

# HR Analytics

## CASE STUDY SUBMISSION

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# Abstract

- **XYZ**, employs over 4000 employees
- Their yearly attrition is around 15%, resulting in
  - Delayed Projects
  - Need for more backup resources
  - More number of trainings and onboarding sessions for New employees
- All the above factors is impacting their operational cost and brand value

# Problem solving methodology

## Data Preparation -

- Test for duplicates
- Fix NA values
- Fix outliers
- Create Segmented variables – like age group, overtime etc.

## EDA -

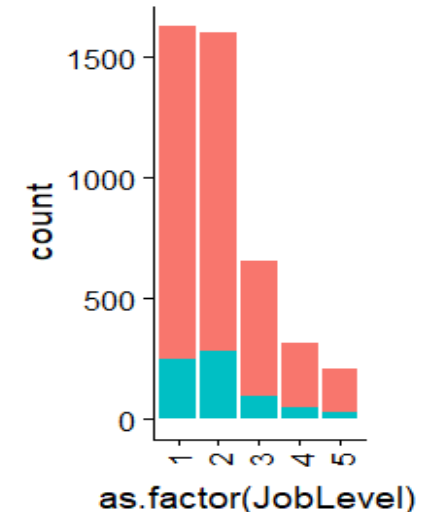
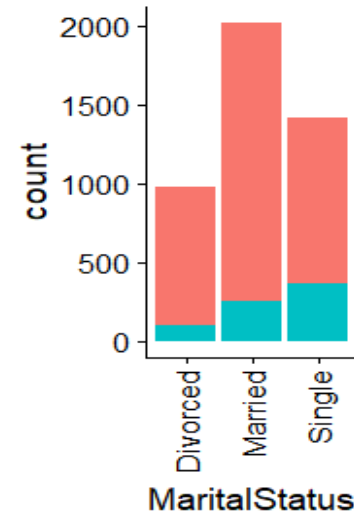
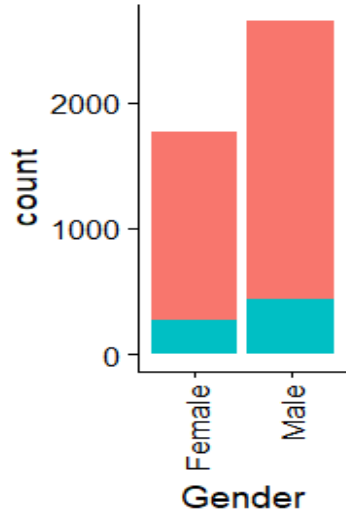
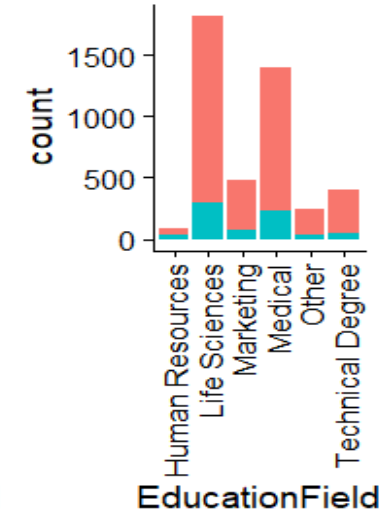
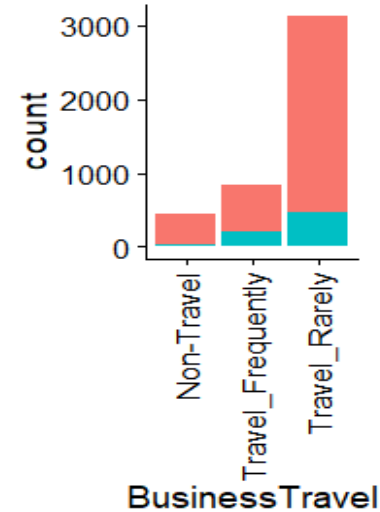
- Perform univariate and multivariate analysis

## Modeling -

- Create train/test data by splitting the data in 70:30 ratio
- Run stepAIC on train data, run iteration to remove columns with High VIF and high p-value
- Take final model and use it to predict attrition in test model
- Evaluate model on sensitivity, specificity and accuracy of model
- Create Gain/Lift charts to validate the model strength
- Find KS-statistics to determine the discriminating power of model

