

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
## <chr>,
## # 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
## 2      48 Female Divorced    HR         Assistant     4
## 3      34 Female Married    Marketing  Manager      4
## 4      27 Female Divorced    HR         Manager      4
## 5      40 Male    Married    HR         Analyst      1
## 6      58 Male    Married    Finance    Executive     3
## # i 19 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>, Performance <dbl>, Performance <chr>,
## # Performance_Rating <dbl>, Training_Hours_Last_Year <dbl>, Overtime
## <chr>,
## # Project_Count <dbl>, Average_Hours_Worked_Per_Week <dbl>,
## # Absenteeism <dbl>, Work_Environment_Satisfaction <dbl>, ...
```

```

## 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 x 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 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 2 2 2 2 ...
## $ Project_Count : num [1:10000] 9 9 3 9 1 1 1 3 8 6 ...

```

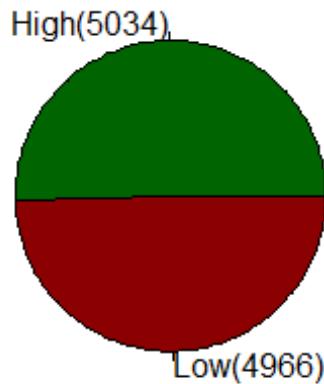
```

## $ 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                    Single   :3295   Marketing:2133
##  3rd Qu.:49.00                    Single   :3295   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.   : 1.000    Min.   : 1.000    Min.   :0.000
## 1st Qu.: 8.000    1st Qu.: 4.000    1st Qu.:2.000
## Median :15.000    Median : 7.000    Median :4.000
## Mean   :14.94     Mean   : 7.451    Mean   :4.472
## 3rd Qu.:22.000    3rd Qu.:11.000   3rd Qu.:7.000
## Max.   :29.000    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

```

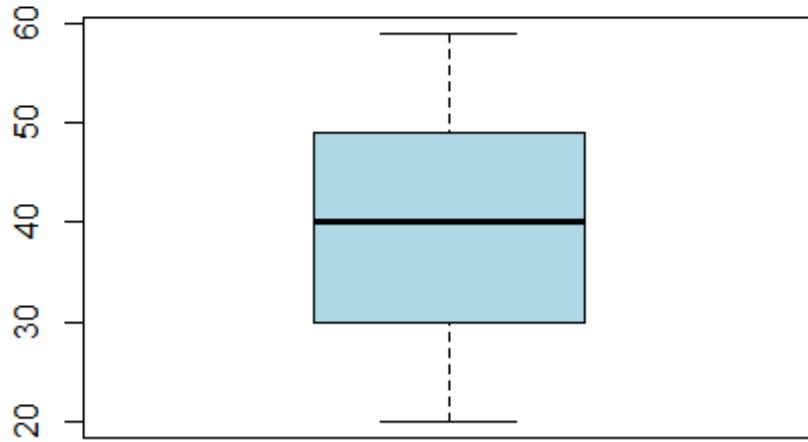
#Draw boxplot for the find outliers

