The transcriptome of the mosquito host *Anopheles gambiae* upon

infection by different selected lines of the microsporidian parasite

Vavraia culicis

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Abstract (198)

Understanding host-parasite interactions is of the utmost importance for disease prediction, prevention and management. Hence, this study assessed the transcriptional response of the primary malaria vector, *Anopheles gambiae*, to infection with several lines of the candidate vector-control parasite, the microsporidian *Vavraia culicis*. These parasitic lines were selected for early or late transmission within this host. Previous studies extensively described them phenotypically, differing in their virulence, infection dynamics and host exploitation. Using RNA sequencing, gene expression profiles were analyzed in *Anopheles gambiae* mosquitoes infected with early-transmission-selected, late-transmission-selected, unselected (*Stock*) *V. culicis* lines and uninfected controls. The results revealed distinct transcriptional changes associated with each parasite line. Early-selected parasites induced a broader immune response than late-selected ones. Differential expression of immune-related genes, including *Toll-interacting protein* and *Protein ERGIC-53* was identified. Additionally, significant changes were observed in pathways related to Golgi membrane function and oxidative stress response, particularly in response to early-selected parasites. These findings highlight the evolutionary pressures shaping host-parasite coevolution and provide insights into how parasite transmission traits can influence mosquito immune response and regulation. This work offers a foundation for future studies on mosquito-microsporidia dynamics and potential applications for vector control strategies, particularly with respect to *Plasmodium* control.

Background (1190)

Mosquitoes encounter numerous parasites throughout their life cycle [1–6]. Among them, one fungal phylum has drawn significant attention from vector biologists due to its potential as a biological control agent against vector-borne diseases [7–9]. This group, Microsporidia, consists of unicellular intracellular parasites that infect a wide range of organisms, from honeybees to humans, and span diverse ecological niches, including soil and aquatic environments [7, 10–14].

Microsporidian species that infect mosquitoes are predominantly aquatic, as they are adapted to target mosquito larvae that feed in water bodies. The most extensively studied species in this group is *Vavraia culicis* (Order: Glugeida, Family: Pleistophoridae). Like other mosquito-infecting microsporidia, *V. culicis* reaches the gut after ingestion, integrating the host gut cells along the extracellular matrix [15]. A key feature of microsporidian spores facilitates this process: the polar tube. This harpoon-like, protein-based apparatus, typically coiled within the spore, enables the parasite to inject the sporoplasm (containing all genetic material) directly into the host cell and start parasite replication. Once the infected cell reaches its maximum capacity of spores, it lyses, releasing the newly formed spores. In *V. culicis*, this cycle repeats multiple times in the gut, followed by the fat

body. These spores will get transmitted into a new host through one of two ways: horizontal

transmission, in mosquitoes that die on water, release the spores, which infect nearby larvae and

restart the infection cycle; Transovum (vertical) transmission, where the laid eggs are covered in

spores and infect other larvae either through cannibalism or contamination of the water body [15].

In a time where strategies to control the public health toll of vector-borne diseases are starting to

come short in efficiency, microsporidian parasites offer a possible solution [7, 8, 16, 17], in particular

V. culicis, with its ease of culture in the lab [18–21] and potentially Microsporidia MB [8, 22].

Several studies have demonstrated that co-infection with *V. culicis* impairs *Plasmodium spp.* (i.e.,

malaria-causing parasite) development within its mosquito vectors, *Anopheles spp.*, and the same has

been hypothesized to vector-borne diseases carried by *Aedes aegypti* (e.g., dengue) [23]. Besides this,

V. culicis also has been shown to improve the efficiency of insecticides against mosquitoes via

restoration of their sensibility to insecticides [16]. Since *V. culicis* is not cleared by *Anopheles*

mosquitoes [24], maintaining its prevalence in mosquito populations. Although a lot has been done to

understand infection by *V. culicis* in *Anopheles* mosquitoes [20, 24–29], research has fallen short on

describing the molecular mechanisms and pathways involved in the infection and response to this

parasite, let alone the underpinnings of its *Plasmodium*-inhibitory effect.

When we consider mosquito infections with *Plasmodium spp.,* we cannot disregard the fact that this

parasite can only be transmitted in mosquitoes that live long enough for the parasite to develop (e.g.,

approximately 20 days for *Plasmodium berghei*). Hence, this setting might also select for co-infecting

parasites that can survive this long within the host. In the case of *V. culicis* a short or a long time

within the host might affect its *Plasmodium*-inhibitory potential. Moreover, despite their small

genome [14, 28, 30], which could hinder the number of genes where mutation can happen, *V. culicis*

has been shown to evolve rapidly [25]. Taking this into consideration, Silva and Koella (2024)

artificially selected *V. culicis* that: i) killed the host *Anopheles gambiae* (Order: Diptera, Family:

Culicidae) faster and therefore spent less time within it (host death and release of spores

approximately at 1 to 7 days of adulthood, hereafter *Early* regime); or ii) allowed the host to live

longer and therefore remained in its system for longer too (host death and release of spores

approximately at 20 to 30 days of adulthood, hereafter *Late* regime). Surprisingly, they found more

time within the host selected the parasite lines of *V. culicis* to be more virulent (i.e., mosquitoes

infected with *Late* parasites had a mean time to death of 15.31 ± 0.22 days; *Early* 18.17 ± 0.01; and

Stock 20.31 ± 0.56), faster parasite development, and higher spore burdens (for day 10 of adulthood:

Late with $239,500 \pm 23,352$; *Early* with $98,080 \pm 14,853$; and *Stock* with $78,900 \pm 11,919$ spores per

mosquito) to throughout their life (which will reflect in higher transmission rates upon death), when

compared to lines selected to spend less time within their host, or the unselected stock laboratory

population of the parasite (hereafter referred to as *Stock*) [25]. These findings reinforce the potential

of this microsporidian species as a biological control agent, as higher replication of *V. culicis* in long-

lived mosquitoes might increase its *Plasmodium*-inhibitory effect. The potential increase of the

inhibitory effect is due to the two main hypotheses for how the inhibitory process takes place, which

are resource competition or collateral immune response, both of which are parasite growth-dependent.

Thus, in this study, I aimed to characterize the response of the host *A. gambiae* to infection with

unselected and selected lines of *V. culicis* through RNA-sequencing (RNA-seq). With this approach, I

had two main aims: i) describe the host response to the unselected *Stock* population of *V. culicis*; and

ii) compare the host response to five *Early* from five *Late* selected lines of *V. culicis* obtained from

Silva and Koella (2024) (Fig. 1). Through the inclusion of these selected lines, variability in parasite

treatments was increased, which may provide insights into which expression patterns are more readily

available or likely to evolve in response to different parasite challenges. An uninfected mosquito

treatment (hereafter *Uninfected)* was also included as a baseline for the *Stock*-infected mosquito

treatment. The time-point selected to assess their transcriptome was day 10 of adulthood for two

reasons: i) previous results have shown all infected mosquitoes, regardless of the parasite treatment,

are actively releasing infective spores in high numbers, and therefore most likely to induce a host

response that can be detected; ii) an assessment of the host response during the early stages of

infection is more likely to show clear differences between the parasites, which would be more

informative. When considering host immunity, mosquitoes do not possess an adaptive immune

system, and instead their immune defences are mediated by innate immunity, with critical pathways,

such as the Toll, Imd and JAK-STAT pathways, involved in parasite recognition and elimination [31,

32].

The findings of this study provide the first insights into the *A. gambiae* transcriptional response to

infection by *V. culicis*, while also demonstrating how parasite evolution drives changes in genome

expression, impacting the ability of the mosquito to respond to infection. Such insights have been

performed in *Aedes aegypti,* a vector of dengue, unveiling a weak immune response, but significant

up-regulation in lectins and CLIP-domain serine proteases [33], both effectors involved in innate

immunity and its signalling [34]. This study complements previous work on these *V. culicis* lines,

providing a more comprehensive understanding of this host-parasite interaction. Moreover, the data

presented here have important implications for a more general understanding of parasite fitness, as

well as mosquito vector competence for vector-borne parasites. Understanding the molecular basis of

interactions is crucial for designing effective vector control strategies that exploit the natural

evolutionary pressures acting on mosquito-parasite systems [35].

Figure 1. Illustration of the experimental workflow and results. Naive *Anopheles gambiae* mosquito larvae were either uninfected or infected with one of three different sets of lines of the parasite *Vavraia culicis,* to create the following four experimental groups: (1) uninfected mosquitoes (*Uninfected*); (2) mosquitoes infected with spores from an unselected *V. culicis* line (*Stock*); (3) mosquitoes infected with spores from a *V. culicis* line previously selected for early transmission from mosquito hosts (*Early*); and (4) mosquitoes infected with spores from a *V. culicis* line previously selected for late transmission from mosquito hosts (*Late*). After larval infection and subsequent

pupation, adult mosquitoes were collected at day 10 post-adult-emergence and pooled to create 10

individuals per selection replicate for RNA extraction (10 females x 5 selection replicates). Then,

differences in host response to the presence/absence of the parasite were assessed. First, the 20 most

and least expressed transcripts in *Stock*-infected mosquitoes, comparatively to *Uninfected* (Table 1),

were identified, as well as the most represented gene ontology (GO) categories and the proportion of

genes DE (Fig. 2). Second, all DE transcripts were presented according to their respective (GO)

biological process. The latter was followed by a heatmap of the GO categories that are

upregulated/downregulated in infection by the *Early* and *Late* parasites, comparatively to the *Stock*

parasite (Fig. 3). Third, selected candidate GO categories from Figure 3 that differed the most

between *Early* and *Late* were compared. The expression of the transcripts from these categories was

then tested for differences between selection regimes (Fig. 4).

Figure 2. Most enriched gene ontology categories in infected mosquitoes compared to

uninfected. Percentage of differentially expressed (DE) transcripts in *A. gambiae* infected with *Stock*

V. culicis, comparatively to *Uninfected* hosts. The plot shows the 50 most enriched gene ontology

categories, the percentage of DE transcripts per category, and the absolute number of transcripts (on

the right to each dot).

Table 1. Most up-and down-regulated genes in mosquitoes upon infection with the *Stock V.*

***culicis* parasite compared to *Uninfected*.** The host transcriptomic profile concerning an unselected

Stock V. culicis parasite during infection was compared to the uninfected host profile. The 20

transcripts with the highest and lowest expression, compared to the control, are presented. Fold

change and *p*-value were calculated using the transcriptional expression of four technical replicates

per treatment. See materials and methods for more details. The *p*-value was adjusted using the

Benjamini-Hochberg correction method for each comparison, controlling for a false discovery rate (p

< 0.05).

#	Gene ID	Gene name	Fold change	Adj. p
1	LOC3291958	Putative uncharacterized protein DDB_G0271606	-1680.95	< 0.001
2	LOC133393242	Uncharacterized	-306.13	< 0.001
3	LOC1281554	Protein CIMAP1D	-250.60	< 0.001
4	LOC1271479	Uncharacterized	-246.11	< 0.001
5	LOC133393316	Uncharacterized	-195.71	< 0.001
6	LOC1281083	Arginine kinase 1	-191.24	< 0.001
7	LOC1273786	Serine/arginine-rich splicing factor 2	-190.72	< 0.001
8	LOC133394019	Nucleosome assembly protein 1-like 1	-164.62	< 0.001
9	LOC133391046	Putative cyclin-dependent serine/threonine-protein kinase DDB_G0272797/DDB_G0274007	-161.63	< 0.001
10	LOC1271272	Uncharacterized	-149.47	< 0.001
11	LOC5667228	Uncharacterized	-133.19	< 0.001
12	LOC1269550	Uncharacterized	-126.11	< 0.001
13	LOC1270351	Uncharacterized	-123.92	< 0.001
14	LOC133391930	Uncharacterized	-121.19	< 0.001
15	LOC1277465	Uncharacterized	-119.77	< 0.001
16	LOC133391419	Uncharacterized	-107.89	< 0.001
17	LOC11175695	Uncharacterized	-107.77	< 0.001
18	LOC1278052	Cytosol aminopeptidase	-103.75	< 0.001
19	LOC1267823	Uncharacterized	-95.91	< 0.001
20	LOC4576455	Dynein regulatory complex subunit 4	-89.14	< 0.001
1	LOC1281665	Uncharacterized	13.30	< 0.001
2	LOC1269260	Stress-activated protein kinase JNK	4.89	< 0.001
3	LOC1280522	Neuropeptide FF receptor 2	4.40	0.039
4	LOC4576208	Protein obstructor-E	4.15	< 0.001
5	LOC1271120	Protein regulator of cytokinesis 1	4.10	< 0.001
6	LOC1279262	CLIP domain-containing serine protease B15	4.10	< 0.001
7	LOC1269964	Chymotrypsin-2-like	3.99	0.007
8	LOC1270108	Uncharacterized	3.97	0.020
9	LOC1281383	Ninjurin-2	3.96	0.002
10	LOC1279267	CLIP domain-containing serine protease B15-like	3.89	0.002
11	LOC1277042	Uncharacterized	3.76	< 0.001
12	LOC133392142	Uncharacterized	3.62	0.030
13	LOC1271472	Uncharacterized	3.62	< 0.001
14	LOC4577422	Uncharacterized	3.43	< 0.001
15	LOC133393335	Uncharacterized	3.39	0.005
16	LOC133393411	Uncharacterized	3.33	0.001
17	LOC1280050	Annulin	3.27	< 0.001

18	LOC1269628	Monocarboxylate transporter 12	3.27	0.011
19	LOC1271676	Uncharacterized	3.18	< 0.001
20	LOC1277959	UDP-glucosyltransferase 2	3.12	< 0.001

a

b   Negative fold change

c   Positive fold change

d

Figure 3. Host transcriptional signatures of infection. (a) Heatmap showing the log-transformed fold change in expression of all detected transcripts (i.e., 9920) by parasite treatment comparison. While the host transcriptome in response to the *Stock* parasite was compared to the *Uninfected* host transcriptome, the response to *Early* and *Late* parasite treatments was compared to the response to the *Stock* parasite. A heatmap of all biological samples can be found in SI Figure S3. **(bc)** Venn diagrams showing the number of differentially expressed (DE) genes (adjusted *p*-value < 0.05) exclusively and commonly found between hosts infected with *Early* or *Late* parasites in comparison to the *Stock* parasite. Negative (b) and positive (c) differential expression are presented. (d) Heatmap illustrating mean fold change in expression per biological processes families of significantly DE transcripts across all replicates of the same selection regime. Gene ontology annotation was performed using PANTHER 19.0 and VectorBase.

Figure 4. Differentially host expressed candidate genes during infection with selected parasites.
Fold difference in host expression of Golgi membrane (GO:0000139), response to reactive oxygen species (GO:0000302) and immune system process (GO:0002376) markers when compared to infection with the *Stock*. *Early* and *Late* selected fold change parasites was then compared to each other using a linear model (see materials and methods for details). Although the expression of all genes presented differed from the infection with *Stock* parasite, only the Toll-interacting and ERGIC-53 transcripts' fold change significantly differed between *Early* and *Late* parasites. *Late* had a lower expression of *Toll-interacting protein* ($df = 1$, $F = 9.70$, $p = 0.01$) and *protein ERGIC-53* ($df = 1$, $F = 7.69$, $p = 0.02$) than *Early* parasites, represented by an asterisk.

Results (814)

In this study, *A. gambiae* transcriptome in response to unselected and selected lines of the microsporidian parasite *V. culicis* was characterized through RNA-seq. The findings suggest infection by each parasite treatment resulted in distinct transcriptomic profiles, which differed in their host immune response to each parasite. Moreover, the involvement of several biological processes, namely Golgi apparatus-associated signalling and response to reactive oxygen species (ROS) were also observed. The following sections describe and discuss these and other findings in detail.

Transcriptional profiles were measured on day 10 of mosquito adulthood, as a previous study has identified this as the time when all *V. culicis* within the infected mosquitoes reach 100% prevalence in spore production, regardless of the *V. culicis* line infecting them [25]. Hence, this time point was chosen as sporulation represents an increase in parasite numbers and, therefore, should elicit a more evident host response to the parasite. Nevertheless, it is noteworthy that, on average, infection by different lines of the parasite reached 100% prevalence at slightly different days of adulthood - i.e., *Stock* at day 10, *Early* at day 8 and *Late* at day 4.

Across the twelve samples (i.e., one *Uninfected*, one *Stock,* infected with unselected *V. culicis*, five *Early*, infected by five selection replicates for early transmission, and five *Late*, infected by five selection replicates for late transmission), 20 to 38 million reads were obtained, covering over 9,900 genes. The reads were successfully aligned to the *A. gambiae* genome (SI Fig. S1), and quality control confirmed their validity (SI Fig. S2). The number of genes detected by regime, selection, and technical replicates can be found in the SI Table S1.

First, the host transcriptional profile of *A. gambiae* infected with the *Stock* parasite was compared to *Uninfected* controls (Fig. 2, Table 1). Out of the 9,900 annotated genes, 863 (8.7%) were significantly down-regulated, 1104 (11.1%) were significantly up-regulated, and 7953 (80.2%) showed no differential expression (DE). A complete list of the transcripts for this comparison can be found in SI

Files S1-S11. Next, the focus turned to comparing to *Uninfected* controls. The 20 most up- and down-
regulated transcripts are listed in Table 1.

For this comparison, although a higher number of up-regulated genes was found, the down-regulated
genes exhibited higher fold changes. For instance, the most down-regulated gene was the *Putative*
uncharacterized protein DDB_G0271606, with a negative fold change of 1680.95. In contrast, the
most up-regulated gene was an uncharacterized gene with a positive fold change of 13.30.

Next, the DE profiles of *A. gambiae* infected by *Early* and *Late* parasite lines were also analyzed and
compared to *Stock* (Fig. 3). A heatmap showing the expression of all transcripts across the different
parasite lines shows a high fold-change in selected parasites (i.e., *Early* and *Late*) compared to a
standard infection by the *Stock* (even when this is compared to the *Uninfected* treatment) (Fig. 3a).
Overall, Figure 3a demonstrates distinct expression profiles across the different parasite treatments.
After comparing transcript expression to the reference *Stock*, transcripts that were uniquely DE in
each selected treatment and those shared between treatments were identified (Fig. 3bc).

For down-regulated genes, 81 genes were poorly expressed by *Early* parasites. In comparison, 71
genes were down-regulated by *Late* parasites, with 28 genes commonly down-regulated in both
treatments (Fig. 3b). For up-regulated genes, 113 genes were up-regulated in *Early* parasites,
compared to 52 in *Late* parasites, with 77 genes commonly up-regulated in both groups (Fig. 3c).
These findings suggest that a higher number of DE genes are found in response to infection by *Early*
parasites compared to late-selected ones, at least during the 10th day of host adulthood. However, a
notable overlap exists between both treatments (Fig. 3bc).

After categorizing the DE genes into their respective biological processes using Gene Ontology (GO)
categories, a heatmap was plotted to show the mean fold-change for each GO category across parasite
treatments (Fig. 3d). Most categories exhibited similar fold changes between *Early* and *Late* parasites.
However, three categories stood out as potential candidates for explaining differences between the
two selected treatments due to their distinct expression patterns and biological relevance: response to
reactive oxygen species (ROS), Golgi membrane, and immune system processes.

Genes with DE and their respective fold-change relative to the *Stock* were plotted for each candidate
GO category (Fig. 4). The Golgi membrane category had the highest number of DE genes and showed
the greatest variation between *Early* and *Late* parasites, followed by the immune process and response
to ROS categories. Fold-change differences between the selected treatments were also compared.
Only two genes showed significant differences: *Protein ERGIC-53* ($df = 1$, $F = 7.69$, $p = 0.02$; Golgi
membrane) and *Toll-interacting protein* ($df = 1$, $F = 9.70$, $p = 0.01$; Immune process). Both genes had
a higher fold change in *Late* parasites than in *Early* ones. However, while *Protein ERGIC-53* was up-
regulated, *Toll-interacting protein* was down-regulated (Fig. 4).

Discussion (1799)

The aim of this study is to provide a brief foundational description of the *A. gambiae* response to the
microsporidian parasite *V. culicis*. To accomplish this, the host response was described and discuss in
respect to infection by: i) an unselected microsporidian *V. culicis* (*Stock*), as well as ii) differently
selected lines of the same microsporidian species, five selected for early transmission (*Early*) and five
for late transmission (*Late*) within the same mosquito host (Fig. 1). The findings discussed below
provide key insights into this host-parasite interaction and might have consequences for this vector's
competence upon co-infection with *Plasmodium spp*.

1. Mosquito response to infection with unselected Stock of V. culicis

When the host's transcriptomic response to infection by *Stock* was compared (Fig. 2, Table 1), a large
number of differentially expressed (DE) genes were observed (see File S2 for the complete list of
transcripts and respective expression). Such a broad response has been reported in the mosquito *Aedes
aegypti* when infected with *V. culicis* [33]. Nevertheless, only a small fraction of the host genome was
DE, with unclassified genes leading the numbers, followed by genes associated with biological and
cellular processes and several other metabolic gene ontology categories (Fig. 2). Overall, metabolic
and biosynthetic processes were some of the most well-represented, which likely suggests changes in
the transcriptional response linked to the energy processing and translation of immune- and non-
immune elements. From the immunity-related elements, only a few DE markers were identified, with
two being notably over-expressed: *Cecropin-B*, an antimicrobial peptide known for its fungicidal
properties *in vitro* [36, 37], and *Ninjurin-2*, a non-apoptotic cell death protein induced by the Toll
pathway and which plays a key role in the dipteran response to microbial infections [38, 39].

Significant over-expression of regeneration markers was also detected. For instance, *Annulin* was
found to be responsible for peritrophic matrix formation in the gut and early resilience to infection
[40], and *Obstructor-E*, a member of a multigene family, was shown to regulate insect cuticle
development [41]. Additionally, *Rab11*, essential for circulatory iron internalization, was over-
expressed [42], further validating previous findings on the role of iron during infection with *V. culicis*,
where iron sequestration by the parasite has been hypothesized [43].

An overall down-regulation of sensory genes was observed, including *Prestin*, which increases gut pH
[44] and *CIMAP1D*, involved in sound-sensory function [45]. *Arginine kinase 1*, known to participate
in ATP-to-ADP conversion and consequent energy release in certain bacteria [46], was found to be
severely down-regulated in infected mosquitoes (Table 1). Interestingly, most Mucin proteins were
up-regulated, except for *Mucin-6*, which was down-regulated. Mucin proteins are recognized for their

277 role in epithelial lubrication, immune modulation and inflammation [47, 48]. Notably, *Mucin-6* is

278 exclusive to the *A. gambiae* gut, which protects against parasites such as *Plasmodium spp* [49].

2. Mosquito response to infection by selected lines of V. culicis

Next, the focus is turned to infections caused by parasite lines that have undergone selection for early

or late transmission within *A. gambiae* [25] (Fig. 3 and 4). As expected, fewer genes were responsible

for differences between selected and unselected parasites, i.e., 299 for *Early* and 228 for *Late* (Fig.

3cd), than *Stock* and *Uninfected*, i.e., 1967 genes. Notably, a broader transcriptional response was

induced by *Early* parasites than by *Late* ones (when they are compared to the *Stock*), as indicated by

the high number of DE genes in both selection treatments (Fig. 3cd). Despite this, a core set of genes

was found to be commonly up- or down-regulated across both parasite treatments (Fig. 3bc).

To better understand the differences in transcriptomic profiles from infections with selected parasites,

DE genes were categorized into biological processes using GO analysis (Fig. 3d). Most of the

differences between *Early* and *Late* parasites were grouped into three GO biological processes: Golgi

membrane, immune processes, and response to reactive oxygen species (ROS). These findings are

unsurprising, given the biological relevance of these categories for infection [50–52], particularly

from the host's perspective. Equally interestingly, the distinct host transcriptomic profile observed in

response to infection between the *Early* and *Late* parasites indicates that selection did affect not only

phenotypic traits, as shown in previous studies [25, 29, 43], but also drove significant parasite

evolution, causing a substantial shift in the host's response to the selected parasites (Fig. 3 and 4).

The three GO categories were treated as candidate categories, along with their associated genes,

which are believed to play crucial roles in modulating host-parasite dynamics and therefore warrant

further study (Fig. 4). First, the Golgi membrane was considered, as it has recently gained attention

for its critical role in innate immune signalling, preceding other organelles such as mitochondria and

the endoplasmic reticulum (ER) during infection [50]. As a result, the apparent difference observed in

the host response to *Early* or *Late* parasite treatments was unsurprising (Fig. 3). The genes associated

with the Golgi membrane include coatomers, galactosyltransferases, and mannosidases, which are

involved in vesicle transport and carbohydrate metabolism with the ER. The richness and differential

expression of Golgi membrane-associated genes suggest that some upregulation of effectors involved

in glycosylation and trafficking processes in the Golgi apparatus occurs across both parasite

treatments (Fig. 4). One gene, *ERGIC-53*, stood out due to its significantly different fold-change

between host responses to infections by *Early* and *Late* parasites. The protein encoded by *ERGIC-53*

is involved in glycoprotein trafficking. The *Early* parasites seem to induce higher production of

ERGIC-53 and, consequently, this suggests an increased investment in glycoprotein trafficking

compared to *Late* (Fig. 4). Although the precise role of ERGIC-53 in the immune response of *A.*

gambiae remains unclear, it can be hypothesized, based on existing literature, that glycoproteins (e.g.,

peptidoglycan recognition proteins or PGRPs) serve as essential information markers for the host, vital for accurate recognition and elimination of parasites [53, 54]. Nevertheless, it is important to mention further studies have to be conducted to validate such claims. Statistical changes in these effectors do not mean they are themselves involved in the process/changes but instead might be the pathway they are in.

Second, variation in the expression of immune-related genes was analyzed, including *peptidoglycan-recognition proteins* (PGRPs), *CLIP domain serine proteases* and *cecropins*, among others (Fig. 3). A notable difference was observed between the host response to *Early* and *Late* parasite treatments, with many of the DE immune related-genes being downregulated upon infection with *Late* parasites (Fig. 3). Interestingly, both *cecropin-B* and *cecropin-C* were overexpressed in response to early-selected parasites compared to *Stock*, suggesting a more robust antimicrobial response. However, this pattern was not observed in response to *Late* parasites, but rather the opposite. In particular, *CLIP domain serine proteases* have been associated with mosquito late resistance to *Plasmodium* via melanisation [55, 56] (Table 1, Fig. 4). In this study, I noticed all lines of *V. culicis* induced an up-regulation of these immune effectors, suggesting that infection by this microsporidian might elicit an immune response that has collateral damage to *Plasmodium* development.

Although it was hypothesized that the immune response to *Late* parasites would be stronger due to their higher virulence, time, and replication within the host, the data might suggest otherwise. Two hypotheses are proposed to explain this finding, although they are merely speculative at this stage.

The first hypothesis involves the recognition of *V. culicis* as a virulent parasite. As discussed in the manuscript, *Early* parasites induce a larger Golgi membrane response, potentially leading to better recognition of PGRPs than infection by Late parasites. This could explain the rapid up-regulation of immune markers in response to *Early* parasites. The second hypothesis concerns the balance between the costs and benefits of the immune response and its association with the presence/absence of damage-associated molecular patterns (DAMPs) [57]. As with any other life history trait, investment in immunity is expected to trade off with other physiological traits [58]. As suggested by Rolff and Lazzaro (2012), the type of intensity of the immune response may result from the recognition of a combination of molecules, alerting the host to the parasite (PGRPs) and/or to the damage inflicted by the infection (DAMPs). If so, infection by the more virulent *Late* parasite could lead to a more costly and damaging infection, potentially overwhelming the host. In this scenario, the host may prioritize reproductive fitness over survival, which aligns with previous observations of increased reproductive investment during infection by *Late* parasites [25]. This phenomenon has been documented in various infections as predicted by life history theory [58]. In extreme cases, the investment shift to fecundity can be lifelong due to a shortening of the host's survival. In this specific case, we refer to this phenomenon as terminal investment, which may occur when a host invests its resources into fecundity

due to an inability to survive the infection [59–61]. It is unclear if this shift in fecundity and immunity is temporary or observed throughout the host life, given that both fecundity and RNA-seq were performed for the same infection time point and did not represent the whole infection process. It is noteworthy to say the individuals in this study were not blood-fed, necessary for their reproduction. Although we cannot confirm there is a trade-off with fecundity or other life history traits, we are aware *A. gambiae* invests in its first clutch of eggs before having a blood meal, which would represent at least a partial investment in fecundity. Moreover, from this category, the *Toll-interacting protein* was found to have significantly different fold changes between the infection by each of the two selection parasite treatments (Fig. 4). This gene, which produces a protein important for parasite recognition [62–65], was slightly more up-regulated in response to *Early* parasites than to *Late* ones (Fig. 4). This might support the idea that *Early* parasites may be more recognizable by the host's immune system. However, the extent to which parasite selection has affected the presence or absence of specific PGRPs that enable host recognition remains unclear.

It is essential to state the limitations of this study at this stage. Here, the aim is to provide the first description of the mosquito response to *V. culicis* and, potentially, other microsporidia by extension. However, the observations in this study are insufficient to fully understand parasite evolution, nor do they reflect in-depth mechanisms or how they change with the progression of infection, which is particularly difficult in the presence of multiple lines of *V. culicis* with distinct growing cycles. Nevertheless, discussing the potential scenarios is of interest and might pave research avenues for future studies.

Third, only a few genes were found to be DE in response to ROS, showing mixed expression patterns (Fig. 4). Overall, *Early* parasites elicited a stronger host response to ROS than *Late* ones, suggesting higher oxidative stress within the host. This is likely due to a heightened immune response to this infection treatment.

Conclusions (447)

In conclusion, the transcriptional response of *A. gambiae* to infection by *V. culicis* has revealed curious and significant insights into this host-parasite interaction and its evolutionary history. Moreover, the findings in this study further explain many of the results measured in previous studies concerning the effects of selecting *V. culicis* for early and late transmission [25, 29, 43]. Differential expression of genes related to immune response, Golgi membrane function, and oxidative stress were observed, with distinct patterns emerging between infections with reference to *Stock*, *Early* and *Late* parasite lines. Overall, *Early* parasites seemed to elicit a broader and more apparent immune response, characterized by the upregulation of immune-related genes such as cecropin and PGRPs and enhanced

glycoprotein trafficking through the Golgi apparatus. In contrast, *Late* parasites appeared to trigger a

weaker immune response and, therefore, potentially evolved immune evasion mechanisms - but such

claims are still up for testing.

Altogether, the results of this study demonstrate that selection pressures on *V. culicis* not only

influence its phenotypic traits but also drive significant changes in the host gene expression, revealing

complex host-parasite co-evolutionary dynamics. Further investigation is needed to elucidate and

quantify the specific molecular mechanisms involved, particularly in relation to immune strategies

and the balance between host survival and reproductive fitness. Additionally, it is important to

position this study within the context of the microsporidian's ability to interfere with *Plasmodium*

development, as previously mentioned. Vector-borne parasites, such as *Plasmodium spp.*, transmitted

by this host, have long transmission cycles within their vectors. It is reasonable to expect that co-

infection of Plasmodium spp. and microsporidia may inadvertently select for late-transmitted

microsporidian parasites, which could have significant consequences for the spread of both parasites

within a population. Although the mechanism by which microsporidians suppress *Plasmodium*

development remains unclear, our study demonstrated that melanisation-based immunity is

upregulated, which might facilitate late collateral melanisation of *Plasmodium* oocysts. Nevertheless,

this hypothesis is still up for testing, as beside *CLIP proteases*, we observed a limited immune

response that could impact *Plasmodium spp.* and hence, resource competition between the parasites

might be a main source of this inhibitory relationship too. Moreover, although conclusions have to be

taken with some consideration, the results suggest a lack of immunity towards late-transmitted *V.*

culicis might highlight the potential of this parasite as a biological vector-control as it suggests

selection to reduce its prevalence in a population is low. Further research is required to understand

this interaction fully, but our study stands as a foundation for them. These findings provide valuable

insights that could inform future vector control strategies, particularly in the context of the

evolutionary pressures exerted by parasites on mosquito populations and the diseases they vector.

Materials and Methods (1063)

1. Experimental model

The Kisumu strain of the mosquito host *Anopheles gambiae (s.s.)* and evolved lines of the

microsporidian parasite *Vavraia culicis floridensis* were used for these experiments. These parasite

lines had previously been selected for seven generations for: i) early transmission and shorter time

within the host (i.e., *Early*); or for ii) late transmission and longer time within the host (i.e., *Late*); or

iii) they come from an unselected reference, our unselected laboratory stock (i.e., *Stock*) [25]. Further

details on the selection experiment and its maintenance are described in Silva and Koella (2024). All

experiments and rearing were conducted under standard laboratory conditions (26 ± 1ºC, 70 ± 5%

relative humidity, and 12 h light/dark).

2. Experiment and samples collection

Freshly hatched *A. gambiae* larvae were reared in groups of 50 individuals per petri dish (120 mm

diameter × 17 mm) containing 80 ml of distilled water. Larvae were fed daily with Tetramin Baby

fish food according to their age: 0.04, 0.06, 0.08, 0.16, 0.32, and 0.6 mg/larva for 0, 1, 2, 3, 4, and 5

days or older, respectively. On the second day of larval development, individuals were exposed to

10,000 spores per larva of their respective *V. culicis* treatment, or not exposed in the case of the

Uninfected treatment. At pupation, individuals were transferred into 50 ml falcon tubes with

approximately 10 ml of distilled water and were allowed to emerge as adults. Female mosquitoes

were transferred to individual cups with humid filter paper and given a constant 6% sucrose solution

until death. On day 10 of adulthood, all live female mosquitoes were collected, snap-frozen in liquid

nitrogen, and stored at -80ºC until RNA extraction. Transcriptomic profiles were assessed at day 10 of

adulthood, as previous studies on these lines have shown this to be when all hosts produce infective

spores [25]. Thus, a stronger host response and more pronounced signatures of parasite selection were

anticipated for this time point.

3. RNA extractions

RNA extractions were performed using Qiagen's RNAeasy Universal Plus kit following the provided

protocol. Before beginning the protocol, ten female mosquitoes per treatment were pooled in a 1.5 mL

microcentrifuge tube with 1000 µl of Qiazol reagent (from the RNAeasy Universal Plus kit by

Qiagen) briefly mixed, and two stainless steel beads (Ø 2 mm) were added. Each tube was

homogenized using a Qiagen TissueLyser LT at a frequency of 30 Hz for three minutes. The Qiagen

kit protocol was then followed, resulting in a 50 µl eluted solution of RNA in nuclease-free water at a

concentration of approximately 300 ng/µl for each treatment. RNA samples were standardized to 10

ng/µl and delivered on dry ice to the Genomic Technologies Facility (University of Lausanne,

Switzerland) for further processing, as described below.

4. RNA sequencing

RNA quality was assessed using a Fragment analyzer (Agilent Technologies). The poly-A tails and

adapters were removed from single-end reads, and quality was trimmed with Cutadapt version 4.8

[66]. Reads matching ribosomal RNA sequences were removed with fastq_screen version 0.11.1 [67].

Reads longer than 40nt were aligned against the concatenated genomes of *Anopheles gambiae* (NCBI:

GCA_943734735.2) and *Vavraia culicis floridensis* (NCBI: GCA_000192795.1) using STAR version

2.7.10a [68]. The number of read counts per gene locus was summarized with htseq-count version

0.11.2 [69] using gene annotation. RNA-seq libraries were prepared from 250 ng of total RNA with

the Illumina Stranded mRNA Prep reagents (Illumina) using a unique dual indexing strategy and

following the official protocol, which was automated on a Sciclone liquid handling robot

(PerkinElmer). Libraries were quantified by a fluorometric method (QubIT, Life Technologies), and

their quality was assessed on a Fragment Analyzer (Agilent Technologies). Sequencing was

performed on an Illumina NovaSeq 6000 for 150 cycles single read. Sequencing data were

demultiplexed using the bcl2fastq2 Conversion Software (version 2.20, Illumina).

5. Statistical analysis

All the analyses were performed with R version 4.3.0. in RStudio version 3034.06.2+561. The

following packages were used for the statistical analysis and diagnosis, and for visualising and

plotting the data: "Glimma" [70], "cowplot" [71], "ggrepel" [72], "RColorBrewer" [73], "maptools"

[74], "sp" [75], "edgeR" [76], "MASS" [77], "limma" [78], "knitr" [79], "ggplot2" [80] and "dplyr"

[81]. Figure panels were created and edited with BioRender.com and Microsoft PowerPoint 16.16.27.

The high quality of the RNA-seq data alignment was verified using RSeQC version 5.0.1 [82].

Alignment rates to both genomes exceed 93.5% for all samples (SI Fig. S1a). A very few number of

reads aligned to *V. culicis*, ranging from 127 to 5,393 (SI Fig. S1b). *A. gambiae* samples exhibit good

sequencing depth (SI Fig. S1c) and diverse gene representation. The gene body coverage is biased

towards the 3' end, suggesting partial RNA degradation, but this pattern is consistent across all

samples.

Genes with low counts were filtered out according to the rule of 3 count(s) per million (cpm) in at

least 4 sample(s). Library sizes were scaled using TMM normalization and log-transformed into

counts per million or CPM (EdgeR package version 3.42.4) [76]. Statistical quality controls were

performed through density gene expression distribution, clustering, and sample PCA (SI Fig. S2).

Differential expression was computed with limma-trend approach [78] by fitting all samples into one

linear model. Next, the following group of comparisons were performed:

▪ Infected with *Stock vs. Uninfected* (1 comparison).

▪ Infected with selected parasite *vs.* infected with *Stock* (10 comparisons, one for each parasite
line).

A moderated t-test was used for each contrast. The Benjamini-Hochberg method computed an

adjusted p-value for each comparison, controlling for false discovery rate (FDR, Adj. p). Then, a host

gene ontology (GO) enrichment analysis was carried out on differentially expressed (DE) genes

through Panther 19.0 [OBJ] and "vectorbase.org" using the *A. gambiae* genome as a reference (Fig. 2-4).

The host response to different biological replicate lines of *Early* and *Late* were then pooled together

per selection regime, and their average expression, compared to *Stock*, was calculated and plotted.

Afterwards, a candidate gene analysis was performed (Fig. 4) taking into account the differently

expressed biological processes in selected parasites, comparatively to *Stock* (Fig. 3). For each DE

transcript of the GO:0000139 (Golgi membrane), GO:0000302 (Response to reactive oxygen species)

and GO:0002376 (Immune system process) classes, a linear model was run with fold-change as a

dependent variable and parasite treatment as an explanatory variable. These comparisons allow us to

test for differential fold change (compared to *Stock*) for *Early* and *Late* parasites.

Author contributions

490 LMS designed and conducted the experiments, analyzed the data and wrote the manuscript.

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Competing interests

502 The author declare that they have no competing interests.

References

[1] Silva LM, Acerbi G, Amann M, et al. Exposure to Pseudomonas spp. increases Anopheles

gambiae insecticide resistance in a host-dependent manner. *Sci Rep* 2024; 14: 29789.

[2] Ranasinghe HAK, Amarasinghe LD. Naturally occurring microbiota associated with mosquito

breeding habitats and their effects on mosquito larvae. *Biomed Res Int*; 2020.

[3] Dong Y, Manfredini F, Dimopoulos G. Implication of the mosquito midgut microbiota in the

defense against malaria parasites. *PLoS Pathog* 2009; 5: e1000423.

[4] Caragata EP, Tikhe C V, Dimopoulos G. Curious entanglements: interactions between

mosquitoes, their microbiota, and arboviruses. *Curr Opin Virol* 2019; 37: 26–36.

[5]   Duncan AB, Agnew P, Noel V, et al. The consequences of co-infections for parasite

transmission in the mosquito A edes aegypti. *J Anim Ecol* 2015; 84: 498–508.

[6]   Bozic J, Capone A, Pediconi D, et al. Mosquitoes can harbour yeasts of clinical significance

and contribute to their environmental dissemination. *Environ Microbiol Rep* 2017; 9: 642–648.

[7]   Becnel JJ, Weiss LM. *Microsporidia: pathogens of opportunity*. John Wiley & Sons

Incorporated, 2014.

[8]   Herren JK, Mbaisi L, Mararo E, et al. A microsporidian impairs Plasmodium falciparum

transmission in Anopheles arabiensis mosquitoes. *Nat Commun* 2020; 11: 1–10.

[9]   Bjørnson S, Oi D. Microsporidia biological control agents and pathogens of beneficial insects.

Microsporidia Pathog Oppor 2014; 635–670.

[10]   Franzen C. Microsporidia: a review of 150 years of research. *Open Parasitol J*; 2.

[11]   Seatamanoch N, Kongdachalert S, Sunantaraporn S, et al. Microsporidia, a highly adaptive

organism and its host expansion to humans. *Front Cell Infect Microbiol* 2022; 12: 924007.

[12]   Keeling PJ, Fast NM, Law JS, et al. Comparative genomics of microsporidia. *Folia Parasitol*

(Praha) 2013; 52: 8–14.

[13]   Keesey IW, Koerte S, Khallaf MA, et al. Pathogenic bacteria enhance dispersal through

alteration of Drosophila social communication. *Nat Commun* 2017; 8: 265.

[14]   Keeling PJ, Fast NM. Microsporidia: biology and evolution of highly reduced intracellular

parasites. *Annu Rev Microbiol* 2002; 56: 93–116.

[15]   Silva LM, Zeferino TG, Koella JC. Vavraia culicis. *Trends Parasitol*. Epub ahead

of print 25 March 2025. DOI: 10.1016/j.pt.2025.02.011.

[16]   Koella JC, Saddler A, Karacs TPS. Blocking the evolution of insecticide-resistant malaria

vectors with a microsporidian. *Evol Appl* 2012; 5: 283–292.

[17]   Koella JC, Lorenz L, Bargielowski I. Microsporidians as evolution-proof agents of malaria

control? *Adv Parasitol* 2009; 68: 315–327.

[18]   Bargielowski I, Koella JC. A possible mechanism for the suppression of Plasmodium berghei

development in the mosquito Anopheles gambiae by the microsporidian Vavraia culicis. *PLoS*

One 2009; 4: e4676.

[19]   Kelly JF, Anthony DW, Dillard CR. A laboratory evaluation of the microsporidian Vavraia

culicis as an agent for mosquito control. *J Invertebr Pathol* 1981; 37: 117–122.

[20]   Lorenz LM, Koella JC. The microsporidian parasite Vavraia culicis as a potential late life–

acting control agent of malaria. *Evol Appl* 2011; 4: 783–790.

[21]   Vyas-Patel N. The Suppression of Plasmodium berghei in Anopheles coluzzii infected later

with Vavraia culicis. *bioRxiv* 2023; 2002–2023.

[22]   Makhulu EE, Onchuru TO, Gichuhi J, et al. Localization and tissue tropism of the symbiont

Microsporidia MB in the germ line and somatic tissues of Anopheles arabiensis. *MBio* 2024;

549   15: e02192-23.

550   [23]   Bukhari T, Pevsner R, Herren JK. Microsporidia: a promising vector control tool for residual
551   malaria transmission. *Front Trop Dis* 2022; 3: 957109.

552   [24]   Zeferino TG, Koella JC. Host-specific effects of a generalist parasite of mosquitoes. *Sci Rep*
553   2024; 14: 18365.

554   [25]   Silva LM, Koella JC. Complex interactions in the life cycle of a simple parasite shape the
555   evolution of virulence. *bioRxiv* 2024; 2024.01.28.577571.

556   [26]   Sy VE, Agnew P, Sidobre C, et al. Reduced survival and reproductive success generates
557   selection pressure for the dengue mosquito Aedes aegypti to evolve resistance against
558   infection by the microsporidian parasite Vavraia culicis. *Evol Appl* 2014; 7: 468–479.

559   [27]   Vyas-Patel N. Anopheles coluzzii infection by the microsporidian, Vavraia culicis: the effect
560   of host age. *bioRxiv* 2020; 2010–2020.

561   [28]   Khalaf A, Francis O, Blaxter ML. Genome evolution in intracellular parasites: Microsporidia
562   and Apicomplexa. *J Eukaryot Microbiol* 2024; e13033.

563   [29]   Silva LM, Koella JC. Virulence evolution: thinking outside of the host. *bioRxiv* 2024; 2005–
564   2024.

565   [30]   Nakjang S, Williams TA, Heinz E, et al. Reduction and expansion in microsporidian genome
566   evolution: new insights from comparative genomics. *Genome Biol Evol* 2013; 5: 2285–2303.

567   [31]   Yassine H, Osta MA. Anopheles gambiae innate immunity. *Cell Microbiol* 2010; 12: 1–9.

568   [32]   Osta MA, Christophides GK, Vlachou D, et al. Innate immunity in the malaria vector
569   Anopheles gambiae: comparative and functional genomics. *J Exp Biol* 2004; 207: 2551–2563.

570   [33]   Desjardins CA, Sanscrainte ND, Goldberg JM, et al. Contrasting host–pathogen interactions
571   and genome evolution in two generalist and specialist microsporidian pathogens of
572   mosquitoes. *Nat Commun* 2015; 6: 7121.

573   [34]   Kanost MR, Jiang H. Clip-domain serine proteases as immune factors in insect hemolymph.
574   *Curr Opin insect Sci* 2015; 11: 47–55.

575   [35]   Waterhouse RM, Kriventseva E V, Meister S, et al. Evolutionary dynamics of immune-related
576   genes and pathways in disease-vector mosquitoes. *Science (80-)* 2007; 316: 1738–1743.

577   [36]   De Lucca AJ, Bland JM, Jacks TJ, et al. Fungicidal and binding properties of the natural
578   peptides cecropin B and dermaseptin. *Med Mycol* 1998; 36: 291–298.

579   [37]   De Lucca AJ, Bland JM, Vigo CB, et al. D-cecropin B: proteolytic resistance, lethality for
580   pathogenic fungi and binding properties. *Med Mycol* 2000; 38: 301–308.

581   [38]   Broderick S, Wang X, Simms N, et al. Drosophila Ninjurin A induces nonapoptotic cell death.

582   [39]   Thomas T, De T Das, Sharma P, et al. Structural and functional prediction analysis of
583   mosquito Ninjurin protein: Implication in the innate immune responses in Anopheles
584   stephensi. *Int J Mosq Res* 2014; 1: 60–65.

[40]   Wang Z, Wilhelmsson C, Hyrsl P, et al. Pathogen entrapment by transglutaminase—a

conserved early innate immune mechanism. *PLoS Pathog* 2010; 6: e1000763.

[41]   Behr M, Hoch M. Identification of the novel evolutionary conserved obstructor multigene

family in invertebrates. *FEBS Lett* 2005; 579: 6827–6833.

[42]   Wilcke M, Johannes L, Galli T, et al. Rab11 regulates the compartmentalization of early

endosomes required for efficient transport from early endosomes to the trans-golgi network. *J*

Cell Biol 2000; 151: 1207–1220.

[43]   Silva LM, Vallat A, Koella JC. Mechanisms of host exploitation by a microsporidian parasite.

bioRxiv; 2024–12. Epub ahead of print 2024. DOI: https://doi.org/10.1101/2024.12.13.628237.

[44]   Hirata T, Czapar A, Brin L, et al. Ion and solute transport by Prestin in Drosophila and

Anopheles. *J Insect Physiol* 2012; 58: 563–569.

[45]   Senthilan PR, Piepenbrock D, Ovezmyradov G, et al. Drosophila auditory organ genes and

genetic hearing defects. *Cell* 2012; 150: 1042–1054.

[46]   Cozzone AJ. ATP-dependent protein kinases in bacteria. *J Cell Biochem* 1993; 51: 7–13.

[47]   Dam TK, Brewer CF. Multivalent lectin—carbohydrate interactions: energetics and

mechanisms of binding. *Adv Carbohydr Chem Biochem* 2010; 63: 139–164.

[48]   Kufe DW. Mucins in cancer: function, prognosis and therapy. *Nat Rev Cancer* 2009; 9: 874–

885.

[49]   Shen Z, Dimopoulos G, Kafatos FC, et al. A cell surface mucin specifically expressed in the

midgut of the malaria mosquito Anopheles gambiae. *Proc Natl Acad Sci* 1999; 96: 5610–5615.

[50]   Tao Y, Yang Y, Zhou R, et al. Golgi apparatus: an emerging platform for innate immunity.

Trends Cell Biol 2020; 30: 467–477.

[51]   Bahia AC, Oliveira JHM, Kubota MS, et al. The role of reactive oxygen species in Anopheles

aquasalis response to Plasmodium vivax infection. *PLoS One* 2013; 8: e57014.

[52]   Andreyev AY, Kushnareva YE, Starkova NN, et al. Metabolic ROS signaling: to immunity

and beyond. *Biochem* 2020; 85: 1650–1667.

[53]   Davicino RC, Eliçabe RJ, Di Genaro MS, et al. Coupling pathogen recognition to innate

immunity through glycan-dependent mechanisms. *Int Immunopharmacol* 2011; 11: 1457–

1463.

[54]   Sperandio M, Gleissner CA, Ley K. Glycosylation in immune cell trafficking. *Immunol Rev*

2009; 230: 97–113.

[55]   Zhang X, Li M, El Moussawi L, et al. CLIPB10 is a terminal protease in the regulatory

network that controls melanization in the African malaria mosquito Anopheles gambiae. *Front*

Cell Infect Microbiol 2021; 10: 585986.

[56]   Zeineddine S, Jaber S, Saab SA, et al. Late sporogonic stages of Plasmodium parasites are

susceptible to the melanization response in Anopheles gambiae mosquitoes. *Front Cell Infect*

Microbiol; 14. Epub ahead of print 1 August 2024. DOI: 10.3389/fcimb.2024.1438019.

[57]   Lazzaro BP, Rolff J. Danger, microbes, and homeostasis. *Science (80-)* 2011; 332: 43–44.

[58] Stearns SC. Trade-offs in life-history evolution. *Funct Ecol* 1989; 3: 259–268.

[59] Hudson AL, Moatt JP, Vale PF. Terminal investment strategies following infection are
dependent on diet. *J Evol Biol* 2020; 33: 309–317.

[60] Clutton-Brock TH. Reproductive effort and terminal investment in iteroparous animals. *Am*
Nat 1984; 123: 212–229.

[61] Duffield KR, Bowers EK, Sakaluk SK, et al. A dynamic threshold model for terminal
investment. *Behav Ecol Sociobiol* 2017; 71: 1–17.

[62] Burns K, Clatworthy J, Martin L, et al. Tollip, a new component of the IL-1RI pathway, links
IRAK to the IL-1 receptor. *Nat Cell Biol* 2000; 2: 346–351.

[63] Zhang G, Ghosh S. Negative regulation of toll-like receptor-mediated signaling by Tollip. *J*
Biol Chem 2002; 277: 7059–7065.

[64] Didierlaurent A, Brissoni B, Velin D, et al. Tollip regulates proinflammatory responses to
interleukin-1 and lipopolysaccharide. *Mol Cell Biol*.

[65] Valanne S, Wang J-H, Rämet M. The Drosophila toll signaling pathway. *J Immunol* 2011;
186: 649–656.

[66] Martin M. Cutadapt removes adapter sequences from high-throughput sequencing reads.
EMBnet J 2011; 17: 10–12.

[67] Wingett SW, Andrews S. FastQ Screen: A tool for multi-genome mapping and quality control.
F1000Research; 7.

[68] Dobin A, Davis CA, Schlesinger F, et al. STAR: ultrafast universal RNA-seq aligner.
Bioinformatics 2013; 29: 15–21.

[69] Anders S, Pyl PT, Huber W. HTSeq—a Python framework to work with high-throughput
sequencing data. *bioinformatics* 2015; 31: 166–169.

[70] Su S, Law CW, Ah-Cann C, et al. Glimma: interactive graphics for gene expression analysis.
Bioinformatics 2017; 33: 2050–2052.

[71] Wilke CO, Wickham H, Wilke MCO. Package 'cowplot'. *Streamlined plot theme plot Annot*
'ggplot2; 1.

[72] Slowikowski K, Schep A, Hughes S, et al. Package ggrepel. *Autom position non-overlapping*
text labels with 'ggplot2.

[73] Neuwirth E, Neuwirth ME. *Package 'RColorBrewer'*. CRAN 2011-06-17 08: 34: 00. Apache
License 2.0, 2011.

[74] Bivand R, Lewin-Koh N, Pebesma E, et al. Package 'maptools'. *Packag 'maptools'*.

[75] Pebesma E, Bivand R, Pebesma ME, et al. Package 'sp'. *Compr R Arch Netw*; 9.

[76] Robinson MD, McCarthy DJ, Smyth GK. edgeR: a Bioconductor package for differential
expression analysis of digital gene expression data. *bioinformatics* 2010; 26: 139–140.

[77] Ripley B, Venables B, Bates DM, et al. Package 'mass'. *Cran r* 2013; 538: 113–120.

[78] Ritchie ME, Phipson B, Wu DI, et al. limma powers differential expression analyses for RNA-

660   sequencing and microarray studies. *Nucleic Acids Res* 2015; 43: e47–e47.

661   [79]   Xie Y. knitr: a comprehensive tool for reproducible research in R. In: *Implementing*

662   *reproducible research*. Chapman and Hall/CRC, 2018, pp. 3–31.

663   [80]   Wickham H. *ggplot2: elegant graphics for data analysis*. springer, 2016.

664   [81]   Wickham H, François R, Henry L, et al. Package 'dplyr'. *A Gramm Data Manip R Packag*

665   *version*; 8.

666   [82]   Wang L, Wang S, Li W. RSeQC: quality control of RNA-seq experiments. *Bioinformatics*

667   2012; 28: 2184–2185.

Supplementary information

Table S1. Number of genes annotated per sample.

Figure S1. Host RNA-sequencing and parasite gene counts.

Figure S2. Quality control clustering and PCAs.

Figure S3. Heatmap of the complete host transcriptome in response to each parasite treatment and

replicate.

File S1. Transcript expression and statistical contrasts for *Stock* infection *vs. Uninfected*.

File S2. Transcript expression and statistical contrasts for *Early* (selection replicate 1) infection *vs.*

Stock infection..

File S3. Transcript expression and statistical contrasts *Early* (selection replicate 2) infection *vs. Stock*

infection.

File S4. Transcript expression and statistical contrasts for *Early* (selection replicate 3) infection *vs.*

Stock infection.

File S5. Transcript expression and statistical contrasts for *Early* (selection replicate 4) infection *vs.*

Stock infection.

File S6. Transcript expression and statistical contrasts for *Early* (selection replicate 5) infection *vs.*

Stock infection.

File S7. Transcript expression and statistical contrasts for *Late* (selection replicate 1) infection *vs.*

Stock infection.

File S8. Transcript expression and statistical contrasts for *Late* (selection replicate 2) infection *vs.*

Stock infection.

File S9. Transcript expression and statistical contrasts for *Late* (selection replicate 3) infection *vs.*

Stock infection.

File S10. Transcript expression and statistical contrasts for *Late* (selection replicate 4) infection *vs.*

Stock infection.

File S11. Transcript expression and statistical contrasts for *Late* (selection replicate 5) infection *vs.*

Stock infection.