Title:

Genomic evidence for population-specific responses to coevolving parasites in a New Zealand
freshwater snail

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Abstract

Reciprocal coevolving interactions between hosts and parasites are a primary source of strong

selection that can promote rapid and often population- or genotype-specific evolutionary change.

These host-parasite interactions are also a major source of disease. Despite their importance,

very little is known about the genomic basis of coevolving host-parasite interactions, particularly

in nature. Here, we use gene expression and molecular sequence evolution approaches to take

critical steps towards characterizing the genomic basis of interactions between the freshwater

snail *Potamopyrgus antipodarum* and its coevolving sterilizing trematode parasite, *Microphallus*

sp., a textbook example of natural coevolution. We found that *Microphallus*-infected *P.*

antipodarum exhibit systematic downregulation of genes relative to uninfected *P. antipodarum*.

The specific genes involved in parasite response differ markedly across lakes, consistent with a

scenario where population-level coevolution is leading to population-specific host-parasite

interactions and evolutionary trajectories. We also identified a set of rapidly evolving loci that

present promising candidates for targets of parasite-mediated selection across lakes as well as

within each lake population. These results constitute the first genomic evidence for population-

specific responses to coevolving infection in the *P. antipodarum-Microphallus* interaction and

provide new insights into the genomic basis of coevolutionary interactions in nature.

Keywords: Host-parasite interactions, coevolution, infection, gene expression, F_{ST}

Introduction

Host-parasite interactions constitute a primary source of natural selection and provide a powerful

means of evaluating the evolutionary response to strong selection (1-2). Reciprocal antagonistic

selection for greater host resistance and parasite infectivity can lead to antagonistic coevolution

between host and parasite (3), which can in turn maintain high genetic diversity (4-6) and drive

rapid evolutionary diversification (4, 7, 8).

Host-parasite interactions often have a genetic basis (*e.g.,* 7-9), are linked to rapid

molecular evolution (*e.g.*, 7, 10), and are associated with gene expression changes (*e.g.,* 11-13).

Because most studies providing important insights into the genetic and genomic basis of host-

parasite coevolutionary interactions have been limited to the laboratory (*e.g.,* 11, 14-15; but also

see 16-17), we know very little about the evolutionary genomics of host-parasite interactions in

natural populations (17). As such, elucidating the processes of coevolution in nature is a

requirement both to characterize a primary driver of selection and to formulate targeted strategies

to fight natural parasite populations (*e.g.* those that infect human populations, 18).

Characterization of parasite-mediated changes in host gene expression provides a powerful

means of deciphering the genomic basis of natural coevolutionary interactions because

expression changes are often genotype or population-specific (*e.g.,* 11), can affect host response

to parasites (*e.g.,* 19), and underlie variation in resistance and susceptibility to infection (*e.g.,*

20).

The New Zealand freshwater snail *Potamopyrgus antipodarum* and its sterilizing

coevolving trematode parasite *Microphallus* sp. ("*Microphallus*") provide an especially

compelling context for characterizing the genomic basis of coevolution in natural populations.

First, selection imposed by both host and parasite is extremely strong: infected snails are

completely sterilized (21), and *Microphallus* that cannot infect the first *P. antipodarum* by which

they are ingested will die (22). Second, different *P. antipodarum* populations experience

consistently high *vs*. consistently low *Microphallus* infection frequencies (23-25), indicating that

infection frequency is a major determinant of the strength of parasite-mediated selection and,

thus, the rate and mode of coevolution within each population. Third, *Microphallus* is locally

adapted to *P. antipodarum* at both within- and among-lake scales (26-27), demonstrating the fine

spatial scale of coevolution in this system. Fourth, the existence of many naturally replicated and

separately evolving *P. antipodarum*-*Microphallus* interactions means that each population can be

treated as a separate evolutionary experiment into the consequences of antagonistic coevolution.

Fifth, rapid coevolution of *P. antipodarum* and *Microphallus* has been documented in both a

natural population (28) and in an experimental coevolution study (29). Finally, other snail-

trematode systems, many of which also involve ingestion by and infection of vertebrate final

hosts, are common sources of human and wildlife diseases (30-31), highlighting the translational

relevance of this study.

Multiple studies suggest that coevolutionary interactions between *P. antipodarum* and

Microphallus fit a "matching alleles" infection genetics model, whereby there are no universally

infective parasites or resistant hosts (26-27, 32-33). Rather, *P. antipodarum* susceptibility

depends on whether the genotype of the *Microphallus* individual matches the *P. antipodarum*

individual at loci involved in resistance. Similar matching-allele mechanisms are thought to

operate in other snail-trematode systems such as the laboratory model *Biomphalaria glabrata*-

Schistosoma mansoni (reviewed in 34). Recent studies suggest that both allelic identity (14) and

gene expression (15) are likely involved in mediating *B. glabrata* susceptibility to *S. mansoni*

infection, emphasizing the potentially central role for gene expression in determining outcomes

of host-parasite interactions in this and other snail-trematode systems (30-31).

Here, we take critical steps towards illuminating the genomic basis of coevolution in this

textbook example of antagonistic coevolution (*e.g.*, 35-36) by using RNA-Seq to perform gene

expression and F_{ST} analyses for three replicated samples of *Microphallus*-infected *vs.* uninfected

P. antipodarum from each of three different lake populations. In light of existing ecological

evidence for population structure and local adaptation in this system, we expected to 1) observe

divergence between populations at both the gene expression and genetic level, and 2), that such

divergence should be especially notable with respect to genes likely to be involved in the

coevolutionary response to *Microphallus* infection.

Materials and Methods

Sample collection and dissection. Adult *P. antipodarum* were collected from shallow water

(depth < 1m) habitats of three New Zealand lakes (Alexandrina, Kaniere, Selfe) known to

contain relatively high *Microphallus* infection frequencies (~10-20%, 25; Table S1). Following

transfer to the University of Iowa, we housed snails in 15 L tanks at 16°C with a 16:8 hour

light:dark cycle and supplied *ad libitum* dried *Spirulina* (*e.g.*, 37) until dissection.

Because *P. antipodarum* is polymorphic for reproductive mode and ploidy level (38, 39),

we confined our RNA-Seq analyses to diploid adult non-brooding (non-reproductively active)

females. These criteria ensured that we limited extraneous biological processes that may conflate

differences in gene expression related to interactions between *P. antipodarum* and *Microphallus*.

We dissected each snail to determine sex (male *vs.* female), *Microphallus* infection status

(infected *vs.* uninfected), and reproductive status (brooding *vs.* non-brooding). We used the

infection data to establish infection frequency for each lake sample (Table S1). Because

Microphallus infection fills the body cavity of *P. antipodarum*, we confined analyses to head

tissue. While the use of head tissue necessarily prevents the analysis of genes that are solely

expressed in body tissue, this approach is the only way to ensure that comparable tissue types are

isolated from both infected and uninfected snails. We stored one half of the head tissue in

RNA*later*® until RNA extraction and used the other head half for flow cytometric determination

of ploidy level (following 40; SI Methods). For each RNA-sequencing replicate, we pooled head

tissue from seven snails to obtain a sufficient amount of tissue for RNA extraction and

sequencing. We obtained three biological replicates of parasite-infected and uninfected *P.*

antipodarum from each of the three lakes, for a total of 18 replicates and 126 snails.

