

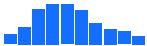
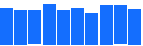


AFAME TECHNOLOGIES: HR DATA ANALYSIS

Importing Libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
import warnings
warnings.simplefilter('ignore')
```

Dataset: HR DATA ANALYSIS

```
data=pd.read_csv("HR Data.csv")
data
```

	Age int64 18 - 60	Attrition object	BusinessTravel obj...	DailyRate int64 102 - 1499	Department object	DistanceFromHome i	Education int64 1 - 5	EducationField obj...
		No 83.9% Yes 16.1%	Travel_Rarely 71% Travel_Freq... 18.8% Non-Travel 10.2%		Research &... 65.4% Sales 30.3% Human Reso... 4.3%			Life Sciences 41.2% Medical 31.6% 4 others 27.2%
0	41	Yes	Travel_Rarely	1102	Sales	1	2	Life Sciences
1	49	No	Travel_Frequently	279	Research & Devel...	8	1	Life Sciences
2	37	Yes	Travel_Rarely	1373	Research & Devel...	2	2	Other
3	33	No	Travel_Frequently	1392	Research & Devel...	3	4	Life Sciences
4	27	No	Travel_Rarely	591	Research & Devel...	2	1	Medical
5	32	No	Travel_Frequently	1005	Research & Devel...	2	2	Life Sciences
6	59	No	Travel_Rarely	1324	Research & Devel...	3	3	Medical
7	30	No	Travel_Rarely	1358	Research & Devel...	24	1	Life Sciences
8	38	No	Travel_Frequently	216	Research & Devel...	23	3	Life Sciences
9	36	No	Travel_Rarely	1299	Research & Devel...	27	3	Medical

1470 rows, 35 cols, showing 10 rows/page

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```
data.describe()
```

	Age float64	DailyRate float64	DistanceFromHome f	Education float64	EmployeeCount flo...	EmployeeNumber f...	EnvironmentSatisf...	HourlyRate float64
cou...	1470	1470	1470	1470	1470	1470	1470	1470
me...	36.92380952	802.4857143	9.192517007	2.91292517	1	1024.865306	2.721768707	65.89115646
std	9.135373489	403.5090999	8.106864436	1.024164945	0	602.0243348	1.093082215	20.32942759
min	18	102	1	1	1	1	1	30
25%	30	465	2	2	1	491.25	2	48
50%	36	802	7	3	1	1020.5	3	66
75%	43	1157	14	4	1	1555.75	4	83.75
max	60	1499	29	5	1	2068	4	100

8 rows, 26 cols, showing 10 rows/page

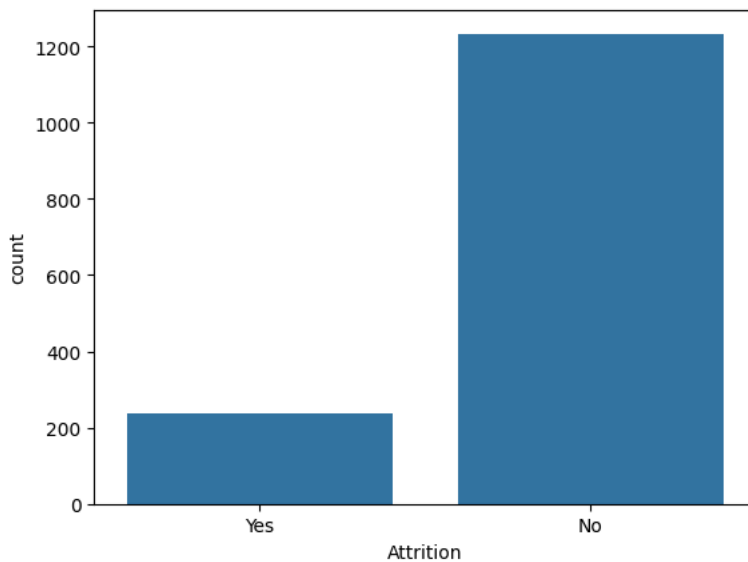
<< < Page 1 of 1 > >>

```
data.info()
```

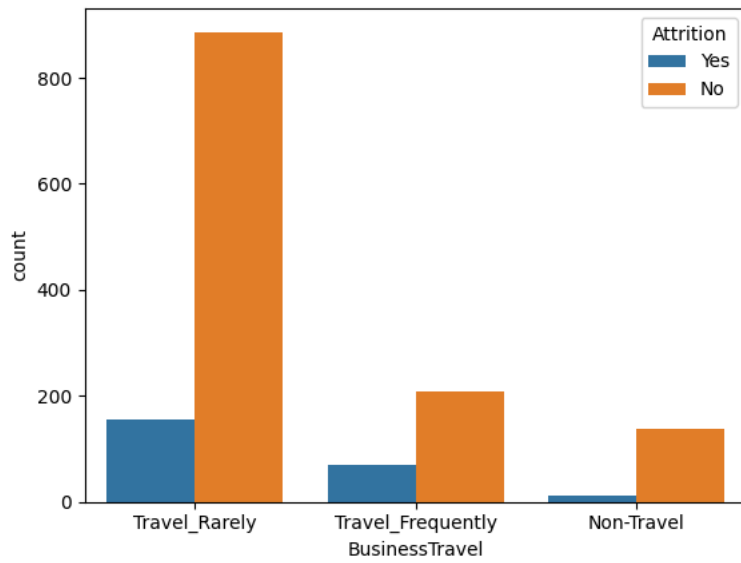
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1470 entries, 0 to 1469
Data columns (total 35 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Age                   1470 non-null   int64
1   Attrition             1470 non-null   object
2   BusinessTravel        1470 non-null   object
3   DailyRate             1470 non-null   int64
4   Department            1470 non-null   object
5   DistanceFromHome      1470 non-null   int64
6   Education             1470 non-null   int64
7   EducationField         1470 non-null   object
8   EmployeeCount         1470 non-null   int64
9   EmployeeNumber        1470 non-null   int64
10  EnvironmentSatisfaction 1470 non-null   int64
11  Gender                1470 non-null   object
12  HourlyRate            1470 non-null   int64
13  JobInvolvement         1470 non-null   int64
14  JobLevel              1470 non-null   int64
15  JobRole               1470 non-null   object
16  JobSatisfaction        1470 non-null   int64
17  MaritalStatus         1470 non-null   object
18  MonthlyIncome          1470 non-null   int64
19  MonthlyRate           1470 non-null   int64
20  NumCompaniesWorked     1470 non-null   int64
21  Over18                1470 non-null   object
22  OverTime               1470 non-null   object
23  PercentSalaryHike      1470 non-null   int64
24  PerformanceRating      1470 non-null   int64
```

Data Visualization

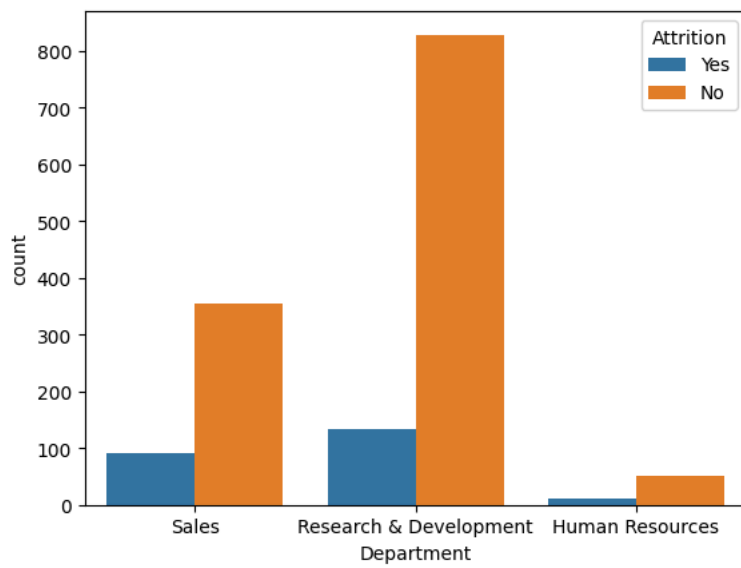
```
sns.countplot(x=data.Attrition)
plt.show()
```



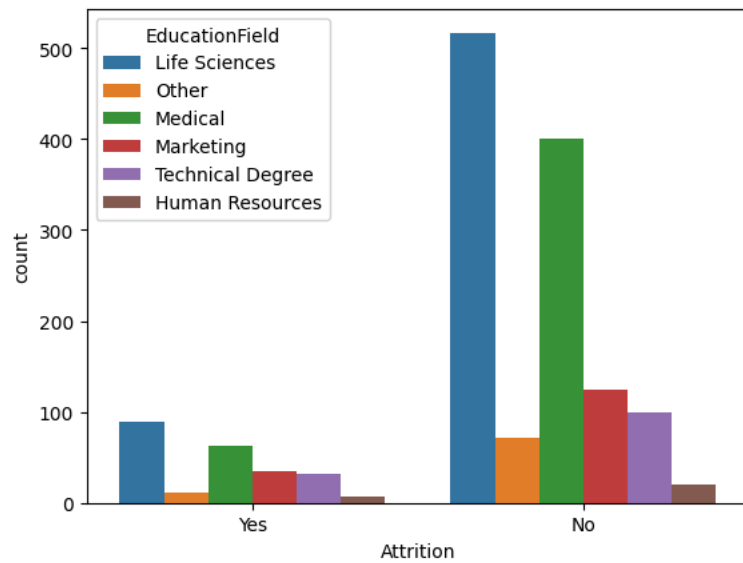
```
sns.countplot(hue=data.Attrition, x = data.BusinessTravel)
plt.show()
```



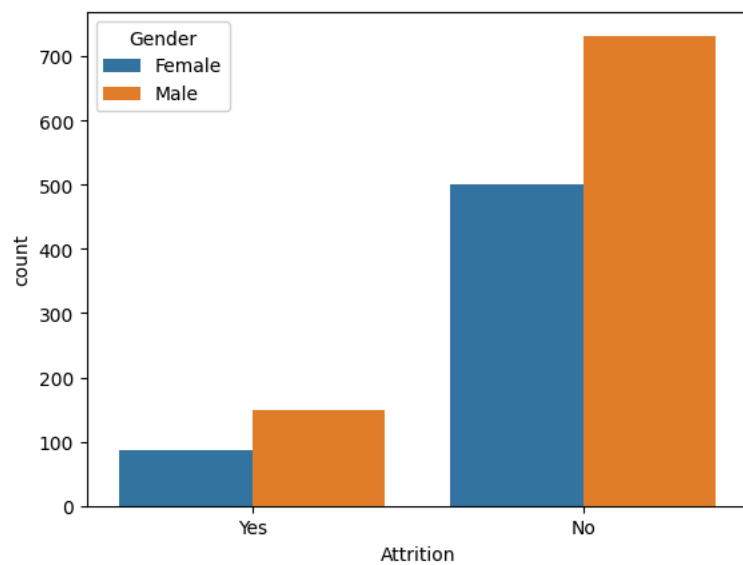
```
sns.countplot(hue=data.Attrition, x=data.Department)
plt.show()
```



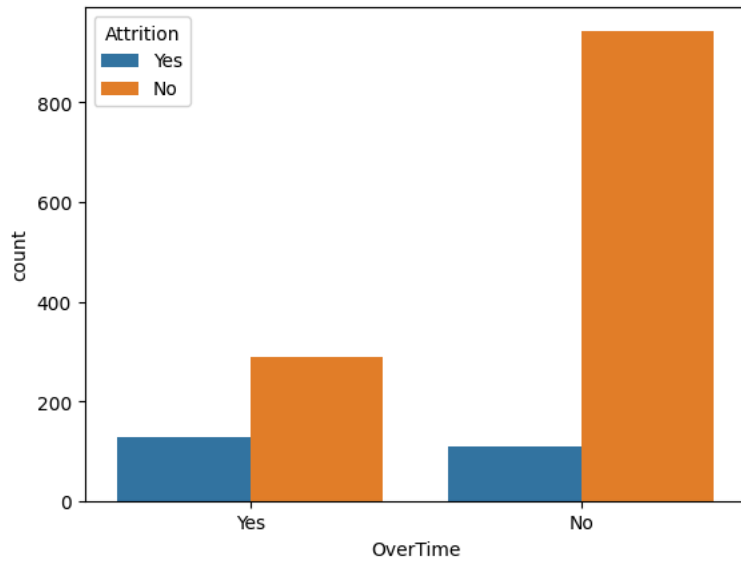
```
sns.countplot(x=data.Attrition, hue=data.EducationField)
plt.show()
```



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

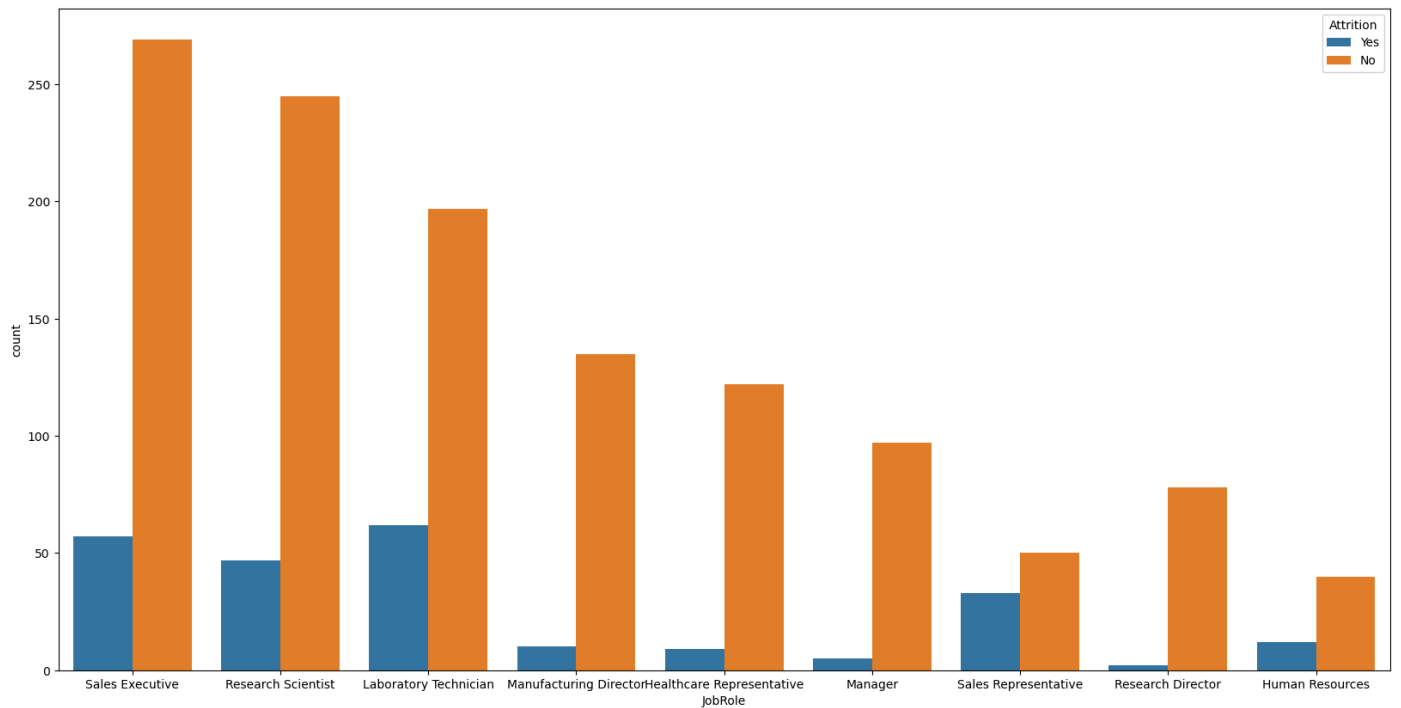


```
sns.countplot(hue=data.Attrition, x=data.OverTime)
plt.show()
```



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

```
Text(0.5, 0, 'JobRole')
```



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

```
Age                0
Attrition          0
BusinessTravel     0
DailyRate          0
Department         0
DistanceFromHome   0
Education          0
EducationField     0
EmployeeCount      0
EmployeeNumber     0
EnvironmentSatisfaction  0
Gender             0
HourlyRate         0
JobInvolvement     0
JobLevel           0
JobRole            0
JobSatisfaction    0
MaritalStatus      0
MonthlyIncome      0
MonthlyRate        0
NumCompaniesWorked 0
Over18             0
OverTime           0
PercentSalaryHike  0
PerformanceRating  0
RelationshipSatisfaction  0
StandardHours      0
StockOptionLevel   0
TotalWorkingYears  0
TrainingTimesLastYear  0
WorkLifeBalance    0
YearsAtCompany     0
```

```
data.shape
```

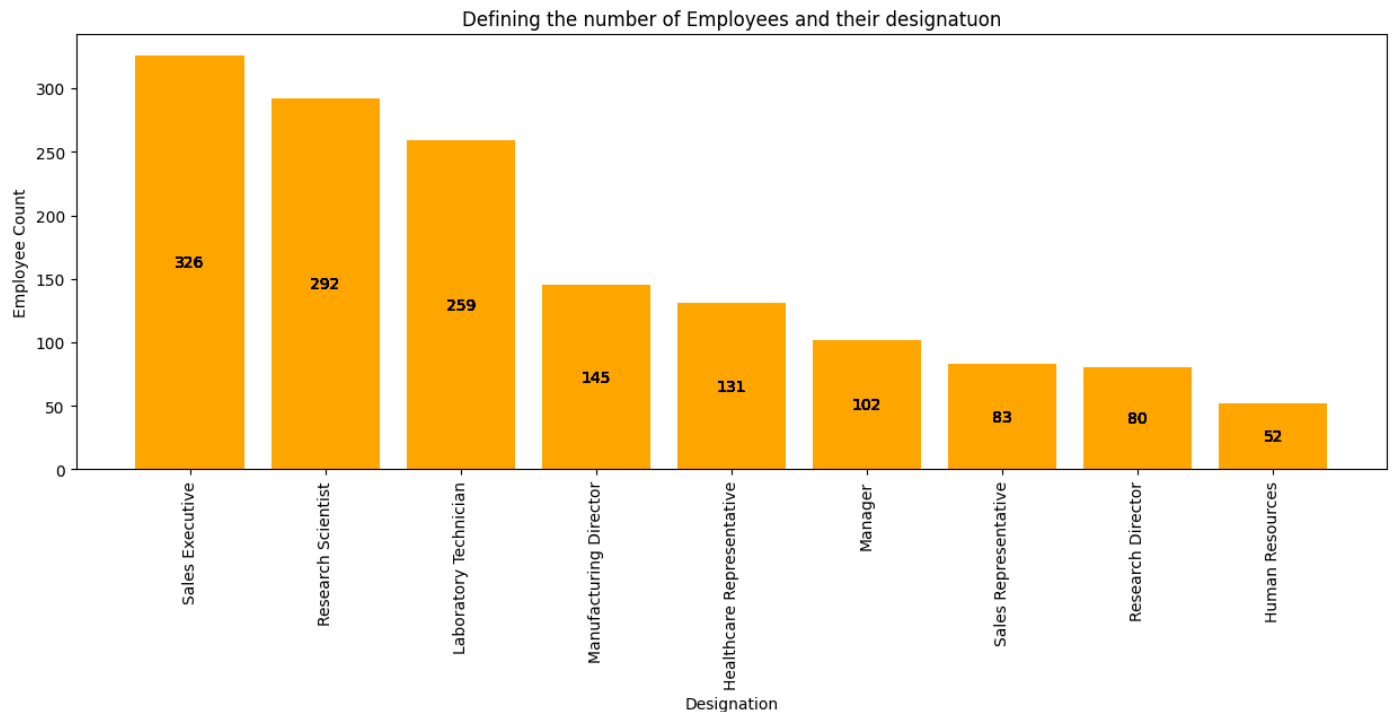
```
(1470, 35)
```

```
job_role_num=data['JobRole'].value_counts()
```

```
job_role_num
```

```
JobRole
Sales Executive      326
Research Scientist   292
Laboratory Technician 259
Manufacturing Director 145
Healthcare Representative 131
Manager              102
Sales Representative  83
Research Director     80
Human Resources       52
Name: count, dtype: int64
```

```
plt.figure(figsize=(15,5))
for i in job_role_num.items():
    p = plt.bar(job_role_num.index, job_role_num.values,color='orange')
    plt.bar_label(p, label_type='center')
plt.xlabel('Designation')
plt.ylabel('Employee Count')
plt.title('Defining the number of Employees and their designatuon')
plt.xticks(rotation='vertical')
plt.show()
```



Data Preprocessing

```
data.columns
```

```
Index(['Age', 'Attrition', 'BusinessTravel', 'DailyRate', 'Department',
      'DistanceFromHome', 'Education', 'EducationField', 'EmployeeCount',
      'EmployeeNumber', 'EnvironmentSatisfaction', 'Gender', 'HourlyRate',
      'JobInvolvement', 'JobLevel', 'JobRole', 'JobSatisfaction',
      'MaritalStatus', 'MonthlyIncome', 'MonthlyRate', 'NumCompaniesWorked',
      'Over18', 'OverTime', 'PercentSalaryHike', 'PerformanceRating',
      'RelationshipSatisfaction', 'StandardHours', 'StockOptionLevel',
      'TotalWorkingYears', 'TrainingTimesLastYear', 'WorkLifeBalance',
      'YearsAtCompany', 'YearsInCurrentRole', 'YearsSinceLastPromotion',
      'YearsWithCurrManager'],
      dtype='object')
```

```
data = data.drop(columns=['EmployeeCount', 'StandardHours', 'Over18', 'EmployeeNumber'])
```

```
data.columns
```

```
Index(['Age', 'Attrition', 'BusinessTravel', 'DailyRate', 'Department',  
      'DistanceFromHome', 'Education', 'EducationField',  
      'EnvironmentSatisfaction', 'Gender', 'HourlyRate', 'JobInvolvement',  
      'JobLevel', 'JobRole', 'JobSatisfaction', 'MaritalStatus',  
      'MonthlyIncome', 'MonthlyRate', 'NumCompaniesWorked', 'OverTime',  
      'PercentSalaryHike', 'PerformanceRating', 'RelationshipSatisfaction',  
      'StockOptionLevel', 'TotalWorkingYears', 'TrainingTimesLastYear',  
      'WorkLifeBalance', 'YearsAtCompany', 'YearsInCurrentRole',  
      'YearsSinceLastPromotion', 'YearsWithCurrManager'],  
      dtype='object')
```

```
data = data.rename(columns={  
    'Attrition': 'Turnover',  
    'BusinessTravel': 'Business_Travel',  
    'DailyRate': 'Daily_Rate',  
    'DistanceFromHome': 'Distance_From_Home',  
    'EducationField': 'Education_Field',  
    'EnvironmentSatisfaction': 'Environment_Satisfaction',  
    'HourlyRate': 'Hourly_Rate',  
    'JobInvolvement': 'Job_Involvement',  
    'JobLevel': 'Job_Level',  
    'JobRole': 'Job_Role',  
    'JobSatisfaction': 'Job_Satisfaction',  
    'MaritalStatus': 'Marital_Status',  
    'MonthlyIncome': 'Monthly_Income',  
    'MonthlyRate': 'Monthly_Rate',  
    'NumCompaniesWorked': 'Num_Companies_Worked',  
    'OverTime': 'Over_Time',  
    'PercentSalaryHike': 'Percent_Salary_Hike',  
    'PerformanceRating': 'Performance_Rating',  
    'RelationshipSatisfaction': 'Relationship_Satisfaction',  
    'StockOptionLevel': 'Stock_Option_Level',  
    'TotalWorkingYears': 'Total_Working_Years',  
    'TrainingTimesLastYear': 'Training_Times_Last_Year',  
    'WorkLifeBalance': 'Work_Life_Balance',  
    'YearsAtCompany': 'Years_At_Company',  
    'YearsInCurrentRole': 'Years_In_Current_Role',  
    'YearsSinceLastPromotion': 'Years_Since_Last_Promotion',  
    'YearsWithCurrManager': 'Years_With_Curr_Manager'  
})
```

```
data['Turnover'] = data['Turnover'].astype('category')  
data['Business_Travel'] = data['Business_Travel'].astype('category')  
data['Department'] = data['Department'].astype('category')  
data['Education_Field'] = data['Education_Field'].astype('category')  
data['Gender'] = data['Gender'].astype('category')  
data['Job_Role'] = data['Job_Role'].astype('category')  
data['Marital_Status'] = data['Marital_Status'].astype('category')  
data['Over_Time'] = data['Over_Time'].astype('category')
```

```
#Eliminate NaN values
```

```
data = data.dropna()
```

```
data = data.drop_duplicates()
```


print(data.info())

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1470 entries, 0 to 1469
Data columns (total 31 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Age                                    1470 non-null   int64
1   Turnover                             1470 non-null   category
2   Business_Travel                       1470 non-null   category
3   Daily_Rate                           1470 non-null   int64
4   Department                           1470 non-null   category
5   Distance_From_Home                   1470 non-null   int64
6   Education                             1470 non-null   int64
7   Education_Field                       1470 non-null   category
8   Environment_Satisfaction              1470 non-null   int64
9   Gender                                1470 non-null   category
10  Hourly_Rate                           1470 non-null   int64
11  Job_Involvement                       1470 non-null   int64
12  Job_Level                             1470 non-null   int64
13  Job_Role                              1470 non-null   category
14  Job_Satisfaction                      1470 non-null   int64
15  Marital_Status                       1470 non-null   category
16  Monthly_Income                       1470 non-null   int64
17  Monthly_Rate                         1470 non-null   int64
18  Num_Companies_Worked                  1470 non-null   int64
19  Over_Time                            1470 non-null   category
20  Percent_Salary_Hike                  1470 non-null   int64
21  Performance_Rating                   1470 non-null   int64
22  Relationship_Satisfaction              1470 non-null   int64
23  Stock_Option_Level                   1470 non-null   int64
24  Total_Working_Years                   1470 non-null   int64
```

data								
	<div>Age int64</div> <div>18 - 60</div> <div></div>	<div>Turnover category</div> <div>No 83.9%</div> <div>Yes 16.1%</div>	<div>Business_Travel ca...</div> <div>Travel_Rarely ... 71%</div> <div>Travel_Freq... 18.8%</div> <div>Non-Travel ... 10.2%</div>	<div>Daily_Rate int64</div> <div>102 - 1499</div> <div></div>	<div>Department categ...</div> <div>Research &... 65.4%</div> <div>Sales 30.3%</div> <div>Human Reso... 4.3%</div>	<div>Distance_From_H...</div> <div>1 - 29</div> <div></div>	<div>Education int64</div> <div>1 - 5</div> <div></div>	<div>Education_Field ca...</div> <div>Life Sciences 41.2%</div> <div>Medical 31.6%</div> <div>4 others 27.2%</div>
0	41	Yes	Travel_Rarely	1102	Sales	1	2	Life Sciences
1	49	No	Travel_Frequently	279	Research & Devel...	8	1	Life Sciences
2	37	Yes	Travel_Rarely	1373	Research & Devel...	2	2	Other
3	33	No	Travel_Frequently	1392	Research & Devel...	3	4	Life Sciences
4	27	No	Travel_Rarely	591	Research & Devel...	2	1	Medical
5	32	No	Travel_Frequently	1005	Research & Devel...	2	2	Life Sciences
6	59	No	Travel_Rarely	1324	Research & Devel...	3	3	Medical
7	30	No	Travel_Rarely	1358	Research & Devel...	24	1	Life Sciences
8	38	No	Travel_Frequently	216	Research & Devel...	23	3	Life Sciences
9	36	No	Travel_Rarely	1299	Research & Devel...	27	3	Medical

1470 rows, 31 cols, showing 10 rows/page

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↓

Checking whether NaN value is present or not.

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

```
Age                0
Turnover           0
Business_Travel    0
Daily_Rate         0
Department         0
Distance_From_Home 0
Education          0
Education_Field     0
Environment_Satisfaction 0
Gender             0
Hourly_Rate        0
Job_Involvement    0
Job_Level          0
Job_Role           0
Job_Satisfaction   0
Marital_Status     0
Monthly_Income     0
Monthly_Rate       0
Num_Companies_Worked 0
Over_Time          0
Percent_Salary_Hike 0
Performance_Rating 0
Relationship_Satisfaction 0
Stock_Option_Level 0
Total_Working_Years 0
Training_Times_Last_Year 0
Work_Life_Balance  0
Years_At_Company   0
Years_In_Current_Role 0
Years_Since_Last_Promotion 0
Years_With_Curr_Manager 0
dtype: int64
```

Cleaned Dataset

```
# cleaned dataset
data.to_csv('cleaned_hr_data.csv', index=False)
```