# Retail Giant Sales Forecasting assignment

Submitted by-

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#### Introduction

- Global Mart is an online supergiant with multiple categories such as consumer, corporate and home office.
- It operates in Different regions and wants us to identify the key market segment and predict sales inmost profitable market segment of next 6 months.

#### Problem Statement

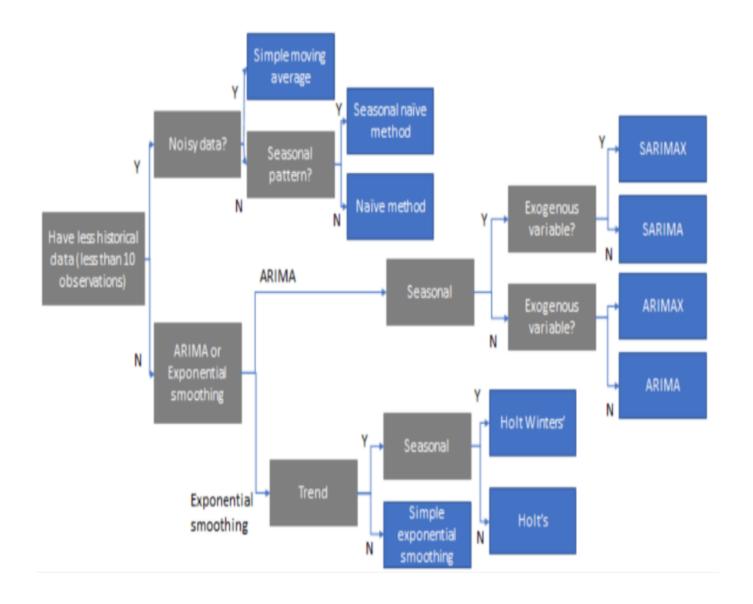
- We need to forecast the Sales of next 6 months for most consistent market Segment for given Global Store.
- We need to follow Data Preparation and Forecasting techniques using various Smoothing and ARIMA models techniques.

#### Coefficient of Variation(CoV) for Market-Segment

segment	cov
Consumer-APAC	0.522725
Consumer-Africa	1.310351
Consumer-Canada	1.250315
Consumer-EMEA	2.652495
Consumer-EU	0.595215
Consumer-LATAM	0.683770
Consumer-US	1.010530
Corporate-APAC	0.530051
Corporate-Africa	1.891744
Corporate-Canada	1.786025
Corporate-EMEA	6.355024
Corporate-EU	0.722076
Corporate-LATAM	0.882177
Corporate-US	1.071829
Home Office-APAC	1.008219
Home Office-Africa	2.012937
Home Office-Canada	2.369695
Home Office-EMEA	7.732073
Home Office-EU	0.938072
Home Office-LATAM	1.169693
Home Office-US	1.124030

- ➤ There are in total 21 Distinct Market Segments and out of which 'Consumer-APAC' Market Segment is having the least value of Covariance.
- ➤ Thus, based on Covariance value we can say that 'Consumer-APAC' is the most Consistent Market Segment.
- The lest consistent market segment is 'Home Office-EMEA' as it have the highest COV value.

#### Optimum technique from Flowchart that might work best for predicting sales.



- As per the flow chart, we have Historical data of more than 10 observations.
- From ARIA techniques, we can observe Seasonality in data, without any exogenous variable.
- Preferred technique can be considered as SARIMA method in case of ARMIA Models.
- From Smoothing techniques, we prefer Holt Winters over other Smoothing method as it captures Seasonality along with Level and Trend in the Global Market data set.

