

M10

_Xarxa_walkforard_normalitzat_multivariate2tempmin_presio_14dies

December 21, 2019

1 Xarxa neuronal

```
In [3]: 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 [4]: daily=pd.read_csv('C:/Users/Laura/Desktop/Smart meters London/workspace R/Dades netes/1')
daily.head(5)
```

```
Out [4]:
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	date	apparentTemperatureMax	apparentTemperatureMin	sunsetTimeHour	\
0	2014-02-08	5.67	2.19	17	
1	2013-12-24	11.93	2.68	15	
2	2012-11-01	11.46	0.85	16	
3	2014-02-05	5.86	1.03	16	
4	2012-04-17	10.01	2.76	19	

	weekday	season	cloudCover	humidity	visibility	month	dewPoint	\
0	6	winter	0.47	0.77	11.20	2	3.99	

1	2	winter	0.40	0.81	10.86	12	5.42
2	4	autumn	0.44	0.85	12.54	11	5.06
3	3	winter	0.73	0.77	10.91	2	4.06
4	2	spring	0.60	0.87	11.86	4	5.74

	pressure	energy_sum
0	979.25	11.569300
1	979.52	11.981672
2	979.63	10.781689
3	982.20	11.415105
4	982.22	10.617443

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

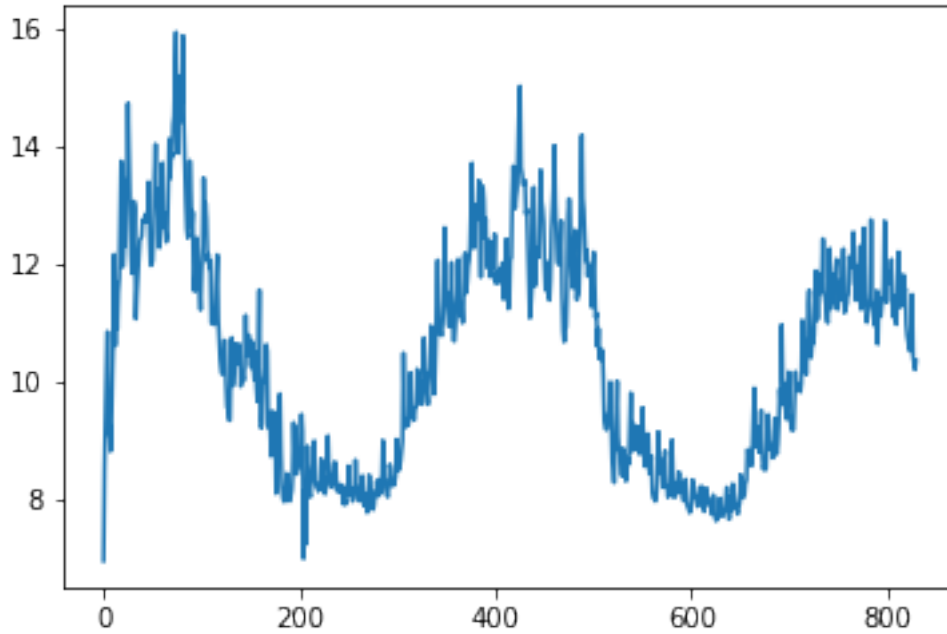
```
Out[5]:
```

	index	date	energy_sum	apparentTemperatureMax	\
0	735	2011-11-23	6.952692	10.36	
1	736	2011-11-24	8.536480	12.93	
2	682	2011-11-25	9.499781	13.03	
3	713	2011-11-26	10.267707	12.96	
4	609	2011-11-27	10.850805	13.54	

	apparentTemperatureMin	humidity	pressure
0	2.18	0.93	1027.12
1	7.01	0.89	1027.22
2	4.84	0.79	1024.47
3	4.69	0.81	1025.80
4	2.94	0.72	1021.11

```
In [18]: plt.plot(daily_dia.energy_sum )
```

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Out[18]: [<matplotlib.lines.Line2D at 0x1d48d92d710>]
```



```
In [6]: 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)
daily_dia['t-9']=daily_dia['energy_sum'].shift(9)
daily_dia['t-10']=daily_dia['energy_sum'].shift(10)
daily_dia['t-11']=daily_dia['energy_sum'].shift(11)
daily_dia['t-12']=daily_dia['energy_sum'].shift(12)
daily_dia['t-13']=daily_dia['energy_sum'].shift(13)
daily_dia['t-14']=daily_dia['energy_sum'].shift(14)

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['temp(t-9)']=daily_dia['apparentTemperatureMax'].shift(9)
daily_dia['temp(t-10)']=daily_dia['apparentTemperatureMax'].shift(10)
daily_dia['temp(t-11)']=daily_dia['apparentTemperatureMax'].shift(11)
```

```
daily_dia['temp(t-12)']=daily_dia['apparentTemperatureMax'].shift(12)
daily_dia['temp(t-13)']=daily_dia['apparentTemperatureMax'].shift(13)
daily_dia['temp(t-14)']=daily_dia['apparentTemperatureMax'].shift(14)
```

```
daily_dia['tempmin(t-1)']=daily_dia['apparentTemperatureMin'].shift(1)
daily_dia['tempmin(t-2)']=daily_dia['apparentTemperatureMin'].shift(2)
daily_dia['tempmin(t-3)']=daily_dia['apparentTemperatureMin'].shift(3)
daily_dia['tempmin(t-4)']=daily_dia['apparentTemperatureMin'].shift(4)
daily_dia['tempmin(t-5)']=daily_dia['apparentTemperatureMin'].shift(5)
daily_dia['tempmin(t-6)']=daily_dia['apparentTemperatureMin'].shift(6)
daily_dia['tempmin(t-7)']=daily_dia['apparentTemperatureMin'].shift(7)
daily_dia['tempmin(t-8)']=daily_dia['apparentTemperatureMin'].shift(8)
daily_dia['tempmin(t-9)']=daily_dia['apparentTemperatureMin'].shift(9)
daily_dia['tempmin(t-10)']=daily_dia['apparentTemperatureMin'].shift(10)
daily_dia['tempmin(t-11)']=daily_dia['apparentTemperatureMin'].shift(11)
daily_dia['tempmin(t-12)']=daily_dia['apparentTemperatureMin'].shift(12)
daily_dia['tempmin(t-13)']=daily_dia['apparentTemperatureMin'].shift(13)
daily_dia['tempmin(t-14)']=daily_dia['apparentTemperatureMin'].shift(14)
```

```
daily_dia['humidity(t-1)']=daily_dia['humidity'].shift(1)
daily_dia['humidity(t-2)']=daily_dia['humidity'].shift(2)
daily_dia['humidity(t-3)']=daily_dia['humidity'].shift(3)
daily_dia['humidity(t-4)']=daily_dia['humidity'].shift(4)
daily_dia['humidity(t-5)']=daily_dia['humidity'].shift(5)
daily_dia['humidity(t-6)']=daily_dia['humidity'].shift(6)
daily_dia['humidity(t-7)']=daily_dia['humidity'].shift(7)
daily_dia['humidity(t-8)']=daily_dia['humidity'].shift(8)
daily_dia['humidity(t-9)']=daily_dia['humidity'].shift(9)
daily_dia['humidity(t-10)']=daily_dia['humidity'].shift(10)
daily_dia['humidity(t-11)']=daily_dia['humidity'].shift(11)
daily_dia['humidity(t-12)']=daily_dia['humidity'].shift(12)
daily_dia['humidity(t-13)']=daily_dia['humidity'].shift(13)
daily_dia['humidity(t-14)']=daily_dia['humidity'].shift(14)
```

```
daily_dia['pres(t-1)']=daily_dia['pressure'].shift(1)
daily_dia['pres(t-2)']=daily_dia['pressure'].shift(2)
daily_dia['pres(t-3)']=daily_dia['pressure'].shift(3)
daily_dia['pres(t-4)']=daily_dia['pressure'].shift(4)
daily_dia['pres(t-5)']=daily_dia['pressure'].shift(5)
daily_dia['pres(t-6)']=daily_dia['pressure'].shift(6)
daily_dia['pres(t-7)']=daily_dia['pressure'].shift(7)
daily_dia['pres(t-8)']=daily_dia['pressure'].shift(8)
daily_dia['pres(t-9)']=daily_dia['pressure'].shift(9)
daily_dia['pres(t-10)']=daily_dia['pressure'].shift(10)
daily_dia['pres(t-11)']=daily_dia['pressure'].shift(11)
daily_dia['pres(t-12)']=daily_dia['pressure'].shift(12)
daily_dia['pres(t-13)']=daily_dia['pressure'].shift(13)
daily_dia['pres(t-14)']=daily_dia['pressure'].shift(14)
```

daily_dia

```
Out [6]:
```

	index	date	energy_sum	apparentTemperatureMax \
0	735	2011-11-23	6.952692	10.36
1	736	2011-11-24	8.536480	12.93
2	682	2011-11-25	9.499781	13.03
3	713	2011-11-26	10.267707	12.96
4	609	2011-11-27	10.850805	13.54
5	641	2011-11-28	9.103382	12.58
6	265	2011-11-29	9.274873	13.47
7	571	2011-11-30	8.813513	11.87
8	199	2011-12-01	9.227707	12.15
9	338	2011-12-02	10.145910	5.33
10	131	2011-12-03	10.780273	11.42
11	100	2011-12-04	12.163127	6.66
12	176	2011-12-05	10.609714	3.13
13	203	2011-12-06	11.673417	3.77
14	240	2011-12-07	10.889362	5.14
15	299	2011-12-08	11.525150	12.89
16	294	2011-12-09	11.759837	3.99
17	455	2011-12-10	12.633801	3.14
18	215	2011-12-11	13.749174	5.72
19	115	2011-12-12	11.951958	5.94
20	22	2011-12-13	11.957446	12.08
21	45	2011-12-14	12.392776	2.88
22	59	2011-12-15	12.307079	4.38
23	11	2011-12-16	13.376080	0.99
24	228	2011-12-17	13.511968	1.72
25	478	2011-12-18	14.732271	1.98
26	412	2011-12-19	13.774471	4.02
27	433	2011-12-20	12.709106	4.98
28	524	2011-12-21	12.148570	12.14
29	689	2011-12-22	11.839403	12.14
..
800	41	2014-01-29	11.800777	2.53
801	105	2014-01-30	11.685169	5.86
802	80	2014-01-31	11.857957	5.27
803	21	2014-02-01	11.710582	6.86
804	163	2014-02-02	12.078164	6.48
805	135	2014-02-03	11.280011	4.59
806	60	2014-02-04	11.095584	5.63
807	3	2014-02-05	11.415105	5.86
808	18	2014-02-06	11.445403	7.34
809	14	2014-02-07	10.972318	8.44
810	0	2014-02-08	11.569300	5.67
811	7	2014-02-09	12.202967	3.91

812	35	2014-02-10	11.264175	7.07
813	57	2014-02-11	11.452649	4.06
814	44	2014-02-12	11.679099	4.73
815	33	2014-02-13	11.285737	3.42
816	23	2014-02-14	11.816914	12.02
817	13	2014-02-15	11.490470	5.79
818	187	2014-02-16	11.582159	7.88
819	218	2014-02-17	10.979566	10.67
820	235	2014-02-18	10.781898	10.13
821	322	2014-02-19	10.674624	10.13
822	101	2014-02-20	10.573835	12.50
823	129	2014-02-21	10.518126	10.15
824	248	2014-02-22	10.776242	11.63
825	285	2014-02-23	11.480411	11.94
826	158	2014-02-24	10.411403	14.23
827	95	2014-02-25	10.294997	11.43
828	360	2014-02-26	10.202945	11.29
829	197	2014-02-27	10.356350	10.31

	apparentTemperatureMin	humidity	pressure	t-1	t-2 \
0	2.18	0.93	1027.12	NaN	NaN
1	7.01	0.89	1027.22	6.952692	NaN
2	4.84	0.79	1024.47	8.536480	6.952692
3	4.69	0.81	1025.80	9.499781	8.536480
4	2.94	0.72	1021.11	10.267707	9.499781
5	1.31	0.86	1022.80	10.850805	10.267707
6	3.39	0.82	1009.70	9.103382	10.850805
7	3.34	0.78	1019.43	9.274873	9.103382
8	5.29	0.82	1007.12	8.813513	9.274873
9	0.46	0.87	1012.12	9.227707	8.813513
10	4.71	0.79	1003.55	10.145910	9.227707
11	1.03	0.82	1001.15	10.780273	10.145910
12	-1.69	0.77	1006.01	12.163127	10.780273
13	-1.61	0.83	1007.32	10.609714	12.163127
14	0.94	0.68	1008.76	11.673417	10.609714
15	0.63	0.81	1010.84	10.889362	11.673417
16	-1.42	0.71	1010.60	11.525150	10.889362
17	-3.42	0.81	1015.58	11.759837	11.525150
18	0.11	0.88	1007.71	12.633801	11.759837
19	-0.64	0.84	1002.47	13.749174	12.633801
20	0.22	0.75	990.27	11.951958	13.749174
21	0.78	0.79	994.48	11.957446	11.951958
22	1.07	0.77	996.75	12.392776	11.957446
23	-2.65	0.88	988.10	12.307079	12.392776
24	-3.56	0.86	1008.46	13.376080	12.307079
25	-4.12	0.84	1016.37	13.511968	13.376080
26	-3.67	0.94	1014.39	14.732271	13.511968
27	1.68	0.81	1015.09	13.774471	14.732271

28	3.84	0.94	1017.91	12.709106	13.774471
29	5.37	0.87	1024.71	12.148570	12.709106
..
800	0.18	0.90	993.99	11.344805	11.753871
801	0.61	0.91	1001.76	11.800777	11.344805
802	0.29	0.91	998.51	11.685169	11.800777
803	1.10	0.76	990.08	11.857957	11.685169
804	3.21	0.72	1005.39	11.710582	11.857957
805	1.96	0.79	1003.89	12.078164	11.710582
806	1.12	0.75	996.87	11.280011	12.078164
807	1.03	0.77	982.20	11.095584	11.280011
808	1.96	0.82	989.90	11.415105	11.095584
809	-0.86	0.79	988.77	11.445403	11.415105
810	2.19	0.77	979.25	10.972318	11.445403
811	1.38	0.66	984.71	11.569300	10.972318
812	0.89	0.84	992.84	12.202967	11.569300
813	-0.57	0.76	996.66	11.264175	12.202967
814	-1.20	0.75	994.27	11.452649	11.264175
815	0.05	0.68	992.43	11.679099	11.452649
816	0.45	0.81	990.31	11.285737	11.679099
817	1.77	0.69	988.63	11.816914	11.285737
818	-1.03	0.76	1006.70	11.490470	11.816914
819	2.84	0.83	1007.80	11.582159	11.490470
820	3.83	0.87	1008.67	10.979566	11.582159
821	2.65	0.87	1011.57	10.781898	10.979566
822	3.95	0.84	1001.54	10.674624	10.781898
823	0.19	0.72	1003.42	10.573835	10.674624
824	1.59	0.71	1009.09	10.518126	10.573835
825	5.53	0.76	1010.37	10.776242	10.518126
826	5.52	0.74	1005.19	11.480411	10.776242
827	3.89	0.78	1000.65	10.411403	11.480411
828	1.67	0.73	1012.73	10.294997	10.411403
829	1.41	0.74	1007.02	10.202945	10.294997

	t-3	...	pres(t-5)	pres(t-6)	pres(t-7)	pres(t-8)	pres(t-9)	\
0	NaN	...	NaN	NaN	NaN	NaN	NaN	
1	NaN	...	NaN	NaN	NaN	NaN	NaN	
2	NaN	...	NaN	NaN	NaN	NaN	NaN	
3	6.952692	...	NaN	NaN	NaN	NaN	NaN	
4	8.536480	...	NaN	NaN	NaN	NaN	NaN	
5	9.499781	...	1027.12	NaN	NaN	NaN	NaN	
6	10.267707	...	1027.22	1027.12	NaN	NaN	NaN	
7	10.850805	...	1024.47	1027.22	1027.12	NaN	NaN	
8	9.103382	...	1025.80	1024.47	1027.22	1027.12	NaN	
9	9.274873	...	1021.11	1025.80	1024.47	1027.22	1027.12	
10	8.813513	...	1022.80	1021.11	1025.80	1024.47	1027.22	
11	9.227707	...	1009.70	1022.80	1021.11	1025.80	1024.47	
12	10.145910	...	1019.43	1009.70	1022.80	1021.11	1025.80	

13	10.780273	...	1007.12	1019.43	1009.70	1022.80	1021.11
14	12.163127	...	1012.12	1007.12	1019.43	1009.70	1022.80
15	10.609714	...	1003.55	1012.12	1007.12	1019.43	1009.70
16	11.673417	...	1001.15	1003.55	1012.12	1007.12	1019.43
17	10.889362	...	1006.01	1001.15	1003.55	1012.12	1007.12
18	11.525150	...	1007.32	1006.01	1001.15	1003.55	1012.12
19	11.759837	...	1008.76	1007.32	1006.01	1001.15	1003.55
20	12.633801	...	1010.84	1008.76	1007.32	1006.01	1001.15
21	13.749174	...	1010.60	1010.84	1008.76	1007.32	1006.01
22	11.951958	...	1015.58	1010.60	1010.84	1008.76	1007.32
23	11.957446	...	1007.71	1015.58	1010.60	1010.84	1008.76
24	12.392776	...	1002.47	1007.71	1015.58	1010.60	1010.84
25	12.307079	...	990.27	1002.47	1007.71	1015.58	1010.60
26	13.376080	...	994.48	990.27	1002.47	1007.71	1015.58
27	13.511968	...	996.75	994.48	990.27	1002.47	1007.71
28	14.732271	...	988.10	996.75	994.48	990.27	1002.47
29	13.774471	...	1008.46	988.10	996.75	994.48	990.27
..
800	12.729659	...	1013.91	1012.39	1006.71	1010.13	1007.71
801	11.753871	...	1012.46	1013.91	1012.39	1006.71	1010.13
802	11.344805	...	1002.10	1012.46	1013.91	1012.39	1006.71
803	11.800777	...	989.87	1002.10	1012.46	1013.91	1012.39
804	11.685169	...	985.33	989.87	1002.10	1012.46	1013.91
805	11.857957	...	993.99	985.33	989.87	1002.10	1012.46
806	11.710582	...	1001.76	993.99	985.33	989.87	1002.10
807	12.078164	...	998.51	1001.76	993.99	985.33	989.87
808	11.280011	...	990.08	998.51	1001.76	993.99	985.33
809	11.095584	...	1005.39	990.08	998.51	1001.76	993.99
810	11.415105	...	1003.89	1005.39	990.08	998.51	1001.76
811	11.445403	...	996.87	1003.89	1005.39	990.08	998.51
812	10.972318	...	982.20	996.87	1003.89	1005.39	990.08
813	11.569300	...	989.90	982.20	996.87	1003.89	1005.39
814	12.202967	...	988.77	989.90	982.20	996.87	1003.89
815	11.264175	...	979.25	988.77	989.90	982.20	996.87
816	11.452649	...	984.71	979.25	988.77	989.90	982.20
817	11.679099	...	992.84	984.71	979.25	988.77	989.90
818	11.285737	...	996.66	992.84	984.71	979.25	988.77
819	11.816914	...	994.27	996.66	992.84	984.71	979.25
820	11.490470	...	992.43	994.27	996.66	992.84	984.71
821	11.582159	...	990.31	992.43	994.27	996.66	992.84
822	10.979566	...	988.63	990.31	992.43	994.27	996.66
823	10.781898	...	1006.70	988.63	990.31	992.43	994.27
824	10.674624	...	1007.80	1006.70	988.63	990.31	992.43
825	10.573835	...	1008.67	1007.80	1006.70	988.63	990.31
826	10.518126	...	1011.57	1008.67	1007.80	1006.70	988.63
827	10.776242	...	1001.54	1011.57	1008.67	1007.80	1006.70
828	11.480411	...	1003.42	1001.54	1011.57	1008.67	1007.80
829	10.411403	...	1009.09	1003.42	1001.54	1011.57	1008.67