```

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

```

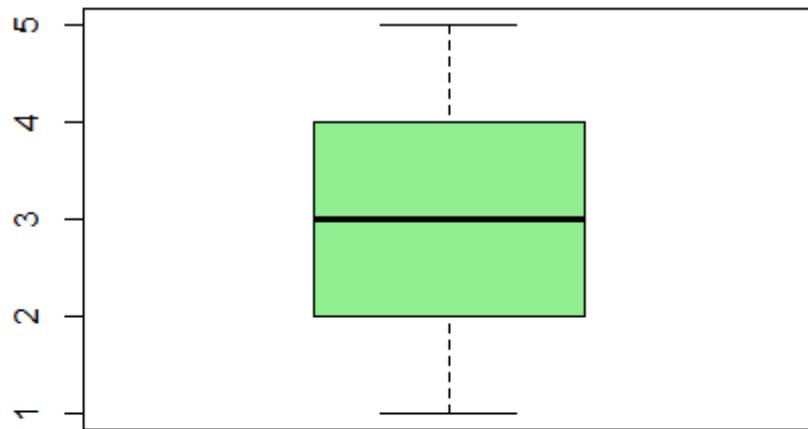
Age



outliers:

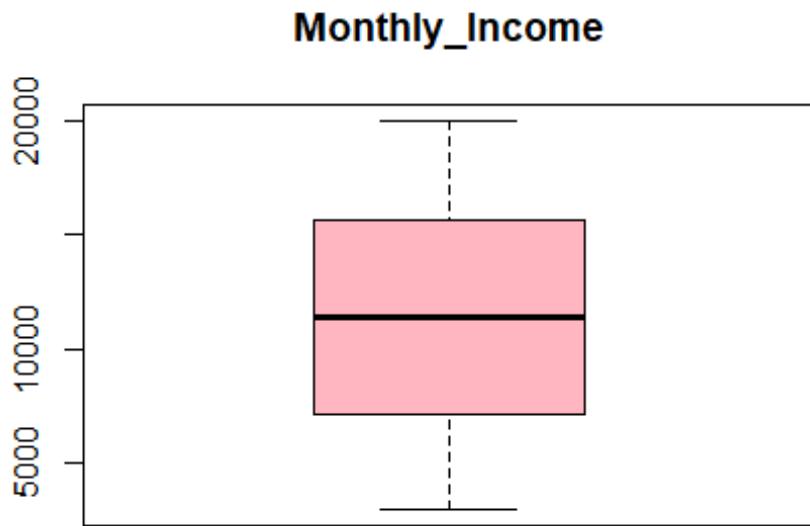
```
boxplot(Employee_Data$Job_Level, main = "Job_Level", col="lightgreen", sub =  
paste("outliers:", boxplot.stats(Employee_Data$Job_Level)$out))
```

Job_Level



outliers:

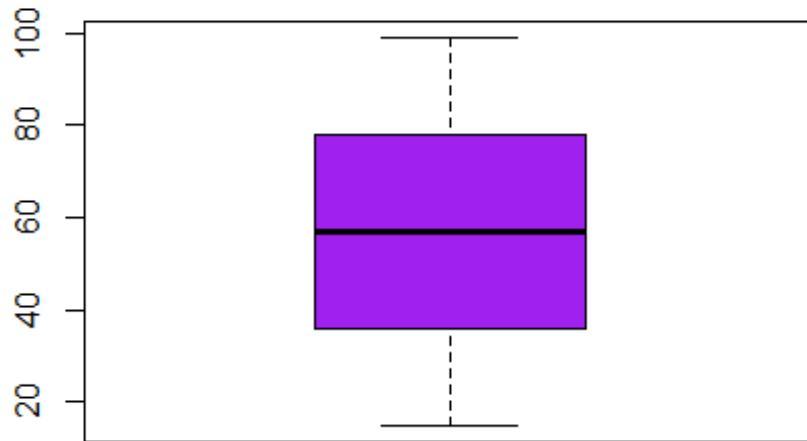
```
boxplot(Employee_Data$Monthly_Income, main =  
"Monthly_Income", col="lightpink", sub = paste("outliers:",  
boxplot.stats(Employee_Data$Monthly_Income)$out))
```



outliers:

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

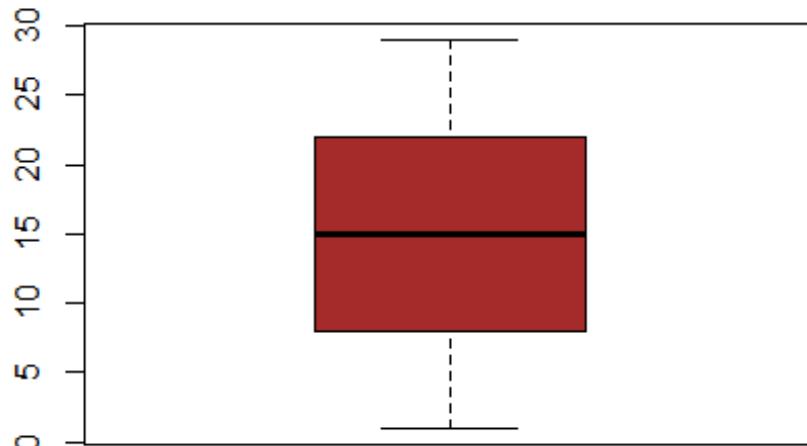
Hourly_Rate



outliers:

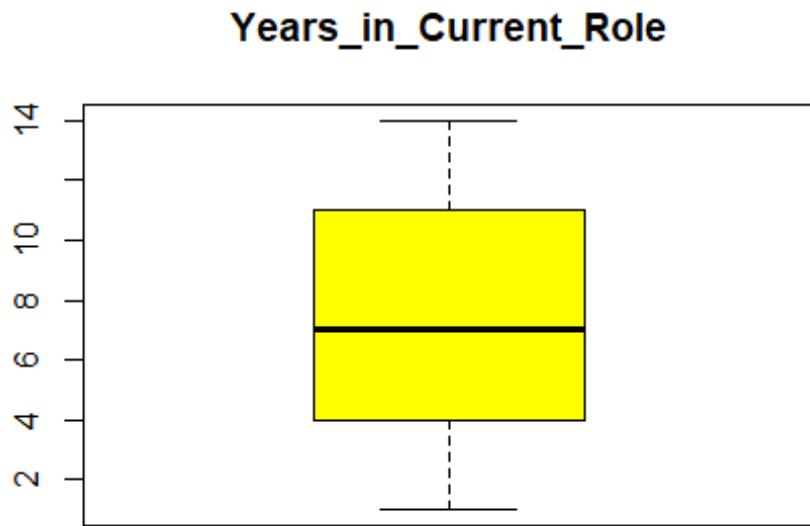
```
boxplot(Employee_Data$Years_at_Company, main = "Years_at_Company",
col="brown",sub = paste("outliers:",
boxplot.stats(Employee_Data$Years_at_Company)$out))
```

Years_at_Company



outliers:

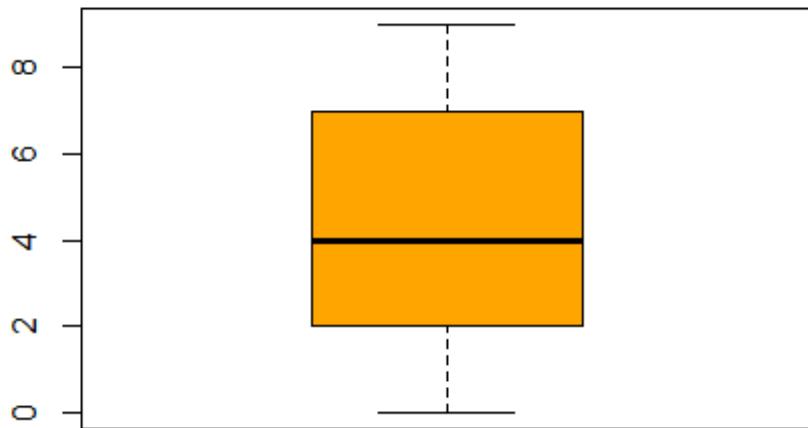
```
boxplot(Employee_Data$Years_in_Current_Role, main =
"Years_in_Current_Role", col="yellow", sub = paste("outliers:",
boxplot.stats(Employee_Data$Years_in_Current_Role)$out))
```



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))
```

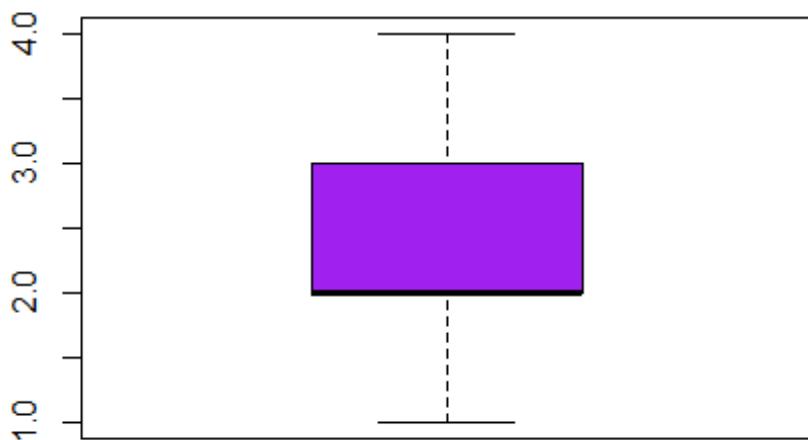
Years_Since_Last_Promotion



outliers:

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

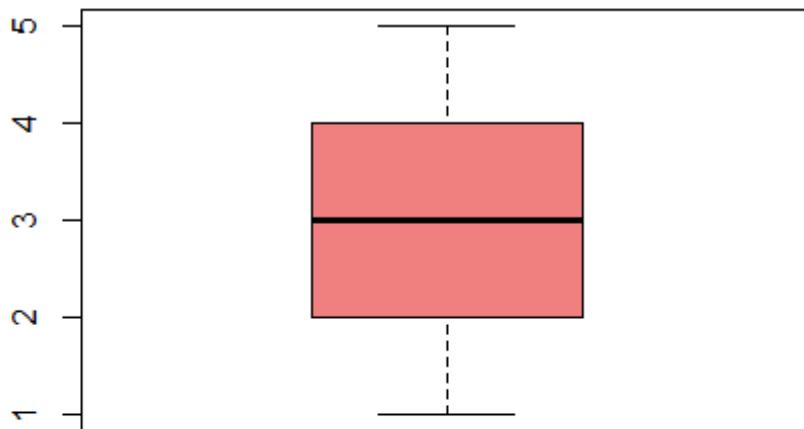
Work_Life_Balance



outliers:

