

# Assignment 8

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Download all python codes from

<https://github.com/Adarsh541/AI1103-prob-and-ranvar/blob/main/Assignment8.1/codes/Assignment8.1.py>

and latex-tikz codes from

<https://github.com/Adarsh541/AI1103-prob-and-ranvar/blob/main/Assignment8.1/Assignment8.1.tex>

where

$$\mathbf{M} = \begin{pmatrix} \frac{4}{5} & -1/5 & -1/5 & -1/5 & -1/5 \\ -1/5 & \frac{4}{5} & -1/5 & -1/5 & -1/5 \\ -1/5 & -1/5 & \frac{4}{5} & -1/5 & -1/5 \\ -1/5 & -1/5 & -1/5 & \frac{4}{5} & -1/5 \\ -1/5 & -1/5 & -1/5 & -1/5 & \frac{4}{5} \end{pmatrix} \quad (2.0.5)$$

$$(2.0.6)$$

we also have

$$\mathbf{M}^2 = \mathbf{M} \quad (2.0.7)$$

## 1 PROBLEM

Let  $X_1, X_2, X_3, X_4, X_5$  be a random sample of size 5 from a population having standard normal distribution. If  $\bar{X} = \frac{1}{5} \sum_{i=1}^5 X_i$  and  $T = \sum_{i=1}^5 (X_i - \bar{X})^2$  then  $E[T^2 \bar{X}^2]$  is equal to

- 1) 3
- 2) 3.6
- 3) 4.8
- 4) 5.2

## 2 SOLUTION

Let  $\mathbf{x} \sim N(\mathbf{0}, \mathbf{I})$  be a standard normal random vector

$$\mathbf{x} = \begin{pmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \\ X_5 \end{pmatrix} \quad (2.0.1)$$

Then  $\bar{X}$  can be written as

$$\bar{X} = \frac{1}{5} \mathbf{u}^T \mathbf{x} \quad (2.0.2)$$

where

$$\mathbf{u} = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} \quad (2.0.3)$$

$$T = \mathbf{x}^T \mathbf{M} \mathbf{x} \quad (2.0.4)$$

## Lemma 2.1.

$$(\mathbf{u}^T \mathbf{x})(\mathbf{x}^T \mathbf{v}) = \mathbf{u}^T (\mathbf{x} \mathbf{x}^T) \mathbf{v} \quad (2.0.8)$$

This lemma is verified using python simulation.

## Definition 2.1 (cross-covariance).

$$\text{Cov}[\mathbf{x}, \mathbf{y}] = E[(\mathbf{x} - E[\mathbf{x}])(\mathbf{y} - E[\mathbf{y}])^T] \quad (2.0.9)$$

**Theorem 2.2.** Let  $\bar{X}$  be the sample mean of size 5 from a standard normal distribution. Then

- 1)  $\bar{X} \sim N(0, \frac{1}{5})$
- 2)  $\bar{X}$  and  $T$  are independent.
- 3)  $T \sim \chi_4^2$

where  $\chi_4^2$  is chi-square distribution with 4 degrees of freedom and  $T$  is defined as

$$T = \sum_{i=1}^5 (X_i - \bar{X})^2 \quad (2.0.10)$$

*Proof.*

$$\bar{X} = \frac{1}{5} \mathbf{u}^T \mathbf{x} \quad (2.0.11)$$

since  $\mathbf{M}$  is symmetric and idempotent we have

$$T = \mathbf{x}^T \mathbf{M} \mathbf{x} \quad (2.0.12)$$

$$= \mathbf{x}^T \mathbf{M} \mathbf{M} \mathbf{x} \quad (2.0.13)$$

$$= \mathbf{x}^T \mathbf{M}^T \mathbf{M} \mathbf{x} \quad (2.0.14)$$

$$= (\mathbf{M} \mathbf{x})^T (\mathbf{M} \mathbf{x}) \quad (2.0.15)$$

From (2.0.11) it can be seen that  $\bar{X}$  depends only on  $\mathbf{u}^T \mathbf{x}$  and from (2.0.15) it can be seen that  $T$  depends

only on  $\mathbf{M}\mathbf{x}$ . So if  $\mathbf{M}\mathbf{x}$  and  $\mathbf{u}^\top \mathbf{x}$  are independent then  $T$  and  $\bar{X}$  are also independent.

$$\begin{aligned} Cov[\mathbf{u}^\top \mathbf{x}, \mathbf{M}\mathbf{x}] &= E[(\mathbf{u}^\top \mathbf{x} - E[\mathbf{u}^\top \mathbf{x}]) \\ &\quad (\mathbf{M}\mathbf{x} - E[\mathbf{M}\mathbf{x}])^\top] \end{aligned} \quad (2.0.16)$$

From the lemma stated above we get

$$Cov[\mathbf{u}^\top \mathbf{x}, \mathbf{M}\mathbf{x}] = \mathbf{u}^\top E[(\mathbf{x} - E[\mathbf{x}]) (\mathbf{x} - E[\mathbf{x}])^\top] \mathbf{M}^\top \quad (2.0.17)$$

$$= \mathbf{u}^\top Var[\mathbf{x}] \mathbf{M} \quad (2.0.18)$$

$$= \mathbf{u}^\top \mathbf{M} \quad (2.0.19)$$

$$= 0 \quad (2.0.20)$$

Two jointly normal vectors are independent if and only if their cross-covariance is zero. So  $\mathbf{M}\mathbf{x}$  and  $\mathbf{u}^\top \mathbf{x}$  are independent. This implies  $\bar{X}$  and  $T$  are independent.  $\square$

$$E[T^2 \bar{X}^2] = E[T^2] E[\bar{X}^2] \quad (2.0.21)$$

$$E[\bar{X}^2] = Var[\bar{X}] + (E[\bar{X}])^2 \quad (2.0.22)$$

$$= \frac{1}{5} \quad (2.0.23)$$

since  $T$  is chi-squared distributed with 4 degrees of freedom

$$E[T] = 4 \quad (2.0.24)$$

$$Var[T] = 8 \quad (2.0.25)$$

$$\implies E[T^2] = Var[T] + (E[T])^2 \quad (2.0.26)$$

$$= 24 \quad (2.0.27)$$

From (2.0.21)

$$E[T^2 \bar{X}^2] = 4.8 \quad (2.0.28)$$