# **Predicting Employee Attrition**

•••

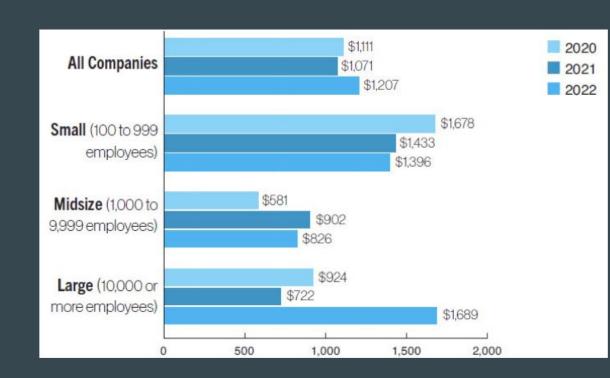
Fatih Demiroz
Capstone 2 Project Presentation

### **Problem: Attrition**

Companies spend 50 - 80% of their operating budgets on employees

Employee attrition costs companies

- money,
- time,
- productivity,
- culture, and
- institutional memory



### Context and Purpose

#### Context:

- CW Research and Analytics LLC is a medical and genetics consulting firm
- CW has experienced 16% employee turnover since last year (industry standard is 4%)

#### Purpose:

- Find the factors contributing to attrition
- Bring attrition to at or below 4% within 2 years

### Data

#### Data:

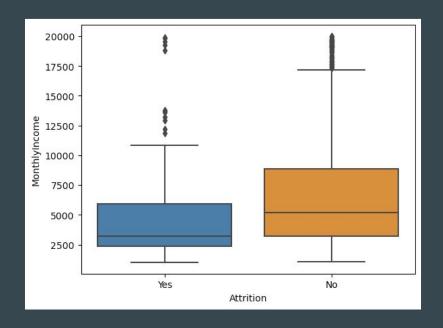
• Single CSV file with 35 columns and 1,470 rows.

#### **Table 1: Feature Properties**

| Column Name             | nn Name Data Value Column Name Type Type |          | Data<br>Type             | Value<br>Type |          |
|-------------------------|--|----------|--------------------------|---------------|----------|
| Age                     | Int64                                    | Discrete | MonthlyIncome            | Int64         | Discrete |
| Attrition               | Object                                   | Binary   | MonthlyRate              | Int64         | Discrete |
| BusinessTravel          | Object                                   | Ordinal  | NumCompaniesWorked       | Int64         | Discrete |
| DailyRate               | Int64                                    | Discrete | Over18                   | Int64         | Discrete |
| Department              | Object                                   | Nominal  | OverTime                 | Int64         | Binary   |
| DistanceFromHome        | Int64                                    | Ordinal  | PercentSalaryHike        | Int64         | Discrete |
| Education               | Int64                                    | Ordinal  | PerformanceRating        | Int64         | Ordinal  |
| EducationField          | Object                                   | Nominal  | RelationshipSatisfaction | Int64         | Ordinal  |
| EmployeeCount           | Int64                                    | Discrete | StandardHours            | Int64         | Discrete |
| EmployeeNumber          | Int64                                    | Discrete | StockOptionLevel         | Int64         | Ordinal  |
| EnvironmentSatisfaction | Int64                                    | Ordinal  | TotalWorkingYears        | Int64         | Discrete |
| Gender                  | Object                                   | Binary   | TrainingTimesLastYear    | Int64         | Discrete |
| HourlyRate              | Int64                                    | Discrete | WorkLifeBalance          | Int64         | Ordinal  |
| Jobinvolvement          | Int64                                    | Ordinal  | YearsAtCompany           | Int64         | Discrete |
| JobLevel                | Int64                                    | Ordinal  | YearsInCurrentRole       | Int64         | Discrete |
| JobRole                 | Object                                   | Nominal  | YearsSinceLastPromotion  | Int64         | Discrete |
| JobSatisfaction         | Int64                                    | Ordinal  | YearsWithCurrManager     | Int64         | Discrete |
| MaritalStatus           | Object                                   | Nominal  |                          |               |          |

# **Exploratory Data Analysis**

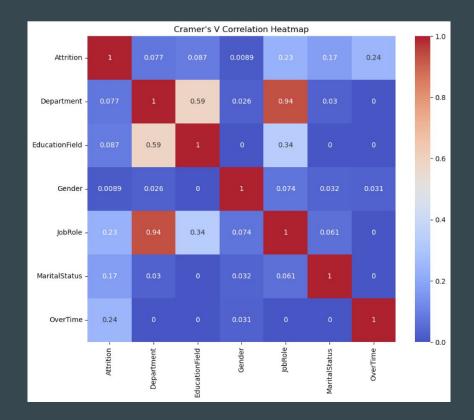
Monthly Income and Attrition



### Correlation Heatmap: Nominal Features

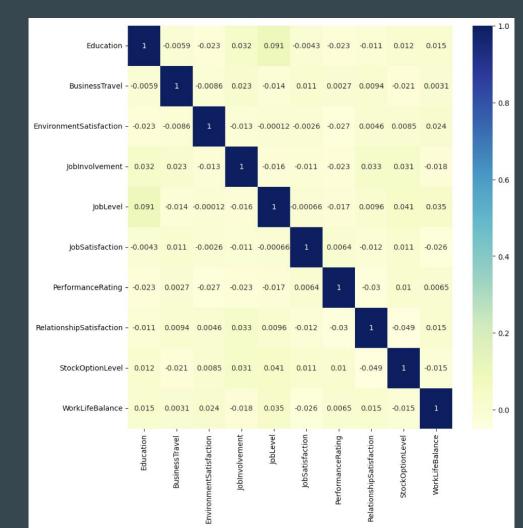
#### Some association

- Attrition JobRole
- Attrition Overtime



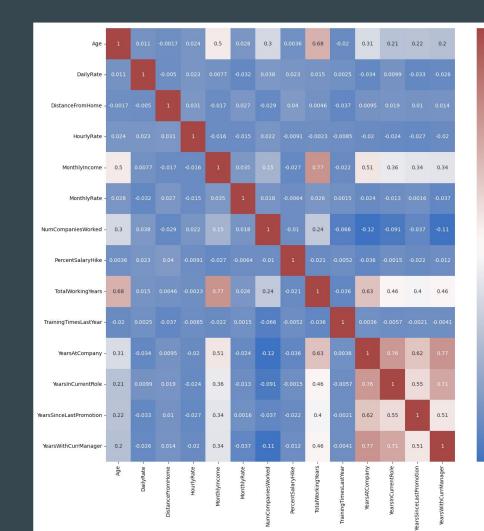
### **Correlation Heatmap: Ordinal**

No visible correlation between ordinal features



### **Correlation Heatmap: Numeric**

Moderate correlation between year related features



### Preprocessing

The dataframe split into three smaller dataframes for preprocessing

- Df for nominal features: converted to dummy variables
- Df for ordinal features: values were mapped and converted to numeric
- Df for numeric features: standardized and scaled using MinMax scaler

Then, these three processed dataframes were combined into one df for modeling phase.

# Modeling

Five separate classification models:

We have tolerance for False Positives

Best results from Logistic Regression

|                |              | Models |                  |                  |          |                        |
|----------------|--------------|--------|------------------|------------------|----------|------------------------|
| METRIC         |              | SVC    | Decision<br>Tree | Random<br>Forest | XG Boost | Logistic<br>Regression |
| Accuracy       |              | 0.89   | 0.83             | 0.87             | 0.88     | 0.88                   |
| Precision      | No Attrition | 0.89   | 0.89             | 0.87             | 0.88     | 0.9                    |
|                | Attrition    | 0.77   | 0.33             | 0.64             | 0.71     | 0.63                   |
| Recall         | No Attrition | 0.99   | 0.92             | 0.99             | 0.99     | 0.97                   |
|                | Attrition    | 0.28   | 0.26             | 0.11             | 0.2      | 0.36                   |
| F1-Score       | No Attrition | 0.94   | 0.9              | 0.93             | 0.93     | 0.93                   |
|                | Attrition    | 0.41   | 0.29             | 0.19             | 0.31     | 0.46                   |
| False Positive |              | 5      | 32               | 4                | 8        | 13                     |
| False Negative |              | 44     | 45               | 54               | 45       | 39                     |
| True Negative  |              | 375    | 348              | 376              | 372      | 367                    |
| True Positive  |              | 17     | 16               | 7                | 16       | 22                     |

# **Features Contributing to Attrition**

### Primary causes

- Years since last promotion
- Number of companies worked
- Overtime
- Job Roles
  - Lab Technicians
  - Sales Reps
  - o HR staff
- Business Travel

| Feature                           | Coefficient | % Increase in Odds of Attrition |
|-----------------------------------|-------------|---------------------------------|
| YearsSinceLastPromotion           | 1.569122    | 380.24                          |
| NumCompaniesWorked                | 1.543734    | 368.20                          |
| OverTime_Yes                      | 1.299739    | 266.83                          |
| MaritalStatus_Single              | 1.064923    | 190.06                          |
| JobRole_Laboratory Technician     | 0.947741    | 157.99                          |
| BusinessTravel                    | 0.855795    | 135.32                          |
| JobRole_Sales Representative      | 0.80235     | 123.08                          |
| EducationField_Technical Degree   | 0.758158    | 113.43                          |
| JobRole_Human Resources           | 0.566221    | 76.16                           |
| Gender_Male                       | 0.560651    | 75.18                           |
| EducationField_Human<br>Resources | 0.501213    | 65.07                           |
| Department_Sales                  | 0.480375    | 61.67                           |
| YearsAtCompany                    | 0.388956    | 47.54                           |

### **Features Reducing Attrition**

#### Factors that reduce attrition are:

- Years in current role
- Total working years
- Age
- No overtime
- Job roles:
  - Healthcare reps
  - Research director

| Feature                              | Coefficient | % Decrease in Odds of<br>Attrition |
|--------------------------------------|-------------|------------------------------------|
| YearsInCurrentRole                   | -1.49803    | -77.64                             |
| TotalWorkingYears                    | -1.34368    | -73.91                             |
| Age                                  | -1.156059   | -68.53                             |
| JobRole_Healthcare<br>Representative | -0.84335    | -56.97                             |
| JobRole_Research Director            | -0.793108   | -54.76                             |
| YearsWithCurrManager                 | -0.752833   | -52.9                              |
| OverTime_No                          | -0.609669   | -45.65                             |
| TrainingTimesLastYear                | -0.59165    | -44.66                             |
| Jobinvolvement                       | -0.464995   | -37.19                             |
| MaritalStatus_Divorced               | -0.423062   | -34.5                              |
| EnvironmentSatisfaction              | -0.3512     | -29.62                             |
| EducationField_Life Sciences         | -0.349265   | -29.48                             |

### Recommendations

- 1. Year gap between Promotions: Invest in employee training and development programs. Find ways to promote employees or show other pathos for career advancement
- 2. Overtime: Minimize overtime
- 3. Job Roles: Investigate complaints about pay and burnout for Lab technicians, Sales Reps, and HR staff.
- 4. Number of Companies Worked: HR shall prioritize job applicants with fewer job changes throughout their careers. HR should redflag the job applicants who change too many jobs in a short time period. Also, the C Suite should consider structural changes for accommodating the work habits of the Millennials and GenZ.

Q & A