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

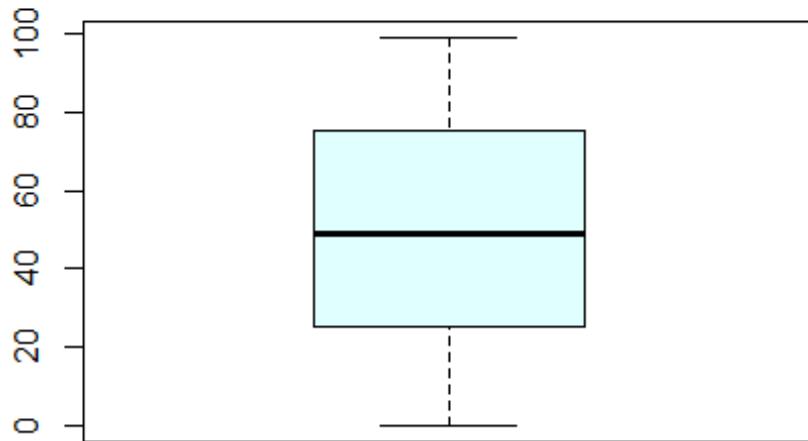
Job_Satisfaction



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))
```

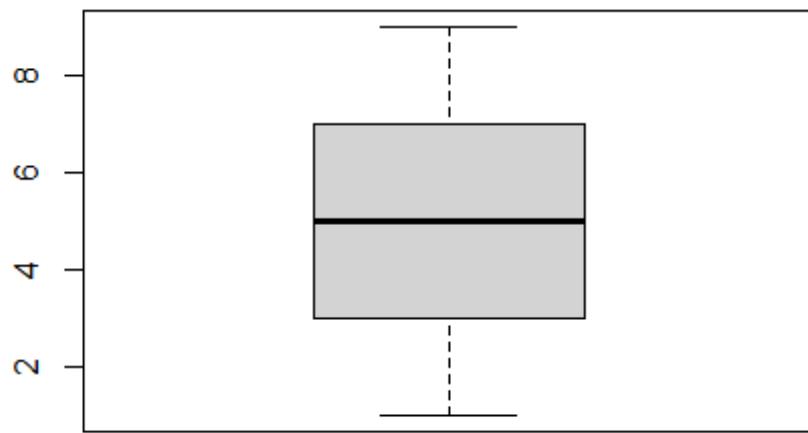
Training_Hours_Last_Year



outliers:

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

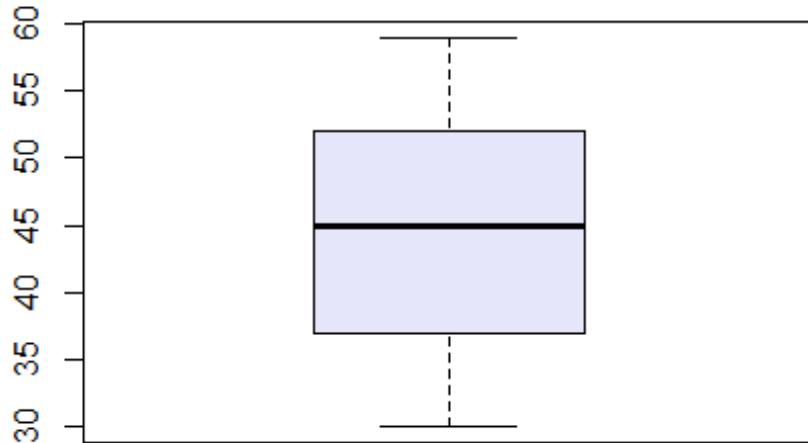
Project_Count



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))
```

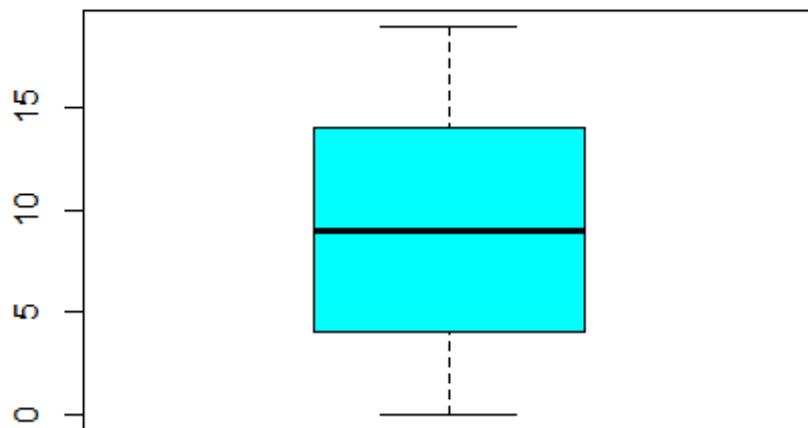
Avg_Hours_Worked



outliers:

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

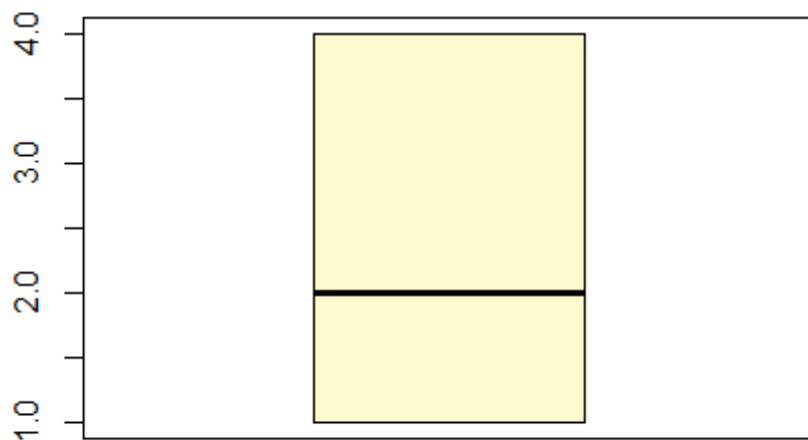
Absenteeism



outliers:

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

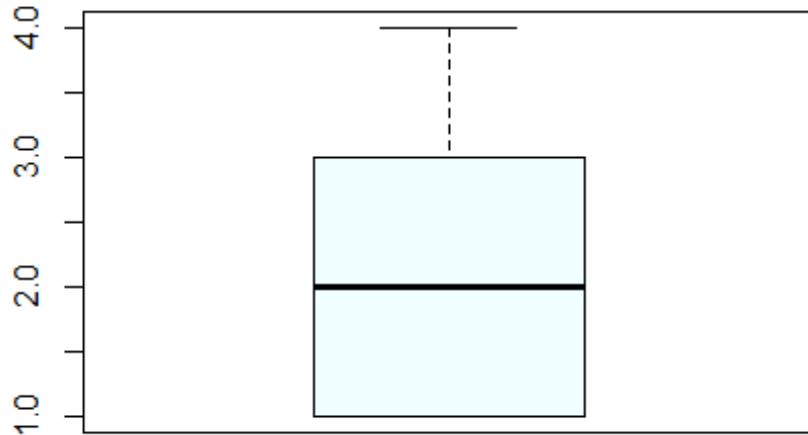
Work_Env_Sat



outliers:

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

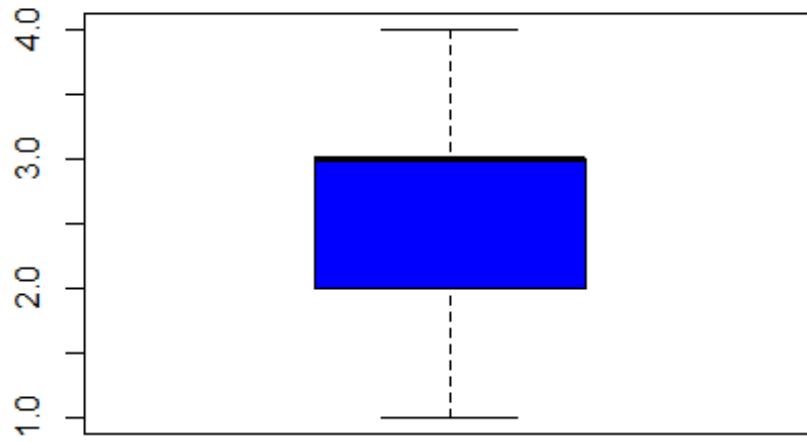
Rel_with_Manager



outliers:

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

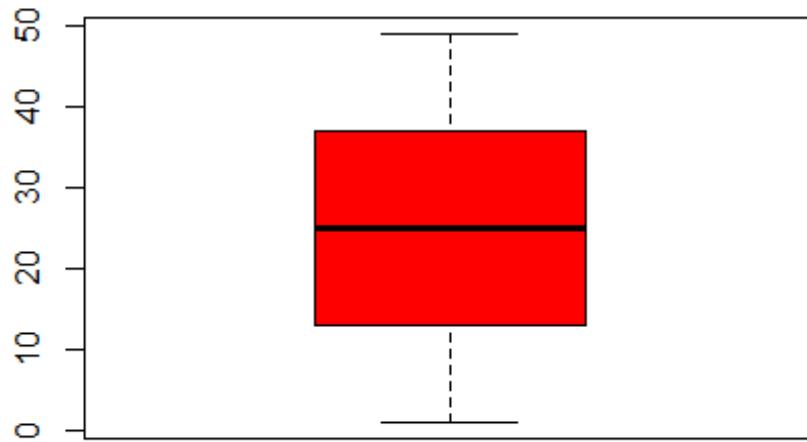
Job_Involvement



outliers:

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

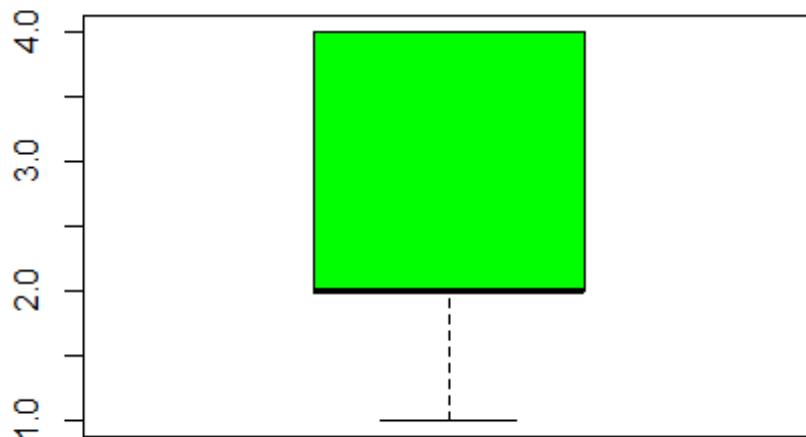
Distance_From_Home



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))
```

Companies_Worked



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))

# Histograms for continuous variables
hist(Job_Level, main="Histogram for Job_Level")
hist(Monthly_Income, main="Histogram for Monthly_Income")
hist(Hourly_Rate, main="Histogram for Hourly_Rate")
hist(Years_in_Current_Role, main="Histogram for Years_in_Current_Role")
hist(Years_Since_Last_Promotion, main="Histogram for Years_Since_Last_Promotion")
hist(Work_Life_Balance, main="Histogram for Work_Life_Balance")
hist(Job_Satisfaction, main="Histogram for Job_Satisfaction")
hist(Training_Hours_Last_Year, main="Histogram for Training_Hours_Last_Year")
hist(Project_Count, main="Histogram for Project_Count")
hist(Average_Hours_Worked_Per_Week, main="Histogram for Average_Hours_Worked_Per_Week")
hist(Absenteeism, main="Histogram for Absenteeism")
hist(Work_Environment_Satisfaction, main="Histogram for Work_Environment_Satisfaction")
hist(Relationship_with_Manager, main="Histogram for Relationship_with_Manager")
hist(Job_Involvement, main="Histogram for Job_Involvement")
hist(Distance_From_Home, main="Histogram for Distance_From_Home")
hist(Number_of_Companies_Worked, main="Histogram for Number_of_Companies_Worked")

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_Worked_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>
## 1 55   Male    Married      Sales     Assistant  4
## 2 27   Female   Divorced    IT        Executive  1
## 3 23   Female   Single      Marketing Manager   1
## 4 53   Male    Divorced    Sales     Executive  2
## 5 49   Female   Married     Marketing Assistant 1
## 6 27   Male    Divorced    IT        Manager   4
## # 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>
## 1 54   Female  Divorced    Finance   Executive  3
## 2 53   Female  Married     HR       Assistant  1
## 3 57   Female  Divorced    IT        Manager   3
## 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>                          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

```

```

##                                     Df Deviance   AIC
## - 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>                         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
## 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

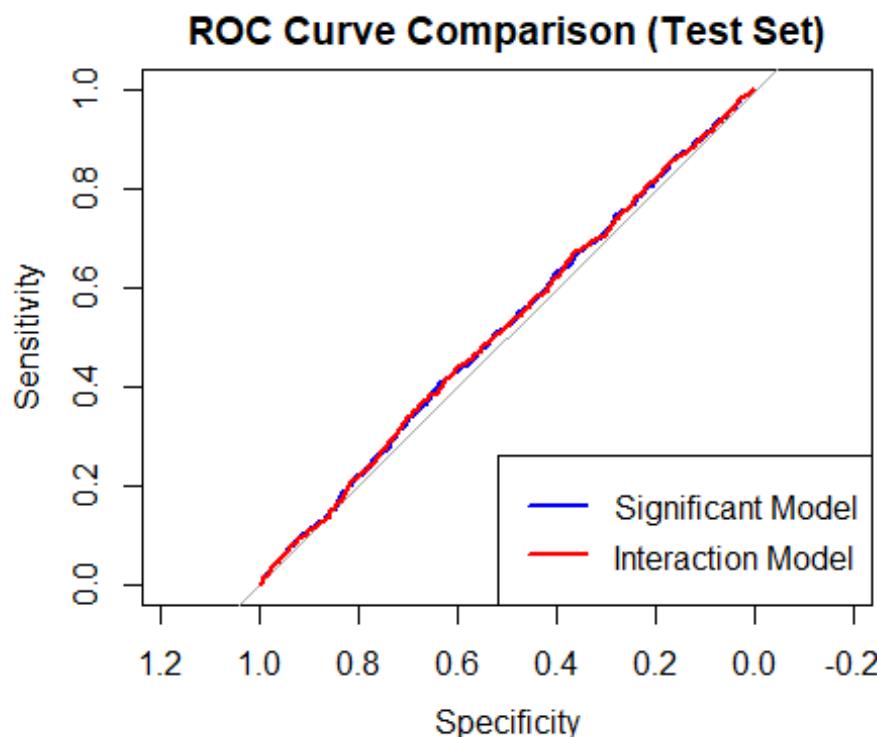
## Setting direction: controls < cases

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
##               1.0820062          0.9983629

```

```

## Work_Environment_Satisfaction           Job_Involvement
##                               1.0433347          0.9585591

#fit <- model2$fitted

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

#r <- (Employee_Data$Attrition_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 ( $H_0$ ): The model fits the data well.
#Alternative hypothesis ( $H_a$ ): 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
0.4830076							
##	369	370	371	372	373	374	375
376							
##	0.4899012	0.4952899	0.4830076	0.5131962	0.4796706	0.5058748	0.4898523
0.4854629							
##	377	378	379	380	381	382	383
384							
##	0.4561482	0.5090329	0.4622515	0.4902617	0.5008905	0.5091266	0.5005702
0.4924028							
##	385	386	387	388	389	390	391
392							
##	0.4984774	0.4808486	0.4975891	0.4723791	0.4907158	0.4826676	0.5180622
0.5012511							
##	393	394	395	396	397	398	399
400							
##	0.5042567	0.5286643	0.4769423	0.5185159	0.4976783	0.4577745	0.5071035
0.5131761							
##	401	402	403	404	405	406	407
408							
##	0.5038226	0.4837569	0.5189249	0.4699296	0.4613683	0.5119235	0.5016406
0.4865771							
##	409	410	411	412	413	414	415
416							
##	0.4609855	0.5171304	0.4956303	0.4601470	0.5127423	0.5355789	0.5234025
0.5017054							
##	417	418	419	420	421	422	423
424							
##	0.4911253	0.4727674	0.4993413	0.4980433	0.5087372	0.5278276	0.4915838
0.4638123							
##	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							

```

## 0.4980923 0.4809176 0.4609855 0.4895810 0.5070099 0.4800551 0.4961335
0.4969283
##      441      442      443      444      445      446      447
448
## 0.4903308 0.5322482 0.4816422 0.5189249 0.4666658 0.4906712 0.5286198
0.5213341
##      449      450      451      452      453      454      455
456
## 0.5191758 0.4983838 0.4622071 0.4789418 0.5261939 0.4609855 0.5017990
0.4903107
##      457      458      459      460      461      462      463
464
## 0.4753075 0.4947910 0.4908094 0.4956102 0.5262140 0.5274192 0.4808730
0.5196293
##      465      466      467      468      469      470      471
472
## 0.4789418 0.5041674 0.5253524 0.5020502 0.4922847 0.5086235 0.5135163
0.4895810
##      473      474      475      476      477      478      479
480
## 0.5155627 0.5180176 0.5154936 0.5054163 0.5200383 0.4690689 0.5017990
0.5327250
##      481      482      483      484      485      486      487
488
## 0.5175395 0.4561482 0.5021395 0.4753075 0.5383618 0.5062153 0.5025290
0.5388377
##      489      490      491      492      493      494      495
496
## 0.4906511 0.4985667 0.4928570 0.4862124 0.5408728 0.4911944 0.4899660
0.5201073
##      497      498      499      500      501      502      503
504
## 0.5282560 0.4638811 0.4976337 0.5143349 0.4691822 0.5119235 0.4902171
0.5102614
##      505      506      507      508      509      510      511
512
## 0.4886485 0.5010043 0.5200829 0.5185159 0.4690445 0.5028896 0.5302081
0.4871002
##      513      514      515      516      517      518      519
520
## 0.5209008 0.4951516 0.5057121 0.5037333 0.4866217 0.4843043 0.4918997
0.4874894
##      521      522      523      524      525      526      527
528
## 0.5225850 0.5033482 0.5152225 0.4792127 0.4791883 0.4924230 0.4768733
0.5505527
##      529      530      531      532      533      534      535
536
## 0.5151290 0.4820758 0.5298444 0.4858276 0.4866908 0.5241510 0.4829386
0.5000223

```

##	537	538	539	540	541	542	543
544							
##	0.5012756	0.4883283	0.4976827	0.5074194	0.5086033	0.5040982	0.4930836
0.5253969							
##	545	546	547	548	549	550	551
552							
##	0.5034173	0.5122192	0.5187667	0.4698851	0.4793508	0.4853737	0.5229248
0.4980232							
##	553	554	555	556	557	558	559
560							
##	0.5066896	0.4863060	0.4727674	0.4873958	0.4874448	0.4638367	0.5115141
0.5026427							
##	561	562	563	564	565	566	567
568							
##	0.4964740	0.4887622	0.4735599	0.5045971	0.4756716	0.5481189	0.4922847
0.4601714							
##	569	570	571	572	573	574	575
576							
##	0.4598087	0.4714935	0.4911253	0.4935133	0.5481189	0.4935868	0.4719952
0.4919242							
##	577	578	579	580	581	582	583
584							
##	0.5395832	0.4792127	0.4944952	0.5127624	0.5062109	0.5005255	0.4859413
0.4993413							
##	585	586	587	588	589	590	591
592							
##	0.5204918	0.4800996	0.5154936	0.5124265	0.4736044	0.5172440	0.5045971
0.5086479							
##	593	594	595	596	597	598	599
600							
##	0.5302526	0.4816667	0.4707014	0.5278031	0.4938783	0.4891270	0.5241955
0.4675503							
##	601	602	603	604	605	606	607
608							
##	0.5095360	0.5208806	0.5180378	0.4740373	0.4918306	0.5269864	0.4711786
0.5062800							
##	609	610	611	612	613	614	615
616							
##	0.4914902	0.4662824	0.5347394	0.4789418	0.4565547	0.5177222	0.5057812
0.4866261							
##	617	618	619	620	621	622	623
624							
##	0.4874894	0.5086724	0.4597644	0.4915593	0.5278276	0.5070790	0.4877807
0.4476260							
##	625	626	627	628	629	630	631
632							
##	0.4883283	0.4812576	0.5189739	0.4686853	0.4931081	0.5028449	0.4952452
0.5095115							
##	633	634	635	636	637	638	639
640							

