Population Genomics of *Plasmodium vivax* in Panama to Assess the Risk of Case Importation on Malaria Elimination
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25 Abstract

Malaria incidence in Panama has plateaued in recent years in spite of elimination efforts, with 26 27 almost all cases caused by *Plasmodium vivax*. Notwithstanding, overall malaria prevalence 28 remains low (fewer than 1 case per 1000 persons). We used selective whole genome amplification to sequence 96 P. vivax samples from Panama collected between 2007 and 2019 to 29 30 study the population structure and transmission dynamics of the parasite. Imported cases resulting from increased levels of human migration could threaten malaria elimination prospects, 31 32 and four of the samples evaluated came from individuals with travel history. We explored 33 patterns of recent common ancestry among the samples and observed that a single highly 34 genetically related lineage was dominant among the samples (47 out of 59 samples with good 35 sequencing coverage), spanning the entire period of the collection (2007-2019) and all regions of the country. We also found a second, smaller clonal lineage of four parasites collected between 36 37 2017 and 2019. To explore the regional context of Panamanian P. vivax we conducted principal 38 components analysis and constructed a neighbor-joining tree using these samples and samples 39 collected worldwide from a previous study. Three of the four samples with travel history clustered with samples collected from their suspected country of origin (consistent with 40 41 importation), while one appears to have been a result of local transmission. The small number of 42 genetically unique Panamanian P. vivax samples clustered with samples collected from Colombia, suggesting they represent the genetically similar ancestral *P. vivax* population in 43 44 Panama or were recently imported from Colombia. The low diversity we observe in Panama 45 indicates that this parasite population has been previously subject to a severe bottleneck and may 46 be eligible for elimination. Additionally, while we confirmed that *P. vivax* is imported to Panama 47 from diverse geographic locations, the lack of impact from imported cases on the overall parasite

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48 population genomic profile suggests that onward transmission from such cases is limited and that49 imported cases may not presently pose a major barrier to elimination.

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51 Author Summary

52 Panama has greatly reduced *P. vivax* incidence, however, this progress has plateaued.

53 Understanding parasite transmission patterns and identifying imported cases is critical to help

54 Panama and other countries succeed in their elimination efforts. Genomic epidemiology and

55 population genomics can help provide information needed to inform malaria control policy. In

this study, we collected 100 Panamanian P. vivax samples from two collection periods (2007-

57 2009 and 2017-2019), of which 59 yielded usable sequencing data. 4 samples had patient travel

history data associated with them. We found that the majority of samples belong to a single

highly related lineage, termed CL1. This lineage has persisted since at least 2007. We also

60 highlight how genomic epidemiology can be used to spotlight parasites that may be imported as

a result of human migration, as well as corroborate or refute the country of origin as suggested

62 by patient travel history. We observe no evidence of outcrossing between these potentially

63 imported parasites and the local Panamanian parasite population, suggesting that imported

64 parasites are not driving ongoing malaria transmission in Panama. The low diversity we observe

65 in Panama indicates that this parasite population has been previously subject to a severe

66 bottleneck and may be eligible for elimination.

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68 Introduction

Malaria is a parasitic disease transmitted by the bite of female *Anopheles* mosquitoes. Malaria
parasites cause approximately 219 million cases and 435,000 deaths each year, the vast majority

