

Quiz

Find a match:

- | | |
|-------------------------------------|---------------------|
| 1. Cumulative distribution function | A. Waiting time |
| 2. Quantile function | B. Separability |
| 3. Pooled variance | C. Approximation |
| 4. Exponential distribution | D. Antiderivative |
| 5. Degrees of freedom | E. Weighted average |
| 6. Central Limit Theorem | F. Goodness-of-fit |
| 7. Independence | G. Constraints |
| 8. Correlation | H. Inverse |

Solution:

1. D (C.d.f. is an antiderivative of p.d.f.)
2. H (Quantile function is an inverse of c.d.f.)
3. E (Pooled variance is a weighted average of several sample variances with weights being degrees of freedom)
4. A (Exponential distribution shows a waiting time in a Poisson process of the next event)
5. G (Degrees of freedom decrease by a number of constraints in a sample)
6. C (The CLT allows approximation of distributions in a large sample with the normal one)
7. B (Independence of random variables is defined as a separability of joint distribution into marginal ones)
8. F (Correlation is a goodness-of-fit metric for a linear regression problem)

Problem 1

A random sample of 400 married couples was selected from a large population of married couples.

- Heights of married men are approximately normally distributed with mean 70 inches and standard deviation 3 inches.
 - Heights of married women are approximately normally distributed with mean 65 inches and standard deviation 2.5 inches.
 - There were 20 couples in which wife was taller than her husband, and there were 380 couples in which wife was shorter than her husband.
1. Find a 95% confidence interval for the proportion of married couples in the population for which the wife is taller than her husband.
 2. Suppose that a married man is selected at random and a married woman is selected at random. Find the approximate probability that the woman will be taller than the man.
 3. Based on your answers to 1 and 2, are the heights of wives and their husbands independent? Explain your reasoning.

Solution:

TODO

Problem 2

Suppose 2000 points are selected independently at a random from the unit square

$S = \{(x, y) : 0 \leq x \leq 1, 0 \leq y \leq 1\}$. Let W be the number of points that fall into the set $A = \{(x, y) : x^2 + y^2 < 1\}$.

1. How is W distributed?
2. Find the mean, variance and standard deviation of W .
3. Estimate probability that W is greater than 1600.

Solution:

TODO

Problem 3

Distribution of X is uniform $\mathcal{U}(-a, a)$. Sample of size $n = 2$ is available.

Consider $\hat{a} = c \cdot (|X_1| + |X_2|)$ as a class of estimators for the parameter a .

Find c such that

1. Estimator \hat{a} is unbiased.
2. Estimator \hat{a} is the most efficient in the class. (In terms of mean square error.)

Solution:

TODO

Problem 4

Consider random variables X and Y with joint density function

$$f(x, y) = \begin{cases} \frac{1}{2} + cx, & x + y \leq 1, \ x \geq 0, \ y \geq 0, \\ 0, & \text{otherwise.} \end{cases}$$

1. Find c .
2. Find $f_X(x)$. Evaluate $E(X)$.
3. Write down an expression for $f_{Y|X}(x, y)$. Find $E(Y \mid X = x)$.

Solution:

TODO

Problem 5

Internal angles $\theta_1, \theta_2, \theta_3, \theta_4$ of a certain quadrilateral, located on the ground, were measured by the aerial system. It is assumed that those observations x_1, x_2, x_3, x_4 were taken with minor and independent errors, which have zero mean and identical variance σ^2 .

1. Find the LSE of $\theta_1, \theta_2, \theta_3, \theta_4$.
2. Find an unbiased estimate of σ^2 in the case, described in part 1.
3. Let's assume now that the considered quadrilateral is a parallelogram with $\theta_1 = \theta_3$ and $\theta_2 = \theta_4$. How values of internal angles LSE would change? Find an unbiased estimate of σ^2 in this particular case.

Solution:

TODO

Problem 6

Suppose that student's grade for a statistics exam, X , has continuous uniform distribution at the interval $[0, 100]$. But less than 25 points means "failed", and more than 80 points is "excellent", hence the final grade Y is calculated as follows:

$$Y = \begin{cases} 0, & X < 25 \\ X, & 25 \leq X < 80 \\ 100, & X \geq 80 \end{cases}$$

1. Find c.d.f. of Y . Sketch the plot.
2. Find p.d.f. of Y . Sketch the plot.
3. Find mean and variance of X and Y .
4. Find $E(Y \mid Y > 0)$.
5. Find $\text{Corr}(X, Y)$.

Solution:

1. A random variable X has following p.d.f. and c.d.f.:

$$f_X(x) = \frac{1}{100} \cdot I_{\{0 \leq x \leq 100\}},$$

$$F_X(x) = \begin{cases} 1, & x \geq 100, \\ \frac{x}{100}, & 0 \leq x < 100, \\ 0, & x < 0. \end{cases}$$

The c.d.f. of Y coincides with that of X for $y \in [25, 80)$:

$$F_Y(y) = \frac{y}{100}, \text{ for } 25 \leq y < 80.$$

$F_Y(y)$ has a sharp increase in $y = 0$ by a value of $P(X < 25) = \frac{1}{4}$, and is maintained on the same level up to $y = 25$:

$$F_Y(y) = \frac{1}{4}, \text{ for } 0 \leq y < 25.$$

Also, $F_Y(y)$ does not change to the right of $y = 80$ until it hits $y = 100$, where it has a sharp increase by a value of $P(X \geq 80) = \frac{1}{5}$:

$$F_Y(y) = \frac{4}{5}, \text{ for } 80 \leq y < 100.$$

Thus, the c.d.f. of Y has a following view:

$$F_Y(y) = \begin{cases} 1, & y \geq 100, \\ \frac{4}{5}, & 80 \leq y < 100, \\ \frac{y}{100}, & 25 \leq y < 80, \\ \frac{1}{4}, & 0 \leq y < 25, \\ 0, & y < 0. \end{cases}$$

A graph of $F_Y(y)$ is shown in the fig. 1.

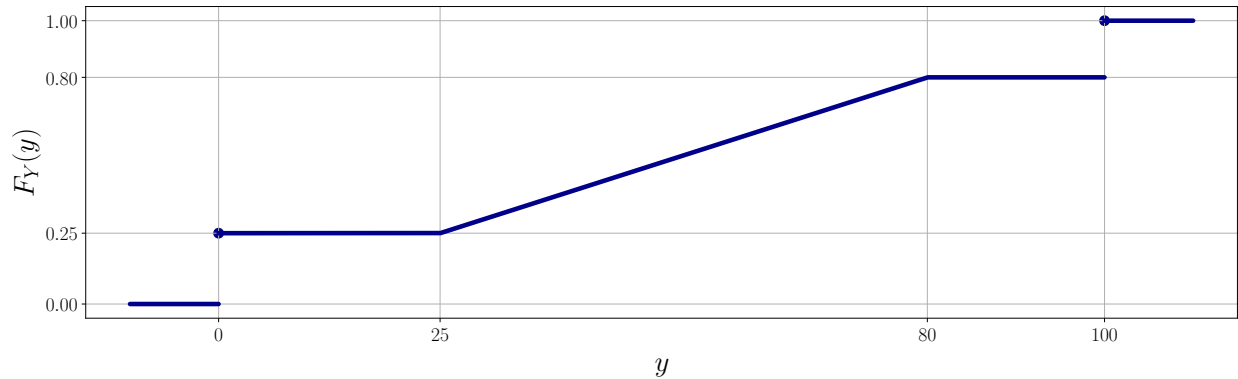


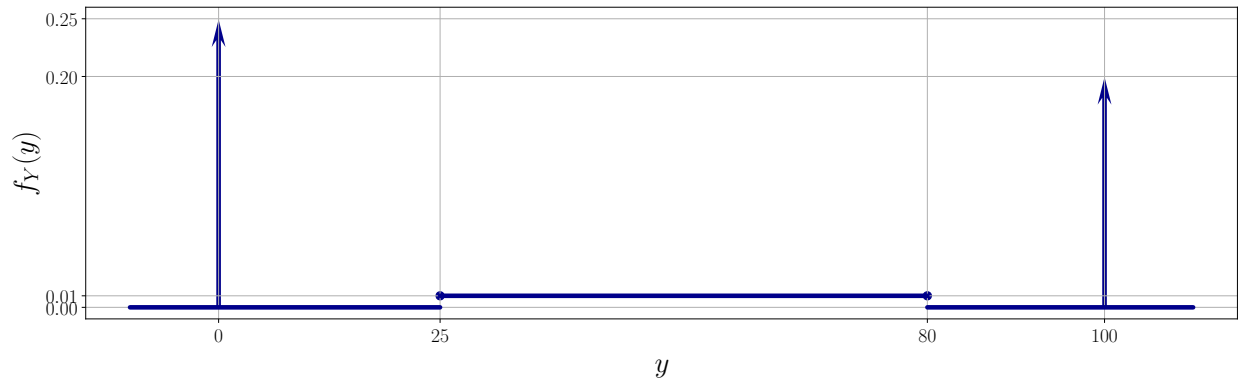
Figure 1: C.d.f. of the random variable Y .

2. Since $F_Y(y)$ has points of discontinuity $y = 0$ and $y = 100$, it only has generalized p.d.f. Taking derivative of c.d.f.:

$$f_Y(y) = \frac{dF_Y(y)}{dy} = \left[\frac{1}{100} \cdot I_{\{25 \leq y \leq 80\}} + \frac{1}{4} \delta(y) + \frac{1}{5} \delta(y - 100) \right],$$

where $\delta(y)$ is a Dirac delta function, defined as:

$$\delta(y) \simeq \begin{cases} +\infty, & y = 0, \\ 0, & y \neq 0, \end{cases} \quad \text{constrained by} \quad \int_{-\infty}^{+\infty} \delta(y) dy = 1.$$

Figure 2: Generalized p.d.f. of the random variable Y .

A graph of $f_Y(y)$ is shown in the fig. 2.

The $f_Y(y)$ possesses the most important property of probability densities – normalization by 1:

$$\begin{aligned} \int_{-\infty}^{+\infty} f_Y(y) dy &= \int_{25}^{80} \frac{1}{100} dy + \int_{-\infty}^{+\infty} \frac{1}{4} \delta(y) dy + \int_{-\infty}^{+\infty} \frac{1}{5} \delta(y - 100) dy = \\ &= \frac{y}{100} \Big|_{25}^{80} + \frac{1}{4} \int_{-\infty}^{+\infty} \delta(y) dy + \frac{1}{5} \underbrace{\int_{-\infty}^{+\infty} \delta(\eta) d\eta}_{\eta=y-100} = \frac{55}{100} + \frac{1}{4} + \frac{1}{5} = 1. \end{aligned}$$

3. The mean and variance of a random variable $X \sim \mathcal{U}(a, b)$:

$$\mathbb{E}(X) = \frac{a+b}{2}, \quad \mathbb{V}(X) = \frac{(b-a)^2}{12}.$$

In our case:

$$\mathbb{E}(X) = \frac{0+100}{2} = \boxed{50}, \quad \mathbb{V}(X) = \frac{(b-a)^2}{12} = \frac{(100-0)^2}{12} = \boxed{\frac{2500}{3} \approx 833.333}.$$

There are 2 ways to calculate mean and variance of Y .

(I) Direct calculation

By definition:

$$\mathbb{E}(Y) = \int_{-\infty}^{+\infty} y f_Y(y) dy = \int_{25}^{80} \frac{y}{100} dy + \frac{1}{4} \int_{-\infty}^{+\infty} y \delta(y) dy + \frac{1}{5} \int_{-\infty}^{+\infty} y \delta(y - 100) dy.$$

Using a sifting property of delta-function:

$$\int_{-\infty}^{+\infty} \varphi(y) \delta(y - b) dy = \varphi(b),$$

we can proceed with calculation expected value of Y :

$$E(Y) = \frac{y^2}{200} \Big|_{25}^{80} + \frac{1}{4} y \Big|_{y=0} + \frac{1}{5} y \Big|_{y=100} = \frac{231}{8} + 0 + 20 = \boxed{\frac{391}{8} = 48.875}.$$

