

R Notebook

1. DEFINING THE QUESTION

a) Specifying the Question

Determining which individuals are most likely to click on the ads.

b) Defining the Metrics of Success

Performing the Exploratory Data Analysis.

c) Understanding the context

Determining the audience the entrepreneur can target.

d) Recording the Experimental Design

1. Defining the question, the metric for success, the context and experimental design.
2. Loading and exploring the dataset.
3. Finding and dealing with outliers, anomalies, and missing data within the dataset.
4. Perform univariate and bivariate analysis.
5. Giving a conclusion and recommendation.

e) Relevance of the data

The data used in this project is for determining which audience should be targeted by the entrepreneur. The dataset link: (<http://bit.ly/IPAdvertisingData>)

2. DATA ANALYSIS

a) Checking the Data

```
library(data.table)
```

```
library(ggplot2)
```

```
library(magrittr)
```

```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:data.table':
##
##   between, first, last

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
# Reading the data
```

```
df <- fread('http://bit.ly/IPAdvertisingData')
df
```

```
##      Daily Time Spent on Site  Age Area Income Daily Internet Usage
##                                <num> <int>          <num>          <num>
##  1:                68.95      35      61833.90          256.09
##  2:                80.23      31      68441.85          193.77
##  3:                69.47      26      59785.94          236.50
##  4:                74.15      29      54806.18          245.89
##  5:                68.37      35      73889.99          225.58
## ---
## 996:                72.97      30      71384.57          208.58
## 997:                51.30      45      67782.17          134.42
## 998:                51.63      51      42415.72          120.37
## 999:                55.55      19      41920.79          187.95
## 1000:                45.01      26      29875.80          178.35
##
##              Ad Topic Line              City Male
##              <char>          <char> <int>
##  1:  Cloned 5thgeneration orchestration  Wrightburgh  0
##  2:  Monitored national standardization  West Jodi    1
##  3:  Organic bottom-line service-desk    Davidton    0
##  4:  Triple-buffered reciprocal time-frame West Terrifurt  1
##  5:  Robust logistical utilization      South Manuel   0
## ---
## 996:  Fundamental modular algorithm      Duffystad    1
## 997:  Grass-roots cohesive monitoring     New Darlene    1
## 998:  Expanded intangible solution       South Jessica  1
## 999:  Proactive bandwidth-monitored policy West Steven  0
## 1000: Virtual 5thgeneration emulation     Ronniemouth    0
##
##      Country      Timestamp Clicked on Ad
##      <char>          <POSct>          <int>
##  1:  Tunisia 2016-03-27 00:53:11          0
##  2:  Nauru 2016-04-04 01:39:02          0
##  3:  San Marino 2016-03-13 20:35:42          0
```

```
## 4: Italy 2016-01-10 02:31:19 0
## 5: Iceland 2016-06-03 03:36:18 0
## ---
## 996: Lebanon 2016-02-11 21:49:00 1
## 997: Bosnia and Herzegovina 2016-04-22 02:07:01 1
## 998: Mongolia 2016-02-01 17:24:57 1
## 999: Guatemala 2016-03-24 02:35:54 0
## 1000: Brazil 2016-06-03 21:43:21 1
```

```
# Viewing the dataset
```

```
View(df)
```

```
# Viewing the column names
```

```
colnames(df)
```

```
## [1] "Daily Time Spent on Site" "Age"
## [3] "Area Income" "Daily Internet Usage"
## [5] "Ad Topic Line" "City"
## [7] "Male" "Country"
## [9] "Timestamp" "Clicked on Ad"
```

```
# Previewing the dataset
```

```
class(df)
```

```
## [1] "data.table" "data.frame"
```

```
# Previewing the top of the dataset
```

```
head(df)
```

```
## Daily Time Spent on Site Age Area Income Daily Internet Usage
## <num> <int> <num> <num>
## 1: 68.95 35 61833.90 256.09
## 2: 80.23 31 68441.85 193.77
## 3: 69.47 26 59785.94 236.50
## 4: 74.15 29 54806.18 245.89
## 5: 68.37 35 73889.99 225.58
## 6: 59.99 23 59761.56 226.74
## Ad Topic Line City Male Country
## <char> <char> <int> <char>
## 1: Cloned 5thgeneration orchestration Wrightburgh 0 Tunisia
## 2: Monitored national standardization West Jodi 1 Nauru
## 3: Organic bottom-line service-desk Davidton 0 San Marino
## 4: Triple-buffered reciprocal time-frame West Terrifurt 1 Italy
## 5: Robust logistical utilization South Manuel 0 Iceland
## 6: Sharable client-driven software Jamieberg 1 Norway
## Timestamp Clicked on Ad
## <POSct> <int>
## 1: 2016-03-27 00:53:11 0
## 2: 2016-04-04 01:39:02 0
## 3: 2016-03-13 20:35:42 0
## 4: 2016-01-10 02:31:19 0
## 5: 2016-06-03 03:36:18 0
## 6: 2016-05-19 14:30:17 0
```

```
# Previewing the bottom of the dataset
tail(df)
```

```
##      Daily Time Spent on Site   Age Area Income Daily Internet Usage
##                               <num> <int>          <num>              <num>
## 1:                          43.70   28    63126.96              173.01
## 2:                          72.97   30    71384.57              208.58
## 3:                          51.30   45    67782.17              134.42
## 4:                          51.63   51    42415.72              120.37
## 5:                          55.55   19    41920.79              187.95
## 6:                          45.01   26    29875.80              178.35
##                               Ad Topic Line      City Male
##                               <char>          <char> <int>
## 1:      Front-line bifurcated ability Nicholasland      0
## 2:      Fundamental modular algorithm  Duffystad      1
## 3:      Grass-roots cohesive monitoring New Darlene      1
## 4:      Expanded intangible solution South Jessica      1
## 5: Proactive bandwidth-monitored policy West Steven      0
## 6:      Virtual 5thgeneration emulation Ronniemouth      0
##                               Country      Timestamp Clicked on Ad
##                               <char>          <POSc>      <int>
## 1:      Mayotte 2016-04-04 03:57:48      1
## 2:      Lebanon 2016-02-11 21:49:00      1
## 3: Bosnia and Herzegovina 2016-04-22 02:07:01      1
## 4:      Mongolia 2016-02-01 17:24:57      1
## 5:      Guatemala 2016-03-24 02:35:54      0
## 6:      Brazil 2016-06-03 21:43:21      1
```

```
# Checking the shape of the dataset
dim(df)
```

```
## [1] 1000   10
```

1000 rows and 10 columns

b) Data Cleaning

Missing Values

```
# Checking for missing values
sum(is.na(df))
```

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

There are no missing values.

```
# Removing all rows with na
na.omit(df)
```

```
##      Daily Time Spent on Site   Age Area Income Daily Internet Usage
##                                <num> <int>          <num>          <num>
##    1:                        68.95   35    61833.90            256.09
##    2:                        80.23   31    68441.85            193.77
##    3:                        69.47   26    59785.94            236.50
##    4:                        74.15   29    54806.18            245.89
##    5:                        68.37   35    73889.99            225.58
##    ---
##  996:                        72.97   30    71384.57            208.58
##  997:                        51.30   45    67782.17            134.42
##  998:                        51.63   51    42415.72            120.37
##  999:                        55.55   19    41920.79            187.95
## 1000:                        45.01   26    29875.80            178.35
##                                Ad Topic Line          City Male
##                                <char>          <char> <int>
##    1:      Cloned 5thgeneration orchestration    Wrightburgh    0
##    2:      Monitored national standardization    West Jodi      1
##    3:      Organic bottom-line service-desk      Davidton      0
##    4:      Triple-buffered reciprocal time-frame West Terrifurt    1
##    5:      Robust logistical utilization         South Manuel     0
##    ---
##  996:      Fundamental modular algorithm         Duffystad      1
##  997:      Grass-roots cohesive monitoring       New Darlene     1
##  998:      Expanded intangible solution         South Jessica   1
##  999:      Proactive bandwidth-monitored policy West Steven     0
## 1000:      Virtual 5thgeneration emulation      Ronniemouth     0
##                                Country      Timestamp Clicked on Ad
##                                <char>          <POSc>          <int>
##    1:      Tunisia 2016-03-27 00:53:11            0
##    2:      Nauru 2016-04-04 01:39:02              0
##    3:      San Marino 2016-03-13 20:35:42          0
##    4:      Italy 2016-01-10 02:31:19              0
##    5:      Iceland 2016-06-03 03:36:18            0
##    ---
##  996:      Lebanon 2016-02-11 21:49:00            1
##  997:      Bosnia and Herzegovina 2016-04-22 02:07:01  1
##  998:      Mongolia 2016-02-01 17:24:57           1
##  999:      Guatemala 2016-03-24 02:35:54          0
## 1000:      Brazil 2016-06-03 21:43:21             1
```

Duplicates

```
# Checking for duplicates
duplicated_rows <- df[duplicated(df),]
duplicated_rows
```

```
## Empty data.table (0 rows and 10 cols): Daily Time Spent on Site, Age, Area Income, Daily Internet Usage
```

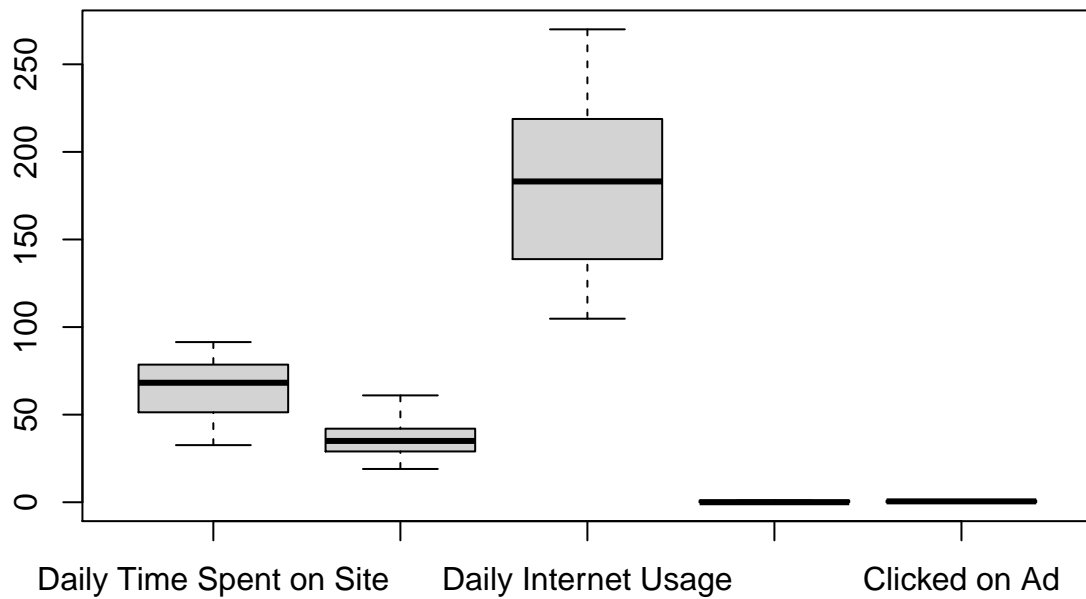
There are no duplicates

```
# Displaying the unique items and assigning unique_items variable
unique_items <- df[!duplicated(df), ]
unique_items
```

```
##      Daily Time Spent on Site   Age Area Income Daily Internet Usage
##              <num> <int>          <num>          <num>
##    1:          68.95    35    61833.90          256.09
##    2:          80.23    31    68441.85          193.77
##    3:          69.47    26    59785.94          236.50
##    4:          74.15    29    54806.18          245.89
##    5:          68.37    35    73889.99          225.58
##    ---
##  996:          72.97    30    71384.57          208.58
##  997:          51.30    45    67782.17          134.42
##  998:          51.63    51    42415.72          120.37
##  999:          55.55    19    41920.79          187.95
## 1000:          45.01    26    29875.80          178.35
##
##              Ad Topic Line              City Male
##              <char>          <char> <int>
##    1:    Cloned 5thgeneration orchestration    Wrightburgh    0
##    2:    Monitored national standardization    West Jodi    1
##    3:    Organic bottom-line service-desk      Davidton    0
##    4:    Triple-buffered reciprocal time-frame West Terrifurt    1
##    5:    Robust logistical utilization    South Manuel    0
##    ---
##  996:    Fundamental modular algorithm    Duffystad    1
##  997:    Grass-roots cohesive monitoring    New Darlene    1
##  998:    Expanded intangible solution    South Jessica    1
##  999:    Proactive bandwidth-monitored policy    West Steven    0
## 1000:    Virtual 5thgeneration emulation    Ronniemouth    0
##
##              Country              Timestamp Clicked on Ad
##              <char>          <POSct>          <int>
##    1:    Tunisia 2016-03-27 00:53:11          0
##    2:    Nauru 2016-04-04 01:39:02          0
##    3:    San Marino 2016-03-13 20:35:42          0
##    4:    Italy 2016-01-10 02:31:19          0
##    5:    Iceland 2016-06-03 03:36:18          0
##    ---
##  996:    Lebanon 2016-02-11 21:49:00          1
##  997:    Bosnia and Herzegovina 2016-04-22 02:07:01          1
##  998:    Mongolia 2016-02-01 17:24:57          1
##  999:    Guatemala 2016-03-24 02:35:54          0
## 1000:    Brazil 2016-06-03 21:43:21          1
```

Outliers

```
# Visualizing outliers using boxplot
df1 <- subset(df, select = c("Daily Time Spent on Site", "Age", "Daily Internet Usage", "Male", "Clicked"))
boxplot(df1)
```



```
# Renaming columns
```

```
df1 <- df1 %>% rename(Daily_Time_Spent_on_Site = "Daily Time Spent on Site")
```

```
df1 <- df1 %>% rename(Daily_Internet_Usage = "Daily Internet Usage")
```

```
df1 <- df1 %>% rename(Clicked_on_Ad = "Clicked on Ad")
```

```
df1
```

```
##      Daily_Time_Spent_on_Site  Age Daily_Internet_Usage  Male Clicked_on_Ad
##      <num> <int>          <num> <int>          <int>
##  1:      68.95   35      256.09    0              0
##  2:      80.23   31      193.77    1              0
##  3:      69.47   26      236.50    0              0
##  4:      74.15   29      245.89    1              0
##  5:      68.37   35      225.58    0              0
##  ---
## 996:      72.97   30      208.58    1              1
## 997:      51.30   45      134.42    1              1
## 998:      51.63   51      120.37    1              1
## 999:      55.55   19      187.95    0              0
##1000:      45.01   26      178.35    0              1
```

3. BIVARIATE AND UNIVARIATE ANALYSIS

a) Univariate Analysis

Measures of Central Tendency

```
# Summary statistics of the dataset
summary(df1)
```

```
##   Daily_Time_Spent_on_Site      Age      Daily_Internet_Usage      Male
##   Min.      :32.60      Min.      :19.00      Min.      :104.8      Min.      :0.000
##   1st Qu.:51.36      1st Qu.:29.00      1st Qu.:138.8      1st Qu.:0.000
##   Median :68.22      Median :35.00      Median :183.1      Median :0.000
##   Mean   :65.00      Mean   :36.01      Mean   :180.0      Mean   :0.481
##   3rd Qu.:78.55      3rd Qu.:42.00      3rd Qu.:218.8      3rd Qu.:1.000
##   Max.   :91.43      Max.   :61.00      Max.   :270.0      Max.   :1.000
##   Clicked_on_Ad
##   Min.      :0.0
##   1st Qu.:0.0
##   Median :0.5
##   Mean   :0.5
##   3rd Qu.:1.0
##   Max.   :1.0
```

```
# Median of age
df1.Age.median <- median(df$Age)
df1.Age.median
```

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

```
# Mean of age
df1.Age.mean <- mean(df$Age)
df1.Age.mean
```

```
## [1] 36.009
```

```
# Mode of age
getmode <- function(v) {
  uniqv <- unique(v)
  uniqv[which.max(tabulate(match(v, uniqv)))]
}
```

```
df1.Age.mode <- getmode(df$Age)
df1.Age.mode
```

```
## [1] 31
```

Measures of Dispersion


```
# Displaying the column names  
colnames(df1)
```

```
## [1] "Daily_Time_Spent_on_Site" "Age"  
## [3] "Daily_Internet_Usage"      "Male"  
## [5] "Clicked_on_Ad"
```

```
# Minimum code of Daily Time Spent on Site  
df1.Daily_Time_Spent_on_Site.min <- min(df1$Daily_Time_Spent_on_Site)  
df1.Daily_Time_Spent_on_Site.min
```

```
## [1] 32.6
```

```
# Minimum code of Daily Internet Usage  
df1.Daily_Internet_Usage.min <- min(df1$Daily_Internet_Usage)  
df1.Daily_Internet_Usage.min
```

```
## [1] 104.78
```

```
# Minimum code of Age  
df1.Age.min <- min(df1$Age)  
df1.Age.min
```

```
## [1] 19
```

```
# Maximum code of age  
df1.Age.max <- max(df1$Age)  
df1.Age.max
```

```
## [1] 61
```

```
# Maximum code of Daily Internet Usage  
df1.Daily_Internet_Usage.max <- max(df1$Daily_Internet_Usage)  
df1.Daily_Internet_Usage.max
```

```
## [1] 269.96
```

```
# Maximum code of Daily Time Spent on Site  
df1.Daily_Time_Spent_on_Site.max <- max(df1$Daily_Time_Spent_on_Site)  
df1.Daily_Time_Spent_on_Site.max
```

```
## [1] 91.43
```

```
# Range code of age  
df1.Age.range <- range(df1$Age)  
df1.Age.range
```

```
## [1] 19 61
```

```
# Range code of Daily Time Spent on Site
df1.Daily_Time_Spent_on_Site.range <- range(df1$Daily_Time_Spent_on_Site)
df1.Daily_Time_Spent_on_Site.range
```

```
## [1] 32.60 91.43
```

```
# Quantile code of Age
df1.Age.quantile <- quantile(df1$Age)
df1.Age.quantile
```

```
##    0%   25%   50%   75%  100%
##   19   29   35   42   61
```

```
# Quantile code of Daily Time Spent on Site
df1.Daily_Time_Spent_on_Site.quantile <- quantile(df1$Daily_Time_Spent_on_Site)
df1.Daily_Time_Spent_on_Site.quantile
```

```
##      0%      25%      50%      75%      100%
## 32.6000 51.3600 68.2150 78.5475 91.4300
```

```
# Variance code of Age
df1.Age.variance <- var(df1$Age)
df1.Age.variance
```

```
## [1] 77.18611
```

```
# Variance code of Daily Time Spent on Site
df1.Daily_Time_Spent_on_Site.variance <- var(df1$Daily_Time_Spent_on_Site)
df1.Daily_Time_Spent_on_Site.variance
```

```
## [1] 251.3371
```

```
# Standard deviation code of age
df1.Age.sd <- sd(df1$Age)
df1.Age.sd
```

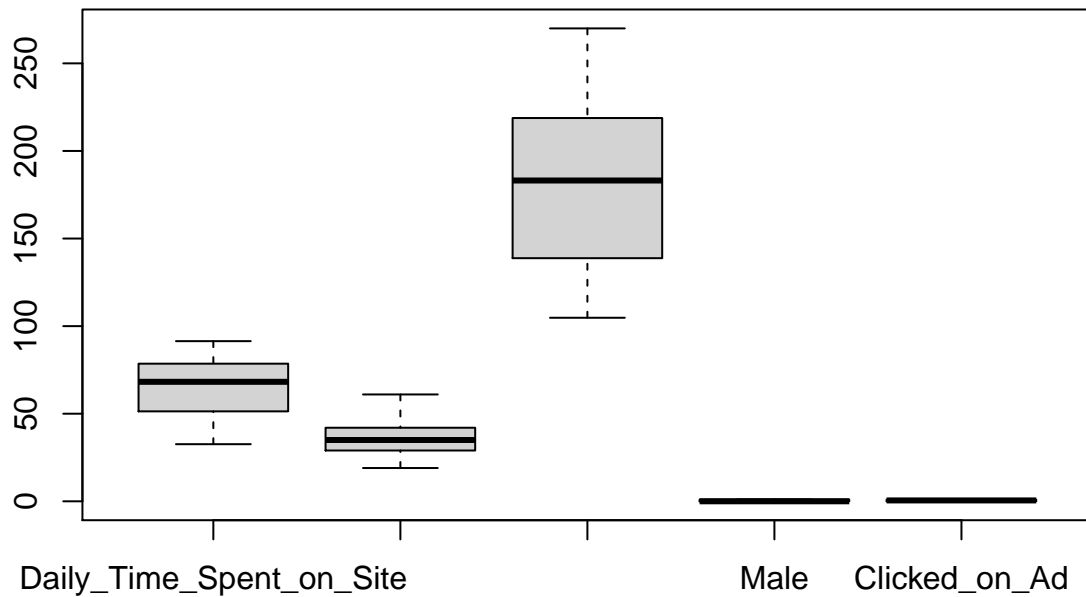
```
## [1] 8.785562
```

```
# Standard deviation code Daily Time Spent on Site
df1.Daily_Time_Spent_on_Site.sd <- sd(df1$Daily_Time_Spent_on_Site)
df1.Daily_Time_Spent_on_Site.sd
```

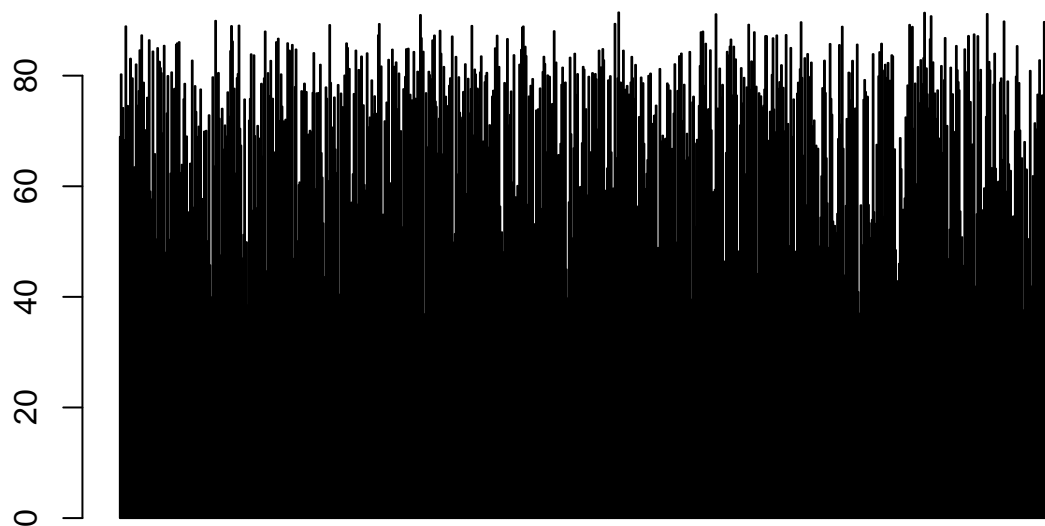
```
## [1] 15.85361
```

Univariate Graphical

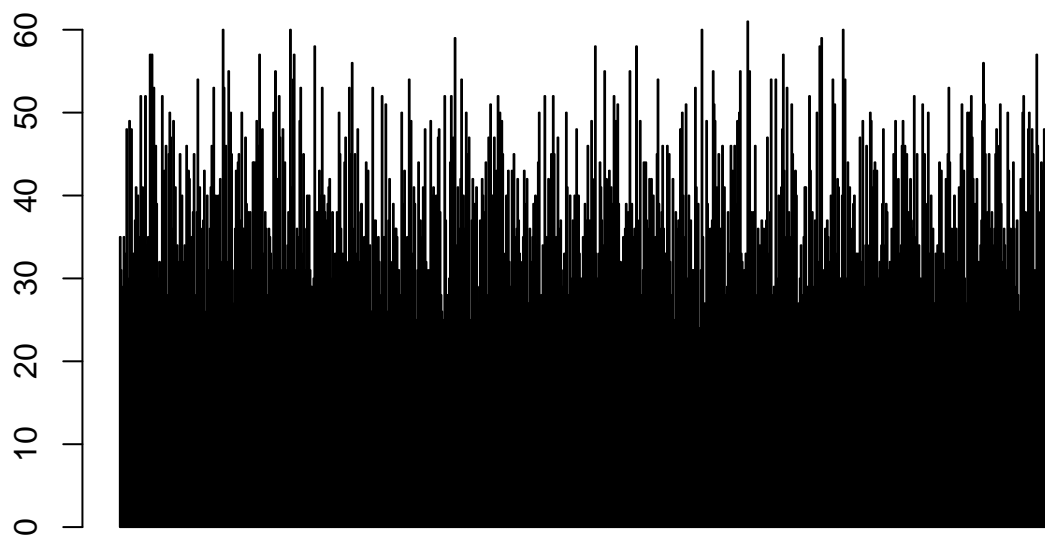
```
# Visualizing a boxplot of numerical values  
boxplot(df1)
```



```
# Assigning the Daily Time Spent on Site column to the variable Daily Time Spent on Site  
Daily_Time_Spent_on_Site <- df1$Daily_Time_Spent_on_Site  
# Frequency Distribution  
Daily_Time_Spent_on_Site_frequency <- table(Daily_Time_Spent_on_Site)  
# Bar plot  
barplot(Daily_Time_Spent_on_Site)
```



```
# Assigning the age column to the variable age  
Age <- df1$Age  
# Frequency Distribution  
Age_frequency <- table(Age)  
# Bar plot  
barplot(Age)
```



```
# Displaying the column names  
colnames(df1)
```

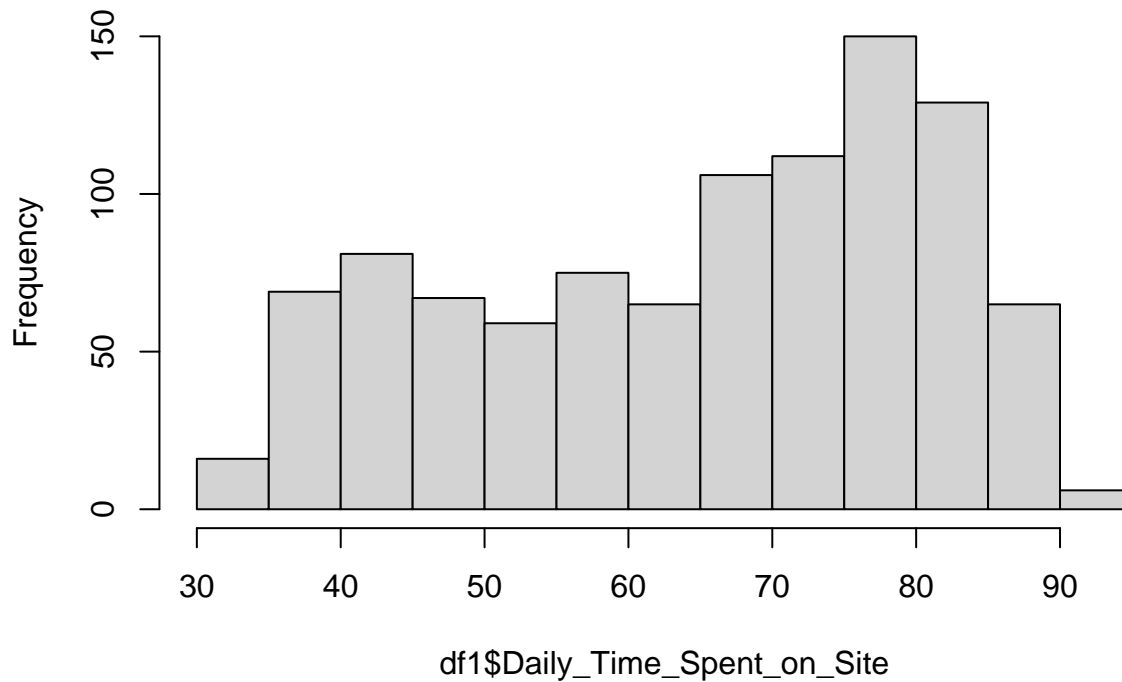
```
## [1] "Daily_Time_Spent_on_Site" "Age"  
## [3] "Daily_Internet_Usage"     "Male"  
## [5] "Clicked_on_Ad"
```

```
# Histogram of age  
hist(df1$Age)
```



```
# Histogram of Daily Time Spent on Site  
hist(df1$Daily_Time_Spent_on_Site)
```

Histogram of df1\$Daily_Time_Spent_on_Site



Bivariate analysis

```
# Assigning the age column to the variable age
Age<- df1$Age
# Covariance
cov(Daily_Time_Spent_on_Site, Age)
```

```
## [1] -46.17415
```

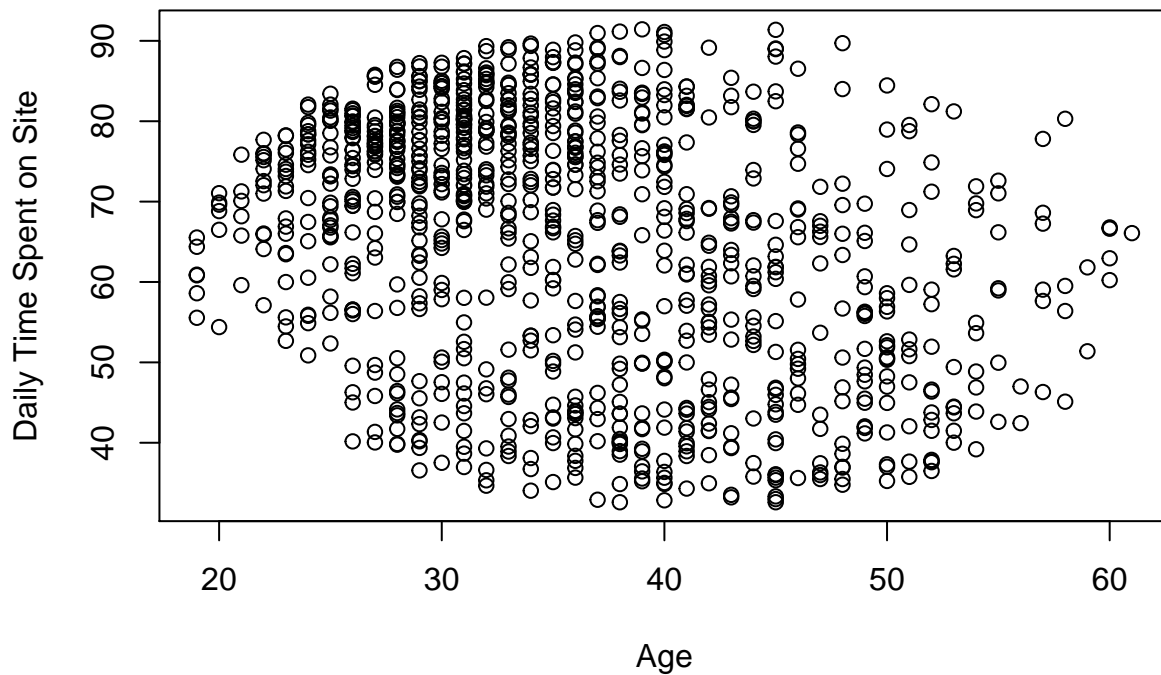
```
# Correlation
cor(Age,Daily_Time_Spent_on_Site)
```

```
## [1] -0.3315133
```

There is a negative correlation.

Graphical Techniques

```
# creating a scatterplot
plot(Age, Daily_Time_Spent_on_Site, xlab="Age", ylab="Daily Time Spent on Site")
```



```
library(corrplot)
```

```
## corrplot 0.92 loaded
```

```
# Rounding the correlation to two decimal places
```

```
res <- cor(df1)
round(res, 2)
```

```
##           Daily_Time_Spent_on_Site   Age Daily_Internet_Usage
## Daily_Time_Spent_on_Site           1.00 -0.33              0.52
## Age                -0.33   1.00              -0.37
## Daily_Internet_Usage           0.52 -0.37              1.00
## Male                -0.02 -0.02              0.03
## Clicked_on_Ad         -0.75  0.49             -0.79
##           Male Clicked_on_Ad
## Daily_Time_Spent_on_Site -0.02   -0.75
## Age                -0.02    0.49
## Daily_Internet_Usage     0.03   -0.79
## Male                   1.00   -0.04
## Clicked_on_Ad          -0.04    1.00
```


4. IMPLEMENTING THE SOLUTION

SUPERVISED LEARNING

```
# Loading Libraries
library(ggplot2)
library(stringr)
```

```
library(rpart)
library(rpart.plot)
```

SVM

```
library(caret)
```

```
## Loading required package: lattice
```

```
str(df1)
```

```
## Classes 'data.table' and 'data.frame':  1000 obs. of  5 variables:
## $ Daily_Time_Spent_on_Site: num  69 80.2 69.5 74.2 68.4 ...
## $ Age : int  35 31 26 29 35 23 33 48 30 20 ...
## $ Daily_Internet_Usage : num  256 194 236 246 226 ...
## $ Male : int  0 1 0 1 0 1 0 1 1 1 ...
## $ Clicked_on_Ad : int  0 0 0 0 0 0 0 1 0 0 ...
## - attr(*, ".internal.selfref")=<externalptr>
```

Summary of df1

```
# Training the dataset
intrain <- createDataPartition(y = df1$Age, p= 0.8, list = FALSE)
training <- df1[intrain,]
head(training)
```

```
##   Daily_Time_Spent_on_Site   Age Daily_Internet_Usage   Male Clicked_on_Ad
##           <num> <int>           <num> <int>           <int>
## 1:           68.95   35           256.09    0             0
## 2:           69.47   26           236.50    0             0
## 3:           74.15   29           245.89    1             0
## 4:           68.37   35           225.58    0             0
## 5:           88.91   33           208.36    0             0
## 6:           66.00   48           131.76    1             1
```

Training the dataset with the age column

```
testing <- df1[-intrain,]
head(testing)
```

```
##   Daily_Time_Spent_on_Site   Age Daily_Internet_Usage   Male Clicked_on_Ad
##                               <num> <int>                <num> <int>         <int>
## 1:                        80.23    31                193.77     1           0
## 2:                        59.99    23                226.74     1           0
## 3:                        42.95    33                143.56     0           1
## 4:                        74.58    40                135.51     1           1
## 5:                        77.22    30                224.44     1           0
## 6:                        84.59    35                226.54     1           0
```

Testing dataset values.

```
# Checking the dimensions of the training dataframe
dim(training);
```

```
## [1] 802    5
```

There are 802 rows and 5 columns

```
# Checking the dimensions of the testing dataframe
dim(testing);
```

```
## [1] 198    5
```

There are 198 rows and 5 columns

```
# Cleaning the data using the anyNA() method which checks for any null values.
anyNA(df1)
```

```
## [1] FALSE
```

There are no null values

```
# Displaying the summary of the data with summary() function
summary(df1)
```

```
##   Daily_Time_Spent_on_Site      Age      Daily_Internet_Usage      Male
##   Min.   :32.60      Min.   :19.00   Min.   :104.8      Min.   :0.000
##   1st Qu.:51.36      1st Qu.:29.00   1st Qu.:138.8      1st Qu.:0.000
##   Median :68.22      Median :35.00   Median :183.1      Median :0.000
##   Mean   :65.00      Mean   :36.01   Mean   :180.0      Mean   :0.481
##   3rd Qu.:78.55      3rd Qu.:42.00   3rd Qu.:218.8      3rd Qu.:1.000
##   Max.   :91.43      Max.   :61.00   Max.   :270.0      Max.   :1.000
##   Clicked_on_Ad
##   Min.   :0.0
##   1st Qu.:0.0
##   Median :0.5
##   Mean   :0.5
##   3rd Qu.:1.0
##   Max.   :1.0
```

Summary of df1 values.

```

# Converting the categorical variables by factorizing them.

training[["Age"]] = factor(training[["Age"]])

# Before training the model you will need to control all the computational overheads.
# Implementing through the trainControl() method.
# We are using setting number =10 and repeats =3

trctrl <- trainControl(method = "repeatedcv", number = 10, repeats = 3)

svm_Linear <- train(Age ~., data = training, method = "svmLinear",
trControl=trctrl,
preProcess = c("center", "scale"),
tuneLength = 10)

# Checking the result of our train() model
svm_Linear

```

```

## Support Vector Machines with Linear Kernel
##
## 802 samples
## 4 predictor
## 43 classes: '19', '20', '21', '22', '23', '24', '25', '26', '27', '28', '29', '30', '31', '32', '33'
##
## Pre-processing: centered (4), scaled (4)
## Resampling: Cross-Validated (10 fold, repeated 3 times)
## Summary of sample sizes: 722, 723, 719, 720, 723, 722, ...
## Resampling results:
##
## Accuracy Kappa
## 0.065696 0.01935521
##
## Tuning parameter 'C' was held constant at a value of 1

```

There are 802 samples.

```

# Using the predict() method for predicting results
# Passing 2 arguments, our trained model and our testing data frame.
test_pred <- predict(svm_Linear, newdata = testing)
test_pred

```

```

## [1] 31 31 45 36 31 31 42 45 42 31 31 31 45 42 45 31 31 45 31 45 42 45 45 31 36
## [26] 36 36 45 50 42 31 31 45 42 31 31 31 31 45 31 31 31 36 31 42 50 36 45 31 45
## [51] 31 45 31 45 31 31 45 50 31 45 31 45 31 31 42 42 36 42 31 45 36 36 31 45 31
## [76] 42 31 31 45 31 31 45 45 42 36 31 31 45 45 31 36 31 31 45 42 31 31 31 42 45
## [101] 36 45 31 31 31 42 50 31 42 42 31 31 31 45 31 31 31 31 45 31 50 31 50 31 31
## [126] 36 31 42 31 31 31 31 45 31 45 42 45 31 45 31 31 31 31 31 36 45 31 31 31 45
## [151] 31 42 42 45 42 31 31 31 31 42 42 42 45 45 31 31 31 31 31 45 45 50 31 45 45
## [176] 31 31 31 45 31 45 42 45 50 31 31 31 31 45 42 42 31 36 45 36 31 36 42
## 43 Levels: 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 ... 61

```

Predicted results of the testing results.

Linear Regression Model

```
# Previewing the dataset
head(df1)
```

```
##      Daily_Time_Spent_on_Site   Age Daily_Internet_Usage   Male Clicked_on_Ad
##                               <num> <int>                <num> <int>        <int>
## 1:                          68.95   35                  256.09    0            0
## 2:                          80.23   31                  193.77    1            0
## 3:                          69.47   26                  236.50    0            0
## 4:                          74.15   29                  245.89    1            0
## 5:                          68.37   35                  225.58    0            0
## 6:                          59.99   23                  226.74    1            0
```

```
# predicting the daily time spent on the site by the users
# Applying the lm() function.
multiple_lm <- lm(Daily_Time_Spent_on_Site ~ ., df1)
```

```
# Generating the anova table
anova(multiple_lm)
```

```
## Analysis of Variance Table
##
## Response: Daily_Time_Spent_on_Site
##              Df Sum Sq Mean Sq F value Pr(>F)
## Age           1  27595   27595 258.4518 < 2e-16 ***
## Daily_Internet_Usage 1  45724   45724 428.2494 < 2e-16 ***
## Male          1    312     312  2.9221 0.08769 .
## Clicked_on_Ad  1  71220   71220 667.0528 < 2e-16 ***
## Residuals    995 106235     107
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
# Performing the prediction
pred2 <- predict(multiple_lm, df1)
```

```
# Printing out the result
pred2
```

```
##           1           2           3           4           5           6           7           8
## 75.13900 77.42549 75.58430 73.73106 77.18840 74.43699 78.15162 53.99553
##           9          10          11          12          13          14          15          16
## 75.46542 77.02978 56.25921 75.51383 55.24760 76.88673 53.26461 50.98072
##          17          18          19          20          21          22          23          24
## 54.60202 80.32467 53.73281 52.96977 75.26861 75.61122 53.67380 76.82366
##          25          26          27          28          29          30          31          32
## 52.44920 76.39506 51.90100 56.06510 54.99764 76.98949 72.89279 76.21437
##          33          34          35          36          37          38          39          40
## 52.34921 47.66112 54.72842 78.31685 54.90304 77.81984 54.63767 52.60100
##          41          42          43          44          45          46          47          48
## 78.15093 74.90589 77.23070 77.69350 77.61286 51.62373 75.79182 79.76648
```

##	49	50	51	52	53	54	55	56
##	53.47152	56.10940	76.45414	76.53075	49.55280	54.09226	56.24845	80.02771
##	57	58	59	60	61	62	63	64
##	53.08125	55.76953	75.69179	53.17868	79.24581	77.77211	80.37497	78.51972
##	65	66	67	68	69	70	71	72
##	54.58259	76.91430	56.51709	49.36195	75.40225	54.15665	53.37678	78.45062
##	73	74	75	76	77	78	79	80
##	51.68679	51.14004	52.45995	76.57908	53.52390	74.58147	54.25273	56.56481
##	81	82	83	84	85	86	87	88
##	75.99875	75.18867	55.60556	55.62178	79.58773	56.61852	78.96643	53.52524
##	89	90	91	92	93	94	95	96
##	55.67808	54.44080	54.58190	51.98360	74.66136	53.61865	50.03164	75.44122
##	97	98	99	100	101	102	103	104
##	51.33950	53.18271	53.53741	76.51466	53.47834	73.86335	79.77788	74.81918
##	105	106	107	108	109	110	111	112
##	77.31056	75.92158	73.67058	53.63612	53.53731	77.90846	53.63153	47.49203
##	113	114	115	116	117	118	119	120
##	76.85179	54.34749	75.67904	74.02864	54.09028	54.72967	50.63563	53.82557
##	121	122	123	124	125	126	127	128
##	76.31783	75.89060	77.27366	48.06826	55.33691	76.44210	55.43633	79.40511
##	129	130	131	132	133	134	135	136
##	76.58583	76.84584	52.77235	54.29437	51.32674	77.28912	55.67548	51.22600
##	137	138	139	140	141	142	143	144
##	53.76304	52.65741	78.05954	81.29863	79.17463	50.15590	54.10432	75.96325
##	145	146	147	148	149	150	151	152
##	78.77502	52.92815	54.28101	54.90364	53.93909	55.13414	78.01992	80.78072
##	153	154	155	156	157	158	159	160
##	52.21817	75.53530	77.72975	76.19898	54.66649	47.33067	76.08540	52.50156
##	161	162	163	164	165	166	167	168
##	77.70898	75.67099	75.09398	80.36554	55.12200	49.72401	55.40214	75.68239
##	169	170	171	172	173	174	175	176
##	55.22741	78.90664	53.92636	78.56884	78.77295	80.51803	55.12736	77.05540
##	177	178	179	180	181	182	183	184
##	56.87245	77.68740	52.25636	78.25709	53.58706	51.01971	54.97093	74.23894
##	185	186	187	188	189	190	191	192
##	77.62363	55.76083	53.37357	74.74670	52.28927	53.40099	54.29433	55.92326
##	193	194	195	196	197	198	199	200
##	55.51562	53.67112	73.56987	54.98352	53.33723	74.84950	76.92238	78.72792
##	201	202	203	204	205	206	207	208
##	78.42170	73.26892	50.47973	78.09183	77.06410	54.39978	77.37845	77.75932
##	209	210	211	212	213	214	215	216
##	55.86631	52.81862	75.51781	52.41694	74.24558	51.89706	77.42756	51.73372
##	217	218	219	220	221	222	223	224
##	54.87013	55.48530	50.13514	51.93935	75.43457	78.09315	54.31319	53.27205
##	225	226	227	228	229	230	231	232
##	81.52704	53.61724	54.47098	52.99663	78.58151	77.67599	75.31767	52.13613
##	233	234	235	236	237	238	239	240
##	53.72275	46.33519	49.65560	53.56158	54.90159	81.24081	54.29838	74.76954
##	241	242	243	244	245	246	247	248
##	53.83967	55.00712	77.58402	78.11936	77.75727	77.58415	52.74008	44.03532
##	249	250	251	252	253	254	255	256
##	53.82966	51.30853	77.50881	54.56042	76.93045	50.04641	56.08791	78.15227
##	257	258	259	260	261	262	263	264
##	76.50592	52.56337	77.53973	54.58317	77.38584	55.72780	51.87418	55.56059

##	265	266	267	268	269	270	271	272
##	76.20436	54.23798	53.03216	76.78537	49.21078	77.81908	52.04490	76.12375
##	273	274	275	276	277	278	279	280
##	75.06309	80.47175	75.38347	52.00580	77.77346	78.59629	79.44664	76.93445
##	281	282	283	284	285	286	287	288
##	55.05484	52.05480	54.15532	76.39511	54.37640	75.27192	53.68237	78.20270
##	289	290	291	292	293	294	295	296
##	54.28298	50.80474	51.18632	77.92928	51.26429	76.78133	79.90951	76.21644
##	297	298	299	300	301	302	303	304
##	76.31517	76.74303	77.52961	74.31139	74.87095	52.69510	53.25390	53.09067
##	305	306	307	308	309	310	311	312
##	52.75417	48.51084	77.73718	76.81963	79.52460	54.11379	75.85101	77.34768
##	313	314	315	316	317	318	319	320
##	50.46484	73.81840	77.30129	56.05360	78.42695	73.99569	79.43526	56.91880
##	321	322	323	324	325	326	327	328
##	53.91551	80.39181	78.91669	75.97730	76.73637	50.03839	48.86028	76.49450
##	329	330	331	332	333	334	335	336
##	76.53409	47.95068	77.99709	77.57394	53.85850	74.22211	78.66553	52.58556
##	337	338	339	340	341	342	343	344
##	79.97937	74.45986	76.53082	76.88404	52.91541	56.61051	77.12058	76.34737
##	345	346	347	348	349	350	351	352
##	51.87278	74.78963	77.98029	50.56713	80.54362	49.13080	75.72134	75.46874
##	353	354	355	356	357	358	359	360
##	76.48777	77.07695	56.21691	76.01425	52.98661	55.99247	54.77210	78.28127
##	361	362	363	364	365	366	367	368
##	51.26559	54.71757	77.13328	52.56873	75.95652	54.01235	77.12522	75.29151
##	369	370	371	372	373	374	375	376
##	77.80840	76.06523	56.91480	52.38138	77.57995	54.54162	76.17539	76.50923
##	377	378	379	380	381	382	383	384
##	76.09218	55.46251	54.28762	74.54389	76.77191	53.89002	80.46565	74.10585
##	385	386	387	388	389	390	391	392
##	52.00377	77.47255	77.82181	53.35467	76.01495	54.19900	75.16386	78.70250
##	393	394	395	396	397	398	399	400
##	76.69801	77.84998	54.39049	75.65949	51.41146	47.66720	77.99045	76.53679
##	401	402	403	404	405	406	407	408
##	51.63109	76.77059	53.87123	78.18384	55.35039	76.25207	53.70133	52.78632
##	409	410	411	412	413	414	415	416
##	47.08361	53.80071	50.78868	77.78552	75.56410	56.32028	77.13334	53.68046
##	417	418	419	420	421	422	423	424
##	54.69334	75.66492	77.34551	78.66487	51.37177	76.98560	53.52390	54.02699
##	425	426	427	428	429	430	431	432
##	52.79779	54.15468	51.38787	76.63758	54.23991	74.86761	76.30642	76.13112
##	433	434	435	436	437	438	439	440
##	46.74167	76.62282	76.35819	54.41330	73.38780	77.16417	54.27959	79.91750
##	441	442	443	444	445	446	447	448
##	52.87501	76.43004	52.05825	54.55702	52.05020	73.31528	47.09157	78.55338
##	449	450	451	452	453	454	455	456
##	56.51513	78.15962	51.62762	52.29192	76.13918	77.54031	49.72144	76.13044
##	457	458	459	460	461	462	463	464
##	54.54558	75.08722	53.54678	74.95960	53.03090	52.14830	78.28794	50.02640
##	465	466	467	468	469	470	471	472
##	78.95439	51.24410	52.38065	50.40574	56.07985	74.75271	51.95210	77.98708
##	473	474	475	476	477	478	479	480
##	77.35425	76.20766	53.22027	75.89131	74.98780	54.12518	52.55466	47.69003

##	481	482	483	484	485	486	487	488
##	75.65480	78.69978	77.04056	52.38062	55.28112	53.56353	80.67059	77.49937
##	489	490	491	492	493	494	495	496
##	53.62676	73.99304	55.70566	54.54221	77.33816	54.34671	53.19540	77.35096
##	497	498	499	500	501	502	503	504
##	75.70589	54.96072	79.91349	54.38573	55.77891	75.52861	76.10823	56.17853
##	505	506	507	508	509	510	511	512
##	56.81135	77.05141	76.05856	55.54916	53.67653	74.86829	51.40932	74.41618
##	513	514	515	516	517	518	519	520
##	76.18082	55.33624	76.12646	53.02282	74.82462	54.13247	55.03336	51.91249
##	521	522	523	524	525	526	527	528
##	54.63093	52.91664	76.90895	55.26638	75.07651	81.81990	54.16004	76.14380
##	529	530	531	532	533	534	535	536
##	54.20042	79.84770	55.00982	54.99970	77.43770	77.11521	77.59140	76.56503
##	537	538	539	540	541	542	543	544
##	79.01471	76.79405	76.46900	75.21826	75.95178	74.56337	79.09135	54.57717
##	545	546	547	548	549	550	551	552
##	77.54437	53.73159	77.43488	76.94191	76.24665	75.37948	77.33081	76.41859
##	553	554	555	556	557	558	559	560
##	54.96621	53.02012	51.62294	77.49270	55.22812	76.42999	78.92472	74.63721
##	561	562	563	564	565	566	567	568
##	56.63533	52.44987	74.83343	81.37117	55.29455	77.50069	56.08519	75.86644
##	569	570	571	572	573	574	575	576
##	78.69509	74.44917	55.14478	74.47131	75.57898	49.79112	56.13156	55.90726
##	577	578	579	580	581	582	583	584
##	52.38269	76.78404	75.67966	77.10042	56.59370	55.12931	51.47379	53.83422
##	585	586	587	588	589	590	591	592
##	49.58828	74.79112	75.69923	55.40275	76.05722	51.59336	52.65398	56.34514
##	593	594	595	596	597	598	599	600
##	75.67967	78.24762	52.79444	51.14134	74.51225	74.46797	78.69707	49.43930
##	601	602	603	604	605	606	607	608
##	53.69661	55.09513	52.97179	76.54685	53.00265	53.84306	77.77811	74.21738
##	609	610	611	612	613	614	615	616
##	54.26215	50.33525	51.13531	51.24744	73.93325	75.06307	75.65889	53.97608
##	617	618	619	620	621	622	623	624
##	54.82438	75.18268	53.33316	76.51595	76.99622	76.31716	53.82767	80.80289
##	625	626	627	628	629	630	631	632
##	77.13931	50.56695	77.17223	52.81130	56.23160	74.09511	74.34500	74.68292
##	633	634	635	636	637	638	639	640
##	76.84177	54.09286	54.70423	51.54581	57.01018	78.00322	53.65896	76.00007
##	641	642	643	644	645	646	647	648
##	51.88902	76.90490	76.61476	78.81454	76.37838	51.64374	55.57871	51.69546
##	649	650	651	652	653	654	655	656
##	78.14890	76.58852	80.86672	75.69920	75.48693	77.72987	74.89780	55.65933
##	657	658	659	660	661	662	663	664
##	78.15967	74.55804	79.20692	74.96766	52.22288	45.35045	54.81436	54.90843
##	665	666	667	668	669	670	671	672
##	77.70292	48.16362	74.77423	76.10356	44.38455	54.41458	77.93260	51.72857
##	673	674	675	676	677	678	679	680
##	78.00586	53.31714	77.53160	75.98941	50.14045	51.81643	81.96629	55.37521
##	681	682	683	684	685	686	687	688
##	76.47566	51.37572	53.27538	78.71247	54.16807	74.82728	74.42700	75.62595
##	689	690	691	692	693	694	695	696
##	76.95331	75.01469	78.90462	75.44129	50.72293	54.64160	77.27705	76.22585

##	697	698	699	700	701	702	703	704
##	53.31512	76.96600	78.90727	76.06862	75.98526	56.05436	47.38374	77.69147
##	705	706	707	708	709	710	711	712
##	78.07711	75.33710	50.56158	77.77619	50.97147	54.95613	55.86353	76.06258
##	713	714	715	716	717	718	719	720
##	80.19566	54.73175	74.96293	55.12394	52.28392	78.83003	79.89214	56.43181
##	721	722	723	724	725	726	727	728
##	75.71600	52.44044	51.38119	78.42774	74.35975	75.79858	74.64724	74.76684
##	729	730	731	732	733	734	735	736
##	79.22973	74.87833	76.50651	74.02194	79.82625	54.02504	53.36533	76.79004
##	737	738	739	740	741	742	743	744
##	74.70974	56.51112	48.19794	77.92322	52.79377	78.83741	77.26427	54.67120
##	745	746	747	748	749	750	751	752
##	54.54823	56.18398	45.82805	54.71752	52.18129	82.00059	53.70474	74.27717
##	753	754	755	756	757	758	759	760
##	78.89317	79.87523	78.15769	75.44798	55.78490	52.96571	49.36261	51.37373
##	761	762	763	764	765	766	767	768
##	77.42958	79.13500	54.94335	50.70601	54.97020	55.12190	48.44101	55.08098
##	769	770	771	772	773	774	775	776
##	50.64632	78.27048	76.75579	77.15807	78.45260	50.68674	51.61026	52.21888
##	777	778	779	780	781	782	783	784
##	52.04541	76.74235	51.12460	77.56926	48.55326	53.37546	76.25871	79.00266
##	785	786	787	788	789	790	791	792
##	51.63104	52.36389	78.45853	49.63065	78.61643	52.24764	49.92551	52.84018
##	793	794	795	796	797	798	799	800
##	76.22309	50.38699	56.40430	76.48978	76.70136	76.38637	78.71527	74.39468
##	801	802	803	804	805	806	807	808
##	52.72990	53.99342	54.99841	55.09714	53.74157	78.00113	55.30063	56.66019
##	809	810	811	812	813	814	815	816
##	51.10238	55.33086	51.39991	74.13073	79.05843	77.96956	76.60803	76.10159
##	817	818	819	820	821	822	823	824
##	53.63213	52.43431	77.94474	79.68115	56.29071	76.35612	44.13736	76.45084
##	825	826	827	828	829	830	831	832
##	76.44007	79.50246	75.48088	55.69427	55.82855	54.05061	51.38044	52.59962
##	833	834	835	836	837	838	839	840
##	53.60186	53.02481	76.38834	73.60682	56.40025	55.06154	53.44198	55.42760
##	841	842	843	844	845	846	847	848
##	55.95890	56.80193	77.79354	75.58302	75.97463	53.30362	53.42988	81.07897
##	849	850	851	852	853	854	855	856
##	72.97137	55.27651	73.83452	50.97341	55.28452	75.60310	45.83749	49.24704
##	857	858	859	860	861	862	863	864
##	75.35190	76.04307	52.30216	75.18540	50.28698	76.97745	80.24665	74.99327
##	865	866	867	868	869	870	871	872
##	75.73558	51.84259	77.37713	78.28198	74.69628	77.52023	49.31221	75.71864
##	873	874	875	876	877	878	879	880
##	78.48211	73.72369	81.00910	50.84978	56.77169	78.16842	76.35888	76.48235
##	881	882	883	884	885	886	887	888
##	50.98817	76.03706	76.83907	53.11084	74.43507	55.79167	53.86189	54.70953
##	889	890	891	892	893	894	895	896
##	76.61473	54.22786	75.93832	48.59075	50.75849	75.16111	77.85609	76.01755
##	897	898	899	900	901	902	903	904
##	74.73327	55.90382	56.03344	53.87729	56.45133	49.69516	55.72852	79.85440
##	905	906	907	908	909	910	911	912
##	76.58582	74.13743	50.04251	78.94961	54.71562	76.06321	50.99434	56.87448


```
##      913      914      915      916      917      918      919      920
## 51.48530 76.81091 53.33585 49.74484 54.62488 77.04731 76.07670 77.68608
##      921      922      923      924      925      926      927      928
## 75.75362 54.22384 54.26690 55.08845 52.85094 53.93976 77.49194 80.20575
##      929      930      931      932      933      934      935      936
## 79.15648 55.27982 76.32055 55.30058 55.58675 51.43017 73.96081 79.36476
##      937      938      939      940      941      942      943      944
## 56.81939 55.24559 50.62548 76.82289 54.33271 53.76380 53.57162 53.02212
##      945      946      947      948      949      950      951      952
## 54.82095 79.12160 77.84667 50.80546 49.09796 51.57805 54.57451 49.90198
##      953      954      955      956      957      958      959      960
## 46.11021 53.01068 75.16182 56.64743 51.60617 75.07179 74.51427 80.17888
##      961      962      963      964      965      966      967      968
## 54.07332 78.33429 76.80208 74.91066 75.56883 51.49139 51.36378 77.21118
##      969      970      971      972      973      974      975      976
## 54.97684 51.64373 52.31344 53.97470 53.64155 78.00785 53.60118 53.48637
##      977      978      979      980      981      982      983      984
## 55.46051 53.38416 80.68598 73.33610 52.40429 76.05730 53.93568 74.76551
##      985      986      987      988      989      990      991      992
## 73.21922 55.56731 74.52166 55.30668 78.35592 74.73125 51.33417 52.42970
##      993      994      995      996      997      998      999     1000
## 55.28114 74.76414 50.80275 47.09424 53.52665 55.05080 78.16833 50.25059
```

```
# Predicting the age of the individuals who click on the site
```

```
# Training the model using the training sets
```

```
linear <- lm(Age ~ ., data=df1)
```

```
summary(linear)
```

```
##
```

```
## Call:
```

```
## lm(formula = Age ~ ., data = df1)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
```

```
## -22.2809  -5.0018  -0.4324   5.1276  20.9440
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept)      24.581484    2.904269   8.464  <2e-16 ***
```

```
## Daily_Time_Spent_on_Site  0.052833    0.023369   2.261   0.024 *
```

```
## Daily_Internet_Usage     0.014085    0.009045   1.557   0.120
```

```
## Male                 0.041153    0.484964   0.085   0.932
```

```
## Clicked_on_Ad         10.876440    1.024908  10.612  <2e-16 ***
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Residual standard error: 7.636 on 995 degrees of freedom
```

```
## Multiple R-squared:  0.2475, Adjusted R-squared:  0.2445
```

```
## F-statistic: 81.83 on 4 and 995 DF, p-value: < 2.2e-16
```

```
# Predicting the Output
```

```
predicted = predict(linear, df1)
```

```
predicted
```

##	1	2	3	4	5	6	7	8
##	31.83142	31.59072	31.58296	32.00363	31.37103	30.98579	32.21366	40.84192
##	9	10	11	12	13	14	15	16
##	31.68031	30.90376	39.69357	32.26334	40.76798	31.80025	39.74918	40.83228
##	17	18	19	20	21	22	23	24
##	40.20711	31.55678	40.05659	41.34804	31.86370	32.28265	39.97164	32.19134
##	25	26	27	28	29	30	31	32
##	40.00002	31.66722	39.95246	40.02283	40.84575	31.55322	31.95717	32.11034
##	33	34	35	36	37	38	39	40
##	41.00251	41.38969	40.42911	31.87830	40.55545	31.18504	39.84409	39.79766
##	41	42	43	44	45	46	47	48
##	31.92504	31.53004	31.77284	31.87133	31.71846	39.92419	31.14961	31.89244
##	49	50	51	52	53	54	55	56
##	39.58379	39.68368	31.25966	31.97822	39.98664	40.03432	40.43031	31.34065
##	57	58	59	60	61	62	63	64
##	41.02348	39.65572	32.05239	39.86977	31.68817	31.87528	30.77713	31.68957
##	65	66	67	68	69	70	71	72
##	39.87525	30.71640	40.31546	40.14468	31.65606	41.46220	39.85310	31.36329
##	73	74	75	76	77	78	79	80
##	40.58082	39.60689	39.58720	31.00951	39.82176	32.35975	40.06421	39.94406
##	81	82	83	84	85	86	87	88
##	31.69676	31.64124	40.08300	41.05399	30.91462	39.99061	31.50136	39.95388
##	89	90	91	92	93	94	95	96
##	40.04266	40.04412	41.09964	40.51362	31.53003	39.91758	40.14854	31.79101
##	97	98	99	100	101	102	103	104
##	40.08401	39.76960	39.56788	31.81510	40.03291	31.29680	32.06747	31.29777
##	105	106	107	108	109	110	111	112
##	30.94183	32.10238	31.83374	39.82425	39.87312	31.45769	41.47099	42.16126
##	113	114	115	116	117	118	119	120
##	31.31687	39.67776	31.38392	32.13184	41.58879	40.34860	42.83785	42.30632
##	121	122	123	124	125	126	127	128
##	32.13787	30.46789	31.68117	40.62495	40.54137	32.03811	41.49074	32.20162
##	129	130	131	132	133	134	135	136
##	31.44289	31.43007	40.41015	39.52648	40.65438	31.61547	39.55615	40.38898
##	137	138	139	140	141	142	143	144
##	39.46354	39.46783	30.85487	30.68649	31.69177	39.77675	40.18851	32.34278
##	145	146	147	148	149	150	151	152
##	31.19754	39.47729	40.27613	40.76944	40.23229	41.19966	30.17941	31.15505
##	153	154	155	156	157	158	159	160
##	39.68247	31.52019	31.40827	32.39936	39.60536	40.55156	31.80275	41.69239
##	161	162	163	164	165	166	167	168
##	31.40584	32.12864	31.92139	30.87581	39.61630	40.03336	40.97451	32.26425
##	169	170	171	172	173	174	175	176
##	40.12324	31.95834	39.76309	31.23225	31.58422	30.80938	40.74115	31.57697
##	177	178	179	180	181	182	183	184
##	40.15461	31.19659	39.72958	32.05069	39.61921	42.29500	40.99571	31.79343
##	185	186	187	188	189	190	191	192
##	32.06625	39.85934	40.11462	32.03199	42.10020	39.61819	39.84910	40.14520
##	193	194	195	196	197	198	199	200
##	40.26344	39.83161	32.27971	40.89585	39.79800	32.14461	31.33590	31.34452
##	201	202	203	204	205	206	207	208
##	31.10976	31.84694	39.81341	30.91263	31.34989	40.23676	31.63985	31.70384
##	209	210	211	212	213	214	215	216
##	40.58564	39.52014	31.74921	39.97235	31.99888	41.24816	32.14135	41.08568

##	217	218	219	220	221	222	223	224
##	40.52589	39.85062	40.12051	40.16284	32.28180	31.41317	40.02084	40.86036
##	225	226	227	228	229	230	231	232
##	31.70110	41.50315	40.17587	39.47938	31.11981	31.43166	31.35688	39.89130
##	233	234	235	236	237	238	239	240
##	40.78319	42.96951	40.27730	39.52060	41.25357	30.27614	41.18977	31.80023
##	241	242	243	244	245	246	247	248
##	41.81076	39.95861	31.87416	31.95458	30.86193	32.05601	39.66857	42.31567
##	249	250	251	252	253	254	255	256
##	40.07029	41.01411	31.76422	40.41007	31.96281	40.25248	40.18787	31.69579
##	257	258	259	260	261	262	263	264
##	32.11581	39.56891	32.17627	41.06412	31.51943	40.18724	40.66432	39.49455
##	265	266	267	268	269	270	271	272
##	32.46612	40.93586	41.11570	31.65029	40.03433	31.62576	40.21594	31.60553
##	273	274	275	276	277	278	279	280
##	32.19403	30.75005	31.85742	40.18085	32.24847	31.84680	31.12571	31.70892
##	281	282	283	284	285	286	287	288
##	39.93987	40.45061	39.48252	31.71648	39.84603	31.59136	40.20958	31.96498
##	289	290	291	292	293	294	295	296
##	39.82914	41.01124	40.54436	32.01811	40.66149	32.11533	30.71267	32.06032
##	297	298	299	300	301	302	303	304
##	31.84895	31.57877	31.85339	31.35965	31.79953	40.72433	39.67765	41.43957
##	305	306	307	308	309	310	311	312
##	39.56085	42.23482	32.08290	32.20545	31.69256	40.62949	31.78784	31.69399
##	313	314	315	316	317	318	319	320
##	41.83549	31.96290	32.31016	40.92070	31.22200	31.82632	31.47123	40.21568
##	321	322	323	324	325	326	327	328
##	40.08470	31.93388	31.60380	31.50929	32.43073	40.16533	40.26931	31.90318
##	329	330	331	332	333	334	335	336
##	31.23565	40.74472	31.67085	31.82825	40.37846	32.27734	31.83817	39.56594
##	337	338	339	340	341	342	343	344
##	31.78379	31.89234	31.65182	31.04066	39.75558	40.89877	32.48955	31.95354
##	345	346	347	348	349	350	351	352
##	40.58139	31.17664	31.70005	40.15579	31.12171	41.28091	31.86001	31.49430
##	353	354	355	356	357	358	359	360
##	31.93534	31.89105	41.20192	32.35050	39.80088	39.66289	40.40584	31.91565
##	361	362	363	364	365	366	367	368
##	39.67887	40.34459	31.56420	41.28338	31.53438	41.27181	31.20208	32.32986
##	369	370	371	372	373	374	375	376
##	31.59151	31.85147	40.15542	39.53879	31.69122	39.70092	31.35921	31.50192
##	377	378	379	380	381	382	383	384
##	32.45536	40.85771	40.79431	32.40374	31.98301	39.54019	31.05292	31.88955
##	385	386	387	388	389	390	391	392
##	40.24690	31.99763	31.72042	39.71658	31.42820	40.76046	32.09243	31.84895
##	393	394	395	396	397	398	399	400
##	31.69721	30.72694	40.79112	31.25291	40.17929	41.22145	31.44401	31.54061
##	401	402	403	404	405	406	407	408
##	39.73141	32.31432	39.63410	32.03701	40.17614	32.38700	40.43382	40.03315
##	409	410	411	412	413	414	415	416
##	41.83107	39.47812	40.53993	31.66882	31.15510	39.99219	32.15239	39.49526
##	417	418	419	420	421	422	423	424
##	40.91284	31.53313	31.30225	30.54242	39.57532	32.15080	40.01037	40.34546
##	425	426	427	428	429	430	431	432
##	39.66569	40.37354	40.55433	32.16555	40.39233	32.36273	32.29422	32.40081

##	433	434	435	436	437	438	439	440
##	42.72678	32.12462	32.15221	40.41520	31.09673	31.62963	40.09726	31.07364
##	441	442	443	444	445	446	447	448
##	39.97385	32.11280	39.50386	39.84191	39.61169	32.12895	40.53162	31.56111
##	449	450	451	452	453	454	455	456
##	39.88367	31.12003	40.01533	40.45583	32.22060	31.61348	42.67097	31.43608
##	457	458	459	460	461	462	463	464
##	39.52431	31.94983	40.75514	31.98472	39.93252	39.70046	31.04608	40.19627
##	465	466	467	468	469	470	471	472
##	31.96672	40.04211	42.05830	40.38384	40.63896	31.25918	40.13228	31.20820
##	473	474	475	476	477	478	479	480
##	31.64145	32.14656	39.95724	31.68461	31.88224	39.74370	39.60518	40.37213
##	481	482	483	484	485	486	487	488
##	30.55798	31.66615	31.25466	40.00182	39.57918	40.71530	31.29040	31.63710
##	489	490	491	492	493	494	495	496
##	39.58337	32.31366	39.76181	40.33558	30.51361	40.30580	40.73968	32.03539
##	497	498	499	500	501	502	503	504
##	31.95671	39.44918	30.93160	40.40885	39.94053	32.36405	31.29875	40.02552
##	505	506	507	508	509	510	511	512
##	40.55413	31.90497	31.98540	39.80434	42.14321	32.12156	40.86511	31.59518
##	513	514	515	516	517	518	519	520
##	32.11947	40.11668	32.40964	39.60444	32.39837	41.69233	40.40808	39.47863
##	521	522	523	524	525	526	527	528
##	39.83860	40.53799	31.66279	40.85256	32.08667	30.40373	39.72010	30.56723
##	529	530	531	532	533	534	535	536
##	40.13172	31.62654	40.60196	39.84607	32.23291	32.40979	31.45983	31.71378
##	537	538	539	540	541	542	543	544
##	31.17366	30.91983	32.37566	31.68119	31.85535	32.11119	31.44813	40.40088
##	545	546	547	548	549	550	551	552
##	31.60897	40.07902	31.72682	32.29793	32.11755	31.76594	31.75400	31.89274
##	553	554	555	556	557	558	559	560
##	40.60954	39.65901	39.57435	31.56757	40.08598	31.98508	30.92080	32.09140
##	561	562	563	564	565	566	567	568
##	40.57257	39.56322	31.30379	30.21169	40.53612	31.45980	39.60181	32.18134
##	569	570	571	572	573	574	575	576
##	30.71709	31.86369	39.88629	31.85610	31.94257	41.67028	39.58742	39.93325
##	577	578	579	580	581	582	583	584
##	39.46806	32.15448	31.08975	31.43660	40.69900	40.62202	39.84926	40.98490
##	585	586	587	588	589	590	591	592
##	40.77968	32.36128	32.26825	39.59073	32.01970	39.83496	40.97343	40.03864
##	593	594	595	596	597	598	599	600
##	31.46105	31.64960	40.03716	40.03311	31.96303	32.04783	31.87590	40.62735
##	601	602	603	604	605	606	607	608
##	41.85559	39.91912	41.20233	31.53418	39.95061	40.86448	31.12452	31.45014
##	609	610	611	612	613	614	615	616
##	40.41558	42.21065	39.74638	39.67328	32.05675	31.87372	31.57840	39.85035
##	617	618	619	620	621	622	623	624
##	39.82345	31.85458	39.46555	31.61489	31.61905	32.22572	40.81568	31.37257
##	625	626	627	628	629	630	631	632
##	31.46076	41.15467	31.85539	39.79817	39.80823	31.85534	31.30852	32.49759
##	633	634	635	636	637	638	639	640
##	31.45557	39.51082	40.46283	40.09485	40.15529	32.22643	40.78451	31.67742
##	641	642	643	644	645	646	647	648
##	41.15041	31.97109	31.91841	31.16523	32.29128	39.88013	39.49911	39.88461

##	649	650	651	652	653	654	655	656
##	30.88425	32.21727	31.46838	32.04717	32.28326	32.44006	31.81911	39.97238
##	657	658	659	660	661	662	663	664
##	31.70146	32.20464	31.79920	31.80340	40.24410	41.16662	39.77353	41.33428
##	665	666	667	668	669	670	671	672
##	31.90940	41.32027	31.82979	32.11122	41.80238	40.52401	31.53539	41.58521
##	673	674	675	676	677	678	679	680
##	32.26008	40.27873	31.63627	32.15801	39.85744	40.10435	31.37803	40.76537
##	681	682	683	684	685	686	687	688
##	31.57119	39.83120	39.56448	31.35642	39.53300	31.84520	32.02631	31.53951
##	689	690	691	692	693	694	695	696
##	31.76870	31.89117	31.96331	32.49463	41.52161	40.55218	31.02948	31.90423
##	697	698	699	700	701	702	703	704
##	39.96591	31.83719	31.84020	31.77524	30.50655	41.00713	42.98700	31.61384
##	705	706	707	708	709	710	711	712
##	31.69027	31.85379	41.61254	32.08431	40.46641	41.72252	39.53860	31.33392
##	713	714	715	716	717	718	719	720
##	31.46714	39.91299	31.43088	39.79039	39.56138	31.78422	30.88322	40.63025
##	721	722	723	724	725	726	727	728
##	32.04729	39.71315	39.88706	30.14799	31.92855	32.04807	31.35717	32.17760
##	729	730	731	732	733	734	735	736
##	31.77145	32.03071	30.76338	31.41555	31.72099	39.53724	40.30565	32.15388
##	737	738	739	740	741	742	743	744
##	32.07663	40.94792	40.96455	31.85518	39.70921	31.01692	31.14387	40.27212
##	745	746	747	748	749	750	751	752
##	40.68800	40.27644	42.48750	39.75074	40.59757	29.76322	40.90518	32.11988
##	753	754	755	756	757	758	759	760
##	30.98991	31.48463	31.63714	31.26530	40.04043	39.88862	40.14867	42.10953
##	761	762	763	764	765	766	767	768
##	31.81495	31.12315	40.11133	40.34401	40.00311	39.81567	40.37019	39.91356
##	769	770	771	772	773	774	775	776
##	41.53854	31.61294	31.50667	30.56599	32.28937	41.82090	39.79268	40.16752
##	777	778	779	780	781	782	783	784
##	40.57662	31.56444	40.93459	31.93467	41.93888	41.23308	32.08302	31.34940
##	785	786	787	788	789	790	791	792
##	39.52846	39.45636	30.86609	41.40988	31.86124	39.57702	39.81734	39.72803
##	793	794	795	796	797	798	799	800
##	30.49089	39.72184	39.75349	31.50386	31.60807	31.97751	30.91657	31.87136
##	801	802	803	804	805	806	807	808
##	40.22740	39.88899	40.13620	39.63082	40.11496	31.60253	40.17862	39.80287
##	809	810	811	812	813	814	815	816
##	39.97820	40.81632	39.79411	32.24337	31.48098	31.37220	32.39226	32.06918
##	817	818	819	820	821	822	823	824
##	39.65727	40.32908	31.91120	31.52276	39.98529	32.20038	42.92492	31.75516
##	825	826	827	828	829	830	831	832
##	31.39124	31.71547	32.17020	39.51164	40.50636	39.92397	39.78413	39.62652
##	833	834	835	836	837	838	839	840
##	39.50625	39.68403	31.39502	31.54783	39.68609	40.09819	39.64027	40.06888
##	841	842	843	844	845	846	847	848
##	40.88529	40.25096	31.52218	31.82215	32.48952	39.71144	40.19534	31.65592
##	849	850	851	852	853	854	855	856
##	32.02513	40.28726	32.29712	39.57994	40.21438	31.94171	43.02176	40.82752
##	857	858	859	860	861	862	863	864
##	32.14104	31.66803	39.76006	32.45919	42.89909	31.28722	31.21108	32.07335

```
##      865      866      867      868      869      870      871      872
## 32.00451 39.59063 31.95664 32.42553 31.37985 32.06350 40.33185 31.96181
##      873      874      875      876      877      878      879      880
## 31.14794 31.96248 31.02401 40.03653 40.52813 30.88557 32.41415 31.13176
##      881      882      883      884      885      886      887      888
## 40.20355 32.10244 32.02076 39.88896 31.88820 39.66820 39.83018 39.88039
##      889      890      891      892      893      894      895      896
## 31.82519 39.92002 31.98948 40.79392 40.14143 31.67202 31.87341 31.82469
##      897      898      899      900      901      902      903      904
## 32.10211 41.17600 39.91065 39.47897 39.85974 40.46807 39.57642 31.57649
##      905      906      907      908      909      910      911      912
## 31.96309 31.97242 40.45584 31.91436 40.52729 31.74681 40.32878 40.57641
##      913      914      915      916      917      918      919      920
## 39.60667 32.22328 39.52231 39.76075 40.00497 32.03510 31.81776 30.63932
##      921      922      923      924      925      926      927      928
## 32.32153 40.02727 39.72117 40.04724 40.80677 39.48278 31.26453 32.00269
##      929      930      931      932      933      934      935      936
## 31.47908 41.65902 32.23740 40.33632 40.43881 39.54947 32.26166 31.75967
##      937      938      939      940      941      942      943      944
## 40.31598 40.04222 39.85719 30.44872 39.60637 41.90156 41.46202 39.58691
##      945      946      947      948      949      950      951      952
## 40.51584 32.11814 31.17177 39.72105 41.13449 42.35018 40.78895 39.97202
##      953      954      955      956      957      958      959      960
## 42.08241 39.85760 30.83149 39.87581 39.77265 31.75636 32.13948 31.40556
##      961      962      963      964      965      966      967      968
## 40.20886 31.49236 31.25893 31.31201 31.31317 40.05904 39.97341 30.86834
##      969      970      971      972      973      974      975      976
## 39.60409 41.02877 39.72115 39.74578 40.41280 31.71762 39.48801 39.56817
##      977      978      979      980      981      982      983      984
## 40.25409 40.04254 30.69627 31.92308 41.42472 32.26510 40.31323 32.30404
##      985      986      987      988      989      990      991      992
## 31.74577 41.07073 31.93855 40.32661 32.24130 31.94198 39.72278 39.53883
##      993      994      995      996      997      998      999      1000
## 40.85359 31.22075 40.20362 42.29220 40.10275 39.92228 30.16368 40.34805
```

Prediction results.

Decision Tree

```
# Loading the party package
library(party)
```

```
## Loading required package: grid
```

```
## Loading required package: mvtnorm
```

```
## Loading required package: modeltools
```

```
## Loading required package: stats4
```

```
## Loading required package: strucchange

## Loading required package: zoo

##
## Attaching package: 'zoo'

## The following objects are masked from 'package:data.table':
##
##   yearmon, yearqtr

## The following objects are masked from 'package:base':
##
##   as.Date, as.Date.numeric

## Loading required package: sandwich

##
## Attaching package: 'strucchange'

## The following object is masked from 'package:stringr':
##
##   boundary
```

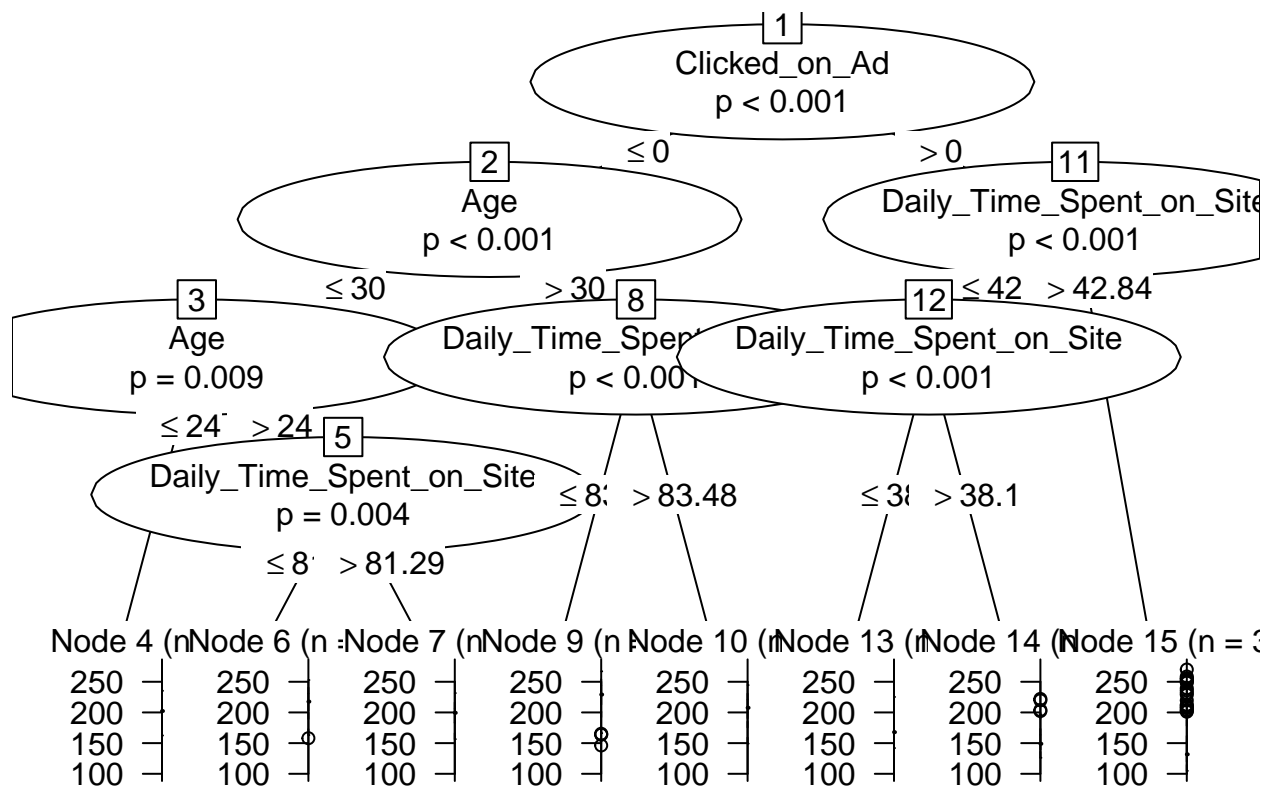
```
# Displaying the top of dataset df1.
head(df1)
```

```
##   Daily_Time_Spent_on_Site  Age Daily_Internet_Usage  Male Clicked_on_Ad
##           <num> <int>           <num> <int>           <int>
## 1:           68.95   35           256.09    0             0
## 2:           80.23   31           193.77    1             0
## 3:           69.47   26           236.50    0             0
## 4:           74.15   29           245.89    1             0
## 5:           68.37   35           225.58    0             0
## 6:           59.99   23           226.74    1             0
```

```
# Creating the input data frame.
input.dat <- df1

# Creating the tree.
output.tree <- ctree( Daily_Internet_Usage ~ Age + Clicked_on_Ad +
  Daily_Time_Spent_on_Site,
  data = input.dat)

# Plotting the tree.
plot(output.tree)
```



Naives Bayes

```
library(ggplot2)
library(caret) #confusionMatrix
library(caretEnsemble)
```

```
##
## Attaching package: 'caretEnsemble'

## The following object is masked from 'package:ggplot2':
##
## autoplot
```

```
library(psych)
```

```
##
## Attaching package: 'psych'

## The following objects are masked from 'package:ggplot2':
##
## %+%, alpha
```



```
library(Amelia) #missmap
```

```
## Loading required package: Rcpp
```

```
## ##  
## ## Amelia II: Multiple Imputation  
## ## (Version 1.8.0, built: 2021-05-26)  
## ## Copyright (C) 2005-2022 James Honaker, Gary King and Matthew Blackwell  
## ## Refer to http://gking.harvard.edu/amelia/ for more information  
## ##
```

```
library(mice) #mice
```

```
##  
## Attaching package: 'mice'
```

