

## Employee Churn Analytics

**PROBLEM STATEMENT:** Companies face the problem that their human resources on whom the company have invested time and money to train them, leave the company voluntarily. It is important for the management and stakeholders to know the variables responsible for employees quitting jobs and also have a prediction that which employees will be quitting their jobs in future.

**Goal:** To predict whether an employee will stay or leave the company within the next year.

```
# Load packages

library('ggplot2') # visualization
## Warning: package 'ggplot2' was built under R version 3.4.4
library('ggthemes') # visualization
## Warning: package 'ggthemes' was built under R version 3.4.4
library('scales') # visualization
## Warning: package 'scales' was built under R version 3.4.4
library('dplyr') # data manipulation
## Warning: package 'dplyr' was built under R version 3.4.3
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
library('mice') # imputation
## Warning: package 'mice' was built under R version 3.4.4
## Loading required package: lattice
##
## Attaching package: 'mice'
```

```

## The following objects are masked from 'package:base':
##
##      cbind, rbind

library('randomForest') # classification algorithm

## Warning: package 'randomForest' was built under R version 3.4.4

## randomForest 4.6-14

## Type rfNews() to see new features/changes/bug fixes.

##
## Attaching package: 'randomForest'

## The following object is masked from 'package:dplyr':
##
##      combine

## The following object is masked from 'package:ggplot2':
##
##      margin

dataNW<-read.csv("WA_Fn-UseC_-HR-Employee-Attrition.csv")
head(dataNW)

##      i..Age Attrition      BusinessTravel DailyRate      Department
## 1      41      Yes      Travel_Rarely      1102      Sales
## 2      49      No Travel_Frequently      279 Research & Development
## 3      37      Yes      Travel_Rarely      1373 Research & Development
## 4      33      No Travel_Frequently      1392 Research & Development
## 5      27      No      Travel_Rarely      591 Research & Development
## 6      32      No Travel_Frequently      1005 Research & Development
##      DistanceFromHome Education EducationField EmployeeCount EmployeeNumber
## 1              1          2 Life Sciences              1              1
## 2              8          1 Life Sciences              1              2
## 3              2          2      Other              1              4
## 4              3          4 Life Sciences              1              5
## 5              2          1      Medical              1              7
## 6              2          2 Life Sciences              1              8
##      EnvironmentSatisfaction Gender HourlyRate JobInvolvement JobLevel
## 1              2 Female          94              3              2
## 2              3 Male           61              2              2
## 3              4 Male           92              2              1
## 4              4 Female          56              3              1
## 5              1 Male           40              3              1
## 6              4 Male           79              3              1
##      JobRole JobSatisfaction MaritalStatus MonthlyIncome
## 1      Sales Executive          4      Single          5993
## 2      Research Scientist          2      Married          5130
## 3 Laboratory Technician          3      Single          2090
## 4      Research Scientist          3      Married          2909

```

```

## 5 Laboratory Technician      2      Married      3468
## 6 Laboratory Technician      4      Single      3068
##      MonthlyRate NumCompaniesWorked Over18 OverTime PercentSalaryHike
## 1      19479      8      Y      Yes      11
## 2      24907      1      Y      No      23
## 3      2396      6      Y      Yes      15
## 4      23159      1      Y      Yes      11
## 5      16632      9      Y      No      12
## 6      11864      0      Y      No      13
##      PerformanceRating RelationshipSatisfaction StandardHours
## 1      3      1      80
## 2      4      4      80
## 3      3      2      80
## 4      3      3      80
## 5      3      4      80
## 6      3      3      80
##      StockOptionLevel TotalWorkingYears TrainingTimesLastYear WorkLifeBalance
## 1      0      8      0      1
## 2      1      10      3      3
## 3      0      7      3      3
## 4      0      8      3      3
## 5      1      6      3      3
## 6      0      8      2      2
##      YearsAtCompany YearsInCurrentRole YearsSinceLastPromotion
## 1      6      4      0
## 2      10      7      1
## 3      0      0      0
## 4      8      7      3
## 5      2      2      2
## 6      7      7      3
##      YearsWithCurrManager
## 1      5
## 2      7
## 3      0
## 4      0
## 5      2
## 6      6

```

```
names(dataNW)[names(dataNW) == 'i..Age'] <- 'Age'
```

```
dim(dataNW)
```

```
## [1] 1470 35
```

```
names(dataNW)
```

```

## [1] "Age"      "Attrition"
## [3] "BusinessTravel" "DailyRate"
## [5] "Department" "DistanceFromHome"
## [7] "Education" "EducationField"
## [9] "EmployeeCount" "EmployeeNumber"

```

```
## [11] "EnvironmentSatisfaction" "Gender"
## [13] "HourlyRate" "JobInvolvement"
## [15] "JobLevel" "JobRole"
## [17] "JobSatisfaction" "MaritalStatus"
## [19] "MonthlyIncome" "MonthlyRate"
## [21] "NumCompaniesWorked" "Over18"
## [23] "OverTime" "PercentSalaryHike"
## [25] "PerformanceRating" "RelationshipSatisfaction"
## [27] "StandardHours" "StockOptionLevel"
## [29] "TotalWorkingYears" "TrainingTimesLastYear"
## [31] "WorkLifeBalance" "YearsAtCompany"
## [33] "YearsInCurrentRole" "YearsSinceLastPromotion"
## [35] "YearsWithCurrManager"
```

```
str(data)
```

```
## function (..., list = character(), package = NULL, lib.loc = NULL,
##     verbose = getOption("verbose"), envir = .GlobalEnv)
```

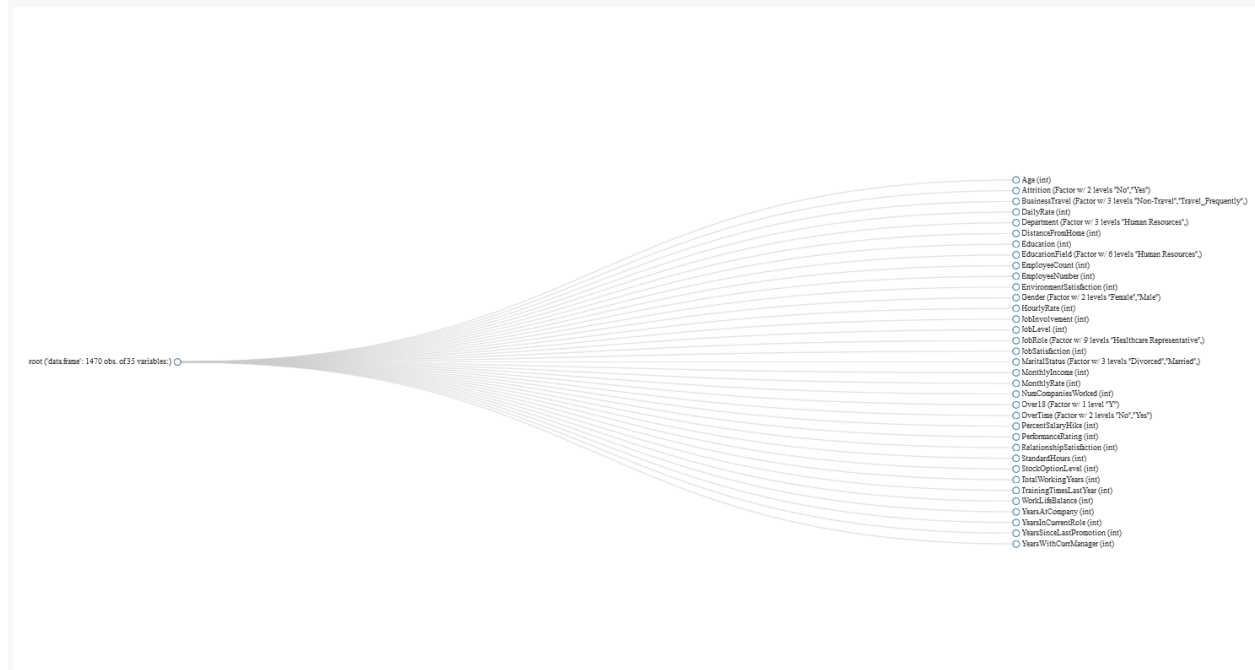
Checking the Missing values:

```
sapply(dataNW, function(x) sum(is.na(x)))
```

```
##           Age           Attrition           BusinessTravel
##           0             0             0
##      DailyRate      Department      DistanceFromHome
##           0             0             0
##      Education      EducationField      EmployeeCount
##           0             0             0
##      EmployeeNumber  EnvironmentSatisfaction      Gender
##           0             0             0
##      HourlyRate      JobInvolvement      JobLevel
##           0             0             0
##      JobRole      JobSatisfaction      MaritalStatus
##           0             0             0
##      MonthlyIncome      MonthlyRate      NumCompaniesWorked
##           0             0             0
##      Over18           OverTime      PercentSalaryHike
##           0             0             0
##      PerformanceRating  RelationshipSatisfaction      StandardHours
##           0             0             0
##      StockOptionLevel      TotalWorkingYears      TrainingTimesLastYear
##           0             0             0
##      WorkLifeBalance      YearsAtCompany      YearsInCurrentRole
##           0             0             0
##      YearsSinceLastPromotion      YearsWithCurrManager
##           0             0
```

Data Exploration

```
library(DataExplorer)
plot_str(dataNW)
```



```
Terminated<-as.factor(dataNW$Attrition)
summary(Terminated)
```

```
##      No      Yes
## 1233    237
```

```
perc_attrition_rate<-sum(dataNW$Attrition)/length(dataNW$Attrition))*100
```

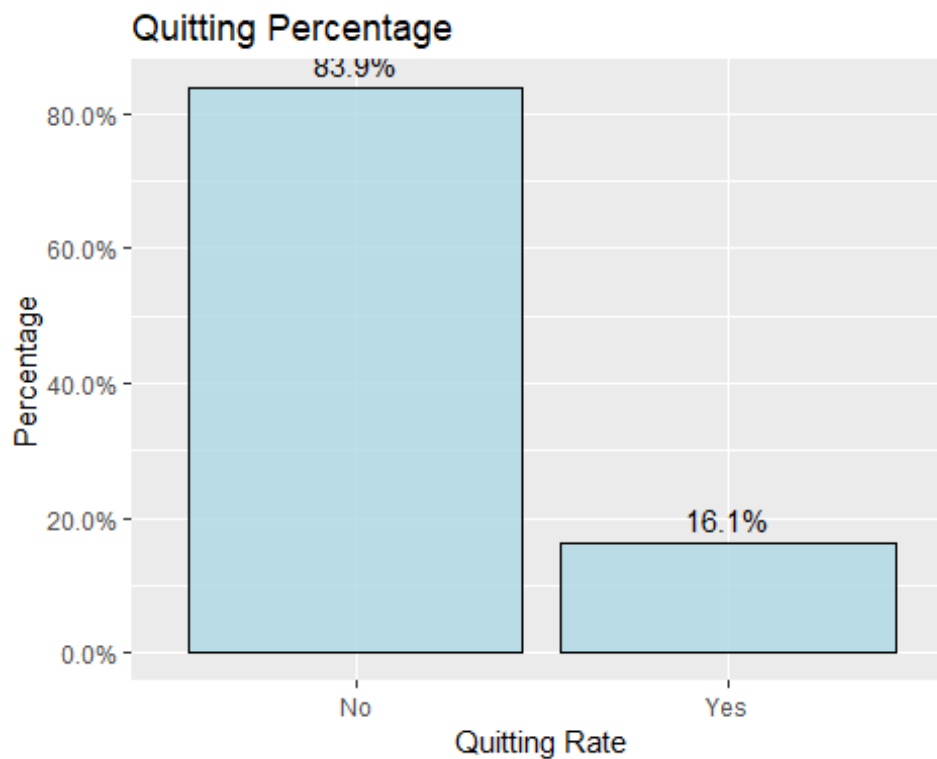
```
## Warning in Ops.factor(dataNW$Attrition, length(dataNW$Attrition)): '/' not
## meaningful for factors
```

```
prop.table(table(dataNW$Attrition))
```

```
##
##           No           Yes
## 0.8387755 0.1612245
```

```
Terminated<- ggplot(dataNW, aes(x=Attrition)) +
  geom_bar(aes(y=(..count..)/sum(..count..)), alpha=0.8, fill="lightblue",
color = "black") +
  scale_y_continuous(labels = scales::percent) +
```

```
geom_text(aes(label = scales::percent(..count../sum(..count..)),
              y= (..count../sum(..count..) ), stat= "count", vjust = -0.5)
+
  ylab("Percentage") + xlab("Quitting Rate")+ ggtitle("Quitting Percentage")
Terminated
```



The Data is

Unbalanced, where the minority class is only 16.1%

```
library(dplyr)
library(tidyr)

## Warning: package 'tidyr' was built under R version 3.4.3
##
## Attaching package: 'tidyr'
##
## The following object is masked from 'package:mice':
##
##   complete

#Transforming Termination Column to Factor with True and False values
dataNW$Attrition<-factor(dataNW$Attrition,labels=c('False',"True"))
SHdf<-dataNW %>% group_by(EducationField,Attrition) %>%
  summarise(count=n())

#replacing NA value with 0
SHdf[is.na(SHdf)]<-0
```

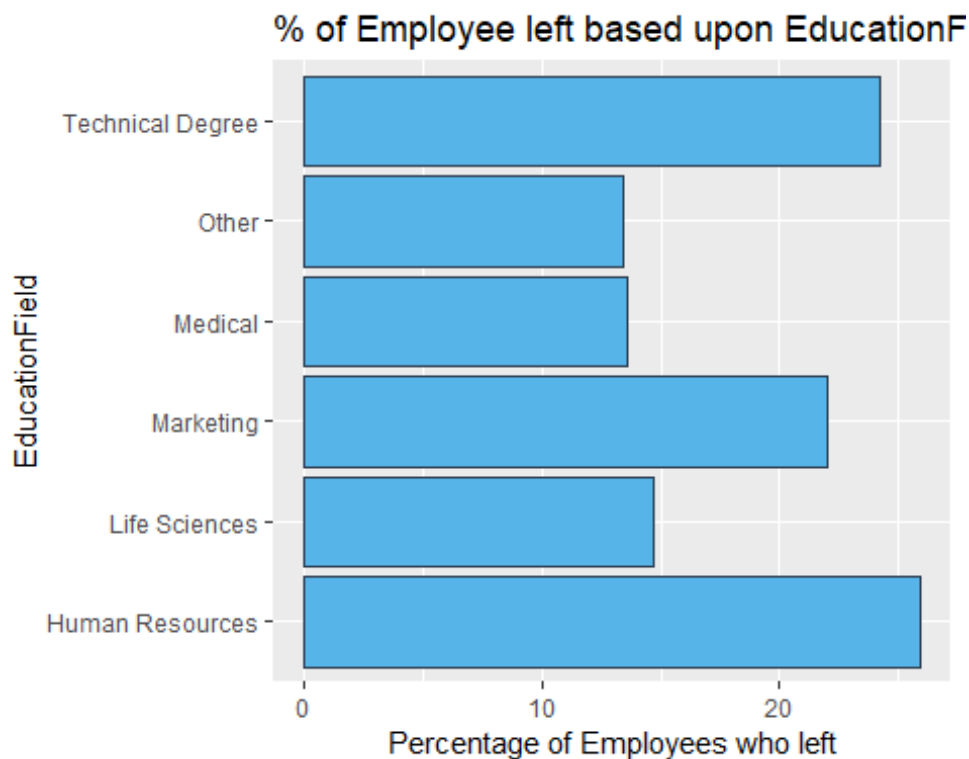
```
#making a data frame of Departments and the count of workers who Left or not
SHdf<-spread(SHdf,Attrition,count)
```

```
SHdf<-transform(SHdf,Perleft=(True/(True+False))*100 ,
PerWork=(False/(True+False))*100)
SHdf
```

```
##      EducationField False True  Perleft  PerWork
## 1   Human Resources    20    7  25.92593  74.07407
## 2    Life Sciences   517   89  14.68647  85.31353
## 3      Marketing    124   35  22.01258  77.98742
## 4        Medical   401   63  13.57759  86.42241
## 5          Other     71   11  13.41463  86.58537
## 6 Technical Degree   100   32  24.24242  75.75758
```

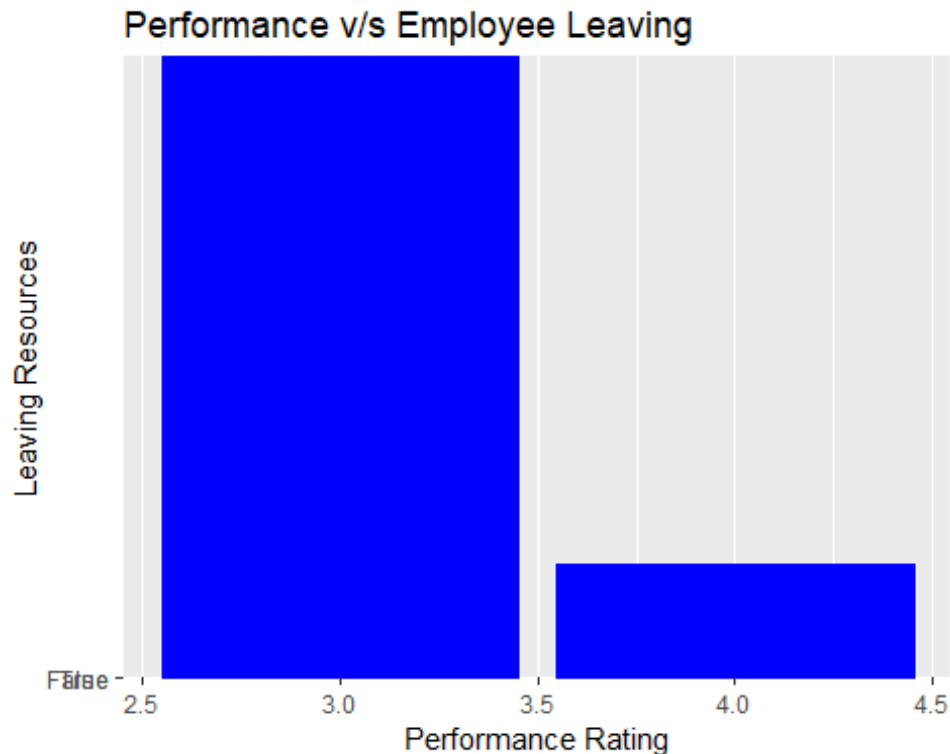
Percentage of employee who left and Employee who are working based upon Source.Of.Hire

```
#Plot of Department vs Percentage of Employees who Left
ggplot(aes(x=EducationField, y = Perleft),data = SHdf) +
  geom_col(fill='#56B4E9',color='#2f3f52') +
  coord_flip()+
  xlab("EducationField") +
  ylab("Percentage of Employees who left") +
  labs(title="% of Employee left based upon EducationField")
```



Employee with Human Resource degree are leaving more.

```
ggplot(dataNW, aes(x = PerformanceRating, y = Attrition)) + geom_bar(stat =
"identity", fill = 'blue', colour = 'blue') + ggtitle("Performance v/s
Employee Leaving") + labs(y = "Leaving Resources", x =
"Performance Rating")
```



RESIGNATION PER DEPARTMENT:

```
dataNW$Attrition <- as.factor(dataNW$Attrition)

dataNW %>%
  select(Department, Attrition) %>%
  group_by(Department, Attrition) %>%
  summarise(count=n()) %>%
  mutate(dep_pct = count/sum(count)) %>%
  ggplot(aes(x=Department, y=dep_pct, fill = Attrition)) +
  geom_bar(stat="identity", alpha = 0.7) +
  geom_text(aes(label = paste0(round(dep_pct*100,0), "%"),
    y=dep_pct+0.02)) +
  scale_fill_brewer(palette="Paired")+
  ylab("Percentage of Employees") + xlab("Department") +
  ggtitle("Resignation per
Department")+theme(axis.text.x=element_text(angle=45,hjust=1))

## Warning: package 'bindrcpp' was built under R version 3.4.3
```

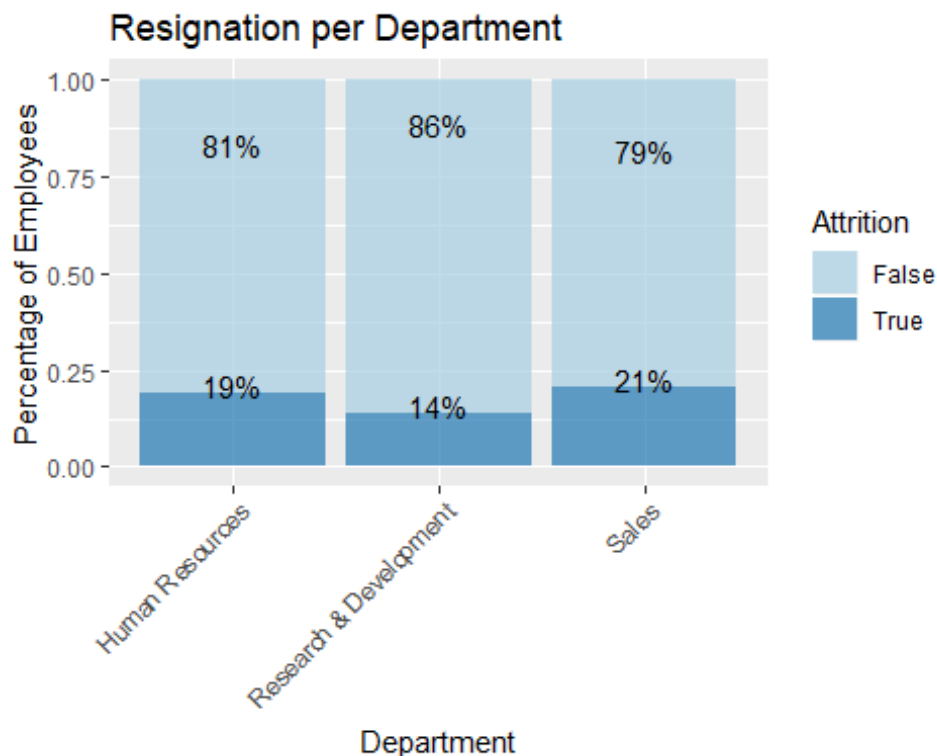


```
## Warning: `as_dictionary()` is soft-deprecated as of rlang 0.3.0.
## Please use `as_data_pronoun()` instead
## This warning is displayed once per session.

## Warning: `new_overscope()` is soft-deprecated as of rlang 0.2.0.
## Please use `new_data_mask()` instead
## This warning is displayed once per session.

## Warning: The `parent` argument of `new_data_mask()` is deprecated.
## The parent of the data mask is determined from either:
##
## * The `env` argument of `eval_tidy()`
## * Quosure environments when applicable
## This warning is displayed once per session.

## Warning: `overscope_clean()` is soft-deprecated as of rlang 0.2.0.
## This warning is displayed once per session.
```

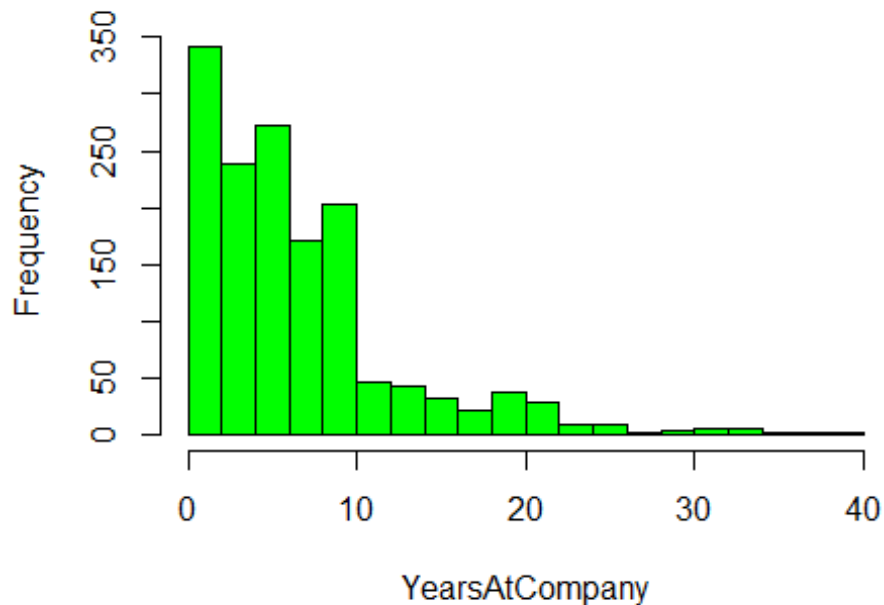


SALES DEPARTMENT ARE LEAVING MORE

CHECKING THE EMPLOYEE TENURE IN THE COMPANY:

```
hist(dataNw$YearsAtCompany, breaks = 15, col = 'green', main = "Analysis of
Years At Company Variable", xlab = "YearsAtCompany")
```

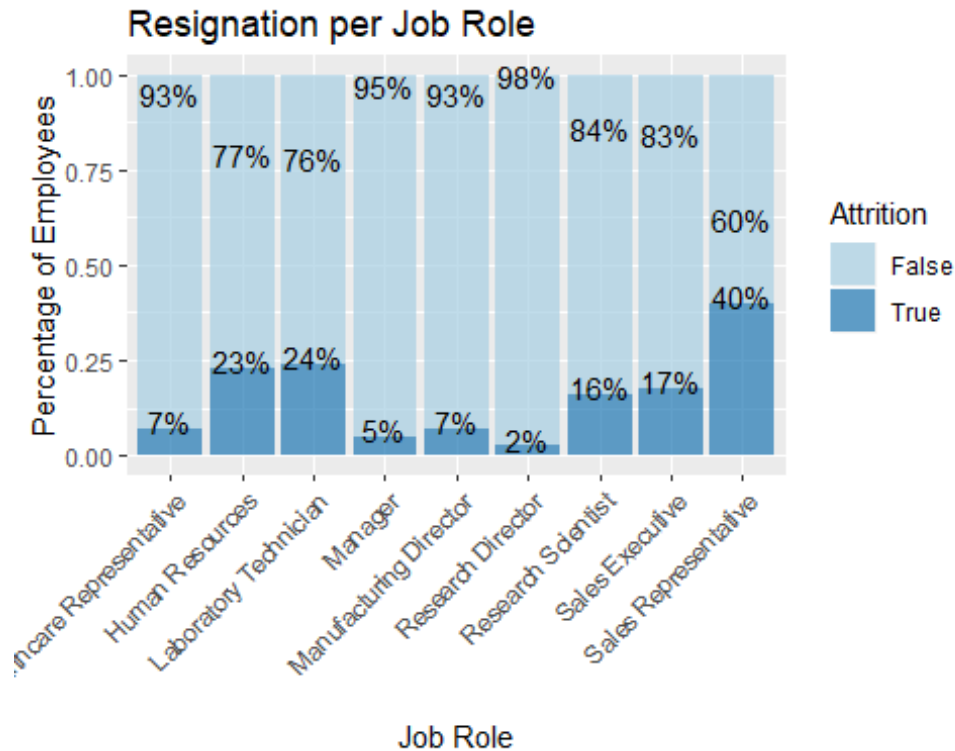
## Analysis of Years At Company Variable



## RESIGNATION BASED UPON JOB ROLE

```
dataNW$Attrition <- as.factor(dataNW$Attrition)

dataNW %>%
  select(JobRole,Attrition) %>%
  group_by(JobRole, Attrition) %>%
  summarise(count=n()) %>%
  mutate(dep_pct = count/sum(count)) %>%
  ggplot(aes(x=JobRole, y=dep_pct, fill = Attrition)) +
  geom_bar(stat="identity", alpha = 0.7) +
  geom_text(aes(label = paste0(round(dep_pct*100,0),"%"),
    y=dep_pct+0.02)) +
  scale_fill_brewer(palette="Paired")+
  ylab("Percentage of Employees") + xlab("Job Role") +
  ggtitle("Resignation per Job
Role")+theme(axis.text.x=element_text(angle=45,hjust=1))
```



SALES REPRESENTATIVE TEND TO LEAVE MORE.

```
str(dataNW)
```

```
## 'data.frame':  1470 obs. of  35 variables:
## $ Age                : int  41 49 37 33 27 32 59 30 38 36 ...
## $ Attrition          : Factor w/ 2 levels "False","True": 2 1 2 1 1
1 1 1 1 1 ...
## $ BusinessTravel     : Factor w/ 3 levels "Non-
Travel","Travel_Frequently",...: 3 2 3 2 3 2 3 3 2 3 ...
## $ DailyRate          : int  1102 279 1373 1392 591 1005 1324 1358
216 1299 ...
## $ Department        : Factor w/ 3 levels "Human Resources",...: 3 2
2 2 2 2 2 2 2 2 ...
## $ DistanceFromHome   : int  1 8 2 3 2 2 3 24 23 27 ...
## $ Education          : int  2 1 2 4 1 2 3 1 3 3 ...
## $ EducationField     : Factor w/ 6 levels "Human Resources",...: 2 2
5 2 4 2 4 2 2 4 ...
## $ EmployeeCount      : int  1 1 1 1 1 1 1 1 1 1 ...
## $ EmployeeNumber     : int  1 2 4 5 7 8 10 11 12 13 ...
## $ EnvironmentSatisfaction : int  2 3 4 4 1 4 3 4 4 3 ...
## $ Gender             : Factor w/ 2 levels "Female","Male": 1 2 2 1 2
2 1 2 2 2 ...
## $ HourlyRate         : int  94 61 92 56 40 79 81 67 44 94 ...
## $ JobInvolvement     : int  3 2 2 3 3 3 4 3 2 3 ...
## $ JobLevel           : int  2 2 1 1 1 1 1 1 3 2 ...
## $ JobRole            : Factor w/ 9 levels "Healthcare
```

```

Representative",...: 8 7 3 7 3 3 3 3 5 1 ...
## $ JobSatisfaction      : int  4 2 3 3 2 4 1 3 3 3 ...
## $ MaritalStatus        : Factor w/ 3 levels "Divorced","Married",...: 3
2 3 2 2 3 2 1 3 2 ...
## $ MonthlyIncome        : int  5993 5130 2090 2909 3468 3068 2670 2693
9526 5237 ...
## $ MonthlyRate          : int  19479 24907 2396 23159 16632 11864 9964
13335 8787 16577 ...
## $ NumCompaniesWorked   : int   8 1 6 1 9 0 4 1 0 6 ...
## $ Over18               : Factor w/ 1 level "Y": 1 1 1 1 1 1 1 1 1 1
...
## $ OverTime             : Factor w/ 2 levels "No","Yes": 2 1 2 2 1 1 2
1 1 1 ...
## $ PercentSalaryHike     : int  11 23 15 11 12 13 20 22 21 13 ...
## $ PerformanceRating     : int   3 4 3 3 3 3 4 4 4 3 ...
## $ RelationshipSatisfaction: int   1 4 2 3 4 3 1 2 2 2 ...
## $ StandardHours         : int  80 80 80 80 80 80 80 80 80 80 ...
## $ StockOptionLevel      : int   0 1 0 0 1 0 3 1 0 2 ...
## $ TotalWorkingYears     : int   8 10 7 8 6 8 12 1 10 17 ...
## $ TrainingTimesLastYear : int   0 3 3 3 3 2 3 2 2 3 ...
## $ WorkLifeBalance       : int   1 3 3 3 3 2 2 3 3 2 ...
## $ YearsAtCompany        : int   6 10 0 8 2 7 1 1 9 7 ...
## $ YearsInCurrentRole    : int   4 7 0 7 2 7 0 0 7 7 ...
## $ YearsSinceLastPromotion : int   0 1 0 3 2 3 0 0 1 7 ...
## $ YearsWithCurrManager  : int   5 7 0 0 2 6 0 0 8 7 ...

df1<-dataNW

