

MEAN :

RAW DATA :  $\bar{x} = \frac{\sum x}{n}$

GROUPED DATA :  $\bar{x} = \frac{\sum f_x}{N} = \frac{\sum fx}{\sum f}$

Note : The sum of the deviation of the observation from the AM is always zero.

$$\sum_{i=1}^n (x_i - \bar{x}) = 0$$

COMBINED Mean :  $\bar{x} = \frac{n_1 \bar{x}_1 + n_2 \bar{x}_2}{n_1 + n_2}$

Weighted Arithmetic Mean :  $\bar{x} = \frac{\sum wx}{\sum w}$

where  $w$  is the weight.

Geometric Mean :  $\sqrt[n]{x_1 \times x_2 \times \dots \times x_n}$

RAW DATA :  $\text{Anti log} \left[ \frac{1}{n} \sum \log(x_i) \right]$

GROUPED DATA :  $\text{Anti log} \left[ \frac{1}{\sum f_i} \sum f_i \log(x_i) \right]$

Harmonic Mean :  $\frac{n}{\sum 1/x}$  where  $x \neq 0$ .

Raw data :  $\frac{n}{\sum 1/x}$   $\{x \neq 0\}$

Grouped data :  $\frac{N}{\sum f/x}$  or  $\frac{F}{\sum f/x}$   $\{x \neq 0\}$

MEDIAN :

$$\text{median (m)} = \left( \frac{n+1}{2} \right)^{\text{th}}$$

where  $n$  is number of observations.

Continuous frequency distribution :

$$L_1 + \frac{(N/2 - Pcf) \cdot C}{f}$$

where  $L_1$  = lower limit of median class

$p_{cf}$  = previous cumulative frequency.

$C$  = class width.

$f$  = class frequency.

$N$  = total number of observation.



Mode :

(Z) = most frequent observation .

Continuous frequency distribution :

$$(Z) = L + \frac{(f_1 - f_0) * C}{(2f_1 - f_0 - f_2)}$$

where,  $L$  : lower limit of modal class

$f_1$  : freq. of modal class

$f_0$  : freq. of preceding class

$f_2$  : freq. of next class

$C$  : class width.

	Discrete	Continuous.
Quantile	$Q_i = \{i(n+1)/4\}^{\text{th}}$	$L + \frac{(iN/4 - Pcf) * C}{F}$
Decile	$D_i = \{i(N+1)/10\}^{\text{th}}$	$L + \frac{(iN/10 - Pcf) * C}{F}$
Percentile	$P_i = \{i(N+1)/100\}^{\text{th}}$	$L + \frac{(iN/100 - Pcf) * C}{F}$

Range :

Highest value - lowest value.

co-efficient of Range :  $\frac{H - L}{H + L}$

Quartile Deviation :  $\frac{Q_3 - Q_1}{2}$

co-efficient of Quartile Deviation :  $\frac{Q_3 - Q_1}{Q_3 + Q_1}$

Interquartile Range :  $Q_3 - Q_1$ .

Mean Deviation about Mean.

RAW DATA :  $\bar{x} = \frac{\sum |x - \bar{x}|}{n}$

tabulated data :  $\bar{x} = \frac{\sum f |x - \bar{x}|}{N}$

Relative Measure :  $\frac{M.D.(\bar{x})}{\bar{x}}$   
co-efficient of M.D.  $(\bar{x})$



## Mean deviation about Median

For Rawdata : 
$$\frac{\sum |x - M|}{n}$$

For tabulated data : 
$$\frac{\sum f |x - M|}{N}$$

Relative Measure : 
$$\frac{M.D.(M)}{M}$$
  
coef. of M.D (M)

## Standard deviation .

	Rawdata	Tabulated data .
for population :	$\sigma = \sqrt{\frac{\sum (x - \bar{x})^2}{n}}$	$\sigma = \sqrt{\frac{\sum f(x - \bar{x})^2}{N}}$

for sample :

$$\sigma = \sqrt{\frac{\sum (x - \bar{x})^2}{n-1}} \quad \sigma = \sqrt{\frac{\sum f(x - \bar{x})^2}{N-1}}$$

## Co-efficient of Variation (C.V)

$$\begin{aligned} C.V &= \frac{\text{Standard deviation}}{\text{Arithmetic Mean}} \times 100 \\ &= \frac{\sigma}{\bar{x}} \times 100. \end{aligned}$$

## Central Moments :

$$\text{Raw data} = \mu_H : \frac{\sum (x - \bar{x})^H}{n}$$

$$\text{Tabulated data} = \mu_H : \frac{\sum f (x - \bar{x})^H}{N}$$

## Raw Moments

$$\text{Raw data} : \mu_H^x = \frac{\sum (x - a)^H}{n}$$

$$\text{tabulated data} : \mu_H^x = \frac{\sum f (x - a)^H}{N}$$

where 'a' is any constant expect 'am'.

$$\therefore \text{mean} = \text{median} = \text{mode} \quad \{ \text{symmetric} \}$$

$$\text{mode} < \text{median} < \text{mean} \quad \{ \text{+ve skewed} \}$$

$$\text{mean} < \text{median} < \text{mode} \quad \{ \text{-ve skewed} \}$$



## Skewness Based on Moments

$$\beta_1 = \frac{\mu_3^2}{\mu_2^3} \quad \text{or} \quad \gamma_1 = +\sqrt{\beta_1}$$

Karl Pearson's co-efficient of skewness

$$S = \frac{\text{mean} - \text{mode}}{\text{S.D.}} = \frac{\bar{x} - Z}{\sigma}$$

If mode is ill defined,

$$S = \frac{3 (\text{mean} - \text{median})}{\text{S.D.}} = \frac{3 (\bar{x} - M)}{\sigma}$$

Since we know,  $Z = 3M = 2\bar{x}$

Bowley's Co-efficient of skewness (Galton's).

$$S = \frac{(Q_3 - Q_2) - (Q_2 - Q_1)}{(Q_3 - Q_2) + (Q_2 - Q_1)}$$

Kurtosis (Based on Moments)

$$\beta_2 = \frac{\mu_4}{\mu_2^2} \quad \text{or} \quad \gamma_2 = \beta_2 - 3$$

$\beta_2$

3

Mesokurtic (Normal)

< 3

Platykurtic

> 3

Leptokurtic

Sampling Distribution :- Mean

$$\sigma_x = \frac{\sigma}{\sqrt{n}}$$

Where the population SD is known

$$\bar{x} - Z_{\alpha/2} \frac{\sigma}{\sqrt{n}} < \mu < \bar{x} + Z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$$

- Where sample size is large, and population SD is not known

$$\bar{x} - Z_{\alpha/2} \frac{S}{\sqrt{n}} < \mu < \bar{x} + Z_{\alpha/2} \frac{S}{\sqrt{n}}$$

- ~~Where~~ where sample size is  $< 30$ , student t-distribution is used (degrees of freedom =  $n-1$ )

$$\bar{x} - t_{\alpha/2} \frac{S}{\sqrt{n}} < \mu < \bar{x} + t_{\alpha/2} \frac{S}{\sqrt{n}}$$

the  
In Case of sampling distribution of the mean, the margin of error is:

$$E = Z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}$$

Therefore

$$n = \left( Z_{\alpha/2} \cdot \sigma / E \right)^2$$



- Variance

$$\sigma^2 = \frac{\sum_{i=1}^N (Y_i - \mu)^2}{N}$$

(Confidence interval :- Variance and SD)

Chi-squared CDF is given by:

$$\chi^2 = \frac{(n-1)s^2}{\sigma^2}$$

C.I.

$$\frac{(n-1)s^2}{\chi^2_{\alpha/2}} < \sigma^2 < \frac{(n-1)s^2}{\chi^2_{1-\alpha/2}}$$

$$\left( \frac{(n-1)s^2}{\chi^2_{\alpha/2}} \right)^{1/2} < \sigma < \left( \frac{(n-1)s^2}{\chi^2_{1-\alpha/2}} \right)^{1/2}$$

Confidence Interval :- Proportions :-  
Test statistics :-

$$Z = \frac{\hat{p} - p}{\sqrt{\hat{p}(1-\hat{p})/n}}$$

Co.J.

$$\hat{p} - Z_{\alpha/2} \sqrt{\hat{p}(1-\hat{p})/n} < p < \hat{p} + Z_{\alpha/2} \sqrt{\hat{p}(1-\hat{p})/n}$$

Co.J. Difference between means  
• population variances known

$$(\bar{X} - \bar{Y}) - Z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

$$< \mu_1 - \mu_2$$

$$< (\bar{X} - \bar{Y}) + Z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

• population variances not known  
; Sample > 30

$$(\bar{X} - \bar{Y}) - Z \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

$$< \mu_1 - \mu_2$$

$$< (\bar{X} - \bar{Y}) + Z \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

• population variances not known,  
Sample < 30

$$(\bar{X} - \bar{Y}) - t \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

$$< \mu_1 - \mu_2$$

$$< (\bar{X} - \bar{Y}) + t \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$



$$D.O.F. = \frac{\left( \frac{S_1^2}{n_1} + \frac{S_2^2}{n_2} \right)^2}{\frac{\left( \frac{S_1^2}{n_1} \right)^2}{(n_1-1)} + \frac{\left( \frac{S_2^2}{n_2} \right)^2}{(n_2-1)}}$$

Confidence Interval

:- Difference between Means

$$S_{pooled}^2 = \frac{(n_1-1)S_1^2 + (n_2-1)S_2^2}{(n_1+n_2-2)}$$

if  $\frac{S_1}{S_2}$  = between 0.5 and 2  
we are using above formula

$$T = \frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)}{S_{pooled} \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

$$D.O.F. = (n_1 + n_2 - 2)$$

C.I. =

$$(\bar{X} - \bar{Y}) - t_{\alpha/2} \cdot S_{pooled} \sqrt{\frac{1}{n_1} + \frac{1}{n_2}} < \mu_1 - \mu_2 < (\bar{X} - \bar{Y}) + t_{\alpha/2} \cdot S_{pooled} \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$$

## C.I. of Difference of Means of "paired samples"

$$d_i = X_i - Y_i, \quad i = 1, 2, \dots, n$$

$$\bar{d} = \left( \frac{1}{n} \right) \sum_{i=1}^n d_i, \text{ and}$$

$$S_d^2 = \left\{ \frac{1}{(n-1)} \right\} \sum_{i=1}^n (d_i - \bar{d})^2$$

$$T = \frac{\bar{d} - (\mu_1 - \mu_2)}{S_d / \sqrt{n}} \quad \text{D.O.F.} = n-1$$

$$\bar{d} - t_{\alpha/2} S_d / \sqrt{n} < \mu_d < \bar{d} + t_{\alpha/2} \frac{S_d}{\sqrt{n}}$$

Confidence Interval: Ratio of Variances

$$\text{Test statistic} = \left( \frac{S_1^2}{\sigma_1^2} \right) / \left( \frac{S_2^2}{\sigma_2^2} \right) = F = \frac{U/r_1}{V/r_2}$$

$$U = \frac{(n_1-1)S_1^2}{\sigma_1^2} \quad V = \frac{(n_2-1)S_2^2}{\sigma_2^2}$$

$$r_1 = (n_1-1) \text{ \& } r_2 = (n_2-1)$$

C.I.

$$\frac{S_2^2}{S_1^2} \times F(\alpha/2, r_2, r_1)$$

$$< \frac{\sigma_2^2}{\sigma_1^2} < \frac{S_2^2}{S_1^2} \times F(1-\frac{\alpha}{2}, r_2, r_1)$$



## Hypothesis Testing about one proportion

Best Estimate  $\hat{p} = x/n$

standard deviation  $\sigma_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}}$

### Conditions

- The sample is a simple random sample
- Sample values are independent of each other.
- $np(1-p) \geq 10$

Test statistic for proportions

$$Z = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}}$$

## Hypothesis Testing: Single parameter

The mean, when variance is known

Test statistic

$$Z = \frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}}$$

The Mean, when variance is unknown

- Large Sample Size  $Z = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$  (normal dist)

- Small sample size  
 $T = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$  (t-distribution)

The Variance

- Test statistic

$$\chi^2 = \frac{(n-1)s^2}{\sigma^2} \text{ (chi-squared)}$$

• The proportion

test statistic

$$Z = \frac{\hat{p} - p_0}{\sqrt{p_0(1-p_0)/n}}$$

Between-column variance

$$\sigma^2 = \frac{\sum n_j (\bar{x}_j - \bar{\bar{x}})^2}{k-1}$$

Variances within the sample

$$\sigma^2 = \sum \left( \frac{n_j - 1}{n_T - k} \right) s_j^2$$

$$F = \frac{MSB}{MSE}$$