Time-Series Analysis Performance Assessment

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D213

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This analysis utilizes Python 3.9.9 and Jupyter 7.0.2.

# Part I. Research Question

A1. Summarize one research question that is relevant to a real-world organizational situation captured in the selected data set and that you will answer using time series modeling techniques.

* ***What is the revenue forecast for the next month?***

A2. Define the objectives or goals of the data analysis. Ensure your objectives or goals are reasonable within the scope of the scenario and are represented in the available data.

Analyze two years’ worth of daily revenue data of the organization and create a predictive model that will forecast the next 30 days of future revenue.

# Part II. Method Justification

B. Summarize the assumptions of a time series model including stationarity and autocorrelated data.

Two assumptions of time series analysis include stationarity and autocorrelation. Stationarity means that "the mean, variance, and autocorrelation structure are constant over time" (Statisticssolutions.com, n.d.). In other words, "the statistical properties of a time series do not change over time" (Statisticssolutions.com, n.d.). The other assumption is no autocorrelation. "Autocorrelation occurs when future values in a time series linearly depend on past values" (Pierre, 2021).

# Part III. Data Preparation

C1. Provide a line graph visualizing the realization of the time series.

A graph showing the growth of the company's revenue

Description automatically generated

C2. Describe the time step formatting of the realization, including any gaps in measurement and the length of the sequence.

There are no gaps in measurement and the sequence is at 700 days.

C3. Evaluate the stationarity of the time series.

The dataset is stationary and as such no further action is necessary to make it so.A close-up of a white envelope

Description automatically generated

C4. Explain the steps you used to prepare the data for analysis, including the training and test set split.

To prepare the data I dropped 0 values of Revenue. I also ensured there are no missing values. I then converted the sequential number of days into proper date formatting and set it as the index. For splitting the train and test sets, I took the last 30 days of the dataset and set it as the test set and assigned the rest as the training set.

C5. Provide a copy of the cleaned data set.

Filenames:

* df.to\_csv('../data/teleco\_cleaned1.csv', index=False)
* train.to\_csv('../data/teleco\_cleaned1\_train.csv', index=False)
* test.to\_csv('../data/teleco\_cleaned1\_test.csv', index=False)

# Part IV. Model Identification and Analysis

D1. Report the annotated findings with visualizations of your data analysis, including the following elements:

* the presence or lack of a seasonal component

A graph showing the time of the year

Description automatically generated with medium confidence

* trendsA graph with a line

  Description automatically generated
* A graph with blue lines and dots

  Description automatically generatedthe autocorrelation function
* the spectral density

A graph with blue lines

Description automatically generated

* A graph of growth and revenue

  Description automatically generated with medium confidencethe decomposed time series
* A graph showing a number of events

  Description automatically generated with medium confidenceconfirmation of the lack of trends in the residuals of the decomposed series

D2. Identify an autoregressive integrated moving average (ARIMA) model that accounts for the observed trend and seasonality of the time series data.

A screenshot of a data report

Description automatically generated

D3. Perform a forecast using the derived ARIMA model identified in part D2.

A graph showing a line

Description automatically generated

D4. Provide the output and calculations of the analysis you performed.

Filename: " D213 Performance Assessment Task 1 (Rev. 3).ipynb"

D5. Provide the code used to support the implementation of the time series model.

Filename: " D213 Performance Assessment Task 1 (Rev. 3).ipynb"

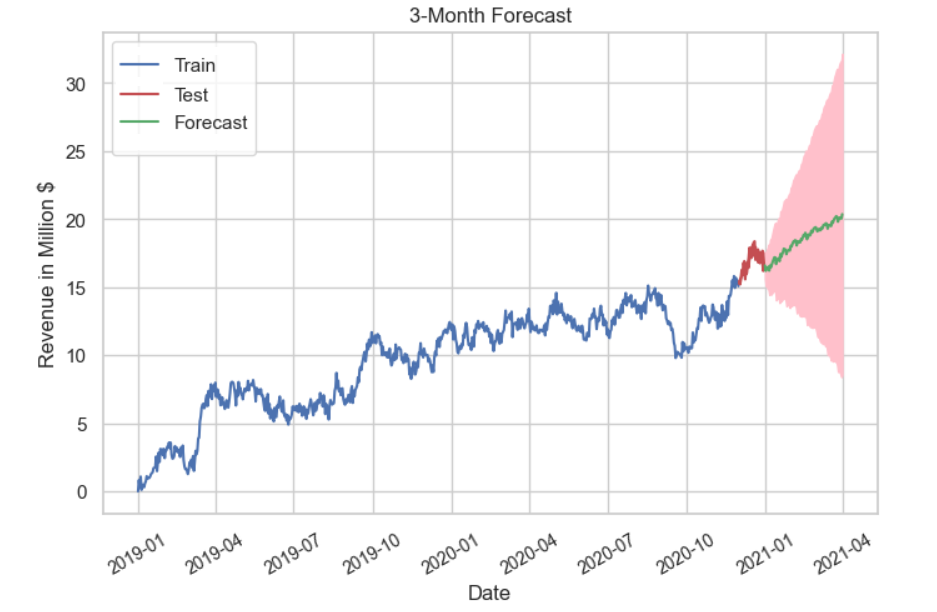
# Part V. Data Summary and Implications

E1. Discuss the results of your data analysis, including the following points:

* the selection of an ARIMA model
* the prediction interval of the forecast
* a justification of the forecast length
* the model evaluation procedure and error metric

The final ARIMA model was based on the results of Auto ARIMA (Best model: ARIMA(1,1,0)(5,1,0)[12]) which considers the trend and seasonality of the data set. The prediction interval of the forecast is 30 days and can be made using the .predict() or .forecast() methods. Forecast length of 30 is enough information to make changes in preparation for the next month. The final model was evaluated with R2. It "measures the strength of the relationship between your model and the dependent variable" (Frost, 2018). Although 11.77 is a low result, a low R2 doesn't necessarily mean the model is bad (Frost, 2018).

E2. Provide an annotated visualization of the forecast of the final model compared to the test set.



E3. Recommend a course of action based on your results.

The forecast data estimates that revenue will be at $17.69 million. Visual inspection of the plot also reveals an upward trend for the forecasted months. As such, I recommend a conservative approach to configuring organization operations for this quarter.

# Part VI. Reporting

F. With the information from part E, create your report using an industry-relevant interactive development environment (e.g., an R Markdown document, a Jupyter Notebook). Include a PDF or HTML document of your executed notebook presentation.

Filename: "D213 Performance Assessment Task 1 (Rev. 3).pdf"

G. Cite the web sources you used to acquire third-party code to support the application.

* https://github.com/ecdedios/code-snippets/blob/main/notebooks/master.ipynb
* https://www.datacamp.com/tutorial/matplotlib-time-series-line-plot
* https://towardsdatascience.com/finding-seasonal-trends-in-time-series-data-with-python-ce10c37aa861
* https://towardsdatascience.com/time-series-decomposition-in-python-8acac385a5b2
* https://analyticsindiamag.com/what-are-autocorrelation-and-partial-autocorrelation-in-time-series-data/
* https://github.com/mkosaka1/AirPassengers\_TimeSeries/blob/master/Time\_Series.ipynb

H. Acknowledge sources, using in-text citations and references, for content that is quoted, paraphrased, or summarized.

Anonymous. (2023, August 30). The Stationary Data Assumption in Time Series Analysis. Statistic Solutions. https://www.statisticssolutions.com/stationary-data-assumption-in-time-series-analysis

Pierre, Sadrach. (2022, October 12). A Guide to Time Series Analysis in Python. Builtin. https://builtin.com/data-science/time-series-python

Frost, Jim. (2018, February 24). How To Interpret R-squared in Regression Analysis. Statistics By Jim. https://statisticsbyjim.com/regression/interpret-r-squared-regression