Common virulence gene expression in naïve and severe malaria cases

Jan Stephan Wichers[1,2], Gerry Tonkin-Hill[3], Thorsten Thye[4], Ralf Krumkamp[4,5], Jan Strauss[1,2,6], Heidrun von Thien[1], Benno Kreuels[4], Judith A.M. Scholz[1], Helle Holm Hansson[7], Rasmus Weisel Jensen[7], Louise Turner[7], Freia-Raphaella Lorenz[8], Anna Schöllhorn[8], Iris Bruchhaus[1,6], Egbert Tannich[4,5], Rolf Fendel[8,9], Thomas D. Otto[10], Thomas Lavstsen[7], Tim-Wolf Gilberger[1,2,6], Michael F. Duffy[11], Anna Bachmann[1,2,5, #]

[1] Molecular Biology and Immunology, Bernhard Nocht Institute for Tropical Medicine, 20359 Hamburg, Germany

[2] Centre for Structural Systems Biology, 22607 Hamburg, Germany

[3] Wellcome Sanger Institute, Hinxton/Cambridge CB10 1SA, UK

[4] Epidemiology and Diagnostics, Bernhard Nocht Institute for Tropical Medicine, 20359 Hamburg, Germany

[5] German Center for Infection Research (DZIF), partner site Hamburg-Borstel-Lübeck-Riems, Germany

[6] Biology Department, University of Hamburg, 22609 Hamburg, Germany

[7] CMP, University of Copenhagen, 2200 Copenhagen, Denmark

[8] Institute of Tropical Medicine, University of Tübingen, 72074 Tübingen, Germany

[9] German Center for Infection Research (DZIF), partner site Tübingen, Germany

[10] Institute of Infection, Immunity and Inflammation, University of Glasgow, Glasgow G12 8TA, UK

[11] Department of Microbiology and Immunology, University of Melbourne, Melbourne/Parkville VIC 3052, Australia

\# corresponding author

Running title: In host expression of *var* genes

Abstract

Sequestration of *Plasmodium falciparum*-infected erythrocytes to host endothelium through the parasite-derived *Pf*EMP1 adhesion proteins, is central to the development of malaria path-ogenesis. *Pf*EMP1 proteins have diversified and expanded to encompass many sequence var-iants conferring the same array of human endothelial receptor binding phenotypes. Here, we analyzed RNA-seq profiles of parasites isolated from 32 infected travelers returning to Ger-many. Patients were categorized into either malaria naïve (n=15) or pre-exposed (n=17), and into severe (n=8) or non-severe (n=24) cases. Expression analysis of *Pf*EMP1-encoding *var* genes showed that severe malaria was associated with expression of *Pf*EMP1 containing the endothelial protein C receptor (EPCR)-binding CIDRα1 domain, whereas CD36-binding *Pf*EMP1 was linked to non-severe malaria outcomes. In addition, gene expression-guided de-termination of parasite age, suggested that circulating parasites from non-severe malaria pa-tients were older than parasites from severe malaria patients. First-time infected patients were also more likely to develop severe symptoms and tended to be infected for a longer period, which thus appeared to select for parasites with more sequestration-efficiency and therefore more pathogenic *Pf*EMP1 variants.

Keywords: *P. falciparum*, *Pf*EMP1, RNA-seq, transcriptomics, variant surface antigens

Introduction

Despite considerable efforts during recent years to combat malaria, this disease remains a major threat to public health in tropical countries. The most severe clinical courses of malaria are due to infections with the protozoan species *Plasmodium falciparum.* In 2018, there were 228 million cases of malaria worldwide, resulting in more than 400,000 deaths (WHO, 2019). Currently, about half of the world population is living at risk of infection and more than 90% of the malaria deaths occur in Africa. In particular, children under the age of five and pregnant women suffer from severe disease. The virulence of *P. falciparum* is linked to the infected erythrocytes binding to endothelial cell surface molecules expressed on blood vessel walls. This phenomenon, known as sequestration, prevents the passage of infected erythrocytes through the spleen, which would otherwise remove the infected erythrocytes from the circula-tion and kill the parasite (Saul, 1999). The membrane proteins mediating sequestration are exposed to the host's immune system and through evolution *P. falciparum* parasites have ac-quired several multi-copy gene families coding for variant surface antigens (VSAs) allowing immune escape through extensive sequence polymorphisms. Endothelial sequestration is me-diated by the *P. falciparum* erythrocyte membrane protein 1 (*Pf*EMP1) family, which members have different binding capacities for host vascular tissue receptors such as CD36, EPCR,

ICAM-1, PECAM1, receptor for complement component C1q (gC1qR) and CSA. *Pf*EMP1 proteins are known to mediate adhesion of infected erythrocytes to the linings of small blood vessels (Magallón-Tejada *et al*, 2016; Turner *et al*, 2013; Rowe *et al*, 2009). The long, variable, extracellular *Pf*EMP1 region responsible for receptor binding contains a single N-terminal segment (NTS main classes A, B and pam) and a variable number of different Duffy-binding like (DBL main classes are DBLα-ζ and pam) and cysteine-rich inter-domain region domains (CIDR main classes are CIDRα-δ and pam) domains (Rask *et al*, 2010). Based on recent findings, the sub-classification within main domain classes, e.g. the DBLβ subclasses 1 – 13, was questioned due to recombination occurring frequently between members of the different subclasses (Otto *et al*, 2019). *Pf*EMP1 molecules have been grouped into four categories (A, B, C and E) depending on the protein domain composition as well as the 5' upstream sequence, the chromosomal localization and the direction of transcription of their encoding *var* genes (Rask *et al*, 2010; Kyes *et al*, 2007; Kraemer & Smith, 2003; Lavstsen *et al*, 2003). Each parasite possesses about 60 *var* genes with approximately the same distribution over the different groups (Rask *et al*, 2010). About 10% of the genes belong to A-type *var* genes, typically encoding longer *Pf*EMP1 proteins with a head-structure containing a DBLα1 and either a CIDRα1 or a CIDRβ/γ/δ domain. In some A-type proteins, an ICAM-1-binding DBLβ1 or 3 domain follows this head-structure (Lennartz *et al*, 2017). Two conserved subfamilies also belong to the group A: the *var1* gene present in two different allele forms in the parasite population (3D7- and IT-type) (Otto *et al*, 2019) and the shortest *var3* gene family with only two extracellularly exposed domains (DBLα1.3 and DBLε8). Most *Pf*EMP1 proteins belong to the B- and C-families and have a DBLα0-CIDRα2-6 head structure attached to a DBLδ1-CIDRβ/γ domain combination. Some B-type proteins possess a DBLα2-CIDRα1 head-structure typically followed by additional domains including DBLβ12 domains suggested to bind gC1qR (Magallón-Tejada *et al*, 2016). Other B-type *Pf*EMP1 have a ICAM-1-binding DBLβ5 domain (Lennartz *et al*, 2019). The VAR2CSA *Pf*EMP1 binds placental CSA and cause pregnancy-associated malaria. The *var2csa* genes constitute the group E and most *P. falciparum* isolates possess a single or two gene copies of this inter-strain conserved *var* gene variant. Based on the head structure domains, *Pf*EMP1 divide into those with DBLα1 (A-type *Pf*EMP1), DBLα2 (B-type *Pf*EMP1) or DBLα0 (B- and C-type *Pf*EMP1) domains as well as those binding CD36 via CIDRα2-6 domains (B- and C-type *Pf*EMP1) or EPCR via CIDRα1 (found in A- and B-type *Pf*EMP1) (Otto *et al*, 2019). Accordingly, the head structure confers mutually exclusive binding properties, either to EPCR, CD36 or to an unknown receptor via the CIDRβ/γ/δ domain of some A-type proteins. Some domain sequence variants are found to often co-occur (Otto *et al*, 2019; Berger *et al*, 2013; Rask *et al*, 2010). For example, one domain cassette (DC), DC8, includes specific CIDRα1.1/8 subtypes capable of binding EPCR.

Group A *var* gene expression has been associated with severe forms of malaria, whereas mild
malaria may be associated with group C expression (Avril *et al*, 2012; Claessens *et al*, 2012;
Lavstsen *et al*, 2012; Falk *et al*, 2009; Warimwe *et al*, 2009; Kyriacou *et al*, 2006; Rottmann *et*
al, 2006; Jensen *et al*, 2004; Kirchgatter & Del Portillo, 2002). Conflicting results were reported
on group B expression during severe disease; however, this is likely explained by the fact that
a subset of B-type *Pf*EMP1 share EPCR and ICAM-1 receptor-binding phenotypes with A-type
*Pf*EMP1. Indeed, consensus from a range of gene expression studies is that severe malaria is
associated with expression of *Pf*EMP1 with EPCR-binding CIDR α1 domains (Jespersen *et al*,
2016; Kessler *et al*, 2017; Storm *et al*, 2019; Shabani *et al*, 2017; Mkumbaye *et al*, 2017;
Bernabeu *et al*, 2016; Magallón-Tejada *et al*, 2016). No other domain has been consistently
associated with severity of disease, but elevated expression of some specific DCs has also
been associated with severe disease, in particular the EPCR-binding DC8 (DBLα2-CIDRα1.1-
DBLβ12-DBLγ4/6), DC13 (DBLα1.7-CIDRα1.4) and DC4 (DBLα1.4-CIDRα1.6-DBLβ3), but
also the DC5 found in A-type *Pf*EMP1 (DBLγ12-DBLδ5-CIDRβ3/5) and DC6 (DBLγ14,-DBLζ5-
DBLε4) found in both A- and B-type *Pf*EMP1 (Bernabeu *et al*, 2016; Magallón-Tejada *et al*,
2016; Berger *et al*, 2013; Avril *et al*, 2013, 2012; Claessens *et al*, 2012; Lavstsen *et al*, 2012).
In this study, we used an RNA-seq based analysis to study gene expression with special em-
phasis on *var* genes in parasites from hospitalized travelers returning from malaria endemic
countries with certified *P. falciparum* infections. Individuals were clustered into i) first-time in-
fected and ii) pre-exposed individuals on the basis of serological data or into iii) severe and iv)
non-severe cases according to medical reports. Our multi-dimensional analysis reveals a clear
association of domain cassettes with EPCR-binding properties with a naïve immune status
and severe malaria, whereas CD36-binding *Pf*EMP1 proteins and the conserved *var1*-3D7
allele were expressed at higher levels in pre-exposed patients and non-severe cases. Inter-
estingly, circulating parasites from severe cases tended to be younger than parasites from
non-severe cases, indicating that the EPCR-binding phenotype confers more efficient seques-
tration of infected erythrocytes.

Results

Cohort characterization

This study is based on a cohort of 32 malaria patients hospitalized in Hamburg, Germany.
Parasitemia, recorded clinical symptoms and patient sub-grouping are summarized in Table
1. For ten patients, the present malaria episode was their first recorded *P. falciparum* infection.
Nine individuals had previously experienced malaria episodes according to the medical re-
ports, whereas malaria exposure was unknown for 13 patients. In order to determine if patients
already had an immune response to *P. falciparum* antigens, indicative of previous exposure to
malaria, plasma samples were analyzed by a Luminex-based assay displaying the antigens

AMA1, MSP1 and CSP (Table S1). Immune responses to AMA1 and MSP1 are known to be
long-lasting and seroconversion to AMA1 is assumed to occur after only a single or very few
malaria infections (Drakeley *et al*, 2005). Principal component analysis of the Luminex data
resulted in separation of the patients into two discrete groups corresponding to first-time in-
fected adults ('naïve cluster') and malaria pre-exposed individuals ('pre-exposed cluster') (Fig-
ure 1A). The 13 patients with unknown malaria exposure status could be grouped into either
the naïve or pre-exposed groups defined by the PCA of the antigen reactivity. The only outlier
in the clustering was a 19-year-old patient (#21) from Sudan, who reported several malaria
episodes during childhood, but clustered with the malaria-naïve patients.
Plasma samples were further subjected to i) a merozoite-directed antibody-dependent respir-
atory burst (mADRB) assay (Kapelski *et al*, 2014), ii) a PEMS-specific ELISA and iii) a protein
microarray with 228 *P. falciparum* antigens (Borrmann, 2020). Analysis of these serological
assays in relation to the patient clustering confirmed the expected higher and broader antigen
recognition by ELISA, protein microarray, and stronger ability to induce burst of neutrophils by
serum from the group of malaria pre-exposed patients (Figure 1B–D, Table S1). Data from all
the serological assays were next used for an unsupervised random forest machine learning
approach to build models predictive of individual's protective status. This algorithm confirmed
the classification of patient #21 as being non-immune (Figure 1E). The calculated variable
importance highlighted the relevance of the mADRB assay, the ELISA and the Luminex to
allocate patients into cluster (Figure 1F). A multidimensional scaling plot was used to visualize
cluster allocation and patient #26 positioned at the borderline to pre-exposed patients (Figure
1E). The patient was grouped into the naïve cluster in accordance with the Luminex data and
the medical report showing that this patient from Jamaica was infected during his first trip to
Africa.
Using protein microarrays, the antibody response against described antigens was analyzed in
detail. As expected, pre-exposed individuals showed significantly elevated IgG antibody re-
sponses against a wide range of parasite antigens, especially typical blood stage markers,
including MSP1, MSP2, MSP4, MSP10, EBA175, ring exported protein 1 (REX1), and AMA1
(Figure 1D, upper panel). Markers for a recent infection, MSP1, MSP4, GLURP and Early
transcribed membrane protein 5 (ETRAMP5) (Van Den Hoogen *et al*, 2019), were significantly
elevated in the pre-exposed individuals in comparison to the defined first-time infected group.
In addition, further members of the ETRAMP family, including ETRAMP10, ETRAMP14,
ETRAMP10.2 and ETRAMP4, and also antibodies against pre-erythrocytic antigens, such as
CSP, STARP and LSA3, were highly elevated. Similar effects were detectable for IgM antibod-
ies; previous exposure to the malaria parasite led to higher antibody levels (Figure 1D, lower
panel).

Based on medical reports eight patients from the malaria-naïve group were considered as
more severe cases, having a substantially higher parasitemia and an impaired function of di-
verse organs especially the brain (Table 1). Patient #1 was included into the severe group due
to increasing parasitemia during hospitalization and circulating schizonts indicative of a very
high sequestering parasite biomass associated with severity (Bernabeu *et al*, 2016). The re-
maining 24 cases were summarized into the non-severe malaria group. Comparing the IgG
antibody response of severe and non-severe cases within the previously malaria-naïve group,
elevated antibody levels were detected in the severe subgroup. The highest fold change was
observed in antibodies directed against intracellular proteins, such as DnaJ protein, GTPase-
activating protein or heat shock protein 70 (Figure S1). Interestingly, IgM antibodies against
ETRAMP5 were detectable in the severely infected individuals, suggesting they were infected
for a prolonged period of time compared to the mild malaria population (Helb *et al*, 2015; Van
Den Hoogen *et al*, 2019).

RNA-seq transcriptomics
Parasites were isolated from the venous blood of all patients for subsequent transcriptional
profiling. Transcriptome libraries were sequenced for all 32 patient samples. The number of
trimmed reads ranged between 29,142,684 and 82,000,248 (median: 41,383,289) within the
individual libraries derived from patients. The proportion of total reads specific for *P. falciparum*
were 87.7% (median; IQR: 76.7–91.3). Lower percentages (12.4% and 15.68%) were obtained
only for patient isolates #1 and #2 (Table S2). These samples were not subjected to globin-
mRNA depletion due to their low RNA content after multiple rounds of DNase treatment. Con-
sequently, less than one million *P. falciparum* reads were obtained for each of these samples.
Therefore, samples from patient #1 and #2 were omitted from assembly due to low coverage,
but included in the differential gene expression analysis.
Variation in parasite ages in the different patient samples was analyzed with a mixture model
in accordance to Tonkin-Hill *et al.* (Tonkin-Hill *et al*, 2018) using published data from López-
Barragán *et al.* as a reference (López-Barragán *et al*, 2011). Parasites from first-time infected
and pre-exposed patients revealed no obvious difference in the proportion of the different par-
asite stages (Figure 2A) or in median age (Figure 2B, Table 2). However, a small bias towards
younger parasites in the severe cases was observed with a median age of 8.2 hpi (IQR: 8.0–
9.8) in comparison to 9.8 hpi in the non-severe cases (IQR: 8.2–11.4) (Figure 2C,D, Table 2).
None of the samples revealed high proportions of late trophozoites (all <3%), schizonts (0%)
or gametocytes (all <6%) (Figure 2A, C).

Differential* var *gene expression on gene, domain and homology block level

To correlate individual *var* genes with a naïve immune status or disease severity differential

var gene expression was analyzed according to Tonkin-Hill *et al.* (Tonkin-Hill *et al*, 2018). First,

var gene assembly was performed by assembling each sample separately, which reduces the

risk for generating false chimeric genes and results in longer contigs compared to a combined

all sample assembly approach. In total, 6,441 contigs with over 500 bp-length were generated

with an N50 of 2,302 bp and a maximum length of 10,412 bp (Data S1). For 5,488 contigs

*Pf*EMP1 domains could be annotated, the remaining contain only homology blocks defined by

Rask *et al.* (Rask *et al*, 2010) (Table S3, S4).

