

# Parameters of ARIMA model

# Recap: Introduction to ARIMA

- Time Series forecasting model
- **ARIMA**: 'Auto Regressive Integrated Moving Average'

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- Time Series forecasting model
- **ARIMA**: 'Auto Regressive Integrated Moving Average'
- ARIMA has three parameters
  - Auto regressive - p
  - Integrated - d
  - Moving Average - q

# Recap: Parameters of ARIMA model

- ARIMA has three parameters

- Auto regressive -  $p$
- Integrated -  $d$
- Moving Average -  $q$

$p, d, q$



# Recap: Parameters of ARIMA model

- ARIMA has three parameters
  - Auto regressive -  $p$
  - Integrated -  $d$
  - Moving Average -  $q$

Differencing

# Parameters of ARIMA model

- ARIMA has three parameters

- Auto regressive - p

- Integrated - d

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# Parameters of ARIMA model

- ARIMA has three parameters

- Auto regressive - p

ACF (autocorrelation function) plot

- Integrated - d

- Moving Average - q

PACF (partial autocorrelation function) plot

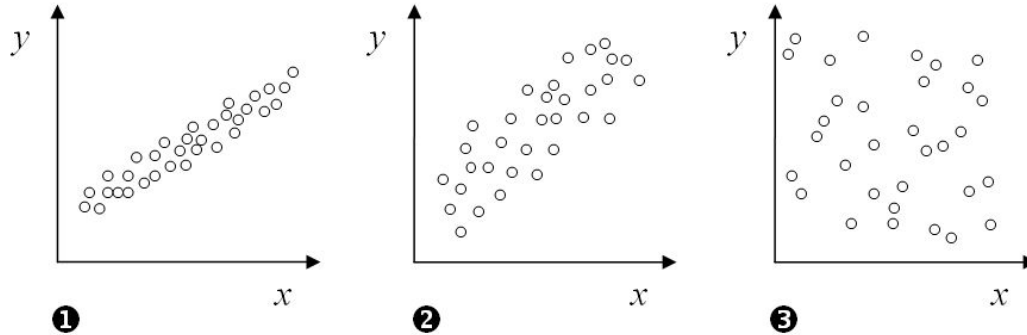
# Parameters of ARIMA model

- Some important concepts before we move over to determining the values of  $p$ , and  $q$ -
  - Correlation
  - Autocorrelation
  - Partial Autocorrelation

