Identifying single origin rare variants in population genomic data

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Abstract

Genomic analyses have shown that some mutations in large population genomic datasets may be the result of repeated, independent events at the same locus. However, the possibility of this recurrent mutation is often ignored, even when it has the potential to introduce errors, such as when assuming co-ancestry through variant sharing for demographic analysis. Even rare variants such as doubletons, which should be particularly informative about recent demography, may have multiple origins despite arising relatively recently in the population. Here, we develop methods first to estimate the frequency of recurrent doubletons in a population genomic dataset from the occurrence of tri-allelic sites with two different singleton mutations, and then to identify a subset of high confidence single origin doubletons based on the presence of a linked rare variant on the surrounding shared haplotype. Applying these methods to data for the malaria mosquito *Anopheles gambiae*, we estimate that up to ∼16% of reported doubletons have independent origins. We then identify a set of doubletons likely (∼99%) to have a single origin, which consists of ∼68% of all the expected single origin doubletons (and ∼57% of all observed doubletons). The effectiveness of our approach is queried, and these doubletons are then used to test population genetic hypotheses about recombination, selection, and isolation by distance. The methods developed here should be useful for demographic inference when populations are large enough such that recurrent mutation cannot be ignored.

Introduction

A frequent assumption underpinning many population genetic models is that all copies of an allele in a sample descend from a single mutational event in a common ancestor in the genealogical past (Wakeley, 2009). Common frameworks that use this assumption include the "infinite alleles" and "infinite sites" models – that is, that each new mutation either introduces a new, unique mutant allele, or that each new mutation occurs at a new, unique position in the genome, and therefore there cannot be multiple copies of the same mutation arising from different sources (Kimura and Crow, 1964; Karlin and McGregor, 1967; Kimura, 1971). However, in reality some variants will be the result of the same mutation occurring on multiple branches of the genealogy, which are then inherited through successive generations into the present (Haldane, 1933; Jenkins and Song, 2011; Wakeley et al., 2023). One set of conditions where this is often observed is sites that come under strong, recent selection from new external factors leading to selective sweeps (Messer and Petrov, 2013; Smadi, 2017); for example, there are thought to be a minimum of 10 separate origins of the L995 mutation in the same codon conferring resistance to pyrethroid insecticides in *Anopheles gambiae* mosquitoes (Clarkson et al., 2021), while target site resistance to herbicides is thought to have arisen on at least seven occasions across three variants in the agricultural weed *Amaranthus tuberculatus* (Kreiner et al., 2022). However, many recurrent mutations may be associated with weakly selected or neutral sites making them difficult to detect in population genomic data, and if not accounted for, can introduce errors into analyses such as the incorrect inference of degrees of relatedness between individuals.

Rare variants - those alleles found at very low frequencies in a population - are expected to be more recent and hence more likely to have a single origin than high frequency variants (Slatkin, 1985; Novembre and Slatkin, 2009; Mathieson and McVean, 2012, 2014; O'Connor et al., 2015). They may therefore be more geographically restricted, making them particularly useful for inferring recent, fine-scale demographic history, and distinguishing populations that are otherwise difficult to differentiate (O'Connor et al., 2015; Cubry et al., 2017). Additionally, rare variants are frequently implicated as contributing to genetic diseases, and so understanding their nature and origin is also important for epidemiology (Lachance, 2010; Nelson et al., 2012; Chen et al., 2022). However, while the assumed status of a single origin may be appropriate for rare variants in some population

genomic datasets, in others, particularly if the species is highly diverse or a great many individuals have been sampled, multiple origins of the same mutation may be common. Demographic inference in these species would be greatly improved by having reliable methods that allow one to identify a subset of rare variants that are highly likely to have a single origin (Johnson et al., 2022).

In this paper we focus on the rarest of variants that can be useful for indicating relatedness, doubleton mutations, which occur in only two individuals in a population genomic dataset (Mathieson and McVean, 2014). We first describe an approach for estimating the proportion of doubletons that have multiple origins, based on an analysis of triallelic sites. We then outline how one can update the probability that a specific doubleton has a single origin based on the presence of another linked rare variant. We apply these methods to published data from the *Anopheles gambiae* species complex (henceforth *An. gambiae* sensu lato, or s.l.), the primary vector of malaria in sub-Saharan Africa and therefore responsible for hundreds of thousands of deaths every year (The Anopheles gambiae 1000 Genomes Consortium, 2017, 2020; World Health Organisation, 2020). Finally, we use single origin doubletons from *An. gambie s.l* to test population genomic hypotheses involving recombination, selection, and isolation by distance.

Theory

For doubletons in a large dataset we will assume that the common allele is ancestral and either (1) the doubleton allele has a common origin and the two sequences are, at that position, each other's closest genealogical relatives - i.e., they are reciprocal closest relatives (henceforth RCR); or (2) the doubleton allele arose twice at that site, independently, and the two sequences are no more closely related at that site than random pairs of sequences in the dataset. In principle there are additional possibilities, such as (3) that the doubleton allele is ancestral, or (4) that it is derived, with a single origin, but then there was a back-mutation to the ancestral state, leaving two sequences with the derived state that are not RCR, but these scenarios are expected to be significantly less likely than the other two and we will not consider them further.

Estimating the proportion of single origin doubletons

Our first goal is to estimate the prior probability that doubletons fit the first scenario above, that they have a single origin and the two sequences sharing the doubleton are RCR at that site, i.e., the conditional probability that the sequences are RCR given they share a doubleton, $P[RCR|d]$. To do so we use the rate of recurrent mutation at a single site, calculated from data on triallelic sites in which one base is found in all but two sequences, and then there are two different singletons (henceforth triallelic singleton sites). While, again, there are multiple possible scenarios that can lead to a polymorphism of this type, the most likely is that the common base is ancestral and there were mutations on two external branches of the genealogy to two different bases, with no mutations on any internal branch. Because the mutation rate from one nucleotide to another depends on the nucleotides involved (e.g., transition rates differ from transversion rates), we will estimate $P[RCR|d]$ separately for each of the 12 types of mutation. However, for simplicity we will assume that the probability of a mutation of a specified type is constant throughout the genome - that is, that there are no neighbourhood effects of local sequence context on mutation rates.

Proceeding by example, suppose the ancestral state at a particular position is A. Let $P[A \rightarrow C]$, $P[A \rightarrow G]$, and $P[A \rightarrow T]$ be the probabilities that a mutation of an A on an external branch of the genealogy results in a C, G, or T, respectively, where the three probabilities sum to 1. Conditional on a site being ancestrally A and that there have been two mutations on external branches, the probability that one of the mutations is to a C and the other to a G is then $P[A \rightarrow C, G] = 2P[A \rightarrow C] \times P[A \rightarrow G]$, and similarly for the other combinations, while the probability that both mutations were to a C (i.e., it is a recurrent doubleton) is $P[A \rightarrow C, C] = P[A \rightarrow C]^2$. This last probability can therefore be expressed as

$$P[A \rightarrow C, C] = \frac{P[A \rightarrow C, G] \, P[A \rightarrow C, T]}{2P[A \rightarrow G, T]} \tag{1}$$

These probabilities can be estimated from the observed occurrences of triallelic singleton sites in the genome, and the estimated number of recurrent $A \rightarrow C$ doubleton sites will be:

$$N_{est}[A \rightarrow C, C] = \frac{N[A \rightarrow C, G] \, N[A \rightarrow C, T]}{2N[A \rightarrow G, T]} \tag{2}$$

where $N[X \to Y, Z]$ is the number of sites in the genome for which the major allele is X with single copies of mutations Y and Z. Finally, if there are $D[A \to C]$ doubleton sites in which the major allele is A and the minor is C, then the estimated proportion that are single origin and RCR is

$$P[RCR|d]_{A \to C} = 1 - \frac{N_{est}[A \to C, C]}{D[A \to C]} \tag{3}$$

Analogous calculations apply for the other 11 types of mutation, and can be used to estimate the prior probability for each one that a doubleton has a single origin and is RCR.

Estimating the probability of having a single origin

Our second goal is to identify a subset of doubletons that are highly likely to have a single origin and are RCR. We do this by updating the prior probability that a doubleton has a single origin using the presence or absence of a linked rare variant (LRV) near the doubleton. The use of $LRVs$ is based on the logic that if the sequences are RCR at the doubleton site, then they are likely to be closely related at linked sites, and therefore have an increased probability of sharing a rare variant. Therefore, the presence of a rare variant should help inform on whether the sequences are RCR. The odds that a doubleton with a linked rare variant is RCR (defined as the probability it is RCR, $P[RCR|d, LRV]$, divided by the probability it is not, $1 - P[RCR|d, LRV]$) is equal to the prior odds that a doubleton is RCR (in the absence of information about linked rare variants) multiplied by the Bayes Factor (BF), which quantifies how much the odds of being RCR change due to the presence of a linked rare variant:

$$PosteriorOdds[RCR|d, LRV] = PriorOdds[RCR|d] \times BF \tag{4}$$

In the Supplementary Information we show that the Bayes Factor can be estimated by calculating the probability that individuals sharing a doubleton have a linked rare variant ($P[LRV|d]$); the probability that random pairs of individuals, which do not share a doubleton at that site, nevertheless have a linked rare variant ($P[LRV|\neg d]$); and the prior probabilities that the doubletons and random pairs are RCR ($P[RCR|d]$ and $P[RCR|\neg d]$, respectively):

$$BF \approx 1 + \frac{P[LRV|d] - P[LRV|\neg d]}{P[RCR|d]\ P[LRV|\neg d] - P[LRV|d]\ P[RCR|\neg d]} \tag{5}$$

Thus we can calculate our probability of interest, that a doubleton with a linked rare variant is RCR, $P[RCR|d, LRV]$, from the prior probability it is RCR, $P[RCR|d]$, estimated as above, and from 3 other probabilities, which can be estimated as follows.

Estimating $P[RCR|\neg d]$

In a sample of m sequences, the probability that 2 of them chosen at random are RCR at a particular site in the absence of a doubleton, $P[RCR|\neg d]$, can be derived by considering the coalescence of lineages back in time. For them to be RCR, they must coalesce with each other before either one does with any other lineage.

Let $P_2[m] = \frac{2}{m(m-1)}$ be the probability that the first coalescence event involves the two sequences, and $P_0[m] = \frac{(m-2)(m-3)}{m(m-1)}$ be the probability that it involves neither of them. If neither sequence is involved in the first coalescence, then the same logic applies to the second coalescent event, except there are now $m-1$ lineages instead of m. The probability that t coalescence events occur before at least one of the two sequences is involved, and that the coalescence involves both sequences (and therefore they are RCR) is

$$P[t] = \left(\prod_{i=0}^{t-1} P_0[m-i]\right)\ P_2[m-t] \tag{6}$$

which can be summed over all possible values of t to get the overall probability they are RCR,

$$P_p[RCR|\neg d] = \sum_{t=0}^{m-3} P[t] = \frac{2}{3(m-1)} \tag{7}$$

where the subscript p emphasises that this expression is for phased (or haploid) data. To avoid the uncertainties associated with statistical phasing of rare variants we analyse unphased diploid genotypes in this paper. For two diploid individuals that do not share a doubleton there are 4 possible combinations of bases that could be RCR, and therefore to get the probability that at a particular site one of the sequences in one individual is RCR with one of the sequences in the other individual we must multiply the above expression by 4:

$$P[RCR|\neg d] = \frac{8}{3(m-1)} \tag{8}$$

$_{145}$ (In principle, for diploid individuals that share a doubleton, it is possible that the doubleton is not

$_{146}$ RCR but one of the other three pairs of bases between individuals is, but this alternative scenario

$_{147}$ will typically be very rare (i.e., $P[RCR|d] >> P[RCR|\neg d]$) and we will not consider it further.)

Estimating $P[LRV|d]$ and $P[LRV|\neg d]$

$_{149}$ The remaining two terms in Eq. (5) are the probabilities of doubletons and random sites having

$_{150}$ a linked rare variant, $P[LRV|d]$ and $P[LRV|\neg d]$, and we must now make precise the notion of

$_{151}$ having a 'linked rare variant'. This could be done in multiple ways. Here we use a method that

$_{152}$ is appropriate for analysing unphased sequence data. In particular, we use the approach described

$_{153}$ by Mathieson and McVean (2014) to define a segment that extends in both directions from a focal

$_{154}$ base (doubleton or otherwise) until an incompatible site is found in which the 2 individuals are

$_{155}$ homozygous for different bases. This typically provides an overestimate of the region shared by

$_{156}$ individuals over which no recombination has occurred. We then ask whether the segment contains

$_{157}$ (besides the focal base, which may be a doubleton), a rare variant that is found only in n sequences

$_{158}$ in the dataset. The probabilities $P[LRV|d]$ and $P[LRV|\neg d]$ can be estimated simply by counting

$_{159}$ the proportion of doubletons and random sites that have a linked rare variant defined in this way.

$_{160}$ We repeat the analysis and calculation of Eq. (4) with different values of n, in order to achieve the

$_{161}$ desired level of confidence in a designation that a doubleton is RCR (e.g., 99% confident).

$_{162}$ Note that sites that are or are not RCR will differ in the probability of having a linked rare

$_{163}$ variant both because the segments defined in this way are expected to be longer in the former than

$_{164}$ the latter and because the density of linked rare variants (i.e., on a per-base-pair basis) will be

$_{165}$ higher. Thus there are 2 contributing factors, which should help the discrimination.

Results

Identifying *RCR* doubletons

Dataset Description

We illustrate our approach using genome sequences from Phase 2 of the *Anopheles gambiae* 1000 Genomes Project (Ag1000G) (The Anopheles gambiae 1000 Genomes Consortium, 2020). We analysed variant data on the 42Mb chromosome arm 3L for 1,142 individuals of the sibling species *An. gambiae* sensu stricto (henceforth *An. gambiae* s.s.) and *An. coluzzii* collected from 13 countries across sub-Saharan Africa. Unphased data were used due to concerns regarding the ability of statistical phasing methods to accurately phase rare variants (Andrés et al., 2007; Snyder et al., 2015; Choi et al., 2018), and 3L was used as there are no reports of common segregating inversions on this chromosome arm (Coluzzi et al., 1985). The data were filtered to remove low quality sites or where any individual had a missing genotype, leaving 10,219,040 sites (see Methods for further details). This dataset contained 1,116,970 doubletons, and for each one we scanned the genome in each direction until finding a site at which the two individuals were homozygous for different bases, resulting in 754,006 unique haplotypes (i.e., some segments carried multiple doubletons) with a median length of 5.5 ± 0.009kb (SE). To calculate $P[RCR|\neg d]$ and $P[LRV|\neg d]$, a dataset of random or "non-doubleton" ($\neg d$) haplotypes was created: for every doubleton we randomly chose two other individuals and, starting from the same site as the doubleton, again scanned out in each direction until finding a site at which the two individuals were homozygous for different bases, resulting in 1,115,755 unique segments. As expected, these randomly chosen haplotypes were considerably shorter, about half the length of those flanking doubletons (median 2.9 ± 0.003 kb).

Estimating the proportion of sites that are *RCR* ($P[RCR|d]$ and $P[RCR|\neg d]$)

A total of 243,668 triallelic singleton sites (where one base is found in all but two sequences and then there are two different singletons at that site) were identified on chromosome 3L (Table 1A). Data from these sites were used along with Eqs. (2 & 3) to estimate the proportion of observed doubletons that are single origin and therefore identify sequences that are *RCR* (Table 1B) (for an example, see Supplementary Information). There is a 5-fold range in the estimated proportion of recurrent doubletons of different types, with C→T and G→A mutations having the highest

proportion ($\sim$25%), and A$\rightarrow$C and T$\rightarrow$G mutations the lowest ($\sim$5%). Overall the estimated proportion of recurrent doubletons is 0.16, thus leaving $P[RCR|d] = 0.84$ doubletons as having a single origin.

To estimate the probability that our random pairs of individuals are RCR at a specified site, we use Eq. (8). With 2,284 haploid genomes in our sample, $P[RCR|\neg d] = 0.0012$. Thus individuals sharing a doubleton are about 700 times more likely to be RCR at that site than a random pair of individuals.

Table 1: (A) Number of triallelic singleton sites[1], and (B) Number of observed doubletons and estimated number of recurrent doubletons[2], on chromosome arm 3L.

(A) Triallelic singletons			(B) Doubletons				
Major allele	Singleton alleles	Count	Major allele	Doubleton allele	Count	Estimated recurrent count (N_{est})	Proportion recurrent (1 - $P[RCR|d]$)
A	C, T	10,620	A	C	40,394	2,002	0.05
A	C, G	8,649	A	G	92,019	9,340	0.09
A	T, G	22,938	A	T	106,529	14,083	0.15
C	A, T	46,232	C	A	69,276	9,923	0.14
C	A, G	10,219	C	G	30,777	2,631	0.09
C	T, G	23,806	C	T	219,184	53,851	0.25
G	A, C	23,608	G	A	220,013	54,534	0.25
G	A, T	45,742	G	C	30,621	2,555	0.08
G	C, T	9,901	G	T	68,860	9,592	0.14
T	A, C	22,846	T	A	107,488	14,095	0.13
T	A, G	10,554	T	C	91,580	9,257	0.10
T	C, G	8,553	T	G	40,229	1,976	0.05
	Total	243,668		Total	1,116,970	183,839	0.16

[1] *E.g., $N[X \rightarrow Y, Z]$*
[2] *Estimated using Eqs.(2) and (3)*

Estimating the probability of sites sharing a linked rare variant ($P[LRV|d]$ and $P[LRV|\neg d]$)

The probability a doubleton has a linked rare variant, $P[LRV|d]$, is given by the proportion of doubletons with at least one rare variant in the flanking segments out to the nearest incompatible sites. Figure 1A shows the fraction of doubletons with an LRV on their flanks as a function of the frequency of the LRV in the dataset (shown both as cumulative counts, where doubletons are included if they have a linked variant present in anywhere from 2 to n copies in the dataset, and as incremental counts). As can be seen, the incremental values trend downwards, with proportions of higher-count $LRVs$ becoming increasingly smaller while the cumulative counts increase at a diminishing rate. This same analysis on the random pair haplotypes shows a relatively flat incremental

plot and the cumulative plot is therefore linear (Fig. 1A). As expected, for all allele counts investigated (up to 100), random pairs are far less likely to have a linked variant than doubletons. Interestingly, random sites are also less likely than doubletons to have a linked singleton site (32% vs. 66%), presumably because the segments out to the nearest incompatible site are longer for doubletons than for random pairs.

Figure 1: (A) Proportion of doubleton (red) or random (blue) pairs sharing a linked rare variant of count f_n, as a function of the rare variant count. These proportions estimate the corresponding probabilities $P[LRV]$. Both groups were randomly downsampled to 750,000 pairs. Note overlap of dashed lines. For both panels, values are shown for variant allele counts of n or less (cumulative; solid lines) and for counts of n but not $< n$ (incremental; dashed lines). (B) Probability that a doubleton with a linked rare variant identifies sequences that are reciprocal closest relatives (RCR; red), and the number of haplotype pairs with at least one RCR doubleton (black), as a function of the allele count of the rare variant f_n. The dashed vertical line shows the f_{18} threshold used to define our high confidence set of RCR doubletons.

Identifying RCR doubletons

The probability that a doubleton with a linked rare variant has a single origin, $P[RCR|d, LRV]$, was calculated using the preceding results and Eq. (4). We used different threshold values of allele count for the rare variant and, as expected, the lower the frequency of the linked mutation, the greater the probability that the associated doubleton had a single origin, but the smaller the number of doubletons that would be included. These opposing trends are shown in Figure 1B. Based on these results we defined our subset of high confidence RCR doubletons as those with linked variants present in 2-18 copies in the dataset (corresponding to a minor allele frequency in the dataset of $< 0.8\%$), as the probability that any additional haplotypes added beyond this are RCR is < 0.95 (Figure 1B; threshold indicated by dashed line). Though doubletons are more likely to have linked singletons than random haplotypes, the presence of a singleton only raises the probability of a doubleton being RCR to 0.92, not enough to be included in our high confidence subset. This threshold resulted in a total of 290,704 unique haplotypes being included, carrying 636,938 doubletons, 57% of the total number called and 68% of the estimated number that are RCR. By comparison, only $P[LRV|\neg d] = 2\%$ of the random pair haplotypes shared a linked variant in this range. The vast majority of individuals (1,113 - or 97%) shared at least one RCR doubleton with another individual. Each individual shared an RCR doubleton with an average of 254 other individuals in the dataset, across 144,542 unique pairs of individuals. As expected, RCR doubletons identified in this way have both longer flanking segments than random pair haplotypes (median: 14.3 ± 0.004 vs. 2.9 ± 0.003 kb; $p < 1e{-}10$, Wilcoxon Rank-Sum test) and a higher density of shared $LRVs$ up to f_{18} (0.3 ± 0.0004 vs. $0 \pm 0 \text{kb}^{-1}$; $p < 1e{-}10$; Fig. 2A). Moreover, doubletons on longer haplotype segments are more likely to be identified as as RCR by our method than those on shorter ones (Fig. 2B).

Figure 2: (A) Haplotype length and density of linked rare variants ($LRVs$; $\leq f_{18}$) on RCR doubleton and random pair haplotypes. Both sets were randomly downsampled to include only 20% of the data to better graphically display differences. Fully 99% of random pair haplotypes had no linked rare variants, constituting the peak at 0. (B) Logistic regression of the probability a doubleton is identified as RCR as a function of haplotype length. Marginal plot shows the distribution of haplotype lengths for RCR and unclassified doubletons.

