

# Geostatistics Basics

SERGIO REY

**GEOG 384**

**Spatial Data Analysis**

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# Outline

- 1 Geostatistical Perspective
  - Spatial Random Field
  - Moment Conditions
- 2 Variogram and Correlogram
  - Semi-Variogram
  - Covariogram

# Spatial Random Field

## Continuous Spatial Process

$$\{Z(s) : s \in D\} \quad (1)$$

- $s$  is spatial index, continuous in  $R^2$  ( $R^3$ )
- Sample of spatial locations
  - ▶  $\{s_1, s_2, \dots, s_n\}$  : sample locations
  - ▶  $\{Z(s_1), Z(s_2), \dots, Z(s_n)\}$  : random variable at sample locations

# Geostatistical Data

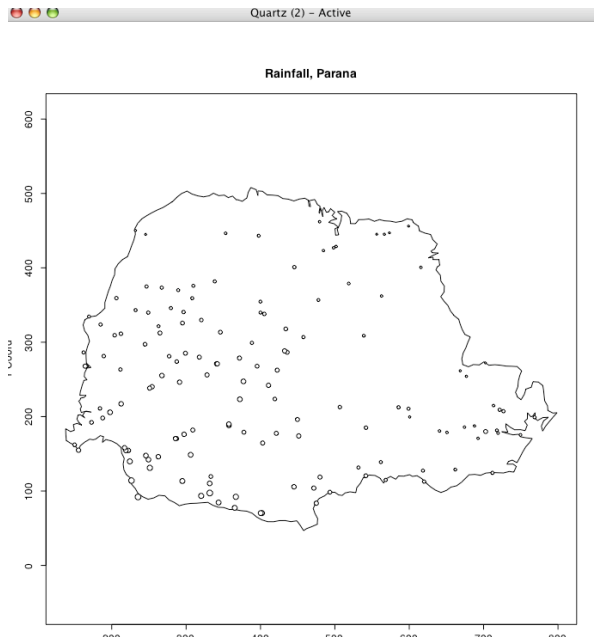
## Spatial Domain: $D$

- A continuous and fixed set.
- Meaning  $Z(s)$  can be observed everywhere within  $D$ .
- Between any two sample locations  $s_i$  and  $s_j$  you can theoretically place an infinite number of other samples.
- By fixed: the points in  $D$  are non-stochastic

## Continuous Variation

- Because of the continuity of  $D$
- Geostatistical data is referred to as “spatial data with continuous variation.”
- Continuity is associated with  $D$ .
- Attribute  $Z$  may, or may not, be continuous.

# Geostatistical Data: Rainfall in Parana State Brazil



# Geostatistical Data

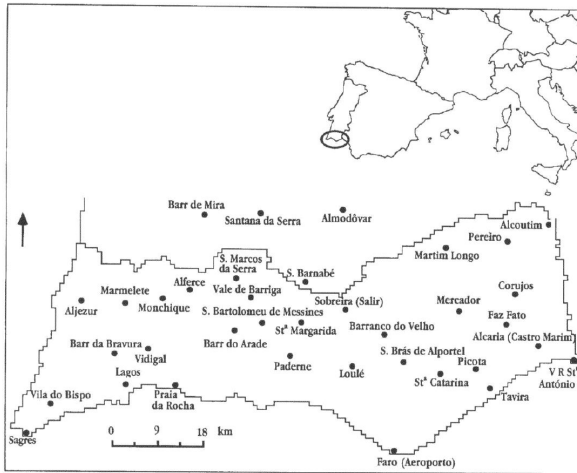
## Continuous variation

- Potentially measurable anywhere in  $D$
- Impossible to sample  $D$  exhaustively

## Reconstruction of the surface from observed sites

- Tessellation based methods
- Interpolation
- Kriging

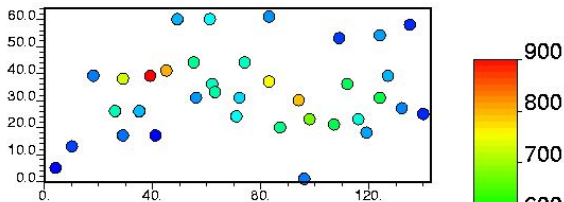
## Surface Reconstruction: Example<sup>1</sup>



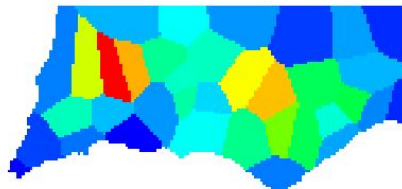
<sup>1</sup>From Goovaerts, P. (1999) "Performance comparison of geostatistical algorithms for incorporating elevation into the mapping of precipitation". *Geocomputation '99*.

# Surface Reconstruction: Tessellation Based Method

Annual rainfall data (mm)

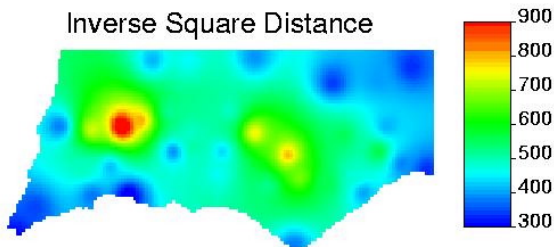


Thiessen Polygons



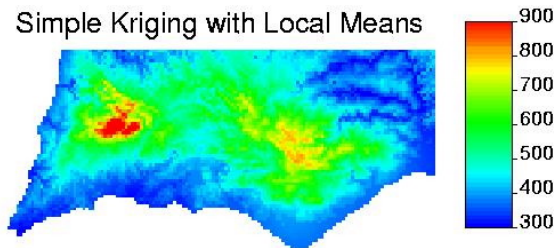


# Surface Reconstruction: Spatial Interpolation

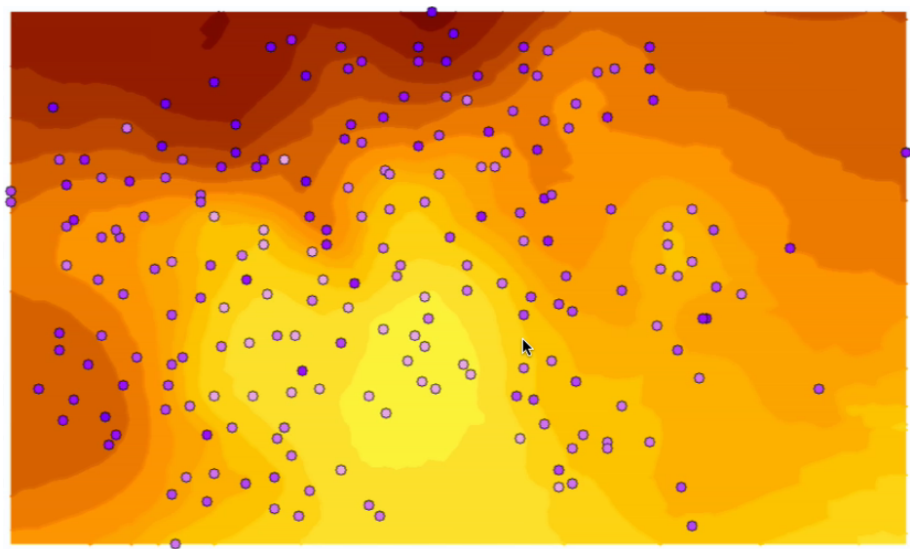


# Surface Reconstruction: Kriging

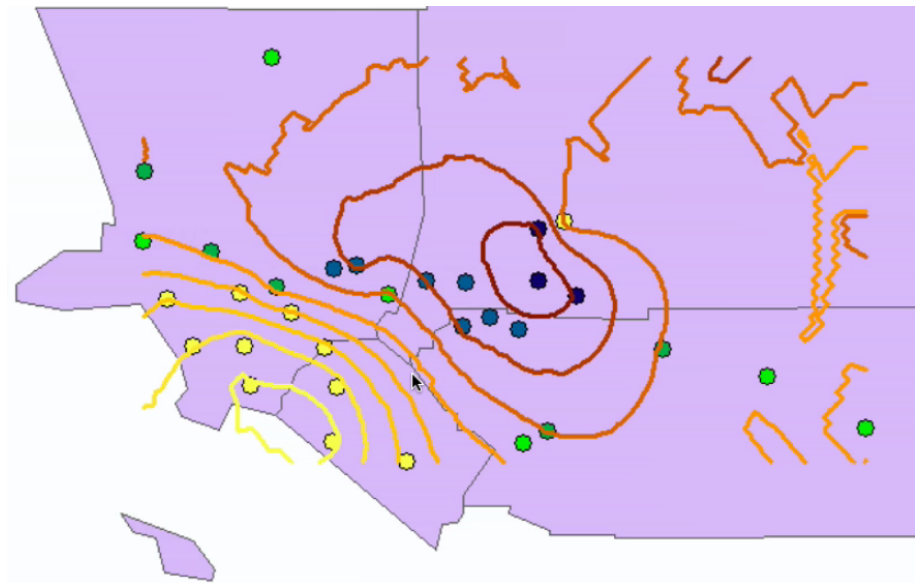
Simple Kriging with Local Means



# Baltimore Sales Price, Baltimore MD



# LA Basin Ozone Levels



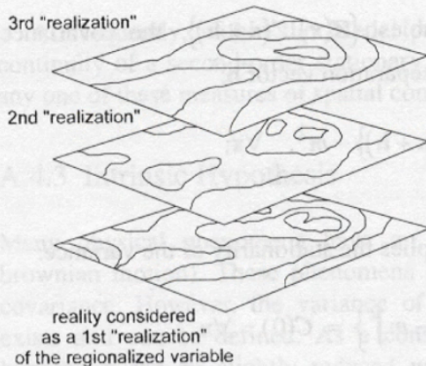
# Conceptual Framework

## Equilibrium

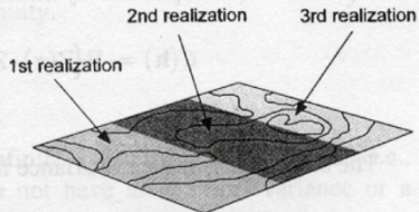
- Equilibrium = Stationarity
- Stochastic Process
  - ▶ not multiple realizations, but a single realization
  - ▶ the map is a single data point
- Notion of Stability
  - ▶ go from a single data point to acting as if there are multiple observations

# Spatial Stationarity

## Concept



## Application (stationarity hypothesis)



# Moment Conditions

- Constrain Variability
- Moments Must Exist
  - ▶ no infinite variance
- Moments Must Be Regular Over Space
  - ▶ restrictions on heterogeneity
  - ▶ restrictions on range of dependence

- Sample Realization is Representative
  - ▶ average obtained using a single realization is same as over all possible realizations
  - ▶ whether you observe one or many maps, the information should be the same



# Strict Stationarity

- Pertains to the Complete Distribution
- Invariance Under Spatial Shift
  - ▶ joint density for two different spatial subsets is same
  - ▶  $\{z(s_l), \dots, z(s_k)\}$  and  $\{z(s_{l+h}), \dots, z(s_{k+h})\}$
  - ▶ information about process is the same no matter where it is obtained
- Very Strict Requirement

# Moment Stationarity

- Moments Invariant Under Spatial Shift
- Mean
  - ▶ no spatial trend
- Variance
  - ▶ no heteroskedasticity (no spatial regimes)
- Covariance
  - ▶ not a function of location
  - ▶ only spatial separation, angle

# Intrinsic Hypothesis

## No Spatial Trend

- if there is a trend, take it out
- residuals have no trend (mean = 0)

## Constant Variance

- Variance of first difference only a function of displacement

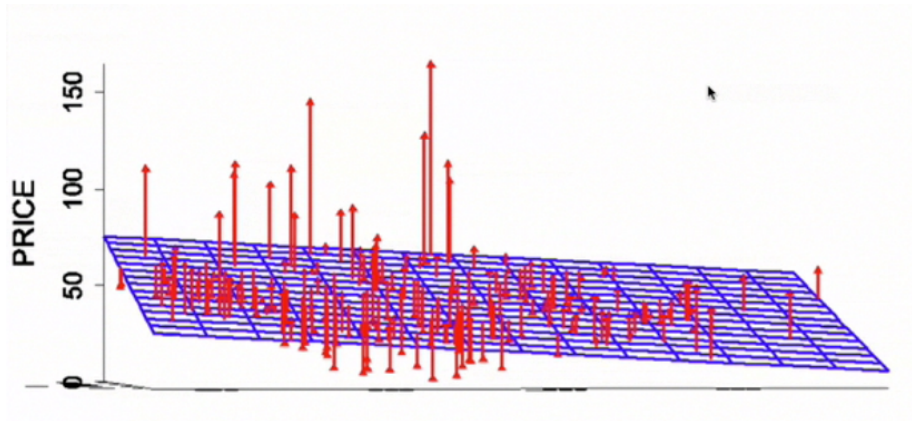
$$V\{Z_{s+h} - Z_s\} \quad (2)$$

a variability of the difference

# Bubble Plot Baltimore House Sales Prices



# Detrending Baltimore House Sales Prices



$$P = -166.02 - 0.148x + 0.634y$$

# Semi-Variogram

## Variogram Function

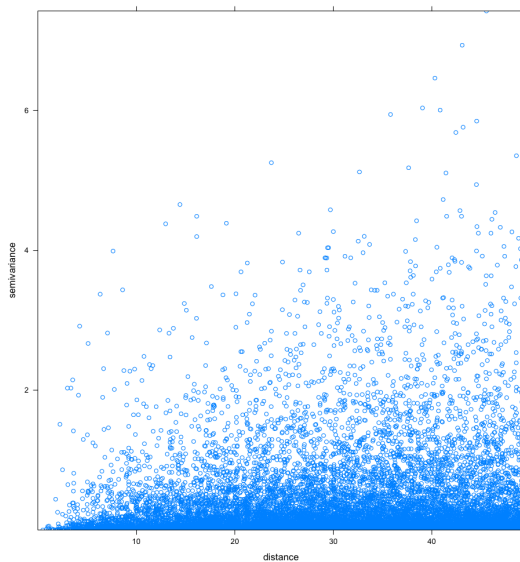
$$2\gamma(h) = V[Z_{s+h} - Z_s] \quad (3)$$

factor 2, so  $\gamma(h)$  is half of the variogram, or semi-variogram

## Constant Mean Assumption

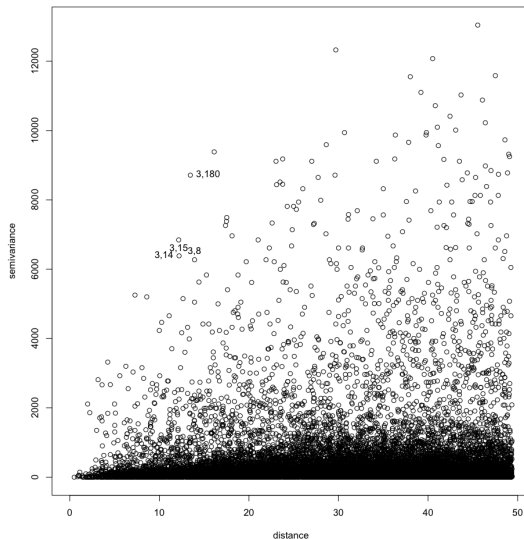
- $E[Z_{s+h} - Z_s] = E[Z_{s+h}] - E[Z_s] = 0$
- $V[Z_{s+h} - Z_s] = E[Z_{s+h} - E[Z_s]]^2 - 0$
- $\gamma(h) = (1/2)E[Z_{s+h} - E[Z_s]]^2$
- average of squared differences

# Variogram Cloud



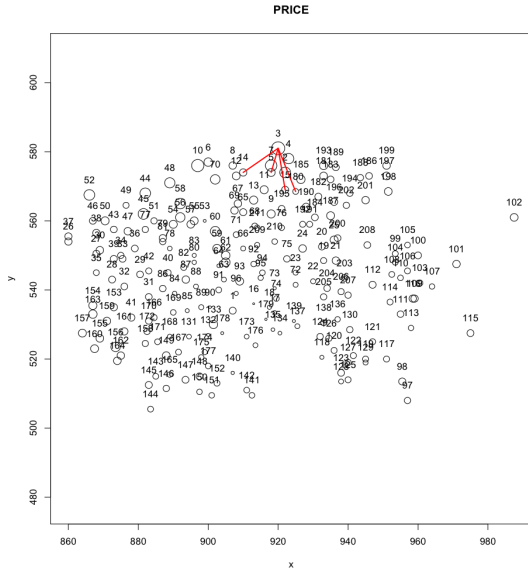
4

# Outliers in Variogram Cloud





# Outliers in Bubble Plot



# Estimating a Variogram

## Methods of Moments

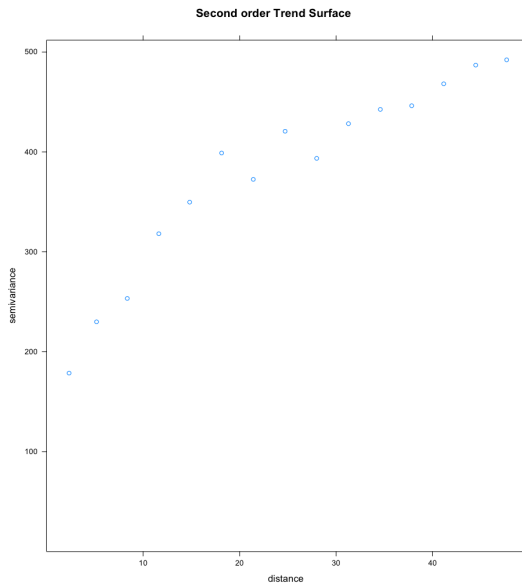
$$2\gamma(h) = (1/|N(h)|) \sum_h [Z_{s+h} - Z_s]^2 \quad (4)$$

- average of squared differences by distance bin  $h$
- $N(h)$  number of pairs in distance bin  $h$

## Rules of Thumb

- at least 30 pairs in each bin
- $\max h < D/2$  ( $D$  is maximum distance)
- distance of reliability

# Variogram on 2nd Order Trend Surface



# Covariogram

## Second Order Stationarity

- covariance regular over space
- stronger assumptions: regularity second order moments, cross-product

## Covariance at $h$

$$C(h) = \text{Cov}[Z_{s+h}, Z_s] \quad (5)$$

## Process Variance

- covariance at distance zero
- $C(0) = V[Z_s]$

# Correlogram

## Autocorrelation Function

$$\rho(h) = C(h)/C(0) \quad (6)$$

covariance standardized by process variance

## Distance Decay

- Correlogram decreases with distance
- Tobler's law

# Semi-Variogram and Covariogram

## Two approaches to same concept

$$2\gamma(h) = E[Z_{s+h} - Z_s]^2 \quad (7)$$

$$= 2E[Z_s^2] - 2E[Z_{s+h}Z_s] \quad (8)$$

## Semi-variance

- $\gamma(h) = C(0) - C(h)$
- semi-variance is variance of process less covariance at distance  $h$