



PREDICTING EMPLOYEES ATTRITION

Innocent Byiringiro

EMPLOYEE ATTRITION

Attrition



Turnover



Unfilled roles left
Roles elimination



OUTLINE

- Data Set
- Data Analysis
- Selected Model
- Show Me
- Interpretation
- References

DATA SET

Position

Role, Level, etc.

Educ. & Experience

Educ., working years

Job satisfaction

Distance from home, WorkLife bal.

Demographics

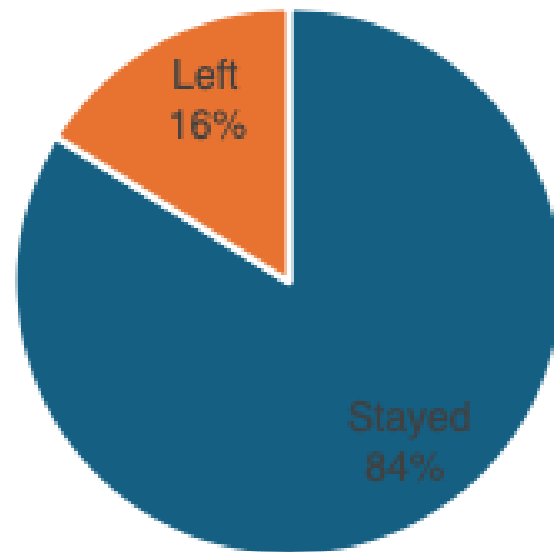
Gender, age, etc.

Financial security

Income, Monthly rate, etc.

DATA ANALYSIS

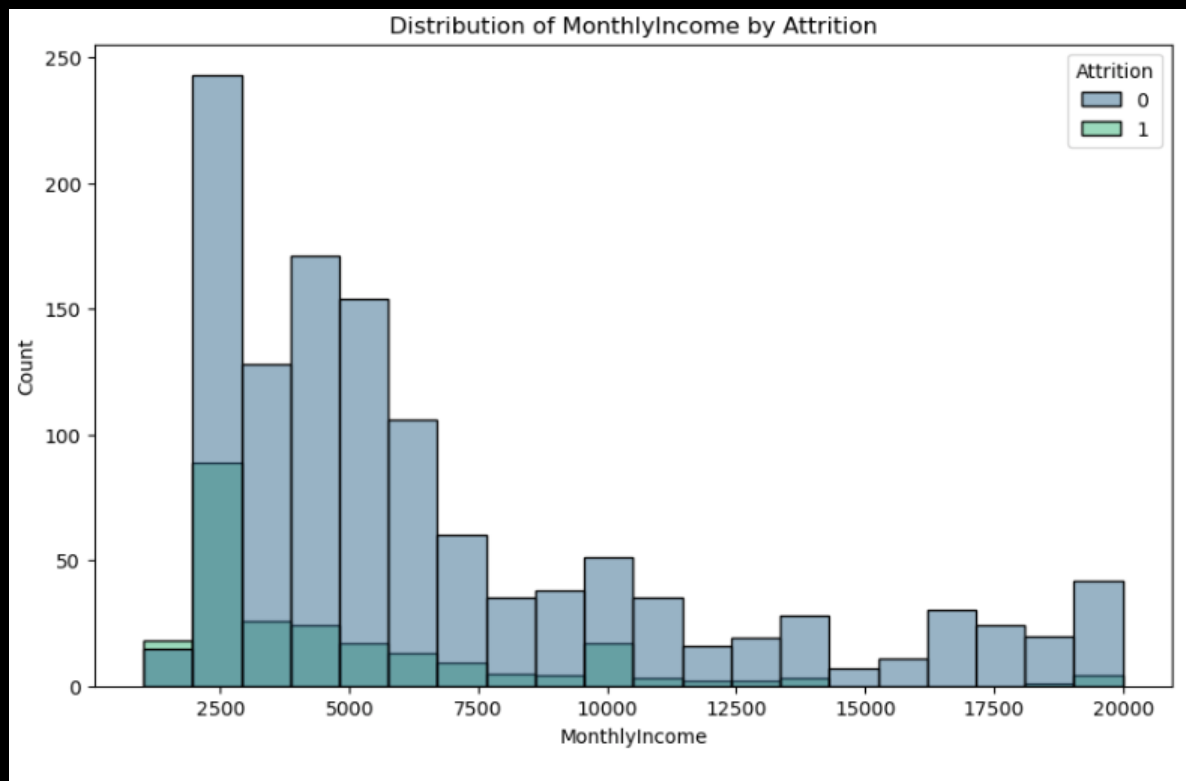
Dataset Employee Attrition



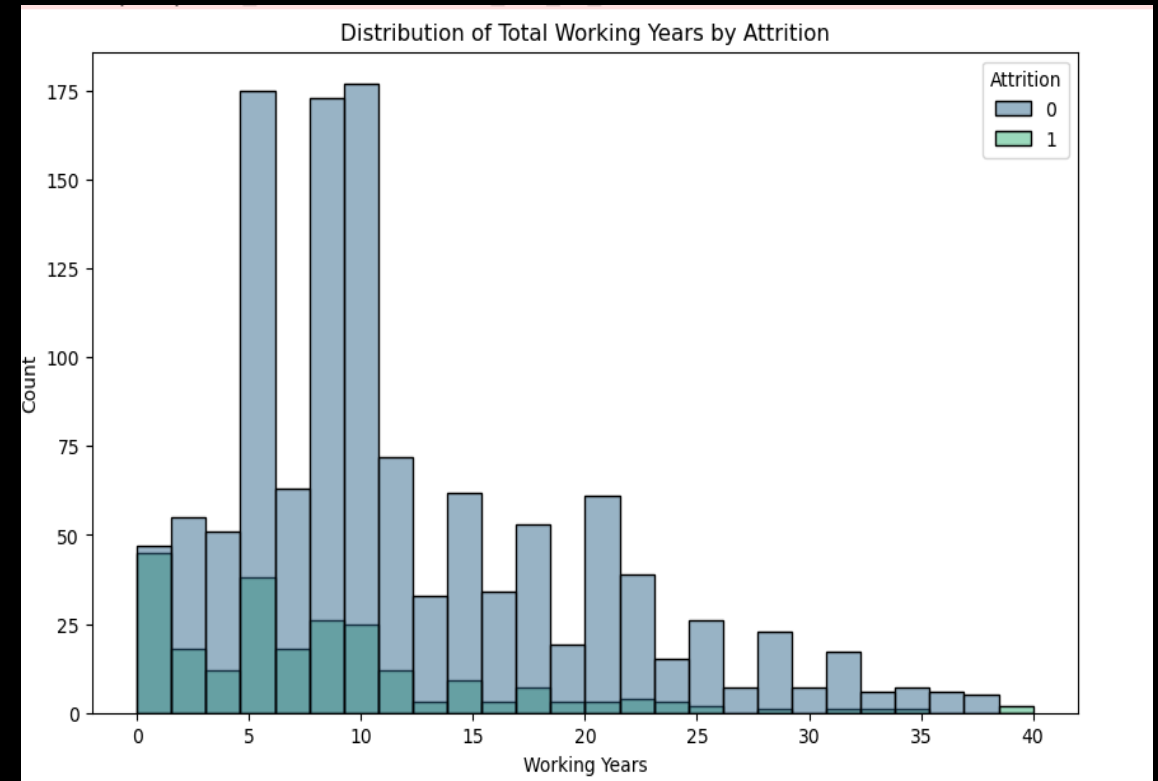
■ Stayed ■ Left

DATA ANALYSIS(CONT.)

ATT. VS MONTHLY INCOME

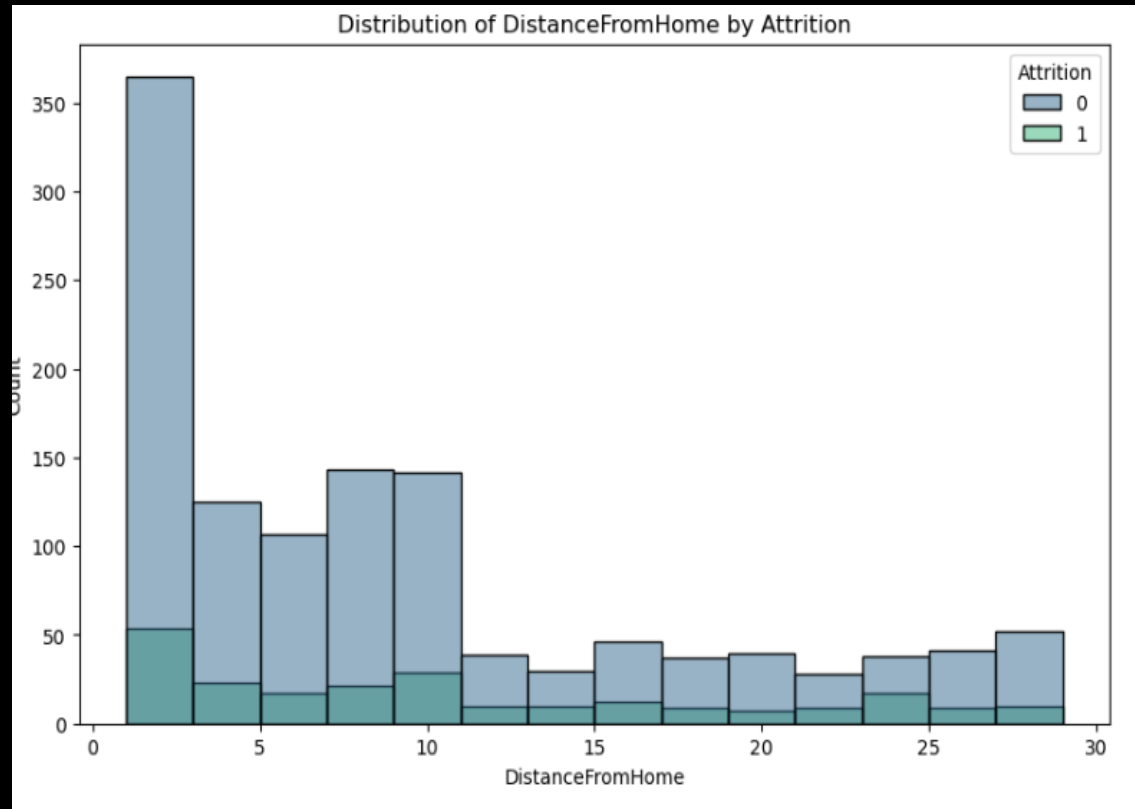


ATT. VS WORKING YEARS

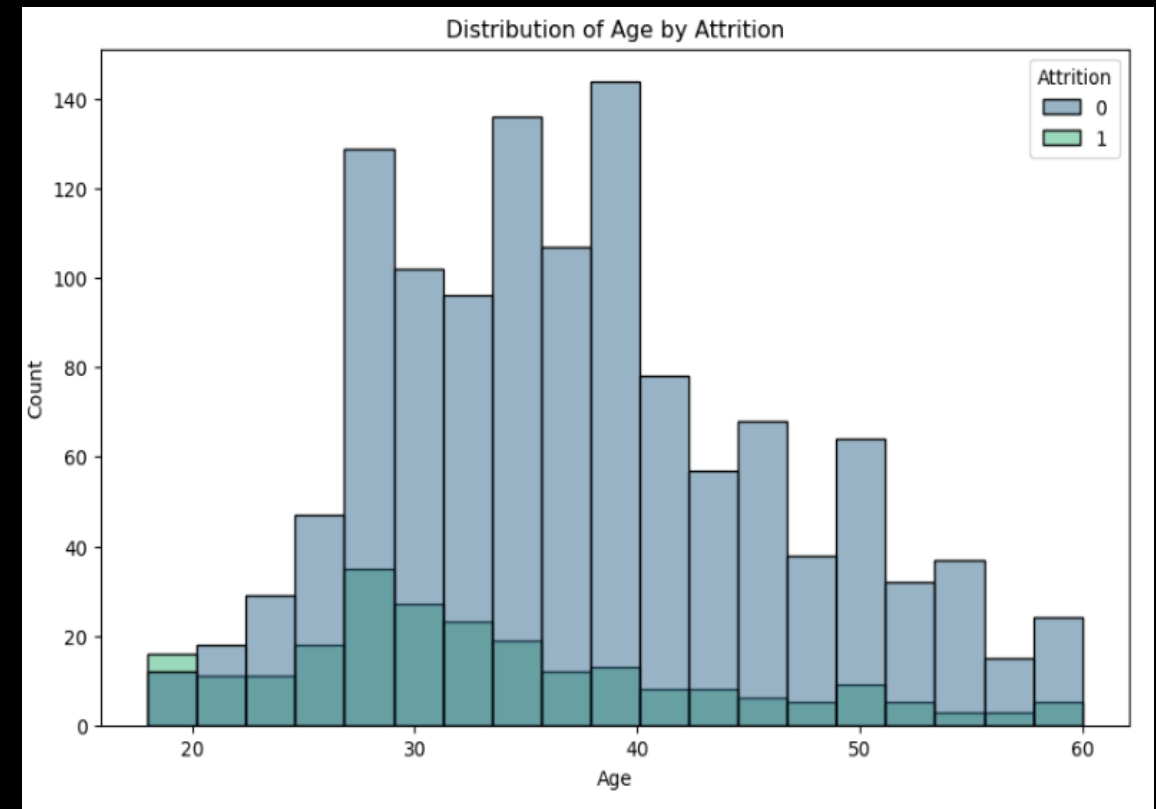


DATA ANALYSIS(CONT.)

ATT. VS DIST. FROM HOME



ATT. VS AGE





SELECTED MODEL

- Random Forest classification
- Handles imbalances without the need to apply other techniques such as oversampling and under-sampling
- These are metrics to evaluate the model: accuracy, precision/recall, f1-score

SHOW ME!

Input

Predicting Employee Attrition

Input Parameters



Factors



	TotalWorkingYears	JobLevel	YearsAtCompany	MonthlyIncome	YearsInCurrentRole	YearsWith
0	10	1	1	3,000	1	

Output

Predict

Prediction

The model predicts that the employee is likely to stay.

MOVING FORWARD

- The model's accuracy is 86%
- The macro averages (0.55 for precision, recall, and F1-score) reflect the overall performance across both classes
- There is imbalance and the model's struggle with class 1 (minority).
- Model is likely to spot if an employee will stay and less likely to spot if they will leave
- Broader use-case

REFERENCES



Inn-ctrl



Innocent Byiringiro



Inn-ctrl

- S. Flowers (1974). Why Employees Stay. Retrieved From: <https://hbr.org/1973/07/why-employees-stay>
- Dataset: Link: <https://www.kaggle.com/datasets/thedevastator/employee-attribution-and-factors>