

1 The within-host ecology of insects and their parasites: integrating experiments and mathematical
2 models

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6 Running Title: Integrating experiments and models

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8

9 **Abstract**

10 The within-host ecology of hosts and their microbes involves complex feedbacks between the
11 host immune system, energetic resources, and microbial growth and virulence, which in turn
12 affect the probability of transmission to new hosts. This complexity can be challenging to
13 address with experiments alone, and mathematical models have traditionally played an essential
14 role in disentangling these processes, making new predictions, and bridging gaps across
15 biological scales. Insect hosts serve as uniquely powerful systems for the integration of
16 experiments and theory in disease biology. In this review, we highlight recent studies in fruit
17 flies, moths, beetles and other invertebrates that have inspired important mathematical models,
18 and present open questions arising from recent modeling efforts that are ripe for testing in
19 insects.

20

21 **Introduction**

22 Insects provide tractable systems to test hypotheses on host-parasite interactions,
23 immunity, and disease at multiple ecological and evolutionary scales. The benefits of testing
24 theories on insects stem from their short generation time, ease of keeping and breeding as well as
25 their relatively small body and genome sizes. For a vast range of species, we have in-depth
26 knowledge of their biology, including their immune systems and life history [1]. These insights
27 have given us tools to understand and even sometimes manipulate insect population and disease
28 dynamics in the spheres of agriculture, health, and ecosystem services [2, 3].

29 However, insects are different from vertebrates – and especially mammals – in almost
30 every aspect of life. Insects can be eusocial, and many undergo a dramatic transformation of
31 morphology during metamorphosis. They are ectothermic, relying on conditions in the external
32 environment for developmental and metabolic processes. Though they lack the classical adaptive
33 immune system found in vertebrates, they benefit from a robust and multi-faceted innate immune
34 system [4] capable of generating an alternate form of immune memory [5, 6]. These immune
35 systems face a vast diversity of parasites, many of which use obligate-killer and vertical
36 transmission modes not commonly observed in mammalian parasites [7]. As a result,
37 mathematical models of infection dynamics designed with vertebrates in mind rarely capture the
38 salient biological processes of insect systems. Fortunately, the unique tractability of insect
39 systems allows continuous iteration between experimental and theoretical approaches to
40 understand host-parasite interactions at multiple levels of biological organization.

41 In this review, we discuss some case studies of within-insect host dynamics, infection
42 outcomes, and transmission that have singularly benefited (or are ripe to benefit) from the
43 integration of experiments and theory. While work in disease vectors has long adopted this
44 approach, the associated literature is reviewed elsewhere (e.g. [8, 9]), and so here we highlight
45 other systems relevant to evolutionary ecology and infection biology.

46 **Within-host dynamics**

47 Scientists have long recognized that the magnitude of an immune response can spell the
48 difference between infected host survival or death, as supported by innumerable experiments in
49 insect strains sporting immune gene mutations [1]. These experiments have also suggested that
50 many infections in insects are a race against time between host control of microbial proliferation
51 and the microbial production of virulence factors that suppress the immune system and
52 pathologically destroy host tissue (**Fig. 1**) [10].

53 A recent study in *Drosophila melanogaster* added quantitative teeth to these observations
54 by demonstrating that, regardless of pathogenic bacterial strain, microbe loads bifurcate over the
55 acute infection phase into predictable trajectories of high load (and host death) or low-level
56 chronic infection [11] (e.g. **Fig. 1**). The authors proposed that variation in the relative frequency
57 of these outcomes could be described mathematically and predicted by a parameter describing
58 the time until onset of immunological control of the infection [11]. The subtle manipulation of
59 this parameter is not trivial from an experimental perspective, as it would require small
60 quantitative changes in immunological signaling rather than the blunter loss-of-function

61 phenotypes accomplished by genetics or RNAi methods. Moreover, there is no guarantee that
62 any given gene target would ultimately alter the rate of immunological induction in the early
63 phase of infection (e.g. **Fig. 1** parameter “*c*”) rather than, for example, the magnitude of effector
64 production when the host is already destined for death.

