

```

import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

# ML
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from sklearn.tree import DecisionTreeClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import classification_report, confusion_matrix,
accuracy_score

# SHAP for explainability
import shap

```

```

import pandas as pd

```

```

df = pd.read_csv("C:/Users/Balambiga2910/Documents/hr_dataset.csv")
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	

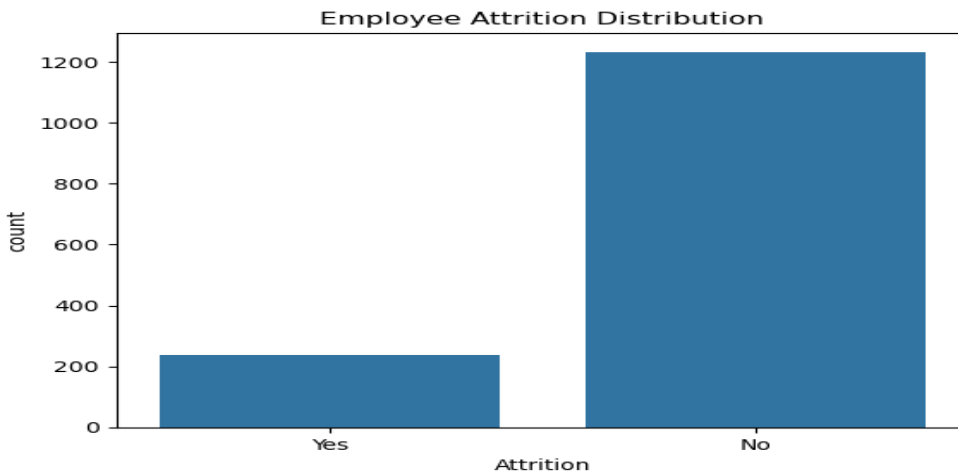
	TotalWorkingYears	TrainingTimesLastYear	WorkLifeBalance	YearsAtCompany	\
0	8		0 1		6
1	10		3 3		10
2	7		3 3		0
3	8		3 3		8

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]

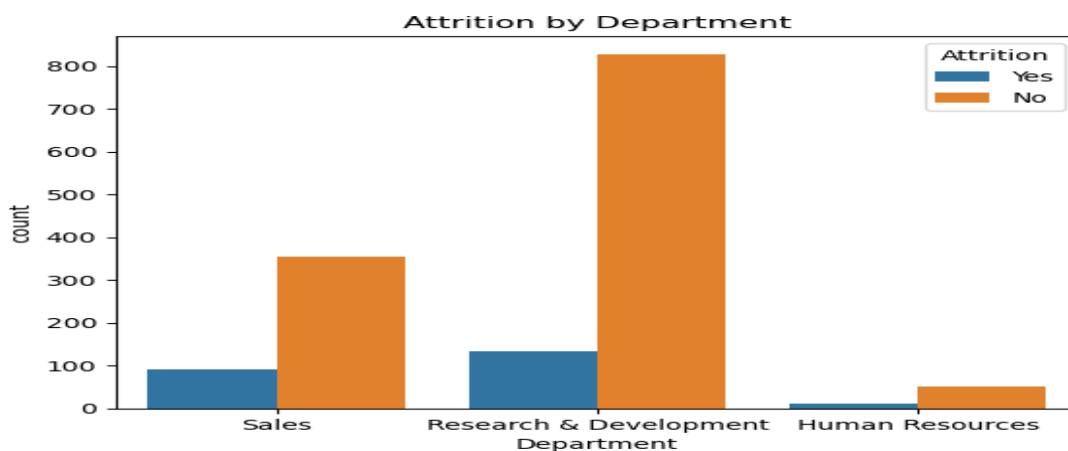
```
import seaborn as sns
import matplotlib.pyplot as plt
sns.countplot(x='Attrition', data=df)
plt.title("Employee Attrition Distribution")
```

Text(0.5, 1.0, 'Employee Attrition Distribution')



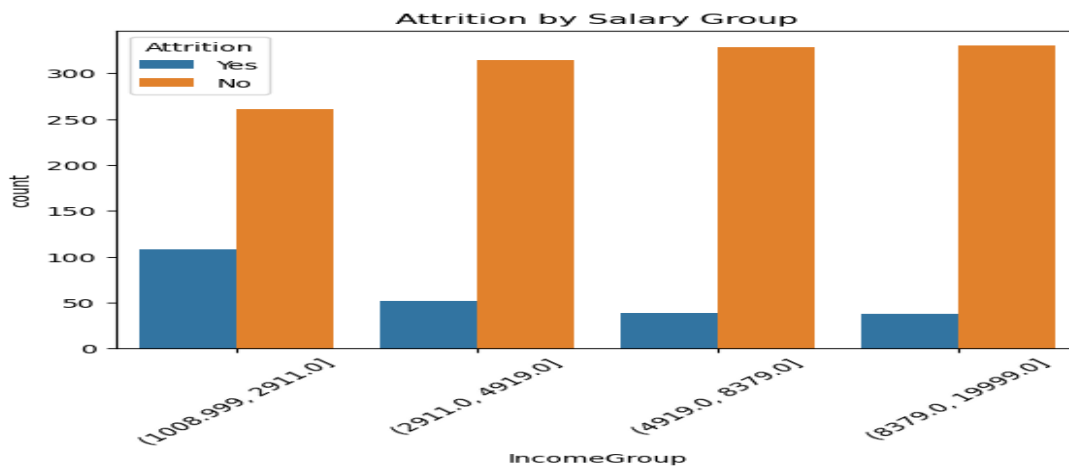
```
sns.countplot(x='Department', hue='Attrition', data=df)
plt.title("Attrition by Department")
```

Text(0.5, 1.0, 'Attrition by Department')



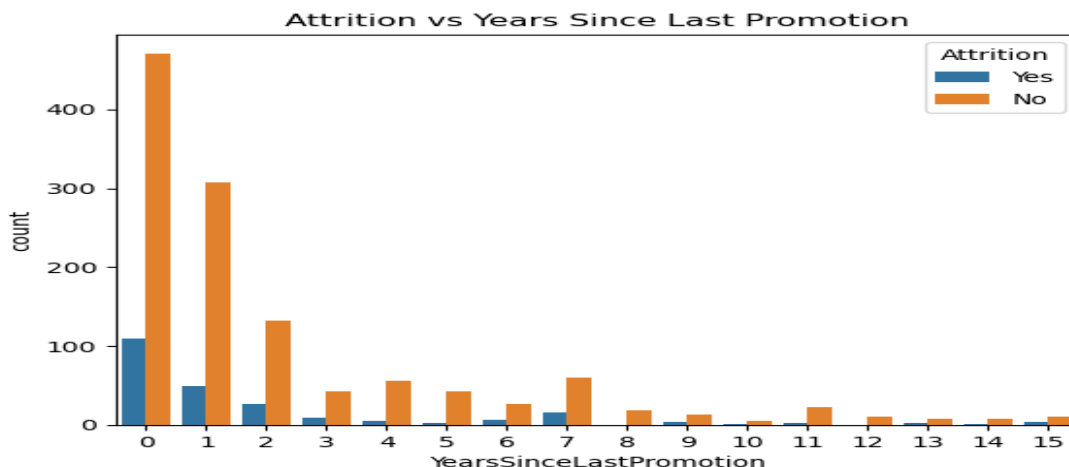
```
df['IncomeGroup'] = pd.qcut(df['MonthlyIncome'], 4)
sns.countplot(x='IncomeGroup', hue='Attrition', data=df)
plt.title("Attrition by Salary Group")
plt.xticks(rotation=45)
```

