

M02_Xarxa_walkforward_normalitzat_multivariate

December 21, 2019

1 Xarxa neuronal

```
In [1]: import pandas as pd
import numpy as np
from pandas import datetime
from matplotlib import pyplot as plt

import keras
from keras.models import Sequential
from keras.layers import Dense
from keras.layers import LSTM

from keras.optimizers import SGD
from sklearn.model_selection import StratifiedKFold
from scipy.stats import uniform as sp_rand
from scipy.stats import randint
from time import time
from sklearn import preprocessing
```

Using TensorFlow backend.

1.1 Consum diari total multivariate one-step

```
In [2]: daily=pd.read_csv('C:/Users/Laura/Desktop/Smart meters London/workspace R/Dades netes/1
daily.head(5)
```

```
Out [2]:
```

	date	apparentTemperatureMax	sunsetTimeHour	weekday	season	\
0	2013-01-16	-0.15	16	3	winter	
1	2013-01-20	-0.46	16	7	winter	
2	2013-01-10	2.36	16	4	winter	
3	2013-01-06	6.98	16	7	winter	
4	2012-01-31	1.13	16	2	winter	

	cloudCover	humidity	visibility	month	energy_sum
0	0.48	0.91	4.12	1	13.147536
1	0.85	0.91	5.10	1	15.021900
2	0.70	0.94	5.21	1	12.066789

3	0.67	0.96	5.50	1	12.422263
4	0.55	0.84	5.62	1	13.890518

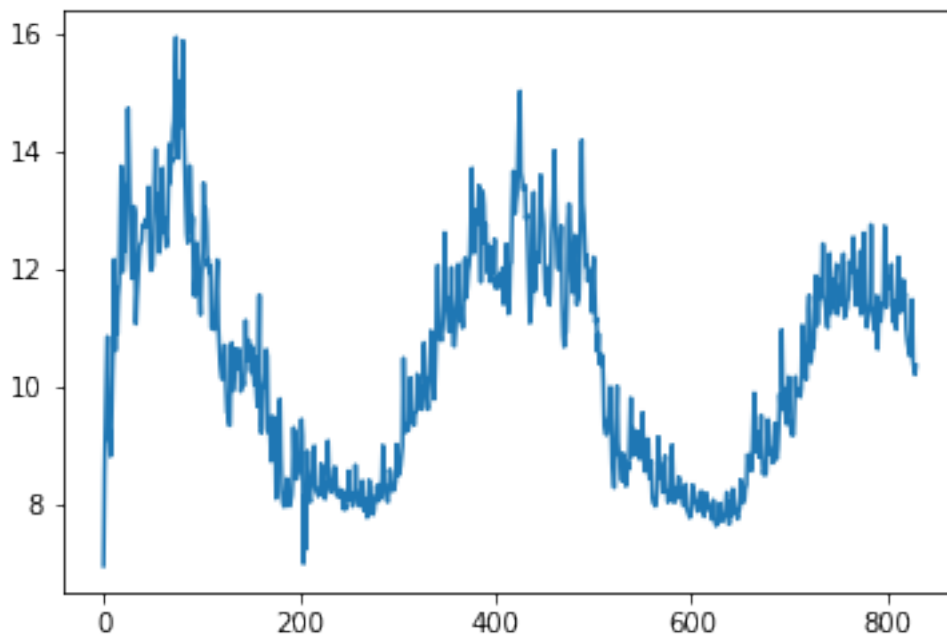
```
In [3]: #Ens quedem amb date i energy_sum, ordenem valors per data i resetejem index
daily_dia=daily[['date','energy_sum','apparentTemperatureMax']].sort_values(by=['date'])
daily_dia.head(5)
```

```
Out[3]:
```

	index	date	energy_sum	apparentTemperatureMax
0	677	2011-11-23	6.952692	10.36
1	691	2011-11-24	8.536480	12.93
2	713	2011-11-25	9.499781	13.03
3	728	2011-11-26	10.267707	12.96
4	729	2011-11-27	10.850805	13.54

```
In [16]: plt.plot(daily_dia )
```

```
Out[16]: [<matplotlib.lines.Line2D at 0x24f9e752240>]
```



```
In [4]: daily_dia['t-1']=daily_dia['energy_sum'].shift(1)
daily_dia['t-2']=daily_dia['energy_sum'].shift(2)
daily_dia['t-3']=daily_dia['energy_sum'].shift(3)
daily_dia['t-4']=daily_dia['energy_sum'].shift(4)
daily_dia['t-5']=daily_dia['energy_sum'].shift(5)
daily_dia['t-6']=daily_dia['energy_sum'].shift(6)
daily_dia['t-7']=daily_dia['energy_sum'].shift(7)
daily_dia['t-8']=daily_dia['energy_sum'].shift(8)
```

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daily_dia['temp(t-1)']=daily_dia['apparentTemperatureMax'].shift(1)
daily_dia['temp(t-2)']=daily_dia['apparentTemperatureMax'].shift(2)
daily_dia['temp(t-3)']=daily_dia['apparentTemperatureMax'].shift(3)
daily_dia['temp(t-4)']=daily_dia['apparentTemperatureMax'].shift(4)
daily_dia['temp(t-5)']=daily_dia['apparentTemperatureMax'].shift(5)
daily_dia['temp(t-6)']=daily_dia['apparentTemperatureMax'].shift(6)
daily_dia['temp(t-7)']=daily_dia['apparentTemperatureMax'].shift(7)
daily_dia['temp(t-8)']=daily_dia['apparentTemperatureMax'].shift(8)

```

daily_dia

```

Out[4]:
   index  date  energy_sum  apparentTemperatureMax  t-1 \
0      677 2011-11-23    6.952692                10.36   NaN
1      691 2011-11-24    8.536480                12.93  6.952692
2      713 2011-11-25    9.499781                13.03  8.536480
3      728 2011-11-26   10.267707                12.96  9.499781
4      729 2011-11-27   10.850805                13.54 10.267707
5      704 2011-11-28    9.103382                12.58 10.850805
6      718 2011-11-29    9.274873                13.47  9.103382
7      727 2011-11-30    8.813513                11.87  9.274873
8      778 2011-12-01    9.227707                12.15  8.813513
9      773 2011-12-02   10.145910                 5.33  9.227707
10     791 2011-12-03   10.780273                11.42 10.145910
11     822 2011-12-04   12.163127                 6.66 10.780273
12     807 2011-12-05   10.609714                 3.13 12.163127
13     813 2011-12-06   11.673417                 3.77 10.609714
14     810 2011-12-07   10.889362                 5.14 11.673417
15     788 2011-12-08   11.525150                12.89 10.889362
16     797 2011-12-09   11.759837                 3.99 11.525150
17     799 2011-12-10   12.633801                 3.14 11.759837
18     776 2011-12-11   13.749174                 5.72 12.633801
19     775 2011-12-12   11.951958                 5.94 13.749174
20     786 2011-12-13   11.957446                12.08 11.951958
21     818 2011-12-14   12.392776                 2.88 11.957446
22     795 2011-12-15   12.307079                 4.38 12.392776
23     763 2011-12-16   13.376080                 0.99 12.307079
24     770 2011-12-17   13.511968                 1.72 13.376080
25     808 2011-12-18   14.732271                 1.98 13.511968
26     757 2011-12-19   13.774471                 4.02 14.732271
27     803 2011-12-20   12.709106                 4.98 13.774471
28     748 2011-12-21   12.148570                12.14 12.709106
29     806 2011-12-22   11.839403                12.14 12.148570
...     ...      ...      ...      ...      ...
800     21 2014-01-29   11.800777                 2.53 11.344805
801     10 2014-01-30   11.685169                 5.86 11.800777
802     12 2014-01-31   11.857957                 5.27 11.685169

```

803	129	2014-02-01	11.710582		6.86	11.857957
804	155	2014-02-02	12.078164		6.48	11.710582
805	145	2014-02-03	11.280011		4.59	12.078164
806	134	2014-02-04	11.095584		5.63	11.280011
807	123	2014-02-05	11.415105		5.86	11.095584
808	118	2014-02-06	11.445403		7.34	11.415105
809	122	2014-02-07	10.972318		8.44	11.445403
810	126	2014-02-08	11.569300		5.67	10.972318
811	149	2014-02-09	12.202967		3.91	11.569300
812	132	2014-02-10	11.264175		7.07	12.202967
813	143	2014-02-11	11.452649		4.06	11.264175
814	131	2014-02-12	11.679099		4.73	11.452649
815	164	2014-02-13	11.285737		3.42	11.679099
816	125	2014-02-14	11.816914		12.02	11.285737
817	141	2014-02-15	11.490470		5.79	11.816914
818	151	2014-02-16	11.582159		7.88	11.490470
819	116	2014-02-17	10.979566		10.67	11.582159
820	128	2014-02-18	10.781898		10.13	10.979566
821	115	2014-02-19	10.674624		10.13	10.781898
822	121	2014-02-20	10.573835		12.50	10.674624
823	174	2014-02-21	10.518126		10.15	10.573835
824	167	2014-02-22	10.776242		11.63	10.518126
825	139	2014-02-23	11.480411		11.94	10.776242
826	162	2014-02-24	10.411403		14.23	11.480411
827	136	2014-02-25	10.294997		11.43	10.411403
828	161	2014-02-26	10.202945		11.29	10.294997
829	133	2014-02-27	10.356350		10.31	10.202945

	t-2	t-3	t-4	t-5	t-6	t-7 \
0	NaN	NaN	NaN	NaN	NaN	NaN
1	NaN	NaN	NaN	NaN	NaN	NaN
2	6.952692	NaN	NaN	NaN	NaN	NaN
3	8.536480	6.952692	NaN	NaN	NaN	NaN
4	9.499781	8.536480	6.952692	NaN	NaN	NaN
5	10.267707	9.499781	8.536480	6.952692	NaN	NaN
6	10.850805	10.267707	9.499781	8.536480	6.952692	NaN
7	9.103382	10.850805	10.267707	9.499781	8.536480	6.952692
8	9.274873	9.103382	10.850805	10.267707	9.499781	8.536480
9	8.813513	9.274873	9.103382	10.850805	10.267707	9.499781
10	9.227707	8.813513	9.274873	9.103382	10.850805	10.267707
11	10.145910	9.227707	8.813513	9.274873	9.103382	10.850805
12	10.780273	10.145910	9.227707	8.813513	9.274873	9.103382
13	12.163127	10.780273	10.145910	9.227707	8.813513	9.274873
14	10.609714	12.163127	10.780273	10.145910	9.227707	8.813513
15	11.673417	10.609714	12.163127	10.780273	10.145910	9.227707
16	10.889362	11.673417	10.609714	12.163127	10.780273	10.145910
17	11.525150	10.889362	11.673417	10.609714	12.163127	10.780273
18	11.759837	11.525150	10.889362	11.673417	10.609714	12.163127

19	12.633801	11.759837	11.525150	10.889362	11.673417	10.609714
20	13.749174	12.633801	11.759837	11.525150	10.889362	11.673417
21	11.951958	13.749174	12.633801	11.759837	11.525150	10.889362
22	11.957446	11.951958	13.749174	12.633801	11.759837	11.525150
23	12.392776	11.957446	11.951958	13.749174	12.633801	11.759837
24	12.307079	12.392776	11.957446	11.951958	13.749174	12.633801
25	13.376080	12.307079	12.392776	11.957446	11.951958	13.749174
26	13.511968	13.376080	12.307079	12.392776	11.957446	11.951958
27	14.732271	13.511968	13.376080	12.307079	12.392776	11.957446
28	13.774471	14.732271	13.511968	13.376080	12.307079	12.392776
29	12.709106	13.774471	14.732271	13.511968	13.376080	12.307079
..
800	11.753871	12.729659	11.620778	11.409880	11.300414	11.109560
801	11.344805	11.753871	12.729659	11.620778	11.409880	11.300414
802	11.800777	11.344805	11.753871	12.729659	11.620778	11.409880
803	11.685169	11.800777	11.344805	11.753871	12.729659	11.620778
804	11.857957	11.685169	11.800777	11.344805	11.753871	12.729659
805	11.710582	11.857957	11.685169	11.800777	11.344805	11.753871
806	12.078164	11.710582	11.857957	11.685169	11.800777	11.344805
807	11.280011	12.078164	11.710582	11.857957	11.685169	11.800777
808	11.095584	11.280011	12.078164	11.710582	11.857957	11.685169
809	11.415105	11.095584	11.280011	12.078164	11.710582	11.857957
810	11.445403	11.415105	11.095584	11.280011	12.078164	11.710582
811	10.972318	11.445403	11.415105	11.095584	11.280011	12.078164
812	11.569300	10.972318	11.445403	11.415105	11.095584	11.280011
813	12.202967	11.569300	10.972318	11.445403	11.415105	11.095584
814	11.264175	12.202967	11.569300	10.972318	11.445403	11.415105
815	11.452649	11.264175	12.202967	11.569300	10.972318	11.445403
816	11.679099	11.452649	11.264175	12.202967	11.569300	10.972318
817	11.285737	11.679099	11.452649	11.264175	12.202967	11.569300
818	11.816914	11.285737	11.679099	11.452649	11.264175	12.202967
819	11.490470	11.816914	11.285737	11.679099	11.452649	11.264175
820	11.582159	11.490470	11.816914	11.285737	11.679099	11.452649
821	10.979566	11.582159	11.490470	11.816914	11.285737	11.679099
822	10.781898	10.979566	11.582159	11.490470	11.816914	11.285737
823	10.674624	10.781898	10.979566	11.582159	11.490470	11.816914
824	10.573835	10.674624	10.781898	10.979566	11.582159	11.490470
825	10.518126	10.573835	10.674624	10.781898	10.979566	11.582159
826	10.776242	10.518126	10.573835	10.674624	10.781898	10.979566
827	11.480411	10.776242	10.518126	10.573835	10.674624	10.781898
828	10.411403	11.480411	10.776242	10.518126	10.573835	10.674624
829	10.294997	10.411403	11.480411	10.776242	10.518126	10.573835

	t-8	temp(t-1)	temp(t-2)	temp(t-3)	temp(t-4)	temp(t-5)	\
0	NaN	NaN	NaN	NaN	NaN	NaN	
1	NaN	10.36	NaN	NaN	NaN	NaN	
2	NaN	12.93	10.36	NaN	NaN	NaN	
3	NaN	13.03	12.93	10.36	NaN	NaN	

4	NaN	12.96	13.03	12.93	10.36	NaN
5	NaN	13.54	12.96	13.03	12.93	10.36
6	NaN	12.58	13.54	12.96	13.03	12.93
7	NaN	13.47	12.58	13.54	12.96	13.03
8	6.952692	11.87	13.47	12.58	13.54	12.96
9	8.536480	12.15	11.87	13.47	12.58	13.54
10	9.499781	5.33	12.15	11.87	13.47	12.58
11	10.267707	11.42	5.33	12.15	11.87	13.47
12	10.850805	6.66	11.42	5.33	12.15	11.87
13	9.103382	3.13	6.66	11.42	5.33	12.15
14	9.274873	3.77	3.13	6.66	11.42	5.33
15	8.813513	5.14	3.77	3.13	6.66	11.42
16	9.227707	12.89	5.14	3.77	3.13	6.66
17	10.145910	3.99	12.89	5.14	3.77	3.13
18	10.780273	3.14	3.99	12.89	5.14	3.77
19	12.163127	5.72	3.14	3.99	12.89	5.14
20	10.609714	5.94	5.72	3.14	3.99	12.89
21	11.673417	12.08	5.94	5.72	3.14	3.99
22	10.889362	2.88	12.08	5.94	5.72	3.14
23	11.525150	4.38	2.88	12.08	5.94	5.72
24	11.759837	0.99	4.38	2.88	12.08	5.94
25	12.633801	1.72	0.99	4.38	2.88	12.08
26	13.749174	1.98	1.72	0.99	4.38	2.88
27	11.951958	4.02	1.98	1.72	0.99	4.38
28	11.957446	4.98	4.02	1.98	1.72	0.99
29	12.392776	12.14	4.98	4.02	1.98	1.72
..
800	11.370601	6.34	4.34	5.99	11.77	5.72
801	11.109560	2.53	6.34	4.34	5.99	11.77
802	11.300414	5.86	2.53	6.34	4.34	5.99
803	11.409880	5.27	5.86	2.53	6.34	4.34
804	11.620778	6.86	5.27	5.86	2.53	6.34
805	12.729659	6.48	6.86	5.27	5.86	2.53
806	11.753871	4.59	6.48	6.86	5.27	5.86
807	11.344805	5.63	4.59	6.48	6.86	5.27
808	11.800777	5.86	5.63	4.59	6.48	6.86
809	11.685169	7.34	5.86	5.63	4.59	6.48
810	11.857957	8.44	7.34	5.86	5.63	4.59
811	11.710582	5.67	8.44	7.34	5.86	5.63
812	12.078164	3.91	5.67	8.44	7.34	5.86
813	11.280011	7.07	3.91	5.67	8.44	7.34
814	11.095584	4.06	7.07	3.91	5.67	8.44
815	11.415105	4.73	4.06	7.07	3.91	5.67
816	11.445403	3.42	4.73	4.06	7.07	3.91
817	10.972318	12.02	3.42	4.73	4.06	7.07
818	11.569300	5.79	12.02	3.42	4.73	4.06
819	12.202967	7.88	5.79	12.02	3.42	4.73
820	11.264175	10.67	7.88	5.79	12.02	3.42

821	11.452649	10.13	10.67	7.88	5.79	12.02
822	11.679099	10.13	10.13	10.67	7.88	5.79
823	11.285737	12.50	10.13	10.13	10.67	7.88
824	11.816914	10.15	12.50	10.13	10.13	10.67
825	11.490470	11.63	10.15	12.50	10.13	10.13
826	11.582159	11.94	11.63	10.15	12.50	10.13
827	10.979566	14.23	11.94	11.63	10.15	12.50
828	10.781898	11.43	14.23	11.94	11.63	10.15
829	10.674624	11.29	11.43	14.23	11.94	11.63

	temp(t-6)	temp(t-7)	temp(t-8)
0	NaN	NaN	NaN
1	NaN	NaN	NaN
2	NaN	NaN	NaN
3	NaN	NaN	NaN
4	NaN	NaN	NaN
5	NaN	NaN	NaN
6	10.36	NaN	NaN
7	12.93	10.36	NaN
8	13.03	12.93	10.36
9	12.96	13.03	12.93
10	13.54	12.96	13.03
11	12.58	13.54	12.96
12	13.47	12.58	13.54
13	11.87	13.47	12.58
14	12.15	11.87	13.47
15	5.33	12.15	11.87
16	11.42	5.33	12.15
17	6.66	11.42	5.33
18	3.13	6.66	11.42
19	3.77	3.13	6.66
20	5.14	3.77	3.13
21	12.89	5.14	3.77
22	3.99	12.89	5.14
23	3.14	3.99	12.89
24	5.72	3.14	3.99
25	5.94	5.72	3.14
26	12.08	5.94	5.72
27	2.88	12.08	5.94
28	4.38	2.88	12.08
29	0.99	4.38	2.88
..
800	4.93	10.02	6.26
801	5.72	4.93	10.02
802	11.77	5.72	4.93
803	5.99	11.77	5.72
804	4.34	5.99	11.77
805	6.34	4.34	5.99

806	2.53	6.34	4.34
807	5.86	2.53	6.34
808	5.27	5.86	2.53
809	6.86	5.27	5.86
810	6.48	6.86	5.27
811	4.59	6.48	6.86
812	5.63	4.59	6.48
813	5.86	5.63	4.59
814	7.34	5.86	5.63
815	8.44	7.34	5.86
816	5.67	8.44	7.34
817	3.91	5.67	8.44
818	7.07	3.91	5.67
819	4.06	7.07	3.91
820	4.73	4.06	7.07
821	3.42	4.73	4.06
822	12.02	3.42	4.73
823	5.79	12.02	3.42
824	7.88	5.79	12.02
825	10.67	7.88	5.79
826	10.13	10.67	7.88
827	10.13	10.13	10.67
828	12.50	10.13	10.13
829	10.15	12.50	10.13

[830 rows x 20 columns]

```
In [5]: #Ens quedem amb energies i temperatures
#No agafem apparent temperature max ja que quan fem la predicció representa que no ho
daily_dia=daily_dia[['energy_sum','t-1','t-2','t-3','t-4','t-5','t-6','t-7','t-8','temp(t-1)']]
daily_dia.head(5)
```

```
Out[5]:
```

	energy_sum	t-1	t-2	t-3	t-4	t-5	t-6	t-7	t-8	\
0	6.952692	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
1	8.536480	6.952692	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
2	9.499781	8.536480	6.952692	NaN	NaN	NaN	NaN	NaN	NaN	
3	10.267707	9.499781	8.536480	6.952692	NaN	NaN	NaN	NaN	NaN	
4	10.850805	10.267707	9.499781	8.536480	6.952692	NaN	NaN	NaN	NaN	

	temp(t-1)	temp(t-2)	temp(t-3)	temp(t-4)	temp(t-5)	temp(t-6)	\
0	NaN	NaN	NaN	NaN	NaN	NaN	
1	10.36	NaN	NaN	NaN	NaN	NaN	
2	12.93	10.36	NaN	NaN	NaN	NaN	
3	13.03	12.93	10.36	NaN	NaN	NaN	
4	12.96	13.03	12.93	10.36	NaN	NaN	

	temp(t-7)	temp(t-8)
0	NaN	NaN

1	NaN	NaN
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN

In [6]: *#Eliminem les 8 primeres files ja que contenen NaN (valors buits)*

```
daily_dia=daily_dia.drop([0,1,2,3,4,5,6,7])
daily_dia.head(5)
```

```
Out [6]:
```

	energy_sum	t-1	t-2	t-3	t-4	t-5	\
8	9.227707	8.813513	9.274873	9.103382	10.850805	10.267707	
9	10.145910	9.227707	8.813513	9.274873	9.103382	10.850805	
10	10.780273	10.145910	9.227707	8.813513	9.274873	9.103382	
11	12.163127	10.780273	10.145910	9.227707	8.813513	9.274873	
12	10.609714	12.163127	10.780273	10.145910	9.227707	8.813513	

	t-6	t-7	t-8	temp(t-1)	temp(t-2)	temp(t-3)	\
8	9.499781	8.536480	6.952692	11.87	13.47	12.58	
9	10.267707	9.499781	8.536480	12.15	11.87	13.47	
10	10.850805	10.267707	9.499781	5.33	12.15	11.87	
11	9.103382	10.850805	10.267707	11.42	5.33	12.15	
12	9.274873	9.103382	10.850805	6.66	11.42	5.33	

	temp(t-4)	temp(t-5)	temp(t-6)	temp(t-7)	temp(t-8)
8	13.54	12.96	13.03	12.93	10.36
9	12.58	13.54	12.96	13.03	12.93
10	13.47	12.58	13.54	12.96	13.03
11	11.87	13.47	12.58	13.54	12.96
12	12.15	11.87	13.47	12.58	13.54

In [7]: `len(daily_dia)`

Out [7]: 822

In [8]: *#normalitzem*

```
scaler=preprocessing.MinMaxScaler(feature_range=(0, 1))
daily_dia_norm=scaler.fit_transform(daily_dia)
```

```
#daily_dia_norm[:,0]
print(daily_dia_norm)
#daily_dia_norm[:,1:]
```

```
[[0.250036  0.20375985 0.25530572 ... 0.46920339 0.46646592 0.39611278]
 [0.35262316 0.250036  0.20375985 ... 0.46728716 0.46920339 0.46646592]
 [0.42349794 0.35262316 0.250036  ... 0.48316452 0.46728716 0.46920339]
 ...
 [0.36928   0.38228562 0.50172153 ... 0.38981659 0.38981659 0.40459896]
 [0.35899548 0.36928   0.38228562 ... 0.45469477 0.38981659 0.38981659]]
```

```
[0.37613476 0.35899548 0.36928      ... 0.39036408 0.45469477 0.38981659]]
```

```
In [47]: X_daily[0]
```

```
Out[47]: array([0.25530572, 0.2361457 , 0.43137821, 0.36623108, 0.28043381,  
                0.17280805, 0.          , 0.48124829, 0.45688475, 0.48316452,  
                0.46728716, 0.46920339, 0.46646592, 0.39611278])
```

```
In [9]: #Seleccionem dades per test i train
```

```
        y_daily=daily_dia_norm[:,0]
```

```
        X_daily=daily_dia_norm[:,1:17]
```

```
        #y_daily=daily_dia['energy_sum']
```

```
        #X_daily=daily_dia.drop(['energy_sum'], axis='columns')
```

```
        #Reshape de [samples,timesteps] a [samples,timesteps,features]
```

```
        #Enlloc de 14 features en son 7 de una feature i 7 duna altre
```

```
        X_daily=np.reshape(X_daily, (X_daily.shape[0], 8,2))
```

```
In [10]: # definim model
```

```
        import tensorflow as tf
```

```
        model =Sequential()
```

```
        model.add(LSTM(50, activation='relu', input_shape=(8, 2)))
```

```
        model.add(Dense(1))
```

```
        model.compile(optimizer='adam', loss='mse', metrics=['accuracy'])
```

WARNING:tensorflow:From c:\users\laura\appdata\local\programs\python\python37\lib\site-packages

Instructions for updating:

Colocations handled automatically by placer.

```
In [11]: import math
```

```
        from sklearn.metrics import mean_squared_error
```

```
        #Walk forward per test i train
```

```
        minim=100
```

```
        n_train=465
```

```
        lenght=len(daily_dia)
```

```
        llista_evaluate=list()
```

```
        llista_prediccions=list()
```

```
        llista_preditrain=list()
```

```
        llista_scores=list()
```

```
        llista_scoretrain=list()
```

```
        sumScores=0
```

```

for i in range(n_train, lenght):
    minim=minim+1
    X_train,X_test= X_daily[minim:i],X_daily[i:i+1]
    y_train,y_test= y_daily[minim:i],y_daily[i:i+1]

    #fem fit al model
    model.fit(X_train, y_train, epochs=50, verbose=0)

    #mostrem score per cada model
    score=model.evaluate(X_test,y_test,verbose=0)
    llista_evaluate.append(score)

    #Predim per cadascun
    preditest=model.predict(X_test)
    llista_prediccions.append(preditest)

    preditrain=model.predict(X_train)
    llista_preditrain.append(preditrain)

    trainScore = math.sqrt(mean_squared_error(y_train, preditrain))
    llista_scoretrain.append(trainScore )

    testScore = math.sqrt(mean_squared_error(y_test, preditest))
    llista_scores.append(testScore)

    sumScores=sumScores+testScore

```

WARNING:tensorflow:From c:\users\laura\appdata\local\programs\python\python37\lib\site-packages\tensorflow\python\ops\math_ops.py:3100: tf.nn.conv2d (from tensorflow.python.ops.nn_ops) taking and returning an array has been deprecated. Instructions for updating:
Use tf.nn.conv2d instead.

```

In [53]: #Dividim la suma de scores de test entre el nombre de prediccions per obtenir la mitjana
        sumScores/(lenght-n_train)

```

```

Out[53]: 0.036842092976021465

```

```

In [66]: sumScores/len(listpredi)

```

```

Out[66]: 0.036842092976021465

```

```

In [68]: print(model.metrics_names)

```

```

['loss', 'acc']

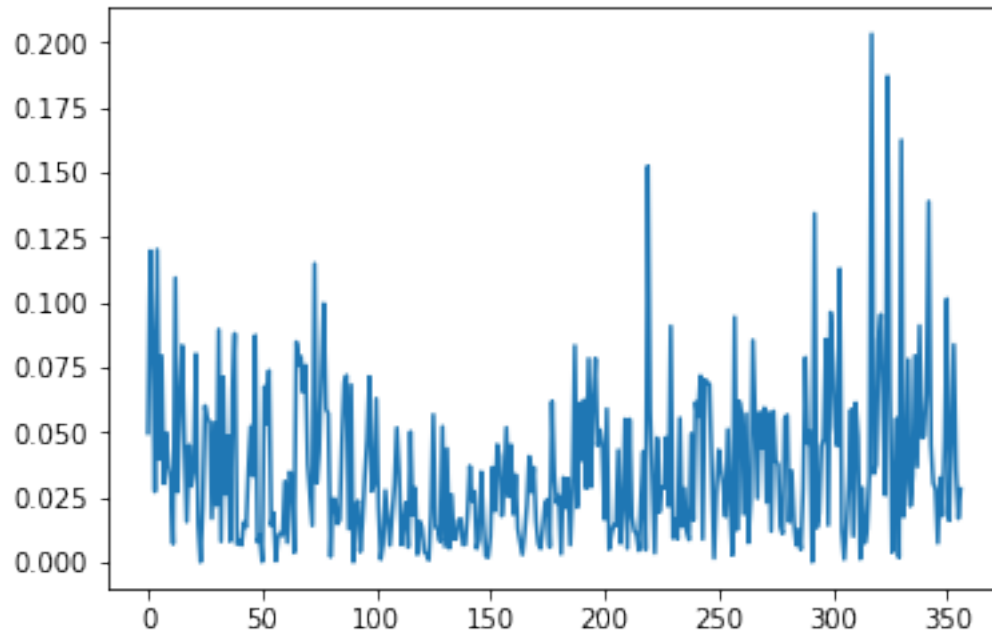
```

```

In [59]: plt.plot(llista_scores)

```

Out[59]: [<matplotlib.lines.Line2D at 0x210feb03898>]



In []:

In [14]: `predis=list()`

```
for i in range(len(llista_prediccions)):  
    predi=llista_prediccions[i].tolist()  
    predis.append(predi)
```

```
predis=np.reshape(predis, (357) )
```

`predis`

Out[14]: array([0.46085009, 0.53195691, 0.59825253, 0.68217349, 0.69342124,
0.57553029, 0.5935061 , 0.55025238, 0.57454509, 0.57619506,
0.45969853, 0.5292812 , 0.61381954, 0.59481907, 0.65893263,
0.72126961, 0.73082519, 0.64654744, 0.66026759, 0.59480816,
0.62077057, 0.61666107, 0.53775054, 0.55210423, 0.53695619,
0.53798634, 0.49620968, 0.63660848, 0.54668939, 0.51269543,
0.44384259, 0.4921385 , 0.44494465, 0.45125276, 0.4250561 ,
0.42657471, 0.3875967 , 0.40787923, 0.34827465, 0.24323215,
0.25302354, 0.24590921, 0.29008964, 0.29254162, 0.30004275,
0.32274878, 0.22045609, 0.25671917, 0.13658869, 0.18956566,
0.28308117, 0.26986459, 0.28494608, 0.28095829, 0.18739031,
0.15616195, 0.15610532, 0.20004138, 0.21367544, 0.17459342,

0.17834769, 0.16153166, 0.22600067, 0.21356571, 0.25172818,
0.23015893, 0.27986604, 0.17786652, 0.18779933, 0.28232181,
0.21208081, 0.21485993, 0.23658533, 0.11186726, 0.22881067,
0.24766654, 0.28773758, 0.38639536, 0.27423066, 0.23538253,
0.17292225, 0.19982389, 0.26149929, 0.24504127, 0.14484641,
0.21251664, 0.12604237, 0.19735295, 0.12779951, 0.1824587 ,
0.1090247 , 0.14129725, 0.16663992, 0.23697135, 0.16126239,
0.19751722, 0.23472074, 0.24496631, 0.16307765, 0.15593161,
0.26808146, 0.15869936, 0.13474932, 0.10942796, 0.10754447,
0.13934474, 0.15247327, 0.20789246, 0.18970646, 0.16762687,
0.1515674 , 0.13117447, 0.12994987, 0.1479177 , 0.15873691,
0.17604882, 0.15874287, 0.10602392, 0.11750112, 0.12529407,
0.1260069 , 0.1467385 , 0.11251688, 0.0968898 , 0.11446655,
0.1440805 , 0.10575226, 0.09352605, 0.14174378, 0.06806809,
0.10941784, 0.07730041, 0.10749334, 0.07307933, 0.10542613,
0.12550946, 0.14918178, 0.12090802, 0.09266201, 0.11687058,
0.1085849 , 0.08405167, 0.11069445, 0.12978655, 0.11814374,
0.09971824, 0.13798106, 0.12649515, 0.0816781 , 0.12291934,
0.12288383, 0.11742289, 0.09164251, 0.12099133, 0.1194756 ,
0.09757482, 0.08556155, 0.14038062, 0.10870832, 0.12573482,
0.07209659, 0.13046399, 0.10082746, 0.12773831, 0.08859462,
0.08631203, 0.09929159, 0.0770894 , 0.11980586, 0.1056841 ,
0.14174503, 0.10596839, 0.10586184, 0.07856441, 0.06176443,
0.12087047, 0.11021381, 0.09785564, 0.14949127, 0.09232625,
0.09066661, 0.12404196, 0.1139628 , 0.12693271, 0.17450196,
0.20769835, 0.16535307, 0.28991252, 0.1972138 , 0.14437976,
0.21796331, 0.26160559, 0.19088972, 0.17337954, 0.18212339,
0.17415878, 0.2806749 , 0.18483448, 0.23094603, 0.24898341,
0.18724485, 0.11625762, 0.17306979, 0.18237498, 0.20995328,
0.28788725, 0.25052544, 0.22284175, 0.19557461, 0.15625358,
0.18604693, 0.18659717, 0.2842918 , 0.20672452, 0.20962366,
0.22256437, 0.23253702, 0.27952942, 0.29607332, 0.2920379 ,
0.28167003, 0.26480588, 0.33692634, 0.32507628, 0.28410167,
0.26491195, 0.32638115, 0.32544938, 0.28100297, 0.33525574,
0.23782894, 0.26176921, 0.28443557, 0.29956496, 0.33674595,
0.35523292, 0.31664503, 0.32426965, 0.36594763, 0.35003546,
0.39140058, 0.31620353, 0.34396929, 0.38085651, 0.41988295,
0.3264496 , 0.41381109, 0.54643726, 0.45057848, 0.40897617,
0.36789387, 0.44404244, 0.45380437, 0.49869251, 0.59765822,
0.43061143, 0.46145371, 0.44688082, 0.50644404, 0.56693125,
0.59066272, 0.58895123, 0.45428407, 0.48997068, 0.50537479,
0.53940779, 0.50098002, 0.48264807, 0.53140885, 0.51865488,
0.43490461, 0.51088828, 0.48160696, 0.52762997, 0.49313605,
0.60768819, 0.51076961, 0.4766359 , 0.51608545, 0.57499653,
0.54848301, 0.51485586, 0.5533579 , 0.48247564, 0.4960584 ,
0.46896282, 0.49273619, 0.53125739, 0.4948459 , 0.59895986,
0.51470107, 0.55755138, 0.48630536, 0.53462493, 0.54072434,
0.53499371, 0.50931495, 0.62471718, 0.56444412, 0.49745327,

```

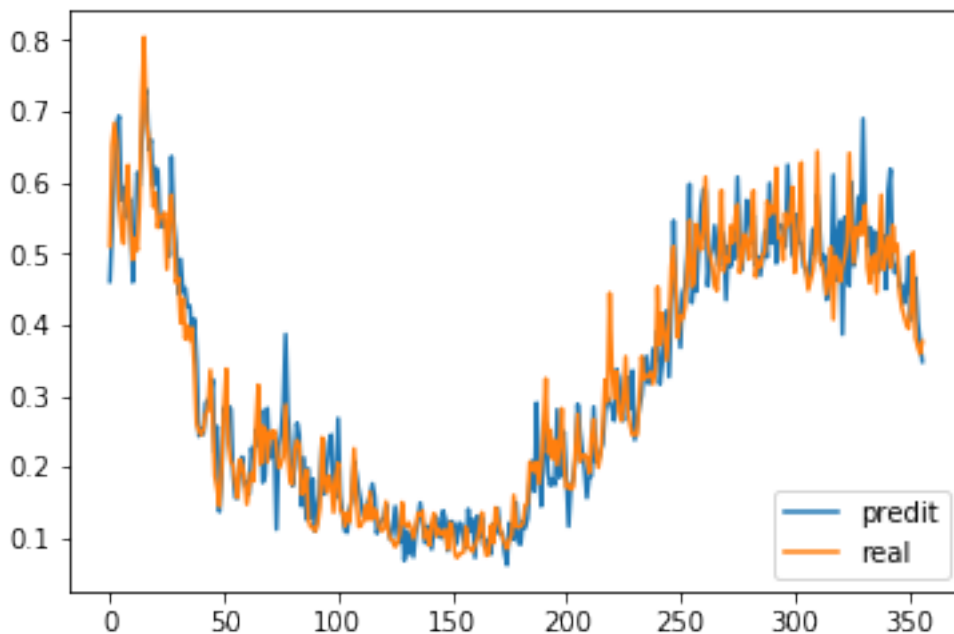
0.54938728, 0.55483735, 0.51443827, 0.51513612, 0.48172978,
0.47681195, 0.46316534, 0.49225408, 0.53438795, 0.49822861,
0.58253014, 0.53703815, 0.48398858, 0.49743167, 0.43505326,
0.45090792, 0.4758442 , 0.61044574, 0.46207792, 0.52703351,
0.54600912, 0.38630357, 0.55108428, 0.49155164, 0.45438111,
0.60053527, 0.48263288, 0.50469762, 0.5799759 , 0.54206264,
0.68978816, 0.55100197, 0.50997245, 0.53674525, 0.47281849,
0.53118455, 0.5244388 , 0.47488239, 0.49147928, 0.52545285,
0.4502126 , 0.58857542, 0.61873007, 0.49416113, 0.47246623,
0.48505291, 0.45323026, 0.45620453, 0.42956594, 0.44696513,
0.49554926, 0.4069531 , 0.45641488, 0.46597323, 0.40472919,
0.37585163, 0.34792179])

```

```

In [15]: ##Mostrem
plt.plot(predis, label="predit")
plt.plot(y_daily[n_train:lenght], label="real")
plt.legend(loc="lower right")
plt.show()

```



```

In [54]: #Creem un dataset amb format (nombre prediccions,17) per tornar les prediccions i els
#El necessitem d'aquesta mida encara que només volguem passar 2 variables ja que al fe
#per fer la inversa necessitem 17 variables
#Com que només en tenim 2, les ajuntem al dataset inicial i ens quedem amb 15 variabl
#Obtenint un dataset amb 15 variables aleatòries i les 2 variables que ens interessen

prova=daily_dia.iloc[n_train:lenght]

```

```

prova
#len(predis)
#lenght-n_train
prova['predi']=predis
prova['y']=y_daily[n_train:lenght]
prova=prova.drop(['energy_sum','t-1'], axis=1)
prova

prova=prova[['predi','y','t-2','t-3','t-4','t-5','t-6','t-7','t-8','temp(t-1)','temp(t-2)']]
prova

```

c:\users\laura\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html>
This is added back by InteractiveShellApp.init_path()
c:\users\laura\appdata\local\programs\python\python37\lib\site-packages\ipykernel_launcher.py:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html>
if sys.path[0] == '':

```

Out[54]:
      predi      y      t-2      t-3      t-4      t-5 \
473  0.460850  0.510600  10.889469  10.675248  10.860481  11.481859
474  0.531957  0.651732  10.930170  10.889469  10.675248  10.860481
475  0.598253  0.683428  11.559878  10.930170  10.889469  10.675248
476  0.682173  0.654997  12.823073  11.559878  10.930170  10.889469
477  0.693421  0.573173  13.106773  12.823073  11.559878  10.930170
478  0.575530  0.535873  12.852295  13.106773  12.823073  11.559878
479  0.593506  0.514061  12.119938  12.852295  13.106773  12.823073
480  0.550252  0.580609  11.786082  12.119938  12.852295  13.106773
481  0.574545  0.624326  11.590859  11.786082  12.119938  12.852295
482  0.576195  0.539280  12.186487  11.590859  11.786082  12.119938
483  0.459699  0.491355  12.577783  12.186487  11.590859  11.786082
484  0.529281  0.522145  11.816573  12.577783  12.186487  11.590859
485  0.613820  0.504442  11.387627  11.816573  12.577783  12.186487
486  0.594819  0.567725  11.663214  11.387627  11.816573  12.577783
487  0.658933  0.719460  11.504756  11.663214  11.387627  11.816573
488  0.721270  0.804631  12.071173  11.504756  11.663214  11.387627
489  0.730825  0.684716  13.429271  12.071173  11.504756  11.663214
490  0.646547  0.662177  14.191591  13.429271  12.071173  11.504756
491  0.660268  0.615194  13.118295  14.191591  13.429271  12.071173
492  0.594808  0.565466  12.916559  13.118295  14.191591  13.429271
493  0.620771  0.585646  12.496044  12.916559  13.118295  14.191591
494  0.616661  0.536523  12.050954  12.496044  12.916559  13.118295

```

495	0.537751	0.552256	12.231576	12.050954	12.496044	12.916559
496	0.552104	0.552256	11.791904	12.231576	12.050954	12.496044
497	0.536956	0.557809	11.932721	11.791904	12.231576	12.050954
498	0.537986	0.477794	11.932721	11.932721	11.791904	12.231576
499	0.496210	0.551195	11.982423	11.932721	11.932721	11.791904
500	0.636608	0.582339	11.266252	11.982423	11.932721	11.932721
501	0.546689	0.529772	11.923226	11.266252	11.982423	11.932721
502	0.512695	0.458904	12.201972	11.923226	11.266252	11.982423
...
800	0.504698	0.537515	11.753871	12.729659	11.620778	11.409880
801	0.579976	0.524598	11.344805	11.753871	12.729659	11.620778
802	0.542063	0.543903	11.800777	11.344805	11.753871	12.729659
803	0.689788	0.527438	11.685169	11.800777	11.344805	11.753871
804	0.551002	0.568506	11.857957	11.685169	11.800777	11.344805
805	0.509972	0.479332	11.710582	11.857957	11.685169	11.800777
806	0.536745	0.458726	12.078164	11.710582	11.857957	11.685169
807	0.472818	0.494425	11.280011	12.078164	11.710582	11.857957
808	0.531185	0.497810	11.095584	11.280011	12.078164	11.710582
809	0.524439	0.444954	11.415105	11.095584	11.280011	12.078164
810	0.474882	0.511653	11.445403	11.415105	11.095584	11.280011
811	0.491479	0.582450	10.972318	11.445403	11.415105	11.095584
812	0.525453	0.477562	11.569300	10.972318	11.445403	11.415105
813	0.450213	0.498620	12.202967	11.569300	10.972318	11.445403
814	0.588575	0.523920	11.264175	12.202967	11.569300	10.972318
815	0.618730	0.479971	11.452649	11.264175	12.202967	11.569300
816	0.494161	0.539318	11.679099	11.452649	11.264175	12.202967
817	0.472466	0.502845	11.285737	11.679099	11.452649	11.264175
818	0.485053	0.513089	11.816914	11.285737	11.679099	11.452649
819	0.453230	0.445764	11.490470	11.816914	11.285737	11.679099
820	0.456205	0.423680	11.582159	11.490470	11.816914	11.285737
821	0.429566	0.411694	10.979566	11.582159	11.490470	11.816914
822	0.446965	0.400434	10.781898	10.979566	11.582159	11.490470
823	0.495549	0.394209	10.674624	10.781898	10.979566	11.582159
824	0.406953	0.423048	10.573835	10.674624	10.781898	10.979566
825	0.456415	0.501722	10.518126	10.573835	10.674624	10.781898
826	0.465973	0.382286	10.776242	10.518126	10.573835	10.674624
827	0.404729	0.369280	11.480411	10.776242	10.518126	10.573835
828	0.375852	0.358995	10.411403	11.480411	10.776242	10.518126
829	0.347922	0.376135	10.294997	10.411403	11.480411	10.776242

	t-6	t-7	t-8	temp(t-1)	temp(t-2)	temp(t-3)	\
473	12.735907	12.308851	12.048499	9.04	7.99	13.14	
474	11.481859	12.735907	12.308851	7.53	9.04	7.99	
475	10.860481	11.481859	12.735907	0.33	7.53	9.04	
476	10.675248	10.860481	11.481859	-4.11	0.33	7.53	
477	10.889469	10.675248	10.860481	-0.56	-4.11	0.33	
478	10.930170	10.889469	10.675248	3.01	-0.56	-4.11	
479	11.559878	10.930170	10.889469	5.17	3.01	-0.56	

480	12.823073	11.559878	10.930170	4.56	5.17	3.01
481	13.106773	12.823073	11.559878	3.91	4.56	5.17
482	12.852295	13.106773	12.823073	5.13	3.91	4.56
483	12.119938	12.852295	13.106773	7.06	5.13	3.91
484	11.786082	12.119938	12.852295	5.81	7.06	5.13
485	11.590859	11.786082	12.119938	3.49	5.81	7.06
486	12.186487	11.590859	11.786082	2.57	3.49	5.81
487	12.577783	12.186487	11.590859	0.07	2.57	3.49
488	11.816573	12.577783	12.186487	-2.27	0.07	2.57
489	11.387627	11.816573	12.577783	-2.86	-2.27	0.07
490	11.663214	11.387627	11.816573	-2.89	-2.86	-2.27
491	11.504756	11.663214	11.387627	-2.29	-2.89	-2.86
492	12.071173	11.504756	11.663214	-0.19	-2.29	-2.89
493	13.429271	12.071173	11.504756	0.31	-0.19	-2.29
494	14.191591	13.429271	12.071173	1.71	0.31	-0.19
495	13.118295	14.191591	13.429271	1.53	1.71	0.31
496	12.916559	13.118295	14.191591	1.29	1.53	1.71
497	12.496044	12.916559	13.118295	1.64	1.29	1.53
498	12.050954	12.496044	12.916559	3.74	1.64	1.29
499	12.231576	12.050954	12.496044	-0.57	3.74	1.64
500	11.791904	12.231576	12.050954	-1.57	-0.57	3.74
501	11.932721	11.791904	12.231576	3.68	-1.57	-0.57
502	11.932721	11.932721	11.791904	8.53	3.68	-1.57
...
800	11.300414	11.109560	11.370601	6.34	4.34	5.99
801	11.409880	11.300414	11.109560	2.53	6.34	4.34
802	11.620778	11.409880	11.300414	5.86	2.53	6.34
803	12.729659	11.620778	11.409880	5.27	5.86	2.53
804	11.753871	12.729659	11.620778	6.86	5.27	5.86
805	11.344805	11.753871	12.729659	6.48	6.86	5.27
806	11.800777	11.344805	11.753871	4.59	6.48	6.86
807	11.685169	11.800777	11.344805	5.63	4.59	6.48
808	11.857957	11.685169	11.800777	5.86	5.63	4.59
809	11.710582	11.857957	11.685169	7.34	5.86	5.63
810	12.078164	11.710582	11.857957	8.44	7.34	5.86
811	11.280011	12.078164	11.710582	5.67	8.44	7.34
812	11.095584	11.280011	12.078164	3.91	5.67	8.44
813	11.415105	11.095584	11.280011	7.07	3.91	5.67
814	11.445403	11.415105	11.095584	4.06	7.07	3.91
815	10.972318	11.445403	11.415105	4.73	4.06	7.07
816	11.569300	10.972318	11.445403	3.42	4.73	4.06
817	12.202967	11.569300	10.972318	12.02	3.42	4.73
818	11.264175	12.202967	11.569300	5.79	12.02	3.42
819	11.452649	11.264175	12.202967	7.88	5.79	12.02
820	11.679099	11.452649	11.264175	10.67	7.88	5.79
821	11.285737	11.679099	11.452649	10.13	10.67	7.88
822	11.816914	11.285737	11.679099	10.13	10.13	10.67
823	11.490470	11.816914	11.285737	12.50	10.13	10.13

824	11.582159	11.490470	11.816914	10.15	12.50	10.13
825	10.979566	11.582159	11.490470	11.63	10.15	12.50
826	10.781898	10.979566	11.582159	11.94	11.63	10.15
827	10.674624	10.781898	10.979566	14.23	11.94	11.63
828	10.573835	10.674624	10.781898	11.43	14.23	11.94
829	10.518126	10.573835	10.674624	11.29	11.43	14.23

	temp(t-4)	temp(t-5)	temp(t-6)	temp(t-7)	temp(t-8)
473	16.06	6.17	5.04	3.13	3.28
474	13.14	16.06	6.17	5.04	3.13
475	7.99	13.14	16.06	6.17	5.04
476	9.04	7.99	13.14	16.06	6.17
477	7.53	9.04	7.99	13.14	16.06
478	0.33	7.53	9.04	7.99	13.14
479	-4.11	0.33	7.53	9.04	7.99
480	-0.56	-4.11	0.33	7.53	9.04
481	3.01	-0.56	-4.11	0.33	7.53
482	5.17	3.01	-0.56	-4.11	0.33
483	4.56	5.17	3.01	-0.56	-4.11
484	3.91	4.56	5.17	3.01	-0.56
485	5.13	3.91	4.56	5.17	3.01
486	7.06	5.13	3.91	4.56	5.17
487	5.81	7.06	5.13	3.91	4.56
488	3.49	5.81	7.06	5.13	3.91
489	2.57	3.49	5.81	7.06	5.13
490	0.07	2.57	3.49	5.81	7.06
491	-2.27	0.07	2.57	3.49	5.81
492	-2.86	-2.27	0.07	2.57	3.49
493	-2.89	-2.86	-2.27	0.07	2.57
494	-2.29	-2.89	-2.86	-2.27	0.07
495	-0.19	-2.29	-2.89	-2.86	-2.27
496	0.31	-0.19	-2.29	-2.89	-2.86
497	1.71	0.31	-0.19	-2.29	-2.89
498	1.53	1.71	0.31	-0.19	-2.29
499	1.29	1.53	1.71	0.31	-0.19
500	1.64	1.29	1.53	1.71	0.31
501	3.74	1.64	1.29	1.53	1.71
502	-0.57	3.74	1.64	1.29	1.53
..
800	11.77	5.72	4.93	10.02	6.26
801	5.99	11.77	5.72	4.93	10.02
802	4.34	5.99	11.77	5.72	4.93
803	6.34	4.34	5.99	11.77	5.72
804	2.53	6.34	4.34	5.99	11.77
805	5.86	2.53	6.34	4.34	5.99
806	5.27	5.86	2.53	6.34	4.34
807	6.86	5.27	5.86	2.53	6.34
808	6.48	6.86	5.27	5.86	2.53

809	4.59	6.48	6.86	5.27	5.86
810	5.63	4.59	6.48	6.86	5.27
811	5.86	5.63	4.59	6.48	6.86
812	7.34	5.86	5.63	4.59	6.48
813	8.44	7.34	5.86	5.63	4.59
814	5.67	8.44	7.34	5.86	5.63
815	3.91	5.67	8.44	7.34	5.86
816	7.07	3.91	5.67	8.44	7.34
817	4.06	7.07	3.91	5.67	8.44
818	4.73	4.06	7.07	3.91	5.67
819	3.42	4.73	4.06	7.07	3.91
820	12.02	3.42	4.73	4.06	7.07
821	5.79	12.02	3.42	4.73	4.06
822	7.88	5.79	12.02	3.42	4.73
823	10.67	7.88	5.79	12.02	3.42
824	10.13	10.67	7.88	5.79	12.02
825	10.13	10.13	10.67	7.88	5.79
826	12.50	10.13	10.13	10.67	7.88
827	10.15	12.50	10.13	10.13	10.67
828	11.63	10.15	12.50	10.13	10.13
829	11.94	11.63	10.15	12.50	10.13

[357 rows x 17 columns]

In [55]: *# Convert predictions back to normal values*

```
predi = scaler.inverse_transform(prova)
print(predi[0][0])
print(predi[0][1])
```

#Les variables en posició 15 i 16 són predicció i y respectivament

11.114592110340794
11.559878061079399

In [57]: *#Fem una llista amb les prediccions i una llista amb y(valor real)*

```
listpredi=list()
for i in range(len(predi)):
    listpredi.append(predi[i][0])
listpredi

listy=list()
for i in range(len(predi)):
    listy.append(predi[i][1])
listy
```

Out [57]: [11.559878061079399,
12.8230727297735,

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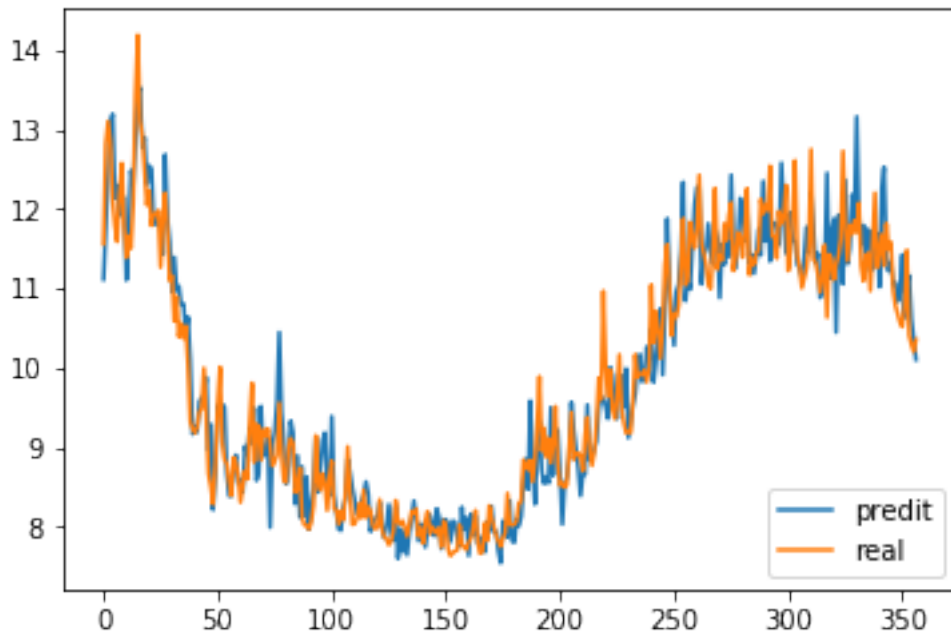
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10.3563498993587]
```

```
In [58]: ##Mostrem  
plt.plot(listpredi, label="predict")  
plt.plot(listy, label="real")  
plt.legend(loc="lower right")  
plt.show()
```



```
In [60]: listy[1]
```

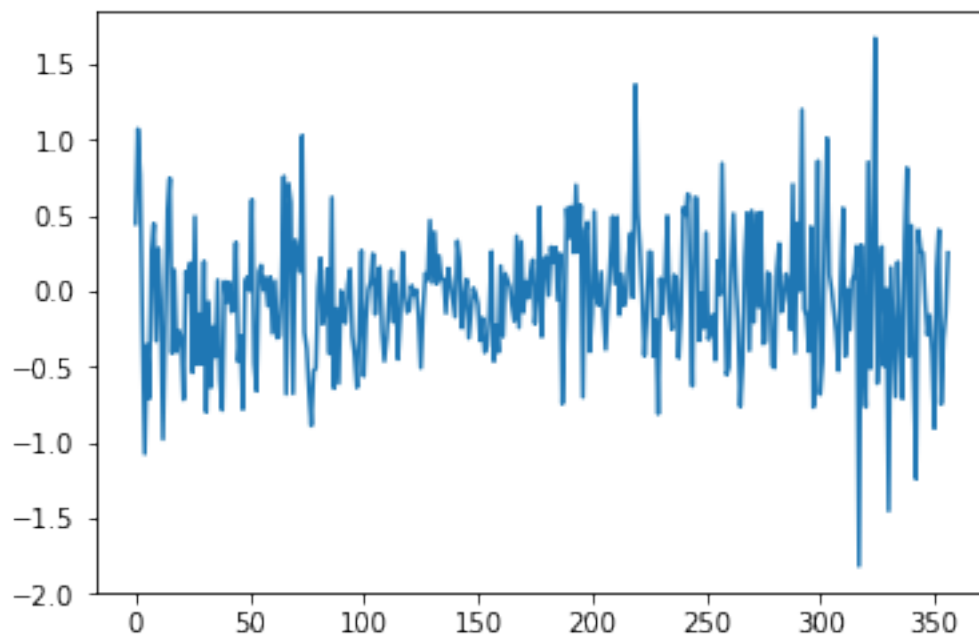
```
Out[60]: 12.8230727297735
```

```
In [62]: llista_errors=list()  
         llista_errorsabs=list()  
         llista_errorsres=list()
```

```
for i in range(len(listpredi)):  
    valor=listy[i]-listpredi[i]  
    valorabs=math.fabs(valor)  
    valorrespecte=valorabs/listy[i]  
    llista_errors.append(valor)  
    llista_errorsabs.append(valorabs)  
    llista_errorsres.append(valorrespecte)
```

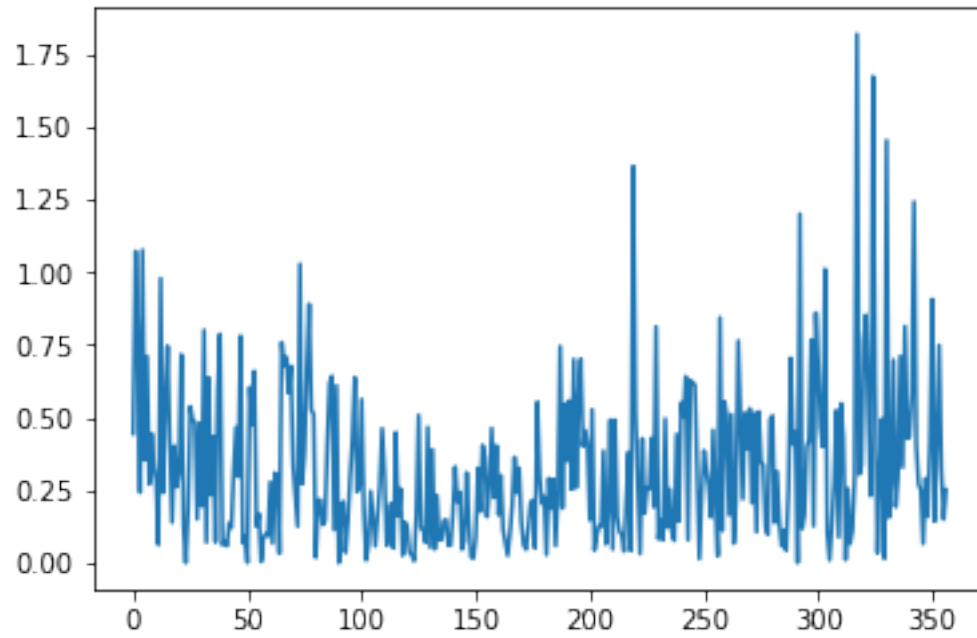
```
In [63]: plt.plot(llista_errors)
```

```
Out[63]: [<matplotlib.lines.Line2D at 0x210febb4ef0>]
```



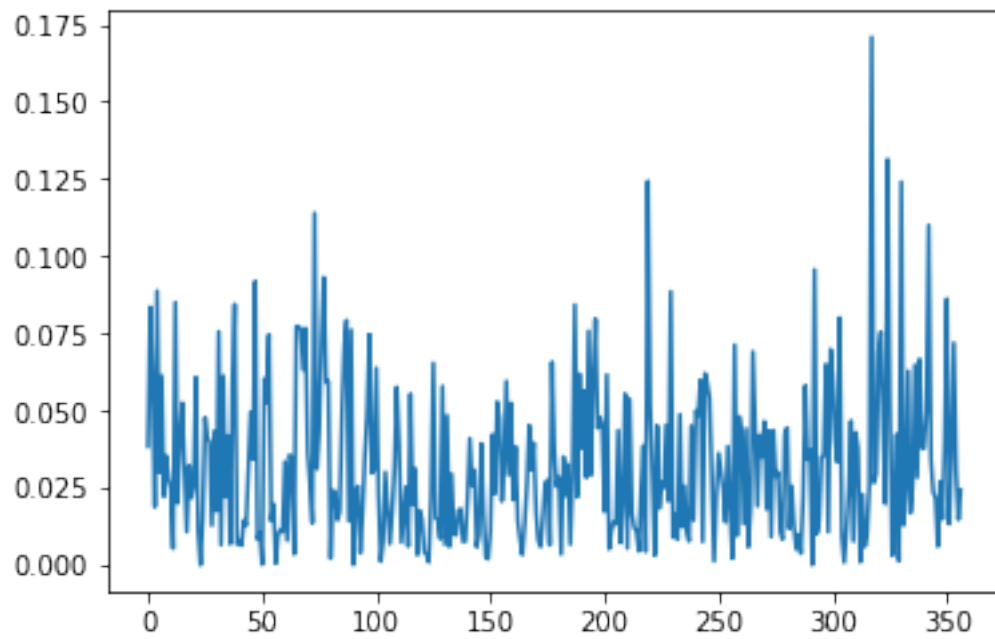
```
In [64]: plt.plot(llista_errorsabs)
```

```
Out[64]: [<matplotlib.lines.Line2D at 0x210fec15ba8>]
```



```
In [69]: plt.plot(llista_errorsres)
```

```
Out[69]: [<matplotlib.lines.Line2D at 0x210fecebf28>]
```



```
In [70]: sum(llista_errorsres)/(len(llista_errorsres))
```

```
Out[70]: 0.0327493809766566
```

```
In [ ]:
```