

Exponential Random Variable

An exponential random variable has PDF of the form

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

where λ is (ve)

$$\begin{aligned} \textcircled{1} \text{ Non (-)ve} \quad \textcircled{2} \quad \int_{-\infty}^{\infty} f_X(x) dx &= \int_0^{\infty} \lambda e^{-\lambda x} dx \\ &= [-e^{-\lambda x}]_0^{\infty} = -\left(\frac{1}{e^{\infty}} - \frac{1}{e^0}\right) \\ &= 1 \end{aligned}$$

- * An exponential random variable is used to model the amount of time until an incident of interest takes place, for example-
 - ① Breaking down of some equipment
 - ② Burning out of light bulb
 - ③ An accident occurring

If X is exponential random variable

$$P(X \geq a) = \int_a^{\infty} f_X(x) dx = \int_a^{\infty} \lambda e^{-\lambda x} dx$$

$$P(X \geq a) = e^{-\lambda a} \quad \textcircled{*}$$

$\textcircled{1}$

The mean and variance of X :-

$$\begin{aligned} E[X] &= \int_{-\infty}^{\infty} x f_X(x) dx = \int_{-\infty}^{\infty} x \lambda e^{-\lambda x} dx \\ &= \lambda \frac{\Gamma(2)}{x^2} = \frac{\Gamma(2)}{\lambda} = \frac{1}{\lambda} \end{aligned}$$

$$\begin{aligned} E[X^2] &= \int_{-\infty}^{\infty} x^2 f_X(x) dx = \lambda \int_{-\infty}^{\infty} x^2 e^{-\lambda x} dx \\ &= \lambda \left[\frac{x^3}{\lambda^3} \right] = \frac{2}{\lambda^2} \end{aligned}$$

$$\text{Var}(X) = E[X^2] - (E[X])^2 = \frac{2}{\lambda^2} - \frac{1}{\lambda^2} = \frac{1}{\lambda^2}$$

ex:- The time until a small meteorite first lands anywhere in Sahara desert is modeled as an exponential random variable with mean of 10 days. The time is currently mid-night. What is the probability that a meteorite first lands some time between 6 AM to 6 PM of the first day?

\Rightarrow Let X be the time elapsed until the

$$\begin{aligned}
 \frac{1}{\lambda} &= 10 \quad \Rightarrow \quad \lambda = \frac{1}{10} \\
 P(X_1 \leq x \leq \frac{3}{4}) &= \int_{y_4}^{\frac{3}{4}} \lambda e^{-\lambda x} dx = \frac{1}{10} \int_{y_4}^{\frac{3}{4}} e^{-\frac{1}{10}x} dx \\
 &= \frac{1}{10} \left[e^{-\frac{x}{10}} \right]_{y_4}^{\frac{3}{4}} = \left(e^{-\frac{3}{40}} - e^{-\frac{y_4}{10}} \right) \\
 P(X \geq y_4) &= 1 - P(X < y_4) = e^{-\frac{y_4}{10}} - e^{-\frac{3}{40}} \\
 &= -e^{-\frac{y_4}{10}} + e^{-\frac{3}{40}} \\
 &= -0.075 \\
 &= 0.04756 \\
 &\approx 0.0476
 \end{aligned}$$

Cumulative Distribution Function (CDF)

The CDF of a random variable X is denoted by F_X & provides the probability $P(X \leq x)$. For every x , we have

$$F_X(x) = P(X \leq x) = \begin{cases} \sum_{k \leq x} p_X(k) & \text{If } X \text{ is discrete} \\ \int_{-\infty}^x f_X(t) dt & \text{If } X \text{ is continuous} \end{cases}$$

\uparrow
 A_x

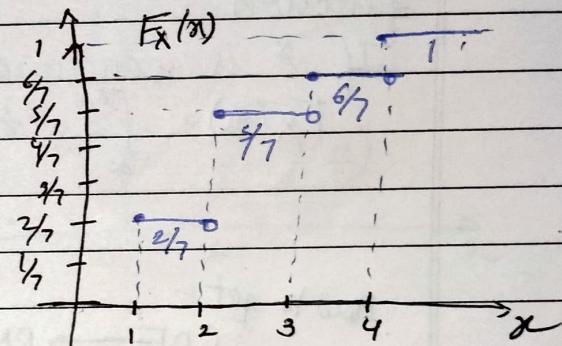
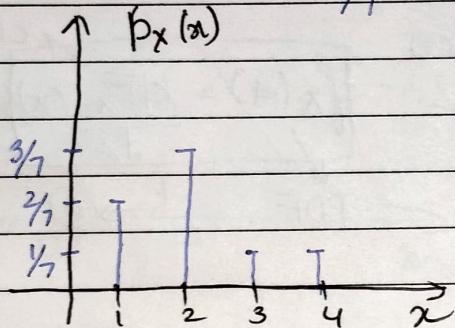
ex:- X is discrete & $x = 1, 2, \dots, 10$

$$\begin{aligned} F_X(5) &= P(X \leq 5) \\ &= p_{X(1)} + p_{X(2)} + p_{X(3)} + p_{X(4)} + p_{X(5)} \\ &= \sum_{x \leq 5} p_{X(x)} \end{aligned}$$

Let X be a discrete random variable

Is it
a PMF? $P_X(x) = \begin{cases} \frac{4}{7} & x=1 \\ \frac{3}{7} & x=2 \\ \frac{1}{7} & x=3 \\ \frac{1}{7} & x=4 \end{cases}$

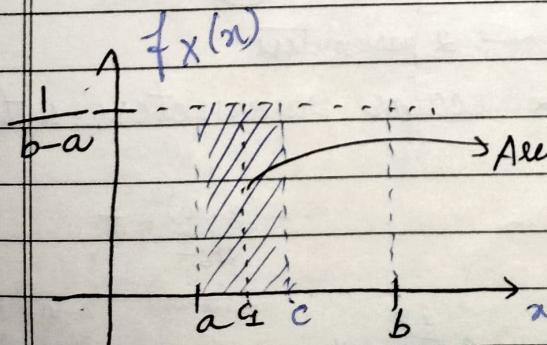
Assume
 $x = 1, 2, 3, 4$



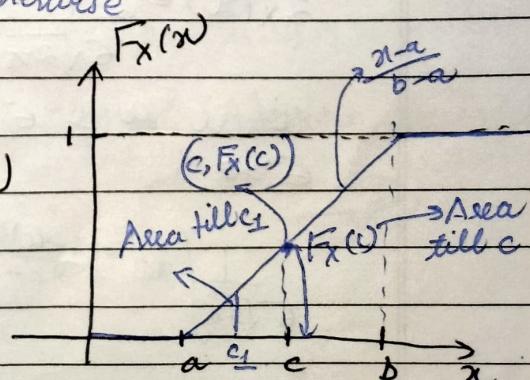
Discrete \rightarrow CDF is step function

Let X be a continuous (uniform) random variable
with PDF

$$f_X(x) = \begin{cases} \frac{1}{b-a} & a \leq x \leq b \\ 0 & \text{otherwise} \end{cases}$$



$$\text{Area} = F_X(c)$$



$E(X^2) = \int_{-\infty}^{\infty} x^2 dt$

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$$F_X(x) = \int_{-\infty}^x \frac{1}{b-a} dt = \int_a^x \frac{1}{b-a} dt = \frac{x-a}{b-a}$$

Properties of CDF:-

* Now If $x \leq y$ $F_X(x) \leq F_Y(y)$

$F_X(x)$ tends to 0 as $x \rightarrow -\infty$ and tends to 1 as $x \rightarrow \infty$.

If X is discrete, then $F_X(x)$ is piecewise constant function.

If X is continuous then $F_X(x)$ is a continuous function.

If X is continuous then

$$F_X(x) = \int_{-\infty}^x f_X(t) dt \quad \text{CDF}$$

$$f_X(x) = \frac{dF_X(x)}{dx} \quad \text{PDF}$$

How to get -
CDF \rightarrow PMF

Normal Random Variable:-

A continuous random variable X is said to be normal (or Gaussian) if it has a PDF of the form

$$f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad \text{--- (1)}$$

where μ & σ are 2 scalar parameters with $\sigma > 0$

$$\frac{1}{\sqrt{2\pi}} \int x e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

$$\frac{1}{\sqrt{2\pi}} \int (x-\mu) e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

$$\frac{2(\mu-x)}{\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$\frac{2(\mu-x)}{\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

exponential is senseless
 \hookrightarrow $v < 0$

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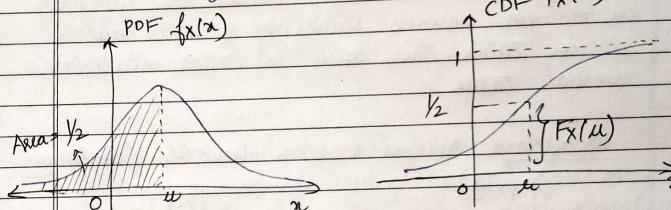
Topic :

$$\frac{1}{\sqrt{2\pi}} \int_{-\infty}^x t e^{-\frac{(t-\mu)^2}{2\sigma^2}} dt + \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} t e^{-\frac{(t-\mu)^2}{2\sigma^2}} dt$$

\Rightarrow

$$\text{The mean & variance } \mathbb{E}[X] = \int_{-\infty}^{\infty} x \cdot 1 \cdot e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx = \mu$$

$$\text{Var}[X] = \int_{-\infty}^{\infty} (x-\mu)^2 \cdot 1 \cdot e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx = \sigma^2$$



The PDF is symmetric around mean μ

Standard Normal random variable

A normal random variable with $\mu=0$ & $\sigma^2=1$ is said to be standard normal random variable (denoted by Z). Its PDF will be $f_Z(y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}}$

Area is 99.9% under $[z_1, z_2]$

Find CDF of Standard normal random variable

$$P(Y) = P(Y \leq y) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^y e^{-\frac{t^2}{2}} dt$$

Phu

(*)

Example:- We know that this integral can be evaluated for all y

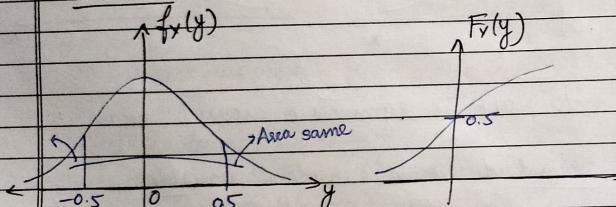
Let Y be a standard normal random variable taking values in $[1, 100]$

$$F_x(s) = P(Y \leq s) = \int_{-\infty}^s f_Y(y) dy$$

$$F_y(s) = P(Y \leq s) = \int_{-\infty}^s f_Y(y) dy$$

People have already calculated the integrals for various values of y and put it in table. This table is called standard normal table.

Standard Normal random variable (continued)



$$\text{Ex}(0.5) = P(Y \leq 0.5) = P(Y \geq 0.5) \\ = 1 - P(Y \leq 0.5)$$

$$\text{In general: } \text{Ex}(-y) = 1 - \text{Ex}(y)$$

Let X be a normal random variable with mean μ and variance σ^2 . Define another random variable

$$Y = \frac{X-\mu}{\sigma}$$

Y is a linear function of X , hence it is also a Normal random variable.
 $E[Y] = E\left[\frac{X-\mu}{\sigma}\right] = \frac{1}{\sigma}(E[X]-\mu) = \frac{1}{\sigma}(\mu-\mu) = 0$

$$V[Y] = V\left[\frac{X-\mu}{\sigma}\right] = \frac{1}{\sigma^2}(V[X] \text{ calc}) = \left(\frac{\sigma^2}{\sigma^2}\right) = 1$$

$$V[ax+b] = a^2 V[aX]$$

Thus Y is a standard normal variable.
 \Rightarrow Prob

Example:- The annual snowfall at a particular location is modeled using normal random variable with mean 60 inches and standard deviation of $\sigma = 20$. What is the probability that this year's snowfall will be atleast 80 inches?

Soln:- Let X be the snowfall. It is given by X is normal random variable with $\mu = 60$ & $\sigma = 20$. Define.

$$Y = \frac{X-60}{20} \quad (Y = \frac{X-\mu}{\sigma})$$

We know that Y is a standard
 $P(X \geq 80) = P\left(\frac{X-60}{20} \geq \frac{80-60}{20}\right)$

$$= P(Y \geq 1) = 1 - P(Y \leq 1) \\ = 1 - 0.8413 \quad \text{From Table} \\ = 0.1587$$

1. $y = x^2$

$$f(x) = \begin{cases} \frac{1}{2} e^{-|x|} & -\infty < x < \infty \\ \int \frac{1}{2} e^{-x} dx & x > 0 \\ \int \frac{1}{2} e^x dx & x < 0 \end{cases}$$

$$\begin{aligned} F_Y(y) &= P(Y \leq y) \\ &= P(X \leq \sqrt{y}) \\ &= P(-\sqrt{y} \leq X \leq \sqrt{y}) \\ &= \int_{-\sqrt{y}}^{\sqrt{y}} f(x) dx = \int_0^{\sqrt{y}} \frac{1}{2} e^{-x} dx \\ F_Y &= \left[\frac{1}{2} e^{-x} \right]_0^{\sqrt{y}} = \left[e^{-|x|} \right]_0^{\sqrt{y}} \\ &= -(e^{-\sqrt{y}} - 1) = 1 - e^{-\sqrt{y}} \end{aligned}$$

2. $y = \frac{2}{x} + 3$

$$f(x) = \begin{cases} x^2(2x+3)^{-2} & 0 < x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned} \text{Var}(Y) &= \text{Var}\left(\frac{2}{x} + 3\right) = \text{Var}\left(\frac{2}{x}\right) \\ &= 4 \text{Var}\left(\frac{1}{x}\right) = 4 \left(E\left(\frac{1}{x}\right)^2 - \left(E\left(\frac{1}{x}\right)\right)^2 \right) \\ E\left(\frac{1}{x}\right) &= \frac{1}{2} \int_0^1 x^{-2} \left(2x+3\right)^{-2} dx = \int_0^1 \frac{2x^2 + 3x}{2} dx \\ &= \frac{2}{3} \left(x^3\right)_0^1 + \frac{3}{2} \left(x^2\right)_0^1 \\ &= \frac{2}{3}(1) + \frac{3}{4}(1) = \frac{8+9}{12} = \frac{17}{12} \end{aligned}$$

$$E\left(\frac{1}{x^2}\right) =$$

5. (a) $P(X \leq a) = P(X > a)$

$$\int_{-\infty}^a 3x^2 dx = 1 - P(X \leq a) = 1 - \int_{-\infty}^a 3x^2 dx$$

$$\Rightarrow \int_{-\infty}^a x^2 dx = 1 \Rightarrow 6 \int_0^a x^2 dx = 1$$

$$\Rightarrow 2[x^3]_0^a = 1 \Rightarrow 2[a^3] = 1 \Rightarrow a^3 = \frac{1}{2}$$

$$\Rightarrow a = (1/2)^{1/3}$$

(b) $P(X \geq b) = 0.05$

$$\Rightarrow 1 - P(X \leq b) = \frac{5}{100}$$

$$\Rightarrow 1 - b^3 = \frac{5}{100} \Rightarrow b^3 = \frac{95}{100} \Rightarrow b = \left(\frac{95}{100}\right)^{1/3}$$

6. (A) $P(0 \leq x \leq 2)$

$$f_{xy}(x, y) \quad \text{Marginal Density Function of } X$$

$$f_x(x) = \int_y f_{xy}(x, y) dy \quad \text{Joint PDF}$$

$$f_{xy}(y) = \int_x f_{xy}(x, y) dx \quad \text{Marginal Density Function of } Y$$

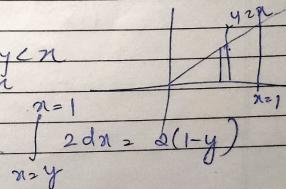
Conditional Density Function :-

$$f_{Y|X}(y|x) = \frac{f_{xy}(x, y)}{f_x(x)} \quad x \text{ given } X$$

6. (a) $f(x, y) = 2 \quad 0 < x < 1 \quad 0 < y < x$

$$(a) f_x(x) = \int_0^x 2 dy = 2x$$

$$(c) f_{Y|X}(y|x) = \frac{f_{xy}(x, y)}{f_x(x)} = \frac{2}{2x} = \frac{1}{x} \quad x = 1$$



(b) $E(XY) = E(X)E(Y)$ X & Y are independent random variables

$$\begin{aligned} f_{Y/X}(x,y) &= \frac{f_{XY}(xy)}{f_X(x)} = \frac{f_{XX}(xy)}{f_X(x)f_{YY}(y)} \\ &= \frac{2}{2x} = \frac{1}{x} \end{aligned}$$

$$f_{XX}(xy) = \frac{f_{XY}(xy)}{f_Y(y)} = \frac{2}{2(1-y)} = \frac{1}{1-y}$$

$$1. f(x) = 1 \quad 0 < x < 1 \\ Y = \frac{x}{1+x}$$

Let X be a continuous random variable with PDF $f_X(x)$ and we have given transformation

$$Y = \sqrt{X}$$

i) If y is continuously differentiable

ii) If y is either non-decreasing or non-increasing for all values of x for which $f_X(x) \neq 0$.

$$g_Y(y) = f_X(x/y) \left| \frac{dx}{dy} \right| \text{ is PDF of } Y$$

$$Y = \frac{x}{1+x} \Rightarrow x = \frac{y}{1-y} \Rightarrow x(1+y) = y \Rightarrow yx + y = y \Rightarrow y(x+1) = y \Rightarrow x = \frac{y}{1-y}$$

X is random variable

$$f(x) = \begin{cases} 1, & 0 < x < 1 \\ 0, & \text{otherwise} \end{cases}$$

$$X = \frac{x}{1+x} \quad Y' = \frac{(1+x)-x}{(1+x)^2} = \frac{1}{(1+x)^2} \geq 0 \quad \text{increasing}$$

$$\begin{aligned} g_Y(y) &= f_X(x(y)) \left| \frac{dx}{dy} \right| \quad X = \frac{y}{1-y} \\ &= f_X\left(\frac{y}{1-y}\right) \cdot \frac{1}{(1-y)^2} \\ &= \frac{1}{(1-y)^2} \end{aligned}$$

Independent

$$8. 20\% \text{ chance of receiving an offer from each company} \\ P(\text{at least one offer}) = 1 - P(\text{no offer}) \\ = 1 - {}^4C_0 \left(\frac{1}{5}\right)^0 \left(\frac{4}{5}\right)^4 \\ = 1 - \left(\frac{4}{5}\right)^4 =$$

No :-

$$9. \mu = 60 \text{ inches} \quad \sigma = 1 \text{ inch} \rightarrow P(|x-\mu| \leq k\sigma) \\ \text{Chebyshev Inequality} \quad 1 - \frac{1}{k^2} \\ P(4' 10'' \leq x \leq 5' 2'') \quad P(58 \leq x \leq 62) \geq A$$

$$\Rightarrow P(58-60 \leq x-60 \leq 62-60) \geq A \\ \Rightarrow P(-2 \leq x-60 \leq 2) \geq A \\ \Rightarrow P(|x-60| \leq 2) \geq 1 - \frac{1}{2^2}$$

$$\frac{\mu \sigma}{2} = \frac{60 \cdot 1}{2} = \frac{1}{2^2} \\ k = 2 \quad 1 - \frac{1}{4} = \frac{3}{4} \\ \text{At least } 75\%$$

Joint PDFs of multiple random variables:-

Let X and Y be continuous random variables associated with same experiment. Then X & Y are jointly continuous if there is a non(-)ve function $f_{X,Y}(x,y)$ such that

$$P(a \leq X \leq b, c \leq Y \leq d) = \int_a^b \int_c^d f_{X,Y}(x,y) dx dy$$

Function $f_{X,Y}$ satisfies following property:-

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx dy = 1$$

This $f_{X,Y}(x,y)$ is joint PDF of X & Y .

Joint CDFs:-

Let X and Y be 2 random variables associated with same experiment, their joint

We have

$$\frac{\partial^2 F_{X,Y}(x,y)}{\partial y \partial x} = f_{X,Y}(x,y)$$

The marginal PDFs of X & Y can be obtained from the joint PDF as follows:-

$$f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dy$$

$$f_Y(y) = \int_{-\infty}^{\infty} f_{X,Y}(x,y) dx$$

Expectation:-

$$E[ax+by+cz] = aE[X]+bE[Y]+cE[Z]$$

Conditioning one random variable

Ex:- The speed of a typical vehicle that drives past a police radar is modeled as an exponential random variable X with mean 50 miles/hr. The police radar's measurement Y of the vehicle's speed has an error which is modeled as a normal random variable with zero mean & standard deviation equal to $1/10$ th of the vehicle's speed. What is the joint PDF of X & Y .

$$\Rightarrow \frac{1}{\lambda} = 50 \Rightarrow \lambda = \frac{1}{50} \quad u=0 \quad S = \frac{x}{10} \quad S = \frac{P}{T}$$

$$\text{variance } \sigma^2 = \frac{\alpha^2}{100} \quad \text{Given that } X=x \quad \left(\frac{(y-x)^2}{2(\frac{x}{10})^2} \right)$$

$$f_{Y|X}(y/x) = \frac{f_{X,Y}(x,y)}{f_X(x)} = \frac{1}{\sqrt{2\pi}} e^{-\frac{(y-x)^2}{2x}}$$

Independence:- 2 continuous random variables X and Y are independent if their joint PDF is the product of the marginal PDFs.

$$f_{X,Y}(x,y) = f_X(x)f_Y(y) \quad \text{for all } x, y$$

Note that,

$$f_{X|Y}(x|y) = \frac{f_{X,Y}(x,y)}{f_Y(y)}$$

Topic :

If random variables are independent

$$f_{X,Y}(x,y) = f_X(x) \quad (2)$$

for all y with $f_X(y) > 0$
and all x

In general, if we have 3 random variables
 x, y and z
then $f_{X,Y,Z}(x,y,z) = f_X(x)f_Y(y)f_Z(z)$

ex - The waiting time in hours between successive
speeders spotted by a radar is continuous
random variable with CDF

$$P(X < x) = F(x) = \begin{cases} 1 - e^{-8x} & x \geq 0 \\ 0 & x < 0 \end{cases}$$

Find the probability of waiting less than 12 min
between successive speeders.

$$\begin{aligned} P(X < 12) &= F(0.2) = 1 - e^{-1.6} \\ f &= \int_0^\infty 8e^{-8x} dx \\ &= \frac{8}{8} (e^{-8x})_0^{0.2} = -1(e^{-1.6} - 1) \\ &= 1 - e^{-1.6} \end{aligned}$$

$$Q- (1) \quad f(x) = \begin{cases} x & 0 \leq x \leq 1 \\ 2-x & 1 \leq x \leq 2 \\ 0 & \text{otherwise} \end{cases}$$

$$F(x) = \int_{-\infty}^x f(a) da = \int_0^x adx = \frac{x^2}{2} \quad 0 \leq x \leq 1$$

$$F(x) = \int_{-\infty}^x f(a) da = \int_0^1 adx + \int_1^x (2-x) dx = \left[\frac{x^2}{2} \right]_0^1 \left[2x - \frac{x^2}{2} \right]_1^x$$

$$= \frac{1}{2} + \left(2x - \frac{x^2}{2} - (2 - \frac{1}{2}) \right)$$

Topic :

$$= \frac{1}{2} + 2x - \frac{x^2}{2} - \frac{3}{2} = \frac{2x - x^2 - 1}{2}$$

$$F(x) = \int_{-\infty}^x f(a) da = \int_0^x \frac{1}{2} dx + \int_x^\infty (2-x) dx + \int_2^\infty f(a) da$$

$$= \frac{1}{2}x - \frac{x^2}{2} \quad \frac{1}{2} + (2x - \frac{x^2}{2})^2$$

$$= \frac{1}{2} + (4 - 2 - (2 - \frac{1}{2})) = \frac{1}{2} + (2 - 2 + \frac{1}{2}) = 1$$

$$F(x) = \begin{cases} \frac{x^2}{2} & 0 \leq x < 1 \\ 2x - \frac{x^2}{2} - 1 & 1 \leq x < 2 \\ 1 & x \geq 2 \end{cases} \quad \begin{matrix} F(x) & \leq 1 & 1 \leq x \leq 2 \\ & & x \geq 2 \end{matrix}$$

 $x \geq 2$ $6-5-4$ $6-5+6$

$$Q- (2) \quad F(x) = \begin{cases} 0 & x < 0 \\ \frac{1}{2} & 0 \leq x < 1 \\ \frac{4}{5} & 1 \leq x < 2 \\ \frac{4}{5} & 2 \leq x < 3 \\ \frac{9}{10} & 3 \leq x < 3.5 \\ 1 & x \geq 3.5 \end{cases} \quad \begin{matrix} P(x) = 0 & x < 0 \\ \frac{1}{2} & 0 \leq x < 1 \\ \frac{4}{5} & 1 \leq x < 2 \\ \frac{4}{5} & 2 \leq x < 3 \\ \frac{9}{10} & 3 \leq x < 3.5 \\ \frac{1}{10} & x \geq 3.5 \end{matrix}$$

Q- (3) The Department of Energy puts projects out on bid & generally estimates what a reasonable bid should be. We call the estimate 'b'. The Department of energy has determined that the density function of winning bid is

$$f(y) = \begin{cases} \frac{3}{8}b & 2b/5 < y \leq 2b \\ 0 & \text{otherwise} \end{cases}$$

Find $F(y)$? What is the probability that winning bid is less than estimate 'b'?

Topic :

$$F(y) = \int_{-\infty}^y \frac{5}{8b} dx = \frac{5}{8b} \log \frac{b}{|x|} = \frac{5}{8b} (x - 2b) = \frac{5x - 1/4}{8b} \quad \begin{matrix} 2b \leq y < 2b \\ 1 \end{matrix}$$

$$= 1 \quad y \geq 2b$$

$$P(y \leq b) = \frac{5x}{8b} - \frac{1}{4} = \frac{5}{8} - \frac{1}{4} = \frac{3}{8}$$

$$\int_{-\infty}^b \frac{5}{8b} dx = \frac{5}{8b} [b - 2b] = \frac{5}{8b} \left(\frac{3b}{2} \right) = \frac{15}{16}$$

Q1 The weekly demand for a certain drink, in thousands of litres, at a chain of convenience stores is a continuous random variable $g(x) = x^2 + x - 2$, where x has the following PDF:-

$$f_x(x) = \begin{cases} 2(n-1) & 1 < n < 2 \\ 0 & \text{otherwise} \end{cases}$$

Find the expected value of the weekly demand for the drink.

$$\begin{aligned} E(g(x)) &= E[x^2 + x - 2] \\ &= E[x^2] + E[x] - 2 \\ &= \int_1^2 x^2 \cdot 2(n-1) dx + \int_1^2 n(n-1) dx - 2 \\ &= 2 \left(\frac{x^4}{4} - \frac{x^3}{3} \right)_1^2 + 2 \left(\frac{x^3}{3} - \frac{x^2}{2} \right)_1^2 - 2 \\ &= 2 \left(4 - \frac{8}{3} - \frac{1}{4} + \frac{1}{3} \right) + 2 \left(\frac{8}{3} - 2 - \frac{1}{3} + \frac{1}{2} \right) - 2 \\ &= 8 - \frac{16}{3} - \frac{1}{2} + \frac{2}{3} + \frac{16}{3} - 4 - \frac{1}{3} + 1 - 2 \\ &= 4 - \frac{9}{2} = \frac{5}{2} = 2.5 \end{aligned}$$

Topic :

 $X = 1, 2, 3$

$$P(Y \leq y) = P(X^2 \leq y) = P(-\sqrt{y} \leq X \leq \sqrt{y})$$

$$= \int_{-\sqrt{y}}^{\sqrt{y}} f_X(x) dx$$

Derived Distribution :- Calculating the PDF of $Y = g(X)$ where X is a continuous random variable:-

ex:- Let X be uniform on $[0, 1]$ & let $Y = \sqrt{X}$. Find PDF of Y .

Soln :- PDF of X

$$f_X(x) = \begin{cases} 1 & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

The range of Y will be $[0, 1]$

First we will find CDF then PDF

$$\begin{aligned} F_Y(y) &= P(Y \leq y) \\ &= P(\sqrt{X} \leq y) \\ &= P(X \leq y^2) \\ &= \int_0^{y^2} f_X(x) dx = \int_0^{y^2} 1 dx = y^2 \end{aligned}$$

$$F_Y(y) = y^2$$

$$\text{PDF of } Y \vdash f_Y(y) = \frac{dF_Y(y)}{dy} = \frac{d(y^2)}{dy} = 2y$$

$$\int_{-\sqrt{y}}^{\sqrt{y}} 1 dx = \int_0^{\sqrt{y}} 1 dx = \sqrt{y} \quad f_Y(y) = \int_0^{\sqrt{y}} 0 \leq y \leq \sqrt{y} \quad \text{otherwise}$$

$$\text{PDF of } Y \vdash \frac{d\sqrt{y}}{dy} = \frac{1}{2\sqrt{y}} = \frac{1}{2\sqrt{g}}$$

Covariance & Correlation :- We will study a quantitative measure of the strength and direction of the relationship b/w 2 random variables.

Let X and Y be 2 random variables. The covariance of X and Y denoted by $\text{Cov}(X, Y)$ is defined as:-

$$\text{Cov}(X, Y) = E[(X - E(X))(Y - E(Y))]$$

A simpler formula is:-

$$\text{Cov}(X, Y) = E[XY] - E[X]E[Y] \quad \textcircled{3}$$

$$\begin{aligned} E[XY - XE[Y] - YE[X] + E[X]E[Y]] \\ E[XY] - E[Y]E[X] - E[X]E[Y] + E[X]E[Y] \\ = E[XY] - E[X]E[Y] \end{aligned}$$

* If X & Y are independent then
 $\text{Cov}(X, Y) = 0$, we say that X & Y are uncorrelated

Reverse not true
 (think example)

The Correlation coefficient $\rho(X, Y)$ of 2 random variables X & Y that have non-zero variances is defined as:-

$$\rho(X, Y) = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X)\text{Var}(Y)}} \quad \textcircled{4}$$

$$\rho(X, Y) = \frac{-1 \leq \rho \leq 1}{\sqrt{\int_{-\infty}^{\infty} f_{X,Y}(x,y) dx dy - \int_{-\infty}^{\infty} f_X(x) dx \int_{-\infty}^{\infty} f_Y(y) dy}}$$

$$\int_{-\infty}^{\infty} x^2 \sin x$$

$$\int_{-\infty}^{\infty} e^{-x} \sin x$$

$$E[XY] - E[X]E[Y]$$

$$\sqrt{E[X - E[X]]^2 E[Y - E[Y]]^2}$$

Moment Generating Function:-

Let X be a random variable, then n th-order moment of X is defined as $E[X^n]$.

The moment generating function (MGF) associated with random variable X is defined as:-

$$M_X(s) = E[e^{sX}] \quad \textcircled{1}$$

where s is a scalar parameter

For a discrete random variable X

$$M(s) = \sum_n e^{sx} p_X(n) \quad \textcircled{2}$$

For a continuous random variable X

$$M(s) = \int_{-\infty}^{\infty} e^{sx} f_X(x) dx$$

or MGF of Poisson Random Variable

$$\begin{aligned} M(s) &= \int_{-\infty}^{\infty} e^{sx} e^{-\lambda} \frac{\lambda^x}{x!} dx = e^{-\lambda} \sum_{x=0}^{\infty} \frac{e^{sx} \lambda^x}{x!} \\ &= e^{-\lambda} \left[1 + e^s + e^{2s} + e^{3s} + \dots \right] \end{aligned}$$

$$1 + e^s + \frac{(e^s \lambda)^2}{2!} = e^s e^s$$

$$e^{s+} = e^{-\lambda} \sum_{n=0}^{\infty} \frac{e^{s+\lambda} \lambda^n}{n!} = e^{\lambda} e^{s+\lambda}$$

$$c = e^{\lambda} e^{s+\lambda} = e^{\lambda(e^s - 1)}$$

MGF of exponential random variable

$$M(s) = \int_{-\infty}^{\infty} e^{sx} \lambda e^{-\lambda x} dx = \left[\lambda e^{(s-\lambda)x} \right]_{-\infty}^{\infty}$$

$$= \lambda \left[\frac{e^0 - e^{-\lambda}}{s-\lambda} \right] = \frac{-\lambda^0}{s-\lambda} \quad s < 1$$

$$M(s) = \frac{x^0}{\lambda-s}$$

From MGF to moments:-

Let X be a continuous random variable
Then MGF of X is defined as
 $M(s) = \int_{-\infty}^{\infty} e^{sx} f(x) dx$

Differentiate both side w.r.t s

$$\frac{d}{ds} M(s) = \frac{d}{ds} \int_{-\infty}^{\infty} e^{sx} f(x) dx \quad \text{when } \int \frac{d}{ds} \frac{d}{ds}$$

$$= \int_{-\infty}^{\infty} \frac{d}{ds} e^{sx} f(x) dx \quad \text{can be interchanged}$$

$$\frac{d}{ds} M(s) = \int_{-\infty}^{\infty} x e^{sx} f(x) dx$$

Consider the special case when $s=0$ we get

$$\frac{d}{ds} M(s) \Big|_{s=0} = \int_{-\infty}^{\infty} x f(x) dx = E[X]$$

In general, we have:

$$\frac{d^n}{ds^n} M(s) \Big|_{s=0} = \int_{-\infty}^{\infty} x^n f(x) dx = E[X^n]$$

Random variable $X \xrightarrow{(1)} \text{MGF of } X$

n th moment

$$E[X^n]$$

Differentiate and put

$$s=0$$

$$E[X] \stackrel{(0.5)}{=} p(X \geq 1) = 0.7$$

example:- Let X be a random variable

$$p(X=n) = \begin{cases} \frac{1}{2} & n=2 \\ \frac{1}{6} & n=3 \\ \frac{1}{3} & n=5 \end{cases}$$

$$E[X] = 1 + \frac{1}{2} + \frac{5}{3} = \frac{6+3+10}{6} = \frac{19}{6}$$

$$M(s) = \int_{-\infty}^{\infty} e^{sx} f(x) dx$$

$$= \int_{-\infty}^{\infty} \frac{e^{2s}}{2} + \frac{e^{3s}}{6} + \frac{e^{5s}}{3}$$

$$\frac{d}{ds} M(s) \Big|_{s=0} = \frac{e^{2s}}{2} + \frac{e^{3s}}{2} + \frac{5e^{5s}}{3}$$

$$= 1 + \frac{1}{2} + \frac{5}{3} = \frac{19}{6}$$

$$\frac{d}{ds} M(s) \Big|_{s=0} = \lambda \left(\frac{1}{\lambda-s} \right)^2 \Big|_{s=0} = \frac{1}{\lambda}$$

$$\boxed{M_X(0) = 1}$$

Markov Inequality :- If a random variable X can take only non-negative values, then
 $P(X \geq a) \leq \frac{E[X]}{a}$ for all $a > 0$

Not for Normal random variable as it is -ve

Chelyshhev Inequality :-

If X is a random variable with mean μ & variance σ^2 , then
 $P(|X-\mu| \geq c) \leq \frac{\sigma^2}{c^2} \quad \forall c > 0$

Ex:- Let X be uniform random variable in the interval $[0, 4]$

Topic :

$$f(x)$$

$$E[X] = 2 = \mu \quad \text{Var} = \frac{1}{3} = \sigma^2$$

$$P(|X-2| \geq 1) \leq \frac{16}{9}$$

$$P((x-2) \leq -1 \text{ and } (x-2) \geq 1)$$

$$P(X \geq 3) = \int_3^4 \frac{1}{4} dx = \frac{1}{4} = 0.25$$

$$P(X-2) \geq 1 \\ P(X \leq 1) = \int_0^1 \frac{1}{4} dx = 1/4 \\ = 0.25$$

$$P(X \geq 3) \text{ and } P(X \leq 1) = 0.5$$

$$C=2$$

$$P(|X-2| \geq 2) = P((x-2) \leq -2 \text{ and } (x-2) \geq 2) \\ = P(X \leq 0 \text{ and } X \geq 4) \\ = 0$$

The Weak law of large numbers :-

If X_1, X_2, \dots be independent identically distributed random variables with mean μ

For every $\epsilon > 0$, we have

$$P\left(\left|\frac{X_1 + X_2 + \dots + X_n}{n} - \mu\right| \geq \epsilon\right) \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

are independent R.V.

$$10^{-3}, 10^{-2}, 10^{-1}$$

$$P(|y-\mu| \geq 10^{-3}) = P(y-\mu \leq -10^{-3} \text{ or } y-\mu \geq 10^{-3}) \\ = P(y \leq -10^{-3} + \mu \text{ or } y \geq 10^{-3} + \mu)$$

Topic :

Explanation:- Let X_1, X_2, \dots have mean μ and variance σ^2 . Define a new random variable

$$M_n = \frac{X_1 + X_2 + \dots + X_n}{n} \quad \text{--- (2)}$$

$$E[M_n] = \frac{1}{n} [E[X_1] + E[X_2] + \dots + E[X_n]]$$

$$= \frac{1}{n} [\mu + \mu + \dots + \mu] = \frac{1}{n} (n\mu) = \mu$$

$$\text{Var}(M_n) = \frac{1}{n^2} [\text{Var}(X_1) + \text{Var}(X_2) + \dots + \text{Var}(X_n)] \\ = \frac{1}{n^2} (\sigma^2 + \sigma^2 + \dots + \sigma^2) = \frac{n\sigma^2 - \sigma^2}{n^2} = \frac{\sigma^2}{n}$$

$$P(|M_n - \mu| \geq \epsilon) \leq \frac{n\sigma^2}{n\epsilon^2} = \frac{\sigma^2}{\epsilon^2} \rightarrow 0 \quad n \rightarrow \infty$$

The Central Limit Theorem (CLT)

$$\text{Cov}(X, Y)$$

Derived Distributions:-

Vectors \rightarrow IID

(Q): We load

Let X_1, X_2, \dots be a sequence of independent identically distributed random variables with common mean μ & variance σ^2 . Define

$$Z_n = \frac{X_1 + X_2 + \dots + X_n - n\mu}{\sigma\sqrt{n}}$$

Then CDF of Z_n converges to standard normal CDF.

$$E[Z_n] = \frac{1}{\sqrt{n}} [E[X_1 + \dots + X_n] - n\mu]$$

Q: We load on a plane 100 packages whose weights are independent random variables that are uniformly distributed b/w 5 & 50 kg. What is the probability that the total wt exceed 3000kg?

$$P(S_{100} > 3000) = ?$$

S_{100} = sum of the weight of 100 packages

$$\Rightarrow p = \frac{1}{50-5} = \frac{1}{45} \quad E = \frac{a+b}{2} = \frac{50+5}{2} = 27.5$$

$$E[X_i] = \mu = 27.5$$

$$Var(X_i) = \frac{(50-5)^2}{12} = \frac{45 \times 45}{12} = 168.75 = \sigma^2$$

$$Z = \frac{X_1 + X_2 + \dots + X_{100} - 100 \times 27.5}{\sqrt{168.75}} = \frac{Y}{\sigma}$$

$$Z = \frac{S_{100} - 100 \times 27.5}{\sqrt{168.75}}$$

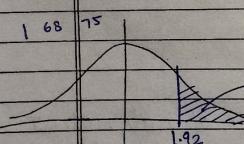
$$P(S_{100} > 3000) = P\left(\frac{S_{100} - 100 \times 27.5}{\sqrt{168.75}} > \frac{3000 - 100 \times 27.5}{\sqrt{168.75}}\right)$$

$$= P\left(Z > \frac{250}{\sqrt{168.75}}\right)$$

$$= P\left(Z > 1.92\right)$$

$$= 0.5 - 0.4726$$

$$= 0.0274$$



Q: A certain type of storage battery lasts on average, 5 yrs with a standard deviation of 0.5 yrs.

Assuming that battery life is normally distributed, find the probability that a given battery

will last less than 2.3 yrs.

$$G = Y_2 \quad \mu = 3$$

$$Z = \frac{X - \mu}{\sigma} = \frac{X - 3}{0.5}$$

$$P(Z < 2.3) = P(X < 2.3) = 0.6 - 0.$$

$$Z = \frac{X - \mu}{\sigma} = \frac{X - 3}{0.5} = \frac{X - 2.3}{0.5}$$

$$P(Z < 2.3) = P\left(\frac{X - 3}{0.5} < 2.3\right) = P\left(\frac{X - 3}{0.5} < \frac{2.3 - 3}{0.5}\right) = P(Z < -1.4)$$

$$P(Z < -1.4)$$

$$= 1 - P(Z < 1.4)$$

$$P(X < 2.3) = 0.0808$$

Multidimensional Random Variable

To develop a vector/matrix notation that will allow us to represent large sequence of random variables with a compact notation.

Joint, Marginal & conditional PMFs, CDFs & PDFs

Defn - For a set of N random variables X_1, X_2, \dots, X_N , the joint PMF, CDF & PDF are given as

$$P_{X_1, X_2, \dots, X_N}(x_{1N}, x_{2N}, \dots, x_{NN}) = P(X_1 = x_{1N}, X_2 = x_{2N}, \dots, X_N = x_{NN})$$

$$F_{X_1, X_2, \dots, X_N}(x_{1N}, x_{2N}, \dots, x_{NN}) = P(X_1 \leq x_{1N}, X_2 \leq x_{2N}, \dots, X_N \leq x_{NN})$$

$$f_{X_1, X_2, \dots, X_N}(x_{1N}, x_{2N}, \dots, x_{NN}) = \frac{\partial^n}{\partial x_{1N} \partial x_{2N} \dots \partial x_{NN}} F_{X_1, X_2, \dots, X_N}(x_{1N}, x_{2N}, \dots, x_{NN})$$

Above notation seems cumbersome for large N .

2 → 1 step Integrate
10 → 9 steps Integrate
Topic :

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Introduce:- Let $X = [x_1, x_2, \dots, x_N]^T$ be a column vector with N random variables & similarly $x = [x_1, x_2, \dots, x_N]$. Then PMF, CDF & PDF are expressed as $P_X(a)$, $F_X(a)$ & $f_X(a)$

$$P_{X_1, X_2, \dots, X_N}(x_1, x_2, \dots, x_N) = \sum_{n_{m+1}, n_{m+2}, \dots, n_N} \dots$$

$$P_{X_1, X_2, \dots, X_N}(x_1, x_2, \dots, x_N)$$

$$F_{X_1, X_2, \dots, X_N}(x_1, x_2, \dots, x_N) = F_{X_1, X_2, \dots, X_N}(x_1, x_2, \dots, x_M, \infty, \dots, \infty)$$

$$f_{X_1, X_2, \dots, X_N}(x_1, x_2, \dots, x_N) = \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} f_{X_1, X_2, \dots, X_N}(x_1, x_2, \dots, x_N) d\pi_{N+1}, d\pi_{N+2}, \dots, d\pi_N$$

Conditional PMF & PDF

$$P_{X_1, X_2, \dots, X_N | X_{M+1}, X_{M+2}, \dots, X_N}(x_1, x_2, \dots, x_M | x_{M+1}, x_{M+2}, \dots, x_N) = \frac{P(X_1 = x_1, X_2 = x_2, \dots, X_N = x_N)}{P(X_{M+1} = x_{M+1}, X_{M+2} = x_{M+2}, \dots, X_N = x_N)}$$

Expectation Correlation / Covariance Matrix:-

Defn:- For a random vector $X = [x_1, x_2, \dots, x_N]^T$ the correlation matrix is defined as $R_{XX} = E[X X^T]$ i.e. the (i, j) th element of $N \times N$ matrix R_{XX} is $E[X_i X_j]$. Similarly, the covariance matrix

C_{XX} is defined as $C_{XX} = E[(X - \mu)(X - \mu)^T]$ i.e. the (i, j) th element of C_{XX} is $\text{Cov}(x_i, x_j)$

$$\mu = E[x_1] E[x_2]$$

Topic : Date :
S. Dhamrait
V.G. Kulkarni
Stochastic Process
Random Page No. :

Theorem:- The matrices R_{XX} & C_{XX} are symmetric & positive definite.

A matrix $Z_{n \times n}$ +ve definite $\Leftrightarrow n \geq n > 0 \forall$

$n \in \mathbb{R}^n \setminus \{0\}$
(which is non-zero)

Proof:-

Gaussian Random Variable in multiple dimensions
(why only Gaussian not others?)

Stochastic Process: → collection of random variables

Consider a system that evolves randomly in time. For example:-

Stock market index

The inventory in warehouse

Queue of customers at a service station

Water level in a reservoir

State of the machines

These systems may be observed at discrete time points $n = 0, 1, \dots$

These systems may be observed continuously in time $X(t) =$

That is - every second, every minute, every hour, every day, every week

X_n = the state of system at time n

or X_n = NSE index at the end of n th day

= # of unsold car on a dealer's lot at the starting of n th day

$X(t)$ = no. of failed machines in a machine stop at the time t
= the amt. of money in a bank account at time t'

2 → Integrate
10 → Integrate
Topic :
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Introduce:- Let $\mathbf{x} = [x_1, x_2, \dots, x_N]^T$ be a column vector with N random variables & similarly $\mathbf{z} = [z_1, z_2, \dots, z_N]$. Then PMF, CDF & PDF are expressed as $P_{\mathbf{x}}(\mathbf{z})$, $F_{\mathbf{x}}(\mathbf{z})$ & $f_{\mathbf{x}}(\mathbf{z})$

$$P_{x_1, x_2, \dots, x_N}(x_1, x_2, \dots, x_N) = \frac{\leq \leq \dots \leq}{\leq \leq \dots \leq} \frac{\leq}{\leq \leq \dots \leq} \frac{\leq}{\leq \leq \dots \leq}$$

$$P_{x_1, x_2, \dots, x_N}(x_1, x_2, \dots, x_N)$$

$$F_{x_1, x_2, \dots, x_N}(x_1, x_2, \dots, x_N) = F_{x_1, x_2, \dots, x_N}(x_1, x_2, \dots, x_M, \infty, \infty)$$

$$f_{x_1, x_2, \dots, x_N}(x_1, x_2, \dots, x_N) = \int \int \dots \int f_{x_1, x_2, \dots, x_N}(x_1, x_2, \dots, x_N) d\pi_{N+1}, d\pi_{N+2}, \dots, d\pi_N$$

Conditional PMF & PDF

$$P_{x_1, x_2, \dots, x_N | X_{M+1}, X_{M+2}, \dots, X_N}(x_1, x_2, \dots, x_M | x_{M+1}, x_{M+2}, \dots, x_N) = \frac{P(x_1 = x_{1N}, x_2 = x_{2N}, \dots, x_M = x_{MN})}{P(x_{M+1} = x_{(M+1)N}, x_{M+2} = x_{(M+2)N}, \dots)}$$

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Defn:- For a random vector $\mathbf{x} = [x_1, x_2, \dots, x_N]^T$ the correlation matrix is defined as $R_{xx} = E[\mathbf{x}\mathbf{x}^T]$ i.e. the (i, j) th element of $N \times N$ matrix R_{xx} is $E[x_i x_j]$. Similarly, the covariance matrix

C_{xx} is defined as $C_{xx} = E[(\mathbf{x} - \mu)(\mathbf{x} - \mu)^T]$ i.e. the (i, j) th element of C_{xx} is $\text{Cov}(x_i, x_j)$

Proof:-

$\mu = E[x_1] E[x_2]$
Topic :

S. Dhanush
V.G. Kulkarni
Stochastic Process Date :
Random Page No. :

Theorem:- The matrices R_{xx} & C_{xx} are symmetric & positive definite.

A matrix $\mathbb{Z}_{n \times n}$ positive definite $\Leftrightarrow \mathbf{z}^T \mathbf{z} \geq 0 \forall \mathbf{z} \in \mathbb{R}^n$
 \Leftrightarrow which is non-zero

Proof:-

Gaussian Random Variable in multiple dimensions
(why only Gaussian not others?)

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Consider a system that evolves randomly in time. For example:-

- Stock market index
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- Water level in a reservoir
- State of the machines

These systems may be observed at discrete time points $n = 0, 1, \dots$

That is - every second, every minute, every hour, every day, every week

X_n = the state of system at time n .

or X_n = NSE index at the end of n th day

= # of unsold car on a dealer's lot at the starting of n th day

These systems may be observed continuously in time $X(t) =$

$X(t)$ = no. of failed machines in a machine

stop at the time t'

= the amt. of money

in a bank account at time t'

$$\begin{array}{l} X_1 = 100 \\ X_2 = 97 \\ X_3 = 101 \\ X_4 = 101 \end{array}$$

Like Natural
Number

Date : Like Real
Page No. : Number

We say that $\{X_n, n \geq 0\}$ is a discrete-time stochastic process describing the system.

We say $\{X(t), t \geq 0\}$ is a continuous time stochastic process.

* A stochastic process is a collection of random variables $\{X(t), t \in T\}$ indexed by the parameter t taking values in the parameter space T (which is usually subset of \mathbb{R})

* The random variables takes values in the set S , called the state-space of the stochastic process.

Following 4 combinations are possible :-

→ Cricket score

1. Discrete-state, discrete-time

a) The no. of individuals in a country at the end of year $t \rightarrow$ stochastic process.
 $\{X_t; t \in T\}$ with $T = \{0, 1, 2, \dots\}$ & $S = \{0, 1, 2, \dots\}$

b) A motor insurance company reviews the status of its customers yearly. Based on this, 3 discounts 5%, 10%, 15% are given depends on the accident record of the driver.

X_t = percentage of discount at the end of year t Then $T = \{0, 1, 2, \dots\}$ & $S = \{5, 10, 15\}$

2. Discrete-state continuous Time stochastic Process

No claim Bonus
(NCB) into heading

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a) The no. of incoming calls X_t in an interval $[0, t]$. Then $T = \{t : 0 \leq t < \infty\}$ & $S = \{0, 1, 2, \dots\}$

b) The no. of cars X_t parked at a commercial center in the time interval $[0, t]$. Again $T = \{t : 0 \leq t < \infty\}$, $S = \{0, 1, 2, \dots\}$

3. Continuous state, Discrete-time

a) The share price of an asset at the close of trading on day t with $T = \{0, 1, 2, \dots\}$ & $S = \{x : 0 \leq x < \infty\}$

4) Continuous state, Continuous-time

b) The value of NSE index at time t where $T = \{t : 0 \leq t < \infty\}$ & $S = \{n : 0 \leq n < \infty\}$

Let $\{X(t), t \in T\}$ be a stochastic process with state space S . Let $x : T \rightarrow S$ be a function. We may think that $\{x(t), t \in T\}$ as a possible evolution (trajectory) of $\{X(t), t \in T\}$. The functions x are called the sample paths of the stochastic process.

infinitely many

Remark:- In general, the set of all possible sample paths, called the sample space of the stochastic process is uncountable. For finite state \rightarrow infinite paths

Independent Increments :- If for all $t_0, t_1, t_2, \dots, t_n$, the random variables $X_{t_0}, X_{t_1} - X_{t_0}, X_{t_2} - X_{t_1}, \dots, X_{t_n} - X_{t_{n-1}}$

are independent (or equivalently), $X_{t_1} - X_{t_0}$ is independent of X_s for $s < t_0$, then the process $\{X_t; t \in T\}$ is said to be a process with independent increments.

Stationary Increments :- A stochastic process $\{X_t; t \in T\}$ is said to have stationary increments if $X_{t_2+c} - X_{t_1+c}$ has the same

distribution as $X_{t_2} - X_{t_1}$, for all choices of t_1, t_2 & $c > 0$.

Stationary Process :- If for arbitrary t_1, t_2, \dots, t_n , $t_1 < t_2 < \dots < t_n$, the joint distributions of the vector random variables $(X_{t_1}, X_{t_2}, \dots, X_{t_n})$ & $(X_{t_1+h}, X_{t_2+h}, \dots, X_{t_n+h})$ are the same

for all $h > 0$. Then the stochastic process is said to be stationary stochastic process of order n (or simply a stationary process). The stochastic process $\{X_t; t \in T\}$ is said to be a strong stationary stochastic process or strictly stationary process if the above property is satisfied for all n .

Markov Process

Consider a system that is modelled by a discrete time stochastic process $\{X_n; n \geq 0\}$ with a countable state-space S , say $S = \{0, 1, 2, \dots\}$. Consider a fixed value of n that we shall call "the present time" or just the "present". Then X_n is called the present (state) of the system, $\{X_0, X_1, X_2, \dots, X_n\}$ is called the past of the system and $\{X_{n+1}, X_{n+2}, \dots\}$ is called the future of the system. If $X_n = i$ & $X_{n+1} = j$. It is said that the system has jumped (or made a transition) from state i to state j from time n to $n+1$.

Markov Property :- If the present state of the system is known, the future of the system is independent of the past.

Or in other words :- the present state of the system contains all the relevant information needed to predict the future of the system in a probabilistic sense.

Tossing a coin regularly
 $X(t) \rightarrow$ no. of heads till time t

Def :- Let $\{X_t; t \geq 0\}$ be a stochastic process, defined over a probability space (Ω, \mathcal{F}, P) & with state space $(\mathbb{R}, \mathcal{B})$. We say that $\{X_t; t \geq 0\}$ is a Markov process if for any $0 \leq t_1 < t_2 < \dots < t_n$ and for any $B \in \mathcal{B}$.

$$P(X_{t_n} \in B | X_{t_1}, \dots, X_{t_{n-1}}) = P(X_{t_n} \in B | X_{t_{n-1}})$$

Remark :- Any stochastic process which has independent increments is a Markov process.

(i) X_n : no. of heads in first n tosses
 ex:- Let Y_0, Y_1, \dots, Y_n be non-negative independent & identically distributed random variables.

The sequence $\{X_n; n \geq 0\}$ with

$$X_0 = Y_0$$

$$X_n = X_0 + Y_1 + \dots + Y_n$$

is a Markov Process

$$P(X_{n+1} = j | X_n = i, X_{n-1} = i_{n-1}, \dots, X_1 = i_1, X_0 = i_0)$$

$$= \frac{P(X_{n+1} = j, X_n = i, X_{n-1} = i_{n-1}, \dots, X_1 = i_1, X_0 = i_0)}{P(X_n = i, X_{n-1} = i_{n-1}, \dots, X_1 = i_1, X_0 = i_0)}$$

$$= \frac{P(X_0 = i_0) P(Y_1 = i_1 - i_0) \dots P(Y_n = i_n - i_{n-1}) P(X_{n+1} = j | i)}{P(X_0 = i_0) P(Y_1 = i_1 - i_0) \dots P(Y_n = i_n - i_{n-1})}$$

$$= \frac{P(Y_{n+1} = j - i)}{P(X_{n+1} = j | X_n = i)}$$

$$\frac{P(X_{n+1} \cap X_n)}{P(X_n)} = P(X_{n+1} | X_n)$$

VG Kulkarni

$$P(X_n = j | X_n = i)$$

$$Y_{n+1} = X_{n+1} - X_n$$

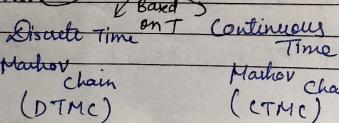
$$\frac{P(X_{n+1})}{P(X_n)} \cdot P(X_{n+1} = j - i)$$

$$\hookrightarrow P(X_{n+1} = i - i_{n-1}) P(X_n = i)$$

$$P(X_n = i)$$

$X_n \rightarrow$ always non-decreasing Why?

A discrete state Markov process is known as markov chain



Discrete Time Markov Chain (DTMC):-

A sequence of random variables $\{X_n\}_{n \in \mathbb{N}}$ with discrete state space is called a DTMC if it satisfies the condition:-

$$P(X_{n+1} = j | X_n = i, X_{n-1} = i_{n-1}, \dots, X_0 = i_0) = P(X_{n+1} = j | X_n = i) \quad (1)$$

for all $n \in \mathbb{N}$ & for all $i, i_1, \dots, i_{n-1}, i, j \in S$ with $P(X_0 = i_0, \dots, X_n = i_n) > 0$

In other words, eqn (1) implies that if we know the present state " $X_n = i$ ", the knowledge of past history " $X_{n-1}, X_{n-2}, \dots, X_0$ " has no influence on the probabilistic structure of the future state X_{n+1} .

Example:- X_n = "no. of heads obtained in the first n tosses"

Do it:- If $\{X_n; n \geq 0\}$ is a Markov chain, then for all $i, i_1, \dots, i_{n-1} \in S$ we have

$$P(X_0 = i_0, X_1 = i_1, \dots, X_n = i_n) = P(X_0 = i_0) P(X_1 = i_1 | X_0 = i_0) \dots P(X_n = i_n | X_{n-1} = i_{n-1})$$

Def of Transition Probability):-

Let $\{X_n; n \geq 0\}$ be a markov chain.

The probabilities $p_{ij} = P(X_{n+1} = j | X_n = i)$ are called transition probabilities.

Def:- A markov chain $\{X_n; n \geq 0\}$ is called homogeneous or a Markov chain with stationary probabilities if the transition probabilities do not depend on n .

Remark:- Because of the

P^n - (Transition Probability Matrix)

The matrix

$$P = (p_{ij}) = \begin{pmatrix} p_{00} & p_{01} & p_{02} & \dots \\ p_{10} & p_{11} & \dots & \dots \\ p_{20} & & & \dots \\ \vdots & & & \dots \end{pmatrix}$$

is called the transition probability matrix or stochastic matrix. Note that

$$\begin{aligned} p_{ij} &\geq 0 \quad \forall i, j \in S \quad \text{if } \textcircled{1} \\ \sum_j p_{ij} &= 1 \quad \forall i \in S \quad \text{if } \textcircled{2} \end{aligned}$$

Explanation of $\textcircled{1}$: p_{ij} is a conditional probability, so $p_{ij} \geq 0$. Now

$$\begin{aligned} \sum_{j \in S} p_{ij} &= \sum_{j \in S} P(X_{n+1}=j | X_n=i) \quad (\text{if } \textcircled{1}) \quad (n \rightarrow n+1) \\ &\quad \xrightarrow{\text{state space}} \rightarrow \text{state space} \\ \text{State } &= P(X_{n+1} \in S | X_n=i) \quad (\xrightarrow{\text{if } \rightarrow \text{H.T.}} P(\text{outcome is } S)) \\ (100, 101) &= 1 \quad (\text{since } X_{n+1} \in S \text{ with probability 1}) \end{aligned}$$

Example:- On any given day, Kuldip is cheerful (C), normal (N) or depressed (D). If he is cheerful today, then he will be C, N or D tomorrow with probabilities 0.5, 0.4 & 0.1 respectively.

If he is feeling so-so today, then he will be C, N or D tomorrow with probabilities 0.3, 0.4, 0.3 respectively.

If he is gloomy today, then he will be C, N or D with probabilities 0.2, 0.3, 0.5

respectively.

Let X_n denote the mood of Kuldip on the n th day. Then $\{X_n, n \geq 0\}$ is a 3 state discrete time Markov chain (state 0 = C, state 1 = N, state 2 = D) with transition probability matrix

$$P = \begin{pmatrix} 0.5 & 0.4 & 0.1 \\ 0.3 & 0.4 & 0.3 \\ 0.2 & 0.3 & 0.5 \end{pmatrix}$$

Ques:- Is a DTMC completely characterised by its transition probability matrix? No

Defn:- Let $a_i = P(X_0=i)$, $i \in S$
 $\& a = [a_i]_{i \in S}$

be a row vector representing the PMF of X_0 . We say that a is the initial distribution of the DTMC.

Next Page continuation

How to characterise a stochastic process?

To provide a mathematically precise method to describe a stochastic process unambiguously.
 Stochastic process - collection of random variables
 To start with - "How one describes a single random variable"

A single random variable is completely described by its CDF $F(x) = P(X \leq x)$ $-\infty < x < \infty$

A multivariate random variable (X_1, X_2, \dots, X_n) is completely described by its joint CDF

$$F(x_1, x_2, \dots, x_n) = P(X_1 \leq x_1, X_2 \leq x_2, \dots, X_n \leq x_n)$$

for all $-\infty < x_i < \infty$ $i = 1, 2, \dots, n$

→ Next Pg.

Theorem:- A DTMC $\{X_n, n \geq 0\}$ is completely described by its initial distribution a & the transition probability matrix P .

Proof:- Do it yourself

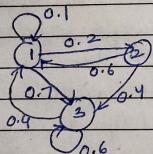
Hint:- Show that the finite dimensional joint PMF $P(X_0 = i_0, X_1 = i_1, \dots, X_n = i_n)$ can be computed in terms of a & P .

Thus if the parameter set T is finite, the stochastic process $\{X_n; n \geq 0\}$ is a multivariate random variable & hence is completely described.

We have already discussed about transition probability matrix. The transition probability matrix of a DTMC can be represented graphically by its transition diagram which is a directed arc from node i to node j if $P_{ij} > 0$ & to itself if $P_{ii} > 0$

state space $\{1, 2, 3\}$

$$P = \begin{bmatrix} 0.1 & 0.2 & 0.7 \\ 0.6 & 0 & 0.4 \\ 0.9 & 0 & 0.6 \end{bmatrix}$$



Example:- Coming back to characterisation, we have shown that a DTMC $\{X_n, n \geq 0\}$ is completely characterised by initial distribution a & transition probability matrix P i.e. we can find

$$\begin{aligned} P(X_0 = i_0, X_1 = i_1, \dots, X_n = i_n) \\ = a_{i_0} b_{i_0 i_1} p_{i_1 i_2} \dots p_{i_{n-1} i_n} p_{i_n i_n}, \text{ where} \\ a = P(X_0 = i_0) \quad \& \quad b_{ij} = P(X_{n+1} = j | X_n = i) \end{aligned}$$

Note that

$$P = \begin{bmatrix} p_{11} & p_{12} & \dots & p_{1n} \\ \vdots & & & \\ p_{n1} & p_{n2} & \dots & p_{nn} \end{bmatrix} \text{ where } S = \begin{bmatrix} f_{1,1}, f_{1,2}, \dots, f_{1,n} \\ \vdots \\ f_{n,1}, f_{n,2}, \dots, f_{n,n} \end{bmatrix}$$

$$a = [a_{i_0}, a_{i_1}, \dots, a_{i_n}]$$

$$P(X_0 = i_0) \quad \downarrow \quad P(X_n = i_n)$$

Let $\{X_n, n \geq 0\}$ be a DTMC on state space $S = \{1, 2, 3, 4\}$

$$S = \{1, 2, 3, 4\}$$

$$P = \begin{bmatrix} 0.1 & 0.2 & 0.3 & 0.4 \\ 0.2 & 0.2 & 0.3 & 0.3 \\ 0.5 & 0 & 0.5 & 0.0 \\ 0.6 & 0.2 & 0.1 & 0.01 \end{bmatrix}$$

$$\& a = [0.25 \ 0.25 \ 0.25 \ 0.25]$$

$$a_1 = P(X_0 = 1) = 0.25 \quad a_2 = P(X_0 = 2) = 0.25$$

$$a_3 = P(X_0 = 3) = 0.25 \quad a_4 = P(X_0 = 4) = 0.25$$

Further:- $i_0 = 1, i_1 = 2, i_2 = 3, i_3 = 4$. For

$$\begin{aligned} P(X_0 = 1, X_1 = 3, X_2 = 1, X_3 = 4) \\ = P(X_0 = i_0, X_1 = i_1, X_2 = i_2, X_3 = i_3) \\ = a_{i_0} b_{i_0 i_1} p_{i_1 i_2} b_{i_2 i_3} p_{i_3 i_4} \end{aligned}$$

$$\begin{aligned} = a_1 b_{13} p_{31} b_{14} p_{14} \\ = 0.25 \times 0.3 \times 0.5 \times 0.4 \\ = 0.015 \end{aligned}$$