MODELING THE RELATIONSHIP BETWEEN THE POPULATION PREVALENCE OF PLASMODIUM FALCIPARUM MALARIA AND ANEMIA

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Abstract. More than half of all young children and pregnant women are affected by anemia. Although its etiology is multi-factorial, malaria is likely to be a major contributor to chronic anemia in endemic areas. Recent reviews have examined the effect of community-based malaria control interventions on anemia. We analyze how the prevalence of anemia depends on that of Plasmodium falciparum malaria by developing models of the excess risk of anemia caused by malaria at a population level in 24 villages in northeastern Tanzania. In that setting, we estimated that the prevalence of a hemoglobin level < 8 g/dL attributable to malaria was 4.6% in infants, 4.1% in children one year of age, 2.7% in children two years of age, and 3.3% in women of childbearing age. Successful validation of our models in other malaria-endemic settings would enable their use for predicting the impact of malaria control interventions on anemia, and for long-term monitoring and surveillance of malaria.

INTRODUCTION

The etiology of anemia is usually multi-factorial. In the developing world, a host of factors play a role, most notably micronutrient deficiencies (especially iron deficiency), parasitic infections (particularly malaria and intestinal nematodes such as hookworms), infection with human immunodeficiency virus (HIV), chronic inflammatory disorders, and hemoglobinopathies.1–6 Recent literature reviews have established the causal relationship between malaria and anemia. Community-based malaria control interventions, including antimalarial chemoprophylaxis, use of insecticide-treated nets (ITNs), and indoor residual spraying, consistently improved anemia-related outcomes in young children.7,8 Therefore, Roll Back Malaria is considering the use of anemia as an additional indicator for measuring the impact of the introduction and/or scaling up of malaria control interventions.8 These include ITNs,9,10 routine intermittent presumptive treatment of infants11 and a malaria vaccine12 distributed alongside the expanded program on immunization. However, there remain gaps in our knowledge about the association between malaria and anemia. Quantification of this association, and its relationship with age, would enable an enhanced understanding of the impact of malaria control interventions on the burden of anemia.

The purpose of this report is to examine how the prevalence of chronic anemia depends on that of Plasmodium falciparum malaria. Parasite and anemia prevalence data, measured in community-based surveys in northeastern Tanzania,13 were summarized by age group and village. We fitted a statistical model to predict the excess risk of anemia for each sub-group as a function of the prevalence of Plasmodium falciparum in that group. We consider the implications for estimates of the burden of chronic anemia attributable to Plasmodium falciparum malaria, and propose that our model, after successful validation in other settings, can be used for predicting the impact of malaria control interventions on anemia.

MATERIALS AND METHODS

Study area and population surveyed. Cross-sectional surveys were carried out in October–November 2001 and April–June 2002 in 24 villages representing 6 altitudinal transects across northeastern Tanzania. In children less than 15 years of age, the prevalence of stunting (height-for-age Z-score < −2) was 40.3% and the prevalence of wasting (weight-for-age Z-score < −2) was 6.8%. The prevalence of alpha thalassemia ranged between 4% and 59%, and the prevalence of sickle cell anemia ranged between 0% and 16%. Both of these hemoglobinopathies appeared to be correlated with altitude and the prevalence of malaria (Drakeley CJ, unpublished data), and alpha thalassemia heterozygotes appeared to have lower hemoglobin (Hb) levels (Ennevald A, unpublished data). The prevalence of a deficiency in glucose-6-phosphate dehydrogenase was approximately 7%.14 The prevalence of HIV is thought to be 6–7% (Mbatia R, unpublished data). Details of the population surveyed and the field and laboratory techniques used have been presented elsewhere.13,15 Briefly, in each village an age-stratified sample of approximately 250 persons (< 5 years, 5–14 years, and 15–45 years of age) was recruited. Blood samples were collected from each individual and analyzed for the presence and density of malaria parasites using a standard, quality-controlled light microscopy technique. Hemoglobin levels were measured by standard hemoglobinometry (HemoCue, Ängleholm, Sweden).

Statistical methods. The parasitemia and anemia data were summarized within each survey and village by age-group, using one-year age groups up to four years and five-year age groups thereafter, leading to a total of 606 sub-groups. For the main analysis, we excluded the data of women 15–45 years of age, among whom anemia is specifically associated with menstruation and pregnancy, and carried out a separate analysis for these women. The prevalence of anemia, as a function of parasite prevalence, density, age group, and village, was analyzed using random effects logistic regression, with the full model of the form

\[
\logit(p_a(a,i,s)) = \beta_0 + f(a)p_s(a,i,s) + \beta_1I(s) + \gamma_i.
\]

where \(p_a(a,i,s)\) is the probability of being anemic in group with age-midpoint \(a\), village \(i\), and survey \(s\), \(p_s(a,i,s)\) is the...
corresponding parasite prevalence. $\beta_{a0}$ is the intercept, $\beta$, the regression coefficient for the survey effect where $I(f(s))$ is an indicator variable taking the value 0 for the first survey, and 1 for the second, and $\gamma$ is a random effect corresponding to village $i$. The observed number of anemic individuals, i.e., $r_a(a,i,s)$, of $n_a(a,i,s)$ individuals tested, was then assumed binomially distributed

$$r_a(a,i,s) \sim \text{Binomial}(n_a(a,i,s), p_a(a,i,s))$$ (2)

where $p_a(a,i,s)$, the observed number of parasitemic individuals (of $n_a(a,i,s)$ tested) was also assumed binomially distributed, i.e.,

$$r_p(a,i,s) \sim \text{Binomial}(n_p(a,i,s), p_p(a,i,s))$$ (3)

Functions $f(\cdot)$ of the forms:

$$f(a, p_a(a,i,s)) = \beta_{a0} \ln(a) + \frac{\beta_p p_p(a,i,s)}{p^* + p_p(a,i,s)}$$ (4)

and

$$f(a, p_p(a,i,s)) = \frac{\beta_{a0} a^*}{a^* + a} + \frac{\beta_p p_p(a,i,s)}{p^* + p_p(a,i,s)}$$ (5)

were evaluated, as were models including age-prevalence interaction terms. The use of sigmoidal functions on the logit scale enabled us to constrain the prevalence of anemia to vary monotonically with age and with parasite prevalence, while allowing for more possible shapes of curves than would a linear logistic model. The parameter $a^*$ corresponds to the age at which 50% of the maximal age effect is observed, and $p^*$ to the parasite prevalence corresponding to 50% of the maximal parasitemia effect. In addition, models were assessed that included terms in the average parasite density for each sub-group.

The models were fitted using a Bayesian Markov chain Monte Carlo algorithm in the software package Winbugs version 1.4.16 with the $\gamma$ assumed normally distributed (centered on 0), and appropriate imprecise priors were assigned for the other parameters. Model fit was assessed using the deviance information criterion (DIC), with lower DICs indicating improved fits.17

The full model contained all the terms we considered for inclusion. Because we required a model that could be applied to general situations, not just to the two surveys that we carried out, we also considered a model without terms in the survey (the reference model), i.e.,

$$\logit(p_A(a,i,s)) = \beta_{a0} + f(a, p_A(a,i,s)) + \gamma_i$$ (6)

This was compared with simpler models from which additional terms were removed if they did not improve the fit.

The excess risk of anemia, $R_A(a,i,s)$, is estimated by the difference between the actual prediction and the prediction that would be made for the same age group, survey, and village at zero prevalence, i.e., in the case of the reference model

$$R_A(a,i,s) = p_A(a,i,s) - \frac{1}{1 + \exp(-\beta_{a0} + f(a,0) + \gamma_i)}$$ (7)

We compute point and interval estimates for $R_A(a,i,s)$ by sampling from the posterior distribution implied by equation 7.

We carried out a separate analysis of anemia in women of childbearing age (i.e., 15–45 years) based on the assumption that pregnancy-associated anemia and malaria account for the higher prevalence in women than in men in this age group. Since there is little age-dependence in the anemia or parasitemia rates in adults, this analysis considered how the anemia prevalence in women of childbearing age compares with that in men, and how this is related to the overall parasite prevalence in the group defined at the level of survey within village. This was estimated using an additive model, i.e., we assume that

$$r_f(i,s) \sim \text{Binomial}(n_f(i,s), p_f(i,s))$$ (8)

where $r_f(i, s)$ is the number of anemic women among $n_f(i,s)$ sampled in village $i$, survey $s$ and

$$p_f(i,s) = p_m(i,s) + (1 - p_m(i,s))\left(p_m(i,s) + (1 - p_m(i,s))\frac{\beta_{mf} p_f(i,s)}{1 + \beta_{mf} p_f(i,s)}\right)$$ (9)

where $p_m(i,s)$ is the corresponding prevalence in men; $p_m(i,s)$ is the excess risk of anemia in women in the absence of $P. falciparum$ (we assume this is pregnancy-related, but not malaria-associated); $p_f(i,s)$ is the overall prevalence of $P. falciparum$ in adults in village $i$, survey $s$; and $\beta_{mf}$ is then a coefficient quantifying the strength of the relationship between parasite prevalence and malaria attributable excess of anemia in women of childbearing age.

RESULTS

The overall prevalence of anemia (Hb levels $< 8$ g/dL and $< 11$ g/dL) was strongly related both to the age of the host and to the parasite prevalence (Figure 1). In both cases the anemia and parasite prevalence were slightly higher at the second survey during the long dry season than at the first survey (Table 1).13

Terms in age and parasite prevalence were both needed to give adequate predictions of the levels of anemia. The age effect on the prevalence of anemia could not be accounted for by the parasite prevalence alone, even when higher order terms in $p_p(a,i,s)$ were introduced into the model, nor could the prevalence effect be explained by age alone. Introducing a term for parasite density did not result in improved model fit, as measured by the DIC.

Testing of different forms of age dependence indicated that the main effect of age was best modeled with a sigmoidal function (equation 5). However, there was an interaction between the effect of age and parasitemia, with the best fit achieved by using a logarithmic transformation of age (equation 10). Models including interactions with sigmoidal functions of age were poorly identifiable and were therefore rejected in favor of the reference model (ii in Table 2) (equation 10).

$$\logit(p_A(a,i,s)) = \beta_{a0} + \frac{\beta_{a1} a^*}{a^* + a} + \frac{\beta_p p_p(a,i,s)}{p^* + p_p(a,i,s)} + \beta_{pf} p_f(a,i,s) \log(a) + \gamma_i$$ (10)

The fit of the model was only improved slightly by including terms in the survey (compare models i and ii in Table 2), indicating that the between-survey variation in anemia was largely explained by the variation in parasite prevalence between survey periods.
We assumed the same relationships of anemia with age and parasite prevalence in all villages, but there was substantial variation between villages in anemia levels even after allowance for the differences in prevalence (compare models ii and iii in Table 2). The best fitting models therefore also included the random effect terms allowing for differences between villages in anemia prevalence. For the model that we plan to generalize to other settings, we use estimates based on the reference model ii but without terms in the survey and with $\gamma_i = 0$ (i.e., no village-level random effect). The parameter estimates for this model are given in Table 3.

We used model ii to estimate fitted values of the moderate anemia prevalence ($\text{Hb level} < 8 \text{ g/dL}$) for each of the 606 sub-groups (Figure 1a and b). Although the age pattern of these predictions showed the expected decrease with age, the predicted anemia prevalence did not increase smoothly with parasite prevalence measured at the level of the group. This was because of the low number of villages with high parasite prevalence.

The fitted values of anemia prevalence ($\text{Hb levels} < 8 \text{ g/dL}$ and $< 11 \text{ g/dL}$) estimated for a hypothetical village with $\gamma_i = 0$ showed the expected smooth patterns of decrease with age (Figure 2) and increase with parasitemia. A fraction of this anemia will be due to causes other than malaria. For this reason, we computed the excess risk associated with parasitemia by subtracting the estimated levels in the absence of

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**TABLE 1**  
Unadjusted estimates of anemia and parasitemia*  

<table>
<thead>
<tr>
<th>Group (age, years)</th>
<th>Survey</th>
<th>No. of individuals</th>
<th>Prevalence of Hb $&lt; 11$ g/dL</th>
<th>Prevalence of Hb $&lt; 8$ g/dL</th>
<th>Prevalence of patent <em>Plasmodium falciparum</em> parasitemia</th>
</tr>
</thead>
<tbody>
<tr>
<td>Children 0–14</td>
<td>1</td>
<td>3,710</td>
<td>34.6%</td>
<td>2.6%</td>
<td>16.9%</td>
</tr>
<tr>
<td>(0–14)</td>
<td>2</td>
<td>3,627</td>
<td>35.0%</td>
<td>2.7%</td>
<td>23.5%</td>
</tr>
<tr>
<td>Men</td>
<td>1</td>
<td>609</td>
<td>7.7%</td>
<td>3.7%</td>
<td>23.5%</td>
</tr>
<tr>
<td>(15–45)</td>
<td>2</td>
<td>589</td>
<td>9.7%</td>
<td>4.0%</td>
<td>11.9%</td>
</tr>
<tr>
<td>Women</td>
<td>1</td>
<td>1,454</td>
<td>24.1%</td>
<td>4.7%</td>
<td>11.5%</td>
</tr>
<tr>
<td>(15–45)</td>
<td>2</td>
<td>1,443</td>
<td>25.8%</td>
<td>3.6%</td>
<td>17.3%</td>
</tr>
</tbody>
</table>

*Hb = hemoglobin.
parasites from our estimates of anemia prevalence (Figure 3). The major impact of malaria on anemia in children has previously been shown to occur in children less than 2–3 years of age, and a similar age-related pattern was seen here. At the average levels of endemicity in the study area we estimated the prevalence of an Hb level < 8 g/dL attributable to malaria to be 4.6% in the first year of life, 4.1% and 2.7% in children 1 and 2 years of age, respectively, which decreased to 0.8% in the 5–9-year-old age group and to approximately 0.1% in adult males.

We fitted an equivalent model to that of equation 10 to predict the prevalence of an Hb level < 11 g/dL. This showed an excellent fit to the age-prevalence relationship (Figure 1c) and to the relationship with prevalence of parasites (Figure 1d).

Unlike an Hb level < 8 g/dL, a level < 11 g/dL is frequent even at very low levels of parasite prevalence. Thus, the proportion attributable to malaria is much lower than that for an Hb level < 8 g/dL. This is especially the case in the youngest children among whom the estimates of the excess risk of an Hb level < 11 g/dL did not increase monotonically with age at any given prevalence of parasitemia. This is because the estimated prevalence of an Hb level < 11 g/dL in the absence of parasitemia is so high in the youngest children that there is little scope for an increase. The excess risk thus reaches a maximum in older children, when the baseline parasitemia has already decreased (Figure 3c).

The prevalence of Hb levels < 8 g/dL and < 11 g/dL in adult women (15–45 years of age) was much higher than in age-matched men, and showed strong relationships with parasite

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**Table 2**

Comparisons of the fit of the model for prevalence of anemia

<table>
<thead>
<tr>
<th>Model</th>
<th>Linear predictor</th>
<th>DIC*</th>
</tr>
</thead>
<tbody>
<tr>
<td>i</td>
<td>$\beta_0 + \frac{\beta_a a}{a^* + a} + \frac{\beta_p p_i(a, i, s)}{p^* + p_i(a, i, s)} + \beta_p p_i(a, i, s) \ln(a) + \beta_p p_i(a, i, s) + \gamma_i$</td>
<td>768.8</td>
</tr>
<tr>
<td>ii</td>
<td>$\beta_0 + \frac{\beta_a a}{a^* + a} + \frac{\beta_p p_i(a, i, s)}{p^* + p_i(a, i, s)} + \beta_p p_i(a, i, s) \ln(a)$</td>
<td>769.5</td>
</tr>
<tr>
<td>iii</td>
<td>$\beta_i + \frac{\beta_a a}{a^* + a} + \frac{\beta_p p_i(a, i, s)}{p^* + p_i(a, i, s)} + \beta_p p_i(a, i, s) \ln(a)$</td>
<td>796.1</td>
</tr>
<tr>
<td>iv</td>
<td>$\beta_i + \frac{\beta_a a}{a^* + a} + \frac{\beta_p p_i(a, i, s)}{p^* + p_i(a, i, s)} + \gamma_i$</td>
<td>771.9</td>
</tr>
<tr>
<td>v</td>
<td>$\beta_i + \frac{\beta_a a}{a^* + a} + \frac{\beta_p p_i(a, i, s)}{p^* + p_i(a, i, s)} + \ln(a)$</td>
<td>778.4</td>
</tr>
</tbody>
</table>

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**Figure 2**

Model predictions of the prevalence of hemoglobin (Hb) levels < 8 g/dL and < 11 g/dL. **a**, Prevalence of an Hb level < 8 g/dL by age. **b**, Prevalence of an Hb level < 11 g/dL by age. The predictions correspond to the fitted values from model ii equation 10 at $\gamma_i = 0$. The lower lines correspond to the lowest parasite prevalence (10%). The remaining lines correspond to predictions at 10% increments in parasitemia.

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**Table 3**

Parameter estimates from the best-fitting model*

<table>
<thead>
<tr>
<th>Parameter description</th>
<th>Estimates from model for Hb &lt; 8 g/dL</th>
<th>Estimates from model for Hb &lt; 11 g/dL</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>2.5%</td>
</tr>
<tr>
<td>$\beta_0$</td>
<td>-4.62</td>
<td>-5.19</td>
</tr>
<tr>
<td>$\beta_p$</td>
<td>8.08</td>
<td>1.51</td>
</tr>
<tr>
<td>$p^*$</td>
<td>6.52</td>
<td>3.74</td>
</tr>
<tr>
<td>$\beta_a$</td>
<td>6.36</td>
<td>5.86</td>
</tr>
<tr>
<td>$a^*$</td>
<td>4.70</td>
<td>3.81</td>
</tr>
<tr>
<td>$\beta_1$</td>
<td>0.50</td>
<td>0.13</td>
</tr>
</tbody>
</table>

* Calculated from the percentiles of >10,000 samples from the posterior distributions, using a model corresponding to equation 10.
prevalence (Figure 4). These relationships were captured by the additive binomial models (equation 9), which showed a good fit to both outcomes (Figure 5). The excess risk of an Hb level < 8 g/dL in adult women compared with men in the absence of *P. falciparum* was estimated to be 1.3% and that of an Hb level < 11 g/dL was estimated to be 6.1% (Table 4). Both of these percentages were estimated to strongly increase with parasite prevalence (Figure 5), so that at the average *P. falciparum* prevalence of 14.4% (Table 1), the difference in risk of an Hb level < 8 g/dL between men and women is 3.3% and of an Hb level < 11 g/dL is as high as 19.1%. Although only women are susceptible to the effects of malaria infection during pregnancy, both pregnancy and menstruation impact anemia independently of any effects of malaria. At the average *P. falciparum* prevalence of 14.4%, 66% of the Hb levels < 8 g/dL in these women can be attributed to malaria exposure, which corresponds to a prevalence of malaria-attributable anemia among women 15–45 years of age of 3.3%.

**DISCUSSION**

The within-host dynamics of malaria anemia are complex, even in the non-immune host. Underlying reasons include malaria infection resulting in destruction of non-parasitized erythrocytes and diserythropoiesis, direct destruction of parasitized erythrocytes, and multiple genetic factors. Thus, anemia secondary to *P. falciparum* infection can result from a long series of events. The parasitologic status of the
host at the time of sampling may be a poor indicator of whether the anemia is mainly due to malaria, rather than nutritional deficiencies, concomitant infections, chronic inflammatory responses, or a combination of etiologic factors.

It is usually appropriate to assess epidemiologic exposures at the level of the individual. Assignment of group-level exposure measurements to individuals can lead to misclassification and to biased estimates of exposure-response relationships. However, the recent history of parasitization may be a better predictor of anemia than the level or presence of concurrent parasitemia.

In a situation where infections are continually turning over in the population, the prevalence at village level represents a reasonable estimate of recent exposure of the individual, and this justifies the use of population level estimates of the burden of parasitism as one approach to predict malaria-associated anemia.

We were not able to predict the age distribution of anemia from either the parasite densities or parasite prevalence data alone. This indicates either age-dependence in other risk factors for anemia, or age-dependence in the vulnerability of the human host to the parasitologic insult as a result of acquired partial immunity, or a combination of these two factors. In malaria-endemic settings of sub-Saharan Africa, age patterns of anemia show a peak between 1 and 1.5 years. Importantly, the effect of malaria parasitemia on anemia has been shown to decrease with age. By including the age-parasitemia interaction term in our model, we hoped to capture this phenomenon. The dependence of our model on the logarithm of the age is potentially inconvenient if it is used to estimate anemia rates for very young babies, but we do not claim validity for the model in the youngest children, since the dynamics of Hb and erythropoiesis in the first few months of life are very different from those later in life.

Estimates of the global burden of *P. falciparum* malaria generally only include direct effects such as clinically relevant malaria fevers, cerebral malaria, severe malaria anemia, and death, with the latter accounting for the bulk of the burden. Inclusion of chronic anemia as an indirect effect of malaria infection is usually avoided because of the complexity of attributing the anemia to malaria. We have attempted to quantify the prevalence of malaria-related chronic anemia (Hb levels < 11 g/dL and < 8 g/dL) by modeling the excess risk of anemia due to malaria at a population level.

### Table 4

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Estimates from model for Hb &lt; 11 g/dL</th>
<th>Estimates from model for Hb &lt; 8 g/dL</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Mean</td>
<td>2.5%</td>
</tr>
<tr>
<td>$p_{90}$</td>
<td>0.061</td>
<td>0.04</td>
</tr>
<tr>
<td>$\beta_m$</td>
<td>1.12</td>
<td>0.86</td>
</tr>
</tbody>
</table>

*Hb = hemoglobin, $p_{90}$ = excess risk of anemia in women (compared to men) in the absence of *Plasmodium falciparum*, $\beta_m$ = increase in risk with prevalence of *P. falciparum*. 
Our estimates suggest that in the epidemiologic setting of northeastern Tanzania where most villages surveyed were mesoendemic for malaria (i.e., parasite prevalence in children 0–4 years of age < 50%), more than 4% of children less than three years of age, and 3.3% of women of childbearing age, who bear the brunt of malaria disease, are affected by malaria-attributable Hb levels < 8 g/dL. The present analyses provide estimates of the prevalence of maternal anemia within the whole population of women of childbearing age without the need to consider what proportion of women are pregnant at any one time, or whether there are differences between primigravidae and multigravidae, although these differences have been documented. The contribution of malarial anemia during pregnancy is generally considered to be very important, resulting in impaired fetal development, premature delivery, low birth weight, maternal and infant anemia, and infant mortality.\textsuperscript{27,28}

The study area was selected specifically to include a wide range of malaria transmission intensities.\textsuperscript{13,15} and therefore provides an average estimate of the proportion of anemia due to malaria in endemic areas. Further investigations in this area support the finding of a high excess risk of anemia due to malaria in very young children.\textsuperscript{29} The models presented here enable us to estimate the burden of malaria-related anemia at any given population prevalence, and can also be used to assess the potential impact on chronic anemia of different malaria intervention efforts.\textsuperscript{7,8}

It is interesting to note that peak anemia prevalence seems to be closely linked to the age of admission for severe anemia in hospitals in the same study area.\textsuperscript{30} Although there is no clear link between moderate anemia and mortality,\textsuperscript{31} there is evidence that anemia can have substantial effects on cognitive and motor development and on growth.\textsuperscript{32,33} It has also been hypothesized that anemia may be associated with impaired immunity and consequently increased susceptibility to infectious disease.\textsuperscript{34} In this way, anemia, as an indirect effect of malaria may contribute to the malaria gap, i.e., the gap between the socioeconomic impact of the disease that can be documented in microeconomic studies, and the massive impact that is implied by macroeconomic studies using cross-country regression analyses that compare malaria-free countries with those where the disease is highly endemic.\textsuperscript{35}

It is therefore essential that estimates of malaria burden and evaluation of malaria interventions include the contribution of malaria to chronic anemia in the community. The models presented here enabled us to attribute a proportion of anemia to malaria in different age groups in northeastern Tanzania. Successful validation of the models in a range of settings across Africa, where the bulk of the malaria burden is currently concentrated, will allow prediction of the impact of different malaria interventions, including the introduction of malaria vaccines, under different transmission intensity scenarios.

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