## RESEARCH

# **Open Access**



# Subnational tailoring of malaria interventions to prioritize the malaria response in Guinea

Ousmane Oumou Diallo<sup>1†</sup>, Abdourahamane Diallo<sup>2†</sup>, Kok Ben Toh<sup>1†</sup>, Nouman Diakité<sup>2</sup>, Mohamed Dioubaté<sup>2</sup>, Manuela Runge<sup>1</sup>, Tasmin Symons<sup>3</sup>, Elhadj Marouf Diallo<sup>2</sup>, Jaline Gerardin<sup>1</sup>, Beatriz Galatas<sup>4</sup> and Alioune Camara<sup>2\*</sup>

## Abstract

**Background** In the context of high malaria burden yet limited resources, Guinea's national malaria programme adopted an innovative subnational tailoring (SNT) approach, including engagement of stakeholders, data review, and data analytics, to update their malaria operational plan for 2024–2026 and identify the most appropriate interventions for each district considering the resources available.

**Methods** Guinea's malaria programme triggered the SNT exercise with a list of decisions that could be informed with local data. The programme established an SNT team, which determined intervention targeting criteria; identified, assembled, and reviewed relevant data sources; stratified malaria risk and its determinants to inform geographical targeting for each intervention; and used mathematical modelling to predict the impact of different intervention mix scenarios. The SNT analysis was performed at the district level, excluding the urban area of Conakry.

**Results** Malaria incidence, malaria prevalence, and all-cause under-5 mortality were used for the epidemiological stratification of Guinea. Additional indicators relevant for decision-making including seasonality patterns, insecticide resistance, historical malaria interventions and vaccine coverage were also stratified. Stratified layers were used to inform the targeting criteria for each intervention to identify districts to prioritize for indoor residual spray, dual-action insecticide-treated nets, seasonal malaria chemoprevention (SMC), including number of cycles for each eligible district, malaria vaccine, and perennial malaria chemoprevention. Results of the SNT analysis were used to mobilize funding from the Global Fund for scale-up of dual-action nets and expansion of SMC.

**Conclusions** SNT allowed Guinea's national malaria programme to adapt their intervention strategy at the health district level, an unprecedented approach in the country. The use of local data to inform eligibility and prioritization allowed the programme to identify the optimal mix of interventions for each district and to successfully mobilize resources to support their plans.

Keywords Subnational tailoring, Malaria, Guinea, Stratification, Interventions, Prioritization

<sup>†</sup>Ousmane Oumou Diallo, Abdourahamane Diallo and Kok Ben Toh have equal contribution.

\*Correspondence: Alioune Camara alioune.camara@pnlp-guinee.org Full list of author information is available at the end of the article



© The Author(s) 2025. **Open Access** This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons licence, and indicate if changes were made. The images or other third party material in this article are included in the article's Creative Commons licence, unless indicated otherwise in a credit line to the material. If material is not included in the article's Creative Commons licence and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this licence, visit http://creativecommons.org/licenses/by/4.0/.

#### Background

Malaria is the primary cause of outpatient visits and hospitalizations in children under five years of age in Guinea despite years of intense malaria control efforts [1, 2]. Guinea's national malaria programme, the Programme National de Lutte contre le Paludisme (PNLP), developed a national strategic plan for malaria covering the period 2023 to 2027, which includes the deployment of pyrethroid-based insecticide-treated bed nets (ITNs), seasonal malaria chemoprevention (SMC), intermittent preventive treatment in pregnancy (IPTp), introduction of a malaria vaccine, and treatment of symptomatic cases. The national strategy development process also considered second-generation ITNs, indoor residual spraying (IRS), and perennial malaria chemoprevention (PMC, then known as intermittent preventive treatment for infants) as interventions of interest should additional resources become available.

Main funders for the malaria response in Guinea include the Global Fund against AIDS, Tuberculosis and Malaria (50%), the US President's Malaria Initiative (PMI) (40%), with World Bank (3%), and domestic sources (7%). As is the case globally [3], the total resources available for malaria response in Guinea are insufficient to fund all interventions necessary to achieve key targets.

In 2023, the PNLP underwent a thorough prioritization exercise to review the malaria control plan for 2024 to 2026, aligned with the Global Fund funding cycle (GC7). This exercise aimed to maximize the impact of the resources available for the fight against malaria in Guinea, following the subnational tailoring of interventions (SNT) approach recommended by the World Health Organization (WHO) [3, 4].

In brief, SNT consists of using local data and additional contextual information available to malaria control programmes to inform decision-making. In the context of strategic planning and prioritization, SNT is used to determine the most appropriate mix of interventions and strategies for a given area. Through SNT, intervention plans are defined that achieve optimum impact on malaria burden either resource-agnostically in a strategic plan, or within a specific resource envelope. SNT can also be used to inform how new tools (such as malaria vaccines) can be integrated most effectively within existing plans, or for dynamic review of plans as additional funding opportunities become available [3].

Guinea's PNLP implemented SNT to respond to six questions related to the prioritization and implementation of five prevention interventions contemplated in their strategy, within the available domestic and international resource envelope, (Fig. 1, Table 1). The questions were: which districts to prioritize for more effective vector control with (1) IRS or (2) dual-action Interceptor

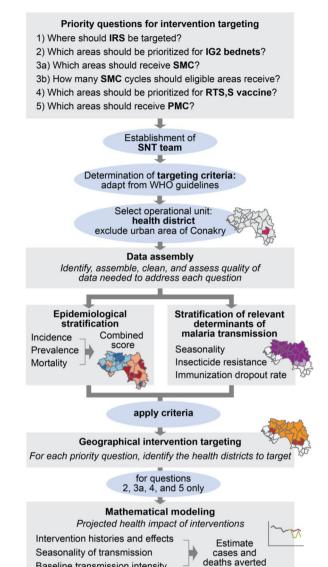


Fig. 1 Overview of the SNT intervention prioritization and modelling process. The Guinea PNLP identified priority questions to inform the prioritization of several interventions within their strategy. An SNT team was established and targeting criteria determined. Data were assembled for each health district. Districts were stratified according to malaria risk and relevant determinants for decision-making. Targeting criteria were applied to identify priority districts per intervention. Mathematical modelling was used to provide further evidence for a subset of questions

Baseline transmission intensity

G2 (IG2) bed nets); (3a) which districts to target with SMC, and (3b) how many cycles of SMC should targeted districts receive; (4) which districts to prioritize for the malaria vaccine RTS,S/AS01 (RTS,S) given the limited supply of doses available in 2023; and (5) which districts to prioritize for PMC. An additional question was raised regarding the tailoring of the malaria response in the capital city Conakry, where the malaria transmission

Question Indicator	All-age malaria clinical case incidence	Malaria infection prevalence estimates in children under 5	Estimates of all- cause mortality in children under 5	Case and rainfall season-ality	Entomological and insecticide resistance indicators	Immunization coverage	Intervention history
Data neec	Data needs Routine, demographic, and care-seeking behaviour information	Household surveys and geospatial modelled estimates	Household surveys and geospatial modelled estimates	Monthly routine case data and rainfall data	Entomological data and vector mortality after 24 h of exposure to pyrethroids and other insecticides	Routine EPI data	Historical intervention implementation data
1: Where should IRS be targeted?	×	×	×		×		×
2: Which areas should be prioritized with IG2 nets?	×	~	~		*		~
3a: Which areas should receive SMC?	×	×		×			×
3b: How many SMC cycles should eligible areas receive?	×	×		×			×
4: Which areas should be prioritized for RTS,S vaccine?	×	×	~			×	
5: Which areas should receive PMC, X and where should it be piloted?	*	×		×		×	

has significantly decreased and become increasingly more heterogeneous and focalized. Given the differential approaches and units of analysis required to address the first six questions and the last question, the SNT team split the exercise into two. This paper presents the activities and results obtained from the prioritization of interventions in areas outside of Conakry.

To address these questions, the SNT process was implemented by the PNLP through the following steps (Fig. 1):

- (1) Establishment of an SNT team. The PNLP created and led a local SNT team under direction of the PNLP programme manager. The SNT team was responsible for the oversight of the entire SNT exercise, including ensuring that consensus was reached by all relevant stakeholders at each step of the process and all actors were ultimately aligned under a single, collectively-discussed plan. The team included members from the PNLP, WHO, Northwestern University, the Global Fund, PMI, Catholic Relief Services, and RTI-Notre Santé. The PNLP regularly convened the SNT team until completion of the exercise.
- (2) For each intervention under consideration, determination of criteria to identify areas to target or prioritize. The first task of the SNT team was to determine the criteria that would inform the targeting of each of the interventions under consideration by the PNLP and to determine the operational unit for SNT. The SNT team took WHO's available guidance and practical information for the implementation of each intervention during the period of this exercise [5] as the basis for the criteria, and adapted them to their specific context. These criteria then established the types of information and data required for review and analysis (Table 1). The SNT team identified the health district (Fig. 2) as the lowest operationally feasible administrative unit at which interventions ought to be prioritized. Together, the criteria and the identification of the unit for SNT analysis established the data needs for the SNT exercise and the analytical outputs required to inform decisions.
- (3) Data collection and stratification of indicators required for decision-making. The SNT team identified and assembled all appropriate databases for analysis, then thoroughly assessed their quality. Next, the key indicators to inform about malaria transmission intensity patterns (clinical malaria incidence, prevalence, and mortality), and its determinants (including seasonality patterns, insecticide resistance, coverage of implemented malaria inter-

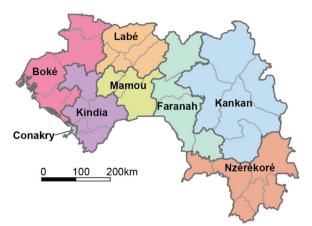


Fig. 2 Regions (bold lines and colors) and health districts (thin lines) of Guinea. Conakry was excluded from all analyses presented here

ventions, immunization coverage) were estimated and stratified into relevant categories for decisionmaking.

- (4) Geographic targeting of each intervention at subnational level. Using the relevant stratified layers obtained in the previous step, the SNT team applied the specified criteria to identify the districts in most need for each intervention. This process led to the development of various scenarios of intervention mixes under different resource conditions.
- (5) Mathematical modelling the impact of each intervention at subnational level. Transmission models were used to predict the impact on future malaria transmission and burden of the different scenarios, for those questions where mathematical modelling was appropriate.

The resulting prioritized mix of interventions was incorporated into Guinea's 2023 funding application to the Global Fund against AIDS, Tuberculosis, and Malaria, and was used to inform the activities considered under PMI's annual Malaria Operational Plans (MOPs). The SNT exercise also identified knowledge gaps for the understanding of the malaria epidemic and its determinants in specific parts of the country and highlighted the need for improved surveillance, data structuring and review systems, and data quality.

#### Methods

Thirty-three health districts were included in the SNT analysis to prioritize the malaria response in Guinea (Fig. 2).

#### Data assembly

Data needs for each priority question were identified (Table 1), and data were assembled from various sources within and outside the PNLP to estimate malaria risk and additional information of interest. Routine surveillance data were obtained from the District Health Information System 2 (DHIS2) of Guinea for the period 2018 to 2022 to estimate clinical malaria incidence and to inform case seasonality patterns. Indicators extracted per health facility included: all-cause outpatient visits, suspected malaria cases, tested malaria cases, confirmed malaria cases, treated malaria cases, all-cause hospitalizations, malaria hospitalizations, treated severe malaria cases, all-cause deaths, and malaria deaths. In total 694 health facilities that received outpatient consultations and hospitalizations, (563 public and 131 private), were included in this analysis. Health facility level data were reviewed for quality and completeness (see Supplementary File 1) and aggregated to the health district level for analysis. Population estimates per health district were obtained from the Institut National de la Statistique of Guinea [6].

Regional level estimates of malaria infection prevalence in children under the age of five (U5) were obtained from the 2012 and 2018 Demographic Health Surveys (DHS) or 2021 Malaria Indicator Survey (MIS), hereafter referred to collectively as DHS [7]. The DHS of 2012 and 2018 also provided regional level estimates of all-cause under-five deaths at the community level. Both of these indicators were used to estimate annual prevalence and all-cause mortality estimates at the district level.

Monthly rainfall data to inform seasonality patterns per district were obtained from ERA5 [8] for the period between January 2015 to December 2021, using the coordinates of the district centroid. Entomological data on pyrethroid resistance in 2022 were obtained from entomological sentinel surveillance sites throughout Guinea [9, 10].

Historical intervention implementation information available to the PNLP included: for SMC, history of SMC implementation per district from 2018 to 2022, including number and timing of SMC rounds; for ITNs, number distributed in the three most recent mass distributions of 2016, 2019 and 2022, number of nets distributed routinely, and type of nets distributed; for immunizations in 2021, number of children vaccinated for each of the three Diphtheria-Tetanus-Pertussis (DTP) doses and for the first measles dose. Data on treatment rate of fever and access to and use of mosquito nets were extracted from the DHS or MIS of 2012, 2018, and 2021 [11-13]. Care-seeking behaviour rates for a fever were calculated per sector of care (public or private) as the proportion of children U5 who had fever within the two weeks prior to the survey and sought care for the fever. Use of mosquito nets was calculated as the proportion of people who slept under a bed net on the night before the survey.

## Epidemiological stratification

#### Clinical malaria incidence estimates

To estimate community-level clinical malaria incidence and thereby measure malaria transmission per health district and year, four incidence estimates were considered following a standard approach used by the WHO in the World Malaria Report [3]. This approach adjusts crude incidence by the main factors that affect the final number of monthly confirmed cases reported per health facility in an additive fashion. The first adjustment aims to correct for varying testing rates of suspected cases. The resulting number of cases is further adjusted by the lack of reporting of data to DHIS2 for some health facilities and months. The final adjustment considers varying levels of care seeking of fevers outside of the public health sector [14]. All adjustments were made separately for each health district between 2018 and 2022 and presented to the SNT team for review. See Supplementary File 1 for full methods on incidence adjustments.

Maps per incidence indicator were produced per year and reviewed by the SNT team with the aim of identifying the most appropriate clinical malaria incidence metric and period of time for decision-making. Upon review of the assumptions and limitations of each estimate, a decision was made to use the median clinical malaria incidence adjusted for testing and reporting rates (second adjustment level) between 2018 and 2022. This was decided due to the geographical and temporal resemblance of the resulting estimates to the local understanding of transmission patterns after expert review.

#### Plasmodium falciparum infection prevalence estimates in children under 5

Estimates of Plasmodium falciparum parasite rate (*Pf*PR) at health-district aggregation were provided by the Malaria Atlas Project (MAP). These estimates were generated using a geospatial model taking as response data the routine case data described above, augmented by geo-located cross-sectional data on malaria parasitaemia by RDT, non-malarial fever incidence in the two weeks prior to survey, and care-seeking for fever from MIS 2021 (Symons et al. pers. commun.). After adjusting the routine data for missingness and care-seeking in steps logically consistent with the approach outlined above, monthly estimates of PfPR January 2015-December 2021 were inferred by reconciling the cross-sectional observations of RDT-positivity in children with the adjusted health-district-aggregated case counts via a semi-mechanistic Bayesian model accounting for (a) varying environmental and socio-economic receptivity

using high-resolution remotely-sensed covariates; (b) the non-linear prevalence-incidence relationship; and (c) the role of non-malarial febrile illness as a driver of care-seeking for incidental fevers. This joint inference procedure ensures biological consistency between estimates of PfPR and clinical incidence, in addition to informing the interpolation of PfPR trends between surveys with routine case data.

