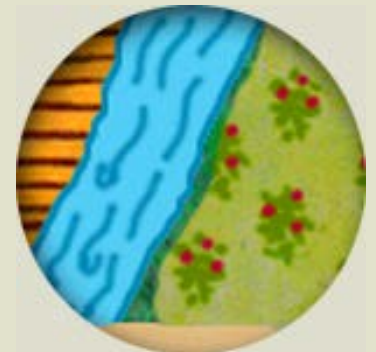


# Spatial linear models in R 2015



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Simoneta Negrete Yankelevich

# Today

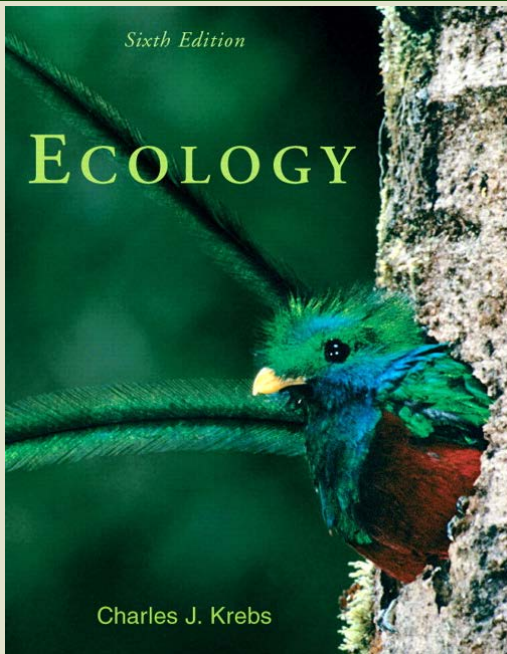
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- ¿Why should we care about space?
- The ABC of space
- How to recognise and measure spatial patterns
- GLS (Generalised least squares): One of the many forms to consider spatial patterns in linear models.

¿Why should we care about  
space?



# ¿What is Ecology?

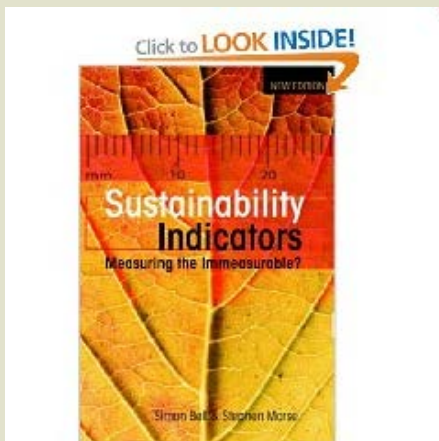


- “We are interested in *where* organisms are, *when* they occur *there* and *why*”.

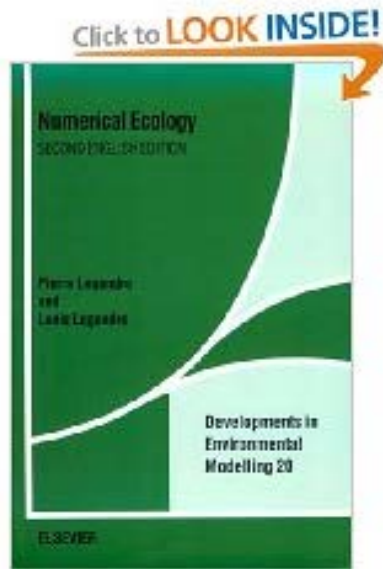
(Krebs, (4th ed) p.3)

- “It can only be considered *sustainable* if the temporal and *spatial scales relevant* to the phenomenon are selected”

(Costanza y Patten, 1995). "Defining and predicting sustainability." *Ecological Economics* **15** (3): 193–196.



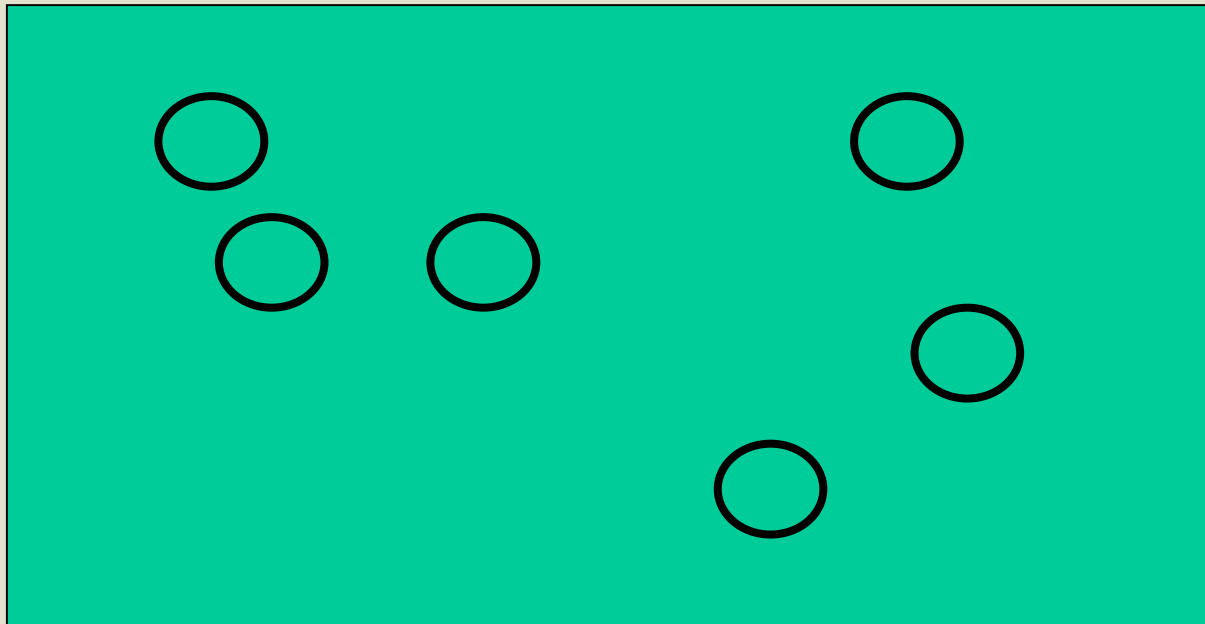
# The selection of numerical methods



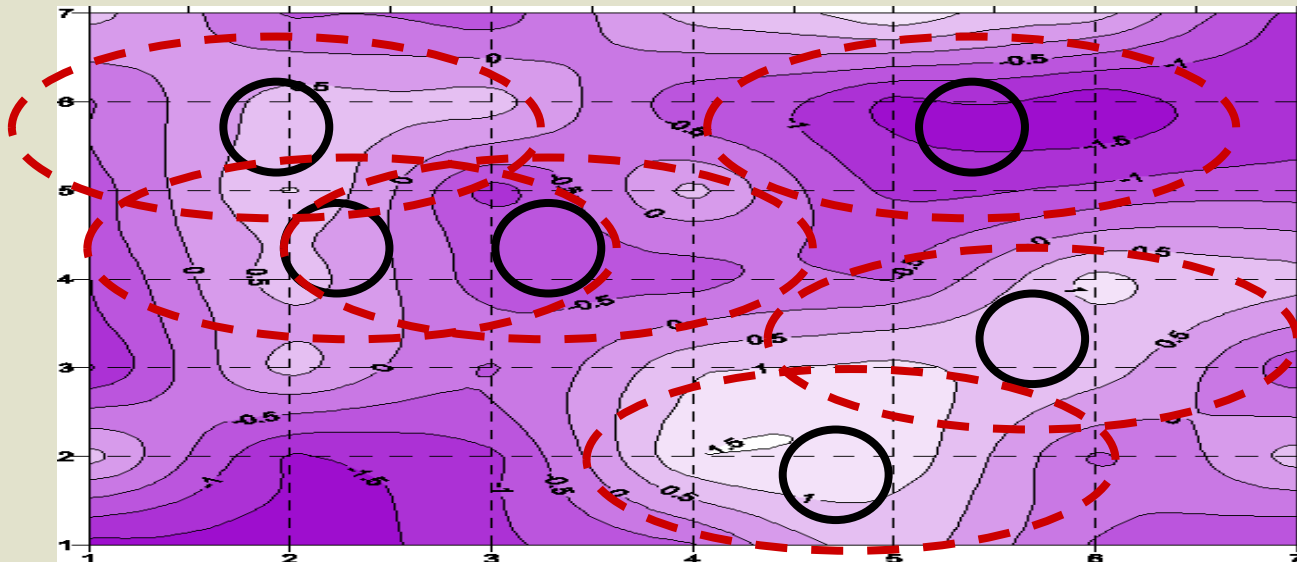
Numerical  
Ecology (Legendre  
y Legendre 1998)

- The statistical methods most commonly used assume **independence** of observations: distribution in space is **uniform** or **random**.
- In nature the rule are **patches** and **gradients**.

# Over estimation of degrees of freedom

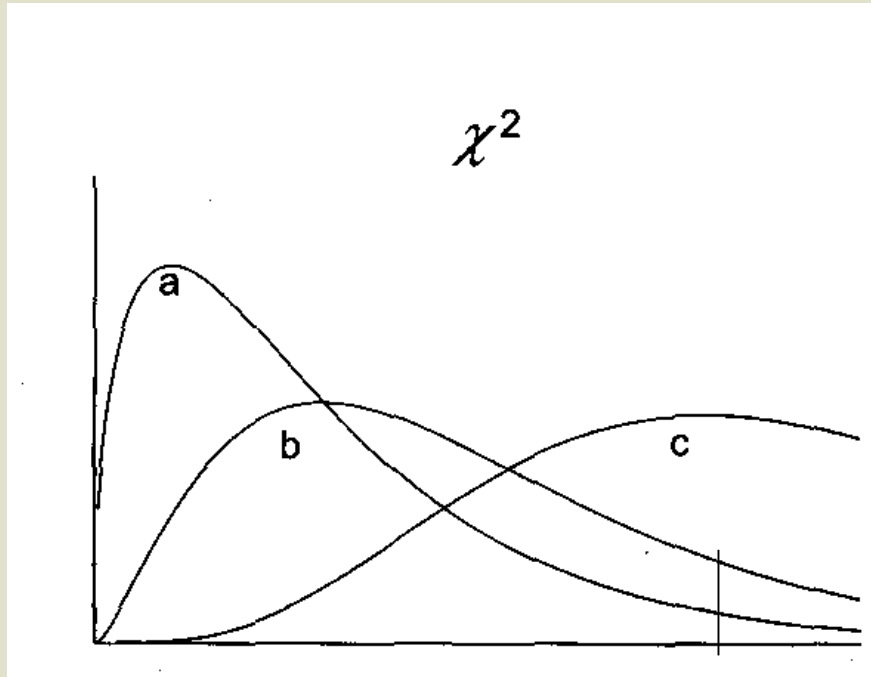


$$\text{d.f.} = (6-1)$$



$$\text{d.f.} = (4.5-1)$$

# Comparison with test statistic ( $\chi^2$ )



$$\chi^2 = 35.5$$

Assuming  
Independence in  
space when there  
isn't one, increases  
the probability of  
**false positives** in  
hypothesis testing.

# Spatial structuring is functional



In homogenous or random ecosystems

- Consumers would not be close to resources
- It would be impossible for an organism to predict where to find a mating partner.
- Soil conditions around a plant would not be particularly suitable or unsuitable for its offspring.

• **Spatial structuring** of ecosystems is **functional** and that means it should be central **focus** of attention in **Ecological research**.



# The ABC of space



# Types of spatial data



- **Point pattern.** The process determines the position of individuals in space.
- **Geostatistical.** The process affects the value of the variable in different points of space.
- **Areal.** The process generates qualitative or quantitative changes in contiguous surfaces, mosaics or grids. The information of which cells are neighbours is available.



## •Patterns

**Patch:** area “with delimited borders” characterized by a higher concentration than expected at random.

**Gradient:** Gradual increase in a single direction.

**uniform** (disperse or overdisperse): More homogeneous than expected at random.

Pattern is used as the opposite of random.

# BEFORE DOING ANYTHING!!!!

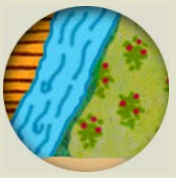


- ¿Does your hypothesis predict who, when and WHERE?
- Define the temporal and spatial **extension** in which your phenomenon expresses itself.
- Determine if the temporal and spatial **resolution** of the study design will capture the process.
- Make sure that the sampling design will produce the type of **data appropriate** for the statistical strategy that you are planning to use

# How to detect and measure spatial pattern?

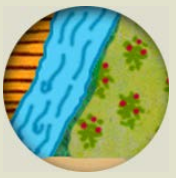


# Spatial dependence

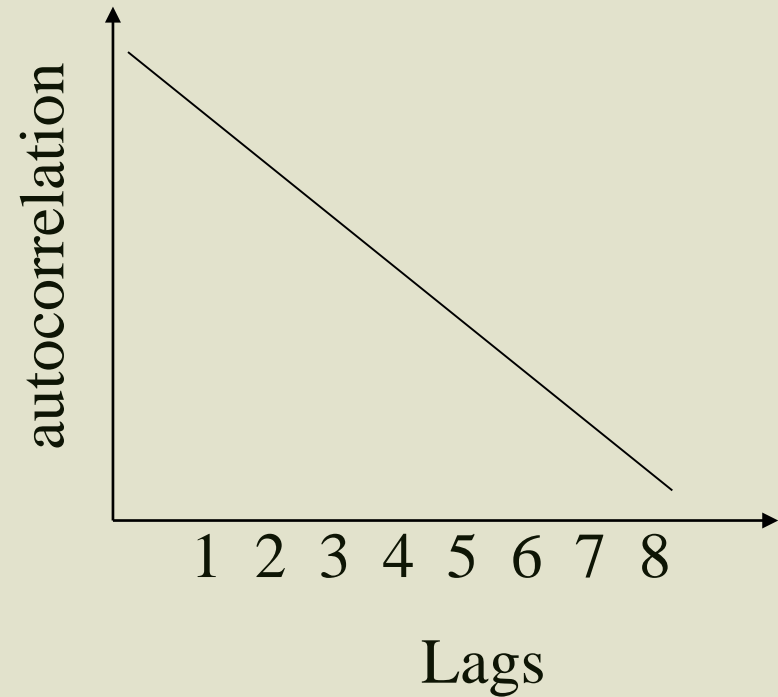
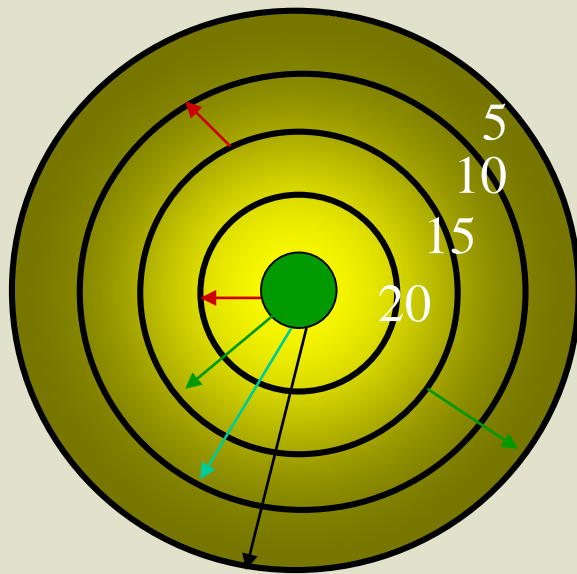


- **Spatial dependence or autocorrelation** is the correlation among the values of a variable given their position in space.
- Two observations that are close together have similar values, as their distance increases, the relationship fades-off until it becomes undetectable. We then **tentatively** conclude that the observations become **spatially independent**.

# Spatial dependence

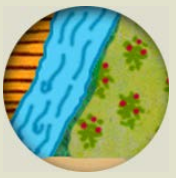


Seeds that fall  
from a tree

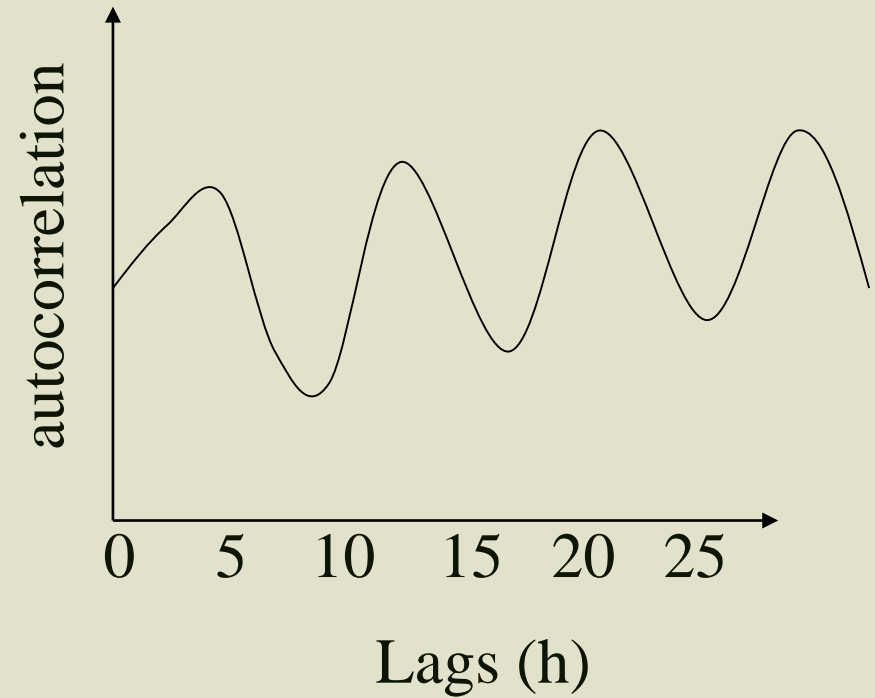
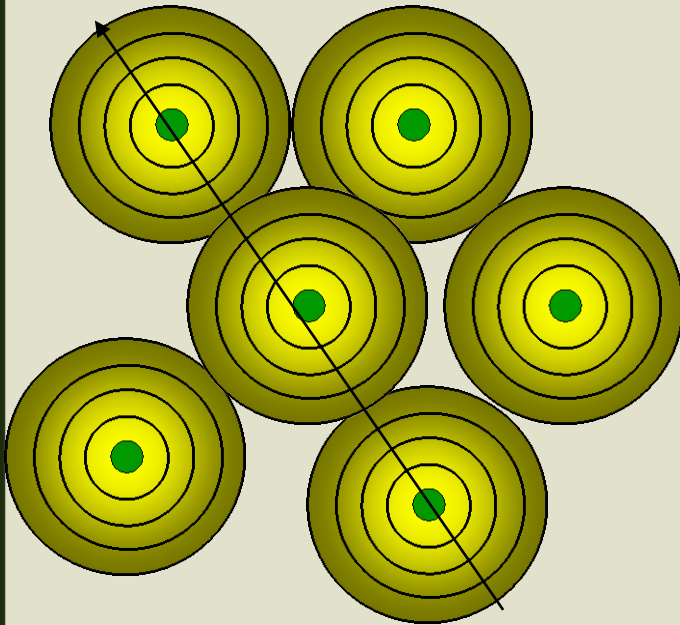


It is measured through the  
comparison of lags or  
distance intervals

# Spatial dependence

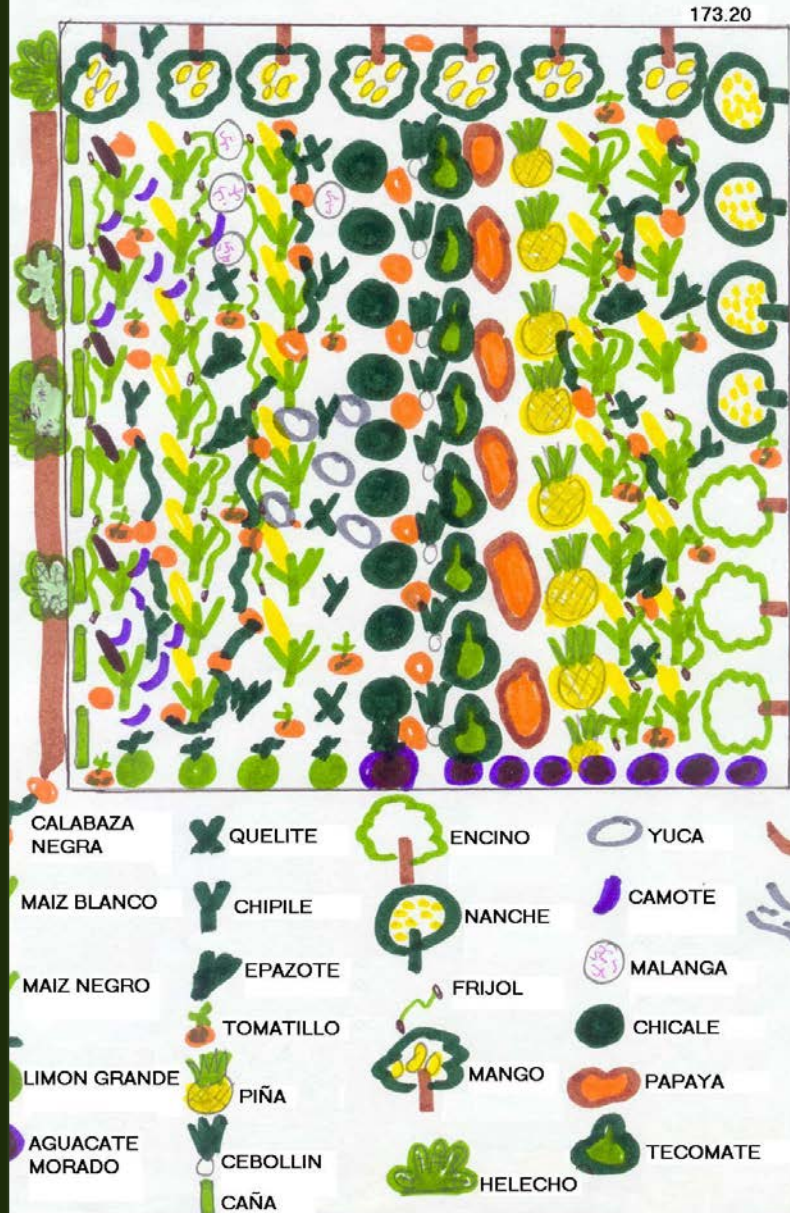


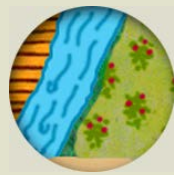
Seed abundance in a  
plantation





# Popoluca Maize polycultures, Los Tuxtlas, Mexico





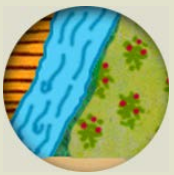
Traditionally d is h  
in variograms

$$\hat{\gamma}(h) = \frac{1}{2W(h)} \sum_{i=1}^{W(h)} [y_i - y_{i+h}]^2$$

Weighted  
mean

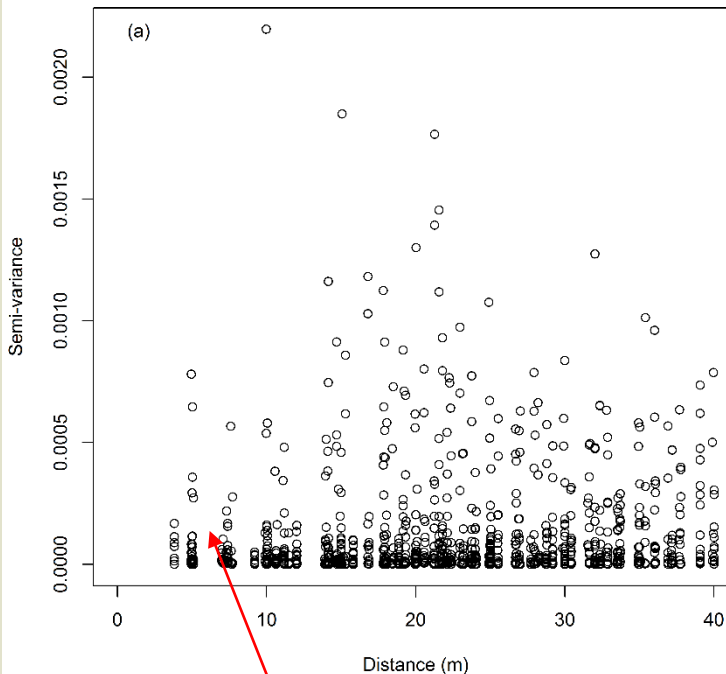
Squared differences

# Autocorrelation functions: Variograms



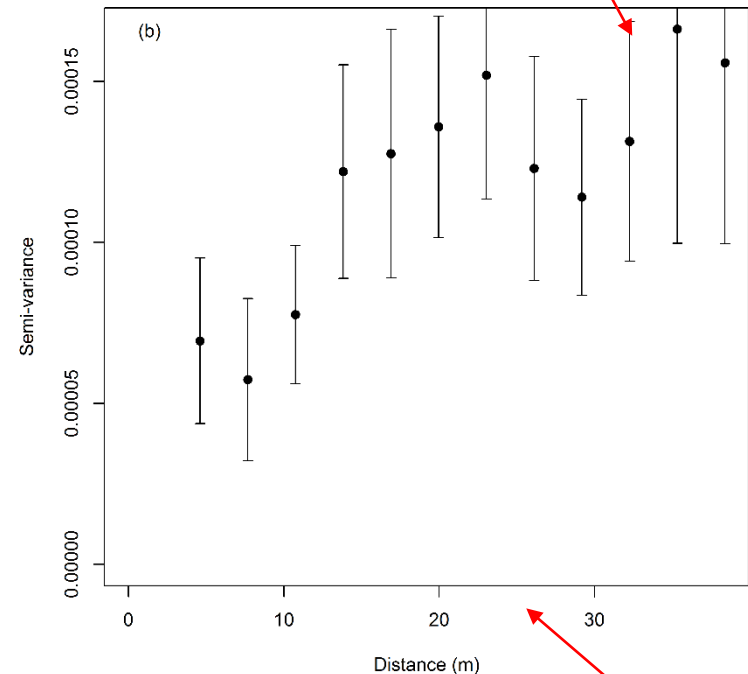
CI increase with distance, confidence in estimations diminishes

## Cloud Variogram

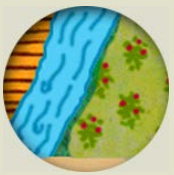


Raw squared differences (each point is a pair of observations)

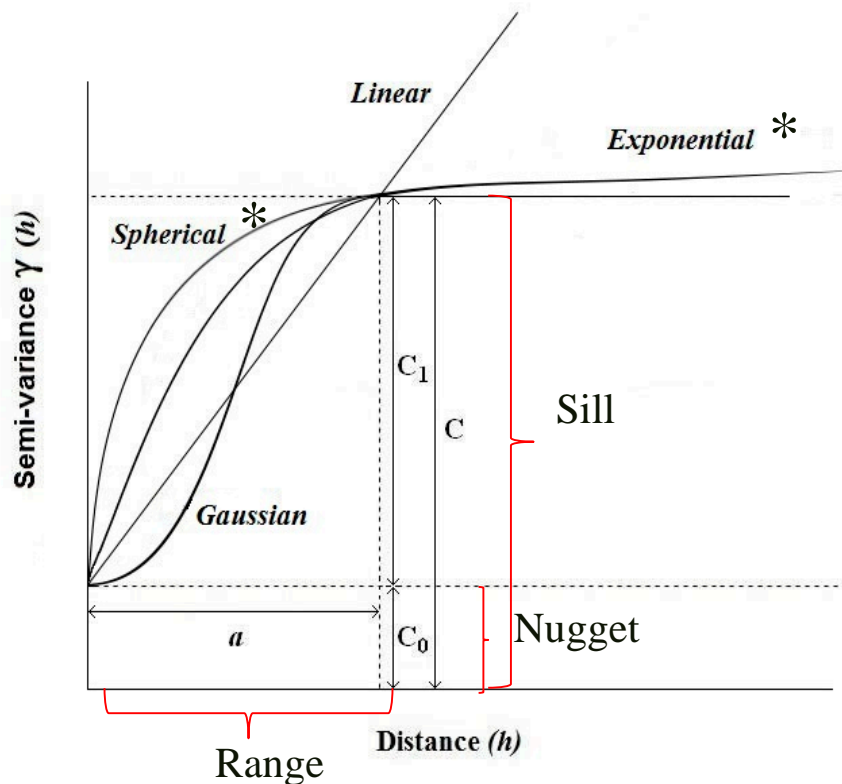
## Variograma



The weighted mean square with exploratory CI.



# Variography: Theoretical Variograms



Nugget effect:  $\gamma(h) = C_0$

Exponential:  $\gamma(h) = C_0 + C_1 \left[ 1 - \exp\left(-3\frac{h}{a}\right) \right]$

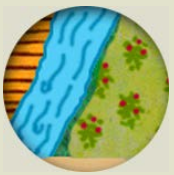
Spherical:  $\gamma(h) = C_0 + C_1 \left[ 1.5\frac{h}{a} - 0.5\left(\frac{h}{a}\right)^3 \right]$   
if  $h \leq a$ ;  $\gamma(h) = C$  if  $h > a$

Linear:  $\gamma(h) = C_0 + bh$

Gaussian:  $\gamma(h) = C_0 + C_1 \left[ 1 - \exp\left(-3\frac{h^2}{a^2}\right) \right]$

\* Asymptotic models: when  $a$  reaches 90% of the asymptote.

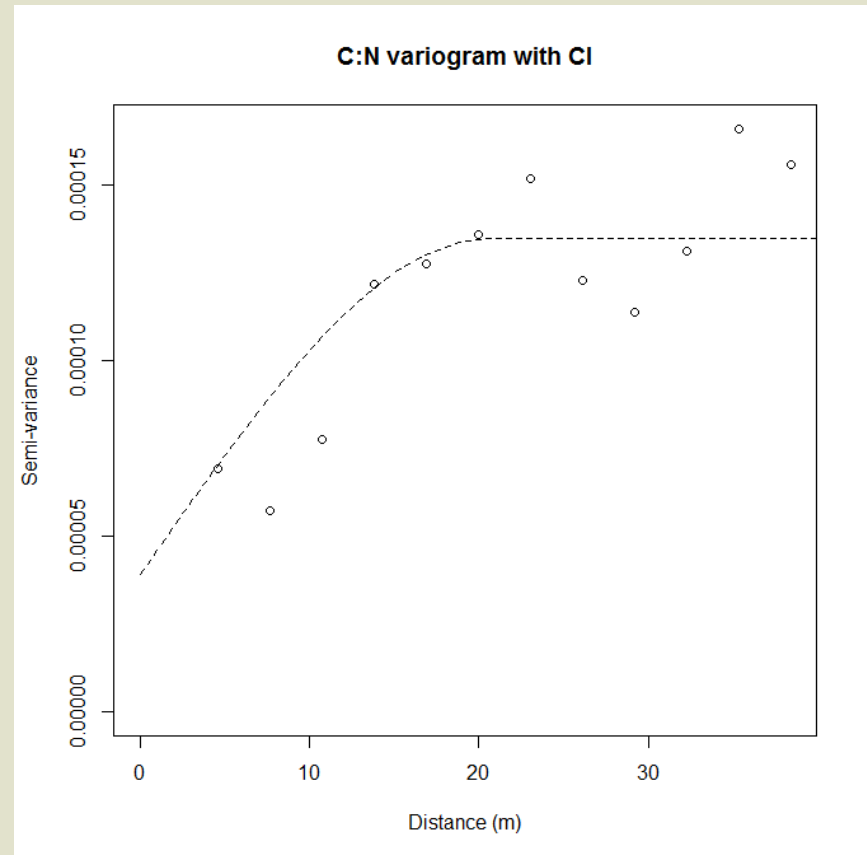




# Then what?

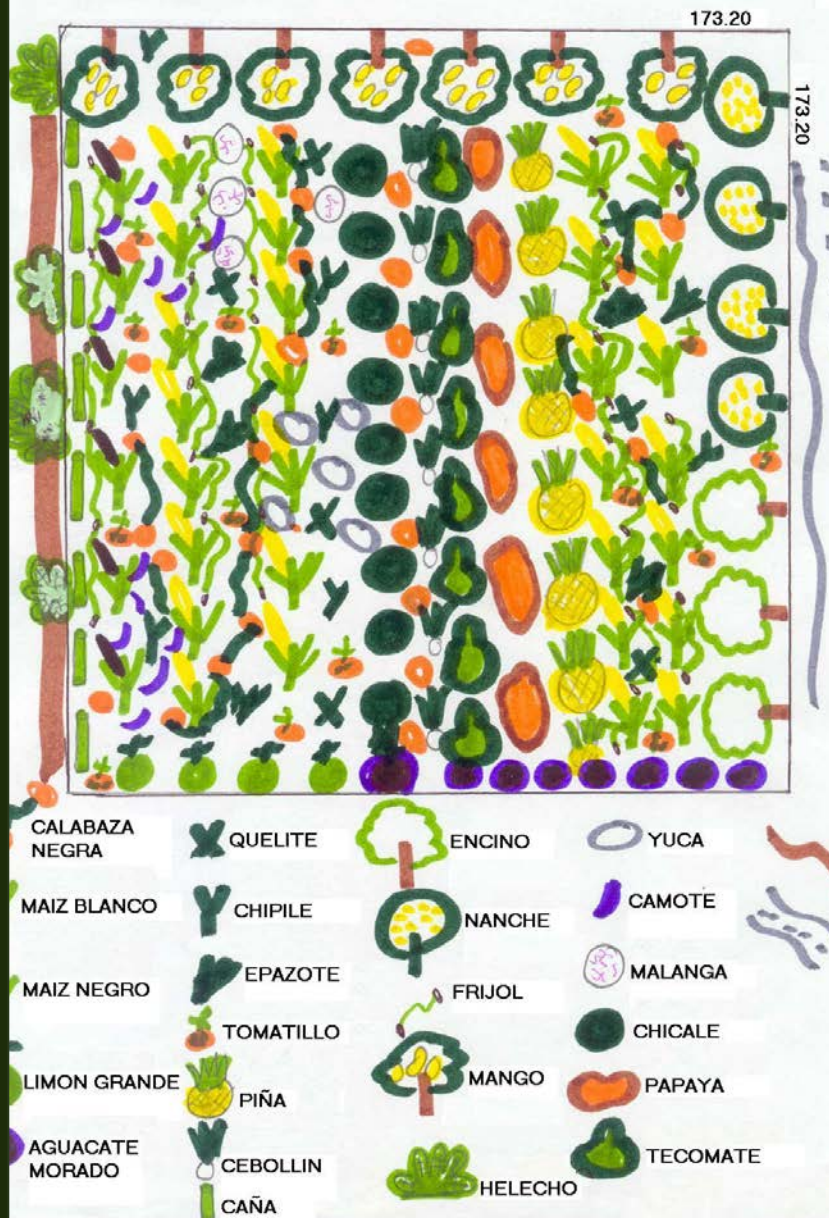
- ML Variogram:

Range: 20.7 m  
nugget:  $3.9 \times 10^{-5}$   
Sill:  $1.0 \times 10^{-4}$



Conclusion: Subtle patchy pattern (20.7 m) within a gradient.

# Popoluca Maize polycultures, Los Tuxtlas, Mexico



# Spatial linear models: GLS



# Spatial linear models



- When one of the important explanatory variables **was not measured** and it prints a spatial pattern on the dependent variable.
- When the dependent variable has a normal distribution there is **GLS (Generalised least squares or kriging regression)**, a type of mixed model with an R-side effect).



# A type of inferential analysis



- Requires significance tests of statistics.
- You have to comply with distributional assumptions once the spatial pattern is accounted for.
- Randomisations of positions (restricted or not) are used to generate the distribution that represents  $H_0$ .

# GLS: a type (strange) of mixed model



There is no random factor!

Sides

G R

$$Y = \beta X + \varepsilon$$

$$\varepsilon \sim N(0, \sigma^2 \Lambda i)$$

variance-covariance matrix. can be calculated with the sill, nugget and range parameters of the variogram.

# What did we learn about the C:N in polycultures?



- There are two spatial structures in its distributions (gradients and patches of 23m).
- When we include the nearest species in the model, the patchy pattern is explained but not the gradient.
- This pattern is mainly associated with a lower C:N close to sweet potatoes (legumes that are symbionts with *Bradyrhizobium* a nitrogen fixing bacteria from the forest!).

# Autoregressive models: SAR



Exogenous autocorrelation (unknown)

$$\text{SARerr: } Y = \beta X + \lambda W u + \varepsilon$$

Endogenous autocorrelation

$$\text{SARlag: } Y = \rho W Y + \beta X + \varepsilon$$

Endogenous and exogenous (known) autocorrelation

$$\text{SARmix: } Y = \rho W Y + \beta X + \theta W X + \varepsilon$$

Autoregression  
coefficients

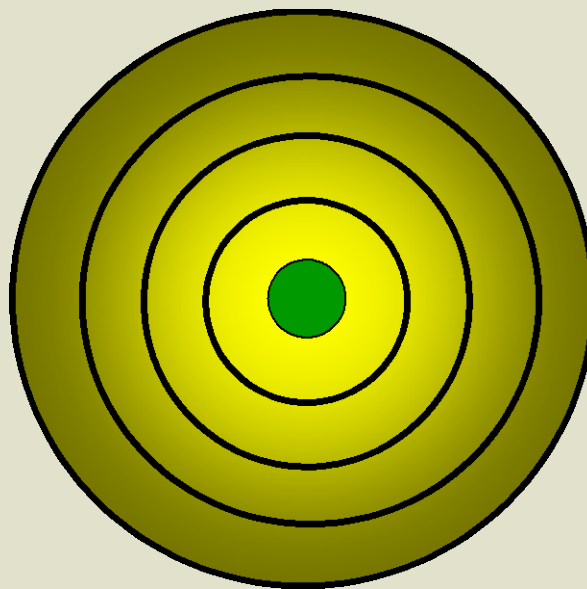
Weighted  
Neighbourhood  
matrix

Spatially  
dependent error

# SARlag and SARmix: direct and indirect impacts

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- Indirect impacts (spill overs) are like an expansive wave (seabird nesting and gregarious behaviour).



## In summary



- There is a world to be explored! And...it is a spatially explicit one.
- Spatially explicit models are complex. They ARE NOT an easy way to get read of design problems.
- Spatially explicit models ARE an elegant and informative way to understand the role of space in Ecology.
- This is only a starter...have a nice spatial trip!