







Ecological challenges and opportunities

Pablo A. Marquet

Tara workshop, Valparíso, May 15-18, 2023

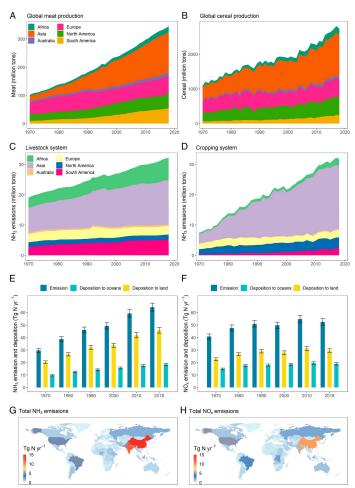
The most salient feature of life has been the stability of its bacterial mode from the beginning of the fossil record until today and, with little doubt, into all future time so long as the earth endures. This is truly the "age of bacteria"—as it was in the beginning is now and ever shall be.

Stephen Jay Gould (1994)

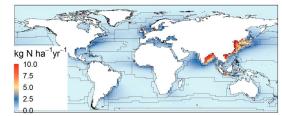
Border conditions.....

Nitrogen coming from the sky!!!

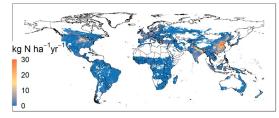
Ocean Nitrogen deposition



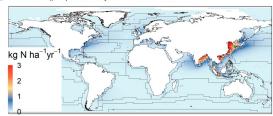
A Diff. in NH_x deposition by food production



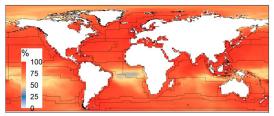
C NH₃ emissions by N fertilizer overuse



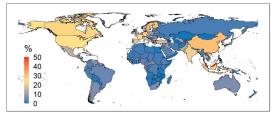
E Diff. in NH_x deposition by overused N fertilizer



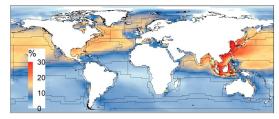
 $B\,$ Changes in NH_x deposition (%) by food production



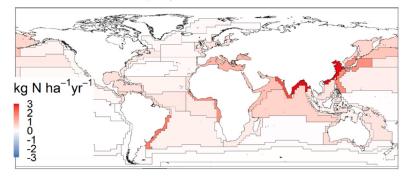
D Contribution of overused N fertilizer to NH₃ emission



F Changes in NH_x deposition (%) by overused N fertilizer



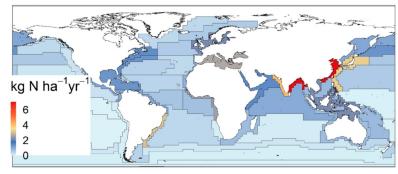
A Difference in total NH_x deposition between 1970-2018

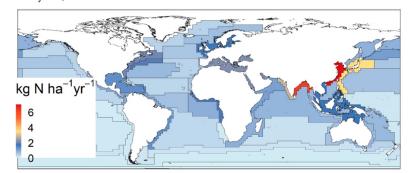


 $C \,$ NH_x deposition in 2018

D NO_y deposition in 2018

-1 -2 -3





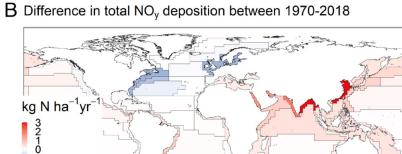


RESEARCH ARTICLE SUSTAINABILITY SCIENCE

Modeling global oceanic nitrogen deposition from food systems and its mitigation potential by reducing overuse of fertilizers

Lei Liu^{a,1} 🕲, Wen Xu^b, Zhang Wen¹ 🕲, Pu Liu^a, Hang Xu^a, Sheng Liu^a, Xiankai Lu^c 🕲, Buqing Zhong^c 🕲, Yixin Guo^a 🕲, Xiao Lu^a, Yuanhong Zhao^f, Xiuying Zhang^e, Songhan Wang^b, Peter M. Vitousek¹ 🕲, and Xuejun Liu^{b,1} 🚳

Edited by Stephen Carpenter, University of Wisconsin-Madison, Madison, WI; received January 9, 2023; accepted March 6, 2023



The Nitrogen cycle works different in polluted areas!!!

nature

Explore content \checkmark About the journal \checkmark Publish with us \checkmark

<u>nature</u> > <u>letters</u> > article

Published: 24 January 2002

Nitrogen loss from unpolluted South American forests mainly via dissolved organic compounds

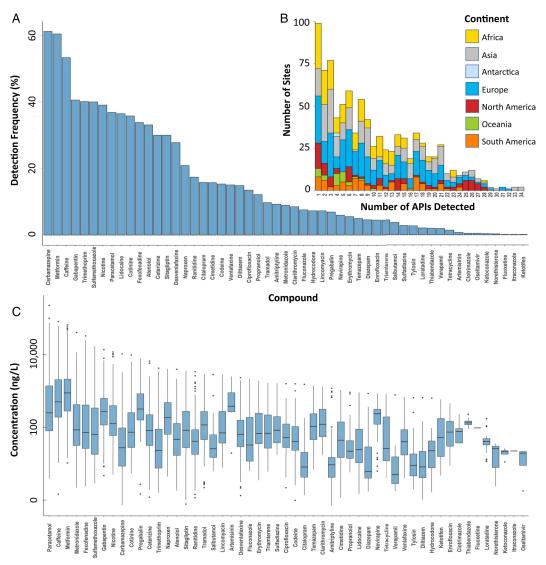
Steven S. Perakis 🗠 & Lars O. Hedin

S@\$%... coming from rivers



Pharmaceutical pollution of the world's rivers

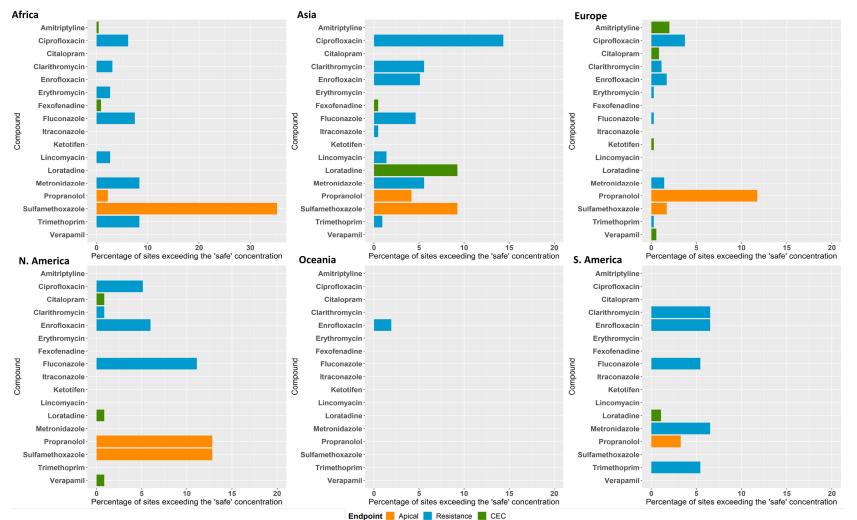
John L. Wilkinson^{a,1}, Alistair B. A. Boxall^a, Dana W. Kolpin^b, Kenneth M. Y. Leung^c, Racliffe W. S. Lai^c,



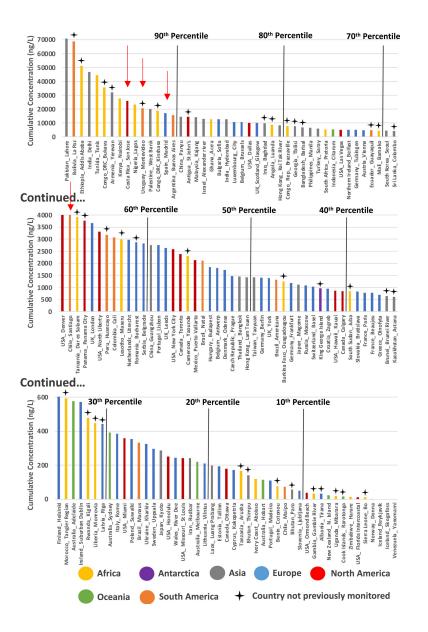
Glossary:

API = Active Pharmaceutical Ingredients

Compound



CEC= Critical Environmental Concentrations



Ecology, 69(2), 1988, pp. 508-515 © 1988 by the Ecological Society of America

THE INDETERMINACY OF ECOLOGICAL INTERACTIONS AS PERCEIVED THROUGH PERTURBATION EXPERIMENTS¹

Peter Yodzis

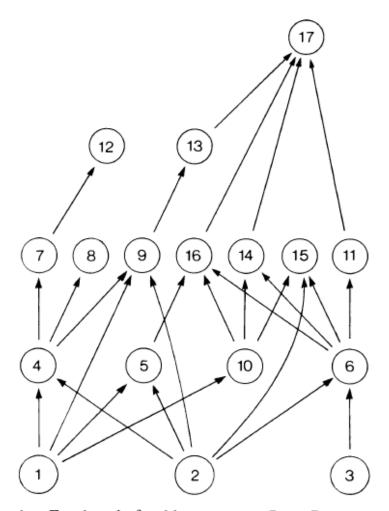


FIG. 1. Food web for Narragansett Bay, Rhode Island. Each vertex (circle) represents a species, and an arrow from vertex i to vertex j means that species j comsumes species i.

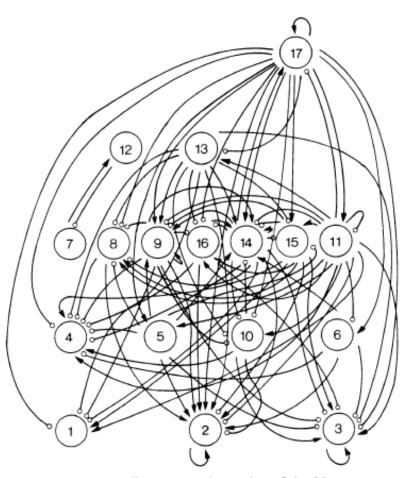
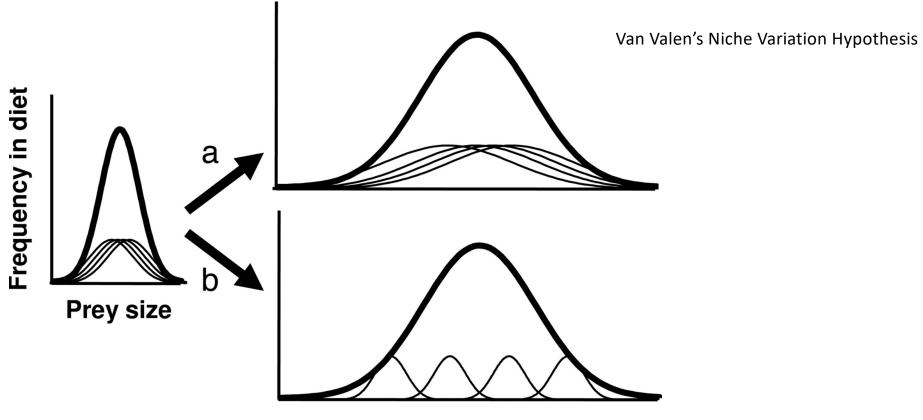


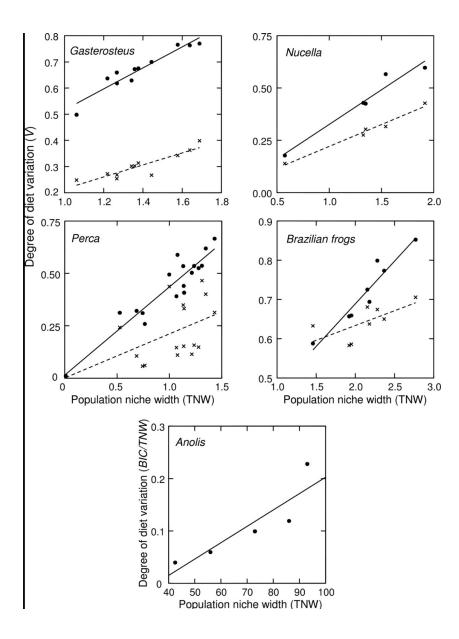
FIG. 4. Major effects on each species of the Narragansett Bay food web, obtained from another randomly generated plausible community matrix. Symbols here have the same meaning as in Fig. 2, except that here they are based on the inverse matrix A⁻¹ rather than on the community matrix itself.

"For those doing practical work with environmental impacts, it is of crucial importance to understand that short-term observations of environmental impacts that can be viewed as press perturbations are close to useless for estimating probable long-term impacts. Moreover, as the present study makes clear, predicting those long-term effects not only requires data on the strengths of many interactions in the system, it requires very accurate data on many interaction strengths. This is a daunting prospect indeed." (Yodzis 1988) Niche complexity



Roughgarden theory of niche width

Niche axis



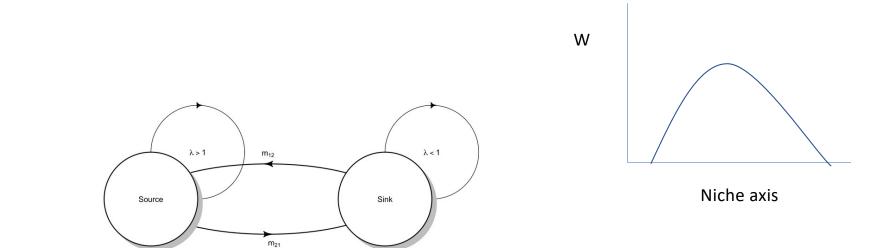


Figure 6.2: The simplest source-sink model, where $m_{21} > m_{12}$ so that net movement of individuals is from the source to the sink.



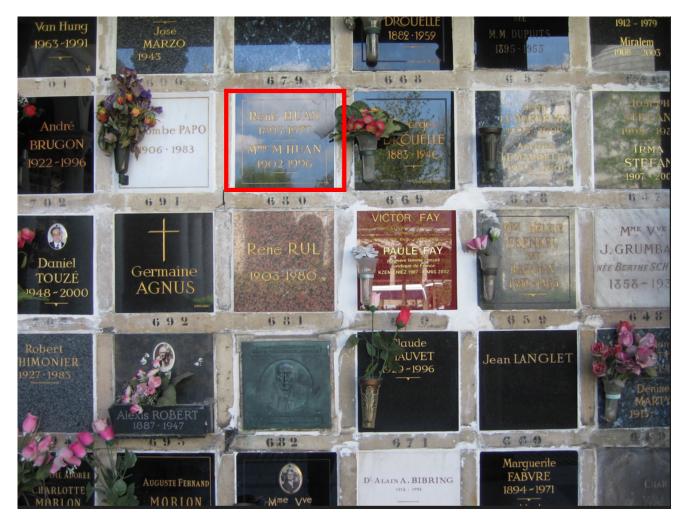
Interspecific competition among metapopulations with space-limited subpopulations

Yoh Iwasa^{*}, Jonathan Roughgarden

https://hankstevens.github.io/Primer-of-Ecology/meta.html

Niche

Colwell and Fuentes (1975)

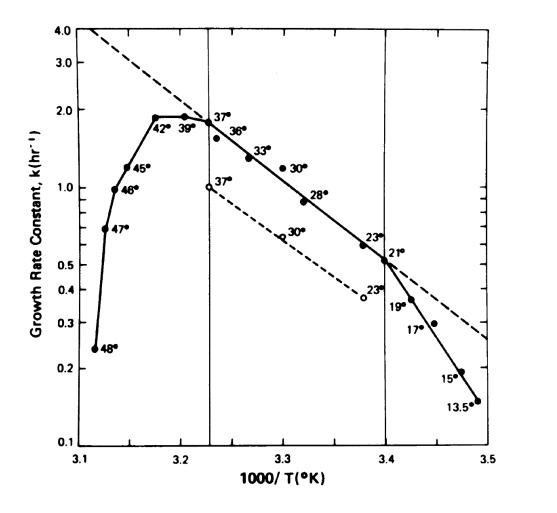


Columbarium

Colwell and Fuentes (1975)



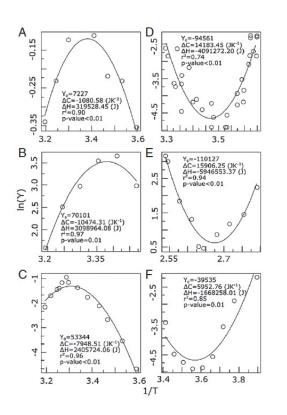
Temperature





Growth rate of E. coli

Herendeen et al



The Eyring–Evans–Polanyi (EEP) transition state theory (TST)

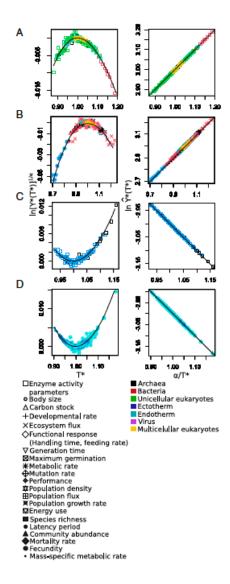
$$k = \frac{k_B}{h} T e^{\Delta S/R} e^{-\Delta H/RT}.$$

PNAS

RESEARCH ARTICLE ECOLOGY

A general theory for temperature dependence in biology

José Ignacio Arroyo^{ab,1,2}, Beatriz Díez^{cd,e}, Christopher P. Kempes^b, Geoffrey B. West^b, and Pablo A. Marquet^{ab,fg,h,2}



$$Y(T) \approx Y_0 \left(\frac{1}{T}\right)^{\frac{-\overline{\Delta C}}{R} - \alpha} e^{\frac{-\overline{\Delta H}}{RT}}.$$

$$Y^{*1/a} = T^* e^{1/T^* - 1}$$

$$\hat{Y}^*(T^*) \equiv (e/T^*)^a Y^*(T^*) = e^{a/T^*}$$

Arroyo et al. (2022) PNAS

General patterns

Diffusion processes



Sewall Wright

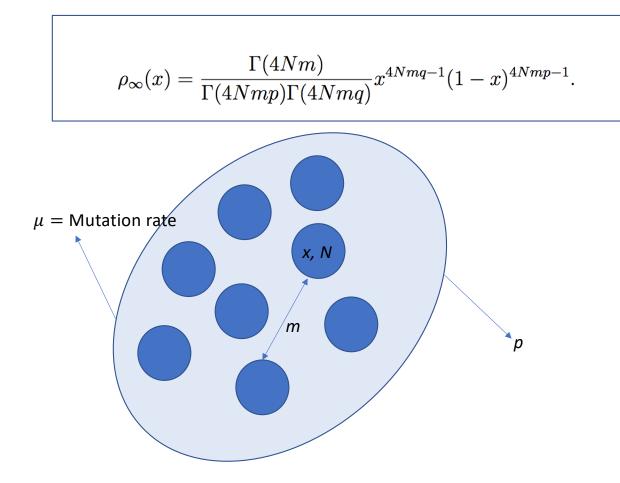




Ronald Fisher

Andrey Kolmogorov

• The frequency of genes in a structured population



- x = Frequency of a given allele in a local population
- N= Effective population size
- *m* = Proportion of migrating individuals among population each generation.
- p= Frequency of a given allele in the total population

Kolmogorov (1935) Dokl. Akad. Nauk SSSR 3 (1935), 129-132.

Large population of **N** individuals consisting of **s** partial populations with **n** individuals each (N=sn) Each generation **k** individuals disperse randomly across the partial populations.

- $ar{p}\,$ = gene frequency in the large population
- p = gen frequency in a local population
- Δp =change in the frequency of p over one generation

q = 1 - p

Following Wright and Fisher:

$$M(p) = E(\Delta p) = \frac{k}{n}(\bar{p} - p) \qquad Var(p) = E(\Delta p)^2 = \frac{pq}{2n}$$

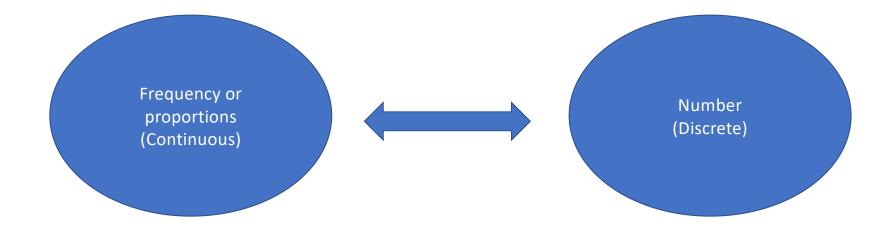
"Since *s* is large, the variation of the total concentration \bar{p} will proceed much more slowly than those of the partial concentrations *p*. Therefore, \bar{p} can temporarily be taken constant. The concentrations *p* in partial populations deviate from \bar{p} in either direction. After sufficiently long time the fluctuations of *p* around \bar{p} result in a certain stationary probability distribution for the concentrations *p*." (Kolmogorov 1935)

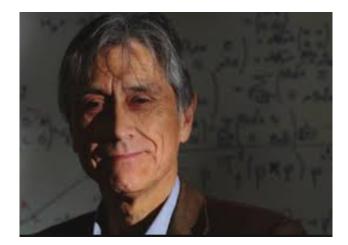
This stationary distribution satisfies the Kolmogorov forward or Fokker-Planck equation:

$$\frac{1}{2}\frac{\partial^2}{\partial p^2}(\operatorname{Var}(\mathbf{p})\boldsymbol{\mu}) - \frac{\partial}{\partial x}(M(p)\boldsymbol{\mu}) = \mathbf{0}$$

Whose solution $\mu(p)$ is:

$$\mu(p) = \frac{1}{B(4k\bar{p}, 4k\bar{q})} p^{4k\bar{p}-1} q^{4k\bar{q}-1}$$

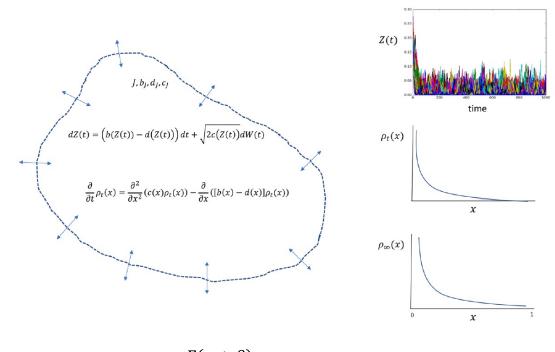




Prof. Rolando Rebolledo

La méthode des martingales appliquée à l'étude de la convergence en loi de processus

Mémoires de la S. M. F., tome 62 (1979), p. I-V+1-125. http://www.numdam.org/item?id=MSMF_1979_62_R1_0



$$\rho_{\infty}(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha) + \Gamma(\beta)} x^{\alpha - 1} (1 - x)^{\beta - 1}$$

 $B_{J}(n) = b_{J}(n) + c_{J}(n)$ $D_{J}(n) = d_{J}(n) + c_{J}(n).$ $b(x) = b_{0} + b_{1}x$ $d(x) = d_{0} + d_{1}x$

 $c(x) = \gamma x(1-x),$



OPEN On the proportional abundance of species: Integrating population genetics and community ecology PabloA. Marguet@123.5, Guillermo Espinoza¹, Sebastian R. Abades⁴, Angela Ganz² & Rolando Rebelfed^{20,4}

$$\rho_{\infty}(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha) + \Gamma(\beta)} x^{\alpha - 1} (1 - x)^{\beta - 1}$$

The Proportional Species Abundance Distribution (PSAD)

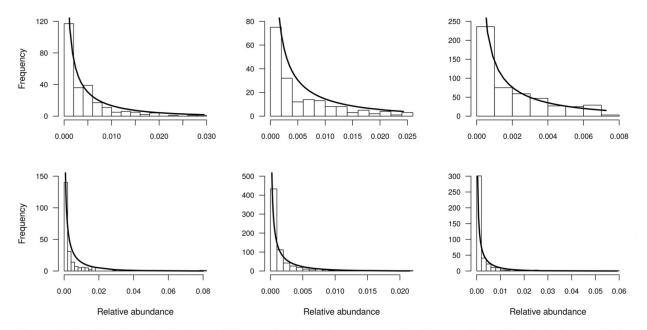
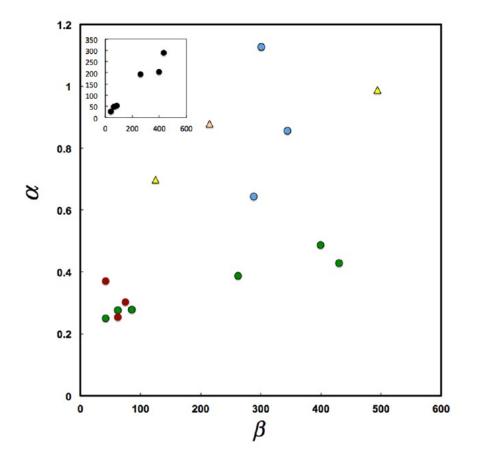


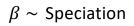
Figure 1. Fit of the Beta distribution to different animal and plant communities. First row, from left to right Amazon birds (community 10 in Table 1), Lepidoptera (12 in Table 1), butterflies (11 in Table 1), second row from left to right Tropical trees (6 in Table 1), Tropical trees (2 in Table 1) and Coral reefs (14 in Table 1)

$$\rho_{\infty}(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha) + \Gamma(\beta)} x^{\alpha - 1} (1 - x)^{\beta - 1}$$



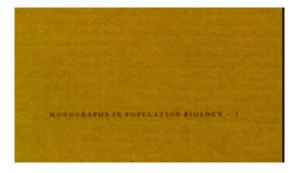


 $\alpha \sim \text{Dispersal}$



(3-4)
$$\frac{\mathrm{d}P_{s}(t)}{\mathrm{d}t} = -(\lambda_{s} + \mu_{s})P_{s}(t) + \lambda_{s-1}P_{s-1}(t) + \mu_{s+1}P_{s+1}(t) + \mu_{s+1}P_{s+1}(t)$$

$$Z(t) = Z(0) + \int_0^t \theta(Z(s)) ds + \int_0^t \sigma(Z(s)) dW_s, \ (t \ge 0).$$



THE THEORY OF

Island Biogeography

ROBERT II. MAC ARTHUR EDWARD O. WILSON

$$\rho_{\infty}(x) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha) + \Gamma(\beta)} x^{\alpha - 1} (1 - x)^{\beta - 1}$$

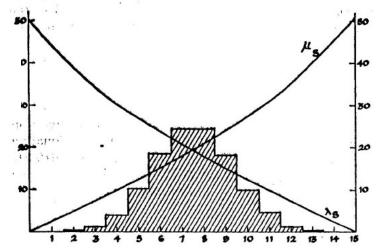
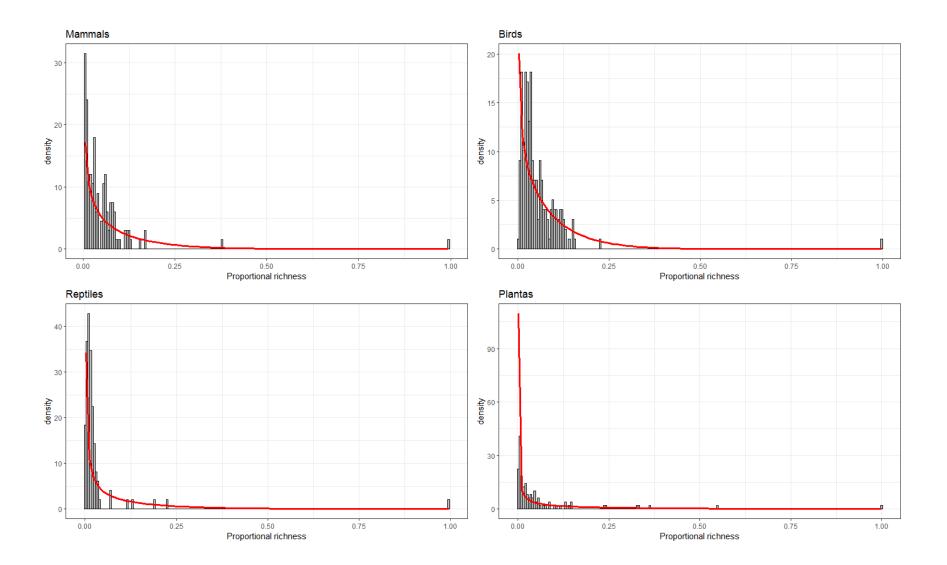


FIGURE 19. A particular case of a predicted distribution of numbers of species on a family of island biotas all with identical extinction and immigration curves and all having had time to reach equilibrium. The histogram represents the number of islands with each number of resident species in an equilibrium situation. The species pool from which the biotas were assembled contained 15 species. If the immigration and extinction curves were straighter, the variance of equilibrial species numbers would be even greater; yet this large variance is still consistent with the equilibrial condition.

"In principle one could solve eq. 3-4....for our purpose is more useful to find the mean M(t) and the variance, var(t), of the number of species at time t. These can be estimated in nature by measuring the mean and variance in number of species of a series of islands of about the same distance and are and hence of the same λ and μ ." pp.33-34



Some Sources of Stochasticity at Different Levels of Organization

Frequency of Gene families within Genomes	Frequency of genes within and among population	Frequency of individuals among species	Frequency of species Among communities
MutationLateral gene transferGene duplication	MutationDriftMigrationLoss	BirthDeathMigration	SpeciationExtinctionMigration

Abundance of gene families within genomes

- Birth-death processes are applicable to understand genome evolution
- Driving processes are similar

Duplication

Mutation

Lateral gene transfer and retroviral infection

Plasmids

Gene extinction

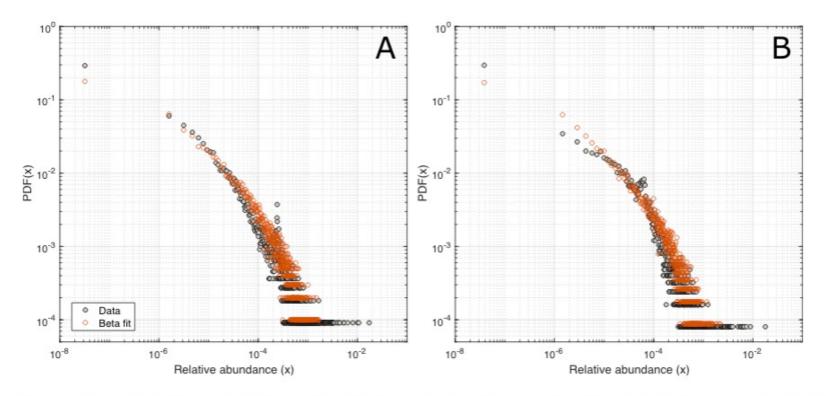


Figure 1. The relative abundance of protein families follows a Beta distribution. The relative abundance of protein families in bacteria and eukarya according to the Pfam classification. **A**. Data from 7,694 bacterial proteomes; 11,106 families; and 30,743,438 genes. Best fit beta distribution parameters: α =0.273 (0.264, 0.283), β =3,031.9 (2823.4, 3240.4). **B**. Data from 1,496 eukaryotic proteomes; 12,579 families; and 25,625,956 genes. Best fit beta distribution parameters: α =0.291 (0.281, 0.300), β =3,655.2 (3424.3, 3886.2).

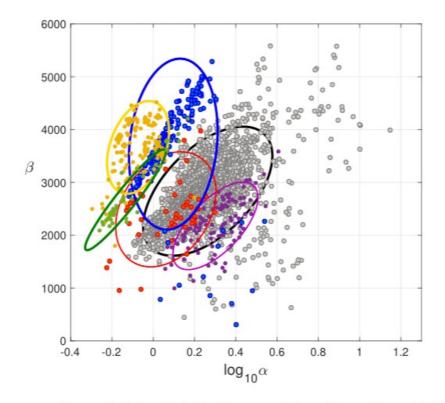


Figure 2. Shape parameters of Beta distributions arising from the relative abundance of gene families in genomes. Annotations of protein families (Pfam v32) in Uniprot reference proteomes (n=17,543). Reference proteomes include Archea (n=285, magenta), Bacteria (n=6,554, grey); unicellular eukarya (n=1,202, red); plants (n=, green); metazoa (n=, yellow); fungi (n=, blue). The total number of protein families classified by Pfam is 15,964.

Gracias