#### All-cause under-five mortality rate

Estimates of the all-cause under-5 mortality rate (U5MR) by health district were obtained from the Institute for Health Metrics and Evaluation (IHME) for the period 2000-2017 using a model that draws on birth history information collected in household surveys, census data, and other various mortality risk factors [15]. Ideally, estimates of malaria mortality would have been more appropriate for the stratification process. However, the precise estimation of malaria mortality in the community is difficult and uncertain due to underreporting of deaths and low sensitivity of the public health and surveillance system to capture the malaria deaths that occur in the community. For this analysis, the distribution and relative intensity of mortality are more important than the magnitude. Therefore, for the purpose of this exercise, the SNT team assumed that areas with higher relative U5MR rates coincided with those with high malaria mortality rates.

#### Risk categorization and combining indicators

A joint indicator of *Pf*PR, clinical malaria incidence adjusted for testing and reporting rates, and U5MR was used to obtain an estimate of overall malaria risk. This approach was carried out in two steps.

First, scores of 1 to 4 were assigned to the relevant categories of prevalence (<10%, 10–20%, 20–40% or >40%) and incidence (<100, 100–250, 250–450 or >450 cases per 1000 population at risk). The sum of the prevalence and incidence scores were reclassified into four morbidity groups on the basis of the combined scores, ranked from "lowest" (score of 4) to "highest" (score of 7).

Second, the morbidity indicator categories were reassigned to the scores from 1 (low) to 4 (high) and combined with the scored categories of U5MR (<9.5, 9-5-12.5, 12.5-15, or >15 deaths per 1000 live births). Health districts were reclassified into four groups on the basis of the combined morbidity and mortality scores, ranked from "lowest" (score of 6) to "highest" (score of 9). The resulting maps were reviewed and discussed with all SNT team members.

# Analysis of other indicators relevant for malaria decision-making

#### SMC eligibility and timing

Malaria case trends are generally seasonal and strongly dependent on rainfall, and SMC is recommended by WHO in areas of seasonal and moderate to high transmission [16]. Ideally, identification of areas with short, intense seasonality suitable for SMC would be done through analysis of seasonality patterns in the number of monthly or weekly cases reported. The preliminary analysis of seasonality of cases revealed ambiguous patterns of seasonality that did not align with the country's reality. The SNT team agreed that routine data was biased by factors such as care-seeking behaviour, reporting rates, and the impact of SMC and other interventions implemented right before the rainy season. Therefore, rainfall trends were used instead to evaluate suitability of seasonality for SMC in each health district. Districts with PfPR of 5% or higher were considered moderate to high transmission, which included all 33 districts in the analysis.

A health district was considered seasonal when at least 60% of cases (or rainfall) occurred during four months of the year. To evaluate which health districts met this definition, the same algorithm was applied to monthly confirmed malaria cases and monthly rainfall data from CHIRPS [17]. For each indicator, a sliding window of four months was applied to each consecutive 4-month block of data. Seasonality peaks were defined as those 4-month blocks where the sum of cases (or rainfall) within the block was at least 60% of the sum of cases (or rainfall) in a 12-month block beginning with the same month. A district was considered seasonal if at least 50% of evaluated years contained one or more consecutive seasonal month-blocks. Results of the case and rainfall evaluations for each district were reviewed by the PNLP prior to final determination of SMC eligibility.

Given that rainfall was used as a proxy to identify malaria case peaks, the trends in rainfall and cases were evaluated among districts with > 5% prevalence with suitable seasonality patterns for analysis. Districts with clear disparities between cases and rainfall trends were discussed to identify additional determinants of malaria trends, such as agricultural practices or presence of other environmental factors, that could explain said disparities. SMC eligible districts according to rainfall trends with additional determinants that contribute to longer or no seasonal trends in cases were considered ineligible for SMC. Finally, districts with previous history of SMC implementation were also considered eligible for SMC.

In Guinea, SMC is targeted to children U5, and monthly routine data on confirmed malaria cases in children U5 were used to determine the timing of SMC cycles for eligible districts. The beginning of the seasonal peak was defined as the month with an increase in cases of over 60% compared with the previous month. The proportion of cases that would be covered by four cycles, i.e. the fraction of annual cases that occur during months with SMC cycles, with SMC beginning in either June or July, was evaluated for all SMC-eligible districts.

Given a 4-cycle schedule beginning in July, two options were explored for the month in which to introduce a 5th cycle in districts where the rainfall analysis found a peak of five months. Ideally, the 5th cycle would be placed in the month before (June) or after (November) the July–October period that would maximize the number of cases averted. Thus, for each district eligible for five cycles, the proportion of cases occurring in June and November was assessed.

#### Immunization dropout rate

The number of children receiving the first, second and third doses of DTP vaccine and the first dose of measles vaccine were obtained for each district for 2021 using routine data from DHIS2. To identify districts with strong immunization implementation where the malaria vaccine or PMC could be successfully implemented, dropout rates were calculated for DTP1 to DTP2, DTP2 to DTP3, and DTP3 to measles. Each dropout rate was calculated by dividing the difference between the number of children receiving the later dose and earlier dose by the number of children receiving the earlier dose. After review, the DTP1 to DTP2 and DTP2 to DTP3 dropout rates were determined to have greater data quality and were considered when determining priority districts for malaria vaccine.

#### Intervention targeting

The criteria for each intervention, adapted from WHO guidelines, were then applied to Guinea's health districts according to the stratifications of epidemiology and malaria determinants described above to generate intervention mix maps relevant for each priority question.

#### Mathematical modelling for impact prediction

Impact predictions were made with EMOD v2.20, an agent-based mathematical model of malaria transmission [18]. The general approach used for adapting EMOD to Guinea was based on an approach used for Nigeria [19]. Full details of the modelling methods are available in Supplementary file 2.

EMOD was parameterized separately for each of the 33 health districts considered for intervention targeting, using district-level data where possible. A fourstep process was used for the modelling. First, past and present interventions in each district were added, including each intervention's schedule, coverage, and effect size. Second, the modelled case seasonality was matched to confirmed case seasonality in the routine data. Third, the modelled baseline transmission intensity in each district was fit to malaria prevalence and incidence data. These three steps are described in detail in Supplementary File 2. Last, impact predictions of intervention scenarios under consideration by the PNLP were modelled for the period 2023–2027 for uncomplicated malaria cases and malaria deaths in children U5 and in individuals of all ages.

The baseline scenario to which other possible intervention mixes were compared was selected by the PNLP as the intervention mix that was, in their opinion, likely to be funded. This base mix assumed that case management, SMC, and ITN usage would remain at 2020 levels. Districts designated as Group 1 priority (see "Results") received IG2 nets, whereas remaining districts received standard pyrethroid nets. SMC was implemented only in districts that were already receiving SMC in 2022, and no PMC or vaccine was implemented in any district.

For a subset of the questions of interest to the PNLP, the base intervention mix was compared to an alternate mix as follows:

- For Question 2: Compare base intervention mix to (1) base mix with pyrethroid nets in all districts and (2) base mix with IG2 nets in all districts.
- For Question 3a: Compare base intervention mix to base mix with SMC implementation in additional districts.
- For Question 4: Compare base intervention mix to base mix with PMC implemented in selected districts, with or without RTS,S vaccine distribution in Yomou district.

For each district, malaria transmission was simulated from 1960 to 2022, including all historical interventions as described earlier, for 15 stochastic realizations of each of the 20 selected parameter sets, a total of 300 model runs per district. For each intervention scenario under consideration, these 300 model runs were continued from 2023 to 2027.

Annual clinical incidence and deaths, for all ages and for children U5, were extracted for each simulation. Modelled incidence included both treated and untreated cases. A set of 1000 national level incidence and death rates were generated from the 300 model runs of each district with a weighted average accounting for the district population (see Supplementary file 2). Mean and 95% predicted intervals were calculated from the 1000 national estimates for each scenario.

#### Results

Α

Results are presented for the data quality assessment, epidemiological stratification of Guinea, and intervention targeting plans developed in response to each of the priority questions. Impact predictions from mathematical models are shown for scenarios of interest to the PNLP.

#### Data quality assessment

Crude incidence

(no adjustment)

The reporting rate was greater than or equal to 80% in most health districts and improved between 2018 and 2022 (Fig. S1.1 in Supplementary File 1). The proportion of missing data for total confirmed cases, suspected cases, tested cases, and treated cases was very low (Fig. S1.2 in Supplementary File 1). The internal consistency check found that in almost all health facilities, the number of suspected cases was equal to the number of tested cases and the number of confirmed cases was lower than the number of tested cases (Fig. S1.4 in Supplementary

**Incidence adjustment 1** 

(testing rate)

File 1), indicating good coherency across these key indicators. However, other comparisons, such as the number of treated cases and the number of confirmed cases, suggested instances of presumptive treatment, which was highlighted as an area for case management improvement.

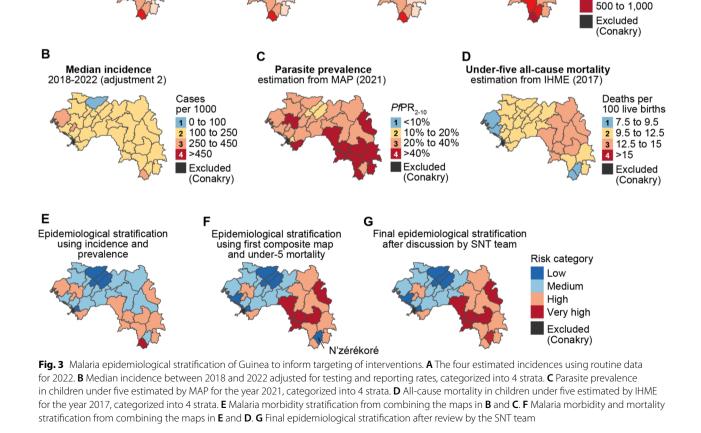
#### **Epidemiological stratification**

Crude and adjusted incidence per health district are shown in Fig. 3A for the year 2022 (see Supplementary file 1 for other years). Crude incidence was between 100 and 300 cases per thousand per year for most health districts and greater than 300 for only two health districts. Adjusting for testing and reporting rates did not have a large impact on estimated incidence (Fig. 3A and Fig. S1.7 in Supplementary file 1), as testing and reporting rates were high in most health districts (Figs S1.1 and S1.10 in Supplementary file 1). Adjusting for care-seeking

Incidence adjustment 3

(testing rate, reporting

rate, and care-seeking)



**Incidence adjustment 2** 

(testing rate and

reporting rate)

Cases per 1000

increased the estimated incidence to over 300 cases per thousand per year in 16 out of 33 health districts, as careseeking in the public sector was reported at 39.6–69.4% in the 2021 MIS, depending on the region (Fig. S1.5 in Supplementary file 1). Care-seeking in the public sector increased between the 2018 DHS and 2021 MIS and, therefore, the magnitude of the adjustment for care-seeking was larger for earlier years.

After review of the four incidence estimates, the PNLP made the choice to use the median incidence adjusted for testing and reporting rates when stratifying health districts by level of malaria risk (Fig. 3B). This choice was driven by the quality of routine data and concerns about the accuracy and district-level representability of the region-level care-seeking rates from the surveys. The selected median adjusted incidence was classified into three categories of transmission intensity following WHO definitions [20]: one health district was very low malaria transmission (less than 100 cases per 1000 per year); 29 health districts were low malaria transmission (between 100 and 250 cases per 1000 per year); and three health districts were moderate malaria transmission (between 250 and 450 cases per 1000 per year).

In 2021, the district-level infection prevalence estimates in children under five was over 40% (high transmission) in 11 health districts, concentrated in the regions of N'zérékoré, Kankan, and Faranah (Fig. 3C). Prevalence was < 10% (low transmission) in only four health districts, three of which were in the region of Labé. All-cause under-five mortality estimates were very high (over 12.5 per 100 live births) in 8 districts in the regions of N'zérékoré, Kankan, and Faranah.

Figure 3E shows a composite map of malaria risk based on morbidity indicators (median adjusted incidence and prevalence in under-fives). In this stratification, five health districts, mainly in the Labé region, were classified in the lowest transmission stratum. Only the district of Yomou, located in the N'zérékoré region, was classified in the very high risk stratum.

The morbidity composite map (Fig. 3E) was combined with stratified mortality (Fig. 3D) to produce a composite risk stratification based on both morbidity and mortality (Fig. 3F). Seven health districts were classified in the lowest risk stratum, in Labé, Boké, and Kindia regions. Districts in the low and medium strata were mostly in central or western Guinea, with the exception of N'zérékoré district in southeast Guinea being classified as low risk. Other districts in eastern and southern Guinea were in the high or very high risk strata. Routine data in N'zérékoré district had previously been shown to be of low quality and not representative of actual malaria incidence in the region [21]. Thus the SNT team decided that N'zérékoré district was more likely to belong to the very high transmission stratum as per their knowledge of the area (Fig. 3G).

#### **Targeting of IRS**

The WHO recommends IRS in areas of high receptivity, defined as *Pf*PR of at least 1% in 2000; all districts in Guinea met this criterion [22]. Since IRS is a costly intervention, the PNLP considered IRS only in districts with very high morbidity and mortality (Fig. 4), where it would have the highest potential impact, and districts in the region of N'zérékoré. Ultimately the PNLP decided not to include IRS in the 2023 funding request, although should resources become available, these ten districts would be the ones prioritized.

#### Targeting and potential impact of IG2 bed nets

Insecticide-treated nets distributed through mass campaigns and routine distribution are the main form of vector control implemented in Guinea. Entomological surveillance suggests that insecticide resistance is widespread across Guinea [9]. Thus, all districts would benefit from the distribution of new-generation mosquito nets, particularly those with a dual-action insecticide, such as IG2 nets. In the context of limited resources, the PNLP prioritized IG2 distribution for districts with high or very high morbidity and mortality (Fig. 3G), as well as the district of Forecariah, which had already received IG2 nets in the past as part of a pilot evaluation project. These districts were prioritized as Group 1 for IG2 nets, and the remaining districts formed Group 2 (Fig. 5A). Districts in Group 2 for prioritization would receive pyrethroid nets unless additional resources were to become available for IG2 nets. The Group 1 targeting scheme for IG2 bed nets, with SMC only in districts that have already been receiving SMC, formed the base scenario for mathematical modelling to which other intervention scenarios were compared (Fig. 5B).

Distribution of IG2 nets in the mass campaign of 2025 was projected to reduce a substantial number of cases in all age groups (Fig. 5C, Table 2), averting 22%



Fig. 4 Target districts for IRS, if it were to be included in Guinea's vector control strategy

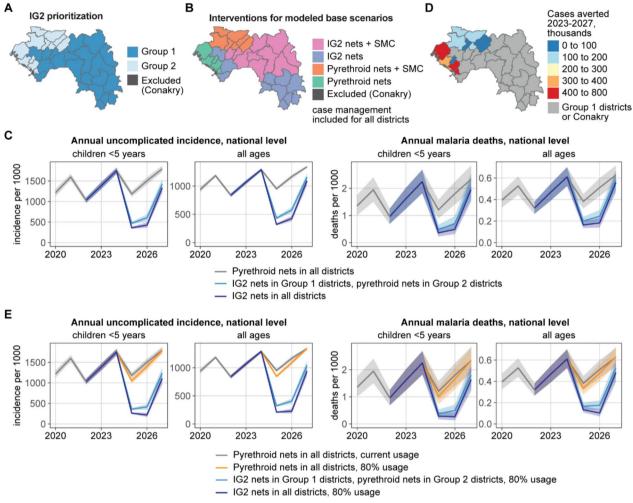


Fig. 5 Prioritization of districts for IG2 implementation and predicted impact of IG2 bed nets in Guinea. A Group 1 districts were prioritized for IG2 nets, and Group 2 districts would receive pyrethroid nets unless additional resources became available. B Base fundable package of interventions for modelled scenarios, to which additional interventions were added to assess impact. C Mean predicted incidence and death rate (solid line) and their 95% interval (shaded area) at the national level from 2020 to 2027 for IG2 in no districts, Group 1 districts only, and all districts. D Total cases averted from 2023 to 2027 in Group 2 districts if IG2 nets were implemented instead of pyrethroid nets. E Mean predicted incidence and death rate (solid line) and their 95% PI (shaded area) at the national level from 2020 to 2027 under various IG2 implementation and usage scenarios

of cases (95% PI: 21-24%) and 19% of deaths (95% PI: 14-25%) across all ages over the period 2023-2027 when distributed in Group 1 districts alone.

Implementation of IG2 nets also in Group 2 districts was predicted to further reduce the number of cases by 6% (95% PI: 5–6%) and the number of deaths by 5% (95% PI: 2–8%) in all ages, a smaller proportional reduction due to the lower malaria risk in Group 2 districts. Among individual districts in Group 2, the projected number of cases averted varied from a mean of 47 thousand to 630 thousand (Fig. 5D) due to differences in population size, transmission intensity, and ITN usage. Districts in the regions of Boké and Kindia were predicted to have the most cases averted if IG2 were to be implemented.

Modelled ITN usage ranged from 42 to 80% (Fig. S2.6 in Supplementary File 2). Under the implementation of pyrethroid nets in all districts, increasing net usage to at least 80% in all districts was predicted to avert 3% of cases (95% PI: 2–4%) in all ages (Fig. 5E, Table 3). Under IG2 implementation in Group 1 districts, increasing usage from current to at least 80% was also beneficial: an additional 7% of cases (95% PI: 5–8%) and 6% of deaths (95% PI: 0.4–12%) averted compared with IG2 implementation under current usage. At the national level, a similar impact was predicted if IG2 was also implemented in

**Table 2**Predicted impact of IG2 nets from 2023 to 2027,compared to the base scenario (Fig. 4B) with pyrethroid netsin all districts, and ITN usage based on the 2021 MIS. The 95%prediction interval is indicated in parentheses

	IG2 nets in Group 1 districts only	IG2 nets in all districts
Children U5 years		
Relative reduction in cases	26% (25–28%)	31% (29–33%)
Cases averted (millions)	3.5 (3.3–3.7)	4.1 (3.9–4.3)
Relative reduction in deaths	23% (16–30%)	28% (20–35%)
Deaths averted (thou- sands)	3.7 (2.5–5.0)	4.4 (3.1–5.9)
Allages		
Relative reduction in cases	22% (21–24%)	28% (26–30%)
Cases averted (millions)	12.4 (11.6–13.1)	15.5 (14.6–16.4)
Relative reduction in deaths	19% (14–25%)	24% (19–29%)
Deaths averted (thou- sands)	5.0 (3.5–6.6)	6.2 (4.7–7.8)

Group 2 or if IG2 was limited to Group 1 but usage was increased to 80%.

# SMC eligibility and potential impact of geographic expansion of SMC

Districts eligible for SMC were those with parasite prevalence over 5% in children under 5 years of age (Fig. 3C) and seasonal transmission suitable for this intervention. 27 districts met these criteria (Fig. 6A). Three observations were made after a visual comparison of trends in cases and rainfall (Fig. 6B, Fig. S1.14–S1.20 in Supplementary File 1): first, there were nearly always case peaks that occurred after the onset of rain even if those peaks were not detected by the seasonality algorithm; second, case peaks were sometimes longer than rainfall peaks; and last, there were districts with no clear case seasonality but clear rainfall peaks.

The district-by-district review of seasonality identified a few districts with highly seasonal rainfall, but where the seasonality of cases was not evident. In some districts of Kindia region, this phenomenon could be explained by agricultural activities and persistent puddles that remain after rainfall, sustaining transmission and widening the seasonal case peaks. In other districts, seasonality could be affected by the presence of mangroves and swamps. These factors were taken into account when determining the final assignment of SMC eligibility, leaving 19 districts eligible for SMC (Fig. 6C).

Three new districts (Télimélé, Kissidougou, and Kérouané) were selected for SMC in addition to the districts where SMC had already been implemented (Fig. 6D). Implementation of SMC over five years in the three expansion districts was predicted to avert 51% of cases (95% PI: 47–56%) and 58% of deaths (95% PI: 28–80%) in children U5 in these districts (Fig. 6E, Table 4). Additional cases were predicted to be averted in individuals over 5 years, as some children over 5 nonetheless receive SMC (see "Methods").

#### Selection of the number and timing of SMC cycles

SMC in Guinea has historically consisted of four monthly cycles. In areas determined to be eligible for SMC (Fig. 6D), the median number of confirmed cases in children U5 between 2018 and 2022 was used to determine the duration of the seasonal peak (Fig. 7A) by

**Table 3** Predicted impact of IG2 deployment if bed net usage were at least 80%, from 2023 to 2027, compared to the base scenario (Fig. 4B) with pyrethroid nets in all districts and ITN usage based on the 2021 MIS. The 95% prediction interval is indicated in parentheses

	Pyrethroid nets only, 80% usage	IG2 nets in Group 1 districts only, 80% usage	lG2 nets in all districts, 80% usage
Children U5			
Relative reduction in cases	3% (2 to 4%)	32% (31 to 34%)	38% (36 to 40%)
Cases averted (millions)	0.43 (0.31 to 0.55)	4.3 (4.1 to 4.5)	5.0 (4.8 to 5.3)
Relative reduction in deaths	4% (- 7 to 13%)	29% (22 to 36%)	34% (27 to 41%)
Deaths averted (thousands)	0.6 (- 1.0 to 2.2)	4.6 (3.3 to 6.0)	5.5 (4.2 to 6.8)
All ages			
Relative reduction in cases	3% (2 to 4%)	29% (28 to 31%)	36% (34 to 38%)
Cases averted (millions)	1.7 (1.3 to 2.2)	16.1 (15.3 to 17.1)	19.9 (18.9 to 20.9)
Relative reduction in deaths	3% (- 4 to 10%)	26% (20 to 31%)	31% (26 to 36%)
Deaths averted (thousands)	0.8 (- 1.0 to 2.6)	6.6 (5.1 to 8.3)	8.1 (6.5 to 9.7)

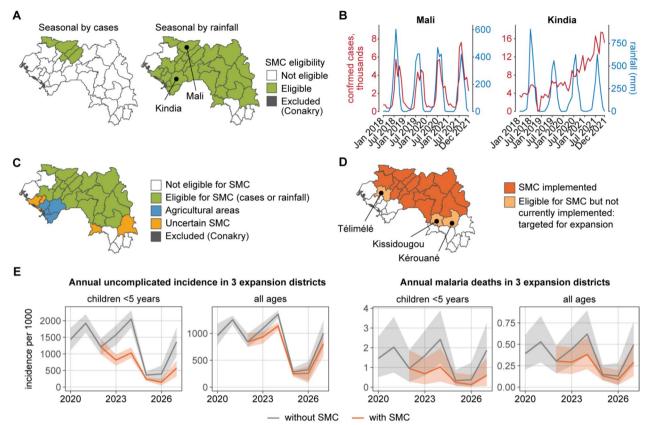


Fig. 6 Assessing eligibility for SMC and potential impact of SMC expansion to three additional districts. A Health districts eligible for SMC according to analysis of routine cases and of rainfall data. B Examples of districts with similar (Mali) and different (Kindia) seasonality in cases and rainfall. C Final seasonality classification for SMC eligibility after review of rainfall and cases trends. D 22 Districts where SMC had been already implemented as of 2022, and three additional districts selected for SMC implementation. E Mean predicted incidence and death rates (solid lines) and their 95% PI (shaded area) from 2020 to 2027 in the three districts targeted for expansion of SMC

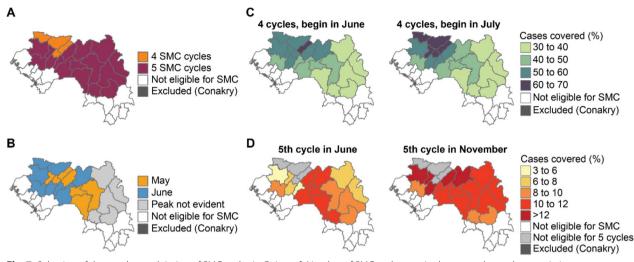
**Table 4**Predicted impact of SMC from 2023 to 2027 in the threeadditional districts targeted for expansion of SMC, comparedwith the base scenario without SMC (Fig. 5B). The 95% predictioninterval is indicated in parentheses

	With SMC
Children U5 years	
Relative reduction in cases	51% (47–56%)
Cases averted (thousands)	450 (350–540)
Relative reduction in deaths	58% (28-80%)
Deaths averted	600 (200–1000)
All ages	
Relative reduction in cases	18% (11–25%)
Cases averted (thousands)	600 (340-880)
Relative reduction in deaths	0.37 (0.10–0.57)
Deaths averted	610 (130–1100)

identifying the 4-month or 5-month window when >60% of cases were reported, which translated into 4 or 5 rounds, respectively. Health districts eligible for SMC

where <60% of cases were reported in a 5-month window were also considered for 5 rounds acknowledging the previously mentioned limitations of the case data to accurately depict seasonality. The median of the confirmed cases in children U5 was also used to determine the month of peak onset per district (Fig. 7B). The impact of adding an additional cycle was evaluated in all health districts targeted for SMC eligible for 5 rounds. The month of peak onset could be detected in the majority of eligible districts as May or June, with the exception of districts in the Kankan region where issues with the quality of routine data made onsets of case peak not detectable with the standard approach.

Considering the predominant month of peak onset and for operational purposes, the most appropriate date for the first SMC cycle in a 4-cycle calendar was identified to be either June or July. In most districts, more cases would be covered with SMC cycles between July and October than between June and September (Fig. 7C). If a fifth SMC cycle were to be implemented in eligible districts, assuming an existing 4-cycle schedule of July



**Fig. 7** Selection of the number and timing of SMC cycles in Guinea. **A** Number of SMC cycles required to cover the peak transmission season, based on the median confirmed cases U5 between 2018 and 2022. **B** Month in which the median number of cases increased by more than 60% compared with the previous month. **C** Proportion of cases covered by four cycles if the first cycle begins in June. **D** Proportion of cases covered by four cycles if the first cycle of SMC if the cycle is implemented in June. **F** Proportion of additional cases covered by the 5th cycle of SMC if the cycle is implemented in June. **F** 

to October, placing the fifth cycle in November would cover more cases than a fifth cycle in June (Fig. 7D). The PNLP decided to have five cycles in all eligible districts, although implementation would be subject to availability of resources.

#### Targeting of the malaria vaccine under vaccine supply constraints

WHO recommends prioritizing the implementation of either of the two recommended malaria vaccines in areas where the intensity of transmission is moderate or high [23]. In Guinea, the majority of health districts met this criterion. At the time when this exercise was conducted, only the RTS,S vaccine was recommended with the availability of very limited supply for Africa (18 million doses). For this purpose, the PNLP followed WHO's Framework for the allocation of limited malaria vaccine supply [23]. Health districts were categorized based on malaria transmission and mortality, where areas with parasite prevalence of at least 20% (Fig. 3C) and mortality of at least 9.5 per 1000 live births (Fig. 3D), or parasite prevalence of at least 40% and mortality of at least 7.5 per 1000 live births, were identified as the those most in need of vaccine (category 1 areas). 26 health districts met the criteria for the highest level of prioritization (Fig. 8). The remaining seven health districts met criteria for second-tier prioritization.

At the time of this analysis, a maximum number of 1 million RTS,S vaccine doses was available per country (to cover approximately 250 thousand children), which

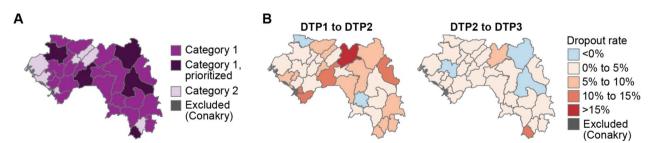


Fig. 8 Prioritization of districts due to limited availability of the RTS,S vaccine in 2022. A Categorization of Guinea's health districts according to their malaria prevalence and all-cause mortality following the WHO framework on allocation of limited doses of malaria vaccine, and final districts chosen for vaccine implementation under supply constraints. B Dropout rates of DTP1 to DTP2 and DTP2 to DTP3 used to inform targeting of malaria vaccine

needed to be prioritized within the 26 highest-need districts. The PNLP initially intended to use DTP1 to DTP2 and DTP2 to DTP3 drop-out rates to identify the districts with higher-functioning immunization programmes that would be more likely to achieve successful implementation of RTS,S with high coverage. However, due to the limited quality of the immunization data the PNLP decided to choose 5 districts from category 1 for pilot malaria vaccine implementation (Gaoual, Mamou, Kankan, Siguiri and Yomou). The districts were selected due to their regional spread, high incidence, and their status as PNLP programmatic priority districts. This plan was since revised to replace RTS,S with R21 after the recommendation of R21, targeting the same districts, with national scale-up to occur subsequently.

# Targeting and potential impact of PMC with or without RTS,S

PMC was targeted to districts with a prevalence of at least 10% and not targeted with SMC. Thirteen districts in the regions of N'zérékoré, Kindia, and Boké met these criteria (Fig. 9A). The PNLP identified the districts in the N'zérékoré region for priority implementation (Fig. 9B) as these were the eligible districts with high or very high transmission according to the risk stratification (Fig. 3G).

The potential impact of PMC if it were to be implemented in all eligible districts (Fig. 9A), with or without co-implementation of RTS,S in Yomou District (one of the selected districts for the piloting of the malaria vaccine), was predicted using mathematical modelling. Other districts prioritized for RTS,S were not also eligible for PMC. Since PMC doses are given at ten weeks, fourteen weeks, and nine months of age, and the protective effect of PMC lasts for around one month, PMC was not expected to have any effect in children older than one year of age, and thus impact was assessed for cases averted in children under the age of 1 year (U1) in addition to children U5. The impact on deaths was not predicted due to high uncertainty in events with very small numbers.

The model predicted that implementation of PMC from 2023 to 2027 could avert 10% of cases (95% PI: 5-14%) in children U1 in the 13 districts eligible for PMC (Fig. 9, Table 5). In Yomou district, addition of RTS,S averted an additional 7% of cases (95% PI: -9% to 21%) in children U1. PMC averted no additional cases in children above the age of one year, as expected due to the duration of protection of PMC. Addition of RTS,S averted 6% of cases (95% PI: -11% to 20%) in children between 1 and 5 years of age. The model stochasticity was larger than the impact of RTS,S, resulting in negative values in the predictive interval.

#### Discussion

Application of the SNT process resulted in improved understanding of recent dynamics in malaria transmission, identification of specific areas to prioritize and to maximize the impact of interventions, and an

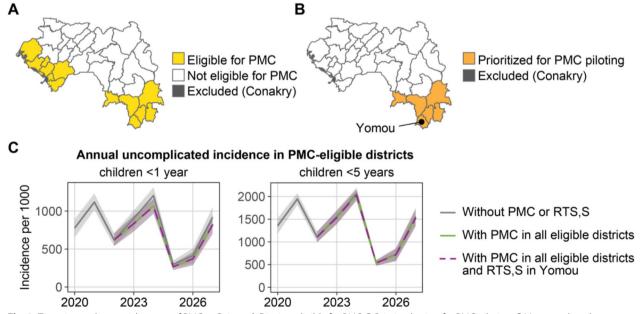


Fig. 9 Targeting and potential impact of PMC in Guinea. A Districts eligible for PMC. B Priority districts for PMC piloting. C Mean predicted incidence (solid line) and their 95% PI (shaded area) from 2020 to 2027 for all 13 districts eligible for PMC. In addition to PMC, malaria vaccine RTS,S may be available to the Yomou district and is factored into the prediction here

	All 13 districts	Yomou district	
	With PMC only	With PMC only	With PMC and RTS,S
Children U1			
Relative reduction in cases	10% (5 to 14%)	11% (- 4 to 27%)	18% (3 to 31%)
Cases averted (thousands)	53 (27 to 78)	2.6 (- 0.8 to 6.4)	4.0 (0.5 to 7.6)
Children U5			
Relative reduction in cases	1% (- 2 to 4%)	1% (– 14 to 13%)	7% (– 7 to 18%)
Cases averted (thousands)	62 (- 90 to 202)	1.8 (- 22.8 to 21.3)	12.7 (- 11.6 to 32.4)

Table 5 Predicted impact of PMC with or without RTS,S from 2023 to 2027, in all 13 districts eligible for PMC and in Yomou district only, compared to the base scenario (Fig. 4B) with no PMC and no RTS,S. The 95% prediction interval is indicated in parentheses

assessment of the availability and quality of data needed for clear decision-making. SNT provided solid support to justify programmatic decisions in Guinea's Global Fund funding request by demonstrating the expected significant impact of the expanded intervention package on malaria transmission. Guinea's Global Fund funding request was ultimately successful at securing funding for IG2 nets in all 33 districts, expanding SMC to three new districts, and implementing five cycles of SMC in all districts eligible for five cycles.

The use of data for sound decision-making relies on the availability of sufficient data of appropriate quality. A surveillance assessment would allow the PNLP to systematically assess the quality of its surveillance system and identify areas for improvement [14], but insights were also gained during the SNT process. Missingness of key malaria indicators in the routine data was low, testing and reporting rates were high, and there was good internal consistency between indicators. This level of quality demonstrates the long-standing efforts made by the PNLP and its partners to avoid stock-outs for diagnostic and treatment products and to improve data collection and reporting via the DHIS2 platform. However, other indicators such as data on inpatient admissions or deaths were not well-reported, and alternative proxies such as modelled all-cause under-five mortality rates had to be used. Furthermore, a substantial effort was required during the SNT process to assemble all routine and non-routine data. The PNLP would benefit from having a more structured malaria data repository containing all relevant data for decision-making.

In Guinea, most facilities in the DHIS2 are in the public sector, and these facilities report more information than those in the private sector. A similar pattern is observed in other countries [24, 25]. In some health districts in Guinea, private health facilities may send their data through public health facilities located in their geographical area. Non-reporting or missing data

from those private facilities would, therefore, not be visible during data quality assessments.

Adjusting for care-seeking behaviour outside of the public health sector substantially increased the estimated incidence. Nevertheless, accurately estimating care-seeking remains difficult: fine-scale spatio-temporal data is lacking, including at the district level, and estimates for care-seeking in individuals above five years of age are often unavailable [26]. Furthermore, the relationship between care-seeking reported in surveys and actual patterns of care-seeking for symptomatic malaria remains unclear. In Guinea, the PNLP did not find the incidence estimated after adjusting for care-seeking to reflect actual epidemiological conditions, and instead chose to use incidence adjusted only for testing and reporting rates to identify relative differences in burden between districts while acknowledging that this incidence was still likely to be an underestimate of the actual incidence. Better data and refined estimation approaches are needed to understand the true burden of malaria.

Until measurements of malaria prevalence or mortality are sufficiently frequent and available at relevant spatial resolution in Guinea, spatio-temporal modelled estimates will be crucial for decision-making. To improve acceptability and utility from the PNLP, estimating these metrics should shift from a global to a local approach, incorporating locally collected subnational routine and non-routine information and adjusting methodologies to fit the country's context through strong partnerships that focus on developing local capacity.

The SNT process highlighted the limited availability and quality of data on determinants of malaria transmission. Under constrained resources, vector control with dual-action insecticide ITNs or with IRS should be targeted at areas where vectors are resistant to lowercost pyrethroid insecticides [22]. Ideally, entomological surveillance would inform the SNT team of the degree of insecticide resistance in each district. However, in practice entomological data is collected in a few sentinel sites across Guinea, so while it is broadly informative of regional levels of insecticide resistance, it is not powered to inform at the district level. Good data on aspects of intervention implementation that could inform effectiveness at the routine and community level were also lacking. This limited the PNLP's ability to interpret epidemiological data, assess gaps, and make accurate predictions of local intervention impact.

Impact predictions from the mathematical model have several key limitations, in addition to the limitations due to data inputs to the model already described above. Model results are limited by uncertainty around intervention coverages and effect sizes, especially at the district level; uncertainty of vector behaviour and susceptibility to insecticides; uncertainty in the true burden of malaria in each district; and uncertainty in case fatality rates of treated and untreated malaria. Due to the time constraints imposed by funding request deadlines, it was not possible to fully explore the sensitivity of model predictions to uncertainties in key parameters. However, future work would benefit from a quantitative understanding of impact of uncertainties, particularly to inform future data collection, via sensitivity analysis or more complex and comprehensive methods to estimate joint uncertainties.

Consensus-building and constant review were essential elements of the SNT process, and consequently, SNT can only be automized to a certain extent. For example, Guinea's analysis of seasonality found that seasonality of rainfall and cases were not always the same, and districts with similar rainfall could have very different case patterns. The SNT team reviewed and discussed the seasonality by rainfall and cases on a district-by-district basis, bringing in local knowledge to guide the interpretation of the analysis results, and together made a determination for each district. A full automation of the SNT process would not have resulted in such a nuanced and informed determination of SMC eligibility, nor would it have given the PNLP and its partners a greater understanding of the utility and limitations of key data sources. Similarly, automation of the epidemiological stratification without adaptation after PNLP review would have misclassified the district of N'zérékoré as low risk.

By including partners at every step of the SNT process, under the leadership of the PNLP, there was team ownership of the results by the entire SNT team. The SNT team reviewed and discussed the data at every step, and decisions for whether and how to include data in the decision-making process were undertaken together, with the ultimate decisions made by the PNLP. Issues with quality or availability of certain data were identified for follow-up and future data collection. Bringing together all partners from the beginning enabled the SNT team to discuss and resolve sensitive issues together, and partners acquired a greater understanding of why decisions were made, resulting in greater buy-in for the decisions.

Capacity does not yet exist within the PNLP to conduct every step of SNT without external support. Strong leadership, capacity to translate evidence and analytical outputs into policy, and thorough understanding of data availability and quality were already present at the initiation of this SNT exercise. For data management and analysis, the PNLP relied on external partners. The PNLP grew its capacity in data management and simpler analysis such as the epidemiological stratification during this SNT, as well as during workshops conducted by WHO after the SNT, and plans to be able to conduct these steps internally in future rounds of SNT. However, other analytical steps such as geostatistical modelling and mathematical modelling may take much longer before capacity is present within the PNLP.

Prior to the use of SNT, Guinea's planning process used a homogeneous approach that applied interventions evenly in all districts regardless of suitability. The SNT process enabled the PNLP to, for the first time, use a targeted approach in their intervention planning and focus resources where they could have the most impact. The success of the process spurred the PNLP to adapt the principles of SNT to new questions after the submission of their funding request, including a microstratification of Conakry, a retrospective analysis of trends in incidence, and an exercise to reprioritize bed nets under further resource constraints.

#### Conclusions

In the face of continued high burden of malaria, Guinea's national malaria programme adopted an innovative datainformed process, guided by local expertise and engaging multiple partners, to prioritize malaria interventions at the district level and successfully obtain funding for expanded intervention plans. This analytic approach was unprecedented in Guinea and allowed the PNLP to effectively decide how to prioritize their limited resources.

#### Abbreviations

- ACT Artemisinin-based combination therapy
- DHIS2 District Health Information System 2
- DHS Demographic and Health Survey
- DTP Diphtheria-Tetanus-Pertussis vaccine
- EPI Expanded Programme on Immunization
- IG2 Interceptor G2
- IHME Institute for Health Metrics and Evaluation
- IPTp Intermittent Preventive Treatment in pregnancy
- IRS Indoor residual spraying
- ITN Insecticide-treated net
- MAP Malaria Atlas Project
- MIS Malaria Indicator Survey
- MOP Malaria Operational Plan
- PfPR Plasmodium falciparum Parasite Rate
- PMC Perennial Malaria Chemoprevention

PMI PNLP	US President's Malaria Initiative Programme National de Lutte contre le Paludisme
RDT	Rapid Diagnostic Test
SMC	Seasonal Malaria Chemoprevention
SNT	Subnational Tailoring of interventions
SP-AQ	Sulfadoxine-Pyrimethamine with Amodiaquine
U1	Under-1 (year of age)
U5	Under-5 (years of age)
U5MR	All-cause under-5 Mortality Rate
WHO	World Health Organization

### **Supplementary Information**

The online version contains supplementary material available at https://doi. org/10.1186/s12936-025-05302-z.

Supplementary Material 1

Supplementary Material 2

#### Acknowledgements

The authors would like to thank PMI Guinea, country teams of the Global Fund and Catholic Relief Services, and RTI/Notre Sante for their support of this work. The authors would also like to thank Northwestern University Research Computing for their assistance with high-performance computing needs. All simulations were run on the Quest computing platform at Northwestern University.

#### Disclaimer

Beatriz Galatas is a staff member of the World Health Organization. This author alone is responsible for the views expressed in this article and does not necessarily represent the decisions, policy or views of the World Health Organization.

#### Author contributions

Study conceptualization: AD, BG, AC. Data assembly and management: OOD, AD, ND, MD, EMD. Data review: OOD, AD, ND, MD, EMD, BG, AC. Data analysis: OOD, JG, BG, AC. Modeling: OOD, KBT, MR, TS, JG. Interpretation: all authors. First draft manuscript: OOD, AD, KBT, MR, TS, JG, BG. All authors have read and approved the final manuscript.

#### Funding

OOD,  $\overline{\text{KBT}}$ , MR, and JG were supported by a grant from the Bill & Melinda Gates Foundation (INV-048410).

#### Availability of data and materials

No datasets were generated or analysed during the current study.

#### Declarations

#### Ethics approval and consent to participate

This study was determined by the Northwestern University Institutional Review Board to not constitute human subjects research (STU00220182).

#### **Consent for publication**

Not applicable.

#### **Competing interests**

The authors declare no competing interests.

#### Author details

<sup>1</sup>Department of Preventive Medicine and Institute for Global Health, Northwestern University Feinberg School of Medicine, Chicago, USA. <sup>2</sup>Programme National de Lutte contre le Paludisme, Conakry, Guinea. <sup>3</sup>Malaria Atlas Project, The Kids Research Institute Australia, Perth, Australia. <sup>4</sup>Global Malaria Programme, World Health Organization, Geneva, Switzerland.

# Received: 18 June 2024 Accepted: 20 February 2025 Published online: 25 February 2025

#### References

- Division Information Sanitaire et Recherche. Annuaire Statistique Sanitaire 2021. Conakry, Guinée, 2022.
- 2. Programme National de Lutte contre le Paludisme de Guinée. Bulletin annuel de surveillance du paludisme. Conakry, Guinée, 2023.
- WHO. World malaria report 2023. Geneva: World Health Organization; 2024.
- WHO. Guiding principles for prioritizing malaria interventions in resourceconstrained country contexts to achieve maximum impact. Geneva: World Health Organization; 2024.
- MAGIC Evidence Ecosystem Foundation. Une Plateforme de création et de publication numérique pour l'écosystème des preuves. https://app. magicapp.org/#/guidelines. Accessed 6 June 2024.
- 6. Institut National de la Statistique. Recensement Général de la Population et de l'Habitat 2014. Conakry, Guinée, 2014.
- Demographic and Health Surveys (DHS) Program. https://dhsprogram. com Accessed 6 June 2024.
- Hersbach H, Bell B, Berrisford P, Biavati G, Horányi A, Muñoz Sabater J, et al. ERA5 hourly data on single levels from 1940 to present. Copernicus Climate Change Service (C3S) Climate Data Store (CDS). European Centre for Medium-Range Weather Forecasts. 2023;147.
- PMI VectorLink Project Guinea. Guinea entomological monitoring annual report, 2022. Rockville, MD, PMI VectorLink Project, Abt Associates Inc., USA; 2022.
- 10. WHO. Malaria threats map: tracking the spread of biological threats to malaria 2023. Geneva: World Health Organization; 2023.
- Institut National de la Statistique et ICF International. Enquête Démographique et de Santé et à Indicateurs Multiples (EDS-MICS). Ministère de la santé publique et Institut national de la statistique, Conakry, Guinée, 2012.
- 12. Institut National de la Statistique et ICF International. Enquête sur les indicateurs du paludisme et de l'anémie en Guinée (EIPAG). Ministère de la santé publique et Institut national de la statistique, Conakry, Guinée, 2021.
- Institut National de la Statistique et ICF International. Enquête Démographique et de Santé (EDS). Ministère de la santé publique et Institut national de la statistique, Conakry, Guinée, 2018.
- 14. WHO. Malaria surveillance, monitoring, and evaluation: a reference manual. Geneva: World Health Organization; 2018.
- GBD 2019 Diseases and Injuries Collaborators. Global burden of 369 diseases and injuries in 204 countries and territories, 1990–2019: a systematic analysis for the Global Burden of Disease Study 2019. Lancet. 2020;396:1204–22.
- WHO. Consolidated guidelines for malaria. Geneva: World Health Organization; 2022.
- 17. Funk C, Peterson P, Landsfeld M, Pedreros D, Verdin J, Shukla S, et al. The climate hazards infrared precipitation with stations—a new environmental record for monitoring extremes. Sci Data. 2015;2: 150066.
- Bershteyn A, Gerardin J, Bridenbecker D, Lorton CW, Bloedow J, Baker RS, et al. Implementation and applications of EMOD, an individual-based multi-disease modeling platform. Pathog Dis. 2018;76:fty059.
- Ozodiegwu ID, Ambrose M, Galatas B, Runge M, Nandi A, Okuneye K, et al. Application of mathematical modelling to inform national malaria intervention planning in Nigeria. Malar J. 2023;22:137.
- WHO. A Framework for malaria elimination. Geneva: World Health Organization; 2017.
- Camara A, Guilavogui T, Keita K, Dioubaté M, Barry Y, Camara D, et al. Rapid epidemiological and entomological survey for validation of reported indicators and characterization of local malaria transmission in Guinea, 2017. Am J Trop Med Hyg. 2018;99:1134–44.
- WHO. Guidelines for malaria. Geneva, World Health Organization. https:// app.magicapp.org/#/guideline/LwRMXj/section/LpOA4j. Accessed 6 June 2024.
- 23. WHO. Framework for the allocation of limited malaria vaccine supply. Geneva: World Health Organization; 2022.
- Githinji S, Oyando R, Malinga J, Ejersa W, Soti D, Rono J, et al. Completeness of malaria indicator data reporting via the District Health Information Software 2 in Kenya, 2011–2015. Malar J. 2017;16:344.
- Muhoza P, Tine R, Faye A, Gaye I, Zeger SL, Diaw A, et al. A data quality assessment of the first four years of malaria reporting in the Senegal DHIS2, 2014–2017. BMC Health Serv Res. 2022;22:18.

26. Ozodiegwu ID, Ambrose M, Battle KE, Bever C, Diallo O, Galatas B, et al. Beyond national indicators: adapting the Demographic and Health Surveys' sampling strategies and questions to better inform subnational malaria intervention policy. Malar J. 2021;20:122.

## **Publisher's Note**

Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.