

SOEE1475 Statistics and Data Analysis

Lecture 5: Spatial statistics



Graeme T. Lloyd



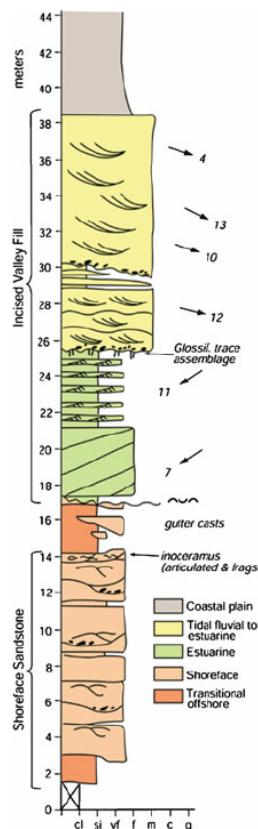
Today

- Spatial patterns of point distributions
- Chi-square test
- Nearest neighbour distances

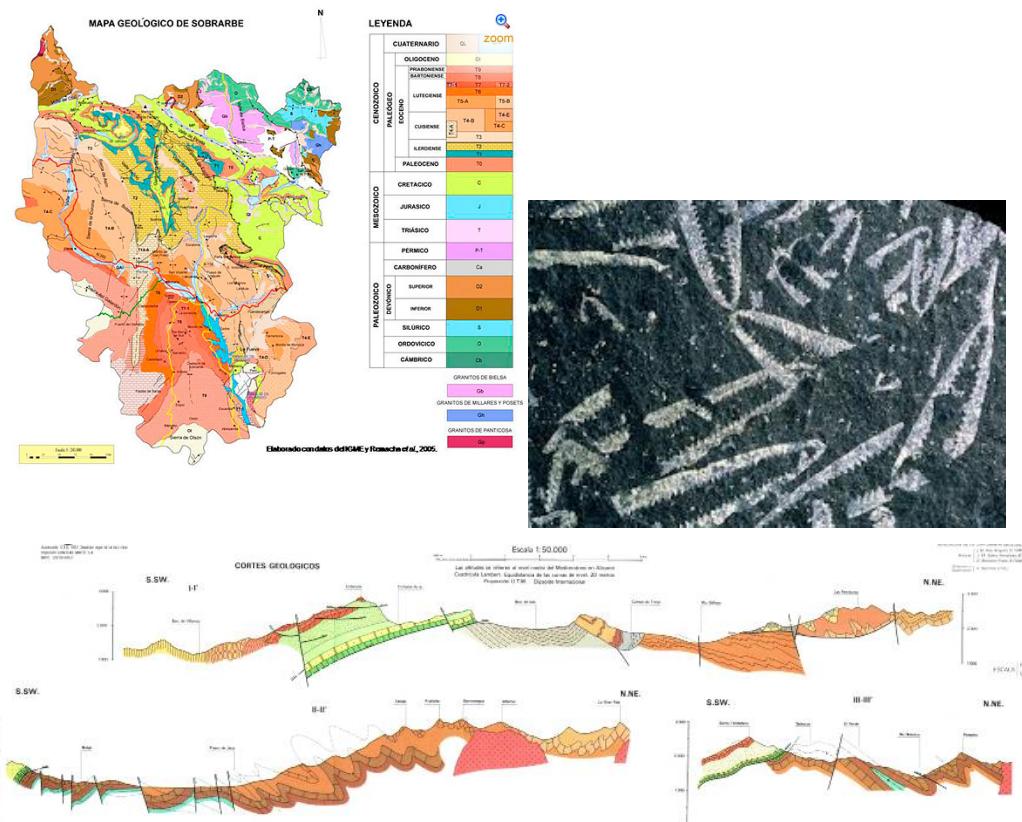


Spatial data in geology

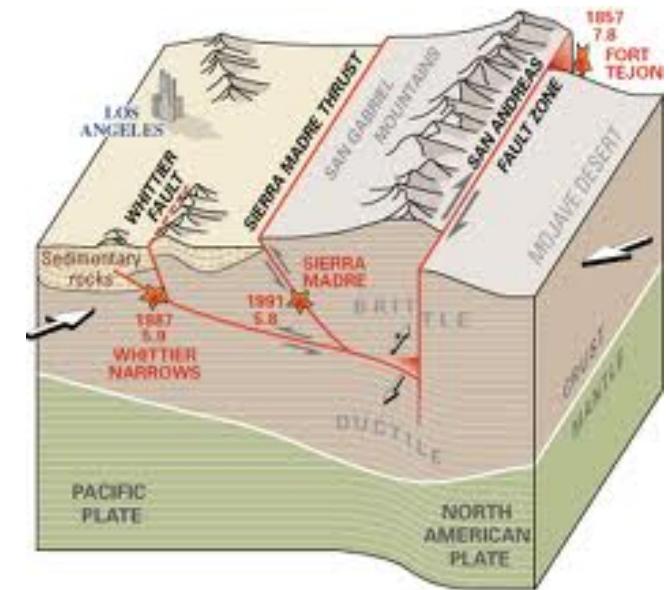
1D



2D

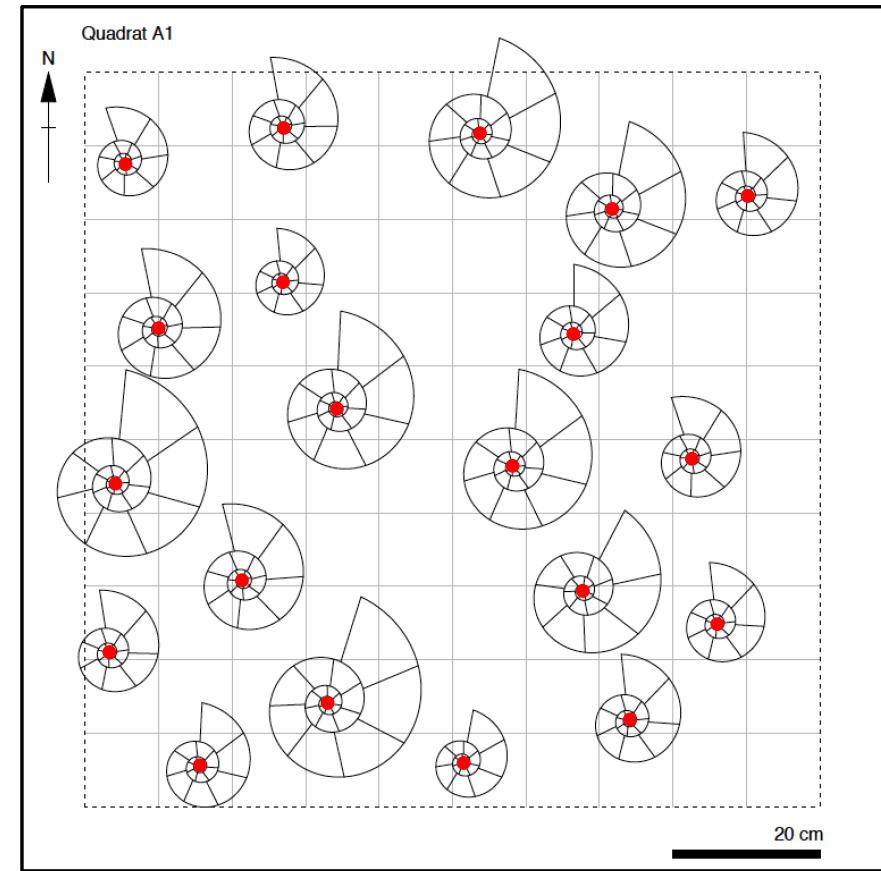
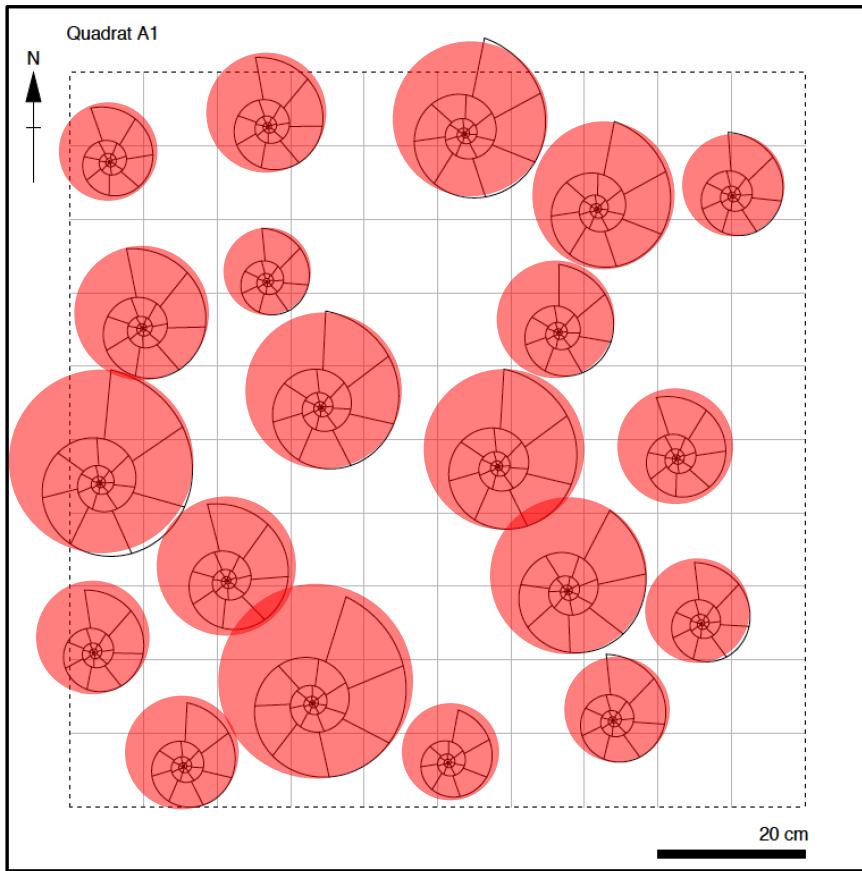


3D





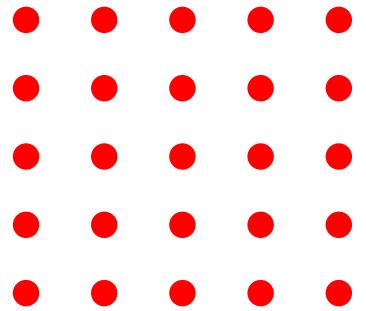
Point distributions



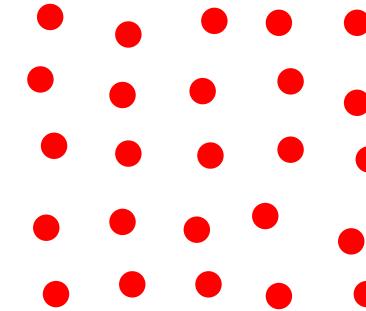


Point distributions

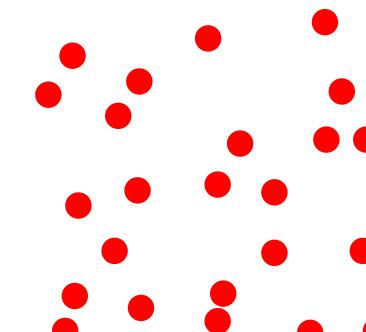
Isotropic



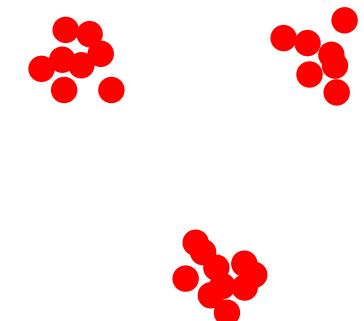
Regular



Uniform

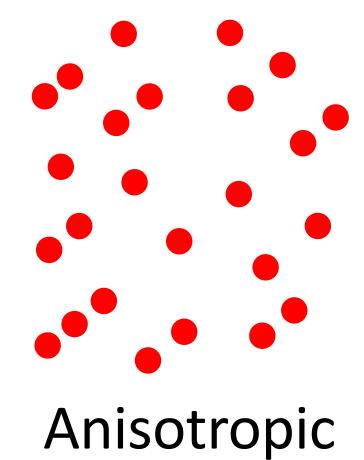
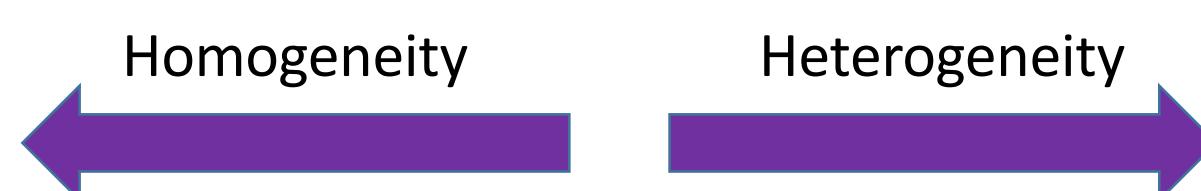


Random



Clustered

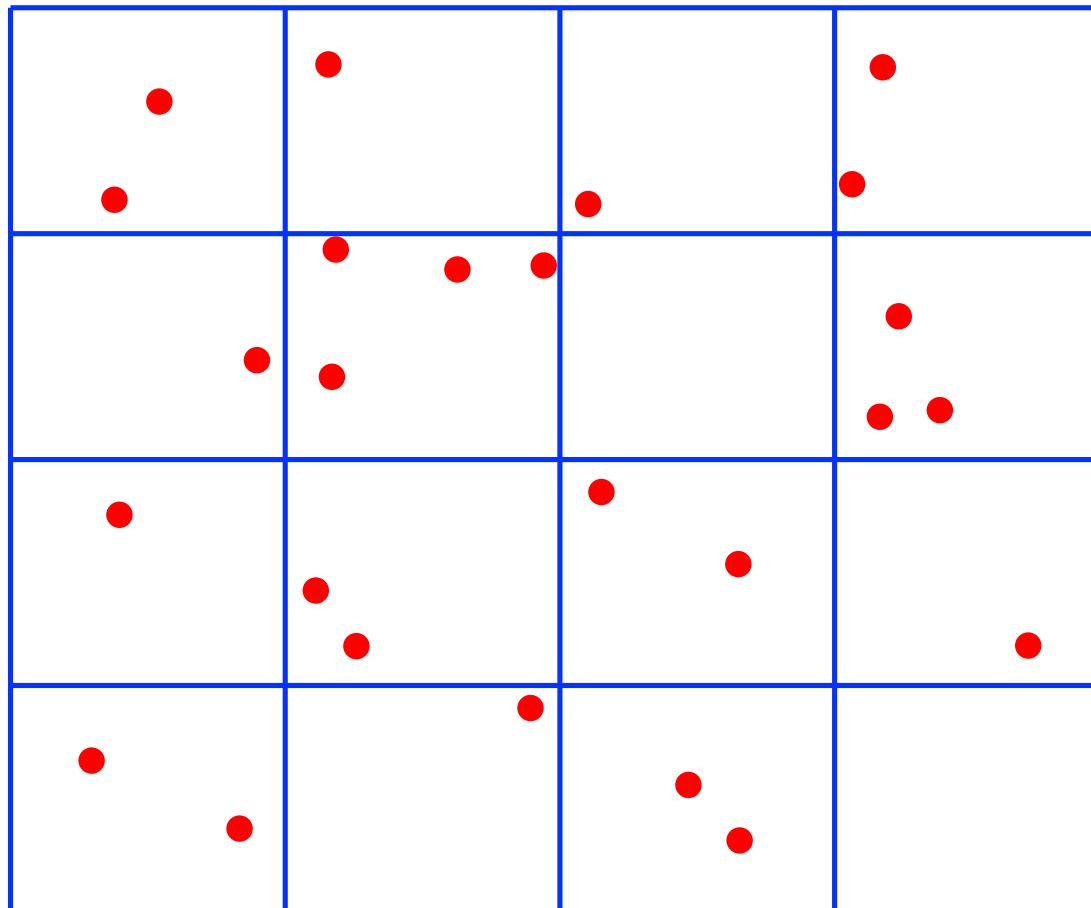
Anisotropic



Anisotropic

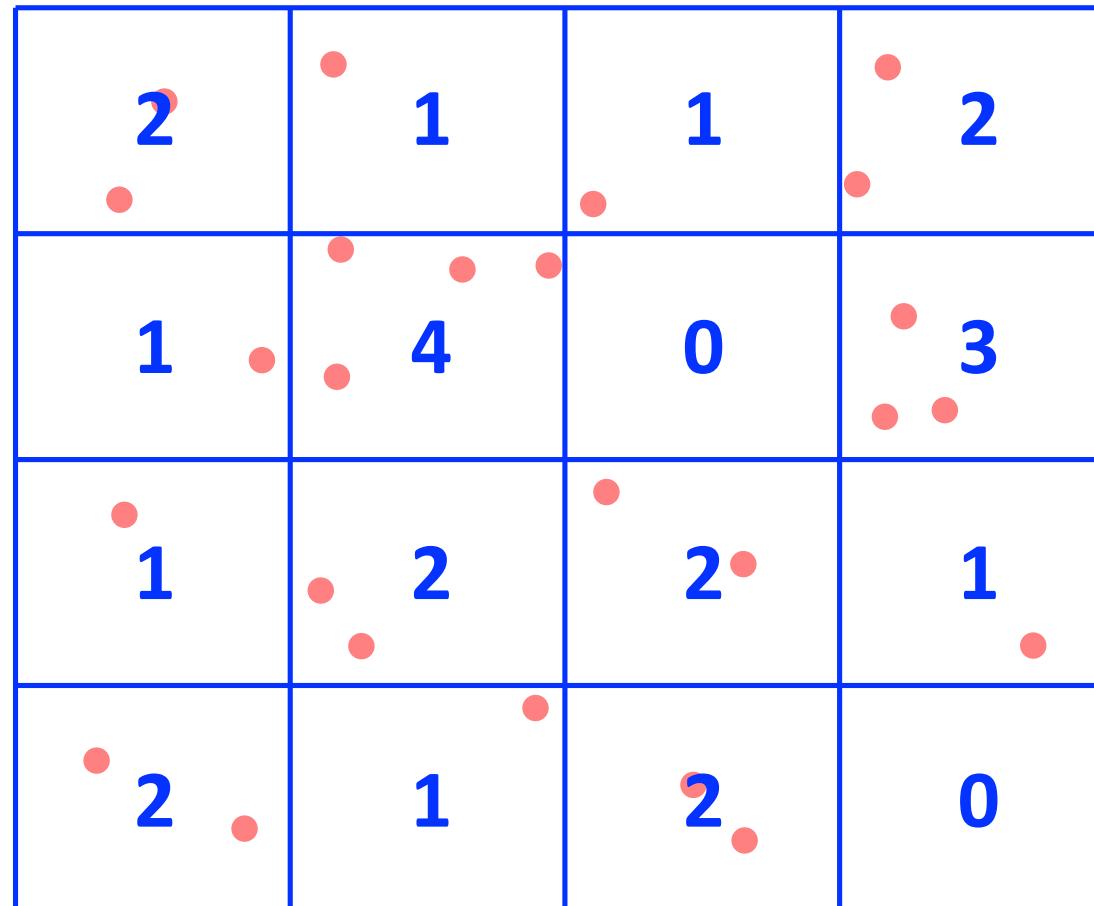


Point distributions



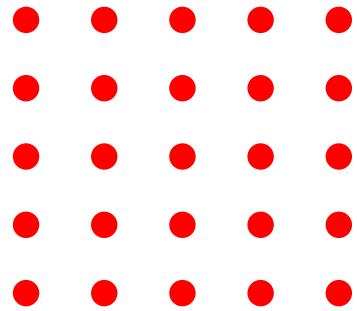


Point distributions

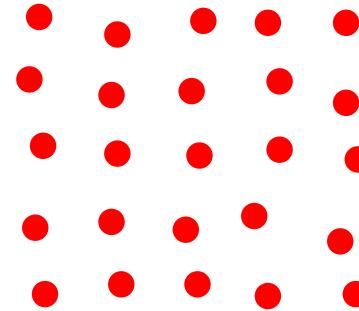




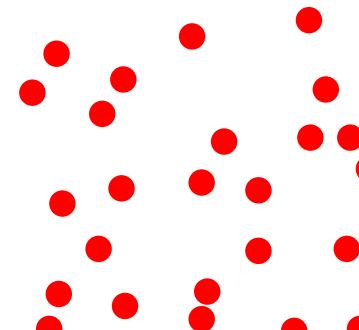
Point distributions



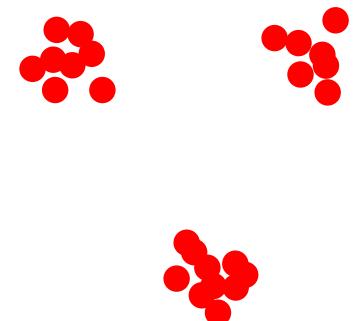
Regular



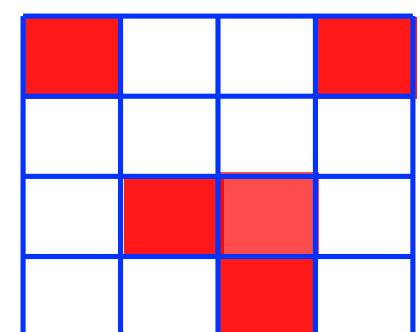
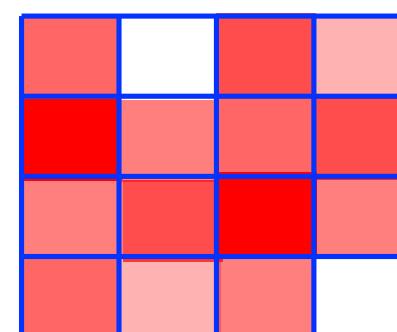
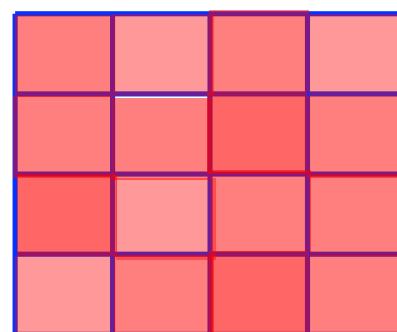
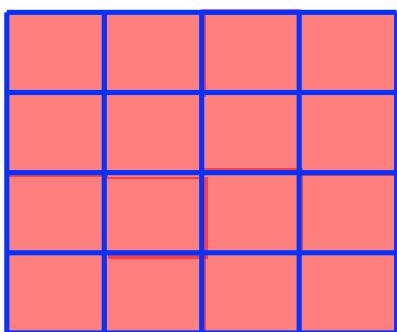
Uniform



Random

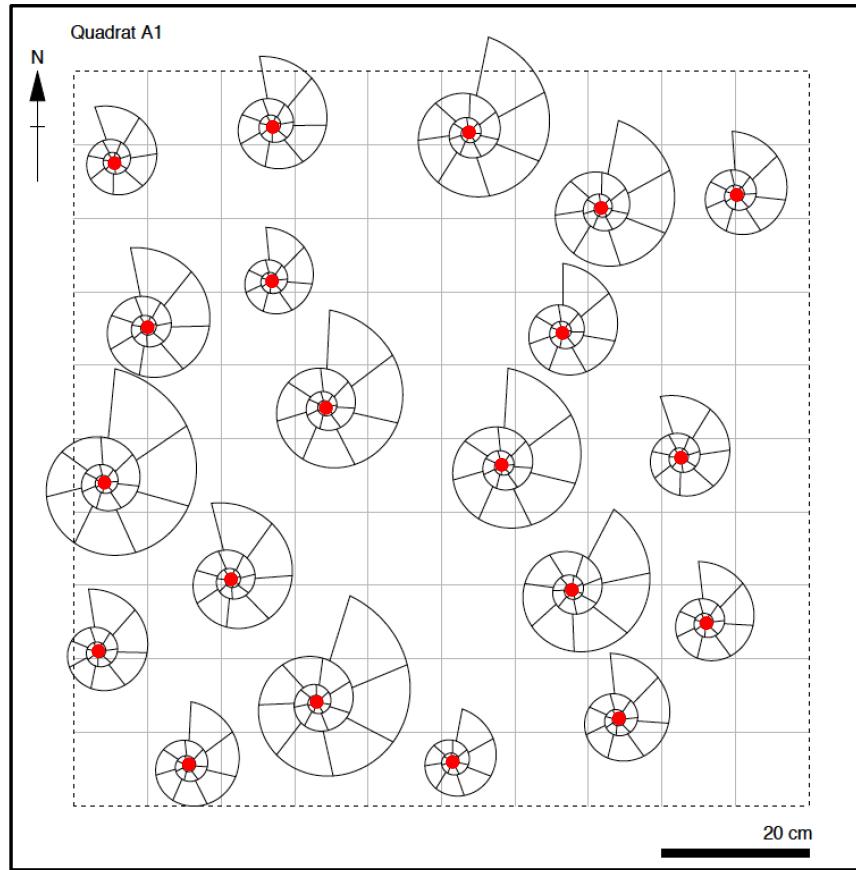


Clustered



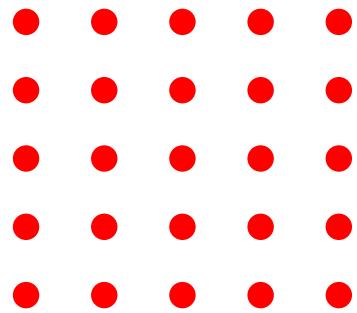


Exercise – part I

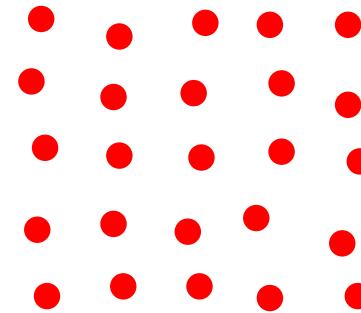




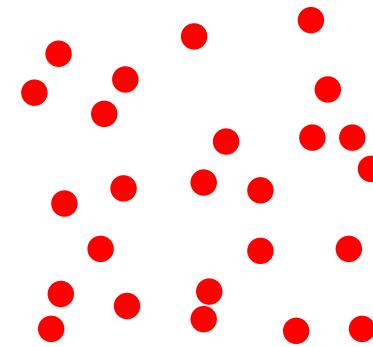
Exercise – part I



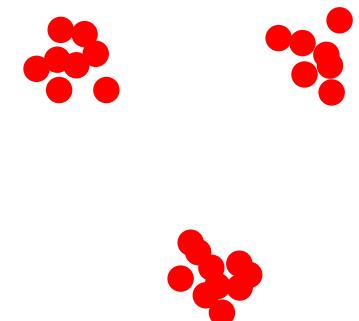
Regular



Uniform



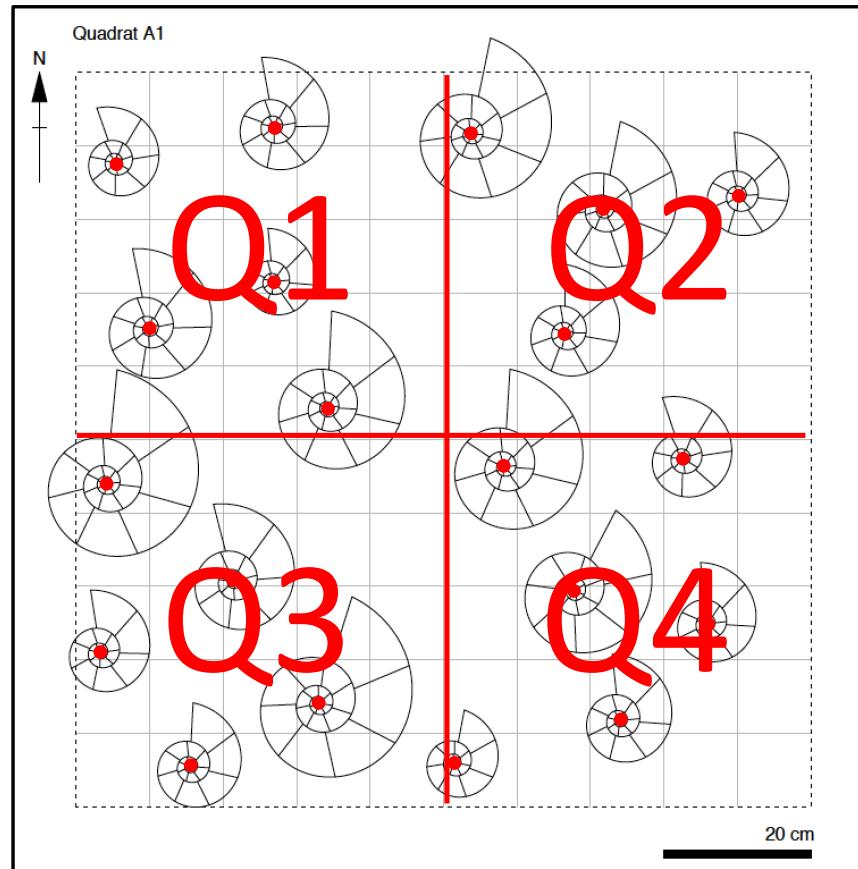
Random



Clustered



Exercise – part I

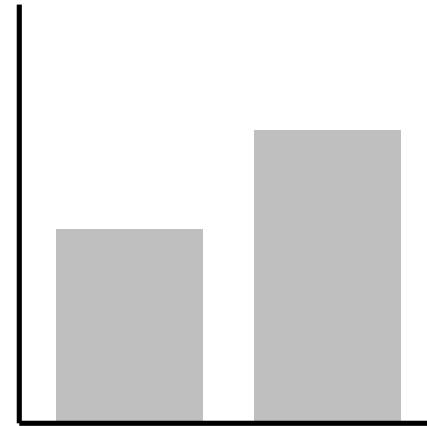




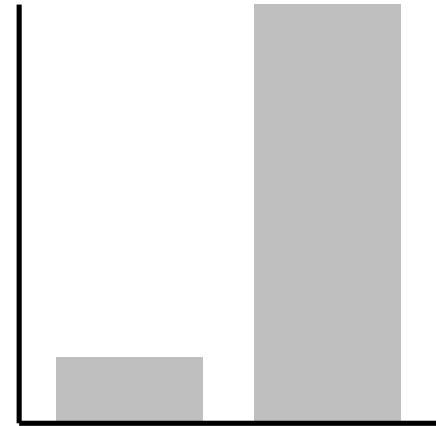
The Chi-squared test – testing for uniform

$$\chi^2 = \sum_{j=1}^k \left[\frac{(O_j - E_j)^2}{E_j} \right]$$

Low χ^2



High χ^2

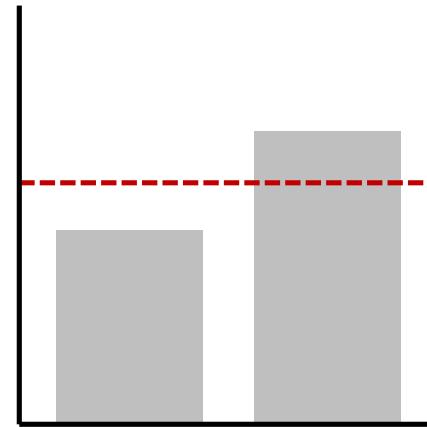




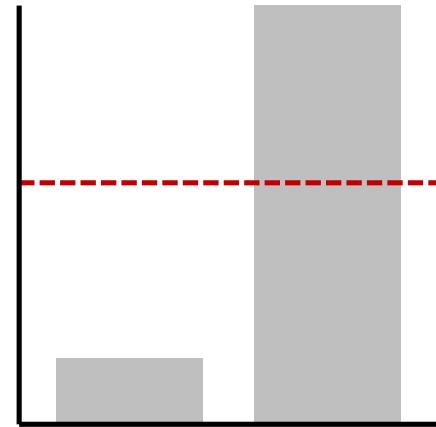
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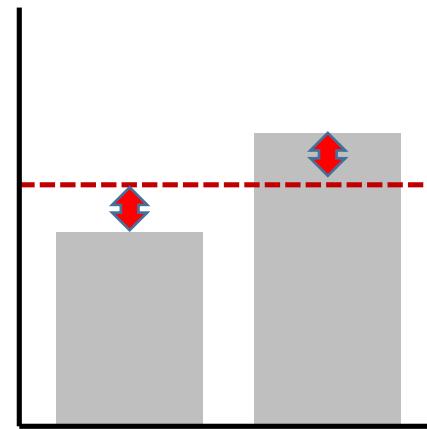




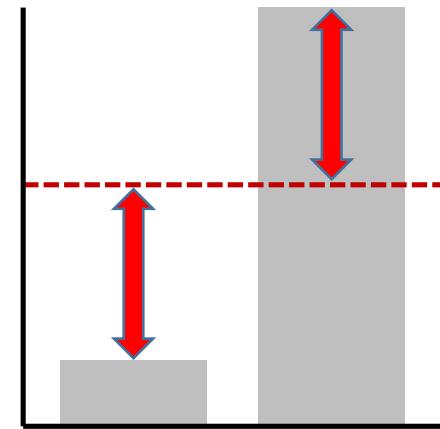
The Chi-squared test – testing for uniform

$$\chi^2 = \sum_{j=1}^k \left[\frac{(O_j - E_j)^2}{E_j} \right]$$

Low χ^2



High χ^2





Exercise – part II

$$\chi^2 = \sum_{j=1}^k \left[\frac{(O_j - E_j)^2}{E_j} \right]$$

$$E_j = \frac{N_{\text{ammonites}}}{N_{\text{quadrants}}}$$

$$\chi^2 = \frac{(Q1-5)^2}{5} + \frac{(Q2-5)^2}{5} + \frac{(Q3-5)^2}{5} + \frac{(Q4-5)^2}{5}$$



The Chi-squared test – testing for uniform

$$0 \leq \chi^2 \leq \infty$$



Degrees of freedom

Sample standard deviation

$$s = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}$$

t-distribution parameter

$$\nu = n - 1$$

Pearson correlation coefficient

$$r = \frac{\sum_{i=1}^n (Z_{xi} \times Z_{yi})}{n - 1}$$

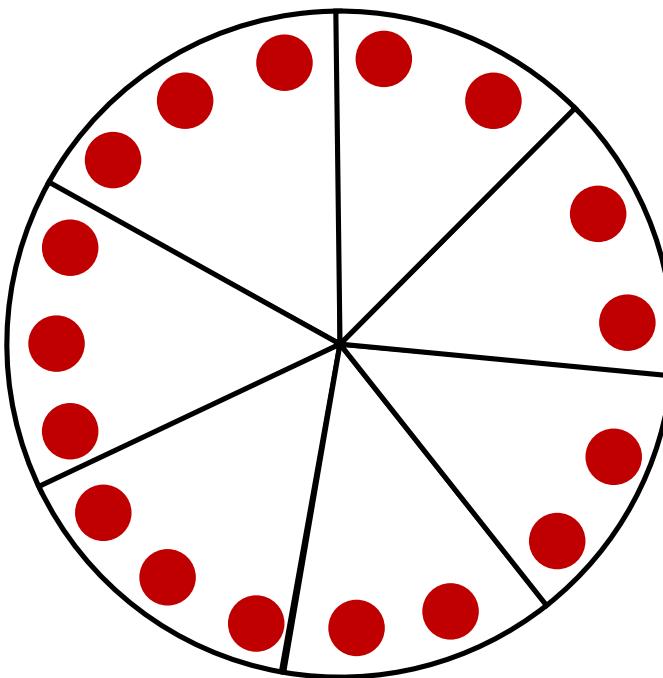
F-distribution parameters

$$d_1 = N_{replicates} - 1$$

$$d_2 = N_{treatments} - 1$$



Degrees of freedom





Degrees of freedom

Person 1: Choice

Person 2: Choice

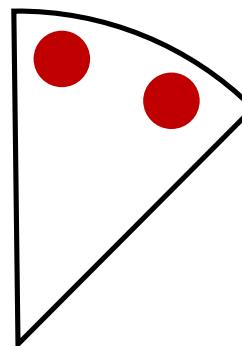
Person 3: Choice

Person 4: Choice

Person 5: Choice

Person 6: Choice

Person 7: No choice





Degrees of freedom

$$x_1 = 5$$

$$x_1 - \bar{x} = -5$$

$$x_2 = 10$$

$$\bar{x} = 10$$

$$x_2 - \bar{x} = 0$$

$$x_3 = 15$$

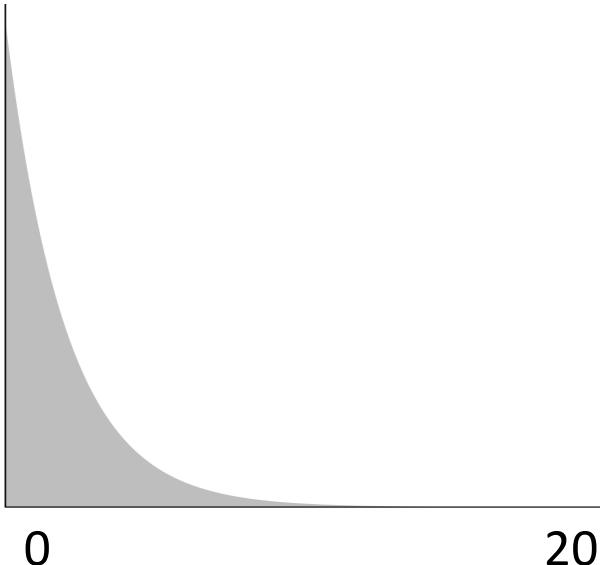
$$x_3 - \bar{x} = ?$$



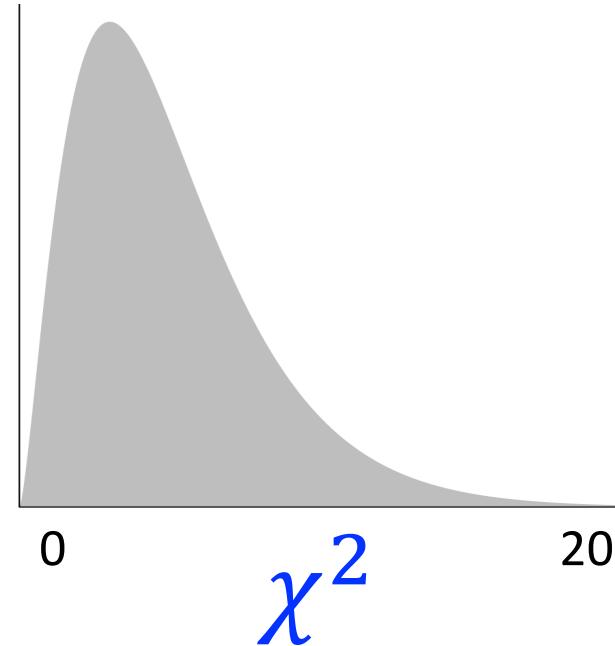
Degrees of freedom

$$0 \leq \chi^2 \leq \infty$$

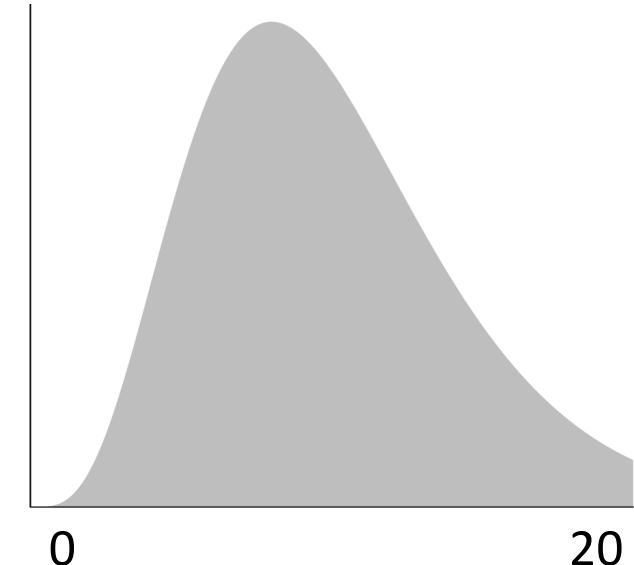
$df = 3 - 1$



$df = 6 - 1$

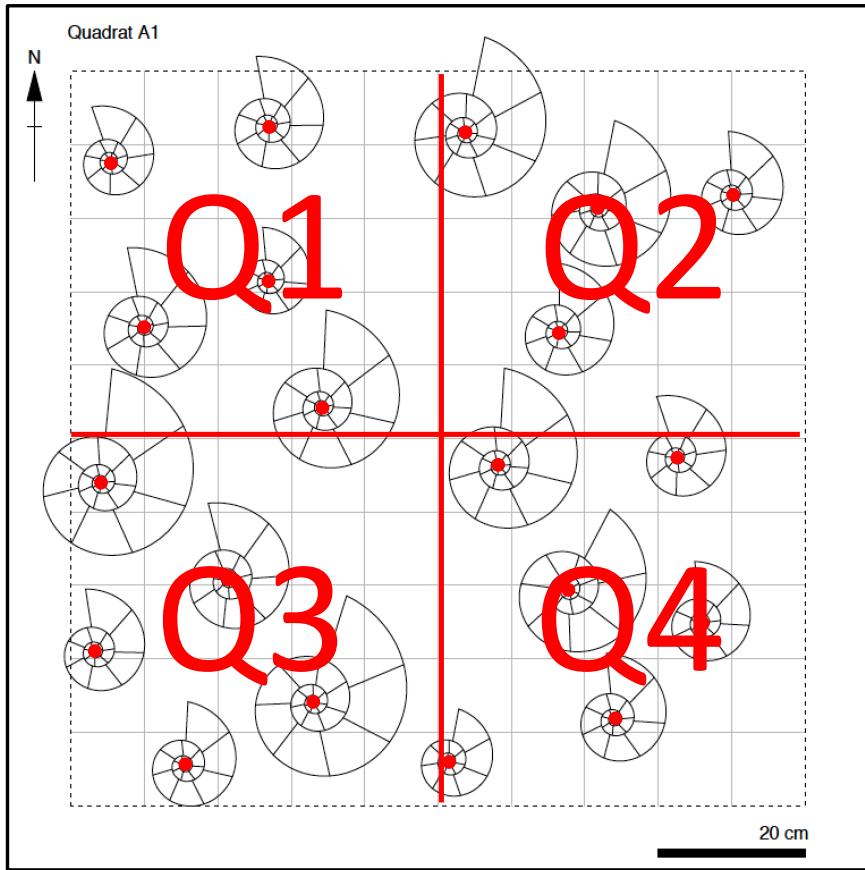


$df = 11 - 1$





Exercise – part III



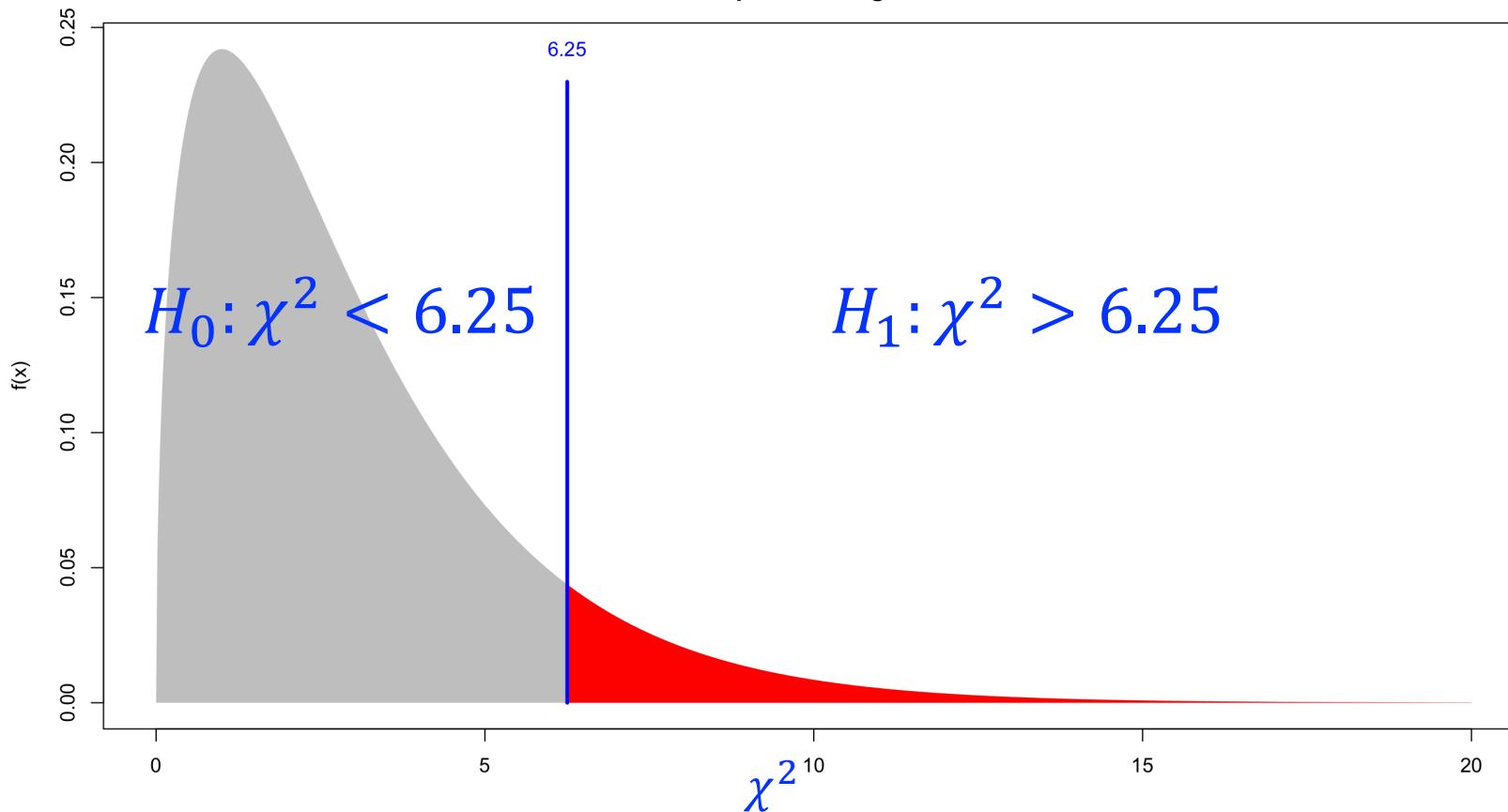
$$\chi^2 = \frac{(Q1-5)^2}{5} + \frac{(Q2-5)^2}{5} + \frac{(Q3-5)^2}{5} + \frac{(Q4-5)^2}{5}$$

df = ?



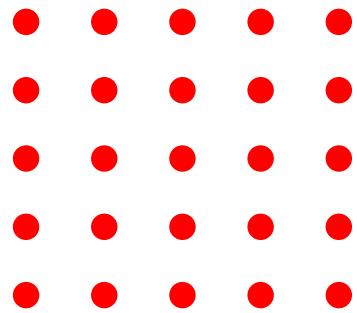
Exercise – part III

Null: data are uniformly distributed
df = 4 - 1; Alpha = 0.1; right-tailed test

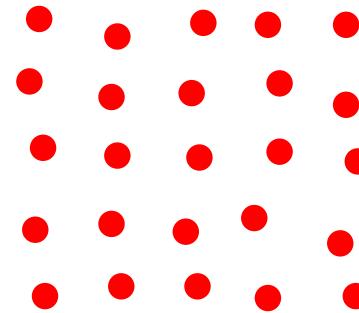




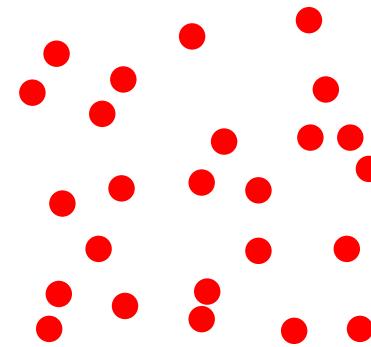
Exercise – part III



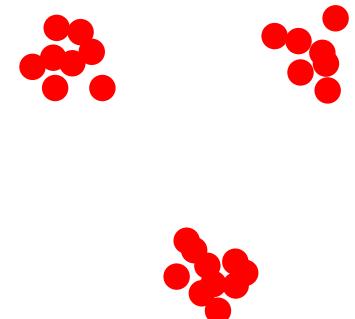
Regular



Uniform



Random



Clustered



Poisson distribution – testing for random

Discrete probability distribution

Has single parameter (λ), the mean

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

Expresses probability of given number of events

Assumes randomness; events can occur in space or time

Good null for random distribution of points in space

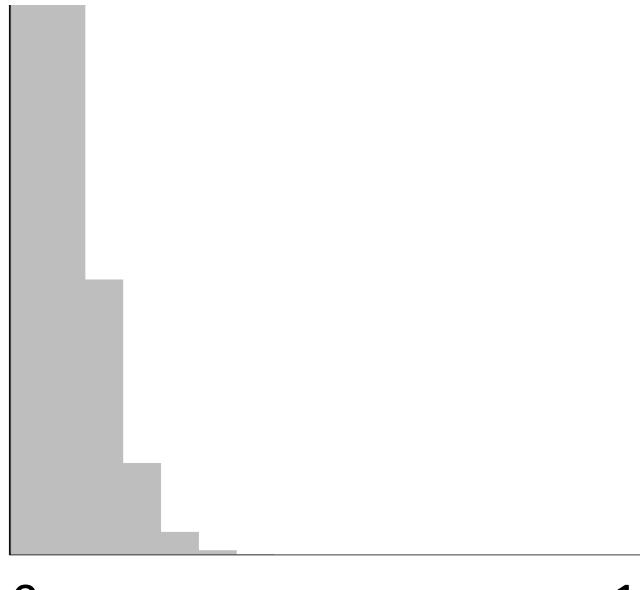
i.e., in ammonite example expected N quadrants with k ammonites



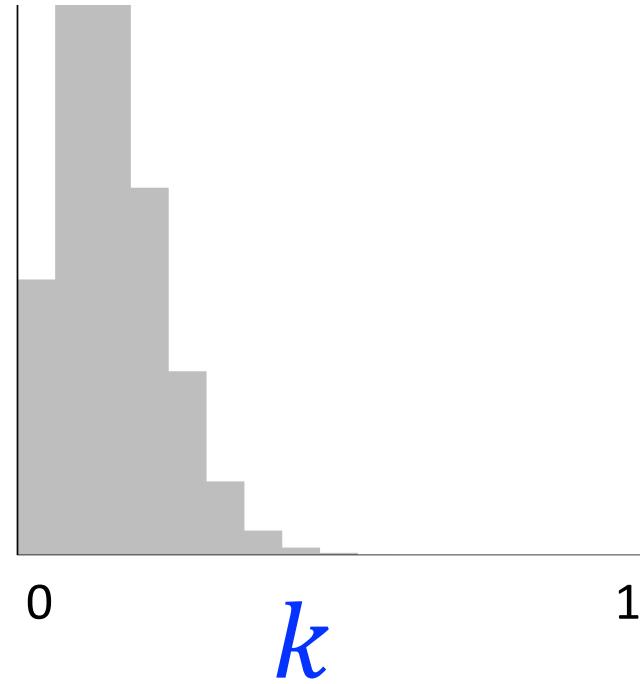
Poisson distribution – testing for random

$$0 < \lambda \leq \infty$$

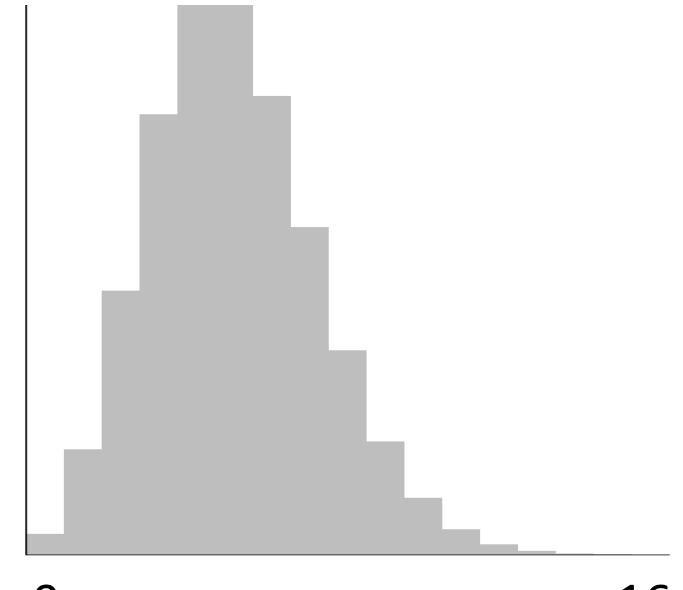
$\lambda = 1$



$\lambda = 2$



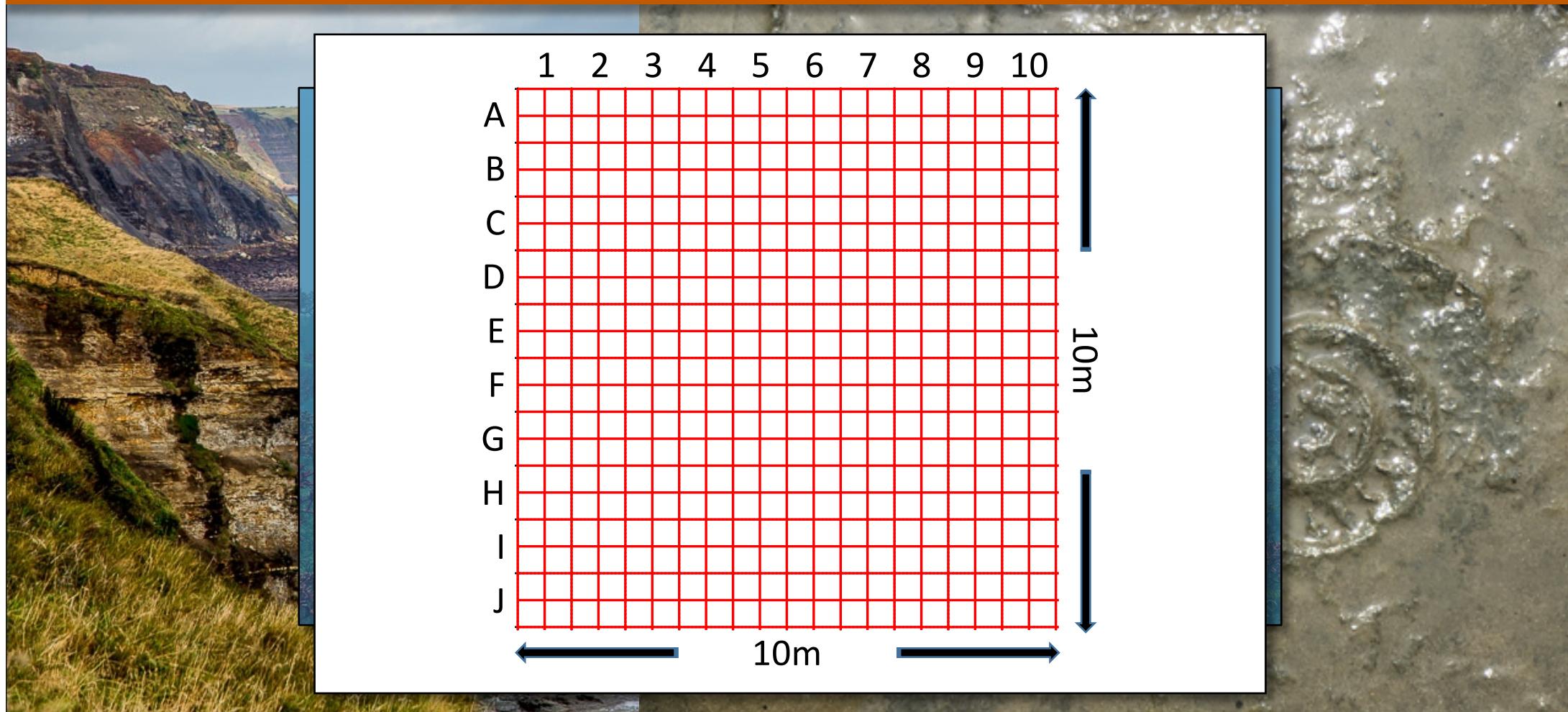
$\lambda = 5$



k



Poisson distribution – testing for random





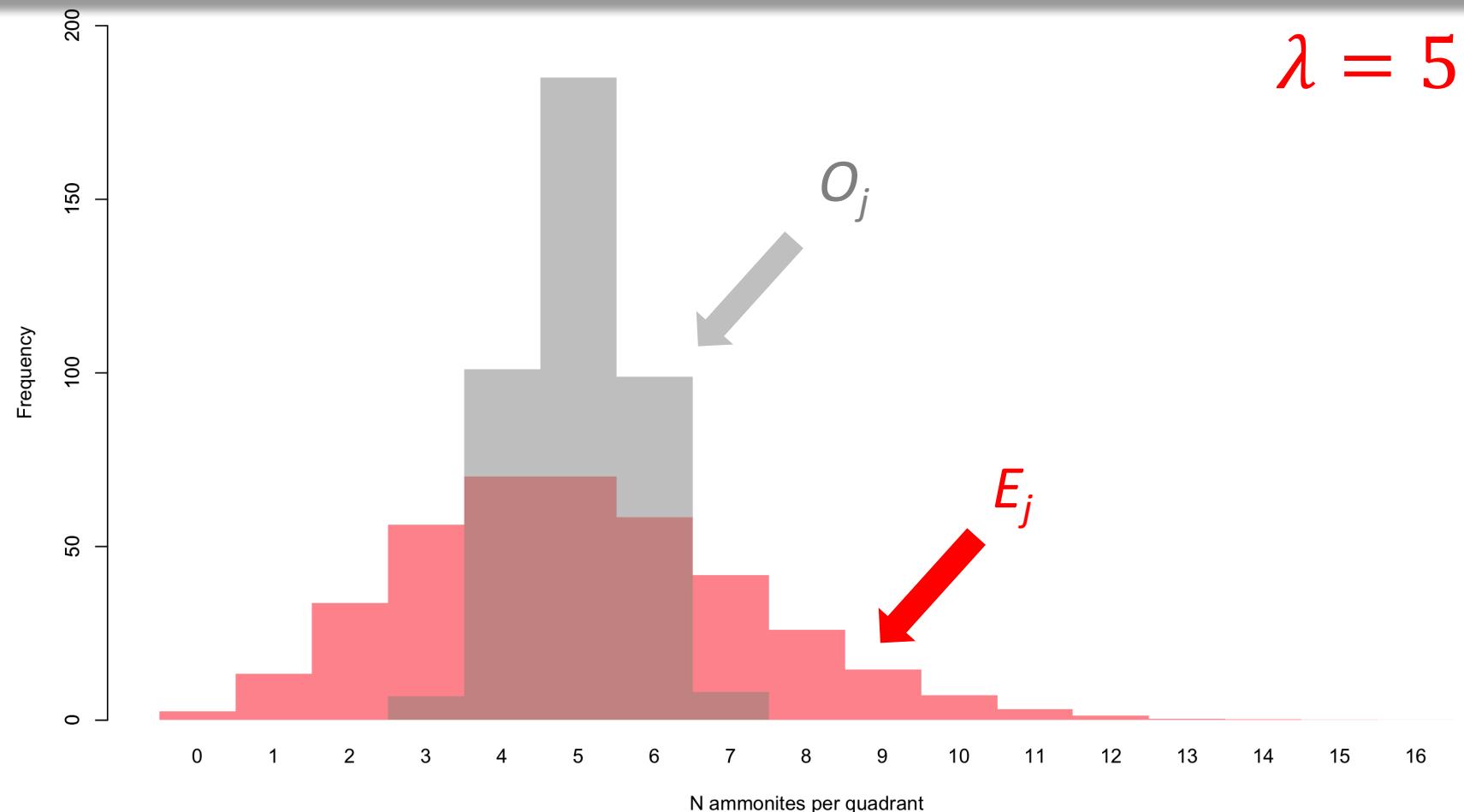
Poisson distribution – testing for random

$$\lambda = \frac{N_{\text{ammonites}}}{N_{\text{quadrants}}}$$

$$\lambda = \frac{2000}{4 \times 100} = 5$$



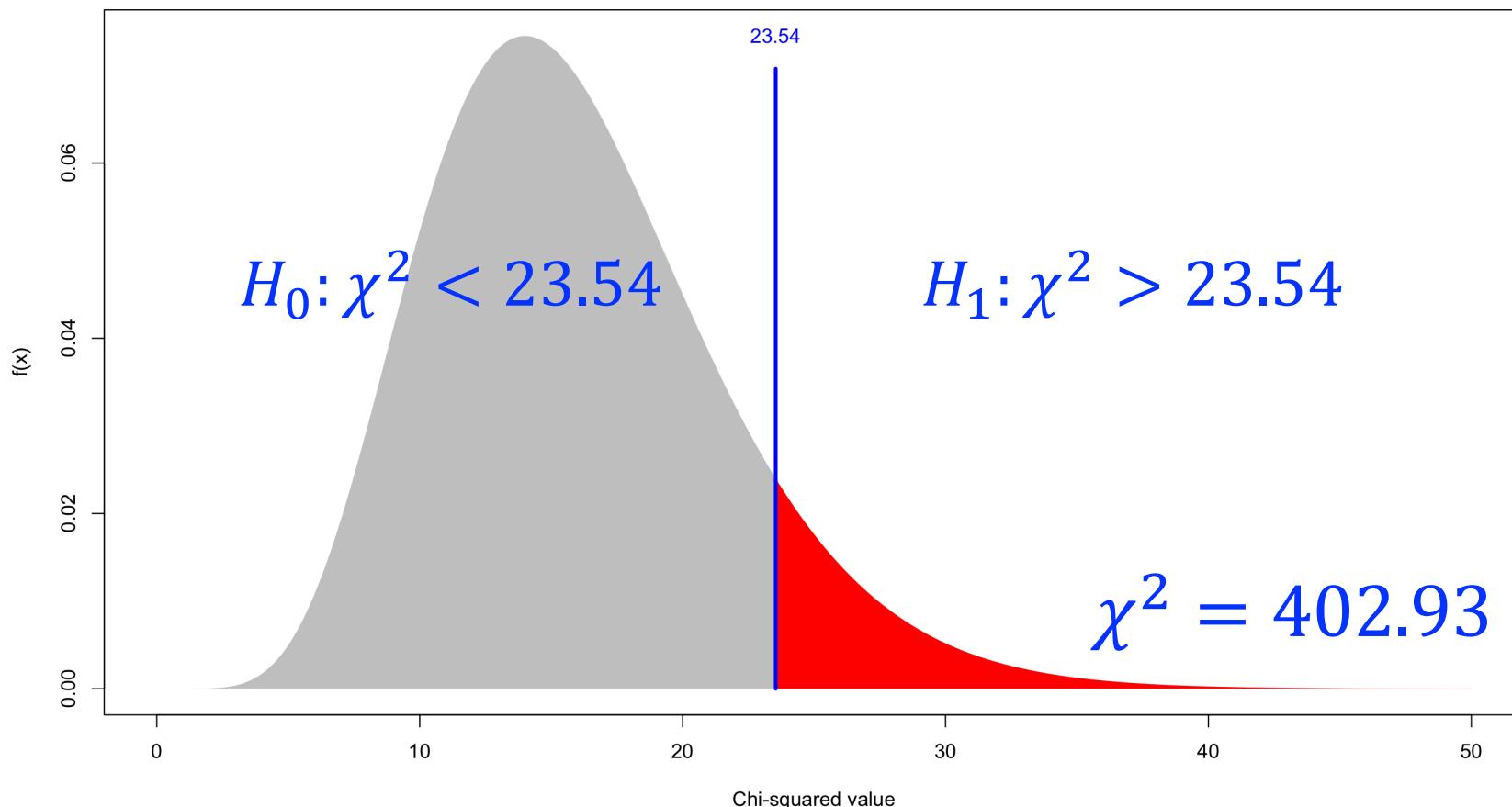
Poisson distribution – testing for random





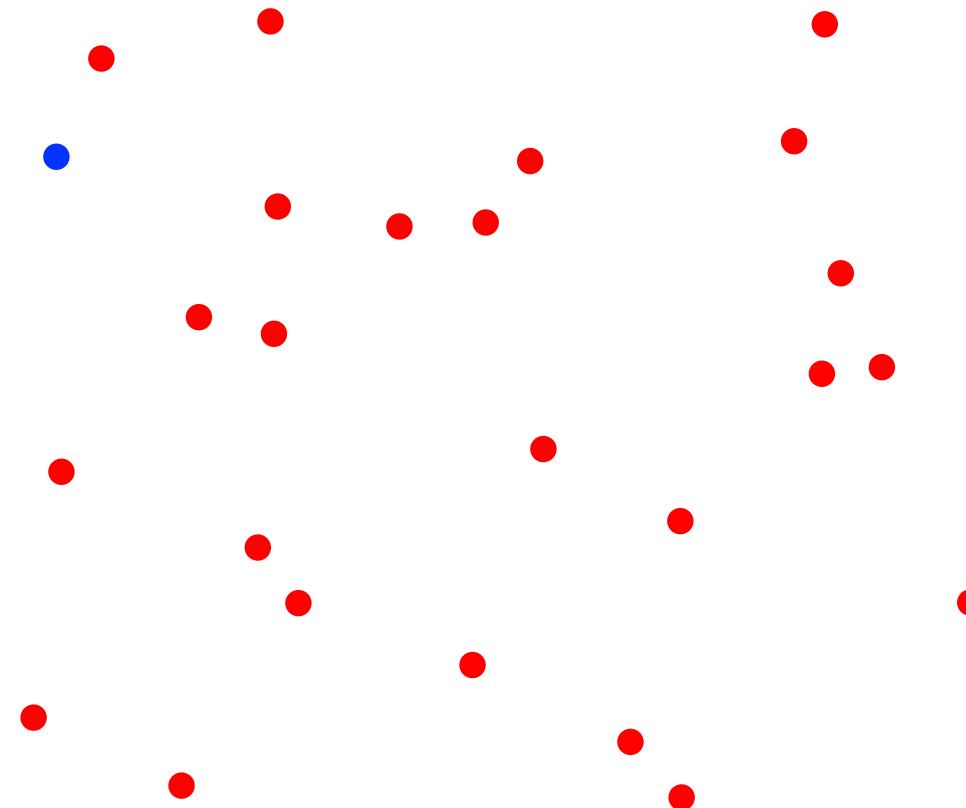
Poisson distribution – testing for random

Null: data are randomly distributed
df = 17 - 1; Alpha = 0.1; right-tailed test



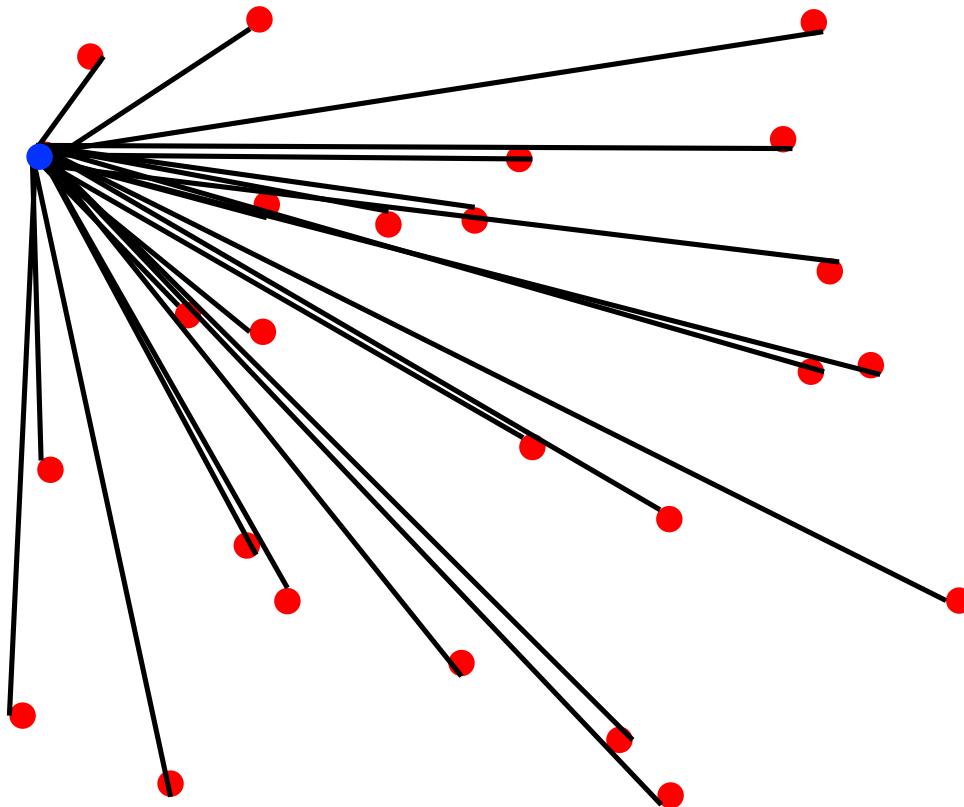


Other tests - nearest neighbour



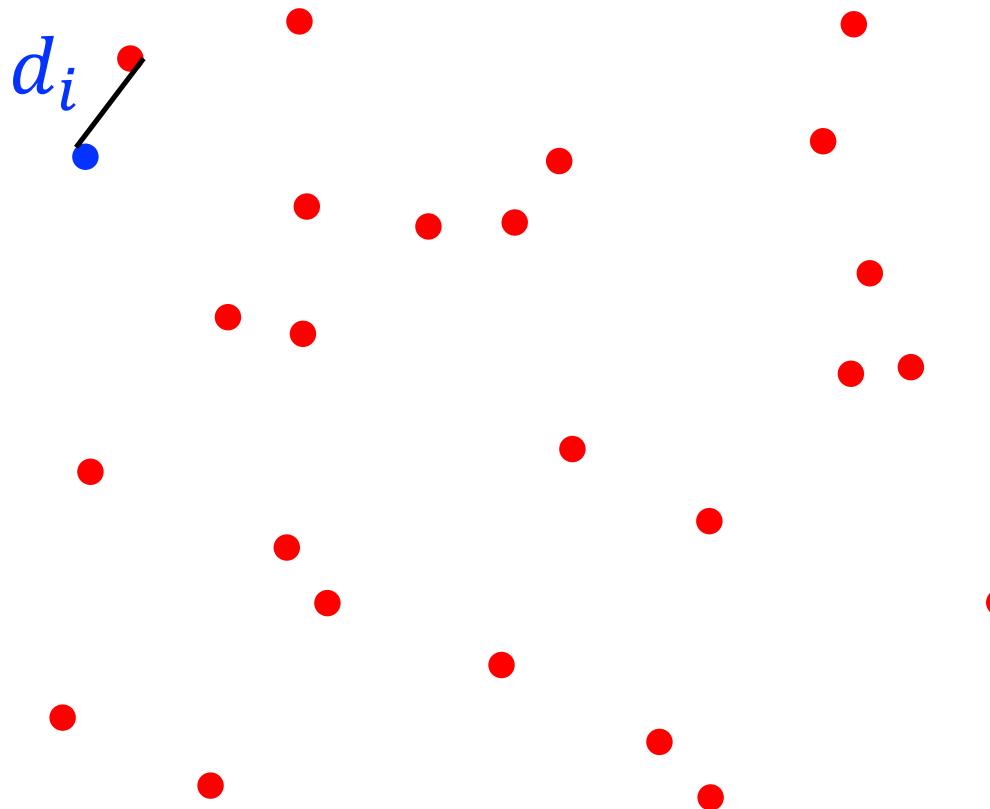


Other tests - nearest neighbour





Other tests - nearest neighbour





Other tests - nearest neighbour

Mean nearest neighbour distance

Captures average distance to nearest point

$$\bar{d} = \frac{1}{n} \sum_{i=1}^n d_i$$

Expected mean nearest neighbour distance

Uses total area (A) and number of objects (n)

$$\bar{\delta} = \frac{1}{2} \sqrt{A/n}$$

Nearest neighbour statistic

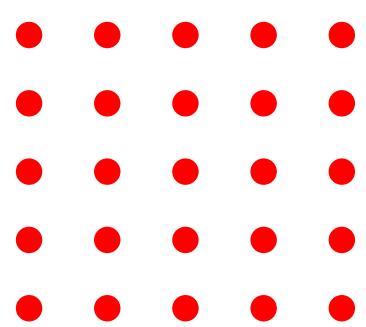
Varies from 0 to 2.15

$$R = \frac{\bar{d}}{\bar{\delta}}$$

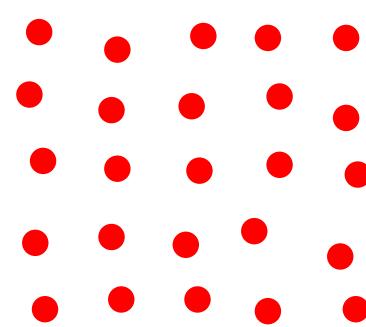


Other tests - nearest neighbour

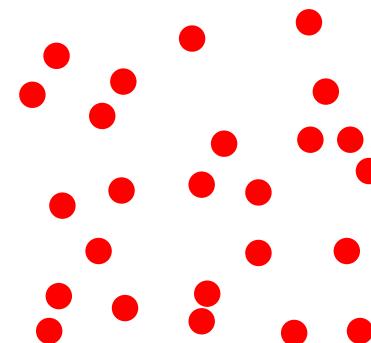
Nearest neighbour statistic



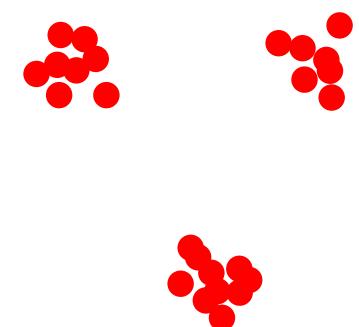
Regular



Uniform



Random



Clustered

$$R = 2.15$$

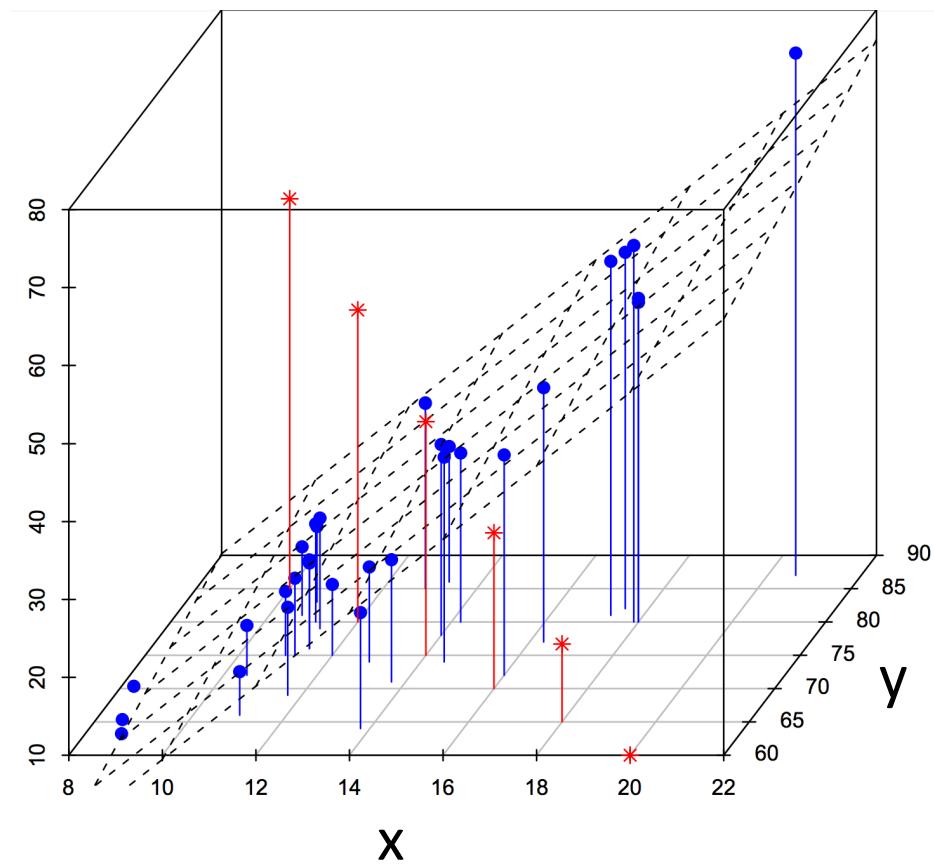
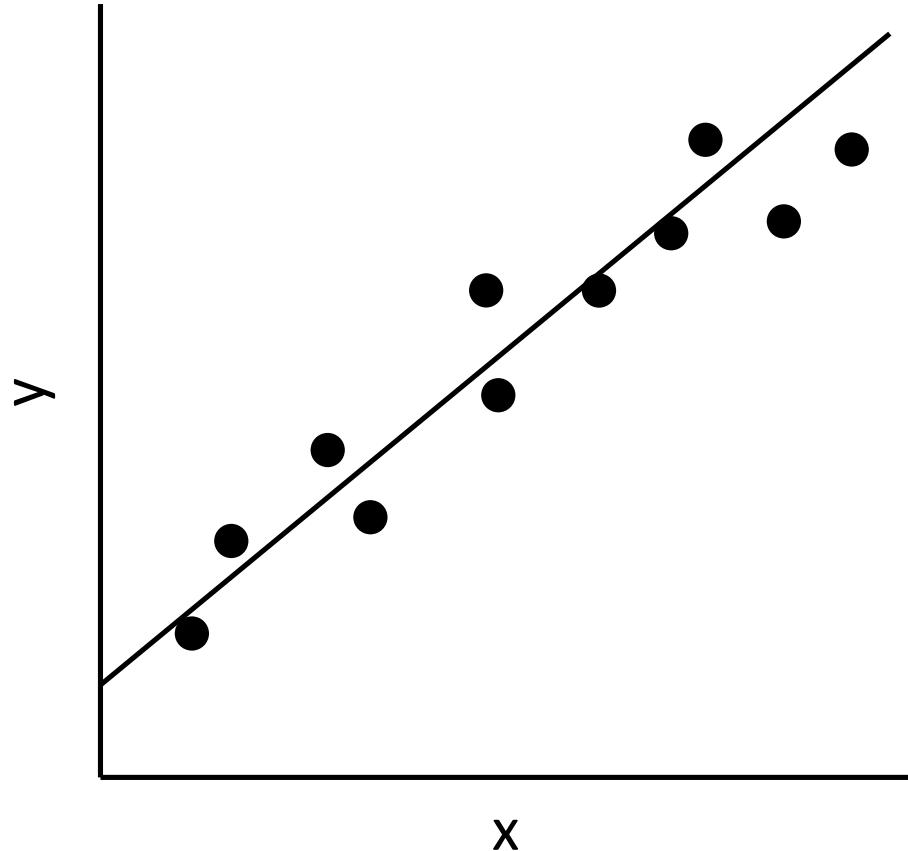
$$2.15 > R > 1$$

$$R \approx 1$$

$$R < 1$$



3D regression





3D regression

Planes

$$z = ax + by + c$$

3D equivalent of straight line

Domes/bowls

$$z = ax + by + cx^2 + dy^2 + exy + f$$

3D equivalent of (simple) curvilinear

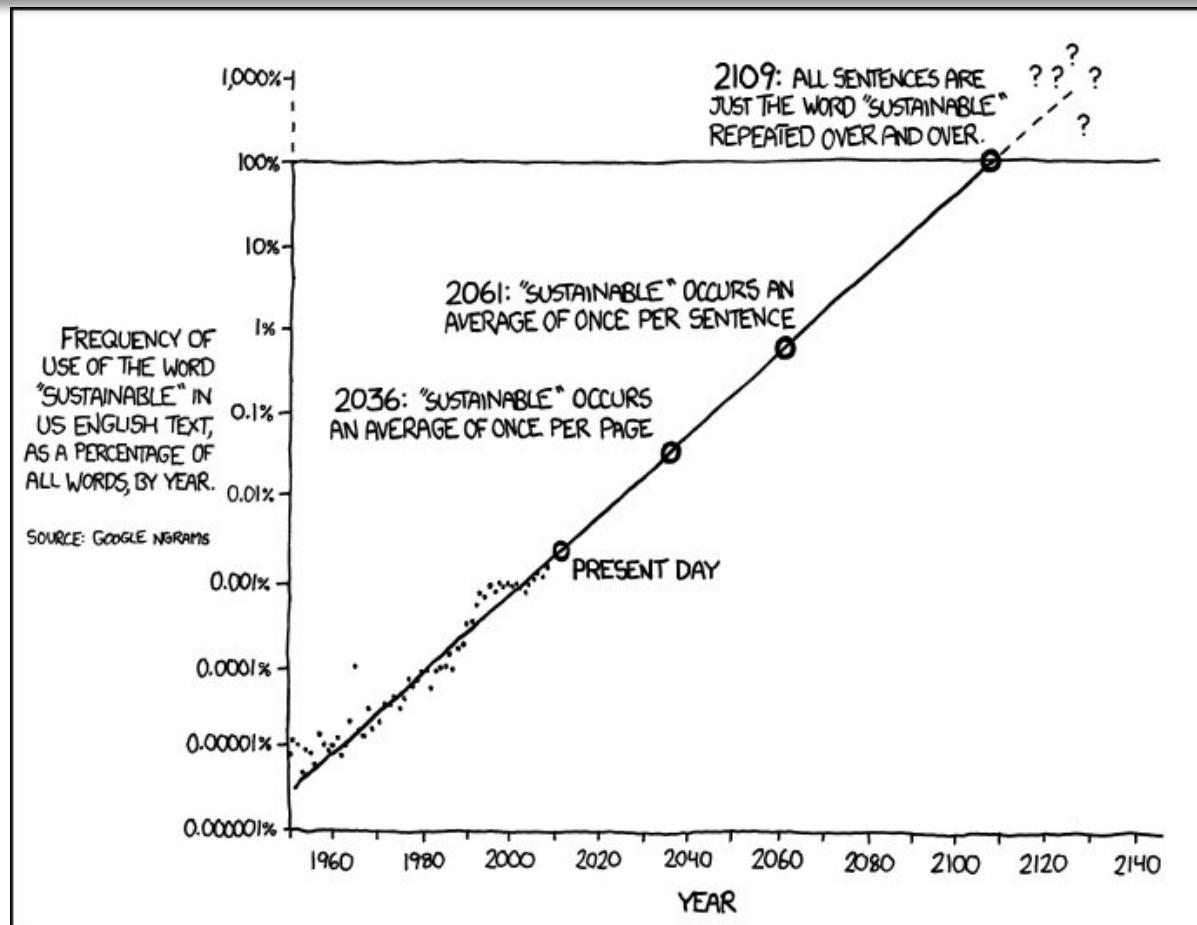
More complex

$$z = ax + by + cx^2 + dy^2 + ex^3 \dots$$

Capture more complex landscapes



The dangers of extrapolation



Source: xkcd