

Name: Satyam Mishra Reg : 21BCE8247 Assignment 4

1. Perform Data Preprocessing
2. Model Building using Logistic Regression and Decision Tree
2. Calculate Performance metrics

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
from sklearn.preprocessing import LabelEncoder, OneHotEncoder
```

```
df = pd.read_csv("dataset/HR-Employee-Attrition.csv")
```

```
df.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
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]

df.shape

(1470, 35)

df.Attrition.value_counts()

Attrition

No 1233

Yes 237

Name: count, dtype: int64

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

df.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	...	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]

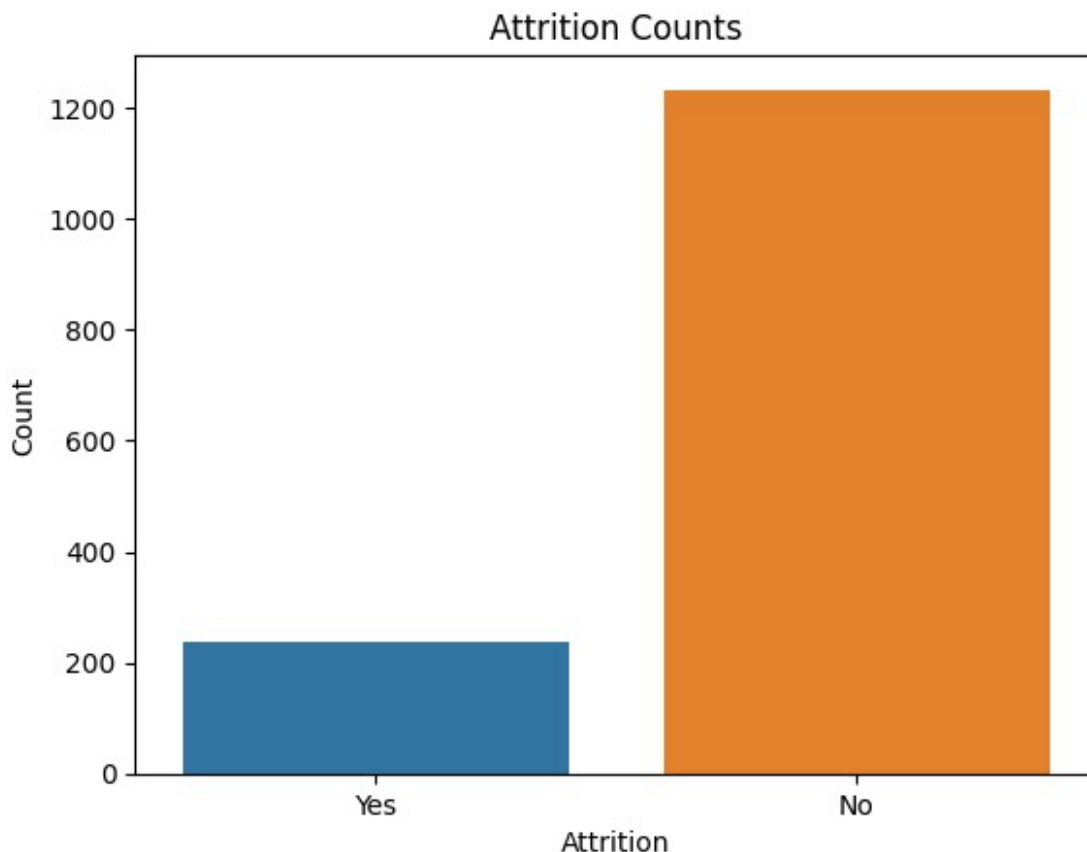
df.isnull().any()

Age	False
Attrition	False
BusinessTravel	False
DailyRate	False
Department	False
DistanceFromHome	False
Education	False
EducationField	False
EmployeeCount	False
EmployeeNumber	False
EnvironmentSatisfaction	False
Gender	False
HourlyRate	False
JobInvolvement	False
JobLevel	False
JobRole	False
JobSatisfaction	False
MaritalStatus	False
MonthlyIncome	False
MonthlyRate	False
NumCompaniesWorked	False
Over18	False
OverTime	False
PercentSalaryHike	False
PerformanceRating	False
RelationshipSatisfaction	False
StandardHours	False
StockOptionLevel	False
TotalWorkingYears	False
TrainingTimesLastYear	False
WorkLifeBalance	False
YearsAtCompany	False
YearsInCurrentRole	False
YearsSinceLastPromotion	False
YearsWithCurrManager	False

dtype: bool

```
sns.countplot(x='Attrition', data=df)
plt.title('Attrition Counts')
plt.xlabel('Attrition')
plt.ylabel('Count')
plt.show()
```

```
E:\Installed softwares\Py 3.10.7\lib\site-packages\seaborn\
_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated
and will be removed in a future version. Use isinstance(dtype,
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```



```
sns.histplot(data=df, x='Age', hue='Attrition', kde=True)
plt.title('Distribution of Age by Attrition')
plt.xlabel('Age')
plt.ylabel('Count')
plt.show()
```

```
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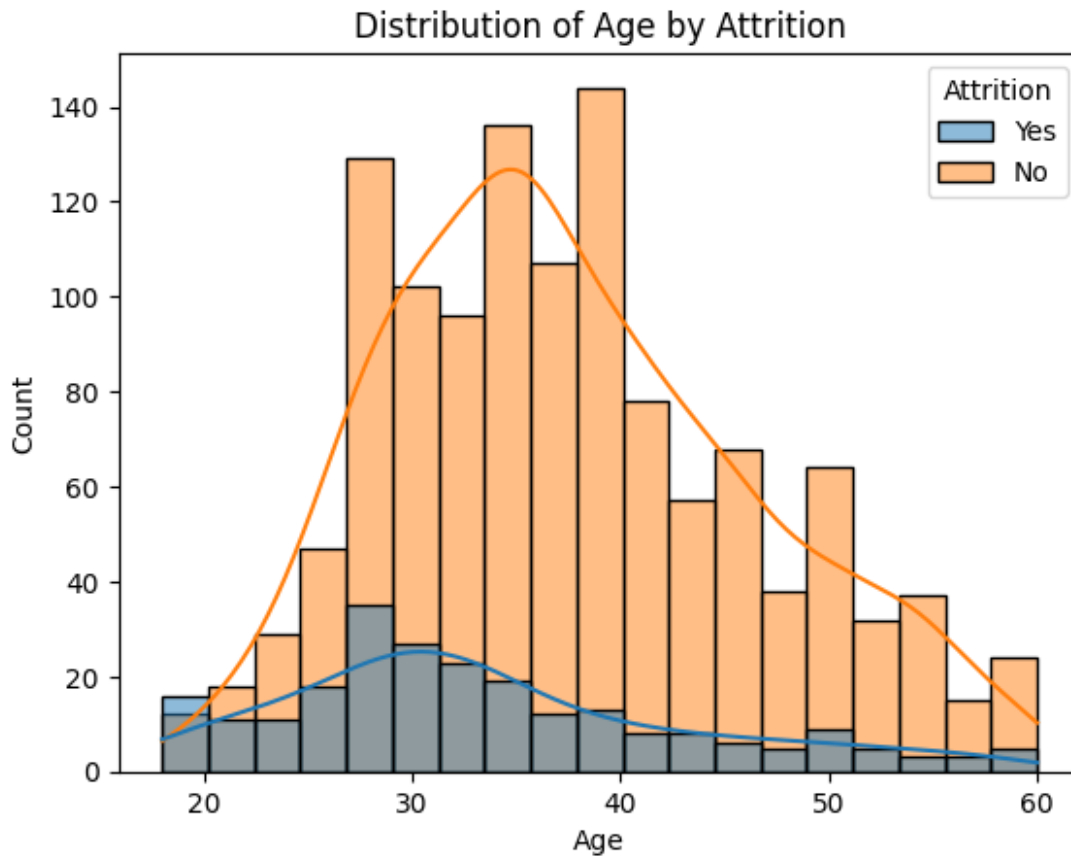
```
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_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated
and will be removed in a future version. Convert inf values to NaN
before operating instead.
```

```
    with pd.option_context('mode.use_inf_as_na', True):
```



```
sns.countplot(x='JobSatisfaction', hue='Attrition', data=df)
plt.title('Job Satisfaction vs. Attrition')
plt.xlabel('Job Satisfaction')
plt.ylabel('Count')
plt.show()
```

```
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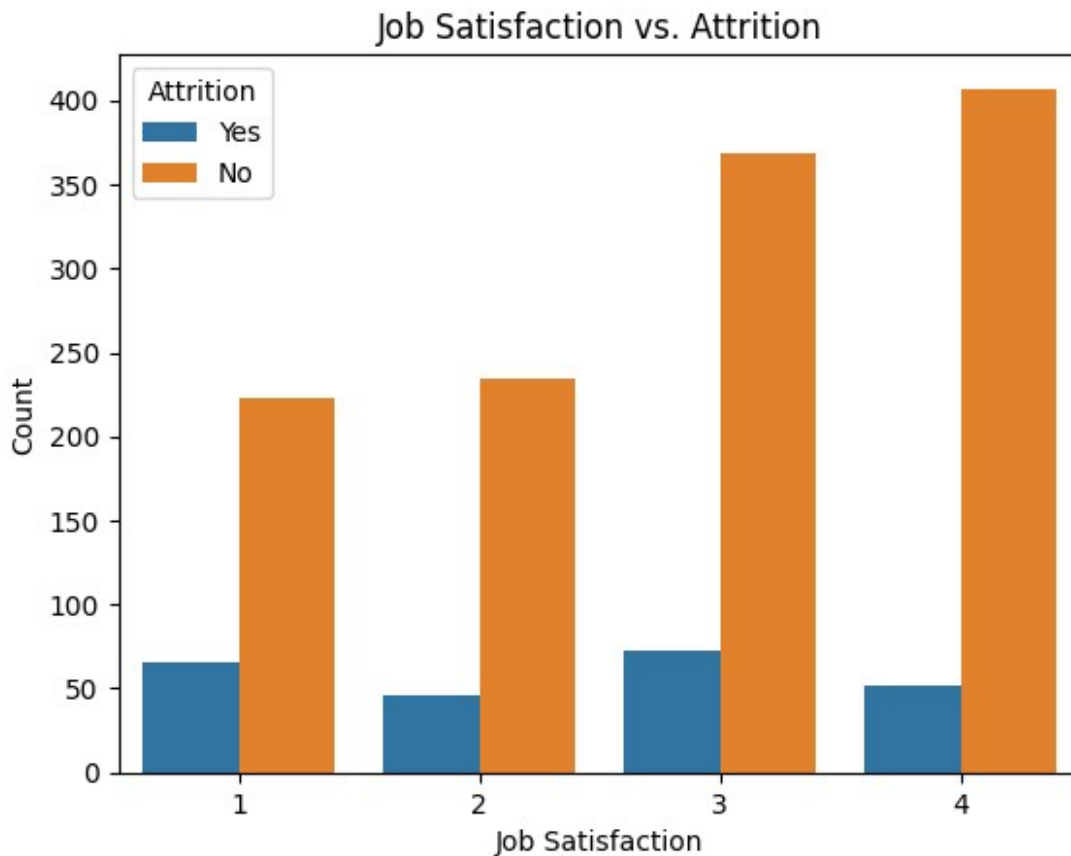
```
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if `pd.api.types.is_categorical_dtype(vector)`:



```
sns.histplot(data=df, x='YearsAtCompany', hue='Attrition', kde=True)
plt.title('Distribution of Years at Company by Attrition')
plt.xlabel('Years at Company')
plt.ylabel('Count')
plt.show()
```

```
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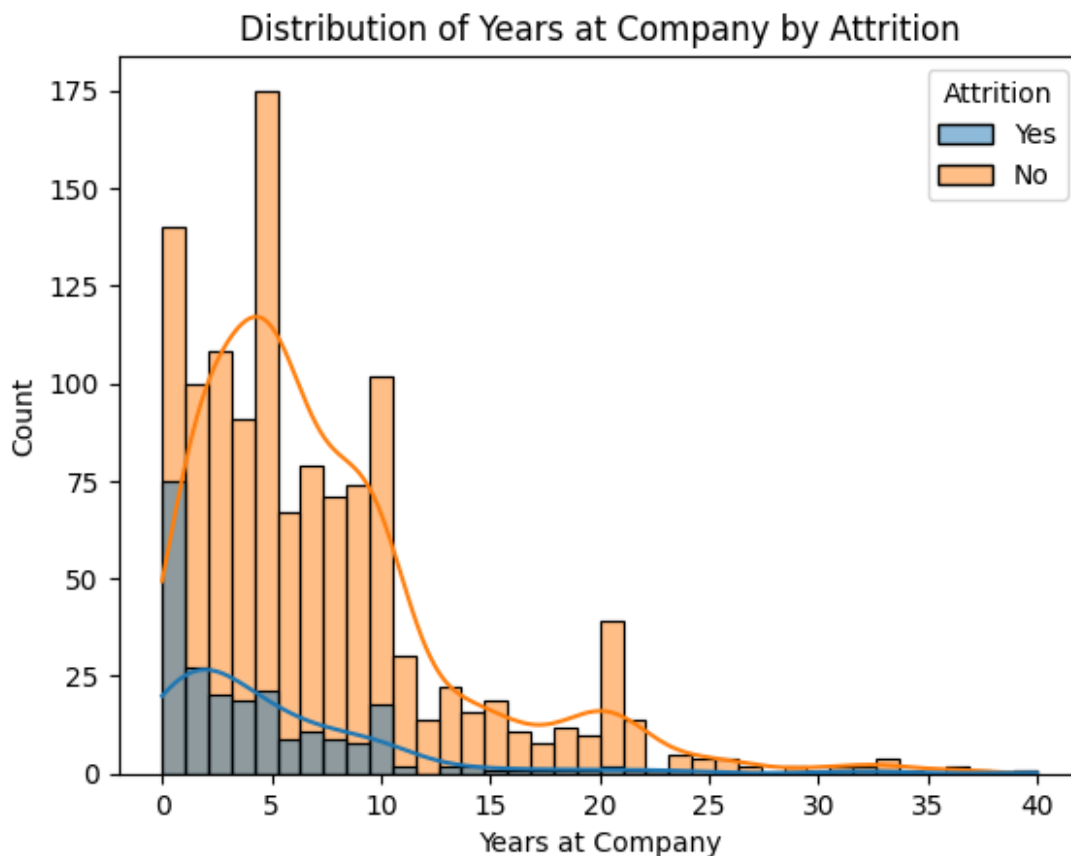
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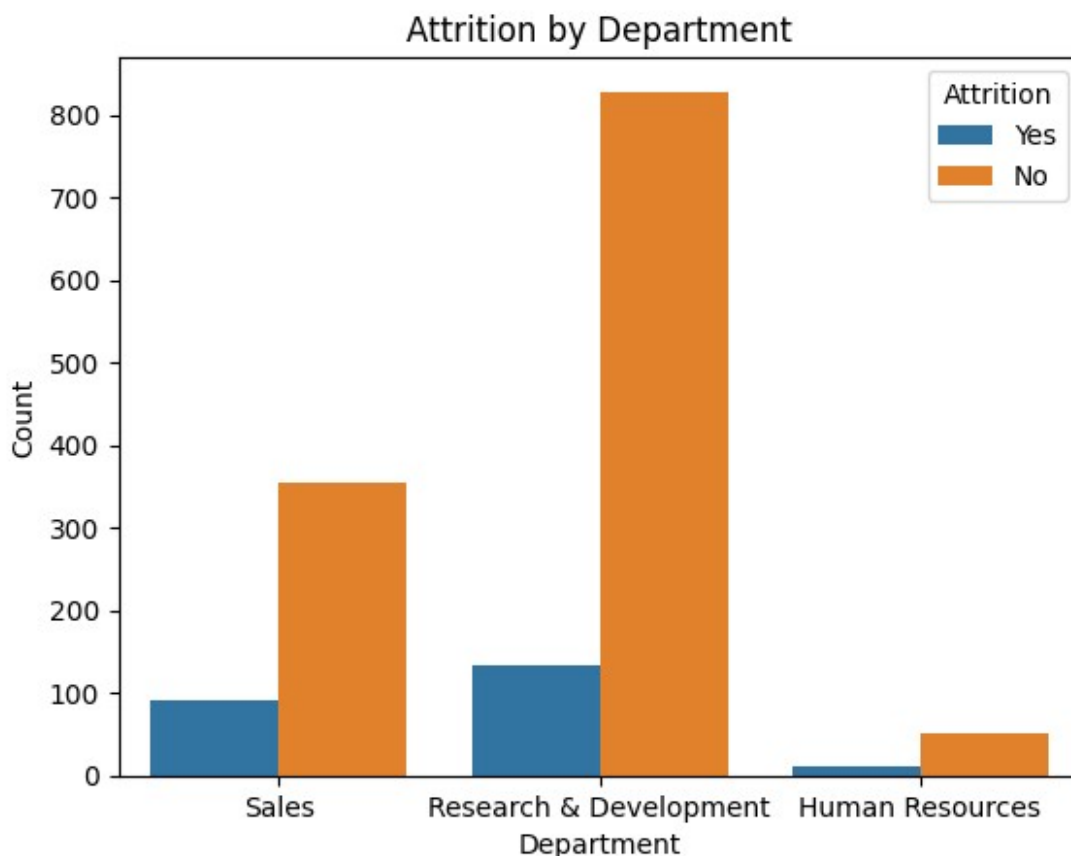
```

sns.countplot(x='Department', hue='Attrition', data=df)
plt.title('Attrition by Department')
plt.xlabel('Department')
plt.ylabel('Count')
plt.show()

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```



```
sns.countplot(x='Education', hue='Attrition', data=df)
plt.title('Attrition by Education Level')
plt.xlabel('Education Level')
```

```
plt.ylabel('Count')  
plt.show()
```

```
E:\Installed softwares\Py 3.10.7\lib\site-packages\seaborn\  
_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated  
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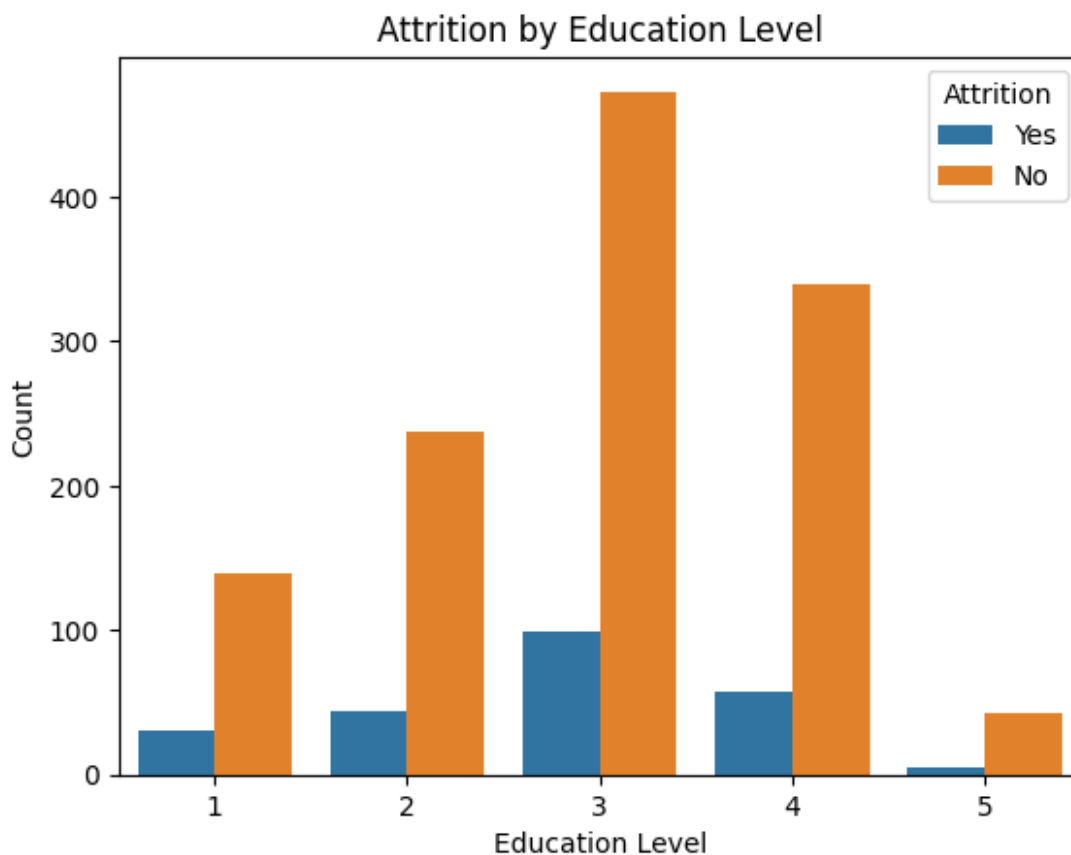
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```

```
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```



```
sns.histplot(data=df, x='MonthlyIncome', hue='Attrition', kde=True)
plt.title('Distribution of Monthly Income by Attrition')
plt.xlabel('Monthly Income')
plt.ylabel('Count')
plt.show()
```

```
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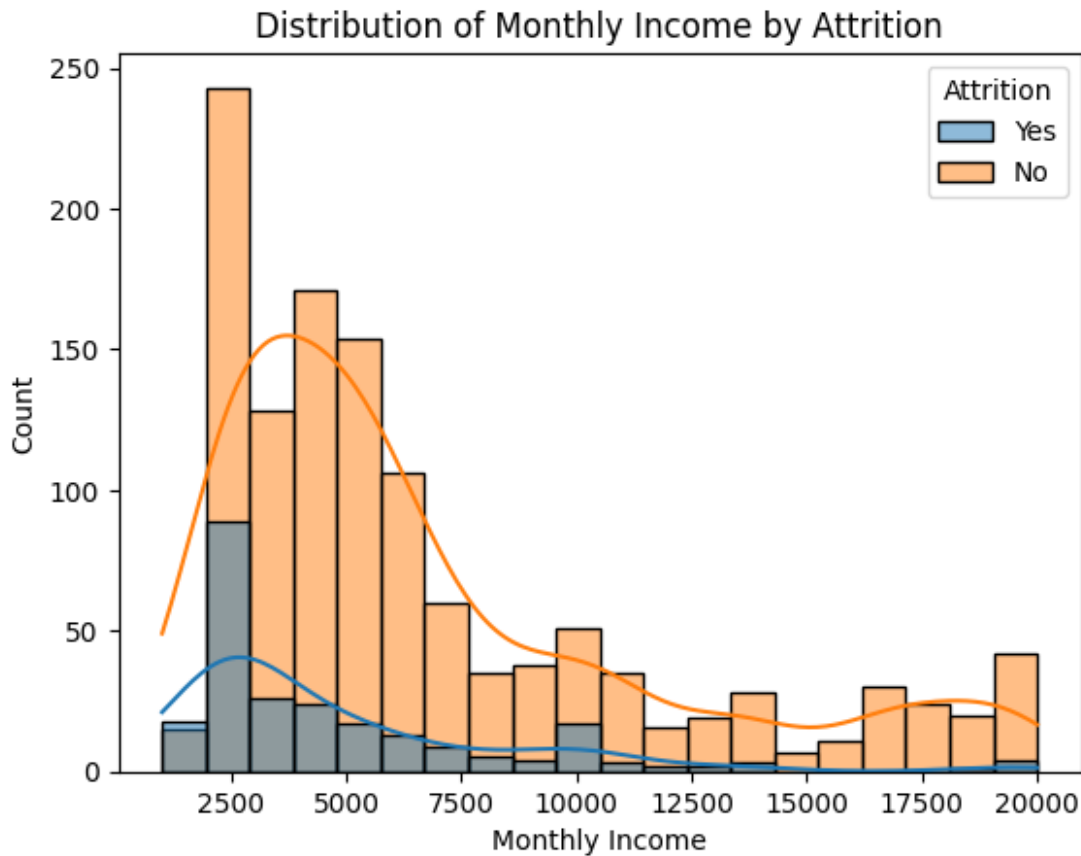
```
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```

```
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```



```
sns.countplot(x='WorkLifeBalance', hue='Attrition', data=df)
plt.title('Attrition by Work-Life Balance')
plt.xlabel('Work-Life Balance')
plt.ylabel('Count')
plt.show()
```

```
E:\Installed softwares\Py 3.10.7\lib\site-packages\seaborn\
_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated
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```
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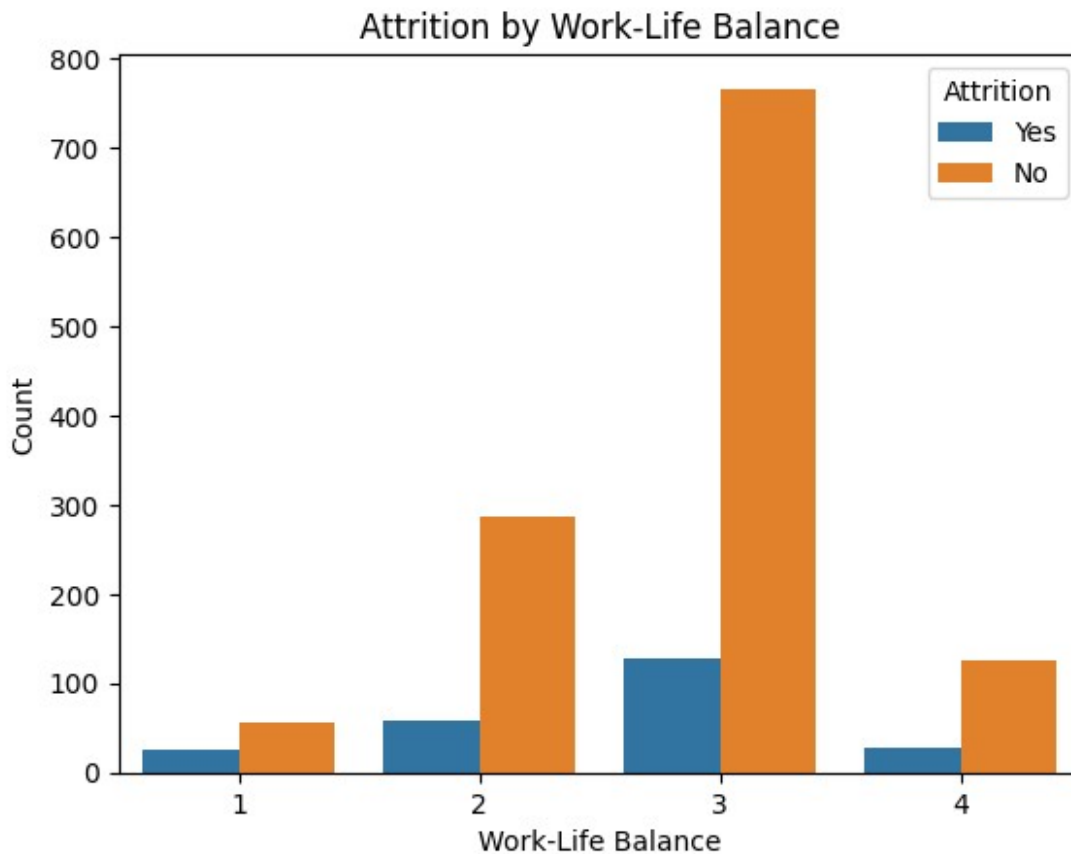
```
E:\Installed softwares\Py 3.10.7\lib\site-packages\seaborn\
_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated
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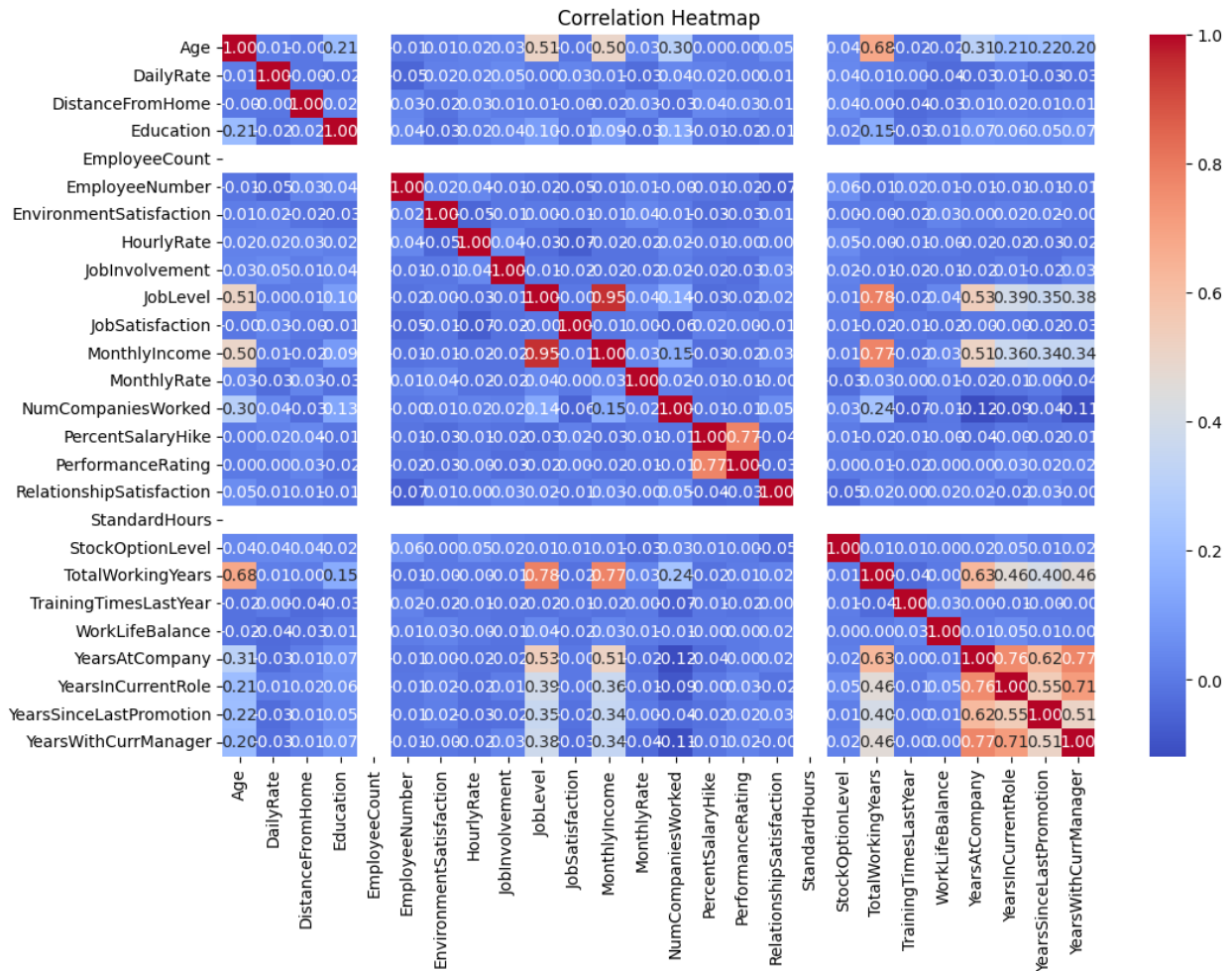
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E:\Installed softwares\Py 3.10.7\lib\site-packages\seaborn\
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```



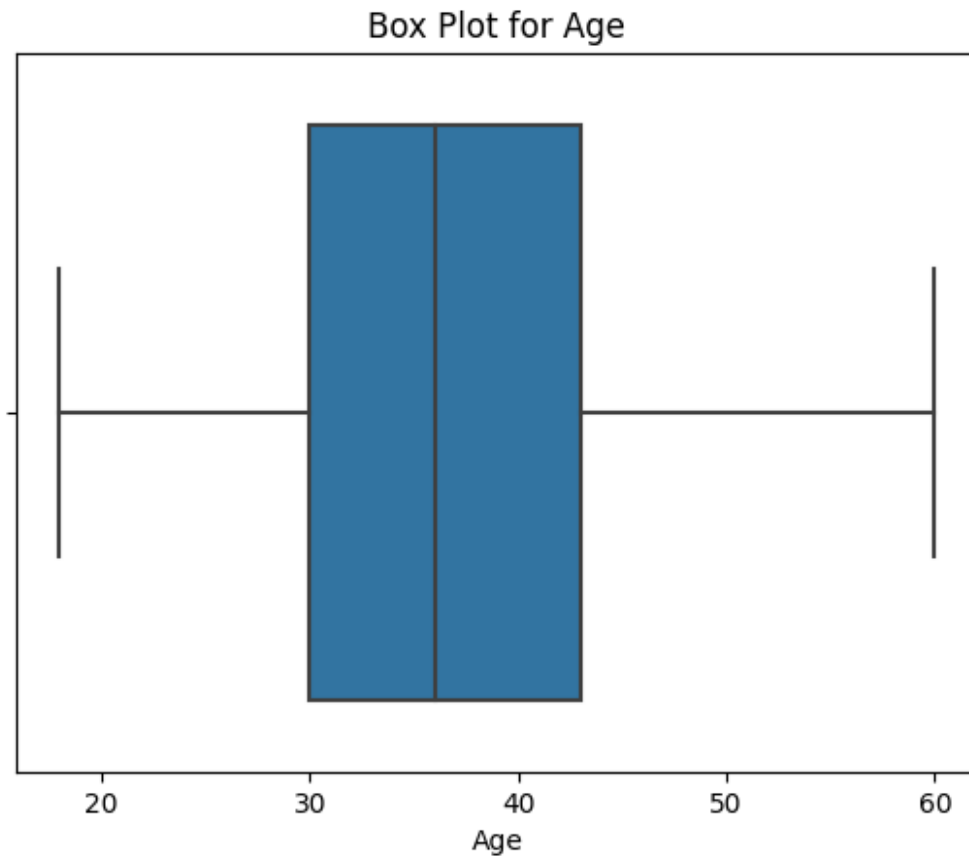
```
correlation_matrix = df.corr(numeric_only=True)
plt.figure(figsize=(12, 8))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm',
            fmt=".2f")
plt.title('Correlation Heatmap')
plt.show()
```



```
sns.boxplot(x='Age', data=df)
plt.title('Box Plot for Age')
plt.show()
```

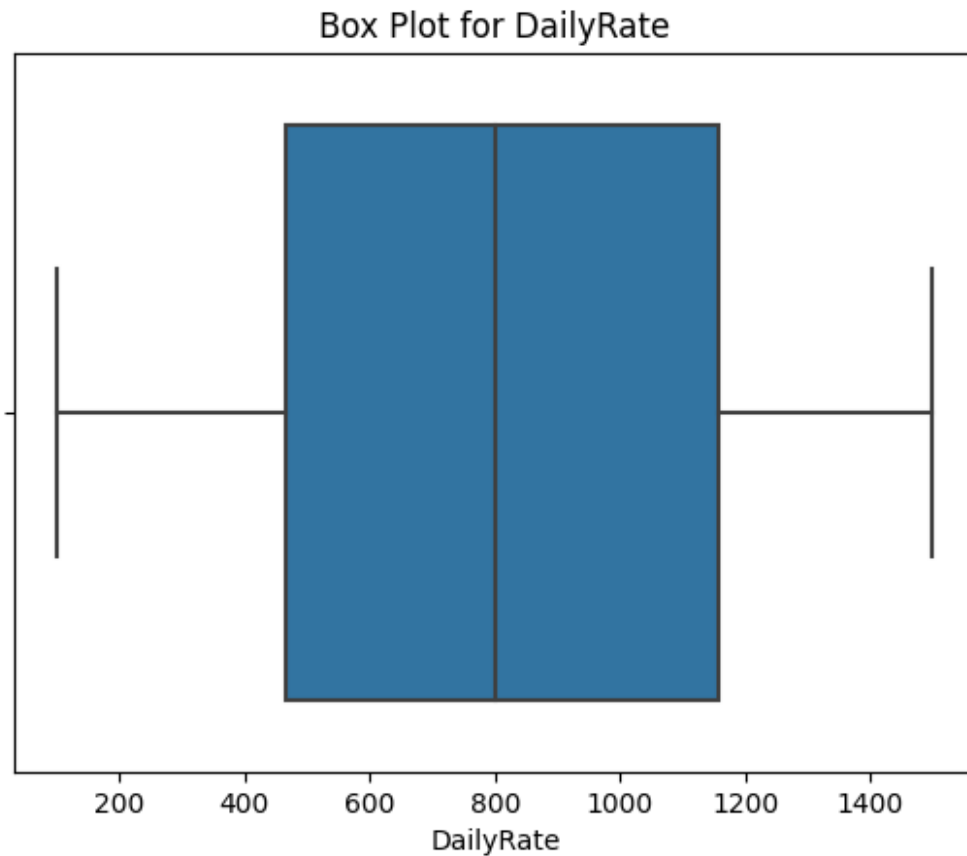
E:\Installed softwares\Py 3.10.7\lib\site-packages\seaborn_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated and will be removed in a future version. Use isinstance(dtype, CategoricalDtype) instead

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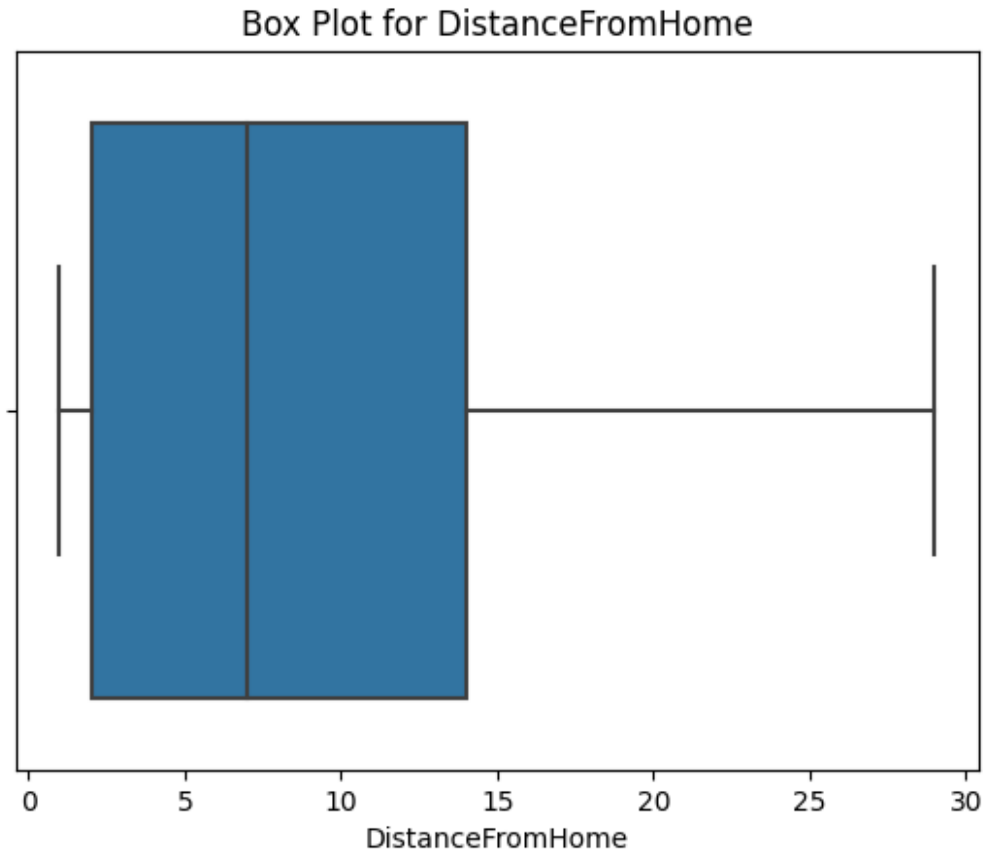
```
sns.boxplot(x='DailyRate', data=df)
plt.title('Box Plot for DailyRate')
plt.show()
```

```
E:\Installed softwares\Py 3.10.7\lib\site-packages\seaborn\
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and will be removed in a future version. Use isinstance(dtype,
CategoricalDtype) instead
  if pd.api.types.is_categorical_dtype(vector):
```



```
sns.boxplot(x='DistanceFromHome', data=df)
plt.title('Box Plot for DistanceFromHome')
plt.show()
```

```
E:\Installed softwares\Py 3.10.7\lib\site-packages\seaborn\
_oldcore.py:1498: FutureWarning: is_categorical_dtype is deprecated
and will be removed in a future version. Use isinstance(dtype,
CategoricalDtype) instead
  if pd.api.types.is_categorical_dtype(vector):
```



```
# dependent variables
```

```
y = df['Attrition']
```

```
# independent variables (features)
```

```
X = df.drop('Attrition', axis=1)
```

```
# one-hot encoding on categorical columns
```

```
X_encoded = pd.get_dummies(X, drop_first=True)
```

```
X_encoded
```

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount	\
0	41	1102	1	2	1	
1	49	279	8	1	1	
2	37	1373	2	2	1	
3	33	1392	3	4	1	
4	27	591	2	1	1	
...	
1465	36	884	23	2	1	
1466	39	613	6	1	1	
1467	27	155	4	3	1	
1468	49	1023	2	3	1	

1469	34	628	8	3	1
EmployeeNumber		EnvironmentSatisfaction	HourlyRate		
JobInvolvement \					
0	1	2	94		
3					
1	2	3	61		
2					
2	4	4	92		
2					
3	5	4	56		
3					
4	7	1	40		
3					
...		
...					
1465	2061	3	41		
4					
1466	2062	4	42		
2					
1467	2064	2	87		
4					
1468	2065	4	63		
2					
1469	2068	2	82		
4					
JobLevel		...	JobRole_Laboratory Technician	JobRole_Manager	\
0	2	...	False	False	
1	2	...	False	False	
2	1	...	True	False	
3	1	...	False	False	
4	1	...	True	False	
...	
1465	2	...	True	False	
1466	3	...	False	False	
1467	2	...	False	False	
1468	2	...	False	False	
1469	2	...	True	False	
JobRole_Manufacturing Director		JobRole_Research Director	\		
0		False	False		
1		False	False		
2		False	False		
3		False	False		
4		False	False		
...			
1465		False	False		
1466		False	False		
1467		True	False		

1468	False	False
1469	False	False

	JobRole_Research Scientist	JobRole_Sales Executive	\
0	False	True	
1	True	False	
2	False	False	
3	True	False	
4	False	False	
...	
1465	False	False	
1466	False	False	
1467	False	False	
1468	False	True	
1469	False	False	

	JobRole_Sales Representative	MaritalStatus_Married	\
0	False	False	
1	False	True	
2	False	False	
3	False	True	
4	False	True	
...	
1465	False	True	
1466	False	True	
1467	False	True	
1468	False	True	
1469	False	True	

	MaritalStatus_Single	OverTime_Yes
0	True	True
1	False	False
2	True	True
3	False	True
4	False	False
...
1465	False	False
1466	False	False
1467	False	True
1468	False	False
1469	False	False

[1470 rows x 47 columns]

```
from sklearn.preprocessing import MinMaxScaler
```

```
scaler = MinMaxScaler()
```

```
X_scaled = scaler.fit_transform(X_encoded)
```

```
# scaled features to DataFrame
```

```
X_scaled_df = pd.DataFrame(X_scaled, columns=X_encoded.columns)
```

```
X_scaled_df
```

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount
\					
0	0.547619	0.715820	0.000000	0.25	0.0
1	0.738095	0.126700	0.250000	0.00	0.0
2	0.452381	0.909807	0.035714	0.25	0.0
3	0.357143	0.923407	0.071429	0.75	0.0
4	0.214286	0.350036	0.035714	0.00	0.0
...
1465	0.428571	0.559771	0.785714	0.25	0.0
1466	0.500000	0.365784	0.178571	0.00	0.0
1467	0.214286	0.037938	0.107143	0.50	0.0
1468	0.738095	0.659270	0.035714	0.50	0.0
1469	0.380952	0.376521	0.250000	0.50	0.0

	EmployeeNumber	EnvironmentSatisfaction	HourlyRate
JobInvolvement \			
0	0.000000	0.333333	0.914286
0.666667			
1	0.000484	0.666667	0.442857
0.333333			
2	0.001451	1.000000	0.885714
0.333333			
3	0.001935	1.000000	0.371429
0.666667			
4	0.002903	0.000000	0.142857
0.666667			
...
...			
1465	0.996613	0.666667	0.157143
1.000000			
1466	0.997097	1.000000	0.171429
0.333333			
1467	0.998065	0.333333	0.814286
1.000000			
1468	0.998549	1.000000	0.471429

0.333333

1469 1.000000

0.333333

0.742857

1.000000

	JobLevel	...	JobRole_Laboratory Technician	JobRole_Manager	\
0	0.25	...	0.0	0.0	
1	0.25	...	0.0	0.0	
2	0.00	...	1.0	0.0	
3	0.00	...	0.0	0.0	
4	0.00	...	1.0	0.0	
...	
1465	0.25	...	1.0	0.0	
1466	0.50	...	0.0	0.0	
1467	0.25	...	0.0	0.0	
1468	0.25	...	0.0	0.0	
1469	0.25	...	1.0	0.0	

	JobRole_Manufacturing Director	JobRole_Research Director	\
0	0.0	0.0	
1	0.0	0.0	
2	0.0	0.0	
3	0.0	0.0	
4	0.0	0.0	
...	
1465	0.0	0.0	
1466	0.0	0.0	
1467	1.0	0.0	
1468	0.0	0.0	
1469	0.0	0.0	

	JobRole_Research Scientist	JobRole_Sales Executive	\
0	0.0	1.0	
1	1.0	0.0	
2	0.0	0.0	
3	1.0	0.0	
4	0.0	0.0	
...	
1465	0.0	0.0	
1466	0.0	0.0	
1467	0.0	0.0	
1468	0.0	1.0	
1469	0.0	0.0	

	JobRole_Sales Representative	MaritalStatus_Married	\
0	0.0	0.0	
1	0.0	1.0	
2	0.0	0.0	
3	0.0	1.0	
4	0.0	1.0	
...	

1465	0.0	1.0
1466	0.0	1.0
1467	0.0	1.0
1468	0.0	1.0
1469	0.0	1.0

	MaritalStatus_Single	OverTime_Yes
0	1.0	1.0
1	0.0	0.0
2	1.0	1.0
3	0.0	1.0
4	0.0	0.0
...
1465	0.0	0.0
1466	0.0	0.0
1467	0.0	1.0
1468	0.0	0.0
1469	0.0	0.0

[1470 rows x 47 columns]

```

from sklearn.model_selection import train_test_split

# data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X_scaled, y,
test_size=0.2, random_state=0)

from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report

logistic_model = LogisticRegression(random_state=0)
decision_tree_model = DecisionTreeClassifier(random_state=0)

# Logistic Regression model
logistic_model.fit(X_train, y_train)
logistic_predictions = logistic_model.predict(X_test)

# Decision Tree model
decision_tree_model.fit(X_train, y_train)
decision_tree_predictions = decision_tree_model.predict(X_test)

# Evaluation of Logistic Regression model
logistic_accuracy = accuracy_score(y_test, logistic_predictions)
logistic_report = classification_report(y_test, logistic_predictions)

print("Logistic Regression Model Accuracy:", logistic_accuracy)
print("Logistic Regression Model Classification Report:")
print(logistic_report)

```


Logistic Regression Model Accuracy: 0.8843537414965986

Logistic Regression Model Classification Report:

	precision	recall	f1-score	support
No	0.89	0.98	0.93	245
Yes	0.80	0.41	0.54	49
accuracy			0.88	294
macro avg	0.85	0.69	0.74	294
weighted avg	0.88	0.88	0.87	294

Evaluation of Decision Tree model

```
decision_tree_accuracy = accuracy_score(y_test,  
decision_tree_predictions)
```

```
decision_tree_report = classification_report(y_test,  
decision_tree_predictions)
```

```
print("Decision Tree Model Accuracy:", decision_tree_accuracy)
```

```
print("Decision Tree Model Classification Report:")
```

```
print(decision_tree_report)
```

Decision Tree Model Accuracy: 0.7891156462585034

Decision Tree Model Classification Report:

	precision	recall	f1-score	support
No	0.87	0.88	0.87	245
Yes	0.36	0.33	0.34	49
accuracy			0.79	294
macro avg	0.61	0.60	0.61	294
weighted avg	0.78	0.79	0.79	294