

Week 12 Core IP

##1. Defining the Question ## a) Specifying the Data Analytic Question

Identify which individuals are most likely to click on ads from a cryptography course website

b) Defining the Metric for Success

For this study, we will perform conclusive Exploratory Data Analysis to enable us identify individuals who are most likely to click on ads. ## c) Understanding 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. Using the data previously collected, she is looking to do a study to identify which individuals are most likely to click on her ads. ## d) Data Relevance

Data is provided was collected in the past but from the same blog hence it is very suitable for this study.

Definition of Variables Daily Time Spent on Site

Age

Area

Income

Daily Internet Usage

Ad Topic Line

City

Male

Country

Timestamp

Clicked on Ad ### 1.4 Drafting the Experimental Design 1. Define the question, set the metric for success, outline the context, drafting the experimental design, and determining the appropriateness of the data. 2. Load the dataset and previewing it. 3. Check for missing and duplicated values and deal with them where necessary. 4. Check for outliers and other anomalies and deal with them where necessary. 5. Perform univariate and bivariate analysis. 6. Create a baseline model and assess its accuracy score. 7. Challenge the solution. 8. Conclude and provide insights on how this project can be improved.

2. Data Preparation and Cleaning

```
# importing and previewing the dataset
df<-read.csv('http://bit.ly/IPAdvertisingData')
head(df)
```

```
##   Daily.Time.Spent.on.Site Age Area.Income Daily.Internet.Usage
## 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
## 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
## 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
```

#Data Dimensions

```
paste("The dimensions of the data frame are ", paste (dim(df), collapse = ','))
```

```
## [1] "The dimensions of the data frame are 1000,10"
```

#Datatypes

```
sapply(df, class)
```

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

```
#We have a mix of datatypes from numeric, integer and character
```

#Summary

```
summary(df)
```

```
##   Daily.Time.Spent.on.Site      Age      Area.Income  Daily.Internet.Usage
## Min.   :32.60      Min.   :19.00  Min.   :13996  Min.   :104.8
## 1st Qu.:51.36      1st Qu.:29.00  1st Qu.:47032  1st Qu.:138.8
## Median :68.22      Median :35.00  Median :57012  Median :183.1
## Mean   :65.00      Mean   :36.01  Mean   :55000  Mean   :180.0
## 3rd Qu.:78.55      3rd Qu.:42.00  3rd Qu.:65471  3rd Qu.:218.8
## Max.   :91.43      Max.   :61.00  Max.   :79485  Max.   :270.0
## Ad.Topic.Line      City      Male      Country
```

```
## Length:1000      Length:1000      Min.   :0.000      Length:1000
## Class :character  Class :character  1st Qu.:0.000      Class :character
## Mode  :character  Mode  :character  Median :0.000      Mode  :character
##                                     Mean  :0.481
##                                     3rd Qu.:1.000
##                                     Max.   :1.000
## Timestamp        Clicked.on.Ad
## Length:1000      Min.   :0.0
## Class :character  1st Qu.:0.0
## Mode  :character  Median :0.5
##                                     Mean  :0.5
##                                     3rd Qu.:1.0
##                                     Max.   :1.0
```

```
#Checking for unique characters
sapply(df, function(x) length(unique(x)))
```

```
## Daily.Time.Spent.on.Site      Age      Area.Income
##                900                43                1000
##      Daily.Internet.Usage      Ad.Topic.Line      City
##                966                1000                969
##                Male      Country      Timestamp
##                2                237                1000
##      Clicked.on.Ad
##                2
```

Data Cleaning

```
# checking for duplicates
anyDuplicated(df)
```

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

There are no duplicated records so there is no need to remove any of them.

```
# looking for missing values
colSums(is.na(df))
```

```
## 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
```

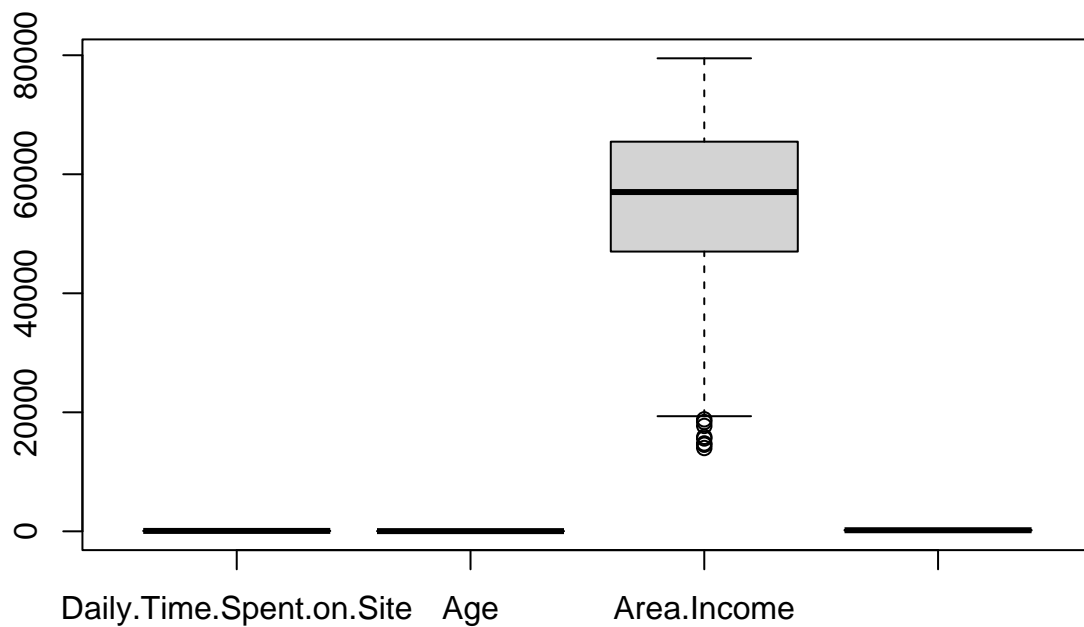
There are no missing values in each column so we don't need to carry out imputation or replacement.

#Checking for outliers *#First we select numeric columns excluding male and clicked.on.ad since they are binary column*

```
df1 <- subset(df, select = -c(Ad.Topic.Line, City, Male, Country, Timestamp, Clicked.on.Ad))
head(df1)
```

```
##   Daily.Time.Spent.on.Site Age Area.Income Daily.Internet.Usage
## 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
```

```
#Plotting boxplots to check for outliers
boxplot(df1
        )
```



```
boxplot.stats(df1$Area.Income)$out
```

```
## [1] 17709.98 18819.34 15598.29 15879.10 14548.06 13996.50 14775.50 18368.57
```

We won't remove the above figures because it concerns income and people earn different amounts of money.

```
#Change datatypes
df$Male <- as.factor(df$Male)
```

```
df$Clicked.on.Ad <- as.factor(df$Clicked.on.Ad)
#Checking datatypes
sapply(df, class)
```

```
## Daily.Time.Spent.on.Site      Age      Area.Income
##           "numeric"           "integer"      "numeric"
##   Daily.Internet.Usage      Ad.Topic.Line      City
##           "numeric"           "character"      "character"
##           Male      Country      Timestamp
##           "factor"           "character"      "character"
##   Clicked.on.Ad
##           "factor"
```

```
# split timestamp column into year, month, day, and hour
# NB: minute and second are irrelevant to our analysis
df$year <- format(as.POSIXct(df$Timestamp, format="%Y-%m-%d %H:%M:%S"), "%Y")
df$month <- format(as.POSIXct(df$Timestamp, format="%Y-%m-%d %H:%M:%S"), "%m")
df$day <- format(as.POSIXct(df$Timestamp, format="%Y-%m-%d %H:%M:%S"), "%d")
df$hour <- format(as.POSIXct(df$Timestamp, format="%Y-%m-%d %H:%M:%S"), "%H")
head(df)
```

```
##   Daily.Time.Spent.on.Site Age Area.Income Daily.Internet.Usage
## 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
## 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 year month day hour
## 1 2016-03-27 00:53:11      0 2016    03 27    00
## 2 2016-04-04 01:39:02      0 2016    04 04    01
## 3 2016-03-13 20:35:42      0 2016    03 13    20
## 4 2016-01-10 02:31:19      0 2016    01 10    02
## 5 2016-06-03 03:36:18      0 2016    06 03    03
## 6 2016-05-19 14:30:17      0 2016    05 19    14
```

```
#Dropping the column Timestamp and Ad.Topic.Line
df_clean = subset(df, select = -c(Timestamp,Ad.Topic.Line))
head(df_clean)
```

```
##   Daily.Time.Spent.on.Site Age Area.Income Daily.Internet.Usage      City
## 1           68.95 35      61833.90      256.09 Wrightburgh
## 2           80.23 31      68441.85      193.77 West Jodi
## 3           69.47 26      59785.94      236.50 Davidton
## 4           74.15 29      54806.18      245.89 West Terrifurt
```

```
## 5          68.37 35      73889.99          225.58 South Manuel
## 6          59.99 23      59761.56          226.74      Jamieberg
##   Male   Country Clicked.on.Ad year month day hour
## 1    0   Tunisia          0 2016    03 27  00
## 2    1     Nauru          0 2016    04 04  01
## 3    0 San Marino          0 2016    03 13  20
## 4    1     Italy          0 2016    01 10  02
## 5    0   Iceland          0 2016    06 03  03
## 6    1    Norway          0 2016    05 19  14
```

```
#Datatypes
sapply(df_clean, class)
```

```
## Daily.Time.Spent.on.Site          Age          Area.Income
##          "numeric"          "integer"          "numeric"
##   Daily.Internet.Usage          City          Male
##          "numeric"          "character"          "factor"
##          Country          Clicked.on.Ad          year
##          "character"          "factor"          "character"
##          month          day          hour
##          "character"          "character"          "character"
```

```
# set the new columns to be of data type Factor
df_clean$year <- as.factor(df_clean$year)
df_clean$month <- as.factor(df_clean$month)
df_clean$day <- as.factor(df_clean$day)
df_clean$hour <- as.factor(df_clean$hour)
```

```
#Datatypes
sapply(df_clean, class)
```

```
## Daily.Time.Spent.on.Site          Age          Area.Income
##          "numeric"          "integer"          "numeric"
##   Daily.Internet.Usage          City          Male
##          "numeric"          "character"          "factor"
##          Country          Clicked.on.Ad          year
##          "character"          "factor"          "factor"
##          month          day          hour
##          "factor"          "factor"          "factor"
```

Exploratory Data Analysis

Univariate Analysis

```
colnames(df_clean)
```

```
## [1] "Daily.Time.Spent.on.Site" "Age"
## [3] "Area.Income"            "Daily.Internet.Usage"
## [5] "City"                   "Male"
## [7] "Country"                "Clicked.on.Ad"
## [9] "year"                   "month"
## [11] "day"                    "hour"
```

```
#Selecting the numeric columns
num <- subset(df_clean, select = -c(City, Male, Country, Clicked.on.Ad, month, day, hour, year))
#Getting the measures of central tendency
summary(num)
```

```
##   Daily.Time.Spent.on.Site      Age      Area.Income   Daily.Internet.Usage
##   Min.      :32.60           Min.      :19.00   Min.      :13996   Min.      :104.8
##   1st Qu.:51.36           1st Qu.:29.00   1st Qu.:47032   1st Qu.:138.8
##   Median :68.22           Median :35.00   Median :57012   Median :183.1
##   Mean   :65.00           Mean   :36.01   Mean   :55000   Mean   :180.0
##   3rd Qu.:78.55           3rd Qu.:42.00   3rd Qu.:65471   3rd Qu.:218.8
##   Max.   :91.43           Max.   :61.00   Max.   :79485   Max.   :270.0
```

Variance and Standard deviation

```
var(df_clean$Age)
```

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

```
sd(df_clean$Age)
```

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

```
var(df_clean$Area.Income)
```

```
## [1] 179952406
```

```
sd(df_clean$Area.Income)
```

```
## [1] 13414.63
```

```
var(df_clean$Daily.Internet.Usage)
```

```
## [1] 1927.415
```

```
sd(df_clean$Daily.Internet.Usage)
```

```
## [1] 43.90234
```

```
var(df_clean$Daily.Time.Spent.on.Site)
```

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

```
sd(df_clean$Daily.Time.Spent.on.Site)
```

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

Conclusions

1. The minimum amount of time spent on the blog is 32.60 and maximum is 91.43 with a mean at 65 and median at 68
2. The mean age of people visiting the site is 36, max age is 61 and min age is 19 which makes sense since the range between 61 and 19 are the people most active online. 3. From data, the maximum income of individuals is 79485 and a min income of 13996 4. The mean daily internet usage on the website is 180 and a median level at 183.1

```
library(moments)
options(warn = -1)
```

```
#Checking for skewness
```

```
paste("Daily Time_Spent_Skewness: ", paste (skewness(df_clean$Daily.Time.Spent.on.Site), collapse = ','))
```

```
## [1] "Daily Time_Spent_Skewness: -0.371202614867441"
```

```
paste("Income_Skewness: ", paste (skewness(df_clean$Area.Income), collapse = ','))
```

```
## [1] "Income_Skewness: -0.649396701694076"
```

```
paste("Age_Skewness: ", paste (skewness(df_clean$Age), collapse = ','))
```

```
## [1] "Age_Skewness: 0.478422676206608"
```

```
paste("Daily_Internet_Usage_Skewness: ", paste (skewness(df_clean$Daily.Internet.Usage), collapse = ','))
```

```
## [1] "Daily_Internet_Usage_Skewness: -0.0334870316434409"
```

```
#Checking for kurtosis
```

```
paste("Daily Time_Spent_Kurtosis: ", paste (kurtosis(df_clean$Daily.Time.Spent.on.Site), collapse = ','))
```

```
## [1] "Daily Time_Spent_Kurtosis: 1.90394215401081"
```

```
paste("Income_Kurtosis: ", paste (kurtosis(df_clean$Area.Income), collapse = ','))
```

```
## [1] "Income_Kurtosis: 2.89469406161926"
```

```
paste("Age_Kurtosis: ", paste (kurtosis(df_clean$Age), collapse = ','))
```

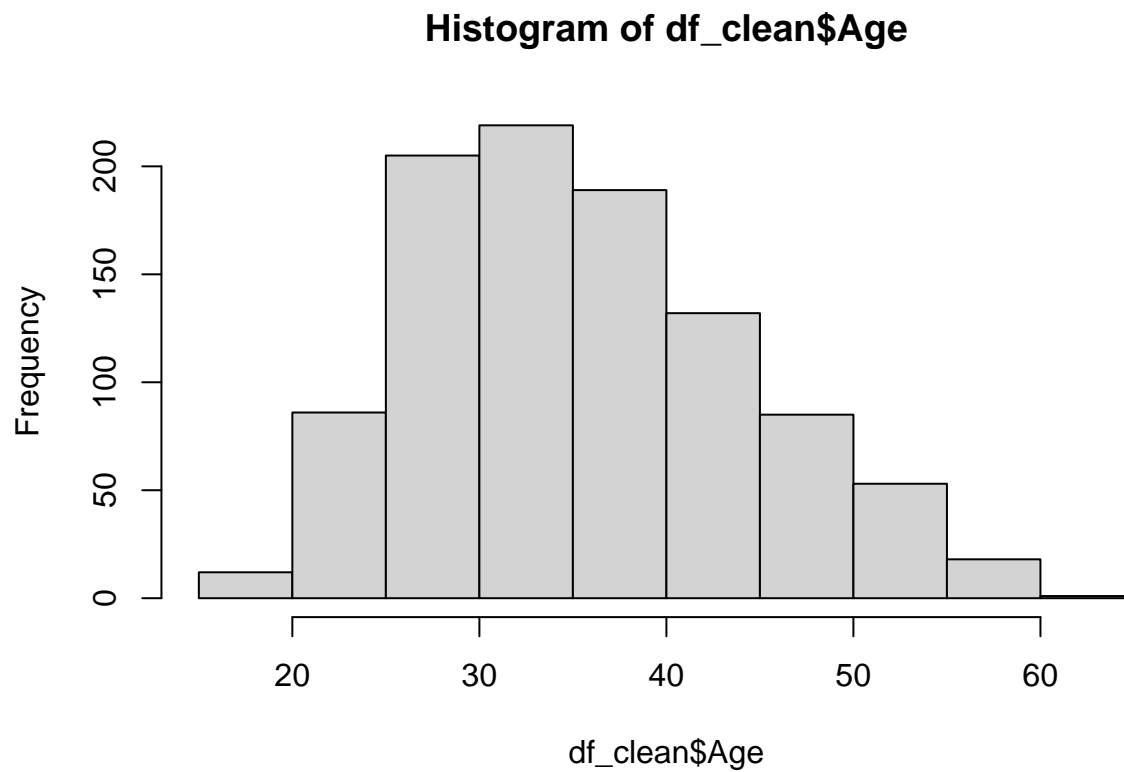
```
## [1] "Age_Kurtosis: 2.59548176807726"
```



```
paste("Daily_Internet_Usage_Kurtosis: ", paste (kurtosis(df_clean$Daily.Internet.Usage), collapse = ',')
```

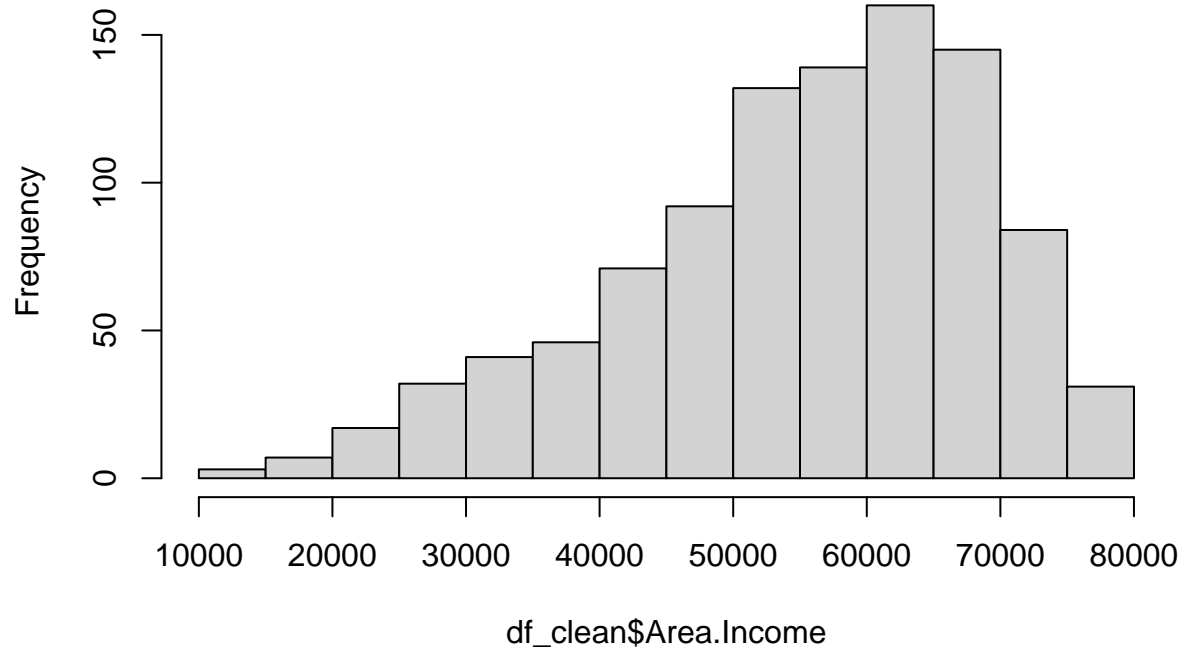
```
## [1] "Daily_Internet_Usage_Kurtosis: 1.72770118094819"
```

```
hist(df_clean$Age)
```



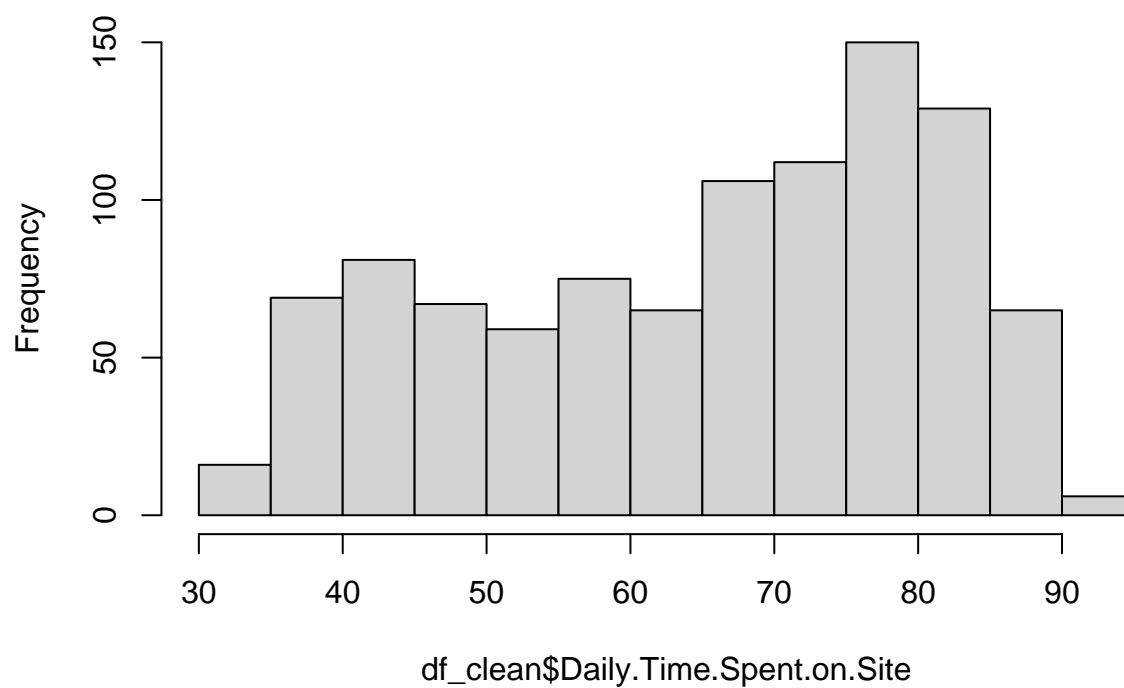
```
hist(df_clean$Area.Income)
```

Histogram of df_clean\$Area.Income



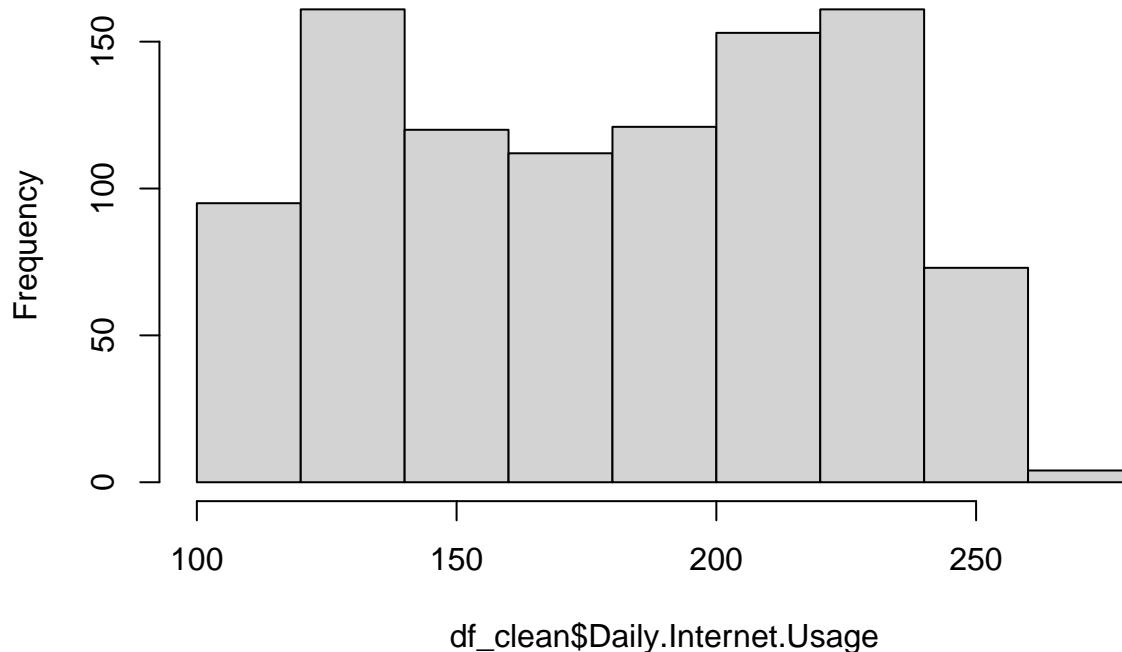
```
hist(df_clean$Daily.Time.Spent.on.Site)
```

Histogram of df_clean\$Daily.Time.Spent.on.Site



```
hist(df_clean$Daily.Internet.Usage)
```

Histogram of df_clean\$Daily.Internet.Usage



Observation -Age: Most people who visit the blog are between 25 and 40 years, data is skewed to the right of the mean. Graph doesn't show a sharp peak. The skewness value implies that the distribution is almost fairly symmetrical, so our initial assumption based on just looking at the visualization of the distribution is slightly wrong.

-Income: Data on income is mostly skewed to the right of the 55,00 mean. A kurtosis value of 2.89 indicates that the distribution is platykurtic although it is getting very close to being mesokurtic. The distribution is negatively skewed.

-Daily internet usage: The distribution is platykurtic. The distribution appears to be relatively uniform and bimodal.

-Time spent on site: There are lots of variations on how much time people spend on the site. A good number does spend between 65 and 85 time on the site.

```
library(plyr)
```

City

```
# displaying the first 6 frequently occurring cities
count_city <- count(df_clean$City)
count_city_head <- head(arrange(count_city, desc(freq)))
count_city_head
```

```
##           x freq
## 1    Lisamouth    3
```

```
## 2    Williamsport    3
## 3 Benjaminchester    2
## 4      East John     2
## 5    East Timothy    2
## 6      Johnstad      2
```

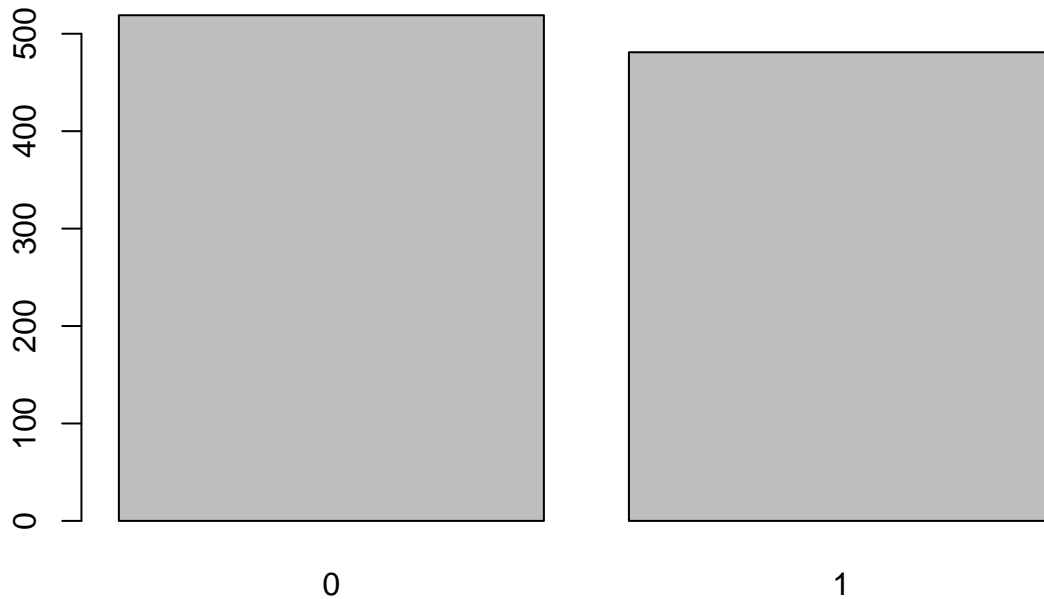
male

```
male_table <- table(df_clean$Male)
male_table
```

```
##
##  0  1
## 519 481
```

We see here that 519 are not male while 481 are. To easily visualize this:

```
barplot(male_table)
```



```
### country
```

```
# displaying the first 10 frequently occurring countries
count_country <- count(df_clean$Country)
count_country_head <- head(arrange(count_country, desc(freq)), 10)
count_country_head
```

```
##           x freq
## 1 Czech Republic 9
## 2      France    9
## 3  Afghanistan  8
## 4    Australia  8
## 5      Cyprus   8
## 6      Greece   8
## 7     Liberia   8
## 8   Micronesia  8
## 9        Peru   8
## 10    Senegal   8
```

month

```
# displaying the months in order of most frequently occurring to least frequently occurring
count_months <- count(df_clean$month)
arrange(count_months, desc(freq))
```

```
##      x freq
## 1 02 160
## 2 03 156
## 3 01 147
## 4 04 147
## 5 05 147
## 6 06 142
## 7 07 101
```

We see here that February is the most frequently occurring month with July being the least frequently occurring month.

day

```
# displaying top 5 frequently occurring days
count_days <- count(df_clean$day)
head(arrange(count_days, desc(freq)), 5)
```

```
##      x freq
## 1 03 46
## 2 17 42
## 3 15 41
## 4 10 37
## 5 04 36
```

The 3rd day is the most frequently occurring day overall. However, to get a more accurate picture of this, we will look at which day occurs most frequently in which month. We will do this in bivariate analysis.

```
tail(arrange(count_days, desc(freq)), 5)
```

```
##      x freq
## 27 02   25
## 28 06   25
## 29 22   24
## 30 25   23
## 31 31   18
```

The 31st day seems to be the least occurring day.

hour

```
# displaying the top 5 hours
count_hours <- count(df_clean$hour)
head(arrange(count_hours, desc(freq)), 5)
```

```
##      x freq
## 1 07   54
## 2 20   50
## 3 09   49
## 4 21   48
## 5 00   45
```

Most frequently occurring time appears to be around 7 AM.

```
tail(arrange(count_hours, desc(freq)), 5)
```

```
##      x freq
## 20 12   38
## 21 02   36
## 22 15   35
## 23 01   32
## 24 10   31
```

Least frequently occurring time appears to be around 10 AM. This is probably because more people get engrossed in the day's work.

clicked on ad

```
ad_table <- table(df_clean$Clicked.on.Ad)
print(ad_table)
```

```
##
##    0    1
## 500 500
```

Looks like the number of people who both clicked on the ad and didn't click on the ad is the same (500 each).

Bivariate Analysis

We will start by looking at the relationship between our target variable (clicked_on_ad) and the other variables.

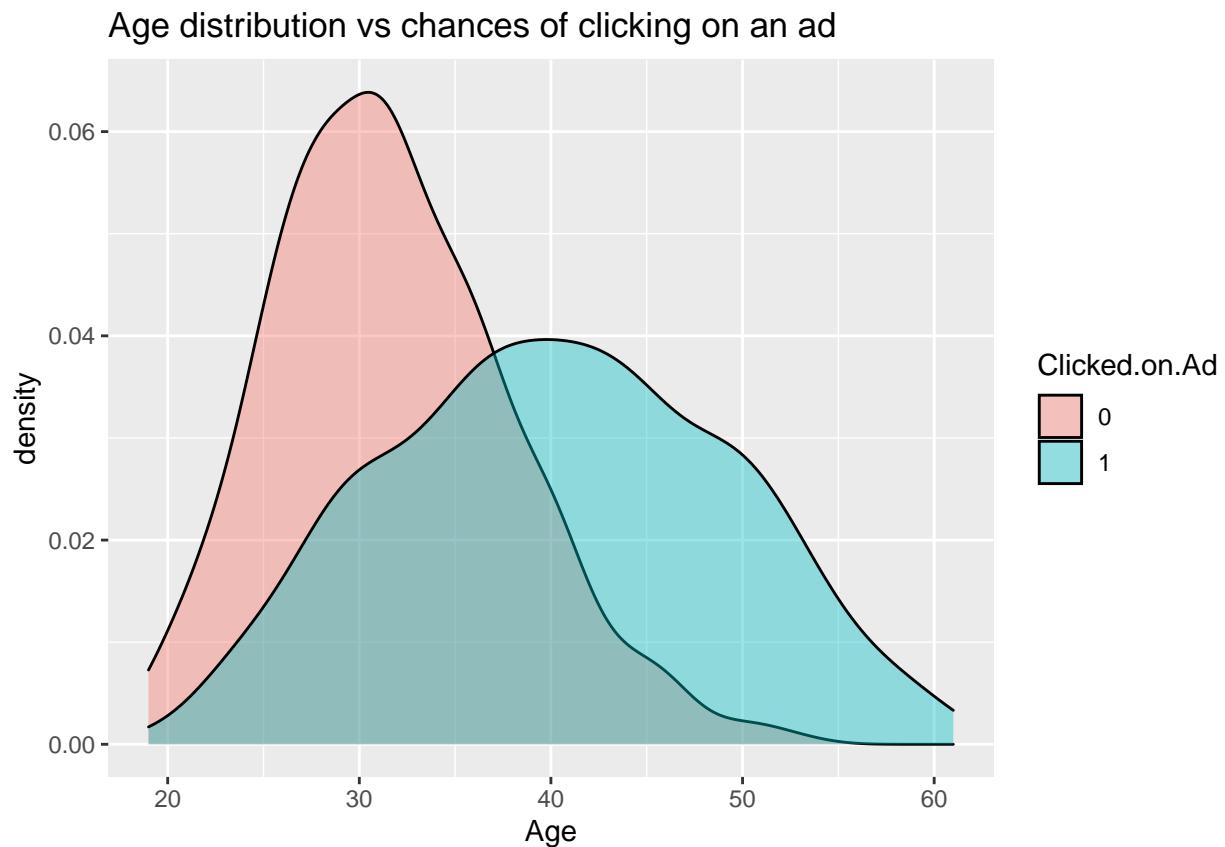
```
# how many males clicked on ads
ad_male.table <- table(df_clean$Clicked.on.Ad, df_clean$Male)
names(dimnames(ad_male.table)) <- c("Clicked on Ad?", "Male?")
ad_male.table
```

```
##           Male?
## Clicked on Ad?  0   1
##           0 250 250
##           1 269 231
```

From this we see that of those who clicked on the ad, 269 were female while 231 were male. There was no difference in gender of those who did not click on the ad.

```
library(ggplot2)
```

```
#Age and it's relationship to clicking an ad
ggplot(df_clean,
       aes(x = Age,
           fill = Clicked.on.Ad)) +
  geom_density(alpha = 0.4) +
  labs(title = "Age distribution vs chances of clicking on an ad")
```



People from all age groups click on ads on the site. People above 40 are more likely to click on an ad as per the graph above. while younger people dont click as often

```
# ad clicked per month
ad_month.table <- table(df_clean$month, df_clean$Clicked.on.Ad)
names(dimnames(ad_month.table)) <- c("Month", "Clicked on Ad?")
ad_month.table
```

```
##      Clicked on Ad?
## Month  0  1
##      01 78 69
##      02 77 83
##      03 82 74
##      04 73 74
##      05 68 79
##      06 71 71
##      07 51 50
```

Looking at this table, we see that February reports the highest number of ads clicked and July the least.

```
# ad clicked per day
ad_day.table <- table(df_clean$day, df_clean$Clicked.on.Ad)
names(dimnames(ad_day.table)) <- c("Day", "Clicked on Ad?")
ad_day.table
```

```
##      Clicked on Ad?
## Day   0  1
##      01 14 19
##      02 15 10
##      03 20 26
##      04 22 14
##      05 17 18
##      06 11 14
##      07 18 14
##      08 20 15
##      09 14 20
##      10 18 19
##      11 17 15
##      12  9 20
##      13 13 17
##      14 12 21
##      15 21 20
##      16 21 14
##      17 24 18
##      18 18 17
##      19 17 12
##      20 22 11
##      21 17 15
##      22 14 10
##      23 13 22
##      24 15 18
##      25  8 15
##      26 21 15
```

```
##    27 19 16
##    28 13 17
##    29 14 15
##    30 14 14
##    31  9  9
```

Day 03 has the highest number of ads clicked. Day 31 has the least.

```
# ad clicked per hour
ad_hour.table <- table(df_clean$hour, df_clean$Clicked.on.Ad)
names(dimnames(ad_hour.table)) <- c("Hour", "Clicked on Ad?")
ad_hour.table
```

```
##      Clicked on Ad?
## Hour  0  1
##    00 19 26
##    01 16 16
##    02 19 17
##    03 19 23
##    04 21 21
##    05 23 21
##    06 16 23
##    07 28 26
##    08 22 21
##    09 21 28
##    10 17 14
##    11 16 24
##    12 22 16
##    13 21 21
##    14 22 21
##    15 16 19
##    16 23 16
##    17 18 23
##    18 16 25
##    19 20 19
##    20 26 24
##    21 29 19
##    22 24 19
##    23 26 18
```

Hour 09 (9 AM) returned the highest number of ads clicked, 28, whereas Hour 10 (10 AM) returned the lowest, 14.

```
# ad clicked per city
ad_city.table <- table(df_clean$City, df_clean$Clicked.on.Ad)
names(dimnames(ad_city.table)) <- c("City", "Clicked on Ad?")
ad_city.table
```

```
##              Clicked on Ad?
## City              0  1
## Adamsbury         0  1
## Adamside           0  1
## Adamsstad         1  0
```

##	Alanview	1 0
##	Alexanderfurt	0 1
##	Alexanderview	0 1
##	Alexandrafort	1 0
##	Alexisland	1 0
##	Aliciatown	0 1
##	Alvaradoport	0 1
##	Alvarezland	0 1
##	Amandafort	0 1
##	Amandahaven	0 1
##	Amandaland	1 0
##	Amyfurt	1 0
##	Amyhaven	1 0
##	Andersonchester	0 1
##	Andersonfurt	0 1
##	Andersonton	1 0
##	Andrewborough	0 1
##	Andrewmouth	1 0
##	Angelhaven	1 0
##	Anthonyfurt	1 0
##	Ashleychester	1 0
##	Ashleymouth	1 0
##	Austinborough	1 0
##	Austinland	1 0
##	Bakerhaven	1 0
##	Barbershire	1 0
##	Beckton	1 0
##	Benjaminchester	2 0
##	Bernardton	0 1
##	Bethburgh	0 1
##	Birdshire	1 0
##	Blairborough	0 1
##	Blairville	1 0
##	Blevinstown	0 1
##	Bowenvue	1 0
##	Boyerberg	0 1
##	Bradleyborough	1 0
##	Bradleyburgh	0 1
##	Bradleyside	0 1
##	Bradshawborough	1 0
##	Bradyfurt	0 1
##	Brandiland	0 1
##	Brandonbury	0 1
##	Brandonstad	1 0
##	Brandymouth	0 1
##	Brendaburgh	1 0
##	Brendachester	0 1
##	Brianabury	1 0
##	Brianfurt	0 1
##	Brianland	0 1
##	Brittanyborough	0 1
##	Brownbury	1 0
##	Brownport	0 1
##	Brownton	0 1

##	Browntown	0 1
##	Brownview	1 0
##	Bruceburgh	1 0
##	Burgessside	0 1
##	Butlerfort	0 1
##	Calebberg	1 0
##	Cameronberg	0 1
##	Campbellstad	1 0
##	Cannonbury	1 0
##	Carsonshire	1 0
##	Carterburgh	1 0
##	Carterland	0 1
##	Carterport	1 0
##	Carterton	1 0
##	Cassandratown	1 0
##	Catherinefort	0 1
##	Cervantesshire	0 1
##	Chapmanland	1 0
##	Chapmanmouth	0 1
##	Charlenetown	0 1
##	Charlesbury	1 0
##	Charlesport	0 1
##	Charlottefort	0 1
##	Chaseshire	0 1
##	Chrismouth	0 1
##	Christinehaven	0 1
##	Christinetown	0 1
##	Christopherchester	1 0
##	Christopherport	0 1
##	Christopherville	1 0
##	Clarkborough	0 1
##	Claytonside	1 0
##	Clineshire	1 0
##	Codyburgh	0 1
##	Coffeytown	1 0
##	Colebury	0 1
##	Colemanshire	1 0
##	Collinsburgh	1 0
##	Combsstad	0 1
##	Contrerasshire	1 0
##	Costaburgh	0 1
##	Courtneyfort	0 1
##	Coxhaven	1 0
##	Cranemouth	1 0
##	Crawfordfurt	0 1
##	Cunninghamhaven	0 1
##	Curtisport	0 1
##	Curtisview	1 0
##	Cynthiaside	1 0
##	Daisymouth	1 0
##	Danielview	0 1
##	Davidmouth	0 1
##	Davidside	0 1
##	Davidstad	0 1

##	Davidton	1 0
##	Davidview	0 1
##	Daviesborough	1 0
##	Davieshaven	1 0
##	Davilachester	0 1
##	Davisfurt	0 1
##	Dayton	1 0
##	Deannaville	1 0
##	Debraburgh	0 1
##	Derrickhaven	1 0
##	Destinyfurt	0 1
##	Dianashire	1 0
##	Dianaville	0 1
##	Donaldshire	1 0
##	Douglasview	1 0
##	Duffystad	0 1
##	Dustinborough	1 0
##	Dustinchester	1 0
##	Dustinmouth	0 1
##	East Aaron	1 0
##	East Anthony	0 1
##	East Barbara	0 1
##	East Benjaminville	1 0
##	East Breannafurt	0 1
##	East Brettton	0 1
##	East Brianberg	1 0
##	East Brittanyville	0 1
##	East Carlos	1 0
##	East Christopher	1 0
##	East Christopherbury	1 0
##	East Connie	1 0
##	East Dana	0 1
##	East Deborahhaven	1 0
##	East Debraborough	1 0
##	East Donna	0 1
##	East Donnatown	1 0
##	East Eric	0 1
##	East Ericport	0 1
##	East Georgeside	0 1
##	East Graceland	1 0
##	East Heatherside	0 1
##	East Heidi	0 1
##	East Henry	1 0
##	East Jason	0 1
##	East Jennifer	1 0
##	East Jessefort	0 1
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##	East Johnport	1 0
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##	East Lindsey	0 1
##	East Maureen	0 1
##	East Michaelland	1 0
##	East Michaelmouth	0 1
##	East Michaeltown	1 0

##	East Michele	0 1
##	East Michelleberg	0 1
##	East Mike	0 1
##	East Paul	1 0
##	East Rachaelfurt	0 1
##	East Rachelview	0 1
##	East Ronald	0 1
##	East Samanthashire	0 1
##	East Sharon	0 1
##	East Shawn	0 1
##	East Shawnchester	1 0
##	East Sheriville	1 0
##	East Stephen	0 1
##	East Susanland	1 0
##	East Tammie	0 1
##	East Theresashire	1 0
##	East Tiffanyport	1 0
##	East Timothy	2 0
##	East Timothyport	1 0
##	East Toddfort	1 0
##	East Troyhaven	1 0
##	East Tylershire	0 1
##	East Valerie	1 0
##	East Vincentstad	0 1
##	East Yvonnechester	0 1
##	Edwardmouth	1 0
##	Edwardsmouth	1 0
##	Edwardsport	0 1
##	Elizabethbury	0 1
##	Elizabethmouth	1 0
##	Elizabethport	0 1
##	Elizabethstad	0 1
##	Emilyfurt	1 0
##	Ericksonmouth	0 1
##	Erikville	1 0
##	Erinmouth	1 0
##	Erinton	0 1
##	Estesfurt	0 1
##	Estradafurt	1 0
##	Estradashire	0 1
##	Evansfurt	1 0
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##	Fraziershire	0 1
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##	Garciaview	0 1
##	Garnerberg	1 0

##	Garrettborough	1 0
##	Garychester	1 0
##	Gilbertville	1 0
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##	Gonzalezburgh	1 0
##	Grahamberg	0 1
##	Gravesport	1 0
##	Greenechester	1 0
##	Greentown	1 0
##	Greerport	0 1
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##	Greghaven	1 0
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##	Hamiltonfort	0 1
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##	Hannaport	0 1
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##	Harperborough	0 1
##	Harrishaven	1 0
##	Harrisonmouth	1 0
##	Hartmanchester	0 1
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##	Harveyport	0 1
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##	Lake Angela	1 0
##	Lake Annashire	1 0
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##	Lake Craigview	0 1
##	Lake Cynthia	1 0
##	Lake Danielle	1 0
##	Lake David	0 2
##	Lake Deannaborough	1 0
##	Lake Deborahburgh	1 0
##	Lake Dustin	0 1
##	Lake Edward	0 1
##	Lake Elizabethside	1 0
##	Lake Evantown	0 1
##	Lake Faith	0 1
##	Lake Gerald	0 1
##	Lake Hailey	1 0
##	Lake Ian	0 1
##	Lake Jacob	1 0
##	Lake Jacqueline	1 0
##	Lake James	0 2
##	Lake Jasonchester	1 0

##	Lake Jennifer	0 1
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##	Lake Jessica	0 1
##	Lake Jessicaville	0 1
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##	Lake Jillville	1 0
##	Lake John	0 1
##	Lake Johnbury	0 1
##	Lake Jonathanview	1 0
##	Lake Jose	1 1
##	Lake Joseph	1 0
##	Lake Josetown	1 0
##	Lake Joshuafurt	0 1
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##	Lake Kurtmouth	1 0
##	Lake Lisa	1 0
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##	Lake Matthewland	1 0
##	Lake Melindamouth	1 0
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##	Lake Michaelport	1 0
##	Lake Michelle	0 1
##	Lake Michellebury	0 1
##	Lake Nicole	1 0
##	Lake Patrick	2 0
##	Lake Rhondaburgh	0 1
##	Lake Stephenborough	0 1
##	Lake Susan	1 1
##	Lake Timothy	1 0
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##	Lake Vanessa	0 1
##	Lake Zacharyfurt	1 0
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##	Laurieside	1 0
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##	Leahside	0 1
##	Leonchester	1 0
##	Lesliebury	0 1
##	Lesliefort	1 0
##	Lewismouth	0 1
##	Lindaside	1 0
##	Lindsaymouth	1 0
##	Lisaberg	1 0
##	Lisafort	1 0
##	Lisamouth	1 2
##	Lopezberg	0 1
##	Lopezmouth	1 0
##	Loriville	0 1
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##	Luischester	1 0
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##	Mclaughlinbury	1 0
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##	Melissafurt	1 0
##	Melissastad	1 0
##	Meyerchester	1 0
##	Meyersstad	0 1
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##	Michaeland	1 0
##	Michaelmouth	1 0
##	Michaelshire	0 1
##	Micheletown	0 1
##	Michellefort	0 1
##	Michelleside	0 2
##	Millerbury	0 2
##	Millerchester	0 1
##	Millerfort	1 0
##	Millerland	1 0
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##	Millertown	1 1
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##	Munozberg	1 0
##	Murphymouth	1 0
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##	New Brandy	1 0
##	New Brendafurt	0 1
##	New Charleschester	0 1
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##	New Cynthia	1 0
##	New Daniellefort	0 1
##	New Darlene	0 1

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##	New Debbiestad	0 1
##	New Denisebury	0 1
##	New Frankshire	1 0
##	New Gabriel	1 0
##	New Henry	0 1
##	New Hollyberg	0 1
##	New James	0 1
##	New Jamestown	1 0
##	New Jasmine	1 0
##	New Jay	0 1
##	New Jeffreychester	1 0
##	New Jessicaport	2 0
##	New Johnberg	1 0
##	New Joshuaport	0 1
##	New Juan	1 0
##	New Julianberg	0 1
##	New Julie	1 0
##	New Karenberg	0 1
##	New Kayla	1 0
##	New Keithburgh	0 1
##	New Lindaberg	0 1
##	New Lucasburgh	0 1
##	New Marcusbury	0 1
##	New Maria	1 0
##	New Matthew	0 1
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##	New Michaeltown	1 0
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##	New Nathan	1 0
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##	New Paul	1 0
##	New Rachel	0 1
##	New Rebecca	0 1
##	New Sabrina	0 1
##	New Sean	1 0
##	New Shane	1 0
##	New Sharon	1 0
##	New Sheila	2 0
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##	New Tammy	0 1
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##	New Theresa	0 1
##	New Thomas	0 1
##	New Timothy	0 1
##	New Tina	0 1
##	New Tinamouth	1 0
##	New Traceystad	1 0
##	New Travis	1 0
##	New Travistown	0 1
##	New Tyler	1 0

##	New Wanda	1 0
##	New Williammouth	0 1
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##	Newmanberg	1 0
##	Nicholasland	0 1
##	Nicholasport	1 0
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##	North Aaronchester	0 1
##	North Alexandra	1 0
##	North Anaport	1 0
##	North Andrew	0 1
##	North Andrewstad	0 1
##	North Angelastad	0 1
##	North Angelatown	0 1
##	North Anna	1 0
##	North April	0 1
##	North Brandon	1 0
##	North Brittanyburgh	0 1
##	North Cassie	0 1
##	North Charlesbury	0 1
##	North Christopher	1 0
##	North Daniel	1 1
##	North Debra	1 0
##	North Debrashire	0 1
##	North Derekville	0 1
##	North Destiny	0 1
##	North Elizabeth	1 0
##	North Frankstad	1 0
##	North Garyhaven	1 0
##	North Isabellaville	1 0
##	North Jenniferburgh	0 1
##	North Jeremyport	1 0
##	North Jessicaville	0 1
##	North Johnside	1 0
##	North Johntown	0 1
##	North Jonathan	0 1
##	North Joshua	1 0
##	North Katie	0 1
##	North Kennethside	1 0
##	North Kevinside	0 1
##	North Kimberly	0 1
##	North Kristine	1 0
##	North Lauraland	0 1
##	North Laurenview	1 0
##	North Leonmouth	1 0
##	North Lisacheater	1 0
##	North Loriburgh	1 0
##	North Mark	0 1
##	North Maryland	0 1
##	North Mercedes	0 1
##	North Michael	0 1
##	North Monicaville	1 0
##	North Randy	1 0
##	North Raymond	1 0

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##	North Ricardotown	0 1
##	North Richardburgh	0 1
##	North Ronaldshire	1 0
##	North Russellborough	0 1
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##	North Sarashire	0 1
##	North Shannon	1 0
##	North Stephanieberg	1 0
##	North Tara	1 0
##	North Tiffany	1 0
##	North Tracyport	1 0
##	North Tylerland	1 0
##	North Virginia	0 1
##	North Wesleychester	1 0
##	Novaktown	1 0
##	Odomville	1 0
##	Olsonside	0 1
##	Olsonstad	0 1
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##	Pamelamouth	2 0
##	Parkerhaven	1 0
##	Patriciahaven	1 0
##	Patrickmouth	1 0
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##	Paulport	1 0
##	Paulshire	1 0
##	Pearsonfort	1 0
##	Penatown	0 1
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##	Perryburgh	0 1
##	Petersonfurt	0 1
##	Phelpschester	1 0
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##	Phillipsbury	0 1
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##	Port Aprilville	0 1
##	Port Beth	0 1
##	Port Blake	0 1
##	Port Brenda	0 1
##	Port Brian	0 1
##	Port Brianfort	1 0
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##	Port Calvintown	1 0
##	Port Cassie	0 1
##	Port Chasemouth	1 0
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##	Port Christinemouth	1 0
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##	Port Derekberg	0 1
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##	Port Eric	0 1
##	Port Erikhaven	0 1
##	Port Erinberg	0 1
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##	Port Georgebury	0 1
##	Port Gregory	1 0
##	Port Jacqueline	1 0
##	Port Jacquelinestad	1 0
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##	Port Jessica	0 1
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##	Port Kathleenfort	0 1
##	Port Kevinborough	1 0
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##	Port Mathew	1 0
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##	Port Michealburgh	0 1
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##	Port Patrickton	0 1
##	Port Paultown	0 1
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##	Port Raymondfort	1 0
##	Port Robin	1 0
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##	Port Sarahshire	0 1
##	Port Sherrystad	0 1
##	Port Stacey	1 0
##	Port Stacy	1 0
##	Port Susan	1 0
##	Port Whitneyhaven	1 0
##	Portermouth	1 0

##	Pottermouth	0 1
##	Princebury	1 0
##	Pruittmouth	1 0
##	Rachelhaven	1 0
##	Ramirezhaven	0 1
##	Ramirezland	1 0
##	Ramirezside	0 1
##	Ramirezton	1 0
##	Ramosstad	1 0
##	Randolphport	1 0
##	Randyshire	1 0
##	Rebeccamouth	0 1
##	Reginamouth	0 1
##	Reneechester	0 1
##	Reyesfurt	1 0
##	Reyesland	1 0
##	Rhondaborough	1 0
##	Richardshire	0 1
##	Richardsland	1 0
##	Richardsonland	0 1
##	Richardsonmouth	1 0
##	Richardsonshire	0 1
##	Richardsontown	1 0
##	Rickymouth	1 0
##	Riggsstad	1 0
##	Rivasland	0 1
##	Robertbury	1 0
##	Robertfurt	0 2
##	Robertmouth	1 0
##	Robertside	0 1
##	Robertsonburgh	0 1
##	Robertstown	0 1
##	Roberttown	0 1
##	Robinsonland	1 0
##	Robinsontown	0 1
##	Rochabury	0 1
##	Rogerburgh	0 1
##	Rogerland	1 0
##	Ronaldport	0 1
##	Ronniemouth	0 1
##	Russellville	0 1
##	Ryanhaven	0 1
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##	Salazarbury	0 1
##	Samanthaland	0 1
##	Samuelborough	1 0
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##	Sanchezmouth	1 0
##	Sandersland	1 0
##	Sanderstown	0 1
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##	Sandrashire	0 1
##	Sandraville	1 0
##	Sarafurt	1 0

##	Sarahland	0 1
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##	Sellerstown	1 0
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##	Sharpberg	1 0
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##	Shawstad	1 0
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##	Sherrishire	1 0
##	Shirleyfort	1 0
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##	Smithside	0 1
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##	South Brian	1 0
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##	South Cynthiashire	0 1
##	South Daniel	0 1
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##	South Denisefurt	1 0
##	South Dianeshire	1 0
##	South George	0 1
##	South Henry	0 1
##	South Jackieberg	0 1
##	South Jade	0 1
##	South Jaimeview	1 0
##	South Jasminebury	0 1
##	South Jeanneport	0 1
##	South Jennifer	1 0
##	South Jessica	0 1
##	South John	0 1
##	South Johnnymouth	0 1
##	South Kyle	0 1
##	South Lauraton	0 1
##	South Lauratown	0 1
##	South Lisa	0 2
##	South Manuel	1 0
##	South Margaret	0 1
##	South Mark	0 1
##	South Meghan	0 1
##	South Meredithmouth	1 0
##	South Pamela	1 0
##	South Patrickfort	1 0
##	South Peter	0 1

##	South Rebecca	0 1
##	South Renee	1 0
##	South Robert	1 0
##	South Ronald	1 0
##	South Stephanieport	1 0
##	South Tiffanyton	0 1
##	South Tomside	1 0
##	South Troy	1 0
##	South Vincentchester	0 1
##	South Walter	0 1
##	Staceyfort	0 1
##	Stephenborough	1 0
##	Stewartbury	1 0
##	Suzannetown	0 1
##	Sylviaview	1 0
##	Tammymouth	0 1
##	Tammyshire	0 1
##	Taylorberg	1 0
##	Taylorhaven	0 1
##	Taylormouth	0 1
##	Taylorport	1 0
##	Teresahaven	1 0
##	Thomasstad	1 0
##	Thomasview	1 0
##	Timothyfurt	0 1
##	Timothymouth	0 1
##	Timothyport	0 1
##	Timothytown	1 0
##	Tinacheater	1 0
##	Tinaton	0 1
##	Townsendfurt	1 0
##	Tracyhaven	0 1
##	Tranland	1 0
##	Troyville	1 0
##	Turnerchester	0 1
##	Turnerview	1 0
##	Turnerville	1 0
##	Tylerport	0 1
##	Valerieland	1 0
##	Vanessastad	0 1
##	Vanessaview	0 1
##	Villanuevastad	1 0
##	Villanuevaton	1 0
##	Wademouth	1 0
##	Wadestad	1 0
##	Wagnerchester	1 0
##	Wallacechester	1 0
##	Walshhaven	1 0
##	Waltertown	0 1
##	Watsonfort	1 0
##	Welchshire	0 1
##	Wendyton	1 0
##	Wendyville	0 1
##	West Alice	1 0

##	West Alyssa	1 0
##	West Amanda	0 2
##	West Andrew	1 0
##	West Angela	1 0
##	West Angelabury	1 0
##	West Annefort	0 1
##	West Aprilport	0 1
##	West Arielstad	1 0
##	West Barbara	1 0
##	West Benjamin	1 0
##	West Brad	0 1
##	West Brandonton	0 1
##	West Brenda	1 0
##	West Carmenfurt	1 0
##	West Casey	0 1
##	West Chloeborough	0 1
##	West Christopher	0 1
##	West Colin	1 0
##	West Connor	0 1
##	West Courtney	1 0
##	West Daleborough	1 0
##	West Dannyberg	1 0
##	West David	0 1
##	West Dennis	1 0
##	West Derekmouth	0 1
##	West Dylanberg	0 1
##	West Eduardotown	0 1
##	West Ericaport	0 1
##	West Ericfurt	0 1
##	West Gabriellamouth	0 1
##	West Gregburgh	1 0
##	West Guybury	1 0
##	West James	0 1
##	West Jane	0 1
##	West Jeremyside	0 1
##	West Jessicahaven	0 1
##	West Jodi	1 0
##	West Joseph	1 0
##	West Julia	0 1
##	West Justin	0 1
##	West Katiefurt	0 1
##	West Kevinfurt	0 1
##	West Lacey	1 0
##	West Leahton	0 1
##	West Lindseybury	0 1
##	West Lisa	1 0
##	West Lucas	1 0
##	West Mariafort	1 0
##	West Melaniefurt	0 1
##	West Melissashire	0 1
##	West Michaelhaven	1 0
##	West Michaelport	1 0
##	West Michaelshire	1 0
##	West Michaelstad	1 0

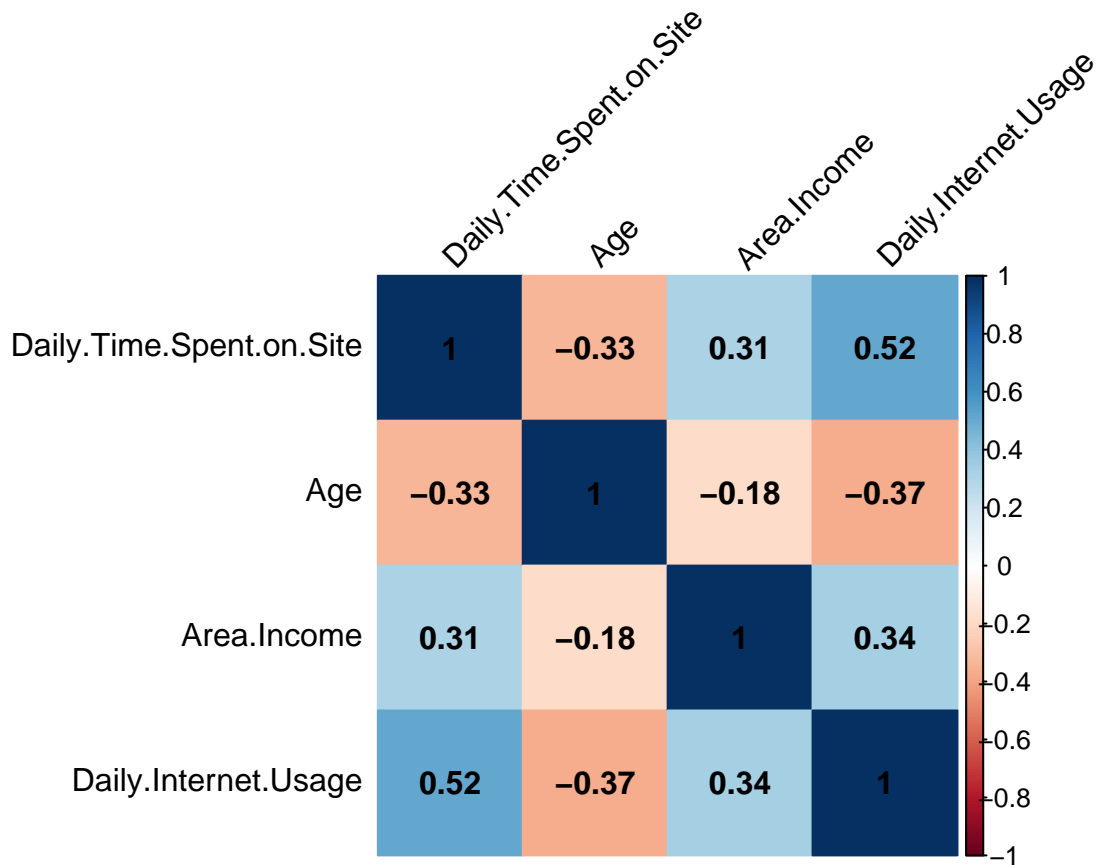
##	West Pamela	0 1
##	West Randy	0 1
##	West Raymondmouth	0 1
##	West Rhondamouth	1 0
##	West Ricardo	0 1
##	West Richard	0 1
##	West Robertside	1 0
##	West Roytown	1 0
##	West Russell	1 0
##	West Ryan	0 1
##	West Samantha	1 0
##	West Shannon	0 2
##	West Sharon	1 0
##	West Shaun	1 0
##	West Steven	2 0
##	West Sydney	1 0
##	West Tanner	1 0
##	West Tanya	0 1
##	West Terrifurt	1 0
##	West Thomas	1 0
##	West Tinashire	0 1
##	West Travismouth	0 1
##	West Wendyland	1 0
##	West William	0 1
##	West Zacharyborough	1 0
##	Westshire	0 1
##	Whiteport	0 1
##	Whitneyfort	1 0
##	Wilcoxport	0 1
##	Williammouth	0 1
##	Williamport	1 0
##	Williamsborough	0 1
##	Williamsfort	0 1
##	Williamsmouth	0 1
##	WilliamSPORT	1 2
##	WilliamSSide	1 0
##	Williamstad	0 1
##	Wilsonburgh	1 0
##	Wintersfort	1 0
##	Wongland	1 0
##	Wrightburgh	2 0
##	Wrightview	0 1
##	Yangside	0 1
##	Youngburgh	1 0
##	Youngfort	0 1
##	Yuton	0 1
##	Zacharystad	1 0
##	Zacharyton	0 1