```

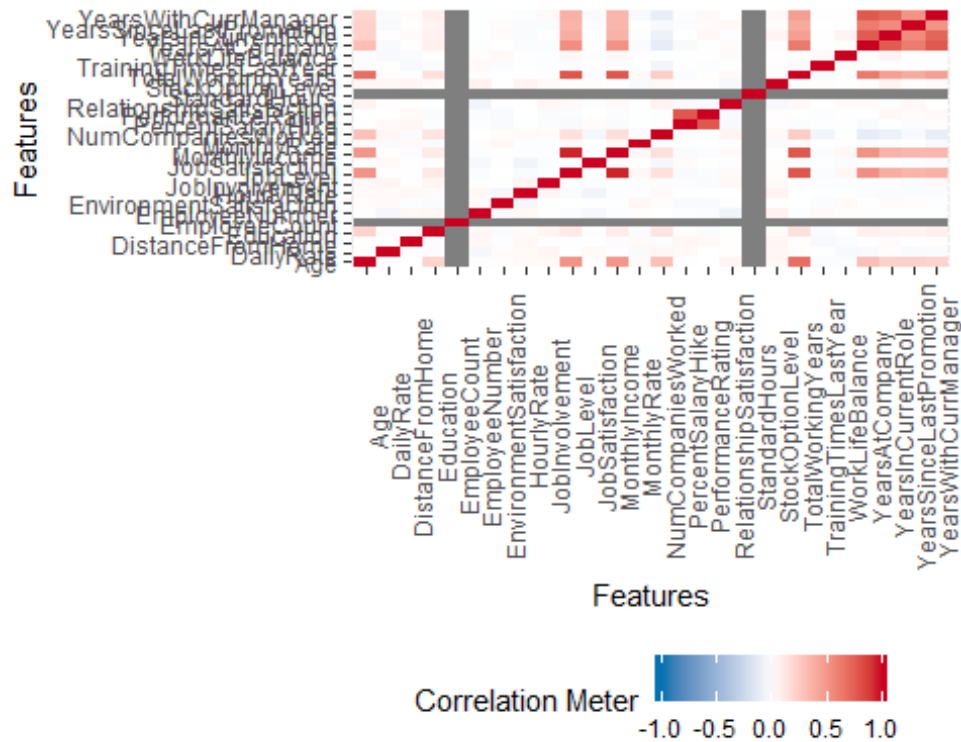
## CORRELATION MATRIX

```
plot_correlation(df1, type = 'continuous')
```

```

## Warning in cor(x = structure(list(Age = c(41L, 49L, 37L, 33L, 27L, 32L, :
## the standard deviation is zero

```



```
names(df1)
```

```
## [1] "Age" "Attrition"
## [3] "BusinessTravel" "DailyRate"
## [5] "Department" "DistanceFromHome"
## [7] "Education" "EducationField"
## [9] "EmployeeCount" "EmployeeNumber"
## [11] "EnvironmentSatisfaction" "Gender"
## [13] "HourlyRate" "JobInvolvement"
## [15] "JobLevel" "JobRole"
## [17] "JobSatisfaction" "MaritalStatus"
## [19] "MonthlyIncome" "MonthlyRate"
## [21] "NumCompaniesWorked" "Over18"
## [23] "OverTime" "PercentSalaryHike"
## [25] "PerformanceRating" "RelationshipSatisfaction"
## [27] "StandardHours" "StockOptionLevel"
## [29] "TotalWorkingYears" "TrainingTimesLastYear"
## [31] "WorkLifeBalance" "YearsAtCompany"
## [33] "YearsInCurrentRole" "YearsSinceLastPromotion"
## [35] "YearsWithCurrManager"
```

DIVIDING THE DATASET INTO TRAIN AND TEST SET

```
library(caTools)
```

```
## Warning: package 'caTools' was built under R version 3.4.3
```

*#Splitting the data*

```
set.seed(123)
```

```
indices = sample.split(df1$Attrition, SplitRatio = 0.7)
```

```
train = df1[indices,]
```

```
validation = df1[!(indices),]
```

MODEL1

LOGISTIC REGRESSION

```
colnames(train)
```

```
## [1] "Age" "Attrition"
## [3] "BusinessTravel" "DailyRate"
## [5] "Department" "DistanceFromHome"
## [7] "Education" "EducationField"
## [9] "EmployeeCount" "EmployeeNumber"
## [11] "EnvironmentSatisfaction" "Gender"
## [13] "HourlyRate" "JobInvolvement"
## [15] "JobLevel" "JobRole"
## [17] "JobSatisfaction" "MaritalStatus"
## [19] "MonthlyIncome" "MonthlyRate"
## [21] "NumCompaniesWorked" "Over18"
## [23] "OverTime" "PercentSalaryHike"
## [25] "PerformanceRating" "RelationshipSatisfaction"
## [27] "StandardHours" "StockOptionLevel"
## [29] "TotalWorkingYears" "TrainingTimesLastYear"
## [31] "WorkLifeBalance" "YearsAtCompany"
## [33] "YearsInCurrentRole" "YearsSinceLastPromotion"
## [35] "YearsWithCurrManager"
```

*#Build the first model using all variables*

```
model_1 = glm(Attrition ~ Age
```

```
+BusinessTravel+DailyRate+Department+DistanceFromHome+Education+EducationField+EnvironmentSatisfaction+Gender+HourlyRate+JobInvolvement+JobLevel+JobRole+JobSatisfaction+MaritalStatus+MonthlyIncome+MonthlyRate+NumCompaniesWorked+OverTime+PercentSalaryHike+PerformanceRating+RelationshipSatisfaction+StandardHours+StockOptionLevel+TotalWorkingYears+TrainingTimesLastYear+WorkLifeBalance+YearsAtCompany+YearsInCurrentRole+YearsSinceLastPromotion+YearsWithCurrManager, data = train, family = "binomial")
```

```
summary(model_1)
```

```
##
```

```
## Call:
```

```
## glm(formula = Attrition ~ Age + BusinessTravel + DailyRate +
```

```
## Department + DistanceFromHome + Education + EducationField +
```

```
## EnvironmentSatisfaction + Gender + HourlyRate + JobInvolvement +
```

```
## JobLevel + JobRole + JobSatisfaction + MaritalStatus + MonthlyIncome +
```

```
## MonthlyRate + NumCompaniesWorked + OverTime + PercentSalaryHike +
```

```
## PerformanceRating + RelationshipSatisfaction + StandardHours +
```

```
## StockOptionLevel + TotalWorkingYears + TrainingTimesLastYear +
```

```

##      WorkLifeBalance + YearsAtCompany + YearsInCurrentRole +
YearsSinceLastPromotion +
##      YearsWithCurrManager, family = "binomial", data = train)
##
## Deviance Residuals:
##      Min        1Q    Median        3Q        Max
## -1.6681   -0.4819   -0.2502   -0.0908    3.4600
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -1.059e+01  7.008e+02  -0.015  0.987940
## Age           -1.804e-02  1.582e-02  -1.140  0.254165
## BusinessTravelTravel_Frequently  1.989e+00  5.396e-01   3.687  0.000227
## BusinessTravelTravel_Rarely     1.037e+00  5.100e-01   2.034  0.041932
## DailyRate     -1.839e-04  2.630e-04  -0.699  0.484450
## DepartmentResearch & Development  1.348e+01  7.008e+02   0.019  0.984649
## DepartmentSales  1.423e+01  7.008e+02   0.020  0.983796
## DistanceFromHome  4.144e-02  1.302e-02   3.184  0.001454
## Education      2.685e-02  1.075e-01   0.250  0.802788
## EducationFieldLife Sciences  -9.295e-01  1.073e+00  -0.866  0.386263
## EducationFieldMarketing  -7.117e-01  1.123e+00  -0.634  0.526292
## EducationFieldMedical  -1.368e+00  1.069e+00  -1.279  0.200766
## EducationFieldOther  -1.014e+00  1.131e+00  -0.896  0.370034
## EducationFieldTechnical Degree -2.719e-01  1.098e+00  -0.248  0.804510
## EnvironmentSatisfaction -4.078e-01  9.939e-02  -4.103  4.09e-05
## GenderMale      3.583e-01  2.204e-01   1.625  0.104058
## HourlyRate      9.030e-04  5.231e-03   0.173  0.862952
## JobInvolvement  -4.798e-01  1.496e-01  -3.207  0.001343
## JobLevel        6.848e-02  3.702e-01   0.185  0.853224
## JobRoleHuman Resources  1.503e+01  7.008e+02   0.021  0.982884
## JobRoleLaboratory Technician  2.157e+00  6.548e-01   3.294  0.000987
## JobRoleManager   7.738e-01  1.121e+00   0.690  0.490189
## JobRoleManufacturing Director  8.420e-01  7.104e-01   1.185  0.235910
## JobRoleResearch Director  -8.839e-01  1.338e+00  -0.661  0.508748
## JobRoleResearch Scientist  8.761e-01  6.704e-01   1.307  0.191231
## JobRoleSales Executive  8.892e-01  1.314e+00   0.676  0.498732
## JobRoleSales Representative  1.985e+00  1.373e+00   1.446  0.148235
## JobSatisfaction  -4.100e-01  9.862e-02  -4.157  3.22e-05
## MaritalStatusMarried  3.086e-01  3.209e-01   0.962  0.336192
## MaritalStatusSingle  9.227e-01  4.069e-01   2.268  0.023357
## MonthlyIncome    -2.048e-05  9.401e-05  -0.218  0.827550
## MonthlyRate      1.907e-05  1.519e-05   1.255  0.209312
## NumCompaniesWorked  1.672e-01  4.741e-02   3.528  0.000419
## OverTimeYes      1.887e+00  2.354e-01   8.014  1.11e-15
## PercentSalaryHike -2.948e-02  4.652e-02  -0.634  0.526186
## PerformanceRating -3.232e-01  5.014e-01  -0.645  0.519218
## RelationshipSatisfaction -2.236e-01  9.700e-02  -2.305  0.021171
## StandardHours      NA           NA           NA           NA
## StockOptionLevel  -2.891e-01  1.925e-01  -1.502  0.133094
## TotalWorkingYears -6.822e-02  3.474e-02  -1.964  0.049547

```

## TrainingTimesLastYear	-1.188e-01	8.587e-02	-1.384	0.166388
## WorkLifeBalance	-5.179e-01	1.491e-01	-3.473	0.000515
## YearsAtCompany	6.794e-02	4.760e-02	1.427	0.153494
## YearsInCurrentRole	-1.445e-01	5.601e-02	-2.581	0.009861
## YearsSinceLastPromotion	1.391e-01	5.060e-02	2.750	0.005968
## YearsWithCurrManager	-8.400e-02	5.929e-02	-1.417	0.156520
##				
## (Intercept)				
## Age				
## BusinessTravelTravel_Frequently	***			
## BusinessTravelTravel_Rarely	*			
## DailyRate				
## DepartmentResearch & Development				
## DepartmentSales				
## DistanceFromHome	**			
## Education				
## EducationFieldLife Sciences				
## EducationFieldMarketing				
## EducationFieldMedical				
## EducationFieldOther				
## EducationFieldTechnical Degree				
## EnvironmentSatisfaction	***			
## GenderMale				
## HourlyRate				
## JobInvolvement	**			
## JobLevel				
## JobRoleHuman Resources				
## JobRoleLaboratory Technician	***			
## JobRoleManager				
## JobRoleManufacturing Director				
## JobRoleResearch Director				
## JobRoleResearch Scientist				
## JobRoleSales Executive				
## JobRoleSales Representative				
## JobSatisfaction	***			
## MaritalStatusMarried				
## MaritalStatusSingle	*			
## MonthlyIncome				
## MonthlyRate				
## NumCompaniesWorked	***			
## OverTimeYes	***			
## PercentSalaryHike				
## PerformanceRating				
## RelationshipSatisfaction	*			
## StandardHours				
## StockOptionLevel				
## TotalWorkingYears	*			
## TrainingTimesLastYear				
## WorkLifeBalance	***			
## YearsAtCompany				



```
## YearsInCurrentRole          **
## YearsSinceLastPromotion     **
## YearsWithCurrManager
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 909.34  on 1028  degrees of freedom
## Residual deviance: 602.56  on  984  degrees of freedom
## AIC: 692.56
##
## Number of Fisher Scoring iterations: 15
```

Using stepAIC for variable selection, which is a iterative process of adding or removing variables, in order to get a subset of variables that gives the best performing model.

```
library(MASS)

##
## Attaching package: 'MASS'

## The following object is masked from 'package:dplyr':
##
##      select

model_2<- stepAIC(model_1, direction="both")

## Start:  AIC=692.56
## Attrition ~ Age + BusinessTravel + DailyRate + Department +
DistanceFromHome +
##      Education + EducationField + EnvironmentSatisfaction + Gender +
##      HourlyRate + JobInvolvement + JobLevel + JobRole + JobSatisfaction +
##      MaritalStatus + MonthlyIncome + MonthlyRate + NumCompaniesWorked +
##      OverTime + PercentSalaryHike + PerformanceRating +
RelationshipSatisfaction +
##      StandardHours + StockOptionLevel + TotalWorkingYears +
TrainingTimesLastYear +
##      WorkLifeBalance + YearsAtCompany + YearsInCurrentRole +
YearsSinceLastPromotion +
##      YearsWithCurrManager
##
##
## Step:  AIC=692.56
## Attrition ~ Age + BusinessTravel + DailyRate + Department +
DistanceFromHome +
##      Education + EducationField + EnvironmentSatisfaction + Gender +
##      HourlyRate + JobInvolvement + JobLevel + JobRole + JobSatisfaction +
##      MaritalStatus + MonthlyIncome + MonthlyRate + NumCompaniesWorked +
##      OverTime + PercentSalaryHike + PerformanceRating +
RelationshipSatisfaction +
```

```

##      StockOptionLevel + TotalWorkingYears + TrainingTimesLastYear +
##      WorkLifeBalance + YearsAtCompany + YearsInCurrentRole +
YearsSinceLastPromotion +
##      YearsWithCurrManager
##
##              Df Deviance      AIC
## - Department          2    604.10  690.10
## - HourlyRate           1    602.59  690.59
## - JobLevel             1    602.59  690.59
## - MonthlyIncome        1    602.60  690.60
## - Education            1    602.62  690.62
## - PercentSalaryHike     1    602.96  690.96
## - PerformanceRating     1    602.97  690.97
## - DailyRate            1    603.05  691.05
## - Age                  1    603.88  691.88
## - EducationField        5    612.11  692.11
## - MonthlyRate           1    604.14  692.14
## - TrainingTimesLastYear 1    604.50  692.50
## - YearsWithCurrManager  1    604.54  692.54
## <none>                  1    602.56  692.56
## - YearsAtCompany        1    604.58  692.58
## - StockOptionLevel      1    604.89  692.89
## - Gender                1    605.24  693.24
## - MaritalStatus         2    608.52  694.52
## - TotalWorkingYears     1    606.58  694.58
## - RelationshipSatisfaction 1    607.91  695.91
## - YearsInCurrentRole    1    609.19  697.19
## - YearsSinceLastPromotion 1    610.31  698.31
## - DistanceFromHome      1    612.63  700.63
## - JobInvolvement        1    612.95  700.95
## - NumCompaniesWorked    1    614.75  702.75
## - WorkLifeBalance       1    614.77  702.77
## - JobRole               8    632.01  706.01
## - EnvironmentSatisfaction 1    619.91  707.91
## - JobSatisfaction       1    620.43  708.43
## - BusinessTravel        2    624.80  710.80
## - OverTime              1    673.26  761.26
##
## Step:  AIC=690.1
## Attrition ~ Age + BusinessTravel + DailyRate + DistanceFromHome +
##      Education + EducationField + EnvironmentSatisfaction + Gender +
##      HourlyRate + JobInvolvement + JobLevel + JobRole + JobSatisfaction +
##      MaritalStatus + MonthlyIncome + MonthlyRate + NumCompaniesWorked +
##      OverTime + PercentSalaryHike + PerformanceRating +
RelationshipSatisfaction +
##      StockOptionLevel + TotalWorkingYears + TrainingTimesLastYear +
##      WorkLifeBalance + YearsAtCompany + YearsInCurrentRole +
YearsSinceLastPromotion +
##      YearsWithCurrManager
##

```

```

##                               Df Deviance    AIC
## - JobLevel                   1    604.12 688.12
## - HourlyRate                 1    604.14 688.14
## - MonthlyIncome              1    604.15 688.15
## - Education                  1    604.15 688.15
## - PercentSalaryHike          1    604.45 688.45
## - PerformanceRating          1    604.53 688.53
## - DailyRate                  1    604.56 688.56
## - Age                        1    605.49 689.49
## - MonthlyRate                1    605.73 689.73
## - TrainingTimesLastYear      1    605.91 689.91
## - YearsWithCurrManager       1    605.96 689.96
## - EducationField             5    614.03 690.03
## <none>                       1    604.10 690.10
## - YearsAtCompany            1    606.12 690.12
## - StockOptionLevel          1    606.61 690.61
## - Gender                     1    606.72 690.72
## - MaritalStatus              2    609.96 691.96
## - TotalWorkingYears          1    608.00 692.00
## + Department                 2    602.56 692.56
## - RelationshipSatisfaction    1    609.53 693.53
## - YearsInCurrentRole         1    610.79 694.79
## - YearsSinceLastPromotion    1    611.87 695.87
## - DistanceFromHome           1    613.88 697.88
## - JobInvolvement             1    615.14 699.14
## - WorkLifeBalance            1    616.16 700.16
## - NumCompaniesWorked         1    616.50 700.50
## - EnvironmentSatisfaction    1    621.45 705.45
## - JobSatisfaction            1    622.01 706.01
## - BusinessTravel             2    626.34 708.34
## - JobRole                    8    642.99 712.99
## - OverTime                   1    676.05 760.05
##
## Step:  AIC=688.12
## Attrition ~ Age + BusinessTravel + DailyRate + DistanceFromHome +
##      Education + EducationField + EnvironmentSatisfaction + Gender +
##      HourlyRate + JobInvolvement + JobRole + JobSatisfaction +
##      MaritalStatus + MonthlyIncome + MonthlyRate + NumCompaniesWorked +
##      OverTime + PercentSalaryHike + PerformanceRating +
##      RelationshipSatisfaction +
##      StockOptionLevel + TotalWorkingYears + TrainingTimesLastYear +
##      WorkLifeBalance + YearsAtCompany + YearsInCurrentRole +
##      YearsSinceLastPromotion +
##      YearsWithCurrManager
##
##                               Df Deviance    AIC
## - MonthlyIncome              1    604.15 686.15
## - HourlyRate                 1    604.16 686.16
## - Education                  1    604.17 686.17
## - PercentSalaryHike          1    604.48 686.48

```

```

## - PerformanceRating      1    604.55 686.55
## - DailyRate              1    604.59 686.59
## - Age                    1    605.51 687.51
## - MonthlyRate            1    605.78 687.78
## - TrainingTimesLastYear  1    605.93 687.93
## - YearsWithCurrManager   1    605.98 687.98
## <none>                   1    604.12 688.12
## - EducationField         5    614.16 688.16
## - YearsAtCompany         1    606.27 688.27
## - StockOptionLevel       1    606.63 688.63
## - Gender                 1    606.75 688.75
## - MaritalStatus         2    610.06 690.06
## + JobLevel              1    604.10 690.10
## - TotalWorkingYears      1    608.11 690.11
## + Department            2    602.59 690.59
## - RelationshipSatisfaction 1    609.58 691.58
## - YearsInCurrentRole     1    611.00 693.00
## - YearsSinceLastPromotion 1    611.87 693.87
## - DistanceFromHome       1    614.01 696.01
## - JobInvolvement         1    615.14 697.14
## - WorkLifeBalance        1    616.17 698.17
## - NumCompaniesWorked     1    616.54 698.54
## - EnvironmentSatisfaction 1    621.49 703.49
## - JobSatisfaction        1    622.03 704.03
## - BusinessTravel         2    626.40 706.40
## - JobRole                8    643.05 711.05
## - OverTime               1    676.05 758.05
##
## Step:  AIC=686.15
## Attrition ~ Age + BusinessTravel + DailyRate + DistanceFromHome +
##      Education + EducationField + EnvironmentSatisfaction + Gender +
##      HourlyRate + JobInvolvement + JobRole + JobSatisfaction +
##      MaritalStatus + MonthlyRate + NumCompaniesWorked + OverTime +
##      PercentSalaryHike + PerformanceRating + RelationshipSatisfaction +
##      StockOptionLevel + TotalWorkingYears + TrainingTimesLastYear +
##      WorkLifeBalance + YearsAtCompany + YearsInCurrentRole +
##      YearsSinceLastPromotion +
##      YearsWithCurrManager
##
##              Df Deviance    AIC
## - HourlyRate      1    604.18 684.18
## - Education        1    604.20 684.20
## - PercentSalaryHike 1    604.51 684.51
## - PerformanceRating 1    604.57 684.57
## - DailyRate        1    604.62 684.62
## - Age              1    605.54 685.54
## - MonthlyRate      1    605.80 685.80
## - TrainingTimesLastYear 1    605.96 685.96
## - YearsWithCurrManager 1    605.98 685.98
## <none>             1    604.15 686.15

```

```

## - EducationField          5   614.17 686.17
## - YearsAtCompany          1   606.27 686.27
## - StockOptionLevel        1   606.66 686.66
## - Gender                   1   606.76 686.76
## - MaritalStatus           2   610.06 688.06
## + MonthlyIncome           1   604.12 688.12
## + JobLevel                 1   604.15 688.15
## + Department               2   602.61 688.61
## - TotalWorkingYears        1   609.14 689.14
## - RelationshipSatisfaction  1   609.58 689.58
## - YearsInCurrentRole       1   611.01 691.01
## - YearsSinceLastPromotion  1   611.88 691.88
## - DistanceFromHome         1   614.05 694.05
## - JobInvolvement           1   615.15 695.15
## - WorkLifeBalance          1   616.25 696.25
## - NumCompaniesWorked       1   616.55 696.55
## - EnvironmentSatisfaction   1   621.59 701.59
## - JobSatisfaction           1   622.14 702.14
## - BusinessTravel           2   626.40 704.40
## - JobRole                   8   648.20 714.20
## - OverTime                  1   676.05 756.05
##
## Step:  AIC=684.18
## Attrition ~ Age + BusinessTravel + DailyRate + DistanceFromHome +
##      Education + EducationField + EnvironmentSatisfaction + Gender +
##      JobInvolvement + JobRole + JobSatisfaction + MaritalStatus +
##      MonthlyRate + NumCompaniesWorked + OverTime + PercentSalaryHike +
##      PerformanceRating + RelationshipSatisfaction + StockOptionLevel +
##      TotalWorkingYears + TrainingTimesLastYear + WorkLifeBalance +
##      YearsAtCompany + YearsInCurrentRole + YearsSinceLastPromotion +
##      YearsWithCurrManager
##
##              Df Deviance    AIC
## - Education          1   604.23 682.23
## - PercentSalaryHike   1   604.54 682.54
## - PerformanceRating    1   604.61 682.61
## - DailyRate           1   604.65 682.65
## - Age                 1   605.56 683.56
## - MonthlyRate         1   605.85 683.85
## - TrainingTimesLastYear 1   606.02 684.02
## - YearsWithCurrManager 1   606.03 684.03
## <none>                604.18 684.18
## - EducationField      5   614.29 684.29
## - YearsAtCompany       1   606.35 684.35
## - StockOptionLevel     1   606.69 684.69
## - Gender               1   606.77 684.77
## - MaritalStatus        2   610.10 686.10
## + HourlyRate           1   604.15 686.15
## + MonthlyIncome        1   604.16 686.16
## + JobLevel             1   604.18 686.18

```

```

## + Department                2    602.64  686.64
## - TotalWorkingYears          1    609.19  687.19
## - RelationshipSatisfaction    1    609.63  687.63
## - YearsInCurrentRole         1    611.07  689.07
## - YearsSinceLastPromotion    1    611.89  689.89
## - DistanceFromHome           1    614.12  692.12
## - JobInvolvement             1    615.21  693.21
## - WorkLifeBalance            1    616.30  694.30
## - NumCompaniesWorked         1    616.56  694.56
## - EnvironmentSatisfaction    1    621.62  699.62
## - JobSatisfaction            1    622.31  700.31
## - BusinessTravel             2    626.41  702.41
## - JobRole                    8    648.22  712.22
## - OverTime                   1    676.10  754.10
##
## Step:  AIC=682.23
## Attrition ~ Age + BusinessTravel + DailyRate + DistanceFromHome +
##      EducationField + EnvironmentSatisfaction + Gender + JobInvolvement +
##      JobRole + JobSatisfaction + MaritalStatus + MonthlyRate +
##      NumCompaniesWorked + OverTime + PercentSalaryHike + PerformanceRating
##      +
##      RelationshipSatisfaction + StockOptionLevel + TotalWorkingYears +
##      TrainingTimesLastYear + WorkLifeBalance + YearsAtCompany +
##      YearsInCurrentRole + YearsSinceLastPromotion + YearsWithCurrManager
##
##              Df Deviance    AIC
## - PercentSalaryHike      1    604.58  680.58
## - PerformanceRating      1    604.67  680.67
## - DailyRate              1    604.71  680.71
## - Age                    1    605.56  681.56
## - MonthlyRate            1    605.86  681.86
## - TrainingTimesLastYear  1    606.09  682.09
## - YearsWithCurrManager   1    606.10  682.10
## <none>                  604.23  682.23
## - EducationField         5    614.40  682.40
## - YearsAtCompany         1    606.41  682.41
## - StockOptionLevel       1    606.70  682.70
## - Gender                 1    606.79  682.79
## + Education              1    604.18  684.18
## + HourlyRate             1    604.20  684.20
## + MonthlyIncome          1    604.21  684.21
## - MaritalStatus          2    610.21  684.21
## + JobLevel               1    604.23  684.23
## + Department             2    602.69  684.69
## - TotalWorkingYears      1    609.22  685.22
## - RelationshipSatisfaction 1    609.64  685.64
## - YearsInCurrentRole     1    611.08  687.08
## - YearsSinceLastPromotion 1    611.96  687.96
## - DistanceFromHome       1    614.31  690.31
## - JobInvolvement         1    615.22  691.22

```

```

## - WorkLifeBalance      1  616.32 692.32
## - NumCompaniesWorked   1  616.69 692.69
## - EnvironmentSatisfaction 1  621.66 697.66
## - JobSatisfaction       1  622.33 698.33
## - BusinessTravel       2  626.49 700.49
## - JobRole              8  648.32 710.32
## - OverTime             1  676.19 752.19
##
## Step: AIC=680.58
## Attrition ~ Age + BusinessTravel + DailyRate + DistanceFromHome +
##      EducationField + EnvironmentSatisfaction + Gender + JobInvolvement +
##      JobRole + JobSatisfaction + MaritalStatus + MonthlyRate +
##      NumCompaniesWorked + OverTime + PerformanceRating +
##      RelationshipSatisfaction +
##      StockOptionLevel + TotalWorkingYears + TrainingTimesLastYear +
##      WorkLifeBalance + YearsAtCompany + YearsInCurrentRole +
##      YearsSinceLastPromotion +
##      YearsWithCurrManager
##
##              Df Deviance    AIC
## - DailyRate      1   605.07 679.07
## - Age            1   605.96 679.96
## - MonthlyRate    1   606.06 680.06
## - YearsWithCurrManager 1   606.44 680.44
## - TrainingTimesLastYear 1   606.46 680.46
## <none>              604.58 680.58
## - YearsAtCompany  1   606.73 680.73
## - EducationField  5   614.92 680.92
## - Gender          1   607.12 681.12
## - StockOptionLevel 1   607.13 681.13
## - PerformanceRating 1   607.29 681.29
## + PercentSalaryHike 1   604.23 682.23
## + Education       1   604.54 682.54
## + HourlyRate      1   604.55 682.55
## + MonthlyIncome   1   604.55 682.55
## - MaritalStatus   2   610.56 682.56
## + JobLevel        1   604.58 682.58
## + Department      2   603.09 683.09
## - TotalWorkingYears 1   609.51 683.51
## - RelationshipSatisfaction 1  610.08 684.08
## - YearsInCurrentRole 1  611.35 685.35
## - YearsSinceLastPromotion 1  612.42 686.42
## - DistanceFromHome 1  614.53 688.53
## - JobInvolvement  1  615.55 689.55
## - WorkLifeBalance  1  616.44 690.44
## - NumCompaniesWorked 1  617.00 691.00
## - EnvironmentSatisfaction 1  621.77 695.77
## - JobSatisfaction  1  622.63 696.63
## - BusinessTravel   2  626.92 698.92
## - JobRole         8  648.61 708.61

```

```

## - OverTime          1    677.31 751.31
##
## Step:  AIC=679.07
## Attrition ~ Age + BusinessTravel + DistanceFromHome + EducationField +
##      EnvironmentSatisfaction + Gender + JobInvolvement + JobRole +
##      JobSatisfaction + MaritalStatus + MonthlyRate + NumCompaniesWorked +
##      OverTime + PerformanceRating + RelationshipSatisfaction +
##      StockOptionLevel + TotalWorkingYears + TrainingTimesLastYear +
##      WorkLifeBalance + YearsAtCompany + YearsInCurrentRole +
##      YearsSinceLastPromotion +
##      YearsWithCurrManager
##
##              Df Deviance    AIC
## - Age          1    606.56 678.56
## - MonthlyRate  1    606.59 678.59
## - YearsWithCurrManager  1    606.85 678.85
## - TrainingTimesLastYear  1    606.94 678.94
## <none>          605.07 679.07
## - YearsAtCompany  1    607.24 679.24
## - Gender          1    607.63 679.63
## - StockOptionLevel  1    607.64 679.64
## - EducationField  5    615.71 679.71
## - PerformanceRating  1    607.88 679.88
## + DailyRate       1    604.58 680.58
## + PercentSalaryHike  1    604.71 680.71
## + Education        1    605.01 681.01
## + MonthlyIncome    1    605.04 681.04
## + HourlyRate       1    605.05 681.05
## + JobLevel         1    605.07 681.07
## - MaritalStatus    2    611.19 681.19
## + Department       2    603.62 681.62
## - TotalWorkingYears  1    610.13 682.13
## - RelationshipSatisfaction  1    610.70 682.70
## - YearsInCurrentRole  1    612.13 684.13
## - YearsSinceLastPromotion  1    613.29 685.29
## - DistanceFromHome  1    615.28 687.28
## - JobInvolvement   1    616.21 688.21
## - WorkLifeBalance  1    616.84 688.84
## - NumCompaniesWorked  1    617.62 689.62
## - EnvironmentSatisfaction  1    622.20 694.20
## - JobSatisfaction   1    623.24 695.24
## - BusinessTravel    2    627.76 697.76
## - JobRole          8    649.00 707.00
## - OverTime         1    677.82 749.82
##
## Step:  AIC=678.56
## Attrition ~ BusinessTravel + DistanceFromHome + EducationField +
##      EnvironmentSatisfaction + Gender + JobInvolvement + JobRole +
##      JobSatisfaction + MaritalStatus + MonthlyRate + NumCompaniesWorked +
##      OverTime + PerformanceRating + RelationshipSatisfaction +

```



```

##      StockOptionLevel + TotalWorkingYears + TrainingTimesLastYear +
##      WorkLifeBalance + YearsAtCompany + YearsInCurrentRole +
YearsSinceLastPromotion +
##      YearsWithCurrManager
##
##
##      Df Deviance    AIC
## - YearsWithCurrManager      1    608.11  678.11
## - MonthlyRate                1    608.17  678.17
## <none>                      606.56  678.56
## - TrainingTimesLastYear      1    608.62  678.62
## - YearsAtCompany             1    608.88  678.88
## - Gender                    1    608.91  678.91
## - EducationField             5    617.02  679.02
## - StockOptionLevel           1    609.06  679.06
## + Age                       1    605.07  679.07
## - PerformanceRating          1    609.39  679.39
## + DailyRate                 1    605.96  679.96
## + PercentSalaryHike          1    606.14  680.14
## + MonthlyIncome             1    606.53  680.53
## + Education                  1    606.55  680.55
## + HourlyRate                 1    606.55  680.55
## + JobLevel                   1    606.55  680.55
## + Department                 2    605.03  681.03
## - MaritalStatus              2    613.09  681.09
## - RelationshipSatisfaction    1    612.51  682.51
## - YearsInCurrentRole         1    613.61  683.61
## - YearsSinceLastPromotion     1    614.49  684.49
## - DistanceFromHome           1    616.56  686.56
## - TotalWorkingYears          1    617.49  687.49
## - WorkLifeBalance            1    617.92  687.92
## - JobInvolvement             1    617.93  687.93
## - NumCompaniesWorked         1    618.15  688.15
## - EnvironmentSatisfaction     1    623.84  693.84
## - JobSatisfaction            1    625.31  695.31
## - BusinessTravel             2    629.84  697.84
## - JobRole                    8    650.35  706.35
## - OverTime                   1    678.86  748.86
##
## Step:  AIC=678.11
## Attrition ~ BusinessTravel + DistanceFromHome + EducationField +
##      EnvironmentSatisfaction + Gender + JobInvolvement + JobRole +
##      JobSatisfaction + MaritalStatus + MonthlyRate + NumCompaniesWorked +
##      OverTime + PerformanceRating + RelationshipSatisfaction +
##      StockOptionLevel + TotalWorkingYears + TrainingTimesLastYear +
##      WorkLifeBalance + YearsAtCompany + YearsInCurrentRole +
YearsSinceLastPromotion
##
##      Df Deviance    AIC
## - YearsAtCompany            1    609.28  677.28
## - MonthlyRate               1    609.87  677.87

```

```

## - TrainingTimesLastYear      1   610.09 678.09
## <none>                        1   608.11 678.11
## + YearsWithCurrManager       1   606.56 678.56
## - Gender                     1   610.60 678.60
## - EducationField             5   618.78 678.78
## - StockOptionLevel           1   610.84 678.84
## + Age                        1   606.85 678.85
## - PerformanceRating          1   611.05 679.05
## + DailyRate                  1   607.60 679.60
## + PercentSalaryHike          1   607.70 679.70
## + Education                  1   608.08 680.08
## + JobLevel                   1   608.10 680.10
## + HourlyRate                 1   608.10 680.10
## + MonthlyIncome              1   608.10 680.10
## - MaritalStatus              2   614.37 680.37
## + Department                 2   606.70 680.70
## - RelationshipSatisfaction    1   614.08 682.08
## - YearsSinceLastPromotion    1   615.36 683.36
## - YearsInCurrentRole         1   617.34 685.34
## - DistanceFromHome           1   618.00 686.00
## - WorkLifeBalance            1   619.03 687.03
## - TotalWorkingYears          1   619.43 687.43
## - NumCompaniesWorked         1   620.12 688.12
## - JobInvolvement             1   620.20 688.20
## - EnvironmentSatisfaction    1   625.21 693.21
## - JobSatisfaction            1   626.50 694.50
## - BusinessTravel             2   631.09 697.09
## - JobRole                    8   652.03 706.03
## - OverTime                   1   680.54 748.54
##
## Step:  AIC=677.28
## Attrition ~ BusinessTravel + DistanceFromHome + EducationField +
##      EnvironmentSatisfaction + Gender + JobInvolvement + JobRole +
##      JobSatisfaction + MaritalStatus + MonthlyRate + NumCompaniesWorked +
##      OverTime + PerformanceRating + RelationshipSatisfaction +
##      StockOptionLevel + TotalWorkingYears + TrainingTimesLastYear +
##      WorkLifeBalance + YearsInCurrentRole + YearsSinceLastPromotion
##
##              Df Deviance    AIC
## - MonthlyRate      1   610.89 676.89
## - TrainingTimesLastYear  1   611.24 677.24
## <none>              1   609.28 677.28
## - EducationField    5   619.36 677.36
## - Gender            1   611.75 677.75
## + Age              1   607.82 677.82
## - StockOptionLevel  1   611.97 677.97
## + YearsAtCompany    1   608.11 678.11
## - PerformanceRating 1   612.24 678.24
## + DailyRate         1   608.71 678.71
## + YearsWithCurrManager 1   608.88 678.88

```

```

## + PercentSalaryHike      1  608.89 678.89
## + JobLevel               1  609.18 679.18
## + HourlyRate             1  609.25 679.25
## + Education              1  609.26 679.26
## + MonthlyIncome          1  609.27 679.27
## - MaritalStatus          2  615.65 679.65
## + Department             2  607.78 679.78
## - RelationshipSatisfaction 1  615.21 681.21
## - YearsInCurrentRole     1  618.06 684.06
## - DistanceFromHome       1  618.92 684.92
## - TotalWorkingYears      1  619.59 685.59
## - YearsSinceLastPromotion 1  620.11 686.11
## - WorkLifeBalance        1  620.15 686.15
## - NumCompaniesWorked     1  620.19 686.19
## - JobInvolvement         1  621.64 687.64
## - EnvironmentSatisfaction 1  626.61 692.61
## - JobSatisfaction        1  627.56 693.56
## - BusinessTravel         2  632.66 696.66
## - JobRole                8  652.61 704.61
## - OverTime               1  681.53 747.53
##
## Step:  AIC=676.89
## Attrition ~ BusinessTravel + DistanceFromHome + EducationField +
##      EnvironmentSatisfaction + Gender + JobInvolvement + JobRole +
##      JobSatisfaction + MaritalStatus + NumCompaniesWorked + OverTime +
##      PerformanceRating + RelationshipSatisfaction + StockOptionLevel +
##      TotalWorkingYears + TrainingTimesLastYear + WorkLifeBalance +
##      YearsInCurrentRole + YearsSinceLastPromotion
##
##
##      Df Deviance    AIC
## - EducationField      5  620.88 676.88
## <none>                 610.89 676.89
## - TrainingTimesLastYear 1  612.98 676.98
## - Gender               1  613.20 677.20
## + MonthlyRate          1  609.28 677.28
## + Age                  1  609.37 677.37
## + YearsAtCompany       1  609.87 677.87
## - StockOptionLevel     1  613.88 677.88
## - PerformanceRating    1  614.18 678.18
## + DailyRate            1  610.31 678.31
## + YearsWithCurrManager  1  610.38 678.38
## + PercentSalaryHike    1  610.65 678.65
## + JobLevel             1  610.75 678.75
## + HourlyRate           1  610.85 678.85
## + MonthlyIncome        1  610.88 678.88
## + Education            1  610.89 678.89
## - MaritalStatus        2  617.20 679.20
## + Department           2  609.39 679.39
## - RelationshipSatisfaction 1  616.90 680.90
## - YearsInCurrentRole   1  620.03 684.03

```

```

## - TotalWorkingYears      1  621.06 685.06
## - DistanceFromHome       1  621.17 685.17
## - YearsSinceLastPromotion 1  621.51 685.51
## - NumCompaniesWorked     1  621.77 685.77
## - WorkLifeBalance        1  621.78 685.78
## - JobInvolvement         1  623.11 687.11
## - EnvironmentSatisfaction 1  627.65 691.65
## - JobSatisfaction        1  629.19 693.19
## - BusinessTravel         2  634.18 696.18
## - JobRole                8  653.70 703.70
## - OverTime               1  683.35 747.35
##
## Step:  AIC=676.88
## Attrition ~ BusinessTravel + DistanceFromHome + EnvironmentSatisfaction +
##      Gender + JobInvolvement + JobRole + JobSatisfaction + MaritalStatus +
##      NumCompaniesWorked + OverTime + PerformanceRating +
##      RelationshipSatisfaction +
##      StockOptionLevel + TotalWorkingYears + TrainingTimesLastYear +
##      WorkLifeBalance + YearsInCurrentRole + YearsSinceLastPromotion
##
##              Df Deviance    AIC
## <none>              620.88 676.88
## + EducationField    5  610.89 676.89
## + MonthlyRate       1  619.36 677.36
## - Gender            1  623.45 677.45
## - TrainingTimesLastYear 1  623.53 677.53
## + Age              1  619.59 677.59
## + YearsWithCurrManager 1  620.00 678.00
## + DailyRate        1  620.03 678.03
## - StockOptionLevel  1  624.17 678.17
## - PerformanceRating 1  624.23 678.23
## + YearsAtCompany    1  620.38 678.38
## + PercentSalaryHike  1  620.46 678.46
## + JobLevel          1  620.48 678.48
## - MaritalStatus     2  626.66 678.66
## + HourlyRate        1  620.77 678.77
## + MonthlyIncome     1  620.78 678.78
## + Education         1  620.85 678.85
## + Department        2  619.24 679.24
## - RelationshipSatisfaction 1  626.95 680.95
## - YearsInCurrentRole 1  629.24 683.24
## - WorkLifeBalance   1  630.94 684.94
## - YearsSinceLastPromotion 1  630.98 684.98
## - DistanceFromHome  1  631.00 685.00
## - TotalWorkingYears 1  631.27 685.27
## - NumCompaniesWorked 1  631.67 685.67
## - JobInvolvement    1  633.48 687.48
## - EnvironmentSatisfaction 1  636.69 690.69
## - JobSatisfaction    1  639.04 693.04
## - BusinessTravel    2  644.84 696.84

```

```
## - JobRole          8   671.13 711.13
## - OverTime         1   692.20 746.20
```

```
summary(model_2)
```

```
##
## Call:
## glm(formula = Attrition ~ BusinessTravel + DistanceFromHome +
##     EnvironmentSatisfaction + Gender + JobInvolvement + JobRole +
##     JobSatisfaction + MaritalStatus + NumCompaniesWorked + OverTime +
##     PerformanceRating + RelationshipSatisfaction + StockOptionLevel +
##     TotalWorkingYears + TrainingTimesLastYear + WorkLifeBalance +
##     YearsInCurrentRole + YearsSinceLastPromotion, family = "binomial",
##     data = train)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.7010  -0.4917  -0.2623  -0.0993   3.8093
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      2.00424    1.48474   1.350 0.177051
## BusinessTravelTravel_Frequently  2.04122    0.53275   3.831 0.000127 ***
## BusinessTravelTravel_Rarely      1.09637    0.50460   2.173 0.029800 *
## DistanceFromHome      0.04036    0.01265   3.190 0.001422 **
## EnvironmentSatisfaction    -0.37848    0.09630  -3.930 8.49e-05 ***
## GenderMale              0.34154    0.21435   1.593 0.111076
## JobInvolvement          -0.51003    0.14446  -3.531 0.000414 ***
## JobRoleHuman Resources      1.94824    0.75083   2.595 0.009465 **
## JobRoleLaboratory Technician  2.10220    0.59416   3.538 0.000403 ***
## JobRoleManager           1.15109    0.81172   1.418 0.156165
## JobRoleManufacturing Director  0.76195    0.70332   1.083 0.278646
## JobRoleResearch Director   -0.65599    1.21877  -0.538 0.590413
## JobRoleResearch Scientist    0.86667    0.60489   1.433 0.151919
## JobRoleSales Executive      1.77094    0.58694   3.017 0.002551 **
## JobRoleSales Representative   2.82936    0.65114   4.345 1.39e-05 ***
## JobSatisfaction          -0.40153    0.09558  -4.201 2.66e-05 ***
## MaritalStatusMarried        0.27415    0.31461   0.871 0.383545
## MaritalStatusSingle         0.86616    0.39500   2.193 0.028323 *
## NumCompaniesWorked          0.14774    0.04445   3.324 0.000887 ***
## OverTimeYes                1.86137    0.23053   8.074 6.79e-16 ***
## PerformanceRating          -0.58703    0.33093  -1.774 0.076083 .
## RelationshipSatisfaction     -0.23271    0.09473  -2.456 0.014031 *
## StockOptionLevel           -0.32569    0.18334  -1.776 0.075664 .
## TotalWorkingYears          -0.07367    0.02384  -3.090 0.001999 **
## TrainingTimesLastYear       -0.13615    0.08428  -1.615 0.106218
## WorkLifeBalance            -0.45746    0.14461  -3.163 0.001559 **
## YearsInCurrentRole         -0.12571    0.04450  -2.825 0.004730 **
## YearsSinceLastPromotion      0.14330    0.04499   3.185 0.001446 **
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 909.34  on 1028  degrees of freedom
## Residual deviance: 620.88  on 1001  degrees of freedom
## AIC: 676.88
##
## Number of Fisher Scoring iterations: 6
```

VIF:

We can use variance inflation factor (vif) to get rid of redundant predictors or the variables that have high multicollinearity between them. Multicollinearity exists when two or more predictor variables are highly related to each other and then it becomes difficult to understand the impact of an independent variable on the dependent variable. The Variance Inflation Factor(VIF) is used to measure the multicollinearity between predictor variables in a model. A predictor having a VIF of 5 or less is generally considered safe and it can be assumed that it is not correlated with other predictor variables. Higher the VIF, greater is the correlation of the predictor variable w.r.t other predictor variables. However, Predictors with high VIF may have high p-value(or highly significant), hence, we need to see the significance of the Predictor variable before removing it from our model.

```
library(car)

##
## Attaching package: 'car'

## The following object is masked from 'package:dplyr':
##
##      recode

vif(model_2)
```

	GVIF	Df	GVIF^(1/(2*Df))
## BusinessTravel	1.159661	2	1.037726
## DistanceFromHome	1.109299	1	1.053232
## EnvironmentSatisfaction	1.058276	1	1.028726
## Gender	1.038340	1	1.018990
## JobInvolvement	1.028626	1	1.014212
## JobRole	2.056495	8	1.046093
## JobSatisfaction	1.066460	1	1.032696
## MaritalStatus	2.032717	2	1.194041
## NumCompaniesWorked	1.260061	1	1.122524
## OverTime	1.240102	1	1.113598
## PerformanceRating	1.037761	1	1.018706
## RelationshipSatisfaction	1.072754	1	1.035738
## StockOptionLevel	1.928597	1	1.388739
## TotalWorkingYears	2.257093	1	1.502363
## TrainingTimesLastYear	1.062476	1	1.030765
## WorkLifeBalance	1.074822	1	1.036736

```
## YearsInCurrentRole      1.789946  1      1.337889
## YearsSinceLastPromotion  1.816638  1      1.347827

final_model <- model_2
```

Accuracy

```
prob_pred=predict(final_model,type='response', newdata = validation[-2])
y_pred = ifelse(prob_pred>0.5,"Yes","No")
```

```
accuracy <- table(y_pred, validation[, "Attrition"])
accuracy
```

```
##
## y_pred False True
##    No      364   42
##    Yes       6   29
```

```
sum(diag(accuracy))/sum(accuracy)
```

```
## [1] 0.8911565
```

```
res=predict(final_model,train, type="response")
```

```
library(ROCR)
```

```
## Warning: package 'ROCR' was built under R version 3.4.3
```

```
## Loading required package: gplots
```

```
##
```

```
## Attaching package: 'gplots'
```

```
## The following object is masked from 'package:stats':
```

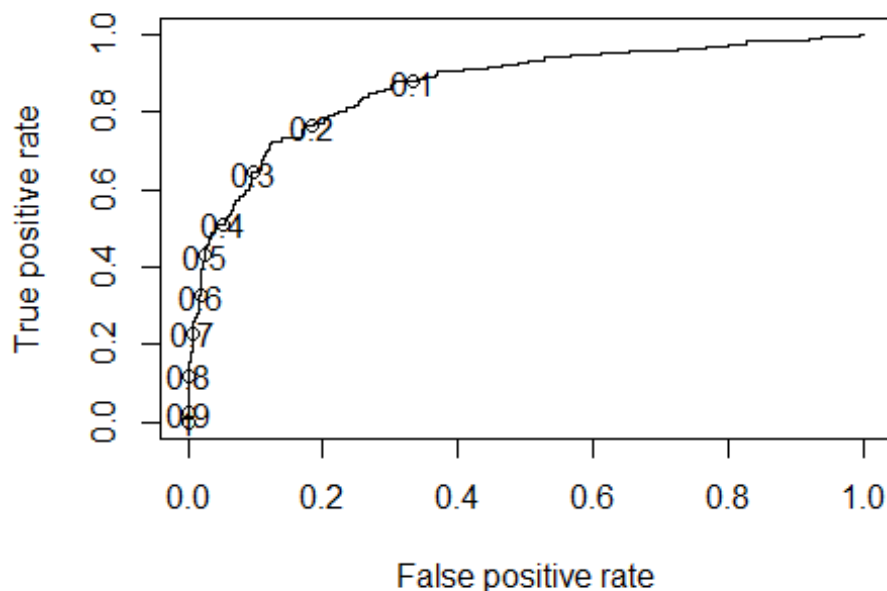
```
##
```

```
##      lowess
```

```
ROCRPred = prediction(res,train$Attrition)
```

```
ROCRPref<- performance(ROCRPred,"tpr","fpr")
```

```
plot(ROCRPref, colorsize=TRUE,print.cutoffs.at=seq(0.1, by=0.1))
```



```
prob_pred1=predict(final_model,type='response', newdata = validation[-2])
y_pred1 = ifelse(prob_pred1>0.2,"Yes","No")

accuracy1 <- table(y_pred1, validation[, "Attrition"])
accuracy1

##
## y_pred1 False True
##      No    306   18
##      Yes    64   53

sum(diag(accuracy1))/sum(accuracy1)

## [1] 0.814059
```

MODEL BUILDING 2 Decision Tree- Splits the data into multiple sets and each set is further split into subsets to arrive at a tree like structure and make a decision. Homogeneity is the basic concept that helps to determine the attribute on which a split should be made. A split that results into the most homogenous subset is often considered better and step by step each attribute is chosen that maximizes the homogeneity of each subset. Further, this homogeneity is measured using different ways such as Gini Index, Entropy and Information Gain. Hide

```
set.seed(123)
df1$Attrition <- as.factor(df1$Attrition)
indices = sample.split(df1$Attrition, SplitRatio = 0.7)
train = df1[indices,]
```



```
validation = df1[!(indices),]
head(validation)
```

```
##      Age Attrition      BusinessTravel DailyRate      Department
## 2    49      False Travel_Frequently      279 Research & Development
## 3    37       True  Travel_Rarely      1373 Research & Development
## 5    27      False  Travel_Rarely      591 Research & Development
## 11   35      False  Travel_Rarely      809 Research & Development
## 13   31      False  Travel_Rarely      670 Research & Development
## 14   34      False  Travel_Rarely      1346 Research & Development
##      DistanceFromHome Education EducationField EmployeeCount EmployeeNumber
## 2              8          1 Life Sciences              1              2
## 3              2          2      Other              1              4
## 5              2          1      Medical              1              7
## 11             16          3      Medical              1             14
## 13             26          1 Life Sciences              1             16
## 14             19          2      Medical              1             18
##      EnvironmentSatisfaction Gender HourlyRate JobInvolvement JobLevel
## 2              3      Male      61              2              2
## 3              4      Male      92              2              1
## 5              1      Male      40              3              1
## 11             1      Male      84              4              1
## 13             1      Male      31              3              1
## 14             2      Male      93              3              1
##      JobRole JobSatisfaction MaritalStatus MonthlyIncome
## 2    Research Scientist      2      Married      5130
## 3    Laboratory Technician      3      Single      2090
## 5    Laboratory Technician      2      Married      3468
## 11   Laboratory Technician      2      Married      2426
## 13    Research Scientist      3      Divorced      2911
## 14   Laboratory Technician      4      Divorced      2661
##      MonthlyRate NumCompaniesWorked Over18 OverTime PercentSalaryHike
## 2      24907              1      Y      No              23
## 3      2396              6      Y      Yes              15
## 5     16632              9      Y      No              12
## 11     16479              0      Y      No              13
## 13     15170              1      Y      No              17
## 14      8758              0      Y      No              11
##      PerformanceRating RelationshipSatisfaction StandardHours
## 2              4              4              80
## 3              3              2              80
## 5              3              4              80
## 11             3              3              80
## 13             3              4              80
## 14             3              3              80
##      StockOptionLevel TotalWorkingYears TrainingTimesLastYear
## 2              1              10              3
## 3              0              7              3
## 5              1              6              3
## 11             1              6              5
```

```

## 13          1          5          1
## 14          1          3          2
##      WorkLifeBalance YearsAtCompany YearsInCurrentRole
## 2          3          10          7
## 3          3          0          0
## 5          3          2          2
## 11         3          5          4
## 13         2          5          2
## 14         3          2          2
##      YearsSinceLastPromotion YearsWithCurrManager
## 2          1          7
## 3          0          0
## 5          2          2
## 11         0          3
## 13         4          3
## 14         1          2

options(repr.plot.width = 10, repr.plot.height = 8)
library(rpart)
library(rpart.plot)

## Warning: package 'rpart.plot' was built under R version 3.4.4

#Training
Dtree = rpart(Attrition ~., data = train, method = "class")
summary(Dtree)

## Call:
## rpart(formula = Attrition ~ ., data = train, method = "class")
##      n= 1029
##
##      CP nsplit rel error   xerror   xstd
## 1 0.03614458      0 1.0000000 1.000000 0.07107939
## 2 0.03012048      3 0.8915663 1.012048 0.07142339
## 3 0.02409639      4 0.8614458 1.006024 0.07125184
## 4 0.01807229      7 0.7891566 1.066265 0.07292739
## 5 0.01000000     13 0.6807229 1.090361 0.07357350
##
## Variable importance
##      JobRole      MonthlyIncome      TotalWorkingYears
##      12      12      10
##      OverTime      JobLevel      MaritalStatus
##      9      8      6
##      Age      DailyRate      Department
##      6      6      5
##      StockOptionLevel      TrainingTimesLastYear      BusinessTravel
##      4      4      3
##      DistanceFromHome      EmployeeNumber      EnvironmentSatisfaction
##      2      2      2
##      YearsAtCompany      PercentSalaryHike      JobInvolvement
##      2      2      1

```

```

##           HourlyRate           MonthlyRate           JobSatisfaction
##                1                1                1
##           EducationField           PerformanceRating           NumCompaniesWorked
##                1                1                1
##
## Node number 1: 1029 observations,      complexity param=0.03614458
##   predicted class=False expected loss=0.1613217 P(node) =1
##   class counts:   863   166
##   probabilities: 0.839 0.161
##   left son=2 (747 obs) right son=3 (282 obs)
##   Primary splits:
##       OverTime           splits as LR,           improve=14.48644, (0
missing)
##       JobRole           splits as LLRLLLLLR,      improve=13.92655, (0
missing)
##       MonthlyIncome     < 2802   to the right, improve=13.85877, (0
missing)
##       TotalWorkingYears < 1.5    to the right, improve=13.53070, (0
missing)
##       JobLevel          < 1.5    to the right, improve=12.38312, (0
missing)
##   Surrogate splits:
##       EmployeeNumber < 22.5     to the right, agree=0.729, adj=0.011, (0
split)
##       DailyRate        < 104.5   to the right, agree=0.727, adj=0.004, (0
split)
##       MonthlyRate      < 26923.5 to the left,  agree=0.727, adj=0.004, (0
split)
##       YearsAtCompany   < 26.5    to the left,  agree=0.727, adj=0.004, (0
split)
##
## Node number 2: 747 observations,      complexity param=0.01807229
##   predicted class=False expected loss=0.1097724 P(node) =0.7259475
##   class counts:   665   82
##   probabilities: 0.890 0.110
##   left son=4 (675 obs) right son=5 (72 obs)
##   Primary splits:
##       TotalWorkingYears < 2.5     to the right, improve=7.964730, (0
missing)
##       YearsAtCompany    < 2.5     to the right, improve=7.802820, (0
missing)
##       YearsWithCurrManager < 0.5   to the right, improve=7.364425, (0
missing)
##       MonthlyIncome     < 2059.5  to the right, improve=6.199431, (0
missing)
##       JobRole           splits as LLRLLLLLR,      improve=5.979504, (0
missing)
##   Surrogate splits:
##       MonthlyIncome < 2009.5  to the right, agree=0.933, adj=0.306, (0
split)

```

```

##      Age          < 21.5    to the right, agree=0.926, adj=0.236, (0
split)
##
## Node number 3: 282 observations,    complexity param=0.03614458
## predicted class=False expected loss=0.2978723 P(node) =0.2740525
##   class counts:   198    84
##   probabilities: 0.702 0.298
##   left son=6 (189 obs) right son=7 (93 obs)
##   Primary splits:
##     MonthlyIncome    < 3751.5  to the right, improve=15.953690, (0
missing)
##     JobLevel         < 1.5     to the right, improve=14.774430, (0
missing)
##     JobRole          splits as LRRLLLLLR,   improve=12.439970, (0
missing)
##     TotalWorkingYears < 1.5     to the right, improve= 7.851383, (0
missing)
##     Age              < 26.5    to the right, improve= 6.737057, (0
missing)
##   Surrogate splits:
##     JobLevel         < 1.5     to the right, agree=0.933, adj=0.796, (0
split)
##     JobRole          splits as LRRLLLRRLR,  agree=0.865, adj=0.591, (0
split)
##     TotalWorkingYears < 3.5     to the right, agree=0.780, adj=0.333, (0
split)
##     YearsAtCompany   < 2.5     to the right, agree=0.723, adj=0.161, (0
split)
##     Age              < 23.5    to the right, agree=0.716, adj=0.140, (0
split)
##
## Node number 4: 675 observations
## predicted class=False expected loss=0.08592593 P(node) =0.6559767
##   class counts:   617    58
##   probabilities: 0.914 0.086
##
## Node number 5: 72 observations,    complexity param=0.01807229
## predicted class=False expected loss=0.3333333 P(node) =0.06997085
##   class counts:    48    24
##   probabilities: 0.667 0.333
##   left son=10 (62 obs) right son=11 (10 obs)
##   Primary splits:
##     BusinessTravel splits as LRL,          improve=5.058065, (0 missing)
##     JobRole        splits as -RR---L-R,    improve=4.500000, (0 missing)
##     HourlyRate     < 58.5    to the right, improve=4.266667, (0 missing)
##     DailyRate      < 258.5   to the right, improve=4.254945, (0 missing)
##     MaritalStatus  splits as LLR,          improve=2.427245, (0 missing)
##   Surrogate splits:
##     EmployeeNumber < 1901.5  to the left,  agree=0.889, adj=0.2, (0
split)

```

```

##      MonthlyRate      < 26306.5 to the left,  agree=0.875, adj=0.1, (0
split)
##
## Node number 6: 189 observations,      complexity param=0.02409639
## predicted class=False expected loss=0.1798942 P(node) =0.1836735
## class counts: 155 34
## probabilities: 0.820 0.180
## left son=12 (114 obs) right son=13 (75 obs)
## Primary splits:
##      JobRole          splits as LLRLLLLRL,  improve=4.881582, (0
missing)
##      MaritalStatus    splits as LLR,        improve=4.654420, (0
missing)
##      StockOptionLevel < 0.5      to the right, improve=4.562328, (0
missing)
##      Department       splits as LLR,        improve=3.841677, (0
missing)
##      EducationField   splits as LLRLLL,     improve=3.565686, (0
missing)
## Surrogate splits:
##      Department       splits as LLR,        agree=0.884, adj=0.707, (0
split)
##      EducationField   splits as LLRLLL,     agree=0.698, adj=0.240, (0
split)
##      Age              < 28.5    to the right, agree=0.651, adj=0.120, (0
split)
##      MonthlyIncome    < 5841.5  to the right, agree=0.646, adj=0.107, (0
split)
##      TotalWorkingYears < 8.5    to the right, agree=0.640, adj=0.093, (0
split)
##
## Node number 7: 93 observations,      complexity param=0.03614458
## predicted class=True  expected loss=0.4623656 P(node) =0.09037901
## class counts: 43 50
## probabilities: 0.462 0.538
## left son=14 (47 obs) right son=15 (46 obs)
## Primary splits:
##      JobRole          splits as -LR---L-R,  improve=4.545532, (0
missing)
##      EnvironmentSatisfaction < 1.5    to the right, improve=4.398897, (0
missing)
##      Age              < 33.5    to the right, improve=3.185902, (0
missing)
##      MonthlyIncome    < 2124    to the right, improve=2.876944, (0
missing)
##      NumCompaniesWorked < 0.5    to the left, improve=2.626303, (0
missing)
## Surrogate splits:
##      Department       splits as LLR,        agree=0.699, adj=0.391, (0
split)

```

```

##      EmployeeNumber    < 674      to the right, agree=0.624, adj=0.239, (0
split)
##      TotalWorkingYears < 1.5      to the right, agree=0.624, adj=0.239, (0
split)
##      Age                < 25      to the right, agree=0.591, adj=0.174, (0
split)
##      DailyRate         < 1285     to the left,  agree=0.591, adj=0.174, (0
split)
##
## Node number 10: 62 observations,    complexity param=0.01807229
##   predicted class=False expected loss=0.2580645 P(node) =0.06025267
##   class counts:    46    16
##   probabilities: 0.742 0.258
##   left son=20 (51 obs) right son=21 (11 obs)
##   Primary splits:
##       DailyRate      < 343.5     to the right, improve=3.827497, (0
missing)
##       HourlyRate     < 58.5      to the right, improve=3.021843, (0
missing)
##       JobRole        splits as  -RR---L-R,    improve=2.391058, (0
missing)
##       StockOptionLevel < 0.5      to the right, improve=2.391058, (0
missing)
##       MaritalStatus  splits as  LLR,          improve=2.341935, (0
missing)
##   Surrogate splits:
##       MonthlyIncome < 1162.5     to the right, agree=0.839, adj=0.091, (0
split)
##
## Node number 11: 10 observations
##   predicted class=True  expected loss=0.2 P(node) =0.009718173
##   class counts:        2      8
##   probabilities: 0.200 0.800
##
## Node number 12: 114 observations
##   predicted class=False expected loss=0.0877193 P(node) =0.1107872
##   class counts:    104    10
##   probabilities: 0.912 0.088
##
## Node number 13: 75 observations,    complexity param=0.02409639
##   predicted class=False expected loss=0.32 P(node) =0.0728863
##   class counts:      51    24
##   probabilities: 0.680 0.320
##   left son=26 (49 obs) right son=27 (26 obs)
##   Primary splits:
##       MaritalStatus  splits as  LLR,          improve=8.870769, (0
missing)
##       StockOptionLevel < 0.5      to the right, improve=6.518571, (0
missing)
##       MonthlyIncome  < 5234.5     to the left,  improve=3.681667, (0

```

```

missing)
##      DistanceFromHome < 23.5      to the left,  improve=3.332308, (0
missing)
##      Gender              splits as  LR,              improve=2.182735, (0
missing)
##      Surrogate splits:
##      StockOptionLevel      < 0.5      to the right, agree=0.880,
adj=0.654, (0 split)
##      HourlyRate            < 82.5      to the left,  agree=0.720,
adj=0.192, (0 split)
##      EmployeeNumber        < 68        to the right, agree=0.693,
adj=0.115, (0 split)
##      NumCompaniesWorked    < 7.5        to the left,  agree=0.693,
adj=0.115, (0 split)
##      TrainingTimesLastYear < 1.5        to the right, agree=0.693,
adj=0.115, (0 split)
##
## Node number 14: 47 observations,      complexity param=0.03012048
## predicted class=False expected loss=0.3829787 P(node) =0.04567541
## class counts:      29      18
## probabilities: 0.617 0.383
## left son=28 (40 obs) right son=29 (7 obs)
## Primary splits:
##      DistanceFromHome      < 16        to the left,  improve=3.698480, (0
missing)
##      EnvironmentSatisfaction < 1.5      to the right, improve=3.470076, (0
missing)
##      BusinessTravel         splits as  -RL,              improve=2.693285, (0
missing)
##      MonthlyRate            < 5384      to the right, improve=2.597381, (0
missing)
##      MonthlyIncome          < 2469.5    to the right, improve=2.183675, (0
missing)
##      Surrogate splits:
##      DailyRate < 159.5      to the right, agree=0.894, adj=0.286, (0 split)
##
## Node number 15: 46 observations,      complexity param=0.01807229
## predicted class=True  expected loss=0.3043478 P(node) =0.0447036
## class counts:      14      32
## probabilities: 0.304 0.696
## left son=30 (20 obs) right son=31 (26 obs)
## Primary splits:
##      Age                    < 33.5      to the right, improve=4.270569, (0
missing)
##      TotalWorkingYears      < 11        to the right, improve=3.846682, (0
missing)
##      EducationField         splits as  -RRLRR,           improve=2.871118, (0
missing)
##      YearsSinceLastPromotion < 2.5      to the right, improve=2.774964, (0
missing)

```

```

##      DailyRate          < 1104    to the right, improve=2.263975, (0
missing)
##  Surrogate splits:
##      TotalWorkingYears < 9        to the right, agree=0.783, adj=0.50, (0
split)
##      DailyRate          < 1315    to the right, agree=0.674, adj=0.25, (0
split)
##      PercentSalaryHike < 19.5     to the right, agree=0.674, adj=0.25, (0
split)
##      PerformanceRating < 3.5      to the right, agree=0.674, adj=0.25, (0
split)
##      StockOptionLevel  < 0.5      to the right, agree=0.674, adj=0.25, (0
split)
##
## Node number 20: 51 observations
##  predicted class=False  expected loss=0.1764706  P(node) =0.04956268
##  class counts:    42    9
##  probabilities: 0.824 0.176
##
## Node number 21: 11 observations
##  predicted class=True   expected loss=0.3636364  P(node) =0.01068999
##  class counts:     4    7
##  probabilities: 0.364 0.636
##
## Node number 26: 49 observations
##  predicted class=False  expected loss=0.1428571  P(node) =0.04761905
##  class counts:    42    7
##  probabilities: 0.857 0.143
##
## Node number 27: 26 observations,    complexity param=0.02409639
##  predicted class=True   expected loss=0.3461538  P(node) =0.02526725
##  class counts:     9   17
##  probabilities: 0.346 0.654
##  left son=54 (8 obs) right son=55 (18 obs)
##  Primary splits:
##      TrainingTimesLastYear < 2.5    to the right, improve=3.769231, (0
missing)
##      MonthlyIncome          < 5791   to the left,  improve=3.211655, (0
missing)
##      JobLevel                < 2.5    to the left,  improve=2.769231, (0
missing)
##      YearsSinceLastPromotion < 2.5    to the left,  improve=2.295547, (0
missing)
##      DistanceFromHome       < 6.5    to the left,  improve=1.054945, (0
missing)
##  Surrogate splits:
##      Department             splits as  -LR,          agree=0.808, adj=0.375, (0
split)
##      JobRole                splits as  --L----R-,    agree=0.808, adj=0.375, (0
split)

```



```

##      JobInvolvement < 1.5      to the left,  agree=0.769, adj=0.250, (0
split)
##      JobSatisfaction < 1.5      to the left,  agree=0.769, adj=0.250, (0
split)
##      MonthlyIncome < 5052      to the left,  agree=0.769, adj=0.250, (0
split)
##
## Node number 28: 40 observations,      complexity param=0.01807229
## predicted class=False expected loss=0.3 P(node) =0.03887269
## class counts:      28      12
## probabilities: 0.700 0.300
## left son=56 (33 obs) right son=57 (7 obs)
## Primary splits:
##      EnvironmentSatisfaction < 1.5      to the right, improve=2.912554, (0
missing)
##      MonthlyRate < 5384      to the right, improve=2.912554, (0
missing)
##      DailyRate < 527.5      to the right, improve=2.190313, (0
missing)
##      BusinessTravel splits as -RL,      improve=1.828213, (0
missing)
##      Gender splits as LR,      improve=1.609524, (0
missing)
## Surrogate splits:
##      MonthlyRate < 3251.5      to the right, agree=0.875, adj=0.286,
(0 split)
##      DistanceFromHome < 12      to the left,  agree=0.850, adj=0.143,
(0 split)
##      JobSatisfaction < 1.5      to the right, agree=0.850, adj=0.143,
(0 split)
##      StockOptionLevel < 1.5      to the left,  agree=0.850, adj=0.143,
(0 split)
##      YearsWithCurrManager < 4      to the left,  agree=0.850, adj=0.143,
(0 split)
##
## Node number 29: 7 observations
## predicted class=True expected loss=0.1428571 P(node) =0.006802721
## class counts:      1      6
## probabilities: 0.143 0.857
##
## Node number 30: 20 observations,      complexity param=0.01807229
## predicted class=False expected loss=0.45 P(node) =0.01943635
## class counts:      11      9
## probabilities: 0.550 0.450
## left son=60 (10 obs) right son=61 (10 obs)
## Primary splits:
##      DailyRate < 1121      to the right, improve=2.500000, (0
missing)
##      YearsSinceLastPromotion < 0.5      to the right, improve=2.500000, (0
missing)

```

```

##      PercentSalaryHike      < 14.5      to the right, improve=2.400000, (0
missing)
##      EducationField        splits as  -RRLRL,      improve=2.031868, (0
missing)
##      MaritalStatus         splits as   RLR,        improve=1.697980, (0
missing)
##      Surrogate splits:
##      MaritalStatus         splits as   RLL,        agree=0.85, adj=0.7,
(0 split)
##      PercentSalaryHike      < 17          to the right, agree=0.80, adj=0.6,
(0 split)
##      TrainingTimesLastYear < 2.5          to the right, agree=0.80, adj=0.6,
(0 split)
##      Department            splits as  -RL,        agree=0.70, adj=0.4,
(0 split)
##      JobInvolvement         < 2.5          to the right, agree=0.70, adj=0.4,
(0 split)
##
## Node number 31: 26 observations
##   predicted class=True   expected loss=0.1153846   P(node) =0.02526725
##   class counts:      3    23
##   probabilities: 0.115 0.885
##
## Node number 54: 8 observations
##   predicted class=False  expected loss=0.25   P(node) =0.007774538
##   class counts:      6    2
##   probabilities: 0.750 0.250
##
## Node number 55: 18 observations
##   predicted class=True   expected loss=0.1666667   P(node) =0.01749271
##   class counts:      3    15
##   probabilities: 0.167 0.833
##
## Node number 56: 33 observations
##   predicted class=False  expected loss=0.2121212   P(node) =0.03206997
##   class counts:     26    7
##   probabilities: 0.788 0.212
##
## Node number 57: 7 observations
##   predicted class=True   expected loss=0.2857143   P(node) =0.006802721
##   class counts:      2    5
##   probabilities: 0.286 0.714
##
## Node number 60: 10 observations
##   predicted class=False  expected loss=0.2   P(node) =0.009718173
##   class counts:      8    2
##   probabilities: 0.800 0.200
##
## Node number 61: 10 observations
##   predicted class=True   expected loss=0.3   P(node) =0.009718173

```

```
##      class counts:      3      7
##      probabilities: 0.300 0.700

#Predicting
DTPred <- predict(Dtree,type = "class", newdata = validation[,-2])

library(caret)

## Warning: package 'caret' was built under R version 3.4.4

confusionMatrix(validation$Attrition, DTPred)

## Confusion Matrix and Statistics
##
##              Reference
## Prediction False True
##      False   352   18
##      True     48   23
##
##              Accuracy : 0.8503
##              95% CI : (0.8136, 0.8823)
##      No Information Rate : 0.907
##      P-Value [Acc > NIR] : 0.9999501
##
##              Kappa : 0.332
##  Mcnemar's Test P-Value : 0.0003575
##
##              Sensitivity : 0.8800
##              Specificity : 0.5610
##              Pos Pred Value : 0.9514
##              Neg Pred Value : 0.3239
##              Prevalence : 0.9070
##              Detection Rate : 0.7982
##              Detection Prevalence : 0.8390
##              Balanced Accuracy : 0.7205
##
##              'Positive' Class : False
##
```

MODEL BUILDING 3: RANDOM FOREST- Often known as an ensemble of a large number of Decision Trees, that uses bootstrapped aggregation technique to choose random samples from a dataset to train each tree in the forest. The final prediction in a RandomForest is an aggregation of prediction of individual trees. One of the advantages of RandomForest is that, it gives out-of-bag(OOB) error estimates, which is the mean prediction error on a training sample, using the trees that do not have that training sample in their bootstrap sample. It may act as a cross validation error and eliminate the need of using test/validation data, thereby increasing the training the data. However, I am still going to use train and validation concept here as well, like I did in the above two Models. Hide

```
library(randomForest)
set.seed(123)
```

```

df1$Attrition <- as.factor(df1$Attrition)
indices = sample.split(df1$Attrition, SplitRatio = 0.7)
train = df1[indices,]
validation = df1[!(indices),]

#Training the RandomForest Model
model.rf <- randomForest(Attrition ~ ., data=train,
proximity=FALSE,importance = FALSE,
                        ntree=500,mtry=4, do.trace=FALSE)
model.rf

##
## Call:
## randomForest(formula = Attrition ~ ., data = train, proximity = FALSE,
## importance = FALSE, ntree = 500, mtry = 4, do.trace = FALSE)
##              Type of random forest: classification
##              Number of trees: 500
## No. of variables tried at each split: 4
##
##              OOB estimate of  error rate: 15.26%
## Confusion matrix:
##      False True class.error
## False   858    5 0.005793743
## True    152   14 0.915662651

#Predicting on the validation set and checking the Confusion Matrix.
testPred <- predict(model.rf, newdata=validation[,-2])
table(testPred, validation$Attrition)

##
## testPred False True
##      False   368    2
##      True     62    9

confusionMatrix(validation$Attrition, testPred)

## Confusion Matrix and Statistics
##
##              Reference
## Prediction False True
##      False   368    2
##      True     62    9
##
##              Accuracy : 0.8549
##              95% CI : (0.8185, 0.8864)
##      No Information Rate : 0.9751
##      P-Value [Acc > NIR] : 1
##
##              Kappa : 0.1843
##      McNemar's Test P-Value : 1.643e-13
##

```

```
##          Sensitivity : 0.8558
##          Specificity : 0.8182
##          Pos Pred Value : 0.9946
##          Neg Pred Value : 0.1268
##          Prevalence : 0.9751
##          Detection Rate : 0.8345
##          Detection Prevalence : 0.8390
##          Balanced Accuracy : 0.8370
##
##          'Positive' Class : False
##
```

Variable Importance Plot: Below is the variable importance plot, that shows the most significant attribute in decreasing order by mean decrease in Gini. The Mean decrease Gini measures how pure the nodes are at the end of the tree. Higher the Gini Index, better is the homogeneity. Hide

```
#Checking the variable Importance Plot
varImpPlot(model.rf)
```



```
library(pROC)

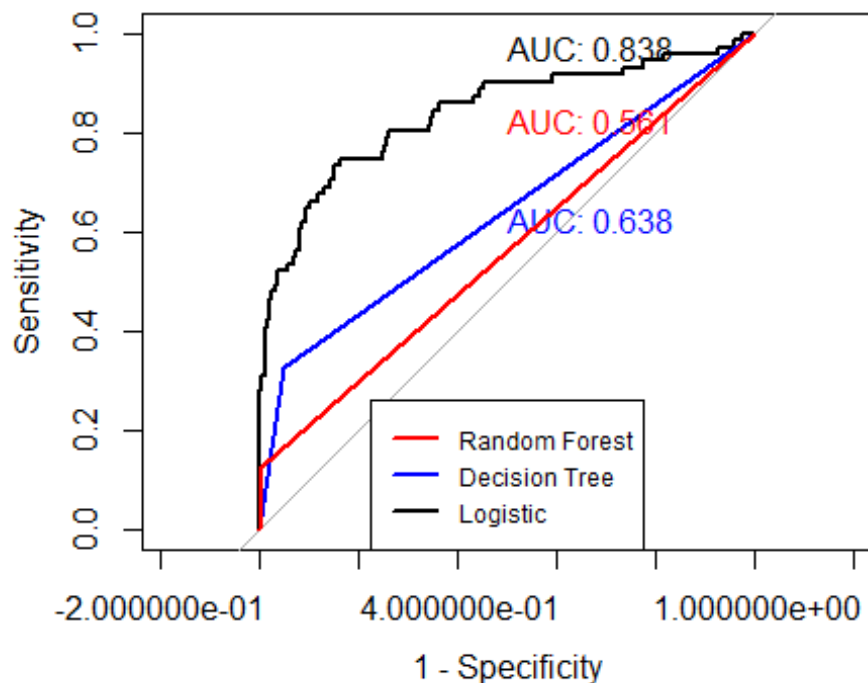
## Warning: package 'pROC' was built under R version 3.4.4
## Type 'citation("pROC")' for a citation.
##
## Attaching package: 'pROC'
```

```
## The following objects are masked from 'package:stats':
##
##      cov, smooth, var

options(repr.plot.width = 10, repr.plot.height = 8)
glm.roc <- roc(response = validation$Attrition, predictor =
as.numeric(prob_pred1))
DT.roc <- roc(response = validation$Attrition, predictor, predictor =
as.numeric(DTPred))

rf.roc <- roc(response = validation$Attrition, predictor =
as.numeric(testPred))
plot(glm.roc,      legacy.axes = TRUE, print.auc.y = 1.0, print.auc = TRUE)
plot(DT.roc, col = "blue", add = TRUE, print.auc.y = 0.65, print.auc = TRUE)

plot(rf.roc, col = "red", add = TRUE, print.auc.y = 0.85, print.auc = TRUE)
legend("bottom", c("Random Forest", "Decision Tree", "Logistic"),
      lty = c(1,1), lwd = c(2, 2), col = c("red", "blue", "black"), cex =
0.75)
```



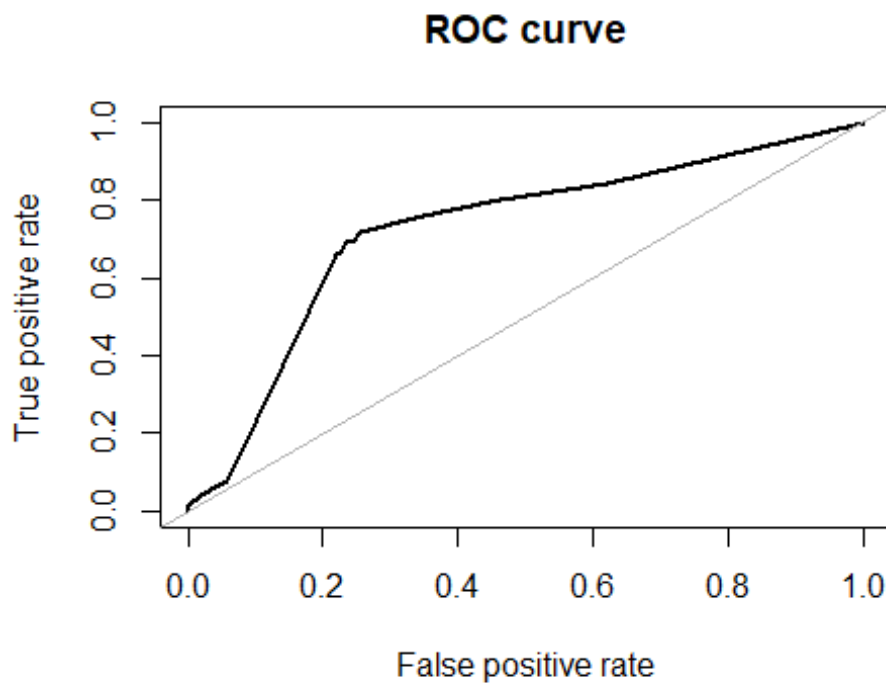
```
library(ROSE)

## Warning: package 'ROSE' was built under R version 3.4.4
## Loaded ROSE 0.0-3
```

```
data.rose<-ROSE(Attrition~., data=train,seed=1)$data
table(data.rose$Attrition)

##
## False  True
##   533   496

library(rpart)
tree.rose <- rpart(Attrition ~ ., data = data.rose)
pred.tree.rose <- predict(tree.rose, newdata = validation)
roc.curve(validation$Attrition, pred.tree.rose[,2])
```



```
## Area under the curve (AUC): 0.725
```

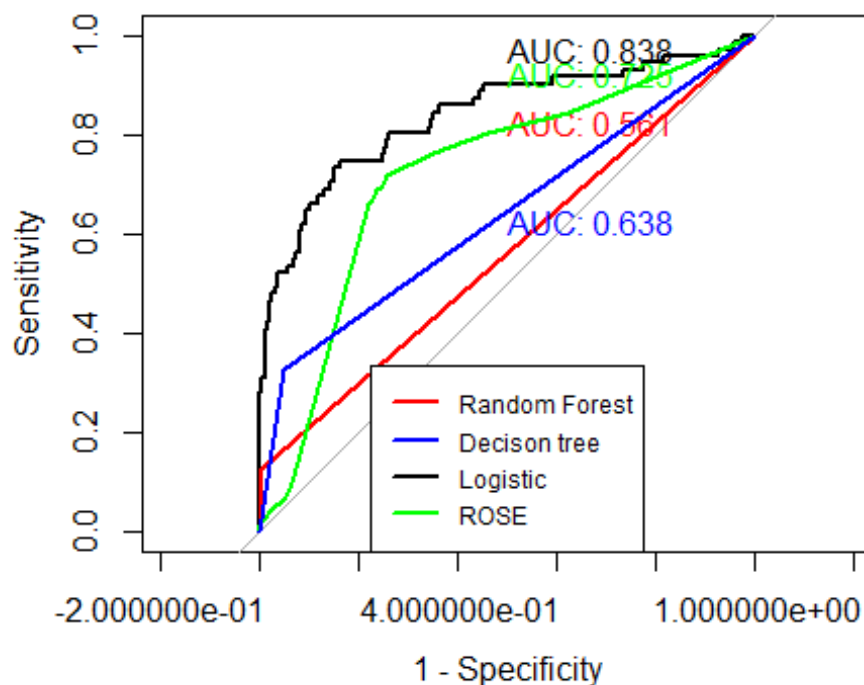
```
library(pROC)
options(repr.plot.width =10, repr.plot.height = 8)
glm.roc <- roc(response = validation$Attrition, predictor =
as.numeric(prob_pred1))
rf.roc <- roc(response = validation$Attrition, predictor =
as.numeric(testPred))
rose.roc<-roc(response = validation$Attrition, predictor =
as.numeric(pred.tree.rose[,2]))
plot(glm.roc,      legacy.axes = TRUE, print.auc.y = 1.0, print.auc = TRUE)
DT.roc <- roc(response = validation$Attrition, predictor, predictor =
as.numeric(DTPred))
```

```

plot(rf.roc, col = "red" , add = TRUE, print.auc.y = 0.85, print.auc = TRUE)
plot(rose.roc, col = "green" , add = TRUE, print.auc.y = 0.95, print.auc = TRUE)
plot(DT.roc, col = "blue", add = TRUE, print.auc.y = 0.65, print.auc = TRUE)

legend("bottom", c("Random Forest","Decison tree" ,"Logistic","ROSE"),
      lty = c(1,1), lwd = c(2, 2), col = c("red", "blue", "black","green"),
      cex = 0.75)

```



So we can see here ROSE(oversampling) on decision tree increases the performance when compared to normal decision tree but still Logistic regression wins the race by best AUC value.

SURVIVAL PROBABILITY:

```

library(survival)

##
## Attaching package: 'survival'

## The following object is masked from 'package:caret':
##
##   cluster

```



```
dataNW$YearsAtCompany=as.numeric(dataNW$YearsAtCompany)
dataNW$Attrition=as.numeric(dataNW$Attrition)
dataNW$Age=as.numeric(dataNW$Age)
```

Assigning the time and event

```
time = dataNW$YearsAtCompany
event= dataNW$Attrition

mySurv<-Surv(time,event)
class(mySurv)

## [1] "Surv"

head(mySurv,20) # plus sign means censored data- there is no informaton

## [1] 6 10+ 0 8+ 2+ 7+ 1+ 1+ 9+ 7+ 5+ 9+ 5+ 2+ 4 10+ 6+
## [18] 1+ 25+ 3+

myfit<-survfit(mySurv~dataNW$OverTime)
myfit

## Call: survfit(formula = mySurv ~ dataNW$OverTime)
##
##              n events median 0.95LCL 0.95UCL
## dataNW$OverTime=No 1054    110    40    32    NA
## dataNW$OverTime=Yes 416    127    24    16    NA
```

Median survival days for employees who does overtime is less(24) than employee who do not.

```
survdif(mySurv~dataNW$OverTime)

## Call:
## survdiff(formula = mySurv ~ dataNW$OverTime)
##
##              N Observed Expected (O-E)^2/E (O-E)^2/V
## dataNW$OverTime=No 1054    110   171.4    22.0    81.7
## dataNW$OverTime=Yes 416    127    65.6    57.5    81.7
##
##  Chisq= 81.7  on 1 degrees of freedom, p= 0

summary(myfit)

## Call: survfit(formula = mySurv ~ dataNW$OverTime)
##
##              dataNW$OverTime=No
##  time n.risk n.event survival std.err lower 95% CI upper 95% CI
##    0   1054     6    0.994 0.00232    0.990    0.999
##    1   1024    31    0.964 0.00578    0.953    0.976
##    2    908    12    0.951 0.00677    0.938    0.965
##    3    816     6    0.944 0.00730    0.930    0.959
##    4    728    11    0.930 0.00836    0.914    0.947
```

```
##      5      648      7      0.920 0.00909      0.902      0.938
##      6      509      4      0.913 0.00971      0.894      0.932
##      7      457      3      0.907 0.01025      0.887      0.927
##      8      388      3      0.900 0.01094      0.879      0.922
##      9      332      6      0.884 0.01260      0.859      0.909
##     10      269      8      0.857 0.01527      0.828      0.888
##     11      181      1      0.853 0.01590      0.822      0.884
##     13      146      2      0.841 0.01770      0.807      0.876
##     14      126      2      0.828 0.01978      0.790      0.867
##     17       89      1      0.818 0.02163      0.777      0.862
##     18       82      1      0.808 0.02356      0.763      0.856
##     20       65      1      0.796 0.02627      0.746      0.849
##     22       35      1      0.773 0.03397      0.709      0.843
##     23       24      1      0.741 0.04532      0.657      0.835
##     32        8      1      0.648 0.09528      0.486      0.865
##     33        6      1      0.540 0.12663      0.341      0.855
##     40        1      1      0.000      NaN      NA      NA
##
##                               dataNW$OverTime=Yes
##  time n.risk n.event survival std.err lower 95% CI upper 95% CI
##    0    416     10    0.976 0.00751    0.961    0.991
##    1    402     28    0.908 0.01423    0.881    0.936
##    2    347     15    0.869 0.01684    0.836    0.902
##    3    312     14    0.830 0.01903    0.793    0.868
##    4    272      8    0.805 0.02034    0.766    0.846
##    5    242     14    0.759 0.02265    0.716    0.804
##    6    185      5    0.738 0.02383    0.693    0.786
##    7    161      8    0.702 0.02593    0.653    0.754
##    8    140      6    0.672 0.02757    0.620    0.728
##    9    116      2    0.660 0.02829    0.607    0.718
##   10     97     10    0.592 0.03254    0.531    0.659
##   11     65      1    0.583 0.03329    0.521    0.652
##   15     50      1    0.571 0.03460    0.507    0.643
##   16     43      1    0.558 0.03626    0.491    0.634
##   19     32      1    0.540 0.03909    0.469    0.623
##   21     22      1    0.516 0.04437    0.436    0.611
##   24     13      1    0.476 0.05595    0.378    0.599
##   31      7      1    0.408 0.07916    0.279    0.597
```

```
library(ggplot2)
require("survival")
library(survival)
library(survminer)
```

```
## Warning: package 'survminer' was built under R version 3.4.4
```

```
## Loading required package: ggpubr
```

```
## Warning: package 'ggpubr' was built under R version 3.4.4
```

```
## Loading required package: magrittr
```

```
##
## Attaching package: 'magrittr'

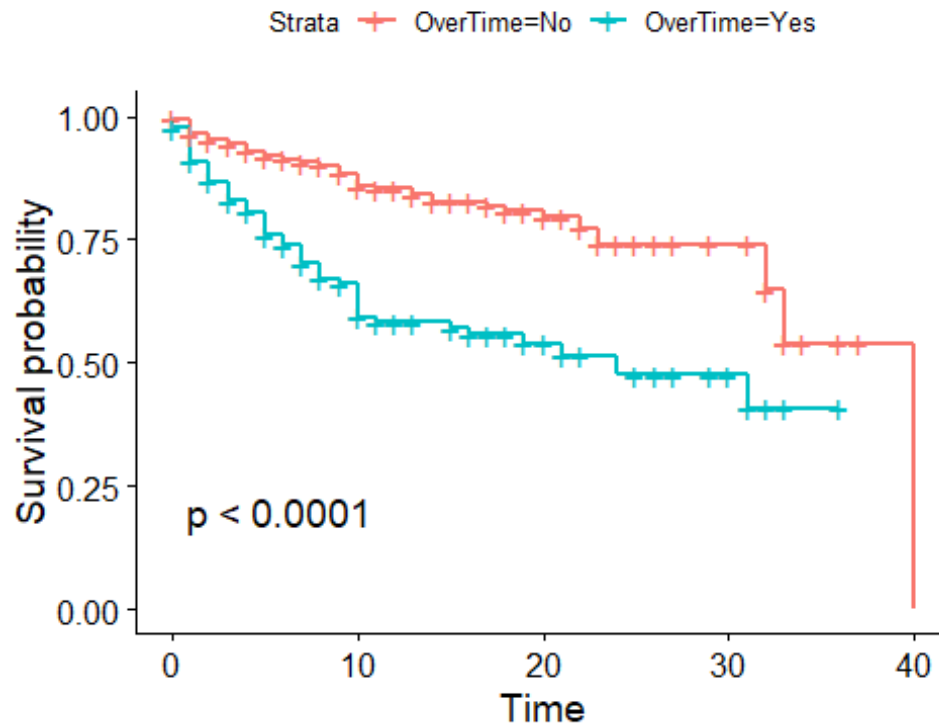
## The following object is masked from 'package:tidyr':
##
##      extract

fit1 <- survfit(mySurv ~ dataNW$OverTime)
summary(fit1)

## Call: survfit(formula = mySurv ~ dataNW$OverTime)
##
##              dataNW$OverTime=No
##   time n.risk n.event survival std.err lower 95% CI upper 95% CI
##    0    1054      6   0.994 0.00232   0.990   0.999
##    1     1024     31   0.964 0.00578   0.953   0.976
##    2      908     12   0.951 0.00677   0.938   0.965
##    3      816      6   0.944 0.00730   0.930   0.959
##    4      728     11   0.930 0.00836   0.914   0.947
##    5      648      7   0.920 0.00909   0.902   0.938
##    6      509      4   0.913 0.00971   0.894   0.932
##    7      457      3   0.907 0.01025   0.887   0.927
##    8      388      3   0.900 0.01094   0.879   0.922
##    9      332      6   0.884 0.01260   0.859   0.909
##   10      269      8   0.857 0.01527   0.828   0.888
##   11      181      1   0.853 0.01590   0.822   0.884
##   13      146      2   0.841 0.01770   0.807   0.876
##   14      126      2   0.828 0.01978   0.790   0.867
##   17       89      1   0.818 0.02163   0.777   0.862
##   18       82      1   0.808 0.02356   0.763   0.856
##   20       65      1   0.796 0.02627   0.746   0.849
##   22       35      1   0.773 0.03397   0.709   0.843
##   23       24      1   0.741 0.04532   0.657   0.835
##   32        8      1   0.648 0.09528   0.486   0.865
##   33        6      1   0.540 0.12663   0.341   0.855
##   40        1      1   0.000      NaN      NA      NA
##
##              dataNW$OverTime=Yes
##   time n.risk n.event survival std.err lower 95% CI upper 95% CI
##    0     416     10   0.976 0.00751   0.961   0.991
##    1     402     28   0.908 0.01423   0.881   0.936
##    2     347     15   0.869 0.01684   0.836   0.902
##    3     312     14   0.830 0.01903   0.793   0.868
##    4     272      8   0.805 0.02034   0.766   0.846
##    5     242     14   0.759 0.02265   0.716   0.804
##    6     185      5   0.738 0.02383   0.693   0.786
##    7     161      8   0.702 0.02593   0.653   0.754
##    8     140      6   0.672 0.02757   0.620   0.728
##    9     116      2   0.660 0.02829   0.607   0.718
##   10      97     10   0.592 0.03254   0.531   0.659
```

##	11	65	1	0.583	0.03329	0.521	0.652
##	15	50	1	0.571	0.03460	0.507	0.643
##	16	43	1	0.558	0.03626	0.491	0.634
##	19	32	1	0.540	0.03909	0.469	0.623
##	21	22	1	0.516	0.04437	0.436	0.611
##	24	13	1	0.476	0.05595	0.378	0.599
##	31	7	1	0.408	0.07916	0.279	0.597

```
ggsurvplot(fit1, data = dataNW, pval = TRUE)
```



After 20 days the survival propability rate for employee who does overtime is 55% and employee who do overtime is 80%