1   **Transcriptomic basis and evolution of the ant social interactome**

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Abstract

Social interactions are a fundamental feature of life for most organisms. Organismal development is often

strongly regulated by interactions among close relatives, yet little is known about how genes expressed in

social partners indirectly affect developmental trajectories and trait expression. In eusocial insects, social

interactions between caregiving worker nurses and larvae strictly regulate larval development and

resultant adult phenotypes. Here, we study the social interactome regulating larval development by

collecting and sequencing interacting nurses and larvae across a time course of larval development. First,

we find that the majority of nurse and larval transcriptomes exhibit parallel expression dynamics across

larval development, providing a strong transcriptomic signature of the social interactome regulating larval

development. Next, we leverage this widespread nurse-larva gene co-expression to infer social gene

regulatory networks acting between nurses and larvae. We find that genes with the strongest social effects

tend to be peripheral elements of within-tissue regulatory networks and are often known to encode

secreted proteins. For example, our data suggest that the gene *giant-lens* expressed by nurses may inhibit

larval epidermal growth factor signaling, which is known to affect various aspects of insect development,

including caste in honey bees. Finally, we find that genes recruited for social regulatory processes tend to

be relatively evolutionarily young and tend to experience relaxed selective constraint. Overall, our study

provides a first glimpse into the molecular details and evolutionary features of the social mechanisms that

regulate all aspects of life in eusocial insects.

Author Summary

From single-celled bacteria to complex plants and animals, most organisms experience and are strongly

impacted by social interactions. Despite their obvious importance, little is known about the genetic basis

of social traits and interactions. In eusocial insects, the development of larvae is strictly regulated by

caregiving nurse workers through a myriad of social interactions occurring across larval development. In

this paper, we study ant nurse-larva social interactions and the effect of interactions on gene expression

by simultaneously collecting and sequencing interacting nurses and larvae across larval development. We

find that broad-scale gene expression patterns reflect ongoing social interactions. Based on nurse-larva

gene co-expression, we identify genes expressed in nurses that exhibit strong putative effects on the

expression of genes in larvae. Interestingly, such genes with strong social roles tend to be less important

for pathways operating within individuals, and such highly social genes tend to experience relatively

weaker natural selection in comparison to less social genes. This suggests that genes underlying social

traits are characterized by distinct evolutionary and network features. Overall, this study represents a

significant advance in our understanding of the intersection of genetics, behavior, and evolution.

Introduction

Social interactions play a prominent role in the lives of nearly all organisms [1] and strongly affect trait

expression as well as fitness [2–4]. Social interactions in the context of development (e.g. parental care)

often strongly regulate developmental trajectories and resultant adult phenotypes, for example via

generationally-transferred compounds such as milk in mammals [5,6], milk-like secretions in arthropods

[7,8], and other forms of nutritional provisioning [9,10]. In many taxa including certain birds, mammals,

and insects, care for offspring and the regulation of offspring development has shifted at least in part from

parents to adult siblings, who perform alloparental care [11]. In eusocial insect societies, sterile nurse

workers strictly regulate the development of their larval siblings by modulating the quantity and quality of

nourishment larvae receive [12–14], as well as through the direct transfer of growth-regulating hormones

and proteins [15,16]. At the same time, larvae manipulate nurse provisioning behavior via pheromones

[17–20] and begging behavior [21,22].

In general, traits such as caregiving behavior that are defined or influenced by social interactions

are the property of the genomes of multiple interacting social partners [2,14]. This has implications for

both the mechanistic (e.g., molecular) underpinnings of development and trait expression as well as the

genetic basis of trait variation at the population level -- i.e. how allelic variation in the genomes of

interacting social partners affects trait variation [2,14]. Furthermore, because social traits are expressed in

one individual but impact the fitness of other individuals, social behavior and socially-influenced traits

experience distinct forms of selection, including kin selection and social selection [23,24]. Altogether,

these distinct genetic features and patterns of selection are often thought to lead to distinct evolutionary

features, such as rapid evolutionary dynamics in comparison to other traits [25–27]. In eusocial insects,

previous studies show that variation in larval developmental trajectories and ultimate adult phenotypes

(including reproductive caste, body size, etc.) depends on the combination of larval and nurse genotypes

[28–34]. However, the identity of specific genes and molecular pathways that are functionally involved in

the expression of social interactions (e.g., genes underlying nurse and larval traits affecting nurse-larva

interactions), and patterns of molecular evolution for these genes, have remained less well studied

[15,16,35,36].

Transcriptomic studies are often used to identify sets of genes underlying the expression of

particular traits by performing RNA-sequencing on individuals that vary in the expression of such traits.

For example, in social insects, recent studies have compared the transcriptomes of workers that perform

nursing versus foraging tasks [37–39], or nurses feeding larvae of different stages or castes [35,40].

