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Carlos A. Guerra
University of Oxford, carlos.guerraloaiza@zoo.ox.ac.uk

Rosalind E. Howes
University of Oxford

Anand P. Patil
University of Oxford

Peter W. Gething
University of Oxford

Thomas P. Van Boeckel
University of Oxford

See next page for additional authors

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The International Limits and Population at Risk of Plasmodium vivax Transmission in 2009

Carlos A. Guerra1*, Rosalind E. Howes1, Anand P. Patil1, Peter W. Gething1, Thomas P. Van Boeckel1,2, William H. Temperley1, Caroline W. Kabaria3, Andrew J. Tatem4,5, Bui H. Manh6, Iqbal R. F. Elyazar7, J. Kevin Baird7,8, Robert W. Snow3,9, Simon I. Hay1*

1 Spatial Ecology and Epidemiology Group, Department of Zoology, University of Oxford, Oxford, United Kingdom, 2 Biological Control and Spatial Ecology, Université Libre de Bruxelles, CP160/12, Brussels, Belgium, 3 Malaria Public Health and Epidemiology Group, Centre for Geographic Medicine, KEMRI - University of Oxford - Wellcome Trust Collaborative Programme, Nairobi, Kenya, 4 Department of Geography, University of Florida, Gainesville, Florida, United States of America, 5 Emerging Pathogens Institute, University of Florida, Gainesville, Florida, United States of America, 6 Oxford University Clinical Research Unit, Bach Mai Hospital, National Institute of Infectious and Tropical Diseases, Ha Noi, Vietnam, 7 Eijkman-Oxford Clinical Research Unit, Jakarta, Indonesia, 8 Centre for Tropical Medicine, Nuffield Department of Clinical Medicine, Oxford University, Oxford, United Kingdom, 9 Centre for Tropical Medicine, Nuffield Department of Clinical Medicine, University of Oxford, CCVTM, Oxford, United Kingdom

Abstract

**Background:** A research priority for Plasmodium vivax malaria is to improve our understanding of the spatial distribution of risk and its relationship with the burden of P. vivax disease in human populations. The aim of the research outlined in this article is to provide a contemporary evidence-based map of the global spatial extent of P. vivax malaria, together with estimates of the human population at risk (PAR) of any level of transmission in 2009.

**Methodology:** The most recent P. vivax case-reporting data that could be obtained for all malaria endemic countries were used to classify risk into three classes: malaria free, unstable (<0.1 case per 1,000 people per annum (p.a.)) and stable (≥0.1 case per 1,000 p.a.) P. vivax malaria transmission. Risk areas were further constrained using temperature and aridity data based upon their relationship with parasite and vector bionomics. Medical intelligence was used to refine the spatial extent of risk in specific areas where transmission was reported to be absent (e.g., large urban areas and malaria-free islands). The PAR under each level of transmission was then derived by combining the categorical risk map with a high resolution population surface adjusted to 2009. The exclusion of large Duffy negative populations in Africa from the PAR totals was achieved using independent modelling of the gene frequency of this genetic trait. It was estimated that 2.85 billion people were exposed to some risk of P. vivax transmission in 2009, with 57.1% of them living in areas of unstable transmission. The vast majority (2.59 billion, 91.0%) were located in Central and South East (CSE) Asia, whilst the remainder were located in America (0.16 billion, 5.5%) and in the Africa+ region (0.10 billion, 3.5%). Despite evidence of ubiquitous risk of P. vivax infection in Africa, the very high prevalence of Duffy negativity throughout Central and West Africa reduced the PAR estimates substantially.

**Conclusions:** After more than a century of development and control, P. vivax remains more widely distributed than P. falciparum and is a potential cause of morbidity and mortality amongst the 2.85 billion people living at risk of infection, the majority of whom are in the tropical belt of CSE Asia. The probability of infection is reduced massively across Africa by the frequency of the Duffy negative trait, but transmission does occur on the continent and is a concern for Duffy positive locals and travellers. The final map provides the spatial limits on which the endemicity of P. vivax transmission can be mapped to support future cartographic-based burden estimations.

