MEAN AND VARIANCE OF FUNCTION OF A RANDOM VARIABLE

Definition: A function of a random variable is a rule that transforms a point from a sample space into a real number line.

If X and Y are random variables on the sample space S, then Y is said to be function of X provided Y can be represented by $Y(S) = \phi$ (X(S)for every $s \in S$. Some real valued function of a real variable: 2X+t, X^2+3X and tX are some example of function of X.

Definition: Let X and Y be random variable on the same space S with $Y = \phi(X)$, then the mean of function of X_n is given by:

$$E(Y) = E(\phi(X)) = \sum_{i=1}^{n} \phi(xi)$$
 f(xi), where f(xi) is the probability mass function of X.

Example 1: An unbiased green octahedral die is tossed. If Y denote twice the number appears and (i) Z = 3 + Y (ii) $Z = \frac{1}{2}Y + 1$, find E(Z)

(i) Here,
$$S = \{1, 2, 3, ---, 8\}$$

 $Y(S) = \{2, 4, 6, ---, 16\}$

The distribution of Y is given below

yi	2	4	6	8	10	12	14	16
f(yi)	1/8	1/8	1/8	1/8	1/8	1/8	1/8	1/8
3 + yi	5	7	9	11	13	15	17	19
$^{1}/_{2}$ yi + 1	2	3	4	5	6	7	8	9

Let ϕ (Y) = 3+ yi :. Z = ϕ (Y) where Z and Y are r.vs on the same sample space.

E(Z) = E(
$$\phi$$
 (Y) = $\Sigma \phi$ (yi)f(yi) = Σ (3 + yi)f(yi)
= 5 x $\frac{1}{8}$ + 7 x $\frac{1}{8}$ + - - + $\frac{19}{8}$ = 12

(ii) Given that
$$\phi(Y) = \frac{1}{2}Y + 1$$

$$E(Z) = E(\phi(Y)) = E(\frac{1}{2} + 1)$$

$$= \sum_{i=1}^{8} (\frac{1}{2} + 1) f(yi)$$

$$= \frac{1}{8} (2 + 3 + 4 + \dots + 9) = \frac{44}{8} = 5.5$$

Exercise: A fair coin is tossed twice. If X denote the number of heads appearing in the sample space S, find the distribution of Z. Given that $Y = \phi(X)$, find E(Y), if (a) $\phi(X) = X^2 - 1$, (b) $\phi(X) = 2 + 3X$. Find the value of Y of (a) and (b) above.

Xi	0	1	2
xi ² -1	-1	0	3
$(xi2-1)^2$	1	0	9

CONTINOUS RANDOM VARIABLE

Definition: A random variable X is said to be continuous if the set of its positive values are contained in an interval, i.e $X \in \{a, b\}$ for $x_1, x_2, \dots, x_n \in X$.

Definition 2: A continuous random variable can also be defined as that random variable which takes on a non-countably infinite number of values.

Definition 3: If X is a continuous random variable, a function f(x) is said to be continuous probability density function if it satisfies the following conditions:

$$(1) F(x) \ge 0$$

(2)
$$\int_{-\infty}^{\infty} f(x) dx = 1.$$

Definition 4: If X lies between a and b i.e. if a < X < b, the probability that X lies

between a and b is defined by
$$P(a \le X \le b) = \int_a^b f(x) dx$$

Note: If X is continuous, we shall assume unless otherwise stated that:

$$P(a \le X \le b) = P(a \le X \le b) = P(a \le X \le b).$$

Definition 5: If X is a continuous random variable, the distribution function F(X) is given by

$$F(X) = P(X \le x) = P(-\infty \le X \le x) = \int_{-\infty}^{x} f(u) du.$$

It should be noted that F(X) is a monotically increasing function.

Example: if X is a continuous random variable with pdf

$$F(X) = \begin{cases} KX/8, & 0 < x < 4 \\ 0, & \text{otherwise} \end{cases}$$

Find (i) the constant K, (ii) $P(1 \le x \le 3)$ (iii) $P(-2 \le x \le 2)$ and (iv) $P(-3 \le x \le 5)$.

Solution

(i)
$$K = 1$$
, $f(x) = KX/8$, $0 < x < 4$
0, otherwise

(ii)
$$\int_{1}^{3} f(x) dx = \int_{1}^{3} \frac{x}{8} dx$$
, $\frac{1}{8} \left[x^{2} / 2 \right]_{1}^{3} = \frac{8}{16} = \frac{1}{2}$

(iii) $P(-2 \le x \le 2) = P(0 \le x \le 2)$ since $-2 \le x \le 0$ is outside the function condition

$$\int_0^2 f(x) dx = \frac{1}{4}.$$

(iv)
$$P(-3 < x < 5) = P(0 < x < 4) = 1$$
.

RELATIONSHIP BETWEEN THE PROBABILITY DENSITY FUNCTION F(X) AND PROBABILITY DISTRIBUTION FUNCTION F(X)

(i) If F(X) is a distribution function of a continuous random variable X,

$$F(X) = P(X \le x) = \int_{-\infty}^{x} f(u) du$$
 where $F(X)$ is a probability density function.

(ii)If f(x) is a probability density function and F(X) is the corresponding distribution function of a continuous random variable X.

$$F(x) = dF(X)/dx$$
 i.e.

The derivative of the probability distribution function is the density function.

Example: Given that
$$f(x) = \begin{cases} tx^3 & 0 < x < 4 \\ 0, \text{ otherwise} \end{cases}$$

Probability density function of a continuous random variable X. finds (a) th constant t and (b) the distribution function.

Solution

(a) Since f(x) is a pdf i.e.
$$\int_0^x f(x)dx = 1$$

 $t = \frac{1}{64}$.
 $F(x) = \frac{1}{64} x^3 \ 0 < x < 4$.

(b) By definition of distribution function F(X)

(c)
$$F(X) = P(X < x) = \int f(u) du$$
 $f(u) = 1/64u^3, 0 < x < 4$

(d)
$$F(X) = P(X < x) = \int_{-\infty}^{x} \frac{1}{64} u^{3} du$$

Using the condition given for $f(x)$ above, if $x < 0$, then

$$F(X) = 0$$
, if $0 < x < 4$, then

(e)
$$F(X) = \int_0^x f(u)du = \int_0^x \frac{1}{64} u^3 du$$

= $\left[u^4 /_{256} \right]_0^4 = x^4 /_{256}$

(f) If
$$x \ge 4$$
, then $F(x) = \int_{4}^{0} f(u) du + \int_{0}^{4} f(x) = \int_{0}^{4} \frac{1}{64} u^{3} du + \int_{4}^{x} du du$

$$= \frac{u4}{256} \Big|_{0}^{4} + 0 = 1$$

Hence the refund probability distribution function is

$$F(X) = \begin{cases} 0 & x^4/_{256} & x < 0 \\ & 0 < x < 4, \\ 1 & x \ge 4. \end{cases}$$

Example: use the above distribution function to find $P(1 \le x \le 2)$

Solution:
$$P(1 \le x \le 2) = P(x \le 2) - P(x \le 1) = F(2) - F(1) = \frac{2^4}{256} - \frac{1^4}{256} = \frac{15}{256}$$

Exercise: The distribution function of a random variable X is given by

$$F(X) = \begin{cases} x^2/_4 & x \le 0 \\ 0 \le x \le 2 \end{cases}$$

Find the pdf and use your result to find $P(1 \le x \le 2)$

Solution

$$f(x) = dF(X)/dx = d(0)/dx + d(x^2/4)/dx + d(1)/dx$$

$$f(x) \int_{0}^{x}/2 \ 0 \le x \le 2 = 0 + x/2 + 0 = x/2$$

0, elsewhere

$$P(1 < x < 2) = \int_{1}^{2} \frac{x}{2} dx = [x^{2}/_{4}]^{2}_{1} = 1 - \frac{1}{4} = \frac{3}{4} = 0.75$$

MEAN AND VARIANCE OF A CONTINUOUS RANDOM VARIABLE

Definition 1: For a continuous r.v. X, having density function F(X) the expectation of X is defined as $E(X) = \int_{-\infty}^{\infty} Xf(X) dx$. As earlier mentioned the expectation of X is often called the mean of X and is denoted by the μ_x or μ when the particular r.v. is understood.

Definition 2: For a continuous random variable X, mean of X is defined as μ =

$$E(X) = \int_{-\infty}^{\infty} Xf(X) dx \text{ and } V(X) = E[(X - \mu)^2] = E(X^2) - \mu^2, \text{ where}$$

$$E(X2) = \int_{-\infty}^{\infty} X2f(x) dx.$$

Example: A continuous r.v. X has probability density function given by

$$F(X) = \begin{cases} de^{-3x} & x>0 \\ 0, & \text{otherwise} \end{cases}$$

Find the constant d and mean of X.

Solution:
$$\int de_{-3x} dx = 1 = \begin{vmatrix} \frac{1}{de^{-3x}} & 0 \\ -3 & 0 \end{vmatrix}$$
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$$0 - (-d/3) = 1, d = 3$$

Hence,

$$F(X) = \begin{cases} 3e^{-3x} & x \ge 0 \\ 0, & \text{otherwise} \end{cases}$$

Mean of $X = E(X) = \int_0^\infty X f(X) dx$ (integration by part)

$$= \int_{0}^{\infty} X (3e^{-3x}) dx$$

$$\frac{3-xe^{-3x}}{3} \Big|_{0}^{\infty} - \frac{e^{-3x}}{9} \Big|_{0}^{\infty}$$

$$E(X) = \frac{1}{3}$$

Example: A continuous random variable has a probability density function

$$f(x) = \begin{cases} x/3 + C & 0 \le x \le 2 \\ 0, & \text{otherwise} \end{cases}$$

Find (a) the constant C, (b) the mean and variance X and (c) the standard deviation of X.

Solution

$$\int \int (x/_3 + C) dx = \frac{x^2 + CX^2}{3}$$
Hence, $F(X) = \int_0^{x/_3 + 1/6} 0 \le x \le 2$

$$0, \text{ otherwise}$$

Mean of X = E(X) =
$$\int_0^2 Xf(X) dx = 11/9$$

Variance of $X = 16/9 - (11/9)^2 = 23/81$

S.D =
$$\sqrt{var(X)} = \frac{\sqrt{23}}{81} = 0.53$$
.

Exercise

(1) Show that the following are probability density function (pdf)

(a)
$$f_1(x) = \ell^{-x}$$
 $x>0$
(b) $f_2(x) = 2 \ell^{-2x}$ $x>0$
(c) $f(x) = (\theta + 1) f_1(x) - \theta f_2(x) 0 < \theta < 1$

(2) Following is a constant K so that the following is a pdf

$$F(x) = Kx^{2} -k < x < k$$

$$\int Xf(X)_{dx} = 1$$

$$\int_{-k}^{k} k_{x^{2}} = 1$$

$$\frac{x^{3}}{3k}\Big|_{-k}^{k} = \frac{k^{3} + k^{3}}{3} = \frac{2k^{4}}{3}$$

- (3) A fair coin is tossed until a head appears. Let X denotes the number of tossed refined.
 - (a) Find the density function of X.
 - (b) Find the mean and variance of X.

DISCRETE DISTRIBUTION: BERNOULLI DISTRIBUTION

Given an experiment whose outcomes can be classified into 1 or 2 classes for example pass or fail, on and off, head or tail. If we let x = 1 for a success and x = 0 for a failure, then the pdf or pmf is given by

$$p(X) = \begin{cases} 1, & \text{with prob p} \\ 2, & \text{with prob 1-p} \end{cases}$$
$$p(x) = p^{x} (1-p)^{1-x}$$

A random variable with the above probability mass function is said to be a Bernoulli random variable and the probability distribution is called a Bernoulli

distribution.

$$E(X) = p$$
$$V(X) = pq.$$

BINOMIAL DISTRIBUTION

This distribution deals with repeated and independent trials of an experiment with two outcomes resulting in success or failure, 0 or 1, true or false, yes or no. If the interest is on the no of successes and not in the order in which they occur, the probability of exactly X successes in n repeated trials is given by

$$f(x) = \begin{cases} \binom{n}{x} p^{x} (1-p)^{n-x}, & x = 0, 1, 2, ---, n \\ 0, & \text{otherwise} \end{cases}$$

p is the prob of success and q = 1-p is the prob of failure and x is the number of successes in repeated trials. F(x) gives the pdf of binomial distribution.

PROPERTIES OF BINOMIAL DISTRIBUTION

- (i) It has n independent trials
- (ii) It has constant prob of success p and prob of failure q = 1-p from trial to trial.
- (iii) There is assigned prob to non-occurrence of events.
- (iv) The mean (μ) = np and the variance (δ^2) = npq.
- (v) Each trial can result in one of only two possible outcomes called success or failure.

NB:
$$\binom{n}{x} = \frac{n!}{x!(n-x)!}$$

Given that X is a random variable with Binomial Distribution

$$f(x) = {n \choose x} p^x (1-p)^{n-x}, x = 0, 1, 2, \dots, n.$$

The mean of the random variable X with Binomial Distribution given above is

$$E(X) = np$$

$$V(X) = npq$$

Application: If X has a binomial distribution with n = 5 and $p = \frac{1}{2}$. Find (i) the P(X=1) (ii) P(X>1).

1.
$$\binom{5}{1}\binom{1}{2}5$$

2. $1 - \binom{5}{0}\binom{1}{2}5 + \binom{5}{1}\binom{1}{2}5$

Example: Assume that boys and girls are equally likely to be born. What is the prob of there being?

- (1) No boys in a family of 3 children
- (2) Only one boy in a family of 3 children.
- (3) At least one boy in a family of 3 children.

POISSON DISTRIBUTION

A random variable X is said to have a poisson distribution if the

$$f(x) = \underline{\pi} \underline{x} \ell - \underline{\pi}$$

$$x!, \qquad x = 0, 1, 2, \dots, \pi > 0$$

The poisson distribution provides a realistic model for many random phenomena. Since the values of a poisson random variable are the non-negative integers, any random phenomenon for which a count of some sort is of interest is a candidate for modeling by assuming a poisson distribution. Such a count might be the no of fatal traffic accidents per week in a given state, the no of radioactive particle emission's per unit of time, the no of telephone calls per hour coming into the switch board of a large business, the no of organizations per unit volume of some fluid, the no of defects per unit of some material.

PROPERTIES OF POISSON DISTRIBUTION

- (1) Random variable X assumes integer values.
- (2) The population average or rate is known.
- (3) There is assigned probability to non-occurrence of events.
- (4) The mean (μ) and the variance (δ^2) are equal. $E(X) = \pi$

$$V(X) = \pi$$

Example: Suppose that the average number of telephone calls arriving at the switch board of a small corporation is 30 calls per hour.

- (i) What is the prob that no calls will arrive in a 3 minute period?
- (ii) What is the prob that more than five calls will arrive in a 5 minute period?

Solution

Let assume X is call arriving at any time has a poisson distribution assume that time is measured in minutes; then 30 calls per hour is equivalent to 0.5 calls per minute, so the mean rate of occurrence is 0.5 per minute.

(i) p(no calls in 3-minutes period) = $\ell^{-vt} = \ell^{-0.5(3)} = 0.223$.

(ii)p(more than 5 calls in 5-minutes interval) = $\sum_{k=6}^{\infty} t_{\frac{-vt}{vt}} (vt)^{x}$

 $\mathbf{x}!$

$$= \sum_{k=6}^{\infty} \frac{t}{\frac{-(0.5)(5)}{x!}} = 0.42$$
or $1 - p(X \le 5) = 1 - \frac{\sum_{x=0}^{5} t}{\frac{-(2.5)}{x!}} = 0.42$

Exercise: Suppose that flaws in plywood occur at random with an average of one flaw per 50 square feet.

- (i) What is the prob that a 4 fool by 8 fool sheet will have no flaws?
- (ii)At most one flaw

Solution

(a) P(no flaw) =
$$\ell^{-1/50(32)} = \ell^{-(.64)} = .527$$

(b) P(at most one flaw) = $\ell^{-.64}$ + $.64 \ell^{-.64}$ = .865

STA 124: INTRODUCTION TO PROBABILITY DISTRIBUTION, NOTE 2 <u>GEOMETRIC DISTRIBUTION</u>

Suppose that independent trials each having a prob of success is performed until a success occur. If we let X be the no of trials required then,

$$F(x) = p(1-p)^{x-1}, x = 1, 2, -----$$

X is said to have a Geometric Random Variable with parameter p.

 $x-1 \Rightarrow$ failure, $x \Rightarrow$ no of trials for the first success. The sample size is not fixed.

PROPERTIES OF GEOMETRIC DISTRIBUTION

- (1) There is a sequence of independent trials (outcome of one trial does not depends on another).
- (2) Only two outcomes e.g. success or failure are possible at each trial.
- (3) There is a constant prob. of success at each trial.
- (4) X is the no of trials for the 1st success to appear (different from that of Binomial distribution).

Show that f(x) is a pdf.

$$\sum_{x=1}^{\infty} p(1-p)_{x-1} = p/1-q = p/p = 1.$$

$$E(X) = \sum_{x=1}^{\infty} X pq_{x-1} = p+2pq+3pq^2 + \cdots$$

$$p(1+2q+3q^2 + \cdots)$$

$$p(1-p)^{-2} = p/(1-p)^2 = p/p^2 1/p$$

$$E(X2) = EX(X-1) + E(X)$$

$$\sum_{x=1}^{\infty} X(X-1)pq_{x-1}$$

$$= 0 + 2pq + 6pq^2 + 12pq^3 + 20pq^4$$

$$= 2pq(1 + 3q + 6q^2 + 10q^3 + \cdots)$$

$$= 2pq(1 + 3q + 3.2q^2 + 5.2q^2 - \cdots)$$

$$= 2pq(1-q)^{-3} = 2pq/(1-q)^3 = 2pq/p^3 = 2q/p^2$$

$$E(X2) = 2q/p^2 + 1/p$$

$$V(X) = 2q/p^{2} + 1/p - 1/p^{2}$$

$$= 2q + p - 1 = 2q - q = q/p^{2}$$

Example: A container has 6 white balls and 4 black balls. Balls are randomly selected one at a time until a black ball is picked. Assure that each ball selected is replaced before the next one selected. What is the prob that

- (1) Exactly 3 draws are needed
- (2) Not more than two draws are needed.

Solution

$$F(x) = p(1-p)^{x-1}$$
, $x = 1, 2, ---$

X is the no of draws for the 1^{st} black ball. P = 0.4

$$P(x) = (0.4)(0.6)^{x-1}$$
, $x = 1, 2, ---$

(i)
$$P(3) = (0.4)(0.6)^2 = 0.144$$

$$(ii)P(1) + p(2) = 0.4 + 0.4 \times 0.6 = 0.64$$

UNIFORM DISTRIBUTION

This is the distribution in which all the possible values have equal probability.

E.g. losing a fair coin; let x = 0 or 1 for head or tail respectively then $p(x) = \frac{1}{2}$, x = 0, 1. Throwing a die let x be the outcome on the die $x = \{1, 2, 3, 4, 5, 6\}$

 $P(x)^{1}/_{6}$. When x = 1 - - - 6., it is an example of uniform distribution

$$P(X=x) = 1/n$$
, where $x = 1, 2, --- n$.

Definition: a discrete random variable X taking values 1, 2, 3, - - -, n such that p(X=x) = 1/n, x=1, 2, ---, n has a discrete uniform distribution.

Show that p(X) is a pmf

$$E(X) = \sum X f(X) = \sum X 1/n = X$$

$$\frac{E(X)}{n} = \frac{(1+2+3+\dots+n)}{n} = \frac{n(1+n)}{2n} = \frac{n+1}{2}$$

$$E(X^2) = \sum X^2 \frac{1}{n} = \frac{1}{n} (1+4+9+\dots) = \frac{n(n+1)(2n+1)}{6n} = \frac{(n+1)(2n+1)}{6}$$

$$V(X) = E(X^2) - [E(X)]^2 = \frac{(n+1)(2n+1)}{6} - \left[\frac{n+1}{2}\right]^2$$

$$= \frac{2n^2 + 3n + 1 - n^2 + 2n + 1}{6}$$

$$= \frac{4n^2 + 6n + 2 - 3n^2 - 6n - 3}{12}$$

$$= \frac{n^2 - 1}{12} = \frac{(n-1)(n-2)}{12}$$

CONTINUOUS DISTRIBUTION

Uniform distribution: A random variable X is said to have a uniform distribution over the interval (a, b) if the prob density function is given by

$$F(X) = \frac{1}{b-a}, a < x < b$$
0, otherwise.

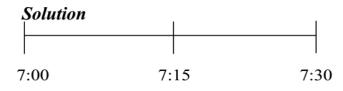
Show that it is a pdf.

$$E(X) = b + a / V(X) = (b - a)^{2} / 12$$

Where
$$b^3 - b^2 = (b - a)(b^2 + ab + a^2)$$
.

Example: Buses arrive at a shop at 15 mins interval starting from 7am. If a passenger arrives at the b/stop at a time that is uniformly distributed between 7 and 7:30 am. Find that the passenger waits

- (a) Less than 5 mins for a bus
- (b) More than 10 mins for a bus.



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(a)
$$P(7:10 < x < 7:15) + p(7:25 < x < 7:30)$$

 $10 < x < 15$ $25 < x < 30$

Uniform between (0, 30) a = 0, b = 30,

$$F(X) = \frac{1}{30 - 0} = \frac{1}{30}$$

$$\int_{10}^{15} \frac{1}{30} dx + \int_{25}^{30} \frac{1}{30} dx$$

$$\frac{x}{30}\Big|_{10}^{15} + \frac{x}{30}\Big|_{25}^{30}$$

$$\frac{(15-10)+(30-25)}{30} = \frac{5+5}{30} = \frac{10}{30} = \frac{1}{30}$$

(b)
$$p(7:00 \le x \le 7:05) + p(7:15 \le x \le 7:20)$$

$$0 < x < 5$$
 $15 < x < 20$

$$\int_{\text{0}}^{\text{5}} \frac{1}{30} \, dx + \int_{\text{15}}^{2\text{0}} \frac{1}{30} \, dx$$

$$\frac{x}{30}\Big|_{0}^{5} + \frac{x}{30}\Big|_{5}^{20}$$

$$= 5/30 + 5/30 = 10/30 = \frac{1}{3}$$

NORMAL DISTRIBUTION

A random variable X is said to have a normal distribution if the

$$F(X) = \frac{1}{\delta\sqrt{2\pi} \ell^{-1/2} \delta^{-2}}, \quad -\infty < X < \infty$$

$$-\infty < \mu < \infty$$

$$\delta^{-2} > 0$$

The mean is μ and the variance is δ^2 . The standard form is $\mu = 0$, and $\delta^2 = 1$. If $X \sim N(0, 1)$, then X has a standard normal distribution.

STA 124: INTRODUCTION TO PROBABILITY DISTRIBUTION, NOTE 2 PROPERTIES OF NORMAL DISTRIBUTION

- (1) The curve is symmetrical about the vertical axis through the mean.
- (2) It has a bell shape
- (3) The mean, mode and median coincide at the point μ .
- (4) δ^2 determine the shape of the curve, when δ^2 is large, the curve tend to be flat and peaked when δ^2 is small.
- (5) The total area under the curve and above the horizontal axis is equal to 1. $\int f(x)dx = 1$

CUMMULATIVE DISTRIBUTION FUNCTION

$$P(X < x) = p(-\infty < x < x) = \int_{-\infty}^{x} f(x) dx$$

This integral cannot be evaluated explicitly, however, the value of integral has been extensively tabulated for $\mu = 0$, $\delta^2 = 1$.

i.e. for the standard normal (standardization) if X ~ $N(\mu, \delta^2)$

then
$$Z = \underline{x - \mu} \sim N(0, 1)$$

If
$$x \sim N(\mu, \delta^2)$$

$$P(-\infty < x < a) = p \begin{bmatrix} -\infty & -\mu & < x - \mu & < a - \mu \\ \delta & \delta & \delta \end{bmatrix}$$

$$= p \left[-\infty < Z < \underline{a - \mu} \right]$$

$$= p(a < x < b)$$

$$= \left[\frac{\mathbf{a} - \boldsymbol{\mu}}{\delta} < \underline{\mathbf{x}} - \boldsymbol{\mu} < \underline{\mathbf{b}} - \underline{\boldsymbol{\mu}} \right]$$

$$= p \left[\frac{a - \mu}{\delta} < Z < \frac{b - \mu}{\delta} \right]$$

$$= \phi$$
 (b) $- \phi$ (a)

Example: find the prob that $p(1.5 \le Z \le 2.3)$

Solution:
$$\phi$$
 (2.3) – ϕ (1.5)
0.9893 – 0.9332
= 0.0561

Exercise: find the prob of

(a)
$$P(0.1 \le Z \le 2.03)$$

(b)
$$P(-1 \le Z \le 2)$$

Example: The score in a test is distributed as normal with mean 50 and standard deviation 10.

- (a) What is the prob of obtaining the score less than 55.
- (b) A score between 45 and 55.

Solution:

(a)
$$P(X < 55) = p \left[\frac{a - \mu}{\delta} < \frac{55 - 50}{10} \right]_{10}$$

$$= p(Z < 0.5) = 0.6915$$

(b)
$$P(45 < x < 55) = p \left[\frac{45 - 50}{10} < Z < \frac{55 - 50}{10} \right]$$

$$= p(-0.5 < Z < 0.5)$$

$$= \phi (0.5) - \phi (-0.5)$$

$$= 0.6915 - 0.3085$$

$$= 0.3830$$

Example: The mean and standard deviation of the weight of boys in a group are known to be 55kg and 2kg respectively. Find the proportion of boys whose weight are:

(i) Between 50kg and 60kg

- (ii) More than 58kg
- (iii) Less than 52kg.

Solution

(i)
$$P[(50-55)/2 < Z < (60-55)/2]$$

 $= P(-2.5 < Z < 2.5) = \phi (2.5) - \phi (-2.5)$
 $= 0.9938 - 0.0062 = 0.9876$
(ii) $P(X > 58) = 1 - P(X \le 58)$
 $= 1 - 0.9332$
 $= 0.0668$.
(iii) $P(X < 52) = P(Z < -1.5) = 0.0668$

Using the same example but put standard deviation to be 4, obtain the weight limit that will contain:

(i) 68% (ii) 95%

Solution

0.68 = 1-0.68 = 0.32/2 = 0.16, 1-0.16 = 0.84
P(-1 < Z < 1)

$$Z = \frac{x - \mu}{\delta}$$

 $x - \mu = Z \delta$
 $x = Z \delta + \mu$
 $x = -1(4) + 55$
=51

:. The weight limit is between 51 and 59.

(ii) 95%
$$Z(-1.96 < Z < 1.96)$$

 $x = -1.96(4) + 55 = 47.16$
 $x = 1.96(4) + 55 = 62.84$

Exercise: the mean and the standard deviation in a mathematical test is 70 and 10 respectively. Find the score in standard units of student receiving (i) 85 (ii) 40 marks.

Solution

(i)
$$Z = (85 - 70)/10 = 1.5$$
 (ii) $Z = (40-70)/10 = -3$

(ii) Find the mark corresponding to standard score (a) 0, (b) 1.70 (c) 1.15

$$x = \mu + Z\delta$$

(a)
$$x = 70 + 0^{\delta} = 70$$

(b)
$$x = 70 + 1.70(10) = 87$$

(c)
$$x = 70 + 1.15(10) = 81.5$$

- (3) A particular storage battery lasts on the average five years. If the battery lives are normally distributed with standard deviation 0.5 years. Find the prob that the given battery will
 - (1) Last less than 6.4 years = 0.9974
 - (2) More than 5.3 years
 - (3) Between 4.8 years and 6 years = 0.6541

BIVARIATE PROBABILITY DISTRIBUTION

Let f(x,y) be the pdf of two r.v. X and Y. A pdf or a distribution function called a joint pdf or a joint distribution when more than one variable is involved. Thus f(x,y) is the joint (bivariate) pdf of the r.v. X and Y.

Consider the event a < x < b. This event can occur when and only when the event a < x < b, $-\infty < y < \infty$ occurs, i.e. $p(a < x < b, -\infty < y < \infty)$

$$= \int_a^b \int_{-\infty}^{\infty} f(x, y) dy dx \qquad --- (1)$$

 $P(a < x < y, -\infty < y < \infty)$

Note that each of equation (1) and (2) is a function of x above $f_1(x)$.

 $f_1(x)$ is called the marginal p.d.f.

$$p(-\infty < x < \infty, -\infty < y < \infty) = \int_{\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dy dx$$
$$= \sum_{x \in \mathcal{X}} \sum_{y} f(x, y) = 1$$

If X and Y are independent then f(x,y) = f(x) f(y).

But if dependent f(x,y) = f(x/y)f(y) or f(y/x)f(x)

Example: Let the joint pdf of X and Y be

$$F(x,y) = \underbrace{x+y}_{21} \quad x = 1, 2, 3.$$

$$y = 1, 2$$

$$0, \quad \text{otherwise}$$

$$P(x=3) = f_1(3)$$

$$= f(3,1) + f(3,2)$$

$$P(x=3,y=1,2) = \sum_{x=3}^{2} \sum_{y=1}^{2} f(x,y)$$

$$4/21 + 5/21 = 9/21 = 3/7$$

$$P(y=2) = f_2(2)$$

$$= f(1, 2) + f(2, 2) + f(3, 2)$$

$$= \underbrace{1+2}_{21} + \underbrace{2+2}_{21} + \underbrace{3+2}_{21} = \underbrace{12}_{21}$$

$$= \underbrace{x+1}_{21} + \underbrace{x+2}_{21} = \underbrace{2x+3}_{21}$$

:. The marginal p.d.f. of X

$$f_1(x) = \sum_{y=1}^{2} \frac{X+Y}{21} = \underbrace{1+y}_{21} + \underbrace{2+y}_{21} + \underbrace{3+y}_{21} = \underbrace{6+3y}_{21} = \underbrace{3(2+y)}_{21} = \underbrace{2+y}_{21}$$

Example: The joint density function of x and y is given as

$$F(x,y) = 2(x+y-2xy), 0 < x < 1$$
$$0 < y < 1$$

Find the marginal of X and Y

Solution

$$F(x) = \int_{Ry}^{1} f(x, y) dy$$

$$= \int_{0}^{1} s(x + y - 2xy) dy$$

$$= 2 \left| \frac{xy}{2} + \frac{y^{2}}{2} - \frac{2xy^{2}}{2} \right|_{0}^{1}$$

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

$$F(y) = 1$$

:. $F_1(x)$ and $f_1(y)$ are uniformly distributed

CONDITIONAL PROBABILITY DISTRIBUTION FUNCTION

Let X and Y denote r.v. of discrete type which have the distribution p.d.f f(x, y).

Let $f_1(x)$ and $f_1(y)$ denote the marginal p.d.f. of X and Y respectively.

$$f(Y/X) = \underline{f(x, y)}$$
, provided $f(x) > 0$
 $f(x)$

Also,

$$f(X/Y) = \underline{f(x, y)}$$
, provided $f(y) > 0$

Example: In the previous example

$$f(x,y) = X + Y , X = 1, 2, 3$$

 $21 Y = 1, 2$
Find $p(X/Y) = f(x,y)$

Find
$$p(X/Y) = \frac{f(x,y)}{f(y)}$$

$$= fl(y) = \sum_{x=1}^{3} \frac{X+Y}{21}$$

$$=(2+y)/7$$

Example:
$$f(x, y) = 4xy, 0 \le x \le 1, 0 \le y \le 1$$

Find the marginal p.d.f. of X and hence f(Y/X).

Solution

$$f_1(x) = \int_0^1 4xy \, dy$$
$$= \underbrace{4^2 x y^2}_{2} \Big|_0^1$$

$$f_1(x) = 2x, 0 \le x \le 1$$

0, otherwise

$$f(Y/X) = f(x, y) = 4^2xy^2 = 2y$$

 $f_1(x) = 2$

Exercise: Given f(x, y) = b(x+2y), $1 \le x \le 3$,

$$0 \le y \le 2$$

Find (i) b (ii) find the marginal of X (iii) find the marginal of Y

Find the conditional p.d.f. of X given Y