

Project_Group_03

2025-06-24

#Import dataset into R-studio

```
library(readxl)
```

```
Employee_Data <- read_xlsx("C:\\Users\\Clone\\Documents\\Campus\\3rd  
year\\lab\\project\\employee_attrition_dataset_10000.xlsx")
```

```
head(Employee_Data)
```

```
## # A tibble: 6 × 28
```

```
##   Employee_ID  Age Gender Marital_Status Department Job_Role  Job_Level  
##         <dbl> <dbl> <chr>   <chr>         <chr>    <chr>      <dbl>  
## 1           1    58 Male    Single         Finance   Manager        5  
## 2           2    48 Female Divorced        HR        Assistant      4  
## 3           3    34 Female Married        Marketing Manager        4  
## 4           4    27 Female Divorced        HR        Manager        4  
## 5           5    40 Male    Married        HR        Analyst        1  
## 6           6    58 Male    Married        Finance   Executive      3
```

```
## # i 21 more variables: Monthly_Income <dbl>, Hourly_Rate <dbl>,  
## #   Years_at_Company <dbl>, Years_in_Current_Role <dbl>,  
## #   Years_Since_Last_Promotion <dbl>, Work_Life_Balance <dbl>,  
## #   Job_Satisfaction <dbl>, Perform <dbl>, Performance <chr>,  
## #   Performance_Rating <dbl>, Training_Hours_Last_Year <dbl>, Overtime  
## #   Project_Count <dbl>, Average_Hours_Worked_Per_Week <dbl>,  
## #   Absenteeism <dbl>, Work_Environment_Satisfaction <dbl>, ...
```

#Identify missing values

```
missing_count <- sum(is.na(Employee_Data))
```

```
missing_count
```

```
## [1] 0
```

#Remove the Employee_ID,Performance_Rating,Perform columns from the dataset

```
Employee_Data$Employee_ID <- NULL
```

```
Employee_Data$Performance_Rating <- NULL
```

```
Employee_Data$Perform <- NULL
```

```
head(Employee_Data)
```

```
## # A tibble: 6 × 25
```

```
##   Age Gender Marital_Status Department Job_Role  Job_Level  
##   <dbl> <chr>   <chr>         <chr>    <chr>      <dbl>  
## 1    58 Male    Single         Finance   Manager        5
```

```
7332
```

```

## 2      48 Female Divorced      HR      Assistant      4
6069
## 3      34 Female Married      Marketing Manager      4
11485
## 4      27 Female Divorced      HR      Manager      4
18707
## 5      40 Male      Married      HR      Analyst      1
16398
## 6      58 Male      Married      Finance      Executive      3
7305
## # i 18 more variables: Hourly_Rate <dbl>, Years_at_Company <dbl>,
## #   Years_in_Current_Role <dbl>, Years_Since_Last_Promotion <dbl>,
## #   Work_Life_Balance <dbl>, Job_Satisfaction <dbl>, Performance <chr>,
## #   Training_Hours_Last_Year <dbl>, Overtime <chr>, Project_Count <dbl>,
## #   Average_Hours_Worked_Per_Week <dbl>, Absenteeism <dbl>,
## #   Work_Environment_Satisfaction <dbl>, Relationship_with_Manager <dbl>,
## #   Job_Involvement <dbl>, Distance_From_Home <dbl>, ...

str(Employee_Data)

## tibble [10,000 × 25] (S3: tbl_df/tbl/data.frame)
##  $ Age                : num [1:10000] 58 48 34 27 40 58 38 42 30
30 ...
##  $ Gender              : chr [1:10000] "Male" "Female" "Female"
"Female" ...
##  $ Marital_Status      : chr [1:10000] "Single" "Divorced"
"Married" "Divorced" ...
##  $ Department          : chr [1:10000] "Finance" "HR" "Marketing"
"HR" ...
##  $ Job_Role            : chr [1:10000] "Manager" "Assistant"
"Manager" "Manager" ...
##  $ Job_Level           : num [1:10000] 5 4 4 4 1 3 5 1 4 4 ...
##  $ Monthly_Income      : num [1:10000] 7332 6069 11485 18707
16398 ...
##  $ Hourly_Rate         : num [1:10000] 81 55 65 28 92 63 63 41 95
53 ...
##  $ Years_at_Company    : num [1:10000] 24 18 6 12 3 25 3 16 17 16
...
##  $ Years_in_Current_Role : num [1:10000] 12 7 4 9 9 2 3 8 10 14 ...
##  $ Years_Since_Last_Promotion : num [1:10000] 3 5 3 1 1 3 4 0 2 4 ...
##  $ Work_Life_Balance    : num [1:10000] 1 1 4 1 3 4 4 2 2 1 ...
##  $ Job_Satisfaction     : num [1:10000] 3 2 5 1 4 5 3 4 3 4 ...
##  $ Performance         : chr [1:10000] "Low" "Low" "Low" "Low"
...
##  $ Training_Hours_Last_Year : num [1:10000] 74 24 63 4 62 84 98 75 51
45 ...
##  $ Overtime            : chr [1:10000] "No" "Yes" "Yes" "No" ...
##  $ Project_Count        : num [1:10000] 9 9 3 9 1 1 1 3 8 6 ...
##  $ Average_Hours_Worked_Per_Week: num [1:10000] 48 57 55 53 54 42 58 45 42
41 ...

```

```
## $ Absenteeism : num [1:10000] 16 10 1 2 11 11 16 9 4 12
...
## $ Work_Environment_Satisfaction: num [1:10000] 4 4 1 3 1 2 3 2 3 4 ...
## $ Relationship_with_Manager : num [1:10000] 1 1 4 4 1 3 3 1 3 2 ...
## $ Job_Involvement : num [1:10000] 1 1 3 1 1 4 4 4 1 1 ...
## $ Distance_From_Home : num [1:10000] 49 25 21 46 43 4 33 3 39 1
...
## $ Number_of_Companies_Worked : num [1:10000] 3 1 1 2 4 3 1 2 4 4 ...
## $ Attrition : chr [1:10000] "No" "No" "Yes" "No" ...
```

#Convert charctor variable into factor type

```
Employee_Data$Gender <- as.factor(Employee_Data$Gender)
Employee_Data$Performance <- as.factor(Employee_Data$Performance)
Employee_Data$Marital_Status <- as.factor(Employee_Data$Marital_Status)
Employee_Data$Department <- as.factor(Employee_Data$Department)
Employee_Data$Job_Role <- as.factor(Employee_Data$Job_Role)
Employee_Data$Overtime <- as.factor(Employee_Data$Overtime)
Employee_Data$Attrition<- as.factor(Employee_Data$Attrition)
```

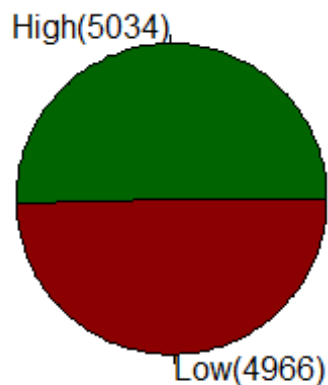
```
str(Employee_Data)
```

```
## tibble [10,000 × 25] (S3: tbl_df/tbl/data.frame)
## $ Age : num [1:10000] 58 48 34 27 40 58 38 42 30
30 ...
## $ Gender : Factor w/ 2 levels "Female","Male": 2 1
1 1 2 2 2 1 1 2 ...
## $ Marital_Status : Factor w/ 3 levels
"Divorced","Married",...: 3 1 2 1 2 2 2 2 1 ...
## $ Department : Factor w/ 5 levels "Finance","HR",...: 1
2 4 2 2 1 5 4 3 5 ...
## $ Job_Role : Factor w/ 4 levels
"Analyst","Assistant",...: 4 2 4 4 1 3 3 3 1 2 ...
## $ Job_Level : num [1:10000] 5 4 4 4 1 3 5 1 4 4 ...
## $ Monthly_Income : num [1:10000] 7332 6069 11485 18707
16398 ...
## $ Hourly_Rate : num [1:10000] 81 55 65 28 92 63 63 41 95
53 ...
## $ Years_at_Company : num [1:10000] 24 18 6 12 3 25 3 16 17 16
...
## $ Years_in_Current_Role : num [1:10000] 12 7 4 9 9 2 3 8 10 14 ...
## $ Years_Since_Last_Promotion : num [1:10000] 3 5 3 1 1 3 4 0 2 4 ...
## $ Work_Life_Balance : num [1:10000] 1 1 4 1 3 4 4 2 2 1 ...
## $ Job_Satisfaction : num [1:10000] 3 2 5 1 4 5 3 4 3 4 ...
## $ Performance : Factor w/ 2 levels "High","Low": 2 2 2 2
1 1 1 1 1 1 ...
## $ Training_Hours_Last_Year : num [1:10000] 74 24 63 4 62 84 98 75 51
45 ...
## $ Overtime : Factor w/ 2 levels "No","Yes": 1 2 2 1 1
1 1 2 2 2 ...
## $ Project_Count : num [1:10000] 9 9 3 9 1 1 1 3 8 6 ...
```

```
## $ Average_Hours_Worked_Per_Week: num [1:10000] 48 57 55 53 54 42 58 45 42
41 ...
## $ Absenteeism : num [1:10000] 16 10 1 2 11 11 16 9 4 12
...
## $ Work_Environment_Satisfaction: num [1:10000] 4 4 1 3 1 2 3 2 3 4 ...
## $ Relationship_with_Manager : num [1:10000] 1 1 4 4 1 3 3 1 3 2 ...
## $ Job_Involvement : num [1:10000] 1 1 3 1 1 4 4 4 1 1 ...
## $ Distance_From_Home : num [1:10000] 49 25 21 46 43 4 33 3 39 1
...
## $ Number_of_Companies_Worked : num [1:10000] 3 1 1 2 4 3 1 2 4 4 ...
## $ Attrition : Factor w/ 2 levels "No","Yes": 1 1 2 1 1
2 2 1 1 1 ...

performance_table<-table(Employee_Data$Performance)
pie(performance_table,main="Pie chart for the
Performance",col=c("darkgreen","darkred"),labels =
paste(names(performance_table), "(" ,performance_table, ")", sep=""))
```

Pie chart for the Performance



```
#summary(Employee_Data)
summary(Employee_Data)
```

```
##      Age      Gender  Marital_Status  Department
## Min.   :20.00  Female:5042  Divorced:3330  Finance   :1990
## 1st Qu.:30.00  Male  :4958  Married :3375  HR        :1953
## Median :40.00                Single  :3295  IT         :1916
## Mean   :39.56                Marketing:2133
## 3rd Qu.:49.00                Sales    :2008
```

```

## Max. :59.00
## Job_Role Job_Level Monthly_Income Hourly_Rate
## Analyst :2572 Min. :1.000 Min. : 3000 Min. :15.00
## Assistant:2538 1st Qu.:2.000 1st Qu.: 7182 1st Qu.:36.00
## Executive:2476 Median :3.000 Median :11402 Median :57.00
## Manager :2414 Mean :2.991 Mean :11437 Mean :57.03
## 3rd Qu.:4.000 3rd Qu.:15680 3rd Qu.:78.00
## Max. :5.000 Max. :19999 Max. :99.00
## Years_at_Company Years_in_Current_Role Years_Since_Last_Promotion
## Min. : 1.00 Min. : 1.000 Min. :0.000
## 1st Qu.: 8.00 1st Qu.: 4.000 1st Qu.:2.000
## Median :15.00 Median : 7.000 Median :4.000
## Mean :14.94 Mean : 7.451 Mean :4.472
## 3rd Qu.:22.00 3rd Qu.:11.000 3rd Qu.:7.000
## Max. :29.00 Max. :14.000 Max. :9.000
## Work_Life_Balance Job_Satisfaction Performance Training_Hours_Last_Year
## Min. :1.000 Min. :1.000 High:5034 Min. : 0.00
## 1st Qu.:2.000 1st Qu.:2.000 Low :4966 1st Qu.:25.00
## Median :2.000 Median :3.000 Median :49.00
## Mean :2.502 Mean :3.038 Mean :49.59
## 3rd Qu.:3.000 3rd Qu.:4.000 3rd Qu.:75.00
## Max. :4.000 Max. :5.000 Max. :99.00
## Overtime Project_Count Average_Hours_Worked_Per_Week Absenteeism
## No :5103 Min. :1.000 Min. :30.00 Min. : 0.00
## Yes:4897 1st Qu.:3.000 1st Qu.:37.00 1st Qu.: 4.00
## Median :5.000 Median :45.00 Median : 9.00
## Mean :4.984 Mean :44.47 Mean : 9.41
## 3rd Qu.:7.000 3rd Qu.:52.00 3rd Qu.:14.00
## Max. :9.000 Max. :59.00 Max. :19.00
## Work_Environment_Satisfaction Relationship_with_Manager Job_Involvement
## Min. :1.000 Min. :1.000 Min. :1.000
## 1st Qu.:1.000 1st Qu.:1.000 1st Qu.:2.000
## Median :2.000 Median :2.000 Median :3.000
## Mean :2.493 Mean :2.491 Mean :2.505
## 3rd Qu.:4.000 3rd Qu.:3.000 3rd Qu.:3.000
## Max. :4.000 Max. :4.000 Max. :4.000
## Distance_From_Home Number_of_Companies_Worked Attrition
## Min. : 1.00 Min. :1.000 No :8003
## 1st Qu.:13.00 1st Qu.:2.000 Yes:1997
## Median :25.00 Median :2.000
## Mean :25.27 Mean :2.517
## 3rd Qu.:37.00 3rd Qu.:4.000
## Max. :49.00 Max. :4.000

```

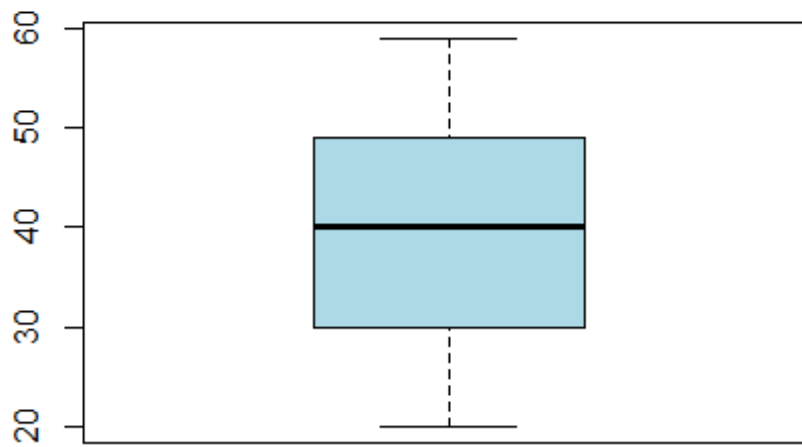
#Draw boxplot for the find outliers

```

boxplot(Employee_Data$Age, main = "Age", col="lightblue", sub =
paste("outliers:", boxplot.stats(Employee_Data$Age)$out))

```

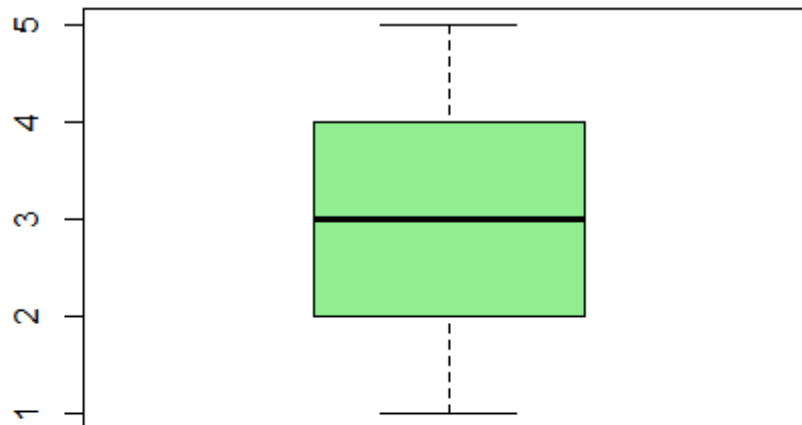
Age



outliers:

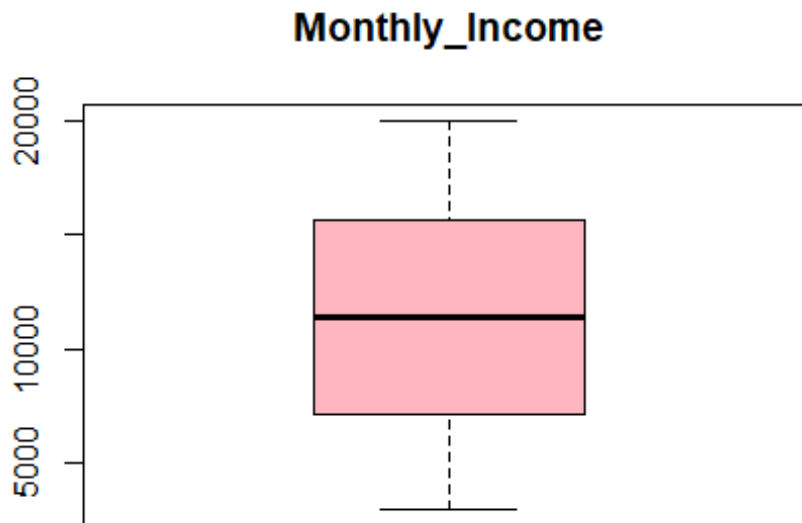
```
boxplot(Employee_Data$Job_Level, main = "Job_Level", col="lightgreen", sub =  
paste("outliers:", boxplot.stats(Employee_Data$Job_Level)$out))
```

Job_Level



outliers:

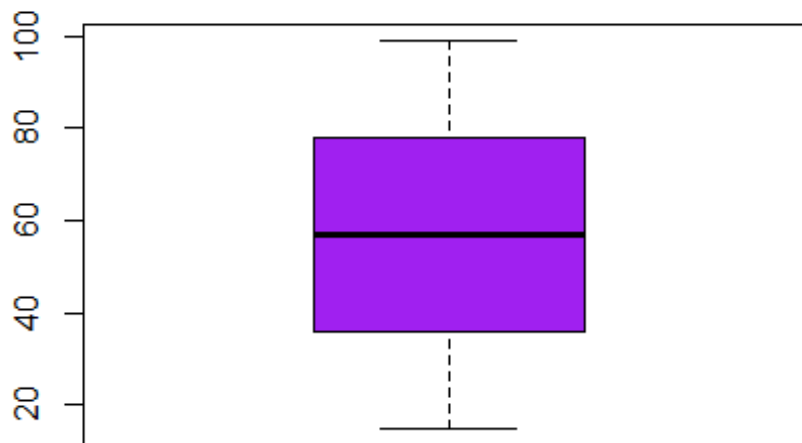
```
boxplot(Employee_Data$Monthly_Income, main =  
"Monthly_Income", col="lightpink", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Monthly_Income)$out))
```



outliers:

```
boxplot(Employee_Data$Hourly_Rate, main = "Hourly_Rate", col="purple", sub =  
paste("outliers:", boxplot.stats(Employee_Data$Hourly_Rate)$out))
```

