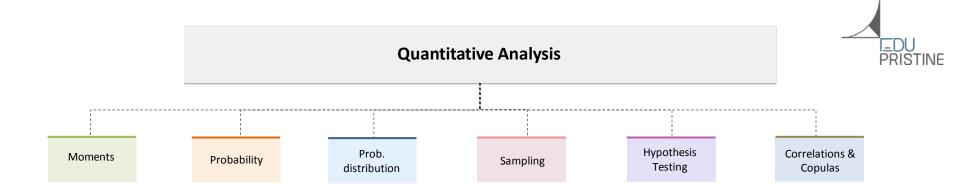
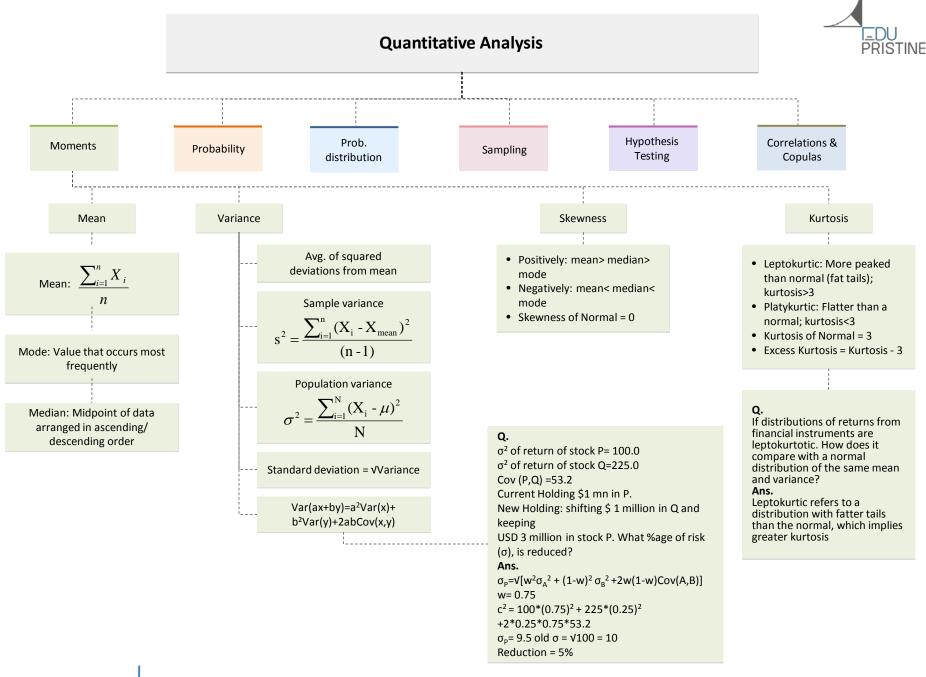
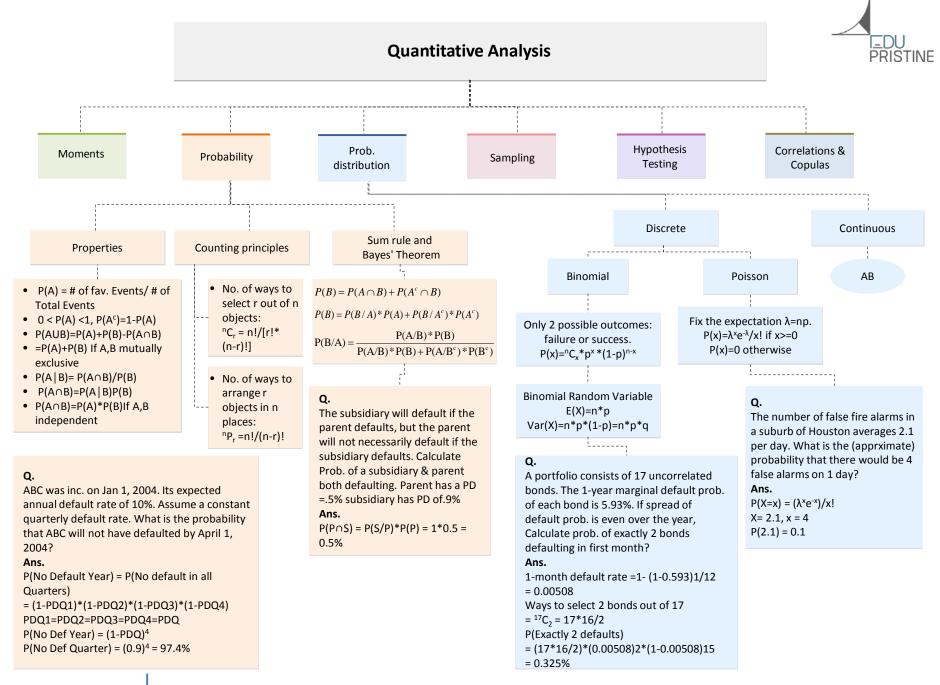
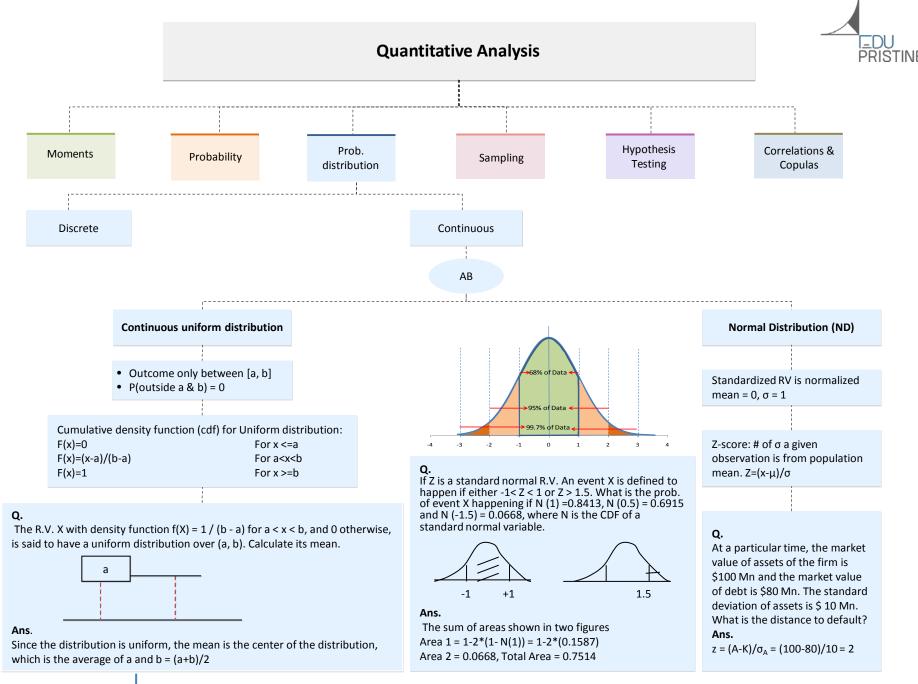


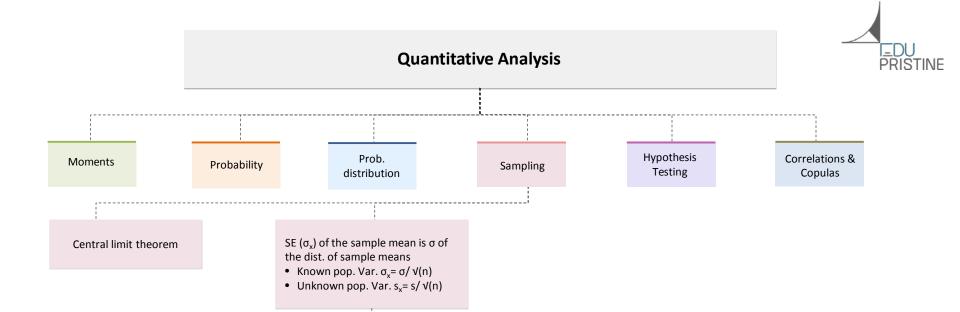
Quantitative Analysis











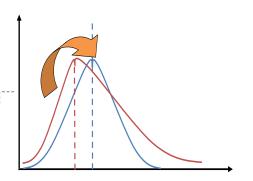
Q.

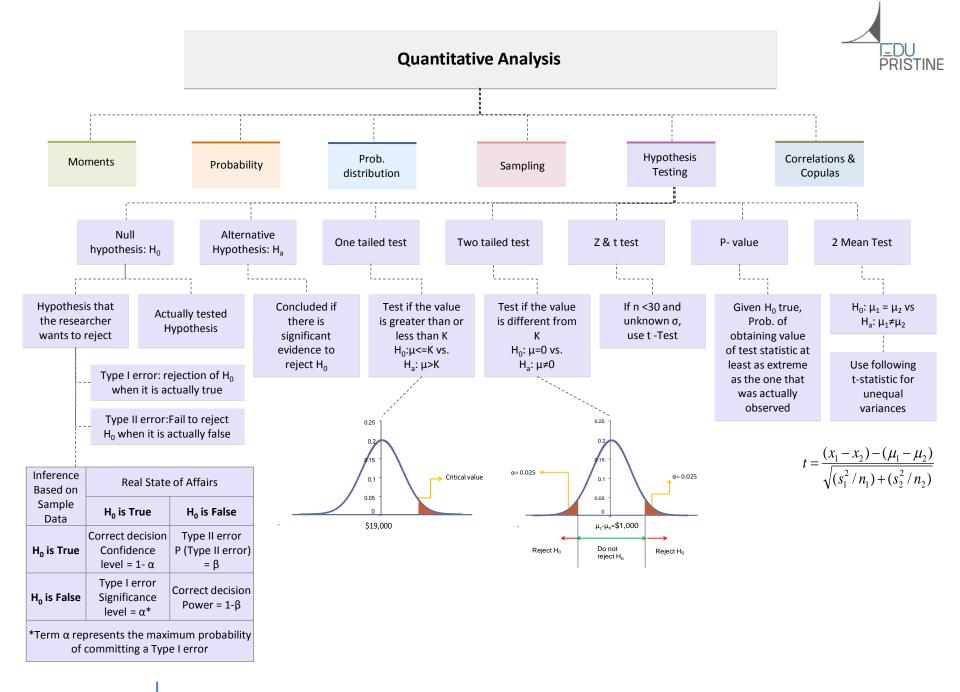
25 observation are taken from a sample of known variance. Sample mean =70 and population σ = 60. You wish to conduct a two - tailed test of null hypothesis that the mean is equal to 50. What is most appropriate test statistic?

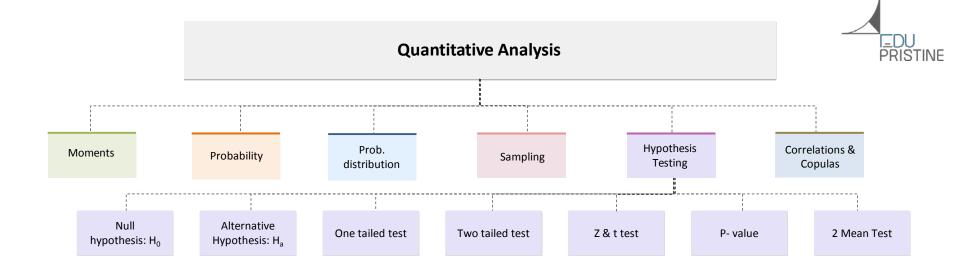
Ans.

Standard Error of mean $(\sigma_x) = \sigma/V(n) = 60/V25 = 12$ Degrees of freedom = 24 Use t- statistic = $(x - \mu)/\sigma_x = (70 - 50)/12 = 1.67$

As Sample Size increases Sampling Distribution Becomes Almost Normal regardless of shape of population







Q

A stock has initial price of \$100. It price one year from now is given by $S = 100 * \exp(r)$, where the rate of return r is normally distributed with mean of 0.1 and a standard deviation of 0.2. What is the range of S in an year with 95% confidence?

Ans.

 $100e^{(0.1-1.96*0.2)} < S < 100e^{(0.1+1.96*0.2)}$

74.68 < S < 163.56

Q.

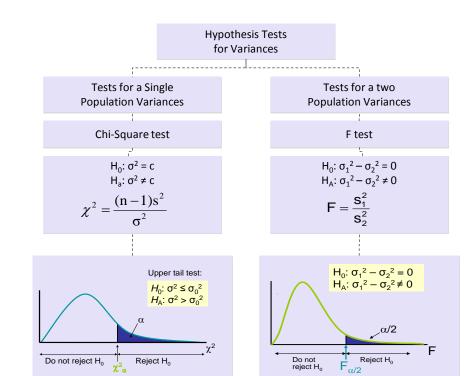
If standard deviation of a normal population is known to be 10 and the mean is hypothesized to be 8. Suppose a sample size of 100 is considered. What is the range of sample means in which hypothesis can be accepted at significance level of 0.05?

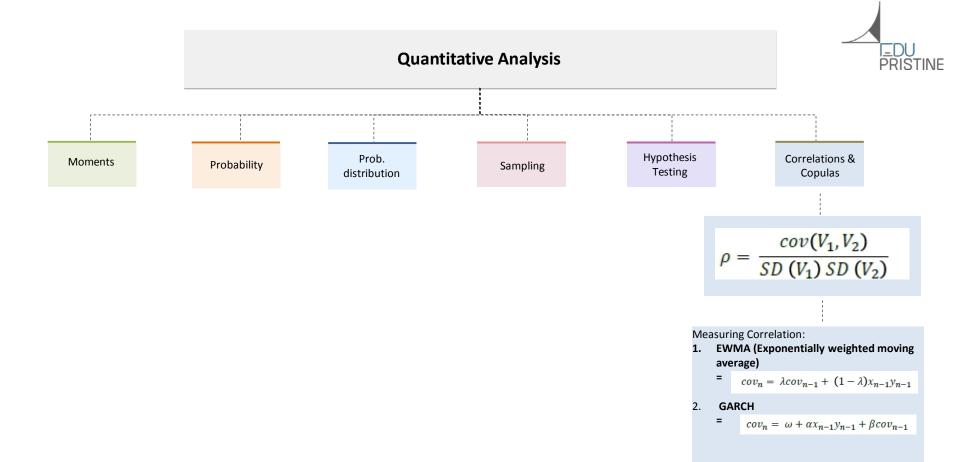
Ans.

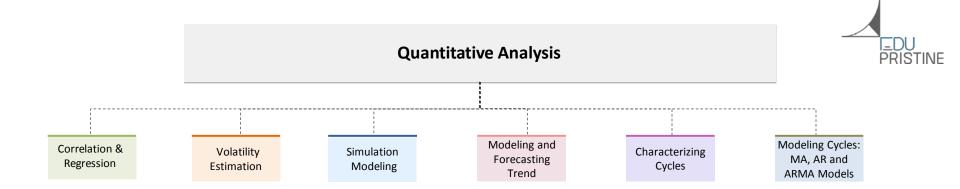
 $s_v = \sigma/v = 10/v = 10$

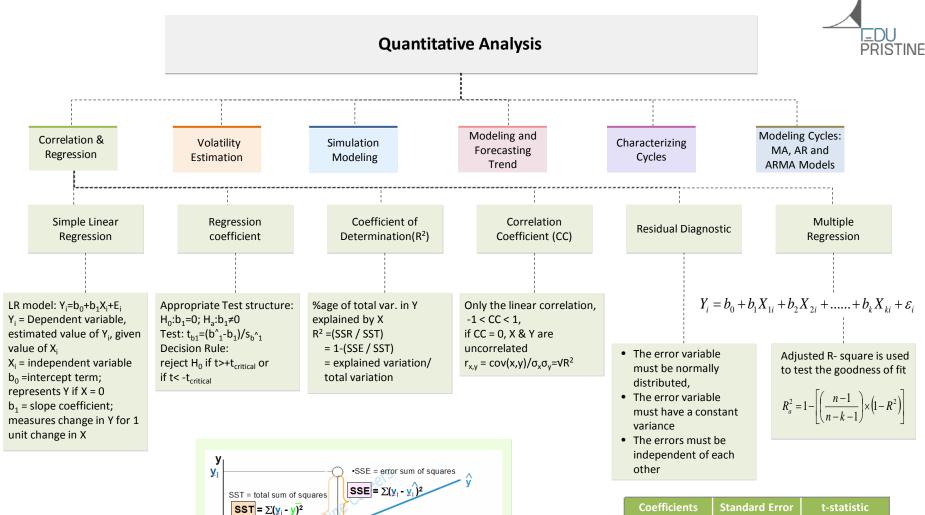
 $z = (x-\mu)/\sigma_x$ = (x-8)/1

At 95% -1.96<z<1.96; So 6.04<x<9.96





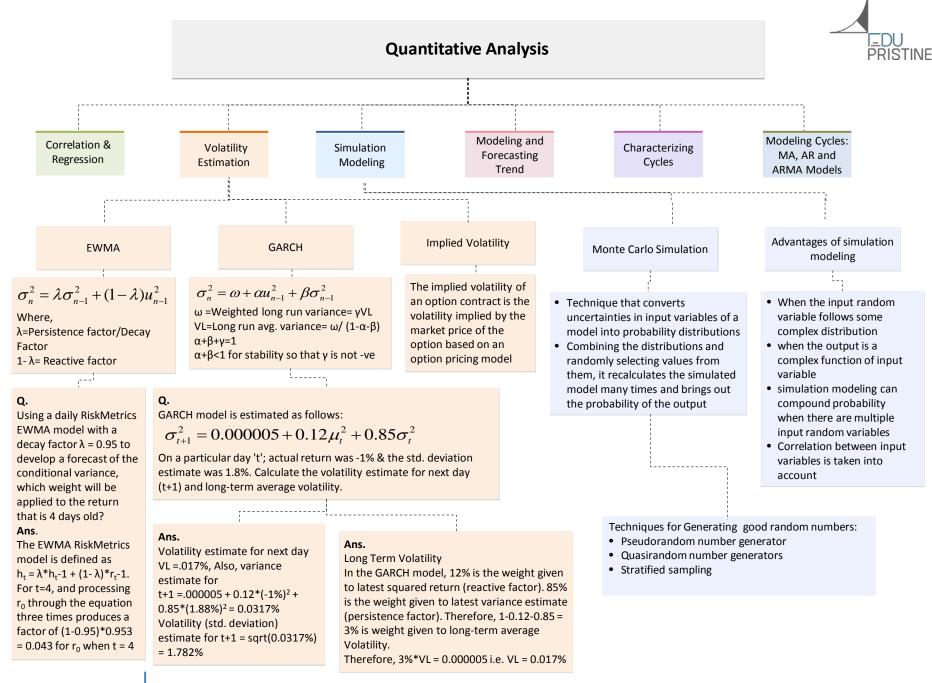




SSR = $\sum (y_i - y)^2$ SSR = regression sum of squares

X

	Coefficients	Standard Error	t-statistic
Intercept	49.94	2.85	17.53
X Variable 1	-38.79	138.93	-0.28
X Variable 2	-431.75	170.50	-2.53
X Variable 3	-70.40	121.06	-0.58





Quantitative Analysis

Correlation & Regression

Volatility Estimation Simulation Modeling Modeling and Forecasting Trend

Characterizing Cycles

Modeling Cycles: MA, AR and ARMA Models

 $MSE = \frac{\sum_{t=1}^{T} e_t^2}{T}$

Where T, is the sample size and $e_t = y_t - \hat{y}_t$

where,
$$\hat{y}_t = \widehat{\beta_0} + \widehat{\beta_1}TIME_t$$

Criteria for panelizing MSE to reflect the df used:

• Akaike information criterion

$$AIC = e^{\left(\frac{2k}{T}\right)} \frac{\sum_{t=1}^{T} e_t^2}{T}$$

• Schwarz information criterion

$$SIC = T^{\left(\frac{k}{T}\right)} \frac{\sum_{t=1}^{T} e_t^2}{T}$$

- Key property for selection criteria of model is "Consistency"
- A model selection criteria is consistent if the following conditions are met:
- When the true model i.e. DGP is among the models considered, the probability of selecting the true DGP approaches 1 as the sample size gets large
- When the true model is not among the models considered, the probability of selecting the best approximation to the true DGP approaches 1 as the sample size gets large



Quantitative Analysis

Correlation &
Regression

Volatility Estimation Simulation Modeling Modeling and Forecasting Trend

Characterizing Cycles Modeling Cycles: MA, AR and ARMA Models

Covariance stationary

Requirements for covariance stationary are:

