

# Critical risk analysis of intraday trading using algorithm like Black-Scholes, Random-Walk and comparison of those with TWAP, MACD strategy, through visual analysis.

By: Team Alpha

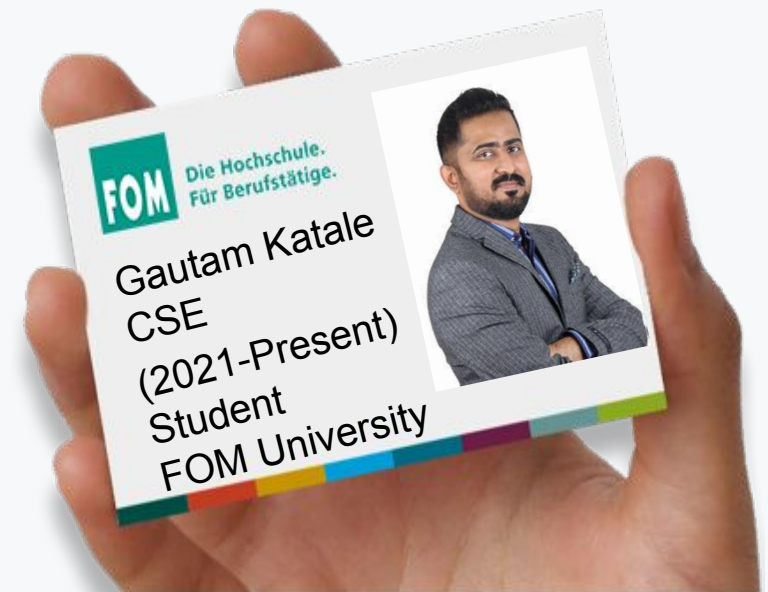
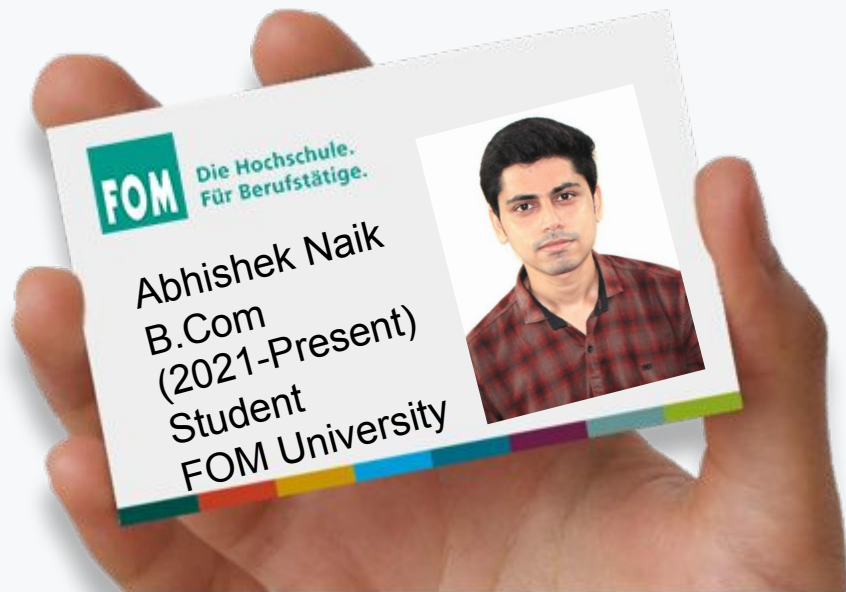
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# About Us:

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# Problem Statement

**Critical risk analysis for intraday trading using widely used algorithm like Black-Scholes, Random-Walk and comparison of those algorithm with the help of TWAP MCAD strategy.**

**The questions we are trying to answer are:**

- How much value do we put at risk by investing in a particular stock?
- Which algorithm gives better results?
- How algorithm perform for intraday trade.
- Does TWAP works well for intraday trade?
- Finding VAR(Value at Risk) for the stocks.
- Analysing performance using Moving Average Convergence Divergence (MACD) indicator strategy and identify entry and exit points.
- See how good option trading is for intraday trades.

# Important Pointers

## Intraday trade:

- Intraday trading means buying and selling stocks on the same trading day.
- Intraday trading is also known as Day Trading.
- Advantages of Intraday Trading: avoid tangible risk in holding shares overnight and reading the pulse of the market.

## Option trading:

- It grants you the right but not the obligation to buy or sell an underlying asset at a set price on or before a certain date.

## European Option:

- It is a version of an options contract that limits execution to its expiration date.

## BSM- Black-Scholes model

- Is a differential equation widely used to price options contracts.
- Published by Myron Scholes and Fischer Black in 1973.

# Continuation...

## Black-Scholes model assumptions:

- Interest rate is known and constant through time and so is the Volatility(%).
- The stock follows a random walk in continuous time, the variance of stock price follow a log-normal distribution.
- Stock pays no dividends or no transaction costs.
- The option can only be exercised at expiration.

## Black Scholes Formula:

$$Call = S_0 N(d_1) - N(d_2) K e^{-rT}$$

$$Put = N(-d_2) K e^{-rT} - N(-d_1) S_0$$

$$d_1 = \frac{\ln(\frac{S}{K}) + (r + \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}$$

$$d_2 = d_1 - \sigma\sqrt{T}$$

$$N(x) = \int_{-\infty}^x \frac{e^{-x^2/2}}{\sqrt{2\pi}}$$

- S : current asset price
- K: strike price of the option
- N: CDF of normal distribution(left to right)
- r: risk free rate
- T : time until option expiration
- $\sigma$ : annualized volatility of the asset's returns
- d1: the factor by which the present value of change of the stock exceeds the current stock price.
- d2: the risk-adjusted probability that the option will be exercised.



## Continuation...

### Black-Scholes model gives us Call and Put values:

- Put Call Parity must hold to avoid market arbitrage.

$$C + Ke^{-rT} = P + S_0$$

### Time-weighted average price (TWAP):

- Is the average price of an asset over a specified time.
- TWAP is calculated by averaging the entire day's price bar, i.e., open, high, low, and close prices of the day.
- Advantages of TWAP: TWAP signals cover a considerable amount of risk with their calculations.

### Formula to calculate TWAP:

- Average of each day's price = (Open + High + Low + Close)/4
- Average of N days = (Average of first day's price + Average of second day's price + ..... + Average of Nth day's price)/N

### Moving Average Convergence/Divergence (MACD):

- To understand this we need to know: Exponential Moving Average (EMA), it automatically allocates greater importance to most recent data point and lesser to the one at a distance.
- MACD is a trend-following leading indicator that is calculated by subtracting two Exponential Moving Averages (one with longer and the other shorter periods).

## MACD components:

- **MACD Line:** This line is the difference between two given Exponential Moving Averages.
  - `MACD LINE = FAST LENGTH EMA - SLOW LENGTH EMA`
- **Signal Line:** This line is the Exponential Moving Average of the MACD line itself for a given period of time.
- **Histogram:** It is a histogram purposely plotted to reveal the difference between the MACD line and the Signal line.

## MACD Strategy:

- `IF MACD LINE > SIGNAL LINE => BUY THE STOCK`
- `IF SIGNAL LINE > MACD LINE => SELL THE STOCK`



# About and Head of Our Dataset

❑ Unfortunately we are using temp data due to connection error currently with Workspace data access.

❑ We are considering two stock data for modeling needs.

❑ Apple and Amazon Data from 2013-2019 with 1 day interval is taken.

❑ The number of rows (observations) is 2376 and column (observations) is 6

❑ **Data Dictionary:**

❑ Volume: Volume of stock purchased in a day.

❑ High: Highest value of the stock in the day.

❑ Low: Lowest value of the stock in the day.

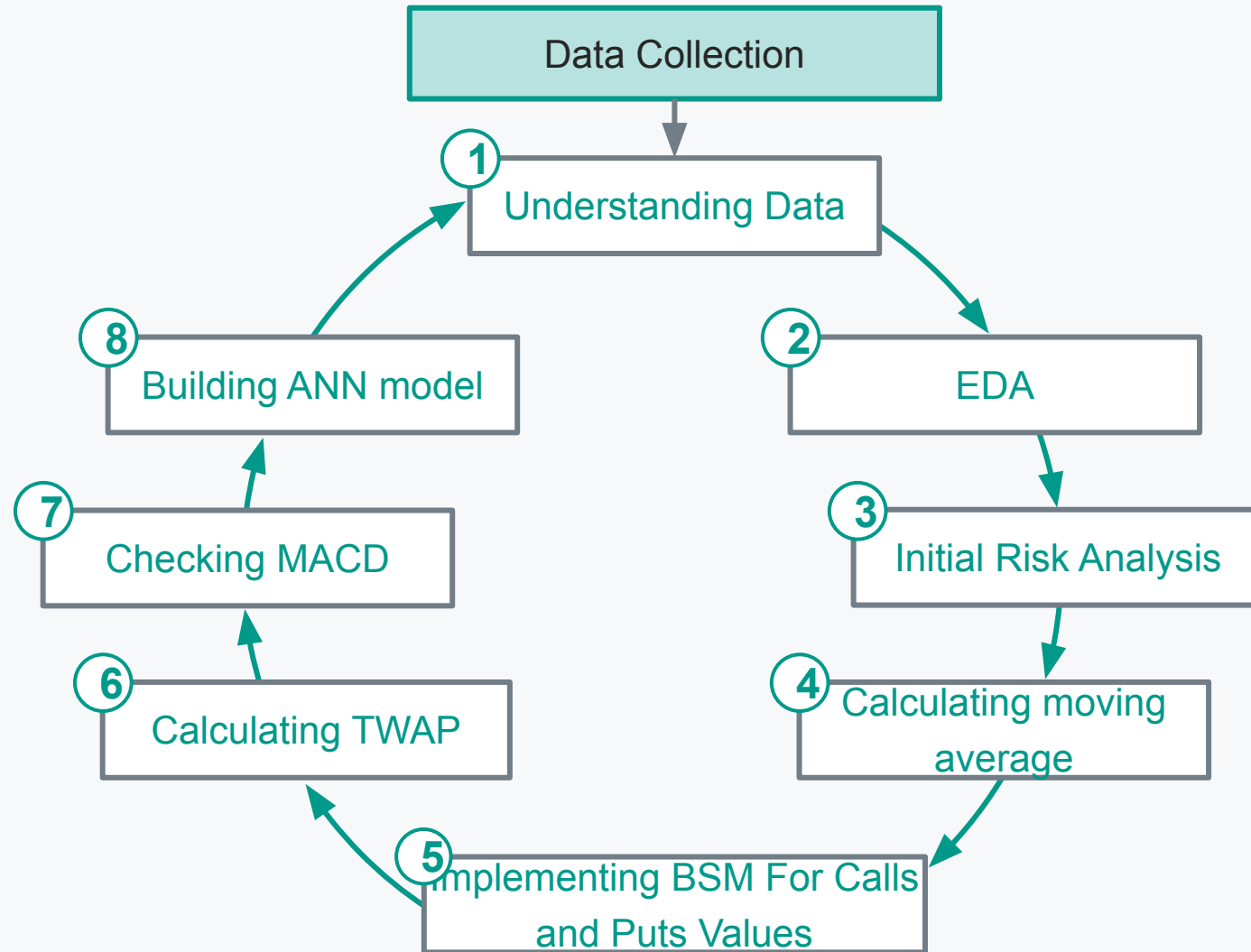
❑ Open: The price at the opening of the trade day.

