Selection on time to parasite transmission shapes the host *Anopheles gambiae* transcriptional response and suggests immune evasion

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Abstract

Understanding host-parasite interactions is of the utmost importance for the correct 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 prominent vector-control parasite, the microsporidian *Vavraia culicis*. These parasitic lines have been 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 mosquitoes infected with early-selected, late-selected, unselected (reference) *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*, suggests enhanced immune evasion in late-selected parasites. 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 *Plasmodium*.

Background

As with any host, mosquitoes are exposed to and harbour numerous parasites [1–6], including microsporidia [7, 8]. Microsporidia are intracellular parasites that infect various hosts and occupy different ecological niches [7, 9–11]. Still, those infecting mosquitoes are predominantly aquatic, as they typically parasitize the aquatic stages of the mosquito life cycle [2]. These parasites have gained attention in the field of vector-borne diseases due to their potential use as biological agents for controlling *Plasmodium* and its mosquito host, *A. gambiae* [12–15]. Studies in both wild and laboratory settings have demonstrated that microsporidians possess *Plasmodium*-inhibitory potential when co-infecting mosquitoes with this parasite [8, 16].

Vavraia culicis is one of the most extensively studied microsporidians infecting mosquitoes and in the context of malaria research [15–18]. Although *A. gambiae* infections by *V. culicis* tend to result in low costs to adult survival, fecundity and developmental time [18, 19], little is known about the molecular basis of the host response to this microsporidian. In contrast, infections by *V. culicis* in the mosquito *Aedes aegypti*, the vector for dengue, are characterized by a weak immune response but significant investment in lectins and CLIP-domain serine proteases, both effectors involved in innate immunity and its signalling [20, 21]. Understanding how *A. gambiae* responds to microsporidian infection at the molecular level can provide valuable insights into host-parasite dynamics, immunity evolution, and the evolutionary pressures exerted by the parasite. Recent research has highlighted the role of host gene expression in response to pathogens, revealing that parasites can modulate or evade host immune responses to facilitate their transmission and survival [22–25]. In mosquitos, immune defences are mediated mainly by innate immunity, with critical pathways such as the Toll, Imd and JAK-STAT pathways involved in response to parasite recognition and elimination [26, 27] — however, the specific role of these pathways in response to microsporidian infections, like *V. culicis*, remains poorly understood. Therefore, this study sought to start by characterizing the *A. gambiae* host response to *V. culicis* at the molecular level.

Despite their small genome [11, 28, 29], microsporidia evolve rapidly [19]. Recent studies have demonstrated that artificial selection for early or late transmission can shape their virulence and several life history traits [19, 30, 31]. In particular, selection for early or late transmission of *V. culicis* in *A. gambiae* has been shown to influence parasite replication, survival outside the host, and virulence [19, 30]. For example, hosts infected with parasites selected for late transmission experienced the most virulent infections, resulting in shortened life cycles for both the host and parasite, increased host exploitation, and greater parasite growth [19]. Interestingly, late-selected parasites did not survive outside the host as well as the early-selected parasites, which could endure the environment for longer [30]. This trade-off is likely due to a better adaptation to the within-host

environment [32]. Thus, these selected lines of *V. culicis* differ significantly in their performance and

virulence, particularly within the host and potentially offer variability in parasite strategies that can be

examined further.

Considering this, the current study aimed to expand knowledge on the drivers of virulence and

transmission evolution. In addition to the infection by the reference parasite, changes specific to

selection for early or late transmission were explored through RNA sequencing (RNA-seq) of infected

hosts. 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.

Hence, in this study, the transcriptional response of *A. gambiae* to infection by different lines of *V.*

culicis - one selected for early transmission, one selected for late transmission and one unselected

reference - was characterized. RNA-seq was used to analyze the differential expression of genes in

mosquitoes infected or not (Fig. 1). 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 gene expression, impacting the mosquito's ability to respond to infection.

This study complements previous work on these parasite lines, providing a multidimensional

perspective of this particular host-parasite interaction. Moreover, the data presented here have

important implications for parasite fitness and vector competence. Understanding the molecular basis

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

evolutionary pressures acting on mosquito-parasite systems [33].