65 In cases like these, mechanistic mathematical models can help to sort through competing
66 hypotheses and identify targets for experimentation, as exemplified by a new modeling study that
67 explains within-host bifurcation trajectories and the persistence of chronic infection in insect
68 hosts [12]. The modeling results suggest that mutual negative feedbacks between the immune
69 response (e.g. antimicrobial peptide (AMP) production) and the production of
70 immunosuppressive molecules (e.g. proteases) by microbes are required to explain the
71 alternative trajectories. The results demonstrate that small differences in initial conditions,
72 including constitutive immunity and bacterial dose, are sufficient to produce bifurcation in
73 infection outcomes, as is modest variation in parameters describing the kinetics of inducible
74 immunity. The model further predicts that the observed chronic infections are only possible if the
75 protease production trades off with microbial growth rates.

76 Many of the model predictions are experimentally tractable and ripe for testing. For
77 example, the modeling results suggest that bistability is only possible if the host responds
78 immunologically to bacterial levels substantially below the lethal load. Previous experiments in
79 insects have already demonstrated variation in immunological sensitivity to microbe density [13-
80 16], providing evidence that this might be an ecologically relevant variable in nature. This
81 prediction could be directly tested using a microbial dose-response approach, or by targeting
82 particular nodes in immunological pathways predicted by more complex mechanistic models
83 (including an dynamic model of IMD pathway signaling presented in the supplement to [12]) to
84 alter sensitivity specifically. Finally, if the bifurcation dynamics do depend on negative
85 feedbacks from the microbe on host immunity, then experimental manipulation or ablation of
86 specific microbial proteases should alter the within-host trajectories in predictable ways.

87 While the crucial first few hours of host-pathogen interactions likely determines
88 dynamics of particularly virulent acute infections, the availability of energetic resources plays an
89 important role in directing within-host trajectories of many other insect host-symbiont
90 associations [17, 18]. There have been several modeling approaches aimed at capturing the
91 within-host ecology of immunity and parasites, often employing a dynamic energy-budget
92 framework to connect host resource acquisition, allocation to different life history traits, and
93 exploitation by parasites [19]. An open question, however, is whether parasites and the immune
94 system are actually competing for the same resources or using different ones, and how the
95 competition mode influences the role of resource acquisition on infection outcomes. To explore
96 this question, Cressler et al. [20] built a set of alternative models using ordinary differential
97 equations to allow immunity and parasites to directly compete for the same resource pool, allow
98 immunity or pathogens priority over resources, or force them to draw from independent energy
99 reserves. The results suggest that, as resource acquisition increases, pathogen load will increase
100 under a direct competition scenario but decrease if the immune system gets priority over
101 resources. These predictions were complemented by a literature search of empirical studies that

102 connect resource acquisition to parasite load. The results suggested that vertebrate animals tend
103 to have lower loads as food increases, favoring the immune priority model, while an increase in
104 food acquisition by invertebrates was most frequently associated with an increase in parasite load
105 [20].

106 Are insect immune systems more likely than vertebrates' to directly compete with
107 parasites for the same energetic resources? This is actually a difficult question to answer because
108 it requires knowledge of the metabolic needs of different arms of the immune system and the
109 particular parasites under investigation. Recent experimental studies are starting to resolve these
110 issues, allowing us to inch closer to connecting dynamic energy budget-style models with
111 empirical observations. For example, one study in *D. melanogaster* identified a metabolic switch,
112 initiated by parasitoid infection, that goads the fat body into mobilizing the carbohydrate
113 trehalose to fuel the cellular immune response [21]. Turning off this switch rendered the host
114 more susceptible to parasitoid infection, suggesting that in this scenario at least, the immune
115 response enjoys priority over competition for trehalose. Future work that quantifies the metabolic
116 needs of hosts and microbes, for example by manipulating specific metabolites within the host or
117 *in vitro*, would allow direct testing of model predictions.

