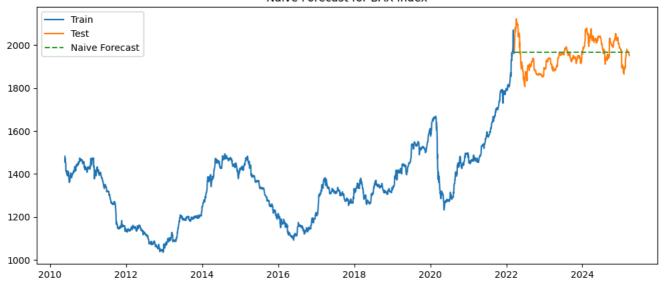
```
1 import pandas as pd
  2 import numpy as np
  3 import matplotlib.pyplot as plt
  5 from sklearn.metrics import mean squared error
  7 from statsmodels.graphics.tsaplots import plot_acf
  8 from statsmodels.tsa.seasonal import seasonal_decompose
 1 df bax m = pd.read csv(r'/content/drive/MyDrive/PRN23039142546/df bax cleaned to view outliers.csv', index col=0,parse dates=True)
 2 df_bax_m.head()
\rightarrow
                                                       Vol. Change %
                                                                         \blacksquare
                  Price
                            Open
                                    High
                                              Low
           Date
                                                                         ıl.
     2010-05-24 1482.42 1491.98 1491.98 1482.42 926980.0
                                                                 -0.64
     2010-05-25 1454.85 1482.42 1482.42 1454.85 1660000.0
                                                                 -1 86
     2010-05-26 1472.29 1456.50 1472.29 1454.85 1500000.0
                                                                  1.20
     2010-05-27 1453.82 1472.29 1478.07 1453.82 2480000.0
                                                                 -1.25
     2010-05-30 1455.16 1453.82 1462.04 1453.72 5910000.0
                                                                  0.09
Next steps: ( Generate code with df_bax_m ) ( View recommended plots
                                                                     New interactive sheet
 1 df_bax_m.info() #check data types
<class 'pandas.core.frame.DataFrame'>
    DatetimeIndex: 3659 entries, 2010-05-24 to 2025-03-27
    Data columns (total 6 columns):
                   Non-Null Count Dtype
     # Column
     0
         Price
                   3659 non-null float64
     1
         0pen
                   3659 non-null
                                  float64
     2
         High
                   3659 non-null
                                  float64
         Low
                   3659 non-null
                                   float64
         Vol.
                   3659 non-null
                                  float64
         Change % 3659 non-null
                                  float64
    dtypes: float64(6)
    memory usage: 200.1 KB
 1 df_naive = df_bax_m.copy()
 1 price = df naive['Price']
 1 split_point = int(len(price)*0.8)
 2 #price[split_point:]
 3 train,test = price[:split_point],price[split_point:]
  1 naive_forecast = pd.Series(train.iloc[-1], index=test.index)
 1 rmse = np.sqrt(mean_squared_error(test, naive_forecast))
  1 print(f"Naive Forecast - RMSE: {rmse:.2f}")
Naive Forecast - RMSE: 64.48
 1 plt.figure(figsize=(12,5))
  2 plt.plot(train, label='Train')
  3 plt.plot(test, label='Test')
  4 plt.plot(naive_forecast,label='Naive Forecast', linestyle='--')
  5 plt.legend()
  6 plt.title("Naive Forecast for BAX Index")
  7 plt.show()
```



Naive Forecast for BAX Index

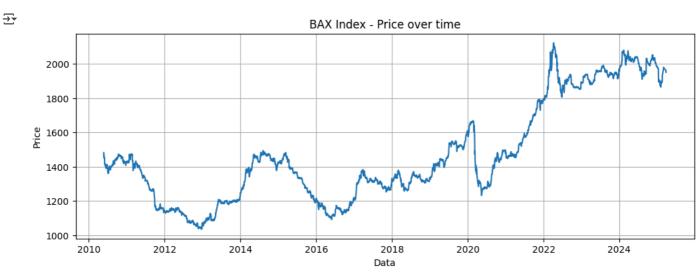


Check seasonality in BAX dataset

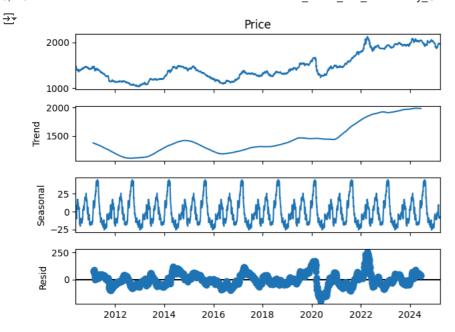
```
1 df = df_bax_m.copy()

1 series = df['Price']

1 # line plot
2
3 plt.figure(figsize=(12,4))
4 plt.plot(series)
5 plt.title("BAX Index - Price over time")
6 plt.xlabel("Data")
7 plt.ylabel("Price")
8 plt.grid(True)
9 plt.show()
```



```
1 # seasonal decomposition (using additive model for now)
2 decomposition = seasonal_decompose(series, model='additive', period=365) # Assuming daily data with yearly seasonality
3 decomposition.plot()
4 plt.tight_layout()
5 plt.show()
```



- 1 # autocorrelation plot
- 2 plot_acf(series, lags=400)
- 3 plt.xlim(340, 380) # Focus on 365 lag
- 4 #plt.title("Autocorrelation Plot (ACF)")
- 5 plt.title("Autocorrelation Plot (ACF) around 365 lag")
- 6 plt.show()

