Workshop 9

Goal

Use python to perform logistic regression.

成功：1，失败：0

A trade is considered as success when it makes money. Success will be represented by 1 and failure by 0.

This strategy works with News and Sentiment. The column news intensity represents the number of news / minute and the column sentiment represents the news sentiment (1: good news…4: bad news).

Create a dataframe df by loading the data using read_csv

```
ts=
```

```
print (df.tail())
```

	success	news intensity	price	sentiment
395	0	620	60.00	2
396	0	560	45.60	3
397	0	460	39.45	2
398	0	700	54.75	2
399	0	600	58.35	3

Analyze the statistics of df

```
print df.XXX()
```

	success	news intensity	price	sentiment
count	400.000000	400.000000	400.000000	400.00000
mean	0.317500	587.700000	50.848500	2.48500
std	0.466087	115.516536	5.708502	0.94446
min	0.000000	220.000000	33.900000	1.00000
25%	0.000000	520.000000	46.950000	2.00000
50%	0.000000	580.000000	50.925000	2.00000
75%	1.000000	660.000000	55.050000	3.00000
max	1.000000	800.000000	60.000000	4.00000

Display the mean of each column separately

```
print df.XXX()
```

```
success          0.3175
news intensity   587.7000
price            50.8485
sentiment        2.4850
```

Since the news sentiment has only 4 levels, draw the following table using crosstab

```
print(df.crosstab(….))
```

```
          sentiment  1   2   3   4
success
0                    28  97  93  55
1                    33  54  28  12
```

Draw the histogram for each column

```
df.XXX()
```

Sentiment is a categorical variable. We are going to transform this variable into 4 dummy variables using the command get_dummies from pandas.

Example:

```
b=pd.DataFrame({'test' : pd.Series([1,2,3,1,2,3,1,1,1])})
 print(pd.get_dummies(b['test'],prefix='test'))
test_1  test_2  test_3
0    1    0    0
1    0    1    0
2    0    0    1
3    0    1    0
4    1    0    0
5    0    1    0
6    0    0    1
```

```
7   0   0   1
8   0   0   1
```

Following the previous example create dummy variables for Sentiment using the function get_dummies. Store the result into *data_dummy*

```

```

Create a joint to keep success, news intensity, price, sentiment_2, sentiment_3 and sentiment_4

```
data = df['XXXXXXXX'].join(.....)
```

	success	news intensity	price	sentiment_2	sentiment_3	sentiment_4
0	0	380	54.15	0	1	0
1	1	660	55.05	0	1	0
2	1	800	60.00	0	0	0
3	1	640	47.85	0	0	1
4	0	520	43.95	0	0	1
5	1	760	45.00	1	0	0
6	1	560	44.70	0	0	0
7	0	400	46.20	1	0	0
8	1	540	50.85	0	1	0
9	0	700	58.80	1	0	0
10	0	800	60.00	0	0	1

Add the intercept manually (a column named intersect will only 1)

```

```

Perform logistic regression.
Step 1: Remove the column name 'success'

```
colnames=
```

Step 2: Create the logistic model

```
Logit_model= sm.Logit(data['success'], data[colnames])
```

Step 3: Fi the model

```
result = logit.fit()
```

Interpret the following result:

```
print (result.summary())
```

```
                           Logit Regression Results
==============================================================================
Dep. Variable:                success   No. Observations:                  400
Model:                          Logit   Df Residuals:                      394
Method:                           MLE   Df Model:                            5
Date:                Sun, 20 Nov 2016   Pseudo R-squ.:                0.08292
Time:                        02:54:02   Log-Likelihood:                -229.26
converged:                       True   LL-Null:                       -249.99
                                        LLR p-value:                 7.578e-08
================================================================================
                   coef    std err          z      P>|z|      [95.0% Conf. Int.]
--------------------------------------------------------------------------------
news intensity     0.0023      0.001      2.070      0.038         0.000     0.004
price              0.0536      0.022      2.423      0.015         0.010     0.097
sentiment_2       -0.6754      0.316     -2.134      0.033        -1.296    -0.055
sentiment_3       -1.3402      0.345     -3.881      0.000        -2.017    -0.663
sentiment_4       -1.5515      0.418     -3.713      0.000        -2.370    -0.733
intersect         -3.9900      1.140     -3.500      0.000        -6.224    -1.756
================================================================================
```

Calculate confidence interval with the function conf_int() associated to result

```

```

Display odds ration (just use np.exp in the params of result)

```

```