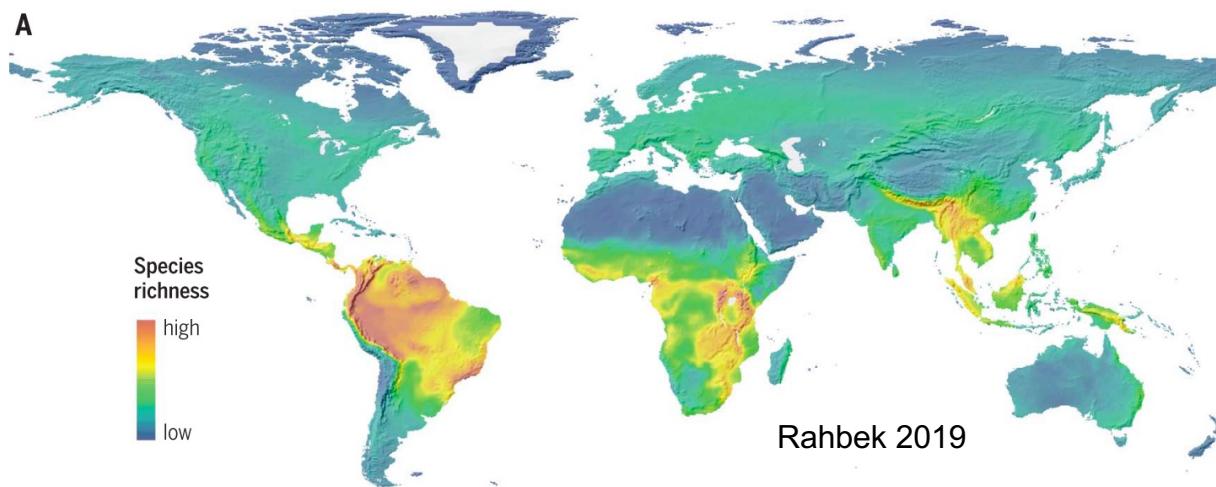


Using graph-based metrics to assess the effect of landscape topography on diversification

Victor Boussange, Loïc Pellissier, and Yaquan Chang

ECBC Amsterdam 2021

Large-scale geographical patterns of species diversity



Mountains represent **25 % of land area**, but **85% of the world's species** of amphibians, birds and mammals, many entirely restricted to mountains (Rahbek 2019)



Topological constraints

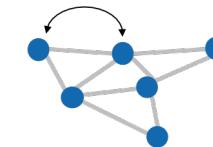


Habitat heterogeneity

Underlying mechanisms?



First principles modelling approach



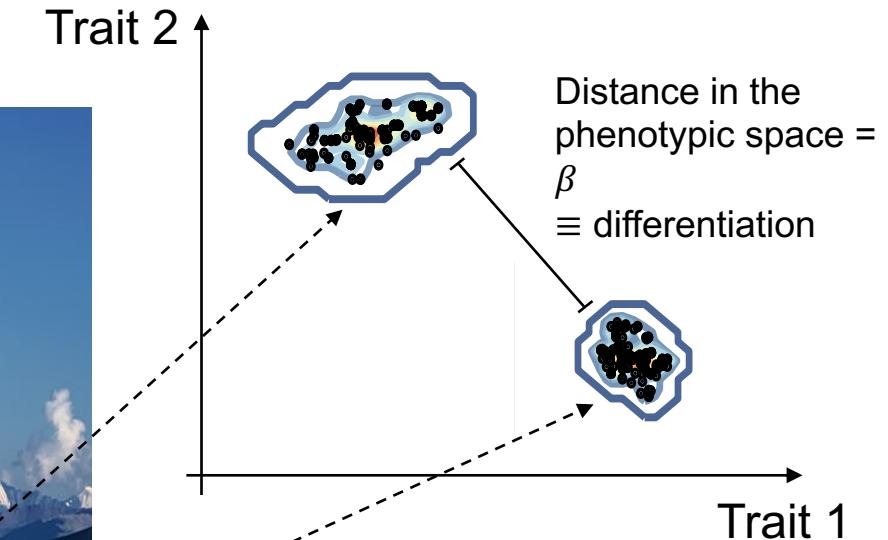
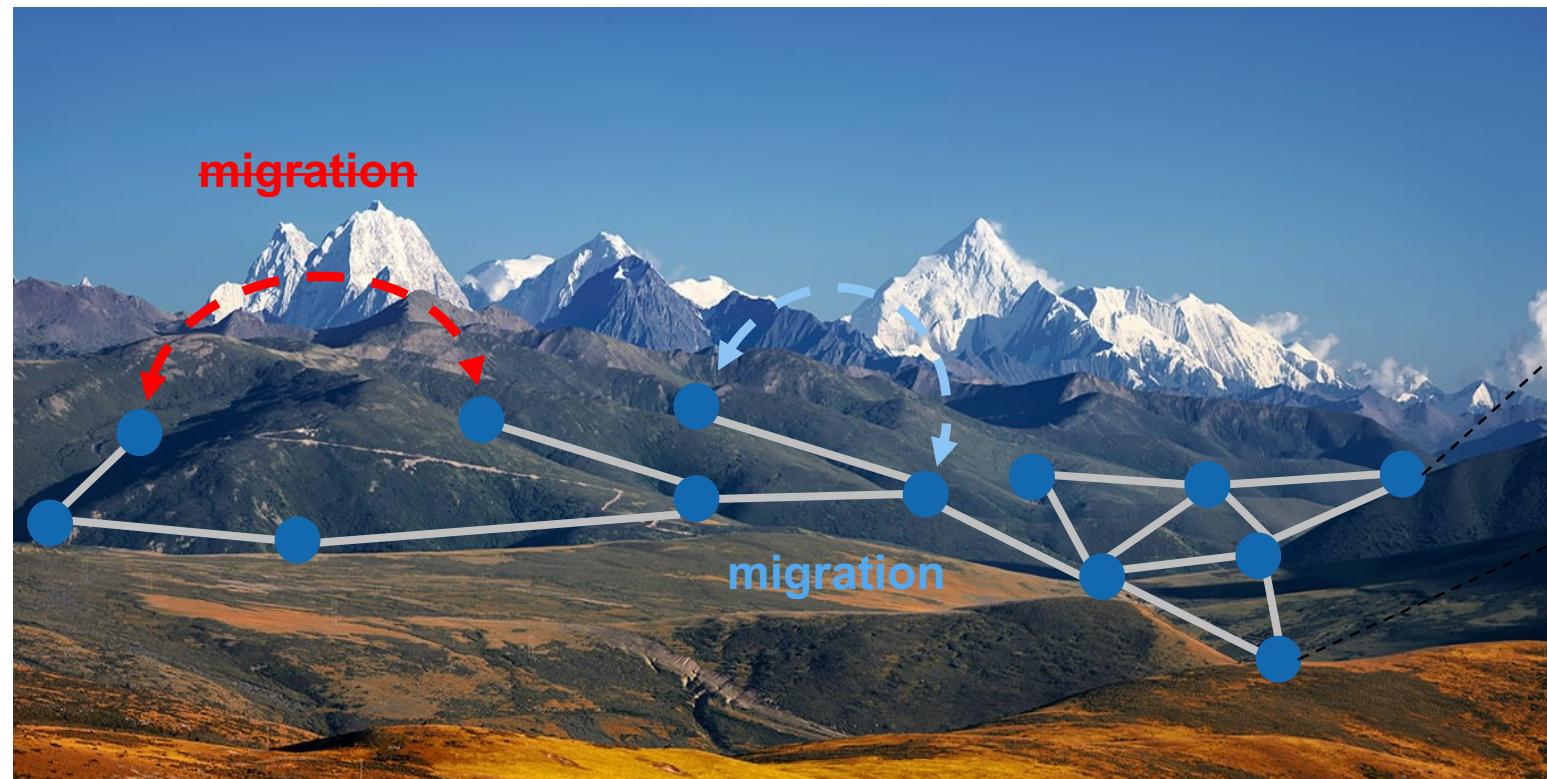
Graph representation of the landscape

