DATA ANALYSIS PROJECT ON H.R DATA-SET

Importing Libraries

```
In [53]: import pandas as pd
   import matplotlib.pyplot as plt
   %matplotlib inline
   import seaborn as sns
   import warnings
   warnings.filterwarnings('ignore')
```

Loading the dataset

```
In [38]: df = pd.read_csv(r'HR Data.csv')
   pd.set_option('display.max_columns', None)
   df
```

Out[38]:

	Attrition	Business Travel	CF_age band	CF_attrition label	Department	Education Field	emp no	Empl Nur
0	Yes	Travel_Rarely	35 - 44	Ex- Employees	Sales	Life Sciences	STAFF- 1	
1	No	Travel_Frequently	45 - 54	Current Employees	R&D	Life Sciences	STAFF- 2	
2	Yes	Travel_Rarely	35 - 44	Ex- Employees	R&D	Other	STAFF- 4	
3	No	Travel_Frequently	25 - 34	Current Employees	R&D	Life Sciences	STAFF- 5	
4	No	Travel_Rarely	25 - 34	Current Employees	R&D	Medical	STAFF- 7	
465	Yes	Non-Travel	25 - 34	Ex- Employees	R&D	Technical Degree	STAFF- 1905	
466	Yes	Travel_Frequently	25 - 34	Ex- Employees	R&D	Life Sciences	STAFF- 1868	
467	Yes	Travel_Frequently	35 - 44	Ex- Employees	Sales	Other	STAFF- 1667	
468	Yes	Travel_Rarely	Under 25	Ex- Employees	R&D	Life Sciences	STAFF- 1878	
469	Yes	Travel_Rarely	Under 25	Ex- Employees	Sales	Life Sciences	STAFF- 1702	

1470 rows × 39 columns

INTRODUCTION OF PROJECT

TARGET COLUMN == ATTRITION

-ATTRITION:-Attrition is the departure of employees from the organization for any reason or in simple words we can say that employees who want to leave the company.

-ATTRITION "YES" :- Employees wants to leave the company.

-ATTRITION "NO" :- Employees don't want to leave.

PROBLEM STATEMENT

This project aims to uncover the factors contributing to employee attrition. Through analysis of various dataset features, we seek to understand their impact on attrition rates within the organization.

Exploratory Data Analysis and Data Cleaning

In [110]: # Visualizing top 5 rows
df.head()

Out[110]:

	Attrition	Business Travel	CF_age band	CF_attrition label	Department	Education Field	emp no	Employed Numbe
0	Yes	Travel_Rarely	35 - 44	Ex- Employees	Sales	Life Sciences	STAFF- 1	
1	No	Travel_Frequently	45 - 54	Current Employees	R&D	Life Sciences	STAFF- 2	:
2	Yes	Travel_Rarely	35 - 44	Ex- Employees	R&D	Other	STAFF- 4	4
3	No	Travel_Frequently	25 - 34	Current Employees	R&D	Life Sciences	STAFF- 5	!
4	No	Travel_Rarely	25 - 34	Current Employees	R&D	Medical	STAFF- 7	•
4								

```
In [111]: # Visualizing bottom 5 rows
df.tail()
```

Out[111]:

	Attrition	Business Travel	CF_age band	CF_attrition label	Department	Education Field	emp no	Emple Nur
1465	Yes	Non-Travel	25 - 34	Ex- Employees	R&D	Technical Degree	STAFF- 1905	
1466	Yes	Travel_Frequently	25 - 34	Ex- Employees	R&D	Life Sciences	STAFF- 1868	
1467	Yes	Travel_Frequently	35 - 44	Ex- Employees	Sales	Other	STAFF- 1667	•
1468	Yes	Travel_Rarely	Under 25	Ex- Employees	R&D	Life Sciences	STAFF- 1878	•
1469	Yes	Travel_Rarely	Under 25	Ex- Employees	Sales	Life Sciences	STAFF- 1702	•
4								

In [112]: df.shape #shows number of rows and column

Out[112]: (1470, 39)

In [113]: df.columns #shows all column name in dataframe

```
Out[113]: Index(['Attrition', 'Business Travel', 'CF_age band', 'CF_attrition labe
          1',
                  'Department', 'Education Field', 'emp no', 'Employee Number', 'Gend
          er',
                  'Job Role', 'Marital Status', 'Over Time', 'Over18',
                  'Training Times Last Year', 'Age', 'CF_current Employee', 'Daily Ra
          te',
                  'Distance From Home', 'Education', 'Employee Count',
                  'Environment Satisfaction', 'Hourly Rate', 'Job Involvement',
                  'Job Level', 'Job Satisfaction', 'Monthly Income', 'Monthly Rate',
                  'Num Companies Worked', 'Percent Salary Hike', 'Performance Ratin
          g',
                  'Relationship Satisfaction', 'Standard Hours', 'Stock Option Leve
          1',
                  'Total Working Years', 'Work Life Balance', 'Years At Company',
                  'Years In Current Role', 'Years Since Last Promotion',
                  'Years With Curr Manager'],
                 dtype='object')
```

In [114]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1470 entries, 0 to 1469
Data columns (total 39 columns):

#	Column	Non-Null Count	Dtype
0	Attrition	1470 non-null	
1	Business Travel	1470 non-null	object
2	CF_age band	1470 non-null	object
3	CF_attrition label	1470 non-null	object
4	Department	1470 non-null	object
5	Education Field	1470 non-null	object
6	emp no	1470 non-null	object
7	Employee Number	1470 non-null	int64
8	Gender	1470 non-null	object
9	Job Role	1470 non-null	object
10	Marital Status	1470 non-null	object
11	Over Time	1470 non-null	object
12	Over18	1470 non-null	object
13	Training Times Last Year	1470 non-null	int64
14	Age	1470 non-null	int64
15	CF_current Employee	1470 non-null	int64
16	Daily Rate	1470 non-null	int64
17	Distance From Home	1470 non-null	int64
18	Education	1470 non-null	object
19	Employee Count	1470 non-null	int64
20	Environment Satisfaction	1470 non-null	int64
21	Hourly Rate	1470 non-null	int64
22	Job Involvement	1470 non-null	int64
23	Job Level	1470 non-null	int64
24	Job Satisfaction	1470 non-null	int64
25	Monthly Income	1470 non-null	int64
26	Monthly Rate	1470 non-null	int64
27	Num Companies Worked	1470 non-null	int64
28	Percent Salary Hike	1470 non-null	int64
29	Performance Rating	1470 non-null	
30	Relationship Satisfaction	1470 non-null	
31	Standard Hours	1470 non-null	int64
32	Stock Option Level	1470 non-null	int64
33	Total Working Years	1470 non-null	int64
34	Work Life Balance	1470 non-null	int64
35	Years At Company	1470 non-null	int64
36	Years In Current Role	1470 non-null	int64
37	Years Since Last Promotion	1470 non-null	int64
38	Years With Curr Manager	1470 non-null	int64
	es: int64(26), object(13)		
memoi	ry usage: 448.0+ KB		

localhost:8888/notebooks/OneDrive/Desktop/HR PROJECT/HR Project.ipynb#Conclusion

```
In [115]: | for col in df.describe(include='object').columns: # check all categorical columns
          print(col)
          print(df[col].unique())
          print('-'*50)
       Attrition
       ['Yes' 'No']
        -----
       Business Travel
       ['Travel_Rarely' 'Travel_Frequently' 'Non-Travel']
        -----
       CF age band
       ['35 - 44' '45 - 54' '25 - 34' 'Over 55' 'Under 25']
           CF_attrition label
       ['Ex-Employees' 'Current Employees']
       Department
       ['Sales' 'R&D' 'HR']
       Education Field
       ['Life Sciences' 'Other' 'Medical' 'Marketing' 'Technical Degree'
        'Human Resources']
       ['STAFF-1' 'STAFF-2' 'STAFF-4' ... 'STAFF-1667' 'STAFF-1878' 'STAFF-1702']
       _____
       Gender
       ['Female' 'Male']
        -----
       Job Role
       ['Sales Executive' 'Research Scientist' 'Laboratory Technician'
        'Manufacturing Director' 'Healthcare Representative' 'Manager'
        'Sales Representative' 'Research Director' 'Human Resources']
        -----
       Marital Status
       ['Single' 'Married' 'Divorced']
        _____
       Over Time
       ['Yes' 'No']
               -----
       Over18
       ['Y']
             -----
       Education
       ['Associates Degree' 'High School' "Master's Degree" "Bachelor's Degree"
        'Doctoral Degree']
```

In [116]: pd.isnull(df) # checking null values

Out[116]:

	Attrition	Business Travel	CF_age band	CF_attrition label	Department	Education Field	emp no	Employee Number	Gei
0	False	False	False	False	False	False	False	False	F
1	False	False	False	False	False	False	False	False	F
2	False	False	False	False	False	False	False	False	F
3	False	False	False	False	False	False	False	False	F
4	False	False	False	False	False	False	False	False	F
1465	False	False	False	False	False	False	False	False	F
1466	False	False	False	False	False	False	False	False	F
1467	False	False	False	False	False	False	False	False	F
1468	False	False	False	False	False	False	False	False	F
1469	False	False	False	False	False	False	False	False	F

1470 rows × 39 columns

```
In [117]:
          pd.isnull(df).sum()
Out[117]: Attrition
                                          0
          Business Travel
                                          0
          CF age band
                                          0
          CF_attrition label
                                          0
          Department
                                          0
          Education Field
                                          0
          emp no
                                          0
          Employee Number
                                          0
          Gender
                                          0
          Job Role
                                          0
          Marital Status
                                          0
          Over Time
                                          0
          0ver18
                                          0
          Training Times Last Year
                                          0
          Age
                                          0
          CF_current Employee
                                          0
          Daily Rate
                                          0
          Distance From Home
                                          0
          Education
                                          0
          Employee Count
                                          0
          Environment Satisfaction
                                          0
          Hourly Rate
                                          0
          Job Involvement
                                          0
          Job Level
                                          0
          Job Satisfaction
                                          0
          Monthly Income
                                          0
          Monthly Rate
                                          0
          Num Companies Worked
                                          0
          Percent Salary Hike
                                          0
          Performance Rating
                                          0
          Relationship Satisfaction
          Standard Hours
                                          0
          Stock Option Level
                                          0
          Total Working Years
                                          0
          Work Life Balance
                                          0
          Years At Company
                                          0
          Years In Current Role
                                          0
          Years Since Last Promotion
                                          0
          Years With Curr Manager
                                          0
          dtype: int64
```

Shows no null value present in the dataframe

In [118]: # Use the describe() method to get summary statistics df.describe()

Out[118]:

	Employee Number	Training Times Last Year	Age	CF_current Employee	Daily Rate	Distance From Home	Emplo Co
count	1470.000000	1470.000000	1470.000000	1470.000000	1470.000000	1470.000000	147
mean	1024.865306	2.799320	36.923810	0.838776	802.485714	9.192517	
std	602.024335	1.289271	9.135373	0.367863	403.509100	8.106864	
min	1.000000	0.000000	18.000000	0.000000	102.000000	1.000000	
25%	491.250000	2.000000	30.000000	1.000000	465.000000	2.000000	
50%	1020.500000	3.000000	36.000000	1.000000	802.000000	7.000000	
75%	1555.750000	3.000000	43.000000	1.000000	1157.000000	14.000000	
max	2068.000000	6.000000	60.000000	1.000000	1499.000000	29.000000	
4							

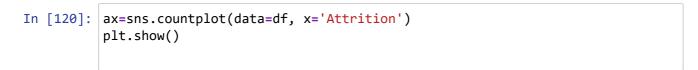
In [119]: df.describe(include='object')

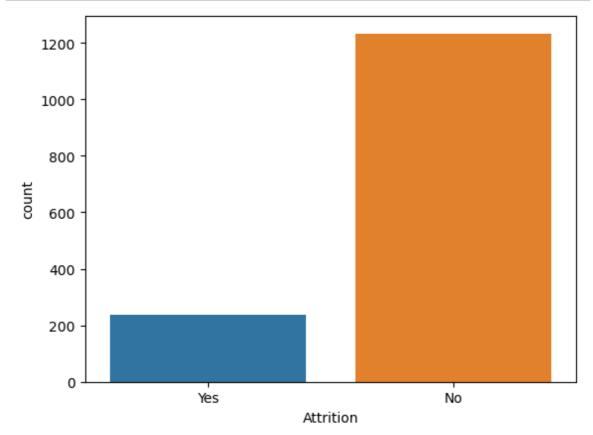
Out[119]:

	Attrition	Business Travel	CF_age band	CF_attrition label	Department	Education Field	emp no	Gender
count	1470	1470	1470	1470	1470	1470	1470	1470
unique	2	3	5	2	3	6	1470	2
top	No	Travel_Rarely	25 - 34	Current Employees	R&D	Life Sciences	STAFF- 1	Male
freq	1233	1043	554	1233	961	606	1	882
4								

Data Analysis and Visulaization

ANALYSIS ON CATEGORICAL COLUMN WRT TARGET COLUMN(ATTRITION)

