

# Retail-Giant Case Study

## SUBMISSION

Group Name: Group PISA

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# Abstract

## **Problem Statement:**

- Global Mart is an online store super giant having worldwide operations. It takes orders and delivers across the globe and deals with all the major product categories - consumer, corporate & home.
- The goal is to finalize the plan for the next 6 months. So, you want to forecast the sales and the demand for the next 6 months, that would help you manage the revenue and inventory accordingly.

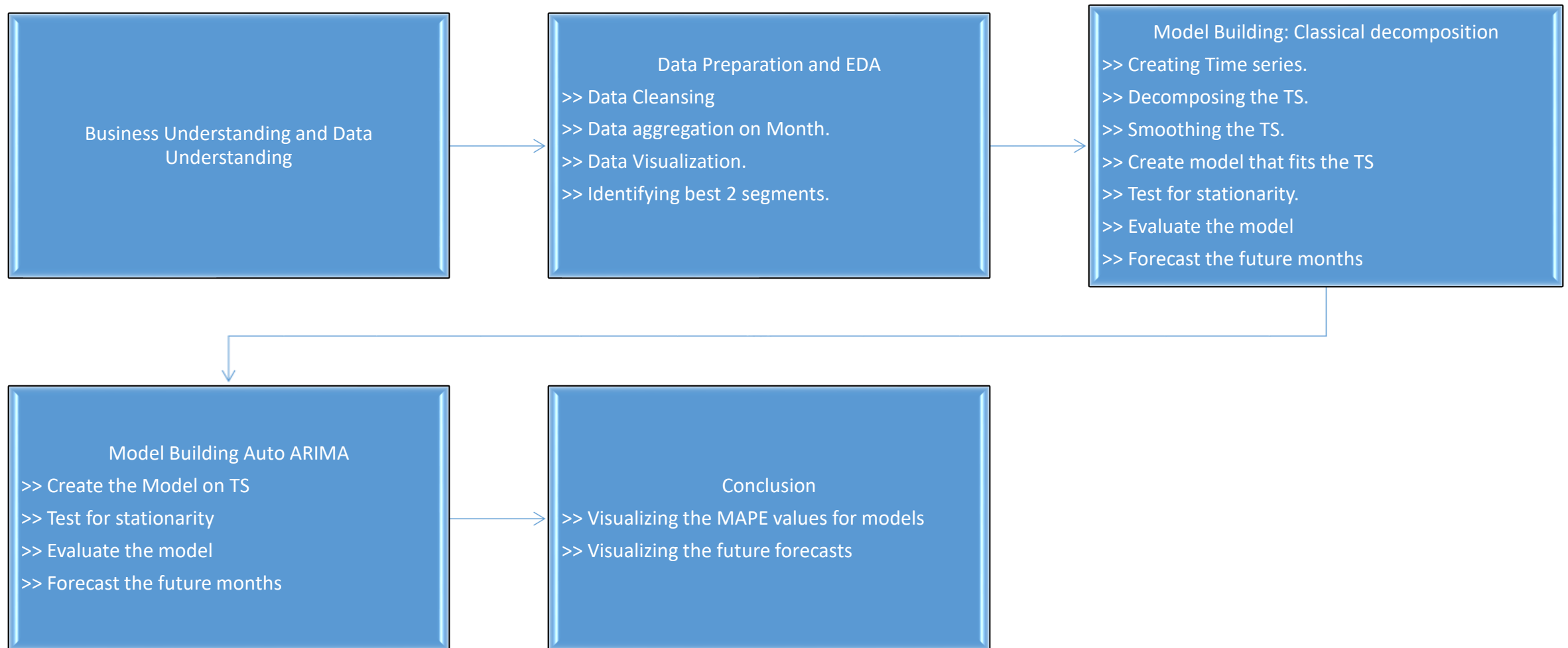
## **Objectives of the Analysis:**

- Identify the 2 most profitable and consistent market segment for the company.
- Forecast the sales and demand for the next 6 months.

## **Solution Approach:**

- R is used for Data preparation, from raw data sources.

# Solution Approach



# Data Understanding and Preparation

## Data Understanding

- Raw data contains 51290 rows of transactional order data.
- Broadly the data is divided in 3 segments
- Further it can be divided to 7 Markets (Geographies).

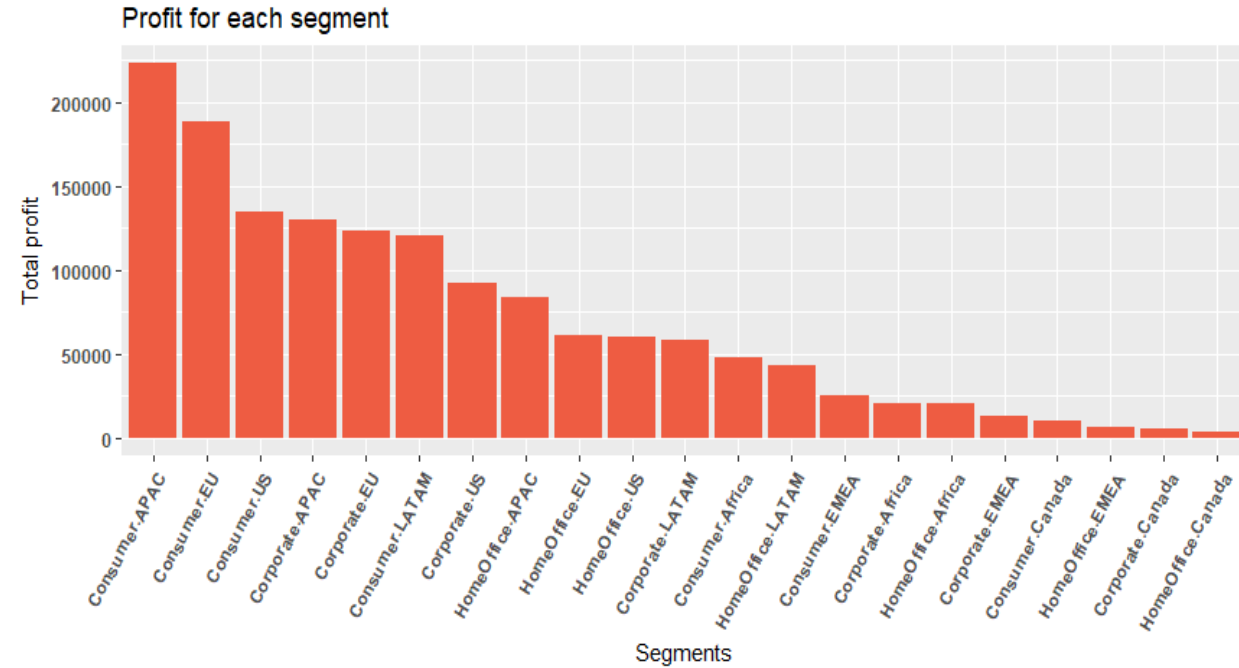
## Data Preparation

- Convert the Order Date column to Date format and extract month.
- Create subsets of data for all 21 market segments.
- Aggregate all data sets on month and calculate Coefficient of variance.
- Identify 2 most consistently profitable segments.
- Create time series for sales and quantity. (Keeping last 6 months as test data).

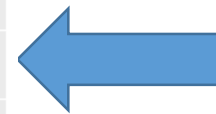
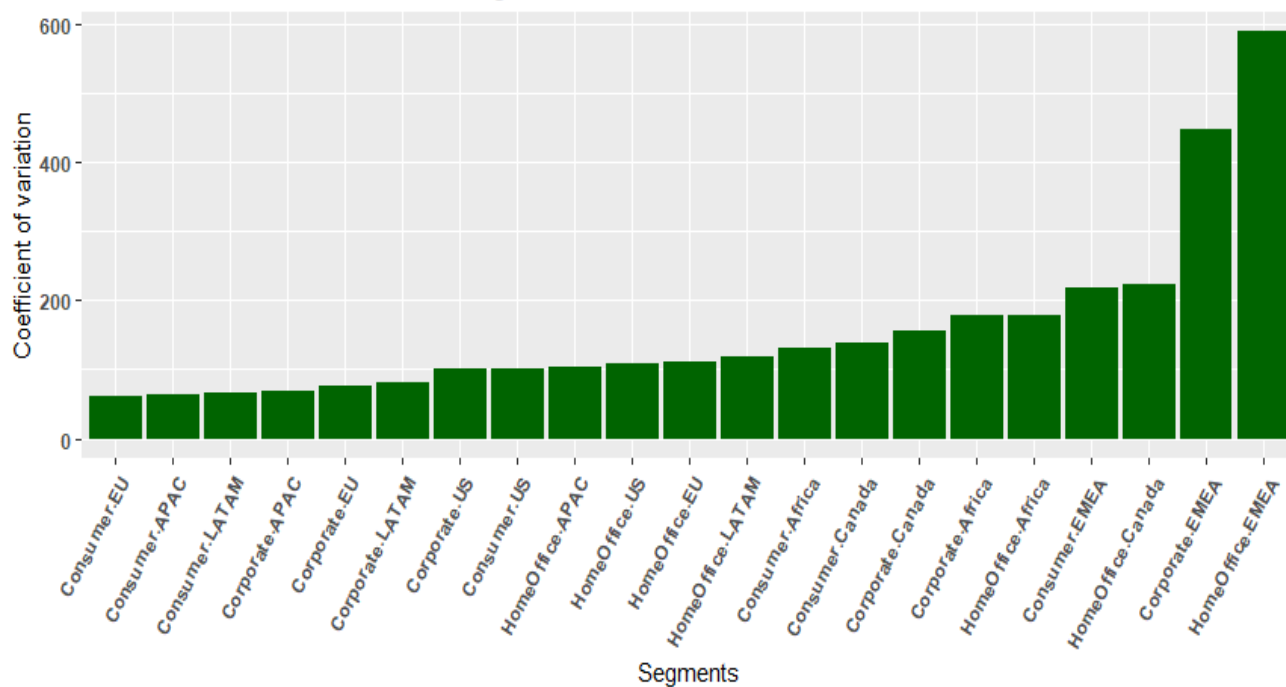
# Identifying best segments

The 2 most profitable segments are

1. Segment Consumer and Market APAC
2. Segment Consumer and Market EU



Coefficient of variation for each segment



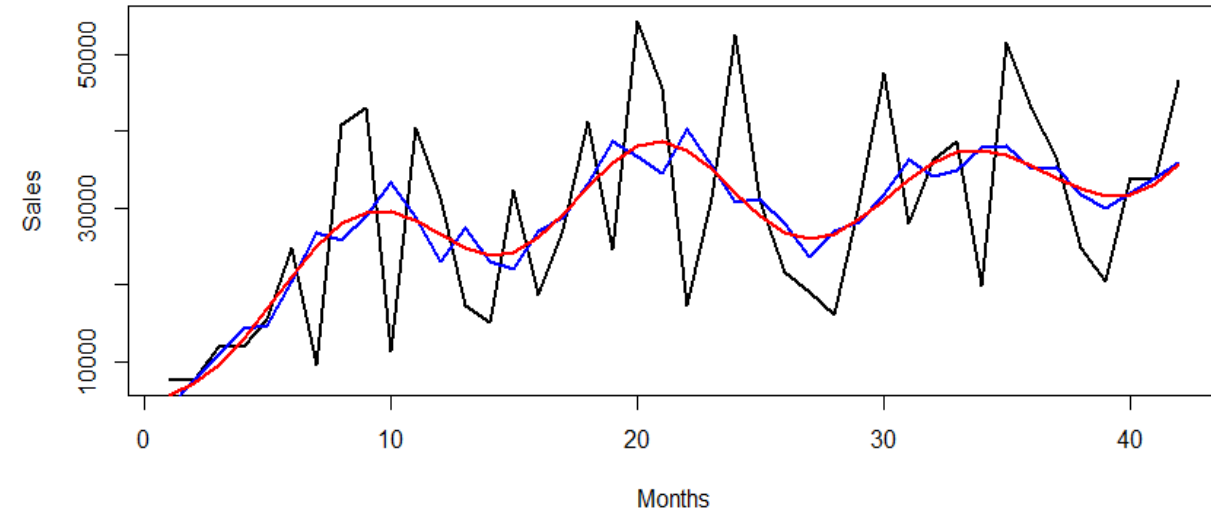
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1. Segment Consumer and Market APAC
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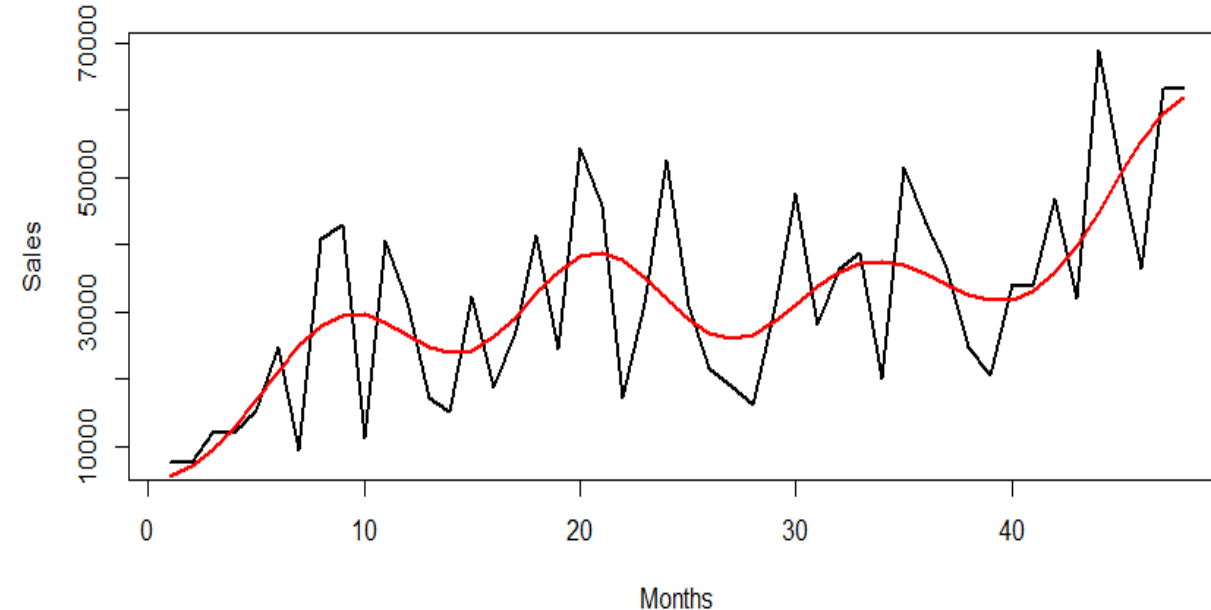
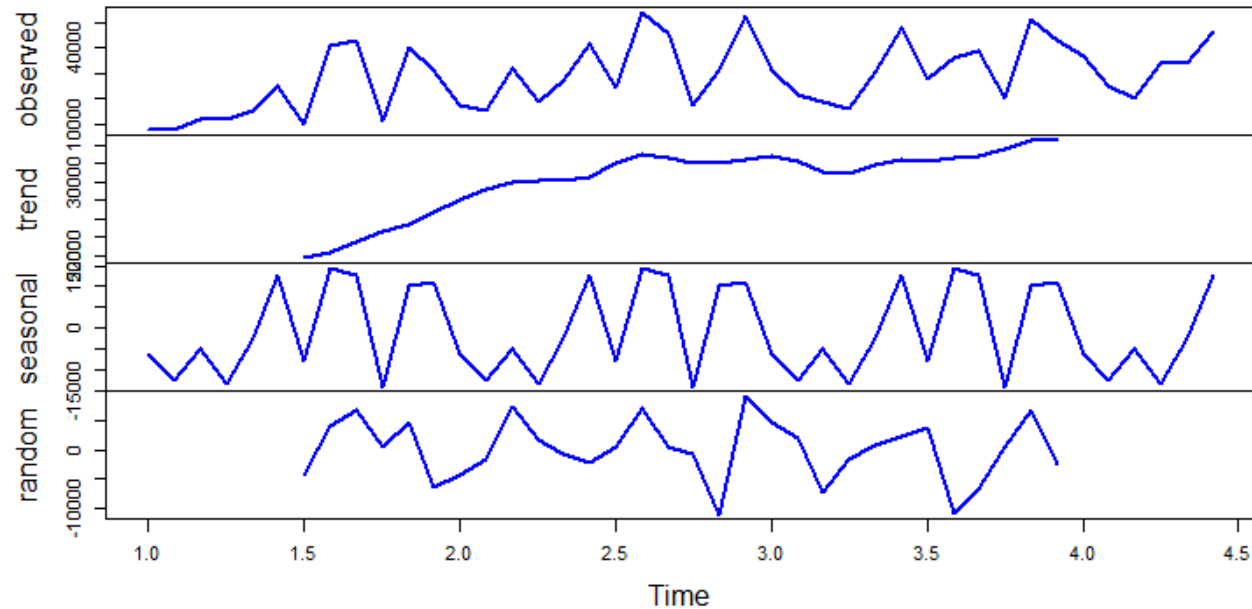
# Classical Decomposition

## Consumer EU Sales Forecast

- The TS shows and additive model.
- The TS shows a linear upward trend
- The seasonality shows a sinusoidal seasons yearly.



**Decomposition of additive time series**

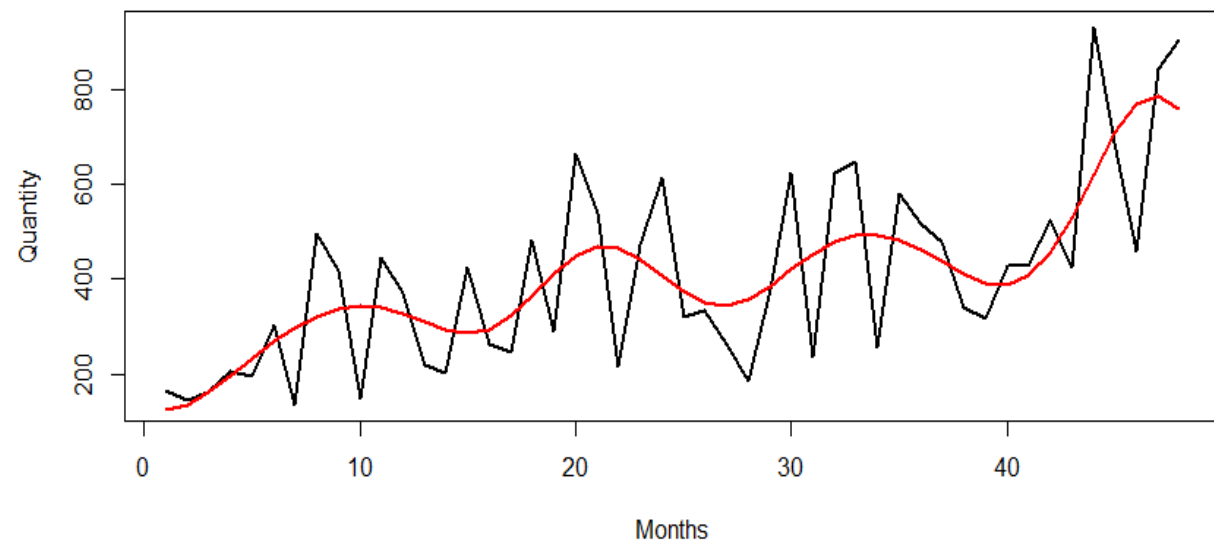
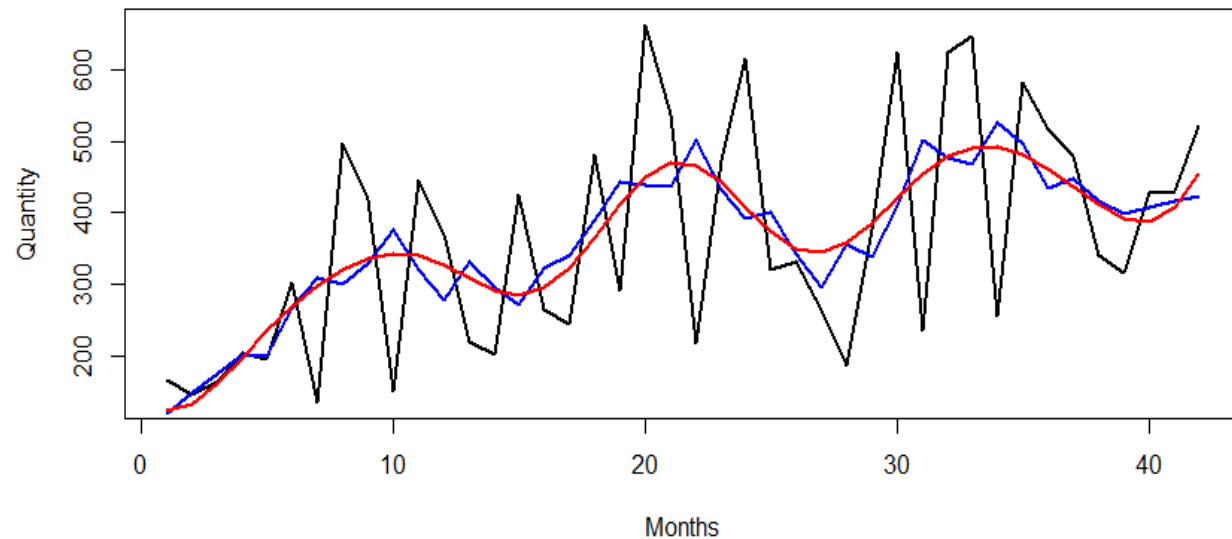
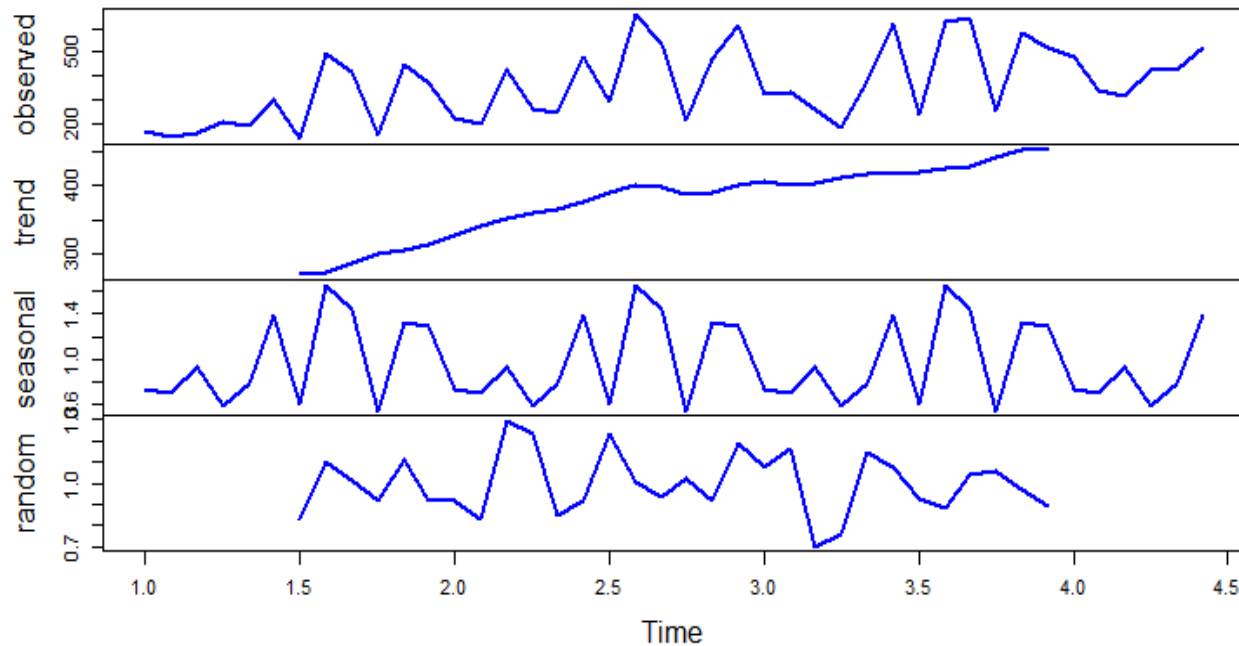


# Classical Decomposition

## Consumer EU Quantity Forecast

- The TS shows a linear upward trend
- The seasonality shows a sinusoidal seasons yearly.

**Decomposition of multiplicative time series**

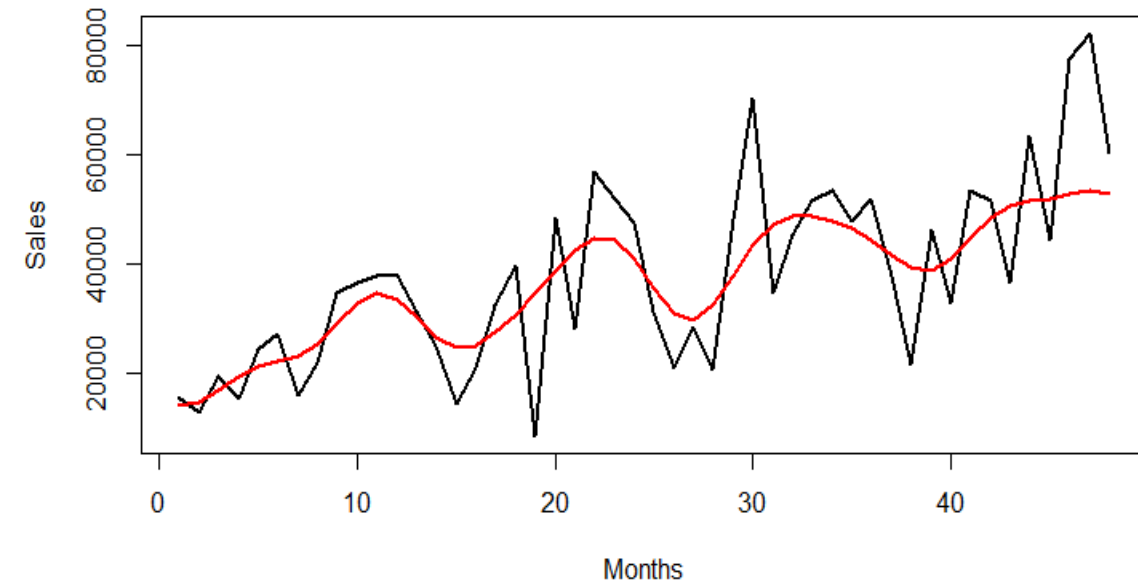
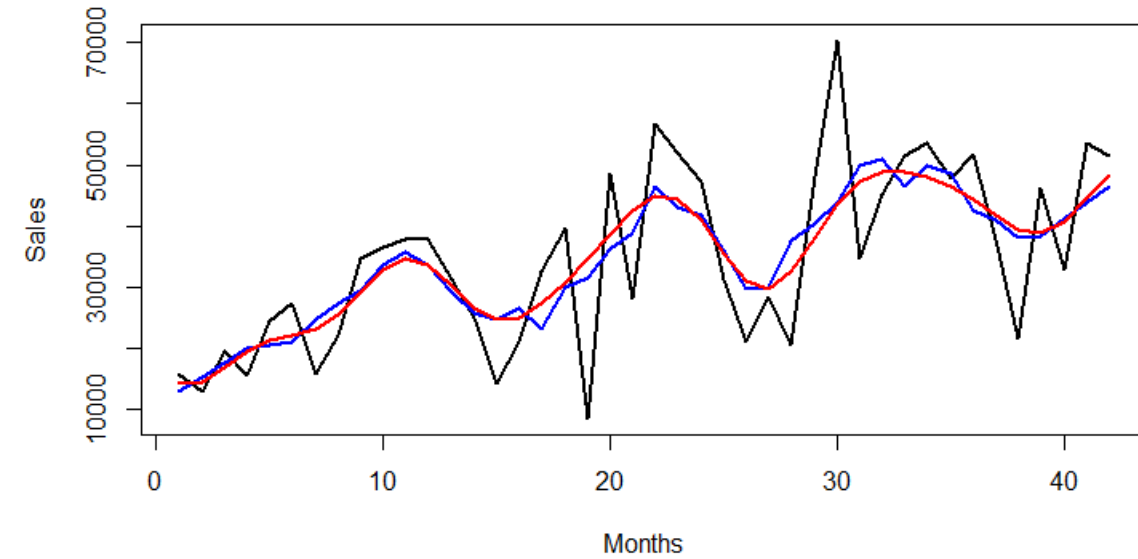
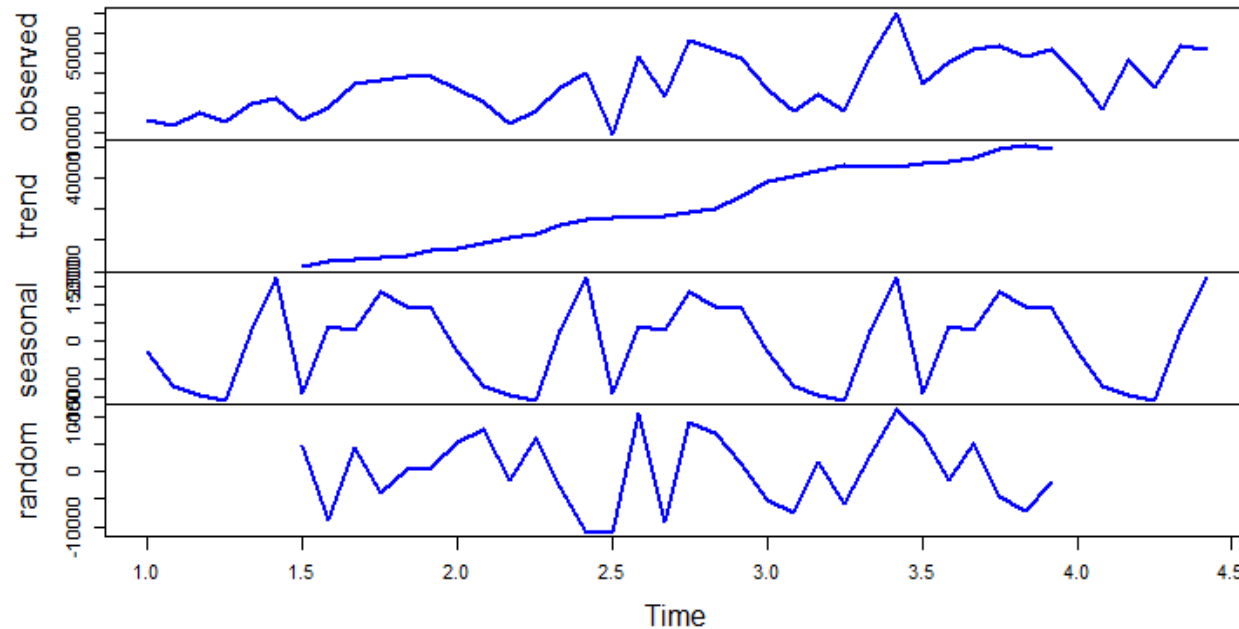


# Classical Decomposition

## Consumer APAC Sales Forecast

- The TS shows a linear upward trend
- The seasonality shows a sinusoidal seasons yearly.

**Decomposition of additive time series**



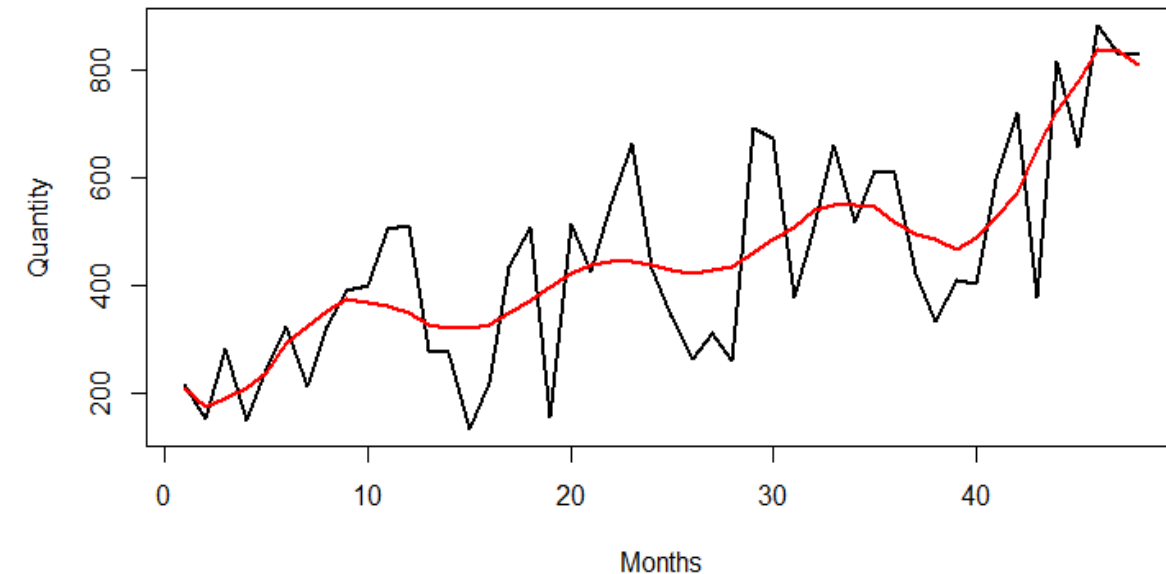
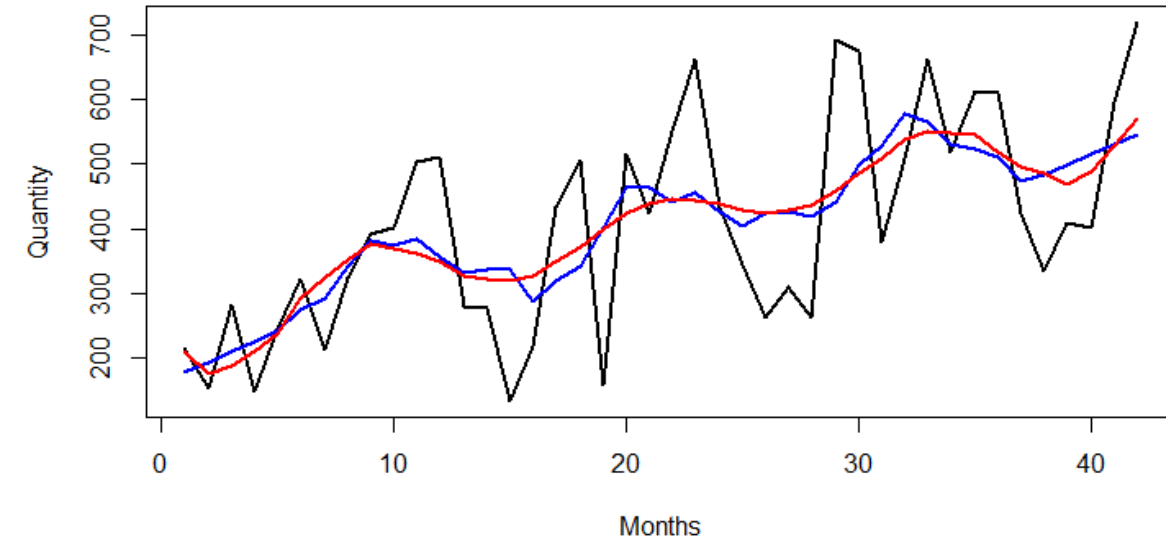
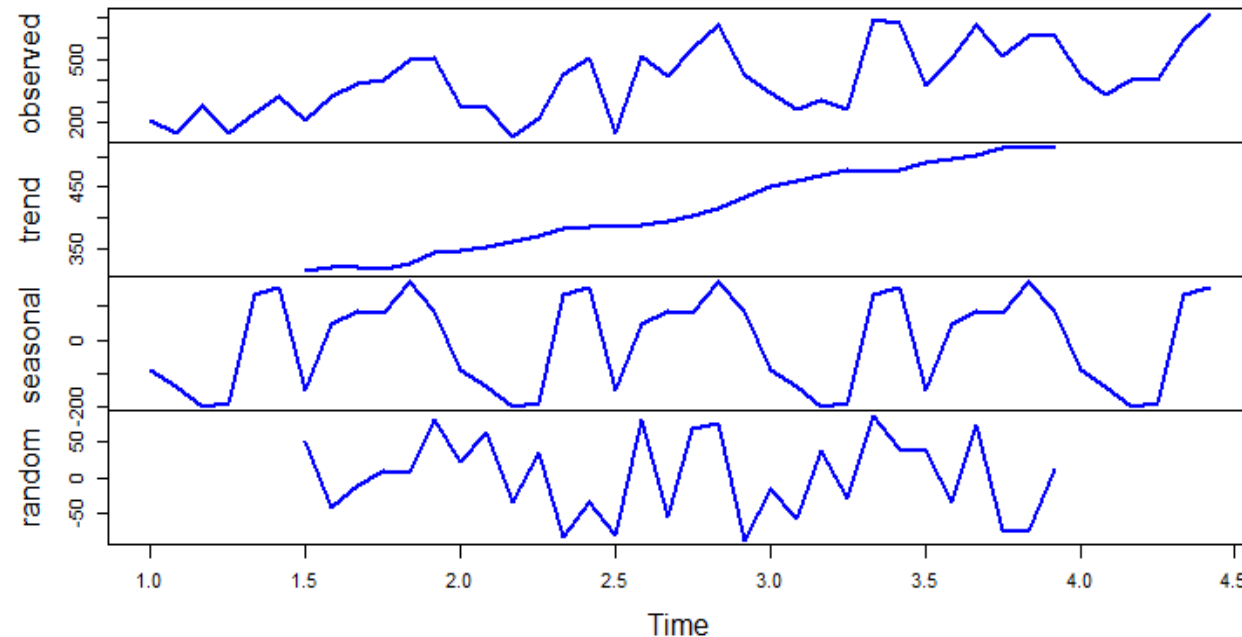


# Classical Decomposition

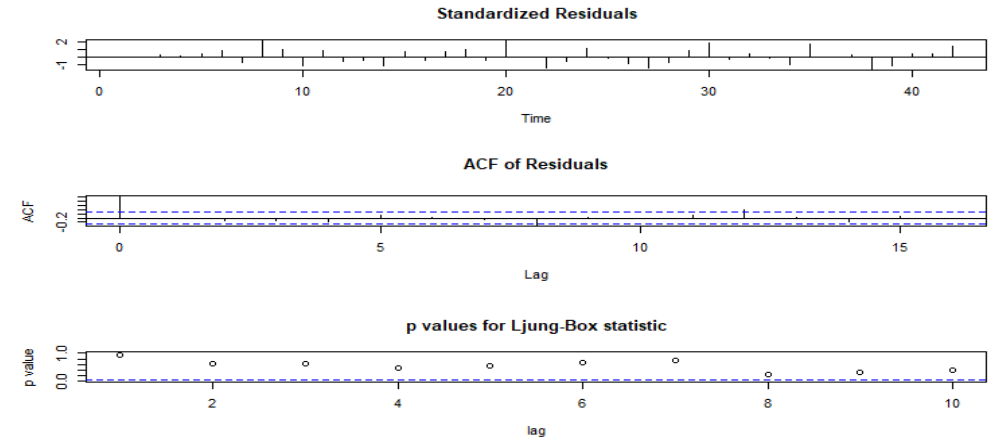
## Consumer APAC Quantity Forecast

- The TS shows a linear upward trend
- The seasonality is yearly.

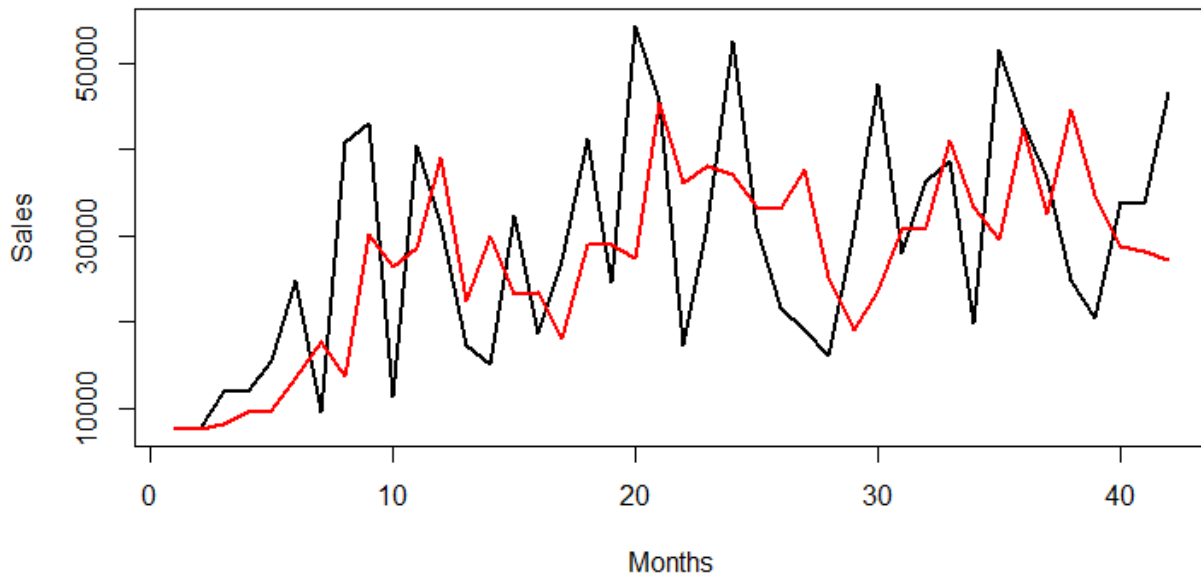
**Decomposition of additive time series**



