

“Disentangling nestedness” disentangled

ARISING FROM A. James, J. W. Pitchford & M. J. Plank *Nature* **487**, 227–230 (2012)

Analytical research indicates that the ‘nestedness’ of mutualistic networks facilitates the coexistence of species by minimizing the costs of competition relative to the benefits of facilitation¹. In contrast, James *et al.*² recently argued that a more parsimonious explanation exists: the persistence of a community and its constituent species depends more on their having many interactions (high connectance and high degree, respectively) than for these interactions to be organized in any particular manner. Here we demonstrate that these conclusions are an unintended consequence of the fact that the methodology of ref. 2 directly changed the number of interactions of each species—and hence their expected persistence. When these changes are taken into account, we find a significant, positive relationship between nestedness and network persistence that reconfirms the importance of nestedness in mutualistic communities^{1,3}. There is a Reply to this Brief Communication Arising by James, A., Pitchford, J. W. & Plank, M. J. *Nature* **500**, <http://dx.doi.org/10.1038/nature12381> (2013).

Given a network, one can robustly quantify the relative numbers of specialist to generalist species via the degree distribution^{4,5}. A network’s degree distribution is of considerable importance, because studies have repeatedly highlighted the significant, positive relationship between a species’ number of mutualistic partners and its survival probability^{1–3,6}. This distribution alone is also capable of driving many higher-order network properties⁷, not to mention the fact that the degrees of species are phylogenetically constrained themselves⁸. For these and other reasons, studies across the ecological-network literature^{4,5,7} have emphasized the need to take the degree distribution into consideration when assessing the significance of the myriad patterns observed in nature^{9–11}.

Unfortunately, when comparing empirically observed networks to random networks, the authors of ref. 2 seem to have overlooked this critical link between changes in the degree distribution and species’ survival. As a direct consequence, the specialists in their random networks became less specialist and the generalists less generalist⁵. Yes, the random networks were observed to be more persistent (Fig. 1a), but this was not in fact an indication that nestedness is unimportant². Instead, this increase in persistence was a result of the random networks having more homogeneous degree distributions^{5,12}, and that the most vulnerable species in the empirical networks almost always had more interactions in the corresponding randomizations. Here this distinction is of critical importance because species’ degrees are, in fact, “a better predictor of individual species survival”². “The more the merrier” indeed¹³.

To quantitatively validate these results, we repeated a key analysis of ref. 2 to measure the relationship between nestedness and persistence while paying explicit attention to changes in the network’s degree distribution (Methods). On taking the small but critical step of controlling for the increased homogeneity of the degree distributions, we observe a significant, positive relationship between nestedness and persistence (Fig. 1b). In addition, we reach the same conclusion whether we account for changes in the degree distribution statistically or by repeating the analysis while generating the randomized networks with a null model that explicitly maintains the observed degree distribution (Fig. 1c, Methods and Appendix). All else being equal, our results here illustrate that, the greater the nestedness of a community, the greater indeed is that community’s persistence.

Given an observed number of species and interactions in a community, a prevailing question across the ecological literature is whether or not some way to structure those interactions (for example, nestedness) lead to more persistent communities. Although the number of

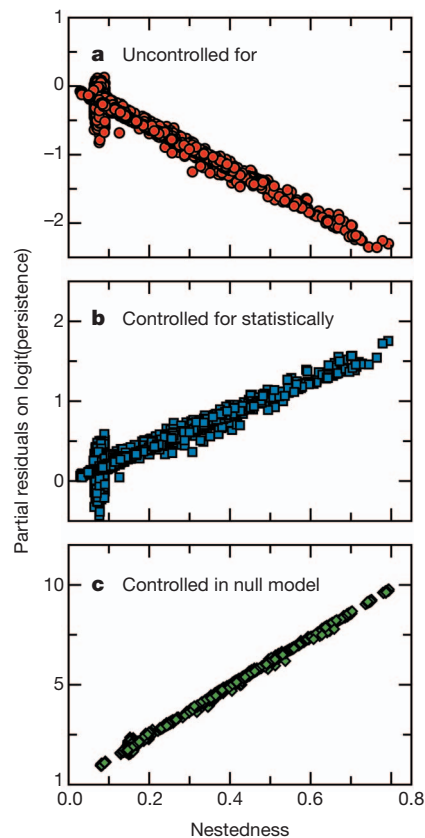


Figure 1 | Within our regression analysis, the relationship between nestedness and persistence in mutualistic networks depends integrally on changes in the degree distributions of the networks. **a**, If these distributions are allowed to change but are uncontrolled for, nestedness appears to be negatively correlated to persistence ($P < 10^{-4}$). **b**, **c**, However, when these changes are appropriately controlled for—either statistically (**b**) or in the null model for randomization (**c**)—there is a significant positive relationship between nestedness and persistence ($P < 10^{-4}$ and $P < 10^{-4}$, respectively). The same general conclusion is reached here for the probabilistic null model hold for other, non-degree-preserving randomizations³.

mutualistic interactions of a species plays an important role in its survival^{2,3,6,13}, we find unambiguous support for the added importance of the way in which mutualistic interactions are organized—the true architecture of biodiversity¹⁴. Echoing ref. 2, our findings re-emphasize the importance of carefully considering the interplay between all potential sources of variation¹¹ in ecological models. Otherwise, one runs the risk of further entangling models that are sufficiently tangled already.

Methods

For 59 empirical networks, we generated 250 randomized networks and for each we simulated persistence (the fraction of surviving species in each simulation) across 250 parameterizations of a dynamic mutualistic model¹². We quantified the relationship between persistence and nestedness with a mixed-effects logistic regression¹⁵ that takes the form $\text{logit}(P_{ijk}) = \beta_0 + \beta_1 M_i + \beta_2 C_i + \beta_3 W_{ij} + \beta_4 N_{ij} + n_i + r_{ij} + \epsilon_{ijk}$. Here the indices i , j , and k indicate the empirical network, network randomization and model parameterization, respectively. β_0 is a constant, the slopes β_1 , β_2 , β_3 , and β_4 quantify the importance of network magnitude², M , connectance², C , relative

degree homogeneity $^{12} W$, and nestedness $^{9} N$, respectively, the random effects n_i and r_{ij} control for variance across networks and randomizations, and ε_{ijk} is the model residual. Variance inflation factors gave no indication of multicollinearity in this model.

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1. Bastolla, U. *et al.* The architecture of mutualistic networks minimizes competition and increases biodiversity. *Nature* **458**, 1018–1020 (2009).
2. James, A., Pitchford, J. W. & Plank, M. J. Disentangling nestedness from models of ecological complexity. *Nature* **487**, 227–230 (2012).
3. Thébaud, E. & Fontaine, C. Stability of ecological communities and the architecture of mutualistic and trophic networks. *Science* **329**, 853–856 (2010).
4. Jordano, P., Bascompte, J. & Olesen, J. M. Invariant properties in coevolutionary networks of plant-animal interactions. *Ecol. Lett.* **6**, 69–81 (2003).
5. Fortuna, M. A. *et al.* Nestedness versus modularity in ecological networks: two sides of the same coin? *J. Anim. Ecol.* **79**, 811–817 (2010).
6. Saavedra, S., Stouffer, D. B., Uzzi, B. & Bascompte, J. Strong contributors to network persistence are the most vulnerable to extinction. *Nature* **478**, 233–235 (2011).
7. Stouffer, D. B., Camacho, J., Guimarães, R., Ng, C. A. & Amaral, L. A. N. Quantitative patterns in the structure of model and empirical food webs. *Ecology* **86**, 1301–1311 (2005).
8. Rezende, E. L., Lavabre, J. E., Guimarães, P. R., Jordano, P. & Bascompte, J. Non-random coextinctions in phylogenetically structured mutualistic networks. *Nature* **448**, 925–928 (2007).

9. Almeida-Neto, M., Guimarães, P., Guimarães, P. R. Jr., Loyola, R. D. & Ulrich, W. A consistent metric for nestedness analysis in ecological systems: reconciling concept and measurement. *Oikos* **117**, 1227–1239 (2008).
10. Ulrich, W., Almeida-Neto, M. & Gotelli, N. J. A consumer's guide to nestedness analysis. *Oikos* **118**, 3–17 (2009).
11. Gotelli, N. J. & Ulrich, W. Statistical challenges in null model analysis. *Oikos* **121**, 171–180 (2012).
12. Williams, R. J. Biology, methodology or chance? the degree distributions of bipartite ecological networks. *PLoS ONE* **6**, e17645 (2011).
13. Allesina, S. Ecology: the more the merrier. *Nature* **487**, 175–176 (2012).
14. Bascompte, J. & Jordano, P. The structure of plant-animal mutualistic networks: the architecture of biodiversity. *Annu. Rev. Ecol. Syst.* **38**, 567–593 (2007).
15. Zuur, A., Ieno, E. N., Walker, N., Saveliev, A. A. & Smith, G. M. *Mixed Effects Models and Extensions in Ecology with R* (Springer, 2009).

Author Contributions S.S. and D.B.S. designed the study, performed the simulations, analysed the data and wrote the manuscript.

Competing Financial Interests Declared none.

doi:10.1038/nature12380

Appendix

We randomized the empirical networks with two null models: the probabilistic and fixed (or swap) algorithms⁵. For our purposes here, the key distinction between the two is that the probabilistic model generates random networks with quantitatively more homogeneous degree distribution than those observed empirically ($W_{ij} > 0$) whereas the degree distribution is strictly conserved in networks generated by the fixed model ($W_{ij} = 0$). The statistical analyses presented here were performed in R version 2.15.3 (<http://R-project.org/>) using the glmer function in package lme4 version 0.999999-0 (<http://lme4.r-forge.R-project.org/>). Code to perform the network randomizations and dynamic simulations in Matlab (<http://www.matlab.com/>) and the mixed-effects logistic regressions in R (<http://R-project.org/>) is available from the Dryad Digital Repository at <http://dx.doi.org/10.5061/dryad.p2gq8>.

