

Linking biodiversity and food-web functions: A bioenergetic approach

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Abstract: Biodiversity drives various aspects of ecosystem functioning, including biomass production and community stability. However, many measures of network structure are also associated with ecosystem functioning. Considering the scaling of network structure with species richness, the effect of biodiversity on the functioning of ecological communities might partly be due to its intimate relationship with network structure. Here I describe the biodiversity-ecosystem functioning (BEF) relationship under the unifying framework of ecological networks. Simulating biomass flows in food webs and comparing different structural equation models, I show that BEF relationships are best described by both direct and indirect effects of biodiversity on ecosystem functioning.

1 Introduction

2 One of the main reasons for preserving biodiversity is to maintain ecosystem services, which
3 are supported by associated ecological processes (Isbell et al. 2017). Perhaps the simplest
4 and most studied measure of biodiversity is species richness (i.e. the number of species in a
5 biological community). Species richness has been associated to various ecological functions,
6 including nutrient retention (Zhang et al. 2010), decomposition (Dangles and Malmqvist 2004),
7 productivity (Mittelbach et al. 2001), and ecosystem stability (Ives, Klug, and Gross 2000).
8 The biodiversity-ecosystem functioning (BEF) relationships describe these associations, and
9 one of the main goals of biodiversity science has been to understand the mechanisms driving
10 these relationships in various systems. Niche complementarity (Loreau 2000), species regional
11 complementarity (Bond and Chase 2002), and mutualistic interactions (Schleuning, Fründ, and
12 García 2015) are examples of well-known mechanisms responsible for the BEF relationships.

13 The structure of ecological networks has also been associated with ecosystem functioning and
14 other ecological processes (Pascual et al. 2006; Gómez, Perfectti, and Jordano 2011). For
15 example, food webs having a high connectance (i.e. highly connected webs) are typically more
16 robust to random losses of species (Dunne, Williams, and Martinez 2002). Trophic interactions
17 can also directly impact ecosystem productivity (Worm and Duffy 2003).

18 Because of the scaling of many measures of network structure with species richness, includ-
19 ing connectance (MacDonald, Banville, and Poisot 2020), it is difficult to disentangle the direct
20 effects of biodiversity on ecosystem functioning from its indirect effects through food-web struc-
21 ture. In other words, are biodiversity-functioning relationships in food webs mainly driven by
22 their total number of species or by their emerging structure? This knowledge will notably help
23 us implement better conservation strategies by targeting more specifically different ecosystem
24 properties to preserve. I expect that both direct and indirect effects should be important path-
25 ways in BEF relationships. In this study, I will examine what drives biomass production and
26 population stability in food webs. To do so, I will simulate biomass flows using a bioenergetic
27 model of food-web dynamics and compare different structural equation models of biodiversity
28 (i.e. species richness), food-web structure, and ecosystem functions.

Methods

Data

All food webs used in this study were queried from mangal.io, an extensive database of published ecological networks worldwide (Poisot et al. 2016). Networks archived on Mangal are multilayer networks, i.e. different types of interactions can be found in the same network. I considered as food webs all networks whose interactions were mainly of trophic type (i.e. predation and herbivory interactions). When writing these lines, 235 networks correspond to that definition. I removed the largest food web ($S = 714$ species) from my dataset since it significantly increased total simulation time. My final dataset thus contained a total of 234 food webs whose number of species ranged between 5 and 106 ($\bar{S} = 32.6$, $s = 19.1$).

Food-web measures and biomass flows

I measured α -diversity by the number of nodes (species) S in each network. Food-web structure was also described by their connectance and nestedness. Connectance represents the proportion of realized links in ecological networks ($C = L/S^2$, where L is the number of interactions), and has been associated with many other properties of network structure (Poisot and Gravel 2014), including nestedness. The nestedness of an ecological network represents the extent to which species interacting with specialists are a subset of species interacting with generalists. Nestedness was estimated by the spectral radius (i.e. the largest eigenvalue) of the adjacency matrix (Staniczenko, Kopp, and Allesina 2013). Because of its tight association with network modularity (Fortuna et al. 2010), computing the nestedness of a network is comparable to estimating its modularity (i.e. network compartmentalization). However, nestedness is perhaps orders of magnitude easier and faster to measure than modularity.

