**Example 1** - A F&B Manager wants to determine whether there is any significant differnce in the diameter of the cutlet between two units. A randomly selected sample of cutlets was colleted from both the units & measured? Analyze the data & draw Inferences at 5% of level of significance. Please state the assumptions & tests that you carried out to check validity of the assumptions.

### **Hypothesis Formulation**

- 1. Null Hypothesis (H0): Diameter of Cultet between 2 units is same that is there is no any sigificant difference between Diameter of Cultet between 2 units.
- 2. Alternative Hypothesis (H1): Diameter of Cultet between 2 units is different that is there is sigificant difference between Diameter of Cultet between 2 units.

#### ▼ 1. Import Necessary Libraries

```
import pandas as pd
import numpy as np
from scipy.stats import f_oneway
from scipy import stats
```

#### → 2. Import Data

```
cutlets_data = pd.read_csv('Cutlets.csv')
cutlets data.head()
```

	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

# → 3. Data Understanding

	Unit A	Unit B
count	35.000000	35.000000
mean	7.019091	6.964297
std	0.288408	0.343401
min	6.437600	6.038000
25%	6.831500	6.753600
50%	6.943800	6.939900
75%	7.280550	7.195000
max	7.516900	7.545900

#### 

```
stats, p_value = f_oneway(cutlets_data['Unit A'],cutlets_data['Unit B'])
p_value

0.47223947245995734
```

#### → 5. Conclusion

```
# At the 5 % significance level

if p_value < 0.05:
    print('Reject the Null Hypothesis because there is significant difference between diameter of cutlets of 2 units. It means Diameter else:
    print('Do not Reject the Null Hypothesis because there is no any significant difference between diameter of cutlets of 2 units. It

Do not Reject the Null Hypothesis because there is no any significant difference between diameter of cutlets of 2 units. It mean

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```

Example 2: A hospital wants to determine whether there is any difference in the average Turn Around Time(TAT) of reports of the laboratories on their preferred list. They Collected a random

sample and recorded TAT for reports of 4 Laboratories. TAT is defined as sample collected to report dispatch.

Analyze the data & determine whether there is any difference in average TAT among the different laboratories at 5% significance level.

### Hypothesis Formulation

- 1. Null Hypothesis (H0): There is no any significant difference between average TAT among the different Laboratories.
- 2. Alternative Hypothesis (H1): There is significant difference between average TAT among the different Laboratories.

### ▼ 1. Import Data

```
tat_data = pd.read_csv('LabTAT.csv')
tat_data
```

	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

tat\_data.describe()

	Laboratory 1	Laboratory 2	Laboratory 3	Laboratory 4
count	120.000000	120.000000	120.000000	120.00000
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

### → Perform ANOVA Test

stats, p\_value\_tat = f\_oneway(tat\_data['Laboratory 1'],tat\_data['Laboratory 2'],tat\_data['Laboratory 3'],tat\_data['Laboratory 4'])
p\_value\_tat

2.1156708949992414e-57

#### Conclusion

```
if p_value_tat < 0.05:
    print('Reject the Null Hypothesis because There is significant difference between average TAT among the different Laboratories.')
else:
    print('Do not Reject the Null Hypothesis because There is no any significant difference between average TAT among the different Lab
    Reject the Null Hypothesis because There is significant difference between average TAT among the different Laboratories.
```

Example 3 - Sales of products in four different regions is tabulated for males and females. Find if male-female buyer ratios are similar across regions.

H0: All proportions are equal.

H1: Not all Proportions are equal

#### ▼ Import Data

```
sales_data = pd.read_csv('BuyerRatio.csv')
sales_data
```

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

```
data = [[50,142,131,70],[435,1523,1356,750]]
```

## ▼ Perform Chi-Squared Test

#### - Conclusion

```
# Level of Significance is 5% so alpha =0.05
if p_value_sales < 0.05:
   print("Reject the Null Hypothesis because there is significant difference between buyer's ratio of male & female")
else:
   print("We do not Reject the Null Hypothesis because there is no significant difference between buyer's ratio of male & female")</pre>
```

We do not Reject the Null Hypothesis because there is no significant difference between buyer's ratio of male & female

Example 4 - TeleCall uses 4 centers around the globe to process customer order forms. They audit a certain % of the customer order forms. Any error in order form renders it defective and has to be reworked before processing. The manager wants to check whether the defective % varies by centre. Please analyze the data at 5% significance level and help the manager draw appropriate inferences.

Hypothesis Formulation

H0 (Null Hypothesis): Customer order forms defective % does not varies by centre

H1 (Alternate Hypothesis): Customer order forms defective % varies by centre

## Import Data

```
customer_data = pd.read_csv('/content/Costomer+OrderForm.csv')
customer data
```

	Phillip	pines	Indonesia	Malta	India	
0	Erro	r Free	Error Free	Defective	Error Free	
1	Erro	r Free	Error Free	Error Free	Defective	
2	e Erro	r Free	Defective	Defective	Error Free	
3	Erro	r Free	Error Free	Error Free	Error Free	
4	Erro	r Free	Error Free	Defective	Error Free	
29	5 Erro	r Free	Error Free	Error Free	Error Free	
customer	_data.Phil	lippine	es.value_co	unts()		
Error Free 271 Defective 29 Name: Phillippines, dtype: int64  customer_data.Indonesia.value_counts()						
Error Free 267 Defective 33 Name: Indonesia, dtype: int64						
<pre>customer_data.Malta.value_counts()</pre>						
Def	ective e: Malta,	269 31 dtype:	int64			
<pre>customer_data.India.value_counts()</pre>						
	or Free ective	280 20				

Name: India, dtype: int64

# → Perform Independent Chi-Squared Test

```
chi score, p value sales, dof, expected table = stats.chi2 contingency(observed = np.array([[271,267,269,280],[29,33,31,20]]))
                        :',chi score.round(5))
print('Chi Score
print('p value sales
                        :',p value sales.round(5))
print('Degree of Freedom :',dof)
print('Expected Table
                        :\n',expected table)
     Chi Score
                       : 3.85896
     p value sales
                      : 0.2771
     Degree of Freedom: 3
     Expected Table
     [[271.75 271.75 271.75]
     [ 28.25 28.25 28.25 28.25]]
```

#### → Conclusion

```
# At Level of Significance is 5% so alpha =0.05
if p_value_sales < 0.05:
    print("Reject the Null Hypothesis because there is significant difference. Thus, customer order forms defective % varies by centre.
else:
    print("Do not reject the Null Hypthesis because there is no any significant difference. Thus, customer order forms defective % does

Do not reject the Null Hypthesis because there is no any significant difference. Thus, customer order forms defective % does no.
```

#### Inference:

As  $(p_value = 0.2771) > (\alpha = 0.05)$ :

Do not reject the Null Hypthesis because there is no any significant difference. Thus, customer order forms defective % does not varies by centre.

THE END!!!!

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