

Time series

* As the name suggests, the Time series deals with the data that is related to time

eg's :

- * stock market price
- * Daily weather Data
- * weekly sales

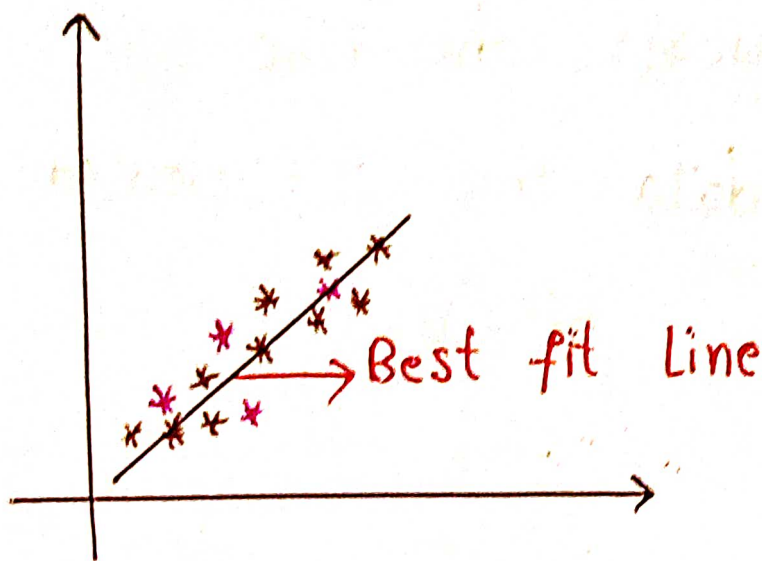
Before, we need to know what actually Time series data is used to predict.

For that, we need to understand two types of data for predictions:

1) Interpolation.

2) Extrapolation.

Interpolation:



* → Training Data

* → New Data

- * In interpolation, New Data will be near to best fit line only.

Extrapolation:

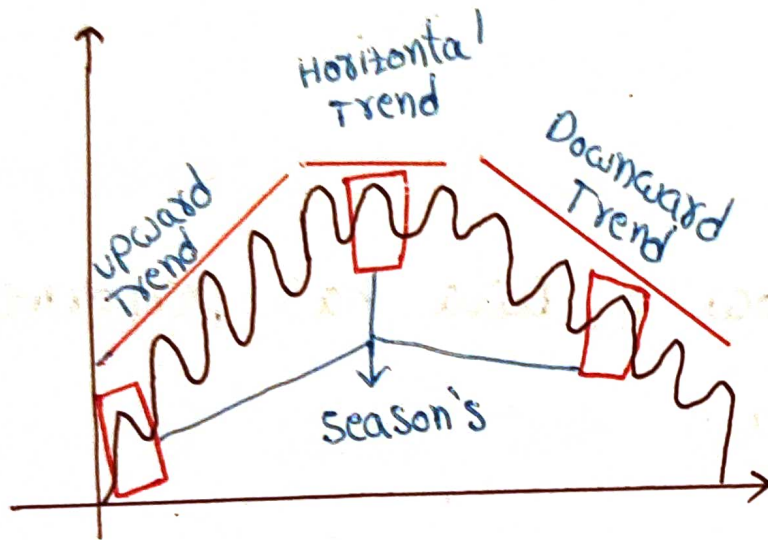


* → Training Data

* → New Data

- * In Extrapolation, New Data is the one which won't be near.
- * i.e, we've to predict past or future outputs also.

components of Time series :



There were 4 components of Time series, they are:

- 1) Trend
- 2) Season
- 3) cycle
- 4) Noise

Trend: It is the process of checking, how a time series data is performing.

* If increasing consequently, then it's upward Trend.

- * If decreasing, then it's Downward Trend.
- * If not changing it's value, then Horizontal Trend.

Season:

It represents how a Data has performed at each Time Interval.

eg's:

- * sales per week.
- * share price for every hour.
- * profits of net every Day.

cycle:

It's about how a data has performed at a certain long period of Time.

eg: How a stock performed during xyz Party's 5 year Ruling.

Noise: Deviation in the continuous Trend (very high, unnormal) due to some sensations.

eg's:

* Hindenburg Report.

* Ukraine/Russia war.

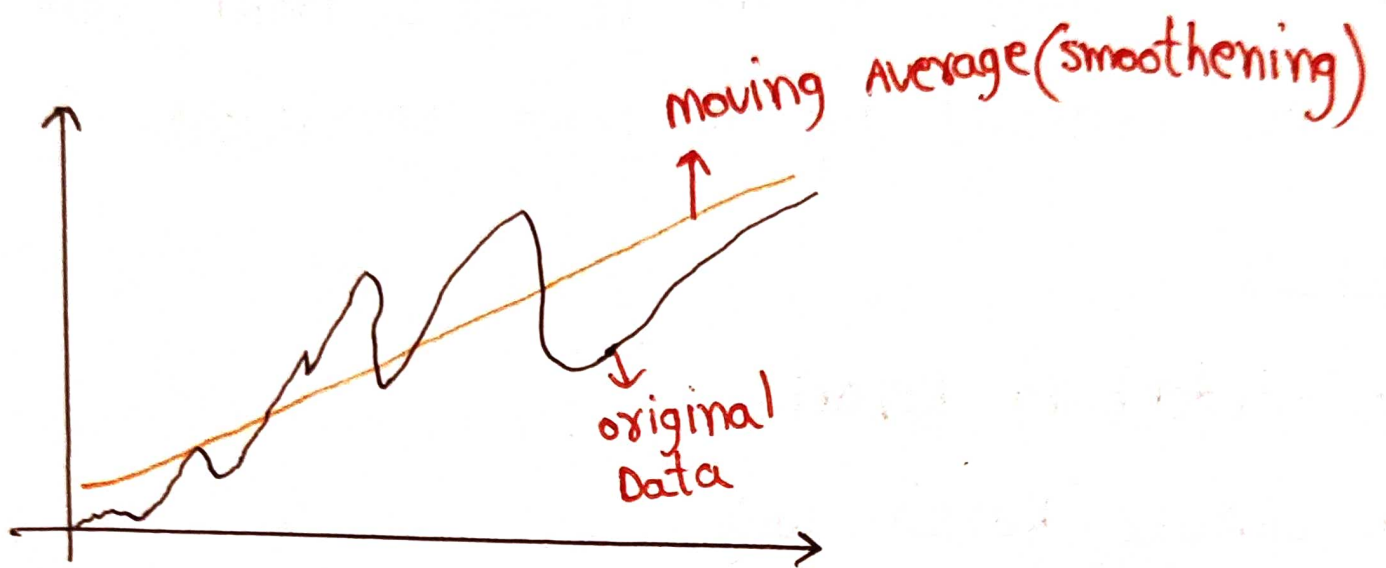
* Elon Musk stating Tesla accepts Bitcoin.

Moving Average:

By using Moving Average, we can draw a smoothening line to understand the patterns.

Types of Moving Averages:

- 1) simple Moving Average.
- 2) cumulative Moving Average.
- 3) Exponential Moving Average.



Simple Moving Average:

Here, we'll use a window size, and then uses the average of that windows to use for moving average line.

For eg Let's window size = 3

c1(Days)	c2(Data)	c3 (simple Moving Average)
D1	1	NAN
D2	5	NAN
D3	10	$10 + 5 + 1 / 3$
D4	12	$12 + 10 + 5 / 3$
D5	14	$14 + 12 + 10 / 3$

cumulative moving Average:

Here, the cumulative average for each column is used, to smooth the data.

c_1 (Day's)	c_2 (Data)	c_3 (Cumulative)
D1	10	$10/1$
D2	15	$10+15/2$
D3	20	$10+15+20/3$
D4	30	$10+15+20+30/4$
D5	50	$10+15+20+30+50/5$
D6	100	$10+15+20+30+50+100/6$
D7	90	$10+15+20+30+50+100+90/7$

Exponential moving Average:

- * Here Recent data will have more data weightage than the old data
- * i.e, old will have less weightage and new will have more weightage.

Stationary and non stationary:

A time series is said to be stationary when moving average and variance are constant overtime.

eg: Flat / Horizontal trend

The Time series is non stationary when moving average and variance are not constant overtime.

eg: upward trend, downward trend, etc...

* we can convert Non stationary Time series to stationary using some operations like

→ Difference

→ Log

→ Root, etc...

ACF and PACF:

Auto

corelation

Function

Partial

Auto

corelation

Function

* ACF is used to find the correlation between Time series data and lag of it.

Time	Data	Lag1	Lag2	Lag3
ω_1	9	NA	NA	NA
ω_2	8	9	NA	NA
ω_3	7	8	9	NA
ω_4	11	7	8	9
ω_5	14	11	7	8
ω_6	5	14	11	7
ω_7	6	5	14	11
	<hr/>	<hr/>	<hr/>	<hr/>
	y_t	y_{t-1}	y_{t-2}	y_{t-3}

* with this we can find $ACF(y_t, y_{t-1})$,
 (y_t, y_{t-2}) etc... and can apply correlations.

* As the name suggests, in PACF we'll use partial lags only, i.e we'll skip some.

year	profit	subs
1	96	-46
2	56	
3	50	19
4	54	
5	69	3
6	70	
7	72	

Finally, we'll use year & subs to find correlations

year	subs
2	-46
4	+19
6	+3