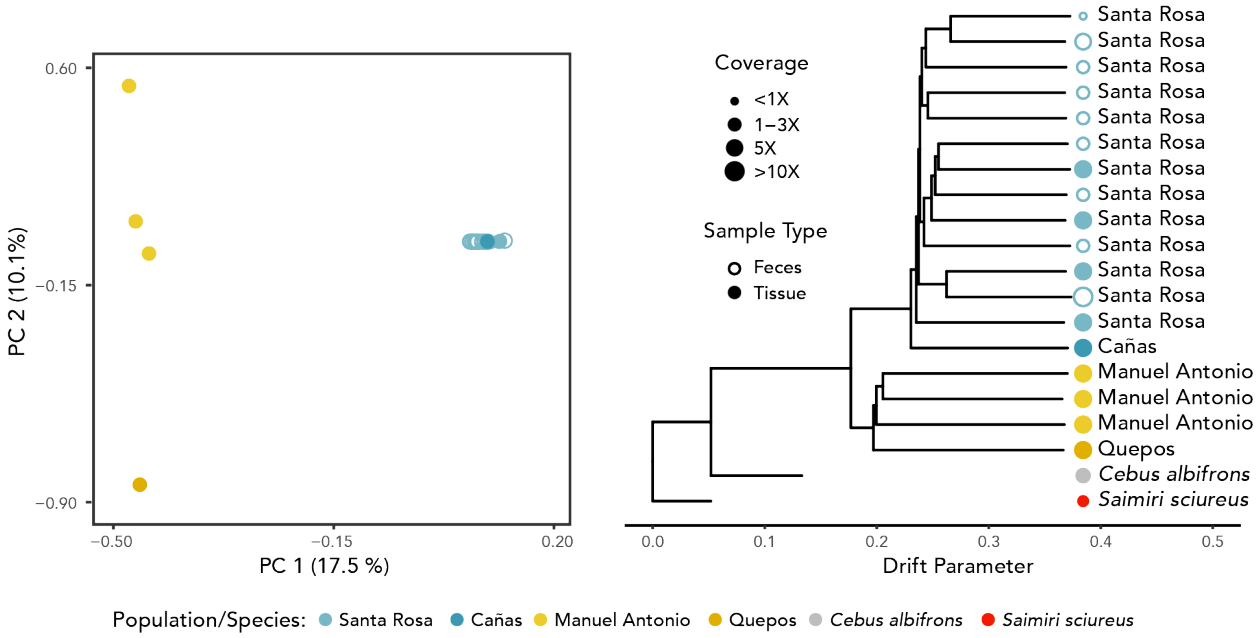

Figure 4: Population subdivision in *Cebus imitator*. Left: Principal components of 13 fecal and 10 blood/tissue libraries from white faced capuchins. Individuals from northern and southern sites separate on PC 1. Low- and high-coverage *C. imitator* samples from Santa Rosa plot in the same cluster. Right: Maximum likelihood tree of 9 fecal and 10 blood/tissue libraries from *C. imitator* (samples with less than 0.5X coverage were excluded). Among the white-faced capuchin samples, individuals from northern (dry forest) and southern (wet forest) regions form the primary split; secondary splits reflect the individuals from different sites within regions. The short branch lengths of the outgroups are a result of only polymorphic positions within *C. imitator* being used to construct the tree.