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71 in sub-Saharan Africa (1). *Plasmodium falciparum*, the most virulent of the five *Plasmodium* 72 species that infect humans (P. falciparum, P. vivax, P. malariae, P. ovale, and P. knowlesi) 73 causes the majority of these cases (1). Though billions of dollars have been devoted to the 74 control and eradication of malaria caused by P. falciparum, comparatively little attention is given 75 to *P. vivax*, the most prevalent malaria parasite outside Africa (2). The impact of *P. vivax* on human health was once considered minimal relative to the more virulent *P. falciparum* (1,2). 76 77 However, recent studies suggest this parasite species causes a significant global health burden (1,2). The *P*.vivax cycle includes a dormant liver "hypnozoite" stage, unique among human 78 79 malaria parasites. (2) The hypnozoite stage can cause a relapse of malaria years after initial 80 infection, thus beginning the cycle of infection and complicating control efforts. (2) 81 Sixty percent of the population of Central and South America live in areas with ongoing 82 malaria transmission, predominantly caused by *P. vivax* (3). The region experiences about 700,000 P. vivax cases each year (1). Between 2000 and 2015 the incidence rate of malaria fell 83 37% globally and 42% in Africa (1). In the Americas during the same period, malaria mortality 84 85 decreased by 72% (1). Unfortunately, recent evidence suggests that this trend has stalled, and in some countries, malaria incidence has even increased (1). Panama eliminated the autochthonous 86 transmission of P. falciparum in 2010, outside of a small outbreak on the Colombian border in 87 2015 (4). Since 2010, P. vivax has caused almost all the malaria cases in Panama (1,5,6). P. 88 vivax cases in Panama have declined precipitously since 2005, from close to 1 case per 1000 89 90 persons, to under 0.25 cases per 1000 persons in 2017 (7). However, malaria incidence in Panama has plateaued since 2008. This plateau in incidence could be due to low levels of 91 transmission and/or imported cases that are re-seeding infections. 92

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93	Human movement leading to parasite migration is a potentially significant
94	epidemiological threat to malaria control in Panama. Parasite importation stemming from human
95	migration is a challenge to elimination programs in other countries around the world (8–10).
96	The unique geographic position of Panama makes it a crossroads for human migration to the
97	United States from South America (5,11). Migrants enter Panama through two paths: through
98	the Darien jungle region on the border with Colombia, and through the Kuna Yala Amerindian
99	reserve ('Comarca') on the Caribbean coast (5,6,11). Previous studies implicate these regions as
100	focal points of ongoing malaria transmission in Panama and suggest this is partly due to
101	imported parasites (5,11). It is estimated that approximately 60,000 continental and extra-
102	continental migrants crossed the southern border of Panama through the Darien jungle region in
103	2015 and 2016 (12).
104	To inform effective malaria elimination strategies in Panama, it is critical to
105	differentiate between imported infections or sustained local infection as the underlying cause of
106	persistent P. vivax incidence in Panama (4,5). Whole-genome sequencing can help paint a
107	detailed picture of parasite movement and transmission within and between countries (10,12,13).
108	However, P. vivax cannot be grown in vitro, and the difficulty of sequencing P. vivax from
109	clinical samples dominated by host DNA has hindered parasite population studies(14). Recent
110	advances such as hybrid capture (15) and selective whole genome amplification (SWGA)
111	mitigate this problem by allowing for parasite DNA to be selectively enriched before
112	sequencing(14). Both methods have allowed for population genomic studies of P. vivax using
113	samples directly from patients.

time span, with the aim of understanding patterns of genetic variation and recent shared ancestry

116 (relatedness) at different geographical and temporal scales. We found the vast majority of P. 117 vivax cases in Panama belong to a single highly related lineage that has persisted for at least a 118 decade. Furthermore, we observed a second smaller clonal lineage concentrated near the 119 Panamanian-Colombian border. We also found several samples that shared no relatedness with 120 any other sample, which may represent either localized pockets of outbred *P. vivax* transmission 121 or imported cases. These findings suggest the Panamanian *P. vivax* population has extremely low 122 genetic diversity and is on the cusp of elimination. Revealing these patterns of relatedness among 123 parasite infections can help inform best strategies for targeting interventions or case investigation 124 methods to increase the likelihood of successful elimination. The results obtained from this study 125 will help inform future elimination efforts in Panama and the rest of Meso-America. 126 **Results** 127 128 Sample Extraction, Sequencing, and Data Filtering 129 We successfully generated usable sequencing data from 35/60 (58%) Panamanian P. vivax 130 samples collected between 2007-2009 and 24/40 (60%) collected between 2017-2019, for a total 131 of 59 samples (Fig 1A). All Panamanian samples had an F_{ws} statistic greater than 0.95, indicating 132 that they were all monoclonal (Fig 1B). 133 We next analyzed the 59 Panamanian genomes in the context of 109 previously published P. vivax genomes, generating a filtered VCF consisting of 168 samples and 62,211 134 135 high-quality biallelic SNPs. 136 **Recent Common Ancestry Analysis Reveals Single Highly Related Lineage of Parasites** 137 We used hmmIBD (23) to estimate the proportion of the genome that is IBD among

138 Panamanian sample pairs to understand patterns of recent common ancestry. We subsetted the

139	VCF to contain only samples collected in Panama and Colombia to estimate pairwise IBD. We
140	strictly filtered sites on the basis of minimum and maximum read depth (five and thirty
141	respectively), resulting in a VCF with 89 samples and 15,788 sites for input into hmmIBD.
142	We observed a bimodal distribution of pairwise IBD in Panamanian samples, with peaks
143	near zero and 0.95 (Fig 2A). Forty-seven of the 59 Panamanian samples shared high IBD
144	(>0.875) with each other, indicating very recent common ancestry. Four other Panamanian
145	samples, all collected in the Kuna Yala Province, shared 100% IBD with each other, and 0-10%
146	IBD with any other sample, Panamanian or Colombian (Fig 2A and S1 Fig). Another four
147	Panamanian samples exhibited no IBD with each other nor any of the other Panamanian samples.
148	All four of these samples were collected in the Darien jungle region or Kuna Yala, which are the
149	two main points of entry for migrants traveling through Panama. These four samples drive the
150	modal peak of pairwise IBD at zero.
151	The variable degree of relatedness among the 47 samples sharing > 0.85 IBD suggested
152	that data quality potentially impacted the estimation of IBD. We plotted the relationship between
153	IBD and sample quality, measured by the average proportion of high coverage sites in each
154	sample pair, to determine if pairwise sample data quality affected the estimation of IBD (Fig
155	2B). We defined high coverage sites as sites with greater than 5x coverage (the cutoff for site
156	filtering). We observed that as the average proportion of high coverage sites for sample pairs
157	increased, pairwise IBD estimates correspondingly increased as well (Fig 2B). This relationship
158	suggests that poor data quality can lead to underestimation of IBD. It is possible that the majority
159	of the pairwise IBD estimations would be closer to one had the overall sample sequence quality
160	been higher. The prevalent highly genetically related lineage is referred to henceforth as cluster
161	one (CL1). CL1 samples share an IBD fraction of at least 0.875 with other samples in this

162	cluster. We also concluded that the four samples that shared 100% IBD with each other
163	constituted a second completely clonal lineage, henceforth referred to as cluster two (CL2).
164	Next, we examined how these two clusters and the other Panamanian samples not
165	belonging to either lineage were geographically distributed in Panama (Fig 3). Samples from
166	CL1 were found across Panama. Notably, samples collected from both 2007-2009 and 2017-
167	2019 were found in this lineage. The inclusion of samples from both collection periods
168	demonstrates that this lineage has persisted throughout Panama for at least a decade. We did not
169	find any evidence of structure in the <i>P. vivax</i> population by region or relative to the Panama
170	Canal, as was previously observed for <i>P. falciparum</i> (4).
171	We only observed samples belonging to CL2 in a specific locality, Puerto Obaldia, in the
172	Kuna Yala Amerindian territory (Comarca) along the Atlantic Coast. We lacked geographic
173	information for one of the four samples in CL2. Collection location data for the other three
174	samples show they were collected in roughly the same location, Puerto Obaldía in Kuna Yala.
175	Additionally, of four samples that shared no recent common ancestry with any sample in the
176	dataset, three were collected in Darien province, along the Colombian border and one was
177	collected in Kuna Yala.
178	After identifying two highly related lineages in Panama, we explored an approach for
179	determining whether the samples excluded from analysis due to low coverage could belong to
180	one of these lineages. We identified a set of 264 genotyped SNPs that were called in at least 80%
181	of samples across both Panama sample collection periods. We then calculated Nei's standard
182	genetic distance on all pairwise sample comparisons. The majority of excluded samples in both

- collection periods (17/21 and 4/16 for the 2007-2009 and 2017-2019 collection periods
- respectively) exhibited very low levels (0-1%) of genetic distance with CL1 samples (S2 Fig).

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Seven of the 2017-2019 samples exhibited 0-1% genetic distance with CL1 also showed 0-1%
genetic distance with CL2, making the identity of these samples uncertain. Three excluded
samples from 2007-2009 collected in the Darien Jungle Region had relatively high genetic
distance from all other samples in the dataset.

189 Exploring the regional context of Panamanian *P. vivax*

190 We built a neighbor-joining tree using the Panamanian samples plus previously 191 sequenced samples (20) (Fig 4A) to understand the Panamanian *P. vivax* population in a global 192 context. As noted in previous studies (20,30), we observed clusters of samples corresponding to 193 different geographic regions, with a large cluster of Central and South American samples. CL1 194 and CL2 formed distinct clusters within the Central and South American cluster with 100% 195 bootstrap support. CL2 is situated in a cluster containing samples from Colombia, with 100% 196 bootstrap support at deep nodes. While these four samples are clustered together with 100% support and short branch length, a long branch connects them to the rest of the Colombian 197 cluster. The Panamanian samples that shared zero IBD with either cluster also grouped with the 198 199 Colombian samples. These samples seem to share distant ancestry with each other and the rest of 200 the Colombian samples. The samples also formed their own sub-cluster within the Colombian 201 cluster.

PCA conducted with all samples worldwide showed tight clustering of all Central and
South American samples, with only one Panamanian sample falling outside this Central and
South American cluster (S3 Fig). PCA restricted to the samples collected from Central and South
America is heavily influenced by covariation among samples within the two clusters (S4 Fig).
PCA performed with a single consensus sequence representing each cluster revealed CL1
clusters with samples from Peru and Brazil and CL2 clusters with the Colombian samples (Fig

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208 4B). All four outbred Panamanian samples that shared no recent ancestry with the other samples 209 also clustered with the Colombian samples. Principal component one differentiated CL1 and 210 samples from Brazil and Peru from CL2 and the rest of the Central and South American samples. 211 212 Genomic data both support and refute travel history data 213 Four of the 59 samples had travel history data associated with them. All of these samples 214 with travel history data were collected during the 2007-2009 period. Travel history information 215 suggested that two samples were originally from Brazil, one sample from India, and one sample 216 from China. The two samples with Brazilian travel history fell within the Brazilian cluster on the

217 NJ tree and clustered with Brazilian samples on PCA (Fig 2A and 2B). The one sample with

218 Indian travel history grouped with the other Indian samples on the NJ tree, and clustered with the

other Indian samples via PCA as well (Fig 2A and S4 Fig). This sample was the only one

collected in Panama to fall outside of the Central and South American cluster in the PCA with

the worldwide sample set. For the two samples with Brazilian travel history and one sample with

Indian travel history, genomic data supported the same country of origin as the travel history

223 information.

The sample with Chinese travel history had a discrepancy between the region of origin suggested by its travel history information and its genomic data. This sample clusters with the Central and South American samples on the worldwide PCA instead of with the samples from China. This sample clustered with the Colombian samples in the PCA conducted with only the Central and South American samples (Fig 4B). Similarly, on the NJ tree, this sample fell within the Colombian cluster with 100% bootstrap support along with the four Panamanian samples that shared zero IBD with other Panamanian samples in the dataset.

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232 **Discussion**

Panama is on the cusp of eliminating malaria after several decades of intervention (5). 233 234 We found extremely low clonal diversity in the Panamanian P. vivax population, observing that 235 the majority of the successfully sequenced samples (47/59) belonged to a single highly related lineage, CL1. CL1 has persisted throughout Panama for at least a decade, in spite of ongoing 236 237 elimination efforts. Sample contamination could not explain this pattern as samples were 238 collected in two collection periods 10 years apart and extracted, amplified, and sequenced 239 separately. Our study suggests that the Panamanian *P. vivax* population has been through a 240 strong bottleneck due to reduced transmission, resulting in the majority of the population belonging to a single highly related lineage. Several studies of the relationship between *P*. 241 242 *falciparum* genetic diversity and transmission intensity from Senegal (30), Thailand (31, 32), 243 Colombia (32), have documented that when there is a decline in parasite transmission, there is 244 an increase in the clonal composition of the population. Furthermore, there is evidence of 245 persistence and transmission of *P. falciparum* clonal lineages in Colombia (33), Ecuador (34), 246 and Haiti (35). This study demonstrates a similar relationship in Panama between low transmission and extremely low genetic diversity of the *P. vivax* population. Almost all CL1 247 248 samples (46/47) share IBD > 0.95 with at least one other CL1 sample, suggesting a substantial 249 fraction of this population is clonal. Interestingly, the CL1 consensus sequence clusters with samples from Brazil and Peru on PCA, suggesting a greater degree of *P. vivax* population 250 251 panmixia across Central and South America than previously understood, and that the population has become more fractured over time (20). Alternatively, CL1 could have been imported into 252 253 Panama at an earlier point in time and has supplanted the ancestral *P. vivax* population in the 254 country, becoming the predominant lineage. Both scenarios point to the need for further

