

Reconstructing the Patterns of Alien Plant Species Habitat Niche Expansion

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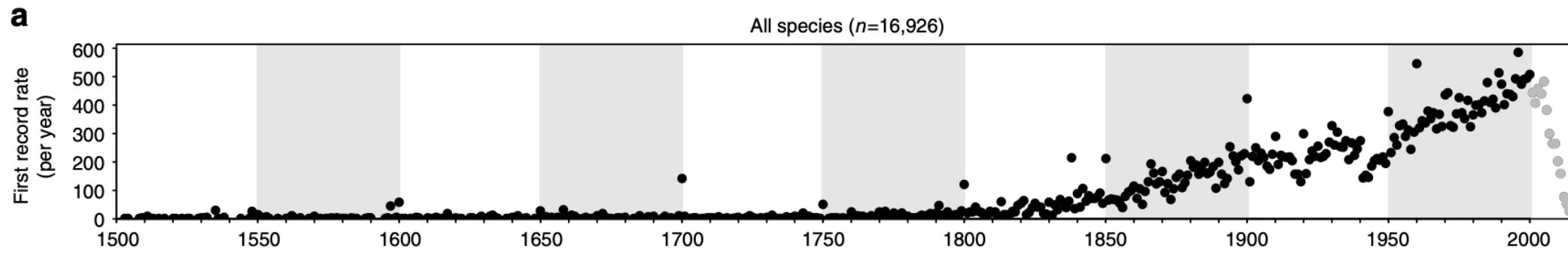
TAC members: Dr. Bernd Lenzner, Dr. Hanno Seebens, Dr. Michael Glaser



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No Signs of Saturation



ARTICLE

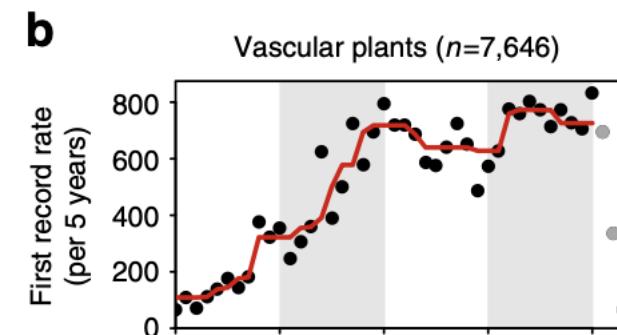
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OPEN

No saturation in the accumulation of alien species worldwide

Hanno Seebens *et al.*[#]

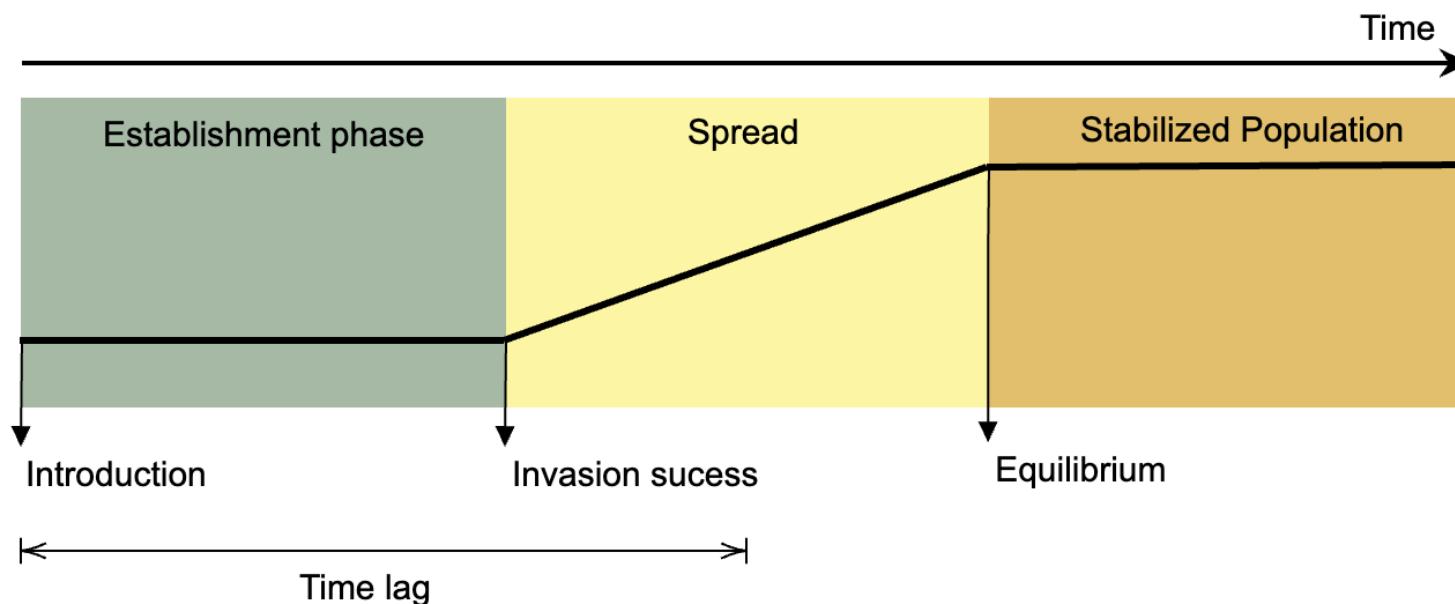


Corollary

- **Regional accumulation** of alien species well-documented (Juozaitienė et al., 2023; Seebens et al., 2017) but **local-scale** effects **less explored**

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- **Regional accumulation** of alien species well-documented (Juozaitienė et al., 2023; Seebens et al., 2017) but **local-scale** effects less explored
- **High numbers** of aliens in **regional** species pools, but **low representation** in **local** communities, mostly a **time-lag phenomenon**



European Perspective - Habitat Origins

- **7'300 alien plants** are recorded across **55 European territories**,
most = **neophytes** (Kalusová et al. 2024)

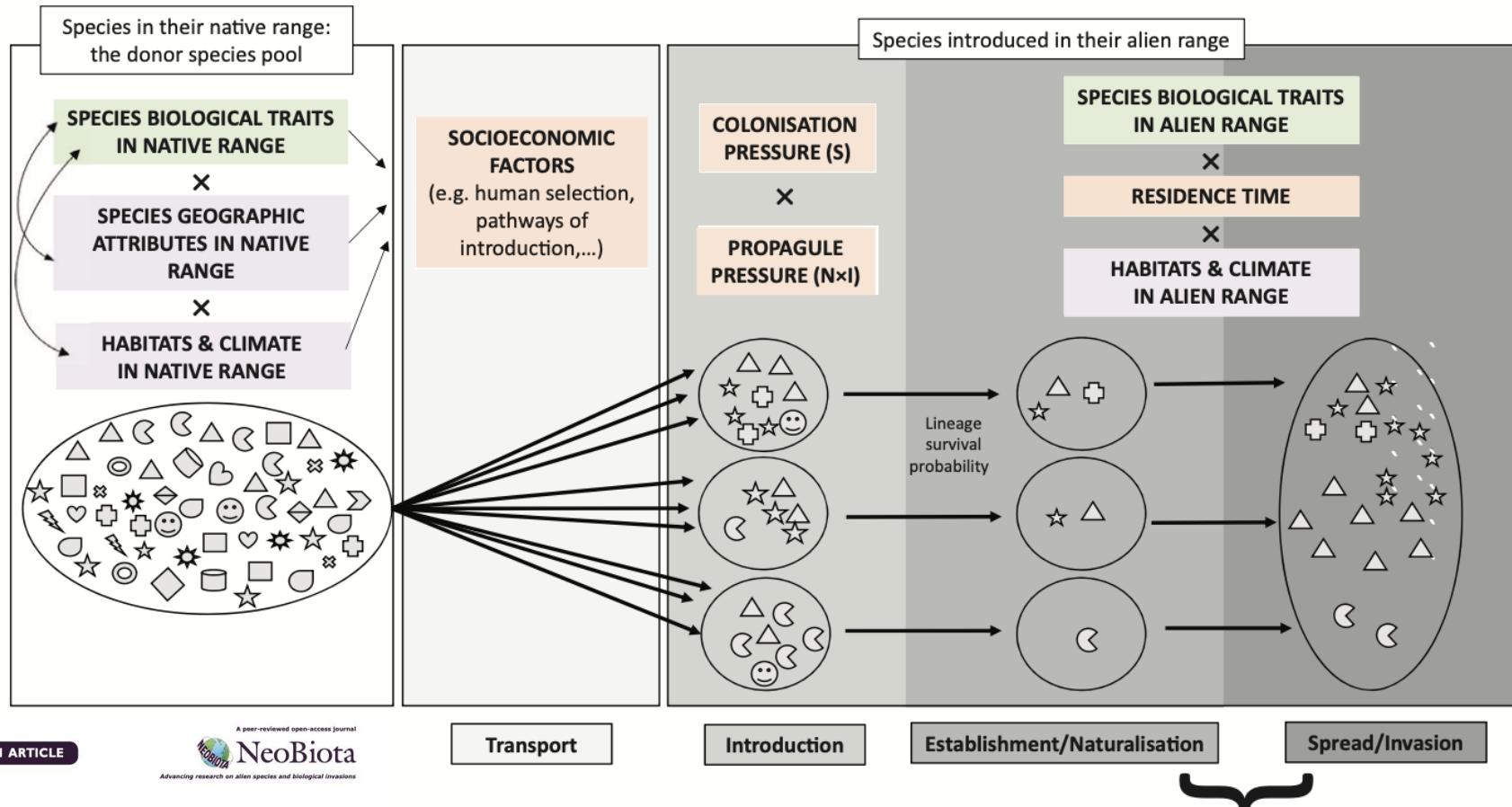
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- **Habitats shaped by disturbance** and environmental **variability**,
(coasts, floodplains, and urban margins) → **main entry points** for
alien species. (Kalusová et al. 2013)
- Interconnected habitats **likely facilitate expansion** of alien species
into new habitat types

