

# Untitled3

November 25, 2019

```
[57]: import pandas as pd
      from pandas.plotting import autocorrelation_plot, lag_plot

[21]: df_brazil = pd.read_csv("sudeste.csv", usecols=["date", "temp"])
      df_madrid = pd.read_csv("weather_madrid_LEMD_1997_2015.csv", usecols=["CET",
      ↪ "Mean TemperatureC"])

[13]: def create_final_df(df1, df2):
      df_brazil_no_dup_date = df1.groupby("date").mean().reset_index()
      df_final = pd.merge(df_brazil_no_dup_date, df2, how="inner",
      ↪ left_on="date", right_on="CET")
      df_final = df_final[["date", "temp", "Mean TemperatureC"]]
      df_final.columns = ["date", "temp_brazil", "temp_madrid"]
      return df_final

[14]: df_final = create_final_df(df_brazil, df_madrid)

[5]: df_final[["temp_brazil", "temp_madrid"]].corr()

[5]:          temp_brazil  temp_madrid
temp_brazil      1.000000     -0.030652
temp_madrid     -0.030652      1.000000

[22]: # Brazil and Madrid average daily temperatures have a negative correlation of
      ↪ -0.03 but that can be ignored.
      # As a result one can say Brazil and Madrid average daily temperatures are
      ↪ independent of each other.

[30]: def prepare_brazil(df):
      temp = df.groupby("date").mean().reset_index()
      date_series = temp["date"]
      temp_series = temp["temp"]
      temp_series.index = pd.DatetimeIndex(date_series)

      start_date, end_date = date_series.head(1).values[0], date_series.tail(1).
      ↪ values[0]
      idx = pd.date_range(start_date, end_date)
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    result = temp_series.reindex(idx, fill_value=0)

    return result

def prepare_madrid(df):
    temp = df
    date_series = temp["CET"]
    temp_series = temp["Mean TemperatureC"]
    temp_series.index = pd.DatetimeIndex(date_series)

    start_date, end_date = date_series.head(1).values[0], date_series.tail(1).
    ↪values[0]
    idx = pd.date_range(start_date, end_date)
    result = temp_series.reindex(idx, fill_value=0)

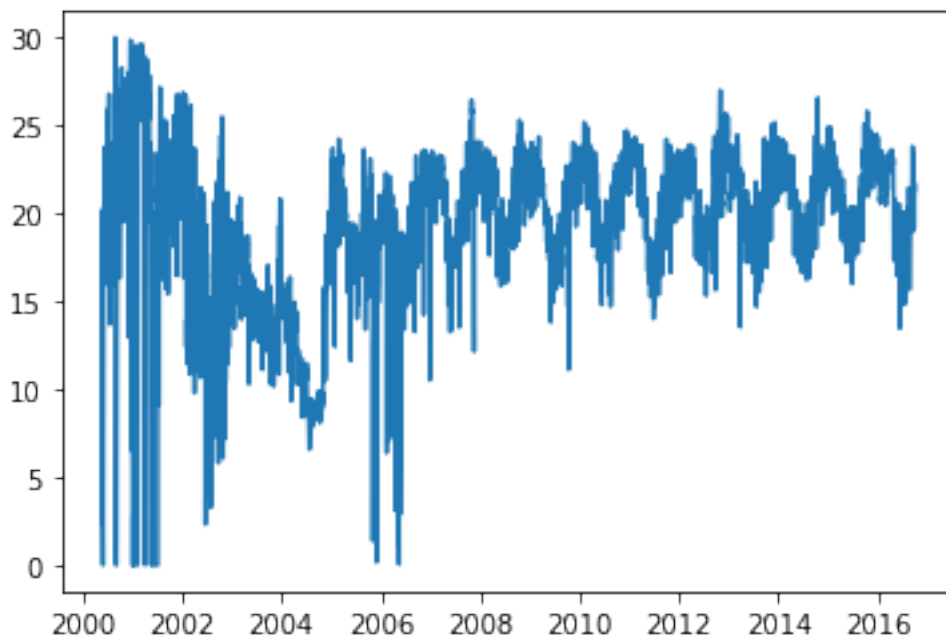
    return result

```

```
[31]: b, m = prepare_brazil(df_brazil), prepare_madrid(df_madrid)
```

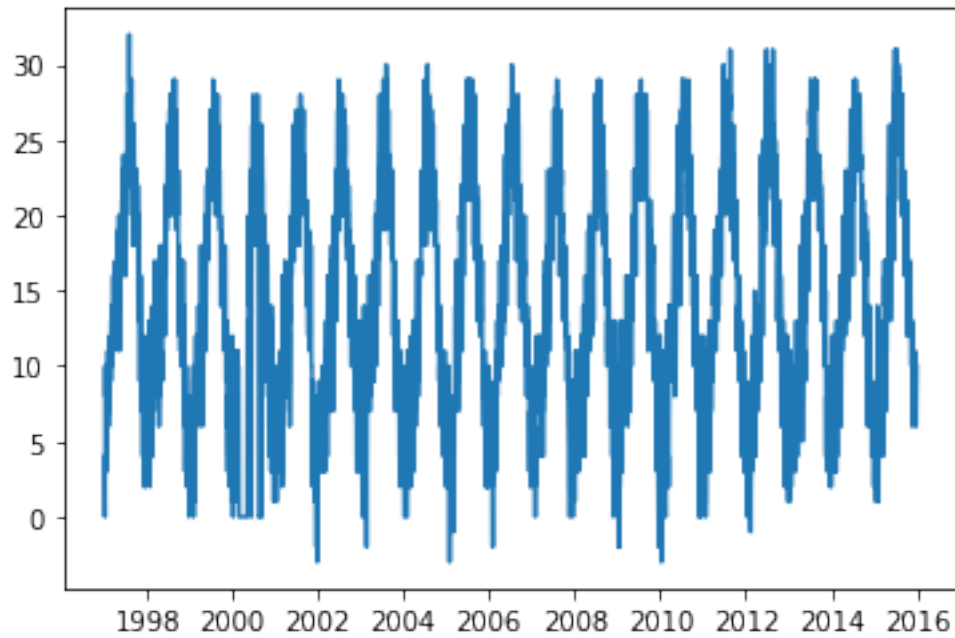
```
[32]: plt.plot(b)
```

```
[32]: [<matplotlib.lines.Line2D at 0x7f9eece95bd0>]
```



```
[33]: plt.plot(m)
```

```
[33]: [<matplotlib.lines.Line2D at 0x7f9eece69410>]
```



```
[6]: from statsmodels.tsa.stattools import adfuller
```

```
[43]: adfuller(b.dropna())
```

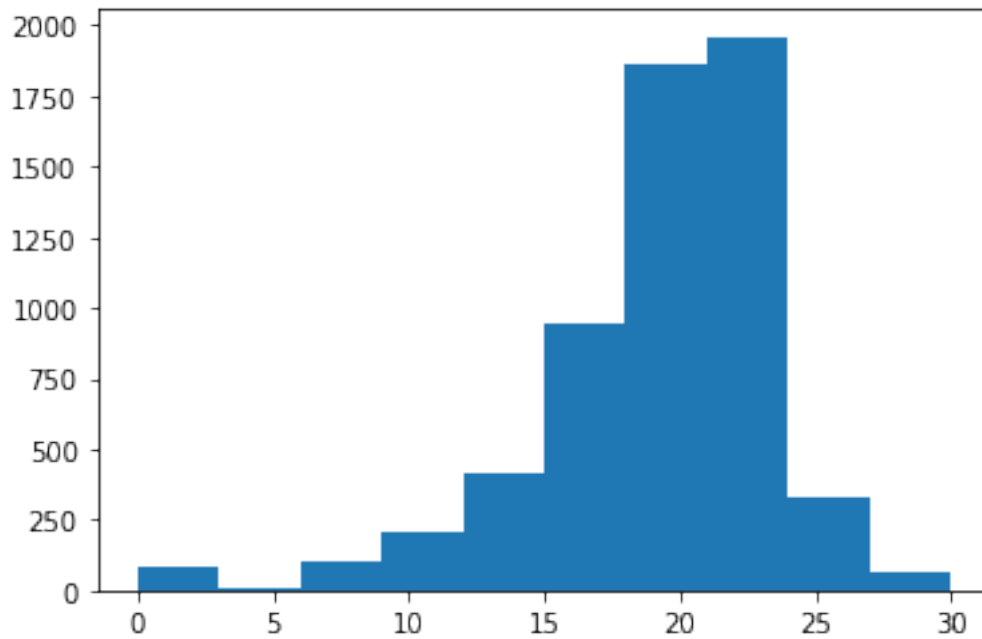
```
[43]: (-6.589743392006552,  
      7.164185170894732e-09,  
      21,  
      5952,  
      {'1%': -3.43144914692048,  
       '5%': -2.8620257211840996,  
       '10%': -2.5670285470333005},  
      23331.562288668385)
```

```
[42]: adfuller(m.dropna())
```

```
[42]: (-5.188311682682534,  
      9.288240716475554e-06,  
      14,  
      6921,  
      {'1%': -3.4312951996865126,  
       '5%': -2.861957701574514,  
       '10%': -2.5669923386600497},  
      29511.769274129747)
```

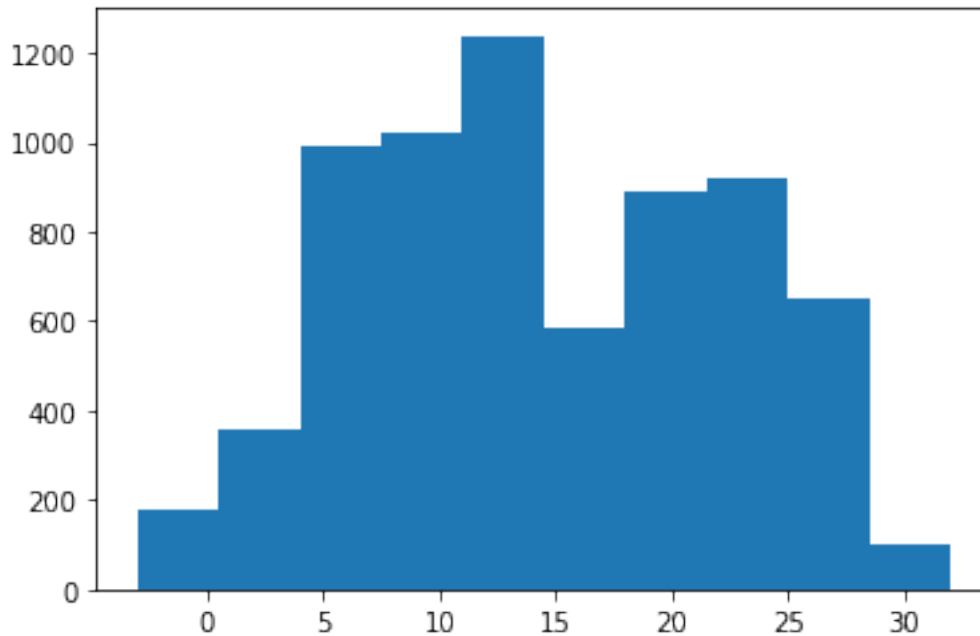
```
[47]: plt.hist(b.dropna())
```

```
[47]: (array([ 83.,  12., 100., 207., 410., 941., 1863., 1960., 330.,
        68.]),
      array([ 0.      ,  2.99875,  5.9975 ,  8.99625, 11.995  , 14.99375,
        17.9925 , 20.99125, 23.99   , 26.98875, 29.9875 ]),
      <a list of 10 Patch objects>)
```



```
[46]: plt.hist(m.dropna())
```

```
[46]: (array([ 180., 358., 989., 1024., 1238., 587., 891., 917., 653.,
        99.]),
      array([-3. ,  0.5,  4. ,  7.5, 11. , 14.5, 18. , 21.5, 25. , 28.5, 32. ]),
      <a list of 10 Patch objects>)
```



```
[50]: X = m.dropna().values
      low, high = X[:len(X)//2], X[len(X)//2:]
      print (low.mean(), high.mean())
      print (low.var(), high.var())
```

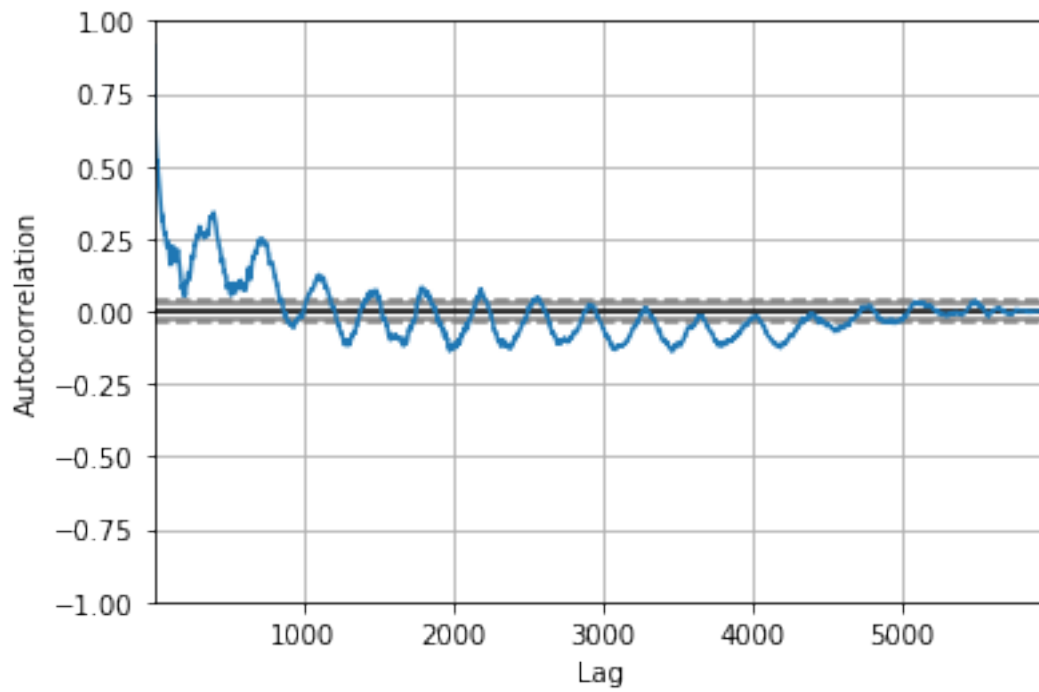
```
13.850346020761245 14.930219146482122
61.389080041745984 58.55856777204402
```

```
[51]: X = b.values
      low, high = X[:len(X)//2], X[len(X)//2:]
      print (low.mean(), high.mean())
      print (low.var(), high.var())
```

```
17.67876992357058 20.766865734120145
30.21849299020105 5.0722175008291615
```

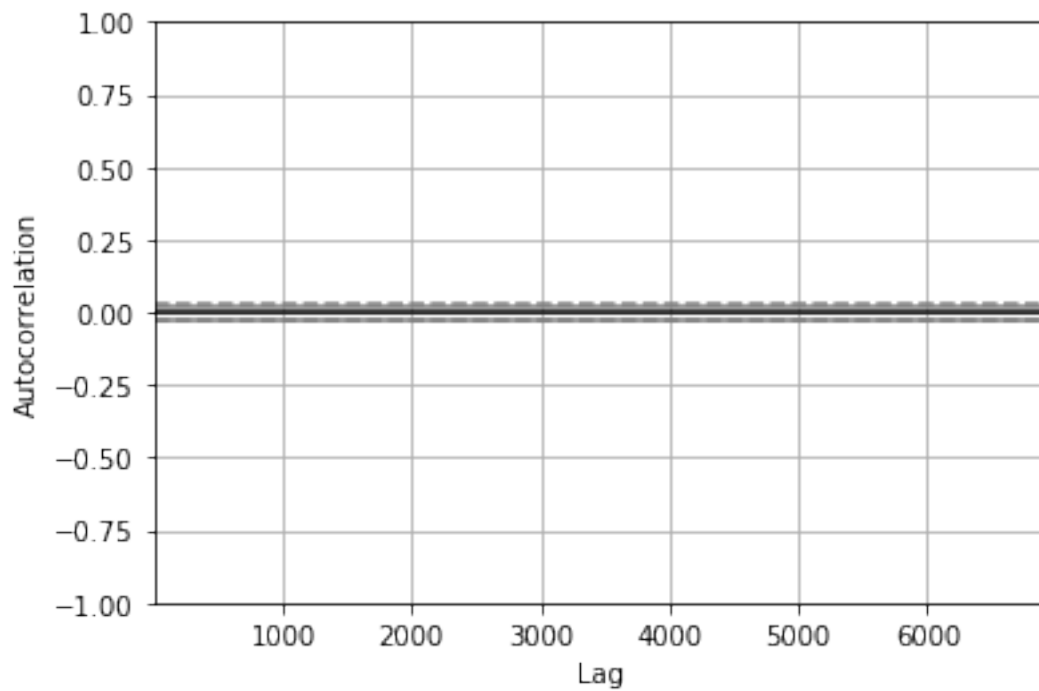
```
[60]: autocorrelation_plot(b)
```

```
[60]: <matplotlib.axes._subplots.AxesSubplot at 0x7f9f0c28b310>
```



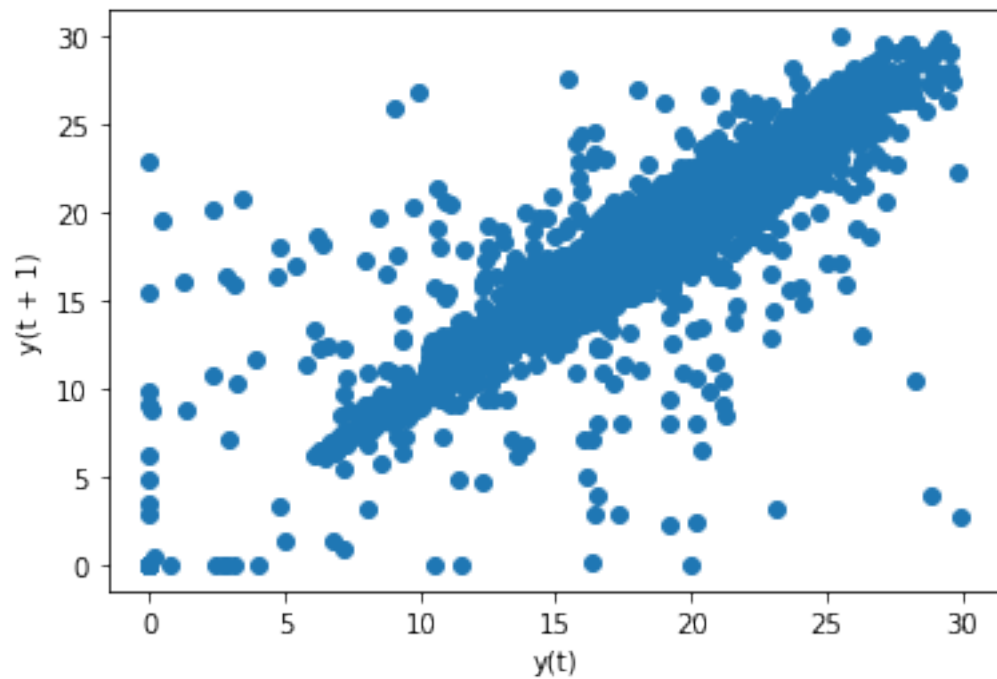
```
[61]: autocorrelation_plot(m)
```

```
[61]: <matplotlib.axes._subplots.AxesSubplot at 0x7f9f0c3c5390>
```



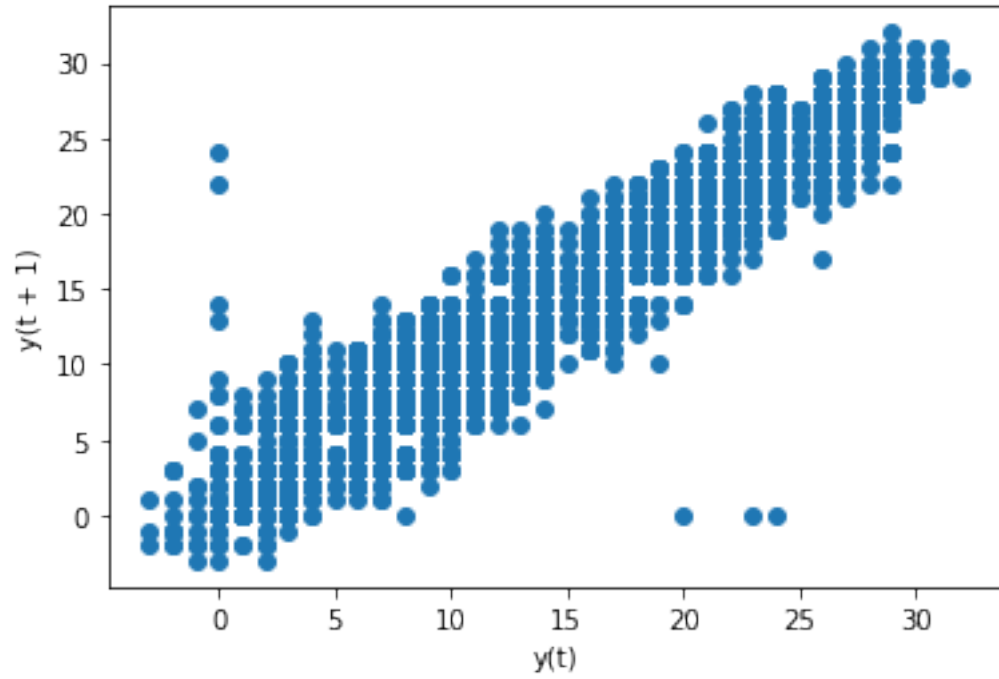
```
[58]: lag_plot(b)
```

```
[58]: <matplotlib.axes._subplots.AxesSubplot at 0x7f9f0fdb6050>
```



```
[59]: lag_plot(m)
```

```
[59]: <matplotlib.axes._subplots.AxesSubplot at 0x7f9f0c1b6550>
```



[65]: *# Brazil and Madrid datasets are both stationary and there no increasing trends*  
*# We can reach to this conclusion by checking autocorrelation, lag, histogram*  
*→ and default plottings.*  
*# In addition we also used adfuller test and for both datasets p-value is lower*  
*→ than 0.05*