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longitudinal genomic studies of *Plasmodium* parasites to better understand population dynamicsover space and time.

257 Previous studies have indicated that Panama has focal transmission in indigenous regions 258 (Comarcas) (5,6,11) Malaria transmission in Panama is increasingly concentrated in the 259 Comarcas, with the proportion of total malaria cases in Panama reported from Comarcas rising 260 from 41.8% in 2005 to 90% in 2019 (36). Prior work shows that low transmission can lead to 261 population structure in part due to variable malaria transmission (31). The finding that CL1 is 262 distributed ubiquitously throughout Panama is unexpected given the concentration of the malaria 263 epidemic within spatially separated regions of the country. The geographic distribution of CL1 264 suggests that parasites have historically moved throughout the country, founding new 265 populations or supplanting small existing ones. Case investigations and understanding human 266 movement patterns throughout Panama will be critical to achieving elimination. This study had some limitations. We were unable to generate high-quality sequencing 267 268 data from ~40% of the samples. Fortunately, we were able to extract a small subset of 264 269 genotypable SNPs from the excluded samples. We used this SNP set to calculate Nei's standard 270 genetic distance to determine if the excluded samples were genetically distant from CL1 or CL2. 271 We found that \sim 57% of the excluded samples were genetically similar to CL1 (S2 Fig). This 272 result indicates that our assessment of relatedness within the Panamanian P. vivax population is 273 not biased by parasitemia or other factors that could have affected sequencing success. This 274 finding also suggests that we did not miss additional genetically distinct circulating Panamanian P. vivax strains by excluding these samples. Seven samples from 2017-2019 exhibited low 275 276 genetic distance from both CL1 and CL2. However, these seven samples were missing ~60-80% 277 of calls even after stringent filtration. The high missing rate in these seven samples potentially

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278 explains the low genetic distance to most other samples in the dataset. Additionally, all samples 279 that did yield usable sequence data represented almost all localities across Panama. The 280 exception to this was a group of samples collected near the Panamanian-Costa-Rican border that 281 we were unable to generate usable sequence data from. However, most ongoing malaria 282 transmission in Panama occurs East of the Panama Canal, where most of the samples that 283 generated usable sequencing data were collected (6,11). Due to the geographic sampling 284 coverage of regions with ongoing malaria transmission, we believe these data are reflective of 285 the current state of the Panamanian P. vivax population. We also lacked geographic collection 286 data for two of the successfully sequenced samples, and they were excluded from the geographic 287 analysis. The lack of geographical data is unlikely to bias our conclusions since these samples 288 came from both different collection periods and regions. The two samples also constitute a small 289 proportion of samples in the final dataset.

290 Genomic epidemiology can help to support malaria elimination efforts in Panama in 291 multiple ways. First, genomic data can help identify genetically distinct cases that may be 292 imported. Panama sits at the crux of migration paths to the United States, and it is possible that 293 genetically distinct samples collected in Panama represent imported cases. Integrating travel history information with genomic data can help solidify the identification of imported cases. 294 295 Four samples had patient travel history information, and genomic data supported the presumed 296 country of origin for three of them. The fourth sample was collected from a subject with travel 297 history from China. However, it clustered with Colombian samples on the NJ tree and the PCA, 298 suggesting the infection was likely acquired somewhere in Central or South America, rather than 299 China.

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300	Second, genomic data will be critical to determine if imported parasites are contributing
301	to local transmission and/or admixing with the local parasite population. For example, we did not
302	observe evidence of admixture between the imported samples from India and Brazil and the
303	samples that comprise CL1, or evidence of onward clonal transmission of the imported samples.
304	Our data cannot distinguish whether CL2 is a native Panamanian parasite lineage or if it has been
305	imported from Colombia. However, the four CL2 samples displayed a genomic and
306	epidemiological pattern consistent with recent local transmission, as all CL2 samples are
307	virtually identical and were collected from the same municipality in 2019.
308	Overall, the existence of one main parasite genetic lineage exhibiting no recent evidence
309	of outcrossing with imported infections suggests that Panama is ripe for the elimination of <i>P</i> .
310	vivax. While case importation remains a threat, the lack of evidence of outcrossing suggests it
311	may not be sufficient to prevent elimination under present circumstances. Ongoing genomic
312	surveillance and case containment efforts will be needed, however, to mitigate the risk of
313	outbreaks resulting from imported cases and reversal of the impressive progress that has been
314	recently made towards malaria elimination in Panama.
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316	Materials and Methods
317	Sample collection
318	We collected 96 P. vivax samples from infected consenting volunteers identified through passive
319	or active surveillance by technicians from the National Department of Vector Control from the
320	Ministry of Health (MINSA) of Panama. The Gorgas Memorial Institute of Health Studies

321 Bioethics Committee gave the study ethical approval (Permits: DIDETEC/06/ET-PA/02 and

322 154/CBI/ICGES/17). Two groups of DNA samples from infected patients were used in this

323	study: 1) 56 DNA samples collected during 2007-2009 and 2) 40 DNA samples collected during
324	2017-2019. The Gorgas Memorial Institutional Animal Use and Care Committee (CIUCAL)
325	(Permit: 002/CIUCAL-ICCES-2012) approved the use of Aotus P. vivax AMRU-1 and SAL-1
326	infected blood samples as a source of control DNA. Patient blood samples were collected by
327	finger-prick with a lancet and spotted into EBF 903 Five Spot Blood Cards (Eastern Business
328	Forms, INC, SC, USA). The samples were then transported at ambient temperature to the
329	laboratory and stored at -20° C until processing. We also created thin and thick blood smears
330	from patient samples. The blood smears were stained with Giemsa for percent parasite density
331	determination, species identification, and stage differential counts. Each volunteer donated ~ 150
332	μL of blood.
333	Information survey
334	We collected demographic, geographic, socioeconomic, and epidemiological info from each
335	study subject using an epidemiological form developed for the Survey123 for ArcGIS online
336	survey program (Esri, Redlands, CA).
337	Malaria microscopy
338	Giemsa stained thick and thin blood smears were examined by light microscopy for parasite
339	density determinations, Plasmodium species confirmation, and parasite lifecycle stage
340	count. Parasite densities were calculated by quantifying the number of malaria-infected red
341	blood cells (iRBCs) among 500 – 2000 RBCs on a thin blood smear and expressing the result as
342	% parasitemia (% parasitemia = parasitized RBCs /total RBCs) x 100).
343	DNA extraction:

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- 344 We extracted DNA from the filter paper blood spots using the Chelex method as described (16)
- for samples obtained during 2007-2009 and with the Qiagen® DNA mini kit for samples
- obtained during 2017-2019.
- 347 Molecular confirmation of *P. vivax* infection
- 348 We confirmed *P. vivax* infection for all samples collected during 2017-2019 by amplification of
- the *P. vivax* PVX_18SrRNA gene using a qRT-PCR assay as described (17).

350 Selective whole genome amplification and sequencing

- 351 We carried out DNA pre-amplification as described (14). Briefly, the thermocycler was
- preheated to 35°C. We dispensed aliquots of 37μl of Power SYBR Green Master Mix, plus 3μl
- phi39 into each PCR tube, next adding DNA and water to achieve a final volume of 47µl.
- Thermocycler settings were as follows: 35° C x 10 min; 34° C x 10 min; 33° C x 10 min; 32° C x 10
- min; 31° C x 10 min 30° C x 16 hours, 65° C x 10 min, 4° C for infinity. SWGA reaction products
- 356 were diluted with 50 µl of water. We purified 50 µl of the diluted product using 50 µl AmPURE
- beads according to the instructions of the manufacturer. We then eluted beads in 30 μ l of water.
- 358 Approximately 60-120 ng/µl of DNA was obtained after bead purification of the SWGA
- 359 reaction. We measured DNA concentration using Nanodrop quantitation.

360 Whole-genome sequencing

- 361 We performed whole-genome sequencing (WGS) on all 96 *P. vivax* samples using Nextera
- 362 libraries and an Illumina HiSeq X platform. Sample reads were aligned to the P01 reference
- 363 genome assembly using BWA-MEM, version 0.7 (18). Illumina sequencing reads are available
- 364 through the NCBI Sequence Read Archive with BioProject accession numbers
- 365 SAMN15722613–SAMN15722671.

366 SNP Discovery and Quality Filtering

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367	We marked duplicate reads using the MarkDuplicates tool from Picard tools. We next
368	performed local realignment around indels using the Genome Analysis Toolkit (GATK)
369	RealignerTargetCreator and GATK IndelRealigner (GATK Version 3.5.0). We called variants
370	using GATK HaplotypeCaller using best practices to call and filter SNPs and generate gVCFs
371	for each sample. We called samples in two batches, one containing samples collected in 2007-
372	2009 and one containing samples collected in 2017-2019. We performed joint variant calling on
373	the sets separately using GATK GenotypeGVCFs tool with GATK hard filters, including calls in
374	subtelomeric regions. The resulting VCFs consisted of 56 samples and 407,554 sites for the
375	2007-2009 samples, and 40 samples and 171,433 variants for the 2017-2019 samples. We
376	retained samples for analysis if they exhibited a minimum mean read depth of five and had calls
377	at more than 80% of variant sites in the VCF corresponding to their collection period, including
378	those in subtelomeric regions. We calculated and evaluated data quality measures using the
379	VCFtools package and custom R scripts (19). Thirty-five samples from 2007-2009 and 24
380	samples from 2017-2019 passed these filters and were kept for further analysis.
381	We next used GenotypeGVCFs tool to construct a joint VCF with the 59 Panamanian
382	samples plus a collection of previously collected global samples (Bioproject numbers
383	PRJNA240356-PRJNA240533 (20)). The joint VCF contained 168 samples and 2,4250,245
384	variants. We filtered sites on the basis of quality ($GQ > 40$), passing VQSR truth sensitivity level
385	of 0.99 or greater, missing rate (having a call at that site in $> 85\%$ of samples). We also
386	excluded any sites that were not bi-allelic and indels. The joint VCF generated after filtering
387	contained 168 samples and 62,211 sites.
388	Lastly, we generated a VCF containing SNPs found jointly in 80% of both the 2007-2009

and 2017-2019 samples. We also filtered sites in this VCF by excluding non-biallelic SNPs and

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390	on the basis of quality ($GQ > 30$), and passing GATK filters. The resultant VCFs contained 56
391	samples and 2335 SNPs for the 2007-2009 samples, and 40 samples and 1301 SNPs for the
392	2017-2019 samples. For these samples, we generated a highly filtered variant set containing
393	biallelic SNPs that passed the GATK filters (GQ $>$ 30, truth sensitivity level $>$ 0.99, Mean DP $>$
394	five) and were called in at least 80% of the samples from both time periods (2007-2009; 2017-
395	2019). Calls from the two sample sets were merged to create a unified VCF of 96 samples and
396	264 genotyped SNPs.

397 Determination of Sample Clonality

398 We estimated sample clonality using the F_{ws} statistic. F_{ws} measures the within-sample 399 genetic diversity (measured by heterozygosity H_w) relative to the overall population genetic 400 diversity (H_s) (21). The underlying theory assumes that a monoclonal (single strain) infection has 401 extremely low genetic diversity relative to overall population genetic diversity. By contrast, a 402 polyclonal (multiple strain) infection has high diversity relative to overall population diversity 403 (compared to a monoclonal infection). By estimating the ratio between within-host diversity and 404 population diversity, we can distinguish between monoclonal and polyclonal infections (21). A 405 sample with an F_{ws} statistic of 0.95 or greater (>=0.95) is considered monoclonal. We calculated 406 F_{ws} using the R package moimix (22).

407 Analysis of Recent Common Ancestry

We used hmmIBD (23) to estimate the proportion of sites identical by descent (IBD) between
sample pairs to ascertain recent common ancestry among Panamanian and Colombian samples
collected previously from the global *P. vivax* population study (20). We estimated minor allele
frequency (MAF) for IBD inference using the genetically distinct Panamanian samples, a
representative sample from each of the two highly related Panamanian clusters, and the

413	Colombian samples. We included Colombian samples to improve MAF estimation given the
414	greater clonal diversity of the Colombian parasite population and presumed historical gene flow
415	with Panama. We subsetted the master VCF file to keep only samples collected in Panama and
416	Colombia. Sites were excluded sites on the basis of minimum and maximum read depth (five and
417	thirty respectively) to ensure that we were using only high-quality SNPs. The input VCF for
418	hmmIBD contained 89 samples (59 Panamanian samples and 30 Colombian samples) and 15,788
419	variant sites. We then re-formatted the data using a custom perl script for input into hmmIBD
420	along with the MAF estimates. We conducted analysis and visualization of the hmmIBD output
421	using custom R scripts.
422	Analysis of Population Structure
423	We employed principal components analysis (PCA) and a neighbor-joining tree to study
424	the population structure of Panamanian samples in the context of the worldwide P. vivax
425	population (20). We used a strictly filtered SNP set for PCA, keeping only variants with calls in
426	at least 95% of samples. This input VCF consisted of 168 samples and 2,428 variants. We used
427	the R package SNPRelate to conduct PCA (24). Covariation within the two clusters heavily
428	influenced the PCA of all samples, so we also performed PCA using a single consensus sequence
429	for each cluster.
430	We used the R packages ape, StAMPP, pegas, and adegenet, (25–28) to generate the
431	neighbor-joining tree and genetic distance statistics. First, we calculated Nei's distance for all
432	pairwise sample combinations using the master VCF consisting of 168 samples and 62,211 sites
433	to generate a distance matrix. The distance matrix was used to generate a tree. We used the

- 434 bootphylo function in the ape package to bootstrap the dataset 100 times to estimate nodal
- 435 support. We then visualized the final tree with support values using the FigTree program

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- 436 (29). We used R software (R version 3.6.1) to carry out statistical analysis and data
- 437 visualization.
- 438
- 439

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24

Fig 1. Sequencing and Sample Assessment at Variant Sites. A) Distribution of variant site read coverage for each sample stratified by the collection period. Coverage values > 100 were censored for visualization purposes. Samples within the red boxes were kept for analysis. B) Distribution of F_{ws} values for all samples, stratified by the collection period. We interpreted F_{ws} values > 0.95 as evidence of sample monoclonality.

556

Fig 2. IBD analysis of the Panamanian samples. A) The distribution of pairwise IBD estimates among the Panamanian samples. IBD values near zero indicate no recent common ancestry. Values closer to one indicate that the sample pair are clonal or essentially clonal. B) Depicts pairwise IBD estimates for all Panamanian sample pairs with IBD > 0.875 plotted against the mean proportion of high coverage sites (sites with > 5x coverage) in each sample pair. The line indicates a linear regression, the box displays the Pearson correlation coefficient between the two axes variables.

564

565 Fig 3. Map of Panamanian sample collection sites. Sample colors show which cluster (or

neither) each sample belongs too. Shape indicates the sample collection period. The dotted line

- shows the location of the Panama Canal. The Blue Line shows the border of the Comarca KunaYala. The Darien Jungle Region is indicated by the green shaded area.
- 569

Fig 4. Population Structure. A) Neighbor-joining tree for all samples worldwide. Node
symbols denote support values: circles indicate 100% support, triangles indicate > 50% support.

572 Branch color indicates the country of collection for each sample. Panamanian samples with

573 travel history are noted with the colored stars. B) PCA of Central and South American samples.

- 574 Circle color indicates country of collection. Consensus sequences for cluster one and cluster two
- are noted as a square and triangle respectively. Panamanian samples with travel history areannotated.
- 577



Figure







- Peru
- Papua New Guinea
- Thailand
- Vietnam





Figure