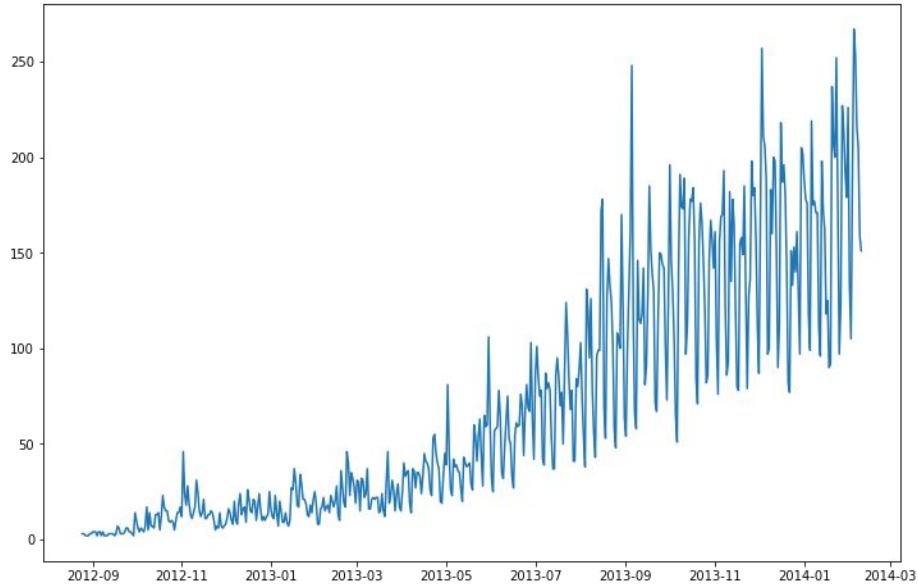


# Correlation

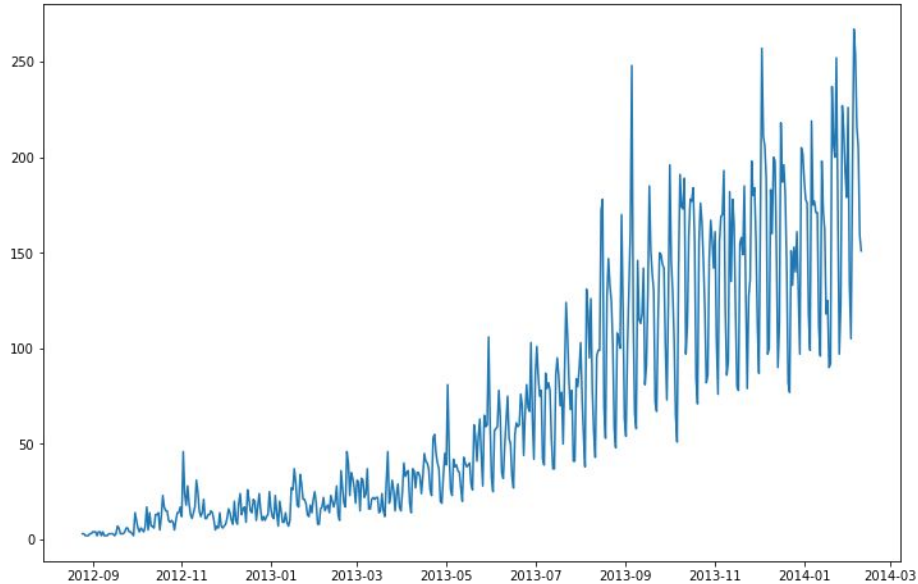
Correlation refers to the strength of mutual relationship between quantities.



# Auto-Correlation



# Auto-Correlation



32

36

44

25

30

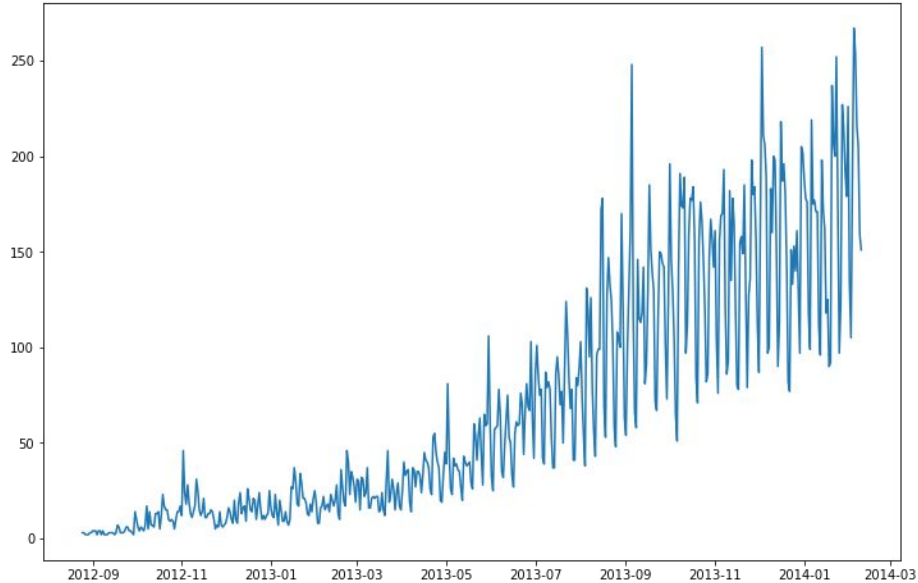
37

27

11

10

# Auto-Correlation



32

36

44

25

30

37

27

11

10

t-1

32

36

44

25

30

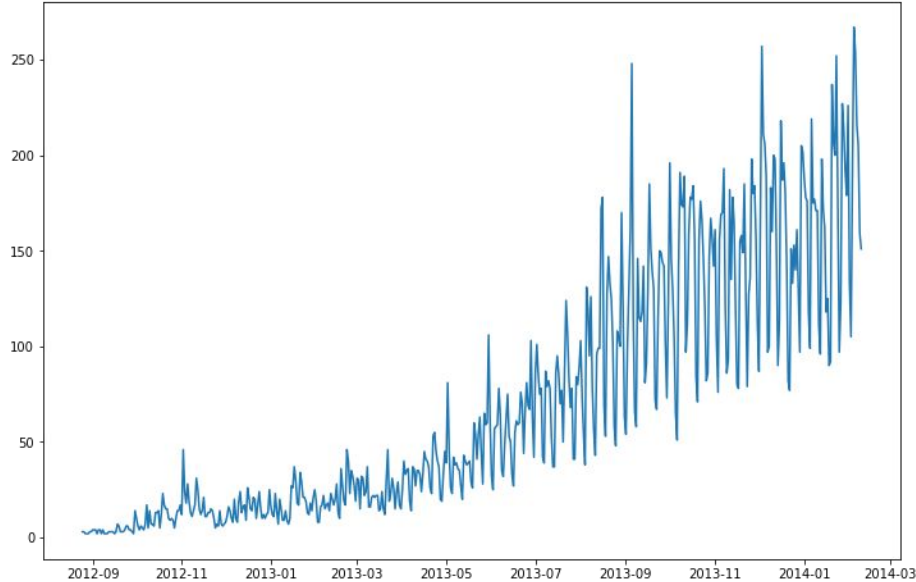
37

27

11

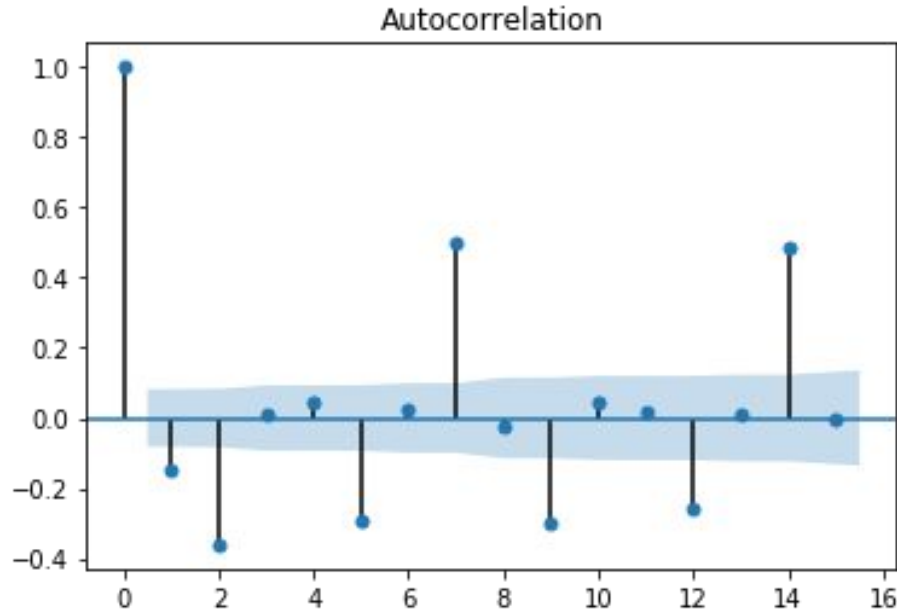
10

# Auto-Correlation

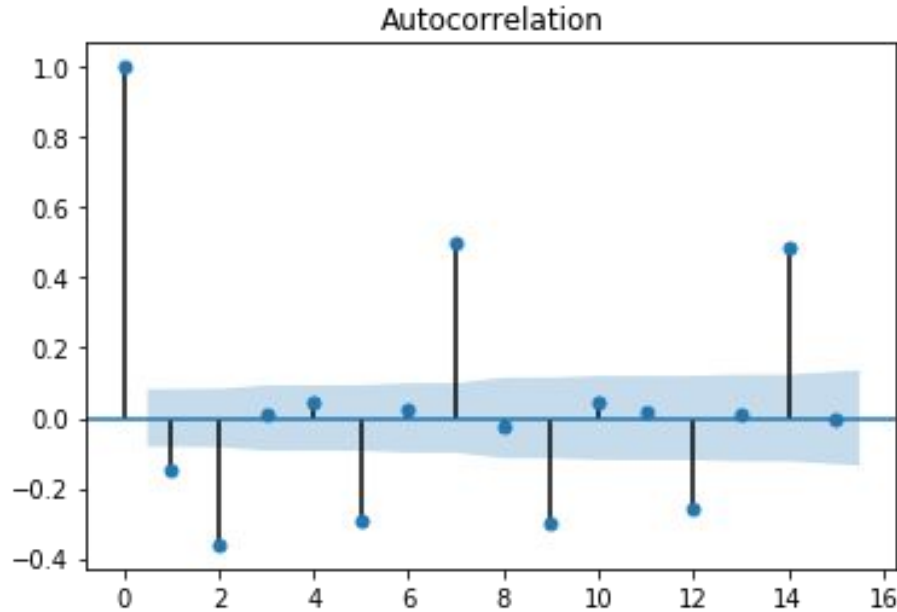


	t-1	t-2
32		
36	32	
44	36	32
25	44	36
30	25	44
37	30	25
27	37	30
11	27	37
10	11	27
10		

# Auto-Correlation Function (ACF)

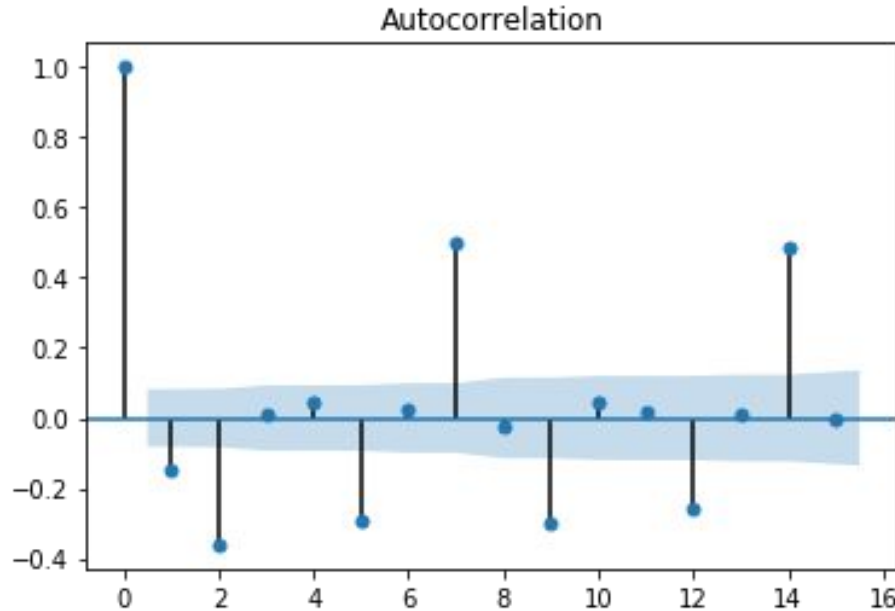


# Auto-Correlation Function (ACF)



$$\frac{1.96}{\sqrt{n}}$$

# Auto-Correlation Function (ACF)



AR(2)

$$y(t) = a_1 y(t-1) + a_2 y(t-2) + u$$

Lag = 2 have significant correlation



# Summarizing - Auto-Correlation Function (ACF)

- Auto Correlation refers to how correlated is the series with its past self
- ACF plot shows the auto correlation values
- ACF plot can be used to select the value of  $p$  for ARIMA

# Partial Auto-Correlation

# Partial Auto-Correlation

AR(1)

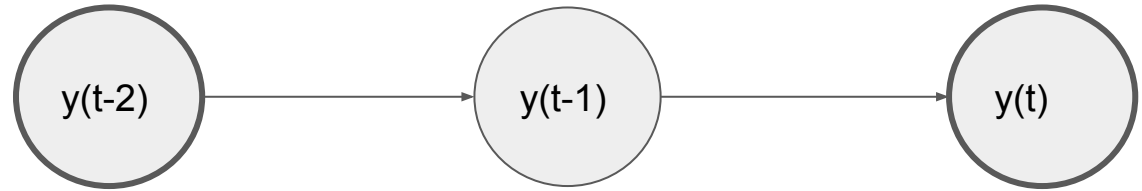
$$y(t) = a_1 y(t-1) + \text{const.}$$



# Partial Auto-Correlation

AR(1)

$$y(t) = a_1 y(t-1) + \text{const.}$$



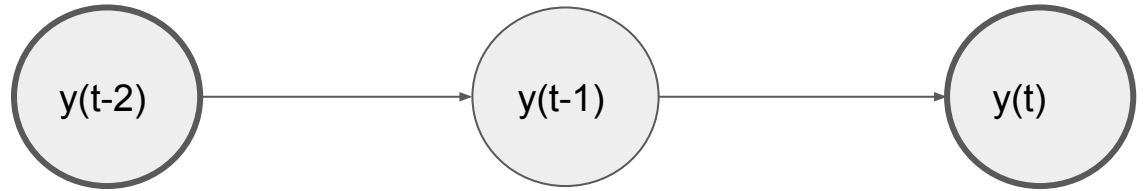
# Partial Auto-Correlation

AR(1)

$$y(t) = a_1 * y(t-1) + \text{const.}$$

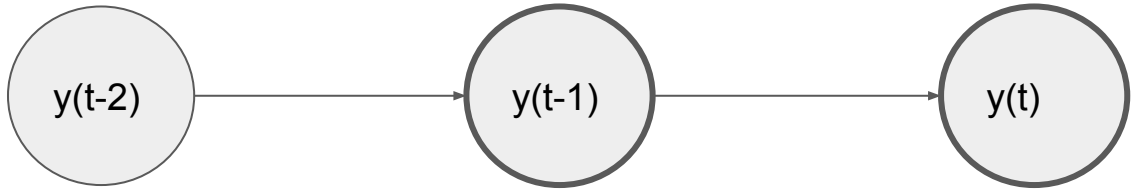
AR(2)

$$y(t) = a_1 * y(t-1) + a_2 * y(t-2) + \text{const.}$$



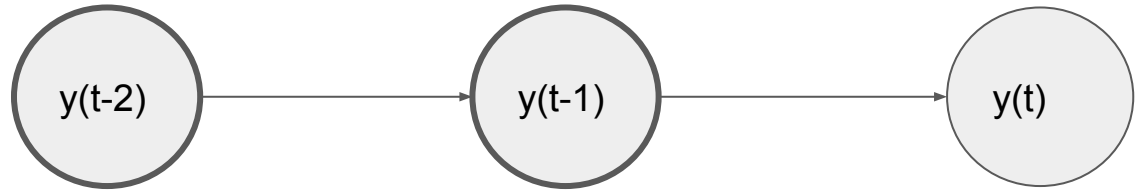
# Partial Auto-Correlation

1. Amount of variance in  $y(t)$  which is not explained by  $y(t-1)$



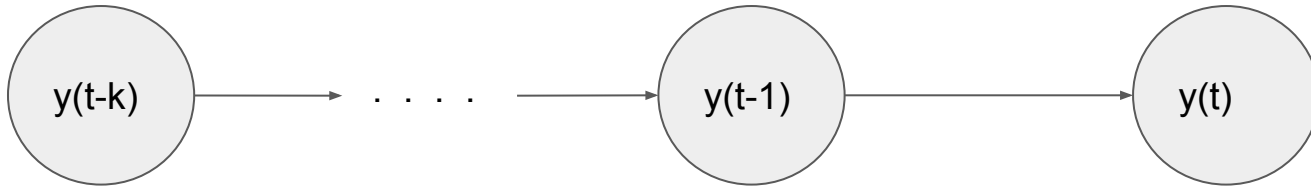
# Partial Auto-Correlation

1. Amount of variance in  $y(t)$  which is not explained by  $y(t-1)$
2. Amount of variance in  $y(t-2)$ , which is not explained by  $y(t-1)$



