TIME SERIES PROJECT

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INTRODUCTION

In today's data-driven world, analyzing and forecasting incident data is crucial for efficient resource allocation and proactive decision-making in various sectors such as public safety, healthcare, and urban planning.

This project aims to leverage advanced time series analysis techniques to predict future incident occurrences based on historical data. By transforming raw incident records into a structured time series format, applying differencing to ensure stationarity, and utilizing auto_arima for model selection, we seek to build a robust forecasting model. The resulting forecasts will help organizations anticipate trends and manage resources more effectively, ultimately enhancing their ability to respond to incidents in a timely manner.

PROBLEM STATEMENT

San Francisco is a combination of natural beauty, cultural richness and urban amenities.

However, it is a nest of crimes and this project aims to leverage machine learning techniques to predict the rate of crime in the city over time

OBJECTIVE

• To determine the rate of crime rate in San Francisco.

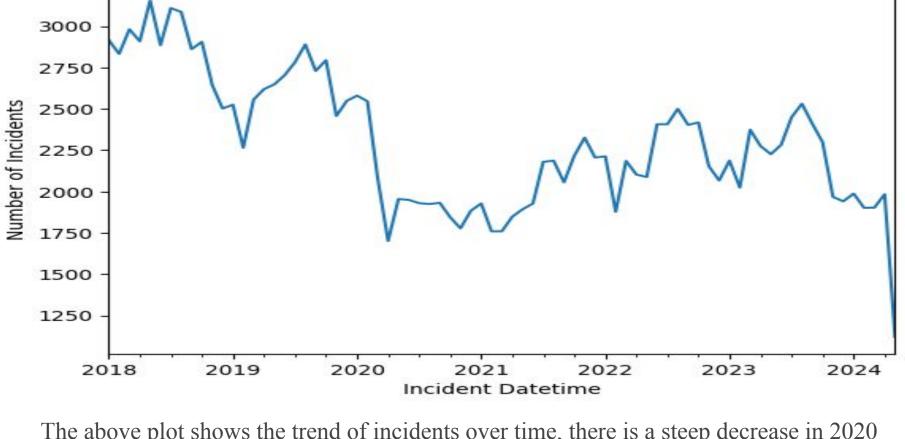
• To show how the crime rate varies over time.

METHODOLOGY

Data Wrangling: In this section, I imported the necessary libraries, loaded the data and cleaned the data by dropping unnecessary columns, rows and handling missing values.

Exploratory Data Analysis: In this section, I used various plots and graphs to have a better understanding of the data and draw insights.

Model Selection and Evaluation: In this section, I chose two time series models; Arima model and the Auto_Arima models. I first evaluated a baseline to see the threshold where my results should fall within. Differencing gave different results for both models.



Yearly Incident Trends

3250

The above plot shows the trend of incidents over time, there is a steep decrease in 2020 owing to the fact that it was during covid 19 era and there were curfews.

CONCLUSIONS

• Seasonality and Trends: The data exhibited clear seasonal patterns and trends, highlighting the importance of incorporating these elements in forecasting models. Weekly seasonality in particular was a significant factor influencing crime rates.

• Actionable Insights: The forecasting results offer actionable insights for city planners, emergency services and policy makers by anticipating periods of high incident rates, proactive measures can be taken to mitigate risks and enhance public safety.

RECOMMENDATIONS

- **Regular Data Updates**: Ensure that the incident data is updated regularly to keep the forecasts accurate and relevant. This will help in timely decision-making and resource allocation.
- Continuous Monitoring and Model Improvement: Establish a system for continuous monitoring and evaluation of the forecasting models. Regularly assess model performance and update parameters or switch to more advanced models as necessary.