```

## 0.4699296 0.4825539 0.4871247 0.5229004 0.5432445 0.4922847 0.4792616
0.5250128
##       641       642       643       644       645       646       647
648
## 0.5232646 0.5073503 0.5057812 0.5477130 0.5278721 0.4842151 0.5150844
0.4699984
##       649       650       651       652       653       654       655
656
## 0.5070790 0.4988625 0.4873713 0.5310687 0.5159028 0.5245596 0.4947464
0.5334718
##       657       658       659       660       661       662       663
664
## 0.5298889 0.5396276 0.4833521 0.5160611 0.5016161 0.4898321 0.4557418
0.5327250
##       665       666       667       668       669       670       671
672
## 0.5286643 0.4805532 0.5005500 0.5283048 0.5095605 0.4911944 0.5208562
0.5237178
##       673       674       675       676       677       678       679
680
## 0.5195848 0.4895566 0.5033928 0.4650349 0.5355789 0.5128113 0.5327006
0.5049174
##       681       682       683       684       685       686       687
688
## 0.5033237 0.4870109 0.5351714 0.5144486 0.4796461 0.4842597 0.5351470
0.4744013
##       689       690       691       692       693       694       695
696
## 0.4569613 0.5136300 0.5026182 0.4764200 0.5071481 0.5049822 0.5192649
0.4946974
##       697       698       699       700       701       702       703
704
## 0.4907202 0.5103305 0.5029386 0.4808730 0.5032992 0.5257854 0.4752630
0.4946974
##       705       706       707       708       709       710       711
712
## 0.4723346 0.5001605 0.4853248 0.5225850 0.5033237 0.4862168 0.4825985
0.4792372
##       713       714       715       716       717       718       719
720
## 0.5303215 0.5221273 0.5314323 0.5160856 0.4761248 0.5082586 0.5083277
0.5152916
##       721       722       723       724       725       726       727
728
## 0.4609855 0.4918752 0.4751940 0.5021150 0.4992275 0.5295052 0.5022086
0.5070099
##       729       730       731       732       733       734       735
736
## 0.4695214 0.5147643 0.5073949 0.5034130 0.5131962 0.4488417 0.5347394
0.5241265

```

##	737	738	739	740	741	742	743
744							
##	0.4476260	0.5257164	0.5079182	0.5217674	0.4914902	0.4561482	0.5024800
0.5034620							
##	745	746	747	748	749	750	751
752							
##	0.4765336	0.5069407	0.4967453	0.4926942	0.5118098	0.5062153	0.4858722
0.4922646							
##	753	754	755	756	757	758	759
760							
##	0.5172685	0.4480312	0.5179486	0.5197919	0.4874448	0.5041184	0.5095360
0.4919443							
##	761	762	763	764	765	766	767
768							
##	0.4773511	0.4858276	0.5298000	0.5013448	0.4993413	0.5229449	0.5164948
0.5184713							
##	769	770	771	772	773	774	775
776							
##	0.4911498	0.5082384	0.4838951	0.4820513	0.5058503	0.5132207	0.4858276
0.4769423							
##	777	778	779	780	781	782	783
784							
##	0.5012310	0.5193340	0.4674815	0.4798043	0.4924028	0.4951070	0.4610299
0.5102167							
##	785	786	787	788	789	790	791
792							
##	0.5229938	0.5075576	0.5168593	0.4992722	0.5029832	0.5237178	0.5066939
0.4804396							
##	793	794	795	796	797	798	799
800							
##	0.5008660	0.4719507	0.4927677	0.4922891	0.5115587	0.4806222	0.4492471
0.4727430							
##	801	802	803	804	805	806	807
808							
##	0.4732203	0.4715624	0.5367770	0.4887622	0.4943569	0.5147888	0.5004564
0.4955857							
##	809	810	811	812	813	814	815
816							
##	0.5017054	0.5091957	0.5114896	0.4997308	0.5016852	0.4814403	0.5016406
0.4816422							
##	817	818	819	820	821	822	823
824							
##	0.5075130	0.4984976	0.5074640	0.4939027	0.5128358	0.4797151	0.5036886
0.4739683							
##	825	826	827	828	829	830	831
832							
##	0.5122638	0.5216740	0.5094871	0.4642198	0.4597644	0.5189048	0.4690689
0.4760803							
##	833	834	835	836	837	838	839
840							