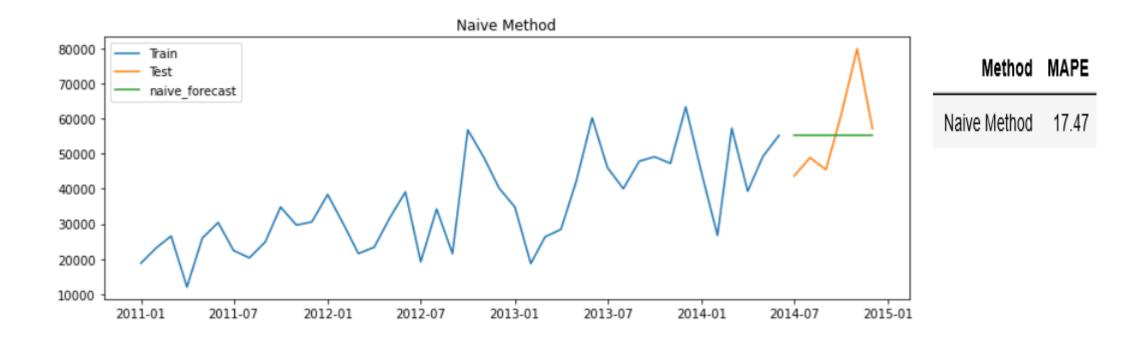
## Smoothing Technique

Following techniques were used to forecast next six month sales.

- Naive method
- Simple average
- Simple Moving Average
- Simple Exponential Smoothing technique
- Holt method
- Holts Winter Additive method
- Holts Winter Multiplicative method

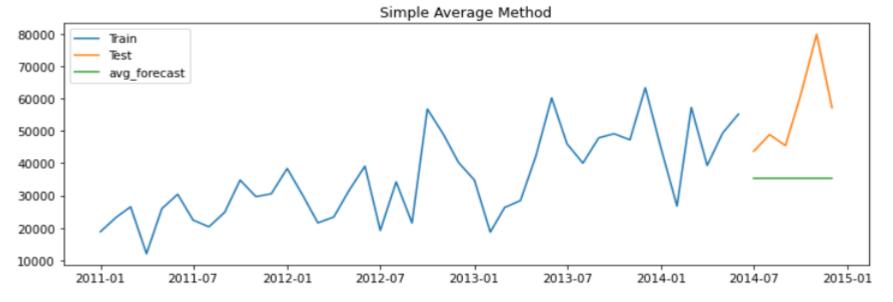
#### Naive method

- > Here, the green Straight line parallel to axis represents the Naive forecast for 6 months duration.
- ➤ In Naive Method, Sales is predicted based on the Last Months Sale.
- ➤ It captures only the LEVEL with the forecast value as 55151.4699.
- > APE value was calculate to be around 17.47



#### Simple Average method

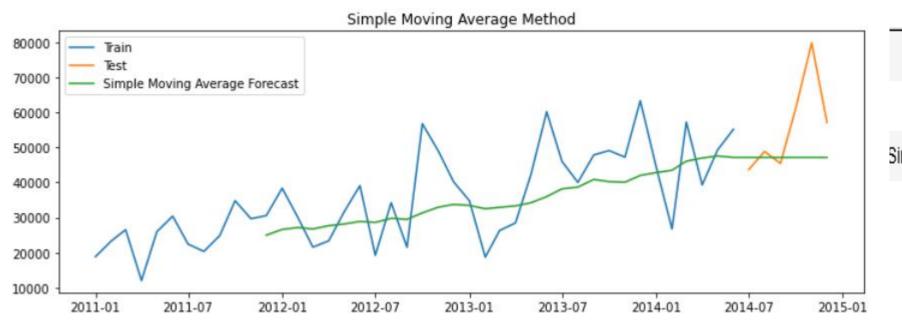
- > Here, the Green Straight line parallel to axis represents the Simple Average forecast for 6 months duration.
- > In Simple Average Method, Predicted Sales is the Average of all the previous months Train data.
- ➤ It captures the Level of average values with forecast value as 35253.632814.
- ➤ MAPE value of Simple Average Method is greater than Naive method in this case.
- > It means Naive method is better among these two methods for the current problem statement.



