

Title: Resources modulate developmental shifts but not infection tolerance upon coinfection in an insect system

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Author Contributions: A.T.T. conceived the project and provided funding; N.S. and A.T.T. designed the experiments. N.S. and S.L. executed all experiments except the gut assay (conducted by A.P.). S.B. and A.W. assembled the *T. confusum* transcriptome. D.A. performed all differential expression and most statistical analyses and made the figures, with contributions from A.T.T. and N.S. A.T.T., D.A., and N.S. wrote the manuscript, with edits from all authors.

2 **Abstract**

3 Energetic resources fuel immune responses and parasite growth within organisms, but it is
4 unclear whether energy allocation is sufficient to explain changes in infection outcomes under
5 the threat of multiple parasites. We manipulated diet in flour beetles (*Tribolium confusum*)
6 infected with two natural parasites to investigate the role of resources in shifting metabolic and
7 immune responses after single and co-infection. Our results suggest that gregarine parasites alter
8 the within-host energetic environment, and by extension juvenile development time, in a diet-
9 dependent manner. Gregarines do not affect host resistance to acute bacterial infection but do
10 stimulate the expression of an alternative set of immune genes and promote damage to the gut,
11 ultimately contributing to reduced survival regardless of diet. Thus, energy allocation is not
12 sufficient to explain the immunological contribution to coinfection outcomes, emphasizing the
13 importance of mechanistic insight for predicting the impact of coinfection across levels of
14 biological organization.

15 **Introduction**

16 If we know how individual stressors like infection affect traits and dynamics at a given
17 biological scale, can we predict how multiple stressors will function in tandem? To unite and
18 explain broad trends of species abundance, persistence, and ecosystem functioning in the face of
19 stress and change, ecological theory leans heavily on metabolic and stoichiometric models that
20 rely on assumptions about the flow and use of resources and energy at different scales of
21 biological organization (Ott *et al.* 2014; Bernot & Poulin 2018). For example, resource-focused
22 theory has promoted recent advances in our understanding of ecology within organisms and the
23 maintenance of microbiomes, symbionts, and parasites (Rynkiewicz *et al.* 2015). Whether
24 drawing on simplified resource allocation assumptions or more complex dynamic energy
25 budgets, within-host models have generated testable predictions for infection outcomes and the
26 oscillation and persistence of parasites (Cressler *et al.* 2014; Ramesh & Hall 2023) as they
27 directly or indirectly compete with immune systems for resources. As these frameworks grow in
28 popularity, however, it is worth asking about the extent to which resource allocation can fully
29 explain infection outcomes, particularly in the face of multiple stressors such as coinfection.

30 Hosts in nature are likely to endure exposure to parasites at multiple periods in their lives;
31 coinfection occurs when these exposure events lead to infection with two (or more) parasite
32 species simultaneously (Tate 2019). Since both species exploit host resources and generally
33 induce or modulate immune responses, they can facilitate or antagonize each other and lead to
34 infection and transmission outcomes that differ from single infection scenarios (reviewed in
35 (Rovenolt & Tate 2022)). To what extent can resource theory explain these outcomes? This
36 question largely depends on the relative sensitivity of parasites, immune dynamics, and damage
37 repair to resource conditions within the host (Graham 2008; Clay *et al.* 2023). After all, resource
38 limitation can alter allocation among life history traits and reconfigure immune system
39 investment (Adamo *et al.* 2016), altering the reception of incoming parasites and their
40 transmission potential (Vale *et al.* 2013). Sometimes, resource availability dominates parasite
41 competition from the bottom-up regardless of immunological regulation. In mice, for example,
42 gut nematodes destroy red blood cells and thereby limit malaria parasite propagation through
43 resource limitation even though the nematodes also suppress immune responses, which should
44 otherwise facilitate parasite replication and transmission (Griffiths *et al.* 2015). On the other

45 hand, helminth co-infection in African buffalo stimulates a T cell polarization state away from
46 the optimal regime needed to fight coccidia. While it is not clear – and indeed not probable – that
47 this immune shift is primarily resource-driven, it does lead to a more hospitable gut habitat for
48 the coccidia that ultimately increases parasite shedding at the (presumably energetic) expense of
49 host reproduction (Seguel *et al.* 2023).

50 The resource sensitivity of these molecular mechanisms can be more directly tested in
51 experiments that manipulate resource availability. Recent examples of coinfection outcomes in
52 insect systems demonstrate limited (Deschodt & Cory 2022) or mixed effects (Zilio & Koella
53 2020) of resource limitation on host and parasite fitness-associated traits, but these studies focus
54 on infection outcomes rather than the mechanisms that drive them. To what extent does a
55 primary infection alter the metabolic and immunological landscape encountered by a second
56 parasite species? Does resource allocation ultimately drive these differences, or should we be
57 selective about generalizing coinfection dynamics from energy budget models?

58 To test these questions, we turned to a model system for host-parasite population biology
59 – the confused flour beetle *Tribolium confusum* (Park 1948) and two of its natural parasites. The
60 first parasite, the eugregarine *Gregarina confusa*, induces a chronic but avirulent infection that
61 provides tractability for these questions by avoiding confounding effects of morbidity and
62 mortality (Detwiler & Janovy 2008; Thomas & Rudolf 2010). The second parasite is the
63 entomopathogenic bacterium *Bacillus thuringiensis* (Bt), which instigates acute mortality
64 (Behrens *et al.* 2014) and is sensitive to the immune dynamics of the host (Jent *et al.* 2019). We
65 divided larvae in a factorial design that included presence or absence of gregarine exposure and a
66 standard or nutrient-limited diet. To quantify the extent to which the primary parasite influences
67 the metabolic and immune landscape experienced by the second parasite, we used mRNA-seq to
68 investigate molecular signatures of metabolic and immunological shifts to gregarine infection
69 and measured diet-dependent development time and metabolite levels in gregarine-infected or
70 uninfected larvae. We then infected larvae with Bt and investigated transcriptomic signatures of
71 gregarine infection on immune and metabolic trajectories during acute Bt infection. We applied
72 these results to disentangle the relative impact of resources and immunity on gut pathology,
73 infection resistance, and disease-induced mortality. Our results suggest that while some
74 coinfection-induced shifts in life history parameters may be approximated with energy budget

75 assumptions alone, others are largely insensitive to resources and rely instead on shifts in
76 pathology associated with immune responses and damage repair. Since these parameters are
77 particularly important for predicting population dynamics and parasite-mediated apparent
78 competition outcomes at the community level (Johnson *et al.* 2015; Cortez & Duffy 2020;
79 Rovenolt & Tate 2022), ecologists should account for basic immunology before relying too
80 heavily on resource theory for coinfection models.

81

82 **Materials and Methods**

83 ***Beetle rearing, handling, and diet***

84 *Tribolium confusum* beetles for this experiment were derived from a stock colony collected in
85 2013 from Pennsylvania, USA (Tate & Graham 2015) and subsequently kept under laboratory
86 conditions (standard diet, 30C, in the dark). To create breeding groups for the experiments, we
87 took 60-80 adults per group from a colony and allowed them to lay eggs in 16g flour for 24
88 hours. We then combined eggs for all breeding groups and distributed them across experimental
89 diets. Besides age-matched eggs from the same egg-laying period, we also created staggered
90 breeding groups two days apart to derive larvae of different ages but equivalent sizes.

91 The *standard diet* for 200 larvae consists of 16g autoclaved whole-wheat flour (Fisher) and 5%
92 w/w brewer's yeast (Fleischmann) in a 100mm petri dish. To reduce the protein and nutrient
93 quality of the diet we excluded the yeast (*no-yeast diet*). Yeast is an important source of protein
94 and other nutrients, and larvae raised on this restricted diet will still develop into adulthood but
95 generally more slowly. In these experiments, all diets were further modified by reducing new
96 flour to 10g and adding 6g flour derived from gregarine-infected or clean mini-colonies as
97 described below.

98 ***Infection protocols***

99 To create infectious flour, we added beetles from gregarine-infected or uninfected stocks to clean
100 flour for four days, providing time to deposit infectious gregarine oocysts. We then removed the
101 beetles and used the flour to create the diets (gregarine exposure: 4g infected flour + 2g
102 uninfected flour; no exposure: 6g uninfected flour). This method resulted in a 60%+ prevalence

103 of gregarine infection in all experiments, as determined by gut microscopy (Thomas & Rudolf
104 2010), (q)PCR (f: CCTCGAGGAAGTTCGAGTCTAT, r: TTGACAGCTTGGGCACTTTAT,
105 400nM efficiency = 99.2%, T_m = 55C), and/or deposited gametocyst counts 18-24 hours after
106 temporary starvation (Janovy *et al.* 2007). Once ingested, gregarine trophozoites attach to the
107 guts for 7-10 days before finding a mate, forming a gametocyst, and evacuating the gut (Janovy
108 *et al.* 2007).

109 To challenge larvae with Bt, we produced bacterial cultures as described in (Jent *et al.* 2019).
110 Plating of the infection culture confirmed a concentration of 1.8×10^9 CFU/mL, which results in
111 an LD50 dose in *T. confusum*. To septicly inoculate larvae, we dipped a micro-dissection
112 needle in the bacterial (or saline control) aliquot and stabbed it into the space between the head
113 and second segments.

114 ***Development and metabolite assays and standardizing age vs stage + stats***

115 We counted freshly emerged pupae and dead larvae daily from day 22-30 post oviposition and
116 removed them to avoid resource-deprived larvae receiving additional protein from cannibalism
117 (Park *et al.* 1970). To collect size-equivalent larvae for the metabolite measurements, we
118 adjusted the collection dates according to the developmental delays in the different treatments,
119 *i.e.* greg+/yeast+ and greg-/yeast- larvae were collected three days later than the greg-/yeast+
120 control treatment, while the slowest greg+/yeast- larvae were collected a week later.

121 To measure the primary metabolites, we froze size-matched larvae after starvation, washed them
122 twice in cold insect saline, and homogenized them (n=18-21 larvae/diet-pathogen treatment). A
123 third of each sample went into each of the three performed measurements. We measured total
124 protein content in a Bradford assay (Schulz *et al.* 2023). For the glucose assay we used the GO
125 Assay Kit (Sigma) and for lipids a Vanillin assay (Abcam) (Barr *et al.* 2023).

126 We analyzed differences in larval development time to pupation among the diet and gregarine
127 treatment groups (N = 77-150 larvae/treatment) using log-rank survival analyses from the
128 “survival” package in R (R_Core_Team 2012; Therneau 2014). We used linear models to
129 confirm treatment-wise mass equivalence among larvae selected for metabolite assays (Bates D.
130 2010). After standardizing lipid, glucose, and protein measurements by individual larval mass,
131 we used generalized linear models with gamma distributions (due to positive values and some

132 right-skew) to evaluate differences among parasite and diet treatments and their interactions for
133 each metabolite (N = 17-64 larvae per treatment).

134 ***Survival assays***

135 To analyze Bt infection-induced mortality of gregarine-infected and uninfected larvae under
136 different resource conditions, we raised larvae from age-staggered breeding groups on
137 contaminated standard or no-yeast diets, size-standardized larvae from each treatment (because
138 smaller larvae generally have higher mortality regardless of treatment), and then infected the
139 beetles with an LD50 dose of bacteria (N = 48 Bt-infected and 8 saline-challenged
140 larvae/treatment/block for 4 experimental blocks). Bt induces rapid-onset mortality, generally
141 between 8 and 14 hours post infection (Jent *et al.* 2019). We excluded larvae that died early from
142 the trauma of inoculation, and then monitored mortality from 6-14 hours, performing a final
143 check at 24 hours as most larvae will have died or recovered from Bt infection by then (Tate *et*
144 *al.* 2017). We analyzed larval survival using Cox proportional hazards (coxme package in R
145 (Therneau & Therneau 2015)). Proportional hazards assumptions were not met (tested using the
146 coxzph function) because survival rates among Bt-infected and saline control beetles were so
147 drastically different. Therefore, we stratified by infection treatment; gregarine exposure, diet, and
148 their interaction served as main effects and experimental block as a random effect.

149 ***Gut integrity assays***

150 To determine whether the damage or immune responses instigated by gregarine parasites
151 accelerate mortality, we evaluated the gut barrier integrity of beetle larvae. We randomly
152 assigned eggs from breeding groups to the four gregarine-by-diet treatment groups in 96-well
153 microplates. After 20 days, we septically exposed larvae to an LD20 dose (1.6×10^6 CFU/mL) of
154 Bt or mock-infected them with insect saline. Subsequently, we placed the larvae on blue dye
155 food prepared with 2.5% FD&C blue dye no.1 (Spectrum Chemicals) using a protocol described
156 by (Zanchi *et al.* 2020). After 20 hours of feeding, we examined the distribution of blue food dye
157 under a microscope to detect any leakage in the gut-intestine barrier, scoring the beetles
158 exhibiting a blue "smurf" phenotype. We analyzed smurf proportions using binomial generalized
159 linear models with block as a main effect and then gregarine status, diet, and Bt infection status
160 as main and interacting effects.

161 ***Time series and RT-qPCR analysis of parasite loads***

162 To evaluate the impact of gregarine infection on host-Bt dynamics, we first collected gut samples
163 (N = 6 pools of 8 guts/gregarine treatment) from gregarine-exposed or clean larvae. We then
164 challenged larvae from these groups with a needle dipped in sterile saline (control) or the
165 experimental dose of live Bt. Beetles were sacrificed every two hours for the 12 hours of the
166 acute infection phase (n = 8-12 Bt-infected and 6 uninfected larvae/time point) and stored
167 individually at -80C. We extracted RNA using Qiagen RNeasy mini-kits, confirmed RNA
168 concentration using the Nanodrop, and then reverse-transcribed RNA into cDNA (VILO
169 mastermix). We quantified Bt load via RT-qPCR (SybrGreen) as previously described (Jent *et al.*
170 2019; Critchlow *et al.* 2024). We validated our qPCR primers (see above) on known gregarine
171 infected and uninfected samples to devise a threshold of detection and used these primers to
172 categorize gregarine-exposed samples as currently infected or not. We log-transformed the
173 linearized dCt values for normality (Jent *et al.* 2019) and used linear models in R (“lm” function)
174 to analyze the impact of Bt exposure, time, gregarine exposure or confirmed infection, and the
175 interaction of time and gregarines on relative Bt loads. Because bacterial loads bifurcate over
176 time, and variation in high-load beetles might not be captured across the entire load distribution,
177 we also used a Bt load threshold on samples from 6-12 hours post infection to characterize
178 beetles as high-load, and performed logistic regression (lme4 package, glm function, family =
179 binomial and link = logit) on high-load status vs gregarine exposure/infection. The results were
180 not sensitive to the chosen threshold or on whether larvae were merely exposed to gregarines or
181 actively infected.

182 ***Transcriptome assembly and annotation***

183 In addition to the gut samples, we chose whole-body larval RNA samples from the time
184 series (sample sizes and time points in **Table S1**) that exhibited a near-median bacterial load for
185 the treatment and time point, to avoid introducing load-induced variance (Tate & Graham 2017).
186 150bp paired end libraries were produced using the Illumina TruSeq kit and sequenced in a
187 single batch at the Vanderbilt VANTAGE core on the Illumina NovaSeq 6000 (complete
188 statistics in **Table S2**). Sequencing data is publicly available on NCBI Sequence Read Archive
189 (accession PRJNA771764). We first assessed RNAseq read quality using fastqc (Andrews 2010).
190 There is currently no published annotated genome for *T. confusum* so using only samples not

191 infected with gregarines, we assembled a *de novo* transcriptome using Trinity with default
192 settings (contig statistics in **Table S3**); quality filtering was performed within Trinity and reads
193 were assembled in paired-end mode (Grabherr *et al.* 2011). Highly similar transcripts were
194 clustered using cd-hit (Fu *et al.* 2012). To assess the quality of the assembly, the reads were
195 realigned to the assembled transcriptome using bowtie2 and the ExN50 statistic was calculated
196 within Trinity (Grabherr *et al.* 2011). To assess the completeness of the assembly, transcripts
197 were analyzed using BUSCO (Benchmarking Universal Single-Copy Orthologs, **Table S4**)
198 against an insect gene set (Manni *et al.* 2021).

199 We used kallisto v 0.48.0 to quantify gene expression (Bray *et al.* 2016) by performing
200 pseudo alignment of RNA-seq reads to the assembled transcriptome of *T. confusum* and
201 summing count or transcript per million (TPM) values across isoforms. Because we were
202 interested in achieving a high degree of accuracy for AMP-specific analyses, we also used
203 Coleoptera AMPs as training sets (**Table S5**) and constructed Hidden Markov Model (HMM)
204 profiles using HMMER (Finn *et al.* 2011) to annotate AMPs in the *T. confusum* proteome. Since
205 some of our analyses relied on annotation data from a well-developed genome (Herndon *et al.*
206 2020), we filtered bit scores to combine results from BlastP and Blastx to identify *T. confusum*
207 orthologs of *T. castaneum* genes (**Figs. S1, S2**). Full methodological details for analytical
208 pipelines and sample processing are described in the Supplementary Methods.

209 ***Differential expression analyses***

210 We used the DEseq2 (v 1.36) package in R (Love *et al.* 2014) to run three differential
211 expression (DE) analyses. In the first analysis, we identified DE genes in the gut upon gregarine
212 infection relative to uninfected guts (sample details in **Tables S1,2**). In the second and third
213 analyses, we identified DE genes upon Bt or coinfection in the whole body samples at six (N=4)
214 or eight (N = 6) hours post infection with Bt relative to uninfected beetles (N = 8) or beetles
215 infected only with gregarines (N=8). Within each time point, we modeled differential gene
216 expression in DEseq2 as expression ~ Bt status + gregarine status + their interaction (false-
217 discovery rate (FDR) corrected *P*-value < 0.05 (Benjamini & Hochberg 1995)). Because we were
218 concerned about type II error after FDR adjustment due to the large number of annotated but
219 low-expressed *T. confusum* genes, we also used the seSeq (2.30) package in R (Hardcastle &
220 Kelly 2010) to investigate the AMPs specifically. To this end, we divided samples into four

221 groups: genes that are not DE across samples, genes that are DE in samples infected by Bt, genes
222 that are DE in the gregarine-infected samples, and genes that are DE in co-infected samples
223 relative to other samples, and reported the posterior probability of differential expression. We
224 tried identifying TF binding sites upstream of relevant AMP genes but we have low confidence
225 in the accuracy so the results are not presented here.

226 We performed weighted gene co-expression network analysis (WGCNA (Langfelder & Horvath
227 2008)) on genes that have orthologs in *T. castaneum* to identify modules of co-expressed genes.
228 Genes with zero count values across all replicates were removed before analyses. Next, we
229 constructed a signed correlation matrix for each analysis (merging threshold = 0.25, minimum
230 module size = 30) using the count data and used them to identify positive or negative correlations
231 between the expression of genes in the network. We calculated the Pearson correlation of the
232 module eigengenes across samples to identify modules of co-expressed genes. Using the
233 associated *T. castaneum* ortholog gene ids, we performed gene ontology (GO) analyses with
234 DAVID (Huang *et al.* 2007) to find functional categories for co-expressed modules identified via
235 WGCNA. In addition, we performed Kyoto Encyclopedia of Genes and Genomes (KEGG)
236 analyses on differentially expressed genes using clusterProfiler package in R (Yu *et al.* 2012).

237 **Results**

238 ***Both restricted diet and gregarine infection prolong development time, but hosts can*** 239 ***compensate metabolically unless stressed by both***

240 In isolation, both yeast restriction (Fig. 1A, Log-rank test, N=150, developmental hazard ratio
241 (HR) = 0.76(0.6-0.96), p = 0.024) and gregarine infection (N=82, HR = 0.73(0.55-0.97), p =
242 0.029) significantly and equivalently prolonged larval development time by approximately one
243 day (Fig. 1B) relative to well-fed and uninfected reference larvae (N = 128). This effect was
244 exacerbated when gregarine infection and yeast restriction were combined, leading to
245 significantly slower development times than all other treatments (N = 77, HR relative to
246 reference = 0.46(0.35-0.62), p < 0.001).

247 To analyze metabolic profiles across the treatments, we minimized the confounding effect of
248 development time discrepancies by controlling for larval mass (Fig. 1C; N = 21 per treatment;
249 mass range 1.3-2.6mg; p > 0.6 among all pairwise treatment comparisons) rather than age or

250 instar, which is indeterminate in flour beetles. Neither diet nor gregarine infection significantly
251 affected mass-corrected lipid levels (Table 1), but the interaction of the two was significant, as
252 the infected and no-yeast group had reduced lipid stores (Fig. 1D; interaction $p = 0.027$). Diet
253 and its interaction with gregarines significantly predicted glucose levels in opposite directions
254 (Fig. 1E), as no-yeast diet larvae had a significantly higher glucose level than the reference group
255 (posthoc BH-corrected $p = 0.003$) but the infected and no-yeast group had significantly lower
256 glucose levels (interaction $p < 0.0001$). Protein (Fig. 1F) made up a significantly higher
257 proportion of larval body mass in no-yeast ($p = 0.005$) and gregarine-infected groups ($p = 0.002$),
258 likely indicating that a higher proportion of total mass is structural rather than stored resources;
259 the interaction effect was not significant. These values were not significantly dependent on
260 individual larval mass within treatments except for lipids in gregarine-exposed individuals (**Fig.**
261 **S3**), which increased for larger larvae in the gregarine-infected standard-diet treatment and
262 decreased in the gregarine-infected-low protein diet.

263 ***Gregarine-infected guts reveal altered metabolic and immunological profiles***

264 A principal component analysis (PCA) of transcriptomic profiles revealed a clear
265 separation of gregarine-infected and uninfected gut samples (**Fig. S4**). Upon DESeq2 analysis,
266 upregulated genes were enriched for ribosomal, cuticular, and glycolytic proteins, while
267 downregulated genes included a bacterial recognition protein (GNBP-1) as well as metabolic and
268 digestive enzymes such as apolipoproteins, lipases, trehalose transporters, cytochrome P450s,
269 cathepsin B, α -L-fucosidase, carboxypeptidase A, and juvenile hormone binding proteins (**Table**
270 **S6**). Zooming in on immune effector responses, we identified the significant differential
271 regulation of four upregulated and tightly co-expressed AMPs in the gut upon gregarine infection
272 ((Fig. 2A, **Table S7**) Defensin-1, Attacin-2, Attacin-3, and Cecropin-3) as well as two
273 downregulated AMPs (Cecropin-1 and PR5-3, which contains a thaumatin domain; Fig. 2B).
274 WGCNA analysis did not identify any specific gene co-expression modules that were
275 significantly associated with gregarine infection (**Table S8**), but KEGG analysis revealed
276 significant enrichment of the ribosome followed by non-significant hits on steroid synthesis and
277 signatures of altered carbohydrate metabolism (**Table S9**).

278

279 *Transcriptomic profiles from gregarine-infected and coinfecting larvae reveal altered*
280 *physiology and unique immune responses to single vs coinfection*

281 We evaluated transcriptomic profiles of whole-body samples from individual gregarine-infected
282 or uninfected larvae six or eight hours post co-infection with *Bacillus thuringiensis* (**Table S1**).
283 PCA showed effects of Bt or gregarine infection at six hours post-infection, but this separation
284 became muddled by eight hours post-infection (**Fig. S4**).

285 Focusing first on immune-specific analyses, we discovered that two AMPs (Coleoptericin-1 and
286 Attacin-1) were highly associated with the main effect of Bt infection at both six and eight hours
287 post infection (Fig. 2C). Meanwhile, gregarine infection was associated with the downregulation
288 of the same cecropin (Cecropin-1) originally downregulated in the gut, although this effect was
289 muted in the coinfecting samples (Fig. 2C). The same AMPs that were upregulated in the gut
290 upon gregarine infection were also uniquely upregulated in the whole-body coinfection samples
291 at both 6 and 8 hour time points (Fig. 1C; Defensin-1, Attacin-2, Attacin-3, and Cecropin-3).
292 While p-values for these four AMPs did not survive multiple corrections (adj $P > 0.05$) (**Tables**
293 **S10, S11, Fig. S5**), we hypothesized that this is due to a type-II error on the FDR because of the
294 unusually large number (>200k) of draft-annotated *T. confusum* genes. Therefore, we ran DE
295 analyses using the Bayseq package and calculated the posterior probabilities instead of P -values
296 (Fig. 2B), which dispenses with the need for multiple corrections. Posterior probabilities of
297 upregulation due to co-infection for these genes were close to one (**Fig. S5**) but were larger at six
298 hours post-infection compared to eight hours post-infection, suggesting that their expression is
299 reduced after six hours (e.g. TPM count in **Fig. 2C**).

300 We ran co-expression network analyses to identify gene modules that are uniquely co-
301 regulated according to each infection treatment (**Tables S12, S13**). We limited our analyses to
302 genes with orthologs in *T. castaneum* because running analyses using the complete dataset is
303 computationally intensive and because GO terms are not annotated for *T. confusum*. Our co-
304 expression results at six hours post infection were generally consistent with our AMP-specific
305 analyses, such that DE AMPs within each analysis belonged to the same modules (e.g., *Attacin-1*
306 and *Coleoptericin-1* both belong to the purple 6 hour WGCNA module; **Table S14**), and the
307 expression of genes in each module was consistent with the direction of differential expression
308 identified by DEseq2 (**Table S10**). Two modules showed significant associations with Bt

309 (purple) and co-infection (cyan) (**Fig. S6**), and the expression of genes within these modules was
310 almost exclusively linked to these specific treatments. The top GO and individual DE terms
311 associated with Bt infection included cell adhesion proteins and wound healing/immune defense
312 (e.g. hemocyanin activity, mucins, PGRP-SC2, AMPs, serine proteases and serpins; **Table S15**).
313 KEGG analyses indicated significant enrichment of the ribosome and oxidative phosphorylation
314 for the main effect of Bt at this time point, while the main effect of gregarines was once again
315 enriched for ribosomes and coinfection for sphingolipid metabolism. GO terms most strongly
316 associated with co-infection (cyan module) were primarily related to ribosomes, translation, and
317 protein synthesis (**Table S15**). Other co-infection associated modules (darkgreen, green,
318 lightcyan, and lightgreen) were enriched for chitin binding and various metabolic processes,
319 although these did not survive FDR correction (**Table S15**). The eight-hour DE, KEGG, and co-
320 expression results generally recapitulated the six-hour and gut results but had fewer modules
321 with weaker associations, possibly due to the resolution of the acute immune response (**Tables**
322 **S11, S13**).

323 ***The net effect of gregarine infection is increased disease-induced mortality upon coinfection***
324 ***associated with a reduction in gut tolerance to damage***

325 In size-matched larvae, gregarine infection significantly increased Bt-induced mortality relative
326 to uninfected and well-fed larvae and exhibited no significant interaction effect with diet (Fig.
327 3A). To determine whether the extra mortality was due to differences in resistance among
328 gregarine-infected and uninfected beetles, we measured Bt load via RT-qPCR across the 12 hour
329 acute phase when most mortality is initiated (Fig. 3B). While bacterial load significantly
330 increased and bifurcated over time as previously described in this and several other insect species
331 (Duneau *et al.* 2017; Tate *et al.* 2017; Franz *et al.* 2023), gregarine samples were equally
332 represented in high bacterial load trajectories (Bernoulli glm; $z = 0.69$, $p = 0.49$) relative to
333 gregarine-uninfected samples, and neither gregarine exposure, confirmed gregarine infection, or
334 the interaction of gregarines and time predicted bacterial load overall (Table 2, **Table S16**).
335 Thus, gregarine-infected individuals are not less resistant to Bt.

336 To understand whether tolerance mechanisms might instead account for the difference in
337 mortality, we employed a smurf assay, which indicates gut leakiness through failure to maintain
338 gut integrity or repair damaged structures (Fig. 3C). We found that both gregarine infection

339 (Table 2) $z = 2.6$, $p = 0.0093$) and Bt infection ($z = 2.7$, $p = 0.0074$) individually predicted
340 greater gut leakiness. Bt infection modestly increased smurf outcomes in coinfecting beetles but
341 not as drastically as in gregarine-free beetles ($z = -2.1$, $p = 0.033$). Neither diet alone nor its
342 interaction terms contributed significantly to smurf status.

343 **Discussion**

344 To what extent does a primary infection alter the metabolic and immunological landscape
345 encountered by a second parasite species, and does resource allocation ultimately drive these
346 differences? These questions are critical for building generalizable frameworks to predict the
347 consequences of coinfection in natural populations and at different levels of biological
348 organization. By manipulating resource availability and monitoring both metabolic and
349 immunological facets of the host response to coinfection, we tested the resource sensitivity of
350 key infection outcome parameters in a model system. Our results suggest that host development
351 time, which contributes to age-structured infection susceptibility (Clay *et al.* 2023) and
352 population intrinsic growth rate (Pearl *et al.* 1941; Park 1948), is exacerbated by the dual effects
353 of gregarines and resource limitation. On the other hand, disease-induced mortality, which
354 influences epidemiological dynamics and competitive outcomes in coinfecting assemblages
355 (Cortez & Duffy 2020; Rovenolt & Tate 2022), is not as sensitive to resources; differences are
356 instead attributable to immune-related pathology during coinfection. Thus, mechanistic models
357 that rely primarily on metabolic theory or energy budgets to predict coinfection dynamics are
358 likely to underestimate the contribution of immunological shifts.

359 In our study, both resource quality and gregarine infection affected development time,
360 presumably through slower storage of resources needed to grow. After accounting for
361 development rate (*i.e.* with mass- rather than age-matched larvae), the metabolic state of diet-
362 restricted and well-fed but gregarine-infected larvae largely catches up to their reference peers,
363 leaving only the dual-stressed group with major metabolic consequences. This indicates that
364 gregarines are capable of starving their hosts or forcing them to purge metabolites to avoid
365 oxidative stress (Li *et al.* 2020), but the effects are dramatic only under resource-limited
366 conditions; otherwise, the larvae appear to compensate by feeding more over a longer
367 developmental window. This result aligns with our general understanding of gregarine infections
368 as ubiquitous but relatively benign resource-exploiting parasites that inflict noticeable costs to

369 their insect hosts only under multiple stressors or high parasite burdens (Randall *et al.* 2013;
370 Wolz *et al.* 2022), and sets the stage for the phenotypes we observe upon Bt infection.

371 The mRNA-seq data suggest that gregarine infection alters the immune environment in
372 the gut through the differential regulation of antimicrobial peptides and other effectors. This
373 largely concurs with our previous study on gut gene expression after gregarine infection in the
374 related flour beetle *T. castaneum*, although the latter exhibited much broader downregulation of
375 antibacterial genes (Critchlow *et al.* 2019). This raises an interesting hypothesis that gregarines
376 may differentially affect susceptibility to coinfection in these two co-occurring and competing
377 host species (Park 1948; Rovenolt & Tate 2022). The transcriptional data also suggest that the
378 gregarines affect the gut metabolic environment more generally, which may be important for
379 nutrient processing and damage repair. The smurf assay indicates that diet does not significantly
380 affect gut integrity, whereas gregarine-infected individuals have a greater baseline ‘leakiness’
381 regardless of diet or coinfection status. This is clearly not enough to kill them in isolation, since
382 naïve and saline-stabbed larvae have low mortality rates regardless of their gregarine status (Fig.
383 3A). Bt infection initiates significant damage to the guts (also shown in (Critchlow *et al.* 2024)),
384 but once everyone is infected Bt, the impact of gregarine status on additional gut leakiness is
385 greatly diminished (Fig. 3C). Thus, there must be another contributor of pathology in gregarine-
386 infected individuals to explain the difference in Bt-induced mortality.

387 Is it the altered immune environment? One set of co-expressed AMPs is specific to
388 gregarine infection (as determined by gut expression) while another set is induced by Bt and not
389 gregarines in the whole body. It is interesting that at the whole-body level, the first set is highly
390 expressed specifically upon coinfection (rather than gregarines alone), suggesting that
391 coinfection activates a separate immune program in the fat body and other tissues beyond the
392 gut. Moving beyond specific AMPs, the WGCNA modules significantly associated with Bt
393 infection feature a lot of the same players previously identified in RNA-seq studies of Bt in flour
394 beetles (e.g. bacterial recognition, immune signaling and defense molecules, cytochrome P450s,
395 serine proteases, glycolysis enzymes (Behrens *et al.* 2014; Tate & Graham 2017)). The modules
396 most significantly associated with coinfection, however, are full of protein synthesis and
397 metabolic genes, suggesting a struggle to effectively manage the physiological response.
398 Interestingly, there is not an observable difference in bacterial load between gregarine-infected

399 and uninfected larvae, suggesting that neither mortality differences nor gene expression patterns
400 are due to differences in resistance. Instead, these coinfection-specific modules point to
401 differences in infection tolerance (Louie *et al.* 2016), possibly due to increased pathology of the
402 alternate immune responses and/or their co-expressed genes or an increased struggle to maintain
403 homeostatic metabolic and tissue repair programs. As we further improve the annotation of the *T.*
404 *confusum* genome, we will be able to test these hypotheses with functional genomics approaches.

405 While we did not directly measure the production and spread of transmission stages, our
406 results do hint at the consequences of coinfection for parasite fitness. Bt is an obligate killer, and
407 relies on making spores in its dying or newly dead host to achieve transmission (Garbutt *et al.*
408 2011). Gregarines, on the other hand, mate in the living host gut to produce oocysts that are shed
409 into the environment, and a dying host also spells a dead end for gregarines (Janovy *et al.* 2007).
410 Thus, the exacerbated mortality in coinfecting hosts undoubtedly hurts gregarine fitness, but it is
411 not entirely clear that Bt benefits because Bt loads were not higher in coinfecting individuals at
412 the time of peak mortality. Future studies would benefit from new protocols for accurately
413 quantifying gregarine transmission so that we can understand how this class of parasites,
414 ubiquitous in the arthropod world (Rueckert *et al.* 2019), influence disease dynamics for
415 biopesticides and vectored infections that preoccupy agricultural and biomedical efforts.

416 In conclusion, resources clearly matter – both the mRNA-seq and phenotype data suggest
417 that the gregarines are indeed acting like parasites in depriving their hosts of resources and
418 altering metabolic efficiency. Based on the metabolite data, the host can compensate for the
419 parasitism when resources aren't strictly limiting, but gregarine presence does change the
420 immunological landscape in the face of secondary infection and may exert additional
421 pathological effects. When it comes to infection mortality outcomes, the shifting immune
422 landscape and physical damage inflicted by the gut parasite overshadow the importance of
423 variance in resources. Thus, mechanistic models should allow for resource-independent
424 contributions of immune responses when predicting or generalizing coinfection dynamics.

425 **Acknowledgements**

426 We thank Justin Buchanan and James Deng for optimizing metabolic assay protocols for
427 *Tribolium*, Jakob Heiser for assistance with the smurf assay, and Jacob Steenwyk for assistance

428 with cd-hit. The experiments in this study were funded by NSF award 1753982 to A.T.T.; the *T.*
429 *confusum* transcriptome assembly was funded by NIH award R35GM138007 to A.T.T.

430 **Data Availability:** RNA-seq data is publicly available in the NCBI Sequence Read Archive
431 (project accession PRJNA771764). Experimental data will be available in Data Dryad (DOI in
432 progress).

433

434 **Supplementary Tables**

435	Table S1	Sample identity and replication for mRNA-seq
436	Table S2	Comprehensive RNA sample information (replicate number, sample description,
437		file size, bowtie statistics)
438	Table S3	Contig statistics
439	Table S4	Busco results
440	Table S5	Training sets used to identify AMPs via HMM
441	Table S6	Differential expression analysis for the gut upon infection with gregarine parasites
442	Table S7	AMP names and expression patterns
443	Table S8	WGCNA analysis results for the gut
444	Table S9	KEGG analysis on differentially expressed genes
445	Table S10	Differential expression analysis for the whole body 6 hours post-infection
446	Table S11	Differential expression analysis for the whole body 8 hours post-infection
447	Table S12	WGCNA analysis for the whole body 6 hours post-infection
448	Table S13	WGCNA analysis for the whole body 8 hours post-infection
449	Table S14	Modules of co-expressed AMPs using WGCNA on whole body samples
450	Table S15	GO analyses for modules significantly associated with Bt or co-infection
451	Table S16	Extended statistics for survival and bacterial load assays

452 **Supplementary Methods and Figures**

453	Figure S1.	Bit score values for BlastP and BlastX results.
454	Figure S2.	Expression of effector genes including antimicrobial peptides (AMPs) and
455		pathogenesis-related (PR) proteins.
456	Figure S3.	Relationship between individual larval mass and mass-normalized metabolite levels.
457	Figure S4.	Principal component analyses (PCAs) for three differential expression analyses.
458	Figure S5.	The posterior probabilities of differential expression using Bayseq.
459	Figure S6.	WGCNA analyses for the whole body 6 and 8 hours post-infection.

460 **References**

- 461 Adamo, S.A., Davies, G., Easy, R., Kovalko, I. & Turnbull, K.F. (2016). Reconfiguration of the
462 immune system network during food limitation in the caterpillar *Manduca sexta*. *The*
463 *Journal of Experimental Biology*.
- 464 Andrews, S. (2010). FastQC: A quality control tool for high throughput sequence data. *Reference*
465 *Source*.
- 466 Barr, J.S., Estevez-Lao, T.Y., Khalif, M., Saksena, S., Yarlagaadda, S., Farah, O. *et al.* (2023).
467 Temperature and age, individually and interactively, shape the size, weight, and body
468 composition of adult female mosquitoes. *Journal of Insect Physiology*, 148, 104525.
- 469 Bates D., M.M. (2010). lme4: linear mixed-effects models using S4 classes.
- 470 Behrens, S., Peuss, R., Milutinovi, B., Eggert, H., Esser, D., Rosenstiel, P. *et al.* (2014). Infection
471 routes matter in population-specific responses of the red flour beetle to the
472 entomopathogen *Bacillus thuringiensis*. *BMC Genomics*, 15, 445.
- 473 Benjamini, Y. & Hochberg, Y. (1995). Controlling the false discovery rate: a practical and
474 powerful approach to multiple testing. *Journal of the Royal statistical society: series B*
475 *(Methodological)*, 57, 289-300.
- 476 Bernot, R.J. & Poulin, R. (2018). Ecological Stoichiometry for Parasitologists. *Trends in*
477 *Parasitology*, 34, 928-933.
- 478 Bray, N.L., Pimentel, H., Melsted, P. & Pachter, L. (2016). Near-optimal probabilistic RNA-seq
479 quantification. *Nature biotechnology*, 34, 525-527.
- 480 Clay, P.A., Gattis, S., Garcia, J., Hernandez, V., Ben-Ami, F. & Duffy, M.A. (2023). Age
481 Structure Eliminates the Impact of Coinfection on Epidemic Dynamics in a Freshwater
482 Zooplankton System. *The American Naturalist*, 202, 785-799.
- 483 Cortez, M.H. & Duffy, M.A. (2020). Comparing the Indirect Effects between Exploiters in
484 Predator-Prey and Host-Pathogen Systems. *The American Naturalist*, 196, E144-E159.
- 485 Cressler, C.E., Nelson, W.A., Day, T. & McCauley, E. (2014). Disentangling the interaction
486 among host resources, the immune system and pathogens. *Ecology Letters*, 17, 284-293.
- 487 Critchlow, J.T., Norris, A. & Tate, A.T. (2019). The legacy of larval infection on immunological
488 dynamics over metamorphosis. *Philosophical Transactions of the Royal Society B:*
489 *Biological Sciences*, 374, 20190066.
- 490 Critchlow, J.T., Prakash, A., Zhong, K.Y. & Tate, A.T. (2024). Mapping the functional form of
491 the trade-off between infection resistance and reproductive fitness under dysregulated
492 immune signaling. *PLOS Pathogens*, 20, e1012049.
- 493 Deschodt, P.S. & Cory, J.S. (2022). Resource limitation has a limited impact on the outcome of
494 virus–fungus co-infection in an insect host. *Ecology and Evolution*, 12, e8707.
- 495 Detwiler, J. & Janovy, J., Jr. (2008). The role of phylogeny and ecology in experimental host
496 specificity: Insights from a eugregarine-host system. *Journal of Parasitology*, 94, 7-12.

