

# 机器学习第二次作业

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## 1. 下列数据时水泥释放的热量与其成分的关系：求其线性依赖关系

y	x1	x2	x3	x4
78.5	7	26	6	60
74.3	1	29	15	52
104.3	11	56	8	20
87.6	11	31	8	47
95.9	7	52	6	33
109.2	11	55	9	22
102.7	3	71	17	6
72.5	1	31	22	44
93.1	2	54	18	22
115.9	21	47	4	26
83.8	1	40	23	34
113.3	11	66	9	12
109.4	10	68	8	12

将上述数据用excel保存，命名为data.xlsx。

编写Python代码：

```
import pandas as pd
import statsmodels.api as sm

data = pd.read_excel('data.xlsx')
data.columns = ['y', 'x1', 'x2', 'x3', 'x4']
# 生成自变量
x = sm.add_constant(data.iloc[:, 1:])
# 生成因变量
y = data['y']
# 生成模型
model = sm.OLS(y, x)
# 模型拟合
result = model.fit()
```

```
# 模型描述
print(result.summary())
```

运行结果:

OLS Regression Results

Dep. Variable:	y	R-squared:	0.982
Model:	OLS	Adj. R-squared:	0.974
Method:	Least Squares	F-statistic:	111.5
Date:	Tue, 11 Oct 2022	Prob (F-statistic):	4.76e-07
Time:	21:24:03	Log-Likelihood:	-26.918
No. Observations:	13	AIC:	63.84
Df Residuals:	8	BIC:	66.66
Df Model:	4		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	62.4054	70.071	0.891	0.399	-99.179	223.989
x1	1.5511	0.745	2.083	0.071	-0.166	3.269
x2	0.5102	0.724	0.705	0.501	-1.159	2.179
x3	0.1019	0.755	0.135	0.896	-1.638	1.842
x4	-0.1441	0.709	-0.203	0.844	-1.779	1.491

Omnibus:	0.165	Durbin-Watson:	2.053
Prob(Omnibus):	0.921	Jarque-Bera (JB):	0.320
Skew:	0.201	Prob(JB):	0.852
Kurtosis:	2.345	Cond. No.	6.06e+03

Notes:

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

[2] The condition number is large, 6.06e+03. This might indicate that there are strong multicollinearity or other numerical problems.

从上述结果描述，我们得到回归模型：

$$y = 62.4054 + 1.5511x_1 + 0.5102x_2 + 0.1019x_3 - 0.1441x_4$$

从结果中还可以看出，Prob (F-statistic)为4.76e-07，其接近于零，说明我们的多元线性方程是显著的，也就是y与x<sub>1</sub>、x<sub>2</sub>、x<sub>3</sub>、x<sub>4</sub>有着显著的线性关系，而R-squared是0.982，也说明这个线性关系比较显著。

2. 经研究发现，学生用于购买书籍及课外读物的支出与本人受教育年限和其家庭收入水平有关，对18名学生进行调查的统计资料如下表所示，求其回归模型

y	x1	x2
450.5	4	171.2

y	x1	x2
507.7	4	174.2
613.9	5	204.3
563.4	4	218.7
501.5	4	219.4
781.5	7	240.4
541.8	4	273.5
611.1	5	294.8
1222.1	10	330.2
793.2	7	333.1
660.8	5	366
792.7	6	350.9
580.8	4	357.9
612.7	5	359
890.8	7	371.9
1121	9	435.3
1094.2	8	523.9
1253	10	604.1

将上述数据用excel保存，命名为data.xlsx。

编写Python代码：

```
import pandas as pd
import statsmodels.api as sm

data = pd.read_excel('data.xlsx')
data.columns = ['y', 'x1', 'x2']
# 生成自变量
x = sm.add_constant(data.iloc[:, 1:])
# 生成因变量
y = data['y']
# 生成模型
model = sm.OLS(y, x)
# 模型拟合
result = model.fit()
# 模型描述
print(result.summary())
```

运行结果：

OLS Regression Results

Dep. Variable:	y	R-squared:	0.980
Model:	OLS	Adj. R-squared:	0.977
Method:	Least Squares	F-statistic:	362.4
Date:	Tue, 11 Oct 2022	Prob (F-statistic):	2.00e-13
Time:	22:22:18	Log-Likelihood:	-89.942
No. Observations:	18	AIC:	185.9
Df Residuals:	15	BIC:	188.6
Df Model:	2		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
const	-0.9756	30.322	-0.032	0.975	-65.606	63.655
x1	104.3146	6.409	16.276	0.000	90.654	117.975
x2	0.4022	0.116	3.457	0.004	0.154	0.650

Omnibus:	0.776	Durbin-Watson:	2.561
Prob(Omnibus):	0.678	Jarque-Bera (JB):	0.728
Skew:	-0.230	Prob(JB):	0.695
Kurtosis:	2.128	Cond. No.	1.13e+03

Notes:

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

[2] The condition number is large, 1.13e+03. This might indicate that there are strong multicollinearity or other numerical problems.

从上述结果描述，我们得到回归模型： $y = -0.9756 + 104.3146x_1 + 0.4022x_2$

从结果中还可以看出，Prob (F-statistic)为2.00e-13，其接近于零，说明我们的多元线性方程是显著的，也就是y与x<sub>1</sub>、x<sub>2</sub>有着显著的线性关系，而R-squared是0.980，也说明这个线性关系比较显著。