	Me	thod	M	APE
/e	е Ме	thod	1	7.47
je	е Ме	thod	3	34.34

#### Simple Moving Average method

- > Here, the Green line represents the Simple Moving Average.
- > In Simple Moving Average Method, we have taken Moving window size of 12 months.
- Predicted Sales is the Average of moving 12 months Train data window.
- > It gives importance to the weight of each window and captures the increasing Trend.
- Based on MAPE value, Simple Moving Average method is performing better than Simple Average method.

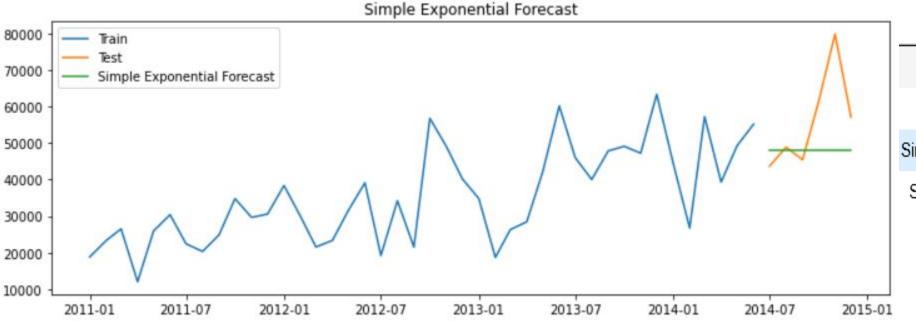


Naive Method	17.47
Simple Average Method	34.34
Simple Moving Average Method	16.10

Method MAPE

#### Simple Exponential Smoothing

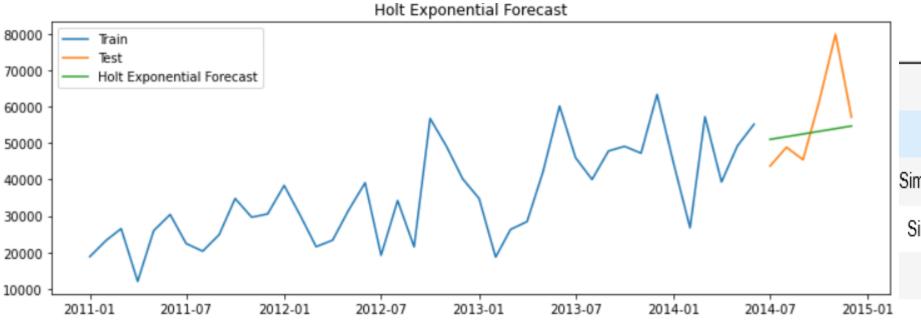
- > Here, the green line parallel to axis represents the Simple Exponential Smoothing.
- In Simple Exponential Smoothing Method, most recent value gets the Higher weight while, the Older values gets the lower weight assigned to them.
- > It captures only the Level of Data, by moving Average window with higher weightage to closest/latest data than older/past data.



Method	MAPE
Naive Method	17.47
Simple Average Method	34.34
Simple Moving Average Method	16.10
Simple Exponential Smoothing	15.83

#### Holt method

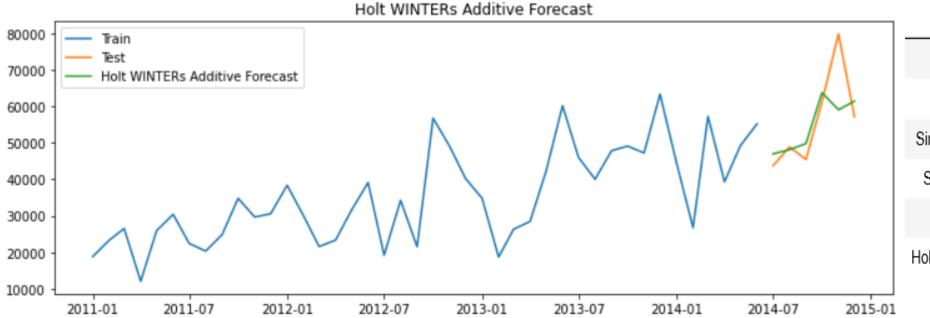
- ➤ Here, the green line represent an Upward trend which is Holts Exponential forecast.
- > It captures both Level as well as Trend in the forecast.
- ➤ An upward increasing trend and level is captured as part of the model.



