Genomic Analysis of Two Phlebotomine Sand Fly Vectors of *Leishmania* from the New and Old World

Short title: Sand fly genomics

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Abstract

Phlebotomine sand flies are of global significance as important vectors of human disease,

transmitting bacterial, viral, and protozoan pathogens, including the devastating kinetoplastid

parasites of the genus *Leishmania*, the causative agents of diseases collectively termed

leishmaniasis. More than 40 pathogenic *Leishmania* species are transmitted to humans by

approximately 35 sand fly species in 98 countries with hundreds of millions of people at risk

around the world. As no approved efficacious vaccine exists, available drugs are expensive and/or

toxic, and resistance is emerging, management of sand fly populations to break transmission is

currently the most effective disease control strategy. To better understand the biology of sand

flies, including the mechanisms involved in their vectorial capacity, insecticide resistance, and

population structures we sequenced the genomes of two of the most important sand fly species:

Phlebotomus papatasi, a cutaneous leishmaniasis vector, (distributed in the Middle East and

North Africa) and *Lutzomyia longipalpis,* a visceral leishmaniasis vector (distributed across

Central and South America). We categorized and curated genes involved in processes important

to their roles as disease vectors, including chemosensation, blood feeding, circadian rhythm,

immunity, and detoxification, as well as mobile genetic elements. We also defined gene orthology

and observed micro-synteny among the genomes. Finally, we present the genetic diversity and

population structure of these species in their respective geographical areas. These genomes will

be a foundation on which to base future efforts to prevent vector-borne transmission of

Leishmania parasites.

Author Summary

The leishmaniases are a group of neglected tropical diseases caused by protist parasites from

the Genus *Leishmania*. Different *Leishmania* species present a wide clinical profile, ranging from

mild, often self-resolving cutaneous lesions that can lead to protective immunity, to severe

metastatic mucosal disease, to visceral disease that is ultimately fatal. *Leishmania* parasites are

transmitted by the bites of sand flies, and as no approved vaccine exists, available drugs are toxic

and/or expensive and resistance is emerging, new dual control strategies to combat these

diseases must be developed, combining interventions on human infections and integrated sand

fly population management. Effective vector control requires a good understanding of the biology

of sand flies. To this end, we sequenced and annotated the genomes of two sand fly species that

are important leishmaniasis vectors from the Old and New Worlds. These genomes allow us to

better understand, at the genetic level, processes important in the vector biology of these species,

such as finding hosts, blood-feeding, immunity, and detoxification. These genomic resources

highlight the driving forces of evolution of two major *Leishmania* vectors and provide foundations

for future research on how to better prevent leishmaniasis by control of the sand fly vectors.

Introduction

Phlebotomine sand flies are a group of vectors that vary widely in their geographic distribution,

ecology, and the pathogens they transmit. Sand flies serve as vectors for several established,

emerging, and re-emerging infectious diseases, transmitting protist, bacterial and viral pathogens.

The most devastating of the sand fly transmitted diseases is leishmaniasis, accounting for an

estimated 2.4 million disability-adjusted life-years (DALYs) [1] and 40,000 deaths annually [2].

These statistics are largely underestimated due to misdiagnosis and lacking surveillance systems

in many of the affected countries. Political instability, urbanization, and climate change are

expanding *Leishmania*-endemic regions and increasing the propensity for epidemics world-wide

[3]. These factors are coupled with the increase of visceral disease due to HIV co-infection,

spurring the World Health Organization to classify leishmaniasis as one of the world's epidemic-

prone diseases [4].

Leishmaniasis is worldwide, occurring in 98 countries and spread over five continents, putting

310 million people at risk of contracting leishmaniasis [2]. Leishmaniasis is a collective term for

a group of distinct clinical manifestations ranging from mild, often self-resolving cutaneous lesions

that can lead to protective immunity, to disseminated lesions that do not heal spontaneously, to

destruction of the mucous membranes of the nose, mouth, and pharynx, to life-threatening

visceral disease. The clinical profile depends on a variety of factors, including vector biology, host

immunity, and parasite characteristics; the *Leishmania* species initiating the infection being the

primary determinant. The two primary clinical forms are cutaneous leishmanisis (CL) and visceral

leishmaniasis (VL). The primary *Leishmania* species that cause CL are *Leishmania major*,

Leishmania infantum, Leishmania tropica, and *Leishmania aethipica* in the Old World and

Leishmania amazonensis, Leishmania braziliensis, Leishmania guyanensis, Le. infantum,

Leishmania mexicana, and *Leishmania panamensis* in the New World. VL is primarily caused by

Leishmania donovani in Asia and Africa and *Le. infantum* in the Middle East, central Asia, South

and Central America, and the Mediterranean Basin.

There are approximately 35 proven, and an additional 63 suspected, vectors of at least 40

different *Leishmania* species to humans [5, 6]. *Phlebotomus* species are the primary *Leishmania*

vectors in the Old World and *Lutzomyia* species are responsible for transmitting leishmaniasis

throughout the Americas [7]. There is a close ecological association, if not co-evolutionary

relationship [8, 9], between *Leishmania* species and their specific vectors such that generally a

single sand fly species transmits a single *Leishmania* species under natural conditions. Some

sand flies, however, can transmit a range of *Leishmania* species under experimental conditions

[10]. This difference has given rise to the concept of "restricted" and "permissive" vectors [11].

For example, *Phlebotomus papatasi* is a restrictive vector, transmitting only *Le. major* parasites

[12]. *Lutzomyia longipalpis* is considered a permissive vector in laboratory conditions, but only

transmits *Le. infantum* naturally [12].

These vectors are part of the Diptera which is an extremely species-rich and ecologically

diverse order of insects and contains the vectors of many of the most important pathogens of man

and his domesticated animals. Both phlebotomine sand flies (family Psychodidae) and

mosquitoes (Culicidae) are specified as members of distinct infra-orders within the suborder

Nematocera. While the Nematocera grouping is paraphyletic, the relationships between infra-

orders remains to be elucidated [13]. Some studies generated topologies with Psychodomorpha

(sand flies) and Culicomorpha (mosquitoes and black flies) as sister groups [14], whereas, others

place sand flies nearer to the muscoid flies (Ephydroidea) [15]. The internal relationships within

the assemblage that includes Psychodidae also remains a matter of debate [16].

It is postulated that this close evolutionary relationship between sand fly species and the

Leishmania species that they transmit may have epidemiological implications for leishmaniasis

[17]. For example, there are three primary zymodemes of *Le. major* that have limited geographical

distributions such that the prevalent zymodeme in a particular area overlaps with the distribution

of one primary population of *Ph. papatasi* [18]. *Ph. papatasi* has a wide geographical distribution,

ranging from Morocco to the Indian subcontinent and from southern Europe to central and eastern

Africa. Given the wide ecological and geographic distribution of *Ph. papatasi* populations [19],

coupled with the low dispersal capacity of these sand flies [12], it is likely that there is limited gene

flow between populations and significant genetic structuring between populations. While previous

studies demonstrated relatively low genetic differentiation between *Ph. papatasi* populations

separated by large geographical distances [9, 20], more recent studies have identified genetic

differentiation between geographically separated populations [18, 21-24] and local differentiation

[25]. Microsatellite analysis, in particular, revealed two distinct genetic clusters of *Ph. papatasi* (A

& B) with further substructure within each population that correlated with geographical origin (A1-5

and B1 &2) [18, 23].

While elucidating the drivers leading to reproductive isolation and speciation remains a

challenge, there is strong evidence that *Lu. longipalpis* is undergoing incipient speciation in Brazil

with various levels of differentiation between siblings of the complex [26]. The Brazilian

populations of *Lu. longipalpis* can be divided into two primary groups. The males of one group

produce Burst-type mating songs and a diterpenoid-like pheromone, Cembrene-1, recently

reclassified as sobralene (pentadeca-E-3,4-Z-8,9-triene) [27]. The second group is more

heterogeneous, with populations producing different subtypes of Pulse-type copulation songs and

different sex pheromones [C_{16} H_{32} types Himachalene (Hima), Germacrene (Germ), and

Sobralene [28]. Crossing experiments of sympatric populations of the two different types revealed

reproductive isolation due to both pre-mating and copulatory mechanisms [29]. Hickner *et al.* 2020

provided genomic insights into the chemoreceptor genome repertoire underlying behavioral

evolution of sexual communication in the *Lu. longipalpis* populations, but whole-genome analyses

could improve the identification of loci related to critical traits such as vectorial capacity, host

preference, and insecticide resistance.

Despite the potential importance of the sand fly immune system for influencing *Leishmania*

development and survival in the sand fly gut the immune response is poorly studied. To date, this

work has been largely restricted to defensins [30-33]. However, gene depletion via RNAi of the

negative regulator of the IMD pathway caspar [34] led to a reduction in *Leishmania* population in

the gut of *Lu. longipalpis*. While the knockout of relish, the transcription factor of IMD pathway,

resulted in the increase of *Leishmania* and bacteria in *Ph. papatasi* [35].

Adaptation to hematophagy presents many challenges to insects, including avoiding the

physiological responses of the host that interfere with obtaining a blood meal, digestion of the

blood, and excretion of the excess water contained in the blood meal. Sand flies have evolved a

complex cocktail of pharmacologically active salivary molecules to facilitate blood feeding that

have been extensively characterized [36]. Although hematophagy and host seeking are

controlled by the circadian clock, the molecular regulation of circadian rhythms is poorly

understood in sand flies. However, Yuan *et al.* 2007 proposed three clock models based on the

presence of the cryptochrome (CRY) proteins, CRY1 and CRY2 [37]. In the *Drosophila* clock

model, only CRY1, which acts as a blue-light photoreceptor [38], is present. In the butterfly model,

CRY1 also acts as a photoreceptor and CRY2, which is a mammalian–like transcriptional

repression, dimers with PER to repress CLK/CYC activity. In the bee model, there is only CRY2,

which seems to act as a repressor together with PER and some other molecule that is not CRY1

that acts as photoreceptor.

A central inquiry of evolutionary biology is the elucidating drivers of speciation, however,

defining species boundaries and identifying the genetic architecture that leads to reproductive

isolation has been a challenge. Understanding of the mechanisms of vectorial capacity,

adaptation to changing ecological environments, and insecticide resistance has epidemiological

consequences for the integrated management of sand fly populations that is the cornerstone of

leishmaniasis control [39]. To explore the driving forces of evolution of two major phlebotomine

sand flies from the *Psychodidae* family (*Phlebotominae* subfamily), *Ph. papatasi* and *Lu.*

longipalpis, that exhibit distinct distributions, behavior, and pathogen specificity, we sequenced

and analyzed their whole-genomes using comparative genomics approaches. We manually

curated a number of gene families with key roles in processes such as immunity, blood-feeding,

chemosensation, detoxification, and circadian biology to provide a basis for studying and

understanding sand flies as *Leishmania* vectors. Moreover, as a better understanding of the

population structure of geographically separated vector populations is necessary, we also

assessed the population structure of *Ph. papatasi* and *Lu. Longipalpis* by collecting and

sequencing individual field-collected specimens sampled over a large geographical range in the

Middle East and North Africa, and Brazil, respectively. Our results provide significant advances in

our understanding of the genetics underlying the population structure and provide a great

foundation for future molecular comparative studies of these two medically important vectors.

Methods

<u>Fly Maintenance</u>

Phlebotomus papatasi

To avoid confounding effects due to genetic polymorphisms, we used an inbred colony of *Ph.*

papatasi (Israeli strain) for the genome assembly. This colony was originally established in the

1970s and given to Walter Reed Army Institute of Research (WRAIR) in 1983 from the Hebrew

University, Jerusalem and transferred to the University of Notre Dame in 2006. At several times

since establishment in the laboratory, the colony has fluctuated in population size and has been

expanded from a relatively low number of files. Sand flies were reared by the method of Modi

and Tesh [40].

Lutzomyia longipalpis

Lu. longipalpis Jacobina strain was used for the genome assembly. This colony was originally

established at the Liverpool School of Tropical Medicine by Richard Ward in 1988 from flies

caught in Jacobina, Bahia State, Brazil. At several times since establishment in the laboratory,

the colony has fluctuated in population size and has been expanded from a small number of flies.

Flies were reared under standardized laboratory conditions [40], *i.e.* under controlled temperature

(27 ± 2°C), humidity (>80%), and photoperiod (8 hours light/16 hours darkness) [40].

Ethics Statement

The study protocol was approved by the Institutional Animal Care and Use Committee at the

University of Notre Dame (#07-052).

Field Collections

Phlebotomus papatasi

Ph. papatasi were collected from three different locations: Tunisia, Egypt, and Afghanistan. In

Tunisia, samples were collected in 2013 from the village of Felta located in an arid

biogeographical area in Central Tunisia (35°16'N, 9°26'E). In North Sinai Egypt, samples were

collected in Om Shikhan (30°50'N, 34°10'E), located approximately 340 km east of Cairo, 80 km

inland from the Mediterranean coast, and 30 km west of the Israeli border in 2007. In Afghanistan,

samples were collected in 2010 in and around a German military camp located near the airport of

Mazar-e Sharif (36°43'N, 67°14'E). This site is located at 400m altitude north of the Hindukush

Mountains and approximately 50km south of the Uzbekistan border. Sand flies were trapped using

CDC-style light traps between 17:00 and 07:00.

Lutzomyia longipalpis

Lu. longipalpis were collected in 2014 from six different locations in Brazil (Fig. 1). Samples

were collected from three allopatric populations: Jacobina, Bahia State (11º10'S 40º31'W),

Lapinha Cave, Minas Gerais State (19º38'S 43º53'W), Marajó Island, Pará State (0°56'S

49°38'W), and two sympatric populations (called 1S and 2S) from Sobral, Ceará State (34º41'S

40º20'W). For comparison of male copulatory courtship songs, flies were also collected from

Olindina (11º 29' S 38° 22' W) and Araci (11º 09' S 39° 01' W), sites near Jacobina. Sand flies

were trapped using CDC-style light traps baited with CO_2 between 18:00 and 06:00 and

transported to the laboratory. Analyses of male copulatory courtship songs was carried out by as

previously described [41]. The recordings were performed by using males and females from

laboratory colonies established from wild-collected flies from Lapinha and Sobral and from Araci

and Olindina.

Figure 1. *Lutzomyia longipalpis* site locations for copulation songs and pheromone types.

Samples were collected from three allopatric populations: Marajó (Pará State; 0°56'S 49°38'W),

Jacobina (Bahia State; 11º 10'S 40º 31'W), and Lapinha Cave (Minas Gerais State; 19º 33'S 43º

57'W); and two sympatric populations (1S and 2S) from Sobral (Ceará State; 3º 41'S 40º 21'W).

Copulation songs: Burst-type (B) and Pulse-types (P1, P2, and P3). Pheromone types: cembrene

(Cemb), 9-methylgermacrene-B (Germ), and himachalene (Hima). For each location, the number

of single nucleotide polymorphisms (SNPs) identified in each population with respect to the

reference genome (VectorBase, LlonJ1) is indicated. A total of 1,937,819 SNPs were identified

among all the populations. Map Source: Google Imagery TerraMetrics.

<u>Nucleic Acid Isolation</u>

Genomic DNA from female sand flies was isolated from pools of flies or from single insects

for population genetics analysis. For pooled insects, DNA was suspended in 50 µl of the hydration

solution using a Tissue DNA isolation kit (GE HealthCare LifeSciences).

Total RNA was obtained from the four larval instars, pupae, and adult males and females one,

three, and ten days post emergence (adults fed only on sugar). Additional RNA was included from

Ph. papatasi females six, 24, and 96 hours post feeding on uninfected and *Le. major* (strain

Friedlin V1) infected mouse blood and *Lu. longipalpis* females 6, 24, and 144 hours post feeding

on uninfected and *Le. infantum* infected mouse blood. Total RNA was extracted using a RNAeasy

Mini Kit (Qiagen).

<u>Genome Sequencing and Assembly</u>

Phlebotomus papatasi

Sequencing and assembly for *Ph. papatasi* were performed by the Genome Institute,

Washington University School of Medicine. The assembly was built with the –het option, using

the Newbler assembler test release 2.6RC02 from an input of ~22.5X total sequence coverage

with Sanger and 454 reads including 15.1X of whole-genome shotgun reads, 4.4X 3 kb clone

inserts, 3.0X 8 kb inserts and 0.01X BAC-end read pairs. The fragment and 3 kb data were

generated from a single sand fly after whole-genome amplification, while the 8kb data were

derived from multiple flies. The 0.1X of Sanger 3,730 BAC end sequences (28,902 reads) were

also derived from multiple flies.

Prior to submission to NCBI, this assembly was screened for contamination and 247 contigs

were removed. Additional contigs were removed or merged by the McDonnell Genome Institute

Washington University School of Medicine, PolyGraph Assembler, which collapses heterozygous

contigs and reduced the assembled genome size from 364 Mb to 345 Mb. A total of 5,661 gaps

were closed and nearly 6.8 Mb of sequence was added by another in house post-assembly

program, PyGap. This program detects and merges overlaps of adjoining contigs and attempts to

close gaps between non-overlapping adjoining contigs with Illumina data. The same Illumina data

used in gap closure was aligned to the assembly to correct 89,378 presumed 454

insertion/deletion errors.

Lutzomyia longipalpis

Three types of *Lu. longipalpis* whole-genome shotgun (WGS) libraries were used: a 454

Titanium fragment library and paired end libraries generated from 3 kb and 8 kb inserts. The 454

data (11.5 million reads; ~24.4X coverage) was derived from the same individual while mate pair

reads (7.4 million 3kb reads, 9.6X; 3.7 million 8kb reads, 4.9X) were derived from a pool of

individuals. In total, approximately 22.6 million reads were generated at the Baylor College of

Medicine Human Genome Sequencing Center (BCM-HGSC) using the Celera CABOG assembler

(version 6.1, 2010/03/22) and represents 38.9X coverage of this sand fly genome. These initial

results were used as a backbone for longer superscaffolds using ATLAS-link [42]. Finally,

discernible gaps were filled (see [43]) with ATLAS-gapfill. The total length of all contigs is 142.7

Mb; however, the total span of the assembly is 154.2 Mb after gaps are included.

<u>Individual population sequencing</u>

To prepare short insert libraries, an Illumina gel-cut paired-end library protocol was used.

Briefly, DNA was extracted from individual adult males or females from inbred lines using the

Qiagen DNAeasy Blood and Tissue kit following the manufacturer's supplementary protocol for

purification of total DNA from insects. Purified DNA was sheared using a Covaris S-2 system

(Covaris, Inc. Woburn, MA). Sheared DNA fragments were purified with Agencourt AMPure XP

beads, end-repaired, dA-tailed, and ligated to Illumina universal adapters. After adapter ligation,

DNA fragments were further size-selected by agarose gel elution and PCR amplified for 6 to 8

cycles using Illumina P1 and Index primer pair and Phusion® High-Fidelity PCR Master Mix (New

England Biolabs). The final library was purified using Agencourt AMPure XP beads and quality

assessed by Agilent Bioanalyzer 2100 (DNA 7500 kit) determining library quantity and fragment

size distribution before sequencing. Sequencing was performed on an Illumina HiSeq2000

platform generating 100 bp paired end reads. Sequenced reads sequence reads were deposited

in the NCBI SRA under Bioproject accession number PRJNA20279 for *Lu. longipalpis* and

PRJNA20293 for *Ph. papatasi*.

<u>RNA-sequencing</u>

RNA-sequencing (RNAseq) was conducted to improve resources available for gene

prediction. RNAseq was performed following standard protocols on an Illumina HiSeq 2000

platform. RNA was obtained from both sexes, during development, and following blood feeding

to generate an extensive RNA-seq coverage to allow for quality gene prediction. Briefly, poly-A$^+$

messenger RNA (mRNA) was extracted from 1 µg total RNA using Oligo(dT)25 Dynabeads (Life

Technologies, cat. no. 61002) followed by fragmentation of the mRNA by heating to 94°C for 3

min [for samples with RNA Integrity Number (RNI) = 3–6] or 4 min (for samples with RIN of >6.0).

First-strand complementary DNA (cDNA) was synthesized using the Superscript III reverse

transcriptase (Life Technologies, cat. no. 18080-044) and purified using Agencourt RNAClean XP

beads (Beckman Coulter, cat. no. A63987). During second-strand cDNA synthesis,

deoxynucleoside triphosphate (dNTP) mix containing deoxyuridine triphosphate was used to

introduce strand specificity. For Illumina paired-end library construction, the resultant cDNA was

processed through end repair and A-tailing, ligated with Illumina PE adapters, and digested with

10 units of uracil–DNA glycosylase (New England Biolabs, Ipswich, MA; cat. no. M0280L).

Amplification of the libraries was performed for 13 PCR cycles using the Phusion High-Fidelity

PCR Master Mix (New England Biolabs, cat. no. M0531L); 6-bp molecular barcodes were also

incorporated during this PCR amplification. These libraries were then purified with Agencourt

AMPure XP beads after each enzymatic reaction, and after quantification using the Agilent

Bioanalyzer 2100 DNA Chip 7500 (cat. no. 5067-1506), libraries were pooled in equimolar

amounts for sequencing. Sequencing was performed on Illumina HiSeq2000s, generating 100-bp

paired-end reads. Sequenced reads were deposited in the NCBI SRA, under BioProject

accession PRJNA81043 for *Lu longipalpis* and PRJNA20293 of *PH. papatasi*.

<u>Annotation</u>

The genome assemblies were initially annotated *ab initio* with gene models derived from

VectorBase annotation MAKER2 [44] pipelines [45]. The automated analyses identified 12,678

gene models for *Ph. papatasi* and 10,429 for *Lu. longipalpis*. Expert curators manually annotated

several gene families of interest (Supplementary Methods) resulting in 11,849 and 10,796 gene

models for *Ph. papatasi* and *Lu. longipalpis*, respectively.

<u>Orthology Delineation</u>

OrthoDB [46] orthology delineation was employed to define orthologous groups of genes

descended from each last common ancestor of the species phylogeny across 43 insects including

the two sand flies - Hemipterodea: *Pediculus humanus* and *Rhodnius prolixus*; Hymenoptera:

Apis mellifera and *Linepithema humile*; Coleoptera: *Tribolium castaneum*; Lepidoptera: *Bombyx*

mori and *Danaus plexippus*; Diptera: *Lu. longipalpis*, *Ph. papatasi* and *Glossina morsitans*, 12

Drosophila species (*D. grimshawi, D. mojavensis, D. virilis, D. willistoni, D. persimilis, D.*

pseudoobscura, D. ananassae, D. erecta, D. yakuba, D. melanogaster, D. sechellia, and D.

simulans), two culicine mosquitoes (*Aedes aegypti* and *Culex quinquefasciatus*), and 19

Anopheles species (*An. darlingi, An. albimanus, An. sinensis, An. atroparvus, An. farauti, An.*

dirus, An. funestus, An. minimus, An. culicifacies, An. maculatus, An. stephensi (SDA-500), *An.*

stephensi (INDIAN), *A. epiroticus, An. christyi, An. melas, An. quadriannulatus, An. arabiensis,*

An. merus, and *An. gambiae* (PEST). The orthology delineation was performed as part of the

Anopheles Genomes Cluster Consortium analyses of 16 newly-sequenced *Anopheles*

mosquitoes [47, 48]. From the complete set of species, the two sand flies were compared to a

symmetrical set of five representative mosquitoes and five representative flies, together with four

outgroup species representing four insect orders. The species compositions of all orthologous

groups defined at the dipteran root were analyzed with custom Perl scripts to count the numbers

of groups and genes shared among the sand flies, mosquitoes, and flies. Pairwise percent amino

acid identities between single-copy and/or multi-copy orthologs among the sand flies, *An.*

gambiae and *D. melanogaster* were extracted from all-against-all protein sequence comparisons

performed with SWIPE [49] as part of the OrthoDB orthology delineation procedure.

<u>Maximum likelihood species phylogeny</u>

To establish species relationships, the maximum likelihood species phylogeny was

determined from concatenated protein sequence alignments [aligned with default MUSCLE [50]

parameters and trimmed with the 'automated1' trimAl [51] setting of 1,627 relaxed single-copy

orthologs (no more than three paralogs in up to three species, longest protein selected)] from the

two sand flies, five mosquitoes, five flies, and four outgroup insect species. These orthologs were

selected from a total of 2,160 orthologous groups and were each required to have an alignment

of more than 50 amino acid columns after trimming and a relative tree certainty (see [52]) of more

than 50% as implemented in RAxML [53]. The concatenated alignment contained 1,065,440

amino acid columns with 627,808 distinct alignment patterns and was used to estimate the

maximum likelihood species phylogeny with RAxML [54] employing the PROTGAMMAJTT model

over 100 bootstrap samples and setting *Pe. humanus* as the outgroup species. The RAxML

phylogenies of individual ortholog groups were analyzed with custom Perl scripts and the Newick

Utilities [55] to partition the phylogenies into the three relevant topologies – i) sand flies with

mosquitoes, ii) sand flies with flies, or iii) sand flies as outgroup to mosquitoes and flies - and all

branch lengths were subsequently averaged.

Synteny Analysis

Macrosynteny

The identifiers, scaffolds, locations, and orientations were extracted from the FASTA headers

for each protein sequence. The protein identifiers were cross-referenced with the OrthoDB v8

Diptera [56] catalog to group the proteins into ortholog groups. Sequences without ortholog

information or not present in all genomes were discarded. Additionally, ortholog groups with

many-to-many and one-to-many relationships were not included in the analysis. Proteins were

sorted along each scaffold by their starting coordinates, while scaffolds were ordered arbitrarily.

Scatter plots were generated by drawing dots at the positions of orthologous proteins.

Synteny Blocks

For each genome, gene identifiers, protein sequences, orientations, scaffolds, and locations

were extracted from FASTA headers and reformatted as input for the programs CHROnicle and

SynChro [57]. SynChro (Δ = 5) was run pairwise for all species. Synteny blocks (lists of gene

identifiers, scaffolds, and positions) were extracted from the OrthBlocks synt files. Three-way

synteny blocks for *An. gambiae*, *Lu. longipalpis*, and *Ph. papatasi* were constructed by finding all

pairs of synteny blocks for *An. gambiae* and *Lu. longipalpis* and *Lu. longipalpis* and *Ph. papatasi*

that overlapped by at least one gene.

Orientation in genome by prediction of protein function

While SynChro predicts synteny, it does not predict the orientation of the blocks in the

genome. For each syntony block, the genes from one species were matched to genes in the

second species using blastp [58] against local BLAST databases created from *An. gambiae*, *Ph.*

papatasi, and *Lu. lutzomyia* protein sequences obtained from VectorBase [60]. The blastp

similarity searches were run using the default algorithm parameters. Two genes were considered

match if they located in the corresponding pair of synteny blocks, were in the top 5 hits for each

other, and 3 criteria were met for the alignment: e-value < 1e-3 and query coverage and identity

> 20%.　Relative alignment of the blocks were determined based on the corresponding

orientations of gene pairs; since some gene pairs had conflicting orientations, block orientation

was determined based on that which gave the greatest agreement across the gene pairs.

Population Genetics Analysis

SNP Calling

We performed alignments and variant calling on the raw reads of whole-genome samples of

Ph. papatasi collected from Tunisia (N=6), Egypt (N=6), and Afghanistan (N=5) to the *Ph. papatasi*

reference genome (Ppap_1.0). We also aligned and called variants for *Phlebotomus bergeroti*

(N=2), and *Phlebotomus duboscqui* (N=1) as outgroups.　In addition, we called variants from the

raw reads of whole-genome samples of *Lu. longipalpis* collected from locations in Brazil [Marajó

(N=9), Lapinha (N=13), Jacobina (N=14), and Sobral (1S N=13; 2S N=16)] and *Nyssomyia*

intermedia (N=2) and *Migonemyia migonei* (N=2) as outgroups aligned to the *Lu. longipalpis*

reference genome (Llon_1.0).

All genomic reads were pre-processed by removing duplicate reads with Picard (v1.113), and

paired-end reads were aligned to the reference genome using bwa-mem [61]. Base position

differences (SNV) were based on the unique convergence from two variant calling software tools,

SAMtools [62] and VarScan 2 [63], using standard variant calling and filtering parameters that are

optimized for whole genome data with moderate coverage (10X-40X). These parameters included

a *P*-value of 0.1, a map quality of 10, a minimum coverage of three reads, and parameters for

filtering by false positives. After alignment and variant detection, we implemented a filter to

exclude variants that were clustered in groups of more than five variants per 500 bp. We finally

implemented backfilling to include homozygous reference calls for each site where a variant is

called in the final multi-sample variant call format (VCF) file for each individual when the coverage

exceeded three reads. Sites that did not exceed this threshold were included as missing diploid

genotypes.

