

## Importing Libraries

```
In [19]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

In [58]: from scipy.stats import f_oneway,kruskal,ttest_ind,levene,shapiro
```

## ANOVA

```
In [3]: a = np.array([25, 25, 27, 30, 23, 20])
b = np.array([30, 30, 21, 24, 26, 28])
c = np.array([18, 30, 29, 29, 24, 26])

In [2]: print("a : ", a)
print("b : ", b)
print("c : ", c)

a : [25 25 27 30 23 20]
b : [30 30 21 24 26 28]
c : [18 30 29 29 24 26]

In [4]: print("Avg(a) :",np.mean(a))
print("Avg(b) :",np.mean(b))
print("Avg(c) :",np.mean(c))

Avg(a) : 25.0
Avg(b) : 26.5
Avg(c) : 28.0

In [5]: (6*((25-25.83)**2))+6*((26.5-25.83)**2)+(6*((26-25.83)**2))

Out[5]: 7.8092

In [9]: np.sum((a-a.mean()))**2+np.sum((b-b.mean()))**2+np.sum((c-c.mean()))**2

Out[9]: 223.5

In [10]: np.sum((a-a.mean()))**2

Out[10]: 58.0

In [14]: p_value=1-f.cdf(0.2348,df=2,dfd=15)
print("p_value : ",p_value)
alpha = 0.05
if p_value< alpha:
    print("Interpretation : Reject Ho")
    print("Conclusion : Gender Affects the buying pattern")
else:
    print("Interpretation : Fail to Reject Ho")

p_value : 0.7935810912142831
Interpretation : Fail to Reject Ho

In [17]: f_stat,p_value=f_oneway(a,b,c)
print("f_stat : ",f_stat)
print("p_value : ",p_value)
alpha = 0.05
if p_value< alpha:
    print("Interpretation : Reject Ho")
    print("Conclusion : Gender Affects the buying pattern")
else:
    print("Interpretation : Fail to Reject Ho")

f_stat : 0.2348993288598604
p_value : 0.793584602720233
Interpretation : Fail to Reject Ho
```

## Aerofit --> Anova

```
In [20]: df=pd.read_csv("aerofit.csv")
df

Out[20]:
```

	Product	Age	Gender	Education	MaritalStatus	Usage	Fitness	Income	Miles
0	KP281	18	Male	14	Single	3	4	29562	112
1	KP281	19	Male	15	Single	2	3	31836	75
2	KP281	19	Female	14	Partnered	4	3	30699	66
3	KP281	19	Male	12	Single	3	3	32973	85
4	KP281	20	Male	13	Partnered	4	2	35247	47
...	...	...	...	...	...	...	...	...	...
175	KP781	40	Male	21	Single	6	5	83416	200
176	KP781	42	Male	18	Single	5	4	89641	200
177	KP781	45	Male	16	Single	5	5	90886	160
178	KP781	47	Male	18	Partnered	4	5	104581	120
179	KP781	48	Male	18	Partnered	4	5	95508	180

180 rows x 9 columns

```
In [21]: df["Product"].unique()

Out[21]: array(['KP281', 'KP481', 'KP781'], dtype=object)

In [22]: sns.boxplot(x="Product",y="Income",data=df)

Out[22]:
```

```
In [31]: df["random_group"]=np.random.choice(["g1","g2","g3"],size=len(df))
df

Out[31]:
```

	Product	Age	Gender	Education	MaritalStatus	Usage	Fitness	Income	Miles	random_group
0	KP281	18	Male	14	Single	3	4	29562	112	g1
1	KP281	19	Male	15	Single	2	3	31836	75	g2
2	KP281	19	Female	14	Partnered	4	3	30699	66	g2
3	KP281	19	Male	12	Single	3	3	32973	85	g2
4	KP281	20	Male	13	Partnered	4	2	35247	47	g1
...	...	...	...	...	...	...	...	...	...	...
175	KP781	40	Male	21	Single	6	5	83416	200	g2
176	KP781	42	Male	18	Single	5	4	89641	200	g2
177	KP781	45	Male	16	Single	5	5	90886	160	g3
178	KP781	47	Male	18	Partnered	4	5	104581	120	g1
179	KP781	48	Male	18	Partnered	4	5	95508	180	g1

180 rows x 10 columns

```
In [32]: sns.boxplot(x="random_group",y="Income",data=df)

Out[32]:
```

```
In [33]: income_g1 = df[df["random_group"]=="g1"]["Income"]
income_g2 = df[df["random_group"]=="g2"]["Income"]
income_g3 = df[df["random_group"]=="g3"]["Income"]

In [34]: # Ho : All have the same means
# Ha : Atleast one of them is different
f_stat,p_value=f_oneway(income_g1,income_g2,income_g3)
print("f_stat : ",f_stat)
print("p_value : ",p_value)
alpha = 0.05
if p_value< alpha:
    print("Interpretation : Reject Ho")
    print("Conclusion : Gender Affects the buying pattern")
else:
    print("Interpretation : Fail to Reject Ho")

f_stat : 0.965286271164961
p_value : 0.684245658417896
Interpretation : Fail to Reject Ho
```

```
In [37]: print("income_g1 Mean :",income_g1.mean())
print("income_g2 Mean :",income_g2.mean())
print("income_g3 Mean :",income_g3.mean())

income_g1 Mean : 53949.056683773584
income_g2 Mean : 52391.07142857143
income_g3 Mean : 55248.228070175435
```

```
In [35]: income_281 = df[df["Product"]=="KP281"]["Income"]
income_481 = df[df["Product"]=="KP481"]["Income"]
income_781 = df[df["Product"]=="KP781"]["Income"]
```

```
In [36]: print("income_281 Mean :",income_281.mean())
print("income_481 Mean :",income_481.mean())
print("income_781 Mean :",income_781.mean())

income_281 Mean : 48118.825
income_481 Mean : 49971.65
income_781 Mean : 75441.575
```

```
In [38]: # Ho : All have the same means
# Ha : Atleast one of them is different
f_stat,p_value=f_oneway(income_281,income_481,income_781)
print("f_stat : ",f_stat)
print("p_value : ",p_value)
alpha = 0.05
if p_value< alpha:
    print("Interpretation : Reject Ho")
    print("Conclusion : Atleast one of them is different")
else:
    print("Interpretation : Fail to Reject Ho")

f_stat : 89.259803546681671
p_value : 1.5644981318342484e-27
Interpretation : Reject Ho
Conclusion : Atleast one of them is different
```

```
In [41]: # Ho : All have the same means
# Ha : Atleast one of them is different
f_stat,p_value=kruskal(income_281,income_481,income_781)
print("f_stat : ",f_stat)
print("p_value : ",p_value)
alpha = 0.05
if p_value< alpha:
    print("Interpretation : Reject Ho")
    print("Conclusion : Atleast one of them is different")
else:
    print("Interpretation : Fail to Reject Ho")

f_stat : 61.43679384567185
p_value : 4.562357014275808e-14
Interpretation : Reject Ho
Conclusion : Atleast one of them is different
```

```
In [42]: sns.kdeplot(x="Income",hue="Product",data=df)

Out[42]:
```

```
In [43]: sns.kdeplot(x="Income",hue="random_group",data=df)

Out[43]:
```

```
In [44]: # Ho : All are similar
# Ha : Atleast one of them is different
f_stat,p_value=kruskal(income_g1,income_g2,income_g3)
print("f_stat : ",f_stat)
print("p_value : ",p_value)
alpha = 0.05
if p_value< alpha:
    print("Interpretation : Reject Ho")
    print("Conclusion : Atleast one of them is different")
else:
    print("Interpretation : Fail to Reject Ho")

f_stat : 0.318749188484798
p_value : 0.8559550163838188
Interpretation : Fail to Reject Ho
```

## Levene Test

```
In [45]: df1=pd.read_csv("weight-height.csv")
df1

Out[45]:
```

	Gender	Height	Weight
0	Male	73.847017	241.893563
1	Male	68.781904	162.310473
2	Male	74.110105	212.740856
3	Male	71.730978	220.042470
4	Male	69.881796	206.348901
...	...	...	...
9995	Female	66.172652	126.777454
9996	Female	67.067155	170.867906
9997	Female	63.867992	128.475319
9998	Female	69.034243	163.852461
9999	Female	61.944248	113.648193

10000 rows x 3 columns

```
In [46]: height_men = df1[df1["Gender"]=="Male"]["Height"]
height_women = df1[df1["Gender"]=="Female"]["Height"]

In [47]: sns.histplot(data=df1, x="Height", hue="Gender")

Out[47]:
```

```
In [49]: height_men.mean()

Out[49]: 69.02634590621737

In [48]: height_women.mean()

Out[48]: 63.788773963424916

In [52]: ttest_ind(height_men,height_women)

Out[52]: Ttest_indResult(statistic=95.68271449148823, pvalue=0.0)

In [54]: height_men.var()

Out[54]: 8.19884325252049

In [53]: height_women.var()

Out[53]: 7.269947493670132

In [56]: # Ho : Variances are Equal
# Ha : Variances are NOT Equal
x_stat,p_value=levene(height_men,height_women)
print("x_stat : ",x_stat)
print("p_value : ",p_value)
alpha = 0.05
if p_value< alpha:
    print("Interpretation : Reject Ho")
    print("Conclusion : Variances are NOT Equal")
else:
    print("Interpretation : Fail to Reject Ho")
    print("Conclusion : Variances are Equal")

x_stat : 12.28491868677703
p_value : 0.8084689346885436178
Interpretation : Reject Ho
Conclusion : Variances are NOT Equal
```

```
In [ ]:
```

## Wilkin Shapiro Test

Take a few samples generally [50-200]

The test may not work if data points are <50 or Greater >200

```
In [57]: height_sample=df1["Height"]
height_sample

Out[57]:
```

3736	71.415165
9496	62.527229
6374	69.878586
2445	66.446420
1954	76.116675
...	...
7382	72.429771
3833	62.202788
2650	71.918912
2974	72.286541
9710	69.107087

Name: Height, Length: 180, dtype: float64

```
In [59]: # Ho : Data is Gaussian
# Ha : Data is NOT Gaussian
shapiro(height_sample)

Out[59]: ShapiroResult(statistic=0.977449454112244, pvalue=0.68472525328397751)
```

```
In [60]: sns.displot(df1["Height"],kde=True)

Out[60]:
```

```
In [62]: sns.kdeplot(df1["Height"])

Out[62]:
```

```
In [ ]:
```

## Doubts

```
In [67]: a = np.array([80,75,82,85,90,82])
b = np.array([57,59,62,85,64,87])
c = np.array([69,71,72,73,66,77])

In [68]: a.mean()

Out[68]: 82.33333333333333

In [69]: b.mean()

Out[69]: 60.5

In [70]: c.mean()

Out[70]: 71.53333333333333

In [72]: sns.kdeplot(a)
sns.kdeplot(b)
sns.kdeplot(c)

Out[72]:
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