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

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

```
library(GGally) #ggpairs
```

```
## Registered S3 method overwritten by 'GGally':  
## method from  
## +.gg ggplot2
```

```
library(rpart)
```

```
# Describing the data  
describe(df1)
```

```
##               vars      n  mean    sd median trimmed  mad   min  
## Daily_Time_Spent_on_Site  1 1000  65.00 15.85  68.22   65.74 17.92  32.60  
## Age                      2 1000  36.01  8.79  35.00   35.51  8.90  19.00  
## Daily_Internet_Usage     3 1000 180.00 43.90 183.13  179.99 58.61 104.78  
## Male                     4 1000   0.48  0.50   0.00    0.48  0.00   0.00  
## Clicked_on_Ad            5 1000   0.50  0.50   0.50    0.50  0.74   0.00  
##               max range skew kurtosis  se  
## Daily_Time_Spent_on_Site 91.43 58.83 -0.37  -1.10 0.50  
## Age                     61.00 42.00  0.48  -0.41 0.28  
## Daily_Internet_Usage    269.96 165.18 -0.03  -1.28 1.39  
## Male                     1.00  1.00  0.08  -2.00 0.02  
## Clicked_on_Ad            1.00  1.00  0.00  -2.00 0.02
```

```
# Displaying the top of the dataset
head(df1)
```

```
##      Daily_Time_Spent_on_Site   Age Daily_Internet_Usage  Male Clicked_on_Ad
##                               <num> <int>                <num> <int>        <int>
## 1:                          68.95   35                  256.09    0            0
## 2:                          80.23   31                  193.77    1            0
## 3:                          69.47   26                  236.50    0            0
## 4:                          74.15   29                  245.89    1            0
## 5:                          68.37   35                  225.58    0            0
## 6:                          59.99   23                  226.74    1            0
```

```
# Converting the output variable into a categorical variable
df1$Clicked_on_Ad <- factor(df1$Clicked_on_Ad)
df1$Clicked_on_Ad
```

```
##      [1] 0 0 0 0 0 0 0 0 1 0 0 1 0 1 0 1 1 1 0 1 1 0 0 1 0 1 0 1 1 0 0 0 1 1 1 0 1
##      [38] 0 1 1 0 0 0 0 0 1 0 0 1 1 0 0 1 1 1 0 1 1 0 1 0 0 0 0 1 0 1 1 0 1 1 0 1 1
##      [75] 1 0 1 0 1 1 0 0 1 1 0 1 0 1 1 1 1 1 0 1 1 0 1 1 1 0 1 0 0 0 0 0 0 1 1 0 1
##     [112] 1 0 1 0 0 1 1 1 1 0 0 0 1 1 0 1 0 0 0 1 1 1 0 1 1 1 1 0 0 0 1 1 0 0 1 1 1
##     [149] 1 1 0 0 1 0 0 0 1 1 0 1 0 0 0 0 1 1 1 0 1 0 1 0 0 0 1 0 1 0 1 0 1 1 1 0 0
##     [186] 1 1 0 1 1 1 1 1 1 0 1 1 0 0 0 0 0 1 0 0 1 0 0 1 1 0 1 0 1 0 1 1 1 1 1 0 0
##     [223] 1 1 0 1 1 1 0 0 0 1 1 1 1 1 1 0 1 0 1 1 0 0 0 0 1 1 1 1 0 1 0 1 1 0 0 1 0
##     [260] 1 0 1 1 1 0 1 1 0 1 0 1 0 0 0 0 1 0 0 0 0 1 1 1 0 1 0 1 0 1 1 1 0 1 0 0 0
##     [297] 0 0 0 0 0 1 1 1 1 1 0 0 0 1 0 0 1 0 0 1 0 0 1 0 0 0 1 1 0 0 0 0 1 1 0 0 1
##     [334] 0 0 1 0 0 0 0 1 1 0 0 1 0 0 1 0 1 0 0 0 0 1 0 1 1 1 0 1 1 0 1 0 1 0 0 0 0
##     [371] 1 1 0 1 0 0 0 1 1 0 0 1 0 0 1 0 0 1 0 1 0 0 0 0 1 0 1 1 0 0 1 0 1 0 1 0 1
##     [408] 1 1 1 1 0 0 1 0 1 1 0 0 0 1 0 1 1 1 1 1 0 1 0 0 0 1 0 0 1 0 0 1 0 1 0 1 1
##     [445] 1 0 1 0 1 0 1 1 0 0 1 0 1 0 1 0 1 1 0 1 0 1 1 1 1 0 1 0 0 0 1 0 0 1 1 1 0
##     [482] 0 0 1 1 1 0 0 1 0 1 1 0 1 1 0 0 1 0 1 1 0 0 1 1 0 0 1 1 0 0 1 0 1 0 1 0 1
##     [519] 1 1 1 1 0 1 0 0 1 0 1 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 1 0 1 0 0 0 0 0 0 1 1 1
##     [556] 0 1 0 0 0 1 1 0 0 1 0 1 0 0 0 1 0 0 1 1 1 1 0 0 0 1 1 1 1 1 0 0 1 0 1 1 1
##     [593] 0 0 1 1 0 0 0 1 1 1 1 0 1 1 0 0 1 1 1 1 0 0 0 1 1 0 1 0 0 0 1 0 0 1 0 1 1
##     [630] 0 0 0 0 1 1 1 1 0 1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 0 0 1 0 0 0 0 1 1 1 1 0 1
##     [667] 0 0 1 1 0 1 0 1 0 0 1 1 0 1 0 1 1 0 1 0 0 0 0 0 0 0 1 1 0 0 1 0 0 0 0 1 1
##     [704] 0 0 0 1 0 1 1 1 0 0 1 0 1 1 0 0 1 0 1 1 0 0 0 0 0 0 0 0 0 0 0 1 1 0 0 1 1 0
##     [741] 1 0 0 1 1 1 1 1 1 0 1 0 0 0 0 0 1 1 1 1 0 0 1 1 1 1 1 1 1 1 0 0 0 0 1 1 1 1
##     [778] 0 1 0 1 1 0 0 1 1 0 1 0 1 1 1 0 1 1 0 0 0 0 0 1 1 1 1 1 0 1 1 1 1 1 0 0 0
##     [815] 0 0 1 1 0 0 1 0 1 0 0 0 0 1 1 1 1 1 1 1 0 0 1 1 1 1 1 1 0 0 0 1 1 0 0 1 0
##     [852] 1 1 0 1 1 0 0 1 0 1 0 0 0 0 1 0 0 0 0 1 0 0 0 0 1 1 0 0 0 1 0 0 1 0 1 1 1
##     [889] 0 1 0 1 1 0 0 0 0 1 1 1 1 1 1 0 0 0 1 0 1 0 1 1 1 0 1 1 1 0 0 0 0 1 1 1 1
##     [926] 1 0 0 0 1 0 1 1 1 0 0 1 1 1 0 1 1 1 1 1 0 0 1 1 1 1 1 1 1 0 1 1 0 0 0 1 0
##     [963] 0 0 0 1 1 0 1 1 1 1 1 0 1 1 1 1 0 0 1 0 1 0 0 1 0 1 0 0 1 1 1 0 1 1 1 1 0
##    [1000] 1
## Levels: 0 1
```

```
# Displaying the top of the dataset
head(df1)
```

```
##      Daily_Time_Spent_on_Site   Age Daily_Internet_Usage  Male Clicked_on_Ad
##                               <num> <int>                <num> <int>        <fctr>
## 1:                          68.95   35                  256.09    0            0
```

```
## 2:      80.23    31      193.77    1      0
## 3:      69.47    26      236.50    0      0
## 4:      74.15    29      245.89    1      0
## 5:      68.37    35      225.58    0      0
## 6:      59.99    23      226.74    1      0
```

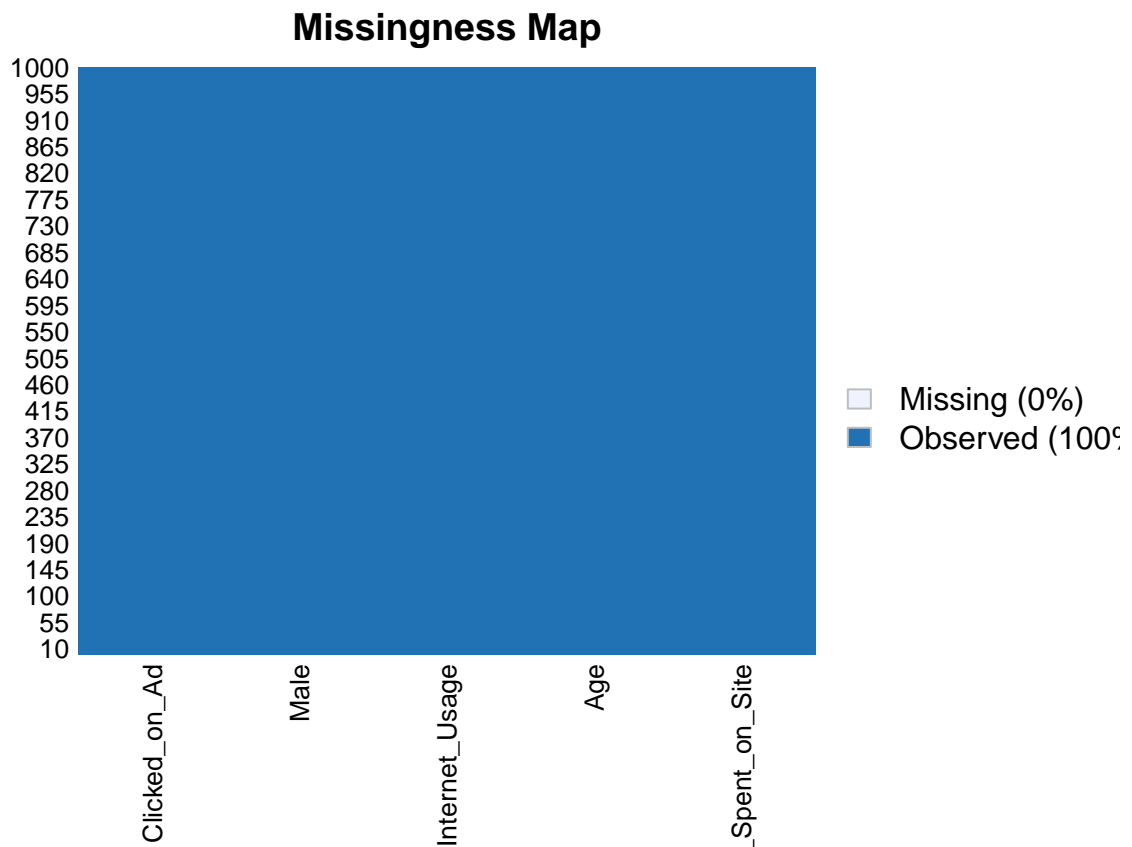
```
# Cleaning the dataset by setting zero values to NA's
```

```
df1[, 1:3][df[, 1:3] == 0] <- NA
```

Replacing the zero's to NA.

```
# Visualizing the dataset by checking for missing values
```

```
missmap(df1)
```



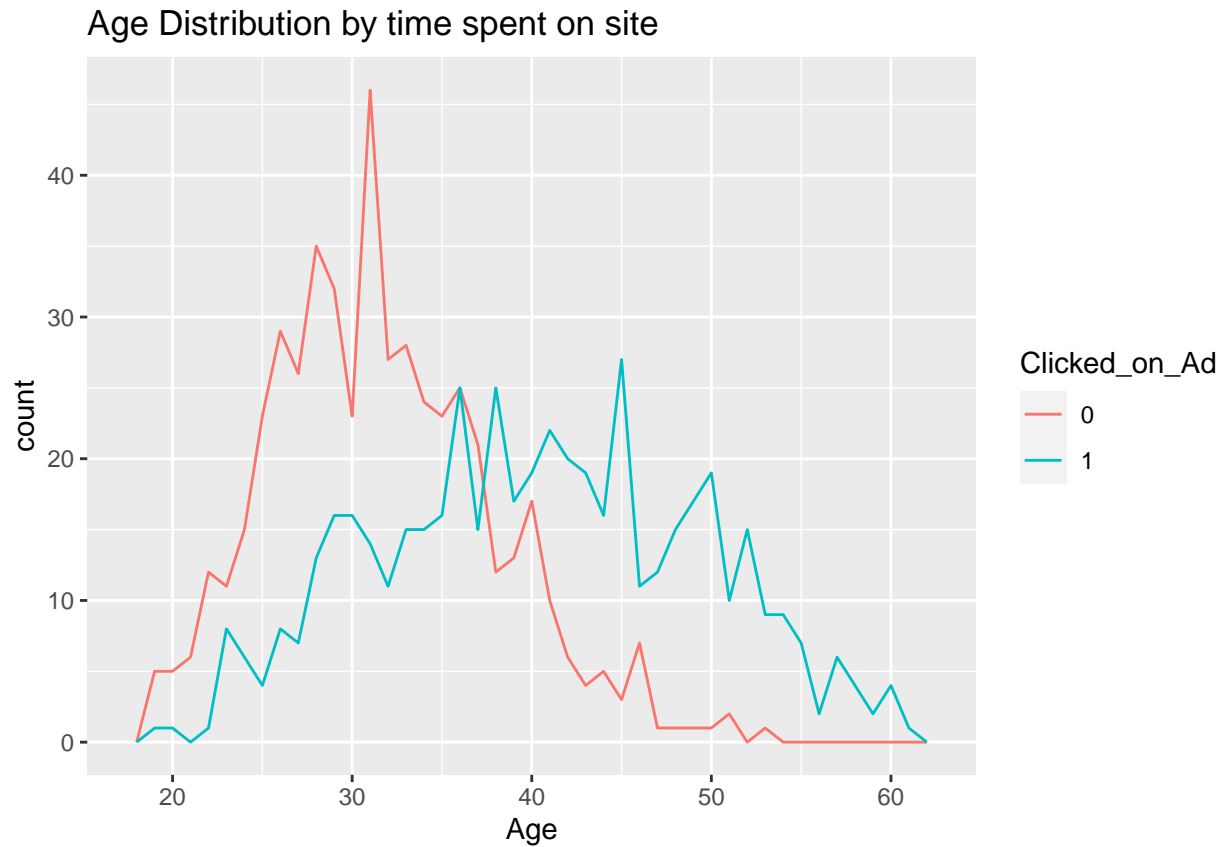
There are many missing values therefore removing them will reduce the dataset.

```
# Displaying the column names
```

```
colnames(df1)
```

```
## [1] "Daily_Time_Spent_on_Site" "Age"
## [3] "Daily_Internet_Usage"    "Male"
## [5] "Clicked_on_Ad"
```

