

Integrating ecological data and patterns across spatial scales, from local plots to continents

Petr Keil

iDiv

German Centre for Integrative Biodiversity Research
Halle-Jena-Leipzig, Germany

petr.keil@idiv.de

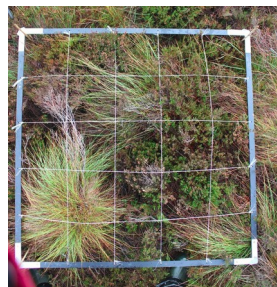


Where on Earth is biodiversity? Why? What drives it?



Biodiversity data and scale

Plots, inventories, checklists



0.062 ha

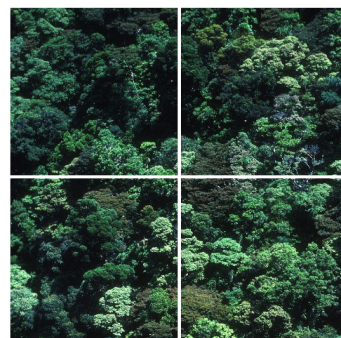
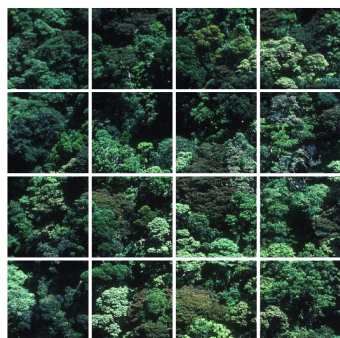
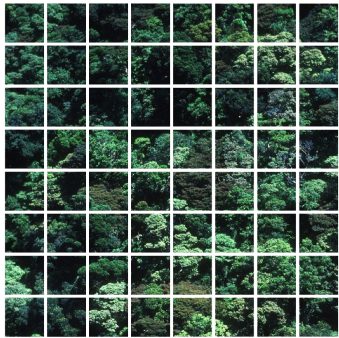
1 ha

50 ha

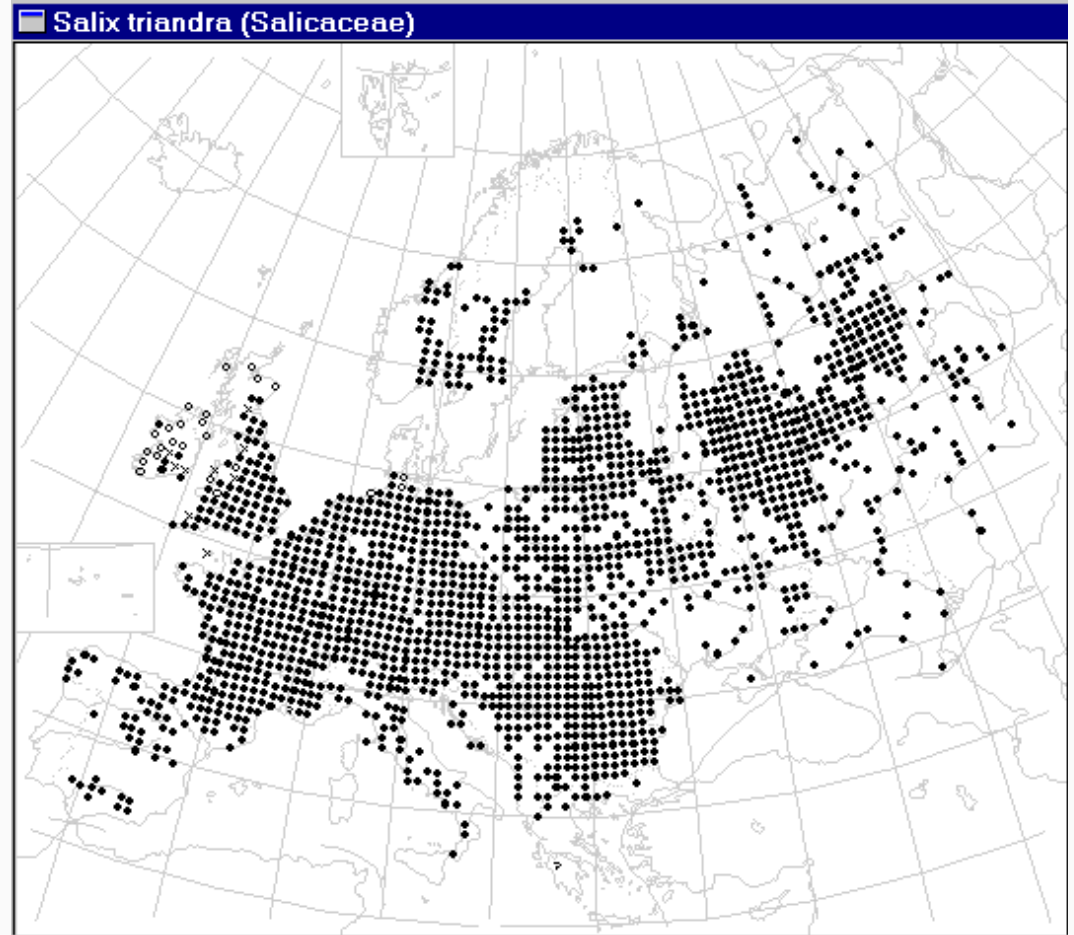
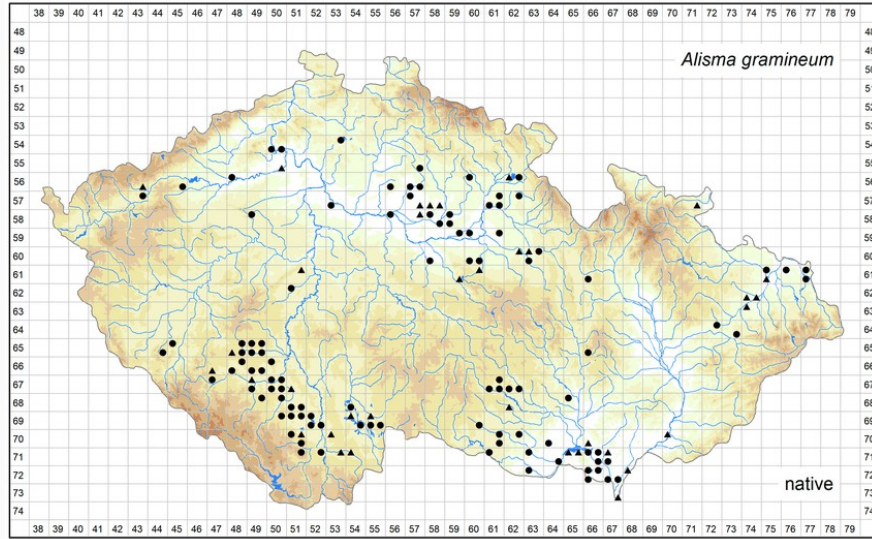
Nature
reserve

Country

Scale = Grain = Resolution = Area of grid cell = Area

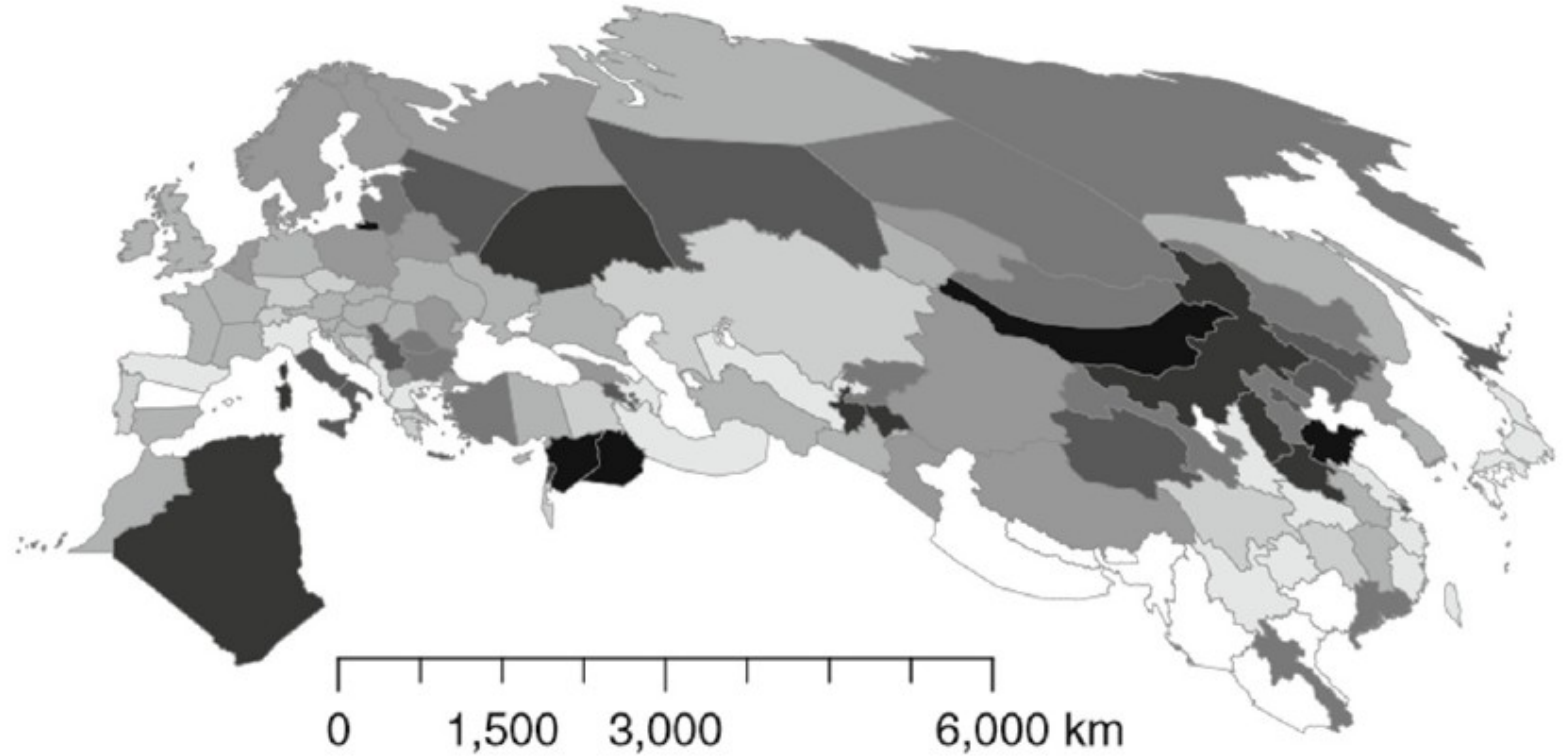


Grid atlases



Kaplan et al. (2017) *Preslia*
Jalas, Suominen, Lampinen, Kurtto: *Atlas Florae Europae*

Plots, inventories, checklists



Point observations, range maps – GBIF, OBIS, MoL, ...



MOL
MAP OF LIFE

Species Home Summary Map Detailed Map

Spring Pasqueflower

Pulsatilla vernalis
Ranunculaceae
Least Concern (IUCN 2014)

Sources

- Expert range maps 2 Q i
- Freiberg (2015) Q i
- Kalwij et al. (2014) Q i
- Point observations 12,886 Q i

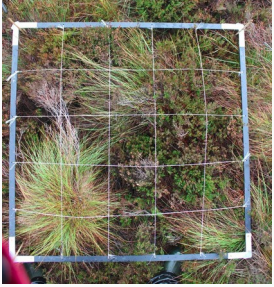
Photo: Tinelot Wittermans

Why integrate?

