

```
In [1]: import pandas as pa
import numpy as np
import matplotlib.pyplot as plt
```

```
In [2]: eustocks = pa.read_csv(r'C:\Users\SACHIN K M\Desktop\python\data\dataset\EuIndices.csv')
eustocks.head()
```

Out[2]:

	Unnamed: 0	DAX	SMI	CAC	FTSE
0	1	1628.75	1678.1	1772.8	2443.6
1	2	1613.63	1688.5	1750.5	2460.2
2	3	1606.51	1678.6	1718.0	2448.2
3	4	1621.04	1684.1	1708.1	2470.4
4	5	1618.16	1686.6	1723.1	2484.7

```
In [3]: eustocks_df = pa.DataFrame(data = eustocks.values, columns = ['0', 'DAX', 'SMI', 'CAC', 'FTSE'],
                                index = pa.DatetimeIndex(start='1991-01-01',
periods = 1860, freq='B'))

eustocks_df = eustocks_df.drop(columns=['0'])
eustocks_df.head()
```

C:\Users\SACHIN K M\Anaconda3\lib\site-packages\ipykernel\_launcher.py:  
2: FutureWarning: Creating a DatetimeIndex by passing range endpoints is deprecated. Use `pandas.date\_range` instead.

Out[3]:

	DAX	SMI	CAC	FTSE
1991-01-01	1628.75	1678.1	1772.8	2443.6

	DAX	SMI	CAC	FTSE
1991-01-02	1613.63	1688.5	1750.5	2460.2
1991-01-03	1606.51	1678.6	1718.0	2448.2
1991-01-04	1621.04	1684.1	1708.1	2470.4
1991-01-07	1618.16	1686.6	1723.1	2484.7

In [4]: `%matplotlib inline`

In [20]: `eustocks_df.describe()`

Out[20]:

	DAX	SMI	CAC	FTSE
count	1860.000000	1860.000000	1860.000000	1860.000000
mean	2530.656882	3376.223710	2227.828495	3565.643172
std	1084.792740	1663.026465	580.314198	976.715540
min	1402.340000	1587.400000	1611.000000	2281.000000
25%	1744.102500	2165.625000	1875.150000	2843.150000
50%	2140.565000	2796.350000	1992.300000	3246.600000
75%	2722.367500	3812.425000	2274.350000	3993.575000
max	6186.090000	8412.000000	4388.500000	6179.000000

In [5]: `#there are no nan-value present in datasets`

```
plt.figure(figsize=(12,7))
plt.plot(eustocks_df)
plt.legend()
```

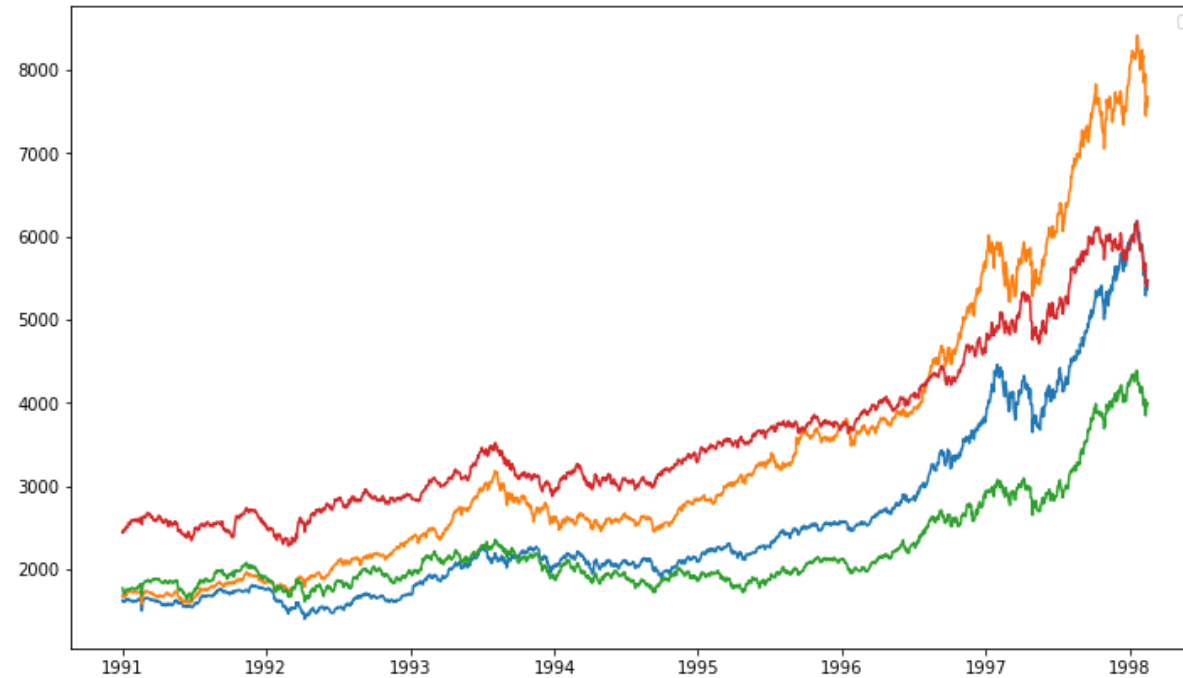
C:\Users\SACHIN K M\Anaconda3\lib\site-packages\pandas\plotting\\_converter.py:129: FutureWarning: Using an implicitly registered datetime converter for a matplotlib plotting method. The converter was registered by pandas on import. Future versions of pandas will require you to explici

```
tly register matplotlib converters.
```

To register the converters:

```
>>> from pandas.plotting import register_matplotlib_converters
>>> register_matplotlib_converters()
warnings.warn(msg, FutureWarning)
No handles with labels found to put in legend.
```

Out[5]: <matplotlib.legend.Legend at 0x2eb90090cf8>



```
In [6]: #testing for stationarity here we have to check for each ciolumns

from statsmodels.tsa.stattools import adfuller
def stationarity_test(timeseries):
    print("results of Dicky fuller test :")
    dftest = adfuller(timeseries, autolag='AIC')
    df_output = pa.Series(dftest[0:4], index=['test statistics', 'P-val
ue', '#lags_used', 'number of observations'])
```

```
for key, value in dfctest[4].items():
    df_output['critical value (%s)'%key] = value
print(df_output)
```

```
In [7]: import functools
eustocks_df.apply(functools.partial(stationarity_test))
```

```
results of Dicky fuller test :
test statistics          2.295811
P-value                 0.998953
#lags_used              24.000000
number of observations   1835.000000
critical value (1%)      -3.433919
critical value (5%)      -2.863116
critical value (10%)     -2.567609
dtype: float64
results of Dicky fuller test :
test statistics          2.235032
P-value                 0.998910
#lags_used              22.000000
number of observations   1837.000000
critical value (1%)      -3.433915
critical value (5%)      -2.863115
critical value (10%)     -2.567608
dtype: float64
results of Dicky fuller test :
test statistics          1.286413
P-value                 0.996532
#lags_used              17.000000
number of observations   1842.000000
critical value (1%)      -3.433905
critical value (5%)      -2.863110
critical value (10%)     -2.567606
dtype: float64
results of Dicky fuller test :
test statistics          0.106116
P-value                 0.966476
#lags_used              13.000000
```

```
number of observations    1846.000000
critical value (1%)      -3.433897
critical value (5%)      -2.863107
critical value (10%)     -2.567604
dtype: float64
```

```
Out[7]: DAX      None
        SMI      None
        CAC      None
        FTSE     None
        dtype: object
```

**the p value should be less than 0.05 if it is more like 0.9 the series is non stationary as in this case**

```
In [8]: eustockdiff = eustocks_df.diff().dropna()
        eustockdiff.head()
```

```
Out[8]:
```

	DAX	SMI	CAC	FTSE
1991-01-02	-15.12	10.4	-22.3	16.6
1991-01-03	-7.12	-9.9	-32.5	-12.0
1991-01-04	14.53	5.5	-9.9	22.2
1991-01-07	-2.88	2.5	15.0	14.3
1991-01-08	-7.55	-15.0	-8.8	-17.9

```
In [9]: eustockdiff.apply(functools.partial(stationarity_test))
```

```
results of Dicky fuller test :
test statistics      -8.293747e+00
P-value              4.193799e-13
#lags_used           1.900000e+01
number of observations 1.839000e+03
```

```
critical value (1%)      -3.433911e+00
critical value (5%)      -2.863113e+00
critical value (10%)     -2.567607e+00
dtype: float64
results of Dicky fuller test :
test statistics          -8.543303e+00
P-value                  9.652044e-14
#lags_used               1.900000e+01
number of observations   1.839000e+03
critical value (1%)      -3.433911e+00
critical value (5%)      -2.863113e+00
critical value (10%)     -2.567607e+00
dtype: float64
results of Dicky fuller test :
test statistics          -1.015705e+01
P-value                  7.680823e-18
#lags_used               1.600000e+01
number of observations   1.842000e+03
critical value (1%)      -3.433905e+00
critical value (5%)      -2.863110e+00
critical value (10%)     -2.567606e+00
dtype: float64
results of Dicky fuller test :
test statistics          -1.079549e+01
P-value                  2.080690e-19
#lags_used               1.200000e+01
number of observations   1.846000e+03
critical value (1%)      -3.433897e+00
critical value (5%)      -2.863107e+00
critical value (10%)     -2.567604e+00
dtype: float64
```

```
Out[9]: DAX      None
        SMI      None
        CAC      None
        FTSE     None
        dtype: object
```

```
In [10]: from statsmodels.tsa.vector_ar.var_model import VAR
```

```
model = VAR(eustockdiff)
```

```
In [11]: models = model.select_order(15)
models.summary()

#here we will get the result which will highlight the lowest value from
all the variables
```

```
Out[11]: VAR Order Selection (* highlights the
minimums)
```

	AIC	BIC	FPE	HQIC
0	25.31	25.32*	9.833e+10	25.32
1	25.28	25.34	9.505e+10	25.30*
2	25.28	25.39	9.553e+10	25.32
3	25.27	25.43	9.458e+10	25.33
4	25.26	25.47	9.382e+10	25.34
5	25.26	25.51	9.357e+10	25.35
6	25.26	25.56	9.332e+10	25.37
7	25.26	25.61	9.361e+10	25.39
8	25.26	25.66	9.371e+10	25.41
9	25.26	25.70	9.314e+10	25.42
10	25.26	25.75	9.358e+10	25.44
11	25.26*	25.79	9.303e+10*	25.45
12	25.26	25.85	9.347e+10	25.48
13	25.26	25.89	9.307e+10	25.49
14	25.26	25.94	9.329e+10	25.51
15	25.27	26.00	9.425e+10	25.54

```
In [12]: results = model.fit(maxlags=15, ic = 'aic')
```

```
results.summary()
```

Out[12]: Summary of Regression Results

```
=====
Model:                                VAR
Method:                               OLS
Date:                                Mon, 15, Jul, 2019
Time:                                23:36:11
-----
No. of Equations:                     4.00000    BIC:                                25.7862
Nobs:                                1848.00    HQIC:                               25.4466
Log likelihood:                       -33638.3    FPE:                               9.23092e+10
AIC:                                25.2484    Det(Omega_mle):                    8.38399e+10
-----
```

Results for equation DAX

```
=====
=====
====
              coefficient      std. error      t-stat
prob
-----
----
const          1.689997          0.772454          2.188
0.029
L1.DAX         -0.002794          0.042406         -0.066
0.947
L1.SMI         -0.106047          0.029764         -3.563
0.000
L1.CAC          0.074580          0.046291          1.611
0.107
L1.FTSE         0.084242          0.036405          2.314
0.021
L2.DAX          0.006051          0.042357          0.143
0.886
L2.SMI         -0.023744          0.029874         -0.795
0.427
L2.CAC          0.082567          0.046370          1.781
0.075
L2.FTSE        -0.082340          0.036647         -2.247
0.025
L3.DAX         -0.070828          0.042131         -1.681
```



0.093			
L3.SMI	0.005490	0.029829	0.184
0.854			
L3.CAC	0.056238	0.046490	1.210
0.226			
L3.FTSE	0.034574	0.036766	0.940
0.347			
L4.DAX	-0.050557	0.042108	-1.201
0.230			
L4.SMI	0.015083	0.029975	0.503
0.615			
L4.CAC	0.107053	0.046670	2.294
0.022			
L4.FTSE	-0.060486	0.036907	-1.639
0.101			
L5.DAX	-0.002764	0.042397	-0.065
0.948			
L5.SMI	-0.090774	0.030195	-3.006
0.003			
L5.CAC	0.082208	0.046651	1.762
0.078			
L5.FTSE	-0.032179	0.036959	-0.871
0.384			
L6.DAX	-0.013454	0.042287	-0.318
0.750			
L6.SMI	0.027158	0.030250	0.898
0.369			
L6.CAC	0.102798	0.046720	2.200
0.028			
L6.FTSE	-0.011240	0.036843	-0.305
0.760			
L7.DAX	0.060914	0.042345	1.439
0.150			
L7.SMI	0.008474	0.030296	0.280
0.780			
L7.CAC	-0.057849	0.046792	-1.236
0.216			
L7.FTSE	-0.047575	0.036888	-1.290
0.197			

L8.DAX 0.342	-0.040198	0.042332	-0.950
L8.SMI 0.002	0.094140	0.030456	3.091
L8.CAC 0.019	-0.109483	0.046762	-2.341
L8.FTSE 0.165	0.051233	0.036867	1.390
L9.DAX 0.743	-0.013915	0.042390	-0.328
L9.SMI 0.682	-0.012519	0.030503	-0.410
L9.CAC 0.003	0.137788	0.046503	2.963
L9.FTSE 0.961	0.001811	0.036969	0.049
L10.DAX 0.142	0.062420	0.042460	1.470
L10.SMI 0.317	0.030617	0.030615	1.000
L10.CAC 0.921	-0.004618	0.046573	-0.099
L10.FTSE 0.026	-0.082541	0.037087	-2.226
L11.DAX 0.045	0.084358	0.042150	2.001
L11.SMI 0.040	0.062522	0.030442	2.054
L11.CAC 0.992	-0.000481	0.046600	-0.010
L11.FTSE 0.240	-0.043432	0.036936	-1.176
=====			
=====			
Results for equation SMI			
=====			
=====			
	coefficient	std. error	t-stat

prob

```
-----  
----  
const          2.666518      0.945949      2.819  
0.005  
L1.DAX         0.015861      0.051931      0.305  
0.760  
L1.SMI        -0.045199      0.036448     -1.240  
0.215  
L1.CAC         0.048672      0.056688      0.859  
0.391  
L1.FTSE        0.140080      0.044582      3.142  
0.002  
L2.DAX        -0.011278      0.051871     -0.217  
0.828  
L2.SMI         0.008591      0.036583      0.235  
0.814  
L2.CAC         0.124562      0.056785      2.194  
0.028  
L2.FTSE       -0.098270      0.044878     -2.190  
0.029  
L3.DAX        -0.154551      0.051594     -2.996  
0.003  
L3.SMI        -0.004518      0.036528     -0.124  
0.902  
L3.CAC         0.106761      0.056931      1.875  
0.061  
L3.FTSE        0.122540      0.045024      2.722  
0.006  
L4.DAX        -0.179588      0.051565     -3.483  
0.000  
L4.SMI         0.027766      0.036707      0.756  
0.449  
L4.CAC         0.134131      0.057153      2.347  
0.019  
L4.FTSE       -0.006755      0.045197     -0.149  
0.881  
L5.DAX       -0.010278      0.051919     -0.198  
0.843
```

L5.SMI 0.027	-0.081579	0.036977	-2.206
L5.CAC 0.148	0.082653	0.057129	1.447
L5.FTSE 0.312	-0.045753	0.045260	-1.011
L6.DAX 0.145	-0.075441	0.051785	-1.457
L6.SMI 0.732	0.012696	0.037044	0.343
L6.CAC 0.033	0.121670	0.057213	2.127
L6.FTSE 0.758	-0.013926	0.045118	-0.309
L7.DAX 0.016	0.124840	0.051856	2.407
L7.SMI 0.966	-0.001572	0.037101	-0.042
L7.CAC 0.077	-0.101293	0.057302	-1.768
L7.FTSE 0.208	-0.056893	0.045173	-1.259
L8.DAX 0.965	0.002253	0.051840	0.043
L8.SMI 0.028	0.082039	0.037297	2.200
L8.CAC 0.126	-0.087675	0.057265	-1.531
L8.FTSE 0.676	-0.018896	0.045147	-0.419
L9.DAX 0.080	0.090922	0.051910	1.752
L9.SMI 0.002	-0.118136	0.037354	-3.163
L9.CAC 0.133	0.085499	0.056948	1.501
L9.FTSE 0.088	0.077151	0.045272	1.704
L10.DAX	0.044624	0.051996	0.858

0.391			
L10.SMI	0.044485	0.037491	1.187
0.235			
L10.CAC	0.047197	0.057033	0.828
0.408			
L10.FTSE	-0.091617	0.045417	-2.017
0.044			
L11.DAX	0.091222	0.051617	1.767
0.077			
L11.SMI	0.074200	0.037280	1.990
0.047			
L11.CAC	-0.002739	0.057067	-0.048
0.962			
L11.FTSE	-0.032980	0.045232	-0.729
0.466			

=====

=====

Results for equation CAC

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=====

	coefficient	std. error	t-stat
prob			
-----			
----			
const	1.259414	0.629391	2.001
0.045			
L1.DAX	-0.004808	0.034552	-0.139
0.889			
L1.SMI	-0.075696	0.024251	-3.121
0.002			
L1.CAC	0.042553	0.037717	1.128
0.259			
L1.FTSE	0.087086	0.029663	2.936
0.003			
L2.DAX	0.010514	0.034512	0.305
0.761			
L2.SMI	-0.032858	0.024341	-1.350
0.177			

L2.CAC 0.020	0.087829	0.037782	2.325
L2.FTSE 0.033	-0.063708	0.029859	-2.134
L3.DAX 0.340	-0.032736	0.034328	-0.954
L3.SMI 0.275	0.026506	0.024304	1.091
L3.CAC 0.129	-0.057546	0.037880	-1.519
L3.FTSE 0.468	0.021721	0.029957	0.725
L4.DAX 0.001	-0.114794	0.034309	-3.346
L4.SMI 0.069	0.044344	0.024423	1.816
L4.CAC 0.056	0.072546	0.038027	1.908
L4.FTSE 0.432	-0.023634	0.030072	-0.786
L5.DAX 0.163	-0.048215	0.034545	-1.396
L5.SMI 0.128	-0.037409	0.024603	-1.521
L5.CAC 0.202	0.048496	0.038011	1.276
L5.FTSE 0.705	-0.011392	0.030114	-0.378
L6.DAX 0.691	-0.013699	0.034455	-0.398
L6.SMI 0.467	0.017909	0.024647	0.727
L6.CAC 0.145	0.055414	0.038067	1.456
L6.FTSE 0.841	-0.006018	0.030019	-0.200
L7.DAX 0.016	0.083216	0.034503	2.412
L7.SMI	-0.006611	0.024685	-0.268

0.789			
L7.CAC	-0.051019	0.038126	-1.338
0.181			
L7.FTSE	-0.061200	0.030056	-2.036
0.042			
L8.DAX	0.009058	0.034492	0.263
0.793			
L8.SMI	0.024778	0.024815	0.998
0.318			
L8.CAC	-0.097555	0.038101	-2.560
0.010			
L8.FTSE	0.048158	0.030039	1.603
0.109			
L9.DAX	0.026921	0.034539	0.779
0.436			
L9.SMI	-0.035057	0.024854	-1.411
0.158			
L9.CAC	0.050274	0.037890	1.327
0.185			
L9.FTSE	-0.004110	0.030122	-0.136
0.891			
L10.DAX	0.040372	0.034596	1.167
0.243			
L10.SMI	0.006593	0.024945	0.264
0.792			
L10.CAC	-0.019308	0.037947	-0.509
0.611			
L10.FTSE	-0.017077	0.030218	-0.565
0.572			
L11.DAX	0.058186	0.034343	1.694
0.090			
L11.SMI	0.013728	0.024804	0.553
0.580			
L11.CAC	-0.015487	0.037970	-0.408
0.683			
L11.FTSE	-0.010573	0.030095	-0.351
0.725			
=====			
=====			

# Results for equation FTSE

=====			
=====			
	coefficient	std. error	t-stat
prob			
-----			
----			
const	1.152197	0.726203	1.587
0.113			
L1.DAX	0.024204	0.039867	0.607
0.544			
L1.SMI	-0.093526	0.027981	-3.342
0.001			
L1.CAC	-0.008986	0.043519	-0.206
0.836			
L1.FTSE	0.174825	0.034225	5.108
0.000			
L2.DAX	0.017229	0.039821	0.433
0.665			
L2.SMI	-0.029392	0.028085	-1.047
0.295			
L2.CAC	0.036309	0.043594	0.833
0.405			
L2.FTSE	-0.031833	0.034452	-0.924
0.355			
L3.DAX	-0.056385	0.039609	-1.424
0.155			
L3.SMI	0.007695	0.028043	0.274
0.784			
L3.CAC	0.059397	0.043706	1.359
0.174			
L3.FTSE	0.007770	0.034565	0.225
0.822			
L4.DAX	-0.050778	0.039586	-1.283
0.200			
L4.SMI	0.038835	0.028180	1.378
0.168			
L4.CAC	0.081026	0.043876	1.847



0.065			
L4.FTSE	-0.091401	0.034697	-2.634
0.008			
L5.DAX	-0.018029	0.039858	-0.452
0.651			
L5.SMI	-0.034902	0.028387	-1.229
0.219			
L5.CAC	0.106804	0.043857	2.435
0.015			
L5.FTSE	-0.066915	0.034746	-1.926
0.054			
L6.DAX	-0.026199	0.039755	-0.659
0.510			
L6.SMI	0.058326	0.028439	2.051
0.040			
L6.CAC	0.105054	0.043922	2.392
0.017			
L6.FTSE	-0.093175	0.034637	-2.690
0.007			
L7.DAX	0.039812	0.039810	1.000
0.317			
L7.SMI	0.060004	0.028482	2.107
0.035			
L7.CAC	-0.054583	0.043990	-1.241
0.215			
L7.FTSE	-0.058763	0.034679	-1.694
0.090			
L8.DAX	-0.014377	0.039797	-0.361
0.718			
L8.SMI	0.058015	0.028633	2.026
0.043			
L8.CAC	-0.068864	0.043962	-1.566
0.117			
L8.FTSE	0.015254	0.034659	0.440
0.660			
L9.DAX	0.008078	0.039851	0.203
0.839			
L9.SMI	-0.025648	0.028677	-0.894
0.371			

L9.CAC	0.073327	0.043718	1.677
0.093			
L9.FTSE	0.013886	0.034755	0.400
0.690			
L10.DAX	0.048346	0.039917	1.211
0.226			
L10.SMI	-0.026559	0.028781	-0.923
0.356			
L10.CAC	-0.002653	0.043784	-0.061
0.952			
L10.FTSE	-0.036373	0.034867	-1.043
0.297			
L11.DAX	0.080742	0.039626	2.038
0.042			
L11.SMI	0.038228	0.028620	1.336
0.182			
L11.CAC	-0.012021	0.043810	-0.274
0.784			
L11.FTSE	0.003188	0.034724	0.092
0.927			

=====

====

Correlation matrix of residuals

	DAX	SMI	CAC	FTSE
DAX	1.000000	0.744557	0.742407	0.670121
SMI	0.744557	1.000000	0.638675	0.609294
CAC	0.742407	0.638675	1.000000	0.669407
FTSE	0.670121	0.609294	0.669407	1.000000

```
In [13]: #which variable we should keep in the final model
#Granger test for casality
#here kind f indicates f test
#we can go for chi-sqr test also
grangers = results.test_causality(['SMI', 'CAC', 'FTSE'], ['DAX'], kind
= 'f')
grangers.summary()
```

Out[13]:

Granger causality F-test. H<sub>0</sub>: DAX does not Granger-cause [SMI, CAC, FTSE]. Conclusion: reject H<sub>0</sub> at 5% significance level.

Test statistic	Critical value	p-value	df
1.681	1.438	0.009	(33, 7212)

```
In [14]: fcast = results.forecast(eustockdiff.values, 50)
```

```
In [15]: pa.DataFrame(fcast).head()
```

```
Out[15]:
```

	0	1	2	3
0	2.741324	0.096875	4.277166	-1.866084
1	-10.667937	1.598237	-3.815663	-8.844529
2	-16.799369	-16.271632	-12.841545	-7.252005
3	-5.835582	-29.049581	-2.570316	-7.261091
4	-42.079109	-27.371264	-17.589227	-18.620932

```
In [16]: DAXvalues = pa.DataFrame(fcast)[0]
DAXvalues.head()
```

```
Out[16]: 0    2.741324
1   -10.667937
2   -16.799369
3    -5.835582
4   -42.079109
Name: 0, dtype: float64
```

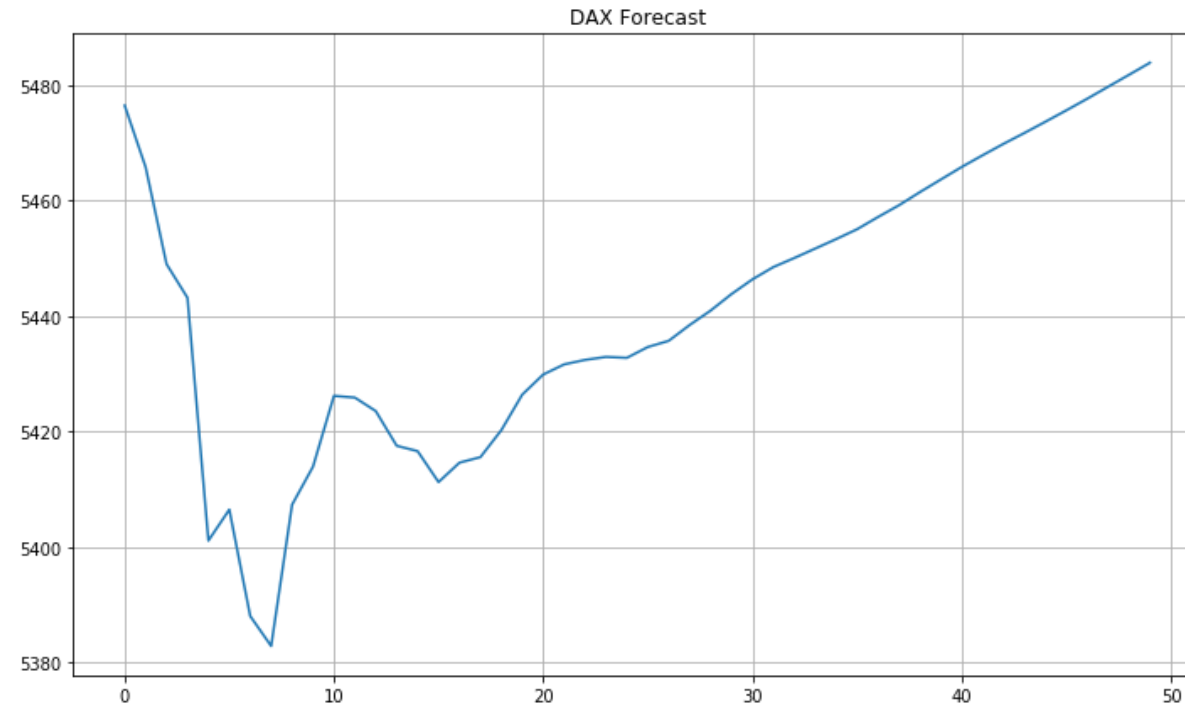
```
In [17]: eustocks_df.tail(1)
```

```
Out[17]:
```

	DAX	SMI	CAC	FTSE
1998-02-16	5473.72	7676.3	3995.0	5455.0

```
In [18]: dediff = np.cumsum(DAXvalues) + 5473.72
```

```
In [19]: plt.figure(figsize=(12,7))  
plt.plot(dediff)  
plt.title('DAX Forecast')  
plt.grid()  
plt.show()
```



```
In [ ]:
```