Day 12 Assignment - Hitik Panchal

In [23]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
from scipy.stats import mannwhitneyu
from scipy.stats import ttest_ind
```

Reading the Data

In [7]:

```
gen_data=pd.read_csv('general_data.csv')
gen_data.head()
```

Out[7]:

| | Age | Attrition | BusinessTravel | Department | DistanceFromHome | Education | EducationFi€ |
|---|-----|-----------|-------------------|------------------------|------------------|-----------|--------------|
| 0 | 51 | No | Travel_Rarely | Sales | 6 | 2 | Life Scienc |
| 1 | 31 | Yes | Travel_Frequently | Research & Development | 10 | 1 | Life Scienc |
| 2 | 32 | No | Travel_Frequently | Research & Development | 17 | 4 | Oth |
| 3 | 38 | No | Non-Travel | Research & Development | 2 | 5 | Life Scienc |
| 4 | 32 | No | Travel_Rarely | Research & Development | 10 | 1 | Medic |

5 rows × 24 columns

In [8]:

```
gen_data.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4410 entries, 0 to 4409
```

| Data | columns (total 24 columns | s): | | | | | |
|-------------------------|---------------------------|----------------|---------|--|--|--|--|
| # | Column | Non-Null Count | Dtype | | | | |
| | | | | | | | |
| 0 | Age | 4410 non-null | int64 | | | | |
| 1 | Attrition | 4410 non-null | object | | | | |
| 2 | BusinessTravel | 4410 non-null | object | | | | |
| 3 | Department | 4410 non-null | object | | | | |
| 4 | DistanceFromHome | 4410 non-null | int64 | | | | |
| 5 | Education | 4410 non-null | int64 | | | | |
| 6 | EducationField | 4410 non-null | object | | | | |
| 7 | EmployeeCount | 4410 non-null | int64 | | | | |
| 8 | EmployeeID | 4410 non-null | int64 | | | | |
| 9 | Gender | 4410 non-null | object | | | | |
| 10 | JobLevel | 4410 non-null | int64 | | | | |
| 11 | JobRole | 4410 non-null | object | | | | |
| 12 | MaritalStatus | 4410 non-null | object | | | | |
| 13 | MonthlyIncome | 4410 non-null | int64 | | | | |
| 14 | NumCompaniesWorked | 4391 non-null | float64 | | | | |
| 15 | Over18 | 4410 non-null | object | | | | |
| 16 | PercentSalaryHike | 4410 non-null | int64 | | | | |
| 17 | StandardHours | 4410 non-null | int64 | | | | |
| 18 | StockOptionLevel | 4410 non-null | int64 | | | | |
| 19 | TotalWorkingYears | 4401 non-null | float64 | | | | |
| 20 | TrainingTimesLastYear | 4410 non-null | int64 | | | | |
| 21 | YearsAtCompany | 4410 non-null | int64 | | | | |
| 22 | YearsSinceLastPromotion | 4410 non-null | int64 | | | | |
| 23 | YearsWithCurrManager | 4410 non-null | int64 | | | | |
| dtype | es: float64(2), int64(14) | , object(8) | | | | | |
| memory usage: 827 0+ KB | | | | | | | |

memory usage: 827.0+ KB

Cleaning the Data

In [9]:

```
gen_data.isnull().any()
```

Out[9]:

False Age Attrition False BusinessTravel False False Department DistanceFromHome False Education False EducationField False EmployeeCount False **EmployeeID** False Gender False JobLevel False JobRole False MaritalStatus False MonthlyIncome False NumCompaniesWorked True Over18 False PercentSalaryHike False StandardHours False StockOptionLevel False TotalWorkingYears True TrainingTimesLastYear False YearsAtCompany False YearsSinceLastPromotion False YearsWithCurrManager False dtype: bool

In [10]:

gen_data.fillna(0 , inplace=True)

In [11]:

```
gen_data.isnull().any()
```

Out[11]:

False Age Attrition False BusinessTravel False False Department DistanceFromHome False Education False EducationField False EmployeeCount False **EmployeeID** False Gender False JobLevel False JobRole False MaritalStatus False MonthlyIncome False NumCompaniesWorked False Over18 False PercentSalaryHike False StandardHours False StockOptionLevel False TotalWorkingYears False TrainingTimesLastYear False YearsAtCompany False YearsSinceLastPromotion False YearsWithCurrManager False dtype: bool

In [12]:

gen_data.duplicated()

Out[12]:

0 False False 1 2 False 3 False 4 False 4405 False 4406 False 4407 False 4408 False 4409 False

Length: 4410, dtype: bool

In [13]:

gen_data.drop_duplicates()

Out[13]:

| | Age | Attrition | BusinessTravel | Department | DistanceFromHome | Education | Education |
|------|-----|-----------|-------------------|------------------------|------------------|-----------|-----------|
| 0 | 51 | No | Travel_Rarely | Sales | 6 | 2 | Life Sci |
| 1 | 31 | Yes | Travel_Frequently | Research & Development | 10 | 1 | Life Sci |
| 2 | 32 | No | Travel_Frequently | Research & Development | 17 | 4 | |
| 3 | 38 | No | Non-Travel | Research & Development | 2 | 5 | Life Sci |
| 4 | 32 | No | Travel_Rarely | Research & Development | 10 | 1 | N |
| | | | | | | | |
| 4405 | 42 | No | Travel_Rarely | Research & Development | 5 | 4 | N |
| 4406 | 29 | No | Travel_Rarely | Research & Development | 2 | 4 | N |
| 4407 | 25 | No | Travel_Rarely | Research & Development | 25 | 2 | Life Sci |
| 4408 | 42 | No | Travel_Rarely | Sales | 18 | 2 | N |
| 4409 | 40 | No | Travel_Rarely | Research & Development | 28 | 3 | N |

4410 rows × 24 columns

In [20]:

```
yes_data = pd.DataFrame(gen_data[gen_data['Attrition']=='Yes'])
yes_data
```

Out[20]:

| | Age | Attrition | BusinessTravel | Department | DistanceFromHome | Education | Educatio |
|------|-----|-----------|-------------------|------------------------|------------------|-----------|----------|
| 1 | 31 | Yes | Travel_Frequently | Research & Development | 10 | 1 | Life Sci |
| 6 | 28 | Yes | Travel_Rarely | Research & Development | 11 | 2 | N |
| 13 | 47 | Yes | Non-Travel | Research & Development | 1 | 1 | N |
| 28 | 44 | Yes | Travel_Frequently | Research & Development | 1 | 2 | N |
| 30 | 26 | Yes | Travel_Rarely | Research & Development | 4 | 3 | N |
| | | ••• | | ••• | | | |
| 4381 | 29 | Yes | Travel_Rarely | Research & Development | 7 | 1 | Life Sci |
| 4386 | 33 | Yes | Travel_Rarely | Sales | 11 | 4 | Mar |
| 4388 | 33 | Yes | Travel_Rarely | Sales | 1 | 3 | Life Sci |
| 4391 | 32 | Yes | Travel_Rarely | Sales | 23 | 1 | Life Sci |
| 4402 | 37 | Yes | Travel_Frequently | Sales | 2 | 3 | Mar |

711 rows × 24 columns

In [22]:

```
no_data = pd.DataFrame(gen_data[gen_data['Attrition']=='No'])
no_data
```

Out[22]:

| | Age | Attrition | BusinessTravel | Department | DistanceFromHome | Education | Educatio |
|------|-----|-----------|-------------------|------------------------|------------------|-----------|----------|
| 0 | 51 | No | Travel_Rarely | Sales | 6 | 2 | Life Sc |
| 2 | 32 | No | Travel_Frequently | Research & Development | 17 | 4 | |
| 3 | 38 | No | Non-Travel | Research & Development | 2 | 5 | Life Sci |
| 4 | 32 | No | Travel_Rarely | Research & Development | 10 | 1 | N |
| 5 | 46 | No | Travel_Rarely | Research & Development | 8 | 3 | Life Sci |
| | ••• | | | | | | |
| 4405 | 42 | No | Travel_Rarely | Research & Development | 5 | 4 | N |
| 4406 | 29 | No | Travel_Rarely | Research & Development | 2 | 4 | N |
| 4407 | 25 | No | Travel_Rarely | Research & Development | 25 | 2 | Life Sci |
| 4408 | 42 | No | Travel_Rarely | Sales | 18 | 2 | N |
| 4409 | 40 | No | Travel_Rarely | Research & Development | 28 | 3 | N |

3699 rows × 24 columns



Seperate T-Test

Test 1 : Attrition vs Distance From Home

In [33]:

```
a1 = yes_data['DistanceFromHome']
a2 = no_data['DistanceFromHome']

tval, p = ttest_ind(a2,a1)

print("The T value is : %.3f" % tval)
print("The p value is : %.3f" % p)
```

The T value is : 0.646 The p value is : 0.518

As the P value is 0.518, which is > than 0.05, the Ha is rejected and H0 is accepted

H0: There is no significant differences in the Distance From Home between Attrition (Y) and Attirition (N)

Ha: There is significant differences in the Distance From Home between Attrition (Y) and Attirition (N)

Test 2: Attrition vs Monthly Income

In [34]:

```
a1 = yes_data['MonthlyIncome']
a2 = no_data['MonthlyIncome']

tval, p = ttest_ind(a2,a1)

print("The T value is : %.3f" % tval)
print("The p value is : %.3f" % p)
```

The T value is : 2.071 The p value is : 0.038

As the P value is 0.038, which is < than 0.05, the Ha is accepted and H0 is rejected

H0: There is no significant differences in the Monthly Income between Attrition (Y) and Attirition (N)

Ha: There is significant differences in the Monthly Income between Attrition (Y) and Attirition (N)

Test 3: Attrition vs Years at Company

In [35]:

```
a1 = yes_data['YearsAtCompany']
a2 = no_data['YearsAtCompany']

tval, p = ttest_ind(a2,a1)

print("The T value is : %.3f" % tval)
print("The p value is : %.3f" % p)
```

The T value is : 9.004
The p value is : 0.000

As the P value is 0.0, which is < than 0.05, the Ha is accepted and H0 is rejected

H0: There is no significant differences in the Years at Company between Attrition (Y) and Attirition (N)

Ha: There is significant differences in the Years at Company between Attrition (Y) and Attirition (N)

Test 4: Attrition vs Years with Current Manager

In [32]:

```
a1 = yes_data['YearsWithCurrManager']
a2 = no_data['YearsWithCurrManager']

tval, p = ttest_ind(a2,a1)

print("The T value is : %.3f" % tval)
print("The p value is : %.3f" % p)
```

The T value is : 10.499
The p value is : 0.000

As the P value is 0.0, which is < than 0.05, the Ha is accepted and H0 is rejected

H0: There is no significant differences in the Years with Current Manager between Attrition (Y) and Attirition (N)

Ha: There is significant differences in the Years with Current Manager between Attrition (Y) and Attirition (N)

Mann-Whitney Test

Test 1: Attrition vs Distance From Home

In [37]:

```
a1 = yes_data['DistanceFromHome']
a2 = no_data['DistanceFromHome']
uval , p = mannwhitneyu(a1,a2)
print("The U value is : %.3f" % uval)
print("The p value is : %.3f" % p)
```

The U value is : 1312110.000 The p value is : 0.463

As the P value is 0.463, which is > than 0.05, the Ha is rejected and H0 is accepted

H0: There is no significant differences in the Distance From Home between Attrition (Y) and Attirition (N)

Ha: There is significant differences in the Distance From Home between Attrition (Y) and Attirition (N)

Test 2: Attrition vs Monthly Income

In [38]:

```
a1 = yes_data['MonthlyIncome']
a2 = no_data['MonthlyIncome']

uval , p = mannwhitneyu(a1,a2)

print("The U value is : %.3f" % uval)
print("The p value is : %.3f" % p)
```

The U value is : 1264900.500 The p value is : 0.054

As the P value is 0.054, which is > than 0.05, the Ha is rejected and H0 is accepted

H0: There is no significant differences in the Monthly Income between Attrition (Y) and Attirition (N)

Ha: There is significant differences in the Monthly Income between Attrition (Y) and Attirition (N)

Test 3: Attrition vs Years at Company

In [39]:

```
a1 = yes_data['YearsAtCompany']
a2 = no_data['YearsAtCompany']

uval , p = mannwhitneyu(a1,a2)

print("The U value is : %.3f" % uval)
print("The p value is : %.3f" % p)
```

The U value is : 923238.000 The p value is : 0.000

As the P value is 0.0, which is < than 0.05, the Ha is accepted and H0 is rejected

H0: There is no significant differences in the Years at Company between Attrition (Y) and Attrition (N)

Ha: There is significant differences in the Years at Company between Attrition (Y) and Attirition (N)

Test 4: Attrition vs Years with Current Manager

In [40]:

```
a1 = yes_data['YearsWithCurrManager']
a2 = no_data['YearsWithCurrManager']
uval , p = mannwhitneyu(a1,a2)
print("The U value is : %.3f" % uval)
print("The p value is : %.3f" % p)
```

The U value is : 957253.500 The p value is : 0.000

As the P value is 0.0, which is < than 0.05, the Ha is accepted and H0 is rejected

H0: There is no significant differences in the Years with Current Manager between Attrition (Y) and Attirition (N)

Ha: There is significant differences in the Years with Current Manager between Attrition (Y) and Attirition (N)

| In []: | | | |
|---------|--|--|--|
| | | | |