

Forest Fires - week 3

June 14, 2016

Regression Modeling in Practice Course
Wesleyan University

Linear Regression Model
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The sample comes from Cortez and Morais study about predicting forest fires using meteorological data [Cortez and Morais, 2007]. The study includes data from 517 forest fires in the Natural Park Montesinho (Trás-os-Montes, in northeastern Portugal) January 2000 to December 2003, including meteorological data, the type of vegetation involved (which determines the six components of the Canadian Forest Fire Weather Index (FWI) system --see below--) and the total burned area in order to generate a model capable of predicting the burned area of small fires, which are more frequent.

Measures

The data contains:

- * X, Y: location of the fire (x,y axis spatial coordinate within the Montesinho park map: from 1 to 9)
- * month, day: month and day of the week the fire occurred (january to december and monday to sunday)
- * FWI system components:
 - FPMC: Fine Fuel Moisture Code (numeric rating of the moisture content of litter and other cured fine fuels: 18.7 to 96.2)
 - DMC: Duff Moisture Code (numeric rating of the average moisture content of loosely compacted organic layers of moderate depth: 1.1 to 291.3)
 - DC: Drought Code (numeric rating of the average moisture content of deep, compact organic layers: 7.9 to 860.6)
 - ISI: Initial Spread Index (numeric rating of the expected rate of fire spread: 0.0 to 56.1)
- * Meteorological variables:
 - temp: temperature (2.2 to 33.3 °C)
 - RH: relative humidity (15 to 100%)
 - wind: wind speed (0.4 to 9.4 Km/h)
 - rain: outside rain (0.0 to 6.4 mm/m²)
- * area: the burned area of the forest as response variable (0.0 to 1090.84 Ha).

1 Forest Fires

1.1 Import required libraries and set global options

```
In [1]: %matplotlib inline

import pandas
```

```

import matplotlib.pyplot as plt
import seaborn
import statsmodels.api as sm
import statsmodels.formula.api as smf
from pandas.tools.plotting import scatter_matrix
from math import ceil

pandas.set_option('display.float_format', lambda x: '%.3f'%x)
#pandas.set_option('display.mpl_style', 'default') # --deprecated
plt.style.use('ggplot') # Make the graphs a bit prettier
plt.rcParams['figure.figsize'] = (15, 5)

```

1.2 Load Forest Fires .csv file

```
In [2]: fires = pandas.read_csv('forestfires.csv')
```

2 Data Exploration

```
In [3]: fires.head() #Show first rows
```

```

Out[3]:   X  Y month  day  FPMC    DMC    DC  ISI  temp  RH  wind  rain  area
0  7  5   mar  fri  86.200  26.200  94.300  5.100  8.200  51  6.700  0.000  0.000
1  7  4   oct  tue  90.600  35.400  669.100  6.700  18.000  33  0.900  0.000  0.000
2  7  4   oct  sat  90.600  43.700  686.900  6.700  14.600  33  1.300  0.000  0.000
3  8  6   mar  fri  91.700  33.300  77.500  9.000  8.300  97  4.000  0.200  0.000
4  8  6   mar  sun  89.300  51.300  102.200  9.600  11.400  99  1.800  0.000  0.000

```

2.1 Get some descriptive statistic of the data

```
In [4]: fires_attributes = fires.columns.values.tolist()
        number_of_columns = len(fires_attributes)
```

```
In [5]: fires.describe() #Original data
```

```

Out[5]:
      count      X      Y  FPMC    DMC    DC  ISI  temp  RH  wind  \
count  517.000  517.000  517.000  517.000  517.000  517.000  517.000  517.000  517.000
mean     4.669   4.300  90.645  110.872  547.940    9.022  18.889   44.288    4.018
std     2.314   1.230   5.520   64.046  248.066    4.559   5.807   16.317    1.792
min     1.000   2.000  18.700    1.100    7.900    0.000    2.200   15.000    0.400
25%     3.000   4.000  90.200   68.600  437.700    6.500   15.500   33.000    2.700
50%     4.000   4.000  91.600  108.300  664.200    8.400   19.300   42.000    4.000
75%     7.000   5.000  92.900  142.400  713.900   10.800   22.800   53.000    4.900
max     9.000   9.000  96.200  291.300  860.600   56.100   33.300  100.000    9.400

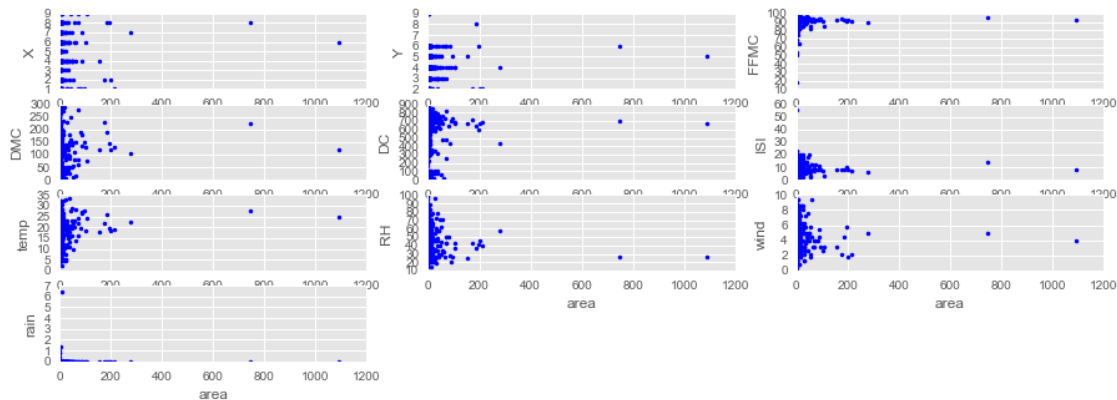
      rain      area
count  517.000  517.000
mean     0.022   12.847
std     0.296   63.656
min     0.000    0.000
25%     0.000    0.000
50%     0.000    0.520
75%     0.000    6.570
max     6.400  1090.840

```

2.2 Display a graph of quantitative variables vs area

```
In [6]: attributes = [0, 1] + list(range(4, number_of_columns - 1))
n_cols = 3
n_rows = int(ceil(len(attributes) / n_cols))
fig = plt.figure()
idx = 1
for attr in attributes:
    plt.subplot(n_rows, n_cols, idx)
    plt.plot(fires['area'], fires[fires_attributes[attr]], 'b.')
    # seaborn.regplot(x = fires['area'], y = fires[fires_attributes[attr]],
    #                 scatter = True, color = 'b', data = fires)
    plt.xlabel('area')
    plt.ylabel(fires_attributes[attr])
    idx += 1

plt.show()
```



There are some data values where the burned area is away from other values:

```
In [7]: fires[fires['area'] > 250]
```

```
Out[7]:
```

	X	Y	month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain
238	6	5	sep	sat	92.500	121.100	674.400	8.600	25.100	27	4.000	0.000
415	8	6	aug	thu	94.800	222.400	698.600	13.900	27.500	27	4.900	0.000
479	7	4	jul	mon	89.200	103.900	431.600	6.400	22.600	57	4.900	0.000


```

area
238 1090.840
415 746.280
479 278.530
```

2.3 Plot some other variables

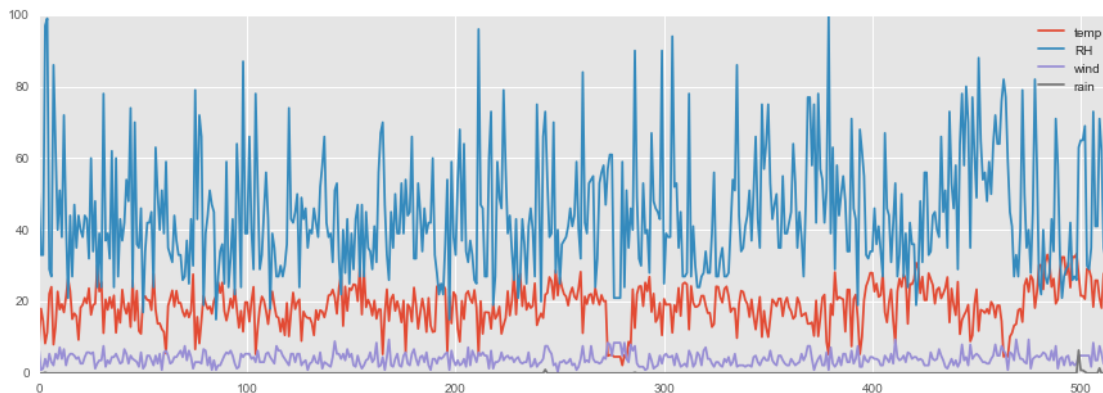
```
In [8]: scatter_matrix(fires, figsize = (15,10))
plt.show()
```



High bias are appreciated in **FFMC**, **DC**, **ISI**, **wind** and **area** variables

```
In [9]: fires[['temp', 'RH', 'wind', 'rain']].plot()    #Plot temperature, relative humidity, wind
                                                #and rain graphs
```

```
Out[9]: <matplotlib.axes._subplots.AxesSubplot at 0x1bae4eb7a20>
```



```
In [10]: fires.corr()    #Show correlation between variables
```

```
Out[10]:
```

	X	Y	FFMC	DMC	DC	ISI	temp	RH	wind	rain
X	1.000	0.540	-0.021	-0.048	-0.086	0.006	-0.051	0.085	0.019	0.065

Y	0.540	1.000	-0.046	0.008	-0.101	-0.024	-0.024	0.062	-0.020	0.033
FFMC	-0.021	-0.046	1.000	0.383	0.331	0.532	0.432	-0.301	-0.028	0.057
DMC	-0.048	0.008	0.383	1.000	0.682	0.305	0.470	0.074	-0.105	0.075
DC	-0.086	-0.101	0.331	0.682	1.000	0.229	0.496	-0.039	-0.203	0.036
ISI	0.006	-0.024	0.532	0.305	0.229	1.000	0.394	-0.133	0.107	0.068
temp	-0.051	-0.024	0.432	0.470	0.496	0.394	1.000	-0.527	-0.227	0.069
RH	0.085	0.062	-0.301	0.074	-0.039	-0.133	-0.527	1.000	0.069	0.100
wind	0.019	-0.020	-0.028	-0.105	-0.203	0.107	-0.227	0.069	1.000	0.061
rain	0.065	0.033	0.057	0.075	0.036	0.068	0.069	0.100	0.061	1.000
area	0.063	0.045	0.040	0.073	0.049	0.008	0.098	-0.076	0.012	-0.007

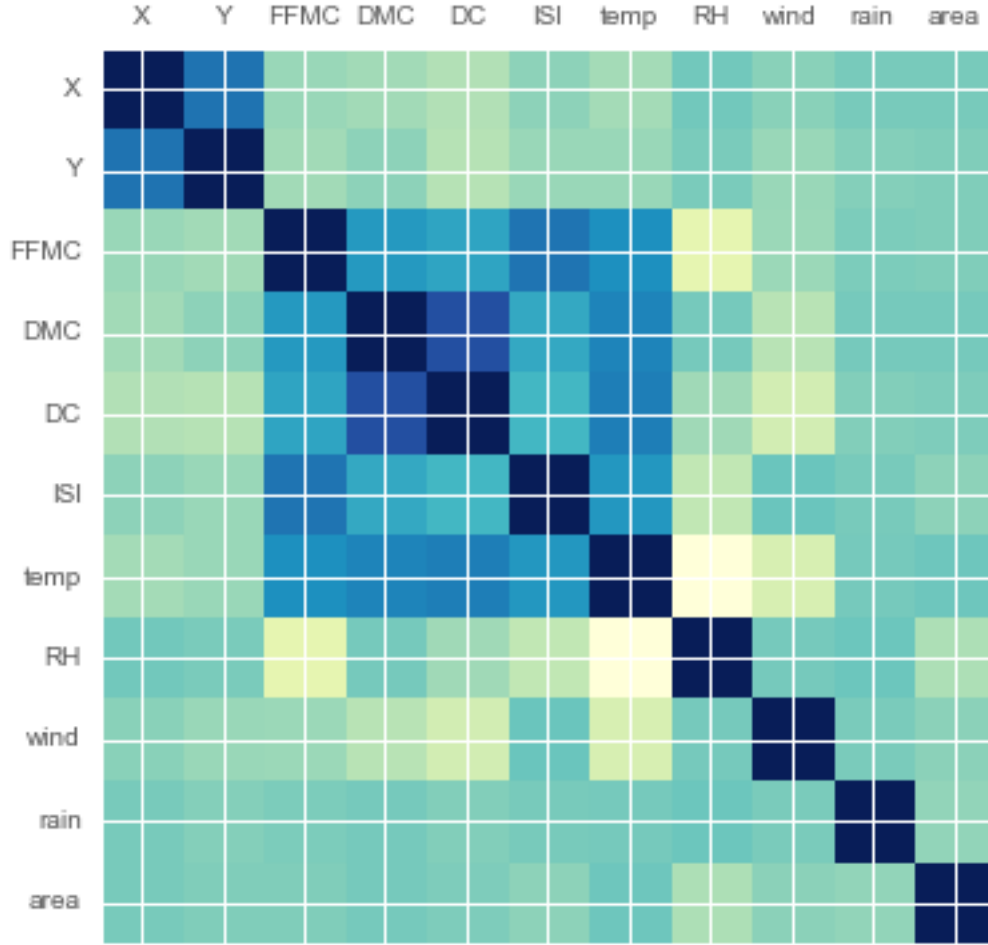
	area
X	0.063
Y	0.045
FFMC	0.040
DMC	0.073
DC	0.049
ISI	0.008
temp	0.098
RH	-0.076
wind	0.012
rain	-0.007
area	1.000

```
In [11]: def plot_corr(df, size=10):
    '''Function plots a graphical correlation matrix for each pair of columns
        in the dataframe, including the names of the attributes
    Input:
        df: pandas DataFrame
        size: vertical and horizontal size of the plot

    Code taken from:
    http://stackoverflow.com/questions/29432629/correlation-matrix-using-pandas
    '''

    corr = df.corr()
    fig, ax = plt.subplots(figsize=(size, size))
    ax.matshow(corr, cmap = 'YlGnBu')
    plt.xticks(range(len(corr.columns)), corr.columns);
    plt.yticks(range(len(corr.columns)), corr.columns);

    #plt.matshow(fires.corr())
    plot_corr(fires, size = 6)
```



There is a medium-high correlation (**0.682**) between **DC** (Drought Code: numeric rating of the average moisture content of deep, compact organic layers) and **DMC** (Duff Moisture Code: numeric rating of the average moisture content of loosely compacted organic layers of moderate depth) and medium correlation (**0.532**) between **ISI** (Initial Spread Index: numeric rating of the expected rate of fire spread) and **FPMC** (Fine Fuel Moisture Code: numeric rating of the moisture content of litter and other cured fine fuels). Also, there is a inverse medium correlation (**-0.527**) between temperature (**temp**) and relative humidity (**RH**). Other relationships are noted between temperature (**temp**) and FWI system components (**FPMC**, **DCM**, **DC** and **ISI**)

3 Linear regression

3.1 Convert categorical variables (months and days) into numerical values

```
In [12]: months_table = ['jan', 'feb', 'mar', 'apr', 'may', 'jun',
                        'jul', 'aug', 'sep', 'oct', 'nov', 'dec']
        days_table = ['sun', 'mon', 'tue', 'wed', 'thu', 'fri', 'sat']

        fires['month'] = [months_table.index(month) for month in fires['month']]
        fires['day'] = [days_table.index(day) for day in fires['day']]
```

```
fires['X'] -= 1
fires['Y'] -= 2

fires.head()
```

```
Out[12]:
```

	X	Y	month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain	area
0	6	3	2	5	86.200	26.200	94.300	5.100	8.200	51	6.700	0.000	0.000
1	6	2	9	2	90.600	35.400	669.100	6.700	18.000	33	0.900	0.000	0.000
2	6	2	9	6	90.600	43.700	686.900	6.700	14.600	33	1.300	0.000	0.000
3	7	4	2	5	91.700	33.300	77.500	9.000	8.300	97	4.000	0.200	0.000
4	7	4	2	0	89.300	51.300	102.200	9.600	11.400	99	1.800	0.000	0.000

3.2 Center each explanatory variable

```
In [13]: for idx in list(range(4, number_of_columns - 1)): #Exclude categorical variables
        fires[fires_attributes[idx]] = fires[fires_attributes[idx]] - \
            fires[fires_attributes[idx]].mean()

In [14]: fires.describe() #Only quantitative explanatory variables (FFMC thru rain) were centered
```

```
Out[14]:
```

	X	Y	month	day	FFMC	DMC	DC	ISI	\
count	517.000	517.000	517.000	517.000	517.000	517.000	517.000	517.000	517.000
mean	3.669	2.300	6.476	2.973	0.000	-0.000	0.000	-0.000	
std	2.314	1.230	2.276	2.144	5.520	64.046	248.066	4.559	
min	0.000	0.000	0.000	0.000	-71.945	-109.772	-540.040	-9.022	
25%	2.000	2.000	6.000	1.000	-0.445	-42.272	-110.240	-2.522	
50%	3.000	2.000	7.000	3.000	0.955	-2.572	116.260	-0.622	
75%	6.000	3.000	8.000	5.000	2.255	31.528	165.960	1.778	
max	8.000	7.000	11.000	6.000	5.555	180.428	312.660	47.078	

	temp	RH	wind	rain	area
count	517.000	517.000	517.000	517.000	517.000
mean	0.000	0.000	-0.000	0.000	12.847
std	5.807	16.317	1.792	0.296	63.656
min	-16.689	-29.288	-3.618	-0.022	0.000
25%	-3.389	-11.288	-1.318	-0.022	0.000
50%	0.411	-2.288	-0.018	-0.022	0.520
75%	3.911	8.712	0.882	-0.022	6.570
max	14.411	55.712	5.382	6.378	1090.840

3.3 Generate models to test each variable

```
In [15]: def print_title(title):
        print('+ ' + "-" * (len(title) + 2) + ' ' + '\n' +
              '| ' + title + ' | ' + '\n' +
              '+ ' + "-" * (len(title) + 2) + ' ')

In [16]: statistics = list()
        for idx in range(0, number_of_columns - 1):
            model = smf.ols(formula = "area ~ " +
                            fires_attributes[idx], data = fires).fit()

            print_title('Model: area ~ ' + fires_attributes[idx])
            print()
            print(model.summary())
```

```
print()
statistics.append([model.f_pvalue, model.rsquared])
```

```
+-----+
| Model: area ~ X |
+-----+
```

OLS Regression Results

```
=====
Dep. Variable:          area    R-squared:                0.004
Model:                  OLS      Adj. R-squared:           0.002
Method:                 Least Squares    F-statistic:          2.077
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):    0.150
Time:                  22:34:25    Log-Likelihood:       -2879.4
No. Observations:      517    AIC:                  5763.
Df Residuals:          515    BIC:                  5771.
Df Model:               1
Covariance Type:       nonrobust
=====
```

```
=====
              coef    std err          t      P>|t|      [95.0% Conf. Int.]
-----
Intercept    6.4487      5.247      1.229     0.220      -3.859      16.756
X             1.7438      1.210      1.441     0.150      -0.633       4.121
=====
```

```
=====
Omnibus:                 981.662    Durbin-Watson:           1.653
Prob(Omnibus):            0.000    Jarque-Bera (JB):        802838.467
Skew:                     12.752    Prob(JB):                 0.00
Kurtosis:                 194.360    Cond. No.                 8.45
=====
```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```
+-----+
| Model: area ~ Y |
+-----+
```

OLS Regression Results

```
=====
Dep. Variable:          area    R-squared:                0.002
Model:                  OLS      Adj. R-squared:           0.000
Method:                 Least Squares    F-statistic:          1.039
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):    0.309
Time:                  22:34:25    Log-Likelihood:       -2879.9
No. Observations:      517    AIC:                  5764.
Df Residuals:          515    BIC:                  5772.
Df Model:               1
Covariance Type:       nonrobust
=====
```

```
=====
              coef    std err          t      P>|t|      [95.0% Conf. Int.]
-----
Intercept    7.5060      5.941      1.263     0.207      -4.165      19.177
Y             2.3225      2.278      1.019     0.309      -2.154       6.799
=====
```



```

Omnibus:                981.970    Durbin-Watson:                1.645
Prob(Omnibus):           0.000    Jarque-Bera (JB):           802937.403
Skew:                    12.761    Prob(JB):                    0.00
Kurtosis:                194.369    Cond. No.                    6.19
=====

```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```

+-----+
| Model: area ~ month |
+-----+

```

OLS Regression Results

```

=====
Dep. Variable:          area    R-squared:                0.003
Model:                  OLS     Adj. R-squared:           0.001
Method:                 Least Squares    F-statistic:             1.649
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):       0.200
Time:                  22:34:26    Log-Likelihood:          -2879.6
No. Observations:      517    AIC:                     5763.
Df Residuals:          515    BIC:                     5772.
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	2.6149	8.445	0.310	0.757	-13.976 19.206
month	1.5801	1.230	1.284	0.200	-0.837 3.997

```

=====
Omnibus:                983.027    Durbin-Watson:                1.647
Prob(Omnibus):           0.000    Jarque-Bera (JB):           807389.375
Skew:                    12.790    Prob(JB):                    0.00
Kurtosis:                194.901    Cond. No.                    21.1
=====

```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```

+-----+
| Model: area ~ day |
+-----+

```

OLS Regression Results

```

=====
Dep. Variable:          area    R-squared:                0.002
Model:                  OLS     Adj. R-squared:           0.000
Method:                 Least Squares    F-statistic:             1.207
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):       0.272
Time:                  22:34:26    Log-Likelihood:          -2879.8
No. Observations:      517    AIC:                     5764.
Df Residuals:          515    BIC:                     5772.
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[95.0% Conf. Int.]	
Intercept	8.5785	4.788	1.792	0.074	-0.829	17.986
day	1.4359	1.307	1.099	0.272	-1.132	4.003
Omnibus:		980.555	Durbin-Watson:			1.636
Prob(Omnibus):		0.000	Jarque-Bera (JB):			794438.352
Skew:		12.725	Prob(JB):			0.00
Kurtosis:		193.346	Cond. No.			6.58

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```

+-----+
| Model: area ~ FPMC |
+-----+

```

OLS Regression Results

```

=====
Dep. Variable:          area    R-squared:                0.002
Model:                  OLS      Adj. R-squared:           -0.000
Method:                 Least Squares    F-statistic:         0.8304
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):    0.363
Time:                  22:34:26    Log-Likelihood:       -2880.0
No. Observations:      517      AIC:                  5764.
Df Residuals:          515      BIC:                  5773.
Df Model:               1
Covariance Type:       nonrobust
=====

```

	coef	std err	t	P> t	[95.0% Conf. Int.]	
Intercept	12.8473	2.800	4.588	0.000	7.346	18.348
FFMC	0.4627	0.508	0.911	0.363	-0.535	1.460
Omnibus:		983.137	Durbin-Watson:			1.649
Prob(Omnibus):		0.000	Jarque-Bera (JB):			808340.065
Skew:		12.793	Prob(JB):			0.00
Kurtosis:		195.015	Cond. No.			5.51

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```

+-----+
| Model: area ~ DMC |
+-----+

```

OLS Regression Results

```

=====
Dep. Variable:          area    R-squared:                0.005
Model:                  OLS      Adj. R-squared:           0.003
Method:                 Least Squares    F-statistic:         2.759
=====

```

```

Date:                Tue, 14 Jun 2016    Prob (F-statistic):        0.0973
Time:                22:34:26           Log-Likelihood:           -2879.1
No. Observations:    517                AIC:                     5762.
Df Residuals:        515                BIC:                     5771.
Df Model:            1
Covariance Type:     nonrobust

```

```

=====
              coef      std err          t      P>|t|      [95.0% Conf. Int.]
-----+-----
Intercept    12.8473      2.795      4.597      0.000       7.357    18.338
DMC           0.0725      0.044      1.661      0.097      -0.013    0.158
=====
Omnibus:                 982.803    Durbin-Watson:                 1.649
Prob(Omnibus):            0.000    Jarque-Bera (JB):            811231.935
Skew:                     12.780    Prob(JB):                     0.00
Kurtosis:                 195.368    Cond. No.                     64.0
=====

```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```

+-----+
| Model: area ~ DC |
+-----+

```

OLS Regression Results

```

=====
Dep. Variable:          area    R-squared:                0.002
Model:                  OLS     Adj. R-squared:            0.001
Method:                 Least Squares    F-statistic:            1.259
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):      0.262
Time:                  22:34:26           Log-Likelihood:          -2879.8
No. Observations:      517            AIC:                     5764.
Df Residuals:          515            BIC:                     5772.
Df Model:              1
Covariance Type:       nonrobust

```

```

=====
              coef      std err          t      P>|t|      [95.0% Conf. Int.]
-----+-----
Intercept    12.8473      2.799      4.590      0.000       7.349    18.346
DC            0.0127      0.011      1.122      0.262      -0.010    0.035
=====
Omnibus:                 982.892    Durbin-Watson:                 1.645
Prob(Omnibus):            0.000    Jarque-Bera (JB):            807312.305
Skew:                     12.786    Prob(JB):                     0.00
Kurtosis:                 194.893    Cond. No.                     248.
=====

```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```

+-----+
| Model: area ~ ISI |
+-----+

```

```

                                OLS Regression Results
=====
Dep. Variable:                  area    R-squared:                  0.000
Model:                          OLS      Adj. R-squared:             -0.002
Method:                        Least Squares  F-statistic:                0.03512
Date:                          Tue, 14 Jun 2016  Prob (F-statistic):        0.851
Time:                          22:34:26    Log-Likelihood:            -2880.4
No. Observations:              517      AIC:                      5765.
Df Residuals:                  515      BIC:                      5773.
Df Model:                      1
Covariance Type:               nonrobust
=====

               coef      std err          t      P>|t|      [95.0% Conf. Int.]
-----
Intercept    12.8473      2.802      4.585      0.000      7.342    18.352
ISI          0.1153      0.615      0.187      0.851     -1.093    1.324
=====

Omnibus:                 983.625    Durbin-Watson:           1.649
Prob(Omnibus):            0.000    Jarque-Bera (JB):        809992.277
Skew:                    12.806    Prob(JB):                0.00
Kurtosis:                195.211    Cond. No.                4.56
=====

```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```

+-----+
| Model: area ~ temp |
+-----+

```

```

                                OLS Regression Results
=====
Dep. Variable:                  area    R-squared:                  0.010
Model:                          OLS      Adj. R-squared:             0.008
Method:                        Least Squares  F-statistic:                4.978
Date:                          Tue, 14 Jun 2016  Prob (F-statistic):        0.0261
Time:                          22:34:26    Log-Likelihood:            -2878.0
No. Observations:              517      AIC:                      5760.
Df Residuals:                  515      BIC:                      5768.
Df Model:                      1
Covariance Type:               nonrobust
=====

               coef      std err          t      P>|t|      [95.0% Conf. Int.]
-----
Intercept    12.8473      2.789      4.607      0.000      7.368    18.326
temp         1.0726      0.481      2.231      0.026      0.128    2.017
=====

Omnibus:                 979.270    Durbin-Watson:           1.650
Prob(Omnibus):            0.000    Jarque-Bera (JB):        793772.021
Skew:                    12.687    Prob(JB):                0.00
Kurtosis:                193.275    Cond. No.                5.80
=====

```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```
+-----+
| Model: area ~ RH |
+-----+
```

OLS Regression Results

```
=====
Dep. Variable:          area    R-squared:                0.006
Model:                  OLS     Adj. R-squared:            0.004
Method:                 Least Squares    F-statistic:        2.954
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):    0.0863
Time:                  22:34:26    Log-Likelihood:       -2879.0
No. Observations:      517    AIC:                  5762.
Df Residuals:          515    BIC:                  5770.
Df Model:               1
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	12.8473	2.794	4.598	0.000	7.358 18.337
RH	-0.2946	0.171	-1.719	0.086	-0.631 0.042

```
=====
Omnibus:                980.422    Durbin-Watson:        1.642
Prob(Omnibus):          0.000    Jarque-Bera (JB):     795947.965
Skew:                   12.720    Prob(JB):              0.00
Kurtosis:               193.531    Cond. No.              16.3
=====
```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

```
+-----+
| Model: area ~ wind |
+-----+
```

OLS Regression Results

```
=====
Dep. Variable:          area    R-squared:                0.000
Model:                  OLS     Adj. R-squared:           -0.002
Method:                 Least Squares    F-statistic:        0.07815
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):    0.780
Time:                  22:34:26    Log-Likelihood:       -2880.4
No. Observations:      517    AIC:                  5765.
Df Residuals:          515    BIC:                  5773.
Df Model:               1
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	12.8473	2.802	4.585	0.000	7.342 18.352
wind	0.4376	1.565	0.280	0.780	-2.638 3.513

```

Omnibus:                983.721    Durbin-Watson:                1.647
Prob(Omnibus):           0.000    Jarque-Bera (JB):         810324.708
Skew:                   12.809    Prob(JB):                 0.00
Kurtosis:               195.251    Cond. No.                 1.79
=====

```

Warnings:

```
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
```

```

+-----+
| Model: area ~ rain |
+-----+

```

OLS Regression Results

```

=====
Dep. Variable:          area    R-squared:                0.000
Model:                  OLS    Adj. R-squared:           -0.002
Method:                 Least Squares    F-statistic:            0.02794
Date:                   Tue, 14 Jun 2016    Prob (F-statistic):      0.867
Time:                   22:34:26    Log-Likelihood:         -2880.4
No. Observations:       517    AIC:                    5765.
Df Residuals:           515    BIC:                    5773.
Df Model:                1
Covariance Type:        nonrobust
=====

```

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	12.8473	2.802	4.585	0.000	7.342 18.352
rain	-1.5842	9.477	-0.167	0.867	-20.203 17.035

```

=====
Omnibus:                983.726    Durbin-Watson:                1.649
Prob(Omnibus):           0.000    Jarque-Bera (JB):         810320.385
Skew:                   12.809    Prob(JB):                 0.00
Kurtosis:               195.250    Cond. No.                 3.38
=====

```

Warnings:

```
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
```

3.3.1 Summary:

```

In [17]: statistics = pandas.DataFrame(statistics,
                                         index=fires_attributes[: number_of_columns - 1],
                                         columns=['p-value', 'R-squared'])

statistics.T

```

```

Out[17]:
           X      Y  month  day  FFMC  DMC  DC  ISI  temp  RH  wind \
p-value  0.150  0.309  0.200  0.272  0.363  0.097  0.262  0.851  0.026  0.086  0.780
R-squared 0.004  0.002  0.003  0.002  0.002  0.005  0.002  0.000  0.010  0.006  0.000

           rain
p-value    0.867
R-squared  0.000

```

```
In [18]: statistics[statistics['p-value'] < 0.05]
```

```
Out[18]:          p-value  R-squared
          temp      0.026      0.010
```

‘temp’ is the only statistically significant variable (p-value = 0.026) but it only explains the 1% of forest fires. Let’s show its linear model summary:

```
In [19]: print((smf.ols(formula = "area ~ temp", data = fires).fit()).summary())
```

OLS Regression Results

```
=====
Dep. Variable:          area    R-squared:                0.010
Model:                  OLS    Adj. R-squared:            0.008
Method:                 Least Squares    F-statistic:        4.978
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):    0.0261
Time:                  22:34:26    Log-Likelihood:        -2878.0
No. Observations:      517    AIC:                  5760.
Df Residuals:          515    BIC:                  5768.
Df Model:              1
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	12.8473	2.789	4.607	0.000	7.368 18.326
temp	1.0726	0.481	2.231	0.026	0.128 2.017

```
=====
Omnibus:                979.270    Durbin-Watson:          1.650
Prob(Omnibus):           0.000    Jarque-Bera (JB):        793772.021
Skew:                   12.687    Prob(JB):                0.00
Kurtosis:               193.275    Cond. No.:               5.80
=====
```

Warnings:

```
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
```

The results of the linear regression models indicated that only temperature (Beta = 1.0726, $p = 0.026$, $R^2 = 0.010$) was significantly and positively associated with the total burned area due to forest fires. ‘p-value’ of other models are greater than threshold value of 0.05 so results are not statistically significant to reject null hypothesis.

3.4 Create a Linear Regression Model for a combination of all variables

```
In [20]: explanatory_variables = "X + Y + month + day + FPMC + DMC + DC + ISI + temp + RH + " + \
          "wind + rain"
          response_variable = "area"

          model = smf.ols(formula = response_variable + " ~ " + explanatory_variables,
                          data = fires).fit()
```

```
In [21]: print(model.summary())
```

OLS Regression Results

```
=====
Dep. Variable:          area    R-squared:                0.025
Model:                  OLS    Adj. R-squared:            0.002
Method:                 Least Squares    F-statistic:        1.092
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):    0.364
```

```

Time:                22:34:27   Log-Likelihood:        -2873.8
No. Observations:    517       AIC:                5774.
Df Residuals:        504       BIC:                5829.
Df Model:            12
Covariance Type:     nonrobust

```

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	-17.5974	19.340	-0.910	0.363	-55.595 20.400
X	1.9002	1.450	1.311	0.191	-0.948 4.748
Y	0.3241	2.754	0.118	0.906	-5.086 5.734
month	2.9004	2.791	1.039	0.299	-2.583 8.384
day	1.3269	1.320	1.005	0.315	-1.267 3.921
FFMC	-0.1127	0.663	-0.170	0.865	-1.415 1.190
DMC	0.0966	0.071	1.369	0.172	-0.042 0.235
DC	-0.0315	0.032	-0.981	0.327	-0.095 0.032
ISI	-0.7305	0.772	-0.947	0.344	-2.247 0.786
temp	0.9546	0.797	1.198	0.232	-0.612 2.521
RH	-0.1758	0.241	-0.730	0.466	-0.649 0.297
wind	1.2321	1.702	0.724	0.470	-2.113 4.577
rain	-3.1958	9.683	-0.330	0.742	-22.220 15.829
Omnibus:	972.663	Durbin-Watson:	1.643		
Prob(Omnibus):	0.000	Jarque-Bera (JB):	769640.593		
Skew:	12.508	Prob(JB):	0.00		
Kurtosis:	190.356	Cond. No.	1.76e+03		

Warnings:

```

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
[2] The condition number is large, 1.76e+03. This might indicate that there are
strong multicollinearity or other numerical problems.

```

p-value of combination model ($p = 0.410$) is bigger than threshold value, so the combination of the Canadian Forest Fire Weather Index (FWI) system plus temperature, humidity, wind and rain are not significantly associated with the total burned area due to forest fires. p-value of temperature in combination model ($p = 0.282$) is not longer statistically significant, a confounder variable?

4 Test a Multiple Regression Model

4.1 Sort explanatory variables by p-value

```
In [22]: statistics = statistics.sort_values(by='p-value')
```

4.2 Define an useful function to plot QQ and Residual plots

```
In [23]: def print_qqplot_and_residuals_plot(model):
# qq-plot
ax1 = plt.subplot(1, 3, 1)
qq_plot = sm.qqplot(model.resid, line = 'r', ax = ax1)

# Residuals plot
ax2 = plt.subplot(1, 3, 2)
```



```

stdres = pandas.DataFrame(model.resid_pearson)
residuals_plot = plt.plot(stdres, 'o', ls = 'None')
plt.axhline(y = 0, color = 'r')
plt.ylabel('Standardized Residual')
plt.xlabel('Observation Number')

plt.show()

```

4.3 Generate linear models adding one explanatory variable a time

```

In [24]: explanatory_variables = None
         response_variable = "area"

         saved_models = list()

         for variable in list(statistics[: number_of_columns - 1].index.values):
             if explanatory_variables == None:
                 explanatory_variables = variable
             else:
                 explanatory_variables += " + " + variable
             model = smf.ols(formula = response_variable + " ~ " + explanatory_variables,
                             data = fires).fit()
             saved_models.append(model)

         print_title('Model: ' + response_variable + " ~ " + explanatory_variables)
         print()
         print(model.summary())
         print_qqplot_and_residuals_plot(model)
         print()

```

```

+-----+
| Model: area ~ temp |
+-----+

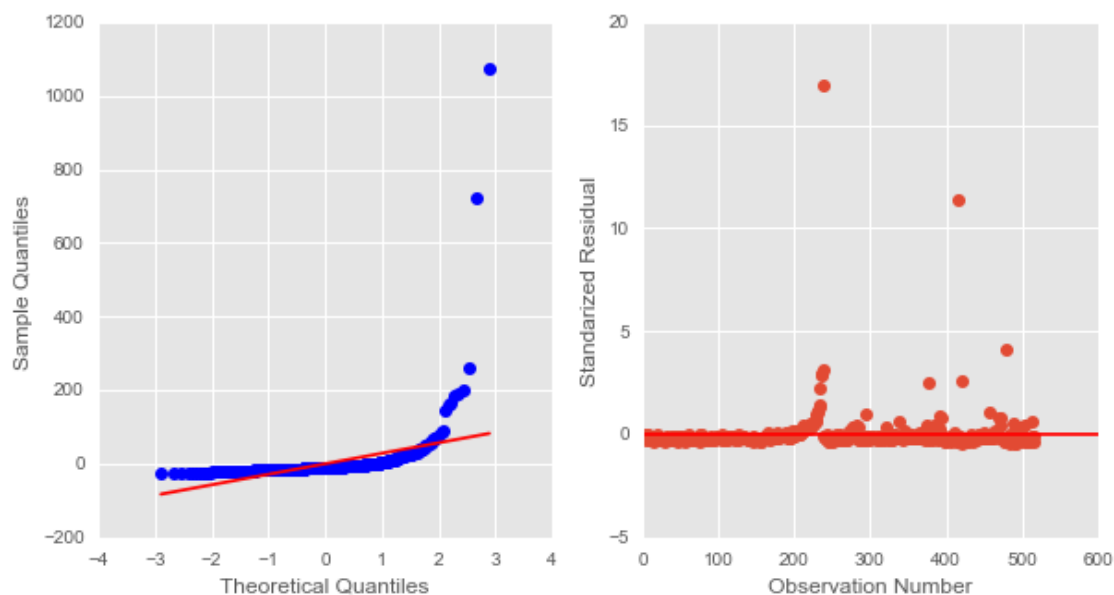
```

OLS Regression Results						
=====						
Dep. Variable:	area	R-squared:	0.010			
Model:	OLS	Adj. R-squared:	0.008			
Method:	Least Squares	F-statistic:	4.978			
Date:	Tue, 14 Jun 2016	Prob (F-statistic):	0.0261			
Time:	22:34:27	Log-Likelihood:	-2878.0			
No. Observations:	517	AIC:	5760.			
Df Residuals:	515	BIC:	5768.			
Df Model:	1					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[95.0% Conf. Int.]	

Intercept	12.8473	2.789	4.607	0.000	7.368	18.326
temp	1.0726	0.481	2.231	0.026	0.128	2.017
=====						
Omnibus:	979.270	Durbin-Watson:	1.650			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	793772.021			
Skew:	12.687	Prob(JB):	0.00			
Kurtosis:	193.275	Cond. No.	5.80			

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.



```
+-----+
| Model: area ~ temp + RH |
+-----+
```

OLS Regression Results

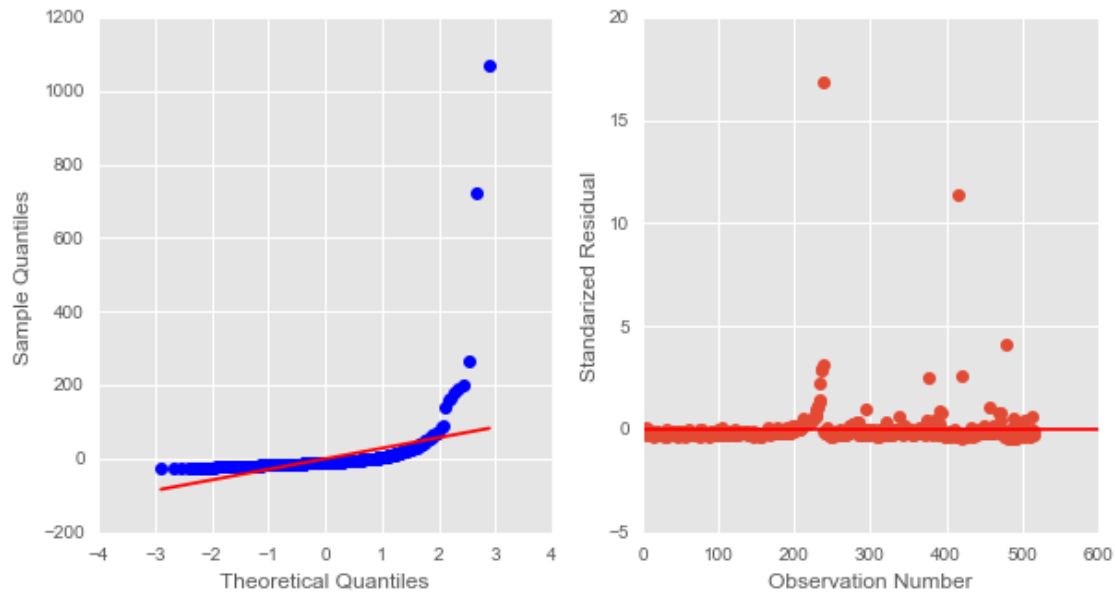
```
=====
Dep. Variable:          area    R-squared:                0.010
Model:                  OLS     Adj. R-squared:           0.007
Method:                 Least Squares    F-statistic:         2.692
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):   0.0687
Time:                  22:34:28    Log-Likelihood:      -2877.8
No. Observations:      517          AIC:                5762.
Df Residuals:          514          BIC:                5774.
Df Model:               2
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[95.0% Conf. Int.]	
Intercept	12.8473	2.790	4.604	0.000	7.365	18.329
temp	0.8811	0.566	1.556	0.120	-0.231	1.993
RH	-0.1293	0.201	-0.642	0.521	-0.525	0.267

```
=====
Omnibus:                978.601    Durbin-Watson:         1.648
Prob(Omnibus):           0.000    Jarque-Bera (JB):      790442.645
Skew:                    12.669    Prob(JB):               0.00
Kurtosis:                192.873    Cond. No.               16.6
=====
```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.



```
+-----+
| Model: area ~ temp + RH + DMC |
+-----+
```

OLS Regression Results

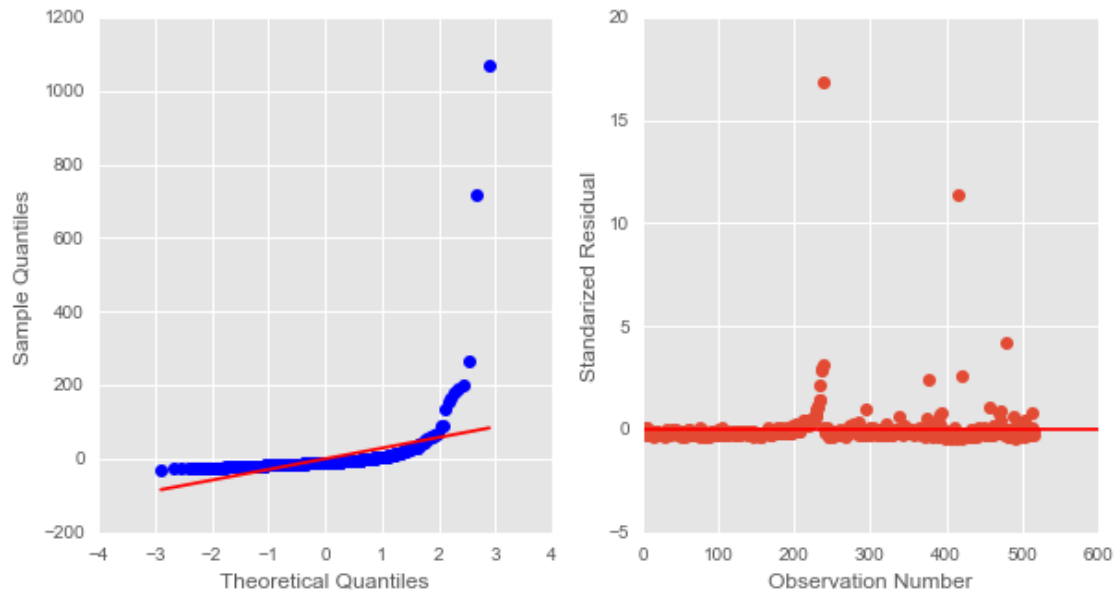
```
=====
Dep. Variable:          area    R-squared:                0.013
Model:                  OLS     Adj. R-squared:           0.007
Method:                 Least Squares   F-statistic:           2.182
Date:                   Tue, 14 Jun 2016   Prob (F-statistic):    0.0892
Time:                   22:34:29   Log-Likelihood:       -2877.2
No. Observations:       517         AIC:                  5762.
Df Residuals:           513         BIC:                  5779.
Df Model:                3
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[95.0% Conf. Int.]	
Intercept	12.8473	2.790	4.605	0.000	7.366	18.329
temp	0.4236	0.708	0.599	0.550	-0.967	1.814
RH	-0.2322	0.223	-1.041	0.298	-0.670	0.206
DMC	0.0589	0.055	1.077	0.282	-0.049	0.166

```
=====
Omnibus:                978.623   Durbin-Watson:          1.644
Prob(Omnibus):           0.000   Jarque-Bera (JB):       793249.169
Skew:                    12.668   Prob(JB):                0.00
Kurtosis:                193.216   Cond. No.                64.1
=====
```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.



```
+-----+
| Model: area ~ temp + RH + DMC + X |
+-----+
```

OLS Regression Results

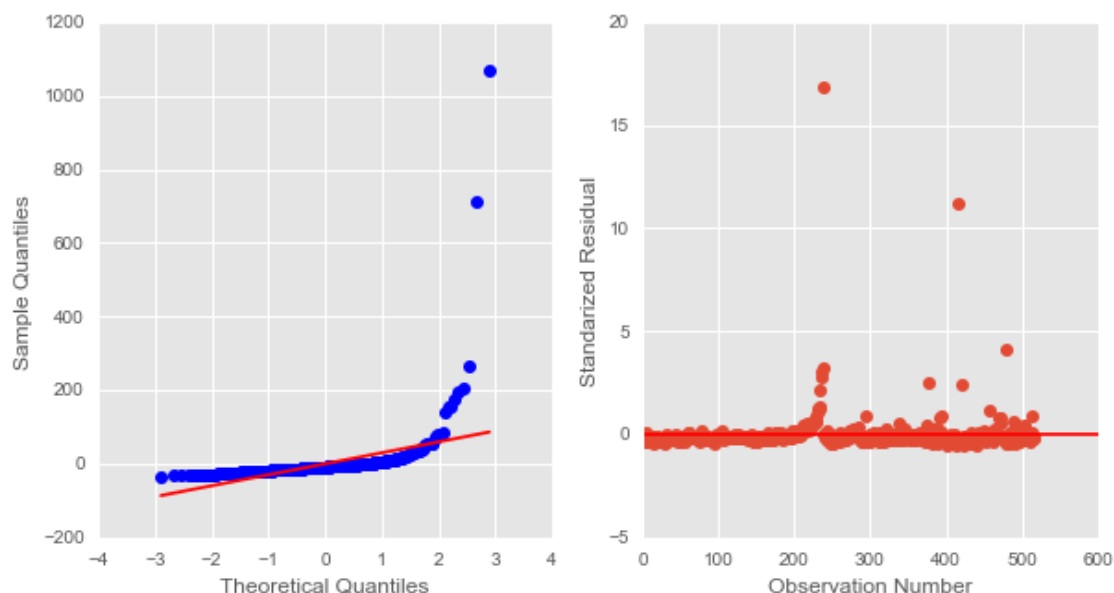
```
=====
Dep. Variable:          area    R-squared:                0.018
Model:                  OLS     Adj. R-squared:           0.010
Method:                 Least Squares   F-statistic:           2.351
Date:                  Tue, 14 Jun 2016   Prob (F-statistic):    0.0532
Time:                  22:34:30    Log-Likelihood:       -2875.7
No. Observations:      517          AIC:                  5761.
Df Residuals:          512          BIC:                  5783.
Df Model:               4
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[95.0% Conf. Int.]	
Intercept	5.3630	5.246	1.022	0.307	-4.944	15.670
temp	0.3855	0.707	0.545	0.586	-1.003	1.774
RH	-0.2656	0.223	-1.189	0.235	-0.705	0.173
DMC	0.0647	0.055	1.183	0.237	-0.043	0.172
X	2.0397	1.212	1.683	0.093	-0.341	4.420

```
=====
Omnibus:                975.911    Durbin-Watson:           1.648
Prob(Omnibus):           0.000     Jarque-Bera (JB):       783930.607
Skew:                   12.593     Prob(JB):               0.00
Kurtosis:               192.095    Cond. No.:              123.
=====
```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.



```
+-----+
| Model: area ~ temp + RH + DMC + X + month |
+-----+
```

OLS Regression Results

```
=====
Dep. Variable:          area    R-squared:                0.018
Model:                  OLS     Adj. R-squared:           0.009
Method:                 Least Squares   F-statistic:          1.896
Date:                   Tue, 14 Jun 2016   Prob (F-statistic):    0.0935
Time:                   22:34:31   Log-Likelihood:       -2875.7
No. Observations:       517         AIC:                  5763.
Df Residuals:           511         BIC:                  5789.
Df Model:                5
Covariance Type:        nonrobust
=====
```

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	2.5422	10.711	0.237	0.812	-18.501 23.585
temp	0.3575	0.713	0.501	0.617	-1.044 1.759
RH	-0.2637	0.224	-1.179	0.239	-0.703 0.176
DMC	0.0588	0.058	1.012	0.312	-0.055 0.173
X	2.0544	1.214	1.693	0.091	-0.330 4.439
month	0.4273	1.414	0.302	0.763	-2.351 3.206

```
=====
Omnibus:                 975.906   Durbin-Watson:           1.647
Prob(Omnibus):            0.000   Jarque-Bera (JB):        783473.752
=====
```

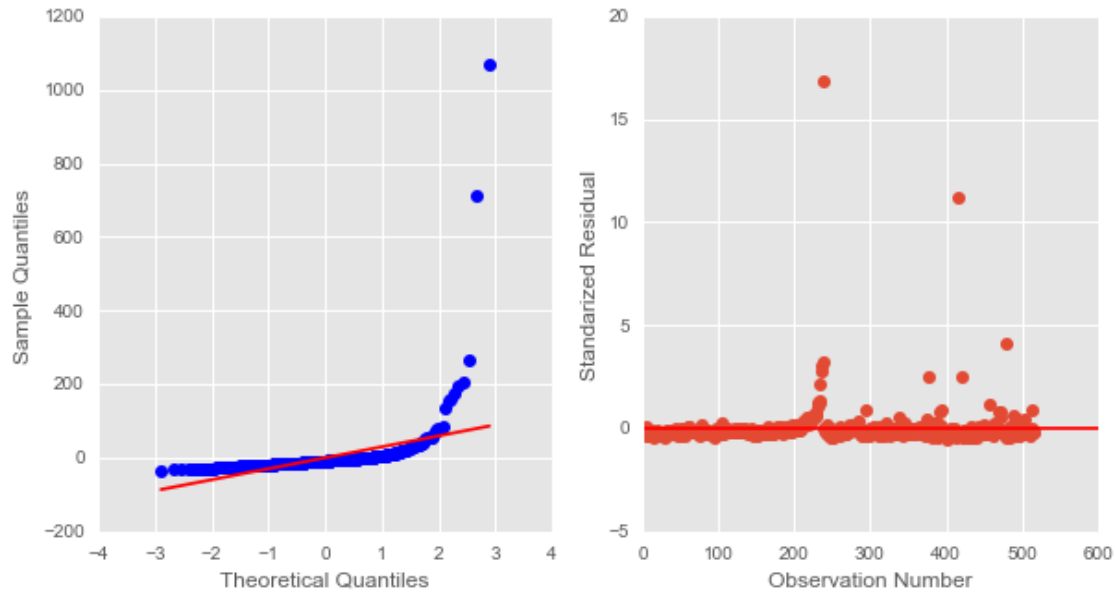
```

Skew:                12.594    Prob(JB):                0.00
Kurtosis:            192.039    Cond. No.                248.
=====

```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.



```

+-----+
| Model: area ~ temp + RH + DMC + X + month + DC |
+-----+

```

OLS Regression Results

```

=====
Dep. Variable:          area    R-squared:                0.020
Model:                  OLS     Adj. R-squared:           0.009
Method:                 Least Squares    F-statistic:          1.758
Date:                   Tue, 14 Jun 2016    Prob (F-statistic):    0.106
Time:                   22:34:32    Log-Likelihood:       -2875.2
No. Observations:       517          AIC:                  5764.
Df Residuals:           510          BIC:                  5794.
Df Model:                6
Covariance Type:        nonrobust
=====

```

	coef	std err	t	P> t	[95.0% Conf. Int.]	
Intercept	-12.4210	18.017	-0.689	0.491	-47.818	22.976
temp	0.5717	0.743	0.770	0.442	-0.888	2.031
RH	-0.2196	0.228	-0.964	0.335	-0.667	0.228
DMC	0.0929	0.067	1.390	0.165	-0.038	0.224
X	1.9615	1.217	1.612	0.108	-0.429	4.352
month	2.7906	2.690	1.037	0.300	-2.494	8.075
DC	-0.0316	0.031	-1.033	0.302	-0.092	0.028

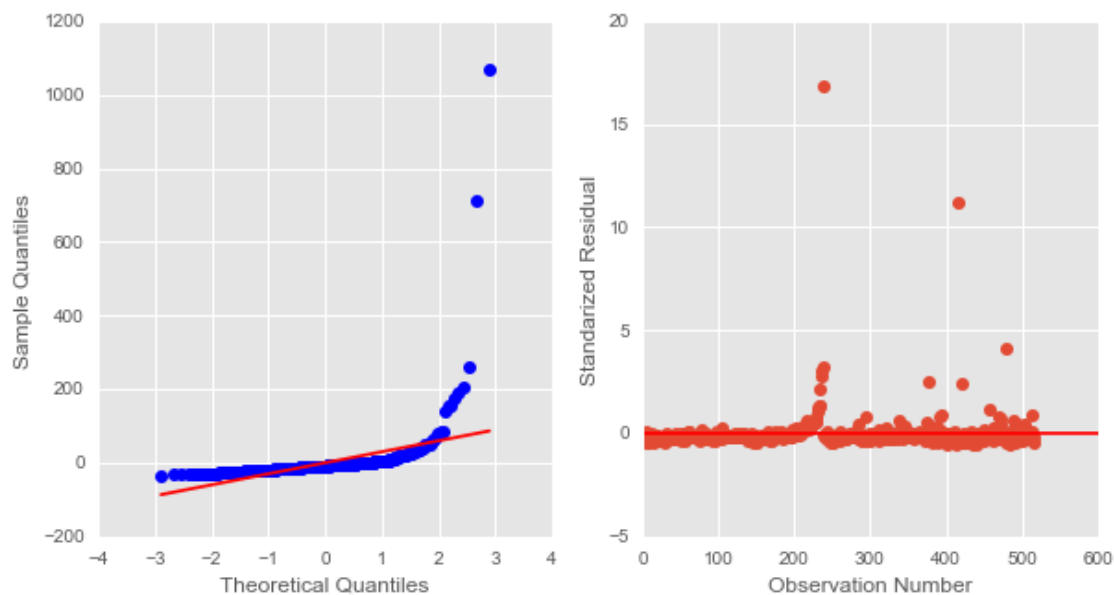
```

=====
Omnibus:                976.248    Durbin-Watson:                1.655
Prob(Omnibus):           0.000    Jarque-Bera (JB):            786247.138
Skew:                    12.602    Prob(JB):                     0.00
Kurtosis:                192.377    Cond. No.                     1.64e+03
=====

```

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
[2] The condition number is large, 1.64e+03. This might indicate that there are strong multicollinearity or other numerical problems.



```

+-----+
| Model: area ~ temp + RH + DMC + X + month + DC + day |
+-----+

```

OLS Regression Results

```

=====
Dep. Variable:          area    R-squared:                0.022
Model:                  OLS     Adj. R-squared:           0.009
Method:                 Least Squares    F-statistic:             1.642
Date:                  Tue, 14 Jun 2016    Prob (F-statistic):       0.121
Time:                  22:34:33    Log-Likelihood:          -2874.7
No. Observations:      517    AIC:                     5765.
Df Residuals:          509    BIC:                     5799.
Df Model:               7
Covariance Type:       nonrobust
=====

```

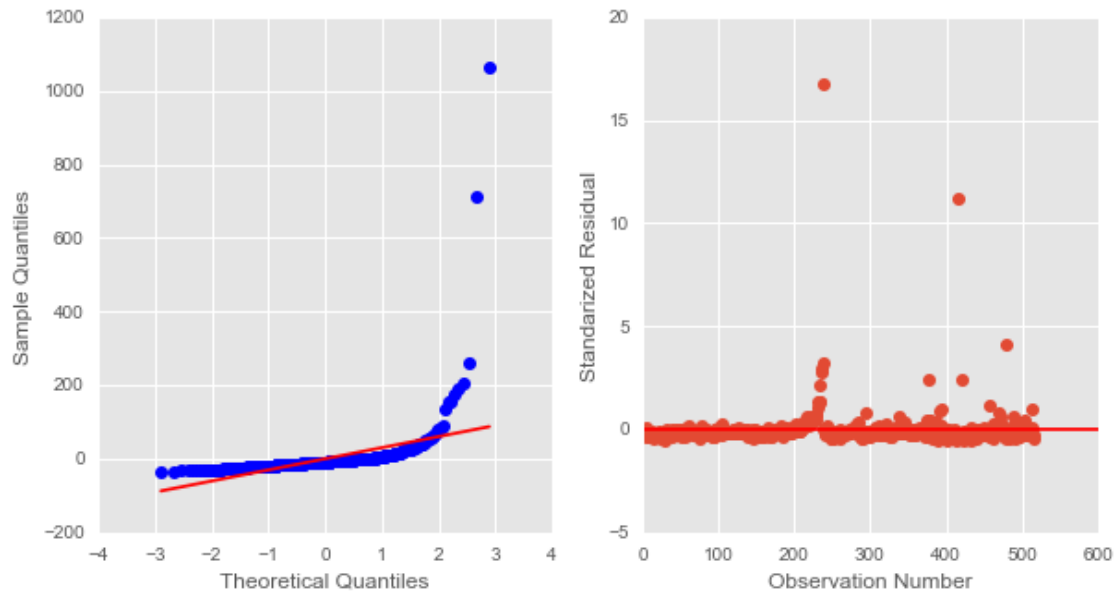
	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	-17.5970	18.784	-0.937	0.349	-54.500 19.306
temp	0.6091	0.744	0.819	0.413	-0.853 2.071

RH	-0.1956	0.229	-0.854	0.393	-0.646	0.254
DMC	0.0907	0.067	1.356	0.176	-0.041	0.222
X	1.9310	1.217	1.586	0.113	-0.461	4.323
month	3.0189	2.700	1.118	0.264	-2.286	8.324
DC	-0.0334	0.031	-1.091	0.276	-0.094	0.027
day	1.2812	1.314	0.975	0.330	-1.300	3.862

```
=====
Omnibus:                973.549    Durbin-Watson:                1.643
Prob(Omnibus):          0.000    Jarque-Bera (JB):          772827.176
Skew:                   12.531    Prob(JB):                  0.00
Kurtosis:               190.744    Cond. No.                  1.71e+03
=====
```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.71e+03. This might indicate that there are strong multicollinearity or other numerical problems.



```
+-----+
| Model: area ~ temp + RH + DMC + X + month + DC + day + Y |
+-----+
```

OLS Regression Results

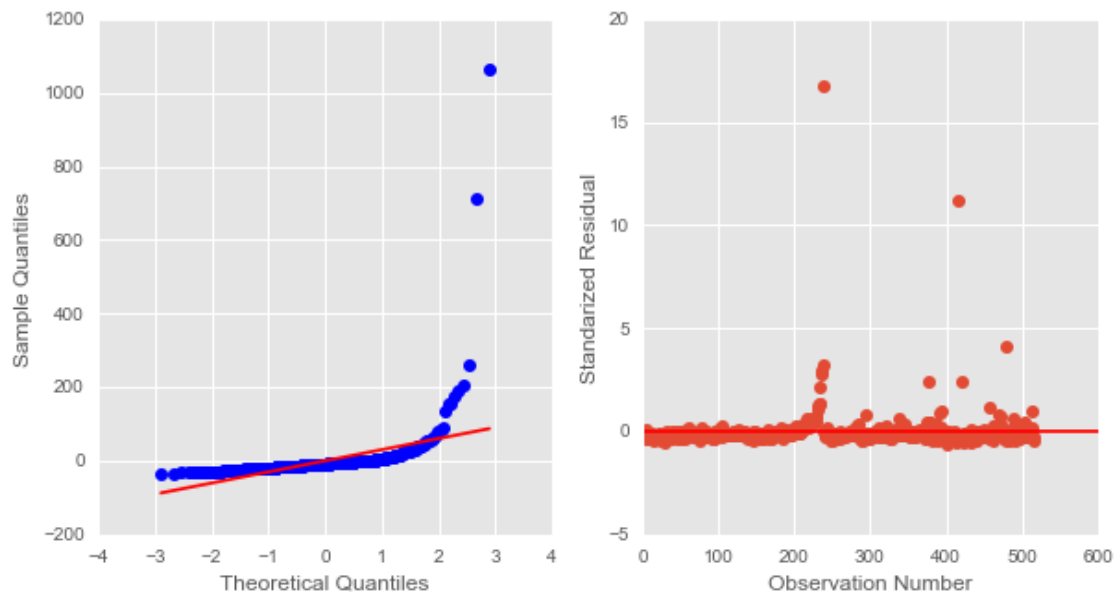
```
=====
Dep. Variable:          area    R-squared:                0.022
Model:                  OLS     Adj. R-squared:           0.007
Method:                 Least Squares    F-statistic:             1.437
Date:                   Tue, 14 Jun 2016    Prob (F-statistic):      0.178
Time:                   22:34:34    Log-Likelihood:          -2874.7
No. Observations:       517    AIC:                     5767.
Df Residuals:           508    BIC:                     5806.
Df Model:                8
```


Covariance Type: nonrobust

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	-17.8818	18.903	-0.946	0.345	-55.020 19.256
temp	0.6048	0.745	0.811	0.417	-0.859 2.069
RH	-0.1967	0.229	-0.857	0.392	-0.647 0.254
DMC	0.0895	0.067	1.327	0.185	-0.043 0.222
X	1.8191	1.441	1.262	0.207	-1.012 4.651
month	2.9834	2.714	1.099	0.272	-2.348 8.315
DC	-0.0328	0.031	-1.058	0.291	-0.094 0.028
day	1.2850	1.315	0.977	0.329	-1.299 3.869
Y	0.3977	2.734	0.145	0.884	-4.974 5.770
Omnibus:	973.369	Durbin-Watson:	1.642		
Prob(Omnibus):	0.000	Jarque-Bera (JB):	771951.439		
Skew:	12.527	Prob(JB):	0.00		
Kurtosis:	190.637	Cond. No.	1.72e+03		

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.72e+03. This might indicate that there are strong multicollinearity or other numerical problems.



```
+-----+
| Model: area ~ temp + RH + DMC + X + month + DC + day + Y + FFMC |
+-----+
```

OLS Regression Results

```
=====
Dep. Variable:          area    R-squared:          0.023
```

```

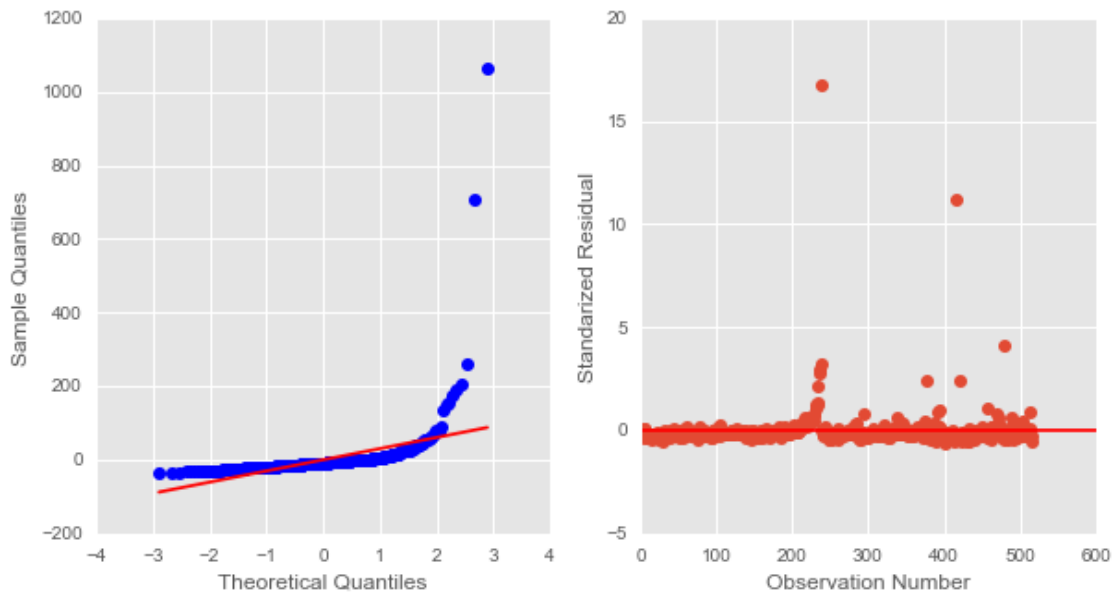
Model:                OLS      Adj. R-squared:      0.006
Method:               Least Squares      F-statistic:      1.320
Date:                 Tue, 14 Jun 2016    Prob (F-statistic):    0.223
Time:                 22:34:34    Log-Likelihood:      -2874.5
No. Observations:     517      AIC:      5769.
Df Residuals:         507      BIC:      5811.
Df Model:              9
Covariance Type:      nonrobust

```

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	-19.0738	19.008	-1.003	0.316	-56.418 18.270
temp	0.6607	0.751	0.880	0.379	-0.815 2.136
RH	-0.2260	0.234	-0.965	0.335	-0.686 0.234
DMC	0.1008	0.070	1.444	0.149	-0.036 0.238
X	1.8695	1.444	1.294	0.196	-0.968 4.707
month	3.1573	2.729	1.157	0.248	-2.205 8.519
DC	-0.0341	0.031	-1.098	0.273	-0.095 0.027
day	1.3289	1.318	1.008	0.314	-1.260 3.918
Y	0.2888	2.741	0.105	0.916	-5.097 5.674
FFMC	-0.3772	0.596	-0.633	0.527	-1.547 0.793
Omnibus:	973.080		Durbin-Watson:	1.643	
Prob(Omnibus):	0.000		Jarque-Bera (JB):	771099.036	
Skew:	12.519		Prob(JB):	0.00	
Kurtosis:	190.534		Cond. No.	1.73e+03	

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.73e+03. This might indicate that there are strong multicollinearity or other numerical problems.



```

+-----+
| Model: area ~ temp + RH + DMC + X + month + DC + day + Y + FFMC + wind |
+-----+

```

OLS Regression Results

```

=====
Dep. Variable:          area    R-squared:                0.023
Model:                  OLS      Adj. R-squared:           0.004
Method:                 Least Squares    F-statistic:         1.214
Date:                   Tue, 14 Jun 2016    Prob (F-statistic):    0.279
Time:                   22:34:35    Log-Likelihood:       -2874.3
No. Observations:       517    AIC:                  5771.
Df Residuals:           506    BIC:                  5817.
Df Model:               10
Covariance Type:        nonrobust
=====

```

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	-17.3156	19.319	-0.896	0.371	-55.271 20.640
temp	0.7126	0.758	0.940	0.348	-0.777 2.202
RH	-0.2247	0.234	-0.959	0.338	-0.685 0.236
DMC	0.0958	0.071	1.359	0.175	-0.043 0.234
X	1.8451	1.446	1.276	0.203	-0.996 4.686
month	2.8663	2.788	1.028	0.304	-2.611 8.344
DC	-0.0301	0.032	-0.939	0.348	-0.093 0.033
day	1.3241	1.319	1.004	0.316	-1.267 3.915
Y	0.3890	2.750	0.141	0.888	-5.014 5.792
FFMC	-0.3932	0.597	-0.659	0.510	-1.566 0.779
wind	0.8642	1.660	0.521	0.603	-2.398 4.126

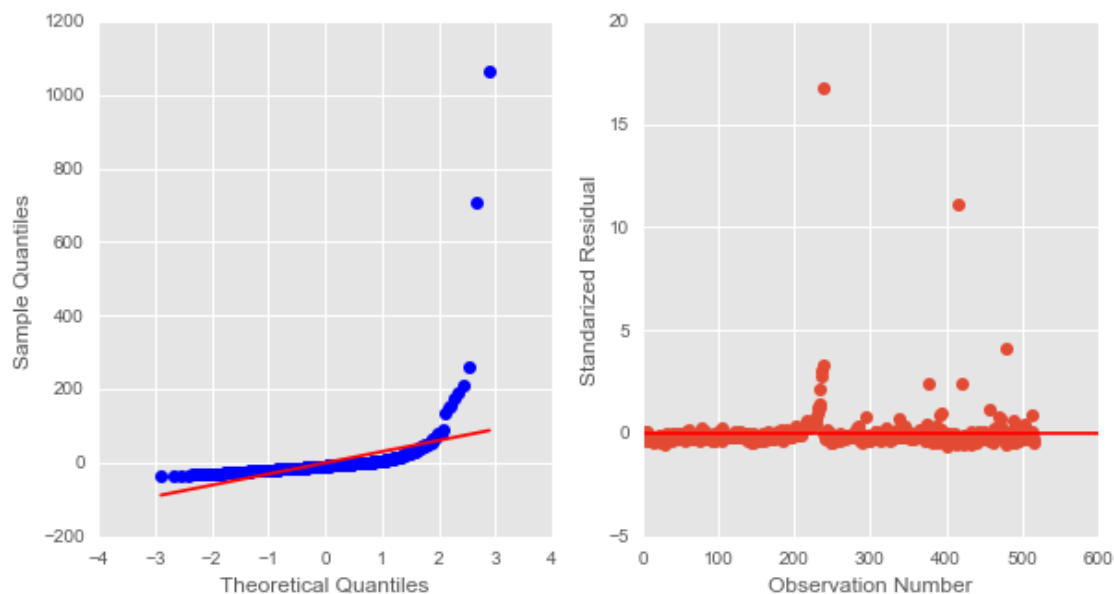
```

=====
Omnibus:                 972.761    Durbin-Watson:           1.637
Prob(Omnibus):           0.000    Jarque-Bera (JB):       769820.664
Skew:                    12.510    Prob(JB):                0.00
Kurtosis:                190.377    Cond. No.                1.76e+03
=====

```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.76e+03. This might indicate that there are strong multicollinearity or other numerical problems.



```
+-----+
| Model: area ~ temp + RH + DMC + X + month + DC + day + Y + FFMC + wind + ISI |
+-----+
```

OLS Regression Results

```
=====
Dep. Variable:          area    R-squared:                0.025
Model:                  OLS     Adj. R-squared:           0.004
Method:                 Least Squares   F-statistic:         1.184
Date:                  Tue, 14 Jun 2016   Prob (F-statistic):   0.295
Time:                  22:34:36   Log-Likelihood:      -2873.9
No. Observations:      517         AIC:                 5772.
Df Residuals:          505         BIC:                 5823.
Df Model:              11
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	-17.5792	19.323	-0.910	0.363	-55.543 20.385
temp	0.9189	0.789	1.165	0.245	-0.631 2.469
RH	-0.1885	0.237	-0.794	0.428	-0.655 0.278
DMC	0.0970	0.071	1.375	0.170	-0.042 0.236
X	1.8773	1.447	1.298	0.195	-0.965 4.720
month	2.9056	2.789	1.042	0.298	-2.573 8.384
DC	-0.0314	0.032	-0.978	0.329	-0.094 0.032
day	1.3346	1.319	1.012	0.312	-1.257 3.926
Y	0.3280	2.751	0.119	0.905	-5.077 5.733
FFMC	-0.1239	0.662	-0.187	0.851	-1.424 1.176
wind	1.1858	1.695	0.700	0.485	-2.145 4.516
ISI	-0.7271	0.771	-0.943	0.346	-2.242 0.787

```
=====
Omnibus:                972.712   Durbin-Watson:          1.642
```

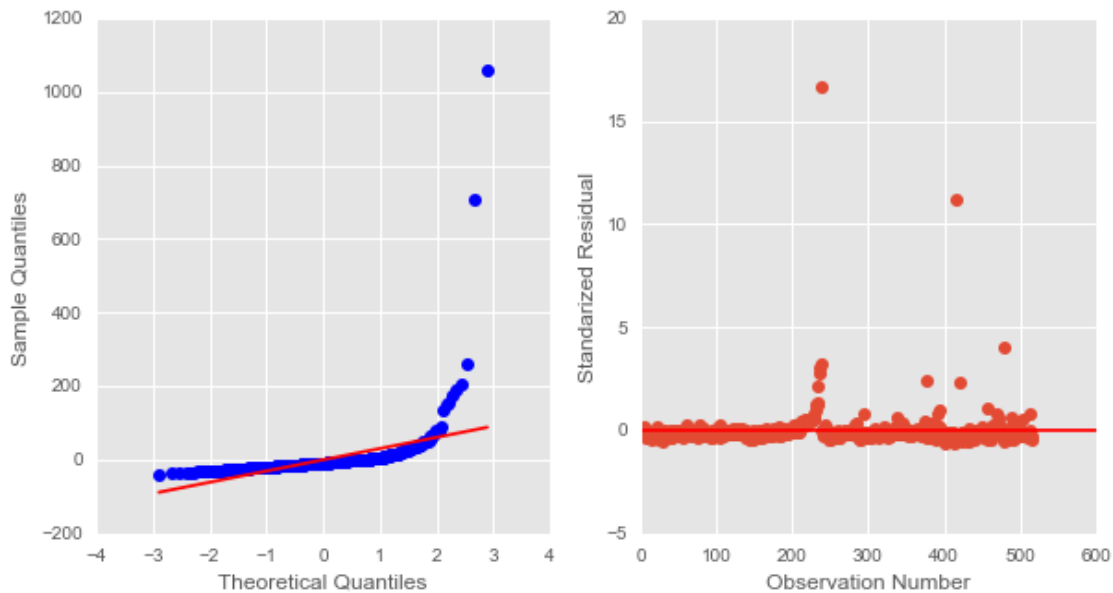
```

Prob(Omnibus):          0.000   Jarque-Bera (JB):          769667.142
Skew:                  12.509   Prob(JB):              0.00
Kurtosis:              190.359   Cond. No.              1.76e+03
=====

```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.76e+03. This might indicate that there are strong multicollinearity or other numerical problems.



```

+-----+
| Model: area ~ temp + RH + DMC + X + month + DC + day + Y + FFMC + wind + ISI + rain |
+-----+

```

OLS Regression Results

```

=====
Dep. Variable:          area   R-squared:              0.025
Model:                  OLS   Adj. R-squared:         0.002
Method:                 Least Squares   F-statistic:           1.092
Date:                   Tue, 14 Jun 2016   Prob (F-statistic):    0.364
Time:                   22:34:37   Log-Likelihood:        -2873.8
No. Observations:       517   AIC:                   5774.
Df Residuals:           504   BIC:                   5829.
Df Model:                12
Covariance Type:        nonrobust
=====

```

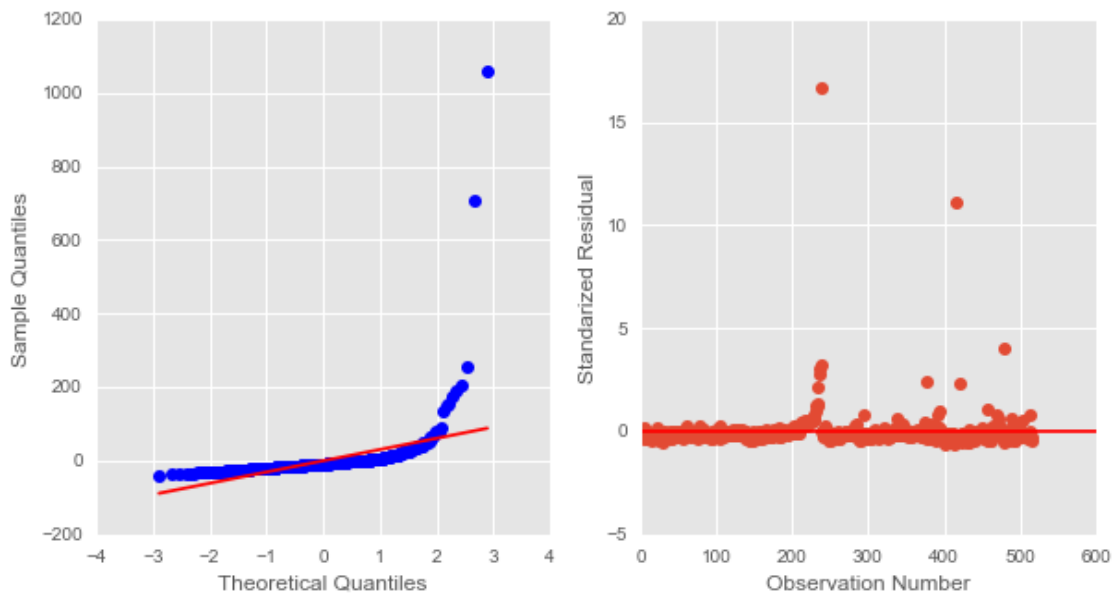
	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	-17.5974	19.340	-0.910	0.363	-55.595 20.400
temp	0.9546	0.797	1.198	0.232	-0.612 2.521
RH	-0.1758	0.241	-0.730	0.466	-0.649 0.297
DMC	0.0966	0.071	1.369	0.172	-0.042 0.235

X	1.9002	1.450	1.311	0.191	-0.948	4.748
month	2.9004	2.791	1.039	0.299	-2.583	8.384
DC	-0.0315	0.032	-0.981	0.327	-0.095	0.032
day	1.3269	1.320	1.005	0.315	-1.267	3.921
Y	0.3241	2.754	0.118	0.906	-5.086	5.734
FFMC	-0.1127	0.663	-0.170	0.865	-1.415	1.190
wind	1.2321	1.702	0.724	0.470	-2.113	4.577
ISI	-0.7305	0.772	-0.947	0.344	-2.247	0.786
rain	-3.1958	9.683	-0.330	0.742	-22.220	15.829

```
=====
Omnibus:                972.663    Durbin-Watson:                1.643
Prob(Omnibus):          0.000    Jarque-Bera (JB):          769640.593
Skew:                   12.508    Prob(JB):                  0.00
Kurtosis:               190.356    Cond. No.                  1.76e+03
=====
```

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.76e+03. This might indicate that there are strong multicollinearity or other numerical problems.



From above results can be seen that after adding a second variable to the model (relative humidity –RH–), temperature –temp– is not longer statistically significant, so its a confounder variable: “temperature increases due to forest fire, or forest fire is helped by high temperatures?”

In all generated models we can see the summary of the multiple regression model, the quantile-quantile plot (qq-plot), left side, which “plots the quantiles of the residuals that we would theoretically see if the residuals followed a normal distribution, against the quantiles for the residuals estimated from our regression model”, and, right side, the Standarized Residuals

plot, which are, “simply, the residual values transformed to have a mean of zero and a standard deviation of one”

The qq-plots of our regression models shows a straight line with high deviations at the lower and higher quantiles, indicating the residuals do not follow a normal distribution.

The Standardized Residuals plot shows there are a group of observations which standardized residuals are greater than 2 standard deviations, and more than 3 standard deviations implying the presence of extreme outliers

```
In [26]: stdres = pandas.DataFrame(saved_models[11].resid_pearson)
        stdres[(stdres[0] < -2.5) | (stdres[0] > 2.5)].T
```

```
Out[26]:      235   236   237   238   415   479
         0  2.760  3.019  3.262 16.681 11.154  4.055
```

In all generated models, more than 1% of our observations has standardized residuals with an absolute value greater than 2.5, then there is evidence that the level of error within our model is unacceptable. That is the model is a fairly poor fit to the observed data.

```
In [27]: pct_outliers = list()
        for idx in range(len(saved_models)):
            stdres = pandas.DataFrame(saved_models[idx].resid_pearson)
            pct_outliers.append(round(len(stdres[(stdres[0] < -2.5) | (stdres[0] > 2.5)]) / len(fires)

        print(pct_outliers)
```

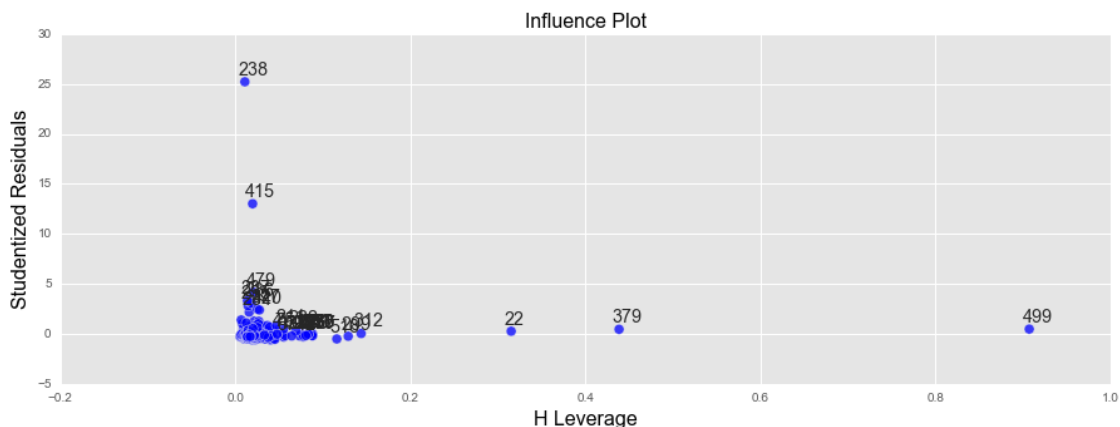
```
[0.015, 0.015, 0.014, 0.012, 0.012, 0.012, 0.012, 0.012, 0.012, 0.012, 0.012, 0.012]
```

The second warning of full model indicates “The condition number is large, 1.76e+03. This might indicate that there are strong multicollinearity or other numerical problems”, as it could be expected.

During data exploration we found that there is a medium to medium-high correlation between the average moisture content of deep, compact organic layers and the average moisture content of loosely compacted organic layers of moderate depth (DC-DMC: 0.682) and between the expected rate of fire spread and the moisture content of litter and other cured fine fuels (ISI- FFMC: 0.532). Also, there is a inverse medium correlation (-0.527) between temperature (temp) and relative humidity (RH). Other relationships are noted between temperature (temp) and FWI system components (FFMC, DCM, DC and ISI)

4.4 Leverage plot

```
In [29]: sm.graphics.influence_plot(model, size=8) #Leverage plot for full model
        plt.show()
```



Leverage plot permits identify observations that have an unusually large influence on the estimation of the predicted value of the response variable, burned area, or that are outliers, or both. The graph of full model shows one observation with a very high influence (observation 499, with near 90%), one with medium influence (observation 379, with near 45%) and one with medium-low influence (observation 22, with near 32%). The rest of the observations have influence under 20%

The graph of full model also show us a group of outliers. Note this extreme outliers are the same observations we found during data exploration: 238, 415, 479, plus a cloud of minor outliers (residuals outside range -2 to 2 standard deviations), but with low influence ($< 5\%$) on the estimation of the regression model

No observations in this data are both high leverage and outliers.

```
In [30]: fires.iloc[[22, 379, 499]]
```

```
Out[30]:
```

	X	Y	month	day	FFMC	DMC	DC	ISI	temp	RH	wind	\
22	6	2	5	0	3.655	-14.572	-347.940	47.078	2.111	-0.288	0.482	
379	3	3	0	0	-71.945	-109.772	-376.540	-9.022	-13.689	55.712	-3.118	
499	6	3	7	2	5.455	70.228	123.260	5.278	8.411	18.712	0.882	

	rain	area
22	-0.022	0.000
379	-0.022	0.000
499	6.378	10.820

```
In [ ]:
```