



Time Series Project

Business Case , Decomposed Linear Regression

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Business Case

- Different models perform differently for different KPIs
- We need to figure out which model is best suited to which KPI
- Models in question :
 - Decomposed Linear Regression
 - ARIMA
 - Auto-ARIMA
 - XGBoost regression



Decomposed Linear Regression Model

- Model Structure
- Model Steps
- Model Implementation



Decomposed Linear Regression Model

Any time series visualization may consist of the following components:

- Trend + Seasonality + Error.

We may have different combinations of trends and seasonality. Depending on the nature of the trends and seasonality, a time series can be modeled as an additive or multiplicative time series.



Decomposed Linear Regression Model

Each observation in the series can be expressed as either a sum or a product of the components.

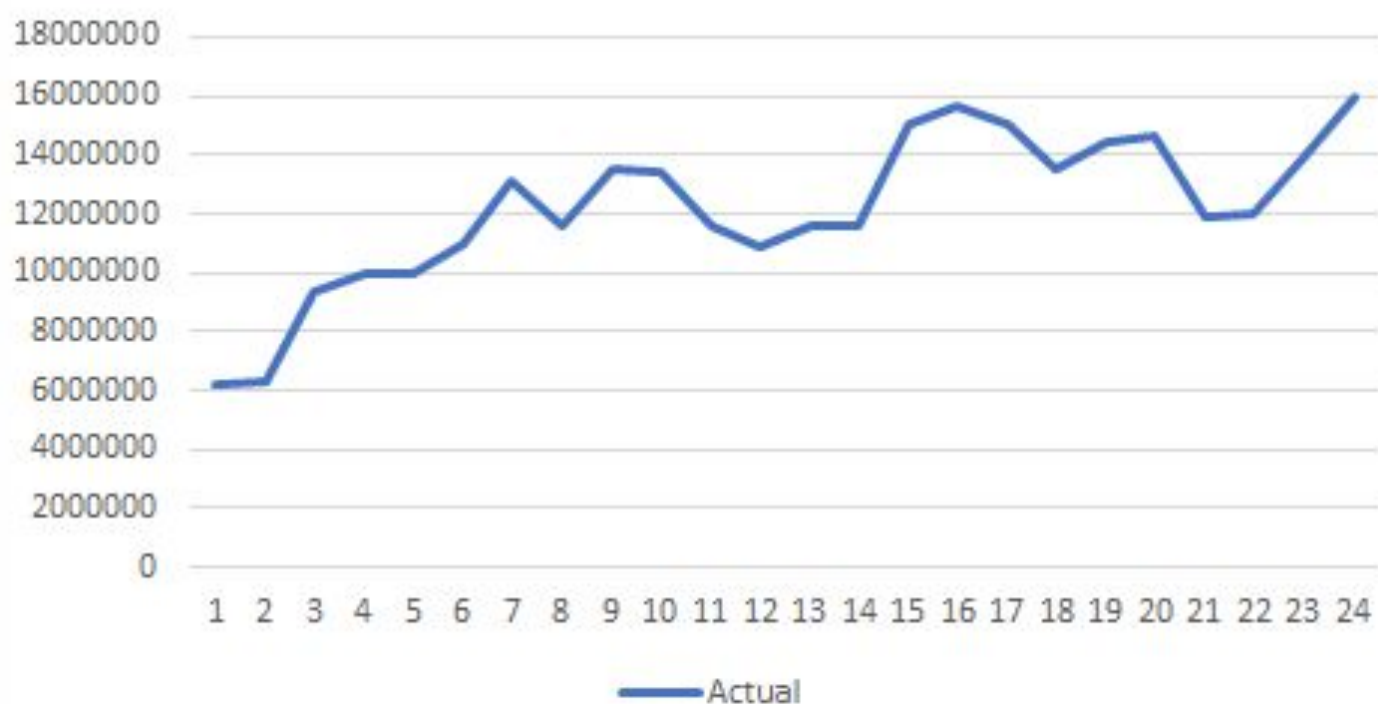
Additive time series:

- $\text{Value} = \text{Trend} + \text{Seasonality} + \text{Error}$

Multiplicative Time Series:

- $\text{Value} = \text{Trend} \times \text{Seasonality} \times \text{Error}$

Actual





Decomposed Linear Regression Model

Model : Trend * Seasonal * Irregular

Step 1 : Get the probable trend component (By Smoothen Y Values)

Step 2 : Get detrended data (Actual / (Seasonal x Irregular))

Step 3 : Get seasonal component (Seasonal by averaging and removing Irregular)

Step 4 : Get deseasonalized component (Actual / Seasonal)



Decomposed Linear Regression Model

Model : $T_t * S_t * I_t$

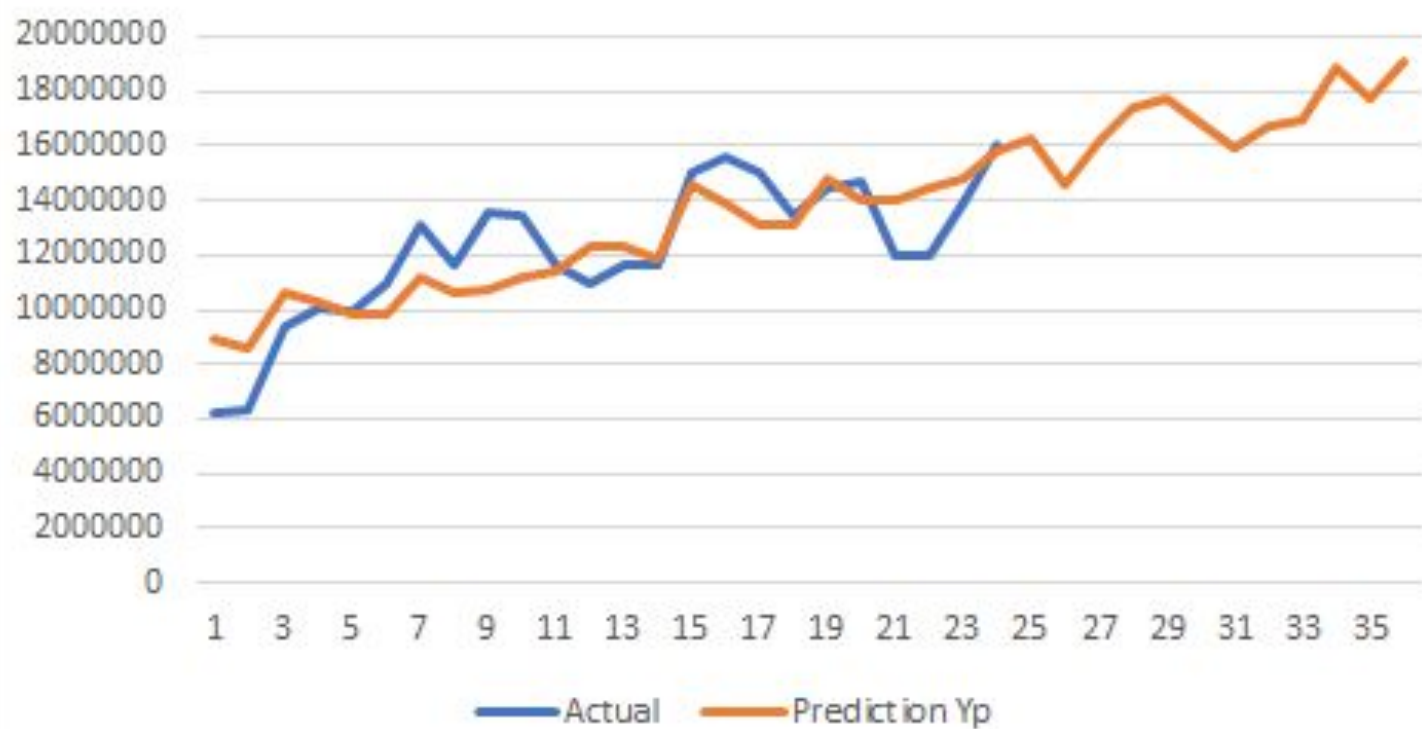
Step 5 : Find regression values for deseasonalized values ($X = T \text{ Code}, Y = DS$)

Step 6 : Find the T_t component ($T_t = \text{Intercept} + \text{Coeff} * T \text{ Code}$)

Step 7 : Calculate Y_p with T_t and S_t ($Y_p = S_t * T_t$)

Step 8 : Use further $T \text{ Code}$, T_t along with S_t for Future Prediction

Actual vs Predicted





Python Codebase :

- **Inputs :**
 - Month Level Data + Marketing Spend Data
- **Model :**
 - Decomposed Multiplicative (LR + MLR)
- **Output :**
 - Prediction at Month Level for Next 12 Months