M24 _Xarxa_walkforard_normalitzat_multivariate2tempmin_weekday_14d walkforwardaugment

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/
        daily.head(5)
```

```
Out[2]:          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
```

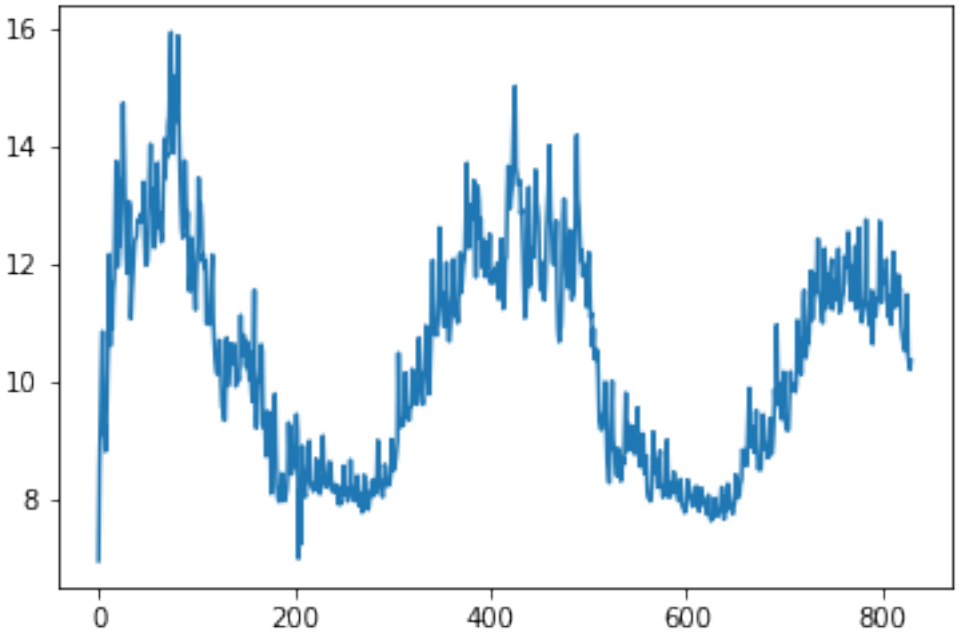
```
In [3]: #Ens quedem amb date i energy_sum, ordenem valors per data i resetejem index
        daily_dia=daily[['date','energy_sum','apparentTemperatureMax','apparentTemperatureMin'
        daily_dia.head(5)
```

```
Out[3]:    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  weekday
0                    2.18      0.93        3
1                    7.01      0.89        4
2                    4.84      0.79        5
3                    4.69      0.81        6
4                    2.94      0.72        7
```

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

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



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

```
daily_dia
```

Out[4]:

	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	weekday	t-1	t-2	\
0	2.18	0.93	3	NaN	NaN	
1	7.01	0.89	4	6.952692	NaN	
2	4.84	0.79	5	8.536480	6.952692	
3	4.69	0.81	6	9.499781	8.536480	
4	2.94	0.72	7	10.267707	9.499781	
5	1.31	0.86	1	10.850805	10.267707	
6	3.39	0.82	2	9.103382	10.850805	
7	3.34	0.78	3	9.274873	9.103382	
8	5.29	0.82	4	8.813513	9.274873	
9	0.46	0.87	5	9.227707	8.813513	
10	4.71	0.79	6	10.145910	9.227707	
11	1.03	0.82	7	10.780273	10.145910	
12	-1.69	0.77	1	12.163127	10.780273	
13	-1.61	0.83	2	10.609714	12.163127	
14	0.94	0.68	3	11.673417	10.609714	
15	0.63	0.81	4	10.889362	11.673417	
16	-1.42	0.71	5	11.525150	10.889362	
17	-3.42	0.81	6	11.759837	11.525150	
18	0.11	0.88	7	12.633801	11.759837	
19	-0.64	0.84	1	13.749174	12.633801	
20	0.22	0.75	2	11.951958	13.749174	
21	0.78	0.79	3	11.957446	11.951958	
22	1.07	0.77	4	12.392776	11.957446	
23	-2.65	0.88	5	12.307079	12.392776	
24	-3.56	0.86	6	13.376080	12.307079	
25	-4.12	0.84	7	13.511968	13.376080	
26	-3.67	0.94	1	14.732271	13.511968	
27	1.68	0.81	2	13.774471	14.732271	

28	3.84	0.94	3	12.709106	13.774471
29	5.37	0.87	4	12.148570	12.709106
..	...	...	...	...	...
800	0.18	0.90	3	11.344805	11.753871
801	0.61	0.91	4	11.800777	11.344805
802	0.29	0.91	5	11.685169	11.800777
803	1.10	0.76	6	11.857957	11.685169
804	3.21	0.72	7	11.710582	11.857957
805	1.96	0.79	1	12.078164	11.710582
806	1.12	0.75	2	11.280011	12.078164
807	1.03	0.77	3	11.095584	11.280011
808	1.96	0.82	4	11.415105	11.095584
809	-0.86	0.79	5	11.445403	11.415105
810	2.19	0.77	6	10.972318	11.445403
811	1.38	0.66	7	11.569300	10.972318
812	0.89	0.84	1	12.202967	11.569300
813	-0.57	0.76	2	11.264175	12.202967
814	-1.20	0.75	3	11.452649	11.264175
815	0.05	0.68	4	11.679099	11.452649
816	0.45	0.81	5	11.285737	11.679099
817	1.77	0.69	6	11.816914	11.285737
818	-1.03	0.76	7	11.490470	11.816914
819	2.84	0.83	1	11.582159	11.490470
820	3.83	0.87	2	10.979566	11.582159
821	2.65	0.87	3	10.781898	10.979566
822	3.95	0.84	4	10.674624	10.781898
823	0.19	0.72	5	10.573835	10.674624
824	1.59	0.71	6	10.518126	10.573835
825	5.53	0.76	7	10.776242	10.518126
826	5.52	0.74	1	11.480411	10.776242
827	3.89	0.78	2	10.411403	11.480411
828	1.67	0.73	3	10.294997	10.411403
829	1.41	0.74	4	10.202945	10.294997

	t-3	...	weekday(t-5)	weekday(t-6)	weekday(t-7)	weekday(t-8)	\
0	NaN	...	NaN	NaN	NaN	NaN	
1	NaN	...	NaN	NaN	NaN	NaN	
2	NaN	...	NaN	NaN	NaN	NaN	
3	6.952692	...	NaN	NaN	NaN	NaN	
4	8.536480	...	NaN	NaN	NaN	NaN	
5	9.499781	...	3.0	NaN	NaN	NaN	
6	10.267707	...	4.0	3.0	NaN	NaN	
7	10.850805	...	5.0	4.0	3.0	NaN	
8	9.103382	...	6.0	5.0	4.0	3.0	
9	9.274873	...	7.0	6.0	5.0	4.0	
10	8.813513	...	1.0	7.0	6.0	5.0	
11	9.227707	...	2.0	1.0	7.0	6.0	
12	10.145910	...	3.0	2.0	1.0	7.0	

13	10.780273	...	4.0	3.0	2.0	1.0
14	12.163127	...	5.0	4.0	3.0	2.0
15	10.609714	...	6.0	5.0	4.0	3.0
16	11.673417	...	7.0	6.0	5.0	4.0
17	10.889362	...	1.0	7.0	6.0	5.0
18	11.525150	...	2.0	1.0	7.0	6.0
19	11.759837	...	3.0	2.0	1.0	7.0
20	12.633801	...	4.0	3.0	2.0	1.0
21	13.749174	...	5.0	4.0	3.0	2.0
22	11.951958	...	6.0	5.0	4.0	3.0
23	11.957446	...	7.0	6.0	5.0	4.0
24	12.392776	...	1.0	7.0	6.0	5.0
25	12.307079	...	2.0	1.0	7.0	6.0
26	13.376080	...	3.0	2.0	1.0	7.0
27	13.511968	...	4.0	3.0	2.0	1.0
28	14.732271	...	5.0	4.0	3.0	2.0
29	13.774471	...	6.0	5.0	4.0	3.0
..	...	...	...	...	...	...
800	12.729659	...	5.0	4.0	3.0	2.0
801	11.753871	...	6.0	5.0	4.0	3.0
802	11.344805	...	7.0	6.0	5.0	4.0
803	11.800777	...	1.0	7.0	6.0	5.0
804	11.685169	...	2.0	1.0	7.0	6.0
805	11.857957	...	3.0	2.0	1.0	7.0
806	11.710582	...	4.0	3.0	2.0	1.0
807	12.078164	...	5.0	4.0	3.0	2.0
808	11.280011	...	6.0	5.0	4.0	3.0
809	11.095584	...	7.0	6.0	5.0	4.0
810	11.415105	...	1.0	7.0	6.0	5.0
811	11.445403	...	2.0	1.0	7.0	6.0
812	10.972318	...	3.0	2.0	1.0	7.0
813	11.569300	...	4.0	3.0	2.0	1.0
814	12.202967	...	5.0	4.0	3.0	2.0
815	11.264175	...	6.0	5.0	4.0	3.0
816	11.452649	...	7.0	6.0	5.0	4.0
817	11.679099	...	1.0	7.0	6.0	5.0
818	11.285737	...	2.0	1.0	7.0	6.0
819	11.816914	...	3.0	2.0	1.0	7.0
820	11.490470	...	4.0	3.0	2.0	1.0
821	11.582159	...	5.0	4.0	3.0	2.0
822	10.979566	...	6.0	5.0	4.0	3.0
823	10.781898	...	7.0	6.0	5.0	4.0
824	10.674624	...	1.0	7.0	6.0	5.0
825	10.573835	...	2.0	1.0	7.0	6.0
826	10.518126	...	3.0	2.0	1.0	7.0
827	10.776242	...	4.0	3.0	2.0	1.0
828	11.480411	...	5.0	4.0	3.0	2.0
829	10.411403	...	6.0	5.0	4.0	3.0

	weekday(t-9)	weekday(t-10)	weekday(t-11)	weekday(t-12)	weekday(t-13)	\
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	3.0	NaN	NaN	NaN	NaN	
10	4.0	3.0	NaN	NaN	NaN	
11	5.0	4.0	3.0	NaN	NaN	
12	6.0	5.0	4.0	3.0	NaN	
13	7.0	6.0	5.0	4.0	3.0	
14	1.0	7.0	6.0	5.0	4.0	
15	2.0	1.0	7.0	6.0	5.0	
16	3.0	2.0	1.0	7.0	6.0	
17	4.0	3.0	2.0	1.0	7.0	
18	5.0	4.0	3.0	2.0	1.0	
19	6.0	5.0	4.0	3.0	2.0	
20	7.0	6.0	5.0	4.0	3.0	
21	1.0	7.0	6.0	5.0	4.0	
22	2.0	1.0	7.0	6.0	5.0	
23	3.0	2.0	1.0	7.0	6.0	
24	4.0	3.0	2.0	1.0	7.0	
25	5.0	4.0	3.0	2.0	1.0	
26	6.0	5.0	4.0	3.0	2.0	
27	7.0	6.0	5.0	4.0	3.0	
28	1.0	7.0	6.0	5.0	4.0	
29	2.0	1.0	7.0	6.0	5.0	
..	...	...	...	...	...	
800	1.0	7.0	6.0	5.0	4.0	
801	2.0	1.0	7.0	6.0	5.0	
802	3.0	2.0	1.0	7.0	6.0	
803	4.0	3.0	2.0	1.0	7.0	
804	5.0	4.0	3.0	2.0	1.0	
805	6.0	5.0	4.0	3.0	2.0	
806	7.0	6.0	5.0	4.0	3.0	
807	1.0	7.0	6.0	5.0	4.0	
808	2.0	1.0	7.0	6.0	5.0	
809	3.0	2.0	1.0	7.0	6.0	
810	4.0	3.0	2.0	1.0	7.0	
811	5.0	4.0	3.0	2.0	1.0	
812	6.0	5.0	4.0	3.0	2.0	
813	7.0	6.0	5.0	4.0	3.0	
814	1.0	7.0	6.0	5.0	4.0	

815	2.0	1.0	7.0	6.0	5.0
816	3.0	2.0	1.0	7.0	6.0
817	4.0	3.0	2.0	1.0	7.0
818	5.0	4.0	3.0	2.0	1.0
819	6.0	5.0	4.0	3.0	2.0
820	7.0	6.0	5.0	4.0	3.0
821	1.0	7.0	6.0	5.0	4.0
822	2.0	1.0	7.0	6.0	5.0
823	3.0	2.0	1.0	7.0	6.0
824	4.0	3.0	2.0	1.0	7.0
825	5.0	4.0	3.0	2.0	1.0
826	6.0	5.0	4.0	3.0	2.0
827	7.0	6.0	5.0	4.0	3.0
828	1.0	7.0	6.0	5.0	4.0
829	2.0	1.0	7.0	6.0	5.0

	weekday(t-14)
0	NaN
1	NaN
2	NaN
3	NaN
4	NaN
5	NaN
6	NaN
7	NaN
8	NaN
9	NaN
10	NaN
11	NaN
12	NaN
13	NaN
14	3.0
15	4.0
16	5.0
17	6.0
18	7.0
19	1.0
20	2.0
21	3.0
22	4.0
23	5.0
24	6.0
25	7.0
26	1.0
27	2.0
28	3.0
29	4.0
..	...

```
800           3.0
801           4.0
802           5.0
803           6.0
804           7.0
805           1.0
806           2.0
807           3.0
808           4.0
809           5.0
810           6.0
811           7.0
812           1.0
813           2.0
814           3.0
815           4.0
816           5.0
817           6.0
818           7.0
819           1.0
820           2.0
821           3.0
822           4.0
823           5.0
824           6.0
825           7.0
826           1.0
827           2.0
828           3.0
829           4.0

[830 rows x 77 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.drop(['index','date','apparentTemperatureMax','apparentTemperatureM
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	

	t-9	...	weekday(t-5)	weekday(t-6)	weekday(t-7)	weekday(t-8)	\
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

   weekday(t-9)  weekday(t-10)  weekday(t-11)  weekday(t-12)  weekday(t-13)  \
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

   weekday(t-14)
0            NaN
1            NaN
2            NaN
3            NaN
4            NaN

[5 rows x 71 columns]
```

In [6]:
```
#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[6]:

	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	...	weekday(t-5)	\
14	9.227707	8.813513	9.274873	9.103382	...	5.0	
15	10.145910	9.227707	8.813513	9.274873	...	6.0	
16	10.780273	10.145910	9.227707	8.813513	...	7.0	
17	12.163127	10.780273	10.145910	9.227707	...	1.0	
18	10.609714	12.163127	10.780273	10.145910	...	2.0	

	weekday(t-6)	weekday(t-7)	weekday(t-8)	weekday(t-9)	weekday(t-10)	\
14	4.0	3.0	2.0	1.0	7.0	
15	5.0	4.0	3.0	2.0	1.0	
16	6.0	5.0	4.0	3.0	2.0	
17	7.0	6.0	5.0	4.0	3.0	
18	1.0	7.0	6.0	5.0	4.0	

	weekday(t-11)	weekday(t-12)	weekday(t-13)	weekday(t-14)
14	6.0	5.0	4.0	3.0

```
        15          7.0         6.0         5.0         4.0
        16          1.0         7.0         6.0         5.0
        17          2.0         1.0         7.0         6.0
        18          3.0         2.0         1.0         7.0

        [5 rows x 71 columns]
```

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

Out[7]: 816

In [7]:
```
#normalitzem
scaler=preprocessing.MinMaxScaler(feature_range=(0, 1))
daily_dia_norm=scaler.fit_transform(daily_dia)
```

In [8]:
```
#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 [9]:
```
# 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'])
```

```
WARNING:tensorflow:From c:\users\laura\appdata\local\programs\python\python37\lib\site-packages
Instructions for updating:
Colocations handled automatically by placer.
```

In [10]:
```
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
Instructions for updating:
Use tf.cast instead.
```

```
In [11]: #Dividim la suma de scores de test entre el nombre de prediccions per obtenir la mitj
         sumScores/(lenght-n_train)
```

```
Out[11]: 0.02920458300042364
```

```
In [12]: llista_scores
```

```
Out[12]: [0.0032813839690737368,
          0.08255193851794984,
```

```
0.0045423211984649825,
0.018203012090584503,
0.01737050278037433,
0.06542844147766158,
0.033085725603486216,
0.01214984558724641,
0.10088646822462355,
0.05175425053279037,
0.03787588832020816,
0.05282771697474975,
0.09515512448993624,
0.08068377226597367,
0.024253962991972466,
0.12456723948404158,
0.11994353720004236,
0.041703484570434934,
0.10958477289329593,
0.07900614090861402,
0.027063930434184202,
0.08668039456569221,
0.052258605764777366,
0.003057256599590996,
0.04910935473916056,
0.029648361342025975,
0.051468917015615734,
0.07469426873470875,
0.025508984924089262,
0.009637255655596144,
0.03198025786178205,
0.11936924995331033,
0.05718683305165384,
0.03419454565852442,
0.03604222238352728,
0.05057381025897967,
0.052208985190217216,
0.0497259735328508,
0.007968141136309947,
0.014890693804721522,
0.03473343778513882,
0.03228406931546346,
0.04971166577904973,
0.020865291577815803,
0.04814991003530977,
0.014553831434167996,
0.008031861240635152,
0.027547887446458086,
0.031709277097019095,
0.004563201519243609,
```

```
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0.03148145408955405,
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0.06399576790629236,
0.04068846279749627,
0.009504133937346726,
0.0012867036391775999,
0.012718831564675459,
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0.00805893264795654,
0.03534474332508941,
0.045989487018471276,
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0.0008970306179710841,
0.02923226411940605,
```

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0.021424881734709578,
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0.019085132348839573,
0.012902481311136471,

```
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```
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0.00983043866300326,
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0.001074426609839474,
0.009799325358235134,
0.035902256115529996,
0.03666392289064513,
0.0844914118696165,
0.07485881311954512,
0.030573665039967723,
0.08202770048982155,
0.06264039494032603,
0.003059263593300443,
0.06715712081616565,
```

```
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0.04447668953756634,
0.041895251448822224,
0.0320814274101382,
0.06310933784365469,
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0.023650041570227653,
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0.016012650381318716,
0.03933629686637552,
0.05918280586479807,
0.1181398532200355,
0.03835526282516932,
0.06649556173528692,
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0.0037624463689744925,
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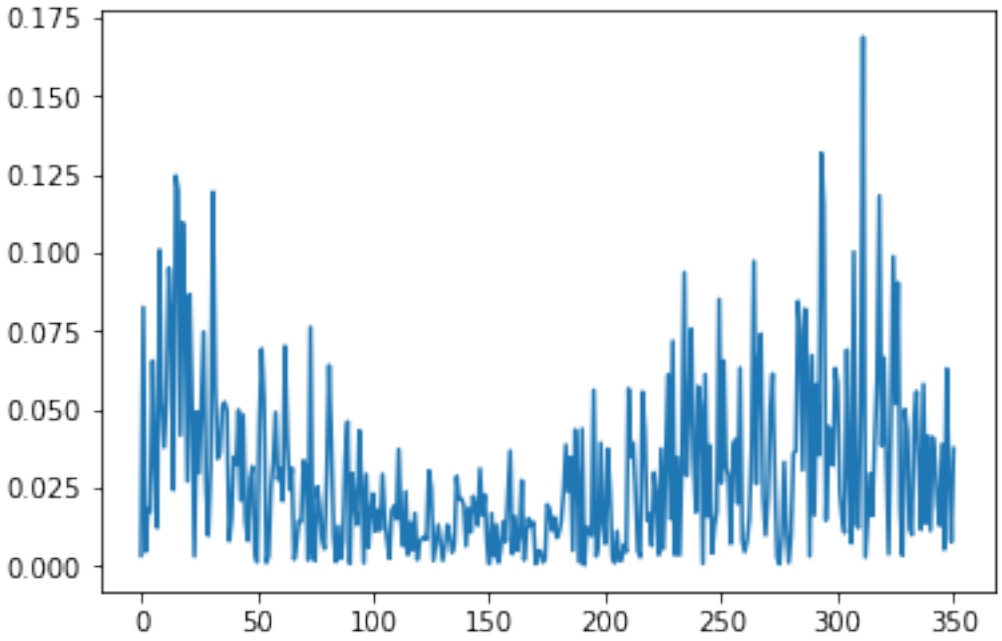
0.043008732491006896,
0.01186932504442062,
0.009918026641999811,
0.04694965229240222,
0.05567768111005389,
0.03071454660564621,
0.011539750927439307,
0.05798329200853414,
```

```
 0.013610855546844602,
 0.04154094913517903,
 0.011244075138834742,
 0.041073928125424874,
 0.03368927085440521,
 0.01999785466872095,
 0.012878730507450697,
 0.03878532742758911,
 0.005266520752180481,
 0.0629330572985618,
 0.027656990847906737,
 0.007598504730579103,
 0.037483932714463375]
```

In [13]:
```
plt.plot(llista_scores)
```

Out[13]: [<matplotlib.lines.Line2D at 0x262abf0d5f8>]



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

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

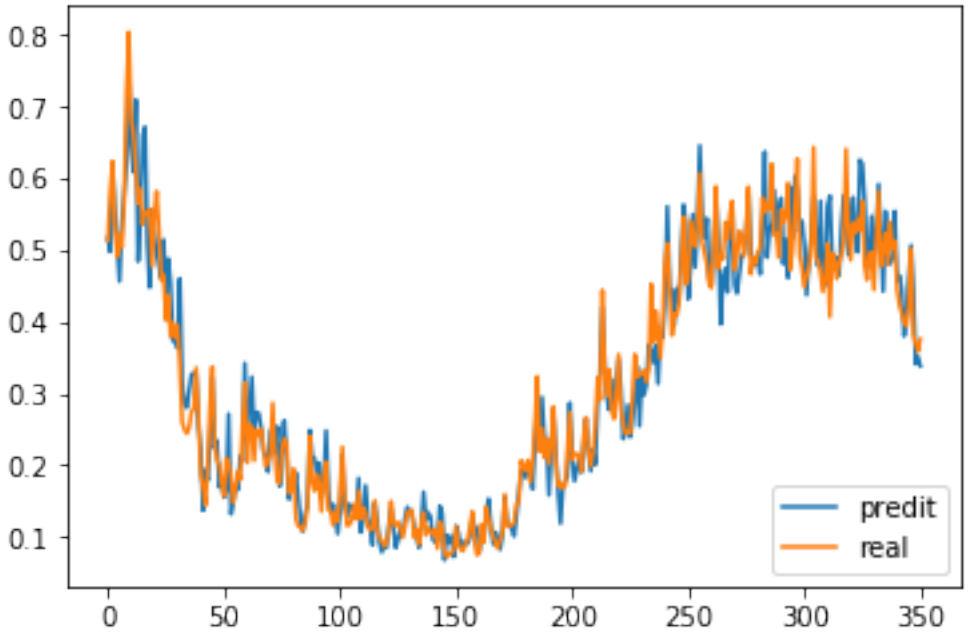
predis=np.reshape(predis, (351)  )

predis
```

```
Out[14]: array([0.51734287, 0.49805662, 0.61978412, 0.55748254, 0.50872564,
                0.45671695, 0.53752726, 0.57987481, 0.61857331, 0.75287652,
                0.6468398 , 0.60934889, 0.71034926, 0.48478228, 0.56139219,
                0.66109061, 0.67219985, 0.51055282, 0.44822454, 0.55680054,
                0.52413154, 0.49565828, 0.47751379, 0.46196163, 0.51484221,
                0.43227005, 0.48838684, 0.4547419 , 0.37335092, 0.38755327,
                0.36373642, 0.46063507, 0.31749803, 0.284473  , 0.2806668 ,
                0.30313057, 0.32697058, 0.32858655, 0.32798174, 0.25568241,
                0.22192994, 0.13712963, 0.19409516, 0.17897432, 0.23519614,
                0.32286501, 0.22360648, 0.23475212, 0.16981491, 0.17970705,
                0.15545374, 0.16444227, 0.27182782, 0.13209872, 0.14815342,
                0.17235743, 0.16611555, 0.21313727, 0.20646378, 0.34248671,
                0.23484926, 0.27775824, 0.32338995, 0.24800047, 0.27420086,
                0.27270243, 0.25275284, 0.2197621 , 0.21260372, 0.19093698,
                0.24738978, 0.2554093 , 0.21711925, 0.25439632, 0.17078863,
                0.22570801, 0.26240736, 0.21126792, 0.15281263, 0.17149338,
                0.16675913, 0.18949229, 0.15567821, 0.12374374, 0.10788213,
                0.13224153, 0.14526522, 0.24908425, 0.21258977, 0.21012086,
                0.18668914, 0.20308459, 0.15534711, 0.17267233, 0.24839813,
                0.16388975, 0.13500889, 0.14568648, 0.12923591, 0.1046499 ,
                0.13600872, 0.21450625, 0.14192058, 0.12726283, 0.1461426 ,
                0.13930501, 0.13745269, 0.12734364, 0.18215007, 0.10664023,
                0.1256015 , 0.17190391, 0.12820143, 0.11584614, 0.08884515,
                0.14734782, 0.12311141, 0.09279351, 0.07902463, 0.08923546,
                0.08475557, 0.11525758, 0.14045322, 0.11193646, 0.08509562,
                0.09611204, 0.11123569, 0.10633893, 0.12742522, 0.1416522 ,
                0.1345237 , 0.13099092, 0.11252519, 0.10098921, 0.0853095 ,
                0.11396189, 0.16285594, 0.12378123, 0.13427377, 0.12781365,
                0.0968092 , 0.09275592, 0.09466843, 0.14323142, 0.13205318,
                0.06801446, 0.10285467, 0.09192734, 0.10137491, 0.0726117 ,
                0.11547431, 0.10530686, 0.0863624 , 0.09362714, 0.08954038,
                0.09108956, 0.09967265, 0.127821  , 0.11407544, 0.11167341,
                0.08098154, 0.10203644, 0.09764989, 0.13246423, 0.1532931 ,
                0.11494999, 0.09024283, 0.10682061, 0.09737348, 0.08277751,
                0.10378493, 0.1549522 , 0.12027851, 0.11404417, 0.11415703,
                0.10125235, 0.12843421, 0.15924923, 0.19121359, 0.19200829,
                0.18273683, 0.19245982, 0.19939731, 0.16694856, 0.23353498,
                0.28947684, 0.22416025, 0.29515398, 0.23841909, 0.23654748,
                0.15857364, 0.22943828, 0.26948112, 0.21527244, 0.16112544,
                0.11910985, 0.17113946, 0.17850552, 0.1849757 , 0.28727019,
                0.21436913, 0.17817216, 0.19289753, 0.20837171, 0.19044679,
                0.23054707, 0.26513633, 0.21705772, 0.19205001, 0.2224697 ,
                0.20100558, 0.28743806, 0.33049518, 0.42141113, 0.34448367,
                0.2935788 , 0.27776784, 0.32001117, 0.27936268, 0.31071615,
                0.34843615, 0.30717713, 0.23712046, 0.24793343, 0.28451473,
                0.23951359, 0.25952134, 0.29393458, 0.33801278, 0.25498164,
                0.32502297, 0.29828459, 0.31300005, 0.36795807, 0.35931879,
                0.34352213, 0.3679705 , 0.31412777, 0.38466042, 0.37855506,
```

```
0.42501819, 0.56124336, 0.44823977, 0.44266233, 0.39540476,
0.44675997, 0.428451  , 0.4940595 , 0.56374592, 0.53808928,
0.43213621, 0.47571498, 0.55033422, 0.47638911, 0.54122639,
0.64620364, 0.55176538, 0.51748949, 0.54474562, 0.46199667,
0.45214826, 0.51529282, 0.57399791, 0.51266271, 0.39702606,
0.46164489, 0.47462228, 0.44161403, 0.52938467, 0.55900872,
0.4509891 , 0.44020736, 0.46581632, 0.49226731, 0.49450558,
0.52977788, 0.57871735, 0.5001533 , 0.47840765, 0.48263082,
0.48788762, 0.46681437, 0.53696299, 0.63778621, 0.49009004,
0.5271517 , 0.53839552, 0.58396399, 0.52506661, 0.55755132,
0.57304358, 0.4810074 , 0.51478398, 0.46179855, 0.58854121,
0.51723719, 0.60370195, 0.58617306, 0.52577049, 0.54112518,
0.5068658 , 0.43775639, 0.49074215, 0.51883596, 0.57488549,
0.5189994 , 0.47805876, 0.56893998, 0.45566204, 0.45141363,
0.56180626, 0.57612109, 0.49340934, 0.47407317, 0.48976287,
0.46559855, 0.53317773, 0.57658702, 0.52315509, 0.49391887,
0.55306637, 0.57382071, 0.52083588, 0.49758816, 0.62624687,
0.62023675, 0.56974477, 0.47642407, 0.49772316, 0.54778451,
0.48796314, 0.52352214, 0.59236789, 0.52451205, 0.44294217,
0.55463475, 0.49151114, 0.48133442, 0.48923451, 0.5546304 ,
0.45700824, 0.46475345, 0.44538346, 0.38043568, 0.40708813,
0.46183294, 0.50698805, 0.44521868, 0.34162301, 0.35139698,
0.33865082])
```

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 [16]:
```
#Creem un dataset amb format (nombre prediccions,17) per tornar les prediccions i els
#El necessitem d'questa 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 aleatories 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[16]:

	predi	y	t-2	t-3	t-4	t-5	\
479	0.517343	0.514061	12.119938	12.852295	13.106773	12.823073	
480	0.498057	0.580609	11.786082	12.119938	12.852295	13.106773	
481	0.619784	0.624326	11.590859	11.786082	12.119938	12.852295	
482	0.557483	0.539280	12.186487	11.590859	11.786082	12.119938	
483	0.508726	0.491355	12.577783	12.186487	11.590859	11.786082	
484	0.456717	0.522145	11.816573	12.577783	12.186487	11.590859	
485	0.537527	0.504442	11.387627	11.816573	12.577783	12.186487	
486	0.579875	0.567725	11.663214	11.387627	11.816573	12.577783	
487	0.618573	0.719460	11.504756	11.663214	11.387627	11.816573	
488	0.752877	0.804631	12.071173	11.504756	11.663214	11.387627	
489	0.646840	0.684716	13.429271	12.071173	11.504756	11.663214	

490	0.609349	0.662177	14.191591	13.429271	12.071173	11.504756
491	0.710349	0.615194	13.118295	14.191591	13.429271	12.071173
492	0.484782	0.565466	12.916559	13.118295	14.191591	13.429271
493	0.561392	0.585646	12.496044	12.916559	13.118295	14.191591
494	0.661091	0.536523	12.050954	12.496044	12.916559	13.118295
495	0.672200	0.552256	12.231576	12.050954	12.496044	12.916559
496	0.510553	0.552256	11.791904	12.231576	12.050954	12.496044
497	0.448225	0.557809	11.932721	11.791904	12.231576	12.050954
498	0.556801	0.477794	11.932721	11.932721	11.791904	12.231576
499	0.524132	0.551195	11.982423	11.932721	11.932721	11.791904
500	0.495658	0.582339	11.266252	11.982423	11.932721	11.932721
501	0.477514	0.529772	11.923226	11.266252	11.982423	11.932721
502	0.461962	0.458904	12.201972	11.923226	11.266252	11.982423
503	0.514842	0.465733	11.731479	12.201972	11.923226	11.266252
504	0.432270	0.402622	11.097177	11.731479	12.201972	11.923226
505	0.488387	0.436918	11.158295	11.097177	11.731479	12.201972
506	0.454742	0.380048	10.593420	11.158295	11.097177	11.731479
507	0.373351	0.398860	10.900388	10.593420	11.158295	11.097177
508	0.387553	0.377916	10.391372	10.900388	10.593420	11.158295
..	...	...	...	...	...	...
800	0.573821	0.537515	11.753871	12.729659	11.620778	11.409880
801	0.520836	0.524598	11.344805	11.753871	12.729659	11.620778
802	0.497588	0.543903	11.800777	11.344805	11.753871	12.729659
803	0.626247	0.527438	11.685169	11.800777	11.344805	11.753871
804	0.620237	0.568506	11.857957	11.685169	11.800777	11.344805
805	0.569745	0.479332	11.710582	11.857957	11.685169	11.800777
806	0.476424	0.458726	12.078164	11.710582	11.857957	11.685169
807	0.497723	0.494425	11.280011	12.078164	11.710582	11.857957
808	0.547785	0.497810	11.095584	11.280011	12.078164	11.710582
809	0.487963	0.444954	11.415105	11.095584	11.280011	12.078164
810	0.523522	0.511653	11.445403	11.415105	11.095584	11.280011
811	0.592368	0.582450	10.972318	11.445403	11.415105	11.095584
812	0.524512	0.477562	11.569300	10.972318	11.445403	11.415105
813	0.442942	0.498620	12.202967	11.569300	10.972318	11.445403
814	0.554635	0.523920	11.264175	12.202967	11.569300	10.972318
815	0.491511	0.479971	11.452649	11.264175	12.202967	11.569300
816	0.481334	0.539318	11.679099	11.452649	11.264175	12.202967
817	0.489235	0.502845	11.285737	11.679099	11.452649	11.264175
818	0.554630	0.513089	11.816914	11.285737	11.679099	11.452649
819	0.457008	0.445764	11.490470	11.816914	11.285737	11.679099
820	0.464753	0.423680	11.582159	11.490470	11.816914	11.285737
821	0.445383	0.411694	10.979566	11.582159	11.490470	11.816914
822	0.380436	0.400434	10.781898	10.979566	11.582159	11.490470
823	0.407088	0.394209	10.674624	10.781898	10.979566	11.582159
824	0.461833	0.423048	10.573835	10.674624	10.781898	10.979566
825	0.506988	0.501722	10.518126	10.573835	10.674624	10.781898
826	0.445219	0.382286	10.776242	10.518126	10.573835	10.674624
827	0.341623	0.369280	11.480411	10.776242	10.518126	10.573835

```
828  0.351397  0.358995  10.411403  11.480411  10.776242  10.518126
829  0.338651  0.376135  10.294997  10.411403  11.480411  10.776242

           t-6        t-7        t-8        t-9  ...  weekday(t-5)  \
479  11.559878  10.930170  10.889469  10.675248  ...           7.0
480  12.823073  11.559878  10.930170  10.889469  ...           1.0
481  13.106773  12.823073  11.559878  10.930170  ...           2.0
482  12.852295  13.106773  12.823073  11.559878  ...           3.0
483  12.119938  12.852295  13.106773  12.823073  ...           4.0
484  11.786082  12.119938  12.852295  13.106773  ...           5.0
485  11.590859  11.786082  12.119938  12.852295  ...           6.0
486  12.186487  11.590859  11.786082  12.119938  ...           7.0
487  12.577783  12.186487  11.590859  11.786082  ...           1.0
488  11.816573  12.577783  12.186487  11.590859  ...           2.0
489  11.387627  11.816573  12.577783  12.186487  ...           3.0
490  11.663214  11.387627  11.816573  12.577783  ...           4.0
491  11.504756  11.663214  11.387627  11.816573  ...           5.0
492  12.071173  11.504756  11.663214  11.387627  ...           6.0
493  13.429271  12.071173  11.504756  11.663214  ...           7.0
494  14.191591  13.429271  12.071173  11.504756  ...           1.0
495  13.118295  14.191591  13.429271  12.071173  ...           2.0
496  12.916559  13.118295  14.191591  13.429271  ...           3.0
497  12.496044  12.916559  13.118295  14.191591  ...           4.0
498  12.050954  12.496044  12.916559  13.118295  ...           5.0
499  12.231576  12.050954  12.496044  12.916559  ...           6.0
500  11.791904  12.231576  12.050954  12.496044  ...           7.0
501  11.932721  11.791904  12.231576  12.050954  ...           7.0
502  11.932721  11.932721  11.791904  12.231576  ...           1.0
503  11.982423  11.932721  11.932721  11.791904  ...           2.0
504  11.266252  11.982423  11.932721  11.932721  ...           3.0
505  11.923226  11.266252  11.982423  11.932721  ...           4.0
506  12.201972  11.923226  11.266252  11.982423  ...           5.0
507  11.731479  12.201972  11.923226  11.266252  ...           6.0
508  11.097177  11.731479  12.201972  11.923226  ...           7.0
..         ...        ...        ...        ...  ...           ...
800  11.300414  11.109560  11.370601  11.430883  ...           5.0
801  11.409880  11.300414  11.109560  11.370601  ...           6.0
802  11.620778  11.409880  11.300414  11.109560  ...           7.0
803  12.729659  11.620778  11.409880  11.300414  ...           1.0
804  11.753871  12.729659  11.620778  11.409880  ...           2.0
805  11.344805  11.753871  12.729659  11.620778  ...           3.0
806  11.800777  11.344805  11.753871  12.729659  ...           4.0
807  11.685169  11.800777  11.344805  11.753871  ...           5.0
808  11.857957  11.685169  11.800777  11.344805  ...           6.0
809  11.710582  11.857957  11.685169  11.800777  ...           7.0
810  12.078164  11.710582  11.857957  11.685169  ...           1.0
811  11.280011  12.078164  11.710582  11.857957  ...           2.0
812  11.095584  11.280011  12.078164  11.710582  ...           3.0
```

813	11.415105	11.095584	11.280011	12.078164	...	4.0
814	11.445403	11.415105	11.095584	11.280011	...	5.0
815	10.972318	11.445403	11.415105	11.095584	...	6.0
816	11.569300	10.972318	11.445403	11.415105	...	7.0
817	12.202967	11.569300	10.972318	11.445403	...	1.0
818	11.264175	12.202967	11.569300	10.972318	...	2.0
819	11.452649	11.264175	12.202967	11.569300	...	3.0
820	11.679099	11.452649	11.264175	12.202967	...	4.0
821	11.285737	11.679099	11.452649	11.264175	...	5.0
822	11.816914	11.285737	11.679099	11.452649	...	6.0
823	11.490470	11.816914	11.285737	11.679099	...	7.0
824	11.582159	11.490470	11.816914	11.285737	...	1.0
825	10.979566	11.582159	11.490470	11.816914	...	2.0
826	10.781898	10.979566	11.582159	11.490470	...	3.0
827	10.674624	10.781898	10.979566	11.582159	...	4.0
828	10.573835	10.674624	10.781898	10.979566	...	5.0
829	10.518126	10.573835	10.674624	10.781898	...	6.0

	weekday(t-6)	weekday(t-7)	weekday(t-8)	weekday(t-9)	weekday(t-10)	\
479	6.0	5.0	4.0	3.0	2.0	
480	7.0	6.0	5.0	4.0	3.0	
481	1.0	7.0	6.0	5.0	4.0	
482	2.0	1.0	7.0	6.0	5.0	
483	3.0	2.0	1.0	7.0	6.0	
484	4.0	3.0	2.0	1.0	7.0	
485	5.0	4.0	3.0	2.0	1.0	
486	6.0	5.0	4.0	3.0	2.0	
487	7.0	6.0	5.0	4.0	3.0	
488	1.0	7.0	6.0	5.0	4.0	
489	2.0	1.0	7.0	6.0	5.0	
490	3.0	2.0	1.0	7.0	6.0	
491	4.0	3.0	2.0	1.0	7.0	
492	5.0	4.0	3.0	2.0	1.0	
493	6.0	5.0	4.0	3.0	2.0	
494	7.0	6.0	5.0	4.0	3.0	
495	1.0	7.0	6.0	5.0	4.0	
496	2.0	1.0	7.0	6.0	5.0	
497	3.0	2.0	1.0	7.0	6.0	
498	4.0	3.0	2.0	1.0	7.0	
499	5.0	4.0	3.0	2.0	1.0	
500	6.0	5.0	4.0	3.0	2.0	
501	7.0	6.0	5.0	4.0	3.0	
502	7.0	7.0	6.0	5.0	4.0	
503	1.0	7.0	7.0	6.0	5.0	
504	2.0	1.0	7.0	7.0	6.0	
505	3.0	2.0	1.0	7.0	7.0	
506	4.0	3.0	2.0	1.0	7.0	
507	5.0	4.0	3.0	2.0	1.0	

508	6.0	5.0	4.0	3.0	2.0
..	...	...	...	...	...
800	4.0	3.0	2.0	1.0	7.0
801	5.0	4.0	3.0	2.0	1.0
802	6.0	5.0	4.0	3.0	2.0
803	7.0	6.0	5.0	4.0	3.0
804	1.0	7.0	6.0	5.0	4.0
805	2.0	1.0	7.0	6.0	5.0
806	3.0	2.0	1.0	7.0	6.0
807	4.0	3.0	2.0	1.0	7.0
808	5.0	4.0	3.0	2.0	1.0
809	6.0	5.0	4.0	3.0	2.0
810	7.0	6.0	5.0	4.0	3.0
811	1.0	7.0	6.0	5.0	4.0
812	2.0	1.0	7.0	6.0	5.0
813	3.0	2.0	1.0	7.0	6.0
814	4.0	3.0	2.0	1.0	7.0
815	5.0	4.0	3.0	2.0	1.0
816	6.0	5.0	4.0	3.0	2.0
817	7.0	6.0	5.0	4.0	3.0
818	1.0	7.0	6.0	5.0	4.0
819	2.0	1.0	7.0	6.0	5.0
820	3.0	2.0	1.0	7.0	6.0
821	4.0	3.0	2.0	1.0	7.0
822	5.0	4.0	3.0	2.0	1.0
823	6.0	5.0	4.0	3.0	2.0
824	7.0	6.0	5.0	4.0	3.0
825	1.0	7.0	6.0	5.0	4.0
826	2.0	1.0	7.0	6.0	5.0
827	3.0	2.0	1.0	7.0	6.0
828	4.0	3.0	2.0	1.0	7.0
829	5.0	4.0	3.0	2.0	1.0

	weekday(t-11)	weekday(t-12)	weekday(t-13)	weekday(t-14)
479	1.0	7.0	6.0	5.0
480	2.0	1.0	7.0	6.0
481	3.0	2.0	1.0	7.0
482	4.0	3.0	2.0	1.0
483	5.0	4.0	3.0	2.0
484	6.0	5.0	4.0	3.0
485	7.0	6.0	5.0	4.0
486	1.0	7.0	6.0	5.0
487	2.0	1.0	7.0	6.0
488	3.0	2.0	1.0	7.0
489	4.0	3.0	2.0	1.0
490	5.0	4.0	3.0	2.0
491	6.0	5.0	4.0	3.0
492	7.0	6.0	5.0	4.0

493	1.0	7.0	6.0	5.0
494	2.0	1.0	7.0	6.0
495	3.0	2.0	1.0	7.0
496	4.0	3.0	2.0	1.0
497	5.0	4.0	3.0	2.0
498	6.0	5.0	4.0	3.0
499	7.0	6.0	5.0	4.0
500	1.0	7.0	6.0	5.0
501	2.0	1.0	7.0	6.0
502	3.0	2.0	1.0	7.0
503	4.0	3.0	2.0	1.0
504	5.0	4.0	3.0	2.0
505	6.0	5.0	4.0	3.0
506	7.0	6.0	5.0	4.0
507	7.0	7.0	6.0	5.0
508	1.0	7.0	7.0	6.0
..	...	...	...	...
800	6.0	5.0	4.0	3.0
801	7.0	6.0	5.0	4.0
802	1.0	7.0	6.0	5.0
803	2.0	1.0	7.0	6.0
804	3.0	2.0	1.0	7.0
805	4.0	3.0	2.0	1.0
806	5.0	4.0	3.0	2.0
807	6.0	5.0	4.0	3.0
808	7.0	6.0	5.0	4.0
809	1.0	7.0	6.0	5.0
810	2.0	1.0	7.0	6.0
811	3.0	2.0	1.0	7.0
812	4.0	3.0	2.0	1.0
813	5.0	4.0	3.0	2.0
814	6.0	5.0	4.0	3.0
815	7.0	6.0	5.0	4.0
816	1.0	7.0	6.0	5.0
817	2.0	1.0	7.0	6.0
818	3.0	2.0	1.0	7.0
819	4.0	3.0	2.0	1.0
820	5.0	4.0	3.0	2.0
821	6.0	5.0	4.0	3.0
822	7.0	6.0	5.0	4.0
823	1.0	7.0	6.0	5.0
824	2.0	1.0	7.0	6.0
825	3.0	2.0	1.0	7.0
826	4.0	3.0	2.0	1.0
827	5.0	4.0	3.0	2.0
828	6.0	5.0	4.0	3.0
829	7.0	6.0	5.0	4.0

```
[351 rows x 71 columns]
```

In [17]:
```
# 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
```

```
[[ 11.62022911  11.59085917 115.46893021 ...  43.          37.
   31.        ]
 [ 11.44760811  12.18648691 112.48075791 ...   7.          43.
   37.        ]
 [ 12.53712663  12.57778255 110.7334244  ...  13.           7.
   43.        ]
 ...
 [ 10.04745348  10.2949966  109.74485905 ...  25.          19.
   13.        ]
 [ 10.13493512  10.20294532 100.17673598 ...  31.          25.
   19.        ]
 [ 10.02085102  10.3563499   99.13484299 ...  37.          31.
   25.        ]]
11.620229105354944
11.590859170709699
```

In [18]:
```
#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[18]:
```
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```

```
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```

```
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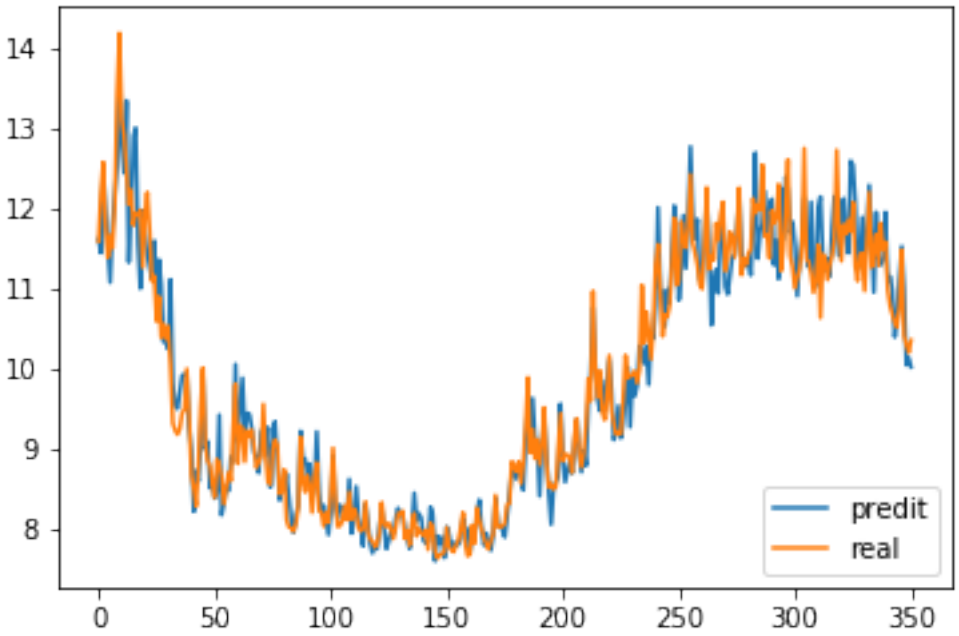
 10.294996596876901,
 10.202945322371301,
 10.3563498993587]
```

In [19]:
```
##Mostrem
plt.plot(listpredi, label="predit")
plt.plot(listy, label="real")
plt.legend(loc="lower right")
plt.show()
```



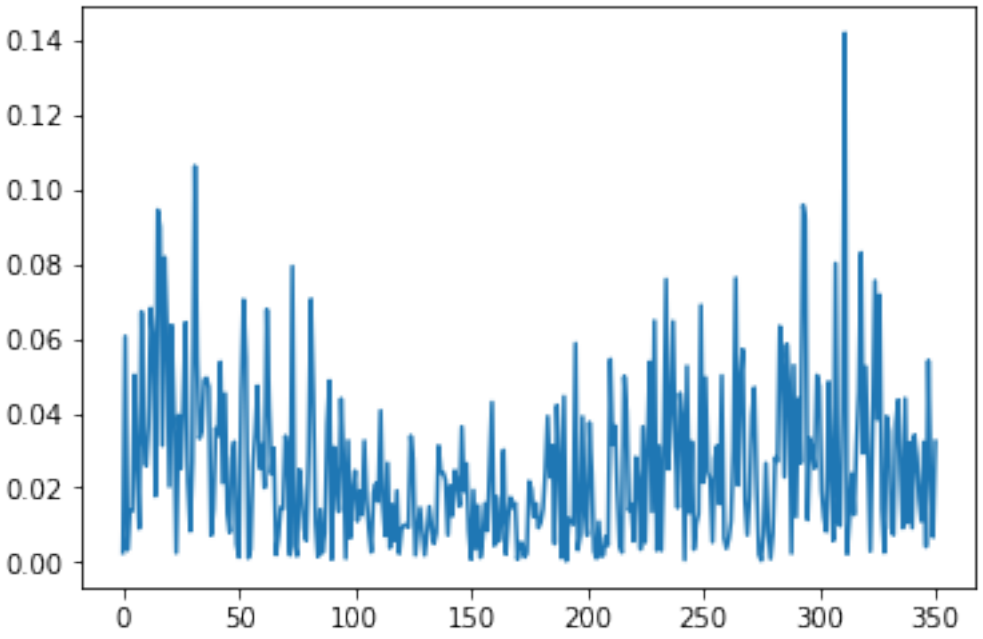
In [20]:
```
llista_errors=list()
llista_errorsabs=list()
llista_errorsres=list()


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 2.5491059334676294 %
```



In [ ]: