- Assessing connectivity despite high diversity in island
- populations of a malaria mosquito
- ³ Christina M. Bergey^{1,2,3,*}, Martin Lukindu^{1,2}, Rachel M. Wiltshire^{1,2},
- Michael C. Fontaine^{4,5}, Jonathan K. Kayondo⁶, and Nora J. Besansky^{1,2,*}
- ¹Department of Biological Sciences, University of Notre Dame, Notre Dame, IN 46556, USA
- ²Eck Institute for Global Health, University of Notre Dame, Notre Dame, IN 46556, USA
- ³Departments of Anthropology and Biology, Pennsylvania State University, University Park, PA
- 8 16802, USA.

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- ⁴Groningen Institute for Evolutionary Life Sciences (GELIFES), University of Groningen, PO
 Box 11103 CC, Groningen, The Netherlands.
- ⁵MIVEGEC, IRD, CNRS, University of Montpellier, Montpellier, France.
- ⁶Department of Entomology, Uganda Virus Research Institute (UVRI), Entebbe, Uganda.
 - * To whom correspondence should be addressed.

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- 16 Corresponding authors: C.M.B. (cxb585@psu.edu) and N.J.B. (nbesansk@nd.edu)

Abstract:

Documenting isolation is notoriously difficult for species with vast polymorphic populations. High proportions of shared variation impede estimation of connectivity, even despite leveraging information from many genetic markers. We overcome these impediments by combining classical analysis of neutral variation with assays of the structure of selected variation, demonstrated using populations of the principal African malaria vector *Anopheles gambiae*. Accurate estimation of mosquito migration is crucial for efforts to combat malaria. Modeling and cage experiments suggest that mosquito gene drive systems will enable malaria eradication, but establishing safety and efficacy requires identification of isolated populations in which to conduct field-testing. We assess Lake Victoria islands as candidate sites, finding one island 30 kilometers offshore is as differentiated from mainland samples as populations from across the continent. Collectively, our results suggest sufficient contemporary isolation of these islands to warrant consideration as field-testing locations and illustrate shared adaptive variation as a useful proxy for connectivity in highly polymorphic species.

The difficulties in estimating migration with genetic methods are exacerbated for large, interconnected populations exhibiting shallow population structure. Large population sizes result in high levels of polymorphism in the genome and impede accurate estimation of connectivity [1] and discernment of demographic independence from panmixia [2]. Population genetic methods for estimating migration using neutral markers may thus have limited utility when such a high proportion of diversity is shared between populations, a failing that is only partially redressed with the high quantity of markers available from massively parallel sequencing. The most powerful window into migration may instead be the distribution of selected variants [3].

The major African malaria vector Anopheles gambiae sensu stricto (henceforth An. gam-42 biae) is among the most genetically diverse eukaryotic species [4], with shallow population 43 structure [4, 5] that complicates efforts to estimate connectivity from genetic data. Overcoming these obstacles to infer migration accurately is crucial for control efforts to reduce 45 the approximately 445,000 annual deaths attributable to malaria [6]. Such vector control 46 efforts include novel methods involving the release of genetically modified mosquitoes. The 47 most promising involve introducing transgenes into the mosquito genome or its endosymbionts that interrupt pathogen transmission coupled with a gene drive system to propagate the effector genes through a population [7–9]. Such systems have recently been successfully engineered in the laboratory [10]. A detailed understanding of population structure and 51 connectivity is essential for effective implementation of any genetic control method, not least 52 a gene drive system designed to spread in a super-Mendelian fashion.

Here, we analyze population structure, demographic history, and migration between populations from genome-wide variation in An. qambiae mosquitoes living near and on the Ssese archipelago of Lake Victoria in Uganda (Fig. 1). We augment these analyses with a demonstration of our framework using selective sweep sharing as a proxy for connectivity when inferring migration in taxa with high variation. Islands present natural laboratories for disentangling the determinants of population structure, as gene flow—likely important in post-dry season recolonization [11]—is reduced. In addition to the high malaria prevalence of the islands (44% in children; 30% in children country-wide; [12]), we were motivated by 61 the potential of such an island to be a field site for future tests of gene-drive vector control strategies: Geographically-isolated islands have been proposed as locales to test the dynam-63 ics of transgene spread while limiting their movement beyond the study population [13–16]. Antecedent studies of population structure and connectivity of potential release sites are cru-65 cial to evaluate the success of such field trials, as well as to quantify the chance of migration of transgenic insects carrying constructs designed to propagate across mosquito populations 68 and country borders.

• Results

The Ssese Islands are approximately 4-50 km from the mainland, and vary in size, infrastructure, and accessibility. Sampled islands range from Banda—a small, largely forested 71 island of approximately 1 square kilometer with a single settlement—to Bugala—296 square 72 kilometers, site of a 10,000 ha oil palm plantation [17], and linked to the mainland via ferry 73 service [18]. To explore the partitioning of An. qambiae genetic variation in the Lake Victoria Basin (LVB), we sequenced the genomes of 116 mosquitoes from 5 island and 4 mainland localities (Fig. 1, Supplementary Table S1). We sequenced 10-23 individuals per site to an average depth of 17.6 \pm 4.6 (Supplementary Table S2). After filtering (detailed in Meth-77 ods), we identified 28.6 million high quality Single Nucleotide Polymorphisms (SNPs). We 78 merged our dataset with that of the An. gambiae 1000 Genomes project (Ag1000G; [4]) for a combined dataset of 12.54 million SNPs (9.86 million after linkage disequilibrium pruning) in 881 individuals.



Figure 1: Map of Lake Victoria Basin study area.

Map of study area showing sampling localities on Ssese Islands (blue) and mainland localities (red) in Lake Victoria Basin. The Ag1000G reference population, Nagongera, Tororo District, is not shown, but lies 111 km NE of Kiyindi, 57 km from the shore of Lake Victoria. Map data copyright 2018 Google.

2 Genetic structure

We analyzed LVB population structure with context from continent-wide populations [4] of An. gambiae and sister species Anopheles coluzzii mosquitoes (formerly known as An. gambiae M molecular form [19]). Both Bayesian clustering ([20]; Fig. 2a) and principal component analysis (PCA; Supplementary Fig. S1) showed LVB individuals closely related 86 to the Ugandan reference population (Nagongera, Tororo; $0^{\circ}46'12.0''N$, $34^{\circ}01'34.0''E$; ~57 87 km from Lake Victoria; Fig. 1). With ≥ 6 clusters (which optimized predictive accuracy in the clustering analysis; Supplementary Fig. S2), island samples had distinct ancestry 89 proportions (Fig. 2a), and with k = 9 clusters, we observed additional subdivision in LVB samples and the assignment of the majority of Ssese individuals' ancestry to a largely island-91 specific component (Figs. 2a, 2b, and Supplementary Fig. S3). 92 PCA of only LVB individuals indicated little differentiation among mainland samples 93 in the first two components and varying degrees of differentiation on islands, with Banda. Sserinya, and Bukasa the most extreme (Fig. 2c). Twelve of 23 individuals from Bugala, the largest, most developed, and most connected island, exhibited affinity to mainland individuals instead of ancestry typical of the islands (Supplementary Fig. S4). As both PCA and clustering analyses revealed this differentiation, we split the Bugala sample into mainlandand island-like subsets for subsequent analyses (hereafter referenced as "Bugala (M)" and "Bugala (I)," respectively). Individuals with partial ancestry attributable to the component prevalent on the mainland and the rest to the island-specific component were present on all 101 islands except Banda. 102 Differentiation concurred with observed population structure. Mean F_{ST} between sam-103 pling localities (range: 0.001-0.034) was approximately $0 \leq 0.003$ for mainland-mainland 104 comparisons and was highest in comparisons involving the small island Banda (Fig. 2d). 105 Geographic distances and F_{ST} were uncorrelated (Mantel p = 0.88; Supplementary Fig. S5). 106 Island samples showed greater within- and between-locality sharing of genomic regions identi-107

cal by descent (IBD), with sharing between nearby islands Sserinya, Banda and Bugala (Fig. 2e). Importantly, Banda Island shared no IBD regions with mainland sites, underscoring its contemporary isolation from the mainland.

111 Genetic diversity

Consistent with the predicted decrease in genetic variation for semi-isolated island populations due to inbreeding and smaller effective population sizes (N_e) , islands displayed slightly lower nucleotide diversity $(\pi; \text{Wilcoxon rank sum test } p < 0.001; \text{Fig. 3a})$, a higher proportion of shared to rare variants (Tajima's D; p < 0.001; Fig. 3b), and more linkage among SNPs (LD; $r^2; p < 0.001; \text{Fig. 3f})$. They were however, similar in inbreeding coefficient (F;p = 0.0719; Fig. 3c), number of long runs of homozygosity $(F_{ROH}; p = 0.182; \text{Fig. 3d})$, and proportions of low frequency SNPs (Fig. 3e). The small island Banda was the most extreme in these measures.

120 Demographic history

To test islands for isolation and demographic independence from the mainland, we inferred 121 the population history of LVB samples by estimating long-term and recent trends in N_e using 122 stairway plots [21] based on the site frequency spectrum (SFS; Fig. 4a) and patterns of IBD 123 sharing ([22]; Fig. 4b), respectively. Short-term final mainland sizes were unrealistically high, likely due to low samples sizes for each locality, but island-mainland differences were 125 nonetheless informative. In both, islands had consistently lower N_e compared to mainland 126 populations extending back 500 generations (\sim 50 years) and often severely fluctuated, 127 particularly in the last 250 generations (~ 22 years). Mainland sites Wamala and Kaazi had 128 island-like recent histories, with Wamala abruptly switching to an island-like pattern around 129 2005. 130

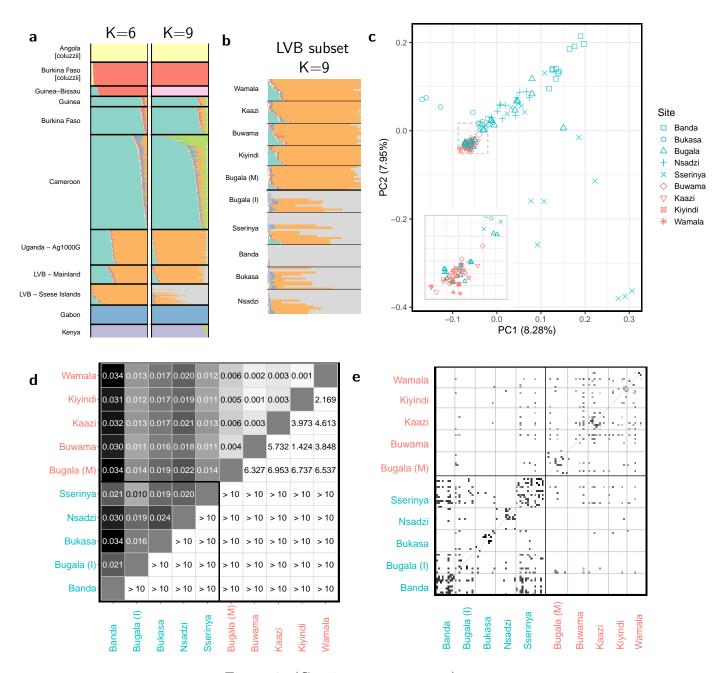


Figure 2: (Caption on next page.)

Figure 2: Population structure in the Lake Victoria Basin.

Analyses are based on chromosome 3 to avoid segregating inversions on other chromosome, unless otherwise noted. (A) ADMIXTURE-inferred ancestry of individuals in Lake Victoria Basin. Results based on analysis of LVB and Ag1000G merged dataset. Analysis is restricted to A. qambiae s. s.. Clustering shown for k=6 clusters, which minimizes cross validation error, and k=9 clusters, the lowest k for which island individuals have the majority of their ancestry assigned to an island-specific cluster. (B) Results of the clustering analysis with k = 9 clusters for LVB individuals, split by sampling locality. (C) Plot of first two components of PCA of Lake Victoria Basin individuals showing locality of origin. Mainland individuals are colored red, while island individuals are blue, and point shape indicates sampling locality. Based on these results and that of ADMIXTURE analysis, the island sample of Bugala was split into mainland- and island-like subpopulations ("Bugala (M)" and "Bugala (I)," respectively) for subsequent analyses (Fig. S4). (D) Heatmap of F_{ST} between sites (lower triangle) and associated z-score computed via block jackknife (upper triangle). "Bugala (M)" and "Bugala (I)" are the mainland- and island-like subpopulations of Bugala. (E) Proportion of genome-wide pairwise IBD sharing between individuals, based on the full genome. Each small square represents a comparison between two individuals, and darker colors indicate a higher proportion of the two genomes is in IBD, shaded on a logarithmic scale. Individuals are grouped by locality.

To all pairs of LVB localities we fit an isolation-with-migration (IM) demographic model 131 using $\delta a \delta i$, in which an ancestral population splits into two populations, allowing exponential 132 growth and continuous asymmetrical migration between the daughter populations (Supple-133 mentary Figs. S6, S7). In all comparisons involving islands and some between mainland 134 sites, the best fitting model as chosen via AIC had zero migration (Supplementary Tables 135 S3, S4, and S5). Time since population split was much more recent for mainland-mainland comparisons (excluding Bugala, median: 361 years) than those involving islands (island-137 island median: 7,128 years; island-mainland median: 4,043 years). Island-island split time 138 confidence intervals typically did not overlap those involving mainland sites.

Selection

As beneficial variants would be the most likely signatures of past gene flow to persist, we next examined signatures of selective sweeps for insight into migration. Identifying signatures of

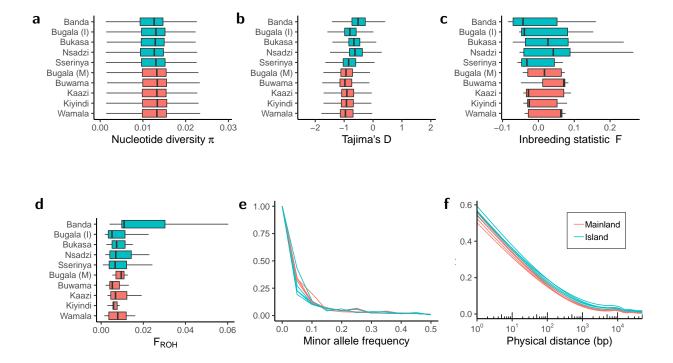
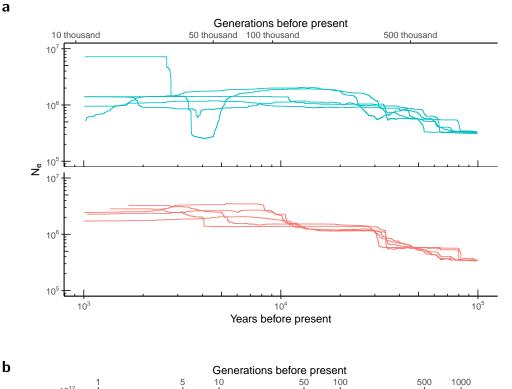


Figure 3: Diversity metrics in the Lake Victoria Basin samples. Shown are a (A) boxplot of nucleotide diversity (π ; in 10 kilobase windows), (B) boxplot of Tajima's D (in 10 kilobase windows), (C) boxplot of inbreeding statistic (F), (D) boxplot of length of runs of homozygosity (F_{ROH}), (E) histogram of Minor Allele Frequency (MAF), and (F) decay in linkage disequilibrium (r^2), all grouped by sampling locality. For all boxplots, outlier points are not shown.



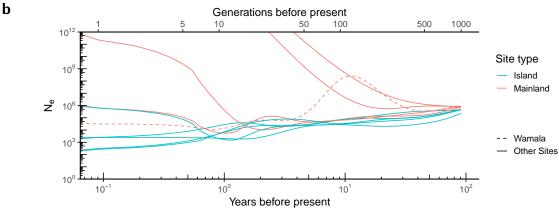


Figure 4: Population history of the Lake Victoria Basin samples.

(A) Long-term evolutionary population histories inferred via stairway plots for island and mainland samples. (B) Contemporary or short-term effective population size (N_e) history inferred using sharing of regions that are identical by descent (IBD). Wamala, a mainland locality showing island-like fluctuations in population size, is indicated with a dashed line. Plot truncated to exclude implausibly high estimates that are likely an artifact of sample size.

selection in the same genomic region in populations with independent lineages would be consistent with several scenarios [23]: (i) independent parallel selective sweeps on de novo mutations, (ii) independent parallel selective sweeps on shared ancestral variation, or (iii) selective sweeps on variants transferred via gene flow. As we were most interested in the transfer of adaptive variants for its insight into migration (iii), we distinguished between the alternative scenarios as follows.

We would expect independent sweeps on novel mutations (i) to exhibit differences in 149 genetic background between the two populations, evidenced by distinct haplotype clusters, 150 each comprising near-identical haplotypes separated by individual haplotypes lacking signa-151 tures of a selective sweep. In both other scenarios (ii-iii), we would instead expect haplotypes 152 with the sweep to group together when clustered by genetic distance. By itself, haplotype 153 information does not differentiate sweeps targeting standing ancestral variation (ii) from 154 those targeting adaptive variants spread through gene flow (iii), but additional information 155 such as geographic distance between the populations, estimates of gene flow inferred from 156 other regions of the genome, and assessment of gene flow between other nearby populations, 157 may suggest that one of these scenarios is the more likely. 158

While the sharing of a sweep may indicate migration between populations, the inverse would be suggestive—though not conclusive—of barriers to gene flow. A lack of sharing of a selective sweep signal between two populations may indicate no migration is occurring. However, it would also be consistent with the occurrence of migration that is subsequently countered by the local effects of selection or lost to genetic drift.

We first compared mainland Uganda and the Ssese Islands, reasoning that shared signatures of sweeps at a genomic location may indicate migration is occurring with the islands, while the absence was suggestive of isolation. We identified sweeps in the LVB using genome scans of between- and within-locality statistics, including F_{ST} ([24], Supplementary Fig. S8), Extended Haplotype Homozygosity (XP-EHH, [25], Supplementary Fig. S8), and haplotype homozygosity (H12, [26], Supplementary Fig. S9). To test for sweeps that were variable within the LVB, we identified locality-specific sweeps (found at only one sampling site in the LVB), sweeps that were found in our island localities but not mainland LVB localities, and sweeps that were found only in our mainland LVB localities (all defined as H12 > 99th percentile). To add additional country-level context, we then intersected these regions with those under putative selection in a mainland Ugandan reference population (H12 > 95th percentile; [4]).

Some genomic locations had heterogeneous selection signals within the LVB and within 176 Uganda, indicative of potential geographic barriers to gene flow or local variation in selective 177 regimes. Locality-specific putative sweeps were more prevalent on island than LVB mainland 178 localities (mean per locality: island = 52.4; mainland = 26.8), concordant with increased 179 isolation of the islands (Supplementary Table S6). Sweeps detected only or primarily in 180 mainland LVB localities were shared with the Ag1000G Ugandan reference population more 181 often (8 of 37; 22%) than those found only or primarily on islands (1 of 21, 5%; Supplementary 182 Tables S7 and S8), again indicative of some barriers to gene flow with the islands. 183

We next reasoned that continent-wide selective sweeps, with broadly distributed selective advantage, would be the most likely to be shared via gene flow. Widespread sweeps that were absent or at extremely low frequency on the islands would be a strong suggestion against contemporary gene flow, and those that conversely were present on the islands would be indicative that gene flow had occurred, if the alternative scenarios could be excluded as outlined above. To identify these regions, we intersected our set of sweeps with those under putative selection in populations across the continent (H12 > 95th percentile in Ag1000G; [4]).

As expected, outlier regions included known selective sweep targets from elsewhere in Africa ([4], Supplementary Table S9). All sweeps found in the reference Uganda population [4] were detected in at least some sampling localities in our LVB dataset, except the

sweep targeting Vqsc, which was likely excluded during filtration of the region adjacent to the centromere. For instance, the large genomic region spanning the cluster of insecticide 196 resistance-associated cytochrome P450s (Cyp6p) on chromosome arm 2R, including Cyp6p3197 which is upregulated in mosquitoes with permethrin and bendiocarb resistance [27], exhib-198 ited low diversity, an excess of low frequency polymorphisms (Tajima's D), and elevated 199 haplotype homozygosity (H12) within the LVB populations (Supplementary Figs. S9, S10). 200 Pairwise statistics (F_{ST} and XP-EHH) indicated low differentiation between LVB localities, 201 as expected for a continent-wide sweep (Supplementary Figs. S8). The signal was found in 202 every LVB site, including all islands. Hierarchical clustering of LVB and Ag1000G haplotypes 203 revealed clades with low inter-haplotype diversity, expected after selection rapidly increases 204 the frequency of a haplotype containing adaptive variation (Supplementary Fig. S11). Con-205 sistent with previous results [4], these clusters of closely related haplotypes on independent 206 lineages indicate that multiple parallel sweeps targeting the Cyp6p region have occurred in 207 several genetic backgrounds at numerous localities across Africa. Within Uganda, since al-208 most all mainland and island individuals carry haplotypes from a single cluster, the selected 209 haplotype of this cluster likely spread to near-fixation via gene flow. 210

In contrast, some sweeps with continent-wide prevalence including the reference Ugan-211 dan population [4] were found at all mainland LVB sites but had colonized the islands incompletely. For example, a region on chromosome arm 2L (2L:2,900,000-3,000,000) was found in all assayed Ag1000G populations and LVB mainland sites, but found on no island 214 but Sserinya (Supplementary Table S8). As in previous studies [4], independent clusters 215 of low-diversity haplotypes in varied genetic backgrounds suggest multiple sweeps targeting 216 the cluster of genes encoding glutathione S-transferases (Gste1-Gste7), including one sweep 217 specific to Uganda. This Ugandan sweep was similarly confined largely to the mainland 218 in the LVB. These sweeps at targets of selection throughout the continent that are largely 219 restricted to the mainland are suggestive of strong barriers to gene flow to the islands, either due to lack of connectivity or the countering effects of selection or drift. Other sweeps
had colonized the islands incompletely. The sweep targeting cytochrome P450 gene Cyp9k1
likely arose multiple times independently, since Ugandan haplotypes do not cluster with low
diversity clusters from elsewhere in Africa. Within the LVB, the sweep signature is found
on some, but not all islands, suggesting some barrier to gene flow or local selection limiting
the spread of the sweep.

Two regions exhibited selection signals similar in amplitude to known insecticide-related 227 loci, with elevated between-locality differentiation, low diversity, and extended homozygos-228 ity (Supplementary Figs. S8, S9, S12, and S13). The first, at 2L:34.1 Mb, contains many 229 genes, including a cluster involved in chorion formation [28] near the signal peak. Haplotype 230 clustering revealed a group of closely-related Ugandan individuals, consistent with a geo-231 graphically bounded selective sweep (Supplementary Fig. S14). The selected variation had 232 not fully colonized the islands or the LVB mainland sites, however, suggesting some barriers 233 to gene flow, loss due to drift at some localities, or local differences in selective pressure 234 within the LVB. Elsewhere in Africa, clustering analysis revealed other low-variation clades 235 in distinct genetic backgrounds in, e.g., Cameroon and Angola, suggesting parallel selection 236 on independent mutations at this locus.

The second putative sweep, at X:9.2 Mb, coincided precisely with eye-specific diacyl-238 glycerol kinase (AGAP000519, chrX:9,215,505-9,266,532). Low diversity haplotypes formed 239 a single cluster including LVB haplotypes overwhelmingly from the islands (Supplementary 240 Fig. S15). Transfer via gene flow between islands but not to the mainland is reasonable, 241 given the connectivity patterns we have inferred from neutral variation. Additionally, lo-242 cal selection may be countering the spread of the sweep to the mainland. However, more 243 surprisingly, these island haplotypes with evidence of a selective sweep were most closely 244 related to haplotypes from distant locations, primarily Gabon and Burkina Faso rather than 245 Uganda. This sharing of extended haplotypes between islands and distant localities is consistent with either gene flow or independent sweeps targeting ancestral standing variation.

Of these alternatives, extremely long distance gene flow that persists only on islands seems

less likely.

Discussion

Understanding the population genetics of island Anopheles quambiae has both evolutionary 251 and practical importance. A limited number of genetic investigations have been conducted 252 on oceanic [29–32] and lacustrine islands [33–36], though the latter have been limited in the 253 type or count of molecular markers used. In contrast to shallow population structure across 254 Africa [4, 5], partitioning of genetic variation on islands suggests varying isolation. Using 255 a genome-wide dataset, we found differentiation between the Ssese Islands to be relatively 256 high in the context of continent-wide structure, with the differentiation between Banda Island 257 (only 30 km offshore) and mainland localities on par with or higher than for populations on 258 opposite sides of the continent (e.g., Banda vs. Wamala, $F_{ST}=0.034$; mainland Uganda vs. 259 Burkina Faso, $F_{ST} = 0.007$ [4]). The Ssese Islands are approximately as differentiated as all 260 but the most outlying oceanic islands tested (e.g. mainland Tanzania vs. Comoros, 690-830 261 km apart, $F_{ST} = 0.199-0.250$ [31]). Patterns of haplotype sharing did include direct evidence for the recent exchange of migrants between nearby islands, but analyses based on haplotype 263 sharing, Bayesian clustering, and demographic reconstruction included no evidence of direct sharing between Banda and the mainland. Banda is nonetheless connected to other islands 265 and thereby indirectly connected to the mainland, and additional sampling may reveal signs 266 of admixture. Additional sampling on Banda and other islands that are disjunct from the 267 rest of the archipelago would be prudent when assessing potential field testing locations. 268 The name "Seese" derives from another arthropod vector, the tsetse fly (Glossina spp.) 269 The tsetse-mediated arrival of sleeping sickness in 1902 brought "enormous mortality" [37, 270

pp. 332 to the 20 thousand residents, who were evacuated in 1909 [37, 38]. Though encouraged to return by 1920, the human population numbered only 4 thousand in 1941 [37] 272 and took until 1980 to double [39], but has since rapidly risen to over 62 thousand (2015, 273 projected; [18, 40]). The impacts on mosquito populations of this prolonged depression in 274 human population size, coupled with water barriers to mosquito migration, are reflected in 275 the distinctive demographic histories of island An. qambiae populations, which were smaller 276 and fluctuated more than mainland localities, echoing previous results [34, 36]. Two main-277 land sites had island-like recent population histories, with Wamala abruptly switching from 278 a mainland-like to island-like growth pattern around 2005. This coincides precisely with a 279 $\geq 20\%$ reduction from 2000-2010 in the Plasmodium falciparum parasite rate (PfPR₂₋₁₀; a 280 measure of malaria transmission intensity) in Mityana, the district containing Wamala [41]. 281 Though previous Anopheles population genetic studies have inferred gene flow even 282 among species [4, 42], we inferred that no genetic exchange had occurred since the split 283 between island sites and between islands and the mainland. Island pairs were inferred to 284 have split far deeper in the past (5,000-14,000 years ago) than mainland sites (typically < 500285 years ago), on par with the inferred split time between Uganda and Kenya (approximately 286 4,000 years ago; [4]). Although bootstrapping-derived confidence intervals permit some certainty, our model fit is not optimal likely due to low sample sizes and high levels of shared ancestral variation, and additional sampling is necessary to clarify population history. Our 289 inferred lack of gene flow to the islands appears contradictory to the presence of individuals 290 who share ancestry with the mainland on all islands but Banda. We cannot dismiss the 291 possibility that this indicates actual migration occurs. If so, effects of migration would have 292 to be sufficiently countered by local selection to limit its effect on allele frequency spectra, 293 rendering effective migration (as estimated in population history inference) zero. The ap-294 parent contradiction can also be resolved if shared ancestry between islands and mainland 295 suggested by the clustering result is interpreted as retention of shared ancestral polymor-296

phism or the existence of inadequately sampled ancestral variation [43], rather than recent admixture. This interpretation is consistent with the affinity we observed between the Ssese Islands and West Africa in the structure of adaptive variation.

Discerning whether the absence of observed gene flow is due to lack of connectivity, the 300 opposition of selection, or the stochasticity of genetic drift is difficult. Instead we must rely 301 on estimates of the strength of selection in the two locales to inform our conclusions. For 302 example, we would expect that an insecticide sweep found all over Africa would spread in 303 island mosquito populations with insecticide treated bed nets, despite the considerable effect 304 of genetic drift in small populations. As insecticide treated bed net usage is present on the 305 islands [18], variation conferring a major selective advantage related to insecticides would 306 be expected to spread to and persist on the islands if migration allows the transfer, and the 307 strongest evidence of a lack of contemporary connectivity is therefore the absence of a sweep 308 on the islands that is widespread on the continent. 309

We found two sweeps on insecticide-related genes that are common targets of selection 310 elsewhere but which have incompletely colonized the Ssese Islands: one on cytochrome P450 311 monooxygenase Cyp9K1 [44, 45] present on some islands, and another on glutathione S-312 transferase genes (Gste1-Gste7; [46-49]) at extremely low frequency on the islands. That the selective sweeps targeting these loci [4] have not fully colonized the islands despite the advantage in detoxifying pyrethroids and DDT suggests a lack of contemporary exchange. 315 However, the sweep targeting the Cyp6p cluster was found on all islands, confirming past 316 gene flow had occurred at some point. Although these distributions confirm that past mi-317 gration from the mainland to islands has occurred and we are unable to exclude low levels of 318 contemporary gene flow, taken together our data are consistent with potentially high degrees 319 of isolation on contemporary timescales for some islands of the Ssese archipelago. 320

Our investigation also identified two previously unknown signatures of selection. For the first, on chromosome arm 2L and encompassing many genes, haplotypes with sweeps in

distinct genetic backgrounds across Africa suggest the region has been affected by multiple independent convergent sweeps. In Uganda, most individuals with the sweep are from the 324 mainland, suggesting a local origin and spread via short distance migration. The putative 325 target of the second sweep is diacylglycerol kinase on the X-chromosome, a homolog of 326 retinal degeneration A (rdqA) in Drosophila. The gene is highly pleiotropic, contributing to 327 signal transduction in the fly visual system [50, 51], but also olfactory [52] and auditory [53] 328 sensory processing. It has been recently implicated in nutritional homeostasis in *Drosophila* 329 [54] and is known to interact with the TOR pathway [55], which has been identified as a 330 target of ecological adaptation in *Drosophila* [56, 57] and *An. qambiae* [58]. The sweep 331 appears largely confined to island individuals in the LVB, but the cluster of haplotypes also 332 includes those from Gabon, Burkina Faso, and Kenya. Shared extended haplotypes suggest 333 a single sweep event spread by gene flow or selection on standing ancestral variation, not 334 independent selection on de novo mutations. Possible explanations include long distance 335 migration of an adaptive variant persisting on only the islands or, more reasonably, selection 336 on standing ancestral variation. We have not found obvious candidate targets of selection, 337 e.g. coding changes, which may be due to imperfect annotation of the genome or the likely 338 possibility that the target is a non-coding regulator of transcription or was filtered from our dataset. Further functional studies would be needed to clarify the selective advantage that these haplotypes confer. Interestingly, the putative sweep coincides with a similar region of low diversity in a cryptic subgroup of Anopheles qambiae sensu lato (GOUNDRY; 342 [42]), suggesting possible parallel selective events on independent mutations or adaptive 343 introgression. 344 345

Population structure investigations are paramount for informing the design and deployment of control strategies, including field trials of transgenic mosquitoes. We demonstrate alternatives to simple extrapolation of migration rates from differentiation, which is fraught particularly given the assumption of equilibrium between the evolutionary forces of migration and drift [59–61], an unlikely state for huge An. gambiae populations [3]. We suggest that future assessments of connectivity include, as we have, the spatial distribution of adaptive variation, identification of recent migrants via haplotype sharing, and demographic history modeling, from which we have inferred the Ssese Islands to be relatively isolated on contemporary time scales. Though we cannot exclude the possibility of a small amount of gene flow over evolutionary time between our most isolated islands and the mainland, the data are consistent with a sufficiently low amount of gene flow that it becomes reasonable to consider these islands as isolated on short time frames.

A completely isolated population of mosquitoes is not a reasonable expectation given 357 mosquitoes' propensity for active and even passive (human-aided or windborne) dispersal 358 [16], potentially up to hundreds of kilometers [11]. Although no island, lacustrine or oceanic, 359 is completely isolated, such localities may still be ideal for initial gene drive field testing, 360 as the geographical barriers maximize isolation to the extent possible [16], and absolute 361 isolation on evolutionary timescales is unnecessary given the relatively short timeframe of 362 small-scale field tests. Thus, the probability of contemporary migration may be sufficiently 363 low to qualify some Ssese Islands as candidate field sites. Additionally, the assessment of the islands' suitability as potential sites for field trials of genetically modified mosquitoes must also consider the logistical ease of access and monitoring that the bounded geography of a small lacustrine island with low human population density affords initial field tests. Due 367 consideration should be provided to these characteristics of small lake islands that may be 368 appealing to regulators, field scientists, local communities, and other stakeholders. Given 369 such features and the probable rarity of migration, the Ssese Islands may be logical and 370 tractable candidates for initial field tests of genetically modified An. qambiae mosquitoes, 371 warranting further entomological study. 372

3 Materials and Methods

Experimental design Mosquitoes were sampled from 5 of the Ssese Islands in Lake Victoria, Uganda (Banda, Bukasa, Bugala, Nsadzi, and Sserinya) and 4 mainland sampling localities (Buwama, Kaazi, Kiyindi, and Wamala) at varying distances from the lake in May and June, 2015. Sampling took place between 4:40 and 8:15 over a 30 day period as follows: Indoor resting mosquitoes were collected from residences via mouth or mechanical aspirators and subsequently identified morphologically to species group. Female mosquitoes assigned to the An. gambiae sensu lato complex based on morphology (N=575) were included in further analyses. All mosquitoes were preserved with silica desiccant and transported to the University of Notre Dame, Indiana, U.S.A. for analysis.

DNA extraction, Library preparation, and Whole Genome Sequencing Animals
were assigned to species level via a PCR-based assay [62] using DNA present in a single leg
or wing. DNA from individual An. gambiae s. s. N=116 mosquitoes was extracted from
the whole body via phenol-chloroform extraction [63] and then quantified via fluorometry
(PicoGreen). Automated library preparation took place at the NYU Langone Medical Center
with the Biomek SPRIWorks HT system using KAPA Library Preparation Kits, and libraries
were sequenced on the Illumina HiSeq 2500 with 100 paired end cycles.

Mapping and SNP calling, filtering Software version information is provided in Supplementary Table S10. After quality filtering and trimming using ea-utils' fastq-mcf (-l 15
-q 15 -w 4; [64]), reads were mapped to the An. gambiae reference genome (AgamP4 PEST;
[65, 66]) using BWA aln and sampe with default parameters [67].

After realignment around indels with GATK's IndelRealigner, variants were called using
GATK's UnifiedGenotyper (with -stand_call_conf 50.0 and -stand_emit_conf 10.0; selected to
be consistent with methods of recent comparison SNP dataset [4]) and filtered for quality [68],

excluding SNPs with QualByDepth (QD) < 2.0, RMSMappingQuality (MQ) < 40.0, Fisher-Strand (FS) > 60.0, HaplotypeScore > 13.0, or ReadPosRankSum < -8.0. All bioinformatic 398 steps for read mapping and variant identification are encapsulated in the NGS-map pipeline 399 (https://github.com/bergeycm/NGS-map). This yielded 33.1 million SNPs. Individuals 400 and variants with high levels of missingness (> 10%) and variants that were not biallelic 401 or exhibited values of HWE that were likely due to sequencing error (p < 0.00001) were 402 excluded from further analysis. For use in population structure inference, the SNP dataset 403 was further pruned for linkage disequilibrium by sliding a window of 50 SNPs across the 404 genome in 5 SNP increments and recursively removing random SNPs in pairs with $r^2 > 0.5$ 405 using PLINK [69, 70]. After filtration, the dataset contained 28,569,621 SNPs before LD 406 pruning and 115 individuals. SNPs unpruned for linkage disequilibrium were phased with 407 SHAPEIT2 [71] using an effective population size (N_e) of 1,000,000 (consistent with pre-408 vious demographic modeling [4]), default MCMC parameters (7 burn-in MCMC iterations, 409 8 pruning iterations, and 20 main iterations), conditioning states for haplotype estimation 410 (K = 100), and window size of 2 Mb. 411

Population structure inference To explore population structure in a larger, continentwide context, we merged our LVB SNP set with a recently published dataset of Anopheles
gambiae individuals (from the Ag1000G project) [4]. Prior to filtering, biallelic SNPs from
the LVB and Ag1000G datasets were merged using bcftools [72]. We excluded any SNP with
greater than 10% missingness in either dataset, any SNPs that did not pass the accessibility
filter of the Ag1000G dataset, and SNPs with MAF < 1%. After this filtration, our merged
SNP dataset contained 12,537,007 SNPs.

After pruning the merged dataset for LD (leaving 9,861,756 SNPs) and excluding laboratory crosses (leaving 881 individuals), we assigned individuals' genomes to ancestry components using ADMIXTURE [20]. We created 10 replicate samples of 100,000 SNPs from

chromosome 3 (prior to LD-pruning), including only biallelic SNPs in euchromatic regions with MAF > 1%. These replicate datasets were pruned for LD by randomly selecting from 423 pairs of SNPs with $r^2 > 0.01$ in sliding windows of size 500 SNPs and with a stepsize of 424 250 SNPs. For each replicate, we ran ADMIXTURE for 5 iterations in five-fold cross val-425 idation mode for values of k from 2 to 10. This resulted in 50 estimates for each value of 426 k. We assessed these results using the online version of CLUMPAK with default settings to 427 ensure the stability of the resulting clustering [73]. CLUMPAK clusters the replicate runs' 428 Q-matrices to produce a major cluster for each value of k, which we then visualized. The 429 lowest cross-validation error was found for k=6 clusters, but we also display ancestry esti-430 mates with k=9 clusters to further explore patterns of structure with a level of subdivision 431 at which the Ssese Island individuals are assigned a unique ancestry component. 432

We visualized population structure via principal components analysis (PCA) with PLINK [69, 70], using the LVB-Ag1000G merged dataset (excluding the outlier Kenyan population; [4]) and 3,212,485 chromosome 3 SNPs (to avoid the well-known inversions on chromosome 2 and the X-chromosome) outside of heterochromatic regions (such as centromeric regions; [66]; Supplementary Table S11). We next performed a PCA on the LVB dataset alone, pruning for LD and low-MAF (< 1%) SNPs on chromosome 3. Based on the results of this analyses, we split individuals from the large island of Bugala into two clusters for subsequent analyses: those that cluster with mainland individuals and those that cluster with individuals from the smaller islands.

We computed the pairwise fixation index (F_{ST}) between locality samples for An. gambiae using the unbiased estimator of Hudson [74] as implemented in smartpca [75, 76]. To obtain overall values between sampling sites, per-SNP values were averaged across the genome excluding known inversions (2La, 2Rb, and 2Rc) and heterochromatic regions. We also computed z-scores via block jackknife, using 42 blocks of size 5 Mb. We tested for isolation by distance, or a correlation between genetic and geographic distances, with a Mantel test

[77] as implemented in the R package ade4 [78], using these F_{ST} estimates and Euclidean geographic distances between localities.

To estimate fine-scale structure and relatedness between individuals, we estimated the proportion of pairs of individuals genomes that are identical by descent (IBD) using PLINK [69, 70]. We excluded heterochromatic and inversion regions, and retained informative pairs of SNPs within 500 kb in the pairwise population concordance test.

Diversity estimation Grouping individuals by site (except for Bugala, which was split 454 based on the results of the PCA), we calculated nucleotide diversity (π) and Tajima's D 455 in nonoverlapping windows of size 10 kb, the inbreeding coefficient (F) estimated with the 456 method of moments, minor allele frequencies (the site frequency spectrum, SFS), and a mea-457 sure of linkage disequilibrium (r^2) using VCFtools (Danecek2011). For r^2 , we computed the measure for all SNPs (unpruned for linkage) within 50 kb of a random set of 100 SNPs 459 with MAF > 10\% and corrected for differences in sample size by subtracting 1/n, where n 460 equaled the number of sampled chromosomes per site. To visualize decay in LD, we plotted 461 r^2 between SNPs against their physical distance in base pairs, first smoothing the data by fitting a generalized additive model (GAM) to them. We also inferred runs of homozygosity using PLINK [69, 70] to compare their length (F_{ROH}) , requiring 10 homozygous SNPs 464 spanning a distance of 100 kb and allowing for 3 heterozygous and 5 missing SNPs in the 465 window. Runs of homozygosity were inferred using LD-pruned SNPs outside of inversions or 466 heterochromatic regions. We tested the significance of differences in these statistics between 467 island and mainland categories using a two-sided Wilcoxon rank sum test. 468

Demographic history inference To estimate the contemporary or short-term N_e for each site, we inferred regions of IBD from unphased data with IBDseq [79] and analyzed them with IBDNe [22]. We restricted our analysis to SNPs from chromosome 3 to avoid inverted regions. We allowed a minimum IBD tract length of 0.005 cM (or 5 kb), scaling

it down from the recommended length for human genomes due to mosquitoes' high level of heterozygosity [4] and assumed a constant recombination rate of 2.0 cM/Mb (after [80]).

To estimate the long-term evolutionary demographic history of mosquitoes on and near the Ssese Islands, including a long-term estimate of N_e [81], we inferred population demographic history for each site via stairway plots using the full site frequency spectra based on SNPs on chromosome 3 with heterochromatic regions and regions within 5 kb of a gene excluded [21].

We also inferred a "two-population" isolation-with-migration (IM) demographic model 480 with $\delta a \delta i$ [82, 83] in which the ancestral population splits to form two daughter populations 481 that are allowed to grow exponentially and exchange migrants asymmetrically, as described 482 in the main text. For $\delta a \delta i$ -based analyses, we used the full dataset of SNPs on chromosome 483 3, not pruned for LD but with heterochromatic regions and regions within 5 kb of a gene 484 masked. We polarized the SNPs using outgroup information from Anopheles merus and An. 485 merus [84]. We fit this two-population model and the same model without migration to all 486 pairs of locality samples, choosing the optimal model using the Godambe Information Matrix 487 and an adjusted likelihood ratio test to compare the two nested models. We compared the 488 test statistic to a χ^2 distribution and rejected the null model if the p-value for the test statistic was < 0.05. For both, singletons and doubletons private to one population were masked from the analysis and a parameter encompassing genotype uncertainty was included in the models 491 and found to be low (mean = 0.0.70%). We assessed the goodness-of-fit visually using the 492 residuals of the comparison between model and data frequency spectra (Supplementary Fig. 493 S7). Using the site frequency spectrum, we projected down to 2-6 fewer chromosomes than 494 the total for the smaller population to maximize information given missing data. We set the 495 grid points to $\{n, n+10, n+20\}$, where n= the number of chromosomes. Bounds for N_e 496 scalars were $\nu \in (0.01, 10, 000)$, for time were $T \in (1e-8, 0.1)$, for migration were $m \in (1e-8, 0.1)$ 497 8, 10), and for genotyping uncertainty were $p_{misid} \in (1e-8, 1)$. Parameters were perturbed

before allowing up to 1000 iterations for optimization. We estimated parameter uncertainty using the Fisher information matrix and 100 bootstrap replicates of 1 Mb from the dataset. If the Hessian was found to be not invertible when computing the Fisher information matrix, the results of that iteration were excluded from the analysis. For population size change parameters, ν , optimized values for one or both populations were often close to the upper limit. Due to this runaway behavior, common in analyses of the SFS [85], we excluded the population size change from our interpretation.

To translate $\delta a \delta i$ - and stairway plot-based estimates of N_e and time to individuals and years respectively, we assumed a generation time of 11 per year and a mutation rate of 3.5e-9 per generation [4].

Selection inference To infer candidate genes and regions with selection histories that varied geographically, we compared allele frequencies and haplotype diversity between the 510 sampling sites. To infer differing selection between sampling sites, we computed F_{ST} between 511 all populations in windows of size 10 kb using the estimator of Weir and Cockerham [24] 512 (as implemented in VCFtools [86]), and H12 (as implemented in SelectionHapStats [26]) 513 and XP-EHH on a per-site basis (as implemented in selscan [87]) to detect long stretches of 514 homozygosity in a given population considered alone or relative to another population [25]. 515 For XP-EHH, EHH was calculated in windows of size 100 kb in each direction from core 516 SNPs, allowing EHH decay curves to extend up to 1 Mb from the core, and SNPs with MAF 517 < 0.05 were excluded from consideration as a core SNP. As we lacked a fine-scale genetic map 518 for Anopheles, we assumed a constant recombination rate of 2.0 cM/Mb (after [80]). Scores 519 were normalized within chromosomal arms and the X-chromosome. The between-locality 520 statistics, F_{ST} and XP-EHH, were summarized using the composite selection score [CSS; 521 [88, 89]]. 522

We plotted these statistics across the genome to identify candidate regions with signa-

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tures of selection, including high differentiation between samples from different localities, reduced variability within a sample, and extended haplotype homozygosity. To identify re-525 gions of the genome showing signatures of selection specific to certain geographic areas, we 526 identified genomic regions with elevated H12 in a subset of localities, and confirmed both ele-527 vated differentiation (as inferred from F_{ST}) and evidence of differing selective sweep histories 528 (as inferred from XP-EHH). Excluding the mainland-like portion of Bugala, we identified 520 putative locality-specific sweeps (H12 over 99th percentile in one population), island-specific 530 sweeps (H12 over 99th percentile in 4 or more of the 5 island localities but 0 or 1 mainland 531 localities), or LVB mainland-specific sweeps (H12 over 99th percentile in 3 or more of the 4 532 island localities but 0 or 1 island localities). To place these putative sweeps in their continen-533 tal context, for the region of each putative locality-, island-, or LVB mainland-specific sweep, 534 we determined if the H12 values of each of the Ag1000G populations (excluding Kenya due 535 to its signatures of admixture and recent population decline; [4]) were in the top 5% for that 536 population, indicating a possible selective sweep at the same location. 537

We further explored the haplotype structure and putative functional impact of loci for 538 which we detected signatures of potential selection to determine the count and geographic 539 distribution of independent selective sweeps. To provide necessary context for the reconstruction of sweeps and quantify long distance haplotype sharing between populations, we included data from several other An. qambiae populations across Africa (Burkina Faso, Cameroon, Gabon, Guinea, Guinea-Bissau, Kenya, and other Ugandan individuals; [4]). We 543 computed the pairwise distance matrix as the raw number of base pairs that differed and 544 grouped haplotypes via hierarchical clustering analysis (implemented in the hclust R func-545 tion) in regions of size 100 kb centered on each peak in pairwise F_{ST} or XP-EHH, or the 546 average of peaks, in the case for multiple nearby spikes. As short terminal branches can 547 result from a beneficial allele and linked variants rising to fixation during a recent selective 548 sweep, we identified such clusters by cutting the tree at a height of 0.4 SNP differences per 550 kb.

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Data and materials availability

- All scripts used in the analysis are available at https://github.com/bergeycm/Anopheles
- 24 _gambiae_structure_LVB and released under the GNU General Public License v3. Sequenc-
- ing read data for the LVB individuals are deposited in the NCBI Short Read Archive (SRA)
- under BioProject accession PRJNA493853.

827 Author Contributions

- 828 C.M.B., J.K.K., and N.J.B. designed the study; C.M.B., M.L., R.M.W., and J.K.K. col-
- lected biological samples; C.M.B. analyzed the data; C.M.B., M.C.F., and N.J.B. wrote the
- manuscript; M.C.F., J.K.K., and N.J.B. supervised the research; C.M.B., M.L., R.M.W.,
- M.C.F., J.K.K., and N.J.B. edited the manuscript.

S22 Conflict of Interest Statement

The authors declare no competing financial interests.

Supplemental Material for: Assessing connectivity despite high diversity in island populations of a malaria mosquito

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Christina M. Bergey, Martin Lukindu, Rachel M. Wiltshire, Michael C. Fontaine, Jonathan

K. Kayondo, and Nora J. Besansky

Tables

Table S1: Sampling sites and coordinates.

Location	Latitude	Longitude	Sample Count
Banda	-0.25893	32.39594	11
Bugala - Bugoma	-0.26697	32.07936	11
Bugala - Lutoboka	-0.31624	32.29246	7
Bugala - Mweena	-0.32806	32.31113	5
Bukasa	-0.48609	32.45091	11
Buwama	0.02077	32.10574	11
Kaazi	-0.31831	31.88183	11
Kiyindi	0.27558	33.14699	10
Nsadzi	-0.08632	32.58895	11
Sserinya	-0.26476	32.37228	16
Wamala	0.40811	31.99609	11
		·	<u> </u>

Table S2: List of individuals included in study with mean depth of sequencing coverage.

ID	Field ID	Island	Site	Mean depth
LVB2015-1	CM-KSB-J5	Nsadzi	Kansambwe	20.40
LVB2015-2	K-KSB-E1	Nsadzi	Kansambwe	24.50
LVB2015-3	RM- KSB - $G1$	Nsadzi	Kansambwe	17.90
LVB2015-4	NKG-F-G3	Bukasa	Nakibanga	4.90
LVB2015-6	NKG-F-H1	Bukasa	Nakibanga	15.80
LVB2015-7	NKG-K-I1	Bukasa	Nakibanga	15.80
LVB2015-8	NKG-K-K1	Bukasa	Nakibanga	19.90
LVB2015-9	NKG-M-C1	Bukasa	Nakibanga	22.30
LVB2015-10	NKG-M-D1	Bukasa	Nakibanga	20.30
LVB2015-11	NKG-M-F1	Bukasa	Nakibanga	23.90
LVB2015-14	MWN-K-A1	Bugala	Mweena	21.20
LVB2015-15	MWN-K-C2	Bugala	Mweena	18.30
LVB2015-16	MWN-P-D1	Bugala	Mweena	5.34
LVB2015-17	MWN-R-E1	Bugala	Mweena	22.60
LVB2015-18	MWN-R-F1	Bugala	Mweena	17.10
LVB2015-19	BDA-K-B1	Banda	Banda	14.40
LVB2015-20	BDA-K-B2	Banda	Banda	18.00

LVB2015-21	BBS-C-M1	Sserinya	Bbosa	19.70
LVB2015-22	BBS-F-F1	Sserinya	Bbosa	20.80
LVB2015-24	BBS-K-J3	Sserinya	Bbosa	22.80
LVB2015-25	BBS-K-J8	Sserinya	Bbosa	17.30
LVB2015-26	BBS-K-K2	Sserinya	Bbosa	18.20
LVB2015-27	BBS-M-L1	Sserinya	Bbosa	22.40
LVB2015-28	BBS-P-I4	Sserinya	Bbosa	23.30
LVB2015-29	BBS-R-A2	Sserinya	Bbosa	19.60
LVB2015-30	BBS-R-C1	Sserinya	Bbosa	16.10
LVB2015-32	KSS-F-E2	Sserinya	Kasisa	21.00
LVB2015-33	LBK-C-F1	Bugala	Lutoboka	21.00
LVB2015-34	LBK-C-F6	Bugala	Lutoboka	18.60
LVB2015-35	LBK-C-G6	Bugala	Lutoboka	20.90
LVB2015-36	LBK-K-E2	Bugala	Lutoboka	22.00
LVB2015-37	LBK-M-A1	Bugala	Lutoboka	18.50
LVB2015-39	LBK-R-O1	Bugala	Lutoboka	20.20
LVB2015-42	BGM-F-D1	Bugala	Bugoma	16.70
LVB2015-43	$\operatorname{BGM-F-E2}$	Bugala	Bugoma	24.50
LVB2015-45	$\operatorname{BGM-K-M2}$	Bugala	Bugoma	23.80
LVB2015-46	BGM-M-G1	Bugala	Bugoma	18.40
LVB2015-47	$\operatorname{BGM-M-H2}$	Bugala	Bugoma	14.10
LVB2015-48	BGM-M-J1	Bugala	Bugoma	20.90
LVB2015-50	BGM-P-F9	Bugala	Bugoma	18.90
LVB2015-51	$\operatorname{BGM-R-O2}$	Bugala	Bugoma	16.80
LVB2015-52	KZI-F-F001	Kaazi	Nabugabo	19.10
LVB2015-53	KZI- F - $G001$	Kaazi	Nabugabo	19.50
LVB2015-54	KZI-F-H001	Kaazi	Nabugabo	19.40
${\rm LVB}2015\text{-}55$	KZI-P-A001	Kaazi	Nabugabo	10.30
LVB2015-56	KZI-P-B005	Kaazi	Nabugabo	16.50
LVB2015-59	KZI-R-C003	Kaazi	Nabugabo	18.10
LVB2015-60	KZI-R-D007	Kaazi	Nabugabo	15.90
LVB2015-61	BWM-C-G001	Buwama	Buwama	16.10
LVB2015-62	BWM-C-H001	Buwama	Buwama	11.10
LVB2015-63	BWM-F-A001	Buwama	Buwama	20.40
LVB2015-64	BWM-F-B001	Buwama	Buwama	21.50
LVB2015-65	BWM-P-J001	Buwama	Buwama	14.80
LVB2015-66	BWM-R-C002	Buwama	Buwama	19.10
		•		

LVB2015-67	$\operatorname{BWM-R-F005}$	Buwama	Buwama	22.30
LVB2015-68	NMA-C-E003	Wamala	Naama	20.30
LVB2015-69	NMA-C-F002	Wamala	Naama	17.80
LVB2015-70	NMA-F-A001	Wamala	Naama	13.10
LVB2015-71	NMA-K-B001	Wamala	Naama	22.10
LVB2015-72	NMA-K-C002	Wamala	Naama	18.00
LVB2015-73	NMA-P-G001	Wamala	Naama	18.20
LVB2015-74	NMA-P-H003	Wamala	Naama	16.60
LVB2015-76	KYD-C-G001	Kiyindi	Kiyindi	16.10
LVB2015-77	KYD-C-H001	Kiyindi	Kiyindi	11.80
LVB2015-78	KYD-C-I001	Kiyindi	Kiyindi	16.40
LVB2015-79	$\mathrm{KYD}\text{-}\mathrm{C}\text{-}\mathrm{J}002$	Kiyindi	Kiyindi	11.50
LVB2015-80	KYD- F - $A003$	Kiyindi	Kiyindi	10.30
LVB2015-81	KYD-F-B004	Kiyindi	Kiyindi	21.50
LVB2015-82	KYD-K-D002	Kiyindi	Kiyindi	18.40
LVB2015-84	KYD-R-K001	Kiyindi	Kiyindi	16.80
LVB2015-89	BDA-K-E2	Banda	Banda	15.10
LVB2015-90	BDA-K-F1	Banda	Banda	25.10
LVB2015-91	BDA-M-N1	Banda	Banda	25.60
LVB2015-92	BDA-M-O4	Banda	Banda	17.60
LVB2015-93	BDA-M-Q1	Banda	Banda	39.20
LVB2015-96	CM-KSB-J2	Nsadzi	Kansambwe	9.22
LVB2015-97	CM- KSB - $J3$	Nsadzi	Kansambwe	10.10
LVB2015-98	CM- KSB - $J6$	Nsadzi	Kansambwe	16.90
LVB2015-100	K-KSB-D1	Nsadzi	Kansambwe	6.05
LVB2015-101	ML- KSB - $M1$	Nsadzi	Kansambwe	4.27
LVB2015-102	$\operatorname{ML-KSB-M2}$	Nsadzi	Kansambwe	19.90
LVB2015-103	RM- KSB - $G2$	Nsadzi	Kansambwe	14.20
LVB2015-104	RM- KSB - $G3$	Nsadzi	Kansambwe	17.50
LVB2015-105	NKG-R-A12	Bukasa	Nakibanga	15.30
LVB2015-106	NKG-C-E1	Bukasa	Nakibanga	16.20
LVB2015-108	NKG-K-C5	Bukasa	Nakibanga	18.50
LVB2015-109	NKG-M-A1	Bukasa	Nakibanga	12.80
LVB2015-112	BDA-K-D4	Banda	Banda	12.70
LVB2015-113	BDA-K-E3	Banda	Banda	12.20
LVB2015-114	BDA-M-N5	Banda	Banda	15.00
LVB2015-115	BDA-M-P1	Banda	Banda	16.80

LVB2015-116	BBS-C-M3	Sserinya	Bbosa	16.60
LVB2015-117	BBS-K-J1	Sserinya	Bbosa	18.80
LVB2015-118	BBS-K-J11	Sserinya	Bbosa	14.60
LVB2015-120	BBS-K-K6	Sserinya	Bbosa	18.10
LVB2015-121	BBS-P-I8	Sserinya	Bbosa	15.00
LVB2015-122	BBS-R-A19	Sserinya	Bbosa	15.50
LVB2015-125	LBK-R-A5	Bugala	Lutoboka	18.10
LVB2015-126	BGM-K-K1	Bugala	Bugoma	15.20
LVB2015-128	BGM-M-H4	Bugala	Bugoma	20.30
LVB2015-129	BGM-P-F4	Bugala	Bugoma	19.00
LVB2015-130	KZI- F - $G005$	Kaazi	Nabugabo	18.60
LVB2015-131	KZI-P-A007	Kaazi	Nabugabo	15.80
LVB2015-132	KZI-R-C012	Kaazi	Nabugabo	15.10
LVB2015-133	KZI-R-E011	Kaazi	Nabugabo	16.30
LVB2015-134	$\operatorname{BWM-P-I001}$	Buwama	Buwama	18.20
LVB2015-135	BWM-P-K002	Buwama	Buwama	19.30
LVB2015-136	BWM-R-D001	Buwama	Buwama	14.40
LVB2015-137	BWM-R-F002	Buwama	Buwama	19.90
LVB2015-138	NMA-C-E006	Wamala	Naama	21.90
LVB2015-139	NMA-C-F003	Wamala	Naama	20.40
LVB2015-140	NMA-P-G003	Wamala	Naama	18.90
LVB2015-141	NMA-R-I001	Wamala	Naama	14.10
LVB2015-142	KYD-F-B006	Kiyindi	Kiyindi	18.10
LVB2015-143	KYD-K-E003	Kiyindi	Kiyindi	14.20

Table S3: Results of two population demographic inference with IM model in $\delta a \delta i$ when comparing island to island localities. Numbers in parentheses are bounds of 95% confidence interval computed using Fisher information matrix and 100 bootstrap replicates of 1 Mb from the dataset.

Localities	N_a	% Pop. 1 in Split	Pop. 1 ν_F	Pop. 2 ν_F	Time since split	m_{12}	m_{21}
Banda - Bugala (I)	531,000	0.603	1.94	9,800	3,290	None	None
	(530,000, 532,000)	(0.596, 0.609)	(1.8, 2.09)	(7,060, 12,500)	(3,190, 3,390)		
Banda - Bukasa	526,000	0.568	14.7	9,080	7,580	None	None
	(525,000, 527,000)	(0.556, 0.581)	(13.7, 15.6)	(7,240, 10,900)	(7,470, 7,690)		
Banda - Nsadzi	527,000	0.518	47.1	9,880	9,340	None	None
	(526,000, 528,000)	(0.502, 0.534)	(39.9, 54.4)	(7,310, 12,400)	(9,160, 9,530)		
Banda - Sserinya	531,000	0.489	10.5	9,840	4,430	None	None
	(530,000, 532,000)	(0.464, 0.514)	(8.69, 12.4)	(7,390, 12,300)	(4,250, 4,610)		
Bugala (I) - Bukasa	527,000	0.49	9,840	536	5,290	None	None
	(526,000, 528,000)	(0.471, 0.509)	(7,850, 11,800)	(447, 624)	(5,170,5,410)		
Bugala (I) - Nsadzi	526,000	0.56	8,980	128	6,680	None	None
	(525,000, 527,000)	(0.536, 0.583)	(6,920, 11,000)	(110, 146)	(6,440, 6,910)		
Bugala (I) - Sserinya	530,000	0.61	5,850	49	2,130	None	None
	(529,000, 531,000)	(0.571, 0.65)	(3,610, 8,090)	(38.3, 59.6)	(1,950, 2,300)		
Bukasa - Nsadzi	527,000	0.499	3,420	335	9,340	None	None
	(526,000, 528,000)	(0.49, 0.507)	(2,860, 3,980)	(293, 377)	(9,190, 9,500)		
Bukasa - Sserinya	525,000	0.503	9,840	6,760	8,930	None	None
	(524,000, 526,000)	(0.495, 0.51)	(7,490, 12,200)	(5,630, 7,880)	(8,780, 9,080)		
Nsadzi - Sserinya	540,000	0.538	893	9,900	9,540	None	None
	(539,000, 541,000)	(0.521, 0.554)	(612, 1,170)	(7,810, 12,000)	(9,280, 9,790)		

Table S4: Results of two population demographic inference with IM model in $\delta a \delta i$ when comparing island to mainland localities. Numbers in parentheses are bounds of 95% confidence interval computed using Fisher information matrix and 100 bootstrap replicates of 1 Mb from the dataset.

Localities	N_a	% Pop. 1 in Split	Pop. 1 ν_F	Pop. 2 ν_F	Time since split	m_{12}	m_{21}
Banda - Bugala (M)	522,000	0.599	3.83	9,900	5,400	None	None
	(521,000, 523,000)	(0.586, 0.612)	(3.59, 4.07)	(8,410, 11,400)	(5,300, 5,500)		
Banda - Buwama	522,000	0.51	1.09	8,520	3,040	None	None
	(521,000, 523,000)	(0.491, 0.529)	(1.01, 1.16)	(6,830, 10,200)	(2,940, 3,150)		
Banda - Kaazi	522,000	0.563	3.99	9,960	5,890	None	None
	(521,000, 523,000)	(0.55, 0.575)	(3.71, 4.28)	(8,370, 11,500)	(5,760, 6,030)		
Banda - Kiyindi	510,000	0.568	1.72	9,910	3,910	None	None
	(509,000, 511,000)	(0.56, 0.577)	(1.69, 1.75)	(8,170, 11,600)	(3,880, 3,950)		
Banda - Wamala	521,000	0.562	3.9	7,970	5,510	None	None
	(520,000, 522,000)	(0.554, 0.57)	(3.66, 4.15)	(6,790, 9,140)	(5,390, 5,620)		
Bugala (I) - Bugala (M)	522,000	0.555	1,440	9,710	4,170	None	None
	(521,000, 523,000)	(0.543, 0.566)	(954, 1,930)	(7,420, 12,000)	(4,040, 4,300)		
Bugala (I) - Buwama	523,000	0.499	0.215	130	190	None	None
	(522,000, 524,000)	(0.496, 0.503)	(0.209, 0.222)	(2.12, 258)	(188, 192)		
Bugala (I) - Kaazi	523,000	0.366	1,420	8,510	5,060	None	None
	(522,000, 524,000)	(0.348, 0.384)	(1,210, 1,620)	(6,610, 10,400)	(4,890, 5,230)		
Bugala (I) - Kiyindi	508,000	0.363	1,350	5,910	3,580	None	None
	(507,000, 509,000)	(0.342, 0.383)	(1,070, 1,620)	(4,470, 7,350)	(3,420, 3,740)		
Bugala (I) - Wamala	520,000	0.486	2,060	9,270	3,700	None	None
	(519,000, 521,000)	(0.462, 0.51)	(1,610, 2,500)	(6,790, 11,800)	(3,570, 3,830)		
Bugala (M) - Bukasa	521,000	0.53	9,000	66.3	4,830	None	None
	(520,000, 522,000)	(0.508, 0.552)	(6,720, 11,300)	(52.2, 80.5)	(4,670, 4,990)		
Bugala (M) - Nsadzi	535,000	0.439	9,220	7.59	4,500	None	None

	(534,000, 536,000)	(0.423, 0.455)	(7,500, 10,900)	(6.91, 8.26)	(4,390, 4,620)		
Bugala (M) - Sserinya	534,000	0.516	9,590	48.8	4,040	None	None
	(533,000, 535,000)	(0.497, 0.535)	(7,640, 11,500)	(42.4, 55.2)	(3,930, 4,160)		
Bukasa - Buwama	522,000	0.501	1.43	9,960	1,690	None	None
	(521,000, 523,000)	(0.496, 0.506)	(1.4, 1.46)	(7,250, 12,700)	(1,670, 1,710)		
Bukasa - Kaazi	522,000	0.501	41.5	9,410	5,490	None	None
	(521,000, 522,000)	(0.491, 0.51)	(33.8, 49.3)	(6,990, 11,800)	(5,320, 5,650)		
Bukasa - Kiyindi	508,000	0.361	18.6	9,220	3,360	None	None
	(507,000, 509,000)	(0.336, 0.386)	(15.8, 21.4)	(6,790, 11,600)	(3,190, 3,540)		
Bukasa - Wamala	520,000	0.39	304	8,060	5,200	None	None
	(519,000, 521,000)	(0.372, 0.409)	(257, 351)	(6,270, 9,850)	(5,050, 5,360)		
Buwama - Nsadzi	524,000	0.54	9,810	1.25	1,930	None	None
	(523,000, 525,000)	(0.502, 0.578)	(6,950, 12,700)	(1.07, 1.44)	(1,800, 2,070)		
Buwama - Sserinya	523,000	0.493	54.2	0.137	187	None	None
	(522,000, 524,000)	(0.488, 0.498)	(24.2, 84.2)	(0.134, 0.14)	(186, 189)		
Kaazi - Nsadzi	524,000	0.489	9,540	19.4	5,730	None	None
	(523,000, 525,000)	(0.47, 0.507)	(7,690, 11,400)	(16, 22.8)	(5,540, 5,910)		
Kaazi - Sserinya	523,000	0.598	6,980	104	4,550	None	None
	(522,000, 524,000)	(0.579, 0.618)	(5,440, 8,510)	(89.7, 119)	(4,410, 4,680)		
Kiyindi - Nsadzi	511,000	0.499	9,940	2.62	2,840	None	None
	(510,000, 512,000)	(0.491, 0.507)	(7,340, 12,500)	(2.26, 2.98)	(2,690, 2,990)		
Kiyindi - Sserinya	509,000	0.5	1,870	0.129	185	None	None
	(508,000, 510,000)	(0.496, 0.503)	(-5,010, 8,750)	(0.126, 0.132)	(183, 187)		
Wamala - Nsadzi	523,000	0.473	8,490	9.72	4,420	None	None
	(522,000, 524,000)	(0.454, 0.492)	(6,720, 10,200)	(8.57, 10.9)	(4,290, 4,560)		
Wamala - Sserinya	521,000	0.649	9,840	39.1	2,790	None	None
	(520,000, 522,000)	(0.643, 0.655)	(7,070, 12,600)	(38.1, 40.1)	(2,760, 2,820)		

Table S5: Results of two population demographic inference with IM model in $\delta a \delta i$ when comparing mainland to mainland localities. Numbers in parentheses are bounds of 95% confidence interval computed using Fisher information matrix and 100 bootstrap replicates of 1 Mb from the dataset.

Localities	N_a	% Pop. 1 in Split	Pop. 1 ν_F	Pop. 2 ν_F	Time since split	m_{12}	m_{21}
Bugala (M) - Buwama	523,000	0.498	9,590	1,090	368	None	None
	(523,000, 524,000)	(0.49, 0.506)	(802, 18,400)	(573, 1,620)	(320, 416)		
Bugala (M) - Kaazi	521,000	0.483	7,570	8,450	2,710	None	None
	(520,000, 522,000)	(0.458, 0.509)	(5,740, 9,400)	(5,860, 11,000)	(2,590, 2,830)		
Bugala (M) - Kiyindi	507,000	0.361	1.88	3,870	355	0.0000169	0
	(506,000, 508,000)	(0.318, 0.404)	(1.74, 2.01)	(1,810, 5,920)	(348, 362)	(-14.3, 14.3)	(-31,100, 31,100)
Bugala (M) - Wamala	521,000	0.51	13.1	323	186	0	0.00226
	(520,000, 521,000)	(0.491, 0.53)	(9.03, 17.1)	(-168, 814)	(179, 194)	(-162, 162)	(-3,980, 3,980)
Buwama - Kaazi	523,000	0.553	706	186	181	3.16	2.12
	(522,000, 524,000)	(0.357, 0.749)	(-764, 2,180)	(-101, 472)	(122, 239)	(-5,690, 5,690)	(-836, 841)
Buwama - Kiyindi	511,000	0.322	9,110	9,810	369	761	178
	(510,000, 512,000)	(0.287, 0.357)	(5,390, 12,800)	(4,390, 15,200)	(302, 436)	(-80,200, 81,700)	(-81,200, 81,500)
Buwama - Wamala	523,000	0.365	31.7	19.7	124	9.27	0
	(522,000, 524,000)	(-1,000, 1,000)	(22, 41.3)	(13.6, 25.8)	(107, 141)	(-578, 596)	(-344, 344)
Kaazi - Kiyindi	510,000	0.448	9,900	1,670	438	923	718
	(509,000, 511,000)	(0.325, 0.571)	(1,100, 18,700)	(749, 2,600)	(365, 510)	(-64,900, 66,800)	(610, 825)
Kaazi - Wamala	521,000	0.581	9,200	4,820	974	None	None
	(520,000, 522,000)	(0.515, 0.647)	(3,680, 14,700)	(2,430, 7,210)	(795, 1,150)		
Kiyindi - Wamala	510,000	0.454	134	40.3	181	0	0.00363
	(509,000, 511,000)	(0.391, 0.518)	(57.9, 209)	(25.4, 55.3)	(166, 197)	(-663, 663)	(-3.69, 3.7)
							<u> </u>

Table S6: Locality-specific (in LVB) putative sweeps based on $\rm H12$ statistic.

Site	Count	Chr.	Putative Sweeps	Other sites ¹
Banda	44	2L	28.6 Mb; 36 Mb; 36.4 Mb; 36.9 Mb; 37.6 Mb;	1 also found in BFS, GNS
			38.1 Mb; 39.1 Mb; 42.2 Mb; 43.4 Mb; 43.8	
			Mb; 44.3 Mb; 44.9 Mb; 45.4 Mb	
		2R	4.2 Mb; 12.3 Mb; 18.3 Mb; 23.6 Mb; 29.4	1 also found in BFM, BFS, CMS, GNS, GWA; 1 also found in
			Mb; 30.3 Mb; 33.7 Mb; 34.8 Mb; 35.8 Mb;	BFM, GWA; 5 also found in GWA
			36.5 Mb; 44.1 Mb; 44.6 Mb; 49.7 Mb	
		3L	18.5 Mb; 21.6 Mb; 23.4 Mb; 23.9 Mb; 32.8	
			Mb	
		3R	$2.6~{\rm Mb};7.9~{\rm Mb};29.2~{\rm Mb};30.5~{\rm Mb};31.3~{\rm Mb};$	1 also found in GNS
			32.1 Mb; 33.2 Mb; 45.3 Mb; 46.4 Mb; 47 Mb	
		X	0.5 Mb; 2.1 Mb; 4.3 Mb	1 also found in AOM
Bugala (I)	24	2L	2.5 Mb; 5.5 Mb; 7.1 Mb; 19 Mb; 31.1 Mb; 43	1 also found in AOM, BFM, BFS, CMS, GAS, GNS, UGS; 1
			Mb; 45.7 Mb	also found in AOM, UGS
		2R	6.7 Mb; 21.1 Mb; 24 Mb; 24.6 Mb; 35.6 Mb;	1 also found in BFM, GWA; 2 also found in GWA
			37.1 Mb; 38.6 Mb; 39 Mb; 55.9 Mb	
		3L	17.2 Mb; 29.5 Mb	
		3R	26 Mb; 35.8 Mb; 37.5 Mb	
		X	3.5 Mb; 5.7 Mb; 10.8 Mb	
Bukasa	112	2L	12.6 Mb; 13.6 Mb; 17.7 Mb; 20.1 Mb; 20.9	1 also found in AOM, BFM, BFS, CMS, GAS, GNS; 1 also
			Mb; 21.6 Mb; 22.7 Mb; 23.6 Mb; 24.7 Mb;	found in BFM, GAS; 1 also found in BFS, GAS, GNS; 1 also
			25.4 Mb; 26.2 Mb; 26.9 Mb; 27.3 Mb; 27.8	found in BFS, GNS; 2 also found in CMS; 2 also found in GAS;
			Mb; 28.4 Mb; 29.1 Mb; 30.1 Mb; 31.5 Mb;	1 also found in GWA
			32.3 Mb; 33.3 Mb; 35.8 Mb; 39.4 Mb; 39.8	
			Mb; 40.6 Mb; 41.4 Mb; 43.1 Mb; 45.6 Mb;	
			48.1 Mb; 49.3 Mb	

	2R	1.3 Mb; 4.7 Mb; 5.3 Mb; 7.2 Mb; 7.6 Mb; 8	1 also found in AOM, GAS, GWA; 2 also found in BFM; 1 also
		Mb; 9.7 Mb; 10.5 Mb; 12 Mb; 12.4 Mb; 13.5	found in BFM, GWA; 3 also found in GAS, GWA; 6 also found
		Mb; 14 Mb; 15.7 Mb; 16.9 Mb; 17.5 Mb; 19.4	in GWA
		Mb; 22.9 Mb; 24.9 Mb; 25.8 Mb; 26.6 Mb;	
		29.9 Mb; 30.8 Mb; 32.4 Mb; 33.4 Mb; 35.5	
		Mb; 37.6 Mb; 43 Mb; 45.6 Mb; 47 Mb; 49.5	
		Mb; 54.8 Mb	
	3L	7.3 Mb; 11.6 Mb; 13.1 Mb; 15.6 Mb; 18.1	2 also found in BFM; 1 also found in GAS
		Mb; 19.1 Mb; 19.8 Mb; 20.6 Mb; 24.2 Mb;	
		25.2 Mb; 27.3 Mb; 28 Mb; 28.7 Mb; 29.7 Mb;	
		30.6 Mb; 33.7 Mb; 34.7 Mb; 35.3 Mb; 36.1	
		Mb; 38.7 Mb; 39.9 Mb; 40.3 Mb; 41.2 Mb	
	3R	5.1 Mb; 5.9 Mb; 7.2 Mb; 8.9 Mb; 12.7 Mb;	1 also found in GWA
		13.3 Mb; 14.1 Mb; 14.9 Mb; 15.9 Mb; 17.2	
		Mb; 22.3 Mb; 23.3 Mb; 23.8 Mb; 24.9 Mb;	
		26.8 Mb; 27.9 Mb; 31.4 Mb; 33 Mb; 35.9 Mb;	
		36.9 Mb	
	X	1.7 Mb; 2.8 Mb; 4.9 Mb; 6 Mb; 7 Mb; 11.5	1 also found in BFM, GAS, GWA; 4 also found in GAS
		Mb; 12.5 Mb; 13.6 Mb; 16.7 Mb	
Buwama 27	2L	14.9 Mb; 15.9 Mb; 25.1 Mb; 26.5 Mb; 31.6	1 also found in BFM, GAS; 1 also found in GWA
		Mb	
	2R	24.4 Mb; 39.5 Mb; 44.5 Mb; 46.3 Mb; 49.1	1 also found in AOM, BFM, CMS; 1 also found in BFS, CMS,
		Mb; 53.7 Mb; 55.3 Mb	GNS; 1 also found in CMS; 1 also found in GWA
	3L	2.4 Mb; 3.1 Mb; 3.6 Mb; 4.1 Mb; 10.6 Mb;	
		16.1 Mb; 21.7 Mb; 29.8 Mb	
	3R	18 Mb; 29.1 Mb; 35.5 Mb; 37.7 Mb; 38.4 Mb;	1 also found in GNS

