



HR Analytics Case Study Decode & Curb Attrition

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Business Objectives



Problem Statement:

• A 4000 employee Company XYZ that has 15% attrition wants identify the factors impacting attrition to reduce the costs as a result of it.

Business Goals:

- Find the factors affecting the 15% attrition rate
- Express the findings in terms of neat visualizations
- Find most important factors leading to attrition to be addressed on priority
- Suggest methods of minimizing the cause of attrition



Problem Solving Methodology



Data Sources

General Data

Employee Survey

Manager Survey Data

In_time_dat,
Outtime Data

Data Preparation

Analysis of In_Out data frame

Calculate mean attendance, Workload, regularity, Stats

Missing Values, treatment

Outliers and Invalid record treatment

Exploratory Data Analyses

Nominal Categories

Ordinal Categories

Interval Variables

Ratio Variables

Bivariate analysis

Variable Transformation

Scaling of variables

Converting variables to factors

Creating Dummy Variables

Model Building

StepAIC method to minimize insignificant variables

Iterating the model variables

Verifying the AIC and p-values of predictors

VIF values are iteratively measured

Model Evaluation

Sens, Spec, Acc at Cutoff

Gain, Lift and KS-Static verification



Data Preparation & Data Cleaning



Data Preparation:

• Employee_survey_data, Manager_survey data, Derived metrics from In_time_data and out_time data, and general_data data frames were merged to collate into employee_master data frame using merge() function.

Data Cleaning:

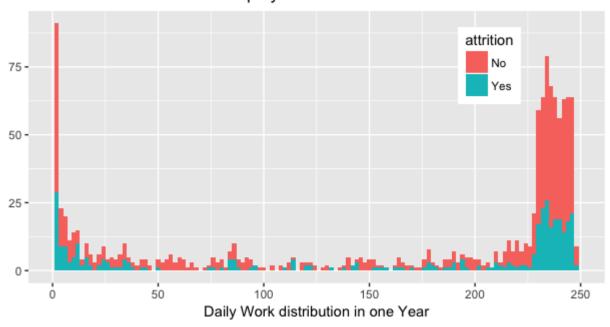
- Preparation of *in_out_duration* dataframe from from *in_data* and *out_data* dataframes.
- Computation of derived metrics of each employees mean_attendance, Workload, and Regularity.
- Identified the NAs in each of the columns and removed the rows wherever NAs were found.
- Performed WOE analysis to impute the NAs "TotalWorkingYears"
- Finally, data quality checks of invalid records were followed to get to the *employee_master_cleaned* dataframe.
- Exploratory Data Analysis on the cleaned data to visualize the behavior of continuous variables, categorical variables, Nominal variables, Ordinals, Ratio Values and Interval variable.



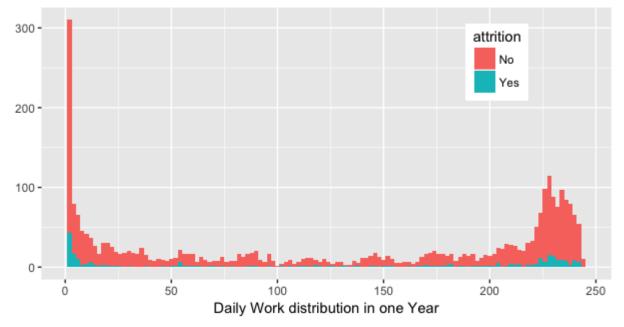
Univariate Analysis – Derived Metrics UpGrad



Workload Distribution of Employees in one Year



Regularity to Work of Employees in one Year



- 50 Percentile of data is spreads across 0 count indicating percentile of overworked
- 75 percentile is spread across 3 times overworked
- Rest fall under 100% overworked.

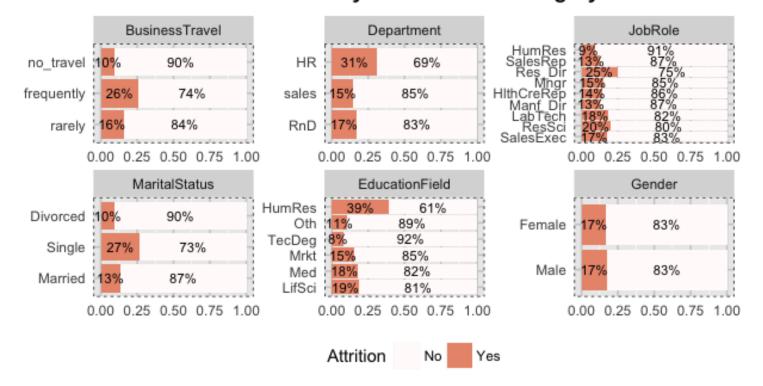
- 50% of data is < 20 times in 249 working days
- 75 % percentile of data is spread across 205 times,
- Rest 100% percentile is 249



Univariate Analysis – Nominal Variables



% Attrition by each Nominal category



Attrition rates:

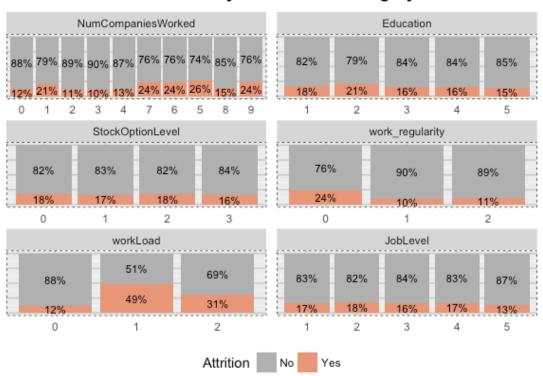
- Single & Married Employees
- Employees travelling Frequently and Rarely
- HR and R & D Departments
- Research Directors, Research Scientists, Lab Technicians
- HR degrees, Life Sciences, Medical degrees



Univariate Analysis – Ordinals



% Attrition by each Ordinal category



% Attrition by each Ordinal category



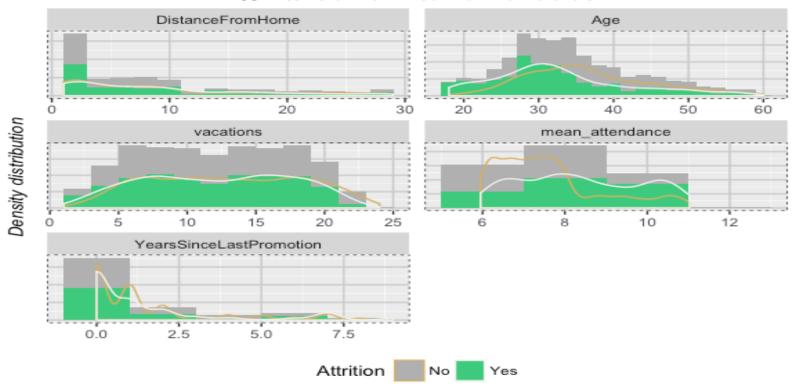
- Categories of employees with Low Performance Rating, Lesser Job Involvement, Low Environment Satisfaction, Job Satisifaction, No training received and lesser WorkLifeBalance are more likely part away
- Education levels, StockOptionLevels, JobLevels variables do not signify any attrition.
- Employes Heavily Worked, and under worked, irregular to work shows higher attrition over other factors indicators.



Univariate Analysis – Interval Variables







Attrition rates from Data:

- Employees in the Age group of 24-35 are highly like to part way
- Employees closest to the work location are also like to part away
- Mean attendance > 10 or < 7 are the ones that frequently have attrition.

Attrition rates from Data:

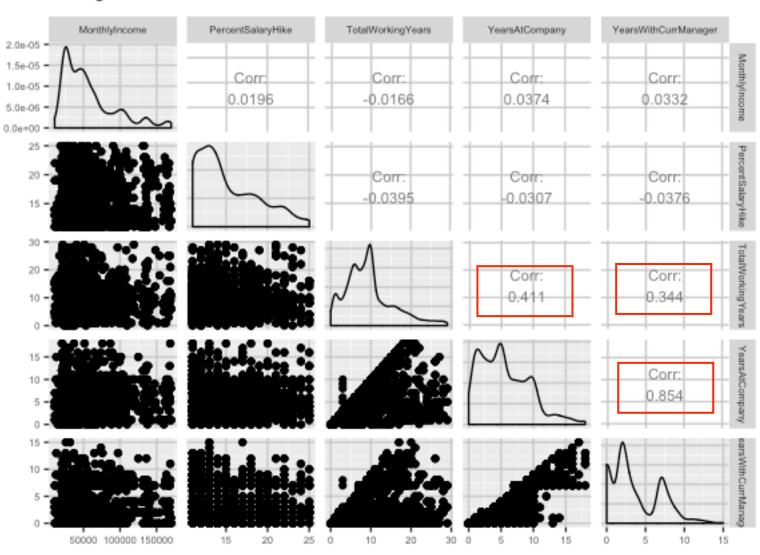
- Employees with no promotion for 7 years, 9 years and 15 years and very likely to part way
- Large portion of this concentrated between0 to 2.5 years



Bivariate Analysis- Ratio Variables



Plotting correlations between variables



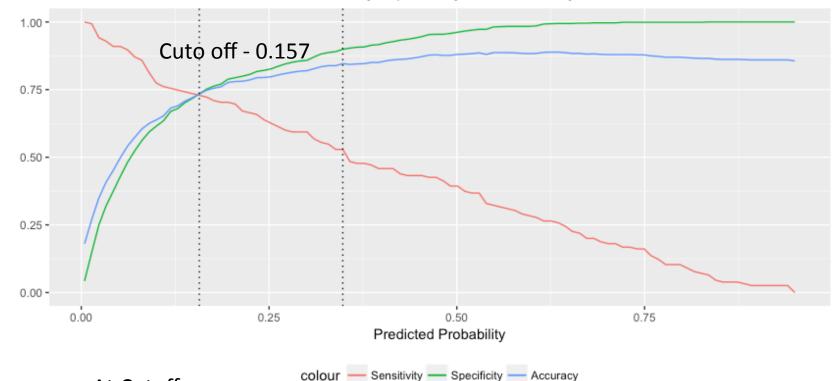
- 85.4% employees with YearsatCompany work as associated with Currentmanager
- 41.4% YearsSinceLastPromotion
 & YearWithCurrManager (44.6%)
- Of the total work experience,
 41.4% employees spent time in the company



Result – Model Performance







- Attrition of True Positive or True Negative is 73.3% accurately Predicted by the model.
- Model predicts "Yes" Value
 72.9% of times among available
 "Yes"
- Also, model predicts "Nos" in 73.43% times of all Nos.

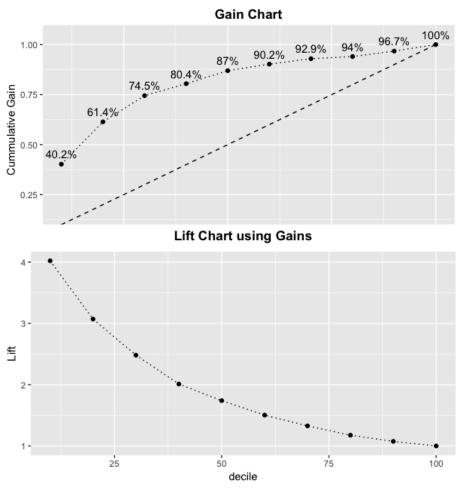
At Cutoff:

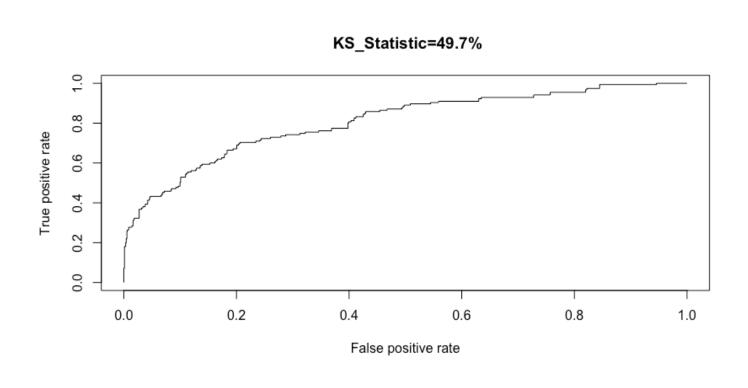
Accuracy : 0.7335
 Sensitivity : 0.7290
 Specificity : 0.7343



Model Evaluation Gain and Lift Charts







- With KS- Statistic 54.7%, Area under the curve is 84%. model has greater ability to predict the Attrition
- With Progressive Lift predictive model scales very well against the random model,
- With each passing by decile, the gradual rise in cumulative decile scales very well with the model.



Conclusion



- Age Higher Attrition for Lesser Age, It is higher fro the age groups beween 26-35.
- Business Travel Frequently and Rarely,
- **R&D** and Sales Department have negative relationship with Attrition. The JobRoles of the Department such as Lab Technicians and Research Director and Scientists are the one that are frequently looking for moving out.
- Employees with Marital_Status Singles are the ones who are frequently look out.
- Resources with NumberCompaniesWorked and Total Working Years are Correlated are directly. This impacts the attrition to some extent and the higher the number of companies a resource works, the higher is the attrition. Specifically, between 1 6,7,8
- Years with Current Manager and Years Since last promotion are significantly correlated with Each other. As an employee works with the CurrentManger attrition is apparently is showing the reduced trend.
- Employees with mean_attendance > 10 hours of workload are more prone for lookout and perhaps the reason be a stressful workenvironment and it is clear from the data variables where workload_1 and WorkLoad_2 that demonstrate higher positive values indictor to strong predictors.
- Further, employees with poor worklife balance, and poor Job Satisifaction, poor values of Environment Satisfaction and and lower Job_involvment_3 have Higher attrition. Thus there variables have a inverse relationship with attrition.