RNA sequencing and *de novo* reference transcriptome assembly and annotation. We

extracted RNA following the Invitrogen TRIzol protocol (41), used the Illumina Truseq LS

protocol for cDNA library preparation, and performed 2x100 bp paired-end RNA sequencing on

an Illumina HiSeq 2000 (Illumina, San Diego, CA, 2012) (SI Methods). Next, we used FASTX

Toolkit (42) and FastQC (43) to trim adapters, assess sequencing quality, and filter out low-

quality reads. We used Trinity v. 2.0.4 (44-45) to generate an initial *de novo* assembly, which we

filtered using TransDecoder (45), CD-HIT-EST (46), and Blobology (47) to produce a *de novo*

reference transcriptome containing 62862 assembled transcripts (SI Methods, Table S2).

Using blastx (48) and Blast2GO (49) (SI Methods), we obtained blastx annotations for

10171 transcripts and both blastx and GO annotations for 15797 transcripts. Nearly 75% of our

transcriptome did not receive GO annotations, meaning that the functions of the majority of

genes in our transcriptome cannot yet be determined. This result is unsurprising in light of a

distinct deficit of mollusc genome sequence data available to aid in annotation (50); though

molluscs are the most species-rich animal phylum after arthropods (51), only eight (< 1.3%) of

the 613 sequenced animal genomes available on NCBI as of February 2016 are from molluscs.

RNA-Seq gene expression analyses, functional enrichment, and F_{ST} outlier analyses. We

used Tophat2 (52) to map RNA-Seq reads to our *de novo* reference transcriptome, followed by

Cufflinks, Cuffmerge, and CuffDiff to estimate transcript abundance and quantify expression

differences (53). We then visualized results with cummeRbund (54). We used our *de novo*

transcriptome as our reference for all gene expression analyses. We used Blast2GO to compare

the functions of differentially expressed genes and quantify the number of transcripts annotated

with each GO term. We performed functional enrichment analyses and Fisher's Exact Tests as

implemented in Blast2GO to identify significantly overrepresented functional groups among

differentially expressed genes (SI Methods, Table S2).

We used F_{ST} outlier analyses to identify those genes that were evolving especially

rapidly in infected *vs*. uninfected snails; these genes are candidates for *Microphallus*-mediated

selection. We filtered our reference transcriptome to include only transcripts with fragments per

kilobase per million reads mapped (FPKM) > 0 in all 18 replicates (30685 transcripts) to ensure

each replicate was represented for all loci in the F_{ST} comparisons. We used Tophat2 (52) to map

RNA-Seq reads to the filtered transcriptome and Picard Tools to prepare mapped reads for

variant discovery (http://picard.sourceforge.net). Next, we used Samtools mpileup (55) to call

SNPs from processed bam files, followed by Popoolation2 (56) to calculate F_{ST} per site. We used

IBM SPSS Statistics v. 23 to perform outlier analyses and identify outlier SNPs between infected

and uninfected snails (SI Methods). Finally, we compared levels of genetic differentiation (mean

transcriptome-wide F_{ST}) between infected and uninfected snails within each lake and mean

transcriptome-wide F_{ST} between lakes using Bonferroni-corrected Welch's t-tests as

implemented within IBM SPSS Statistics v. 23.

We used three complementary analytical approaches to quantify gene expression and

genetic differentiation (Fig. S1). First, for the "inclusive analysis," we identified differentially

expressed genes likely to be broadly important for *P. antipodarum* response to *Microphallus*

infection. We quantified expression for each transcript for all pooled replicates under both

conditions (infected and uninfected) to determine expression differences between all infected *vs.*

all uninfected snails (Fig. S1a). We then compared the annotated functions of differentially

expressed genes and performed functional enrichment analyses to determine the types of genes

broadly important for infection response. Next, we calculated mean transcriptome-wide F_{ST} and

performed F_{ST} outlier analyses to measure genetic differentiation between infected and

uninfected snails and identified functional annotations and expression patterns of each transcript

containing one or more F_{ST} outlier SNPs.

Second, for the "within-lake analysis", we compared gene expression patterns between

infected *vs.* uninfected snails from each lake to characterize local (lake level) gene expression

responses to *Microphallus* infection. This analysis included pairwise comparisons between

infected and uninfected snails from each lake population (Fig. S1b). We then compared the

functions of differentially expressed genes between infected and uninfected snails from each

lake. We also calculated mean transcriptome-wide F_{ST} and performed F_{ST} outlier analyses

between infected and uninfected snails within each lake. Finally, we determined the functional

annotation and expression pattern of each outlier-containing transcript.

Third, in the "across-lake analysis", we compared replicates by lake and infection status.

Here, our goal was to use patterns of gene expression within and across lakes to evaluate

evidence for local adaptation and identify genes likely to be locally important for response to

Microphallus infection. We conducted every possible pairwise comparison of gene expression

between all replicates from all three lakes (Fig. S1c), allowing us to differentiate between genes

that were differentially expressed in infected *vs*. uninfected replicates across the three lakes (*i.e.*,

evidence for population-specific infection responses) and genes expressed differently across

lakes regardless of infection status (*i.e.*, expression difference due to lake of origin rather than

infection status). Annotation and functional enrichment analyses allowed us to identify putative

functions of genes that were significantly differentially expressed between infected and

uninfected snails in more than one lake population *vs*. within a single lake (SI Methods). We

visualized these comparisons with Euler diagrams generated with eulerAPE v3 (57). Finally, we

evaluated whether transcripts containing F_{ST} outliers between infected and uninfected snails

contained outliers in multiple lake populations *vs*. only a single lake population.

Results

Inclusive analysis: Greater downregulation of transcripts in infected snails. We identified

1408 significantly differentially expressed transcripts (FDR: 5%, Benjamini-Hochberg) between

Microphallus-infected and uninfected snails (Figs. 1a, S1a). A significantly higher proportion of

these transcripts were downregulated (1057, ~75%) *vs*. upregulated (351, ~25%) in infected *vs*.

uninfected snails (Fisher's Exact Test: $p < 0.0001$; Table 1, Fig. 1a), indicating that infected *P*.

antipodarum experience systematic reduction in gene expression.

We obtained gene ontology (GO) annotations for 447 of these 1408 transcripts, 216 of

which were significantly upregulated (hereafter "upregulated") and 231 significantly

downregulated (hereafter "downregulated") in infected snails. 16 upregulated genes were

annotated as involved in immune system processes (Fig. S2), making these loci our strongest

candidates for direct involvement in response to *Microphallus* infection. The 10 genes with

putative brain/behaviour functions that were upregulated in infected snails (Fig. S2) are

interesting in light of the well-characterized influence of *Microphallus* infection on *P.*

antipodarum behaviour (*e.g*., 58). Functional enrichment revealed that in infected snails, antigen

processing and presentation were overrepresented among upregulated genes. Processes related to

transcription and translation were overrepresented among downregulated genes, consistent with

our observation that infected snails had significantly more downregulated genes (Table S3).

Our inclusive F_{ST} analysis, comparing genetic differentiation between infected and

uninfected snails, revealed a mean transcriptome-wide F_{ST} (SD) of 6.34367e-5 (7.35667e-6) and

identified 58 outlier SNPs (mean F_{ST} of outliers (SD) of 0.497 (0.28)) from 46 transcripts. Of

these 46 transcripts, three were downregulated and two were upregulated in infected snails,

meaning the majority of F_{ST} outlier-containing genes were not significantly differentially

expressed (Table 1). We obtained functional annotations for 21 of these 46 transcripts. We

annotated the three genes that contained F_{ST} outlier SNPs and were downregulated in infected

snails with functions related to immune and stress responses (Tables S4, S5).

