

Question 41.2023

EE22BTECH11051

Question:

Suppose that X_1, X_2, \dots, X_{10} are independent and identically distributed random vectors each having $N_2(\mu, \Sigma)$ distribution, where Σ is non-singular. If

$$U = \frac{1}{1 + (\bar{X} - \mu)^T \Sigma^{-1} (\bar{X} - \mu)}, \quad (1)$$

where $\bar{X} = \frac{1}{10} \sum_{i=1}^{10} X_i$ then the value of $\log_e \Pr(U \leq \frac{1}{2})$ equals

- 1) -5
- 2) -10
- 3) -2
- 4) -1

GATE ST 2023

Solution:

We are given a bivariate distribution;

$$X_i \sim N_2(\mu, \Sigma) \quad (2)$$

$$(3)$$

The distribution of \bar{X} can be given as:

The mean of \bar{X} will be the average of the means of X_1, X_2, \dots, X_{10} , which is:

$$\mu_{\bar{X}} = \frac{\mu + \mu + \mu + \dots + \mu}{10} = \frac{10\mu}{10} = \mu \quad (4)$$

And since the distributions X_1, X_2, \dots, X_{10} are independent, the covariance between them is zero. Hence we can find the new covariance as:

The variance of a sum of i.i.d random variables is calculated as

$$\text{var}(Y_n) = E \left[\left(\frac{1}{n} \sum_{i=1}^n X_i \right)^2 \right] - \left(E \left[\frac{1}{n} \sum_{i=1}^n X_i \right] \right)^2 \quad (5)$$

$$= \frac{1}{n^2} \left\{ E \left[\left(\sum_{i=1}^n X_i \right)^2 \right] - \left(E \left[\sum_{i=1}^n X_i \right] \right)^2 \right\} \quad (6)$$

But

$$E \left[\left(\sum_{i=1}^n X_i \right)^2 \right] = E \left[\sum_{i=1}^n \sum_{j=1}^n X_i X_j \right] \quad (7)$$

$$= \sum_{i=1}^n \sum_{j=1}^n E[X_i X_j] \quad (8)$$

and

$$\left(E \left[\sum_{i=1}^n X_i \right] \right)^2 = \left(\sum_{i=1}^n E[X_i] \right)^2 \quad (9)$$

$$= \sum_{i=1}^n \sum_{j=1}^n E[X_i] E[X_j] \quad (10)$$

Putting (8) and (10) in (6) , and using the definition of covariance,

$$\text{var}(Y_n) = \frac{1}{n^2} \left\{ \sum_{i=1}^n \sum_{j=1}^n (E[X_i X_j] - E[X_i] E[X_j]) \right\} \quad (11)$$

$$= \frac{1}{n^2} \left\{ \sum_{i=1}^n \sum_{j=1}^n \text{cov}(X_i, X_j) \right\} \quad (12)$$

As all the variables are i.i.d's and are thus uncorrelated,

$$\text{cov}(X_i, X_j) = \begin{cases} 0 & \text{if } i \neq j \\ \text{var}(X_i) & \text{if } i = j \end{cases} \quad (13)$$

$$\text{var}(Y_n) = \frac{1}{n^2} \left(\sum_{i=1}^n \text{cov}(X_i, X_i) \right) \quad (14)$$

$$= \frac{1}{n^2} \left(\sum_{i=1}^n \text{var}(X_i) \right) \quad (15)$$

$$= \frac{1}{n^2} \left(\sum_{i=1}^n \Sigma \right) \quad (16)$$

$$= \frac{\Sigma}{n} \quad (17)$$

hence;

$$\bar{X} \sim N_2 \left(\mu, \frac{\Sigma}{n} \right) \quad (18)$$

Now let us convert this normal distribution into χ^2_2 distribution; where χ^2_k distribution is given as;

$$f(x) = \frac{x^{\frac{k}{2}-1}}{2^{\frac{k}{2}} \Gamma(\frac{k}{2})} \quad x \geq 0 \quad (19)$$

$$(20)$$

hence we get;

$$\frac{(\bar{X} - \mu)^T (\bar{X} - \mu)}{\left(\frac{\Sigma}{n} \right)} \sim \chi^2_2 \quad (21)$$

we can now write this as;

$$n(\bar{X} - \mu)^T \Sigma^{-1} (\bar{X} - \mu) \sim \chi^2_2 \quad (22)$$

Let Y be;

$$Y = n(\bar{X} - \mu)^T \Sigma^{-1} (\bar{X} - \mu) \quad (23)$$

which follows χ^2_2 distribution
Now;

$$\Pr\left(U \leq \frac{1}{2}\right) = \Pr\left(\frac{1}{1 + \frac{Y}{n}} \leq \frac{1}{2}\right) \quad (24)$$

$$= \Pr\left(\frac{Y}{n} \geq 1\right) \quad (25)$$

$$= \Pr(Y \geq 10) \quad (26)$$

Since $k = 2$ for Y , we get;

$$= \int_{10}^{\infty} \frac{1}{2} e^{-\frac{y}{2}} dy \quad (27)$$

$$= e^{-5} \quad (28)$$

Hence;

$$\log_e \Pr\left(U \leq \frac{1}{2}\right) = -5 \quad (29)$$

Simulation Steps:

- 1) Set the number of simulations
- 2) Run a loop for number of simulations specified
- 3) Generate random vectors from a bivariate normal distribution
- 4) Calculate the sample mean and set the covariance matrix
- 5) Using the given formula, calculate U for each simulation
- 6) Calculate the simulated probability and take the natural log of that value

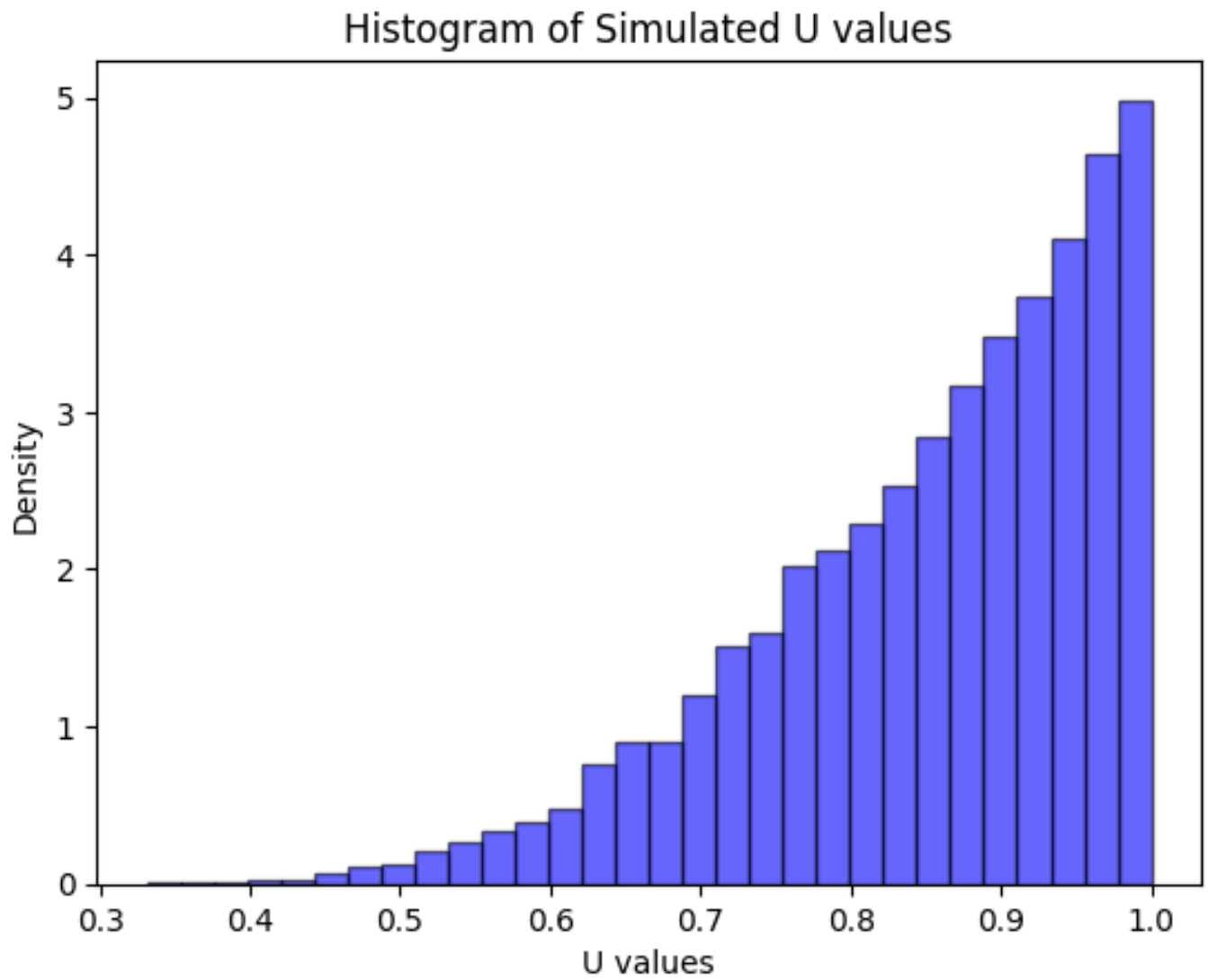


Fig. 1. Plot of $p_X(n)$. Simulations are close to the analysis.