	pres(t-10)	pres(t-11)	pres(t-12)	pres(t-13)	pres(t-14)
0	NaN	NaN	NaN	NaN	NaN
1	NaN	NaN	NaN	NaN	NaN
2	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN
4	NaN	NaN	NaN	NaN	NaN
5	NaN	NaN	NaN	NaN	NaN
6	NaN	NaN	NaN	NaN	NaN
7	NaN	NaN	NaN	NaN	NaN
8	NaN	NaN	NaN	NaN	NaN
9	NaN	NaN	NaN	NaN	NaN
10	1027.12	NaN	NaN	NaN	NaN
11	1027.22	1027.12	NaN	NaN	NaN
12	1024.47	1027.22	1027.12	NaN	NaN
13	1025.80	1024.47	1027.22	1027.12	NaN
14	1021.11	1025.80	1024.47	1027.22	1027.12
15	1022.80	1021.11	1025.80	1024.47	1027.22
16	1009.70	1022.80	1021.11	1025.80	1024.47
17	1019.43	1009.70	1022.80	1021.11	1025.80
18	1007.12	1019.43	1009.70	1022.80	1021.11
19	1012.12	1007.12	1019.43	1009.70	1022.80
20	1003.55	1012.12	1007.12	1019.43	1009.70
21	1001.15	1003.55	1012.12	1007.12	1019.43
22	1006.01	1001.15	1003.55	1012.12	1007.12
23	1007.32	1006.01	1001.15	1003.55	1012.12
24	1008.76	1007.32	1006.01	1001.15	1003.55
25	1010.84	1008.76	1007.32	1006.01	1001.15
26	1010.60	1010.84	1008.76	1007.32	1006.01
27	1015.58	1010.60	1010.84	1008.76	1007.32
28	1007.71	1015.58	1010.60	1010.84	1008.76
29	1002.47	1007.71	1015.58	1010.60	1010.84
...
800	995.52	991.61	990.80	990.50	998.57
801	1007.71	995.52	991.61	990.80	990.50
802	1010.13	1007.71	995.52	991.61	990.80
803	1006.71	1010.13	1007.71	995.52	991.61
804	1012.39	1006.71	1010.13	1007.71	995.52
805	1013.91	1012.39	1006.71	1010.13	1007.71
806	1012.46	1013.91	1012.39	1006.71	1010.13
807	1002.10	1012.46	1013.91	1012.39	1006.71
808	989.87	1002.10	1012.46	1013.91	1012.39
809	985.33	989.87	1002.10	1012.46	1013.91
810	993.99	985.33	989.87	1002.10	1012.46
811	1001.76	993.99	985.33	989.87	1002.10
812	998.51	1001.76	993.99	985.33	989.87
813	990.08	998.51	1001.76	993.99	985.33
814	1005.39	990.08	998.51	1001.76	993.99

815	1003.89	1005.39	990.08	998.51	1001.76
816	996.87	1003.89	1005.39	990.08	998.51
817	982.20	996.87	1003.89	1005.39	990.08
818	989.90	982.20	996.87	1003.89	1005.39
819	988.77	989.90	982.20	996.87	1003.89
820	979.25	988.77	989.90	982.20	996.87
821	984.71	979.25	988.77	989.90	982.20
822	992.84	984.71	979.25	988.77	989.90
823	996.66	992.84	984.71	979.25	988.77
824	994.27	996.66	992.84	984.71	979.25
825	992.43	994.27	996.66	992.84	984.71
826	990.31	992.43	994.27	996.66	992.84
827	988.63	990.31	992.43	994.27	996.66
828	1006.70	988.63	990.31	992.43	994.27
829	1007.80	1006.70	988.63	990.31	992.43

[830 rows x 77 columns]

```
In [7]: #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.drop(['index', 'date', 'apparentTemperatureMax', 'apparentTemperature'])
daily_dia.head(5)
```

```
Out [7]:
```

	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	

	t-9	...	pres(t-5)	pres(t-6)	pres(t-7)	pres(t-8)	pres(t-9)	\
0	NaN	...	NaN	NaN	NaN	NaN	NaN	
1	NaN	...	NaN	NaN	NaN	NaN	NaN	
2	NaN	...	NaN	NaN	NaN	NaN	NaN	
3	NaN	...	NaN	NaN	NaN	NaN	NaN	
4	NaN	...	NaN	NaN	NaN	NaN	NaN	

	pres(t-10)	pres(t-11)	pres(t-12)	pres(t-13)	pres(t-14)
0	NaN	NaN	NaN	NaN	NaN
1	NaN	NaN	NaN	NaN	NaN
2	NaN	NaN	NaN	NaN	NaN
3	NaN	NaN	NaN	NaN	NaN
4	NaN	NaN	NaN	NaN	NaN

[5 rows x 71 columns]

```
In [8]: #Eliminem les 14 primeres files ja que contenen NaN (valors buits)
```

```
daily_dia=daily_dia.drop([0,1,2,3,4,5,6,7,8,9,10,11,12,13])
daily_dia.head(5)
```

```
Out [8]:
```

	energy_sum	t-1	t-2	t-3	t-4	t-5	\
14	10.889362	11.673417	10.609714	12.163127	10.780273	10.145910	
15	11.525150	10.889362	11.673417	10.609714	12.163127	10.780273	
16	11.759837	11.525150	10.889362	11.673417	10.609714	12.163127	
17	12.633801	11.759837	11.525150	10.889362	11.673417	10.609714	
18	13.749174	12.633801	11.759837	11.525150	10.889362	11.673417	

	t-6	t-7	t-8	t-9	...	pres(t-5)	pres(t-6)	\
14	9.227707	8.813513	9.274873	9.103382	...	1012.12	1007.12	
15	10.145910	9.227707	8.813513	9.274873	...	1003.55	1012.12	
16	10.780273	10.145910	9.227707	8.813513	...	1001.15	1003.55	
17	12.163127	10.780273	10.145910	9.227707	...	1006.01	1001.15	
18	10.609714	12.163127	10.780273	10.145910	...	1007.32	1006.01	

	pres(t-7)	pres(t-8)	pres(t-9)	pres(t-10)	pres(t-11)	pres(t-12)	\
14	1019.43	1009.70	1022.80	1021.11	1025.80	1024.47	
15	1007.12	1019.43	1009.70	1022.80	1021.11	1025.80	
16	1012.12	1007.12	1019.43	1009.70	1022.80	1021.11	
17	1003.55	1012.12	1007.12	1019.43	1009.70	1022.80	
18	1001.15	1003.55	1012.12	1007.12	1019.43	1009.70	

	pres(t-13)	pres(t-14)
14	1027.22	1027.12
15	1024.47	1027.22
16	1025.80	1024.47
17	1021.11	1025.80
18	1022.80	1021.11

[5 rows x 71 columns]

```
In [7]: len(daily_dia)
```

```
Out [7]: 816
```

```
In [9]: #normalitzem
```

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

```
In [10]: #Seleccionem dades per test i train
```

```
y_daily=daily_dia_norm[:,0]
X_daily=daily_dia_norm[:,1:72]
```

```
#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], 14,5))

In [26]: # definim model
import tensorflow as tf
model =Sequential()
model.add(LSTM(50, activation='relu', input_shape=(14, 5)))
model.add(Dense(1))
model.compile(optimizer='adam', loss='mse', metrics=['accuracy'])

In [27]: 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_predittrain=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_predittrain.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
```

```
In [28]: #Dividim la suma de scores de test entre el nombre de prediccions per obtenir la mitja  
sumScores/(length-n_train)
```

```
Out [28]: 0.03389682158314454
```

```
In [29]: llista_scores
```

```
Out [29]: [0.02999627663921789,  
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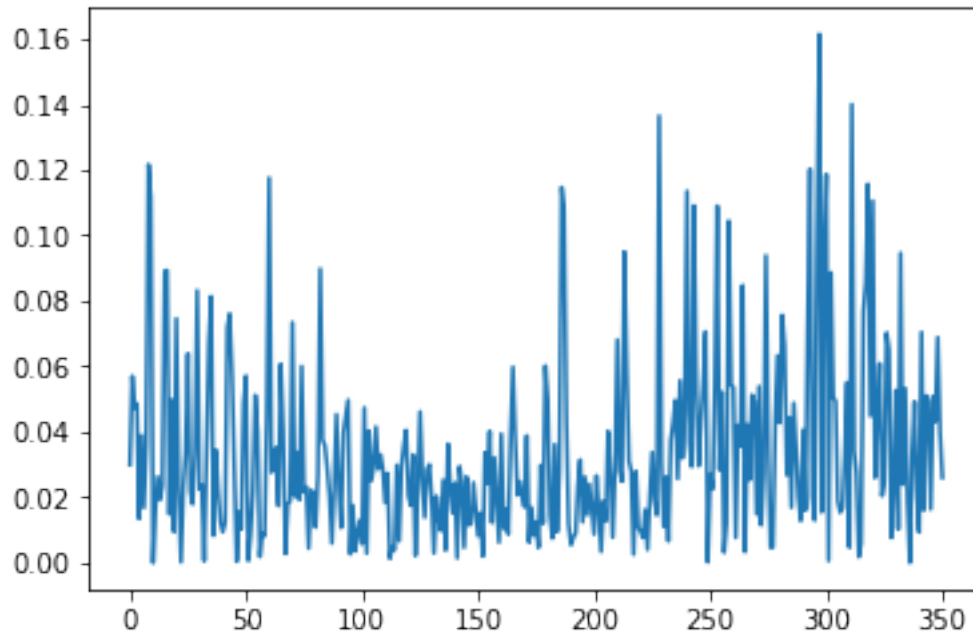
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```

```
In [30]: plt.plot(llista_scores)
```

```
Out[30]: [<matplotlib.lines.Line2D at 0x21bc00a9a90>]
```



```
In [31]: predis=list()
```

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

```
predis=np.reshape(predis, (351) )
```

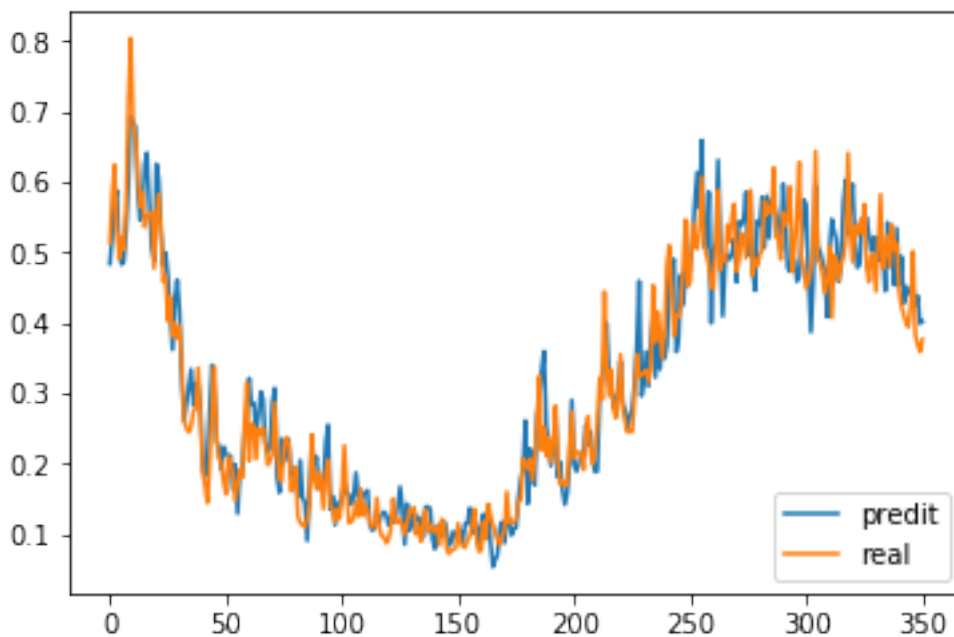
```
predis
```

```
Out[31]: array([0.4840652 , 0.52356768, 0.57689756, 0.58793586, 0.50484443,
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```

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0.48551872, 0.51864594, 0.57278627, 0.61401409, 0.56261849,
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0.50132203, 0.51499629, 0.63117224, 0.51043153, 0.40955055,
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0.5673883 , 0.46207535, 0.38702446, 0.45760429, 0.59448099,
0.50768769, 0.50074291, 0.48552531, 0.47568303, 0.40846264,
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```
0.59711504, 0.51151478, 0.47859955, 0.48303425, 0.54792476,
0.5438568 , 0.54960155, 0.52391565, 0.50213158, 0.52185225,
0.49765229, 0.52192467, 0.48783928, 0.50152427, 0.44532877,
0.54269534, 0.48004264, 0.50618845, 0.45356494, 0.53447115,
0.45516458, 0.4941124 , 0.42782843, 0.4514764 , 0.44352078,
0.43946874, 0.45082676, 0.42542899, 0.43814427, 0.39908499,
0.40217802])
```

```
In [32]: ##Mostrem
plt.plot(predis, label="predit")
plt.plot(y_daily[n_train:lenght], label="real")
plt.legend(loc="lower right")
plt.show()
```



```
In [33]: #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','t-9','t-10','t-11']]
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>
if sys.path[0] == '':
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>
del sys.path[0]

```
Out[33]:
```

	predi	y	t-2	t-3	t-4	t-5	\
479	0.484065	0.514061	12.119938	12.852295	13.106773	12.823073	
480	0.523568	0.580609	11.786082	12.119938	12.852295	13.106773	
481	0.576898	0.624326	11.590859	11.786082	12.119938	12.852295	
482	0.587936	0.539280	12.186487	11.590859	11.786082	12.119938	
483	0.504844	0.491355	12.577783	12.186487	11.590859	11.786082	
484	0.483200	0.522145	11.816573	12.577783	12.186487	11.590859	
485	0.487606	0.504442	11.387627	11.816573	12.577783	12.186487	
486	0.531441	0.567725	11.663214	11.387627	11.816573	12.577783	
487	0.597795	0.719460	11.504756	11.663214	11.387627	11.816573	
488	0.693004	0.804631	12.071173	11.504756	11.663214	11.387627	
489	0.684740	0.684716	13.429271	12.071173	11.504756	11.663214	
490	0.678917	0.662177	14.191591	13.429271	12.071173	11.504756	
491	0.588885	0.615194	13.118295	14.191591	13.429271	12.071173	
492	0.546107	0.565466	12.916559	13.118295	14.191591	13.429271	
493	0.561951	0.585646	12.496044	12.916559	13.118295	14.191591	
494	0.625583	0.536523	12.050954	12.496044	12.916559	13.118295	
495	0.641593	0.552256	12.231576	12.050954	12.496044	12.916559	
496	0.567145	0.552256	11.791904	12.231576	12.050954	12.496044	
497	0.508001	0.557809	11.932721	11.791904	12.231576	12.050954	
498	0.487344	0.477794	11.932721	11.932721	11.791904	12.231576	
499	0.625780	0.551195	11.982423	11.932721	11.932721	11.791904	
500	0.604753	0.582339	11.266252	11.982423	11.932721	11.932721	
501	0.529538	0.529772	11.923226	11.266252	11.982423	11.932721	
502	0.481994	0.458904	12.201972	11.923226	11.266252	11.982423	
503	0.499510	0.465733	11.731479	12.201972	11.923226	11.266252	
504	0.466515	0.402622	11.097177	11.731479	12.201972	11.923226	

505	0.414933	0.436918	11.158295	11.097177	11.731479	12.201972
506	0.362091	0.380048	10.593420	11.158295	11.097177	11.731479
507	0.432896	0.398860	10.900388	10.593420	11.158295	11.097177
508	0.461112	0.377916	10.391372	10.900388	10.593420	11.158295
..
800	0.511515	0.537515	11.753871	12.729659	11.620778	11.409880
801	0.478600	0.524598	11.344805	11.753871	12.729659	11.620778
802	0.483034	0.543903	11.800777	11.344805	11.753871	12.729659
803	0.547925	0.527438	11.685169	11.800777	11.344805	11.753871
804	0.543857	0.568506	11.857957	11.685169	11.800777	11.344805
805	0.549602	0.479332	11.710582	11.857957	11.685169	11.800777
806	0.523916	0.458726	12.078164	11.710582	11.857957	11.685169
807	0.502132	0.494425	11.280011	12.078164	11.710582	11.857957
808	0.521852	0.497810	11.095584	11.280011	12.078164	11.710582
809	0.497652	0.444954	11.415105	11.095584	11.280011	12.078164
810	0.521925	0.511653	11.445403	11.415105	11.095584	11.280011
811	0.487839	0.582450	10.972318	11.445403	11.415105	11.095584
812	0.501524	0.477562	11.569300	10.972318	11.445403	11.415105
813	0.445329	0.498620	12.202967	11.569300	10.972318	11.445403
814	0.542695	0.523920	11.264175	12.202967	11.569300	10.972318
815	0.480043	0.479971	11.452649	11.264175	12.202967	11.569300
816	0.506188	0.539318	11.679099	11.452649	11.264175	12.202967
817	0.453565	0.502845	11.285737	11.679099	11.452649	11.264175
818	0.534471	0.513089	11.816914	11.285737	11.679099	11.452649
819	0.455165	0.445764	11.490470	11.816914	11.285737	11.679099
820	0.494112	0.423680	11.582159	11.490470	11.816914	11.285737
821	0.427828	0.411694	10.979566	11.582159	11.490470	11.816914
822	0.451476	0.400434	10.781898	10.979566	11.582159	11.490470
823	0.443521	0.394209	10.674624	10.781898	10.979566	11.582159
824	0.439469	0.423048	10.573835	10.674624	10.781898	10.979566
825	0.450827	0.501722	10.518126	10.573835	10.674624	10.781898
826	0.425429	0.382286	10.776242	10.518126	10.573835	10.674624
827	0.438144	0.369280	11.480411	10.776242	10.518126	10.573835
828	0.399085	0.358995	10.411403	11.480411	10.776242	10.518126
829	0.402178	0.376135	10.294997	10.411403	11.480411	10.776242

	t-6	t-7	t-8	t-9	...	pres(t-5)	pres(t-6)	\
479	11.559878	10.930170	10.889469	10.675248	...	1005.94	1002.41	
480	12.823073	11.559878	10.930170	10.889469	...	1007.26	1005.94	
481	13.106773	12.823073	11.559878	10.930170	...	1008.83	1007.26	
482	12.852295	13.106773	12.823073	11.559878	...	1010.59	1008.83	
483	12.119938	12.852295	13.106773	12.823073	...	1015.57	1010.59	
484	11.786082	12.119938	12.852295	13.106773	...	1007.60	1015.57	
485	11.590859	11.786082	12.119938	12.852295	...	996.08	1007.60	
486	12.186487	11.590859	11.786082	12.119938	...	990.34	996.08	
487	12.577783	12.186487	11.590859	11.786082	...	991.02	990.34	
488	11.816573	12.577783	12.186487	11.590859	...	997.51	991.02	
489	11.387627	11.816573	12.577783	12.186487	...	1006.74	997.51	

490	11.663214	11.387627	11.816573	12.577783	...	1015.96	1006.74
491	11.504756	11.663214	11.387627	11.816573	...	1009.80	1015.96
492	12.071173	11.504756	11.663214	11.387627	...	1009.85	1009.80
493	13.429271	12.071173	11.504756	11.663214	...	1012.85	1009.85
494	14.191591	13.429271	12.071173	11.504756	...	1014.28	1012.85
495	13.118295	14.191591	13.429271	12.071173	...	1013.39	1014.28
496	12.916559	13.118295	14.191591	13.429271	...	1011.63	1013.39
497	12.496044	12.916559	13.118295	14.191591	...	1012.30	1011.63
498	12.050954	12.496044	12.916559	13.118295	...	1010.88	1012.30
499	12.231576	12.050954	12.496044	12.916559	...	1011.38	1010.88
500	11.791904	12.231576	12.050954	12.496044	...	1011.52	1011.38
501	11.932721	11.791904	12.231576	12.050954	...	1014.56	1011.52
502	11.932721	11.932721	11.791904	12.231576	...	1014.63	1014.56
503	11.982423	11.932721	11.932721	11.791904	...	1017.40	1014.63
504	11.266252	11.982423	11.932721	11.932721	...	1014.22	1017.40
505	11.923226	11.266252	11.982423	11.932721	...	1015.19	1014.22
506	12.201972	11.923226	11.266252	11.982423	...	1023.34	1015.19
507	11.731479	12.201972	11.923226	11.266252	...	1020.37	1023.34
508	11.097177	11.731479	12.201972	11.923226	...	1005.84	1020.37
..
800	11.300414	11.109560	11.370601	11.430883	...	1013.91	1012.39
801	11.409880	11.300414	11.109560	11.370601	...	1012.46	1013.91
802	11.620778	11.409880	11.300414	11.109560	...	1002.10	1012.46
803	12.729659	11.620778	11.409880	11.300414	...	989.87	1002.10
804	11.753871	12.729659	11.620778	11.409880	...	985.33	989.87
805	11.344805	11.753871	12.729659	11.620778	...	993.99	985.33
806	11.800777	11.344805	11.753871	12.729659	...	1001.76	993.99
807	11.685169	11.800777	11.344805	11.753871	...	998.51	1001.76
808	11.857957	11.685169	11.800777	11.344805	...	990.08	998.51
809	11.710582	11.857957	11.685169	11.800777	...	1005.39	990.08
810	12.078164	11.710582	11.857957	11.685169	...	1003.89	1005.39
811	11.280011	12.078164	11.710582	11.857957	...	996.87	1003.89
812	11.095584	11.280011	12.078164	11.710582	...	982.20	996.87
813	11.415105	11.095584	11.280011	12.078164	...	989.90	982.20
814	11.445403	11.415105	11.095584	11.280011	...	988.77	989.90
815	10.972318	11.445403	11.415105	11.095584	...	979.25	988.77
816	11.569300	10.972318	11.445403	11.415105	...	984.71	979.25
817	12.202967	11.569300	10.972318	11.445403	...	992.84	984.71
818	11.264175	12.202967	11.569300	10.972318	...	996.66	992.84
819	11.452649	11.264175	12.202967	11.569300	...	994.27	996.66
820	11.679099	11.452649	11.264175	12.202967	...	992.43	994.27
821	11.285737	11.679099	11.452649	11.264175	...	990.31	992.43
822	11.816914	11.285737	11.679099	11.452649	...	988.63	990.31
823	11.490470	11.816914	11.285737	11.679099	...	1006.70	988.63
824	11.582159	11.490470	11.816914	11.285737	...	1007.80	1006.70
825	10.979566	11.582159	11.490470	11.816914	...	1008.67	1007.80
826	10.781898	10.979566	11.582159	11.490470	...	1011.57	1008.67
827	10.674624	10.781898	10.979566	11.582159	...	1001.54	1011.57

828	10.573835	10.674624	10.781898	10.979566	...	1003.42	1001.54
829	10.518126	10.573835	10.674624	10.781898	...	1009.09	1003.42

	pres(t-7)	pres(t-8)	pres(t-9)	pres(t-10)	pres(t-11)	pres(t-12)	\
479	997.56	996.96	1000.69	1004.57	1013.74	1023.52	
480	1002.41	997.56	996.96	1000.69	1004.57	1013.74	
481	1005.94	1002.41	997.56	996.96	1000.69	1004.57	
482	1007.26	1005.94	1002.41	997.56	996.96	1000.69	
483	1008.83	1007.26	1005.94	1002.41	997.56	996.96	
484	1010.59	1008.83	1007.26	1005.94	1002.41	997.56	
485	1015.57	1010.59	1008.83	1007.26	1005.94	1002.41	
486	1007.60	1015.57	1010.59	1008.83	1007.26	1005.94	
487	996.08	1007.60	1015.57	1010.59	1008.83	1007.26	
488	990.34	996.08	1007.60	1015.57	1010.59	1008.83	
489	991.02	990.34	996.08	1007.60	1015.57	1010.59	
490	997.51	991.02	990.34	996.08	1007.60	1015.57	
491	1006.74	997.51	991.02	990.34	996.08	1007.60	
492	1015.96	1006.74	997.51	991.02	990.34	996.08	
493	1009.80	1015.96	1006.74	997.51	991.02	990.34	
494	1009.85	1009.80	1015.96	1006.74	997.51	991.02	
495	1012.85	1009.85	1009.80	1015.96	1006.74	997.51	
496	1014.28	1012.85	1009.85	1009.80	1015.96	1006.74	
497	1013.39	1014.28	1012.85	1009.85	1009.80	1015.96	
498	1011.63	1013.39	1014.28	1012.85	1009.85	1009.80	
499	1012.30	1011.63	1013.39	1014.28	1012.85	1009.85	
500	1010.88	1012.30	1011.63	1013.39	1014.28	1012.85	
501	1011.38	1010.88	1012.30	1011.63	1013.39	1014.28	
502	1011.52	1011.38	1010.88	1012.30	1011.63	1013.39	
503	1014.56	1011.52	1011.38	1010.88	1012.30	1011.63	
504	1014.63	1014.56	1011.52	1011.38	1010.88	1012.30	
505	1017.40	1014.63	1014.56	1011.52	1011.38	1010.88	
506	1014.22	1017.40	1014.63	1014.56	1011.52	1011.38	
507	1015.19	1014.22	1017.40	1014.63	1014.56	1011.52	
508	1023.34	1015.19	1014.22	1017.40	1014.63	1014.56	
..	
800	1006.71	1010.13	1007.71	995.52	991.61	990.80	
801	1012.39	1006.71	1010.13	1007.71	995.52	991.61	
802	1013.91	1012.39	1006.71	1010.13	1007.71	995.52	
803	1012.46	1013.91	1012.39	1006.71	1010.13	1007.71	
804	1002.10	1012.46	1013.91	1012.39	1006.71	1010.13	
805	989.87	1002.10	1012.46	1013.91	1012.39	1006.71	
806	985.33	989.87	1002.10	1012.46	1013.91	1012.39	
807	993.99	985.33	989.87	1002.10	1012.46	1013.91	
808	1001.76	993.99	985.33	989.87	1002.10	1012.46	
809	998.51	1001.76	993.99	985.33	989.87	1002.10	
810	990.08	998.51	1001.76	993.99	985.33	989.87	
811	1005.39	990.08	998.51	1001.76	993.99	985.33	
812	1003.89	1005.39	990.08	998.51	1001.76	993.99	

813	996.87	1003.89	1005.39	990.08	998.51	1001.76
814	982.20	996.87	1003.89	1005.39	990.08	998.51
815	989.90	982.20	996.87	1003.89	1005.39	990.08
816	988.77	989.90	982.20	996.87	1003.89	1005.39
817	979.25	988.77	989.90	982.20	996.87	1003.89
818	984.71	979.25	988.77	989.90	982.20	996.87
819	992.84	984.71	979.25	988.77	989.90	982.20
820	996.66	992.84	984.71	979.25	988.77	989.90
821	994.27	996.66	992.84	984.71	979.25	988.77
822	992.43	994.27	996.66	992.84	984.71	979.25
823	990.31	992.43	994.27	996.66	992.84	984.71
824	988.63	990.31	992.43	994.27	996.66	992.84
825	1006.70	988.63	990.31	992.43	994.27	996.66
826	1007.80	1006.70	988.63	990.31	992.43	994.27
827	1008.67	1007.80	1006.70	988.63	990.31	992.43
828	1011.57	1008.67	1007.80	1006.70	988.63	990.31
829	1001.54	1011.57	1008.67	1007.80	1006.70	988.63

	pres(t-13)	pres(t-14)
479	1027.67	1029.18
480	1023.52	1027.67
481	1013.74	1023.52
482	1004.57	1013.74
483	1000.69	1004.57
484	996.96	1000.69
485	997.56	996.96
486	1002.41	997.56
487	1005.94	1002.41
488	1007.26	1005.94
489	1008.83	1007.26
490	1010.59	1008.83
491	1015.57	1010.59
492	1007.60	1015.57
493	996.08	1007.60
494	990.34	996.08
495	991.02	990.34
496	997.51	991.02
497	1006.74	997.51
498	1015.96	1006.74
499	1009.80	1015.96
500	1009.85	1009.80
501	1012.85	1009.85
502	1014.28	1012.85
503	1013.39	1014.28
504	1011.63	1013.39
505	1012.30	1011.63
506	1010.88	1012.30
507	1011.38	1010.88

508	1011.52	1011.38
...
800	990.50	998.57
801	990.80	990.50
802	991.61	990.80
803	995.52	991.61
804	1007.71	995.52
805	1010.13	1007.71
806	1006.71	1010.13
807	1012.39	1006.71
808	1013.91	1012.39
809	1012.46	1013.91
810	1002.10	1012.46
811	989.87	1002.10
812	985.33	989.87
813	993.99	985.33
814	1001.76	993.99
815	998.51	1001.76
816	990.08	998.51
817	1005.39	990.08
818	1003.89	1005.39
819	996.87	1003.89
820	982.20	996.87
821	989.90	982.20
822	988.77	989.90
823	979.25	988.77
824	984.71	979.25
825	992.84	984.71
826	996.66	992.84
827	994.27	996.66
828	992.43	994.27
829	990.31	992.43

[351 rows x 71 columns]

In [34]: *# Convert predictions back to normal values*

```

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

```

#Les variables en posició 0 i 1 són predicció i y respectivament

```

[[1.13223783e+01 1.15908592e+01 1.15468930e+02 ... 6.40997284e+04
 6.43556589e+04 6.44487806e+04]
[1.16759441e+01 1.21864869e+01 1.12480758e+02 ... 6.34965958e+04
 6.40997284e+04 6.43556589e+04]

```

```

[1.21532717e+01 1.25777826e+01 1.10733424e+02 ... 6.29310819e+04
 6.34965958e+04 6.40997284e+04]
...
[1.09113643e+01 1.02949966e+01 1.09744859e+02 ... 6.21824081e+04
 6.22958809e+04 6.24432722e+04]
[1.05617653e+01 1.02029453e+01 1.00176736e+02 ... 6.20516677e+04
 6.21824081e+04 6.22958809e+04]
[1.05894494e+01 1.03563499e+01 9.91348430e+01 ... 6.19480621e+04
 6.20516677e+04 6.21824081e+04]]
11.3223783428646
11.590859170709699

```

In [35]: *#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 [35]: [11.590859170709699,
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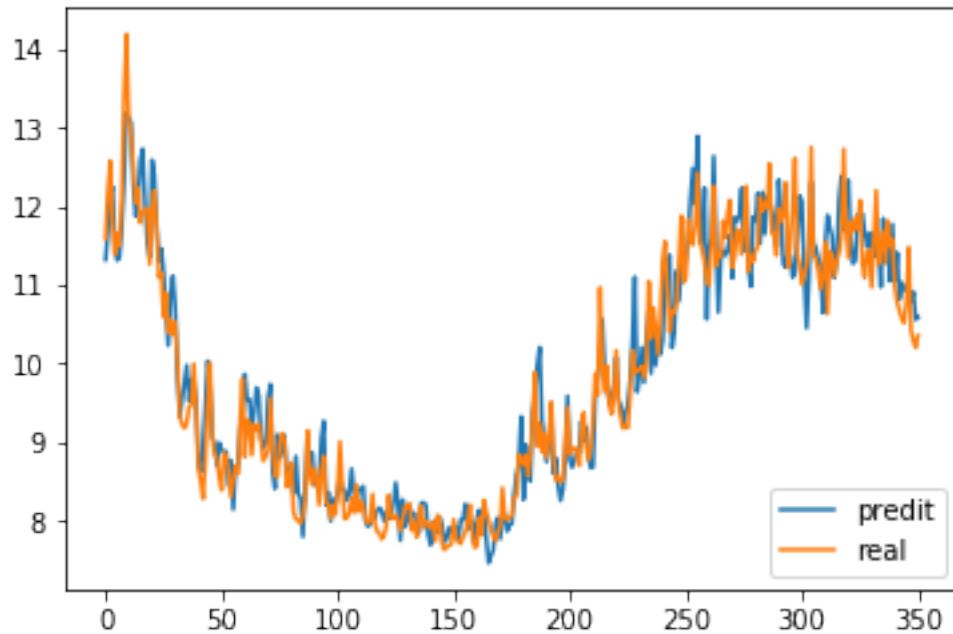
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```

```
In [36]: ##Mostrem  
plt.plot(listpredi, label="predict")  
plt.plot(listy, label="real")  
plt.legend(loc="lower right")  
plt.show()
```



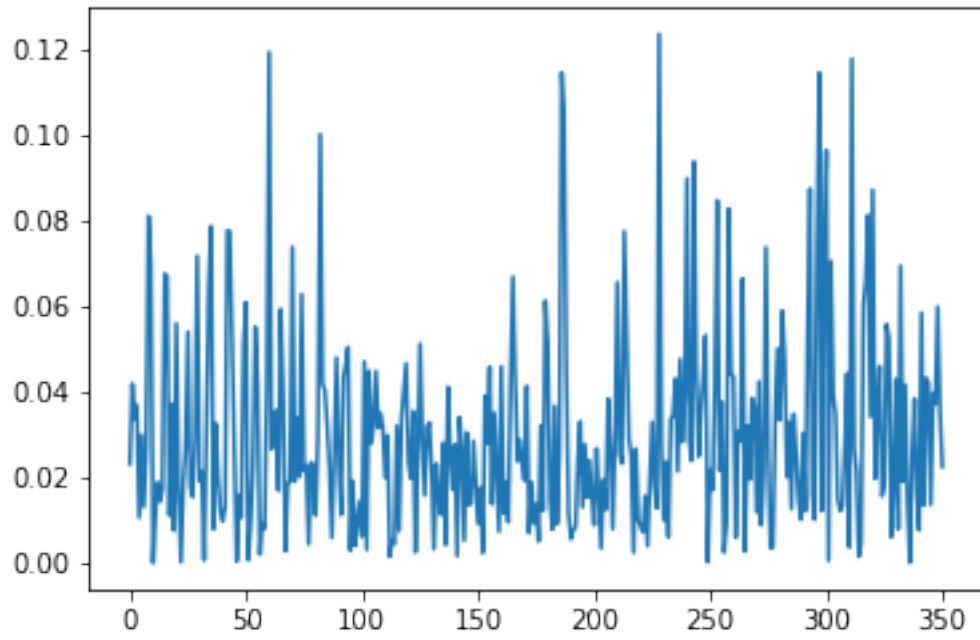
```
In [37]: llista_errors=list()
         llista_errorsabs=list()
         llista_errorsres=list()

         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)

         plt.plot(llista_errorsres)

         error_mitja=sum(llista_errorsres)/(len(llista_errorsres))*100
         print("L'error mitjà és de {} % ".format(error_mitja))
```

L'error mitjà és de 3.024031977841125 %



In []: