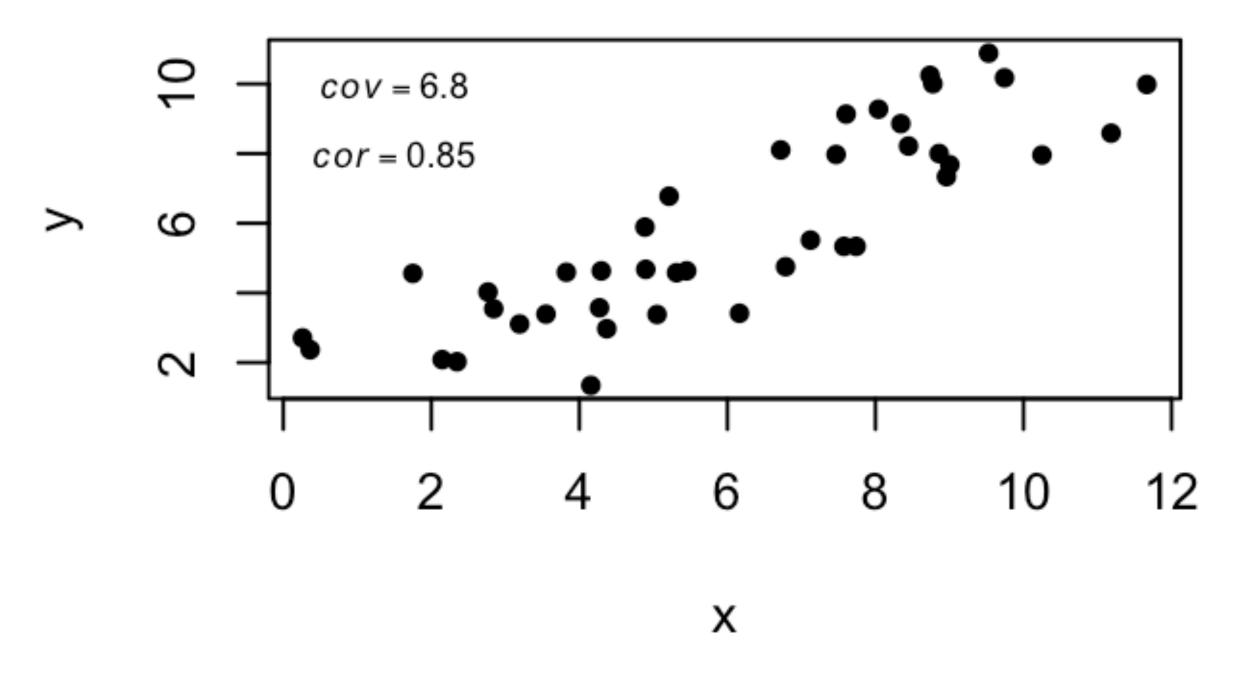
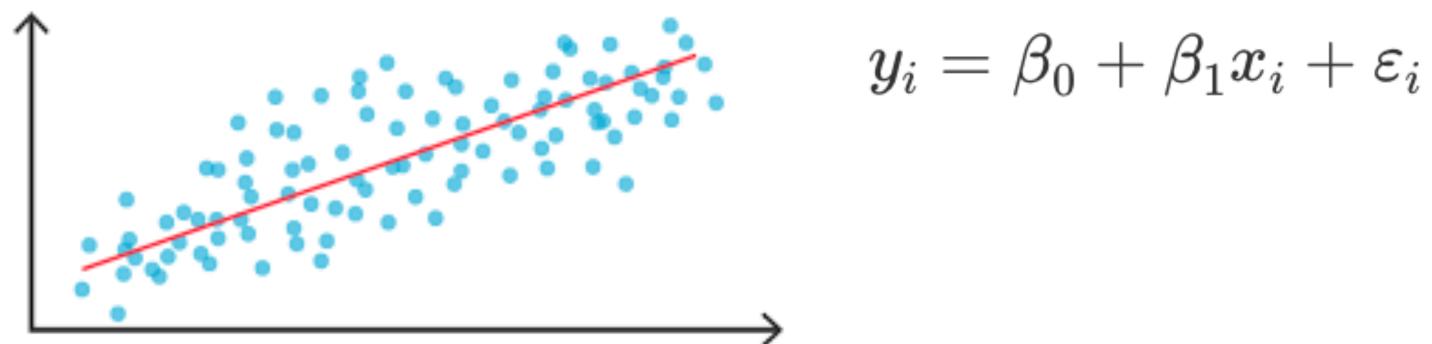
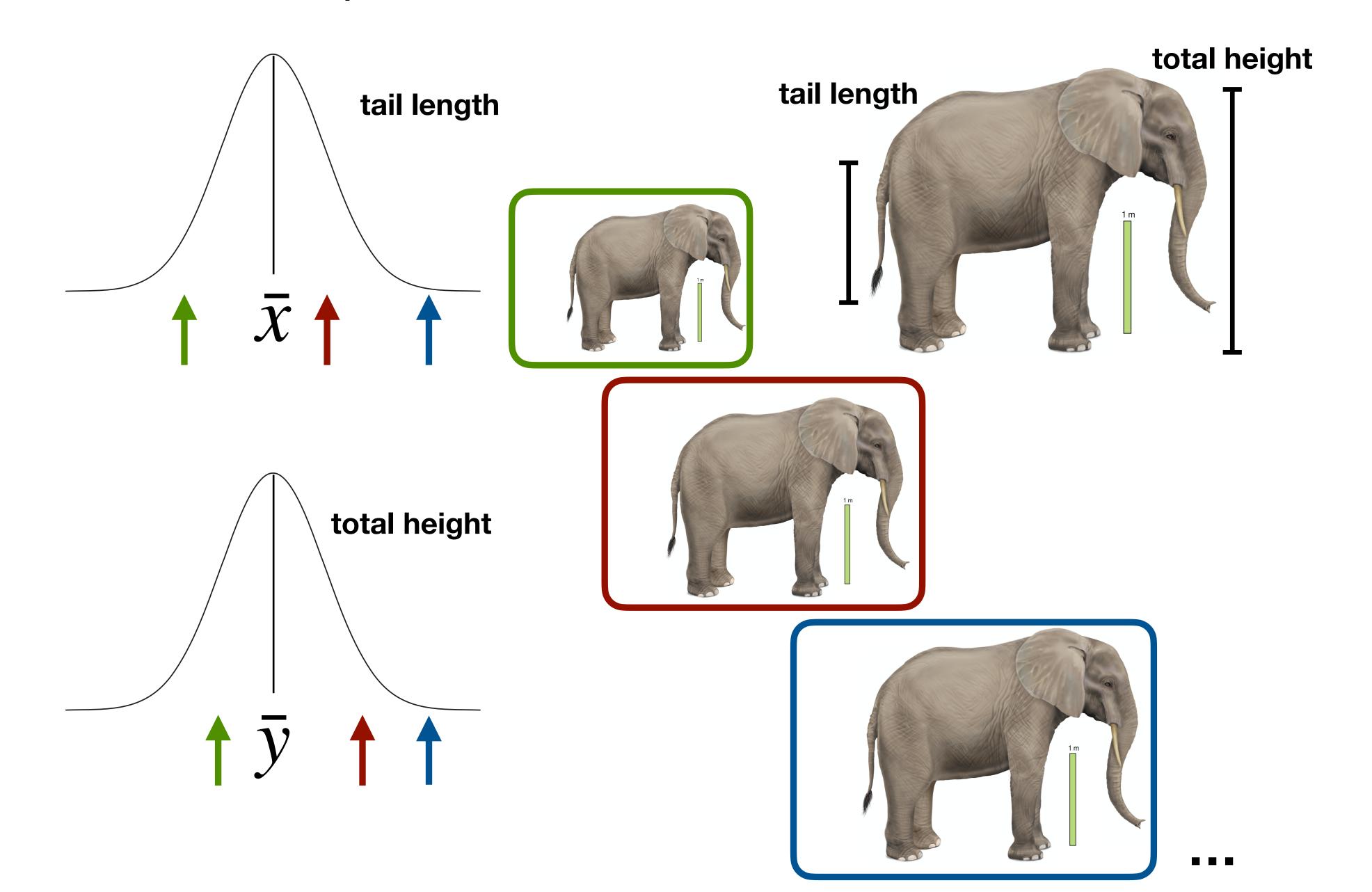
Foundational Statistics Covariance and Correlation





Relationships between numeric variables



Relationships between numeric variables

How do we quantify whether one variable is systematically related to another variable?

The sample covariance:

$$cov(x,y) = s_{xy} = rac{\sum_{i=1}^{n} (x_i - ar{x})(y_i - ar{y})}{n-1}$$

vars deviate from their means in same dir.: <u>product is positive</u>

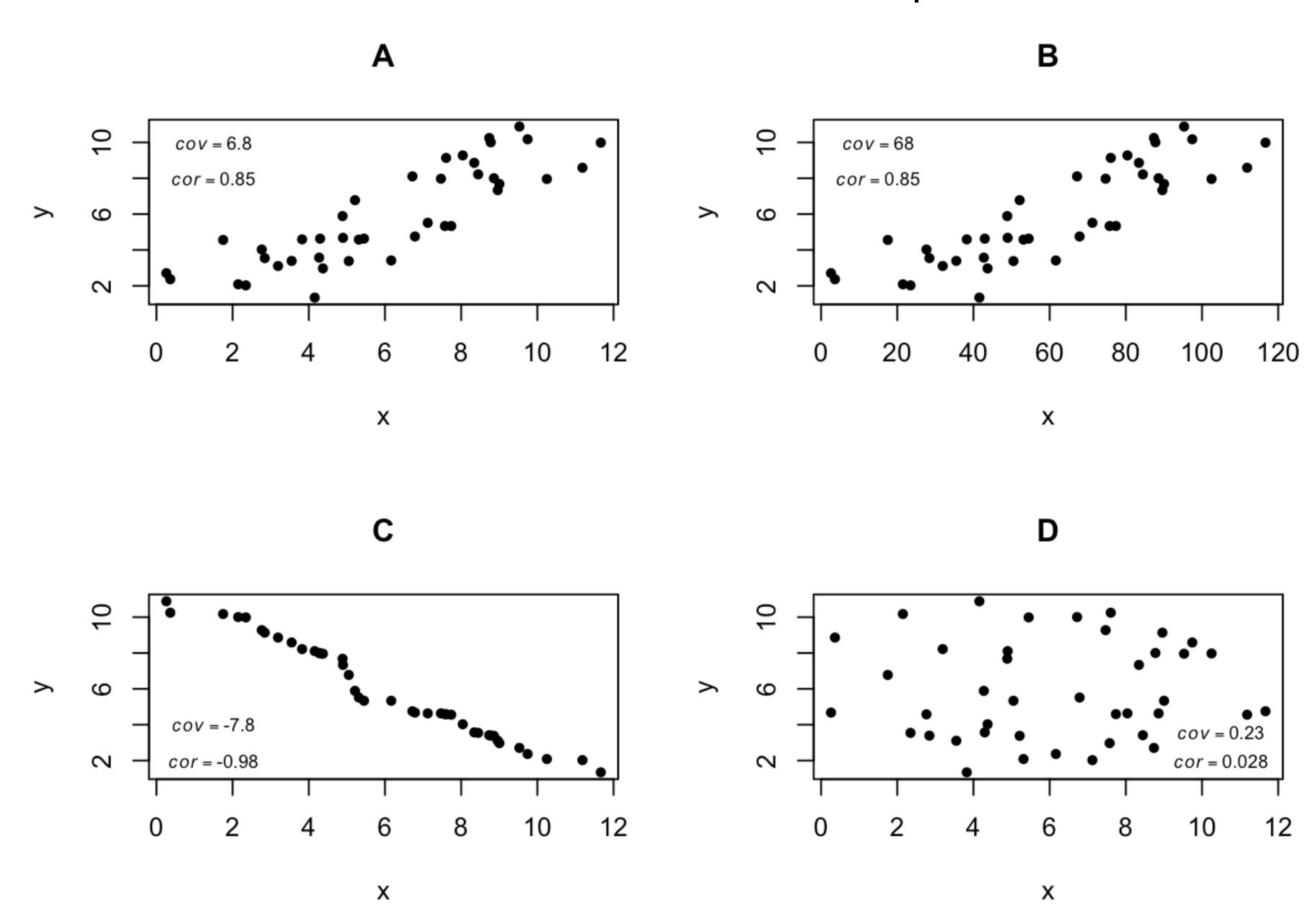
vars deviate from their means in opposite dir.: <u>product is negative</u>

The sample correlation coefficient:

$$r_{xy} = rac{\sum_{i=1}^{n}(x_i - ar{x})(y_i - ar{y})}{\sqrt{\sum_{i=1}^{n}(x_i - ar{x})^2}\sqrt{\sum_{i=1}^{n}(y_i - ar{y})^2}}$$

The denominator scales *r* so that it ranges from -1 to +1

Covariance and correlation: examples



Hypothesis tests for correlation

$$H_0: \rho_1 = 0$$

$$H_A: \rho_1 \neq 0$$

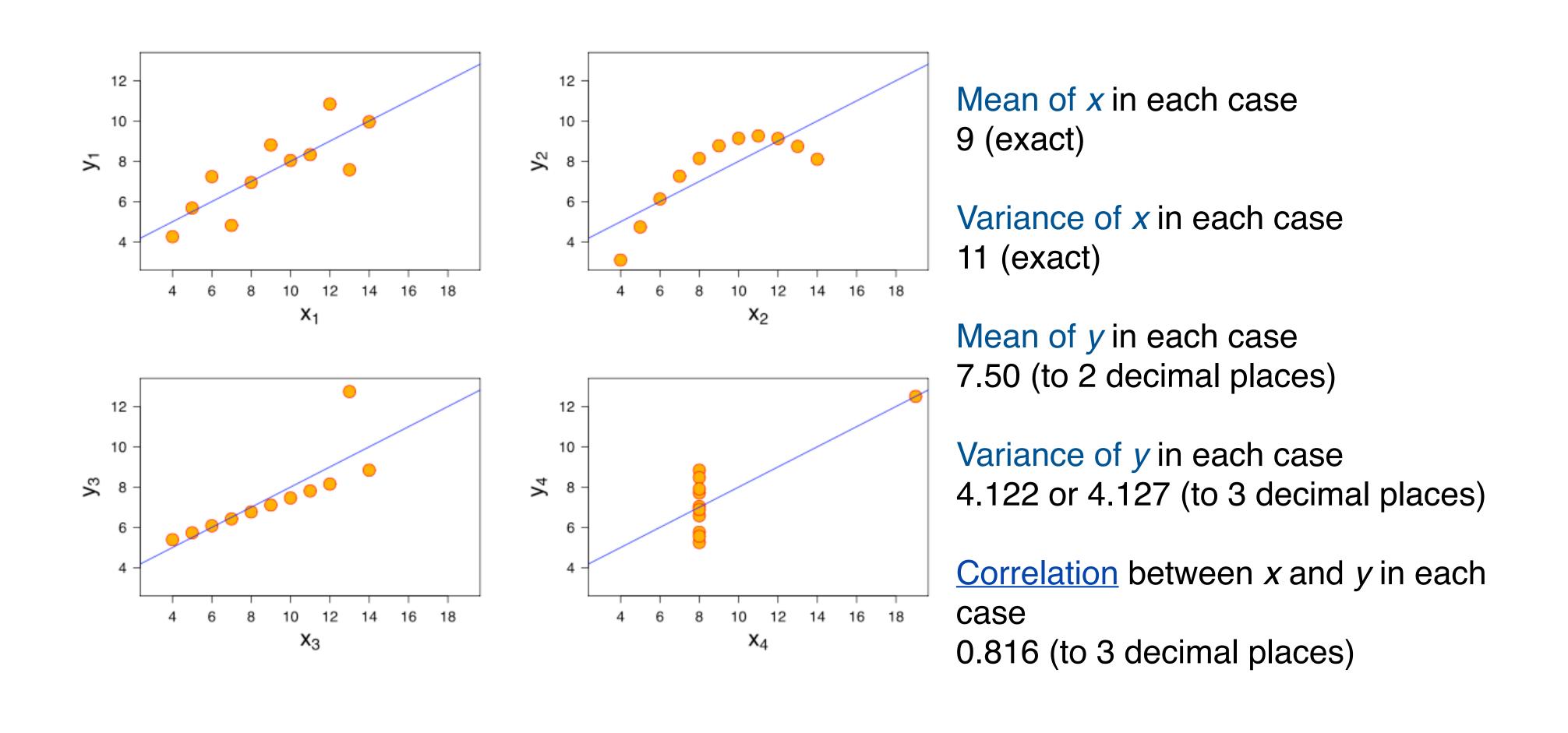
One test statistic for this hypothesis test:

$$t=r\sqrt{rac{n-2}{1-r^2}}$$
 Compare to a t -distribution with n - 2 df cor.test() function in R will carry out

Assumptions of the test:

- 1. Relationship mostly linear (no strong curvilinearity)
- 2. "Bivariate normal": both variables normally dist.

Anscombe's Quartet: Always plot your data to assess assumptions and guide interpretation!



Hypothesis tests for correlation: Nonparametric alternatives

- 1. Spearman's rank: Rank-based, for n < 30
- 2. Kendall's tau: Rank-based, for larger sample sizes
- 3. Randomization or resampling test

Key properties of correlation analysis

- 1. Indicates directionality
- 2. Strength of relationship is scaled by the variances
- 3. Does not say anything about causation
- 4. Does not say anything about the steepness of the relationship (need regression for that!)