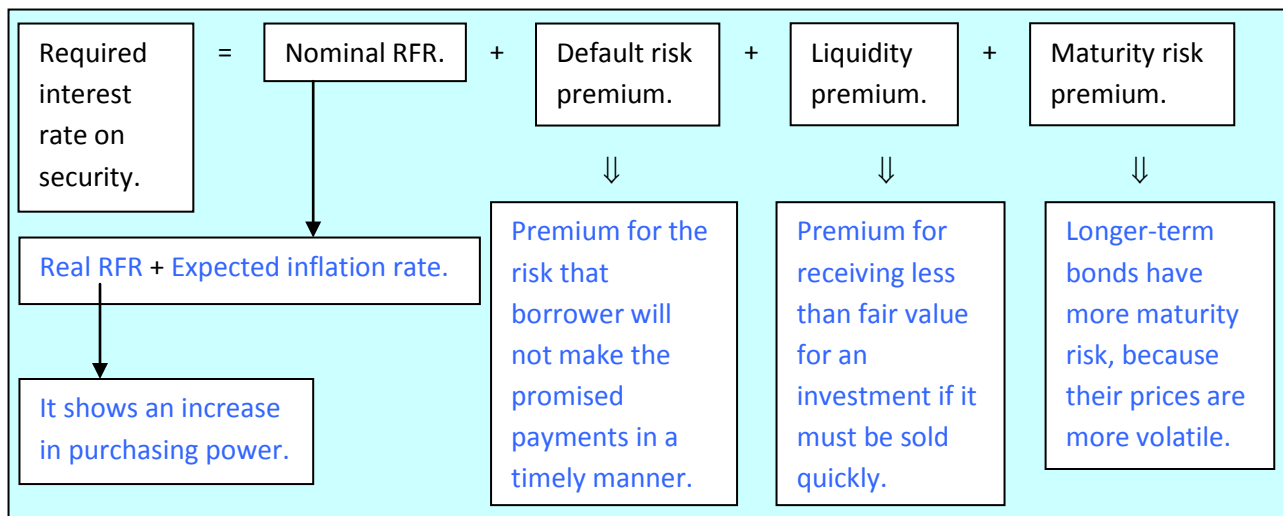


# “The Time Value of Money”

<p><b><u>Compound Interest or Interest on Interest</u></b></p> <p>Growth in the value of investment includes, interest earned on:</p> <ul style="list-style-type: none"> <li>• Original principal.</li> <li>• Previous period’s interest earnings.</li> </ul>	<p><b><u>Time Line</u></b></p> <p>Diagram of the cash flows associated with a TVM problem.</p>	<p><b><u>Discounting</u></b></p> <p>Moving CF to the beginning of an investment period to calculate PV.</p> $PV = \frac{FV}{(1+i)^N}$ <p style="text-align: center;"><math>\frac{1}{(1+i)^N}</math> is PV factor</p>	<p><b><u>Compounding</u></b></p> <p>Moving cash flow to the end of the investment period to calculate FV.</p> $FV = PV (1+i)^N$ <p>(1+i)<sup>N</sup> is FV factor</p>
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<p><b><u>Loan Amortization</u></b></p> <p>Process of paying off a loan with a series of periodic loan payments, whereby a portion of the outstanding loan amount is paid off, or amortized, with each payment.</p>	<p><b><u>Perpetuity</u></b></p> <ul style="list-style-type: none"> <li>• Perpetual annuity.</li> <li>• Fixed payment at set intervals over an infinite time period.</li> <li>• <math>\frac{1}{i}</math> is the discounting factor for perpetuity.</li> </ul>	<p><b><u>Annuity</u></b></p> <p>Stream of equal cash flows accruing at equal intervals.</p>
<p><b><u>Cash flow Additivity Principle</u></b></p> <p>PV of any stream of cash flows equals the sum of PV of each cash flow.</p>	<p>PV of annuity due &gt; PV of ordinary annuity.</p>	<p style="text-align: center;">↓</p> <p style="text-align: center;">Two types</p> <p style="text-align: center;">⇐      ⇒</p> <div style="display: flex; justify-content: space-between;"> <div style="width: 45%; padding: 5px;"> <p><b><u>Annuity Due</u></b></p> <p>Cash flows occur at the beginning of each period.</p> </div> <div style="width: 45%; padding: 5px;"> <p><b><u>Ordinary Annuity</u></b></p> <p>Cash flows occur at the end of each period.</p> </div> </div>

<p style="text-align: center;"><b><u>Interpretations of Interest Rate</u></b></p> <ul style="list-style-type: none"> <li>• Required rate of return.</li> <li>• Discount rate.</li> <li>• Opportunity cost.</li> </ul>
<p style="text-align: center;"><b><u>Effective Annual Rate (EAR)</u></b></p> <ul style="list-style-type: none"> <li>• Rate of return actually being earned after adjustments have been made for different compounding periods.</li> <li>• <math>EAR = (1 + \text{periodic rate})^m - 1</math></li> <li>• Stated rate will be equal to the actual (effective) rate only when it is compounded annually.</li> </ul>

# “Discounted Cash Flow Applications”

D.R. = Discount Rate.

### NPV:

• PV of expected cash inflows – PV of expected cash outflows.

$$NPV = \sum_{t=0}^N \frac{CF_t}{(1+r)^t}$$

- D.R used is the market based opportunity cost of capital.
- NPV assumes reinvestment at D.R.

### Decision rule:

NPV	Decision	Impact
+ve (IRR > r)	Accept	Increase shareholder's wealth.
0	-	Size of the company rises but shareholder's wealth is not affected.
-ve (IRR < r)	Reject	Decreases shareholder's wealth.

### IRR

- The D.R. at which NPV = 0.
- The rate of return at which; PV inflows = PV outflows.
- It assumes reinvestment at IRR.

#### For Single Project:

IRR / NPV rules lead to exactly the same accept /reject decision

#### Decision rule

- $IRR > r \Rightarrow$  Accept
- $IRR < r \Rightarrow$  Reject

#### For Mutually Exclusive Project:

Select the project with the greatest NPV.

### Problems in IRR

For mutually exclusive projects, NPV & IRR may give conflicting project rankings due to:

Different sizes of project's initial cost

Differences in timings of cash flows.

### Money-Weighted Return (MWR)

- IRR of an investment.
- It is highly sensitive to the timing & amount of withdrawals from & additions to a portfolio.
- If one has complete control over the timings of cash flows then it is more appropriate.

### Time-Weighted Return (TWR)

- It measures compound growth (Geometric return).
- It is not affected by the timings of cash flows.
- It is preferred over MWR.
- TWR can't be calculated if we don't know the period end values of investment.

Funds contributed prior to a period of relatively..

Poor returns

MWR < TWR

High returns

MWR > TWR

### Holding Period Yield or Holding Period Return (HPR)

- Total return an investor earns between the purchase date & the sale or maturity date.

$$HPY \text{ or } HPR = \frac{P_1 - P_0 + D_1}{P_0} = \frac{P_1 + D_1}{P_0} - 1$$

- It is the actual return an investor receives.