Within-lake analysis: Lake-specific responses to *Microphallus* infection. Similar to the

inclusive analysis, we observed a significantly greater proportion of downregulated *vs*.

upregulated transcripts in infected *P. antipodarum* in two of the three lakes (Table 1, Figs. 1b,

S1b). A total of 1539 transcripts were significantly differentially expressed between infected and

uninfected snails in at least one of the three pairwise tests (Fig. 1b).

Functional enrichment for the within-lake analysis revealed that various metabolic

processes are overrepresented among upregulated genes in infected snails from Alexandrina

(Table S3). We did not detect significant functional enrichment for genes upregulated in infected

snails from lakes Selfe or Kaniere. Upregulated genes in uninfected snails from Alexandrina and

Selfe are enriched for GO terms related to transcription and translation. Upregulated genes in

uninfected snails from Kaniere are enriched for pantothenate metabolism (Table S3).

The mean transcriptome-wide F_{ST} between infected and uninfected snails within each

lake was significantly different from each of the other lakes (Fig. 2), indicating population-

specific levels of genetic differentiation between infected and uninfected snails. We identified

45, 40, and 51 F_{ST} outlier SNPs between infected and uninfected snails (mean F_{ST} of outliers

(SD) of 0.51 (0.28), 0.43 (0.24), and 0.52 (0.25)) from Alexandrina, Kaniere, and Selfe,

respectively. Similar to the inclusive analysis, the vast majority of outlier-containing transcripts

(96%) were not significantly differentially expressed (Table 1).

Across-lake analysis: Lake of origin strongly influences genetic differentiation and gene

expression. We observed 6228 significantly differentially expressed transcripts within and

across lakes (Figs. 3, S1c). ~75% (4689) of these 6228 transcripts were significantly

differentially expressed across the three lakes regardless of infection status, indicating that lake

of origin has a markedly stronger influence on gene expression than infection status (Figs. 1b, 3,

4, S3). Similarly, when replicates are clustered based on expression profile, they group first by

lake of origin, followed by infection status. This result indicates both that population of origin is

a key determinant of gene expression in *P. antipodarum* (Figs. 3, S3) and that infection results in

predictable changes in expression (Fig. 3) that are detectable on this background of population

divergence. These findings are consistent with and extend to the gene expression level evidence

for population-specific phenotypes (40, 59) and marked population genetic structure (39, 60-61)

in *P. antipodarum*. These results also demonstrate that infection has marked consistent and

systematic consequences for gene expression in this species.

We used the total number of transcripts identified as significantly differentially expressed

between infected and uninfected snails (1539) in at least one of the three within-lake analyses

(Fig. 1b) to compare the number of significantly up or downregulated transcripts in multiple *vs*.

single populations, allowing us to determine whether snails from different lakes have similar

gene expression responses to *Microphallus* infection. This comparison revealed that nearly all of

the differentially expressed transcripts from the within-lake analysis (1447 transcripts, 94%) are

only significantly differentially expressed in a single lake. Only 6% (92) of the differentially

expressed transcripts showed significant differential expression in more than one population

(Fig. 4). In summary, the vast majority of differentially expressed transcripts between infected

and uninfected snails show significant up or downregulation in only one population, suggesting a

distinct local (lake specific) gene expression response to parasite infection.

Next, we identified the types of genes that were significantly differentially expressed in

single *vs*. multiple populations to parse out general *vs*. lake-specific infection responses. Of the

92 genes that were significantly differentially expressed in more than one lake, 22 genes were

upregulated in infected snails (Table S6). We obtained GO annotations for seven of these genes,

which had functions related to immune response, nervous system function, and metabolism. 10

of the 70 genes found to be significantly downregulated in infected snails in more than one lake

received GO annotations, which included immune function, response to stimulus, and

transcription/translation (Table S6). 396 of the 1447 genes that were significantly differentially

expressed in only one lake received GO annotations (Fig. S2). Even though the particular genes

experiencing differential expression differ on a lake-by-lake basis, these different genes often

belong to similar GO categories (*e.g.*, immune system processes, response to stimulus, and

behaviour).

We also evaluated how many transcripts contained F_{ST} outlier SNPs in single *vs.* multiple

populations. Similar to our gene expression analyses, we found that very few transcripts (6

transcripts, ~4%) containing F_{ST} outlier SNPs between infected and uninfected snails in one

population also contained outlier SNPs in another population (3 transcripts each between

Alexandrina and Kaniere and between Kaniere and Selfe). This result indicates that the focal loci

of *Microphallus*-mediated selection are likely to often be population specific. Of the transcripts

that contained outlier SNPs, 14, 12, and 15 transcripts were annotated for Alexandrina, Kaniere,

and Selfe, respectively. 10 of these transcripts were annotated as relevant to immune

response/function and 4 for neurological processes (Tables S4, S5).

Finally, we compared mean transcriptome-wide F_{ST} across lakes as a whole, for infected

replicates only, and for uninfected replicates only. The mean F_{ST} was significantly different ($p <$

0.0001, Fig. 2) in all possible pairwise comparisons. We also found that F_{ST} between infected

and uninfected snails within each lake was significantly lower than the mean F_{ST} for any across-

lake comparison (Fig. 2), indicating greater levels of genetic differentiation among lake

populations than between infected and uninfected snails within a population. This result is

consistent with and extends to the genome level the outcome of marker-based studies (*e.g.,* 61)

that have documented strong across-lake genetic structure in *P. antipodarum*.

Discussion

We used gene expression and F_{ST} analyses to shed light on the genomic underpinnings of

coevolution in a natural context. Most importantly, we found that infection of *P. antipodarum* by

its coevolving trematode parasite *Microphallus* elicits a marked, systematic, and population-

specific gene expression response, with important potential implications for our understanding of

the dynamics of coevolution. Our F_{ST} analyses allowed us to evaluate levels of genetic

differentiation within and among lakes and revealed promising candidate genes for the focus of

Microphallus-mediated selection. These findings provide a qualitative advance by extending

evidence for local adaptation and coevolution in this textbook coevolutionary interaction to the

genomic level, demonstrating distinct local genetic and gene expression responses by *P.*

antipodarum to *Microphallus* infection. Together, these results illuminate the unique and

important insights that can come from sampling multiple natural populations of interacting hosts

and parasites.

Systematic downregulation and response to parasite infection. Our inclusive analysis

revealed that the majority of transcripts that are significantly differentially expressed in infected

snails are downregulated relative to uninfected snails. This pattern could reflect several non-

mutually exclusive phenomena, ranging from tissue/organ destruction and/or overall poor

condition of infected snails (*e.g.,* 11) to reallocation of resources to genes needed for defence

against and response to *Microphallus* infection (*e.g.,* 62) to suppression of *P. antipodarum* gene

expression by *Microphallus* as a means of evading host immune and defence systems (*e.g.,* 11,

13). Regardless of the specific mechanism(s) involved, the transcripts that are differentially

expressed between infected and uninfected snails in the inclusive analysis represent a set of

genes that are most likely to be of general importance to the *P. antipodarum-Microphallus*

interaction. Our inclusive F_{ST} analyses yielded additional candidate genes for the focus of

parasite-mediated selection, with the 58 SNPs from 46 genes that are evolving especially rapidly

providing a particularly strong set of candidate loci for *Microphallus* response. The presence of

multiple genes annotated with immune/stress responses and with neurological and behaviour-

related functions strengthens this conclusion.

Genomic evidence for population-level responses to infection. Our results demonstrate that *P.*

antipodarum response to *Microphallus* infection is largely population specific. For example, in

Alexandrina and Kaniere, the majority of significantly differentially expressed transcripts are

downregulated in infected relative to uninfected snails (73% for Alexandrina and 66% for

Kaniere). By contrast, the proportion of significantly upregulated and downregulated transcripts

is much more similar in snails from Selfe (52% *vs*. 58%, respectively). This pattern is interesting

in light of data from Vergara *et al.* (2013) suggesting that lake Selfe *P. antipodarum* have

historically experienced lower frequencies of *Microphallus* infection (3.5x lower than

Alexandrina, 1.4x lower than Kaniere; Table S1)(25), suggesting that lake Selfe snails are likely

to have experienced a somewhat weaker intensity of *Microphallus*-mediated selection.

The vast majority (~87%) of differentially expressed transcripts were only significantly

differentially expressed in a single lake population, and ~75% of significantly differentially

expressed transcripts were expressed differently across lakes regardless of infection status. We

also found significantly lower levels of genetic differentiation between snails within *vs.* across

lakes, regardless of infection status. This result is consistent with previous evidence for

population structure (*e.g.* 39, 60-61) and/or negative frequency-dependent selection driving

divergence between populations (*e.g.*, 63). Together, these results demonstrate that lake of origin

has a major influence on gene expression and genetic differentiation and provide the first

genomic evidence consistent with local adaptation between *Microphallus* and *P. antipodarum*.

Candidate loci for future research. Our results provide a strong set of candidate loci for the

genomic basis of *P. antipodarum* and *Microphallus* interactions. Three of the 16 upregulated

immune-relevant genes are homologous to genes involved in response to trematode infection in

the *Biomphalaria glabrata-Schistosoma mansoni* host-parasite system (*e.g.,* 64; reviewed in 65),

suggesting these genes might play an important role in *P. antipodarum* response to *Microphallus*

infection. We also identified seven downregulated genes with immune-related functions,

including myosin light-chain kinase and fibrinogen-related proteins, which have also been shown

to contribute to resistance to infection in the *Biomphalaria-Schistosoma* system (10, 64). Other

significantly upregulated genes with immune-related functions in infected snails represent further

candidates for response to *Microphallus* infection, while significantly downregulated genes with

immune-related functions have the potential to be involved in resisting infection.

Among the genes upregulated in infected snails, we observed functional enrichment for

actin, myosin, and genes with other cytoskeletal functions. Genes involved in cytoskeletal

function, including actin and myosin, are upregulated upon parasite exposure in the

Biomphalaria-Schistosoma system (reviewed in 65), indicating that these genes may also

contribute to response to parasite infection in *P. antipodarum*. Oxidative processes have also

been implicated in schistosome defence response in *Biomphalaria* (66, reviewed in 65); a

preliminary line of evidence that oxidative processes may also be involved in the *P. antipodarum*

trematode response is provided by significant functional enrichment of genes involved in

oxidative processes (*e.g.*, NADH oxidation, fatty acid beta-oxidation, energy derivation by

oxidation of organic compounds) in infected snails.

The inclusive analysis revealed 21 upregulated and 16 downregulated genes with

potential roles in important behavioural traits (*e.g.*, foraging, locomotion, mating) (Fig. S2).

These results are consistent with evidence that exposure to (67) and infection by *Microphallus*

(58, 68) affects *P. antipodarum* behaviour. In particular, infected snails forage at a higher

frequency than uninfected snails during the time of day when the waterfowl that are

Microphallus's final host are active, rendering infected snails more vulnerable to predation (68).

The implications are that these genes are a set of candidates for potential genetic mechanisms

and pathways involved in *Microphallus*-induced alterations to *P. antipodarum* behaviour that

could influence transmission probability. Future study of snails from natural populations

featuring little to no *Microphallus* infection (23, 38) as well as manipulative experiments that

allow comparisons between exposed *vs.* unexposed and infected *vs.* uninfected individuals will

provide valuable additional steps forward by enabling differentiation between exposed but

uninfected *vs.* naïve individuals.

Genes involved in the regulation of gene expression and ribosome structure and function

were overrepresented among significantly downregulated genes. These results are consistent

with our overall observation that infected snails had significantly more downregulated *vs.*

upregulated genes, suggesting that *Microphallus* infection leads to decreased overall gene

expression in *P. antipodarum*. These results are strikingly similar to a recent laboratory study

showing that bumblebees (*Bombus terrestris*) exposed to particularly infective genotypes of a

trypanosome parasite (*Crithidia bombi*) downregulated more genes than unexposed bumblebees

(11). Similar results have been reported in other lab-based studies (*e.g.,* 14-15). Our results thus

extend to natural coevolving populations the growing body of evidence that infected hosts

experience systematic downregulation of gene expression.

Summary and conclusions. We present novel evidence for systematic but population-specific

genetic and gene expression responses to parasite infection in the *P. antipodarum-Microphallus*

coevolutionary interaction. These results are the first genome-level evidence for the type of

population-specific response expected under local adaptation and coevolution in this important

host-parasite system. We also identified genes with functions related to immune and defence

response and behaviour that are likely involved in the *P. antipodarum* response to *Microphallus*

infection, providing a set of candidate genes for involvement in the genomic basis of

coevolution. The targeted exploration of these genes made possible by these results will help

illuminate the genetic and genomic mechanisms that determine the outcome of interactions

between coevolving hosts and parasites in nature.

Data accessibility: All RNA-Seq data will be made publically available on the NCBI GEO

database upon article acceptance.

Competing interests: We have no competing interests.

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Table and Figure Legends

Table 1: Summary of inclusive and within-lake analyses. Left: number of significantly upregulated, downregulated, and total number of significantly differentially expressed genes in infected *vs.* uninfected snails (proportion of differentially expressed genes). Right: number of F_{ST} outlier SNPs between infected and uninfected snails, number of transcripts containing outlier SNPs, and the number of transcripts containing outlier SNPs that were significantly upregulated or downregulated in infected *vs.* uninfected snails.

Fig. 1. Heatmaps representing significantly differentially expressed transcripts (higher FPKM = higher expression). Significance was assessed with a FDR of 5% and Benjamini-Hochberg multiple test correction. a) Inclusive analysis: 1408 significantly differentially expressed transcripts in infected *vs.* uninfected snails. Transcripts above the black line demarcated with "*" are downregulated in infected snails (1057), and transcripts below this line are upregulated in infected snails (351). b) Within-lake analysis: dendrogram and heatmap representing the 6228 significantly differentially expressed transcripts across infection status and lakes. "Ax" = Alexandrina, "Se" = Selfe, "Kn" = Kaniere, red = infected, blue = uninfected. Snails clustered by lake of origin rather than infection status, indicating spatially structured gene expression patterns. Highlighted regions of heatmap contain examples of: expression differences between infected and uninfected snails within a lake (green), lake-specific expression regardless of infection status (downregulation in Ax, blue), and across-lake upregulation in infected snails (purple).

Fig. 2. Mean transcriptome-wide F_{ST} (SD) calculated based on F_{ST} per site. From left to right, the first panel is mean pairwise F_{ST} between lakes (*e.g.,* AxKn = mean pairwise F_{ST} between Alexandrina and Kaniere), the second panel is mean pairwise F_{ST} for infected replicates only, the third panel is mean F_{ST} for uninfected replicates only, and the fourth panel is mean F_{ST} between infected and uninfected snails within each lake. Lake acronyms follow Fig. 1. We used Welch's t-tests to compare mean F_{ST} (+/- SD) for all possible pairwise comparisons with Bonferroni multiple test correction ($p < 0.0001$; all comparisons were statistically significant).

Fig. 3. A multidimensional scaling (MDS) plot depicting how expression profiles are affected by lake population and infection status. Closer proximity of replicates in plot space indicates a more similar expression profile. Red = infected, blue = uninfected, diamonds = Alexandrina, triangles = Kaniere, circles = Selfe. The elliptical outlines (red = infected replicates from each lake; blue = uninfected replicates from each lake) delineate clustering for a particular replicate type and do

not represent formal statistical tests; these clustering patterns are further supported by a dendrogram (Fig. S3). While the expression profiles that characterize uninfected and infected snails within each lake reveal lake-specific gene expression responses to infection, the consistent shift to the right along the "M1" axis for infected *vs.* uninfected snails for each lake population (highlighted with arrows) also demonstrates across-lake commonalities in expression profiles for infected snails.

Fig. 4. Euler diagrams depicting the number of significantly differentially expressed genes in infected *vs.* uninfected snails for the within-lake analysis. Left: number of upregulated genes in single *vs.* multiple lake populations (*e.g.*, 47 genes upregulated in Kaniere; 0 genes upregulated in all three lakes). Right: number of downregulated genes in single *vs.* multiple populations. Circle size is proportional to the number of genes in that category.

Table 1. Summary of gene expression and F_{ST} outlier analyses.

| Population | Expression analyses | | | | | F_{ST} outlier analyses | | | |
	Upregulated		Downregulated	Total	FET p-value	F_{ST} outliers	Transcripts	Upregulated	Downregulated
Alexandrina	267 (0.27)	<	728 (0.73)	995	$p < 0.0001$	45	37	2	6
Kaniere	53 (0.34)	<	103 (0.66)	156	$p = 0.0011$	40	31	0	0
Selfe	267 (0.52)	$\cong$	246 (0.48)	513	$p = 0.4821$	51	35	1	1
Inclusive	351 (0.25)	<	1057 (0.75)	1408	$p < 0.0001$	58	46	1	3

Supplementary Methods

Sample preparation and RNA sequencing. Because *Microphallus* infection fills the

body cavity of *P. antipodarum*, we confined analyses to head tissue to ensure that

comparable tissue types were isolated from infected and uninfected individuals. We first

separated the head tissue of dissected snails, which does not contain *Microphallus*

metacercariae, from body tissue. We then split the dissected head tissue into two halves.

One head half was immediately submerged in 100 μL RNA*later*® Solution (Life

Technologies Corporation) and stored at 4°C for 24 hours followed by storage at -80°C

(according to manufacturer protocol) until RNA extraction. The other head half was

immediately snap-frozen in liquid nitrogen and stored at -80°C for flow cytometric

determination of ploidy level following (40).

We extracted RNA from pooled head tissue following the Invitrogen TRIzol

protocol (41). RNA quantity and quality were assessed with a Bio-Rad Experion

Automated Electrophoresis Station, following manufacturer protocol for the Experion

RNA analysis kit with an RQI ≥ 8 and a minimum of 2μg of total RNA. RNA shearing

and cDNA library preparation were completed following the Illumina Truseq LS protocol

(Illumina, San Diego, CA, 2012). Following library preparation, we used an Illumina

HiSeq 2000 for 2x100 bp paired-end RNA sequencing. Each RNA sample was given a

unique indexed adapter sequence (Illumina, San Diego, CA, 2012) and was then

separated into two halves, allowing us to sequence each sample twice in two different

lanes in order to eliminate bias due to sequencing lane. We obtained a mean read length

(SD) of 100.7 bp (0.96) and a mean number of paired-end reads/replicated sample/lane

(SD) of 10070420 (1474539.10), for ~20000000 paired-end reads/replicate.

De novo reference transcriptome assembly. We used the FASTX Toolkit (42) and

FastQC (43) to trim adapter sequences, assess sequencing quality, and remove poor-

quality reads from the raw RNA-Seq data (see Table S2 for specific parameters). Quality

filtering resulted in a mean number of reads/replicate/lane (SD) of 9871440.4

(1468645.04), for a total of ~18000000 paired-end reads/replicate, about twice the

coverage required to quantify expression differences across a wide range of expression

levels (69). We assembled a _de novo_ transcriptome using Trinity v. 2.0.4 and all of the

combined filtered RNA-Seq data. First, we performed the recommended Trinity _in silico_

normalization for each replicate in order to reduce the amount of memory required for the

assembly process while maintaining a representative read set (following 44-45). We then

used Trinity to assemble the normalized reads (45), generating a transcriptome assembly

with 462736 transcripts (Table S2). We annotated protein-coding regions in the

transcriptome in order to identify long ORFs of putative genes that could be missed by

homology searches and to filter out miscalled isoforms by using the Trinity plugin

TransDecoder (45) (Table S2). This step also lessens the influence of sequencing errors,

misassembled transcripts, chimeric sequences, and other common assembly issues (70).

Next, we used hierarchical clustering based on sequence identity to further reduce

redundancy in the transcriptome assembly, as implemented in CD-HIT-EST (46) (Table

S2). Finally, we identified and eliminated potential contaminant transcripts using the

'blast_taxonomy_report.pl' script from the Blobology pipeline (47), blastx (48), and a

custom python script (available upon request). We used Blobology to filter out transcripts

with non-metazoan top blast hits, as well as Platyhelminthes (representing potential

Microphallus contamination). These last filtering steps provided a final reference

transcriptome assembly with 62862 transcripts.

RNA-Seq gene expression analyses. We used our *de novo* transcriptome as our

reference assembly for all gene expression analyses in the Tuxedo pipeline. First, we

mapped filtered reads to the *de novo* transcriptome assembly using TopHat2 (52). Next,

we assembled mapped reads into transcripts and estimated transcript abundance with

Cufflinks (53). We merged Cufflinks GTF and Tophat bam files for use in CuffDiff with

Cuffmerge (53). Finally, we identified and quantified significant changes in gene

expression using an FDR of 5% and Benjamini-Hochberg multiple test correction, as

implemented in CuffDiff (53). We required that fragments per kilobase per million reads

mapped (FPKM) exceeded zero to ensure that we were only comparing expression

patterns among genes transcribed in all replicates (Table S2). We generated heatmaps and

dendrograms with cummeRbund, allowing us to visualize significantly differentially

expressed transcripts. We also used cummeRbund to generate MDS plots (54).

GO annotation of differentially expressed transcripts. We annotated the reference

transcriptome by using blastx (48) with an E-value cutoff of 1e-5. We then imported

these blastx results into Blast2GO to assign gene ontology ("GO") terms to blastx-

annotated transcripts (49). Next, we compared the functions of differentially expressed

transcripts between infected and uninfected replicates overall ("inclusive analysis") and

within and across lakes ("within-lake analysis" and "across-lake analysis", respectively;

Fig. S1) using the combined graph function in Blast2GO. With this analysis, we

generated a list of all GO terms present in the datasets and quantified the number of

transcripts annotated with each GO term. This approach allowed us to identify the

putative functions of transcripts that were significantly differentially expressed between

infected and uninfected replicates in more than one lake population compared to

transcripts that were only significantly differentially expressed between infected and

uninfected replicates in a single lake. These comparisons were visualized with Euler

diagrams generated with the program eulerAPE v3 (57). Finally, we used functional

enrichment analyses and Fisher's Exact Tests implemented in Blast2GO to identify

significantly over *vs.* underrepresented functional groups among the differentially

expressed transcripts between infected and uninfected snails in the inclusive and within-

lake analyses.

SNP calling and F_{ST} outlier detection. We performed F_{ST} outlier analyses in order to

identify genes evolving especially rapidly in infected *vs.* uninfected snails; these genes

can be considered candidates for the focus of *Microphallus*-mediated selection. We first

filtered our *de novo* reference transcriptome down to only the transcripts with an FPKM >

zero in all 18 replicates (30685 transcripts) to ensure that each replicate was represented

for all loci in the F_{ST} comparisons. We used Tophat2 (52) with the same parameters as in

the gene expression analyses to map RNA-Seq reads to the filtered reference

transcriptome. We then used Picard Tools to prepare mapped reads for variant discovery

(http://picard.sourceforge.net), applying the AddOrReplaceReadGroups script to add read

groups to each mapped bam file. We then merged technical replicate bam files with the

MergeSamFiles script, resulting in a single bam file for each of the 18 replicates. Finally,

we used the MarkDuplicates script to identify duplicate reads, removing reads that

mapped to more than one location in the transcriptome. These processed bam files were

then used to generate mpileup files using Samtools mpileup (55), which calls SNPs from

the processed bam files (default parameters). Analyses of levels of nucleotide variation

from pooled RNA-sequencing data with programs like Popoolation2 (56), which was

developed to perform variant calling and calculate F_{ST} from pooled sequencing data, can

effectively identity SNPs and candidate focal genes of selection and perform F_{ST}-based

measures of genetic differentiation (*e.g.*, 71-72).

We used Popoolation2 for an inclusive analysis (all pooled infected replicates *vs.*

all pooled uninfected replicates), within-lake analyses (Alexandrina infected *vs.*

uninfected, Kaniere infected *vs.* uninfected, Selfe infected *vs.* uninfected), and across-

lake analyses (Alexandrina *vs.* Kaniere, Alexandrina *vs.* Selfe, Kaniere *vs.* Selfe) (Fig.

S1). The Popoolation2 pipeline begins by using the mpileup2sync.pl script to generate

synchronized mpileup files that were filtered for base quality (Q20) and then applies the

fst-sliding.pl script to calculate F_{ST} per site. In all cases, we required a minimum coverage

of 10 (to account for sequencing errors and reduce the likelihood of false positives by

ensuring that each SNP was sequenced multiple times), maximum coverage of 200 (to

reduce memory usage and biases introduced by inter-locus variation in gene expression)

(*e.g.*, 71-72), and otherwise used default parameters. We chose these parameters to

reduce the likelihood of obtaining false positives at the risk of not detecting rare alleles

(*e.g.,* 72). Finally, we used IBM SPSS Statistics v. 23 to perform outlier analyses to

identify outlier SNPs between infected and uninfected snails as a whole (inclusive

analysis) and between infected and uninfected snails within and between each lake and

lake pair. We also identified the expression patterns of each transcript containing F_{ST}

outlier SNPs based on the expression analyses detailed above and evaluated whether

transcripts containing F_{ST} outliers between infected and uninfected snails contained

outliers in multiple lake populations *vs.* only a single lake population. Finally, we

compared mean transcriptome-wide F_{ST} between infected and uninfected snails within

each lake and mean transcriptome-wide F_{ST} between lakes using Welch's t-tests as

implemented in IBM SPSS Statistics v. 23 (Bonferroni corrected).

Additional References (SI Methods only)

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Table S1. Location of sample sites and infection frequencies.

Lake	Location	Infection frequency (Vergara *et al.* 2013)	Infection frequency (present study)
Alexandrina	-43.936736º, 170.460260º	25.4%	20.90%
Kaniere	-42.806435º, 171.127667º	10.3% (SD +/- 6.06)	10.60%
Selfe	-43.243705º, 171.518647º	7.36% (SD +/- 6.53)	11.50%

Geographical coordinates of lakes sampled in this study and *Microphallus* infection frequencies as reported by Vergara *et al.* (2013) and in this study. Means and standard deviations are given for data from Vergara *et al.* (2013) when multiple years of sampling were available for the shallow region of a given lake (26). We determined infection frequency for this study by dissecting 116, 107, and 100 snails from Alexandrina, Kaniere, and Selfe, respectively.

Table S2. Programs and associated parameters used in transcriptome assembly and quantification and comparisons of patterns of gene expression.

Program	Purpose	Parameters	References
FASTX Toolkit	Quality filtering of raw RNA-Seq reads	Mean sequence Phred quality score cutoff = 20	53
Trinity (*in silico* normalization)	Reduces memory requirement for assembly process, recommended for large datasets	Minimum kmer coverage = 2, maximum kmer coverage = 30	55, 56
Trinity (assembly)	*De novo* assembly of RNA-Seq data using normalized read set	Default parameters	55, 56
TransDecoder	ORF annotation and miscalled isoform identification and filtering	Default parameters	56
CD-HIT-EST	Hierarchical clustering-based method to reduce assembly redundancy	Minimum sequence identity = 0.95, word size = 8 nucleotides	57
Blobology (blast taxonomy report)	Identification of potential contaminant transcripts	Default parameters	58
Tophat2	Map filtered RNA-Seq reads to *de novo* reference transcriptome	3% mismatch between reads to account for polymorphism and sequencing error	63, 64
Cufflinks	Assemble mapped reads into transcripts and estimate their abundances	Upper-quartile normalization; multi-read correction	63, 64
Cuffmerge	Merge Cufflinks gtf files and Tophat2 bam files for use in CuffDiff	Default parameters	64
CuffDiff	Quantify significant changes in gene expression	FDR = 0.05, Benjamini-Hochberg multiple test correction, FPKM > 0	64

Table S3. Functionally enriched GO terms for significantly differentially expressed genes between infected and uninfected snails for the inclusive and within lake analyses.

	GO-ID	Term	FDR	p-value	Percent in Test	Percent in Reference
		Functionally enriched GO terms for genes upregulated in infected snails in the inclusive analysis				
	GO:0032555	purine ribonucleotide binding	7.52E-08	5.73E-11	16.52	10.68
	GO:0030554	adenyl nucleotide binding	1.56E-06	1.90E-09	13.68	8.62
	GO:0005524	ATP binding	1.80E-06	2.33E-09	13.39	8.38
	GO:0019752	carboxylic acid metabolic process	2.68E-06	3.77E-09	8.55	3.89
	GO:0015629	actin cytoskeleton	4.24E-05	8.72E-08	4.27	1.16
	GO:0005737	cytoplasm	1.17E-04	2.51E-07	28.77	29.87
	GO:0002479	antigen processing and presentation of exogenous peptide antigen via MHC class I, TAP-dependent	1.56E-04	3.69E-07	1.71	0.10
	GO:0005829	cytosol	3.83E-04	9.92E-07	10.26	6.82
	GO:0030016	myofibril	1.61E-03	4.80E-06	2.85	0.68
	GO:0006508	proteolysis	2.53E-03	8.10E-06	6.55	3.67
	GO:0006977	DNA damage response, signal transduction by p53 class mediator resulting in cell cycle arrest	3.43E-03	1.23E-05	1.42	0.11
	GO:0051436	negative regulation of ubiquitin-protein ligase activity involved in mitotic cell cycle	3.43E-03	1.23E-05	1.42	0.11
	GO:0044262	cellular carbohydrate metabolic process	3.50E-03	1.39E-05	2.85	0.78
	GO:0006094	gluconeogenesis	3.50E-03	1.55E-05	1.71	0.22
	GO:0006096	glycolytic process	3.66E-03	1.64E-05	1.99	0.34
	GO:0008092	cytoskeletal protein binding	4.27E-03	2.08E-05	4.84	2.32
	GO:0022624	proteasome accessory complex	9.68E-03	5.68E-05	1.42	0.16
	GO:0004613	phosphoenolpyruvate carboxykinase (GTP) activity	1.35E-02	8.36E-05	0.85	0.03
	GO:0006521	regulation of cellular amino acid metabolic process	1.41E-02	8.79E-05	1.14	0.09
	GO:0051437	positive regulation of ubiquitin-protein ligase activity involved in mitotic cell cycle	1.42E-02	8.99E-05	1.42	0.18
	GO:0015980	energy derivation by oxidation of organic compounds	1.75E-02	1.13E-04	2.85	1.01
	GO:0044449	contractile fiber part	1.85E-02	1.21E-04	2.28	0.64
	GO:0006116	NADH oxidation	2.63E-02	1.84E-04	0.57	0.00
	GO:0031145	anaphase-promoting complex-dependent proteasomal ubiquitin-dependent protein catabolic process	3.34E-02	2.52E-04	1.42	0.23
	GO:0005977	glycogen metabolic process	3.65E-02	2.82E-04	1.42	0.24

| | GO:0006635 | fatty acid beta-oxidation | 3.65E-02 | 2.86E-04 | 1.14 | 0.13 |
| | GO:0036464 | cytoplasmic ribonucleoprotein granule | 4.81E-02 | 3.88E-04 | 1.42 | 0.26 |

Functionally enriched GO terms for genes upregulated in uninfected snails (downregulated in infected snails) in the inclusive analysis

	GO-ID	Term	FDR	p-value	Percent in Test	Percent in Reference
	GO:0003735	structural constituent of ribosome	3.62E-10	1.40E-13	2.37	1.46
	GO:0008810	cellulase activity	9.86E-09	6.01E-12	0.66	0.01
	GO:0030247	polysaccharide binding	2.36E-08	1.98E-11	0.66	0.02
	GO:0016985	mannan endo-1,4-beta-mannosidase activity	2.10E-07	2.72E-10	0.66	0.04
	GO:0046355	mannan catabolic process	2.10E-07	2.72E-10	0.66	0.04
	GO:0006412	translation	3.07E-05	4.91E-08	2.84	3.89
	GO:0022626	cytosolic ribosome	2.80E-04	5.76E-07	1.14	0.73
	GO:0042742	defense response to bacterium	4.93E-03	1.09E-05	0.57	0.18
	GO:0030833	regulation of actin filament polymerization	1.06E-02	2.66E-05	0.66	0.32
	GO:0009506	plasmodesma	1.65E-02	4.77E-05	0.47	0.14
	GO:0045493	xylan catabolic process	2.14E-02	6.70E-05	0.28	0.02
	GO:0015935	small ribosomal subunit	2.27E-02	7.25E-05	0.76	0.53
	GO:0030301	cholesterol transport	3.34E-02	1.12E-04	0.47	0.17
	GO:0015934	large ribosomal subunit	4.86E-02	1.67E-04	0.66	0.44

Functionally enriched GO terms for genes upregulated in infected snails in the within lake analysis

	GO-ID	Term	FDR	p-value	Percent in Test	Percent in Reference
Alexandrina	GO:0004613	phosphoenolpyruvate carboxykinase (GTP) activity	4.92E-03	7.50E-07	6.67	0.03
	GO:0006089	lactate metabolic process	1.76E-02	7.93E-06	4.44	0.00
	GO:0042593	glucose homeostasis	1.76E-02	8.59E-06	8.89	0.27
	GO:0010817	regulation of hormone levels	2.00E-02	1.22E-05	11.11	0.64
	GO:0030073	insulin secretion	2.22E-02	1.52E-05	8.89	0.31
	GO:0046327	glycerol biosynthetic process from pyruvate	2.84E-02	2.38E-05	4.44	0.01
Kaniere	None					
Selfe	None					

Functionally enriched GO terms for genes upregulated in uninfected snails in the within lake analysis

	GO-ID	Term	FDR	p-value	Percent in Test	Percent in Reference
Alexandrina	GO:0003735	structural constituent of ribosome	2.64E-07	6.04E-11	9.63	1.51
	GO:0006412	translation	2.02E-04	7.70E-08	12.84	3.96
	GO:0022626	cytosolic ribosome	2.98E-03	1.59E-06	5.05	0.75
	GO:0016798	hydrolase activity, acting on glycosyl bonds	2.18E-02	1.65E-05	5.96	1.37
	GO:0015935	small ribosomal subunit	4.39E-02	3.68E-05	3.67	0.53

Kaniere	GO:0017159	pantetheine hydrolase activity	1.61E-02	1.23E-06	11.11	0.00
	GO:0015939	pantothenate metabolic process	2.41E-02	3.68E-06	11.11	0.01
Selfe	GO:0006418	tRNA aminoacylation for protein translation	3.26E-09	4.97E-13	13.75	0.48
	GO:0004812	aminoacyl-tRNA ligase activity	3.26E-09	4.97E-13	13.75	0.48
	GO:0000049	tRNA binding	7.01E-06	4.81E-09	7.50	0.14
	GO:0043022	ribosome binding	5.46E-04	6.23E-07	6.25	0.18
	GO:0032797	SMN complex	1.19E-02	2.53E-05	2.50	0.00
	GO:0032575	ATP-dependent 5'-3' RNA helicase activity	1.19E-02	2.53E-05	2.50	0.00
	GO:0005829	cytosol	1.40E-02	3.36E-05	21.25	6.97
	GO:0000166	nucleotide binding	2.36E-02	6.39E-05	35.00	16.75
	GO:0022403	cell cycle phase	4.06E-02	1.14E-04	3.75	0.10

Functionally enriched GO terms for the inclusive analysis and the within-lake analysis; enrichment was determined via significant outcome of Fisher's Exact tests with FDR < 0.05 as implemented in Blast2GO. "Percent in Test" refers to the percent of transcripts in the test dataset (*e.g.*, upregulated in infected snails in the inclusive analysis) that were assigned a particular GO term. "Percent in Reference" refers to the percent of transcripts in the reference transcriptome that were assigned a particular GO term.

Table S4. Functional annotation of transcripts containing at least one F_{ST} outlier SNP between infected and uninfected snails and the expression pattern of these transcripts.

Comparison	Annotation	Expression Pattern
Inclusive analysis	cathepsin l	downregulated
	probable serine threonine-protein kinase fhkb	n.s.
	cellulase egx3	n.s.
	protein deglycase dj-1	n.s.
	alkali-sensitive linkage protein 1-like	n.s.
	pancreatic triacylglycerol lipase-like	upregulated
	lambda-crystallin homolog	n.s.
	upf0764 protein c16orf89 homolog	n.s.
	PREDICTED: uncharacterized protein LOC106079933	n.s.
	PREDICTED: uncharacterized protein C9orf78 homolog	n.s.
	alpha-aminoadipic semialdehyde dehydrogenase	n.s.
	alpha-l-fucosidase-like	n.s.
	deleted in malignant brain tumors 1 partial	downregulated
	kielin chordin-like protein	n.s.
	intraflagellar transport protein 88 homolog isoform x1	n.s.
	solute carrier family 26 member 6-like	n.s.
	exonuclease 3 -5 domain-containing protein 1	n.s.
	cell wall integrity and stress response component 4-like	n.s.
	signal recognition particle 19 kda protein	n.s.
	contactin-like	n.s.
	c-type lectin	n.s.
Within lake		
Alexandrina	transcription elongation factor b polypeptide 2-like	n.s.
	chymotrypsin-like serine proteinase	downregulated
	perlucin-like protein	n.s.
	mammalian ependymin-related protein 1-like	n.s.
	uncharacterized oxidoreductase -like	n.s.
	amyloid beta a4 protein isoform x2	n.s.
	tyrosinase-like protein tyr-1	n.s.
	pleckstrin homology domain-containing family b member 2-like	n.s.
	microtubule-actin cross-linking factor isoforms 1 2 3 5-like	n.s.
	cytochrome p450 3a7-like	n.s.
	neurexin-4 isoform x2	downregulated
	complement c1q-like protein 2	n.s.
	v-type proton atpase 16 kda proteolipid subunit	n.s.
	adenylate kinase 9	n.s.

Kaniere	proteasome subunit beta type-1	n.s.
	60s ribosomal protein l23a	n.s.
	electron transfer flavoprotein subunit mitochondrial	n.s.
	chitin-binding protein	n.s.
	cd109 antigen-like	n.s.
	kielin chordin-like protein	n.s.
	hemk methyltransferase family member 1	n.s.
	ganglioside gm2 activator	n.s.
	proteasome inhibitor pi31 subunit	n.s.
	aconitate mitochondrial-like	n.s.
	ependymin-related protein 1-like	n.s.
	protein pif-like	n.s.
Selfe	39s ribosomal protein mitochondrial	n.s.
	neuropilin-2 isoform x2	n.s.
	transmembrane protease serine 2	n.s.
	goose-type lysozyme 2	n.s.
	carboxypeptidase b-like	n.s.
	thioester-containing protein	n.s.
	mucin-5ac-like isoform x1	n.s.
	peptidoglycan recognition protein s1l	n.s.
	mucin-5ac- partial	n.s.
	papilin-like isoform x1	n.s.
	neurocalcin homolog	n.s.
	signal peptidase complex subunit 3-like	n.s.
	sperm-associated antigen 6	n.s.
	cartilage matrix	n.s.
	blastula protease 10	n.s.

Annotations for transcripts containing at least one F_{ST} outlier SNP between infected and uninfected snails and that received functional annotations using Blast2GO. The expression pattern of each transcript is based on the expression analyses for the analogous comparison. Upregulated means the transcript was upregulated in infected relative to uninfected snails, downregulated means that the transcript was downregulated in infected relative to uninfected snails, and "n.s." means that the transcript was not significantly differentially expressed between infected and uninfected snails.

Table S5. Level 2 GO terms assigned to annotated transcripts that also contained at least one F_{ST} outlier SNP as well as the number of transcripts that received each GO term within a particular comparison.

Comparison	GO-id	GO-term	Number of Transcripts
Inclusive analysis	GO:0051179	localization	5
	GO:0071840	cellular component organization or biogenesis	4
	GO:0023052	signaling	3
	GO:0009987	cellular process	9
	GO:0051704	multi-organism process	2
	GO:0032502	developmental process	2
	GO:0044699	single-organism process	10
	GO:0008152	metabolic process	13
	GO:0065007	biological regulation	4
	GO:0022414	reproductive process	2
	GO:0050896	response to stimulus	3
	GO:0002376	immune system process	1
	GO:0032501	multicellular organismal process	3
	GO:0007610	behavior	2
	GO:0000003	reproduction	2
Within lake Alexandrina	GO:0051179	localization	2
	GO:0071840	cellular component organization or biogenesis	2
	GO:0023052	signaling	1
	GO:0009987	cellular process	5
	GO:0051704	multi-organism process	2
	GO:0032502	developmental process	3
	GO:0040011	locomotion	1
	GO:0044699	single-organism process	7
	GO:0008152	metabolic process	8
	GO:0065007	biological regulation	2
	GO:0022610	biological adhesion	2
	GO:0022414	reproductive process	1
	GO:0050896	response to stimulus	2
	GO:0040007	growth	1
	GO:0007610	behavior	1
	GO:0032501	multicellular organismal process	3
	GO:0000003	reproduction	1
Kaniere	GO:0051179	localization	1
	GO:0009987	cellular process	8
	GO:0044699	single-organism process	3
	GO:0008152	metabolic process	10

	GO:0065007	biological regulation	5
	GO:0022610	biological adhesion	1
	GO:0007610	behavior	1
	GO:0032501	multicellular organismal process	1
Selfe	GO:0051179	localization	3
	GO:0071840	cellular component organization or biogenesis	3
	GO:0023052	signaling	3
	GO:0051704	multi-organism process	2
	GO:0009987	cellular process	9
	GO:0032502	developmental process	2
	GO:0040011	locomotion	2
	GO:0044699	single-organism process	5
	GO:0008152	metabolic process	11
	GO:0065007	biological regulation	7
	GO:0022610	biological adhesion	2
	GO:0050896	response to stimulus	6
	GO:0040007	growth	1
	GO:0032501	multicellular organismal process	3

Table S6. Genes with significant up *vs*. downregulation in more than one population in the within-lake analysis that received both blastx and GO annotations.

Populations	Upregulated in infected snails			Downregulated in infected snails		
	Total # of transcripts	# of transcripts annotated	Annotations	Total # of transcripts	# of transcripts annotated	Annotations
Alexandrina and Kaniere	1	0	N/A	8	2	ribosomal protein l44e uncharacterized oxidoreductase yrbe-like
Alexandrina and Selfe	15	6	protein jagged-2 tryptophan -dioxygenase annexin a7 tyramine beta-hydroxylase phosphoenolpyruvate cytosolic phosphoenolpyruvate carboxykinase	44	11	tyrosinase-like protein tyr-3 achain keyhole limpet hemocyanin hemocyanin isoform 1 beta- -galactosyl-o-glycosyl-glycoprotein beta- -n-acetylglucosaminyltransferase 4 von willebrand factor type a 5 -nucleotidase elongation factor 2 gtp-binding protein 1 lysine--trna ligase-like eukaryotic translation initiation factor 2 subunit 3-like low quality protein: hemicentin-1
Kaniere and Selfe	5	0	N/A	15	2	myosin heavy chain collagen alpha-1 chain-like
Alexandrina, Kaniere, and Selfe	0	0	N/A	3	0	N/A

Fig. S1. Schematic depicting our three hierarchical analyses. Ax = Alexandrina, Kn = Kaniere, Se = Selfe; numbers after lake names refer to biological replicates. Black lines are drawn between the groups/samples being compared. a) "inclusive analysis," comparing all infected samples pooled *vs.* all uninfected samples pooled, b) "within-lake analysis," comparing infected *vs.* uninfected samples within each lake, c) "across-lake analysis," performing all possible pairwise comparisons between samples.

Fig. S2. Distribution of GO terms identified by Blast2GO for genes that showed significant differential expression; numbers indicate the number of genes annotated within each particular GO category. From left to right: inclusive analysis, and the three within-lake analyses. Many transcripts are represented by more than one GO term. The number of annotated transcripts that each pie chart represents is in the top left corner of each box.

Fig. S3. Dendrogram (generated with cummeRbund) depicting how expression profile is influenced by lake of origin and infection status. Samples are clustered based on their expression profile using the Jensen-Shannon distance. Samples that have more similar expression profiles are grouped more closely together. These results recapitulate and strengthen those depicted in the MDS plot (Fig. 4).