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* E-mail: carlos.guerraloaiza@zoo.ox.ac.uk (CAG); simon.hay@zoo.ox.ac.uk (SIH)
Introduction

The bulk of the global burden of human malaria is caused by two parasites: *Plasmodium falciparum* and *P. vivax*. Existing research efforts have focussed largely on *P. falciparum* because of the mortality it causes in Africa [1,2]. This focus is increasingly regarded as untenable [3–6] because the following factors indicate that the public health importance of *P. vivax* may be more significant than traditionally thought: i) *P. vivax* has a wider geographical range, potentially exposing more people to risk of infection [7,8]; ii) it is less amenable to control [9,10]; and, most importantly, iii) infections with *P. vivax* can cause severe clinical syndromes [5,11–16].

A key research priority for *P. vivax* malaria is to improve the basic understanding of the geographical distribution of risk, which is needed for adequate burden estimation [6]. Recent work by the Malaria Atlas Project (MAP; www.map.ox.ac.uk) [17] has shown *P. falciparum* malaria mapping to be a fundamental step in understanding the epidemiology of the disease at the global scale [18,19], in appraising the equity of global financing for control [20] and in forming the basis for burden estimation [21,22]. The benefits of a detailed knowledge of the spatial distribution of *P. vivax* transmission, and its clinical burden within these limits, are identical to those articulated for *P. falciparum*; establishing a benchmark against which control targets may be set, budgeted and monitored. Such maps do not exist for *P. vivax*, making any strategic planning problematic. In addition, information about the global extent of *P. vivax* transmission and population at risk (PAR) is crucial for many nations that are re-evaluating their prospects for malaria elimination [23,24].

This paper documents the global spatial limits of *P. vivax* malaria using a combination of national case-reporting data from health management information systems (HMIS), biological rules of transmission exclusion and medical intelligence combined in a geographical information system. The output is an evidence-based map from which estimates of PAR are derived. The resulting map also provides the global template in which contemporary *P. vivax* endemicity can be estimated and it contributes to a cartographic basis for *P. vivax* disease burden estimation.

Methods

Analyses Outline

A schematic overview of the analyses is presented in Figure 1. Briefly, *P. vivax* malaria endemic countries (*PvMECs*) were first identified and the following layers were progressively applied within a geographical information system to constrain risk areas and derive the final *P. vivax* spatial limits map: i) a *P. vivax* annual parasite incidence (*PvAPI*) data layer; biological exclusion layers comprising of ii) temperature and iii) aridity data layers; iv) a medical intelligence exclusion layer; and v) a predicted Duffy negativity layer. A detailed description of these steps follows.

Identifying *PvMECs*

Those countries that currently support *P. vivax* transmission were first identified. The primary sources for defining national risk were international travel and health guidelines [25,26] augmented with national survey information, pertinent published sources and personal communication with malariologists. Nations were grouped into three regions, as described elsewhere [19]: i) America; ii) Africa, Saudi Arabia and Yemen (Africa+); and iii) Central and South East (CSE) Asia. To further resolve PAR estimates, the CSE Asia region was sub-divided into West Asia, Central Asia and East Asia (Protocol S1).

Mapping case-reporting data

Methods described previously for mapping the global spatial limits of *P. falciparum* malaria [18] were used to constrain the area defined at risk within the *PvMECs* using *PvAPI* data (the number of confirmed *P. vivax* malaria cases reported per administrative unit per 1,000 people per annum (p.a.)). The *PvAPI* data were obtained mostly through personal communication with individuals and institutions linked to malaria control in each country (Protocol S1). The format in which these data were available varied considerably between countries. Ideally, the data would be available by administrative unit and by year, with each record presenting the estimated population for the administrative unit and the number of confirmed autochthonous malaria cases by the two main parasite species (*P. falciparum* and *P. vivax*). This would allow an estimation of species-specific API. These requirements, however, were often not met. Population data by administrative unit were sometimes unavailable, in which cases these data were sourced separately or extrapolated from previous years. An additional problem was the lack of parasite species-specific case or API values. In such cases, a parasite species ratio was inferred from alternative sources and applied to provide an estimate of species-specific API. There was, thus, significant geographical variation in the ability to look at the relative frequency of these parasites between areas and this was not investigated further. Finally, although a differentiation between confirmed and suspected cases and between autochthonous and imported cases was often provided, whenever this was not available it was assumed that the cases in question referred to confirmed and autochthonous occurrences.

The aim was to collate data for the last four years of reporting (ideally up to 2009) at the highest spatial resolution available (ideally at the second administrative level (ADMIN2) or higher). A geo-database was constructed to archive this information and link it to digital administrative boundaries of the world available from the 2009 version of the Global Administrative Unit Layers (GAUL) data set, implemented by the Food and Agriculture Organization of the United Nations (FAO) within the EC FAO Food Security for Action Programme [27]. The *PvAPI* data were averaged over the period available and were used to classify areas...
as malaria free, unstable (<0.1 case per 1,000 p.a.) or stable (≥0.1 case per 1,000 p.a.) transmission, based upon metrics advised during the Global Malaria Eradication Programme [28–30]. These data categories were then mapped using ArcMAP 9.2 (ESRI 2006).

Biological masks of exclusion of risk

To further constrain risk within national territories, two “masks” of biological exclusion were implemented [Protocol S2]. First, risk was constrained according to the relationship between temperature and the duration of sporogony, based upon parameters specific to *P. vivax* [31]. Synoptic mean, maximum and minimum monthly temperature records were obtained from 30-arcsec (~1 x 1 km) spatial resolution climate surfaces [32]. For each pixel, these values were converted, using spline interpolation, to a continuous time series representing a mean temperature profile across an average year. Diurnal variation was represented by adding a sinusoidal component to the time series with a wavelength of 24 hours and the amplitude varying smoothly across the year determined by the difference between the monthly

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**Figure 1. Flow chart of the various data and exclusion layers used to derive the final map.** The pink rectangle denotes the surface area and populations of *PvMECs*, whilst the pink ovoid represents the resulting trimmed surface area and PAR after the exclusion of risk by the various input layers, denoted by the blue rhomboids. Orange rectangles show area and PAR exclusions at each step to illustrate how these were reduced progressively. The sequence in which the exclusion layers are applied does not affect the final PAR estimates.

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minimum and maximum values. For *P. vivax* transmission to be biologically feasible, a cohort of anopheline vectors infected with *P. vivax* must survive long enough for sporogony to complete within their lifetime. Since the rate of parasite development within anopheles is strongly dependent on ambient temperature, the time required for sporogony varies continuously as temperatures fluctuate across a year [31]. For each pixel, the annual temperature profile was used to determine whether any periods existed in the year when vector lifespan would exceed the time required for sporogony, and hence when transmission was not precluded by temperature. This was achieved via numerical integration whereby, for cohorts of vectors born at each successive 2-hour interval across the year, sporogony rates varying continuously as a function of temperature were used to identify the earliest time at which sporogony could occur. If this time exceeded the maximum feasible vector lifespan, then the cohort was deemed unable to support transmission. If sporogony could not complete for any cohort across the year, then the pixel was classified as being at zero risk. Vector lifespan was defined as 31 days since estimates of the longevity of the main dominant vectors [33] indicate that 99% of anopheles die in less than a month and, therefore, would be unable to support parasite development in the required time. The exceptions were areas that support the longer-lived *Anopheles sergentii* and *An. superpictus*, where 62 days were considered more appropriate (Protocol S2) [18].

The second mask was based on the effect of arid conditions on anopheline development and survival [34]. Limited surface water reduces the availability of sites suitable for oviposition and reduces the survival of vectors at all stages of their development through the process of desiccation [35]. The ability of adult vectors to survive long enough to contribute to parasite transmission and of pre-adult stages to ensure minimum population abundance is, therefore, dependent on the levels of aridity and species-specific resilience to arid conditions. Extremely arid areas were identified using the global GlobCover Land Cover product (ESA/ESA GlobCover Project, led by MEDIAS-France/POSTEL) [36]. GlobCover products are derived from data provided by the Medium Resolution Imaging Spectrometer (MERIS), on board the European Space Agency’s (ESA) ENVironmental SATellite (ENVISAT), for the period between December 2004 and June 2006, and are available at a spatial resolution of 300 meters [36]. The layer was first resampled to a 1 km grid using a majority filter, and all pixels classified as “bare areas” by GlobCover were overlaid onto the PAPI surface. The aridity mask was treated differently from the temperature mask to allow for the possibility of the adaptation of human and vector populations to arid environments [37–39]. A more conservative approach was taken, which down-regulated risk by one class. In other words, GlobCover’s bare areas defined originally as at stable risk by PAPI were stepped down to unstable risk and those classified initially as unstable to malaria free.

**Medical intelligence modulation of risk**

Medical intelligence contained in international travel and health guidelines [25,26] was used to inform risk exclusion and down-regulation in specific urban areas and sub-national territories, which are cited as being free of malaria transmission (Protocol S3). Additional medical intelligence and personal communication with malaria experts helped identify further sub-national areas classified as malaria free in Cambodia, Vanuatu and Yemen. Specified urban areas were geo-positioned and their urban extents were identified using the Global Rural Urban Mapping Project (GRUMP) urban extents layer [40]. Rules of risk modulation within these urban extents were as follows: i) risk within urban extents falling outside the range of the urban vector *An. stephensi* [41] (Protocol S3) was excluded; ii) risk within urban areas inhabited by *An. stephensi* was down-regulated by one level from stable to unstable and from unstable to free (Protocol S5). Specified sub-national territories were classified as malaria free if not already identified as such by the PAPI layer and the biological masks. These territories were mapped using the GAUL data set [27].

**Duffy negativity phenotype**

Since Duffy negativity provides protection against infection with *P. vivax* [42], a continuous map of the Duffy negativity phenotype was generated from a geostatistical model fully described elsewhere (Howes et al., manuscript in preparation). The model was informed by a database of Duffy blood group surveys assembled from thorough searches of the published literature and supplemented with unpublished data by personal communication with relevant authors. Sources retrieved were added to existing Duffy blood group survey databases [43,44]. The earliest inclusion date for surveys was 1950, when the Duffy blood group was first described [45].

To model the Duffy system and derive a global prediction for the frequency of the homozygous Duffy negative phenotype (Fy[a-b-]), which is encoded by the homozygous FY*BES/FY*BES genotype), the spatially variable frequencies of the two polymorphic loci determining Duffy phenotypes were modelled: i) nucleotide −33 in the gene’s promoter region, which defines positive/negative expression (T-33C); ii) the coding region locus (G125A) determining the antigen type expressed: FyA or FyB [46]. Due to the wide range of diagnostic methods used to describe Duffy blood types in recent decades, data were recorded in a variety of forms, each providing differing information about the frequency of variants at both loci. For example, some molecular studies sequenced only the gene’s promoter region, and thus could not inform the frequency of the coding region variant; serological diagnoses only testing for the FyA antigen could not distinguish FyB from the Duffy negative phenotype. As part of the larger dataset, however, these incomplete data types can indirectly inform frequencies of negativity. Therefore, despite only requiring information about the promoter locus to model the negativity phenotype, variant frequencies at both polymorphic sites were modelled. This allowed the full range of information contained in the dataset to be used rather than just the subset specifically reporting Duffy negativity frequencies.

The model’s general architecture and Bayesian framework will be described elsewhere (Howes et al., manuscript in preparation). Briefly, the dataset of known values at fixed geographic locations was used to predict expression frequencies at each locus in all geographic sites where no data were available, thereby generating continuous global surfaces of the frequency of each variant. From the predicted frequency of the promoter region variant encoding null expression (-33C), a continuous frequency map of the Duffy negative population was derived.

**Estimating the population at risk of *P. vivax* transmission**

The GRUMP beta version provides gridded population counts and population density estimates for the years 1990, 1995, and 2000, both adjusted and unadjusted to the United Nations’ national population estimates [40]. The adjusted population counts for the year 2000 were projected to 2009 by applying national, medium variant, urban and rural-specific growth rates by country [47]. These projections were undertaken using methods described previously [48], but refined with urban growth rates being applied solely to populations residing within the GRUMP
urban extents, while the rural growth rates were applied to the remaining population. This resulted in a 2009 population count surface of approximately 1×1 km spatial resolution, which was used to extract PAR figures. The PAR estimates in Africa were corrected for the presence of the Duffy negativity phenotype by multiplying the extracted population by \([1 - \text{frequency of Duffy negative individuals}]\).

**Results**

*Plasmodium vivax* malaria endemic countries

A total of 109 potentially endemic countries and territories listed in international travel and health guidelines were identified [25,26]. Ten of these countries: Algeria, Armenia, Egypt, Jamaica (*P. falciparum* only), Mauritius, Morocco, Oman, Russian Federation, Syrian Arab Republic and Turkmenistan have either interrupted transmission or are extremely effective at dealing with minor local outbreaks. These nations were not classified as *Pv* MECs and are all considered to be in the elimination phase by the Global Malaria Action Plan [24]. Additionally, four malaria endemic territories report *P. falciparum* transmission only: Cape Verde [49], the Dominican Republic [50], Haiti [50,51] and Mayotte [52]. This resulted in a global total of 95 *Pv* MECs. Figure 1 summarises the various layers applied on the 95 *Pv* MECs in order to derive the limits of *P. vivax* transmission. The results of these different steps are described below.

**Defining the spatial limits of *P. vivax* transmission at sub-national level**

*PvAPI* data were available for 51 countries. Data for four countries were available up to 2009. For 29 countries the last year of reporting was 2008, whilst 2007 and 2006 were the last years available for 11 and six countries, respectively. For Colombia the last reporting year was 2005. No HMIS data could be obtained for Kyrgyzstan and Uzbekistan, for which information contained in the most recent travel and health guidelines [23,26] was used to map risk. With the exception of Namibia, Saudi Arabia, South Africa and Swaziland, which were treated like all other nations, no HMIS data were solicited for countries in the Africa+ region, where stable risk of *P. vivax* transmission was assumed to be present throughout the country territories. In Botswana, stable risk was assumed in northern areas as specified by travel and health guidelines [25,26]. Amongst those countries for which HMIS data were available, 16 reported at ADMIN1 and 29 at ADMIN2 level. For Southern China, Myanmar, Nepal and Peru, data were available at ADMIN3 level. Data for Namibia and Venezuela were resolved at ADMIN1 and ADMIN2 levels. In total, 17,591 administrative units were populated with *PvAPI* data. Protocol S1 describes these data in detail. Figure 2 shows the spatial extent of *P. vivax* transmission as defined by the *PvAPI* data, with areas categorised as malaria free, unstable (*PvAPI* <0.1 case per 1,000 p.a.) or stable (*PvAPI* $\leq$0.1 case per 1,000 p.a.) transmission [29].

**Biological masks to refine the limits of transmission**

Figure 3 shows the limits of *P. vivax* transmission after overlaying the temperature mask on the *PvAPI* surface. The *P. vivax*-specific temperature mask was less exclusive of areas of risk than that derived for *P. falciparum* [18]. Exclusion of risk was mainly evident in the Andes, the southern fringes of the Himalayas, the eastern fringe of the Tibetan plateaux, the central mountain ridge of New Guinea and the East African, Malagasy and Afghan highlands. There was a remarkable correspondence between *PvAPI* defined risk in the Andean and Himalayan regions and the temperature mask, which trimmed pixels of no risk at very high spatial resolution in these areas.

The aridity mask used here [36] was more contemporary and derived from higher spatial resolution imagery than the one used to define the limits of *P. falciparum* [18]. Figure 4 shows that the
The two international travel and health guidelines consulted [25,26] cite 59 specific urban areas in 31 countries as being malaria free, in addition to urban areas in China, Indonesia (those southern Iran and throughout Pakistan were largely reduced to unstable risk by the aridity mask.

Medical intelligence used to refine risk

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found in Sumatra, Kalimantan, Nusa Tenggara Barat and Sulawesi) and the Philippines (Protocol S3). A total of 42 of these cities fell within areas classified as malarious and amongst these, eight were found within the range of *An. stephensi*, as were some urban areas in south-western Yunnan, China. Risk in the latter was down-regulated from stable to unstable and from unstable to free due to the presence of this urban vector. In the remaining 34 cities and other urban areas in China, Indonesia and the Philippines, risk was excluded. In addition, 36 administrative units, including islands, are cited as being malaria free (Protocol S3). These territories were excluded as areas of risk, if not already classified as such by the *Pv* API surface and biological masks. In addition, the island of Aneityum, in Vanuatu [53], the area around Angkor Watt, in Cambodia, and the island of Socotra, in Yemen [54], were classified as malaria free following additional medical intelligence and personal communication with malaria experts from these countries.

Frequency of Duffy negativity

From the assembled library of references, 821 spatially unique Duffy blood type surveys were identified. Globally the data points were spatially representative, with 265 in America, 213 in Africa+ (167 sub-Saharan), 207 in CSE Asia and 136 in Europe. The total global sampled population was 131,187 individuals, with 24,816 (18.9%) in Africa+ and 33 African countries represented in the final database.

The modelled global map of Duffy negativity (Figure 5) indicates that the *P. vivax* resistant phenotype is rarely seen outside of Africa, and, when this is the case, it is mainly in localised New World migrant communities. Within Africa, the predicted prevalence was strikingly high south of the Sahara. Across this region, the silent Duffy allele was close to fixation in 31 countries with 95% or more of the population being Duffy negative. Frequencies fell sharply into southern Africa and into the Horn of Africa. For instance, the frequency of Duffy negativity in the South African population was 62.7%, increasing to 65.0% in Namibia and 73.5% across Madagascar. The situation was predicted to be highly heterogeneous across Ethiopia, with an estimated 50.0% of the overall population being Duffy negative.

Populations at risk of *P. vivax* transmission

The estimated *P. vivax* endemic areas and PAR for 2009 are presented in Table 1, stratified by unstable (*Pv* API < 0.1 per 1,000 p.a.) and stable (*Pv* API ≥ 0.1 per 1,000 p.a.) risk of transmission, globally and by region and sub-region. It was estimated that there were 2.85 billion people at risk of *P. vivax* transmission worldwide in 2009, the vast majority (91.0%) inhabiting the CSE Asia region, 5.5% living in America and 3.4% living in Africa+, after accounting for Duffy negativity. An estimated 57.1% of the *P. vivax* PAR in 2009 lived in areas of unstable transmission, with a population of 1.63 billion.

Country level PAR estimates are provided in Protocol S4. The ten countries with the highest estimated PAR, in descending order, were: India, China, Indonesia, Pakistan, Viet Nam, Philippines, Brazil, Myanmar, Thailand and Ethiopia. PAR estimates in India accounted for 41.9% of the global PAR estimates, with 60.3% of the more than one billion PAR (1.19 billion) living in stable transmission areas. The situation in China was different as, according to the *Pv* API input data, areas of stable transmission were only found in the southern provinces of Yunnan and Hainan, and in the north-eastern province of Anhui, which reported an unusually high number of cases up to 2007. The latter is in accordance with a recent report documenting the resurgence of malaria in this province [55]. Transmission in the rest of China was largely negligible, with *Pv* API values well below 0.1 case per 1,000 people p.a. Given the reported cases, however, these were classified as unstable transmission areas and the total PAR estimated within them, after urban exclusions, was 583 million.
people. All other countries reporting the highest PAR were in CSE
Asia, with the exception of Brazil and Ethiopia.

Discussion

We present a contemporary evidence-based map of the global
distribution of *P. vivax* transmission developed from a combination
of mapped sub-national HMIS data, biological rules of transmis-
sion exclusion and medical intelligence. The methods used were
developed from those implemented for *P. falciparum* malaria [18]
and can be reproduced following the sequence of data layer
assemblies and exclusions illustrated in Figure 1.

*Plasmodium vivax* is transmitted within 95 countries in tropical,
sub-tropical and temperate regions, reaching approximately 43
degrees north in China and approximately 30 degrees south in
Southern Africa. The fact that *P. vivax* has a wider range than *P.
falciparum* [18] is facilitated by two aspects of the parasite’s biology
[56]: i) its development at lower temperatures during sporogony
[31]; and ii) its ability to produce hypnozoites during its life cycle
in the human host [57]. The sporogonic cycle of *P. vivax* is shorter
(i.e. a lower number of degree days required for its completion)
and the parasite’s sexual stage is active at lower temperatures than
other human malaria parasites (Protocol S2) [31]. Consequently,
generation of sporozoites is possible at higher altitudes and more
extreme latitudes. In the human host, hypnozoites of *P. vivax*
temperate strains can relapse anywhere between months and
years after the initial infection, often temporally coincident
with optimal climatic conditions in a new transmission season
[10,57].

The resulting maps produced an estimate of 2.85 billion people
living at risk of *P. vivax* malaria transmission in 2009. The
distribution of *P. vivax* PAR is very different from that of *P.
falciparum* [18], due to the widespread distribution of *P. vivax* in
Asia, up to northern China, and the high prevalence of the Duffy
negativity phenotype in Africa. China accounts for 22.0% of the
global estimated *P. vivax* PAR, although 93.1% of these people live
in areas defined as unstable transmission (Protocol S4). An
important caveat is that PvAPI data from central and northern
China could only be accessed at the lowest administrative level
(ADMIN1) [Protocol S1]. The very high population densities
found in this country exacerbate the problem, inevitably biasing
PAR estimates, despite urban areas in China being excluded from
the calculations following information from the sources of medical
intelligence that were consulted [25,26]. Malaria transmission in
most of these unstable transmission areas in China is probably
negligible given the very few cases reported between 2003 and
2007. It is important to stress the necessity to access PvAPI data at
a higher spatial resolution from China (i.e. at the county level) in
order to refine these estimates and minimise biases.

### Table 1. Regional and global areas and PAR of *Plasmodium vivax* malaria in 2009.

<table>
<thead>
<tr>
<th>Region</th>
<th>Area (km²)</th>
<th>PAR (millions)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Unstable</td>
<td>Stable</td>
</tr>
<tr>
<td>Africa+</td>
<td>4,812,618</td>
<td>17,980,708</td>
</tr>
<tr>
<td>America</td>
<td>1,368,380</td>
<td>8,087,335</td>
</tr>
<tr>
<td>CSE Asia</td>
<td>5,848,939</td>
<td>6,127,549</td>
</tr>
<tr>
<td>West Asia</td>
<td>2,007,247</td>
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</tr>
<tr>
<td>Central Asia</td>
<td>3,156,574</td>
<td>1,277,219</td>
</tr>
<tr>
<td>East Asia</td>
<td>685,118</td>
<td>2,049,717</td>
</tr>
<tr>
<td>World</td>
<td>12,029,937</td>
<td>32,195,600</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Country</th>
<th>References*</th>
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</thead>
<tbody>
<tr>
<td>Angola</td>
<td>[68–73]</td>
</tr>
<tr>
<td>Benin</td>
<td>[68,70,71,74]</td>
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<tr>
<td>Botswana</td>
<td>[72]</td>
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<tr>
<td>Burkina Faso</td>
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</tr>
<tr>
<td>Burundi</td>
<td>[70–73]</td>
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<tr>
<td>Cameroon</td>
<td>[68,69,71–79]</td>
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<tr>
<td>Chad</td>
<td>[74]</td>
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<tr>
<td>Comoros</td>
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<tr>
<td>Congo</td>
<td>[68,70,71,73,74,76,77,80]</td>
</tr>
<tr>
<td>Côte d’Ivoire</td>
<td>[68–71,73,74,76,78]</td>
</tr>
<tr>
<td>Congo (DR)</td>
<td>[68,81]</td>
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<tr>
<td>Djibouti</td>
<td>[68,78]</td>
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<tr>
<td>Equatorial Guinea</td>
<td>[82]</td>
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<td>Eritrea</td>
<td>[71,73,76,77,83,84]</td>
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<tr>
<td>Ethiopia</td>
<td>[68–74,76–79,85]</td>
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<tr>
<td>Gabon</td>
<td>[68,71,86]</td>
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<tr>
<td>Gambia</td>
<td>[71,72,76,78]</td>
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<tr>
<td>Ghana</td>
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<td>Rwanda</td>
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<tr>
<td>São Tome and Príncipe</td>
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<td>Senegal</td>
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<td>Somalia</td>
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<tr>
<td>South Africa</td>
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<td>Sudan</td>
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<td>[70,71]</td>
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<td>Uganda</td>
<td>[69–74,76–79,95]</td>
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<tr>
<td>Tanzania</td>
<td>[68–72,76,77,79]</td>
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<tr>
<td>Zambia</td>
<td>[69–72,78,96]</td>
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<tr>
<td>Zimbabwe</td>
<td>[68,69,71]</td>
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</table>

*The cited references mostly document imported cases from Africa. Evidence of transmission of *P. vivax* in Guinea Bissau and Swaziland could not be found in the published literature.

doi:10.1371/journal.pntd.0000774.t001

doii.org/10.1371/journal.pntd.0000774.t002

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doi:10.1371/journal.pntd.0000774.t002
In Africa, the modelled prevalence of Duffy negativity shows that very high rates of this phenotype are present in large swathes of West and Central Africa (Figure 3). One of the functions of the Duffy antigen is being a receptor of *P. vivax* [46] and its absence has been shown to preclude infection with this parasite [38,59], although there are some reports that suggest the opposite [60–63]. There is no doubt that the African continent has a climate highly conducive to *P. vivax* transmission (Protocol S2). Moreover, dominant African *Anopheles* have been shown to be competent vectors of this parasite [62,64,65]. In addition, there is a plethora of evidence of *P. vivax* transmission in Africa, mostly arising from travel-acquired infections due to a combination of morphological similarity and the prevailing biogeo-graphical barriers driven by the high prevalence of Duffy negativity [60]. Despite the fact that the risk of *P. vivax* is cosmopolitan, PAR estimates in Africa were modulated according to the high limitations placed on infection by the occurrence of the Duffy negative trait. Consequently, the PAR in the Africa+ region accounts for only 3.5% of the global estimated *P. vivax* PAR. Although recent work has shown 42 *P. vivax* infections amongst 476 individuals genotyped as Duffy negative across eight sites in Madagascar [63], we have taken a conservative approach and consider it premature to relax the Duffy exclusion of PAR across continental Africa until this study has been replicated elsewhere.

Mapping the distribution of *P. vivax* malaria has presented a number of unique challenges compared to *P. falciparum*, some of which have been addressed by the methods used here. The influence of climate on parasite development has been allowed for by implementing a temperature mask parameterised specifically for the *P. vivax* life cycle. The question of Duffy negativity and *P. vivax* transmission has also been addressed by modelling the distribution of this phenotype and by allowing the predicted prevalence to modulate PAR. It is also worth noting that the accuracy of HMIS for *P. vivax* clinical cases, particularly in areas of coincidental *P. falciparum* risk, is notoriously poor [66], in part because microscopists are less likely to record the presence of a parasite assumed to be clinically less important. Here, HMIS data were averaged over a period of up to four years and used to differentiate malaria free areas from those that are malarious. Within the latter, a conservative threshold was applied to classify risk areas as being of unstable (*PvAPI*<0.1 per 1,000 p.a.) or stable (*PvAPI*≥0.1 per 1,000 p.a.) transmission [29]. We believe that this conservative use of HMIS data balances, to some extent, anomalies introduced by *P. vivax* underreporting and the correspondence of the biological masks and *PvAPI* data in many areas is reassuring.

The intensity of transmission within the defined stable limits of *P. vivax* risk will vary across this range and this will be modelled using geostatistical techniques similar to those developed recently for *P. falciparum* [19]. This modelling work will be cognisant of the unique epidemiology of *P. vivax*. First, in areas where *P. vivax* infection is coincidental with *P. falciparum*, prevalence of the former may be suppressed by cross-species immunity [67] or underestimated by poor diagnostics [66]. Second, there is the ability of *P. vivax* to generate hypnozoites that lead to relapses. These characteristics render the interpretation of prevalence measures more problematic [5]. Third, the prevalence of Duffy negativity provides protection against infection in large sections of the population in Africa [38,59]. An appropriate modelling framework is under development and will be the subject of a subsequent paper mapping *P. vivax* malaria endemicity using parasite prevalence data. These data are being collated in the MAP database, with nearly 9,000 *P. vivax* parasite rate records archived by 01 March 2010.

### Supporting Information

**Protocol S1** Defining risk of transmission of *Plasmodium vivax* using case reporting data. Document describing more extensively one of the layers used to create the final map. Found at: doi:10.1371/journal.pntd.0000774.s001 (2.87 MB DOC)

**Protocol S2** Defining the global biological limits of *Plasmodium vivax* transmission. Document describing more extensively two of the layers used to create the final map. Found at: doi:10.1371/journal.pntd.0000774.s002 (0.42 MB DOC)

**Protocol S3** Risk modulation based upon medical intelligence. Document describing more extensively one of the layers used to create the final map. Found at: doi:10.1371/journal.pntd.0000774.s003 (0.36 MB DOC)

**Protocol S4** Country level area and population at risk of *Plasmodium vivax* malaria in 2009. Country-level table of the estimated area and populations at risk of *P. vivax* malaria in 2009 Found at: doi:10.1371/journal.pntd.0000774.s004 (0.16 MB DOC)

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### Author Contributions

Conceived and designed the experiments: SIH. Performed the experiments: CAG REH APP PWG TPVB WHT CWK AJT BHM IRFE JKB RWS. Wrote the paper: CAG SIH.

### References


