

Project 3 : "HR Analytics"

```
# Importing the Dependencies
import pandas as pd
import numpy as np
import seaborn as sns
from matplotlib import pyplot as plt
import warnings
warnings.filterwarnings("ignore")

# Importing the dataset
data=pd.read_csv(r"C:\Users\hp\Desktop\FIRST PROJECT\HR-Employee-Attrition.csv")
pd.set_option('display.max_columns',None)
data.head()
```

	Age	Attrition	BusinessTravel	DailyRate	Department
0	41	Yes	Travel_Rarely	1102	Sales
1	49	No	Travel_Frequently	279	Research & Development
2	37	Yes	Travel_Rarely	1373	Research & Development
3	33	No	Travel_Frequently	1392	Research & Development
4	27	No	Travel_Rarely	591	Research & Development

	DistanceFromHome	Education	EducationField	EmployeeCount
0	1	2	Life Sciences	1
1				
1	8	1	Life Sciences	1
2				
2	2	2	Other	1
4				
3	3	4	Life Sciences	1
5				
4	2	1	Medical	1
7				

	EnvironmentSatisfaction	Gender	HourlyRate	JobInvolvement
0		2 Female	94	3
2				
1		3 Male	61	2
2				

2		4	Male	92	2
1					
3		4	Female	56	3
1					
4		1	Male	40	3
1					
	JobRole	JobSatisfaction	MaritalStatus	MonthlyIncome	
\					
0	Sales Executive	4	Single	5993	
1	Research Scientist	2	Married	5130	
2	Laboratory Technician	3	Single	2090	
3	Research Scientist	3	Married	2909	
4	Laboratory Technician	2	Married	3468	
	MonthlyRate	NumCompaniesWorked	Over18	OverTime	PercentSalaryHike
\					
0	19479	8	Y	Yes	11
1	24907	1	Y	No	23
2	2396	6	Y	Yes	15
3	23159	1	Y	Yes	11
4	16632	9	Y	No	12
	PerformanceRating	RelationshipSatisfaction	StandardHours	\	
0	3		1	80	
1	4		4	80	
2	3		2	80	
3	3		3	80	
4	3		4	80	
	StockOptionLevel	TotalWorkingYears	TrainingTimesLastYear	\	
0	0	8		0	
1	1	10		3	
2	0	7		3	
3	0	8		3	
4	1	6		3	
	WorkLifeBalance	YearsAtCompany	YearsInCurrentRole	\	
0	1	6		4	
1	3	10		7	
2	3	0		0	

3	3	8	7
4	3	2	2

	YearsSinceLastPromotion	YearsWithCurrManager
0	0	5
1	1	7
2	0	0
3	3	0
4	2	2

data.tail()

	Age	Attrition	BusinessTravel	DailyRate	Department
DistanceFromHome \					
1465	36	0	NaN	884	2
23					
1466	39	0	2.0	613	2
6					
1467	27	0	2.0	155	2
4					
1468	49	0	NaN	1023	1
2					
1469	34	0	2.0	628	2
8					

	Education	EducationField	EmployeeCount	EmployeeNumber	\
1465	2	Medical	1	2061	
1466	1	Medical	1	2062	
1467	3	Life Sciences	1	2064	
1468	3	Medical	1	2065	
1469	3	Medical	1	2068	

	EnvironmentSatisfaction	Gender	HourlyRate	JobInvolvement
JobLevel \				
1465	3	Male	41	4
2				
1466	4	Male	42	2
3				
1467	2	Male	87	4
2				
1468	4	Male	63	2
2				
1469	2	Male	82	4
2				

	JobRole	JobSatisfaction	MaritalStatus
MonthlyIncome \			
1465	Laboratory Technician	4	Married
2571			
1466	Healthcare Representative	1	Married

9991			
1467	Manufacturing Director	2	Married
6142			
1468	Sales Executive	2	Married
5390			
1469	Laboratory Technician	3	Married
4404			

	MonthlyRate PercentSalaryHike	NumCompaniesWorked	Over18	OverTime
1465	12290	4	Y	No
17				
1466	21457	4	Y	No
15				
1467	5174	1	Y	Yes
20				
1468	13243	2	Y	No
14				
1469	10228	2	Y	No
12				

	PerformanceRating	RelationshipSatisfaction	StandardHours
1465	3	3	80
1466	3	1	80
1467	4	2	80
1468	3	4	80
1469	3	1	80

	StockOptionLevel	TotalWorkingYears	TrainingTimesLastYear
1465	1	17	3
1466	1	9	5
1467	1	6	0
1468	0	17	3
1469	0	6	3

	WorkLifeBalance	YearsAtCompany	YearsInCurrentRole
1465	3	5	2
1466	3	7	7
1467	3	6	2
1468	2	9	6
1469	4	4	3

	YearsSinceLastPromotion	YearsWithCurrManager
1465	0	3
1466	1	7
1467	0	3
1468	0	8
1469	1	2

data.info()

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 1470 entries, 0 to 1469
```

```
Data columns (total 35 columns):
```

#	Column	Non-Null Count	Dtype
0	Age	1470 non-null	int64
1	Attrition	1470 non-null	object
2	BusinessTravel	1470 non-null	object
3	DailyRate	1470 non-null	int64
4	Department	1470 non-null	object
5	DistanceFromHome	1470 non-null	int64
6	Education	1470 non-null	int64
7	EducationField	1470 non-null	object
8	EmployeeCount	1470 non-null	int64
9	EmployeeNumber	1470 non-null	int64
10	EnvironmentSatisfaction	1470 non-null	int64
11	Gender	1470 non-null	object
12	HourlyRate	1470 non-null	int64
13	JobInvolvement	1470 non-null	int64
14	JobLevel	1470 non-null	int64
15	JobRole	1470 non-null	object
16	JobSatisfaction	1470 non-null	int64
17	MaritalStatus	1470 non-null	object
18	MonthlyIncome	1470 non-null	int64
19	MonthlyRate	1470 non-null	int64
20	NumCompaniesWorked	1470 non-null	int64
21	Over18	1470 non-null	object
22	Overtime	1470 non-null	object
23	PercentSalaryHike	1470 non-null	int64
24	PerformanceRating	1470 non-null	int64
25	RelationshipSatisfaction	1470 non-null	int64
26	StandardHours	1470 non-null	int64
27	StockOptionLevel	1470 non-null	int64
28	TotalWorkingYears	1470 non-null	int64
29	TrainingTimesLastYear	1470 non-null	int64
30	WorkLifeBalance	1470 non-null	int64
31	YearsAtCompany	1470 non-null	int64
32	YearsInCurrentRole	1470 non-null	int64
33	YearsSinceLastPromotion	1470 non-null	int64
34	YearsWithCurrManager	1470 non-null	int64

