

Members:

2.





OBJECTIVE:

• To derive the factors which is the main cause for the increase of Attrition in the organization and to provide solution to curb the same.

APPROACH STRATEGY:

To analyze historical data and identify employees who might leave the organization.

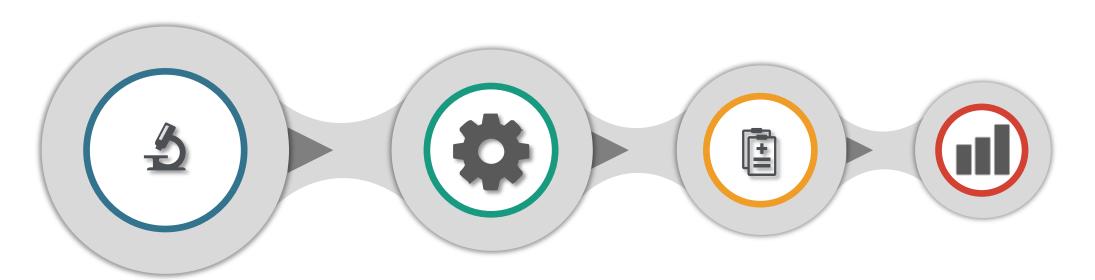
•PROBLEM SOLVING METHODOLOGY:

- Data Cleaning
 - •Eliminating column having NA's, single value & reducing the data file for further analysis.
 - •Standardizing Date format and removing special characters symbols like "%".
 - •Removing Outliers.
 - •Combining the employee data from different data sources
- Applying logistic regression methodology to gain the insights



DATA FLOW





DATA UNDERSTANDING AND PREPARATION

(Identifying unique keys for given data frames, Collating files and checking for Data Quality issues)

DATA CLEANING

(DataCorrection, Data Conversion and Data Manipulation)

DATA COMPUTATION

(Computing the attrition ratio)

DATA ANALYSIS

(Analyzing various factors related to employee leading to high attrition)





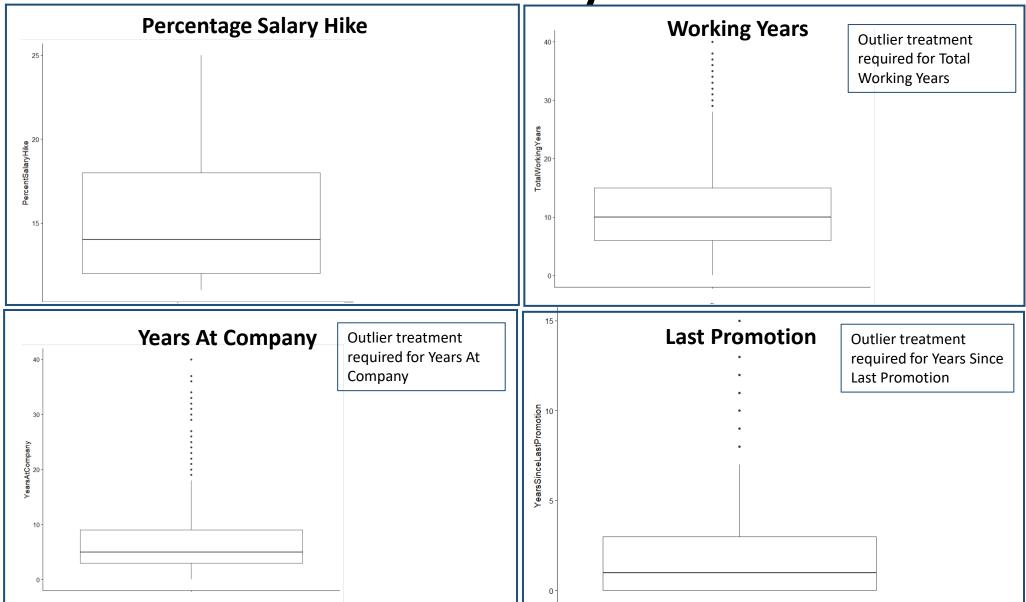
Outlier Analysis







Outlier Analysis



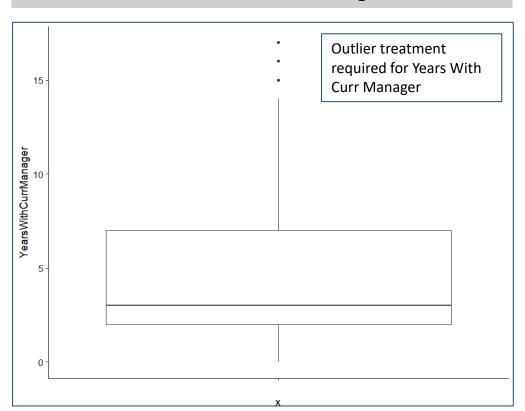
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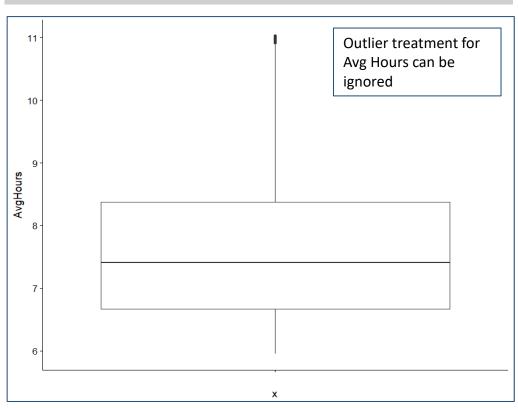
Outlier Analysis