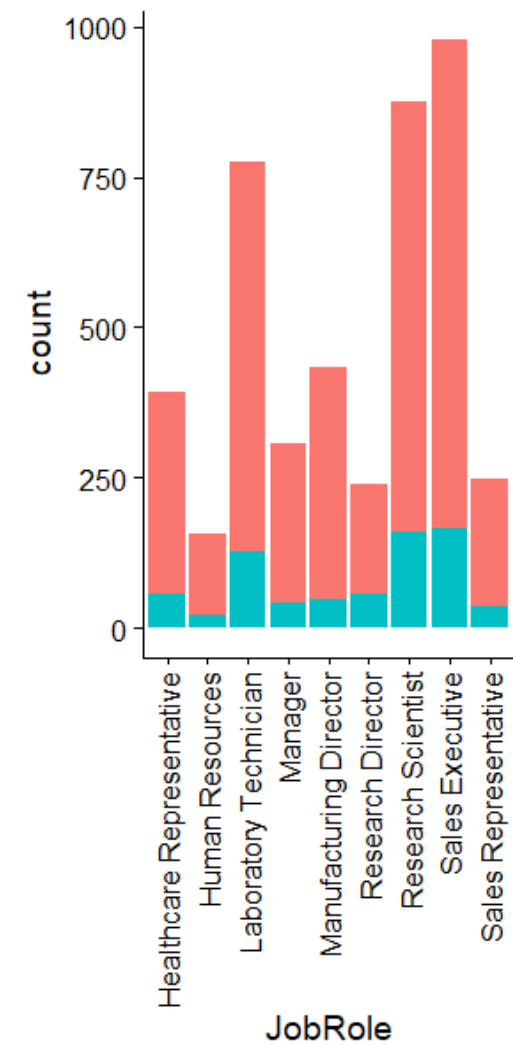
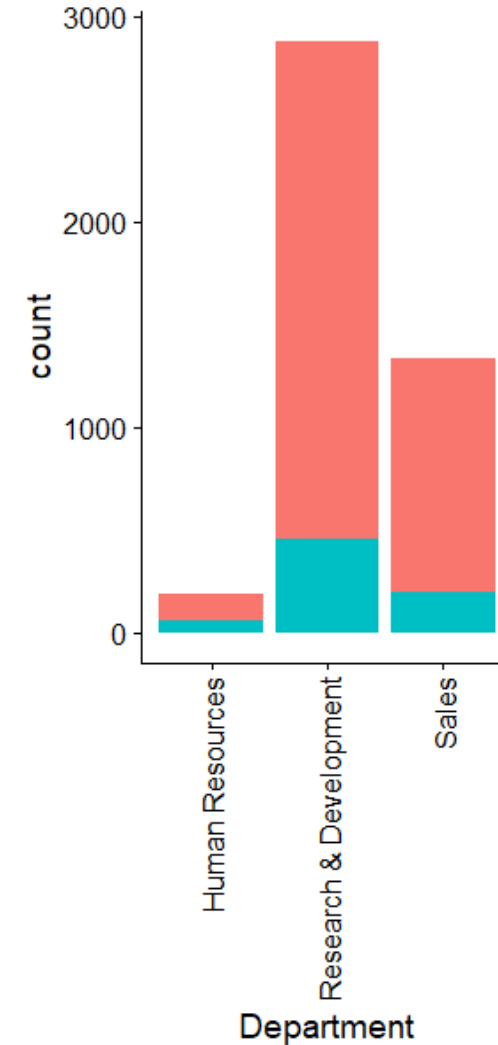
# Variables across attrition

- BusinessTravel has a clear impact on attrition
- Other variables too are showing significant spread for attrition
- Numberwise More Males leave job but percentagewise women leave more
- Singles tend to quit jobs more than married and divorced



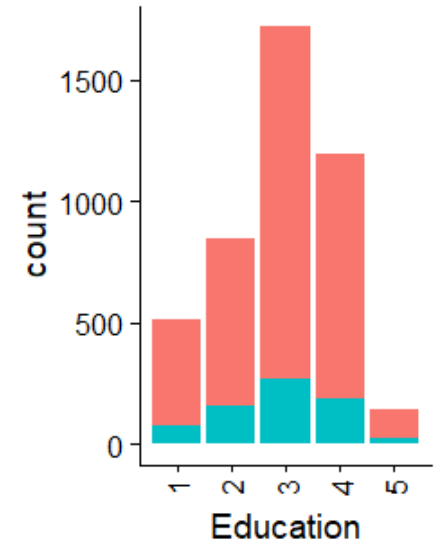
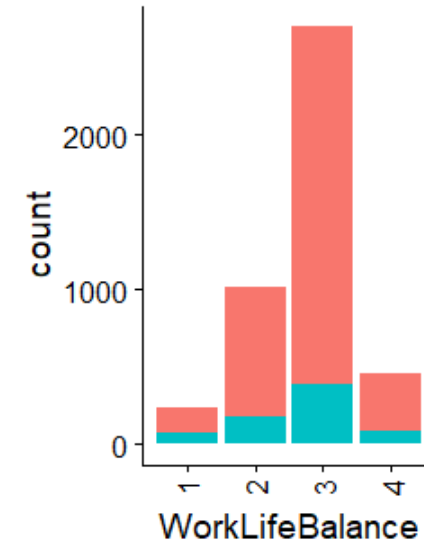
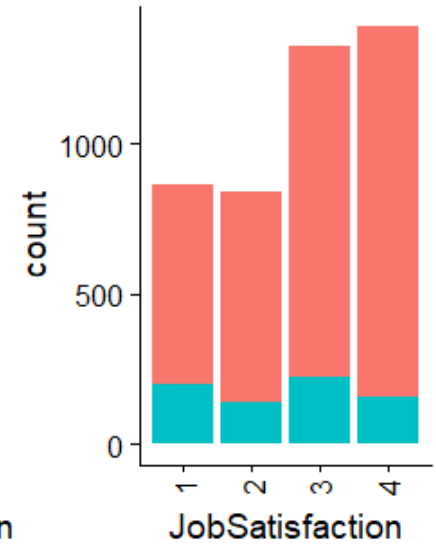
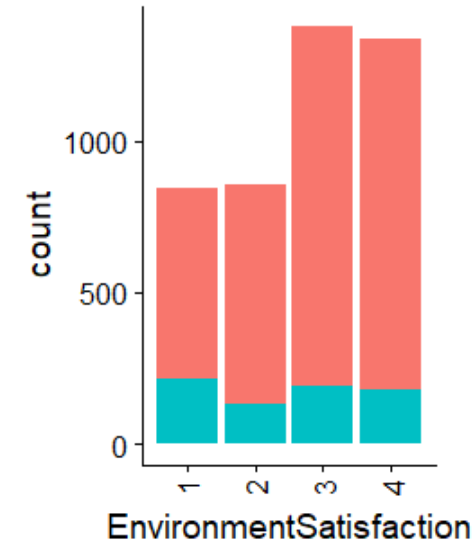
# Variables across attrition

- People working in Research be it Research Scientist or Lab Technician quits job more than others
- Sales executives are also attrition prone more than others