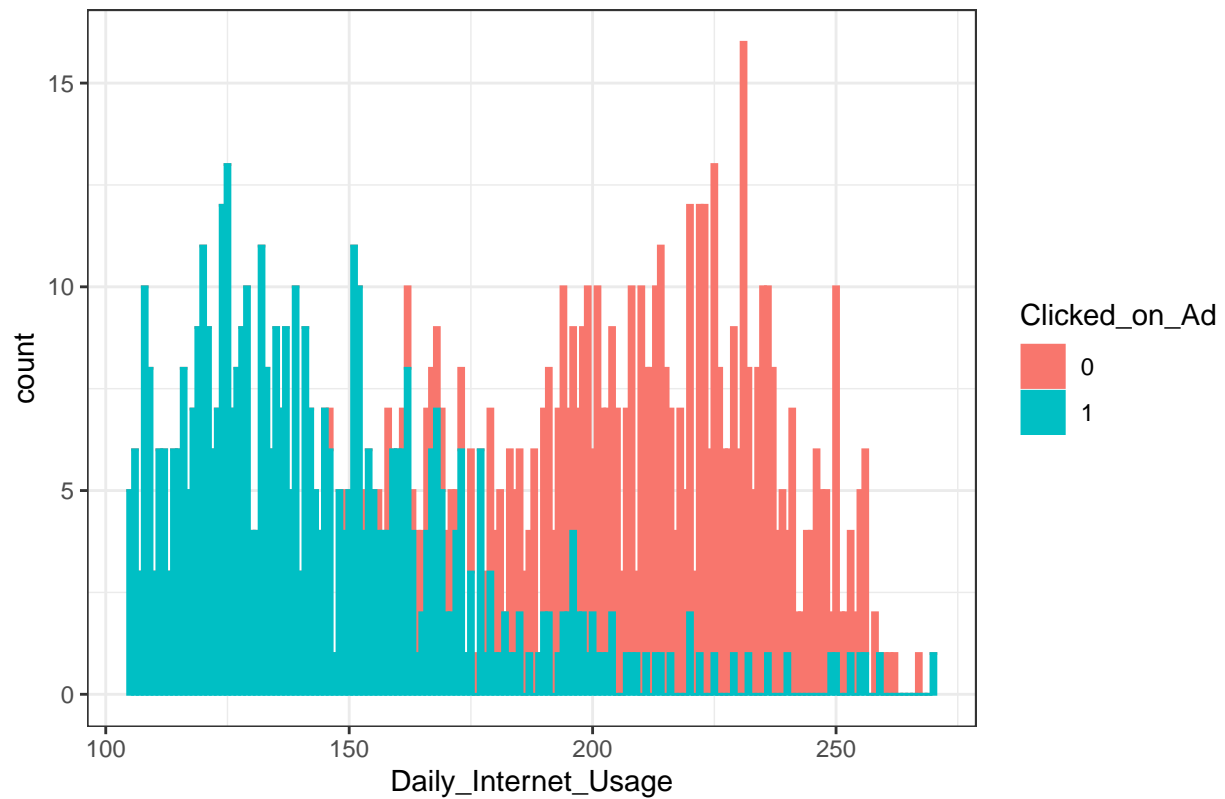
```
# Creating some visualisations to check each variable
# Visualisation 1
ggplot(df1, aes(Age, colour = Clicked_on_Ad)) +
  geom_freqpoly(binwidth = 1) + labs(title="Age Distribution by time spent on site")
```



As the age increases the count reduces.

```
# Visualisation 2
c <- ggplot(df1, aes(x=Daily_Internet_Usage, fill=Clicked_on_Ad, color=Clicked_on_Ad)) +
  geom_histogram(binwidth = 1) + labs(title="Pregnancy Distribution by time spent on site")
c + theme_bw()
```

Pregnancy Distribution by time spent on site

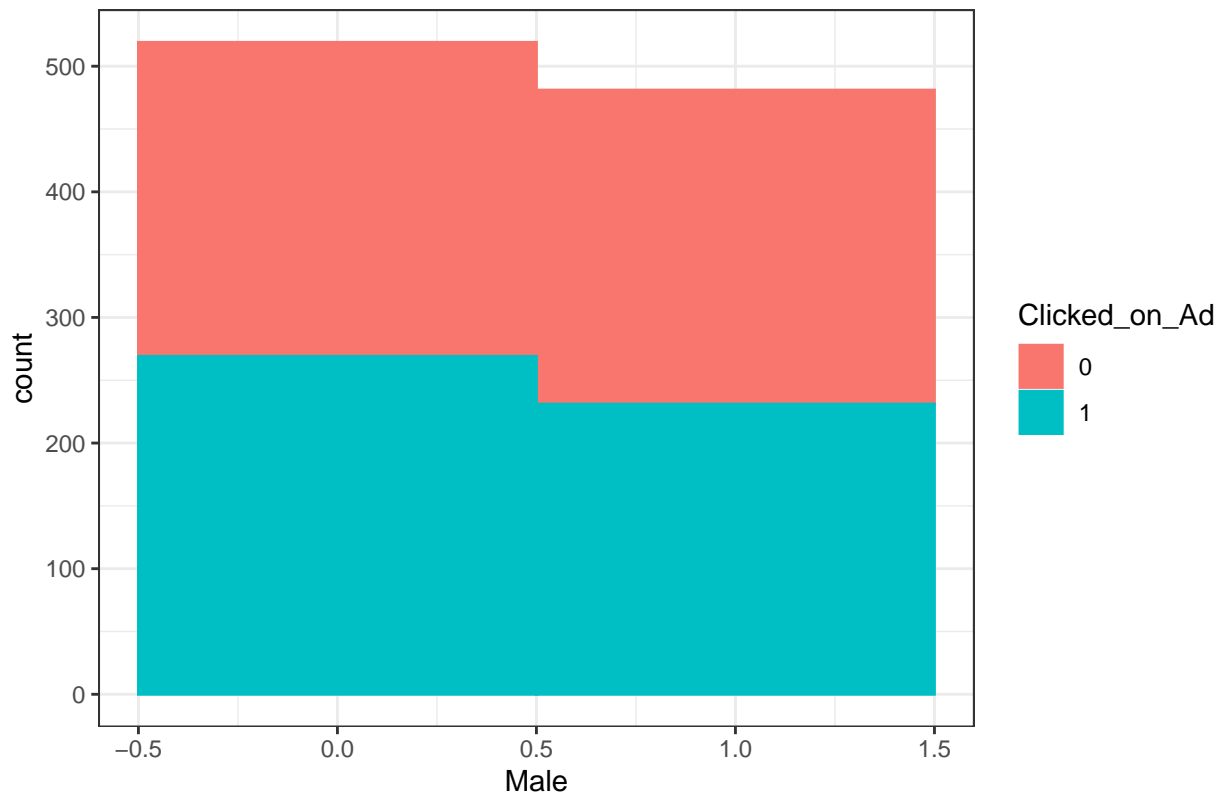


The daily internet usage increases per count.

Visualisation 3

```
P <- ggplot(df1, aes(x=Male, fill=Clicked_on_Ad, color=Clicked_on_Ad)) +
  geom_histogram(binwidth = 1) + labs(title="BMI Distribution by time spent on site")
P + theme_bw()
```

BMI Distribution by time spent on site



Splitting data into training and test data sets

```
indxTrain <- createDataPartition(y = df1$Clicked_on_Ad,p = 0.75,list = FALSE)
training <- df1[indxTrain,]
training
```

```
##      Daily_Time_Spent_on_Site  Age Daily_Internet_Usage  Male Clicked_on_Ad
##      <num> <int>          <num> <int>          <fctr>
##  1:      68.95   35          256.09    0              0
##  2:      74.15   29          245.89    1              0
##  3:      68.37   35          225.58    0              0
##  4:      59.99   23          226.74    1              0
##  5:      88.91   33          208.36    0              0
##  ---
## 746:      43.70   28          173.01    0              1
## 747:      72.97   30          208.58    1              1
## 748:      51.30   45          134.42    1              1
## 749:      51.63   51          120.37    1              1
## 750:      55.55   19          187.95    0              0
```

```
testing <- df1[-indxTrain,]
testing
```

```
##      Daily_Time_Spent_on_Site  Age Daily_Internet_Usage  Male Clicked_on_Ad
##      <num> <int>          <num> <int>          <fctr>
```

```
## 1:      80.23   31      193.77   1      0
## 2:      69.47   26      236.50   0      0
## 3:      66.00   48      131.76   1      1
## 4:      74.53   30      221.51   1      0
## 5:      82.03   41      187.53   0      0
## ---
## 246:     40.18   29      151.96   0      1
## 247:     50.48   50      162.43   0      1
## 248:     54.97   31      116.38   1      1
## 249:     67.26   57      168.41   1      1
## 250:     45.01   26      178.35   0      1
```

Displaying the test dataset,

```
# Checking dimensions of the split

prop.table(table(df1$Clicked_on_Ad)) * 100
```

```
##
##  0  1
## 50 50
```

The dimensions are 50/50 split.

```
prop.table(table(testing$Clicked_on_Ad)) * 100
```

```
##
##  0  1
## 50 50
```

The dimensions for the testing clicked ad are 50/50.

```
prop.table(table(training$Clicked_on_Ad)) * 100
```

```
##
##  0  1
## 50 50
```

The dimensions for the training clicked ad are 50/50.

```
# Creating objects x which holds the predictor variables and y which holds the response variables

x = training[,-5]
y = training$Clicked_on_Ad
```

x and y predictor variables.

```
# Loading our inbuilt e1071 package that holds the Naive Bayes function.

library(e1071)
```

```
# Building the model
```

```
model = train(x,y,'nb',trControl=trainControl(method='cv',number=10))
```

```
## Warning in FUN(X[[i]], ...): Numerical 0 probability for all classes with  
## observation 18
```

```
## Warning in FUN(X[[i]], ...): Numerical 0 probability for all classes with  
## observation 8
```

```
## Warning in FUN(X[[i]], ...): Numerical 0 probability for all classes with  
## observation 30
```

```
# Model Evalution
```

```
# Predicting our testing set
```

```
Predict <- predict(model, newdata = testing)
```

```
# Confusion matrix to see accuracy value and other parameter values
```

```
confusionMatrix(Predict, testing$Clicked_on_Ad )
```

```
## Confusion Matrix and Statistics
```

```
##
```

```
##           Reference
```

```
## Prediction  0    1
```

```
##           0 123    9
```

```
##           1   2 116
```

```
##
```

```
##           Accuracy : 0.956
```

```
##           95% CI : (0.9226, 0.9778)
```

```
## No Information Rate : 0.5
```

```
## P-Value [Acc > NIR] : < 2e-16
```

```
##
```

```
##           Kappa : 0.912
```

```
##
```

```
## McNemar's Test P-Value : 0.07044
```

```
##
```

```
##           Sensitivity : 0.9840
```

```
##           Specificity : 0.9280
```

```
##           Pos Pred Value : 0.9318
```

```
##           Neg Pred Value : 0.9831
```

```
##           Prevalence : 0.5000
```

```
##           Detection Rate : 0.4920
```

```
##           Detection Prevalence : 0.5280
```

```
##           Balanced Accuracy : 0.9560
```

```
##
```

```
##           'Positive' Class : 0
```

```
##
```

The confusion matrix of the model.

KNN

```
set.seed(1234)
```

```
# Randomizing the rows, creates a uniform distribution of 150
random <- runif(150)
df1_random <- df1[order(random),]
df1_random
```

```
##      Daily_Time_Spent_on_Site    Age Daily_Internet_Usage  Male Clicked_on_Ad
##                                <num> <int>                <num> <int>        <fctr>
##  1:                        88.91    33                   208.36    0            0
##  2:                        86.06    32                   178.92    1            0
##  3:                        55.35    39                   153.17    1            1
##  4:                        39.94    41                   156.30    0            1
##  5:                        41.49    53                   169.18    0            1
## ---
## 146:                       46.98    50                   175.37    0            1
## 147:                       49.78    46                   152.24    0            1
## 148:                       37.51    30                   163.00    1            1
## 149:                       50.43    46                   119.32    1            1
## 150:                       77.65    27                   212.79    0            0
```

```
# Selecting the first 6 rows from iris_random
head(df1_random)
```

```
##      Daily_Time_Spent_on_Site    Age Daily_Internet_Usage  Male Clicked_on_Ad
##                                <num> <int>                <num> <int>        <fctr>
## 1:                        88.91    33                   208.36    0            0
## 2:                        86.06    32                   178.92    1            0
## 3:                        55.35    39                   153.17    1            1
## 4:                        39.94    41                   156.30    0            1
## 5:                        41.49    53                   169.18    0            1
## 6:                        74.02    32                   210.54    0            0
```

```
normal <- function(x) (
  return( ((x - min(x)) / (max(x)-min(x))) )
)
normal(1:5)
```

```
## [1] 0.00 0.25 0.50 0.75 1.00
```

```
df1_new <- as.data.frame(lapply(df1_random[,5], normal))
summary(df1_new)
```

```
##      Daily_Time_Spent_on_Site    Age    Daily_Internet_Usage
## Min.      :0.0000      Min. :0.0000  Min.      :0.0000
## 1st Qu.:0.2986      1st Qu.:0.2500  1st Qu.:0.1903
## Median :0.5959      Median :0.4000  Median :0.4430
## Mean     :0.5425      Mean   :0.4138  Mean     :0.4427
```

```
## 3rd Qu.:0.7752          3rd Qu.:0.5750   3rd Qu.:0.6633
## Max.    :1.0000          Max.    :1.0000   Max.    :1.0000
##      Male
## Min.    :0.0000
## 1st Qu.:0.0000
## Median :0.0000
## Mean    :0.4533
## 3rd Qu.:1.0000
## Max.    :1.0000
```

Normalization of the dataset.

```
# Lets now create test and train data sets
```

```
train <- df1_new[1:130,]
test  <- df1_new[131:150,]
train_sp <- df1_random[1:130,5]
test_sp <- df1_random[131:150,5]
```

Testing and training the dataset.

```
df1_train_labels <- df1[1:65, 1]
df1_test_labels  <- df1[66:100, 1]
```

5. CONCLUSION

The people that clicked on the ads on the blog were aged between 19 yrs and 61 years old.

The internet usage ranged between 104.8 to 269 units with the time spent on the blog was between 32 to 91 minutes.

There was a negative correlation between age and daily time spent on Site of the individuals.

The ads were mostly viewed by the young and middle aged audience.

6. RECOMMENDATIONS

The ads that should be placed on the blog should be relevant to the ages so that the individuals can click on the ads.

For the older people they can minimize the ads and for the younger people they can maximize the ads so that each can relate to ads accordingly.