V/S Modellieren in der Landschaftsarchäologie

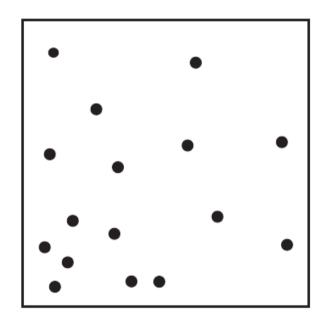
Freie Universität Berlin M.Sc. Landschaftsarchäologie

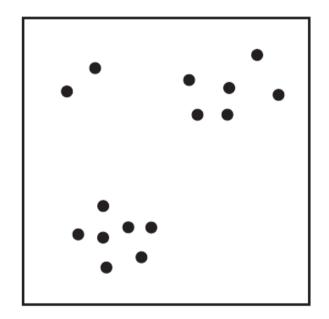
Dr. Daniel Knitter

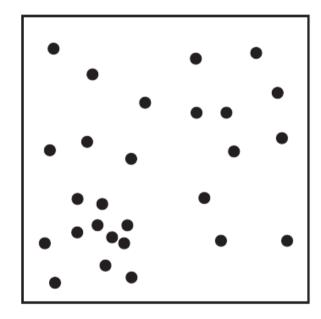


Point Pattern

What caused these patterns? First- or Second order effects?







Point pattern analysis aims to distinguish between both effects

– helping you to understand the genuine process creating the pattern –



Point Pattern

Simple measures:

- mean center
- standard distance
- intensity of a pattern

$$\bar{\mathbf{s}} = \left(\mu_x, \mu_y\right) = \left(\frac{\sum_{i=1}^n x_i}{n}, \frac{\sum_{i=1}^n y_i}{n}\right)$$

$$d = \sqrt{\frac{\sum_{i=1}^{n} (x_i - \mu_x)^2 + (y_i - \mu_y)^2}{n}}$$

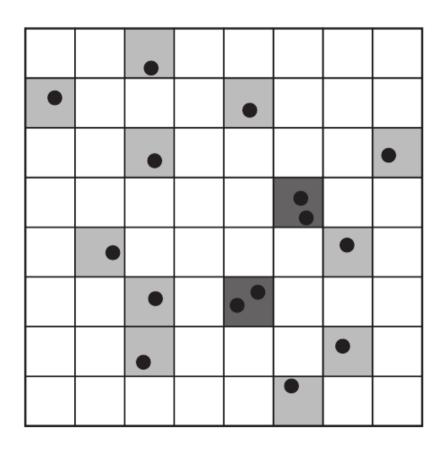
$$\hat{\lambda}=rac{n}{a}=rac{\#(S\in A)}{a}$$
 events in the pattern found in study region



Intensity – more advanced Independent Random Process (IRP)

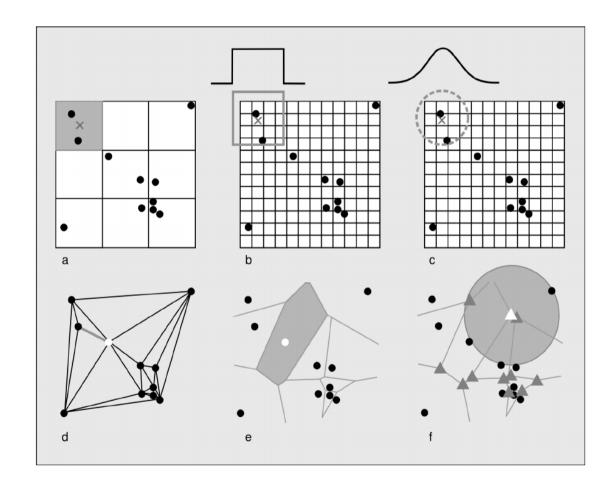
- = Complete Spatial Randomness (CSR)
- → any event has equal probability to be located a quadrant
- → the occurrence of points is independent of the Positioning of other events

P (event in a quadrant) = 1/64





Intensity – possibilities to calculate

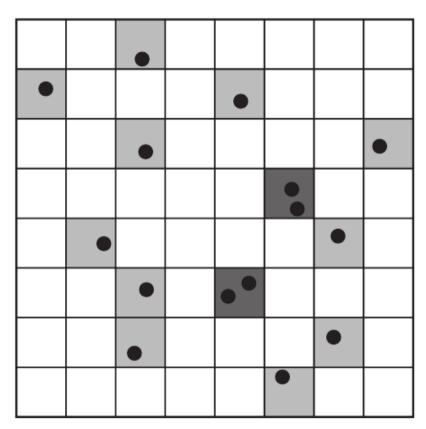




Intensity – more advanced
Independent Random Process (IRP)
= Complete Spatial Randomness (CSR)

Probability of k events in a quadrant is calculated with Poisson distribution

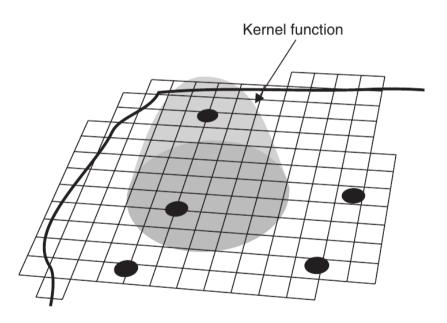
$$P(k) = \frac{\lambda^k e^{-\lambda}}{k!}$$
 intensity

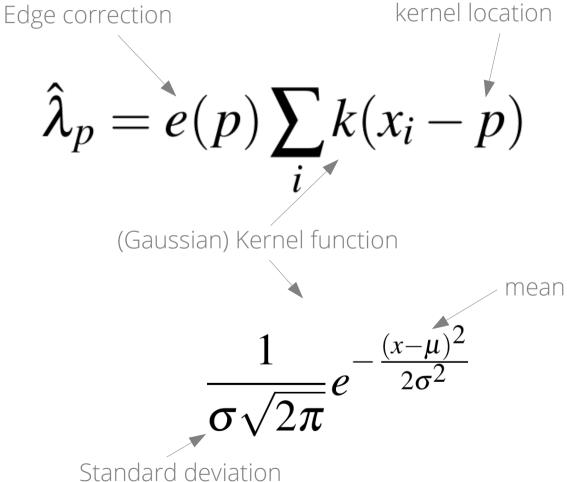


Base of natural logarithm (2.78...)



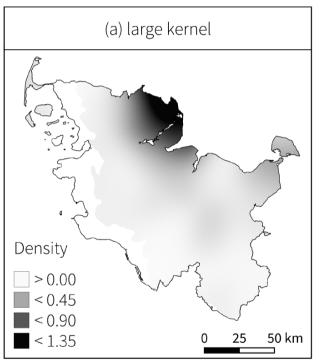
Intensity – more advanced Kernel density estimation (KDE)

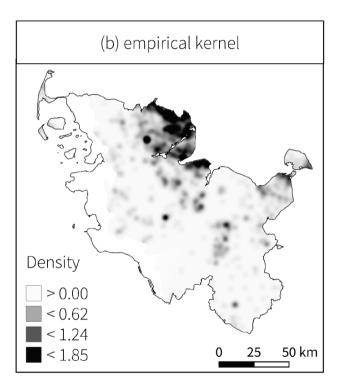




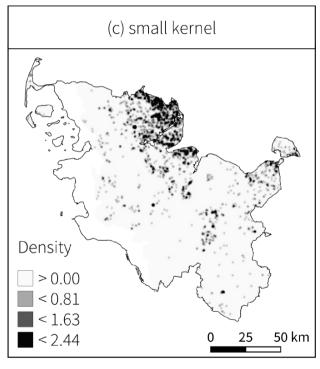


Sigma large





Sigma small



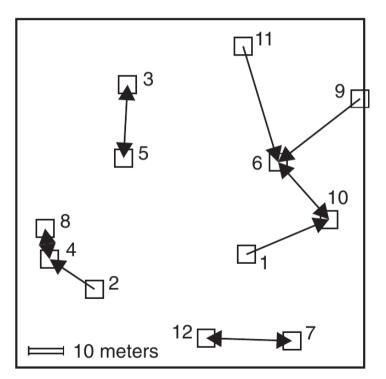
Local trend





Point patterns described by distance measures

- nearest-neighbor distance → euclidean distance



mean nearest-neighbor distance

$$ar{d}_{\min} = rac{\sum_{i=1}^n d_{\min}(\mathbf{s}_i)}{n}$$

Are the points clustered or dispersed?
Use Clark and Evan's *R* statistic of nearest neighbor distances