Method	MAPE
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Simple Exponential Smoothing	15.83
HOLT Exponential Forecast	14.67

#### Holt Winters Additive method

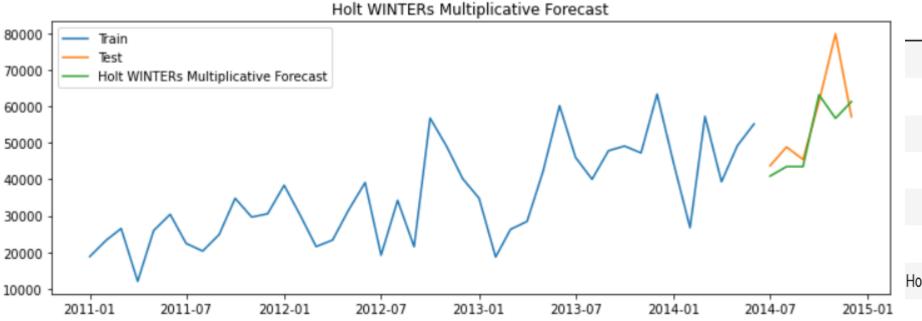
- ➤ Here the green line, represents the Holts Winter Additive forecast, obtained by following an Additive approach.
- ➤ We have captured the Level and an Upward Increasing Trend and Seasonality, which is quite close to actual Test data.



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Holt WINTERs Additive Forecast	9.39

#### Holt Winters Multiplicative method

- > Here, the green Line is the Holts Winter Multiplicative forecast .
- > It captures the Trend and level and Seasonality using Multiplicative approach in calculation.
- > We observe the Forecast value is quite close to the test data.



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Holt WINTERs Multiplicative Forecast	10.20

#### Inference

- ➤ Based on the Smoothing techniques, we can see lowest MAPE value is for Holts Winter Additive Forecast- which means its best suited for prediction.
- ➤ Visually, the forecast for Holts Winter Additive and Multiplicative method appears much closer to actual 6 months data than other smoothing techniques.
- ➤ We were able to capture the Seasonality in data in Holt Winters method.

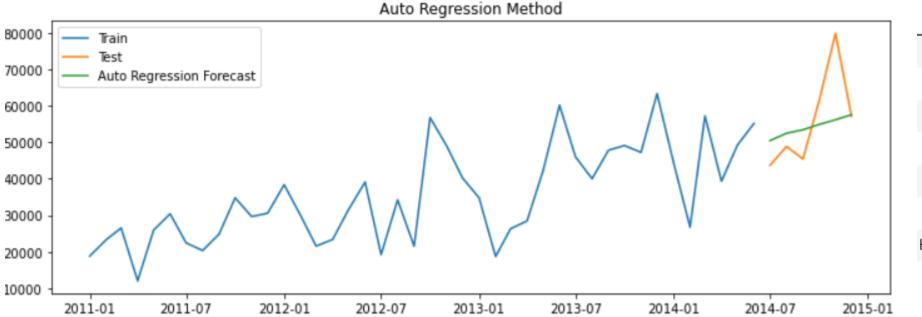
#### ARIMA techniques

Following is the list of ARIMA techniques executed:

- Simple Auto Regressive(AR)
- Moving Average(MA)
- Auto Regressive Moving Average(ARMA)
- Auto Regressive Integrated Moving Average(ARIMA)
- Seasonal Auto Regressive Integrated Moving Average(SARIMA)

#### Simple Auto Regressive(AR) method

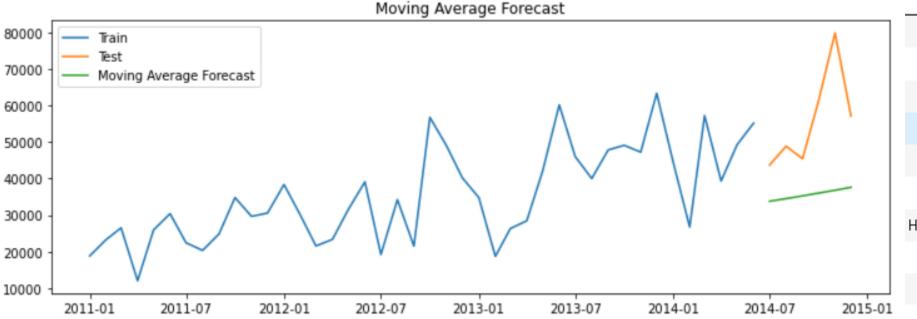
- ➤ Here, the green line indicates the Auto Regression method
- > Auto regression method predicts the future observation as linear regression of past Sales data.
- In this case, value of Lag order i.e. p is taken as 1



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Auto Regressive (AR)method	13.56

#### Moving Average(MA) method

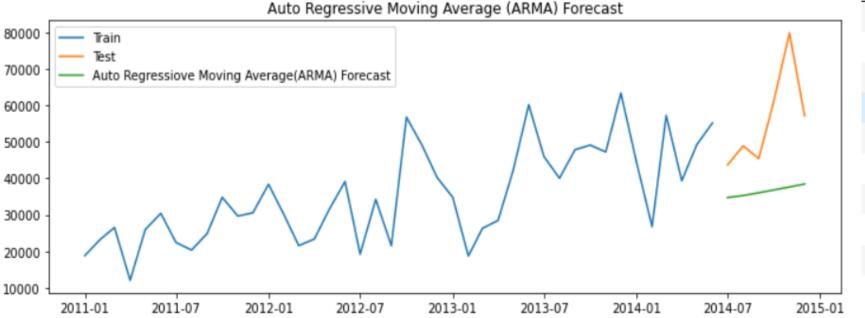
- ➤ Here, the green line indicates the Moving Average Forecast.
- ➤ Moving Average Forecast method predicts the Forecast value ,using the past forecast errors.
- ➤ It has single parameter window size i.e. q=1



MAPE	Method
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13.56	Auto Regressive (AR)method
33.93	Moving Average (MA) method

#### ARMA(Auto Regressive Moving Average) method

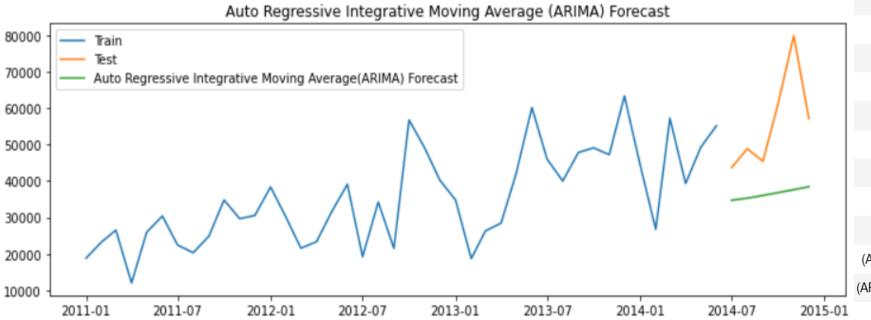
- ➤ Here, the Green line indicates the Auto Regressive Moving Average Forecast
- > ARMA method predicts the Forecast value by combining both AR and MA method.
- ➤ It has two parameters : Lag Order(p=1) and window size(q=1).



#### Method MAPE Naive Method 17.47 Simple Average Method 34.34 Simple Moving Average Method 16.10 Simple Exponential Smoothing 15.83 **HOLT Exponential Forecast** 14.67 Holt WINTERs Additive Forecast 9.39 Holt WINTERs Multiplicative Forecast 10.20 Auto Regressive (AR)method 13.56 Moving Average (MA) method 33.93 (ARMA) Auto Regressive Moving Average method

#### ARIMA(Auto Regressive Integrate Moving Average) method

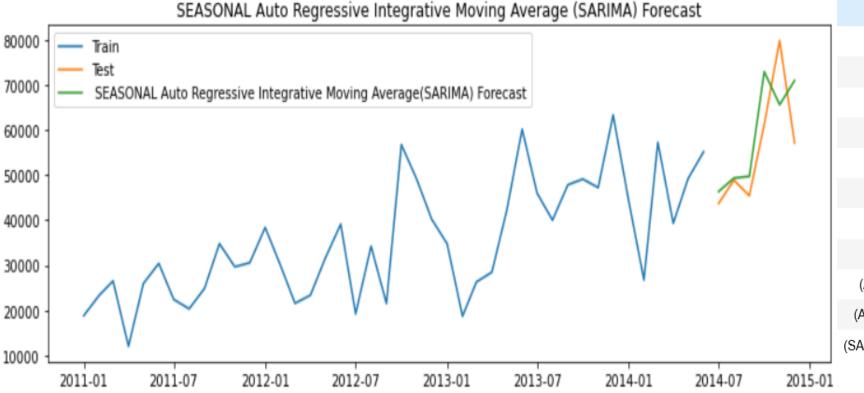
- ➤ Here, the green line represents the ARIMA Forecast.
- ARIMA captures the Level and trend, but no Seasonality.



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Auto Regressive (AR)method	13.56
Moving Average (MA) method	33.93
(ARMA) Auto Regressive Moving Average method	32.40
ARIMA) Auto Regressive Integrative Moving Ave	32.40

#### SARIMA(Seasonal Auto Regressive Integrated Moving Average) method

- ➤ Here, the Green line represent the SARIMA forecast.
- > In the model, all 3 factors: Level, Trend and Seasonality are captured and observed.
- > Forecast values observed to be very close to Actual values of next 6 months.



MAPE
17.47
34.34
16.10
15.83
14.67
9.39
10.20
13.56
33.93
32.40
32.40
12.95

Method MADE

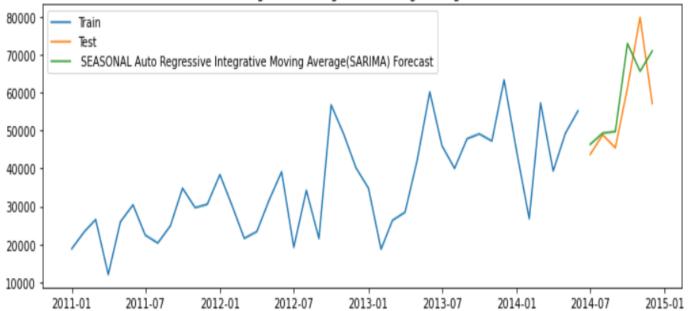
### ARIMA Technique inference:

 Based on ARIMA techniques, we can observe the lowest value of MAPE is for SARIMA method- thus it's the best suited of prediction of sale of nest 6 months.

#### Overall Inference:

- ➤ Overall, MAPE value is observed to have least value for Holts Winter and SARIMA model amongst all the models.
- ➤ Visually, SARIMA model is giving us better result and it have pretty decent MAPE value on lower side as well. So, we can prefer it.





#### Method MAPE

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Auto Regressive (AR)method	13.56
Moving Average (MA) method	33.93
(ARMA) Auto Regressive Moving Average method	32.40
(ARIMA) Auto Regressive Integrative Moving Ave	32.40
(SARIMA)Seasonal Auto Regressive Integrated Mo	12.95

## Thank you.