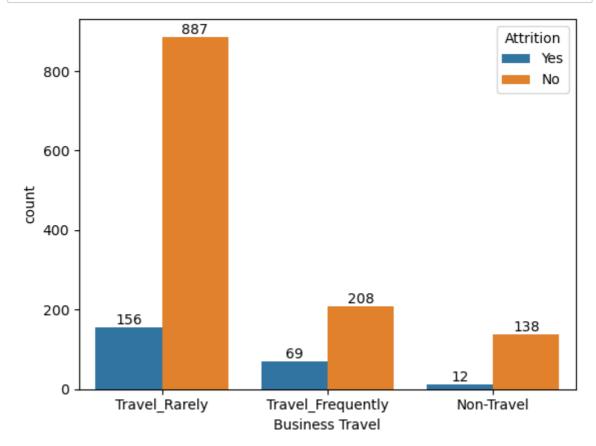




Observation:- Data of attrition say that it has more number of "No" values and less number of "Yes"

1. IMPACT OF BUSINESS TRAVEL ON ATTRITION

In [121]: ax=sns.countplot(data=df, x='Business Travel', hue='Attrition')
for bars in ax.containers:
 ax.bar_label(bars)
plt.show()



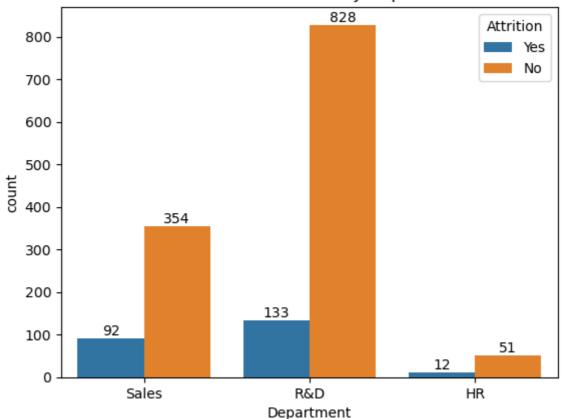
Observation:-

- -Graph tells us that company has more count or more number of employes whos travels rarely it means travel rate of company is less
- -There are more employees which travel rarely and are not satisfied with their job.
- -Non traveller have least count as well as least attrition.

2. IMPACT OF DEPARTMENT ON ATTRITION

```
In [54]: ax2=sns.countplot(data=df, x='Department', hue='Attrition')
    plt.title('Distribution of Attrition by Department ')
    for bars in ax2.containers:
        ax2.bar_label(bars)
    plt.show()
```

Distribution of Attrition by Department

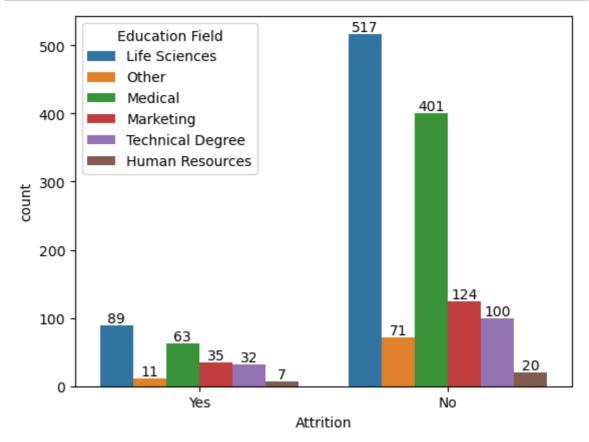


observation:-

- Research and Development (R&D) department have more number attrition (133 employees) as compared to other 2 department
- After R & D department second highest attrition(92 employees) sales department.
- HR department have least attrition with just 12 employees

3. IMPACT OF EDUCATION FIELD ON ATTRITION

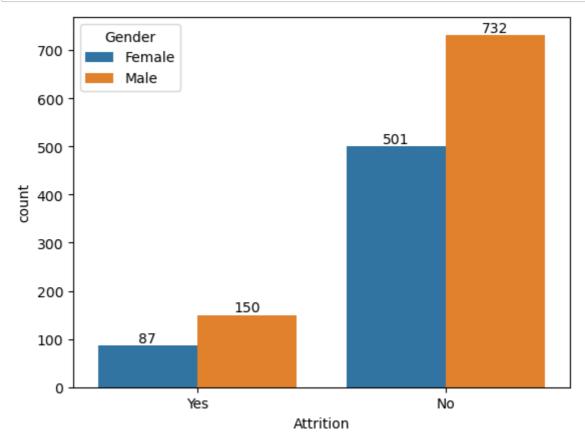
In [73]: ax3=sns.countplot(data=df, x='Attrition', hue='Education Field')
for bars in ax3.containers:
 ax3.bar_label(bars)
plt.show()



