

① Arithmetic Mean of Ungrouped Data

$$\bar{x} = \frac{\sum_{i=1}^n x_i}{n} \quad (\text{no. of obs.})$$

② Mean of Grouped Data:

$$\bar{x} = \frac{\sum f_i \cdot x_i}{N}$$

$$N = \sum f_i$$

$$x_i = \frac{\text{lower class limit} + \text{upper class limit}}{2}$$

$$\text{class limit} = l - 0.5 \text{ to } h + 0.5$$

③ Median

Ungrouped

$$n = \text{odd} \quad M = \left(\frac{n+1}{2} \right)^{\text{th}} \text{ observ.}$$

$$n = \text{even} \quad M = \frac{\left(\frac{n}{2} \right)^{\text{th}} + \left(\frac{n}{2} + 1 \right)^{\text{th}}}{2} \text{ obsv.}$$

Grouped

$$M = l + \frac{h}{f} \left(\frac{N}{2} - CF \right)$$

l = lower limit of median class

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 h = width of " "
 $N = \sum f$
 f = freq of " "
 CF = cumulative freq. of class preceding median class.

Median class = The class having "CF" just greater or equal to $\frac{N}{2}$.

CF = Continuous addition of freq.
table

④ Mode

Grouped / Cont. Freq. Distribution

Modal class = class with max freq.

$$\text{Mode} = l + h \left[\frac{f_m - f_1}{2f_m - f_1 - f_2} \right]$$

l = lower limit of modal class

h = width of " "

f_m = freq. of modal class

f_1 = freq. of class just "above" modal class

f_2 = " " " " "below" " "

⑤ Standard Deviation

$$SD = \sigma = \sqrt{\frac{\sum_{i=1}^N (x_i - \bar{x})^2 \cdot f}{N}}$$

$$\text{Variance} = \sigma^2$$

$$\bar{x} = \frac{\sum x}{N}$$

↓
no. of obs.

$N = \sum f_i$ in grouped data

⑥ Coeff. of Variation

$$CV = \frac{SD}{\bar{x}} \times 100$$

→ measure of dispersion.

⑦ Moments

— Measure of force w.r.t its tendency to provide rotation.

→ Moment about mean

$$\mu_r = \frac{\sum f(x - \bar{x})^r}{\sum f}$$

$$\therefore \mu_1 = \frac{\sum f(x - \bar{x})}{\sum f}; \mu_2 = \frac{\sum f(x - \bar{x})^2}{\sum f}$$

$$\mu_2 = \sigma^2$$

→ Moment about arbitrary origin

$$\mu_r' = \frac{\sum f(x-a)^r}{\sum f}$$

$$\therefore \mu_0' = \frac{\sum f}{\sum f} = 1$$

$$\mu_1' = \frac{\sum f(x-a)}{\sum f} = \bar{x} - a$$

→ Relation b/w moment about mean & moment about arbitrary value

$$(i) \mu_1 = \mu_1' - \mu_1' = 0$$

$$(ii) \mu_2 = \mu_2' - (\mu_1')^2$$

$$(iii) \mu_3 = \mu_3' - 3\mu_2'\mu_1' + 2(\mu_1')^3$$

$$(iv) \mu_4 = \mu_4' - 4\mu_3'\mu_1' + 6\mu_2'\mu_2' - 3(\mu_1')^4$$

⑧ Skewness

— Measure of degree of non symmetry of freq. distribution.

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coeff of skewness = γ

$$\gamma_1 = \sqrt{\beta_1}$$

$$\beta_1 = \frac{\mu_3^2}{\mu_2^3}$$

⑨ Kurtosis

- Measure of thickness of distribution

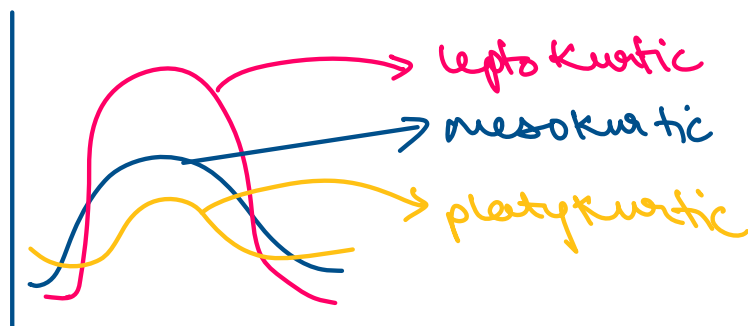
$$\beta_2 = \frac{\mu_4}{\mu_2^2}$$

→ defines peakness of curve

$\beta_2 = 3$ Mesokurtic (Normal)

$\beta_2 > 3$ Leptokurtic (More peak)

$\beta_2 < 3$ Platykurtic (Less peak)



freq. distribution

freq. distribution

10) Stark Deviation Method

Suppose $d = \frac{x-a}{h}$ [h = height of class]

$$\therefore \mu_r' = \frac{\sum f(x-a)^r}{\sum f} \quad \rightarrow x-a = d \cdot h$$

$$\therefore \mu_r' = \frac{\sum f d^r}{\sum f} \cdot h^r$$

$$\mu_1' = h \frac{\sum f d}{\sum f} ; \mu_2' = h^2 \frac{\sum f d^2}{\sum f}$$

$$\mu_3' = h^3 \frac{\sum f d^3}{\sum f}$$

Suppose

$$u_1 = \frac{\sum f d}{\sum f} ; u_2 = \frac{\sum f d^2}{\sum f} ; u_3 = \frac{\sum f d^3}{\sum f}$$

$$\therefore \mu_1' = h u_1 ; \mu_2' = h^2 u_2 ; \mu_3' = h^3 u_3$$

$$\begin{aligned}
 (i) \mu_1 &= 0 \\
 (ii) \mu_2 &= h^2 (u_2 - u_1^2) \\
 (iii) \mu_3 &= h^3 (u_3 - 3u_2u_1 + 2u_1^3) \\
 (iv) \mu_4 &= h^4 (u_4 - 4u_3u_1 + 6u_2u_2 - 3u_1^4)
 \end{aligned}$$

11 Mathematical Expectations

Random variable = A real value assigned to each sample point of the sample space.
 Discrete RV = Random variable takes finite or countable values. Eg coin tossing
 Continuous RV = Random variable taking all possible values. Eg weight of all college students

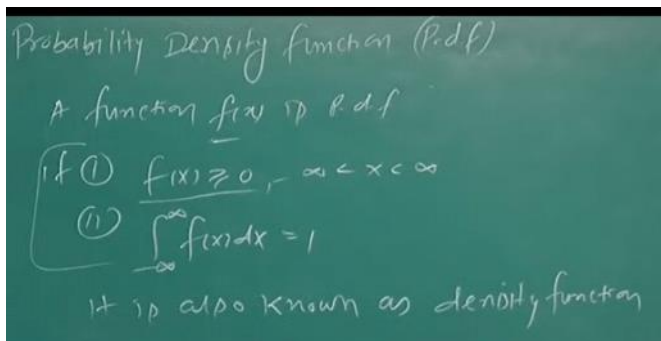
Probability function of $x = f(x)$.

$P(X=x) = f(x)$ is the probability function if it satisfies the following conditions

1. $f(x) \geq 0$ for all values of x
2. Sum of all values of $f(x) = 1$

If X is discrete RV then its probability function is called = Probability mass function

If X is continuous RV then its probability function is called = Probability density function



Mathematical Expectation:

If X is random variable and $f(x)$ is probability function then mathematical expectation is denoted by $E(x)$.

$E(x) = \sum x_i f_i$ from $i=1$ to n

$E(x) = \sum x_i \cdot P(x_i)$ for DRV

$E(x) =$

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

for CRV

$E(x^2) = x^2 \text{ times } P(x)$ for DRV

If $P(X) = x$

$\bar{x} = \text{mean} = E(X)$

$\begin{cases} \text{D.R.V} & E(X) = \sum x P(X) \\ \text{C.R.V} & E(X) = \int_{-\infty}^{\infty} x f(x) dx \end{cases}$

$E(X) = \bar{x}$

Variance = $E(x - \bar{x})^2 = E$

$= E(x^2) - (E(x))^2$

Correlation coefficient = karls

$r = \frac{\text{cov}(x, y)}{\sigma_x \sigma_y}$

$\text{cov}(x, y) = \frac{\sum (x - \bar{x})(y - \bar{y})}{n}$

$\text{cov}(x, y) = E(xy) - E(x)E(y)$
 $= \frac{\sum xy}{n} - \frac{\sum x}{n} \frac{\sum y}{n}$

$\sigma_x = \sqrt{E(x^2) - (E(x))^2} = \sqrt{\frac{\sum x^2}{n} - \left(\frac{\sum x}{n}\right)^2}$

$\sigma_y = \sqrt{E(y^2) - (E(y))^2} = \sqrt{\frac{\sum y^2}{n} - \left(\frac{\sum y}{n}\right)^2}$

$r = \frac{n \sum xy - \sum x \sum y}{\sqrt{n \sum x^2 - (\sum x)^2} \sqrt{n \sum y^2 - (\sum y)^2}}$

Rank correlation = spearsman

$\rho_{xy} = 1 - \frac{6 \sum d^2}{n(n^2 - 1)}$

rank $r = 1 - \frac{6 \sum d^2}{n(n^2 - 1)}$

$d = (R_x - R_y)^2$

Regression = statistical technique to estimate or predict a variable from another known variable.
statistical technique that defines relationship between two variables

Regression lines = when a graph is plotted with two variables which are linearly correlated to each other. So the lines have 2 regression equations as well which are

1. Y on x i.e., y depends on x $\Rightarrow y = a + bx$
2. X on y i.e., x depends on y $\Rightarrow x = a + by$

→ Regression eqⁿ

① y on x : $y - \bar{y} = b_{yx}(x - \bar{x})$

$b_{yx} = r \frac{\sigma_y}{\sigma_x} \rightarrow \text{Regression coeff.}$

$$\boxed{b_{yx} = r \frac{\sigma_y}{\sigma_x}} \rightarrow \text{Regression Coeff.}$$

② x on y :
$$x - \bar{x} = b_{xy} (y - \bar{y})$$

$$\boxed{b_{xy} = r \frac{\sigma_x}{\sigma_y}}$$

$$b_{yx} = \frac{n \sum xy - \sum x \sum y}{n \sum x^2 - (\sum x)^2} = \frac{\text{cov}(x, y)}{\sigma_x^2}$$

$$b_{xy} = \frac{n \sum xy - \sum x \sum y}{n \sum y^2 - (\sum y)^2} = \frac{\text{cov}(x, y)}{\sigma_y^2}$$

→ Properties of Regression Coeff.

① Geometric mean r b/w 2 reg. coeff. is given by:

$$r^2 = b_{xy} \cdot b_{yx}$$

$$\boxed{r = \pm \sqrt{b_{xy} \cdot b_{yx}}}$$

② Both RC will have same sign.

③ Individual values of both PC
can't exceed 1.

* 2 regression lines always intersect
② mid pts. of x & y i.e., (\bar{x}, \bar{y})