

**RE: 07/18 DM課上午場**

洪子軒

**Sent:** Monday, July 18, 2016 4:09 PM**To:** 洪子軒[http://bse.nchu.edu.tw/new\\_page\\_535.htm](http://bse.nchu.edu.tw/new_page_535.htm)[http://statsmodels.sourceforge.net/devel/example\\_formulas.html](http://statsmodels.sourceforge.net/devel/example_formulas.html)

X變數用愈多，R square 會越大，造成解釋能力虛化

所以要改用 → Adjusted R Square

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降低維度（避免模型失效、視覺化、屬性不獨立造成重複解釋）

屬性重複

屬性不相關

方法一、Feature Selection

方法二、Feature Extraction

&gt; pca 主成份分析 Principle Component Analysis

&gt; svd Singular Value Decomposition

&gt; mds Multi-dimensional scaling

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機率密度 cdf pdf

empirical normal probability 常態分配

<http://docs.scipy.org/doc/scipy-0.16.0/reference/generated/scipy.stats.norm.html>

poisson distribution 離散分配

二項分配

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對不同迴歸模型作 anova 測試比較

[http://statsmodels.sourceforge.net/devel/examples/generated/example\\_interactions.html](http://statsmodels.sourceforge.net/devel/examples/generated/example_interactions.html)**from statsmodels.stats.anova import** anova\_lm

table1 = anova\_lm(lm, interX\_lm)

**print** table1p 值  $\geq 0.05$  代表檢定的對象彼此相同（不拒絕虛無假設）p 值  $< 0.05$  有顯著差異（拒絕虛無假設—兩者相同）

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用來衡量迴歸模型複雜度的方式

參數用得越多，模型越複雜，成本就越高

aic [https://en.wikipedia.org/wiki/Akaike\\_information\\_criterion](https://en.wikipedia.org/wiki/Akaike_information_criterion)bic [https://en.wikipedia.org/wiki/Bayesian\\_information\\_criterion](https://en.wikipedia.org/wiki/Bayesian_information_criterion)說明 <https://read01.com/zPjKQK.html>

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統計共線性檢定（X變數夠不夠獨立）

變異數影響因子（Variance Inflation Factor，VIF）

[http://statsmodels.sourceforge.net/devel/modules/statsmodels/stats/outliers\\_influence.html#variance\\_inflation\\_factor](http://statsmodels.sourceforge.net/devel/modules/statsmodels/stats/outliers_influence.html#variance_inflation_factor)

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X vs. Y

- 類別/類別：卡方檢定
- 類別/數值：ANOVA

例：abc 教學法影響成績

- 數值/類別：均值T檢定（先做變異數F檢定）

例：平均身高對於性別有無顯著；好公司和壞公司的平均固定資產周轉率有無不同

- 數值/數值：相關係數

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[http://scikit-learn.org/stable/auto\\_examples/classification/plot\\_classifier\\_comparison.html#example-classification-plot-classifier-comparison-py](http://scikit-learn.org/stable/auto_examples/classification/plot_classifier_comparison.html#example-classification-plot-classifier-comparison-py)

```
from sklearn.preprocessing import StandardScaler
```

```
X = StandardScaler().fit_transform(X)
```

<http://docs.scipy.org/doc/scipy/reference/stats.html>

雙尾、單尾

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**From:** 洪子軒

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**Subject:** RE: 07/18 DM課上午場

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**From:** 洪子軒

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```
>>> x = pandas.Series(np.random.randn(10))
>>> stats.skew(x)
-0.17644348972413657
>>> x.skew()
-0.20923623968879457
>>> stats.skew(x, bias=False)
-0.2092362396887948
>>> stats.kurtosis(x)
0.6362620964462327
>>> x.kurtosis()
2.0891062062174464
>>> stats.kurtosis(x, bias=False)
2.089106206217446
```

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常態分配

<http://stackoverflow.com/questions/13865596/quantile-quantile-plot-using-scipy>

```
import numpy as np
import pylab
import scipy.stats as stats

measurements = np.random.normal(loc = 20, scale = 5, size=100)
stats.probplot(measurements, dist="norm", plot=pylab)
pylab.show()
```

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shapiro-wilk normality test 常態分配檢定測試

<http://docs.scipy.org/doc/scipy-0.14.0/reference/generated/scipy.stats.shapiro.html>

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雙群樣本 Avg 平均數檢定 = T test (相減) : var 變異數檢定 F test (相除)

```
>>> import pandas as pd
>>> import scipy.stats
>>> import numpy as np
>>> df_a = pd.read_clipboard()
>>> df_b = df_a + np.random.randn(5, 7)
>>> df_c = df_a + np.random.randn(5, 7)
>>> t_b, p_b = scipy.stats.ttest_ind(df_a.dropna(axis=0), df_b.dropna(axis=0))
>>> t_c, p_c = scipy.stats.ttest_ind(df_a.dropna(axis=0), df_c.dropna(axis=0))
>>> pd.DataFrame([p_b, p_c], columns = df_a.columns, index = ['df_b', 'df_c'])
      VSPD1_perc  VSPD2_perc  VSPD3_perc  VSPD4_perc  VSPD5_perc  VSPD6_perc \
df_b      0.425286      0.987956      0.644236      0.552244      0.432640      0.624528
df_c      0.947182      0.911384      0.189283      0.828780      0.697709      0.166956

      VSPD7_perc
df_b      0.546648
df_c      0.206950
```

p 值<0.05 對立 (不相等) 假設成立

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ANOVA 檢定：多群平均值是否彼此相等

<http://www.marsja.se/four-ways-to-conduct-one-way-anovas-using-python/>

```
# compute one-way ANOVA P value
from scipy import stats

f_val, p_val = stats.f_oneway(treatment1, treatment2, treatment3)

print "One-way ANOVA P =", p_val

One-way ANOVA P = 0.381509481874
```

If  $P > 0.05$ , we can claim with high confidence that the means of the results of all three experiments are not significantly different.

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卡方檢定/比例檢定：離散、數目count 檢定

<http://codereview.stackexchange.com/questions/96761/chi-square-independence-test-for-two-pandas-df-columns>

```
def chi_square_of_df_cols(df,col1,col2):
    return scs.chi2_contingency([
        [
            len(df[(df[col1] == cat) & (df[col2] == cat2)])
            for cat2 in range(int(df[col1].min()), int(df[col1].max()) + 1)
        ]
        for cat in range(int(df[col2].min()), int(df[col2].max()) + 1)
    ])
```

<http://stackoverflow.com/questions/11725115/p-value-from-chi-sq-test-statistic-in-python>

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Outlier 離群值 mahalanobis distance (欄位屬性彼此不獨立)

<http://stackoverflow.com/questions/29817090/is-there-a-python-equivalent-to-the-mahalanobis-function-in-r-if-not-how-can>

```
from scipy.spatial.distance import mahalanobis
import scipy as sp
import pandas as pd

x = pd.read_csv('IrisData.csv')
x = x.ix[:,1:]

Sx = x.cov().values
Sx = sp.linalg.inv(Sx)

mean = x.mean().values

def mahalanobisR(X,meanCol,IC):
    m = []
    for i in range(X.shape[0]):
        m.append(mahalanobis(X.ix[i,:],meanCol,IC) ** 2)
    return(m)

mR = mahalanobisR(x,mean,Sx)

stats.chi2.cdf()
```

To calculate probability of null hypothesis given chisquared sum, and degrees of freedom you can also call `chisqprob` :

```
>>> from scipy.stats import chisqprob
>>> chisqprob(3.84, 1)
0.050043521248705189
```

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<http://stackoverflow.com/questions/19991445/run-an-ols-regression-with-pandas-data-frame>

R-squared / adjusted R-squared

```

>>> import pandas as pd
>>> import statsmodels.formula.api as sm
>>> df = pd.DataFrame({"A": [10,20,30,40,50], "B": [20, 30, 10, 40, 50], "C": [32, 234, 23, 23,
>>> result = sm.ols(formula="A ~ B + C", data=df).fit()
>>> print result.params
Intercept    14.952480
B             0.401182
C             0.000352
dtype: float64
>>> print result.summary()

```

OLS Regression Results

```

=====
Dep. Variable:          A      R-squared:            0.579
Model:                  OLS      Adj. R-squared:       0.158
Method:                 Least Squares      F-statistic:       1.375
Date:                  Thu, 14 Nov 2013      Prob (F-statistic):    0.421
Time:                  20:04:30      Log-Likelihood:      -18.178
No. Observations:      5      AIC:                42.36
Df Residuals:          2      BIC:                41.19
Df Model:              2
=====

```

	coef	std err	t	P> t	[95.0% Conf. Int.]
Intercept	14.9525	17.764	0.842	0.489	-61.481 91.386
B	0.4012	0.650	0.617	0.600	-2.394 3.197
C	0.0004	0.001	0.650	0.583	-0.002 0.003

```

=====
Omnibus:                nan      Durbin-Watson:       1.061
Prob(Omnibus):          nan      Jarque-Bera (JB):     0.498
Skew:                  -0.123     Prob(JB):             0.780
Kurtosis:               1.474     Cond. No.             5.21e+04
=====

```

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