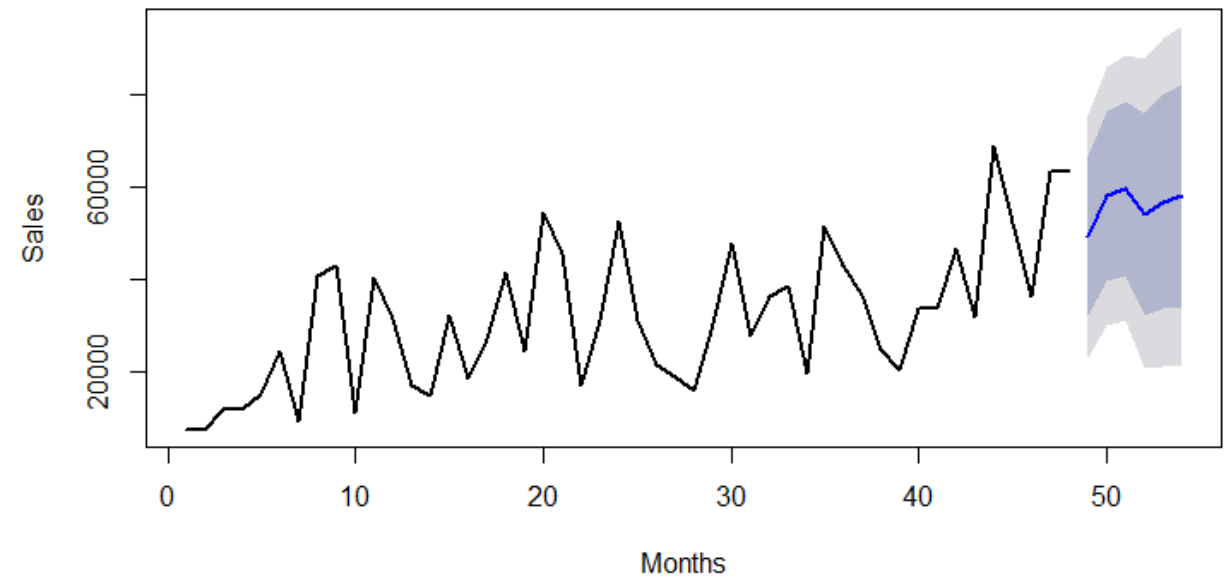
- Auto ARIMA predicts a ARIMA (2,1,0) model.
- MAPE = 28.9226



**Actual V/S Predicted**



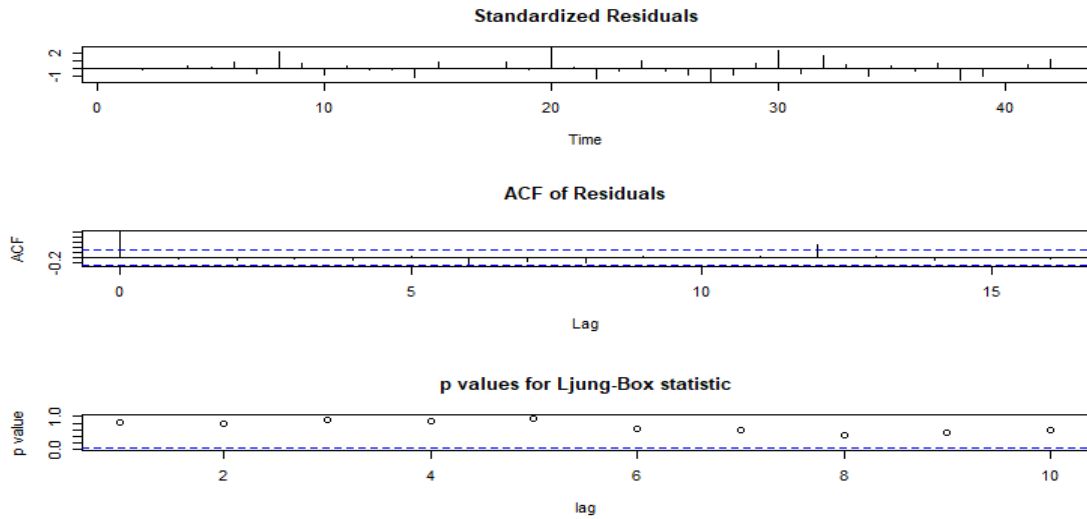
**Forecasts from ARIMA(2,1,0)**



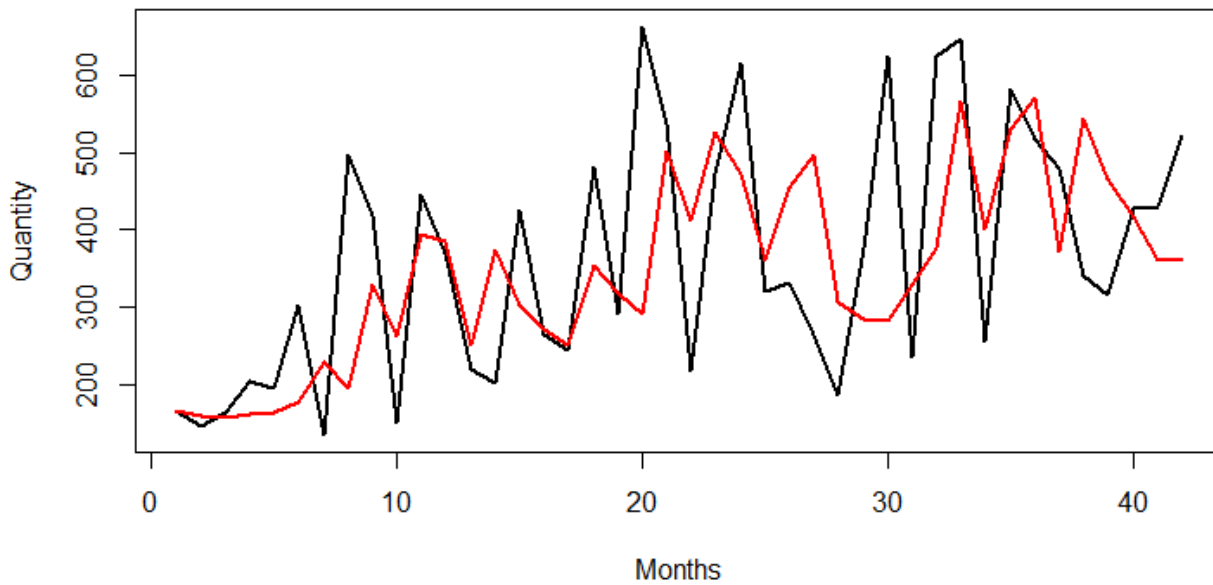
# Auto ARIMA

## Consumer EU Quantity Forecast

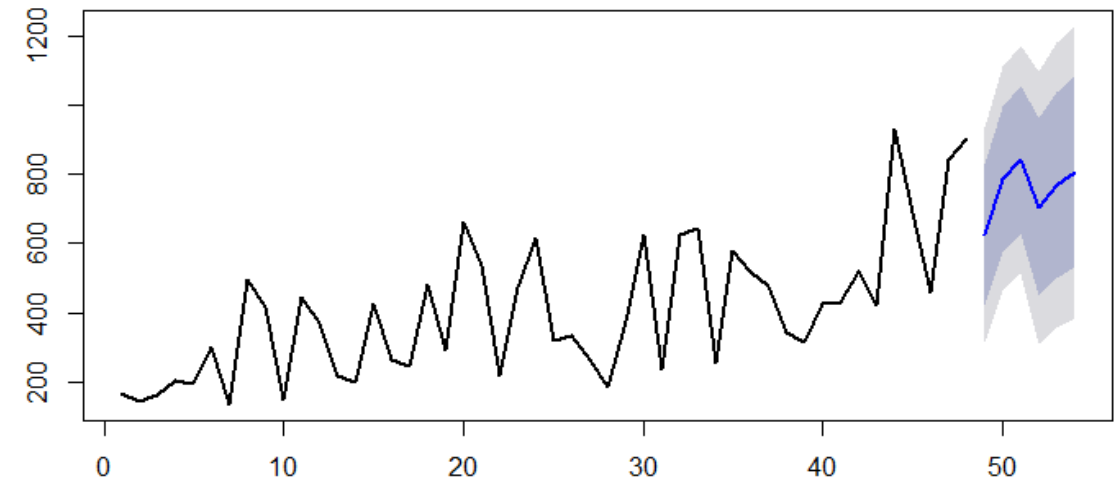
- Auto ARIMA predicts a ARIMA (2,1,0) model.
- MAPE = 30.13319



Actual V/S Predicted

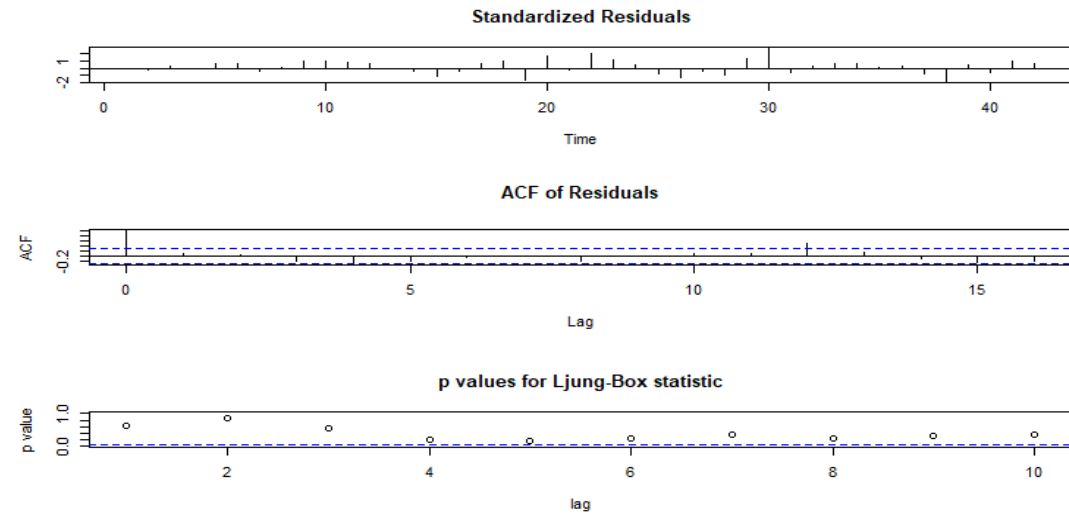


Forecasts from ARIMA(2,1,0)

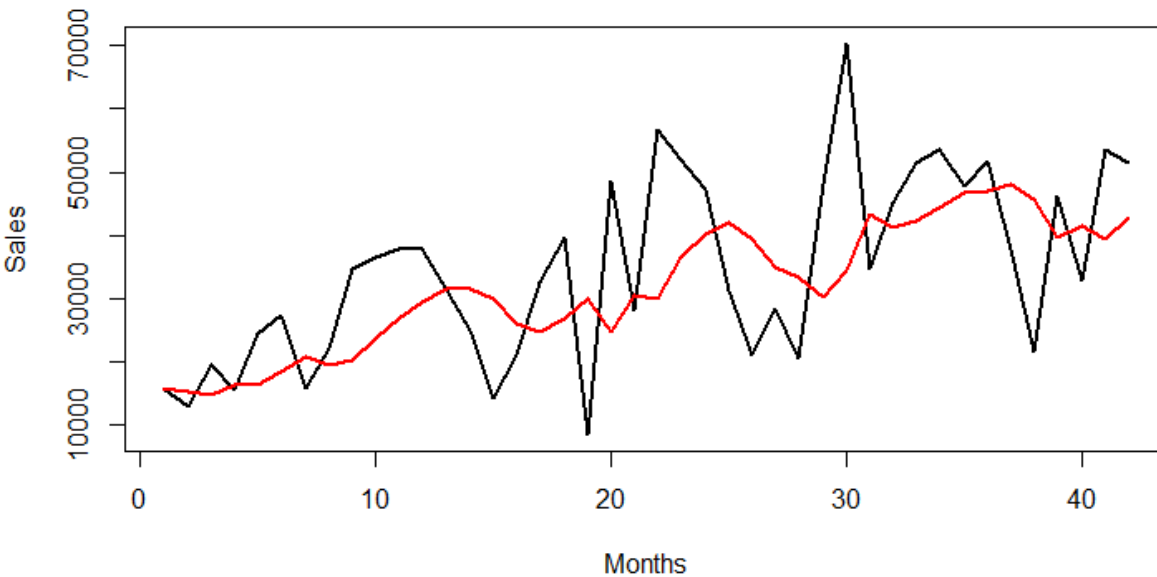


# Auto ARIMA Consumer APAC Sales Forecast

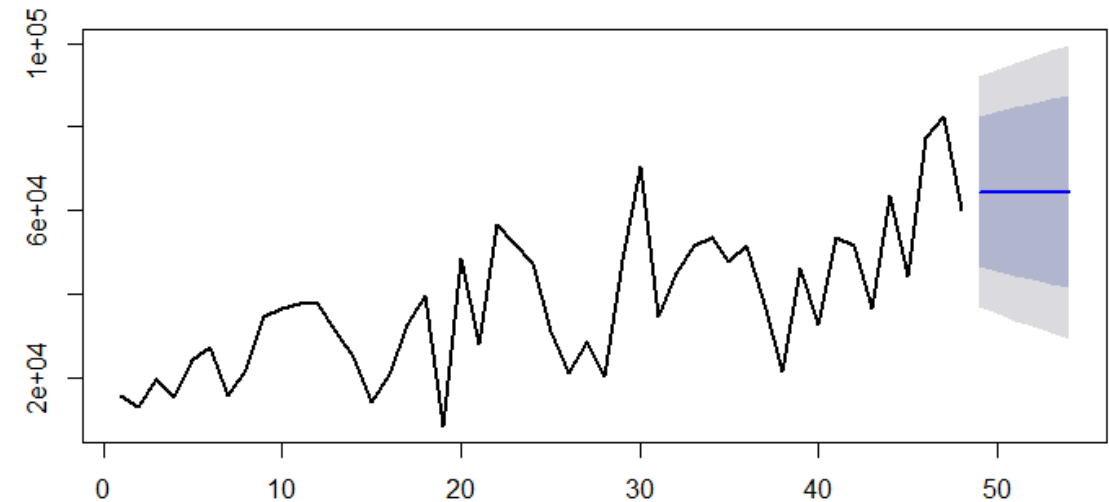
- Auto ARIMA predicts a ARIMA (0,1,1) model.
- MAPE = 27.68952



**Actual V/S Predicted**



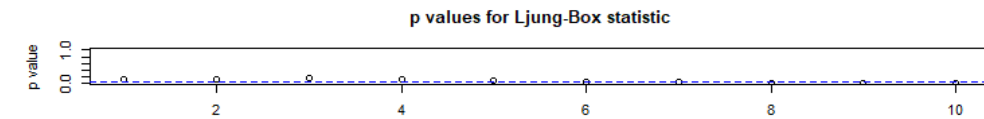
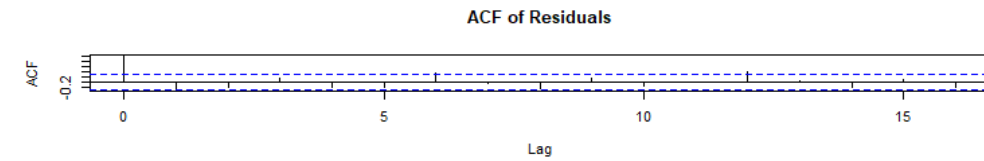
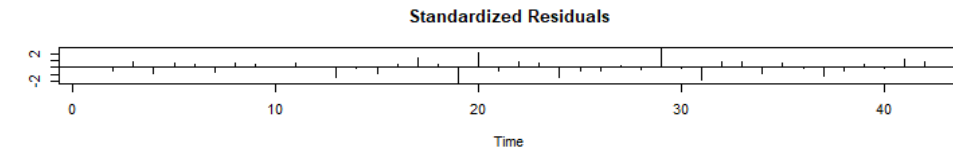
**Forecasts from ARIMA(0,1,1)**



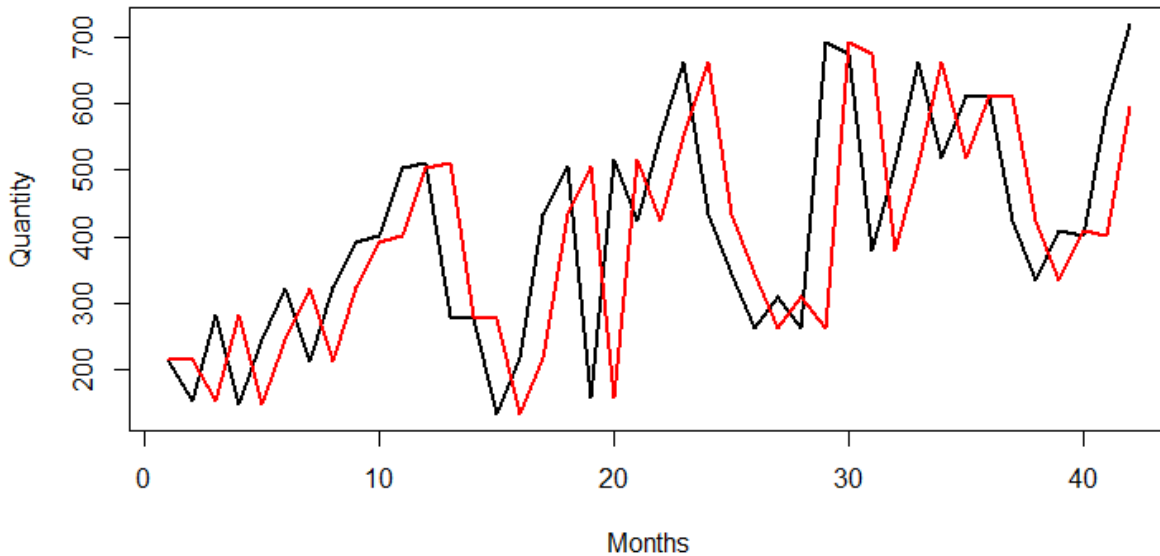
# Auto ARIMA

## Consumer APAC Quantity Forecast

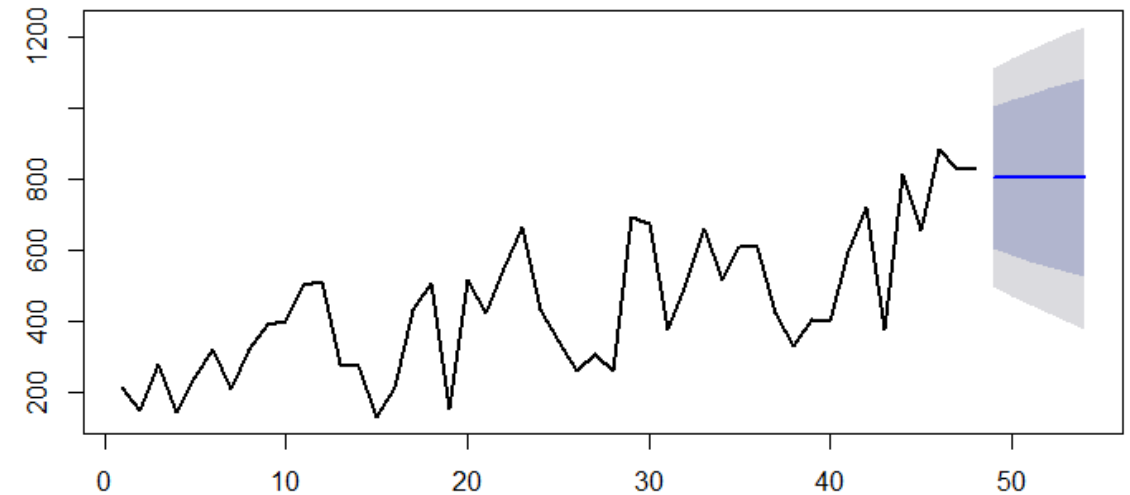
- Auto ARIMA predicts a ARIMA (0,1,0) model.
- MAPE = 26.24458



Actual V/S Predicted



Forecasts from ARIMA(0,1,1)



# Future Forecast from Jan-2015 to June-2015

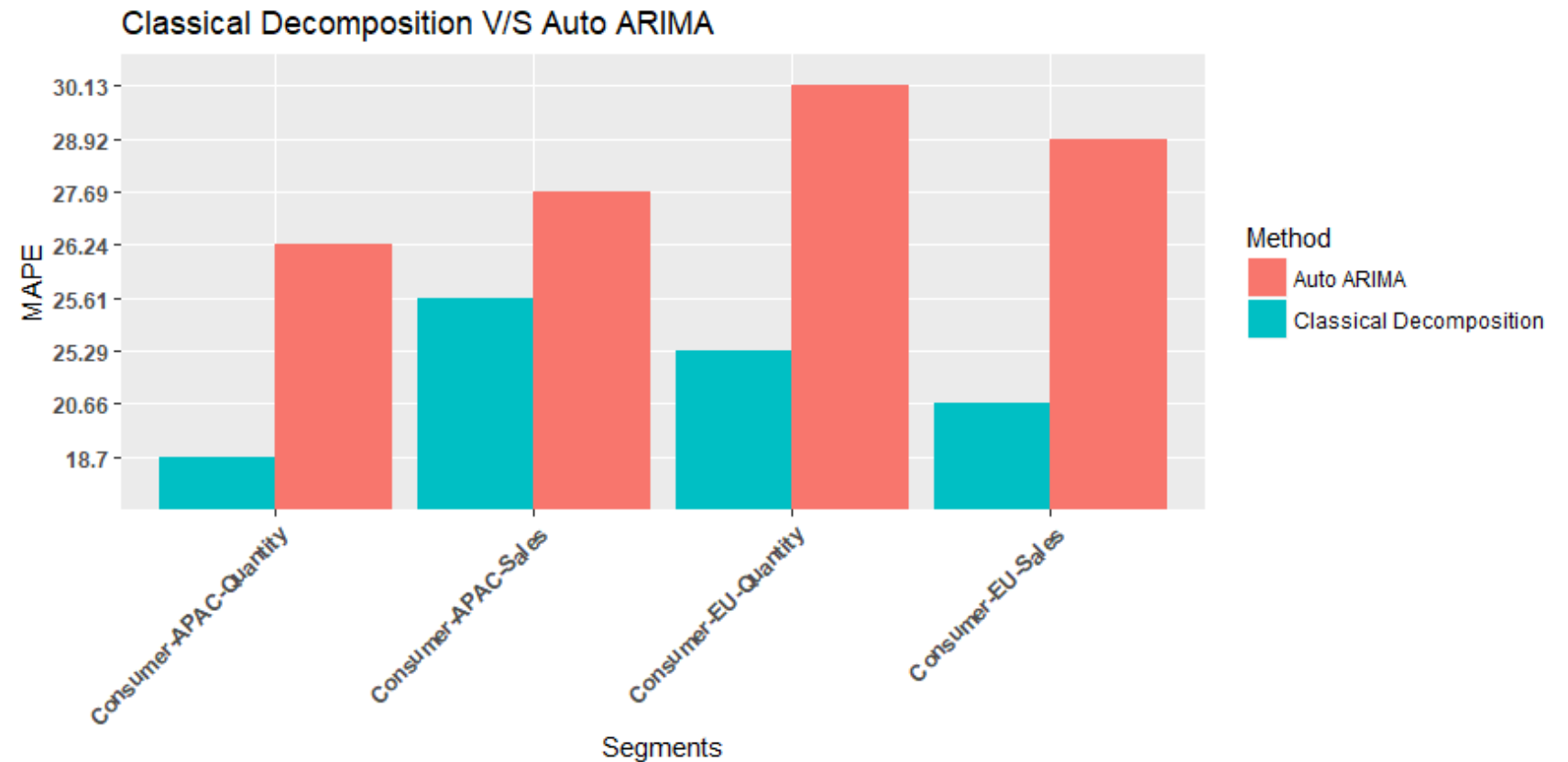
Month	Method	Consumer-EU-Sales	Consumer-EU-Quantity	Consumer-APAC-Sales	Consumer-APAC-Quantity
Jan-15	Classical Decomposition	62252.97	697.77	49717.10	766.45
Feb-15	Classical Decomposition	61033.13	635.00	44433.38	659.01
Mar-15	Classical Decomposition	59042.48	610.43	39283.80	575.44
Apr-15	Classical Decomposition	57575.83	659.55	37150.33	493.35
May-15	Classical Decomposition	58050.25	799.23	39718.73	414.26
June-15	Classical Decomposition	61598.35	1018.22	46435.21	424.94

Month	Method	Consumer-EU-Sales	Consumer-EU-Quantity	Consumer-APAC-Sales	Consumer-APAC-Quantity
<b>Jan-15</b>	Auto ARIMA	49358.71	626.20	64494.89	804.41
<b>Feb-15</b>	Auto ARIMA	58063.62	786.61	64494.89	804.41
<b>Mar-15</b>	Auto ARIMA	59714.33	842.92	64494.89	804.41
<b>Apr-15</b>	Auto ARIMA	54191.79	704.83	64494.89	804.41
<b>May-15</b>	Auto ARIMA	56811.55	768.63	64494.89	804.41
<b>Jun-15</b>	Auto ARIMA	58010.84	807.65	64494.89	804.41

# Conclusion – 1

## Evaluating Classical Decomposition Vs Auto ARIMA

In all cases classical decomposition shows a better accuracy as compared to Auto ARIMA process.



# Conclusion – 2

## Predictions Classical Decomposition Vs Auto ARIMA

Below is the comparison of predictions made by classical decomposition and Auto ARIMA.