118 A central challenge still facing our understanding of within-host dynamics in insects,
119 from both theoretical and empirical perspectives, is temporal structure. Models of within-host
120 dynamics frequently assume that there are constitutive and inducible components to immunity,
121 with the latter generated only after recognition of infection [22, 23]. We know from experiments
122 that the inducible response, potentially to minimize energetic and immunopathological costs
123 [24], is not just turned on all at once, but layered over time and in response to variable cues like
124 microbe density and damage [13]. In bumblebees (*Bombus terrestris*) and mealworm beetles
125 (*Tenebrio molitor*), for example, activated phenoloxidase appears to be depleted within minutes
126 of an immune challenge and not subsequently upregulated, while AMPs are induced gradually
127 but steadily after infection [25, 26]. When juxtaposed with microbial temporal structure,
128 including colonization bottlenecks [27], density-dependent quorum-sensing [28], and other forms
129 of plasticity, the relative importance of a particular immune effector for infection outcomes can
130 change rapidly. We recommend that future models of immune system dynamics and evolution
131 account for the constraints, costs and benefits associated with the temporal structure and
132 regulation of multiple immune effectors.

133 **Connecting within-host to between-host dynamics**

134 Since Anderson and May demonstrated with a simple model that baculoviruses could
135 plausibly act as key regulators of moth population cycles [29], a large body of work integrating
136 experiments and theory has embraced insect systems to answer fundamental questions in disease
137 ecology. Currently, one of the most intriguing research avenues in the field is to understand how
138 differences among individuals, for example due to immunological variation, infection history,
139 sex, or life stage, scale up to influence disease transmission and prevalence in host populations.
140 Given the tractability of their generation times, population sizes, and ease of manipulation
141 relative to vertebrates, insects provide ideal systems for conducting carefully controlled
142 experiments that integrate within- and between-host processes.

143 Moths and their baculoviruses remain central to this effort. Empirical work in the gypsy
144 moth (*Lymantria dispar*) and the Indian meal moth (*Plodia interpunctella*) has demonstrated that
145 host susceptibility and transmission potential both depend on the larval instar of the host [30,
146 31], reinforcing the need for subsequent mathematical models of insect disease dynamics to
147 account for life stage [32]. A recent study in the gypsy moth employed stochastic modeling of
148 empirical data to demonstrate that both inter-host variation and viral demographic stochasticity
149 can explain observed variation in infection outcomes [27], emphasizing that deterministic models
150 alone may not be sufficient to capture relevant dynamics. Studies in these systems have also
151 integrated experiments and theory at multiple spatial scales to demonstrate that the behavior of
152 moth individuals can influence disease dynamics through structuring contact networks [33], and
153 that local networks cannot always explain disease dynamics at larger scales [34]. These insects
154 are likely to remain powerful model systems for connecting disease dynamics across biological
155 levels of organization due to their tractability, natural history, and tradition of attracting
156 researchers interested in the marriage of theory and experiments.

157 The naturally occurring symbionts of traditional model insects have also lately inspired
158 integrative studies connecting within- and between-host dynamics. For example, recent work
159 using *Drosophila C* virus in *D. melanogaster* took advantage of its enviable genetic resources to
160 demonstrate variation in susceptibility and viral shedding dynamics of hosts from different
161 genetic backgrounds and sexes [35, 36]. Stochastic individual-based models inspired by this data
162 suggest that individual heterogeneity can substantially alter the subsequent dynamics of
163 outbreaks in theoretical fruit fly populations [35]. Moreover, insect hosts are commonly infected
164 with more than one type of parasite at a time, and the within-host interactions among parasites,
165 including facilitation or antagonistic competition, can alter infection outcomes for both hosts and
166 parasites. Inspired by parasite experiments in flour beetle model systems (*Tribolium* spp.) [37-
167 39], a recent modeling study [40] bridged ecological scales to explore the impact of within-host
168 interactions among parasite species (e.g. bacteria and protozoa) on disease transmission and
169 competition dynamics among beetle host species. The results suggest that asymmetrical mortality
170 rates among coinfecting host species could alter the transmission of each parasite, flipping
171 competitive exclusion outcomes from one host species to the other [40]. Finally, while water
172 fleas (*Daphnia* spp.) are not insects, the extensive body of integrative empirical and theoretical
173 studies in this system have enriched our understanding of concepts – e.g. susceptibility [41, 42],
174 parasite transmission modes [43], and multiple infections [44, 45] – directly relevant to
175 connecting within- and between-host dynamics in insects.

176 **Conclusions**

177 While we have focused here on the ecology of host-parasite interactions, insects and
178 other invertebrates have also served as powerful models for the evolution (e.g. [46, 47]) and
179 coevolution (e.g. [48-51]) of hosts and symbionts. Insects offer unique and alluring advantages to
180 both modelers and experimentalists, including a simpler (though still not simple) and tractable
181 innate immune system, an intriguing distribution of symbiont transmission modes not seen in
182 vertebrates, and ecological breadth and diversity abetted by metamorphosis. Their importance as
183 agricultural pests and pollinators, disease vectors, and contributors to ecosystem services adds

184 instant applied relevance to basic research on insect disease ecology. As we attempt to scale
185 within-host dynamics from controlled laboratory conditions to multi-host, multi-parasite natural
186 systems, the tight integration of theory and experiments will be essential to make sense of
187 complex interactions and outcomes at different levels of biological organization.

188 **Declarations of Interest:** This work was supported in part by a NSF DEB award #1753982 to
189 A.T.T.

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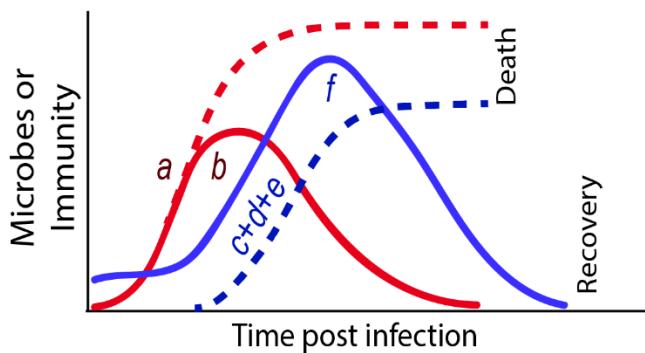
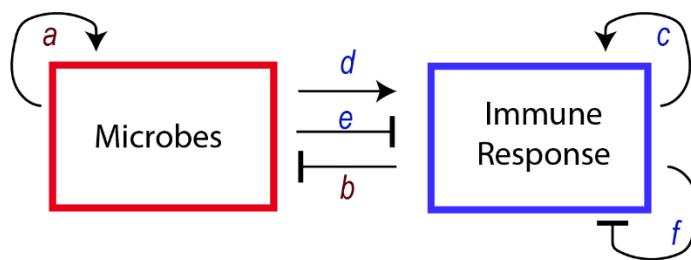
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337 **Figure 1:** Mathematical models help formalize empirical observations of within-host processes
 338 to explain observed variance in infection trajectories and transmission. The parameters
 339 associated with variation in infection outcomes among hosts (solid = recovered, dashed =
 340 infection-induced mortality, red = microbes, blue = immunity) can be expressed mathematically.
 341 For example, rates of microbial growth (a), immune-induced microbial death (b), microbe-
 342 independent (c) and microbe-dependent (d) inducible immune induction, microbial suppression
 343 of immunity (e) and negative feedback on immunity (f) exert individual and combined effects on
 344 the slope and magnitude of microbial and immunological dynamics (as suggested by color-coded
 345 letter placement on trajectories over time), although many of these parameters affect multiple
 346 aspects of the within-host dynamics and are not mutually exclusive. Initial conditions, including
 347 levels of constitutive immunity and microbial dose (as represented by the intercept at time = 0),
 348 can also affect outcomes by altering the dynamics of host-microbe interactions early in infection.
 349 The connection between microbe load, infection outcome, and transmission rate depends on
 350 microbial transmission mode, as obligate killer transmission (common for insect pathogens)
 351 requires host death induced by high microbe loads, while host death would stymy the
 352 transmission of directly transmitted microbes.



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