```
([0, 1, 2, 3],
 [Text(0, 0, '(1008.999, 2911.0]'),
  Text(1, 0, '(2911.0, 4919.0]'),
  Text(2, 0, '(4919.0, 8379.0]'),
  Text(3, 0, '(8379.0, 19999.0]')])
```



```
sns.countplot(x='YearsSinceLastPromotion', hue='Attrition', data=df)
plt.title("Attrition vs Years Since Last Promotion")
```

```
Text(0.5, 1.0, 'Attrition vs Years Since Last Promotion')
```



```
import pandas as pd
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split
# Drop irrelevant columns
```

```
cols_to_drop = ['EmployeeNumber', 'Over18', 'StandardHours', 'EmployeeCount']
existing_cols = [col for col in cols_to_drop if col in df.columns]
df.drop(existing_cols, axis=1, inplace=True)
```

```
# Encode categorical columns
```

```
le = LabelEncoder()
for column in df.select_dtypes(include='object').columns:
    df[column] = le.fit_transform(df[column])
```

```
# Split data
```

```
X = df.drop('Attrition', axis=1)
y = df['Attrition']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)
```

```
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score, classification_report,
confusion_matrix
tree_model = DecisionTreeClassifier(max_depth=5, random_state=42)
tree_model.fit(X_train, y_train)
y_pred_tree = tree_model.predict(X_test)
```

```
print("Decision Tree Accuracy:", accuracy_score(y_test, y_pred_tree))
print(classification_report(y_test, y_pred_tree))
```

```
Decision Tree Accuracy: 0.826530612244898
```

	precision	recall	f1-score	support
0	0.88	0.93	0.90	255
1	0.25	0.15	0.19	39
accuracy			0.83	294
macro avg	0.56	0.54	0.55	294
weighted avg	0.79	0.83	0.81	294

