

1 **Highlights**

2 Integrating microorganisms into community-level ecology remains a critical challenge
3 despite their important influence on macro-organisms.

4 Insects and their symbionts are a tractable and powerful system to uncover how
5 microbial associations reshape ecological interactions.

6 Network analysis provides transformative tools to integrate symbionts into ecological
7 frameworks and to test key hypotheses on their community-level roles.

8 Embedding symbionts into networks can expose hidden drivers and advance
9 mechanistic understanding of community functioning under different scenarios.

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11 **Mapping microbial symbiont impacts using ecological networks**

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19 indirect effects, multilayer networks

20 **Abstract**

21 Microbial symbionts can dramatically reshape host biology, triggering cascading effects
22 that alter the strength, direction, or even presence of interactions among macro-
23 organisms. However, integrating microbes into community-level understanding remains
24 a major ecological challenge. Using insect–facultative symbiont systems as a model, we
25 show how ecological network analyses can reveal these hidden layers of complexity for
26 a deeper, mechanistic understanding of community functioning. By combining testable
27 hypotheses with tools such as multilayer networks or stochastic co-extinction models
28 and interaction rewiring, we provide a roadmap for uncovering how symbionts drive
29 ecological and evolutionary dynamics, opening new possibilities for empirical studies
30 that link microbial traits to ecosystem-level impacts.

31 *Integrating Microbial Symbionts into Ecological Communities*

32 Ecosystems consist of dynamic networks of interacting species, in which macro-
33 organisms associate with diverse internal and external microorganisms [1]. These
34 microbes can influence their hosts' biology both directly (e.g., facilitating nutrient
35 acquisition, causing disease) [2, 3] and indirectly (e.g., modifying interactions among
36 host organisms or between organisms and their environment) [4]. Relationships with
37 micro-organisms are particularly important for insects, many of which have evolved
38 specialised associations with bacteria, protists, fungi or viruses [5,6]. Disentangling
39 microbial symbionts' effects on insect community ecology requires incorporating the
40 hidden presence of symbionts within the food web context.

41 *Insect–Symbiont Relationships: Dynamics and Evolution*

42 The extent of interdependence between insects and their symbiotic partners varies
43 greatly. Obligate symbionts are essential for the survival and reproduction of both parties
44 [7]. These associations are evolutionarily ancient, maintained through strict maternal
45 transmission [8], and often have a nutritional function to supplement imbalanced diets
46 [9]. Others mediate antagonistic interactions, as in the symbiotic viruses of ichneumonid
47 and braconid parasitoids that suppress host immune systems [10]. Owing to their
48 extreme functional integration, obligate symbionts and their hosts operate as a single
49 biological unit within ecological communities and can be treated as single entities when
50 analysing species interactions.

51 Facultative symbionts are not required for host growth and reproduction and are not
52 universally present across natural insect populations [11]. The symbionts rely on hosts
53 for transmission, but the evolutionary interests of host and symbiont are not necessarily

54 aligned, resulting in interactions that span the parasitism–mutualism spectrum. Some
55 facultative symbionts can confer benefits, such as defending their hosts against
56 predators, parasitoids and pathogens [12] or protecting from abiotic challenges such as
57 heat shock [13]. However, another common strategy is manipulating host reproduction
58 to promote their own transmission. Maternally-transmitted facultative symbionts favour
59 a female bias in host offspring sex ratio [14], producing a variety of manipulations—
60 feminisation (see Glossary), parthenogenesis, male killing and cytoplasmic
61 incompatibility [15, 16]. Reproductive manipulation and symbiont-mediated protection
62 are not mutually exclusive [17], and many facultative symbionts function as conditional
63 mutualists: beneficial in certain ecological situations, and costly in others.

64 Facultative symbionts can influence the outcome of insect pairwise interactions, such
65 as those between prey species and their natural enemies [18,19]. Recent studies have
66 broadened this perspective by examining effects of symbionts within tri-trophic systems,
67 such as plant–herbivore–parasitoid (e.g. [20,21]) and host–parasitoid–hyperparasitoid
68 [22] interactions. Symbiont-mediated traits could propagate through ecosystems,
69 modifying the intensity, direction, or the existence of interactions between macro-
70 organisms [23]. We propose using tools from ecological network theory as a framework
71 for quantifying and analysing impact of facultative symbiotic interactions at the
72 community level.

73 *The Role of Symbionts in Ecological Networks*

74 **Application of network analysis in insect–endosymbiont communities**

75 Ecological network analysis quantifies interactions among species to determine
76 structure, function and dynamics of ecological systems [24]. An ecological network is a

77 map of nodes (e.g. species, populations, functional groups) connected by edges (e.g.
78 trophic interactions, mutualism or competition), providing a visual and mathematical
79 representation of a community. The construction and analysis of host–symbiont
80 networks involves the collection of interaction data, typically weighted by the strength or
81 frequency of these interactions [25]; their representation as matrices detailing the links
82 between nodes (Box 1; Box 2); extraction of metrics to interpret the structure [26]; and
83 mathematical modelling to predict responses to change [27,28].

84 *Approaches to Integrating Symbionts in a Network*

85 Incorporating microbial symbionts into ecological network analyses requires careful
86 consideration of the research question and study system. We suggest four approaches
87 to describe different structural representations; outline how hosts, symbionts, and their
88 associations can be treated in a network; and discuss how each approach can be used
89 to explore processes such as coextinction, infection spread, interaction rewiring, or
90 responses to change (Figure 1). Several processes (e.g. responses to change) can be
91 analysed across approaches; what differs is how each framework represents these
92 processes structurally.