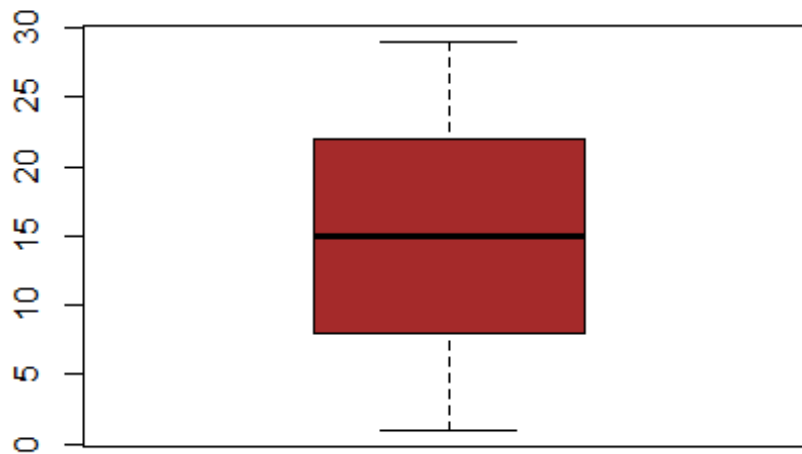
Hourly_Rate



outliers:

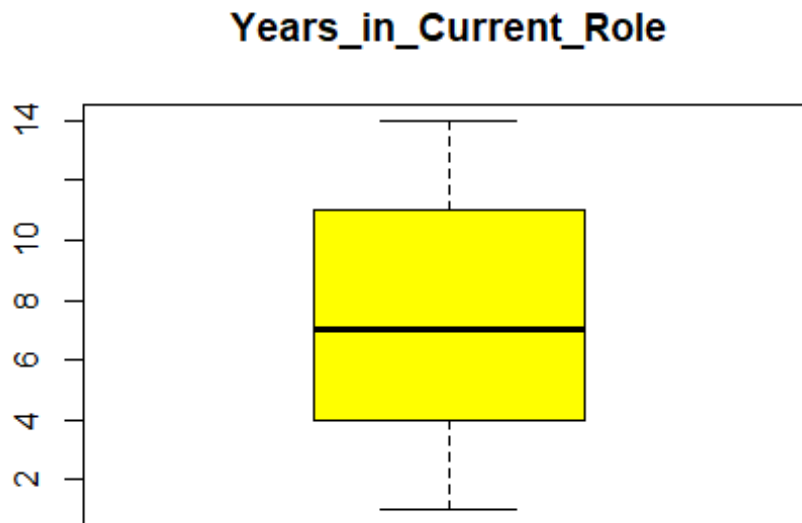
```
boxplot(Employee_Data$Years_at_Company, main = "Years_at_Company",  
col="brown", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Years_at_Company)$out))
```

Years_at_Company



outliers:

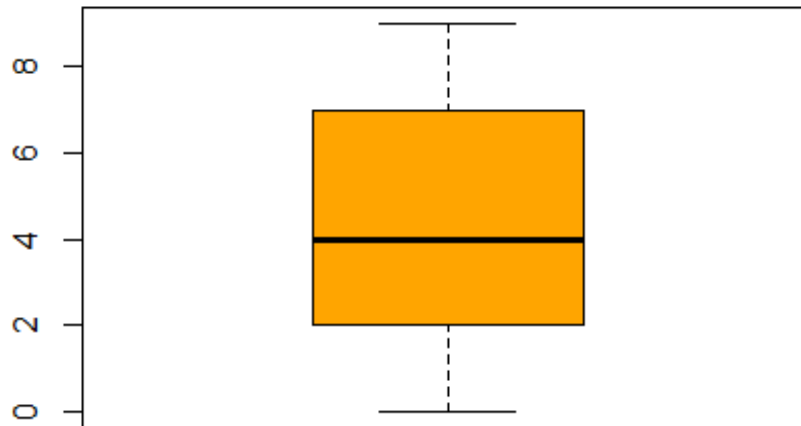

```
boxplot(Employee_Data$Years_in_Current_Role, main =  
"Years_in_Current_Role", col="yellow", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Years_in_Current_Role)$out))
```



outliers:

```
boxplot(Employee_Data$Years_Since_Last_Promotion, main =  
"Years_Since_Last_Promotion", col="orange", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Years_Since_Last_Promotion)$out))
```

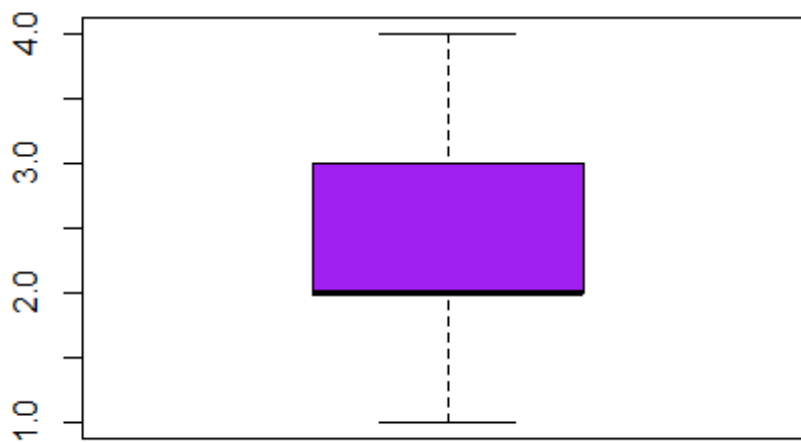
Years_Since_Last_Promotion



outliers:

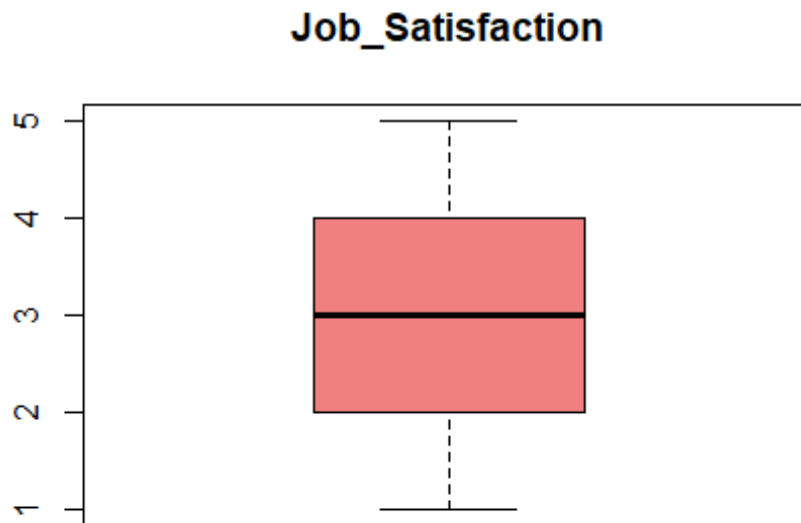
```
boxplot(Employee_Data$Work_Life_Balance, main =  
"Work_Life_Balance", col="purple", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Work_Life_Balance)$out))
```

Work_Life_Balance



outliers:

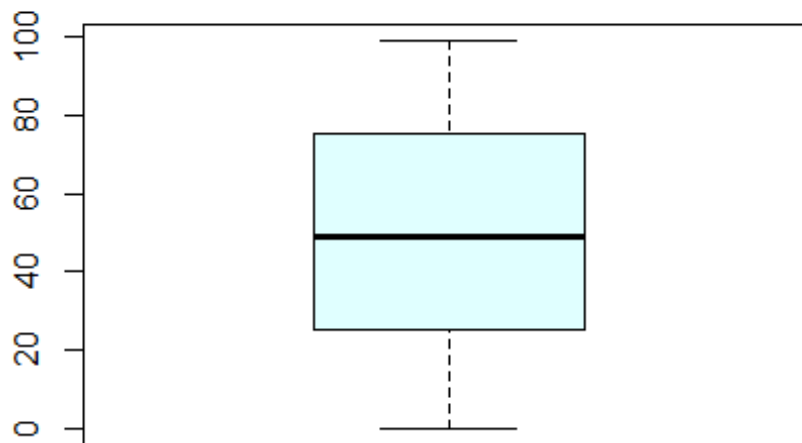
```
boxplot(Employee_Data$Job_Satisfaction, main = "Job_Satisfaction",  
col="lightcoral", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Job_Satisfaction)$out))
```



outliers:

```
boxplot(Employee_Data$Training_Hours_Last_Year, main =  
"Training_Hours_Last_Year", col="lightcyan", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Training_Hours_Last_Year)$out))
```

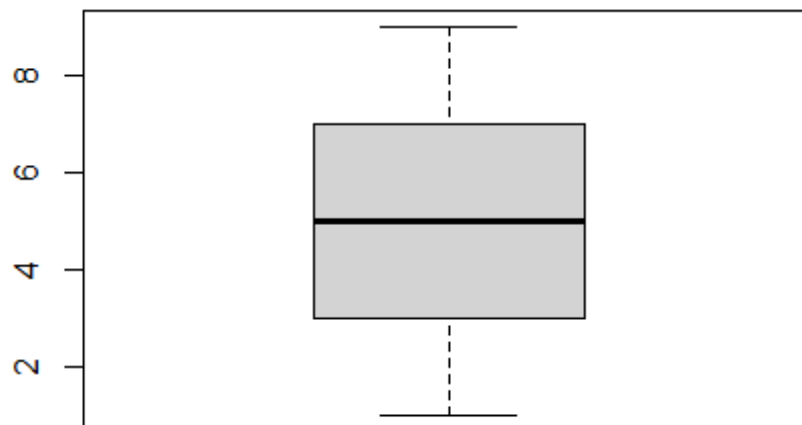
Training_Hours_Last_Year



outliers:

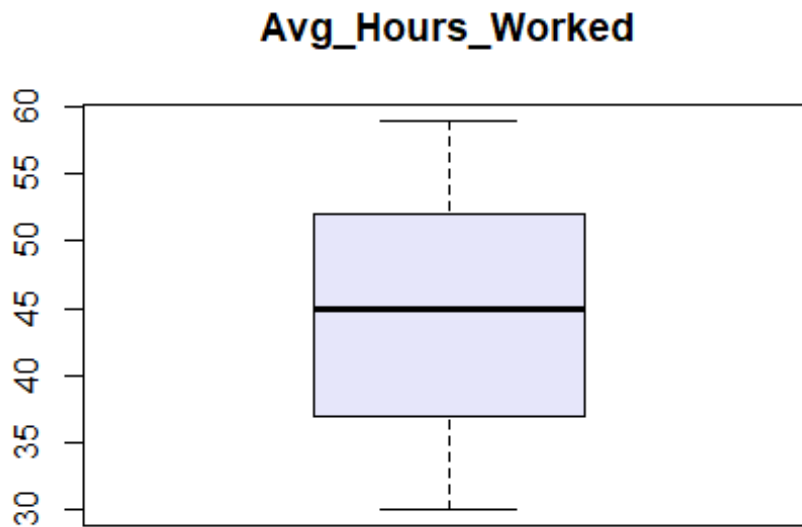
```
boxplot(Employee_Data$Project_Count, main = "Project_Count",  
col="lightgray", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Project_Count)$out))
```

Project_Count



outliers:

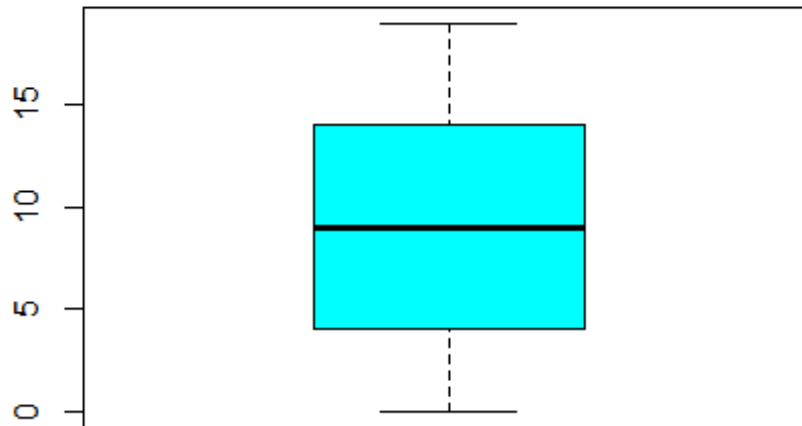
```
boxplot(Employee_Data$Average_Hours_Worked_Per_Week, main =  
"Avg_Hours_Worked", col="lavender", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Average_Hours_Worked_Per_Week)$out))
```



outliers:

```
boxplot(Employee_Data$Absenteeism, main = "Absenteeism", col="cyan", sub =  
paste("outliers:", boxplot.stats(Employee_Data$Absenteeism)$out))
```

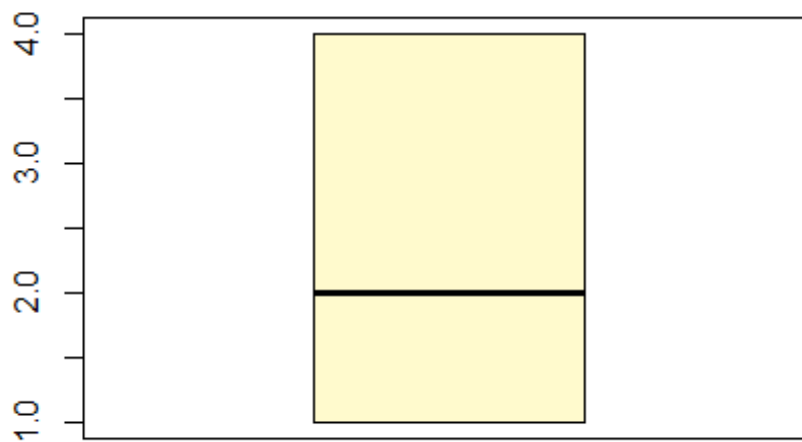
Absenteeism



outliers:

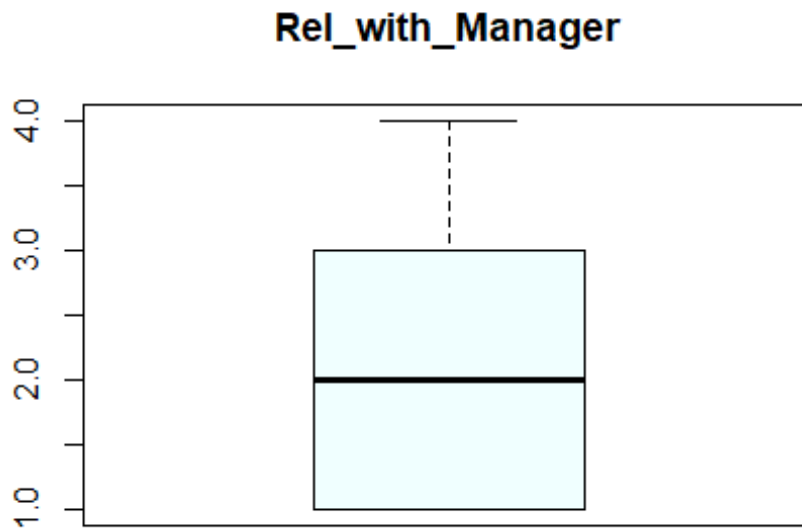
```
boxplot(Employee_Data$Work_Environment_Satisfaction, main = "Work_Env_Sat",  
col="lemonchiffon", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Work_Environment_Satisfaction)$out))
```

Work_Env_Sat



outliers:

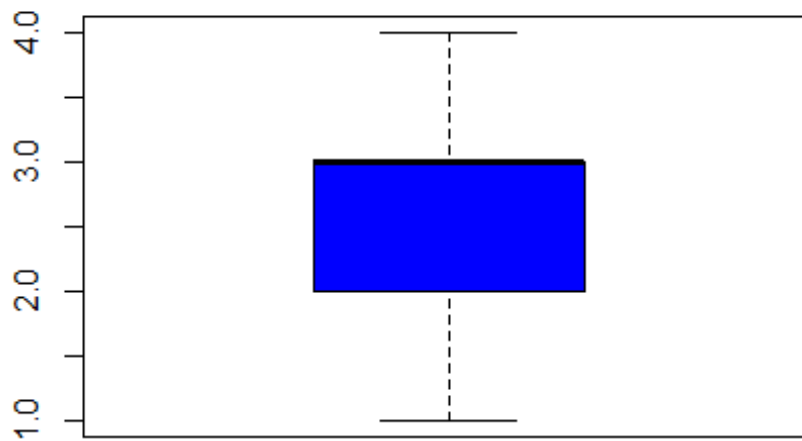
```
boxplot(Employee_Data$Relationship_with_Manager, main =  
"Rel_with_Manager", col="azure", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Relationship_with_Manager)$out))
```



outliers:

```
boxplot(Employee_Data$Job_Involvement, main = "Job_Involvement",  
col="blue", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Job_Involvement)$out))
```

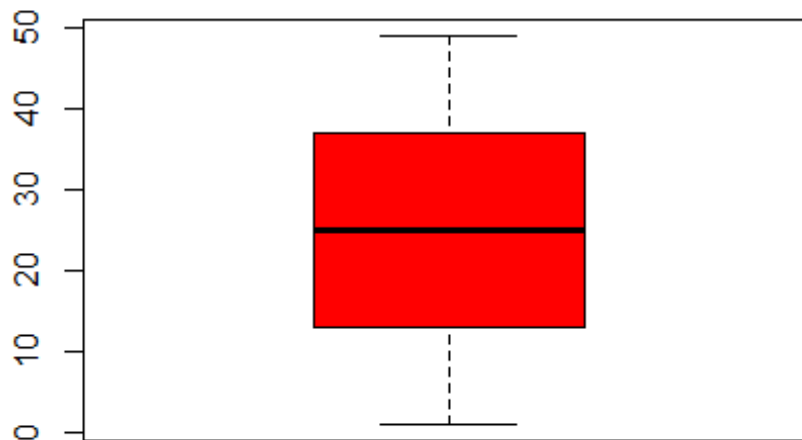
Job_Involvement



outliers:

```
boxplot(Employee_Data$Distance_From_Home, main =  
"Distance_From_Home", col="red", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Distance_From_Home)$out))
```

Distance_From_Home

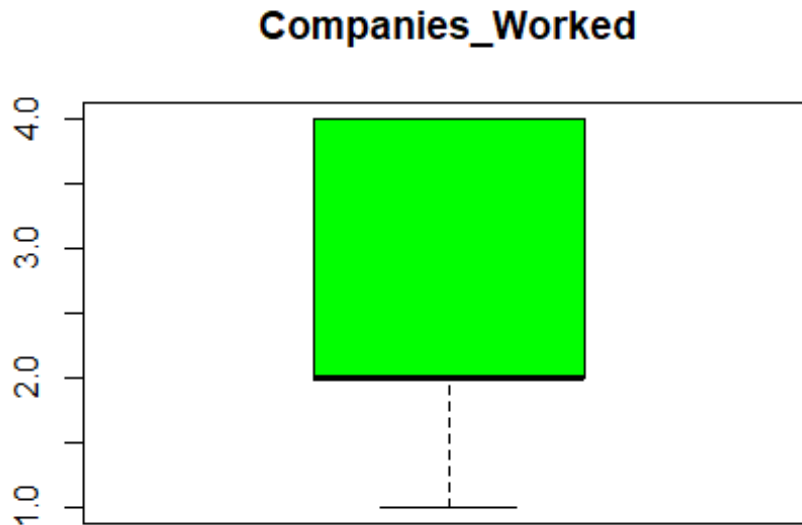


outliers:


```

boxplot(Employee_Data$Number_of_Companies_Worked, main = "Companies_Worked",
col="green",sub = paste("outliers:",
boxplot.stats(Employee_Data$Number_of_Companies_Worked)$out))