$$m\partial_t v = \sum_i F_i$$

Eco-evolutionary individual based model

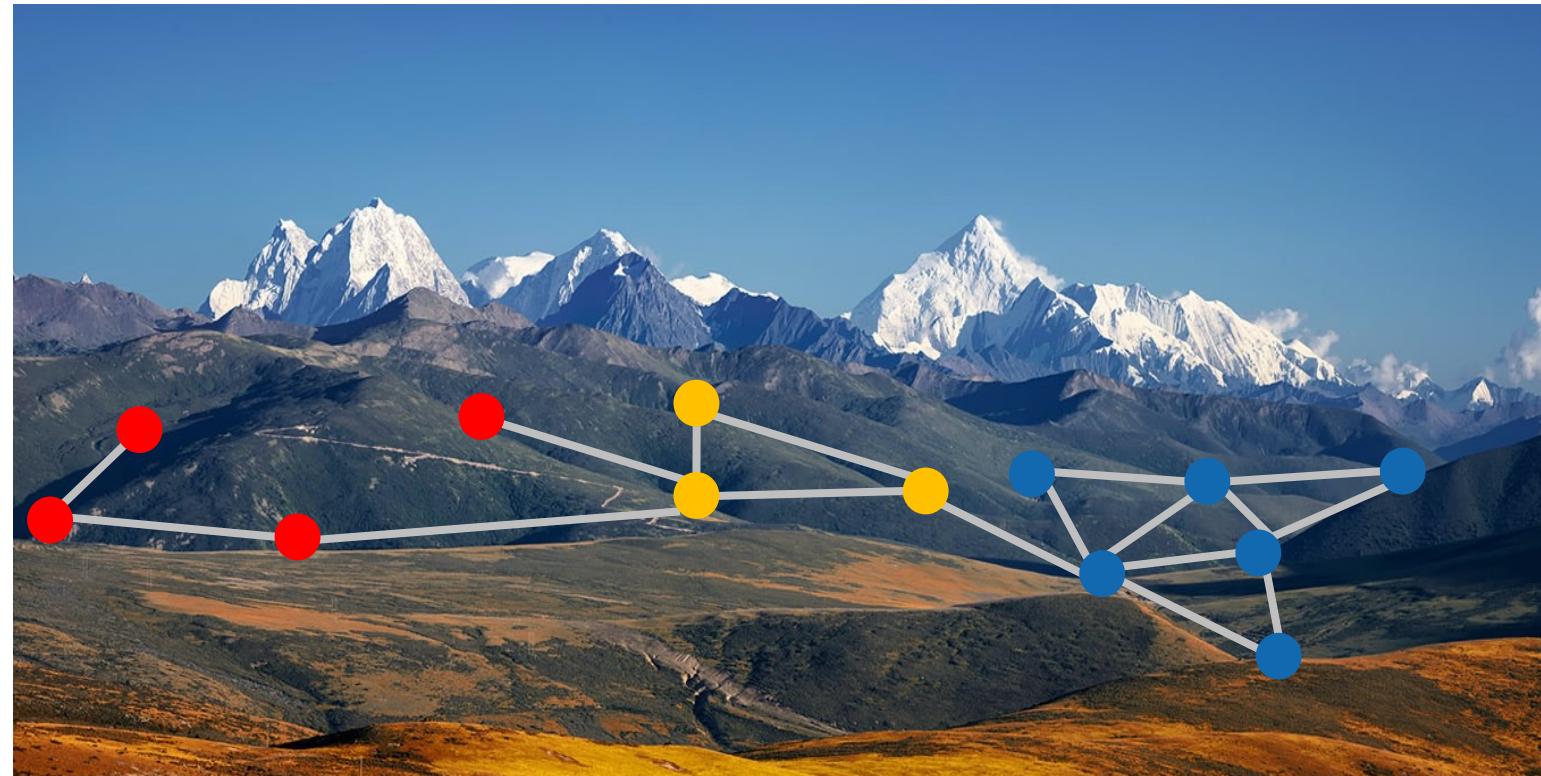
Capturing complex connectivity patterns with graphs

Graphs, to capture dispersal patterns
→ Geometric constraints



Capturing habitat heterogeneity with graphs

Graphs, to capture
environmental heterogeneity



- Habitat 1
- Habitat 2
- Habitat 3

Graph-based eco-evolutionary model

Numerical simulations



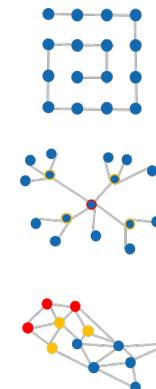
&

Analytical results

$$\begin{aligned}
 L\phi(\nu_t^{(i)}) = & \int_{\mathcal{X}} \left\{ b_i(\mathbf{x})(1-\mu)(1-m)(\phi(\nu_t^{(i)} + \delta_{\mathbf{x}}) - \phi(\nu_t^{(i)})) \right\} \nu_t^{(i)}(d\mathbf{x}) \\
 & + \int_{\mathcal{X}} \left\{ \mu(1-m) \int_{\mathcal{X}} b_i(y)(\phi(\nu_t^{(i)} + \delta_z) - \phi(\nu_t^{(i)})) \mathcal{M}(\mathbf{x}, y) dy \right\} \nu_t^{(i)}(d\mathbf{x}) \\
 & + \iint_{\mathcal{X}} \left\{ \frac{1}{K} (\phi(\nu_t^{(i)} - \delta_{\mathbf{x}}) - \phi(\nu_t^{(i)})) \nu_t^{(i)}(dy) \nu_t^{(i)}(dx) \right\} \\
 & + \sum_{j \neq i} \frac{a_{i,j}}{d_j} \int_{\mathcal{X}} \mu m \left\{ \int_{\mathcal{X}} b_j(y)(\phi(\nu^{(j)} + \delta_{\mathbf{x}}) - \phi(\nu^{(j)})) \mathcal{M}(\mathbf{x}, y) dy \right\} \nu_t^{(j)}(d\mathbf{x}) \\
 & + \sum_{j \neq i} \frac{a_{i,j}}{d_j} \int_{\mathcal{X}} \left\{ b_j(\mathbf{x})(1-\mu)m(\phi(\nu^{(j)} + \delta_{\mathbf{x}}) - \phi(\nu^{(j)})) \right\} \nu_t^{(j)}(d\mathbf{x}).
 \end{aligned}$$



- Characteristic length
- Heterogeneity in degree
- Environmental assortativity

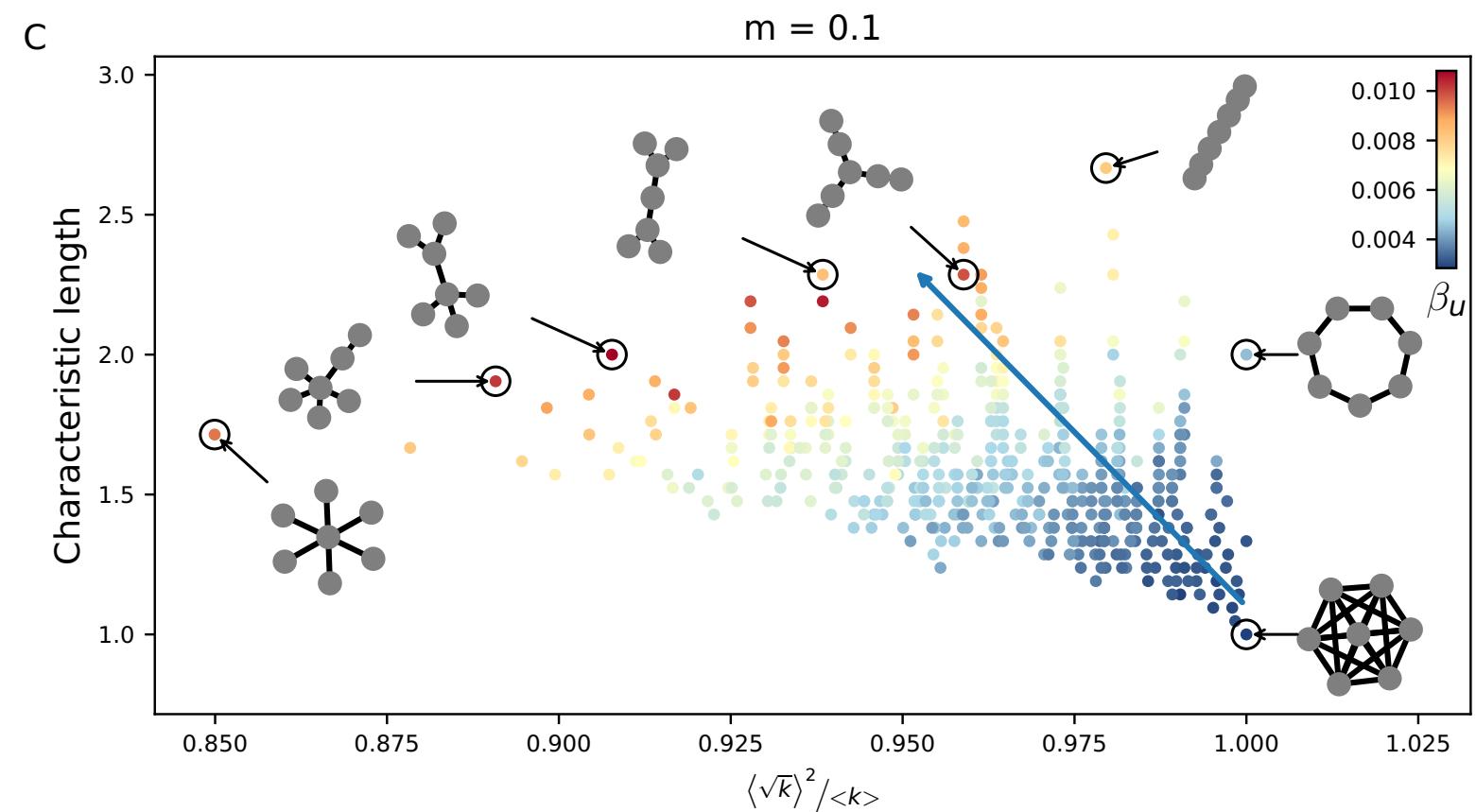
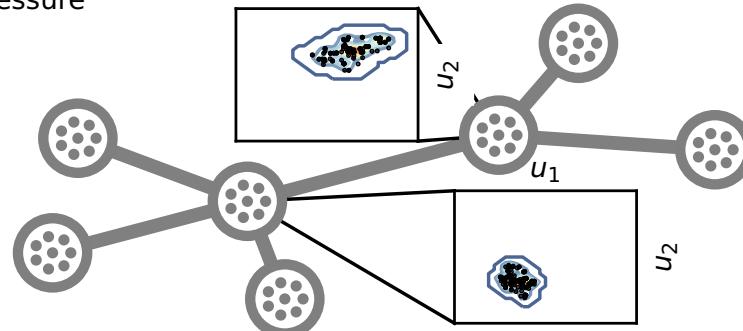


Mechanisms

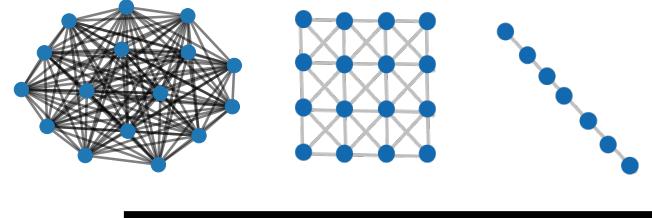


Experiment #1 – Effect of topology on differentiation

Setting (1):
no selection pressure



Characteristic length and heterogeneity in degree explain 88% of diversity variation



Characteristic length
~ landscape dimensionality

average shortest path between all pairs of nodes in the graph

Nodes with relatively high degree



High influx of migrants



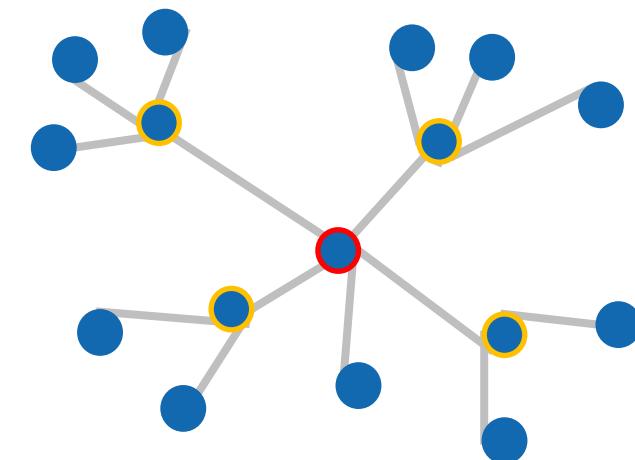
Increased competition



Higher death rate

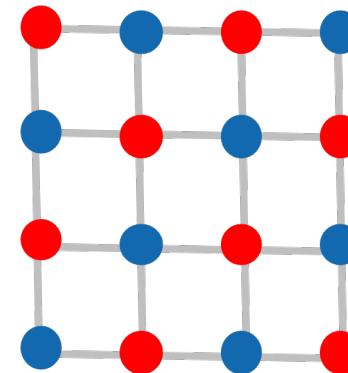
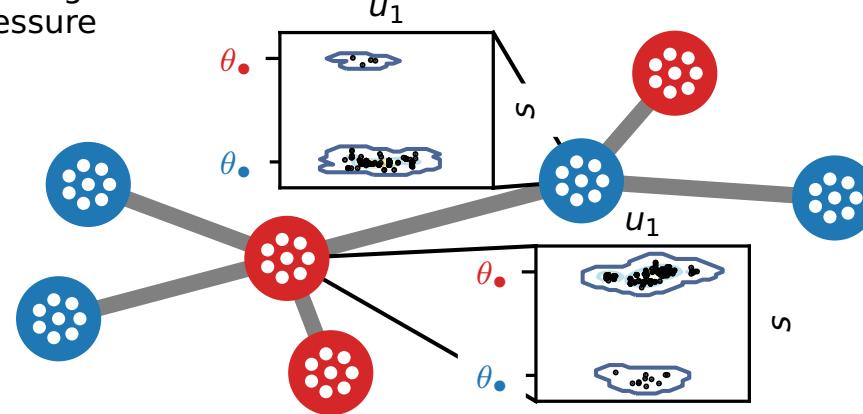


Bottlenecks
Barriers to dispersal

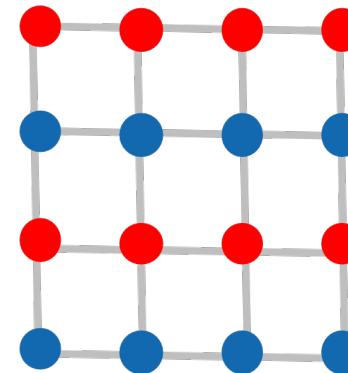


Experiment #2 – Effect of topology on adaptive differentiation

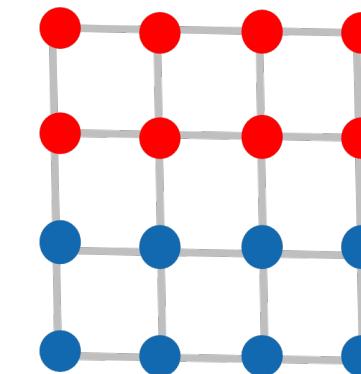
Setting (2):
heterogeneous selection pressure



$$r_\theta \approx -1$$



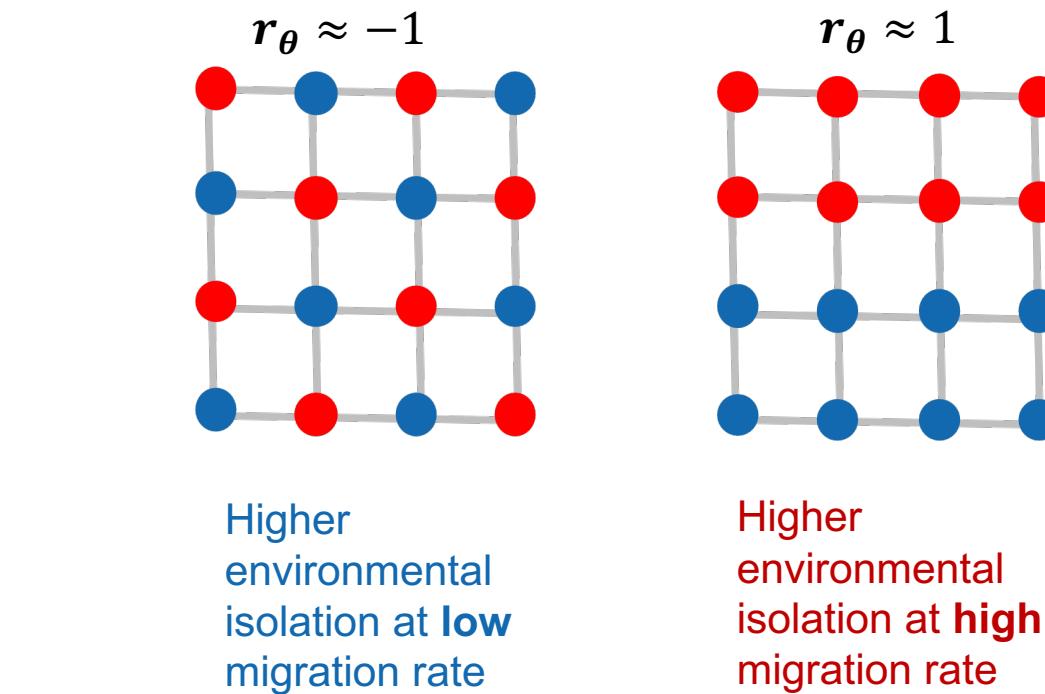
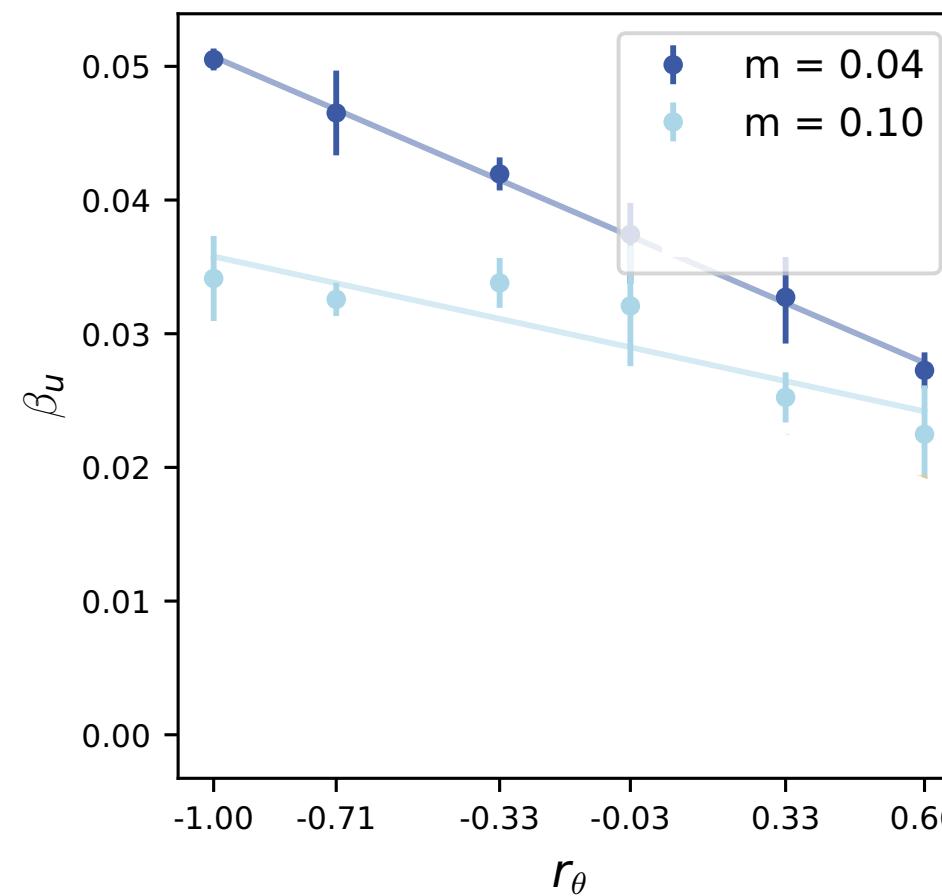
$$r_\theta \approx 0$$



$$r_\theta \approx 1$$

Environmental assortativity $r_\theta \sim$ Environmental spatial autocorrelation

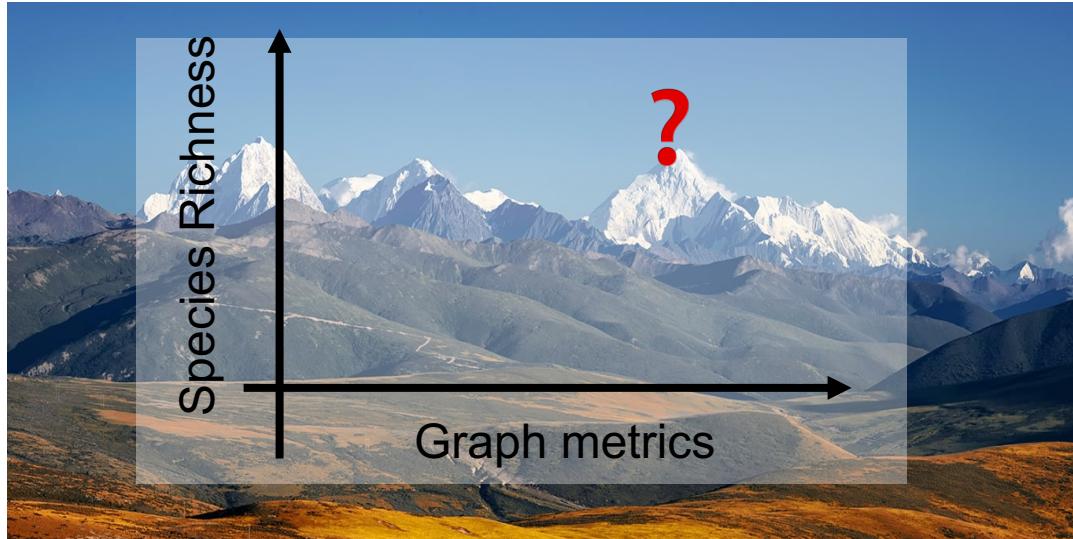
Experiment #2 – Effect of environmental assortativity on neutral differentiation



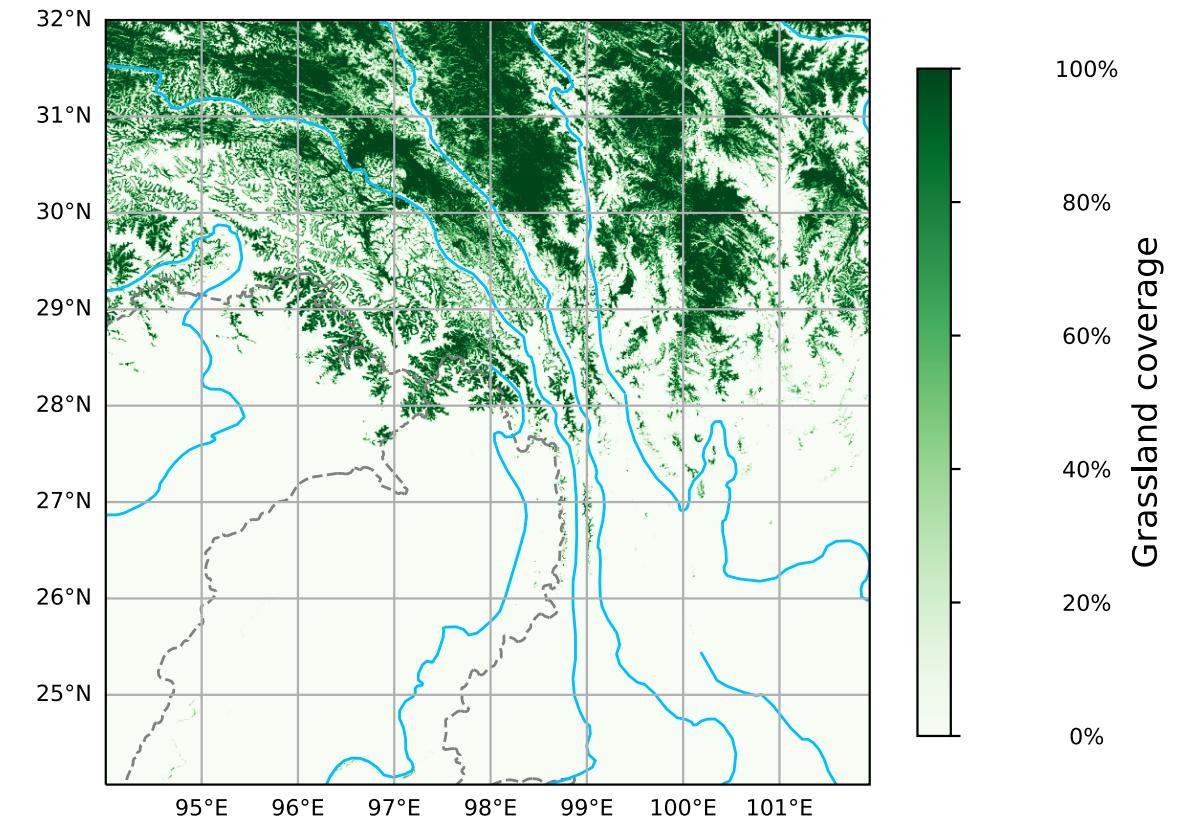
Boussange, V., & Pellissier, L. (2021). Topology and habitat assortativity drive neutral and adaptive diversification in spatial graphs. *BioRxiv*.

In revision @ Communications Biology

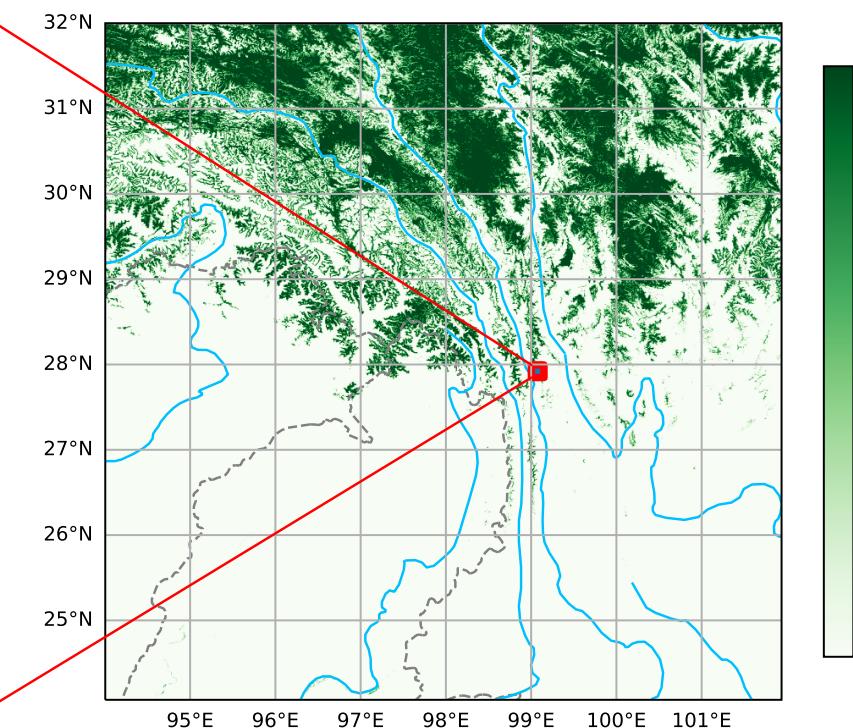
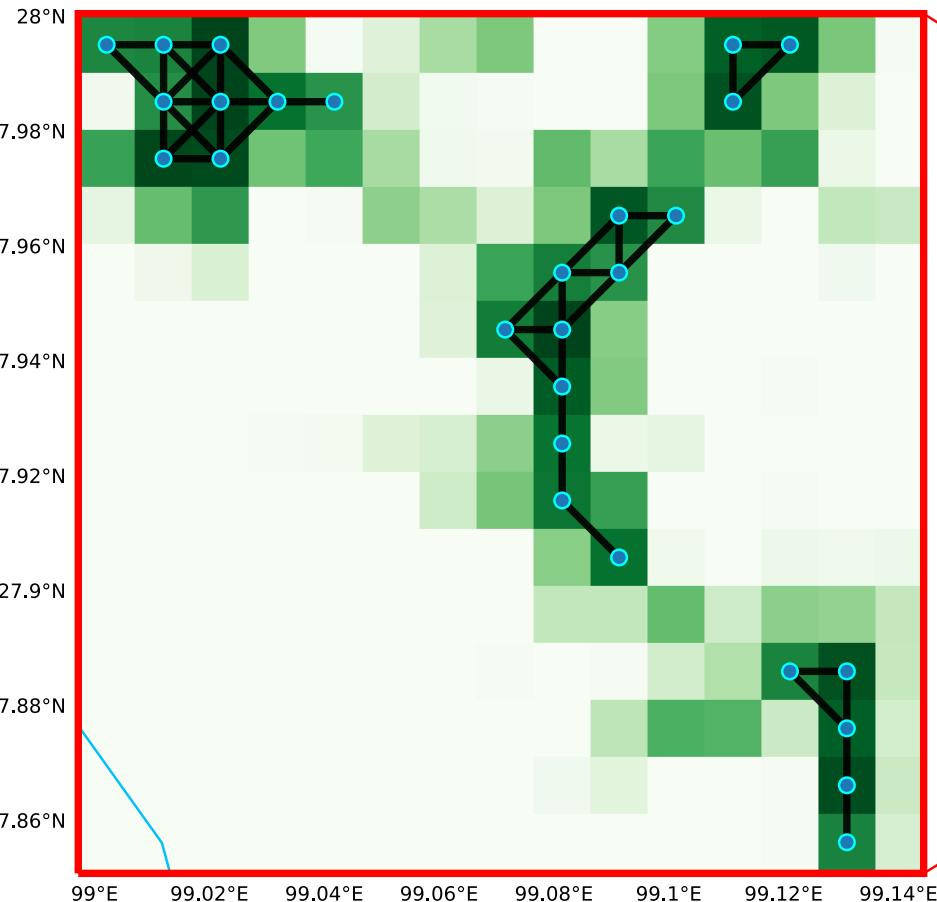
Theory validation and application



