Problem Statement

- People Charm', a growing company is facing a high attrition rate among their employees which in turn affects their business due to lack of expertise and experience.
- Their HR department is assigned the task to reduce the attrition rate by retaining employees who are about to churn out.
- They need to recommend special plans or strategies which will help them to retain their employees which in turn will help them to grow bigger as a company

- ATTRITION = NO. OF EMPLOYEES LEAVING THE COMPANY / TOTAL NUMBER OF EMPLOYEES
 - EMPLOYEES LEAVING= FIRED EMPLOYEE, RESIGNED EMPLOYEE
- If the experienced employee leave the company and new employee joins, it takes ample amount of time to get trained for new employee.
- DATA IS OF 14999 EMPLOYEES



Project Introduction

- Title: Analyze Employee Attrition Using Python
- Objective: Identify the reasons for employee attrition (leaving the company) and explore strategies to minimize attrition using Python libraries like NumPy, Pandas, and Matplotlib.

Dataset Overview

1

Total size :14999 x 10 Data file : People Charm

Variables Description		
satisfactoryLevel	Scores given by the employees, scaling 0 to 1	
lastEvaluation	Last evaluation points given, scaling 0 to 1	
numberOfProjects	Number of projects involved	
avgMonthlyHours	Average monthly hours	
timeSpent.company	Time spent at the company, in years	
workAccident	Whether he/she had a work accident	
left	if the employee is about to leave or not, about to leave(serving notice period) - 1 and 0 otherwise	
promotionInLast5years	Whether he/she had a promotion in the last 5 years	
dept Department he/she belongs to		
Salary	Salary as high, medium or low	

Dataset Overview

- Dataset Columns:
 - Satisfaction Level: Employee satisfaction score.
 - Last Evaluation: Latest performance evaluation score.
 - **Number of Projects**: The number of projects assigned to the employee.
 - Average Monthly Hours: Employee's average working hours per month.
 - Years at Company: Time spent at the company (in years).
 - Work Accident: If the employee had a work accident (0: No, 1: Yes).
 - Left: Whether the employee left the company (0: No, 1: Yes).
 - **Promotion in Last 5 Years**: Whether the employee received a promotion (0: No, 1: Yes).
 - **Department**: The department where the employee worked.
 - Salary: Employee salary level (low, medium, high).



Dataset Overview

import pandas as pd

```
# Load the dataset
df = pd.read_csv('your_dataset.csv')
```

Display first few rows df.head()

Tools and Libraries







NUMPY: FOR NUMERICAL OPERATIONS.

PANDAS: FOR DATA MANIPULATION AND ANALYSIS.

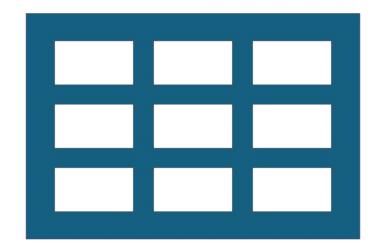
MATPLOTLIB: FOR DATA VISUALIZATION TO EXPLORE TRENDS. import pandas as pd import numpy as np import matplotlib.pyplot as plt

```
# Loading the dataset
df = pd.read_csv('people.csv')
df.head()
```

Data Preprocessing

•Tasks:

- Handle missing values.
- Convert categorical data (salary levels) to numerical form.
- Filter relevant columns for analysis.



```
# Drop rows with missing values df.dropna(inplace=True)
```

```
# Convert 'Attrition' column to binary (Yes=1, No=0)
df['Attrition'] = df['Attrition'].apply(lambda x: 1 if x == 'Yes' else 0)
```

df.head()



Key Factors for Attrition

- •Satisfaction Level: Dissatisfied employees are more likely to leave.
- •Number of Projects: Employees with either too few or too many projects have a higher attrition rate.
- •Average Monthly Hours: Overworked employees are more likely to leave.
- •Promotion: Employees without promotions in the last 5 years are more likely to leave.

Group data by Attrition and calculate mean for each feature attrition_factors = df.groupby('Attrition').mean()

Display the key statistics for attrition analysis attrition_factors[['YearsAtCompany', 'MonthlyIncome', 'NumCompaniesWorked', 'JobSatisfaction']]

Visualizing Data 1. Attrition vs Salary Range

```
plt.figure(figsize=(10, 5))
plt.hist(df[df['Attrition'] == 1]['MonthlyIncome'], bins=20, alpha=0.7,
label='Attrition=Yes', color='red')
plt.hist(df[df['Attrition'] == 0]['MonthlyIncome'], bins=20, alpha=0.7,
label='Attrition=No', color='green')
plt.title('Monthly Income Distribution by Attrition')
plt.xlabel('Monthly Income')
plt.ylabel('Number of Employees')
plt.legend()
plt.show()
Insight: Employees with lower monthly income tend to leave more
frequently.
```

2. Attrition vs Years at Company

```
plt.figure(figsize=(10, 5))
plt.hist(df[df['Attrition'] == 1]['YearsAtCompany'], bins=10, alpha=0.7,
label='Attrition=Yes', color='red')
plt.hist(df[df['Attrition'] == 0]['YearsAtCompany'], bins=10, alpha=0.7,
label='Attrition=No', color='green')
plt.title('Years at Company by Attrition')
plt.xlabel('Years at Company')
plt.ylabel('Number of Employees')
plt.legend()
plt.show()
Insight: Employees with 3-6 years at the company are more likely to leave.
```

3. Attrition vs Number of Projects

- plt.figure(figsize=(10, 5))
- plt.hist(df[df['Attrition'] == 1]['NumCompaniesWorked'], bins=8, alpha=0.7, label='Attrition=Yes', color='red')
- plt.hist(df[df['Attrition'] == 0]['NumCompaniesWorked'], bins=8, alpha=0.7, label='Attrition=No', color='green')
- plt.title('Number of Companies Worked vs Attrition')
- plt.xlabel('Number of Companies Worked')
- plt.ylabel('Number of Employees')
- plt.legend()
- plt.show()
- Insight: Employees with either very few or too many previous companies worked are more likely to leave.

Conclusion

• Key Takeaways:

- Satisfaction Level: Employees with low satisfaction are more prone to leaving.
- Workload: Balancing the number of projects can reduce attrition.
- Average Working Hours: Overwork contributes significantly to employee turnover.
- Promotion: Regular promotions help retain employees.



Recommendation

<u>5</u>

- Suggestions to Reduce Attrition:
 - Improve employee satisfaction through engagement initiatives.
 - Balance workload by assigning optimal project numbers.
 - Provide fair promotion opportunities to deserving employees.
 - Introduce incentives for overworked employees to improve retention.