

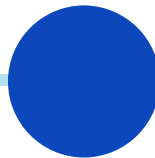
# Contents

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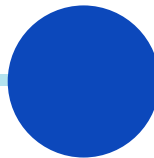
## Introduction

Project Overview and  
Goal"



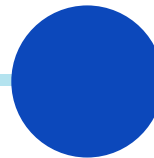
## Data Exploration

Exploring the Dataset



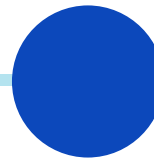
## Feature Engineering

Root Cause Analysis



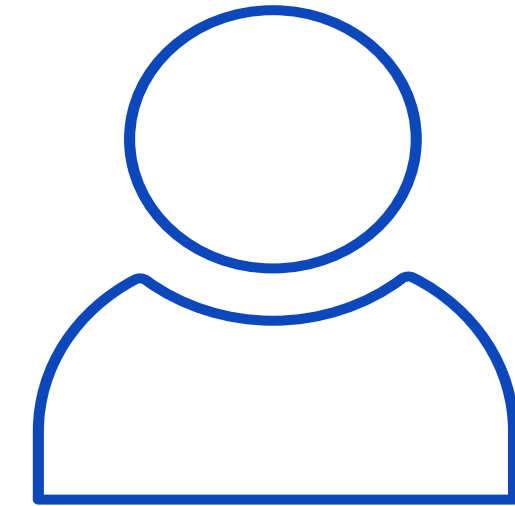
## Models

Evaluating  
Algorithm  
Performance



## Future

Data-Driven Insights  
for Business Decisions



## Situation

IBM is facing significant losses due to high employee turnover.

## Task

As consultants, our mission is to develop a machine learning model that predicts whether or not an employee will leave the company based on various factors.

## Action

We approached the problem as a binary classification task, with the model outputting either 0 (employee stays) or 1 (employee leaves).

## Result

Help IBM take proactive steps to reduce employee departures.

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# Understanding The Data

## Corporate Workforce Profile



- No missing values
- No duplicate values

- Average Age: **36 years**
- Gender Ratio: **1.7** Male vs. Female
- Job Satisfaction: **2.8** out of a range of 1 to 4
- Distribution of Employees by Departments :
  - Sales: **28%**
  - Human Resources: **2 %**
  - Research & Development: **70 %**

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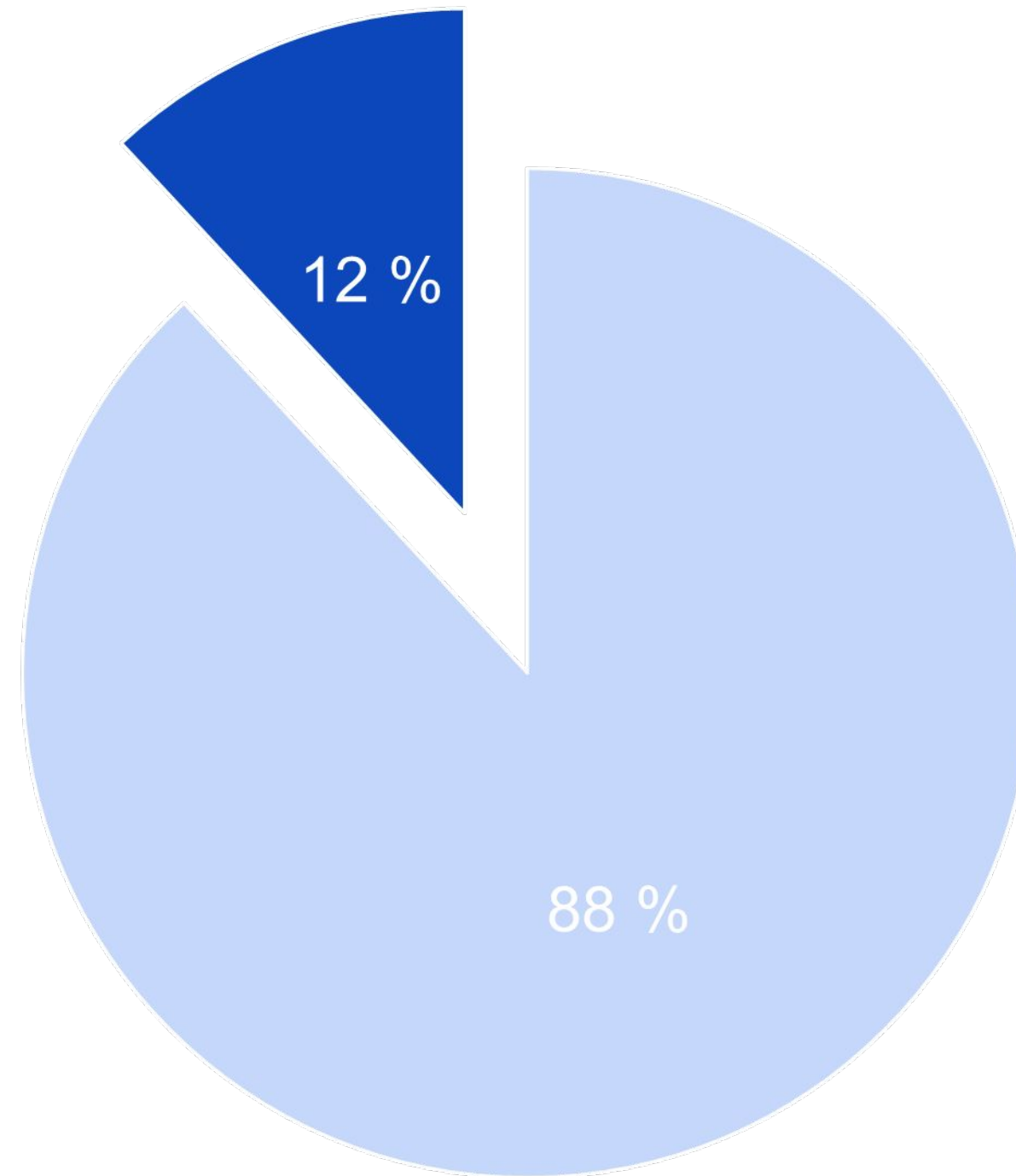
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# Understanding The Data



12 % of Employees Leave

■ Employees who stay  
■ Leavers

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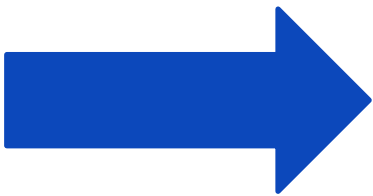
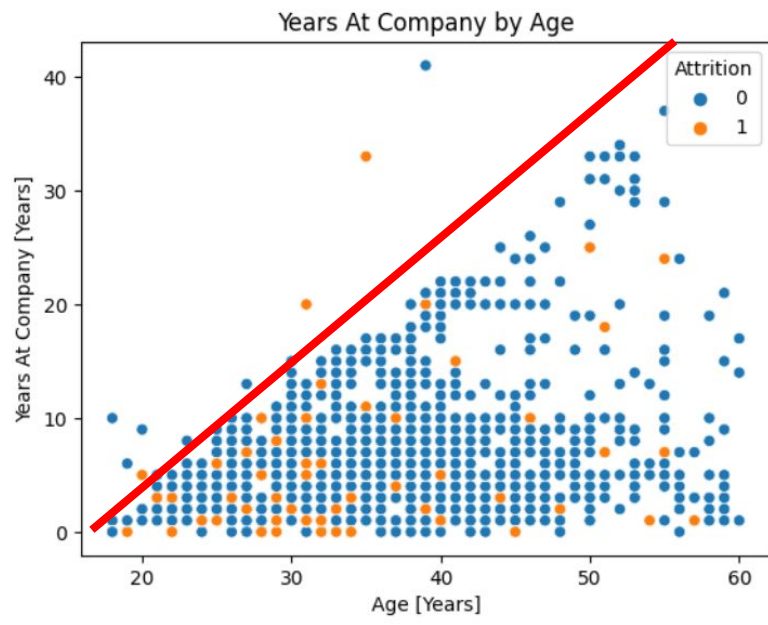
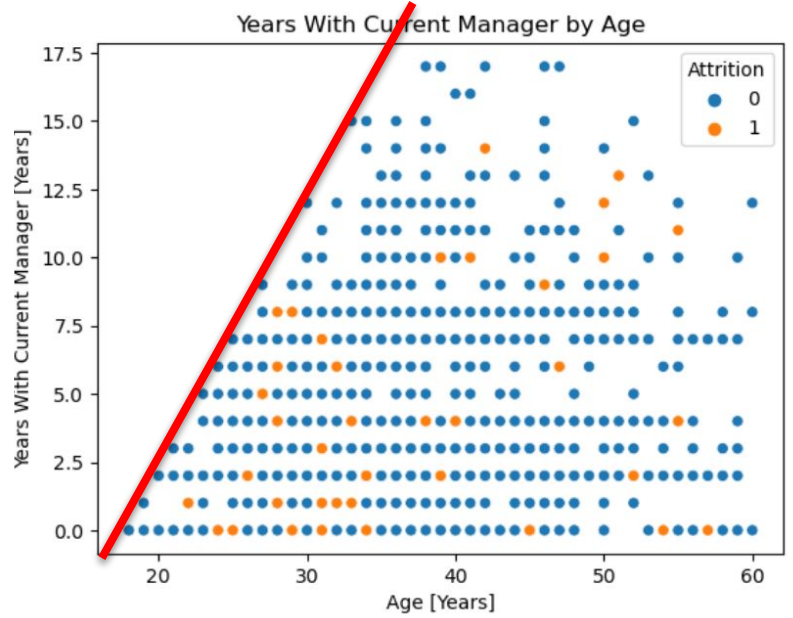
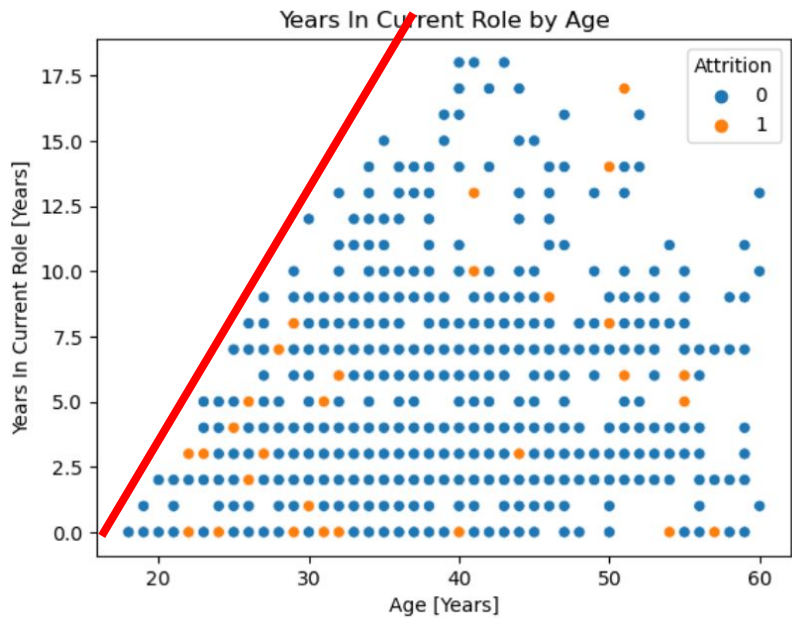
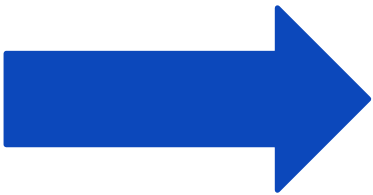
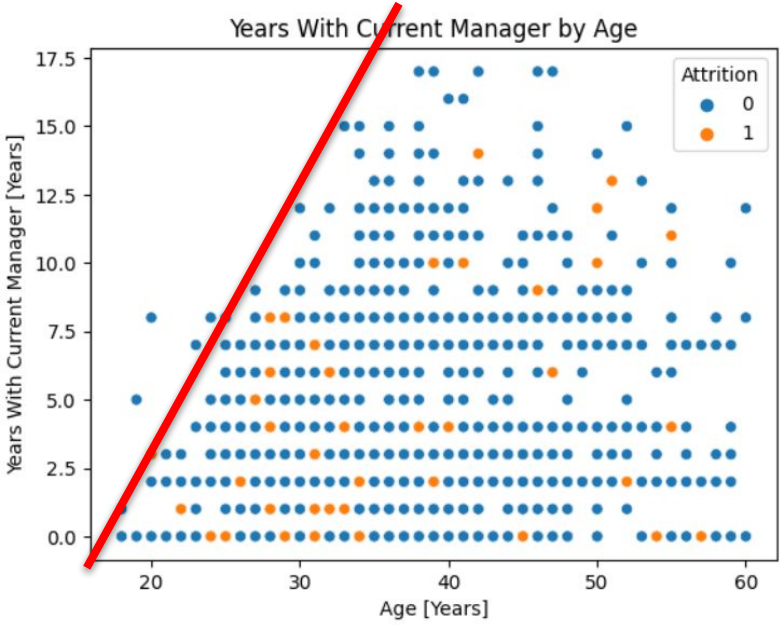
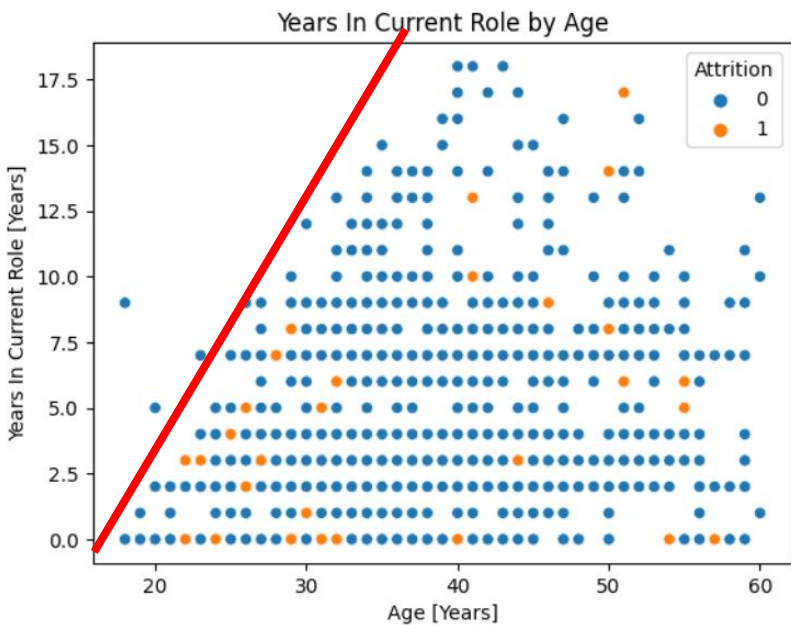
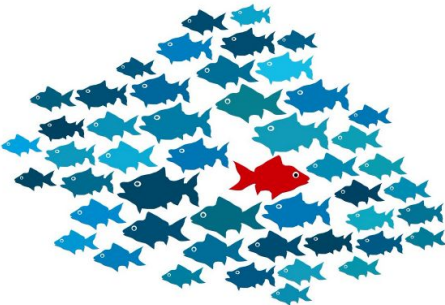
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# Handling Outliers



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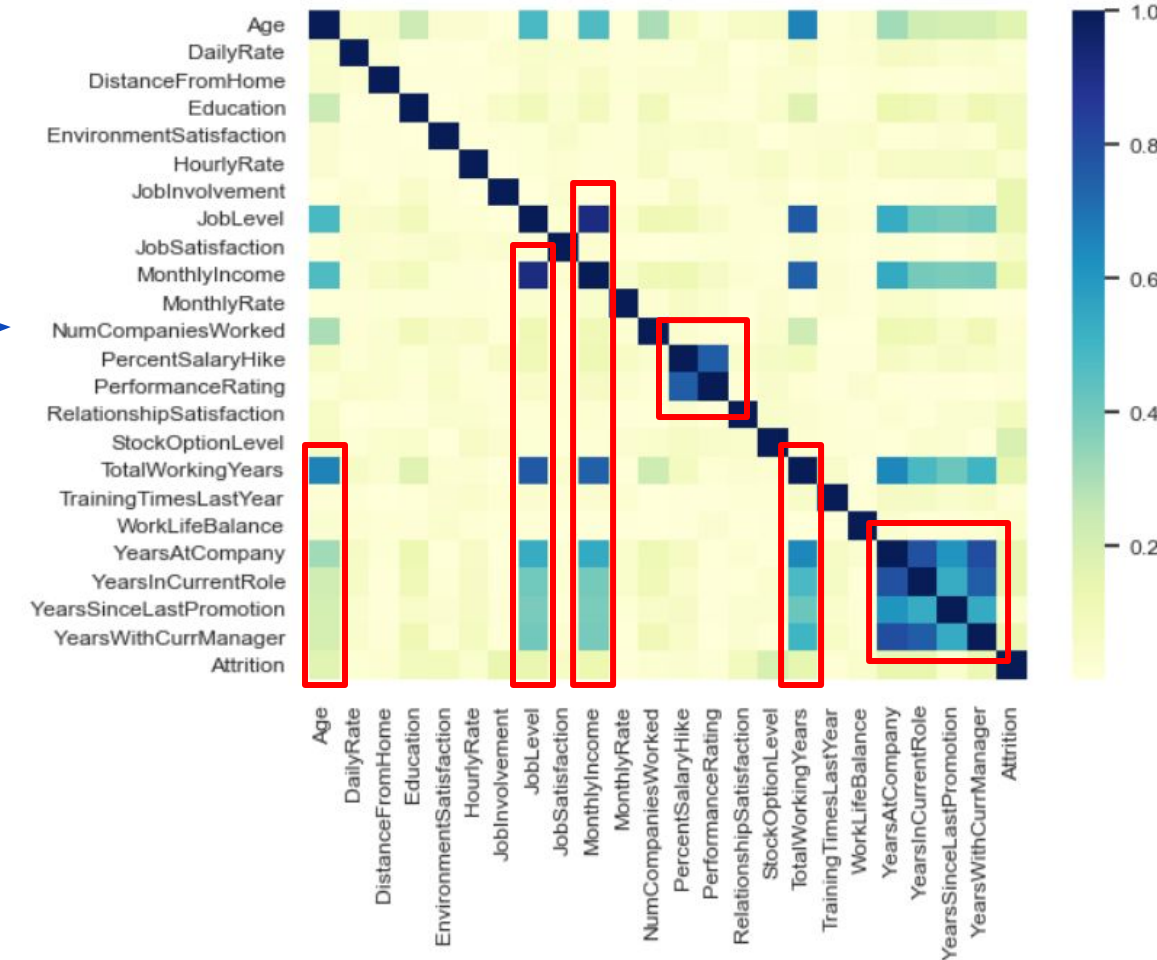


# Feature Selection

Correlation Matrix

Drop features whose domain is a single value:

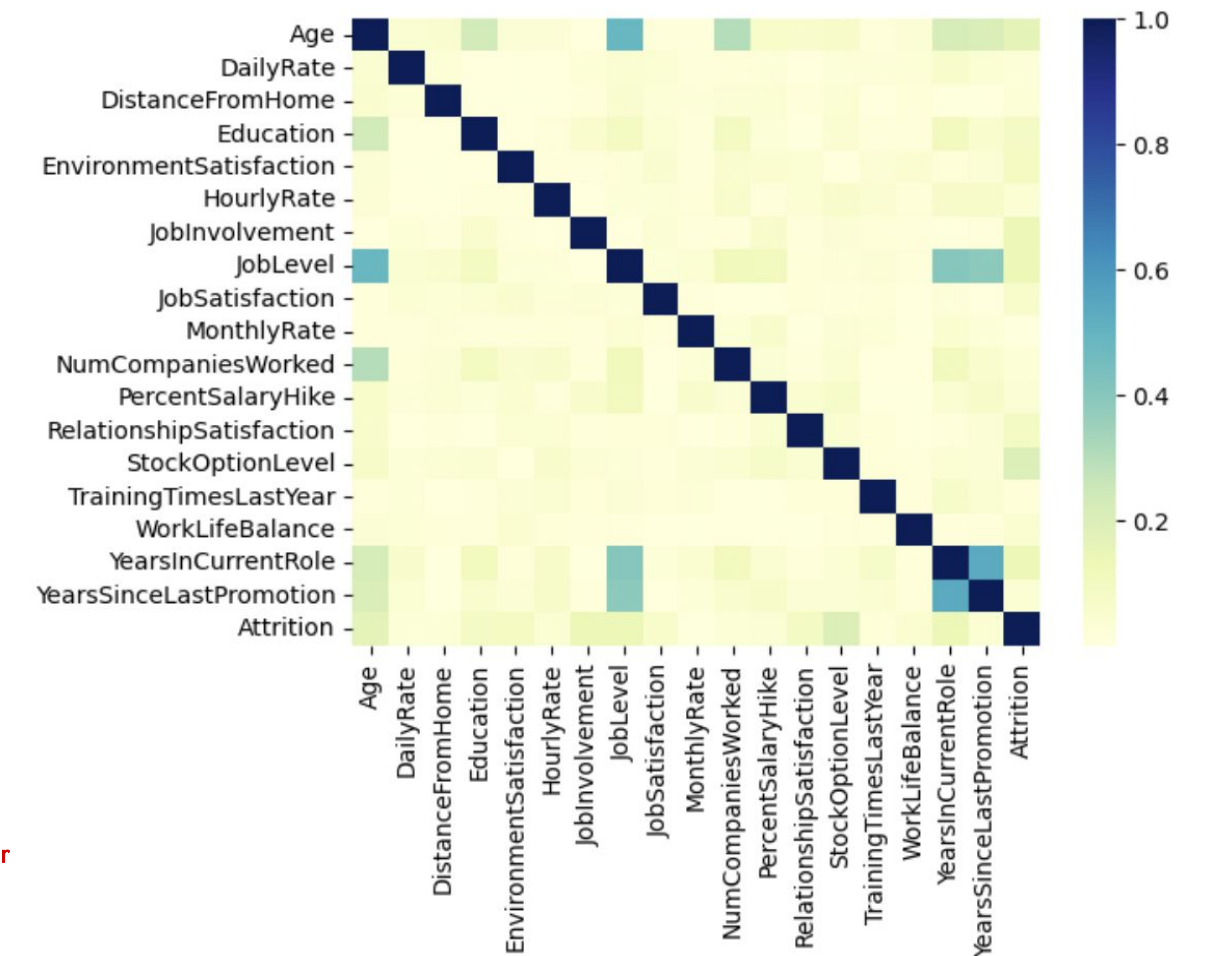
- **EmployeeCount**
- **Over18**
- **StandardHours**



Final Features

Drop:

- MonthlyIncome**
- TotalWorkingYears**
- PerformanceRating**
- YearsWithCurrManager**
- YearsAtCompany**



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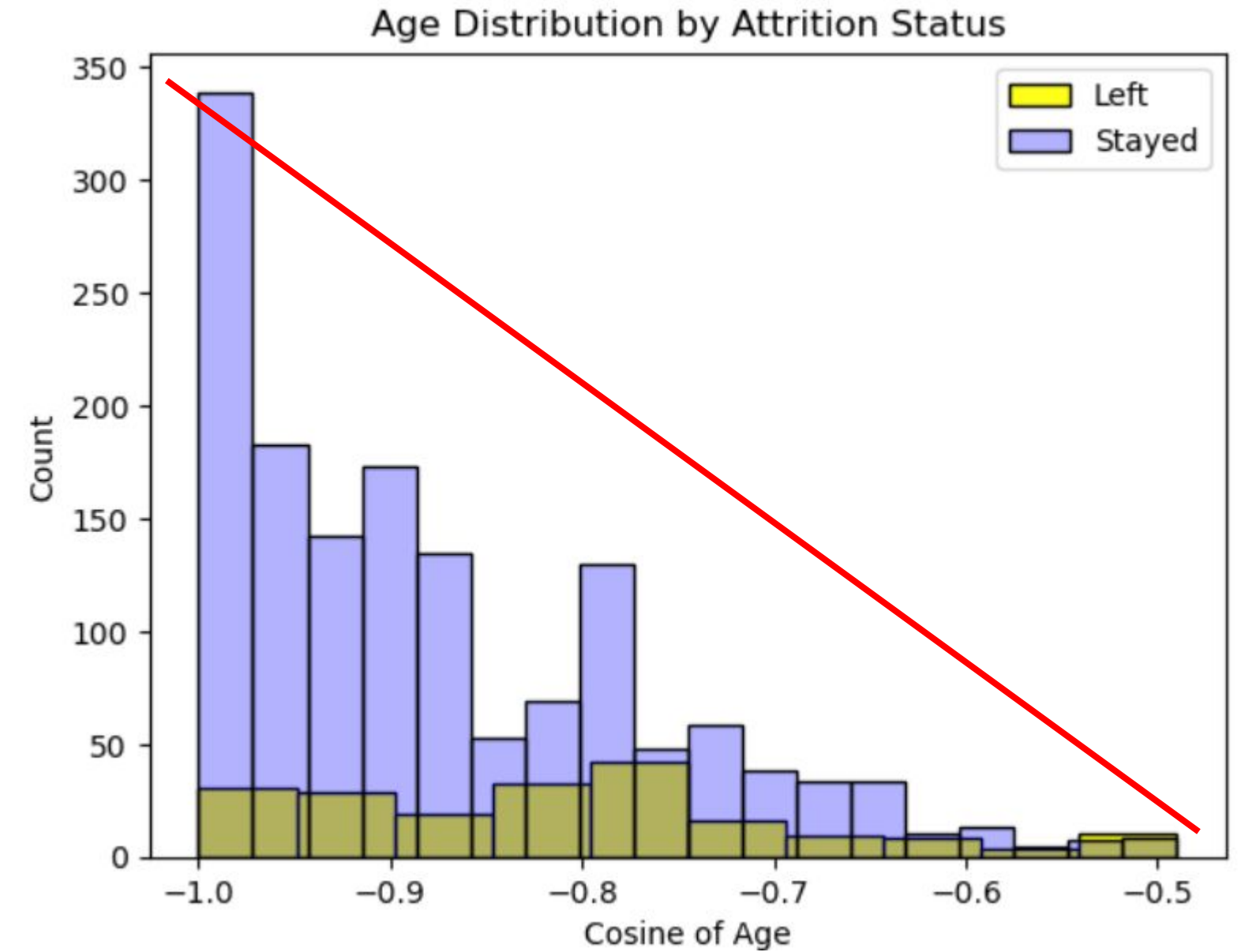
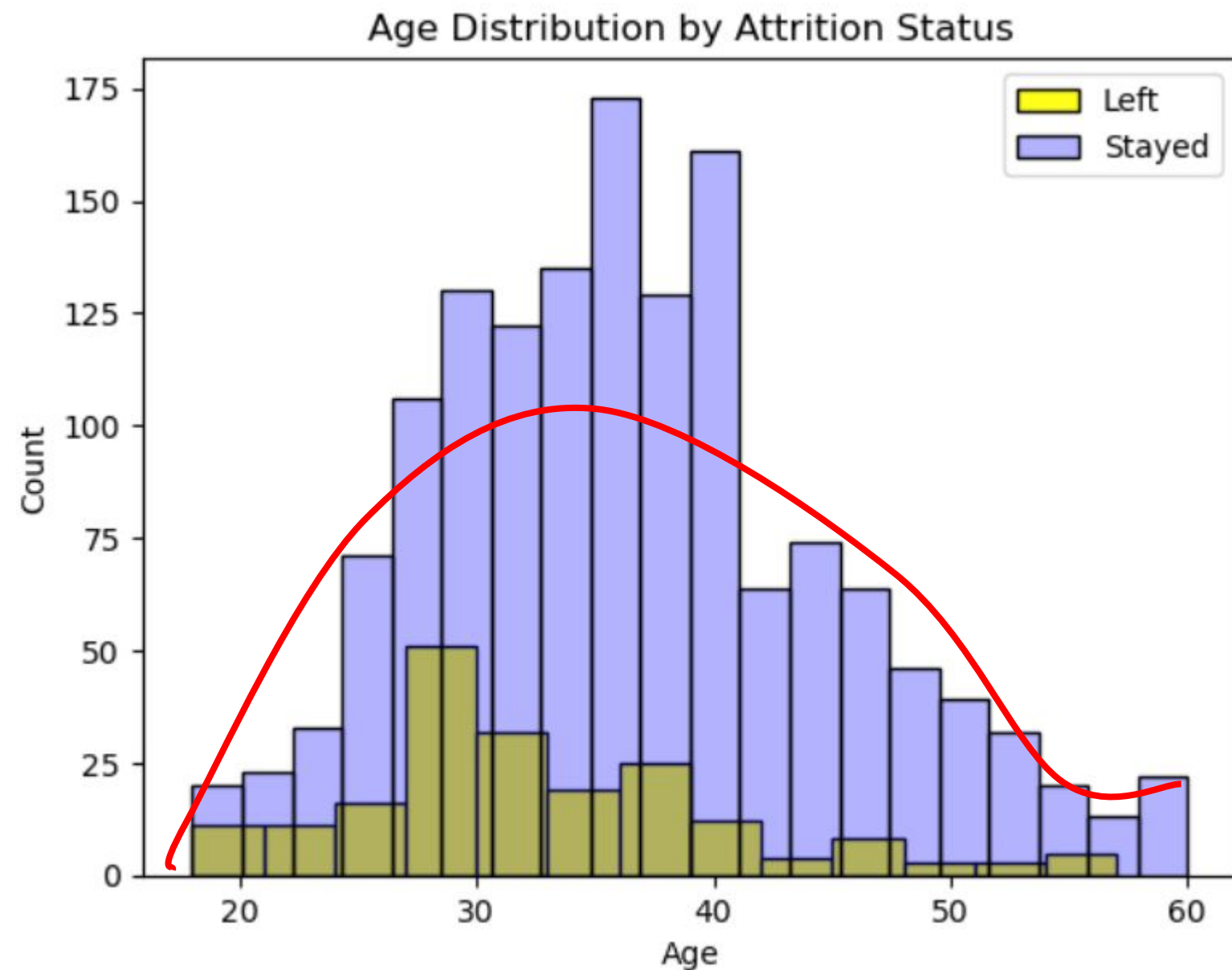
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# Feature Creation

Converting the Age into a Linear Feature



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# Feature Creation

## ONE-HOT-ENCODING (OHE) AND STANDARD SCALING

Data may be represented as **words**, **letters**, or **symbols**

Scaling is a common preprocessing step as most Machine Learning algorithms only process numerical data and require **standardized** Numerical Variables.

We did **OHE** on **Categorical Variables** and **Standard Scaling** on **Numerical Variables**

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# Models Trained

Logistic Regression

**Baseline Model**

Gaussian Naive  
Bayes

**Simple - Efficient**

Random Forest

**Flexible - Powerful -  
For High Number of Features**

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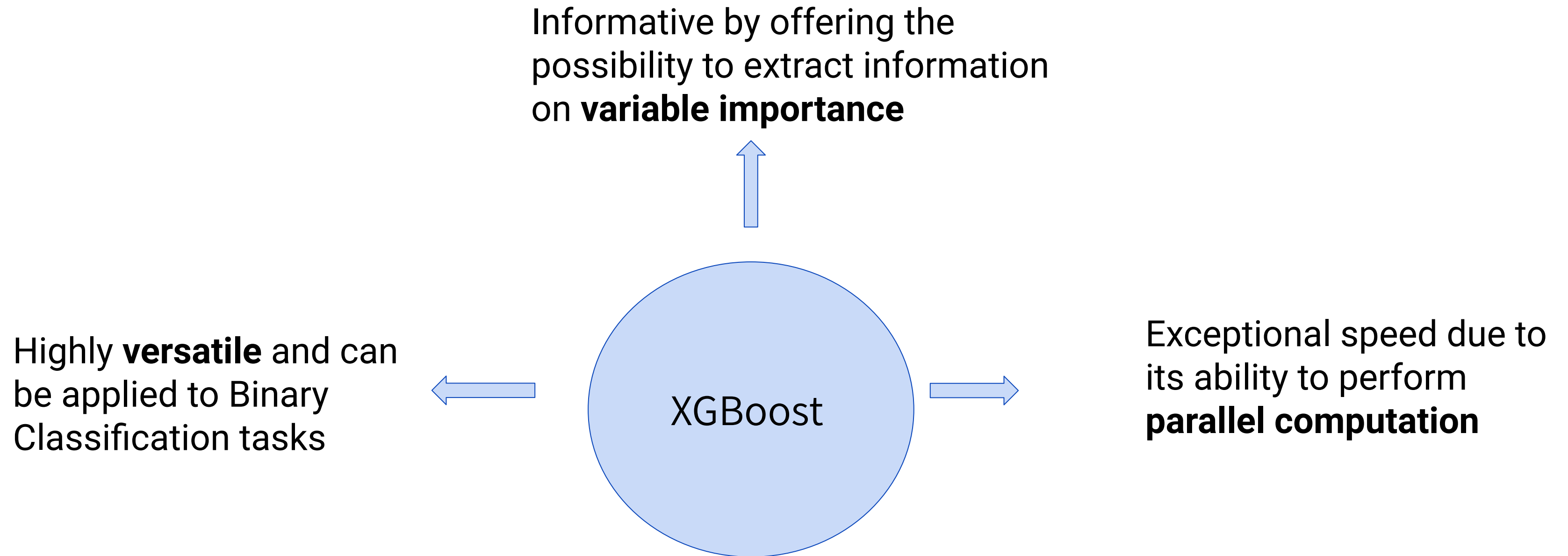
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# Why did We Choose the Model?



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# How Many Employees will Leave Next Year?

Predictions	
ID	
1677	1
1678	0
1679	0
1680	0
1681	1

→  
THRESHOLD = 0.12

Amount	
Predictions	
0	808
1	311

**311** is the Number of workers leaving the company

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# Final Conclusions

## Profile of **predicted leavers**:

- Average Age: **32 years**
- Gender Ratio: **1.9** Male vs. Female
- Job Satisfaction: **2.6** out of a range of 1 to 4
- Distribution of Employees by Departments
  - Sales : **27 %**
  - Human Resources : **4 %**
  - Research & Development : **70 %**

## Proposed Actions:

1. Implement **employee benefits** especially **targeting young professional in R&D**, such as company phones for private usage, a company gym, or free meals
2. Invest in employee development by creating a **career plan** for young professionals
3. Foster a **positive work environment** by defining strong company values

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Thank you !