Exp no: 2 Implement programs for visualizing time series data

Date: 31/1/25

Objectives:

The goal of this preprocessing pipeline is to prepare time series sales data for accurate predictive modeling. This involves handling missing values, removing anomalies, ensuring correct time order, and potentially transforming features. The aim is to create a clean and consistent dataset that improves the performance of time series forecasting models, enabling better predictions of future sales trends and the identification of seasonality patterns.

Background/Scope:

Time series sales data is a sequence of observations indexed by time. Preprocessing is crucial because missing values, outliers, incorrect time order, and other data quality issues can negatively impact forecasting models. This process addresses these challenges by filling gaps, removing outliers, ordering data chronologically, and potentially creating new features. Proper preprocessing is essential for time series models to accurately capture trends, predict future sales, and support informed business decisions.

Steps for Time Series Sales Data Preprocessing:

Step 1: Load the Dataset and libraries

1. **Load** the dataset from a CSV file into a Pandas DataFrame.

Step 1: Load the Dataset

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from statsmodels.graphics.tsaplots import plot_acf
from statsmodels.tsa.stattools import adfuller
df=pd.read_csv("/content/NFLX.csv")
```

Step 2: Visualizing techniques

Autocorrelation Plot: Create an autocorrelation plot to identify potential seasonality using the acf_plot

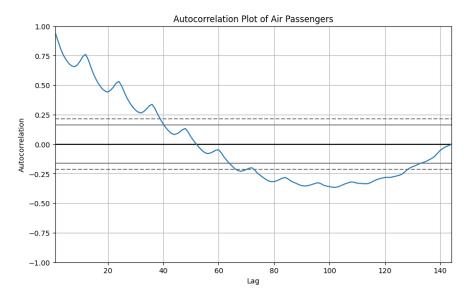
```
if 'Date' not in df.columns:
    print("'Date' is already the index or not present in the DataFrame.")
else:
    df.set index('Date', inplace=True)
```

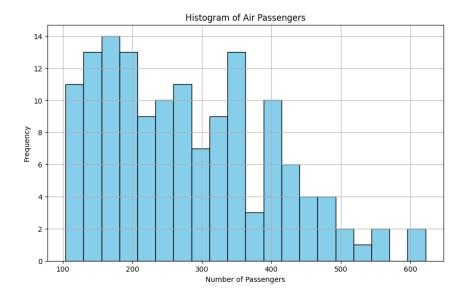
```
# Plot the ACF
plt.figure(figsize=(12, 6))
plot_acf(df['Volume'], lags=40) # You can adjust the number of lags as
needed
plt.xlabel('Lag')
plt.ylabel('Autocorrelation')
plt.title('Autocorrelation Function (ACF) Plot')
plt.show()
```

Histogram Plot: Create a histogram plot to visualize autocorrelation.

```
plt.figure(figsize=(10, 6))
plt.hist(df['#Passengers'], bins=20, color='skyblue', edgecolor='black')
plt.title('Histogram of Air Passengers')
plt.xlabel('Number of Passengers')
plt.ylabel('Frequency')
plt.grid(True)
plt.show()
```

output:





Result:

Thus the time series dataset is visualized successfully.