Figure 1. Illustration of the experimental workflow and results. Naive *Anopheles gambiae* larvae
were infected, or not, with one of three lines of the parasite *Vavraia culicis,* totalizing in the following
treatments: 1) infection with spores selected for early transmission; 2) infection with spores selected
for late transmission; 3) infection with unselected reference spores; 4) uninfected. Adult mosquitoes
were collected at day 10 in pools of 10 individuals per selection replicate for RNA extraction (10
females x 5 selection replicates). We then assessed differences in host response to the
presence/absence of the parasite. First, we compared the 20 most and least expressed transcripts in
reference-infected mosquitoes, comparatively to the uninfected ones (Table 1). Second, we
categorized all DE transcripts according to their respective gene ontology (GO) biological process.
We then presented which GO categories are upregulated/downregulated in infection by the Early- and
Late-selected parasite, comparatively to the reference parasite (Fig. 2). Third, we selected candidate
GO categories from Figure 2 that differ the most between early- and late-selected and we compared
the expression of the transcripts in for each of the categories (Fig. 3) and if they differ between
selected regimes.

**Table 1. Most up-and down-regulated genes in mosquitoes are infected with the reference
parasite compared to those uninfected.** The host transcriptomic profile concerning an unselected
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

135 and methods for more details. The *p*-value was adjusted using the Benjamini-Hochberg correction
136 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

Figure 2. Host transcriptional signatures of infection by selected parasites compared to unselected parasites. 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-selected parasites, in comparison to the unselected reference parasite. Negative (a) and positive (b) differential expression are presented. (c) 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 3. 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 unselected control. 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 unselected parasite, only the Toll-interacting and
ERGIC-53 transcripts' fold change significantly differed between early and late-selected parasites.
Late-selected 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.
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Results

This study aimed to characterize the transcriptional response of *A. gambiae* to different lines of the
microsporidian parasite *V. culicis*. Infections were performed using an unselected reference parasite
and two treatment groups, each with five selection replicates, which had been artificially selected for
early or late transmission within this mosquito host [19]. Our findings suggest infection by each
parasite treatment resulted in distinct transcriptomic profiles, which differed in their host immune
response to each parasite. Moreover, we also observed the involvement of several biological
processes, namely Golgi apparatus-associated signalling and response to reactive oxygen species
(ROS). The following sections describe and discuss these and other findings in detail.
168

169 Transcriptional profiles were measured on day 10 of adulthood. Across the twelve samples, 20 to 38
170 million reads were obtained, covering over 9,900 genes. The reads were successfully aligned to the *A.*
171 *gambiae* genome (SI Fig. S1), and quality control confirmed their validity (SI Fig. S2).
172

First, the host transcriptional profile of *A. gambiae* infected with the reference parasite was compared
to uninfected controls (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. Compared to uninfected controls, the 20 most up- and down-regulated transcripts were
then focused on (Table 1).
179

180 For this comparison, although a higher number of up-regulated genes was found, the down-regulated
181 genes exhibited higher fold changes. For instance, the most down-regulated gene was the *Putative*
182 *uncharacterized protein DDB_G0271606*, with a negative fold change of 1680.95. In contrast, the
183 most up-regulated gene was an uncharacterized gene with a positive fold change of 13.30.
184

Next, the DE profiles of *A. gambiae* infected by early—and late-selected parasite lines were analyzed

and compared to the reference parasite (Fig. 2). After comparing transcript expression to the reference

control, transcripts that were uniquely DE in each selected treatment and those shared between

treatments were identified (Fig. 2ab).

For down-regulated genes, 81 genes were suppressed by early-selected parasites. In comparison, 71

genes were down-regulated by late-selected parasites, with 28 genes commonly down-regulated in

both treatments (Fig. 2a). For up-regulated genes, 113 genes were up-regulated in early-selected

parasites, compared to 52 in late-selected parasites, with 77 genes commonly up-regulated in both

groups (Fig. 2b). These findings suggest that a higher number of DE genes are driven by early-

selected parasites compared to late-selected ones, at least during the 10th day of host adulthood.

However, a notable overlap exists between both treatments (Fig. 2ab).

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. 2c). Most categories exhibited similar fold changes between early- and late-selected

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 reference control were plotted for each

candidate GO category (Fig. 3). The Golgi membrane category had the highest number of DE genes

and showed the greatest variation between early- and late-selected 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-selected parasites than in early-selected ones.

However, while *Protein ERGIC-53* was up-regulated, *Toll-interacting protein* was down-regulated

(Fig. 3).

Discussion

Our study highlights distinct transcriptional changes when infected with different lines of this parasite

(Fig. 1). The findings discussed below provide key insights into this host-parasite interaction and

might have consequences for this vector's competence.

1. Mosquito response to reference V. culicis infection

When the host's transcriptomic response to infection was compared (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* [20]. From this response, only a few DE immune markers were identified, with two being notably over-expressed: *Cecropin-B*, an antimicrobial peptide known for its fungicidal properties *in vitro* [34, 35], 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 [36, 37].

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 [38], and *Obstructor-E*, a member of a multigene family, was shown to regulate insect cuticle development [39]. Additionally, *Rab11*, essential for circulatory iron internalization, was over-expressed [40], further validating previous findings on the role of iron during infection with *V. culicis*, where iron sequestration by the parasite has been hypothesized [31].

An overall down-regulation of sensory genes was observed, including *Prestin*, which increases gut pH [41] and *CIMAP1D*, involved in sound-sensory function [42]. *Arginine kinase 1*, known to participate in ATP-to-ADP conversion and consequent energy release in certain bacteria [43], 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 role in epithelial lubrification, immune modulation and inflammation [44, 45]. Notably, *Mucin-6* is exclusive to the *A. gambiae* gut, which has a protective function against parasites such as *Plasmodium spp* [46].

2. Mosquito response to 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* [19] (Fig. 2 and 3). As expected, fewer genes were responsible for differences between selected and unselected parasites, i.e., 299 for early-selected and 228 for late-selected (Fig. 2ab), than unselected and uninfected, i.e., 1967 genes. Notably, a broader transcriptional response was induced by early-selected parasites than by late-selected ones (when they are compared to the reference control), as indicated by the high number of DE genes in both selection treatments (Fig. 2ab). Despite this, a core set of genes was found to be commonly up- or down-regulated across both parasite treatments (Fig. 2ab).

To better understand the differences in transcriptomic profiles from infections with selected parasites, DE genes were categorized into biological processes using GO analysis (Fig. 2c). Most of the differences between early- and late-selected 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 [47–
49], particularly from the host's perspective. Equally interestingly, the distinct host transcriptomic
profile observed in response to infection between the early and late selected parasites indicates that
selection did affect not only phenotypic traits, as shown in previous studies [19, 30, 31], but also
drove significant parasite evolution, causing a substantial shift in the host's response to the selected
parasites (Fig. 2 and 3).
265
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. 3). 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 [47]. As a result, the apparent difference observed in
the host response to early- or late-selected 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. 3). One gene, *ERGIC-53*, stood out due to its significantly different fold-
change between host responses to infections by early-selected and late-selected parasites. The protein
encoded by *ERGIC-53* is involved in glycoprotein trafficking. Early transmission-selected parasites
seem to induce higher production of ERGIC-53 and, consequently, increased investment in
glycoprotein trafficking compared to late-selected parasites (Fig. 3). 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
[50, 51].
285
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-selected parasite
treatments, with many of the DE immune related-genes being downregulated upon infection with late-
selected parasites (Fig. 3). Interestingly, both *cecropin-B* and *cecropin-C* were overexpressed in
response to early-selected parasites compared to the unselected reference, suggesting a more robust
antimicrobial response. However, this pattern was not observed in response to late-selected parasites,
but rather the opposite. In particular, *CLIP domain serine proteases* have been associated with
mosquito late resistance to *Plasmodium* via melanisation [52, 53] (Table 1, Fig. 3). In our study, we

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-selected parasites would be stronger

due to their higher virulence, time and replication within the host, the data suggests otherwise. Two

hypotheses are proposed to explain this finding. The first hypothesis involves the recognition of *V.*

culicis as a virulent parasite. As discussed in the manuscript, early-selected parasites induce a larger

Golgi membrane response, potentially leading to better recognition of PGRPs than infection by late-

selected parasites. This could explain the rapid up-regulation of immune markers in response to early-

selected 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) [54]. As with any other life history trait, investment in immunity is expected to

trade off with other physiological traits [55]. 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-selected 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-selected parasites [19]. This phenomenon has been documented in various

infections as predicted by life history theory [55]. 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 [56–58]. 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. 3). This gene, which produces a protein important for parasite

recognition [59–62], was slightly more up-regulated in response to early-selected parasites than to

late-selected ones (Fig. 3). This supports the idea that early-selected 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.

Nonetheless, it is reasonable to assume that late-selected parasites might have evolved higher levels of
immune evasion within *A. gambiae*.

Third, only a few genes were found to be DE in response to ROS, showing mixed expression patterns
(Fig. 3). Overall, the early-selected parasites elicited a stronger host response to ROS than late-
selected ones, suggesting higher oxidative stress within the host. This is likely due to a heightened
immune response to this infection treatment.

Conclusions

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 this particular *V. culicis* selection lines for early and late transmission [19, 30, 31].
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 unselected,
early-selected and late-selected parasite lines. Overall, early-selected parasites elicited a broader and
more intense 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-selected parasites appeared to trigger a weaker immune response and, therefore, potentially
evolved immune evasion mechanisms.

