

knnChurn.R

sigsp

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```
## Author: Stephen E. Porter
## Title: k-Nearest-Neighbor (Task 1)
## Course: WGU D209: Data Mining I
## Instructor: Dr. Festus Elleh
options(warn=-1)

# Libraries
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.1 --
## v ggplot2 3.3.5      v purrr  0.3.4
## v tibble  3.1.5      v dplyr  1.0.7
## v tidyr   1.1.4      v stringr 1.4.0
## v readr   2.0.2      v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

library(fastDummies)
library(class)
library(caret)

## Loading required package: lattice

##
## Attaching package: 'caret'

## The following object is masked from 'package:purrr':
##
##     lift

library(pROC)

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

##
## Attaching package: 'pROC'

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

# Import CSV as data frame
df <- read.csv(file = 'C:/WGU/D209 Data Mining I/churn_clean.csv')
```

```
# Checking for nulls
```

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

```
##          CaseOrder      Customer_id      Interaction
##          0          0          0
##          UID          City          State
##          0          0          0
##          County      Zip          Lat
##          0          0          0
##          Lng          Population      Area
##          0          0          0
##          TimeZone      Job          Children
##          0          0          0
##          Age          Income      Marital
##          0          0          0
##          Gender      Churn      Outage_sec_perweek
##          0          0          0
##          Email      Contacts      Yearly_equip_failure
##          0          0          0
##          Techie      Contract      Port_modem
##          0          0          0
##          Tablet      InternetService      Phone
##          0          0          0
##          Multiple      OnlineSecurity      OnlineBackup
##          0          0          0
##          DeviceProtection      TechSupport      StreamingTV
##          0          0          0
##          StreamingMovies      PaperlessBilling      PaymentMethod
##          0          0          0
##          Tenure      MonthlyCharge      Bandwidth_GB_Year
##          0          0          0
##          Item1      Item2      Item3
##          0          0          0
##          Item4      Item5      Item6
##          0          0          0
##          Item7      Item8
##          0          0
```

```
dim(df)
```

```
## [1] 10000    50
```

```
str(df)
```

```
## 'data.frame':    10000 obs. of  50 variables:
## $ CaseOrder      : int  1 2 3 4 5 6 7 8 9 10 ...
## $ Customer_id    : chr  "K409198" "S120509" "K191035" "D90850" ...
## $ Interaction     : chr  "aa90260b-4141-4a24-8e36-b04ce1f4f77b" "fb76459f-c047-4a9d-8af9-e0f7d4..."
## $ UID            : chr  "e885b299883d4f9fb18e39c75155d990" "f2de8bef964785f41a2959829830fb8a" ...
## $ City           : chr  "Point Baker" "West Branch" "Yamhill" "Del Mar" ...
## $ State          : chr  "AK" "MI" "OR" "CA" ...
## $ County         : chr  "Prince of Wales-Hyder" "Ogemaw" "Yamhill" "San Diego" ...
## $ Zip            : int  99927 48661 97148 92014 77461 31030 37847 73109 34771 45237 ...
## $ Lat            : num  56.3 44.3 45.4 33 29.4 ...
## $ Lng            : num  -133.4 -84.2 -123.2 -117.2 -95.8 ...
```

```
## $ Population      : int  38 10446 3735 13863 11352 17701 2535 23144 17351 20193 ...
## $ Area            : chr   "Urban" "Urban" "Urban" "Suburban" ...
## $ TimeZone        : chr   "America/Sitka" "America/Detroit" "America/Los_Angeles" "America/Los_Angeles" ...
## $ Job             : chr   "Environmental health practitioner" "Programmer, multimedia" "Chief Financial Officer" ...
## $ Children        : int   0 1 4 1 0 3 0 2 2 1 ...
## $ Age             : int   68 27 50 48 83 83 79 30 49 86 ...
## $ Income          : num   28562 21705 9610 18925 40074 ...
## $ Marital         : chr   "Widowed" "Married" "Widowed" "Married" ...
## $ Gender          : chr   "Male" "Female" "Female" "Male" ...
## $ Churn           : chr   "No" "Yes" "No" "No" ...
## $ Outage_sec_perweek : num   7.98 11.7 10.75 14.91 8.15 ...
## $ Email           : int   10 12 9 15 16 15 10 16 20 18 ...
## $ Contacts        : int   0 0 0 2 2 3 0 0 2 1 ...
## $ Yearly equip_failure: int   1 1 1 0 1 1 1 0 3 0 ...
## $ Techie          : chr   "No" "Yes" "Yes" "Yes" ...
## $ Contract        : chr   "One year" "Month-to-month" "Two Year" "Two Year" ...
## $ Port_modem      : chr   "Yes" "No" "Yes" "No" ...
## $ Tablet          : chr   "Yes" "Yes" "No" "No" ...
## $ InternetService : chr   "Fiber Optic" "Fiber Optic" "DSL" "DSL" ...
## $ Phone           : chr   "Yes" "Yes" "Yes" "Yes" ...
## $ Multiple        : chr   "No" "Yes" "Yes" "No" ...
## $ OnlineSecurity  : chr   "Yes" "Yes" "No" "Yes" ...
## $ OnlineBackup    : chr   "Yes" "No" "No" "No" ...
## $ DeviceProtection : chr   "No" "No" "No" "No" ...
## $ TechSupport     : chr   "No" "No" "No" "No" ...
## $ StreamingTV     : chr   "No" "Yes" "No" "Yes" ...
## $ StreamingMovies : chr   "Yes" "Yes" "Yes" "No" ...
## $ PaperlessBilling : chr   "Yes" "Yes" "Yes" "Yes" ...
## $ PaymentMethod   : chr   "Credit Card (automatic)" "Bank Transfer(automatic)" "Credit Card (automatic)" ...
## $ Tenure          : num   6.8 1.16 15.75 17.09 1.67 ...
## $ MonthlyCharge   : num   172 243 160 120 150 ...
## $ Bandwidth_GB_Year : num   905 801 2055 2165 271 ...
## $ Item1           : int   5 3 4 4 4 3 6 2 5 2 ...
## $ Item2           : int   5 4 4 4 4 3 5 2 4 2 ...
## $ Item3           : int   5 3 2 4 4 3 6 2 4 2 ...
## $ Item4           : int   3 3 4 2 3 2 4 5 3 2 ...
## $ Item5           : int   4 4 4 5 4 4 1 2 4 5 ...
## $ Item6           : int   4 3 3 4 4 3 5 3 3 2 ...
## $ Item7           : int   3 4 3 3 4 3 5 4 4 3 ...
## $ Item8           : int   4 4 3 3 5 3 5 5 4 3 ...
```

```
# Renaming unclear columns named Item1 through Item8 for improved readability &
# confirming they have been renamed correctly
```

```
df <- df %>%
  rename(
    Response = Item1,
    Fix = Item2,
    Replacement = Item3,
    Reliability = Item4,
    Options = Item5,
    Respectful = Item6,
    Courteous = Item7,
    Listening = Item8
```

```

)

colnames(df)

## [1] "CaseOrder"      "Customer_id"      "Interaction"
## [4] "UID"            "City"             "State"
## [7] "County"         "Zip"              "Lat"
## [10] "Lng"            "Population"       "Area"
## [13] "TimeZone"       "Job"              "Children"
## [16] "Age"            "Income"           "Marital"
## [19] "Gender"         "Churn"            "Outage_sec_perweek"
## [22] "Email"          "Contacts"         "Yearly_equip_failure"
## [25] "Techie"         "Contract"         "Port_modem"
## [28] "Tablet"         "InternetService"  "Phone"
## [31] "Multiple"       "OnlineSecurity"   "OnlineBackup"
## [34] "DeviceProtection" "TechSupport"      "StreamingTV"
## [37] "StreamingMovies" "PaperlessBilling" "PaymentMethod"
## [40] "Tenure"         "MonthlyCharge"    "Bandwidth_GB_Year"
## [43] "Response"       "Fix"              "Replacement"
## [46] "Reliability"    "Options"          "Respectful"
## [49] "Courteous"      "Listening"