In order to calculate variance $V(Y)$, we need to find $E(Y^2)$:

$$\begin{aligned} E(Y^2) &= \int_{-\infty}^{+\infty} y^2 f_Y(y) dy = \int_{25}^{80} \frac{y^2}{100} dy + \frac{1}{4} \int_{-\infty}^{+\infty} y^2 \delta(y) dy + \frac{1}{5} \int_{-\infty}^{+\infty} y^2 \delta(y - 100) dy = \\ &= \frac{y^3}{300} \Big|_{25}^{80} + \frac{1}{4} y^2 \Big|_{y=0} + \frac{1}{5} y^2 \Big|_{y=100} = \frac{19855}{12} + 0 + 2000 = \\ &= \frac{43855}{12} \approx 3654.583. \end{aligned}$$

Variance then:

$$V(Y) = E(Y^2) - E(Y)^2 = \frac{43855}{12} - \left(\frac{391}{8}\right)^2 = \boxed{\frac{243037}{192} \approx 1265.818}.$$

(II) Total expectation

A random variable Y consists of 3 parts exhaustive, which is reflected in the total expectation equation:

$$\begin{aligned} E(Y) &= E(Y | X < 25) \cdot P(X < 25) + E(Y | 25 \leq X < 80) \cdot P(25 \leq X < 80) + \\ &+ E(Y | X \geq 80) \cdot P(X \geq 80) = 0 \cdot \frac{1}{4} + \frac{25+80}{2} \cdot \frac{55}{100} + 100 \cdot \frac{1}{5} = \\ &= 0 + \frac{231}{8} + 20 = \boxed{\frac{391}{8} = 48.875}. \end{aligned}$$

The same for $E(Y^2)$:

$$\begin{aligned} E(Y^2) &= E(Y^2 | X < 25) \cdot P(X < 25) + \\ &+ E(Y^2 | 25 \leq X < 80) \cdot P(25 \leq X < 80) + E(Y^2 | X \geq 80) \cdot P(X \geq 80). \end{aligned}$$

While the first and the third terms are clear, the second one in the equation above requires more calculations:

$$\begin{aligned} E(Y^2 | 25 \leq X < 80) &= E(X^2 | 25 \leq X < 80) = \\ &= V(X | 25 \leq X < 80) + E(X | 25 \leq X < 80)^2 = \\ &= \frac{(80 - 25)^2}{12} + \left(\frac{25 + 80}{2}\right)^2 = \frac{9025}{3} \approx 3008.333. \end{aligned}$$

$E(Y^2)$ then:

$$E(Y^2) = 0 \cdot \frac{1}{4} + \frac{9025}{3} \cdot \frac{55}{100} + 100^2 \cdot \frac{1}{5} = \frac{43855}{12} \approx 3654.583.$$

Variance then:

$$V(Y) = E(Y^2) - E(Y)^2 = \frac{43855}{12} - \left(\frac{391}{8}\right)^2 = \boxed{\frac{243037}{192} \approx 1265.818}.$$

4. Total expectation:

$$E(Y) = E(Y | Y > 0) \cdot P(Y > 0) + E(Y | Y \leq 0) \cdot P(Y \leq 0).$$

Since $E(Y | Y \leq 0) = 0$:

$$E(Y | Y > 0) = \frac{E(Y)}{P(Y > 0)} = \frac{391}{8} : \left(1 - \frac{1}{4}\right) = \boxed{\frac{391}{6} \approx 65.167}.$$

5. Using the definition of correlation coefficient:

$$\text{Corr}(X, Y) = \frac{\text{Cov}(X, Y)}{\sqrt{V(X) \cdot V(Y)}} = \frac{E(XY) - E(X) \cdot E(Y)}{\sqrt{V(X) \cdot V(Y)}}.$$