```
import shap
```

```
explainer = shap.TreeExplainer(tree_model)
shap_values = explainer.shap_values(X_test)
```

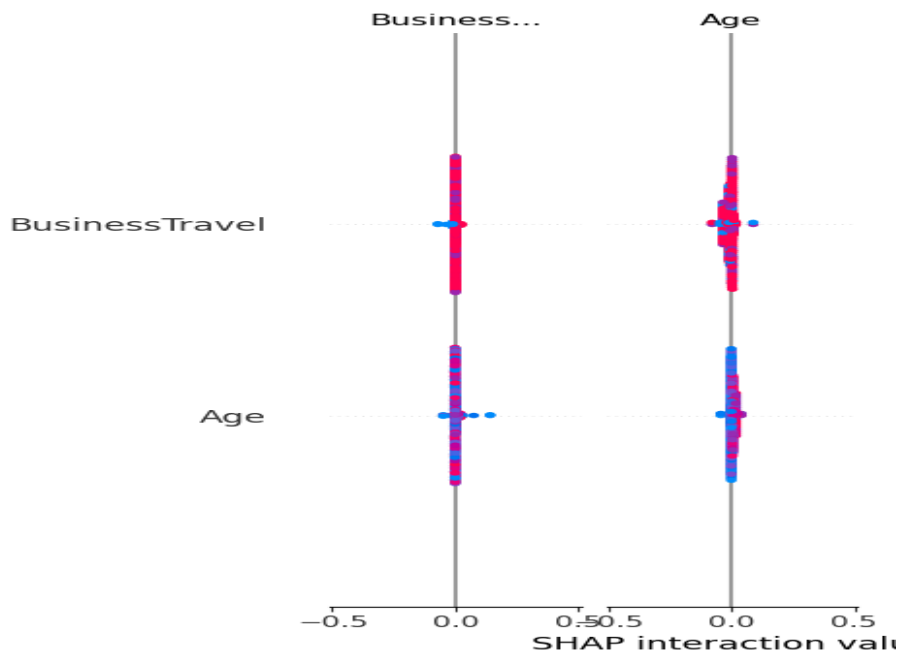
```
# Check structure
```

```
print("Length of shap_values:", len(shap_values)) # Debug print
```

```
# If it's a list of 2 elements, use [1]
```

```
if isinstance(shap_values, list) and len(shap_values) == 2:
    shap.summary_plot(shap_values[1], X_test)
else:
    shap.summary_plot(shap_values, X_test)
```

Length of shap\_values: 294



# 1. Import Libraries

```
import pandas as pd
```

```
import numpy as np
```

```
import seaborn as sns
```

```
import matplotlib.pyplot as plt
```

```
from sklearn.model_selection import train_test_split
```

```
from sklearn.preprocessing import LabelEncoder
```

```
from sklearn.tree import DecisionTreeClassifier
```

```
from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
```

# 2. Load the Dataset

```
df = pd.read_csv("C:/Users/Balambiga2910/Documents/hr dataset.csv")
```

### # 3. Drop Irrelevant Columns

```
cols_to_drop = ['EmployeeNumber', 'Over18', 'StandardHours', 'EmployeeCount']  
df.drop([col for col in cols_to_drop if col in df.columns], axis=1, inplace=True)
```

### # 4. Encode Categorical Columns

```
le = LabelEncoder()  
for column in df.select_dtypes(include='object').columns:  
    df[column] = le.fit_transform(df[column])
```

### # 5. Split into Features and Target

```
X = df.drop('Attrition', axis=1)  
y = df['Attrition']
```

### # 6. Train/Test Split

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

### # 7. Train the Model

```
tree_model = DecisionTreeClassifier(max_depth=5, random_state=42)  
tree_model.fit(X_train, y_train)
```

### # 8. Make Predictions

```
y_pred_tree = tree_model.predict(X_test)
```

### # 9. Print Accuracy & Classification Report

```
print("Decision Tree Accuracy:", accuracy_score(y_test, y_pred_tree))  
print(classification_report(y_test, y_pred_tree))
```

## # 10. Confusion Matrix Plot

```
cm = confusion_matrix(y_test, y_pred_tree)

plt.figure(figsize=(6, 4))

sns.heatmap(cm, annot=True, fmt='d', cmap='Blues',
            xticklabels=['No', 'Yes'], yticklabels=['No', 'Yes'])

plt.xlabel('Predicted')

plt.ylabel('Actual')

plt.title('Confusion Matrix - Decision Tree')

plt.show()
```

Decision Tree Accuracy: 0.826530612244898

	precision	recall	f1-score	support
0	0.88	0.93	0.90	255
1	0.25	0.15	0.19	39

accuracy		0.83		294
macro avg	0.56	0.54	0.55	294
weighted avg	0.79	0.83	0.81	294

