

## Topics Covered today

1. ACF/PACF Contd.
2. AR
3. MA
4. ARMA
5. ARIMA
6. SARIMA
7. Exogenous Variable
8. SARIMAX

# ⇒ ARIMA Family of Models

Agenda

Start at 9:05

① TS Decomposition → trend, seasonality, error  
Simple method →

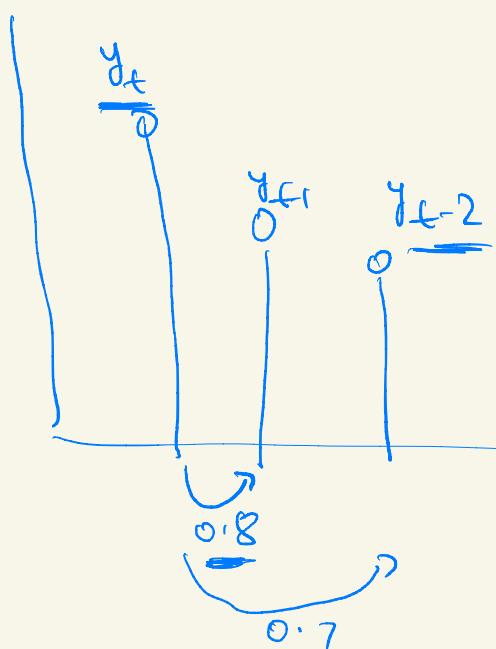
↓  
Smoothing method → SES, DSES, TES

Stationarity & ACF →

Holt's Whinters  
method

⇒ Auto-correlation → Correlation b/w  $y_t$  & any other prev value.

$y_t$        $y_{t-1}$   
 $y_{t-2}$



$$y_t \quad y_{t-1} = 0.8$$

$$y_t \quad y_{t-2} = 0.7$$

$$y_{t-1} \quad y_{t-2}$$

$A \rightarrow B$   
 $B \rightarrow C$ .



Auto

Partial Correlation  $\rightarrow$  Correlation b/w 2 time values after isolating them from impact of everything else.

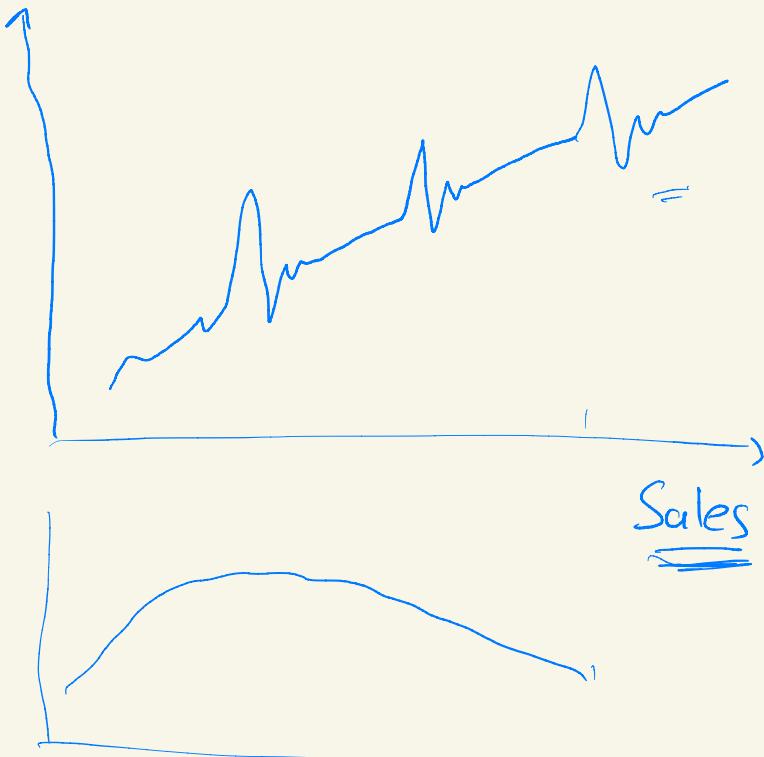
$$y_t = \alpha_1 y_{t-1} + \alpha_2 y_{t-2} + \dots$$

$$\Rightarrow y_{t-1} = \alpha_1 y_{t-2} + \alpha_3 y_{t-3} + \dots$$

$$y_{t-2} = \frac{1}{\alpha_1} y_{t-1} - \left( \frac{\alpha_3}{\alpha_1} y_{t-3} + \dots \right)$$

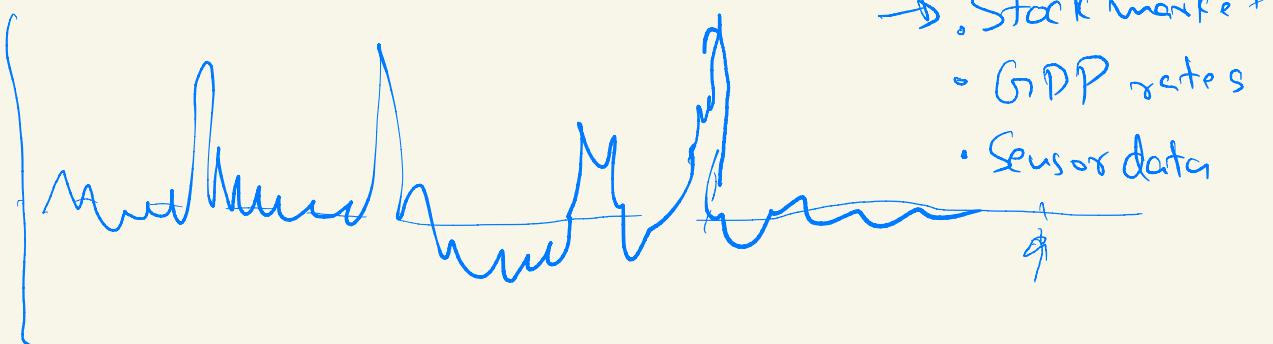
$$y_t = \alpha_1 y_{t-1} + \frac{\alpha_2}{\alpha_1} y_{t-1} - \left( \dots \right) + \left( \dots \right)$$

$\Rightarrow \text{TES} \rightarrow 3.6\% \text{ Error}$



Smoothing method  
↳ Trended  
Seasonal

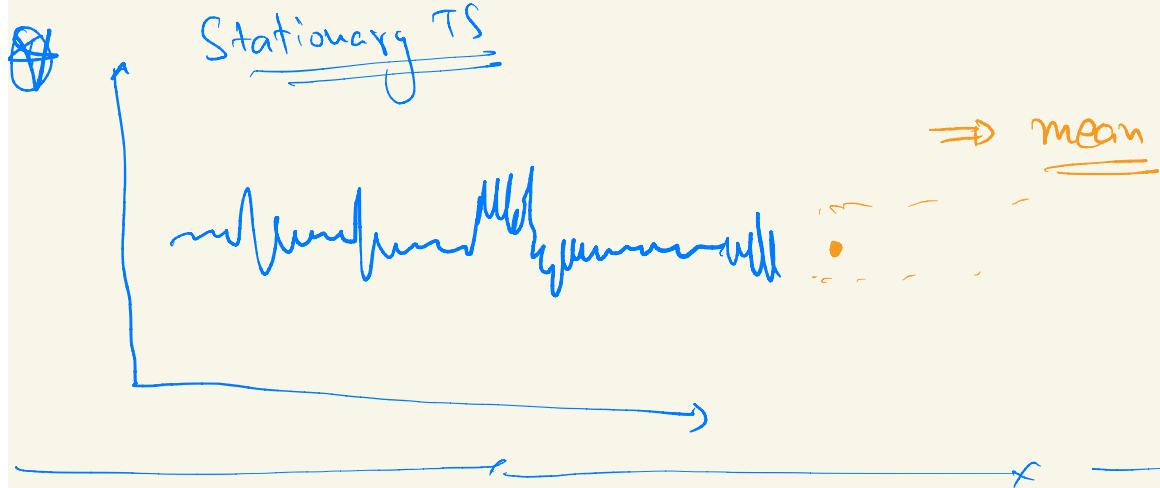
TES  $\rightarrow 2-4\%$



$\Rightarrow$  Stock market +  
• GDP rates  
• Sensor data

$$y = b(t) + s(t) + \underbrace{\delta(t)}_{z(t)} + e(t)$$

??



=> ARIMA Models

- 1) AR
- 2) MA
- 3) ARMA
- 4) ARIMA
- 5) SARIMA
- 6) SARIMAX

① AR → Auto regression  
on self

$$\hat{y}_t = \frac{\alpha_1}{q} y_{t-1} + \frac{\alpha_2}{q} y_{t-2} + \frac{\alpha_0}{q}$$

linear regression

$$3y_{t-1} - 2y_{t-2} + 1$$

$$\hat{y}_t = \alpha_1 y_{t-1} + \alpha_2 y_{t-2} + \dots + \alpha_{10} y_{t-10} + \alpha_0$$

Smoothing weights  
 → sum to 1  
 → decline rapidly

→ Auto Regressor parameters

$AR(p) \rightarrow AR(3), AR(4), \dots$

$$\hat{y}_t = [AR(3) - \alpha_1 y_{t-1} - \alpha_2 y_{t-2} - \alpha_3 y_{t-3} + \alpha_0]$$

$\Rightarrow 10, 11, 12, 13, 17, 20, 25, 27, 28, 29$  ~~29~~  $\stackrel{\text{TS Data}}{=}$

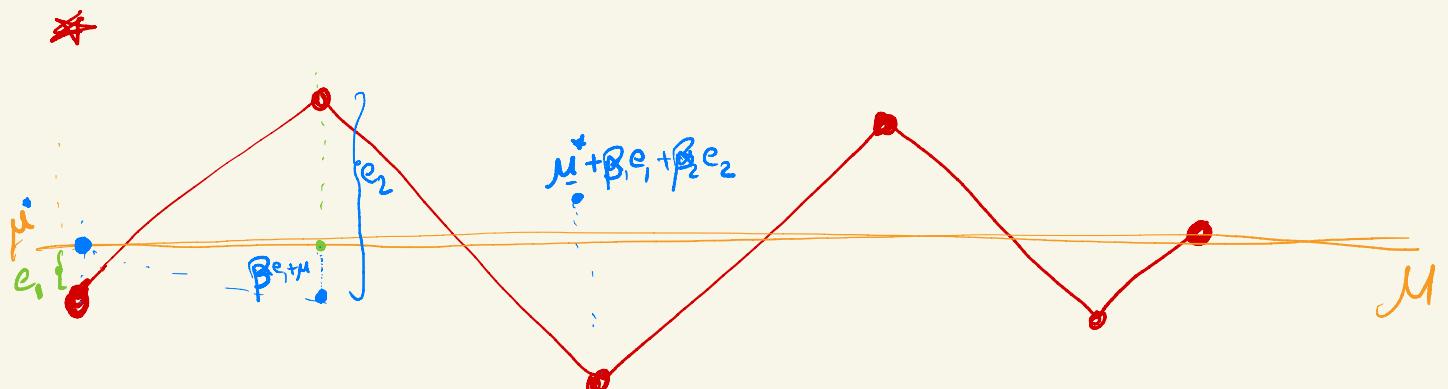
$y_{t-3}$	$y_{t-2}$	$y_{t-1}$	$y_t$
10	13	17	20
13	14	20	25
14	20	25	27
20	25	27	28
25	27	28	29

$X$        $Y$

normal LR problem

Auto Regression

② MA(q)  $\rightarrow$  ~~Moving Avg method~~ (Simple method)  
~~rolling avg~~



MA key Idea  $\Rightarrow$  I should learn from each error.

$$\hat{y}_t = \mu + \beta_1 e_1 + \beta_2 e_2 + \dots + \beta_q e_{t-q}$$

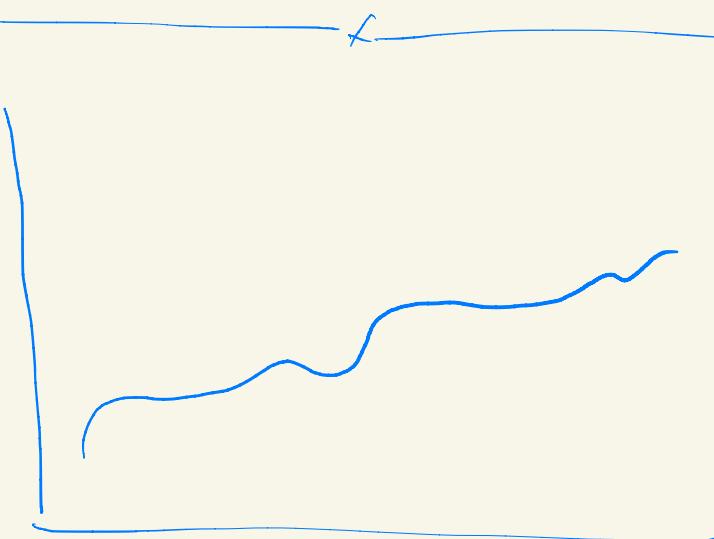
$\Rightarrow \mu_{t+1} \rightarrow$  data till  $t+1$   
 to predict  $t \rightarrow \hat{y}_t - \hat{y}_f$

$$\mu_t + \beta^+ e_t \rightarrow \hat{y}_{t+1}$$

③ ARMA  $\rightarrow$  AR + MA

$$\hat{y}_t = \zeta_0 + \underbrace{\alpha_1 y_{t-1} + \alpha_2 y_{t-2} + \dots + \alpha_p y_{t-p}}_{\text{AR component}} + \underbrace{\beta_1 e_{t-1} + \beta_2 e_{t-2} + \dots + \beta_q e_{t-q}}_{\text{MA component}}$$

$y(t)$



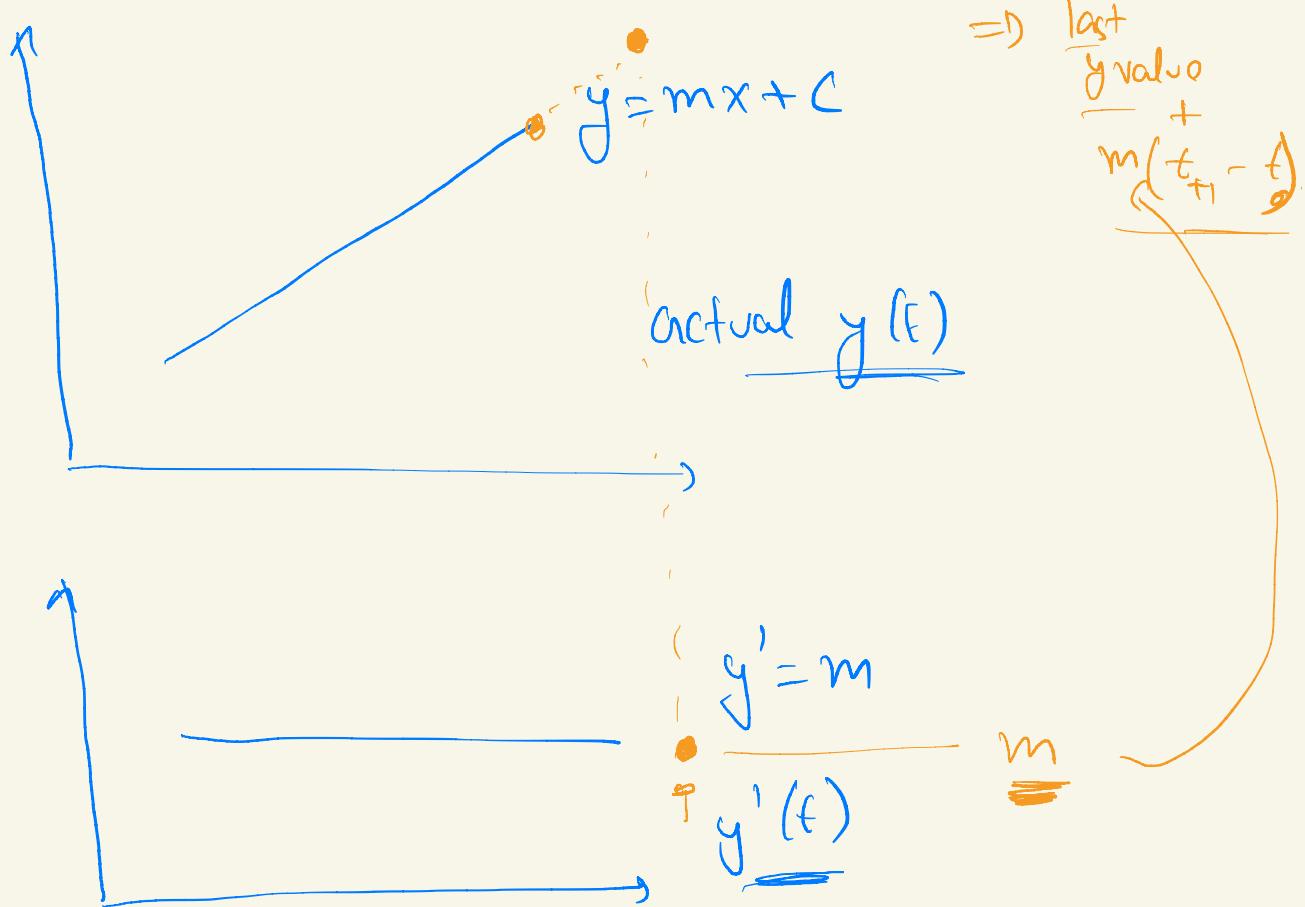
Actual 100, 120, 110, 115, 98, 127

Dift(y(t)) +20, -10, +5, -17, +29,

$\hat{y}'(t)$

19  $\rightarrow$  +4  $- 13$  +20

$\Rightarrow$  Differentiate  $\rightarrow$  Integrate



$$y_t = mx_t + C$$

$$x_{t+1} = x_t + 1$$

$$\underline{y_{t+1}} = mx_{t+1} + C$$

$\hat{y} \Rightarrow$  Last Value + Cumsum of Prediction of  $y'$

⇒ Take TS

↓  
→ Differentiate → decompose  
diff (detrend) =&

Predict using AR, MA, ARMA

↓  
→ Integrate → (Cumsum) &

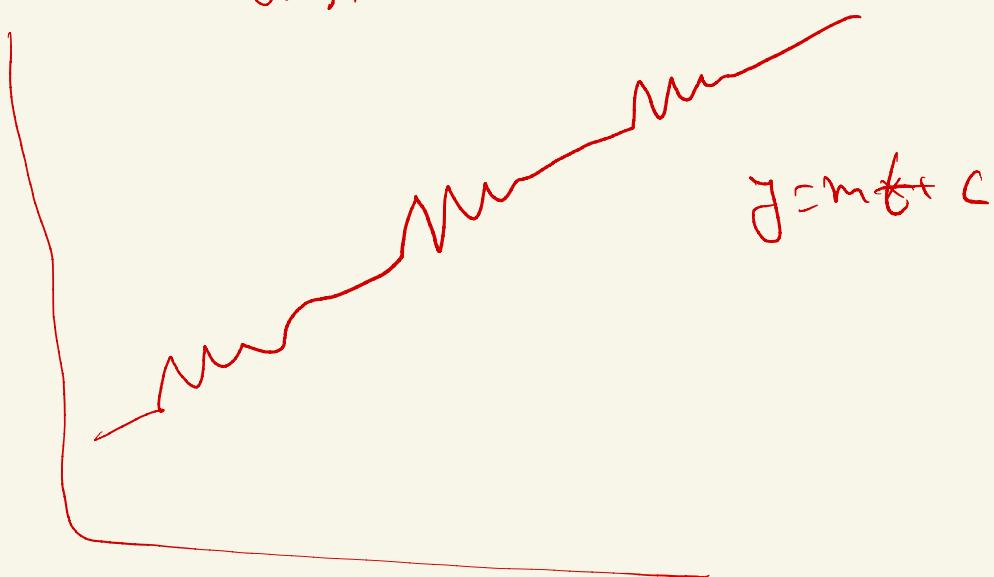
Final Answers

ARIMA

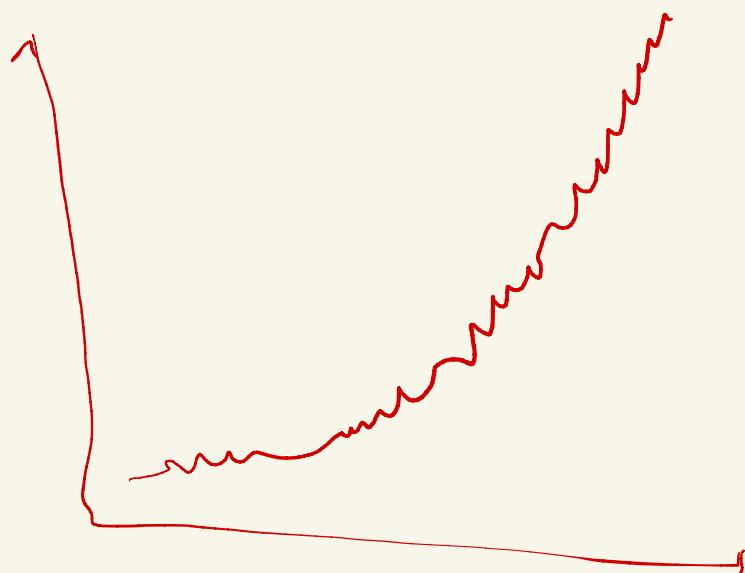
↓  
Integration → Auto

$$\text{Trend} \rightarrow y = mx + c$$

• diff



$$y = mx + c$$



$$y = bt^2 + ct + d$$

Single differentiation  
2 differentiation

$$t \Rightarrow 100 \quad 120 \quad 115 \quad 117 \quad 98 \quad 123$$

$$y' \Rightarrow 20 \quad -5 \quad +2 \quad -19 \quad +25$$

$$\underline{y''} \Rightarrow 25 \quad -7 \quad 21 \quad -44$$

Order =  $(P, d, q)$

→  $P \rightarrow AR$  ~~lag~~ terms

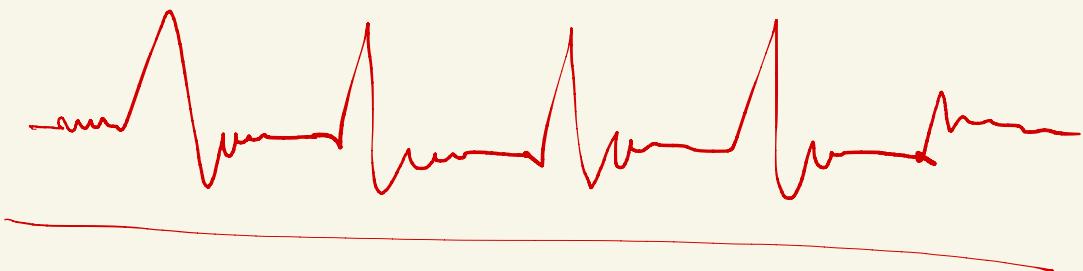
→  $d \rightarrow$  no. of differentiation to do

→  $q \rightarrow$  Count of MA terms

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⇒ SARIMA  $\leftarrow (P, d, q, P, D, Q, S\right)$

S → Seasonal  $\rightarrow$  Impact of very old/  
Seasonal



Jan Sale → Seasonality →

$$AR(P) = \alpha_1 y_{t-1} + \alpha_2 y_{t-2} + \dots + \alpha_p y_{t-p} + \alpha_0$$

$$\underline{SAR(P)} = \gamma_1 y_{t-S} + \gamma_2 y_{t-2S} + \dots + \gamma_p y_{t-PS} + \gamma_0$$

$$\underline{\text{SMA}(\ell)} = \delta_1 e_{t-s} + \delta_2 e_{t-2s} + \dots + \delta_\ell e_{t-\ell s}$$

$$\boxed{\text{SARIMA}} = C + \alpha_1 y_{t-1} + \alpha_2 y_{t-2} + \dots + \alpha_p y_{t-p} \\ + \beta_1 e_{t-1} + \dots + \beta_q e_{t-q} \\ + \gamma_1 y_{t-s} + \dots + \gamma_p y_{t-ps} \\ + \delta_1 e_{t-s} + \dots + \delta_\ell e_{t-\ell s}$$

$P \rightarrow \text{AR}$

$q \rightarrow \text{MA}$

$d \rightarrow \text{ARIMA}$

$P \rightarrow \text{SAR}$

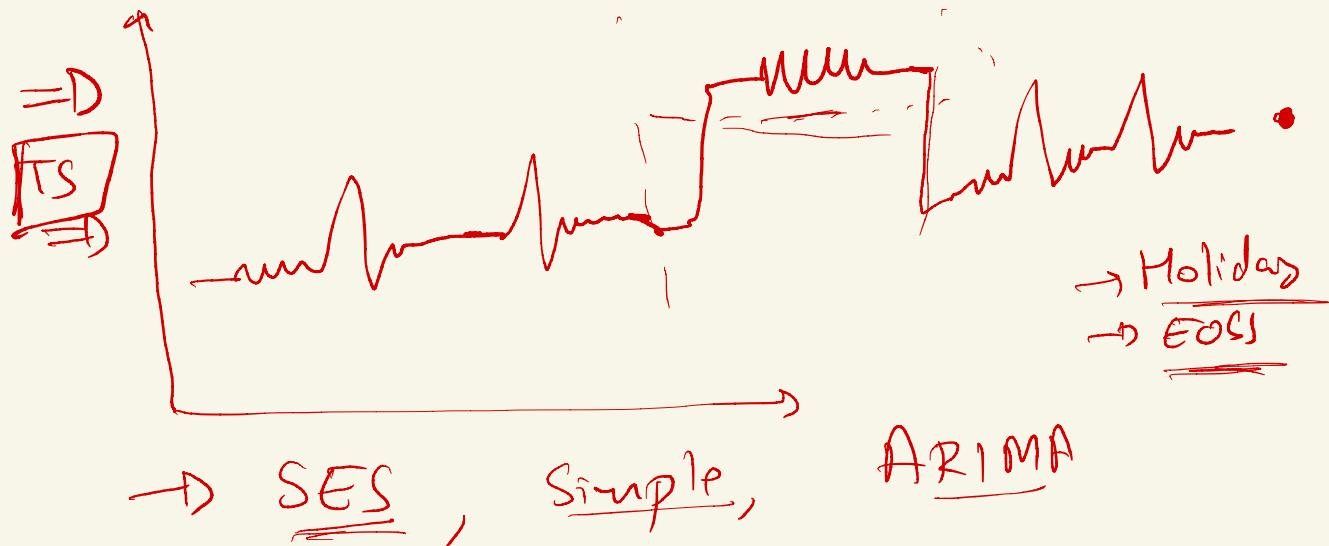
$Q \rightarrow \text{SMA}$

$D \rightarrow \text{SARIMA}$

$S \rightarrow \underline{\text{Seasonality}}$

# SARIMAX

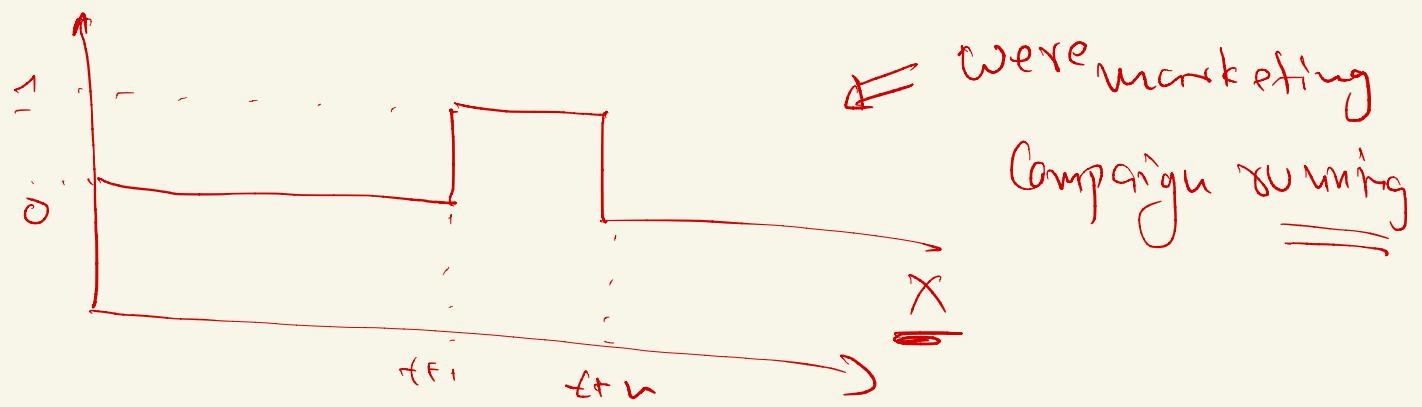
~~↳ Exogenous Variables → ??~~



=D Seasonality → D-Winter

① Holiday EOSS →

② we give extra information to the model.



$$\Rightarrow \text{SARIMA}_{\underline{MAX}} \rightarrow \alpha_1 \\ \beta_1 \\ \gamma_1 \\ \delta_1 \\ \text{weights} \\ \text{extra variable} \rightarrow w_1 x_1 + w_2 x_2 + \dots + w_n x_n$$

### Practical Approach

Data  $\rightarrow$  EDA  $\rightarrow$  missing values  
+ outliers

$\rightarrow$  TS decompose

$\hookrightarrow$  TES  $\rightarrow$

2-1%

( )

$\rightarrow$  PACF Plot

$\hookrightarrow P, q, P, Q, S$

$\hookrightarrow$  SARIMA model  $\rightarrow$  2-4%