Voori	15	OT.	0 5 Mb. 24 6 Mb	1 also found in AOM
Kaazi	15	2L	8.5 Mb; 34.6 Mb	1 also found in AOM
		2R	8.3 Mb; 23 Mb	1 also found in AOM; 1 also found in GAS
		3L	3.5 Mb; 4.8 Mb; 8.6 Mb; 11.8 Mb; 13 Mb;	1 also found in BFM; 1 also found in BFM, GNS
			15.8 Mb; 25 Mb; 26.8 Mb	
		3R	14.7 Mb; 15.6 Mb; 46 Mb	1 also found in GNS
Kiyindi	40	2L	$2~{\rm Mb};~10.6~{\rm Mb};~17.8~{\rm Mb};~22.1~{\rm Mb};~23.9~{\rm Mb};$	1 also found in AOM, BFM, BFS, CMS, GNS, UGS; 1 also
			$26~{\rm Mb};~28.7~{\rm Mb};~29.9~{\rm Mb};~34.8~{\rm Mb}$	found in BFS, GNS; 3 also found in GAS
		2R	19.1 Mb; 20.2 Mb; 25.9 Mb; 35.3 Mb; 36.6	1 also found in AOM; 1 also found in AOM, BFS, CMS, GNS,
			Mb; 38.1 Mb; 40 Mb; 41.7 Mb; 42.4 Mb; 45.3	$\operatorname{GWA};$ 1 also found in BFM; 2 also found in GWA
			Mb; 48.2 Mb; 48.6 Mb; 50.1 Mb; 52.2 Mb;	
			53.6 Mb; 54.7 Mb; 55.1 Mb	
		3L	1.2 Mb; 8.9 Mb; 12.1 Mb; 12.6 Mb; 13.5 Mb;	1 also found in BFM; 1 also found in GWA
			14.8 Mb; 15.4 Mb; 16 Mb; 16.8 Mb; 19.7 Mb;	
			26.7 Mb	
		3R	38 Mb; 41.9 Mb; 48.3 Mb	
Nsadzi	47	2L	23.2 Mb; 27 Mb; 45.5 Mb	
		2R	1.6 Mb; 2.3 Mb; 3.2 Mb; 4 Mb; 8.8 Mb; 10.2	1 also found in BFM, GAS; 1 also found in BFM, GWA; 1 also
			Mb; 13.2 Mb; 16.1 Mb; 20 Mb; 21.3 Mb; 24.7	found in BFS, CMS, GNS; 1 also found in CMS; 2 also found
			Mb; 30.5 Mb; 34.2 Mb; 37.3 Mb; 41.2 Mb;	in GWA
			43.5 Mb; 52 Mb	
		3L	10.5 Mb; 11 Mb; 24.3 Mb; 35 Mb; 35.4 Mb;	1 also found in GAS
			36.8 Mb; 37.6 Mb	
		3R	3.8 Mb; 6 Mb; 7.4 Mb; 19.9 Mb; 20.5 Mb;	1 also found in BFS, GNS
			21.4 Mb; 23 Mb; 24.2 Mb; 27.7 Mb; 41.6 Mb;	
			42.2 Mb; 48.2 Mb; 49.8 Mb; 50.4 Mb	
		X	0.7 Mb; 2.3 Mb; 5.2 Mb; 7.7 Mb; 11.9 Mb;	1 also found in BFM, GAS, GWA; 1 also found in GAS
			17.9 Mb	

Sserinya	35	2L	22.2 Mb; 24.2 Mb; 25.7 Mb; 33.2 Mb; 34.9	1 also found in BFM, GNS; 1 also found in CMS, GAS; 1 also
			Mb; 35.4 Mb; 40.2 Mb; 41.1 Mb; 45.1 Mb;	found in GAS, GNS
			45.9 Mb; 46.8 Mb	
		2R	0.4 Mb; 7.7 Mb; 21.5 Mb; 30 Mb; 32 Mb; 36.1	3 also found in GWA
			Mb	
		3L	10.1 Mb; 10.9 Mb; 14.6 Mb; 34.5 Mb; 41.8	1 also found in BFS, CMS, GNS, GWA, UGS
			Mb	
		3R	1.9 Mb; 10 Mb; 15 Mb; 24.8 Mb; 26.2 Mb; 27	1 also found in GAS
			Mb; 29 Mb	
		X	5.8 Mb; 12.7 Mb; 13.1 Mb; 18.1 Mb; 18.8 Mb;	1 also found in BFM, CMS, GAS, GWA; 1 also found in BFM,
			21.3 Mb	GAS, GWA; 1 also found in CMS, GNS, GWA; 2 also found in
				GAS
Wamala	25	2L	13.4 Mb; 15.5 Mb; 17.1 Mb; 19.1 Mb; 20 Mb	2 also found in GAS
		2R	21.2 Mb; 22.2 Mb; 29.6 Mb; 38.8 Mb; 39.7	2 also found in AOM; 2 also found in GWA
			Mb; 47.6 Mb; 48.3 Mb; 48.9 Mb	
		3L	3.3 Mb; 7.6 Mb; 8.2 Mb	
		3R	5 Mb; 39.2 Mb; 43.2 Mb; 46.5 Mb; 47.5 Mb;	
			48.5 Mb; 50.5 Mb; 50.9 Mb; 51.8 Mb	

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¹ Ag1000G site codes: AOM: Angola [coluzzii]; BFM: Burkina Faso [coluzzii]; BFS: Burkina Faso [gambiae]; CMS: Cameroon [gambiae]; GAS: Gabon [gambiae]; GNS: Guinea [gambiae]; GWA: Guinea-Bissau; UGS: Uganda [gambiae]

Table S7: Putative sweeps based on H12 statistic present on islands but rare or absent on LVB mainland.

Chr.	Region Start	Region End	Island Sites with Putative Sweep	Mainland Sites with Putative Sweep	Outlier Island Localities	Outlier Mainland Localities	Ag1000G Populations with Putative Sweep
2R	16,200,000	16,300,000	4 / 5	0 / 4	Banda; Bukasa; Nsadzi;	None None	Guinea-Bissau
					Sserinya		
2R	17,300,000	17,500,000	4 / 5	1 / 4	Banda; Bugala (I); Bukasa;	Buwama	Guinea-Bissau
					Sserinya		
2R	21,000,000	21,100,000	5 / 5	1 / 4	Banda; Bugala (I); Bukasa;	Buwama	None
					Nsadzi; Sserinya		
2R	40,400,000	40,800,000	4 / 5	1 / 4	Bugala (I); Bukasa;	Wamala	Burkina Faso [gambiae],
					Nsadzi; Sserinya		Cameroon $[gambiae],$
							Gabon [gambiae], Guinea-
							Bissau
2R	41,100,000	41,200,000	4 / 5	1 / 4	Banda; Bukasa; Nsadzi;	Wamala	Cameroon [gambiae]
					Sserinya		
2R	55,800,000	55,900,000	4 / 5	1/4	Banda; Bugala (I); Bukasa;	Kiyindi	Angola [coluzzii]
					Sserinya		
2L	7,700,000	7,800,000	4 / 5	1 / 4	Banda; Bugala (I); Bukasa;	Buwama	Guinea-Bissau, Uganda
					Nsadzi		$oxed{[gambiae]}$
2L	8,100,000	8,200,000	4 / 5	0 / 4	Banda; Bugala (I); Nsadzi;	None	None
					Sserinya		
2L	42,400,000	42,500,000	4 / 5	0 / 4	Banda; Bugala (I); Nsadzi;	None	None
					Sserinya		
2L	43,500,000	43,600,000	5 / 5	1 / 4	Banda; Bugala (I); Bukasa;	Buwama	None
					Nsadzi; Sserinya		
2L	49,000,000	49,100,000	4 / 5	1 / 4	Banda; Bugala (I); Bukasa;	Wamala	None
					Sserinya		

3R	26,600,000	26,700,000	5 / 5	0 / 4	Banda; Bugala (I); Bukasa;	None	None
					Nsadzi; Sserinya		
3R	36,700,000	36,800,000	4 / 5	0 / 4	Banda; Bugala (I); Bukasa;	None	None
					Nsadzi		
3R	44,200,000	44,300,000	4 / 5	1 / 4	Banda; Bukasa; Nsadzi;	Kiyindi	Angola [coluzzii]
					Sserinya		
3R	46,200,000	46,300,000	4 / 5	0 / 4	Banda; Bukasa; Nsadzi;	None	None
					Sserinya		
X	6,600,000	7,000,000	4 / 5	0 / 4	Banda; Bugala (I); Bukasa;	None	Burkina Faso [coluzzii],
					Nsadzi; Sserinya		Gabon [gambiae], Guinea-
							Bissau
X	8,100,000	10,700,000	4 / 5	0 / 4	Banda; Bugala (I); Bukasa;	Kiyindi	Burkina Faso [coluzzii],
					Nsadzi; Sserinya		Burkina Faso [gambiae],
							Gabon [gambiae]
X	11,300,000	11,800,000	5 / 5	0 / 4	Banda; Bugala (I); Bukasa;	None	Gabon [gambiae]
					Nsadzi; Sserinya		
X	12,900,000	13,000,000	4 / 5	0 / 4	Banda; Bugala (I); Bukasa;	None	Gabon $[gambiae]$
					Sserinya		
X	14,300,000	14,400,000	5 / 5	1 / 4	Banda; Bugala (I); Bukasa;	Kaazi	Gabon [gambiae]
					Nsadzi; Sserinya		
X	16,200,000	16,300,000	4 / 5	1 / 4	Banda; Bugala (I); Bukasa;	Kaazi	Burkina Faso [coluzzii],
					Sserinya		Gabon [gambiae]

Table S8: Putative sweeps based on H12 statistic present on LVB mainland but rare or absent on islands.

			Island Sites with	Mainland Sites with	Outlier	Outlier	Ag1000G Populations
Chr.	Region Start	Region End	Putative Sweep	Putative Sweep	Island Localities	Mainland Localities	with Putative Sweep
2R	27,600,000	27,700,000	1 / 5	3 / 4	Nsadzi	Buwama; Kiyindi; Wamala	None
2R	38,000,000	38,100,000	1 / 5	3 / 4	Bugala (I)	Buwama; Kiyindi; Wamala	None
2R	42,700,000	42,800,000	0 / 5	3 / 4	None	Buwama; Kiyindi; Wamala	None
2R	45,400,000	45,500,000	1 / 5	3 / 4	Sserinya	Buwama; Kiyindi; Wamala	None
2R	46,800,000	46,900,000	1 / 5	3 / 4	Banda	Buwama; Kiyindi; Wamala	Cameroon [gambiae]
2R	48,000,000	48,100,000	1 / 5	3 / 4	Bukasa	Buwama; Kaazi; Wamala	Angola [coluzzii],
							Cameroon [gambiae]
2R	48,800,000	48,900,000	1 / 5	3 / 4	Nsadzi	Buwama; Kaazi; Wamala	None
2R	50,900,000	51,000,000	1 / 5	3 / 4	Bukasa	Kaazi; Kiyindi; Wamala	Burkina Faso [gambiae],
							Guinea [gambiae]
2R	51,500,000	51,600,000	0 / 5	3 / 4	None	Kaazi; Kiyindi; Wamala	None
2R	57,500,000	57,600,000	1 / 5	3 / 4	Banda	Buwama; Kaazi; Kiyindi	Angola [coluzzii], Guinea-
							Bissau
2L	2,900,000	3,000,000	1 / 5	4 / 4	Sserinya	Buwama; Kaazi; Kiyindi;	Angola [coluzzii], Burk-
						Wamala	ina Faso [coluzzii], Burkina
							Faso [gambiae], Cameroon
							[gambiae], Gabon [gam-
							biae], Guinea [gambiae],
							Uganda [gambiae]
2L	4,200,000	4,300,000	1 / 5	4 / 4	Bugala (I)	Buwama; Kaazi; Kiyindi;	Angola [coluzzii],
						Wamala	Cameroon $[gambiae],$
							Gabon [gambiae], Uganda
							$ \boxed{ [gambiae] }$

			T				
2L	5,700,000	5,800,000	1 / 5	3 / 4	Bugala (I)	Buwama; Kaazi; Kiyindi	Angola [coluzzii], Guinea
							[gambiae], Uganda [gam-
							biae]
2L	6,200,000	6,300,000	1 / 5	3 / 4	Bugala (I)	Kaazi; Kiyindi; Wamala	Uganda [gambiae]
2L	6,600,000	6,800,000	1 / 5	3 / 4	Bugala (I)	Kaazi; Kiyindi; Wamala	Angola [coluzzii],
							Cameroon $[gambiae],$
							Gabon [gambiae], Guinea
							[gambiae], Uganda [gam-
							biae]
2L	10,000,000	10,100,000	1 / 5	3 / 4	Sserinya	Kaazi; Kiyindi; Wamala	None
2L	10,800,000	10,900,000	0 / 5	3 / 4	None	Kaazi; Kiyindi; Wamala	None
2L	11,300,000	11,400,000	1 / 5	3 / 4	Bugala (I)	Kaazi; Kiyindi; Wamala	Guinea-Bissau
2L	12,000,000	12,100,000	1 / 5	3 / 4	Bugala (I)	Kaazi; Kiyindi; Wamala	None
2L	12,400,000	13,000,000	0 / 5	3 / 4	Bukasa	Buwama; Kaazi; Kiyindi;	None
						Wamala	
2L	14,500,000	14,900,000	1 / 5	3 / 4	Sserinya	Buwama; Kiyindi; Wamala	Gabon [gambiae], Uganda
							[gambiae]
2L	16,000,000	16,300,000	1 / 5	3 / 4	Bukasa	Buwama; Kaazi; Wamala	Gabon [gambiae]
^{2}L	16,600,000	16,700,000	1 / 5	4 / 4	Bugala (I)	Buwama; Kaazi; Kiyindi;	None
						Wamala	
2L	18,700,000	18,800,000	1 / 5	3 / 4	Nsadzi	Kaazi; Kiyindi; Wamala	None
2L	33,600,000	33,700,000	1 / 5	3 / 4	Bugala (I)	Buwama; Kaazi; Kiyindi	Angola [coluzzii]
2L	34,400,000	34,500,000	1 / 5	3 / 4	Sserinya	Buwama; Kaazi; Wamala	None

3R	28,500,000	28,700,000	1 / 5	4 / 4	Sserinya	Buwama; Kaazi; Kiyindi;	Burkina Faso [coluzzii],
						Wamala	Burkina Faso [gambiae],
							Cameroon $[gambiae],$
							Gabon [gambiae], Guinea
							[gambiae], Uganda [gam-
							biae]
3R	36,500,000	36,900,000	0 / 5	3 / 4	Nsadzi	Buwama; Kiyindi; Wamala	None
3R	43,000,000	43,100,000	0 / 5	3 / 4	None	Buwama; Kiyindi; Wamala	None
3R	43,700,000	44,100,000	0 / 5	3 / 4	Nsadzi	Buwama; Kiyindi; Wamala	Angola [coluzzii], Burk-
							ina Faso [gambiae], Guinea
							[gambiae], Uganda [gam-
							biae]
3R	48,800,000	48,900,000	0 / 5	3 / 4	None	Buwama; Kiyindi; Wamala	None
3R	50,000,000	50,100,000	1 / 5	3 / 4	Sserinya	Kaazi; Kiyindi; Wamala	None
3L	7,000,000	7,100,000	1 / 5	4 / 4	Sserinya	Buwama; Kaazi; Kiyindi;	None
						Wamala	
3L	11,500,000	11,600,000	1 / 5	3 / 4	Sserinya	Buwama; Kiyindi; Wamala	Burkina Faso [coluzzii]
3L	12,200,000	12,300,000	0 / 5	3 / 4	None	Kaazi; Kiyindi; Wamala	None
3L	13,400,000	13,500,000	0 / 5	3 / 4	None	Kaazi; Kiyindi; Wamala	None
3L	16,300,000	16,400,000	1 / 5	3 / 4	Sserinya	Buwama; Kiyindi; Wamala	Uganda [gambiae]

Table S9: Signatures of selective sweeps on known insecticide genes by site based on H12 statistic.

		Insecticide	Island Sites with	Mainland Sites with	Outlier		Outlier	Ag1000G Populations
Chr.	Location	Gene	Putative Sweep	Putative Sweep	Island Lo	ocalities	Mainland Locali-	with Putative Sweep
							ties	
-2R	28,497,407	Сурбр	5 / 5	4 / 4	Banda;	Bugala	Buwama; Kaazi;	Angola [coluzzii], Burkina Faso
					(I);	Bukasa;	Kiyindi; Wamala	[coluzzii], Burkina Faso [gambiae],
					Nsadzi; S	Sserinya		Cameroon [gambiae], Guinea [gam-
								biae], Uganda $[gambiae]$
3R	28,598,038	Gste	1 / 5	4 / 4	Sserinya		Buwama; Kaazi;	Burkina Faso [coluzzii], Burkina
							Kiyindi; Wamala	Faso [gambiae], Cameroon [gam-
								biae], Gabon [gambiae], Guinea
								[gambiae], Uganda [gambiae]
X	15,241,718	Cyp9k1	3 / 5	4 / 4	Banda;	Bukasa;	Buwama; Kaazi;	Burkina Faso [coluzzii], Burkina
					Sserinya		Kiyindi; Wamala	Faso [gambiae], Gabon [gambiae],
								Guinea [gambiae]

Table S10: Software and versions used for major parts of analysis.

Software	Version	Citation
ea-utils	-	[64]
BWA	0.7.16a	[67]
GATK	3.8	[68]
PLINK	1.90b4.6	[69, 70]
SHAPEIT2	2.837	[71]
${\rm SAM tools/BCF tools}$	1.5	[72, 90]
ADMIXTURE	1.3.0	[20]
CLUMPAK	-	[73]
VCFtools	0.1.15	[86]
$\delta a \delta i$ (python package)	1.7.0	[82, 83]
Stairway plot - Jpopgen	2-beta	[21]
selscan	1.2.0a	[87]
adegenet (R package)	2.1.0	[91]
ape (R package)	5.0	[92]
RColorBrewer (R package)	1.1-2	[93]
dendextend (R package)	1.6.0	[94]
rehh (R package)	2.0.2	[95]
eigensoft	7.2.1	[75, 76]
GNU parallel	20170422	[96]
tabix	1.5	[90]
bedtools	2.26.0	[97]

Table S11: Genomic coordinates of heterochromatic and inverted regions.

Chromosome arm	Start	End	Information
2L	20,524,058	$42,\!165,\!532$	2La inversion [66]
2R	18,575,300	26,767,588	2Rb inversion [66]
2L	1	2,431,617	Heterochromatic region [66]
2L	5,078,962	5,788,875	Heterochromatic region [66]
2R	58,984,778	61,545,105	Heterochromatic region [66]
3L	1	1,815,119	Heterochromatic region [66]
3L	4,264,713	5,031,692	Heterochromatic region [66]
3R	38,988,757	41,860,198	Heterochromatic region [66]
3R	52,161,877	53,200,684	Heterochromatic region [66]
X	20,009,764	24,393,108	Heterochromatic region [66]

Figures

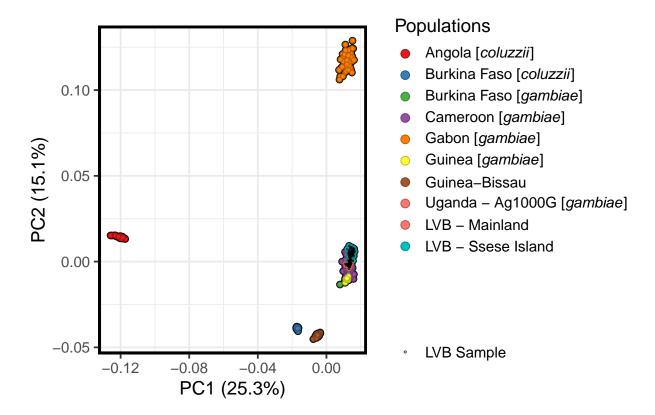


Figure S1: PCA plot of study individuals and *A. gambiae* and *A. coluzzii* individuals from reference Ag1000G populations, showing the first and second components. Kenyan population is not included, and analysis is based on chromosome 3.

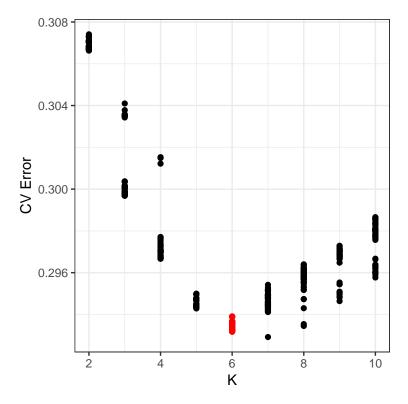


Figure S2: ADMIXTURE cross-validation error. Cross-validation error for range of k values for ADMIXTURE analysis of Lake Victoria Basin individuals and $A.\ qambiae$ and $A.\ coluzzii$ Ag1000G reference populations.

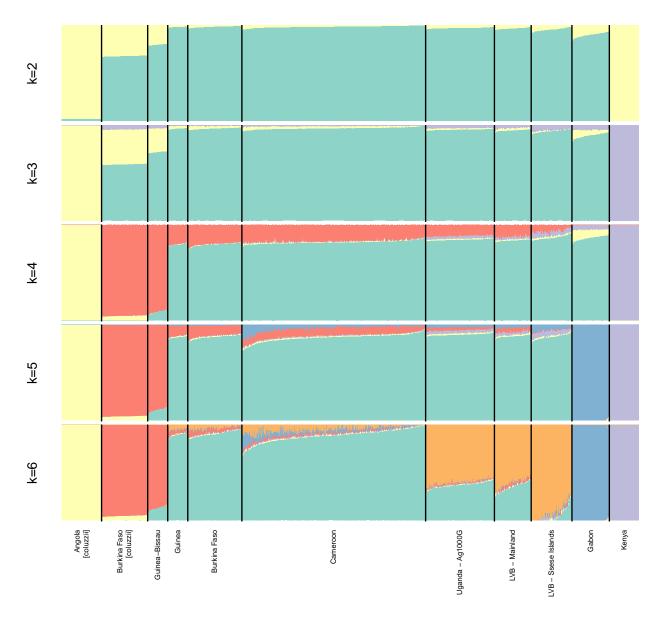


Figure S3: (Caption on next page.)

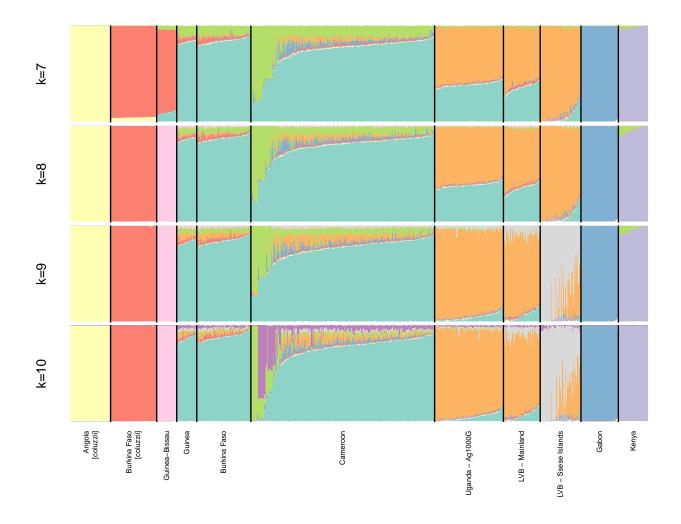


Figure S3: ADMIXTURE-inferred ancestry.

Ancestry of individuals in Lake Victoria Basin and of Ag1000G reference populations as inferred by ADMIXTURE clustering method. Samples are A. gambiae unless noted, and analysis is based on chromosome 3. Using k=6 clusters minimizes cross validation error (Fig. S2).

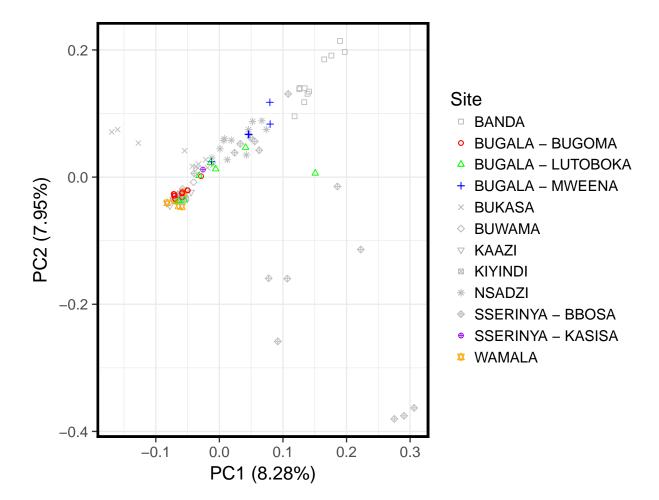


Figure S4: PCA showing Bugala subdivision.

PCA colored by sampling locations. Based on this analysis, individuals from Bugala were split into mainland- and island-like subpopulations. Samples from Sserinya Island, though sampled from two localities, were not split.

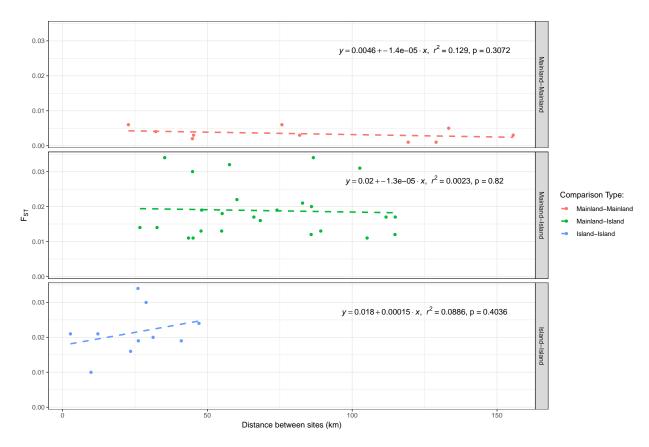


Figure S5: Correlations between genetic distance (F_{ST}) and geographic distance between localities. The p-values are for the test that the slope is significantly different from zero.

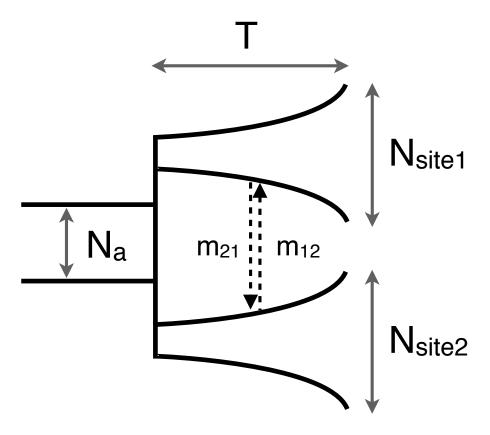
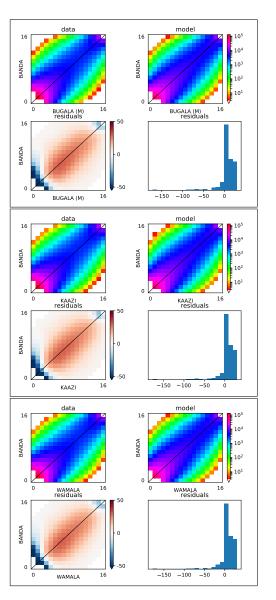
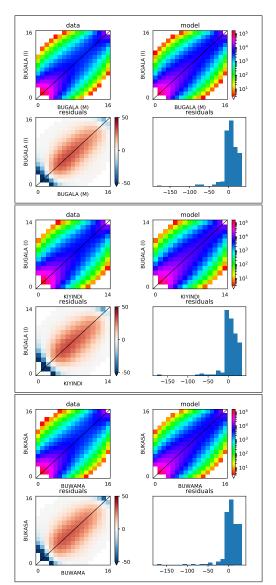


Figure S6: IM model schematics.

Schematic of model fit to data with $\delta a \delta i$ for population history inference between all pairs of sampled sites using IM model.





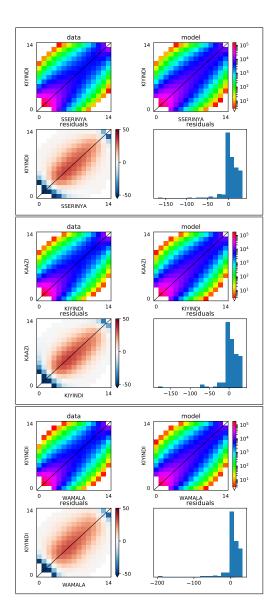


Figure S7: (Caption on next page.)

Figure S7: Two population $\delta a \delta i$ optimization results.

Comparison between best fitting model and data frequency spectra for two population $\delta a \delta i$ inference. Of the pairwise comparisons for which the best model included migration, a randomly selected set of nine are shown here. Two-dimensional frequency spectra are plotted as logarithmic colormaps for the data (upper left) and model (upper right), and the bottom row plots show the residuals between model and data. Positive residuals in red indicate the model predicts too many SNPs in that entry while negative residuals in blue indicate the model predicts too few.



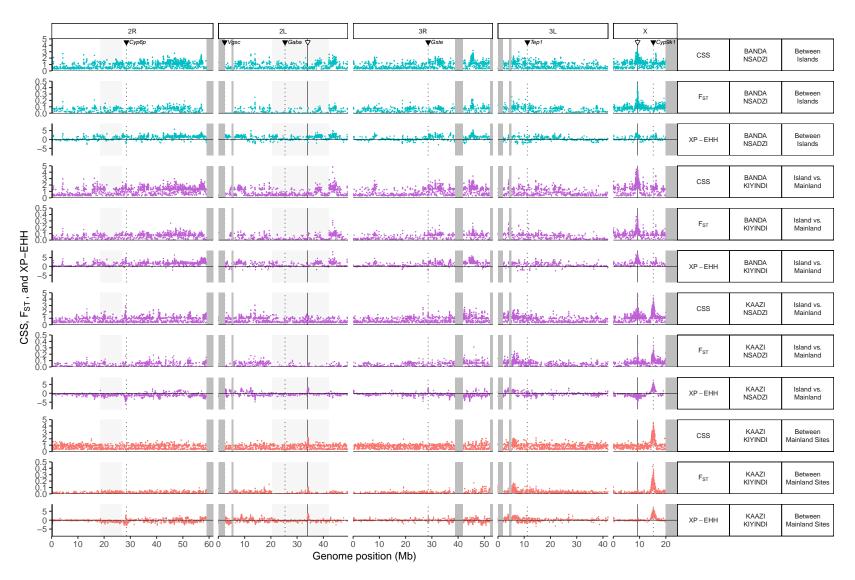


Figure S8: F_{ST} , XP-EHH, and Composite Selection Score (CSS) across genome.

 F_{ST} , XP-EHH, and CSS averaged in windows of size 10 kb plotted across genome for pairwise comparisons of island and mainland localities. Shaded regions indicate inversions or heterochromatic regions (excluded from analysis) and dotted lines indicate known insecticide genes while solid lines indicate the two putative sweeps identified in the present study. Only several exemplar pairs of populations shown.

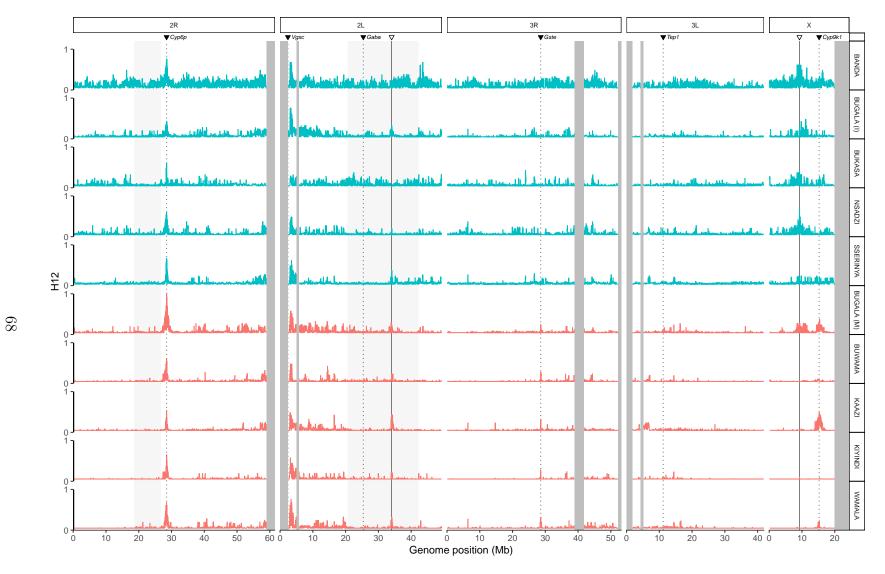


Figure S9: H12 across genome.

Values of H12, a measure of haplotype homozygosity, plotted across genome. Shaded regions indicate inversions or heterochromatic regions (excluded from analysis) and dotted lines indicate known insecticide genes while solid lines indicate the two putative sweeps identified in the present study.

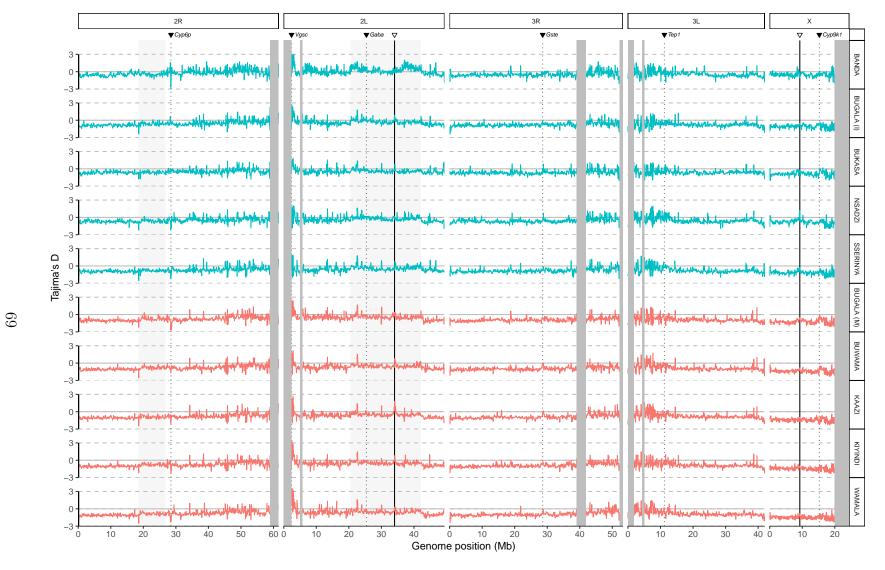


Figure S10: Tajima's D across genome.

Tajima's D plotted across genome. Shaded regions indicate inversions or heterochromatic regions (excluded from analysis) and dotted lines indicate known insecticide genes while solid lines indicate the two putative sweeps identified in the present study.

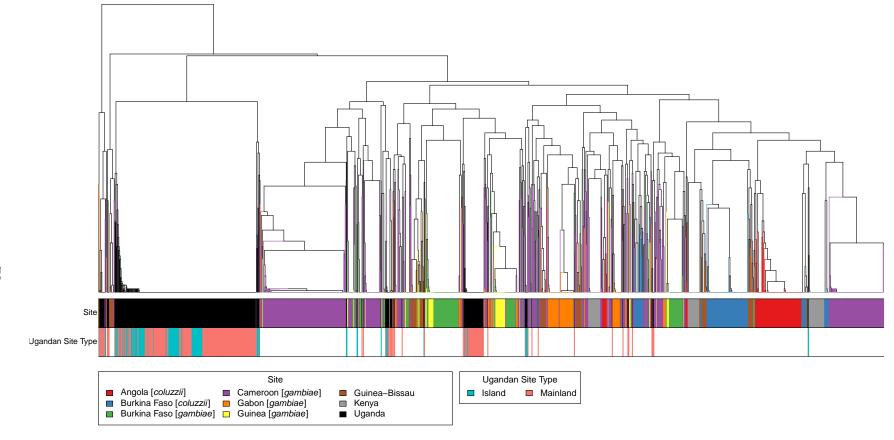


Figure S11: Tree for sweep on Cyp6p gene cluster on chromosome arm 2R.

Distance-based tree of haplotypes near sweep at Cyp6p gene cluster on chromosome arm 2R. Region shown is 10 kb up- and downstream of sweep target, centered at chr2R:28,501,972 (the approximate location of the peaks in pairwise statistics). Top color bar indicates locality, with all Ugandan individuals, from both the Ag1000G reference population and the LVB, in black. The bottom color bar differentiates the Ugandan individuals into mainland (red) and island (blue) individuals.

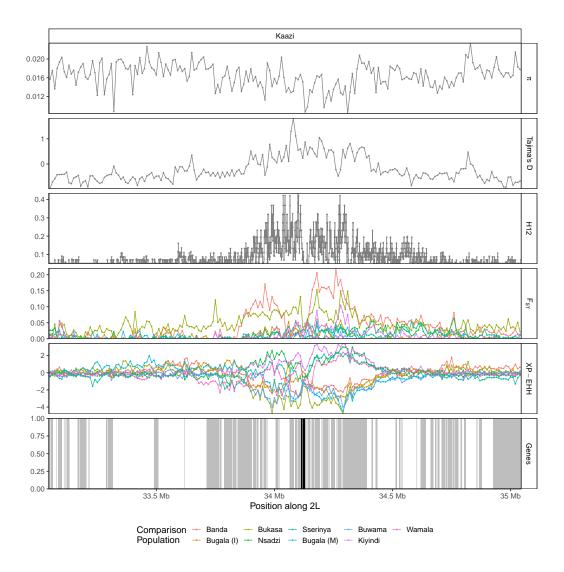


Figure S12: Selective sweep signal on chromosome 2L.

Population genetic statistics plotted near putative sweep on chromosome 2L. Focus population for all pairwise F_{ST} and XP-EHH comparisons is mainland site Kaazi, chosen to maximize peak height in these statistics. Region shown is 1 Mb up- and downstream of sweep target, centered at chr2L:34,044,820. Several genes involved in chorion formation (AGAP006549, AGAP006550, AGAP006551, AGAP006553, AGAP006554, AGAP006555 and AGAP006556) are highlighted with black vertical lines, while other genes are indicated with gray vertical lines.

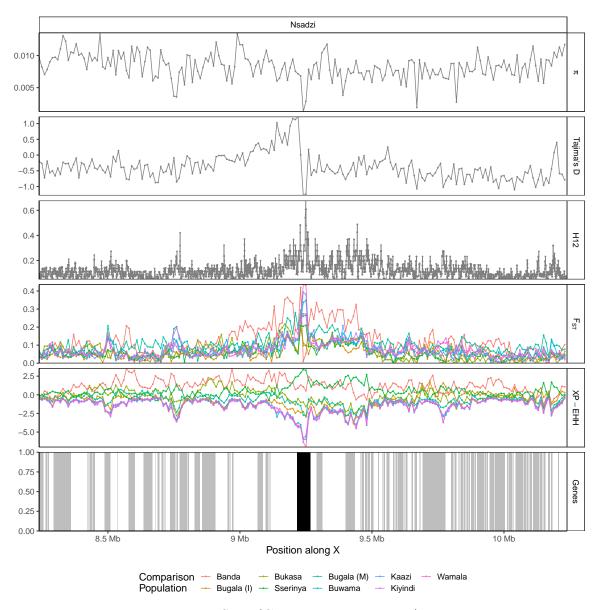


Figure S13: (Caption on next page.)

Figure S13: Selective sweep signal on X-chromosome near rdgA ortholog.

Population genetic statistics plotted near putative sweep on X-chromosome. Focus population for all pairwise F_{ST} and XP-EHH comparisons is island site Nsadzi, chosen to maximize peak height in these statistics. Region shown is 1 Mb up- and downstream of sweep target, centered at chrX:9,238,942 (the approximate peak in pairwise statistics). The gene eye-specific diacylglycerol kinase (AGAP000519, chrX:9,215,505-9,266,532) is highlighted with a black vertical line, while other genes are indicated with gray vertical lines.

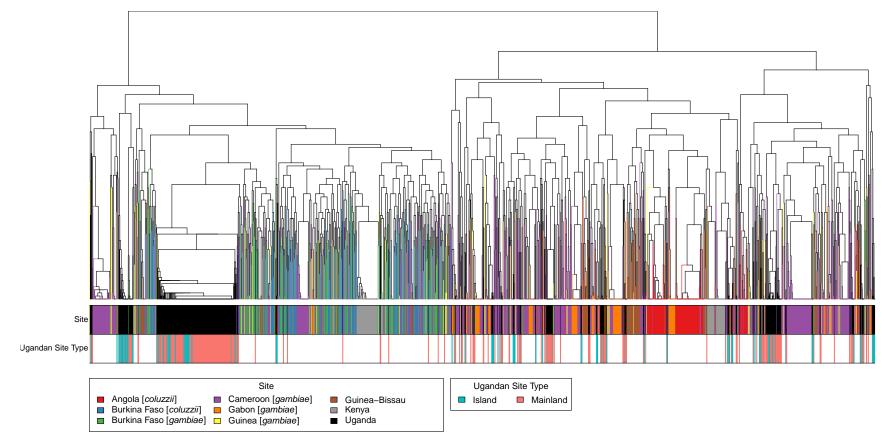


Figure S14: Tree for putative sweep on chromosome 2L.

Distance-based tree of haplotypes near putative sweep on chromosome 2L. Region shown is 10 kb up- and downstream of sweep target, centered at chr2L:34,044,820 (the approximate location of the peaks in pairwise statistics). Top color bar indicates locality, with all Ugandan individuals, from both the Ag1000G reference population and the LVB, in black. The bottom color bar differentiates the Ugandan individuals into mainland (red) and island (blue) individuals.

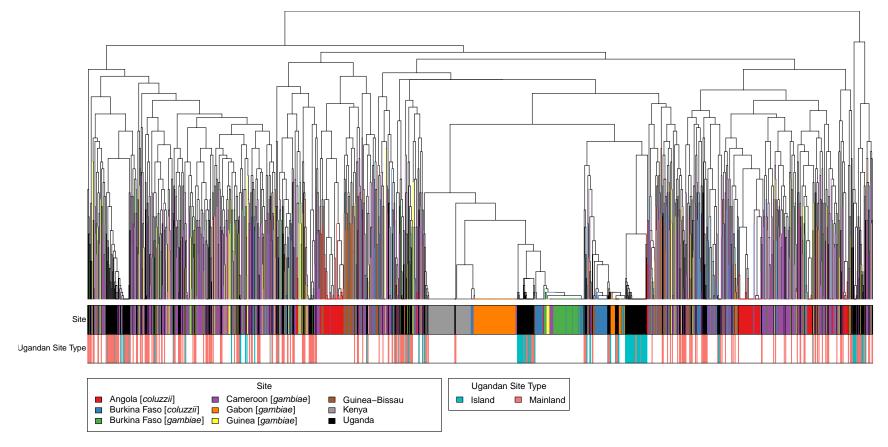


Figure S15: Tree for putative sweep on X-chromosome near rdgA ortholog.

Distance-based tree of haplotypes near putative sweep on X-chromosome. Region shown is 10 kb up- and downstream of sweep target, centered at chrX:9,238,942 (the approximate location of the peaks in pairwise statistics). Top color bar indicates locality, with all Ugandan individuals, from both the Ag1000G reference population and the LVB, in black. The bottom color bar differentiates the Ugandan individuals into mainland (red) and island (blue) individuals.