

# Reconstructing the Patterns of Alien Plant Species Habitat Niche Expansion

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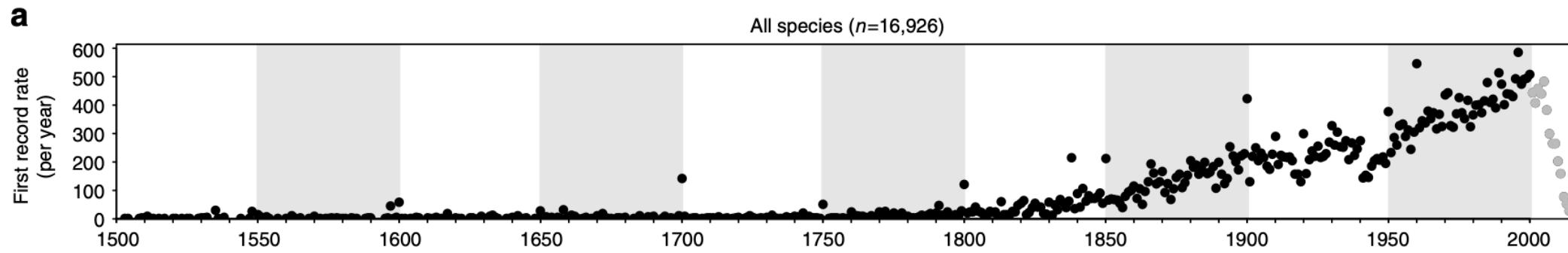
**TAC members:** Dr. Bernd Lenzner, Dr. Hanno Seebens, Dr. Michael Glaser



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# No Signs of Saturation



ARTICLE

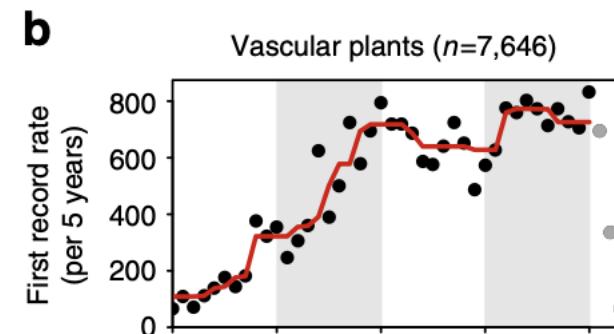
Received 16 Feb 2016 | Accepted 28 Dec 2016 | Published 15 Feb 2017

DOI: 10.1038/ncomms14435

OPEN

## No saturation in the accumulation of alien species worldwide

Hanno Seebens et al.<sup>#</sup>

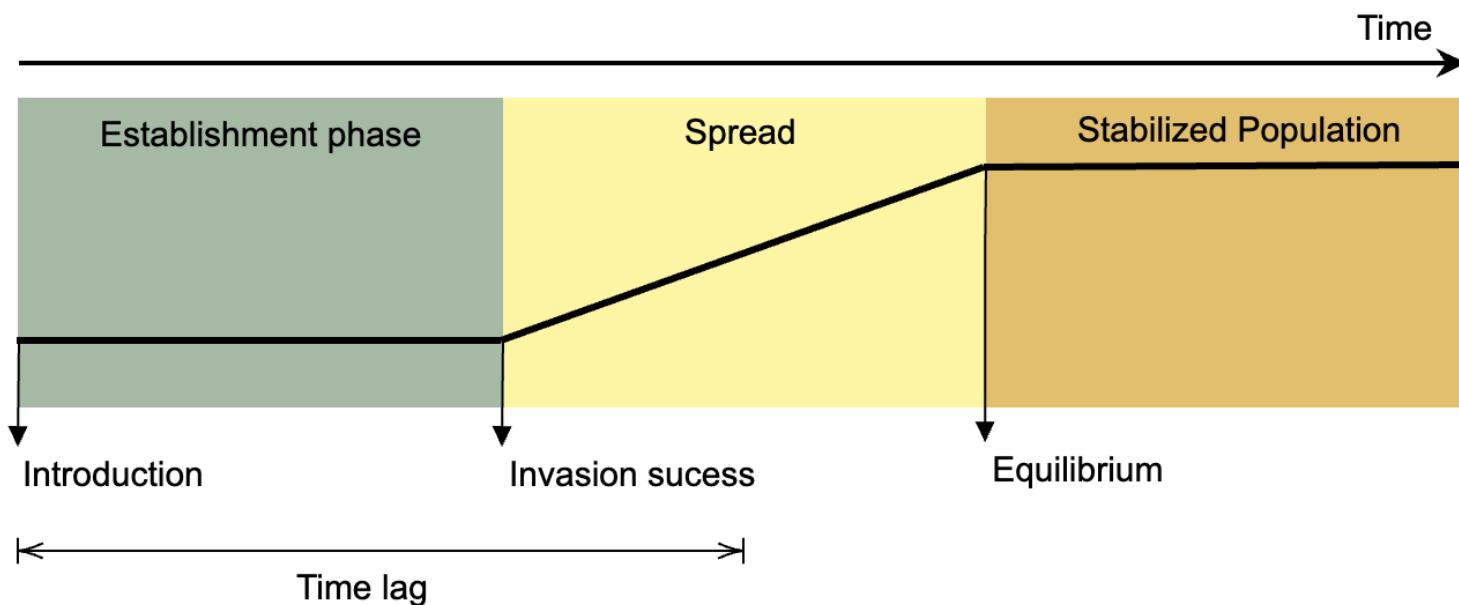


# Corollary

- **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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- 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)

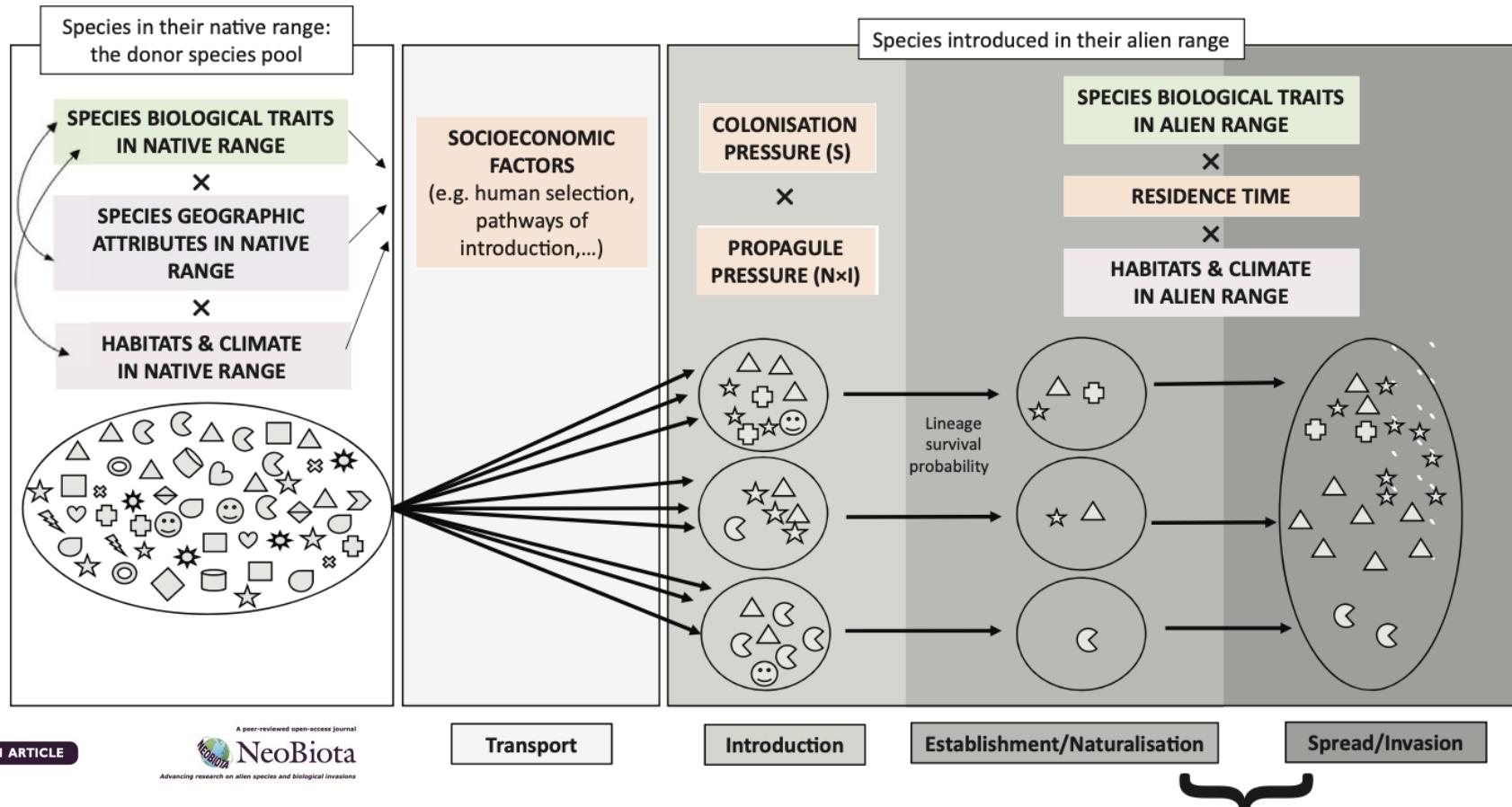
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(coasts, floodplains, and urban margins) → **main entry points** for  
alien species. (Kalusová et al. 2013)
- Interconnected habitats **likely facilitate expansion** of alien species  
**into new habitat types**

# Residence Time and Habitat Expansion



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

RESEARCH ARTICLE

A peer-reviewed open-access journal  
**NeoBiota**  
Advancing research on alien species and biological invasions

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

# Hypotheses

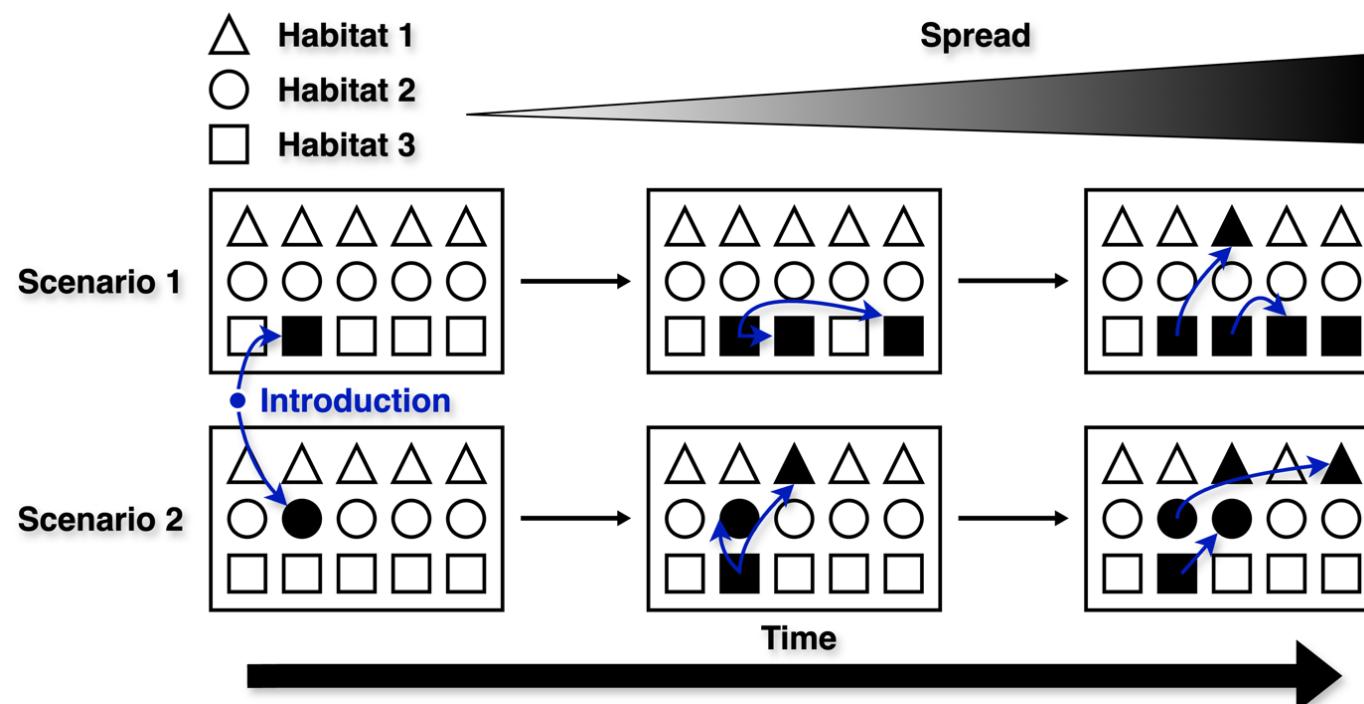
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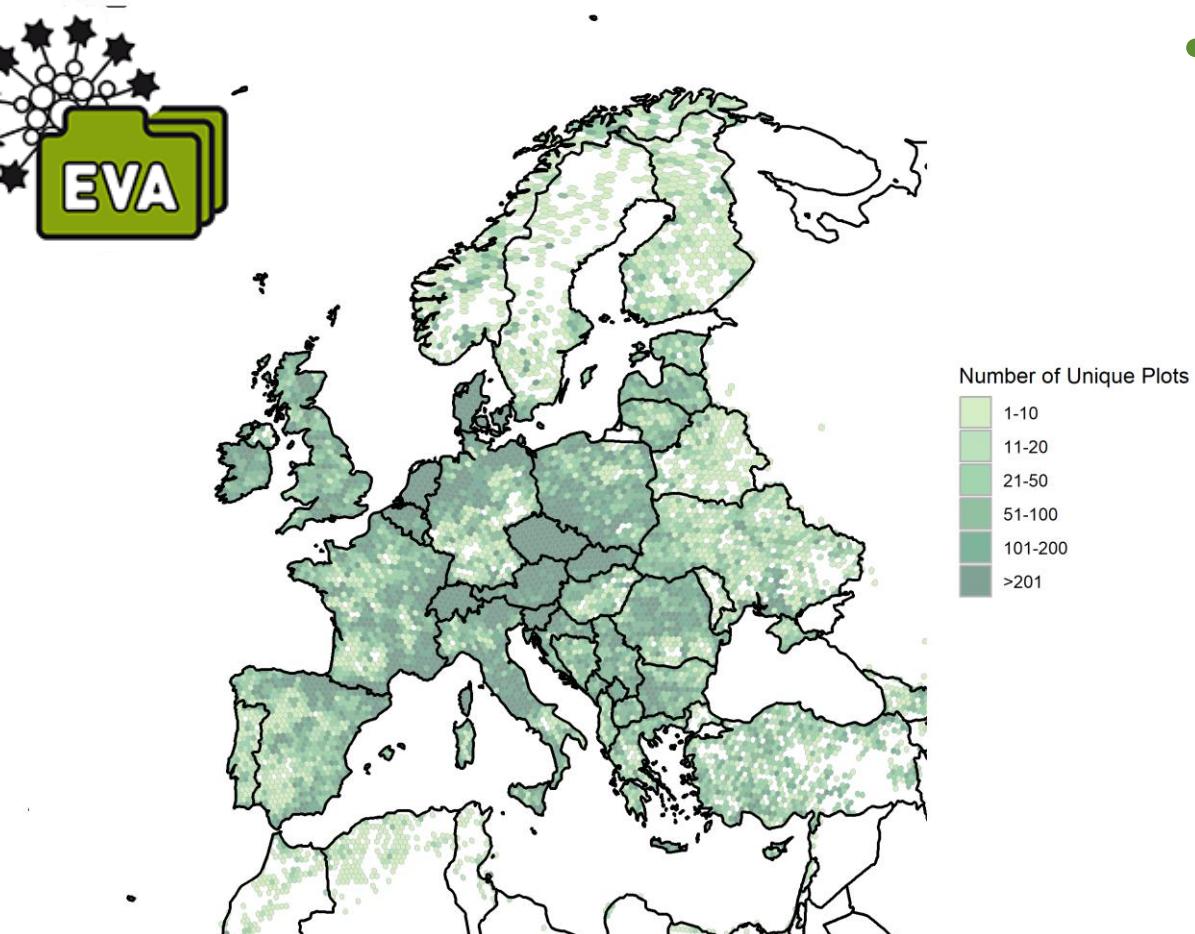
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  - (b) **more local communities of different habitat types.**



# Dataset

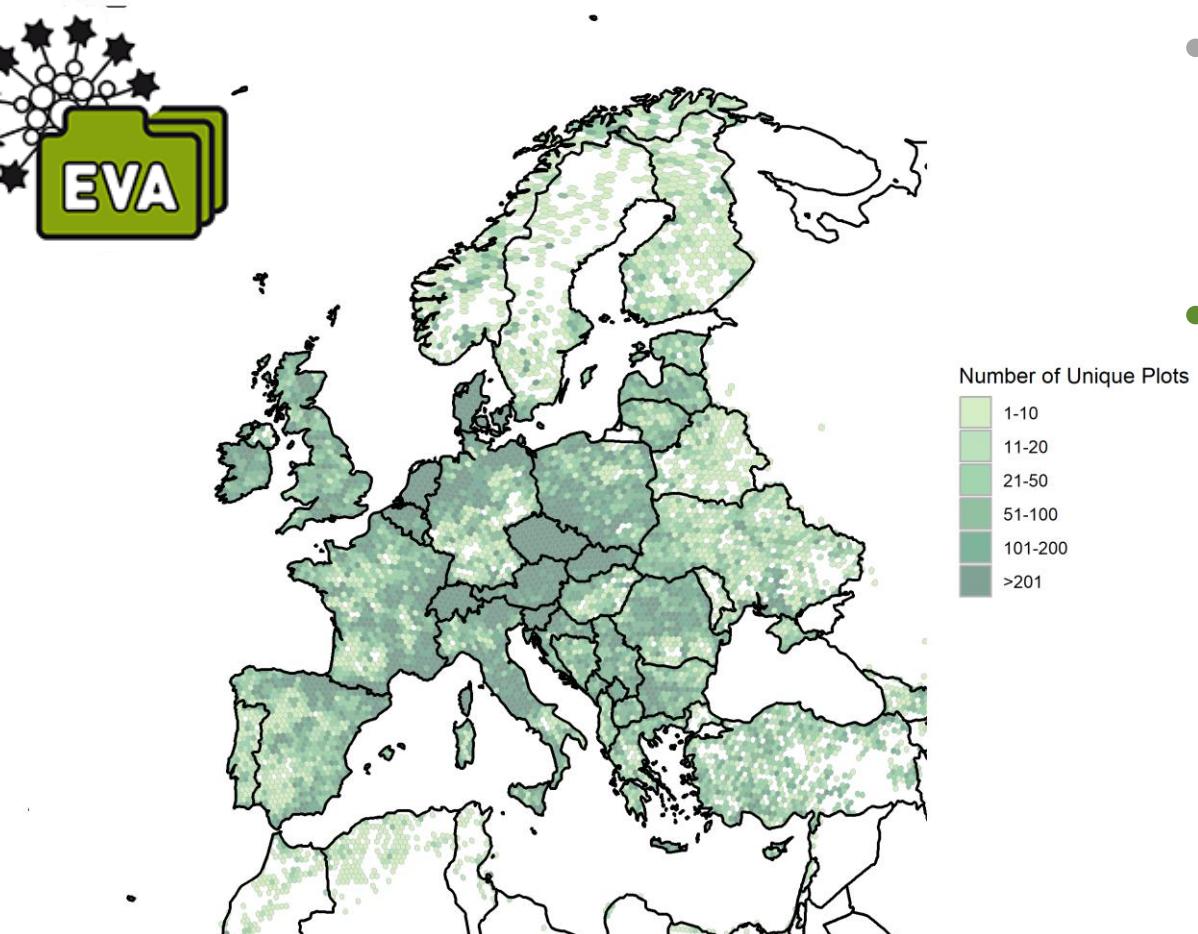
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- **Data for ASAAS compiles 1.9 million vegetation plots from more than 100 regional databases across Europe**

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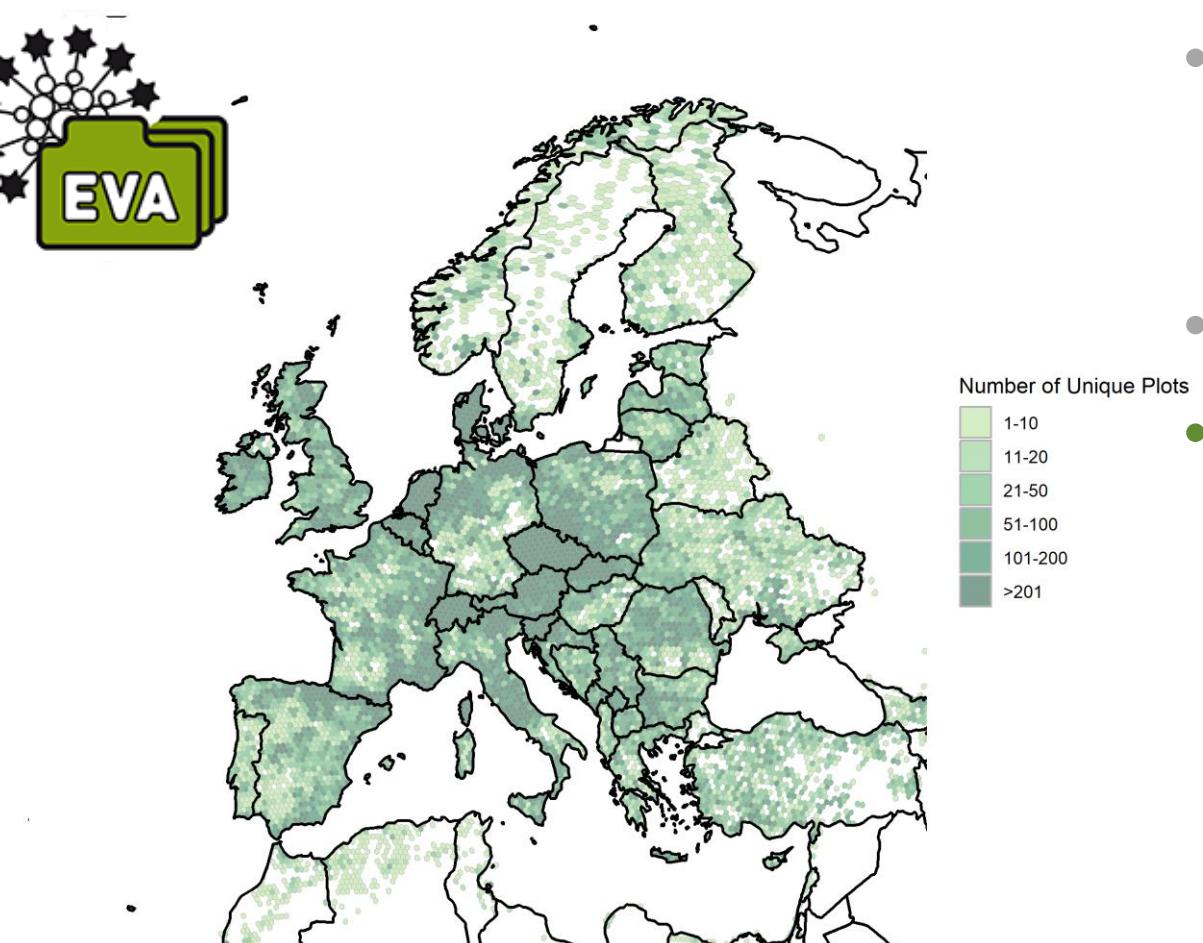
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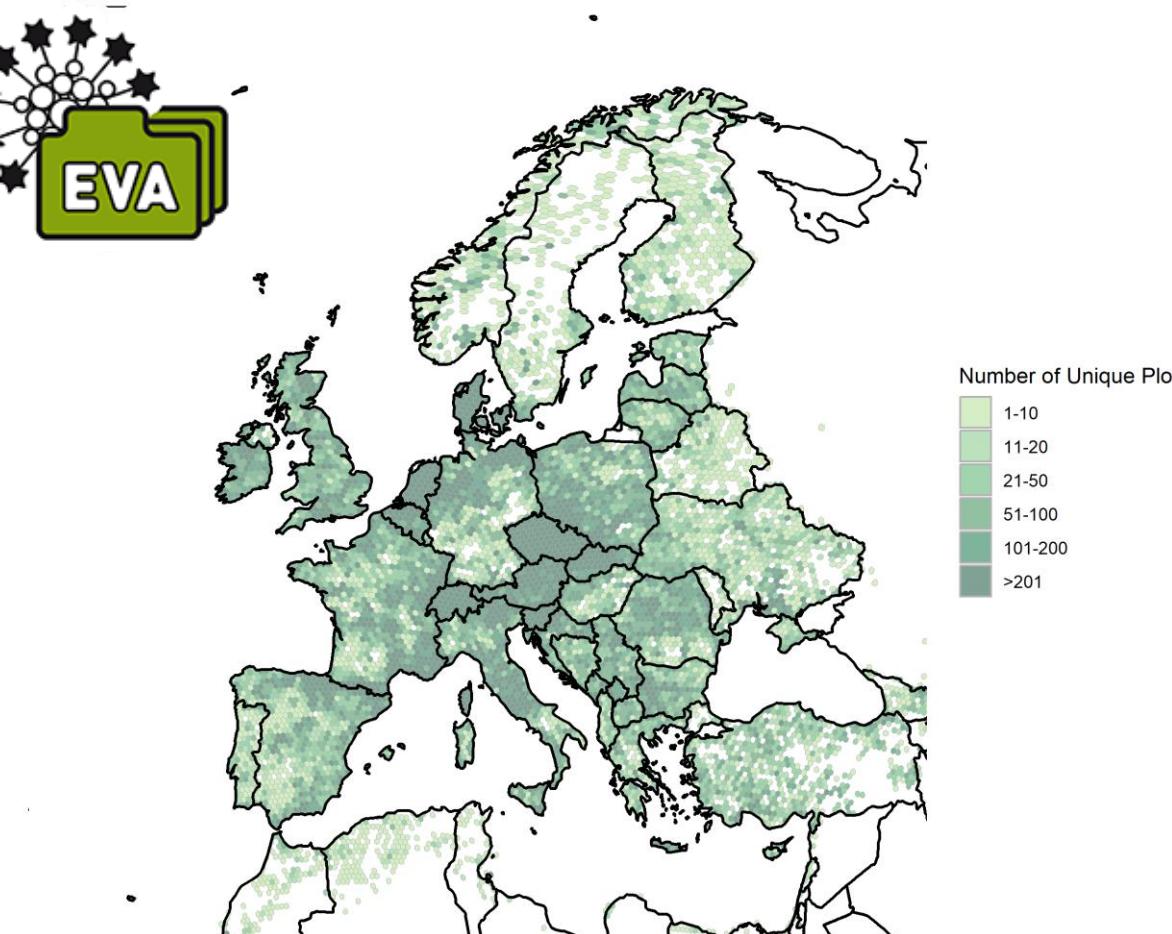
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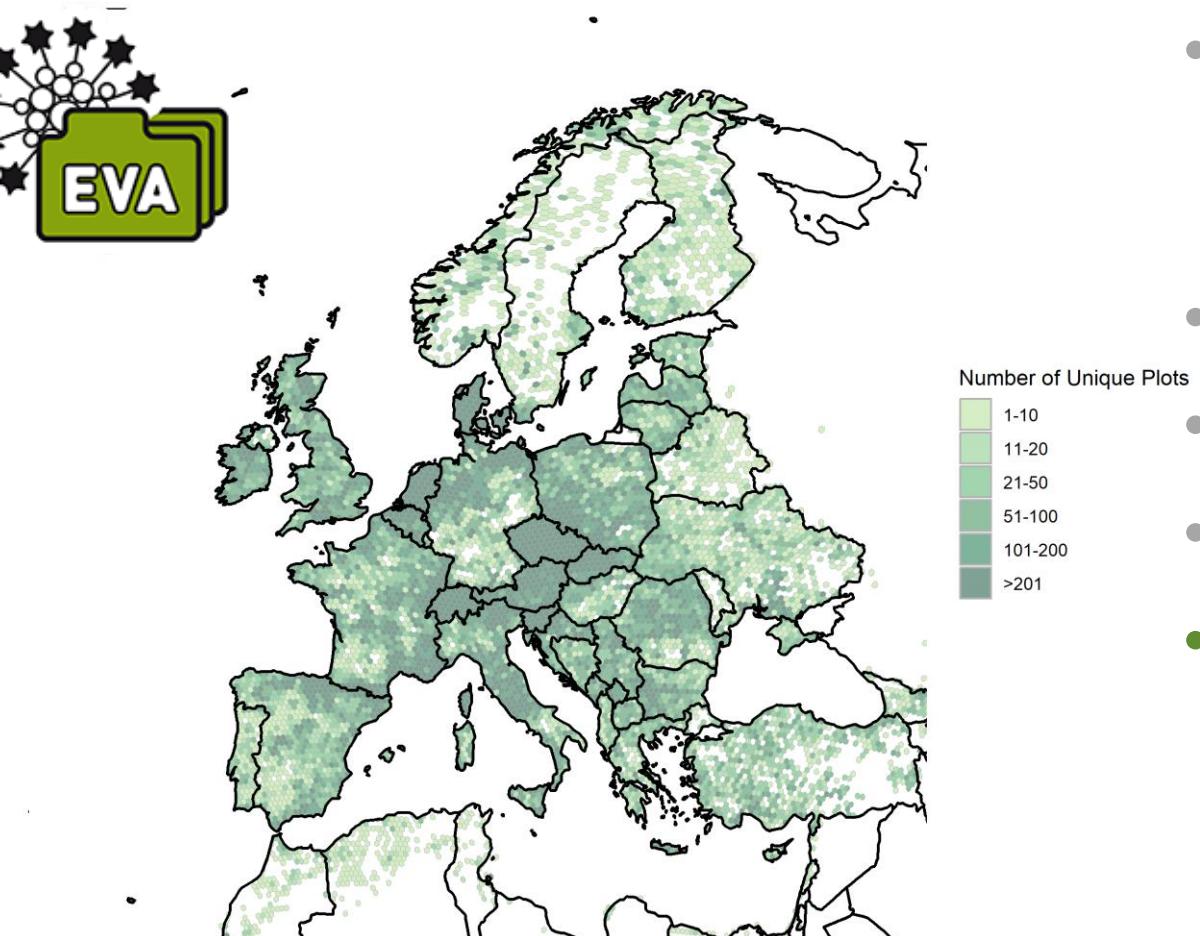
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  - **EUNIS classification**

