

1) Objective:

Forecasting number of anti-diabetic drug prescriptions

2) checking for through visualizations

'Trend' ✓ & 'Seasonality' ✓

3) checking for stationarity

ADF Statistic \rightarrow should be less
& p-value < 0.05

ours is not stationary

4) So we apply transformations to make it stationary

tried {
i) 1st diff \rightarrow again not stationary
ii) Seasonal diff $\rightarrow m=12$
Now it is stationary ✓

& $d=1, D=1, m=12$

$$4 \quad d=1, D=1, m=12$$

5) Model selection 4 parameters till now

$$\text{SARIMA} \rightarrow (p, 1, q) \overbrace{(P, 1, Q)}^{\text{Seasonal}} m=12$$

(No. exogenous variables)

6) Optimizing the model

to get the best possible combination of parameters through 'AIC' criterion

$$\left. \begin{array}{l} \text{Minimum} \\ \text{AIC} \end{array} \right\} \rightarrow p=2, q=3, P=1, Q=3$$

\therefore Final Model

$$\text{SARIMA } (2, 1, 3) (1, 1, 3) 12$$

7) Residuals Analysis

✓ i) Variance: Constant Variance

(No Heteroskedasticity problem)

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✓ ii) Distribution: Normal (Histogram + Density plot)

✓ iii) Normal Q-Q plot: Straight line

✓ iv) Correlogram:

No significant co-efficients after lag 0
just like white-noise

So, with this analysis, residuals seem to closely resemble white noise

8) Ljung-Box test:

to check whether the
residuals

→ Independent ??
(or)
Correlated

9) Forecasting & Comparison



Actual values	Seasonal forecasts (Baseline model)	SARIMA model
MAPE:	12.69	7.9

Since the SARIMA model achieves the lowest MAPE, we can conclude that the SARIMA(2,1,3)(1,1,3)₁₂ model should be used to forecast the monthly number of antidiabetic drug prescriptions.