

```
import numpy as np
import pandas as pd
import seaborn as sns
```

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

	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					
...	
...					
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	\
0	1	2	Life Sciences	1	
1	8	1	Life Sciences	1	
2	2	2	Other	1	
3	3	4	Life Sciences	1	
4	2	1	Medical	1	
...	
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	\
0	1	...	1	80	
1	2	...	4	80	
2	4	...	2	80	
3	5	...	3	80	

4	7	...	4	80
...
1465	2061	...	3	80
1466	2062	...	1	80
1467	2064	...	2	80
1468	2065	...	4	80
1469	2068	...	1	80

	StockOptionLevel	TotalWorkingYears	TrainingTimesLastYear	\
0	0	8	0	
1	1	10	3	
2	0	7	3	
3	0	8	3	
4	1	6	3	
...	
1465	1	17	3	
1466	1	9	5	
1467	1	6	0	
1468	0	17	3	
1469	0	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	
...	
1465	3	5	2	
1466	3	7	7	
1467	3	6	2	
1468	2	9	6	
1469	4	4	3	

	YearsSinceLastPromotion	YearsWithCurrManager
0	0	5
1	1	7
2	0	0
3	3	0
4	2	2
...
1465	0	3
1466	1	7
1467	0	3
1468	0	8
1469	1	2

[1470 rows x 35 columns]

df.describe()

		Age	DailyRate	DistanceFromHome	Education
EmployeeCount \					
count	1470.000000	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	
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	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.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.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

df.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
```

```
corr = df.corr()
corr
```

```
C:\Users\hp\AppData\Local\Temp\ipykernel_12144\2438084875.py:1:
FutureWarning: The default value of numeric_only in DataFrame.corr is
deprecated. In a future version, it will default to False. Select only
valid columns or specify the value of numeric_only to silence this
warning.
```

```
    corr = df.corr()
```

	Age	DailyRate	DistanceFromHome
Education \			
Age	1.000000	0.010661	-0.001686
0.208034			
DailyRate	0.010661	1.000000	-0.004985
0.016806			
DistanceFromHome	-0.001686	-0.004985	1.000000
0.021042			
Education	0.208034	-0.016806	0.021042
1.000000			
EmployeeCount	NaN	NaN	NaN
NaN			
EmployeeNumber	-0.010145	-0.050990	0.032916
0.042070			
EnvironmentSatisfaction	0.010146	0.018355	-0.016075
0.027128			
HourlyRate	0.024287	0.023381	0.031131
0.016775			
JobInvolvement	0.029820	0.046135	0.008783
0.042438			
JobLevel	0.509604	0.002966	0.005303
0.101589			
JobSatisfaction	-0.004892	0.030571	-0.003669
0.011296			
MonthlyIncome	0.497855	0.007707	-0.017014
0.094961			
MonthlyRate	0.028051	-0.032182	0.027473
0.026084			
NumCompaniesWorked	0.299635	0.038153	-0.029251
0.126317			
PercentSalaryHike	0.003634	0.022704	0.040235
0.011111			
PerformanceRating	0.001904	0.000473	0.027110
0.024539			
RelationshipSatisfaction	0.053535	0.007846	0.006557
0.009118			
StandardHours	NaN	NaN	NaN

NaN			
StockOptionLevel	0.037510	0.042143	0.044872
0.018422			
TotalWorkingYears	0.680381	0.014515	0.004628
0.148280			
TrainingTimesLastYear	-0.019621	0.002453	-0.036942
0.025100			
WorkLifeBalance	-0.021490	-0.037848	-0.026556
0.009819			
YearsAtCompany	0.311309	-0.034055	0.009508
0.069114			
YearsInCurrentRole	0.212901	0.009932	0.018845
0.060236			
YearsSinceLastPromotion	0.216513	-0.033229	0.010029
0.054254			
YearsWithCurrManager	0.202089	-0.026363	0.014406
0.069065			

	EmployeeCount	EmployeeNumber \
Age	NaN	-0.010145
DailyRate	NaN	-0.050990
DistanceFromHome	NaN	0.032916
Education	NaN	0.042070
EmployeeCount	NaN	NaN
EmployeeNumber	NaN	1.000000
EnvironmentSatisfaction	NaN	0.017621
HourlyRate	NaN	0.035179
JobInvolvement	NaN	-0.006888
JobLevel	NaN	-0.018519
JobSatisfaction	NaN	-0.046247
MonthlyIncome	NaN	-0.014829
MonthlyRate	NaN	0.012648
NumCompaniesWorked	NaN	-0.001251
PercentSalaryHike	NaN	-0.012944
PerformanceRating	NaN	-0.020359
RelationshipSatisfaction	NaN	-0.069861
StandardHours	NaN	NaN
StockOptionLevel	NaN	0.062227
TotalWorkingYears	NaN	-0.014365
TrainingTimesLastYear	NaN	0.023603
WorkLifeBalance	NaN	0.010309
YearsAtCompany	NaN	-0.011240
YearsInCurrentRole	NaN	-0.008416
YearsSinceLastPromotion	NaN	-0.009019
YearsWithCurrManager	NaN	-0.009197

	EnvironmentSatisfaction	HourlyRate
JobInvolvement \		
Age	0.010146	0.024287

0.029820			
DailyRate	0.018355	0.023381	
0.046135			
DistanceFromHome	-0.016075	0.031131	
0.008783			
Education	-0.027128	0.016775	
0.042438			
EmployeeCount	NaN	NaN	
NaN			
EmployeeNumber	0.017621	0.035179	-
0.006888			
EnvironmentSatisfaction	1.000000	-0.049857	-
0.008278			
HourlyRate	-0.049857	1.000000	
0.042861			
JobInvolvement	-0.008278	0.042861	
1.000000			
JobLevel	0.001212	-0.027853	-
0.012630			
JobSatisfaction	-0.006784	-0.071335	-
0.021476			
MonthlyIncome	-0.006259	-0.015794	-
0.015271			
MonthlyRate	0.037600	-0.015297	-
0.016322			
NumCompaniesWorked	0.012594	0.022157	
0.015012			
PercentSalaryHike	-0.031701	-0.009062	-
0.017205			
PerformanceRating	-0.029548	-0.002172	-
0.029071			
RelationshipSatisfaction	0.007665	0.001330	
0.034297			
StandardHours	NaN	NaN	
NaN			
StockOptionLevel	0.003432	0.050263	
0.021523			
TotalWorkingYears	-0.002693	-0.002334	-
0.005533			
TrainingTimesLastYear	-0.019359	-0.008548	-
0.015338			
WorkLifeBalance	0.027627	-0.004607	-
0.014617			
YearsAtCompany	0.001458	-0.019582	-
0.021355			
YearsInCurrentRole	0.018007	-0.024106	
0.008717			
YearsSinceLastPromotion	0.016194	-0.026716	-
0.024184			

YearsWithCurrManager
0.025976

-0.004999 -0.020123

	JobLevel	...	RelationshipSatisfaction	\
Age	0.509604	...	0.053535	
DailyRate	0.002966	...	0.007846	
DistanceFromHome	0.005303	...	0.006557	
Education	0.101589	...	-0.009118	
EmployeeCount	NaN	...	NaN	
EmployeeNumber	-0.018519	...	-0.069861	
EnvironmentSatisfaction	0.001212	...	0.007665	
HourlyRate	-0.027853	...	0.001330	
JobInvolvement	-0.012630	...	0.034297	
JobLevel	1.000000	...	0.021642	
JobSatisfaction	-0.001944	...	-0.012454	
MonthlyIncome	0.950300	...	0.025873	
MonthlyRate	0.039563	...	-0.004085	
NumCompaniesWorked	0.142501	...	0.052733	
PercentSalaryHike	-0.034730	...	-0.040490	
PerformanceRating	-0.021222	...	-0.031351	
RelationshipSatisfaction	0.021642	...	1.000000	
StandardHours	NaN	...	NaN	
StockOptionLevel	0.013984	...	-0.045952	
TotalWorkingYears	0.782208	...	0.024054	
TrainingTimesLastYear	-0.018191	...	0.002497	
WorkLifeBalance	0.037818	...	0.019604	
YearsAtCompany	0.534739	...	0.019367	
YearsInCurrentRole	0.389447	...	-0.015123	
YearsSinceLastPromotion	0.353885	...	0.033493	
YearsWithCurrManager	0.375281	...	-0.000867	

	StandardHours	StockOptionLevel	
TotalWorkingYears \			
Age	NaN	0.037510	
0.680381			
DailyRate	NaN	0.042143	
0.014515			
DistanceFromHome	NaN	0.044872	
0.004628			
Education	NaN	0.018422	
0.148280			
EmployeeCount	NaN	NaN	
NaN			
EmployeeNumber	NaN	0.062227	-
0.014365			
EnvironmentSatisfaction	NaN	0.003432	-
0.002693			
HourlyRate	NaN	0.050263	-
0.002334			

JobInvolvement	NaN	0.021523	-
0.005533			
JobLevel	NaN	0.013984	
0.782208			
JobSatisfaction	NaN	0.010690	-
0.020185			
MonthlyIncome	NaN	0.005408	
0.772893			
MonthlyRate	NaN	-0.034323	
0.026442			
NumCompaniesWorked	NaN	0.030075	
0.237639			
PercentSalaryHike	NaN	0.007528	-
0.020608			
PerformanceRating	NaN	0.003506	
0.006744			
RelationshipSatisfaction	NaN	-0.045952	
0.024054			
StandardHours	NaN	NaN	
NaN			
StockOptionLevel	NaN	1.000000	
0.010136			
TotalWorkingYears	NaN	0.010136	
1.000000			
TrainingTimesLastYear	NaN	0.011274	-
0.035662			
WorkLifeBalance	NaN	0.004129	
0.001008			
YearsAtCompany	NaN	0.015058	
0.628133			
YearsInCurrentRole	NaN	0.050818	
0.460365			
YearsSinceLastPromotion	NaN	0.014352	
0.404858			
YearsWithCurrManager	NaN	0.024698	
0.459188			
	TrainingTimesLastYear	WorkLifeBalance	\
Age	-0.019621	-0.021490	
DailyRate	0.002453	-0.037848	
DistanceFromHome	-0.036942	-0.026556	
Education	-0.025100	0.009819	
EmployeeCount	NaN	NaN	
EmployeeNumber	0.023603	0.010309	
EnvironmentSatisfaction	-0.019359	0.027627	
HourlyRate	-0.008548	-0.004607	
JobInvolvement	-0.015338	-0.014617	
JobLevel	-0.018191	0.037818	
JobSatisfaction	-0.005779	-0.019459	

MonthlyIncome	-0.021736	0.030683
MonthlyRate	0.001467	0.007963
NumCompaniesWorked	-0.066054	-0.008366
PercentSalaryHike	-0.005221	-0.003280
PerformanceRating	-0.015579	0.002572
RelationshipSatisfaction	0.002497	0.019604
StandardHours	NaN	NaN
StockOptionLevel	0.011274	0.004129
TotalWorkingYears	-0.035662	0.001008
TrainingTimesLastYear	1.000000	0.028072
WorkLifeBalance	0.028072	1.000000
YearsAtCompany	0.003569	0.012089
YearsInCurrentRole	-0.005738	0.049856
YearsSinceLastPromotion	-0.002067	0.008941
YearsWithCurrManager	-0.004096	0.002759

	YearsAtCompany	YearsInCurrentRole	\
Age	0.311309	0.212901	
DailyRate	-0.034055	0.009932	
DistanceFromHome	0.009508	0.018845	
Education	0.069114	0.060236	
EmployeeCount	NaN	NaN	
EmployeeNumber	-0.011240	-0.008416	
EnvironmentSatisfaction	0.001458	0.018007	
HourlyRate	-0.019582	-0.024106	
JobInvolvement	-0.021355	0.008717	
JobLevel	0.534739	0.389447	
JobSatisfaction	-0.003803	-0.002305	
MonthlyIncome	0.514285	0.363818	
MonthlyRate	-0.023655	-0.012815	
NumCompaniesWorked	-0.118421	-0.090754	
PercentSalaryHike	-0.035991	-0.001520	
PerformanceRating	0.003435	0.034986	
RelationshipSatisfaction	0.019367	-0.015123	
StandardHours	NaN	NaN	
StockOptionLevel	0.015058	0.050818	
TotalWorkingYears	0.628133	0.460365	
TrainingTimesLastYear	0.003569	-0.005738	
WorkLifeBalance	0.012089	0.049856	
YearsAtCompany	1.000000	0.758754	
YearsInCurrentRole	0.758754	1.000000	
YearsSinceLastPromotion	0.618409	0.548056	
YearsWithCurrManager	0.769212	0.714365	

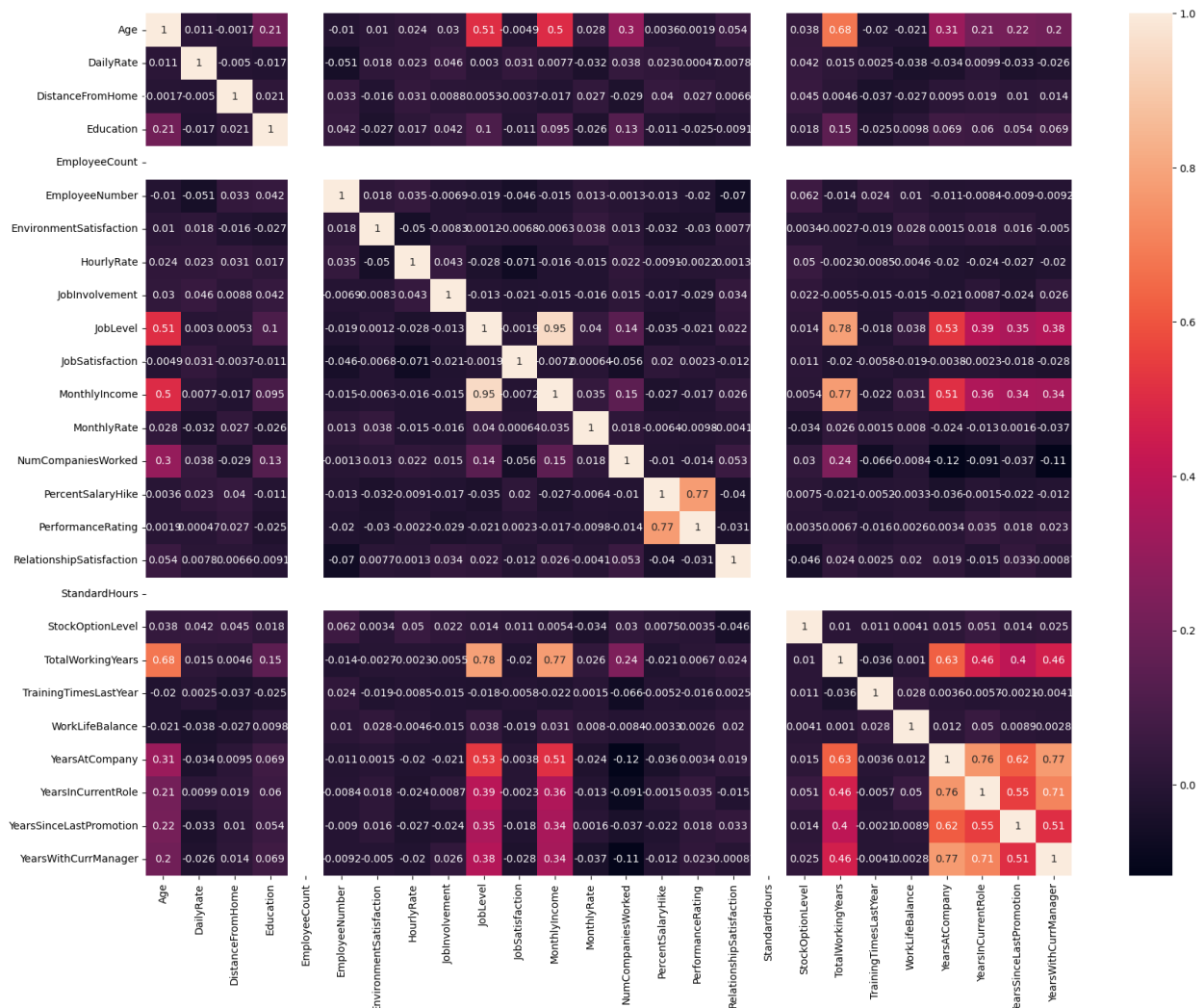
	YearsSinceLastPromotion	
YearsWithCurrManager		
Age	0.216513	
0.202089		
DailyRate	-0.033229	-

0.026363		
DistanceFromHome	0.010029	
0.014406		
Education	0.054254	
0.069065		
EmployeeCount	NaN	
NaN		
EmployeeNumber	-0.009019	-
0.009197		
EnvironmentSatisfaction	0.016194	-
0.004999		
HourlyRate	-0.026716	-
0.020123		
JobInvolvement	-0.024184	
0.025976		
JobLevel	0.353885	
0.375281		
JobSatisfaction	-0.018214	-
0.027656		
MonthlyIncome	0.344978	
0.344079		
MonthlyRate	0.001567	-
0.036746		
NumCompaniesWorked	-0.036814	-
0.110319		
PercentSalaryHike	-0.022154	-
0.011985		
PerformanceRating	0.017896	
0.022827		
RelationshipSatisfaction	0.033493	-
0.000867		
StandardHours	NaN	
NaN		
StockOptionLevel	0.014352	
0.024698		
TotalWorkingYears	0.404858	
0.459188		
TrainingTimesLastYear	-0.002067	-
0.004096		
WorkLifeBalance	0.008941	
0.002759		
YearsAtCompany	0.618409	
0.769212		
YearsInCurrentRole	0.548056	
0.714365		
YearsSinceLastPromotion	1.000000	
0.510224		
YearsWithCurrManager	0.510224	
1.000000		

[26 rows x 26 columns]

