

```
In [1]: import numpy as np
import pandas as pd
import scipy.stats as stats
import math
from scipy.stats import norm
from scipy.stats import chi2_contingency
```

```
In [2]: data=pd.read_csv('cutlets.csv')
```

```
In [3]: data
```

```
Out[3]:
```

	Unit A	Unit B
0	6.8090	6.7703
1	6.4376	7.5093
2	6.9157	6.7300
3	7.3012	6.7878
4	7.4488	7.1522
5	7.3871	6.8110
6	6.8755	7.2212
7	7.0621	6.6606
8	6.6840	7.2402
9	6.8236	7.0503
10	7.3930	6.8810
11	7.5169	7.4059
12	6.9246	6.7652
13	6.9256	6.0380
14	6.5797	7.1581
15	6.8394	7.0240
16	6.5970	6.6672
17	7.2705	7.4314
18	7.2828	7.3070
19	7.3495	6.7478
20	6.9438	6.8889
21	7.1560	7.4220
22	6.5341	6.5217
23	7.2854	7.1688

24	6.9952	6.7594
25	6.8568	6.9399
26	7.2163	7.0133
27	6.6801	6.9182
28	6.9431	6.3346
29	7.0852	7.5459
30	6.7794	7.0992
31	7.2783	7.1180
32	7.1561	6.6965
33	7.3943	6.5780
34	6.9405	7.3875

**Assume H_0 =There is no difference in diameter of cutlets between two units.
 H_1 =There is significant Difference in diameter of cutlets between two units.**

```
In [4]: UnitA=pd.Series(data.iloc[:,0])  
UnitA
```

```
Out[4]: 0      6.8090  
1      6.4376  
2      6.9157  
3      7.3012  
4      7.4488  
5      7.3871  
6      6.8755  
7      7.0621  
8      6.6840  
9      6.8236  
10     7.3930  
11     7.5169  
12     6.9246  
13     6.9256  
14     6.5797  
15     6.8394  
16     6.5970  
17     7.2705  
18     7.2828  
19     7.3495  
20     6.9438  
21     7.1560  
22     6.5341  
23     7.2854  
24     6.9952  
25     6.8568  
26     7.2163  
27     6.6801  
28     6.9431  
29     7.0852  
30     6.7794  
31     7.2783  
32     7.1561  
33     7.3943  
34     6.9405  
Name: Unit A, dtype: float64
```

```
In [5]: UnitB=pd.Series(data.iloc[:,1])
UnitB
```

```
Out[5]: 0      6.7703
1      7.5093
2      6.7300
3      6.7878
4      7.1522
5      6.8110
6      7.2212
7      6.6606
8      7.2402
9      7.0503
10     6.8810
11     7.4059
12     6.7652
13     6.0380
14     7.1581
15     7.0240
16     6.6672
17     7.4314
18     7.3070
19     6.7478
20     6.8889
21     7.4220
22     6.5217
23     7.1688
24     6.7594
25     6.9399
26     7.0133
27     6.9182
28     6.3346
29     7.5459
30     7.0992
31     7.1180
32     6.6965
33     6.5780
34     7.3875
Name: Unit B, dtype: float64
```

```
In [6]: p_value=stats.ttest_ind(UnitA,UnitB)
p_value
```

```
Out[6]: Ttest_indResult(statistic=0.7228688704678063, pvalue=0.4722394724599501)
```

```
In [7]: p_value[1]
```

```
Out[7]: 0.4722394724599501
```

```
In [8]: if p_value[1] < 0.05:
        print('There is significant Difference in diameter of cutlets b
        else:
        print('There is no difference in diameter of cutlets between two
```

There is no difference in diameter of cutlets between two units.

Conclusion is as p value is 0.4722 which is greater than $\alpha = 0.05$, we accept null hypothesis. therefore, $\text{mean}_1 = \text{mean}_2$ which means there is no significant difference in diameters of the cutlets between the two units.

```
In [9]: data=pd.read_csv('LabTAT.csv')
```

```
In [10]: data
```

```
Out[10]:
```

	Laboratory 1	Laboratory 2	Laboratory 3	Laboratory 4
0	185.35	165.53	176.70	166.13
1	170.49	185.91	198.45	160.79
2	192.77	194.92	201.23	185.18
3	177.33	183.00	199.61	176.42
4	193.41	169.57	204.63	152.60
...
115	178.49	170.66	193.80	172.68
116	176.08	183.98	215.25	177.64
117	202.48	174.54	203.99	170.27
118	182.40	197.18	194.52	150.87
119	182.09	215.17	221.49	162.21

120 rows × 4 columns

In [11]: `data.describe()`

Out [11]:

	Laboratory 1	Laboratory 2	Laboratory 3	Laboratory 4
count	120.000000	120.000000	120.000000	120.000000
mean	178.361583	178.902917	199.913250	163.68275
std	13.173594	14.957114	16.539033	15.08508
min	138.300000	140.550000	159.690000	124.06000
25%	170.335000	168.025000	188.232500	154.05000
50%	178.530000	178.870000	199.805000	164.42500
75%	186.535000	189.112500	211.332500	172.88250
max	216.390000	217.860000	238.700000	205.18000

Null hypothesis(H_0)=mean TAT for 4 laboratory is equal($\text{mean}_1=\text{mean}_2=\text{mean}_3=\text{mean}_4$), There is no significance difference between TAT of laboratories.

Alternate hypothesis(H_a)= If mean is not equal, there is significance difference between TAT of laboratories.

In [14]: `#Anova test`
`p_value = stats.f_oneway(data.iloc[:,0],data.iloc[:,1],data.iloc[:,2],data.iloc[:,3])`
`print("The p-value is ",p_value[1])`

The p-value is 2.1156708949992414e-57

In [15]: `#compare p value with $\alpha = 0.05$`
`print(p_value[1]>0.05)`

False

conclusion: p - value is less than 0.05. Reject null hypothesis. There is significance difference between TAT laboratories.

In [17]: `data=pd.read_csv('BuyerRatio.csv')`
`data`

Out [17]:

	Observed Values	East	West	North	South
0	Males	50	142	131	70
1	Females	435	1523	1356	750

```
In [18]: obs=np.array([[50,142,131,70],[435,1523,1356,750]])
obs
```

```
Out[18]: array([[ 50, 142, 131,  70],
                [ 435, 1523, 1356,  750]])
```

Assume H_0 =All Proportions are equal. H_1 =Not all Proportions are equal.

```
In [19]: chi2_contingency(obs)
```

```
Out[19]: (1.595945538661058,
          0.6603094907091882,
          3,
          array([[ 42.76531299, 146.81287862, 131.11756787,  72.3042405
2],
                [ 442.23468701, 1518.18712138, 1355.88243213,  747.6957594
8]]))
```

```
In [20]: p_value=chi2_contingency(obs)[1]
p_value
```

```
Out[20]: 0.6603094907091882
```

```
In [21]: #Since p_value>0.05 we accept the null hypothesis i.e. All proporti
```

```
In [22]: from scipy.stats import chi2_contingency
```

```
In [23]: data=pd.read_csv('downloads/Custommer+OrderForm.csv')
```

```
In [24]: data.head(10)
```

```
Out[24]:
```

	Phillippines	Indonesia	Malta	India
0	Error Free	Error Free	Defective	Error Free
1	Error Free	Error Free	Error Free	Defective
2	Error Free	Defective	Defective	Error Free
3	Error Free	Error Free	Error Free	Error Free
4	Error Free	Error Free	Defective	Error Free
5	Error Free	Error Free	Error Free	Error Free
6	Error Free	Defective	Error Free	Error Free
7	Error Free	Error Free	Error Free	Error Free
8	Error Free	Error Free	Error Free	Error Free
9	Error Free	Error Free	Error Free	Error Free

```
In [25]: data.Phillippines.value_counts()
```

```
Out[25]: Error Free      271
         Defective       29
         Name: Phillippines, dtype: int64
```

```
In [26]: data.Indonesia.value_counts()
```

```
Out[26]: Error Free      267
         Defective       33
         Name: Indonesia, dtype: int64
```

```
In [27]: data.Malta.value_counts()
```

```
Out[27]: Error Free      269
         Defective       31
         Name: Malta, dtype: int64
```

```
In [28]: data.India.value_counts()
```

```
Out[28]: Error Free      280
         Defective       20
         Name: India, dtype: int64
```

```
In [29]: ot=np.array([[271,267,269,280],[29,33,31,20]])
         ot
```

```
Out[29]: array([[271, 267, 269, 280],
                [ 29,  33,  31,  20]])
```

Assume H_0 =Independance of categorical variable. H_1 =Dependence of categorical variable

```
In [30]: chi2_contingency(ot)
```

```
Out[30]: (3.858960685820355,
         0.2771020991233135,
         3,
         array([[271.75, 271.75, 271.75, 271.75],
                [ 28.25,  28.25,  28.25,  28.25]]))
```

```
In [31]: p_value=0.277
```

```
In [32]: if p_value < 0.05:
         print('dependance of categorical variable')
         else:
         print('Independance of categorical variable')
```

Independance of categorical variable

In []: