

# R-Fundamentals

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## Defining the Question

To identify which individuals are most likely to click on her ads.

## Metric for success

To be able to identify who is likely to click on the ads

## The Context

A Kenyan entrepreneur has created an online cryptography course and would want to advertise it on her blog. She currently targets audiences originating from various countries. In the past, she ran ads to advertise a related course on the same blog and collected data in the process. She would now like to employ your services as a Data Science Consultant to help her identify which individuals are most likely to click on her ads.

## Experimental Design

1. Loading dataset into R
2. Cleaning the dataset
3. Perform EDA
4. Perform Univariate and Bivariate Analysis
5. Provide conclusions and recommendations

## Loading the dataset

```
#library(knitr)  
#setwd("C:/Users/Desktop/MORINGA_CORE/R PROGRAMMING/R WEEK 1/R WEEK ONE EXERCISES") # Change working directory  
  
# Properly import data  
library(data.table)  
advert <- fread("http://bit.ly/IPAdvertisingData")
```

```
# Loading the dataset
# If .csv file, use this
# advertising.csv
#library(data.table)
#advert <- read.csv(file.choose())

#advert <- read.csv("advertising.csv")
```

## Previewing the dataset

```
#previewing the first 6 rows of the dataset
head(advert)
```

```
##      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
##      <POS<      <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
```

## EDA

### Exploring the Dataset

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

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

```
#The dataset has 1000 rows and 10 columns
```

```
#Finding the datatypes of the dataset  
str(advert)
```

```
## Classes 'data.table' and 'data.frame':  1000 obs. of  10 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 ...  
## $ Area Income : num  61834 68442 59786 54806 73890 ...  
## $ Daily Internet Usage : num  256 194 236 246 226 ...  
## $ Ad Topic Line : chr  "Cloned 5thgeneration orchestration" "Monitored national standardi  
## $ City : chr  "Wrightburgh" "West Jodi" "Davidton" "West Terrifurt" ...  
## $ Male : int  0 1 0 1 0 1 0 1 1 1 ...  
## $ Country : chr  "Tunisia" "Nauru" "San Marino" "Italy" ...  
## $ Timestamp : POSIXct, format: "2016-03-27 00:53:11" "2016-04-04 01:39:02" ...  
## $ Clicked on Ad : int  0 0 0 0 0 0 0 1 0 0 ...  
## - attr(*, ".internal.selfref")=<externalptr>
```

## Data Cleaning

### Editing column names

```
# assigning new names to the columns of the data frame  
colnames(advert) <- c('Daily.Time.Spent.on.Site', 'Age', 'Area.Income', 'Daily.Internet.Usage', 'Ad.Topic.L  
# printing new data frame  
print("New data frame : ")
```

```
## [1] "New data frame : "
```

```
print(advert)
```

```
##      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> <POS< <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
```

## Missing Values

```
#Checking for the sum of Missing values
colSums(is.na(advert))
```

```
## Daily.Time.Spent.on.Site Age Area.Income
## 0 0 0
## Daily.Internet.Usage Ad.Topic.Line City
## 0 0 0
## Male Country Timestamp
## 0 0 0
## Clicked.on.Ad
## 0
```

```
#There are no missing values in this dataset
```

## Duplicates

```
#Checking for duplicates in the dataset
advert.duplicates <- advert[duplicated(advert),]

#printing duplicated rows
advert.duplicates
```

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

```
#There are no duplicated rows in the dataset
```

## Outliers

```
#checking for dataframe class  
class(advert)
```

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

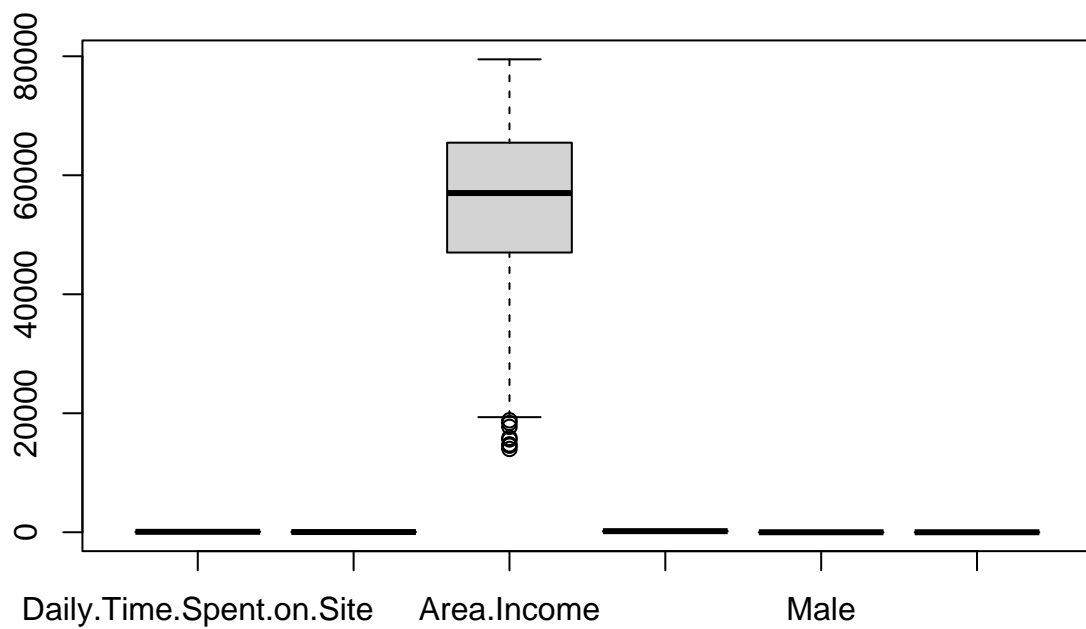
```
#Extracting numeric columns to analyse for outliers  
num.cols <- unlist(lapply(advert, is.numeric))  
  
#printing numeric columns  
num.cols
```

```
## Daily.Time.Spent.on.Site      Age      Area.Income  
##           TRUE                TRUE          TRUE  
##   Daily.Internet.Usage      Ad.Topic.Line      City  
##           TRUE                FALSE          FALSE  
##           Male              Country            Timestamp  
##           TRUE                FALSE          FALSE  
##           Clicked.on.Ad  
##           TRUE
```

```
#creating a dataframe with numeric columns only so as to plot a boxplot  
advert.numeric <-advert[, ..num.cols]  
  
#checking the data types, previewing  
str(advert.numeric)
```

```
## Classes 'data.table' and 'data.frame':  1000 obs. of  6 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 ...  
## $ Area.Income : num  61834 68442 59786 54806 73890 ...  
## $ 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>
```

```
#Plotting a boxplot to check for outliers  
#library(data.table)  
boxplot(advert.numeric)
```



```
#there are outliers in Area.income column
```

```
#Plotting a boxplot to check for outliers in Area.Income column
```

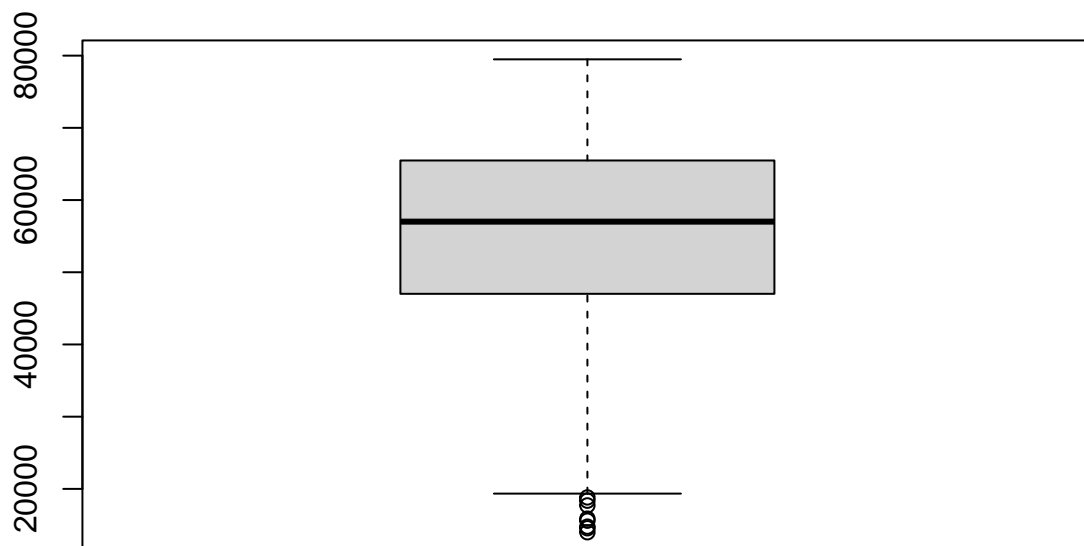
```
#for(y in 1:ncol(advert.numeric)){
```

```
  # if (is.na(advert.numeric[1,y])) advert.numeric[1,y] = 0
```

```
  #}
```

```
#+
```

```
boxplot(advert.numeric$Area.Income)#,ylim=c(0,300), main = 'Boxplot of Area Income')$out
```



```
#there are some records appearing as outliers in the lower quartile of the Area.Income column
#These will be removed before we begin analysis
```

```
#Removing outliers in the lower quartile of the Area.Income
```

```
Q1 <- quantile(advert$Area.Income, .25)
```

```
Q3 <- quantile(advert$Area.Income, .75)
```

```
IQR <- IQR(advert$Area.Income)
```

```
#Keeping values above 1.5*IQR of Q1
```

```
no.outliers <- subset(advert, advert$Area.Income > (Q1 - 1.5*IQR)) #& advert$Area.Income < (Q3 + 1.5*IQR)
dim(no.outliers)
```

```
## [1] 991 10
```

```
dim(advert)
```

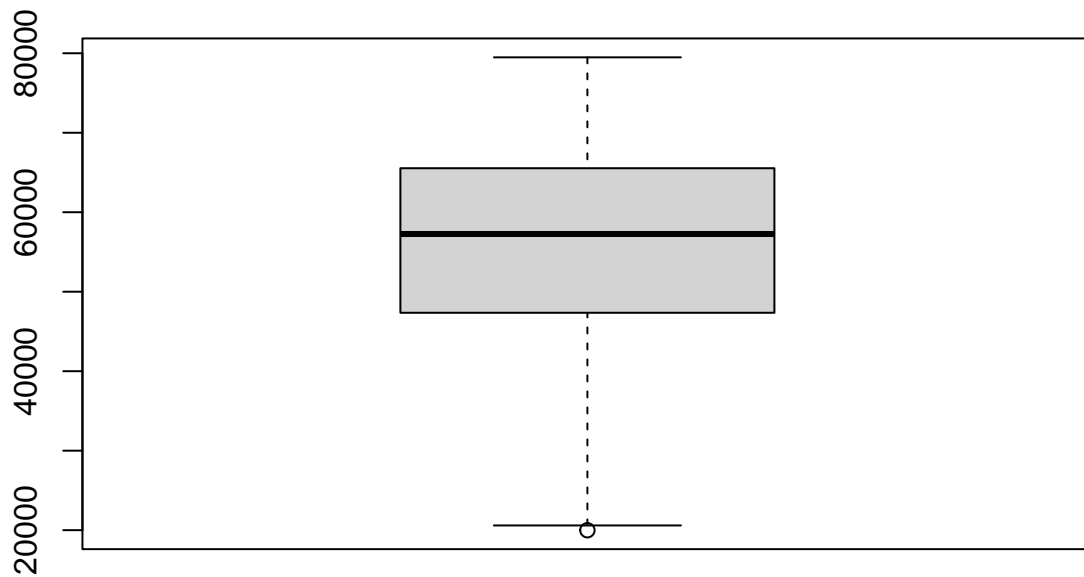
```
## [1] 1000 10
```

```
#9 records were dropped
```

```
#Plotting a boxplot to check if outliers in Area.Income column have been dropped
```

```
boxplot(no.outliers$Area.Income, main = 'Boxplot of Area Income')
```

## Boxplot of Area Income



*#the boxplot indicates that the outliers have been removed*

## Univariate Analysis

*#previewing the new dataset without outliers*  
head(no.outliers)

```
##      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
##          <POS<          <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
```

```
#checking summary statistics
summary(no.outliers)
```

```
##   Daily.Time.Spent.on.Site      Age      Area.Income  Daily.Internet.Usage
##   Min.   :32.60             Min.   :19.00   Min.   :19992   Min.   :104.8
##   1st Qu.:51.34             1st Qu.:29.00   1st Qu.:47348   1st Qu.:138.6
##   Median :68.41             Median :35.00   Median :57260   Median :183.4
##   Mean   :65.06             Mean   :35.99   Mean   :55349   Mean   :180.0
##   3rd Qu.:78.59             3rd Qu.:42.00   3rd Qu.:65538   3rd Qu.:218.9
##   Max.   :91.43             Max.   :61.00   Max.   :79485   Max.   :270.0
##   Ad.Topic.Line      City      Male      Country
##   Length:991      Length:991      Min.   :0.0000   Length:991
##   Class :character  Class :character  1st Qu.:0.0000   Class :character
##   Mode  :character  Mode  :character  Median :0.0000   Mode  :character
##                                     Mean   :0.4793
##                                     3rd Qu.:1.0000
##                                     Max.   :1.0000
##   Timestamp      Clicked.on.Ad
##   Min.   :2016-01-01 02:52:10.00   Min.   :0.0000
##   1st Qu.:2016-02-17 22:51:14.50   1st Qu.:0.0000
##   Median :2016-04-07 03:56:16.00   Median :0.0000
##   Mean   :2016-04-10 02:20:21.53   Mean   :0.4955
##   3rd Qu.:2016-05-31 01:37:57.50   3rd Qu.:1.0000
##   Max.   :2016-07-24 00:22:16.00   Max.   :1.0000
```

```
#Extracting a numeric subset from the no outliers dataset
no.out.num.cols <-unlist(lapply(no.outliers, is.numeric))
#Extracting numeric columns to analyse for outliers
#num.cols <- unlist(lapply(advert, is.numeric))

#printing numeric columns
no.out.num.cols
```

```
##   Daily.Time.Spent.on.Site      Age      Area.Income
##   TRUE                        TRUE      TRUE
##   Daily.Internet.Usage      Ad.Topic.Line      City
##   TRUE                        FALSE      FALSE
##   Male                        Country      Timestamp
##   TRUE                        FALSE      FALSE
##   Clicked.on.Ad
##   TRUE
```

```
#creating a dataframe with numeric columns only so as to plot a boxplot
no.outliers.numeric <-no.outliers[, ..no.out.num.cols]
```

```
#previewing
head(no.outliers.numeric)
```

```
##      Daily.Time.Spent.on.Site      Age Area.Income Daily.Internet.Usage  Male
##                                <num> <int>         <num>           <num> <int>
## 1:                68.95      35      61833.90           256.09      0
## 2:                80.23      31      68441.85           193.77      1
## 3:                69.47      26      59785.94           236.50      0
## 4:                74.15      29      54806.18           245.89      1
## 5:                68.37      35      73889.99           225.58      0
## 6:                59.99      23      59761.56           226.74      1
##      Clicked.on.Ad
##                <int>
## 1:                0
## 2:                0
## 3:                0
## 4:                0
## 5:                0
## 6:                0
```

```
#checking the data types, previewing
#str(no.outliers.numeric)
```

## Measures of Central Tendency

### i) Mean

```
#means of all numeric columns in the dataset
#this has been extracted from the dataset and named no.outliers.numeric
#the variable for the column means is no.out.col.means
```

```
no.out.col.means <- colMeans(data.frame(no.outliers.numeric))
```

```
# Printing out
# ---
#
no.out.col.means
```

```
## Daily.Time.Spent.on.Site      Age      Area.Income
##      6.505689e+01      3.598587e+01      5.534910e+04
##      Daily.Internet.Usage      Male      Clicked.on.Ad
##      1.799846e+02      4.793138e-01      4.954591e-01
```

The average daily time spent on site was 65.05 units. The average area income was 55,349 units. The average age of respondents was 35.98 years. The average daily internet usage was 179.98 units.

## ii) Median

```
#median of all numeric columns in the dataset
#this has been extracted from the dataset and named no.outliers.numeric
#the variable for the column means is no.out.col.median
library(matrixStats)

no.out.col.median <- colMedians(as.matrix.data.frame(no.outliers.numeric))

# Printing out
# ---
#
print(no.out.col.median)
```

```
## [1]    68.41    35.00 57260.41   183.43     0.00     0.00
```

The median of the daily time spent on site was 68.41 units. The median of the area income was 57,260.41 units. The median of the ages of the respondents was 35 years. The median of the daily internet usage was 183.43 units.

## iii) Mode

```
# We create the mode function that will perform our mode operation for us
# The mode will give us values that appeared the most number of times
# ---
# library(purrr)
FindMode <- function(no.outliers) {
  uniqv <- unique(no.outliers)
  uniqv[which.max(tabulate(match(no.outliers, uniqv)))]
}

# Calculating the mode using out getmode() function
# ---
#
#no.out.col.mode <- getmode(as.matrix(no.outliers.numeric))
no.out.col.mode <- data.frame(no.outliers)

# Printing out
# ---
#
apply(no.out.col.mode, 2, FindMode)
```

```
##           Daily.Time.Spent.on.Site           Age
##                               "62.26"          "31"
##           Area.Income           Daily.Internet.Usage
##                               "61833.90"         "167.22"
##           Ad.Topic.Line           City
## "Cloned 5thgeneration orchestration"         "Lisamouth"
##                               Male           Country
##                               "0"           "Czech Republic"
```

```
##                                Timestamp                                Clicked.on.Ad
##                                "2016-03-27 00:53:11"                                "0"
```

```
#The modes of all the variables, both categorical and numerical are as follows:
# For factors male and clicked on ad, 0 = no and 1= yes
# There were lesser male respondents
# Most respondents did not click on the adverts
```

## Measures of Dispersion

We will use the numeric data-frame while calculating measures of dispersion

### i)Minimum

```
# Minimum
#min <-colMins(as.matrix(no.outliers.numeric[sapply(no.outliers.numeric, is.numeric)]))

#printing
#min

sapply(no.outliers.numeric, min)
```

```
## Daily.Time.Spent.on.Site      Age      Area.Income
##                32.60      19.00      19991.72
##      Daily.Internet.Usage      Male      Clicked.on.Ad
##                104.78      0.00                0.00
```

The minimum of the daily time spent on site was 32.60 units. The minimum of the area income was 19,991.72 units. The minimum of the ages of the respondents was 19 years. The minimum of the daily internet usage was 104.78 units. The minimum value of whether male or not is 0. The minimum value of whether clicked on advert or not is 0.

### ii)Maximum

```
#minimum, maximum, range, quantile, variance
# and standard deviation
#max <-colMaxs(as.matrix(data.frame(no.outliers.numeric[sapply(no.outliers.numeric, is.numeric)]))

# previewing
#max
#max

sapply(no.outliers.numeric, max)
```

```
## Daily.Time.Spent.on.Site      Age      Area.Income
##                91.43      61.00      79484.80
##      Daily.Internet.Usage      Male      Clicked.on.Ad
##                269.96      1.00                1.00
```

The maximum of the daily time spent on site was 91.43 units. The maximum of the area income was 79,484.80 units. The maximum of the ages of the respondents was 61 years. The maximum of the daily internet usage was 269.96 units. The maximum value of whether male or not is 1. The maximum value of whether clicked on advert or not is 1. ### iii) Variance

```
# Finding the variance of all the variables
# area <-sd(no.outliers.numeric$Area.Income)
sapply(no.outliers.numeric, var)
```

##	Daily.Time.Spent.on.Site	Age	Area.Income
##	2.528258e+02	7.752303e+01	1.680004e+08
##	Daily.Internet.Usage	Male	Clicked.on.Ad
##	1.940743e+03	2.498242e-01	2.502319e-01

The variance of the daily time spent on site was 252.82. The variance of the area income was 168,000,385. The variance of the ages of the respondents was 77.52. The variance of the daily internet usage was 1940.74. The variance of male column is 0.2498 . The variance of whether ad was clicked or not 0.2502 .

#### iv) Standard Deviation

```
# Finding the standard deviation for all numeric variables
sapply(no.outliers.numeric, sd)
```

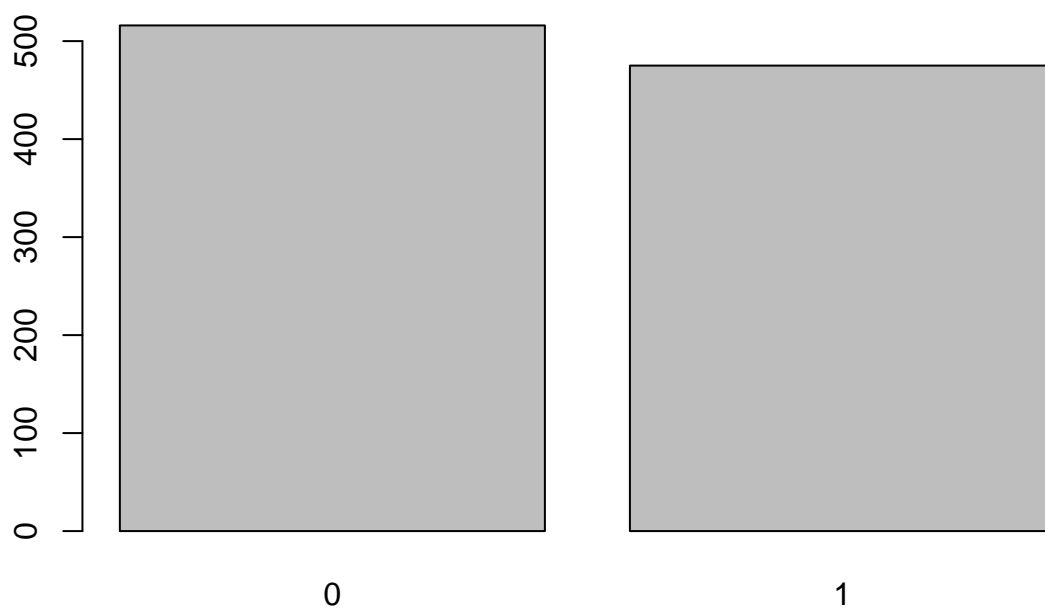
##	Daily.Time.Spent.on.Site	Age	Area.Income
##	1.590050e+01	8.804716e+00	1.296150e+04
##	Daily.Internet.Usage	Male	Clicked.on.Ad
##	4.405386e+01	4.998241e-01	5.002318e-01

The standard deviation of the daily time spent on site was 15.90. The standard deviation of the area income was 12,961.5. The standard deviation of the ages of the respondents was 8.80. The standard deviation of the daily internet usage was 44.05.

## Univariate Graphicals

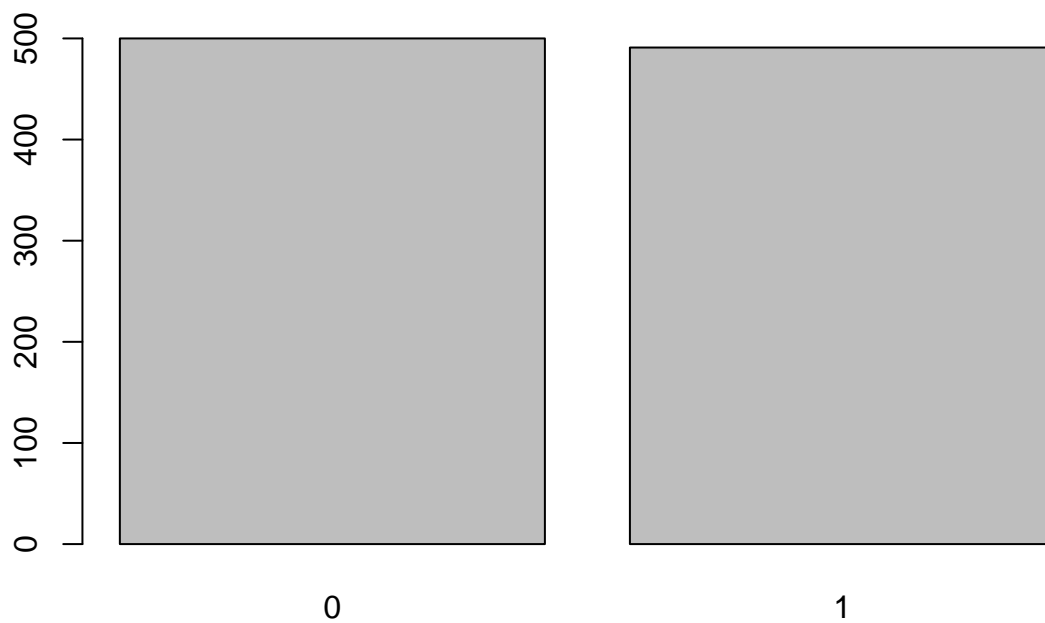
```
# Plotting a bar-graph to see the frequency of the categorical variables
# The table() function computes the frequency distribution of the categorical variables

# for the male column
barplot(table(no.outliers.numeric$Male))
```



```
# The respondents who were not males were fewer than those who were males
```

```
# for the male column  
barplot(table(no.outliers.numeric$Clicked.on.Ad))
```



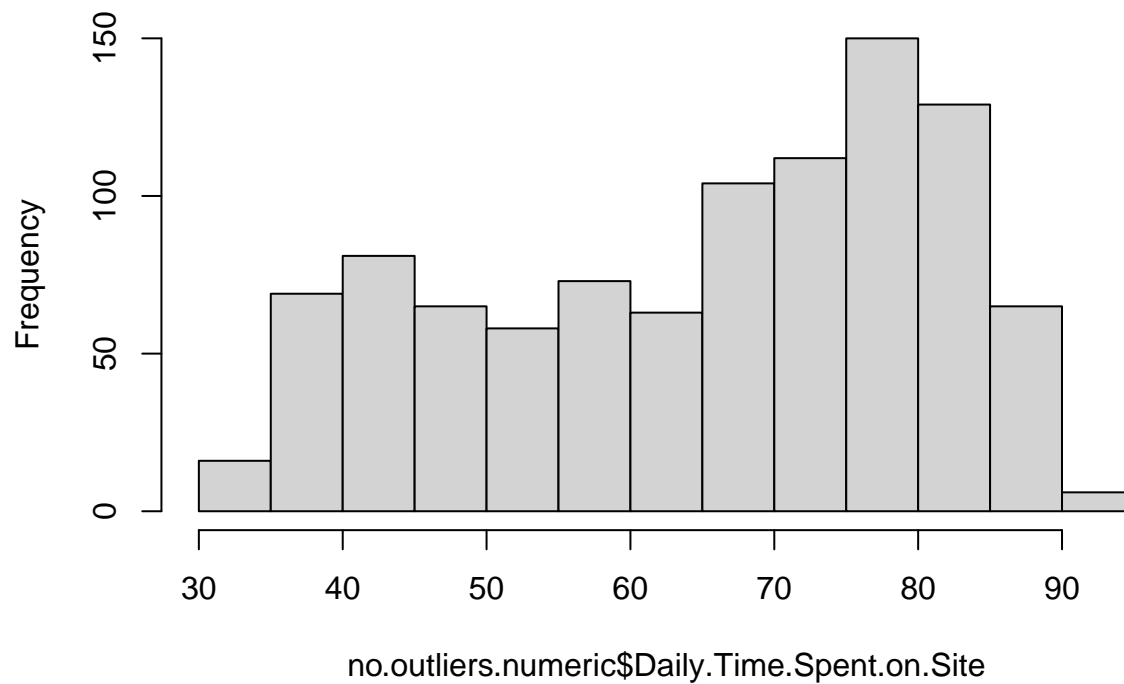
```
# The number of respondents who clicked and who did not click on adverts were almost the same
```

```
# Plotting histograms to show the distribution of the numerical variables
```

```
# Histogram of time spent on site
```

```
hist(no.outliers.numeric$Daily.Time.Spent.on.Site, main = "Histogram of Time spent on Site")
```

## Histogram of Time spent on Site



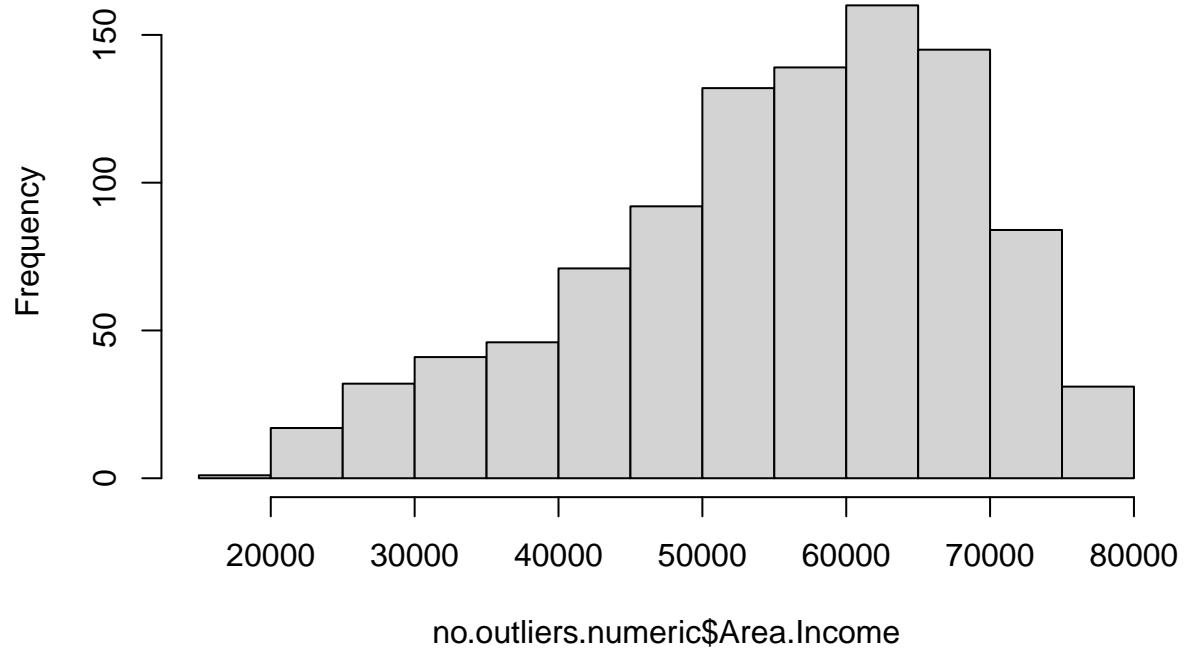
*# The time spent on sight is not skewed, meaning the data points tend to be evenly distributed*

*# Histogram of area income*

```
hist(no.outliers.numeric$ Area.Income, main = "Histogram of Area Income")
```

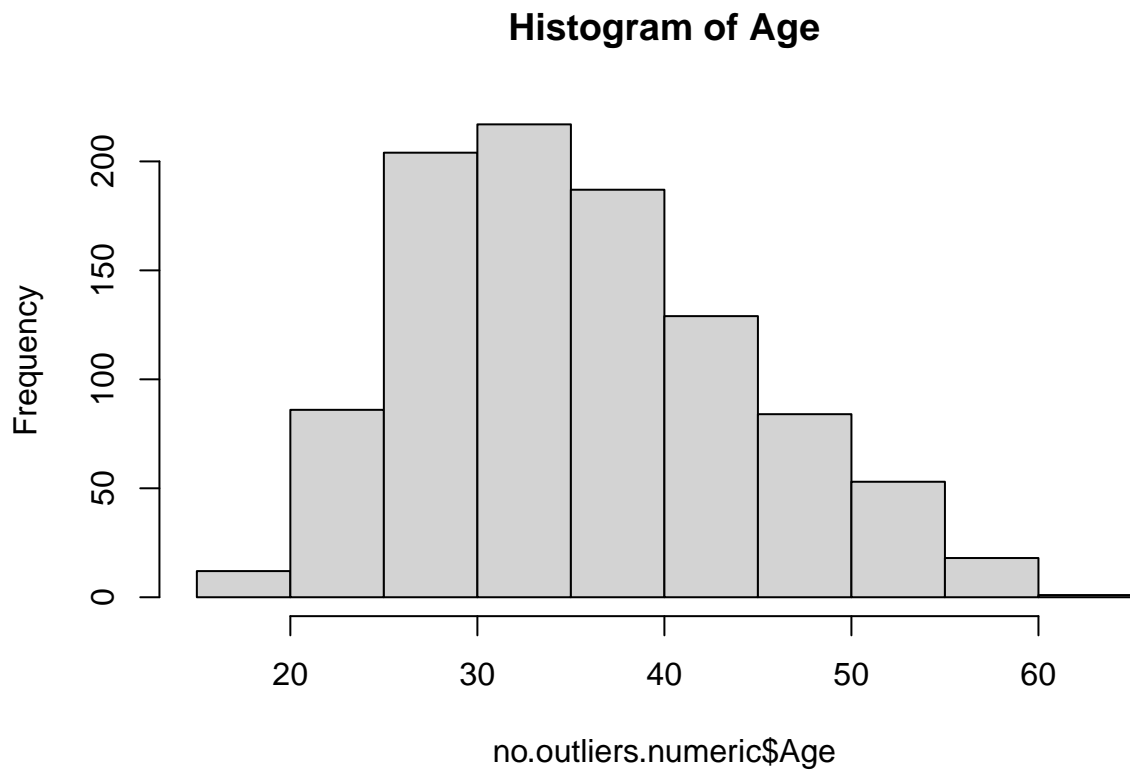


## Histogram of Area Income



*# The area income is left skewed, meaning the data points extend to the left of the distribution*

```
# Histogram of Age  
hist(no.outliers.numeric$Age, main = "Histogram of Age")
```

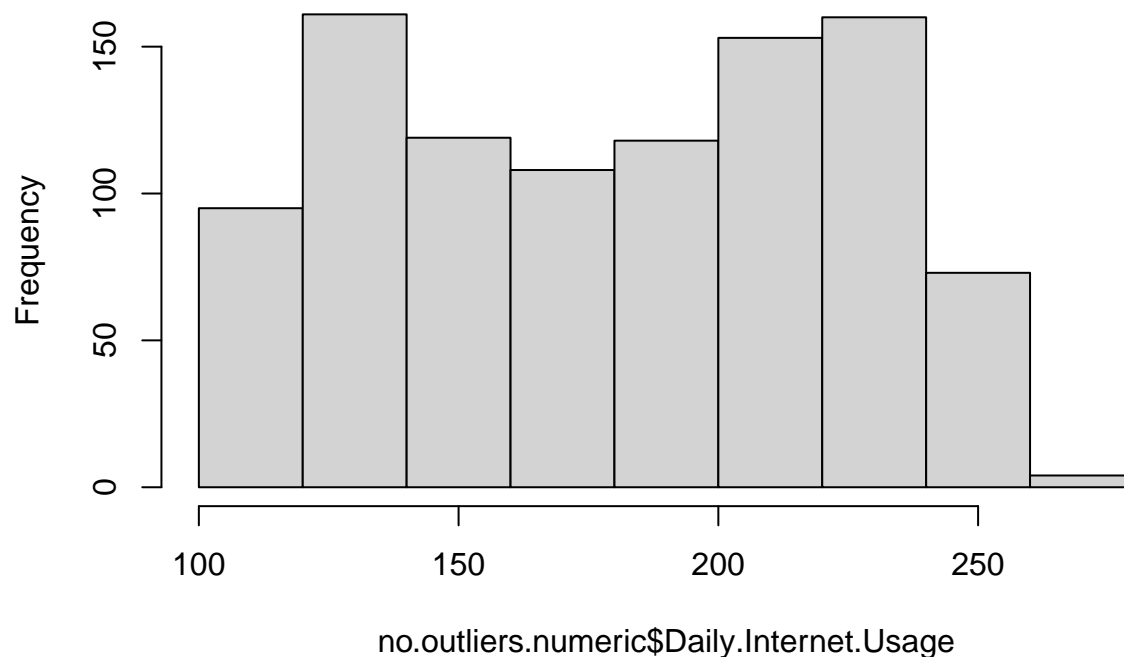


*# The age variable is right skewed, meaning the data points extend to the right of the data points distribution*

*# Histogram of daily internet usage*

```
hist(no.outliers.numeric$Daily.Internet.Usage, main = "Histogram of Daily Internet Usage")
```

## Histogram of Daily Internet Usage



*# Daily internet usage is not skewed, meaning the data points tend to be normally distributed*

### Categorical data analysis

```
# Checking the number of countries  
# Checking the unique entries  
countries <-unique(no.outliers$Country)  
  
# printing the number of unique countries  
# we will use the length function to do a unique value count  
length(countries)
```

```
## [1] 237
```

*# there are 237 countries in the dataset*

```
# Checking the number of cities  
# Checking the unique entries  
cities <-unique(no.outliers$City)  
  
# printing the number of unique cities  
# we will use the length function to do a unique value count  
length(cities)
```

```
## [1] 960
```

```
#there are 960 cities in the dataset
```

## Bivariate Analysis

### i) Covariance

```
# previewing  
head(no.outliers)
```

```
##      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  
##      <POS<      <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
```

```
# finding the covariance of the target variable and other numerical variables  
# we assign different variables for the specific columns
```

```
# Assigning Daily.Time.Spent.on.Site column to variable time.site  
time.site <- no.outliers$Daily.Time.Spent.on.Site
```

```
# Assigning Age column to variable age  
age <-no.outliers$Age
```

```
# Assigning Area.income column to variable area.income  
area.income <-no.outliers$Area.Income
```

```
# Assigning Daily.Internet.Usage column to variable daily.internet  
daily.internet <-no.outliers$Daily.Internet.Usage
```

```
# Assigning Male column to variable male
male <-no.outliers$Male

# Assigning clicked on ads column to variable clicks.target
clicks.target <-no.outliers$Clicked.on.Ad
```

```
# Finding co-variances of the numerical variables

# covariance of age and time spent on site
cov(time.site,age )
```

```
## [1] -46.59899
```

```
# negative linear relationship between the variables
```

```
# Finding co-variances of the numerical variables

# covariance of age and time spent on site
cov(time.site,area.income )
```

```
## [1] 64600.67
```

```
# strong positive linear relationship between the variables
```

```
# Finding co-variances of the numerical variables

# covariance of age and time spent on site
cov(time.site,daily.internet)
```

```
## [1] 364.2711
```

```
#positive linear relationship between the variables
```

```
# Finding co-variances of the numerical variables

# covariance of age and time spent on site
cov(age,area.income )
```

```
## [1] -20744.22
```

```
# strong negative linear relationship between the variables
```

```
# covariance of age and time spent on site
cov(age,daily.internet )
```

```
## [1] -142.7226
```

```
# negative linear relationship between the variables
```

```
#covariance
```

```
cov(area.income,daily.internet )
```

```
## [1] 201115
```

```
# strong positive linear relationship between the variables
```

## ii) Correlation

We will use the numeric dataframe

```
# correlation matrix
```

```
ad_cor <- cor(no.outliers.numeric, use="pairwise.complete.obs",method = "pearson")  
round(ad_cor, 2)
```

```
##           Daily.Time.Spent.on.Site   Age Area.Income  
## Daily.Time.Spent.on.Site           1.00 -0.33      0.31  
## Age                               -0.33  1.00     -0.18  
## Area.Income                       0.31 -0.18      1.00  
## Daily.Internet.Usage               0.52 -0.37      0.35  
## Male                             -0.02 -0.02      0.01  
## Clicked.on.Ad                     -0.75  0.49     -0.47  
##           Daily.Internet.Usage   Male Clicked.on.Ad  
## Daily.Time.Spent.on.Site         0.52 -0.02     -0.75  
## Age                             -0.37 -0.02      0.49  
## Area.Income                      0.35  0.01     -0.47  
## Daily.Internet.Usage              1.00  0.03     -0.79  
## Male                             0.03  1.00     -0.04  
## Clicked.on.Ad                    -0.79 -0.04      1.00
```

```
# gives correlation co-efficients in pairs and rounding them off to decimal places
```

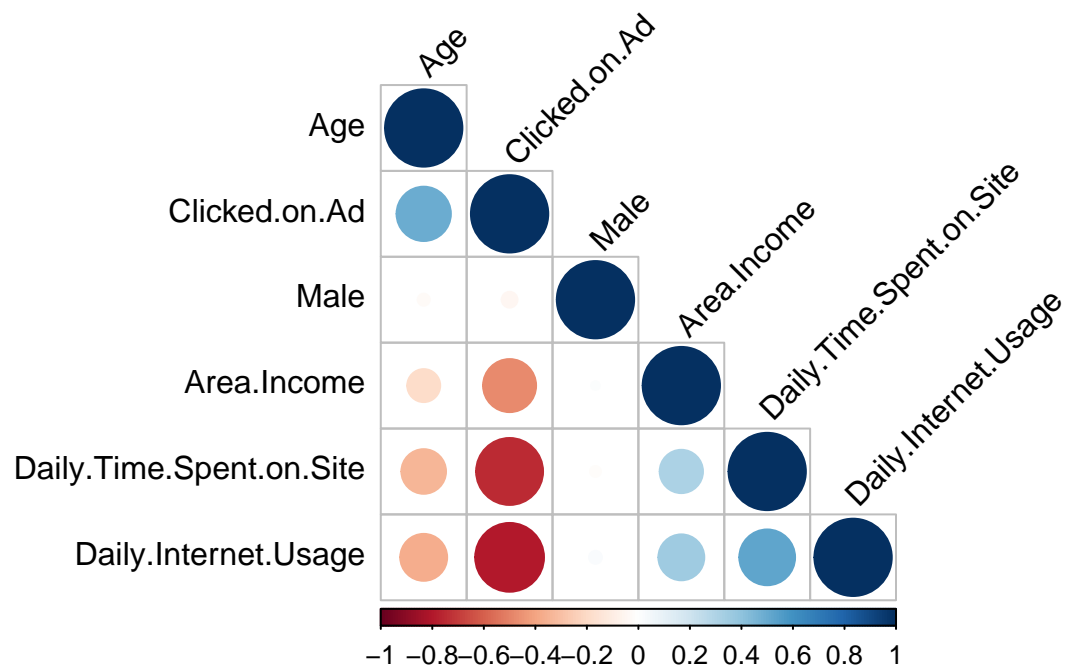
```
# When the correlation the coefficient value is next to 1 it shows a positive linear relationship,  
# when next to -1, it indicates that the variables are negatively linearly related  
# When close to zero, it would indicate a weak linear relationship between the variables.
```

```
# Visualizing the correlation matrix
```

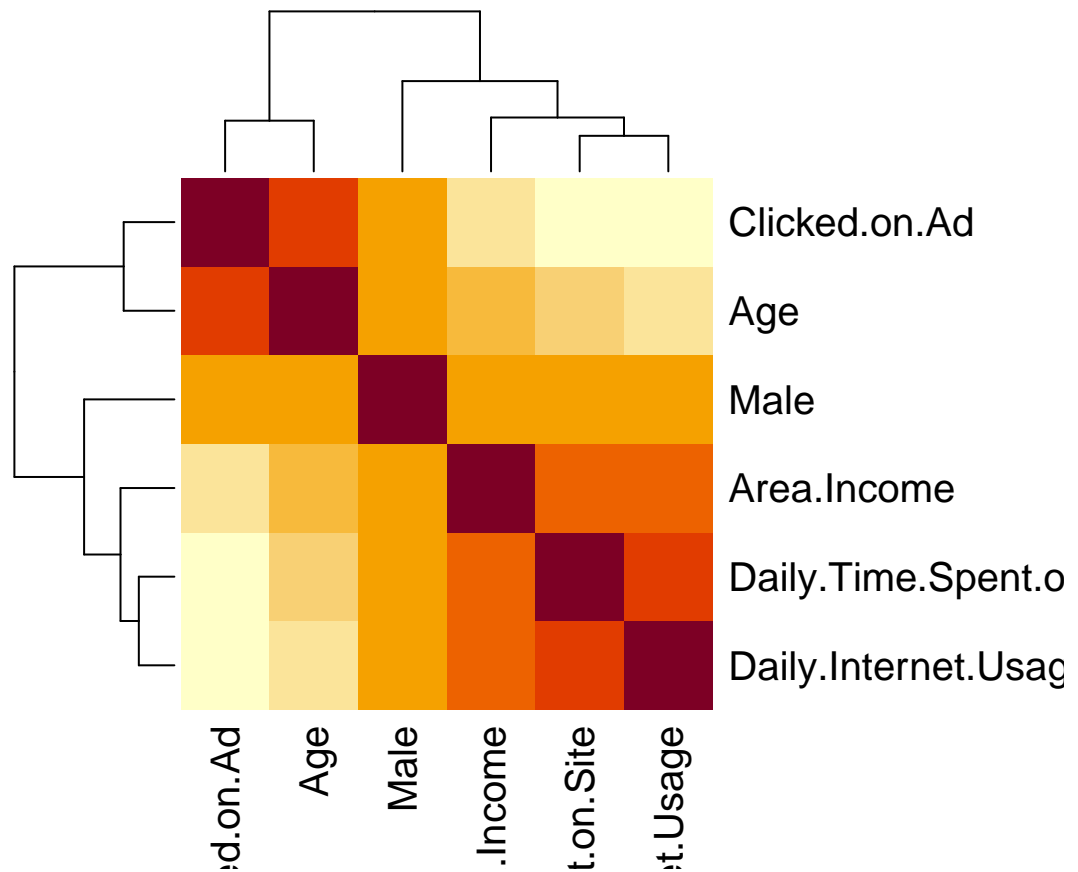
```
library(corrplot)
```

```
## corrplot 0.92 loaded
```

```
corrplot(ad_cor, type = "lower", order = "hclust",  
         tl.col = "black", tl.srt = 45)
```



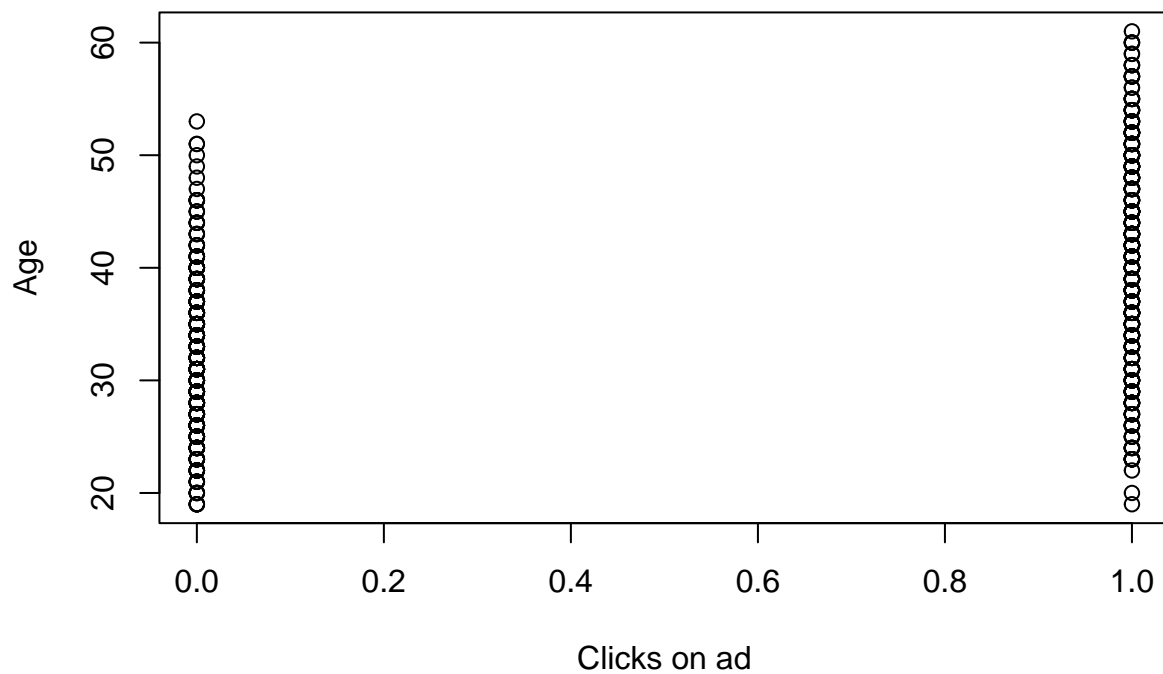
```
# Plotting a correlation Heatmap
# Get some colors
#col<- colorRampPalette(c("blue", "white", "red"))(20)
heatmap(x = ad_cor, symm = TRUE)
```



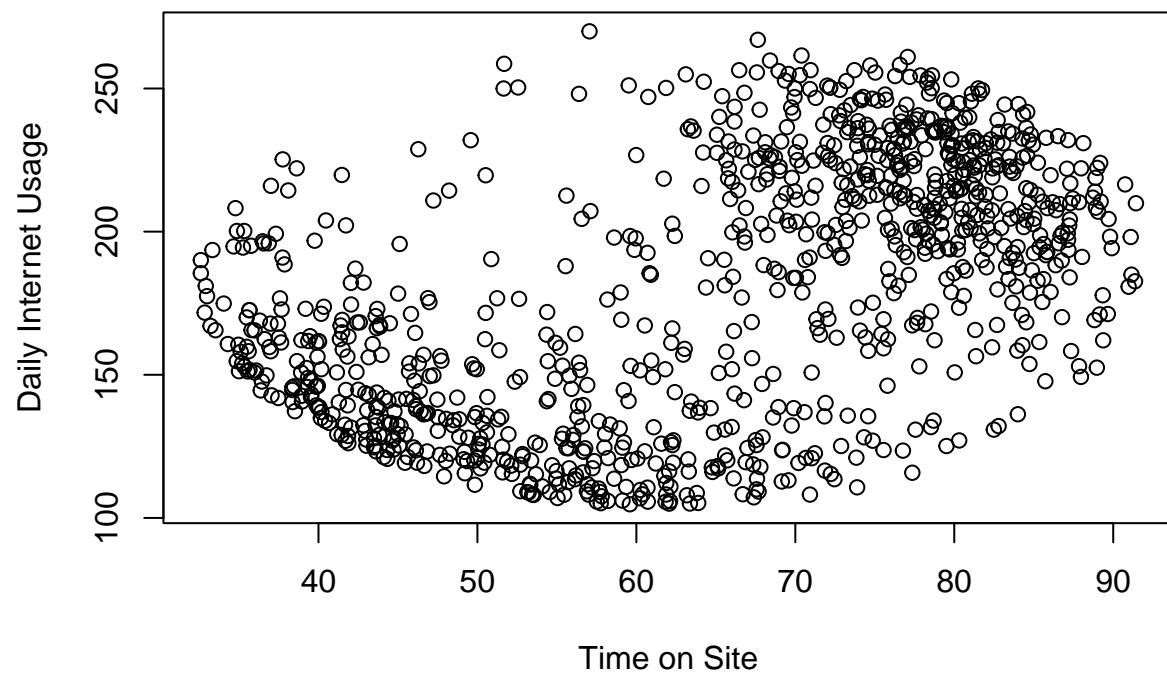
```
# Plotting scatter plots
# we will use the variables we assigned earlier
#time.site
#age
#area.income
#daily.internet
#male
#clicks.target
```

```
# plotting
plot(clicks.target, age, xlab="Clicks on ad", ylab="Age")
```

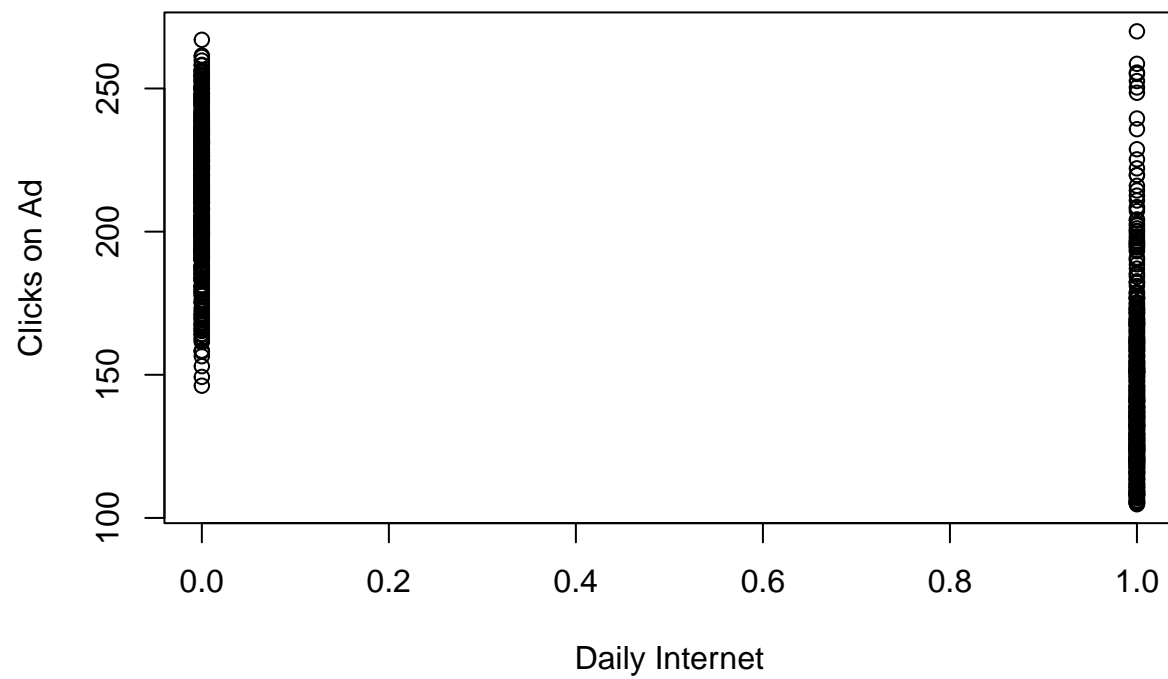




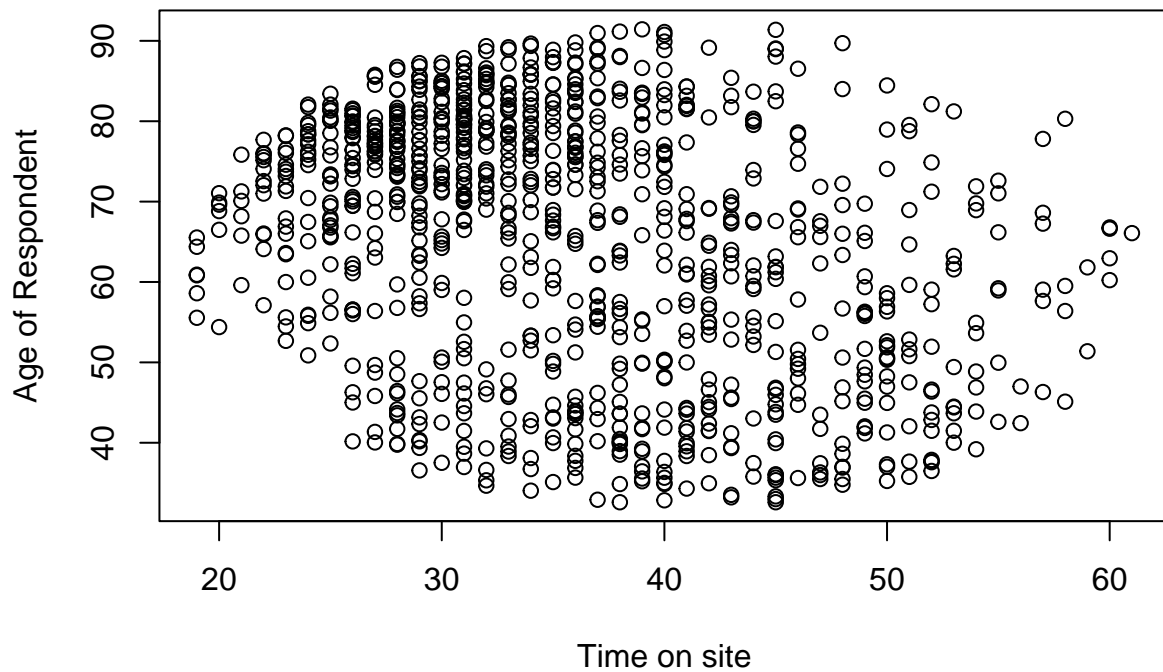
```
# plotting  
plot(time.site, daily.internet, xlab="Time on Site", ylab="Daily Internet Usage")
```



```
# plotting  
plot(clicks.target, daily.internet, xlab="Daily Internet", ylab="Clicks on Ad")
```

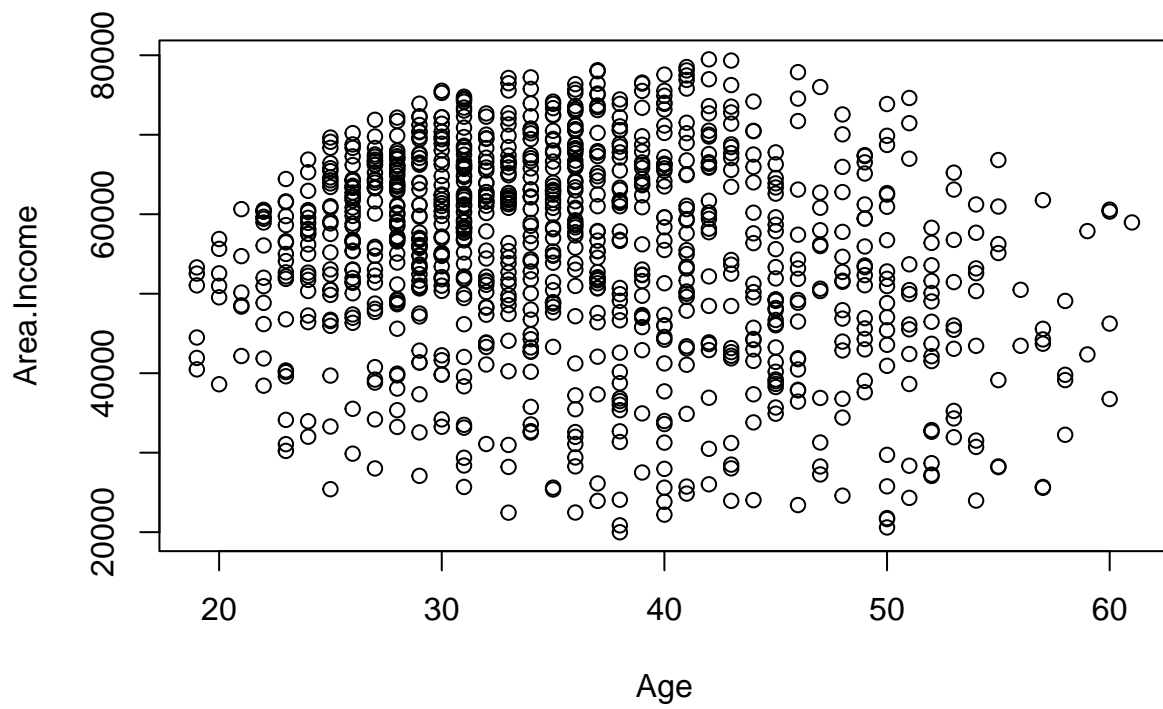


```
# plotting  
plot(age, time.site, xlab="Time on site", ylab="Age of Respondent")
```

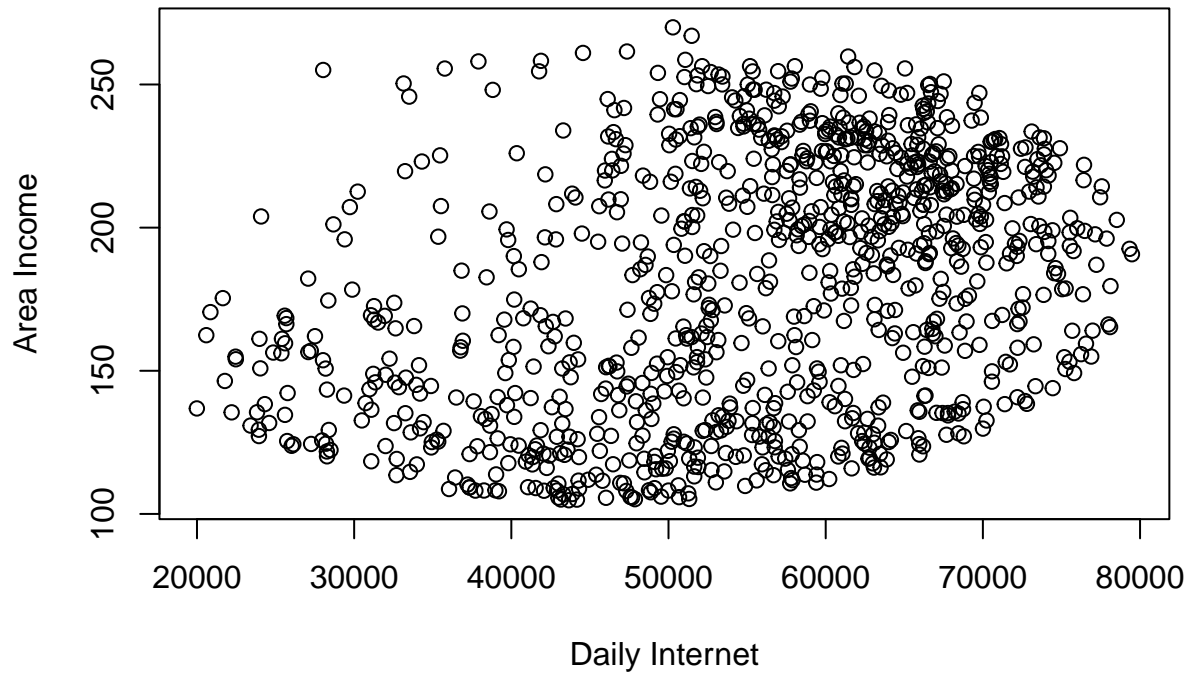


```
# there is no relationship between time spent on site and age
```

```
# plotting  
plot(age, area.income, xlab="Age", ylab="Area.Income")
```



```
# plotting  
plot(area.income, daily.internet, xlab="Daily Internet", ylab="Area Income")
```



The relationship between the ad being clicked on and other variables are as below Clicked.on.Ad  
Daily.Time.Spent.on.Site -0.75 Age 0.49 Area.Income -0.47 Daily.Internet.Usage -0.79 Male -0.04  
Clicked.on.Ad 1.00

## Conclusion and Recommendation

Gender has the least influence on whether the ad is being clicked on or not. Age has a moderately high positive influence on an ad being clicked on, with a mean of about 35 years old, the entrepreneur is advised to custom the advert to target this age group. This data is however skewed and hence could be causing this observation. Area Income has a moderately high negative influence on an ad being clicked on. However since this data is skewed to the right, this could have an influence on this analysis. Daily internet usage and Daily time spent on the site has high negative correlations, this means that when these measurements increase, the chances of an ad being clicked go down. A more balanced data-set could lead to better results.