The only term we don't know is $E(XY)$. Since the joint distribution is unknown, let's use total expectation:

$$\begin{aligned} E(XY) &= E(XY | X < 25) \cdot P(X < 25) + E(XY | 25 \leq X < 80) \cdot P(25 \leq X < 80) + \\ &+ E(XY | X \geq 80) \cdot P(X \geq 80) = E(0 | X < 25) \cdot P(X < 25) + \\ &+ E(X^2 | 25 \leq X < 80) \cdot P(25 \leq X < 80) + E(100X | X \geq 80) \cdot P(X \geq 80). \end{aligned}$$

From paragraph 3II we know that

$$E(X^2 | 25 \leq X < 80) = \frac{9025}{3} \approx 3008.333,$$

and by linearity of expected value:

$$E(100X \mid X \geq 80) = 100E(X \mid X \geq 80) = 100 \cdot \frac{80 + 100}{2} = 100 \cdot 90 = 9000.$$

Overall

$$E(XY) = 0 \cdot \frac{1}{4} + \frac{9025}{3} \cdot \frac{55}{100} + 9000 \cdot \frac{1}{5} = \frac{41455}{12} \approx 3454.583.$$

Correlation coefficient then:

$$\text{Corr}(X, Y) = \frac{\frac{41455}{12} - 50 \cdot \frac{391}{8}}{\sqrt{\frac{2500}{3} \cdot \frac{243037}{192}}} \approx \boxed{0.984}.$$

Problem 7

Let X and Y be two independent standard normal random variables. Find

1. $P(|X + Y| > |X - Y|)$.
2. $P(|X + Y| > 2|X - Y|)$.

Solution:

1. Using independence:

$$\begin{aligned}
 P(|X + Y| > |X - Y|) &= P((X + Y)^2 > (X - Y)^2) = \\
 &= P(X^2 + 2XY + Y^2 > X^2 - 2XY + Y^2) = P(4XY > 0) = \\
 &= P(X > 0) \cdot P(Y > 0) + P(X < 0) \cdot P(Y < 0) = \\
 &= \frac{1}{2} \cdot \frac{1}{2} + \frac{1}{2} \cdot \frac{1}{2} = \boxed{\frac{1}{2}}.
 \end{aligned}$$

2. There are 2 ways to calculate $P(|X + Y| > 2|X - Y|)$.

(I) Strict calculation

$$\begin{aligned}
 P(|X + Y| > 2|X - Y|) &= P((X + Y)^2 > 4(X - Y)^2) = \\
 &= P((X + Y + 2(X - Y)) \cdot (X + Y - 2(X - Y)) > 0) = \\
 &= P((3X - Y) \cdot (3Y - X) > 0) = \\
 &= P(3X - Y > 0 \cap 3Y - X > 0) + \\
 &\quad + P(3X - Y < 0 \cap 3Y - X < 0).
 \end{aligned}$$

Random variables $3X - Y$ and $3Y - X$ are not independent, so the probabilities of intersections can not be separated. Let's consider a region $(3X - Y) \cdot (3Y - X) > 0$. Its probability is a volume under a joint p.d.f. inside this region. This is illustrated in the fig. 3 and in the fig. 4.

Since X and Y are independent components of the vector $(X \ Y)^T \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$, their joint p.d.f. is radially symmetrical. The volume of a considered region is determined by the angle those lines cover.

The angle of one sector is $\arctan 3 - \arctan \frac{1}{3}$, so the required probability:

$$P(|X + Y| > 2|X - Y|) = \frac{\arctan 3 - \arctan \frac{1}{3}}{\pi} = \boxed{\frac{\arctan \frac{4}{3}}{\pi} \approx 0.295}.$$

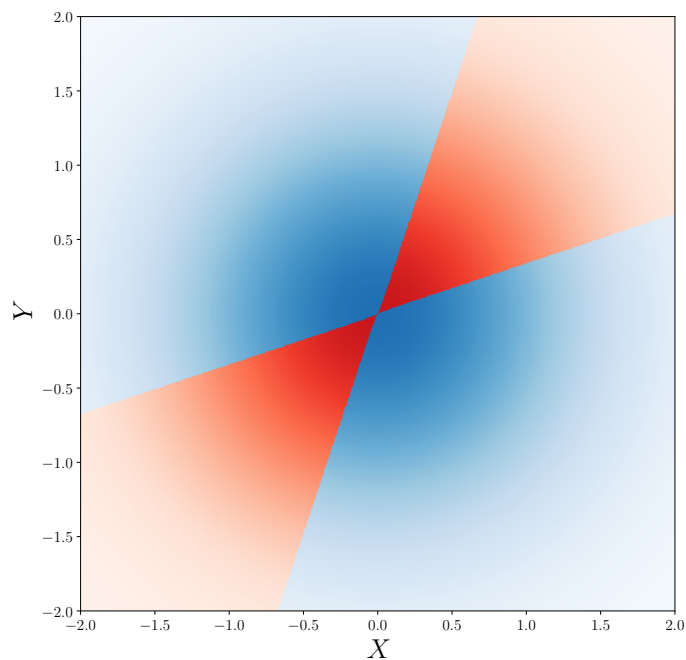


Figure 3: Probability density function of $(X \ Y)^\top \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$ (top view).

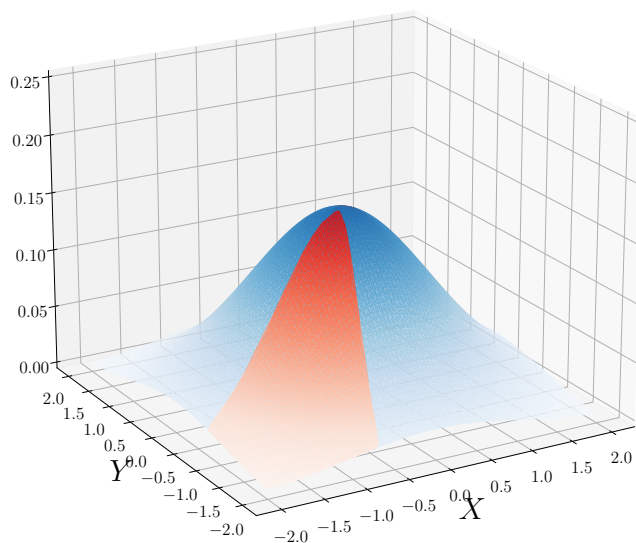


Figure 4: Probability density function of $(X \ Y)^\top \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$ (side view).

(II) Approximation

Let's consider variables $X + Y$ and $X - Y$. They have identical distribution:

$$X \pm Y \sim \mathcal{N}(0, 1^2 + (\pm 1)^2) = \mathcal{N}(0, 2).$$

Inherently, they are uncorrelated:

$$\begin{aligned}\text{Cov}(X + Y, X - Y) &= \text{Cov}(X, X) - \text{Cov}(X, Y) + \text{Cov}(Y, X) - \text{Cov}(Y, Y) = \\ &= \text{V}(X) - \text{V}(Y) = 2 - 2 = 0.\end{aligned}$$

Since X and Y are components of the bivariate normal vector $(X \ Y)^\top \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$, $X + Y$ and $X - Y$ are also components of a vector with bivariate normal distribution. In such case uncorrelatedness means independence – this is the unique property of multivariate normal distributions.

Thus, $X + Y$ and $X - Y$ are independent. The squares of standardized variables are also independent, and moreover, are χ^2 -distributed:

$$Q_1 = \left(\frac{X + Y}{\sqrt{2}} \right)^2 \sim \chi_1^2, \quad Q_2 = \left(\frac{X - Y}{\sqrt{2}} \right)^2 \sim \chi_1^2.$$

The initial probability can be rewritten with F -distribution:

$$\begin{aligned}\text{P}(|X + Y| > 2 |X - Y|) &= \text{P}((X + Y)^2 > 4(X - Y)^2) = \text{P}(Q_1 > 4Q_2) = \\ &= \text{P}\left(\frac{Q_1/1}{Q_2/1} > 4\right) = \text{P}(F_{1,1} > 4) \in \boxed{(0.25, 0.5)}.\end{aligned}$$

The exact result was found in 2I.

Problem 8

Two random variables are given: $X \sim \mathcal{N}(0, 9)$ and $Y \sim \mathcal{N}(0, 4)$. $\text{Corr}(X, Y) = -1$.
Evaluate $P(2X + Y > 3)$.

Solution:

Fig. 5 shows how a bivariate distribution changes with the increase of correlation coefficient between components.

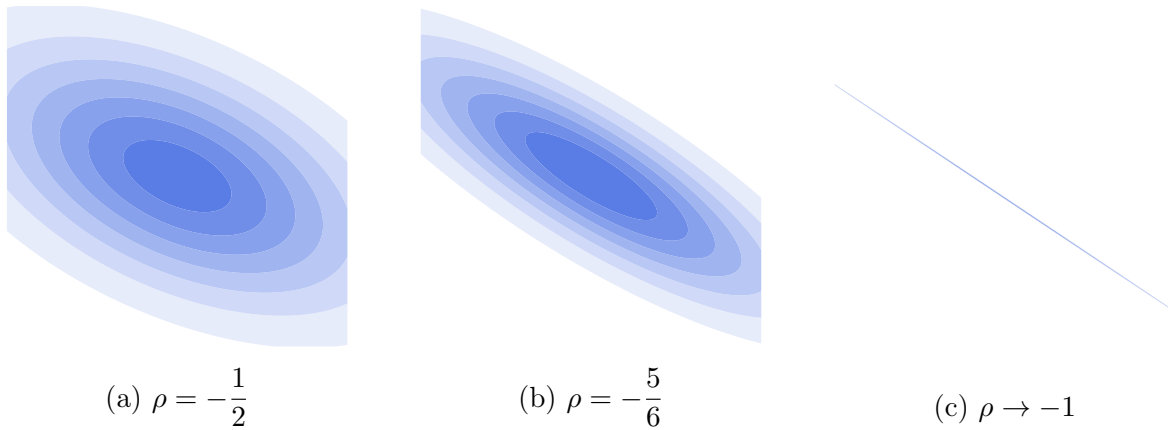


Figure 5: P.d.f. of a bivariate normal distribution with $\sigma_X = 3$ and $\sigma_Y = 2$.

The correlation of -1 means that there is a linear dependency between X and Y , but the change happens in opposite directions.

It means that standardized variables are opposite in sign:

$$\frac{X - \mu_X}{\sigma_X} = -\frac{Y - \mu_Y}{\sigma_Y},$$

$$Y = \mu_Y - \frac{\sigma_Y}{\sigma_X}(X - \mu_X).$$

From problem statement:

$$Y = -\frac{2}{3}X.$$

Thus, the probability:

$$\begin{aligned} P(2X + Y > 3) &= P\left(2X - \frac{2}{3}X > 3\right) = P\left(\frac{4}{3}X > 3\right) = P\left(X > \frac{9}{4}\right) = \\ &= 1 - \Phi(2.25) \approx 1 - 0.988 = \boxed{0.012}. \end{aligned}$$

Problem 9

The sample from bivariate normal distribution with random variables X and Y is following:

X	1.59	-2.20	-0.06	-1.45	-1.02	-2.59	-1.14	-3.25
Y	3.24	0.44	-1.14	5.40	2.09	5.33	1.25	8.72

Find 90% confidence interval for a population correlation coefficient ρ .

Solution:

$(1 - \alpha) \cdot 100\%$ confidence interval for ρ :

$$CI_{1-\alpha}(\rho) = \left(\tanh \left(\operatorname{artanh}(\hat{\rho}) - z_{\alpha/2} \cdot \frac{1}{\sqrt{n-3}} \right); \tanh \left(\operatorname{artanh}(\hat{\rho}) + z_{\alpha/2} \cdot \frac{1}{\sqrt{n-3}} \right) \right),$$

Values of sample mean:

$$\bar{x} = \frac{1}{8} \sum_{i=1}^8 x_i \approx -1.265, \quad \bar{y} = \frac{1}{8} \sum_{i=1}^8 y_i \approx 3.166.$$

Values of corrected sums:

$$\begin{aligned} SS_{xx} &= \sum_{i=1}^8 x_i^2 - 8 \cdot (-1.265)^2 \approx 16.283, \\ SS_{yy} &= \sum_{i=1}^8 y_i^2 - 8 \cdot 3.166^2 \approx 71.340, \\ SS_{xy} &= \sum_{i=1}^8 x_i y_i - 8 \cdot (-1.265) \cdot 3.166 \approx -17.240. \end{aligned}$$

Value of sample correlation coefficient:

$$\hat{\rho} = \frac{SS_{xy}}{\sqrt{SS_{xx} \cdot SS_{yy}}} = \frac{-17.240}{\sqrt{16.283 \cdot 71.340}} \approx -0.506.$$

Value of Fisher-transformed correlation coefficient:

$$\operatorname{artanh}(\hat{\rho}) = \frac{1}{2} \ln \left(\frac{1 + \hat{\rho}}{1 - \hat{\rho}} \right) = \frac{1}{2} \ln \left(\frac{1 - 0.506}{1 + 0.506} \right) \approx -0.557.$$

Confidence interval for Fisher-transformed ρ :

$$\begin{aligned}\text{CI}_{90\%}(\text{artanh}(\rho)) &= -0.557 \pm z_{0.05} \cdot \frac{1}{\sqrt{8-3}} = \\ &= -0.557 \pm 1.645 \cdot \frac{1}{\sqrt{8-3}} = \\ &= -0.557 \pm 0.736 = (-1.293; 0.179).\end{aligned}$$

Applying inverse Fisher-transform to the confidence interval above gives required interval:

$$\begin{aligned}\text{CI}_{90\%}(\rho) &= \tanh(\text{CI}_{90\%}(\text{artanh}(\rho))) = \frac{e^{2\text{CI}_{90\%}(\text{artanh}(\rho))} - 1}{e^{2\text{CI}_{90\%}(\text{artanh}(\rho))} + 1} = \\ &= \left(\frac{e^{2 \cdot (-1.293)} - 1}{e^{2 \cdot (-1.293)} + 1}, \frac{e^{2 \cdot 0.179} - 1}{e^{2 \cdot 0.179} + 1} \right) = \boxed{(-0.860; 0.177)}.\end{aligned}$$

Problem 10

Consider observations in the table below:

x	0	2	6	-3	4	1	-2	5	-1
y	8	2	0	6	1	5	7	3	4

1. Find Spearman's rank correlation coefficient r_s .
2. Find sample correlation coefficient r and compare it with r_s .

Solution:

1. Let's rank our sample and calculate differences d :

x	0	2	6	-3	4	1	-2	5	-1
y	8	2	0	6	1	5	7	3	4
rank(x)	4	6	9	1	7	5	2	8	3
rank(y)	9	3	1	7	2	6	8	4	5
d	-5	3	8	-6	5	-1	-6	4	-2
d^2	25	9	64	36	25	1	36	16	4

Spearman's rank correlation coefficient:

$$r_s = 1 - \frac{6 \sum_{i=1}^n d_i^2}{n(n^2 - 1)} = 1 - \frac{6 \cdot 216}{9 \cdot (81 - 1)} = \boxed{-0.8}.$$

2. Values of sample mean:

$$\bar{x} = \frac{1}{9} \sum_{i=1}^9 x_i = \frac{4}{3}, \quad \bar{y} = \frac{1}{9} \sum_{i=1}^9 y_i = 4.$$

Values of corrected sums:

$$SS_{xx} = \sum_{i=1}^9 x_i^2 - 9 \cdot \left(\frac{4}{3}\right)^2 = 80,$$

$$SS_{yy} = \sum_{i=1}^9 y_i^2 - 9 \cdot 4^2 = 60,$$

$$SS_{xy} = \sum_{i=1}^9 x_i y_i - 9 \cdot \frac{4}{3} \cdot 4 = -56.$$

Value of sample correlation coefficient:

$$r = \frac{SS_{xy}}{\sqrt{SS_{xx} \cdot SS_{yy}}} = \frac{-56}{\sqrt{80 \cdot 60}} \approx \boxed{-0.808}.$$

Clearly, correlation coefficients are close due to the absence of prominent outliers:

$$\boxed{r \approx r_s}.$$