1 **Genetic variation at the *Cyp6m2* putative insecticide resistance locus in**

2 ***Anopheles gambiae* and *Anopheles coluzzii***

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21 Abstract

22 Background

23 The emergence of insecticide resistance is a major threat to malaria control programmes in Africa,
 24 with many different factors contributing to insecticide resistance in its vectors, *Anopheles* mosquitoes.
 25 *CYP6M2* has previously been recognized as an important candidate in cytochrome P450-mediated
 26 detoxification in *Anopheles* mosquitoes. As it has been implicated in resistance against pyrethroids,
 27 organochlorines and carbamates, its broad metabolic activity makes it a potential agent in insecticide
 28 cross-resistance. Currently, allelic variation within the *Cyp6m2* gene remains unknown.

29 Results

30 Here, we use Illumina whole-genome sequence data from Phase 2 of the *Anopheles gambiae* 1000
 31 Genomes Project (Ag1000G) to examine genetic variation in the *Cyp6m2* gene across 16 populations
 32 in 13 countries comprising *Anopheles gambiae* and *Anopheles coluzzii* mosquitoes. We find 15
 33 missense biallelic substitutions at high frequency (defined as >5% frequency in one or more
 34 populations), that fall into five distinct haplotype groups that carry the main high frequency variants:
 35 A13T, D65A, E328Q, Y347F, I359V and A468S. We examine whether these alleles show evidence of
 36 selection either through potentially modified enzymatic function or by being linked to variants that
 37 change the transcriptional profile of the gene. Despite consistent reports of *Cyp6m2* upregulation and
 38 metabolic activity in insecticide resistant Anophelines, we find no evidence of directional selection
 39 occurring on these variants or on the haplotype clusters in which they are found.

40 Conclusion

41 Our results imply that emerging resistance associated with *Cyp6m2* is potentially driven by distant
 42 regulatory loci such as transcriptional factors rather than by its missense variants, or that other genes
 43 are playing a more significant role in conferring metabolic resistance.

44 Keywords

45 Mosquito; *Cyp6m2*; metabolic resistance; allelic variants; selection

46 Background

47 Malaria remains a pernicious public health problem that plagues the African region, which has over
48 90% of the world's malaria cases and deaths [1]. Although concerted vector control interventions such
49 as long lasting insecticidal nets (LLINs) and indoor residual spraying (IRS) have led to the attainment
50 of key milestones, global progress has stagnated and case numbers are stable or on the rise in many
51 countries in Africa [1-3]. This is due to multiple factors, including the emergence of insecticide
52 resistance, which threaten the effectiveness of vector control interventions [4].

53 The most well understood mechanisms of insecticide resistance are classified into two main functional
54 categories depending on the underlying genes involved: target-site insensitivity and metabolic
55 sequestration and detoxification. Both types may occur concurrently within a single population or even
56 within a single mosquito [5-7]. These mechanisms have led to increasing resistance to all four
57 common insecticide classes — pyrethroids, organochlorines, carbamates and organophosphates — in
58 all major malaria vectors across Africa [7, 8].

59 Metabolic detoxification occurs mainly through the elevated activity of large and functionally diverse
60 multigene enzyme families: glutathione S-transferases (GSTs), carboxylesterases (COEs) and
61 cytochrome P450 monooxygenases (P450s) [7, 9]. Although a few candidates in these enzyme
62 families have been directly associated with resistance, our understanding of metabolic resistance has
63 lagged far behind that of pyrethroid target-site resistance, chiefly due to its complexity and the lack of
64 associated causal mutations [10]. This is despite the fact that metabolic resistance is often considered
65 a greater threat to mosquito control [9], especially since the only widely accepted occurrence of
66 malaria vector control failure was attributed to the elevated expression of resistance-associated
67 P450s in *An. funestus* [11-13]. A comprehensive understanding of metabolic resistance must
68 therefore involve disambiguating the roles that individual enzymes play and the genetic backgrounds
69 that underlie their significance in vector populations.

70 The CYP6M2 enzyme exhibits complex insecticide metabolism associated with multiple binding
71 modes for insecticides [14]. Its gene is located within a cluster of 14 Cyp6 P450 genes on
72 chromosome 3R of *An. gambiae* [15], and is among the 111 known P450 genes across the *An.*

73 *gambiae* genome [16, 17]. In this genomic region, *Cyp6m2* is nested within a sub-cluster of P450s
74 containing *Cyp6m3* and *Cyp6m4* which have also been associated with xenobiotic detoxification[18].

75 *Cyp6m2* is notably one of the few specific P450s that have shown a consistent association with
76 metabolic resistance [5]. Metabolic resistance is mainly assessed through transcriptional profiling of
77 genes involved in xenobiotic detoxification. Transcriptomic experiments such as quantitative PCR and
78 microarray assays have established a link between *Cyp6m2* overexpression and the resistance
79 phenotype in field populations of *An. gambiae*, *An. coluzzii*, *An. arabiensis* and *An. sinensis*,
80 irrespective of the presence of knock-down resistance (*kdr*) mutations such as L995F or L995S in the
81 voltage gated sodium channel (VGSC) [5, 19-21]. In DDT resistant *An. gambiae* in Ghana, *Cyp6m2*
82 has been found to be overexpressed 3.2 to 5.2-fold in combination with the upregulation of additional
83 P450s like *Cyp6z2* [18]. In DDT resistant *An. coluzzii* collected in Benin, *Cyp6m2* was also found to
84 be overexpressed 1.2 to 4.6-fold in combination with *Gste2* from the *GST* gene family and in the
85 presence of fixed *kdr* alleles in the *Vgsc* gene [22]. In Nigeria, the 2.4 to 2.7-fold upregulation of
86 *Cyp6m2* was found to be associated with high levels of permethrin resistance [5] and *An. gambiae*
87 that exhibited a strong resistance to bendiocarb in Côte d'Ivoire also had an elevated (up to 8-fold)
88 expression of the *Cyp6m2* gene [20]. In the same study, transgenic expression of *Cyp6m2* in
89 *Drosophila melanogaster* was shown to produce resistance to both DDT and bendiocarb. *In vivo*
90 functional analysis of multi-tissue overexpression induced by genetic modification has also shown
91 *Cyp6m2* to be sufficient in conferring resistance to permethrin and deltamethrin [23]. However, this
92 overexpression also increased the mosquitoes' susceptibility to the organophosphate malathion.
93 Collectively, these studies indicate that *Cyp6m2* can confer metabolic resistance against insecticides
94 in 3 of the 4 known classes: both type I and type II pyrethroids [14, 18, 23], organochlorines [24], and
95 carbamates [20]. It therefore has a high potential for cross-resistance, which may make the problem
96 of malaria vector control even more intractable by limiting the options available to malaria control
97 programs for insecticide rotation or combination. The negative cross-resistance associated with
98 malathion hereby points to potential mitigating strategies [23].

99 The frequent association of *Cyp6m2* with insecticide resistance described above warrants further
100 investigation into whether there is evidence of copy number variation (CNV) or missense mutations at
101 the locus. CNVs have been implicated in augmenting gene dosage leading to increased transcription

of metabolic enzymes [25, 26]. A genome-wide CNV analysis conducted on the Ag1000G dataset and described in detail elsewhere [25] found CNVs to be significantly enriched in metabolic resistance-associated gene families and to be undergoing positive selection. These CNVs were identified across P450s (such as *Cyp9k1* and at both the *Cyp6z3–Cyp6z1* and the *Cyp6aa1–Cyp6p2* gene clusters) and GSTs (at the *Gstu4–Gste3* cluster). However, CNVs across the *Cyp6m2* locus were found to be rare, even in populations that are known to exhibit *Cyp6m2*-mediated resistance [25]. This indicates that CNVs alone are not sufficient to explain the widespread occurrence of the *Cyp6m2*-associated resistance phenotype: additional factors such as allelic variation might contribute to resistance associated with *Cyp6m2* activity.

Allelic variation can play an additional role in P450-mediated resistance by modifying either enzyme catalytic activity or gene expression levels [27]. Allelic variation has been shown to be key in inducing high metabolic efficiency of *Cyp6P9b* and in conferring metabolic resistance to *An. funestus* [28]. Allelic variants in metabolic genes have also been identified to reliably and reproducibly associate with resistance, such as in *Cyp4J5* and *Coeae1d* in *An. gambiae*, and can serve as diagnostic markers of phenotypic resistance [29]. However, there is still a paucity of information about allelic variation associated with metabolic resistance when compared to the well-characterized target-site mutations [29]. Mutations that may modulate metabolic resistance by either altering function or modifying expression in *Cyp6m2* are yet to be described.

Following the consistent association of *Cyp6m2* with insecticide resistance in many populations, we examine whole-genome Illumina sequence data from phase 2 of the *Anopheles gambiae* 1000 Genomes Project (Ag1000G) [30] which consists of 1,142 wild-caught mosquitoes sequenced to a mean depth above 14x, and report a comprehensive analysis of genetic variation within the *Cyp6m2* gene. We also examine the wider haplotypes around *Cyp6m2* spanning across the *Cyp6m* sub cluster and the larger *Cyp6* supercluster for signatures of selection.

Results

Cyp6m2 non-synonymous nucleotide variation

Short-read whole-genome sequence data from the Ag1000G phase 2 data resource [30] were used to investigate genetic variation at the *Cyp6m2* locus across 16 populations of *An. gambiae* and *An.*

coluzzii (n = 1,142 total individuals) collected between 2000 and 2012 [Table 1, Additional file 1]. The single nucleotide polymorphisms (SNPs) we studied here were discovered and QC'd using methods described elsewhere [31]. We focused on SNPs that change the amino acid sequence of the *CYP6M2* enzyme as they have a potential functional role in *Cyp6m2*-associated insecticide resistance (n = 193) [Additional file 2]. As putative resistance variants under selection pressure from insecticides are expected to increase in frequency over time, we subsequently computed allele frequencies for every non-synonymous SNP in each population with reference to species and country of origin. We filtered the list to focus only on those variants that were at high frequency within populations or across populations (defined as >5% frequency in one or more populations). In total, this resulted in 15 non-synonymous variants that we further explored [Table 1].

Analysis of the patterns of polymorphism of *Cyp6m2* from different populations showed both relative homogeneity within some geographical regions and distinct variants across different regions. The variants with the highest overall frequency were **I359V** (16%) and **D65A** (6%) [Table 1]. The most widespread variant was **I359V**, which was present in West, Central and East African populations of both *An. gambiae* and *An. coluzzii*. Populations with the highest frequency of **I359V** were Gabon (49%) and Ghana (25%) for *An. gambiae*, and Guinea (37.5%) for *An. coluzzii*. Another mutation, **E328Q**, was found across West Africa's *An. coluzzii* populations in Burkina Faso, Côte d'Ivoire, Ghana, Guinea and The Gambia and ranged in frequency from 6.2 to 13.6%. Several variants were found to exceed the 5% threshold only in one or two populations: **A13T** and **Y347F**, in Angola's *An. coluzzii* (39.7%) and in Kenya (52.1%) respectively and **D65A** only in Gabon's *An. gambiae* and in Kenya's populations at 42.8% and 52.1%, respectively [Table 1].

Table1. Allele frequencies of common *Cyp6m2* variants.

Variants			Population allele frequency (%)																
Positio		Ag ²	All	AOco	GHc	BFc	CIc	GNc	G	G	CMga	GHga	BFga	GNga	GAga	UGga	GQga	FRga	KE
n ¹				I ³	ol	ol	ol	ol	W	M	m	m	m	m	m	m	m	m	
69289	G>	A13T	2.8	39.7	0	0.7	0	0	0	0	0	0	0	0	0	0	0	0	0
45	A																		
69290	G>	G47R	0.6	9	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
47	C																		
69290	T>	S48T	2.8	9	0.9	4	2.1	0	3.3	2.	3.2	4.2	2.7	2.5	0	2.2	0	0	0
50	A									3									

69291	A>	D65	6.0	0	0	0	0	0	0.5	0	1.9	4.2	2.7	3.8	42.8	2.7	0	0	52.
02	C	A																	1
69293	A>	K156	0.4	0	0	0	0	0	0	0	0	0	0	0	1.4	0	44.4	0	0
75	T	I																	
69295	A>	N20	0.4	0	0	0	0	0	0	0	0	0	0	0	1.4	0	44.4	0	0
06	G	0D																	
69297	T>	S288	1.1	0	5.5	0.7	0.7	0	0	0	0.8	4.2	2.7	1.3	0.7	1.3	0	0	0
70	A	T																	
69298	G>	E325	1.2	0	7.3	3.3	2.1	0	3.8	2.	0	0	0.5	0	0	0	0	0	0
81	A	K								3									
69298	G>	E328	2.9	0.6	13.6	7.3	7	12.5	9.9	6.	0.3	0	0	0	0	0	0	0	0
90	C	Q								2									
69299	A>	Y347	2.4	0	0	0	0	0	0	0	0.2	0	0	0	0	1.8	0	0	52.
48	T	F																	1
69299	A>	I359	16.	0	4.5	0	7	37.5	16.	20	19.4	25	17.4	16.3	48.6	19.6	0	0	0
83	G	V	0						5										
69302	C>	P407	0.7	0	2.7	0.7	7	0	0	0	0.2	0	0	0	0	0	0	0	0
06	T	L																	
69302	A>	E419	0.4	5.1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
42	T	V																	
69302	A>	K428	2.0	0	0	0.7	0	0	0.5	0.	3.5	0	2.7	2.5	0.7	6.3	0	0	0
69	G	R							8										
69303	G>	A468	2.6	0	0	0	0	0	0	0	0.3	0	0	0	42	0	0	0	0
88	T	S																	

¹Position relative to the AgamP4 reference sequence, chromosome 3R.

²Codon numbering according to *Anopheles gambiae* AGAP008212-RA transcript in geneset AgamP4.12.

³AOcol=Angola *coluzzii*; GHcol=Ghana *coluzzii*; BFcol=Burkina Faso *coluzzii*; CIcol=Côte d'Ivoire *coluzzii*; GNcol=Guinea *coluzzii*; GW=Guinea Bissau; GM = The Gambia; CMgam=Cameroon; GHgam = Ghana *gambiae*; BFgam = Burkina Faso *gambiae*; NGam = Guinea *gambiae*; GAgam=Gabon *gambiae*; UGgam=Uganda *gambiae*; GQgam = Equatorial Guinea *gambiae*; FRgam=Mayotte *gambiae*; KE=Kenya.

Haplotypic backgrounds of non-synonymous alleles

The Ag1000G data resource contains data that not only spans across exonic regions of any given gene, but also intronic and intergenic regions. This enables a comprehensive analysis of haplotypes that contain putative insecticide resistance alleles, but is constrained by the fact that this resource does not contain samples whose resistance status or *Cyp6m2* expression levels are known. Selection pressure acting upon missense variants or linked cis regulatory variants is likely to affect the haplotype structure of the gene. To study haplotype structure at *Cyp6m2*, we extracted biallelic SNPs across the entire 1689bp *Cyp6m2* gene to calculate the number of SNP differences between all pairs of 2,284 haplotypes derived from the mosquitoes. We identified a clustering threshold of seven SNPs

where the haplotype clusters corresponded to the haplotypes carrying the high frequency alleles [Table 1, Figure 1]. We found that these haplotypes could mostly be grouped into five distinct clusters (labelled C1-C5): C1 contained haplotypes carrying **A13T**; C2 contained most haplotypes carrying **D65A**, **A468S**, and some haplotypes carrying **I359V**; C3 contained most haplotypes carrying both **D65A** and **Y347F**, and C5 contained haplotypes carrying **E328Q**. C4 contained haplotypes with no signature missense mutation [Figure 1].

Figure 1. Hierarchical clustering of *Cyp6m2* haplotypes.

Top: a dendrogram showing hierarchical clustering of haplotypes derived from wild-caught mosquitoes. The colour bar indicates the population of origin for each haplotype. Bottom: high frequency (>5%) alleles identified within each haplotype (white = reference allele; black = alternative allele). The lowest margin labels the major haplotype clusters.

Overall, haplotype cluster distribution resembled the whole genome groupings of individuals described elsewhere using our dataset [30]: Cluster C5 contained haplotypes from West African *An. coluzzii*; C4 contained *An. gambiae* from West, Central and near-East Africa; and the rest of the clusters contained haplotypes from samples from a single country and species [Figure 2]. The variation across the haplotypes largely showed no strict or systematic difference between the two species or across broad geographic regions, which is in line with recent whole genome sequencing reports [31].

Figure 2. Map of haplotype cluster frequencies and distribution.

Each pie chart indicates the haplotype group frequencies within specific sampling populations. The sizes of the wedges within the pies are proportional to haplotype group frequencies within the populations. Haplotypes in group C1 carry the **A13T** allele. Haplotypes in group C2 carry **D65A**, **I359V** and **A468S** alleles. Haplotypes in group C3 carry **D65A** and **Y347F** alleles. Haplotypes in group C5 carry the **E328Q** allele. Haplotypes in group C4 had no defining non-synonymous variant, and wild type (wt) haplotypes were all those that did not fall within the C1-C5 clusters.

We investigated patterns of association among these non-synonymous variants by computing the normalized coefficient of linkage disequilibrium (D') using haplotypes from the Ag1000G phase 2 resource. Of the two highest frequency variants, **I359V** was found to be in perfect linkage with **A468S** but this was driven only by one population (Gabon) with most backgrounds carrying **I359V** not

showing linkage with any other missense mutations [Figure 1 & Supplementary Figure 1]. **D65A** was in perfect linkage with **A468S** and **Y347F**, showing that **D65A** was almost only ever found on haplotypes carrying either **A468S** or **Y347F**. **I359V** and **D65A**, the highest frequency mutations across all populations, were found to be only in moderate linkage disequilibrium (0.36) [Supplementary Figure 1]. Other variants were found to be in weak linkage disequilibrium with the six main high frequency alleles and segregated independently within their own populations. While we observed some strong associations through linkage disequilibrium analysis across all populations, a deeper investigation revealed that these associations were driven by population specific dynamics in populations (such as Kenya) where we know bottlenecking has been an issue [31]. It is therefore unlikely that the identified variants are conferring some selective advantage against existing insecticide pressures.

We next explored whether the surrounding genomic region showed a similar hierarchical clustering pattern to *Cyp6m2*, which might be indicative of either dominant demographic effects or selection acting at other linked loci that is having a major impact on variation within *Cyp6m2*. The downstream genes we selected coded for proteins that were 1-to-1 orthologs with *D. melanogaster* genes. We selected *ODR2* [32], *HAM* [33] and *SH2* [34], which were 81280 bases, 457164 bases and 1198636 bases downstream of the *Cyp6m2* gene respectively. The distinctive haplotype clustering pattern observed for *Cyp6m2* in the Kenya, Angola and Gabon populations persisted across these genes, indicating that in these populations, the diversity reduction in and downstream of *Cyp6m2* is more likely driven by demography rather than by a selective sweep [Supplementary Fig. 2-4]. We also extracted biallelic SNPs across the *Cyp6m* sub cluster of 3 genes (*Cyp6m2*, *Cyp6m3* and *Cyp6m4*) and across the *Cyp6* supercluster of 14 genes within which the *Cyp6m* sub cluster is located (*Cyp6s2*, *Cyp6s1*, *Cyp6r1*, *Cyp6n2*, *Cyp6y2*, *Cyp6y1*, *Cyp6m1*, *Cyp6n1*, *Cyp6m2*, *Cyp6m3*, *Cyp6m4*, *Cyp6z3*, *Cyp6z2* and *Cyp6z1*), and performed hierarchical clustering across these regions as described above. The typical geographical stratification of haplotypes persisted, suggesting the absence of a selective sweep across this region [Supplementary Fig.5 & 6].

We examined the genetic backgrounds carrying these alleles further by constructing median joining networks (MJNs) [35] using the Ag1000G Phase 2 haplotype data. This enabled us to resolve the

radiation of DNA substitutions arising on haplotypes carrying the identified variants. It also allowed us to reconstruct and position intermediate haplotypes while revealing the non-hierarchical relationships between haplotypes that could not be resolved by hierarchical clustering alone. The MJNs were constructed with reference to a maximum edge distance of two SNPs. This ensured that the connected components captured only closely related haplotypes. The resulting MJNs had a close correspondence with the hierarchical clustering output in assignment of haplotypes to clusters (88% overall concordance across all clusters).

The median joining networks showed more clearly the distinctive demographic stratification of the high frequency variants that was highlighted by the hierarchical clustering networks [Figure 3]. Most nodes containing secondary variants arising from the main nodes were small, which is inconsistent with directional selection where larger nodes are expected. Only one of the **I359V** nodes contained haplotypes from mosquitoes of both species, however the secondary nodes did not contain haplotypes from more than one species. This indicates that although **I359V** is shared by both *An. gambiae* and *An. coluzzii*, it is unlikely that this is because of an introgression event across the *Cyp6m2* gene.

Figure 3. Haplotype networks.

Median joining network for haplotypes carrying **A13T**, **D65A**, **E328Q**, **Y347F**, **I359V** and **A468S**, with a maximum edge distance of two SNPs. Node size indicates haplotype counts and node colour indicates the population/species of haplotypes. AO=Angola; GH=; BF=Burkina Faso; CI=Côte d'Ivoire; GN=Guinea; CM=Cameroon; GW=Guinea Bissau; GM = The Gambia; GA=Gabon; UG=Uganda; FR=Mayotte; GQ=Equatorial Guinea; KE=Kenya.

Positive selection of non-synonymous alleles

Extended Haplotype Heterozygosity (EHH) decay [36] was calculated to explore evidence for directional selection on the haplotypes carrying high frequency non-synonymous variants. It is expected that the presence of ongoing or recent directional selection pressure would lead to the increase in frequency of haplotypes, which on average will have longer regions of haplotype homozygosity relative to haplotypes that are not under selection. This diversity reduction would produce signatures of selection that would be conspicuous across a large genomic region. EHH

analysis would therefore be able to detect diversity reduction caused by ongoing directional selection being driven either by amino acid substitutions identified within the gene or by mutations within *cis*-acting elements next to the gene that may be under selection.

To perform the EHH decay analysis, we defined a core region of 1689 bases that spans across the entire gene. This was identical to what was used to differentiate the identified haplotype groups though hierarchical clustering. This region contained multiple distinct haplotypes above 1% frequency within the cohort, including haplotypes corresponding to the C1-C5 haplotype clusters. All haplotypes that did not correspond to C1-C5 were considered to be wild type (wt). Although there were several different haplotypes in each population that fit this description, we do not distinguish between them and call all these wild type, as *Cyp6m2* has no known resistance alleles and a true wild type remains to be discovered. EHH decay was then computed for each core haplotype up to 200 kilobases upstream and downstream [Supplementary Fig. 7]: beyond 200 kb, the EHH had decayed to zero.

We noted that haplotype clusters containing high frequency variants (C1-C5) did not exhibit a significantly slower EHH decay relative to the wild types, showing no evidence of positive selection. However, one Kenyan wild type haplotype group had a dramatically slower EHH decay relative to wild type haplotypes from other populations. In order to account for this difference within wild type groups across multiple populations and to reveal potential signs of selection that would be obscured by a collective analysis across all populations, we separated the haplotypes by population and species and recomputed EHH decay for each core haplotype as above.

Figure 4. Extended haplotype homozygosity per population.

No evidence for drastic difference in linkage disequilibrium within populations around core haplotypes across *Cyp6m2*. Extended Haplotype Heterozygosity (EHH) decay was calculated around cluster (C1 to C5) and non-cluster (wt) haplotypes using SNPs across and flanking the *Cyp6m2* region. KE=Kenya, GAgam=Gabon *An. gambiae*, AOcol= Angola *An. coluzzii*, GW=Guinea Bissau, CIcol=Côte d'Ivoire *An. coluzzii*, GHcol=Ghana *An. coluzzii*.

Kenyan mosquito populations are known to have an extreme demographic history, as they have experienced a severe recent bottleneck, and the Angola and Gabon populations are known to be geographically unique populations which are strongly differentiated from all other populations[31]. Hence, their haplotypes exhibited a considerably slower decay than West African haplotypes [first three panels: Figure 4]. However, the putative resistance haplotypes C1-C5 did not experience a

slower EHH decay relative to their wild type haplotypes, showing no evidence of positive selection acting upon those haplotypes in those populations.

As expected, the West African *An. coluzzii* haplotypes exhibited a much faster decay of EHH than specimens from Kenya, Angola, or Gabon, highlighting the demographic differences previously observed for these collections [31] [last three panels, Figure 4]. The C5 haplotype was a promising candidate for potential selection as it occurred within a more diverse population, and it was interesting to note that some wild type haplotypes in Côte d'Ivoire's *An. coluzzii* had a slightly slower decay than others within West Africa [fifth panel, Figure 4]. However, these haplotypes were not part of the C5 cluster, and did not carry the widespread **E328Q** mutation. The C5 haplotype did not exhibit a dramatically slower decay of EHH than wild type haplotypes in the populations in which it was found, suggesting that it is not under positive selection.