Assessing RCR classification

As an independent test of the effectiveness of our identification method for high likelihood RCR doubletons, we compared the relative frequencies of the different types of mutational change in doubletons with those observed for singletons. For example, C→T singletons are 3.75 times more frequent in the genome than A→C singletons, and if all doubletons have a single origin we would expect C→T doubletons to also be 3.75 times more frequent than the A→C doubletons, whereas if all doubletons are recurrent then C→T doubletons should be $3.75^2 = 14.1$ times more frequent. A 2 x 12 contingency table analysis comparing the frequencies of the 12 different types of mutational change for singletons and our chosen set of RCR doubletons shows the two sets are significantly different (G=261.41, $p < 2.2e{-}10$), but the differences are very small, with points very close to the 1:1 line (Fig. 3). By contrast, points for the unclassified doubletons (of which 63% are expected to be RCR and 37% recurrent) are closer to the expectation for recurrent mutations.

Figure 3: Expected proportions of different classes of doubleton mutations (e.g., A→C) derived from a single origin (i.e., identity-by-descent; 1:1 solid line) and multiple origins (recurrent mutation; dashed line), together with the observed proportions for doubletons classified as *RCR* and those not classified. Expected proportions were calculated from the observed proportions of singleton mutations.

One would expect that, on filtering to include only *RCR* doubletons, much of the 'noise' in the data would be removed, enhancing genuine biological signals due to shared ancestry. To test this, differences in patterns of doubleton sharing within and between countries were investigated before and after filtering to include only high-confidence *RCR* doubletons (Fig. 4). It would be expected that sharing would be more pronounced between locations in close geographical proximity and between samples from the same species, when considering only *RCR* doubletons. This was broadly observed, with doubleton sharing more pronounced in location-species clusters (e.g., West African *An. coluzzii*) and reduced between more distant locations (Fig. 4B), with these differences highlighted in Figure 4C.

Figure 4: (A) and (B) Doubleton sharing within and between countries, before and after *RCR* filtration. Bar heights are proportional to the number of doubletons shared between individuals of each country combination, and are standardised such that the total height of any given row is fixed. For both panels a random subset of individuals was chosen, equal to the size of the smallest sample (Mayotte, $n = 24$) to remove effects caused by unequal sampling. Colour groupings are determined by both location and species. (C) Change in sharing proportions after *RCR* filtration. Bars are coloured according to direction of change.

Using *RCR* doubletons for population genetic analyses

Having defined a set of high likelihood *RCR* doubletons we now illustrate how they can be used to test population genetic hypotheses involving recombination, selection, and isolation by distance. In many instances these tests involve analysing haplotype lengths, the combined length of the segments from the doubleton out to the first incompatible site on each side.

Recombination

In many species, including *An. gambiae* s.l., rates of recombination are thought to be lower near the centromere (Choo, 1998), leading to the expectation that doubletons near the centromere will be flanked by longer haplotypes than those further away. Figure 5 shows that haplotypes around both *RCR* doubletons and random pairs are indeed about an order of magnitude longer near the centromere compared to the rest of the chromosome, with this difference decaying over the first 20Mb. The local peaks observed along the chromosome are associated with areas of the genome with no coverage, which increase the distance to the nearest detected inconsistent homozygous site if the haplotype ends in this region. As random pair haplotypes represent the expected distance

to an inconsistent homozygous site from a given position in the genome, these should give a rough indicator of the length of the overestimate generated by our approach to haplotype detection. The difference between these and the *RCR* haplotypes, plotted in black, should therefore provide an estimate of the true length of *RCR* haplotypes in a given region.

Figure 5: Effects of chromosome position on the length of haplotypes flanking *RCR* doubletons. Median lengths of *RCR* and random pair haplotypes are shown along chromosome arm 3L, and the difference between the two groups at each position, highlighting elevated haplotype length close to the centromere, a likely effect of reduced recombination. Position is defined by the midpoint of the haplotype span, with each section of the chromosome binned into 10kb regions (points) and then smoothed via a rolling median spanning 100 bins (lines).

Selection

At least some of the doubleton mutations are likely to reduce fitness and be selected against, and the stronger the selection against the mutation, the more recent the mutation is expected to be, and therefore the longer the haplotype (Maruyama, 1974; Kiezun et al., 2013; Mathieson and McVean, 2014). To test whether the data fit these expectations we first compared the length of haplotypes associated with doubletons occurring at zero-fold and four-fold degenerate sites, as mutations in the former are more likely to be deleterious than those in the latter (Li et al., 1985). As expected, haplotypes flanking zero-fold doubletons are significantly longer than those flanking four-fold doubletons (median 20.5 ± 0.33 vs. 16.0 ± 0.30kb, $p < 2.2e{-}10$, Wilcoxon Rank-Sum test), though there is considerable overlap in the distributions (Table 2).

Nucleotide bases that are conserved across species are also more likely to be under stronger

purifying selection than sites that are not conserved, and therefore we used an alignment of sequences across 21 species of *Anopheles* mosquitoes and categorised doubletons as being at (i) non-conserved sites; (ii) completely conserved sites across the 21 species; and (iii) completely conserved sites that are also inside a region of at least 18bp that is also completely conserved. Again, haplotype lengths showed small but statistically significant differences in the expected direction (median 16.6 ± 0.48 vs. 17.7 ± 0.24 vs. 21.4 ± 2.41kb, all $p < 0.05$, Wilcoxon Rank-Sum tests (Table 2).

Table 2: Haplotypes associated with doubletons at positions of different codon degeneracy or nucleotide conservation.

	Number of haplotypes	Mean length $\pm$ SE (kb)	Median length $\pm$ SE (kb)	Length quartiles (lower; upper - kb)
Codon Degeneracy				
Four-fold	5,737	46.7±1.52	16.0±0.30	7.6; 40.1
Zero-fold	9,631	62.6±1.49	20.5±0.33	9.0; 55.4
Nucleotide Conservation				
Non-conserved site[1]	2,893	52.8±2.46	16.6±0.48	7.7; 43.8
Conserved site[2]	16,618	55.3±1.08	17.7±0.24	8.1; 46.9
Conserved region[3]	274	80.2±13.6	21.4±2.41	9.8; 54.6

[1] Not conserved across 21 species of *Anopheles*
[2] Conserved across 21 species of *Anopheles*.
[3] Inside of a region of at least 18 contiguous conserved sites.

Isolation by distance

The sequences in the Ag1000 Phase 2 dataset are from 13 countries across a broad swath of sub-Saharan Africa, spanning distances much larger than the typical dispersal distance of an individual mosquito, and therefore individuals collected near to each other are expected to be more closely related than those collected further apart. This higher relatedness should be manifest both in a higher probability of them sharing an *RCR* doubleton and in the haplotypes flanking those doubletons being longer. The data support these predictions. First, 48% of *RCR* doubletons are shared between individuals from the same country, compared to 21% of the unclassified doubletons, and a random null expectation of 11% ($G = 175,560, p < 2.2e-10$; Fig. 6A). Second, *RCR* doubletons from individuals from the same country have significantly longer flanking haplotypes than those from different countries (Fig. 6B). This difference is found for doubletons in both *An. coluzzii* (median; 25.2 ± 0.23 vs. 15.2 ± 0.14kb; $p < 1e-10$, Wilcoxon Rank-Sum test) and *An. gambiae* s.s. (median; 19.7 ± 0.12 vs. 10.7 ± 0.04kb; $p < 1e-10$, Wilcoxon Rank-Sum test). Finally, considering

only RCR doubletons from individuals from different countries, there is a negative correlation in both species between the geographical distance separating the sampling locations and haplotype length (Fig. 6C). Here, the haplotype length considered is the total length of all unique haplotypes shared between each pair of individuals, to account for some pairs of individuals who share multiple RCR haplotypes with each other (38% of those between countries), often in close genomic proximity. Interestingly, *An. coluzzii* has both longer haplotypes than *An. gambiae* s.s. within countries ($p < 1e-10$, Wilcoxon Rank-Sum test), and a steeper decline in haplotype length with geographical distance between countries ($p = 0.021$). Random pair haplotype lengths also decrease with geographical distance, as expected with isolation by distance, though with a shallower slope.

Figure 6: Relationship between genetic relatedness inferred from RCR haplotypes and geographical distance. (A) Probability of a doubleton being found between individuals within a country (private) for RCR and unclassified doubletons. Also shown is the null expectation if doubletons were randomly distributed among pairs of individuals. Error bars were too small to be displayed on the graph. (B) Distribution of haplotype lengths for doubletons shared within and between countries, for *An. coluzzii* and *An. gambiae* s.s. Significance of pairwise Wilcoxon Rank-Sum tests is displayed. (C) Mean total length of shared RCR haplotypes per pair of individuals, between countries, as a function of the geographical distance between sites for *An. coluzzii* (red solid points and line) and *An. gambiae* s.s. (blue); the slopes of Ln[length] on distance (in km) are -0.00081 ± 0.00014, $R^2 = 0.808, p = 6.1e{-}4$ and -0.00035 ± 0.00008, $R^2 = 0.423$, $p = 4.6e{-}4$, respectively. Also shown are the mean lengths of shared haplotypes for the random pairs of individuals (open circles and dashed lines; slopes 0.00035 ± 0.00008, $R^2 = 0.679, p = 2.1e{-}3$ and 0.00015 ± 0.00002, $R^2 = 0.631, p = 2.7e{-}7$ for *An. coluzzii* and *An. gambiae* s.s., respectively).

Discussion

Identifying *RCR* doubletons

The usefulness of a molecular marker for population genetic inference is in large part a function of how likely it is to have a single origin, and therefore identifies individuals that are genealogically related. Rare variants are of interest because they are more likely than common variants to have a single origin and to have arisen relatively recently, and may therefore allow for finer discrimination of populations (e.g. Gravel et al., 2011; O'Connor et al., 2015). In this paper we have focused on doubletons, the rarest of rare variants that can be informative about relatedness between individuals, and used them to analyse published data from *An. gambiae* s.l. mosquitoes. While it is commonly assumed that their rarity ensures derivation from a single common ancestor, this may not be the case and could confound analyses if not properly accounted for. We first devised an approach for estimating the frequency of doubletons that have single origins, based on the frequency of triallelic singleton sites. We then developed a method to identify a subset of doubletons with a high likelihood (ca. 99%) of being single origin, based on the presence of a linked rare variant. Applying these methods to published genomic data from *An. gambiae* s.l. mosquitoes, we estimated that 5-25% of doubletons are due to recurrent events, depending on the type of mutation, with an overall estimate across all doubletons of $\sim$16%. We then identified a subset of $\sim$637,000 doubletons (out of a total of $\sim$1,100,000) that are highly likely to have single origin, which define $\sim$291,000 unique haplotype pairs. The mutational spectrum of these doubletons matches that of singleton mutations, and an analysis of doubleton sharing within and between countries show enhanced localised intra-species sharing, consistent with the removal of 'noise' due to recurrent mutations, further enhancing confidence in the methodology.

A number of factors are expected to affect the proportion of doubletons that are single origin in a population genomic dataset. A recurrent doubleton occurs if the same mutation arises on two different external branches of the genealogy at a site (and on no internal branch). Under the standard neutral model of constant population size, random mating and no selection, the average or expected total length of the external branches of a genealogy is 4Ne (where Ne is the effective population size) and is independent of sample size (Fu and Li, 1993). Therefore, the probability of there being two mutations on external branches will be higher when the effective population size

is larger and/or the mutation rate is higher, but will be (largely) independent of sample size (see Wakeley et al. (2023) for a more formal treatment). However, if the population size has not been constant, then sample size can have an effect: increasing the sample size reduces the average length of external branches (i.e., their first coalescent event occurs more recently than if the sample size was smaller), and the rate of those coalescent events will be more influenced by recent population size. Thus, in populations that have been expanding (like *An. gambiae* s.l.; (Khatri and Burt, 2019; The Anopheles gambiae 1000 Genomes Consortium, 2020)) the total length of external branches is expected to increase with sample size, and the proportion of recurrent doubletons as well (Wakeley et al., 2023). Note too that a high frequency of sequencing errors could also increase the proportion of (apparent) doubletons that are recurrent. For this study we followed the relatively stringent quality thresholds employed by the Ag1000G Project (2020). To the extent that sequencing errors can be considered an additional source of mutations on external branches, and would lead to triallelic singleton sites as well as doubleton sites, our estimate for the number of single origin doubletons should be largely unaffected.

To identify a subset of doubletons that are highly likely to have a single origin, we developed a method using the formalism of Bayes' Theorem and the different proportions of doubleton haplotypes and matched random pair haplotypes that have a linked rare variant (LRV). We defined 'linked' based on the haplotype detection approach of Mathieson and McVean (2014), and varied the definition of 'rare' to arrive at a desired confidence level. RCR doubletons are expected to differ from matched random pairs of individuals both in the length of their flanking haplotypes and in the density of $LRVs$ found on them, and, by using the presence of an LRV anywhere in the haplotype, our method indirectly combines both features to identify high likelihood RCR doubletons. The only previous method we are aware of for distinguishing between rare variants with single or multiple origins is that proposed by Johnson et al. (2022), which is based on haplotype length alone. Their method also differs in that it requires a demographic model of the study population - specifically assuming a single, exponentially growing population - whereas ours does not.

The median length of our high confidence RCR haplotypes was 14.3kb. Using a genome-wide average recombination rate of 1.8 cM/Mb (Pombi et al., 2006), this translates to about 0.03cM. These values are much shorter than were found in Mathieson and McVean's (2014) analysis of doubleton haplotypes in human data (600kb and 0.7cM), with roughly comparable sample size

and recombination rates. The much smaller values in mosquitoes are consistent with haplotypes being much older (i.e., more time back to the common ancestor), which in turn is consistent with mosquitoes having a much larger population size. Note too that 0.02cM is 100-fold shorter than the limit of confident detection of IBD segments in standard human analyses (e.g., Browning and Browning, 2013) (though see Shetty et al. (2020) who use LRVs to improve confidence for shorter segments). One way to recover longer RCR haplotypes, with more recent common ancestors, would be to increase the sample size as this would increase the relatedness of the closest relatives. In the standard neutral model, if the sample size is doubled, then the average length of external branches is halved, and so is the distance to the nearest relative (Fu and Li, 1993). In *An. gambiae* s.l., where the population has likely been expanding, the effect of doubling the sample size will not be as large, but still one can probably expect more than twice as many detectable haplotypes for downstream analyses.

Analyses using RCR doubletons

Having defined a high confidence subset of single origin doubletons, we then used them to test a series of population genetic hypotheses. We find that RCR doubletons near the centromere are associated with longer haplotypes, as expected if recombination rates there are reduced. Second, RCR doubletons at sites likely to be under purifying selection are associated with longer haplotypes, consistent with them being more recent, a difference also seen in human data (Kiezun et al., 2013; Mathieson and McVean, 2014). And third, the data are consistent with an isolation by distance model: RCR doubletons are more likely than random pairs to be derived from two individuals in the same country; RCR doubleton haplotypes in the same country are longer than those from different countries; and for those shared between countries, the combined lengths of their haplotypes decreases with geographical distance between the locations.

We also found differences in haplotype lengths between the two *An. gambiae* s.l. species. Within countries, haplotypes are longer in *An. coluzzii*, consistent with a smaller average local population size. Haplotypes also shorten more rapidly in *An. coluzzii* with geographical distance, indicating stronger isolation by distance. Previous work has shown that F_{ST} between countries increases more rapidly with geographical distance in *An. coluzzii* than in *An. gambiae* s.s. (The Anopheles gambiae 1000 Genomes Consortium, 2020), an analysis based on a completely different aspect of the same

population genomic data (differences in allele frequencies, independent of linkage). Differences in population differentiation with distance could be due to differences in population density or dispersal distances, with these parameters in principle separable in haplotype-based analyses (Barton et al., 2013). Our analyses appear to suggest that *An. coluzzii* not only has a lower average local population size, but also a shorter average dispersal distance. Shorter dispersal distances for *An. coluzzii* could make sense ecologically, as they are thought to prefer permanent water sources, as opposed to *An. gambiae* s.s. which prefer more ephemeral sources (Kamdem et al., 2012; Akpodiete et al., 2019). Moreover, there is a greater evidence that *An. coluzzii* aestivates in the Sahelian dry season and therefore may not be so reliant on dispersal to persist (Dao et al., 2014). On the other hand, of the two species, *An. coluzzii* has thus far been found more often at high altitude during airborne sampling, where it is thought upper atmospheric currents may facilitate long range dispersal (Huestis et al., 2019).

Future developments

In principle, it is possible that our methods could be modified to recover a greater proportion of high confidence *RCR* doubletons. Our analysis of triallelic data showed that different classes of doubletons have different prior probabilities of having a single origin, but in the subsequent analysis incorporating linked rare variants, for simplicity, we only used the overall prior probability and derived a single threshold frequency. We could instead have used mutation-specific prior probabilities of being single origin, and calculated separate thresholds for each. One could also go beyond this to incorporate differences in mutation rate due to local sequence context (Harris and Pritchard, 2017; DeWitt et al., 2021). A further extension could also take into account the number of linked rare variants (e.g., if the current threshold is f_{18}, maybe having 4 f_{25}'s would also be sufficient). Our method identified 68% of the estimated number of *RCR* doubletons based on the analysis of triallelic sites. It is possible this is an under-estimate, if the expected number was itself an over-estimate, and it is possible that the modifications above might increase the number identified. However, there is presumably an upper limit to the fraction of single origin doubletons that are detectable by testing for linked rare variants - for example, if one copy has been transferred to a different genetic background by a gene conversion event with no other rare variant in the conversion tract, then it will be missed. Such short haplotypes would in any case seem difficult to date (i.e., would contain

little information on the degree of relatedness), and therefore a failure to identify them may not be such a loss.

The analysis could also be extended in different ways. First, to avoid uncertainties associated with statistically phasing the sequence data prior to analysis, we used unphased data (though the method of analysis does mean that, for single origin doubletons, the doubleton and linked rare variant are highly likely to be on the same chromosome). Some studies have found that errors in inference introduced by incorrect imputation can be greater than that introduced by using unphased and therefore less accurately identified haplotypes (Janzen and Miró Pina, 2022). If robust phasing data is available the method could be modified to incorporate it. Second, one could go beyond doubletons to mutations present in 3, 4, or more individuals. One approach may be to do the same analysis for all possible doubletons in a tripleton, and look for congruence in sharing of $LRVs$ — if all pairs of a tripleton are judged as single origin, then that would suggest that the tripleton is a clade. Interestingly, for a fixed mutation rate, the distribution of the number of origins of a rare variant appears to rapidly become independent of the allele count once it is above a relatively small number (Wakeley et al., 2023).

The other main extension is in parameter estimation. We have shown that the length of the haplotype on which an RCR doubleton resides varies in expected ways with chromosomal position, with whether or not the doubleton is likely to be under purifying selection, and with geography. The data thus contains information on recombination rates, fitness differences, and movement, and a worthwhile next step, beyond the scope of this paper, would be to develop the appropriate inferential methods for using these data to estimate the parameters of a model. Such estimates will be especially useful for some purposes because they should reflect relatively recent parameter values. In the study of human population genetics, IBD haplotypes have typically been defined based on more common variants, excluding rare variants (e.g., Browning and Browning, 2013; but see Shetty et al. (2020)), and have been very useful for demographic inference in humans (e.g., Palamara et al., 2012; Barton et al., 2013; Ralph and Coop, 2013; Baharian et al., 2016; Ringbauer et al., 2017). The challenge is that defining IBD haplotypes will be more difficult when the ratio of recombination rate to mutation rate is high, and while in humans this ratio is about 1, in *An. gambiae* s.l. it is about 10 (Nelson et al., 2021). Using rare variants to define haplotypes, as demonstrated here, may be a useful way forwards. Given how much harm is caused by this species (World Health Organisation,

2020), and that population genetic interventions including using gene drive are being actively called for (African Union, 2018; World Health Organisation, 2021) and developed (e.g., Hammond et al., 2021; Hoermann et al., 2022; Carballar-Lejarazú et al., 2023), further investigation of appropriate methods to extract as much information as possible from population genomic data is much to be desired.

Materials and Methods

Data acquisition

Genomic data was acquired from the Phase 2 AR1 data release of the *Anopheles gambiae* 1000 Genomes project, in VCF format, for chromosome arm 3L (https://www.malariagen.net/data/ag1000g-phase-2-ar1). This release consisted of 1,142 wild-caught individuals from 13 countries across sub-Saharan Africa, belonging to the sister species *An. gambiae* s.s. and *An. coluzzii*, who comprise two of the primary species in the *Anopheles gambiae* species complex (*An. gambiae* s.l.). As described in the main text, unphased reads were used due to concerns over phasing of rare variants. In addition to variant calls, sample metadata were also accessed from the data release.

Data filtering

Initial filtering for quality was performed using the pre-prepared PASS filter from Ag1000G. This removed any SNP that was not classified as 'accessible', removing sites subject to poor sequencing quality, allelic dropout, poor alignment, and with problematic repeats and structural variation. Furthermore, it provided additional quality control to account for allelic position on reads, mapping of repeats, strand bias, and coverage relative to depth. For more detail on the exact filters used, see Supplementary Information 3.4 & 3.5 in The Anopheles gambiae 1000 Genomes Consortium (2017). Following quality filtering, SNPs were further filtered to remove any site where there was missing data – that is, every sampled individual had two SNP calls at that position. For haplotype detection, for both doubleton and random pairs of individuals, SNPs were finally filtered to include only those at biallelic sites.

Haplotype detection

Haplotype detection was performing using modified versions of the scripts from Mathieson and McVean (2014), found at https://github.com/mathii/f2. In short, VCFtools (Danecek et al., 2011) was used to identify sites with doubletons and the individuals carrying them; subsequently, the combined homologous chromosomes of each pair were parsed in both directions from the focal doubleton until a site where the two individuals had inconsistent homozygous genomes was encountered, which was inferred to be an estimate of the end of the haplotype. Any doubletons which were found in a single homozygous individual, and those haplotypes reaching either end of the chromosome arm, were discarded. About 5% of haplotypes had directly adjacent start/end points with another haplotype and it is possible these do not represent separate haplotypes, but we did not investigate them further.

For the random pair haplotypes, two individuals were randomly sampled at each position where a doubleton was identified, but that did not carry the doubleton themselves. Shared haplotypes were then identified in the same manner as for doubleton pairs. For both groups, lengths were identified as the difference between the positions of the two inconsistent homozygous genotypes.

Allele counting

To determine what linked variants lay on each shared haplotype, VCFtools was used to identify the position and individuals carrying variants of counts from 3 to 100 in the sample. For each haplotype pair, the number of variants of each count shared by both individuals, within the bounds of the shared haplotypes, was recorded. Density of linked rare variants was calculated using the number of variants of counts 2-18 (our threshold value) shared on each haplotype, divided by haplotype length. Counting of singletons, doubletons, and triallelic singletons of each specific base change was performed using scikit-allel (Miles et al., 2021).

Site classifications

Positional filters to determine the fold degeneracy of sites in the genome were generated using scripts supplied by Alistair Miles (unpublished), and then applied to doubleton sites. Filters to determine the conservation of sites (non-conserved, conserved site, conserved region) were supplied by Saman-

tha O'Loughlin (personal communication). Conserved sites were identified via the alignment of 21 species within the *Anopheles* genus, with conserved regions referring to those sites which lie in contiguous segments of conserved sites 18bp or more in length (O'Loughlin et al., 2021). For both degeneracy and conservation, where there were multiple doubletons on a single haplotype, each doubleton was considered its own entity.

Isolation by distance

Geographical distances between sample sites were calculated using associated latitude-longitude coordinates from Ag1000G sample metadata. Where collections had been performed at multiple sites in an area, an average position of these coordinates was used. For comparisons between different countries, average total haplotype lengths were calculated on a per unique pair basis – that is, for every pair of individuals that shared a haplotype, the length of all of the haplotypes they shared between them were combined, to give a total length of shared haplotype. Pairs of countries for which there were <10 unique pairings of individuals with shared haplotypes were excluded from further analysis. Pairwise comparisons of haplotype lengths were computed using pairwise Wilcoxon Rank-Sum tests with a Holm-Bonferroni correction. Difference in slope was tested via the interaction term of a linear regression model.

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References

African Union. 2018. Gene Drives for Malaria Control and Elimination in Africa — AUDA-NEPAD.

Akpodiete NO, Diabate A, Tripet F. 2019. Effect of water source and feed regime on development and phenotypic quality in Anopheles gambiae (s.l.): Prospects for improved mass-rearing techniques towards release programmes. Parasites & Vectors. 12:210.

Andrés AM, Clark AG, Shimmin L, Boerwinkle E, Sing CF, Hixson JE. 2007. Understanding the accuracy of statistical haplotype inference with sequence data of known phase. Genetic Epidemiology. 31:659–671.

Baharian S, Barakatt M, Gignoux CR, Shringarpure S, Errington J, Blot WJ, Bustamante CD, Kenny EE, Williams SM, Aldrich MC, Gravel S. 2016. The Great Migration and African-American Genomic Diversity. PLOS Genetics. 12:e1006059.

Barton NH, Etheridge AM, Kelleher J, Véber A. 2013. Inference in two dimensions: Allele frequencies versus lengths of shared sequence blocks. Theoretical Population Biology. 87:105–119.

Browning BL, Browning SR. 2013. Detecting Identity by Descent and Estimating Genotype Error Rates in Sequence Data. The American Journal of Human Genetics. 93:840–851.

Carballar-Lejarazú R, Dong Y, Pham TB, Tushar T, Corder RM, Mondal A, Sánchez C HM, Lee HF, Marshall JM, Dimopoulos G, James AA. 2023. Dual effector population modification gene-drive strains of the African malaria mosquitoes, Anopheles gambiae and Anopheles coluzzii. Proceedings of the National Academy of Sciences. 120:e2221118120.

Chen W, Coombes BJ, Larson NB. 2022. Recent advances and challenges of rare variant association analysis in the biobank sequencing era. Frontiers in Genetics. 13.

Choi Y, Chan AP, Kirkness E, Telenti A, Schork NJ. 2018. Comparison of phasing strategies for whole human genomes. PLOS Genetics. 14:e1007308.

Choo KHA. 1998. Why Is the Centromere So Cold? Genome Research. 8:81–82.

Clarkson CS, Miles A, Harding NJ, O'Reilly AO, Weetman D, Kwiatkowski D, Donnelly MJ, Consortium TAgG. 2021. The genetic architecture of target-site resistance to pyrethroid insecticides in the African malaria vectors Anopheles gambiae and Anopheles coluzzii. Molecular Ecology. 30:5303–5317.

Coluzzi M, Petrarca V, di Deco MA. 1985. Chromosomal inversion intergradation and incipient speciation in Anopheles gambiae. Bollettino di zoologia. 52:45–63.

Cubry P, Vigouroux Y, François O. 2017. The Empirical Distribution of Singletons for Geographic Samples of DNA Sequences. Frontiers in Genetics. 8.

Danecek P, Auton A, Abecasis G, Albers CA, Banks E, DePristo MA, Handsaker RE, Lunter G, Marth GT, Sherry ST, McVean G, Durbin R, 1000 Genomes Project Analysis Group. 2011. The variant call format and VCFtools. Bioinformatics. 27:2156–2158.

Dao A, Yaro AS, Diallo M, Timbiné S, Huestis DL, Kassogué Y, Traoré AI, Sanogo ZL, Samaké D, Lehmann T. 2014. Signatures of aestivation and migration in Sahelian malaria mosquito populations. Nature. 516:387–390.

DeWitt WS, Harris KD, Ragsdale AP, Harris K. 2021. Nonparametric coalescent inference of mutation spectrum history and demography. Proceedings of the National Academy of Sciences of the United States of America. 118:e2013798118.

Fu YX, Li WH. 1993. Statistical tests of neutrality of mutations. Genetics. 133:693–709.

Gravel S, Henn BM, Gutenkunst RN, Indap AR, Marth GT, Clark AG, Yu F, Gibbs RA, 1000 Genomes Project, Bustamante CD. 2011. Demographic history and rare allele sharing among human populations. Proceedings of the National Academy of Sciences of the United States of America. 108:11983–11988.

Haldane JBS. 1933. The Part Played by Recurrent Mutation in Evolution. The American Naturalist. 67:5–19. 2457127.

Hammond A, Pollegioni P, Persampieri T, North A, Minuz R, Trusso A, Bucci A, Kyrou K, Morianou I, Simoni A, Nolan T, Müller R, Crisanti A. 2021. Gene-drive suppression of mosquito populations in large cages as a bridge between lab and field. Nature Communications. 12:4589.

Harris K, Pritchard JK. 2017. Rapid evolution of the human mutation spectrum. eLife. 6:e24284.

Hoermann A, Habtewold T, Selvaraj P, Del Corsano G, Capriotti P, Inghilterra MG, Kebede TM, Christophides GK, Windbichler N. 2022. Gene drive mosquitoes can aid malaria elimination by retarding Plasmodium sporogonic development. Science Advances. 8:eabo1733.

Huestis DL, Dao A, Diallo M, Sanogo ZL, Samake D, Yaro AS, Ousman Y, Linton YM, Krishna A, Veru L, Krajacich BJ, Faiman R, Florio J, Chapman JW, Reynolds DR, Weetman D, Mitchell R, Donnelly MJ, Talamas E, Chamorro L, Strobach E, Lehmann T. 2019. Windborne long-distance migration of malaria mosquitoes in the Sahel. Nature. 574:404–408.

Janzen T, Miró Pina V. 2022. Estimating the time since admixture from phased and unphased molecular data. Molecular Ecology Resources. 22:908–926.

Jenkins PA, Song YS. 2011. The effect of recurrent mutation on the frequency spectrum of a segregating site and the age of an allele. Theoretical Population Biology. 80:158–173.

Johnson KE, Adams CJ, Voight BF. 2022. Identifying rare variants inconsistent with identity-by-descent in population-scale whole-genome sequencing data. Methods in Ecology and Evolution. 13:2429–2442.

Kamdem C, Fossog BT, Simard F, Etouna J, Ndo C, Kengne P, Boussès P, Etoa FX, Awono-Ambene P, Fontenille D, Antonio-Nkondjio C, Besansky NJ, Costantini C. 2012. Anthropogenic Habitat Disturbance and Ecological Divergence between Incipient Species of the Malaria Mosquito Anopheles gambiae. PLOS ONE. 7:e39453.

Karlin S, McGregor J. 1967. The number of mutant forms maintained in a population. Proceedings of the Fifth Berkeley Symposium on Mathematical Statistics and Probability, Volume 4: Biology and Problems of Health. 5.4:415–439.

Khatri BS, Burt A. 2019. Robust Estimation of Recent Effective Population Size from Number of Independent Origins in Soft Sweeps. Molecular Biology and Evolution. 36:2040–2052.

Kiezun A, Pulit SL, Francioli LC, van Dijk F, Swertz M, Boomsma DI, van Duijn CM, Slagboom PE, van Ommen GJB, Wijmenga C, Consortium GotN, de Bakker PIW, Sunyaev SR. 2013. Deleterious Alleles in the Human Genome Are on Average Younger Than Neutral Alleles of the Same Frequency. PLOS Genetics. 9:e1003301.

Kimura M. 1971. Theoretical foundation of population genetics at the molecular level. Theoretical Population Biology. 2:174–208.

Kimura M, Crow JF. 1964. The Number of Alleles That Can Be Maintained in a Finite Population. Genetics. 49:725–738.

Kreiner JM, Sandler G, Stern AJ, Tranel PJ, Weigel D, Stinchcombe JR, Wright SI. 2022. Repeated origins, widespread gene flow, and allelic interactions of target-site herbicide resistance mutations. eLife. 11:e70242.

Lachance J. 2010. Disease-associated alleles in genome-wide association studies are enriched for derived low frequency alleles relative to HapMap and neutral expectations. BMC Medical Genomics. 3:57.

Li WH, Wu CI, Luo CC. 1985. A new method for estimating synonymous and nonsynonymous rates of nucleotide substitution considering the relative likelihood of nucleotide and codon changes. Molecular Biology and Evolution. 2:150–174.

Maruyama T. 1974. The age of a rare mutant gene in a large population. American Journal of Human Genetics. 26:669–673.

Mathieson I, McVean G. 2012. Differential confounding of rare and common variants in spatially structured populations. Nature Genetics. 44:243–246.

Mathieson I, McVean G. 2014. Demography and the Age of Rare Variants. PLOS Genetics. 10:e1004528.

Messer PW, Petrov DA. 2013. Population genomics of rapid adaptation by soft selective sweeps. Trends in Ecology & Evolution. 28:659–669.

Miles A, io bot p, R M, Ralph P, Harding N, Pisupati R, Rae S, Millar T. 2021. Cggh/scikit-allel: V1.3.3. Zenodo. doi:10.5281/zenodo.4759368.

Nelson JT, Cornejo OE, Consortium A. 2021. Evolutionary implications of recombination differences across diverging populations of Anopheles. doi:10.1101/2021.02.04.429659.

Nelson MR, Wegmann D, Ehm MG, Kessner D, Jean PS, Verzilli C, Shen J, Tang Z, Bacanu SA, Fraser D, Warren L, Aponte J, Zawistowski M, Liu X, Zhang H, Zhang Y, Li J, Li Y, Li L, Woollard P, Topp S, Hall MD, Nangle K, Wang J, Abecasis G, Cardon LR, Zöllner S, Whittaker

JC, Chissoe SL, Novembre J, Mooser V. 2012. An Abundance of Rare Functional Variants in 202 Drug Target Genes Sequenced in 14,002 People. Science. 337:100–104.

Novembre J, Slatkin M. 2009. Likelihood-Based Inference in Isolation-by-Distance Models Using the Spatial Distribution of Low-Frequency Alleles. Evolution. 63:2914–2925.

O'Connor TD, Fu W, NHLBI GO Exome Sequencing Project, ESP Population Genetics and Statistical Analysis Working Group ET, Mychaleckyj JC, Logsdon B, Auer P, Carlson CS, Leal SM, Smith JD, Rieder MJ, Bamshad MJ, Nickerson DA, Akey JM. 2015. Rare Variation Facilitates Inferences of Fine-Scale Population Structure in Humans. Molecular Biology and Evolution. 32:653–660.

O'Loughlin SM, Forster AJ, Fuchs S, Dottorini T, Nolan T, Crisanti A, Burt A. 2021. Ultraconserved sequences in the genomes of highly diverse Anopheles mosquitoes, with implications for malaria vector control. G3 Genes—Genomes—Genetics. 11:jkab086.

Palamara PF, Lencz T, Darvasi A, Pe'er I. 2012. Length Distributions of Identity by Descent Reveal Fine-Scale Demographic History. The American Journal of Human Genetics. 91:809–822.

Pombi M, Stump AD, Della Torre A, Besansky NJ. 2006. Variation in recombination rate across the X chromosome of Anopheles gambiae. The American Journal of Tropical Medicine and Hygiene. 75:901–903.

Ralph P, Coop G. 2013. The Geography of Recent Genetic Ancestry across Europe. PLOS Biology. 11:e1001555.

Ringbauer H, Coop G, Barton NH. 2017. Inferring Recent Demography from Isolation by Distance of Long Shared Sequence Blocks. Genetics. 205:1335–1351.

Shetty AC, Consortium NTOfPMT, Group TPGW, O'Connell J, Mitchell BD, O'Connor TD. 2020. Rare variant enriched identity-by-descent enables the detection of distant relatedness and older divergence between populations. doi:10.1101/2020.05.05.079541.

Slatkin M. 1985. Rare Alleles as Indicators of Gene Flow. Evolution. 39:53–65.

Smadi C. 2017. The Effect of Recurrent Mutations on Genetic Diversity in a Large Population of Varying Size. Acta Applicandae Mathematicae. 149:11–51.

Snyder MW, Adey A, Kitzman JO, Shendure J. 2015. Haplotype-resolved genome sequencing: Experimental methods and applications. Nature Reviews Genetics. 16:344–358.

The Anopheles gambiae 1000 Genomes Consortium. 2017. Genetic diversity of the African malaria vector Anopheles gambiae. Nature. 552:96–100.

The Anopheles gambiae 1000 Genomes Consortium. 2020. Genome variation and population structure among 1142 mosquitoes of the African malaria vector species Anopheles gambiae and Anopheles coluzzii. Genome Research. 30:1533–1546.

Wakeley J. 2009. Coalescent Theory: An Introduction. Greenwood Village, Colo. :. Roberts & Co. Publishers,. c2009.. xii, 326 p. :.

Wakeley J, Fan WTL, Koch E, Sunyaev S. 2023. Recurrent mutation in the ancestry of a rare variant. Genetics. 224:iyad049.

World Health Organisation. 2020. World Malaria Report 2020.

World Health Organisation. 2021. Guidance framework for testing genetically modified mosquitoes, second edition.