- Different data have different strengths, weaknesses, and gaps
- Larger N, geographic extent, environmental coverage
- Improved **inference**, e.g. about species niches, predictors of diversity, ...
- Improved **predictions** of occurrence, abundance, diversity, and community composition, both in space and time

All is one

Can we integrate the data?



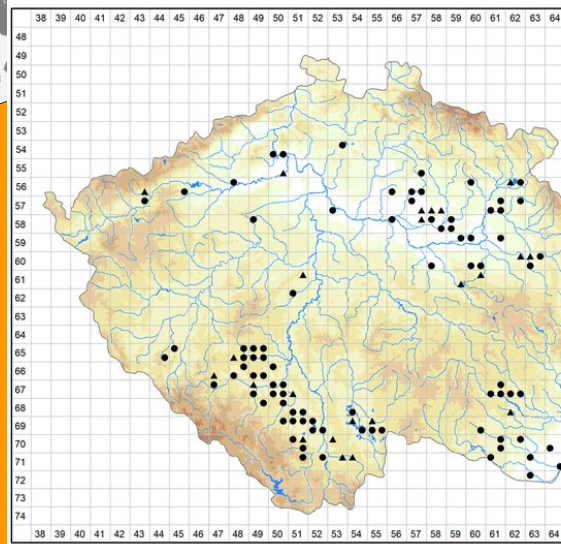
0.062 ha

1 ha

50 ha



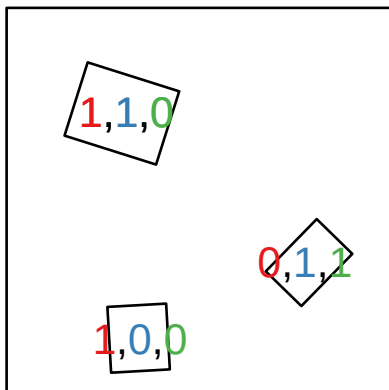
Nature reserve



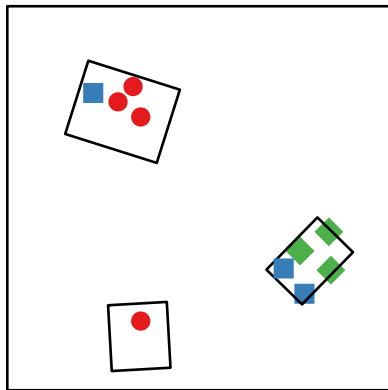
Country

Can we integrate the data?

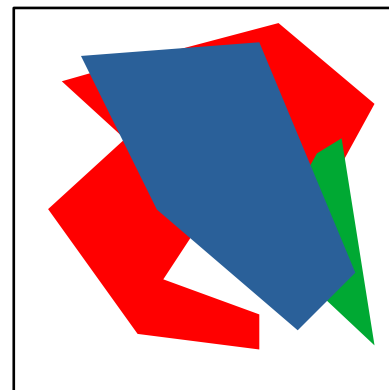
Incidences at sites



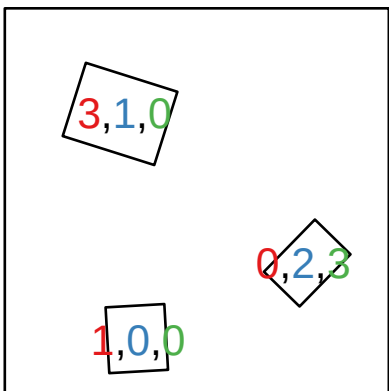
Fully mapped sites



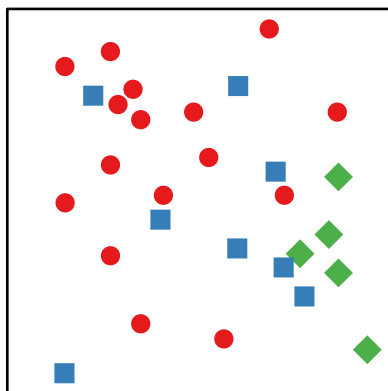
Range maps



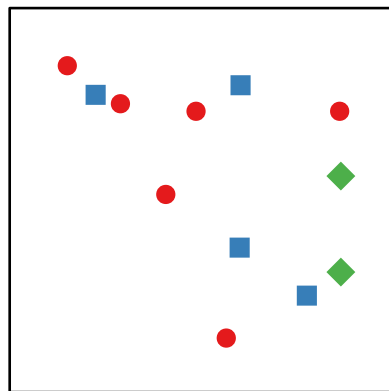
Abundances at sites



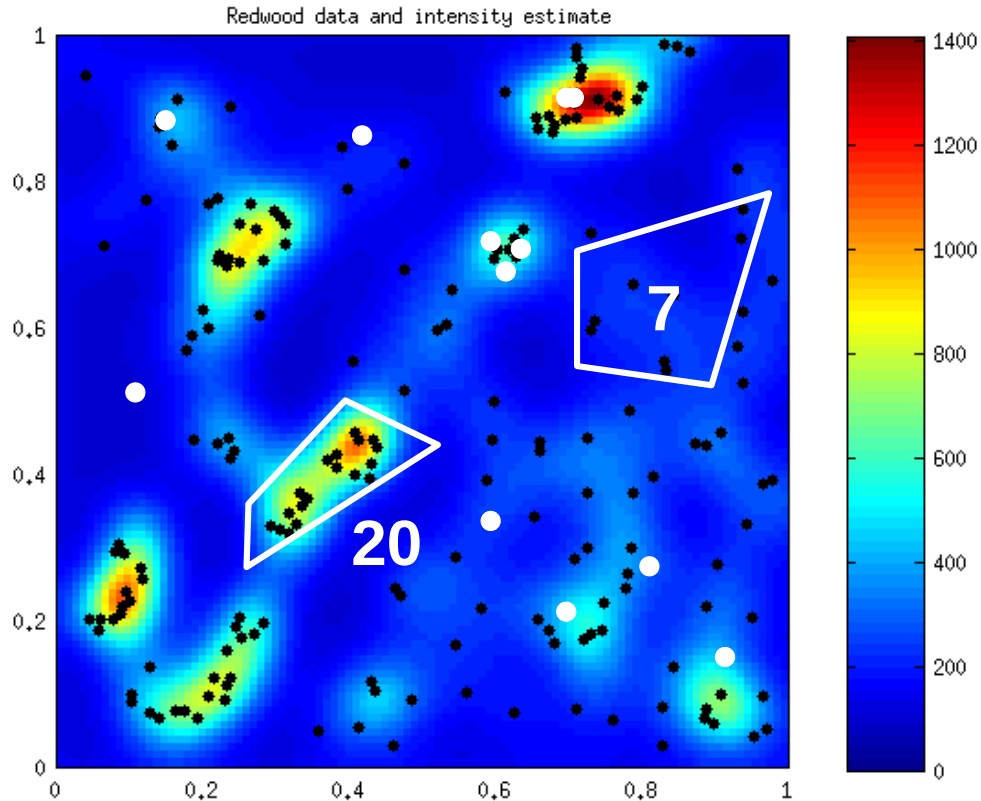
Real world



Point observations

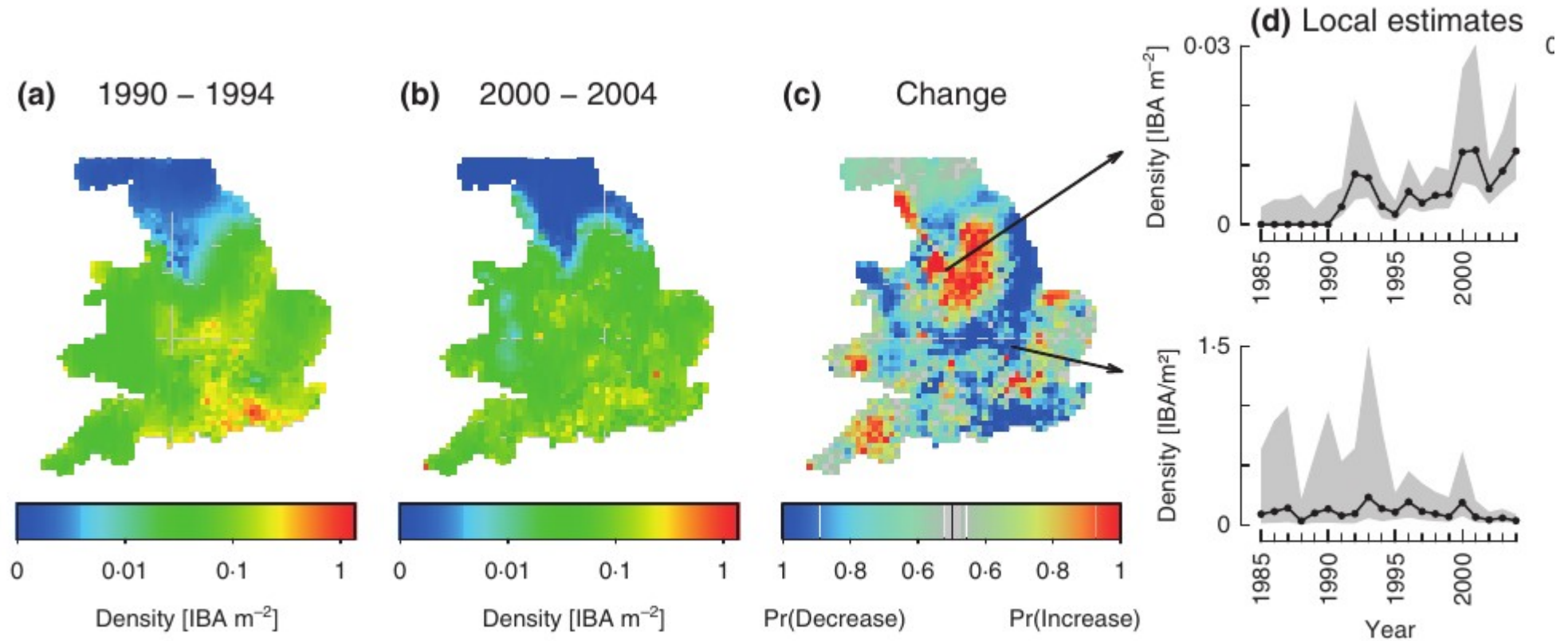


Poisson point process as a common denominator



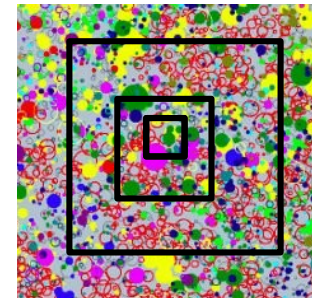
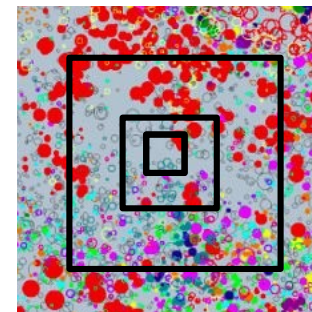
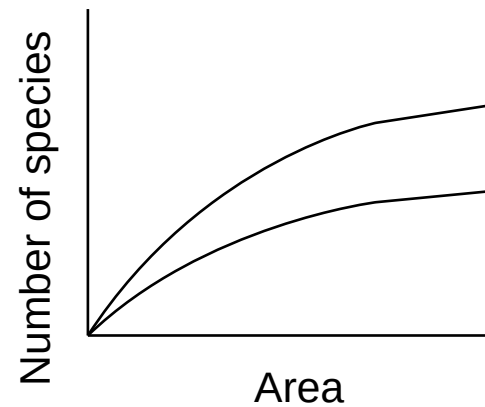
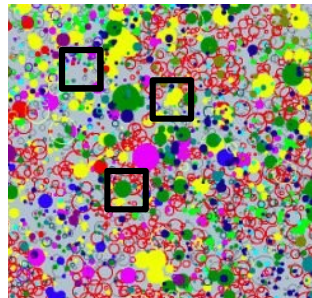
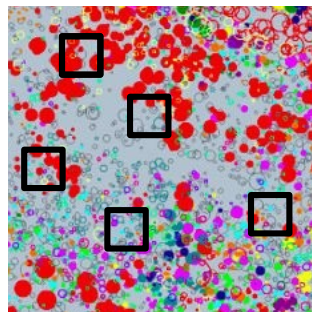
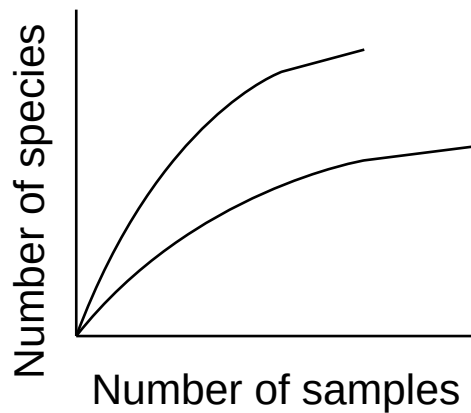
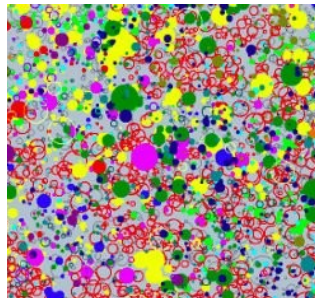
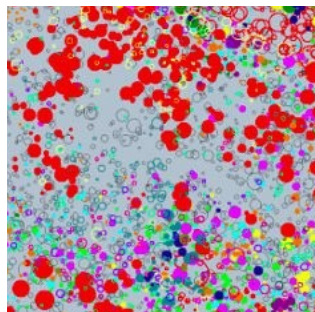
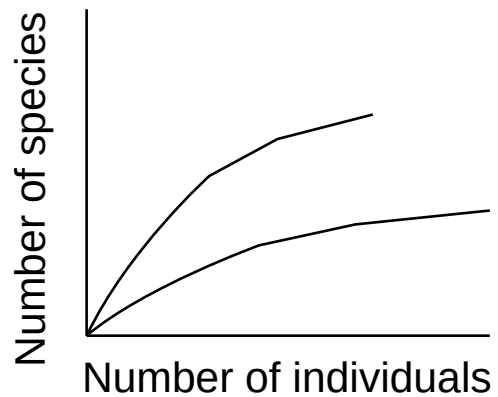
- MaxEnt
- Poisson GLM, cloglog regression
- Occupancy models, hierarchical models
- Geostatistics
- Point pattern analysis

Atlas data + point observations → population trends



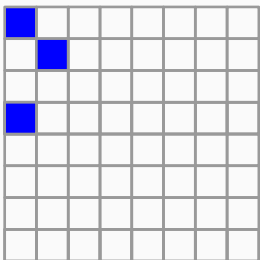
Scaling

Comparing surveys with varying sampling effort

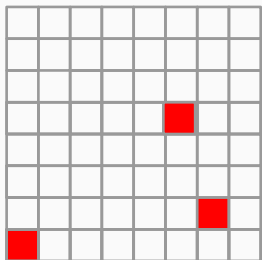


Occupancy-area and species-area relationships

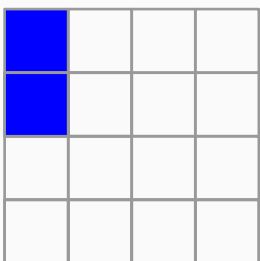
$$P = 3/64 = 0.05$$



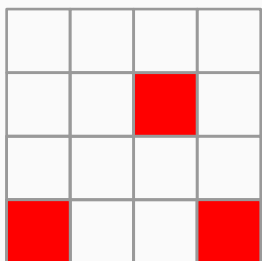
$$P = 3/64 = 0.05$$



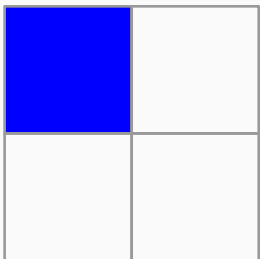
$$P = 2/16 = 0.13$$



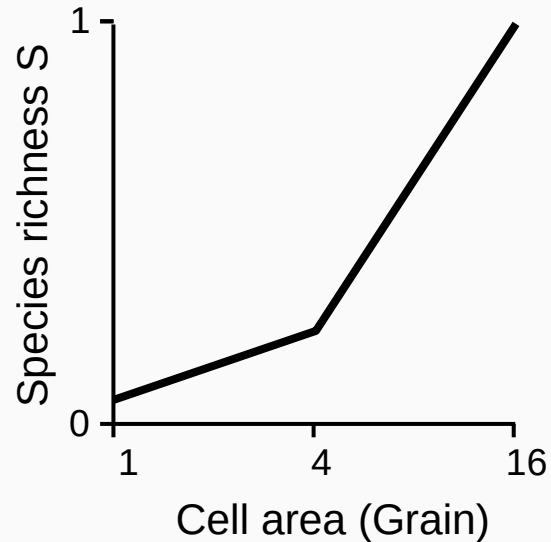
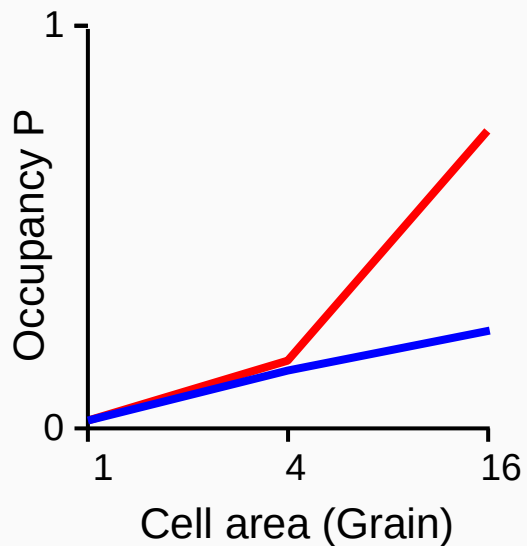
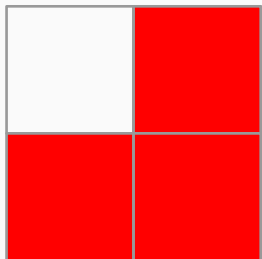
$$P = 3/16 = 0.19$$



$$P = 1/4 = 0.25$$



$$P = 3/4 = 0.75$$



- Occupancy-area relationship (He & Condit 2007)
- Scale-area relationship (Kunin 1998)
- Range-area relationship (Harte et al. 2005)
- Area-area curve (IUCN 2011)
- Scaling pattern of occupancy (Hui et al. 2009)

