Expression profiling of the schizont and trophozoite stages of *Plasmodium falciparum* with a long-oligonucleotide microarray

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Abstract

**Background:** The worldwide persistence of drug-resistant *Plasmodium falciparum*, the most lethal variety of human malaria, is a global health concern. The *P. falciparum* sequencing project has brought new opportunities for identifying molecular targets for antimalarial drug and vaccine development.

**Results:** We developed a software package, ArrayOligoSelector, to design an open reading frame (ORF)-specific DNA microarray using the publicly available *P. falciparum* genome sequence. Each gene was represented by one or more long 70mer oligonucleotides selected on the basis of uniqueness within the genome, exclusion of low-complexity sequence, balanced base composition and proximity to the 3’ end. A first-generation microarray representing approximately 6,000 ORFs of the *P. falciparum* genome was constructed. Array performance was evaluated through the use of control oligonucleotide sets with increasing levels of introduced mutations, as well as traditional northern blotting. Using this array, we extensively characterized the gene-expression profile of the intra-erythrocytic trophozoite and schizont stages of *P. falciparum*. The results revealed extensive transcriptional regulation of genes specialized for processes specific to these two stages.

**Conclusions:** DNA microarrays based on long oligonucleotides are powerful tools for the functional annotation and exploration of the *P. falciparum* genome. Expression profiling of trophozoites and schizonts revealed genes associated with stage-specific processes and may serve as the basis for future drug targets and vaccine development.

Background

*Plasmodium falciparum*, a parasitic protozoan, is the causative agent of the most lethal form of human malaria. It is responsible for 300-500 million infections per year in some 90 countries and regions throughout the tropical and subtropical world. Of these clinical cases, approximately 2.1 million result in death annually [1]. In areas where mosquito abatement has failed, chemotherapy, consisting of a limited selection of antimalarial agents, is the only defense against this disease. The increase in drug resistance throughout the malaria endemic regions is cause for great concern and calls for the development of new antimalarial measures, which would involve a larger variety of drug targets as well as a wider array of vaccine strategies (reviewed in [2,3]).
The study of malaria will be greatly helped by the publicly available complete genome sequence of \textit{P. falciparum}. The sequencing project, driven by the Sanger Centre, the Institute for Genomic Research (TIGR), and Stanford University is essentially complete \cite{4}. The sequence of the completed chromosomes are available for download from each sequencing center and from the \textit{Plasmodium} Genome Resource, PlasmoDB \cite{5,6}. Preliminary analysis of the 23 megabase-pair (Mbp) \textit{P. falciparum} genome indicates the presence of approximately 5,400 genes spread across 14 chromosomes, a circular plastid genome and a mitochondrial genome. Strikingly, more than 60\% of the predicted open reading frames (ORFs) lack orthologs in other genomes \cite{4}. This fact underscores the need to elucidate gene function, yet many of the tools that have propelled the study of model organisms remain inefficient or nonexistent in \textit{Plasmodium}. Despite recent improvements in \textit{P. falciparum} transformation techniques, \cite{7} the efficiency of stable transfection under a direct drug selection remains approximately 10\(^{-6}\), making knockout and gene replacement experiments difficult, and genetic complementation strategies nearly impossible. Genome-wide expression profiling by microarray technology provides an easy alternative for the functional genomic exploration of \textit{P. falciparum}.

In organisms ranging from bacteria to humans, expression profiling has proved a powerful tool. Profiling has been used to gain important insights into processes such as development, responses to environmental perturbations, gene mutation, pathogen and host response, and cancer \cite{8-14}. Expression profiling has already been successfully applied to the partial genome sequence of \textit{P. falciparum}, and has been used to characterize the role of previously unannotated genes \cite{15-17}.

Here we present the design and assembly of a long-oligonucleotide \textit{P. falciparum} gene-specific microarray using the currently available genomic sequence generated by the Malaria Genome Consortium \cite{18-20}. During the course of this work, we have developed software, improved by experimental data and an open-source policy, for rapidly selecting unique sequences from predicted ORFs of any genome. Subsequently, we constructed a long-oligonucleotide-based \textit{P. falciparum} microarray, which we used to evaluate changes in the global expression profile between two distinct stages of \textit{P. falciparum} erythrocytic-stage asexual development - merozoite and mid-schizont. The large number of differentially expressed genes detected in this analysis suggests that extensive transcriptional regulation has a major role in the functional specialization of parasite development.

\textbf{Results and discussion}

\textbf{\textit{P. falciparum} ORF predictions}

At the outset of these studies, a total of 27.6 Mbp of \textit{P. falciparum} genomic sequence was obtained from the publicly available sources presented by the Malaria Genome Consortium \cite{18-20} in October 2000. The sequence comprised two completely assembled chromosomes, the complete mitochondrial and plastid genomes, and the sum of all the partial contigs from the remaining chromosomes. ORF predictions were carried out using GlimmerM, a gene-finding tool trained with \textit{P. falciparum} specific sequences \cite{21,22}. Using default parameters, GlimmerM frequently yielded a large number of overlapping predictions (competing gene models) and thus additional filtering of the initial prediction output was required. As slight overprediction of ORFs is generally desirable for the purpose of expression array building, the post-prediction filtering of the GlimmerM output was modified with respect to the process used by the Malaria Genome Consortium \cite{21}. Briefly, individual predictions that overlapped and were on opposite strands or in different reading frames were retained. For competing predictions within a given GlimmerM gene model, ORFs that were extended downstream by at least 300 bp and were within 300 bp of the total size compared to the size of the largest prediction were chosen. In all other cases, the largest predicted ORF was selected. This selection method resulted in 290 ORF predictions for chromosome 2, whereas the Malaria Genome Consortium selected 210 for the same chromosome \cite{21}.

\textbf{ArrayOligoSelector: array element design}

To construct a gene-specific microarray of the \textit{P. falciparum} genome, we designed 70mer oligonucleotide array elements. We chose this length for a number of reasons. Long oligonucleotides are a highly sensitive alternative to PCR products and provide a means to readily distinguish between genes with high degrees of sequence similarity \cite{25}. In addition, the presence of various types of repetitive sequences and highly homologous gene families in the AT-rich \textit{P. falciparum} genome contributes to a high rate of PCR failure \cite{17} and J.L.D., unpublished results). A software program, ArrayOligoSelector, was developed specifically for the purpose of systematically selecting gene-specific long oligonucleotide probes for entire genomes. The latest version and complete source code for ArrayOligoSelector is freely available at \cite{26}.

For each ORF, the program optimizes the oligonucleotide selection on the basis of several parameters, including uniqueness in the genome, sequence complexity, lack of self-binding, and GC content (Figure 1). Similar approaches to oligonucleotide design have previously been described, but...
the exact algorithms, source code, and/or accompanying hybridization data are not available [25,27,28].

ArrayOligoSelector helps ensure complete genome coverage and optimal array hybridization while avoiding several potential problems originating from the peculiar characteristics of the *P. falciparum* genome. The algorithm attempts to minimize cross-hybridization between the oligonucleotide and other regions of the genome. To evaluate the potential for cross-hybridization, early versions of ArrayOligoSelector used a simple BLASTN alignment identity [29]. Although this method prevents the selection of troublesome sequences, it does not take into account the effect of mismatch distribution or base composition. Subsequent versions of ArrayOligoSelector were improved by calculating a theoretical energy of binding between the oligonucleotide and its most probable cross-hybridization target in the genome (‘second best target’). The binding energy (kcal/mol) is calculated using a nearest-neighbor model using established thermodynamic parameters [30-35]. Thus, a sequence with high cross-hybridization potential will have a more stable binding energy with a larger absolute value. In contrast, a sequence unique in the genome will yield a smaller absolute value for the binding energy. A representative plot of the calculated binding energies for all possible 70 bp oligonucleotides from a putative *var* gene (PlasmoDB v4.0 annotated gene ID PF08_0140) is shown in Figure 2a.

An important aspect of oligonucleotide design for microarray hybridization is avoiding secondary structures within the oligonucleotide, as these are likely to be detrimental to hybridization performance. To avoid selecting oligonucleotides with secondary complex structure, ArrayOligoSelector uses the Smith-Waterman algorithm with the PAM47 DNA matrix to calculate the optimal alignment score between the candidate oligonucleotide sequence and the reverse complement of that sequence [24]. A high Smith-Waterman score indicates the potential to create secondary structures (Figure 2b).

The presence of low-complexity sequence could also result in significant nonspecific cross-hybridization. For example, the *P. falciparum* genome contains a large number of low-complexity sequence elements as a result of a high frequency...
of continuous stretches of A and T nucleotides in both the non-coding and the coding regions. ArrayOligoSelector automatically detects such sequences by subjecting candidate oligonucleotide sequences to a lossless compression [36]. The compression score, calculated as the difference in bytes between the original sequence and the compressed version,
is inversely proportional to complexity (Figure 2c). Using this score, repeats of essentially any nature are detected in a computationally efficient manner.

In addition, in order to avoid specific sequence features, ArrayOligoSelector supports filtering based on user-defined patterns. This feature can be used to implement filtering rules based on empirically derived data. Finally, the melting temperature of an oligonucleotide is largely determined by its GC content. As is the case with most ORFs, there exists a large range of %GC values (<10 to >60%) over a 70 bp window (Figure 2d). For this reason, a user-defined %GC target range is used by ArrayOligoSelector such that the majority of the array elements will share a similar base composition and hybridization properties across the array.

Given the above parameters, ArrayOligoSelector evaluates every 70mer sequence within an ORF and chooses an optimal set on the following criteria. The uniqueness-filter requires oligonucleotides to satisfy two simultaneous threshold criteria based on the calculation of the binding energy to their second-best target (the best target is itself). First, the oligonucleotide must rank among the top 5% of the unique or almost unique 70mers in the entire ORF. Second, its binding energy must be within 5 kcal/mol of the best candidate for the ORF. In addition, an optional user-defined energy threshold can operate in conjunction with the default threshold. Initial settings for the low-complexity and the self-binding terms allow the top-scoring 33% of 70mers to pass to the next selection step. Finally, an optional ‘user-defined sequence filter’ simply eliminates the 70mer candidates containing the defined sequences. These four filters operate on the entire set of 70mer candidates for a particular ORF and generate four independent output sets. The intersection of the four outputs is then subjected to the final selection. If no common oligonucleotide is identified in the first intersection, the self-binding and complexity filters are incrementally relaxed until an intersection becomes available. The final selection of candidate oligonucleotides depends upon the %GC filter and 3’-end proximity ranking. Initially, oligonucleotides are allowed to pass if they meet the user-specified %GC. If no oligonucleotide with the desired GC content is found, the target %GC range is relaxed by one percentage point in each direction until one or more oligonucleotides pass. As a final step, a single candidate, closest to the 3’ end of the gene is chosen. Finally, ArrayOligoSelector generates an output file containing the oligonucleotide selections for each putative ORF.

From our initial set of predictions, a total of 6,272 70mer oligonucleotides were selected and synthesized. For our first pass of malaria oligonucleotide selections, the earlier version of ArrayOligoSelector utilizing the BLASTN-based identity threshold was used. The identity cutoff was adjusted to a very conservative value of <30 bp of identity. The initial setting of the GC content filter was set to 28% GC (73°C T_m).

Subsequently, with the release of additional sequence information, a new set of predictions was generated in April 2002 and an additional 1,025 oligonucleotides were selected using the upgraded version of ArrayOligoSelector. In this selection, the user-defined uniqueness threshold was set at -35 kcal/mol, the value at which cross-hybridization is essentially eliminated (Figure 3). The GC content target was set at 28%. The sequence and location of each oligonucleotide is available online [23]. The experiments described in the following section were conducted with the first set of predictions only. As additional annotations become available for the whole genome sequence, additional oligonucleotides will be selected and added to the existing collection. We expect the final set to contain approximately 8,500 oligonucleotides.

**Oligonucleotide performance**

Hughes et al. [25] showed that 60mer oligonucleotides make highly sensitive specific microarray elements for expression profiling of *Saccharomyces cerevisiae* [25]. The oligonucleotides used in that study were synthesized in situ using ink-jet technology whereas the oligonucleotides used in our study were commercially synthesized and subsequently printed using mechanical deposition. Similarly to

![Figure 3](http://genomebiology.com/2003/4/2/R9)

**Figure 3**

Relationship of calculated binding energy to relative hybridization intensity. For each of the oligonucleotides shown in Figure 4, a binding energy (kcal/mol) to the perfect match sequence was calculated using ArrayOligoSelector and plotted against relative hybridization intensity. The Pearson correlation coefficient |r| = 0.91 indicates a strong correlation between the calculated binding energy and hybridization performance.
the experiments of Hughes et al., we wished to test experimentally the effect of mismatches on sensitivity and specificity of 70mer oligonucleotides in the context of a complex hybridization mixture (P. falciparum total RNA). Ten separate malaria ORF predictions were arbitrarily selected for analysis and for each of these ORFs a set of ten oligonucleotides were synthesized. The first oligonucleotide in each set represents the original 70mer selection from ArrayOligoSelector. Each successive oligonucleotide within a set contains an increasing number of mutations made in increments of 10%. Thus, the second oligonucleotide in each set had seven bases (10%) altered, while the last oligonucleotide had 63 bases (90%) mutated. For the first set of five ORFs (Figure 4a-e), which is referred to as the ‘distributed set’, both the position and the identity of each mutation was random. For the second set of five ORFs (Figure 4f-k), referred to as the ‘anchored set’, the mutations in each oligonucleotide were limited to the ends of the sequence. In this manner, a contiguous stretch of perfectly matched bases was always preserved in the center of each oligonucleotide.

Figure 4 summarizes normalized hybridization intensities of control oligonucleotides obtained from the global gene-expression comparisons between trophozoite and schizont stages. The results originate from the six microarray hybridizations presented in Figure 5 and 10 additional hybridizations available as additional data files [23]. The resulting hybridization intensity measurements for each oligonucleotide were averaged across all hybridizations and scaled as a fraction of the average intensity of the perfect-match oligonucleotide for each set. As is evident from Figure 4a-e, the presence of internal mismatches (bubbles and bulges) had a large effect on hybridization performance: oligonucleotides with 10% mismatches (7 bases) suffered an average reduction of 64% in hybridization intensity when compared to the perfect match, while oligonucleotides with 20% (14 bases) or more mismatches were reduced by an average of 97% (Figure 4f). For the anchored set (Figure 4g-k), a more gradual hybridization trend was observed. Mutating the terminal 14 bases (7 bases at each end) resulted in an average loss of 49% of the maximal hybridization intensity. Not until 42 bases had been mutated (21 bases at each end) did the relative intensity of hybridization drop by an average of 97.5% (Figure 4l). In agreement with the findings of Hughes et al. [25], the data from the anchored set of oligonucleotides reveal a strong relationship between the length of contiguous match (the equivalent of oligonucleotide length) and overall hybridization performance.

To measure the extent to which the energy calculation implemented in ArrayOligoSelector matches reality, we have plotted in Figure 3 the calculated energy of the 100 control oligos shown in Figure 4 and their relative intensities of hybridization. The calculated energy and relative intensity of hybridization correlate at $|r| = 0.91$. For comparison, the relative intensity of hybridization and number of nucleotide identities correlate at $|r| = 0.72$. This indicates that a calculated binding energy approach may be used to estimate the potential for cross-hybridization for any sequence relative to the rest of the genome. The specificity for each oligonucleotide is uniquely and computationally determined and expressed as a binding energy (kcal/mol).

To further address the question of specificity of oligonucleotide hybridization to their targets in a complex sample we introduced a set of probes targeting a set of 19 non-repetitive sequences from S. cerevisiae to the microarray. To control for the nucleotide bias of the malaria genome relative to yeast, the selection criteria for this set were identical to selection of the plasmodial microarray elements. The average GC content of the S. cerevisiae oligonucleotides was 31.5%, whereas the average GC content of plasmodial oligonucleotides is 32.5%. The average signal-to-background ratios across all hybridizations for these negative control spots was less than twofold, which is well below the conservative fivefold signal-to-background threshold used to filter data (see Materials and methods). In addition, a series of 10 hybridizations was carried out where total RNA from an asynchronous parasite culture was hybridized against PCR products corresponding to the negative control S. cerevisiae sequences. In these hybridizations the yeast PCR fragment hybridized strictly to its cognate sequence, while the average signal to background value for plasmodial elements in the same channel was 1.17±0.05. In no individual case did a plasmoidal element yield a signal greater than 2.3% of the target hybridization signal intensity. The results of these microarray hybridizations are available as additional data files [23].

To assess whether separate oligonucleotides designed to represent the same target gene perform in a similar manner, we examined three distinct situations: elements dispersed over a long single exon ORF (Figure 6a), overlapping oligonucleotides (Figure 6b), and oligonucleotides representing multiple exons of a single gene (Figure 6c). In each case we observed consistent oligonucleotide performance.

**Gene-expression profiling of trophozoites and schizonts**

We chose a direct comparison of the trophozoite and schizont stages of the P. falciparum asexual intra-erythrocytic life cycle as a first step toward comprehensively profiling all life-cycle stages of this parasite. The trophozoite and schizont represent two distinct developmental stages within the 48-hour plasmodial erythrocytic life cycle. These stages vary greatly in morphology, biochemical properties, and transcriptional activity (reviewed in [15,37]). The mid-trophozoite stage, 18-24 hours post-invasion, contains a highly transcriptionally active nucleus with abundant euchromatin. In addition, trophozoites are characterized by massive hemoglobin ingestion, intake of nutrients from the surrounding medium,
increasing concentration of cytoplasmic ribosomes and rapid formation of organelles. In contrast, the mid-schizonts, at 36-42 hours post-invasion, are characterized by DNA replication (16-32 copies) and compaction into newly formed nuclei. In addition, maturation of merozoite cells begins at the schizont stage and is characterized by the appearance of merozoite organelles such as the rhoptry and dense granules. The several trophozoite- and schizont-specific genes identified previously provide an excellent source of positive controls for the experiments described below.

For microarray hybridization, total RNA was prepared from synchronized in vitro P. falciparum cultures representing the trophozoite stage and the schizont stage (see Materials and methods). Six independent hybridizations were carried out; in three, the trophozoite-derived cDNA was labeled with Cy3 and the schizont-derived cDNA with Cy5. In the other three hybridizations, the fluorophore assignment was reversed. Of the genes assayed, 854 features displayed a differential expression greater than twofold (Figure 5): 525 showed higher relative transcript

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**Figure 4**
The hybridization performance of long oligonucleotides in relation to internal versus terminal mismatches. Ten putative genes were chosen arbitrarily. Each gene is represented by a series of ten oligonucleotides. The first (rightmost) oligonucleotide in each series represents a perfect match to the target sequence. Mismatches were introduced in increasing numbers (10% steps) into each subsequent oligonucleotide in each series. The pattern of matches and mismatches within each oligonucleotide is shown below each graph, where a black bar indicates a perfect match and a gray bar indicates a mismatch. (a-e) Mismatch positions were randomized (‘distributed set’). (g-k) Mismatch positions were chosen such that a contiguous stretch of perfect-match sequence remained in the middle of the oligonucleotide (‘anchored set’). The average hybridization performance, measured as the normalized total intensity for 16 independent experiments, is plotted for each oligonucleotide. (f) The average plot for a-e; (l) the average plot for g-k.
abundance in trophozoites than in schizonts, whereas 326 had greater relative transcript abundance in schizonts. Linear regression ratios were calculated for each possible pair of microarray hybridizations using the filtered dataset. The correlation between hybridizations with the same Cy3/Cy5 assignment was $r = 0.94 \pm 0.02$, while correlation of

...
hybridizations with the opposite Cy3/Cy5 order was \( r = 0.89 \pm 0.03 \).

**Northern blot hybridizations**

To confirm the microarray results, we examined six genes by northern blot analysis. In the microarray hybridization, the expression levels of two of the selected genes were unchanged (< 2-fold) while four additional genes showed a differential expression between the trophozoite and schizont stage (> 2-fold). An equal mass of total RNA from both the trophozoite and schizont stages was hybridized with PCR-generated DNA probes corresponding to the selected genes. Subsequently, each northern blot was stripped and rehybridized with a probe specific for the 41 kD antigen (p41), fructose-bisphosphate aldolase (PfALDO; PlasmoDB v4.0 ID PF14_0425; Oligo ID M11919_1) [38], as a loading control. While the relative amount of PfALDO transcript differs by more than twofold between trophozoite and schizont stages when equal masses of total RNA are blotted, we found that the relative amount of PfALDO to be essentially equivalent when equal masses of poly(A)+ RNA were used for the northern blot (Figure 7a). The discrepancy between northern blots with total RNA and poly(A)+ mRNA are probably due to changes in the relative amounts of mRNA and ribosomal RNA during the intra-erythrocytic life cycle. The poly(A)+ northern blot measurements agree well with the replicate array hybridizations, in which PfALDO was consistently less than 1.5-fold differentially expressed (Figure 7b). To make northern blot measurements comparable to the normalized expression array ratios, the ratio between the two stages was measured using a phosphoimager and divided by the ratio obtained for the PfALDO control in each case. The normalized ratios of the radiolabel signal were highly consistent with the averaged ratios from the six microarray hybridizations (Figure 7b).

**Biological significance of the gene-expression results**

The genome-wide expression data summarized by hierarchical cluster analysis (Figure 5) resulted in two main gene categories, corresponding to genes differentially expressed between trophozoite and schizont stages. Serving as internal positive controls, a number of previously well-characterized plasmodial genes were detected in both categories. In addition, an evaluation of the homology-based gene identities within these categories revealed several functional gene groups. All data from these experiments are available at PlasmoDB and the DeRisi Lab website [23].
Trophozoite genes

The predominant group of features with elevated expression in the trophozoite stage comprised genes encoding various components of the eukaryotic translation machinery. This group contained 24 and 33 proteins of the 40S and 60S ribosomal subunits (RPS and RPL), respectively.

Figure 7

Northern blot validation of the microarray results. (a) Comparison of total RNA and poly(A)+ RNA for northern analysis. 12 μg total RNA and 2 μg poly(A)+ mRNA from both the schizont (s) and trophozoite (t) stages of \textit{P. falciparum} were blotted and probed with a PCR fragment overlapping the genomic sequence corresponding to oligonucleotide ID M11919_1, which represents the 41kD antigen (p41), fructose-bisphosphate aldolase (PlasmoDB v4.0 annotated gene ID PF14_0425). (b) Total RNA northern blots were probed with PCR probes overlapping the corresponding oligonucleotide elements (gene specific) for each gene. The selected gene set included a probable calcium-transporting ATPase (PlasmoDB MAL13P1.61), plasmodial heat-shock protein (PlasmoDB PF10875w), a member of a family of conserved hypothetical proteins (PlasmoDB PFI0875w) and an ORF with sequence similarity to a sodium-and chloride-dependent taurine transporter (PlasmoDB MAL13P1.130), lactate dehydrogenase (PlasmoDB PF13_0141), and a heat-shock protein homolog of Hsp70-3 (PlasmoDB PF11_0351). The northern blot membranes were stripped and reprobed with the fructose bisphosphate aldolase control probe. Northern blot expression levels were measured by a phosphorimager. For each gene-specific probe, the measured ratio of trophozoite to schizont was divided by the ratio measured for aldolase on the same membrane. The average ratio from six replicate hybridizations on the microarray, including fluorophore reversal, is shown in the lower panel. The ratios express fold enrichment in either the trophozoite stage versus schizont stage (t) or schizont stage versus trophozoite stage (s).
addition, nine orthologs of aminoacyl-tRNA synthetases, and 10 initiation and seven elongation translation factors were detected among trophozoite-specific genes. Several previously identified plasmidal genes were present in this group, including Asp-tRNA synthetase, two plasmodial elongation factors (PfEF1A and PfEF2) and one ribosome-releasing factor, PRF1 [21,39]. Consistent with our findings, PF04-1A has been previously shown to have peak expression during the trophozoite stage [40]. Two additional gene groups whose functions are linked to the process of protein synthesis were present among the trophozoite genes: five DEAD-box RNA helicases, including a close homolog of P. cynomolgi RNA helicases-1 [41] and 23 molecular chaperone-like molecules, including two P. falciparum heat-shock proteins such as PfHSP70 (GenBank accession number M19753) and PfHSP66 (accession number L34028), and a homolog of a DnaJ-domain-containing protein family, DNJ/SISI homolog [42]. These data agree with previous studies that found a group of DEAD-box RNA helicases to be overexpressed during the trophozoite stage in P. falciparum [41]. Along with the genes for the translation machinery a number of genes involved in various steps of RNA synthesis and processing were located among the ‘trophozoite genes’, including 16 ORFs belonging to various RNA polymerase complexes and 11 splicing factors (Figure 5). Two previously identified plasmoidal RNA polymerase components were found in this group, including the largest subunit of P. falciparum RNA polymerase II, PfRNApolIIA (M73770), and a homolog RNApolK (14 kD) [39]. The expression characteristics revealed are also consistent with several previous studies that suggested that the plasmodial transcription and translation machinery is active through the late ring and early trophozoite stage before decaying during the late schizont stage [15,40].