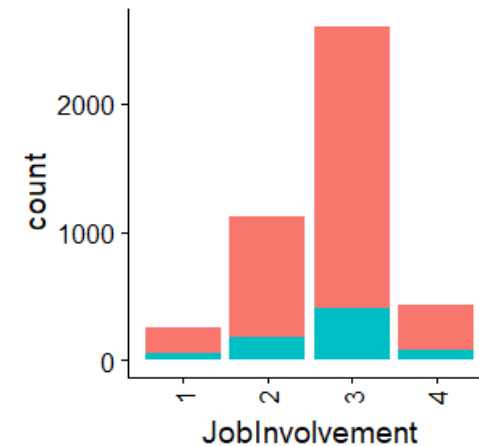
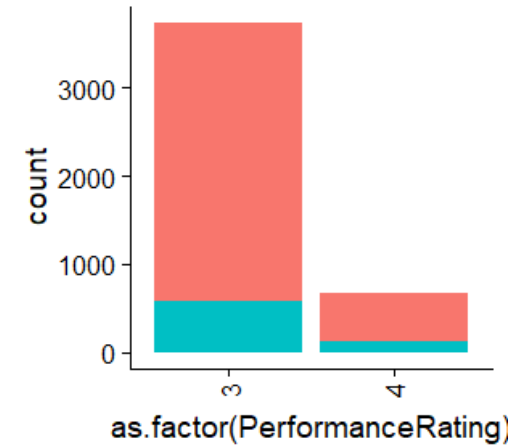
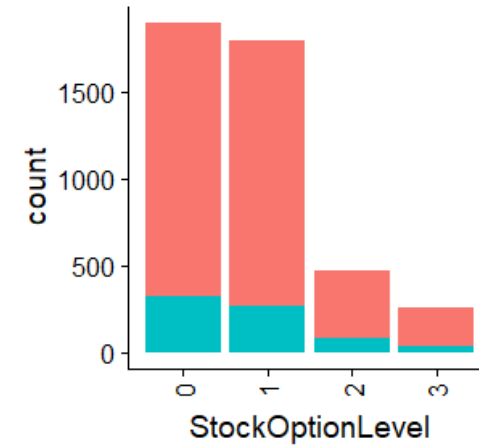
# Factor (ordinal though) but have numerical value

- WorkLifeBalance & Education is an important Factor
- Looks like EnvironmentSatisfaction and JobSatisfaction explains variance in similar fashion



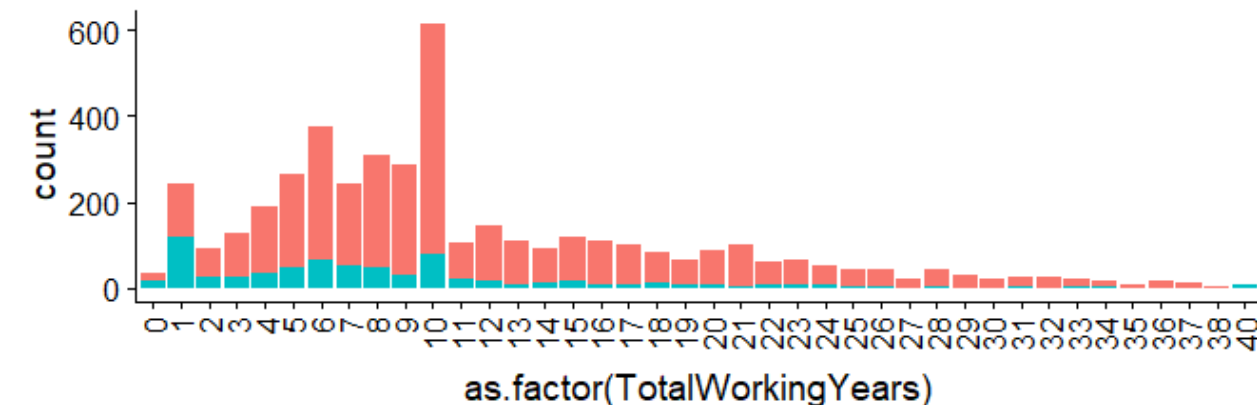
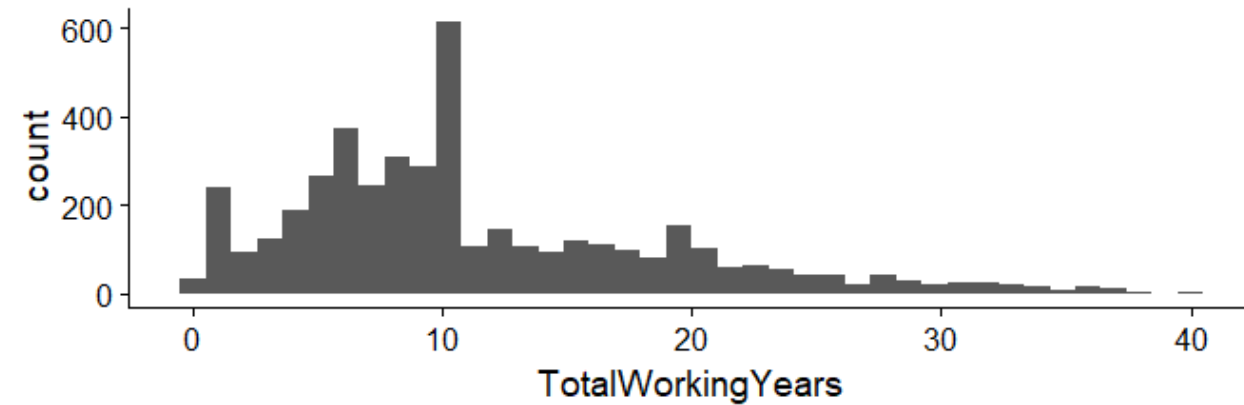
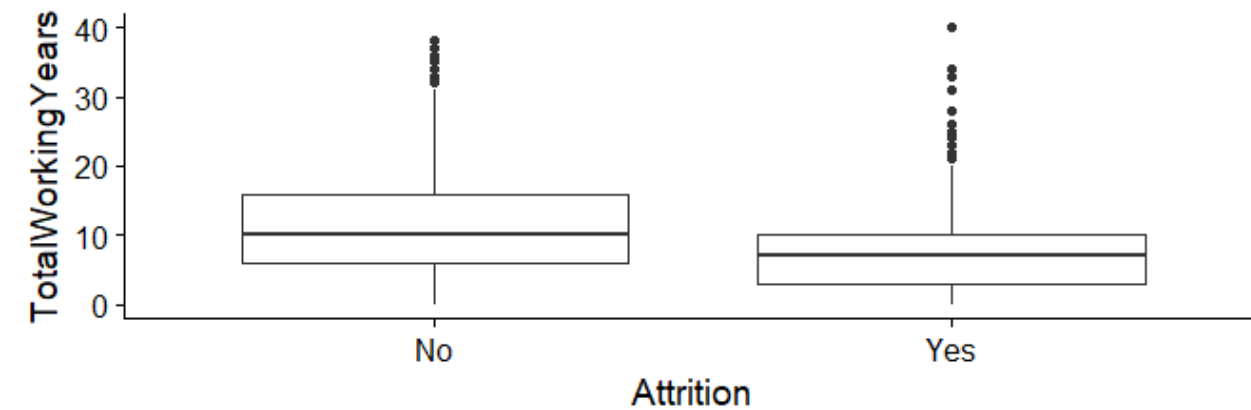
# Variables categorical (ordinal) in nature

- All of these variables looks important as they have significant attrition spread
- StockOptionLevel and JobInvolvement looks like key



# Total Work Experience

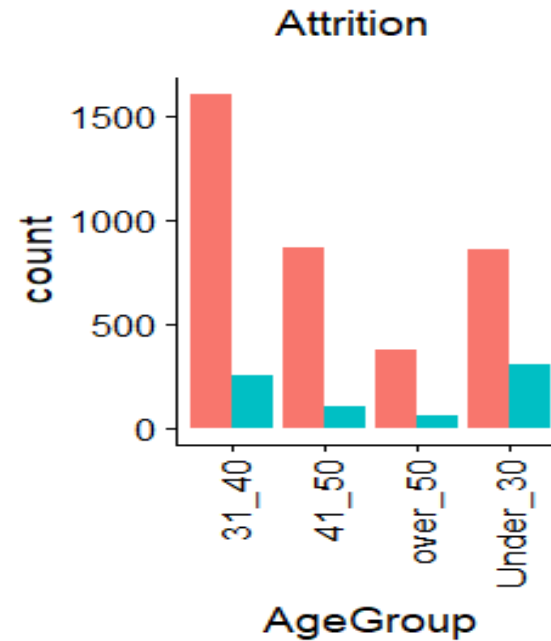
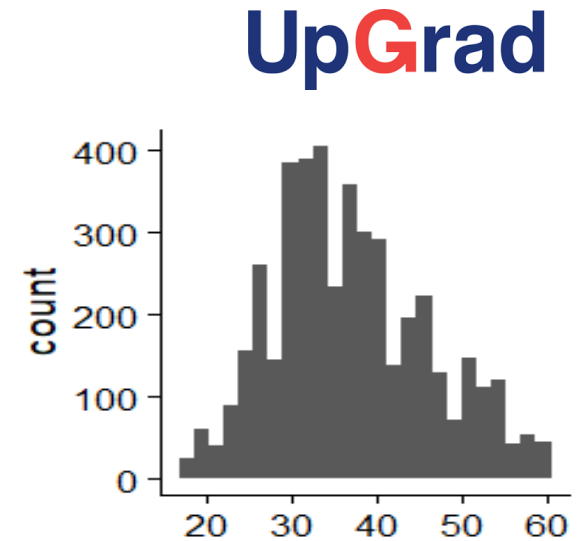
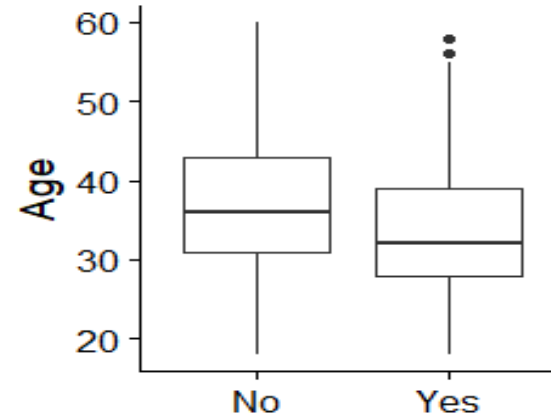
- People leaving company have median 7 years of experience and have relatively lower overall experience
- This might mean as people get more experienced they tend to stay at same company for longer time





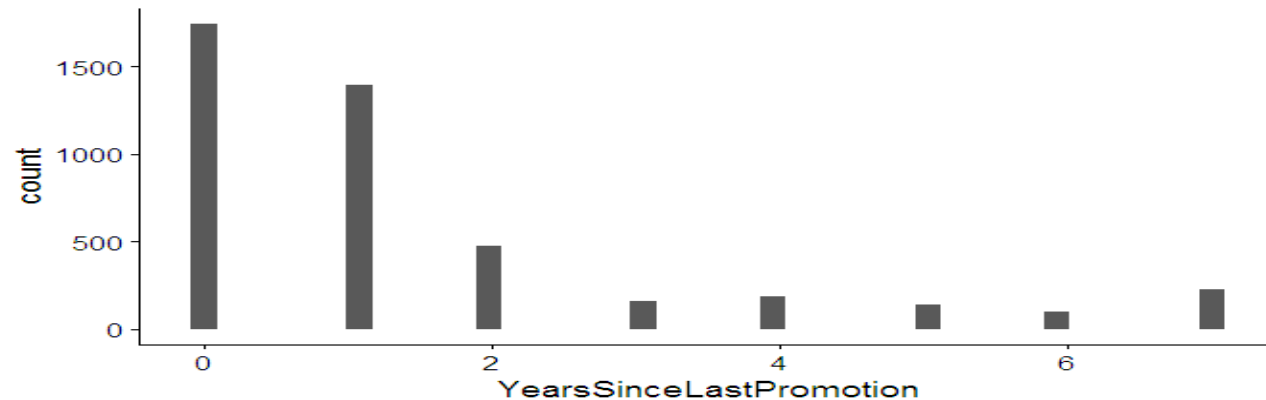
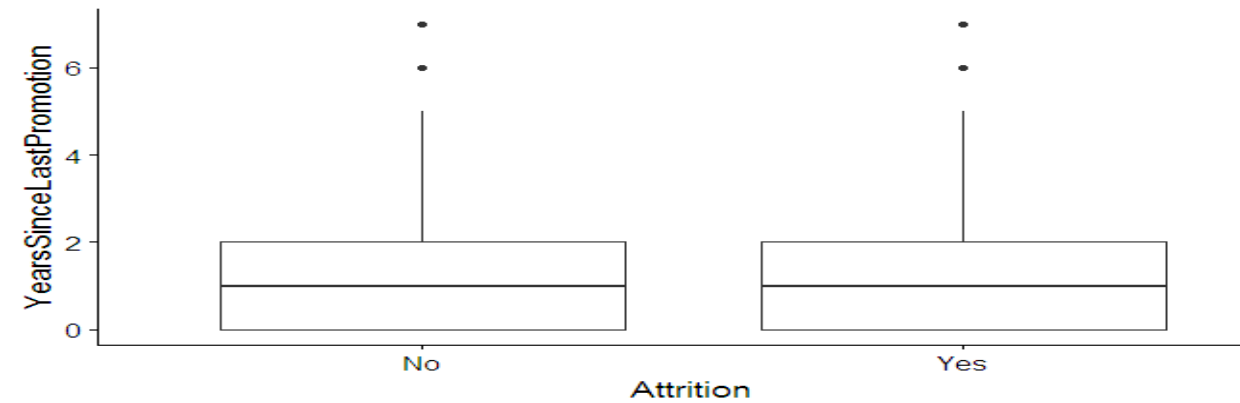
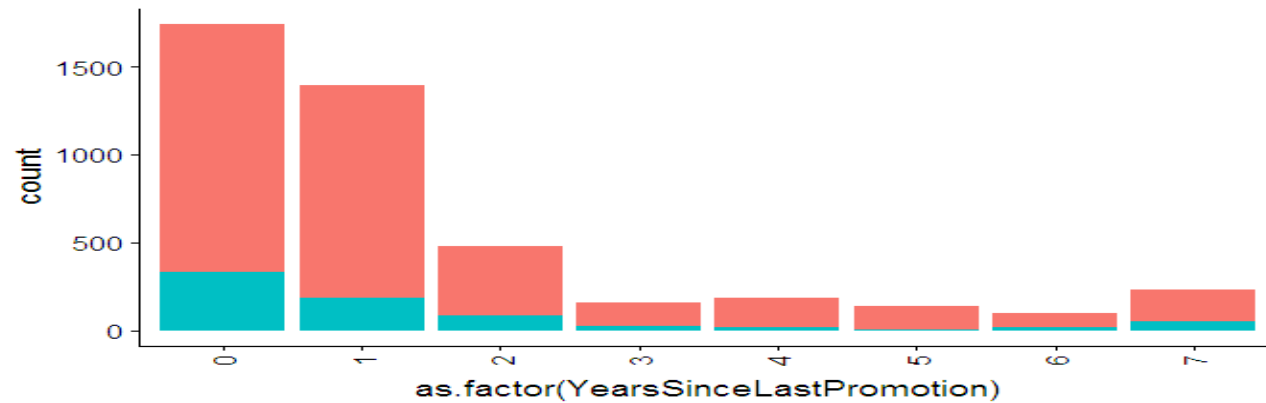
# Age and AgeGroup

- Mostly work force has age between 25 to 50
- People resigning appears to be relatively younger ones
- We will use AgeGroup for modeling as it gives a clearer picture, So dropping Age



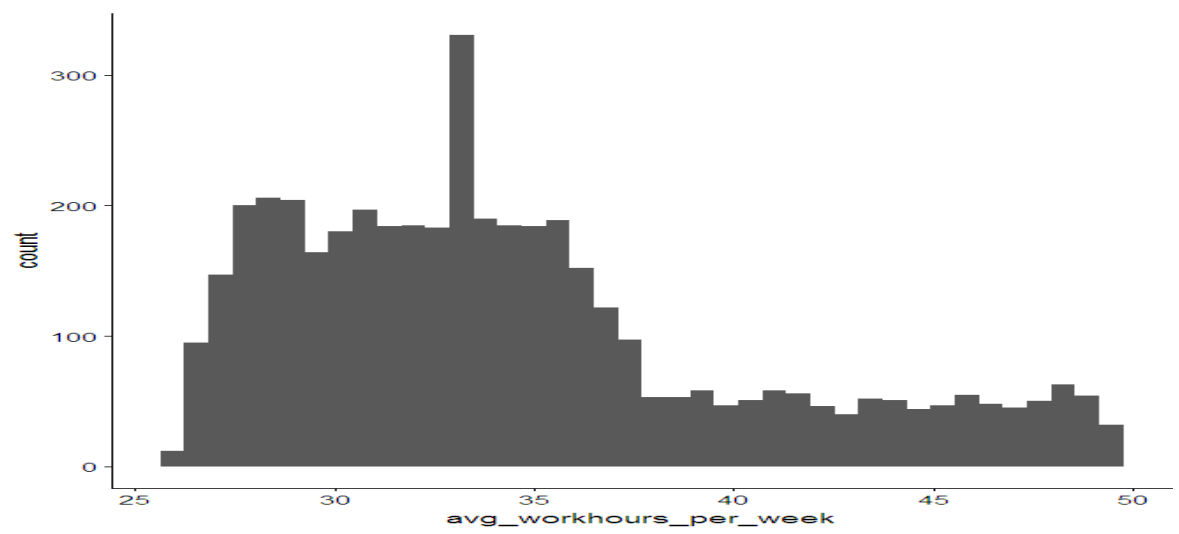
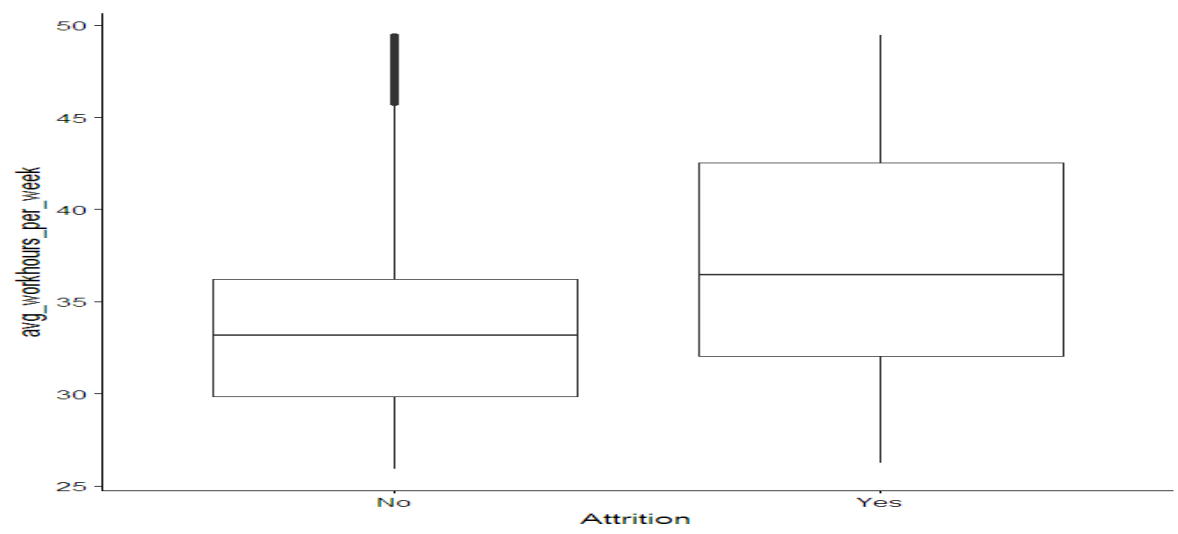
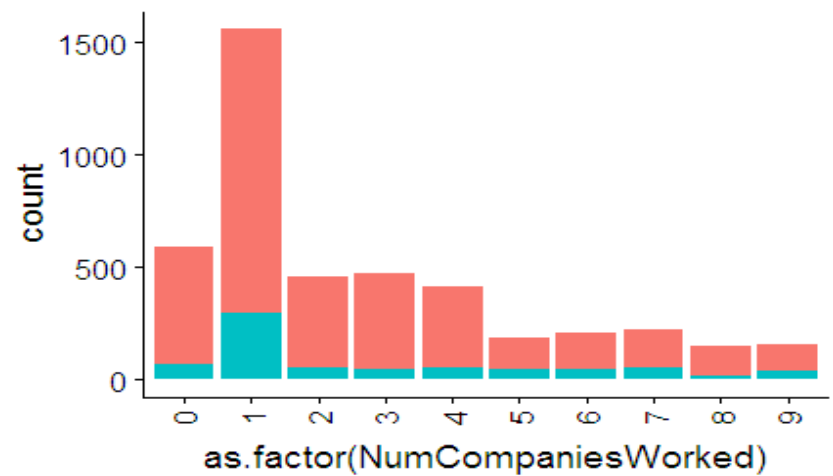
# Years Since Last Promotion

- people are trying to change job soon after getting promotion
- that might make sense as well because that means reaching higher salary/ higher grade jump in relatively lower time span



# Other Variables having impact on Attrition

- Most people leaving have worked in 1 company or if it is their 1st company
- People resigning are those who are generally overworked



# Model Creation

- We did a 70-30 split for Training & Testing respectively
- Post stepAIC and 29 Iterations over Train dataset, we reached a model that takes 13 input variables
- AIC – 2226.1, Null deviance: 2747.7, Residual deviance: 2198.1

## Final Model Equation –

```
model_29 <- glm(formula = Attrition ~ NumCompaniesWorked + TotalWorkingYears +
  YearsAtCompany + YearsSinceLastPromotion +
  avg_workhours_per_week + AgeGroup.xUnder_30 +
  EnvironmentSatisfaction.x2 + EnvironmentSatisfaction.x3 +
  EnvironmentSatisfaction.x4 + JobSatisfaction.x4 +
  BusinessTravel.xTravel_Frequently + JobRole.xManufacturing.Director +
  MaritalStatus.xSingle,
  family = "binomial", data = train)
```

# Model Evaluation

- We predicted the Attrition on test dataset -

```
summary(test_pred)
```

```
      Min.   1st Qu.   Median     Mean   3rd Qu.     Max.
0.002391 0.046552 0.110468 0.171465 0.238789 0.862327
```

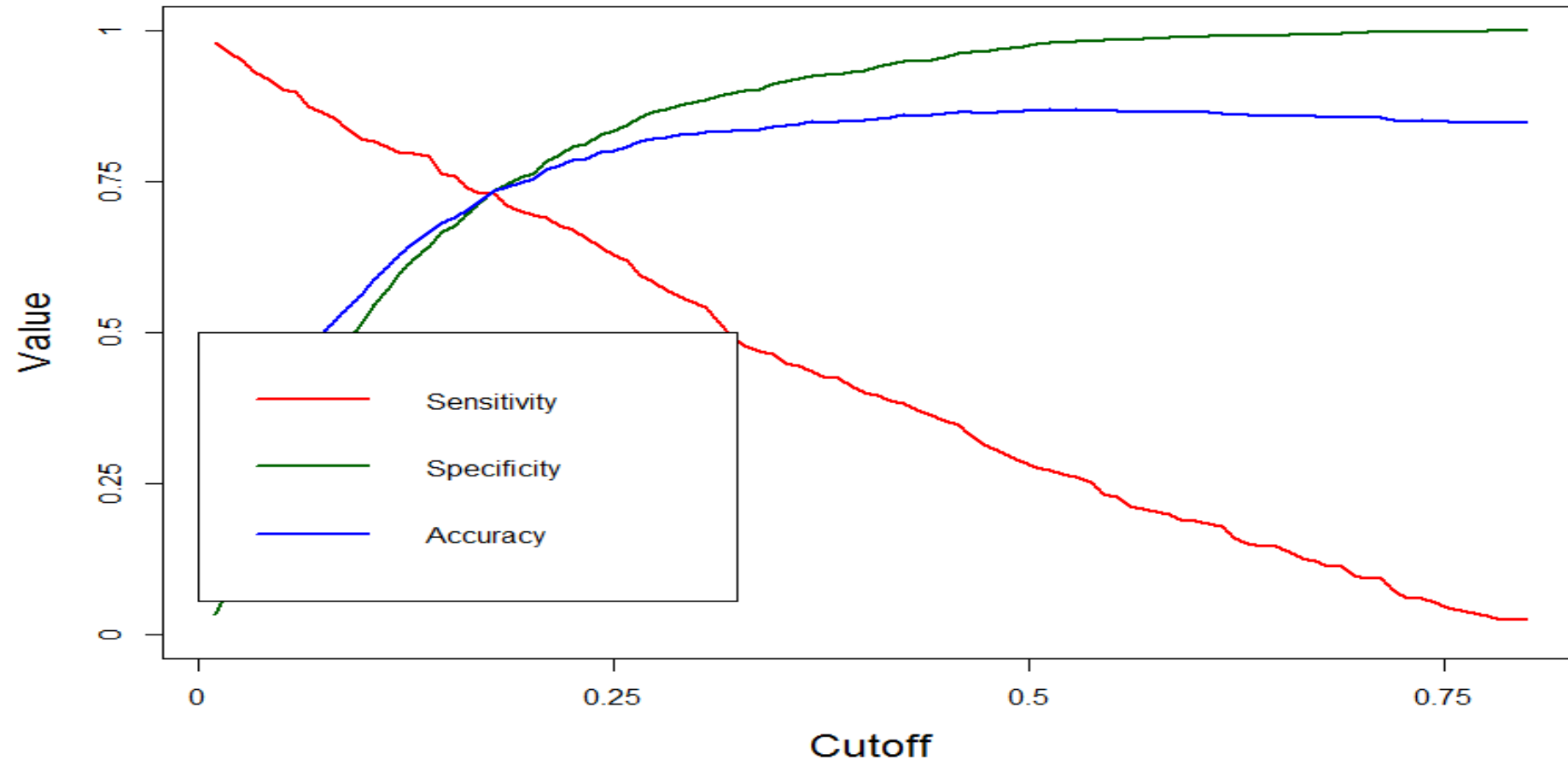
- Then we plotted sensitivity, specificity and accuracy of model, and finalized a cutoff of 0.1775758

```
+-----+
|Accuracy   | Sensitivity | Specificity |
+-----+
|0.733938   | 0.7294686  | 0.734767    |
+-----+
```

- Plot of sensitivity, specificity and accuracy on next page
- KS-Statistics of our model is 0.4642356 – which is a satisfactory discrimination power

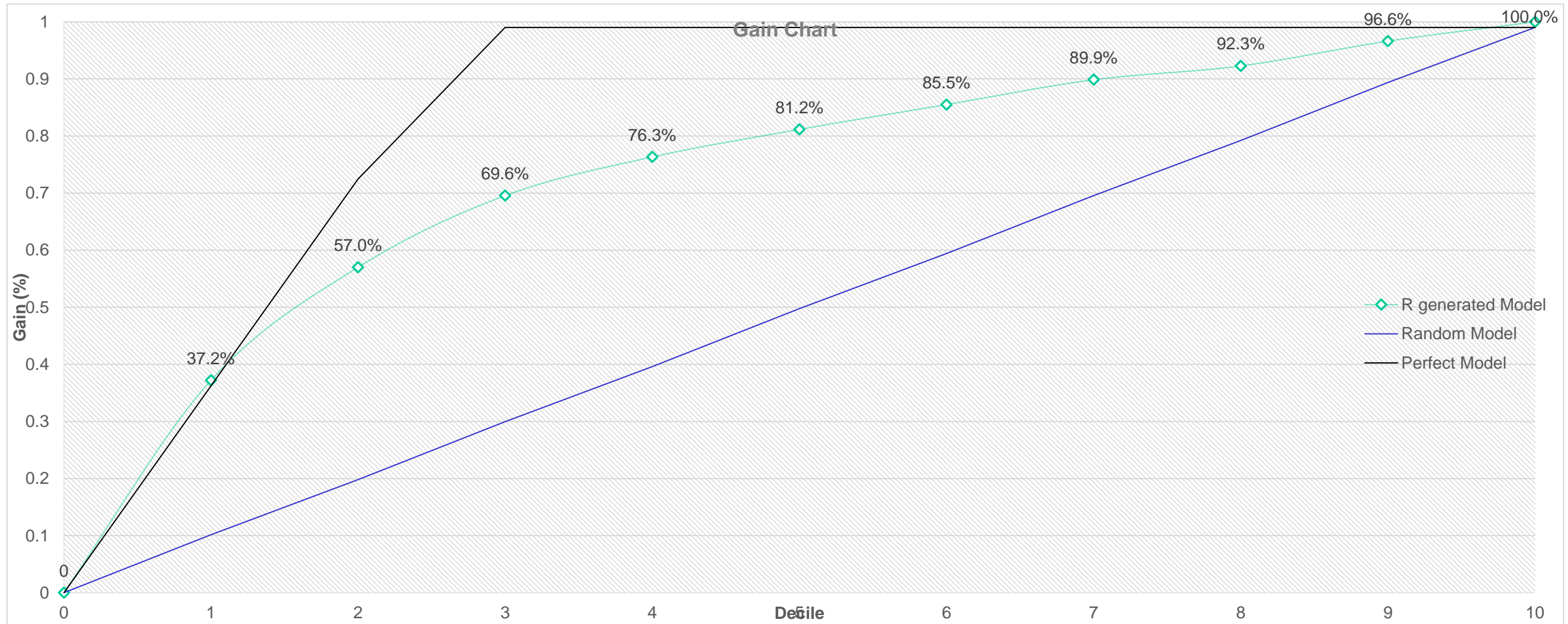
# Model Evaluation – Contd.

Plot of sensitivity, specificity and accuracy



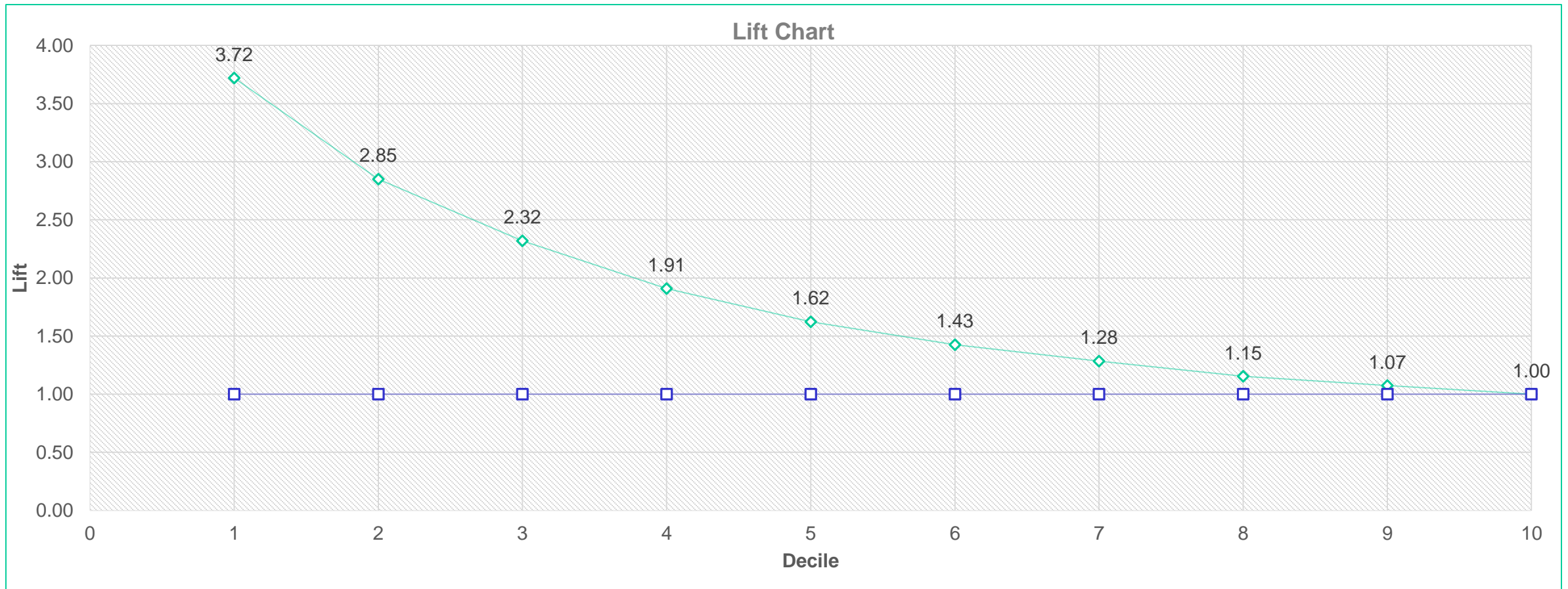
# Model Evaluation – Gain Chart

From Gain Chart it is clearly visible that in 4<sup>th</sup> decile using our model we identify 76.3% of attrition.



# Model Evaluation – Lift Chart

From Lift Chart it is clearly visible that in 3<sup>rd</sup> decile our model is outperforming a random model by 2.32 and in 4<sup>th</sup> by 1.91





# Conclusion

**Factors that affect adversely and attribute to increase in attrition are:**

- **NumCompaniesWorked** – People who have worked in many companies before have higher chances of leaving.
- **YearsSinceLastPromotion** – People who have not been promoted for a long time have higher chances of leaving.
- **avg\_workhours\_per\_week** – Has strong impact on attrition, people who are working more hours, i.e. overworked employee, are more likely to leave company.
- **AgeGroup.xUnder\_30** – It tells that people with age under 30 are more likely to change job than others.
- **BusinessTravel.xTravel\_Frequently** – People traveling frequently is coming out a factor for attrition, as it may be affecting their work life balance.
- **MaritalStatus.xSingle** – the employees who are single have higher chances of leaving than the ones married or divorced

# Conclusion – contd.

## Factors that keeps attrition in control:

- **TotalWorkingYears** – This variable has a very low p-value and has stronger impact on attrition. The employees with higher work experience tend to stay longer at company.
- **YearsAtCompany** – People who have spent more time at company tends to stay with the company.
- **EnvironmentSatisfaction.x2/ x3 / x4**- People who rated work environment as Medium/ High/ Very High are happy with the company environment and do not leave.
- **JobSatisfaction.x4** –People only who are very highly satisfied with their job are going to stay with the organization.
- **JobRole.xManufacturing.Director** – Employees who are Director in manufacturing are satisfied with their role and tend to stay with the organization