Downscaling occupancy

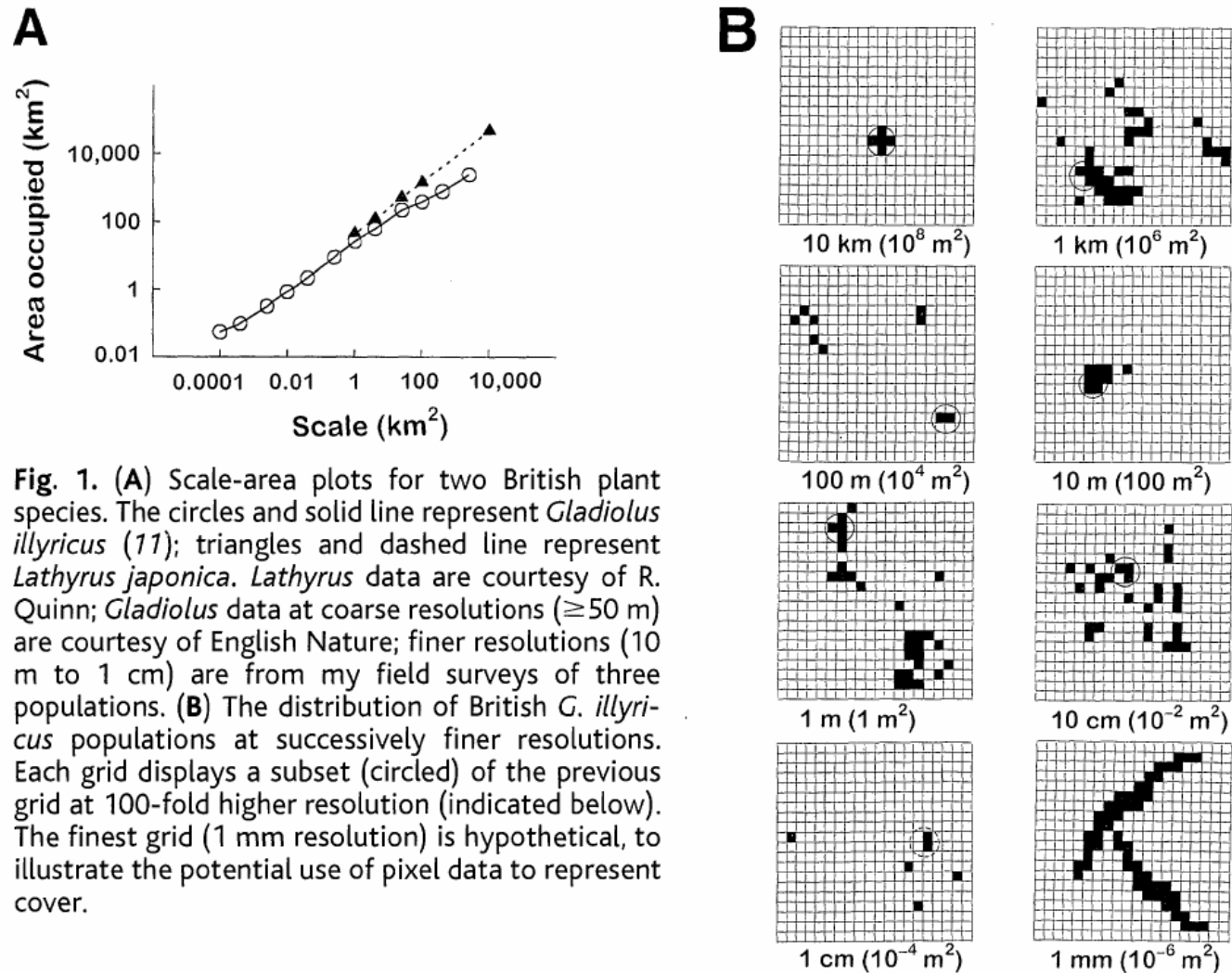
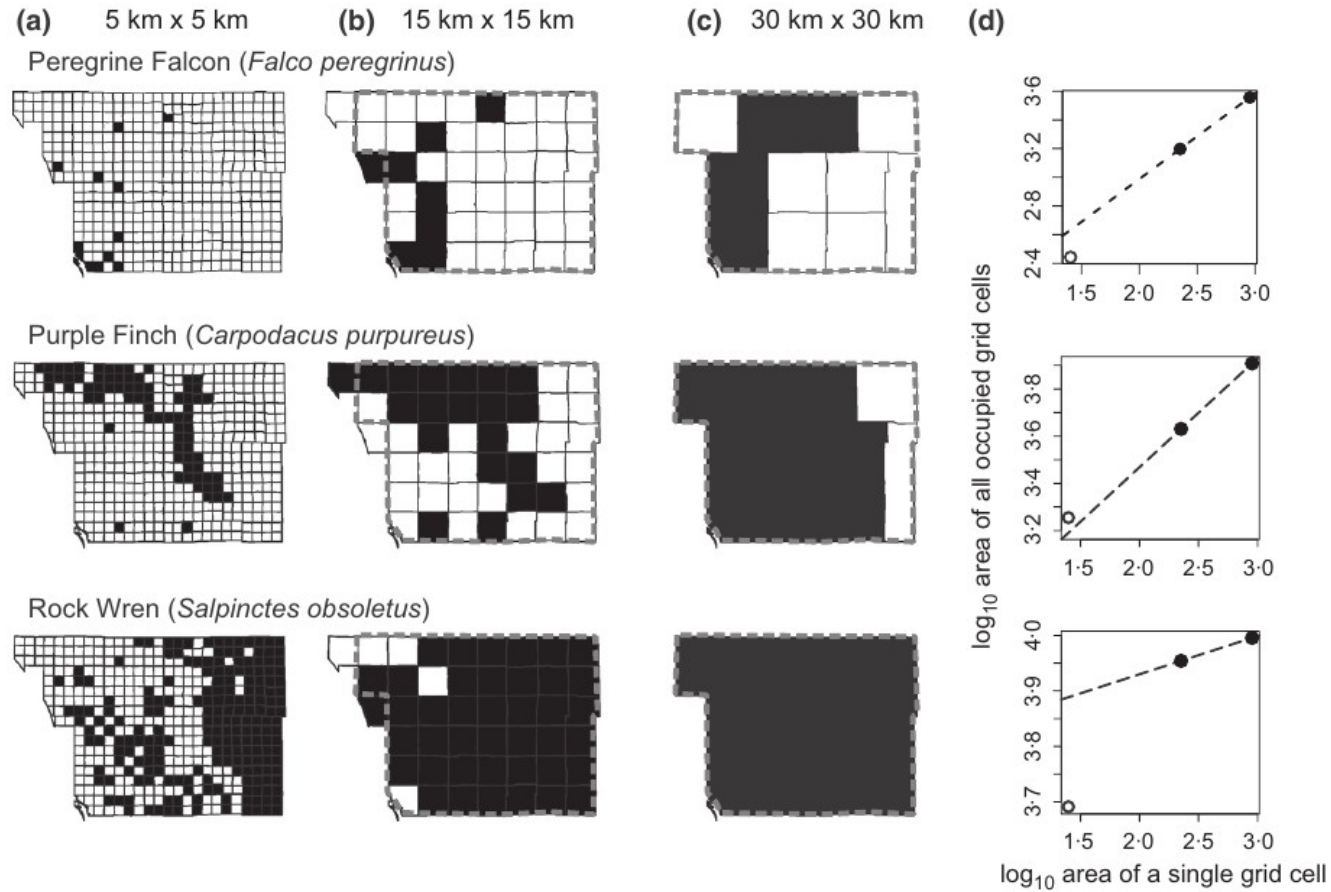
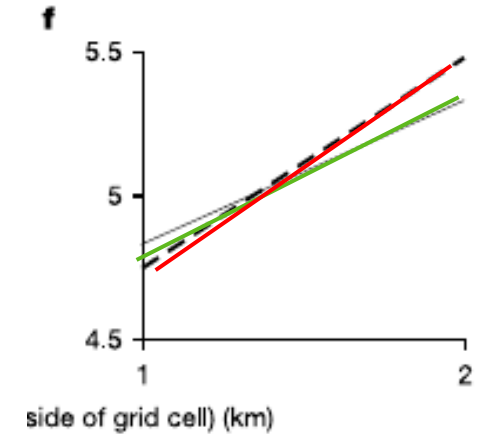
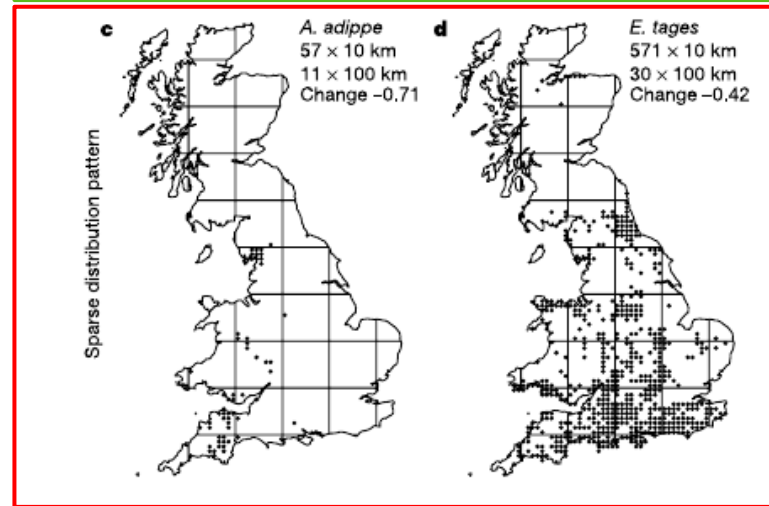
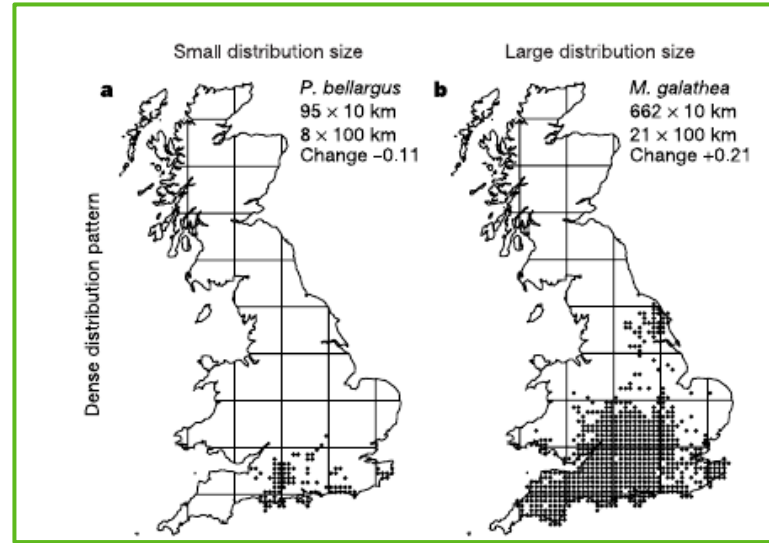
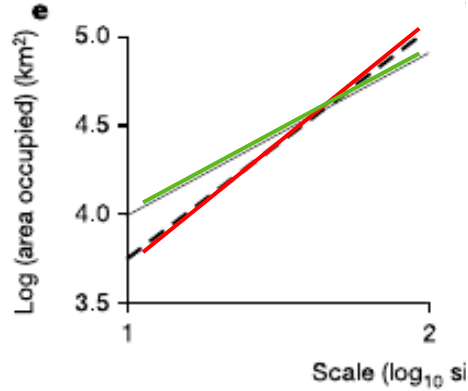


Fig. 1. (A) Scale-area plots for two British plant species. The circles and solid line represent *Gladiolus illyricus* (11); triangles and dashed line represent *Lathyrus japonica*. *Lathyrus* data are courtesy of R. Quinn; *Gladiolus* data at coarse resolutions (≥ 50 m) are courtesy of English Nature; finer resolutions (10 m to 1 cm) are from my field surveys of three populations. **(B)** The distribution of British *G. illyricus* populations at successively finer resolutions. Each grid displays a subset (circled) of the previous grid at 100-fold higher resolution (indicated below). The finest grid (1 mm resolution) is hypothetical, to illustrate the potential use of pixel data to represent cover.

