

Probability Questions

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1. There are 30 questions in a certain multiple choice examination paper. Each question has 4 options and exactly one is to be marked by the candidate. Three candidates A,B,C mark each of the 30 questions at random independently. The probability that all the 30 answers of the three students match each other perfectly is?
2. Consider a Markov Chain with state space $\{0,1,2,3,4\}$ and transition matrix

$$P = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 1/3 & 1/3 & 1/3 & 0 & 0 \\ 0 & 1/3 & 1/3 & 1/3 & 0 \\ 0 & 0 & 1/3 & 1/3 & 1/3 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix} \end{matrix}$$

3. Consider the function $f(x)$ defined as $f(x) = ce^{-x^4}, x \in \mathbb{R}$. For what value of c is f a probability density function?
 - a) $\frac{2}{\Gamma(\frac{1}{4})}$
 - b) $\frac{4}{\Gamma(\frac{1}{4})}$
 - c) $\frac{3}{\Gamma(\frac{1}{3})}$
 - d) $\frac{1}{4\Gamma(4)}$
4. A random sample of size 7 is drawn from a distribution with p.d.f.

$$f_{\theta}(x) = \begin{cases} \frac{1+x^2}{3\theta(1+\theta^2)}, & -2\theta \leq x \leq \theta, \theta > 0 \\ 0, & \text{otherwise} \end{cases}$$

and the observations are 12, -54, 26, -2, 24, 17, -39. What is the maximum likelihood estimation of θ .

- a) 12
 - b) 24
 - c) 26
 - d) 27
5. Let X_1, X_2, X_3, X_4, X_5 be i.i.d random variables having a continuous distribution function. Then $P(X_1 > X_2 > X_3 > X_4 > X_5 | X_1 = \max(X_1, X_2, X_3, X_4, X_5))$ equals
 - a) $\frac{1}{4}$
 - b) $\frac{1}{5}$
 - c) $\frac{1}{4!}$
 - d) $\frac{1}{5!}$
 6. Suppose (X, Y) follows bivariate normal distribution with means μ_1, μ_2 , standard deviations σ_1, σ_2 and correlation coefficient ρ , where all parameters are un-known. Then, testing $H_0: \sigma_1 = \sigma_2$ is equivalent to testing the independence of
 - a) X and Y
 - b) X and X-Y
 - c) X+Y and Y
 - d) X+Y and X-Y

Solution:

Definition of Bivariate Gaussian and its independency Bi-variate random variables are distribution of normal distribution to two coordinates. are said to be bivariate normal or jointly normal, if $aX + bY$ has normal distribution $\forall a, b \in \mathbb{R}$.

Random normal vector

$$\mathbf{z} = \begin{pmatrix} X \\ Y \end{pmatrix} \quad (1)$$

is Bi-variate when it is jointly normal
Joint PDF of Z is given as

$$f_z(Z) = \frac{1}{2\pi \sqrt{\det \Sigma}} \exp \left\{ \frac{-1}{2} (z - m)^T \Sigma^{-1} (z - m) \right\} \quad (2)$$

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Where

$$\mathbf{m} = \begin{pmatrix} \mu_x \\ \mu_y \end{pmatrix} \quad (3)$$

$$\Sigma = [(\mathbf{Z} - \mathbf{E}(\mathbf{Z}))(\mathbf{Z} - \mathbf{E}(\mathbf{Z}))^T] \quad (4)$$

$$\Sigma = \begin{bmatrix} \sigma_x^2 & \rho\sigma_x\sigma_y \\ \rho\sigma_y\sigma_x & \sigma_y^2 \end{bmatrix} \quad (5)$$

If X, Y, which are independent, then they are un-correlated or their co-variances are $\rho\sigma_y\sigma_x = 0$ then co-variance matrix becomes a diagonal matrix

$$\Sigma = \begin{bmatrix} \sigma_x^2 & \rho\sigma_x\sigma_y \\ \rho\sigma_y\sigma_x & \sigma_y^2 \end{bmatrix} = \begin{bmatrix} \sigma_x^2 & 0 \\ 0 & \sigma_y^2 \end{bmatrix} \quad (6)$$

0.1 Co-variance Matrix

$$\Sigma = [(\mathbf{Z} - \mathbf{E}(\mathbf{Z}))(\mathbf{Z} - \mathbf{E}(\mathbf{Z}))^T] \quad (7)$$

When $\mathbf{Z}' = \mathbf{TZ}$, Where \mathbf{T} is a transformation

$$\Sigma_{TZ} = [(\mathbf{TZ} - \mathbf{E}(\mathbf{TZ}))(\mathbf{TZ} - \mathbf{E}(\mathbf{TZ}))^T] \quad (8)$$

$$= [(\mathbf{TZ} - \mathbf{TE}(\mathbf{Z}))(\mathbf{TZ} - \mathbf{TE}(\mathbf{Z}))^T] \quad (9)$$

$$= [\mathbf{T}(\mathbf{Z} - \mathbf{E}(\mathbf{Z}))(\mathbf{Z} - \mathbf{E}(\mathbf{Z}))^T \mathbf{T}^T] \quad (10)$$

$$= [\mathbf{T}(\mathbf{Z} - \mathbf{E}(\mathbf{Z}))(\mathbf{Z} - \mathbf{E}(\mathbf{Z}))^T \mathbf{T}^T] \quad (11)$$

$$= [\mathbf{T}\Sigma\mathbf{T}^T] \quad (12)$$

Where , $[(\mathbf{Z} - \mathbf{E}(\mathbf{Z}))(\mathbf{Z} - \mathbf{E}(\mathbf{Z}))^T] = \Sigma$

$$\Sigma_{TZ} = [\mathbf{T}\Sigma\mathbf{T}^T] \quad (13)$$

0.2 Few observations:

- \mathbf{T} is diagonal matrix
- σ_{TZ} is diagonal matrix.
- σ is symmetric.
- coloums of \mathbf{T} are eigen vectors.

Now if condition 1 and 4 satisfies then we can say a transformation is independent.

0.3 Evaluating option 1

: Given $\sigma_x = \sigma_y$

$$\Sigma = \begin{bmatrix} \sigma_x^2 & \rho\sigma_x\sigma_x \\ \rho\sigma_x\sigma_x & \sigma_y^2 \end{bmatrix} = \begin{bmatrix} \sigma_x^2 & \rho\sigma_x^2 \\ \rho\sigma_x^2 & \sigma_x^2 \end{bmatrix} = \sigma_x^2 \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix}$$

Σ is not a diagonal matrix so components of \mathbf{Z} in option 1 are not independent.

0.4 Evaluating option 2

$$\mathbf{X}, \mathbf{X}-\mathbf{Y} \text{ can be written as } \begin{bmatrix} 1 & 0 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} \mathbf{X} \\ \mathbf{Y} \end{bmatrix}$$

$$\text{Where } \mathbf{T} = \begin{bmatrix} 1 & 0 \\ 1 & -1 \end{bmatrix}, \mathbf{Z} = \begin{bmatrix} \mathbf{X} \\ \mathbf{Y} \end{bmatrix}$$

Co-variance matrix Σ for $\mathbf{X}, \mathbf{X}-\mathbf{Y}$ From Eq. 1 $\Sigma_{TZ} = [\mathbf{T}\Sigma\mathbf{T}^T]$

$$\Sigma_{TZ} = \begin{bmatrix} 1 & 0 \\ 1 & -1 \end{bmatrix} \sigma_x^2 \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & -1 \end{bmatrix}$$

Question 53) Suppose (\mathbf{X}, \mathbf{Y}) follows bivariate normal distribution with means μ_1, μ_2 , standard deviations σ_1, σ_2 and correlation coefficient ρ , where all parameters are un-known. Then, testing $H_0: \sigma_1 = \sigma_2$ is equivalent to testing the independence of

- 1.) \mathbf{X} and \mathbf{Y}
- 2.) \mathbf{X} and $\mathbf{X}-\mathbf{Y}$
- 3.) $\mathbf{X}+\mathbf{Y}$ and \mathbf{Y}
- 4.) $\mathbf{X}+\mathbf{Y}$ and $\mathbf{X}-\mathbf{Y}$

1 QUESTION 54

(52) Suppose $\mathbf{X} \sim \text{Cauchy}(0,1)$. Then the distribution of $\frac{1-\mathbf{X}}{1+\mathbf{X}}$ is?

2 SOLUTION

A continuous random variable \mathbf{X} follows **Cauchy distribution** with parameters μ and λ if its pdf is given by

$$f(x) = \begin{cases} \frac{\lambda}{\pi} \cdot \frac{1}{\lambda^2 + (x-\mu)^2}, & -\infty < x < \infty; \\ 0, & -\infty < \mu < \infty, \lambda > 0; \\ & \text{Otherwise.} \end{cases}$$

The parameter μ and λ are location and scale parameters respectively.

When $\mu=0$ and $\lambda=1$, then the distribution is called **Standard Cauchy Distribution**. The pdf of

standard Cauchy distribution is

$$f(x) = \begin{cases} \frac{1}{\pi} \cdot \frac{1}{1+x^2}, & -\infty < x < \infty; \\ 0, & \text{Otherwise.} \end{cases}$$

Let,

$$Y = \frac{1-X}{1+X} \quad (14)$$

Then, cdf of Y is

$$F_Y(y) = P(Y \leq y) \quad (15)$$

$$= P\left(\frac{1-X}{1+X} \leq y\right) \quad (16)$$

$$= P((1-X) \leq y(1+X)) \quad (17)$$

$$= P((1-X) \leq (y+yX)) \quad (18)$$

$$= P((1-y) \leq (X+yX)) \quad (19)$$

$$= P\left(\frac{1-y}{1+y} \leq X\right) \quad (20)$$

$$= 1 - P\left(X < \frac{1-y}{1+y}\right) \quad (21)$$

$$= 1 - \int_{-\infty}^{\frac{1-y}{1+y}} f(x) dx \quad (22)$$

$$= 1 - \int_{-\infty}^{\frac{1-y}{1+y}} \frac{1}{\pi} \cdot \frac{1}{1+x^2} dx \quad (23)$$

$$= 1 - \frac{1}{\pi} \cdot \left[\tan^{-1} x \right]_{-\infty}^{\frac{1-y}{1+y}} \quad (24)$$

$$= 1 - \frac{1}{\pi} \cdot \left[\left[\tan^{-1} x \right]_{-\infty}^0 + \left[\tan^{-1} x \right]_0^{\frac{1-y}{1+y}} \right] \quad (25)$$

$$= 1 - \frac{1}{\pi} \cdot \left[-\frac{\pi}{2} + \left[\tan^{-1} x \right]_0^{\frac{1-y}{1+y}} \right] \quad (26)$$

$$F_Y(y) = 1 - \frac{1}{\pi} \cdot \left[-\frac{\pi}{2} + \tan^{-1} \left(\frac{1-y}{1+y} \right) \right] \quad (27)$$

The pdf of Y is

$$f_Y(y) = \frac{dF_Y(y)}{dy} = -\frac{1}{\pi} \cdot \frac{d\left(\tan^{-1} \left(\frac{1-y}{1+y} \right)\right)}{dy} = -\frac{1}{\pi} \cdot \frac{1}{1 + \left(\frac{1-y}{1+y} \right)^2} \cdot \frac{d\left(\frac{1-y}{1+y} \right)}{dy} = -\frac{1}{\pi} \cdot \frac{1}{1 + \left(\frac{1-y}{1+y} \right)^2} \cdot \left(-\frac{2}{(y+1)^2} \right) = \frac{1}{\pi} \cdot \frac{1}{1+y^2} \quad (28)$$

Hence, $Y \sim \text{Cauchy}(0,1)$.