Title: Employee Attrition & Performance Analysis using Python

Objective:

The objective of this project is to analyze employee attrition trends and identify key factors influencing employees to leave the organization. The goal is to provide actionable insights to the HR department to reduce attrition and improve employee retention.

Tools & Technologies Used:

- Python
- Pandas (Data Handling)
- Seaborn & Matplotlib (Data Visualization)
- Jupyter Notebook
- CSV File (Dataset)

Dataset Overview:

The dataset contains employee records with demographic, job role, compensation, and satisfaction-related information.

Total Records: 1470 employees

Key Columns:

Age, Gender, Department, EducationField, JobRole, MonthlyIncome, TotalWorkingYears, WorkLifeBalance, JobSatisfaction, MaritalStatus, Attrition (Yes/No)

Data Exploration & Visual Analysis:

import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
sns.set_style('whitegrid')
plt.rcParams['figure.figsize'] = (8,5)

Load Data

```
df = pd.read_csv(r"E:/HR_Analytics_Python_Project/data/HR_Attrition.csv")
print(df.head())
print(df.info())
```

											geIndex: 1470 entries, 0 to a columns (total 35 columns		
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1				4	80		1			16	JobSatisfaction	1470 non-null	int64
2				2	80		0			17	MaritalStatus	1470 non-null	object
3				3	80		0			18	MonthlyIncome	1470 non-null	int64
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2		7			3	3		0		25	RelationshipSatisfaction	1470 non-null	int64
3		8			3	3		8		26	StandardHours	1470 non-null	int64
4		6	5		3	3		2		27	StockOptionLevel	1470 non-null	int64
										28	TotalWorkingYears	1470 non-null	int64
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0		4	ı		0		- 5			30	WorkLifeBalance	1470 non-null	int64
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										memo	ory usage: 402.1+ KB		
[5	rows x	35 columns	1							None	:		

Basic Stats

print("Total Employees:", len(df))

print("Attrition Rate:\n", df['Attrition'].value_counts(normalize=True) * 100)

Total Employees: 1470

Attrition Rate:

Attrition

No 83.877551 Yes 16.122449

Name: proportion, dtype: float64

Gender-wise Attrition

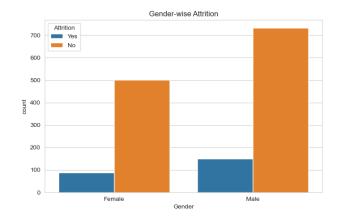
```
sns.countplot(x='Gender', hue='Attrition', data=df)

plt.title("Gender-wise Attrition")

plt.savefig(r'E:/HR_Analytics_Python_Project/output/graphs/gender_attrition.png')

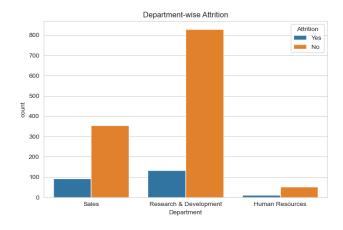
plt.show()

plt.clf()
```



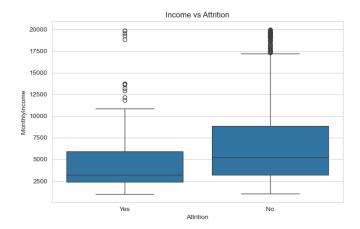
Department-wise Attrition

sns.countplot(x='Department', hue='Attrition', data=df)
plt.title("Department-wise Attrition")
plt.savefig(r'E:/HR_Analytics_Python_Project/output/graphs/department_attrition.png')
plt.show()
plt.clf()



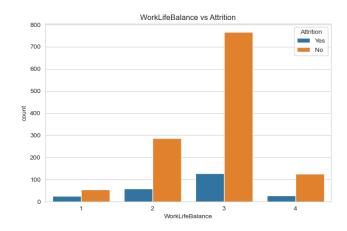
Income vs Attrition

```
sns.boxplot(x='Attrition', y='MonthlyIncome', data=df)
plt.title("Income vs Attrition")
plt.savefig(r'E:/HR_Analytics_Python_Project/output/graphs/income_attrition.png')
plt.show()
plt.clf()
```



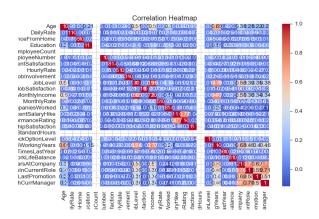
WorkLifeBalance impact

sns.countplot(x='WorkLifeBalance', hue='Attrition', data=df)
plt.title("WorkLifeBalance vs Attrition")
plt.savefig(r'E:/HR_Analytics_Python_Project/output/graphs/wlb_attrition.png')
plt.show()
plt.clf()



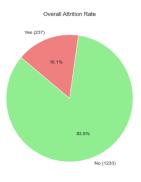
Correlation Heatmap

```
corr = df.corr(numeric_only=True)
sns.heatmap(corr, annot=True, cmap='coolwarm')
plt.title("Correlation Heatmap")
plt.savefig(r'E:/HR_Analytics_Python_Project/output/graphs/correlation.png')
plt.show()
plt.clf()
```



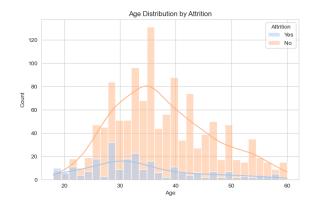
Overall Attrition Pie Chart

attrition_counts = df['Attrition'].value_counts()
plt.figure(figsize=(6,6))
plt.pie(attrition_counts, labels=[f"{label} ({count})" for label, count in zip(attrition_counts.index,
attrition_counts.values)], autopct='%1.1f%%', colors=['lightgreen','lightcoral'], startangle=140)
plt.title('Overall Attrition Rate')
plt.savefig(r'E:/HR_Analytics_Python_Project/output/graphs/overall_attrition_pie.png')
plt.show()
plt.clf()

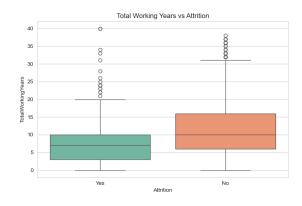


Age Distribution with Attrition

```
plt.figure(figsize=(8,5))
sns.histplot(data=df, x='Age', hue='Attrition', kde=True, palette='pastel', bins=30)
plt.title('Age Distribution by Attrition')
plt.savefig(r'E:/HR_Analytics_Python_Project/output/graphs/age_distribution.png')
plt.show()
plt.clf()
```

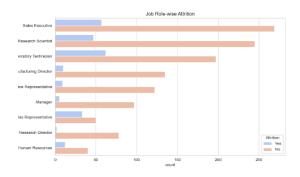


Total Working Years vs Attrition



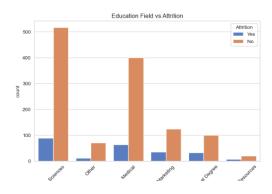
Job Role-wise Attrition

```
plt.figure(figsize=(10,6))
sns.countplot(y='JobRole', hue='Attrition', data=df, palette='coolwarm')
plt.title('Job Role-wise Attrition')
plt.savefig(r'E:/HR_Analytics_Python_Project/output/graphs/jobrole_attrition.png')
plt.show()
plt.clf()
```



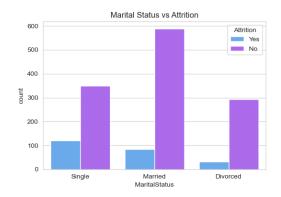
Education Field ka Impact Analysis

```
plt.figure(figsize=(8,5))
sns.countplot(x='EducationField', hue='Attrition', data=df, palette='muted')
plt.title("Education Field vs Attrition")
plt.xticks(rotation=45)
plt.savefig(r'E:/HR_Analytics_Python_Project/output/graphs/education_field_attrition.png')
plt.show()
plt.clf()
```



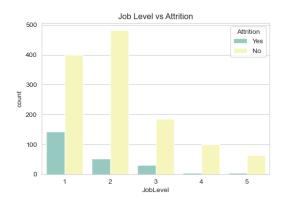
Marital Status vs Attrition

```
plt.figure(figsize=(6,4))
sns.countplot(x='MaritalStatus', hue='Attrition', data=df, palette='cool')
plt.title("Marital Status vs Attrition")
plt.savefig(r'E:/HR_Analytics_Python_Project/output/graphs/marital_status_attrition.png')
plt.show()
plt.clf()
```



Attrition by Job Level

plt.figure(figsize=(6,4))
sns.countplot(x='JobLevel', hue='Attrition', data=df, palette='Set3')
plt.title("Job Level vs Attrition")
plt.savefig(r'E:/HR_Analytics_Python_Project/output/graphs/job_level_attrition.png')
plt.show()
plt.clf()



Summary

Higher attrition observed among employees in Sales department. Female employees show slightly higher attrition trends.. Employees with lower Work-Life Balance tend to leave more Low-income and low Job Satisfaction employees are at greater risk. Education field and job level also play a role in attrition patterns.

Conclusion

- Improve work-life balance programs, especially in Sales department
- Review compensation for employees in lower income brackets
- Conduct regular employee engagement and satisfaction surveys
- Focus retention efforts on departments with historically high attrition
- Provide career growth paths to employees with low job levels