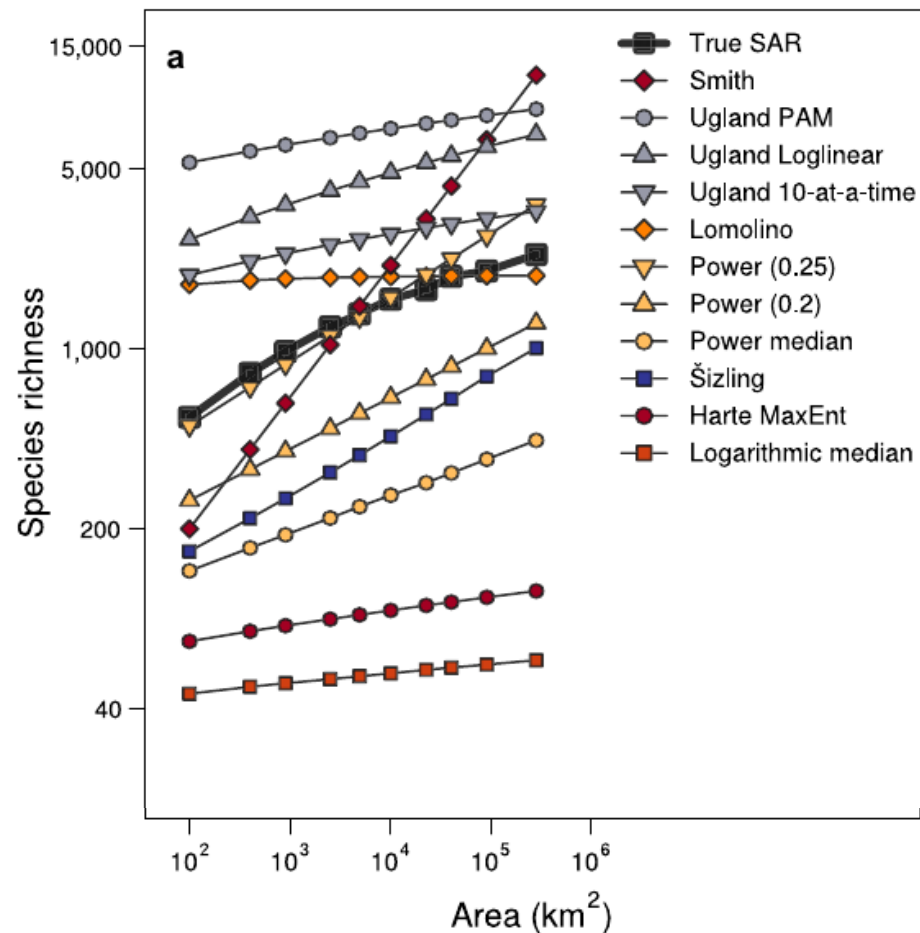
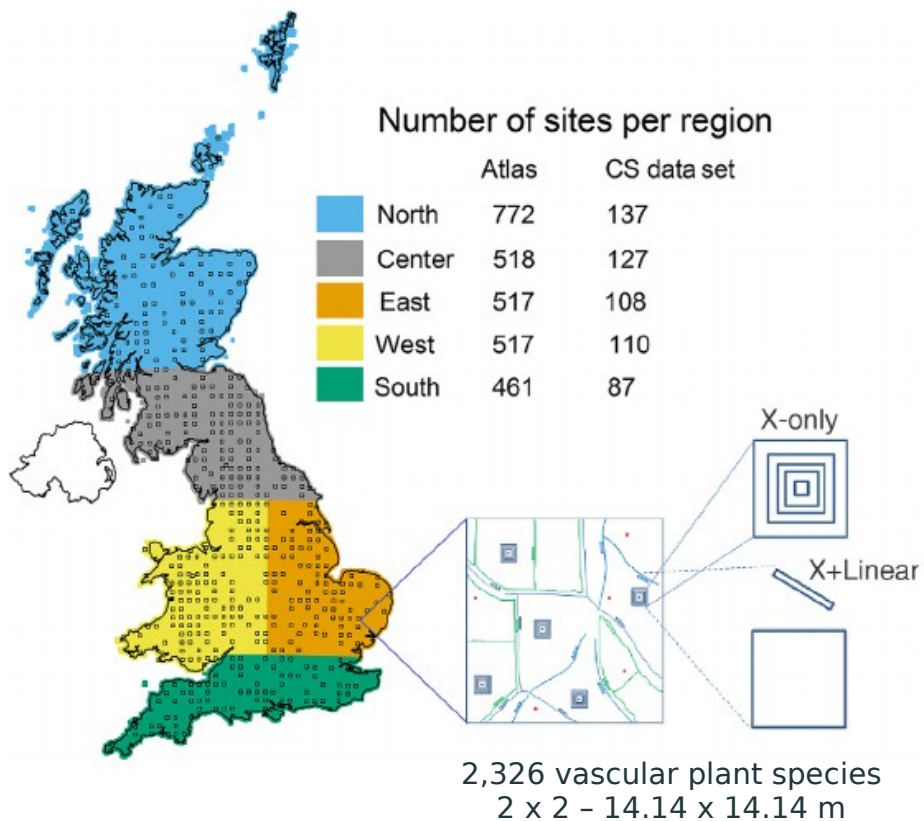
Downscaling occupancy



Inferring change from static patterns

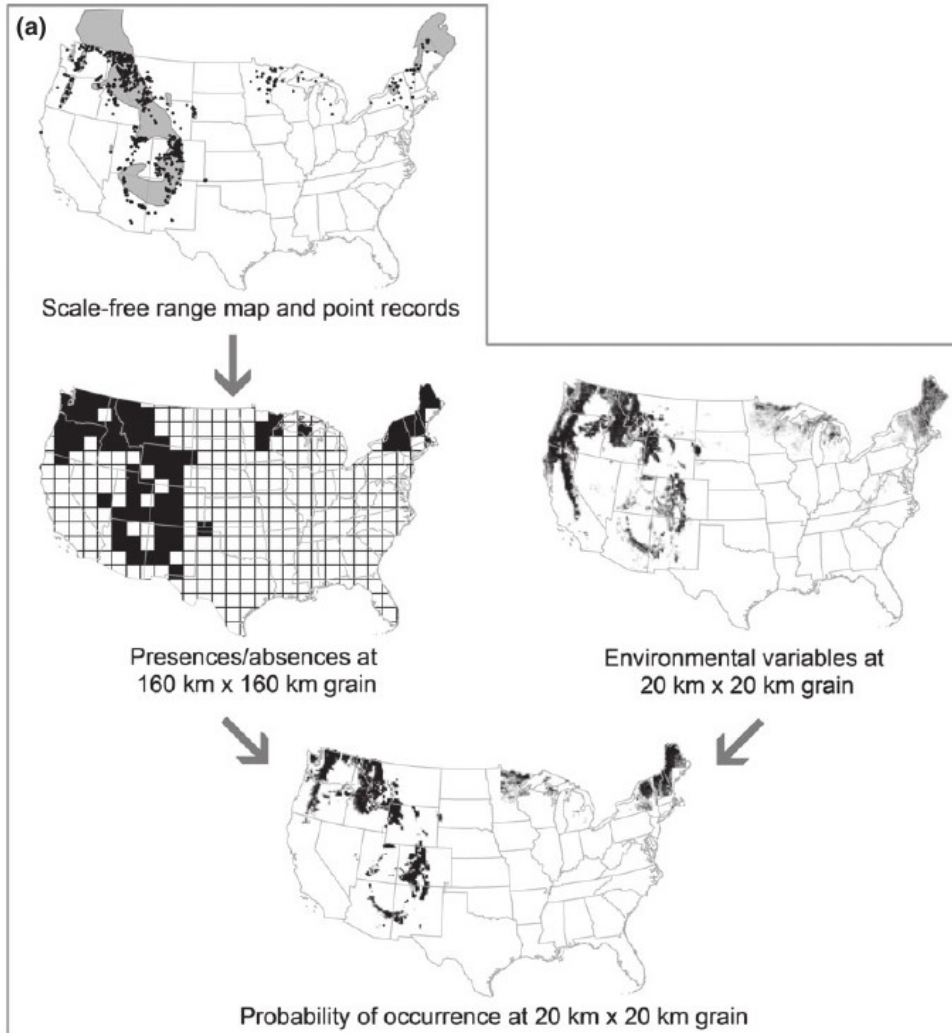


Upscaling richness



Environmental conditions

Downscaling species-distributions



Coarse grain

$P_1 = 0.78$
 $Y_1 = 1$
 (species present)

$P_2 = 0.18$
 $Y_2 = 0$
 (species absent)

$$P_i = 1 - \prod_{j=1}^m (1 - p_{ij})$$

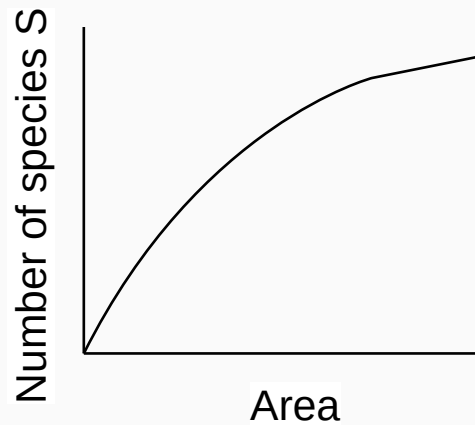
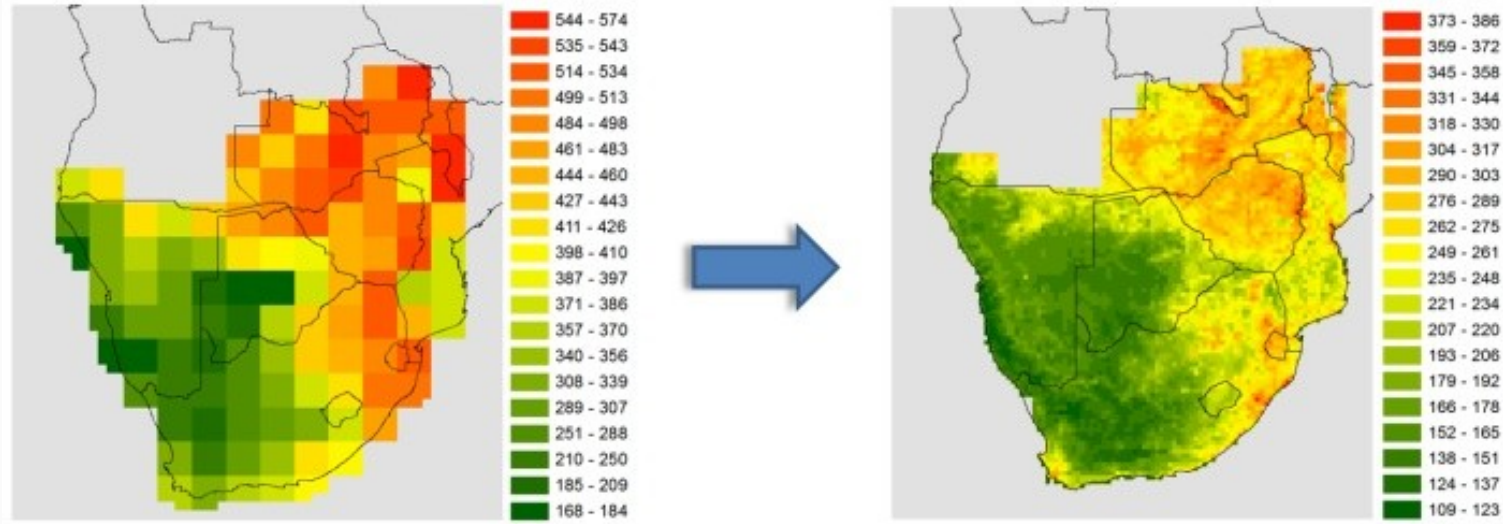
$$Y_i \sim \text{Bernoulli}(P_i)$$

Fine grain

$p_{1,1} = 0.5$	$p_{1,2} = 0.01$	$p_{1,3} = 0.4$	$p_{2,1} = 0.01$	$p_{2,2} = 0.01$	$p_{2,3} = 0.01$
$p_{1,4} = 0.2$	$p_{1,5} = 0.01$	$p_{1,6} = 0.02$	$p_{2,4} = 0.01$	$p_{2,5} = 0.02$	$p_{2,6} = 0.01$
$p_{1,7} = 0.03$	$p_{1,8} = 0.01$	$p_{1,9} = 0.01$	$p_{2,7} = 0.01$	$p_{2,8} = 0.01$	$p_{2,9} = 0.01$

$$P_{ij} = f(\text{env}1_{ij}, \text{env}2_{ij})$$

Downscaling spatial patterns of species richness



Scale as a statistical interaction

What drives global variation of species diversity?



