

Formula Sheet

Range (R) = *highest(data value) - lowest(data value)*

Class width = $R / (\text{no. of Class})$

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

For pie graph: Degrees = $\frac{f}{n} \times 360^\circ$ Or % = $\frac{f}{n} \times 100\%$

Sample Mean $\bar{X} = \frac{\sum_{i=1}^n X_i}{n}$

Population Mean $\mu = \frac{\sum_{i=1}^n X_i}{N}$

Sample Mean (with frequency) $\bar{X} = \frac{\sum_{i=1}^n f_i X_i}{n}$

Midrange MR = $\frac{\text{lowest value} + \text{highest value}}{2}$

Weighted Mean $\bar{X} = \frac{\sum_{i=1}^n w_i X_i}{\sum_{i=1}^n w_i}$

Harmonic Mean HM = $\frac{n}{\sum_{i=1}^n (1/X_i)}$

Geometric Mean GM = $\sqrt[n]{(X_1)(X_2)(X_3) \cdots (X_n)}$

Quadratic Mean QM = $\sqrt{\left(\frac{\sum_{i=1}^n X_i^2}{n}\right)}$

Sample Variance $s^2 = \frac{\sum_{i=1}^n X_i^2 - [(\sum_{i=1}^n X_i)^2 / n]}{n-1}$

Standard Deviation $s = \sqrt{s^2}$

Sample Variance (with frequency) $s^2 = \frac{\sum_{i=1}^n f_i X_i^2 - [(\sum_{i=1}^n f_i X_i)^2 / n]}{n-1}$

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

Coefficient of Variation CV = $\frac{\text{standard deviation}}{\text{mean}} \times 100\%$

Chebyshev's theorem: The proportion of values from a data set that will fall within k standard deviations of the mean will be at least $1 - \frac{1}{k^2}$, where k is a number greater than 1

Z-score = $\frac{\text{data value} - \text{mean}}{\text{standard deviation}}$

percentile rank of X = $\frac{(\text{no. of data values below X})}{\text{total no. of data values}} \times 100\%$

Finding k^{th} percentile data value, $L = (\frac{k}{100}) \times (n)$, where $n = \text{total no. of values}$ and $k = \text{percentile}$. If L is a whole number then k^{th} percentile is the average of the L^{th} and $(L + 1)^{\text{th}}$ data values. If L has a decimal value, round L up then k^{th} percentile is the rounded L^{th} data value.

Interquartile Range (IQR), $IQR = Q_3 - Q_1$, where Q_1 is the first quartile and Q_3 is the third quartile

To check the outliers, data value which is smaller than $Q_1 - 1.5 \times (IQR)$ or larger than $Q_3 + 1.5 \times (IQR)$

Permutation: The arrangements of n objects in a specific order using r objects at a time is called a permutation of n objects taking r objects at a time, i.e., $nPr = \frac{n!}{(n-r)!}$

Combination: The number of combinations of r objects selected from n objects is obtained by $nCr = \frac{n!}{r!(n-r)!}$

The conditional probability of an event B given an event A is $P(B|A) = \frac{P(A \text{ and } B)}{P(A)}$

Expected value of a discrete random variable is

$$E(X) = \sum XP(X)$$

$$V(X) = \sum X^2P(X) - [\sum XP(X)]^2$$

In a binomial experiment, the probability of getting exactly X successes in n trials is

$$P(X) = \frac{n!}{(n-X)!X!} \cdot p^X q^{n-X}, \quad X = 0, 1, \dots, n, \quad E(X) = np, \quad V(X) = npq$$

The Poisson distribution with parameter λ is

$$P(X; \lambda) = \frac{e^{-\lambda} \lambda^X}{X!}, \quad \text{where } X = 0, 1, 2, \dots, \quad E(X) = \lambda, \quad V(X) = \lambda$$

For Hypergeometric distribution

$$P(X) = \frac{\binom{a}{X} \binom{b}{n-X}}{\binom{a+b}{n}}$$

Normal distribution of a continuous random variable y

$$f(y|\mu, \sigma) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{y-\mu}{\sigma}\right)^2}$$

$$E(y) = \mu$$

$$V(y) = \sigma^2$$

Standard Normal distribution of a continuous random variable y

$$f(y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}}$$

Large sample $100(1 - \alpha)\%$ confidence interval for μ is $\bar{X} \pm z_{\alpha/2} (\sigma/\sqrt{n})$

Small sample $100(1 - \alpha)\%$ confidence interval for μ is $\bar{X} \pm t_{\alpha/2, n-1} (s/\sqrt{n})$

Large sample $100(1 - \alpha)\%$ confidence interval for p is $\hat{p} \pm z_{\alpha/2} (\sqrt{\hat{p}\hat{q}/n})$

Sample size $n = \left(\frac{z_{\alpha/2} \cdot \sigma}{E}\right)^2$, E is the minimum error of estimation

When σ is known the test-statistic $Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$

When σ is unknown the test-statistic $t = \frac{\bar{X} - \mu}{s/\sqrt{n}}$

Test-statistic with a specific population proportion $Z = \frac{\hat{p} - p}{\sqrt{pq/n}}$

Confidence interval for σ^2 is $\frac{(n-1)s^2}{\chi_R^2} < \sigma^2 < \frac{(n-1)s^2}{\chi_L^2}$

Test-statistic for testing a claim about σ or σ^2 is $\chi^2 = \frac{(n-1)s^2}{\sigma^2}$

When population variances are known the test-statistic $Z = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\left(\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}\right)}}$

Confidence interval for the difference between two population means is $(\bar{X}_1 - \bar{X}_2) \pm z_{\alpha/2} \sqrt{\left(\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}\right)}$

Z-test for the difference between two population proportions, $Z = \frac{(\hat{p}_1 - \hat{p}_2) - (p_1 - p_2)}{\sqrt{\hat{p}\hat{q}\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}}$

where $\bar{p} = \frac{X_1 + X_2}{n_1 + n_2}$,

$\hat{p}_1 = \frac{X_1}{n_1}$, $\hat{p}_2 = \frac{X_2}{n_2}$, $\bar{q} = 1 - \bar{p}$

Confidence interval for the difference between two proportions is $(\hat{p}_1 - \hat{p}_2) \pm z_{\alpha/2} \sqrt{\left(\frac{\hat{p}_1\hat{q}_1}{n_1} + \frac{\hat{p}_2\hat{q}_2}{n_2}\right)}$

Correlation Co-efficient $r = \frac{n(\sum xy) - (\sum x)(\sum y)}{\sqrt{[n(\sum x^2) - (\sum x)^2][n(\sum y^2) - (\sum y)^2]}}$

$t = r \sqrt{\frac{n-2}{1-r^2}}$ for testing $H_0 : \rho = 0$ and $H_1 : \rho \neq 0$.

For Regression $a = \frac{(\sum y)(\sum x^2) - (\sum x)(\sum xy)}{n(\sum x^2) - (\sum x)^2}$ $b = \frac{n(\sum xy) - (\sum x)(\sum y)}{n(\sum x^2) - (\sum x)^2}$