93 *Approach 1: Treating symbionts as independent nodes.* This approach is suitable for
94 complex ecological systems, in which individual symbiont infection status cannot be
95 monitored (e.g. observational studies, manipulated mixed infected–uninfected
96 populations). Symbionts are considered separate nodes interacting directly and
97 indirectly with other community members, capturing their effects without individual-
98 level data. For example, the bacterium *Hamiltonella defensa* infecting the whitefly
99 *Bemisia tabaci* provides several phenotypic advantages on tomato plants, suppressing

100 plant defence and increasing whitefly reproduction [29, 30]. In *B.tabaci* biotype B, an *H.*
101 *defensa*-produced protein interacts with the *Tomato yellow leaf curl virus*, facilitating its
102 transmission to the plant [31]. These changes can potentially benefit other herbivores
103 and impact predator–prey dynamics. Representing the symbiont as an independent
104 node linked to the host by infection frequency (Figure 1) would capture its community-
105 level effects, such as altered predatory dynamics, or changes in **apparent mutualism**. A
106 similar approach has been applied in the analysis of parasites as independent nodes
107 [32]. In estuarine ecosystems, parasites can form structural units different from free-
108 living species and significantly affect food web structure and decrease network
109 **robustness** due to their complex interactions [33].

110 *Approach 2: Distinguishing between symbiont-infected and uninfected insect nodes.*
111 Infection status (whether a host is infected or symbiont-free) can be treated as a node
112 property. This approach applies when performing experimental manipulation of
113 endosymbiont infection, such as comparing infected and cured communities. Hosts with
114 different infection states can be represented as separate nodes within the same network,
115 or within different networks for infected and symbiont-free populations (Figure 1). This
116 approach could be particularly useful for identifying indirect effects and testing
117 **interaction rewiring** scenarios empirically (see following sections). For example,
118 infection by the symbiont bacterium *Wolbachia* is often heterogeneous within *Drosophila*
119 populations. *Wolbachia* can alter host–parasitoid dynamics by influencing parasitoid
120 developmental success [34]. Separating *Drosophila* nodes by infection status would
121 account for this specificity when considering interactions such as competition and
122 apparent competition. More broadly, incorporating intraspecific variation can reveal
123 hidden mechanisms: in pollination networks, for instance, accounting for variation in

124 pollen loads and deposition in communities invaded by the alien plant Himalayan
125 balsam (*Impatiens glandulifera*) helped disentangle the ecological processes underlying
126 invasion dynamics [35].

127 Producing separate ecological networks allows comparison before and after the spread
128 of a symbiont, or in different geographic locations. Comparing networks with uninfected
129 or infected nodes enables quantitative evaluation of network properties such as
130 **modularity** (the extent of sub-structuring within the network), **connectance**, and
131 robustness to species loss, thereby revealing the ecological consequences of symbiont-
132 driven host variation. Similar approaches have been applied to investigate how the
133 pathogen *Anaplasma phagocytophilum* affects the microbiota of *Ixodes scapularis* ticks
134 and their colonisation resistance, finding that *A. phagocytophilum* infection reduces
135 microbial network robustness, making ticks more vulnerable to microbial invasions [36].

136 *Approach 3: k-partite networks.* **k-partite networks** focus on direct interactions between
137 two discrete groups (bipartite) (e.g., host and symbiont or host and parasitoid) or more
138 (tri-, quadripartite, etc.), with no connections between components of the same trophic
139 level [26]. These networks are, therefore, most appropriate when the focus is on
140 interactions across trophic levels, such as when symbionts perform a protective or
141 nutritional function. For instance, this approach could analyse how environmental
142 stressors (e.g., heat waves, drought, insecticides) affect systems composed of hosts
143 classified by infection status (infected versus uninfected; see Approach 2), interacting
144 with other groups, such as basal resources, parasitoids, predators, or pathogens (Figure
145 1). In such systems, stressors can act as external modifiers of interaction strength [37].
146 For example, heat waves disrupt endosymbiont-mediated benefits in parasitoids,

147 weakening their suppression of host immune responses and increasing parasitoid
148 mortality [38]. These disruptions can lead to a decrease in key network properties—such
149 as **interaction frequency**, connectance, or modularity—particularly if parasitoids shift
150 to alternative hosts and appear less specialised [39] or are excluded from parts of the
151 network under stress. *k*-partite networks have been widely applied in plant–pollinator
152 and plant–pest–natural enemy systems to assess how environmental drivers alter
153 interaction structure and network stability [40].

154 *Approach 4: Multilayer networks.* These networks integrate different entities, locations,
155 time points and/or interaction types in a single network by distinguishing between types
156 of network edges in different layers [41]. This approach provides a flexible framework for
157 disentangling the context-dependent (e.g. multiple interactions, time, space, etc.) roles
158 of symbionts within ecological systems [42]. Inter-layer edges are key components
159 differentiating multilayer networks from multiple discrete ecological networks (Box 1).
160 These edges link ecological processes (e.g. reduction in parasitism due to host infection
161 status) to interaction networks, allowing species interactions and their drivers to be
162 analysed jointly [43].

163 Layers could represent distinct interaction types (e.g. symbiosis, parasitism,
164 competition), with interlayer edges connecting the same host across layers (Figure 1).
165 This would enable researchers to analyse (i) how facultative symbionts modify host
166 ecological roles across contexts and (ii) how disturbances (e.g., symbiont loss or
167 acquisition) propagate between layers. As an example, mushroom-feeding *Drosophila*
168 communities often share fungal breeding sites and are commonly infected by the
169 parasitic nematode *Howardula aoronymphium*, which substantially reduces female

170 fertility and increases mortality. Strong interspecific aggregation on mushrooms leads to
171 high cross-species transmission, such that high parasitism in one species can increase
172 infection in others, leading to apparent competition [44]. Within this system, *Drosophila*
173 *neotestacea* can harbour a defensive *Spiroplasma* strain that protects females from
174 nematode-induced sterility [45], altering infection dynamics at the community level. In a
175 multilayer framework, such effects could be represented through interlayer edges
176 weighted by symbiont effects on nematode infection and linking symbiosis, parasitism,
177 and competition layers (Figure 1).

178 Changing community structure can be captured as shifts in **interlayer connectivity** or
179 modularity, helping quantify the impact of symbiont-mediated indirect effects on overall
180 network structure and stability. Similar approaches have revealed cascading impacts in
181 other systems; for example, invasive ungulates disrupted pollinator-seed dispersal
182 networks by decoupling interaction layers (i.e. pollination and seed dispersal connected
183 via shared plants), reducing robustness and accelerating the spread of disturbance [46].

184 Multilayer network analyses could model systems with multiple infections within an
185 individual host. The costs of carrying multiple symbionts may combine additively, or
186 there may be negative or positive interference. For example, testing all possible
187 combinations of genotypes of the microbial symbionts *Hamiltonella defensa* and
188 *Fukatsuia symbiotica* in a single pea aphid genotype resulted in a high diversity of
189 phenotypes when exposed to different ecological challenges [47]. In this instance,
190 multilayer networks could generate a more integrative view of direct and indirect effects
191 of symbionts at different levels of the trophic network by analysing the effect of co-
192 infections with different symbiont genotypes on more than one environmental challenge

193 at a time (e.g. host plants and natural enemies). Multilayer networks can be constructed
194 with each layer representing the interactions of symbiont hosts and other partners, such
195 as natural enemies or plants, under a distinct symbiont infection state. Interlayer edges
196 reflect interaction pattern changes across infection states, connecting the same host
197 species or natural enemy across layers. For hosts, these edges can indicate changes in
198 interaction frequency (e.g. attack rate), fitness or interacting partner identity. For natural
199 enemies, interlayer edges can represent changes in attack efficiency across layers. This
200 structure allows for the application of multilayer-specific metrics—such as **versatility** or
201 **interlayer flexibility** to dissect direct and indirect ecological impacts of multiple
202 infections.

203 Multilayer networks can also capture temporal dynamics (e.g. infection rates by insect
204 symbionts varying seasonally [48]), spatial variation (e.g. symbiont infections correlating
205 with geographic areas [49]) and **horizontal transmission** of symbionts (e.g. symbionts
206 spreading between parasitoid species sharing a single host [50]).

207 *Using ecological network tools to evaluate effects of symbionts at community level*

208 Endosymbiont-mediated host phenotypic changes can lead to emerging indirect effects
209 such as **exploitative competition**, **trophic cascades**, apparent competition or
210 **apparent mutualism** [21,51]. These effects have typically been explored through
211 experimental approaches, where the presence of symbionts is manipulated to observe
212 consequences for third-party species (e.g. [20,21,52]). While effective, such studies only
213 capture a fraction of the indirect interactions present in complex natural communities.

214 To incorporate the wider community, interaction matrices trace indirect interactions [53]
215 by capturing multi-step relationships (e.g. a symbiont affecting predator populations via

216 changes in host traits). However, identifying underlying mechanisms, such as
217 behavioural or immunological changes in the host, requires integration with other
218 complementary methods that quantify the relative importance of direct and indirect
219 interactions (e.g. **structural equation modelling, path analysis** [54,55]).

220 Host–symbiont networks have community structures defined as clusters or groups of
221 nodes that are more tightly connected internally than with the rest of the network. In
222 insect–symbiont networks, detecting community patterns could highlight distinct
223 assemblages [56], such as clustering of insect species in well-documented complex
224 networks that rely on similar types of symbionts (e.g. nutritional symbionts in one cluster,
225 defensive symbionts in another). Species that bridge clusters, such as generalist hosts
226 or symbionts with multiple functions, can play a crucial role in facilitating indirect effects.
227 Changes in their interactions can propagate across clusters, influencing species in
228 distant parts of the network [57]. **Community detection methods** provide a wide variety
229 of algorithms (e.g. modularity, **spectral clustering**, or **overlapping community**
230 **detection**) to explore indirect interactions by identifying and revealing patterns of
231 ecological specialisation and interaction strength [58].

232 *Symbiont Loss, Extinction Cascades, and Network Resilience*

233 Fluctuations in infection frequency in insect-symbiont associations can profoundly
234 reshape community structure and dynamics [59]. By simulating species (node) losses,
235 stochastic co-extinction models offer critical insights into community resilience [60]. In
236 insect–symbiont networks, modelling symbiont extinction could reveal the cascading
237 effects of symbionts on hosts and their broader ecological interactions (Box 3). However,
238 coextinction risks are often mitigated through interaction rewiring [61]: higher-level

239 components consumers switch to alternative resources following the loss of their
240 original partners (e.g., [62]) (Box 3). The introduction of defensive symbionts into an insect
241 community illustrates how biological constraints determine rewiring potential. Defensive
242 endosymbionts can lead to insect hosts becoming unsuitable as resources, thus
243 disrupting interactions with higher-level consumers. Parasitoids can respond by shifting
244 host species, targeting different developmental stages or exploiting hosts on alternative
245 plants [63,64]. Incorporating rewiring into **extinction modelling** would permit exploring
246 both the direct consequences of endosymbiont loss and the conditional capacity of
247 communities to adapt through interaction reconfiguration.

248 **Concluding Remarks**

249 Understanding of the ecological processes involving symbionts from a community
250 perspective remains limited. Ecological network analyses can disentangle complex
251 processes, and freely available R-packages, including bipartite [65], igraph [66],
252 NetworkExtinction [67], EMLN [68], and econullnetr [69], now support network
253 construction, extracting metrics and modelling extinctions or resource selection. By
254 explicitly mapping symbiont presence within a network of interactions, researchers
255 could detect how symbionts influence community stability, clustering, and energy flow—
256 patterns that remain hidden when using observational or experimental techniques alone.
257 Incorporating symbionts into network models would therefore not only reveal their
258 ecological importance, but also open the door to a deeper, mechanistic understanding
259 of community functioning under both natural and disturbed conditions. Integrating
260 theoretical models with empirical data from field and laboratory studies, will advance our
261 understanding of host–microbe–environment interactions and help consider microbial

262 symbionts not as peripheral elements, but as active and dynamic components of
263 ecological networks (see Outstanding Questions).

264

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436 **Acknowledgements**

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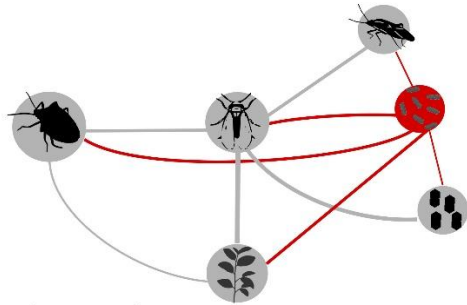
441 **Declaration of interests**

442 The authors declare no competing interests.

443

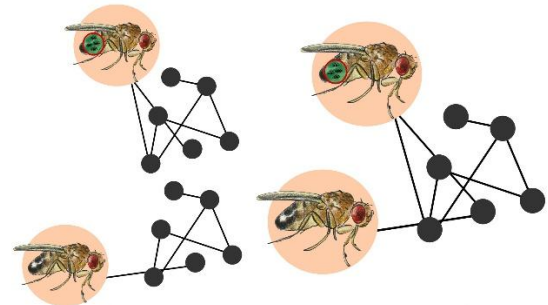
(i) Symbionts as independent nodes

Revealing overlooked interactions and emergent properties that shape community structure and function



Unipartite network

(ii) Integrating symbionts in the network



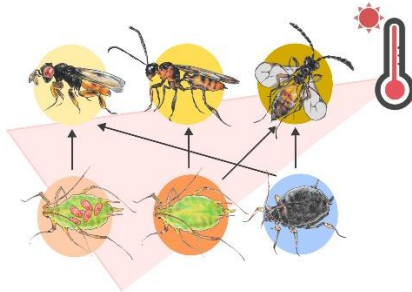
Independent networks

Same network

Representing infection states in different networks or within the same network depending on data availability and study design

(iii) Symbionts and ecosystem change

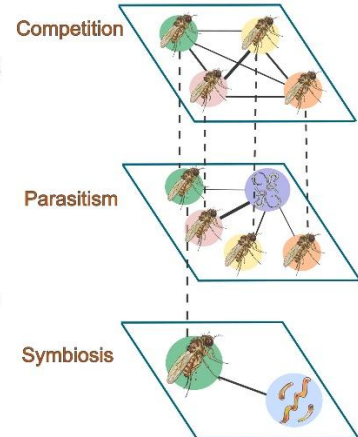
k-partite network



Uncovering how perturbations such as environmental stress shift the architecture of symbiont-insect networks

(iv) Symbionts in complex systems

Disentangling the complex and context-dependent roles of symbionts in ecological systems across time, ecological interactions and space.



Multilayer network

444

445 **Figure 1.** Integrating symbionts into ecological networks. Symbionts can be represented (i) as independent
446 nodes in unipartite networks to reveal hidden interactions; (ii) as a node property to compare infection
447 states in experimental studies; (iii) in *k*-partite networks to map host–symbiont–partner interactions,
448 across trophic interactions; (iv) in multilayer networks to capture multiple interaction types, temporal
449 changes, or infection states, highlighting context-dependent roles (e.g. the effect of spreading
450 *Spiroplasma*-infected *D. neotestacea* on *Drosophila* communities parasitised by the nematode *Howardula*
451 *aoronymphium*. Green node: *D. neotestacea*, pink node: *D. putrida*, yellow node: *D. falleni*, orange node:
452 *D. recens*). Network icons drawn by Biliana Todorova.

453

454 **Box 1. Mathematical Representation of Ecological Networks**

455 Complex communities consist of multiple species—each represented as a node—
456 connected by interactions such as parasitism, mutualism, or predation. These
457 interactions are best captured using networks, where any pair of nodes can interact
458 (unipartite), interactions have a direction (directed), and their strength can vary
459 (weighted). The fundamental mathematical representation of such an interaction
460 network is an adjacency matrix, where rows are resources, columns are consumers, and
461 entry a_{ij} reflects interactions from node j to node i —such as infection frequency or
462 parasitism rate. **Self-interactions** (e.g. cannibalism) appear on the diagonal (Figure IA)
463 [27].

464 Two species can be connected by more than one interaction type—**multi-edges** (Figure
465 IA) that can be mathematically represented by: (i) separate matrices for each interaction
466 type; (ii) multilayer network models.

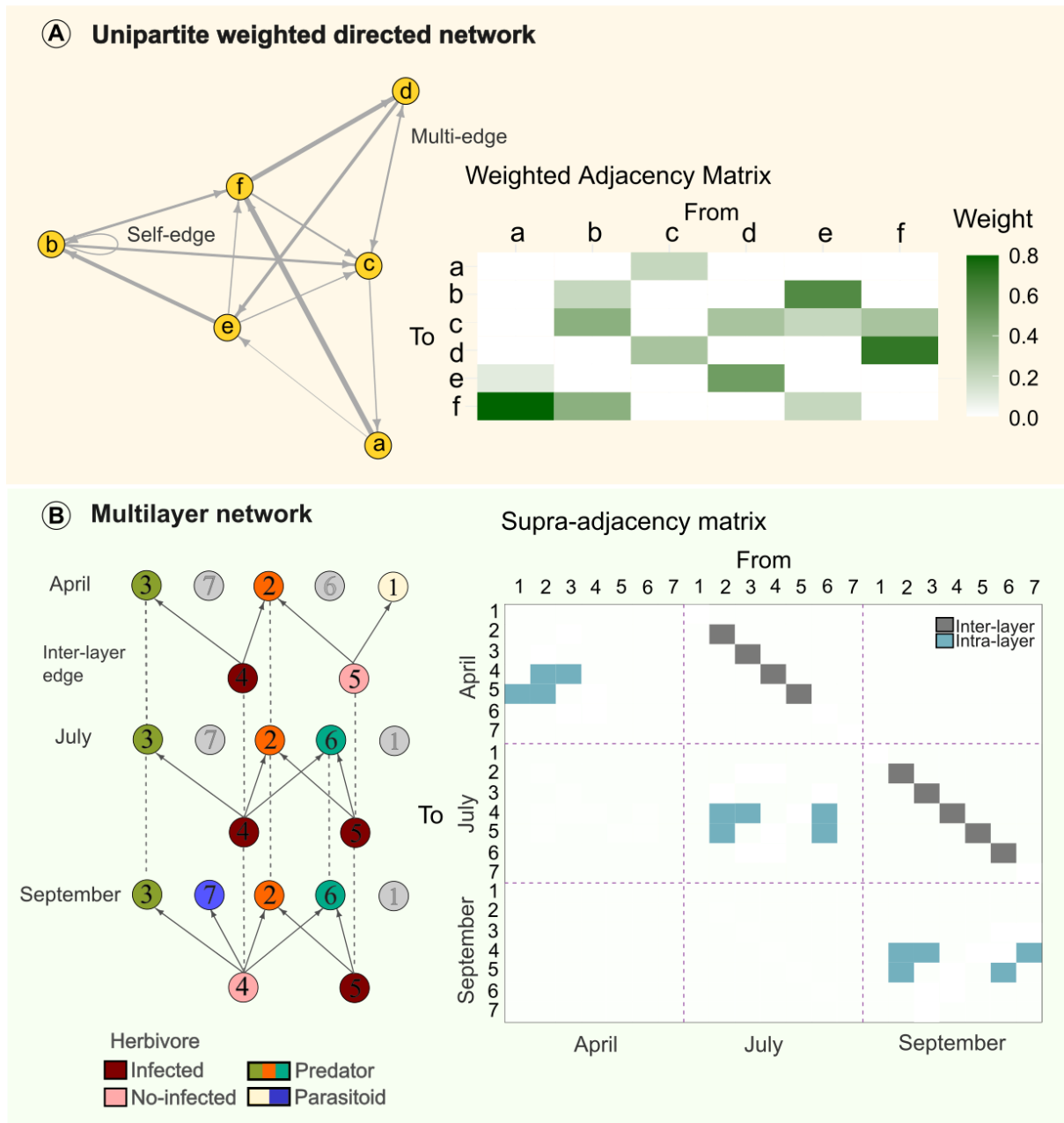
467 To examine structured subsets of complex communities, k -partite networks offer a
468 focused representation of interactions restricted to distinct trophic or functional groups.
469 Species are grouped into k distinct trophic or functional levels (e.g. plants, herbivores,
470 parasitoids), and interactions occur only between species in different levels.

471 Multilayer networks extend traditional network representations by incorporating multiple
472 interaction types or contexts (e.g., habitats, temporal dynamics) in separate matrices
473 (layers). Nodes may occur in one or several layers, with inter-layer edges linking them
474 across contexts—e.g. how symbiont-mediated protection alters parasitism (Figure IB)
475 [42]. These networks are represented by a supra-adjacency matrix integrating all layers
476 and their connections. Each layer α has its own $n \times n$ intra-layer adjacency matrix (A^α)

477 while inter-layer matrices ($B^{\alpha\beta}$), sized $n_\alpha \times n_\beta$, describe node connections across layers α
478 and β . Just as with edges in a monolayer network, inter-layer edges may take several
479 forms. Inter-layer edges with the same weight (uniform) describe the assumed
480 dependence of layers on each other. In contrast, inter-layer edges with specific weights
481 to each connection (non-uniform) represent the actual or estimated linkage between two
482 nodes and add substantial resolution to the network [43]. For example, in temporal
483 networks, non-uniform inter-layer edges, representing fluctuations in species
484 abundance, detect the species-specific responses to seasonality.

485 Once a network is constructed, its structure can be analysed using key metrics
486 describing size, interaction density, and connectivity [27]. These metrics underpin
487 models of co-extinction or rewiring, and other analyses of community robustness,
488 dynamics, and evolution.

489



490

491 Figure I. Graphical and mathematical representation of (A) a unipartite weighted directed network and
 492 host-symbiont network dynamics across seasons (April, July and October) in a multilayer network where
 493 every layer represents a bipartite network (in this case herbivore–natural enemy) at different times. The
 494 arrow thickness in (A) represents the weight of the link. Dashed lines in (B) indicate interlayer edges and
 495 nodes that are in grey-scale are absent in that layer.

496

497 **Box 2. Identifying Nodes and Links in Insect–Symbiont Ecological Networks**

498 In insect–endosymbiont networks, each species could be represented as a node. Node
499 identification begins with the taxonomic classification of insect hosts. Symbionts are
500 identified primarily through molecular tools such as metabarcoding, single or multiplex
501 PCR, and qPCR. Whether symbionts should be treated as independent nodes or
502 attributes of hosts depends on the ecological question and system (see main text). The
503 choice of symbiont identification technique—quantitative or qualitative—also depends
504 on the study context (e.g., natural communities, experimental systems, or manipulated
505 populations).

506 Direct interactions between hosts and symbionts can be revealed through molecular
507 detection, while broader ecological interactions—trophic and non-trophic—can be
508 inferred from empirical observations, literature, or expert knowledge [27]. Methods
509 include observational data (e.g. parasitism rates), stable isotope analysis, gut content
510 identification, and controlled experiments (e.g. feeding or parasitism trials). High-
511 resolution molecular tools such as gut content PCR and metabarcoding can further
512 enhance the identification of within-community interactions [70]. However, not all
513 interactions are direct. Indirect effects—often mediated by symbionts—require
514 specialised approaches to infer (see main text).

515 Understanding the strength (weight) and direction of interactions is essential for
516 ecological interpretation. In host–symbiont networks, infection frequency—quantified
517 via PCR—can serve as a proxy for interaction strength. For symbiont-mediated defensive
518 interactions, weights may be based on parasitism rates or the proportion of prey DNA in
519 predator diets.

520 Symbiont-mediated effects are highly context-dependent. Their phenotypic outcomes
521 vary based on host genotype, symbiont strain and identity, environmental conditions [71],
522 and co-infection dynamics [72]. For instance, in pea aphids, combining a set of different
523 isolates of *Hamiltonella defensa* and *Fukatsuia symbiotica* in coinfections resulted in
524 diverse symbiont-mediated phenotypes when exposed to different ecological challenges
525 [47]. Therefore, symbiont detection alone is insufficient to assume ecological function.
526 Currently, establishing functional links often requires experimental exposure of hosts
527 with known symbionts to specific challenges. Nevertheless, advances in symbiont
528 genomics are providing promising insights that may in future aid in predicting symbiont-
529 mediated host phenotypes from symbiont genotypes, such as the identification of toxin
530 cassettes on mobile genetic elements. For example, different phage **(APSE)** toxins in
531 *H. defensa* correlate with different levels of protection against parasitism [73].

532

533 **Box 3. Modelling Symbiont-Driven Network Dynamics**

534 We used data from a field experiment by Rothacher et al. [74], who tested whether the
535 facultative endosymbiont *Hamiltonella defensa* reduces parasitism in the black bean
536 aphid (*Aphis fabae*). The authors provide a full, species-level record of aphid–primary
537 parasitoid–secondary parasitoid interactions.

538 To illustrate network analysis in symbiont–insect systems, we constructed a simplified
539 weighted tri-trophic network using the most abundant five primary parasitoids and three
540 secondary parasitoids. Interaction strengths were derived from emergence counts and
541 normalised to represent proportional dependencies among nodes (Fig II A).

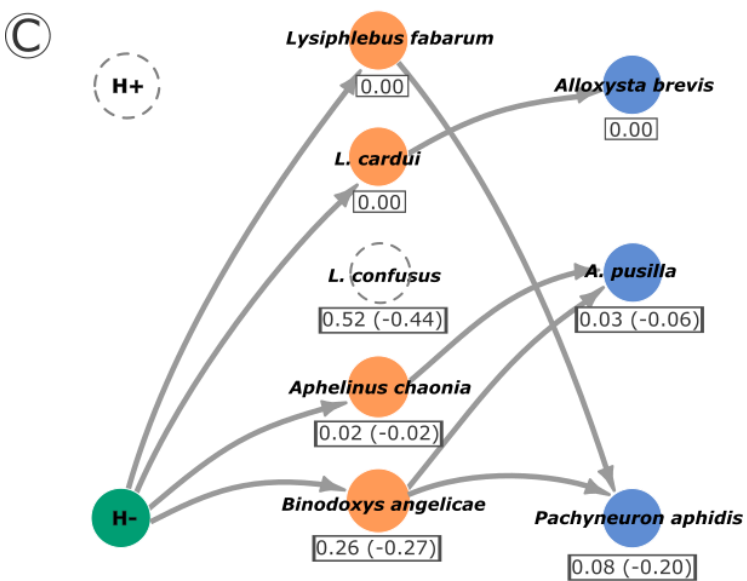
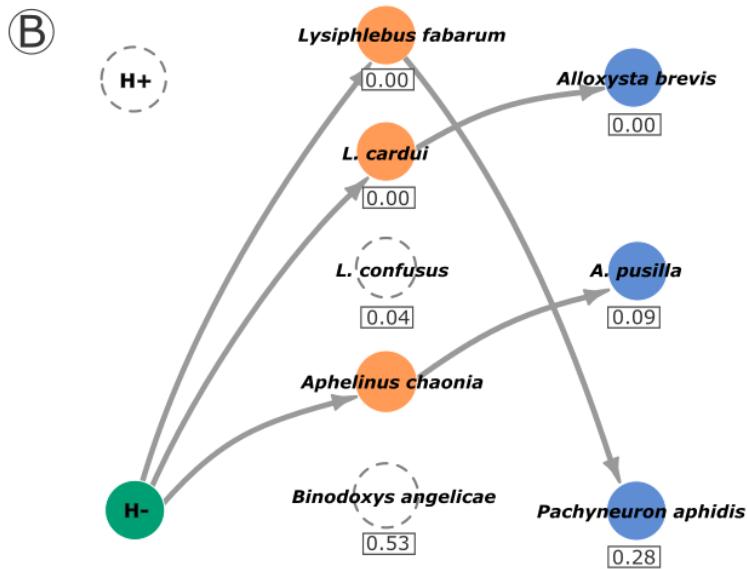
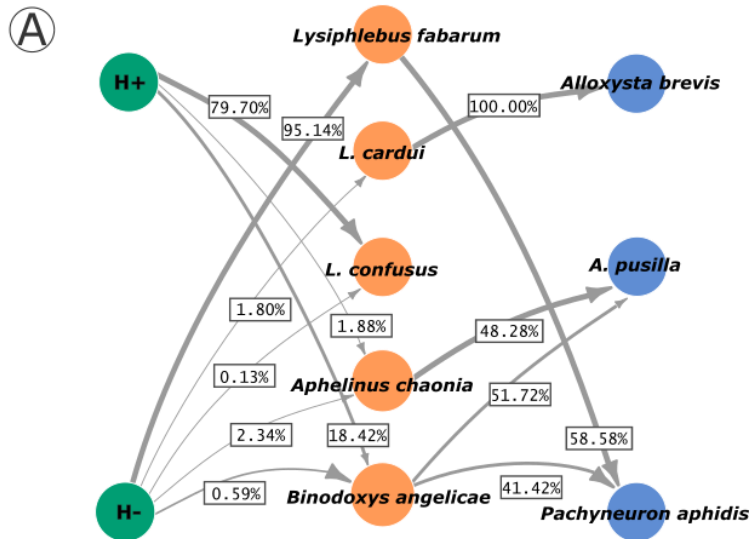
542 We applied a probabilistic co-extinction model [75] to explore cascading consequences
543 of symbiont-mediated host loss. We removed the symbiont-infected H+ aphid
544 phenotype, and, at each step, the algorithm propagates secondary extinctions based on
545 species' proportional dependence on lost resources (Fig. II B). This framework moves
546 beyond deterministic thresholds by incorporating interaction-strength-dependent
547 extinction probabilities (Fig. II B), such that specialists face higher extinction risks than
548 generalists following resource loss.

549 To increase ecological realism, we extended the model to include rewiring capacity,
550 allowing species to shift resource use after perturbation. Rewiring probabilities integrate:
551 (i) prior use of alternative hosts, (ii) degree of generalism, and (iii) network position, with
552 highly connected species exhibiting greater adaptive potential (Fig. II C). The model
553 tracks cumulative extinctions across cascade stages while recording successful rewiring
554 events, enabling comparisons between rewiring and no-rewiring scenarios. This

555 highlights how behavioural plasticity can buffer symbiont-induced perturbations and
556 influence overall network robustness.

557 The removal of H⁺ aphids generated substantial secondary losses in the network, with
558 extinction risks closely tracking species' proportional dependence on the H⁺ phenotype.
559 Under the probabilistic co-extinction model, two primary parasitoids were lost: *L.*
560 *confusus*, which relied almost entirely on H⁺ aphids, and *B. angelicae*, which, despite
561 also attacking H⁻ aphids, was strongly affected by the disappearance of its preferred
562 host. Secondary parasitoids, however, largely persisted because many primary hosts
563 remained available. Incorporating rewiring reduced these losses: specialists were able to
564 shift to alternative hosts, allowing *B. angelicae* to survive via increased use of H⁻ aphids.
565 Overall, rewiring increased final community richness and boosted system robustness,
566 illustrating how adaptive flexibility can buffer symbiont-driven perturbations.

567 This framework is intended as a conceptual demonstration. Applying it to real systems
568 would require empirical data on probabilistic dependencies, temporal changes in
569 resource availability, and context-dependent attack rates to parameterise extinction and
570 rewiring processes accurately.



■ Aphids ■ Primary parasitoids
■ Secondary parasitoids Extinct

572 Figure II: Symbiont-driven tri-trophic “toy” network dynamics based on Rothacher et al. [74]. (A)
573 Weighted aphid–primary parasitoid–secondary parasitoid network with proportional
574 dependencies shown in squares. (B) Network after removal of H⁺ aphids under a probabilistic co-
575 extinction model; squares indicate extinction probabilities. (C) Network including rewiring;
576 squares show extinction probabilities with changes relative to no rewiring in parentheses.

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585 **Outstanding questions**

586 **1. What role do symbionts play in shaping the structure and dynamics of**
587 **ecological networks?**

588 Despite their well-known effects on host traits and their direct interactions, symbionts
589 are never included as explicit nodes in ecological network studies. As a result, their
590 potential to modulate structure, stability, and resilience at the community level remains
591 largely unexplored. Integrating symbionts into complex systems research could reveal
592 hidden interaction pathways, uncover overlooked drivers of network dynamics, and
593 fundamentally change our understanding of how ecological communities function.

594 **2. How do symbiont–host coevolutionary dynamics shape long-term patterns of**
595 **specialisation, diversification, and interaction reallocation in ecological networks?**

596 Symbionts can influence host range, defence traits, and resource use, potentially
597 accelerating or constraining evolutionary trajectories. Incorporating symbionts into
598 network analyses could help reveal how coevolution—between hosts, symbionts, and
599 their natural enemies—drives patterns of modularity, nestedness, and reallocation
600 through time, improving predictions of community reorganisation under changing
601 ecological and evolutionary pressures.

602 **3. Do symbiont-mediated phenotypic shifts driven by global change scale up**
603 **from individual hosts to alter food-web structure and stability under climatic stress?**

604 Facultative symbionts can modify key host traits—from thermal tolerance to defence
605 against natural enemies and resource use—that often vary with environmental
606 conditions. As warming, altered seasonality, and landscape change reshape ecological

607 contexts, shifts in symbiont composition, density, and genotype may influence host
608 performance and population dynamics in complex, context-dependent ways. Whether
609 these trait-level responses propagate to reshape species interactions is crucial for
610 predicting community resilience under global change.

611 **4. Can symbiont-mediated traits in invasive insect hosts reshape native**
612 **ecological networks?**

613 Invasive insects may carry symbionts that confer novel traits—such as resistance to
614 enemies, thermal tolerance, or altered behaviour—which can disrupt native species
615 interactions. These shifts may trigger indirect effects on uninfected native species,
616 ultimately altering the structure and dynamics of local ecological networks.

617

618 **Glossary**

619 **Apparent competition:** Indirect negative interaction between species mediated
620 through shared natural enemies

621 **Apparent mutualism:** Indirect positive interaction between species mediated through
622 reduction in shared natural enemies

623 **APSE:** A phage known from the insect bacterial symbiont *Hamiltonella defensa*
624 (*Acyrtosiphon pisum* Secondary Endosymbiont phage)

625 **Community detection methods:** Procedures for identifying groups of highly connected
626 nodes in a network based upon network structural properties

627 **Connectance:** Measure of community complexity based on the proportion of realised
628 interactions from the pool of all possible interactions between the species of a network

629 **Cytoplasmic incompatibility:** Post-mating incompatibility causing F1 inviability when
630 females are infected with different symbiont strains to males (or males are uninfected)

631 **Exploitative competition:** Species compete for the same limited resource

632 **Extinction modelling:** Use of network science to evaluate how the extinction of one
633 species can trigger the extinction of other species

634 **Feminisation:** Genotypic males develop as functional phenotypic females

635 **Horizontal transmission:** Transfer of genetic material within or between host species
636 that does not occur via inheritance from parents to offspring

637 **Interaction frequency:** Rate at which different species interact with each other

638 **Interaction rewiring:** Process by which higher-level components of the food web re-
639 allocate to alternative resources following the loss of their original partners

640 **Interlayer connectivity:** Structural property of multilayer networks, defined by links
641 connecting species across layers, used to quantify how processes in one layer relate to
642 others

643 **Interlayer flexibility:** The ability of a node to be active across different layers of the
644 multilayer network

645 **k -partite network:** A network composed of a defined number, k , of non-overlapping
646 node sets, with edges allowed only between (and not within) those sets.

647 **Male killing:** Symbiont infection of a female results in the production of female-biased
648 broods because male progeny dies post-fertilization

649 **Modularity:** Metric quantifying how strongly a network is divided into groups (modules),
650 comparing interactions within modules to those between modules.

651 **Multi-edges:** The same two nodes are linked by more than one type of edge

652 **Overlapping community detection:** Set of algorithms to identify the multiple
653 community memberships of a node

654 **Parthenogenesis:** Form of asexual reproduction where an embryo develops from an
655 unfertilized egg

656 **Path analysis:** Statistical method to evaluate the direct and indirect effects of variables
657 on a specified outcome and the relative strength of each causal pathway

658 **Robustness:** Tolerance of the network to species extinctions, quantified as the fraction
659 of species that had to be removed to result in a total loss of $\geq 50\%$ of the species

660 **Self-interaction:** A node is linked to itself; for example, autophagy or cannibalism

661 **Spectral clustering:** Set of algorithms that divide a graph into clusters using the
662 eigenvectors of the input data matrix

663 **Structural equation modelling:** Multivariate statistical framework used to model
664 complex relationships between directly and indirectly observed variables

665 **Trophic cascade:** Indirect species interactions that originate with the removal/addition
666 of components in one trophic level and propagate down a food web (sometimes also
667 used for similar upward consequences)

668 **Versatility:** Node role in the cohesion of the multilayer network structure, building
669 bridges between different types of relationships

670