Var allele level:

A median of 200 contigs (IQR: 137–279) with >500 bp was assembled per sample. Almost half

of the transcripts, for which a *Pf*EMP1 domain annotation could be made (47.2%: 2,592),

showed >98% nucleotide identity for at least 80% of the length of a *var* gene in the varDB

(Otto, 2019). Moreover, 203 transcripts matched with over 1 kb overlap and 98% identity to

var genes from the 15 reference genomes (Otto *et al*, 2018a) (Table S4). The Salmon RNA-

seq quantification pipeline, which identifies equivalence classes allowing reads to contribute

to the expression estimates of multiple transcripts, was used to estimate expression levels for

each transcript. As a result, it accounts for the redundancy present in our whole set of *var* gene

contigs from all separate sample-specific assemblies. We compared this approach with Corset

which previously has been used to investigate differential expression of *var* genes in severe

malaria and found it gave similar results (see methods) (Tonkin-Hill *et al*, 2018). Due to the

high diversity in *var* genes both of these approaches are only able to identify significant asso-

ciations between transcripts and phenotypes when there is sufficient similarity within the as-

sociated sequences.

By comparing transcripts expressed in first-time infected patients with those from parasites

isolated from pre-exposed patients using the Salmon approach, transcript levels were higher

for twelve and lower for three genes in malaria-naïve hosts (Table 3, Figure 3A, B, Table S6).

Assembled alleles of the conserved subfamilies, *var1*, *var2csa* and *var3*, were also found at

higher frequencies in first-time infected patients. Notably, the *var1-IT* allele was expressed at

higher levels in parasites from first-time infected patients, whereas the *var1-3D7* allele was

expressed at higher levels in parasites from pre-exposed and non-severe patients (Figure 3,

Table 3, 4). This was confirmed by mapping normalized reads from all patients to the *var1-*

3D7 and *var1-IT* allele forms (Figure S2). Several *var* fragments from B- or C-type *var* genes

were associated with a naïve immune status and three transcripts from A, DC8 and B-type *var*

genes as well as *var2csa* were linked to severe malaria patients (Figure 3, Table 3, 4, Table

S6, S7).

Domain level:

Similar to Tonkin-Hill *et al.* (Tonkin-Hill *et al*, 2018), we quantified domain level expression by aligning reads to their single best 'hit' in the combined assembly. Domains were identified within each transcript and the sum of the read counts corresponding to domains with the same classification was calculated to provide domain level read counts. This showed that different EPCR-binding CIDRα1 domain variants and other domains found in DCs with CIDRα1 domains were expressed at significantly higher levels in first-time infected patients (Table 5, Figure 4A, B, Table S6). Thus, besides domains from DC8 (DBLα2, CIDRα1.1, DBLβ12) and DC13 (DBLα1.7, CIDRα1.4), a single domain from DC15 (DBLα1.2) were increased upon infection of malaria-naïve individuals. The DBLα1.2 domain was in all of the 32 gene assemblies with adjacent CIDR domains annotated linked to an EPCR binding CIDRα1 domain, 56% of these were a CIDRα1.5 domain (Table S3, S4). The CIDRα1.6 domain from DC4 failed to reach statistical significance (p_{adj}=0.07), but was 2.25x times higher expressed in parasites infecting naïve patients (Table S6). In addition to domains associated with EPCR-binding *Pf*EMP1, parasites from first-time infected patients showed significantly more transcripts encoding the CIDRδ1 domain of DC16 (DBLα1.5/6-CIDRδ1/2) (Table 5, Figure 4). In general, all domains associated with the same domain cassettes showed the same trend even if some domains failed to reach statistical significance (Table S6). Moreover, the DBLβ6 domain was among the top hits of significantly higher expressed domains in the naïve patient cluster. The DBLβ6 is associated with A-type *var* genes and often found adjacent to DC15 and DC16 (Otto *et al*, 2019) thus supporting the association of both domain cassettes with a naïve immune status. Domain types found expressed at lower levels in malaria-naïve included several domains from the *var1-3D7* allele (DBLα1.4, DBLγ15, DBLε5) as well as NTSB and N-terminal head domains from B- and C-type *Pf*EMP1 (DBLα0.13/22/23) with CD36-binding CIDR domains (CIDRα2.8/9,6) and the C-terminal CIDRγ11 domain (Table 5, Figure 4A, B).

When comparing the severe sample set to the non-severe, domains of DC8 (DBLα2, CIDRα1.1, DBLβ12) and DC15 (DBLα1.2) were found associated with severe disease (Table 6, Figure 4C, D, Table S7). The DC16 consists either of a DBLα1.5 or 1.6 attached to a CIDRδ1 domain. The DBLα1.6 domain was found expressed at higher levels in severe malaria patients whereas the DBLα1.5 domain was found to be highly expressed in non-severe cases. The CIDRδ1 domain showed no association with disease group, and the DBLα1.5 domain type was generally expressed at a very high level in multiple patient isolates (Table 6, Figure 4). As observed for pre-exposed individuals, domain types expressed at significantly higher levels in non-severe cases included the CIDRα1.3 domain from the *var1-3D7* allele as well as N-terminal head domains from B- and C-type *Pf*EMP1 with CD36-binding capacity (DBLα0.23, CIDRα2.4/9)(Table 6, Figure 4C, D).

Since the subclassification of domains is debatable (Otto *et al*, 2019) and different domain

subclasses confer the same binding phenotype (Lau *et al*, 2015), the domains of the N-terminal

head structure were grouped according to their binding phenotype and the normalized read

counts (TPM) were summarized for each patient (Figure 4E, F). This showed clear differences

were observed for DBL and CIDR domains associated with EPCR- or CD36-binding *Pf*EMP1.

As expected, the EPCR-binding phenotype as well as the CIDRγ3 domain were associated

with the naïve and more severe cases, the CD36-binding phenotype with the pre-exposed and

non-severe patients.

Homology block level:

Within *Pf*EMP1 sequences 628 homology blocks were defined (Rask *et al*, 2010) and 613 were

available for download and subsequent analysis from the VARDOM server (Tonkin-Hill *et al*,

2018). Homology block expression levels were obtained by aggregating read counts for each

block after first identifying all occurrences of the block within the combined transcript assembly.

Transcripts encoding blocks number 255, 584 and 614, all typically located within DBLβ do-

mains of DC8 and CIDRα1-containing type A *Pf*EMP1 (Table 7, Figure 5A, B, Table S6), num-

ber 557, located in the inter-domain region between DBLβ and a DBLγ domains (no *Pf*EMP1

type association) and block number 155 found in NTSA, were found associated with a naïve

immune status. All these blocks are found in A-, B/A- or B-type genes. Conversely, transcripts

encoding block 88 from DBLα0 domains and 269 from ATSB were found at lower levels in

malaria-naïve patients indicating that B- and C-type genes are more frequently expressed in

pre-exposed individuals (Table 7, Figure 5A, B). Two homology blocks, 591 and 559, associ-

ated with B-type *Pf*EMP1 were found to be lower expressed in severe malaria cases (Table 8,

Figure 5C, D, Table S7).

A summary of the differential *var* gene expression data on the multi-, single- and subdomain

level can be found in Figure 6.

Var expression profiling by DBLα-tag sequencing

To supplement our RNA-seq analysis with an orthogonal analysis, we performed DBLα-tag

RT-PCR combined with deep amplicon sequencing on 30 of our patient samples (Lavstsen *et*

al, 2012). 851 to 3,368 reads with a median of 1,666 over all samples were analyzed. Identical

DBLα-tag sequences were clustered to generate relative expression levels of each unique *var*

gene tag. Overall, the relative expression levels were similar for sequences found in both the

RNA-seq and the DBLα-tag approach with a mean log2(DBLα-PCR/RNA-seq) of 0.4 (CI of

95%: -2.5–3.3) determined by Bland-Altman plotting (Figure S3). In median, 82.6% of all de-

tected individual DBLα-tag sequence clusters with >10 reads (92.9% of all DBLα-tag reads)

were found in the RNA-seq approach; and 81.8% of the upper 75th percentile of RNA-seq
contigs (with DBLα tag sequences) were found by the DBLα-tag approach.
Unique DBLα-tag sequences were searched for near identical sequences among all known
var genes on varDB Version V1.1 (Otto, 2019) using Varia tool (Mackenzie *et al*, 2020). Nearly
identical database sequences were found for ~85% of the DBLα-tag sequences allowing pre-
diction of these query gene's domain annotation (Table S10). In line with the RNA-seq data
we found DBLα1 and DBLα2 sequences enriched in first-time infected patients and the severe
malaria patients. Conversely, a significant higher proportion of DBLα0 sequences was found
in pre-exposed individuals and mild cases (Figure 7A, B). No difference was observed in the
number of reads or unique DBLα-tags detected between patient groups, although a trend to-
wards more DBLα-tag clusters could be observed in first-time infected patients and severe
cases (Figure 7A, B). A prediction of the NTS and CIDR domains adjacent to the DBLα domain
showed a significant higher proportion of NTSA in severe cases as well as EPCR-binding
CIDRα1 domains in first-time infected and severe cases. Expression of *var* genes encoding
NTSB and CIDRα2-6 domains were significantly associated with in pre-exposed and non-se-
vere cases (Figure 7A, B). Analysis of *var* expression in relation to other domains, showed *var*
transcripts with DBLβ, γ and ζ or CIDRγ domains were more frequently expressed in first-time
infected and severe malaria patients whereas those encoding DBLδ and CIDRβ were less
frequent (Figure 7C, D).
Assessing expression in relation to domain subtype, CIDRα1.1/5, DBLβ12, DBLγ2/12 and
DBLα2 were associated with severe malaria whereas CIDRα3.1/4, DBLα0.12/16, and DBLδ1
associated with non-severe cases (Figure 7D). Overall, these data corroborated the main ob-
servations from the RNA-seq analysis, confirming the association of EPCR-binding *Pf*EMP1
variants with development of severe malaria symptoms and CD36-binding *Pf*EMP1 variants
with establishment of less severe infections in semi-immune individuals.
349
Correlation of* var *gene expression with antibody levels against head structure CIDR
domains
A detailed analysis of the antibody repertoire of the patients against head structure CIDR do-
mains of *Pf*EMP1 was carried out using a panel of 19 different EPCR-binding CIDRα1 do-
mains, 12 CD36-binding CIDRα2–6 domains, three CIDRδ1 domains as well as a single
CIDRγ3 domain (Obeng-Adjei *et al*, 2020; Bachmann *et al*, 2019). Additionally, the minimal
binding region of VAR2CSA was included (Figure 8). Generally, plasma samples from malaria-
naïve as well as severe cases showed lower MFI values for all antigens tested in comparison
to samples from pre-exposed or non-severe cases (Figure 8A, B). Mann-Whitney U testing
revealed significant differences for CIDRα2–6, CIDRδ1 and CIDRγ3, but not for EPCR-binding
CIDRα1 domains.

Another way of analyzing the samples is to take the average MFI from an unrelated control
cohort plus two standard deviations as a cut off for seropositivity for calculation of the coverage
of antigen recognition (Cham *et al*, 2010). By doing so, almost half of the tested antigens were
recognized by pre-exposed (median: 46.3%) and non-severe patients (median: 45.1%), but
only 1/4 of the antigens were recognized by first-time infected patients (median: 24.4%) and
1/20 by severely ill patients (median: 4.9%). Apart from controls, antigens recognized by over
60% of the pre-exposed and/or non-severe patient sera were i) four CIDRα1 domains capable
of EPCR-binding (CIDRα1.5, CIDRα1.6, CIDRα1.7 and the DC8 domain CIDRα1.8), ii) two
CD36-binding CIDRα domains (CIDRα2.10, CIDRα3.1) and iii) two CIDR domains with un-
known binding phenotype (CIDRδ1 and CIDRγ3) (Figure 8C, Table S1).

Differential gene expression analysis of core genes
Global gene expression analysis (Tonkin-Hill *et al*, 2018) identified 420 genes to be higher and
236 to be lower expressed in first-time infected patients, together corresponding to 11.3% of
all *P. falciparum* genes (Table S9). Analysis of gene set enrichment analysis (GSEA) of GO
terms and KEGG pathways showed a significantly lower expression level for genes with sev-
eral GO terms involved in antigenic variation and host cell remodeling in first-time infected
patients (Figure 9A, Table S9). These analysis results may be distorted, since variant surface
antigens like *var*, *rif*, *stevor*, *surf* and *pfmc-2tm* are largely clone-specific and reads from the
clinical isolates would map only to homologous regions in 3D7 genes, which may actually not
be present in the clinical isolate. Therefore, we manually screened differentially expressed
genes known to be involved in *var* gene regulation or correct display of *Pf*EMP1 at the host
cell surface (Table S9). On single gene level, genes encoding the Maurer's cleft proteins
MAHRP1 and REX2 (Spycher *et al*, 2003; Spielmann *et al*, 2006), REX4 and MSRP7 located
within the host cell cytosol (Spielmann *et al*, 2006; Heiber *et al*, 2013), the erythrocyte mem-
brane-located glycophorin binding protein GBP130 (Perkins, 1988) as well as Sir2a and SWIB
known to be involved in *var* gene regulation (Tonkin *et al*, 2009; Wang & Zhang, 2020) were
expressed at significantly lower levels in malaria-naïve patients. MAHRP1 is essential for
translocation of *Pf*EMP1 to the surface of infected erythrocytes (Spycher *et al*, 2006, 2008)
and was suggested to be part of the *Pf*EMP1 loading hub (McHugh *et al*, 2020); on the con-
trary, deletion of the REX2 and REX4-encoding genes via chromosome breakage was asso-
ciated with the loss of cytoadherence, but not with aberrant trafficking of *Pf*EMP1 (Nacer *et al*,
2011; Chaiyaroj *et al*, 1994; Day *et al*, 1993). Genetic ablation of GBP130 increased the mem-
brane rigidity of infected erythrocytes without negative impact on cytoadherence under flow
conditions (Maier *et al*, 2008).
In contrast, several PHIST encoding genes were found expressed at higher levels in first-time
infected patients including the lysine-rich membrane-associated PHISTb protein (LyMP),

previously reported to interact with the ATS domain (Oberli *et al*, 2014, 2016). Furthermore,

the exported proteins FIKK9.6 (Nunes *et al*, 2007), MSRP5 and MSRP6 (Heiber *et al*, 2013)

as well as PF3D7_0721100 showed a significant increase compared to pre-exposed patients.

The conserved *Plasmodium* protein of unknown function PF3D7_0721100 detected in deter-

gent-resistant fractions of the red blood cell membrane (Yam *et al*, 2013) was also found in a

putative *Pf*EMP1 unloading hub together with REX1, MAHRP2, PTP5 and PF3D7_1353100

using protein interaction network analysis (McHugh *et al*, 2020). On the level of *var* gene ex-

pression regulation, the chromatin-associated exoribonuclease *Pf*RNase II (PF3D7_0906000)

was increased in first-time infected patients (Zhang *et al*, 2014).

Most of the associations with a naïve immune status were also observed in the comparison of

severe versus non-severe cases (Table S10). Significantly higher expresses genes included

several PHIST and HYP proteins, MSRP5, PfRNase II, PF3D7_0721100 and, additionally,

SET1, PTP1, SBP1, PTP5 and PF3D7_1353100. Overall, three proteins defined by McHugh

et al. as members of the *Pf*EMP1 unloading hub were significantly higher expressed in severe

malaria cases (PTP5, PF3D7_0721100 and PF3D7_1353100) and all remaining proteins

showed the same trend for upregulation (PIESP2, PfJ23, REX1 and MAHRP2) (McHugh *et al*,

2020).

Furthermore, the KEGG pathway 03410 'base excision repair' facilitating the maintenance of

the genome integrity by repairing small bases lesions in the DNA was expressed at significantly

higher levels in first-time infected patient samples (Figure 9A, Figure S4). In total, six out of 15

P. falciparum genes included into the KEGG pathway were found to be statistically significant

enriched upon first-time infection, more precisely the putative endonuclease III

(PF3D7_0614800) from the short-patch pathway and the putative A-/G-specific adenine gly-

cosylase (PF3D7_1129500), the putative apurinic/apyrimidinic endonuclease Apn1

(PF3D7_1332600), the proliferating cell nuclear antigens 1 (PF3D7_1361900), the catalytic

(PF3D7_1017000) and small (PF3D7_0308000) subunits from the DNA polymerase delta from

the long-patch pathway (Figure 9B).

