

**ACST 890- TAKE HOME QUIZ 2 SOLUTIONS: 45535531-ROHAN MARCUS DEANS**

**GITHUB Username:** 45535531

**Repository Name:** Quiz1 (<https://github.com/45535531/Quiz1.git>)

**File Name:** 45535531DeansRohanMarcusTHQ2.pdf

**Other Attached files on GITHUB:** codings.txt

**1.a.**

```
In [25]: %%bash

echo "using bash from inside Jupyter!" > test-file.txt
ls
echo ""
cat test-file.txt
rm test-file.txt
```

Couldn't find program: 'bash'

```
In [28]: # Source: https://github.com/catherinedevlin/ipython-sql
# do pip install ipython-sql in the terminal
%load_ext sql
```

```
-----
ModuleNotFoundError                                Traceback (most recent call last)
<ipython-input-28-e943513a1a1d> in <module>
      1 # Source: https://github.com/catherinedevlin/ipython-sql
      2 # do pip install ipython-sql in the terminal
----> 3 get_ipython().run_line_magic('load_ext', 'sql')
```

**ModuleNotFoundError:** No module named 'sql'

```
In [30]: # pip install version_information
# https://github.com/jrjohansson/version_information
#
# alternate option: https://github.com/rasbt/watermark

%load_ext version_information
%version_information requests, numpy, pandas, matplotlib, seaborn, sklearn

-----
ModuleNotFoundError                                Traceback (most recent call last)

ModuleNotFoundError: No module named 'version_information'
```

```
In [5]: # Setting plot appearance
# See here for more options: https://matplotlib.org/users/customizing.html

%config InlineBackend.figure_format='retina'
sns.set() # Revert to matplotlib defaults
plt.rcParams['figure.figsize'] = (9, 6)
plt.rcParams['axes.labelpad'] = 10
sns.set_style("darkgrid")
# sns.set_context("poster", font_scale=1.0)

-----
NameError                                Traceback (most recent call last)
<ipython-input-5-b9c4f8741526> in <module>
      3
      4 get_ipython().run_line_magic('config', "InlineBackend.figure_format='retina'")
----> 5 sns.set() # Revert to matplotlib defaults
      6 plt.rcParams['figure.figsize'] = (9, 6)
      7 plt.rcParams['axes.labelpad'] = 10

NameError: name 'sns' is not defined
```

1.b.

```
In [11]: # Source: https://github.com/catherinedevlin/ipython-sql
# do pip install ipython-sql in the terminal
%load_ext sql

The sql extension is already loaded. To reload it, use:
%reload_ext sql
```

```
In [15]: # pip install version_information
# https://github.com/jrjohansson/version_information
#
# alternate option: https://github.com/rasbt/watermark

%load_ext version_information
%version_information requests, numpy, pandas, matplotlib, seaborn, sklearn
```

```
Out[15]:
```

Software	Version
Python	3.7.3 64bit [MSC v.1915 64 bit (AMD64)]
IPython	7.4.0
OS	Windows 10 10.0.17134 SP0
requests	2.21.0
numpy	1.16.2
pandas	0.24.2
matplotlib	3.0.3
seaborn	0.9.0
sklearn	0.20.3

2.

```
In [1]: import pandas as pd
```

```
In [2]: from pandas import DataFrame
```

```
In [3]: Data = {'Country': ['Output, Q', 'Labour, L', 'Capital, K'],
               'B': [80, 60, 50],
               'C': [150, 100, 100],
               'D': [135, 100, 80],
               'E': [165, 120, 100],
               'F': [95, 70, 60],
               'G': [130, 90, 80],
               'H': [110, 80, 70],
               }
```

```
In [4]: Table = DataFrame(Data, columns= ['Country', 'B', 'C', 'D', 'E', 'F', 'G', 'H'])
```

```
In [5]: print(Table)
```

	Country	B	C	D	E	F	G	H
0	Output, Q	80	150	135	165	95	130	110
1	Labour, L	60	100	100	120	70	90	80
2	Capital, K	50	100	80	100	60	80	70

a.

```
In [6]: Table.mean(axis=1)
```

```
Out[6]: 0    123.571429  
        1     88.571429  
        2     77.142857  
        dtype: float64
```

```
In [7]: Table.var(axis=1)
```

```
Out[7]: 0    914.285714  
        1    414.285714  
        2    357.142857  
        dtype: float64
```

```
In [8]: Table.std(axis=1)
```

```
Out[8]: 0    30.237158  
        1    20.354010  
        2    18.898224  
        dtype: float64
```

```
In [9]: Table.median(axis=1)
```

```
Out[9]: 0    130.0  
        1     90.0  
        2     80.0  
        dtype: float64
```

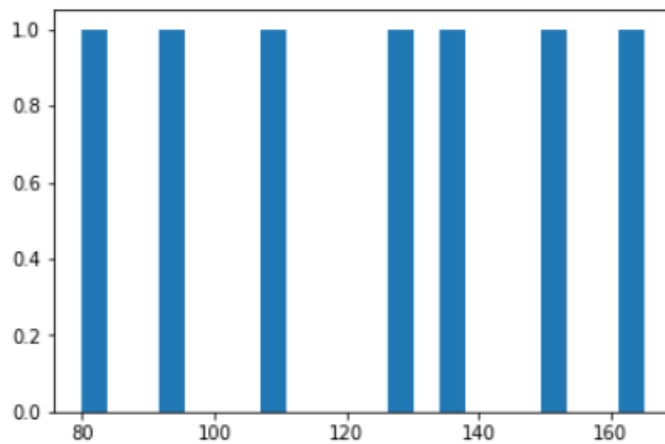
b.

```
In [10]: import matplotlib.pyplot as plt
```

```
In [11]: Q=[80,150,135,165,95,130,110]
```

```
In [12]: plt.hist(Q,bins=22)
```

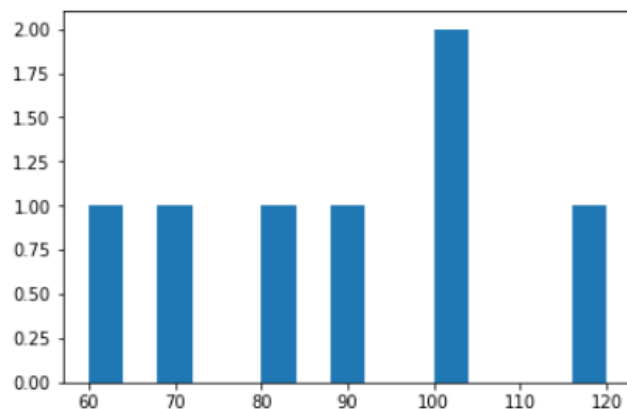
```
Out[12]: (array([1., 0., 0., 1., 0., 0., 0., 1., 0., 0., 0., 0., 1., 0., 1., 0., 0.,  
                0., 1., 0., 0., 1.]),  
          array([ 80.,      83.86363636,  87.72727273,  91.59090909,  
                95.45454545,  99.31818182, 103.18181818, 107.04545455,  
                110.90909091, 114.77272727, 118.63636364, 122.5      ,  
                126.36363636, 130.22727273, 134.09090909, 137.95454545,  
                141.81818182, 145.68181818, 149.54545455, 153.40909091,  
                157.27272727, 161.13636364, 165.      ]),  
          <a list of 22 Patch objects>)
```



```
In [13]: L=[60,100,100,120,70,90,80]
```

```
In [14]: plt.hist(L,bins=15)
```

```
Out[14]: (array([1., 0., 1., 0., 0., 1., 0., 1., 0., 0., 2., 0., 0., 0., 1.]),  
          array([ 60.,  64.,  68.,  72.,  76.,  80.,  84.,  88.,  92.,  96., 100.,  
                104., 108., 112., 116., 120.]),  
          <a list of 15 Patch objects>)
```



C.

```
In [17]: import numpy as np

In [18]: K=[50,100,80,100,60,80,70]

In [19]: logQ=np.log(Q)

In [20]: logL=np.log(L)

In [21]: logK=np.log(K)

In [22]: import statsmodels.api as sm

In [23]: print(logQ,logL,logK)

[4.38202663 5.01063529 4.90527478 5.10594547 4.55387689 4.86753445
 4.70048037] [4.09434456 4.60517019 4.60517019 4.78749174 4.24849524 4.49980967
 4.38202663] [3.91202301 4.60517019 4.38202663 4.60517019 4.09434456 4.38202663
 4.24849524]
```

d.

```
In [24]: import statsmodels.api as sm
```

```
#Q2d. fitting multiple regression model
y=logdata['Output, Q']
x1=logdata['Labour, L']
x2=logdata['Capital, K']
model=sm.ols(formula="y~x1+x2",data=logdata).fit()
print(model.params)
```

```
Intercept    0.146233
x1            0.548427
x2            0.508741
dtype: float64
```

f.

```
#Q2f. calculating adjusted R square of model
print(model.summary())
```

```

=====
                    OLS Regression Results
=====
Dep. Variable:          y      R-squared:          0.998
Model:                  OLS    Adj. R-squared:      0.997
Method:                 Least Squares    F-statistic:      958.4
Date:                   Thu, 16 May 2019    Prob (F-statistic):  4.34e-06
Time:                   01:11:31    Log-Likelihood:     21.712
No. Observations:       7      AIC:              -37.42
Df Residuals:           4      BIC:              -37.59
Df Model:                2
Covariance Type:        nonrobust
=====
                    coef      std err          t      P>|t|      [0.025      0.975]
-----
Intercept      0.1462      0.114      1.282      0.269      -0.170      0.463
x1              0.5484      0.090      6.127      0.004      0.300      0.797
x2              0.5087      0.083      6.150      0.004      0.279      0.738
=====
Omnibus:          nan    Durbin-Watson:      2.801
Prob(Omnibus):    nan    Jarque-Bera (JB):    2.261
Skew:             1.352    Prob(JB):            0.323
Kurtosis:         3.665    Cond. No.            160.
=====

```