- 1. The mean of the series to be stable over time
- The covariance structure of the series to be stable over time
- 3. The variance of the series the auto covariance at displacement 0, (0) be finite

- Autocorrelation function
- We work with autocorrelation function p(τ) rather than auto covariance function y(τ)

$$p(\tau) = \frac{\gamma(\tau)}{\gamma(0)}$$

White Noise:

- When shock is uncorrelated then y_t is serially correlated and when it is not mentioned we assume this process with zero mean, constant variance and no correlation is called White Noise
- When y is serially independent then y is independent white noise
- When y is independently and identically distributed with zero mean and constant variance and uncorrelated then y is Gaussian White House





Correlation & Regression

Volatility Estimation Simulation Modeling Modeling and Forecasting Trend

Characterizing Cycles

Modeling Cycles: MA, AR and ARMA Models

Moving Average Models:

• First order MA=

$$y_t = \varepsilon_t + \theta \varepsilon_{t-1} = (1 + \theta L)\varepsilon_t$$

Types:

- Population autocorrelation function
- Population Partial Autocorrelation Function

Autoregressive Models (AR)

• The first order AR=

$$y_t = \varphi y_{t-1} + \varepsilon_t$$

• In lag operator form, it is

$$(1 - \varphi L)y_t = \varepsilon_t$$

Types:

- Population autocorrelation function
- Population Partial Autocorrelation Function

Autoregressive Moving Average models

- MA and AR processes are combined to get better parsimonious approximations
- The ARMA (1,1) Process:
- 1. The process is

$$y_t = \varphi y_{t-1} + \varepsilon_t + \theta \varepsilon_{t-1}$$

2. In lag operator form, it is

$$(1 - \varphi L)y_t = (1 + \theta L)\varepsilon_t$$



Thank you!

Contact:

E: <u>help@edupristine.com</u>

Ph: +1 347 647 9001