Observation:-

- In this graph shows the life science and medical education fileds are more employees as compared to other
- Nearly 89 number of employees are there who are from life science education background will leave the company anf followed by medical education employees
- As we conclude from analysis of department and attrition, here also HR education background employees have leat attrition

4. IMPACT OF GENDER ON ATTRITION

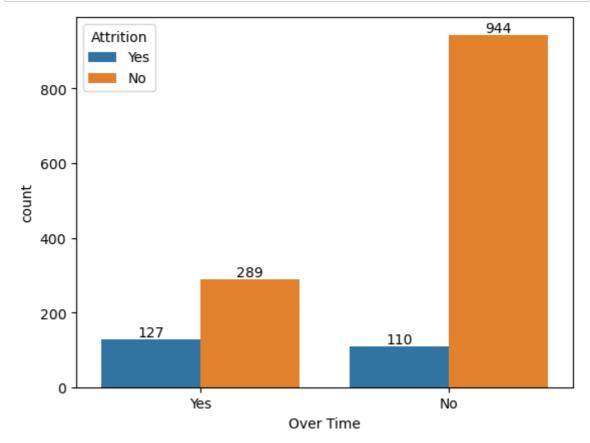


observation:-

- Male employee are more as compared to female
- Males are more likely to quit the job rather than females

5. IMPACT OF OVERTIME ON ATTRITION

```
In [72]: ax5=sns.countplot(data=df,x='Over Time', hue='Attrition')
for bars in ax5.containers:
    ax5.bar_label(bars)
plt.show()
```

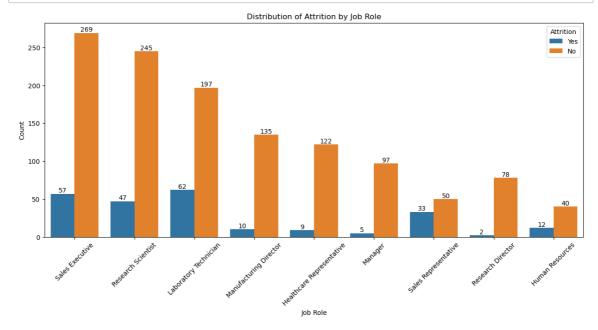


observation:-

- As far as 'Attrition yes' there is a minor difference between the employees who are doing overtime and those who are not doing overtime.
- so we can say that the overtime feature does not much affect attrition.
- But we can conclude that most of employees are not doing overtime.

6. IMPACT OF JOB ROLE ON ATTRITION

```
In [82]: plt.figure(figsize=(15,6))
    ax1=sns.countplot(data=df, x='Job Role', hue='Attrition')
    plt.title('Distribution of Attrition by Job Role')
    plt.xlabel('Job Role')
    plt.ylabel('Count')
    plt.xticks(rotation=45)
    plt.legend(title='Attrition')
    for bars in ax1.containers:
        ax1.bar_label(bars)
    plt.show()
```



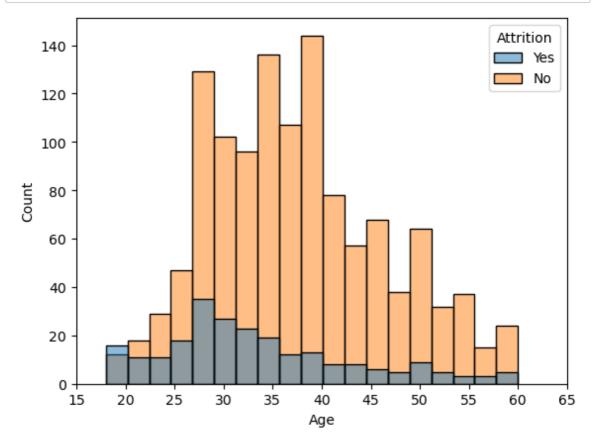
observation:-

- In this graph shows laboratory technician(62 employees) , sales
 executive(57 employees) and research scientist(47 employees) are top 3 job
 roles in which employees have their 'Attrition yes'
- -Apart from this it can also seen that there are more number of employees in sales executive job role
- There are less number of Research director(2 employees) who leave company $% \left(1\right) =\left(1\right) \left(1\right$

ANALYSIS ON CONTINUOS DATA W.R.T TARGET COLUMN

1. IMPACT OF AGE ON ATTRITION

```
In [91]: sns.histplot(data=df, x='Age', hue='Attrition')
    plt.xticks(range(15,70,5))
    plt.show()
```

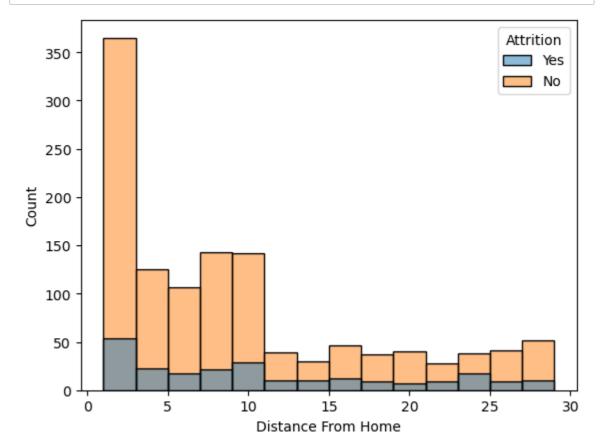


observation:-

- In the histplot shows employess in range of 20 to 35 are more likely to leave the job.
- After 40 age, the distribution tells us that 'Higher the age lesser will be attrition'

2. IMPACT OF DISTANCE FROM HOME AND ATTRITION

In [93]: sns.histplot(data=df, x='Distance From Home', hue='Attrition')
plt.show()



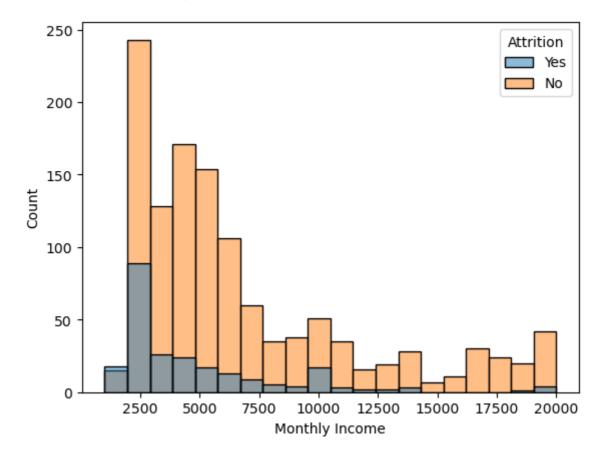
Observation:-

- In this histplot shows 0 to 10 km range more number of employees are working.
- In 0 to 5 km range distance employees are more likely to leave the job

3. HOW MONTHLY INCOME GIVES TRENDS W.R.T ATTRITION

In [94]: sns.histplot(data=df, x='Monthly Income', hue='Attrition')
plt.show

Out[94]: <function matplotlib.pyplot.show(close=None, block=None)>

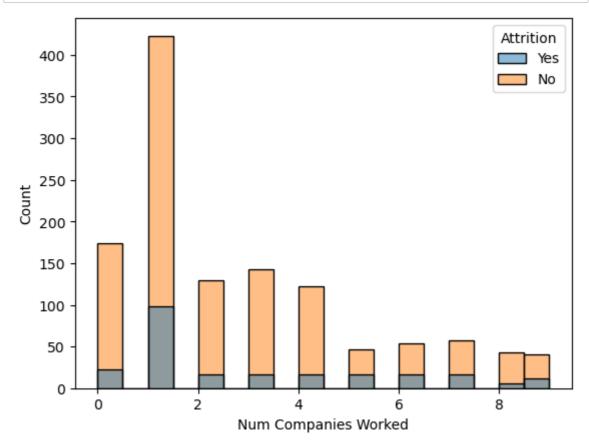


Observation:-

- In this histplot shows employees who have income 2500 are more likely to quit their job, because 2500 is the least range of income .
- That means Lesser the income attrition is High and Higher the monthly income less attrition

4. IMPACT OF NUMBER OF COMPANIES WORKED ON ATTRITION

In [95]: sns.histplot(data=df, x='Num Companies Worked', hue='Attrition')
plt.show()

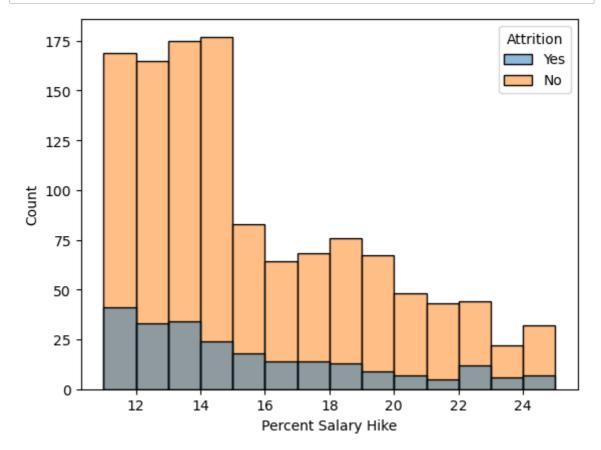


In []: observation:-

- In this hisplot shows number of past company workde 1 are more attrition
- Thats way it is not majorly affect Attrition

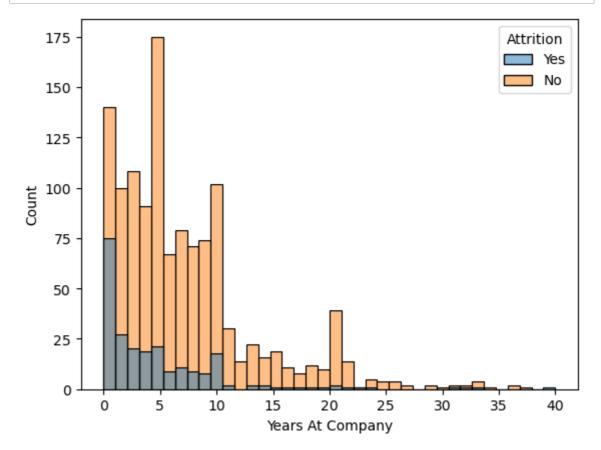
5. IMPACT OF SALARY HIKE ON ATTRITION

In [96]: sns.histplot(data=df, x='Percent Salary Hike', hue='Attrition')
plt.show()



6. IMPACT OF YEARS AT THE COMPANY ON ATTRITION

In [97]: sns.histplot(data=df, x='Years At Company', hue='Attrition')
plt.show()

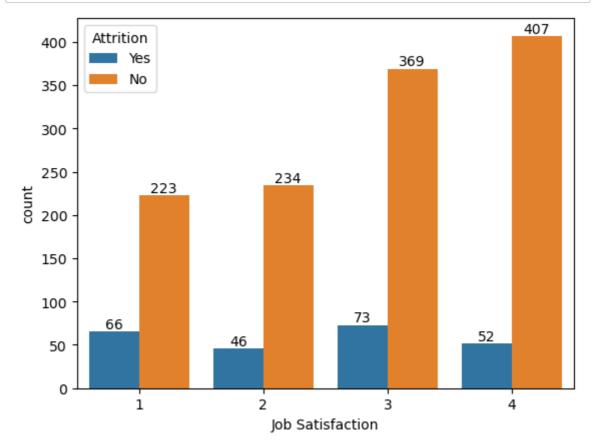


Observation:-

- Among freshers (employees with 0 years of experience), more than half of them have left the company ("attrition yes").
- Employees who have been with the company for 0 to 10 years are also likely to leave their jobs

ANALYSIS ON DISCRETE DATA W.R.T TARGET COLUMN

1. IMPACT OF JOB SATISFACTION ON ATTRITION

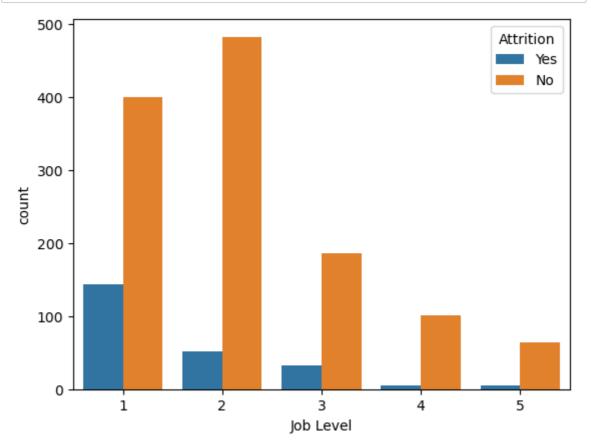


observation:-

- In this graph shows job satisfaction rating 3 and 1 'attration yes' is more $\frac{1}{2}$
- Increase the rate of job satisfaction gives me to increase 'attrition no'

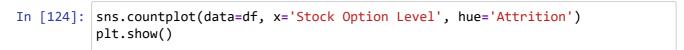
2. IMPACT OF JOB LEVEL ON ATTRITION

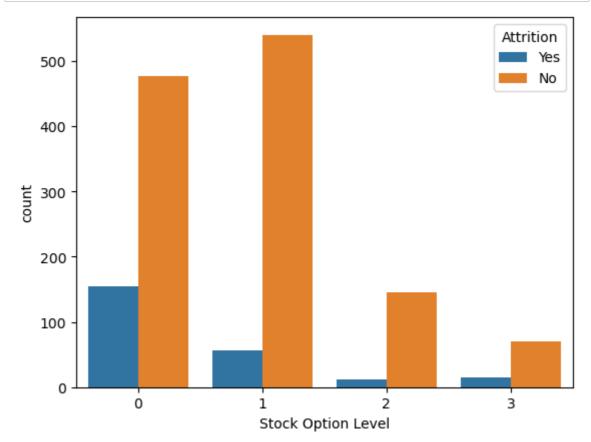
In [123]: sns.countplot(data=df, x='Job Level', hue='Attrition')
plt.show()



observation:- Increase in job level, Decrease in chance of leaving the company for Employees