Discussion

Analysis of blood samples from travelers returning to Germany and hospitalized with *P. falci-*

parum malaria for the first time, allows studies of parasites' development and gene expression

unaffected by host immune responses elicited in previous Plasmodia infections. Here we an-

alyzed gene expression in 32 such patients using direct *ex vivo* RNA-seq to exclude transcrip-

tional adaptation through *in vitro* cultivation. Most of these returning travelers studied were

infected with a single or very few parasite genotypes, which likely simplified the analysis of *var*

gene expression due to the restricted genomic repertoire. We were able to distinguish

retrospectively between first-time infected and pre-exposed patients by medical history and
assessing the presence of antibodies to *P. falciparum* antigens. Eight out of 15 first-time in-
fected patients were classified as severe malaria cases due to impaired organ function accord-
ing to their medical reports. Despite the relatively low number of patient samples, the RNA-
seq approach confirmed previously reported associations between transcripts encoding type
A and B EPCR-binding *Pf*EMP1 and infections in naïve hosts and disease severity (Table S6,
S7)(Duffy *et al*, 2019; Tonkin-Hill *et al*, 2018; Kessler *et al*, 2017; Bernabeu *et al*, 2016;
Jespersen *et al*, 2016; Lavstsen *et al*, 2012).

Overall, there was a high degree of consensus between analyses of the *var* transcriptome data
in relation to the different levels of *Pf*EMP1 domain annotation. Stratifying *var* gene expression
according to different main and subtype of DBL and CIDR domains, showed only A- and DC8-
type *Pf*EMP1 domains, and predominantly those linked to EPCR-binding *Pf*EMP1, to be asso-
ciated with first-time infections. Conversely, domains typical for CD36-binding *Pf*EMP1 pro-
teins was found at higher levels in malaria-experienced patients. Specifically, expression of
*Pf*EMP1 domains included in DC8, DC13 and DC15 as well as all EPCR-binding CIDRα1 do-
mains were associated with first time infections, whereas DBLα0 and CD36-binding CIDRα2-
6 domains were linked to pre-exposed individuals. These differences were in large due to the
differential expression between the first-time infected patients with more severe symptoms and
patients with non-severe malaria. Here, domains of DC8 and DC15 as well as all DBLα1/2 and
CIDRα1 domains were associated with severe symptoms, while NTSB, DBLα0, CIDRα2-6 do-
mains including specific subsets of CIDRα2 were linked to non-severe symptoms. These con-
clusions were closely mirrored in the DBLα tag analysis, and was further corroborated by the
differential RNA-seq expression stratified according to the smaller homology blocks, which
identified mainly homology blocks of DBLβ1, 3, 5 and 12 DBLβ domains to be associated with
first- time infected patients. These DBLβ domains are parts of DCs associated with EPCR-
binding, so it is hard to distinguish between co-occurring domains and clear associations.
DBLβ domains do not segregate distinctly by sequence similarity into groups reflecting their
observed binding to ICAM-1 and gC1qR, (Otto *et al*, 2019). Best defined is ICAM-1 binding
DBLβ5 domains found in CD36-binding B-type *Pf*EMP1 (Janes *et al*, 2011; Lennartz *et al*,
2019) and ICAM1-binding of and DBLβ1 and 3 domains found in EPCR-binding type-A
*Pf*EMP1 (Lennartz *et al*, 2017). The relative importance of ICAM1 binding to CD36 or EPCR
binding *Pf*EMP1 is not well understood, but ICAM-1-binding is believed to contribute to disease
severity by either tethering endothelial binding (Bernabeu *et al*, 2019) or initiating or securing
endothelial sequestration on inflamed endothelium, which is likely to shed EPCR (Jensen *et*
al, 2020)

In addition, three other A-type-associated domains, CIDRγ3, CIDRδ from the DC16 and
DBLβ9 from DC5, were found associated with first-time infected patients. DC5 could have

been detected due to its presence C-terminally to some EPCR-binding PfEMP1. However, the

CIDRδ domain of DC16 (DBLα1.5/6-CIDRδ1/2) constitute a different subset of A-type PfEMP1,

which together with A-type PfEMP1 with CIDRβ2 (found in DC11) or CIDRγ3 domains may be

associated with rosetting (Carlson et al, 1990; Ghumra et al, 2012). Direct evidence that any

of these CIDR domains have intrinsic rosetting properties is lacking (Rowe et al, 2002). Rather,

their association with rosetting may be related to their tandem expression with DBLα1 at the

N-terminal head (Ghumra et al, 2012). The CIDRδ domain was not associated with severe

malaria patient group and the two different DC16-associated DBLα1 domains were found as-

sociated with severe and non-severe malaria, respectively. However, CIDRγ3 expression was

low, but it was found at higher levels in severe malaria patients.

The DC16 group A signature was not associated with severe disease outcome in previous

DBLα-tag studies or qPCR studies by Lavstsen et al. (Lavstsen et al, 2012) and Bernabeu et

al. (Bernabeu et al, 2016), but DBLα1.5/6 and CIDRδ of DC16 were enriched in cerebral ma-

laria cases with retinopathy in the study of Shabani et al. (Shabani et al, 2017) and Kessler et

al. (Kessler et al, 2017) using the same qPCR primer set. Also, association of DC11 with se-

vere malaria in Indonesia was found using the same RNA-seq approach as used here (Tonkin-

Hill et al, 2018). Rosetting is thought to enhance microvascular obstruction but the role of ro-

setting in severe malaria pathogenesis remains unclear (McQuaid & Rowe, 2020). Together

with previous observations, our data suggest that pediatric cerebral malaria infections are dom-

inated by the expansion of parasites expressing EPCR-binding domains accompanied by par-

asites expressing other group A PfEMP1, possibly rosetting variants.

To the best of our knowledge, this study is the first description of expression differences be-

tween the two var1 alleles, 3D7 and IT. At the transcript level the var1-IT allele was found to

be enriched in parasites from first-time infected patients; conversely, several transcripts cov-

ering almost the full-length protein and in total half of the domains from the var1-3D7 allele

were increased in pre-exposed and non-severely ill patients. Expression of the var1 gene was

previously observed to be elevated in malaria cases imported to France with an uncomplicated

disease phenotype (Argy et al, 2017). In general, the var1 subfamily is ubiquitously transcribed

(Winter et al, 2003; Duffy et al, 2006), atypically late in the cell cycle after transcription of var

genes encoding the adhesion phenotype (Kyes et al, 2003; Duffy et al, 2002) and is annotated

as a pseudogene in 3D7 due to its premature truncation. Similarly, numerous isolates display

frame-shift mutations often in exon 2 in the full gene sequences (Rask et al, 2010). However,

none of these studies addressed differences in the two var1-alleles just recently described by

comparing var gene sequences from 714 P. falciparum genomes (Otto et al, 2019) and to date

it is unclear if both allele forms fulfill the same function or harbor the same characteristics

previously described. Overall, the var1 gene – and the first 3.2 kb of the 3D7 allele in particular

– seems to be under high evolutionary pressure (Otto et al, 2019) and the bi-allelic pattern can

509 be traced back before the split of *P. reichenowi* from *P. praefalciparum* and *P. falciparum* (Otto

510 *et al*, 2018b). Our data indicate that the two alleles, 3D7 and IT, may have different roles during

511 disease, however, this remains to be determined in future studies.

512 In general, data from immunologically naïve malaria patients are rather limited, restricting our

513 comparison mainly to the severe phenotype described in numerous previous studies. How-

514 ever, a recent study explored the *var* gene expression during infancy in Kenyan children and

515 could correlate the waning of maternal antibodies with increasing transcription of DC8, DC13

516 and A-type *var* genes in general. After the first year of life the amount of these transcripts

517 decreases with age and acquired immunity (Kivisi *et al*, 2019). A high expression of A-type *var*

518 genes in naïve malaria cases imported to France and an association of DC4, 8 and 13 with

519 disease severity has also been reported (Argy *et al*, 2017). Both qPCR studies are in agree-

520 ment with the RNA-seq data from our cohort of immunologically malaria-naïve adults, but we

521 could extend the list with DC15 and DC16, that are presumably involved in binding of infected

522 erythrocytes to EPCR and that may also mediate binding to uninfected red blood cells by ro-

523 setting.

524 Overall, most studies – including this one – are looking at differentially expressed genes. From

525 CHMI studies we already know that at the early onset of infection the parasite population ex-

526 presses a wide range of different *var* genes located in the subtelomeric regions (A- and mainly

527 B-type *var* genes). Since B-types are also highly expressed in pre-exposed cases domains of

528 these *Pf*EMP1s may be missing in pattern of first-time infected patients. Furthermore, EPCR-

529 and ICAM-1 binding and rosetting-mediating variants may confer a parasite growth advantage

530 in malaria naïve hosts, and in some circumstances increase the risk for severe malaria, so that

531 a selection towards these variants may have been already occurred in our patients. This would

532 fit nicely to the observation that the severe patients within the malaria-naïve patient group

533 seem to be infected for a longer time period. Moreover, patients with preformed immunity rec-

534 ognize several CIDRα1 domains capable of EPCR-binding as well as the two atypical CIDR

535 head domains CIDRδ1 and CIDRγ3 more frequently than CIDRα2-6 domains with CD36-

536 binding affinity. This is in agreement with studies from malaria endemic setting indicating that

537 IgG against EPCR-binding domains were acquired first followed by domains with unknown

538 binding phenotypes associated with rosetting and CD36-binding domains. Resulting in the ear-

539 lier acquisition of antibodies against DBL and CIDR domains of group A and B/A, associated

540 with EPCR binding, than against B- and C-type domains (Obeng-Adjei *et al*, 2020; Cham *et al*,

541 2009, 2010; Turner *et al*, 2015).

542 Genes involved in *Pf*EMP1 biology were found to be expressed at lower levels in severe ma-

543 laria patients (Tonkin-Hill *et al*, 2018). In concordance, GO term analysis revealed a general

544 lower expression of genes involved in antigenic variation or host cell remodeling during first-

545 time infection and severe disease, but the expression analysis of clonally variant genes is

complicated by the existence of multiple diverse families that have strain specific members
leading to mis-mapping of reads to genes present in the 3D7 strain that do not exist in the
clinical samples analyzed. Therefore, manually selected genes encoding regulators of *var*
gene expression and *Pf*EMP1 trafficking were additionally inspected for association with a na-
ïve immune status and severity. The NAD$^+$-dependent histone deacetylases Sir2a and Sir2b
remove acetyl groups from the N-terminal tails of histone 3 and 4 and are therefore considered
as *var* silencing factors. Sir2b regulates the most telomeric B-type *var* genes, Sir2a-regulated
var genes are of type A, C and E by deacetylation of H3K9ac, H3K14ac and H4K16ac (Duffy
et al, 2014; Tonkin *et al*, 2009). A down-regulation of Sir2a indicates an elevated expression
of these *var* types, which is in concordance with our data on A- and E-type *var* genes found
more frequently in first-time infected patients. Contradictory to a previous study showing that
the exoribonuclease PfRNase II controls the silencing of A-type *var* genes and was negatively
associated with *var*-A expression in severe malaria, PfRNase II expression was significantly
enriched in first-time infected and severely ill patients (Zhang *et al*, 2014). Our results also
contradicted down-regulation of A- and partially B-type *var* genes by conditional knockout of
SWIB (Wang & Zhang, 2020). However, high expression levels of the *var1*-3D7 allele, also an
A-type *var* gene, in pre-exposed and non-severe cases may be responsible for these opposite
results. On the other hand, expression of genes encoding other factors was also found inverted
in comparison to other studies. GBP was found to be highly enriched in severe relative to
uncomplicated malaria cases (Lee *et al*, 2018), but is significantly lower expressed in first-time
infected patients and has the same trend in severe cases in this study. Contrarily, LyMP, SBP1
and SET proteins were expressed on a significantly lower level in severe cases in the study of
Tonkin-Hill *et al.* (Tonkin-Hill *et al*, 2018), but elevated levels were found in our study.
A recent study showed that parasites isolated from symptomatic infections were on average
younger than blood-circulating parasites from asymptomatic infections presumably due to a
more efficient sequestration of parasites in the symptomatic cases (Andrade *et al*, 2020).
Based on this and our findings we hypothesize that parasites circulating in severely ill patients
are younger due to the expression of EPCR-/ICAM-1-binding *Pf*EMP1 variants, whereas par-
asites in non-severe patients may circulate longer due to their expression of *Pf*EMP1 binding
to CD36. Although affinity of the CIDR domains to each of these receptors was shown to be
similar with median K_d values of 12 nM for CD36 (Hsieh *et al*, 2016) and 16.5 nM for EPCR
(Lau *et al*, 2015), recent papers describe a rolling binding phenotype of infected red blood cells
over CD36 and a static binding for EPCR and ICAM-1 under flow conditions (Lubiana *et al*,
2020; Bernabeu *et al*, 2019; Dasanna *et al*, 2017; Herricks *et al*, 2013). Due to their biconcave
shape trophozoites seem to roll faster but less stable by flipping over CD36-expressing cells,
whereas schizonts roll over longer distances at different shear stresses applied. This might
explain why young trophozoites are found in blood samples from non-severe cases, but not

583 older trophozoites or schizont stages as also previously described (Tonkin-Hill *et al*, 2018).

584 The parasite population in first-time infected individuals may have broader binding potential

585 after liver release as there is no pre-existing immunity to clear previously experienced *Pf*EMP1

586 variants, but during the blood stage infection variants with high-affinity binding to EPCR and

587 ICAM-1 binding are selected which may lead to severe symptoms. This hypothesis is sup-

588 ported by (1) the difference in parasite age between severe and non-severe malaria cases

589 calculated by matching RNA-seq data to a reference data set (López-Barragán *et al*, 2011).

590 (2) This correlates with a higher *var1* expression in parasite from non-severe and pre-exposed

591 patients, which expression is not suppressed in trophozoites (Kyes *et al*, 2003). (3) The ex-

592 pression of EPCR- & ICAM-1-binding variants in parasites from severe and first-time infected

593 patients is significantly increased, conversely transcripts of CD36-binding variants are found

594 more frequently in parasites from non-severe and pre-exposed patients. For parasites survival

595 and transmission, it may be highly beneficial to have more less virulent *Pf*EMP1 variants able

596 to bind CD36. This interaction may not, or is less likely to, lead to obstruction of blood flow,

597 inflammation and organ failure at least of the brain, where CD36 is nearly absent.

599 **Material and methods**

601 *Ethics statement*

602 The study was conducted according to the principles of the Declaration of Helsinki in its 6th

603 revision as well as International Conference on Harmonization–Good Clinical Practice (ICH-

604 GCP) guidelines. All patients, aged 19 to 70 years, provided written informed consent for this

605 study, which was approved by the Ethical Review Board of the Medical Association of Ham-

606 burg (reference number PV3828).

608 *Blood sampling and processing*

609 Blood samples from *P. falciparum* malaria patients collected either at the diagnostic unit of the

610 Bernhard Nocht Institute for Tropical Medicine, Hamburg, Germany, or at the Medical Clinic

611 and Polyclinic of University Clinic Hamburg-Eppendorf, Hamburg, Germany, were used in this

612 study. EDTA blood samples (1–30 mL) were obtained from the patients. The plasma was sep-

613 arated by centrifugation and immediately stored at -20°C. Erythrocytes were isolated by Ficoll

614 gradient centrifugation followed by filtration through Plasmodipur filters (EuroProxima) to clear

615 the remaining granulocytes. An aliquot of red blood cells (about 50–100 µl) was separated and

616 further processed for gDNA purification. At least 400 µl of purified erythrocytes were rapidly

617 lysed in 5 volumes pre-warmed TRIzol (ThermoFisher Scientific) and stored at -80°C until fur-

618 ther processing (*ex vivo* samples, n = 32).

Serological assays

Luminex assay

The Luminex assay was conducted as previously described using the same plex of antigens tested (Bachmann *et al*, 2019). In brief, plasma samples from patients were screened for individual recognition of 19 different CIDRα1, 12 CIDRα2–6, three CIDRδ1 domains and a single CIDRγ3 domain as well as of the controls AMA1, MSP1, CSP, VAR2CSA (VAR2), tetanus toxin (TetTox) and BSA. The data are shown as mean fluorescence intensities (MFI) allowing comparison between different plasma samples, but not between different antigens. Alternatively, the breadth of antibody recognition (%) was calculated using MFI values from Danish controls plus two standard deviations (SD) as cut off.

Merozoite-triggered antibody-dependent respiratory burst (mADRB)

The assay to determine the mADRB activity of the patients was set up as described before (Kapelski *et al*, 2014). Polymorphonuclear neutrophil granulocytes (PMNs) from one healthy volunteer were isolated by a combination of dextran-sedimentation and Ficoll-gradient centrifugation. Meanwhile, 1.25×10^6 merozoites were incubated with 50 μl of 1:5 diluted plasma (decomplemented) for 2 h. The opsonized merozoites were pelleted (20 min, 1500 g), re-suspended in 25 μl HBSS and then transferred to a previously blocked well of an opaque 96 well high-binding plate (Greiner Bio-One). Chemiluminescence was detected in HBSS using 83.3 μM luminol and 1.5×10^5 PMNs at 37°C for 1 h to characterize the PMN response, with readings taken at 2 min intervals using a multiplate reader (CLARIOstar, BMG Labtech). PMNs were added in the dark, immediately before readings were initiated.

ELISA

Antibody reactivity against parasitophorous vacuolar membrane-enclosed merozoite structures (PEMS) was estimated by ELISA. PEMS were isolated as described before (Llewellyn *et al*, 2015). For the ELISA, 0.625×10^5 PEMS were coated on the ELISA plates in PBS. Plates were blocked using 1% Casein (Thermo Scientific #37528) and incubated for 2 h at 37°C. After washing using PBS/0.1% Tween, plasma samples were added at two-fold dilutions of 1:200 to 1:12800 in PBS/0.1% Casein. The samples were incubated for 2 h at room temperature (RT). IgG was quantified using HRP-conjugated goat anti-human IgG at a dilution of 1:20,000 and incubated for 1 h. For the color reaction, 50 μl of TMB substrate was used and stopped by adding 1 M HCl after 20 min. Absorbance was quantified at 450 nm using a multiplate reader (CLARIOstar, BMG Labtech).

Protein microarray

Microarrays were produced at the University of California Irvine, Irvine, California, USA (Doolan *et al*, 2008). In total, 262 *P. falciparum* proteins representing 228 unique antigens were expressed using an *E. coli* lysate *in vitro* expression system and spotted on a 16-pad ONCYTE AVID slide. The selected *P. falciparum* antigens are known to frequently provide a positive signal when tested with sera from individuals with sterile and naturally acquired immunity against the parasite (Obiero *et al*, 2019; Dent *et al*, 2016; Doolan *et al*, 2008; Felgner *et al*, 2013). For the detection of binding antibodies, secondary IgG antibody (goat anti-human IgG QDot™800, Grace Bio-Labs #110635), secondary IgM antibody (biotin-SP-conjugated goat anti-human IgM, Jackson ImmunoResearch #109-065-043) and Qdot™585 Streptavidin Conjugate (Invitrogen #Q10111MP) were used (Taghavian *et al*, 2018).

Study serum samples as well as the European control serum were diluted 1:50 in 0.05X Super G Blocking Buffer (Grace Bio-Labs, Inc.) containing 10% *E. coli* lysate (GenScript, Piscataway, NJ) and incubated for 30 minutes on a shaker at RT. Meanwhile, microarray slides were rehydrated using 0.05X Super G Blocking buffer at RT. Rehydration buffer was subsequently removed and samples added onto the slides. Arrays were incubated overnight at 4°C on a shaker (180 rpm). Serum samples were removed the following day and microarrays were washed using 1X TBST buffer (Grace Bio-Labs, Inc.). Secondary antibodies were then applied at a dilution of 1:200 and incubated for two hours at RT on the shaker, followed by another washing step and a one-hour incubation in a 1:250 dilution of Qdot™585 Streptavidin Conjugate. After a final washing step, slides were dried by centrifugation at 500 g for 10 minutes. Slide images were taken using the ArrayCAM® Imaging System (Grace Bio-Labs) and the ArrayCAM 400-S Microarray Imager Software.

Microarray data were analyzed in R statistical software package version 3.6.2. All images were manually checked for any noise signal. Each antigen spot signal was corrected for local background reactivity by applying a normal-exponential convolution model (McGee & Chen, 2006) using the RMA-75 algorithm for parameter estimation (available in the LIMMA package v3.28.14) (Silver *et al*, 2009). Data was log2-transformed and further normalized by subtraction of the median signal intensity of mock expression spots on the particular array to correct for background activity of antibodies binding to *E. coli* lysate. After log2 transformation data approached normal distribution. Differential antibody levels (protein array signal) in the different patient groups were determined by Welch-corrected Student's t-test. Antigens with $p < 0.05$ and a fold change >2 of mean signal intensities were defined as differentially recognized between the tested sample groups. Volcano plots were generated using the PAA package (Turewicz *et al*, 2016) and GraphPad Prism 8. Individual antibody breadths were defined as number of seropositive features with signal intensities exceeding an antigen-specific threshold set at six standard deviations above the mean intensity in negative control samples.

Unsupervised random forest model

An unsupervised random forest (RF) model, a machine learning method based on multiple classification and regression trees, was calculated to estimate proximity between patients. Variable importance was calculated, which shows the decrease in prediction accuracy if values of a variable are permuted randomly. The *k*-medoids clustering method was applied on the proximity matrix to group patients according to their serological profile. Input data for random forest were Luminex measurements for MSP1, AMA1 and CSP reduced by principal component analysis (PCA; first principal component selected), mADRB, ELISA, and antibody breadth of IgG and IgM determined by protein microarray were used to fit the RF model. Multidimensional scaling was used to display patient cluster. All analyses were done with R (4.02) using the packages randomForest (4.6-14) to run RF models and cluster (2.1.0) for k-medoids clustering.

DNA purification and MSP1 genotyping

Genomic DNA was isolated using the QIAamp DNA Mini Kit (Qiagen) according to the manufacturer's protocol. To assess the number of *P. falciparum* genotypes present in the patient isolates, MSP1 genotyping was carried out as described elsewhere (Robert *et al*, 1996).

RNA extraction, RNA-seq library preparation, and sequencing

TRIzol samples were thawed, mixed rigorously with 0.2 volumes of cold chloroform and incubated for 3 min at room temperature. After centrifugation for 30 min at 4°C and maximum speed, the supernatant was carefully transferred to a new tube and mixed with an equal volume of 70% ethanol. Afterwards the manufacturer's instruction from the RNeasy MinElute Kit (Qiagen) were followed with DNase digestion (DNase I, Qiagen) for 30 min on column. Elution of the RNA was carried out in 14 µl. Human globin mRNA was depleted from all samples except from samples #1 and #2 using the GLOBINclear kit (ThermoFisher Scientific). The quality of the RNA was assessed using the Agilent 6000 Pico kit with the Bioanalyzer 2100 (Agilent) (Figure S5), the RNA quantity using the Qubit RNA HA assay kit and a Qubit 3.0 fluorometer (ThermoFisher Scientific). Upon arrival at BGI Genomics Co. (Hong Kong), the RNA quality of each sample was double-checked before sequencing. The median RIN value over all *ex vivo* samples was 6.75 (IQR: 5.93–7.40) (Figure S5), although this measurement has only limited significance for samples containing RNA of two species. Customized library construction in accordance to Tonkin-Hill *et al.* (Tonkin-Hill *et al*, 2018) including amplification with KAPA polymerase and HiSeq 2500 100 bp paired-end sequencing was also performed by BGI Genomics Co. (Hong Kong).

RNA-seq read mapping and data analysis

Var *gene assembly*

Var genes were assembled using the pipeline described in Tonkin-Hill *et al.* (Tonkin-Hill *et al*, 2018). Briefly, non-*var* reads were first filtered out by removing reads that aligned to *H. sapiens*, *P. vivax* or non-*var P. falciparum*. Assembly of the remaining reads was then performed using a pipeline combining SOAPdenovo-Trans and Cap3 (Xie *et al*, 2014; Huang & Madan, 1999; Liao *et al*, 2013). Finally, contaminants were removed from the resulting contigs and they were then translated into the correct reading frame. Reads were mapped to the contigs using BWA-MEM (Li, 2013) and RPKM values were calculated for each *var* transcript to compare individual transcript levels in each patient. Although transcripts might be differentially covered by RNA-seq due to their variable GC content, this seems not to be an issue between *var* genes (Tonkin-Hill *et al*, 2018).

Var *transcript differential expression*

Expression for the assembled *var* genes was quantified using Salmon v0.14.1 (Patro *et al*, 2017) for 531 transcripts with five read counts in at least 3 patient isolates. Both the naïve and pre-exposed groups as well as the severe and non-severe groups were compared. The combined set of all *de novo* assembled transcripts was used as a reference in addition to the coding regions of the 3D7 reference genome. The Salmon algorithm identifies equivalence sets between transcripts allowing a single read to support the expression of multiple transcripts. This enables it to account for the redundancy present in our dataset. To confirm the suitability of this approach we also ran the Corset algorithm as used in Tonkin-Hill *et al.*, (Tonkin-Hill *et al*, 2018; Davidson & Oshlack, 2014). Unlike Salmon which attempts to quantify the expression of transcripts themselves, Corset copes with the redundancy present in *de novo* transcriptome assemblies by clustering similar transcripts together using both the sequence identify of the transcripts as well as multi-mapping read alignments. Of the transcripts identified using the Salmon analysis 5/15 in the naïve versus pre-exposed and 4/13 in the severe versus non-severe were identified in the significant clusters produced using Corset. As the two algorithms take very different approaches and as Salmon is quantifying transcripts rather than the 'gene' like clusters of Corset this represents a fairly reasonable level of concordance between the two methods. In both the Salmon and Corset pipelines differential expression analysis of the resulting *var* expression values was performed using DESeq2 v1.26 (Love *et al*, 2014). The Benjamini-Hochberg method was used to control for multiple testing (Benjamini & Hochberg, 1995).

To check differential expression of the conserved *var* gene variants *var1*-3D7, *var1*-IT and *var2csa* raw reads were mapped with BWA-MEM (AS score >110) to the reference genes from the 3D7 and the IT strains. The mapped raw read counts (bam files) were normalized with the

number of 3D7 mappable reads in each isolate using bamCoverage by introducing a scaling
factor to generate bigwig files displayed in Artemis (Carver *et al*, 2012).

Var *domain and segment differential expression*
Differential expression analysis at the domain and segment level was performed using a similar
approach to that described previously (Tonkin-Hill *et al*, 2018). Initially, the domain families
and homology blocks defined in Rask *et al.* were annotated to the assembled transcripts using
HMMER v3.1b2 (Rask *et al*, 2010; Eddy, 2011). Domains and segments previously identified
to be significantly associated with severe disease in Tonkin-Hill *et al.,* 2018 were also anno-
tated by single pairwise comparison in the assembled transcripts using USEARCH v11.0.667
(Tonkin-Hill *et al*, 2018; Edgar, 2010). Overall, 336 contigs (5.22% of all contigs >500 bp) pos-
sess partial domains in an unusual order, e.g. an NTS in an internal region or a tandem ar-
rangement of two DBLα or CIDRα domains. This might be caused by *de novo* assembly errors,
which is challenging from transcriptome data. Therefore, in both cases the domain or segment
with the most significant alignment was taken as the best annotation for each region of the
assembled *var* transcripts (E-value cutoff of 1e$^-$8), once with the additional requirement that at
least 60% of the domain was found. The expression at each of these annotations was then
quantified using featureCounts v1.6.4 before the counts were aggregated to give a total for
each domain and segment family in each sample. Finally, similar to the transcript level analy-
sis, DESeq2 was used to test for differences in expression levels of both domain and segment
families in the naïve versus pre-exposed groups as well as the severe versus non-severe
groups. Again, more than five read counts in at least three patient isolates were required for
inclusion into differential expression analysis.

Differential expression of core genes
Differential gene expression analysis of *P. falciparum* core genes was done in accordance to
Tonkin-Hill *et al.* (Tonkin-Hill *et al*, 2018) by applying the script as given in the Github repository
(https://github.com/gtonkinhill/falciparum_transcriptome_manuscript/tree/master/all_gene_an
alysis). In brief, subread-align v1.4.6 (Liao *et al*, 2013) were used to align the reads to the
H. sapiens and *P. falciparum* reference genomes. Read counts for each gene were obtained
with FeatureCounts v1.20.2 (Liao *et al*, 2014). To account for parasite life cycle, each sample
is considered as a composition of six parasite life cycle stages excluding the ookinete stage
(López-Barragán *et al*, 2011). Unwanted variations were determined with the 'RUV' (Remove
Unwanted Variation) algorithm implemented in the R package ruv v0.9.6 (Gagnon-Bartsch &
Speed, 2012) adjusting for systematic errors of unknown origin by using the genes with the
1009 lowest p-values as controls as described in (Vignali *et al*, 2011). The gene counts and

estimated ring-stage factor, and factors of unwanted variation were then used as input for the
Limma /Voom (Law *et al*, 2014; Smyth, 2005) differential analysis pipeline.
806
Functional enrichment analysis of differentially expressed core genes
Genes that were identified as significantly differentially expressed (defined as -1<logFC>1,
p<0.05) during prior differential gene expression analysis were used for functional enrichment
analysis using the R package gprofiler2 (Kolberg *et al*, 2020). Enrichment analysis was per-
formed on multiple input lists containing genes expressed significantly higher (logFC > 1, P <
0.05) and lower (logFC < - 1, P < 0.05) between different patient cohorts. All *var* genes were
excluded from the enrichment analysis. For custom visualization of results, gene set data
sources available for *P. falciparum* were downloaded from gprofiler (Raudvere *et al*, 2019).
Pathway data available in the KEGG database (https://www.kegg.jp/kegg/) was accessed via
the KEGG API using KEGGREST (Tenenbaum, 2020) to supplement gprofiler data sources
and build a custom data source in Gene Matrix Transposed file format (*.gmt) for subsequent
visualization. Functional enrichment results were then output to a Generic Enrichment Map
(GEM) for visualization using the Cytoscape EnrichmentMap app (Merico *et al*, 2010) and
RCy3 (Gustavsen *et al*, 2019). Bar plots of differential gene expression values for genes of
selected KEGG pathways were generated using ggplot2 (Wickham, 2016) and enriched KEGG
pathways were visualized using KEGGprofile (Zhao S, Guo Y, 2020).
823
DBLα-tag sequencing
For DBLα-tag PCR the forward primer varF_dg2 (5'-tcgtcggcagcgtcagatgtgtataagaga-
cagGCAMGMAGTTTYGCNGATATWGG-3') and the reverse primer brlong2 (5'-
gtctcgtgggctcggagatgtgtataagagacagTCTTCDSYCCATTCVTCRAACCA-3') were used result-
ing in an amplicon size of 350-500 bp (median 422 bp) plus the 67 bp overhang (small type).
Template cDNA (1 µl) was mixed with 5x KAPA HiFi buffer, 0.3 µM of each dNTP, 2 µM of
each primer and 0.5 U KAPA HiFi Hotstart Polymerase in a final reaction volume of 25 µl.
Reaction mixtures were incubated at 95°C for 2 min and then subjected to 35 cycles of 98°C
for 20 s, 54°C for 30 s and 68°C for 75 s with a final elongation step at 72°C for 2 min. For the
first 5 cycles cooling from denaturation temperature is performed to 65°C at a maximal ramp
of 3°C per second, then cooled to 54°C with a 0.5°C per second ramp. Heating from annealing
temperature to elongation temperature was performed with 1°C per second, all other steps
with a ramp of 3°C per second. Agarose gel images taken afterwards showed clean amplicons.
The DBLα-tag primers contain an overhang, which was used to conduct a second indexing
PCR reaction using sample-specific indexing primers as described in Nag *et al.* (Nag *et al*,
2017). The overhang sequence also serves as annealing site for Illumina sequencing primers
and indexing primers include individual 8-base combinations and adapter sequences that will

allow the final PCR product to bind in MiSeq Illumina sequencing flow cells. Indexing PCR
reactions were performed with a final primer concentration of 0.065 µM and 1 µl of first PCR
amplicon in a final volume of 20 µl; and by following steps: Heat activation at 95 °C, 15 min, 20
cycles of 95 °C for 20 s, 60 °C for 1 min and 72 °C for 1 min, and one final elongation step at
72 °C for 10 min. Indexing PCR amplicons were pooled (4 µl of each) and purified using AM-
Pure XP beads (Beckman Coulter, California, United States) according to manufacturer's pro-
tocol, using 200 µl pooled PCR product and 0.6 x PCR-pool volume of beads, to eliminate
primer dimers. The purified PCR pool were analyzed on agarose gels and Agilent 2100 Bio-
analyser to verify elimination of primer dimers, and correct amplicon sizes. Concentration of
purified PCR pools was measured by Nanodrop2000 (Thermo Fisher Scientific, Waltham, MA,
USA) and an aliquot adjusted to 4 nM concentration was pooled with other unrelated DNA
material and added to an Illumina MiSeq instrument for paired end 300 bp reads using a MiSeq
v3 flow cell.
854

DBLα-tag sequence analysis

The paired-end DBLα-tag sequences were identified and partitioned into correct sample origin
based on unique index sequences. Each indexed raw sequence-pair were then processed
through the *Galaxy* webtool (usegalaxy.eu). Read quality checks was first performed with
FastQC to ensure a good NGS run (sufficient base quality, read length, duplication etc.). Next,
the sequences were trimmed by the Trimmomatic application, with a four base sliding window
approach and a *Phred* quality score above 20 to ensure high sequence quality output. The
trimmed sequences were then paired and converted, following analysis using the Varia tool for
quantification and prediction of the domain composition of the full-length *var* sequences from
which the DBLα-tag originated (Mackenzie *et al*, 2020). In brief, Varia clusters DBLα-tags with
99% sequence identity using *Vsearch* program (v2.14.2), and each unique tag is used to
search a database consisting of roughly 235,000 annotated *var* genes for near identical *var*
sequences (95% identity over 200 nucleotides). The domain composition of all "hit" sequences
is checked for conflicting annotations and the most likely domain composition is retuned. The
tool validation indicated prediction of correct domain compositions for around 85% of randomly
selected *var* tags, with higher hit rate and accuracy of the N-terminal domains. An average of
2,223.70 reads per patient sample was obtained and clusters consisting of less than 10 reads
were excluded from the analysis. The raw Varia output file is given in Table S8. The proportion
of transcripts encoding a given *Pf*EMP1 domain type or subtype was calculated for each pa-
tient. These expression levels were used to first test the hypothesis that N-terminal domain
types associated with EPCR are found more frequently in first-time infections or upon severity
of disease, while those associated with non-EPCR binding were associated with pre-exposed
or mild cases. Secondly, quantile regression was used to calculate median differences (with

95%-confidence intervals) in expression levels for all main domain classes and subtypes be-
tween severity and exposure groups. All analyses were done with R (4.02) using the package
quantreg (5.73) for quantile regression.
For the comparison of both approaches, DBLα-tag sequencing and RNA-seq, only RNA-seq
contigs spanning the whole DBLα-tag region were considered. All conserved variants, the sub-
families *var1*, *var2csa* and *var3*, detected by RNA-seq were omitted form analysis since they
were not properly amplified by the DBLα-tag primers. To scan for the occurrence of DBLα-tag
sequences within the contigs assembled from the RNA-seq data we applied BLAST (basic
local alignment search tool) v2.9.0 software (Altschul *et al*, 1990). Therefore, we created a
BLAST database from the RNA-seq assemblies and screened for the occurrence of those
DBLα-tag sequence with more than 97% percent sequence identity using the "megablast op-
tion".
Calculation of the proportion of RNA-seq data covered by DBLα-tag was done with the upper
75th percentile based on total RPKM values determined for each patient. Vice versa, only
DBLα-tag clusters with more than 10 reads were considered and percent coverage of reads
and clusters calculated for each individual patient.
For all samples the agreement between the two molecular methods DBLα-tag sequencing and
RNA-seq was analyzed with a Bland-Altman plot, each individually and summarized. The ratio
between %-transformed measurements are plotted on the y-axis and the mean of the respec-
tive DBLα-tag and RNA-seq results are plotted on the x-axis. The bias and the 95% limits of
agreement were calculated using GraphPad Prism 8.4.2.
899

900   **Acknowledgments**
901
902   We thank all the patients providing an extra blood sample for our research purposes. Further-
903   more, we thank Jürgen May for critical reading of the manuscript and Tobias Spielmann for
904   helpful discussions. We thank Marlene Danner Dalgaard, Kathrine Hald Langhoff and Sif Ravn
905   Søeborg technical assistance with DBLα-tag sequencing.
906

907   **References**
908

Altschul SF, Gish W, Miller W, Myers EW & Lipman DJ (1990) Basic local alignment search
tool. *J Mol Biol* 215: 403–410
Andrade CM, Fleckenstein H, Thomson-Luque R, Doumbo S, Lima NF, Anderson C, Hibbert
J, Hopp CS, Tran TM, Li S, *et al* (2020) Increased circulation time of Plasmodium
falciparum underlies persistent asymptomatic infection in the dry season. *Nat Med*
Argy N, Kendjo E, Augé-Courtoi C, Cojean S, Clain J, Houzé P, Thellier M, Hubert V,

Deloron P & Houzé S (2017) Influence of host factors and parasite biomass on the
severity of imported Plasmodium falciparum malaria. *PLoS One* 12: e0175328

Avril M, Brazier AJ, Melcher M, Sampath S & Smith JD (2013) DC8 and DC13 var Genes
Associated with Severe Malaria Bind Avidly to Diverse Endothelial Cells. *PLoS Pathog* 9

Avril M, Tripathi AK, Brazier AJ, Andisi C, Janes JH, Soma VL, Sullivan DJ, Bull PC, Stins
MF & Smith JD (2012) A restricted subset of var genes mediates adherence of
Plasmodium falciparum-infected erythrocytes to brain endothelial cells. *Proc Natl Acad*
Sci U S A 109: E1782–E1790

Bachmann A, Bruske E, Krumkamp R, Turner L, Wichers JS, Petter M, Held J, Duffy MF,
Sim BKL, Hoffman SL, *et al* (2019) Controlled human malaria infection with Plasmodium
falciparum demonstrates impact of naturally acquired immunity on virulence gene
expression. *PLOS Pathog* 15: e1007906

Benjamini Y & Hochberg Y (1995) Controlling the False Discovery Rate: A Practical and
Powerful Approach to Multiple Testing. *J R Stat Soc Ser B* 57: 289–300

Berger SS, Turner L, Wang CW, Petersen JEV, Kraft M, Lusingu JPA, Mmbando B,
Marquard AM, Bengtsson DBAC, Hviid L, *et al* (2013) Plasmodium falciparum
Expressing Domain Cassette 5 Type PfEMP1 (DC5-PfEMP1) Bind PECAM1. *PLoS One*
8: 69117

Bernabeu M, Danziger SA, Avril M, Vaz M, Babar PH, Brazier AJ, Herricks T, Maki JN,
Pereira L, Mascarenhas A, *et al* (2016) Severe adult malaria is associated with specific
PfEMP1 adhesion types and high parasite biomass. *Proc Natl Acad Sci U S A* 113:
E3270–E3279

Bernabeu M, Gunnarsson C, Vishnyakova M, Howard CC, Nagao RJ, Avril M, Taylor TE,
Seydel KB, Zheng Y & Smith JD (2019) Binding Heterogeneity of Plasmodium
falciparum to Engineered 3D Brain Microvessels Is Mediated by EPCR and ICAM-1.
MBio 10: e00420-19

Borrmann S (2020) Mapping of safe and early chemo-attenuated live Plasmodium falciparum
immunization identifies immune signature of vaccine efficacy. *bioRxiv*:
2020.09.14.296152

Carlson J, Helmby H, Wahlgren M, Carlson J, Helmby H, Wahlgren M, Hill A, Brewster D &
Greenwood BM (1990) Human cerebral malaria: association with erythrocyte rosetting
and lack of anti-rosetting antibodies. *Lancet* 336: 1457–1460

Carver T, Harris SR, Berriman M, Parkhill J & McQuillan JA (2012) Artemis: An integrated
platform for visualization and analysis of high-throughput sequence-based experimental
data. *Bioinformatics* 28: 464–469

Chaiyaroj SC, Coppel RL, Magowan C & Brown G V. (1994) A Plasmodium falciparum
isolate with a chromosome 9 deletion expresses a trypsin-resistant cytoadherence

molecule. *Mol Biochem Parasitol* 67: 21–30

Cham GKK, Turner L, Kurtis JD, Mutabingwa T, Fried M, Jensen ATR, Lavstsen T, Hviid L,

Duffy PE & Theander TG (2010) Hierarchical, domain type-specific acquisition of

antibodies to Plasmodium falciparum erythrocyte membrane protein 1 in Tanzanian

children. *Infect Immun* 78: 4653–4659

Cham GKK, Turner L, Lusingu J, Vestergaard L, Mmbando BP, Kurtis JD, Jensen ATR,

Salanti A, Lavstsen T & Theander TG (2009) Sequential, Ordered Acquisition of

Antibodies to Plasmodium falciparum Erythrocyte Membrane Protein 1 Domains . *J*

Immunol 183: 3356–3363

Claessens A, Adams Y, Ghumra A, Lindergard G, Buchan CC, Andisi C, Bull PC, Mok S,

Gupta AP, Wang CW, *et al* (2012) A subset of group A-like var genes encodes the

malaria parasite ligands for binding to human brain endothelial cells. *Proc Natl Acad Sci*

U S A 109: E1772

Dasanna AK, Lansche C, Lanzer M & Schwarz US (2017) Rolling Adhesion of Schizont

Stage Malaria-Infected Red Blood Cells in Shear Flow. *Biophys J* 112: 1908–1919

Davidson NM & Oshlack A (2014) Corset: Enabling differential gene expression analysis for

de novo assembled transcriptomes. *Genome Biol* 15: 410

Day KP, Karamalis F, Thompson J, Barnes DA, Peterson C, Brown H, Brown G V. & Kemp

DJ (1993) Genes necessary for expression of a virulence determinant and for

transmission of Plasmodium falciparum are located on a 0.3-megabase region of

chromosome 9. *Proc Natl Acad Sci U S A* 90: 8292–8296

Dent AE, Malhotra I, Wang X, Babineau D, Yeo KT, Anderson T, Kimmel RJ, Angov E, Lanar

DE, Narum D, *et al* (2016) Contrasting Patterns of Serologic and Functional Antibody

Dynamics to Plasmodium falciparum Antigens in a Kenyan Birth Cohort. *Clin Vaccine*

Immunol 23: 104–116

Doolan DL, Mu Y, Unal B, Sundaresh S, Hirst S, Valdez C, Randall A, Molina D, Liang X,

Freilich DA, *et al* (2008) Profiling humoral immune responses to P. falciparum infection

with protein microarrays. *Proteomics* 8: 4680–4694

Drakeley CJ, Corran PH, Coleman PG, Tongren JE, McDonald SLR, Carneiro I, Malima R,

Lusingu J, Manjurano A, Nkya WMM, *et al* (2005) Estimating medium- and long-term

trends in malaria transmission by using serological markers of malaria exposure. *Proc*

Natl Acad Sci U S A 102: 5108–5113

Duffy F, Bernabeu M, Babar PH, Kessler A, Wang CW, Vaz M, Chery L, Mandala WL,

Rogerson SJ, Taylor TE, *et al* (2019) Meta-analysis of Plasmodium falciparum var

Signatures Contributing to Severe Malaria in African Children and Indian Adults. *MBio*

10: e00217-19

Duffy MF, Brown G V., Basuki W, Krejany EO, Noviyanti R, Cowman AF & Reeder JC (2002)

Transcription of multiple var genes by individual, trophozoite-stage Plasmodium

falciparum cells expressing a chondroitin sulphate A binding phenotype. *Mol Microbiol*

43: 1285–1293

Duffy MF, Caragounis A, Noviyanti R, Kyriacou HM, Choong EK, Boysen K, Healer J, Rowe

JA, Molyneux ME, Brown G V., *et al* (2006) Transcribed var genes associated with

placental malaria in Malawian women. *Infect Immun* 74: 4875–4883

Duffy MF, Selvarajah SA, Josling GA & Petter M (2014) Epigenetic regulation of the

plasmodium falciparum genome. *Brief Funct Genomics* 13: 203–216

Eddy SR (2011) Accelerated profile HMM searches. *PLoS Comput Biol* 7: 1002195

Edgar RC (2010) Search and clustering orders of magnitude faster than BLAST.

Bioinformatics 26: 2460–2461

Falk N, Kaestli M, Qi W, Ott M, Baea K, Cortés A & Beck HP (2009) Analysis of plasmodium

falciparum var genes expressed in children from Papua New Guinea. *J Infect Dis* 200:

347–356

Felgner PL, Roestenberg M, Liang L, Hung C, Jain A, Pablo J, Nakajima-Sasaki R, Molina D,

Teelen K, Hermsen CC, *et al* (2013) Pre-erythrocytic antibody profiles induced by

controlled human malaria infections in healthy volunteers under chloroquine

prophylaxis. *Sci Rep* 3

Gagnon-Bartsch JA & Speed TP (2012) Using control genes to correct for unwanted

variation in microarray data. *Biostatistics* 13: 539–552

Ghumra A, Semblat J-P, Ataide R, Kifude C, Adams Y, Claessens A, Anong DN, Bull PC,

Fennell C, Arman M, *et al* (2012) Induction of Strain-Transcending Antibodies Against

Group A PfEMP1 Surface Antigens from Virulent Malaria Parasites. *PLoS Pathog* 8:

e1002665

Gustavsen JA, Pai S, Isserlin R, Demchak B & Pico AR (2019) RCy3: Network biology using

Cytoscape from within R. *F1000Research* 8: 1774

Heiber A, Kruse F, Pick C, Grüring C, Flemming S, Oberli A, Schoeler H, Retzlaff S, Mesén-

Ramírez P, Hiss JA, *et al* (2013) Identification of new PNEPs indicates a substantial

non-PEXEL exportome and underpins common features in Plasmodium falciparum

protein export. *PLoS Pathog* 9: e1003546

Helb DA, Tetteh KKA, Felgner PL, Skinner J, Hubbard A, Arinaitwe E, Mayanja-Kizza H,

Ssewanyana I, Kamya MR, Beeson JG, *et al* (2015) Novel serologic biomarkers provide

accurate estimates of recent Plasmodium falciparum exposure for individuals and

communities. *Proc Natl Acad Sci U S A* 112: E4438–E4447

Herricks T, Avril M, Janes J, Smith JD & Rathod PK (2013) Clonal variants of Plasmodium

falciparum exhibit a narrow range of rolling velocities to host receptor CD36 under

dynamic flow conditions. *Eukaryot Cell* 12: 1490–1498

Van Den Hoogen LL, Walk J, Oulton T, Reuling IJ, Reiling L, Beeson JG, Coppel RL, Singh
SK, Draper SJ, Bousema T, *et al* (2019) Antibody responses to antigenic targets of
recent exposure are associated with low-density parasitemia in controlled human
plasmodium falciparuminfections. *Front Microbiol* 10

Hsieh FL, Turner L, Bolla JR, Robinson C V., Lavstsen T & Higgins MK (2016) The structural
basis for CD36 binding by the malaria parasite. *Nat Commun* 7

Huang X & Madan A (1999) CAP3: A DNA sequence assembly program. *Genome Res* 9:
868–877

Janes JH, Wang CP, Levin-Edens E, Vigan-Womas I, Guillotte M, Melcher M, Mercereau-
Puijalon O & Smith JD (2011) Investigating the host binding signature on the
Plasmodium falciparum PfEMP1 protein family. *PLoS Pathog* 7

Jensen AR, Adams Y & Hviid L (2020) Cerebral Plasmodium falciparum malaria: The role of
PfEMP1 in its pathogenesis and immunity, and PfEMP1-based vaccines to prevent it.
Immunol Rev 293: 230–252 doi:10.1111/imr.12807

Jensen ATR, Magistrado P, Sharp S, Joergensen L, Lavstsen T, Chiucchiuini A, Salanti A,
Vestergaard LS, Lusingu JP, Hermsen R, *et al* (2004) Plasmodium falciparum
Associated with Severe Childhood Malaria Preferentially Expresses PfEMP1 Encoded
by Group A var Genes. *J Exp Med* 199: 1179–1190

Jespersen JS, Wang CW, Mkumbaye SI, Minja DT, Petersen B, Turner L, Petersen JE,
Lusingu JP, Theander TG & Lavstsen T (2016) *Plasmodium falciparum var* genes
expressed in children with severe malaria encode CIDRα1 domains. *EMBO Mol Med* 8:
839–850

Kapelski S, Klockenbring T, Fischer R, Barth S & Fendel R (2014) Assessment of the
neutrophilic antibody-dependent respiratory burst (ADRB) response to Plasmodium
falciparum . *J Leukoc Biol* 96: 1131–1142

Kessler A, Dankwa S, Bernabeu M, Harawa V, Danziger SA, Duffy F, Kampondeni SD,
Potchen MJ, Dambrauskas N, Vigdorovich V, *et al* (2017) Linking EPCR-Binding
PfEMP1 to Brain Swelling in Pediatric Cerebral Malaria. *Cell Host Microbe* 22: 601-
614.e5

Kirchgatter K & Del Portillo HA (2002) Association of severe noncerebral Plasmodium
falciparum malaria in Brazil with expressed PfEMP1 DBL1α sequences lacking cysteine
residues. *Mol Med* 8: 16–23

Kivisi CA, Muthui M, Hunt M, Fegan G, Otto TD, Githinji G, Warimwe GM, Rance R, Marsh
K, Bull PC, *et al* (2019) Exploring Plasmodium falciparum Var Gene Expression to
Assess Host Selection Pressure on Parasites During Infancy. *Front Immunol* 10: 2328

Kolberg L, Raudvere U, Kuzmin I, Vilo J & Peterson H (2020) gprofiler2 -- an R package for
gene list functional enrichment analysis and namespace conversion toolset g:Profiler.

F1000Research 9: 709

Kraemer SM & Smith JD (2003) Evidence for the importance of genetic structuring to the

structural and functional specialization of the Plasmodium falciparum var gene family.

Mol Microbiol 50: 1527–1538

Kyes SA, Christodoulou Z, Raza A, Horrocks P, Pinches R, Rowe JA & Newbold CI (2003) A

well-conserved Plasmodium falciparum var gene shows an unusual stage-specific

transcript pattern. *Mol Microbiol* 48: 1339–1348

Kyes SA, Kraemer SM & Smith JD (2007) Antigenic variation in Plasmodium falciparum:

Gene organization and regulation of the var multigene family. *Eukaryot Cell* 6: 1511–

1520 doi:10.1128/EC.00173-07

Kyriacou HM, Stone GN, Challis RJ, Raza A, Lyke KE, Thera MA, Koné AK, Doumbo OK,

Plowe C V. & Rowe JA (2006) Differential var gene transcription in Plasmodium

falciparum isolates from patients with cerebral malaria compared to hyperparasitaemia.