```
dtypes: int64(26), object(9)
```

```
memory usage: 402.1+ KB
```

```
data.describe()
```

	Age	DailyRate	DistanceFromHome	Education
EmployeeCount \				
count	1470.000000	1470.000000	1470.000000	1470.000000
1470.0				
mean	36.923810	802.485714	9.192517	2.912925

1.0				
std	9.135373	403.509100	8.106864	1.024165
0.0				
min	18.000000	102.000000	1.000000	1.000000
1.0				
25%	30.000000	465.000000	2.000000	2.000000
1.0				
50%	36.000000	802.000000	7.000000	3.000000
1.0				
75%	43.000000	1157.000000	14.000000	4.000000
1.0				
max	60.000000	1499.000000	29.000000	5.000000
1.0				

	EmployeeNumber	EnvironmentSatisfaction	HourlyRate
JobInvolvement \			
count	1470.000000	1470.000000	1470.000000
1470.000000			
mean	1024.865306	2.721769	65.891156
2.729932			
std	602.024335	1.093082	20.329428
0.711561			
min	1.000000	1.000000	30.000000
1.000000			
25%	491.250000	2.000000	48.000000
2.000000			
50%	1020.500000	3.000000	66.000000
3.000000			
75%	1555.750000	4.000000	83.750000
3.000000			
max	2068.000000	4.000000	100.000000
4.000000			

	JobLevel	JobSatisfaction	MonthlyIncome	MonthlyRate \
count	1470.000000	1470.000000	1470.000000	1470.000000
mean	2.063946	2.728571	6502.931293	14313.103401
std	1.106940	1.102846	4707.956783	7117.786044
min	1.000000	1.000000	1009.000000	2094.000000
25%	1.000000	2.000000	2911.000000	8047.000000
50%	2.000000	3.000000	4919.000000	14235.500000
75%	3.000000	4.000000	8379.000000	20461.500000
max	5.000000	4.000000	19999.000000	26999.000000

	NumCompaniesWorked	PercentSalaryHike	PerformanceRating \
count	1470.000000	1470.000000	1470.000000
mean	2.693197	15.209524	3.153741
std	2.498009	3.659938	0.360824
min	0.000000	11.000000	3.000000
25%	1.000000	12.000000	3.000000
50%	2.000000	14.000000	3.000000

75%	4.000000	18.000000	3.000000
max	9.000000	25.000000	4.000000

	RelationshipSatisfaction	StandardHours	StockOptionLevel \
count	1470.000000	1470.0	1470.000000
mean	2.712245	80.0	0.793878
std	1.081209	0.0	0.852077
min	1.000000	80.0	0.000000
25%	2.000000	80.0	0.000000
50%	3.000000	80.0	1.000000
75%	4.000000	80.0	1.000000
max	4.000000	80.0	3.000000

	TotalWorkingYears	TrainingTimesLastYear	WorkLifeBalance \
count	1470.000000	1470.000000	1470.000000
mean	11.279592	2.799320	2.761224
std	7.780782	1.289271	0.706476
min	0.000000	0.000000	1.000000
25%	6.000000	2.000000	2.000000
50%	10.000000	3.000000	3.000000
75%	15.000000	3.000000	3.000000
max	40.000000	6.000000	4.000000

	YearsAtCompany	YearsInCurrentRole	YearsSinceLastPromotion \
count	1470.000000	1470.000000	1470.000000
mean	7.008163	4.229252	2.187755
std	6.126525	3.623137	3.222430
min	0.000000	0.000000	0.000000
25%	3.000000	2.000000	0.000000
50%	5.000000	3.000000	1.000000
75%	9.000000	7.000000	3.000000
max	40.000000	18.000000	15.000000

	YearsWithCurrManager
count	1470.000000
mean	4.123129
std	3.568136
min	0.000000
25%	2.000000
50%	3.000000
75%	7.000000
max	17.000000

data.describe(include="0")

	Attrition	BusinessTravel	Department	EducationField
Gender \				
count	1470	1470	1470	1470
1470				
unique	2	3	3	6

2					
top	No	Travel_Rarely	Research & Development	Life Sciences	
Male					
freq	1233	1043		961	606
882					

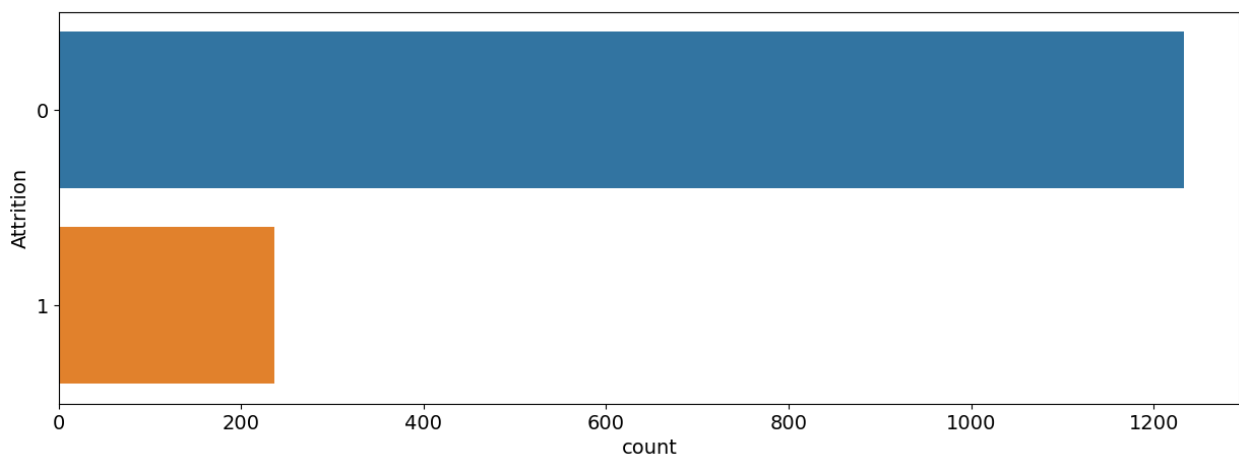
	JobRole	MaritalStatus	Over18	OverTime
count	1470	1470	1470	1470
unique	9	3	1	2
top	Sales Executive	Married	Y	No
freq	326	673	1470	1054

EDA(EXPLORATORY DATA ANALYSIS)

Analysis on categorical columns with respect to the target column(Attrition)

Target column==Attrition

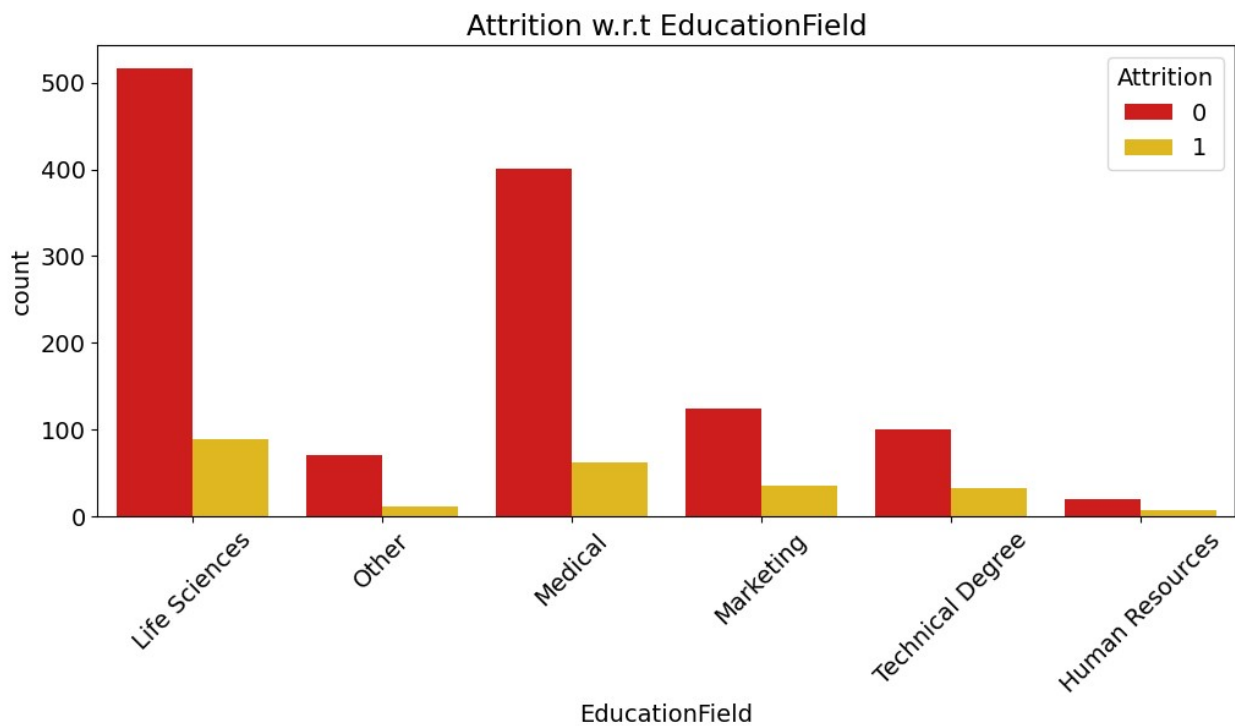
```
plt.figure(figsize=(15,5))
plt.rc("font",size=14)
sns.countplot(y = 'Attrition',data=data)
plt.show()
```



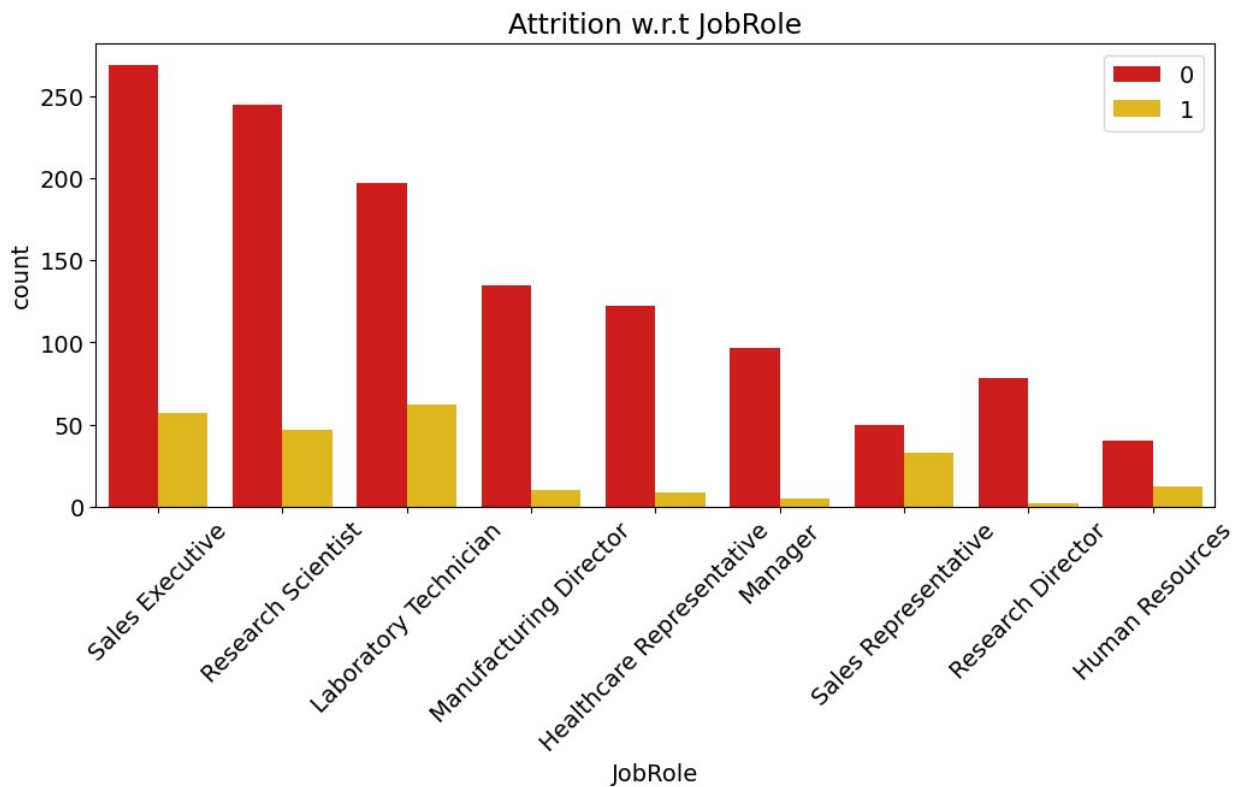
```
#Department w.r.t Attrition
plt.figure(figsize=(12,5))
sns.countplot(x='Department',hue='Attrition',data=data,palette='hot')
plt.title("Attrition w.r.t Department")
plt.show()
```



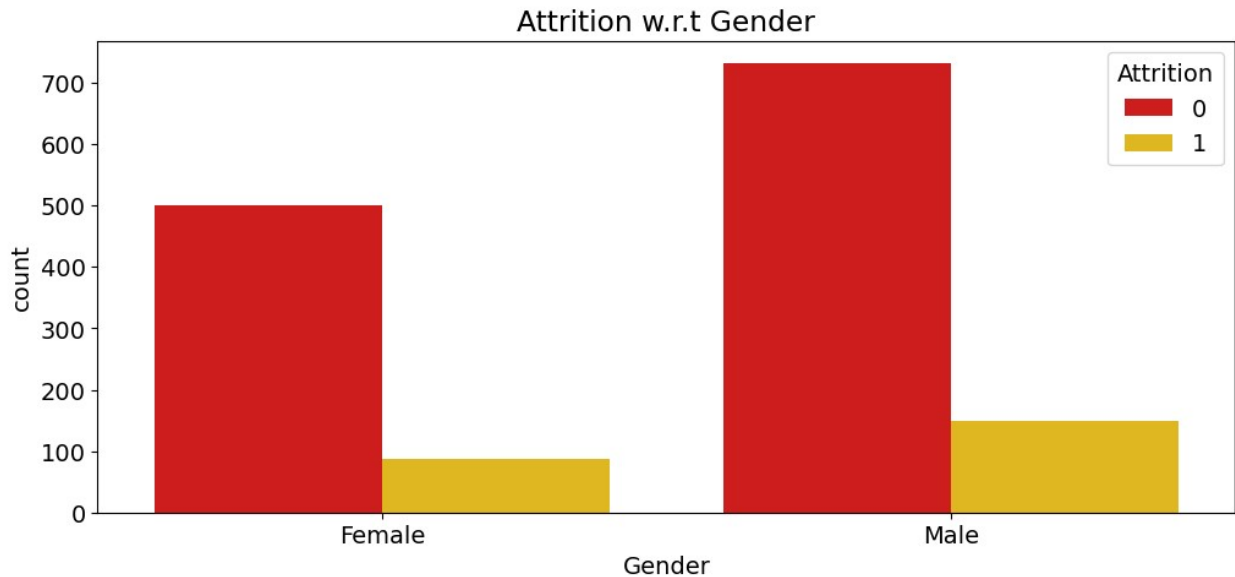

```
#Department wrt Attrition
plt.figure(figsize=(12,5))
sns.countplot(x='EducationField',hue='Attrition',data=data,palette='hot')
plt.title("Attrition w.r.t EducationField")
plt.xticks(rotation=45)
plt.show()
```



```
#Jobrole
plt.figure(figsize=(12,5))
sns.countplot(x='JobRole',hue='Attrition',data=data,palette='hot')
plt.title("Attrition w.r.t JobRole")
plt.legend(loc='best')
plt.xticks(rotation=45)
plt.show()
```



```
#Gender
plt.figure(figsize=(12,5))
sns.countplot(x='Gender',hue='Attrition',data=data,palette='hot')
plt.title("Attrition w.r.t Gender")
plt.xticks()
plt.show()
```



```
data['Gender'].value_counts()
```

```
Male      882
```

```
Female    588
```

```
Name: Gender, dtype: int64
```

```
data['count']=1
```

```
data.groupby(["Gender", "Attrition"]).agg({"count": "sum"})
```

		count
Gender	Attrition	
Female	0	501
	1	87
Male	0	732
	1	150

```
87/(501+87)
```

```
0.14795918367346939
```

As There is 14% Attrition in Female

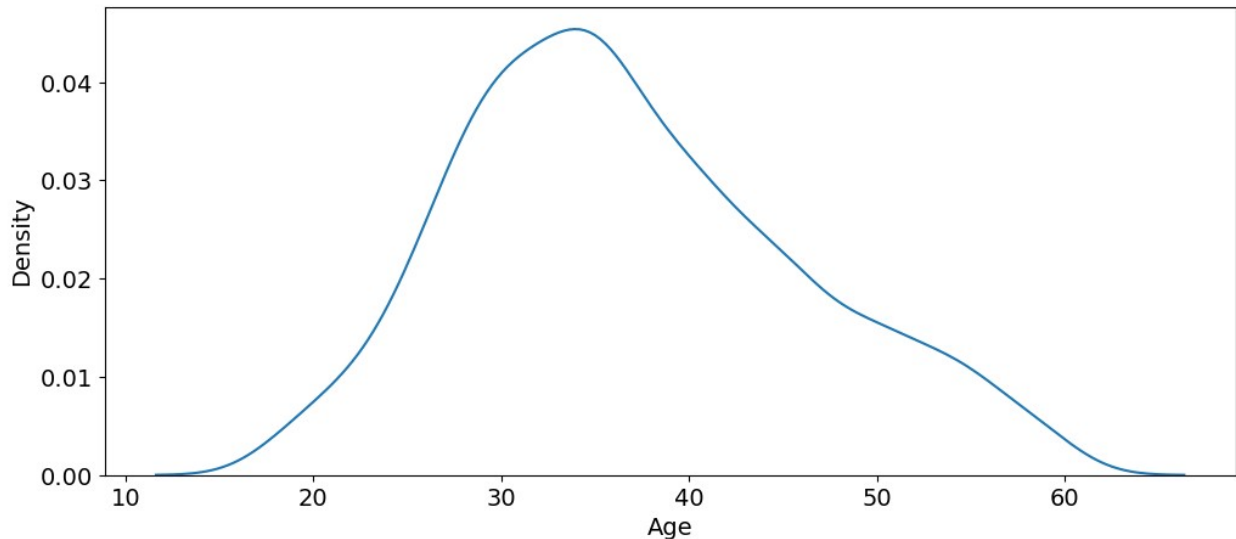
```
150/(732+150)
```

```
0.17006802721088435
```

As there is 17% Attrition in Male

```
#Dropping the column which was created for exter analytics part  
data.drop(columns = ["count"], inplace = True)
```

```
#Distribution of age
plt.figure(figsize=(12,5))
sns.distplot(data['Age'],hist=False)
plt.show()
```



```
#This all factor are in numerical form
#Remove leading and trailing whitespaces in column names and correct
typos
ordinal_features = ['Education',
                    'EmployeeCount',
                    'EnvironmentSatisfaction',
                    'JobInvolvement',
                    'JobLevel',
                    'JobSatisfaction',
                    'NumCompaniesWorked',
                    'PerformanceRating',
                    'RelationshipSatisfaction',
                    'StandardHours',
                    'StockOptionLevel',
                    'TrainingTimesLastYear',
                    'WorkLifeBalance']
#Select the columns from the DataFrame
data[ordinal_features].head()
```

	Education	EmployeeCount	EnvironmentSatisfaction
JobInvolvement \			
0	2	1	2 3
1	1	1	3 2
2	2	1	4 2

3	4	1	4	3
4	1	1	1	3

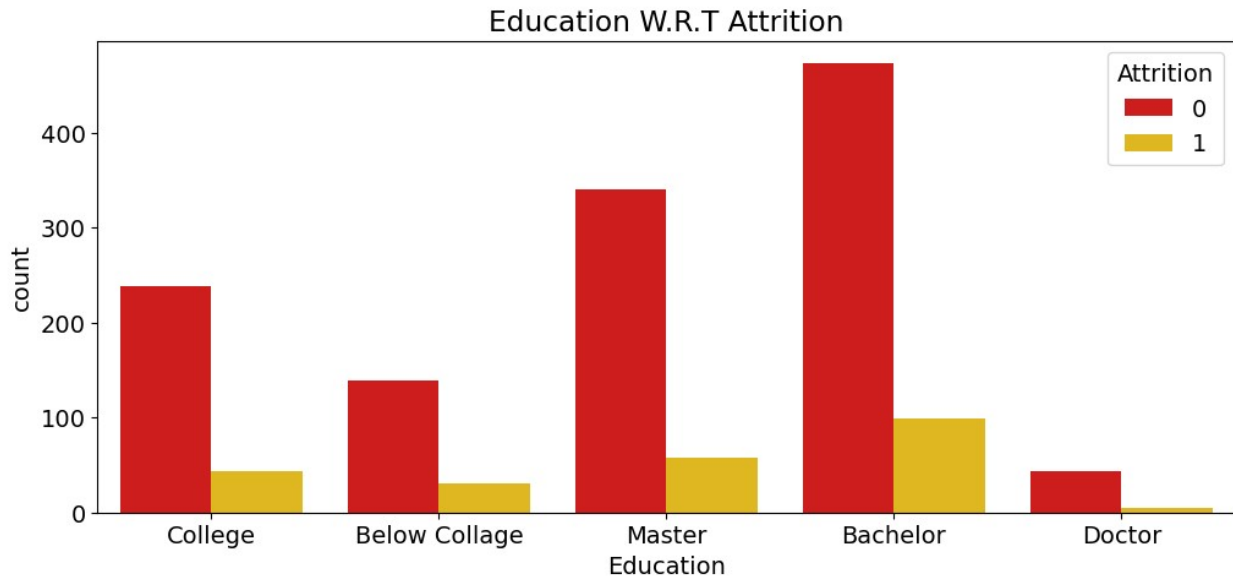
	JobLevel	JobSatisfaction	NumCompaniesWorked	PerformanceRating	\
0	2	4	8	3	
1	2	2	1	4	
2	1	3	6	3	
3	1	3	1	3	
4	1	2	9	3	

	RelationshipSatisfaction	StandardHours	StockOptionLevel	\
0	1	80	0	
1	4	80	1	
2	2	80	0	
3	3	80	0	
4	4	80	1	

	TrainingTimesLastYear	WorkLifeBalance
0	0	1
1	3	3
2	3	3
3	3	3
4	3	3

#This all factor are in non-numerical form

```
edu_map = {1:'Below
Collage',2:'College',3:'Bachelor',4:'Master',5:'Doctor'}
plt.figure(figsize=(12,5))
sns.countplot(x=data['Education'].map(edu_map),hue='Attrition',data=da
ta,palette='hot')
plt.title("Education W.R.T Attrition")
plt.show()
```



```
#List of columns you want to get value counts for
columns_to_count = ['BusinessTravel', 'Department', 'EducationField',
                    'JobRole', 'MaritalStatus']
```

```
#Loop through each column and print its value counts
```

```
for column in columns_to_count:
    counts = data[column].value_counts()
    print(f"Value counts for {column}:\{counts}\n")
```

```
Value counts for BusinessTravel:\2.0      1043
0.0      150
Name: BusinessTravel, dtype: int64
```

```
Value counts for Department:\2      961
1      446
0      63
Name: Department, dtype: int64
```

```
Value counts for EducationField:\Life Sciences      606
Medical      464
Marketing      159
Technical Degree      132
Other      82
Human Resources      27
Name: EducationField, dtype: int64
```

```
Value counts for JobRole:\Sales Executive      326
Research Scientist      292
Laboratory Technician      259
Manufacturing Director      145
Healthcare Representative      131
Manager      102
```

```
Sales Representative      83
Research Director        80
Human Resources          52
Name: JobRole, dtype: int64
```

```
Value counts for MaritalStatus:\Married    673
Single      470
Divorced    327
Name: MaritalStatus, dtype: int64
```

#Target Variable (Attrition)

```
data['Attrition'] = data['Attrition'].replace({'No':0,'Yes':1})
```

#Encode Binary Variable

```
data['OverTime']=data['OverTime'].map({'No':0,'Yes':1})
```

#Map 'Male' to 0 and 'Female' to 1

```
data['Gender'] = data['Gender'].map({'Male': 0, 'Female': 1})
```

#Encode Categorical Column which are ordinal, use LabelEncoding

#apple Label Encoder to df_categorical

```
from sklearn.preprocessing import LabelEncoder
```

```
encoding_cols = ['BusinessTravel', 'Department', 'EducationField',
'JobRole', 'MaritalStatus']
```

```
label_encoders = {}
```

```
for column in encoding_cols:
```

```
    label_encoders[column] = LabelEncoder()
```

```
    data[column] = label_encoders[column].fit_transform(data[column])
```

```
data.head(10)
```

	Age	Attrition	BusinessTravel	DailyRate	Department
DistanceFromHome \					
0	41	1	1	1102	1
1					
1	49	0	2	279	2
8					
2	37	1	1	1373	2
2					
3	33	0	2	1392	2
3					
4	27	0	1	591	2
2					
5	32	0	2	1005	2
2					
6	59	0	1	1324	2
3					
7	30	0	1	1358	2

24					
8	38	0	2	216	2
23					
9	36	0	1	1299	2
27					

	Education	EducationField	EmployeeCount	EmployeeNumber	\
0	2	1	1	1	
1	1	1	1	2	
2	2	4	1	4	
3	4	1	1	5	
4	1	3	1	7	
5	2	1	1	8	
6	3	3	1	10	
7	1	1	1	11	
8	3	1	1	12	
9	3	3	1	13	

	EnvironmentSatisfaction	Gender	HourlyRate	JobInvolvement	
JobLevel \					
0	2	1	94	3	
2					
1	3	0	61	2	
2					
2	4	0	92	2	
1					
3	4	1	56	3	
1					
4	1	0	40	3	
1					
5	4	0	79	3	
1					
6	3	1	81	4	
1					
7	4	0	67	3	
1					
8	4	0	44	2	
3					
9	3	0	94	3	
2					

	JobRole	JobSatisfaction	MaritalStatus	MonthlyIncome	MonthlyRate
\					
0	7	4	2	5993	19479
1	6	2	1	5130	24907
2	2	3	2	2090	2396
3	6	3	1	2909	23159

4	2	2	1	3468	16632
5	2	4	2	3068	11864
6	2	1	1	2670	9964
7	2	3	0	2693	13335
8	4	3	2	9526	8787
9	0	3	1	5237	16577
<div> <div>NumCompaniesWorked</div> <div>Over18</div> <div>OverTime</div> <div>PercentSalaryHike</div> <div>PerformanceRating \</div> </div>					
0	8	Y	1	11	
3					
1	1	Y	0	23	
4					
2	6	Y	1	15	
3					
3	1	Y	1	11	
3					
4	9	Y	0	12	
3					
5	0	Y	0	13	
3					
6	4	Y	1	20	
4					
7	1	Y	0	22	
4					
8	0	Y	0	21	
4					
9	6	Y	0	13	
3					
<div> <div>RelationshipSatisfaction</div> <div>StandardHours</div> <div>StockOptionLevel \</div> </div>					
0	1	80	0		
1	4	80	1		
2	2	80	0		
3	3	80	0		
4	4	80	1		
5	3	80	0		
6	1	80	3		
7	2	80	1		
8	2	80	0		
9	2	80	2		
<div> <div>TotalWorkingYears</div> <div>TrainingTimesLastYear</div> <div>WorkLifeBalance</div> </div>					

YearsAtCompany \			
0	8	0	1
6			
1	10	3	3
10			
2	7	3	3
0			
3	8	3	3
8			
4	6	3	3
2			
5	8	2	2
7			
6	12	3	2
1			
7	1	2	3
1			
8	10	2	3
9			
9	17	3	2
7			

	YearsInCurrentRole	YearsSinceLastPromotion	YearsWithCurrManager
0	4	0	5
1	7	1	7
2	0	0	0
3	7	3	0
4	2	2	2
5	7	3	6
6	0	0	0
7	0	0	0
8	7	1	8
9	7	7	7

```
data['Over18']=data['Over18'].map({'Y':1,'No':0})
```

```
data.head(10)
```

	Age	Attrition	BusinessTravel	DailyRate	Department
DistanceFromHome \					
0	41	1	1	1102	1
1					
1	49	0	2	279	2
8					
2	37	1	1	1373	2
2					
3	33	0	2	1392	2
3					
4	27	0	1	591	2
2					

5	32	0	2	1005	2
2					
6	59	0	1	1324	2
3					
7	30	0	1	1358	2
24					
8	38	0	2	216	2
23					
9	36	0	1	1299	2
27					
Education EducationField EmployeeCount EmployeeNumber \					
0	2	1	1	1	1
1	1	1	1	2	2
2	2	4	1	4	4
3	4	1	1	5	5
4	1	3	1	7	7
5	2	1	1	8	8
6	3	3	1	10	10
7	1	1	1	11	11
8	3	1	1	12	12
9	3	3	1	13	13
EnvironmentSatisfaction Gender HourlyRate JobInvolvement JobLevel \					
0		2	1	94	3
2					
1		3	0	61	2
2					
2		4	0	92	2
1					
3		4	1	56	3
1					
4		1	0	40	3
1					
5		4	0	79	3
1					
6		3	1	81	4
1					
7		4	0	67	3
1					
8		4	0	44	2
3					
9		3	0	94	3
2					
JobRole JobSatisfaction MaritalStatus MonthlyIncome MonthlyRate \					
0	7	4	2	5993	19479

1	6	2	1	5130	24907
2	2	3	2	2090	2396
3	6	3	1	2909	23159
4	2	2	1	3468	16632
5	2	4	2	3068	11864
6	2	1	1	2670	9964
7	2	3	0	2693	13335
8	4	3	2	9526	8787
9	0	3	1	5237	16577
NumCompaniesWorked		Over18	OverTime	PercentSalaryHike	
PerformanceRating	\				
0	8	1	1	11	
3					
1	1	1	0	23	
4					
2	6	1	1	15	
3					
3	1	1	1	11	
3					
4	9	1	0	12	
3					
5	0	1	0	13	
3					
6	4	1	1	20	
4					
7	1	1	0	22	
4					
8	0	1	0	21	
4					
9	6	1	0	13	
3					
RelationshipSatisfaction		StandardHours	StockOptionLevel		
			\		
0		1	80	0	
1		4	80	1	
2		2	80	0	
3		3	80	0	
4		4	80	1	
5		3	80	0	
6		1	80	3	

7	2	80	1
8	2	80	0
9	2	80	2

	TotalWorkingYears	TrainingTimesLastYear	WorkLifeBalance
YearsAtCompany \			
0	8	0	1
6			
1	10	3	3
10			
2	7	3	3
0			
3	8	3	3
8			
4	6	3	3
2			
5	8	2	2
7			
6	12	3	2
1			
7	1	2	3
1			
8	10	2	3
9			
9	17	3	2
7			

	YearsInCurrentRole	YearsSinceLastPromotion	YearsWithCurrManager
0	4	0	5
1	7	1	7
2	0	0	0
3	7	3	0
4	2	2	2
5	7	3	6
6	0	0	0
7	0	0	0
8	7	1	8
9	7	7	7

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Int64Index: 1470 entries, 0 to 1469
```

```
Data columns (total 35 columns):
```

#	Column	Non-Null Count	Dtype
---	-----	-----	-----
0	Age	1470 non-null	int64
1	Attrition	1470 non-null	int64
2	BusinessTravel	1470 non-null	int64
3	DailyRate	1470 non-null	int64

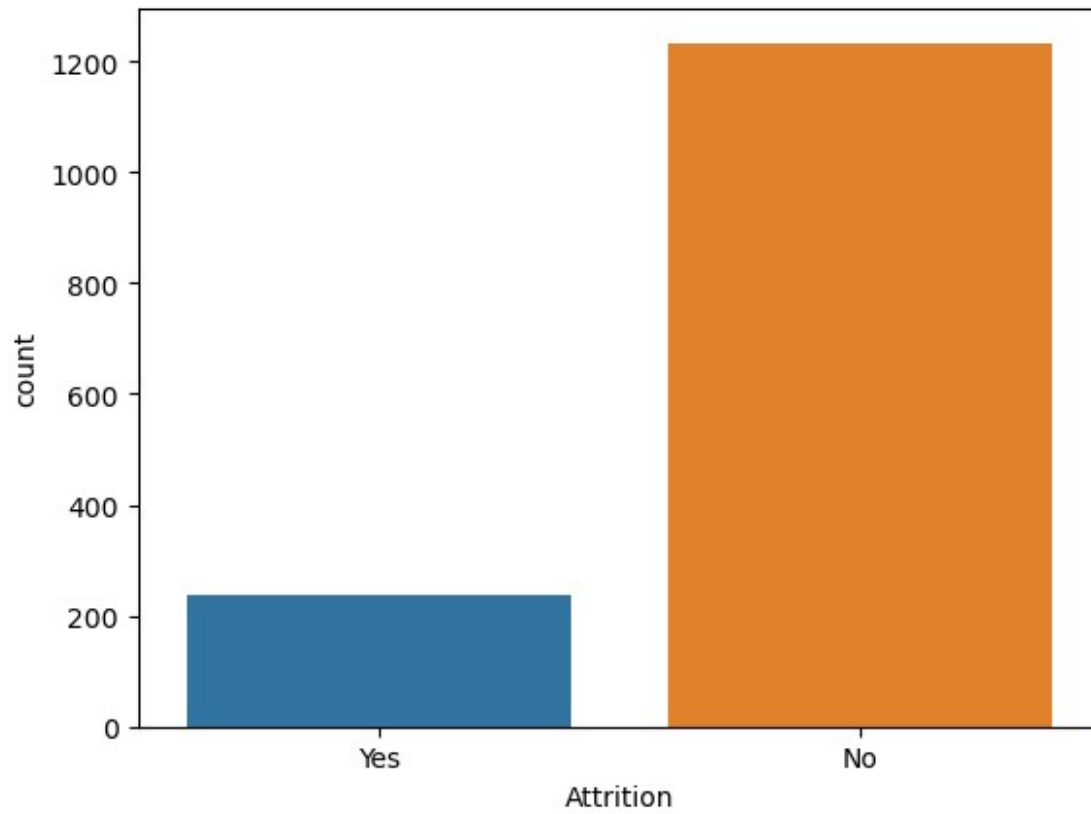
4	Department	1470	non-null	int64
5	DistanceFromHome	1470	non-null	int64
6	Education	1470	non-null	int64
7	EducationField	1470	non-null	int32
8	EmployeeCount	1470	non-null	int64
9	EmployeeNumber	1470	non-null	int64
10	EnvironmentSatisfaction	1470	non-null	int64
11	Gender	1470	non-null	int64
12	HourlyRate	1470	non-null	int64
13	JobInvolvement	1470	non-null	int64
14	JobLevel	1470	non-null	int64
15	JobRole	1470	non-null	int32
16	JobSatisfaction	1470	non-null	int64
17	MaritalStatus	1470	non-null	int32
18	MonthlyIncome	1470	non-null	int64
19	MonthlyRate	1470	non-null	int64
20	NumCompaniesWorked	1470	non-null	int64
21	Over18	1470	non-null	int64
22	Overtime	1470	non-null	int64
23	PercentSalaryHike	1470	non-null	int64
24	PerformanceRating	1470	non-null	int64
25	RelationshipSatisfaction	1470	non-null	int64
26	StandardHours	1470	non-null	int64
27	StockOptionLevel	1470	non-null	int64
28	TotalWorkingYears	1470	non-null	int64
29	TrainingTimesLastYear	1470	non-null	int64
30	WorkLifeBalance	1470	non-null	int64
31	YearsAtCompany	1470	non-null	int64
32	YearsInCurrentRole	1470	non-null	int64
33	YearsSinceLastPromotion	1470	non-null	int64
34	YearsWithCurrManager	1470	non-null	int64

dtypes: int32(3), int64(32)

memory usage: 396.2 KB

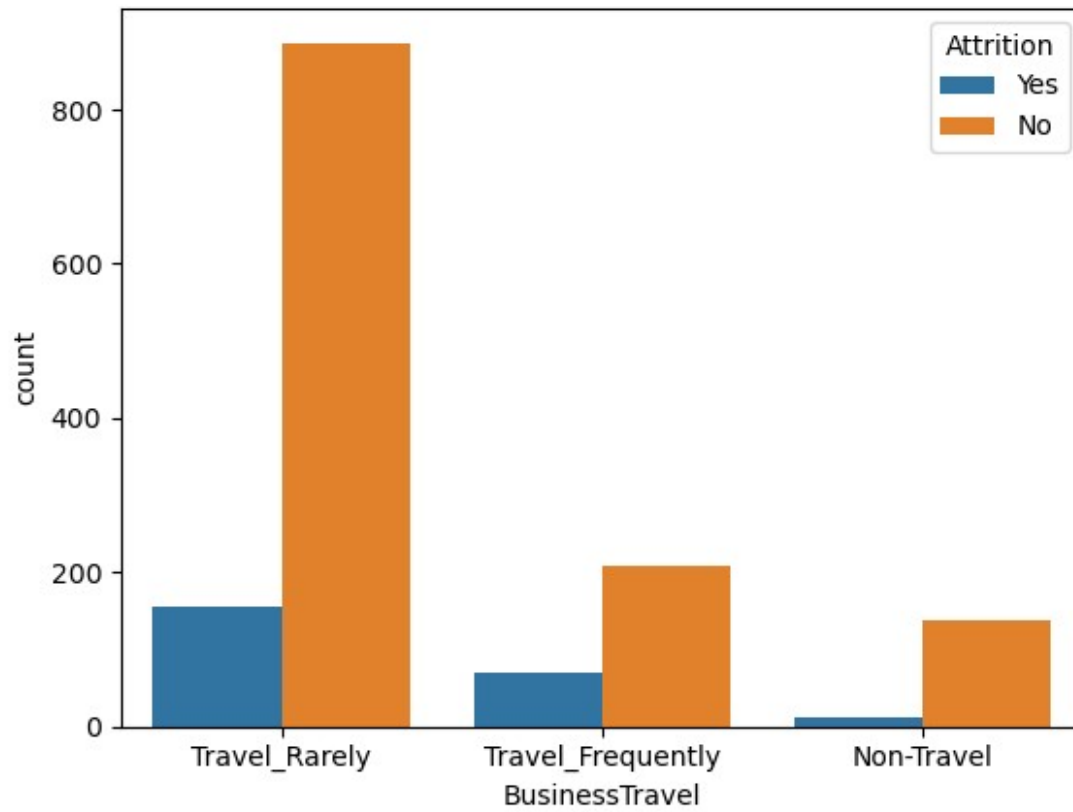
```
x = data.drop(['Attrition', 'Over18'], axis=1) #Input Features
y = data['Attrition'].values #Output Features
```

```
sns.countplot(x=data.Attrition)
plt.show()
```



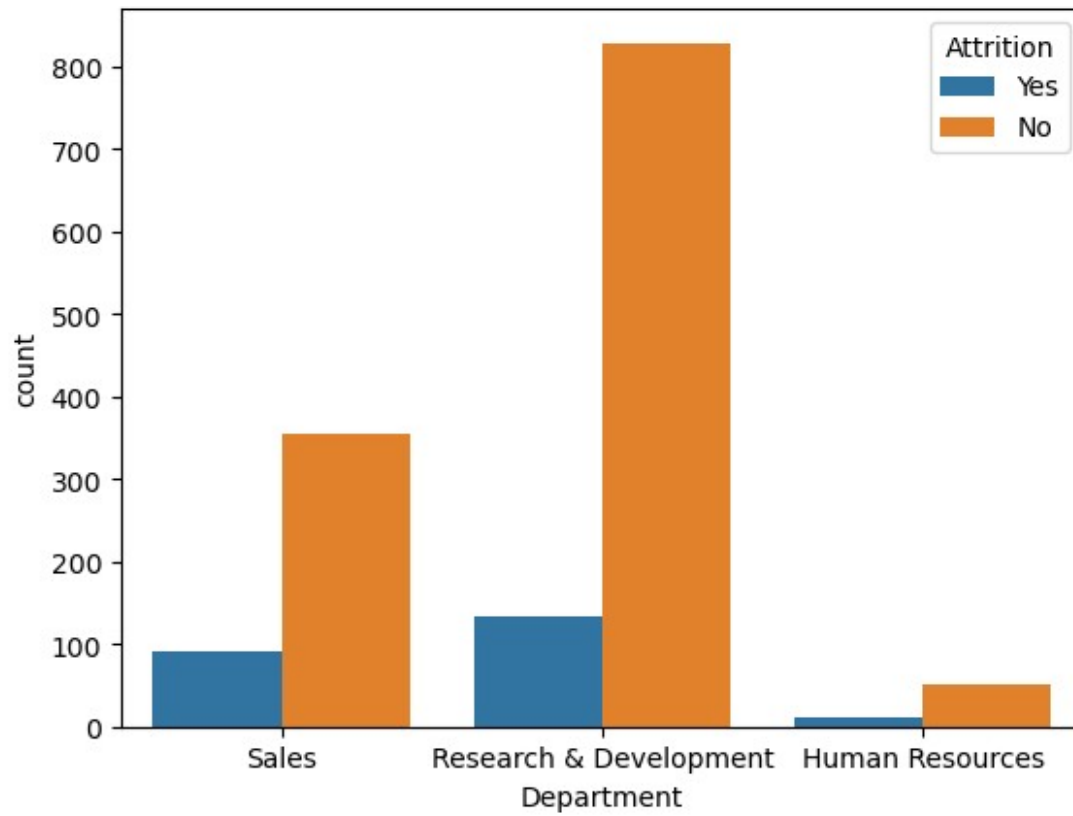
1. Impact Of Business Travel On Attrition

```
sns.countplot(hue=data.Attrition,x=data.BusinessTravel)  
plt.show()
```



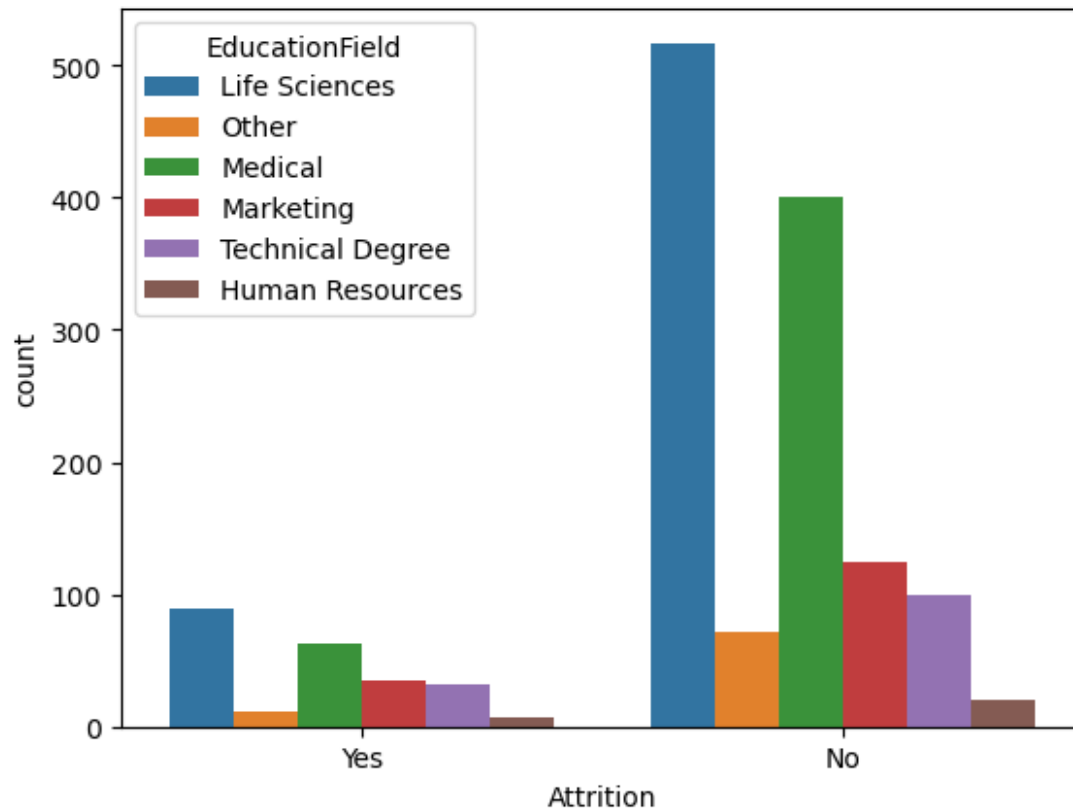
1. Impact Of Department On Attrition

```
sns.countplot(hue=data.Attrition,x=data.Department)  
plt.show()
```

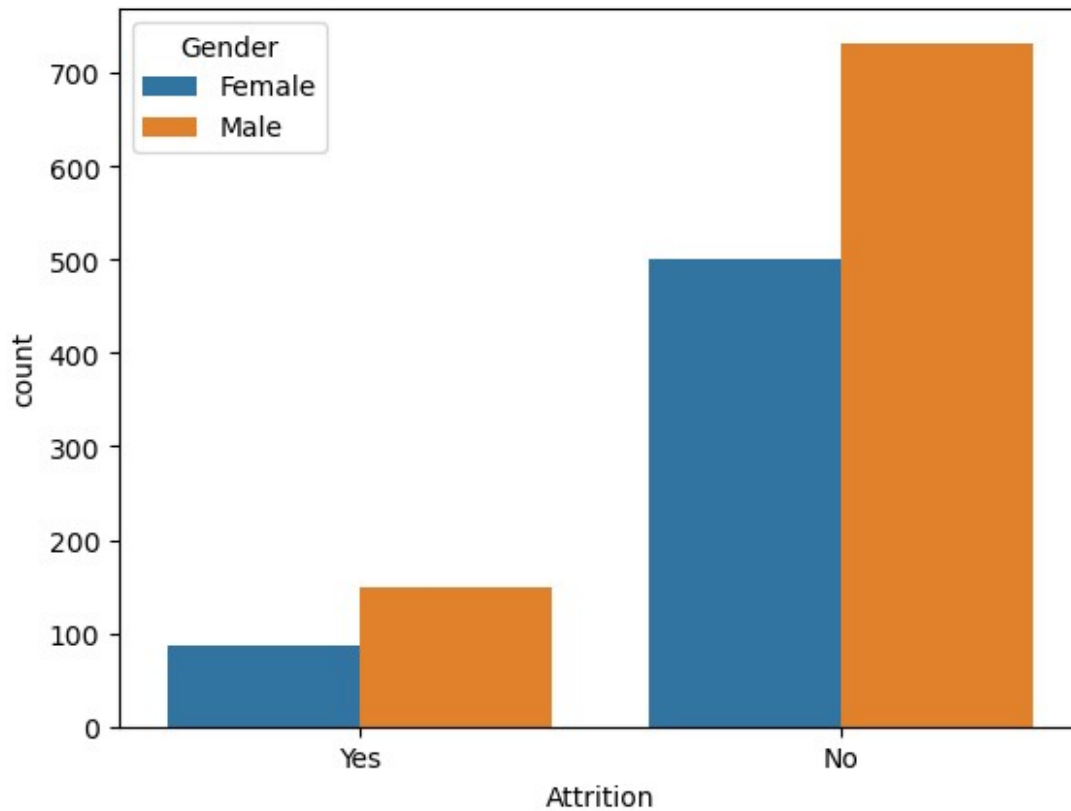
1. Impact Of Education Field On Attrition

```
sns.countplot(x=data.Attrition,hue=data.EducationField)  
plt.show()
```



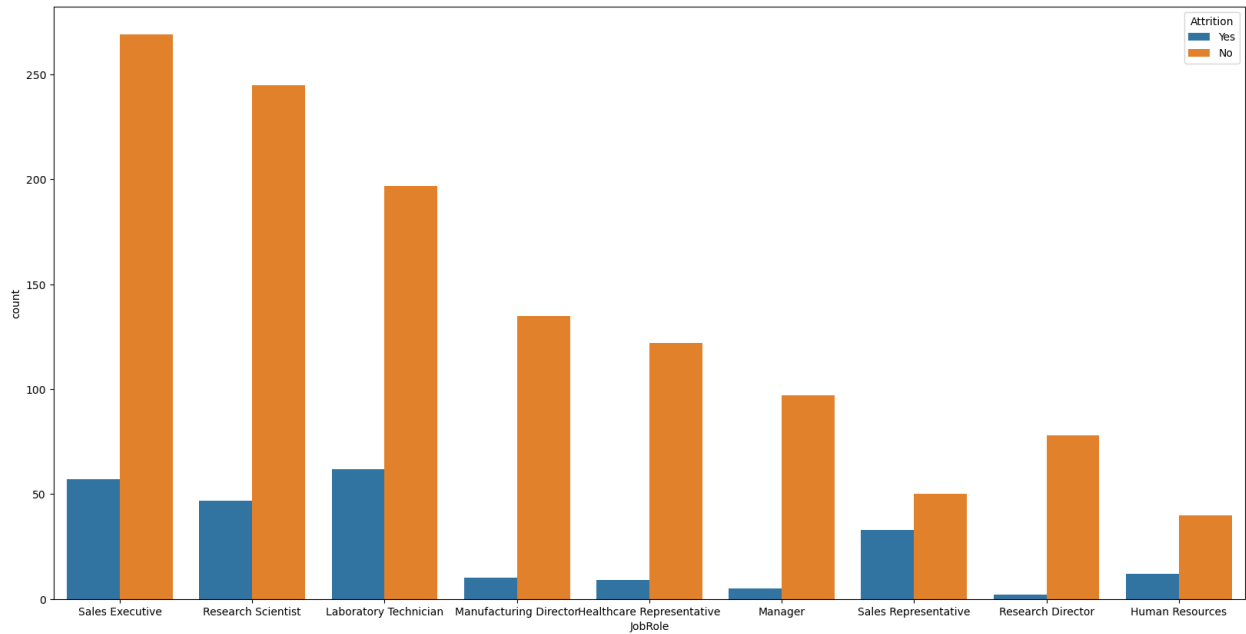
1. Gender And Attrition

```
sns.countplot(x=data.Attrition,hue=data.Gender)  
plt.show()
```



1. Impact Of Job Role On Attrition

```
plt.figure(figsize=(20,10), facecolor='white')
sns.countplot(x='JobRole',hue='Attrition',data=data)
plt.xlabel('JobRole',fontSize=10)
Text(0.5, 0, 'JobRole')
```



Analysis On Continuos Data With Respect To Target Column

```
numerical_col = []
for column in data.columns:
    if data[column].dtype == "int64" and len(data[column].unique())>=
10:

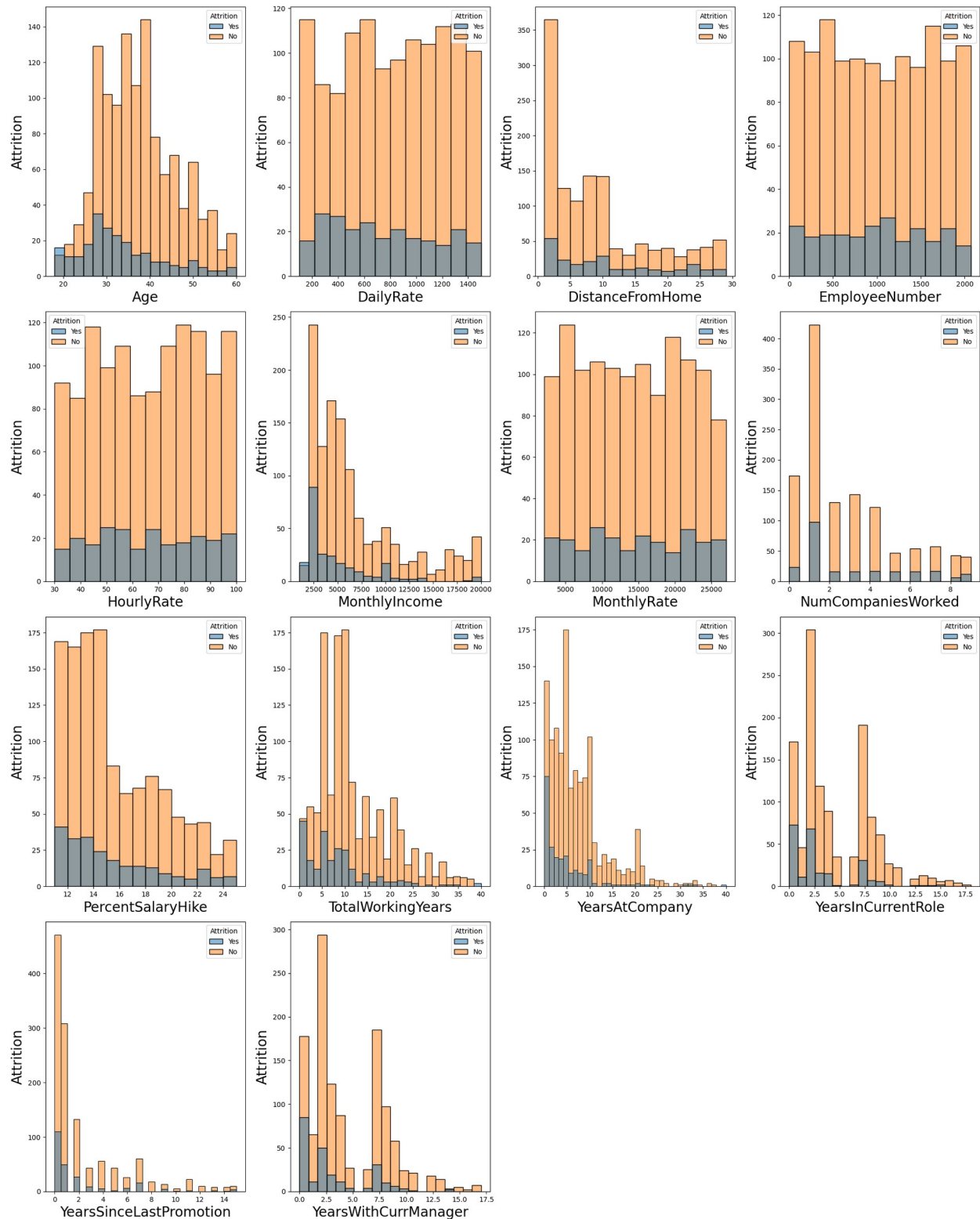
        numerical_col.append(column)
numerical_col
['Age',
'DailyRate',
'DistanceFromHome',
'EmployeeNumber',
'HourlyRate',
'MonthlyIncome',
'MonthlyRate',
'NumCompaniesWorked',
'PercentSalaryHike',
'TotalWorkingYears',
'YearsAtCompany',
'YearsInCurrentRole',
'YearsSinceLastPromotion',
'YearsWithCurrManager']
```

Graphical Representation Of Continuous Data

```
data2=data[['Age',  
            'DailyRate',  
            'DistanceFromHome',  
            'EmployeeNumber',  
            'HourlyRate',  
            'MonthlyIncome',  
            'MonthlyRate',  
            'NumCompaniesWorked',  
            'PercentSalaryHike',  
            'TotalWorkingYears',  
            'YearsAtCompany',  
            'YearsInCurrentRole',  
            'YearsSinceLastPromotion',  
            'YearsWithCurrManager']]
```

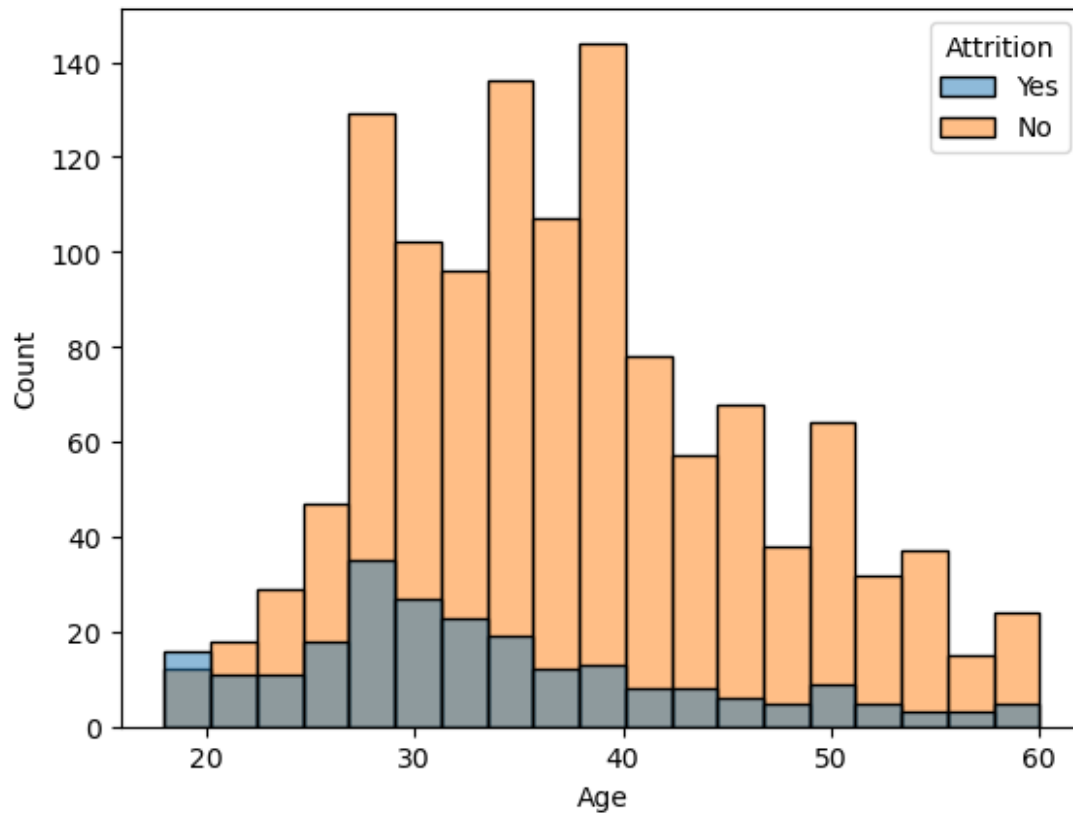
Data Visualization

```
plt.figure(figsize=(20,25), facecolor='white')  
plotnumber = 1  
  
for column in data2:  
    if plotnumber<=16 :  
        ax = plt.subplot(4,4,plotnumber)  
        sns.histplot(x=data2[column].dropna(axis=0)  
                    ,hue=data.Attrition)  
        plt.xlabel(column,fontsize=20)  
        plt.ylabel('Attrition',fontsize=20)  
        plotnumber+=1  
plt.tight_layout()
```



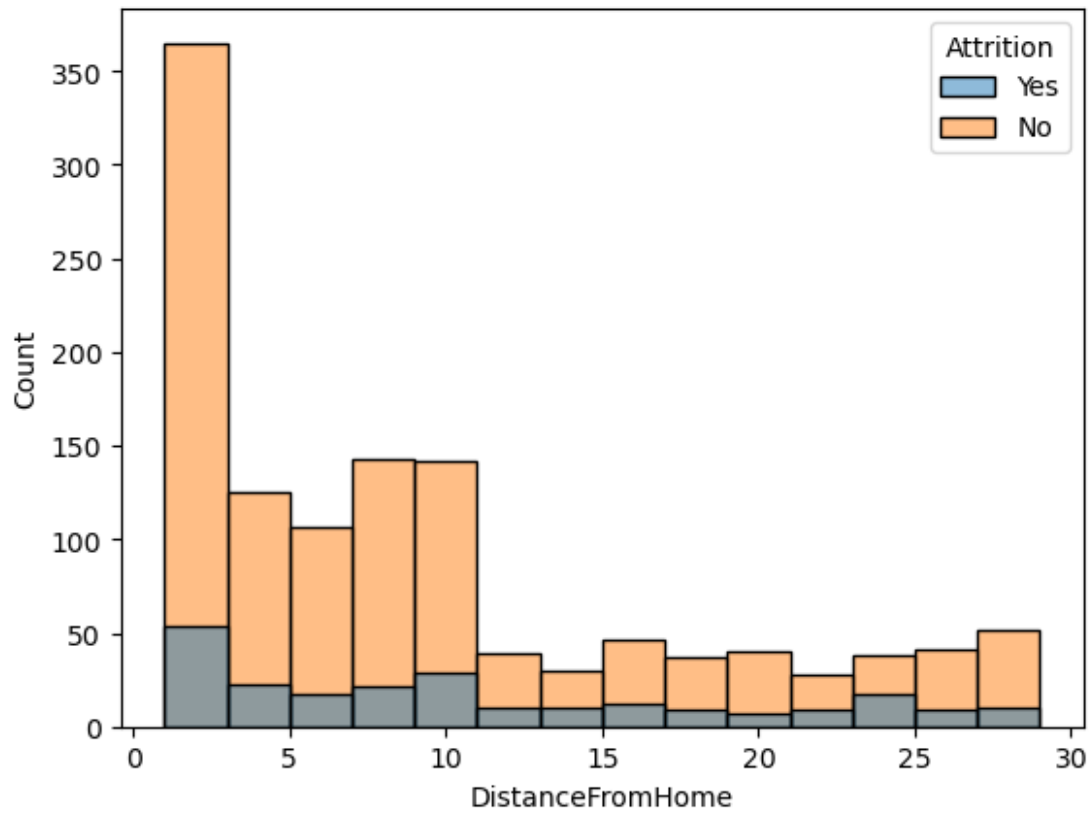
1. Impact Of Age On Attrition

```
sns.histplot(hue=data.Attrition,x=data.Age)
plt.show()
```



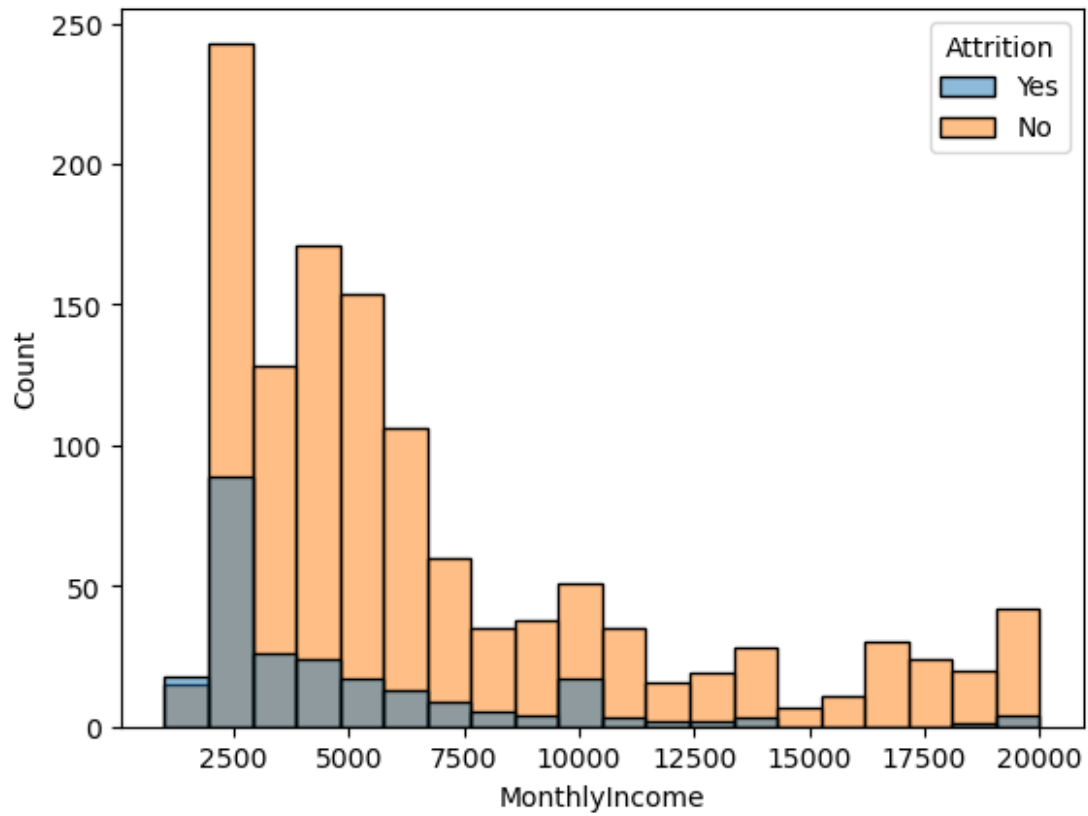
1. Distance From Home And Attrition

```
sns.histplot(hue=data.Attrition,x=data.DistanceFromHome)  
plt.show()
```



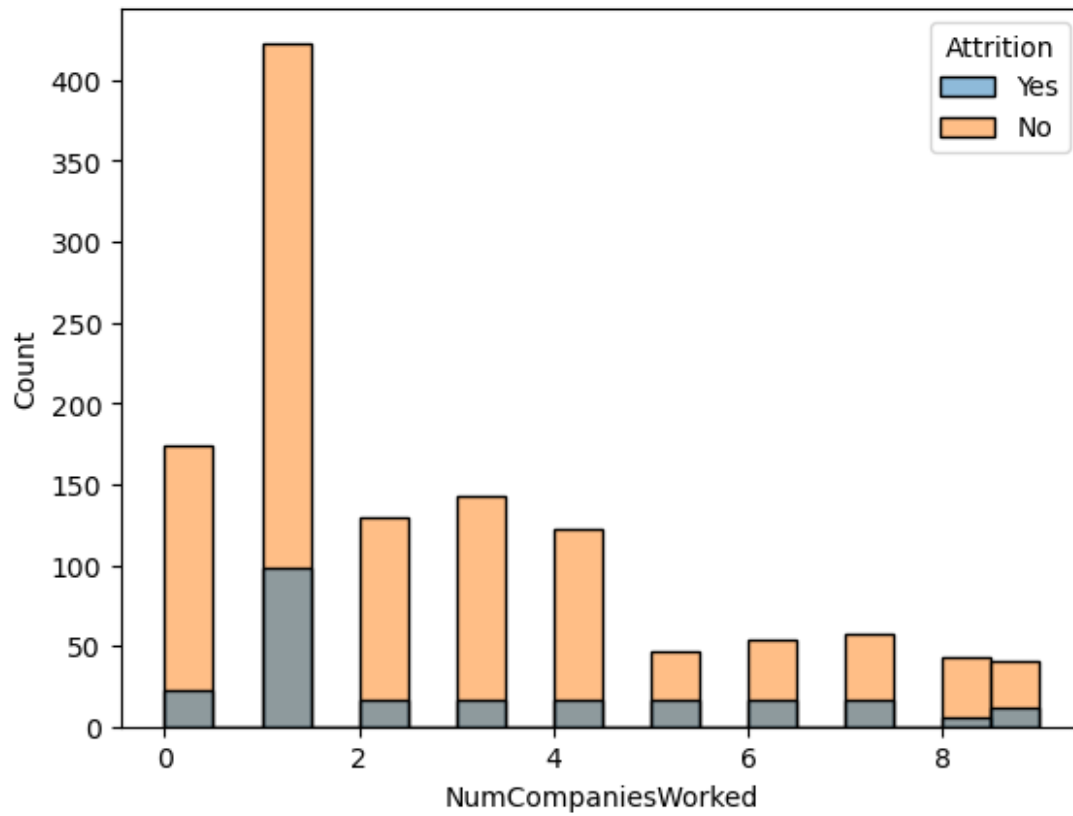
1. How Monthly Income Gives Trends w.r.t Attrition

```
sns.histplot(x=data.MonthlyIncome,hue=data.Attrition)
plt.show()
```

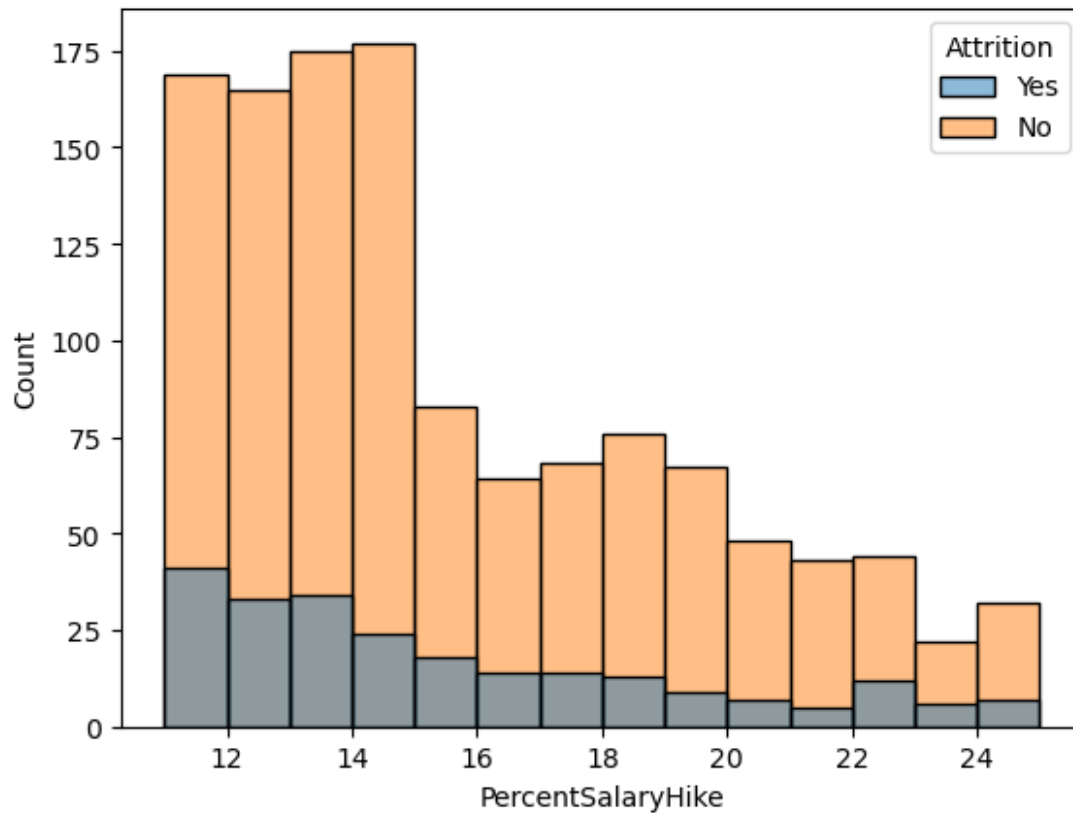
1. Impact Of No. Of Companies Worked

```
sns.histplot(hue=data.Attrition,x=data.NumCompaniesWorked)  
plt.show()
```



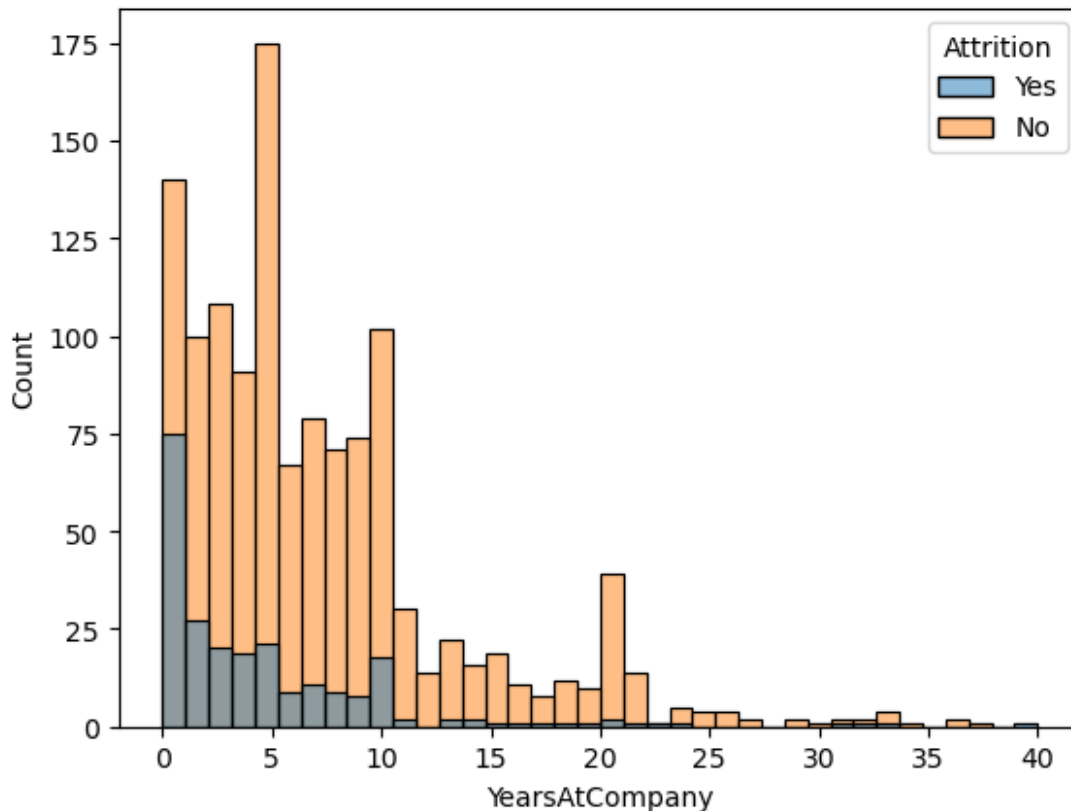
1. How Salary Hike Is Impacting The Attrition

```
sns.histplot(hue=data.Attrition,x=data.PercentSalaryHike)  
plt.show()
```



1. Years At The Company

```
sns.histplot(hue=data.Attrition,x=data.YearsAtCompany)  
plt.show()
```



Analysis Of Discrete Data W.R.T Target Column

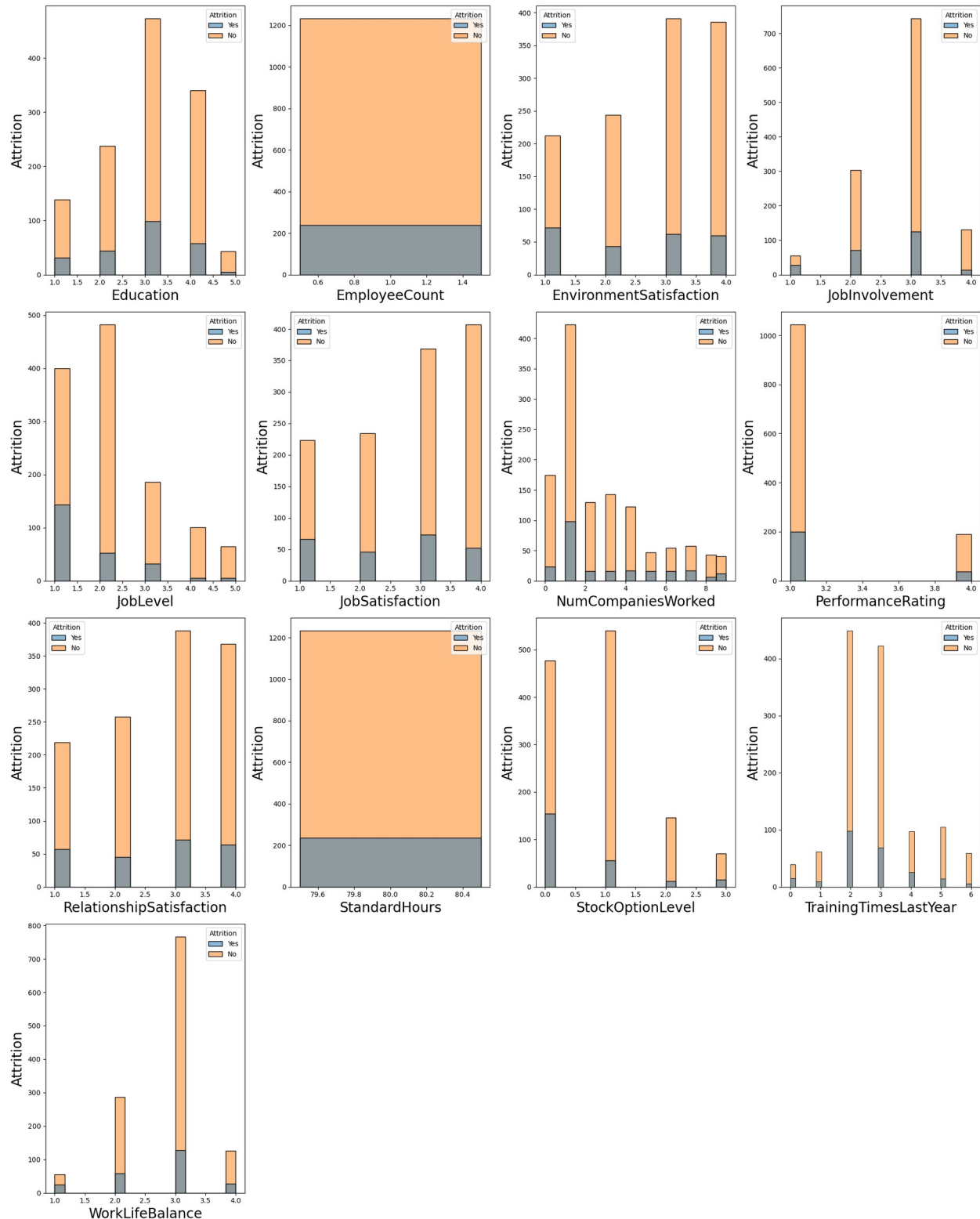
```
discrete_col = []
for column in data.columns:
    if data[column].dtype == "int64" and len(data[column].unique()) <=
10:
        discrete_col.append(column)

data3=data[['Education',
            'EmployeeCount',
            'EnvironmentSatisfaction',
            'JobInvolvement',
            'JobLevel',
            'JobSatisfaction',
            'NumCompaniesWorked',
            'PerformanceRating',
            'RelationshipSatisfaction',
            'StandardHours',
            'StockOptionLevel',
            'TrainingTimesLastYear',
            'WorkLifeBalance']]
```

Graphical Representation

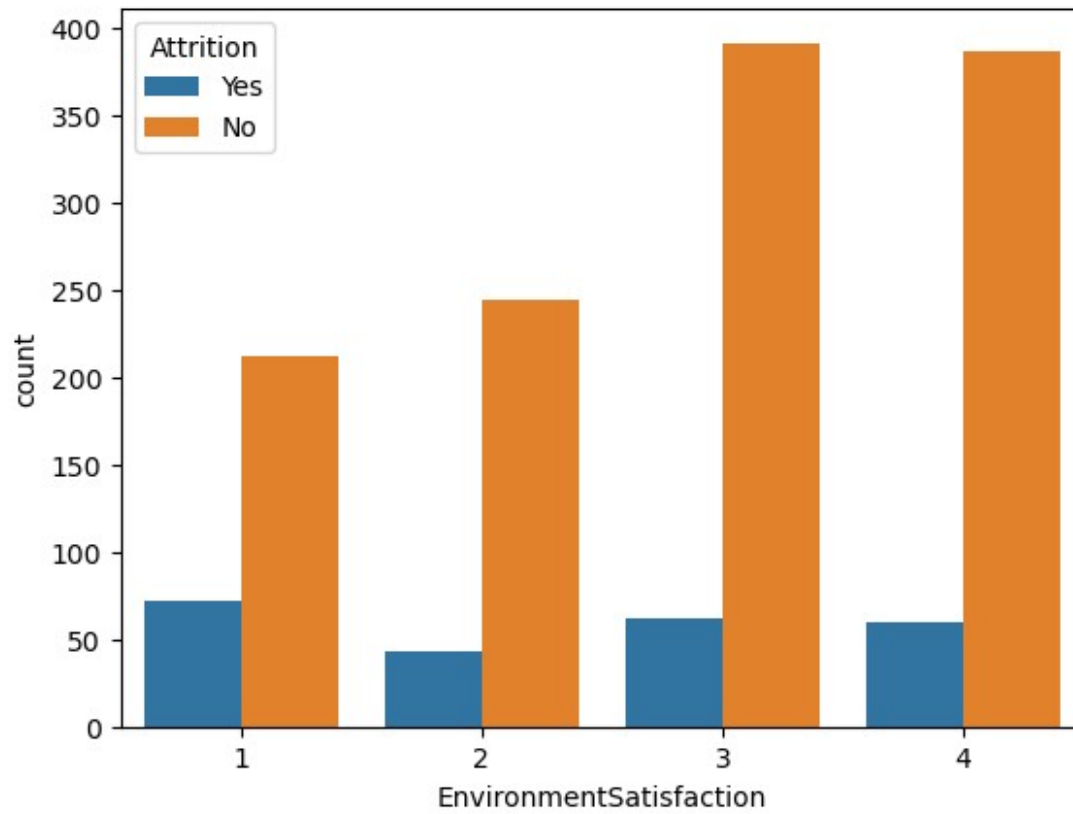
```
plt.figure(figsize=(20,25), facecolor='white')
plotnumber = 1

for column in data3:
    if plotnumber<=16 :
        ax = plt.subplot(4,4,plotnumber)
        sns.histplot(x=data3[column].dropna(axis=0)
                    ,hue=data.Attrition)
        plt.xlabel(column,fontsize=20)
        plt.ylabel('Attrition',fontsize=20)
        plotnumber+=1
plt.tight_layout()
```



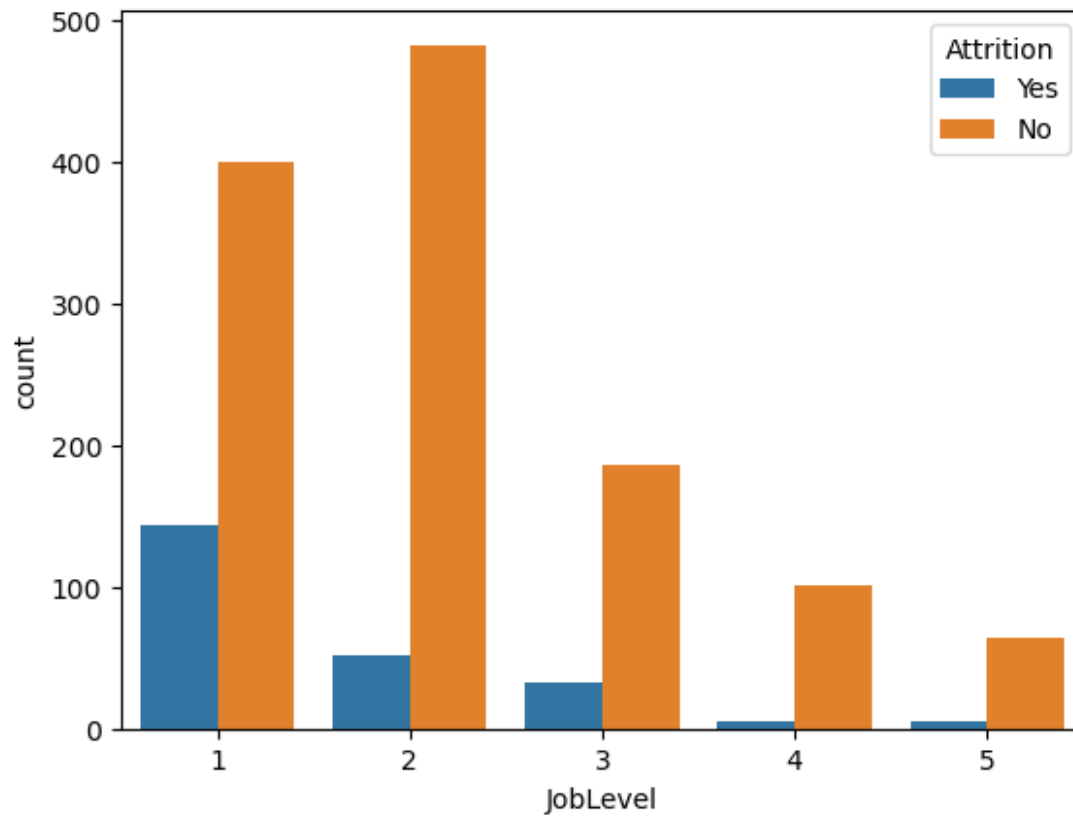
1. Impact Of Environment And Job Satisfaction On Attrition

```
sns.countplot(hue=data.Attrition,x=data.EnvironmentSatisfaction)
plt.show()
```



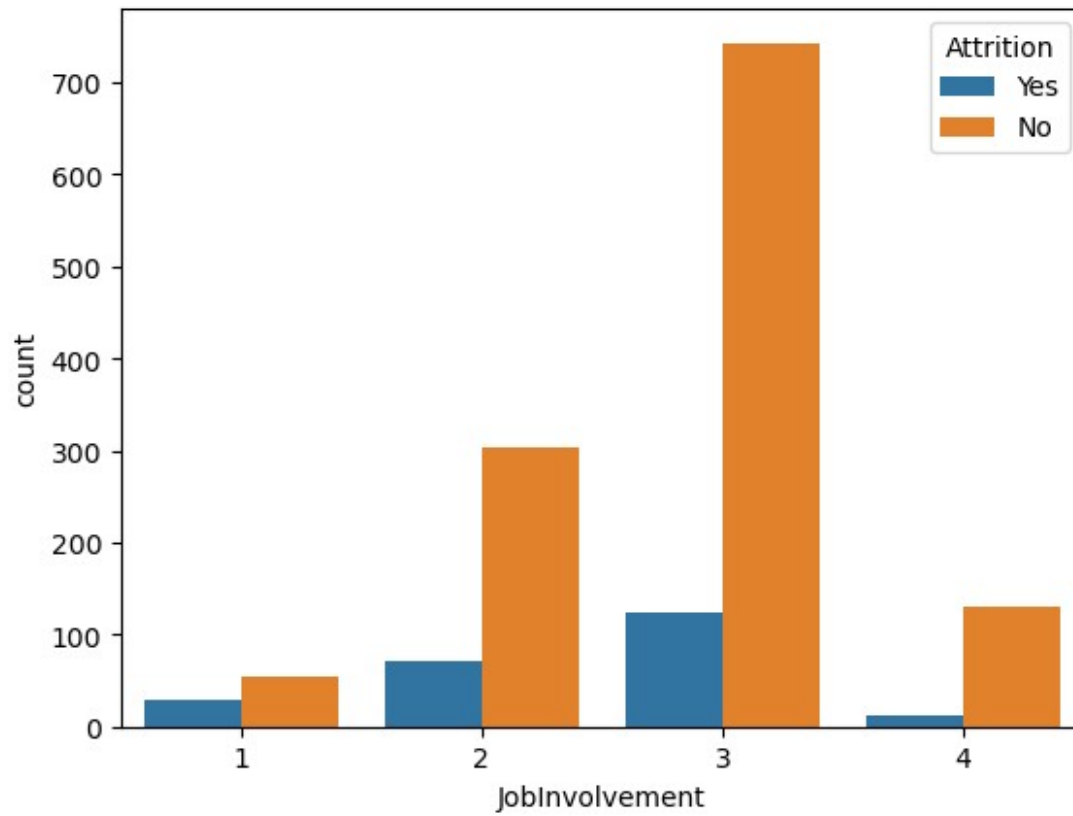
1. How Job Level Is Affecting Attrition

```
sns.countplot(hue=data.Attrition,x=data.JobLevel)  
plt.show()
```



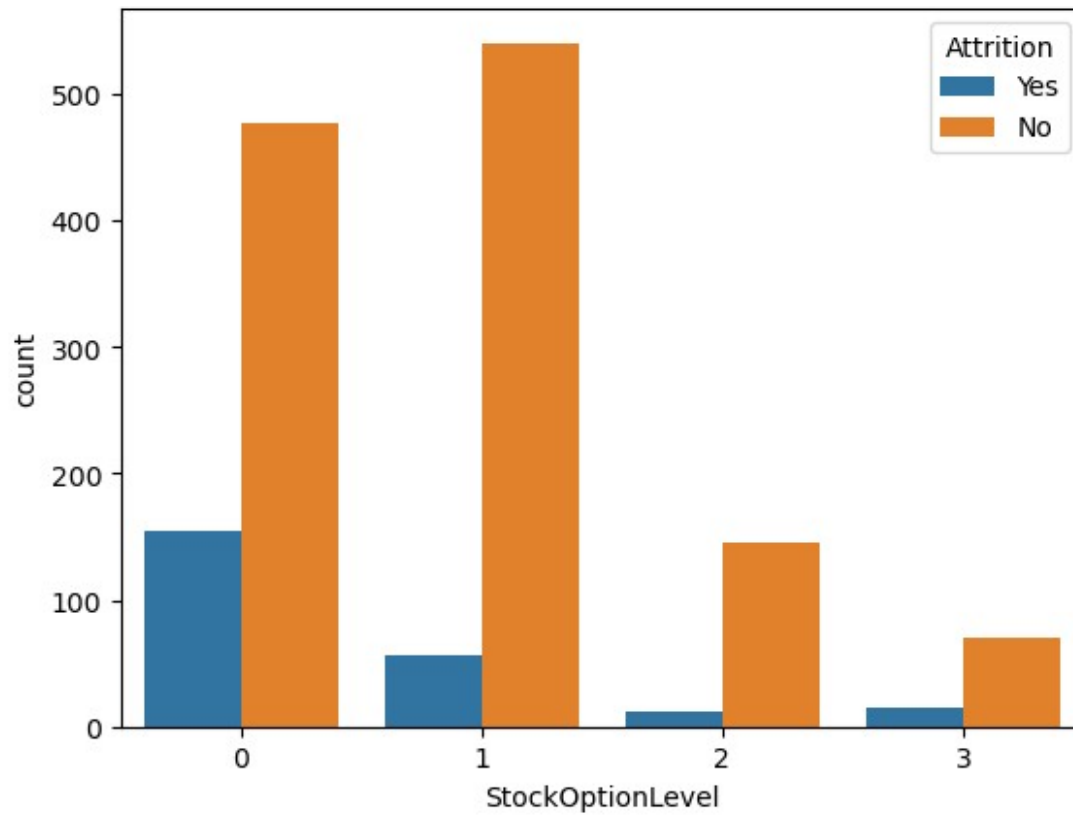
1. Job Involvement Impacting On Attrition

```
sns.countplot(hue=data.Attrition,x=data.JobInvolvement)  
plt.show()
```

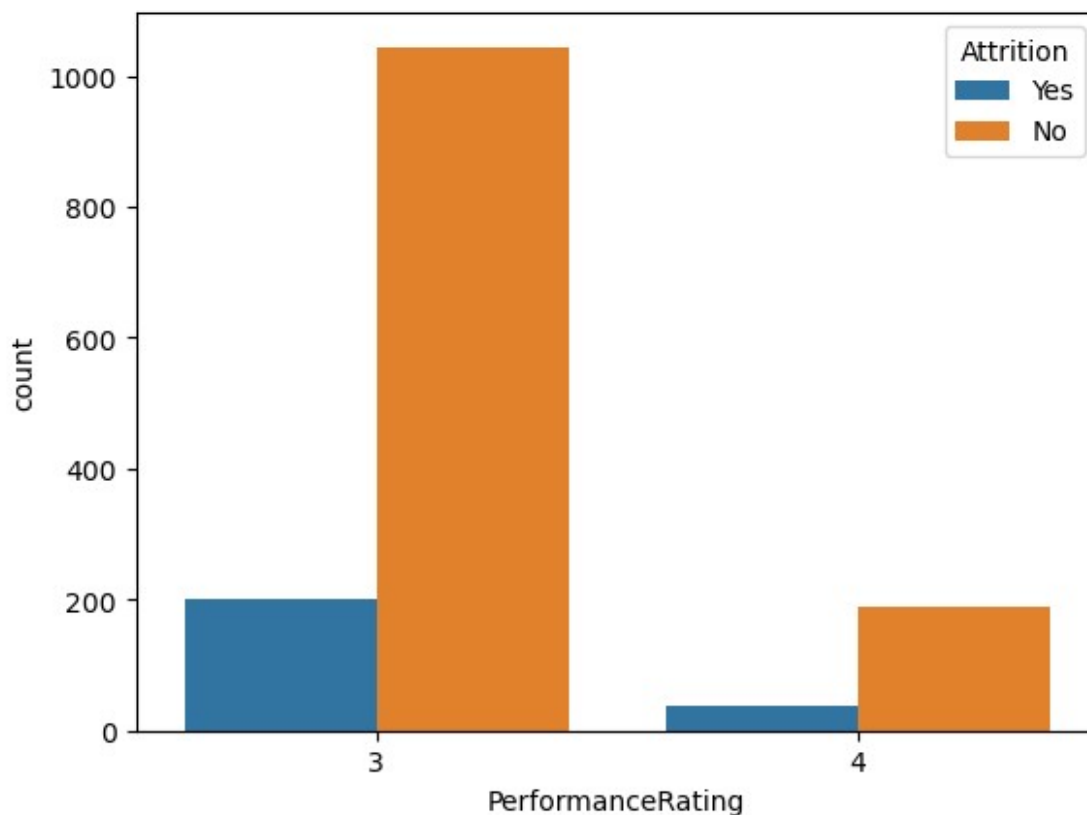
1. Impact Of Stock Option Level On Attrition

```
sns.countplot(hue=data.Attrition,x=data.StockOptionLevel)  
plt.show()
```



1. Performance Rating And Attrition

```
sns.countplot(hue=data.Attrition,x=data.PerformanceRating)  
plt.show()
```



```
data.isnull().sum()
```

Age	0
Attrition	0
BusinessTravel	0
DailyRate	0
Department	0
DistanceFromHome	0
Education	0
EducationField	0
EmployeeCount	0
EmployeeNumber	0
EnvironmentSatisfaction	0
Gender	0
HourlyRate	0
JobInvolvement	0
JobLevel	0
JobRole	0
JobSatisfaction	0
MaritalStatus	0
MonthlyIncome	0
MonthlyRate	0
NumCompaniesWorked	0
Over18	0

```
OverTime          0
PercentSalaryHike 0
PerformanceRating 0
RelationshipSatisfaction 0
StandardHours     0
StockOptionLevel  0
TotalWorkingYears 0
TrainingTimesLastYear 0
WorkLifeBalance   0
YearsAtCompany    0
YearsInCurrentRole 0
YearsSinceLastPromotion 0
YearsWithCurrManager 0
dtype: int64
```

```
print(data.duplicated().value_counts())
data.drop_duplicates(inplace=True)
print(len(data))
```

```
False    1470
dtype: int64
1470
```