```

## 0.4934687 0.5176085 0.4817113 0.5180867 0.4894183 0.4800795 0.5224916
0.5464951
##     841      842      843      844      845      846      847
848
## 0.4723791 0.4751940 0.5274192 0.5180867 0.5346949 0.5269419 0.4617755
0.4617999
##     849      850      851      852      853      854      855
856
## 0.4833276 0.5053716 0.5140393 0.5196739 0.5241510 0.4833477 0.5147643
0.5197185
##     857      858      859      860      861      862      863
864
## 0.4828695 0.5114695 0.4922891 0.4784395 0.5270553 0.4923092 0.5180867
0.5000914
##     865      866      867      868      869      870      871
872
## 0.4931974 0.5189739 0.5026182 0.4956548 0.4818048 0.4753075 0.5359177
0.4980188
##     873      874      875      876      877      878      879
880
## 0.4919933 0.4931773 0.5051003 0.4851227 0.5375473 0.5091021 0.4926741
0.4903107
##     881      882      883      884      885      886      887
888
## 0.5008415 0.5025938 0.5299377 0.4793953 0.4740373 0.4975891 0.5299133
0.4841705
##     889      890      891      892      893      894      895
896
## 0.4833521 0.5448703 0.4821693 0.5323171 0.5278276 0.4870556 0.4914902
0.5192448
##     897      898      899      900      901      902      903
904
## 0.4923538 0.5144040 0.4996371 0.5343317 0.5278276 0.4830321 0.5099700
0.4752630
##     905      906      907      908      909      910      911
912
## 0.4597400 0.5199937 0.4702932 0.5119480 0.5020257 0.5177222 0.4816422
0.4744904
##     913      914      915      916      917      918      919
920
## 0.4940611 0.4687742 0.5208562 0.5327006 0.4801442 0.4935825 0.5171750
0.5059195
##     921      922      923      924      925      926      927
928
## 0.4992275 0.4885995 0.5440575 0.4961091 0.4910116 0.4682973 0.5004564
0.4776707
##     929      930      931      932      933      934      935
936
## 0.4894918 0.4984284 0.5115587 0.4870800 0.5139458 0.5042365 0.4894428
0.5302770

```

##	937	938	939	940	941	942	943
944							
##	0.4927879	0.4845998	0.4891471	0.4829631	0.4956347	0.5093978	0.4829185
0.5103103							
##	945	946	947	948	949	950	951
952							
##	0.5069854	0.5266469	0.4808241	0.5368013	0.4922891	0.5069854	0.4984328
0.5265779							
##	953	954	955	956	957	958	959
960							
##	0.4724035	0.5061461	0.5216294	0.4866908	0.4735354	0.5173375	0.4699095
0.5152427							
##	961	962	963	964	965	966	967
968							
##	0.5367326	0.4670736	0.4870354	0.5058950	0.5009351	0.4841215	0.4898523
0.4727186							
##	969	970	971	972	973	974	975
976							
##	0.5185159	0.5029386	0.4512749	0.4984083	0.4878542	0.5179932	0.5208116
0.4968836							
##	977	978	979	980	981	982	983
984							
##	0.5029832	0.5053716	0.5046907	0.4480312	0.5372087	0.5302081	0.5225606
0.4959708							
##	985	986	987	988	989	990	991
992							
##	0.5001360	0.4841705	0.5085788	0.4927677	0.5278477	0.4899214	0.5244906
0.4989317							
##	993	994	995	996	997	998	999
1000							
##	0.5034620	0.5248993	0.4764890	0.5046662	0.4874649	0.4706770	0.5071236
0.4808975							
##	1001	1002	1003	1004	1005	1006	1007
1008							
##	0.5017097	0.5220828	0.4753075	0.5099210	0.5066248	0.4878297	0.5155627
0.5440575							
##	1009	1010	1011	1012	1013	1014	1015
1016							
##	0.4944016	0.4858075	0.4646273	0.4793263	0.5144731	0.5306851	0.5302081
0.4809421							
##	1017	1018	1019	1020	1021	1022	1023
1024							
##	0.5278477	0.5160856	0.4622071	0.4902617	0.5184914	0.5001850	0.5106262
0.4837368							
##	1025	1026	1027	1028	1029	1030	1031
1032							
##	0.4699095	0.5274881	0.4894183	0.4922401	0.4638123	0.4821693	0.4902416
0.4886931							
##	1033	1034	1035	1036	1037	1038	1039
1040							

```

## 0.5176977 0.4940410 0.5436510 0.5042322 0.4715423 0.5078736 0.5147643
0.5009553
##      1041      1042      1043      1044      1045      1046      1047
1048
## 0.4952006 0.4857830 0.4912189 0.5049174 0.5062354 0.4826230 0.4973379
0.5042322
##      1049      1050      1051      1052      1053      1054      1055
1056
## 0.5053716 0.4655114 0.5033482 0.5156519 0.4732649 0.4732203 0.5013894
0.4484364
##      1057      1058      1059      1060      1061      1062      1063
1064
## 0.5163811 0.5062800 0.4972486 0.4881901 0.5135365 0.5213096 0.5273503
0.5159474
##      1065      1066      1067      1068      1069      1070      1071
1072
## 0.5176977 0.5380233 0.4764890 0.4565547 0.5111983 0.5428379 0.5306851
0.5050513
##      1073      1074      1075      1076      1077      1078      1079
1080
## 0.5045280 0.5144486 0.5233537 0.5063290 0.4874649 0.5050758 0.4850781
0.4719952
##      1081      1082      1083      1084      1085      1086      1087
1088
## 0.4760113 0.4940611 0.4821693 0.4980923 0.4904000 0.4967943 0.4976092
0.5029141
##      1089      1090      1091      1092      1093      1094      1095
1096
## 0.5168349 0.4776707 0.5111047 0.4833276 0.4930592 0.4658503 0.5014139
0.4874203
##      1097      1098      1099      1100      1101      1102      1103
1104
## 0.4655114 0.4829185 0.4911944 0.4899660 0.4977720 0.5070790 0.5464951
0.5204027
##      1105      1106      1107      1108      1109      1110      1111
1112
## 0.5306851 0.5058258 0.4780997 0.4813957 0.4987934 0.5164012 0.4752185
0.5266224
##      1113      1114      1115      1116      1117      1118      1119
1120
## 0.4857585 0.4951070 0.5376161 0.4944952 0.4655114 0.4557418 0.4959751
0.4784884
##      1121      1122      1123      1124      1125      1126      1127
1128
## 0.5130825 0.4858031 0.4712231 0.5404416 0.5347394 0.5169284 0.4866261
0.5020704
##      1129      1130      1131      1132      1133      1134      1135
1136
## 0.5024353 0.5061663 0.5290725 0.4674815 0.4891716 0.5266713 0.4728564
0.5017789

```

##	1137	1138	1139	1140	1141	1142	1143
1144							
##	0.5363939	0.5400589	0.5183577	0.4663268	0.5216740	0.4948155	0.4816667
0.4691822							
##	1145	1146	1147	1148	1149	1150	1151
1152							
##	0.4943123	0.5302971	0.4914657	0.5221072	0.4952251	0.5017054	0.5168593
0.4878743							
##	1153	1154	1155	1156	1157	1158	1159
1160							
##	0.5184023	0.5089883	0.5038471	0.5139947	0.5139458	0.5283048	0.5119480
0.4944261							
##	1161	1162	1163	1164	1165	1166	1167
1168							
##	0.4825539	0.5493361	0.4601714	0.4573679	0.5069854	0.5197185	0.4846444
0.4845552							
##	1169	1170	1171	1172	1173	1174	1175
1176							
##	0.4960198	0.4861678	0.5176977	0.5509581	0.4789864	0.4764646	0.5123329
0.4508692							
##	1177	1178	1179	1180	1181	1182	1183
1184							
##	0.5111493	0.5371400	0.5193584	0.5025938	0.5114004	0.4903553	0.4838951
0.5005702							
##	1185	1186	1187	1188	1189	1190	1191
1192							
##	0.4847135	0.5509581	0.5126976	0.4805532	0.4788484	0.4955612	0.5157008
0.4923538							
##	1193	1194	1195	1196	1197	1198	1199
1200							
##	0.4985019	0.4955410	0.4732649	0.4969283	0.5205163	0.4809176	0.5220382
0.4915838							
##	1201	1202	1203	1204	1205	1206	1207
1208							
##	0.5139012	0.4655114	0.4883283	0.5026427	0.5246286	0.4926942	0.5294363
0.4727875							
##	1209	1210	1211	1212	1213	1214	1215
1216							
##	0.4977273	0.4976582	0.4969081	0.4545228	0.4932219	0.4837569	0.5367326
0.5021193							
##	1217	1218	1219	1220	1221	1222	1223
1224							
##	0.5091467	0.5004809	0.5008415	0.4936314	0.4877807	0.5221518	0.5087171
0.5049376							
##	1225	1226	1227	1228	1229	1230	1231
1232							
##	0.5090574	0.5436510	0.5000914	0.4698607	0.4943325	0.5111292	0.5016161
0.4699984							
##	1233	1234	1235	1236	1237	1238	1239
1240							