Mol Biochem Parasitol 150: 211–218

Lau CKY, Turner L, Jespersen JS, Lowe ED, Petersen B, Wang CW, Petersen JEV, Lusingu

J, Theander TG, Lavstsen T, *et al* (2015) Structural conservation despite huge

sequence diversity allows EPCR binding by the pfemp1 family implicated in severe

childhood malaria. *Cell Host Microbe* 17: 118–129

Lavstsen T, Salanti A, Jensen ATR, Arnot DE & Theander TG (2003) Sub-grouping of

Plasmodium falciparum 3D7 var genes based on sequence analysis of coding and non-

coding regions. *Malar J* 2: 1–14

Lavstsen T, Turner L, Saguti F, Magistrado P, Rask TS, Jespersen JS, Wang CW, Berger

SS, Baraka V, Marquard AM, *et al* (2012) Plasmodium falciparum erythrocyte

membrane protein 1 domain cassettes 8 and 13 are associated with severe malaria in

children. *Proc Natl Acad Sci U S A* 109: E1791–E1800

Law CW, Chen Y, Shi W & Smyth GK (2014) Voom: Precision weights unlock linear model

analysis tools for RNA-seq read counts. *Genome Biol* 15

Lee HJ, Georgiadou A, Walther M, Nwakanma D, Stewart LB, Levin M, Otto TD, Conway DJ,

Coin LJ & Cunnington AJ (2018) Integrated pathogen load and dual transcriptome

analysis of systemic host-pathogen interactions in severe malaria. *Sci Transl Med* 10:

eaar3619

Lennartz F, Adams Y, Bengtsson A, Olsen RW, Turner L, Ndam NT, Ecklu-Mensah G,

Moussiliou A, Ofori MF, Gamain B, *et al* (2017) Structure-Guided Identification of a

Family of Dual Receptor-Binding PfEMP1 that Is Associated with Cerebral Malaria. *Cell*

Host Microbe 21: 403–414

Lennartz F, Smith C, Craig AG & Higgins MK (2019) Structural insights into diverse modes of

ICAM-1 binding by Plasmodium falciparum-infected erythrocytes. *Proc Natl Acad Sci U*

S A 116: 20124–20134

Li H (2013) Aligning sequence reads, clone sequences and assembly contigs with BWA-MEM. arXiv:1303.3997

Liao Y, Smyth GK & Shi W (2013) The Subread aligner: Fast, accurate and scalable read mapping by seed-and-vote. *Nucleic Acids Res* 41

Liao Y, Smyth GK & Shi W (2014) FeatureCounts: An efficient general purpose program for assigning sequence reads to genomic features. *Bioinformatics* 30: 923–930

Llewellyn D, Miura K, Fay MP, Williams AR, Murungi LM, Shi J, Hodgson SH, Douglas AD, Osier FH, Fairhurst RM, *et al* (2015) Standardization of the antibody-dependent respiratory burst assay with human neutrophils and Plasmodium falciparum malaria. *Sci Rep* 5

López-Barragán MJ, Lemieux J, Quiñones M, Williamson KC, Molina-Cruz A, Cui K, Barillas-Mury C, Zhao K & Su X (2011) Directional gene expression and antisense transcripts in sexual and asexual stages of Plasmodium falciparum. *BMC Genomics* 12: 587

Love MI, Huber W & Anders S (2014) Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. *Genome Biol* 15: 550

Lubiana P, Bouws P, Roth LK, Dörpinghaus M, Rehn T, Brehmer J, Wichers JS, Bachmann A, Höhn K, Roeder T, *et al* (2020) Adhesion between P. falciparum infected erythrocytes and human endothelial receptors follows alternative binding dynamics under flow and febrile conditions. *Sci Rep* 10: 4548

Mackenzie G, Jensen R, Lavstsen T & Otto T (2020) Varia: Prediction of variable genes. (https://github.com/GCJMackenzie/Varia)

Magallón-Tejada A, Machevo S, Cisteró P, Lavstsen T, Aide P, Rubio M, Jiménez A, Turner L, Valmaseda A, Gupta H, *et al* (2016) Cytoadhesion to gC1qR through Plasmodium falciparum Erythrocyte Membrane Protein 1 in Severe Malaria. *PLOS Pathog* 12: e1006011

Maier AG, Rug M, O'Neill MT, Brown M, Chakravorty S, Szestak T, Chesson J, Wu Y, Hughes K, Coppel RL, *et al* (2008) Exported Proteins Required for Virulence and Rigidity of Plasmodium falciparum-Infected Human Erythrocytes. *Cell* 134: 48–61

McGee M & Chen Z (2006) Parameter estimation for the exponential-normal convolution model for background correction of affymetrix GeneChip data. *Stat Appl Genet Mol Biol* 5

McHugh E, Carmo OMS, Blanch A, Looker O, Liu B, Tiash S, Andrew D, Batinovic S, Low AJY, Cho HJ, *et al* (2020) Role of plasmodium falciparum protein GEXP07 in Maurer's Cleft morphology, Knob architecture, and P. Falciparum EMP1 trafficking. *MBio* 11

McQuaid F & Rowe JA (2020) Rosetting revisited: A critical look at the evidence for host erythrocyte receptors in Plasmodium falciparum rosetting. *Parasitology* 147: 1–11

doi:10.1017/S0031182019001288

Merico D, Isserlin R, Stueker O, Emili A & Bader GD (2010) Enrichment Map: A Network-

Based Method for Gene-Set Enrichment Visualization and Interpretation. *PLoS One* 5:

e13984

Mkumbaye SI, Wang CW, Lyimo E, Jespersen JS, Manjurano A, Mosha J, Kavishe RA,

Mwakalinga SB, Minja DTR, Lusingu JP, *et al* (2017) The severity of *Plasmodium*

falciparum infection is associated with transcript levels of *var* genes encoding EPCR-

binding PfEMP1. *Infect Immun*: IAI.00841-16

Nacer A, Roux E, Pomel S, Scheidig-Benatar C, Sakamoto H, Lafont F, Scherf A & Mattei D

(2011) clag9 is not essential for PfEMP1 surface expression in non-cytoadherent

Plasmodium falciparum parasites with a chromosome 9 deletion. *PLoS One* 6

Nag S, Dalgaard MD, Kofoed PE, Ursing J, Crespo M, Andersen LOB, Aarestrup FM, Lund

O & Alifrangis M (2017) High throughput resistance profiling of Plasmodium falciparum

infections based on custom dual indexing and Illumina next generation sequencing-

technology. *Sci Rep* 7: 1–13

Nunes MC, Goldring JPD, Doerig C & Scherf A (2007) A novel protein kinase family in

Plasmodium falciparum is differentially transcribed and secreted to various cellular

compartments of the host cell. *Mol Microbiol* 63: 391–403

Obeng-Adjei N, Larremore DB, Turner L, Ongoiba A, Li S, Doumbo S, Yazew TB, Kayentao

K, Miller LH, Traore B, *et al* (2020) Longitudinal analysis of naturally acquired PfEMP1

CIDR domain variant antibodies identifies associations with malaria protection. *JCI*

Insight

Oberli A, Slater LM, Cutts E, Brand F, Mundwiler-Pachlatko E, Rusch S, Masik MFG, Erat

MC, Beck HP & Vakonakis I (2014) A Plasmodium falciparum PHIST protein binds the

virulence factor PfEMP1 and comigrates to knobs on the host cell surface. *FASEB J* 28:

4420–4433

Oberli A, Zurbrügg L, Rusch S, Brand F, Butler ME, Day JL, Cutts EE, Lavstsen T,

Vakonakis I & Beck HP (2016) Plasmodium falciparum Plasmodium helical interspersed

subtelomeric proteins contribute to cytoadherence and anchor P. falciparum erythrocyte

membrane protein 1 to the host cell cytoskeleton. *Cell Microbiol* 18: 1415–1428

Obiero JM, Campo JJ, Scholzen A, Randall A, Bijker EM, Roestenberg M, Hermsen CC,

Teng A, Jain A, Davies DH, *et al* (2019) Antibody Biomarkers Associated with Sterile

Protection Induced by Controlled Human Malaria Infection under Chloroquine

Prophylaxis. *mSphere* 4

Otto TD (2019) ThomasDOtto/varDB: First release of varDB.

Otto TD, Assefa SA, Böhme U, Sanders MJ, Kwiatkowski D, Berriman M, Newbold C &

Newbold C (2019) Evolutionary analysis of the most polymorphic gene family in

falciparum malaria. *Wellcome Open Res* 4: 193

Otto TD, Böhme U, Sanders M, Reid A, Bruske EI, Duffy CW, Bull PC, Pearson RD, Abdi A,
Dimonte S, *et al* (2018a) Long read assemblies of geographically dispersed
Plasmodium falciparum isolates reveal highly structured subtelomeres. *Wellcome open*
Res 3: 52

Otto TD, Gilabert A, Crellen T, Böhme U, Arnathau C, Sanders M, Oyola SO, Okouga AP,
Boundenga L, Willaume E, *et al* (2018b) Genomes of all known members of a
Plasmodium subgenus reveal paths to virulent human malaria. *Nat Microbiol* 3: 687–
697

Patro R, Duggal G, Love MI, Irizarry RA & Kingsford C (2017) Salmon provides fast and bias-
aware quantification of transcript expression. *Nat Methods* 14: 417–419

Perkins M (1988) Stage-dependent processing and localization of a Plasmodium falciparum
protein of 130,000 molecular weight. *Exp Parasitol* 65: 61–68

Rask TS, Hansen DA, Theander TG, Gorm Pedersen A & Lavstsen T (2010) Plasmodium
falciparum Erythrocyte Membrane Protein 1 Diversity in Seven Genomes – Divide and
Conquer. *PLoS Comput Biol* 6: e1000933

Raudvere U, Kolberg L, Kuzmin I, Arak T, Adler P, Peterson H & Vilo J (2019) G:Profiler: A
web server for functional enrichment analysis and conversions of gene lists (2019
update). *Nucleic Acids Res* 47: W191–W198

Robert F, Ntoumi F, Angel G, Candito D, Rogier C, Fandeur T, Sarthou JL & Mercereau-
Puijalon O (1996) Extensive genetic diversity of Plasmodium falciparum isolates
collected from patients with severe malaria in Dakar, Senegal. *Trans R Soc Trop Med*
Hyg 90: 704–711

Rottmann M, Lavstsen T, Mugasa JP, Kaestli M, Jensen ATR, Müller D, Theander T & Beck
HP (2006) Differential expression of var gene groups is associated with morbidity
caused by Plasmodium falciparum infection in Tanzanian children. *Infect Immun* 74:
3904–3911

Rowe JA, Claessens A, Corrigan RA & Arman M (2009) Adhesion of Plasmodium
falciparum-infected erythrocytes to human cells: Molecular mechanisms and therapeutic
implications. *Expert Rev Mol Med* 11: e16 doi:10.1016/0925-4005(95)85135-6

Rowe JA, Obiero J, Marsh K & Raza A (2002) Short report: Positive correlation between
rosetting and parasitemia in Plasmodium falciparum clinical isolates. *Am J Trop Med*
Hyg 66: 458–460

Saul A (1999) The role of variant surface antigens on malaria-infected red blood cells.
Parasitol Today 15: 455–7

Shabani E, Hanisch B, Opoka RO, Lavstsen T & John CC (2017) Plasmodium falciparum
EPCR-binding PfEMP1 expression increases with malaria disease severity and is

elevated in retinopathy negative cerebral malaria. *BMC Med* 15: 183

Silver JD, Ritchie ME & Smyth GK (2009) Microarray background correction: Maximum
likelihood estimation for the normal-exponential convolution. *Biostatistics* 10: 352–363

Smyth GK (2005) limma: Linear Models for Microarray Data. In *Bioinformatics and*
Computational Biology Solutions Using R and Bioconductor pp 397–420. Springer-
Verlag

Spielmann T, Hawthorne PL, Bixon MWA, Hannemann M, Klotz K, Kemp DJ, Klonis N, Tilley
L, Trenholme KR & Gardiner DL (2006) A cluster of ring stage-specific genes linked to a
locus implicated in cytoadherence in Plasmodium falciparum codes for PEXEL-negative
and PEXEL-positive proteins exported into the host cell. *Mol Biol Cell* 17: 3613–3624

Spycher C, Klonis N, Spielmann T, Kump E, Steiger S, Tilley L & Beck H-P (2003) MAHRP-
1, a novel Plasmodium falciparum histidine-rich protein, binds ferriprotoporphyrin IX and
localizes to the Maurer's clefts. *J Biol Chem* 278: 35373–83

Spycher C, Rug M, Klonis N, Ferguson DJP, Cowman AF, Beck H-P & Tilley L (2006)
Genesis of and Trafficking to the Maurer's Clefts of Plasmodium falciparum-Infected
Erythrocytes. *Mol Cell Biol* 26: 4074–4085

Spycher C, Rug M, Pachlatko E, Hanssen E, Ferguson D, Cowman AF, Tilley L & Beck H
(2008) The Maurer ' s cleft protein MAHRP1 is essential for trafficking of PfEMP1 to the
surface of Plasmodium falciparum-infected erythrocytes. *Mol Microbiol* 68: 1300–1314

Storm J, Jespersen JS, Seydel KB, Szestak T, Mbewe M, Chisala N V, Phula P, Wang CW,
Taylor TE, Moxon CA, *et al* (2019) Cerebral malaria is associated with differential
cytoadherence to brain endothelial cells. *EMBO Mol Med* 11

Taghavian O, Jain A, Joyner CJ, Ketchum S, Nakajima R, Jasinskas A, Liang L, Fong R,
King C, Greenhouse B, *et al* (2018) Antibody Profiling by Proteome Microarray with
Multiplex Isotype Detection Reveals Overlap between Human and Aotus nancymaae
Controlled Malaria Infections. *Proteomics* 18

Tenenbaum D (2020) KEGGREST: Client-side REST access to KEGG. R package version
1.30.0.

Tonkin-Hill GQ, Trianty L, Noviyanti R, Nguyen HHT, Sebayang BF, Lampah DA, Marfurt J,
Cobbold SA, Rambhatla JS, McConville MJ, *et al* (2018) The Plasmodium falciparum
transcriptome in severe malaria reveals altered expression of genes involved in
important processes including surface antigen–encoding var genes. *PLoS Biol* 16:
e2004328

Tonkin CJ, Carret CK, Duraisingh MT, Voss TS, Ralph SA, Hommel M, Duffy MF, Da Silva
LM, Scherf A, Ivens A, *et al* (2009) Sir2 paralogues cooperate to regulate virulence
genes and antigenic variation in Plasmodium falciparum. *PLoS Biol* 7: 0771–0788

Turewicz M, Ahrens M, May C, Marcus K & Eisenacher M (2016) PAA: An R/bioconductor

1248   package for biomarker discovery with protein microarrays. *Bioinformatics* 32: 1577–

1249   1579

1250 Turner L, Lavstsen T, Berger SS, Wang CW, Petersen JE V., Avril M, Brazier AJ, Freeth J,

1251   Jespersen JS, Nielsen MA, *et al* (2013) Severe malaria is associated with parasite

1252   binding to endothelial protein C receptor. *Nature* 498: 502–505

1253 Turner L, Lavstsen T, Mmbando BP, Wang CW, Magistrado PA, Vestergaard LS, Ishengoma

1254   DS, Minja DTR, Lusingu JP & Theander TG (2015) IgG antibodies to endothelial protein

1255   C receptor-binding cysteine-rich interdomain region domains of Plasmodium falciparum

1256   erythrocyte membrane protein 1 are acquired early in life in individuals exposed to

1257   malaria. *Infect Immun* 83: 3096–3103

1258 Vignali M, Armour CD, Chen J, Morrison R, Castle JC, Biery MC, Bouzek H, Moon W, Babak

1259   T, Fried M, *et al* (2011) NSR-seq transcriptional profiling enables identification of a gene

1260   signature of Plasmodium falciparum parasites infecting children. *J Clin Invest* 121:

1261   1119–1129

1262 Wang WF & Zhang YL (2020) PfSWIB, a potential chromatin regulator for var gene

1263   regulation and parasite development in Plasmodium falciparum. *Parasites and Vectors*

1264   13: 48

1265 Warimwe GM, Keane TM, Fegan G, Musyoki JN, Newton CRJC, Pain A, Berriman M, Marsh

1266   K & Bull PC (2009) Plasmodium falciparum var gene expression is modified by host

1267   immunity. *Proc Natl Acad Sci U S A* 106: 21801–21806

1268 WHO (2019) WHO World Malaria Report 2019

1269 Wickham H (2016) Ggplot2 : elegant graphics for data analysis Springer

1270 Winter G, Chen Q, Flick K, Kremsner P, Fernandez V & Wahlgren M (2003) The 3D7var5.2

1271   (varCOMMON) type var gene family is commonly expressed in non-placental

1272   Plasmodium falciparum malaria. *Mol Biochem Parasitol* 127: 179–191

1273 Xie Y, Wu G, Tang J, Luo R, Patterson J, Liu S, Huang W, He G, Gu S, Li S, *et al* (2014)

1274   SOAPdenovo-Trans: De novo transcriptome assembly with short RNA-Seq reads.

1275   *Bioinformatics* 30: 1660–1666

1276 Yam XY, Birago C, Fratini F, Di Girolamo F, Raggi C, Sargiacomo M, Bachi A, Berry L, Fall

1277   G, Currà C, *et al* (2013) Proteomic analysis of detergent-Resistant membrane

1278   microdomains in trophozoite blood stage of the human malaria parasite plasmodium

1279   falciparum. *Mol Cell Proteomics* 12: 3948–3961

1280 Zhang Q, Siegel TN, Martins RM, Wang F, Cao J, Gao Q, Cheng X, Jiang L, Hon CC,

1281   Scheidig-Benatar C, *et al* (2014) Exonuclease-mediated degradation of nascent RNA

1282   silences genes linked to severe malaria. *Nature* 513: 431–435

1283 Zhao S, Guo Y SY (2020) KEGGprofile: An annotation and visualization package for multi-

1284   types and multi-groups expression data in KEGG pathway. R package version 1.32.0.