- 497 Duneau, D., Ferdy, J.-B., Revah, J., Kondolf, H., Ortiz, G.A., Lazzaro, B.P. *et al.* (2017).
498 Stochastic variation in the initial phase of bacterial infection predicts the probability of
499 survival in *D. melanogaster*. *eLife*, 6, e28298.
- 500 Finn, R.D., Clements, J. & Eddy, S.R. (2011). HMMER web server: interactive sequence
501 similarity searching. *Nucleic acids research*, 39, W29-W37.
- 502 Franz, M., Armitage, S.A., Rolff, J. & Regoes, R.R. (2023). Virulence decomposition for
503 bifurcating infections. *Proceedings of the Royal Society B*, 290, 20230396.
- 504 Fu, L., Niu, B., Zhu, Z., Wu, S. & Li, W. (2012). CD-HIT: accelerated for clustering the next-
505 generation sequencing data. *Bioinformatics*, 28, 3150-3152.
- 506 Garbutt, J., Bonsall, M.B., Wright, D.J. & Raymond, B. (2011). Antagonistic competition
507 moderates virulence in *Bacillus thuringiensis*. *Ecology Letters*, 14, 765-772.
- 508 Grabherr, M.G., Haas, B.J., Yassour, M., Levin, J.Z., Thompson, D.A., Amit, I. *et al.* (2011).
509 Full-length transcriptome assembly from RNA-Seq data without a reference genome.
510 *Nature biotechnology*, 29, 644-652.
- 511 Graham, A.L. (2008). Ecological rules governing helminth–microparasite coinfection.
512 *Proceedings of the National Academy of Sciences*, 105, 566-570.
- 513 Griffiths, E.C., Fairlie-Clarke, K., Allen, J.E., Metcalf, C.J.E. & Graham, A.L. (2015). Bottom-
514 up regulation of malaria population dynamics in mice co-infected with lung-migratory
515 nematodes. *Ecology Letters*, 18, 1387-1396.
- 516 Hardcastle, T.J. & Kelly, K.A. (2010). baySeq: empirical Bayesian methods for identifying
517 differential expression in sequence count data. *BMC bioinformatics*, 11, 1-14.
- 518 Herndon, N., Shelton, J., Gerischer, L., Ioannidis, P., Ninova, M., Dönitz, J. *et al.* (2020).
519 Enhanced genome assembly and a new official gene set for *Tribolium castaneum*. *BMC*
520 *Genomics*, 21, 47.
- 521 Huang, D.W., Sherman, B.T., Tan, Q., Collins, J.R., Alvord, W.G., Roayaei, J. *et al.* (2007). The
522 DAVID Gene Functional Classification Tool: a novel biological module-centric
523 algorithm to functionally analyze large gene lists. *Genome biology*, 8, 1-16.
- 524 Janovy, J., Jr., Detwiler, J., Schwank, S., Bolek, M.G., Knipes, A.K. & Langford, G.J. (2007).
525 New and emended descriptions of gregarines from flour beetles (*Tribolium* spp. and
526 *Palorus subdepressus*: Coleoptera, Tenebrionidae). *Journal of Parasitology*, 93, 1155-
527 1170.
- 528 Jent, D., Perry, A., Critchlow, J. & Tate, A.T. (2019). Natural variation in the contribution of
529 microbial density to inducible immune dynamics. *Molecular Ecology*, 28, 5360-5372.
- 530 Johnson, P.T.J., de Roode, J.C. & Fenton, A. (2015). Why infectious disease research needs
531 community ecology. *Science*, 349, 1259504.
- 532 Langfelder, P. & Horvath, S. (2008). WGCNA: an R package for weighted correlation network
533 analysis. *BMC Bioinformatics*, 9, 559.
- 534 Li, X., Rommelaere, S., Kondo, S. & Lemaitre, B. (2020). Renal Purge of Hemolymphatic
535 Lipids Prevents the Accumulation of ROS-Induced Inflammatory Oxidized Lipids and
536 Protects *Drosophila* from Tissue Damage. *Immunity*, 52, 374-387.e376.

- 537 Louie, A., Song, K.H., Hotson, A., Thomas Tate, A. & Schneider, D.S. (2016). How Many
538 Parameters Does It Take to Describe Disease Tolerance? *PLoS Biol*, 14, e1002435.
- 539 Love, M.I., Huber, W. & Anders, S. (2014). Moderated estimation of fold change and dispersion
540 for RNA-seq data with DESeq2. *Genome Biology*, 15, 550.
- 541 Manni, M., Berkeley, M.R., Seppey, M., Simão, F.A. & Zdobnov, E.M. (2021). BUSCO update:
542 novel and streamlined workflows along with broader and deeper phylogenetic coverage
543 for scoring of eukaryotic, prokaryotic, and viral genomes. *Molecular biology and
544 evolution*, 38, 4647-4654.
- 545 Ott, D., Digel, C., Rall, B.C., Maraun, M., Scheu, S. & Brose, U. (2014). Unifying elemental
546 stoichiometry and metabolic theory in predicting species abundances. *Ecology Letters*,
547 17, 1247-1256.
- 548 Park, T. (1948). Interspecies Competition in Populations of *Trilobium confusum* Duval and
549 *Trilobium castaneum* Herbst. *Ecological Monographs*, 18, 265-307.
- 550 Park, T., Nathanson, M., Ziegler, J.R. & Mertz, D.B. (1970). Cannibalism of Pupae by Mixed-
551 Species Populations of Adult *Tribolium*. *Physiological Zoology*, 43, 166-184.
- 552 Pearl, R., Park, T. & Miner, J.R. (1941). Experimental Studies on the Duration of Life. XVI. Life
553 Tables for the Flour Beetle *Tribolium confusum* Duval. *The American Naturalist*, 75, 5-
554 19.
- 555 R_Core_Team (2012b). R: a language and environment for statistical computing. In: *Vienna,
556 Austria: R foundation for statistical computing*.
- 557 Ramesh, A. & Hall, S.R. (2023). Niche theory for within-host parasite dynamics: Analogies to
558 food web modules via feedback loops. *Ecology Letters*, 26, 351-368.
- 559 Randall, J., Cable, J., Guschina, I.A., Harwood, J.L. & Lello, J. (2013). Endemic infection
560 reduces transmission potential of an epidemic parasite during co-infection. *Proceedings
561 of the Royal Society of London B: Biological Sciences*, 280, 20131500.
- 562 Rovenolt, F. & Tate, A.T. (2022). The impact of coinfection dynamics on host competition and
563 coexistence. *The American Naturalist*, 199, 91-107.
- 564 Rueckert, S., Betts, E.L. & Tsaousis, A.D. (2019). The Symbiotic Spectrum: Where Do the
565 Gregarines Fit? *Trends in Parasitology*, 35, 687-694.
- 566 Rynkiewicz, E.C., Pedersen, A.B. & Fenton, A. (2015). An ecosystem approach to understanding
567 and managing within-host parasite community dynamics. *Trends in Parasitology*, 31,
568 212-221.
- 569 Schulz, N.K.E., Stewart, C.M. & Tate, A.T. (2023). Female investment in terminal reproduction
570 or somatic maintenance depends on infection dose. *Ecological Entomology*, 48, 714-724.
- 571 Seguel, M., Budischak, S.A., Jolles, A.E. & Ezenwa, V.O. (2023). Helminth-associated changes
572 in host immune phenotype connect top-down and bottom-up interactions during co-
573 infection. *Functional Ecology*, 37, 860-872.
- 574 Tate, A.T. (2019). Role of Multiple Infections on Immunological Variation in Wild Populations.
575 *mSystems*, 4, e00099-00019.

- 576 Tate, A.T., Andolfatto, P., Demuth, J.P. & Graham, A.L. (2017). The within-host dynamics of
577 infection in trans-generationally primed flour beetles. *Molecular Ecology*, 26, 3794–
578 3807.
- 579 Tate, A.T. & Graham, A.L. (2015). Trans-generational priming of resistance in wild flour beetles
580 reflects the primed phenotypes of laboratory populations and is inhibited by co-infection
581 with a common parasite. *Functional Ecology*, 29, 1059-1069.
- 582 Tate, A.T. & Graham, A.L. (2017). Dissecting the contributions of time and microbe density to
583 variation in immune gene expression. *Proceedings of the Royal Society B: Biological*
584 *Sciences*, 284.
- 585 Therneau, T. (2014). A Package for Survival Analysis in S. R package version 2.37-7. 2014.
- 586 Therneau, T.M. & Therneau, M.T.M. (2015). Package ‘coxme’. *R package version*, 2.
- 587 Thomas, A.M. & Rudolf, V.H. (2010). Challenges of metamorphosis in invertebrate hosts:
588 Maintaining parasite resistance across life-history stages. *Ecological Entomology*, 35,
589 200-205.
- 590 Vale, P., F., Choisy, M. & Little Tom, J. (2013). Host nutrition alters the variance in parasite
591 transmission potential. *Biology Letters*, 9, 20121145.
- 592 Wolz, M., Rueckert, S. & Müller, C. (2022). Fluctuating Starvation Conditions Modify Host-
593 Symbiont Relationship Between a Leaf Beetle and Its Newly Identified Gregarine
594 Species. *Frontiers in Ecology and Evolution*, 10.
- 595 Yu, G., Wang, L.-G., Han, Y. & He, Q.-Y. (2012). clusterProfiler: an R package for comparing
596 biological themes among gene clusters. *Omics: a journal of integrative biology*, 16, 284-
597 287.
- 598 Zanchi, C., Lindeza, A.S. & Kurtz, J. (2020). Comparative mortality and adaptation of a smurf
599 assay in two species of tenebrionid beetles exposed to *Bacillus thuringiensis*. *Insects*, 11,
600 261.
- 601 Zilio, G. & Koella, J.C. (2020). Sequential co-infections drive parasite competition and the
602 outcome of infection. *Journal of Animal Ecology*, 89, 2367-2377.
- 603

Table 1. The impact of diet, gregarine infection, and their interaction on within-host metabolites in mass-matched larvae

<i>Mass</i>	estimate	se	t	p
No yeast/uninfected	0.047	0.097	-0.49	0.63
Standard/infected	-0.03	0.097	-0.31	0.76
No yeast/infected	-0.011	0.097	-0.11	0.91
<i>Lipid</i>				
Intercept	0.926	0.100	9.3	<0.0001
Gregarine exposure	-0.088	0.140	-0.6	0.53
No-yeast diet	0.067	0.140	0.5	0.63
Gregarine: No-yeast	-0.412	0.197	-2.1	0.039
<i>Glucose</i>				
Intercept	0.965	0.087	11.2	<0.0001
Gregarine exposure	0.216	0.122	1.8	0.082
No-yeast diet	0.392	0.122	3.2	0.0020
Gregarine: No-yeast	-0.960	0.173	-5.5	<0.0001
<i>Protein</i>				
Intercept	1.423	0.051	28.0	<0.0001
Gregarine exposure	0.134	0.039	3.4	0.00064
No-yeast diet	0.121	0.039	3.1	0.0021

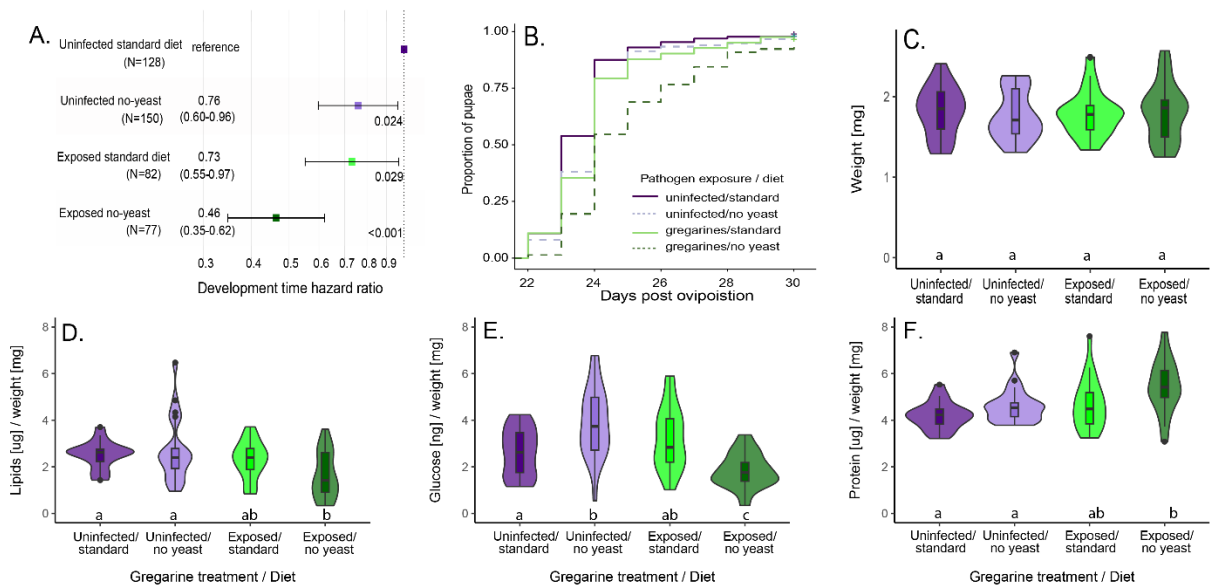
Notes: individual larval mass was analyzed using an lm evaluating all groups relative to the standard diet uninfected reference group. All metabolites were modeled using gamma-distributed glms and diet, gregarines, and their interactions as main effects. The interaction effect was dropped from the protein analysis as it was not significant and its exclusion yielded a better model fit.

Table 2. The impact of coinfection on survival, Bt resistance, and gut integrity

<i>Survival post Bt or saline challenge</i> ¹				
Factor	exp(coef)	se(coef)	z	p
Bt challenge	9.47	0.25	9.05	<0.0001
Gregarine exposure	1.39	0.14	2.39	0.017
No-yeast diet	1.15	0.14	0.98	0.32
Gregarines:no-yeast	0.93	0.19	-0.39	0.7
<i>Bacterial load over time by gregarine exposure status</i> ²				
	estimate	se	t	p
Intercept	-19.30	1.04	-18.57	< 0.001
Time post Bt challenge	0.33	0.13	2.56	0.012
Gregarine exposure	-0.77	1.42	-0.54	0.59
Time:gregarines	0.08	0.18	0.45	0.65
<i>Bacterial load over time by confirmed greg infection status</i> ²				
	estimate	se	t	p
Intercept	-19.43	0.83	-23.32	< 0.001
Time post Bt challenge	0.34	0.11	3.12	0.0023
Gregarine infection	-1.02	1.56	-0.66	0.51
Time:gregarines	0.15	0.20	0.76	0.45
<i>Proportion smurfs by gregarine exposure, Bt infection, and diet</i> ³				
	estimate	se	z	p
Intercept	-2.18	0.75	-2.90	0.0037
Experimental block	-0.39	0.31	-1.28	0.20
Gregarine exposure	2.21	0.85	2.60	0.0093
Bt challenge	2.23	0.83	2.68	0.0074
No-yeast diet	0.80	0.92	0.86	0.39
Gregarines:Bt	-2.14	1.01	-2.13	0.033
Gregarines:diet	-0.46	1.09	-0.42	0.67
Bt:diet	-0.93	1.07	-0.87	0.39
Gregarines:Bt:Diet	1.54	1.35	1.14	0.25

Notes: 1. Full Cox proportional hazards model is survival ~ Bt + gregarines*diet + (1|block). Exp(coef) is the hazard ratio. Stratifying by bacterial vs saline challenge did not qualitatively change results. 2. Full linear model is log2(relative Bt load) ~ time*gregarines. Includes only those larvae challenged with Bt; background amplification rates in controls were not significantly impacted by any factors (Table SJJ). Gregarine infection was confirmed by qPCR. 3. Full logistic model is smurf status ~ block +gregarines*diet*Bt. Reduced models did not yield a significantly better fit so the full factorial model is presented here.

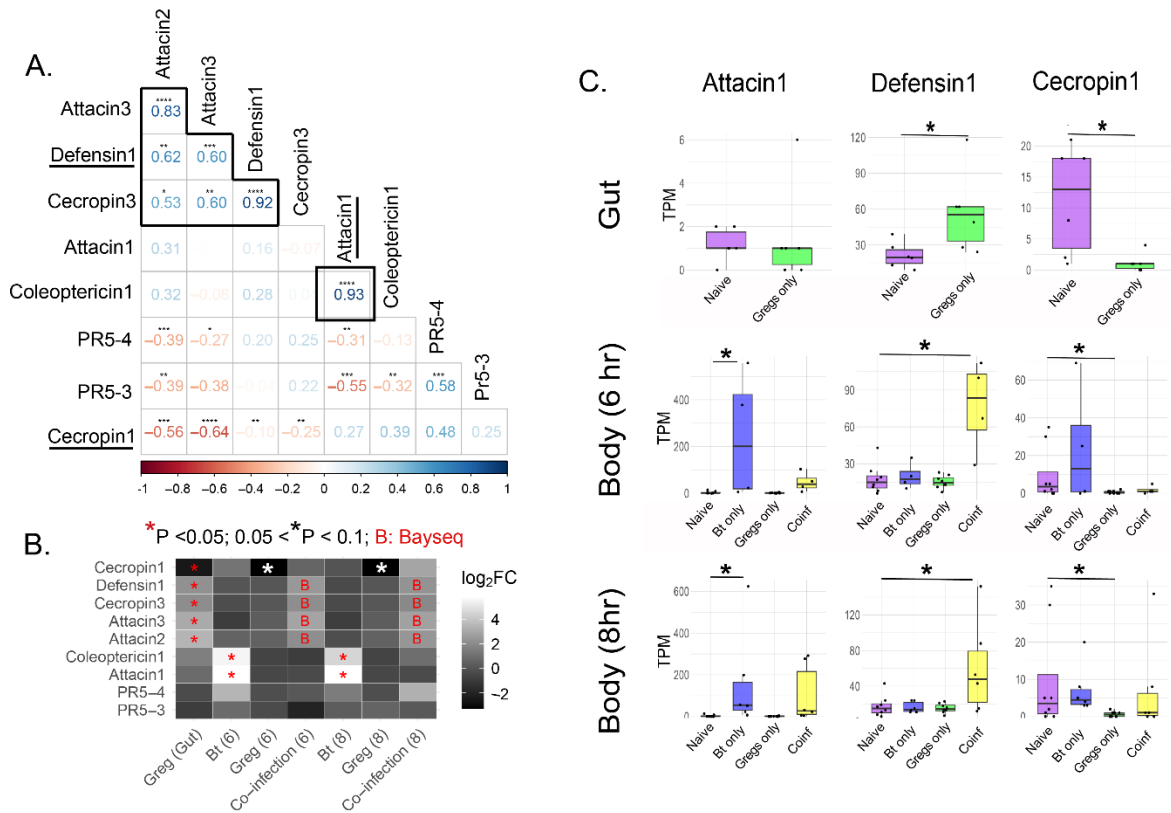
606 **Figure 1.** Effects of chronic gregarine infection, resource limitation, and their interaction on
 607 larval development and metabolic profiles. A) Log-rank statistics (hazard ratio of
 608 development time to pupation, 95% CI, and p values; note that a smaller HR means they
 609 develop more slowly) and B) development curves for the rate of larval development to the
 610 pupal stage for gregarine-infected larvae under standard or yeast-restricted diets relative to
 611 uninfected larvae on standard diets. In mass-standardized larvae (C), the effect of gregarine
 612 infection and diet on lipid (D), glucose (E), and protein content (F) were analyzed using
 613 GLMs with gamma distributions (Table 1); post-hoc pairwise test (BH-corrected) bins appear
 614 in lowercase letters.



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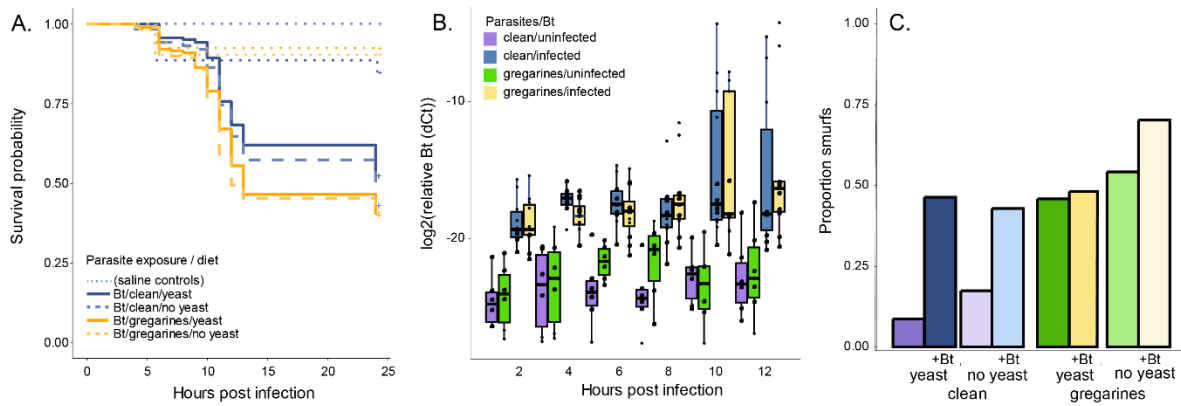
617 **Figure 2.** Effector protein (AMP) expression patterns upon infection with Bt, gregarines, and
 618 co-infection. A) The Pearson correlation matrix of expressed AMPs across all samples
 619 includes highlighted clusters of co-expressed effectors. B) Differential expression analysis of
 620 effectors across treatments showing log₂-fold change of effectors in the gut and at 6 and 8
 621 hours post Bt infection relative to naïve reference; DE genes delineated with asterisks. Four
 622 AMPs identified as DE upon co-infection using Bayseq are shown with red “B”. C)
 623 Expression patterns over time and treatments for three AMPs that each represent a cluster of
 624 co-expressed effectors (underlined in A). The Y-axis shows normalized counts (transcripts
 625 per million; TPM) and the x-axis shows treatments. Significant differences relative to naïve
 626 are shown with asterisks.



627

628

629 **Figure 3:** The impact of gregarines on host outcomes after Bt infection. A) Survival curves
630 illustrate that saline-challenged larvae (dotted lines; same colors as Bt-infected legend) have
631 high rates of survival regardless of treatment, but gregarine-infected larvae (yellow) are more
632 likely to die than clean larvae (blue) after Bt infection regardless of diet (dashed: no yeast).
633 B) Bacterial load relative to housekeeping gene (18s) was quantified via qPCR in Bt-infected
634 (blue: no gregs, yellow: gregs) and saline control larvae (purple: no gregs, green: gregs),
635 revealing no difference in Bt load based on gregarine status; all stats in Table SJJ. C) The
636 proportion of larvae revealing smurf (leaky gut) phenotypes by gregarine, diet, and Bt
637 infection treatment.



638