# Several columns will not be useful in analysis and therefore will be dropped.
to_drop <- c('CaseOrder', 'Customer_id', 'Interaction', 'UID', 'City',
             'County', 'Zip', 'Lat', 'Lng', 'TimeZone', 'Job')

dfDropped = df[,(!(names(df) %in% to_drop)]
str(dfDropped)

## 'data.frame': 10000 obs. of 39 variables:
## $ State : chr "AK" "MI" "OR" "CA" ...
## $ Population : int 38 10446 3735 13863 11352 17701 2535 23144 17351 20193 ...
## $ Area : chr "Urban" "Urban" "Urban" "Suburban" ...
## $ Children : int 0 1 4 1 0 3 0 2 2 1 ...
## $ Age : int 68 27 50 48 83 83 79 30 49 86 ...
## $ Income : num 28562 21705 9610 18925 40074 ...
## $ Marital : chr "Widowed" "Married" "Widowed" "Married" ...
## $ Gender : chr "Male" "Female" "Female" "Male" ...
## $ Churn : chr "No" "Yes" "No" "No" ...
## $ Outage_sec_perweek : num 7.98 11.7 10.75 14.91 8.15 ...
## $ Email : int 10 12 9 15 16 15 10 16 20 18 ...
## $ Contacts : int 0 0 0 2 2 3 0 0 2 1 ...
## $ Yearly_equip_failure: int 1 1 1 0 1 1 1 0 3 0 ...
## $ Techie : chr "No" "Yes" "Yes" "Yes" ...
## $ Contract : chr "One year" "Month-to-month" "Two Year" "Two Year" ...
## $ Port_modem : chr "Yes" "No" "Yes" "No" ...
## $ Tablet : chr "Yes" "Yes" "No" "No" ...
## $ InternetService : chr "Fiber Optic" "Fiber Optic" "DSL" "DSL" ...
## $ Phone : chr "Yes" "Yes" "Yes" "Yes" ...
## $ Multiple : chr "No" "Yes" "Yes" "No" ...
## $ OnlineSecurity : chr "Yes" "Yes" "No" "Yes" ...
## $ OnlineBackup : chr "Yes" "No" "No" "No" ...
## $ DeviceProtection : chr "No" "No" "No" "No" ...
## $ TechSupport : chr "No" "No" "No" "No" ...
## $ StreamingTV : chr "No" "Yes" "No" "Yes" ...

```

```
## $ StreamingMovies      : chr "Yes" "Yes" "Yes" "No" ...
## $ PaperlessBilling     : chr "Yes" "Yes" "Yes" "Yes" ...
## $ PaymentMethod       : chr "Credit Card (automatic)" "Bank Transfer(automatic)" "Credit Card (aut
## $ Tenure               : num 6.8 1.16 15.75 17.09 1.67 ...
## $ MonthlyCharge       : num 172 243 160 120 150 ...
## $ Bandwidth_GB_Year   : num 905 801 2055 2165 271 ...
## $ Response            : int 5 3 4 4 4 3 6 2 5 2 ...
## $ Fix                 : int 5 4 4 4 4 3 5 2 4 2 ...
## $ Replacement         : int 5 3 2 4 4 3 6 2 4 2 ...
## $ Reliability         : int 3 3 4 2 3 2 4 5 3 2 ...
## $ Options             : int 4 4 4 5 4 4 1 2 4 5 ...
## $ Respectful          : int 4 3 3 4 4 3 5 3 3 2 ...
## $ Courteous           : int 3 4 3 3 4 3 5 4 4 3 ...
## $ Listening            : int 4 4 3 3 5 3 5 5 4 3 ...
```

```
# Creating dummy variables for categorical columns
```

```
dfReg <- dummy_cols(dfDropped, remove_selected_columns = TRUE)
names(dfReg) <- gsub(" ", "_", names(dfReg))
names(dfReg) <- gsub("-", "_", names(dfReg))
names(dfReg) <- gsub("[()]", "_", names(dfReg))
str(dfReg)
```

```
## 'data.frame': 10000 obs. of 118 variables:
## $ Population           : int 38 10446 3735 13863 11352 17701 2535 23144 17351 201
## $ Children             : int 0 1 4 1 0 3 0 2 2 1 ...
## $ Age                  : int 68 27 50 48 83 83 79 30 49 86 ...
## $ Income               : num 28562 21705 9610 18925 40074 ...
## $ Outage_sec_perweek   : num 7.98 11.7 10.75 14.91 8.15 ...
## $ Email                : int 10 12 9 15 16 15 10 16 20 18 ...
## $ Contacts             : int 0 0 0 2 2 3 0 0 2 1 ...
## $ Yearly_equip_failure : int 1 1 1 0 1 1 1 0 3 0 ...
## $ Tenure               : num 6.8 1.16 15.75 17.09 1.67 ...
## $ MonthlyCharge       : num 172 243 160 120 150 ...
## $ Bandwidth_GB_Year   : num 905 801 2055 2165 271 ...
## $ Response            : int 5 3 4 4 4 3 6 2 5 2 ...
## $ Fix                 : int 5 4 4 4 4 3 5 2 4 2 ...
## $ Replacement         : int 5 3 2 4 4 3 6 2 4 2 ...
## $ Reliability         : int 3 3 4 2 3 2 4 5 3 2 ...
## $ Options             : int 4 4 4 5 4 4 1 2 4 5 ...
## $ Respectful          : int 4 3 3 4 4 3 5 3 3 2 ...
## $ Courteous           : int 3 4 3 3 4 3 5 4 4 3 ...
## $ Listening            : int 4 4 3 3 5 3 5 5 4 3 ...
## $ State_AK            : int 1 0 0 0 0 0 0 0 0 0 ...
## $ State_AL            : int 0 0 0 0 0 0 0 0 0 0 ...
## $ State_AR            : int 0 0 0 0 0 0 0 0 0 0 ...
## $ State_AZ            : int 0 0 0 0 0 0 0 0 0 0 ...
## $ State_CA            : int 0 0 0 1 0 0 0 0 0 0 ...
## $ State_CO            : int 0 0 0 0 0 0 0 0 0 0 ...
## $ State_CT            : int 0 0 0 0 0 0 0 0 0 0 ...
## $ State_DC            : int 0 0 0 0 0 0 0 0 0 0 ...
## $ State_DE            : int 0 0 0 0 0 0 0 0 0 0 ...
## $ State_FL            : int 0 0 0 0 0 0 0 0 0 1 ...
## $ State_GA            : int 0 0 0 0 0 1 0 0 0 0 ...
## $ State_HI            : int 0 0 0 0 0 0 0 0 0 0 ...
## $ State_IA            : int 0 0 0 0 0 0 0 0 0 0 ...
```

## \$ State_ID	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_IL	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_IN	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_KS	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_KY	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_LA	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_MA	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_MD	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_ME	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_MI	: int	0 1 0 0 0 0 0 0 0 0 0 ...
## \$ State_MN	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_MO	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_MS	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_MT	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_NC	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_ND	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_NE	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_NH	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_NJ	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_NM	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_NV	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_NY	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_OH	: int	0 0 0 0 0 0 0 0 0 0 1 ...
## \$ State_OK	: int	0 0 0 0 0 0 0 0 1 0 0 ...
## \$ State_OR	: int	0 0 1 0 0 0 0 0 0 0 0 ...
## \$ State_PA	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_PR	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_RI	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_SC	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_SD	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_TN	: int	0 0 0 0 0 0 1 0 0 0 0 ...
## \$ State_TX	: int	0 0 0 0 1 0 0 0 0 0 0 ...
## \$ State_UT	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_VA	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_VT	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_WA	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_WI	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_WV	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ State_WY	: int	0 0 0 0 0 0 0 0 0 0 0 ...
## \$ Area_Rural	: int	0 0 0 0 0 0 0 0 0 0 1 ...
## \$ Area_Suburban	: int	0 0 0 1 1 0 1 1 1 0 ...
## \$ Area_Urban	: int	1 1 1 0 0 1 0 0 0 0 ...
## \$ Marital_Divorced	: int	0 0 0 0 0 0 0 0 0 0 ...
## \$ Marital_Married	: int	0 1 0 1 0 0 0 1 0 1 ...
## \$ Marital_Never_Married	: int	0 0 0 0 0 1 0 0 0 0 ...
## \$ Marital_Separated	: int	0 0 0 0 1 0 0 0 1 0 ...
## \$ Marital_Widowed	: int	1 0 1 0 0 0 1 0 0 0 ...
## \$ Gender_Female	: int	0 1 1 0 0 1 0 1 0 1 ...
## \$ Gender_Male	: int	1 0 0 1 1 0 1 0 0 0 ...
## \$ Gender_Nonbinary	: int	0 0 0 0 0 0 0 0 1 0 ...
## \$ Churn_No	: int	1 0 1 1 0 1 0 0 1 1 ...
## \$ Churn_Yes	: int	0 1 0 0 1 0 1 1 0 0 ...
## \$ Techie_No	: int	1 0 0 0 1 1 0 0 1 1 ...
## \$ Techie_Yes	: int	0 1 1 1 0 0 1 1 0 0 ...

```
## $ Contract_Month_to_month      : int  0 1 0 0 1 0 1 1 1 0 ...
## $ Contract_One_year            : int  1 0 0 0 0 1 0 0 0 0 ...
## $ Contract_Two_Year            : int  0 0 1 1 0 0 0 0 0 1 ...
## $ Port_modem_No                : int  0 1 0 1 0 0 1 1 0 0 ...
## $ Port_modem_Yes               : int  1 0 1 0 1 1 0 0 1 1 ...
## $ Tablet_No                    : int  0 0 1 1 1 1 1 1 1 1 ...
## $ Tablet_Yes                   : int  1 1 0 0 0 0 0 0 0 0 ...
## $ InternetService_DSL          : int  0 0 1 1 0 0 1 1 1 0 ...
## $ InternetService_Fiber_Optic : int  1 1 0 0 1 0 0 0 0 1 ...
## $ InternetService_None         : int  0 0 0 0 0 1 0 0 0 0 ...
## $ Phone_No                     : int  0 0 0 0 1 0 0 1 0 0 ...
## $ Phone_Yes                    : int  1 1 1 1 0 1 1 0 1 1 ...
## $ Multiple_No                  : int  1 0 0 1 1 0 1 1 1 1 ...
## [list output truncated]
```

```
dim(dfReg)
```

```
## [1] 10000  118
```

```
# Split dfReg into training and testing subsets
```

```
set.seed(22)
```

```
trainId = createDataPartition(dfReg$Churn_Yes, times = 1, p = 0.7, list = FALSE)
```

```
dfTrain = dfReg[trainId,]
```

```
dfTest = dfReg[-trainId,]
```

```
# Summary Statistics
```

```
summary(dfTrain)
```

```
##      Population      Children      Age      Income
## Min.   : 0.0      Min.   : 0.00      Min.   :18.00      Min.   : 643.2
## 1st Qu.: 733.8    1st Qu.: 0.00      1st Qu.:35.00      1st Qu.: 19349.8
## Median : 2867.0    Median : 1.00      Median :53.00      Median : 33304.9
## Mean   : 9683.6    Mean   : 2.06      Mean   :52.89      Mean   : 40080.0
## 3rd Qu.: 12933.0   3rd Qu.: 3.00      3rd Qu.:71.00      3rd Qu.: 53249.4
## Max.   :111850.0   Max.   :10.00      Max.   :89.00      Max.   :256998.4
## Outage_sec_perweek      Email      Contacts      Yearly_equip_failure
## Min.   : 0.09975      Min.   : 1.00      Min.   :0.000      Min.   :0.0000
## 1st Qu.: 8.02346      1st Qu.:10.00      1st Qu.:0.000      1st Qu.:0.0000
## Median : 9.98095      Median :12.00      Median :1.000      Median :0.0000
## Mean   : 9.97626      Mean   :12.02      Mean   :1.004      Mean   :0.3939
## 3rd Qu.:11.93163      3rd Qu.:14.00      3rd Qu.:2.000      3rd Qu.:1.0000
## Max.   :21.20723      Max.   :23.00      Max.   :7.000      Max.   :6.0000
##      Tenure      MonthlyCharge      Bandwidth_GB_Year      Response
## Min.   : 1.000      Min.   : 79.98      Min.   : 155.5      Min.   :1.00
## 1st Qu.: 7.851      1st Qu.:139.98      1st Qu.:1233.9      1st Qu.:3.00
## Median :28.095      Median :169.94      Median :3018.9      Median :3.00
## Mean   :34.282      Mean   :172.84      Mean   :3377.1      Mean   :3.49
## 3rd Qu.:61.354      3rd Qu.:202.44      3rd Qu.:5570.3      3rd Qu.:4.00
## Max.   :71.994      Max.   :290.16      Max.   :7159.0      Max.   :7.00
##      Fix      Replacement      Reliability      Options      Respectful
## Min.   :1.00      Min.   :1.000      Min.   :1.000      Min.   :1.000      Min.   :1.000
## 1st Qu.:3.00      1st Qu.:3.000      1st Qu.:3.000      1st Qu.:3.000      1st Qu.:3.000
## Median :4.00      Median :3.000      Median :4.000      Median :3.000      Median :4.000
## Mean   :3.51      Mean   :3.491      Mean   :3.507      Mean   :3.485      Mean   :3.516
```

##	3rd Qu.:4.00	3rd Qu.:4.000	3rd Qu.:4.000	3rd Qu.:4.000	3rd Qu.:4.000
##	Max. :7.00	Max. :8.000	Max. :7.000	Max. :7.000	Max. :8.000
##	Courteous	Listening	State_AK	State_AL	
##	Min. :1.000	Min. :1.000	Min. :0.000000	Min. :0.00000	
##	1st Qu.:3.000	1st Qu.:3.000	1st Qu.:0.000000	1st Qu.:0.00000	
##	Median :4.000	Median :4.000	Median :0.000000	Median :0.00000	
##	Mean :3.517	Mean :3.508	Mean :0.008143	Mean :0.01857	
##	3rd Qu.:4.000	3rd Qu.:4.000	3rd Qu.:0.000000	3rd Qu.:0.00000	
##	Max. :7.000	Max. :8.000	Max. :1.000000	Max. :1.00000	
##	State_AR	State_AZ	State_CA	State_CO	
##	Min. :0.00000	Min. :0.00000	Min. :0.00000	Min. :0.00000	
##	1st Qu.:0.00000	1st Qu.:0.00000	1st Qu.:0.00000	1st Qu.:0.00000	
##	Median :0.00000	Median :0.00000	Median :0.00000	Median :0.00000	
##	Mean :0.01843	Mean :0.01214	Mean :0.05486	Mean :0.01586	
##	3rd Qu.:0.00000	3rd Qu.:0.00000	3rd Qu.:0.00000	3rd Qu.:0.00000	
##	Max. :1.00000	Max. :1.00000	Max. :1.00000	Max. :1.00000	
##	State_CT	State_DC	State_DE	State_FL	
##	Min. :0.000000	Min. :0.000000	Min. :0.000000	Min. :0.000	
##	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000	
##	Median :0.000000	Median :0.000000	Median :0.000000	Median :0.000	
##	Mean :0.007286	Mean :0.001429	Mean :0.002143	Mean :0.033	
##	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000	
##	Max. :1.000000	Max. :1.000000	Max. :1.000000	Max. :1.000	
##	State_GA	State_HI	State_IA	State_ID	
##	Min. :0.00000	Min. :0.000	Min. :0.00000	Min. :0.000000	
##	1st Qu.:0.00000	1st Qu.:0.000	1st Qu.:0.00000	1st Qu.:0.000000	
##	Median :0.00000	Median :0.000	Median :0.00000	Median :0.000000	
##	Mean :0.02314	Mean :0.003	Mean :0.02714	Mean :0.008714	
##	3rd Qu.:0.00000	3rd Qu.:0.000	3rd Qu.:0.00000	3rd Qu.:0.000000	
##	Max. :1.00000	Max. :1.000	Max. :1.00000	Max. :1.000000	
##	State_IL	State_IN	State_KS	State_KY	
##	Min. :0.00000	Min. :0.00000	Min. :0.00000	Min. :0.00000	
##	1st Qu.:0.00000	1st Qu.:0.00000	1st Qu.:0.00000	1st Qu.:0.00000	
##	Median :0.00000	Median :0.00000	Median :0.00000	Median :0.00000	
##	Mean :0.04143	Mean :0.02414	Mean :0.01929	Mean :0.02357	
##	3rd Qu.:0.00000	3rd Qu.:0.00000	3rd Qu.:0.00000	3rd Qu.:0.00000	
##	Max. :1.00000	Max. :1.00000	Max. :1.00000	Max. :1.00000	
##	State_LA	State_MA	State_MD	State_ME	
##	Min. :0.00000	Min. :0.00000	Min. :0.000	Min. :0.00000	
##	1st Qu.:0.00000	1st Qu.:0.00000	1st Qu.:0.000	1st Qu.:0.00000	
##	Median :0.00000	Median :0.00000	Median :0.000	Median :0.00000	
##	Mean :0.01386	Mean :0.01586	Mean :0.012	Mean :0.01171	
##	3rd Qu.:0.00000	3rd Qu.:0.00000	3rd Qu.:0.000	3rd Qu.:0.00000	
##	Max. :1.00000	Max. :1.00000	Max. :1.000	Max. :1.00000	
##	State_MI	State_MN	State_MO	State_MS	
##	Min. :0.000	Min. :0.000	Min. :0.00000	Min. :0.00000	
##	1st Qu.:0.000	1st Qu.:0.000	1st Qu.:0.00000	1st Qu.:0.00000	
##	Median :0.000	Median :0.000	Median :0.00000	Median :0.00000	
##	Mean :0.028	Mean :0.027	Mean :0.02929	Mean :0.01229	
##	3rd Qu.:0.000	3rd Qu.:0.000	3rd Qu.:0.00000	3rd Qu.:0.00000	
##	Max. :1.000	Max. :1.000	Max. :1.00000	Max. :1.00000	
##	State_MT	State_NC	State_ND	State_NE	
##	Min. :0.000000	Min. :0.00000	Min. :0.000	Min. :0.00000	
##	1st Qu.:0.000000	1st Qu.:0.00000	1st Qu.:0.000	1st Qu.:0.00000	

##	Median :0.000000	Median :0.000000	Median :0.000	Median :0.000000
##	Mean :0.009429	Mean :0.02843	Mean :0.012	Mean :0.01986
##	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.000000	Max. :1.000	Max. :1.000000
##	State_NH	State_NJ	State_NM	State_NV
##	Min. :0.000000	Min. :0.00	Min. :0.000000	Min. :0.000000
##	1st Qu.:0.000000	1st Qu.:0.00	1st Qu.:0.000000	1st Qu.:0.000000
##	Median :0.000000	Median :0.00	Median :0.000000	Median :0.000000
##	Mean :0.008857	Mean :0.02	Mean :0.01157	Mean :0.005429
##	3rd Qu.:0.000000	3rd Qu.:0.00	3rd Qu.:0.000000	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.00	Max. :1.000000	Max. :1.000000
##	State_NY	State_OH	State_OK	State_OR
##	Min. :0.000000	Min. :0.000000	Min. :0.000	Min. :0.00
##	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000	1st Qu.:0.00
##	Median :0.000000	Median :0.000000	Median :0.000	Median :0.00
##	Mean :0.05471	Mean :0.03514	Mean :0.021	Mean :0.01
##	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000	3rd Qu.:0.00
##	Max. :1.000000	Max. :1.000000	Max. :1.000	Max. :1.00
##	State_PA	State_PR	State_RI	State_SC
##	Min. :0.000000	Min. :0.000	Min. :0.000	Min. :0.000000
##	1st Qu.:0.000000	1st Qu.:0.000	1st Qu.:0.000	1st Qu.:0.000000
##	Median :0.000000	Median :0.000	Median :0.000	Median :0.000000
##	Mean :0.05471	Mean :0.003	Mean :0.002	Mean :0.01314
##	3rd Qu.:0.000000	3rd Qu.:0.000	3rd Qu.:0.000	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.000	Max. :1.000	Max. :1.000000
##	State_SD	State_TN	State_TX	State_UT
##	Min. :0.000000	Min. :0.000000	Min. :0.000000	Min. :0.000000
##	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000
##	Median :0.000000	Median :0.000000	Median :0.000000	Median :0.000000
##	Mean :0.008571	Mean :0.01943	Mean :0.05971	Mean :0.006714
##	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.000000	Max. :1.000000	Max. :1.000000
##	State_VA	State_VT	State_WA	State_WI
##	Min. :0.000000	Min. :0.000	Min. :0.000000	Min. :0.000000
##	1st Qu.:0.000000	1st Qu.:0.000	1st Qu.:0.000000	1st Qu.:0.000000
##	Median :0.000000	Median :0.000	Median :0.000000	Median :0.000000
##	Mean :0.02857	Mean :0.007	Mean :0.01714	Mean :0.02329
##	3rd Qu.:0.000000	3rd Qu.:0.000	3rd Qu.:0.000000	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.000	Max. :1.000000	Max. :1.000000
##	State_WV	State_WY	Area_Rural	Area_Suburban
##	Min. :0.000	Min. :0.000	Min. :0.0000	Min. :0.0000
##	1st Qu.:0.000	1st Qu.:0.000	1st Qu.:0.0000	1st Qu.:0.0000
##	Median :0.000	Median :0.000	Median :0.0000	Median :0.0000
##	Mean :0.024	Mean :0.004	Mean :0.3391	Mean :0.3324
##	3rd Qu.:0.000	3rd Qu.:0.000	3rd Qu.:1.0000	3rd Qu.:1.0000
##	Max. :1.000	Max. :1.000	Max. :1.0000	Max. :1.0000
##	Area_Urban	Marital_Divorced	Marital_Married	Marital_Never_Married
##	Min. :0.0000	Min. :0.0000	Min. :0.0000	Min. :0.0000
##	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:0.0000
##	Median :0.0000	Median :0.0000	Median :0.0000	Median :0.0000
##	Mean :0.3284	Mean :0.2136	Mean :0.1889	Mean :0.1943
##	3rd Qu.:1.0000	3rd Qu.:0.0000	3rd Qu.:0.0000	3rd Qu.:0.0000
##	Max. :1.0000	Max. :1.0000	Max. :1.0000	Max. :1.0000
##	Marital_Separated	Marital_Widowed	Gender_Female	Gender_Male

##	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000	Min.	:0.000
##	1st Qu.	:0.0000	1st Qu.	:0.0000	1st Qu.	:0.0000	1st Qu.	:0.000
##	Median	:0.0000	Median	:0.0000	Median	:0.0000	Median	:0.000
##	Mean	:0.2024	Mean	:0.2009	Mean	:0.4981	Mean	:0.479
##	3rd Qu.	:0.0000	3rd Qu.	:0.0000	3rd Qu.	:1.0000	3rd Qu.	:1.000
##	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000	Max.	:1.000
##	Gender_Nonbinary		Churn_No		Churn_Yes		Techie_No	
##	Min.	:0.00000	Min.	:0.0000	Min.	:0.0000	Min.	:0.000
##	1st Qu.	:0.00000	1st Qu.	:0.0000	1st Qu.	:0.0000	1st Qu.	:1.000
##	Median	:0.00000	Median	:1.0000	Median	:0.0000	Median	:1.000
##	Mean	:0.02286	Mean	:0.7267	Mean	:0.2733	Mean	:0.833
##	3rd Qu.	:0.00000	3rd Qu.	:1.0000	3rd Qu.	:1.0000	3rd Qu.	:1.000
##	Max.	:1.00000	Max.	:1.0000	Max.	:1.0000	Max.	:1.000
##	Techie_Yes		Contract_Month_to_month		Contract_One_year		Contract_Two_Year	
##	Min.	:0.000	Min.	:0.000	Min.	:0.0000	Min.	:0.0000
##	1st Qu.	:0.000	1st Qu.	:0.000	1st Qu.	:0.0000	1st Qu.	:0.0000
##	Median	:0.000	Median	:1.000	Median	:0.0000	Median	:0.0000
##	Mean	:0.167	Mean	:0.548	Mean	:0.2051	Mean	:0.2469
##	3rd Qu.	:0.000	3rd Qu.	:1.000	3rd Qu.	:0.0000	3rd Qu.	:0.0000
##	Max.	:1.000	Max.	:1.000	Max.	:1.0000	Max.	:1.0000
##	Port_modem_No		Port_modem_Yes		Tablet_No		Tablet_Yes	
##	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000
##	1st Qu.	:0.0000	1st Qu.	:0.0000	1st Qu.	:0.0000	1st Qu.	:0.0000
##	Median	:1.0000	Median	:0.0000	Median	:1.0000	Median	:0.0000
##	Mean	:0.5173	Mean	:0.4827	Mean	:0.7063	Mean	:0.2937
##	3rd Qu.	:1.0000	3rd Qu.	:1.0000	3rd Qu.	:1.0000	3rd Qu.	:1.0000
##	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000
##	InternetService_DSL		InternetService_Fiber_Optic		InternetService_None			
##	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000		
##	1st Qu.	:0.0000	1st Qu.	:0.0000	1st Qu.	:0.0000		
##	Median	:0.0000	Median	:0.0000	Median	:0.0000		
##	Mean	:0.3544	Mean	:0.4344	Mean	:0.2111		
##	3rd Qu.	:1.0000	3rd Qu.	:1.0000	3rd Qu.	:0.0000		
##	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000		
##	Phone_No		Phone_Yes		Multiple_No		Multiple_Yes	
##	Min.	:0.00000	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000
##	1st Qu.	:0.00000	1st Qu.	:1.0000	1st Qu.	:0.0000	1st Qu.	:0.0000
##	Median	:0.00000	Median	:1.0000	Median	:1.0000	Median	:0.0000
##	Mean	:0.09586	Mean	:0.9041	Mean	:0.5407	Mean	:0.4593
##	3rd Qu.	:0.00000	3rd Qu.	:1.0000	3rd Qu.	:1.0000	3rd Qu.	:1.0000
##	Max.	:1.00000	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000
##	OnlineSecurity_No		OnlineSecurity_Yes		OnlineBackup_No		OnlineBackup_Yes	
##	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000
##	1st Qu.	:0.0000	1st Qu.	:0.0000	1st Qu.	:0.0000	1st Qu.	:0.0000
##	Median	:1.0000	Median	:0.0000	Median	:1.0000	Median	:0.0000
##	Mean	:0.6446	Mean	:0.3554	Mean	:0.5426	Mean	:0.4574
##	3rd Qu.	:1.0000	3rd Qu.	:1.0000	3rd Qu.	:1.0000	3rd Qu.	:1.0000
##	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000
##	DeviceProtection_No		DeviceProtection_Yes		TechSupport_No		TechSupport_Yes	
##	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000
##	1st Qu.	:0.0000	1st Qu.	:0.0000	1st Qu.	:0.0000	1st Qu.	:0.0000
##	Median	:1.0000	Median	:0.0000	Median	:1.0000	Median	:0.0000
##	Mean	:0.5626	Mean	:0.4374	Mean	:0.6237	Mean	:0.3763
##	3rd Qu.	:1.0000	3rd Qu.	:1.0000	3rd Qu.	:1.0000	3rd Qu.	:1.0000

```
## Max. :1.0000      Max. :1.0000      Max. :1.0000      Max. :1.0000
## StreamingTV_No StreamingTV_Yes StreamingMovies_No StreamingMovies_Yes
## Min. :0.000      Min. :0.000      Min. :0.0000      Min. :0.0000
## 1st Qu.:0.000      1st Qu.:0.000      1st Qu.:0.0000      1st Qu.:0.0000
## Median :1.000      Median :0.000      Median :1.0000      Median :0.0000
## Mean :0.503      Mean :0.497      Mean :0.5107      Mean :0.4893
## 3rd Qu.:1.000      3rd Qu.:1.000      3rd Qu.:1.0000      3rd Qu.:1.0000
## Max. :1.000      Max. :1.000      Max. :1.0000      Max. :1.0000
## PaperlessBilling_No PaperlessBilling_Yes
## Min. :0.000      Min. :0.000
## 1st Qu.:0.000      1st Qu.:0.000
## Median :0.000      Median :1.000
## Mean :0.414      Mean :0.586
## 3rd Qu.:1.000      3rd Qu.:1.000
## Max. :1.000      Max. :1.000
## PaymentMethod_Bank_Transfer_automatic_ PaymentMethod_Credit_Card__automatic_
## Min. :0.000      Min. :0.000
## 1st Qu.:0.000      1st Qu.:0.000
## Median :0.000      Median :0.000
## Mean :0.225      Mean :0.202
## 3rd Qu.:0.000      3rd Qu.:0.000
## Max. :1.000      Max. :1.000
## PaymentMethod_Electronic_Check PaymentMethod_Mailed_Check
## Min. :0.0000      Min. :0.0000
## 1st Qu.:0.0000      1st Qu.:0.0000
## Median :0.0000      Median :0.0000
## Mean :0.3399      Mean :0.2331
## 3rd Qu.:1.0000      3rd Qu.:0.0000
## Max. :1.0000      Max. :1.0000
```

```
# Normalize Function
normalize = function(x) {
  result = (x - min(x)) / (max(x) - min(x))
  return(result)
}

# Normalize training set
dfTrainNorm <- dfTrain
for (i in 1:19) {
  dfTrainNorm[i] <- normalize(dfTrainNorm[i])
}

# Normalize testing set
dfTestNorm <- dfTest

for (i in 1:19) {
  dfTestNorm[i] <- normalize(dfTestNorm[i])
}

# Summary Statistics
summary(dfTrainNorm)
```

```
## Population      Children      Age      Income
## Min. :0.00000      Min. :0.000      Min. :0.0000      Min. :0.00000
## 1st Qu.:0.00656      1st Qu.:0.000      1st Qu.:0.2394      1st Qu.:0.07297
## Median :0.02563      Median :0.100      Median :0.4930      Median :0.12741
```

##	Mean	:0.08658	Mean	:0.206	Mean	:0.4914	Mean	:0.15384
##	3rd Qu.	:0.11563	3rd Qu.	:0.300	3rd Qu.	:0.7465	3rd Qu.	:0.20521
##	Max.	:1.00000	Max.	:1.000	Max.	:1.0000	Max.	:1.00000
##	Outage_sec_perweek		Email		Contacts		Yearly_equip_failure	
##	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000	Min.	:0.00000
##	1st Qu.	:0.3754	1st Qu.	:0.4091	1st Qu.	:0.0000	1st Qu.	:0.00000
##	Median	:0.4681	Median	:0.5000	Median	:0.1429	Median	:0.00000
##	Mean	:0.4679	Mean	:0.5008	Mean	:0.1434	Mean	:0.06564
##	3rd Qu.	:0.5606	3rd Qu.	:0.5909	3rd Qu.	:0.2857	3rd Qu.	:0.16667
##	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000	Max.	:1.00000
##	Tenure		MonthlyCharge		Bandwidth_GB_Year		Response	
##	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000
##	1st Qu.	:0.0965	1st Qu.	:0.2855	1st Qu.	:0.1540	1st Qu.	:0.3333
##	Median	:0.3817	Median	:0.4280	Median	:0.4088	Median	:0.3333
##	Mean	:0.4688	Mean	:0.4418	Mean	:0.4600	Mean	:0.4150
##	3rd Qu.	:0.8501	3rd Qu.	:0.5827	3rd Qu.	:0.7732	3rd Qu.	:0.5000
##	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000
##	Fix		Replacement		Reliability		Options	
##	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000
##	1st Qu.	:0.3333	1st Qu.	:0.2857	1st Qu.	:0.3333	1st Qu.	:0.3333
##	Median	:0.5000	Median	:0.2857	Median	:0.5000	Median	:0.3333
##	Mean	:0.4183	Mean	:0.3558	Mean	:0.4179	Mean	:0.4141
##	3rd Qu.	:0.5000	3rd Qu.	:0.4286	3rd Qu.	:0.5000	3rd Qu.	:0.5000
##	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000
##	Respectful		Courteous		Listening		State_AK	
##	Min.	:0.0000	Min.	:0.0000	Min.	:0.0000	Min.	:0.000000
##	1st Qu.	:0.2857	1st Qu.	:0.3333	1st Qu.	:0.2857	1st Qu.	:0.000000
##	Median	:0.4286	Median	:0.5000	Median	:0.4286	Median	:0.000000
##	Mean	:0.3595	Mean	:0.4195	Mean	:0.3582	Mean	:0.008143
##	3rd Qu.	:0.4286	3rd Qu.	:0.5000	3rd Qu.	:0.4286	3rd Qu.	:0.000000
##	Max.	:1.0000	Max.	:1.0000	Max.	:1.0000	Max.	:1.000000
##	State_AL		State_AR		State_AZ		State_CA	
##	Min.	:0.00000	Min.	:0.00000	Min.	:0.00000	Min.	:0.00000
##	1st Qu.	:0.00000	1st Qu.	:0.00000	1st Qu.	:0.00000	1st Qu.	:0.00000
##	Median	:0.00000	Median	:0.00000	Median	:0.00000	Median	:0.00000
##	Mean	:0.01857	Mean	:0.01843	Mean	:0.01214	Mean	:0.05486
##	3rd Qu.	:0.00000	3rd Qu.	:0.00000	3rd Qu.	:0.00000	3rd Qu.	:0.00000
##	Max.	:1.00000	Max.	:1.00000	Max.	:1.00000	Max.	:1.00000
##	State_CO		State_CT		State_DC		State_DE	
##	Min.	:0.00000	Min.	:0.000000	Min.	:0.000000	Min.	:0.000000
##	1st Qu.	:0.00000	1st Qu.	:0.000000	1st Qu.	:0.000000	1st Qu.	:0.000000
##	Median	:0.00000	Median	:0.000000	Median	:0.000000	Median	:0.000000
##	Mean	:0.01586	Mean	:0.007286	Mean	:0.001429	Mean	:0.002143
##	3rd Qu.	:0.00000	3rd Qu.	:0.000000	3rd Qu.	:0.000000	3rd Qu.	:0.000000
##	Max.	:1.00000	Max.	:1.000000	Max.	:1.000000	Max.	:1.000000
##	State_FL		State_GA		State_HI		State_IA	
##	Min.	:0.000	Min.	:0.00000	Min.	:0.000	Min.	:0.00000
##	1st Qu.	:0.000	1st Qu.	:0.00000	1st Qu.	:0.000	1st Qu.	:0.00000
##	Median	:0.000	Median	:0.00000	Median	:0.000	Median	:0.00000
##	Mean	:0.033	Mean	:0.02314	Mean	:0.003	Mean	:0.02714
##	3rd Qu.	:0.000	3rd Qu.	:0.00000	3rd Qu.	:0.000	3rd Qu.	:0.00000
##	Max.	:1.000	Max.	:1.00000	Max.	:1.000	Max.	:1.00000
##	State_ID		State_IL		State_IN		State_KS	
##	Min.	:0.000000	Min.	:0.00000	Min.	:0.00000	Min.	:0.00000

##	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000
##	Median :0.000000	Median :0.000000	Median :0.000000	Median :0.000000
##	Mean :0.008714	Mean :0.04143	Mean :0.02414	Mean :0.01929
##	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.000000	Max. :1.000000	Max. :1.000000
##	State_KY	State_LA	State_MA	State_MD
##	Min. :0.000000	Min. :0.000000	Min. :0.000000	Min. :0.000000
##	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000
##	Median :0.000000	Median :0.000000	Median :0.000000	Median :0.000000
##	Mean :0.02357	Mean :0.01386	Mean :0.01586	Mean :0.012
##	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.000000	Max. :1.000000	Max. :1.000000
##	State_ME	State_MI	State_MN	State_MO
##	Min. :0.000000	Min. :0.000000	Min. :0.000000	Min. :0.000000
##	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000
##	Median :0.000000	Median :0.000000	Median :0.000000	Median :0.000000
##	Mean :0.01171	Mean :0.028	Mean :0.027	Mean :0.02929
##	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.000000	Max. :1.000000	Max. :1.000000
##	State_MS	State_MT	State_NC	State_ND
##	Min. :0.000000	Min. :0.000000	Min. :0.000000	Min. :0.000000
##	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000
##	Median :0.000000	Median :0.000000	Median :0.000000	Median :0.000000
##	Mean :0.01229	Mean :0.009429	Mean :0.02843	Mean :0.012
##	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.000000	Max. :1.000000	Max. :1.000000
##	State_NE	State_NH	State_NJ	State_NM
##	Min. :0.000000	Min. :0.000000	Min. :0.00	Min. :0.000000
##	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.00	1st Qu.:0.000000
##	Median :0.000000	Median :0.000000	Median :0.00	Median :0.000000
##	Mean :0.01986	Mean :0.008857	Mean :0.02	Mean :0.01157
##	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.00	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.000000	Max. :1.00	Max. :1.000000
##	State_NV	State_NY	State_OH	State_OK
##	Min. :0.000000	Min. :0.000000	Min. :0.000000	Min. :0.000000
##	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000
##	Median :0.000000	Median :0.000000	Median :0.000000	Median :0.000000
##	Mean :0.005429	Mean :0.05471	Mean :0.03514	Mean :0.021
##	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.000000	Max. :1.000000	Max. :1.000000
##	State_OR	State_PA	State_PR	State_RI
##	Min. :0.00	Min. :0.000000	Min. :0.0000	Min. :0.0000
##	1st Qu.:0.00	1st Qu.:0.000000	1st Qu.:0.0000	1st Qu.:0.0000
##	Median :0.00	Median :0.000000	Median :0.0000	Median :0.0000
##	Mean :0.01	Mean :0.05471	Mean :0.003	Mean :0.002
##	3rd Qu.:0.00	3rd Qu.:0.000000	3rd Qu.:0.0000	3rd Qu.:0.0000
##	Max. :1.00	Max. :1.000000	Max. :1.0000	Max. :1.0000
##	State_SC	State_SD	State_TN	State_TX
##	Min. :0.000000	Min. :0.000000	Min. :0.000000	Min. :0.000000
##	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000	1st Qu.:0.000000
##	Median :0.000000	Median :0.000000	Median :0.000000	Median :0.000000
##	Mean :0.01314	Mean :0.008571	Mean :0.01943	Mean :0.05971
##	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000	3rd Qu.:0.000000
##	Max. :1.000000	Max. :1.000000	Max. :1.000000	Max. :1.000000

##	State_UT	State_VA	State_VT	State_WA
##	Min. :0.000000	Min. :0.00000	Min. :0.000	Min. :0.00000
##	1st Qu.:0.000000	1st Qu.:0.00000	1st Qu.:0.000	1st Qu.:0.00000
##	Median :0.000000	Median :0.00000	Median :0.000	Median :0.00000
##	Mean :0.006714	Mean :0.02857	Mean :0.007	Mean :0.01714
##	3rd Qu.:0.000000	3rd Qu.:0.00000	3rd Qu.:0.000	3rd Qu.:0.00000
##	Max. :1.000000	Max. :1.00000	Max. :1.000	Max. :1.00000
##	State_WI	State_WV	State_WY	Area_Rural
##	Min. :0.00000	Min. :0.000	Min. :0.000	Min. :0.0000
##	1st Qu.:0.00000	1st Qu.:0.000	1st Qu.:0.000	1st Qu.:0.0000
##	Median :0.00000	Median :0.000	Median :0.000	Median :0.0000
##	Mean :0.02329	Mean :0.024	Mean :0.004	Mean :0.3391
##	3rd Qu.:0.00000	3rd Qu.:0.000	3rd Qu.:0.000	3rd Qu.:1.0000
##	Max. :1.00000	Max. :1.000	Max. :1.000	Max. :1.0000
##	Area_Suburban	Area_Urban	Marital_Divorced	Marital_Married
##	Min. :0.0000	Min. :0.0000	Min. :0.0000	Min. :0.0000
##	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:0.0000
##	Median :0.0000	Median :0.0000	Median :0.0000	Median :0.0000
##	Mean :0.3324	Mean :0.3284	Mean :0.2136	Mean :0.1889
##	3rd Qu.:1.0000	3rd Qu.:1.0000	3rd Qu.:0.0000	3rd Qu.:0.0000
##	Max. :1.0000	Max. :1.0000	Max. :1.0000	Max. :1.0000
##	Marital_Never_Married	Marital_Separated	Marital_Widowed	Gender_Female
##	Min. :0.0000	Min. :0.0000	Min. :0.0000	Min. :0.0000
##	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:0.0000
##	Median :0.0000	Median :0.0000	Median :0.0000	Median :0.0000
##	Mean :0.1943	Mean :0.2024	Mean :0.2009	Mean :0.4981
##	3rd Qu.:0.0000	3rd Qu.:0.0000	3rd Qu.:0.0000	3rd Qu.:1.0000
##	Max. :1.0000	Max. :1.0000	Max. :1.0000	Max. :1.0000
##	Gender_Male	Gender_Nonbinary	Churn_No	Churn_Yes
##	Min. :0.000	Min. :0.00000	Min. :0.0000	Min. :0.0000
##	1st Qu.:0.000	1st Qu.:0.00000	1st Qu.:0.0000	1st Qu.:0.0000
##	Median :0.000	Median :0.00000	Median :1.0000	Median :0.0000
##	Mean :0.479	Mean :0.02286	Mean :0.7267	Mean :0.2733
##	3rd Qu.:1.000	3rd Qu.:0.00000	3rd Qu.:1.0000	3rd Qu.:1.0000
##	Max. :1.000	Max. :1.00000	Max. :1.0000	Max. :1.0000
##	Techie_No	Techie_Yes	Contract_Month_to_month	Contract_One_year
##	Min. :0.000	Min. :0.000	Min. :0.000	Min. :0.0000
##	1st Qu.:1.000	1st Qu.:0.000	1st Qu.:0.000	1st Qu.:0.0000
##	Median :1.000	Median :0.000	Median :1.000	Median :0.0000
##	Mean :0.833	Mean :0.167	Mean :0.548	Mean :0.2051
##	3rd Qu.:1.000	3rd Qu.:0.000	3rd Qu.:1.000	3rd Qu.:0.0000
##	Max. :1.000	Max. :1.000	Max. :1.000	Max. :1.0000
##	Contract_Two_Year	Port_modem_No	Port_modem_Yes	Tablet_No
##	Min. :0.0000	Min. :0.0000	Min. :0.0000	Min. :0.0000
##	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:0.0000
##	Median :0.0000	Median :1.0000	Median :0.0000	Median :1.0000
##	Mean :0.2469	Mean :0.5173	Mean :0.4827	Mean :0.7063
##	3rd Qu.:0.0000	3rd Qu.:1.0000	3rd Qu.:1.0000	3rd Qu.:1.0000
##	Max. :1.0000	Max. :1.0000	Max. :1.0000	Max. :1.0000
##	Tablet_Yes	InternetService_DSL	InternetService_Fiber_Optic	
##	Min. :0.0000	Min. :0.0000	Min. :0.0000	
##	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:0.0000	
##	Median :0.0000	Median :0.0000	Median :0.0000	
##	Mean :0.2937	Mean :0.3544	Mean :0.4344	