Discussion

Cyp6m2 has been implicated in many *Anopheles* populations as a key P450 that contributes to the insecticide resistance phenotype [5, 14, 20, 24]. It has been reported that allelic variants across some P450s can affect enzyme conformational dynamics and substrate binding affinity [28], offering potential mechanisms that may modulate enzyme activity and efficiency, and thus account for additional *Cyp6m2* resistance where CNVs alone may not suffice. However, little is also known about *Cyp6m2* allelic variation across Africa.

In this study, we report a comprehensive account of the distribution of amino acid substitutions occurring within the *Cyp6m2* gene. We also examine the haplotype structure of the gene to probe for selective sweeps by performing hierarchical clustering of haplotypes. We also examine the genetic background upon which the missense variants are found by plotting both median joining networks and decay of extended haplotype homozygosity, which are useful for revealing signatures of selection. We note that the distinct haplotype groups therein are stratified demographically and largely correspond to signature missense variants found in specific populations. This is in contrast to the strong signals of recent positive selection at other cytochrome P450 gene loci such as at *Cyp6p3* [31] which is often upregulated in tandem with *Cyp6m2* in multiple pyrethroid resistant populations [5, 37, 38].

It is still unclear how the identified non-synonymous variants may modulate *Cyp6m2* binding activity, in either the presence or absence of multiple competitive substrates and metabolites. The two aromatic residues (Phe 108 and Phe 121) that have been previously identified to be vital in deltamethrin orientation in the *Cyp6m2* active site[14] were not found to contain high frequency variants in our dataset.

None of the haplotype groups identified that carried missense variants were found to be under directional selection. This is despite the existence of a widespread variant (**E328Q**) linked to a geographic region (West Africa) where *Cyp6m2* upregulation has been associated with emerging metabolic resistance [20, 37]. In *An. coluzzii* originating from both Côte d'Ivoire and Ghana, the C5 haplotype that carried **E328Q** was shown to have an even faster decay of EHH than the wild type haplotypes, further indicating an absence of directional selection. The stratification of other main haplotype clusters from Angola (C1), Gabon (C2) and Kenya (C3) was also consistent with the strong demographic differentiation and overall reduced heterozygosity of these populations described elsewhere [31].

While the genomic data quality across the *Cyp6m2* gene and its putative promoter region was satisfactory, there was a ~10,000 base region of inaccessibility upstream of *Cyp6m2* that cut across the intergenic region into *Cyp6n1* [39]. A similar inaccessible region was also present 1 kb downstream of the gene in the intergenic region between *Cyp6m2* and *Cyp6m3*, which is likely caused by the presence of repeats that inhibit read mapping. Although it is possible that the upstream region of inaccessibility could contain a regulatory variant that is susceptible to selection, it is unlikely to obscure signatures of selection.

It has been shown in multiple studies that target-site resistance (i.e. VGSC-*kdr*) provides a strong persistent baseline of resistance as it rises towards fixation within populations [40]. In the presence of insecticide selection pressure, target-site mutations and metabolic resistance have also been shown to act synergistically to confer a stronger resistance phenotype to pyrethroids [29, 41]. While signatures of selection have previously been identified in some metabolic gene clusters within populations that have a high *kdr* frequency[31], further studies need to examine whether directional selection occurring on one locus can obscure selection on another locus. To resolve this conundrum, genomic analysis must be performed on populations sampled across generations and whose

transcriptomic and phenotypic characteristics are known, in order to tease out the individual contributions of specific sources of resistance.

Independent studies employing different experimental designs have also shown that metabolic resistance manifests as a cascade of multiple upregulated genes [42]. These genes, like *Cyp6m2*, are part of the normal cellular mechanism for xenobiotic detoxification that involves a linked, coordinated response of large multi-gene enzyme families in complicated pathways. Therefore, it is likely that identifying signatures of selection due to insecticide pressure will involve thorough analysis across this vast network. The Cap 'n' Collar isoform-C (CncC) transcription factor sub-family has been shown to work in tandem with other transcription factors to regulate the transcription of phase I, II and III detoxification loci of multiple insects such as *Culex quinquefasciatus* and *D. melanogaster* [43, 44]. *CncC* knockdown or upregulation has been shown to directly affect phenotypic resistance in *Anopheles gambiae* as well, modulating the expression of key P450s enzymes such as *Cyp6z2*, *Cyp6z3* and *Cyp6m2* that are located in the same genomic region[43, 44]. Given that we have detected no evidence of selection on amino acid variants in the *Cyp6m2* gene, it is possible that the emergence of *Cyp6m2* associated resistance is being driven by selection pressures acting upon genes coding for distant regulatory proteins such as transcription factors. These transcription factors can regulate downstream gene expression across large genomic distances. These transcription factors have also been implicated in the differential expression of other detoxification enzyme families also associated with insecticide resistance (GSTs, COEs, UDP-glucuronosyltransferases (UGTs) and ABC transporters). It is therefore likely that the centre of selection leading to the *Cyp6m2* associated resistance phenotype will be identified through whole genome selection scans of susceptible and resistant populations rather than by single loci analysis. Further research on Anopheline epigenomics, transcriptomics, proteomics and systems biology will also be game changers in mapping the complex regulatory network of insecticide resistance, aiding the identification of critical targets and the development of new strategies to control the spread of metabolic insecticide resistance.