3. IMPACT OF STOCK OPTION LEVEL ON ATTRITION

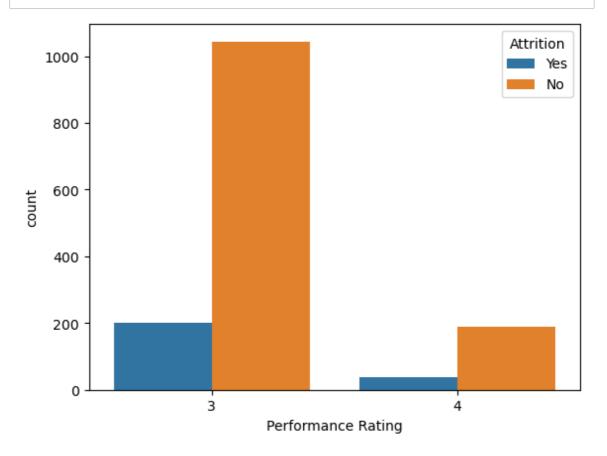




observation:- For the employees who's not having stock option are likely to quit

4. IMPACT OF PERFORMANCE RATING ON ATTRITION

In [125]: sns.countplot(data=df, x='Performance Rating', hue='Attrition')
plt.show()



observation:-

- In this countplot shows performance rating range lies on 3 -4
- However performance rating of employee 3 have more likely to quit or we can say that company wants to fire the employees.

IDENTITY HOTSPOTS

This are top 10 factors which are affecting attrition "Yes" (Employees wants to leave the company)

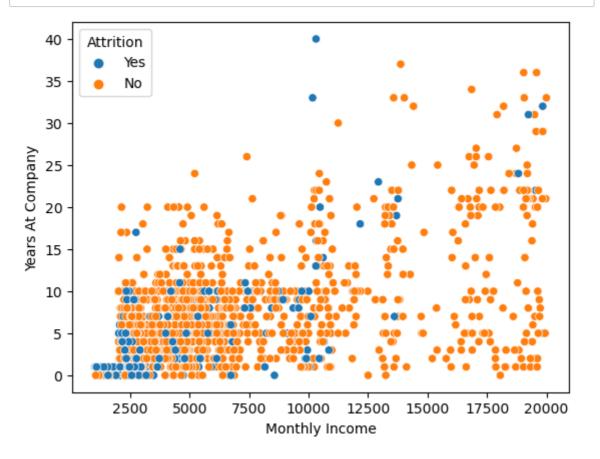
- Impact of years at the company on attrition (Major Factors of affrecting attrition (employees want to leave the company))
- Impact of Monthly Income on attrition (Major Factors of affecting attrition (employees want to leave the company)
- Impact of salary hike on attrition (Major Factors of affecting attrition (employees want to leave the company)
- · Impact of age on attrition
- · Impact of performance rating on attrition
- · Impact of stock level option on attrition
- · Impact of job satisfaction on attrition
- · Impact of job role on attrition
- · Impact of Department on attrition

Impact of Business Travel on attrition

Explore Relationships

Relationship 2 major factors of attrition monthly income and years at company

In [130]: sns.scatterplot(data=df, x='Monthly Income', y='Years At Company', hue='Att
plt.show()



observation:- In this relationship in the scatterplot suggests that fresher whos experience 0-5 yeas and monthly income 2500 are more likely to leave jobs.

Conclusion and Recommendation:

- Invest in Employee Development: Implement comprehensive training and development programs to nurture employees' skills and facilitate career advancement opportunities.
- Enhance Compensation and Benefits: Review and adjust compensation packages to remain competitive in the market and ensure employees feel valued for their contributions.
- Feedback Mechanisms: Establish regular feedback mechanisms to solicit input from fresher employees regarding their experiences, challenges, and suggestions for improvement.
- Career Path Development: Create clear and transparent career paths for fresher employees, outlining opportunities for growth, advancement, and skill development.

• Salary Adjustment: Consider adjusting the salary structure for fresher employees to

	can help improve employee satisfaction and reduce turnover rates among this demographic.
In []:	