Lecture 36 Exploring Bivariate Data Using Correlation

BIO210 Biostatistics

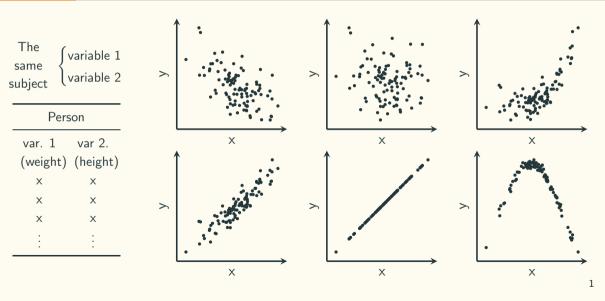
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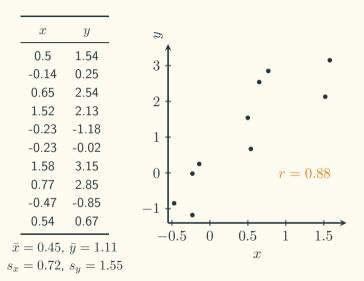
School of Life Sciences
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Scatter Plot



Pearson's Correlation Coefficient (r)



$$r = \frac{1}{n-1} \sum_{i=1}^{n} \left(\frac{x_i - \bar{x}}{s_x} \right) \left(\frac{y_i - \bar{y}}{s_y} \right)$$

$$r = \frac{1}{n-1} \sum_{i=1}^{n-1} Z_{x_i} Z_{y_i}$$

$$r = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\left[\sum_{i=1}^{n} (x_i - \bar{x})^2\right] \left[\sum_{i=1}^{n} (y_i - \bar{y})^2\right]}}$$

 $-1 \le r \le 1$

Covariance

$$\begin{split} \sigma(\boldsymbol{X}, \boldsymbol{Y}) &= E[(\boldsymbol{X} - E[\boldsymbol{X}]) \cdot (\boldsymbol{Y} - E[\boldsymbol{Y}])] \\ &= E[\boldsymbol{X}\boldsymbol{Y} - \boldsymbol{X} \cdot E[\boldsymbol{Y}] - \boldsymbol{Y} \cdot E[\boldsymbol{X}] + E[\boldsymbol{X}] \cdot E[\boldsymbol{Y}]] \\ &= E[\boldsymbol{X}\boldsymbol{Y}] - E[\boldsymbol{X} \cdot E[\boldsymbol{Y}]] - E[\boldsymbol{Y} \cdot E[\boldsymbol{X}]] + E[E[\boldsymbol{X}] \cdot E[\boldsymbol{Y}]] \\ &= E[\boldsymbol{X}\boldsymbol{Y}] - E[\boldsymbol{Y}] \cdot E[\boldsymbol{X}] - E[\boldsymbol{X}] \cdot E[\boldsymbol{Y}] + E[\boldsymbol{X}] \cdot E[\boldsymbol{Y}] \\ &= E[\boldsymbol{X}\boldsymbol{Y}] - E[\boldsymbol{X}] \cdot E[\boldsymbol{Y}] \end{split}$$

$$Cov(x,y) = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{n-1}$$

If
$$\boldsymbol{X}$$
 and \boldsymbol{Y} are independent: $\sigma(\boldsymbol{X},\boldsymbol{Y})=0$

$$r = \frac{\sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\left[\sum_{i=1}^{n} (x_i - \bar{x})^2\right] \left[\sum_{i=1}^{n} (y_i - \bar{y})^2\right]}}$$

$$= \frac{Cov(x,y)}{\sqrt{s_x^2 \cdot s_y^2}} = \frac{Cov(x,y)}{\sqrt{Cov(x,x) \cdot Cov(y,y)}}$$

Variance of The Sum of Two Random Variables

$$var(\mathbf{X} + \mathbf{Y}) = E[(\mathbf{X} + \mathbf{Y})^{2}] - (E[\mathbf{X} + \mathbf{Y}])^{2}$$

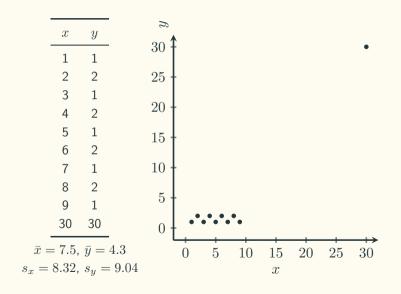
$$= E[\mathbf{X}^{2} + 2\mathbf{X}\mathbf{Y} + \mathbf{Y}^{2}] - (E[\mathbf{X}] + E[\mathbf{Y}])^{2}$$

$$= E[\mathbf{X}^{2}] + 2 \cdot E[\mathbf{X}\mathbf{Y}] + E[\mathbf{Y}^{2}] - (E[\mathbf{X}])^{2} - 2 \cdot E[\mathbf{X}]E[\mathbf{Y}] - (E[\mathbf{Y}])^{2}$$

$$= \left(E[\mathbf{X}^{2}] - (E[\mathbf{X}])^{2}\right) + \left(E[\mathbf{Y}^{2}] - (E[\mathbf{Y}])^{2}\right) + 2\left(E[\mathbf{X}\mathbf{Y}] - E[\mathbf{X}]E[\mathbf{Y}]\right)$$

$$= var(\mathbf{X}) + var(\mathbf{Y}) + 2 \cdot \sigma(\mathbf{X}, \mathbf{Y})$$

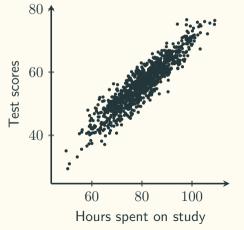
Pearson's Correlation Coefficient (r)



r = 0.95Be careful about outliers!

Hypothesis testing of Pearson's \boldsymbol{r}

We suspect that there is a linear relationship between the number of hours spent on study and the test scores. To find out if this is the case, we can draw a random sample and conduct a hypothesis testing.



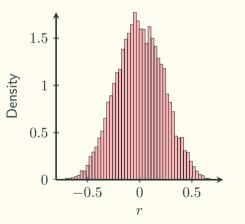
Population correlation coefficient: ρ Sample correlation coefficient: r

$$\begin{cases} H_0: \text{no linear relationship} \\ H_1: \text{some linear relationship} \end{cases} \Leftrightarrow \begin{cases} H_0: \rho = 0 \\ H_1: \rho \neq 0 \end{cases}$$

What is the sampling distribution of r ?

Sampling Distribution of Pearson's \boldsymbol{r}

10,000 simulations under H_0 is true



Under H_0 (no linear relationship) is true:

$$r\sqrt{\frac{n-2}{1-r^2}} = \frac{r}{\sqrt{(1-r^2)/(n-2)}} \sim t_{n-2}$$

Hypothesis testing of Pearson's r

To investigate whether there is a linear relationship between the number of hours spent on study and the test scores, 20 students were randomly selected, and Pearson's r was calculated to be r=0.69.

Test statistic:
$$t = r\sqrt{\frac{n-2}{1-r^2}} = 0.69 \times \sqrt{\frac{20-2}{1-0.69^2}} = 4.04$$

Two-tailed *p*-value:
$$P(|t| \ge 4.04) = 2 \times P(t \ge 4.04) = 0.000768$$