Conclusion

The scale up of insecticide-based interventions has caused increased selection pressure and higher levels of insecticide resistance across Africa. While the *CYP6M2* enzyme has been associated with emerging metabolic resistance in Africa, our data indicates that allelic variation within the *Cyp6m2*

gene itself or across its Cyp6 supercluster has not been subject to recent positive selection in any of the populations sampled. This is in contrast to other Cytochrome P450 genes where CNV alleles are clearly under strong selection. Our results do not rule out a role for *Cyp6m2* in insecticide resistance in natural populations, but highlight the need for a deeper understanding of the regulatory networks affecting Cytochrome P450 gene expression in malaria vectors. This will require large-scale, holistic experimental work that collects genomic, transcriptomic and phenotypic datasets which when juxtaposed can resolve the complexities of metabolic resistance.

Methods

Data collection and analysis

In this study, we followed the species nomenclature of Coetzee *et al* [45] where *An. gambiae* refers to *An. gambiae sensu stricto* (S form) and *An. coluzzii* refers to *An. gambiae sensu stricto* (M form). A detailed description of the Ag1000G sample collection, DNA extraction, sequencing, variant calling, quality control and phasing can be found here [31]. Briefly, Anopheline samples were collected from 33 sampling sites across 16 populations in 13 countries in sub-Saharan Africa [Table 1 & Additional file 1]. The sampling procedure covered different ecosystems and aimed at collecting a minimum of 30 specimens per country. The specimens consisted of *An. gambiae* and *An. coluzzii*: only *An. coluzzii* were sampled from Angola, both *An. gambiae* and *An. coluzzii* were sampled from Burkina Faso, while all other populations consisted of *An. gambiae*, except Kenya and Guinea Bissau where the species identity was indeterminate.

Whole genome sequencing of all mosquitoes was performed on the Illumina HiSeq 2000 platform. The generated 100 base paired-end reads were aligned to the *An. gambiae* AgamP3 reference genome assembly [46] and variants were called using GATK UnifiedGenotyper. Samples with mean coverage $\leq 14\times$ and variants with attributes that correlated with Mendelian error in genetic crosses were removed during quality control.

The SnpEff v4.1b software was used for the functional annotation of Ag1000G variant data [47] using locations from geneset AgamP4.12. All variants in transcript AGAP008212-RA with a SnpEff

annotation of “missense” were regarded as nonsynonymous variants. The *Cyp6m2* gene has not been shown to exhibit alternative splicing, and no alternative transcripts have been reported.

Haplotype clustering, linkage disequilibrium and mapping of haplotype clusters

To reveal the haplotype structure at *Cyp6m2*, *Cyp6m* sub-cluster, *Cyp6* supercluster, *HAM*, *ODR-2* and *SH2*, we computed the Hamming distance between all haplotype pairs and performed hierarchical clustering of haplotypes. We worked through arbitrary clustering threshold values to cut the dendrograms at genetic distances that would best highlight the most relevant clusters. We used Lewontin's *D'* [48] to compute the linkage disequilibrium (LD) between all pairs of missense *Cyp6m2* mutations. Image rendering for the haplotype clustering, linkage disequilibrium and haplotype cluster frequencies map was performed using the matplotlib Python package [49]. Geography handling for the haplotype cluster frequencies map was done using cartopy [50].

Haplotype Networks

We constructed haplotype networks using the median-joining algorithm [35] implemented in Python [51]. Haplotypes carrying the main high frequency mutations were analysed with a maximum edge distance of two SNPs. The Graphviz library was used to render the networks and the composite figure was constructed in Inkscape [52].

Extended haplotype homozygosity

We defined the core haplotype on a 1689 base region spanning the *Cyp6m2*, from chromosome arm 3R, starting at position 6928858 and ending at position 6930547. We selected this region to ensure a 1:1 haplotype correspondence with that used in the hierarchical clustering analysis. We computed extended haplotype homozygosity (EHH) across all core haplotypes in all populations as described in Sabeti et al. [36] using scikit-allel version 1.1.9 [53]. EHH composite plots were made using the matplotlib Python package [49].

List of abbreviations

CncC: Cap ‘n’ Collar isoform-C
CNV: Copy Number Variation

425 COEs: Carboxylesterases
 426 DDT: Dichlorodiphenyltrichloroethane
 427 EHH: Extended Haplotype Heterozygosity
 428 GSTs: Glutathione S-Transferases
 429 HAM: Transcription Factor Hamlet
 430 IRS: Indoor Residual Spraying
 431 *kdr*: Knock-Down Resistance
 432 LLINs: Long Lasting Insecticidal Nets
 433 MJNs: Median-Joining Networks
 434 ODR2: Odd-Skipped Related
 435 SH2: SRC Homology 2
 436 SNPs: Single Nucleotide Polymorphisms
 437 P450s: Cytochrome P450 Monooxygenases
 438 UGTs: UDP-glucuronosyltransferases
 439 VGSC: Voltage Gated Sodium Channel
 440 *wt*: wild type

441

442 **Declarations**

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 445 facilities for their contributions.

