

# Improving the diagnosis of severe malaria in African children using platelet counts and plasma *Pf*HRP2 concentrations

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Severe falciparum malaria is difficult to diagnose accurately in children in high transmission settings. Using data from 2,649 patients enrolled in four studies of severe illness in three countries (Bangladesh, Kenya, and Uganda), we fitted Bayesian latent class models using two diagnostic biomarkers: the platelet count and the plasma *Pf*HRP2 concentration. In severely ill patients with clinical features consistent with severe malaria, a combined platelet count  $\leq 150,000$  per  $\mu\text{L}$  and a plasma *Pf*HRP2 concentration  $\geq 1,000$  ng/mL had an estimated sensitivity of 74% and specificity of 93% in identifying ‘true’ severe falciparum malaria. Patients with true severe malaria had higher parasite densities, lower hematocrit, lower rates of invasive bacterial disease, and a lower prevalence of both HbAS and HbSS than children misdiagnosed. We estimate one third of African children enrolled in clinical studies of severe malaria in high transmission settings had another cause of severe illness.

## 1 Introduction

2 Severe falciparum malaria is defined clinically as vital organ dysfunction in the presence of  
3 circulating *Plasmodium falciparum* parasites (1). The primary objective of this definition is to  
4 identify severely ill patients rapidly and provide life-saving clinical management (notably par-  
5 enteral artesunate treatment). This prioritizes diagnostic sensitivity over specificity. In areas of  
6 moderate and high *P. falciparum* transmission, many apparently healthy children have malaria  
7 parasitemia if their blood is examined either by light microscopy or a rapid diagnostic test. It  
8 follows that many hospitalized children will also be parasitemic. As the major clinical features  
9 of severe malaria are not specific, it is difficult to differentiate clinically between severe faldi-  
10 parum malaria caused by extensive sequestration of malaria parasites in the microvasculature  
11 (i.e. ‘true’ severe malaria (2)) and other causes of severe febrile illness accompanied by either

12 coincidental asymptomatic parasitemia or uncomplicated malaria (3, 4). We have shown pre-  
13 viously that complete blood counts (platelet counts and total white blood cell counts) provide  
14 substantial discriminating value in distinguishing true severe malaria from other causes of se-  
15 vere illness (5). Approximately one third of a cohort of 2,220 Kenyan children diagnosed with  
16 severe malaria in a moderate transmission area were estimated to have had another cause of  
17 their severe illness (5). The discriminative value of complete blood counts was validated using  
18 genetic polymorphism data in two different ways. First, the distribution of sickle trait (HbAS)  
19 was correlated very strongly with the estimated probability of severe malaria. The prevalence  
20 of HbAS was around five times higher in the subgroup of patients likely to have been mis-  
21 diagnosed compared to the subgroup with a likely correct diagnosis (5). Second, we showed  
22 that genome-wide false discovery rates could be reduced substantially in a case-control whole-  
23 genome association study in which the output probability weights were used in a ‘data-tilting’  
24 framework to adjust for patient mis-classification (5).

25 The diagnostic value of complete blood counts in severe malaria has great operational utility  
26 as blood counts are widely measured in routine practice at low cost. Moderate thrombocytope-  
27 nia is a consistent feature of all human malaria infections (6–8), although the diagnostic utility  
28 of the platelet count has been debated (9, 10). Platelets are activated in malaria infections and  
29 have increased turnover. In severe falciparum malaria there is endothelial activation with re-  
30 lease of platelet aggregating activated high-multimeric von Willebrand Factor from specialized  
31 secretory vesicles in endothelial cells (the Weibel-Palade bodies). Platelets may also contribute  
32 to, and be consumed by, parasitised erythrocyte cytoadherence and autoagglutination (11, 12).  
33 Severe thrombocytopenia is associated with increased mortality in severe malaria (10, 13, 14).  
34 The total white blood cell count is also informative (although much less so than thrombocy-  
35 topenia), with a greater prognostic than diagnostic value (very high or very low white counts in  
36 severe disease are associated with high case fatality ratios) (5, 8). In addition, total white counts

37 vary according to age and ethnicity (15), confounding cross-study assessments.

38 The main pathophysiological process in severe falciparum malaria is the extensive seques-  
39 tration of parasitised erythrocytes in the vascular beds of vital organs (1, 2). These parasites,  
40 which cause potentially lethal pathology, have stopped circulating and are therefore not repre-  
41 sented in the peripheral blood smear. In African children the peripheral parasite density is a  
42 poor indicator of disease severity and a poor diagnostic biomarker (16). The parasite protein  
43 *Plasmodium falciparum* histidine-rich protein-2 (*PfHRP2*), the basis for most rapid diagnostic  
44 tests, is liberated mainly at schizont rupture - and so the amount released is proportional to  
45 the extent of recent schizogony. Plasma *PfHRP2* is a much better discriminant of severe falciparum  
46 malaria than the peripheral blood parasite count (17, 18). An example of how *PfHRP2*  
47 can help interpret clinical trial data comes from the large multi-center AQUAMAT trial which  
48 was a randomized comparison of parenteral artesunate versus parenteral quinine in African chil-  
49 dren clinically diagnosed as having severe falciparum malaria. Overall, artesunate reduced the  
50 mortality by approximately 22% (19). However, there was evidence for treatment effect het-  
51 erogeneity across *PfHRP2* strata. In the subgroup of children in the highest tertile of *PfHRP2*  
52 artesunate reduced the mortality by one third (a similar proportion to that observed earlier in  
53 the SEAQUAMAT randomized comparison conducted in Southeast Asia (20)), whereas there  
54 was no substantial difference in mortality in the subgroup of children in the lowest tertile, i.e.  
55 the group with likely other causes of severe illness (18).

56 Defining optimal cutoff values for diagnostic biomarkers in severe malaria is difficult be-  
57 cause in high-transmission areas there is no ‘gold-standard’ for diagnosis against which to cali-  
58 brate thresholds. In the absence of a gold-standard, latent class models can be used to assess the  
59 sensitivity and specificity of threshold values for diagnostic indices (21). Latent class models  
60 rely typically on having multiple biomarkers measured in the same individuals across different  
61 populations with varying disease prevalences, thus allowing for triangulation (21, 22). We ap-

62 plied Bayesian parametric latent class models to admission platelet counts and plasma *PfHRP2*  
63 concentrations in combination in order to estimate their diagnostic operating characteristics  
64 and to estimate the proportion of misdiagnosed patients in studies of severe malaria. We used  
65 data from four prospective studies of severe malaria or severe illness from Uganda, Kenya and  
66 Bangladesh, reflecting a range of *P. falciparum* transmission intensities from high to low, and  
67 thus a range of false diagnosis rates. The results suggest that the high rates of misdiagnosis  
68 could be reduced substantially by incorporating measurement of platelet counts and plasma  
69 *PfHRP2* concentrations in the diagnostic criteria.

## 70 **Results**

### 71 **Platelet counts and *PfHRP2* concentrations in severe febrile illness**

72 We pooled individual patient data from 2,649 severely ill African children and Asian adults,  
73 for whom platelet counts and measured plasma *PfHRP2* concentrations were available. The  
74 patients were from four separate studies in 3 countries:

- 75 • An observational study of severe falciparum malaria in Bangladesh ( $n = 172$ ; all patients  
76 were clinically diagnosed with severe falciparum malaria).
- 77 • The Ugandan sites of the FEAST trial ( $n = 567$ , a randomized controlled trial of fluid  
78 resuscitation approaches in severe childhood illness not specific to severe malaria) (23).
- 79 • An observational study of cerebral malaria and severe malarial anemia in Kampala, Uganda  
80 ( $n = 492$ ) (24).
- 81 • A large cohort of children diagnosed with severe malaria in Kilifi, Kenya ( $n = 1,418$ )  
82 (25).

83 Malaria transmission intensity is generally low in Bangladesh, moderate in Kilifi and Kampala,  
84 and high around the other Ugandan sites.

85 In total 27 patients (1%: 1 in Bangladesh, 8 in the FEAST sites in Uganda, and 18 from  
86 Kenya) had no detectable *PfHRP2* in the ELISA assay but had a parasite density  $>1,000$  per  
87  $\mu\text{L}$  by microscopy. These could have either been assay errors or parasites with *HRP2/3* gene  
88 deletions. They were removed from the analysis, leaving a total of 2,622 samples. A summary  
89 of the patient characteristics in the analyzed data set is shown in Table 1.

90  $\text{Log}_{10}$ -transformed platelet counts and  $\text{log}_{10}$ -transformed *PfHRP2* concentrations were strongly  
91 inversely correlated in both the studies of severely ill African children ( $\rho = -0.54$ ), and in the  
92 Bangladeshi adults with severe malaria ( $\rho = -0.36$ ). Patients with platelet counts in the normal  
93 range ( $>150,000$  per  $\mu\text{L}$ ) generally had low to non-measurable plasma *PfHRP2* concentrations  
94 (median concentration was 269 ng/mL, IQR: 24 to 1,043), while patients with thrombocytopenia  
95 ( $\leq 150,000$  per  $\mu\text{L}$ ) had a median *PfHRP2* concentration of 3,031 ng/mL (IQR: 1,261 to  
96 6,035).

## 97 **Discriminative value of platelet counts and plasma *PfHRP2* in patients di-** 98 **agnosed with severe malaria**

99 We first assessed the diagnostic value of platelet counts and plasma *PfHRP2* concentrations in  
100 patients who were diagnosed clinically with severe malaria (a total of  $n=2,063$ ; this excluded  
101 patients from the FEAST study which explicitly included non malarial causes of severe febrile  
102 illness). We fitted a two component parametric Bayesian latent class model (with the two la-  
103 tent classes representing ‘severe malaria’ and ‘not severe malaria’) using the  $\text{log}_{10}$  transformed  
104 biomarkers (see Methods for description of the informative priors used and the key assump-  
105 tions). Fig. 1 shows the model estimated sensitivities and specificities for all cutoff values  
106 of the platelet count and the *PfHRP2* concentration. As expected, the platelet count and the

	Kilifi (Kenya)	Kampala (Uganda)	FEAST (Uganda)	Bangladesh
<i>n</i>	1,400	492	559	171
Age (years, IQR)	2.4 (1.4-3.7)	3.3 (2.2-4.6)	2.0 (1.2-3.3)	30 (23-45)
Proportion parasite positive (%)	100	100	59.4	100
Parasite density* (per $\mu\text{L}$ , IQR)	69,824 (6,099-316,350)	42,530 (10,635-198,540)	37,600 (3,640-153,680)	148,874 (23,550-348,540)
Platelet count ( $10^3/\mu\text{L}$ , IQR)	111 (64-215)	96 (49-170)	165 (75-326)	50 (27-139)
<i>Pf</i> HRP2 (ng/mL, IQR)	2,207 (419-5,072)	1,838 (588-4,097)	175 (0-1,953)	2,667 (1,083-6,128)
White blood cell count ( $10^3/\mu\text{L}$ , IQR)	12.6 (8.9-19)	10.4 (7.5-15.3)	12.0 (8.4-18.7)	9.0 (6.9-11.0)
Mortality (%)	11.1	6.7	11.4	26.9
HbAS ( <i>n</i> , %)	41 (2.9)	4 (0.8)	46 (8.2)	-
HbSS ( <i>n</i> , %)	7 (0.5)	23 (4.7)	21 (3.8)	-

Table 1: Patient characteristics across the four studies. For age, parasite densities, platelet counts, total white blood cell counts and *Pf*HRP2 concentrations we show the median values and inter-quartile ranges. No hemoglobin S (HbS) genotyping was done for the patients from Bangladesh as HbS is absent from that population. \*For FEAST, this excludes patients who had a negative RDT. IQR: inter-quartile range. HbAS: sickle trait; HbSS: homozygous sickle cell anemia.

107 *PfHRP2* concentration both had high discriminative value (illustrated by the receiver operating  
108 characteristic (ROC) curves). Overall, for thresholds giving the same sensitivity, the plasma  
109 *PfHRP2* concentration had a higher specificity. For example, in these populations, a lower  
110 limit of 1,000 ng/mL for the *PfHRP2* concentration had an estimated sensitivity of 87% and a  
111 specificity of 83%. In comparison, an upper limit for the platelet count of 150,000 per  $\mu\text{L}$  had  
112 an estimated sensitivity of 83% but a specificity of 71%. A series of sensitivity analyses (using  
113 non-informative priors, different choices for the parametric model, or using only data from the  
114 two studies of African children with severe malaria) showed near identical results (Fig. S1).

### 115 **Joint diagnostic thresholds for platelet counts and plasma *PfHRP2***

116 For clinical and epidemiological studies, and in contrast to clinical practice, specificity is usu-  
117 ally more important than sensitivity. We used both biomarkers in combination to improve the  
118 precision of the definition of severe malaria, in order to achieve low false positive rates. Multi-  
119 ple combinations of the platelet count and *PfHRP2* concentration have approximately the same  
120 operating characteristics for diagnosis, so the optimal choice is subjective and can be made  
121 based on operational simplicity (Fig. S2). In severe illness with clinical features consistent  
122 with severe malaria, platelet counts  $\leq 150,000$  per  $\mu\text{L}$  and plasma *PfHRP2* concentrations  $\geq$   
123 1,000 ng/mL in combination have an estimated diagnostic specificity of 93% and a sensitivity  
124 of 74%. If this definition was applied to a hospital cohort with a prevalence of severe malaria of  
125 60% (approximately the prevalence estimated for the Kenyan cohort of children with clinically  
126 diagnosed severe malaria), we would expect over 94% of the resulting population identified by  
127 these two biomarkers to have ‘true’ severe malaria (positive predictive value).

## 128 **Estimating the probability of severe malaria**

129 We fitted a three component parametric Bayesian latent class model to all available platelet  
130 count and *Pf*HRP2 concentration data from patients from the four studies of severe febrile  
131 illness (a total of  $n=2,622$ ) in order to estimate the individual patient probabilities that severe  
132 malaria was the true cause of their severe illness, and to calculate the prevalences of ‘true’  
133 severe malaria among those diagnosed. The additional third component captured a cluster of  
134 patients in FEAST trial who had no detectable plasma *Pf*HRP2 and normal platelet counts  
135 (Fig. 2). Under this model, we estimated that the prevalence of true severe malaria was  
136 96% (95% CI: 91-99) in Bangladeshi adults; 37% (95% CI: 31-42) in the Ugandan children  
137 enrolled in the FEAST trial (FEAST intentionally enrolled severely ill children with and without  
138 malaria, although 66% of all patients had a clinical diagnosis of severe malaria (23)); 74% (95%  
139 CI: 67-79) in the children enrolled in Kampala, Uganda (24); and 66% (95% CI: 61-70) in the  
140 children diagnosed with severe malaria in Kilifi, Kenya.

141 Fig. 2 shows scatter plots of the platelet counts versus the plasma *Pf*HRP2 concentra-  
142 tions, colored by the probability that the patient had severe malaria and grouped by study (dark  
143 blue: high probability of severe malaria; dark red: low probability of severe malaria). The model  
144 estimated a geometric mean platelet count in severe malaria across the four studies of 74,000  
145 per  $\mu\text{L}$  (95% of patients are predicted to have platelet counts between 17,000 and 312,000) and a  
146 geometric mean *Pf*HRP2 of 3,135 ng/mL (95% prediction interval: 402 to 24,452). The model-  
147 based probabilities of severe malaria were highly concordant with our previously published  
148 model which used platelet counts and total white blood cell counts ( $\rho=0.63$ , Fig. S3) (5).

149 We compared the estimated false diagnosis rates for cerebral malaria versus severe malaria  
150 without coma, and for severe malarial anemia versus severe malaria without severe anemia  
151 (where severe anemia is defined as a hematocrit  $\leq 15\%$ ) in the two cohorts of children clinically  
152 diagnosed with severe malaria. In the Kenyan cohort, the estimated false diagnosis rate was

153 slightly higher in the cerebral malaria group relative to the non-cerebral malaria groups: 36%  
154 of patients with coma, and 26% of patients without coma were classified as “falsely diagnosed”  
155 with severe malaria ( $p=0.001$ ). For the subgroup with severe anemia, the false diagnosis rate  
156 was 15%, compared to 36% in the subgroup without severe anemia. In the Ugandan cohort,  
157 both of these trends were reversed. A false diagnosis of severe malaria was estimated for only  
158 12% of patients with coma, but for 40% of patients without coma. For severe anemia, these  
159 proportions were 36% and 15%, respectively (the Ugandan cohort only recruited patients with  
160 either severe malarial anemia or patients with cerebral malaria).

161 Mortality rates varied substantially as a function of the estimated probability of severe  
162 malaria and across the studies (Fig. 3). In African children with a high probability of having  
163 severe malaria, the mean mortality was consistently around 10% for the three studies included.  
164 Apart from the Bangladeshi adults, all patients were treated with initial intravenous quinine.  
165 Mortality in severe malaria in adults from Bangladesh was substantially higher (~30%), con-  
166 sistent with previous studies (20). In Kenyan children, the mortality in the mis-classified group  
167 of patients was higher than in the correctly classified patients (14% versus 10%), whereas in the  
168 Ugandan study the trend was reversed (1% versus 10%). This is largely explained by the study  
169 populations: the group with a low probability of having severe malaria in the Ugandan study  
170 was predominantly composed of patients with severe anemia without other features of severity.

## 171 **Relationship with other admission variables**

172 We explored the relationship between the model estimated probability of severe malaria and  
173 the admission parasite densities, admission hematocrit, the total white blood cell counts, and  
174 the blood culture positivity rates (i.e. cultures growing a likely pathogen). In the three African  
175 studies, parasite densities were between 12 and 16 times higher in patients with a high proba-  
176 bility of severe malaria versus those with a low probability of severe malaria (Fig. 4). After

177 adjusting for study differences, the parasite densities were estimated to be 12.8-fold higher  
178 (95% C.I 9.9-16.5) in severe malaria versus not severe malaria.

179 The admission hematocrit were also highly correlated with the model estimated probability  
180 of severe malaria (Fig. S4). In the three African studies, children with a high probabilit-  
181 ity of having severe malaria had median admission hematocrit between 16 and 20% (FEAST:  
182 19%; Kampala: 16%; Kilifi: 20%). The hematocrit distributions in this group were unimodal  
183 (Fig. S5). In contrast, the hematocrit distributions in patients with low probabilities of hav-  
184 ing severe malaria ( $<0.2$ ) were strongly bi-modal, with the majority of patients having higher  
185 hematocrit (the median hematocrit in this group were: FEAST: 30%; Kampala: 12%; Kilifi:  
186 28%), but a substantial minority had low hematocrit of around 10%. The few Bangladeshi  
187 adults with a low probability of severe malaria also had higher hematocrit.

188 Blood cultures were done for all 1,400 Kenyan children and for 298 out of 332 (90%)  
189 Ugandan children in the FEAST study who had malaria parasitemia on admission. Overall  
190 51 and 35 patients respectively had positive blood cultures after removing contaminants. The  
191 probability of having severe malaria was highly predictive of the blood culture result, with an  
192 adjusted odds-ratio of 0.43 (95% CI: 0.27-0.66;  $p=0.0002$ ) for a positive culture in patients  
193 likely to have severe malaria versus those unlikely to have severe malaria (after adjustment for  
194 study). This difference in blood culture positivity rates was also reflected in the total white blood  
195 cell counts (Fig. S6). Across the four studies, after adjustment for age, in patients likely  
196 to have severe malaria (probability  $>0.5$ ) compared to those unlikely to have severe malaria  
197 (probability  $<0.5$ ) the odds-ratio for having a total white count  $>15,000$  per  $\mu\text{L}$  was 0.50 (95%  
198 CI: 0.46-0.56;  $p=10^{-12}$ ).

## 199 **Gene-dose relationship for hemoglobin S and severe malaria**

200 In the three studies in African children, the prevalence of both HbAS and HbSS were strongly  
201 inversely correlated with the model estimated probability of severe malaria (data are shown  
202 in Fig. 2). Pooling the three African studies and adjusting for differences in the estimated  
203 prevalence of severe malaria, the odds-ratio for being classified as severe malaria (probability  
204  $>0.5$ ) for HbAS patients relative to HbAA patients was 0.24 (95% CI: 0.15 to 0.40,  $p = 10^{-8}$ ),  
205 and for HbSS relative to HbAA the odds-ratio was 0.08 (95% CI: 0.04 to 0.18,  $p = 10^{-10}$ ).  
206 Under an additive model of association, each additional hemoglobin S (HbS) allele was asso-  
207 ciated with an odds-ratio for severe malaria of 0.27 (95% CI: 0.19 to 0.37,  $p = 10^{-16}$ ). This  
208 association between HbS genotypes and the biomarker-based severe malaria classification was  
209 highly concordant across the three studies.

## 210 **Discussion**

211 The diagnosis of severe malaria in African children is imprecise (3, 4, 26). This is because it is  
212 difficult to distinguish clinically between severe illness caused by malaria from severe illness  
213 with incidental asymptomatic or uncomplicated malaria since they share many clinical charac-  
214 teristics. This is a substantial problem for studies of one of the most important life threatening  
215 infections in childhood. It dilutes and distorts the results of epidemiology (27), pathophysiol-  
216 ogy (3), genetic association (5) and therapeutic investigations (18). In areas of moderate and  
217 high levels of malaria transmission asymptomatic parasitemia is very common. At any time a  
218 high proportion of children have detectable malaria parasitemia. Malaria parasitemia is there-  
219 fore a sensitive but not specific indicator that malaria is the cause of illness (26). Other simple  
220 laboratory tests provide valuable diagnostic information. Thrombocytopenia is a common fea-  
221 ture of all symptomatic malarias (6–8). A low platelet count therefore supports (but does not

222 prove) attributing malaria as the cause of illness. We estimate that less than one in five patients  
223 with severe malaria will have platelet counts in the normal range ( $>150,000$  per  $\mu\text{L}$ ). This  
224 analysis of data from large prospective studies of severe illness in African children shows that  
225 measurement of platelet counts and the plasma concentration of the parasite protein *PfHRP2*  
226 can together substantially improve the specificity of a diagnosis of severe falciparum malaria.

227 A key strength of this work is that we can validate the discriminant power of the platelet  
228 count and the plasma *PfHRP2* concentration by comparing the prevalence of HbS genotypes  
229 (HbAS and HbSS) and the blood culture positivity rates across in the inferred subgroups.  
230 HbAS is the genotype that provides the strongest known protection against severe falciparum  
231 malaria (28, 29). The prevalence of HbAS was four times lower in children considered likely to  
232 have severe malaria compared with those considered less likely to have severe malaria. It is im-  
233 portant to note that HbAS also protects against complications of malaria, such as an increased  
234 risk of bacterial infections, so the HbAS prevalence in children with presumed bacterial infec-  
235 tions is still expected to be lower than that in the healthy population (30). Recent work suggests  
236 that the protective effect of HbAS may vary according to parasite genotype (31). It may be  
237 that the few HbAS individuals who have “true” severe malaria have parasite genotypes that can  
238 evade the HbAS defense mechanisms. Future work will assess the relationship between these  
239 probabilistic classifications and the parasite genotypes. Whereas people with the AS genotype  
240 are essentially hematologically normal, those with homozygous SS suffer from sickle cell dis-  
241 ease. This causes anemia, and may also cause thrombocytopenia and leukocytosis. This could  
242 confound the use of full blood count data in the probabilistic assessment of severe malaria.  
243 However, the interpretation of the plasma *PfHRP2* concentration (or other parasite biomass in-  
244 dicators) should not be affected by sickle cell disease. In this study there was strong evidence  
245 for an additive effect between the number of HbS alleles and a decreased probability of severe  
246 malaria under the platelet/HRP2 model. This suggests that HbSS is strongly protective against a

247 high parasite biomass (32). However as sickle crises may be severe, and they are often triggered  
248 by infections, it is possible a low parasite biomass could trigger severe events. Whether or not  
249 this should be described as ‘severe malaria’ is a semantic question (33). A recent in-depth anal-  
250 ysis of the Ugandan cohort from Kampala showed that the HbSS children considered to have  
251 severe malaria had a lower parasite biomass but higher levels of endothelial dysregulation (32).  
252 The main differential diagnosis in suspected severe malaria is bacterial sepsis. The relationship  
253 is complicated as severe malaria does predispose to bacterial sepsis (34). Blood culture has low  
254 sensitivity, however we observed a three fold increase in positivity rates in the patients likely  
255 have been mis-diagnosed compared to those with a likely correct diagnosis.

256 The main limitation of this study is the absence of a gold-standard diagnosis, and thus the  
257 reliance on parametric latent class models. The application of latent class models to multiple  
258 populations with varying disease prevalence provides a powerful framework for estimating the  
259 diagnostic accuracy of imperfect tests, but the model outputs rely on the validity of key assump-  
260 tions, notably the underlying parametric models used for the biomarker distributions. We used  
261 log-normal distributions for the platelet count and plasma *PfHRP2* concentration in both ‘true’  
262 severe malaria and ‘not severe malaria’. Visual data plots suggest that this is a good approxi-  
263 mation of the true underlying distributions. Each individual biomarker has its own limitations.  
264 Thrombocytopenia can be caused by other infections (e.g. severe arbovirus infection) and may  
265 occur in sepsis, although it is much less prevalent in sepsis than malaria (8). The main limitation  
266 of plasma *PfHRP2* as a marker of total parasite biomass is that it requires specialist measure-  
267 ment (although rapid point-of-care tests can be applied to plasma (35) and modifications are  
268 under development). As plasma *PfHRP2* accumulates each asexual parasite cycle, an indi-  
269 vidual with a sustained low parasite multiplication rate at high parasite densities can have the  
270 same *PfHRP2* concentration as an individual with a fulminant infection and a high sequestered  
271 biomass (17). In addition mutations in the *PfHRP2* gene causing changes in antigenicity may

272 affect immunoassays, although these are rare outside of the Horn of Africa (36, 37).

273 In these very large prospective series of patients hospitalised with a diagnosis of severe  
274 malaria, combining platelet counts and plasma *PfHRP2* concentrations provided good discrim-  
275 ination between “true” severe falciparum malaria and other severe illnesses (likely to be bacter-  
276 aemia in many of the cases) with concomitant incidental malaria. A major strength of this study  
277 is in combining two measures which are mechanistically distinct (platelet counts are measuring  
278 the host response to acute malaria illness; *PfHRP2* is measuring the parasite biomass). We show  
279 that in combination these biomarkers provide a high level of discrimination between patients  
280 who were likely to have had true severe malaria and those with a different illness aetiology.  
281 This proportion ranged from one third in the FEAST study which intentionally studied children  
282 with both severe malaria and other severe illnesses (likely mainly sepsis) requiring fluid re-  
283 suscitation to >90% in Bangladesh -a low transmission area where the diagnosis of falciparum  
284 malaria as the cause of illness is highly specific. Overall it suggests that approximately one third  
285 of African children diagnosed with severe malaria have another cause of their severe illness (5).  
286 Our results suggest that severe malaria with a normal platelet count is unusual and a platelet  
287 count >150,000 per  $\mu\text{L}$  in a child with suspected severe malaria should motivate further ex-  
288 amination for a possible different cause of illness, while giving prompt antimalarial treatment.  
289 The high diagnostic accuracy of the plasma *PfHRP2* concentration should motivate future work  
290 on simple modifications to the currently used *PfHRP2* rapid tests, in order to estimate plasma  
291 *PfHRP2* at the bedside. Finally, we recommend that future studies of severe falciparum malaria  
292 should include only children with platelet counts of  $\leq 150,000$  per  $\mu\text{L}$  and plasma *PfHRP2*  
293 concentrations of  $\geq 1,000$  ng/mL.

## 294 **Methods**

### 295 **Study design**

296 This study is a retrospective analysis of platelet counts and plasma *PfHRP2* concentrations in  
297 patients with severe febrile illness. We merged data from four separate studies, three of which  
298 were studies of severe malaria and one was a study of fluid resuscitation in severe febrile ill-  
299 ness (FEAST trial). The main goal of the analysis was to use Bayesian latent class modelling  
300 to estimate the proportion of mis-diagnosed patients in the three severe malaria studies, and to  
301 assess the operating characteristics of different cutoff values for the platelet count and *PfHRP2*  
302 concentrations for future clinical studies. For the analysis and reporting we followed the EQUA-  
303 TOR guidelines for the reporting of diagnostic accuracy studies that use Bayesian latent class  
304 models (STARD-BLCM, see Supplementary Materials).

### 305 **Data**

306 All the clinical studies were prospective studies of severe illness and had appropriate ethical  
307 approval.

#### 308 **Bangladesh**

309 We included data from observational studies in severe falciparum malaria conducted by the  
310 Mahidol Oxford Tropical Medicine Research Unit in Bangladesh between 2003 and 2019.  
311 These pooled data have been described previously (38). Malaria transmission is seasonal and of  
312 low intensity in this location. In brief, adults and children with microscopy evidence for asex-  
313 ual stage malaria parasites (thick and thin blood films) who met the WHO definition for severe  
314 malaria were enrolled in study after written informed consent was obtained from the patient or  
315 an attending relative.

316 Criteria for severe malaria included coma (Glasgow Coma Score <11 or Blantyre Coma

317 Score <3), pulmonary oedema, repeated convulsions ( $\geq 2$  in 24 hours), severe anemia (hemat-  
318 ocrit level <20%, plus a parasite count >100 000 parasites/ $\mu$ L) or jaundice (bilirubin level >3.0  
319 mg/dL, plus a parasite count >100 000 parasites/ $\mu$ L), renal failure (serum creatinine level >3  
320 mg/dL), hypoglycemia (blood glucose level <40 mg/dL), shock (systolic blood pressure <80  
321 mm Hg with cool extremities), hyperparasitemia (peripheral asexual stage parasitemia >10%),  
322 hyperlactemia (venous plasma lactate >4 mmol/L), and/or acidemia (venous plasma bicarbon-  
323 ate level <15 mmol/L).

324 Platelet counts and *PfHRP2* concentrations were jointly measured in a total of 172 patients.  
325 The majority of these patients received intravenous artesunate.

### 326 **Kilifi (Kenya)**

327 The Kenyan case-control cohort has been described in detail previously (25). Severe malaria  
328 cases consisted of all children aged <14 years who were admitted with clinical features of  
329 severe falciparum malaria to the high dependency ward of Kilifi County Hospital between  
330 June 11th 1999 and June 12th 2008. Severe malaria was defined as a positive blood-film  
331 for *P. falciparum* along with: prostration (Blantyre Coma Score of 3 or 4); cerebral malaria  
332 (Blantyre Coma Score of <3); respiratory distress (abnormally deep breathing); severe ane-  
333 mia (hemoglobin <5 g/dL). The standard of care antimalarial treatment during this period was  
334 intravenous quinine.

### 335 **FEAST (Uganda)**

336 FEAST was a multi-center randomized controlled trial comparing fluid boluses for severely  
337 ill children with shock ( $n = 3,161$ ) that was not specific to severe malaria (23). Children  
338 between 2 months and 12 years of age were eligible for the trial if they presented with a severe  
339 febrile illness complicated by impaired consciousness (prostration or coma), respiratory distress  
340 (increased work of breathing), or both, and with impaired perfusion, as evidenced by one or

341 more of the following: a capillary refill time of 3 or more seconds, lower-limb temperature  
342 gradient, 19 weak radial-pulse volume, or severe tachycardia (>180 beats per minute in children  
343 younger than 12 months of age, >160 beats per minute in children 1 to 5 years of age, or  
344 >140 beats per minute in children older than 5 years of age). Exclusion criteria were severe  
345 malnutrition, gastroenteritis, noninfectious causes of shock (e.g., trauma, surgery, or burns).

346 Platelet counts and plasma *PfHRP2* concentrations were measured in 502 children in the  
347 Ugandan sites (Mulago National Referral Hospital, Mbale and Soroti Regional Referral Hos-  
348 pitals and St Mary's Hospital, Lacor). The standard of care antimalarial treatment during this  
349 period was intravenous quinine.

### 350 **Kampala (Uganda)**

351 The trial by Brand *et al* was an observational study of cerebral malaria and severe malarial ane-  
352 mia in Mulago Hospital, Kampala, Uganda (24). Children were enrolled if they were between  
353 18 months and 12 years of age. Cerebral malaria was defined as coma (Blantyre Coma Score  
354 of <3 or Glasgow Coma Score <8) in presence of asexual parasites on a blood smear. Severe  
355 malarial anemia was defined as presence of *Plasmodium falciparum* parasites on a blood smear  
356 in children with a hemoglobin level  $\leq 5$ g/dL. The standard of care antimalarial treatment during  
357 this study was intravenous quinine.

### 358 **Procedures**

359 For the Kilifi cohort and the Ugandan sites in the FEAST trial, pediatric blood samples for  
360 bacterial cultures were collected in BACTEC Peds Plus bottles and processed with a BACTEC  
361 automated blood culture instrument (Becton Dickinson) for initial detection of bacteria in the  
362 blood. BACTEC-positive samples were subcultured on standard media by routine microbio-  
363 logical techniques. Either biochemical test kits (API, BioMérieux), serological tests, or both

364 were used to confirm suspected pathogens. Good Clinical Laboratory Practice was audited by  
365 Qualogy and external quality assurance was provided by the UK National External Quality As-  
366 sessment service. The following organisms were considered as contaminants: *Bacillus* species,  
367 coryneforms, *Micrococcus* species, coagulase- negative *Staphylococcus* and *Citrobacter*.

368 In all four studies plasma *Pf*HRP2 levels were quantitated using the same previously pub-  
369 lished methodology (18). The lower limit of detection of the ELISA plasma assay is approxi-  
370 mately 2 ng/mL. Patients in the Kilifi study and the FEAST trial were genotyped for the rs334  
371 SNP (HbS) using DNA extracted from fresh or frozen samples of whole blood as described in  
372 detail previously (25, 39).

### 373 **Statistical analysis**

374 We fitted a series of Bayesian parametric latent class models to the available biomarker data  
375 (21). The key assumptions of the main models are summarized as follows:

- 376 1. The marginal distribution of each biomarker in each latent class is log-normal.
- 377 2. The marginal distribution of each biomarker in each latent class is the same across the  
378 different studies and countries.
- 379 3. Informative Bayesian priors on all parameters.

380 Sensitivity analyses relaxed assumptions 1-3 (using bivariate t-distributions, only using data  
381 from African children, and using weakly informative priors). We compared models with and  
382 without correlation between the biomarkers within each latent class. The models with correla-  
383 tion performed better and were more conservative.

384 In an initial exploratory analysis, we used the R package *mclust* (40) (Gaussian Mixture  
385 Modelling for Model-Based Clustering - this uses the EM algorithm for parameter estimation)  
386 to estimate the number of latent components in the data (Fig. S7). This suggested three

387 major underlying clusters, one of which was only present in the FEAST trial (patients with no  
388 measurable plasma *PfHRP2* and normal platelet counts). We then used full Bayesian inference  
389 with informative priors to determine the distributions of the latent clusters. In this initial ex-  
390 ploratory analysis, we also looked at the relationship between the peripheral parasite density  
391 and the two biomarkers (Fig. S8). This preliminary analysis confirmed that the parasite  
392 density is a poor biomarker for discriminating between ‘true’ severe malaria and other causes  
393 of illness as it was not possible to robustly fit clustering models with either the platelet count or  
394 the plasma *PfHRP2*.

395 The first set of models used data from the three severe malaria studies only (Kenyan children  
396 from Kilifi, Ugandan children from Kampala, and Bangladeshi adults). For each biomarker,  
397 the marginal distribution in each latent class (severe malaria versus not severe malaria) was  
398 assumed to be log-normal with the following informative priors (weakly informative). Means  
399 and standard deviations are given on the  $\log_{10}$  scale.

- 400 • The mean platelet count:  $N(\log_{10} 75, 0.1)$  [weakly informative:  $N(\log_{10} 100, 0.5)$ ] for  
401 severe malaria and  $N(\log_{10} 200, 0.1)$  [weakly informative:  $N(\log_{10} 200, 0.5)$ ] for not se-  
402 vere malaria. The platelet count geometric mean of 100,000 per  $\mu\text{L}$  versus 200,000 per  
403  $\mu\text{L}$  in severe malaria versus not severe malaria is informed by the results of (3) (autopsy  
404 study in patients who had died with a diagnosis of cerebral malaria: the patients with  
405 evidence of parasite sequestration all had very low platelet counts compared to those who  
406 did not; we expect slightly higher platelet counts in all patients as the platelet count is a  
407 prognostic factor as well as diagnostic).
- 408 • The mean *PfHRP2*:  $N(\log_{10} 2500, 0.2)$  for severe malaria and  $N(\log_{10} 500, 0.2)$  for not  
409 severe malaria. The weakly informative prior used a standard deviation of 0.5 instead of  
410 0.2. The mean plasma *PfHRP2* values are informed from previous modeling of data from

411 the AQUAMAT trial (18).

- 412 • The standard deviation on the  $\log_{10}$  scale for each log-normal distribution was given a  
413 exponential prior with rate parameter set to 2 (i.e. a mean standard deviation of 0.5 on the  
414  $\log_{10}$  scale)

415 We used informative beta priors for the prevalence of severe malaria in the three studies:  
416 Beta(19,1) for Bangladesh; Beta(14,6) for Kilifi; Beta(14,6) for Kampala (the weakly infor-  
417 mative beta priors: Beta(4,1), Beta(2,1), and Beta(2,1), respectively).

418 To assess the robustness of the model outputs, we performed three sensitivity analyses: (i)  
419 changing the parametric form to a bivariate student  $t$ -distribution with 10 degrees of freedom  
420 (robustness to assumption 1); (ii) by fitting the model to data only from the studies in African  
421 children with severe malaria (Kampala and Kilifi; robustness to assumption 2 as the Bangladeshi  
422 adults could plausibly be different from African children in terms of parasite biomass and host  
423 response); (iii) by using weakly informative priors (larger standard deviations around the prior  
424 mean values and weaker beta priors on the prevalence parameters).

425 For the second set of models, we included data from the FEAST trial (23). FEAST inten-  
426 tionally enrolled severely ill patients with and without malaria, thus there is a subset of patients  
427 who are clearly distinct from all patients in the studies of severe malaria (this subset is char-  
428 acterized on average by a negative malaria blood slide, no measurable plasma *PfHPR2* and a  
429 normal platelet count). The latent class model needed to include a third component in order to fit  
430 the data. Under this model, the biomarker distributions were also assumed to be log-normal in  
431 the severe malaria class, but were a mixture of two log-normal distributions for the ‘not severe  
432 malaria’ class.

433 We used the same priors for the second set of models with the addition of the following  
434 priors for the third additional component (which can be summarized as severe illness with no

435 evidence of acute malarial infection):

436 • The mean platelet count:  $N(\log_{10} 300, 0.1)$  for not severe malaria.

437 • The mean *Pf*HRP2:  $N(\log_{10} 1, 0.2)$  for not severe malaria.

438 For the three component model, Dirichlet priors were used for the prevalence parameters for  
439 each latent class: Dirichlet(19,1,0.1) for the Bangladesh study (with hyperparameters corre-  
440 sponding to the severe malaria and the two not severe malaria classes, respectively); Dirich-  
441 let(3,3,3) for the FEAST trial; Dirichlet(14,7,1) for the Kampala and Kilifi studies.

442 In order for patients with non-measurable plasma *Pf*HRP2 concentrations to be included  
443 in this analysis, we set all non-measurable concentrations to 1 ng/mL (approximately half the  
444 lower limit of detection of the ELISA assay). Patients who had non-measurable plasma *Pf*HRP2  
445 concentrations but parasite densities above 1,000 per  $\mu\text{L}$  were excluded from the analysis as  
446 these could represent *P. falciparum* parasites with *HRP2/3* gene deletions. The median platelet  
447 count in these samples was 142,000 per  $\mu\text{L}$  (range: 28,000 to 808,000 per  $\mu\text{L}$ ).

448 All posterior distributions were estimated using Monte Carlo methods. All models were  
449 implemented in the *rstan* language (41). For each model we ran 4 independent chains for  
450 2000 iterations, discarding half for burn-in. Convergence was checked by visually assessing  
451 traceplots. Convergence problems due to ‘label switching’ were avoided by using parameter  
452 constraints: the mean value of each likelihood distribution was set as increasing (for the platelet  
453 count) or decreasing (for the *Pf*HRP2 and parasite density). This uses the *ordered* parameter  
454 class in *stan*. The *stan* language does not support discrete parameters, however posterior dis-  
455 tributions of latent class models can be sampled by using the *logsumexp* trick, computing the  
456 marginal likelihood over the different class combinations.

## 457 **Code and data availability**

458 All code in the form of an RMarkdown document along with a minimal clinical dataset is avail-  
459 able online via the following github repository: [https://github.com/jwatowatson/  
460 SevereMalariaDiagnosis](https://github.com/jwatowatson/SevereMalariaDiagnosis). A static version can be found at [https://doi.org/10.  
461 5281/zenodo.5602924](https://doi.org/10.5281/zenodo.5602924) (42).

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## 509 **Acknowledgements**

510 This research was funded by Wellcome. A CC BY or equivalent licence is applied to the author  
511 accepted manuscript arising from this submission, in accordance with the grant's open access  
512 conditions.

513 This work was done as part of SMAART (Severe Malaria Africa – A consortium for Re-  
514 search and Trials) funded by a Wellcome Collaborative Award in Science grant (209265/Z/17/Z)  
515 held in part by KM, NPJD and AMD.

516 JAW is a Sir Henry Dale Fellow jointly funded by the Wellcome Trust and the Royal Society  
517 (223253/Z/21/Z). TNW and NJW are senior and principal research fellows respectively funded  
518 by the Wellcome Trust (202800/Z/16/Z and 093956/Z/10/C, respectively). ECG acknowledges  
519 funding from a core grant to the MRC CTU at UCL from the MRC (MC\_UU\_12023/26). CCJ  
520 and ROO acknowledge grant R01 NS055349 from the National Institutes for Neurologic Dis-  
521 orders and Stroke. The FEAST trial was supported by a grant (G0801439) from the Medical  
522 Research Council, United Kingdom provided through the (MRC) DFID concordat. KM and  
523 ECG were supported by this grant. This paper is published with permission from the Director  
524 of the Kenya Medical Research Institute (KEMRI).

525 JAW conceived the study, designed the experiments, analysed the data and wrote the first  
526 draft of the paper. NJW, TNW, KM supervised the work. SU, PW, JM, GMN, NM, ECG, CW,  
527 NPJD, PB, ROO, AMD, CJ, KM, TNW, NJW contributed data from the clinical studies. All  
528 authors read and revised the paper.

529 No competing interests are declared.

## 530 **Supplementary materials**

531 EQUATOR STARD-BLCM checklist

532 Figs. S1 to S7

533

534 **Fig. 1.** The diagnostic value of platelet counts (pink) and plasma *PfHRP2* concentrations  
535 (green) estimated using data from 2,063 patients diagnosed with severe falciparum malaria in  
536 three studies. Sensitivities (dotted lines) and specificities (dashed lines) are estimated under  
537 a Bayesian parametric two-component latent class model which assumes the same sensitivity  
538 and specificity for each biomarker across the three studies (African children: Kilifi, Kenya and  
539 Kampala, Uganda; Asian adults: Bangladesh). For the platelet count, thresholds correspond to  
540 upper limits, whereas for the *PfHRP2* concentration the thresholds correspond to lower limits.

541 **Fig. 2.** Probabilistic model of severe falciparum malaria using platelet counts and plasma  
542 *PfHRP2* concentrations in 2,622 severely ill patients based on a Bayesian parametric latent class  
543 model with three latent classes (a severe malaria class and two ‘not severe malaria’ classes). The  
544 colors correspond to the probability of severe malaria under the model (dark blue: high proba-  
545 bility; dark red: low probability). Triangles show the HbAS individuals; crosses show the HbSS  
546 individuals. In order to show data points with non-measurable plasma *PfHRP2*, non-measurable  
547 concentrations were set to  $1 \text{ ng/mL} \pm \text{random jitter}$  on the  $\log_{10}$  scale (approximately half the  
548 lower limit of quantification of the assay).

549 **Fig. 3.** Mortality as a function of the probability of having severe malaria under the Bayesian  
550 latent class model (based on platelet counts and *PfHRP2* concentrations). The lines (shaded  
551 areas) show mean (95% confidence intervals) mortality estimates from logistic regression fits.

552 **Fig. 4.** Admission parasite densities as a function of the probability of severe malaria under the  
553 Bayesian latent class model. Data from the FEAST trial includes only the malaria RDT positive  
554 patients. The thick lines show the additive linear model fit (spline based).