```



outliers:

```

# Pick numeric columns only
numeric_cols <- sapply(Employee_Data, is.numeric)

# Subset numeric data
numeric_data <- Employee_Data[, numeric_cols]

# Calculate mean and standard deviation for each numeric column
means <- sapply(numeric_data, mean)
sds <- sapply(numeric_data, sd)

# Print means and standard deviations
print(data.frame(variable = names(means), mean = means, sd = sds))

##               variable      mean
## Age                  Age    39.5618
## Job_Level            Job_Level  2.9908
## Monthly_Income       Monthly_Income 11436.7167
## Hourly_Rate           Hourly_Rate  57.0323
## Years_at_Company      Years_at_Company 14.9362
## Years_in_Current_Role Years_in_Current_Role  7.4513
## Years_Since_Last_Promotion Years_Since_Last_Promotion  4.4719
## Work_Life_Balance     Work_Life_Balance  2.5024
## Job_Satisfaction      Job_Satisfaction  3.0380

```

## Training_Hours_Last_Year	Training_Hours_Last_Year	49.5889
## Project_Count	Project_Count	4.9844
## Average_Hours_Worked_Per_Week	Average_Hours_Worked_Per_Week	44.4735
## Absenteeism	Absenteeism	9.4102
## Work_Environment_Satisfaction	Work_Environment_Satisfaction	2.4931
## Relationship_with_Manager	Relationship_with_Manager	2.4914
## Job_Involvement	Job_Involvement	2.5054
## Distance_From_Home	Distance_From_Home	25.2720
## Number_of_Companies_Worked	Number_of_Companies_Worked	2.5166
##	sd	
## Age	11.454986	
## Job_Level	1.410643	
## Monthly_Income	4926.528302	
## Hourly_Rate	24.703261	
## Years_at_Company	8.431657	
## Years_in_Current_Role	4.042903	
## Years_Since_Last_Promotion	2.891617	
## Work_Life_Balance	1.112348	
## Job_Satisfaction	1.414764	
## Training_Hours_Last_Year	28.801393	
## Project_Count	2.580043	
## Average_Hours_Worked_Per_Week	8.611662	
## Absenteeism	5.760335	
## Work_Environment_Satisfaction	1.120927	
## Relationship_with_Manager	1.115101	
## Job_Involvement	1.116018	
## Distance_From_Home	14.219474	
## Number_of_Companies_Worked	1.113934	

draw Histograms

Select numeric columns

```
numeric_cols <- sapply(Employee_Data, is.numeric)
```

Set up plotting window: adjust rows and cols based on number of variables

```
num_vars <- sum(numeric_cols)
```

```
rows <- ceiling(sqrt(num_vars))
```

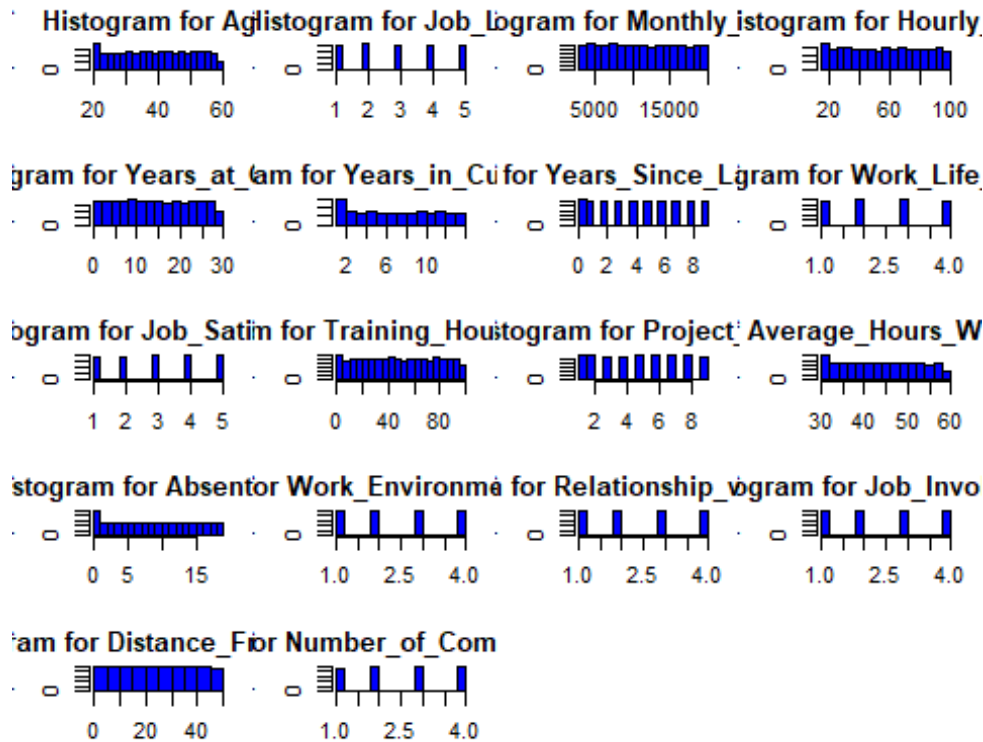
```
cols <- ceiling(num_vars / rows)
```

```
par(mfrow = c(rows, cols), mar = c(3, 3, 2, 1)) # margins: bottom, left, top, right
```

Loop to plot histograms for each numeric variable

```
for (var_name in names(Employee_Data)[numeric_cols]) {
  hist(Employee_Data[[var_name]],
    main = paste("Histogram for", var_name),
    col = "blue",
    xlab = var_name)
}
```

```
# Reset plotting layout to default
par(mfrow = c(1, 1))
```



```
library(e1071)
continuous_vars<- c("Job_Level","Monthly_Income","Hourly_Rate",
"Years_in_Current_Role","Years_Since_Last_Promotion","Work_Life_Balance","Job
_Satisfaction","Training_Hours_Last_Year","Project_Count","Average_Hours_Work
ed_Per_Week","Absenteeism","Work_Environment_Satisfaction","Relationship_with
_Manager","Job_Involvement","Distance_From_Home","Number_of_Companies_Worked"
)
```

```
# Calculate skewness for each variable
skew_values <- sapply(Employee_Data[continuous_vars], skewness, na.rm = TRUE)
```

```
# Print skewness values
print(skew_values)
```

```
##           Job_Level           Monthly_Income
##           0.015217678           0.015038630
##           Hourly_Rate           Years_in_Current_Role
##           0.007873296           0.003241685
##           Years_Since_Last_Promotion           Work_Life_Balance
##           -0.001873735           0.001810751
##           Job_Satisfaction           Training_Hours_Last_Year
##           -0.034540544           -0.005605580
##           Project_Count           Average_Hours_Worked_Per_Week
##           -0.008448485           -0.003639286
```

```

##          Absenteeism Work_Environment_Satisfaction
##          0.030083060          0.013832390
## Relationship_with_Manager          Job_Involvement
##          0.014011848          -0.011169225
##          Distance_From_Home Number_of_Companies_Worked
##          -0.012448076          -0.005456899

#all are closer to 0 ,so no need to apply transformation(fixed skewness)

# ----- CATEGORICAL VARIABLES -----

# Gender
chisq_gender <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Gender))
print(chisq_gender)

##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data:  table(Employee_Data$Performance, Employee_Data$Gender)
## X-squared = 1.5734, df = 1, p-value = 0.2097

# Marital_Status
chisq_MaritalStatus <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Marital_Status))
print(chisq_MaritalStatus)

##
## Pearson's Chi-squared test
##
## data:  table(Employee_Data$Performance, Employee_Data$Marital_Status)
## X-squared = 0.4812, df = 2, p-value = 0.7862

# Department
chisq_Depaertment <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Department))
print(chisq_Depaertment)

##
## Pearson's Chi-squared test
##
## data:  table(Employee_Data$Performance, Employee_Data$Department)
## X-squared = 3.3905, df = 4, p-value = 0.4947

# Job_Role
chisq_JobRole <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Job_Role))
print(chisq_JobRole)

##
## Pearson's Chi-squared test
##

```

```

## data:  table(Employee_Data$Performance, Employee_Data$Job_Role)
## X-squared = 2.6129, df = 3, p-value = 0.4552

# Job_Level (numeric, but ordinal, so still categorical)
chisq_JobLevel <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Job_Level))
print(chisq_JobLevel)

##
##  Pearson's Chi-squared test
##
## data:  table(Employee_Data$Performance, Employee_Data$Job_Level)
## X-squared = 10.68, df = 4, p-value = 0.03041

# Work_Life_Balance
chisq_WorklifeBalance <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Work_Life_Balance))
print(chisq_WorklifeBalance)

##
##  Pearson's Chi-squared test
##
## data:  table(Employee_Data$Performance, Employee_Data$Work_Life_Balance)
## X-squared = 4.1495, df = 3, p-value = 0.2458

# Job_Satisfaction
chisq_Jobsatisfaction <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Job_Satisfaction))
print(chisq_Jobsatisfaction)

##
##  Pearson's Chi-squared test
##
## data:  table(Employee_Data$Performance, Employee_Data$Job_Satisfaction)
## X-squared = 4.2625, df = 4, p-value = 0.3716

# Overtime
chisq_Overtime <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Overtime))
print(chisq_Overtime)

##
##  Pearson's Chi-squared test with Yates' continuity correction
##
## data:  table(Employee_Data$Performance, Employee_Data$Overtime)
## X-squared = 0.20622, df = 1, p-value = 0.6497

# Work_Environment_Satisfaction
chisq_EnviromentSatisfaction <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Work_Environment_Satisfaction))
print(chisq_EnviromentSatisfaction)

```

```

##
## Pearson's Chi-squared test
##
## data: table(Employee_Data$Performance,
Employee_Data$Work_Environment_Satisfaction)
## X-squared = 8.9978, df = 3, p-value = 0.02932

# Relationship_with_Manager
chisq_Relationwith_Manager <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Relationship_with_Manager))
print(chisq_Relationwith_Manager)

##
## Pearson's Chi-squared test
##
## data: table(Employee_Data$Performance,
Employee_Data$Relationship_with_Manager)
## X-squared = 6.4532, df = 3, p-value = 0.09153

# Job_Involvement
chisq_JobInvolvement <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Job_Involvement))
print(chisq_JobInvolvement)

##
## Pearson's Chi-squared test
##
## data: table(Employee_Data$Performance, Employee_Data$Job_Involvement)
## X-squared = 4.0665, df = 3, p-value = 0.2544

# Attrition
chisq_Attrition <- chisq.test(table(Employee_Data$Performance,
Employee_Data$Attrition))
print(chisq_Attrition)

##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: table(Employee_Data$Performance, Employee_Data$Attrition)
## X-squared = 2.7605, df = 1, p-value = 0.09662

library(dplyr)

##
## 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

set.seed(250)
split_ratio <- 0.8
df1_bound <- ceiling(nrow(Employee_Data) * split_ratio)

train1 <- Employee_Data %>% slice_sample(n = df1_bound, replace = FALSE)
head(train1)

## # A tibble: 6 × 25
##   Age Gender Marital_Status Department Job_Role Job_Level
##   <dbl> <fct> <fct>          <fct>    <fct>    <dbl>
##   <dbl>
## 1    55 Male    Married        Sales    Assistant    4
##   3362
## 2    27 Female Divorced        IT      Executive    1
##   6217
## 3    23 Female Single          Marketing Manager      1
##   3859
## 4    53 Male    Divorced        Sales    Executive    2
##   17808
## 5    49 Female Married          Marketing Assistant    1
##   14551
## 6    27 Male    Divorced        IT      Manager      4
##   7732
## # i 18 more variables: Hourly_Rate <dbl>, Years_at_Company <dbl>,
## #   Years_in_Current_Role <dbl>, Years_Since_Last_Promotion <dbl>,
## #   Work_Life_Balance <dbl>, Job_Satisfaction <dbl>, Performance <fct>,
## #   Training_Hours_Last_Year <dbl>, Overtime <fct>, Project_Count <dbl>,
## #   Average_Hours_Worked_Per_Week <dbl>, Absenteeism <dbl>,
## #   Work_Environment_Satisfaction <dbl>, Relationship_with_Manager <dbl>,
## #   Job_Involvement <dbl>, Distance_From_Home <dbl>, ...

test1 <- Employee_Data[-as.numeric(rownames(train1)),]
head(test1)

## # A tibble: 6 × 25
##   Age Gender Marital_Status Department Job_Role Job_Level
##   <dbl> <fct> <fct>          <fct>    <fct>    <dbl>
##   <dbl>
## 1    54 Female Divorced        Finance    Executive    3
##   5064
## 2    53 Female Married          HR      Assistant    1
##   12679
## 3    57 Female Divorced        IT      Manager      3
##   5345
## 4    24 Male    Single          HR      Assistant    2
```

```

5640
## 5    37 Male    Divorced      HR          Executive      3
4050
## 6    35 Male    Single        HR          Executive      5
15470
## # i 18 more variables: Hourly_Rate <dbl>, Years_at_Company <dbl>,
## #   Years_in_Current_Role <dbl>, Years_Since_Last_Promotion <dbl>,
## #   Work_Life_Balance <dbl>, Job_Satisfaction <dbl>, Performance <fct>,
## #   Training_Hours_Last_Year <dbl>, Overtime <fct>, Project_Count <dbl>,
## #   Average_Hours_Worked_Per_Week <dbl>, Absenteeism <dbl>,
## #   Work_Environment_Satisfaction <dbl>, Relationship_with_Manager <dbl>,
## #   Job_Involvement <dbl>, Distance_From_Home <dbl>, ...

#Univariable Logistic regression for each predictor
uni_Age <- glm(Performance ~ Age, data = train1, family = binomial)
summary(uni_Age)

##
## Call:
## glm(formula = Performance ~ Age, family = binomial, data = train1)
##
## Coefficients:
##             Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.093734   0.080243  -1.168   0.243
## Age          0.002295   0.001949   1.177   0.239
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11089  on 7998  degrees of freedom
## AIC: 11093
##
## Number of Fisher Scoring iterations: 3

uni_Job_Level <- glm(Performance ~ Job_Level, data = train1, family =
binomial)
summary(uni_Job_Level)

##
## Call:
## glm(formula = Performance ~ Job_Level, family = binomial, data = train1)
##
## Coefficients:
##             Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.05235    0.05229  -1.001   0.317
## Job_Level    0.01655    0.01585   1.044   0.296
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11089  on 7998  degrees of freedom

```



```

## AIC: 11093
##
## Number of Fisher Scoring iterations: 3

uni_Monthly_Income <- glm(Performance ~ Monthly_Income, data = train1, family
= binomial)
summary(uni_Monthly_Income)

##
## Call:
## glm(formula = Performance ~ Monthly_Income, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -2.793e-02  5.657e-02  -0.494    0.621
## Monthly_Income  2.179e-06  4.541e-06   0.480    0.631
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11090  on 7998  degrees of freedom
## AIC: 11094
##
## Number of Fisher Scoring iterations: 3

uni_Hourly_Rate <- glm(Performance ~ Hourly_Rate, data = train1, family =
binomial)
summary(uni_Hourly_Rate)

##
## Call:
## glm(formula = Performance ~ Hourly_Rate, family = binomial, data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.0306141  0.0563093  -0.544    0.587
## Hourly_Rate  0.0004842  0.0009062   0.534    0.593
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11090  on 7998  degrees of freedom
## AIC: 11094
##
## Number of Fisher Scoring iterations: 3

uni_Years_at_Company <- glm(Performance ~ Years_at_Company, data = train1,
family = binomial)
summary(uni_Years_at_Company)

```

```
##
## Call:
## glm(formula = Performance ~ Years_at_Company, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.037142   0.045346   0.819   0.413
## Years_at_Company -0.002696   0.002649  -1.018   0.309
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11089  on 7998  degrees of freedom
## AIC: 11093
##
## Number of Fisher Scoring iterations: 3

uni_Years_in_Current_Role <- glm(Performance ~ Years_in_Current_Role, data =
train1, family = binomial)
summary(uni_Years_in_Current_Role)

##
## Call:
## glm(formula = Performance ~ Years_in_Current_Role, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -0.029259   0.046813  -0.625   0.532
## Years_in_Current_Role  0.003523   0.005518   0.638   0.523
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11090  on 7998  degrees of freedom
## AIC: 11094
##
## Number of Fisher Scoring iterations: 3

uni_Years_Since_Last_Promotion <- glm(Performance ~
Years_Since_Last_Promotion, data = train1, family = binomial)
summary(uni_Years_Since_Last_Promotion)

##
## Call:
## glm(formula = Performance ~ Years_Since_Last_Promotion, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
```

```
## (Intercept)          0.005775    0.041282    0.140    0.889
## Years_Since_Last_Promotion -0.001959    0.007749   -0.253    0.800
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11090  on 7998  degrees of freedom
## AIC: 11094
##
## Number of Fisher Scoring iterations: 3

uni_Work_Life_Balance <- glm(Performance ~ Work_Life_Balance, data = train1,
family = binomial)
summary(uni_Work_Life_Balance)

##
## Call:
## glm(formula = Performance ~ Work_Life_Balance, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -0.025843   0.055150  -0.469    0.639
## Work_Life_Balance  0.009102   0.020088   0.453    0.650
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11090  on 7998  degrees of freedom
## AIC: 11094
##
## Number of Fisher Scoring iterations: 3

uni_Job_Satisfaction <- glm(Performance ~ Job_Satisfaction, data = train1,
family = binomial)
summary(uni_Job_Satisfaction)

##
## Call:
## glm(formula = Performance ~ Job_Satisfaction, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.023761   0.052951   0.449    0.654
## Job_Satisfaction -0.008812   0.015805  -0.558    0.577
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11090  on 7998  degrees of freedom
```

```

## AIC: 11094
##
## Number of Fisher Scoring iterations: 3

uni_Training_Hours_Last_Year <- glm(Performance ~ Training_Hours_Last_Year,
data = train1, family = binomial)
summary(uni_Training_Hours_Last_Year)

##
## Call:
## glm(formula = Performance ~ Training_Hours_Last_Year, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.0782021   0.0446381    1.752   0.0798 .
## Training_Hours_Last_Year -0.0016329   0.0007768   -2.102   0.0356 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11086  on 7998  degrees of freedom
## AIC: 11090
##
## Number of Fisher Scoring iterations: 3

uni_Project_Count <- glm(Performance ~ Project_Count, data = train1, family =
binomial)
summary(uni_Project_Count)

##
## Call:
## glm(formula = Performance ~ Project_Count, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  -0.035685   0.048790  -0.731   0.465
## Project_Count  0.006528   0.008660   0.754   0.451
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11090  on 7998  degrees of freedom
## AIC: 11094
##
## Number of Fisher Scoring iterations: 3

```

```

uni_Average_Hours_Worked_Per_Week <- glm(Performance ~
Average_Hours_Worked_Per_Week, data = train1, family = binomial)
summary(uni_Average_Hours_Worked_Per_Week)

##
## Call:
## glm(formula = Performance ~ Average_Hours_Worked_Per_Week, family =
binomial,
## data = train1)
##
## Coefficients:
## Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.084926 0.118128 -0.719 0.472
## Average_Hours_Worked_Per_Week 0.001839 0.002603 0.706 0.480
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 11090 on 7999 degrees of freedom
## Residual deviance: 11090 on 7998 degrees of freedom
## AIC: 11094
##
## Number of Fisher Scoring iterations: 3

uni_Absenteeism <- glm(Performance ~ Absenteeism, data = train1, family =
binomial)
summary(uni_Absenteeism)

##
## Call:
## glm(formula = Performance ~ Absenteeism, family = binomial, data = train1)
##
## Coefficients:
## Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.0050472 0.0431001 -0.117 0.907
## Absenteeism 0.0002164 0.0038944 0.056 0.956
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 11090 on 7999 degrees of freedom
## Residual deviance: 11090 on 7998 degrees of freedom
## AIC: 11094
##
## Number of Fisher Scoring iterations: 3

uni_Work_Environment_Satisfaction <- glm(Performance ~
Work_Environment_Satisfaction, data = train1, family = binomial)
summary(uni_Work_Environment_Satisfaction)

##
## Call:
## glm(formula = Performance ~ Work_Environment_Satisfaction, family =

```

```

binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -0.10901    0.05456  -1.998   0.0457 *
## Work_Environment_Satisfaction  0.04245    0.01993   2.130   0.0331 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11086  on 7998  degrees of freedom
## AIC: 11090
##
## Number of Fisher Scoring iterations: 3

uni_Relationship_with_Manager <- glm(Performance ~ Relationship_with_Manager,
data = train1, family = binomial)
summary(uni_Relationship_with_Manager)

##
## Call:
## glm(formula = Performance ~ Relationship_with_Manager, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.07794    0.05476   1.423   0.155
## Relationship_with_Manager -0.03238    0.02000  -1.619   0.105
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11088  on 7998  degrees of freedom
## AIC: 11092
##
## Number of Fisher Scoring iterations: 3

uni_Job_Involvement <- glm(Performance ~ Job_Involvement, data = train1,
family = binomial)
summary(uni_Job_Involvement)

##
## Call:
## glm(formula = Performance ~ Job_Involvement, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)

```

```

## (Intercept)      0.10558      0.05513      1.915      0.0555 .
## Job_Involvement -0.04324      0.02007     -2.155      0.0312 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11086  on 7998  degrees of freedom
## AIC: 11090
##
## Number of Fisher Scoring iterations: 3

uni_Distance_From_Home <- glm(Performance ~ Distance_From_Home, data =
train1, family = binomial)
summary(uni_Distance_From_Home)

##
## Call:
## glm(formula = Performance ~ Distance_From_Home, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -0.047147   0.045649  -1.033   0.302
## Distance_From_Home  0.001748   0.001576   1.109   0.267
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11089  on 7998  degrees of freedom
## AIC: 11093
##
## Number of Fisher Scoring iterations: 3

uni_Number_of_Companies_Worked <- glm(Performance ~
Number_of_Companies_Worked, data = train1, family = binomial)
summary(uni_Number_of_Companies_Worked)

##
## Call:
## glm(formula = Performance ~ Number_of_Companies_Worked, family = binomial,
##      data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -0.021307   0.055193  -0.386   0.699
## Number_of_Companies_Worked  0.007261   0.020014   0.363   0.717
##
## (Dispersion parameter for binomial family taken to be 1)
##

```

```
## Null deviance: 11090 on 7999 degrees of freedom
## Residual deviance: 11090 on 7998 degrees of freedom
## AIC: 11094
##
## Number of Fisher Scoring iterations: 3
```

#Fitting binary logistic regression using "glm" function

```
data<-train1
```

```
model1<-glm(Performance ~.,data=data,family="binomial"(link=logit))
```

```
summary(model1)
```

```
##
## Call:
## glm(formula = Performance ~ ., family = binomial(link = logit),
## data = data)
##
## Coefficients:
## Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.632e-01 2.414e-01 -0.676 0.4991
## Age 2.085e-03 1.957e-03 1.065 0.2869
## GenderMale 4.921e-02 4.491e-02 1.096 0.2732
## Marital_StatusMarried -7.768e-02 5.488e-02 -1.416 0.1569
## Marital_StatusSingle -5.179e-02 5.520e-02 -0.938 0.3481
## DepartmentHR 3.610e-02 7.165e-02 0.504 0.6144
## DepartmentIT -5.269e-02 7.186e-02 -0.733 0.4634
## DepartmentMarketing 3.234e-02 6.996e-02 0.462 0.6438
## DepartmentSales 6.882e-02 7.104e-02 0.969 0.3327
## Job_RoleAssistant 3.899e-02 6.292e-02 0.620 0.5354
## Job_RoleExecutive 4.652e-02 6.317e-02 0.736 0.4615
## Job_RoleManager -9.540e-04 6.361e-02 -0.015 0.9880
## Job_Level 1.829e-02 1.591e-02 1.149 0.2504
## Monthly_Income 2.753e-06 4.560e-06 0.604 0.5461
## Hourly_Rate 4.592e-04 9.094e-04 0.505 0.6136
## Years_at_Company -2.677e-03 2.659e-03 -1.007 0.3140
## Years_in_Current_Role 3.481e-03 5.540e-03 0.628 0.5298
## Years_Since_Last_Promotion -2.701e-03 7.775e-03 -0.347 0.7283
## Work_Life_Balance 9.211e-03 2.017e-02 0.457 0.6480
## Job_Satisfaction -9.868e-03 1.587e-02 -0.622 0.5339
## Training_Hours_Last_Year -1.625e-03 7.804e-04 -2.083 0.0373 *
## OvertimeYes -4.915e-03 4.489e-02 -0.109 0.9128
## Project_Count 7.024e-03 8.697e-03 0.808 0.4193
## Average_Hours_Worked_Per_Week 1.635e-03 2.613e-03 0.626 0.5314
## Absenteeism 4.713e-04 3.908e-03 0.121 0.9040
## Work_Environment_Satisfaction 3.980e-02 2.000e-02 1.989 0.0466 *
## Relationship_with_Manager -3.302e-02 2.009e-02 -1.644 0.1002
## Job_Involvement -4.342e-02 2.013e-02 -2.157 0.0310 *
## Distance_From_Home 1.801e-03 1.582e-03 1.138 0.2551
```



```
## Number_of_Companies_Worked      6.459e-03  2.009e-02  0.321  0.7479
## AttritionYes                    -1.027e-01  5.635e-02  -1.822  0.0685 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11056  on 7969  degrees of freedom
## AIC: 11118
##
## Number of Fisher Scoring iterations: 3
```

```
library(car)
```

```
## Loading required package: carData
```

```
##
```

```
## Attaching package: 'car'
```

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

```
##
```

```
##      recode
```

```
print(vif(model1))
```

```
##              GVIF Df GVIF^(1/(2*Df))
## Age              1.004510  1      1.002253
## Gender            1.003941  1      1.001969
## Marital_Status    1.005901  2      1.001472
## Department        1.011895  4      1.001479
## Job_Role           1.014737  3      1.002441
## Job_Level          1.003363  1      1.001680
## Monthly_Income     1.004296  1      1.002145
## Hourly_Rate        1.002746  1      1.001372
## Years_at_Company   1.002920  1      1.001459
## Years_in_Current_Role 1.003898  1      1.001947
## Years_Since_Last_Promotion 1.002473  1      1.001236
## Work_Life_Balance  1.004363  1      1.002179
## Job_Satisfaction   1.003548  1      1.001773
## Training_Hours_Last_Year 1.005468  1      1.002730
## Overtime           1.002513  1      1.001256
## Project_Count      1.004298  1      1.002147
## Average_Hours_Worked_Per_Week 1.003357  1      1.001677
## Absenteeism        1.002704  1      1.001351
## Work_Environment_Satisfaction 1.004086  1      1.002041
## Relationship_with_Manager 1.005147  1      1.002570
## Job_Involvement    1.002396  1      1.001197
## Distance_From_Home 1.004135  1      1.002066
## Number_of_Companies_Worked 1.003903  1      1.001950
## Attrition          1.003702  1      1.001849
```

```

step_model <- step(model1, direction = "both")

## Start:  AIC=11118.24
## Performance ~ Age + Gender + Marital_Status + Department + Job_Role +
##   Job_Level + Monthly_Income + Hourly_Rate + Years_at_Company +
##   Years_in_Current_Role + Years_Since_Last_Promotion + Work_Life_Balance
## +
##   Job_Satisfaction + Training_Hours_Last_Year + Overtime +
##   Project_Count + Average_Hours_Worked_Per_Week + Absenteeism +
##   Work_Environment_Satisfaction + Relationship_with_Manager +
##   Job_Involvement + Distance_From_Home + Number_of_Companies_Worked +
##   Attrition
##
##
##           Df Deviance   AIC
## - Job_Role      3    11057 11113
## - Department    4    11060 11114
## - Overtime       1    11056 11116
## - Absenteeism    1    11056 11116
## - Marital_Status 2    11058 11116
## - Number_of_Companies_Worked 1    11056 11116
## - Years_Since_Last_Promotion 1    11056 11116
## - Work_Life_Balance 1    11056 11116
## - Hourly_Rate    1    11056 11116
## - Monthly_Income 1    11057 11117
## - Job_Satisfaction 1    11057 11117
## - Average_Hours_Worked_Per_Week 1    11057 11117
## - Years_in_Current_Role 1    11057 11117
## - Project_Count  1    11057 11117
## - Years_at_Company 1    11057 11117
## - Age            1    11057 11117
## - Gender          1    11057 11117
## - Distance_From_Home 1    11058 11118
## - Job_Level       1    11058 11118
## <none>           11056 11118
## - Relationship_with_Manager 1    11059 11119
## - Attrition       1    11060 11120
## - Work_Environment_Satisfaction 1    11060 11120
## - Training_Hours_Last_Year 1    11061 11121
## - Job_Involvement 1    11061 11121
##
## Step:  AIC=11113.18
## Performance ~ Age + Gender + Marital_Status + Department + Job_Level +
##   Monthly_Income + Hourly_Rate + Years_at_Company +
##   Years_in_Current_Role +
##   Years_Since_Last_Promotion + Work_Life_Balance + Job_Satisfaction +
##   Training_Hours_Last_Year + Overtime + Project_Count +
##   Average_Hours_Worked_Per_Week +
##   Absenteeism + Work_Environment_Satisfaction +
##   Relationship_with_Manager +
##   Job_Involvement + Distance_From_Home + Number_of_Companies_Worked +

```

```

##      Attrition
##
##
##              Df Deviance   AIC
## - Department      4    11060 11108
## - Overtime         1    11057 11111
## - Absenteeism      1    11057 11111
## - Number_of_Companies_Worked 1    11057 11111
## - Marital_Status   2    11059 11111
## - Years_Since_Last_Promotion 1    11057 11111
## - Work_Life_Balance 1    11057 11111
## - Hourly_Rate      1    11057 11111
## - Monthly_Income   1    11058 11112
## - Job_Satisfaction 1    11058 11112
## - Average_Hours_Worked_Per_Week 1    11058 11112
## - Years_in_Current_Role 1    11058 11112
## - Project_Count    1    11058 11112
## - Years_at_Company 1    11058 11112
## - Age              1    11058 11112
## - Gender            1    11058 11112
## - Job_Level         1    11058 11112
## - Distance_From_Home 1    11058 11112
## <none>              11057 11113
## - Relationship_with_Manager 1    11060 11114
## - Attrition         1    11060 11114
## - Work_Environment_Satisfaction 1    11061 11115
## - Training_Hours_Last_Year 1    11061 11115
## - Job_Involvement   1    11062 11116
## + Job_Role          3    11056 11118
##
## Step:  AIC=11108.47
## Performance ~ Age + Gender + Marital_Status + Job_Level + Monthly_Income +
##      Hourly_Rate + Years_at_Company + Years_in_Current_Role +
##      Years_Since_Last_Promotion + Work_Life_Balance + Job_Satisfaction +
##      Training_Hours_Last_Year + Overtime + Project_Count +
##      Average_Hours_Worked_Per_Week +
##      Absenteeism + Work_Environment_Satisfaction +
##      Relationship_with_Manager +
##      Job_Involvement + Distance_From_Home + Number_of_Companies_Worked +
##      Attrition
##
##              Df Deviance   AIC
## - Absenteeism      1    11060 11106
## - Overtime         1    11060 11106
## - Marital_Status   2    11062 11106
## - Number_of_Companies_Worked 1    11061 11107
## - Years_Since_Last_Promotion 1    11061 11107
## - Work_Life_Balance 1    11061 11107
## - Hourly_Rate      1    11061 11107
## - Monthly_Income   1    11061 11107
## - Job_Satisfaction 1    11061 11107

```

```

## - Years_in_Current_Role      1      11061 11107
## - Average_Hours_Worked_Per_Week 1      11061 11107
## - Project_Count              1      11061 11107
## - Years_at_Company           1      11062 11108
## - Age                        1      11062 11108
## - Job_Level                  1      11062 11108
## - Gender                     1      11062 11108
## - Distance_From_Home         1      11062 11108
## <none>                       11060 11108
## - Relationship_with_Manager  1      11063 11109
## - Attrition                  1      11064 11110
## - Work_Environment_Satisfaction 1      11064 11110
## - Training_Hours_Last_Year   1      11065 11111
## - Job_Involvement            1      11065 11111
## + Department                 4      11057 11113
## + Job_Role                   3      11060 11114
##
## Step: AIC=11106.48
## Performance ~ Age + Gender + Marital_Status + Job_Level + Monthly_Income +
##   Hourly_Rate + Years_at_Company + Years_in_Current_Role +
##   Years_Since_Last_Promotion + Work_Life_Balance + Job_Satisfaction +
##   Training_Hours_Last_Year + Overtime + Project_Count +
##   Average_Hours_Worked_Per_Week +
##   Work_Environment_Satisfaction + Relationship_with_Manager +
##   Job_Involvement + Distance_From_Home + Number_of_Companies_Worked +
##   Attrition
##
##                               Df Deviance  AIC
## - Overtime                   1      11060 11104
## - Marital_Status              2      11062 11104
## - Number_of_Companies_Worked  1      11061 11105
## - Years_Since_Last_Promotion  1      11061 11105
## - Work_Life_Balance           1      11061 11105
## - Hourly_Rate                 1      11061 11105
## - Monthly_Income              1      11061 11105
## - Job_Satisfaction            1      11061 11105
## - Years_in_Current_Role       1      11061 11105
## - Average_Hours_Worked_Per_Week 1      11061 11105
## - Project_Count               1      11061 11105
## - Years_at_Company            1      11062 11106
## - Age                         1      11062 11106
## - Job_Level                   1      11062 11106
## - Gender                      1      11062 11106
## - Distance_From_Home          1      11062 11106
## <none>                       11060 11106
## - Relationship_with_Manager   1      11063 11107
## - Attrition                   1      11064 11108
## + Absenteeism                 1      11060 11108
## - Work_Environment_Satisfaction 1      11064 11108
## - Training_Hours_Last_Year    1      11065 11109

```

```

## - Job_Involvement          1    11065 11109
## + Department              4    11057 11111
## + Job_Role                 3    11060 11112
##
## Step:  AIC=11104.5
## Performance ~ Age + Gender + Marital_Status + Job_Level + Monthly_Income +
##   Hourly_Rate + Years_at_Company + Years_in_Current_Role +
##   Years_Since_Last_Promotion + Work_Life_Balance + Job_Satisfaction +
##   Training_Hours_Last_Year + Project_Count +
##   Average_Hours_Worked_Per_Week +
##   Work_Environment_Satisfaction + Relationship_with_Manager +
##   Job_Involvement + Distance_From_Home + Number_of_Companies_Worked +
##   Attrition
##
##                                     Df Deviance   AIC
## - Marital_Status                   2    11062 11102
## - Number_of_Companies_Worked       1    11061 11103
## - Years_Since_Last_Promotion       1    11061 11103
## - Work_Life_Balance                1    11061 11103
## - Hourly_Rate                     1    11061 11103
## - Monthly_Income                   1    11061 11103
## - Job_Satisfaction                 1    11061 11103
## - Years_in_Current_Role            1    11061 11103
## - Average_Hours_Worked_Per_Week    1    11061 11103
## - Project_Count                    1    11061 11103
## - Years_at_Company                 1    11062 11104
## - Age                             1    11062 11104
## - Job_Level                        1    11062 11104
## - Gender                           1    11062 11104
## - Distance_From_Home               1    11062 11104
## <none>                             1    11060 11104
## - Relationship_with_Manager        1    11063 11105
## - Attrition                        1    11064 11106
## + Overtime                         1    11060 11106
## + Absenteeism                      1    11060 11106
## - Work_Environment_Satisfaction    1    11064 11106
## - Training_Hours_Last_Year         1    11065 11107
## - Job_Involvement                  1    11065 11107
## + Department                       4    11057 11109
## + Job_Role                         3    11060 11110
##
## Step:  AIC=11102.51
## Performance ~ Age + Gender + Job_Level + Monthly_Income + Hourly_Rate +
##   Years_at_Company + Years_in_Current_Role + Years_Since_Last_Promotion
## +
##   Work_Life_Balance + Job_Satisfaction + Training_Hours_Last_Year +
##   Project_Count + Average_Hours_Worked_Per_Week +
##   Work_Environment_Satisfaction +
##   Relationship_with_Manager + Job_Involvement + Distance_From_Home +
##   Number_of_Companies_Worked + Attrition

```

```

##
##
## - Number_of_Companies_Worked      1    11063 11101
## - Years_Since_Last_Promotion      1    11063 11101
## - Work_Life_Balance                1    11063 11101
## - Hourly_Rate                     1    11063 11101
## - Job_Satisfaction                 1    11063 11101
## - Monthly_Income                   1    11063 11101
## - Years_in_Current_Role            1    11063 11101
## - Average_Hours_Worked_Per_Week    1    11063 11101
## - Project_Count                    1    11063 11101
## - Years_at_Company                 1    11064 11102
## - Age                             1    11064 11102
## - Job_Level                        1    11064 11102
## - Gender                           1    11064 11102
## - Distance_From_Home               1    11064 11102
## <none>                             11062 11102
## - Relationship_with_Manager         1    11065 11103
## - Attrition                         1    11066 11104
## + Marital_Status                    2    11060 11104
## + Overtime                          1    11062 11104
## + Absenteeism                       1    11062 11104
## - Work_Environment_Satisfaction     1    11066 11104
## - Training_Hours_Last_Year          1    11067 11105
## - Job_Involvement                   1    11067 11105
## + Department                        4    11059 11107
## + Job_Role                          3    11062 11108
##
## Step:  AIC=11100.61
## Performance ~ Age + Gender + Job_Level + Monthly_Income + Hourly_Rate +
##      Years_at_Company + Years_in_Current_Role + Years_Since_Last_Promotion
##      +
##      Work_Life_Balance + Job_Satisfaction + Training_Hours_Last_Year +
##      Project_Count + Average_Hours_Worked_Per_Week +
##      Work_Environment_Satisfaction +
##      Relationship_with_Manager + Job_Involvement + Distance_From_Home +
##      Attrition
##
##
##      Df Deviance  AIC
## - Years_Since_Last_Promotion      1    11063 11099
## - Work_Life_Balance                1    11063 11099
## - Hourly_Rate                     1    11063 11099
## - Job_Satisfaction                 1    11063 11099
## - Monthly_Income                   1    11063 11099
## - Years_in_Current_Role            1    11063 11099
## - Average_Hours_Worked_Per_Week    1    11063 11099
## - Project_Count                    1    11063 11099
## - Years_at_Company                 1    11064 11100
## - Age                             1    11064 11100
## - Job_Level                        1    11064 11100