446

447 **Availability of data and materials**

448 Jupyter Notebooks and scripts containing all analyses, tables and figures can be found in the GitHub
 449 repository [51]. Variant calls and phased haplotype data from the Ag1000G Phase 2 AR3 data release
 450 were used, and can be found here [54].

451

452 **Authors contribution**

453 AM and MKNL designed the study. AM and CC developed the base code. MGW and AM performed
454 all analyses. MGW drafted the manuscript. All authors read and approved the final manuscript.

455

456 **Competing interests statement**

457 The authors declare no competing interests.

458

459 **Consent for publication**

460 Not applicable

461

462 **Ethics approval and consent to participate**

463 Not applicable.

464

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469

470

471 **Supplementary Figures**

472 **Supplementary Figure 1. Linkage disequilibrium (D') between non-synonymous variants.**

473

474 A value of 1 shows perfect linkage between the alleles. A value of -1 shows that the alleles are never found conjointly. The bar

475 plot indicates allele frequencies within the Ag1000G phase 2 cohort.

476

477 **Supplementary Figure 2. Hierarchical clustering and missense mutations for *ODR2*.**

478 Top: a dendrogram showing hierarchical clustering of haplotypes across the *ODR2* gene. The gene is located at position

479 7,059,422 to 7,119,244: 128,875 bases downstream of *Cyp6m2*.

480 The colour bar indicates the population of origin for each haplotype.

481 Bottom: high frequency (>5%) alleles identified within each haplotype (white = reference allele; black = alternative allele).

482

483 **Supplementary Figure 3. Hierarchical clustering and missense mutations for *HAM*.**

484 Top: a dendrogram showing hierarchical clustering of haplotypes across the *HAM* gene. The gene is located at position

485 7,435,306 to 7,485,012: 504,759 bases downstream of *Cyp6m2*.

486 The colour bar indicates the population of origin for each haplotype.

487 Bottom: high frequency (>5%) alleles identified within each haplotype (white = reference allele; black = alternative allele).

488

489 **Supplementary Figure 4. Hierarchical clustering and missense mutations for *SH2*.**

490 Top: a dendrogram showing hierarchical clustering of haplotypes across the *SH2* gene. The gene is located at position

491 8,176,778 to 8,183,084: 1,246,231 bases downstream of *Cyp6m2*.

492 The colour bar indicates the population of origin for each haplotype.

493 Bottom: high frequency (>5%) alleles identified within each haplotype (white = reference allele; black = alternative allele).

494

495 **Supplementary Figure 5. Hierarchical clustering and missense mutations for *Cyp6m* sub**

496 **cluster.**

Top: a dendrogram showing hierarchical clustering of haplotypes across the Cyp6m sub cluster of genes containing Cyp6m2, Cyp6m3 and Cyp6m4. The genes are located at position 6928858 to 6935721. The colour bar indicates the population of origin for each haplotype. Bottom: high frequency (>5%) alleles identified within each haplotype (white = reference allele; black = alternative allele).

Supplementary Figure 6. Hierarchical clustering and missense mutations for Cyp6 supercluster.

Top: a dendrogram showing hierarchical clustering of haplotypes across the Cyp6 supercluster of 14 P450 genes containing Cyp6s2, Cyp6s1, Cyp6r1, Cyp6n2, Cyp6y2, Cyp6y1, Cyp6m1, Cyp6n1, Cyp6m2, Cyp6m3, Cyp6m4, Cyp6z3, Cyp6z2 and Cyp6z1. The genes are located at position 6903106 to 6978142. The colour bar indicates the population of origin for each haplotype. Bottom: high frequency (>70%) alleles identified within each haplotype (white = reference allele; black = alternative allele).

Supplementary Figure 7. Extended haplotype homozygosity across all populations.

A rapid decay of EHH in comparison to other haplotypes implies absence of positive selection.

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- 632 34. **Chromosome 3R: 8,176,778 - 8,183,084 - Region in detail - *Anopheles gambiae* -**
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Additional files

1. Additional file 1.

- 694 a. File name = Additional file 1
- 695 b. Title = List of *An. gambiae* and *An. coluzzii* genome samples and haplotypes from
- 696 Ag1000G Phase 2-AR3.
- 697 c. Format = csv
- 698 d. Description = Table showing Ag1000G Phase 2-AR3 sample properties such as
- 699 population, country, region, sex, species identity and haplotype cluster.
- 700 2. Additional file 2.
- 701 a. File name = Additional file 2
- 702 b. Title = List of synonymous and non-synonymous genetic variants in *Cyp6m2*.
- 703 c. Format = csv
- 704 d. Description = Table showing Ag1000G Phase 2-AR3 *Cyp6m2* variant calls and
- 705 variant properties stratified by population and effect.
- 706
- 707