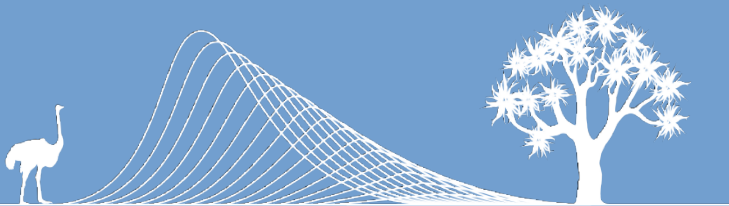


SEEC Stats Toolbox

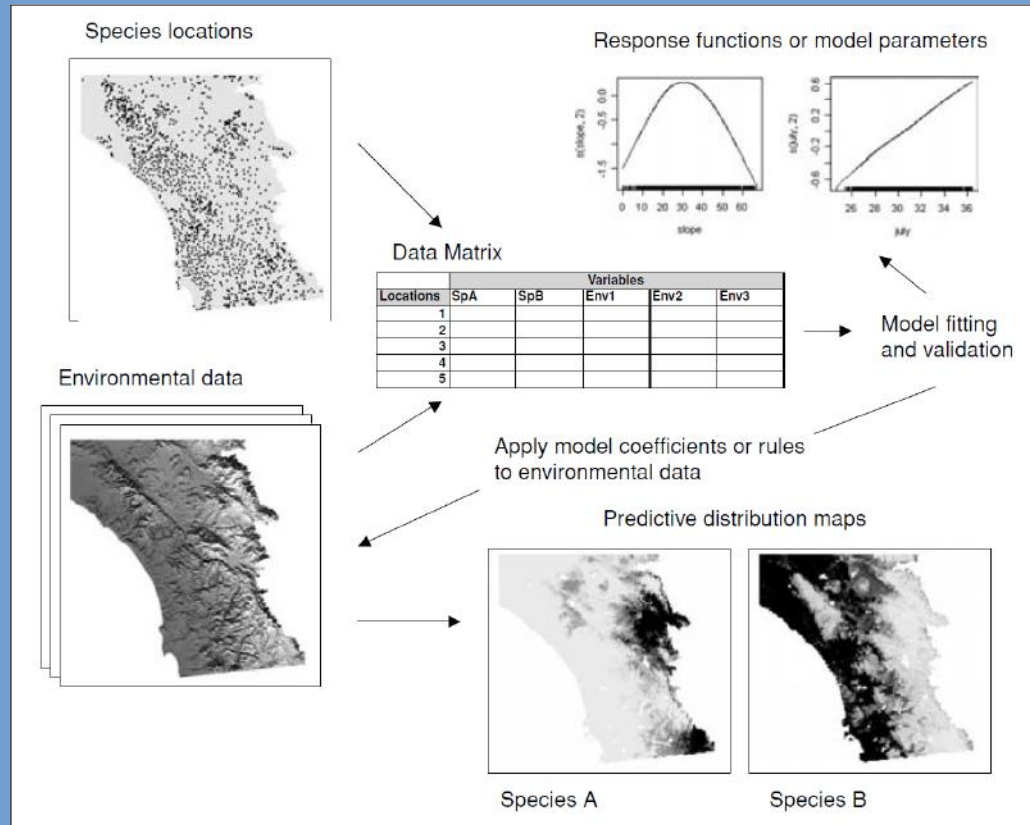
Species distribution modelling II:
using expert range maps and other spatial
information to supplement biased
occurrence data



Setting the scene

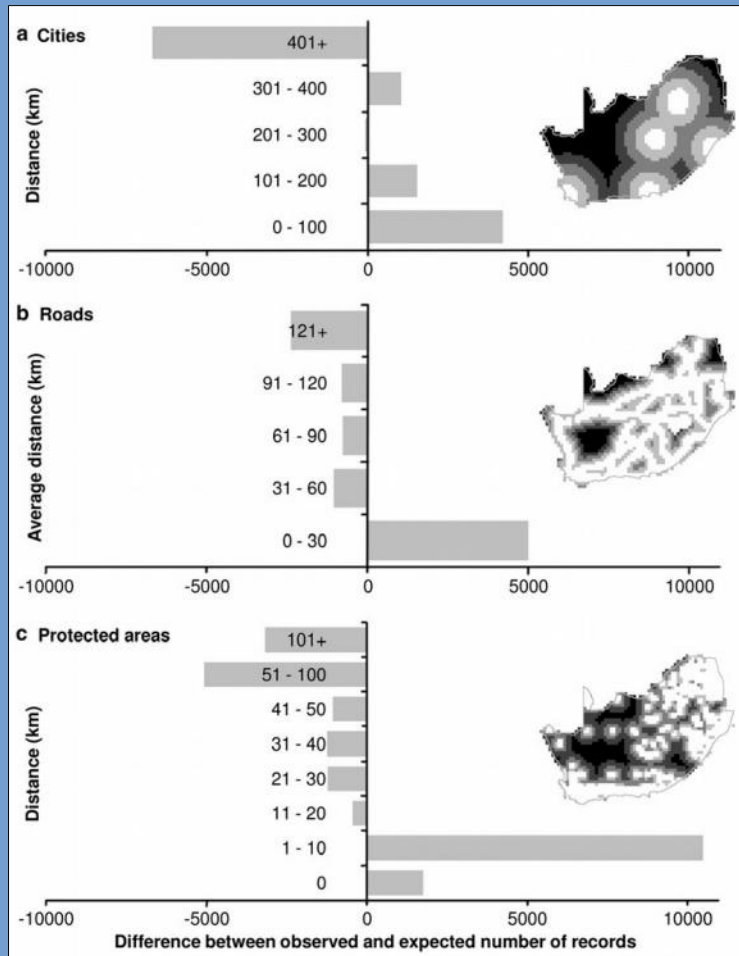


What is a species distribution model?



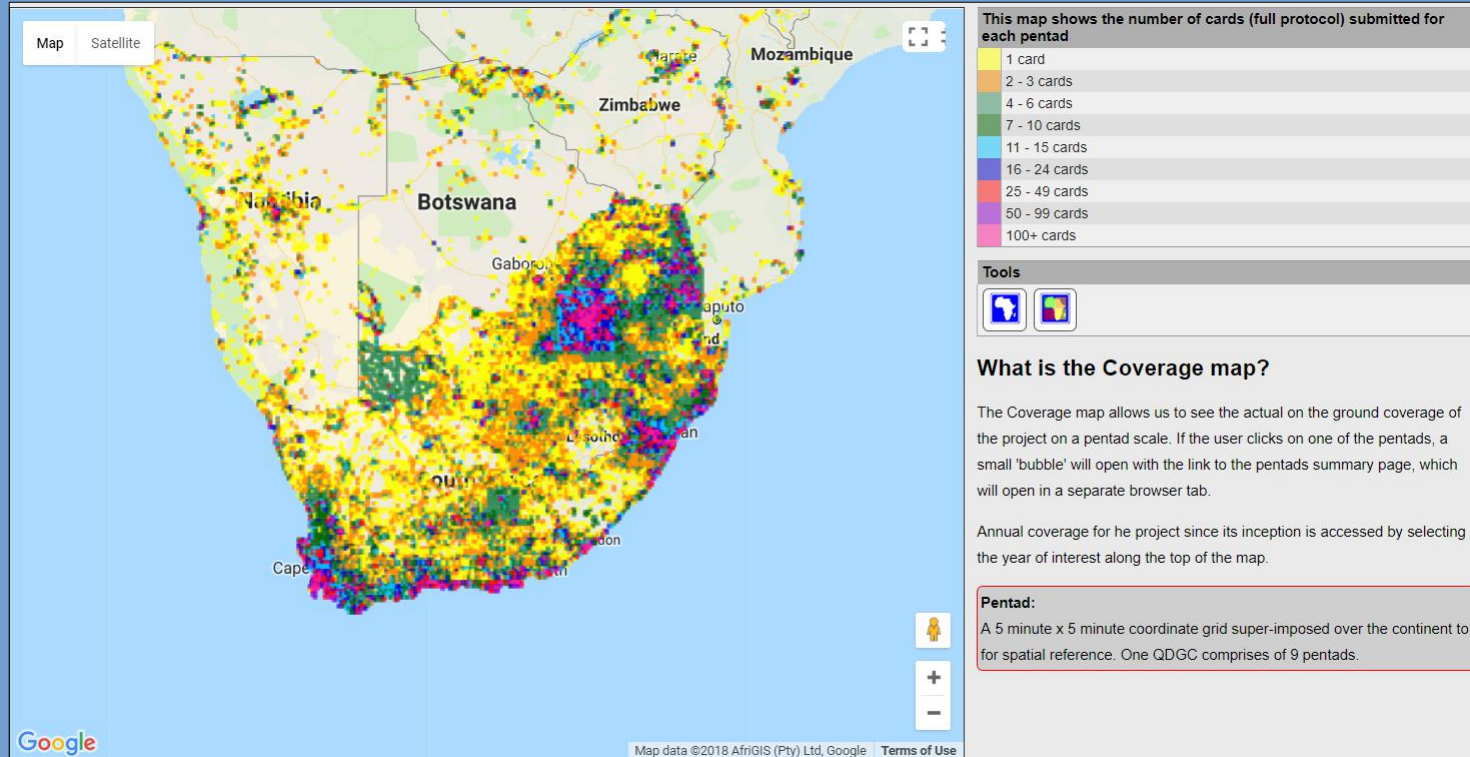
Franklin (2009). Mapping species distributions. Cambridge University Press

Biased occurrence data



Botts et al. (2011) *Biodiv Cons* 20

Biased occurrence data



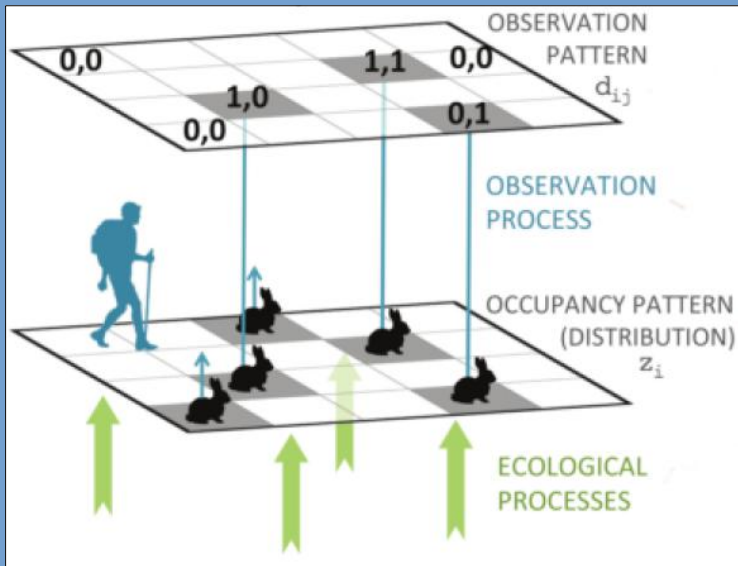
Occupancy

$$\text{Occupancy} = \psi \times p$$

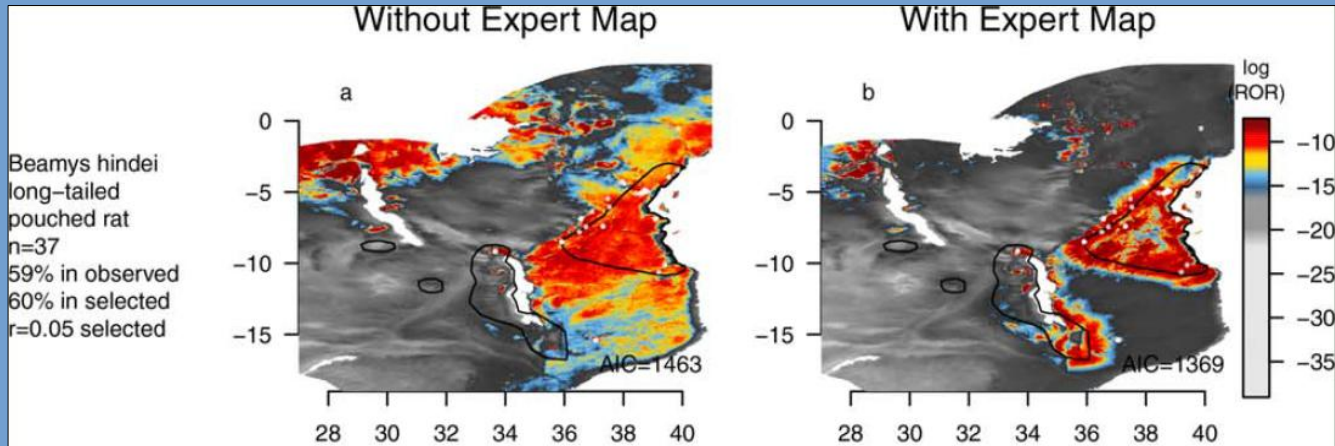
ψ is the probability a site is occupied

p is the probability of observing a species

SDMs usually model the probability a location contains a presence ($p \neq 1$)



SDMs tend to overpredict

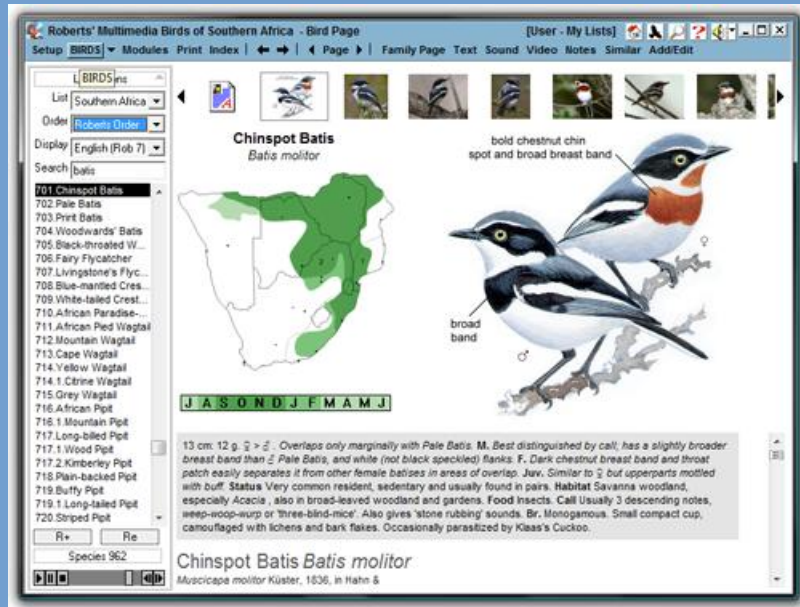


Dispersal limitations

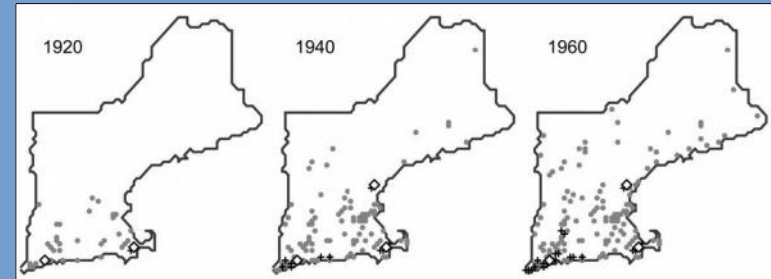
Invasions

Equilibrium with the environment

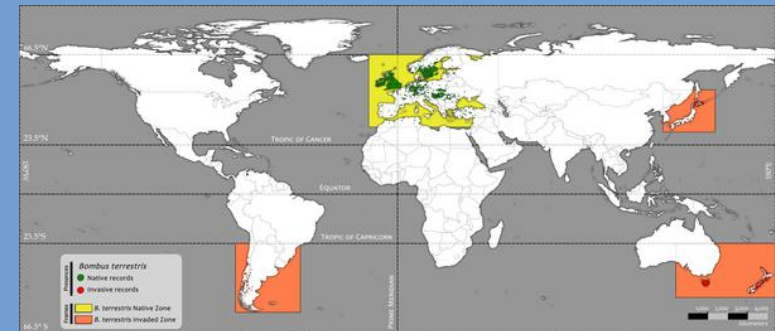
Incorporating other useful spatial information



Expert range maps



Dispersal info



Native or "other" range info

The approach



The approach

Global Ecology and Biogeography, (Global Ecol. Biogeogr.) (2016) 25, 1022–1036

MACROECOLOGICAL METHODS



Improving niche and range estimates with Maxent and point process models by integrating spatially explicit information

Cory Merow^{1*}, Jenica M. Allen², Matthew Aiello-Lammens^{3,4} and John A. Silander, Jr⁵

Global Ecology and Biogeography, (Global Ecol. Biogeogr.) (2017) 26, 243–258

MACROECOLOGICAL METHODS



Integrating occurrence data and expert maps for improved species range predictions

Cory Merow^{1,2*}, Adam M. Wilson^{1,3} and Walter Jetz^{1,4}

General approach - Minxent

Just a generalisation of Maxent

$$\ln[P^*(z(x_i))] = \ln(Q(x_i)) + z(x_i)\lambda - \ln(C)$$

Relative
occurrence rate
(ROR)

Offset

Vector of
environmental
predictors

Fitted
coefficients

Constant

General approach - Minxent

Just a generalisation of Maxent

$$\ln[P^*(z(x_i))] = \ln(Q(x_i)) + z(x_i)\lambda - \ln(C)$$

Relative
occurrence rate
(ROR)

Offset

Fitted
coefficients

Vector of
environmental
predictors

Constant

General approach - Minxent

- 1) Build an offset (Maxent model using prior spatial information)
 - a) Nuisance offset (factor out of the prediction)
 - b) Informative offset (include in the prediction)
- 2) Build a normal Maxent SDM
- 3) Build Minxent SDM (by factoring offset in or out)
- 4) Compare 3) and 4) (optional)

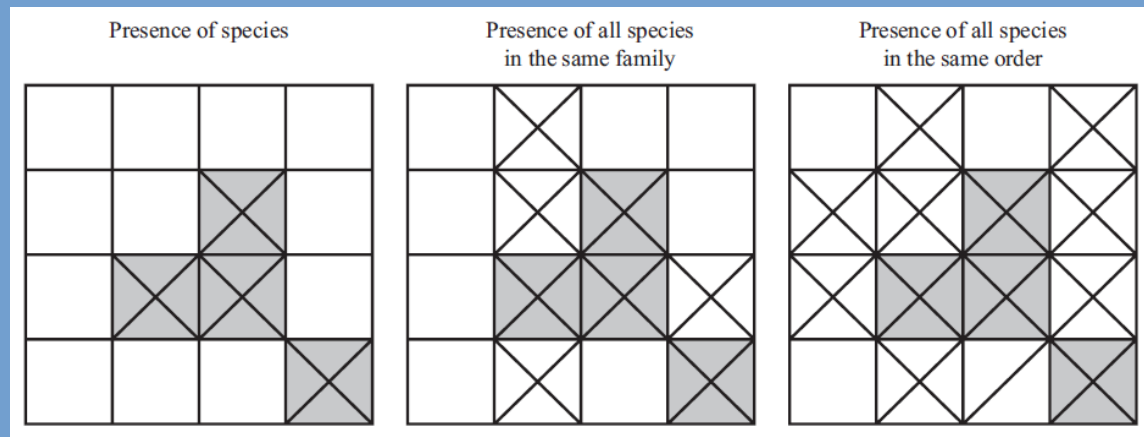
Example 1: Accounting for sampling bias



App1_Sampling_Bias.r in Merow *et al.* (2016)

Accounting for sampling bias

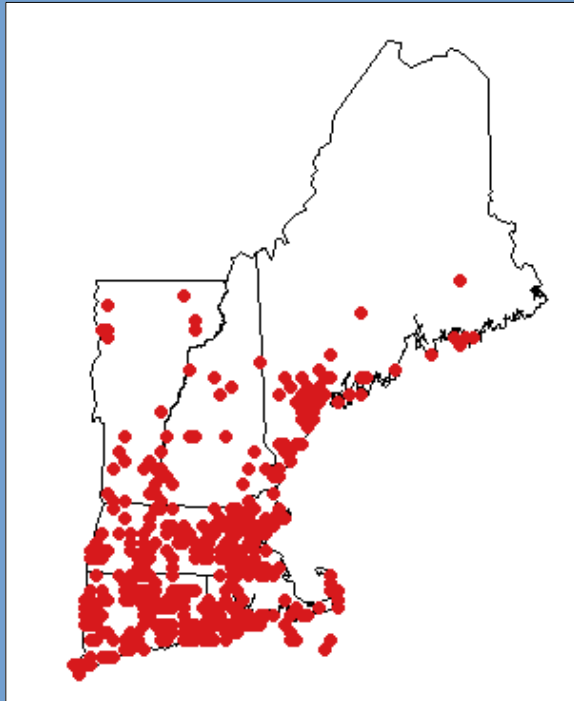
Approach used for a long time – Target Group Sampling (TGS)



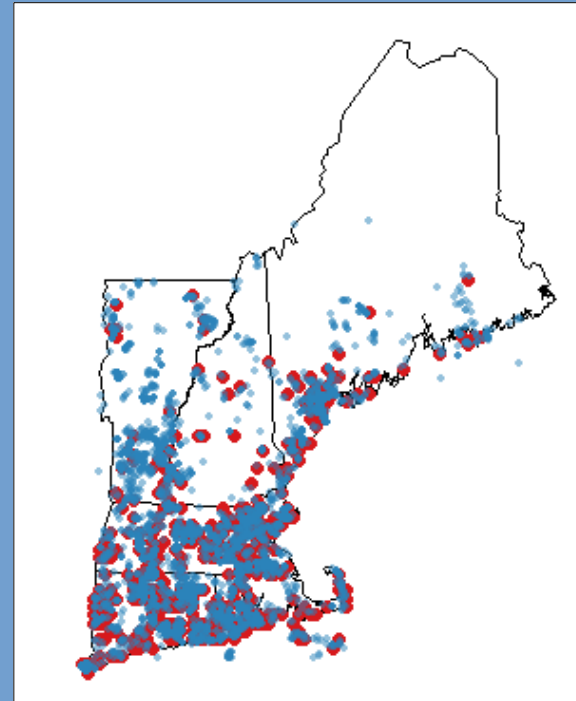
Niamir et al. (2016) *Glob Ecol Biogeog* 25

Step-by-step sampling bias

Step 1: Get species occs and ID target group samples



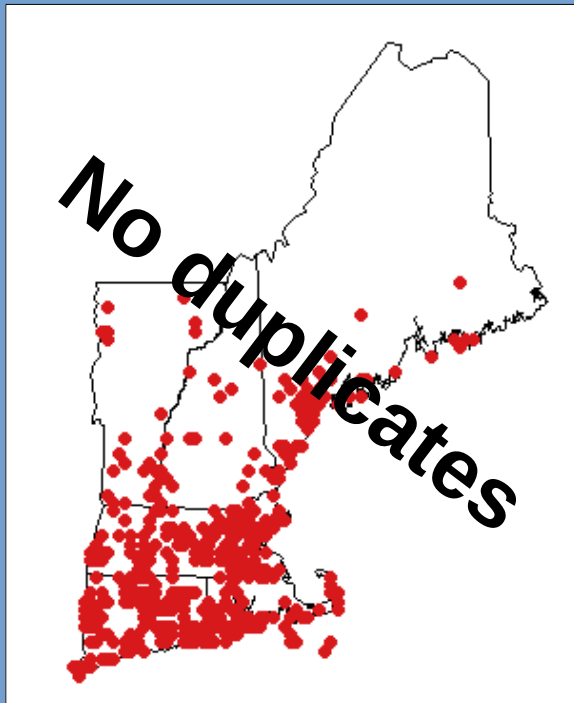
Celastrus orbiculatus



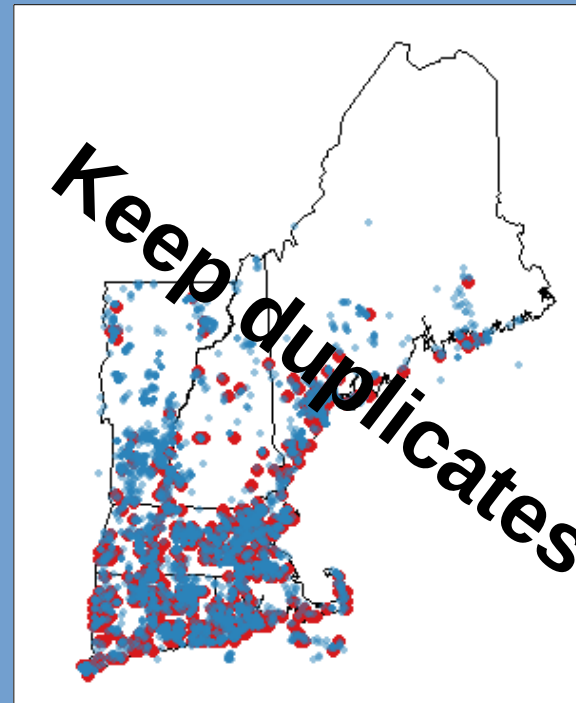
Other invasive
species' occurrences

Step-by-step sampling bias

Step 1: Get species occs and ID target group samples

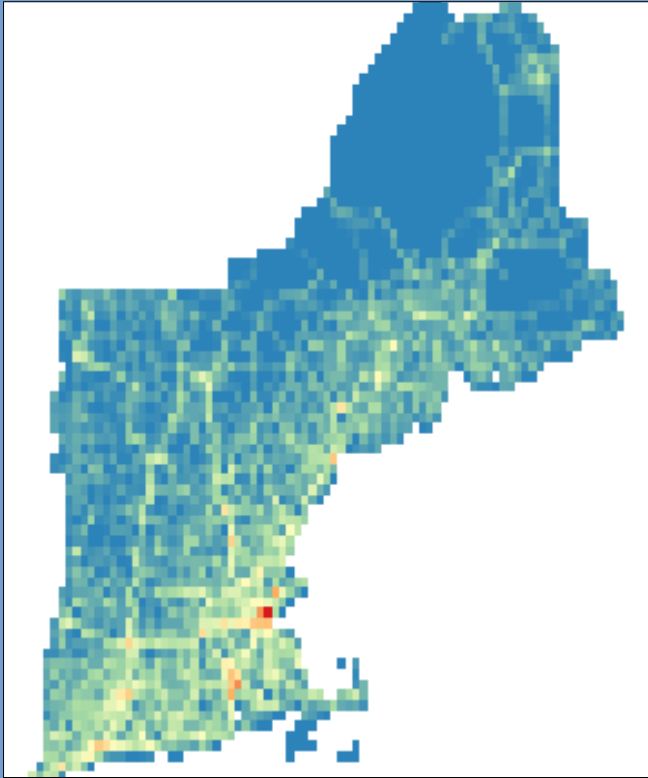


Celastrus orbiculatus

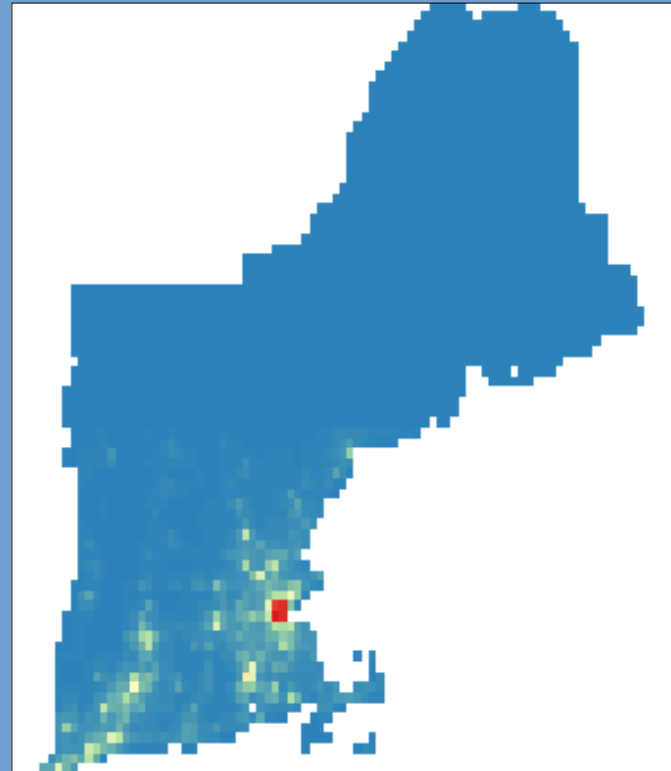


Other invasive
species' occurrences

Step 2: Get predictors of sampling bias

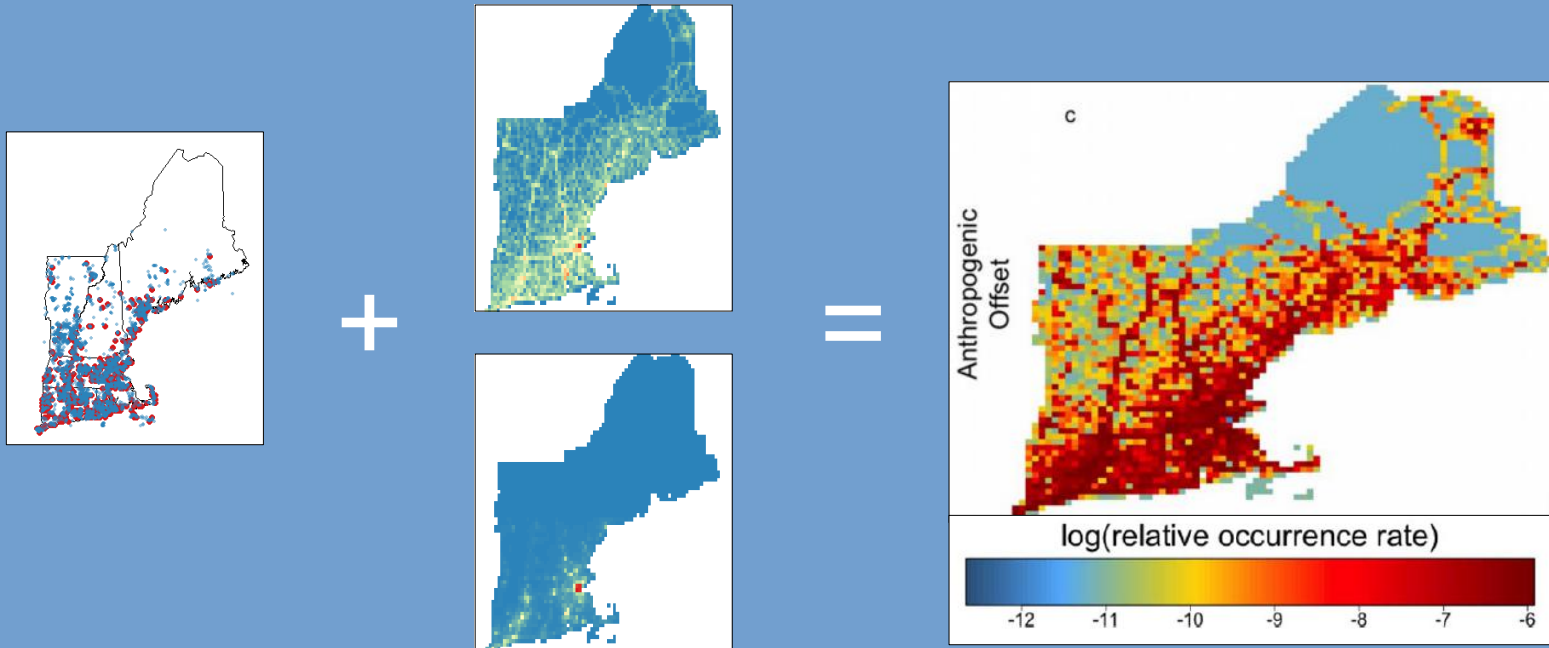


Road density

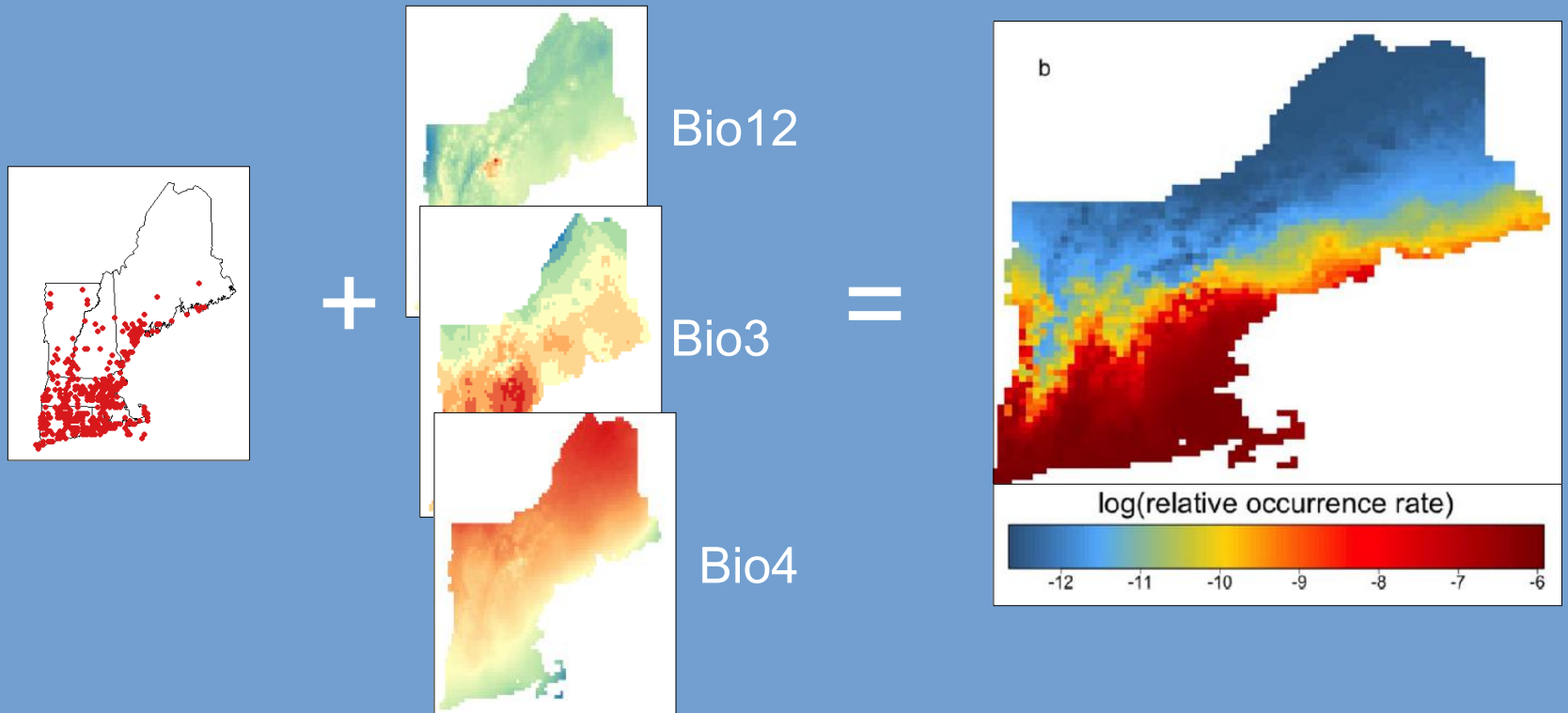


Population density

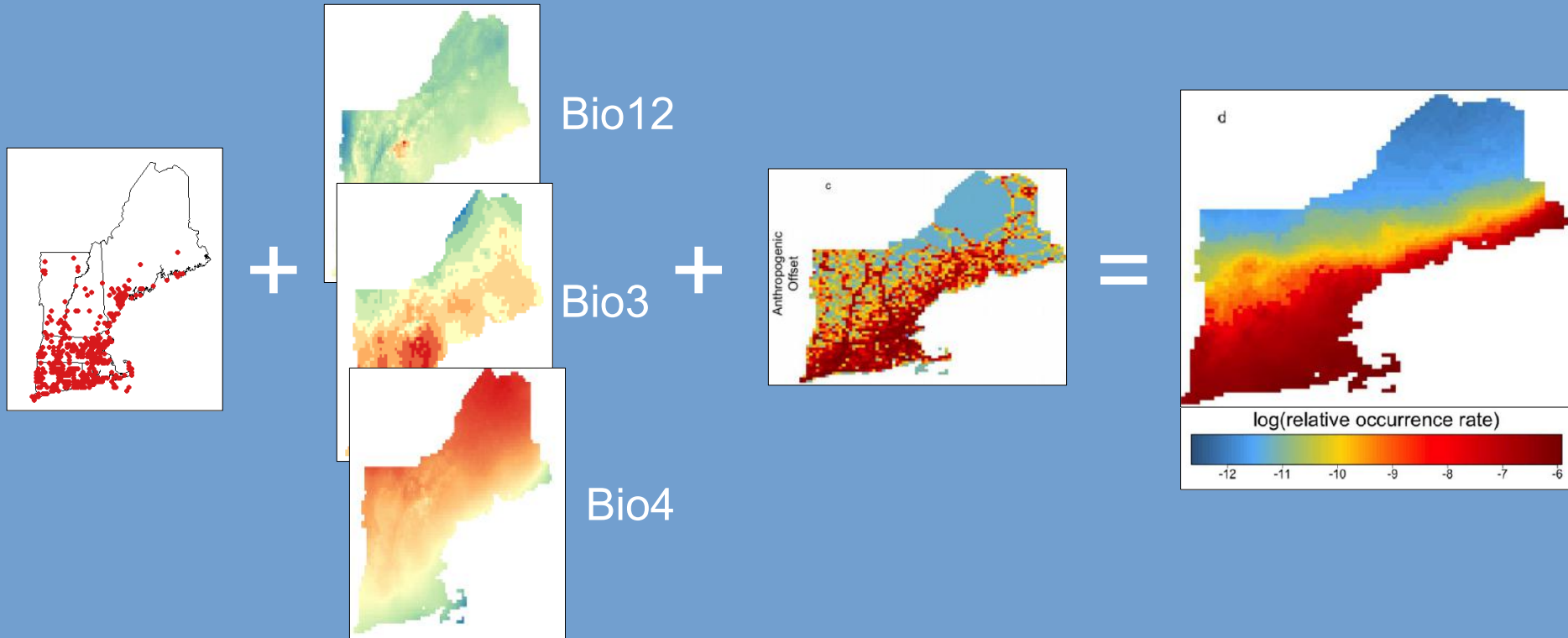
Step 3: Build a sampling bias model



Step 4: Build a normal Maxent model for your species



Step 5: Build Maxent model with an offset (your sampling bias model as a bias grid)



Example 2: Using expert range maps



App5_Expert_Maps.r in Merow *et al.* (2016)
Merow *et al.* (2017)
bossMaps R package

Expert knowledge vs messy data

Roberts' Multimedia Birds of Southern Africa - Bird Page

Setup: BIRDS | Modules | Print Index | Page | Family Page Text Sound Video Notes Similar Add/Edit

List: Southern Africa

Order: Roberts Order

Display: English (Rob 7)

Search: batis

701 Chinspot Batis

702 Pale Batis

703 Print Batis

704 Woodwards' Batis

705 Black-throated W...

706 Fairy Flycatcher

707 Livingstone's Flyc...

708 Blue-mantled Crea...

709 White-tailed Crest...

710 African Paradise...

711 African Pied Wagtail

712 Mountain Wagtail

713 Cape Wagtail

714 Yellow Wagtail

714.1 Citrine Wagtail

715 Grey Wagtail

716 African Pipit

716.1 Mountain Pipit

717 Long-billed Pipit

717.1 Wood Pipit

717.2 Kimberley Pipit

718 Plain-backed Pipit

719 Buffy Pipit

719.1 Long-tailed Pipit

720 Striped Pipit

Chinspot Batis *Batis molitor*

bold chestnut chin spot and broad breast band

broad band

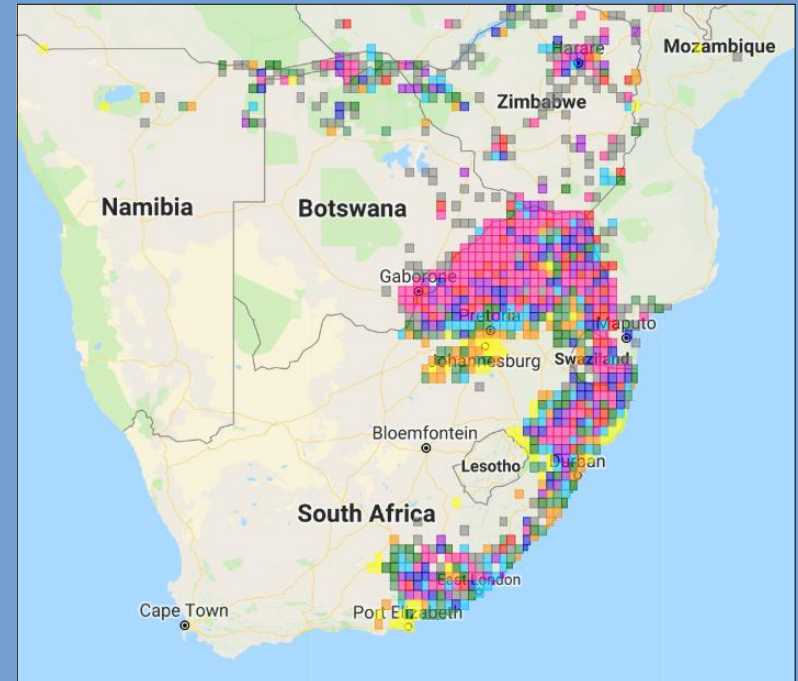
J A S O N D J F M A M J

13 cm. 12 g. ♂ > ♀. Overlaps only marginally with Pale Batis. M. Best distinguished by call; has a slightly broader breast band than ♀. Pale Batis, and white (not black speckled) flanks. F. Dark chestnut breast band and throat patch easily separates it from other female batises in areas of overlap. Juvs. Similar to ♀ but upperparts mottled with buff. Status: Very common resident, sedentary and usually found in pairs. Habitat: Savanna woodland, especially Acacia, also in broad-leaved woodland and gardens. Food: Insects. Call: Usually 3 descending notes, weep-woop-woop or 'three-blind-mice'. Also gives 'stone rubbing' sounds. Br. Monogamous. Small compact cup, camouflaged with lichens and bark flakes. Occasionally parasitized by Klaas's Cuckoo.

Chinspot Batis *Batis molitor*

Muscicapa molitor Küster, 1836, in Hahn &

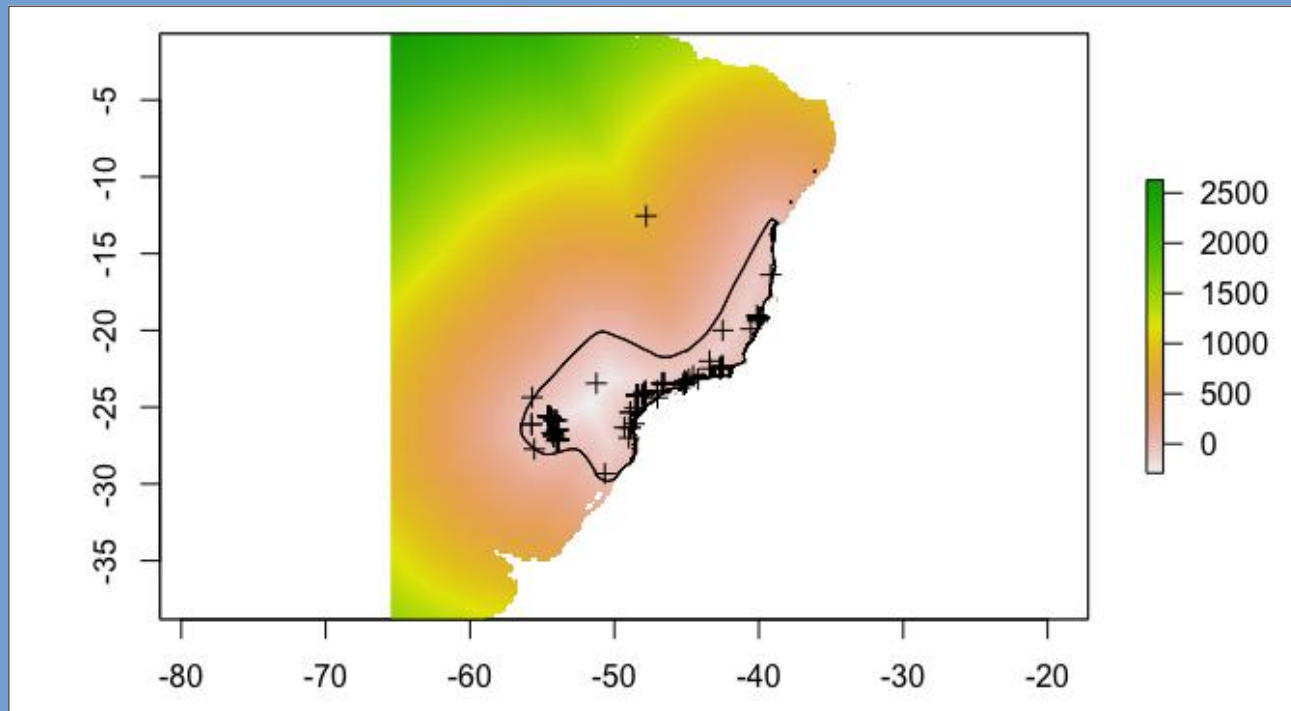
Roberts' Birds



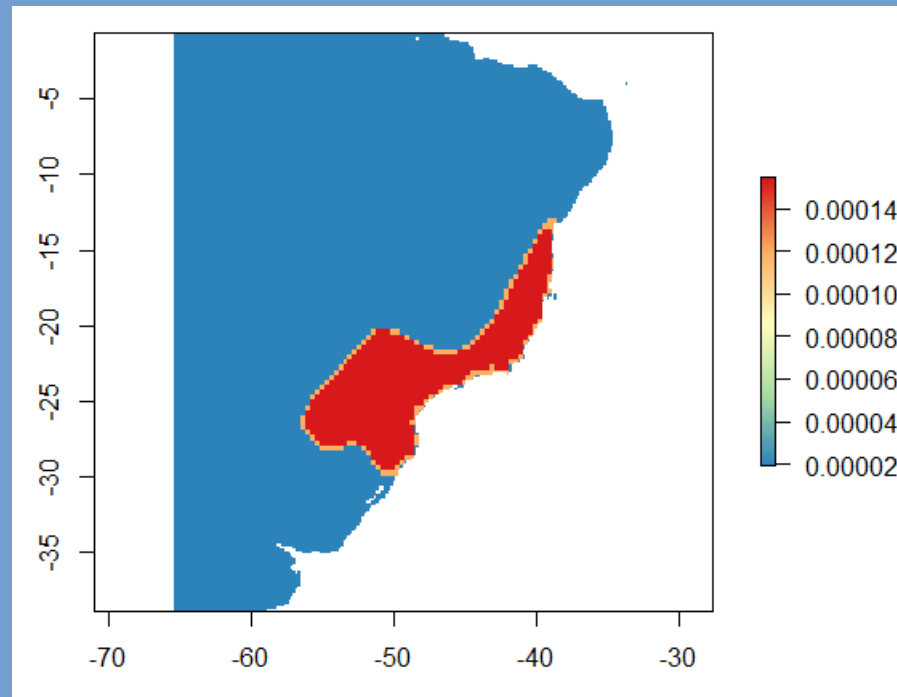
SABAP2

Step-by-step expert maps

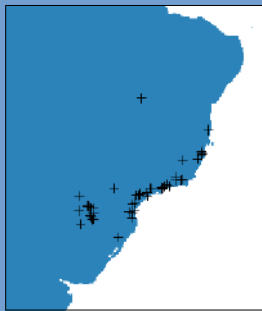
Step 1: Get species occurrences and an expert map



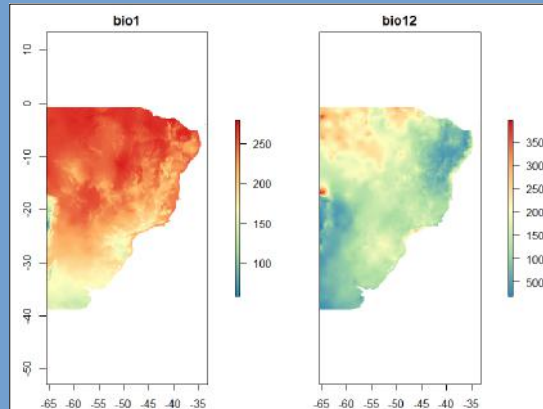
Step 2: Assign probabilities to areas inside and outside of expert range map (expert prior)



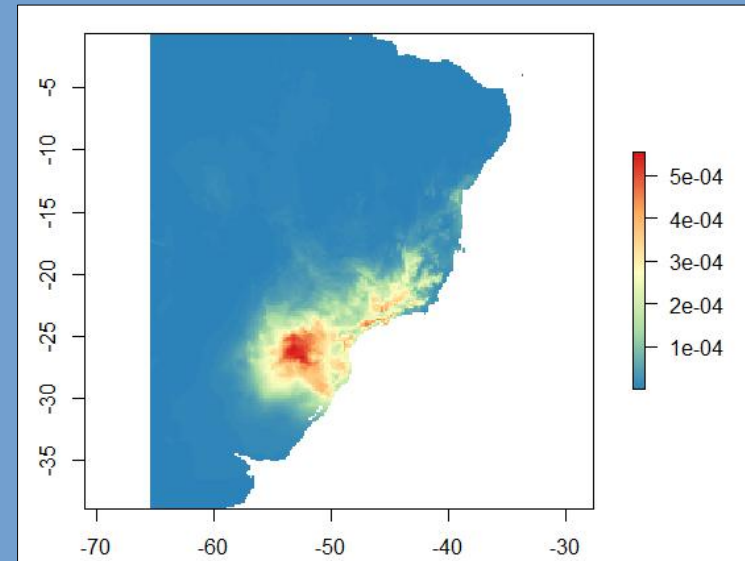
Step 3: Build a normal Maxent model for your species



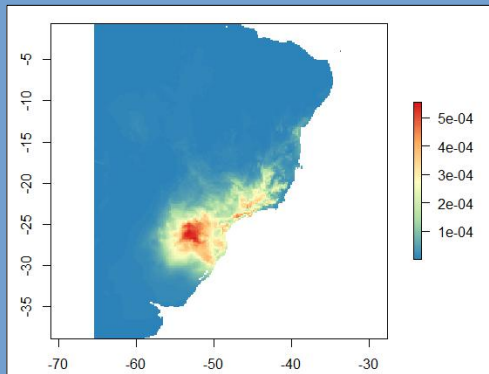
+



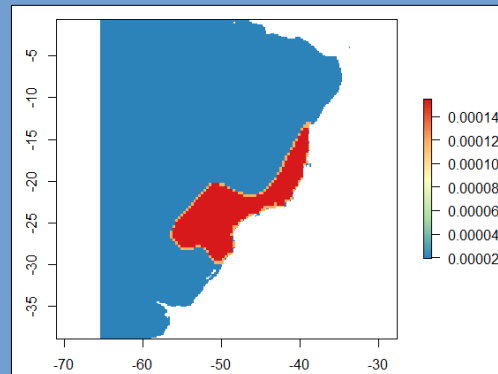
=



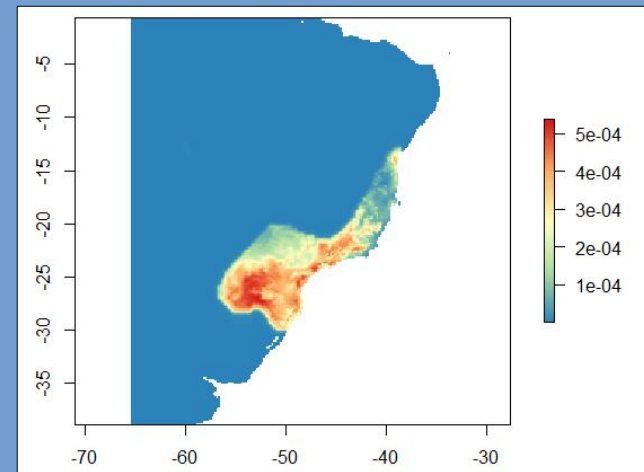
Step 4: Multiply normal Maxent model by expert prior



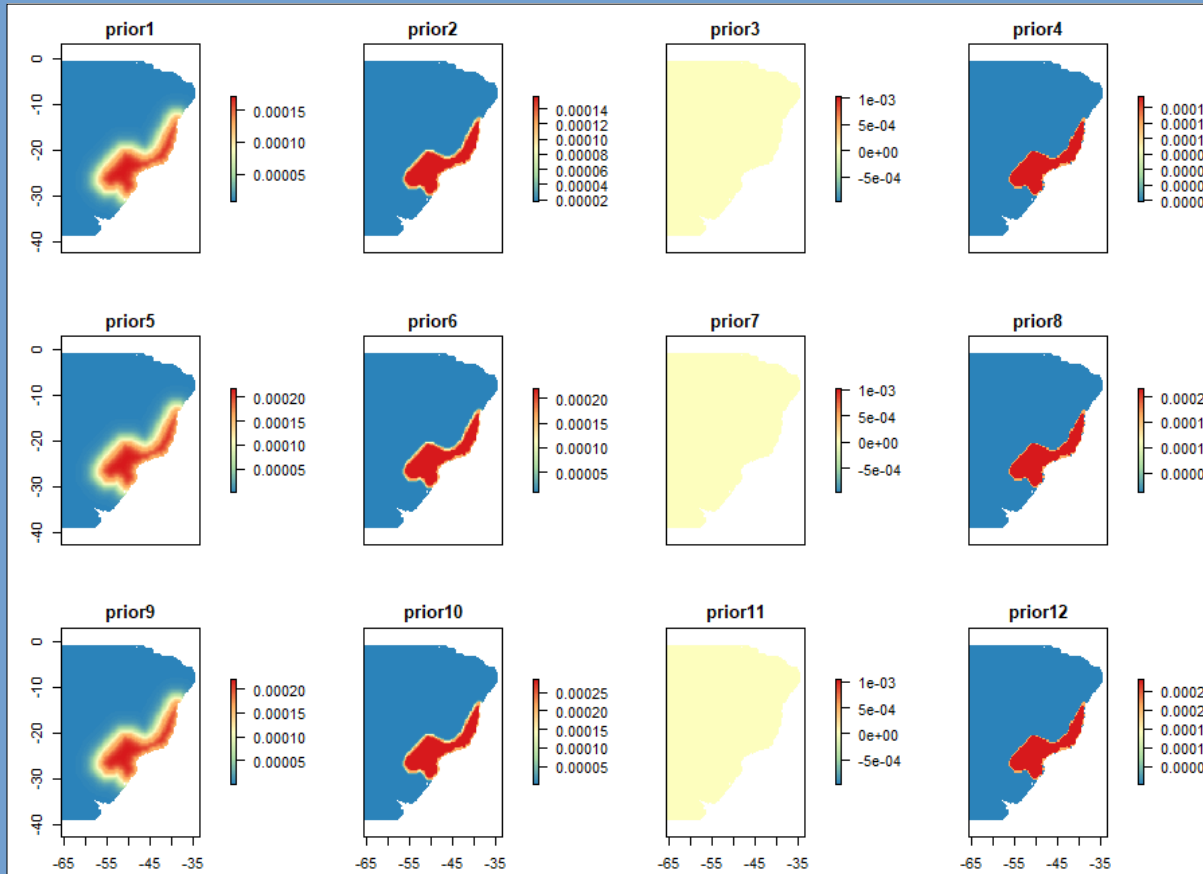
X



=



Extra steps!



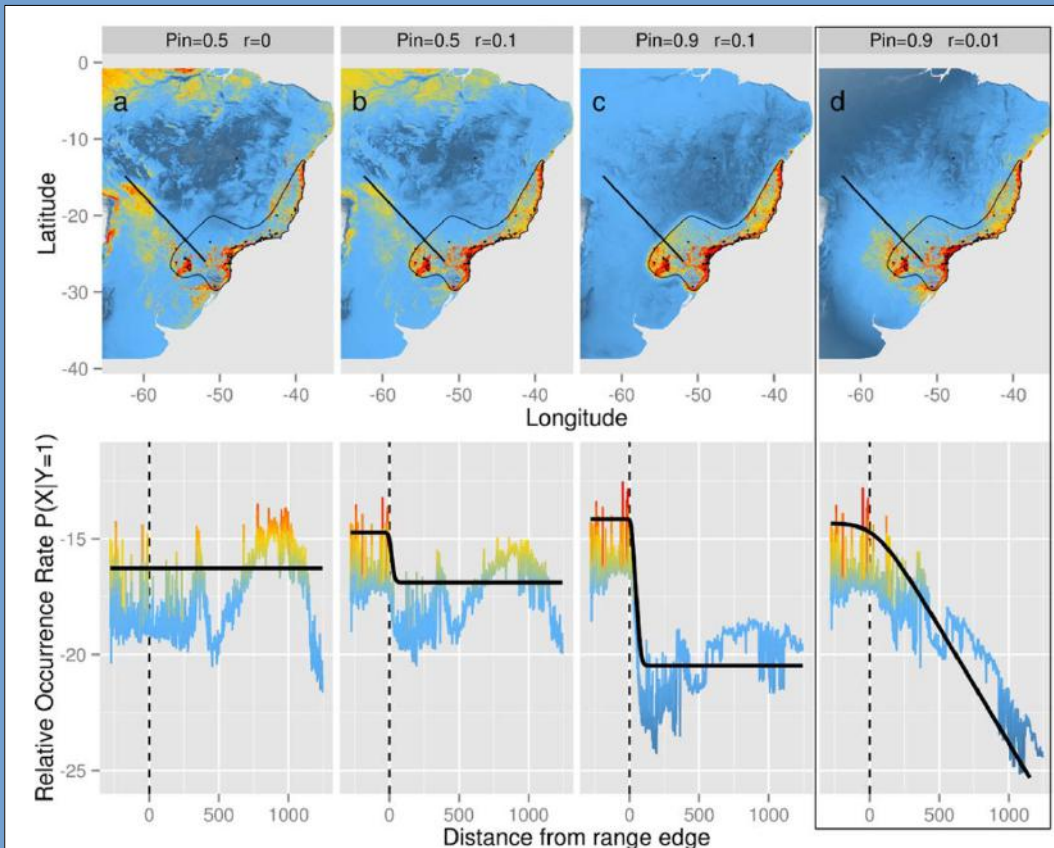
1) Assigning the probability values inside and outside the expert map

2) Sharpen or relax the expert map boundary

1) Assigning expert map probabilities

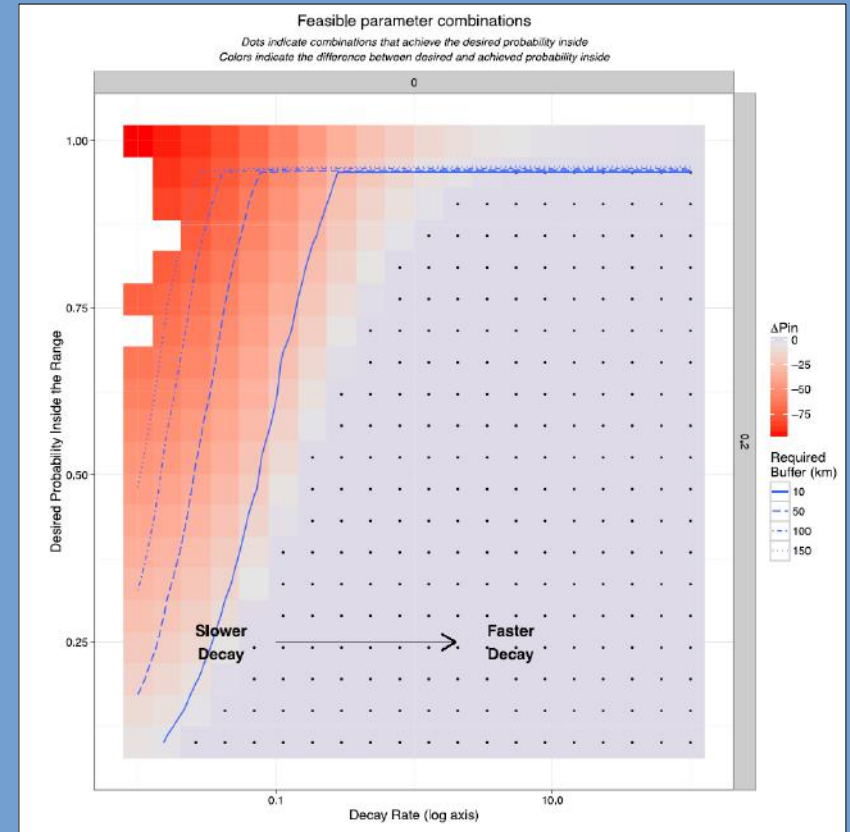
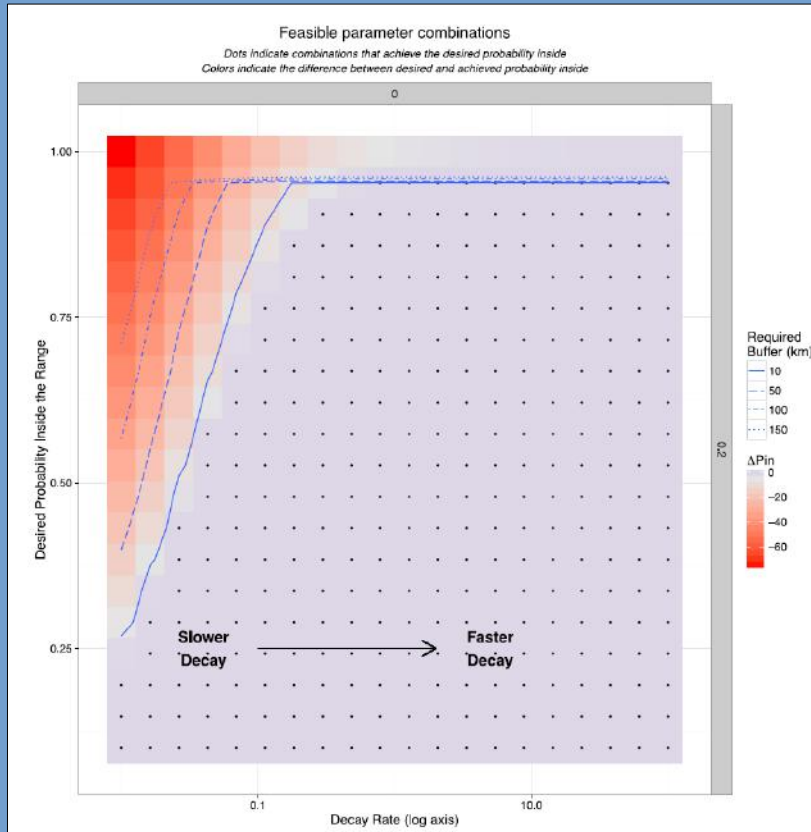
- Higher values inside expert map (P_{in}) = higher certainty
- Can use omission rate (prop. of observed presences outside the expert map)
- Should ideally use independent data for this

2) Expert map boundary transition “shape”

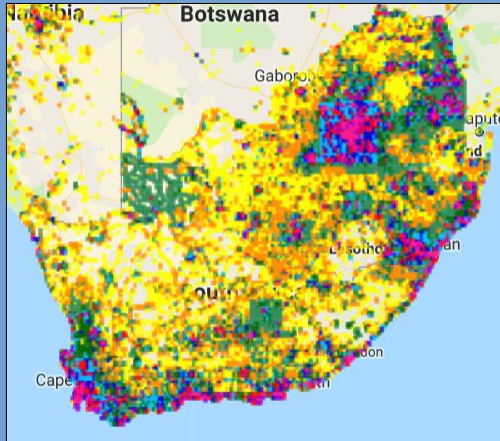


- Simplest = step function
- Logistic curve that can be parameterized
- Decay parameter (r): determines steepness

Not all combinations of P_{in} and r are possible



Summary



- Occurrence data usually biased
- SDMs overpredict
- Using offsets can help to address these problems by incorporating other sources of spatial info
- Other applications