Another functional group of genes that encode enzymes of cellular biosynthetic pathways was distinguished within the trophozoite category. This gene set includes 16 enzymes of carbohydrate metabolism, 10 ORFs likely to be involved in nucleotide metabolism, and 11 ORFs involved in the biosynthetic pathways of several amino acids. Several well-characterized plasmodial genes were identified in this metabolic collection, including P. falciparum lactate dehydrogenase (PfLDH; Q27743), enolase (U00152), triose-phosphate isomerase (PTPI; L01654), glucose-6-phosphate isomerase (PG6PI; J05544), hypoxanthine-guanine phosphoribosyltransferase (PHGPT; X16279) and dihydropterate synthetase (PDHPS; U07706). In addition, a group of 11 proteolytic enzymes potentially involved in hemoglobin degradation was detected among the trophozoite genes; these include a cysteine protease, falcipain-2 (AF25199), a metalloprotease falcisyn, (AF123458), and a member of an aspartic protease family, plasmepsin-2 (L10740). Falcipain-2 and plasmspepin-2 have been the targets of recent drug discovery research [43,44].

Overall, the emergent gene clusters suggest that the trophozoite stage, a central phase of plasmodial intra-erythrocytic development, is characterized by the activation of general cellular growth functions such as transcription, translation and hemoglobin degradation and biosynthesis of basic metabolites.

**Schizont genes**

A large number of ORFs found in the schizont-expressed category correspond to genes previously associated with the various steps by which newly released merozoites invade new host cells. The initial step of this process, adhesion of the merozoite to the surface of an erythrocyte, is facilitated by several classes of proteins exposed on the surface of the parasite. Eighteen ORFs, identical or homologous to proteins associated with the merozoite surface, were present among the schizont-enriched genes. This group included four merozoite surface proteins (MSP): MSP1 (M19753), MSP4, MSP5, (AF033037) and MSP6 (AY007721). Additional members of this group include two ORFs containing Duffy-like binding domains, erythrocyte-binding antigen, EBA 175 (L07755), a putative erythrocyte-binding protein, EBL1 (AF131999), and proteins known to be delivered to the surface from apical organelles, including apical membrane antigen, AMA1 (U65407), and finally two rhoptry-associated proteins (RAP1 (U20985) and RAP2).

Initial attachment of the merozoite is followed by reorientation of the parasite cell with its apical part toward the erythrocyte membrane followed by invagination of the membrane. Previous studies suggested that both steps are facilitated by the action of actomyosin, which requires ATP hydrolysis [45]. Consistent with these findings, we found five proteins previously associated with this process, pf-actin1 (M19146), pf-tyoA (AF255909), and merozoite cap protein-1 (U14189), and two subtilisin-like proteases (PISUB1 and PISUB2 (AJ132422)) differentially enriched in schizonts. Interestingly, one additional homolog of PISUB1 was identified among the schizont genes. Moreover, the expression levels of a set of plasmodial protein kinases were previously found to be augmented during the late stages of the malarial erythrocytic life cycle [15]. Our findings confirm and extend this report: 26 unique ORFs sharing a high to medium level of homology with protein kinases and phosphorylases had elevated mRNA levels during the schizont stage (Figure 5). Two previously identified representatives were present in this set: a cAMP-dependent protein kinase, PIPKac (AF126719), and a plasmodial serine/threonine protein phosphatase, PIPFP (AF126719).

A second functional group of genes with increased expression in schizonts encodes proteins that are thought to function on the periphery of a newly infected erythrocyte at the early stages of asexual development. Representatives include: the genes for ring-infected erythrocyte surface antigen (RESA) (X04572) and several close RESA homologs, CLAG9

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The related gene CLAG3.1, and two members of the serine-repeats rich protein (SERA) family [21]. In addition to these well-characterized surface proteins, the schizont-enriched set of transcripts contained a number of ORFs identical or homologous to proteins recognized by antibodies present in plasmodium immune sera obtained either from model organisms [46] or from acute and/or convalescent patients [47]. In summary, the schizont stage of plasmodial development featured genes predominantly occupied with the process of merozoite function as well as the advance synthesis of transcripts for proteins that facilitate parasite establishment within the newly infected erythrocyte.

Taken together, these results suggest that the parasite cell in the trophozoite stage is dedicated to cell growth, and the predominant function of the mid-late schizont stage is maturation of the next generation of merozoites. Of particular interest is the large number of ORFs within both categories (39% in trophozoites and 61% in schizonts) with no putative functions assigned. These ORFs have little to no homology to any other known genes and may possibly represent highly specialized functions not likely to be shared outside this family of parasites.

Conclusions
In this study, we present a P. falciparum ORF-specific microarray utilizing 70mer oligonucleotides as individual microarray elements. This approach helped to overcome potential problems originating from low PCR amplification and allowed us to select probes with a high specificity, thereby minimizing potential cross-hybridization. Moreover, the oligonucleotide-selection algorithm allowed a balanced GC content (around 28%) across the entire microarray set, which is significantly higher than the plasmodial genome average, which is 19.4% with 23.7% in coding regions [4].

Application of the ArrayOligoSelector is not restricted to the P. falciparum genome, but is broadly useful for the automated selection of hybridization probes for a range of species. The flexibility of the selection parameters controlling stringency of uniqueness, self-binding, complexity, user-defined filters and GC content, allows the selection of oligonucleotides appropriate for any genome.

Evaluation of results from derivative control oligonucleotides showed that long oligonucleotides could tolerate 10% mismatches; however, alteration of the target sequence by more than 20% eliminated most of the hybridization signal. Therefore, small sequencing errors and natural variation among isolates are not likely to impact on sensitivity. These performance characteristics imply that the array design for this effort can accommodate the study of essentially any P. falciparum strain with a high degree of specificity.

At present, the P. falciparum microarray used in this study consists of approximately 6,000 gene-specific elements corresponding to the majority of the total coding content predicted for the P. falciparum genome. As new sequence and improved gene predictions arise, additional elements will be added to this evolving platform. Moreover, the present oligonucleotide representation could be further extended for investigation of several unusual P. falciparum genetic and transcriptional phenomena, including antisense mRNA transcription [48] and alternative splicing and/or transcriptional initiation [49,50]. This may be achieved by designing exon-specific array features, as well as antisense oligonucleotides. The oligonucleotide collection could also be expanded by sequences corresponding to intergenic genomic regions. Inclusion of such elements was found to be extremely useful for identifying protein-binding DNA regions by chromatin-immunoprecipitation as well as genes not detected by automated gene-prediction algorithms [51].

Within both the trophozoite and schizont categories, large numbers of genes belong to functionally related processes. These include genes encoding ribosomal subunits, multiple factors for transcription and translation, enzymes of biosynthetic and catabolic pathways, or merozoite adherence and invasion machinery. These results are consistent with predictions that a large number of plasmodial genes undergo strict stage-specific transcriptional regulation, and that such (co-)regulation is shared among functionally related genes [15,52]. Naturally, a ‘fine-resolution’ global gene-expression profile including the different steps of the plasmodial life cycle for multiple divergent strains will be necessary to characterize fully the intra-erythrocytic life of the parasite. At present, our laboratory is analyzing a global gene-expression profile of the 48-hour erythrocytic life cycle with 1-hour resolution for three strains of P. falciparum.

In a number of model organisms, high-resolution gene-expression maps have served as extremely powerful tools for discovery and characterization of novel genes as well as exploration of multiple cellular functions [9,11]. The gene-expression maps typically comprise genome-wide expression profiles at a number of different stages of cellular development, profiles of multiple strains and genetic variants, and global expression responses to number of growth perturbations and growth-inhibitory drugs. Following a similar approach in P. falciparum is most likely to provide substantial information about the many ORFs that lack functional annotation. Further understanding of cellular physiology of this parasite including basic metabolic functions and the intricate interactions between the parasite cell and human host immune system will be a key step in uncovering new targets for antimalarial drug discoveries and vaccine development.

Materials and methods
Microarray fabrication
The 70-mer oligonucleotides were synthesized (Operon Technologies, CA), resuspended in 3x SSC to a final concentration...
of 60 pmol/μl, and spotted onto poly-L-lysine-coated microscopic slides, as previously described [53]. All oligo sequences are available at [23].

Cell cultures

*P. falciparum* parasite cells (W2 strain) were cultured as described [54] with slight modifications: 2% suspension of purified human red blood cells in RPMI1640 media supplemented with 0.25% Albumax I (GIBCO/Invitrogen, San Diego, CA), 2 g/l sodium bicarbonate, 0.1 mM hypoxanthine, 25 mM HEPES pH 7.4, and 50 μg/l gentamycin. Cells were synchronized by two consecutive sorbitol treatments on two consecutive cell cycles (a total of four treatments) and harvested at the subsequent trophozoite stage (18-24 h post-invasion) and schizont stage (36-42 h post-invasion). For the trophozoite stage collection, visual inspection of the Giemsa stains show a nearly pure trophozoite population with less than 1% schizonts. For the schizont stage collection, we estimate the amount of ring contamination to be around 3%. The cells were harvested in prewarmed PBS at 37°C, and spun at 1,500 g for 5 min. Cell pellets were rapidly frozen in liquid nitrogen and stored at -80°C.

RNA preparation and microarray hybridization

Total RNA was prepared directly from the frozen pellets of parasitized erythrocytes, where approximately 1 ml of cell pellet was lysed in 7.5 ml Trizol (GIBCO) and RNA was extracted according to the manufacturer’s instructions. mRNA was isolated from total RNA preparations using the Oligotex mRNA Mini Kit (Qiagen, Valencia, CA). For the hybridization experiments, 1.5 μg total RNA was used for first-strand cDNA synthesis as follows: RNA was mixed with a mixture of random hexamer (pdN6) oligo nucleotides and oligo-(dT20) at final concentration 125 μg/μl for each oligonucleotide. The mixture was heated to 70°C for 10 min and then incubated on ice for 10 min. Reverse transcription was started by adding dNTPs to a final concentration of 1 mM dATP and 500 μM each: dCTP, dGTP, dTTP and 5-(3-deoxyadenosinyl)-2'-deoxyuridine-5'-triophosphate, (aa-dUTP) (Sigma), with 150 units of StrataScript (Stratagene, La Jolla, CA). The reaction was carried out at 42°C for 120 min and the residual RNA was hydrolyzed with 1.0 mL EDTA and 0.2 M NaOH at 65°C for 15 min. The resulting aa-dUTP-containing cDNA was coupled to CyScribe Cy3 or Cy5 (Amerham, Piscataway, NJ) monofunctional dye in the presence of 0.2 M NaHCO3 pH 9.0. Coupling reactions were incubated for a minimum of 1 h at room temperature. The labeled product was purified using QIAquick PCR purification system (Qiagen). Hybridizations and final washing procedures were carried out as described [9] with slight modifications. Briefly, the hybridization medium contained 3x SSC, 1.5 μg/μl poly(A) DNA (Pharmacia Biotech, Uppsala), and 0.5% SDS. Hybridizations were incubated at 65°C for 8-16 h. Arrays were washed in 2x SSC/o.2% SDS and then 0.1x SSC at room temperature. The microarrays were scanned with a GenePix Pro 3.0 software (Axon Instruments, Union City, CA). Subsequently, the data were normalized using the AMAD microarray database and subjected to the cluster analysis using the CLUSTER and TREVIEW software, as described [53]. For the CLUSTER analysis, low-quality features and features with a signal level less than fivefold the background were filtered from the initial raw data set, yielding 4,737 elements. Subsequently, features with an arbitrary twofold fluorescence signal difference in at least four experiments were considered. All programs and microarray-related protocols are available online [35].

Probe preparation and northern blot analysis

The northern blot probes were generated by PCR using the following oligonucleotide sequences:

FWD-M11919_1: 5’-TAGAACACAGACAGCTACGAG; REV-M11919_1: 5’-AGTTGGTTTTCTTGCGCTGTG; FWD-M12812_7: 5’-CTGTAGGCTGTATCCTTTCACAG; REV-M12812_7: 5’-GACAATAAATGCCTCATTACAG; FWD-11261_2: 5’-AAATCCTTGTATTGCCTC; REV-11261_2: 5’-GCTCTTTTGTAGCTTTAATGAG; FWD-F5910_2: 5’-AAACACTTGTCTGTCTAT; REV-F5910_2: 5’-GGCAGATTTGCTATATAT; FWD-M38757_7: 5’-TGGAGATATATCATTCCGAAAGTG; REV-M38757_7: 5’-GTAGAAGCTTCAATATAACAGC; FWD-M12812_7: 5’-CTGTAGGCTGTATCCTTTCACAG; REV-M12812_7: 5’-GCTAATGCTTCCATTTCTAT; FWD-Ks44_1: 5’-GCGAAGCTCATAAACAATTCGAG; REV-Ks44_1: 5’-GCTAAAGCGGCGACGTTCT.

Additional data files

The predicted ORFs and GenePix results (GPR) files containing raw data for Figure 5 and from 10 additional hybridizations are available as additional data files with the online version of this paper and from [23]. Data for Figure 5: three hybridizations (1, 2, 3) with trophozoite RNA labeled with Cy3 and schizont RNA labeled with Cy5; Four hybridizations (4,5,6,7) with trophozoite RNA labeled with Cy5 and schizont RNA labeled with Cy3; Additional hybridizations: Six hybridizations (8,9,10,11,12,13) with trophozoite RNA labeled with Cy3 and schizont RNA labeled with Cy5; Four hybridizations (14,15,16,17) with trophozoite RNA labeled with Cy5 and schizont RNA labeled with Cy3.

ORF predictions of August 2000 were predicted from contig sequences available in August 2000, using GlimmerM software. These predictions were used to design the first set of 70mer oligonucleotides and includes genes from the plastid
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20. TIGR Plasmodium falciparum Genome Database (PFDB) [http://www.tigr.org/tdb/d2eb2/plf/htmls].