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Assessment Report

on

“Predict Employee Attrition”

submitted as partial fulfillment for the award of

**BACHELOR OF TECHNOLOGY
DEGREE**

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Introduction

Employee attrition is a key HR metric for organizations, and being able to predict it can help businesses retain top talent and reduce costs associated with turnover. In this project, we use a classification approach to predict attrition based on various employee features such as environment satisfaction, job level, and years at the company. Visual aids such as a confusion matrix heatmap and feature importance plot are used to interpret the results.

Methodology

1. **Data Loading & Cleaning:** Loaded the dataset and dropped non-informative columns such as EmployeeCount, EmployeeNumber, StandardHours, and Over18.
2. **Encoding:** Label-encoded categorical variables and the binary target Attrition.
3. **Splitting:** Data was split into training and testing sets using an 80-20 split while stratifying on the target.
4. **Model Training:** A Random Forest Classifier with 100 trees was trained.
5. **Evaluation:** Calculated confusion matrix, accuracy, precision, recall, and visualized results using seaborn heatmaps and bar charts.

Code:-

```
import pandas as pd

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report,
confusion_matrix, accuracy_score

import matplotlib.pyplot as plt

import seaborn as sns

# Load the dataset
df = pd.read_csv('6. Predict Employee Attrition.csv')

# Drop irrelevant or constant columns
df.drop(['EmployeeCount', 'EmployeeNumber',
'StandardHours', 'Over18'], axis=1, inplace=True)

# Encode target variable
df['Attrition'] = df['Attrition'].map({'Yes': 1, 'No': 0})
```

```
# Encode categorical features

categorical_cols =
df.select_dtypes(include='object').columns

le = LabelEncoder()

for col in categorical_cols:
    df[col] = le.fit_transform(df[col])


# Split data into features and target
X = df.drop('Attrition', axis=1)
y = df['Attrition']


# Train/test split
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42, stratify=y)


# Train Random Forest Classifier
clf = RandomForestClassifier(n_estimators=100,
random_state=42)

clf.fit(X_train, y_train)


# Predictions
```

```
y_pred = clf.predict(X_test)
```

```
# Evaluation
```

```
conf_matrix = confusion_matrix(y_test, y_pred)
```

```
print("Confusion Matrix:\n", conf_matrix)
```

```
print("\nClassification Report:\n",  
      classification_report(y_test, y_pred))
```

```
print("Accuracy Score:", accuracy_score(y_test, y_pred))
```

```
# Confusion Matrix Heatmap
```

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

```
sns.heatmap(conf_matrix, annot=True, fmt='d',  
            cmap='Blues',
```

```
            xticklabels=['No Attrition', 'Yes Attrition'],
```

```
            yticklabels=['No Attrition', 'Yes Attrition'])
```

```
plt.title('Confusion Matrix Heatmap')
```

```
plt.xlabel('Predicted')
```

```
plt.ylabel('Actual')
```

```
plt.tight_layout()
```

```
plt.show()
```

```
# Feature Importance Plot
```

```
importances = pd.Series(clf.feature_importances_,  
index=X.columns)
```

```
importances.sort_values(ascending=False).head(10).plot  
(kind='barh')
```

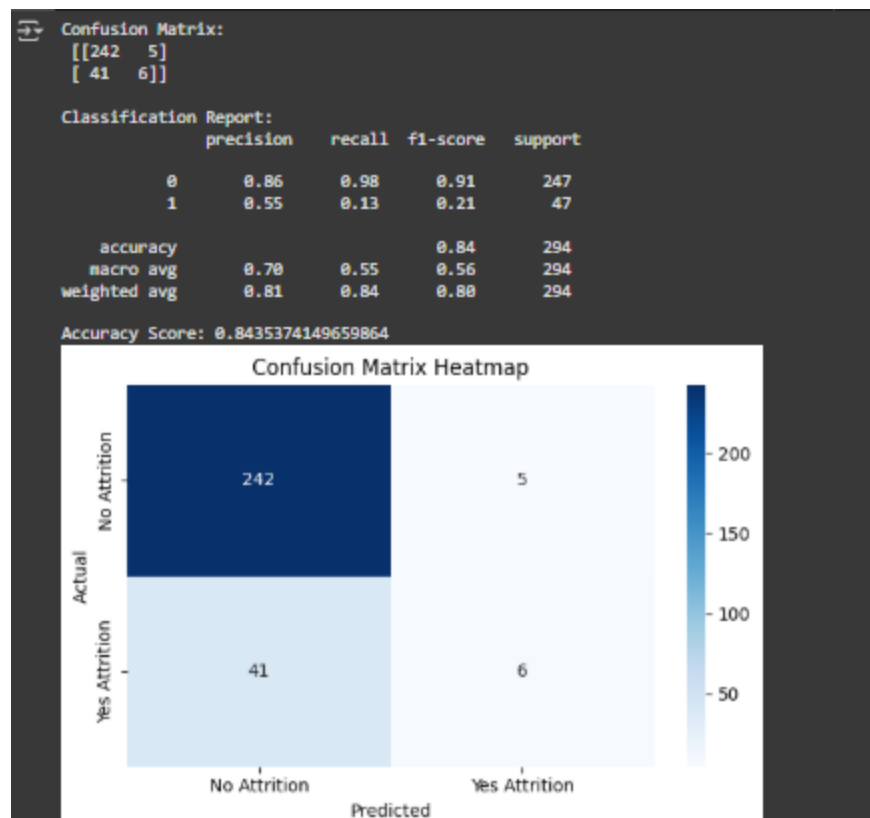
```
plt.title('Top 10 Important Features')
```

```
plt.gca().invert_yaxis()
```

```
plt.tight_layout()
```

```
plt.show()
```

Output:-



References/Credits

- Dataset Source: IBM HR Analytics Employee Attrition Dataset
- Libraries Used: pandas, sklearn, seaborn, matplotlib
- Classifier: RandomForestClassifier from scikit-learn