```

## 3rd Qu.:1.0000 3rd Qu.:1.0000 3rd Qu.:1.0000
## Max. :1.0000 Max. :1.0000 Max. :1.0000
## InternetService_None Phone_No Phone_Yes Multiple_No
## Min. :0.0000 Min. :0.00000 Min. :0.0000 Min. :0.0000
## 1st Qu.:0.0000 1st Qu.:0.00000 1st Qu.:1.0000 1st Qu.:0.0000
## Median :0.0000 Median :0.00000 Median :1.0000 Median :1.0000
## Mean :0.2111 Mean :0.09586 Mean :0.9041 Mean :0.5407
## 3rd Qu.:0.0000 3rd Qu.:0.00000 3rd Qu.:1.0000 3rd Qu.:1.0000
## Max. :1.0000 Max. :1.00000 Max. :1.0000 Max. :1.0000
## Multiple_Yes OnlineSecurity_No OnlineSecurity_Yes OnlineBackup_No
## Min. :0.0000 Min. :0.0000 Min. :0.0000 Min. :0.0000
## 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.0000
## Median :0.0000 Median :1.0000 Median :0.0000 Median :1.0000
## Mean :0.4593 Mean :0.6446 Mean :0.3554 Mean :0.5426
## 3rd Qu.:1.0000 3rd Qu.:1.0000 3rd Qu.:1.0000 3rd Qu.:1.0000
## Max. :1.0000 Max. :1.0000 Max. :1.0000 Max. :1.0000
## OnlineBackup_Yes DeviceProtection_No DeviceProtection_Yes TechSupport_No
## Min. :0.0000 Min. :0.0000 Min. :0.0000 Min. :0.0000
## 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.0000
## Median :0.0000 Median :1.0000 Median :0.0000 Median :1.0000
## Mean :0.4574 Mean :0.5626 Mean :0.4374 Mean :0.6237
## 3rd Qu.:1.0000 3rd Qu.:1.0000 3rd Qu.:1.0000 3rd Qu.:1.0000
## Max. :1.0000 Max. :1.0000 Max. :1.0000 Max. :1.0000
## TechSupport_Yes StreamingTV_No StreamingTV_Yes StreamingMovies_No
## Min. :0.0000 Min. :0.000 Min. :0.000 Min. :0.0000
## 1st Qu.:0.0000 1st Qu.:0.000 1st Qu.:0.000 1st Qu.:0.0000
## Median :0.0000 Median :1.000 Median :0.000 Median :1.0000
## Mean :0.3763 Mean :0.503 Mean :0.497 Mean :0.5107
## 3rd Qu.:1.0000 3rd Qu.:1.000 3rd Qu.:1.000 3rd Qu.:1.0000
## Max. :1.0000 Max. :1.000 Max. :1.000 Max. :1.0000
## StreamingMovies_Yes PaperlessBilling_No PaperlessBilling_Yes
## Min. :0.0000 Min. :0.000 Min. :0.000
## 1st Qu.:0.0000 1st Qu.:0.000 1st Qu.:0.000
## Median :0.0000 Median :0.000 Median :1.000
## Mean :0.4893 Mean :0.414 Mean :0.586
## 3rd Qu.:1.0000 3rd Qu.:1.000 3rd Qu.:1.000
## Max. :1.0000 Max. :1.000 Max. :1.000
## PaymentMethod_Bank_Transfer_automatic_ PaymentMethod_Credit_Card__automatic_
## Min. :0.000 Min. :0.000
## 1st Qu.:0.000 1st Qu.:0.000
## Median :0.000 Median :0.000
## Mean :0.225 Mean :0.202
## 3rd Qu.:0.000 3rd Qu.:0.000
## Max. :1.000 Max. :1.000
## PaymentMethod_Electronic_Check PaymentMethod_Mailed_Check
## Min. :0.0000 Min. :0.0000
## 1st Qu.:0.0000 1st Qu.:0.0000
## Median :0.0000 Median :0.0000
## Mean :0.3399 Mean :0.2331
## 3rd Qu.:1.0000 3rd Qu.:0.0000
## Max. :1.0000 Max. :1.0000

```

```
# Export prepared data sets
```

```
write.csv(dfTrainNorm, "C:\\WGU\\D209 Data Mining I\\PA Task 1\\D209_dfTrainNorm.csv", row.names = FALSE)
```

```

write.csv(dfTestNorm, "C:\\WGU\\D209 Data Mining I\\PA Task 1\\D209_dfTestNorm.csv", row.names = FALSE)

dfTrainChurn <- dfTrainNorm[, -which(names(dfTrainNorm) %in% c("Churn_No"))]
dfTestChurn <- dfTestNorm[, -which(names(dfTestNorm) %in% c("Churn_No"))]

# Finding the best k-value for accuracy

# Initial guess
kVal <- sqrt(nrow(dfTrainChurn)) # kval = 83.666

predFloor <- knn(train = dfTrainChurn, test = dfTestChurn, cl = dfTrainChurn$Churn_Yes, k = floor(kVal))
outcomesFloor <- table(predFloor, dfTestChurn$Churn_Yes)
confFloor <- confusionMatrix(outcomesFloor)
confFloor # Accuracy 0.94