# 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?

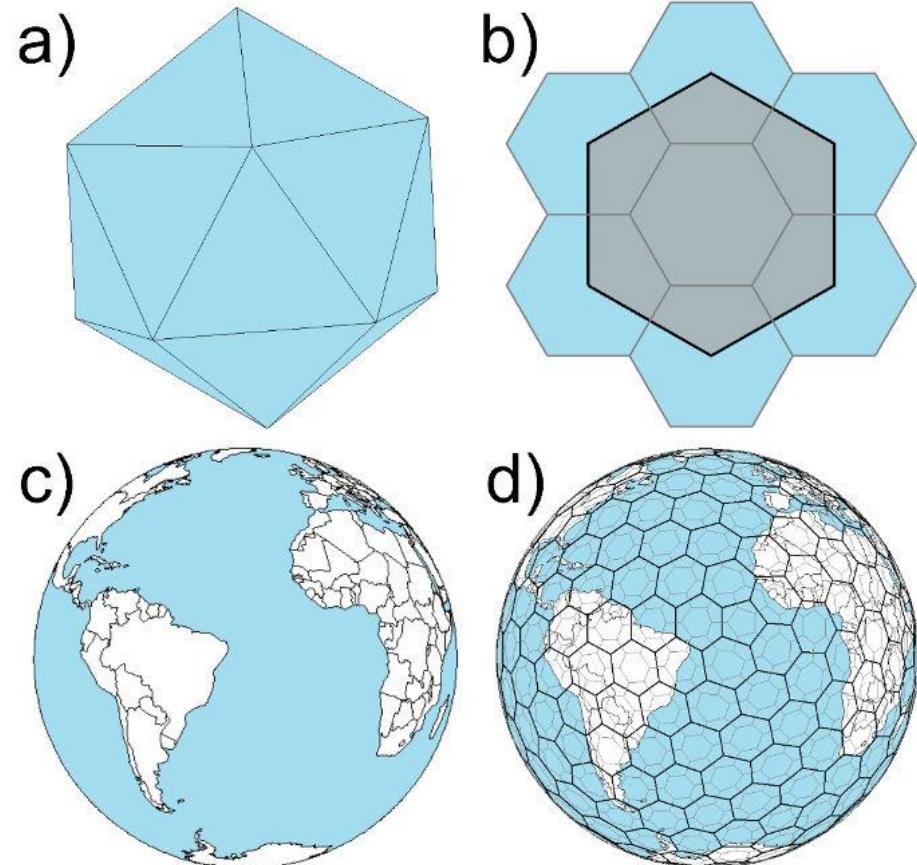
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# How to measure Habitat profiles?

- **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

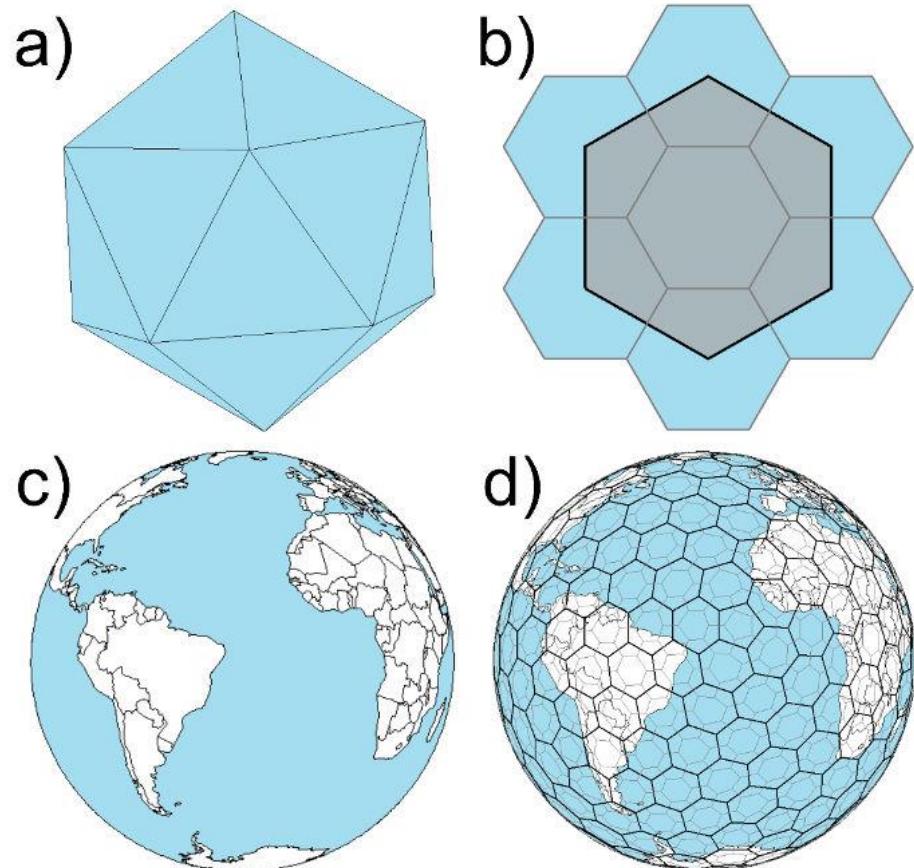
## Comparing Habitat use across 863 km<sup>2</sup> x 10 years hexcells



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

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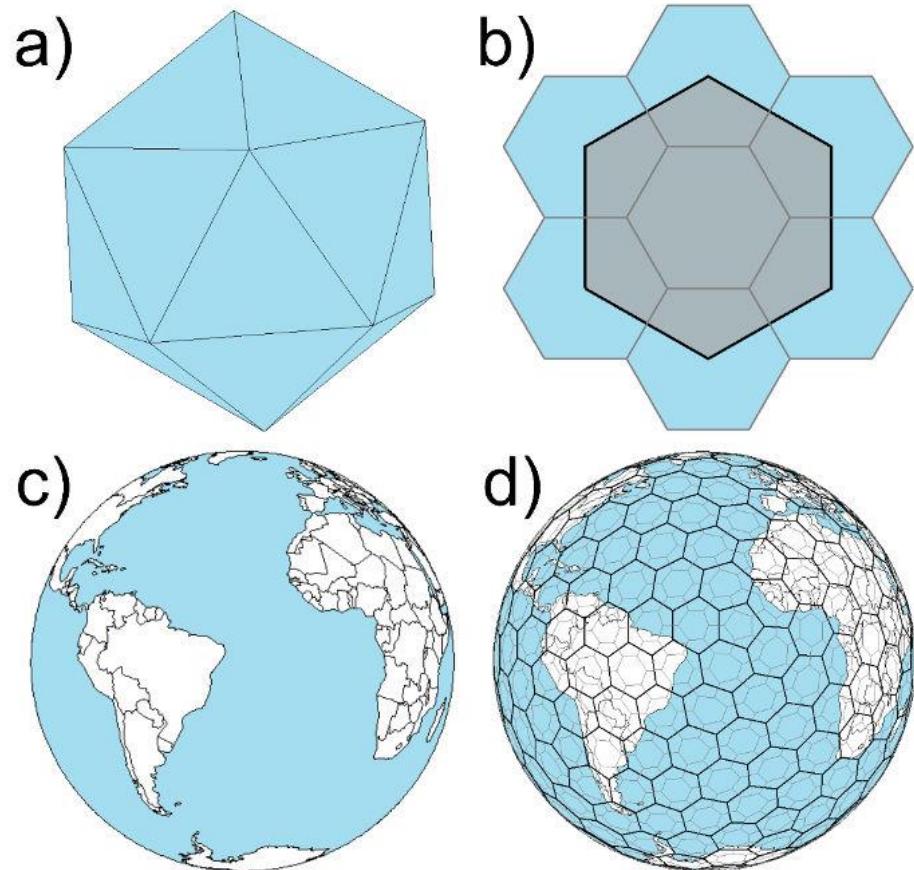
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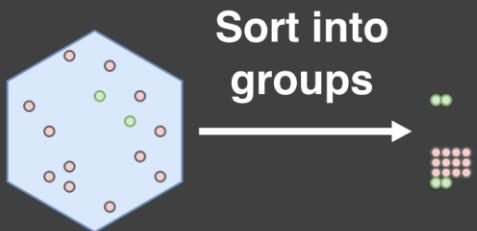
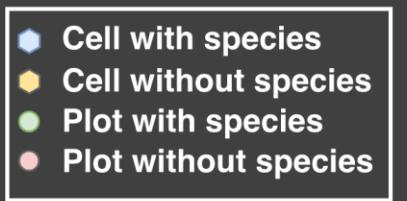
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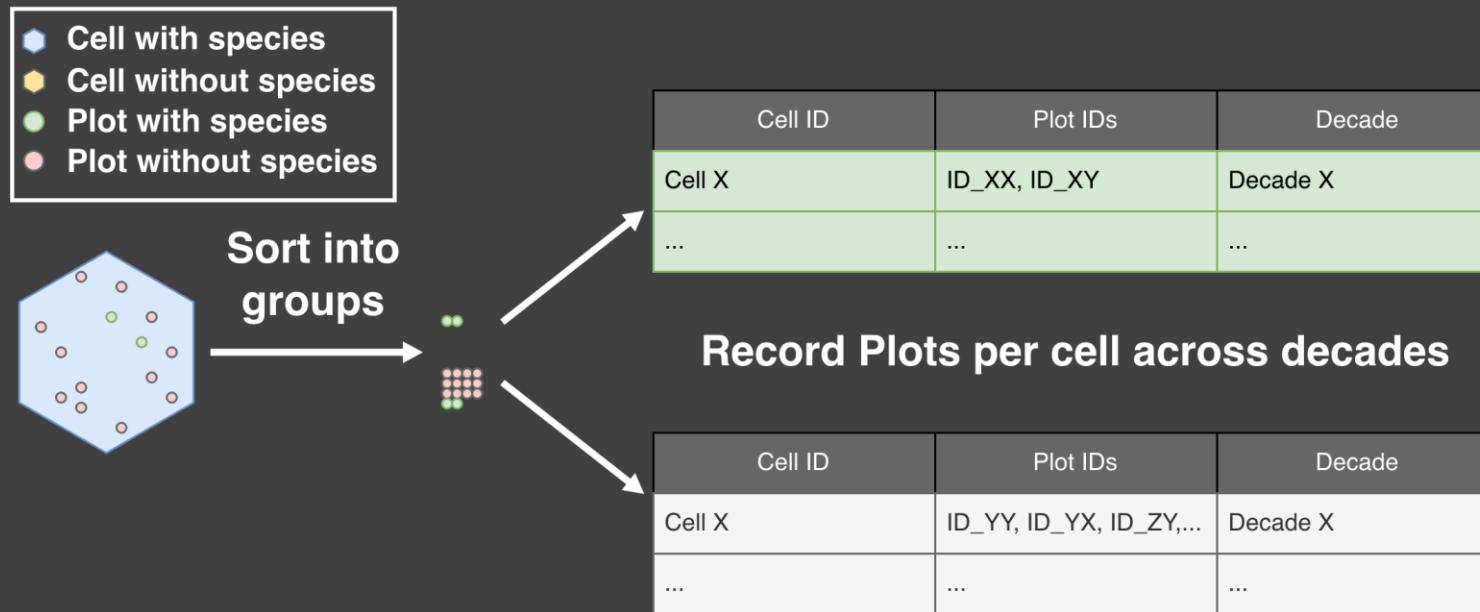


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- Within each **hexcell** we collect:
  - **Neophyte** habitat distribution (**p**)
  - **Background** habitat distribution (**q**)
- Within each habitat (per decade):
  - Null-model:  $p \approx q$
  - Overrepresented:  $p > q$

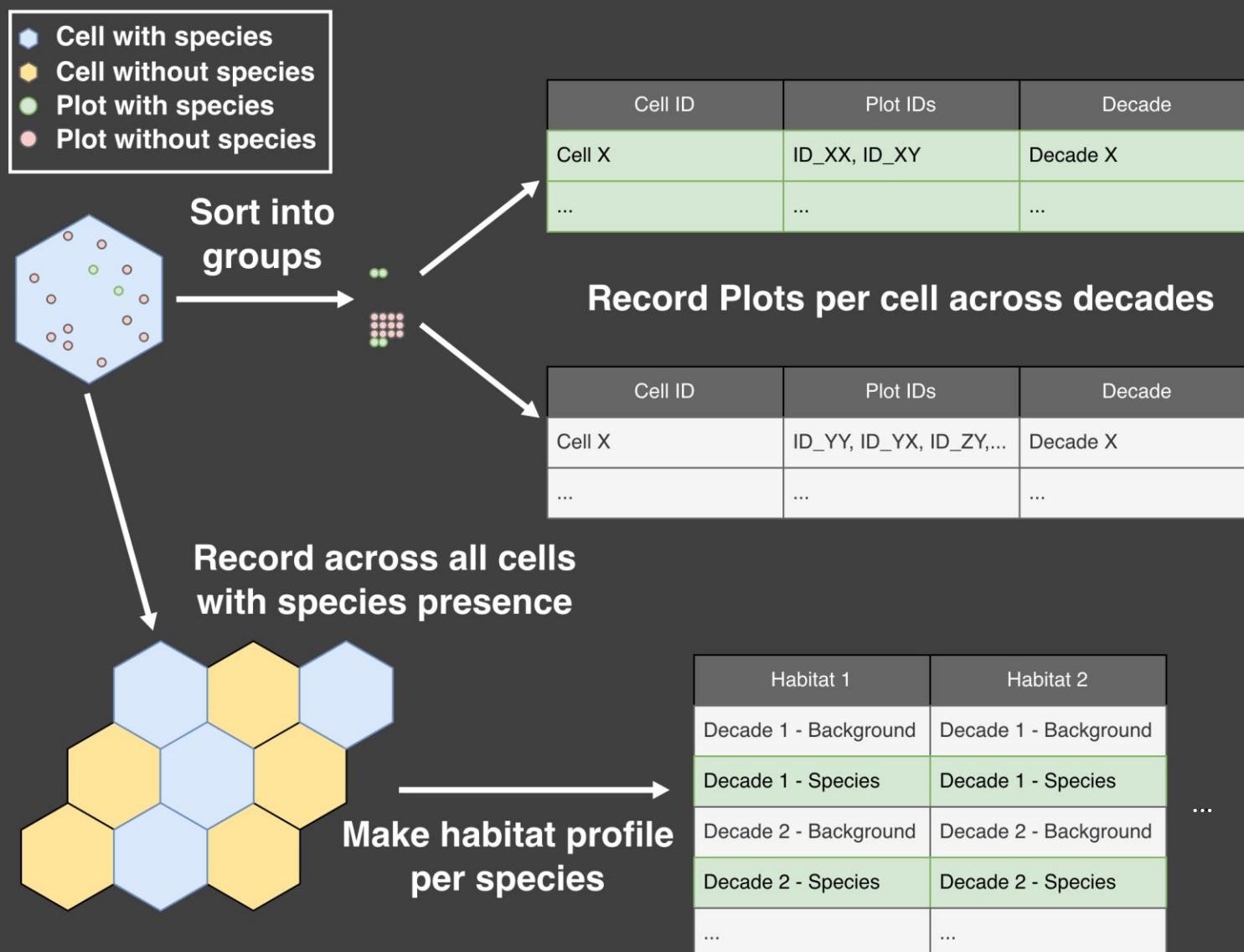
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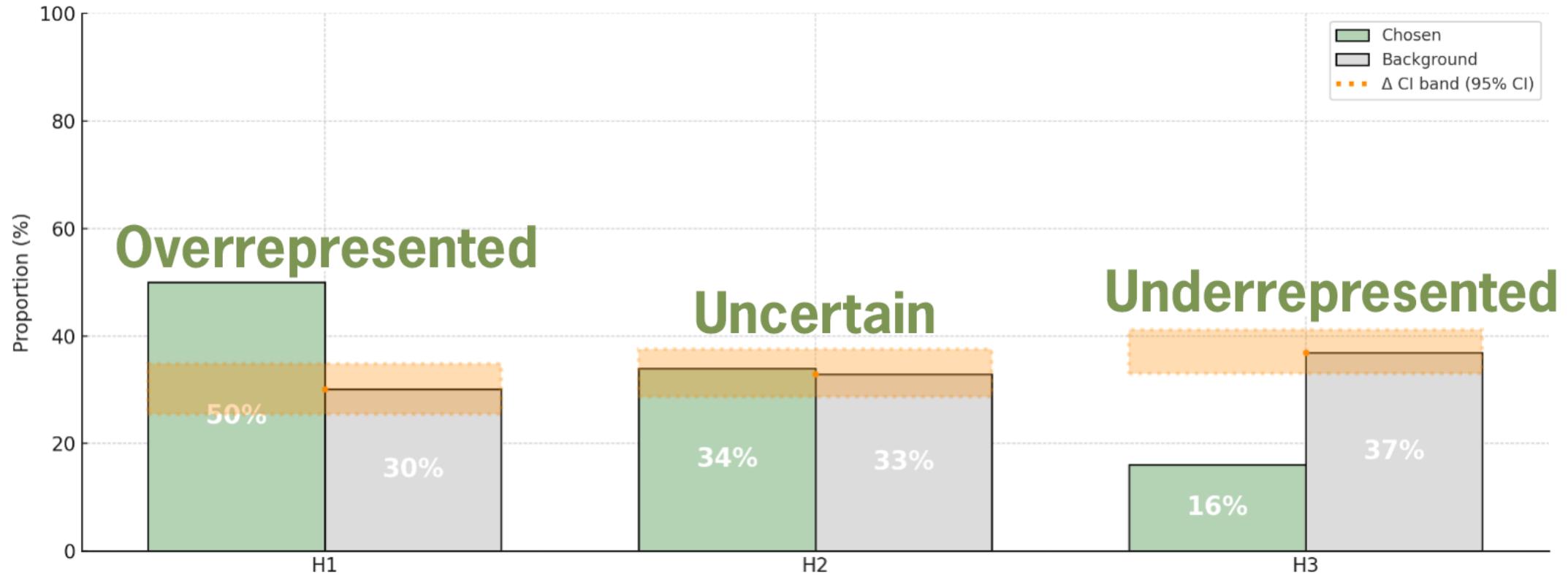
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**Species (green) vs background (grey) with 95% CI band**

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  - Interpretation: “**Does the species prefer this habitat?**”

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- **Local comparison**
  - 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)
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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)

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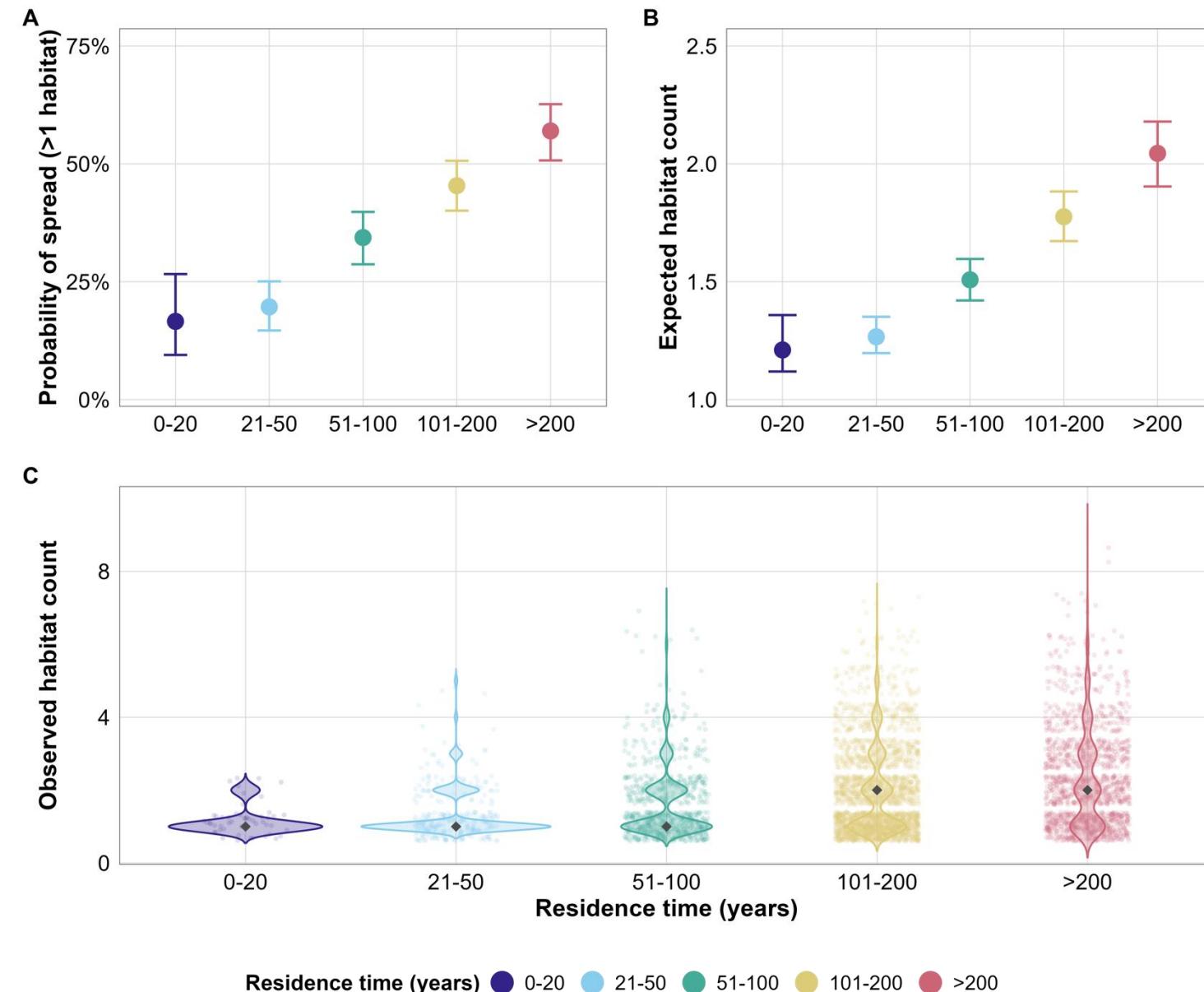
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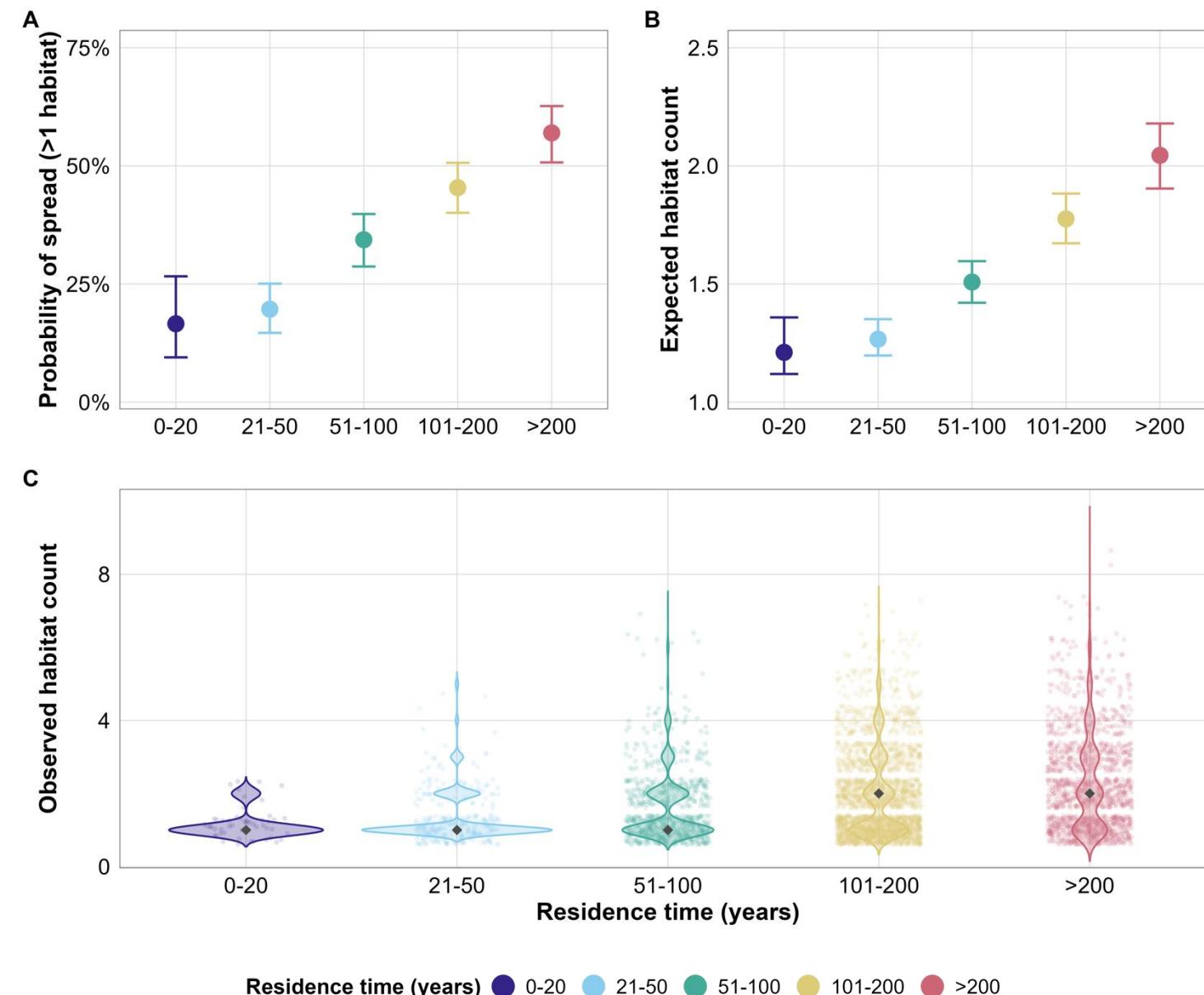
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- **Random effects:** country intercepts

# Results



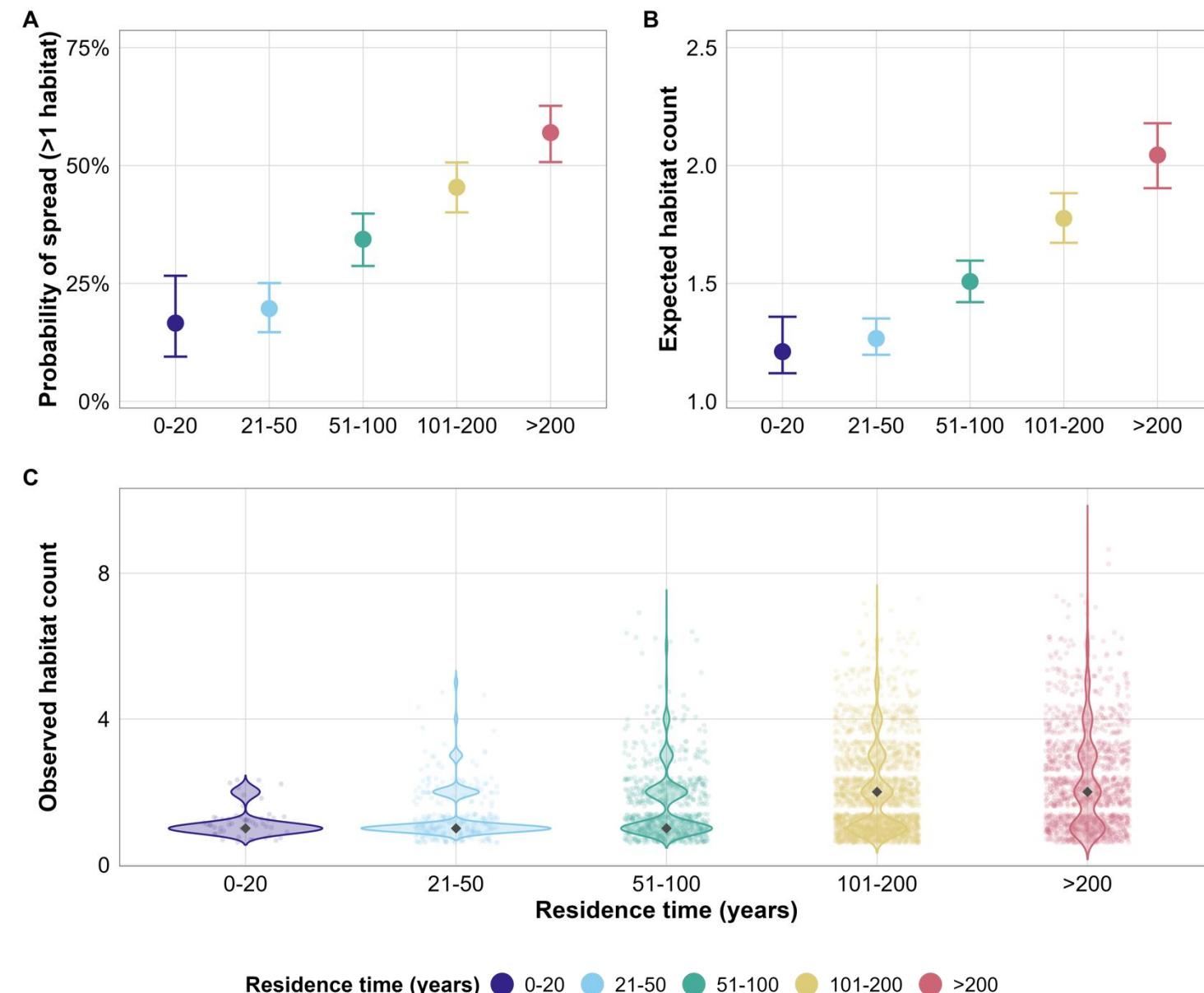
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  - Most pairwise contrasts significant

# 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)

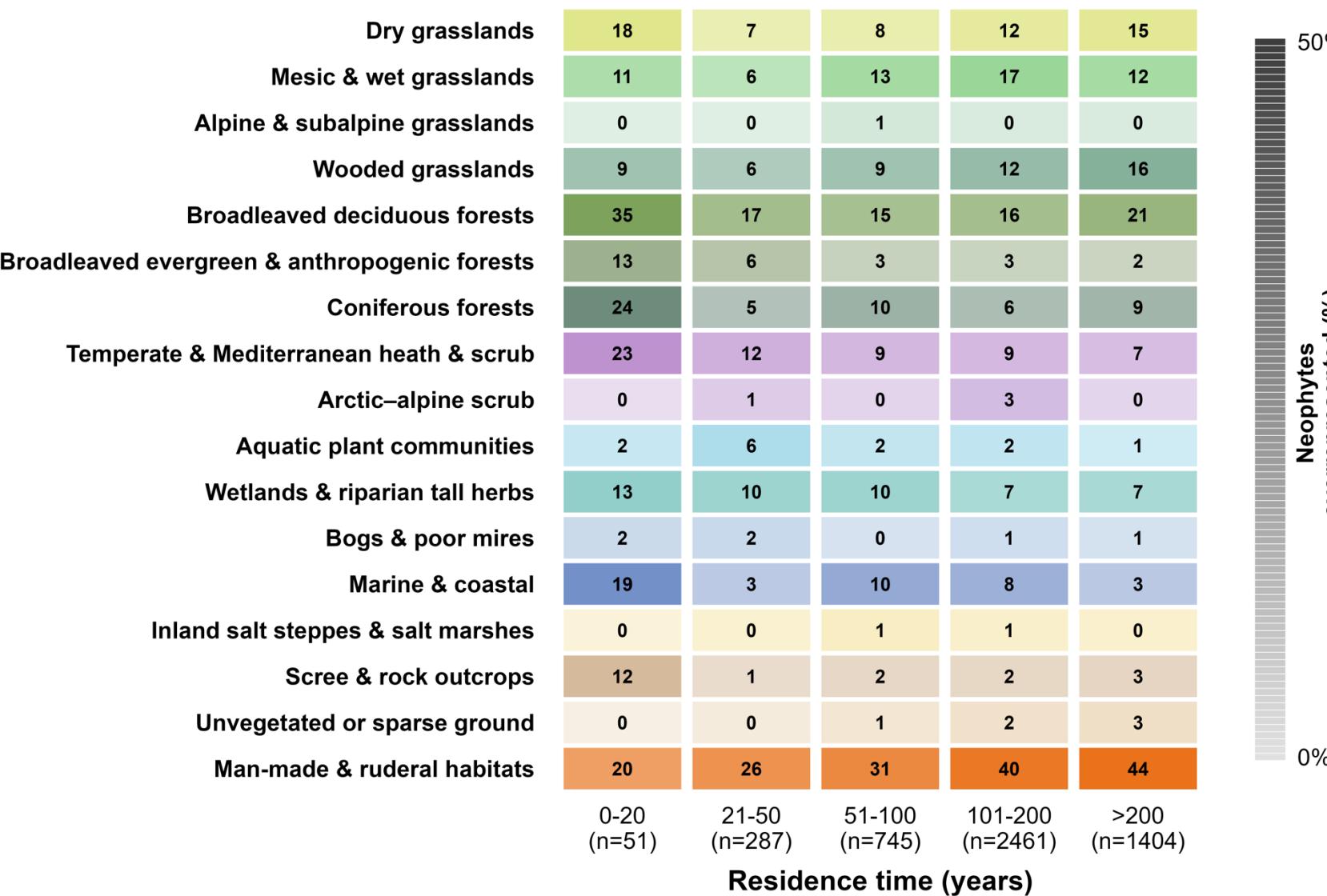
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  - Long-lived perennials: reduced prob. (OR 0.55)
  - Human-mediated dispersal: reduced prob. (OR 0.73)
- **Count component (expected habitats given > 1)**
  - Long-lived perennials: +43% more habitats (IRR 1.43)
  - Medium-lived perennials: +20% more habitats (IRR 1.20)

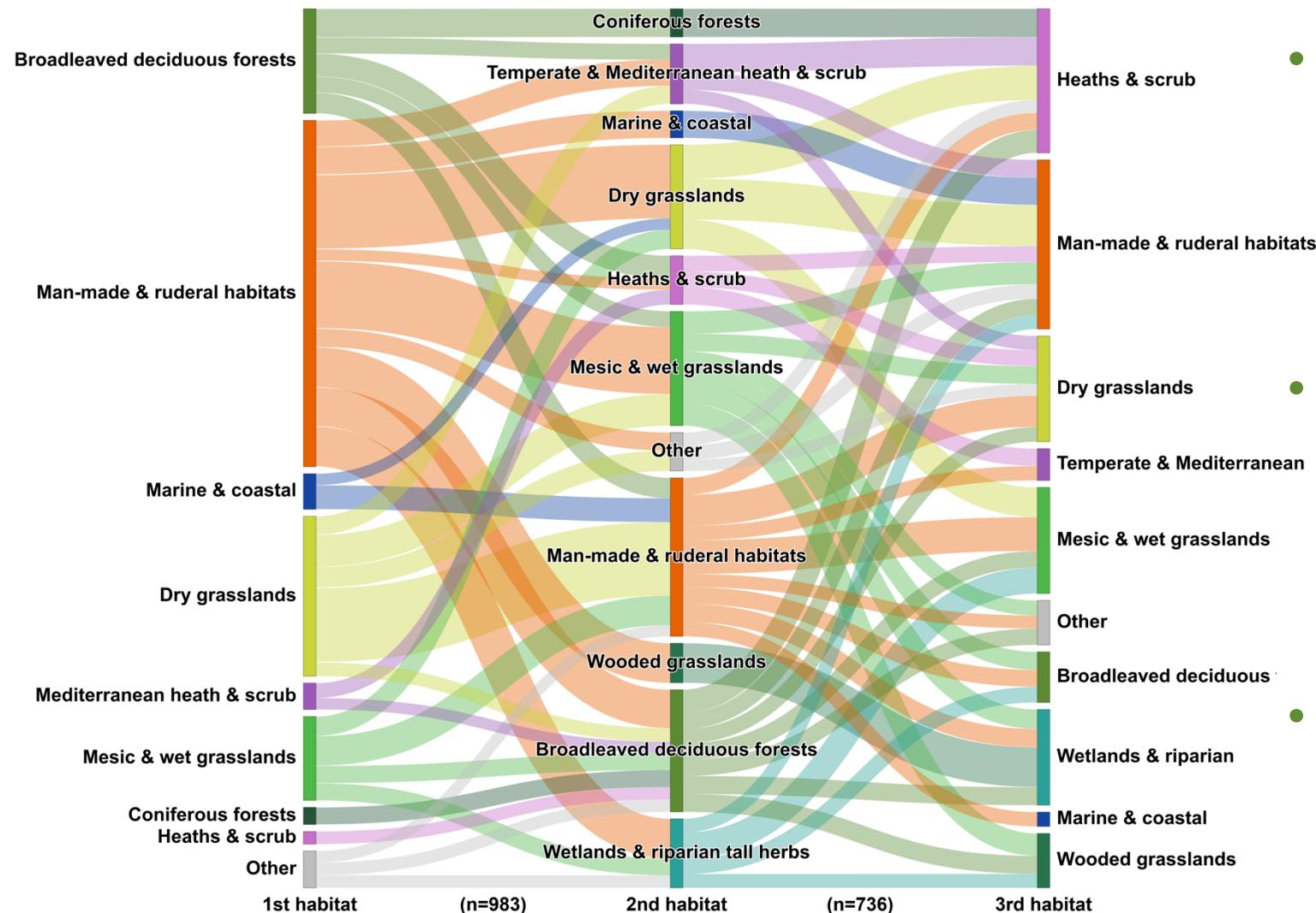
\*OR = Odds ratio, IRR = Incidence Rate Ratio

# 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

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

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  - 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 reaches upwards towards a bright sky. The stalks are light-colored with dark, horizontal bands where they meet the ground or other stalks. The perspective is from a low angle, looking up through the canopy.

Thank you