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, confirm
and quantify the specific molecular mechanisms involved, particularly in relation to immune evasion
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, the 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

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; or for ii) late transmission and longer time within the host; or iii) they come from an

unselected reference, our laboratory stock [19]. 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 ($26 \pm 1ºC$, $70 \pm 5\%$ relative humidity, and 12 h light/dark).

2. 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 [19]. 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

[63]. Reads matching ribosomal RNA sequences were removed with fastq_screen version 0.11.1 [64].

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

0.11.2 [66] 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" [67], "cowplot" [68], "ggrepel" [69], "RColorBrewer" [70], "maptools"

[71], "sp" [72], "edgeR" [73], "MASS" [74], "limma" [75], "knitr" [76], "ggplot2" [77] and "dplyr"

[78]. 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 [79].

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) [73]. 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 [75] by fitting all samples into one linear model. Next, the following group of comparisons were performed:

- Infected with unselected reference *vs.* uninfected (1 comparison).

- Infected with selected parasite *vs.* infected with unselected reference (10 comparisons).

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, biological processes) enrichment analysis was carried out on differentially expressed (DE) genes through Panther 19.0 ⌗ and "vectorbase.org" using the A. gambiae genome as a reference (Fig. 3 and 4).

Afterwards, a candidate gene analysis was performed (Fig. 4) taking into account the differently expressed biological processes in selected parasites, comparatively to the unselected reference (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 the unselected reference) for early- and late-selected parasites.

Author contributions

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

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Competing interests

The author declare that they have no competing interests.

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Supplementary information

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

Figure S2. Quality control clustering and PCAs.

File S1. Transcript expression and statistical contrasts for reference infection *vs.* no infection.

File S2. Transcript expression and statistical contrasts for early-selected (selection replicate 1)
infection *vs.* reference infection.

File S3. Transcript expression and statistical contrasts for early-selected (selection replicate 2)
infection *vs.* reference infection.

File S4. Transcript expression and statistical contrasts for early-selected (selection replicate 3)
infection *vs.* reference infection.

File S5. Transcript expression and statistical contrasts for early-selected (selection replicate 4)
infection *vs.* reference infection.

File S6. Transcript expression and statistical contrasts for early-selected (selection replicate 5)
infection *vs.* reference infection.

File S7. Transcript expression and statistical contrasts for late-selected (selection replicate 1) infection
vs. reference infection.

File S8. Transcript expression and statistical contrasts for late-selected (selection replicate 2) infection
vs. reference infection.

File S9. Transcript expression and statistical contrasts for late-selected (selection replicate 3) infection
vs. reference infection.

File S10. Transcript expression and statistical contrasts for late-selected (selection replicate 4)
infection *vs.* reference infection.

File S11. Transcript expression and statistical contrasts for late-selected (selection replicate 5)
infection *vs.* reference infection.

a Negative fold change

Early   Late

81   28   71
(45%)   (15.6%)   (39.4%)

b Positive fold change

Early   Late

113   77   52
(46.7%)   (31.8%)   (21.5%)

c

Biological process description

Response to reactive oxygen species
Single-stranded DNA endodeoxyrubinuclease activity
Translation repressor activity, mRNA regulatory element binding
Long-chain fatty acid metabolic process
mRNA 5'splice-site recognition
mRNA splicing, via spliceosome
Endonucleolytic cleave in ITS1 to separate SSU-rRNA from 5.8S rRNA and LSU-rRNA
Regulation of alternative mRNA splicing, via spliceosome
Basal RNA polymerase II transcription machinery binding
Maturation of LSU-rRNA
Damaged DNA binding
Cytoplasmic translation
Maturation of SSU-rRNA from tricistronic tRNA transcript
Structural constituent of ribosome
Ribosomal small subunit assembly
Protein peptidyl-prolyl isomerization
RNA binding
Golgi membrane
Lysozyme activity
Immune system process
Authophagosome activity
DNA binding
Catalytic activity
Nucleic acid binding
Nucleotide binding
Formation of cytoplasmatic translation initiation complex
Actin binding
Transcription coactivator activity
Acyl-carrier activity
RNA polymerase II cis-regulatory region sequence-specific DNA binding
Obsolete activation of MAPKKK activity
RNA polymerase II transcription regulatory region sequence-specific DNA binding

Mean fold change
5.0
2.5
0.0
-2.5
-5.0

Early   Late

Parasite treatment compared to the reference parasite