```

## 0.5164012 0.5185403 0.5094669 0.5036886 0.4809867 0.5266023 0.5115832
0.4707703
## 1241      1242      1243      1244      1245      1246      1247
1248
## 0.4932420 0.5469011 0.5126732 0.4829429 0.5053918 0.5363939 0.4618443
0.4808932
## 1249      1250      1251      1252      1253      1254      1255
1256
## 0.4946974 0.4870354 0.5270108 0.4813267 0.5008415 0.4924028 0.5030034
0.5163811
## 1257      1258      1259      1260      1261      1262      1263
1264
## 0.5168795 0.4866261 0.5074640 0.5057567 0.4756472 0.4759869 0.5118990
0.5135854
## 1265      1266      1267      1268      1269      1270      1271
1272
## 0.5323171 0.4858276 0.4524924 0.4963847 0.4646517 0.5293918 0.5131717
0.4825295
## 1273      1274      1275      1276      1277      1278      1279
1280
## 0.4907158 0.5168349 0.4940410 0.5030279 0.5146751 0.4841661 0.5290969
0.4858031
## 1281      1282      1283      1284      1285      1286      1287
1288
## 0.4866016 0.5160611 0.4719707 0.5066003 0.4654670 0.5024353 0.5160165
0.4622071
## 1289      1290      1291      1292      1293      1294      1295
1296
## 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" "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

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##	257	258	259	260	261	262	263	264	265	266	267	268	269	270	271
272															
##	"0"	"0"	"0"	"0"	"1"	"0"	"0"	"0"	"1"	"1"	"1"	"0"	"0"	"0"	"0"
"0"															
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##	"0"	"0"	"1"	"0"	"1"	"1"	"0"	"1"	"1"	"0"	"0"	"1"	"0"	"0"	"0"
"1"															
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##	"1"	"1"	"0"	"0"	"1"	"1"	"1"	"0"	"0"	"1"	"0"	"1"	"1"	"1"	"1"
"1"															
305	306	307	308	309	310	311	312	313	314	315	316	317	318	319	320
##	"1"	"1"	"1"	"0"	"0"	"1"	"0"	"1"	"1"	"1"	"0"	"0"	"1"	"1"	"0"
"0"															
321	322	323	324	325	326	327	328	329	330	331	332	333	334	335	336
##	"1"	"1"	"0"	"1"	"1"	"0"	"0"	"0"	"0"	"0"	"0"	"1"	"0"	"1"	"0"
"1"															
337	338	339	340	341	342	343	344	345	346	347	348	349	350	351	352
##	"1"	"1"	"1"	"1"	"0"	"0"	"0"	"0"	"1"	"1"	"0"	"1"	"0"	"1"	"1"
"1"															
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##	"0"	"0"	"1"	"0"	"1"	"1"	"1"	"1"	"1"	"0"	"1"	"1"	"0"	"1"	"1"
"0"															
369	370	371	372	373	374	375	376	377	378	379	380	381	382	383	384
##	"0"	"0"	"0"	"1"	"0"	"1"	"0"	"0"	"0"	"1"	"0"	"0"	"1"	"1"	"1"
"0"															
385	386	387	388	389	390	391	392	393	394	395	396	397	398	399	400
##	"0"	"0"	"0"	"0"	"0"	"0"	"1"	"1"	"1"	"1"	"0"	"1"	"0"	"0"	"1"
"1"															
401	402	403	404	405	406	407	408	409	410	411	412	413	414	415	416
##	"1"	"0"	"1"	"0"	"0"	"1"	"1"	"0"	"0"	"1"	"0"	"0"	"1"	"1"	"1"
"1"															
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##	"0"	"0"	"0"	"0"	"1"	"1"	"0"	"0"	"1"	"0"	"0"	"0"	"1"	"0"	"0"
"1"															
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##	"0"	"0"	"0"	"0"	"1"	"0"	"0"	"0"	"0"	"1"	"0"	"1"	"0"	"0"	"1"
"1"															
449	450	451	452	453	454	455	456	457	458	459	460	461	462	463	464

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

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"1"
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1024
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1040
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"1"
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## 1057 1058 1059 1060 1061 1062 1063 1064 1065 1066 1067 1068 1069 1070 1071
1072
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1088
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1104
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"1"
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1120
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"0"
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1136
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"0"
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1216
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"1"
## 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" "0" "0" "1" "1" "1"
"0"
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1248
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"0"
## 1249 1250 1251 1252 1253 1254 1255 1256 1257 1258 1259 1260 1261 1262 1263
1264

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1296
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1344
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"1"
## 1361 1362 1363 1364 1365 1366 1367 1368 1369 1370 1371 1372 1373 1374 1375
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"1"
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1392
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"1"
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1424
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## 1441 1442 1443 1444 1445 1446 1447 1448 1449 1450 1451 1452 1453 1454 1455
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## 1457 1458 1459 1460 1461 1462 1463 1464 1465 1466 1467 1468 1469 1470 1471
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1488
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1504
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1616
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"1"
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1632
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"0"
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1664

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1712
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## "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" "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" "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" "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" "0" "1" "0" "0" "0" "1" "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" "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" "0" "0" "1" "0" "1" "1" "1" "0" "1" "1"
"1"
## 1889 1890 1891 1892 1893 1894 1895 1896 1897 1898 1899 1900 1901 1902 1903
1904
## "1" "0" "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" "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" "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" "0" "1" "0" "0" "1"
"0"
## 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"
"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" "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"

# 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
```