Moreover, ecosystem functioning was simulated using the model of Yodzis and Innes (1992) of biomass flows in food webs, implemented in Julia v0.7.0 or newer (Delmas et al. 2021). Indeed, to the best of my knowledge, no extensive dataset of species interactions and ecosystem functioning were available. In this context, I had to rely on first-order simulations to conduct my analysis. All model parameters (e.g. carrying capacities, body-mass ratios, growth rates,

coefficients of interspecific competition, *etc.*) were kept to their default values (Delmas et al. 2017, 2021). In the absence of empirical data on species biomass, the initial biomass of every species in the model were sampled from a uniform distribution $U(0, 1)$. This ensured that no further biases regarding initial values were introduced arbitrarily in the simulation process. Simulations of biomass flows were performed for 500 timesteps, and repeated 50 times for each food web. Because 13 food webs could not be simulated appropriately, I ended up with 221 simulated webs whose total biomass and population stability were measured and averaged over the last 10 timesteps. Total biomass was estimated as the sum of all species biomass at the end of simulation, whereas population stability was measured as the mean of the negative coefficient of variation of species biomass (Delmas et al. 2017). I used the total biomass at the end of the simulation as a proxy of biomass production. I identified the median of both simulated measures for each web, as well as their 95% percentile intervals (2.5% and 97.5% quantiles).

Structural equation models

To identify which emerging ecosystem properties drove the functioning of food webs, I evaluated the fit of seven structural equation models (SEMs) against my measured and simulated data. The most saturated model (model 6) is represented by eq. 1.

$$Y = \beta_0 + \beta_{YS}S + \beta_{YC}C + \beta_{YN}N$$

$$N = \beta_1 + \beta_{NS}S + \beta_{NC}C$$

$$C = \beta_2 + \beta_{CS}S, \tag{1}$$

where Y is the predicted variable (i.e. either total biomass or population stability), S is species richness, C the connectance, and N the nestedness of all food webs. The subscripts ij of the parameters indicate the hypothesized causal relationship between the corresponding variables i and j (i.e. $j \rightarrow i$). The parameters β_0 , β_1 , and β_2 are the model intercepts. All parameters were estimated in model 6 using the lavaan package v.0.6-7 in R. I imposed further structure on the other six models (see tbl. 1).

Table 1: **Model sets.** Every parameter was either estimated (value=1) or not (value=0) in each of the seven structural equation models. The subscripts ij of the parameters indicate the hypothesized causal relationship between the corresponding variables i and j (i.e. $j \rightarrow i$). Y is the predicted variable (either total biomass or population stability), S the number of species, C the connectance, and N the nestedness of a food web. The parameters β_0 , β_1 , and β_2 are the intercepts of Y , N , and C , respectively.

Model	β_0	β_{YS}	β_{YC}	β_{YN}	β_1	β_{NS}	β_{NC}	β_2	β_{CS}
model 0	1	0	0	0	0	0	0	0	0
model 1	1	1	0	0	0	0	0	0	0
model 2	1	0	1	0	0	0	0	0	0
model 3	1	1	1	0	0	0	0	1	1
model 4	1	0	1	1	1	0	1	0	0
model 5	1	1	1	1	1	0	1	1	1
model 6	1	1	1	1	1	1	1	1	1

These seven structural equation models were fit against both Y variables (i.e. total biomass and population stability) independently. I performed model selection using the Akaike information criterion (AIC) for both variables.

Data and code availability

All code and data to reproduce this project are available on GitHub (<https://github.com/FrancisBanville/BIO860M-Project>). Simulations and analyses were all performed in Julia v.1.5.3, except the structural equation modeling which was performed in R v.3.6.1.

Results

Model selection

The most saturated model (model 6) was the best fit model for both the prediction of total biomass and population stability (tbl. 2). However, model 5 was also a really good fit to the data, with

a Δ AIC of 4.1 for the two predicted variables. In both cases, model 6 had an Akaike weight of 88.7% and model 5 had the remaining 11.3%. Model 6 only differed from model 5 by the imposed relationship between species richness and nestedness. Therefore, using an ensemble model did not seem necessary to me. Model 6 was thus selected as the unique best structural equation model.

Table 2: **Model selection.** The difference of AIC scores between each model and the best model were computed for the two sets of models (i.e. the predictive models of total biomass and of population stability). Model 6 was the best model for both sets according to the information criterion.

Model	df	Δ AIC _{biomass}	Δ AIC _{stability}
6	12	0	0
5	11	4.1	4.1
4	7	718.2	650.3
3	7	722.5	687.3
2	3	1441.3	1333.3
1	3	1363.2	1332.2
0	2	1475.8	1333.1

Path analysis

All direct and indirect relationships between measures of ecosystem functioning, biodiversity and food-web structure were analyzed using model 6. Regression coefficients between pairs of variables are listed in tbl. 3 for the prediction of total biomass and in tbl. 4 for the prediction of population stability.

The total biomass of the community after simulation was positively associated with the total number of species and the nestedness of food webs at the beginning of the simulation, but negatively associated with connectance (tbl. 3). However, since both measures of food-web structure were associated with species richness, species richness also shaped total biomass through these indirect pathways.

Table 3: **Results of the path analysis for the prediction of total biomass using model 6.** All regression coefficients were significant at the significance level of $\alpha = 0.05$ (Y = total biomass, S = species richness, C = connectance, N = nestedness). Model 6 was selected according to the information criterion.

Coefficient	estimate	z-value	p-value
β_{YS}	0.12	9.5	< 0.001***
β_{YC}	-36.3	-6.0	< 0.001***
β_{YN}	24.8	6.2	< 0.001***
β_{NS}	0.001	2.5	0.01*
β_{NC}	1.2	20.4	< 0.001***
β_{CS}	-0.002	-8.7	< 0.001***

106 These relationships were weaker when model 6 was used to predict population stability (tbl. 4).

107 Population stability was not significantly associated with either measure of network structure,

108 but was only weakly associated with species richness.

Table 4: **Results of the path analysis for the prediction of population stability using model 6.** All regression coefficients were significant at the significance level of $\alpha = 0.05$, except for β_{YC} and β_{YN} (Y = population stability, S = species richness, C = connectance, N = nestedness). Model 6 was selected according to the information criterion.

Coefficient	estimate	z-value	p-value
β_{YS}	0.002	2.8	0.005**
β_{YC}	0.587	1.8	0.067
β_{YN}	-0.082	-0.4	0.700
β_{NS}	0.001	2.5	0.01*
β_{NC}	1.2	20.4	< 0.001***
β_{CS}	-0.002	-8.7	< 0.001***

109 Bivariate relationships between measures of ecosystem functioning and measures of biodiversity

110 and food-web structure were plotted in fig. 1. We can identify the positive relationship between

111 total biomass and species richness, and its negative relationship with connectance. However, the

confidence intervals around the point estimates of total biomass were wider for food webs with more species and narrower for networks with high connectance. On the other hand, the weak relationship between population stability and species richness is not easily observable in fig. 1.

[Figure 1 about here.]

Discussion

Biodiversity and ecosystem functioning: direct pathways

My study suggests that biodiversity remains an important driver of ecosystem functioning, independently of food-web structure. First, the total biomass of a biological community at the end of my simulations increased with the total number of species. The positive relationship between biomass productivity and species richness uncovered in my study was previously observed in various systems and taxa (Mittelbach et al. 2001) and at different spatial scales (Gonzalez et al. 2020). More diverse systems might indeed use resources and nutrients more efficiently than homogeneous communities because of niche complementarities (Tilman, Wedin, and Knops 1996).

Moreover, I found that the temporal stability of species biomass also increased with species richness. This is in agreement with the work of Ives, Klug, and Gross (2000) and Gross et al. (2014). Species typically have different levels of tolerance to the variation of environmental conditions and to the variation of other species' abundances. A decrease of biomass from a given species might be compensated by an increase of biomass from another species in the community. A high number of species might promote ecosystem stability because of a higher probability of finding complementary species. However, the positive relationship between population stability and species richness uncovered in this study seems to be in disagreement with the work of May (1972), Allesina and Tang (2012), and MacDonald, Banville, and Poisot (2020), who found that the probability of a network to be stable decreases with species richness. This might however be due to differences in the definitions and measures used to describe ecosystem stability (biomass stability vs food-web stability).

The relationship between other measures of biodiversity, such as species evenness and measures of β -diversity, and ecosystem functioning should also be investigated alongside measures of network structure. This might unveil intriguing aspects of the BEF relationship in the light of food-web structure.

Biodiversity and ecosystem functioning: indirect pathways

Species richness also drove food-web functioning through indirect pathways. For example, species richness was negatively associated with food-web connectance, which was positively associated with total biomass. This was however in contrast with the relationship between species richness and population stability, which did not have any indirect pathways through food-web structure.

The bioenergetic model of biomass flows of Yodzis and Innes (1992) and Delmas et al. (2017) allowed me to simulate ecosystem functioning of many published food webs. This model was useful to make first order predictions of total biomass and population stability. However, many parameter values were set to their default values or sampled from a non-biased distribution (i.e. a uniform distribution), although they were probably system specific. To the best of my knowledge, no empirical data was available to validate this choice of parameter values in a broad range of food webs. When collecting data on species interactions, ecologists should estimate as much as possible the biomass of individual species. This would enable the conduction of more extensive studies on biodiversity and ecosystem functioning. Overall, my results nevertheless highlight the importance of considering network structure when describing the biodiversity-ecosystem functioning (BEF) relationship, especially with regards to biomass production.

References

- Allesina, Stefano, and Si Tang. 2012. "Stability Criteria for Complex Ecosystems." *Nature* 483 (7388, 7388): 205–8. <https://doi.org/10.1038/nature10832>.
- Bond, Emily M., and Jonathan M. Chase. 2002. "Biodiversity and Ecosystem Functioning at Local and Regional Spatial Scales." *Ecology Letters* 5 (4): 467–70. <https://doi.org/10.1017/S1461024802001511>.

1046/j.1461-0248.2002.00350.x.

Dangles, Olivier, and Björn Malmqvist. 2004. "Species Richness–Decomposition Relationships Depend on Species Dominance." *Ecology Letters* 7 (5): 395–402. <https://doi.org/10.1111/j.1461-0248.2004.00591.x>.

Delmas, Eva, Azenor Bideault, Tom Clegg, and Timothée Poisot. 2021. *Simulations of Biomass Dynamics and Extinctions in Food Webs*. Zenodo. <https://doi.org/10.5281/zenodo.4426680>.

Delmas, Eva, Ulrich Brose, Dominique Gravel, Daniel B. Stouffer, and Timothée Poisot. 2017. "Simulations of Biomass Dynamics in Community Food Webs." *Methods in Ecology and Evolution* 8 (7): 881–86. <https://doi.org/10.1111/2041-210X.12713>.

Dunne, Jennifer A., Richard J. Williams, and Neo D. Martinez. 2002. "Network Structure and Biodiversity Loss in Food Webs: Robustness Increases with Connectance." *Ecology Letters* 5 (4): 558–67. <https://doi.org/10.1046/j.1461-0248.2002.00354.x>.

Fortuna, Miguel A., Daniel B. Stouffer, Jens M. Olesen, Pedro Jordano, David Mouillot, Boris R. Krasnov, Robert Poulin, and Jordi Bascompte. 2010. "Nestedness Versus Modularity in Ecological Networks: Two Sides of the Same Coin?" *Journal of Animal Ecology* 79 (4): 811–17. <https://doi.org/10.1111/j.1365-2656.2010.01688.x>.

Gonzalez, Andrew, Rachel M. Germain, Diane S. Srivastava, Elise Filotas, Laura E. Dee, Dominique Gravel, Patrick L. Thompson, et al. 2020. "Scaling-up Biodiversity-Ecosystem Functioning Research." *Ecology Letters* 23 (4): 757–76. <https://doi.org/10.1111/ele.13456>.

Gómez, José M., Francisco Perfectti, and Pedro Jordano. 2011. "The Functional Consequences of Mutualistic Network Architecture." *PLOS ONE* 6 (1): e16143. <https://doi.org/10.1371/journal.pone.0016143>.

Gross, Kevin, Bradley J. Cardinale, Jeremy W. Fox, Andrew Gonzalez, Michel Loreau, H. Wayne Polley, Peter B. Reich, and Jasper van Ruijven. 2014. "Species Richness and the Temporal Stability of Biomass Production: A New Analysis of Recent Biodiversity Experiments." *The American Naturalist* 183 (1): 1–12. <https://doi.org/10.1086/673915>.

- Isbell, Forest, Andrew Gonzalez, Michel Loreau, Jane Cowles, Sandra Díaz, Andy Hector, Georgina M. Mace, et al. 2017. “Linking the Influence and Dependence of People on Biodiversity Across Scales.” *Nature* 546 (7656, 7656): 65–72. <https://doi.org/10.1038/nature22899>.
- Ives, A. R., J. L. Klug, and K. Gross. 2000. “Stability and Species Richness in Complex Communities.” *Ecology Letters* 3 (5): 399–411. <https://doi.org/10.1046/j.1461-0248.2000.00144.x>.
- Loreau, Michel. 2000. “Biodiversity and Ecosystem Functioning: Recent Theoretical Advances.” *Oikos* 91 (1): 3–17. <https://doi.org/10.1034/j.1600-0706.2000.910101.x>.
- MacDonald, Arthur Andrew Meahan, Francis Banville, and Timothée Poisot. 2020. “Revisiting the Links-Species Scaling Relationship in Food Webs.” *Patterns* 0 (0). <https://doi.org/10.1016/j.patter.2020.100079>.
- May, Robert M. 1972. “Will a Large Complex System Be Stable?” *Nature* 238 (5364, 5364): 413–14. <https://doi.org/10.1038/238413a0>.
- Mittelbach, Gary G., Christopher F. Steiner, Samuel M. Scheiner, Katherine L. Gross, Heather L. Reynolds, Robert B. Waide, Michael R. Willig, Stanley I. Dodson, and Laura Gough. 2001. “What Is the Observed Relationship Between Species Richness and Productivity?” *Ecology* 82 (9): 2381–96. [https://doi.org/10.1890/0012-9658\(2001\)082%5B2381:WITORB%5D2.0.CO;2](https://doi.org/10.1890/0012-9658(2001)082%5B2381:WITORB%5D2.0.CO;2).
- Pascual, Department of Ecology and Evolutionary Biology Mercedes, Mercedes Pascual, Jennifer A. Dunne, Jennifer A. Dunne, and Visiting Professor Jennifer A. Dunne. 2006. *Ecological Networks: Linking Structure to Dynamics in Food Webs*. Oxford University Press, USA. <http://books.google.com?id=YpQRDAAQBAJ>.
- Poisot, Timothée, Benjamin Baiser, Jennifer A. Dunne, Sonia Kéfi, François Massol, Nicolas Mouquet, Tamara N. Romanuk, Daniel B. Stouffer, Spencer A. Wood, and Dominique Gravel. 2016. “Mangal – Making Ecological Network Analysis Simple.” *Ecography* 39 (4): 384–90. <https://doi.org/10.1111/ecog.00976>.

220 Poisot, Timothée, and Dominique Gravel. 2014. “When Is an Ecological Network Complex?
221 Connectance Drives Degree Distribution and Emerging Network Properties.” *PeerJ* 2 (Febru-
222 ary): e251. <https://doi.org/10.7717/peerj.251>.

223 Schleuning, Matthias, Jochen Fründ, and Daniel García. 2015. “Predicting Ecosystem Func-
224 tions from Biodiversity and Mutualistic Networks: An Extension of Trait-Based Concepts
225 to Plant–Animal Interactions.” *Ecography* 38 (4): 380–92. [https://doi.org/10.1111/](https://doi.org/10.1111/ecog.00983)
226 [ecog.00983](https://doi.org/10.1111/ecog.00983).

227 Staniczenko, Phillip P. A., Jason C. Kopp, and Stefano Allesina. 2013. “The Ghost of Nested-
228 ness in Ecological Networks.” *Nature Communications* 4 (1, 1): 1391. [https://doi.org/](https://doi.org/10.1038/ncomms2422)
229 [10.1038/ncomms2422](https://doi.org/10.1038/ncomms2422).

230 Tilman, David, David Wedin, and Johannes Knops. 1996. “Productivity and Sustainability
231 Influenced by Biodiversity in Grassland Ecosystems.” *Nature* 379 (6567, 6567): 718–20.
232 <https://doi.org/10.1038/379718a0>.

233 Worm, Boris, and J. Emmett Duffy. 2003. “Biodiversity, Productivity and Stability in Real Food
234 Webs.” *Trends in Ecology & Evolution* 18 (12): 628–32. [https://doi.org/10.1016/j.](https://doi.org/10.1016/j.tree.2003.09.003)
235 [tree.2003.09.003](https://doi.org/10.1016/j.tree.2003.09.003).

236 Yodzis, P., and S. Innes. 1992. “Body Size and Consumer-Resource Dynamics.” *The American*
237 *Naturalist* 139 (6): 1151–75. <https://doi.org/10.1086/285380>.

238 Zhang, Chong-Bang, Jiang Wang, Wen-Li Liu, Si-Xi Zhu, Dong Liu, Scott X Chang, Jie Chang,
239 and Ying Ge. 2010. “Effects of Plant Diversity on Nutrient Retention and Enzyme Activities
240 in a Full-Scale Constructed Wetland.” *Bioresource Technology* 101 (6): 1686–92. [https:](https://doi.org/10.1016/j.biortech.2009.10.001)
241 [/doi.org/10.1016/j.biortech.2009.10.001](https://doi.org/10.1016/j.biortech.2009.10.001).

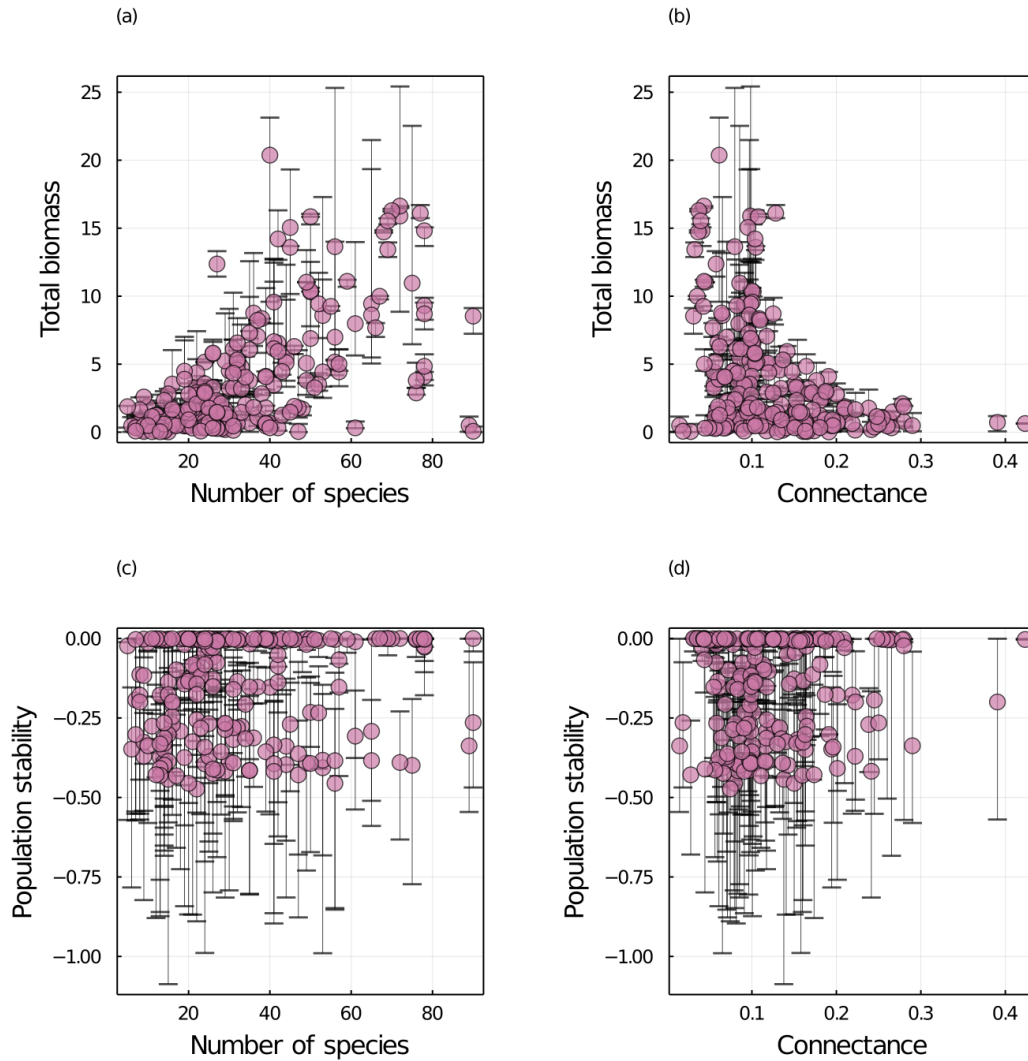


Figure 1: **Bivariate relationships between food-web measures.** (a-b) Relationship between total biomass and (a) species richness and (b) connectance. The total biomass of a community represents the sum of biomass for all species averaged over the last 10 timesteps of the simulations. (c-d) Relationship between population stability and (c) species richness and (d) connectance. The temporal stability of population biomass was estimated using the negative coefficient of variation of biomass across populations averaged over the last 10 timesteps of the simulations. Species richness and connectance were measured before simulating. All simulations ($n=221$ webs) were run over 500 timesteps and were repeated 50 times for every food web archived on mangal.io (except for the largest web and 13 webs where simulation failed). In each panel, error bars indicate the 95% percentile intervals of total biomass and population stability, respectively.