Figure legends

1287

Figure 1: Subgrouping of patients into first-time infected and pre-exposed individuals based on antibody levels against *P. falciparum*. In order to further characterize the patient cohort plasma samples (n=32) were subjected to Luminex analysis with the *P. falciparum* antigens AMA1, MSP1, CSP known to induce a strong antibody response in humans. With exception of patient #21 unsupervised clustering of the PCA-reduced data clearly discriminates between first-time infected (naïve) and pre-exposed patients with higher antibody levels against tested *P. falciparum* antigens and also assigns plasma samples from patients with unknown immune status into naïve and pre-exposed clusters (**A**). Classification of patient #21 into the naïve subgroup was confirmed using different serological assays assessing antibody levels against *P. falciparum* on different levels: a merozoite-directed antibody-dependent respiratory burst (mADRB) assay (Kapelski *et al*, 2014) (**B**), a PEMS-specific ELISA (**C**) and a 262-feature protein microarray covering 228 well-known *P. falciparum* antigens detecting reactivity with individual antigens and the antibody breadth of IgG (upper panel) and IgM (lower panel) (**D**). The boxes represent medians with IQR; the whiskers depict minimum and maximum values (range) with outliers located outside the whiskers. Serological assays revealed significant differences between patient groups (Mann Whitney U test). Reactivity of patient plasma IgG and IgM with individual antigens in the protein microarray is presented as volcano plot highlighting the significant hits in red. Box plots represent antibody breadths by summarizing the number of recognized antigens out of 262 features tested. Data from all assays were used for an unsupervised random forest approach (**E**). The variable importance plot of the random forest model shows the decrease in prediction accuracy if values of a variable are permuted randomly. The decrease in accuracy was determined for each serological assay indicating that the mADRB, ELISA and Luminex assays are most relevant in the prediction of patient clusters (**F**). Patients with known immune status based on medical reports were marked in all plots with filled circles in blue (naïve) and grey (pre-exposed), samples from patients with unknown immune status are shown as open circles. ELISA: Enzyme-linked immunosorbent assay, IQR: interquartile range, PCA: Principal component analysis

1315

Figure 2: Estimated stage proportions for each sample. Patient samples consist of a combination of different parasite stages. To estimate the proportion of different life cycle stages in each sample a constrained linear model was fit using data from López-Barragán *et al.* (López-Barragán *et al*, 2011). The proportions of rings (8 hpi), early trophozoites (19 hpi), late trophozoites (30 hpi), schizonts (42 hpi) and gametocytes stages shown in the columns of the bar plots must add to 1 for each sample. Shown are the comparisons between first-time infected

(naïve; blue) and pre-exposed samples (grey) (**A**) and severe (red) and non-severe cases (grey) (**B**). A bias towards the early trophozoite appears in the non-severe malaria sample group, which was confirmed by calculating the age in hours post infection (hpi) for each parasite sample. The boxes represent medians with IQR; the whiskers depict minimum and maximum values (range) with outliers located outside the whiskers (**C**, **D**). IQR: interquartile range

Figure 3: Analysis of RNA-seq data at the level of *var* gene transcripts using the separate assembly approach. RNA-seq reads of each patient sample were matched to *de novo* assembled *var* contigs with varying length, domain and homology block composition. Shown are significant differently expressed *var* gene contigs with an adjusted p-value of <0.05 in first-time infected (blue) and pre-exposed patient samples (grey) (**A**, **B**) as well as severe (red) and non-severe cases (grey) (**C**, **D**). Data are displayed as heatmaps showing expression levels in log transformed normalized Salmon read counts for each individual sample (**A**, **C**) or as box plot with median log transformed normalized Salmon read counts and interquartile range (IQR) for each group of samples (**B**, **D**). Normalized Salmon read counts for all assembled transcripts are available in Table S4.

Figure 4: Analysis of RNA-seq data via *de novo* assembly at the level of *var* gene domains. RNA-seq reads of each patient sample were matched to *de novo* assembled *var* contigs with varying length, domain and homology block composition. Shown are significantly differently expressed *Pf*EMP1 domain subfamilies from Rask *et al.* (Rask *et al*, 2010) with an adjusted p-value of <0.05 in first-time infected (blue) and pre-exposed patient samples (grey) (**A**, **B**) as well as severe (red) and non-severe cases (grey) (**C**, **D**) using HMMER3 models. The N-terminal head structure (NTS-DBLα -CIDRα/β/γ/δ) confers a mutually exclusive binding phenotype either to EPCR-, CD36-, CSA- or an unknown receptor. Expression values of the N-terminal domains were summarized for each patient and differences in the distribution among patient groups were tested using the Mann-Whitney U test (**E**, **F**). Data are displayed as heatmaps showing expression levels in log transcripts per million (TPM) for each individual sample (**A**, **C**) or as box plot with median log TPM and interquartile range (IQR) for each group of samples (**B**, **D, E, F**).

Figure 5: Analysis of RNA-seq data via *de novo* assembly at the level of *var* gene homology blocks. RNA-seq reads of each patient sample were matched to *de novo* assembled *var* contigs with varying length, domain and homology block composition. Shown are significantly differently expressed homology blocks from Rask *et al.* (Rask *et al*, 2010) with an adjusted p-value of <0.05 in first-time infected (blue) and pre-exposed patient samples (grey) (**A**, **B**) as well as severe (red) and non-severe cases (grey) (**C**, **D**). Data are displayed as heatmaps showing expression levels in log transcripts per million (TPM) for each individual sample (**A**,

1360   **C**) or as box plot with median log TPM and interquartile range (IQR) for each group of samples

1361   (**B**, **D**).

1362

Figure 6: Summary of *Pf*EMP1 transcripts, domains, and homology blocks that were

found more or less frequently in malaria-naïve and severely ill patients. A schematic

presentation of typical *Pf*EMP1 domain compositions (**A**). The N-terminal head structure con-

fers mutually exclusive receptor binding phenotypes: EPCR (yellow: CIDRα1.1/4-8), CD36

(salmon CIDRα2-6), CSA (VAR2CSA) and yet unknown phenotypes (orange: CIDRβ/γ/δ, red:

CIDRα1.2/3 from VAR1, VAR3). Group A includes the conserved subfamilies VAR1 and VAR3,

EPCR binding variants and those with unknown binding phenotypes sometimes associated

with rosetting. Group B *Pf*EMP1 can have EPCR-binding capacities, but most variants share

a four-domain structure with group C-type variants capable of CD36-binding. Dual binder can

be found within group A and B with an DBLβ domain responsible for ICAM-1- (DBLβ1/3/5) or

gC1qr-binding (DBLβ12). Inter-strain conserved tandem arrangements of domains, so called

domain cassettes (DC), can be found within all groups as selectively indicated.

Transcripts, domains and homology blocks according to Rask *et al.* (Rask *et al*, 2010) found

significant differently expressed (p-value <0.05) between patient groups of both comparisons:

first-time infected (blue) versus pre-exposed (black) cases and severe (red) versus non-severe

(black) cases (**B**).

ATS: acidic terminal sequence, CIDR: cysteine-rich interdomain region, CSA: chondroitin sul-

phate A, DBL: Duffy binding-like, DC: domain cassette, EPCR: endothelial protein C receptor,

gC1qr: receptor for complement component C1q, ICAM-1: intercellular adhesion molecule 1,

NTS: N-terminal segment, TM: transmembrane domain

1383

Figure 7: Verification of RNA-seq results using DBLα-tag sequencing. Amplified DBLα-

tag sequences were blasted against the ~2,400 genomes on varDB (Otto, 2019) to obtain

subclassification into DBLα0/1/2 and prediction of adjacent head-structure NTS and CIDR do-

mains and their related binding phenotype. Proportion of each NTS and DBLα subclass as well

as CIDR domains grouped according to binding phenotype (CIDRα1.1/4-8: EPCR-binding,

CIDRα2-6: CD36-binding, CIDRβ/γ/δ: unknown binding phenotype/rosetting) was calculated

and shown separately on the left, number of total reads and individual sequence cluster with

n ≥10 sequences are shown on the right. Differences in the distribution among first-time in-

fected (blue) and pre-exposed individuals (grey) (**A**) as well as severe (red) and non-severe

cases (grey) (**B**) were tested using the Mann-Whitney U test. The boxes represent medians

with IQR; the whiskers depict minimum and maximum values (range) with outliers located out-

side the whiskers. Quantile regression was applied to look for differences between patient

groups on the level of domain main classes (left) and subdomains (right). Shown are median

differences with 95%-confidence intervals of domains with values unequal 0. Domains with
positive values tend to be higher expressed in naïve (**C**) and severe patients (**D**).

1399

Figure 8: Correlation of *var* gene expression with antibody levels against head structure
CIDR domains. Patient plasma samples (n=32) were subjected to Luminex analysis with 35
*Pf*EMP1 head structure CIDR domains. The panel includes EPCR-binding CIDRα1 domains
(n = 19), CD36-binding CIDRα2–6 domains (n = 12) and CIDR domains with unknown binding
phenotype (CIDRγ3: n = 1, CIDRδ1: n = 3) as well as the minimal binding region of VAR2CSA
(VAR2). Box plots showing mean fluorescence intensities (MFI) extending from the 25th to the
75th percentiles with a line at the median indicate higher reactivity of the pre-exposed (**A**) and
non-severe cases (**B**) with all *Pf*EMP1 domains tested. Significant differences were observed
for recognition of CIDRα2–6, CIDRδ1 and CIDRγ3; VAR2CSA recognition differed only be-
tween severe and non-severe cases (Mann Whitney U test). Furthermore, the breadth of IgG
recognition (%) of CIDR domains for the different patient groups was calculated and shown as
a heat map (**C**).

1412

Figure 9: Differential all gene expression analysis. Gene set enrichment analysis (GSEA)
of GO terms and KEGG pathways indicate gene sets deregulated in first-time infected malaria
patients. GO terms related to antigenic variation and host cell remodeling are significantly
down-regulated, only the KEGG pathway 03410 'base excision repair' shows a significant up-
regulation in malaria-naïve patients (**A**). Log fold changes (logFC) for the 15 *P. falciparum*
genes assigned to the KEGG pathway 03410 'base excision repair' are plotted with the six
significant hits marked with * for p<0.05 and ** for p<0.01 (**B**).

1420

1421   **Table 1: Characteristics and classification of malaria patients.**

1422

1423   **Table 2: Patient groups data.**

1424

1425   **Table 3: *var* transcripts up- and down-regulated in first-time infected patients.**

1426

1427   **Table 4: *var* transcripts up- and down-regulated in severe cases.**

1428

1429   **Table 5: *var* domains defined by Rask *et al.* (Rask *et al*, 2010) up- and down-regulated**
1430   **in first-time infected patients.**

1431

1432   **Table 6: *var* domains defined by Rask *et al.* . (Rask *et al*, 2010) up- and down-regulated**
1433   **in severe cases.**

Table 7: *var* blocks defined by Rask *et al.* . (Rask *et al*, 2010) up- and down-regulated in first-time infected patients.

Table 8: *var* blocks defined by Rask *et al.* . (Rask *et al*, 2010) down-regulated in severe cases.

Supplement

Supplement figure 1: Early immune response in mild and severe malaria within the naïve patient cluster. Antibody reactivity against individual antigens within the three subgroups 'naïve with mild symptoms', 'naïve with severe symptoms' and 'pre-exposed with mild symptoms'. Sera from all volunteers were assessed on protein microarrays and data normalized to control spots containing no antigen (no DNA control spots). Median reactivity of the mild infected malaria-naïve, severely infected malaria-naïve as well as the mild infected with pre-exposure to malaria are represented as bar-charts. IgG data is given for all 262 *P. falciparum* proteins spotted on the microarray representing 228 unique antigens (**A**). To estimate differences in immune response in mild and severe malaria within the malaria-naïve population, normalized IgG (**B**) and IgM (**C**) antibody responses were compared in the two subpopulations. Differentially recognized antigens (p <0.05 and fold change >2) are depicted in red.

Supplement figure 2: Differential expression of the *var1*-allele forms and *var2csa* between patient groups. RNA-seq reads from each patient were normalized against the number of mappable reads to the 3D7 genome and aligned to the *var1*-3D7 and *var1*-IT allele forms as well as *var2csa*. The resulting bigwig files were displayed in Artemis (Carver *et al*, 2012). Individual samples are colored according to the patient group: first-time infected in blue (A), severe in red (B) and the respective pre-exposed or non-severe samples in grey.

Supplement figure 3: Comparison of the DBLα-tag sequencing with RNA-seq analysis. DBLα-tag sequencing and RNA-seq data compared in Bland-Altman plots for all patients summarized (**A**) and for each individual patient (**B**), where the mean log expression of each gene is indicated on the X-axis and the log ratio between normalized DBLα-tag (% of reads) and RNA-seq values (% of RPKM from all contigs containing both DBLα-tag primer binding sites) on the y-axis. The mean (equal to bias) of all ratios (line) and the confidence interval (CI) of 95% (dotted lines) are indicated. Data points with negative values for one of the approaches are displayed in dependence of their mean log expression on top (DBLα-tag sequence clusters

1470   not detected by RNA-seq) or bottom (RNA-seq contigs not found within DBLα-tag sequence

1471   cluster) of the graph.

1472

1473   **Supplement figure 4: The base excision repair (KEGG:03410) in _P. falciparum_.**

1474   Orthologues present in _P. falciparum_ are indicated by gene IDs, log fold changes (logFC) are

1475   indicated by color code (red: up-regulated, blue: down-regulated) (**A**). Summary of logFC in

1476   gene expression in first-time infected relative to pre-exposed patients and p-values for the

1477   logFC.

1478

Supplement figure 5: RNA quality. The Bioanalyzer automated RNA electrophoresis system

was used to characterize the total RNA quality prior library synthesis. The calculated RIN value

is provided, although this measurement is questionable for samples from mixed species. From

the four rRNA peaks visible in all samples, the inner peaks represent _P. falciparum_ 18S and

28S rRNA, the outer peaks are of human origin.

Data S1: Sequences of assembled _var_ contigs from all patient isolates.

1485

1486   **Supplement table 1: Data from Luminex, mADRB, ELISA and protein microarray.** Sero-

1487   prevalence of head-structure CIDR domains determined by applying a cut off from Danish

1488   controls (mean + 2 STD) to the Luminex data.

1489

1490   **Supplement table 2: Raw read counts by sample for _H. sapiens_, _P. falciparum_, _var_ exon**

1491   **1 and percentage of reads that mapped either to _P. falciparum_ or _var_ exon 1 as well as**

1492   **the number of assembled _var_ contigs >500 bp in length.**

1493

Supplement table 3: Features of the assembled _var_ fragments annotated in accordance

with Rask _et al._ . (Rask _et al_, 2010) and Tonkin-Hill _et al._ (Tonkin-Hill _et al_, 2018). The

reading frame used for translation is given after the contig ID, the position of each annotation

is provided by starting and ending amino acid followed by the p-value from the blast search

against the respective database. For annotations in accordance with Tonkin-Hill _et al._ (Tonkin-

Hill _et al_, 2018) either the short ID or 'NA' (not applicable) is listed at the end. Short IDs are

only available for significant differently expressed domains and blocks between severe and

non-severe cases (Tonkin-Hill _et al_, 2018).

1502

Supplement table 4: Summary of _var_ gene fragments assembled for each patient isolate

showing length, raw read counts, RPKM, blast hits, domain and block annotations in

accordance with Rask _et al._ . (Rask _et al_, 2010). The RPKM for the contigs was calculated

as number of mapped reads and normalized by the number of mapped reads against all

transcript in each isolate, respectively. Therefore, RPKM expression values are only valid to
compare within a single sample since RNA-seq reads were mapped only to the contigs of the
respective patient isolate using BWA-MEM (Li, 2013). Further, the amount of blast hits with
500 bp or 80% of overlap against the ~2400 samples from varDB (Otto, 2019) with an identity
cutoff of 98%. Further hits of 1 kb (>98% identity) against the *var* genes from the 15 reference
genomes (Otto *et al*, 2018a) are listed. The last two column show the annotations from Rask
et al. (Rask *et al*, 2010) associated to each contig.

1514

1515   **Supplement table 5: Log transformed normalized Salmon read counts for assembled**
1516   ***var* transcripts, TPM for collapsed domains and homology blocks from each patient iso-**
1517   **late.** Normalized counts and TPM values calculated for transcripts, domains and blocks with
1518   expression in at least three patient isolates with more than five read counts.

1519

1520   **Supplement table 6: Differently expressed *var* transcripts, domains and homology**
1521   **blocks between first-time infected and pre-exposed patient samples.**

1522

1523   **Supplement table 7: Differently expressed *var* transcripts, domains and homology**
1524   **blocks between severe and non-severe patient samples.**

1525

1526   **Supplement table 8: Data from DBLα-tag sequencing.**

1527

1528   **Supplement table 9: Differentially expressed genes excluding *var* genes (all gene anal-**
1529   **ysis) between first-time infected and pre-exposed patient samples.**

1530

1531   **Supplement table 10: Differentially expressed genes excluding *var* genes (all gene anal-**
1532   **ysis) between severe and non-severe patient samples.**

A

first-time infected   pre-exposed

B

C

severe   non-severe

D

A

B

Binding phenotype	DC	first-time infected – pre-exposed			severe – non-severe		
		Transcripts	Domains	Homology blocks	Transcripts	Domains	Homology blocks
Rosetting/ unknown	DC16		CIDRδ1 DBLβ6	155 (NTSA)		DBLα1.6, DBLα1.5 DBLβ6	
EPCR	DC8, DC15		DBLα1.2, DBLα2 CIDRα1.1, CIDRα1.7 DBLβ6, DBLβ12 DBLα1.7	155 (NTSA)	S_L7_20_Contig5132	DBLα1.2, DBLα2 CIDRα1.1 DBLβ6, DBLβ12	
EPCR & ICAM-1	DC13		CIDRα1.4	584 (DBLβ3)			
CD36 & ICAM-1		S_L8_68_Contig4767			S_L8_68_Contig4767 S_L3_13_Contig2037		
CD36		NS_L1_88_Contig17086 NS_L7_29_Contig31477	NTSB DBLα0.13, DBLα0.22, DBLα0.23 CIDRα2.8, CIDRα2.9, CIDRα6	88 (DBLα0) 69 (ATSB)		DBLα0.23 CIDRα2.4, CIDRα2.9	591 (CIDRβ/γ-ATSB)
unknown	DC3	S_L7_16_Contig3710		155 (NTSA)			
unknown	DC1-3D7	NS_L8_65_Contig2160 NS_L6_18_Contig1987	DBLα1.4 DBLγ15 DBLε5		NS_L6_24_Contig8626 NS_L6_18_Contig1987 S_L8_56_Contig7448 NS_L8_65_Contig2160 NS_L8_59_Contig4567	CIDRα1.3	
	DC1-IT	S_L7_16_Contig1649 S_L8_56_Contig2546 NS_L1_80_Contig36393		155 (NTSA)			
CSA	DC2	S_L7_42_Contig2378, NS_L8_59_Contig2262			S_L7_42_Contig2378 NS_L8_50_Contig1847		
others		NS_L1_88_Contig54310 NS_L7_17_Contig4771 NS_L7_27_Contig7034	DBLβ9 CIDRγ11	557 (DBLβ-DBLγ) 614 (DBLβ12/3) 255 (DBLβ12/3/1/5)	S_L6_19_Contig4243		559 (DBLζ6)

A

B

Table 1: Characteristics of each patient.

ID #	Parasitemia	Symptoms	Classification
1	1%	High fever (40,2°C), liver dysfunction (transaminases ↑↑), developed 7% parasitemia with circulating schizonts upon hospitalization	Naïve, severe
2	1%	Fever (38.5°)	Pre-exposed, non-severe
3	>5%	Fever, chill	Naïve, non-severe
4	1.5%	Fever, chill, headache, joint pain	Pre-exposed, non-severe
5	1.5%	Fever, chill	Pre-exposed, non-severe
6	0.5%	Fever (39°C), diarrhea, headache, liver dysfunction (transaminases ↑↑)	Naïve, non-severe
7	2.5%	Fever, 'malaria-associated symptoms'	Pre-exposed, non-severe
9	7%	Fever, chill, joint pain, headache, liver dysfunction (transaminases ↑↑)	Naïve, non-severe
10	2%	Fever (39°C)	Pre-exposed, non-severe
11	3%	Fever, chill	Pre-exposed, non-severe
12	0.5%	Fever, chill, joint pain, headache, diarrhea, abdominal pain	Pre-exposed, non-severe
13	35%	Fever, multiple organ failure, metabolic acidosis, death	Naïve, severe
14	<1%	Fever, diarrhea	Naïve, non-severe
15	35%	Fever, cerebral and liver dysfunction (transaminases ↑↑)	Naïve, severe
16	8%	Fever, chill, headache	Naïve, non-severe
17	2.5%	Fever, diarrhea, headache	Pre-exposed, non-severe
18	7%	Fever, diarrhea, headache, abdominal pain	Pre-exposed, non-severe
19	0.8%	No clinical data	Pre-exposed, non-severe
20	40%	Fever, diarrhea, cerebral dysfunction, renal failure (creatinine ↑↑↑), anemia	Naïve, severe
21	7%	Fever, headache	Naïve, non-severe
22	3%	Fever (up to 41°C), headache, abdominal cramps	Naïve, non-severe
23	12%	Fever, headache, multiple organ dysfunction (renal: creatinine ↑↑, liver: transaminases ↑, cerebrum: aphasia, ataxia)	Naïve, severe
24	3%	Fever, diarrhea	Pre-exposed, non-severe
25	48%	Fever, multiple organ failure, liver dysfunction (transaminases ↑), lactate acidosis	Naïve, severe
26	7%	Fever, headache, cerebral dysfunction	Naïve, severe
27	0.2%	Fever, diarrhea	Pre-exposed, non-severe
28	3.5%	Fever, chill	Pre-exposed, non-severe
29	8%	Fever, headache, diarrhea	Pre-exposed, non-severe
30	11%	Fever, headache, diarrhea, cerebral dysfunction, metabolic acidosis	Naïve, severe
31	1%	No clinical data	Pre-exposed, non-severe
32	2%	Fever, chill, diarrhea, headache	Pre-exposed, non-severe
33	3.5%	Fever, diarrhea, abdominal pain	Pre-exposed, non-severe

Table 2: Patient data.

	First-time infected (naïve) (n=15)	Pre-exposed (n=17)	Severe malaria (n=8)	Non-severe malaria (n=24)
Female sex [n (%)]	6 (40%)	3 (18%)	3 (38%)	6 (25%)
Patient age in years [median (IQR)]	34 (26–53)	38 (31–45)	47 (27–59)	35 (31–46)
Hb g/dl [median (IQR)]*	13.1 (12.1–14.6)	12.2 (11.8–13.1)	12.1 (11.6–13.0)	13.2 (12.0–14.3)
Parasitemia % [median (IQR)]	7.0 (4.0–23.5)	2.0 (1.0–3.0)	23.5 (10.0–36.3)	2.5 (1.0–3.9)
MSP1 [n (%)]	1: 10 (66%) 2: 1 (7%) 3: 3 (20%) 4: 1 (7%)	1: 12 (71%) 2: 3 (18%) 3: 2 (12%) 4: 0 (0%)	1: 5 (63%) 2: 1 (13%) 3: 1 (13%) 4: 1 (13%)	1: 17 (71%) 2: 3 (13%) 3: 4 (17%) 4: 0 (0%)
Total reads [median (IQR)]	41,341,958 (37,804,417–43,659,324)	41,259,082 (36,921,362–43,904,892)	42,458,431 (38,520,154–49,561,881)	41,050,568 (36,920,201–44,030,863)
P. falciparum reads [median (IQR)]	35,940,843 (34,099,395–39,090,313)	37,065,150 (28,707,096–38,070,441)	37,980,501 (35,195,959–45,563,701)	35,559,157 (29,711,534–37,774,576)
Number of assembled *var* contigs (>500 bp) [median (IQR)]	220.5 (169.3–320.8)	165.5 (121.3–251.5)	292 (210–404)	174 (121–259)
Parasite age [median (IQR)]	9.4 (8.0–10.3)	9.8 (8.0–10.6)	8.2 (8.0–9.8)	9.8 (8.2–11.4)

Sides of infection: Ghana (n=10), Nigeria (n=6), Guinea (n=3), Tansania (n=2), Kongo (n=2), other African countries (n=10), Germany (n=1), unknown (n=1)

* n=21

Table 3: *Var* transcripts up- and downregulated in first-time infected patients.

	Transcript	Length (bp)	Log2 fold change	p_{adj}	Comment	Associated *var* group	Rask domain composition
↑ naïve	NS_L1_80_Contig36393	2,536	23.04	2.28e-11	*var1*	A	DBLγ8-DBLζ1-DBLε5
	S_L7_42_Contig2378#	8,504	4.38	0.0009	*var2csa*	E	NTSpam-DBLpam1-DBLpam2-CIDRpam-DBLpam3-DBLεpam4-DBLεpam5-DBLε10
	NS_L1_88_Contig54310	2,975	6.31	0.0027		B	DBLδ1-CIDRβ1
	S_L7_16_Contig1649	2,244	6.31	0.0050	*var1-IT*	A	NTSA-DBLα1.1-CIDRα1.2
	S_L8_68_Contig4767#	6,006	8.45	0.0068	ICAM-1 binding?	B	DBLβ5-DBLβ5-DBLδ1-CIDRβ1
	NS_L7_17_Contig4771	2,561	2.24	0.0243		B	DBLδ1-CIDRγ5
	NS_L1_88_Contig17086	3,568	5.13	0.0243	CD36-binding	B	CIDRα3.4-DBLδ1-CIDRβ5
	S_L7_16_Contig3710	3,037	3.74	0.0319	DC3, *var3*	A	DBLα1.3-DBLε8
	NS_L8_59_Contig2262	5,189	2.15	0.0323	*var2csa*	E	DBLpam3-DBLεpam4-DBLεpam5-DBLε10
	S_L8_56_Contig2546	804	6.58	0.0339	*var1-IT*	A	NTSA-DBLα1.1
	NS_L7_27_Contig7034	1,410	1.26	0.0366		B	CIDRβ5
	NS_L7_27_Contig6972#	724	4.50	0.0403			not applicable
↓ naïve	NS_L8_65_Contig2160#	7,254	-22.85	2.28e-11	*var1-3D7*	A	NTSA-DBLα1.4-CIDRα1.3-DBLβ1-DBLγ15-DBLε1-DBLγ8
	NS_L6_18_Contig1987#	549	-9.63	0.0027	*var1-3D7*	A	NTSA-DBLα1.4
	NS_L7_29_Contig31477	4,497	-8.92	0.0479	CD36-binding	B/C	DBLα0.20-CIDRα3.1-DBLδ1-CIDRβ6

\# also found in severe cases up-/downregulated

Table 4: *Var* transcripts up- and downregulated in severe cases.

	Transcript	Length (bp)	Log2 fold change	p_{adj}	Comment	Associated *var* group	Rask domain composition
↑ severe	S_L7_42_Contig2378#	8,504	4.79	4.61e-5	*var2csa*	E	NTSpam-DBLpam1-DBLpam2-CIDRpam-DBLpam3-DBLεpam4-DBLεpam5-DBLε10
	S_L7_20_Contig5132	6,276	10.72	0.0007	EPCR- & gC1qR binding	A	DBLα1.2-CIDRα1.6-DBLβ12-DBLγ6-DBLδ1
	S_L8_68_Contig4767#	6,006	7,85	0.0076	ICAM-1 binding?	B	DBLβ5-DBLβ5-DBLδ1-CIDRβ1
	S_L3_13_Contig2037	1,807	8,75	0.0234	ICAM-1 binding?	A/B	DBLβ5
	NS_L8_50_Contig1847	8,254	12.91	0.0319	*var2csa*	E	NTSpam-DBLpam1-DBLpam2-CIDRpam-DBLpam3-DBLεpam4-DBLεpam5-DBLε10
	NS_L7_27_Contig6972#	724	5.09	0.0347			not applicable
	S_L6_19_Contig4243	5,681	7.64	0.0364		A/B	DBLγ17-DBLγ10-DBLδ4-CIDRδ1
↓ severe	NS_L6_24_Contig8626	1,175	-20.35	9.20e-17	*var1-3D7*	A	NTSA-DBLα1.4
	NS_L6_18_Contig1987#	549	-25.45	7.13e-15	*var1-3D7*	A	NTSA-DBLα1.4
	S_L8_56_Contig7448	3,570	-23.55	1.00e-8	*var1-3D7*	A	NTSA-DBLα1.4-CIDRα1.3-DBLβ1
	NS_L8_65_Contig2160#	7,254	-22.46	1.03e-8	*var1-3D7*	A	NTSA-DBLα1.4-CIDRα1.3-DBLβ1-DBLγ15-DBLε1-DBLγ8
	NS_L8_59_Contig4567	10,201	-5.04	0.0234	*var1-3D7*	A	NTSA-DBLα1.4-CIDRα1.3-DBLβ1-DBLγ15-DBLε1-DBLγ8-DBLζ1-DBLε5

also found in naïve cases up-/downregulated

Table 5: *Var* domains up- and downregulated in first-time infected patients.

	Rask domain	Log2 fold change	p_{adj}	Comment	Associated *var* group
↑ naïve	DBLα1.7	3.52	6.36e-8	DC13	A
	CIDRδ1	3.95	0.0003	DC16, rosetting	A
	DBLβ6#	2.45	0.0016	DC15, DC16	A
	DBLα1.2#	2.30	0.0092	DC15	A
	DBLβ9	2.13	0.0092	DC5, PECAM1-binding	A
	CIDRα1.1#	2.38	0.0229	DC8, EPCR-binding	B
	CIDRα1.4	2.28	0.0298	DC13, EPCR-binding	A
	DBLβ12#	1.77	0.0381	DC8, gC1qr-binding	B
	CIDRα1.7	1.64	0.0425	EPCR-binding	A
	DBLα2#	1.86	0.0425	DC8	B
↓ naïve	DBLα0.22	-2.54	0.0035		B, C
	DBLα0.23#	-3.03	0.0035		B, C
	CIDRα6	-1.68	0.0092	DC22, CD36-binding	B, C
	CIDRγ11	-3.11	0.0092		B, C
	NTSB	-1.55	0.0105		B, C
	DBLα0.13	-2.28	0.0216		B
	DBLγ15	-2.77	0.0381	DC1, *var1-3D7*	A, *var1*
	CIDRα2.8	-2.41	0.0425	CD36-binding	B
	CIDRα2.9#	-2.15	0.0425	CD36-binding	B
	DBLα1.4	-2.58	0.0425	DC1/4, *var1-3D7*	A, *var1*
	DBLε5	-1.35	0.0499	DC1, *var1*	A, *var1*

also found in severe cases up-/downregulated

Table 6: *Var* domains up- and downregulated in severe cases.

	Rask domain	Log2 fold change	p_{adj}	Comment	Associated *var* group
↑ severe	DBLβ6[#]	2.87	0.001	DC15, DC16	A
	CIDRα1.1[#]	3.02	0.009	DC8, EPCR-binding	B
	DBLα1.2[#]	2.49	0.022	DC15/(DC8)	A
	DBLβ12[#]	2.19	0.023	DC8, gC1qr-binding	B
	DBLα1.6	2.89	0.023	DC16	A
	DBLα2[#]	2.23	0.042	DC8	B
↓ severe	DBLα0.23[#]	-3.61	0.006		B, C
	CIDRα1.3	-2.09	0.006	DC1, *var1-3D7*	A, *var1*
	CIDRα2.4	-2.31	0.009	CD36-binding	B
	CIDRα2.9[#]	-2.74	0.039	CD36-binding	B
	DBLα1.5	-2.64	0.042	DC16, rosetting	A

also found in naïve cases up-/downregulated

Table 7: *Var* blocks up- and downregulated in first-time infected patients.

	Rask block	Log2 fold change	p_{adj}	Comment	Associated *var* group	Frequence on VarDom 1.0 server
	557	4.33	0.0003	interdomain (DBLβ-DBLγ)	B	5
	584	4,90	0.0003	DBLβ3	A	5
↑ naïve	614	4.59	0.0174	DBLβ (6x DBLβ12, 1x DBLβ3)	A, B/A	7
	155	3.03	0.0427	NTSA	A	37
	255	2.79	0.0427	DBLβ (8x DBLβ12, 6x DBLβ3, 3x DBLβ1, 1x DBLβ5)	A, B/A	18
↓ naïve	88	-1.41	0.0427	DBLα0	B, C	75
	269	-1.26	0.0446	ATSB	B, C	16

Table 8: *Var* blocks downregulated in severe cases.

	Rask block	Log2 fold change	p_{adj}	Comment	Associated *var* group	Frequence on VarDom 1.0 server
↓ severe	591	-23.93	3.90E-12	Interdomain (CIDRβ/γ-ATSB)	B	5
	559	-4.62	0.03999292	DBLζ6	B	5