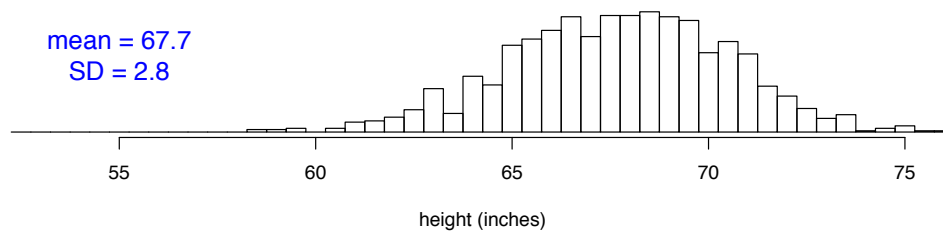


# Correlation

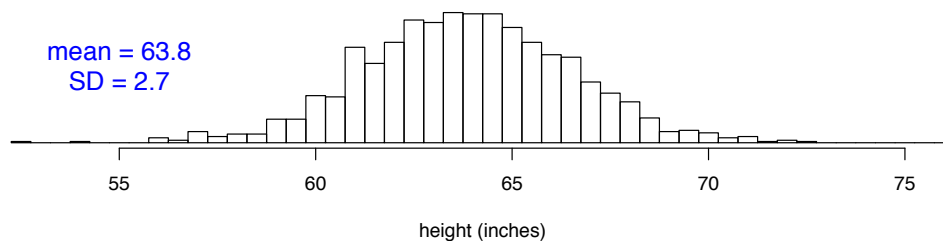
## Fathers' and daughters' heights

---

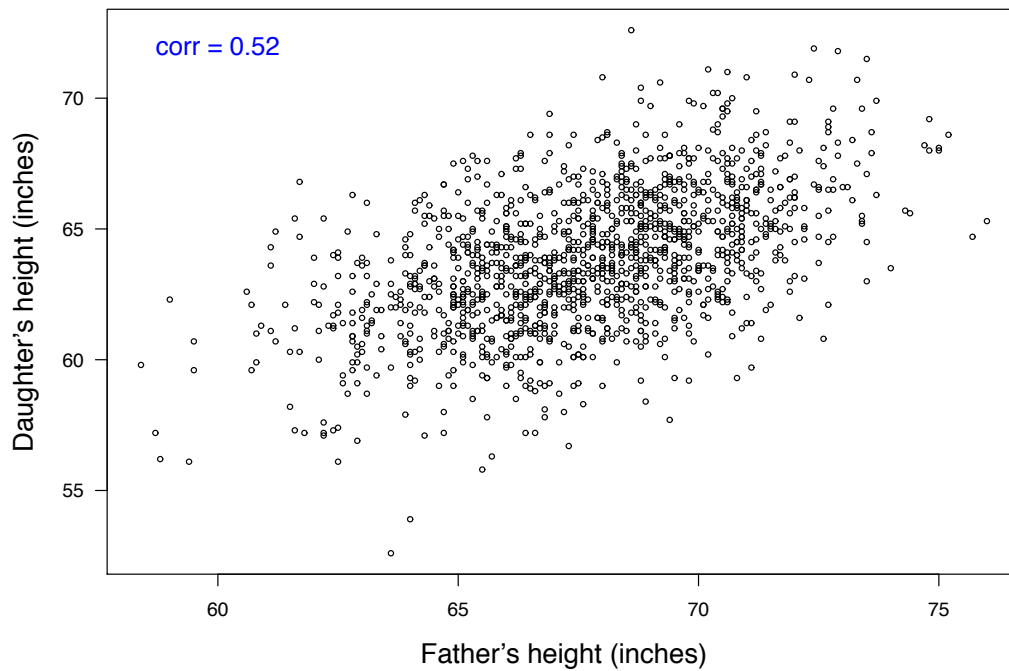
Fathers' heights



Daughters' heights



# Fathers' and daughters' heights



Reference: Pearson and Lee (1906) Biometrika 2:357-462

1376 pairs

## Covariance and correlation

Let  $X$  and  $Y$  be random variables with

$$\mu_X = E(X), \mu_Y = E(Y), \sigma_X = SD(X), \sigma_Y = SD(Y)$$

For example, sample a father/daughter pair and let

$X$  = the father's height and  $Y$  = the daughter's height.

Covariance

$$\text{cov}(X, Y) = E\{(X - \mu_X)(Y - \mu_Y)\}$$

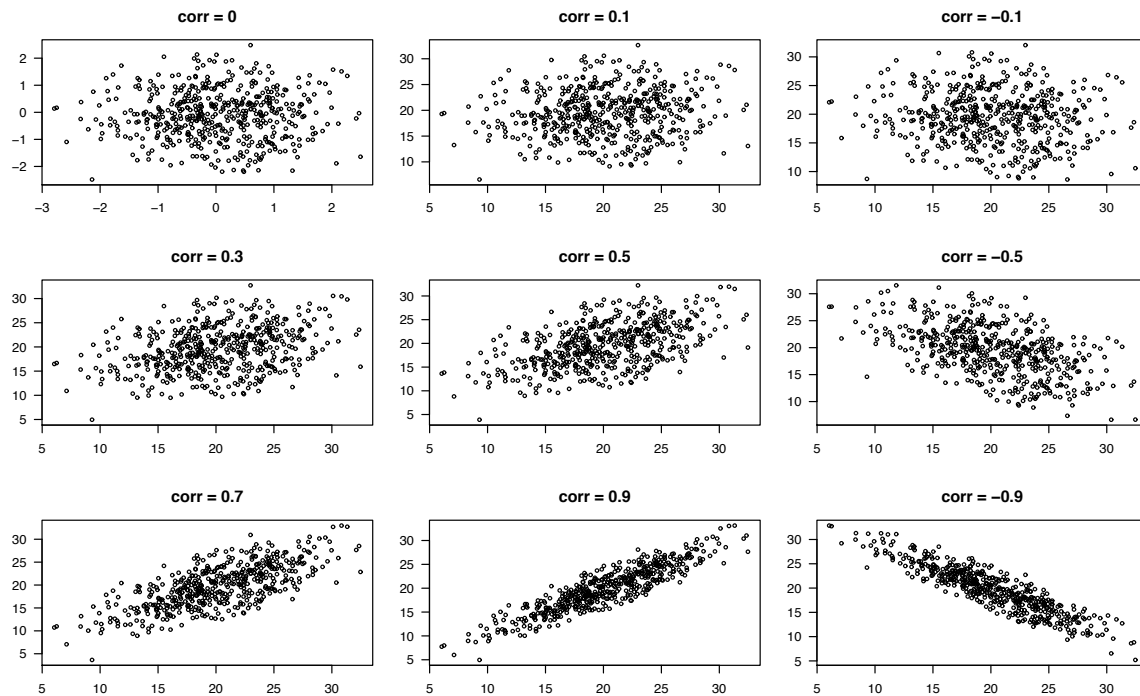
→  $\text{cov}(X, Y)$  can be any real number

Correlation

$$\text{cor}(X, Y) = \frac{\text{cov}(X, Y)}{\sigma_X \sigma_Y}$$

→  $-1 \leq \text{cor}(X, Y) \leq 1$

# Examples



## Estimated correlation

Consider  $n$  pairs of data:  $(x_1, y_1), (x_2, y_2), (x_3, y_3), \dots, (x_n, y_n)$

We consider these as independent draws from some bivariate distribution.

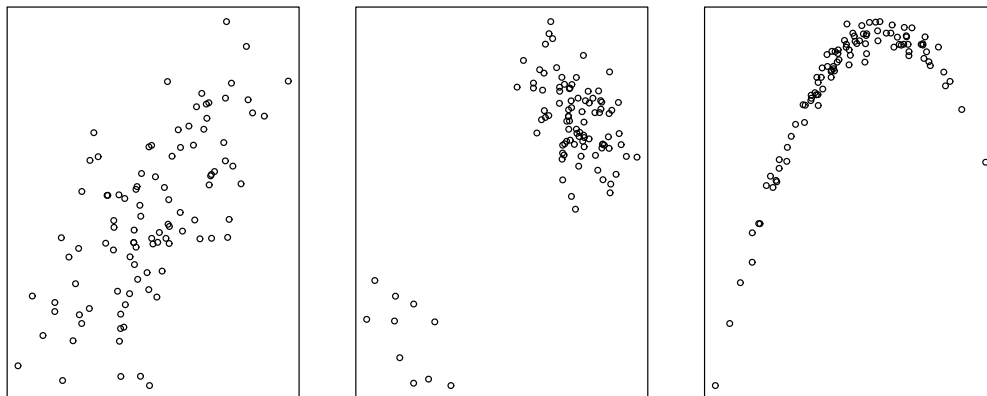
We estimate the correlation in the underlying distribution by:

$$r = \frac{\sum_i (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_i (x_i - \bar{x})^2 \sum_i (y_i - \bar{y})^2}}$$

This is sometimes called the **correlation coefficient**.

## Correlation measures **linear** association

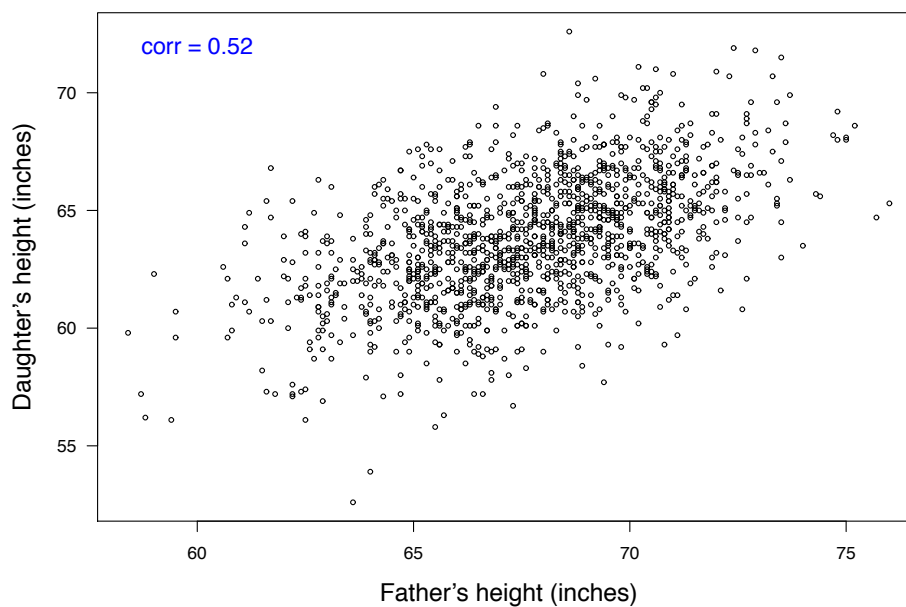
---



→ All three plots have correlation  $\approx 0.7$ !

## Fathers' and daughters' heights

---



# Pearson and Spearman

---