SNP filtering

To aid in the quality assessment of variants, we excluded the genotypes having a genotype

quality (GQ) lower than 30 (i.e., minimum accuracy of 99.9%). We also applied hard filters on the

variants, excluding any variants having an average depth lower than 10 or higher than 200, a

Hardy-Weinberg equilibrium (HWE) *P*-value lower than 0.001, levels of missing genotypes higher

than 20%, and having a minor allele frequency (MAF) lower than 1%. The dataset used in

population structure inferences was pruned for linkage disequilibrium, excluding variants above

an r^2 threshold of 0.5 in sliding windows of 50 variants with a step size of 5 variants using PLINK

v.1.90 [64, 65]. Variants in linkage disequilibrium were pruned from the 6,390,876 sites using a

sliding window of 500 kb and a linkage disequilibrium threshold of 0.2 using SNPRelate v.x [66].

Genomics structure

Although low powered due to limited sampling, we made an initial attempt to identify regions

in the genome that may be contributing to differentiation between the populations. For the *Ph.*

papatasi samples, VCFtools v.0.1.15 [67] was used to run a sliding window analysis with a 5 kb

sliding window size, a 500 bp step size, and at least 10 variants per window [67]. After calculating

Tajima's D for each window within each population [Tunisia (TUN), Egypt (EGP); Afghanistan

(AFG)], we calculated pairwise population divergence using Wright's fixation index (F_{ST}). We

made three pairwise comparisons: i) TUN to NS; ii) TUN to AFG; and iii) NS to AFG. The

distributions of these results were not normal, so we relied on a percentile approach and selected

all 5 kb windows that met the 1st percentile for Tajima's D and the 99th percentile for F_{ST}. Windows

with fewer than 10 SNPs and windows with coordinates from 1-500 were eliminated. We then

searched for 5 kb windows that passed the following thresholds: i) low within-population Tajima's

D and ii) high F_{ST}. We looked for direct overlapping windows of high F_{ST} with low Tajima's D scores

and indirect overlap, allowing for a 10kb buffer on either end of each window we identified.

Individual ancestry was estimated using Admixture v.1.9 [74]. The analysis was performed for *K*

values (ranging from two to seven with 30 iterations per *K*). In order to better understand the

different solutions reported by Admixture, post processing of the Admixture results was performed

in CLUMPAK v.1.1 [75]. Principal component analysis (PCA) was performed in scikit-allel v.1.1.10

[76], following the methods described in [77]. Weir and Cockerham's F_{ST}, Nei's D_{xy}, and Tajima's

D were calculated using VCFtools, and using the python script popGenWindows.py

(https://github.com/simonhmartin/genomics_general). Single marker FLK test [79] was performed

using HapFLK v.1.4 [80].

Phylogenetic analysis

We explored ancestral phylogenetic relationships between individuals by building a neighbor-

joining (NJ) tree across the genome using the R packages *adegenet v.2.1.1* [68, 69], *ape v.5.1*

[70], *poppr v.2.7.1 [71]*, and *vcfR v.1.7.0 [72]* . For *Ph. papatasi,* we included both *Ph. bergeroti*

and *Ph. duboscqi* and used the later to root the trees. For *Lu. longipalpis* phylogenetic analysis

we included both *N. intermedia* and *M. migonei* and used *M. migonei* to root the NJ trees. We

evaluated node support using 1,000 bootstrap replicates [73].

dN/dS

Selective constraints on gene sequence evolution were estimated using the dN/dS statistic

calculated for orthologous group multiple sequence alignments. Protein sequences were

assigned to ortholog groups by cross-referencing the OrthoDB v8 catalog [56]. For ortholog

groups with one-to-many and many-to-many orthologs, a single protein sequence was chosen for

each species by choosing randomly, with uniform probabilities, from the sequences for each

species. Protein sequence multiple alignments were generated first using Clustal-Ω [81], and then

used to inform CDS alignments with the codon-aware PAL2NAL alignment program [82]. The

yn00 program from PAML v4.8 [83] was used to calculate dN/dS ratios for each pair of sequences

in each aligned orthologous group.

Results and Discussion

Sequencing and Genome Characteristics

The genome of *Ph. papatasi* is ~350 Mb and was completed in 2012 for community analysis and population comparisons (S1 Table). The assembly was built from an input of ~22.5X total sequence coverage and resulted in 139,199 contigs with an N50 of 5.8 kb and 106,826 scaffolds with an N50 of 28 kb. The draft genome of *Lu. longipalpis* (Llon_1.0) was also completed in 2012 and is approximately 154.2 Mb, more than two time smaller than the size of the *Ph. papatasi* genome, representing 38.9X coverage (S1 Table). There are 35,696 contigs with an N50 of 7.5 kb and 11,532 scaffolds with an N50 of 85.1 kb. Based on automated and manual annotations, the *Ph. papatasi* and *Lu. longipalpis* genomes are estimated to contain 11,216 and 10,311 protein-coding genes, respectively. The BUSCO analysis [84] indicated 86.5% and 86.1% completeness for the *Ph. papatasi* and *Lu. longipalpis* genomes respectively (S2 Table). Annotation was augmented with RNA-seq expression evidence from different life-cycle stages, multiple days post adult emergence, and after blood-feeding in uninfected and *Le. major*-infected blood for *Ph. papatasi* and *Le. infantum*-infected blood for *Lu. longipalpis* (S3 Table).

Orthology

To improve our understanding of the phylogenetic relationships, we generated a maximum likelihood phylogenetic tree using orthologous genes selected from an orthology dataset comprising 43 insect species, including 36 dipterans. Consistent with [55], the phylogenetic tree supported clustering of sand flies with Culicomorpha infraorder (mosquitoes and black flies) rather than with the Muscomorpha infraorder (*Drosophila* and *Glossina)* (Fig 2A). In addition, percent identity between orthologs is higher between sand flies and mosquitoes than between sand flies and fruit files, in agreement with the maximum likelihood phylogeny (Fig 2A). Sand flies and culicines have more than three times as many exclusively-shared orthologous groups than sand flies do with muscoids, also consistent with the maximum likelihood phylogeny (Fig 2B). Analysis

of individual gene phylogenies, however, shows great uncertainty with almost equal proportions

of phylogenies supporting clustering of sand flies with mosquitoes and with muscoids (S1 Fig.).

Figure 2: Molecular species phylogeny and ortholog sharing. (A) The quantitative maximum

likelihood species phylogeny computed from the concatenated superalignment of 1,627

orthologous protein-coding genes places the sand flies (Psychodomorpha) as a sister group to

the mosquitoes (Culicomorpha) rather than the flies (Muscomorpha), with all branches showing

100% bootstrap support. The Culicomorpha are represented by four *Anopheles* mosquito species

and *Culex quinquefasciatus* and the Muscomorpha include four *Drosophila* fruit fly species and

the tsetse fly, *Glossina morsitans*. Outgroup species represent Lepidoptera (*Bombyx mori*),

Coleoptera (*Tribolium castaneum*), Hymenoptera (*Apis mellifera*), and the phylogeny is rooted

with the phthirapteran human body louse, *Pe. humanus*. The inset boxplots show that single-copy

(1:1) and multi-copy (X:X) ortholog amino acid percent identity is higher between each sand fly

(Ll, *Lu. longipalpis*; Pp, *Ph. papatasi*) and *An. gambiae* (Ag) than *D. melanogaster* (Dm). Boxplots

show median values with boxes extending to the first and third quartiles of the distributions. (B)

The Venn diagram summarizes the numbers of orthologous groups and mean number of genes

per species (in parentheses) shared among the two sand flies (L. lon., *Lu. longipalpis*; P. pap.,

Ph. papatasi) and/or the Culicomorpha and/or the Muscomorpha. Analysis of ortholog sharing

shows that the sand flies share more than three times as many orthologous groups exclusively

with the Culicomorpha (*Anopheles* and *Culex*) compared to the Muscomorpha (*Drosophila,*

Glossina) (subsets highlighted with thin and thick dashed lines). Numbers of unique genes are in

italics. Colors in panel A and panel B match species and sets of species analyzed.

Synteny

Comparisons of the relative ordering of ortholog genome positions between *Lu. longipalpis*

and *Ph. papatasi* did not indicate evidence of macrosynteny (S2 Fig.). The comparison of ortholog

genome positions is complicated by the fragmented nature of the assembled genomes, with *Lu.*

longipalpis's 10,331 genes spread across 11,532 scaffolds and the 11,216 genes of *Ph. papatasi*

spread across 106,826 scaffolds. Synteny analysis is likely also affected by the high rate of

evolutionary divergence among insects [85] and *Ph. papatasi* and *Lu. longipalpis* are separated

by at least 100 million years [86]. Previous studies of anopheline species have likewise shown

that synteny is conserved over relatively short time spans compared with other organisms [48].

Microsynteny, however, was observed between the sand flies and between each sand fly

species with *An. gambiae* (S4 Table). Microsynteny between *Lu. longipalpis* and *Ph. papatasi* is

similar to what was previously reported for *An. gambiae* and *D. melanogaster* [87]. We identified

499 synteny blocks with a maximum of 24 genes, average of 4.2 genes, and a median of 3 genes.

By comparison, we found 1,037 synteny blocks with a maximum of 10 genes, an average of 2.5

genes, and a median of 2 genes between *An. gambiae* and *D. melanogaster.* Previous studies

reported 948 synteny blocks with a maximum of 39 genes between *An. gambiae* and *D.*

melanogaster [87].

<u>Transposable Elements</u>

Transposable elements (TEs) are ubiquitous repetitive sequences present in eukaryotic

genomes that can be an important factor affecting genome sizes and are thought to be one of the

driving forces of evolution [45]. Some insect genomes have less than 3% of TEs, while others

contain as much as 50% or more of TEs, associated with large genomic size differences [88]. Our

analysis indicated that the *Ph. papatasi* genome is composed of 5.65% of TE derived sequences

while the *Lu. longipalpis* genome contains only 0.57%, corresponding to the genome size

difference between two sand fly species. This difference in TE-derived sequence could be due to

the result of divergent evolutionary dynamics of some TE families or superfamilies, affecting either

their distribution (presence or absence of specific superfamilies) or their abundance (copy number

per superfamily) in the genome. Higher abundance of TE derived sequences, presence of full-

length TEs and the genome size expansion in the *Ph. papatasi* genome also could be due to

recently active TEs. Alternatively, genomic differences in TE content might be the result of intrinsic

genomic deletion patterns in *Lu. longipalpis*, due to the effective recognition and elimination

machinery removing these foreign sequences from the genome, as has been shown to occur in

Drosophila species [89].

The comparison of the TE content in both sand fly genomes indicated an expansion of all the

TE classes and orders in the *Ph. papatasi* genome. This multiplication was more pronounced in

elements belonging to the class II, or "cut-and-paste" TEs, and especially in non-autonomous

miniature inverted-repeat transposable elements (MITEs), representing up to 29-fold differences

between the two genomes. Expansion of MITEs suggests the recent activity of class II TEs in the

Ph. papatasi genome. On the other hand, class I elements, or "copy-paste" elements, including

the Long Terminal Repeat (LTRs) and non-LTRs, which traditionally are accountable for the

genome expansion, showed more subtle changes between the two sand fly genomes,

representing up to 4-fold difference. (Table 1).

Table 1: Transposable Elements

	Phlebotomus papatasi % genome	*Lutzomyia longipalpis* % genome
LTR retrotransposons	**0.41%**	**0.21%**
Bel	0.17%	0.09%
Mag	0.07%	0.04%
Pao	0.06%	0.01%
Mdg3	0.05%	0.02%
Gypsy	0.03%	0.02%
Mdg1	0.02%	0.01%
Osvaldo	0.01%	0.01%
Copia	0.01%	0.00%

Non-LTR retrotransposons		**0.95%**	**0.22%**
	L2	0.25%	0.04%
	RTE	0.21%	0.02%
	Jck	0.18%	0.04%
	CR1	0.16%	0.03%
	LOA	0.07%	0.02%
	I	0.05%	0.05%
	Loner	0.03%	0.01%
	Ocas	0.01%	0.01%
DNA transposons		**1.13%**	**0.05%**
	Tc1/mariner	0.96%	0.04%
	piggyBac	0.18%	0.00%
	Helitron	0.00%	0.01%
MITEs		**4.11%**	**0.14%**
	mTA	2.63%	0.01%
	m2bp	0.54%	0.01%
	m8bp	0.29%	0.10%
	m3bp	0.24%	0.00%
	otherMITEs	0.23%	0.00%
	m4bp	0.18%	0.03%
Unclassified TE sequences		**0.00%**	**0.01%**
Total, percent TE in genome		**5.65%**	**0.57%**

Immunity Genes

Several immune pathways are conserved among insects. These include the Toll, Immune

Deficiency (IMD), Janus Kinase/Signal Transducer and Activator of Transcription (JAK/STAT),

lectin, and encapsulation pathways. We found the Toll signaling pathway highly conserved in the

genomes of both sand fly species, including homologues of the upstream peptidoglycan

recognition proteins (PGRPs), and glucan binding protein (GNBPs) (S5 Table). Similarly, the IMD

and JAK/STAT pathways appear to be conserved among dipterans, including *Drosophila*, *Aedes*,

Anopheles and both sand fly species analysed in this study (S6 and S7 Tables).

Galactose-binding proteins (galectins) are a diverse family of proteins playing roles in

development and immunity [90]. Comparing the sand flies' galectin protein sequences with other

Diptera, both shared and independent orthologs were identified (S3 Fig.; S8 Table). Future

analyses evaluating *Leishmania* parasite interactions with the *Ph. Papatasi* and *Lu. longipalpis*

galectins may provide a better understanding of the mechanisms that influence restrictive versus

permissive vectorial competence due to the key role some galectins play in parasite establishment

and survival [91].

Fourteen genes related to TGF-beta or TGF-beta pathways were found in each of the sand

fly genomes (S9 Table) and 16 and 15 different MAPK gene loci were identified in the genome of

Lu. longipalpis and *Ph. Papatasi* genome, respectively (S10 Table). Interestingly, two

prophenoloxidase homologues were identified in each species (S11 Table). A TEP-1-like protein,

and a COX-like ortholog were also found in the genomes of both *Lu. longipalpis* and *Ph. Papatasi*

(S6 Table).

Blood Feeding Genes

We mapped *Ph. Papatasi* and *Lu. longipalpis* putative salivary genes deposited at the NCBI

to the sand fly assemblies (S12 Table). Equally well studied in phlebotomine sand flies are the

genes associated with digestive properties [93-102]. Here, we characterized the following

digestive gene families: Peptidases (S13 Table); Glycoside Hydrolase Family 13 (S14 Table);

Chitinase and Chitinase-like protein family (S15 Table); N-acetylhexosaminidases (S16 Table; S4

Fig.); Chitin deacetylases (S17 Table; S5 Fig.); and Peritrophin-like proteins (S18 Table; S6 Fig.).

A detailed analysis of GH13 genes, including amylases, maltases and sucrases has been

published elsewhere [103]. Aquaporins (AQPs) are required for the transportation of water and

other small solutes across cell membranes and are important for excreting water from the blood

meal. We have identified six AQP genes from both species of sand flies (S19 Table; S7 Fig.).

This is similar to the number present in mosquitoes (N = 6), but two and four less than *Drosophila*

and *Glossina*, respectively [104]. Members of each AQP group previously identified from insects

are present in the sand fly genomes.

<u>Circadian Rhythm Genes</u>

Many important aspects in sand fly biology such as hematophagy and host seeking are

controlled by the biological clock [105]. In *Lu. longipalpis*, the main clock genes and their

expression pattern throughout the day have been previously characterized [106, 107].

Orthologues of all the core circadian clock genes were found in the genome of both *Lu. longipalpis*

(S20 Table) and *Ph*. *papatasi* (S21 Table). Interestingly, cryptochrome evolution has been a

matter of great interest [105] and we similarly found compelling features in sand fly CRY gene

structure. We found both *CRY1 and CRY2* genes in *Ph. papatasi* but, surprisingly, we did not find

a *cry1* gene in *Lu. longipalpis* genome assembly (S8 Fig.). Although both sand fly species are

closely related, these data suggest that whereas *Ph. papatasi* seems to have a functional

mammalian-like clock closer to butterflies, mosquitoes, and other dipterans, with *CRY1* and *CRY2*

genes, *Lu. longipalpis* may have a circadian clock working with a mechanism more similar to that

found in triatomines, bees and beetles, presenting only *CRY2* gene. We can speculate that the

possible loss of *CRY1* in *Lu. longipalpis* genome may be related to a better adaptation of these

insects to living in caves and dark places or alternatively, is just missing in the current assembly.

Chemosensory Receptors

The sand fly olfactory receptor (OR) repertoires comprise 139 canonical ORs in *Lu. longipalpis*

and *Ph. papatasi*, plus one copy each of the odorant receptor co-receptor, *Orco* (S22 Table). We

identified 82 genes encoding 91 gustatory receptors (GRs) in *Lu. longipalpis*, and 77 genes

encoding 88 GRs in *Ph. papatasi* (S22 Table). Based on homology to *D. melanogaster*, these

include seven to eight sugar receptors and two carbon dioxide receptors [108]. Among the other

conserved GRs are orthologs to *DmelGr28a/b*, *DmelGr33a,* and *DmelGr66a.* The ionotropic (IR)

repertoires are relatively small, with 23 and 28 in *Lu. longipalpis* and *Ph. papatasi*, respectively

(S22 Table). Among these are orthologs of 13 of the 16 IRs expressed in *D. melanogaster*

antennae [109]. We focused on antennal IRs that are conserved among Diptera [110, 111] and

did not attempt to identify glutamate receptors or highly divergent IRs, therefore, the true number

of IRs could be greater than reported here. Three ORs and three IRs suspected to be missing in

the *Lu. longipalpis* references assembly were found in *de novo* assemblies of the field isolates

Jacobina 14 and/or Jacobina 23. A comprehensive analysis of the sand fly chemoreceptors was

published elsewhere [112].

Nine and ten members of the transient receptor potential (TRP) cation channel family are

found in *Lu. longipalpis* and *Ph. papatasi* genomes, respectively, and the phylogenetic tree

showed a separation of the different TRP subfamilies (S9 Fig.). The pickpocket (PPK) receptor

phylogenetic tree demonstrated a division of the six different PPK subfamilies (S10 Fig.) with 14

and 13 family members in *Lu. longipalpis* and *Ph. papatasi* genomes, respectively.

G-Protein Coupled Receptors

G-Protein Coupled Receptors (GPCRs) are a large family of membrane-bound proteins that

operate in cellular signal transduction and interact with a wide variety of chemistries including

small molecules, neuro-peptides, and proteins. These proteins play roles in essential invertebrate

functions [113]. We utilized a novel classifier to identify insect GPCRs [114] in both *Ph. papatasi*

and *Lu. longipalpis*, followed by validation and manual annotation of identified genes. Ninety-four

and 92 GPCRs from *Ph. papatasi* and *Lu. longipalpis,* respectively, were compared with other

insects with well characterized GPCRs such as *D. melanogaster, An. gambiae, Ae. aegypti and*

Pe. humanus (S23 Table). Class A (rhodopsin-like) is the most numerous class with ~50 genes

in each sand fly, and includes the opsins that are thought to function in visual processes and

circadian rhythm. Both sand flies have one opsin gene for each functional group, the long-

wavelength, short-wavelength, ultraviolet, rh7-like, and pteropsin. Classes B (secretin-like), C

(metabotropic glutamate-like) and D (atypical GPCRs) have fewer members, with ~20, ~10 and

~10 in each sand fly, respectively. Sand flies include GPCR genes absent from *D. melanogaster*

(ocular albinism) and absent in *Ae. aegypti* and *An. gambiae* (parathyroid hormone receptor);

both genes from class B.

Cytochrome P450 Monoxygenase Genes

Cytochrome P450s (CYPs or P450s) constitute a conserved enzyme superfamily with a

diverse array of functions, ranging from core developmental pathways to the detoxification of

xenobiotics [115]. The CYP gene repertoire (CYPome) plays an important role in insect

physiology and in the development of resistance to insecticides used for vector control. Here we

identified and manually curated 104 CYP genes in *Lu. longipalpis* (Data File S1) and 93 CYP

genes in *Ph. papatasi* (Data File S2). These numbers are similar to the number of CYPs identified

in the mosquito *An. gambiae* (n =100). In *Lu. longipalpis* all 104 CYPs are full-length genes,

compared to 34 full-length and 59 fragmented genes in *Ph. papatasi*, likely reflecting the more

fragmented genome assembly of *Ph. papatasi* compared to *Lu. longipalpis*.

The identified sand fly CYP genes belong to the four clans typically found in insects;

mitochondrial (Mito), CYP2, CYP3, and CYP4 clan [115]. Remarkably, both sand fly species have

an expanded CYP3 clan compared to *An. gambiae* (S11 Fig.). This expansion is mostly caused

by gains in the CYP9J/9L, CYP6AG, and CYP6AK subfamilies (S11 Fig.).

Other Groups

In addition, we identified core genes as well as non-coding RNAs in the siRNA, miRNA, and

piRNA pathways, suggesting that these regulatory mechanisms are fully functional in sand flies

(S24 Table). We have also annotated heat shock and hypoxia proteins (S25 Table), cuticular

proteins (S26 Table), hormonal signaling (S27 Table), insulin signaling (S28 Table), and

antioxidant (S29 Table) genes, as well as genes involved in vitamin metabolism (S30 Table).

Additional information about annotated gene families can be found in the Supplementary Results.

Population Structure

Genetic structure across the range of *Ph. papatasi*

Average genome coverage ranged from 8X-16X (mean=12X; S31 Table). A total of 6,390,876

sites passed the thresholds using variant calling methods, where at least one sample displayed

a variant at a reference coordinate. As expected, the *Ph. papatasi* samples showed the lowest

count of Single Nucleotide Variants (SNVs) (1.84-1.99M SNVs) while the two *Ph. bergeroti*

samples (mean SNVs = 3.26M), and the *Ph. duboscqi* sample (4.01M SNVs) contained a higher

variant count (S32 Table). We found a small percentage of singletons (unique SNV's) in the *Ph.*

papatasi samples (3.0%-4.3%) and 3,482 shared variant alleles among the *Ph. papatasi* samples.

We also calculated the transition: transversion ratios, inbreeding coefficients, and pairwise

relatedness (S32 Table).

For phylogenetic analysis the dataset was filtered by keeping only variants of the highest

quality, leaving 1,084,952 total variants: 284,341 for Afghanistan, 435,972 for Egypt, and 439,446

for Tunisia. The dataset used in population structure inferences was further pruned for linkage

disequilibrium, creating a final dataset containing 423,236 total variants.

We explored ancestral phylogenetic relationships between individuals by building a NJ tree

across the genome. The NJ tree clustered the *Ph. papatasi* individuals into three clades that

correlated to geographical location, with bootstrap values of 100 (S12A Fig.).

Admixture was used to estimate the individual ancestries. Admixture cross-validation errors

(CV) suggest that the number of genetic clusters that best explains the observed population

structure as *K*=2 (S13A Fig.), where the Afghanistan samples were distinct from the Tunisian and

Egyptian samples (S12C Fig.).

We next performed a PCA, which does not depend on any model assumption and can thus

provide a useful validation of the results of Admixture analysis. The PCA supported the

phylogenetic analysis, separating the individuals into three distinct clusters, with all individuals

from the collection site clustering together. Principal components 1 and 2 accounted for 20.1% of

the total variation (S12B Fig.).

We found no direct overlapping windows of high F_{ST} with low Tajima's D scores for any of the

Ph. *papatasi* populations. Next, we searched for windows that met the above criteria but included

a 20 kb (10 kb on either side of the window) to identify indirect overlaps. We identified 29 genes

that fell within in the indirect overlapping windows (S33 Table). Functional annotation revealed 3

tRNAs, 3 putative transcription factors, and a snoRNA as possibly under selective pressure, as

well as 9 genes involved in metabolic pathways.

Genomic evidence of cryptic species within *Lu. longipalpis sensu lato*

The average genome coverage ranged from 8X to 105X (mean = 47X; S34 Table). We

identified 4,821,847 variants across all the individuals. To aid in quality assessment of variants,

filtration was performed as described for *Ph. papatasi.* After filtration, 1,937,819 variants

remained, ranging from 206,588 for Marajó to 633,519 for Jacobina (Fig. 1). After filtration and

LD pruning, the dataset contained 1,059,627 variants.

Consistent with the phylogeny based on the chemoreceptor repertoire [112], the full genome

phylogeny clustered the populations into two clades based on song and pheromone type, where

Marajó and Sobral 2S (Burst, Sobralene) cluster together and Lapinha and Sobral 1S (Pulse,

Germ) cluster together (Fig. 3A). An analysis of the male copulatory courtship songs of males

collected from Lapinha and Sobral are in agreement with those previously recorded [41]. In these

three resampled populations, we observed the sub-type P2 in Lapinha, the sub-type P3 in Sobral

1S, and the burst-type in Sobral 2S.

Figure 3. *Lutzomyia longipalpis* population structure. Inferred population structure of *Lu.*

longipalpis individuals collected from Marajó (MAR; pink), Lapinha (LAP; blue), from Jacobina

(JAC; red), and Sobral, including Sobral 1S (S1S; orange) and 16 Sobral 2S (S2S; green). (A)

Rooted neighbor joining (NJ) radial tree. We included both *N. intermedia* (INT; yellow) and *M.*

migonei (MIG; purple) and used *M. migonei* to root the trees. Bootstrap values represent the

percentage of 1,000 replicates. (B) Principal component analysis (PCA). Individuals were plotted

according to their coordinates on the first two principal components (PC1 and PC2). (C) Admixture

analysis. Ancestry proportions for Admixture models from $K = 2$ to $K = 7$ ancestral populations.

Each individual is represented by a thin vertical line, partitioned into K coloured segments

representing the individual's estimated membership fractions to the K clusters. These data are

the average of the major q-matrix clusters derived by CLUMPAK analysis.

Interestingly, the phylogenetic analysis separated the Jacobina population into two groups.

As expected, because flies from Jacobina are known to express a $C_{16} H_{32}$ pheromone and pulse-

like copulatory songs, some individuals clustered with Sobral 1S and Lapinha. Unexpectedly,

however, six individuals clustered with the diterpenoid-like pheromone and burst song expressing

individuals, suggesting that there is more than one population living in sympatry at the Jacobina

site. Male copulatory songs were not recorded for the Jacobina samples. However, sand flies

collected from two localities near to Jacobina, Araci and Olindina (Bahia state) (S14A Fig.) exhibit

different copulatory song patterns, suggesting the possible existence of two groups in Jacobina,

as observed in molecular data. Males from Araci exhibit the P1 copulation song pattern (S14B

Fig.), composed of train of similar pulses as previously described in males collected in Jacobina

[41]. Males from the nearby locality, Olindina, produced burst-type songs (S14B Fig.) with similar

pattern as Sobral 2S males [41]. The mean values of all song parameters observed from these

flies (S35 Table) are similar as previously reported [118].

In addition, the phylogenetic analysis indicated sub-structure within the Sobral 1S population.

The song tracings, however, did not suggest any sort of split. Although the analysis suggests

seven distinct populations, there is not enough statistical support to separate the six Jacobina

individuals from the Sobral 2S population, resulting in support for six populations.

The PCA based on the whole-genome clustered the individuals into six groups as well (Fig.

3B). Contrary to the phylogenetic analysis, however, the six Jacobina individuals were closely

clustered, but separate from the Marajó and Sobral 2S populations, which were indistinguishable.

PC1 explained 5.3% of the variation and separated the individuals collected from Jacobina into

two populations. Consistent with the NJ tree, the Sobral 1S population also exhibited some

population structure, the two clusters distinguishable through PC1 and PC2. PC2 accounted for

4.6% of the total variation and distinguished Lapinha from the other populations. The sympatric

Sobral 1S and 2S populations separate by both PC1 and PC2. Interestingly, while consistent with

Hickner *et al.* 2020, the whole-genome PCA allowed higher discriminating power among clusters

than the PCA based on the chemoreceptor repertoire which only identified 3-4 discrete clusters

[112].

Seven groups are clearly distinguishable from the Admixture analysis at *K*=7, consistent with

the PCA, NJ tree (Fig. 3C), and [112]. However, the cross-validation error analysis indicates 3-4

populations (S13B Fig.), one population consisting of all Marajó and Sobral 2S individuals and six

Jacobina individuals, one population made up of 8 Jacobina individuals and another population

with 7 Sobral 1S individuals. In contrast to the NJ tree that suggests that the individuals from

Lapinha make up a single population, the Admixture analysis indicates that all Lapinha individuals

and six Sobral 1S individuals are of similar ancestry. The analysis suggests no introgression

between the sympatric Sobral 1S and 2S individuals.

To identify candidate genomic regions contributing to reproductive isolation and to distinguish

between the two models of speciation, that with and without gene flow, pairwise measures of

divergence were calculated for Marajó, Sobral 1S, Sobral 2S, and Lapinha. Relative (Weir and

Cockerham's F_{ST}) and absolute (Nei's D_{xy}) measures of divergence were calculated for 1 kb non-

overlapping windows for all population comparisons, excluding Jacobina. Mean weighted F_{ST}

values indicate that genome wide differentiation is greater among population comparisons of

different pheromone and song types (Lapinha- Marajó, 0.214; Lapinha-Sobral 2S, 0.211; Sobral

1S-Sobral 2S, 0.116 compared to Lapinha – Sobral 1S, 0.154; Marajó-Sobral 2S, 0.114) and

allopatric populations (Lapinha- Marajó, 0.214; Lapinha-Sobral 2S, 0.211; Lapinha – Sobral 1S,

0.154 compared to Sobral 1S-Sobral 2S, 0.116) (S15 Fig.).

We identified genomic regions possibly contributing to population differentiation as F_{ST} outlier

windows that were in the top 2.5% quantile for each sympatric and allopatric comparison of

differing pheromone/song phenotype (S16 Fig.). There were 170 differentiation regions in

common among all of the different pheromone and song type comparisons (Fig. 4A). The mean

F_{ST} estimates were higher in the genomic regions shared by more than one comparison than in

those unique to each comparison, suggesting that these regions are being targeted by selection

in each case. Supporting the hypothesis that the Sobral populations have more recently diverged

from one another, the F_{ST} outlier windows had a mean value less than the allopatric populations

(Fig. 4B).

Figure 4. Genomic regions with high pairwise F_{ST} between the different populations of

***Lutzomyia longipalpis*.** (A) Venn diagram depicting the number of 1 kb non-overlapping genomic

windows having F_{ST} values in the top 2.5% quantile (outlier) among the different population

comparisons. (B) Box plots of outlier F_{ST} windows shared with another population comparison

(blue) or unique to a population comparison (pink). Box plots show the medians (lines) and

interquartile ranges (boxes); the whiskers extend out from the box plots to 1.5 times the

interquartile range, and values outside this limit are represented by dots. Mean F_{ST} values are

represented by open diamonds.

We further characterized the genomic regions by computing additional statistics in each

window. We tested if these regions were enriched for signatures of selection by computing

Tajima's D in the 1 kb non-overlapping windows, negative values of Tajima's D indicating a

potential selective sweep. As with the F_{ST} values, we considered outlier windows as those that

were in the lower or upper 2.5% quantiles (S17 Fig.). The vast majority of Tajima's D outlier

windows were unique to each population (S18 Fig.). No positive outlier windows overlapped

among the four populations (S18A Fig.) and only four negative outlier windows were shared

among all of the populations (S18B Fig.), of which only one contained a gene, LLOJ005792 of

unknown function. None of the Tajima's D outlier windows overlapped with the outlier F_{ST}

windows.

As absolute measures of divergence are less affected by within population levels of

polymorphism than relative measures of divergence, like F_{ST} [119], we calculated Nei's measure

of absolute divergence, D_{xy}, as an additional signature of selection. As expected, because these

populations are thought to have recently diverged from one another, the top 2.5% of D_{xy} values

were substantially lower than the outlier F_{ST} windows (Fig. 5A). The majority of outlier F_{ST} values

did not fall in the upper quantile of D_{xy} values (Table 2) and the windows with the highest D_{xy}

values did not overlap with the F_{ST} outlier windows (Fig. 5B), suggesting that there may be varying

levels of genetic diversity within each population.

Figure 5. Measures of divergence in 1 kb non-overlapping genomic windows between the

different populations of *Lutzomyia longipalpis*. (A) Box plots of D_{xy} and F_{ST} values in the top

2.5% quantile (outlier) of each population comparison. (B) Box plots of D_{xy} and F_{ST} values for

windows having both high D_{xy} and high F_{ST} (differentiation islands). (C) Box plots of Tajimas' D

values for the differentiation islands. (D) Box plots of FLK values for sites within differentiation

islands. Box plots show the medians (lines) and interquartile ranges (boxes); the whiskers extend

out from the box plots to 1.5 times the interquartile range, and values outside this limit are

represented by dots. Mean values are represented by open diamonds.

Table 2: *Lutzomyia longipalpis* Differentiation Island (DI) Statistics

	Lapinha-Marajó	Lapinha-Sobral 2S	Marajó-Sobral 1S	Sobral 1S-Sobral 2S

# 1 kb Windows	127,065	129,513	127,065	131,430
# F_{ST} Outlier Windows	3,176	3,237	3,176	3,285
# Regions of Interest	729	841	740	1,023
% F_{ST} Outlier Non-DI	77.05	74.03	76.70	68.87
Mean F_{ST} DI	0.93	0.85	0.85	0.86
Mean D_{xy} DI	0.30	0.29	0.29	0.29

There were 92 1 kb windows that fell in the upper 2.5% of F_{ST} and D_{xy} values and shared among all the comparisons. These windows were defined as Differentiation Islands (DI).

To identify genomic loci that may be contributing to the reproductive isolation of these

populations [29], we defined 'regions of interest' as those windows that fell in both the upper 2.5%

quantile of F_{ST} and D_{xy} values. There were 729, 841, 740, and 1023 regions of interest between

Lapinha-Marajó, Lapinha-Sobral 2S, Marajó-Sobral 1S, and Sobral 1S-Sobral 2S, respectively

(Table 2). The 92 regions of interest shared among all the population comparisons we interpreted

as 'differentiation islands' (DI).

We tested whether the DIs were enriched for signatures of selection by calculating Tajima's

D for these windows and performing a single marker FLK test [79] with HapFLK v. 1.4 [80]. The

Tajima's D (Fig. 5C) and FLK (Fig. 5D) values do not provide evidence that selection (either

balancing or positive) has led to the genomic divergence in these regions.

The genes present in the DIs are candidates that might explain the reproductive isolation of

the populations. The 92 DIs contained 35 genes, 25 of which had orthologues in *An. gambiae*

(S36 Table). Thirty-two of these genes were uncharacterized, LLOJ001208 is a protein MAK16

homolog, LLOJ009447 a rRNA adenine N(6)-methyltransferase, and LLOJ009732 a Lipase

maturation factor. No enrichment of gene ontology terms was identified using the *An. gambiae*

orthologs.

Conclusions

Our study provides the genome assembly and annotation of two divergent sand fly species that

will facilitate molecular and comparative studies of these medically important vectors. These

results provide a foundation for annotating and analyzing future chromosome length assemblies

generated from single sand flies. Global comparisons between sand fly vectors will greatly inform

the evolutionary relationships among these species and lead to advances in our understanding

of genes involved in important phenomena such as vectorial capacity, host-specificity, blood-

feeding, insecticide resistance, and immune system modulation.

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Supporting Information

Supplemental Methods: Methods used for manual annotation

Supplemental Results: Detailed information of annotated gene families

S1 Table: Assembly statistics

S2 Table: BUSCO analysis

S3 Table: RNAseq samples

S4 Table: Microsyntenny

S5 Table: Toll pathway annotation

S6 Table: Insect immune deficiency pathway annotation

S7 Table: JakStat pathway annotation

S8 Table: Galectin family annotation

S9 Table: Transforming growth factor-beta family annotation

S10 Table: Mitogen activated protein kinase family annotation

S11 Table: Prophenoloxidase family annotation

S12 Table: Salivary protein annotation

S13 Table: Peptidase annotation

S14 Table: Glycosidase Hydrolase family 13 annotation

S15 Table: Chitinase family annotation

S16 Table: Hexosaminidase family annotation

S17 Table: Chitinase deacetylase family annotation

S18 Table: Peritrophin family annotation

S19 Table: Aquoporin family annotation

S20 Table: *Lutzomyia longipalpis* circadian rhythm pathway annotation

S21 Table: *Phlebotomus papatasi* circadian rhythm pathway annotation

S22 Table: Number of chemoreceptor family genes

S23 Table: G-protein coupled receptor family annotation

S24 Table: MicroRNA annotation

S25 Table: Heat shock and hypoxia gene family annotation

S26 Table: Cuticular protein gene family annotation

S27 Table: Juvenile hormone family annotation

S28 Table: Insulin signaling pathway annotation

S29 Table: Antioxidant family annotation

S30 Table: Vitamin metabolism pathway annotation

S31 Table: *Phlebotomus papatasi* population sequencing median coverage depth

S32 Table: *Phlebotomus papatasi* population variant summary statistics

S33 Table: *Phlebotomus papatasi* F_{ST}-Tajima's D overlap (including 10kb upstream and

downstream)

S34 Table: *Lutzomyia longipalpis* population sequencing median coverage depth

S35 Table: Parameter values of male copulatory songs from *Lutzomyia longipalpis* from

Araci and Olindina

S36 Table: *Lutzomyia longipalpis* genes within differentiation islands

S1 Figure: Conflicting phylogenetic signals. Analysis of the gene phylogenies of individual

orthologous groups identified three major topologies with sand fly-mosquito (41%), sand fly-fly

(37%), or mosquito-fly (22%) sister clades. Comparisons of average branch lengths for each

topology suggest that, although substitution rates in flies are always higher, orthologs that support

the sand fly-mosquito topology show the lowest substitution rates in flies and the smallest

differences in substitution rates among the fly, sand fly, and mosquito clades. In contrast, the

sand fly-fly and mosquito-fly topologies show much higher substitution rates in flies and much

greater differences in substitution rates among the three clades.

