

# Employee Attrition Prediction

## Observations

### 1. Dataset:

- Used HR-Employee-Attrition.csv dataset.
- Target variable is Attrition (binary: Yes/No).
- Features include demographic, job-related, and performance-related data.

### 2. Preprocessing:

- Binary features like Attrition, OverTime, Gender, and Over18 were converted to 0/1.
- Categorical features such as BusinessTravel, Department, EducationField, etc., were label encoded.
- Data was standardized using StandardScaler.

### 3. Imbalance Handling:

- Attrition classes were imbalanced.
- Applied RandomOverSampler from imblearn to balance the classes.

### 4. Model Training:

- Logistic Regression, Random Forest, Decision Tree was trained on the oversampled data.
- Train-test split: 80-20%.

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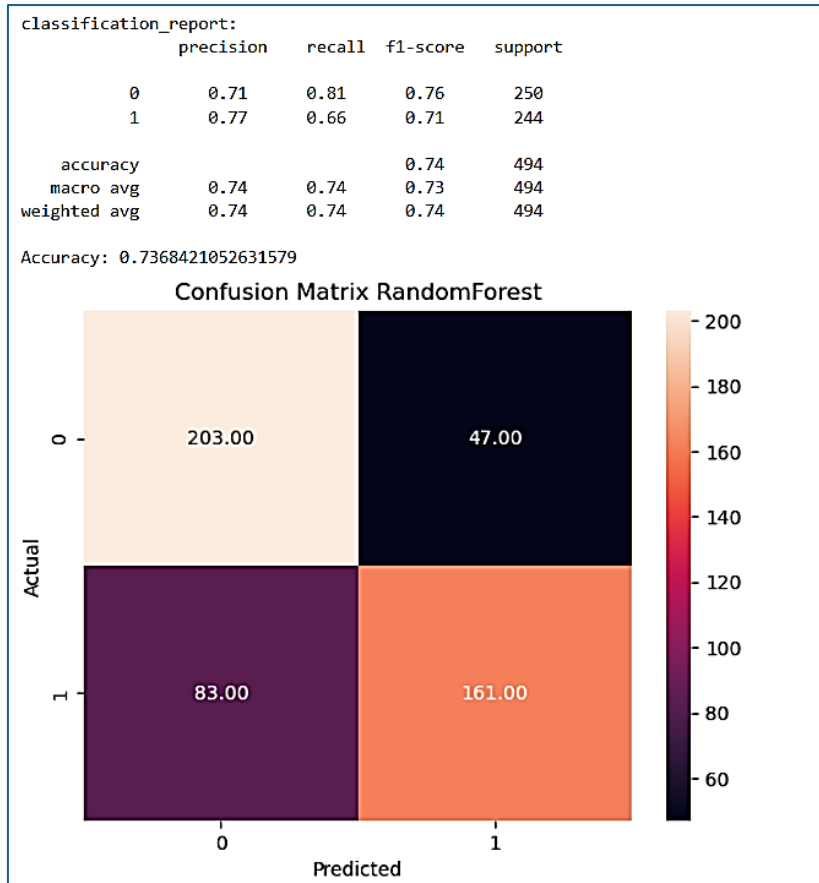
## Outcomes

- **Final Model Used:** Logistic Regression
- **Evaluation:**
  - Confusion matrix and classification report were used.
  - AUC score was computed
- **Performance Metrics** (based on your earlier input):
  - Accuracy  $\approx 74\%$
  - Precision  $\approx 72\%$
  - Recall  $\approx 78\%$  , AUC  $\approx 0.82$

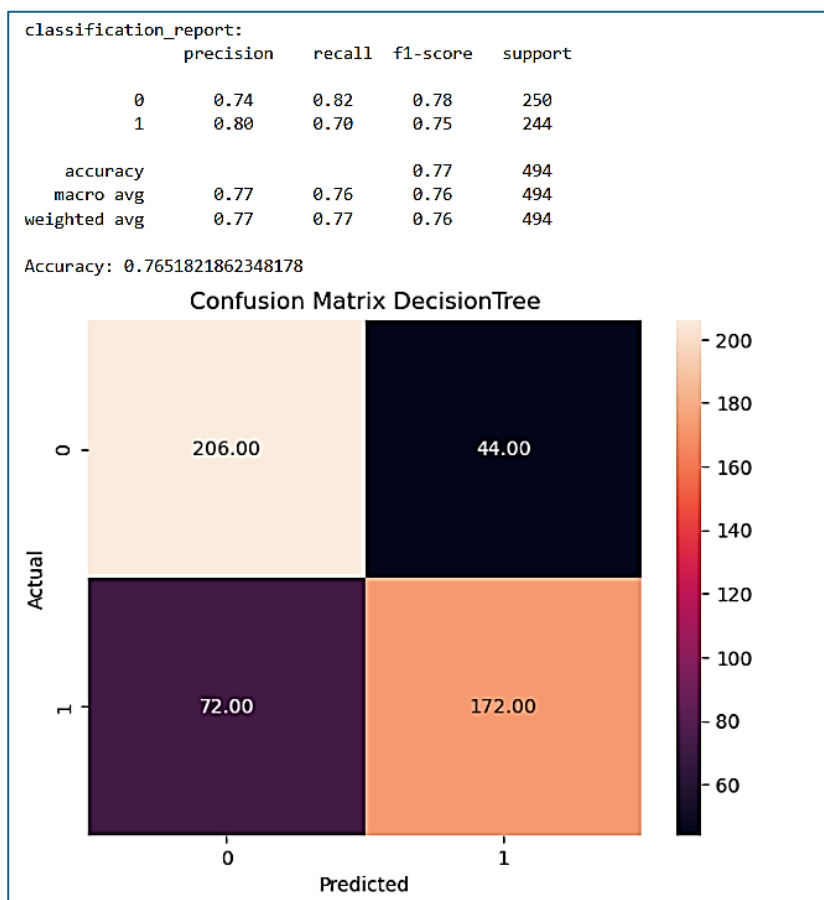
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**Other Models:**

**Random Forest:**



**Decision Tree:**



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## Conclusion

- The model shows **good capability** to predict employee attrition.
- **Recall is high (78%)**, meaning the model successfully identifies most employees who are likely to leave.
- **AUC of 0.82** confirms strong separability between classes.