

IMPORT LIBRARIES

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

IMPORT DATASET

```
df=pd.read_csv("WA_Fn-UseC_-HR-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.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.tail()

	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]

df.shape

(1470, 35)

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]

corr=df.corr()

corr

```
<ipython-input-11-7d5195e2bf4d>: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 0.009819	-0.021490	-0.037848	-0.026556
YearsAtCompany 0.069114	0.311309	-0.034055	0.009508
YearsInCurrentRole 0.060236	0.212901	0.009932	0.018845
YearsSinceLastPromotion 0.054254	0.216513	-0.033229	0.010029
YearsWithCurrManager 0.069065	0.202089	-0.026363	0.014406

	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 0.029820	0.010146	0.024287
DailyRate 0.046135	0.018355	0.023381
DistanceFromHome 0.008783	-0.016075	0.031131
Education 0.042438	-0.027128	0.016775

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		-0.004999	-0.020123	
0.025976				
	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	

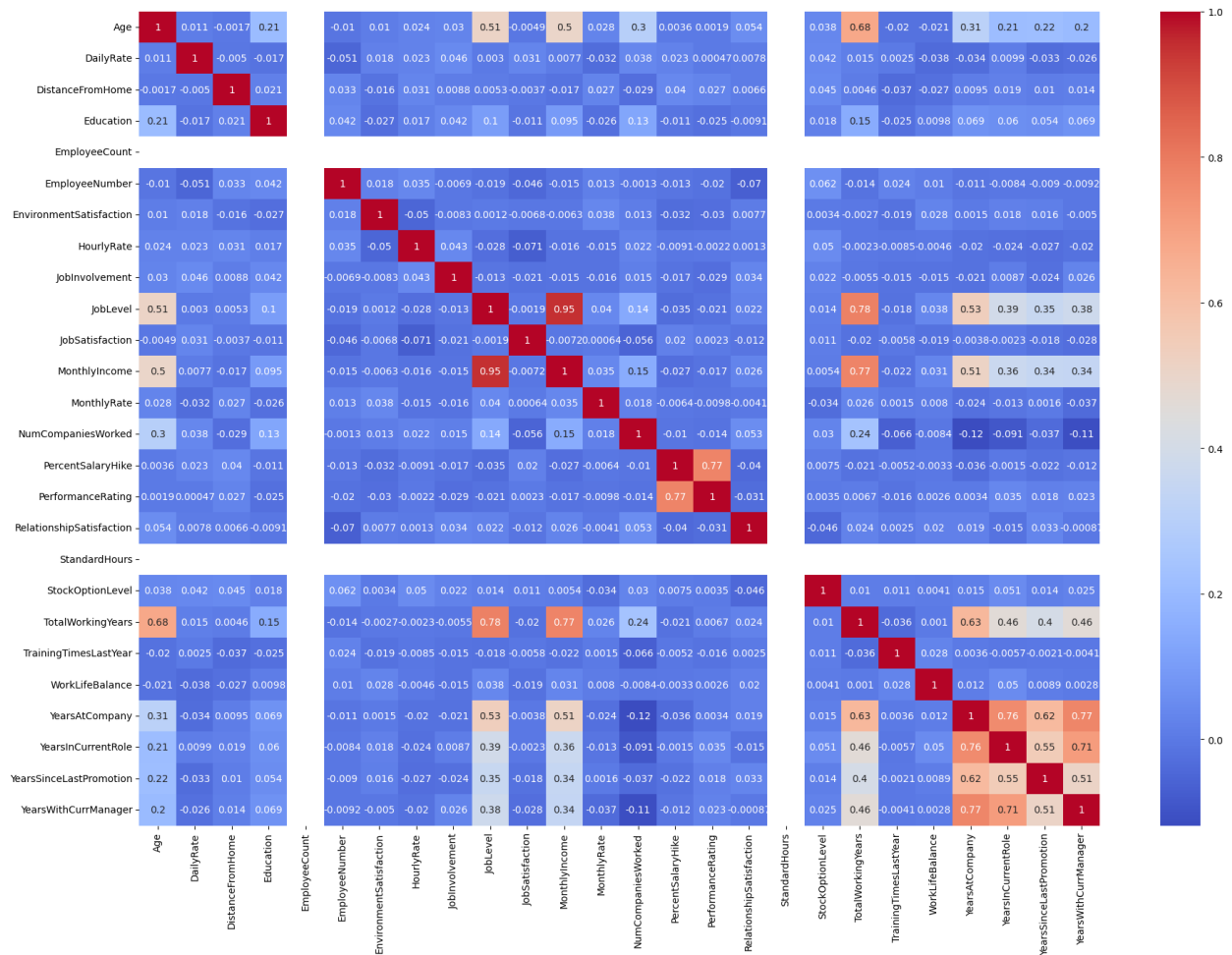
	YearsSinceLastPromotion
YearsWithCurrManager	
Age	0.216513
0.202089	
DailyRate	-0.033229
0.026363	
DistanceFromHome	0.010029
0.014406	
Education	0.054254
0.069065	
EmployeeCount	NaN
NaN	

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]

```
plt.subplots(figsize=(22,15))
sns.heatmap(corr,annot=True,cmap="coolwarm")
```

<Axes: >



```
df.Attrition.value_counts()
```

No	1233
Yes	237

Name: Attrition, dtype: int64

Checking for NULL Values

```
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

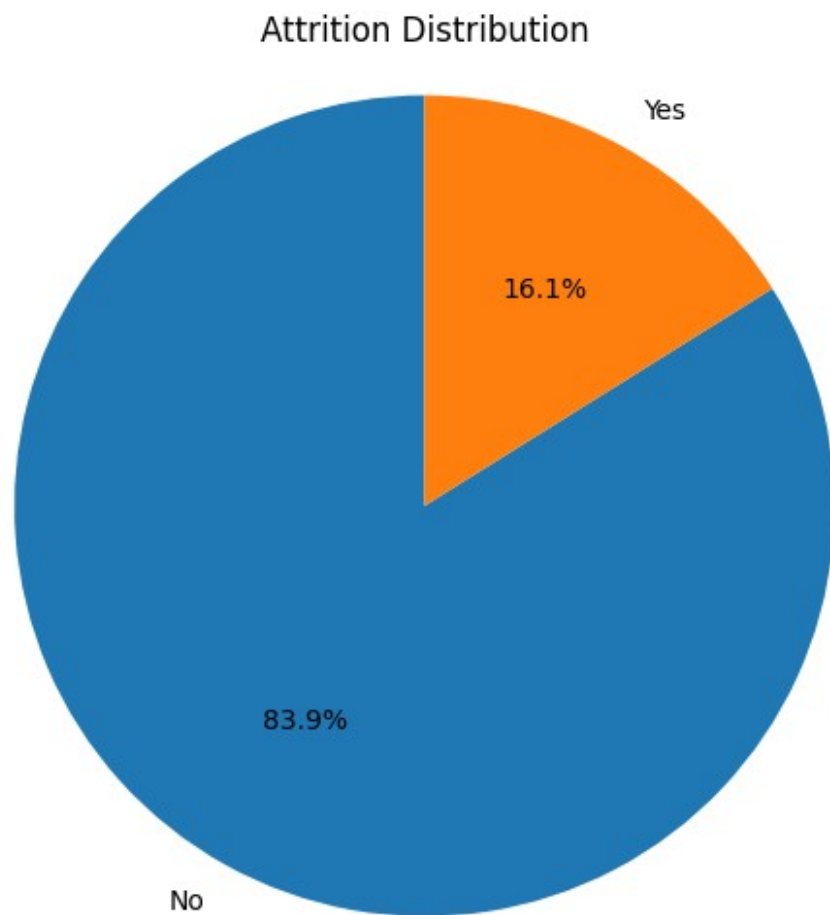
Data Visualization

```

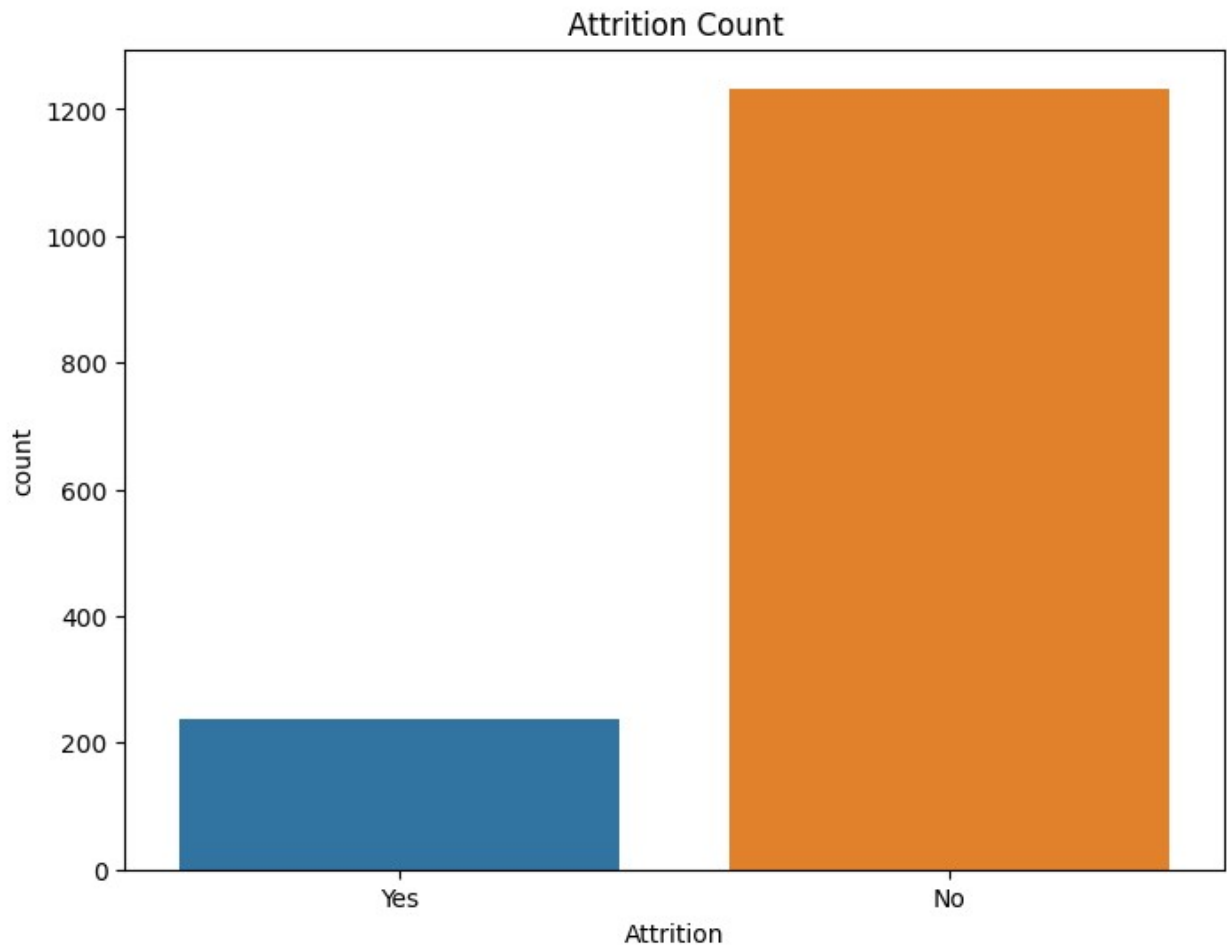
attrition_counts = df['Attrition'].value_counts()
plt.figure(figsize=(6, 6))
plt.pie(attrition_counts, labels=attrition_counts.index,
autopct='%1.1f%%', startangle=90)
plt.title('Attrition Distribution')
plt.axis('equal')

plt.show()

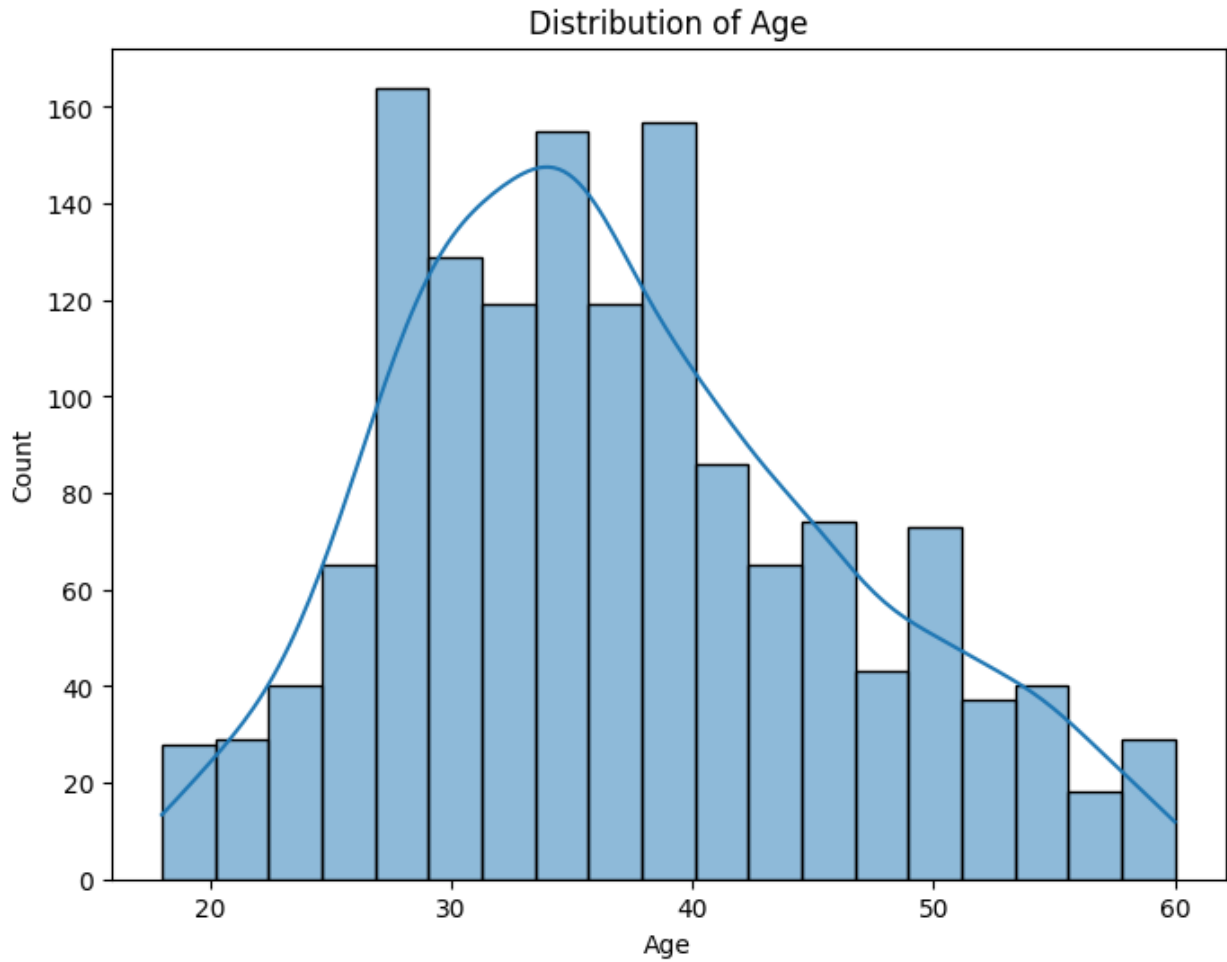
```

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

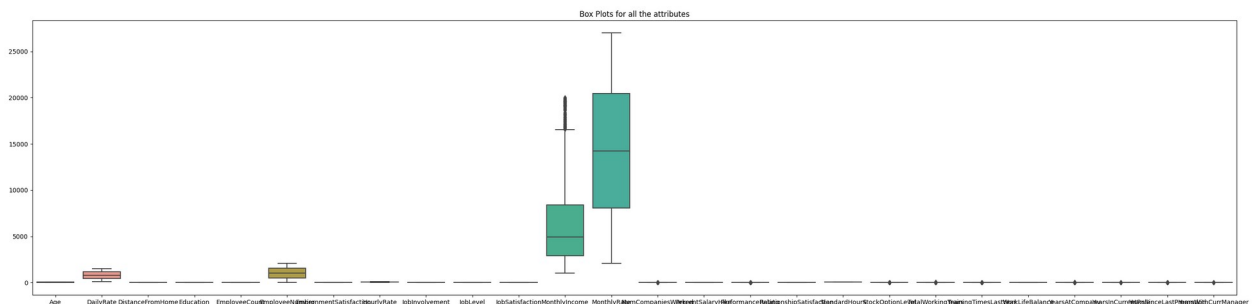


```
plt.figure(figsize=(8, 6))  
sns.histplot(data=df, x="Age", kde=True)  
plt.title("Distribution of Age")  
plt.show()
```



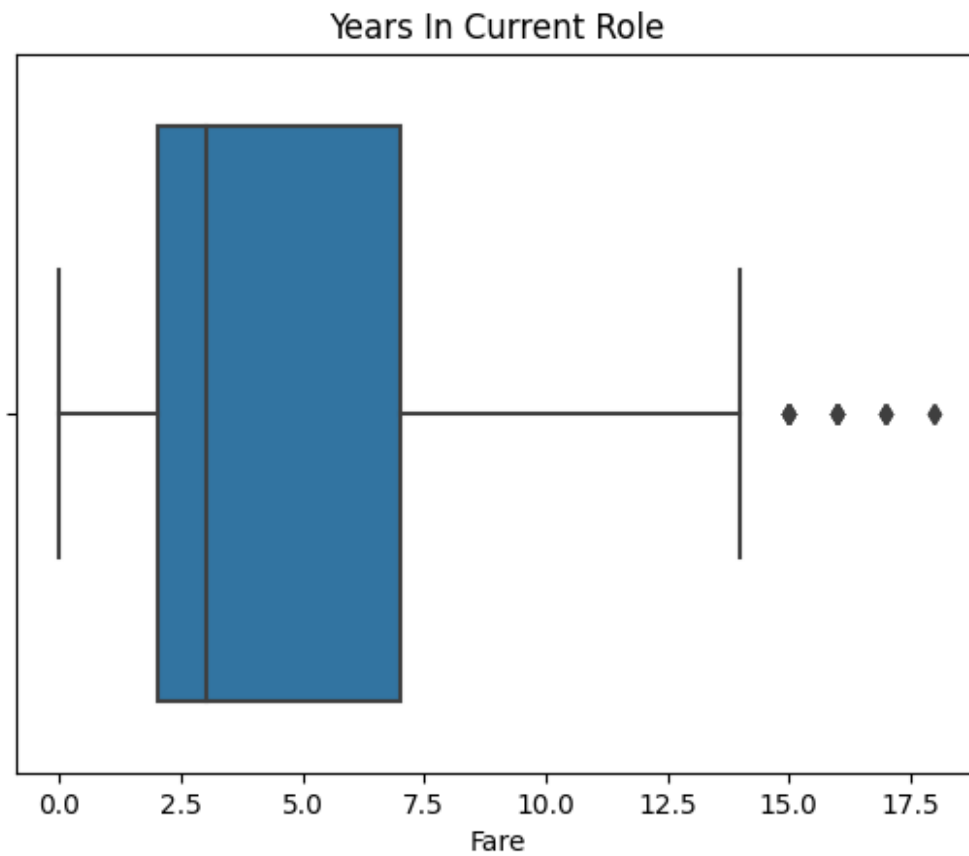
Outlier Detection

```
plt.figure(figsize=(35, 8))
sns.boxplot(data=df)
plt.title('Box Plots for all the attributes')
plt.show()
```

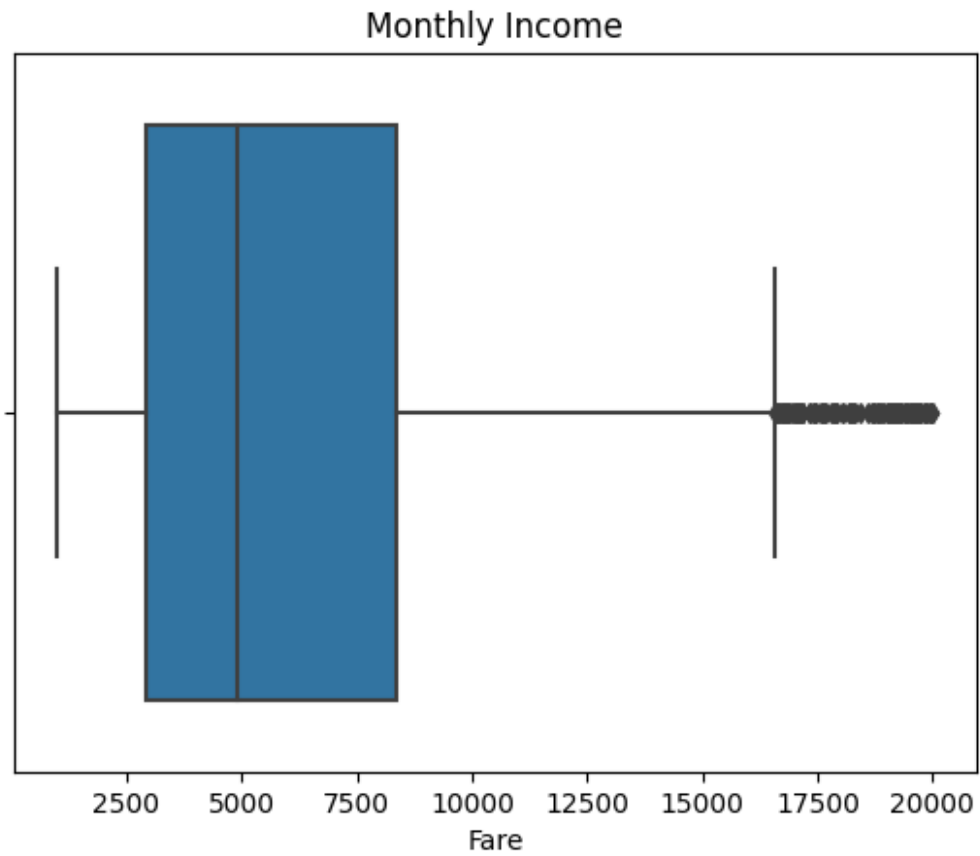


```
sns.boxplot(data=df, x='YearsInCurrentRole')
plt.title('Years In Current Role')
```

```
plt.xlabel('Fare')  
plt.show()
```



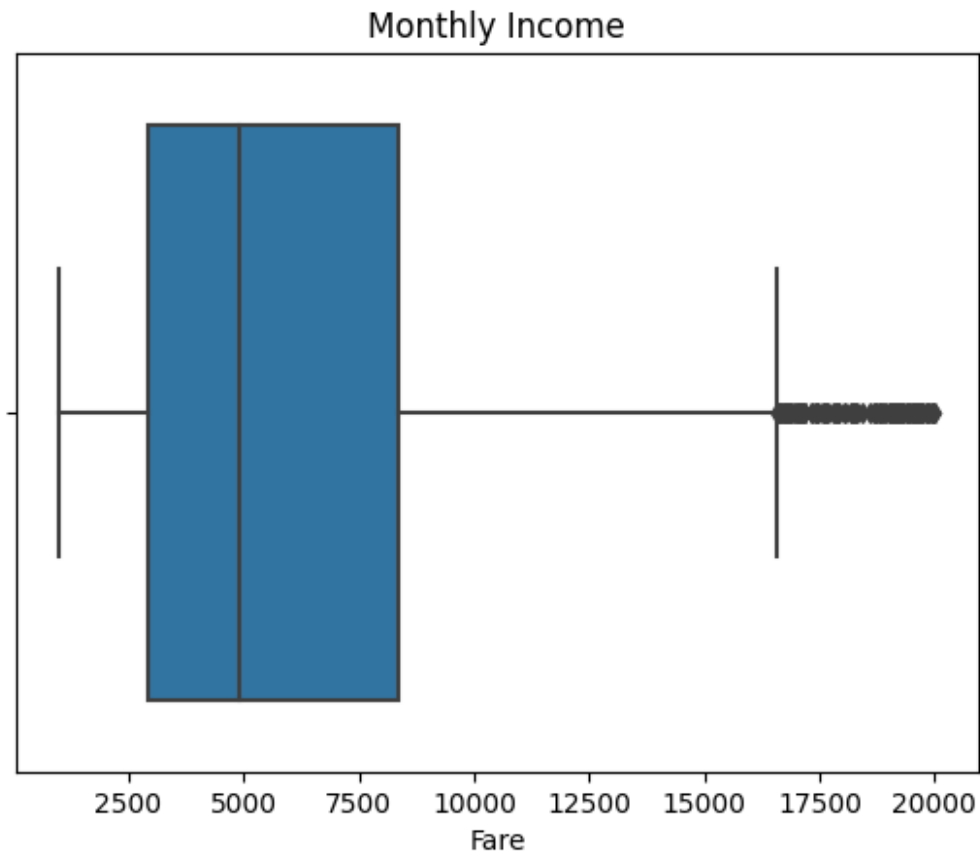
```
sns.boxplot(data=df, x='MonthlyIncome')  
plt.title('Monthly Income')  
plt.xlabel('Fare')  
plt.show()
```



```
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()
```



So the outliers are in large quantity, and they are inside the threshold, so let us not remove the outliers

SPLITTING INDEPENDENT AND DEPENDENT VARIABLES

```
x= df.drop(columns=["Attrition"])
```

```
y = df["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
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

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.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
```

Train and test split

```
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]]
```