```
import matplotlib.pyplot as plt
plt.subplots(figsize=(20,15))
sns.heatmap(corr,annot=True)
```

<Axes: >



```
df['BusinessTravel'].value_counts()
```

```
Travel_Rarely      1043
Travel_Frequently   277
Non-Travel         150
Name: BusinessTravel, dtype: int64
```

```
df['Department'].value_counts()
```

```
Research & Development    961
Sales                     446
Human Resources           63
Name: Department, dtype: int64
```

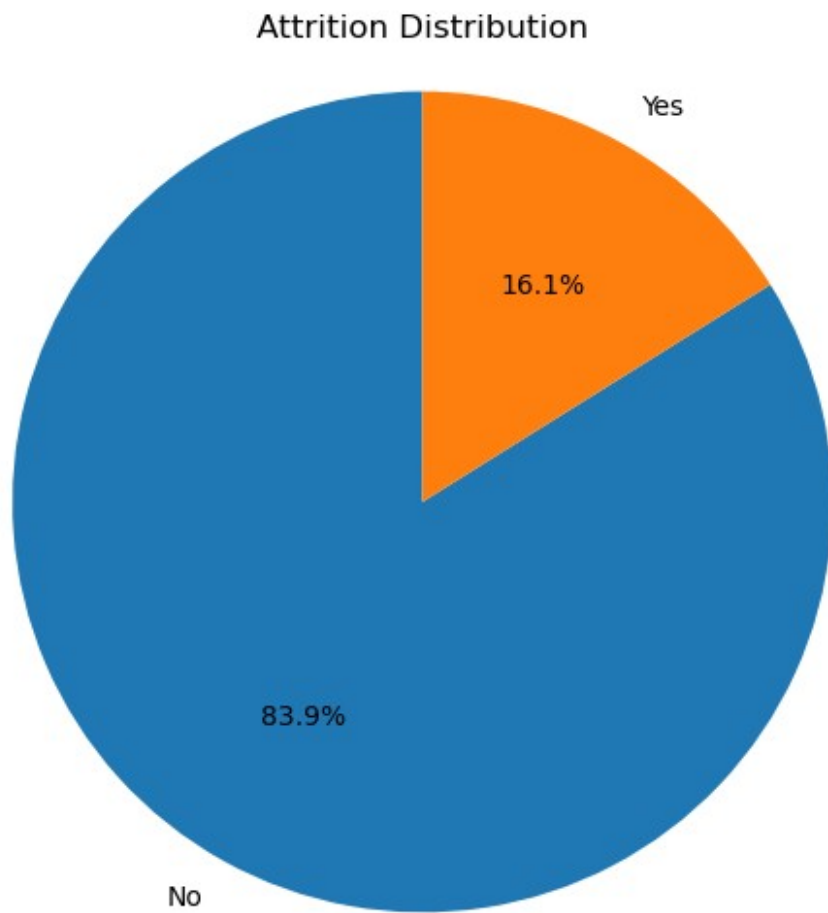
```
df['EducationField'].value_counts()
```

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

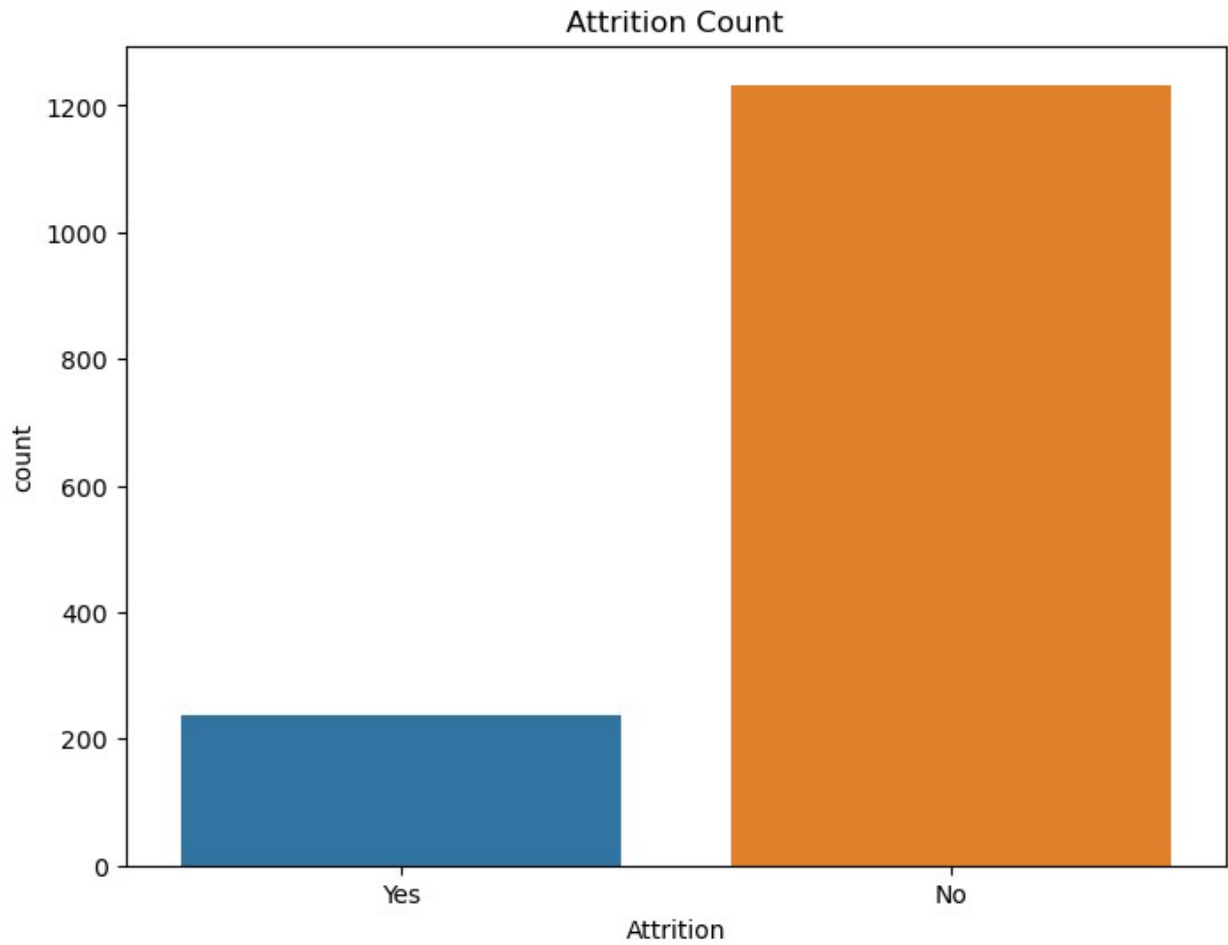
Data Visualization

```
attrition_counts = df['Attrition'].value_counts()
plt.figure(figsize=(6, 6))
plt.pie(attrition_counts, labels=attrition_counts.index,
autopct='%1.1f%%', startangle=90)
plt.title('Attrition Distribution')
plt.axis('equal')

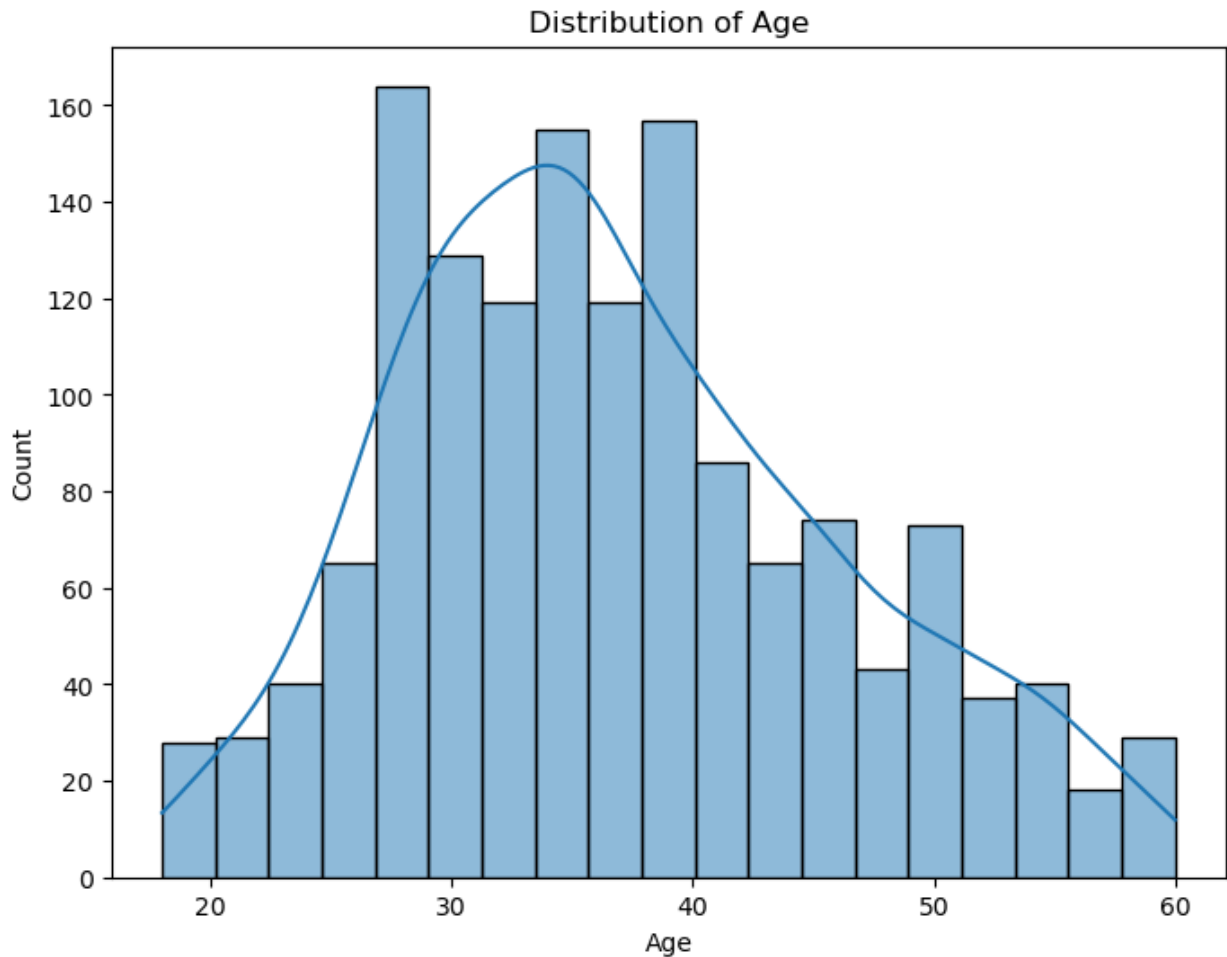
plt.show()
```



```
plt.figure(figsize=(8, 6))  
sns.countplot(x="Attrition", data=df)  
plt.title("Attrition Count")  
plt.show()
```

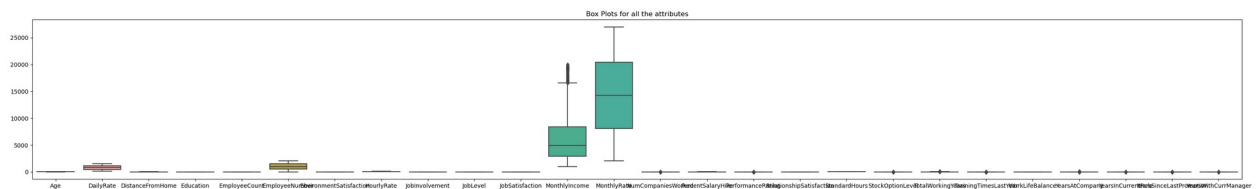



```
plt.figure(figsize=(8, 6))  
sns.histplot(data=df, x="Age", kde=True)  
plt.title("Distribution of Age")  
plt.show()
```

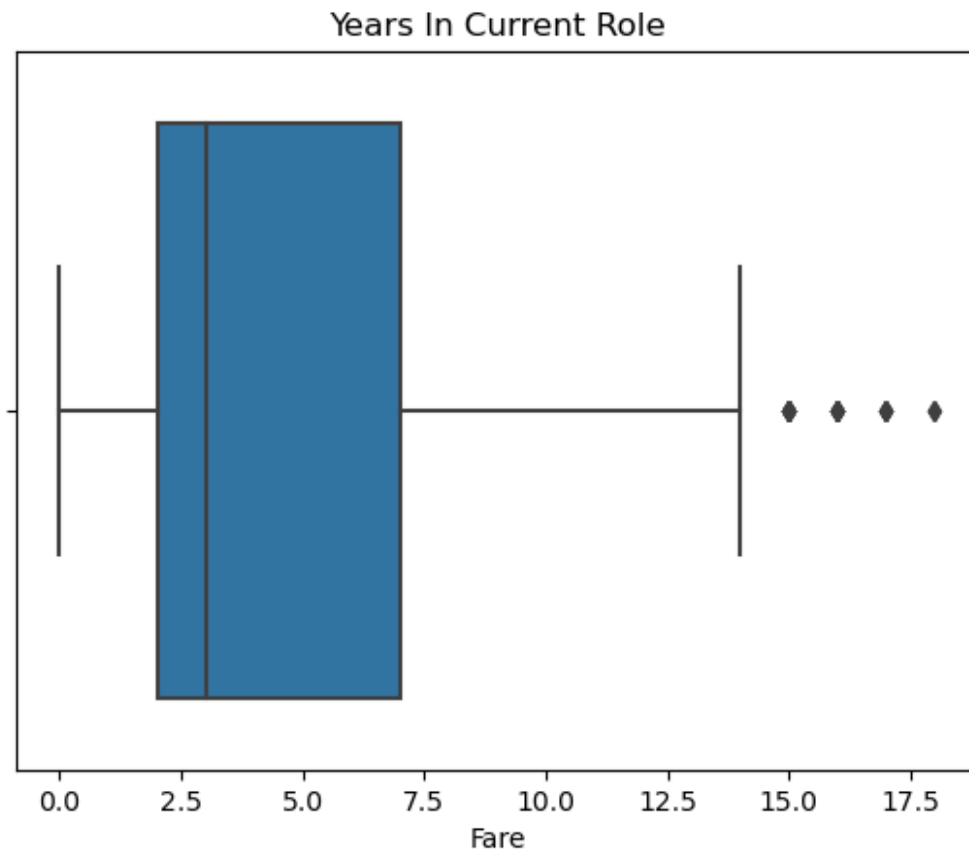


Outlier Detection

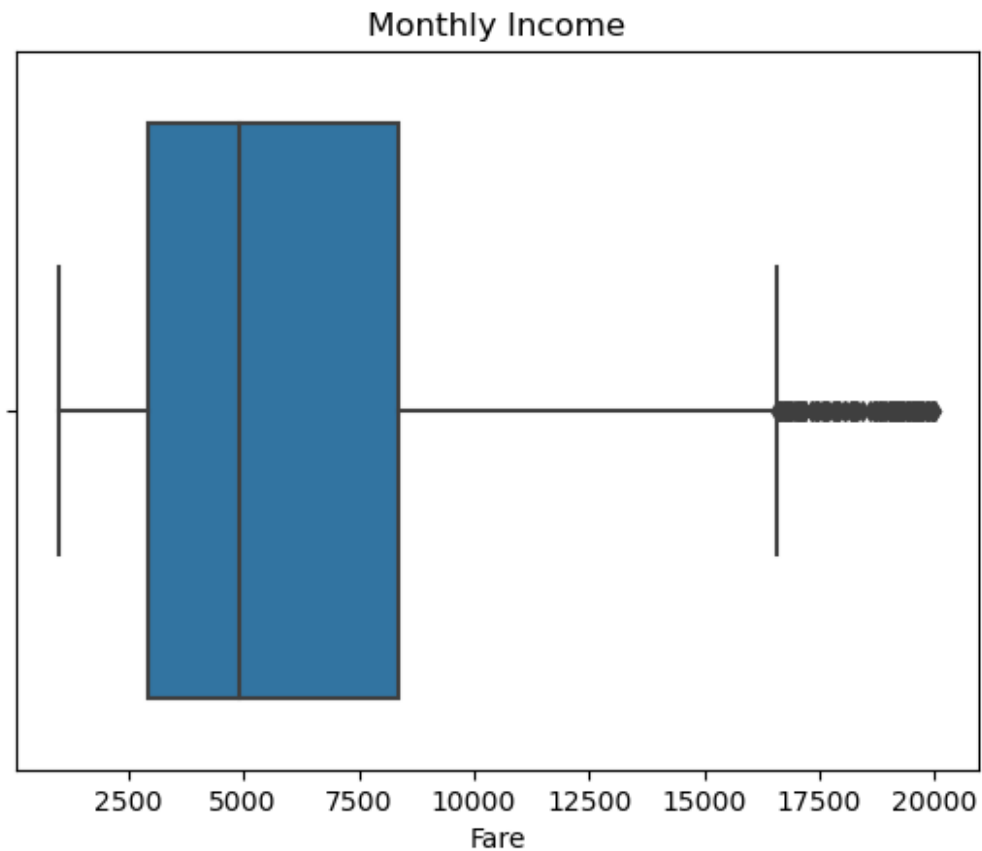
```
plt.figure(figsize=(38, 5))
sns.boxplot(data=df)
plt.title('Box Plots for all the attributes')
plt.show()
```



```
sns.boxplot(data=df, x='YearsInCurrentRole')
plt.title('Years In Current Role')
plt.xlabel('Fare')
plt.show()
```



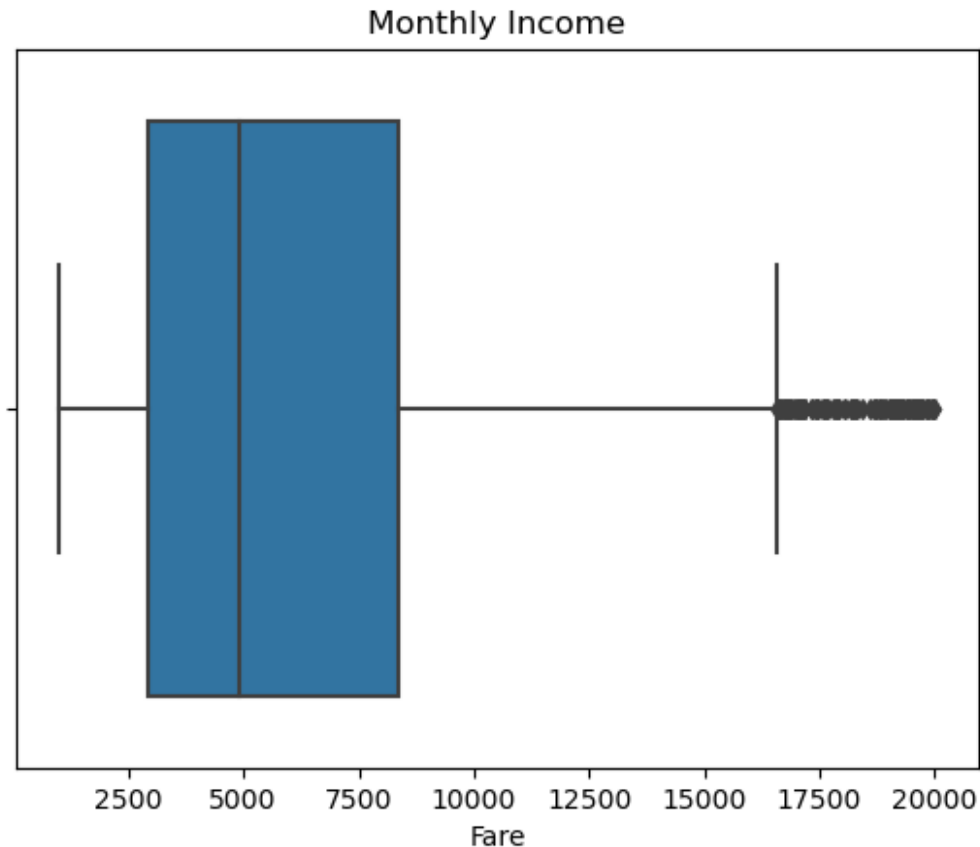
```
sns.boxplot(data=df, x='MonthlyIncome')  
plt.title('Monthly Income')  
plt.xlabel('Fare')  
plt.show()
```



```
from scipy import stats

z_scores = stats.zscore(df['MonthlyIncome'])
z_score_threshold = 3
df_cleaned = df[(np.abs(z_scores) <= z_score_threshold)]

sns.boxplot(data=df_cleaned, x='MonthlyIncome')
plt.title('Monthly Income')
plt.xlabel('Fare')
plt.show()
```



Splitting Independent and dependent variables

```
x = df.drop(columns= 'Attrition')
```

```
x.head()
```

	Age	BusinessTravel	DailyRate	Department
0	41	Travel_Rarely	1102	Sales
1	49	Travel_Frequently	279	Research & Development
2	37	Travel_Rarely	1373	Research & Development
3	33	Travel_Frequently	1392	Research & Development
4	27	Travel_Rarely	591	Research & Development

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

7

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

	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 34 columns]

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

```
y.head()
```

0	Yes
1	No

```

2     Yes
3     No
4     No
Name: Attrition, dtype: object

```

Encoding

```

categorical_features =
x.select_dtypes(include=['object']).columns.tolist()
x_encoded = pd.get_dummies(x, columns=categorical_features,
drop_first=True)

```

```
x_encoded.head()
```

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount
EmployeeNumber \					
0	41	1102	1	2	1
1					
1	49	279	8	1	1
2					
2	37	1373	2	2	1
4					
3	33	1392	3	4	1
5					
4	27	591	2	1	1
7					

	EnvironmentSatisfaction	HourlyRate	JobInvolvement	JobLevel	...
\					
0		2	94	3	2 ...
1		3	61	2	2 ...
2		4	92	2	1 ...
3		4	56	3	1 ...
4		1	40	3	1 ...

	JobRole_Laboratory Technician	JobRole_Manager	\
0	0	0	
1	0	0	
2	1	0	
3	0	0	
4	1	0	

	JobRole_Manufacturing Director	JobRole_Research Director	\
0	0	0	
1	0	0	

2	0	0
3	0	0
4	0	0

	JobRole_Research Scientist	JobRole_Sales Executive \
0	0	1
1	1	0
2	0	0
3	1	0
4	0	0

	JobRole_Sales Representative	MaritalStatus_Married
MaritalStatus_Single \		
0	0	0
1		
1	0	1
0		
2	0	0
1		
3	0	1
0		
4	0	1
0		

	OverTime_Yes
0	1
1	0
2	1
3	1
4	0

[5 rows x 47 columns]

Feature Scaling

```
from sklearn.preprocessing import StandardScaler
```

```
scaler = StandardScaler()
x_scaled = pd.DataFrame(scaler.fit_transform(x_encoded),
                        columns=x_encoded.columns)
```

```
x_scaled.head()
```

	Age	DailyRate	DistanceFromHome	Education	EmployeeCount \
0	0.446350	0.742527	-1.010909	-0.891688	0.0
1	1.322365	-1.297775	-0.147150	-1.868426	0.0
2	0.008343	1.414363	-0.887515	-0.891688	0.0
3	-0.429664	1.461466	-0.764121	1.061787	0.0
4	-1.086676	-0.524295	-0.887515	-1.868426	0.0

	EmployeeNumber	EnvironmentSatisfaction	HourlyRate	JobInvolvement
\				
0	-1.701283	-0.660531	1.383138	0.379672
1	-1.699621	0.254625	-0.240677	-1.026167
2	-1.696298	1.169781	1.284725	-1.026167
3	-1.694636	1.169781	-0.486709	0.379672
4	-1.691313	-1.575686	-1.274014	0.379672
	JobLevel ...	JobRole_Laboratory Technician	JobRole_Manager	\
0	-0.057788 ...	-0.462464	-0.273059	
1	-0.057788 ...	-0.462464	-0.273059	
2	-0.961486 ...	2.162331	-0.273059	
3	-0.961486 ...	-0.462464	-0.273059	
4	-0.961486 ...	2.162331	-0.273059	
	JobRole_Manufacturing Director	JobRole_Research Director	\	
0	-0.330808	-0.239904		
1	-0.330808	-0.239904		
2	-0.330808	-0.239904		
3	-0.330808	-0.239904		
4	-0.330808	-0.239904		
	JobRole_Research Scientist	JobRole_Sales Executive	\	
0	-0.497873	1.873287		
1	2.008543	-0.533821		
2	-0.497873	-0.533821		
3	2.008543	-0.533821		
4	-0.497873	-0.533821		
	JobRole_Sales Representative	MaritalStatus_Married		
MaritalStatus_Single \				
0	-0.244625	-0.918921		
1.458650				
1	-0.244625	1.088232		-
0.685565				
2	-0.244625	-0.918921		
1.458650				
3	-0.244625	1.088232		-
0.685565				
4	-0.244625	1.088232		-
0.685565				
	OverTime_Yes			
0	1.591746			
1	-0.628241			

```
2      1.591746
3      1.591746
4     -0.628241
```

```
[5 rows x 47 columns]
```

```
x=x_scaled
```

Training and Testing

```
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y,
test_size=0.2, random_state=42)
```

Model Building

```
# Import the necessary libraries
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score, classification_report,
confusion_matrix
from joblib import dump

logreg_model = LogisticRegression(random_state=42)
dt_model = DecisionTreeClassifier(random_state=42)

logreg_model.fit(x_train, y_train)
dt_model.fit(x_train, y_train)

DecisionTreeClassifier(random_state=42)

logreg_predictions = logreg_model.predict(x_test)

dt_predictions = dt_model.predict(x_test)

logreg_accuracy = accuracy_score(y_test, logreg_predictions)
print("Logistic Regression Accuracy:", logreg_accuracy)

dt_accuracy = accuracy_score(y_test, dt_predictions)
print("Decision Tree Accuracy:", dt_accuracy)

logreg_report = classification_report(y_test, logreg_predictions)
print("Classification Report for Logistic Regression:\n",
logreg_report)

dt_report = classification_report(y_test, dt_predictions)
print("Classification Report for Decision Tree Classifier:\n",
dt_report)

logreg_conf_matrix = confusion_matrix(y_test, logreg_predictions)
```

```
print("Confusion Matrix for Logistic Regression:\n",  
logreg_conf_matrix)
```

```
dt_conf_matrix = confusion_matrix(y_test, dt_predictions)  
print("Confusion Matrix for Decision Tree Classifier:\n",  
dt_conf_matrix)
```

Logistic Regression Accuracy: 0.8809523809523809

Decision Tree Accuracy: 0.7721088435374149

Classification Report for Logistic Regression:

	precision	recall	f1-score	support
No	0.92	0.95	0.93	255
Yes	0.56	0.46	0.51	39
accuracy			0.88	294
macro avg	0.74	0.70	0.72	294
weighted avg	0.87	0.88	0.88	294

Classification Report for Decision Tree Classifier:

	precision	recall	f1-score	support
No	0.87	0.86	0.87	255
Yes	0.17	0.18	0.17	39
accuracy			0.77	294
macro avg	0.52	0.52	0.52	294
weighted avg	0.78	0.77	0.78	294

Confusion Matrix for Logistic Regression:

```
[[241  14]
```

```
[ 21  18]]
```

Confusion Matrix for Decision Tree Classifier:

```
[[220  35]
```

```
[ 32   7]]
```