However, given the phenotypic co-regulation known to occur between interacting social partners (here,

nurses and larvae), it is likely that genes expressed in one social partner affect the expression of genes in

the other social partner, and vise-versa, such that interacting social partners are connected by "social"

gene regulatory networks [14,32,41,42]. Thus, identifying the genes important for *social interactions* such

as nurse-larva interactions is only possible by studying the transcriptomic dynamics of both interacting

social partners across a time series of interactions. Similarly, host-symbiont interactions result in gene

regulation between interacting partners, and recent studies have sought to reconstruct gene regulatory

networks acting between hosts and symbionts by collecting and profiling the transcriptomes of each social

partner across a time series of interactions [43–47].

Here, we use analogous methodology to study transcriptomic signatures of nurse-larva

interactions in the pharaoh ant, *Monomorium pharaonis*. We sample a developmental time series of larvae

as well as the nurses that feed each larval stage in this series, collecting individuals at the moment of

interaction in order to identify genes involved in the expression of nurse-larva interactions, as well as

genes affected by these interactions (i.e. the full "social interactome" [14]). Given that pharaoh ant nurses

tend to specialize on feeding young versus old larvae and nurses feeding young versus old larvae show

different transcriptomic profiles [40], and larval transcriptomic profiles also change over development

[48,49], we predicted that if nurse-larva interactions have functional implications as described above, we

would observe concerted changes in broad-scale gene expression in larvae and their nurses across larval

development (Fig 1). Based on our dual RNA-seq data, we infer social gene regulatory networks acting

between nurses and larvae to identify genes predicted to have important social regulatory effects. Finally,

we combine our measures of social regulatory effects with available population genomic data [48] to

characterize the patterns of molecular evolution of genes underlying nurse-larva interactions.

Results

Transcriptome-wide signatures of nurse-larva co-expression across larval development

To elucidate transcriptomic signatures of nurse-larva interactions, we performed RNA-sequencing on

worker-destined larvae across five developmental stages and nurses that fed larvae of each developmental

stage (termed "stage-specific" nurses; see Fig S1 for sampling scheme, Table S1 for list of samples),

building upon a previously published dataset focused on caste development in *M. pharaonis* [48]. We

hypothesized that if genes expressed in larvae regulate the expression of genes in nurse and vise versa, we

would observe correlated expression profiles across development in larvae and nurses (Fig 1). As a

biological control, we also collected "random nurses" that we observed feeding any stage of larvae in the

colony, and hence would not be expected to show correlated expression dynamics with larvae across the

five larval developmental stages. We also concurrently collected reproductive-destined larvae across the

last four developmental stages (reproductive-destined larvae can only be identified after the first larval

stage [50]), but unless clearly stated otherwise, all analyses were conducted only on worker-destined

larvae. To minimize experimental variance, we pooled ten individuals per sample. We sequenced whole

bodies of larvae but separated nurse heads and abdomens prior to sequencing.

We grouped genes into co-expression profiles or "modules" using an algorithm designed to

characterize gene co-expression dynamics across a short time series [51]. Each module represents a

standardized pre-defined expression profile, consisting of five values that each represent the $\log_2$ fold-

change between the given developmental stage and the initial stage (see Fig S2; this results in a total of 81

possible modules). We sorted genes into the module that most closely represented their expression profile

by Pearson correlation. We identified modules containing a greater than expected number of genes, where

we formed null expectations using permutation tests across developmental stage [51]. We identified such

significantly-enriched modules separately for larvae, stage-specific nurse heads, stage-specific nurse

abdomens, random nurse heads, and random nurse abdomens. We focused on both parallel (i.e. positive

regulation or activation) and anti-parallel (i.e. inhibitory) correlated expression patterns by identifying

significantly-enriched modules that were shared in both larvae and nurses (parallel), as well as

significantly-enriched modules for which the inverse of the module was identified as significantly-

enriched in the social partner (anti-parallel).

Larvae and stage-specific nurses shared many significantly-enriched modules (Table S2). These

modules contained the majority of genes expressed in nurses (stage-specific nurse head: 65%; abdomen:

76%) and a substantial proportion of the larval transcriptome (shared with stage-specific nurse head: 22%;

abdomen: 60%), such that there was a general widespread signature of correlated transcriptional patterns

between stage-specific nurses and larvae across larval development (Fig 2A-D). These coordinated

dynamics were dominated by parallel associations in nurse abdomens (possibly reflecting shared

metabolic pathways) but anti-parallel associations in nurse heads (possibly reflecting the social regulation

of larval growth). In contrast to stage-specific nurses, random nurses (our biological control) shared few

significantly-enriched modules with larvae (Table S2), and modules shared between random nurses and

larvae contained significantly fewer genes than modules shared between stage-specific nurses and larvae

(Fig 2E).

Identification of genes putatively involved in social interactions

Given that we observed transcriptome-wide patterns consistent with nurse-larva transcriptional co-

regulation across larval development, we next identified the genes that might be driving these patterns

(see Fig S3). We performed differential expression analysis to identify genes that varied in larval

expression according to larval developmental stage, as well as genes that varied in nurse expression

according to the developmental stage of larvae they fed. We identified 8125 differentially expressed

genes (DEGs) in larvae (78% of 10446 total genes). We identified 2057 and 1408 DEGs in stage-specific

nurse heads and abdomens, respectively, compared to 599 and 520 DEGs in random nurse heads and

abdomens, respectively. We removed genes differentially expressed in both stage-specific and random

nurses (N = 272 DEGs in heads, N = 140 DEGs in abdomens), which might differ among our colony

replicates due to random colony-specific effects that were not consistently associated with social

regulation of larval development. After this removal, we retained the top 1000 DEGs, sorted by P-value,

for each sample type other than random nurses (larvae, stage-specific nurse heads, stage-specific nurse

abdomens) for social gene regulatory network reconstruction.

Reconstruction of social gene regulatory networks

To infer gene-by-gene social regulatory relationships between nurses and larvae, we reconstructed gene

regulatory networks acting within and between nurses and larvae (Fig S3). To identify the most highly

connected (i.e. centrally located, upstream) genes of regulatory networks, we calculated within-tissue

connectivity and social connectivity by averaging the strength of connections across each connection a

gene made, differentiating between within-tissue (nurse-nurse or larva-larva) and social connections

(nurse-larva) (Fig 1B). On average, within-tissue connectivity was higher than social connectivity

(Wilcoxon rank-sum test; P < 0.001 in all tissues), and within-tissue connectivity was negatively

correlated with social connectivity in each tissue (Fig S4). The top enriched gene ontology terms based on

social connectivity in nurses were entirely dominated by metabolism (Tables S3,S4).

Secreted proteins and social gene regulation

While based on our data it is not possible to distinguish between genes that code for protein products that

are actually exchanged between nurses and larvae versus genes that affect behavior or physiology within

organisms (Fig 1A), proteins that are known to be cellularly secreted represent promising candidates for

the social regulation of larval development (40). Genes coding for proteins that are cellularly secreted in

Drosophila melanogaster had higher social connectivity than genes coding for non-secreted orthologs in

nurse heads (Fig 3A; Wilcoxon rank-sum test; $P = 0.011$), though not for nurse abdomens ($P = 0.094$).

While many of the genes with the highest social connectivity have unknown function, one of the genes

with the highest social connectivity within nurse heads was the protein *giant-lens* (Table S5). *Giant-lens*

(also known as *argos*) is a secreted protein that is known to inhibit EGFR signalling in *D. melanogaster*

[52]. *Giant-lens* expression in nurse heads was negatively correlated with the expression of the homolog

of *eps8*, human EGFR substrate 8 (Fig 3B; note however that this pattern was not seen for all genes in the

EGFR pathway). Furthermore, *giant-lens* expression in larvae steadily drops throughout development

(Fig S5), suggesting that *giant-lens* expressed in nurses and passed to larvae may inhibit EGFR signaling

in larvae, even as EGFR inhibition within larvae via *giant-lens* decreases over the course of development.

Interestingly, *eps8* does not exhibit a similar peak and drop in expression level in reproductive-destined

larvae in comparison to worker-destined larvae (Fig S6), suggesting that this pathway could be involved

in the regulation of caste development.

Molecular evolution of social gene regulatory networks

To investigate the selective pressures shaping social regulatory networks, we used population genomic

data from 22 resequenced *M. pharaonis* workers, using one *M. chinense* worker as an outgroup [48].

Using polymorphism and divergence data, we estimated gene-specific values of selective constraint,

which represents the intensity of purifying selection that genes experience [53]. To identify genes

disproportionately recruited to the core of social regulatory networks, we calculated "sociality index" as

the difference between social connectivity and within-tissue connectivity for each gene. Sociality index

was negatively correlated to selective constraint due to a positive correlation between within-tissue

connectivity and constraint and a negative correlation between social connectivity and constraint (Fig 4A-

C). Additionally, genes differed in sociality index according to their estimated evolutionary age, with

ancient genes exhibiting lower sociality indices than genes in younger age categories (Fig 4D). Finally,

while evolutionary age and evolutionary rate appear to be somewhat empirically confounded [54],

selective constraint and evolutionary age were each independently associated with sociality index, based

on a model including both variables as well as tissue type (GLM; LRT; evolutionary age: $\chi^2 = 21.536$, P <

0.001; selective constraint: $\chi^2 = 22.191$, P < 0.001).

Discussion

In organisms with extended offspring care, developmental programs are controlled in part by socially-

acting gene regulatory networks that operate between caregivers and developing offspring [14,41]. In this

study, we sequenced the transcriptomes of ant nurses and larvae as they interacted across larval

development to assess the effects of social interactions on gene expression dynamics. We found that large

sets of genes (i.e. modules) expressed in ant larvae and their caregiving adult nurses show correlated

changes in expression across development (Fig 2). The majority of nurse and larval transcriptomes was

represented in these correlated modules, indicating that the tight phenotypic co-regulation characterizing

nurse-larva interactions over the course of larval development is strongly reflected at the molecular level.

To characterize the overall network and evolutionary patterns of genes involved in nurse-larva

interactions, we reverse engineered nurse-larva gene regulatory networks and calculated the "social

connectivity" for each gene, defined as the sum of inferred social regulatory effects on all genes

expressed in social partners. We found that genes with high social connectivity tended to have low

within-individual connectivity (Fig S4; where within-individual connectivity is defined as the sum of

inferred regulatory effects acting within a given tissue). Nurse-expressed genes with higher sociality

indices (i.e disproportionately higher social connectivity than within-individual connectivity) tended to be

evolutionarily young and rapidly evolving due to relaxed selective constraint (Fig 4). Genes with high

social connectivity were enriched for a number of Gene Ontology (GO) categories associated with

metabolism (Tables S3,S4), consistent with the idea that molecular pathways associated with metabolism

are involved in the expression of social behavior [55,56]. Finally, we found that genes encoding for

orthologs of cellularly-secreted proteins important for intercellular signaling in *Drosophila melanogaster*

tended to exhibit higher levels of social connectivity than their non-secreted counterparts (Fig 3A).

One gene that stands out in terms of being cellularly secreted and exhibiting a relatively high

social connectivity is the gene *giant-lens*, which inhibits epidermal growth factor receptor (EGFR)

signaling [52]. EGFR signaling affects eye and wing development [57] and body size in *D. melanogaster*

[58], caste development in the honey bee *Apis mellifera* [58,59] via the transfer of royalactin from nurses

to larvae [58], and worker body size variation in the ant *Camponotus floridanus* [60]. While further

experimental work is necessary to ascertain whether *giant-lens* is actually orally secreted by nurses and

transferred to larvae, gene expression dynamics suggest that *giant-lens* transferred from nurses to larvae

may inhibit EGFR signaling at the end of larval development in worker-destined larvae (Fig 3B), but not

in reproductive-destined larvae (Fig S6). While caste in *M. pharaonis* is socially regulated in the first

larval stage [49], social inhibition of EGFR signaling could play a role in the regulation of worker body

size [60] or secondary caste phenotypes such as wings [61,62].

In terms of broad evolutionary patterns, our study complements previous results suggesting genes

with worker-biased expression (in comparison to genes with queen-biased expression) tend to be rapidly

evolving, evolutionarily young, and loosely connected in regulatory networks [37,48,49,63–65]. Because

workers are obligately sterile, their traits are shaped indirectly by kin selection, based on how they affect

the reproductive success of fertile relatives (i.e. queens and males) [23,66]. As a result, all-else-equal,

genes associated with worker traits are expected to evolve under relaxed selection relative to genes

associated with queen traits [67,68]. In general, the suite of genic characteristics commonly associated

with worker-biased genes (rapidly evolving, evolutionarily young, loosely connected) are all consistent

with relaxed selection acting on genes associated with workers [49]. Here, we show that within the

worker caste, genes that appear to be functionally involved in the expression of social behavior (i.e.

nursing) experience relaxed selective constraint relative to genes important for within-worker processes.

Therefore, the combination of kin selection as well rapid evolution thought to be characteristic of social

traits [25] likely act in concert to shape the labile evolutionary patterns commonly associated with

worker-biased genes.

In this study, we reconstructed regulatory networks acting between nurses and larvae, which of

course assumes that nurse gene expression changes as a function of larval stage fed. This is much more

likely to be the case when nurses are specialized on feeding particular larval stages. About half of all

feeding events were performed by specialist nurses in a previous study in *M. pharaonis* [40], so our

pooled nurse samples likely contained a mix of generalist and specialist nurses. This would serve to

effectively dilute the signal of nurse-larva interactions. Clearly, future studies would ideally focus entirely

on specialists, as well as on tissues such as brains and the specific exocrine glands [69] known to be

important for social behavior and communication. Despite these limitations, we were still able to observe

strong transcriptomic signatures of social interactions.

Conclusions

In this study, we uncovered strong transcriptomic signatures of social interactions (i.e. co-regulation of

nurse and larval genes and modules) and identified distinct evolutionary features of genes that were

inferred to have strong social regulatory roles. Because we simultaneously collected nurses and larvae

over a time series of interactions, we were able to elucidate the molecular basis of social interactions. This

is a promising approach that could be readily extended to study the molecular underpinnings of all forms

of social regulation in social insect colonies, including regulation of foraging, regulation of reproduction,

etc.. Furthermore, by adapting the methodology presented here (i.e. simultaneous collection over the

course of interactions followed by sequencing), the molecular mechanisms and evolutionary features of

genes underlying a diverse array of social interactions, including courtship behavior, dominance hierarchy

formation, and regulation of biofilm production could all be investigated. Overall, this study provides

empirical, theoretical, and methodological foundations on which future research can build upon to realize

the full power of the genomics revolution for understanding the genetic underpinnings and evolution of

interacting phenotypes.

Methods

This study builds on previous work investigating genomic signatures of kin selection in which we

characterized transcriptomic profiles from adult queens and workers, as well as queen- and worker-

destined larvae [48]. While stage-specific nurses were used in the previous analysis, the knowledge of the

developmental stage of larvae they fed was not, as they were simply treated as adult workers. This study

complements the past dataset with new data from random nurses, which were collected concurrently with

previous samples.

Study Design

We created thirty total replicate experimental colonies of approximately equal sizes (~300-400 workers,

~300-400 larvae) from a mixture of many genetically heterogeneous colonies of the ant *Monomorium*

pharaonis. We removed queens from ½ the study colonies to promote the production of reproductive-

destined larvae. In this study, we pooled samples across colonies with and without queens, as queen

presence appeared to have a negligible effect on gene expression (0 DEGs at FDR < 0.1 for each sample

type). We pre-assigned colonies to one of five larval developmental stages (labeled L1-L5, where L1 and

L2 refer to 1st-instar and 2nd-instar larvae and L3, L4, and L5 refer to small, medium, and large 3rd-

instar larvae [50]). We sampled individuals (larvae as well as nurses) across larval development time (3-4

days between each time point based on the developmental stage of the youngest larvae in colonies without

queens). From each colony, we sampled stage-specific nurses and worker-destined larvae, as well as

random nurses from colonies with queens and reproductive-destined larvae from colonies without queens

(starting at the L2 stage, because at L1 caste cannot be distinguished [50,70]. See Table S1 for full sample

list.

For each time point in each assigned colony, we collected stage-specific nurses, nurses feeding

larvae of the specified developmental stage (L1, L2, etc). Concurrently, we haphazardly collected random

nurses, nurses we observed feeding a larva of any developmental stage. After collecting nurses, we

anaesthetized the colony using carbon dioxide and collected larvae of the specified developmental stage.

We separated heads and abdomens of nurses to sequence separately and sequenced whole bodies of

larvae. We collected ten samples of each type to pool into single samples (separated by sample type and

tissue) for each colony.

We performed RNA-sequencing on all samples concurrently using Illumina HiSeq 2000 at

Okinawa Institute of Science and Technology Sequencing Center. Reads were mapped to the NCBI

version 2.0 *M. pharaonis* assembly [37], and we used RSEM [71] to estimate counts per locus and

fragments per kilobase mapped (FPKM) for each locus. For further details on RNA extraction and library

preparation, see [48].

Transcriptome-wide signatures of nurse-larva co-expression across larval development

We used an algorithm that categorizes genes based on their expression dynamics over time into a number

of modules represented by pre-defined expression profiles [51]; see Fig S2 for workflow). To create

modules, we started at 0 and either doubled, halved, or kept the expression level the same at each

subsequent stage, resulting in 81 possible modules ($3*3*3*3 = 81$; four stages after L1). To generate

gene-specific expression profiles based on real results, we calculated the average $\log_2$ fold change in

expression (FPKM) of the gene at each developmental stage compared to the initial expression level at

stage L1. We then assigned each gene to the closest module by Pearson correlation between gene

expression profile and module expression profile [51]. To identify significantly-enriched modules, we

generated null distributions of the number of genes present in each module (based on permutation of

expression over time), and retained modules with a significantly greater than expected number of genes

based on these null distributions (FDR < 0.05 after Bonferroni multiple correction [51]).

Identification of genes putatively involved in social interactions

We used the package EdgeR [72] to construct models including larval developmental stage and replicate

and performed differential expression analysis for each sample type separately. We retained genes

differentially expressed according to a nominal P-value of less than 0.05 (i.e. no false discovery

correction), as the purpose of this step was simply to identify genes that could be involved in interactions

that shape larval development (rather than spurious interactions arising from replicate-specific effects).

Social regulatory network reconstruction

We normalized expression for each gene using the inverse hyperbolic sine transformation of FPKM. As

input to the algorithm, we constructed "meta-samples" by combining expression data within the same

replicate and time point from nurses and larvae and labeling genes according to the tissue they were

expressed in, along the lines of previous host-symbiont studies [46,47]. We utilized the program GENIE3

[73], the top-performing individual program in a wide survey of GRN reconstruction programs [74], to

construct two types of networks: those acting between larvae and nurse heads, and those acting between

larvae and nurse abdomens. We repeated the reconstruction process 1000 times and averaged pairwise

connection strengths across runs. To capture the total effect of each gene on the transcriptome dynamics

within tissues, we averaged the regulatory effects each gene had on all other 999 genes expressed in the

same tissue ("within-individual connectivity"). Similarly, to capture the effect each gene had on the

transcriptome of social partners, we averaged regulatory effects each gene had on the 1000 genes

expressed in social partners ("social connectivity").

Estimation of selective constraint, and evolutionary rate

Previously, we performed whole-genome resequencing on 22 diploid *M. pharaonis* workers as well as

one diploid *M. chinense* worker to serve as an outgroup [48]. We estimated selective constraint using

MKtest2.0 [75], assuming an equal value of alpha (an estimate of the proportion of nonsynonymous

substitutions fixed by positive selection) across all genes. Selective constraint is the estimate of the

proportion of nonsynonymous mutations that are strongly deleterious and thereby do not contribute to

polymorphism or divergence [75]. Selective constraint is estimated using polymorphism data, so it

represents the strength of purifying selection genes experience within the study population [53].

Phylostratigraphic Analysis

Phylostrata are hierarchical taxonomic categories, reflecting the most inclusive taxonomic grouping for

which an ortholog of the given gene can be found [76–79]. We focused on distinguishing between genes

that were evolutionarily "ancient", present in non-insect animals, versus genes present in only insects,

hymenopterans, or ants [49]. We constructed a database containing 48 hymenopteran genomes, 10 insect

non-hymenopteran genomes, and 10 non-insect animal genomes, and estimated evolutionary age based on

the most evolutionarily distant identified BLASTp hit (E-value 10^{-10}).

Gene Set Enrichment Analysis

We performed gene set enrichment analysis based on social connectivity for each gene in each tissue

separately using the R package topGO [80]. We identified enriched gene ontology terms using

Kolmogorov-Smirnov tests ($P < 0.05$).

General Analyses

We performed all statistical analyses and generated all plots using R version in R version 3.4.0 [81], aided

by the packages "reshape2" [82], "plyr" [83], and "ggplot2" [84].

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571   **Fig 1. Social regulation of gene expression between ant nurses and larvae.**
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Fig 2. Nurse and larval transcriptomes show strong signatures of gene co-expression across larval development.

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597 **Fig 3. Genes encoding secreted proteins such as *giant-lens* are important for social gene regulation.**

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612 **Fig 4. Highly social genes tend to be less evolutionarily constrained.**
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L1
- larvae — x10
- stage-specific nurses — x10
- random nurses ? — x10

L2
- larvae — x10
- stage-specific nurses — x10
- random nurses ? — x10

L3
- larvae — x10
- stage-specific nurses — x10
- random nurses ? — x10

L4
- larvae — x10
- stage-specific nurses — x10
- random nurses ? — x10

L5
- larvae — x10
- stage-specific nurses — x10
- random nurses ? — x10

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616 **Fig S1. Diagram of sampling scheme.**
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Fig S2. Identification of significantly-enriched modules shared between larvae and nurses.

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644 **Figure S3. Workflow of preliminary differential expression analysis and gene regulatory network**
645 **reconstruction.**
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648   **Figure S4. Genes highly connected in within-tissue networks tended to have low values of social**
649   **connectivity.**

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657 **Fig S5. Expression of *giant-lens* in nurse heads and worker-destined larvae.**
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661 **Fig S6. Expression of *eps8* (epidermal growth factor receptor substrate 8) in worker-destined and**
662 **reproductive-destined larvae.**

development stage	sample type	number of samples
L1	larva (W/R)	5
	stage-specific nurse head	5
	stage-specific nurse abdomen	6
	random nurse head	3
	random nurse abdomen	2
L2	larva (W)	6
	larva (R)	3
	stage-specific nurse head	6
	stage-specific nurse abdomen	5
	random nurse head	3
	random nurse abdomen	3
L3	larva (W)	6
	larva (R)	3
	stage-specific nurse head	5
	stage-specific nurse abdomen	6
	random nurse head	3
	random nurse abdomen	3
L4	larva (W)	6
	larva (R)	3
	stage-specific nurse head	4
	stage-specific nurse abdomen	5
	random nurse head	2
	random nurse abdomen	2
L5	larva (W)	6
	larva (R)	3
	stage-specific nurse head	5
	stage-specific nurse abdomen	5
	random nurse head	3
	random nurse abdomen	3

673 **Table S1. Description of samples included in study.**
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	number of significantly-enriched modules	modules positively shared with larvae	modules negatively shared with larvae	number of genes in shared modules
stage-specific nurse head	9	0	5	6838
random nurse head	9	0	2	209
stage-specific nurse abdomen	21	13	4	7943
random nurse abdomen	10	0	1	1400

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677 **Table S2. Number of nurse significantly-enriched modules shared with larvae.**

GO.ID	Term	Annotated	Significant	Expected	P-value
GO:0044710	single-organism metabolic process	81	14	7.82	0.00011
GO:0055114	oxidation-reduction process	45	11	4.35	0.00230
GO:0019637	organophosphate metabolic process	18	1	1.74	0.00506
GO:0044711	single-organism biosynthetic process	18	0	1.74	0.00506
GO:0006629	lipid metabolic process	12	2	1.16	0.00565
GO:0009117	nucleotide metabolic process	14	0	1.35	0.00755
GO:0006812	cation transport	18	2	1.74	0.00781
GO:0015672	monovalent inorganic cation transport	11	1	1.06	0.00898
GO:0090407	organophosphate biosynthetic process	11	0	1.06	0.01033
GO:0055086	nucleobase-containing small molecule metabolic process	16	0	1.55	0.01156

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704 **Table S3. Nurse head social connectivity GO terms based on GSEA of social connectivity.**

GO.ID	Term	Annotated	Significant	Expected	P-value
GO:0055114	oxidation-reduction process	52	7	4.96	0.022
GO:0008152	metabolic process	246	25	23.46	0.033

723
724 **Table S4. Nurse abdomen GO terms based on GSEA of social connectivity.**

nurse head	nurse abdomen
Basement membrane-specific heparan sulfate proteoglycan core protein	Procollagen-lysine 2-oxoglutarate 5-dioxygenase 3
Collagen alpha-1(IV) chain	Collagen alpha-1(IV) chain
Spondin-1	Tubulointerstitial nephritis antigen-like
Serine proteinase stubble	Papilin
Angiotensin-converting enzyme (Fragment)	Semaphorin-2A
Thrombospondin-4	Transferrin
Protein giant-lens	Basement membrane-specific heparan sulfate proteoglycan core protein
Protein lev-9	Protein NPC2 homolog
Papilin	Testican-2
Glypican-6	Lysozyme

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746 **Table S5. Swiss-Prot annotations for the top genes top genes coding for secreted proteins, sorted by**
747 **social connectivity.**

 Figure Captions

Fig 1. Social regulation of gene expression between ant nurses and larvae.

(A) Cartoon depicting positive gene regulation (i.e. activation) between larvae and nurses, where gene 1 is

expressed in nurses and genes 2 and 3 are expressed in larvae. After the expression of gene 1 increases,

the expression of gene 2 increases as a result of the social interaction of nursing (depicted in [B]). This

can occur if gene 1 itself codes for a protein passed to larvae, if the mRNA transcript is passed directly, or

if gene 1 activates the expression of some other gene in nurses, which in turn is passed as mRNA (or

codes for a protein that is passed) to larvae. Following the increase in expression of gene 2, the expression

of gene 3, which is shown to be activated by gene 2, also increases. While we have depicted a time-lag in

this social regulation of gene expression, the time lags are likely too short to observe in our data, as larvae

were collected every 3-4 days across development. Therefore, correlated transcriptome dynamics over

development (see Fig 2) would reflect mechanisms shown here. (B) Gene regulatory networks act

between and within individuals engaged in social interactions. Blue boxes are genes expressed in larvae,

and red boxes are genes expressed in nurses. Solid lines depict regulatory interactions within tissues

(here, within larvae or within nurses), while dashed lines represent social connections (nurse-larva or vise

versa).

Fig 2. Nurse and larval transcriptomes show strong signatures of gene co-expression across larval

development.

Plots (A-D) depict the expression profiles of individual genes (light lines) as expressed in (A) nurse head,

and (B) nurse abdomens, as well as larvae (C) larvae, shared with nurse heads, and (D) larvae, shared

with nurse abdomens. Dark lines indicate the median expression values of all genes sorted into modules,

with pre-defined expression profiles of modules depicted in plot insets. Only the five shared modules

containing the most nurse-expressed genes are shown for clarity. Larval expression profiles are divided

by the nurse tissue they are shared with, such that (C) depicts larval gene expression shared with nurse

heads (A), while (D) depicts larval gene expression shared with nurse abdomens (B). Note that nurse

heads and larvae shared inversely-related expression profiles. (E) Stage-specific nurses have more genes

than random nurses in modules shared with larvae than do random nurses, reflecting more broad-scale co-

expression across development. Error bars indicate 95% confidence intervals derived from systematic

drop-1 jackknifing of nurse samples. N = 10944 genes total.

Fig 3. Genes encoding secreted proteins such as *giant-lens* are important for social gene regulation.

(A) Genes encoding for proteins that are secreted in *Drosophila melanogaster* exhibit higher social

connectivity (i.e. more strongly socially regulate larval expression) in nurse heads than genes encoding

for non-secreted proteins (Wilcoxon rank-sum test). (B) The protein *giant-lens* is one of the genes coding

for secreted proteins with the highest social connectivity in nurse heads. Based on our data, *giant-lens*

expressed in stage-specific nurse heads (red) appears to inhibit the expression of the homolog of human

EGFR substrate 8 (*eps8*) expressed in worker-destined larvae (blue). Expression at stage i is equal to

$\log_2(\text{expression}_i/\text{expression}_1)$, i.e. the ratio of expression at the given stage to expression at the initial

stage.

Fig 4. Highly social genes tend to be less evolutionarily constrained.

Selective constraint, estimated from whole-genome polymorphism data, is (A) positively correlated with

within-tissue connectivity (Spearman correlation; head: r = 0.122, P < 0.001; abdomen: r = 0.217, P <

0.001), but negatively correlated with (B) social connectivity (head: r = -0.090, P = 0.009; abdomen: r = -

0.150, P < 0.001) and (C) sociality index (head: r = -0.132, P < 0.001; abdomen: r = -0.223, P < 0.001),

where sociality index is the difference between social and within-tissue connectivity per gene. Each point

in (a-c) indicates a single gene, as expressed in nurse heads or abdomens. Lines indicate trendline from

linear model. (D) Highly social genes also tend to be taxonomically restricted. Individual points depict

average values across nurse heads and abdomens for all genes within each estimated age class, indicated

by labels on points. Error bars depict 95% confidence intervals from bootstrapping. Numbers in

parentheses indicate number of genes in each age class.

Fig S1. Diagram of sampling scheme.

We collected ten worker-destined larvae, ten stage-specific nurses, and ten random nurses from each

colony (six colonies per time point, where time points represent larval developmental stages L1, L2, etc).

We collected stage-specific nurses when we observed them feeding larvae of the given developmental

stage. We collected random nurses when we observed them feeding any stage of larvae. We identified the

five separate stages of larvae as in [50].

Fig S2. Identification of significantly-enriched modules shared between larvae and nurses.

