

Should we build a macro-scale theory for coevolution? Nope.

Timothée Poisot Daniel B. Stouffer

Oct. 31, 2014

1 Coevolutionary dynamics acting on both species and their interactions
2 are a key driving force behind the structure of ecological communi-
3 ties. The Geographic Mosaic Theory of Coevolution (GMTC) provides
4 a spatial perspective to these dynamics by proposing the existence of
5 feedbacks between local and regional scales. It remains unclear, how-
6 ever, how the structure of communities at larger spatial scales either in-
7 fluences or is influenced by local coevolutionary processes. Despite this,
8 ecological networks are known to have an ecological structure, which
9 suggest that coevolution may play a role in their evolutionary dynamics.
10 Here we show that the coevolutionary significance of individual interac-
11 tions is maintained when moving between the local and regional scale.
12 Importantly, this occurs despite the fact that community variation at
13 the local scale tends to weaken or remove community-wide coevolution-
14 ary signal. This apparent mismatch between our interaction-level and
15 community-level results strongly suggests that interactions provide the
16 correct scale to study coevolution at small spatial scales while commu-
17 nities are the relevant scale only at larger spatial extents. We provide

1 a new perspective on the interplay between coevolutionary theory and
2 community ecology, by establishing the organisational scales at which
3 the different theories have relevance. Although it has been tempting
4 so far to understand how coevolution relates to network structure, our
5 results suggest that the way forward is to understand how network
6 structure may affect coevolution over space instead.

7 Ecological interactions introduce selective pressures on the species involved. At
8 large organisational scales, this results in taxonomic boundaries delineating groups
9 of interacting species ¹, invariant structures in some ecological communities ², and
10 conservatism of both the distribution of community modules ⁴ and the role of
11 species occupy within them ⁵. Although the evolutionary dynamics for a pair of
12 interacting species has been well described ^{???}, attempts to understand how these
13 mechanisms cascade up to generate species diversity observed in large ecological
14 networks have been unconvincing ^{6,7}, despite the well known effect of antagonistic
15 coevolution on genomic diversification ^{8,9}. The scales at which these diversities
16 happen are hard to reconcile: coevolution is expressed within patches connected
17 by gene-flow ¹⁰⁻¹², whereas the species diversity of complex networks is typically
18 observed at spatial scales matching the species distribution ^{13,14}. Because these
19 scales differ by orders of magnitude, one must question the relevance of previous
20 calls to scale the theory on coevolution up to multi-species systems covering large
21 spatial extents ^{???}.

22 Network-based approaches ^{15,16}, on the other hand, were designed to study and
23 describe species-rich systems. Previous empirical findings revealed the impact
24 of evolutionary dynamics on overall network structure in food webs ^{1,17}, pollina-

tion networks ^{18,19}, and host-parasite networks ^{20,21}. Both micro ^{3,22} and macro ^{23,24} evolutionary dynamics have been measured or modelled in species-rich communities, but there is no understanding of how, or even of whether, local/micro-evolutionary and regional/macro-evolutionary feedback into one another. Another layer of complexity is that ecological networks are known to vary in their structure over time and space ²⁵: the same two species will not interact in a consistent way locally, either because of local environmental contingencies, by chance, or because their phenologies do not positively covary in space ²⁶. As a consequence, *locally*, the evolutionary signal on network structure is expected to be buried under much ecological noise, and the effect of coevolution can only be inferred *regionally*.

In multi-species systems that typically span a large taxonomic range, coevolution is often measured as the matching between the phylogenies of two sets of interacting organisms ^{20,27}. This build on the century-old ideas that extant species interact in a way similar to the way their ancestors did ²⁸. “Coevolved” systems should (i) have approximately similar phylogenetic trees and (ii) species at matching positions in either trees should interact. It is not clear, however, how this idea relates to dynamics occurring at smaller scales ²⁹: many ecological and evolutionary processes that occur locally, or over small spatial scales, can disturb this expected structure. Notably, it has been shown that species interactions are not consistent through space ^{25,30}. Local loss of both interactions and species from the regional pool is most likely to result in observed communities that do not appear to have been shaped by coevolution.

We use data on ectoparasites of rodents from Western to Eastern Europe ³¹ to test the following four hypotheses. First, local species assemblages do not

1 show evidence of coevolution even though the system has a whole does. Second,
2 interaction-level coevolutionary signal is conserved. Third, interaction-level co-
3 evolutionary signal does not predict the spatial consistency of interactions. We
4 do so by coupling two novel methods: the *PACO* algorithm for detection of phy-
5 logenetic congruence ³², and a general framework for the variation of ecological
6 networks ³⁰.

7 Local observations on the 51 localities (*Supp. Mat. 1*) are aggregated into a
8 regional metanetwork ³⁰. The two phylogenetic trees have been rendered ultra-
9 metric. We use *PACO* ³² to measure the congruence between trees knowing the
10 matrix of interaction. *PACO* yields a network-level significance value for the likeli-
11 hood that hosts and parasites have coevolved. For each local network, we measure
12 the strength of coevolution using (i) only local observations and (ii) all possible
13 interactions between local species (as known from the regional aggregation of all
14 local networks). This allows us to separate the effect of species sorting (regional
15 interactions) and interaction sorting (local interactions). At the regional scale,
16 coevolutionary signal is extremely strong ($p \leq 10^{-4}$), as established by previous
17 analysis of this system ³³. Most local networks, on the other hand, show very little
18 evidence of phylogenetic congruence. Out of 51 local networks, 35 show no signal
19 of coevolution, 11 show coevolution when using the regional interactions, and 12
20 show coevolution using the local interactions (see *Supp. Mat. 1* for network-level
21 significance values). **Figure?**

22 This suggests that macro-evolutionary processes (such as co-diversification) have
23 consequences at the macro-ecological level ³⁴, but may not be detected at finer
24 spatial scales due to a stronger effect of ecological processes locally. *PACO* per-

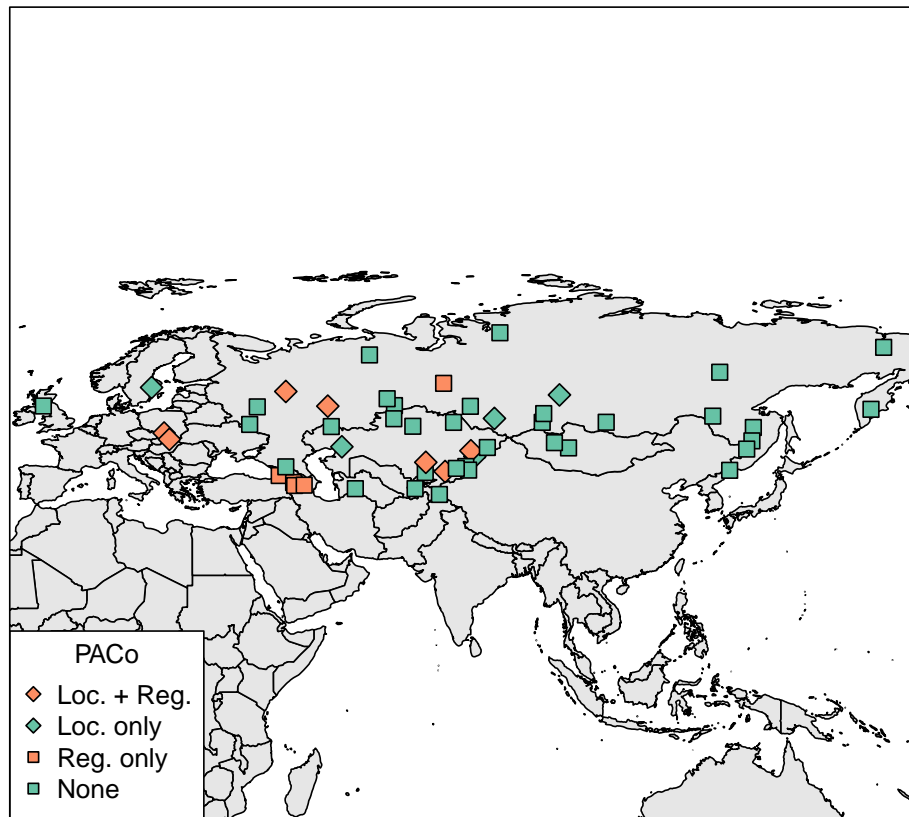


Figure 1: figure1

1 mits the analysis of *how strongly* each interaction contribute to coevolution, in a
 2 way that is as independent as possible from other interactions. As interactions
 3 vary only insofar that there are some locations in which they *do not* happen, we
 4 expect that the overall contribution of interactions will be the same in the local
 5 and regional networks. For the 5 networks that show evidences of coevolution
 6 accounting both for species and interactions sorting, we measured the contribu-
 7 tion of each interaction locally, and compared it to its contribution to the regional
 8 network. Results are presented in **FIG. one-sentence summary**. This is a key
 9 result, as it establishes that although coevolution does not leave an imprint on
 10 local networks, it is still detectable in *interactions*. This is in line with recent re-
 11 sults that established that, although networks are composed of interactions, both
 12 objects seem to have uncoupled behaviors ³⁵.

13 **Point 2 – Interactions respond locally to micro-** 14 **processes**

- 15 • Review of mechanisms involved in interactions locally: neutrality and trait-
 16 based mechanisms
- 17 • Key refs: Oikos paper, Canard, Olesen, Combes
- 18 • These mechanism can only *filter* from a pool of regional interactions
- 19 • Key refs: Biol Lett paper, Olito & Fox, Lounnas, ...
- 20 • Key points: i) even though they will change the network-level signal, con-
 21 tribution of interactions should not vary locally vs. regionally ii) this is the

- 1 expectation because (a) past knowledge about motifs/evolution and (b) in-
- 2 teractions are the unit in which selection takes place, not networks
- 3 • Illustration: distribution of the significance in each interaction locally (mea-
- 4 sured vs. metaweb) vs. in the regional network.

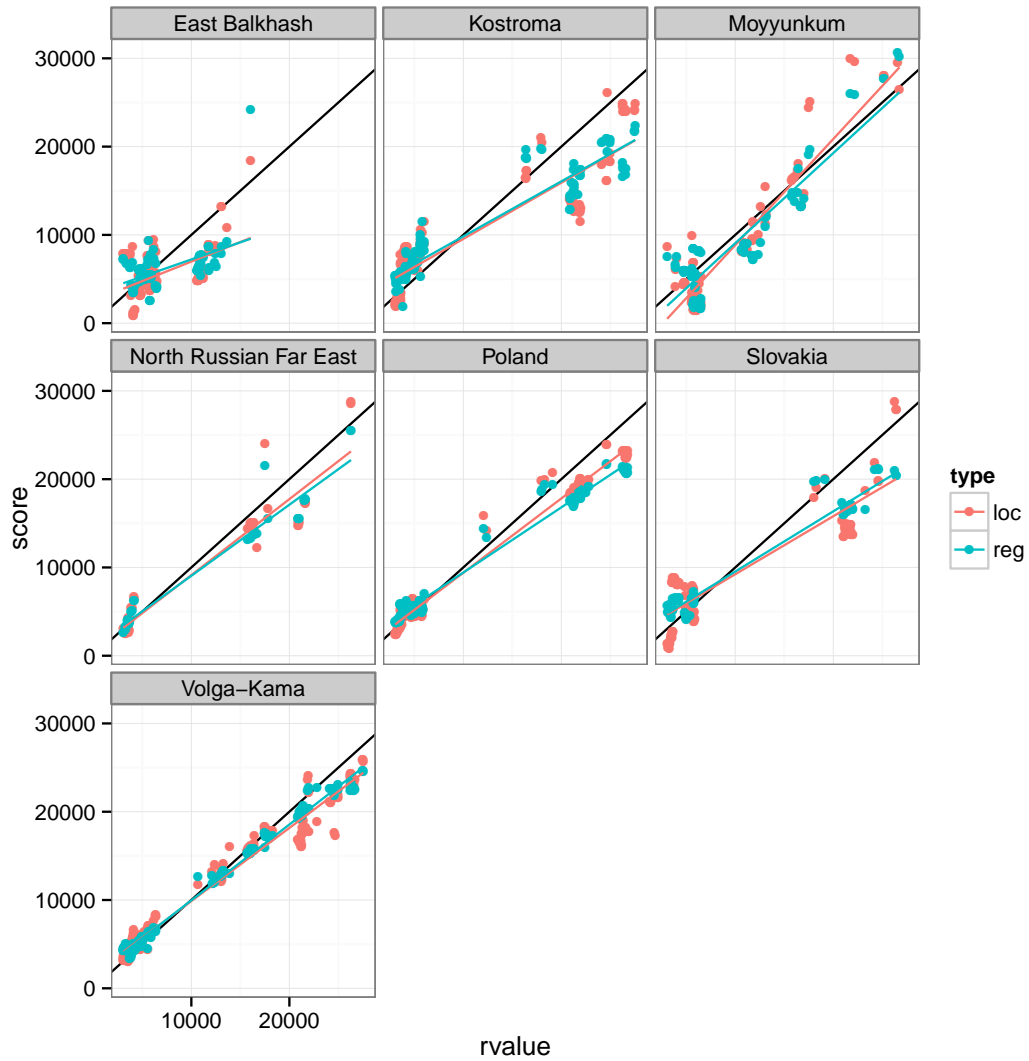


Figure 2: figure2

1 Point 3 – The spatial replication of networks is im- 2 perfect

- 3 • Ecological processes = sorting of interactions + species
- 4 • Key refs: Ecol Lett -div
- 5 • Key points: observation of interactions result from filtering through two
6 processes, and not clear they are related to past coevolution = we expect a
7 loss of signal locally
- 8 • Illustration: jackknife score vs. observed/possible : there is no relationship
9 between how important an interaction is (coevo) and how frequent it will be.
10 suggests that ecological >> evolutionary drivers of local network structure
- 11 • This is caused by local ecological processes, but also influences local evolu-
12 tionary dynamics
- 13 • Key refs: Gomulkiewicz, Hochberg, Benkman, Siepilski
- 14 • Key points: i) Opportunity to integrate more network theory to understand
15 between-sites differences ii) Even though coevo theory recognized variation
16 in effects, very little attention to variation in presence/absence of int and
17 its implications

18 Conclusions

- 19 • Need to bridge ecological networks with co-evolution (understand the impact
20 of ecology vs. evolution in the local observations)

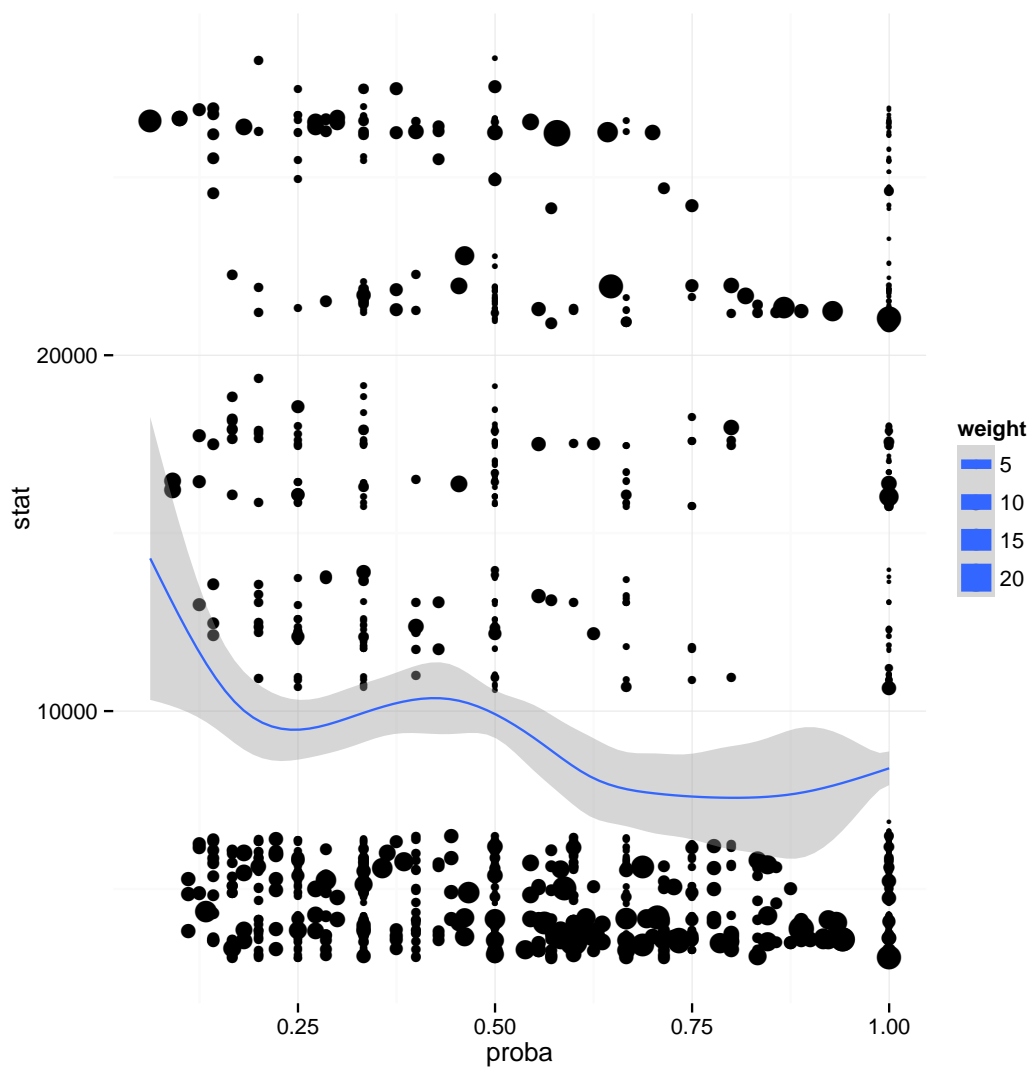


Figure 3: figure3

- 1 • Main advance: coevolutionary dynamics are important at the continen-
2 tal/regional scale, much less at the local scale
 - 3 • Focus on datasets that are spatially replicated, with as good as possible
4 phylogenies. Ref Buckley, Hearn.
 - 5 • Currently local variability is not accounted for by statistical methods, need
6 to do a better job. See Desdevices chapter on ParaFit 2.0.
- 7 1.Eklof, A., Helmus, M. R., Moore, M., Allesina, S. & Eklöf, A.*et al.* Relevance
8 of evolutionary history for food web structure. *Proc. R. Soc. B Biol. Sci.* **279**,
9 1588–1596 (2011).
 - 10 2.Nuismer, S. L., Jordano, P. & Bascompte, J.*et al.* Coevolution and the archi-
11 tecture of mutualistic networks. *Evolution* **67**, 338–354 (2013).
 - 12 3.Jordano, P., Bascompte, J. & Olesen, J. M.*et al.* Invariant properties in coevo-
13 lutionary networks of plant-animal interactions. *Ecol. Lett.* **6**, 69–81 (2003).
 - 14 4.Stouffer, D. B., Camacho, J., Guimera, R., Ng, C. A. & Amaral, L. A. N.*et al.*
15 Quantitative patterns in the structure of model and empirical food webs. *Ecology*
16 **86**, 1301–1311 (2005).
 - 17 5.Stouffer, D. B., Sales-Pardo, M., Sirer, M. I. & Bascompte, J.*et al.* Evolutionary
18 Conservation of Species’ Roles in Food Webs. *Science* **335**, 1489–1492 (2012).
 - 19 6.Hembry, D. H., Yoder, J. B. & Goodman, K. R.*et al.* Coevolution and the
20 Diversification of Life. *The American Naturalist* **184**, 425–438 (2014).
 - 21 7.Yoder, J. B. & Nuismer, S. L. *et al.* When does coevolution promote diversifi-
22 cation? *Am. Nat.* **176**, 802–817 (2010).

- 1 8. Paterson, S. *et al.* Antagonistic coevolution accelerates molecular evolution.
2 *Nature* **464**, 275–278 (2010).
- 3 9. Buckling, A., Wei, Y., Massey, R. C., Brockhurst, M. A. & Hochberg, M. E. *et*
4 *al.* Antagonistic coevolution with parasites increases the cost of host deleterious
5 mutations. *Proc. R. Soc. B* **273**, 45–49 (2006).
- 6 10. Brockhurst, M. A., Buckling, A., Poullain, V. & Hochberg, M. E. *et al.* The
7 impact of migration from parasite-free patches on antagonistic host-parasite co-
8 evolution. *Evolution* **61**, 1238–1243 (2007).
- 9 11. Nuismer, S. L., Thompson, J. N. & Gomulkiewicz, R. *et al.* Coevolution be-
10 tween hosts and parasites with partially overlapping geographic ranges. *J. Evol.*
11 *Biol.* **16**, 1337–1345 (2003).
- 12 12. Benkman, C. W., Parchman, T. L., Favis, A. & Siepielski, A. M. *et al.* Recip-
13 rocal selection causes a coevolutionary arms race between crossbills and lodgepole
14 pine. *Am. Nat.* **162**, 182–194 (2003).
- 15 13. Havens, K. *et al.* Scale and structure in natural food webs. *Science* **257**,
16 1107–1109 (1992).
- 17 14. Bascompte, J. *et al.* Disentangling the Web of Life. *Science* **325**, 416–419
18 (2009).
- 19 15. Proulx, S., Promislow, D. & Phillips, P. *et al.* Network thinking in ecology and
20 evolution. *Trends Ecol. Evol.* **20**, 345–353 (2005).
- 21 16. Dunne, J. A. *et al.* in *Ecological networks: Linking structure and dynamics*
22 (eds. Dunne, J. A. & Pascual, M.) 27–86 (Oxford University Press, 2006).
- 23 17. Stouffer, D. B., Camacho, J., Jiang, W. & Amaral, L. A. N. *et al.* Evidence for

- 1 the existence of a robust pattern of prey selection in food webs. *Proc. R. Soc. B*
2 *Biol. Sci.* **274**, 1931–40 (2007).
- 3 18. Chamberlain, S. A. *et al.* Traits and phylogenetic history contribute to network
4 structure across Canadian plant–pollinator communities. *Oecologia* 1–12 (2014).
- 5 19. Vamosi, J. C., Armbruster, W. S. & Renner, S. S. *et al.* Evolutionary ecology
6 of specialization: insights from phylogenetic analysis. *Proc. R. Soc. B* **281**,
7 20142004 (2014).
- 8 20. Desdevises, Y., Morand, S., Jousson, O. & Legendre, P. *et al.* Coevolution
9 between *Lamellodiscus* (Monogenea: Diplectanidae) and Sparidae (Teleostei): the
10 study of a complex host–parasite system. *Evolution* **56**, 2459–2471 (2002).
- 11 21. Hafner, M. S. & Page, R. D. M. *et al.* Molecular phylogenies and host–parasite
12 cospeciation: gophers and lice as a model system. *Philos. Trans. R. Soc. B Biol.*
13 *Sci.* **349**, 77–83 (1995).
- 14 22. Hall, A. R., Scanlan, P. D. & Buckling, A. *et al.* Bacteria–Phage coevolution
15 and the emergence of generalist pathogens. *Am. Nat.* **177**, 44–53 (2010).
- 16 23. Roopnarine, P. D. & Angielczyk, K. D. *et al.* The evolutionary palaeoecology
17 of species and the tragedy of the commons. *Biol. Lett.* **8**, 147–50 (2012).
- 18 24. Dunne, J. a, Williams, R. J., Martinez, N. D., Wood, R. a & Erwin, D. H. *et*
19 *al.* Compilation and network analyses of cambrian food webs. *PLoS Biol.* **6**, e102
20 (2008).
- 21 25. Poisot, T., Stouffer, D. B. & Gravel, D. *et al.* Beyond species: why ecological
22 interaction networks vary through space and time. *Oikos* (2014).
- 23 26. Devictor, V. *et al.* Differences in the climatic debts of birds and butterflies at

1 a continental scale. *Nat. Clim. Change* **2**, 121–124 (2012).

2 27.Legendre, P., Desdevises, Y. & Bazin, E.*et al.* A statistical test for host-
3 parasite coevolution. *Syst. Biol.* **51**, 217–234 (2002).

4 28.Fahrenholz, H. *et al.* Ectoparasiten und abstammungslehre. *Zool. Anz.* **41**,
5 371–374 (1913).

6 29.Poisot, T. *et al.* in *Evolutionary Ecology of Host-Parasite Systems* (eds. Morand,
7 S., Littlewood, D. T. & Poulin, R.) (Cambridge University Press, 2015).

8 30.Poisot, T., Canard, E., Mouillot, D., Mouquet, N. & Gravel, D.*et al.* The
9 dissimilarity of species interaction networks. *Ecol. Lett.* **15**, 1353–1361 (2012).

10 31.Data from: Phylogenetic signal in module composition and species connectivity
11 in compartmentalized host-parasite networks. (2012).

12 32.Balbuena, J. A., Míguez-Lozano, R. & Blasco-Costa, I.*et al.* PACo: A Novel
13 Procrustes Application to Cophylogenetic Analysis. *PLoS ONE* **8**, e61048 (2013).

14 33.Krasnov, B. R. *et al.* Phylogenetic signal in module composition and species
15 connectivity in compartmentalized host-parasite networks. *Am. Nat.* **179**, 501–
16 11 (2012).

17 34.Price, P. W. *et al.* *Macroevolutionary Theory on Macroecological Patterns*.
18 (Cambridge University Press, 2003).

19 35.Olito, C. & Fox, J. W. *et al.* Species traits and abundances predict metrics of
20 plant–pollinator network structure, but not pairwise interactions. *Oikos* n/a–n/a
21 (2014).