❑ Close: Closing price at the end of the trade.

```
-----  
ModuleNotFoundError                                Traceback (most recent call last)  
Input In [2], in <cell line: 1>()  
----> 1 import refinitiv.dataplatform.eikon as ek  
      3 ek.set_app_key('DEFAULT_CODE_BOOK_APP_KEY')  
      5 df, err = ek.get_data(  
      6     instruments = ['DA00.0'],  
      7     fields = [  
      (...)  
     14 ]  
     15 )  
  
ModuleNotFoundError: No module named 'refinitiv'
```

TimeStamp	Company	volume	high	low	open	close
2013-01-02	Apple	140.13M	79.29	77.38	79.12	78.43
2013-01-03	Apple	88.24M	78.52	77.29	78.27	77.44
2013-01-04	Apple	148.58M	76.95	75.12	76.71	75.29
2013-01-05	Apple	121.04M	75.61	73.6	74.57	74.84
2013-01-06	Apple	114.68M	75.98	74.46	75.6	75.04

# Workflow



# Progress so far

## ▼ Analysing Closing value of stock per day.

```
✓ [22] 1 df['close'].describe()
```

```
count    2376.000000
mean      148.431098
std        67.475508
min        55.790000
25%       106.887500
50%       135.345000
75%       174.102500
max       506.090000
Name: close, dtype: float64
```

```
[26] 1 # The following code is to set to show the flow
      2
      3 from pylab import rcParams
      4 rcParams['figure.figsize'] = 20,5
      5 df[['close', 'high', 'low']].plot(grid=True)
```

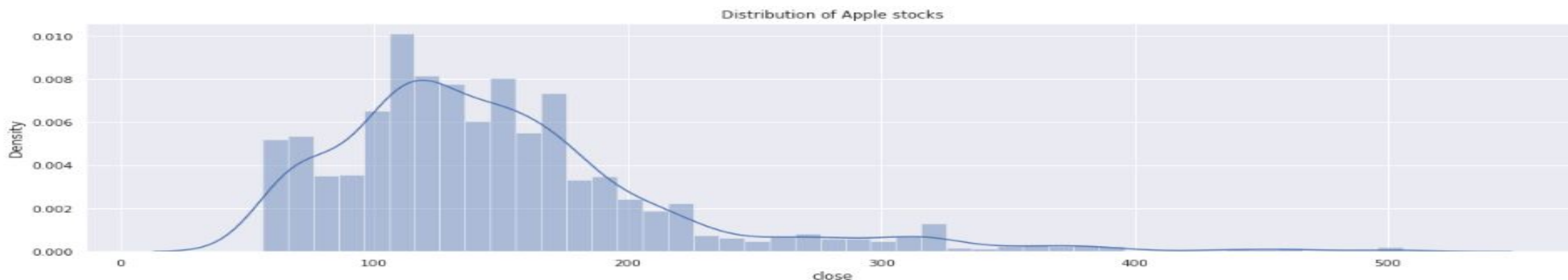
<matplotlib.axes.\_subplots.AxesSubplot at 0x7fde2dc68c50>



A good thing for our model is that for now the min and max for our data is positive.

# Progress so far

```
[31] 1 sns.distplot(df['close'])  
      2 plt.title('Distribution of Apple stocks')  
      3 plt.show()
```



We get almost a highly right-skewed, let's do a Log transformation data.

```
[32] 1 df['close_log'] = np.log(df['close'])  
      2 sns.distplot(df['close_log'])  
      3 plt.title('Log Distribution of Apple stocks')  
      4 plt.show()
```



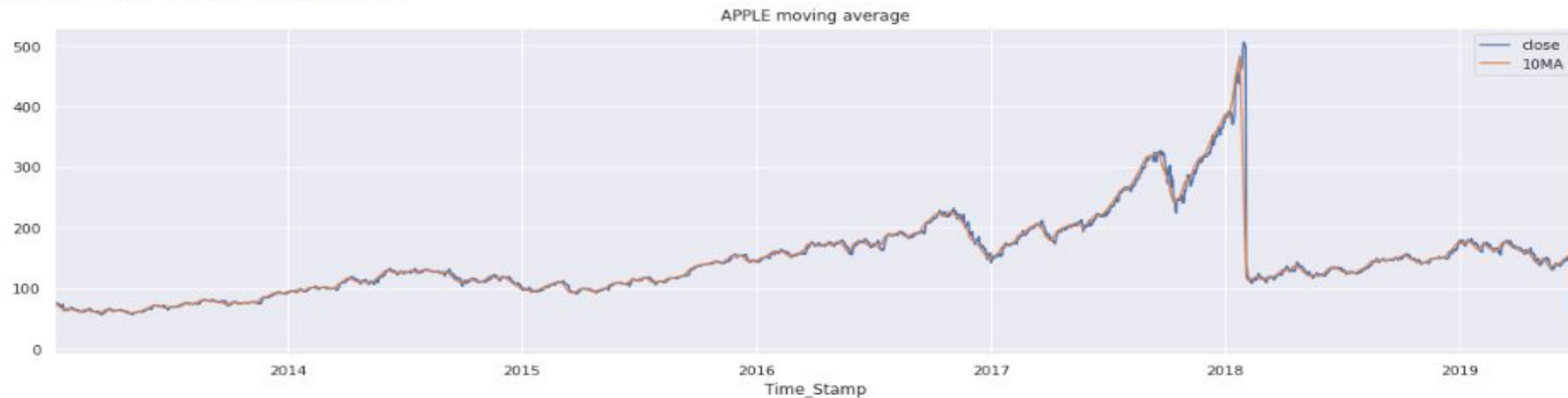
# Progress so far

```
[41] 1 df['10MA'] = moving_averages
     2 df.head()
```

	Company	volume	high	low	open	close	close_log	10MA
Time_Stamp								
2013-01-02	Apple	140.13M	79.29	77.38	79.12	78.43	4.362207	74.51
2013-01-03	Apple	88.24M	78.52	77.29	78.27	77.44	4.349503	73.90
2013-01-04	Apple	148.58M	76.95	75.12	76.71	75.29	4.321347	73.34
2013-01-05	Apple	121.04M	75.61	73.60	74.57	74.84	4.315353	72.95
2013-01-06	Apple	114.68M	75.98	74.46	75.60	75.04	4.318021	72.68

```
1 df[['close', '10MA']].plot().set_title('APPLE moving average')
```

```
Text(0.5, 1.0, 'APPLE moving average')
```



# Progress so far

```
1 df['rolling_sigma'] = df['Change/return'].dropna().rolling(20).std() * np.sqrt(255)
2 df.head()
```

	Company	volume	high	low	open	close	close_log	10MA	20MA	50MA	Change/return	rolling_sigma
Time_Stamp												
2013-01-02	Apple	140.13M	79.29	77.38	79.12	78.43	4.362207	74.51	71.42	66.94	NaN	NaN
2013-01-03	Apple	88.24M	78.52	77.29	78.27	77.44	4.349503	73.90	70.75	66.64	-0.002912	NaN
2013-01-04	Apple	148.58M	76.95	75.12	76.71	75.29	4.321347	73.34	70.12	66.40	-0.006473	NaN
2013-01-05	Apple	121.04M	75.61	73.60	74.57	74.84	4.315353	72.95	69.52	66.19	-0.001387	NaN
2013-01-06	Apple	114.68M	75.98	74.46	75.60	75.04	4.318021	72.68	69.04	65.98	0.000618	NaN

```
[67] 1 df.rolling_sigma.plot(linestyle='--', marker='o')
      2 plt.ylabel('$\sigma$')
      3 plt.title('Apple Rolling Volatility')
```

Text(0.5, 1.0, 'Apple Rolling Volatility')



**We see that the probability of deviation is very less, except for 2018**

**Thank you.**  
**Feedback pointers & Questions.**