**EX:No.2 221501190**

**01/02/25**

**IMPLEMENTING DIFFERENT VISUALIZATION TECHNIQUE USING TIME SERIES DATA**

**AIM:**

To implementing different visualization technique using time series dataset.

**PROCESS:**

**#Importing libraries**

import pandas as pd

import numpy as np

import matplotlib.pyplot as plt

import seaborn as sns

from statsmodels.tsa.seasonal import seasonal\_decompose

**# Generate Synthetic Dataset**

np.random.seed(42)

n = 200

dates = pd.date\_range(start='2022-01-01', periods=n)

close\_prices = np.random.normal(loc=150, scale=10, size=n) # Normal distribution

outliers = np.random.choice(n, size=5, replace=False)

close\_prices[outliers] += np.random.normal(loc=50, scale=5, size=5) # Inject outliers

**# Create DataFrame**

data = pd.DataFrame({'Date': dates, 'Close': close\_prices})

**# Box Plot to Check Outliers**

plt.plot(timeseries, color='blue', label='Original')

plt.plot(movingAverage, color='red', label='Rolling Mean')

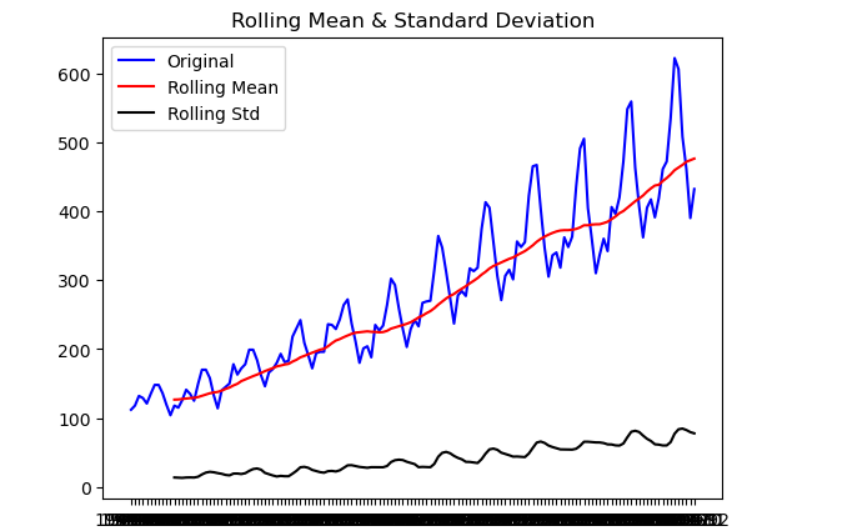
plt.plot(movingSTD, color='black', label='Rolling Std')

plt.legend(loc='best')

plt.title('Rolling Mean & Standard Deviation')

plt.show(block=False)

**OUTPUT:**



**# Scatter Plot to Check Distribution**

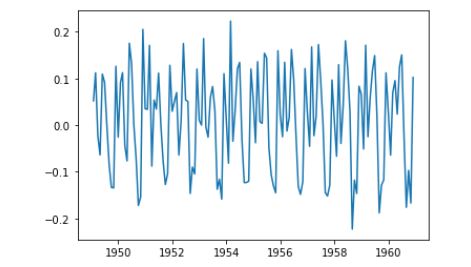
plt.plot(airpass\_log, color='blue', label='Original')

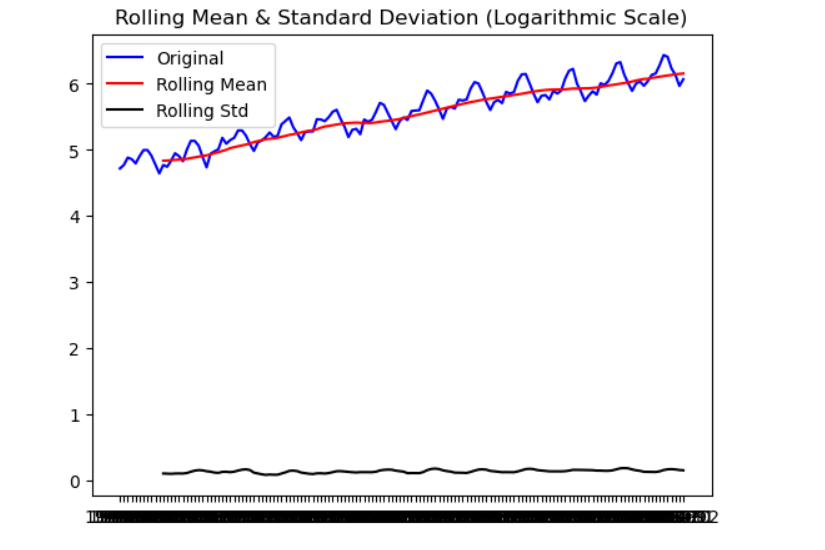
plt.plot(rollmean\_log, color='red', label='Rolling Mean')

plt.plot(rollstd\_log, color='black', label='Rolling Std')

plt.legend(loc='best')

plt.title('Rolling Mean & Standard Deviation (Logarithmic Scale)')



**OUTPUT:**

**# Reverse the order to maintain chronological order**

data = data.iloc[::-1].reset\_index(drop=True)

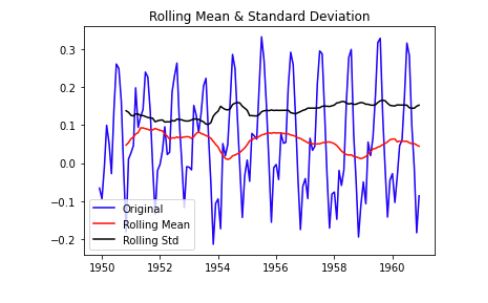
**# Handling Missing Values**

data.dropna(inplace=True) # Drop rows with missing values

data['Close'].fillna(data['Close'].mean(), inplace=True) # Fill NaNs in 'Close'

**# Extract Close Prices**

close\_prices = data['Close'].values



data['Normalized\_Close'] = close\_prices / np.max(close\_prices) # Normalize data

def plot\_time\_series(data, title='Time Series Data', xlabel='Time', ylabel='Value'):

plt.figure(figsize=(12, 6))

plt.plot(data, label='Close Prices', color='blue')

plt.xlabel(xlabel)

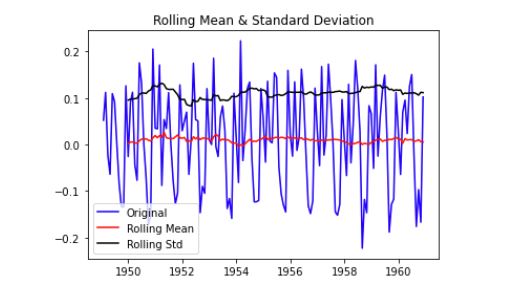
plt.ylabel(ylabel)

plt.title(title)

plt.legend()

plt.grid(True)

plt.show()



**# Simple Line Plot of Closing Prices**

plot\_time\_series(data['Close'], title='Synthetic Stock Close Prices')

**# Seasonal Decomposition**

result = seasonal\_decompose(data['Close'], model='additive', period=30)

plt.figure(figsize=(12, 8))

result.plot()

plt.show()

**RESULT:**

The implementing different visualization technique using time series dataset is successfully implemented.