Residence Time and Habitat Expansion



NeoBiota 62: 407–461 (2020)
doi: 10.3897/neobiota.62.52787
<http://neobiota.pensoft.net>

MAcroecological Framework for Invasive Aliens
(MAFIA): disentangling large-scale context
dependence in biological invasions

ALIEN MACROECOLOGY:
Richness, distribution, abundance, spatial & trait relationships
of alien biota at large scales

Hypotheses

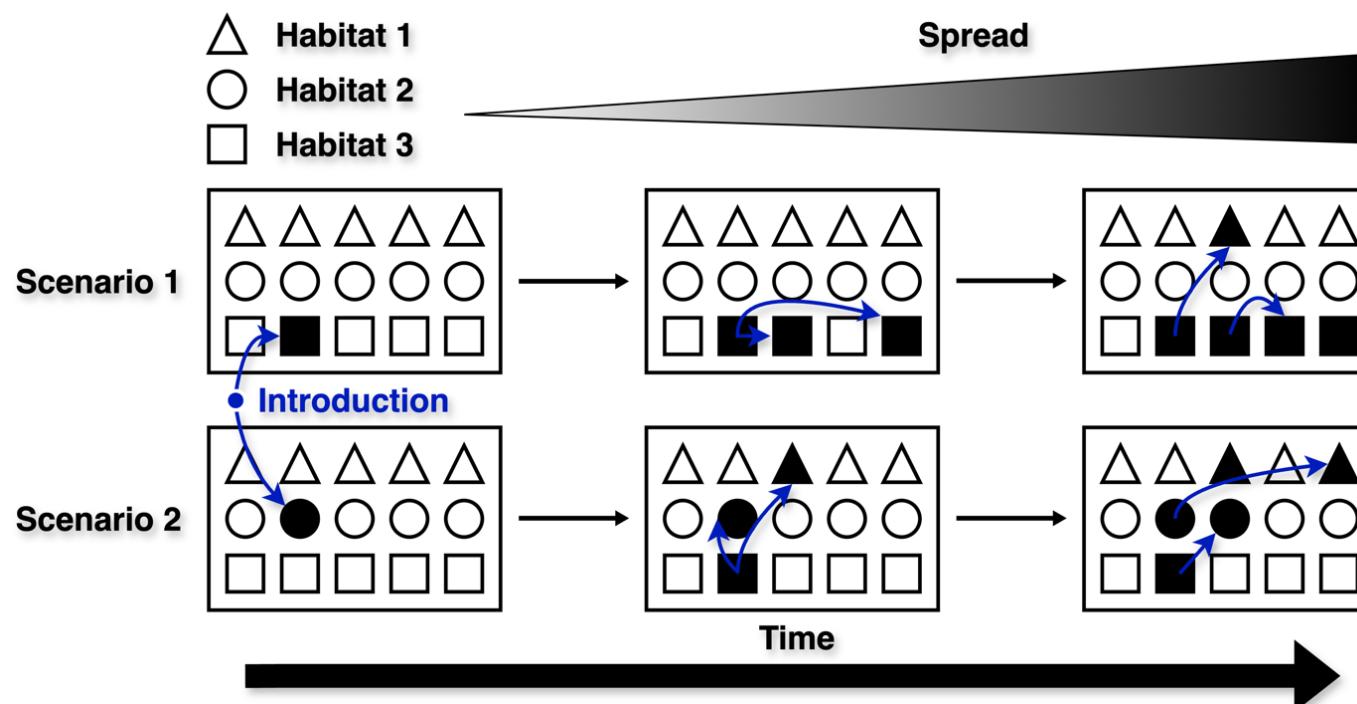
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 - (a) **more local communities of the same habitat type or**

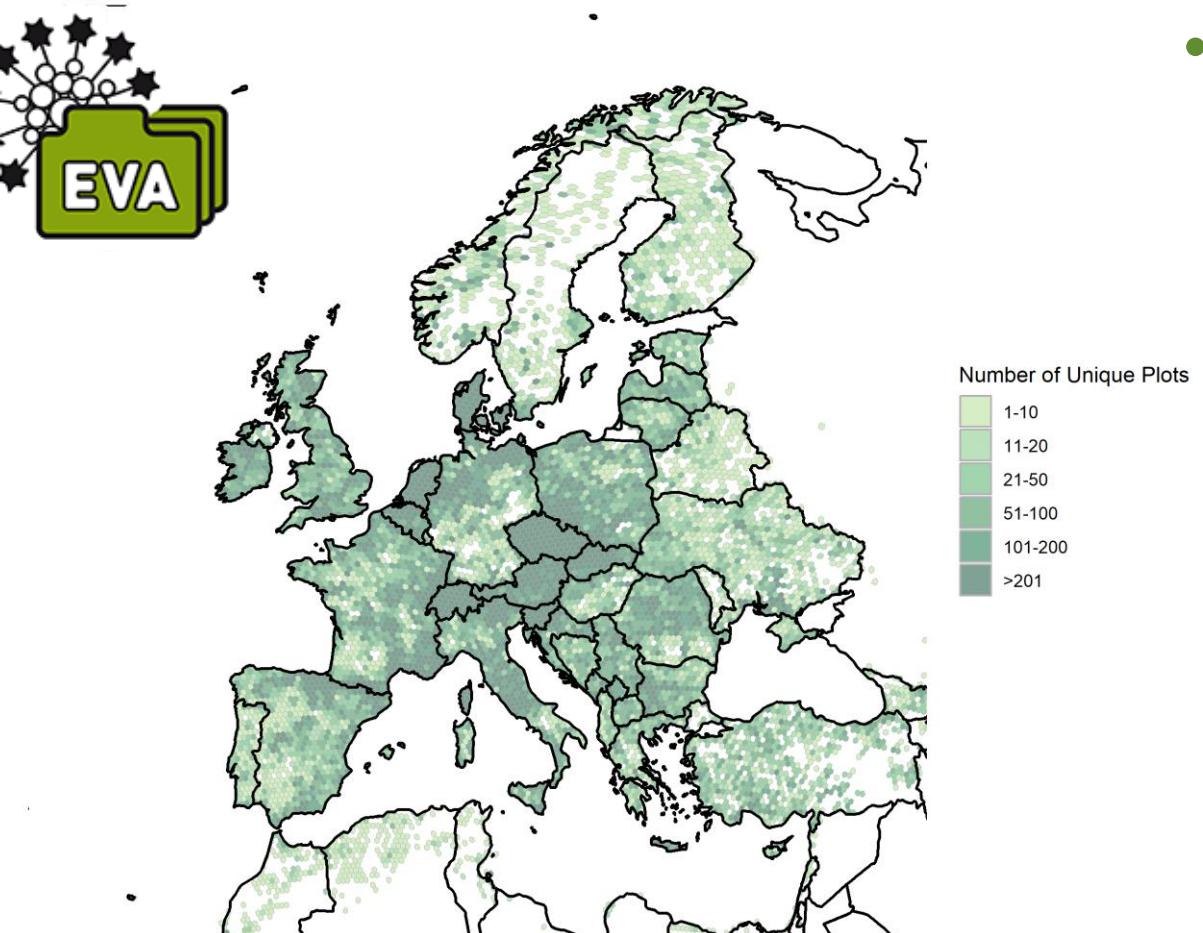
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- To test whether alien species with longer residence times in the regional species pool occur across
 - (a) **more local communities of the same habitat type** or
 - (b) **more local communities of different habitat types.**



Dataset

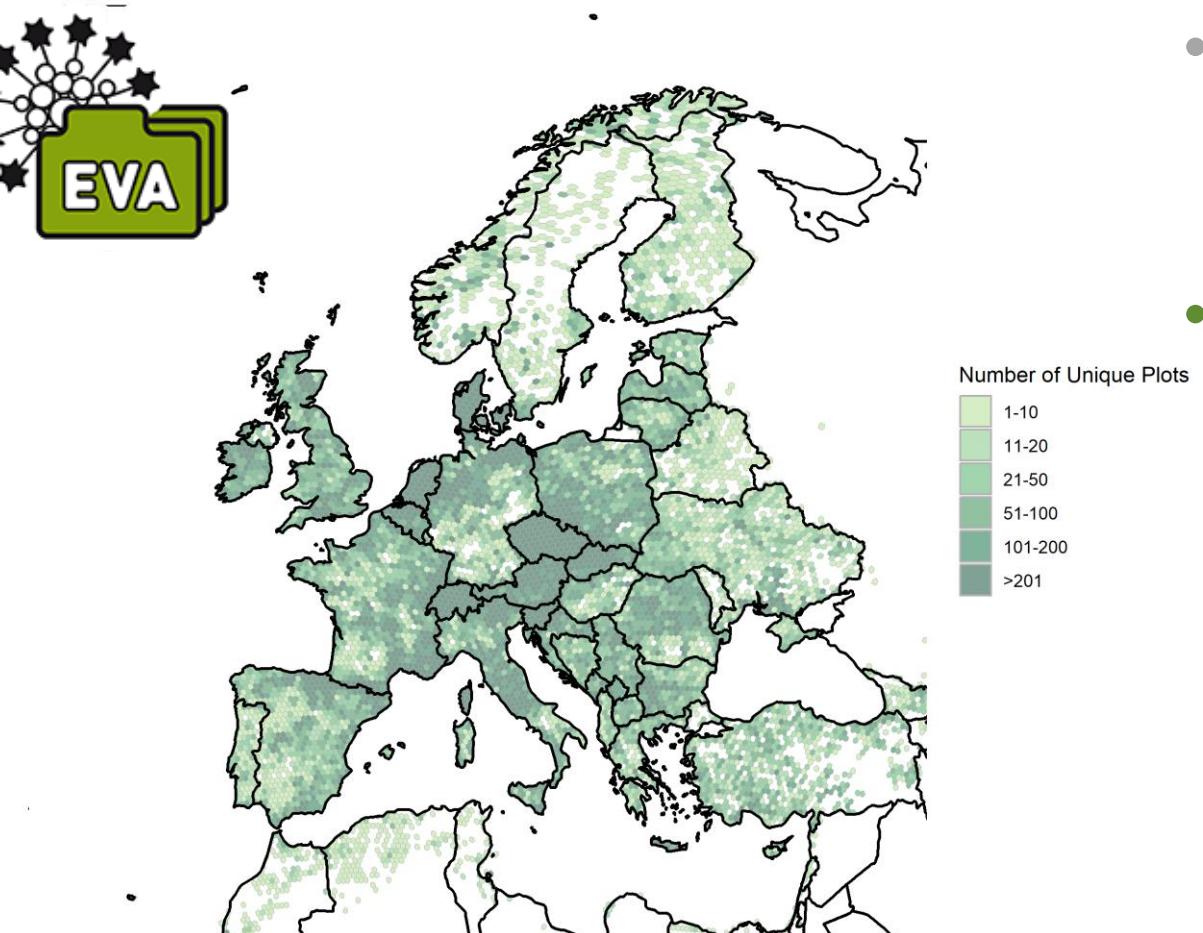
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- **Data for ASAAS compiles 1.9 million vegetation plots from more than 100 regional databases across Europe**

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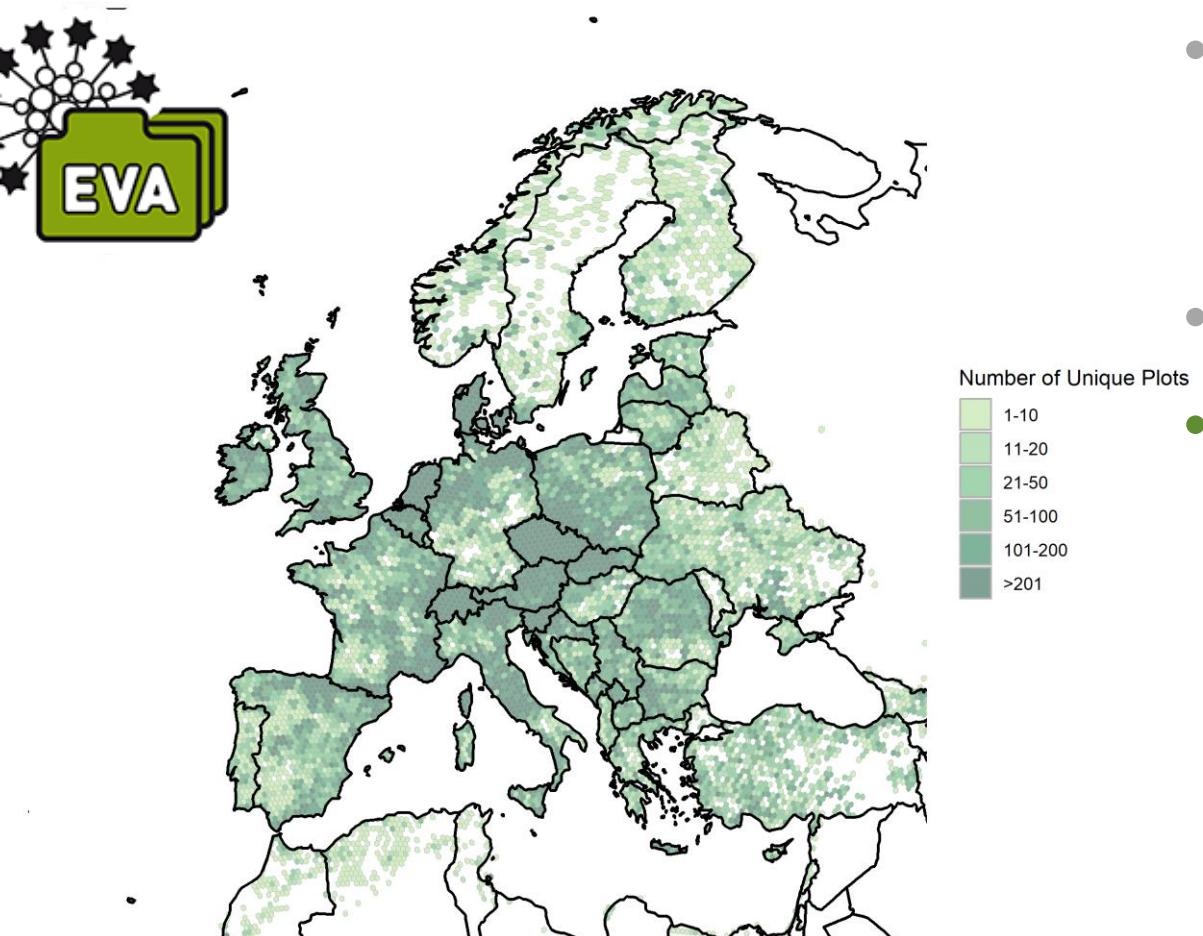
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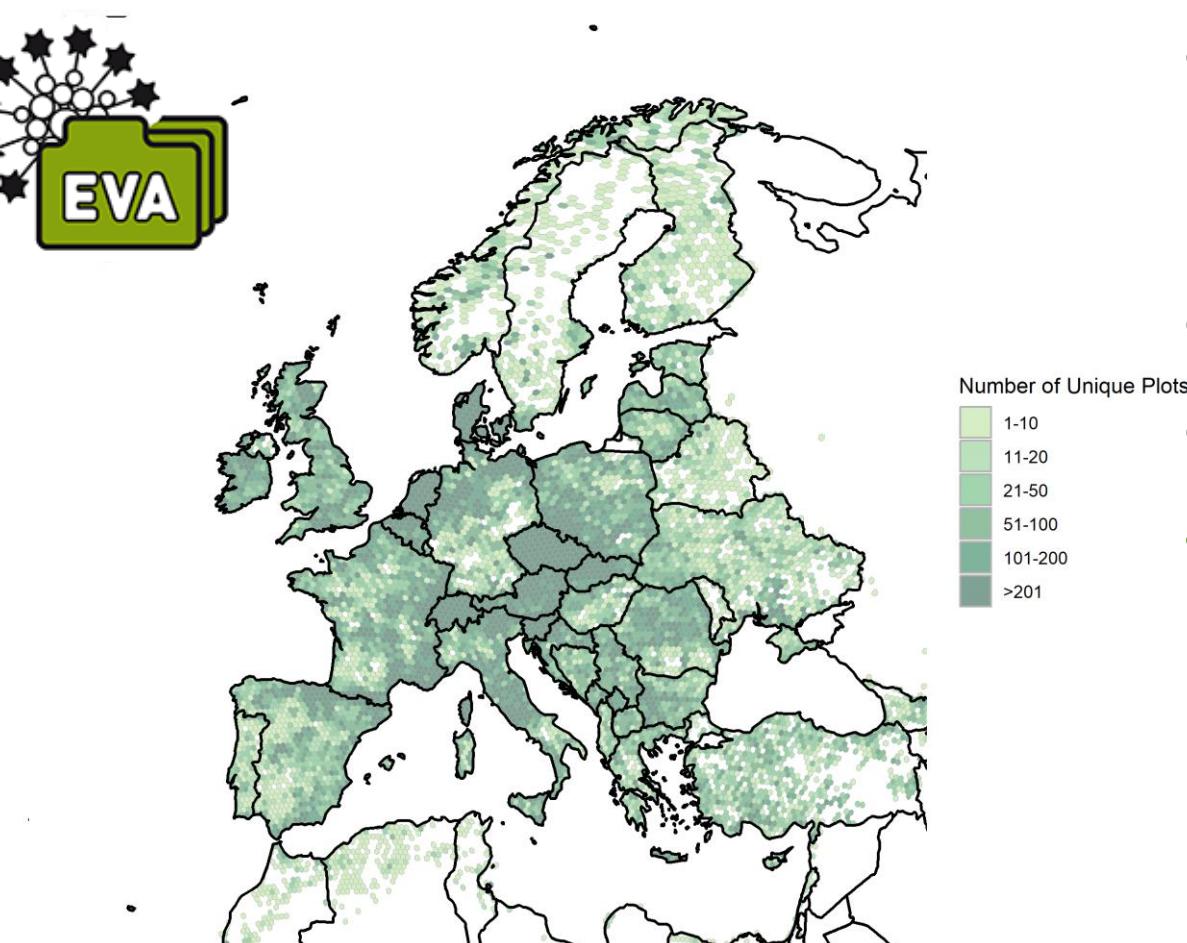
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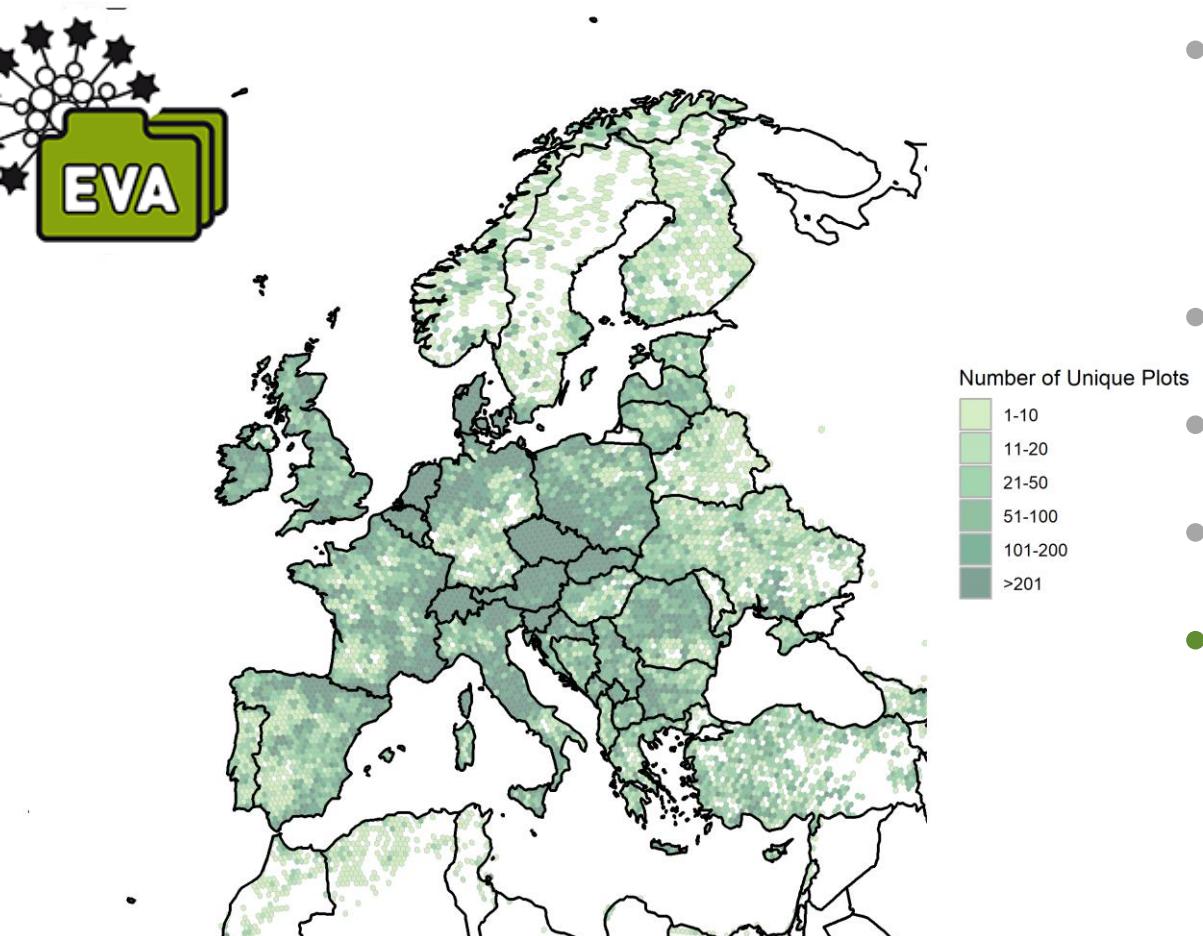
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 - **EUNIS classification**

Habitat Classification

- EUNIS Level 2 aggregated into 18 habitats

Group	Examples
Man-made & ruderal	Arable, gardens, artificial grasslands
Broadleaved deciduous	Temperate deciduous forests
Mesic & wet grasslands	Hay meadows, wet grasslands
Dry grasslands	Xerophytic grasslands
Alpine & subalpine	High-altitude grasslands
Bogs & poor mires	Peat-accumulating wetlands

Intermediate resolution: Level 1 too coarse, Level 2 too detailed

The Problem with Counting Habitats

- If we *observe* a species in **grasslands and wetlands...**
 - a) is it: **Habitat preference**, or **sampling** where it is common?
 - b) Raw **occurrence** conflates distribution with **habitat availability**
 - c) EVA is **sampled opportunistically**: some habitats heavily, some sparse

Need: Compare species to *local habitat availability*

How to measure Habitat profiles?

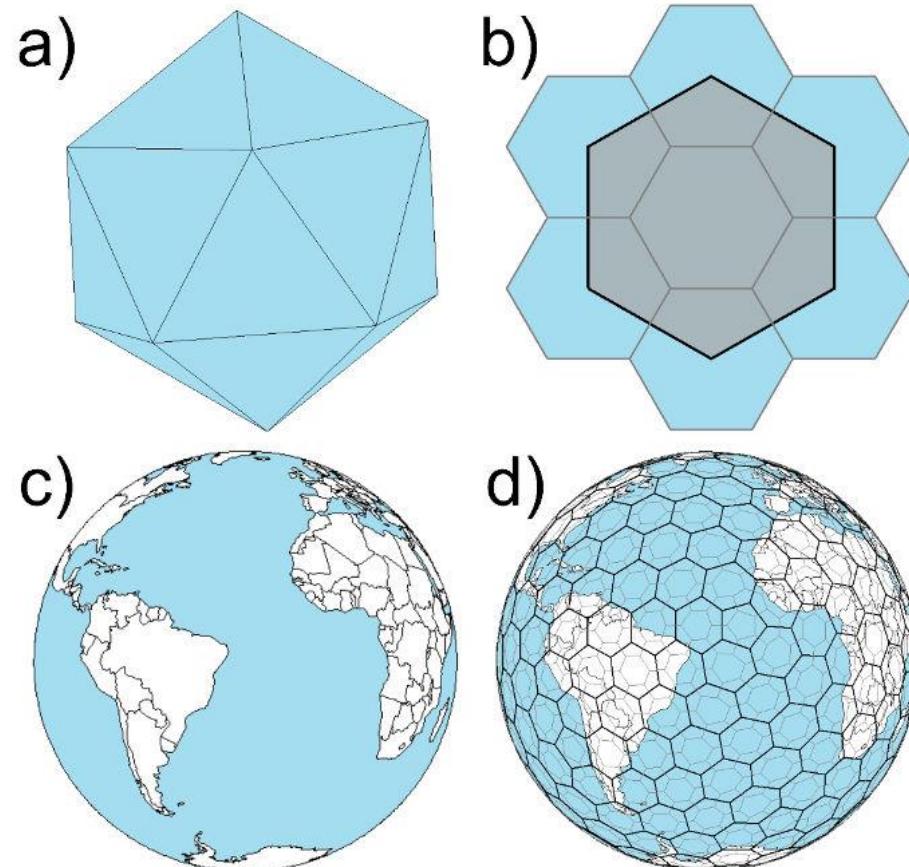
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How to measure Habitat profiles?

- **Fidelity, IndVal, diversity indices** = designed for **systematic surveys**
- **Problems** within EVA data:
 - **Denominators unstable** when habitats **sampled unevenly**
 - **Shannon diversity increases** with **sample size** regardless of true distribution and isn't well defined for “habitat diversity”
 - Does **not answer**: “Is the **species more common than random chance** given local availability?”

Methods

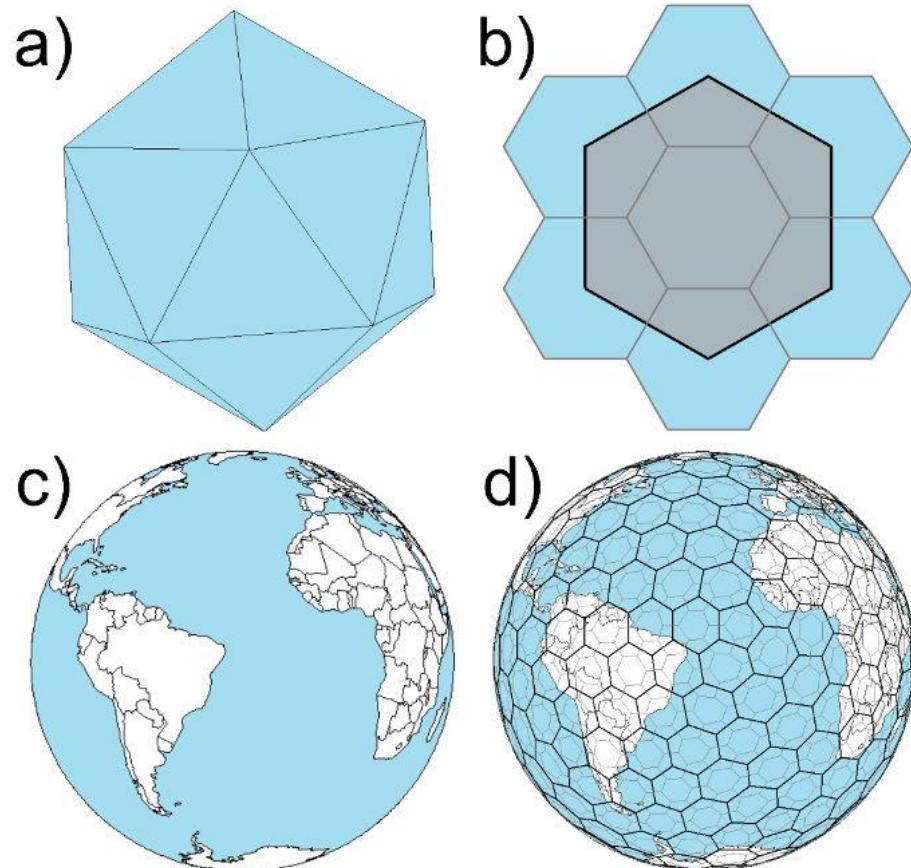
Comparing Habitat use across 863 km² x 10 years hexcells



- **Accounting for sampling:**
→ Reorganize data into hexagonal grid cells

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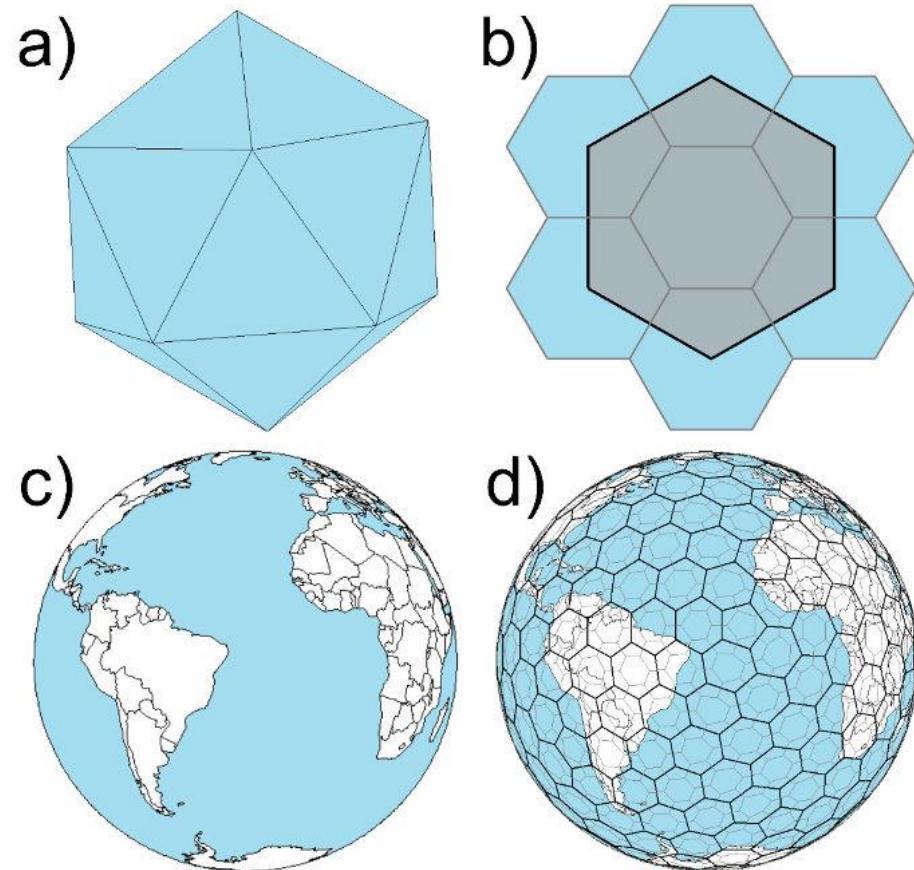
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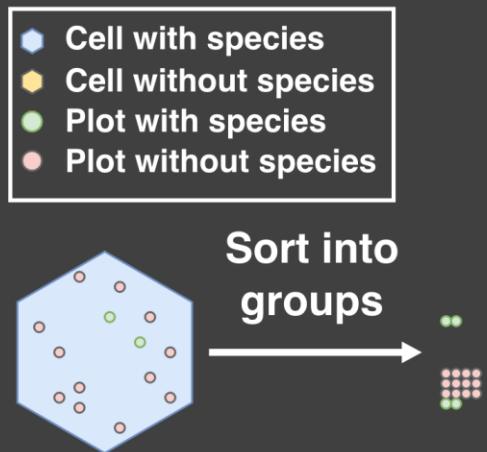
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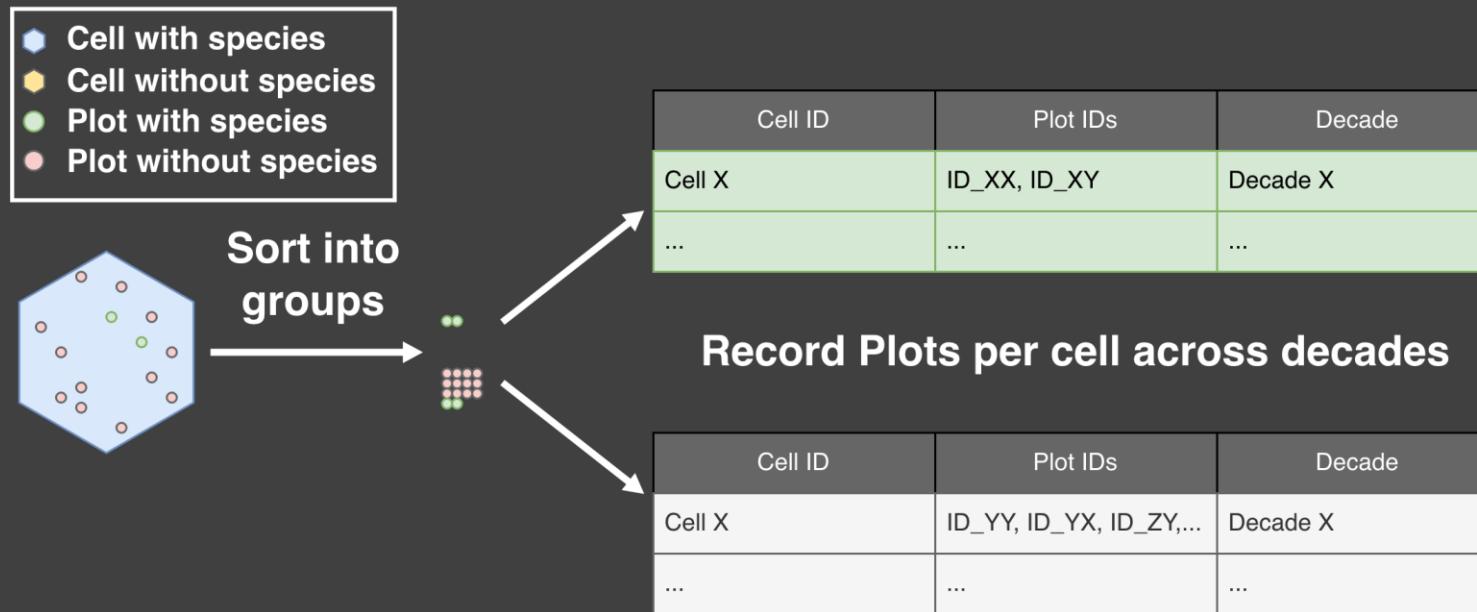


- Accounting for sampling:
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- Within each **hexcell** we collect:
 - **Neophyte** habitat distribution (**p**)
 - **Background** habitat distribution (**q**)
- Within each habitat (per decade):
 - Null-model: $p \approx q$
 - Overrepresented: $p > q$

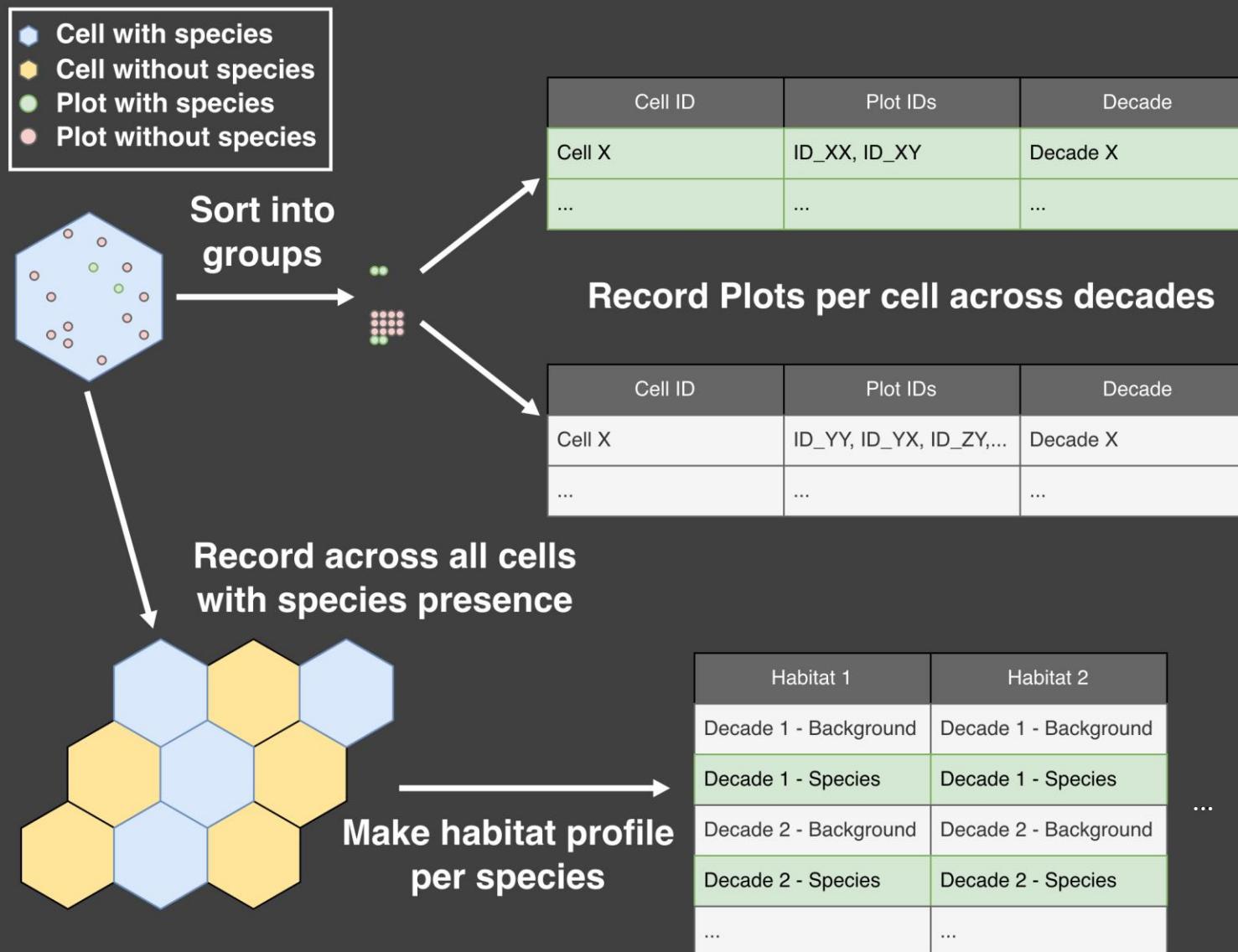
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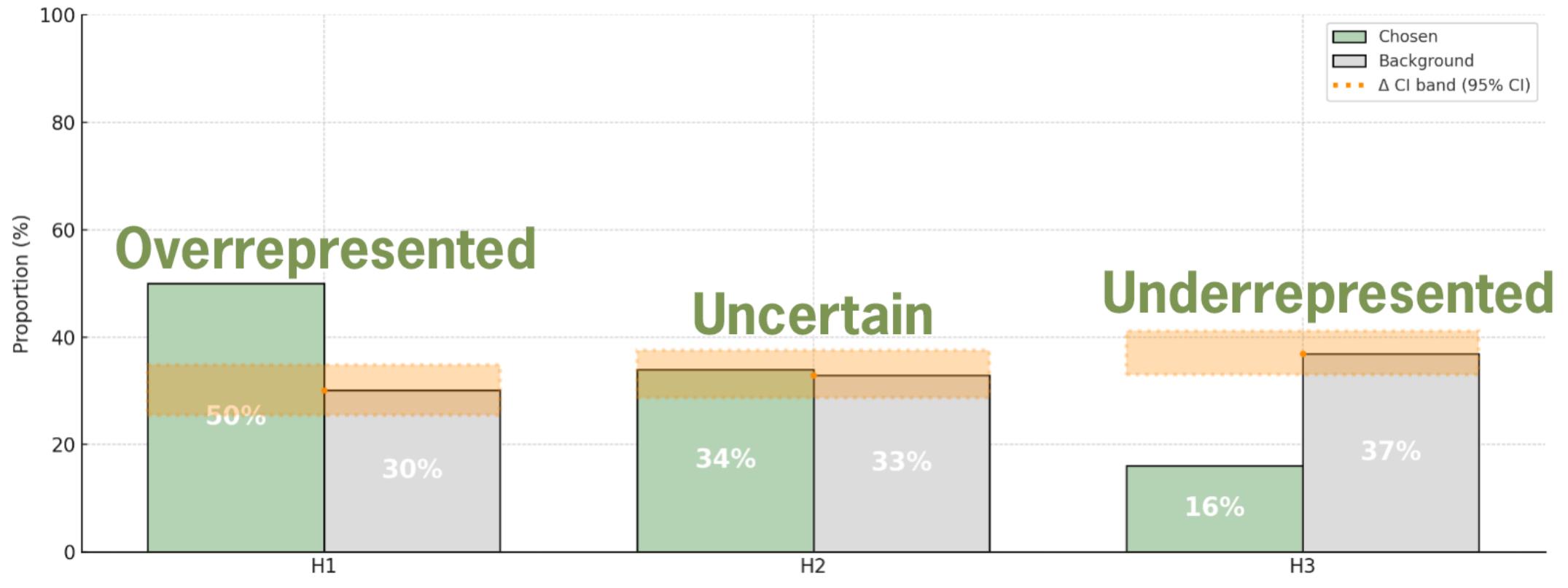
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Species (green) vs background (grey) with 95% CI band

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- **Clear null model**
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 - Each hexcell x decade is independent of global sampling
 - Oversampled regions don't dominate the signal
- **Categorical output**
 - Species is overrepresented **(1) or not (0)** per habitat
 - **Directly usable as response**

Methods

Biologically, two distinct processes are involved:

1. **Can the species extend beyond one habitat?** (generalist potential vs. specialist constraint)
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1. **Can the species extend beyond one habitat?** (generalist potential vs. specialist constraint)
 - Many species remain constraint
 - Outcome: Binary (single vs multiple)
2. **How many additional habitats** does it occupy once it expands?
 - Given expansion, what determines breadth?
 - Outcome: Counts (number of additional habitats)

Methods

$$Y_i \sim \text{TruncHurdleNegBin}(\pi_i, \mu_i, \phi)$$

- **Hurdle** component (logit link)

$$\text{logit}(\pi_i) = \alpha^{(\text{hur.})} + X_i \beta^{(\text{hur.})} + u_{\text{county}[i]}^{(\text{hur.})}$$

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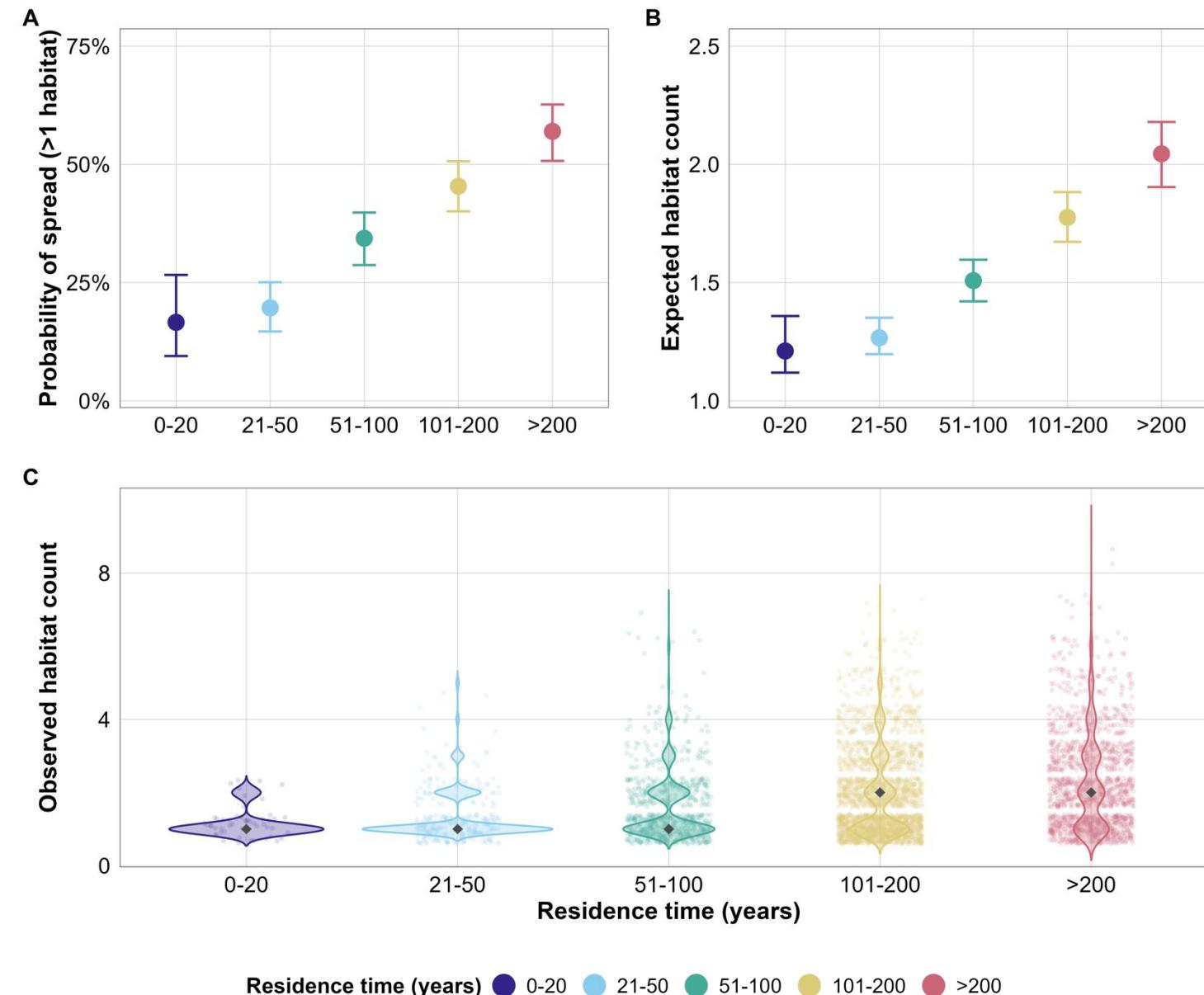
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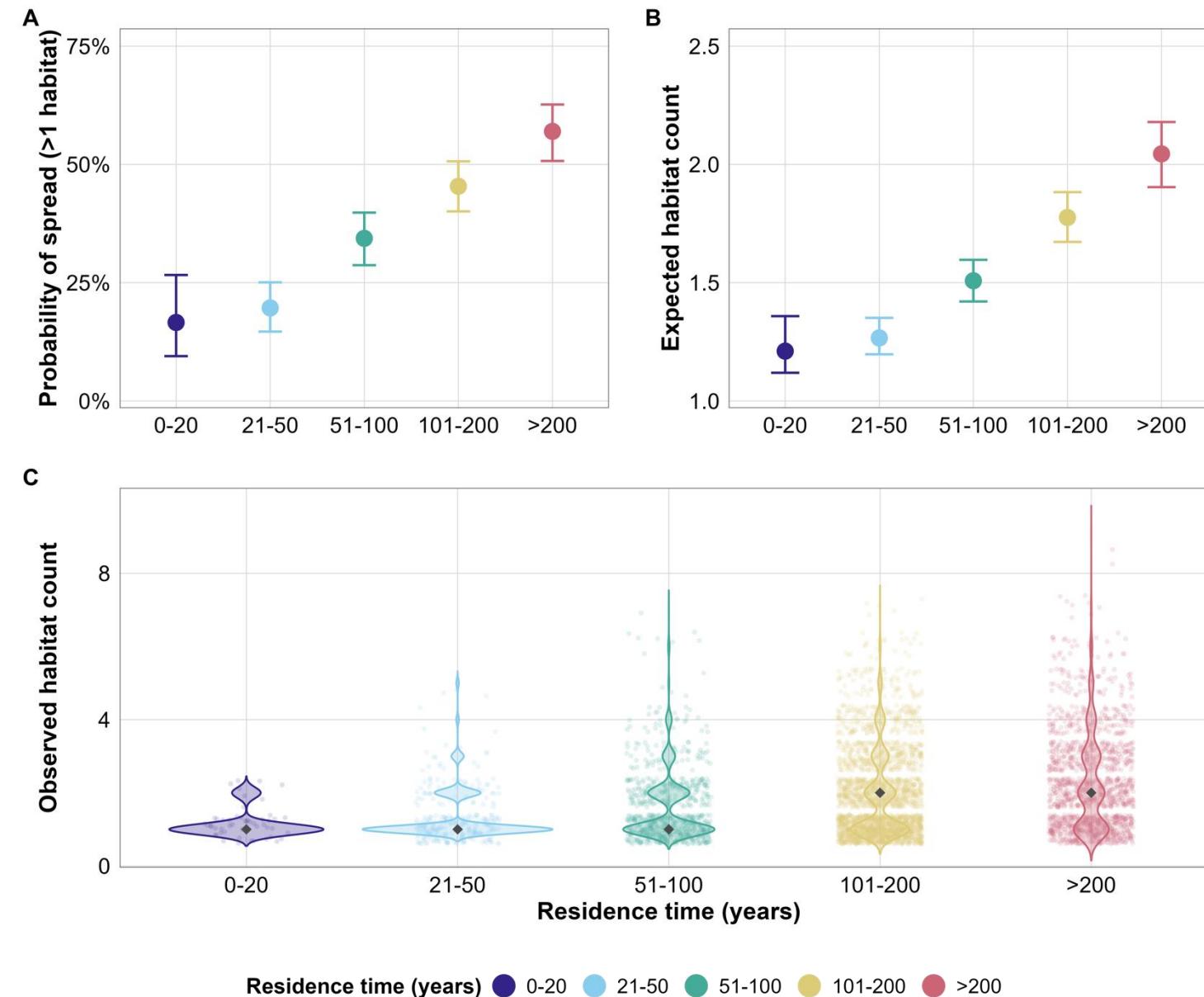
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- **Random effects:** country intercepts

Results



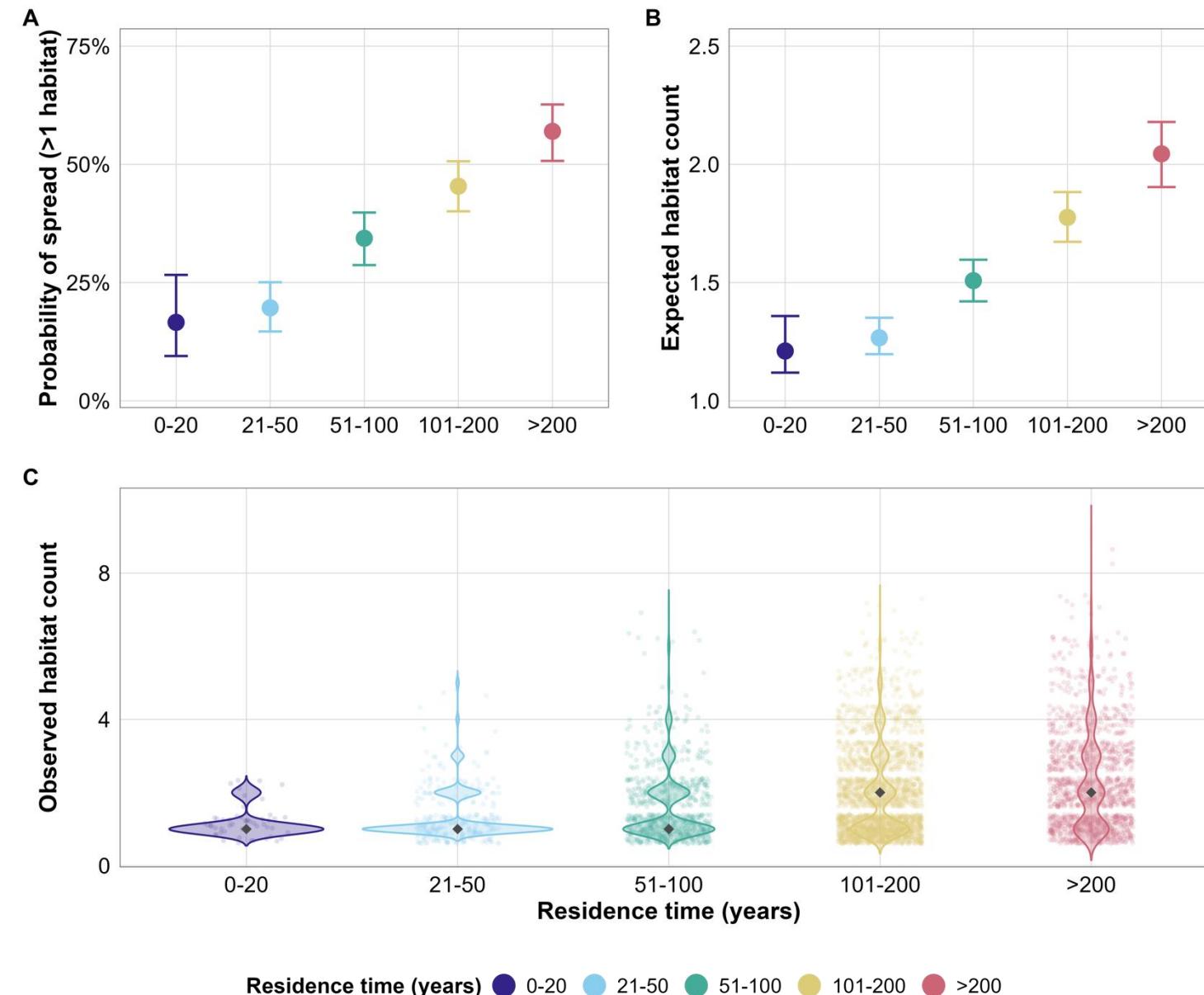
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- **Most pairwise contrasts significant**

Results

- **Hurdle component (P of > 1 habitat)**
 - Trees, vines: reduced prob. (OR 0.57-0.69)
 - Long-lived perennials: reduced prob. (OR 0.55)
 - Human-mediated dispersal: reduced prob. (OR 0.73)

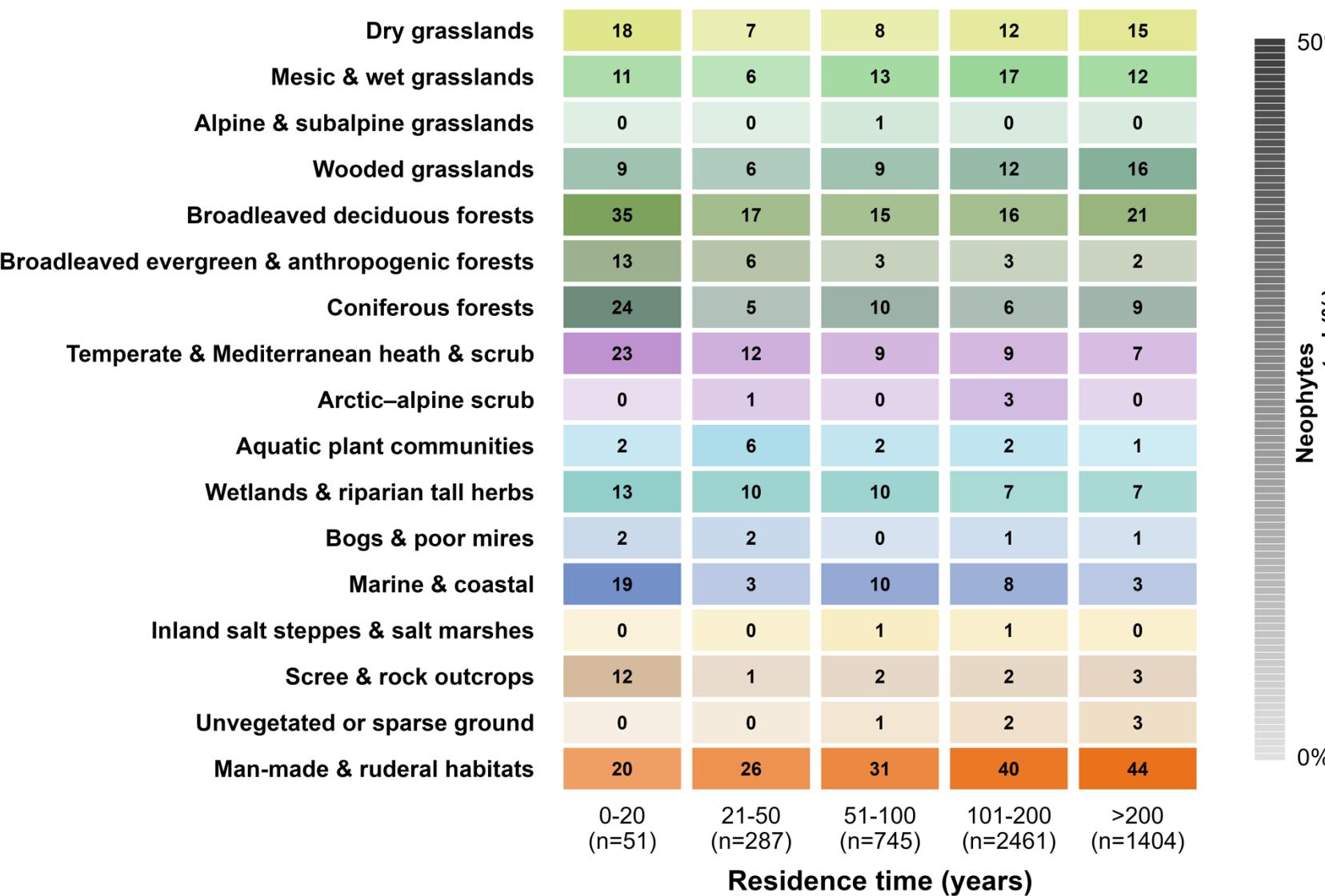
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- **Count component (expected habitats given > 1)**
 - Long-lived perennials: +43% more habitats (IRR 1.43)
 - Medium-lived perennials: +20% more habitats (IRR 1.20)

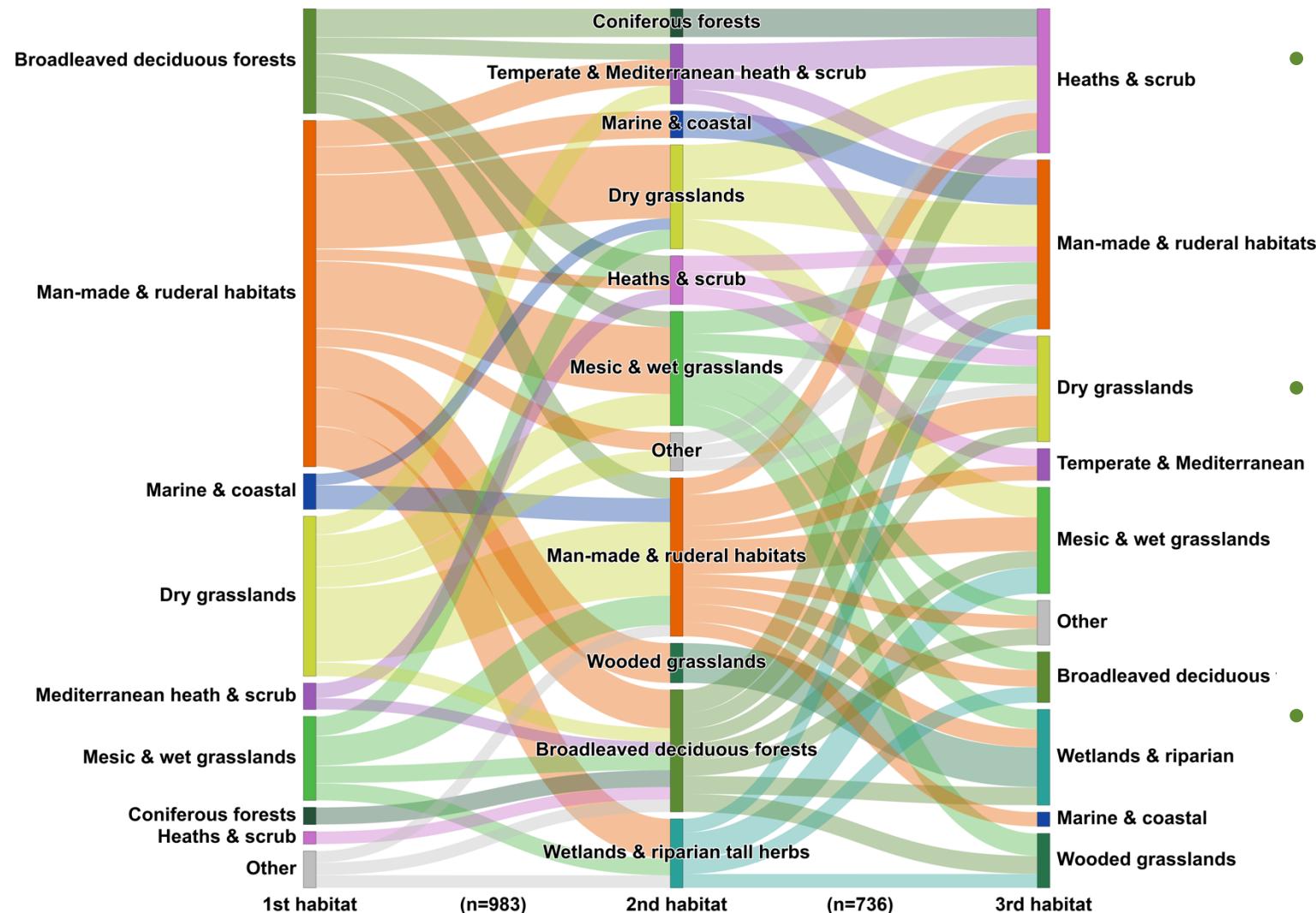
*OR = Odds ratio, IRR = Incidence Rate Ratio

Results



- **High Invasion:**
 - Man made & ruderal
- **High Resistance:**
 - Alpine grasslands
 - Bogs & poor mires
- **Intermediate**
 - Mesic grasslands
 - Deciduous forests

Results



- **First habitat:**
 - Man-made & ruderal (31%)
 - Dry grasslands (16%)
 - Broadleaved dec. (12%)
- **Common Establishment Patterns:**
 - Dry grasslands ↔ man-made
 - Man-made → mesic
 - Man-made → wetlands
- **Third habitat**
 - Mesic & wet grasslands = common tertiary site

Caveats

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- **Transition analysis**
 - Continental aggregation, no local colonization pathways
 - Species are introduced independent across regions

Take Home

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- **Ecological**
 - Habitat breadth grows with residence time → strong evidence of ongoing spread
- **Invasion debt**
 - The residence time effect continues up to >200 years → no sign of equilibrium

A dense forest of tall, slender bamboo stalks with green leaves at the top.

Thank you