### Money Market Yield (CD equivalent yield) $r_{MM}$

- Annualized HPY.
- Assumes 360-day year.
- Does not incorporate effects of compounding.
- $r_{MM} = HPY \times \frac{360}{t}$

### Effective Annual Yield (EAY)

- Annualized HPY.
- Assumes 365-day year.
- Incorporates effects of compounding.
- $EAY = (1 + HPY)^{365/t} - 1$

### Bank Discount Yield ( $r_{BD}$ )

- Dollar discount from the face (par) value as a fraction of the face value.
- Based on face value.
- Not based on market or purchase price.
- $r_{BD} = \frac{D}{F} \times \frac{360}{t}$

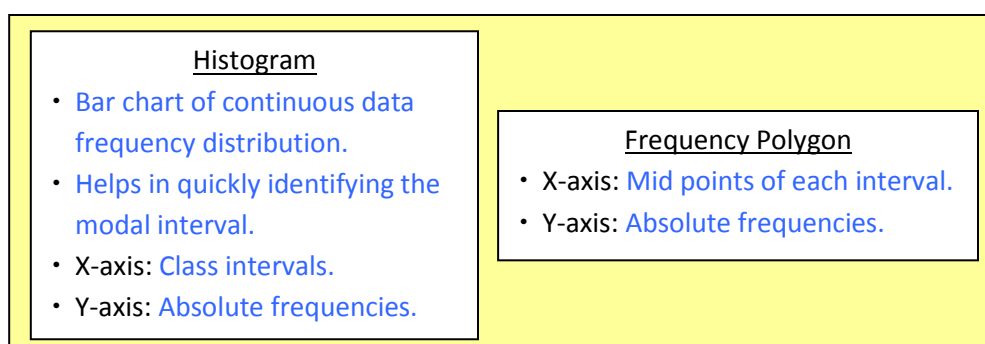
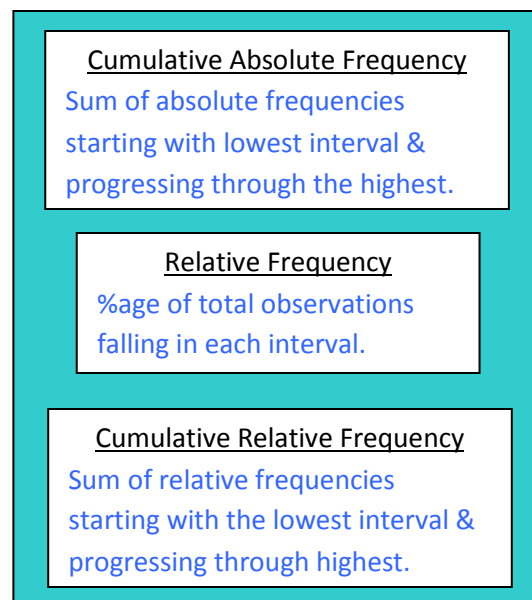
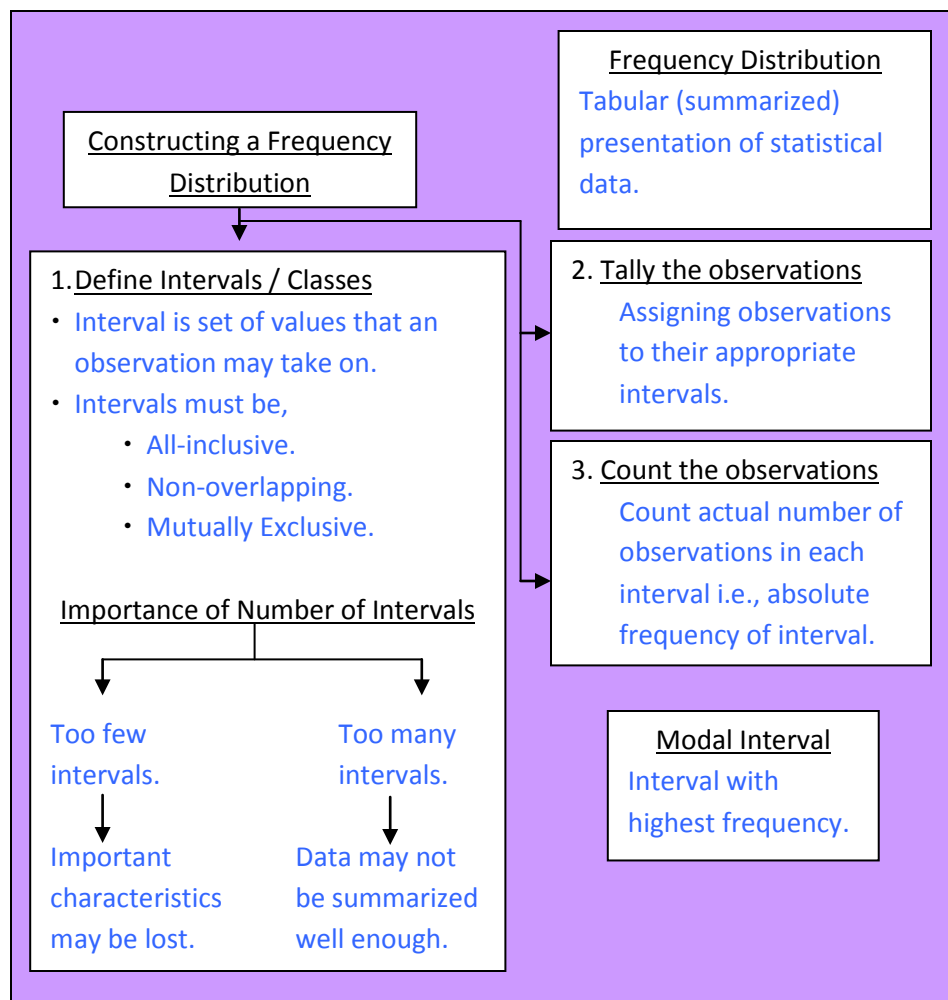
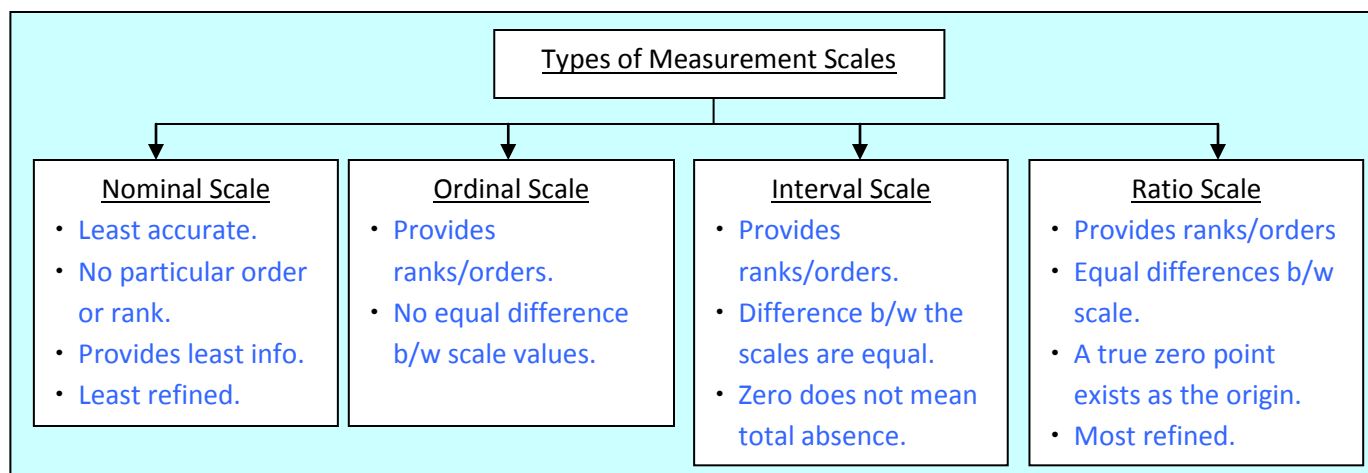
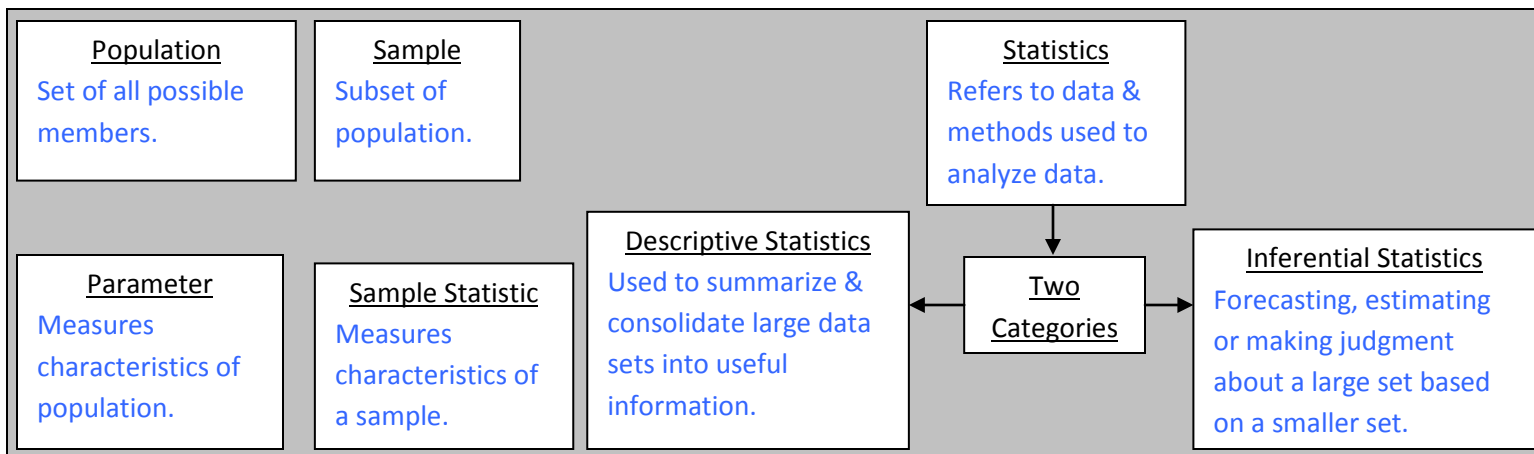
### Converting $r_{BD}$ into $r_{MM}$

$$r_{MM} = \frac{360 \times r_{BD}}{360 - (t \times r_{BD})}$$

### Bond -equivalent yield (BEY)

$$BEY = 2 \times (\text{semi annual effective yield.})$$

# “Statistical Concepts & Market Returns”



**Measures of Central Tendency**

- Identify centre of data set.
- Measure of reward.
- Used to represent typical or expected values in data set.

**Weighted Mean**

It recognizes the disproportionate influence of different observations on mean.

$$\bar{X}_w = \sum_{i=1}^n w_i X_i; \sum_{i=1}^n w_i = 1$$

**Geometric Mean (GM)**

- Calculating multi periods return.
- Measuring compound growth rates.

$$G = \sqrt[n]{X_1 \times X_2 \dots \times X_n}$$

(applicable only to non-negative values)

$$1 + R_G = \sqrt[n]{(1 + R_1) \times (1 + R_2) \times \dots \times (1 + R_n)}$$

**Mean**

- Sum of all values divided by total number of values.
- Population =  $\frac{\sum X}{N} = \mu$
- Sample =  $\frac{\sum x}{n} = \bar{x}$

**Properties:**

- Mean includes all values of data set.
- Mean is unique for each data.
- All intervals & ratio data sets have a Mean.
- Sum of deviations from Mean is always zero i.e.,  $\sum(x - \bar{x}) = 0$
- Mean is the best estimate of true mean & the value of next observation.

**Shortcoming:**

- Mean is affected by extremely large & small values.

**Median**

- Midpoint of an arranged data set.
- Divides data into two equal halves.
- It is not affected by extreme values; hence it is a better measure of central tendency in the presence of extremely large or small values.

**Mode**

- Most frequent value in the data set.

**No. of Modes**      **Names of Distributions**

- One      → Unimodal
- Two     → Bimodal
- Three   → Trimodal

**Harmonic Mean (H.M)**

$$H.M = \frac{N}{\sum \left[ \frac{1}{x} \right]}$$

**H.M is used:**

- When time is involved.
- Equal \$ investment at different times.

**For values that are not all equal**

$$H.M < GM < AM$$

**Quantiles:**

$$L_y = [n - 1] \frac{y}{100}$$

**Quartiles:** Distribution divided into 4 parts (quarters).

**Quintiles:** Distribution divided into 5 parts.

**Deciles:** Distribution divided into 10 parts.

**Percentiles:** Distribution divided into 100 parts (percents).

Measures of Location      Measures of Central + Quantiles Tendency  
 ⇒

**Dispersion**

- Variability around the central tendency.
- Measure of risk.

**Mean Absolute Deviation (MAD)**

Average of absolute deviations from mean:

$$MAD = \frac{\sum |X - \bar{X}|}{n}$$

In general S.D > MAD

**Relative Dispersion**

Amount of variability relative to a reference point.

**Range**

Max Value — Min Value

**Sample Variance**

$$s^2 = \frac{\sum (X - \bar{X})^2}{n - 1}$$

**Sample Standard Deviation**

$$s = \sqrt{\frac{\sum (X - \bar{X})^2}{n - 1}}$$

**Population Variance ( $\sigma^2$ )**

Averaged squared deviations from mean.

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

**Population Standard Deviation (S.D) ' $\sigma$ '.**

Square root of population variance.

$$\sigma = \sqrt{\frac{\sum (x - \mu)^2}{N}}$$

**Using 'n-1' observations**

Using entire number of observations 'n' will systemically underestimate the population parameter & cause the sample variance & S.D to be referred to as **biased estimator.**

**Coefficient of Variation**

- CV =  $\frac{s}{\bar{x}}$  i.e., risk per unit of expected return.
- Helps make direct comparisons of dispersion across different data sets.

**Sharpe Ratio**

- Measures excess return per unit of risk.

$$\frac{\bar{r}_p - r_f}{\sigma_p}$$

- Sharpe ratio =
- Higher sharpe ratios are preferred.

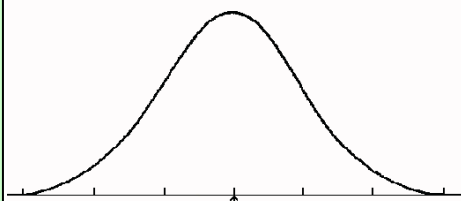
**Chebyshev's Inequality**

Gives the %age of observations that lie within 'k' standard deviations of the mean is at least  $1 - \frac{1}{k^2}$  for all  $k > 1$ , regardless of the shape of the distribution.

± 1.25 s.d	⇒	36% obs.
± 1.5 s.d	⇒	56% obs.
± 2 s.d	⇒	75% obs.
± 3 s.d	⇒	89% obs.
± 4 s.d	⇒	94% obs.

**Symmetrical Distribution**

- Identical on both sides of the mean.
- Intervals of losses & gains exhibit the same frequency.
- Mean = Median = Mode.



Mean = Median = Mode.

**Skewness**

Non Symmetrical.

**Sample Skewness**

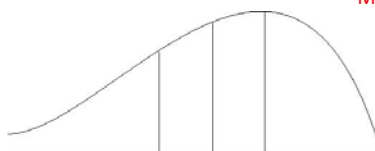
- Sum of cubed deviations from mean divided by number of observations & cubed standard deviation.

$$s_k = \frac{1}{n} \frac{\sum (x - \bar{x})^3}{s^3}$$

- $|s_k| > 0.5$  is considered significant level of skewness.

**Negatively Skewed**

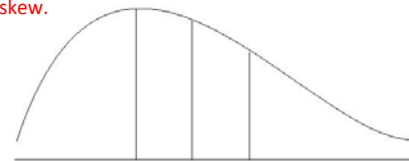
- Longer tail towards left.
- More outliers in the lower region.
- More -ve deviations.
- Mean < Median < Mode



Mean Median Mode

**Positively Skewed**

- Longer tail towards right.
- More outliers in the upper region.
- More +ve deviations.
- Mean > Median > Mode



Mode Median Mean

**Hint**

- Median is always in the center.
- Mean is in the direction of skew.

**Kurtosis**

- Measure of degree of more or less peaked than a normal distribution.
- Kurtosis of normal distribution is 3.
- Excess kurtosis = sample kurtosis-3

where, sample kurtosis =  $\frac{1}{n} \frac{\sum (x - \bar{x})^4}{s^4}$

- Excess kurtosis value exceeding absolute 1 is considered large.

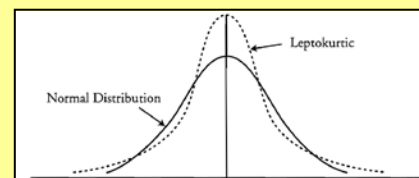
**Distribution**

**Excess Kurtosis**

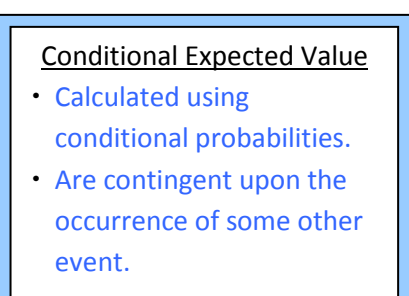
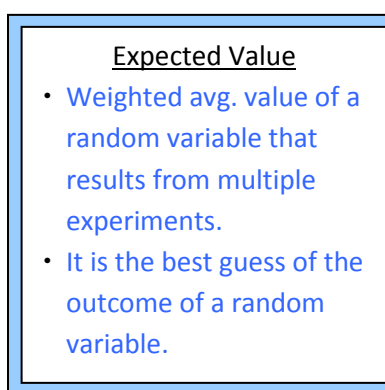
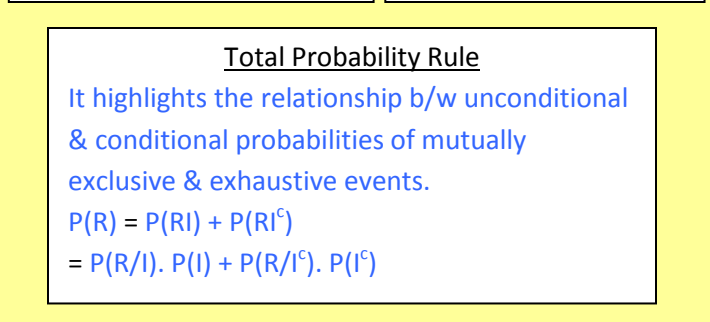
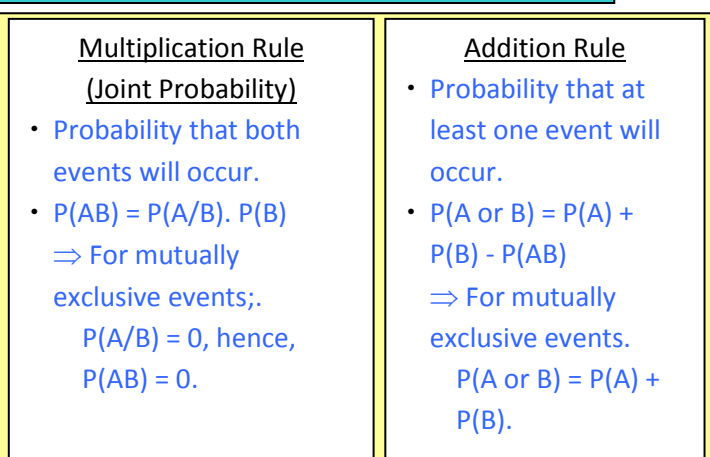
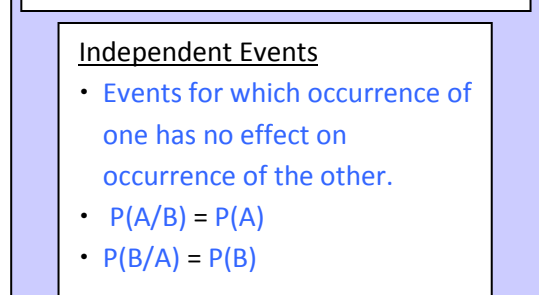
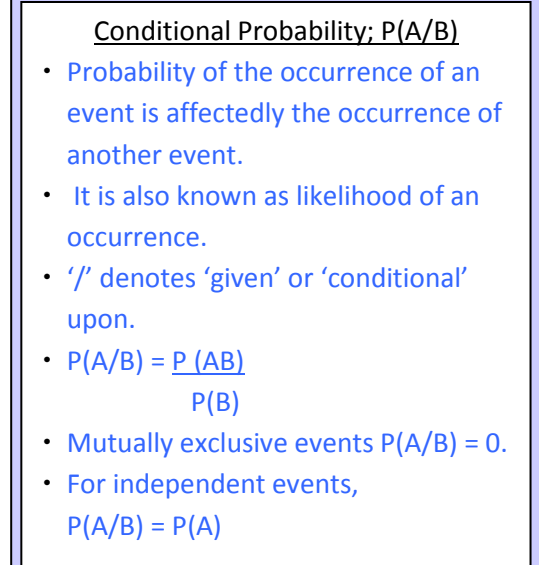
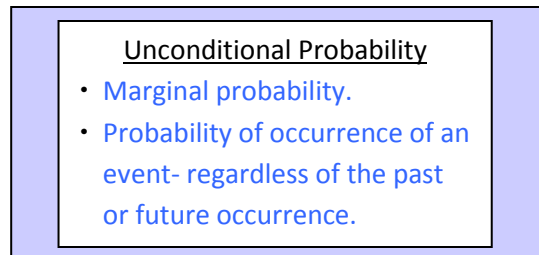
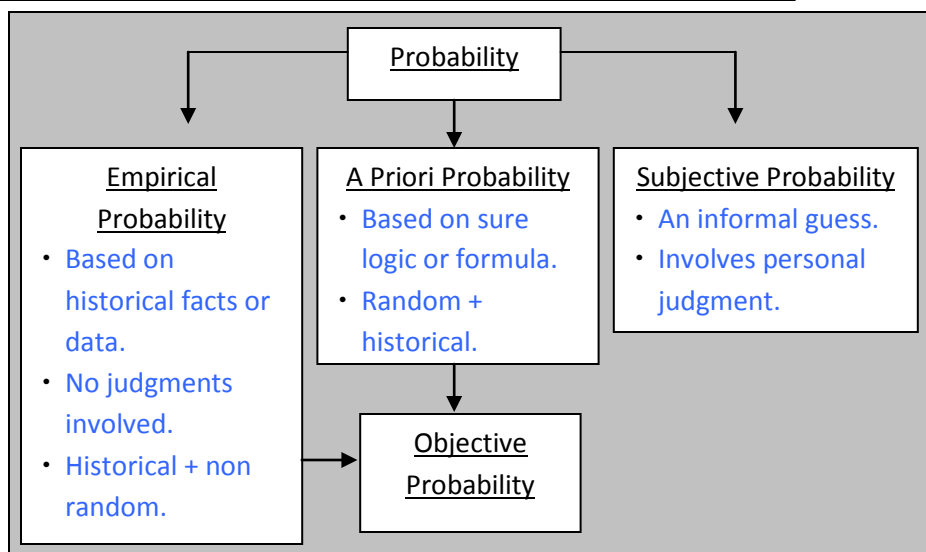
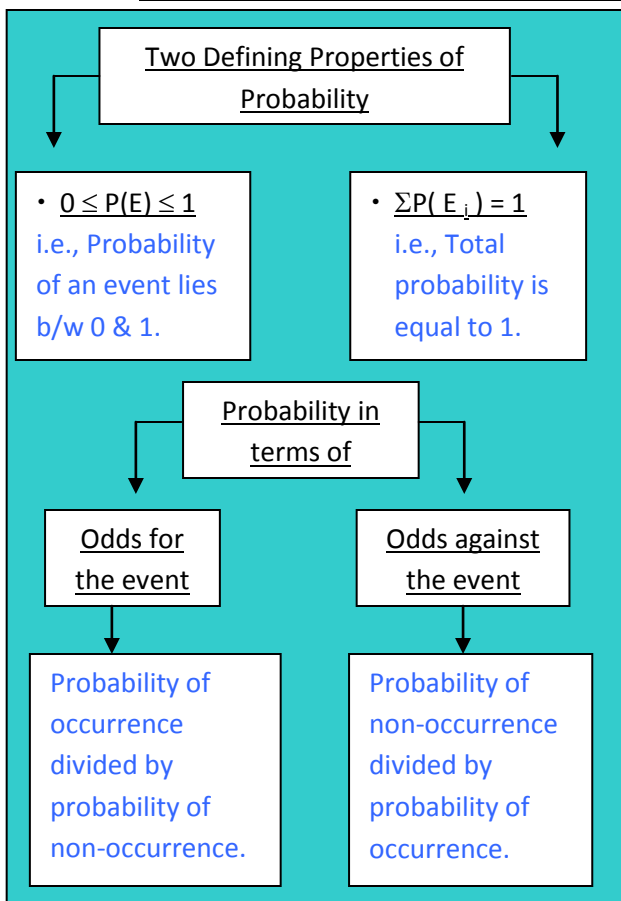
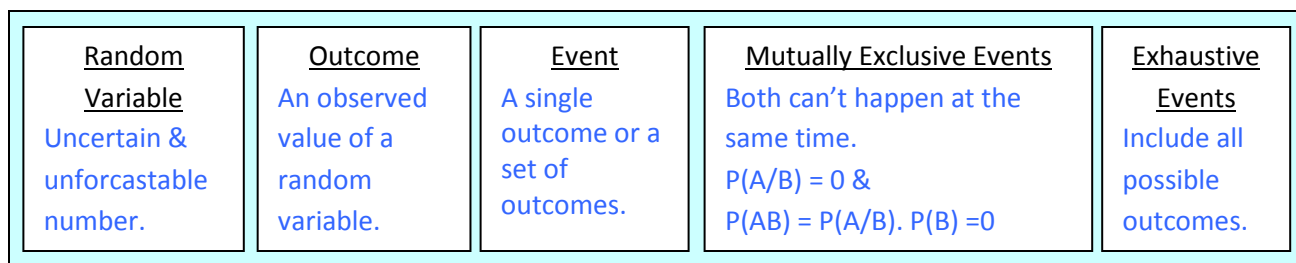
Leptokurtic	⇒	>0
Mesokurtic (Normal)	⇒	=0
Platykurtic	⇒	<0

Greater +ve kurtosis & more -ve skew. ⇒ Increased Risk.

In Leptokurtic distribution, there is a greater probability of both, very small & very large deviation from mean.



# "Probability Concepts"



**Covariance**

- Measure of how two assets move together.
- It measures only direction.
- $-\infty \leq \text{Cov}(x, y) \leq +\infty$  (property).
- It is measured in squared units.
- $\text{Cov}(R_i, R_j) = E \{ [R_i - E(R_i)] [R_j - E(R_j)] \}$   
 $= \sum P(S) [R_i - E(R_i)] [R_j - E(R_j)]$
- $\text{Cov}(R_A, R_A) = \text{variance}(R_A)$  (property).

Covariance	Variables tend to
+ve $\Rightarrow$	Move in same direction.
-ve $\Rightarrow$	Move in opposite direction.
'0' $\Rightarrow$	No linear relationship.

**Correlation**

- Measures the direction as well as the magnitude.
- It is a standardized measure of co-movement.
- It has no units.
- $-1 \leq \text{corr}(R_i, R_j) \leq +1$ .

Value	Correlation	Variables tend to
+1 $\Rightarrow$	Perfectly positive $\Rightarrow$	Move proportionally in the same direction.
-1 $\Rightarrow$	Perfectly negative $\Rightarrow$	Move proportionally in the opposite direction.
0 $\Rightarrow$	Uncorrelated $\Rightarrow$	No linear relationship.

$$\text{Corr}(R_i, R_j) = \frac{\text{Cov}(R_i, R_j)}{\sigma(R_i) \sigma(R_j)}$$

**Portfolio**

**Expected Value**

$$E(R_p) = \sum_{i=1}^N w_i E(R_i)$$

**Variance**

$$\text{Var}(R_p) = \sum_{i=1}^N w_i^2 \sigma_i^2 + \sum_{i=1}^N \sum_{j=1}^N w_i w_j \text{Cov}_{ij}$$

$\Rightarrow$  Where  $w_i = \frac{\text{market value of investment in asset 'i'}}{\text{market value of the portfolio}}$

**Baye's Formula**

$\Rightarrow$  Used to update a given set of prior probabilities in response to the arrival of new information.

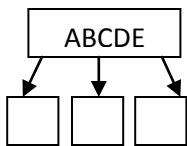
$$\text{Updated Probability} = \frac{\text{probability of new info.} \times \text{prior probability of unconditional event.}}{\text{probability of new info.}}$$

**Counting Methods**

**Labeling Formula**

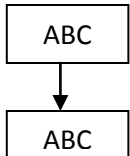
$$\frac{n!}{n_1! \dots n_k!}$$

Assigning each element of the entire group in one of the three or more subgroups.



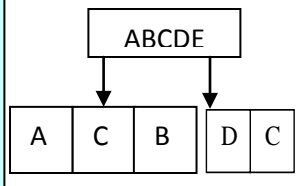
**Factorial [!]**

- Arranging a given set of 'n' items.
- No subgroups.
- There are n! ways of arranging 'n' items.



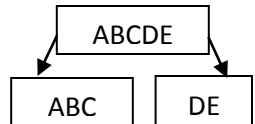
**Permutation [<sup>n</sup>P<sub>r</sub>]**

Specific ordering of a group of objects into only two groups of predetermined size.



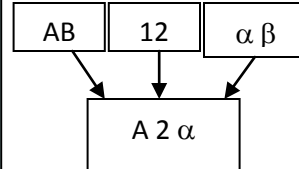
**Combination [<sup>n</sup>C<sub>r</sub>]**

- Choosing 'r' items from a set of 'n' items when order does not matter.
- It applies to only two groups of predetermined size.



**Multiplication Rule**

- Selecting only one item from each of the two or more groups.



# “Common Probability Distributions”

## Probability Distribution

- Describes the probabilities of all possible outcomes for a random variable.
- Sum of probabilities of all possible outcomes is 1.

## Probability Function

Probability of a random variable being equal to a specific value.

### Properties:

- $0 \leq p(x) \leq 1$
- $\sum p(x) = 1$

	<u>Discrete</u>	<u>Continuous</u>
<u>Random Variable</u>	Finite (measurable) # of possible outcomes.	Infinite (immeasurable) # of possible outcomes.
<u>Distribution</u>	<ul style="list-style-type: none"> <li>• P(x) can't be 0 if 'x' can occur.</li> <li>• We can find the probability of a specific point in time.</li> </ul>	<ul style="list-style-type: none"> <li>• P(x) can be zero even if 'x' can occur.</li> <li>• We can't find the probability of a specific point in time.</li> </ul>

## Probability Density Function (PDF)

- It is used for continuous distribution.
- Denoted by f(x).
- Finds the probability of an outcome within a particular range (b/w two values).
- Probability of any one particular outcome is zero.

## Cumulative Distribution Function (CDF)

- Calculates the probability of a random variable 'x' taking on the value less than or equal to a specific value 'x'.
- $F(x) = P(X \leq x)$

## Binomial Distribution

### Properties:

- Two outcomes (success & failure).
- 'n' number of independent trials.
- With replacement.
- Probability of success remains constant.
- $E(x) = np$ .
- $p(x) =$

$$= \frac{n!}{(n-x)!x!} p^x (1-p)^{n-x}$$

$$= {}^n C_x \cdot p^x \cdot q^{n-x}$$

## Binominal Tree

- Shows all possible combinations of up & down moves over a number of successive periods.
- **Node:** Each of the possible values along the tree.
- U is up-move factor.
- D is down-move factor (1/U).
- p is probability of up move.
- (1-p) is probability of down move.

Discrete uniform random variable

⇒ All outcomes have the same probability.

## Uniform Probability Distribution

### Discrete

- Has a finite number of specified outcomes.
- $P(x) \times k$ . K is the probability for 'k' number of possible outcomes in a range.
- **cdf:**  $F(x_n) = n \cdot p(x)$ .

### Continuous

- Defined over a range with parameters 'b' (upper limit) & 'a' (lower limit).
- **cdf:** It is linear over the variable's range.
- **Properties:**
- $P(x < a \text{ or } x > b) = 0$
- $P(x_1 \leq X \leq x_2) = \frac{x_2 - x_1}{b - a}$
- For all  $a \leq x_1 < x_2 \leq b$ .

## Univariate Distribution

Probability distribution of a single random variable.

## Multivariate Distribution

Specifies the probabilities associated with a group of random variables.

## Normal Distribution

- Completely described by  $\mu$  &  $\sigma^2$ .
- Stated as  $X \sim N(\mu, \sigma^2)$ .
- Symmetric about its mean, Skewness = 0.  $P(X \leq \mu) = 0.5 = P(X \geq \mu)$ .
- Kurtosis = 3.
- Linear combination of normally distributed random variables is also normally distributed.
- Large deviations from mean are less likely than small deviations.
- Probability becomes smaller & smaller as we move away from mean, but never becomes '0'.

## Multivariate Normal Distribution

It can be completely described by three parameters

- n means.
- n variances.
- $\frac{n(n-1)}{2}$  pair wise correlations.

## Log-Normal Distribution

- Generated by function  $e^x$  where 'x' is normally distributed.
- Skewed to the right.
- Bounded from below by 0.

## Standard Normal Distribution

- Has  $\mu = 0$  &  $\sigma = 1$  i.e.,  $N \sim (0,1)$ .
- **Standardization:** Process of converting an observed value of a random variable to its z-value.
- $z = \frac{x - \mu}{\sigma}$

**Confidence Interval**

Range of values around the expected value within which actual outcome is expected to be some specified percentage of time.

Confidence Interval	%age
$x \pm 1s$	68%
$x \pm 1.65s$	90%
$x \pm 1.96s$	95%
$x \pm 2s$	95.45%
$x \pm 2.58s$	99%
$x \pm 3s$	99.73%

**Roy's Safety First Criterion**

- Optimal portfolio minimizes the probability that the return of the portfolio falls below some minimum acceptable level.
- Minimize  $P(R_p < R_L)$ .
- SFR is the number of standard deviations below the mean.
- SFRatio =  $\frac{[E(R_p) - R_L]}{\sigma_p}$
- Choose the portfolio with greatest SFRatio.

**Shortfall Risk**

Probability that portfolio value will fall below some minimum level at a future date.

**Monte-Carlo Simulation**

- Repeated generation of one or more factors (e.g. risk) that affect required value (e.g., stock price) in order to generate a distribution of the values (stock price).
- We have the flexibility of providing the data.

**Compounded Rates of Returns**

**Discrete**

Daily, annually, weekly, monthly compounding

**Continuous**

- $\ln(S_1/S_0) = \ln(1+HPR)$
- These are additive for multiple periods.
- Effective annual rate based on continuous compounding is given as:  $EAR = e^{R_{CC}} - 1$

**Simulation Procedure for Stock Option Valuation**

Specify prob. dist. of stock prices & relevant interest rate as well as their parameters.

Randomly generate values of stock prices & interest rates.

Value the options for each pair of risk factors.

Calculate mean option value performing many iterations & use it as estimated option value.

**Historical Simulation**

- Based on actual values & actual distributions of the factors i.e., based on historical data.

**Limitation:**

- History does not repeat itself.
- Historical data does not provide flexibility.

**Uses**

- Valuing complex securities.
- Simulating gains / losses from trading strategy.
- Estimating value at risk (VAR).
- Examining variability of the difference b/w assets & liabilities of pension funds.
- Valuing portfolio with non-normal return distribution.

**Limitations**

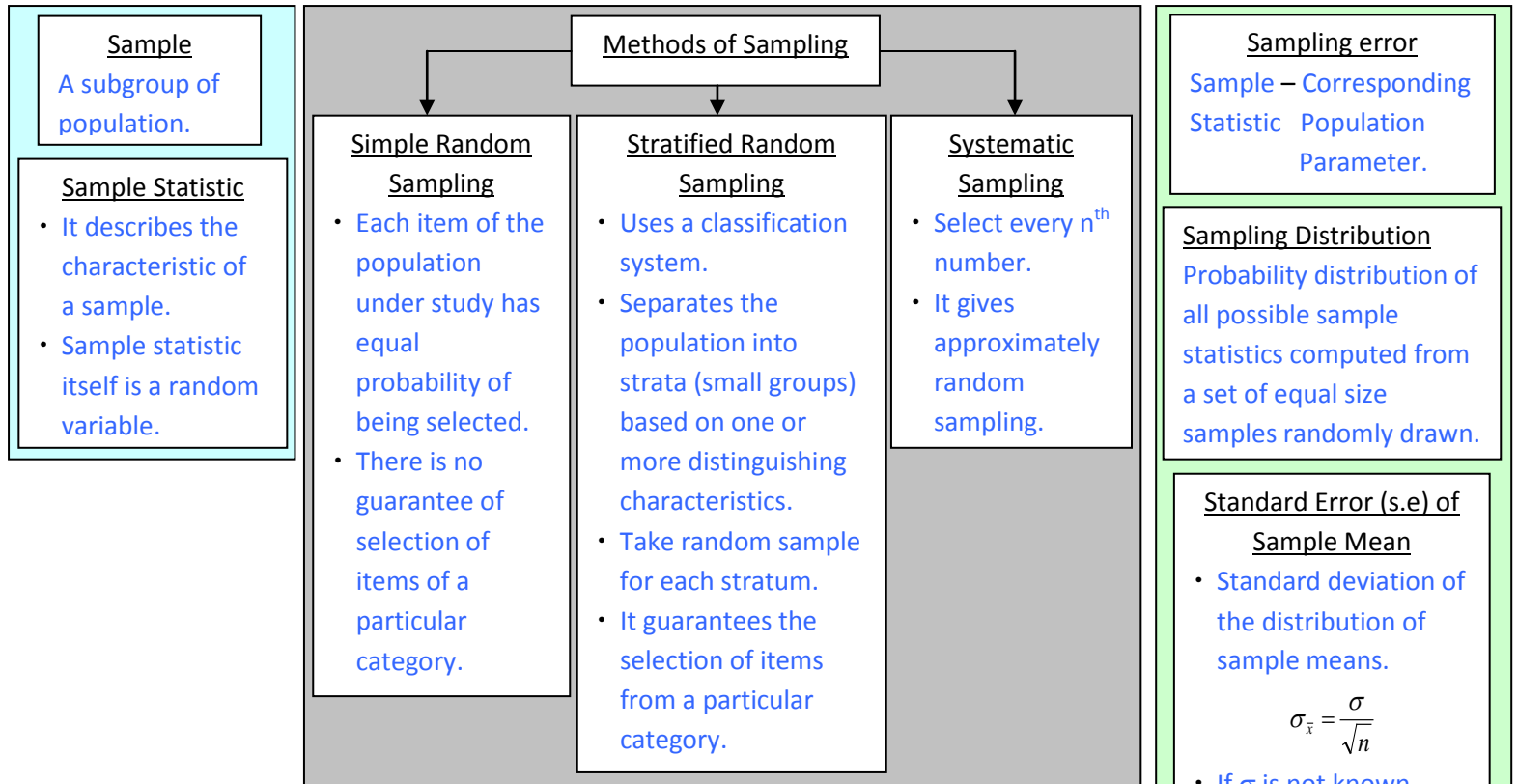
- Complex procedure.
- Highly dependent on assumed distributions.
- Based on a statistical rather than an analytical method.

# "Sampling & Estimation"

s.e = standard error

↑ = rises

→ = approaches to  
d.f = degrees of freedom  
n = sample size



Data	Time	Variables
Time series	Observations taken over equally spaced time interval	Single
Cross-sectional	Single point estimate	Different variables

Data	Entity	Characteristics
Longitudinal	Same	Multiple
Panel	Multiple	Same

**Central Limit Theorem (CLT)**

For a random sample of size 'n' with;

- population mean  $\mu$ ,
- finite variance  $\sigma^2$ , the sampling distribution of sample mean  $\bar{x}$  approaches a normal probability distribution with mean ' $\mu$ ' & variance as ' $n$ ' becomes large. i.e, As  $n \uparrow$ ;  $\bar{x} \sim$

**Properties of CLT**

- For  $n \geq 30 \Rightarrow$  sampling distribution of mean is approx. normal.
- For  $n \leq 30 \Rightarrow$  Not normal.
- Mean of distribution of all possible samples = population mean ' $\mu$ '.
- Variance of distribution =  $\frac{\sigma^2}{n}$

CLT applies only when samples is random.

**Point Estimate (P.E)**

- Single (sample) value used to estimate population parameter.

$$\bar{x} = \frac{\sum X}{n}$$

**Estimator:** Formula used to compute P.E.

**Confidence Interval (CI) Estimates**

- Results in a range of values within which actual parameter value will fall.
- $P.E \pm (\text{reliability factor} \times \text{s.e})$
- $\alpha =$  level of significance.
- $1 - \alpha =$  degree of confidence.

**Desirable properties of an estimator**

**Unbiased**

Expected value of estimator equals parameter e.g.,  $E(\bar{x}) = \mu$  i.e, sampling error is zero.

**Efficient**

If  $\text{var}(\bar{x}_1) < \text{var}(\bar{x}_2)$  then  $\bar{x}_1$  is efficient than  $\bar{x}_2$

**Consistent**

As  $n \uparrow$ , value of estimator approaches parameter & sample error approaches '0' e.g., As  $n \rightarrow \infty$   $\bar{x} \rightarrow \mu$  & s.e.  $\rightarrow 0$

Mean satisfies all 3 properties but median & mode don't.

**If sample is not random then**

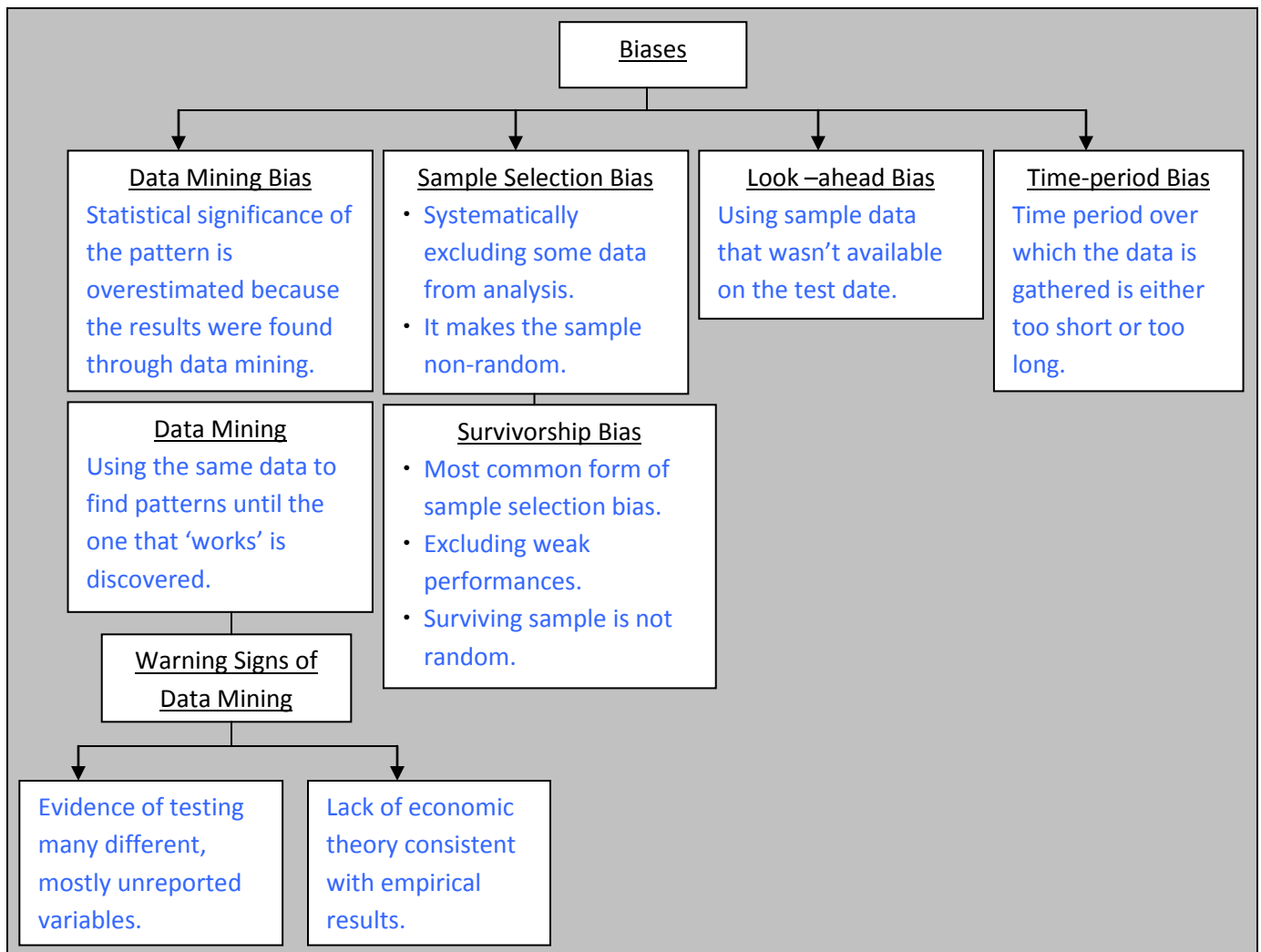
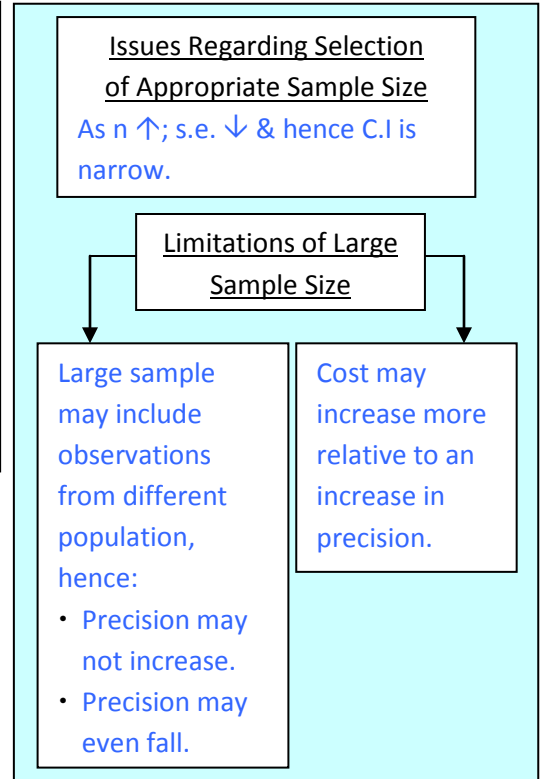
- CLT does not apply.
- CI is not unbiased.
- Estimates will not have the desirable properties.

**Student's T-Distribution**

- Bell shaped.
- Shape is based on d.f.
- d.f. is based on 'n'.
- t-distribution depends on n (d.f) [normal distribution does not depend on n (d.f)].
- Symmetrical about it's mean.
- Less peaked than normal distribution.
- Has fatter tails.
- More probability in tails i.e., more observation are away from the centre of the distribution & more outliers.
- More difficult to reject  $H_0$  using t distribution.
- C.I for a r.v. using t distribution must be wider when d.f are less.

Distribution		Variance		Sample		Test Statistic	
Normal	Non normal	Known	Unknown	Small (n<30)	Large (n≥30)	t	z
✓	x	✓	x	✓	x	x	✓
✓	x	✓	x	x	✓	x	✓
✓	x	x	✓	✓	x	✓	x
✓	x	x	✓	x	✓	✓*	x
x	✓	✓	x	✓	x	x	x
x	✓	✓	x	x	✓	x	✓
x	✓	x	✓	✓	x	x	x
x	✓	x	✓	x	✓	✓*	x

\*The z-statistic is theoretically acceptable here, but use of the t-statistic is more conservative.



# “Hypothesis Testing”

$\alpha$  = Level of significance  
 t.s = Test statistics  
 t.v = Table Value

t.v provides the critical values called as rejection points

s.s = Sample statistic  
 c.v = Critical value  
 s.e = Standard error

**Hypothesis**  
Statement about parameter value developed for testing.

**Hypothesis Testing Procedure**

- It is based on sample statistics & probability theory.
- It is used to determine whether a hypothesis is a reasonable statement or not.

```

                State the hypothesis
                ↓
                Select the appropriate test statistic
                ↓
                Specify the level of significance
                ↓
                State the decision rule regarding the hypothesis
                ↓
                Collect the sample and calculate the sample statistics
                ↓
                Make a decision regarding the hypothesis
                ↓
                Make a decision based on the results of the test
            
```

**One Tailed Test**  
Alternative hypothesis having one side.

- Upper Tail**  
 $H_0: \mu \leq \mu_0$  vs  $H_a: \mu > \mu_0$ .
- Decision rule**  
Reject  $H_0$  if  $t.s > t.v$ .
- Lower Tail**  
 $H_0: \mu \geq \mu_0$  vs  $H_a: \mu < \mu_0$ .
- Decision rule**  
Reject  $H_0$  if  $t.s < -t.v$ .

**Two Tailed Test**

- Alternative hypothesis having two sides.
- $H_0: \mu = \mu_0$  vs  $H_a: \mu \neq \mu_0$ .
- Reject  $H_0$  if  $|t.s| > t.v$ .

**Null Hypothesis  $H_0$**

- Tested for possible rejection.
- Always includes '=' sign.

**Two Types**

**Alternative Hypothesis  $H_a$**   
The one that we want to prove.

(Source: Wayne W. Daniel and James C. Terrell, *Business Statistics, Basic Concepts and Methodology*, Houghton Mifflin, Boston, 1997.)

**Test Statistics (t.s)**  
Hypothesis testing involves two statistics:

- t.s calculated from sample data.
- critical values of t.s.

t.s is a random variable that follows some distribution.

**Two Types of Errors**

**Type I Error**  
Rejecting a true null hypothesis.

**Type II Error**  
Failing to reject a false null hypothesis.

**Decision Rule**

- It is based on distribution of t.s.
- It is specific & quantitative.

**Significance Level ( $\alpha$ )**

- Probability of making a type I error.
- Denoted by Greek letter alpha ( $\alpha$ ).
- Used to identify critical values.

**Power of a Test**

- P(type II error).
- Probability of correctly rejecting a false null hypothesis.

**p- value**

- Probability of obtaining a critical value that would lead to a rejection of a true null hypothesis.
- Reject  $H_0$  if  $p\text{-value} < \alpha$ .

**Statistical Significance vs Economical Significance**

- Statistically significant results may not necessarily be economically significant.
- A very large sample size may result in highly statistically significant results that may be quite small in absolute terms.

**Relationship b/w Confidence Intervals & Hypothesis Tests**

- Related because of critical value.

**C.I**

- $[(s.s) - (c.v)(s.e)] \leq \text{parameter} \leq [(s.s) + (c.v)(s.e)]$ .
- It gives the range within which parameter value is believed to lie given a level of confidence.

**Hypothesis Test**

- $-c.v \leq t.s \leq +c.v$ .
- range within which we fail to reject null hypothesis of two tailed test given level of significance.

$\sigma^2$  = population variance  
 N.dist = Normally distributed  
 N.N.dist = Non Normally distributed

n = sample size  
 $n \geq 30$  = large sample  
 $n < 30$  = small sample

t.s. = Test statistics  
 t.v = table value  
 d.f = degree of freedom

Testing	Conditions	Test Statistics	Decision Rule
Population Mean	<ul style="list-style-type: none"> <li><math>\sigma^2</math> known</li> <li>N. dist.</li> </ul>	$z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}}$	<ul style="list-style-type: none"> <li><math>H_0: \mu \leq \mu_0</math> vs <math>H_a: \mu &gt; \mu_0</math> Reject <math>H_0</math> if t.s. &gt; t.v</li> <li><math>H_0: \mu \geq \mu_0</math> vs <math>H_a: \mu &lt; \mu_0</math> Reject <math>H_0</math> if t.s. &lt; -t.v</li> <li><math>H_0: \mu = \mu_0</math> vs <math>H_a: \mu \neq \mu_0</math> Reject <math>H_0</math> if   t.s.   &gt; t.v</li> </ul>
	<ul style="list-style-type: none"> <li><math>n \geq 30</math></li> <li><math>\sigma^2</math> unknown</li> </ul>	$z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}}$ or $t_{n-1}^* = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$ *(more conservative)	
	<ul style="list-style-type: none"> <li><math>\sigma^2</math> unknown</li> <li><math>n &lt; 30</math></li> <li>N. dist.</li> </ul>	$t_{n-1} = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}}$ ; d.f = n-1	
Equality of the Means of Two Normally Distributed Populations based on Independent Samples.	Unknown variances assumed equal.	$t_{(n_1+n_2-2)} = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$ where; $s_p = \sqrt{\frac{(n_1-1)s_1^2 + (n_2-1)s_2^2}{n_1 + n_2 - 2}}$ d.f = $n_1 + n_2 - 2$	<ul style="list-style-type: none"> <li><math>H_0: \mu_1 - \mu_2 \leq 0</math> vs <math>H_a: \mu_1 - \mu_2 &gt; 0</math> Reject <math>H_0</math> if t.s. &gt; t.v</li> <li><math>H_0: \mu_1 - \mu_2 \geq 0</math> vs <math>H_a: \mu_1 - \mu_2 &lt; 0</math> Reject <math>H_0</math> if t.s. &lt; -t.v</li> <li><math>H_0: \mu_1 - \mu_2 = 0</math> vs <math>H_a: \mu_1 - \mu_2 \neq 0</math> Reject <math>H_0</math> if   t.s.   &gt; t.v</li> </ul>
	Unequal unknown variances.	$t = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$ $d.f = \frac{\left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right)^2}{\frac{\left(\frac{s_1^2}{n_1}\right)^2}{n_1} + \frac{\left(\frac{s_2^2}{n_2}\right)^2}{n_2}}$	

**Paired Comparisons Test**

T.S  $t_{(n-1)} = \frac{\bar{d} - \mu_{d0}}{s_d/\sqrt{n}}$

$\bar{d} = \frac{1}{n} \cdot \sum d$

$s_{\bar{d}} = \frac{s_d}{\sqrt{n}}$

$s_d = \sqrt{\frac{\sum (d - \bar{d})^2}{n - 1}}$

**Decision Rule**

- $H_0: \mu_d \leq \mu_{d0}$  vs  $H_a: \mu_d > \mu_{d0}$   
Reject  $H_0$  if t.s. > t.v.
- $H_0: \mu_d \geq \mu_{d0}$  vs  $H_a: \mu_d < \mu_{d0}$   
Reject  $H_0$  if t.s. < -t.v.
- $H_0: \mu_d = \mu_{d0}$  vs  $H_a: \mu_d \neq \mu_{d0}$   
Reject  $H_0$  if t.s. > t.v.

**Testing Variance of a N.dist. Population**

T.s  $\chi_{(n-1)}^2 = \frac{(n-1)s^2}{\sigma_0^2}$

**Decision Rule**  
Reject  $H_0$  if t.s. > t.v

**Chi-Square Distribution**

- Asymmetrical.
- Bounded from below by zero.
- Chi-Square values can never be -ve.

**Testing Equality of Two Variances from N.dist. Population**

T.s  $F = \frac{s_1^2}{s_2^2}; s_1^2 > s_2^2$

**Decision Rule**  
Reject  $H_0$  if t.s. > t.v

**F- Distribution**

- Right skewed.
- Bounded by zero.

**Parametric Test**

- Specific to population parameter.
- Relies on assumptions regarding the distribution of the population.

**Non-Parametric Test**

- Don't consider a particular population parameter.
- Or
- Have few assumptions regarding population.

# "TECHNICAL ANALYSIS"

T.A = Technical Analysis  
 F.S = Financial Statements  
 C.F = Cash Flows  
 ROC = Rate of Change  
 RSI = Relative Strength Index

S.D = Standard Deviation  
 M.A = Moving Average  
 M.V = Market Value  
 MACD = Moving Average  
 Convergence/divergence

12. a

- Study of collective market sentiment.
- Prices are determined by interaction of supply & demand.
- Key assumption of T.A is that EMH does not hold.
- Usefulness is limited in illiquid & outside manipulation markets.

Comparison	
Technical Analysis	Fundamental Analysis
• Share price & trading volume	• Intrinsic value
• Data is observable	• Use F.S & other information
• Can applied on assets without C.F	