```

```

## - Gender 1 11064 11100
## - Distance_From_Home 1 11064 11100
## <none> 11063 11101
## - Relationship_with_Manager 1 11065 11101
## - Attrition 1 11066 11102
## + Number_of_Companies_Worked 1 11062 11102
## + Marital_Status 2 11061 11103
## + Overtime 1 11063 11103
## + Absenteeism 1 11063 11103
## - Work_Environment_Satisfaction 1 11067 11103
## - Training_Hours_Last_Year 1 11067 11103
## - Job_Involvement 1 11067 11103
## + Department 4 11059 11105
## + Job_Role 3 11062 11106
##
## Step: AIC=11098.73
## Performance ~ Age + Gender + Job_Level + Monthly_Income + Hourly_Rate +
## Years_at_Company + Years_in_Current_Role + Work_Life_Balance +
## Job_Satisfaction + Training_Hours_Last_Year + Project_Count +
## Average_Hours_Worked_Per_Week + Work_Environment_Satisfaction +
## Relationship_with_Manager + Job_Involvement + Distance_From_Home +
## Attrition
##
## Df Deviance AIC
## - Work_Life_Balance 1 11063 11097
## - Hourly_Rate 1 11063 11097
## - Job_Satisfaction 1 11063 11097
## - Monthly_Income 1 11063 11097
## - Years_in_Current_Role 1 11063 11097
## - Average_Hours_Worked_Per_Week 1 11063 11097
## - Project_Count 1 11063 11097
## - Years_at_Company 1 11064 11098
## - Age 1 11064 11098
## - Job_Level 1 11064 11098
## - Gender 1 11064 11098
## - Distance_From_Home 1 11064 11098
## <none> 11063 11099
## - Relationship_with_Manager 1 11066 11100
## - Attrition 1 11066 11100
## + Years_Since_Last_Promotion 1 11063 11101
## + Number_of_Companies_Worked 1 11063 11101
## + Marital_Status 2 11061 11101
## + Overtime 1 11063 11101
## + Absenteeism 1 11063 11101
## - Work_Environment_Satisfaction 1 11067 11101
## - Training_Hours_Last_Year 1 11067 11101
## - Job_Involvement 1 11067 11101
## + Department 4 11060 11104
## + Job_Role 3 11062 11104
##

```

```

## Step:  AIC=11096.94
## Performance ~ Age + Gender + Job_Level + Monthly_Income + Hourly_Rate +
##     Years_at_Company + Years_in_Current_Role + Job_Satisfaction +
##     Training_Hours_Last_Year + Project_Count +
Average_Hours_Worked_Per_Week +
##     Work_Environment_Satisfaction + Relationship_with_Manager +
##     Job_Involvement + Distance_From_Home + Attrition
##
##
##           Df Deviance   AIC
## - Hourly_Rate           1    11063 11095
## - Monthly_Income        1    11063 11095
## - Job_Satisfaction       1    11063 11095
## - Years_in_Current_Role  1    11063 11095
## - Average_Hours_Worked_Per_Week 1    11063 11095
## - Project_Count         1    11064 11096
## - Years_at_Company      1    11064 11096
## - Age                   1    11064 11096
## - Job_Level             1    11064 11096
## - Gender                1    11064 11096
## - Distance_From_Home    1    11064 11096
## <none>                  1    11063 11097
## - Relationship_with_Manager 1    11066 11098
## - Attrition             1    11066 11098
## + Work_Life_Balance     1    11063 11099
## + Years_Since_Last_Promotion 1    11063 11099
## + Number_of_Companies_Worked 1    11063 11099
## + Marital_Status        2    11061 11099
## + Overtime              1    11063 11099
## + Absenteeism           1    11063 11099
## - Work_Environment_Satisfaction 1    11067 11099
## - Training_Hours_Last_Year 1    11067 11099
## - Job_Involvement       1    11068 11100
## + Department            4    11060 11102
## + Job_Role              3    11062 11102
##
## Step:  AIC=11095.18
## Performance ~ Age + Gender + Job_Level + Monthly_Income + Years_at_Company
+
##     Years_in_Current_Role + Job_Satisfaction + Training_Hours_Last_Year +
##     Project_Count + Average_Hours_Worked_Per_Week +
Work_Environment_Satisfaction +
##     Relationship_with_Manager + Job_Involvement + Distance_From_Home +
##     Attrition
##
##
##           Df Deviance   AIC
## - Job_Satisfaction       1    11064 11094
## - Monthly_Income        1    11064 11094
## - Years_in_Current_Role  1    11064 11094
## - Average_Hours_Worked_Per_Week 1    11064 11094
## - Project_Count         1    11064 11094

```



```

## - Years_at_Company      1      11064 11094
## - Age                   1      11064 11094
## - Job_Level             1      11064 11094
## - Gender                1      11064 11094
## - Distance_From_Home    1      11064 11094
## <none>                  11063 11095
## - Relationship_with_Manager 1      11066 11096
## - Attrition              1      11066 11096
## + Hourly_Rate            1      11063 11097
## + Work_Life_Balance      1      11063 11097
## + Years_Since_Last_Promotion 1      11063 11097
## + Number_of_Companies_Worked 1      11063 11097
## + Marital_Status         2      11061 11097
## + Overtime               1      11063 11097
## + Absenteeism            1      11063 11097
## - Work_Environment_Satisfaction 1      11067 11097
## - Training_Hours_Last_Year 1      11068 11098
## - Job_Involvement        1      11068 11098
## + Department             4      11060 11100
## + Job_Role               3      11062 11100
##
## Step:  AIC=11093.52
## Performance ~ Age + Gender + Job_Level + Monthly_Income + Years_at_Company
+
##   Years_in_Current_Role + Training_Hours_Last_Year + Project_Count +
##   Average_Hours_Worked_Per_Week + Work_Environment_Satisfaction +
##   Relationship_with_Manager + Job_Involvement + Distance_From_Home +
##   Attrition
##
##
##              Df Deviance   AIC
## - Monthly_Income      1      11064 11092
## - Years_in_Current_Role 1      11064 11092
## - Average_Hours_Worked_Per_Week 1      11064 11092
## - Project_Count        1      11064 11092
## - Years_at_Company      1      11065 11093
## - Age                  1      11065 11093
## - Job_Level            1      11065 11093
## - Gender               1      11065 11093
## - Distance_From_Home    1      11065 11093
## <none>                 11064 11094
## - Relationship_with_Manager 1      11066 11094
## - Attrition            1      11067 11095
## + Job_Satisfaction      1      11063 11095
## + Hourly_Rate           1      11063 11095
## + Work_Life_Balance     1      11063 11095
## + Years_Since_Last_Promotion 1      11063 11095
## + Number_of_Companies_Worked 1      11063 11095
## + Marital_Status        2      11062 11096
## + Overtime              1      11064 11096
## + Absenteeism           1      11064 11096

```

```

## - Work_Environment_Satisfaction 1 11068 11096
## - Training_Hours_Last_Year 1 11068 11096
## - Job_Involvement 1 11068 11096
## + Department 4 11060 11098
## + Job_Role 3 11063 11099
##
## Step: AIC=11091.84
## Performance ~ Age + Gender + Job_Level + Years_at_Company +
Years_in_Current_Role +
## Training_Hours_Last_Year + Project_Count +
Average_Hours_Worked_Per_Week +
## Work_Environment_Satisfaction + Relationship_with_Manager +
## Job_Involvement + Distance_From_Home + Attrition
##
##
## Df Deviance AIC
## - Years_in_Current_Role 1 11064 11090
## - Average_Hours_Worked_Per_Week 1 11064 11090
## - Project_Count 1 11064 11090
## - Years_at_Company 1 11065 11091
## - Age 1 11065 11091
## - Job_Level 1 11065 11091
## - Gender 1 11065 11091
## - Distance_From_Home 1 11065 11091
## <none> 11064 11092
## - Relationship_with_Manager 1 11067 11093
## - Attrition 1 11067 11093
## + Monthly_Income 1 11064 11094
## + Job_Satisfaction 1 11064 11094
## + Hourly_Rate 1 11064 11094
## + Work_Life_Balance 1 11064 11094
## + Years_Since_Last_Promotion 1 11064 11094
## + Number_of_Companies_Worked 1 11064 11094
## + Marital_Status 2 11062 11094
## + Overtime 1 11064 11094
## + Absenteeism 1 11064 11094
## - Work_Environment_Satisfaction 1 11068 11094
## - Training_Hours_Last_Year 1 11068 11094
## - Job_Involvement 1 11068 11094
## + Department 4 11061 11097
## + Job_Role 3 11063 11097
##
## Step: AIC=11090.21
## Performance ~ Age + Gender + Job_Level + Years_at_Company +
Training_Hours_Last_Year +
## Project_Count + Average_Hours_Worked_Per_Week +
Work_Environment_Satisfaction +
## Relationship_with_Manager + Job_Involvement + Distance_From_Home +
## Attrition
##
##
## Df Deviance AIC

```

```

## - Average_Hours_Worked_Per_Week 1 11065 11089
## - Project_Count 1 11065 11089
## - Years_at_Company 1 11065 11089
## - Age 1 11065 11089
## - Job_Level 1 11065 11089
## - Gender 1 11066 11090
## - Distance_From_Home 1 11066 11090
## <none> 11064 11090
## - Relationship_with_Manager 1 11067 11091
## - Attrition 1 11067 11091
## + Years_in_Current_Role 1 11064 11092
## + Job_Satisfaction 1 11064 11092
## + Monthly_Income 1 11064 11092
## + Hourly_Rate 1 11064 11092
## + Work_Life_Balance 1 11064 11092
## + Years_Since_Last_Promotion 1 11064 11092
## + Number_of_Companies_Worked 1 11064 11092
## + Marital_Status 2 11062 11092
## + Absenteeism 1 11064 11092
## + Overtime 1 11064 11092
## - Work_Environment_Satisfaction 1 11068 11092
## - Training_Hours_Last_Year 1 11069 11093
## - Job_Involvement 1 11069 11093
## + Department 4 11061 11095
## + Job_Role 3 11063 11095
##
## Step: AIC=11088.67
## Performance ~ Age + Gender + Job_Level + Years_at_Company +
Training_Hours_Last_Year +
## Project_Count + Work_Environment_Satisfaction +
Relationship_with_Manager +
## Job_Involvement + Distance_From_Home + Attrition
##
## Df Deviance AIC
## - Project_Count 1 11065 11087
## - Years_at_Company 1 11066 11088
## - Job_Level 1 11066 11088
## - Gender 1 11066 11088
## - Age 1 11066 11088
## - Distance_From_Home 1 11066 11088
## <none> 11065 11089
## - Relationship_with_Manager 1 11067 11089
## - Attrition 1 11068 11090
## + Average_Hours_Worked_Per_Week 1 11064 11090
## + Years_in_Current_Role 1 11064 11090
## + Monthly_Income 1 11064 11090
## + Job_Satisfaction 1 11064 11090
## + Hourly_Rate 1 11064 11090
## + Work_Life_Balance 1 11064 11090
## + Years_Since_Last_Promotion 1 11065 11091

```

```

## + Number_of_Companies_Worked      1      11065 11091
## + Absenteeism                      1      11065 11091
## + Overtime                        1      11065 11091
## + Marital_Status                  2      11063 11091
## - Work_Environment_Satisfaction    1      11069 11091
## - Training_Hours_Last_Year        1      11069 11091
## - Job_Involvement                 1      11069 11091
## + Department                      4      11062 11094
## + Job_Role                        3      11064 11094
##
## Step:  AIC=11087.25
## Performance ~ Age + Gender + Job_Level + Years_at_Company +
Training_Hours_Last_Year +
##      Work_Environment_Satisfaction + Relationship_with_Manager +
##      Job_Involvement + Distance_From_Home + Attrition
##
##
##              Df Deviance   AIC
## - Years_at_Company      1      11066 11086
## - Job_Level              1      11066 11086
## - Gender                 1      11066 11086
## - Age                   1      11066 11086
## - Distance_From_Home    1      11067 11087
## <none>                  11065 11087
## - Relationship_with_Manager 1      11068 11088
## - Attrition              1      11068 11088
## + Project_Count          1      11065 11089
## + Average_Hours_Worked_Per_Week 1      11065 11089
## + Years_in_Current_Role   1      11065 11089
## + Monthly_Income         1      11065 11089
## + Job_Satisfaction        1      11065 11089
## + Hourly_Rate             1      11065 11089
## + Work_Life_Balance       1      11065 11089
## + Number_of_Companies_Worked 1      11065 11089
## + Years_Since_Last_Promotion 1      11065 11089
## + Marital_Status         2      11063 11089
## + Overtime                1      11065 11089
## + Absenteeism             1      11065 11089
## - Work_Environment_Satisfaction 1      11069 11089
## - Training_Hours_Last_Year 1      11070 11090
## - Job_Involvement        1      11070 11090
## + Department              4      11062 11092
## + Job_Role                3      11064 11092
##
## Step:  AIC=11086.32
## Performance ~ Age + Gender + Job_Level + Training_Hours_Last_Year +
##      Work_Environment_Satisfaction + Relationship_with_Manager +
##      Job_Involvement + Distance_From_Home + Attrition
##
##
##              Df Deviance   AIC
## - Job_Level              1      11068 11086

```

```

## - Gender 1 11068 11086
## - Age 1 11068 11086
## - Distance_From_Home 1 11068 11086
## <none> 11066 11086
## - Relationship_with_Manager 1 11069 11087
## + Years_at_Company 1 11065 11087
## - Attrition 1 11069 11087
## + Project_Count 1 11066 11088
## + Average_Hours_Worked_Per_Week 1 11066 11088
## + Years_in_Current_Role 1 11066 11088
## + Monthly_Income 1 11066 11088
## + Job_Satisfaction 1 11066 11088
## + Hourly_Rate 1 11066 11088
## + Work_Life_Balance 1 11066 11088
## + Number_of_Companies_Worked 1 11066 11088
## + Years_Since_Last_Promotion 1 11066 11088
## + Marital_Status 2 11064 11088
## + Overtime 1 11066 11088
## + Absenteeism 1 11066 11088
## - Work_Environment_Satisfaction 1 11070 11088
## - Training_Hours_Last_Year 1 11071 11089
## - Job_Involvement 1 11071 11089
## + Department 4 11063 11091
## + Job_Role 3 11065 11091
##
## Step: AIC=11085.52
## Performance ~ Age + Gender + Training_Hours_Last_Year +
Work_Environment_Satisfaction +
## Relationship_with_Manager + Job_Involvement + Distance_From_Home +
## Attrition
##
## Df Deviance AIC
## - Gender 1 11069 11085
## - Age 1 11069 11085
## - Distance_From_Home 1 11069 11085
## <none> 11068 11086
## - Relationship_with_Manager 1 11070 11086
## + Job_Level 1 11066 11086
## + Years_at_Company 1 11066 11086
## - Attrition 1 11071 11087
## + Project_Count 1 11067 11087
## + Average_Hours_Worked_Per_Week 1 11067 11087
## + Years_in_Current_Role 1 11067 11087
## + Monthly_Income 1 11067 11087
## + Job_Satisfaction 1 11067 11087
## + Hourly_Rate 1 11067 11087
## + Work_Life_Balance 1 11067 11087
## + Number_of_Companies_Worked 1 11067 11087
## + Years_Since_Last_Promotion 1 11067 11087
## + Overtime 1 11068 11088

```

```

## + Absenteeism                1    11068 11088
## + Marital_Status              2    11066 11088
## - Work_Environment_Satisfaction 1    11072 11088
## - Training_Hours_Last_Year     1    11072 11088
## - Job_Involvement              1    11072 11088
## + Department                   4    11064 11090
## + Job_Role                     3    11066 11090
##
## Step:  AIC=11084.72
## Performance ~ Age + Training_Hours_Last_Year +
Work_Environment_Satisfaction +
##     Relationship_with_Manager + Job_Involvement + Distance_From_Home +
##     Attrition
##
##
##              Df Deviance   AIC
## - Age                1    11070 11084
## - Distance_From_Home 1    11070 11084
## <none>                11069 11085
## - Relationship_with_Manager 1    11071 11085
## + Gender              1    11068 11086
## + Job_Level            1    11068 11086
## + Years_at_Company     1    11068 11086
## - Attrition            1    11072 11086
## + Project_Count        1    11068 11086
## + Average_Hours_Worked_Per_Week 1    11068 11086
## + Years_in_Current_Role 1    11068 11086
## + Job_Satisfaction     1    11068 11086
## + Monthly_Income       1    11068 11086
## + Hourly_Rate          1    11068 11086
## + Work_Life_Balance     1    11069 11087
## + Number_of_Companies_Worked 1    11069 11087
## + Years_Since_Last_Promotion 1    11069 11087
## + Absenteeism          1    11069 11087
## + Overtime             1    11069 11087
## + Marital_Status       2    11067 11087
## - Work_Environment_Satisfaction 1    11073 11087
## - Training_Hours_Last_Year 1    11073 11087
## - Job_Involvement      1    11073 11087
## + Department           4    11066 11090
## + Job_Role             3    11068 11090
##
## Step:  AIC=11083.95
## Performance ~ Training_Hours_Last_Year + Work_Environment_Satisfaction +
##     Relationship_with_Manager + Job_Involvement + Distance_From_Home +
##     Attrition
##
##
##              Df Deviance   AIC
## - Distance_From_Home 1    11071 11083
## <none>                11070 11084
## - Relationship_with_Manager 1    11073 11085

```

```

## + Age 1 11069 11085
## + Gender 1 11069 11085
## + Job_Level 1 11069 11085
## + Years_at_Company 1 11069 11085
## - Attrition 1 11073 11085
## + Project_Count 1 11069 11085
## + Average_Hours_Worked_Per_Week 1 11069 11085
## + Years_in_Current_Role 1 11070 11086
## + Job_Satisfaction 1 11070 11086
## + Monthly_Income 1 11070 11086
## + Hourly_Rate 1 11070 11086
## + Work_Life_Balance 1 11070 11086
## + Number_of_Companies_Worked 1 11070 11086
## + Years_Since_Last_Promotion 1 11070 11086
## + Absenteeism 1 11070 11086
## + Overtime 1 11070 11086
## + Marital_Status 2 11068 11086
## - Work_Environment_Satisfaction 1 11074 11086
## - Training_Hours_Last_Year 1 11074 11086
## - Job_Involvement 1 11074 11086
## + Department 4 11067 11089
## + Job_Role 3 11069 11089
##
## Step: AIC=11083.24
## Performance ~ Training_Hours_Last_Year + Work_Environment_Satisfaction +
## Relationship_with_Manager + Job_Involvement + Attrition
##
## Df Deviance AIC
## <none> 11071 11083
## - Relationship_with_Manager 1 11074 11084
## + Distance_From_Home 1 11070 11084
## + Age 1 11070 11084
## + Gender 1 11070 11084
## + Job_Level 1 11070 11084
## - Attrition 1 11074 11084
## + Years_at_Company 1 11070 11084
## + Project_Count 1 11071 11085
## + Average_Hours_Worked_Per_Week 1 11071 11085
## + Years_in_Current_Role 1 11071 11085
## + Job_Satisfaction 1 11071 11085
## + Monthly_Income 1 11071 11085
## + Hourly_Rate 1 11071 11085
## + Work_Life_Balance 1 11071 11085
## + Number_of_Companies_Worked 1 11071 11085
## + Years_Since_Last_Promotion 1 11071 11085
## + Absenteeism 1 11071 11085
## + Overtime 1 11071 11085
## + Marital_Status 2 11069 11085
## - Work_Environment_Satisfaction 1 11076 11086
## - Job_Involvement 1 11076 11086