Years with Current Manager



Average Hours



After the outlier treatment we have the final set of data on which we start applying our modelling.





Modeling

Building Model:

- Continuous variables are scaled for better analysis
- Categorical variables are converted to factors and then dummy levels are created for them
- Using the in_time & out_time data source's we are deriving a new metric AverageWorking hours.
- Data is divided into 70-30 ratio where 70% of the data is the training set and 30% of the data is Test set on which we predict our outcome.
- Applying Logistic Regression techniques for the outcome

Evaluating Model:

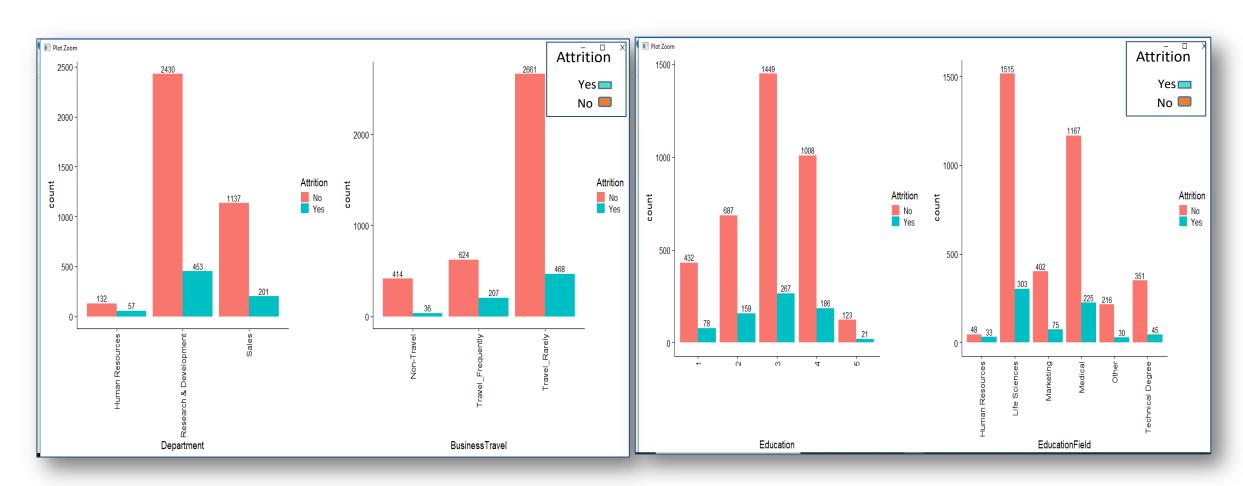
We fetch Accuracy, Sensitivity, Specificity for various cut-off's and predict the built model







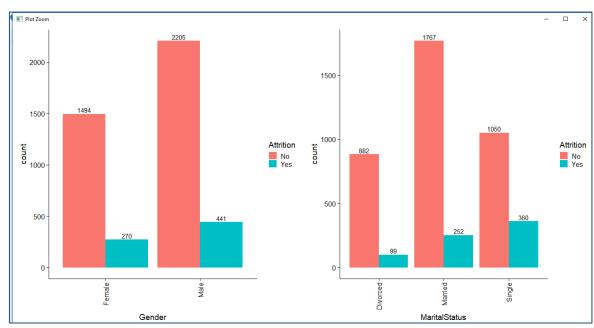
The below plots reveal the attrition rate with respect to various categorical variables

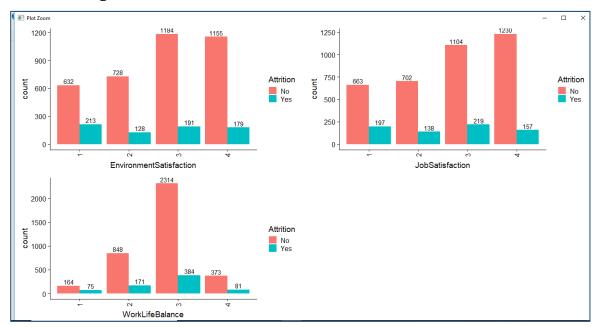


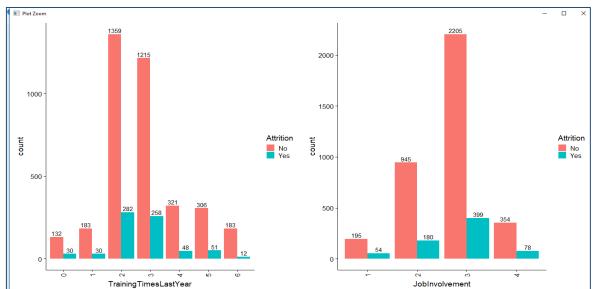


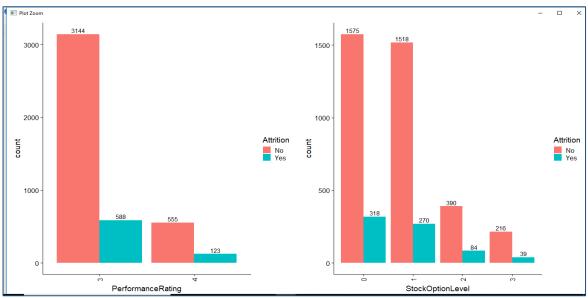
Univariate Analysis









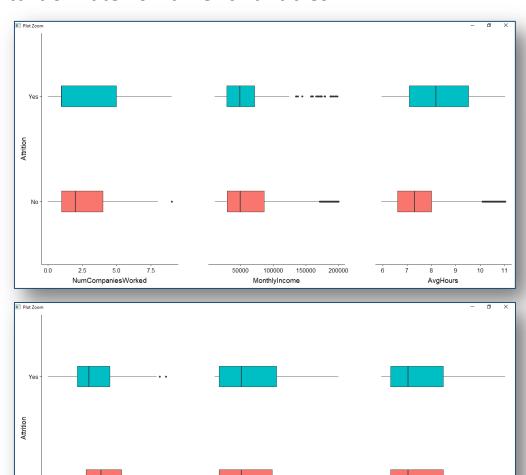




Bivariate Analysis



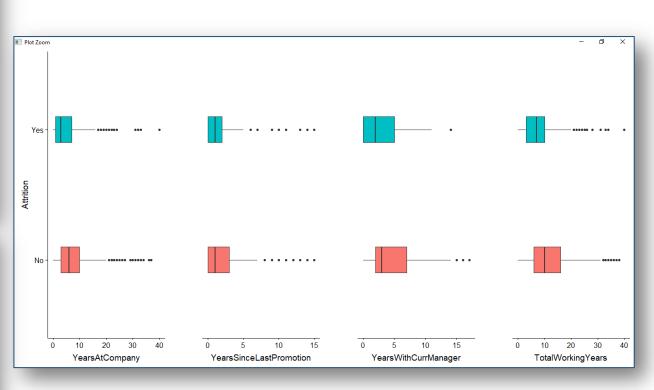
Attrition rate Vs Numeric variables



20

DistanceFromHome

15 20 PercentSalaryHike







Correlation & Significance

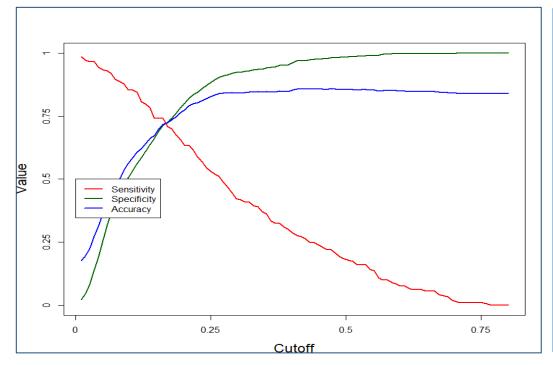
- We apply the glm() function and create accurate models at every step by eliminating variables based on P-value & VIF().
- After 31 models we finally get an accurate model with 9 variables, which are as follows:
 - NumCompaniesWorked
 - TotalWorkingYears
 - AvgHours
 - BusinessTravel_XTravel_Frequently
 - MaritalStatus.xSingle
 - EnvironmentSatisfaction.x2
 - EnvironmentSatisfaction.x3
 - EnvironmentSatisfaction.x4





Accuracy, Specificity & Sensitivity of various cut-off's UpGrad

Cut-off	Accuracy	Sensitivity	Specificity
0.5	0.8556	0.1784	0.98559
0.4	0.8534	0.28638	0.96216
Calculated by R: 0.1616162	0.7150416	0.7230047	0.7135135



Based on the intersection of the 3 components we R calculates the optimal cut-off which keeps the sensitivity and specificity at a controlled level. i.e. 0.1616162





CONCLUSION

- Company has to take care of the 9 factors mentioned in the previous slide to predict if their employee would stay or leave.
- As per the analysis, Performance rating to the 3rd level is the most contributing factor for employees to leave the company .Therefore the performance to every employee at third level must be scrutinized in future.