

# SDM

## GeoComput & ML

### 05 May 2022

# ReCap

- Joint Distr. :  $[A, B]$
- Cond. Prob. :  $[A | B] = [A, B] / [B]$
- LTP :  $[B] = \sum [B | A_i][A_i]$
- Bayes :  $[B_i | A] = \frac{[A | B_i][B_i]}{\sum [A | B_i][B_i]} = \frac{[B_i, A]}{[A]}$

**Spat. Stat.**

# Set Up

$$Z(s)$$

where location  $s$  vary over the index set  $D \subset \mathbb{R}^d$

- random
- dependency

# Stationarity

$$E(Z(s + h) - Z(s))$$

$$V(Z(s + h) - Z(s)) = 2\gamma(h)$$

$2\gamma(h)$  : variogram

# Process

$$[Y, \theta | Z] = \frac{[Z | Y, \theta][Y | \theta][\theta]}{[Z]}$$

# Data update

$$(Z^{(1)}, Z^{(2)})$$

$$[Z^{(1)} | Y, \theta][Y | \theta][\theta]$$

$$\begin{aligned} [Y, \theta | Z^{(1)}, Z^{(2)}] &= \frac{[Z^{(1)}, Z^{(2)} | Y, \theta][Y, \theta]}{[Z^{(1)}, Z^{(2)}]} \\ &= \frac{[Z^{(2)} | Z^{(1)}, Y, \theta][Y, \theta | Z^{(1)}]}{[Z^{(2)} | Z^{(1)}]} \end{aligned}$$

# Process Decomposition

$$[\cap_{i=1}^T Y_i] = [Y_1] \prod_{i=1}^{T-k} [Y_{T-k+1} | \cap_{i=1}^{T-k} Y_i], \quad k \in [1, T-1]$$

1st order Markov Assumption

$$[Y_T | \cap_{i=1}^{T-1} Y_i] = [Y_T | Y_{T-1}]$$

Therefore :

$$[\cap_{i=1}^T Y_i] = [Y_1] \prod_{i=2}^T [Y_i | Y_{i-1}]$$



# Cautions

$$[Y_2 | Y_1]$$

$Y = (Y_1, Y_2, F)$ , where  $F$  is the level of spat.  
aggregation

$$[Y] = [Y_2, Y_1 | F][F] = [Y_2 | Y_1, F][Y_1, F][F]$$

# Cautions

likelihood inference  $[Z | \theta]$

missing the fundamental importance of  $Y$  which is  
rooted in physics, chemistry, biology, economics,  
etc.

# Cautions

$$[\tilde{Z}] = f(Z) \approx \langle Z \rangle$$

$$[Y | \tilde{Z}, \theta_D, \theta_p] \propto [\tilde{Z} | Y, \theta_D][Y | \theta_p]$$

$$[\tilde{Z} | \theta_D, \theta_p] = \int [\tilde{Z} | Y, \theta_D][Y | \theta_p] dY$$

# SDM

# Domain Knowledge

*Statisticians can not evade their responsibilities for understanding the processes they apply or recommend.*

*– Ronald Fisher*

# Community Ecology

Whittaker : Multiplicative Law of Diversity

$$\gamma = \alpha \beta$$

- $\gamma$  : tot. # species
- $\alpha$  : ave. # species in a locality
- $\beta$  : variation in the species composition between locations

# Community Ecology

- organismic : traits determining persistence
- individualistic : species-specific response to env.

# Community Ecology

## Niche Theory

- Hutchinson, 1960s
- n-d hypervolume, resources and env.
- hard to define, species interaction w/ env.



# Community Ecology

## Equilibrium Theory of Island Biogeography

- MacArthur and Wilson, 1960s
- # species determined by the balance of immigration and extinction

# Community Ecology

## Unified Neutral Theory of Biodiversity and Biogeography

- Hubbell, 2000s
- diversity rising and organised at random
- all individuals ecologically identical
- stochastic random process

# Community Ecology

## Theory of Ecological Community

- selection : fitness
- drift : random
- dispersal : movement
- speciation : variation

# SDM

types

- correlative
- mechanistic

# SDM

- tendency

predictors ↑ → model complexity ↑ → overfitting ↑

- right-way

var. selection based on ecol. relevance

# Linear Models

$$\langle Z \rangle = \beta X, \quad Z \sim (\beta X, \sigma^2 I)$$

# Linear Models

$$\text{GLM} : g(f(\mathbf{Z})) = \boldsymbol{\beta} \mathbf{X}$$

link function  $g(\cdot)$

- identity
- logit :  $\log\left(\frac{p}{1-p}\right) \Rightarrow \text{logit}^{-1}(p) = \frac{1}{1 + e^{-x}}$
- log

# Linear Models

Mixed models

$$Z_i = \sum X_i \beta_i + \alpha_{p(i)} + \epsilon_i$$

- $r_i = \epsilon_i + \alpha_{p(i)}$
- $C(i, i) = \sigma_p^2; C(i, j) = 0$



# Presence-Absence Data

## Challenges

- detectability
- sample bias

# Presence-Absence Data

MaxEnt

$$\pi(s)$$

$$[Z = 1 | s] = \frac{[s | Z = 1][Z = 1]}{[s]}$$

Jaynes principle : max entropy

$$H(x) = - \sum \pi(x) \ln \pi(x)$$

# Presence-Absence Data

Feature space :  $\{f_1, \dots, f_n\}$

$$\hat{\pi}(x) = \sum \pi(s) f_j(x)$$

$$\tilde{\pi}(x) = \langle f_j \rangle$$

$$\hat{\pi}(x) \sim q_\lambda(s) = \exp(\lambda f(x)) / Z_\lambda$$

$\min(-\ln(q_\lambda))$  with regularisation

# Presence-Absence Data

Pseudo absence data

- random
- random with exclusion : weighted by env. or geo.

core question : spat. extent

# Joint SDM

$$\mathbf{Z} = \mathbf{XB}$$

# Acknowledgement

Thanks for Your Attention

# References

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