1373

1374 **S2 Figure: Scatter plots of ortholog locations in each pair of genomes.** 1- to -1 ortholog

1375 groups were associated utilizing OrthoDB and scaffolds with only one gene were removed prior

1376   to generating plots. (A) *P. papatasi* vs. *Lu. longipalpis*. (B) *Lu. longipalpis* vs. *An. gambiae*. (C) *P.*

1377   *papatasi* vs. *An. gambiae*. (D) *Lu. longipalpis* vs. *D. melanogaster*. (E) *P. papatasi* vs. *D.*

1378   *melanogaster*. (F) *D. melanogaster* vs. *D. simulans*. (F) *D. melanogaster* vs. *An. gambiae*.

1379

S3 Figure: Clustering of sand fly galectin protein sequences. Condensed Neighbor-Joining

tree depicting clustering among galectin protein sequences of sand flies (*P. papatasi* and *Lu.*

longipalpis; open and filled squares, respectively), mosquitoes (*Ae. aegypti* and *An. gambiae*;

open and filled circles, respectively), fly (*D. melanogaster;* filled triangle), eastern oyster (*C.*

virginica; upside-down open triangle), and freshwater snail (*B. glabrata*; upside-down filled

triangle). Branches encompassing shared orthologs are highlighted by blue shades. Sand fly

specific clusters and genes are highlighted by orange shades. The evolutionary distances were

computed using the p-distance method and are in the units of the number of amino acid

differences per site. One thousand bootstrap replicates were performed, and only branches

displaying at least 50% confidence are shown.

S4 Figure: Condensed Neighbor-Joining tree depicting clustering among n-

acetylhexosaminidase protein sequences of sand flies (*P. papatasi* and *Lu. longipalpis*; open

and filled squares, respectively), mosquitoes (*Ae. aegypti* and *An. gambiae*; open and filled

circles, respectively), fly (*D. melanogaster*; filled triangle), and beetle (*T. castaneum*; filled

diamond). Branches encompassing sequences belonging to group I-IV n-acetylhexosaminidases

are highlighted by a blue shade. The sand fly specific cluster is highlighted by an orange shade.

The evolutionary distances were computed using the p-distance method and are in the units of

the number of amino acid differences per site. One thousand bootstrap replicates were performed,

and only branches displaying at least 50% confidence are shown.

1399

S5 Figure: Condensed Neighbor-Joining tree depicting clustering among chitin

deacetylase catalytic domain sequences of sand flies (*P. papatasi* and *Lu. longipalpis*; open

and filled squares, respectively), mosquitoes (*Ae. aegypti* and *An. gambiae*; open and filled

circles, respectively), fly (*D. melanogaster*; filled triangle), and beetle (*T. castaneum*; filled

diamond). Branches encompassing sequences belonging to group 1-5 and 9 CDA are highlighted

by blue shades. The evolutionary distances were computed using the p-distance method and are

in the units of the number of amino acid differences per site. One thousand bootstrap replicates

were performed, and only branches displaying at least 50% confidence are shown.

1408

S6 Figure: Condensed Maximum likelihood tree depicting peritrophin CBD domain

similarities among the sand flies *P. papatasi* and *Lu. longipalpis* and the red flour beetle *T.*

castaneum. Open squares, filled squares, and filled diamonds represent *P. papatasi*, *Lu.*

longipalpis, and *T. castaneum* domains, respectively. Branches exclusive to *T. castaneum* were

color-coded in magenta; those specific to sand flies were highlighted in blue. The branch

encompassing the CBD-like domain "CBDput" is highlighted in green. The branches shared by

sand flies and RFB CBD domains are color-coded in orange. Maximum likelihood tree was

constructed using the Whelan and Goldman (WAG) model with Gamma distributed among

Invariant sites (G+I), as suggested by the Model test function of the Mega6 software. One

thousand bootstrap replicates were performed, and only branches displaying at least 50%

confidence are shown.

1420

1421 **S7 Figure: Comparison of predicted aquaporins from other flies.** Neighbor-joining tree was

1422 produced using MEGA6 using Dayhoff Model and pairwise matching; branch values indicate

1423 support following 3000 bootstraps; values below 50% are omitted.

1424

S8 Figure: Molecular phylogenetic analysis of vertebrate and invertebrate photolyases

containing *Lu. longipalpis* and *P. papatasi* gene models. The different photoyases are

displayed on the right. The evolutionary history was inferred by using the Maximum Likelihood

method based on the Jones-Taylor-Thorton + four gamma categories with 1000 bootstrap

replicates (showing only above 65). Sequences with squares are vertebrate cryptochromes (black

– cry-4, white – cry-1, cry-2, and cry-3); sequences with black traingles represent (6-4) insect

photolyases; sequences with inverted black triangles are reprenting all insect photolyase repir

proteins; and sequences with a dot symbol show insect cryptochromes (black –cry-1, white –cry-

2). Dashed arrows point to *P. papatasi* photolyase sequences and straight arrows to *Lu.*

longipalpois photolyase sequences.

1435

1436 **S9 Figure: Molecular phylogenetic analysis of *Lu. longipalpis, P. papatasi* and *D.***

1437 ***melanogaster* TRP channel sequences.** The different TRP subfamilies are displayed on the

1438 right. The evolutionary history was inferred by using the Maximum Likelihood method based on

1439 the Whelan and Goldman +Freq. model with 1000 bootstrap replicates.

1440

1441 **S10 Figure: Molecular phylogenetic analysis of *Lu. longipalpis, P. papatasi* and *D.***

1442 ***melanogaster* PPK sequences.** The different PPK subfamilies are displayed on the right. The

1443 evolutionary history was inferred by using the Maximum Likelihood method based on the Whelan

1444 and Goldman +Freq. model with 1000 bootstrap replicates.

1445

1446 **S11 Figure: Comparison of predicted aquaporins from other flies.** Neighbor-joining tree was

1447 produced using MEGA6 using Dayhoff Model and pairwise matching; branch values indicate

1448 support following 3000 bootstraps; values below 50% are omitted.

1449

S12 Figure: *Phlebotomus papatasi* population structure. Inferred population structure of *P.*

papatasi individuals collected from Afghanistan (PPAFG; green), North Sinai - Egypt (PPNS;

purple), and Tunisia (PPTUN; orange). (A) Phylogenetic Analysis. Rooted neighbor joining (NJ)

radial tree generated with the Adegenet and ape packages of R. We included both *P. bergeroti*

(PBRG; black) and *P. duboscqi* (PDMA; gray), and used *P. duboscqi* to root the trees. Bootstrap

values represent the percentage of 1,000 replicates. (B) Principle component analysis (PCA).

Individuals are plotted according to their coordinates on the first two principal components (PC1

and PC2). (C) Admixture analysis. Ancestry proportions for Admixture models from *K* = 2 to *K* =

7 ancestral populations. Each individual is represented by a thin vertical line, partitioned into *K*

coloured segments representing the individual's estimated membership fractions to the *K* clusters.

These data are the average of the major *q*-matrix clusters derived by CLUMPAK analysis.

1461

1462   **S13 Figure: Admixture cross validation error**. Violin plot of the cross-validation error for each

1463   of 30 replicates for each *K* value. (A) *Phlebotomus papatasi* populations. (B) *Lutzomyia*

1464   *longipalpis* populations.

1465

1466   **S14 Figure: Male copulatory courtship songs from Araci and Olinda.** (A) Approximate

1467   distance of Araci and Olinda from Jacobina (B). Male copulatory courtship song tracings of

1468   *Lutzomyia longialpis* males collected from Araci and Olindina. The figure shows ~1 s of song in

1469   each case. Map Source: Google Imagery TerraMetrics.

1470

1471   **S15 Figure : Distribution plots of the pairwise F_{ST} between the different populations of**

1472   ***Lutzomyia longipalpis*.** Weighted F_{ST} values for 1kb non-overlapping windows were calculated

1473   across the genome for each population comparison.

1474

1475 **S16 Figure: Manhattan plots of the pairwise F_{ST} between the different populations of**

1476 ***Lutzomyia longipalpis.*** The red horizontal lines indicate the upper 0.05% of F_{ST} distribution over

1477 the entire genome.

1478

1479 **S17 Figure: Manhattan plot of Tajimas'D for each population of *Lutzomyia longipalpis*.** The

1480 red and blue horizontal lines indicate the upper and lower 0.05% of Tajima's D distribution,

1481 respectively.

1482

1483 **S18 Figure: Genomic regions with high (outlier) Tajimas'D for different populations of**

1484 ***Lutzomyia longipalpis.*** (A) The Venn diagram summarizes the numbers of 1kb genomic

1485 windows with Tajimas'D values in the upper 2.5% of the different populations. (B) The Venn

1486 diagram summarizes the numbers of 1kb genomic windows with Tajimas'D values in the lower

1487 2.5% of the different populations.

1488

1489 **S1 Data File:** *Phlebotomus papatasi* CYPome Fasta File

1490

1491 **S2 Data File:** *Lutzomyia longipalpis* CYPome Fasta File

1492

1493

Marajó
Sobralene
206,588
B

Sobral 2S
Sobralene
284,174
B

Sobral 1S
Germ
412,409
P3

Jacobina
Hima
633,519
P1

Lapinha
Germ
272,604
P2

300 km

Figure 1

Figure 2

Figure 3

Figure 4

Figure 5