$$R=ar{d}_{
m min}igg/rac{1}{\left(2\sqrt{\lambda}
ight)}$$

more clustered < 1 > more dispersed

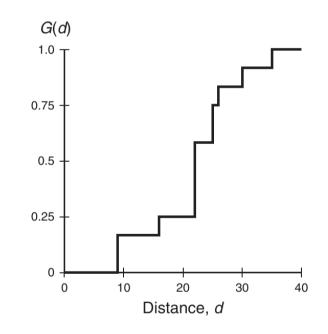


Point patterns described by distance measures

→ cumulative frequency distribution of the nearest-neighbor distances = G(d)

$$G(d) = \frac{\#(d_{\min}(\mathbf{s}_i) < d)}{n}$$

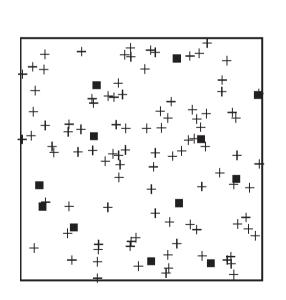
→ G(d) tells us what fraction of all n-n distances is less than d

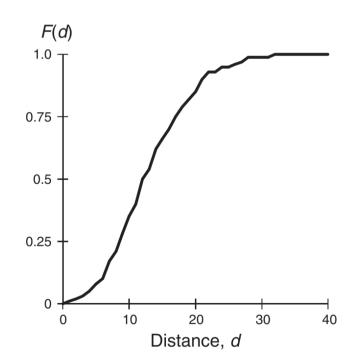




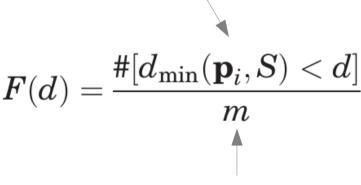
Point patterns described by distance measures

→ cumulative frequency distribution of the nearest-neighbor distances of arbitrary events to known events = F(d)



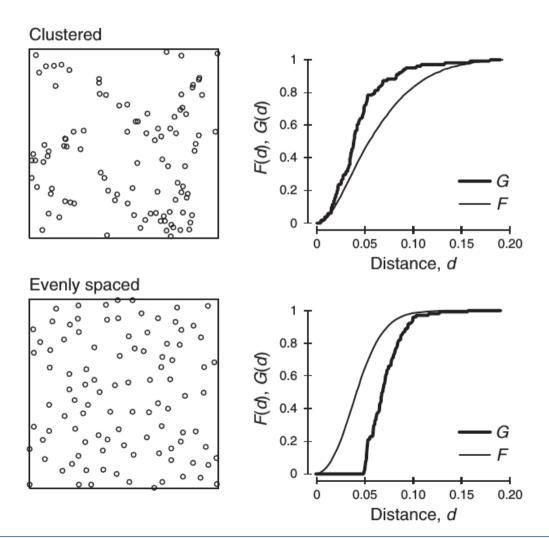


minimum distance from random point pi to an event

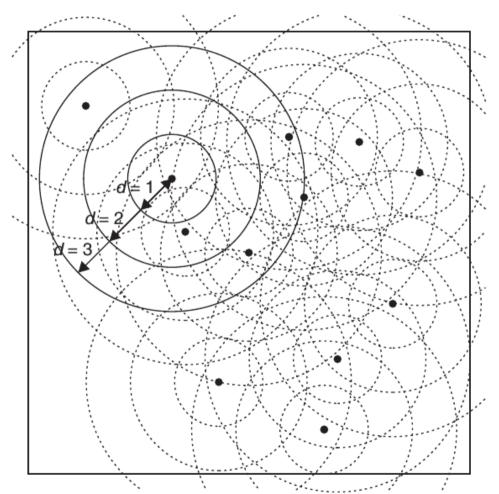


set of randomly selected locations









To get rid of the nearest-neighbor limitation: use the K Function

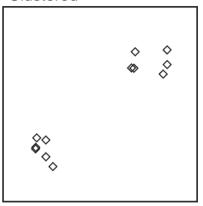
Events in circle radius d centered at s

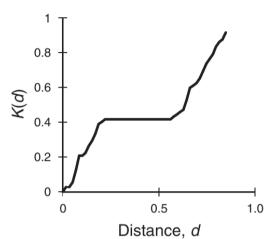
$$K(d) = rac{\sum_{i=1}^{n} \#[S \in C(\mathbf{s}_i, d)]}{n\lambda}$$

Event density in the study area

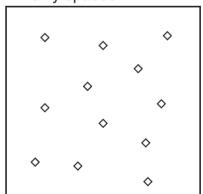


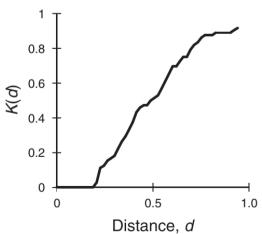






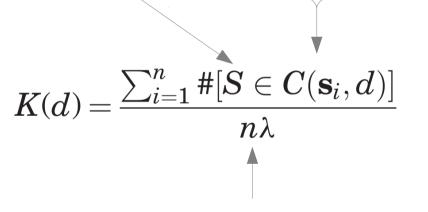
Evenly spaced





To get rid of the nearest-neighbor limitation: use the K Function

Events in circle radius d centered at s

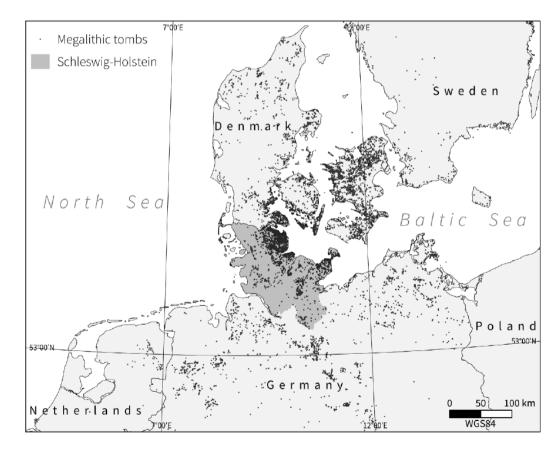


Event density in the study area



Shows the point pattern clustering/dispersion; does it deviate from CSR?

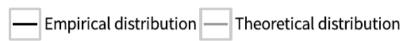
→ Simulation approach based on Monte Carlo simulations





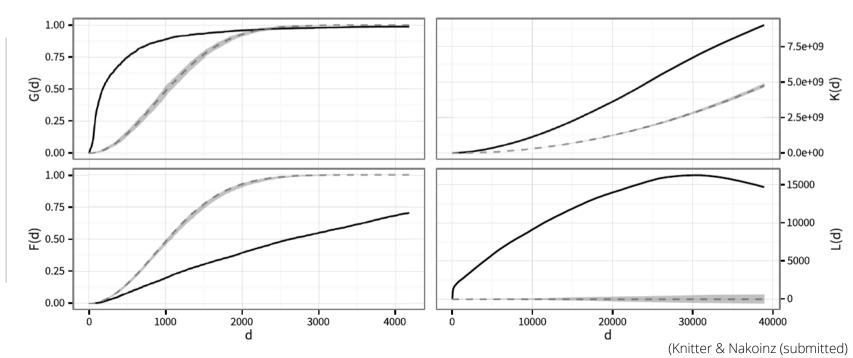
Shows the point pattern clustering/dispersion; does it deviate from CSR?

→ Simulation approach based on Monte Carlo simulations



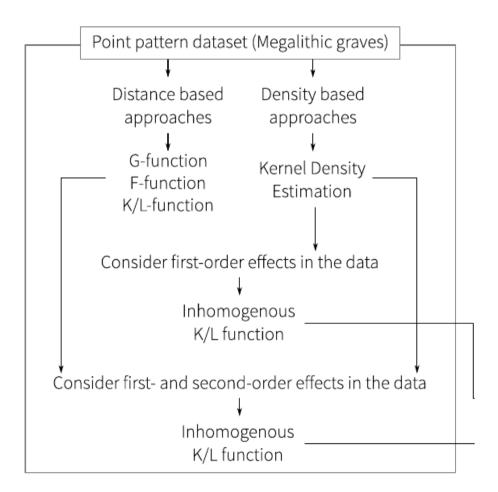
$$L(d) = \sqrt{\frac{K(d)}{\pi}}$$

Just a square root transformation of K(d)





Point Pattern – a look ahead





Point Pattern – a look ahead

