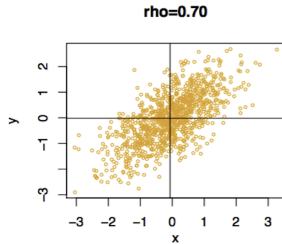


Covariance and correlation



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 - * The converse is **not** true! If covariance $\text{Cov}(X, Y)$ is zero, random variables X and Y might not be independent.

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HW Confirm this result in R.

For computing the covariance use the function `cov`.

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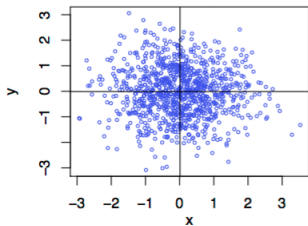
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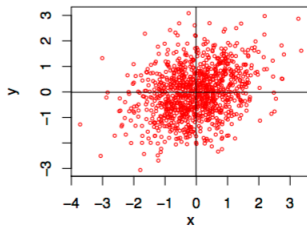
$$\begin{array}{lll} \rho(X, Y) = 1 & \iff & Y = aX + b \text{ with } a > 0 \\ \rho(X, Y) = -1 & \iff & Y = aX + b \text{ with } a < 0 \end{array}$$

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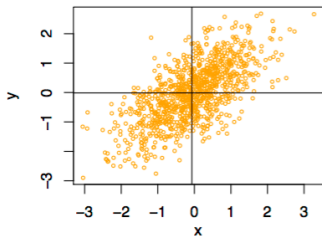
$\rho=0.00$



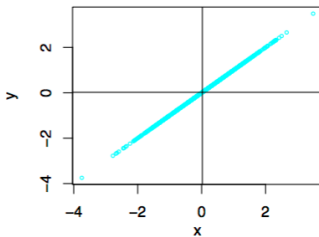
$\rho=0.30$



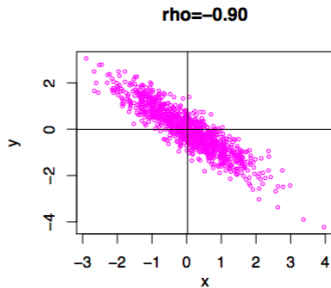
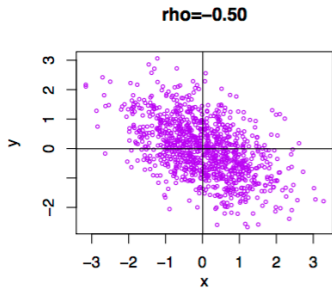
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Examples

HW Let X be a random variable that takes values $-2, -1, 0, 1, 2$, each with probability $\frac{1}{5}$. Let $Y = X^2$. Show that $\text{Cov}(X, Y) = 0$, but X and Y are not independent.

HW Let X_1, \dots, X_{10} be an independent and identically distributed random variables with $\mathcal{N}(3, 12)$. Let

$$Y_1 = \frac{1}{6} \sum_{i=1}^6 X_i \quad \text{and} \quad Y_2 = \frac{1}{4} \sum_{i=7}^{10} X_i$$

- (i) What is the distribution of $Y_1 + Y_2$?
- (ii) Compute $\text{Cov}(2Y_1 - 5, Y_2 + 4)$ and $\rho(Y_1, Y_2)$.

Examples

HW Let $X \sim \text{bern}(p)$.

Let Y_1 be the indicator that $X = 1$ and Y_0 be the indicator that $X = 0$.

Compute $\text{Cov}(Y_0, Y_1) = 0$.

HW Let $Z \sim \mathcal{N}(0, 1)$.

Compute $\text{Cov}(Z^2, Z^3)$.

HW Let $T \sim \exp(\lambda)$ and $S = T + 10$.

(i) Does S have an exponential distribution?

Use the pdf of S to answer this question.

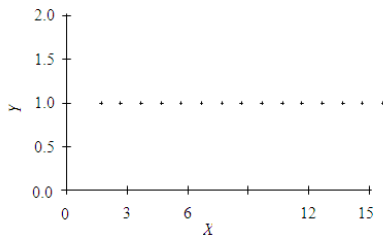
(ii) Compute $\rho(T, S)$.

HW Consider $G \sim U(-3, -1)$, $V \sim U(-1, 1)$ and $W \sim U(1, 3)$ and assume that all are independent. Compute $\text{Cov}(W - V, V - G)$.

Questions

Multiple-choice questions

- (1) A scatterplot of a variable Y versus a variable X produced the scatterplot below. The value of Y for all values of X is exactly 1.0. The correlation coefficient between Y and X is:



- a. 1
- b. -1
- c. 0
- d. either 1 or -1

Multiple-choice questions

- (2) Let X and Z be independent random variables and both $N(0, 1)$ -distributed. Let

$$Y = \frac{1}{2} (X + Z).$$

Then the correlation between X and Y is

- a. smaller than the correlation between X and Z
- b. equal to the correlation between X and Z
- c. larger than the correlation between X and Z
- d. not defined

Next week we start with [Statistics](#)

Thank you for your attention!