Inset tables depict expression profiles of sample modules genes can be assigned to. First, we construct

modules using all possible expression profiles (top left bubble). Expression profiles consist of five values,

starting at zero, that indicate the $\log_2$ fold-change in expression from the initial value (at stage L1). At

each subsequent stage, we either double, halve, or keep the expression level the same. This process is

repeated to produce 81 (four stages after L1; $3*3*3*3 = 81$) total modules. Next, for each tissue

separately (here we depict workflow in larvae with yellow bubbles), we calculate individual gene

expression profiles as the $\log_2$ fold-change in expression from the initial value and assign genes to the

closest related module by Pearson correlation. Concurrently, we permute the developmental stage labels

for each gene and assign the stage-permuted genes to modules (repeated 1000 times). From these stage-

permuted results, we calculate the mean number of genes assigned to each module and treat this number

as a null expectation (as each expression profile is not equally likely to occur by chance). We then

identify significantly-enriched modules using a one-way binomial test (with the calculated mean as the

null), with a Bonferroni-corrected false discovery rate of 0.05. This entire process is repeated in a nurse

tissue and significantly-enriched modules are found (blue bubble). Finally, we compare significantly-

enriched modules between larvae and nurses and retain identical and inverse modules as shared profiles.

An example of an inversely related profile is shown in red, where larvae exhibit the enriched module [0,

0, -1, -2, -3] and nurses exhibit the inverse module, [0, 0, 1, 2, 3].

Figure S3. Workflow of preliminary differential expression analysis and gene regulatory network

reconstruction.

On the left, we identify putatively socially-acting genes through differential expression analysis. First, for

nurse heads and abdomens separately, we perform differential expression analysis in stage-specific and

random nurses to identify genes differentially expressed according to larval stage fed, using a nominal P-

value of 0.05. We remove genes differentially expressed in random nurses, as these correspond to colony-

specific environmental effects unrelated to social regulation of larval development. Next, we select the

top 1000 differentially expressed genes by P-value in stage-specific nurses (after removing those DE in

random nurses) as well as the top 1000 differentially expressed genes in larvae. From these genes, we

create "meta-samples" by combining gene expression of larvae and stage-specific nurses collected from

the same colony (separately for heads and abdomens), and labeling genes by the tissue they are expressed

in. Using these meta-samples, we perform gene regulatory reconstruction (right) to identify genes

expressed in nurses that regulate larval gene expression, and vise-versa. We repeat gene regulatory

reconstruction 1000 times and average connection strength across runs, as the algorithm is non-

deterministic. The output of gene regulatory reconstruction is a matrix of regulatory connections acting

between genes. From this matrix, we calculate the average connectivity for each gene, separating within-

tissue (larva-larva or nurse head-nurse head) from social (nurse-larva) connections. Genes with high

connectivity are predicted to interact with many genes, i.e. are central to the network. Finally, we

calculate each genes' sociality index as the difference between social connectivity and within-tissue

connectivity.

Figure S4. Genes highly connected in within-tissue networks tended to have low values of social

connectivity.

Connectivity is representative of the number and strength of regulatory connections each gene makes.

Points indicate the average connectivity for a given gene, as measured within-tissue (x-axis; i.e. larva-

larva or nurse-nurse) or socially (y-axis; i.e. larva-nurse). Points are colored by tissue the connectivity is

measured in (e.g., dark blue indicates genes expressed in larvae, with connectivity measured in networks

constructed with nurse abdomens). Spearman r = -0.166, -0.374, -0.276, -0.342 for the four tissues as

ordered in legend; P < 0.001 in all cases.

Fig S5. Expression of *giant-lens* in nurse heads and worker-destined larvae. Expression at stage i is

equal to $\log_2(\text{expression}_i/\text{expression}_1)$, i.e. the ratio of expression at the given stage to expression at the

initial stage.

Fig S6. Expression of *eps8* (epidermal growth factor receptor substrate 8) in worker-destined and

reproductive-destined larvae. Expression at stage i is equal to $\log_2(\text{expression}_i/\text{expression}_1)$, i.e. the ratio

of expression at the given stage to expression at the initial stage.

Table S1. Description of samples included in study. Worker-destined larvae are indicated by larva (W),

and reproductive-destined larvae are indicated by larva (R). Larval caste cannot be distinguished at the L1

stage, so L1 larvae are labeled larva (W/R). For network reconstruction, "meta" samples were used as

input for network reconstruction, in which genes were labeled by sample type and grouped such that each

gene contained a measurement of expression in worker-destined larvae, nurse heads, and nurse abdomens.

Because sampling was uneven after removing low-quality samples, we used the minimum number of

samples contained across tissues at a given stage for stage-specific nurse heads and abdomens, and

randomly dropped excess samples. Overall, 24 "aggregate" samples were used as input for gene

regulatory network reconstruction.

Table S2. Number of nurse significantly-enriched modules shared with larvae.

Significantly-enriched modules are defined as modules with a statistically significant number of genes

assigned, as determined by a permutation test (FDR < 0.05). Left column is the total number of

significant modules for each tissue, while the second and third columns indicate number shared with

larvae (out of 24 larval significantly-enriched modules). The last column indicates the total number of

genes in these shared modules.