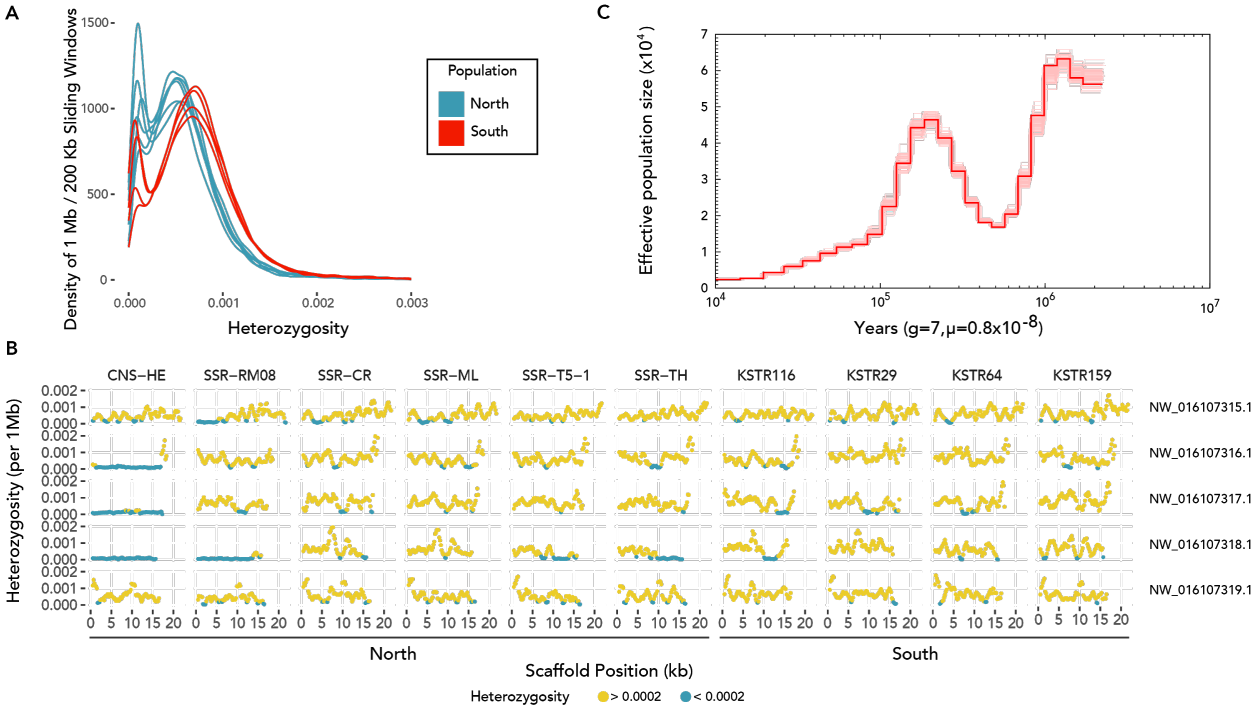

Figure 5: A: Density plot of 1 Mb windows with a slide of 200 Kb in northern and southern populations. The distribution of windows from the northern population indicates lower heterozygosity than the southern distribution.The individuals from the southern population show consistently higher values. B: Long runs of homozygosity in the 5 largest scaffolds. Blue dots represent windows with depleted heterozygosity. The individuals with the longest runs of homozygosity come from the northern population. C: PSMC plot of effective population size over time.

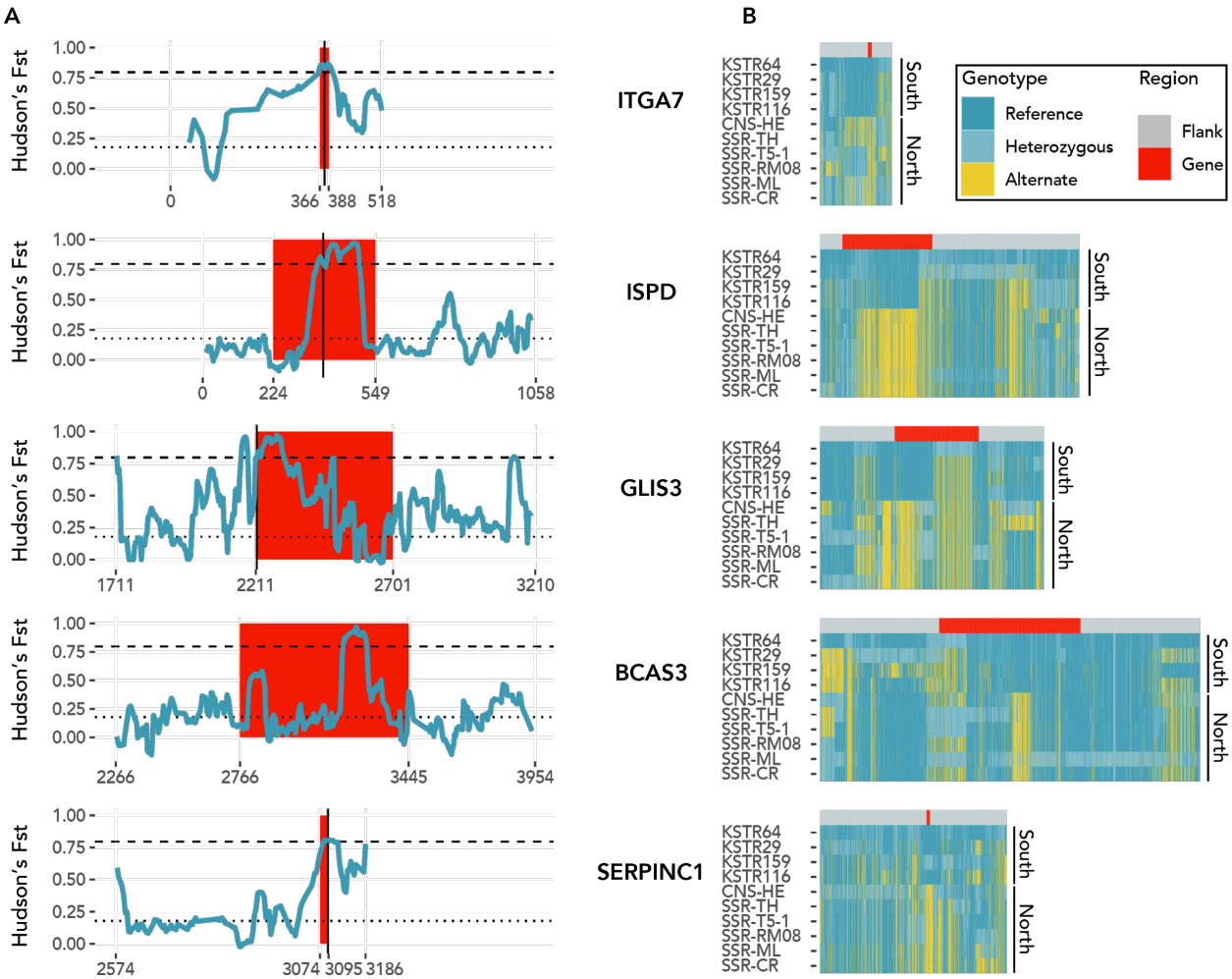

Figure 6: Highly differentiated genes between wet and dry forest populations involved in diabetes, kidney function, and creatinine levels. A: Hudson's F_{ST} within windows of 20kb with a 4 kb slide. Gene regions are in red, flanked by 500 kb (or length to beginning or end of scaffold) of sequence. X-axis values correspond to position along the scaffold. The dotted line indicates average F_{ST} value across all windows (F_{ST} = 0.178), and the dashed line represents the top 0.5% of values (F_{ST} = 0.797). Vertical black lines indicate a non-synonymous SNP with an Fst >= 0.750, excluding BCAS3 (see Results). B: Heatmaps indicating the pattern of SNP variation within and surrounding highly divergent genes. SNVs within the genes are located under the red band and those within 200 kb of flanking region under the gray bands.

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