Connecticut, USA



Czech Republic, Europe

Region effects, “historical” effects, dispersal limitation



Ricklefs & Schluter (1993); Map from NASA (public domain)

Environmental effects

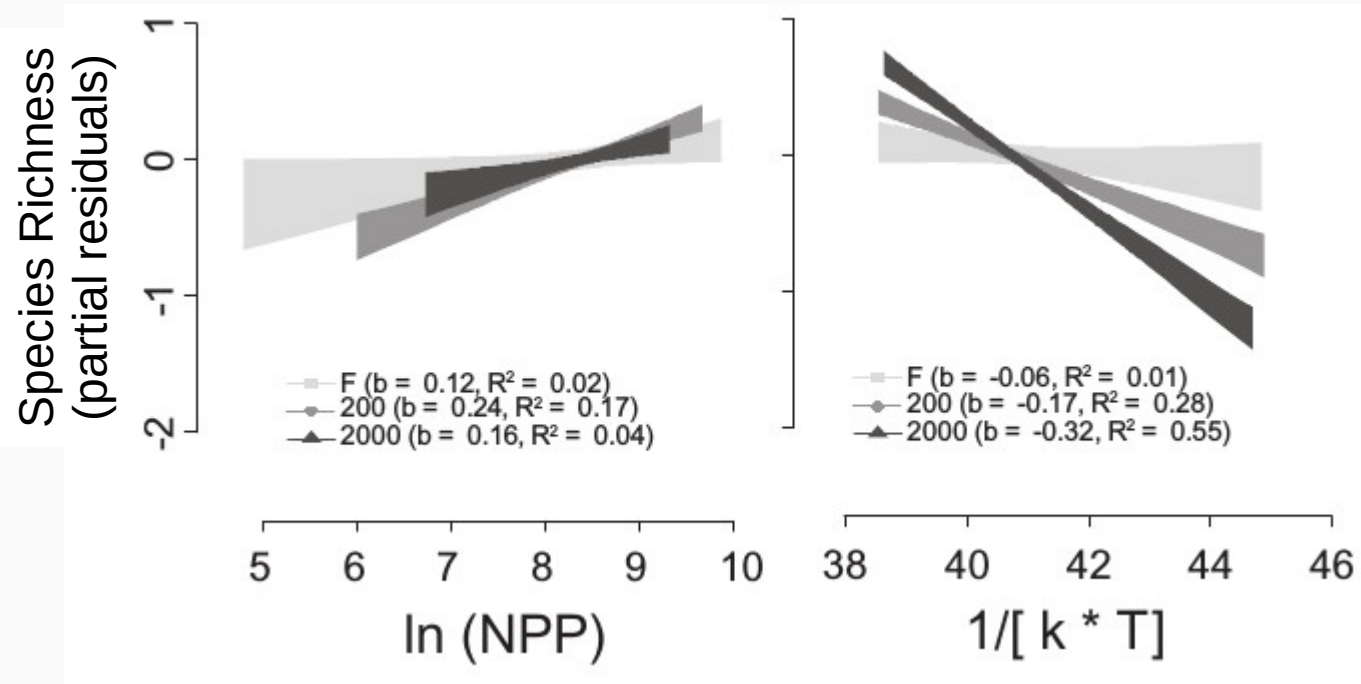
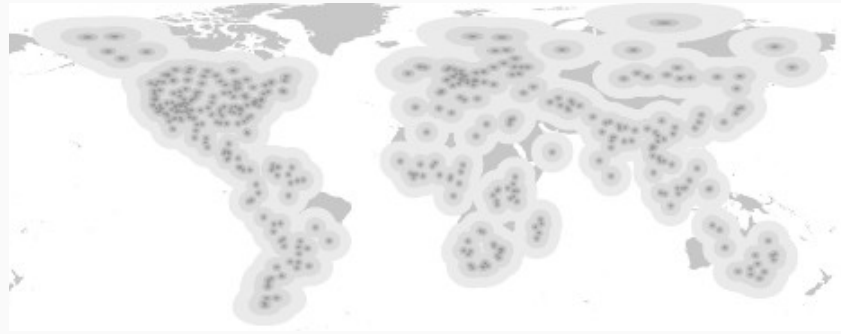
Habitat heterogeneity

Climatic stability

Temperature

Productivity

Drivers of species diversity are grain dependent

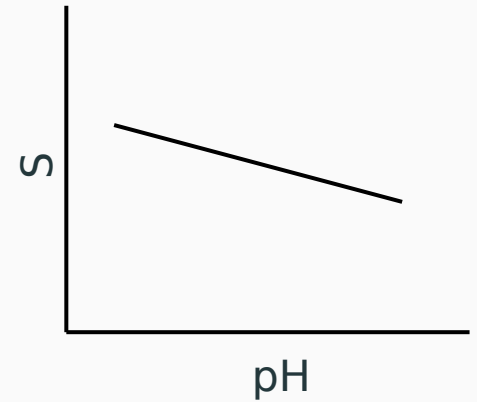
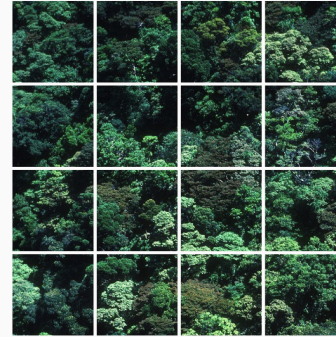
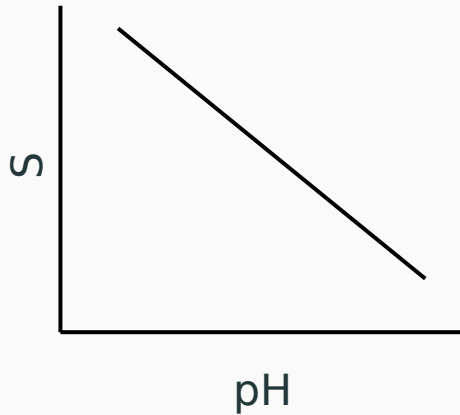
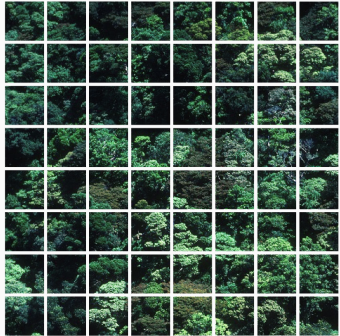


Grain dependence of an effect

$$\log(S) = a + \mathbf{b} \cdot \text{pH}$$

$$\mathbf{b} = c + d \cdot \log(\text{Area})$$

$$\log(S) = a + c \cdot \text{pH} + d \cdot \log(\text{Area}) \cdot \text{pH}$$



Grain dependence of an effect

$$\log(S) = a + c \cdot \text{pH} + d \cdot \log(\text{Area}) \cdot \text{pH} + g \cdot \log(\text{Area})$$



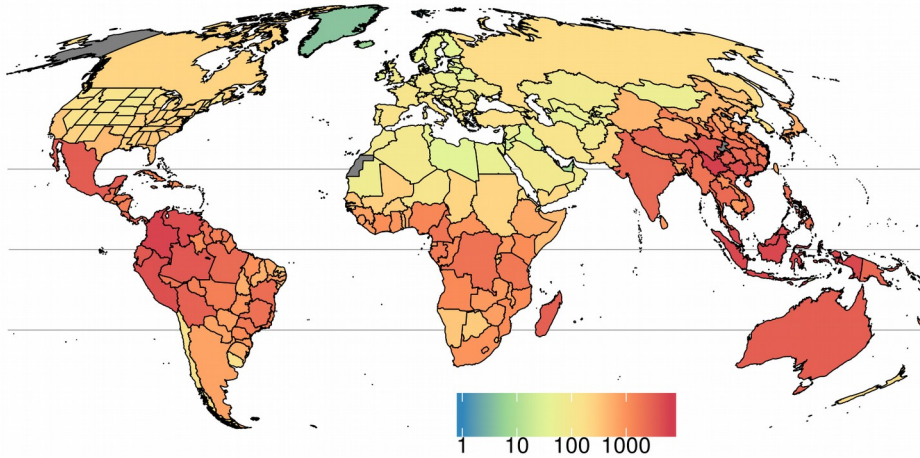
Goals

- Put together global dataset of heterogeneous biodiversity data
- Fit a model estimating drivers and patterns of global species diversity across scales

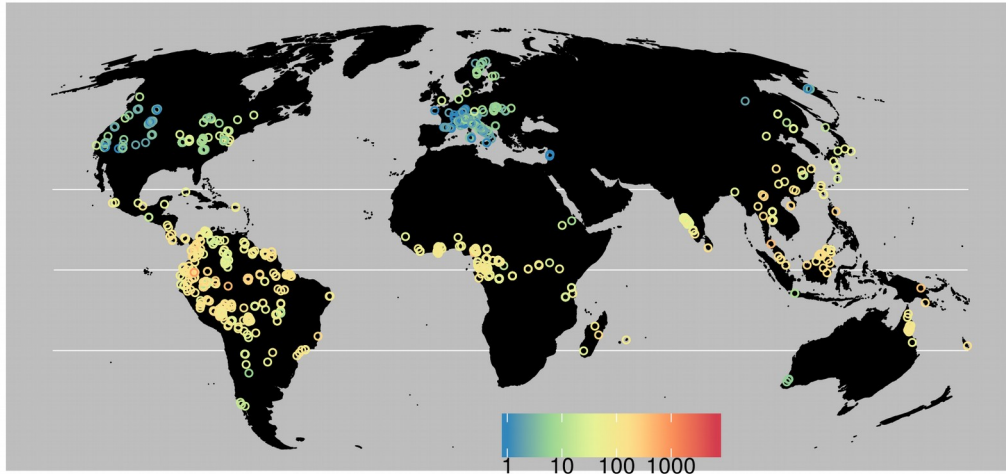


Tree species richness in 286 countries and 1332 plots

a



b



Species richness

Area

Number of trees

Minimum DBH measured

Geographic coordinates

Biogeographic region

Topographic heterogeneity

Insularity

Primary productivity

Annual temperature

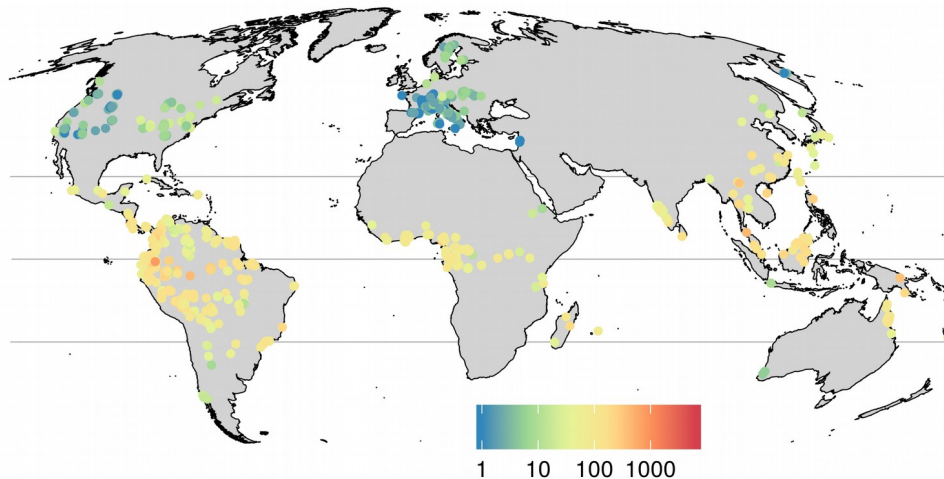
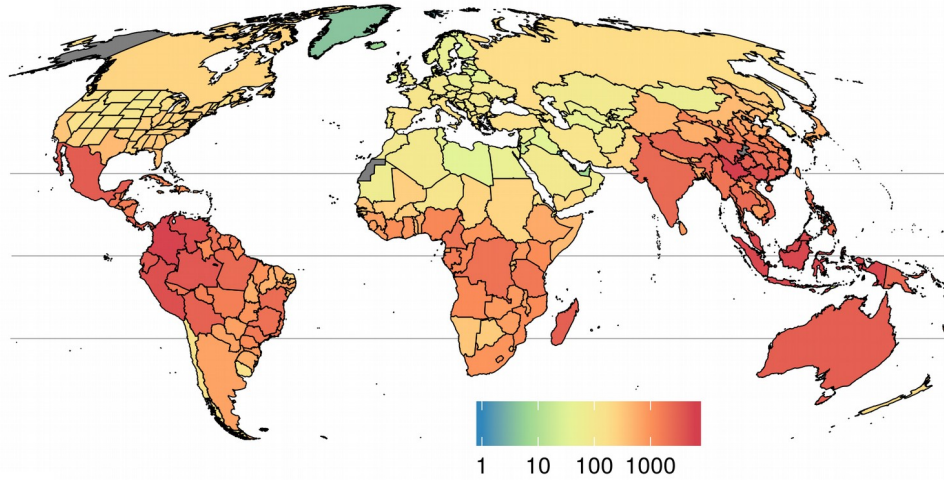
Isothermality

Precipitation in driest quarter

Precipitation seasonality

Keil & Chase (2018) *BioRxiv*

Models



Generalized Additive Models (GAM) with Negative Binomial errors

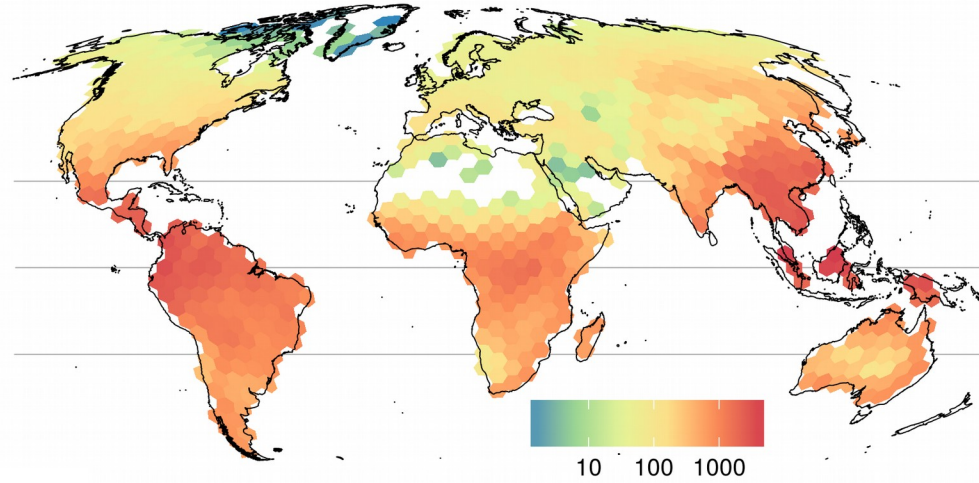
Smooth spatial term

All predictors interact with area

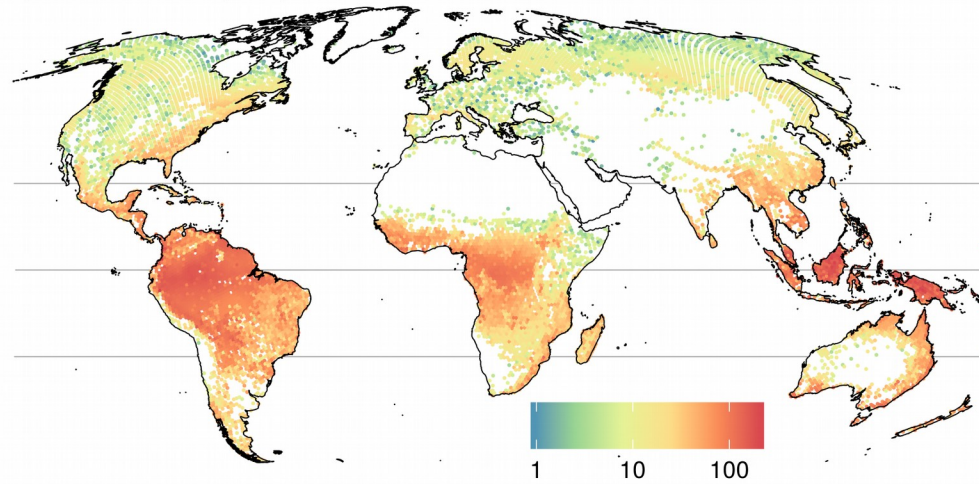
R packages mgcv, brms (interface to STAN)

Predicted species richness

\bar{S}_{hex} (richness in 209,903 km² hexagons)

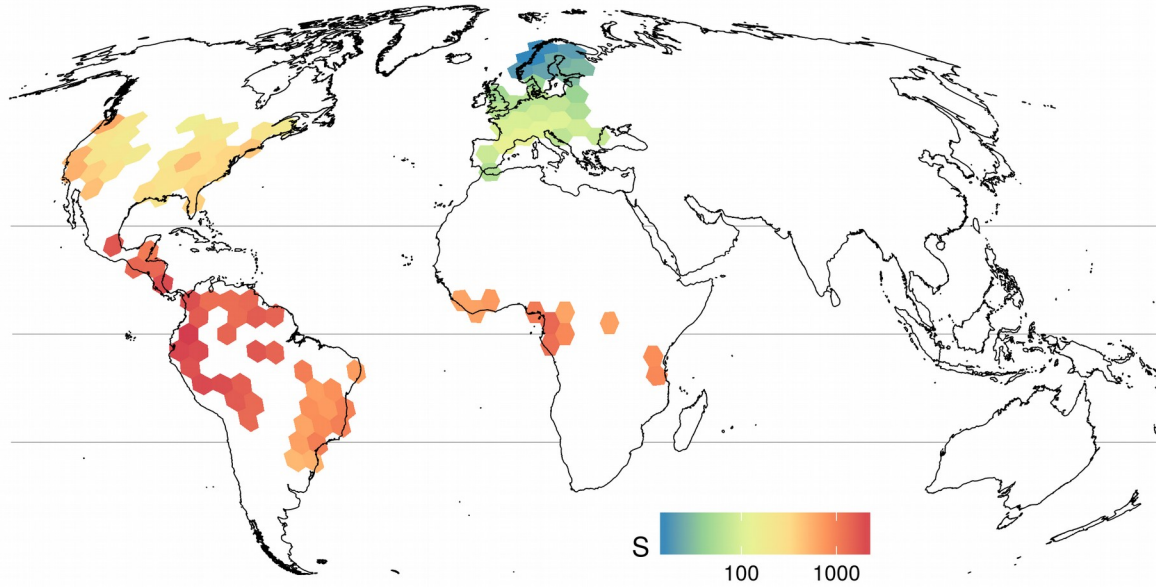


S_{plot} (richness in 1 ha plots)