```

```

## - Training_Hours_Last_Year      1      11076 11086
## + Department                    4      11068 11088
## + Job_Role                      3      11070 11088

cat("\nFinal model after stepwise selection:\n")

##
## Final model after stepwise selection:

summary(step_model)

##
## Call:
## glm(formula = Performance ~ Training_Hours_Last_Year +
Work_Environment_Satisfaction +
##     Relationship_with_Manager + Job_Involvement + Attrition,
##     family = binomial(link = logit), data = data)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.1815903   0.0988512   1.837   0.0662 .
## Training_Hours_Last_Year -0.0016465   0.0007777  -2.117   0.0342 *
## Work_Environment_Satisfaction 0.0417642   0.0199487   2.094   0.0363 *
## Relationship_with_Manager -0.0327912   0.0200212  -1.638   0.1015
## Job_Involvement -0.0422470   0.0200911  -2.103   0.0355 *
## AttritionYes -0.0957194   0.0562007  -1.703   0.0885 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##     Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11071  on 7994  degrees of freedom
## AIC: 11083
##
## Number of Fisher Scoring iterations: 3

Significant_model <- glm(Performance ~ Training_Hours_Last_Year +
                        Work_Environment_Satisfaction +
                        Job_Involvement,
                        data = train1, family = binomial(link = "logit"))
summary(Significant_model)

##
## Call:
## glm(formula = Performance ~ Training_Hours_Last_Year +
Work_Environment_Satisfaction +
##     Job_Involvement, family = binomial(link = "logit"), data = train1)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)

```



```

## (Intercept)          0.0788169  0.0836515  0.942  0.3461
## Training_Hours_Last_Year -0.0016385  0.0007774 -2.108  0.0351 *
## Work_Environment_Satisfaction 0.0424220  0.0199398  2.128  0.0334 *
## Job_Involvement      -0.0423240  0.0200819 -2.108  0.0351 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11077  on 7996  degrees of freedom
## AIC: 11085
##
## Number of Fisher Scoring iterations: 3

model_interactions <- glm(Performance ~ Training_Hours_Last_Year +
  Work_Environment_Satisfaction * Job_Involvement,
  data = train1, family = binomial)

summary(model_interactions)

##
## Call:
## glm(formula = Performance ~ Training_Hours_Last_Year +
  Work_Environment_Satisfaction *
  Job_Involvement, family = binomial, data = train1)
##
## Coefficients:
##
## Estimate Std. Error z
value
## (Intercept)          0.0534053  0.1406421
0.380
## Training_Hours_Last_Year -0.0016372  0.0007774 -
2.106
## Work_Environment_Satisfaction 0.0525211  0.0491603
1.068
## Job_Involvement      -0.0322270  0.0492064 -
0.655
## Work_Environment_Satisfaction:Job_Involvement -0.0040326  0.0179419 -
0.225
##
## Pr(>|z|)
## (Intercept)          0.7041
## Training_Hours_Last_Year 0.0352 *
## Work_Environment_Satisfaction 0.2854
## Job_Involvement      0.5125
## Work_Environment_Satisfaction:Job_Involvement 0.8222
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)

```

```
##
##      Null deviance: 11090  on 7999  degrees of freedom
## Residual deviance: 11077  on 7995  degrees of freedom
## AIC: 11087
##
## Number of Fisher Scoring iterations: 3

library(pROC)

## Type 'citation("pROC")' for a citation.

##
## Attaching package: 'pROC'

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

# Predict on test set
prob_step_model_test <- predict(Significant_model, newdata = test1, type =
"response")
prob_interaction_test <- predict(model_interactions, newdata = test1, type =
"response")

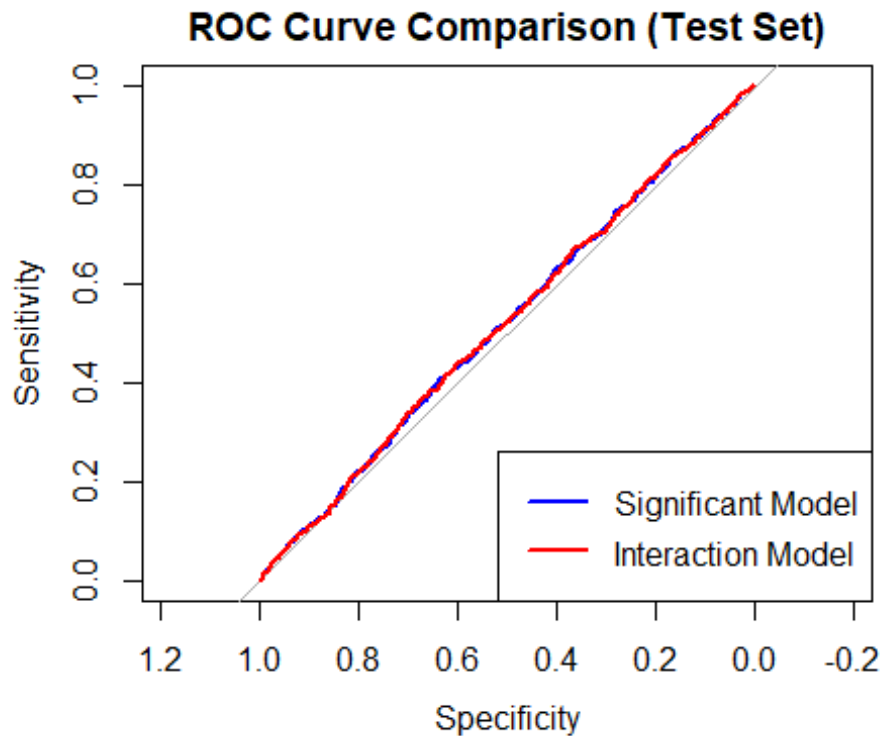
# Generate ROC curves
roc_step_model_test <- roc(test1$Performance, prob_step_model_test)

## Setting levels: control = High, case = Low
## Setting direction: controls < cases

roc_interaction_test <- roc(test1$Performance, prob_interaction_test)

## Setting levels: control = High, case = Low
## Setting direction: controls < cases

# Plot
plot(roc_step_model_test, col = "blue", main = "ROC Curve Comparison (Test
Set)")
lines(roc_interaction_test, col = "red")
legend("bottomright", legend = c("Significant Model", "Interaction Model"),
      col = c("blue", "red"), lwd = 2)
```



```
# Print AUCs
auc(roc_step_model_test)

## Area under the curve: 0.5177

auc(roc_interaction_test)

## Area under the curve: 0.518

library(car)
cat("\nVariance Inflation Factors (VIF):\n")

##
## Variance Inflation Factors (VIF):

##
## Variance Inflation Factors (VIF):
#Check the multicollinearity
print(vif(Significant_model))

##      Training_Hours_Last_Year Work_Environment_Satisfaction
##                1.000231                1.000312
##           Job_Involvement
##                1.000220

exp(coef(Significant_model))

##              (Intercept)      Training_Hours_Last_Year
##                1.082062                0.9983629
```

```

## Work_Environment_Satisfaction          Job_Involvement
##                               1.0433347          0.9585591

#fit <- model2$fitted

#Attrion_num<-as.numeric(as.character(Employee_Data$Attrition))

#r <- (Employee_Data$Attrion_num - fit)/(sqrt(fit*(1-fit))) # Pearson
residuals
#sum(r^2) # Pearson Chi-Square statistic

library(ResourceSelection)

## ResourceSelection 0.3-6    2023-06-27

#Goodness of fit

#Null hypothesis (H0): The model fits the data well.
#Alternative hypothesis (Ha): The model does not fit the data well.
hoslem <- hoslem.test(as.numeric(train1$Performance)-1,
fitted(Significant_model))
print(hoslem)

##
## Hosmer and Lemeshow goodness of fit (GOF) test
##
## data: as.numeric(train1$Performance) - 1, fitted(Significant_model)
## X-squared = 3.9359, df = 8, p-value = 0.8629

#Model Evaluation
predictions <- predict(Significant_model, newdata = test1, type = "response")
print(predictions)

##          1          2          3          4          5          6          7
8
## 0.5053961 0.4899660 0.4968591 0.4861433 0.5042322 0.4695014 0.4712231
0.4956548
##          9         10         11         12         13         14         15
16
## 0.4829429 0.4850537 0.4985466 0.4698851 0.5240820 0.5036641 0.5122638
0.4727674
##         17         18         19         20         21         22         23
24
## 0.5201073 0.4997063 0.5116077 0.5343317 0.5070545 0.4879434 0.4809867
0.5143105
##         25         26         27         28         29         30         31
32
## 0.4727186 0.4613927 0.5065557 0.4825985 0.5164502 0.4832787 0.5327006
0.5177222
##         33         34         35         36         37         38         39
40

```

0.5176287 0.5143349 0.4940856 0.4866261 0.5250372 0.4703621 0.5087617
0.5213585
41 42 43 44 45 46 47
48
0.5196293 0.5030523 0.5444639 0.4875786 0.4719707 0.5334718 0.4883082
0.4951560
49 50 51 52 53 54 55
56
0.4866462 0.4899459 0.5037333 0.5200628 0.5160611 0.5136300 0.5094669
0.4793953
57 58 59 60 61 62 63
64
0.4943325 0.4818494 0.4748300 0.4960644 0.5038024 0.4841460 0.4993168
0.5139702
65 66 67 68 69 70 71
72
0.5156764 0.4830076 0.5046662 0.5177222 0.4973624 0.4874894 0.4780796
0.4687742
73 74 75 76 77 78 79
80
0.4931528 0.4788484 0.4617755 0.5172195 0.5078534 0.5115386 0.5033683
0.5159028
81 82 83 84 85 86 87
88
0.4533044 0.5001850 0.5160165 0.5147197 0.4840971 0.4992520 0.5233781
0.5278721
89 90 91 92 93 94 95
96
0.5038024 0.4923092 0.4752185 0.4792817 0.4699540 0.4830076 0.5237178
0.4993658
97 98 99 100 101 102 103
104
0.5073503 0.5021640 0.5069407 0.4821002 0.5025045 0.5028449 0.5298444
0.5042120
105 106 107 108 109 110 111
112
0.4988625 0.4613927 0.5061663 0.5412110 0.4985912 0.4944707 0.5197919
0.4911253
113 114 115 116 117 118 119
120
0.5144486 0.5065312 0.5041429 0.5144486 0.5054609 0.4760559 0.5044833
0.4480312
121 122 123 124 125 126 127
128
0.4752185 0.5295296 0.4942878 0.4997308 0.5103060 0.4886931 0.5020949
0.4955612
129 130 131 132 133 134 135
136
0.5037578 0.5028694 0.5185159 0.5012310 0.4751940 0.4585194 0.5022331
0.4602157

##	137	138	139	140	141	142	143
144							
##	0.4822584	0.4956303	0.4825985	0.5045971	0.5229694	0.5045525	0.5171304
0.4923092							
##	145	146	147	148	149	150	151
152							
##	0.5327250	0.5074395	0.4768733	0.5045078	0.4899012	0.5063045	0.4866016
0.5127869							
##	153	154	155	156	157	158	159
160							
##	0.5069854	0.4780551	0.4825295	0.5266713	0.5318403	0.5123574	0.4939229
0.4826676							
##	161	162	163	164	165	166	167
168							
##	0.5033237	0.5127667	0.5253078	0.4984328	0.4699984	0.4906511	0.5160410
0.4830076							
##	169	170	171	172	173	174	175
176							
##	0.4898278	0.4968390	0.4780307	0.4968836	0.5099210	0.4904000	0.5274881
0.5062800							
##	177	178	179	180	181	182	183
184							
##	0.5128358	0.4814403	0.4830076	0.5004809	0.5152671	0.4980188	0.4918997
0.5119681							
##	185	186	187	188	189	190	191
192							
##	0.4712231	0.4808486	0.4944016	0.4906021	0.5302526	0.5016607	0.5030034
0.5351714							
##	193	194	195	196	197	198	199
200							
##	0.4952251	0.4744458	0.5091957	0.5127869	0.4967943	0.5154936	0.5289836
0.4723791							
##	201	202	203	204	205	206	207
208							
##	0.5126285	0.4703176	0.5351026	0.5372087	0.5074439	0.5391761	0.4904000
0.5197429							
##	209	210	211	212	213	214	215
216							
##	0.4629974	0.4960644	0.4938783	0.4739439	0.5229449	0.5128113	0.4695903
0.4804640							
##	217	218	219	220	221	222	223
224							
##	0.5156764	0.5000468	0.4702688	0.4796216	0.4760803	0.4976783	0.5083277
0.5229694							
##	225	226	227	228	229	230	231
232							
##	0.5028896	0.4585437	0.5164257	0.5229248	0.5473071	0.4967943	0.4891025
0.5094914							
##	233	234	235	236	237	238	239
240							

0.4887377 0.4752185 0.4959506 0.5106507 0.5135811 0.4883283 0.5110601
0.4911297
241 242 243 244 245 246 247
248
0.5497417 0.4879880 0.4972687 0.5209252 0.4802132 0.4723346 0.4683662
0.4822584
249 250 251 252 253 254 255
256
0.5049174 0.4903064 0.5046216 0.4805331 0.4910605 0.5368013 0.5017789
0.4936516
257 258 259 260 261 262 263
264
0.4899459 0.4976337 0.4973624 0.4972932 0.5005010 0.4690933 0.4959261
0.4927432
265 266 267 268 269 270 271
272
0.5232646 0.5078289 0.5229248 0.4977475 0.4924230 0.4944952 0.4988380
0.4695458
273 274 275 276 277 278 279
280
0.4938783 0.4780997 0.5266713 0.4593331 0.5062354 0.5032545 0.4755782
0.5099455
281 282 283 284 285 286 287
288
0.5323171 0.4723590 0.4822584 0.5090329 0.4829185 0.4920379 0.4985667
0.5489304
289 290 291 292 293 294 295
296
0.5033438 0.5073748 0.4740129 0.4878988 0.5081693 0.5067141 0.5071035
0.4935825
297 298 299 300 301 302 303
304
0.4836879 0.5098965 0.4764200 0.5050758 0.5119436 0.5054653 0.5163566
0.5046216
305 306 307 308 309 310 311
312
0.5070991 0.5033482 0.5017544 0.4984976 0.4610299 0.5054609 0.4948803
0.5293918
313 314 315 316 317 318 319
320
0.5074395 0.5176977 0.4997509 0.4984328 0.4878297 0.5083277 0.4956303
0.4903553
321 322 323 324 325 326 327
328
0.5124020 0.5184713 0.4956548 0.5094669 0.5053471 0.4959261 0.4715868
0.4593574
329 330 331 332 333 334 335
336
0.4975646 0.4850781 0.4659191 0.5012310 0.4650593 0.5197919 0.4805331
0.5065758

##	337	338	339	340	341	342	343
344							
##	0.5061216	0.5066205	0.5156318	0.5102858	0.4914657	0.4977720	0.4952452
0.4924230							
##	345	346	347	348	349	350	351
352							
##	0.5009596	0.5044833	0.4956793	0.5164502	0.4939229	0.5245596	0.5330640
0.5287131							
##	353	354	355	356	357	358	359
360							
##	0.4955612	0.4980433	0.5103103	0.4916039	0.5106708	0.5034173	0.5314768
0.5184958							
##	361	362	363	364	365	366	367
368							
##	0.5140393	0.4654670	0.5241955	0.5444639	0.4828940	0.5303459	0.5262140
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##	369	370	371	372	373	374	375
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##	0.4899012	0.4952899	0.4830076	0.5131962	0.4796706	0.5058748	0.4898523
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##	377	378	379	380	381	382	383
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##	0.4561482	0.5090329	0.4622515	0.4902617	0.5008905	0.5091266	0.5005702
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##	385	386	387	388	389	390	391
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##	0.4984774	0.4808486	0.4975891	0.4723791	0.4907158	0.4826676	0.5180622
0.5012511							
##	393	394	395	396	397	398	399
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##	0.5042567	0.5286643	0.4769423	0.5185159	0.4976783	0.4577745	0.5071035
0.5131761							
##	401	402	403	404	405	406	407
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##	0.5038226	0.4837569	0.5189249	0.4699296	0.4613683	0.5119235	0.5016406
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##	409	410	411	412	413	414	415
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##	0.4609855	0.5171304	0.4956303	0.4601470	0.5127423	0.5355789	0.5234025
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##	417	418	419	420	421	422	423
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##	0.4911253	0.4727674	0.4993413	0.4980433	0.5087372	0.5278276	0.4915838
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##	425	426	427	428	429	430	431
432							
##	0.5025491	0.4854183	0.4932219	0.4748099	0.4866217	0.5193584	0.4918306
0.5388377							
##	433	434	435	436	437	438	439
440							

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##	545	546	547	548	549	550	551
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##	0.5034173	0.5122192	0.5187667	0.4698851	0.4793508	0.4853737	0.5229248
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##	553	554	555	556	557	558	559
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##	561	562	563	564	565	566	567
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##	0.4598087	0.4714935	0.4911253	0.4935133	0.5481189	0.4935868	0.4719952
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##	577	578	579	580	581	582	583
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##	0.5395832	0.4792127	0.4944952	0.5127624	0.5062109	0.5005255	0.4859413
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##	0.5204918	0.4800996	0.5154936	0.5124265	0.4736044	0.5172440	0.5045971
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##	0.4914902	0.4662824	0.5347394	0.4789418	0.4565547	0.5177222	0.5057812
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##	617	618	619	620	621	622	623
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##	633	634	635	636	637	638	639
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##	0.5172685	0.4480312	0.5179486	0.5197919	0.4874448	0.5041184	0.5095360
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##	761	762	763	764	765	766	767
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##	0.5012310	0.5193340	0.4674815	0.4798043	0.4924028	0.4951070	0.4610299
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##	801	802	803	804	805	806	807
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##	1033	1034	1035	1036	1037	1038	1039
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0.4678894 0.4871693 0.4834413 0.4605784 0.5054163 0.5062153 0.4760113
0.4886240
1297 1298 1299 1300 1301 1302 1303
1304
0.5033482 0.5185403 0.4940611 0.5368013 0.5368013 0.4968635 0.5025981
0.4797597
1305 1306 1307 1308 1309 1310 1311
1312
0.4798043 0.5107154 0.5163811 0.4691822 0.4763956 0.5041429 0.5152225
0.5126285
1313 1314 1315 1316 1317 1318 1319
1320
0.5173375 0.5144731 0.4935868 0.4939920 0.4679582 0.4691378 0.5395832
0.5273948
1321 1322 1323 1324 1325 1326 1327
1328
0.5201318 0.5090574 0.5132452 0.4849155 0.5229694 0.4842151 0.5359864
0.4768043
1329 1330 1331 1332 1333 1334 1335
1336
0.4915838 0.4769423 0.5013448 0.5183577 0.4890334 0.4869620 0.4967453
0.4964294

##	1337	1338	1339	1340	1341	1342	1343
1344							
##	0.4842597	0.5148133	0.5017990	0.4947219	0.4829429	0.5156318	0.5041429
0.4992477							
##	1345	1346	1347	1348	1349	1350	1351
1352							
##	0.4793508	0.5225606	0.5123574	0.4989072	0.5041630	0.5083522	0.4784884
0.5066939							
##	1353	1354	1355	1356	1357	1358	1359
1360							
##	0.5209252	0.4752185	0.5098764	0.5005702	0.4964049	0.5090128	0.4802132
0.5107154							
##	1361	1362	1363	1364	1365	1366	1367
1368							
##	0.5065758	0.4634048	0.4891025	0.5071035	0.4883082	0.4793062	0.5066003
0.4922401							
##	1369	1370	1371	1372	1373	1374	1375
1376							
##	0.4545228	0.5179932	0.4863060	0.5086235	0.4646517	0.5290725	0.5331328
0.5008905							
##	1377	1378	1379	1380	1381	1382	1383
1384							
##	0.4915593	0.4989562	0.4912189	0.5040982	0.5359177	0.5124020	0.4492471
0.4797597							
##	1385	1386	1387	1388	1389	1390	1391
1392							
##	0.5004319	0.5054653	0.4524924	0.4944016	0.5371400	0.5073748	0.5204717
0.4735354							
##	1393	1394	1395	1396	1397	1398	1399
1400							
##	0.5144241	0.5151981	0.4878542	0.4667346	0.4642886	0.5050758	0.4772131
0.5322927							
##	1401	1402	1403	1404	1405	1406	1407
1408							
##	0.5079182	0.4968390	0.5159028	0.4883283	0.5497417	0.4731959	0.4756472
0.5046662							
##	1409	1410	1411	1412	1413	1414	1415
1416							
##	0.5200628	0.5193584	0.5045726	0.5179932	0.4936314	0.5079225	0.5192649
0.4708148							
##	1417	1418	1419	1420	1421	1422	1423
1424							
##	0.4918997	0.4927634	0.4879434	0.4589949	0.5033683	0.5383618	0.5460890
0.4915348							
##	1425	1426	1427	1428	1429	1430	1431
1432							
##	0.5335162	0.4824604	0.4476260	0.5132452	0.5028449	0.4679138	0.4804842
0.4910605							
##	1433	1434	1435	1436	1437	1438	1439
1440							

0.4837814 0.4849645 0.4812576 0.5205407 0.5053025 0.5062153 0.4964985
0.4789864
1441 1442 1443 1444 1445 1446 1447
1448
0.4816667 0.5371400 0.5139947 0.5351470 0.5310243 0.5201318 0.4837569
0.5384306
1449 1450 1451 1452 1453 1454 1455
1456
0.4711341 0.4993168 0.5213096 0.5090776 0.5050067 0.4707258 0.5082831
0.5187667
1457 1458 1459 1460 1461 1462 1463
1464
0.4804842 0.5106708 0.4923092 0.4901926 0.5339240 0.4931729 0.5318403
0.4813511
1465 1466 1467 1468 1469 1470 1471
1472
0.5066450 0.5384062 0.5025736 0.4992967 0.4951516 0.5061216 0.5021395
0.4727186
1473 1474 1475 1476 1477 1478 1479
1480
0.4723590 0.5302971 0.4830767 0.5372087 0.5319091 0.4890579 0.4691822
0.5245797
1481 1482 1483 1484 1485 1486 1487
1488
0.5363696 0.4891716 0.4858967 0.4727430 0.4818494 0.5008214 0.4662580
0.5233537
1489 1490 1491 1492 1493 1494 1495
1496
0.5444639 0.5086235 0.4979742 0.4928570 0.4634048 0.4796461 0.5067386
0.5379990
1497 1498 1499 1500 1501 1502 1503
1504
0.4813022 0.5032545 0.5403972 0.4972932 0.5306607 0.4936516 0.4927634
0.5343317
1505 1506 1507 1508 1509 1510 1511
1512
0.4801442 0.5123329 0.4906266 0.4820513 0.4846243 0.5172886 0.5199937
0.5160165
1513 1514 1515 1516 1517 1518 1519
1520
0.5030034 0.5351470 0.4825985 0.4911052 0.5160165 0.4698851 0.5017544
0.4858276
1521 1522 1523 1524 1525 1526 1527
1528
0.5025491 0.5126732 0.5197429 0.4634048 0.4939719 0.5164257 0.5148133
0.4859413
1529 1530 1531 1532 1533 1534 1535
1536
0.5054609 0.4928325 0.5236733 0.5152671 0.4964495 0.4870311 0.4919242
0.4924028

##	1537	1538	1539	1540	1541	1542	1543
1544							
##	0.5087862	0.5033683	0.4910605	0.5029342	0.5071236	0.4989317	0.5150844
0.5224470							
##	1545	1546	1547	1548	1549	1550	1551
1552							
##	0.5048929	0.4789864	0.4940165	0.5026427	0.5070344	0.5098519	0.4557418
0.5017544							
##	1553	1554	1555	1556	1557	1558	1559
1560							
##	0.4723346	0.4943368	0.4808730	0.5205407	0.5070300	0.5440575	0.4955857
0.4638123							
##	1561	1562	1563	1564	1565	1566	1567
1568							
##	0.4861923	0.5157008	0.5061216	0.5069652	0.5009596	0.4768532	0.4813712
0.4894673							
##	1569	1570	1571	1572	1573	1574	1575
1576							
##	0.5071481	0.4846243	0.5032992	0.5265334	0.5017990	0.4898278	0.5070545
0.4853248							
##	1577	1578	1579	1580	1581	1582	1583
1584							
##	0.5074395	0.4703377	0.5185159	0.4846444	0.4936516	0.5041630	0.5102412
0.5057366							
##	1585	1586	1587	1588	1589	1590	1591
1592							
##	0.5070099	0.5295052	0.5237624	0.5070344	0.4955166	0.5013448	0.5083320
0.5040737							
##	1593	1594	1595	1596	1597	1598	1599
1600							
##	0.5278477	0.5204473	0.4711542	0.4939474	0.4838951	0.4804151	0.5164502
0.4748300							
##	1601	1602	1603	1604	1605	1606	1607
1608							
##	0.4980880	0.4865527	0.5388134	0.4951560	0.4800551	0.5030034	0.4951070
0.5375473							
##	1609	1610	1611	1612	1613	1614	1615
1616							
##	0.5177222	0.4629974	0.5140149	0.5059195	0.5118990	0.5169039	0.4960889
0.5412110							
##	1617	1618	1619	1620	1621	1622	1623
1624							
##	0.5238113	0.4812821	0.4944952	0.4989317	0.5135163	0.4800752	0.5053270
0.5057366							
##	1625	1626	1627	1628	1629	1630	1631
1632							
##	0.5199937	0.4936070	0.4756027	0.5143349	0.4606228	0.4988870	0.4944952
0.4702688							
##	1633	1634	1635	1636	1637	1638	1639
1640							

0.4861923 0.4837613 0.4679582 0.5094669 0.5208562 0.5265334 0.5293918
0.5066450
1641 1642 1643 1644 1645 1646 1647
1648
0.4772821 0.5148579 0.4862124 0.4883082 0.5400589 0.5034375 0.4816422
0.4792127
1649 1650 1651 1652 1653 1654 1655
1656
0.5106262 0.4817113 0.5086926 0.5041630 0.5078534 0.5078980 0.4899459
0.4980923
1657 1658 1659 1660 1661 1662 1663
1664
0.4931773 0.5093978 0.5151535 0.4968145 0.4976092 0.5179486 0.4601470
0.4804640
1665 1666 1667 1668 1669 1670 1671
1672
0.4849846 0.4793062 0.5029386 0.5371400 0.4977028 0.5030077 0.5106708
0.4992520
1673 1674 1675 1676 1677 1678 1679
1680
0.5241955 0.4707014 0.5172685 0.5367770 0.5310243 0.4736288 0.4832787
0.4711786
1681 1682 1683 1684 1685 1686 1687
1688
0.4956303 0.4939920 0.4748300 0.5118544 0.4919688 0.5314323 0.5078980
0.5331328
1689 1690 1691 1692 1693 1694 1695
1696
0.4979742 0.5140149 0.5290969 0.4581369 0.5302770 0.4736733 0.5062844
0.4545228
1697 1698 1699 1700 1701 1702 1703
1704
0.4793508 0.5070344 0.4904000 0.4712231 0.4984774 0.4904000 0.4887622
0.4870109
1705 1706 1707 1708 1709 1710 1711
1712
0.4866261 0.4951761 0.5197228 0.4867599 0.5061908 0.4812821 0.5229694
0.5026427
1713 1714 1715 1716 1717 1718 1719
1720
0.4846243 0.4956793 0.4690445 0.4992477 0.5448703 0.4857585 0.5086033
0.4813511
1721 1722 1723 1724 1725 1726 1727
1728
0.4906021 0.5042120 0.5017097 0.4902372 0.5144486 0.5030523 0.5315012
0.5314323
1729 1730 1731 1732 1733 1734 1735
1736
0.4589949 0.4989072 0.5343561 0.5094871 0.5021885 0.4708148 0.4862614
0.5295052

##	1737	1738	1739	1740	1741	1742	1743
1744							
##	0.5228558	0.5130825	0.5119681	0.5037534	0.5180823	0.4842151	0.5310687
0.5375917							
##	1745	1746	1747	1748	1749	1750	1751
1752							
##	0.4826676	0.4838951	0.5128113	0.4642442	0.5070991	0.4977720	0.4687053
0.5327006							
##	1753	1754	1755	1756	1757	1758	1759
1760							
##	0.4952006	0.4903107	0.5034173	0.4955857	0.5077843	0.5168593	0.4670736
0.4926942							
##	1761	1762	1763	1764	1765	1766	1767
1768							
##	0.4719262	0.4776707	0.4739439	0.4996126	0.4993212	0.5187667	0.4948601
0.5269864							
##	1769	1770	1771	1772	1773	1774	1775
1776							
##	0.4838059	0.4854428	0.4898278	0.4932420	0.4837123	0.5179486	0.5265334
0.5115343							
##	1777	1778	1779	1780	1781	1782	1783
1784							
##	0.5033683	0.4744458	0.4651037	0.4817156	0.5152671	0.4862614	0.4931729
0.5054609							
##	1785	1786	1787	1788	1789	1790	1791
1792							
##	0.5159474	0.4830767	0.4736288	0.4715624	0.4796216	0.5241265	0.4589262
0.5041674							
##	1793	1794	1795	1796	1797	1798	1799
1800							
##	0.4927432	0.4907158	0.4621828	0.5028449	0.5135163	0.5155828	0.5069854
0.4800752							
##	1801	1802	1803	1804	1805	1806	1807
1808							
##	0.4736288	0.4903308	0.5074640	0.4867153	0.5412110	0.5171994	0.4979742
0.4955655							
##	1809	1810	1811	1812	1813	1814	1815
1816							
##	0.4808486	0.5148334	0.5099700	0.4866217	0.4760803	0.4980678	0.4776218
0.5041184							
##	1817	1818	1819	1820	1821	1822	1823
1824							
##	0.4869865	0.4984529	0.4931974	0.4768532	0.5009798	0.5302526	0.5016607
0.4837569							
##	1825	1826	1827	1828	1829	1830	1831
1832							
##	0.5298444	0.4706770	0.4585194	0.4804151	0.4821247	0.5127178	0.4818048
0.5042365							
##	1833	1834	1835	1836	1837	1838	1839
1840							

0.4946974 0.4752386 0.5371400 0.5090329 0.4658746 0.5020704 0.5016161
0.4760803
1841 1842 1843 1844 1845 1846 1847
1848
0.5005010 0.5021640 0.5044833 0.5127178 0.4855320 0.4732649 0.4979987
0.5143105
1849 1850 1851 1852 1853 1854 1855
1856
0.4845552 0.4927634 0.5359177 0.4914211 0.4948357 0.4609612 0.5274393
0.5237423
1857 1858 1859 1860 1861 1862 1863
1864
0.4928325 0.4670980 0.5086724 0.5071481 0.4887176 0.4936516 0.4784395
0.4985466
1865 1866 1867 1868 1869 1870 1871
1872
0.5062109 0.4675503 0.4788484 0.4981124 0.4804842 0.4585194 0.5164257
0.5225850
1873 1874 1875 1876 1877 1878 1879
1880
0.4683217 0.4853492 0.5299133 0.5432445 0.5040982 0.5346949 0.5111493
0.4837613
1881 1882 1883 1884 1885 1886 1887
1888
0.4964539 0.5054408 0.4918795 0.5135811 0.5050067 0.4904000 0.5032545
0.5041184
1889 1890 1891 1892 1893 1894 1895
1896
0.5065758 0.4926942 0.5095360 0.5269864 0.5164948 0.5036641 0.5270553
0.4886240
1897 1898 1899 1900 1901 1902 1903
1904
0.5001605 0.5090329 0.5318403 0.5151736 0.4977475 0.4904000 0.4862815
0.4845753
1905 1906 1907 1908 1909 1910 1911
1912
0.4911253 0.5224916 0.4822584 0.4764445 0.4776218 0.5030034 0.5237868
0.5306851
1913 1914 1915 1916 1917 1918 1919
1920
0.5093978 0.4805776 0.5004117 0.5323171 0.5139012 0.4821693 0.4878542
0.4923582
1921 1922 1923 1924 1925 1926 1927
1928
0.5176531 0.5452766 0.4744214 0.5184713 0.4699095 0.5049822 0.4957240
0.4757162
1929 1930 1931 1932 1933 1934 1935
1936
0.5028694 0.5189739 0.5175395 0.5181557 0.4801687 0.5131271 0.4813022
0.4825295

```
##      1937      1938      1939      1940      1941      1942      1943
1944
## 0.5310243 0.4727430 0.5452766 0.4826230 0.5106953 0.4808486 0.5143105
0.4512749
##      1945      1946      1947      1948      1949      1950      1951
1952
## 0.4699984 0.4561482 0.5001605 0.4955166 0.4963602 0.5091467 0.4952452
0.5138565
##      1953      1954      1955      1956      1957      1958      1959
1960
## 0.5005457 0.5399903 0.4841661 0.4638367 0.4944952 0.4959506 0.4894183
0.4849889
##      1961      1962      1963      1964      1965      1966      1967
1968
## 0.5099009 0.4801442 0.5059195 0.5172641 0.5185403 0.5058748 0.5152427
0.4617999
##      1969      1970      1971      1972      1973      1974      1975
1976
## 0.4820513 0.4924475 0.4891270 0.5192448 0.4992030 0.5228558 0.5163566
0.4816667
##      1977      1978      1979      1980      1981      1982      1983
1984
## 0.5266469 0.5050312 0.5253524 0.5079225 0.4791883 0.4907158 0.4890334
0.4488417
##      1985      1986      1987      1988      1989      1990      1991
1992
## 0.5114896 0.5147197 0.5318847 0.4887622 0.5030279 0.4992275 0.4846689
0.4939474
##      1993      1994      1995      1996      1997      1998      1999
2000
## 0.4968145 0.5008415 0.4715179 0.4996616 0.5001404 0.5270108 0.5074194
0.5293918
```

```
predictions_category <- ifelse(predictions >= 0.5, "1","0")
print( predictions_category )
```

```
##      1      2      3      4      5      6      7      8      9     10     11     12     13     14     15
16
## "1"  "0"  "0"  "0"  "1"  "0"  "0"  "0"  "0"  "0"  "0"  "0"  "1"  "1"  "1"
"0"
## 17     18     19     20     21     22     23     24     25     26     27     28     29     30     31
32
## "1"  "0"  "1"  "1"  "1"  "0"  "0"  "1"  "0"  "0"  "1"  "0"  "1"  "0"  "1"
"1"
## 33     34     35     36     37     38     39     40     41     42     43     44     45     46     47
48
## "1"  "1"  "0"  "0"  "1"  "0"  "1"  "1"  "1"  "1"  "1"  "0"  "0"  "1"  "0"
"0"
## 49     50     51     52     53     54     55     56     57     58     59     60     61     62     63
64
```


#	0	0	1	1	1	1	1	0	0	0	0	0	1	0	0
1															
##	65	66	67	68	69	70	71	72	73	74	75	76	77	78	79
80															
##	1	0	1	1	0	0	0	0	0	0	0	1	1	1	1
1															
##	81	82	83	84	85	86	87	88	89	90	91	92	93	94	95
96															
##	0	1	1	1	0	0	1	1	1	0	0	0	0	0	1
0															
##	97	98	99	100	101	102	103	104	105	106	107	108	109	110	111
112															
##	1	1	1	0	1	1	1	1	0	0	1	1	0	0	1
0															
##	113	114	115	116	117	118	119	120	121	122	123	124	125	126	127
128															
##	1	1	1	1	1	0	1	0	0	1	0	0	1	0	1
0															
##	129	130	131	132	133	134	135	136	137	138	139	140	141	142	143
144															
##	1	1	1	1	0	0	1	0	0	0	0	1	1	1	1
0															
##	145	146	147	148	149	150	151	152	153	154	155	156	157	158	159
160															
##	1	1	0	1	0	1	0	1	1	0	0	1	1	1	0
0															
##	161	162	163	164	165	166	167	168	169	170	171	172	173	174	175
176															
##	1	1	1	0	0	0	1	0	0	0	0	0	1	0	1
1															
##	177	178	179	180	181	182	183	184	185	186	187	188	189	190	191
192															
##	1	0	0	1	1	0	0	1	0	0	0	0	1	1	1
1															
##	193	194	195	196	197	198	199	200	201	202	203	204	205	206	207
208															
##	0	0	1	1	0	1	1	0	1	0	1	1	1	1	0
1															
##	209	210	211	212	213	214	215	216	217	218	219	220	221	222	223
224															
##	0	0	0	0	1	1	0	0	1	1	0	0	0	0	1
1															
##	225	226	227	228	229	230	231	232	233	234	235	236	237	238	239
240															
##	1	0	1	1	1										

## 272 ## "0"	257	258	259	260	261	262	263	264	265	266	267	268	269	270	271
## "0"	"0"	"0"	"0"	"0"	"1"	"0"	"0"	"0"	"1"	"1"	"1"	"0"	"0"	"0"	"0"
## 288 ## "1"	273	274	275	276	277	278	279	280	281	282	283	284	285	286	287
## "1"	"0"	"0"	"1"	"0"	"1"	"1"	"0"	"1"	"1"	"0"	"0"	"1"	"0"	"0"	"0"
## 304 ## "1"	289	290	291	292	293	294	295	296	297	298	299	300	301	302	303
## "1"	"1"	"1"	"0"	"0"	"1"	"1"	"1"	"0"	"0"	"1"	"0"	"1"	"1"	"1"	"1"
## 320 ## "0"	305	306	307	308	309	310	311	312	313	314	315	316	317	318	319
## "0"	"1"	"1"	"1"	"0"	"0"	"1"	"0"	"1"	"1"	"1"	"0"	"0"	"0"	"1"	"0"
## 336 ## "1"	321	322	323	324	325	326	327	328	329	330	331	332	333	334	335
## "1"	"1"	"1"	"0"	"1"	"1"	"0"	"0"	"0"	"0"	"0"	"0"	"1"	"0"	"1"	"0"
## 352 ## "1"	337	338	339	340	341	342	343	344	345	346	347	348	349	350	351
## "1"	"1"	"1"	"1"	"1"	"0"	"0"	"0"	"0"	"1"	"1"	"0"	"1"	"0"	"1"	"1"
## 368 ## "0"	353	354	355	356	357	358	359	360	361	362	363	364	365	366	367
## "0"	"0"	"0"	"1"	"0"	"1"	"1"	"1"	"1"	"1"	"0"	"1"	"1"	"0"	"1"	"1"
## 384 ## "0"	369	370	371	372	373	374	375	376	377	378	379	380	381	382	383
## "0"	"0"	"0"	"0"	"1"	"0"	"1"	"0"	"0"	"0"	"1"	"0"	"0"	"1"	"1"	"1"
## 400 ## "1"	385	386	387	388	389	390	391	392	393	394	395	396	397	398	399
## "1"	"0"	"0"	"0"	"0"	"0"	"0"	"1"	"1"	"1"	"1"	"0"	"1"	"0"	"0"	"1"
## 416 ## "1"	401	402	403	404	405	406	407	408	409	410	411	412	413	414	415
## "1"	"1"	"0"	"1"	"0"	"0"	"1"	"1"	"0"	"0"	"1"	"0"	"0"	"1"	"1"	"1"
## 432 ## "1"	417	418	419	420	421	422	423	424	425	426	427	428	429	430	431
## "1"	"0"	"0"	"0"	"0"	"1"	"1"	"0"	"0"	"1"	"0"	"0"	"0"	"0"	"1"	"0"
## 448 ## "1"	433	434	435	436	437	438	439	440	441	442	443	444	445	446	447
## "1"	"0"	"0"	"0"	"0"	"1"	"0"	"0"	"0"	"0"	"1"	"0"	"1"	"0"	"0"	"1"
## 464	449	450	451	452	453	454	455	456	457	458	459	460	461	462	463

#	1	0	0	0	1	0	1	0	0	0	0	0	1	1	0
##	465	466	467	468	469	470	471	472	473	474	475	476	477	478	479
##	0	1	1	1	0	1	1	0	1	1	1	1	1	0	1
#	481	482	483	484	485	486	487	488	489	490	491	492	493	494	495
##	1	0	1	0	1	1	1	1	0	0	0	0	1	0	0
#	497	498	499	500	501	502	503	504	505	506	507	508	509	510	511
##	1	0	0	1	0	1	0	1	0	1	1	1	0	1	1
#	513	514	515	516	517	518	519	520	521	522	523	524	525	526	527
##	1	0	1	1	0	0	0	0	1	1	1	0	0	0	0
#	529	530	531	532	533	534	535	536	537	538	539	540	541	542	543
##	1	0	1	0	0	1	0	1	1	0	0	1	1	1	0
#	545	546	547	548	549	550	551	552	553	554	555	556	557	558	559
##	1	1	1	0	0	0	1	0	1	0	0	0	0	0	1
#	561	562	563	564	565	566	567	568	569	570	571	572	573	574	575
##	0	0	0	1	0	1	0	0	0	0	0	0	1	0	0
#	577	578	579	580	581	582	583	584	585	586	587	588	589	590	591
##	1	0	0	1	1	1	0	0	1	0	1	1	0	1	1
#	593	594	595	596	597	598	599	600	601	602	603	604	605	606	607
##	1	0	0	1	0	0	1	0	1	1	1	0	0	1	0
#	609	610	611	612	613	614	615	616	617	618	619	620	621	622	623
##	0	0	1	0	0	1	1	0	0	1	0	0	1	1	0
#	625	626	627	628	629	630	631	632	633	634	635	636	637	638	639
##	0	0	1	0	0	1	0	1	0	0	0	1	1	0	0
#	641	642	643	644	645	646	647	648	649	650	651	652	653	654	655
##	1	1	1	1	1	0	1	0	1	0	0	1	1	1	0

672	657	658	659	660	661	662	663	664	665	666	667	668	669	670	671
## "1"	"1"	"1"	"0"	"1"	"1"	"0"	"0"	"1"	"1"	"0"	"1"	"1"	"1"	"0"	"1"
688	673	674	675	676	677	678	679	680	681	682	683	684	685	686	687
## "0"	"1"	"0"	"1"	"0"	"1"	"1"	"1"	"1"	"1"	"0"	"1"	"1"	"0"	"0"	"1"
704	689	690	691	692	693	694	695	696	697	698	699	700	701	702	703
## "0"	"0"	"1"	"1"	"0"	"1"	"1"	"1"	"0"	"0"	"1"	"1"	"0"	"1"	"1"	"0"
720	705	706	707	708	709	710	711	712	713	714	715	716	717	718	719
## "1"	"0"	"1"	"0"	"1"	"1"	"0"	"0"	"0"	"1"	"1"	"1"	"1"	"0"	"1"	"1"
736	721	722	723	724	725	726	727	728	729	730	731	732	733	734	735
## "1"	"0"	"0"	"0"	"1"	"0"	"1"	"1"	"1"	"0"	"1"	"1"	"1"	"1"	"0"	"1"
752	737	738	739	740	741	742	743	744	745	746	747	748	749	750	751
## "0"	"0"	"1"	"1"	"1"	"0"	"0"	"1"	"1"	"0"	"1"	"0"	"0"	"1"	"1"	"0"
768	753	754	755	756	757	758	759	760	761	762	763	764	765	766	767
## "1"	"1"	"0"	"1"	"1"	"0"	"1"	"1"	"0"	"0"	"0"	"1"	"1"	"0"	"1"	"1"
784	769	770	771	772	773	774	775	776	777	778	779	780	781	782	783
## "1"	"0"	"1"	"0"	"0"	"1"	"1"	"0"	"0"	"1"	"1"	"0"	"0"	"0"	"0"	"0"
800	785	786	787	788	789	790	791	792	793	794	795	796	797	798	799
## "0"	"1"	"1"	"1"	"0"	"1"	"1"	"1"	"0"	"1"	"0"	"0"	"0"	"1"	"0"	"0"
816	801	802	803	804	805	806	807	808	809	810	811	812	813	814	815
## "0"	"0"	"0"	"1"	"0"	"0"	"1"	"1"	"0"	"1"	"1"	"1"	"0"	"1"	"0"	"1"
832	817	818	819	820	821	822	823	824	825	826	827	828	829	830	831
## "0"	"1"	"0"	"1"	"0"	"1"	"0"	"1"	"0"	"1"	"1"	"1"	"0"	"0"	"1"	"0"
848	833	834	835	836	837	838	839	840	841	842	843	844	845	846	847
## "0"	"0"	"1"	"0"	"1"	"0"	"0"	"1"	"1"	"0"	"0"	"1"	"1"	"1"	"1"	"0"
864	849	850	851	852	853	854	855	856	857	858	859	860	861	862	863

#	"0"	"1"	"1"	"1"	"1"	"0"	"1"	"1"	"0"	"1"	"0"	"0"	"1"	"0"	"1"
## 880	865	866	867	868	869	870	871	872	873	874	875	876	877	878	879
#	"0"	"1"	"1"	"0"	"0"	"0"	"1"	"0"	"0"	"0"	"1"	"0"	"1"	"1"	"0"
#	881	882	883	884	885	886	887	888	889	890	891	892	893	894	895
## 912	"1"	"1"	"1"	"0"	"0"	"0"	"1"	"0"	"0"	"1"	"0"	"1"	"1"	"0"	"0"
#	897	898	899	900	901	902	903	904	905	906	907	908	909	910	911
#	"0"	"1"	"0"	"1"	"1"	"0"	"1"	"0"	"0"	"1"	"0"	"1"	"1"	"1"	"0"
#	913	914	915	916	917	918	919	920	921	922	923	924	925	926	927
#	"0"	"0"	"1"	"1"	"0"	"0"	"1"	"1"	"0"	"0"	"1"	"0"	"0"	"0"	"1"
#	929	930	931	932	933	934	935	936	937	938	939	940	941	942	943
#	"0"	"0"	"1"	"0"	"1"	"1"	"0"	"1"	"0"	"0"	"0"	"0"	"0"	"1"	"0"
#	945	946	947	948	949	950	951	952	953	954	955	956	957	958	959
#	"1"	"1"	"0"	"1"	"0"	"1"	"0"	"1"	"0"	"1"	"1"	"0"	"0"	"1"	"0"
#	961	962	963	964	965	966	967	968	969	970	971	972	973	974	975
#	"1"	"0"	"0"	"1"	"1"	"0"	"0"	"0"	"1"	"1"	"0"	"0"	"0"	"1"	"1"
#	977	978	979	980	981	982	983	984	985	986	987	988	989	990	991
#	"1"	"1"	"1"	"0"	"1"	"1"	"1"	"0"	"1"	"0"	"1"	"0"	"1"	"0"	"1"
#	993	994	995	996	997	998	999	1000	1001	1002	1003	1004	1005	1006	1007
#	"1"	"1"	"0"	"1"	"0"	"0"	"1"	"0"	"1"	"1"	"0"	"1"	"1"	"0"	"1"
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#	"0"	"0"	"0"	"0"	"1"	"1"	"1"	"0"	"1"	"1"	"0"	"0"	"1"	"1"	"1"
#	1025	1026	1027	1028	1029	1030	1031	1032	1033	1034	1035	1036	1037	1038	1039
#	"0"	"1"	"0"	"0"	"0"	"0"	"0"	"0"	"1"	"0"	"1"	"1"	"0"	"1"	"1"
#	1041	1042	1043	1044	1045	1046	1047	1048	1049	1050	1051	1052	1053	1054	1055
#	"0"	"0"	"0"	"1"	"1"	"0"	"0"	"1"	"1"	"0"	"1"	"1"	"0"	"0"	"1"

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## 1057 1058 1059 1060 1061 1062 1063 1064 1065 1066 1067 1068 1069 1070 1071
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## "1" "1" "0" "0" "1" "1" "1" "1" "1" "1" "0" "0" "1" "1" "1"
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## 1073 1074 1075 1076 1077 1078 1079 1080 1081 1082 1083 1084 1085 1086 1087
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## 1089 1090 1091 1092 1093 1094 1095 1096 1097 1098 1099 1100 1101 1102 1103
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1168
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## 1169 1170 1171 1172 1173 1174 1175 1176 1177 1178 1179 1180 1181 1182 1183
1184
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"1"
## 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197 1198 1199
1200
## "0" "1" "1" "0" "0" "0" "1" "0" "0" "0" "0" "0" "1" "0" "1"
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## 1201 1202 1203 1204 1205 1206 1207 1208 1209 1210 1211 1212 1213 1214 1215
1216
## "1" "0" "0" "1" "1" "0" "1" "0" "0" "0" "0" "0" "0" "0" "1"
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## 1217 1218 1219 1220 1221 1222 1223 1224 1225 1226 1227 1228 1229 1230 1231
1232
## "1" "1" "1" "0" "0" "1" "1" "1" "1" "1" "1" "0" "0" "1" "1"
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## 1233 1234 1235 1236 1237 1238 1239 1240 1241 1242 1243 1244 1245 1246 1247
1248
## "1" "1" "1" "1" "0" "1" "1" "0" "0" "1" "1" "0" "1" "1" "0"
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## 1249 1250 1251 1252 1253 1254 1255 1256 1257 1258 1259 1260 1261 1262 1263
1264
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## "0" "0" "1" "0" "1" "0" "1" "1" "1" "0" "1" "1" "0" "0" "1"
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## 1281 1282 1283 1284 1285 1286 1287 1288 1289 1290 1291 1292 1293 1294 1295
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## 1297 1298 1299 1300 1301 1302 1303 1304 1305 1306 1307 1308 1309 1310 1311
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1328
## "1" "1" "0" "0" "0" "0" "1" "1" "1" "1" "1" "0" "1" "0" "1"
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## 1329 1330 1331 1332 1333 1334 1335 1336 1337 1338 1339 1340 1341 1342 1343
1344
## "0" "0" "1" "1" "0" "0" "0" "0" "0" "1" "1" "0" "0" "1" "1"
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## 1345 1346 1347 1348 1349 1350 1351 1352 1353 1354 1355 1356 1357 1358 1359
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## 1361 1362 1363 1364 1365 1366 1367 1368 1369 1370 1371 1372 1373 1374 1375
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## 1377 1378 1379 1380 1381 1382 1383 1384 1385 1386 1387 1388 1389 1390 1391
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## 1393 1394 1395 1396 1397 1398 1399 1400 1401 1402 1403 1404 1405 1406 1407
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## 1409 1410 1411 1412 1413 1414 1415 1416 1417 1418 1419 1420 1421 1422 1423
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## 1425 1426 1427 1428 1429 1430 1431 1432 1433 1434 1435 1436 1437 1438 1439
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## 1441 1442 1443 1444 1445 1446 1447 1448 1449 1450 1451 1452 1453 1454 1455
1456
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## "0" "1" "0" "0" "1" "0" "1" "0" "1" "1" "1" "0" "0" "1" "1"
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## 1585 1586 1587 1588 1589 1590 1591 1592 1593 1594 1595 1596 1597 1598 1599
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## "1" "1" "1" "1" "0" "1" "1" "1" "1" "1" "0" "0" "0" "0" "1"
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## 1601 1602 1603 1604 1605 1606 1607 1608 1609 1610 1611 1612 1613 1614 1615
1616
## "0" "0" "1" "0" "0" "1" "0" "1" "1" "0" "1" "1" "1" "1" "0"
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## 1617 1618 1619 1620 1621 1622 1623 1624 1625 1626 1627 1628 1629 1630 1631
1632
## "1" "0" "0" "0" "1" "0" "1" "1" "1" "0" "0" "1" "0" "0" "0"
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## 1633 1634 1635 1636 1637 1638 1639 1640 1641 1642 1643 1644 1645 1646 1647
1648
## "0" "0" "0" "1" "1" "1" "1" "1" "0" "1" "0" "0" "1" "1" "0"
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## 1649 1650 1651 1652 1653 1654 1655 1656 1657 1658 1659 1660 1661 1662 1663
1664
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## "0" "0" "1" "1" "1" "0" "0" "1" "1" "1" "1" "1" "1" "0" "1"
"1"
## 1745 1746 1747 1748 1749 1750 1751 1752 1753 1754 1755 1756 1757 1758 1759
1760
## "0" "0" "1" "0" "1" "0" "0" "1" "0" "0" "1" "0" "1" "1" "0"
"0"
## 1761 1762 1763 1764 1765 1766 1767 1768 1769 1770 1771 1772 1773 1774 1775
1776
## "0" "0" "0" "0" "0" "1" "0" "1" "0" "0" "0" "0" "0" "0" "1" "1"
"1"
## 1777 1778 1779 1780 1781 1782 1783 1784 1785 1786 1787 1788 1789 1790 1791
1792
## "1" "0" "0" "0" "1" "0" "0" "1" "1" "0" "0" "0" "0" "0" "1" "0"
"1"
## 1793 1794 1795 1796 1797 1798 1799 1800 1801 1802 1803 1804 1805 1806 1807
1808
## "0" "0" "0" "1" "1" "1" "1" "0" "0" "0" "1" "0" "1" "1" "0"
"0"
## 1809 1810 1811 1812 1813 1814 1815 1816 1817 1818 1819 1820 1821 1822 1823
1824
## "0" "1" "1" "0" "0" "0" "0" "1" "0" "0" "0" "0" "1" "1" "1"
"0"
## 1825 1826 1827 1828 1829 1830 1831 1832 1833 1834 1835 1836 1837 1838 1839
1840
## "1" "0" "0" "0" "0" "1" "0" "1" "0" "0" "1" "1" "0" "1" "1"
"0"
## 1841 1842 1843 1844 1845 1846 1847 1848 1849 1850 1851 1852 1853 1854 1855
1856
## "1" "1" "1" "1" "0" "0" "0" "1" "0" "0" "1" "0" "0" "0" "1"
"1"
```

```

## 1857 1858 1859 1860 1861 1862 1863 1864 1865 1866 1867 1868 1869 1870 1871
1872
## "0" "0" "1" "1" "0" "0" "0" "0" "1" "0" "0" "0" "0" "0" "1"
"1"
## 1873 1874 1875 1876 1877 1878 1879 1880 1881 1882 1883 1884 1885 1886 1887
1888
## "0" "0" "1" "1" "1" "1" "1" "0" "0" "1" "0" "1" "1" "0" "1"
"1"
## 1889 1890 1891 1892 1893 1894 1895 1896 1897 1898 1899 1900 1901 1902 1903
1904
## "1" "0" "1" "1" "1" "1" "1" "0" "1" "1" "1" "1" "1" "0" "0" "0"
"0"
## 1905 1906 1907 1908 1909 1910 1911 1912 1913 1914 1915 1916 1917 1918 1919
1920
## "0" "1" "0" "0" "0" "1" "1" "1" "1" "0" "1" "1" "1" "1" "0" "0"
"0"
## 1921 1922 1923 1924 1925 1926 1927 1928 1929 1930 1931 1932 1933 1934 1935
1936
## "1" "1" "0" "1" "0" "1" "0" "0" "1" "1" "1" "1" "1" "0" "1" "0"
"0"
## 1937 1938 1939 1940 1941 1942 1943 1944 1945 1946 1947 1948 1949 1950 1951
1952
## "1" "0" "1" "0" "1" "0" "1" "0" "0" "0" "1" "0" "0" "1" "1" "0"
"1"
## 1953 1954 1955 1956 1957 1958 1959 1960 1961 1962 1963 1964 1965 1966 1967
1968
## "1" "1" "0" "0" "0" "0" "0" "0" "1" "0" "1" "1" "1" "1" "1" "1"
"0"
## 1969 1970 1971 1972 1973 1974 1975 1976 1977 1978 1979 1980 1981 1982 1983
1984
## "0" "0" "0" "1" "0" "1" "1" "0" "1" "1" "1" "1" "1" "0" "0" "0"
"0"
## 1985 1986 1987 1988 1989 1990 1991 1992 1993 1994 1995 1996 1997 1998 1999
2000
## "1" "1" "1" "0" "1" "0" "0" "0" "0" "1" "0" "0" "1" "1" "1" "1"
"1"

```

Creating a confusion matrix (to see how well your model performed on the test set)

```

confusion_matrix <- table(test1$Performance, predictions_category)
print(confusion_matrix)

```

```

##      predictions_category
##      0      1
## High 516 484
## Low  487 513

```

```

accuracy <- (sum(diag(confusion_matrix)) / sum(confusion_matrix))
accuracy

```

```
## [1] 0.5145
```