```
In [1]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        from statsmodels.tsa.stattools import adfuller
        import statsmodels.api as sm
        from statsmodels.tsa.api import VAR
        from statsmodels.tsa.base import datetools
        from sklearn.metrics import mean_squared_error
        # importing all standard library for Vector Autoregression
        df = pd.read csv('SPDJ.csv')
        # read data of S&P500, Dow Jones Index from 11/25/2013 to 11/24/2023
        def adf test(series,title=''):
            Pass in a time series and an optional title, returns an ADF report
            print(f'Augmented Dickey-Fuller Test: {title}')
            result = adfuller(series.dropna(),autolag='AIC') # .dropna() handles differenced d
            labels = ['ADF test statistic','p-value','# lags used','# observations']
            out = pd.Series(result[0:4],index=labels)
            for key,val in result[4].items():
                out[f'critical value ({key})']=val
            print(out.to_string())
                                           # .to_string() removes the line "dtype: float64"
            if result[1] <= 0.05:
                print("Strong evidence against the null hypothesis")
                print("Reject the null hypothesis")
                print("Data has no unit root and is stationary")
            else:
                print("Weak evidence against the null hypothesis")
                print("Fail to reject the null hypothesis")
                print("Data has a unit root and is non-stationary")
                # Augmented Dickey Fuller test for Unit root to decide stationarity
                adf_test(df['SP'])
                adf_test(df['DJ'])
        df difference = df.diff()
        # differencing daily closing as most of financial data itself are not stationary
        adf_test(df_difference['SP'])
        adf test(df difference['DJ'])
        new_df_difference = df_difference.dropna(axis=0, how='any')
        #dropping first line as it is n/a
        test obs = 12
        train = new_df_difference[:-test_obs]
        test = new_df_difference[-test_obs:]
        # Last 12 data point as test points
        for i in [1,2,3,4,5,6,7,8,9,10]:
            model = VAR(train)
            results = model.fit(i)
            print('Order =', i)
            print('AIC: ', results.aic)
            print('BIC: ', results.bic)
```

```
print()
# Run AIC and BIC of VAR(1) to VAR(10)
result = model.fit(1)
result.summary()
#VAR(1) is most logical choice for simple and clean model
lagged_Values = train.values[-8:]
pred = result.forecast(y=lagged_Values, steps=12)
#forecasting 12 steps ahead
idx = pd.date_range('2023-11-08', periods=12, freq='1D')
df_forecast=pd.DataFrame(data=pred, index=idx, columns=['SP_2d', 'DJ_2d'])
#adding forecasted difference back to original index closing
df_forecast['SP1d'] = (df['SP'].iloc[-test_obs-1]-df['SP'].iloc[-test_obs-2]) + df_for
df_forecast['SPForecast'] = df['SP'].iloc[-test_obs-1] + df_forecast['SP1d'].cumsum()
df_forecast['DJ1d'] = (df['DJ'].iloc[-test_obs-1]-df['DJ'].iloc[-test_obs-2]) + df_for
df_forecast['DJForecast'] = df['DJ'].iloc[-test_obs-1] + df_forecast['DJ1d'].cumsum()
test_original = df[-test_obs:]
test_original.index = pd.to_datetime(test_original.index)
f = plt.figure()
f.set_figwidth(12)
f.set_figheight(5)
plt.plot(df_forecast['SPForecast'].index, test_original['SP'], label ='SP')
plt.plot(df_forecast['SPForecast'].index, df_forecast['SPForecast'], '-.', label ='SPF
plt.xlabel("Date")
plt.ylabel("Index Closing")
plt.legend()
plt.title('Actual vs Forecasting')
plt.show()
MSE_SP = mean_squared_error(test_original['SP'], df_forecast['SPForecast'])
print(" Mean Squared Error of S&P Forecasting vs actual = ", MSE_SP)
f = plt.figure()
f.set figwidth(12)
f.set_figheight(5)
plt.plot(df_forecast['DJForecast'].index, test_original['DJ'], label ='DJ')
plt.plot(df_forecast['DJForecast'].index, df_forecast['DJForecast'], '-.', label ='DJF
plt.xlabel("Date")
plt.ylabel("Index Closing")
plt.legend()
plt.title('Actual vs Forecasting')
plt.show()
MSE_DJ = mean_squared_error(test_original['DJ'], df_forecast['DJForecast'])
print(" Mean Squared Error of DJ Index Forecasting vs actual = ", MSE_DJ)
# plotting forecasted valued of S&P and Dow Index against actual indices
# calculating mean squared errors of S&P & Dow foreastings vs actuals
```

Augmented Dickey-Fuller Test: ADF test statistic -1.572177e+01 p-value 1.321288e-28 # lags used 9.000000e+00 # observations 2.506000e+03 critical value (1%) -3.432962e+00 critical value (5%) -2.862694e+00 critical value (10%) -2.567384e+00 Strong evidence against the null hypothesis Reject the null hypothesis Data has no unit root and is stationary Augmented Dickey-Fuller Test: ADF test statistic -1.565104e+01 p-value 1.596371e-28 # lags used 9.000000e+00 # observations 2.506000e+03 critical value (1%) -3.432962e+00 critical value (5%) -2.862694e+00 critical value (10%) -2.567384e+00 Strong evidence against the null hypothesis Reject the null hypothesis Data has no unit root and is stationary Order = 1AIC: 16.086664317766086 BIC: 16.100628149867003 Order = 2AIC: 16.081833051120153 BIC: 16.105113809277736 Order = 3AIC: 16.084552300518894 BIC: 16.11715615618954 Order = 4AIC: 16.08428213729433 BIC: 16.126215268572498 Order = 5AIC: 16.08515356323937 BIC: 16.13642215486737 Order = 6AIC: 16.079894194369633 BIC: 16.140504437747463 Order = 7AIC: 16.06642149452195 BIC: 16.13637958771714 Order = 8AIC: 16.063054049984775 BIC: 16.14236619774224 Order = 9AIC: 16.054513415021052 BIC: 16.143185828772978

Order = 10

AIC: 16.05608462475179

BIC: 16.154123522627522

C:\Users\New User\anaconda3\lib\site-packages\statsmodels\tsa\base\tsa_model.py:471: ValueWarning: An unsupported index was provided and will be ignored when e.g. forecas ting.

self._init_dates(dates, freq)

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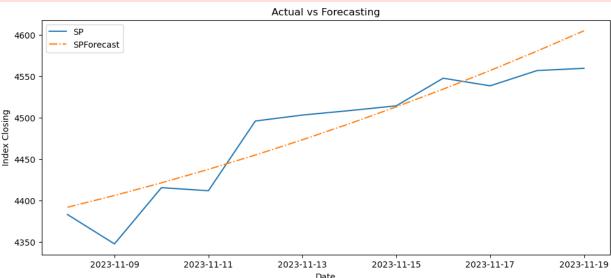
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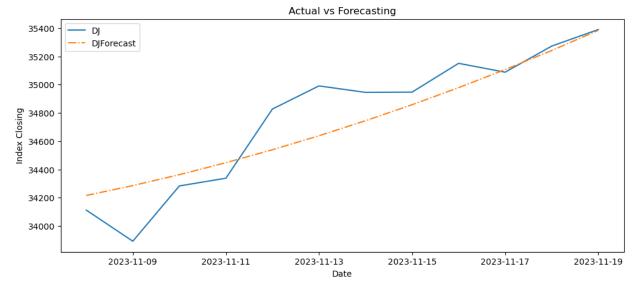
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Mean Squared Error of S&P Forecasting vs actual = 845.6290282401527



Mean Squared Error of DJ Index Forecasting vs actual = 39237.28143171107

In []: