

Project 1 HR Analytics (MeriSkill)

October 20, 2023

1 HR Analytics Project

1.1 Project Description

This Jupyter Notebook contains the analysis and findings of the HR Analytics project. The project's main objective is to analyze employee attrition within the organization and identify factors that contribute to attrition. We explore various aspects such as employee satisfaction, career progression, work-life balance, and more to gain insights into attrition patterns.

The analysis includes data cleaning, data visualization, and recommendations for HR strategies based on the findings.

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GitHub: [ckrazaditya](#)

1.2 Import Libraries

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

Load the Data

```
[2]: df = pd.read_csv("C:/Users/Lenovo/Desktop/Meri skill Project/
↳drive-download-20231003T163556Z-001/Project 3 - HR Analytics/Data P3_
↳MeriSKILL/HR-Employee-Attrition.csv")
```

1.3 Data Exploration

```
[3]: df.head()
```

```
[3]:   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	EmployeeNumber	\
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	

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

	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	

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

[5 rows x 35 columns]

```
[4]: df.tail()
```

```
[4]:
```

	Age	Attrition	BusinessTravel	DailyRate	Department	\
1465	36	No	Travel_Frequently	884	Research & Development	
1466	39	No	Travel_Rarely	613	Research & Development	
1467	27	No	Travel_Rarely	155	Research & Development	
1468	49	No	Travel_Frequently	1023	Sales	
1469	34	No	Travel_Rarely	628	Research & Development	

	DistanceFromHome	Education	EducationField	EmployeeCount	\
1465	23	2	Medical	1	
1466	6	1	Medical	1	
1467	4	3	Life Sciences	1	
1468	2	3	Medical	1	
1469	8	3	Medical	1	

EmployeeNumber	...	RelationshipSatisfaction	StandardHours	\
----------------	-----	--------------------------	---------------	---

1465	2061	...	3	80
1466	2062	...	1	80
1467	2064	...	2	80
1468	2065	...	4	80
1469	2068	...	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

[5 rows x 35 columns]

```
[5]: df.describe()
```

```
[5]:
```

	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	...	RelationshipSatisfaction	StandardHours	\
count	1470.000000	...	1470.000000	1470.0	
mean	2.063946	...	2.712245	80.0	
std	1.106940	...	1.081209	0.0	
min	1.000000	...	1.000000	80.0	
25%	1.000000	...	2.000000	80.0	
50%	2.000000	...	3.000000	80.0	
75%	3.000000	...	4.000000	80.0	
max	5.000000	...	4.000000	80.0	

	StockOptionLevel	TotalWorkingYears	TrainingTimesLastYear	\
count	1470.000000	1470.000000	1470.000000	
mean	0.793878	11.279592	2.799320	
std	0.852077	7.780782	1.289271	
min	0.000000	0.000000	0.000000	
25%	0.000000	6.000000	2.000000	
50%	1.000000	10.000000	3.000000	
75%	1.000000	15.000000	3.000000	
max	3.000000	40.000000	6.000000	

	WorkLifeBalance	YearsAtCompany	YearsInCurrentRole	\
count	1470.000000	1470.000000	1470.000000	
mean	2.761224	7.008163	4.229252	
std	0.706476	6.126525	3.623137	
min	1.000000	0.000000	0.000000	
25%	2.000000	3.000000	2.000000	
50%	3.000000	5.000000	3.000000	
75%	3.000000	9.000000	7.000000	
max	4.000000	40.000000	18.000000	

	YearsSinceLastPromotion	YearsWithCurrManager
count	1470.000000	1470.000000
mean	2.187755	4.123129
std	3.222430	3.568136
min	0.000000	0.000000
25%	0.000000	2.000000
50%	1.000000	3.000000
75%	3.000000	7.000000
max	15.000000	17.000000

[8 rows x 26 columns]

```
[6]: df.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

```

1.4 Data Cleaning and Preprocessing

```
[7]: df.dropna(inplace=True)
```

```
[8]: df.isnull().sum()
```

```
[8]: 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
```

```
[9]: df = df.drop(["EmployeeCount", "Over18", "StandardHours"], axis=1)
```

```
[10]: df = df.rename(columns={"Attrition": "Attrition", "JobSatisfaction": "Job_Satisfaction",})
```

```
[11]: df = df.dropna()
```

```
[12]: df.head()
```

```

[12]: 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 EmployeeNumber \
0              1          2 Life Sciences          1
1              8          1 Life Sciences          2
2              2          2      Other          4
3              3          4 Life Sciences          5
4              2          1      Medical          7

      EnvironmentSatisfaction ... PerformanceRating RelationshipSatisfaction \
0              2 ...              3              1
1              3 ...              4              4
2              4 ...              3              2
3              4 ...              3              3
4              1 ...              3              4

      StockOptionLevel TotalWorkingYears TrainingTimesLastYear WorkLifeBalance \
0              0              8              0              1
1              1             10              3              3
2              0              7              3              3
3              0              8              3              3
4              1              6              3              3

      YearsAtCompany YearsInCurrentRole YearsSinceLastPromotion \
0              6              4              0
1             10              7              1
2              0              0              0
3              8              7              3
4              2              2              2

      YearsWithCurrManager
0              5
1              7
2              0
3              0
4              2

[5 rows x 32 columns]

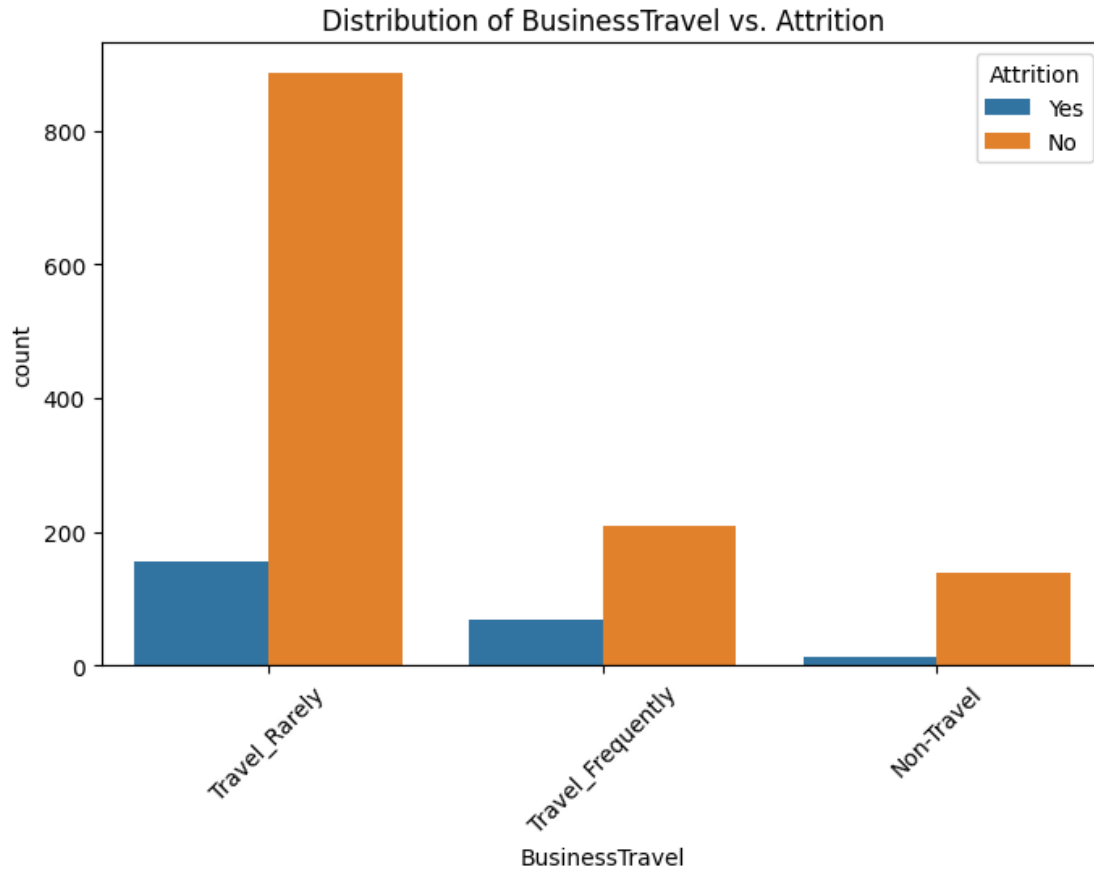
```

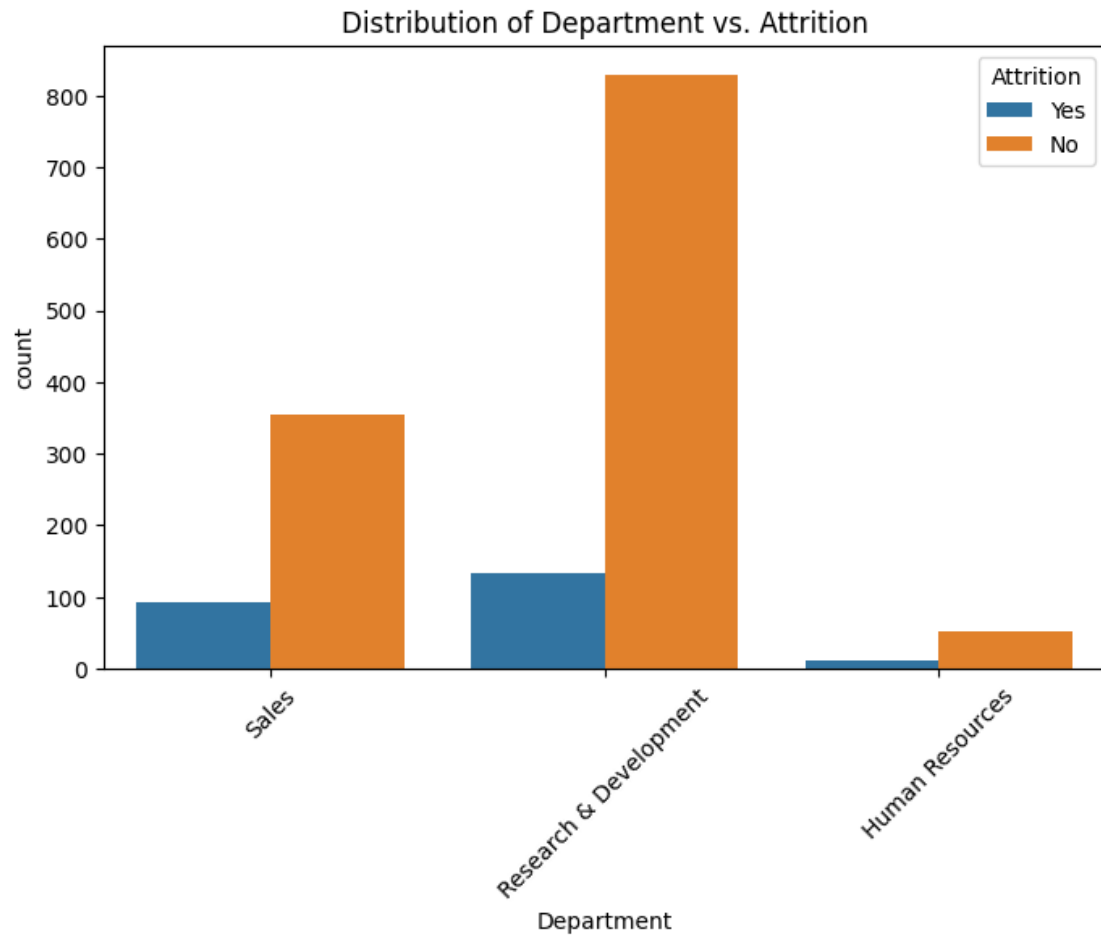
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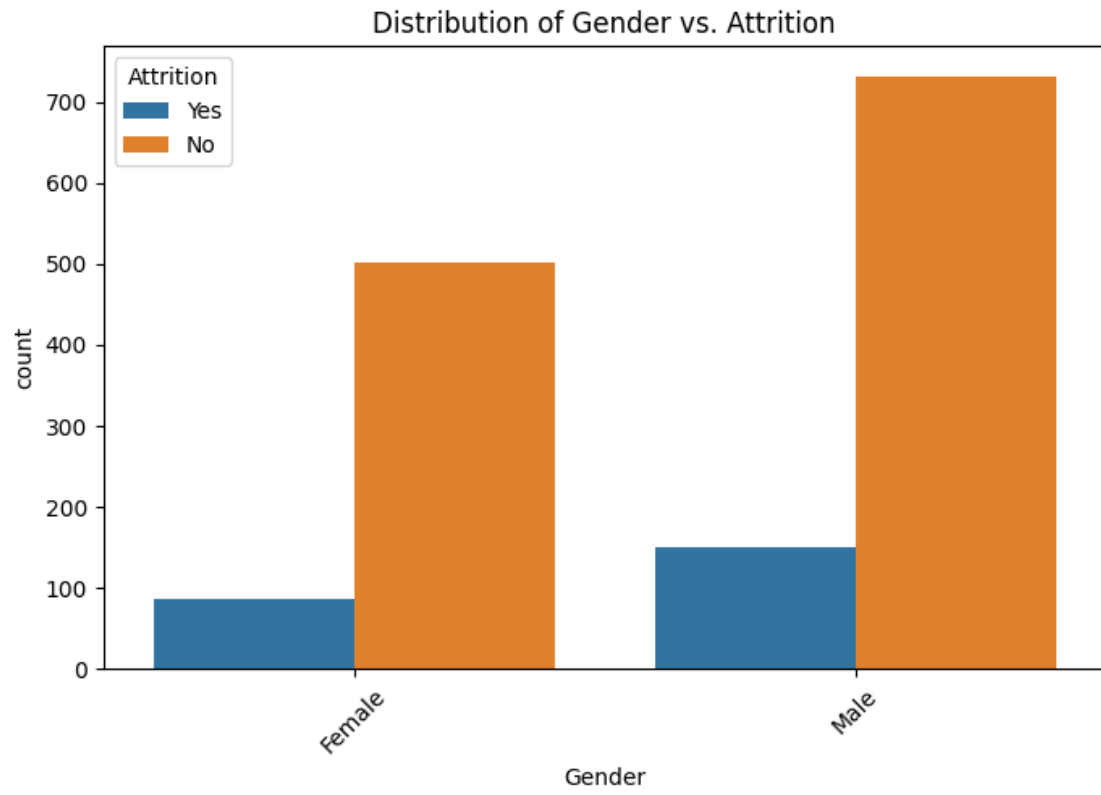
[13]: import matplotlib.pyplot as plt
import seaborn as sns

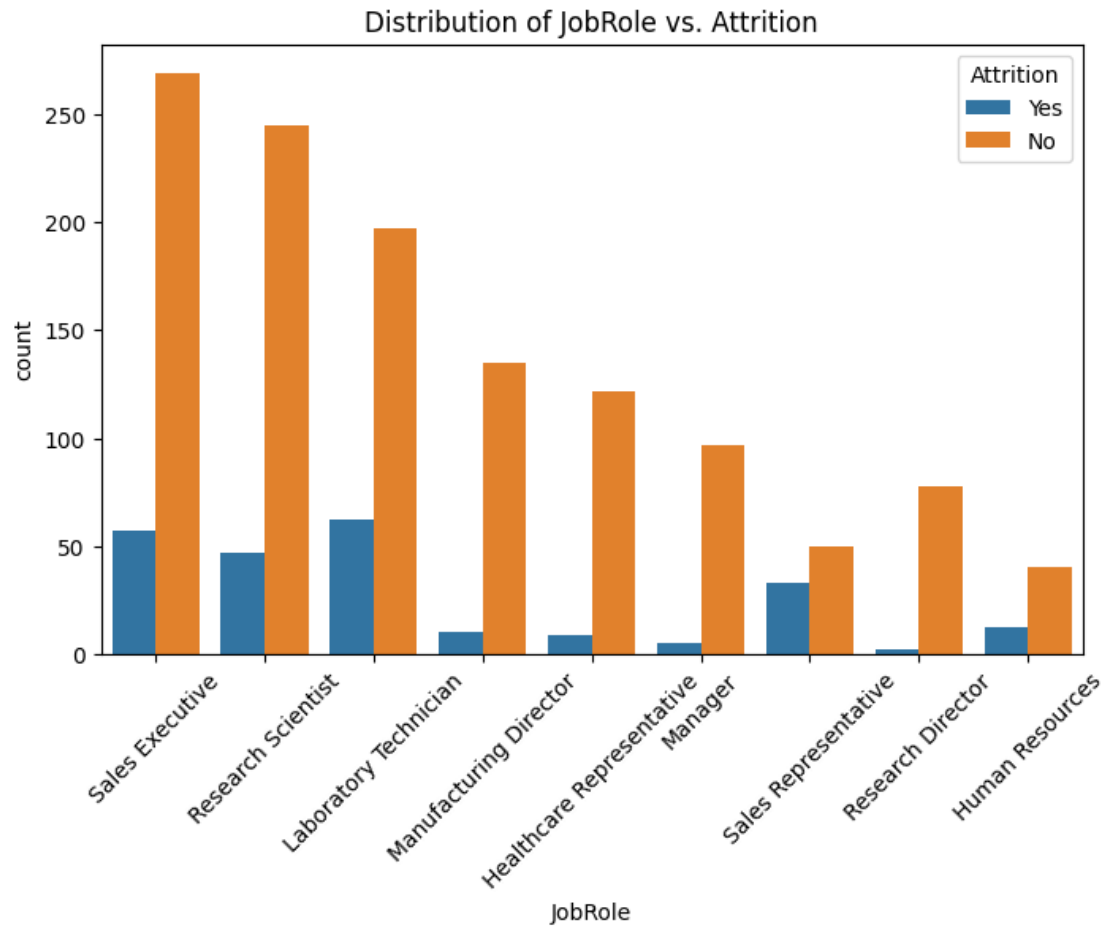
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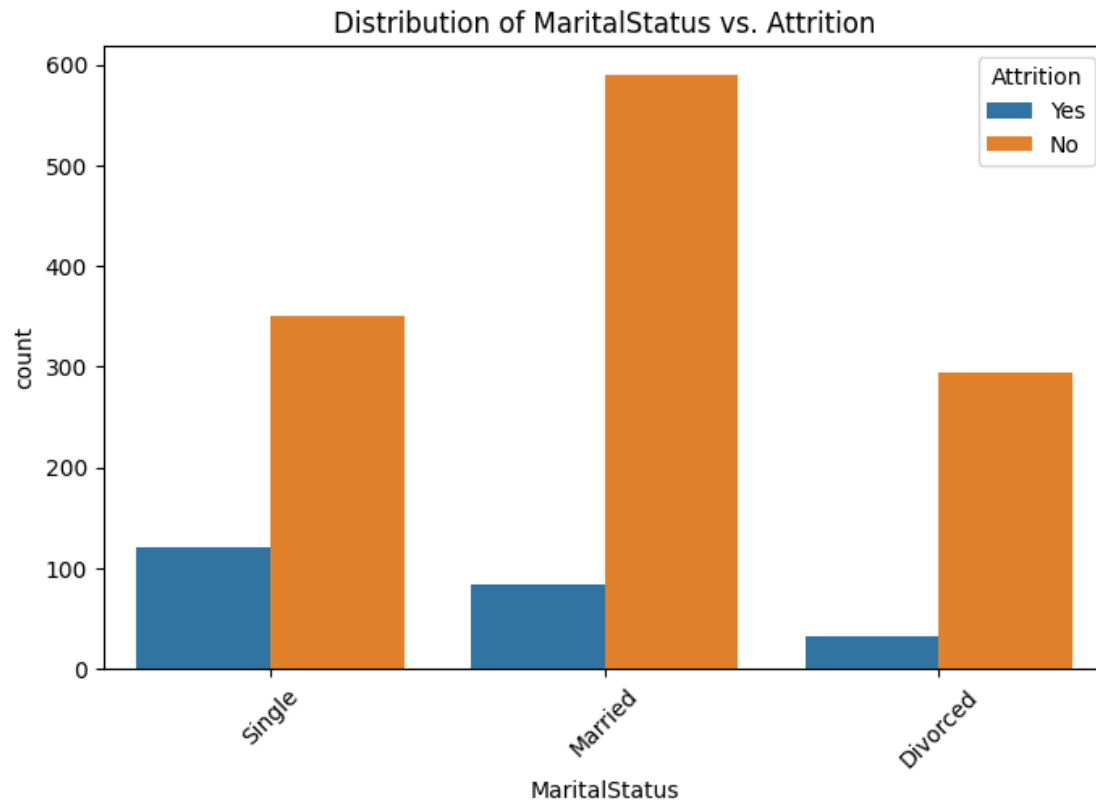
```
[14]: categorical_vars = ["BusinessTravel", "Department", "Gender", "JobRole", "
    ↳ "MaritalStatus"]
for var in categorical_vars:
    plt.figure(figsize=(8, 5))
    sns.countplot(data=df, x=var, hue="Attrition")
    plt.title(f"Distribution of {var} vs. Attrition")
    plt.xticks(rotation=45)
    plt.show()
```



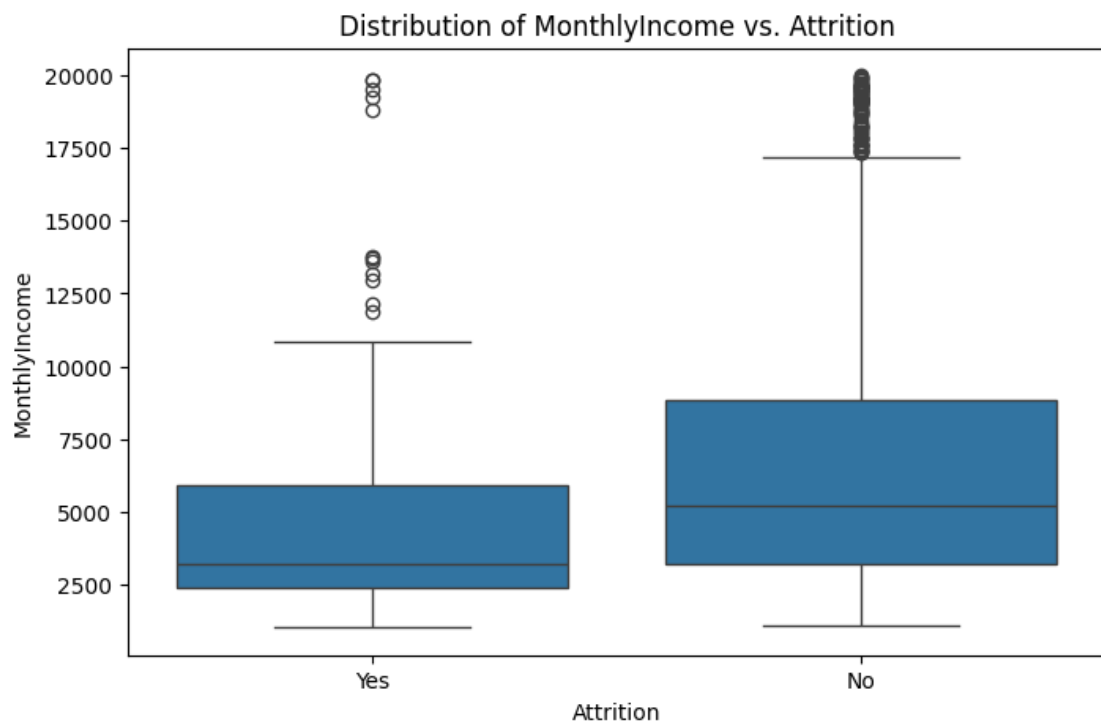
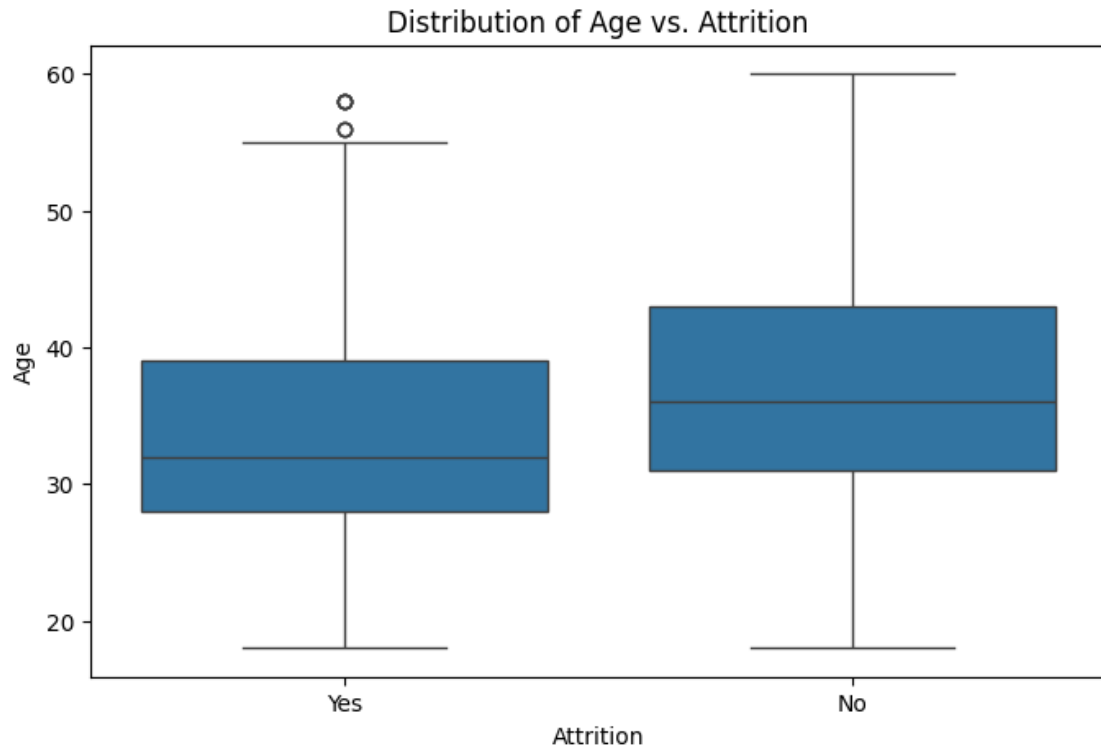


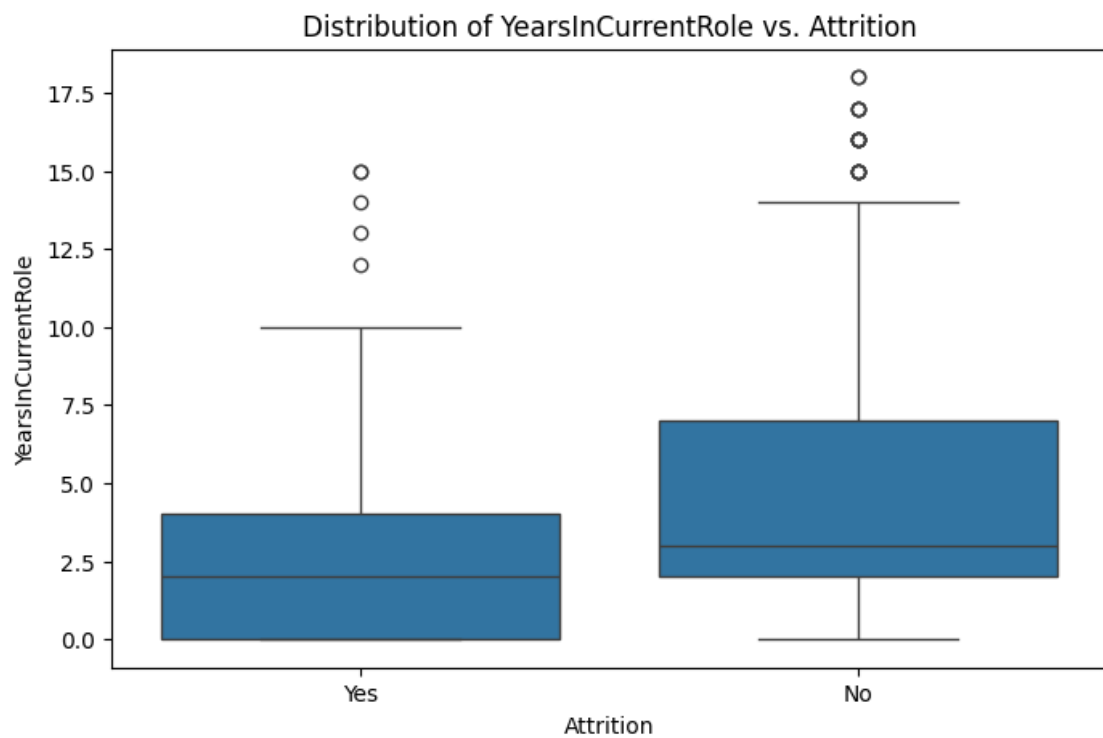
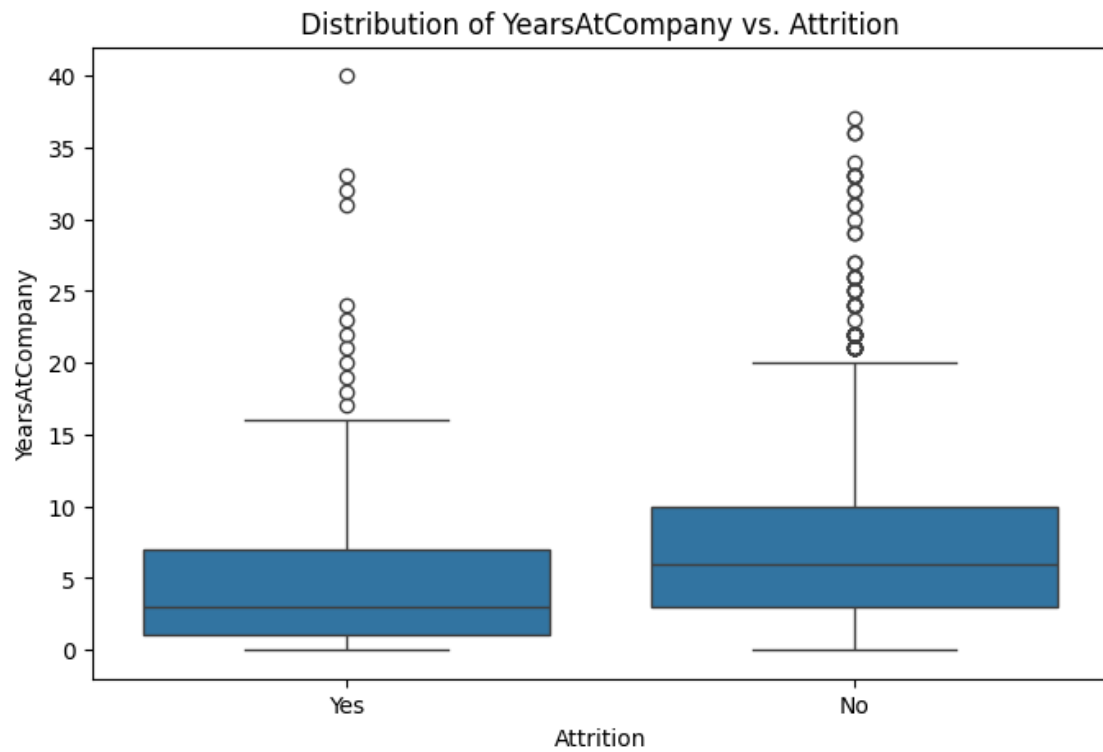




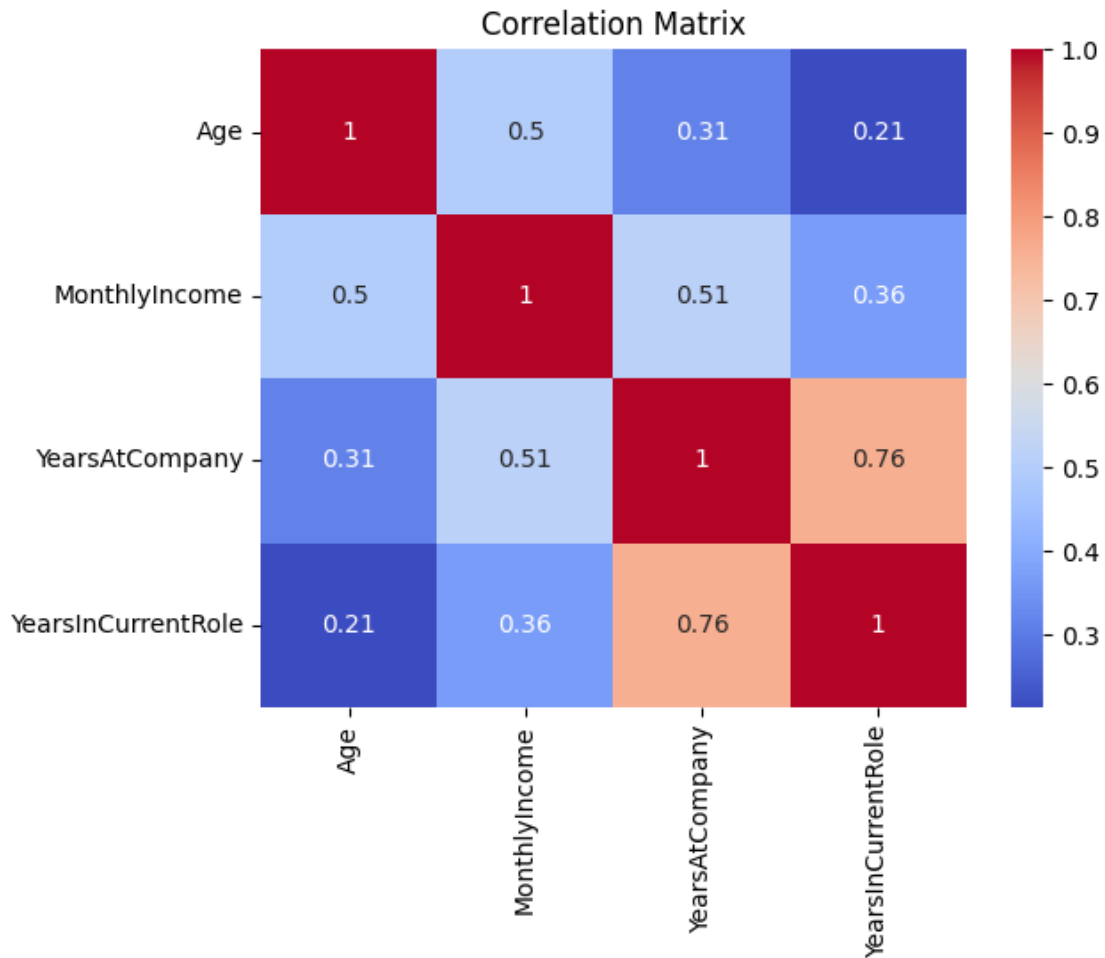


```
[15]: # Explore the distribution of numeric variables
numeric_vars = ["Age", "MonthlyIncome", "YearsAtCompany", "YearsInCurrentRole"]
for var in numeric_vars:
    plt.figure(figsize=(8, 5))
    sns.boxplot(data=df, y=var, x="Attrition")
    plt.title(f"Distribution of {var} vs. Attrition")
    plt.show()
```



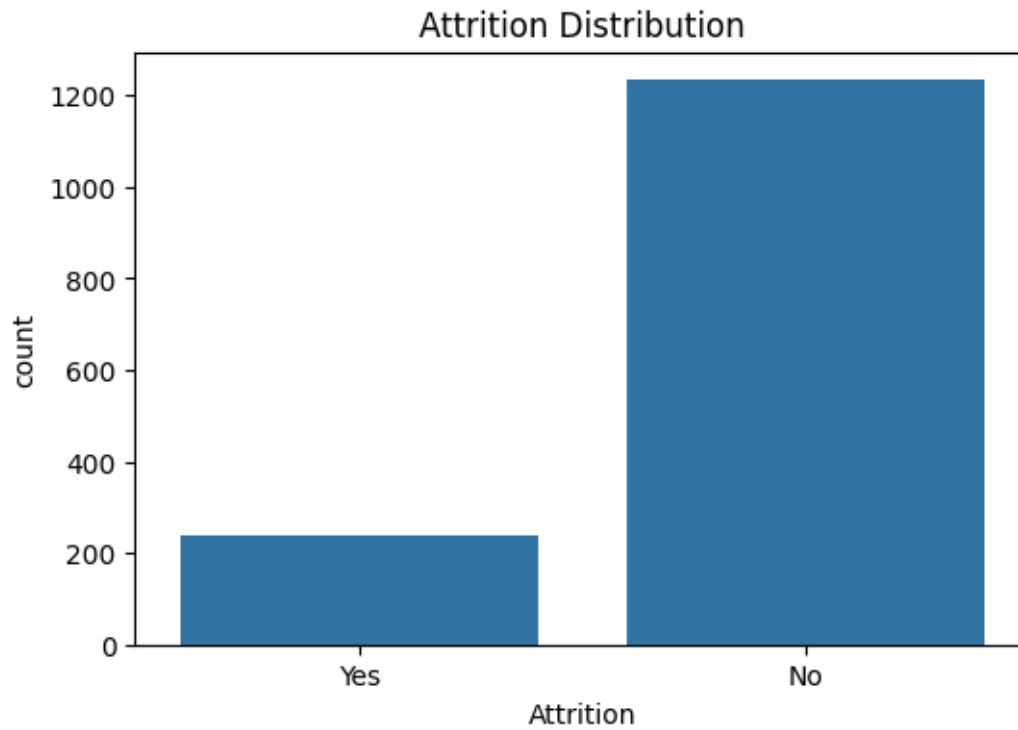


```
[16]: # Calculate and visualize the correlation between numeric variables
correlation_matrix = df[numeric_vars].corr()
sns.heatmap(correlation_matrix, annot=True, cmap="coolwarm")
plt.title("Correlation Matrix")
plt.show()
```

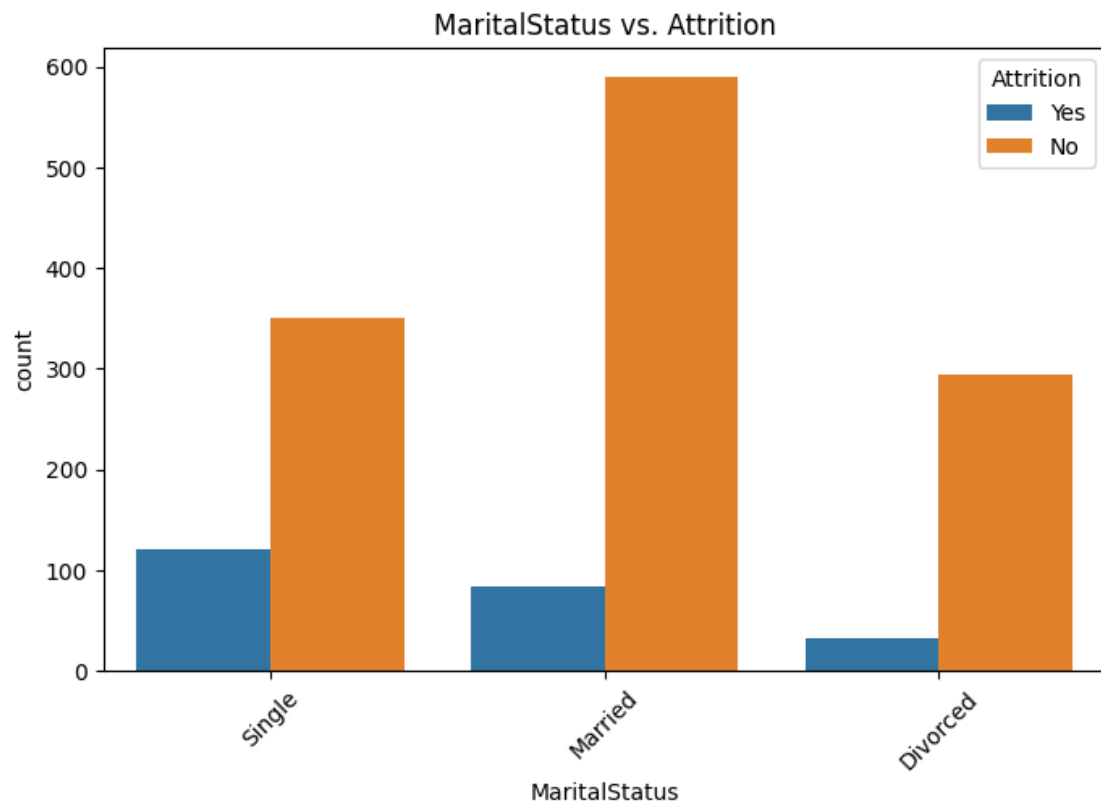


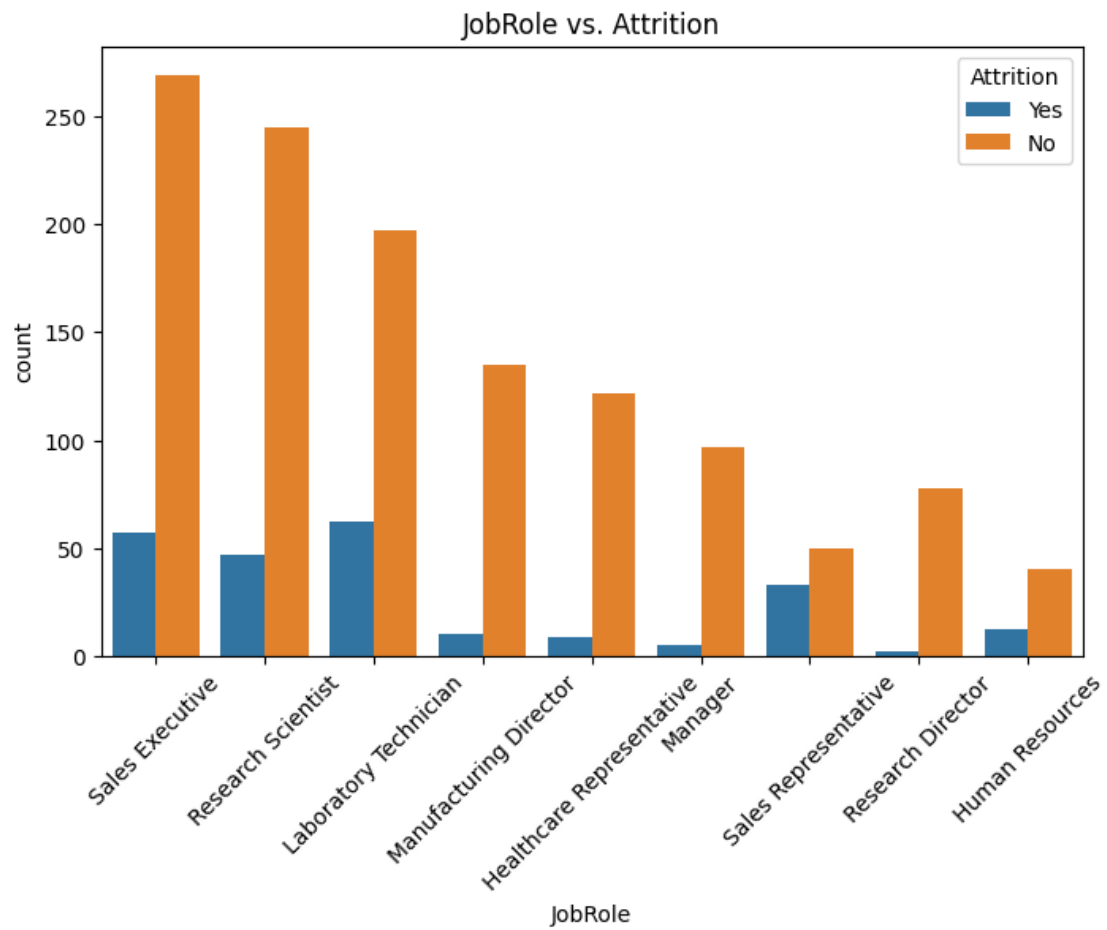
1.5 Analyze Attrition

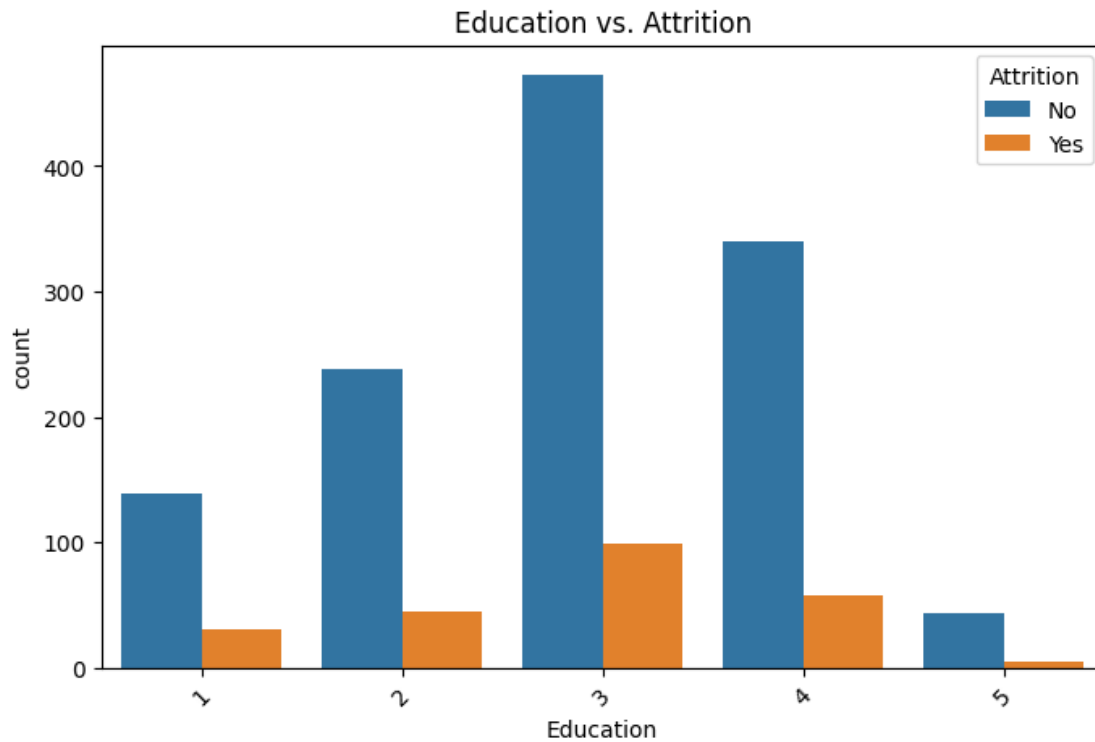
```
[17]: # Understand the distribution of Attrition
plt.figure(figsize=(6, 4))
sns.countplot(data=df, x="Attrition")
plt.title("Attrition Distribution")
plt.show()
```



```
[18]: # Explore factors related to attrition
attrition_factors = ["MaritalStatus", "JobRole", "Education"]
for var in attrition_factors:
    plt.figure(figsize=(8, 5))
    sns.countplot(data=df, x=var, hue="Attrition")
    plt.title(f"{var} vs. Attrition")
    plt.xticks(rotation=45)
    plt.show()
```

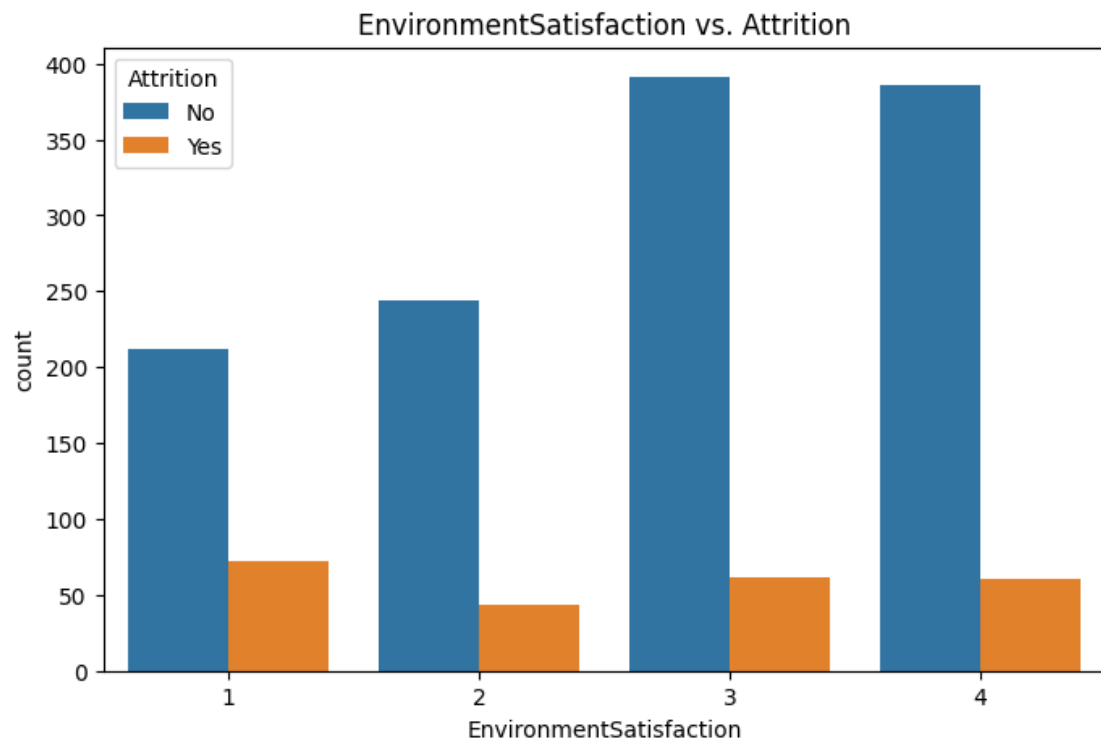
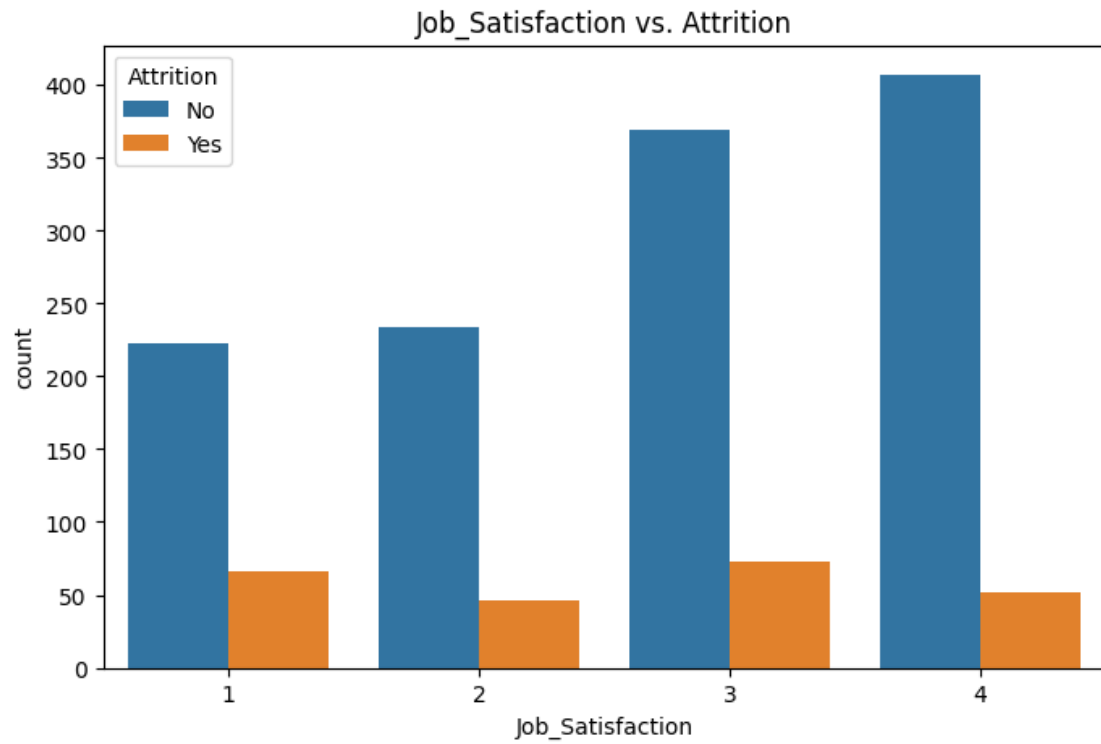



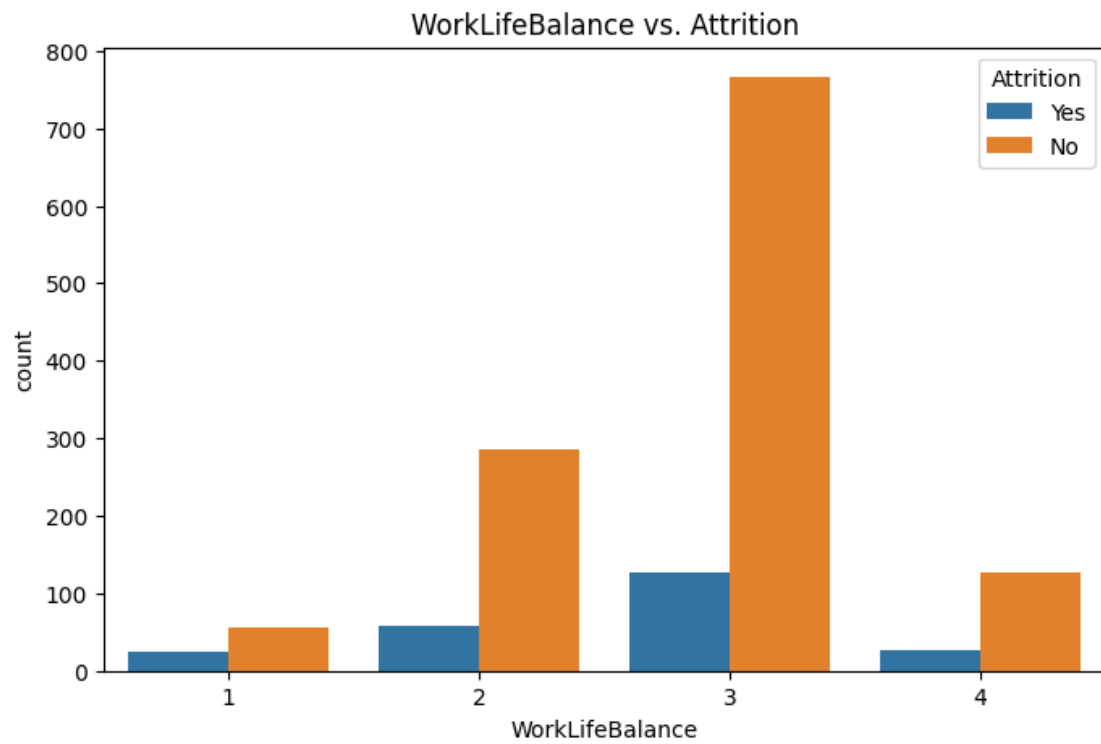
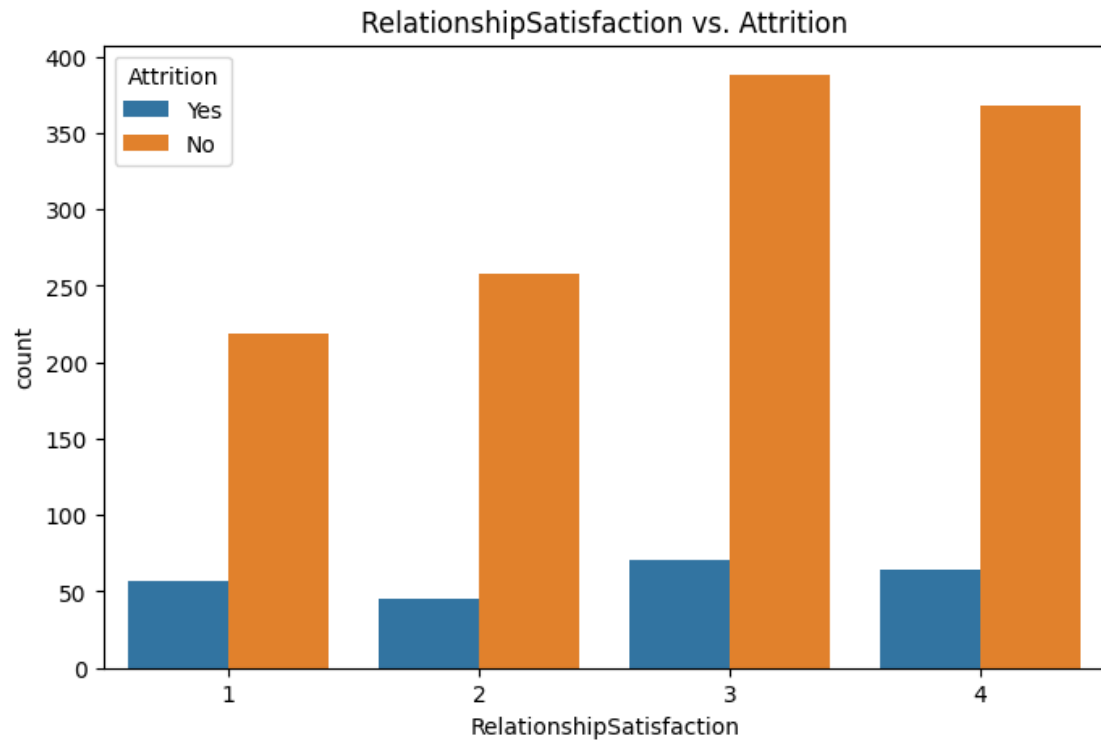




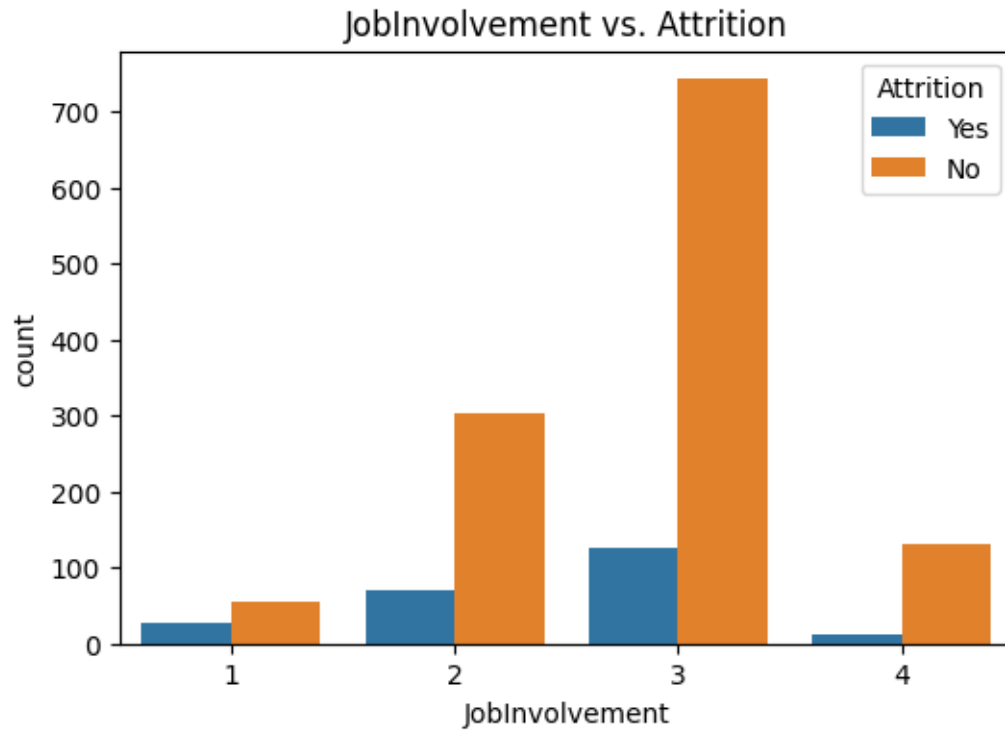
1.6 Employee Satisfaction and Engagement

```
[19]: # Analyze employee satisfaction and engagement
satisfaction_vars = ["Job_Satisfaction", "EnvironmentSatisfaction",
                    ↪ "RelationshipSatisfaction", "WorkLifeBalance"]
for var in satisfaction_vars:
    plt.figure(figsize=(8, 5))
    sns.countplot(data=df, x=var, hue="Attrition")
    plt.title(f"{var} vs. Attrition")
    plt.show()
```



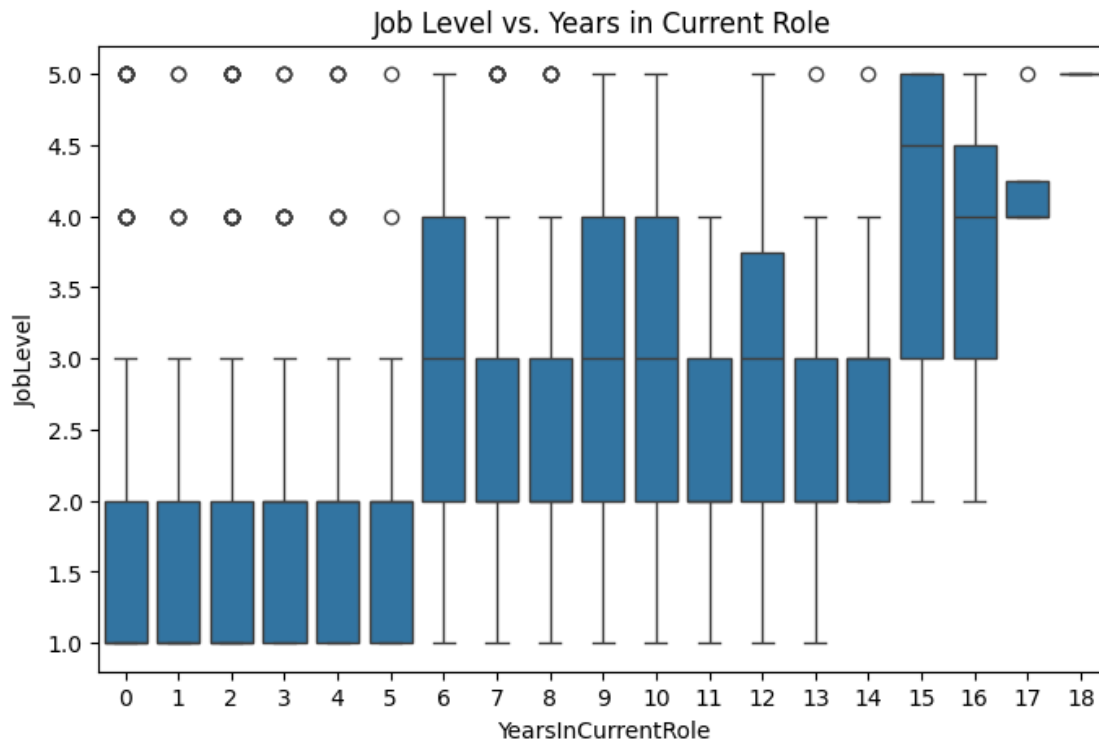


```
[20]: # Investigate the relationship between JobInvolvement and Attrition
plt.figure(figsize=(6, 4))
sns.countplot(data=df, x="JobInvolvement", hue="Attrition")
plt.title("JobInvolvement vs. Attrition")
plt.show()
```

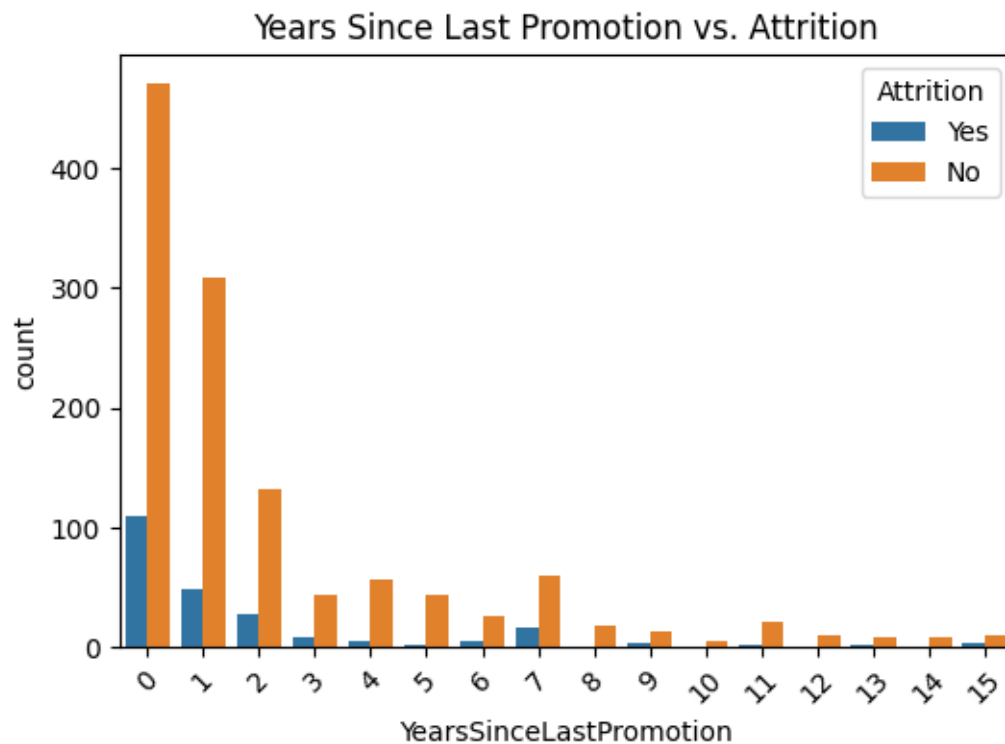


1.7 Career Progression

```
[21]: # Analyze employee career progression
plt.figure(figsize=(8, 5))
sns.boxplot(data=df, y="JobLevel", x="YearsInCurrentRole")
plt.title("Job Level vs. Years in Current Role")
plt.show()
```

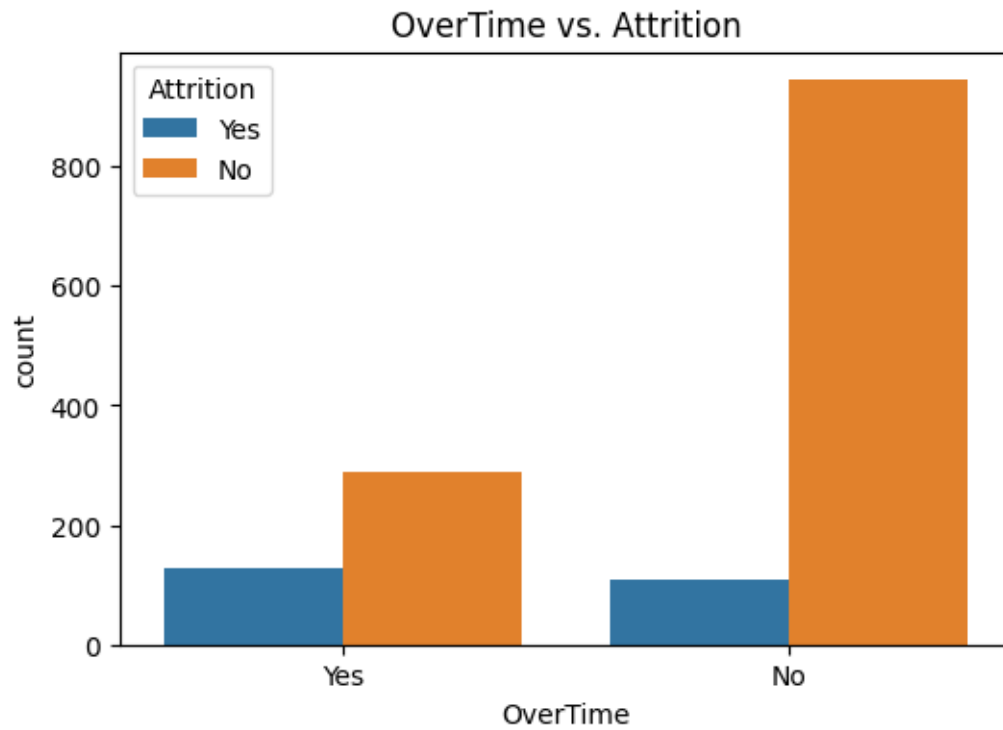


```
[22]: # Investigate the relationship between promotions and attrition
plt.figure(figsize=(6, 4))
sns.countplot(data=df, x="YearsSinceLastPromotion", hue="Attrition")
plt.title("Years Since Last Promotion vs. Attrition")
plt.xticks(rotation=45)
plt.show()
```

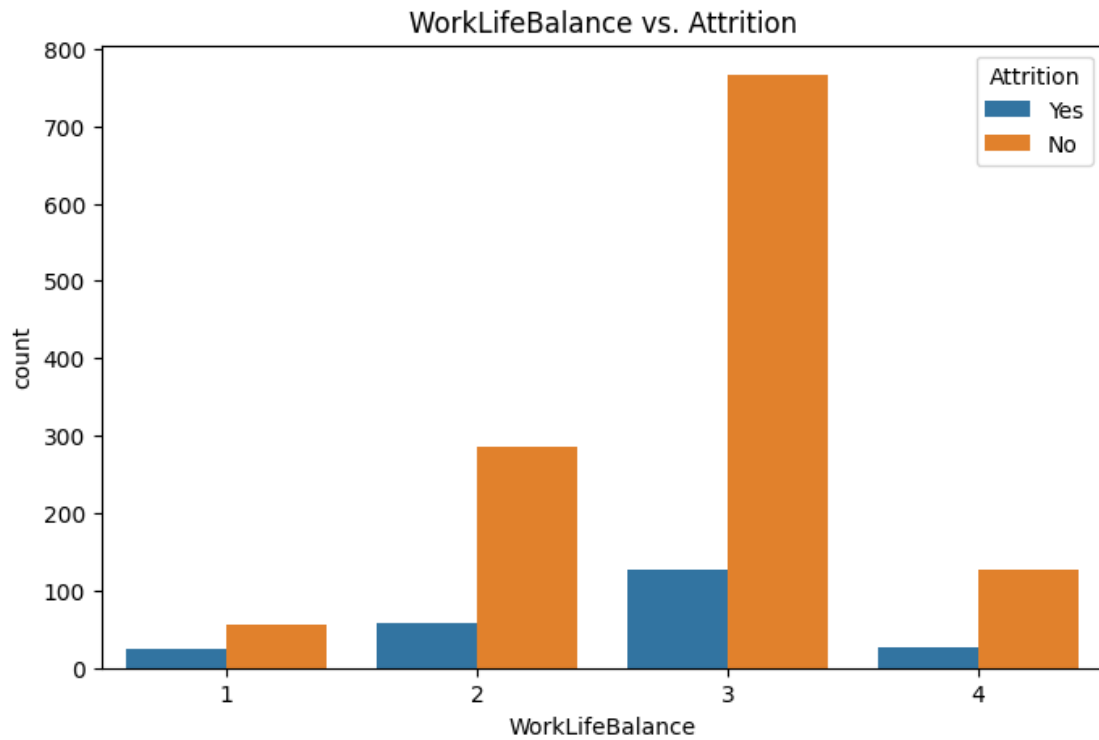


1.8 Work-Life Balance and Overtime

```
[23]: # Analyze the impact of work-life balance and overtime on attrition
plt.figure(figsize=(6, 4))
sns.countplot(data=df, x="OverTime", hue="Attrition")
plt.title("OverTime vs. Attrition")
plt.show()
```

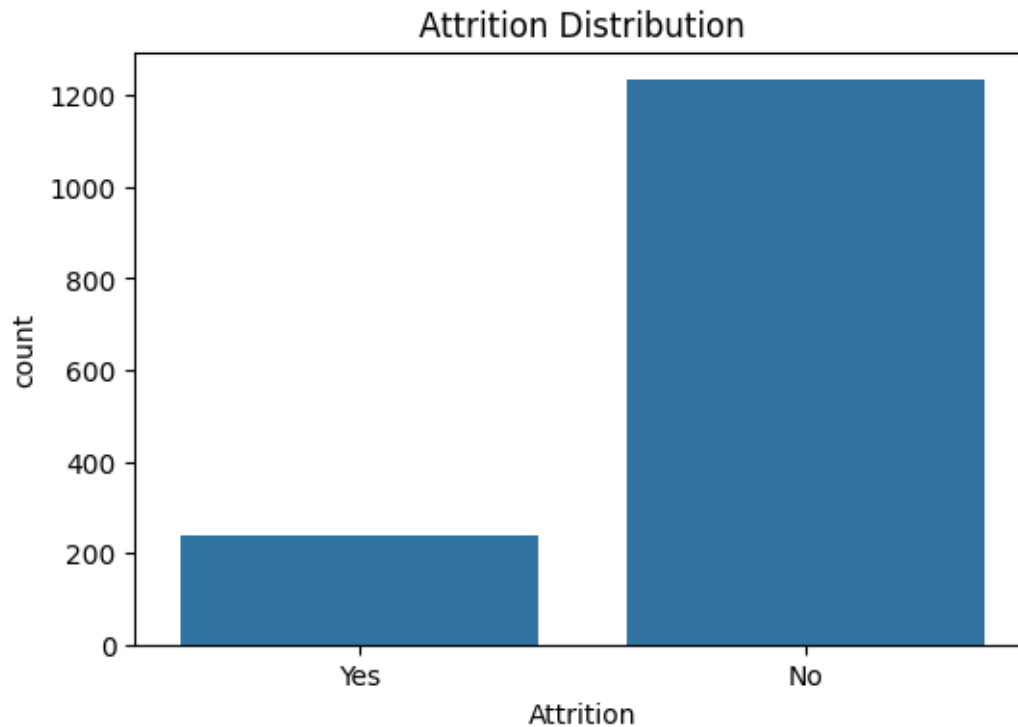



```
[24]: plt.figure(figsize=(8, 5))
sns.countplot(data=df, x="WorkLifeBalance", hue="Attrition")
plt.title("WorkLifeBalance vs. Attrition")
plt.show()
```



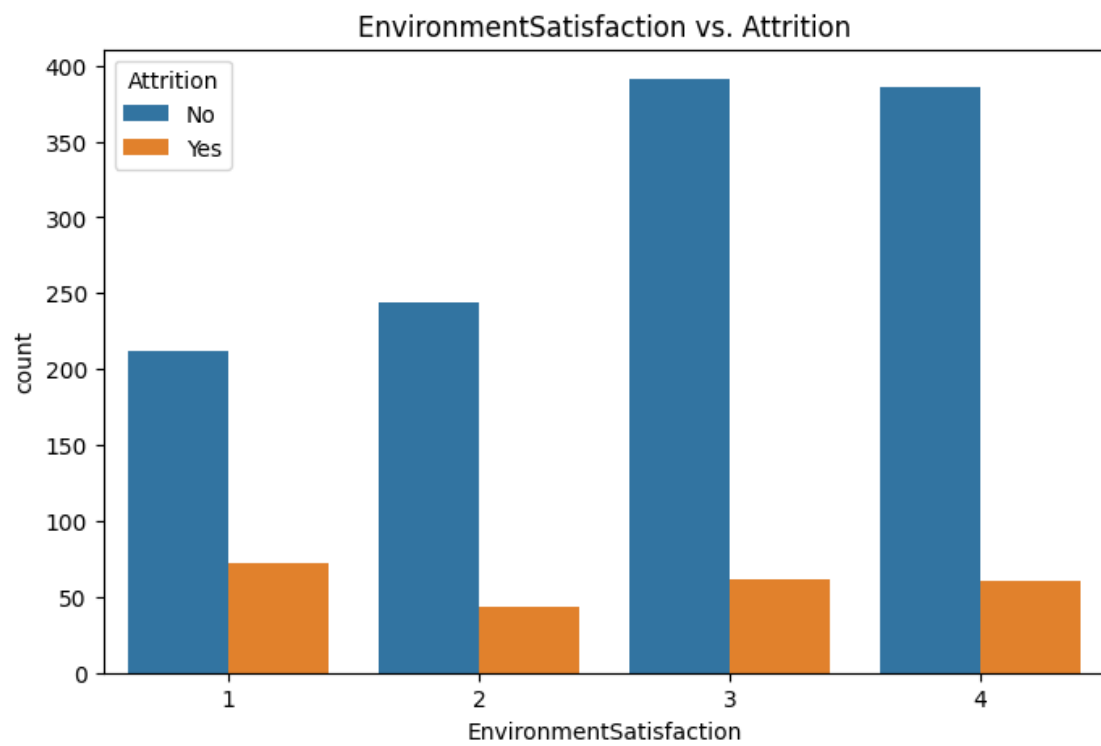
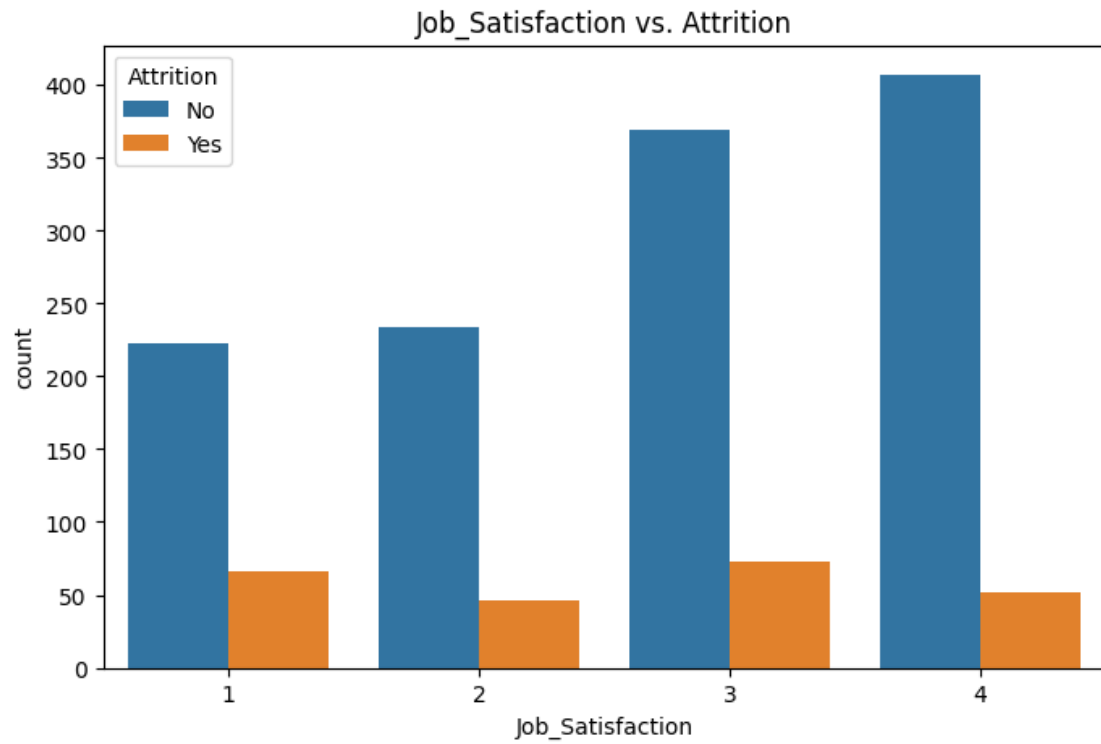
1.9 Analyze Attrition

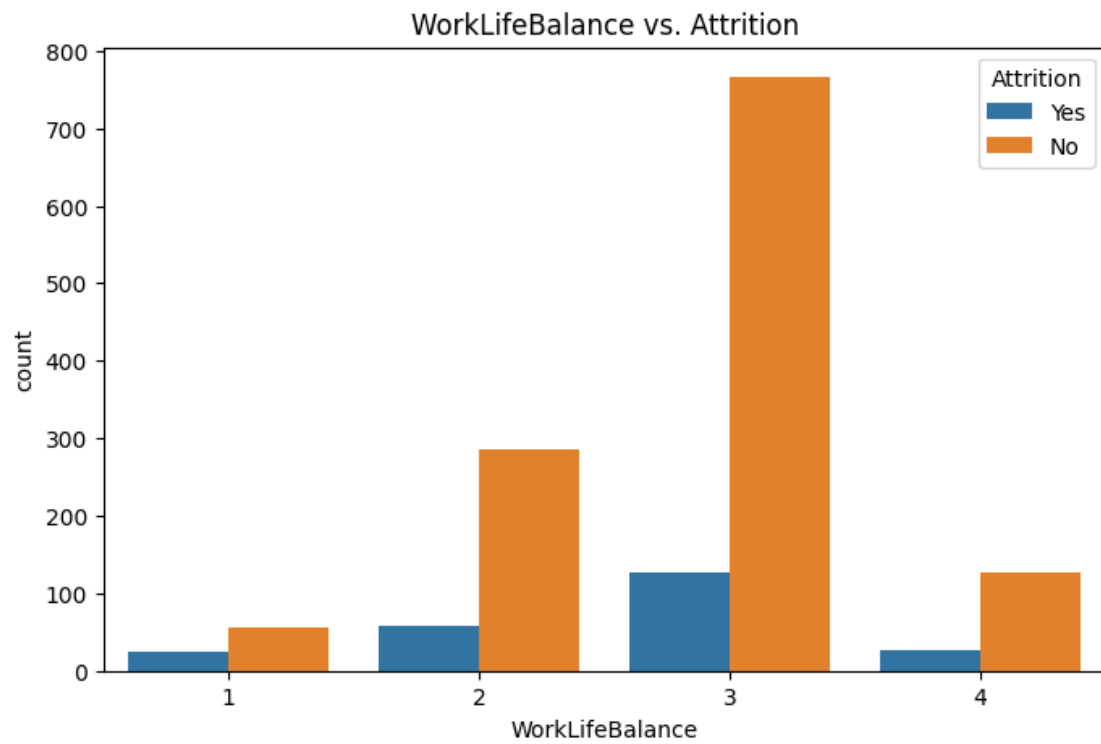
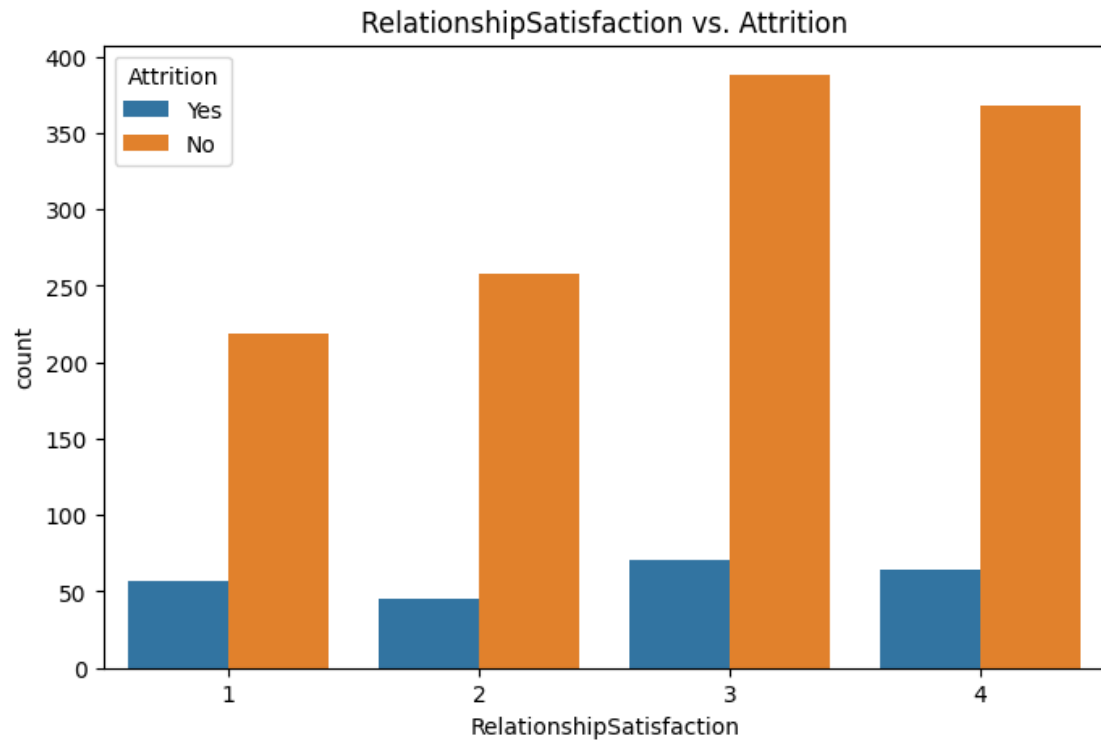
```
[25]: # Understand the distribution of Attrition
plt.figure(figsize=(6, 4))
sns.countplot(data=df, x="Attrition")
plt.title("Attrition Distribution")
plt.show()
```



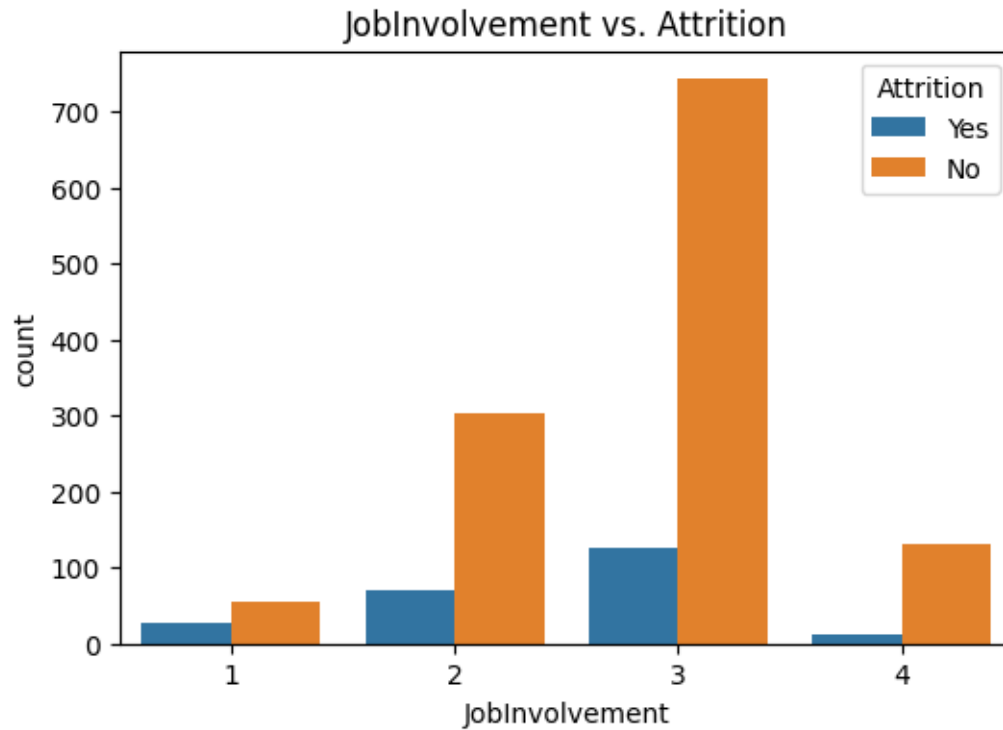
2 Explore factors related to attrition

```
[26]: # Analyze employee satisfaction and engagement
satisfaction_vars = ["Job_Satisfaction", "EnvironmentSatisfaction",
                    ↪ "RelationshipSatisfaction", "WorkLifeBalance"]
for var in satisfaction_vars:
    plt.figure(figsize=(8, 5))
    sns.countplot(data=df, x=var, hue="Attrition")
    plt.title(f"{var} vs. Attrition")
    plt.show()
```



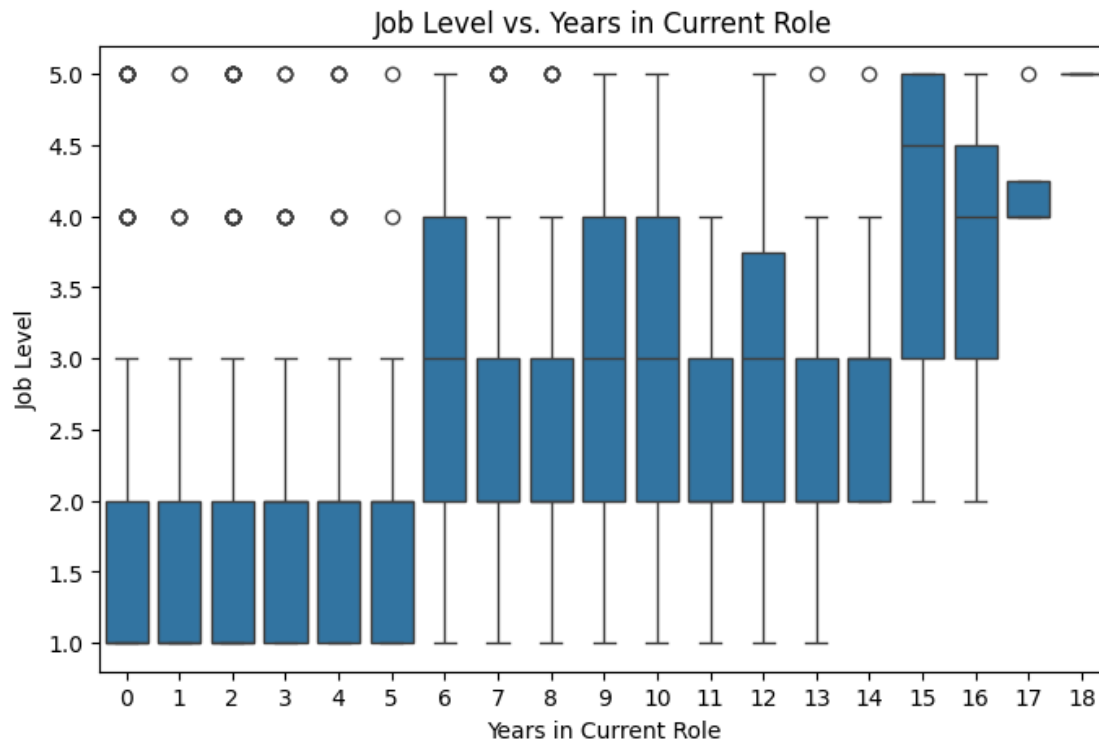


```
[27]: # Investigate the relationship between JobInvolvement and Attrition
plt.figure(figsize=(6, 4))
sns.countplot(data=df, x="JobInvolvement", hue="Attrition")
plt.title("JobInvolvement vs. Attrition")
plt.show()
```

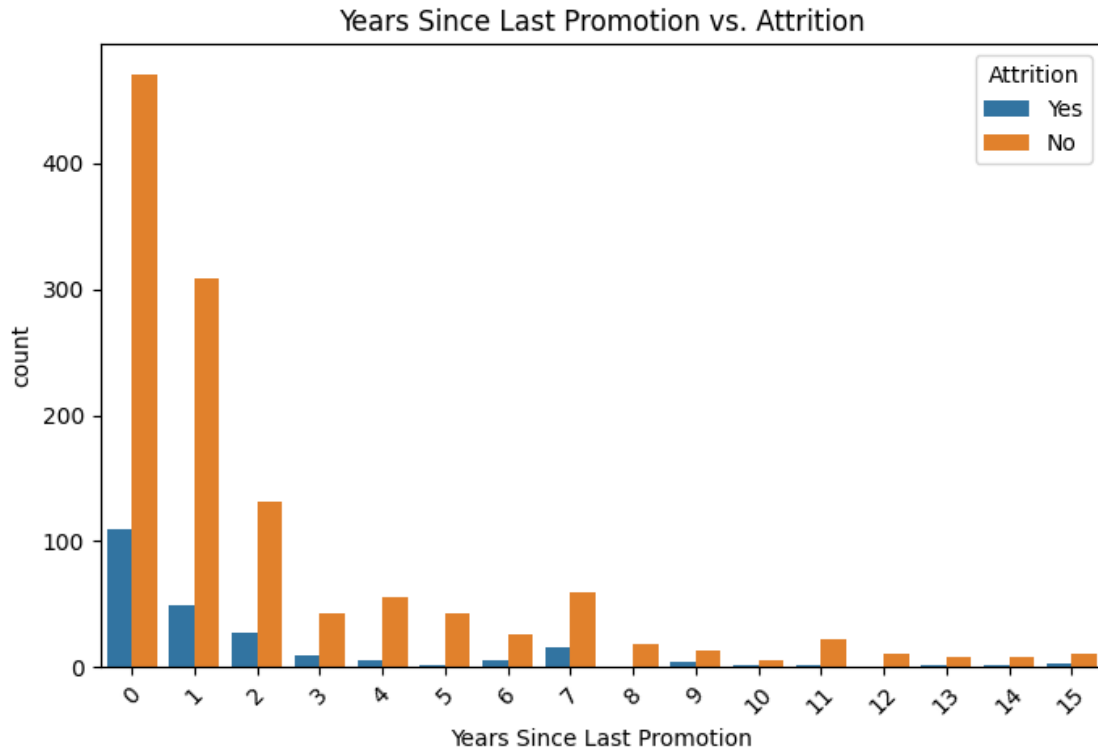


2.1 Career Progression

```
[28]: # Analyze employee career progression
plt.figure(figsize=(8, 5))
sns.boxplot(data=df, y="JobLevel", x="YearsInCurrentRole")
plt.title("Job Level vs. Years in Current Role")
plt.xlabel("Years in Current Role")
plt.ylabel("Job Level")
plt.show()
```

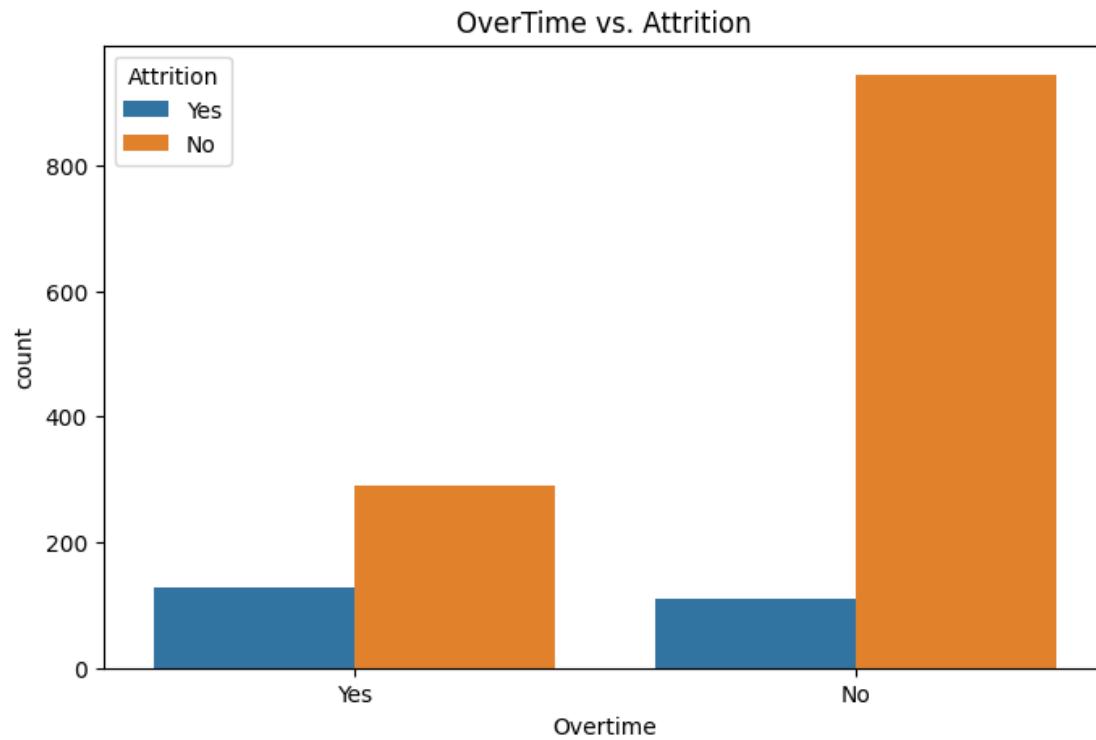


```
[29]: # Investigate the relationship between promotions and attrition
plt.figure(figsize=(8, 5))
sns.countplot(data=df, x="YearsSinceLastPromotion", hue="Attrition")
plt.title("Years Since Last Promotion vs. Attrition")
plt.xlabel("Years Since Last Promotion")
plt.xticks(rotation=45)
plt.show()
```

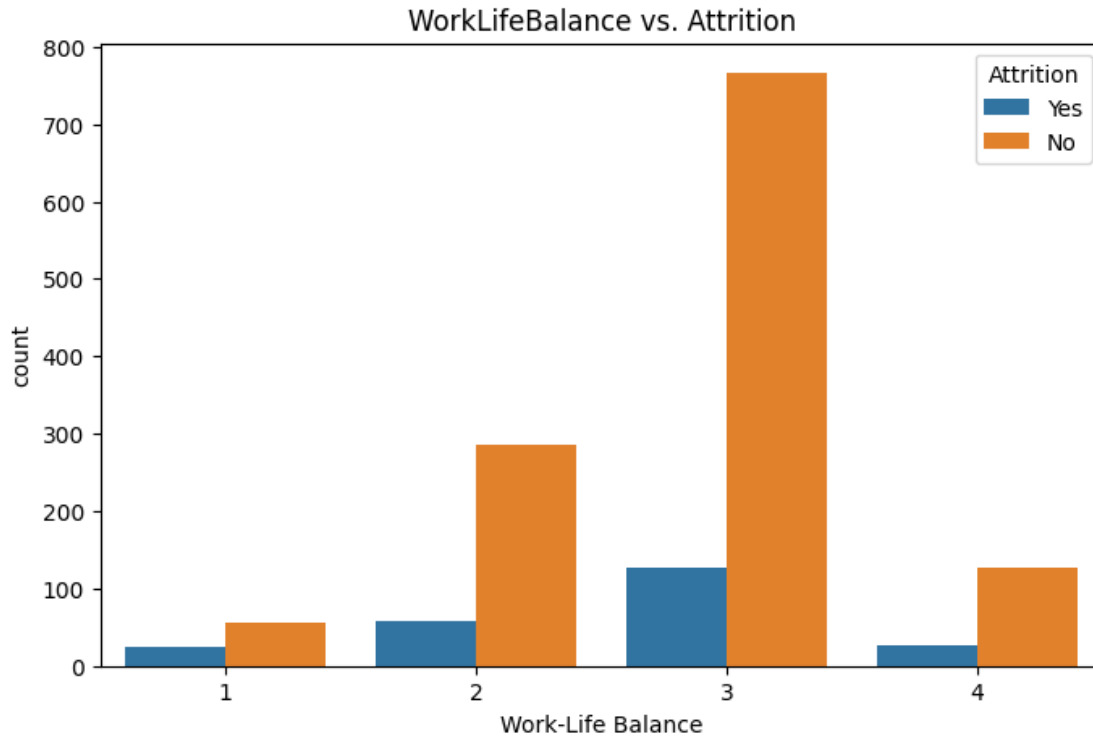


2.2 Work-Life Balance and Overtime

```
[30]: # Analyze the impact of work-life balance on attrition
plt.figure(figsize=(8, 5))
sns.countplot(data=df, x="OverTime", hue="Attrition")
plt.title("OverTime vs. Attrition")
plt.xlabel("Overtime")
plt.show()
```

```
[31]: # Investigate the relationship between work-life balance and attrition
plt.figure(figsize=(8, 5))
sns.countplot(data=df, x="WorkLifeBalance", hue="Attrition")
plt.title("WorkLifeBalance vs. Attrition")
plt.xlabel("Work-Life Balance")
plt.show()
```



2.3 Conclusion and Recommendations

```
[32]: # Summarize your findings
print("Summary of Findings:")
print("- Attrition Distribution:")
attrition_counts = df['Attrition'].value_counts()
print(attrition_counts)
```

```
Summary of Findings:
- Attrition Distribution:
Attrition
No      1233
Yes      237
Name: count, dtype: int64
```

```
[33]: # Provide recommendations
print("\nRecommendations:")
print("- Consider improving work-life balance to reduce attrition among_
      ↪employees.")
print("- Monitor the impact of overtime work on attrition and take necessary_
      ↪actions to manage workload.")
print("- Focus on career development opportunities, such as promotions and_
      ↪skill development, to enhance job satisfaction.")
```

```
print("- Conduct exit interviews with departing employees to gather more_
↳insights into attrition reasons.")
```

Recommendations:

- Consider improving work-life balance to reduce attrition among employees.
- Monitor the impact of overtime work on attrition and take necessary actions to manage workload.
- Focus on career development opportunities, such as promotions and skill development, to enhance job satisfaction.
- Conduct exit interviews with departing employees to gather more insights into attrition reasons.

```
[34]: # Overall Conclusion
print("\nOverall Conclusion:")
print("Based on the analysis, we have identified several factors that are_
↳related to attrition within the organization. It's important for the company_
↳to address these factors in order to improve employee retention and_
↳satisfaction. By implementing the recommended actions, the company can work_
↳towards reducing attrition and creating a more positive work environment.")
```

Overall Conclusion:

Based on the analysis, we have identified several factors that are related to attrition within the organization. It's important for the company to address these factors in order to improve employee retention and satisfaction. By implementing the recommended actions, the company can work towards reducing attrition and creating a more positive work environment.

```
[ ]:
```