## Confusion Matrix and Statistics
##
##
## predFloor    0    1
##           0 2263  180
##           1    0  557
##
##               Accuracy : 0.94
##               95% CI : (0.9309, 0.9482)
##      No Information Rate : 0.7543
##      P-Value [Acc > NIR] : < 2.2e-16
##
##               Kappa : 0.8236
##
##  Mcnemar's Test P-Value : < 2.2e-16
##
##           Sensitivity : 1.0000
##           Specificity : 0.7558
##      Pos Pred Value : 0.9263
##      Neg Pred Value : 1.0000
##           Prevalence : 0.7543
##      Detection Rate : 0.7543
##  Detection Prevalence : 0.8143
##      Balanced Accuracy : 0.8779
##
##      'Positive' Class : 0
##
predCeil <- knn(train = dfTrainChurn, test = dfTestChurn, cl = dfTrainChurn$Churn_Yes, k = ceiling(kVal))
outcomesCeil <- table(predCeil, dfTestChurn$Churn_Yes)
confCeil <- confusionMatrix(outcomesCeil)
confCeil # Accuracy 0.939

## Confusion Matrix and Statistics
##
##
## predCeil     0    1
##           0 2263  186
##           1    0  551

```



```

##
##          Accuracy : 0.938
##          95% CI : (0.9288, 0.9464)
##    No Information Rate : 0.7543
##    P-Value [Acc > NIR] : < 2.2e-16
##
##          Kappa : 0.8172
##
##    McNemar's Test P-Value : < 2.2e-16
##
##          Sensitivity : 1.0000
##          Specificity : 0.7476
##          Pos Pred Value : 0.9241
##          Neg Pred Value : 1.0000
##          Prevalence : 0.7543
##          Detection Rate : 0.7543
##    Detection Prevalence : 0.8163
##          Balanced Accuracy : 0.8738
##
##    'Positive' Class : 0
##
# Checking other values
pred100 <- knn(train = dfTrainChurn, test = dfTestChurn, cl = dfTrainChurn$Churn_Yes, k = 100)
outcomes100 <- table(pred100, dfTestChurn$Churn_Yes)
conf100 <- confusionMatrix(outcomes100)
conf100 # Accuracy 0.938

## Confusion Matrix and Statistics
##
##
## pred100    0    1
##      0 2262  188
##      1    1  549
##
##          Accuracy : 0.937
##          95% CI : (0.9277, 0.9454)
##    No Information Rate : 0.7543
##    P-Value [Acc > NIR] : < 2.2e-16
##
##          Kappa : 0.8141
##
##    McNemar's Test P-Value : < 2.2e-16
##
##          Sensitivity : 0.9996
##          Specificity : 0.7449
##          Pos Pred Value : 0.9233
##          Neg Pred Value : 0.9982
##          Prevalence : 0.7543
##          Detection Rate : 0.7540
##    Detection Prevalence : 0.8167
##          Balanced Accuracy : 0.8722
##
##    'Positive' Class : 0
##

```

```

pred75 <- knn(train = dfTrainChurn, test = dfTestChurn, cl = dfTrainChurn$Churn_Yes, k = 75)
outcomes75 <- table(pred75, dfTestChurn$Churn_Yes)
conf75 <- confusionMatrix(outcomes75)
conf75 # Accuracy 0.9403

```

```

## Confusion Matrix and Statistics
##
##
## pred75      0      1
##      0 2263   179
##      1      0  558
##
##              Accuracy : 0.9403
##              95% CI : (0.9313, 0.9485)
##      No Information Rate : 0.7543
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.8247
##
##  Mcnemar's Test P-Value : < 2.2e-16
##
##              Sensitivity : 1.0000
##              Specificity : 0.7571
##              Pos Pred Value : 0.9267
##              Neg Pred Value : 1.0000
##              Prevalence : 0.7543
##              Detection Rate : 0.7543
##      Detection Prevalence : 0.8140
##              Balanced Accuracy : 0.8786
##
##              'Positive' Class : 0
##

```

```

pred50 <- knn(train = dfTrainChurn, test = dfTestChurn, cl = dfTrainChurn$Churn_Yes, k = 50)
outcomes50 <- table(pred50, dfTestChurn$Churn_Yes)
conf50 <- confusionMatrix(outcomes50)
conf50 # Accuracy 0.9513

```

```

## Confusion Matrix and Statistics
##
##
## pred50      0      1
##      0 2261   148
##      1      2  589
##
##              Accuracy : 0.95
##              95% CI : (0.9416, 0.9575)
##      No Information Rate : 0.7543
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.8554
##
##  Mcnemar's Test P-Value : < 2.2e-16
##

```

```

##          Sensitivity : 0.9991
##          Specificity : 0.7992
##          Pos Pred Value : 0.9386
##          Neg Pred Value : 0.9966
##          Prevalence : 0.7543
##          Detection Rate : 0.7537
##          Detection Prevalence : 0.8030
##          Balanced Accuracy : 0.8992
##
##          'Positive' Class : 0
##

pred25 <- knn(train = dfTrainChurn, test = dfTestChurn, cl = dfTrainChurn$Churn_Yes, k = 25)
outcomes25 <- table(pred25, dfTestChurn$Churn_Yes)
conf25 <- confusionMatrix(outcomes25)
conf25 # Accuracy 0.948

## Confusion Matrix and Statistics
##
##
## pred25    0    1
##      0 2258  151
##      1    5  586
##
##          Accuracy : 0.948
##          95% CI : (0.9394, 0.9557)
##          No Information Rate : 0.7543
##          P-Value [Acc > NIR] : < 2.2e-16
##
##          Kappa : 0.8497
##
##          Mcnemar's Test P-Value : < 2.2e-16
##
##          Sensitivity : 0.9978
##          Specificity : 0.7951
##          Pos Pred Value : 0.9373
##          Neg Pred Value : 0.9915
##          Prevalence : 0.7543
##          Detection Rate : 0.7527
##          Detection Prevalence : 0.8030
##          Balanced Accuracy : 0.8965
##
##          'Positive' Class : 0
##

# k=50 gave the best for a random guess. Checking the 10 on either side of 50.

## The code below was borrowed in part from LearnByMarketing.com
## https://www.learnbymarketing.com/tutorials/k-nearest-neighbors-in-r-example/#::~text=K-Nearest-Neighbors

churn_acc <- numeric() #Holding variable

# This takes about 2.5-3 minutes to complete.
for(i in 40:60){
  #Apply knn with k = i

```

```

predict <- knn(dfTrainChurn,dfTestChurn,dfTrainChurn$Churn_Yes,k=i)
churn_acc <- c(churn_acc, mean(predict==dfTestChurn$Churn_Yes))
}

## end third party block
knnAcc <- tibble(x=40:60, y=churn_acc)
knnAcc[which.max(knnAcc$y),]

## # A tibble: 1 x 2
##       x     y
##   <int> <dbl>
## 1    43 0.953

# k=43 is the best for accuracy.
pred43 <- knn(train = dfTrainChurn, test = dfTestChurn, cl = dfTrainChurn$Churn_Yes, k = 43, prob = TRUE)
outcomes43 <- table(pred43, dfTestChurn$Churn_Yes)
conf43 <- confusionMatrix(outcomes43)
conf43 # Accuracy 0.9527

## Confusion Matrix and Statistics
##
##
## pred43      0      1
##      0 2261  140
##      1      2  597
##
##              Accuracy : 0.9527
##              95% CI : (0.9444, 0.96)
##      No Information Rate : 0.7543
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.8637
##
##      Mcnemar's Test P-Value : < 2.2e-16
##
##              Sensitivity : 0.9991
##              Specificity : 0.8100
##      Pos Pred Value : 0.9417
##      Neg Pred Value : 0.9967
##      Prevalence : 0.7543
##      Detection Rate : 0.7537
##      Detection Prevalence : 0.8003
##      Balanced Accuracy : 0.9046
##
##      'Positive' Class : 0
##

pred43Numeric <- as.numeric(as.character(pred43))

rocPlot <- roc(response = dfTestChurn$Churn_Yes,
               predictor = pred43Numeric,
               percent = TRUE,
               ci=TRUE,
               boot.n=100,
               ci.alpha=-0.9,

```

```
stratified=FALSE,  
plot=TRUE,  
grid=TRUE,  
print.auc=TRUE,  
show.thres=TRUE)
```

```
## Setting levels: control = 0, case = 1
```

```
## Setting direction: controls < cases
```