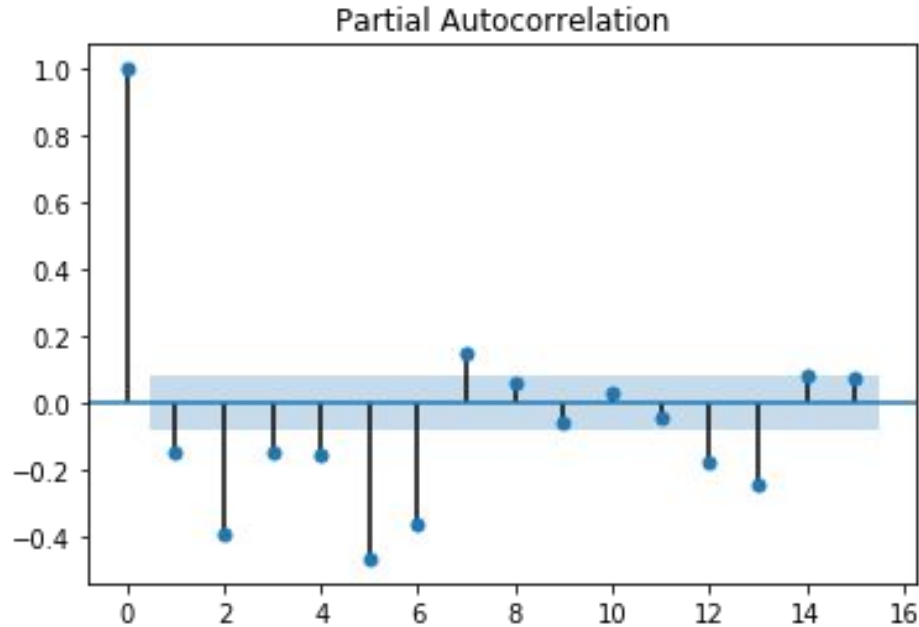
# Partial Auto-Correlation

Partial Auto-Correlation: Correlation between  $y(t)$  and  $y(t-k)$  after removing correlation of time steps in between

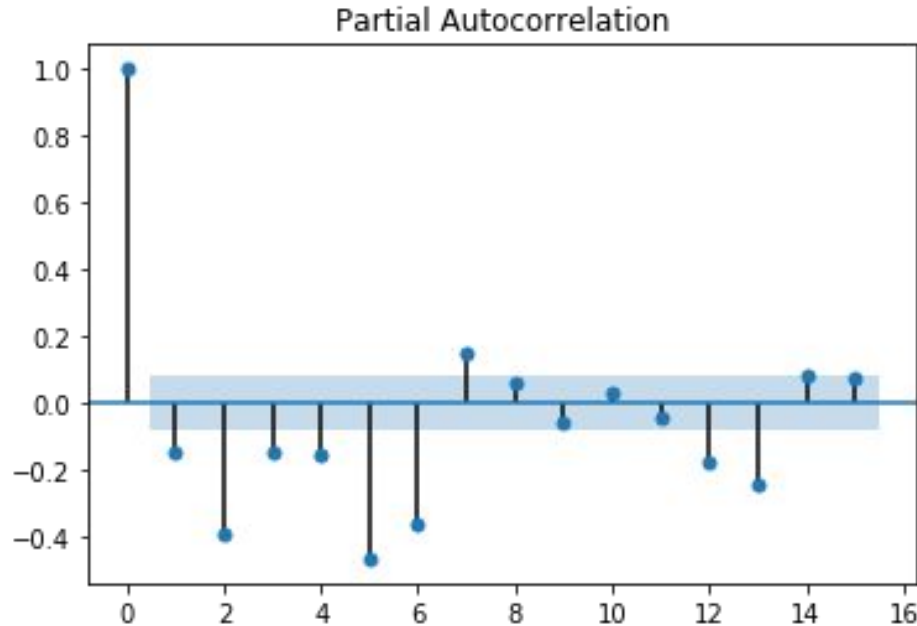




# Partial Auto-Correlation Function (PACF)



# Partial Auto-Correlation Function (PACF)



MA(2)

$$y(t) = E(t) + a_1 \cdot E(t-1) + a_2 \cdot E(t-2) + u$$

Lag = 2

# Summarizing - Partial Auto-Correlation Function (PACF)

- Partial Auto Correlation refers to the correlation with the residuals of the lags.
- PACF plot shows the partial auto correlation values
- PACF plot can be used to select the value of  $q$  for ARIMA

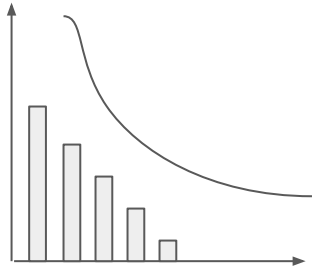
Thank You

# Autoregressive Models - ACF

$$y(t) = a_1 * y(t-1) + a_2 * y(t-2) + \dots + a_p * y(t-p) + E(t)$$

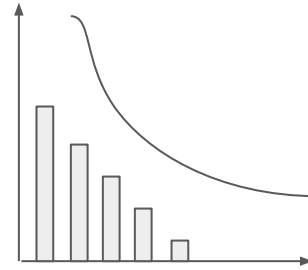
AR(1)

$$y(t) = a_1 * y(t-1) + E(t)$$



AR(2)

$$y(t) = a_1 * y(t-1) + a_2 * y(t-2) + E(t)$$

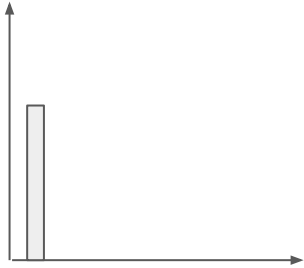


# Autoregressive Models - PACF

$$y(t) = a_1 * y(t-1) + a_2 * y(t-2) + \dots + a_p * y(t-p) + E(t)$$

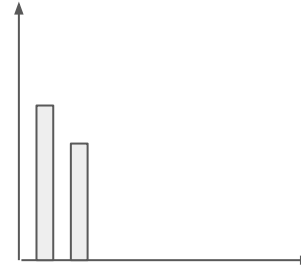
AR(1)

$$y(t) = a_1 * y(t-1) + E(t)$$



AR(2)

$$y(t) = a_1 * y(t-1) + a_2 * y(t-2) + E(t)$$

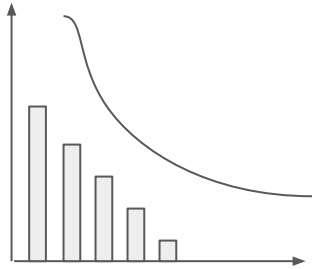


# Moving Average Models - PACF

$$y(t) = u + E(t) + a_1 * E(t-1) + a_2 * E(t-2) + \dots + a_q * E(t-q)$$

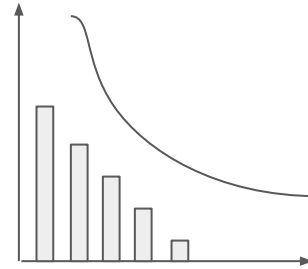
MA(1)

$$y(t) = u + E(t) + a_1 * E(t-1)$$



MA(2)

$$y(t) = u + E(t) + a_1 * E(t-1) + a_2 * E(t-2)$$

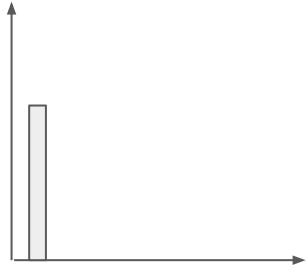


# Moving Average Models - ACF

$$y(t) = a_1 * y(t-1) + a_2 * y(t-2) + \dots + a_p * y(t-p) + E(t)$$

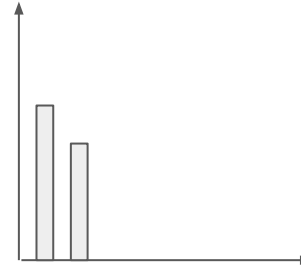
AR(1)

$$y(t) = a_1 * y(t-1) + E(t)$$

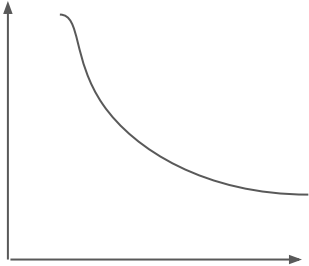
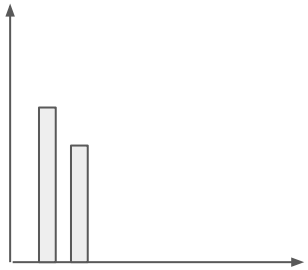
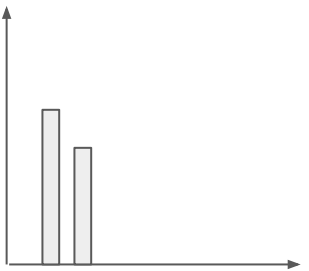
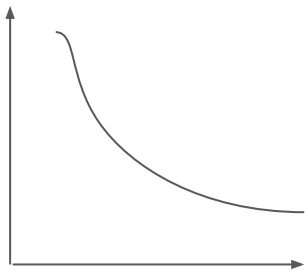


AR(2)

$$y(t) = a_1 * y(t-1) + a_2 * y(t-2) + E(t)$$





	ACF	PACF
AR		
MA		

AR

ACF

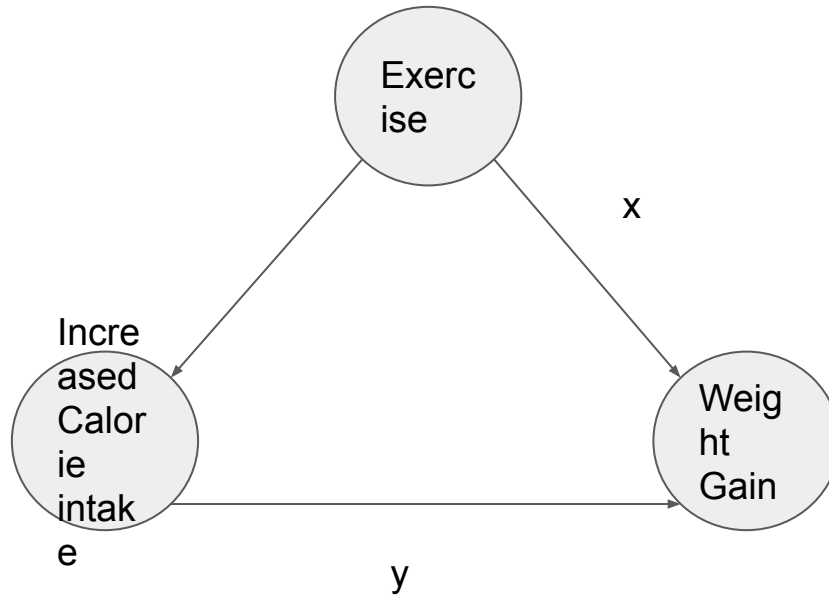
PACF

MA

# Summarizing - Auto-Correlation Function (ACF)

- Lag 1 has positive autocorrelation? AR model
- -ve correlation at lag 1? MA model
- ACF drops off at lag  $k$  = AR( $k$ ) model
- PACF Gradual decrease? MA model

# Partial Autocorrelation



# Differencing - d

- Differencing
  - This is done to stabilize the mean by removing changes in the level
  - Differencing is performed by subtracting the previous observation from the present observation

$$[ y(t) = x(t) - x(t-1) ]$$