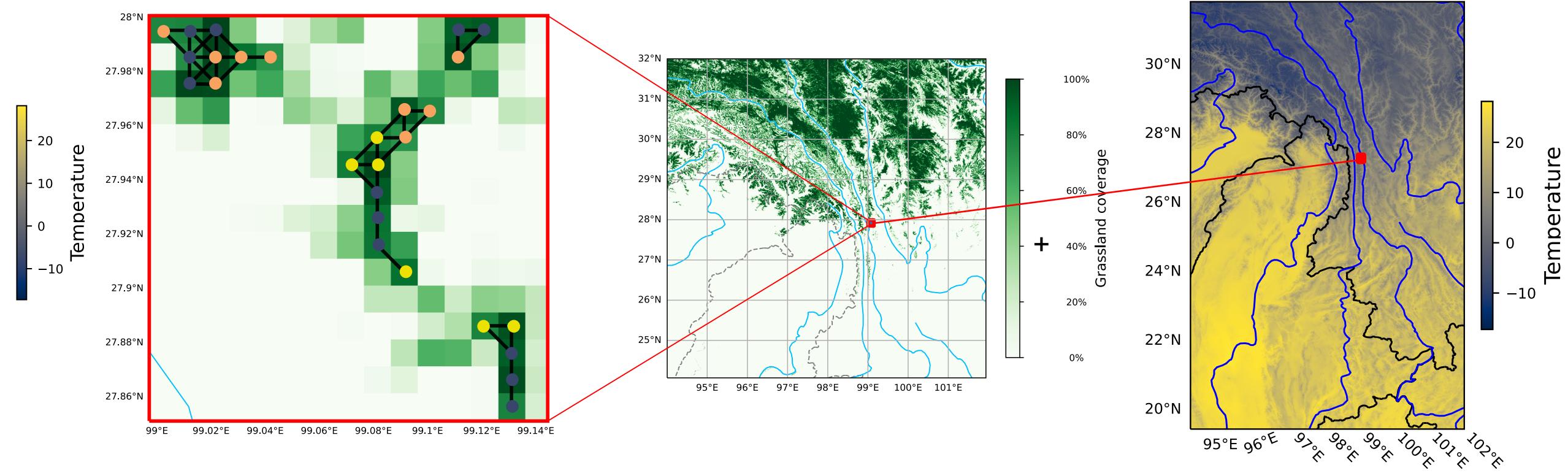
→ Project real landscapes on graphs



Graph representation

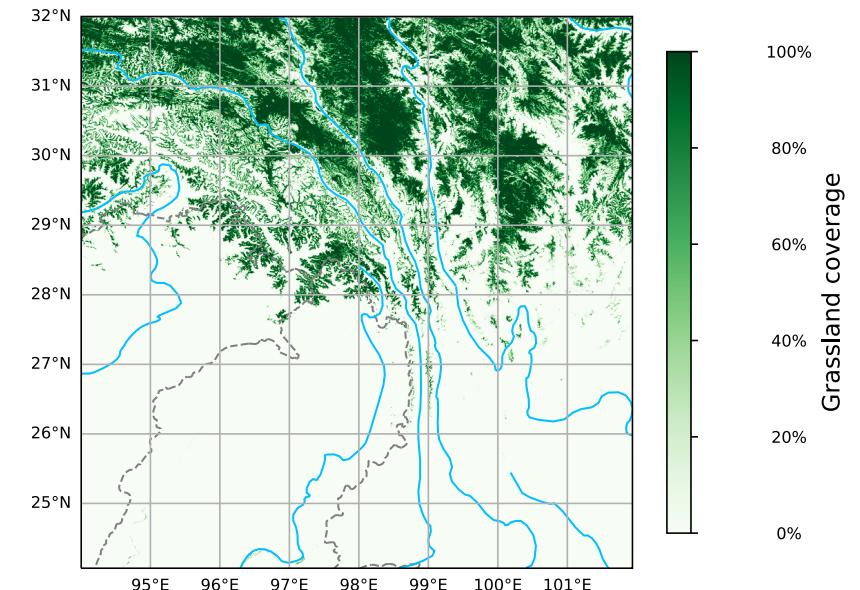


Graph representation with environmental conditions

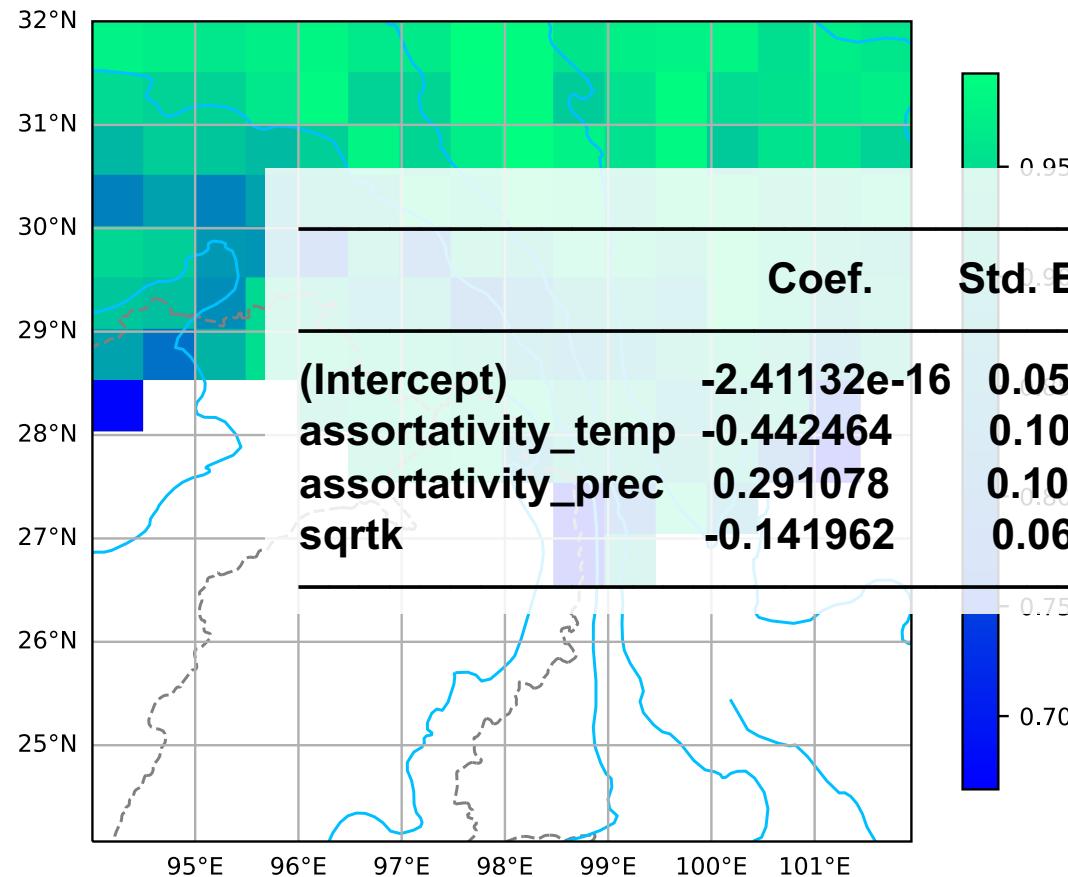


Systematic mapping of graph-based metrics

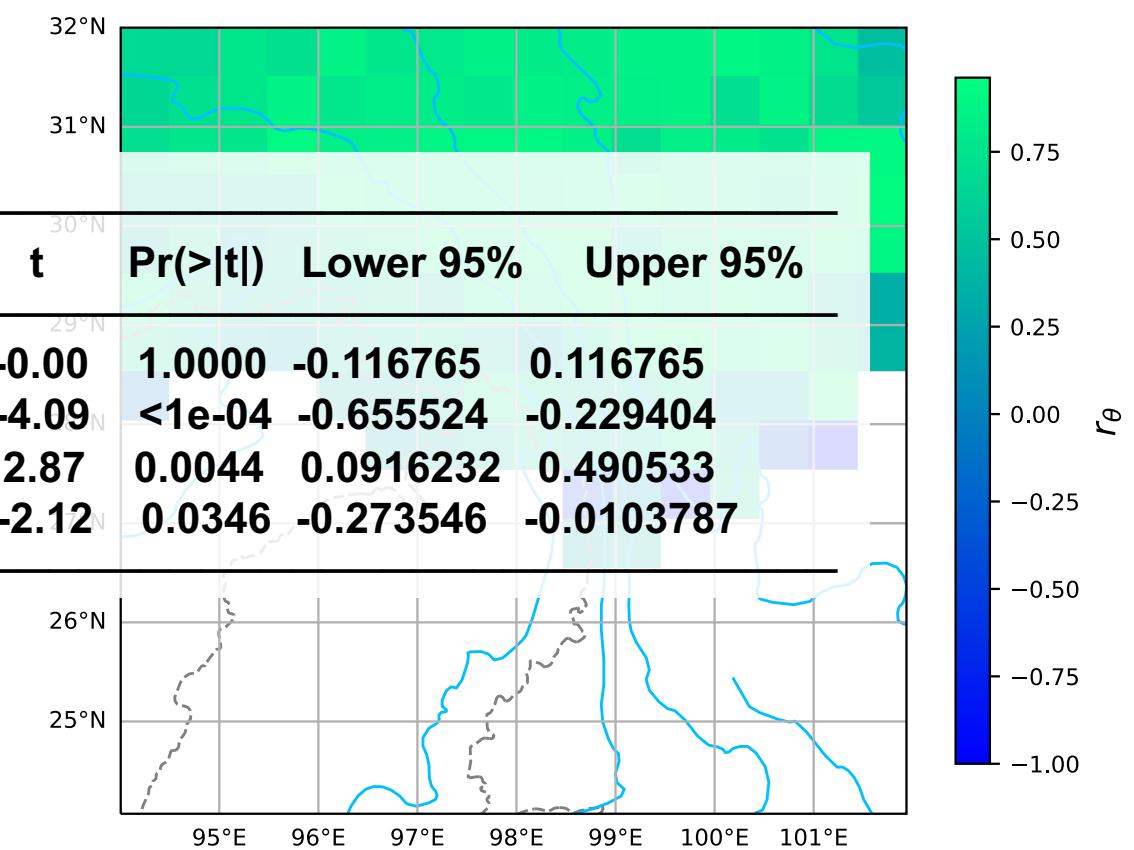
- Compute metrics for all graphs within window $0.25^\circ \times 0.25^\circ$
- Average
- Reproject on map with pixel size $0.25^\circ \times 0.25^\circ$
- Correlate with empirical data



Heterogeneity in degree



Temperature assortativity



Acknowledgements



Loïc Pellissier

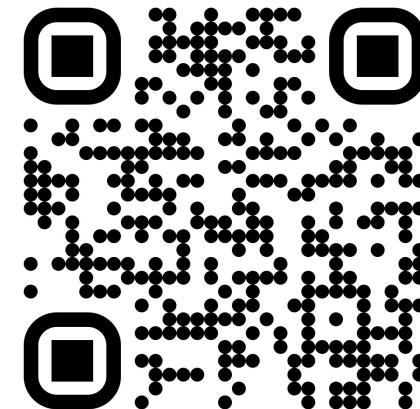


Yaquan Chang



**Thanks!
(looking for a position
next year ☺)**

Check out my personal website



to discover more
about my research