
GATE PROBABILITY

Through Simulations

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Contents

Introduction	iii
1 Axioms	1
2 Distributions	3
3 Conditional Probability	11
4 Moments	13
5 Random Algebra	21
6 Hypothesis Testing	25
7 Bivariate Random Variables	33
8 Random Processes	35
9 Convergence	37
10 Information Theory	43
11 Markov chain	49

Introduction

This book solves probability problems in GATE question papers.

Chapter 1

Axioms

Chapter 2

Distributions

2.1 Let $\phi(\cdot)$ denote the cumulative distribution function of a standard normal random variable. If the random variable X has the cumulative distribution function

$$F(x) = \begin{cases} \phi(x), & x < -1 \\ \phi(x+1), & x \geq -1 \end{cases} \quad (2.1)$$

then which one of the following statements is true?

(a) $P(X \leq -1) = \frac{1}{2}$

(b) $P(X = -1) = \frac{1}{2}$

(c) $P(X < -1) = \frac{1}{2}$

(d) $P(X \leq 0) = \frac{1}{2}$

(GATE ST 2023)

Solution: Gaussian

Q function is defined

$$Q(x) = \frac{1}{\sqrt{2\pi}} \int_x^\infty e^{-\frac{u^2}{2}} du \quad (2.2)$$

From question and (2.2);

$$F_X(x) = \begin{cases} Q(-x), & x < -1 \\ 1 - Q(x+1), & x \geq -1 \end{cases} \quad (2.3)$$

From (2.3);

(a)

$$\Pr(X \leq -1) = F_X(-1) = 1 - Q(0) \quad (2.4)$$

$$= 0.5 \quad (2.5)$$

So Option A i.e., $P(X < -1) = \frac{1}{2}$ is correct

(b) The pdf of X can be defined in terms of cdf as

$$\Pr(X = b) = F_X(b) - \lim_{x \rightarrow b^-} F_X(x) \quad (2.6)$$

From (2.6);

$$\Pr(X = -1) = F_X(-1) - \lim_{x \rightarrow -1^-} F_X(x) \quad (2.7)$$

$$= 1 - Q(0) - Q(-(-1)) \quad (2.8)$$

$$= 0.341 \quad (2.9)$$

So Option B i.e., $P(X = -1) = \frac{1}{2}$ is incorrect

(c)

$$\Pr(X < -1) = \lim_{x \rightarrow -1^-} F_X(x) = F_X(-1) \quad (2.10)$$

$$= Q(-(-1)) \quad (2.11)$$

$$= 0.159 \quad (2.12)$$

So Option C i.e., $P(X < -1) = \frac{1}{2}$ is incorrect

(d)

$$\Pr(X \leq 0) = F_X(0) = 1 - Q(1) \quad (2.13)$$

$$= 0.8413 \quad (2.14)$$

So Option D i.e., $P(X \leq 0) = \frac{1}{2}$ is incorrect

Gaussian CDF plot of X is given in fig2.1

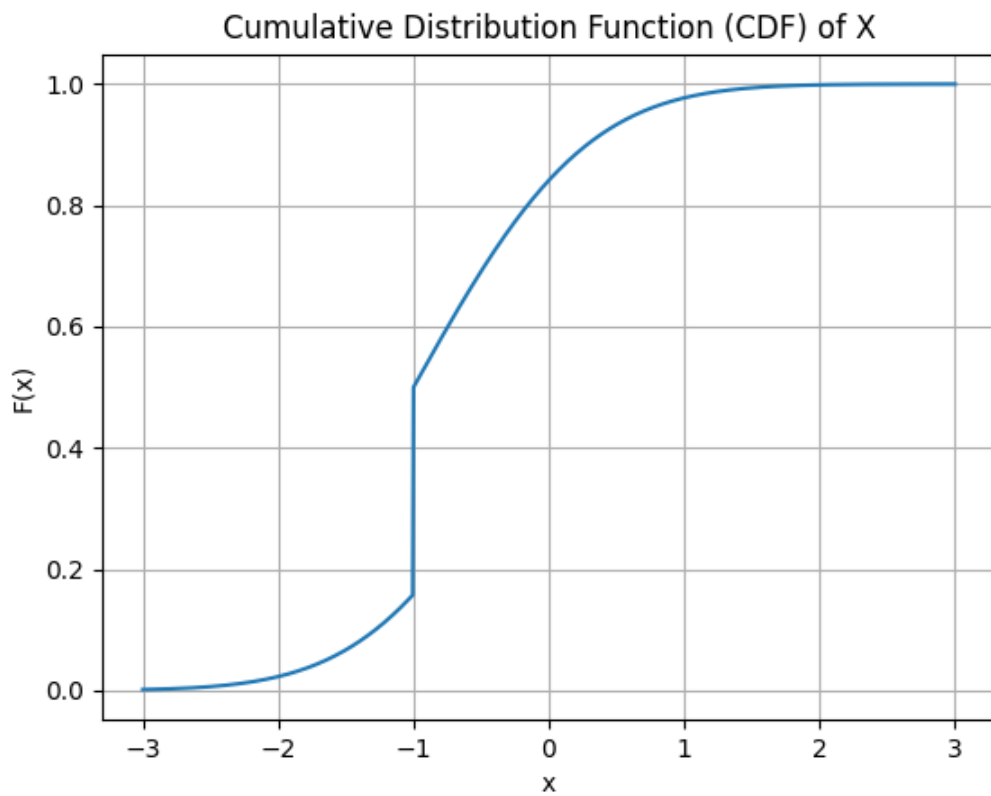


Figure 2.1:

2.2 Let X be a random variable with the probability density function $f(x)$ such that

$$f(x) = \begin{cases} \frac{1}{2\sqrt{3}}, & -\sqrt{3} \leq x \leq \sqrt{3} \\ 0, & \text{otherwise} \end{cases} \quad (2.15)$$

Then the variance of X is?

(GATE XH-C1 2023)

Solution:

The mean of X

$$\mu_X = \int_{-\infty}^{\infty} x f(x) dx \quad (2.16)$$

As the integrand is odd

$$\implies \mu_X = 0 \quad (2.17)$$

The variance of X is:

$$\sigma_X^2 = \mathbb{E} (X - \mu_X)^2 \quad (2.18)$$

From (2.17)

$$\implies \sigma_X^2 = \mathbb{E} (X^2) \quad (2.19)$$

$$= \frac{1}{2\sqrt{3}} \int_{-\sqrt{3}}^{\sqrt{3}} x^2 dx \quad (2.20)$$

$$= 1 \quad (2.21)$$

2.3 Two defective bulbs are present in a set of five bulbs. To remove the two defective bulbs, the bulbs are chosen randomly one by one and tested. If X denotes the minimum number of bulbs that must be tested to find out the two defective bulbs, then $\Pr(X = 3)$ (rounded off to two decimal places) equals
(GATE ST 2023)

Solution:

RV	Values	Description
A	0	1 st Bulb defective
	1	1 st Bulb non-defective
B	0	2 nd Bulb defective
	1	2 nd Bulb non-defective
C	0	3 rd Bulb defective
	1	3 rd Bulb non-defective

Table 2.1: Random variable declaration.

Here, the word "minimum" does not signify anything. Therefore we get

$$p_X(2) = p_{AB}(0, 0) \quad (2.22)$$

$$= \frac{2}{5} \times \frac{1}{4} \quad (2.23)$$

$$= \frac{1}{10} \quad (2.24)$$

$$p_X(3) = p_{ABC}(1, 0, 0) + p_{ABC}(0, 1, 0) + p_{ABC}(1, 1, 1) \quad (2.25)$$

$$= \frac{3}{5} \times \frac{2}{4} \times \frac{1}{3} + \frac{2}{5} \times \frac{3}{4} \times \frac{1}{3} + \frac{3}{5} \times \frac{2}{4} \times \frac{1}{3} \quad (2.26)$$

$$= \frac{3}{10} \quad (2.27)$$

$$p_X(4) = p_{ABC}(0, 1, 1) + p_{ABC}(1, 0, 1) + p_{ABC}(1, 1, 0) \quad (2.28)$$

$$= \frac{2}{5} \times \frac{3}{4} \times \frac{2}{3} + \frac{3}{5} \times \frac{2}{4} \times \frac{2}{3} + \frac{3}{5} \times \frac{2}{4} \times \frac{2}{3} \quad (2.29)$$

$$= \frac{6}{10} \quad (2.30)$$

Hence, The pmf of X is

$$p_X(k) = \begin{cases} 0 & k = 1 \\ \frac{1}{10} & k = 2 \\ \frac{3}{10} & k = 3 \\ \frac{6}{10} & k = 4 \\ 1 & k = 5 \end{cases} \quad (2.31)$$

Chapter 3

Conditional Probability

Chapter 4

Moments

4.1 Suppose that X has the probability density function

$$f(x) = \begin{cases} \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} & \lambda > 0 \\ 0 & \text{otherwise} \end{cases} \quad (4.1)$$

where $\alpha > 0$ and $\lambda > 0$. Which one of the following statements is NOT true?

- (a) $E(X)$ exists for all $\alpha > 0$ and $\lambda > 0$
- (b) Variance of X exists for all $\alpha > 0$ and $\lambda > 0$
- (c) $E(\frac{1}{X})$ exists for all $\alpha > 0$ and $\lambda > 0$
- (d) $E(\ln(1 + X))$ exists for all $\alpha > 0$ and $\lambda > 0$

(GATE ST 2023)

Solution:

(a)

$$E(X) = \int_{-\infty}^{\infty} xp_X(x)dx \quad (4.2)$$

$$= \int_0^{\infty} x \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} \quad (4.3)$$

$$= \frac{\lambda^\alpha}{\Gamma(\alpha)} \int_0^{\infty} x^\alpha e^{-\lambda x} \quad (4.4)$$

$$(4.5)$$

since we know that

$$\int_0^{\infty} x^{\alpha-1} e^{-\lambda x} dx = \frac{\Gamma(\alpha)}{\lambda^\alpha} \quad \text{for } \lambda > 0, \alpha > 0 \quad (4.6)$$

$$E(X) = \frac{\lambda^\alpha}{\Gamma(\alpha)} \frac{\Gamma(\alpha+1)}{\lambda^{\alpha+1}} \quad (4.7)$$

Using the relation

$$\Gamma(x+1) = \Gamma(x)x \quad (4.8)$$

$$E(X) = \frac{\alpha}{\lambda} \quad (4.9)$$

Thus $E(X)$ exists for all $\alpha > 0$ and $\lambda > 0$.

(b)

$$Var(X) = E(X^2) - E(X)^2 \quad (4.10)$$

$$E(X^2) = \int_0^{\infty} x^2 \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} dx \quad (4.11)$$

$$= \int_0^{\infty} \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{(\alpha+2)-1} e^{-\lambda x} dx \quad (4.12)$$

$$= \int_0^{\infty} \frac{1}{\lambda^2} \frac{\lambda^{\alpha+2}}{\Gamma(\alpha)} x^{(\alpha+2)-1} e^{-\lambda x} dx \quad (4.13)$$

$$E(X^2) = \int_0^{\infty} \frac{\alpha(\alpha+1)}{\lambda^2} \frac{\lambda^{\alpha+2}}{\Gamma(\alpha+2)} x^{(\alpha+2)-1} e^{-\lambda x} dx \quad (4.14)$$

using the density of the gamma distribution, we get

$$E(X^2) = \frac{\alpha(\alpha+1)}{\lambda^2} \quad (4.15)$$

$$Var(X) = \frac{\alpha^2 + \alpha}{\lambda^2} - \frac{\alpha^2}{\lambda} \quad (4.16)$$

$$= \frac{\alpha}{\lambda^2} \quad (4.17)$$

Thus, Variance of X exists for all $\alpha > 0$ and $\lambda > 0$

(c)

$$E\left(\frac{1}{X}\right) = \int_0^{\infty} \frac{1}{x} \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} dx \quad (4.18)$$

$$= \frac{\lambda^\alpha}{\Gamma(\alpha)} \int_0^{\infty} x^{\alpha-2} e^{-\lambda x} dx \quad (4.19)$$

For this, $\alpha > 1$ is a must condition. Hence C is not a correct option. Hence C is the answer.

(d) For $E(\ln(1 + X))$,

$$E(\ln(1 + X)) = E(X) - \frac{E(X^2)}{2} + \frac{E(X^4)}{4} - .. \quad (4.20)$$

We write the general expression for $E(X^n)$

$$E(X^n) = \frac{(\alpha)(\alpha + 1) \dots (\alpha + n - 1)}{\lambda^n} \quad (4.21)$$

So by applying the ratio test to check the convergence of the sequence

$$\lim_{n \rightarrow \infty} \left| \frac{a_{n+1}}{a_n} \right| = L \quad (4.22)$$

$$\left| \frac{E(X^{n+2})}{E(X^n)} \right| = \frac{\frac{(\alpha)(\alpha+1)\dots(\alpha+n+1)}{\lambda^{n+2}}}{\frac{(\alpha)(\alpha+1)\dots(\alpha+n-1)}{\lambda^n}} \quad (4.23)$$

$$= \frac{(\alpha + n)(\alpha + n + 1)}{\lambda^2} \quad (4.24)$$

$$\lim_{n \rightarrow \infty} \left| \frac{E(X^{n+2})}{E(X^n)} \right| = \infty \quad (4.25)$$

Thus $E(\ln(1 + X))$ generates a divergent function and hence $E(\ln(1 + X))$ does not exist for all $\alpha > 0$ and $\lambda > 0$.

4.2 Let X be a random variable with probability density function

$$f(x; \lambda) = \begin{cases} \frac{1}{\lambda} e^{-\frac{x}{\lambda}} & \text{if } x > 0 \\ 0 & \text{otherwise} \end{cases} \quad (4.26)$$

where $\lambda > 0$ is an unknown parameter. Let Y_1, Y_2, \dots, Y_n be a random sample of size

n from a population having the same distribution as X^2 . If

$$\bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i \quad (4.27)$$

then which of the following statements is true?

- (a) $\sqrt{\frac{\bar{Y}}{2}}$ is a method of moments estimator of λ
- (b) $\sqrt{\bar{Y}}$ is a method of moments estimator of λ
- (c) $\frac{1}{2}\sqrt{\bar{Y}}$ is a method of moments estimator of λ
- (d) $2\sqrt{\bar{Y}}$ is a method of moments estimator of λ (GATE ST 2023)

Solution:

- (a) Using PDF in (4.26) we need to find an estimator for the unknown parameter λ in terms of sample mean \bar{Y}

we know $Y_i = X_i^2$ then,

$$E(Y_i) = E(X_i^2) \quad (4.28)$$

$$= \int_0^\infty x^2 \frac{1}{\lambda} e^{-\frac{x}{\lambda}} \quad (4.29)$$

$$= 2\lambda^2 \quad (4.30)$$

Method of moment is defined by (4.27) which gives,

$$\bar{Y} = E(Y_i) \quad (4.31)$$

$$= 2\lambda^2 \quad (4.32)$$

where

$$\lambda = \sqrt{\frac{\bar{Y}}{2}} \quad (4.33)$$

∴ Option (4.2a) is correct.

(b) The simulation steps to estimate λ using method of moment estimator in python.

- i. Generate a random value of λ within the specified range using **np.random.uniform(1,10)**
- ii. Use the generated λ to create a random sample of X values following the given PDF using **np.random.exponential()**
- iii. Then, generate Y as $Y = X^2$
- iv. calculate the mean (\bar{Y}) as **np.mean(Y)**
- v. Hence, the estimated value of λ is **np.sqrt($\frac{\bar{Y}}{2}$)**

Graph of simulated CDF vs Theoretical CDF

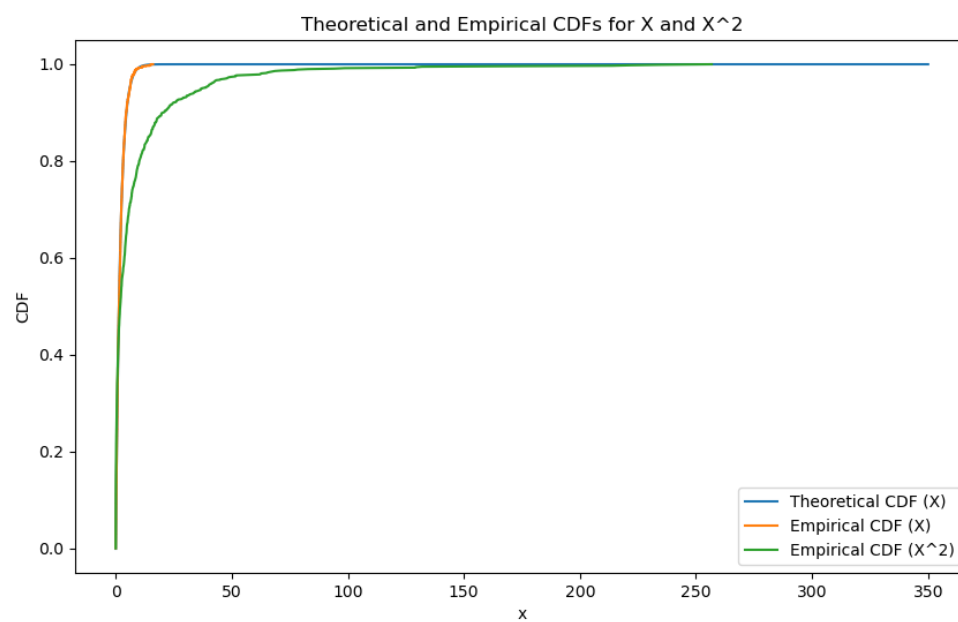


Figure 4.1: Figure1

Chapter 5

Random Algebra

1. Let (X, Y) have joint probability density function

$$p_{XY}(x, y) = \begin{cases} 8xy & \text{if } 0 < x < y < 1 \\ 0 & \text{otherwise} \end{cases} \quad (5.1)$$

if $E(X|Y = y_0) = \frac{1}{2}$, then y_0 equals

(a) $\frac{3}{4}$

(b) $\frac{1}{2}$

(c) $\frac{1}{3}$

(d) $\frac{2}{3}$

(GATE ST 2023)

Solution:

$$E(X|Y) = \int_{-\infty}^{\infty} xp_{X|Y}dx \quad (5.2)$$

where

$$p_{X|Y} = \frac{p_{XY}(x, y)}{p_Y(y)} \quad (5.3)$$

$$p_Y(y) = \int_0^y p_{X|Y}(x, y) dx \quad (5.4)$$

for $0 < y < 1$

$$= \int_0^y 8xy dx \quad (5.5)$$

$$= 8y \left[\frac{x^2}{2} \right]_0^y \quad (5.6)$$

$$= 4y^3 \quad (5.7)$$

For $0 < x < y < 1$, on substituting $p_Y(y)$ we get

$$p_{X|Y} = \frac{8xy}{4y^3} \quad (5.8)$$

$$= \frac{2x}{y^2} \quad (5.9)$$

and

$$E(X|Y = y_0) = \int_0^{y_0} x \cdot \frac{2x}{y_0^2} dx \quad (5.10)$$

$$= \frac{2}{y_0^2} \left[\frac{x^3}{3} \right]_0^{y_0} \quad (5.11)$$

$$= \frac{2y_0}{3} \quad (5.12)$$

$$\implies \frac{2y_0}{3} = \frac{1}{2} \quad (5.13)$$

$$y_0 = \frac{3}{4} \quad (5.14)$$

Chapter 6

Hypothesis Testing

6.1 Suppose that x is an observed sample of size 1 from a population with probability density function $f(\cdot)$. Based on x , consider testing

$$H_0 : f(y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}}; \quad y \in \mathbb{R}$$

against

$$H_1 : f(y) = \frac{1}{2} e^{-|y|}; \quad y \in \mathbb{R}.$$

Then which one of the following statements is true?

- (a) The most powerful test rejects H_0 if $|x| > c$ for some $c > 0$
- (b) The most powerful test rejects H_0 if $|x| < c$ for some $c > 0$
- (c) The most powerful test rejects H_0 if $||x| - 1| > c$ for some $c > 0$
- (d) The most powerful test rejects H_0 if $||x| - 1| < c$ for some $c > 0$

(GATE ST 2023) **Solution:**

$$L = \prod_{i=1}^1 f(x) = f(x) \tag{6.1}$$

To determine the most powerful test, we need to consider the likelihood ratio test

$$\frac{L(H_1)}{L(H_0)} \underset{H_0}{\overset{H_1}{\geq}} k \quad (6.2)$$

$$\implies \frac{\frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}}{\frac{1}{2} e^{-2|x|}} \underset{H_0}{\overset{H_1}{\geq}} k \quad (6.3)$$

$$\implies e^{\frac{x^2 - 2|x|}{2}} \underset{H_0}{\overset{H_1}{\geq}} k \frac{\sqrt{\pi}}{\sqrt{2}} \quad (6.4)$$

$$(|x| - 1)^2 \underset{H_0}{\overset{H_1}{\geq}} 2 \log \left(\frac{k\sqrt{\pi}}{\sqrt{2}} \right) + 1 \quad (6.5)$$

Taking square root on both sides,

$$||x| - 1| \underset{H_0}{\overset{H_1}{\geq}} \sqrt{2 \log \left(\frac{k\sqrt{\pi}}{\sqrt{2}} \right) + 1} \quad (6.6)$$

$$\implies |x| \underset{H_0}{\overset{H_1}{\geq}} 1 + \sqrt{2 \log \left(\frac{k\sqrt{\pi}}{\sqrt{2}} \right) + 1} \quad (6.7)$$

Hence, the correct answer is (6.1c)

6.2 Suppose that X_1, X_2, \dots, X_n are independent and identically distributed random variables, each having probability density function $f(\cdot)$ and median θ . We want to test $H_0 : \theta = \theta_0$ against $H_1 : \theta > \theta_0$

Consider a test that rejects H_0 if $S > c$ for some c depending on the size of the test, where S is the cardinality of the set $\{i : X_i > \theta_0, 1 \leq i \leq n\}$. Then which one of the following statements is true?

- (a) Under H_0 , the distribution of S depends on $f(\cdot)$.
- (b) Under H_1 , the distribution of S does not depend on $f(\cdot)$.
- (c) The power function depends on θ .
- (d) The power function does not depend on θ .

(GATE ST 2023)

Solution:

Definition 6.1: Median θ is defined as

$$\Pr(X_i \leq \theta) = 0.5 \text{ for all } i \text{ from } 1 \text{ to } n.$$

Definition 6.2: S is defined as

$$S = \sum_{i=1}^n I(X_i > \theta_0)$$

where $I(X_i > \theta_0)$ represents an indicator function.

$$I(X_i > \theta_0) = \begin{cases} 1, & \text{if } X_i > \theta_0 \\ 0, & \text{if } X_i \leq \theta_0 \end{cases} \quad (6.8)$$

$$E(S) = E\left(\sum_{i=1}^n I(X_i > \theta_0)\right) \quad (6.9)$$

$$= \sum_{i=1}^n E(I(X_i > \theta_0)) \quad (6.10)$$

Since,

$$E(I(X_i > \theta_0)) = P(X_i > \theta_0) = \int_{\theta_0}^{\infty} f(x) dx \quad (6.11)$$

Therefore,

$$E(S) = \sum_{i=1}^n \int_{\theta_0}^{\infty} f(x) dx \quad (6.12)$$

(a) From (6.12), under H_0 , the distribution of S depends on $f(\cdot)$.

(b) The power function can be expressed as:

$$\pi(\theta) = \Pr(\text{Reject } H_0 \mid H_1 \text{ is true}) \quad (6.13)$$

$$= \Pr(S > c \mid \theta) \quad (6.14)$$

Therefore, power function depends on value of θ .

6.3 Let $X_1, X_2, X_3, \dots, X_n$ be a random sample of size $n (\geq 2)$ from a population having probability density function

$$f(x; \theta) = \begin{cases} \frac{2}{\theta x} (\log_e x) e^{-\frac{(\log_e x)^2}{\theta}} & , 0 < x < 1 \\ 0 & , \text{otherwise} \end{cases}$$

where $\theta > 0$ is an unknown parameter. Then which of the following statements is true,

- (A) $\frac{1}{n} \sum_{i=1}^n (\ln X_i)^2$ is the maximum likelihood estimator of θ
- (B) $\frac{1}{n-1} \sum_{i=1}^n (\ln X_i)^2$ is the maximum likelihood estimator of θ
- (C) $\frac{1}{n} \sum_{i=1}^n \ln X_i$ is the maximum likelihood estimator of θ
- (D) $\frac{1}{n-1} \sum_{i=1}^n \ln X_i$ is the maximum likelihood estimator of θ

(GATE ST 2023)

Solution:

$$L(\theta) = f(x_1, x_2, \dots, x_n; \theta) \quad (6.15)$$

The product of pdfs can be used to approximate the likelihood function even if the variables are dependent. This is a general approach that is often used in practice to estimate MLE of θ . Therefore,

$$L(\theta) = \prod_{i=1}^n f(x_i; \theta) \quad (6.16)$$

Maximizing $L(\theta)$ is equivalent to maximizing the $\ln L(\theta)$ as \ln is a monotonically increasing function.

$$l(\theta) = \ln L(\theta) \quad (6.17)$$

$$= \ln \left(\prod_{i=1}^n f(x_i; \theta) \right) \quad (6.18)$$

$$= \sum_{i=1}^n \ln f(x_i; \theta) \quad (6.19)$$

$$= -n \ln 2 - n \ln \theta + \sum_{i=1}^n \ln(-\ln x_i) - \sum_{i=1}^n (\ln x_i) - \sum_{i=1}^n \frac{(\ln x_i)^2}{\theta} \quad (6.20)$$

Maximizing $l(\theta)$ with respect to θ gives the MLE estimation, therefore

$$\frac{\partial l(\theta)}{\partial \theta} = 0 \quad (6.21)$$

$$\frac{-n}{\theta} + \frac{1}{(\theta)^2} \sum_{i=1}^n (\ln x_i)^2 = 0 \quad (6.22)$$

$$\theta = \frac{1}{n} \sum_{i=1}^n (\ln x_i)^2 \quad (6.23)$$

Hence (A) is the true statement.

6.4 Suppose that (X, Y) has joint probability mass function

$$P(X = 0, Y = 0) = P(X = 1, Y = 1) = \theta, \quad (6.24)$$

$$P(X = 1, Y = 0) = P(X = 0, Y = 1) = \frac{1}{2} - \theta. \quad (6.25)$$

where $0 \leq \theta \leq \frac{1}{2}$ is an unknown parameter. Consider testing $H_0 : \theta = \frac{1}{4}$ against $H_1 : \theta = \frac{1}{3}$; based on a random sample $(X_1, Y_1), (X_2, Y_2), \dots, (X_n, Y_n)$ from the above probability mass function. Let M be the cardinality of the set $\{i : X_i = Y_i, 1 \leq i \leq n\}$. If m is the observed value of M , then which one of the following statements is true?

- (a) The likelihood ratio test rejects H_0 if $m > c$ for some c .
- (b) The likelihood ratio test rejects H_0 if $m < c$ for some c .
- (c) The likelihood ratio test rejects H_0 if $c_1 < m < c_2$ for some c_1 and c_2 .
- (d) The likelihood ratio test rejects H_0 if $m < c_1$ or $m > c_2$ for some c_1 and c_2 .

(GATE ST 2023)

Solution: Given that,

$$H_0 : \quad \theta = \theta_0 = \frac{1}{4}, \quad (6.26)$$

$$H_1 : \quad \theta = \theta_1 = \frac{1}{3}. \quad (6.27)$$

and the pmf is given by

$$p_{XY}(0, 0) = p_{XY}(1, 1) = \theta \quad (6.28)$$

$$p_{XY}(0, 1) = p_{XY}(1, 0) = \frac{1}{2} - \theta \quad (6.29)$$

Then for the given random sample of data,

$$p_{X_i, Y_i}(x, y) = \begin{cases} 2\theta & x = y \\ 1 - 2\theta & x \neq y \end{cases} \quad (6.30)$$

$$(6.31)$$

Then the likelihood of the data under H_0 is given by:

$$L(\theta_0 \mid data) = \prod_{i=1}^n p_{X_i, Y_i}(x, y) \quad (6.32)$$

$$= (2\theta_0)^m (1 - 2\theta_0)^{n-m} \quad (6.33)$$

$$= \left(\frac{1}{2}\right)^m \left(\frac{1}{2}\right)^{n-m} \quad (6.34)$$

Then the likelihood of the data under H_1 is given by:

$$L(\theta_1 \mid data) = \prod_{i=1}^n p_{X_i, Y_i}(x, y) \quad (6.35)$$

$$= (2\theta_1)^m (1 - 2\theta_1)^{n-m} \quad (6.36)$$

$$= \left(\frac{2}{3}\right)^m \left(\frac{1}{3}\right)^{n-m} \quad (6.37)$$

The likelihood ratio will be

$$\lambda(data) = \frac{L(\theta_1 \mid x)}{L(\theta_0 \mid x)} \quad (6.38)$$

$$= \frac{\left(\frac{2}{3}\right)^m \left(\frac{1}{3}\right)^{n-m}}{\left(\frac{1}{2}\right)^m \left(\frac{1}{2}\right)^{n-m}} = (2)^m \left(\frac{2}{3}\right)^n \quad (6.39)$$

Let the critical value be denoted by c_1 , then the likelihood ratio test rejects H_0 if

$$\implies \lambda(data) \underset{H_0}{\overset{H_1}{\geq}} c_1 \quad (6.40)$$

$$(6.41)$$

From (6.39),

$$\implies (2)^m \left(\frac{2}{3}\right)^n \underset{H_0}{\overset{H_1}{\geq}} c_1 \quad (6.42)$$

$$\implies (2)^m \underset{H_0}{\overset{H_1}{\geq}} c_1 \left(\frac{2}{3}\right)^n \quad (6.43)$$

$$\implies m \underset{H_0}{\overset{H_1}{\geq}} \log_2 \left(c_1 \left(\frac{2}{3}\right)^n \right) \quad (6.44)$$

$$\implies m \underset{H_0}{\overset{H_1}{\geq}} c \quad \exists c \in \mathbb{R} \quad (6.45)$$

where,

$$c = \log_2 \left(c_1 \left(\frac{2}{3}\right)^n \right) \quad (6.46)$$

\therefore From (6.45), Option A is correct and Options B,C,D are incorrect

Chapter 7

Bivariate Random Variables

Chapter 8

Random Processes

Chapter 9

Convergence

9.1 Let $\{X_n\}_{n \geq 1}$ and $\{Y_n\}_{n \geq 1}$ be two sequences of random variables and X and Y be two random variables, all of them defined on the same probability space. Which one of the following statements is true?

- (A) If $\{X_n\}_{n \geq 1}$ converges in distribution to a real constant c , then $\{X_n\}_{n \geq 1}$ converges in probability to c .
- (B) If $\{X_n\}_{n \geq 1}$ converges in probability to X , then $\{X_n\}_{n \geq 1}$ converges in 3^{rd} mean to X .
- (C) If $\{X_n\}_{n \geq 1}$ converges in distribution to X and $\{Y_n\}_{n \geq 1}$ converges in distribution to Y , then $\{X_n + Y_n\}_{n \geq 1}$ converges in distribution to $X + Y$.
- (D) If $\{E(X_n)\}_{n \geq 1}$ converges to $E(X)$, then $\{X_n\}_{n \geq 1}$ converges in 1^{st} mean to X .

(GATE ST 2023) **Solution:**

- (a) X_n converges in distribution to X , $X_n \xrightarrow{d} X$, then for all x ,

$$\lim_{n \rightarrow \infty} F_{X_n}(x) = F_X(x) \quad (9.1)$$

(b) X_n converges in probability to X , $X_n \xrightarrow{p} X$, then for all $\epsilon > 0$,

$$\lim_{n \rightarrow \infty} \Pr(|X_n - X| > \epsilon) = 0 \quad (9.2)$$

(c) X_n converges in p^{th} mean to X , then we have

$$\lim_{n \rightarrow \infty} E(|X_n - X|^p) = 0 \quad (9.3)$$

(A) For $\epsilon > 0$, B be defined as

$$B = \{x : |x - c| \geq \epsilon\} \quad (9.4)$$

Now,

$$\Pr(|X_n - c| \geq \epsilon) = \Pr(X_n \in B) \quad (9.5)$$

Using Portmanteau Lemma, if $X_n \xrightarrow{d} c$, we have

$$\limsup_{n \rightarrow \infty} \Pr(X_n \in B) \leq \Pr(c \in B) \quad (9.6)$$

$$\leq \Pr(|0 - 0| \geq \epsilon) \quad (9.7)$$

$$\leq \Pr(0 \geq \epsilon) \quad (9.8)$$

$$\leq 0 \quad (9.9)$$

$$= 0 \quad (9.10)$$

$$\lim_{n \rightarrow \infty} \Pr(|X_n - c| > \epsilon) = 0 \quad (9.11)$$

From (9.2), $X_n \xrightarrow{p} c$. So, we have

$$X_n \xrightarrow{d} c \implies X_n \xrightarrow{p} c \quad (9.12)$$

Option (A) is correct.

(B) Statement (B) is may or may not correct. Counter Example: Consider distribution

X_n	0	n
$\Pr(X_n)$	$1 - \frac{1}{n}$	$\frac{1}{n}$

For $\epsilon > 0$, X_n converges in probability to $X = 0$

$$\lim_{n \rightarrow \infty} \Pr(|X_n - X| > \epsilon) = \lim_{n \rightarrow \infty} \Pr(X_n > \epsilon) \quad (9.13)$$

$X_n > \epsilon$ is subset of $X_n = n$ since every time X_n equals n , it's also true that X_n is greater than ϵ . But there may be times when X_n is greater than ϵ without X_n being equal to n . So,

$$\Pr(X_n > \epsilon) \leq \Pr(X_n = n) \quad (9.14)$$

$$\lim_{n \rightarrow \infty} \Pr(|X_n - X| > \epsilon) \leq \lim_{n \rightarrow \infty} \Pr(X_n = n) \quad (9.15)$$

$$\leq \lim_{n \rightarrow \infty} \frac{1}{n} \quad (9.16)$$

$$\leq 0 \quad (9.17)$$

$$= 0 \quad (9.18)$$

But X_n does not converges in 3^{rd} mean to $X = 0$.

$$\lim_{n \rightarrow \infty} E(|X_n - X|^3) = \lim_{n \rightarrow \infty} E(X_n^3) \quad (9.19)$$

$$= \lim_{n \rightarrow \infty} 0^3 \left(1 - \frac{1}{n}\right) + n^3 \left(\frac{1}{n}\right) \quad (9.20)$$

$$= \lim_{n \rightarrow \infty} n^2 \neq 0 \quad (9.21)$$

(C) Statement (C) is may or may not correct. Counter Example: Consider distribution

$$Z \sim \mathcal{N}(0, 1) \quad (9.22)$$

Let $\{X_n\}_{n \geq 1}$ and $\{Y_n\}_{n \geq 1}$ be sequences of random variables such that they both converge in distribution as Z and $(-1)^n Z$. Proof that Y_n converges in distribution.

For n even

$$\lim_{n \rightarrow \infty} F_{Y_n}(x) = \Pr(Z \leq x) \quad (9.23)$$

For n odd

$$\lim_{n \rightarrow \infty} F_{Y_n}(x) = \Pr(-Z \leq x) \quad (9.24)$$

$$= \Pr(Z \leq x) \quad (9.25)$$

Proved. So, we have

$$F_{X_n+Y_n}(x) = \Pr(X_n + Y_n \leq x) \quad (9.26)$$

$$= \Pr(Z + (-1)^n Z \leq x) \quad (9.27)$$

For n is even

$$F_{X_n+Y_n}(x) = \Pr(2Z \leq x) \quad (9.28)$$

$$= \Pr\left(Z \leq \frac{x}{2}\right) \quad (9.29)$$

$$= 1 - \Pr\left(Z > \frac{x}{2}\right) \quad (9.30)$$

$$\approx 1 - Q\left(\frac{x}{2}\right) \quad (9.31)$$

For n is odd

$$F_{X_n+Y_n}(x) = \Pr(0 \leq x) \quad (9.32)$$

$$= \begin{cases} 1 & \text{if } x \geq 0 \\ 0 & \text{if } x < 0 \end{cases} = H(x) \quad (9.33)$$

So, on generalizing

$$F_{X_n+Y_n}(x) = \begin{cases} 1 - Q\left(\frac{x}{2}\right) & \text{if } n \text{ is even} \\ H(x) & \text{if } n \text{ is odd} \end{cases} \quad (9.34)$$

$\lim_{n \rightarrow \infty} F_{X_n+Y_n}(x)$ oscillate between $1 - Q\left(\frac{x}{2}\right)$ and $H(x)$. This doesnot imply convergence.

(D) Statement (D) is may or may not correct. Counter Example: Consider

X_n	0	n
$\Pr(X_n)$	$1 - \frac{1}{n}$	$\frac{1}{n}$

$$\lim_{n \rightarrow \infty} E(X_n) = 0 \left(1 - \frac{1}{n}\right) + n \left(\frac{1}{n}\right) \quad (9.35)$$

$$= 1 \quad (9.36)$$

As $n \rightarrow \infty$, $E(X_n)$ converges to $E(X) = 1$.

$$\lim_{n \rightarrow \infty} X_n = 0 = X \quad (9.37)$$

To find 1st mean convergence of X_n . From (9.36)

$$\lim_{n \rightarrow \infty} E(|X_n - X|) = \lim_{n \rightarrow \infty} E(X_n) \quad (9.38)$$

$$= 1 \neq 0 \quad (9.39)$$

So, X_n does not converge in 1st mean to X .

Chapter 10

Information Theory

1. The frequency of occurrence of 8 symbols (a-h) is shown in the table below. A symbol is chosen and it is determined by asking a series of “yes/no” questions which are assumed to be truthfully answered. The average number of questions when asked in the most efficient sequence, to determine the chosen symbol, is

Symbols	Frequency of occurrence
a	$\frac{1}{2}$
b	$\frac{1}{4}$
c	$\frac{1}{8}$
d	$\frac{1}{16}$
e	$\frac{1}{32}$
f	$\frac{1}{64}$
g	$\frac{1}{128}$
h	$\frac{1}{128}$

Solution:

Parameter	Value	Description
X	$1 \leq X \leq 8$	number of symbols
l	2	base of algorithm
$H(X)$	$\sum_i p_X(i) \log_l \left(\frac{1}{p_X(i)} \right)$	average number of question

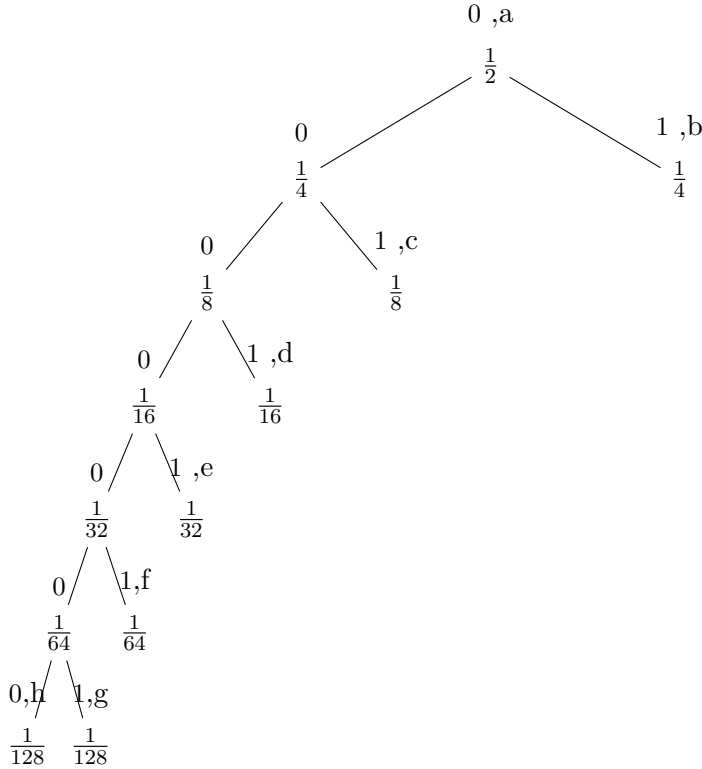
$$H(X) = \sum_i p_X(i) \log_b \left(\frac{1}{p_X(i)} \right) \quad (10.1)$$

$$= \frac{1}{2} \log_2 (2) + \frac{1}{4} \log_2 (4) + \dots + \frac{1}{128} \log_2 (128) \quad (10.2)$$

$$= 0.5 + 0.5 + 0.375 + \dots + 0.0078125 \quad (10.3)$$

$$= 1.984375 \quad (10.4)$$

Now, finding the average using Huffman code,



Using the above binary table following code is generated;

2. The transition diagram of a discrete memoryless channel with three input symbols and three output symbols is shown in the figure. The transition probabilities are as marked.

The parameter α lies in the interval $[0.25, 1]$. The value of α for which the capacity

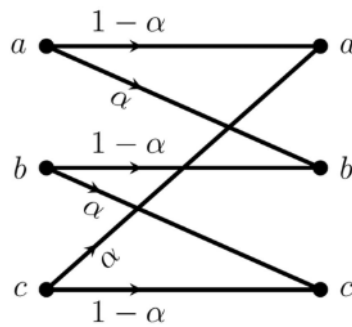
Symbols	Frequency	Code	Size
a	$\frac{1}{2}$	1	0.5
b	$\frac{1}{4}$	01	0.25
c	$\frac{1}{8}$	001	0.125
d	$\frac{1}{16}$	0001	0.0625
e	$\frac{1}{32}$	00001	0.03125
f	$\frac{1}{64}$	000001	0.015625
g	$\frac{1}{128}$	0000001	0.0078125
h	$\frac{1}{128}$	0000000	0.0078125

Table 10.1: Huffman table

The average number of question = Weighted path length = 1.9844

of this channel is maximized, is

(GATE EC 2022) **Solution:**



Variable	Description	Value
x_i	Input	x_0, x_1, x_2
y_i	Output	y_0, y_1, y_2
p_i	Input probability	p_0, p_1, p_2
q_i	Output probability	q_0, q_1, q_2
C	Channel Capacity	C
I	Mutual Information	I
H	Entropy	H

$$C = \sup_{p_X(x)} I(X, Y) \quad (10.5)$$

$$I(X, Y) = \sum_{x,y} p(x, y) \log_2 \frac{p(x, y)}{p(x)p(y)} \quad (10.6)$$

$$= \sum_{x,y} p(x, y) \log_2 \frac{p(y|x)}{p(y)} \quad (10.7)$$

$$= - \sum_{x,y} p(x, y) \log_2 p(y) + \sum_{x,y} p(x, y) \log_2 p(y|x) \quad (10.8)$$

$$= - \sum_y p(y) \log_2 p(y) - \left(- \sum_{x,y} p(x, y) \log_2 p(y|x) \right) \quad (10.9)$$

$$= H(Y) - H(Y|X) \quad (10.10)$$

Now,

$$\sum_{i=0}^2 p_i = 1 \quad (10.11)$$

$$\sum_{j=0}^2 q_j = 1 \quad (10.12)$$

$$H(\mathbf{q}) = - \sum_{i=0}^2 q_i \log_2 q_i \quad (10.13)$$

$$= -(q_0 \log_2 q_0 + q_1 \log_2 q_1 + q_2 \log_2 q_2) \quad (10.14)$$

$$H(Y|X) = - \sum_{i=0}^2 \sum_{j=0}^2 p_i p_{Y|X}(y_j|x_i) \log_2 (p_{Y|X}(y_j|x_i)) \quad (10.15)$$

$$\begin{aligned} &= -p_0 ((1-\alpha) \log_2 (1-\alpha) + \alpha \log_2 \alpha) \\ &\quad - p_1 ((1-\alpha) \log_2 (1-\alpha) + \alpha \log_2 \alpha) \\ &\quad - p_2 ((1-\alpha) \log_2 (1-\alpha) + \alpha \log_2 \alpha) \end{aligned} \quad (10.16)$$

Using (10.14) and (10.16) in (10.10)

$$\begin{aligned} I(X, Y) &= -(q_0 \log_2 q_0 + q_1 \log_2 q_1 + q_2 \log_2 q_2) \\ &\quad + p_0 ((1-\alpha) \log_2 (1-\alpha) + \alpha \log_2 \alpha) \\ &\quad + p_1 ((1-\alpha) \log_2 (1-\alpha) + \alpha \log_2 \alpha) \\ &\quad + p_2 ((1-\alpha) \log_2 (1-\alpha) + \alpha \log_2 \alpha) \end{aligned} \quad (10.17)$$

$$\begin{aligned} \Rightarrow \frac{d}{d\alpha} I(X, Y) = p_0 \log_2 \left(\frac{\alpha}{1-\alpha} \right) + p_1 \log_2 \left(\frac{\alpha}{1-\alpha} \right) \\ + p_2 \log_2 \left(\frac{\alpha}{1-\alpha} \right) \end{aligned} \quad (10.18)$$

For Maxima or minima $\frac{d}{d\alpha} I(X, Y) = 0$

$$\log_2 \left(\frac{\alpha}{1-\alpha} \right) (p_0 + p_1 + p_2) = 0 \quad (10.19)$$

$$\Rightarrow \alpha = \frac{1}{2} \quad (10.20)$$

Chapter 11

Markov chain

11.1 Let $X_{n \geq 1}$ be a Markov chain with state space $\{ 1, 2, 3 \}$ and transition probability matrix

$$\begin{pmatrix} \frac{1}{2} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ 0 & \frac{1}{2} & \frac{1}{2} \end{pmatrix}$$

Then $\Pr(X_2 = 1 | X_1 = 1, X_3 = 2)$ equals

(GATE ST 2023)

Solution: Consider transition matrix as:

$$\begin{pmatrix} p_{11} & p_{12} & p_{13} \\ p_{21} & p_{22} & p_{23} \\ p_{31} & p_{32} & p_{33} \end{pmatrix} \tag{11.1}$$

$$\Pr(X_2 = 1 | X_1 = 1, X_3 = 2) = \Pr(X_2 = 1 | X_1 = 1) \tag{11.2}$$

$$= p_{11} \tag{11.3}$$

$$= 0.5 \tag{11.4}$$

(by markov's property and using transition probability matrix)

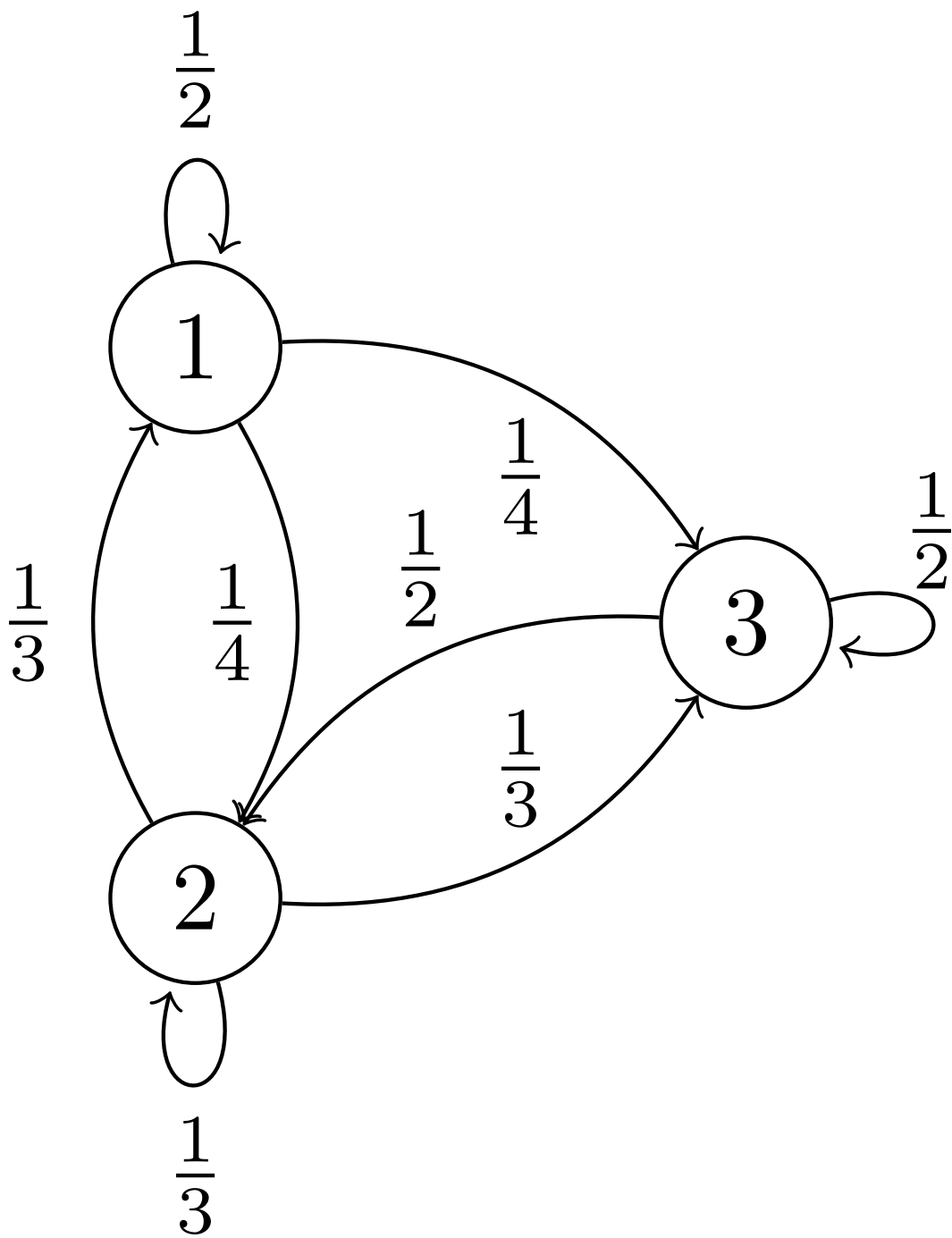


Figure 11.1: Markov Chain diagram