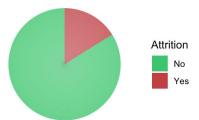
DDSAnalytics: Attrition Study

MSDS6306 - Lavonnia Newman, Jeff Washburn, & Joseph Caguioa



Agenda

- Business objectives
- Data source
- Methodology
- Evaluation/results
- Summary

Business Objectives

DDSAnalytics

- Analytics company that specializes in talent management solutions for Fortune 1000 companies
- Talent management is defined as the iterative process of developing and retaining employees

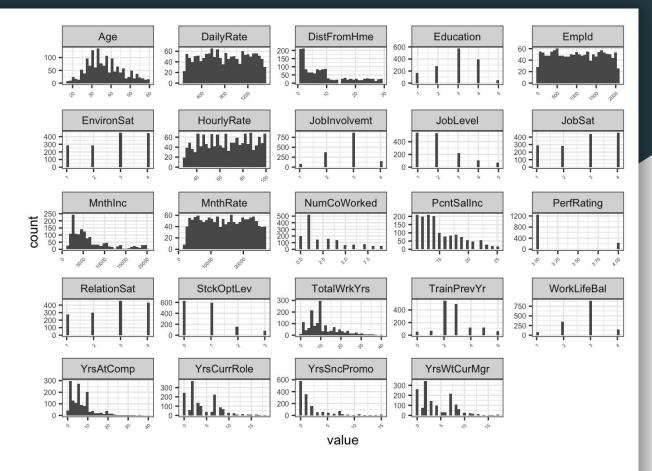
Objective

- To gain a competitive edge over its competition, DDSAnalytics is planning to leverage data science for talent management.
- The executive leadership has identified predicting employee turnover as its first application of data science for talent management

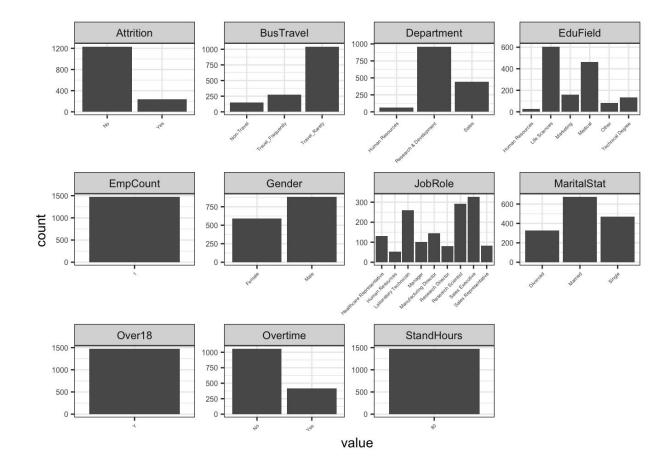
Data Source

Our team has been given a dataset (CaseStudy2-Data.xlsx) to conduct exploratory data analysis (EDA) to determine factors that lead to attrition

Raw Employee data set provided in csv format

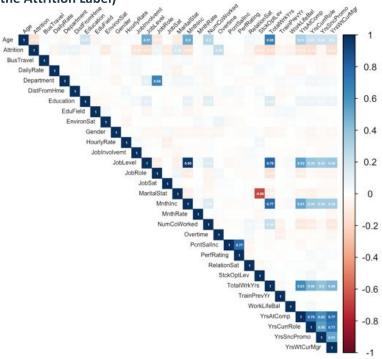


Data Source

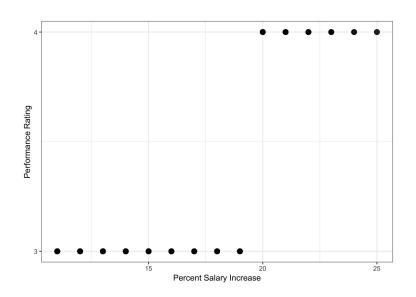


Methodology

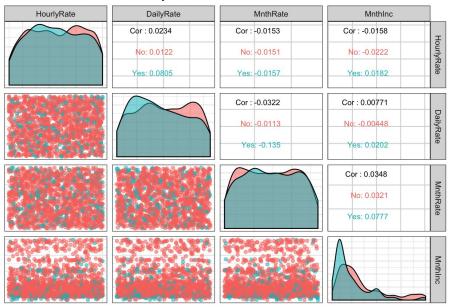
- Steps/Workflow
 - Variable Identification
 - Data size and shape (1470 observations, 35 Features including the Attrition Label)
 - Of the 35 Features
 - 1 Label: Attrition
 - 26 Numerical Features
 - 9 Categorical Features (1 being the Label of Attrition)
 - Univariate Analysis
 - Label
 - Numeric Features
 - Categorical Features
 - Bivariate Analysis
 - Attrition to Age
 - Attrition to Income
 - Insights using correlogram
 - Data Manipulation
 - **■** Feature Manipulation (log transformation on Income)
 - Create new features (life satisfaction)
 - Model creation / prediction
 - Linear Regression / Decision Tree / Random Forest



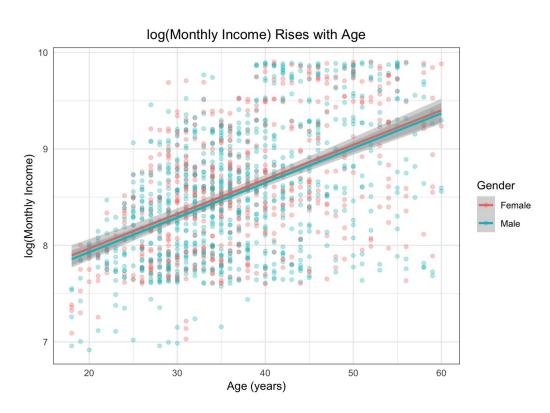
Results: Misc Trends



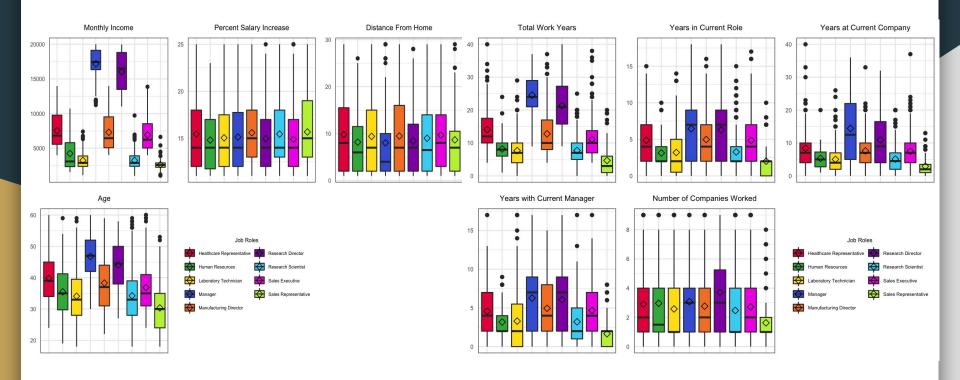
Correlation Matrix for Monetary Variables and Attrition



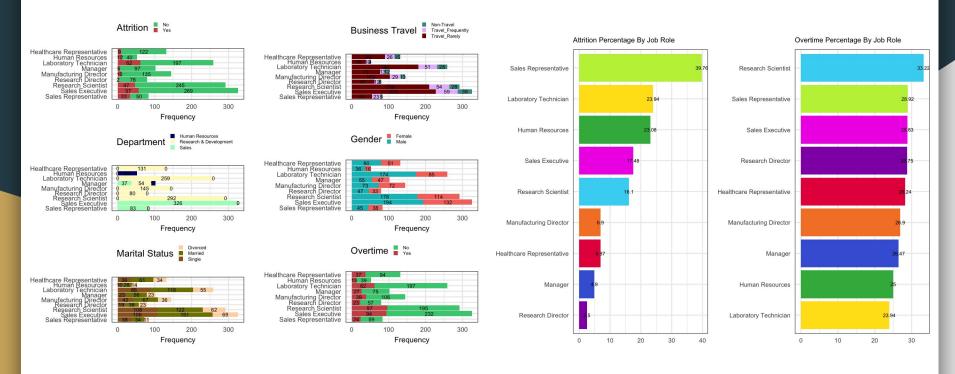
Results: Age & Income



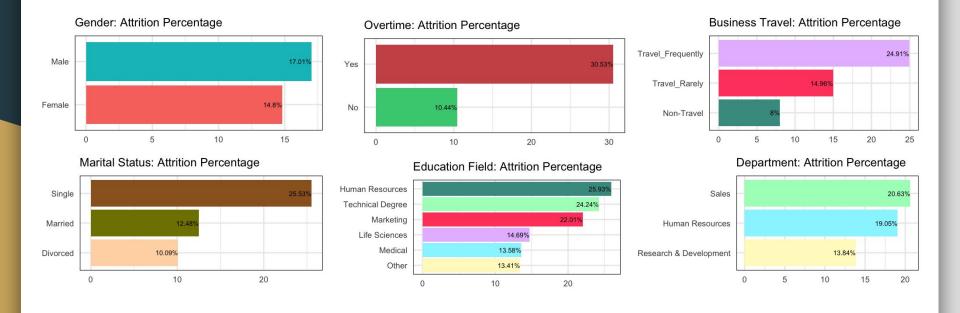
Results: Job Trends



Results: Job Trends



Results: Developing an Attrition Profile



Evaluation: Building a Predictive Model

Model	Accuracy	Sensitivity	Specificity
Logistic Regression	80.38%	67.80%	82.79%
k-Nearest Neighbors	82.83%	6.78%	97.40%
Decision Tree	85.29%	32.20%	95.46%
Random Forest	86.92%	27.12%	99.35%

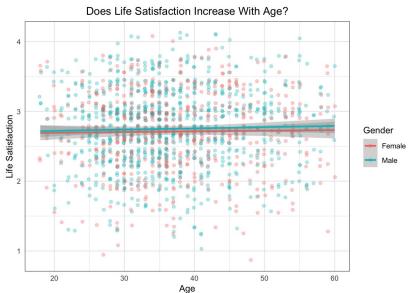
Results: Predicting Attrition

Random Forest: Top Predictors for Attrition=Yes

	Predictor	Level / Trend
1	Overtime	Yes
2	Job Level	1, 2
3	Stock Option Level	0, 1
4	Job Satisfaction	1, 2
5	Age	younger
6	Monthly Income	lower

Results: Life Satisfaction

Defined as μ(Job Satisfaction,
Relationship Satisfaction, Work-Life
Balance)





Summary - Insights

- 1. Component that we discovered during our analysis is that overtime was quite high among those employees that left.
- 2. The employees that had lower job level (1, 2) most likely the ones to leave
- 3. Those employees with stock options at level (0, 1) contributed to attrition
- 4. Those employees that rated their job satisfaction level 0, 1 most likely to leave
- 5. The younger aged employees fit the profile of those employees leaving
- 6. Employees that make a lower income are contributing to attrition

Summary - Recommendation

- Reduce Overtime
- 2. Create career Technical Track for employees not wanting to be managers
- 3. Offer stock options for new hires and refresh packages
- 4. Increase job satisfaction with meal incentives/travel incentives
- 5. Encourage 20% projects that allow employees to engage in other parts of the business
- 6. Be more competitive and aggressive by increasing starting salary for new employees

Summary - Recommendation

- Provide concrete definition for Life Satisfaction so that it can be included in future analysis
- 2. Define the Stock Option levels in more detail and what that means

Questions

Comments? Questions?