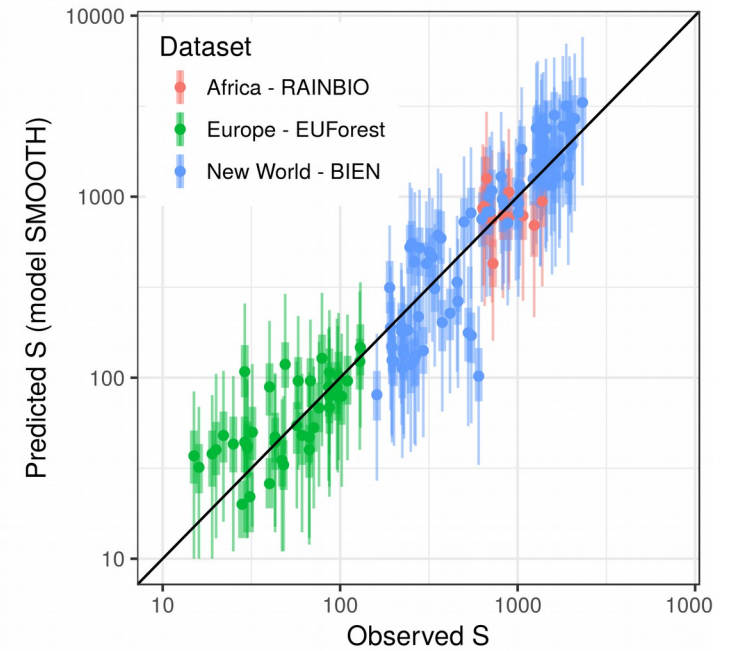


External validation

a

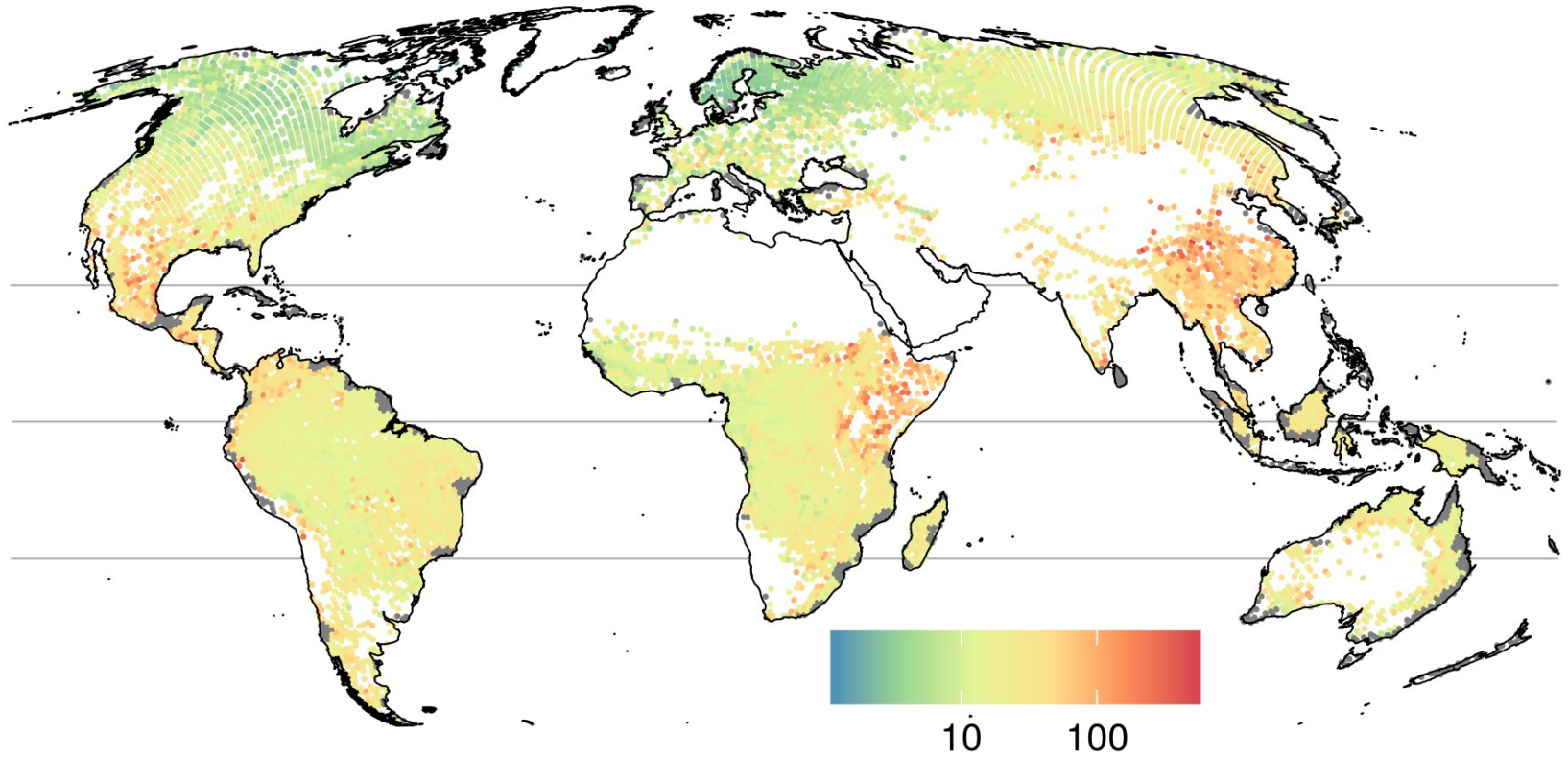


b



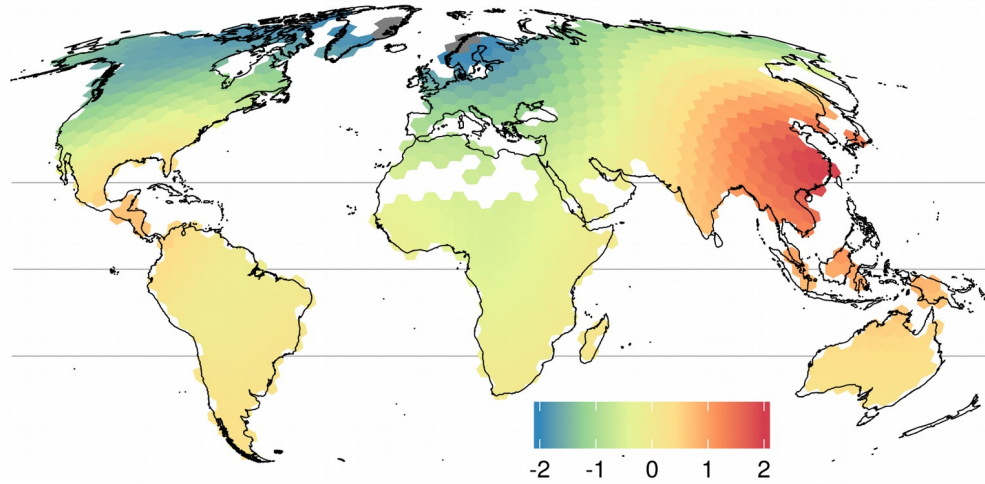
Predicted beta diversity

$$\beta = S_{\text{hex}}/S_{\text{plot}}$$

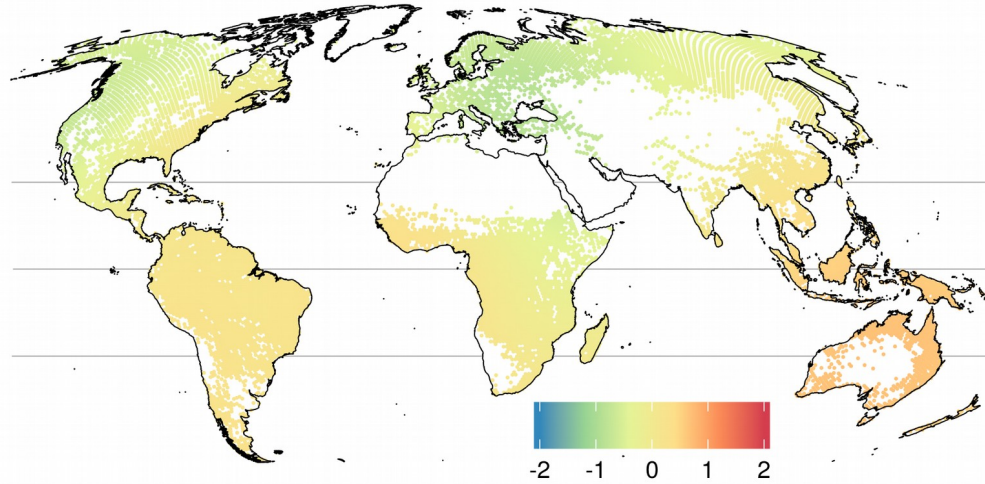


Grain-dependent effects of regions

RE_{hex} (region effects in 209,903 km² hexagons)

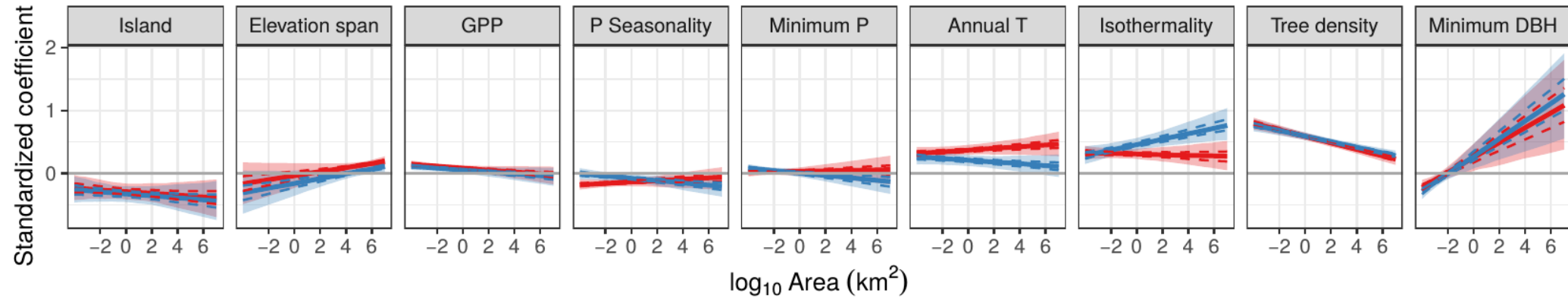


RE_{plot} (region effects in 1 ha plots)

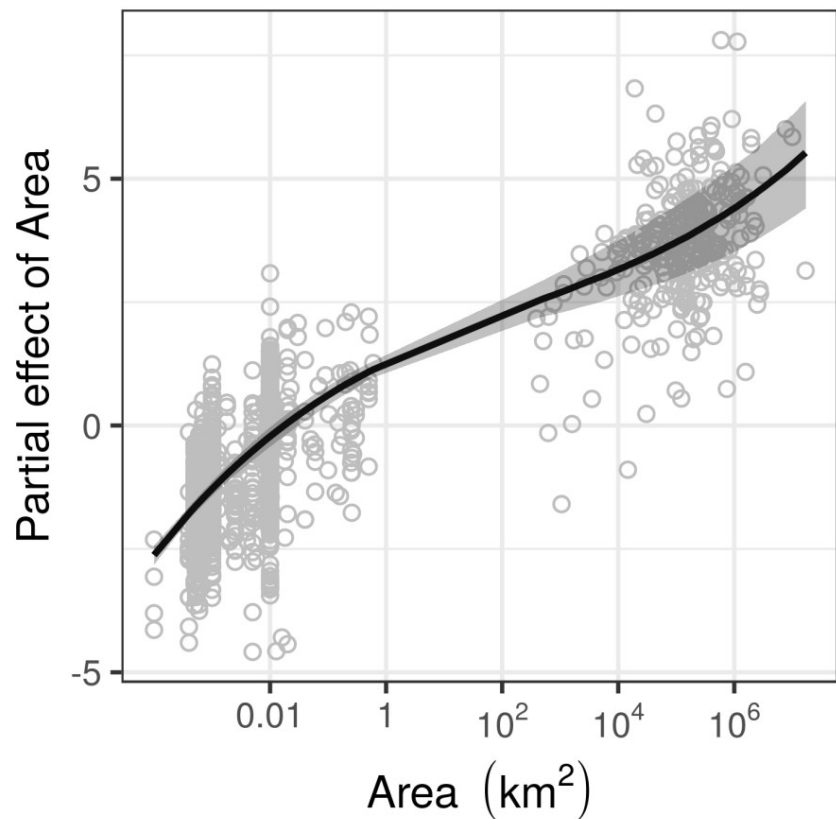


Grain-dependent effects of environment

Model
— REALM
— SMOOTH



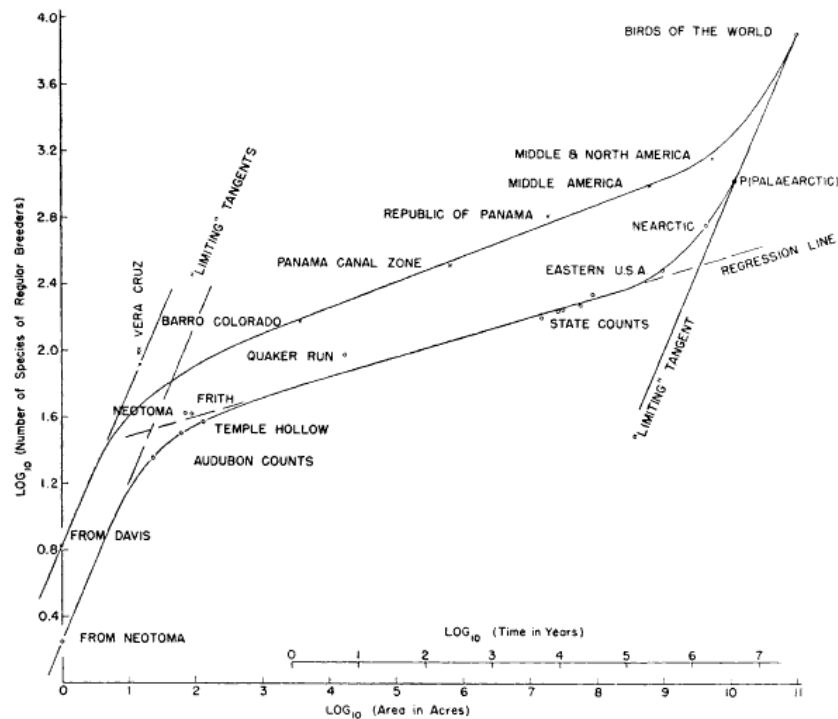
Tri-phasic species-area relationship



616

F. W. PRESTON

Ecology, Vol. 41, No. 4



Preston (1960) *Ecology*
Storch, Keil & Jetz (2012) *Nature*

Summary

Summary

- Various types of biodiversity data reflect **one thing**: Point occurrences of individuals in space, which can be seen as a point pattern.
- Abstraction to **point pattern** unifies techniques such as MaxEnt, Poisson regression, or geostatistics, and enables integration of data, patterns, and hypotheses.
- **Scaling relationships**, such as species-area or occupancy-area are a key to both data integration, and understanding of biodiversity.
- Scaling relationships can be used for **upscaling** or **downscaling**.
- Scale-dependence of “effects” can be modeled as a statistical **interaction with area**.



DFG

Deutsche
Forschungsgemeinschaft

German Centre for Integrative Biodiversity Research (iDiv)

Jonathan Chase, Walter Jetz, David Storch, Bill Kunin
Arnost Sizling, Irena Simova, Jonathan Belmaker,
Hugh Sturrock, Adam M. Wilson, Nick Isaac, Bob O'Hara,
Dylan Craven, Shane Blowes

Thank you! Questions?