###Improving the solution: creating a function that returns the highest and lowest values of a specific column so that you do not have to manually go through each individual record.

```
library(corrplot)
```

```
## corrplot 0.90 loaded
```

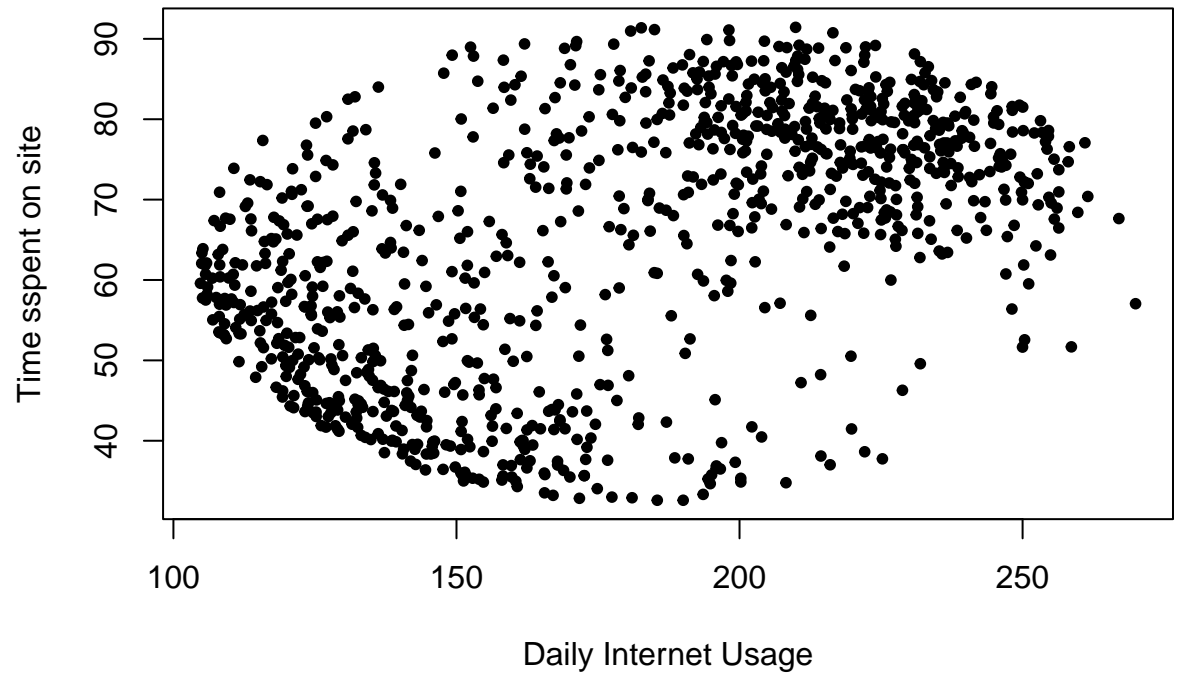
```
#Get the correlation matrix
res = cor(num)
#Plotting a correlation plot
corrplot(res, method="color", addCoef.col = "black",
          tl.col="black", tl.srt=45)
```



There is a fare correlation between amount spent on site and the Daily internet usage.

```
x <-df_clean$Daily.Internet.Usage
y <- df_clean$Daily.Time.Spent.on.Site
# Plot with main and axis titles
# Change point shape (pch = 19) and remove frame.
plot(x, y, main = "Time spent on site vs Daily Internet Usage",
      xlab = "Daily Internet Usage", ylab = "Time sspent on site",
      pch = 20)
```

Time spent on site vs Daily Internet Usage

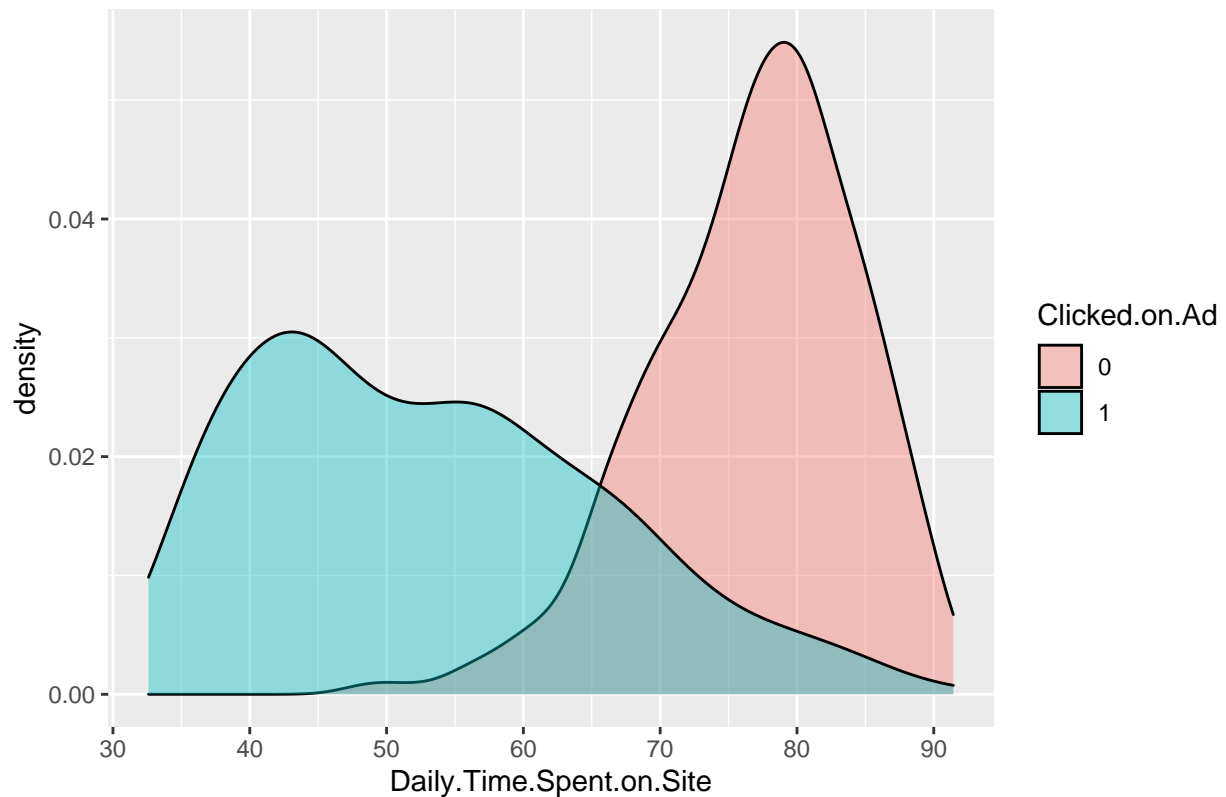


Scatter Plot

The points are all over but our data points are not highly correlated which explains this. But we can see that people who spend less time on site use less internet. Also, most of the people who use a lot of internet per day seem to spend a considerable amount of time on the site.

```
#Time Spent on internet and it's relationship to clicking an ad
ggplot(df_clean,
  aes(x = Daily.Time.Spent.on.Site,
      fill = Clicked.on.Ad)) +
  geom_density(alpha = 0.4) +
  labs(title = "Relationship between time spent on site and chances of clicking on an ad")
```

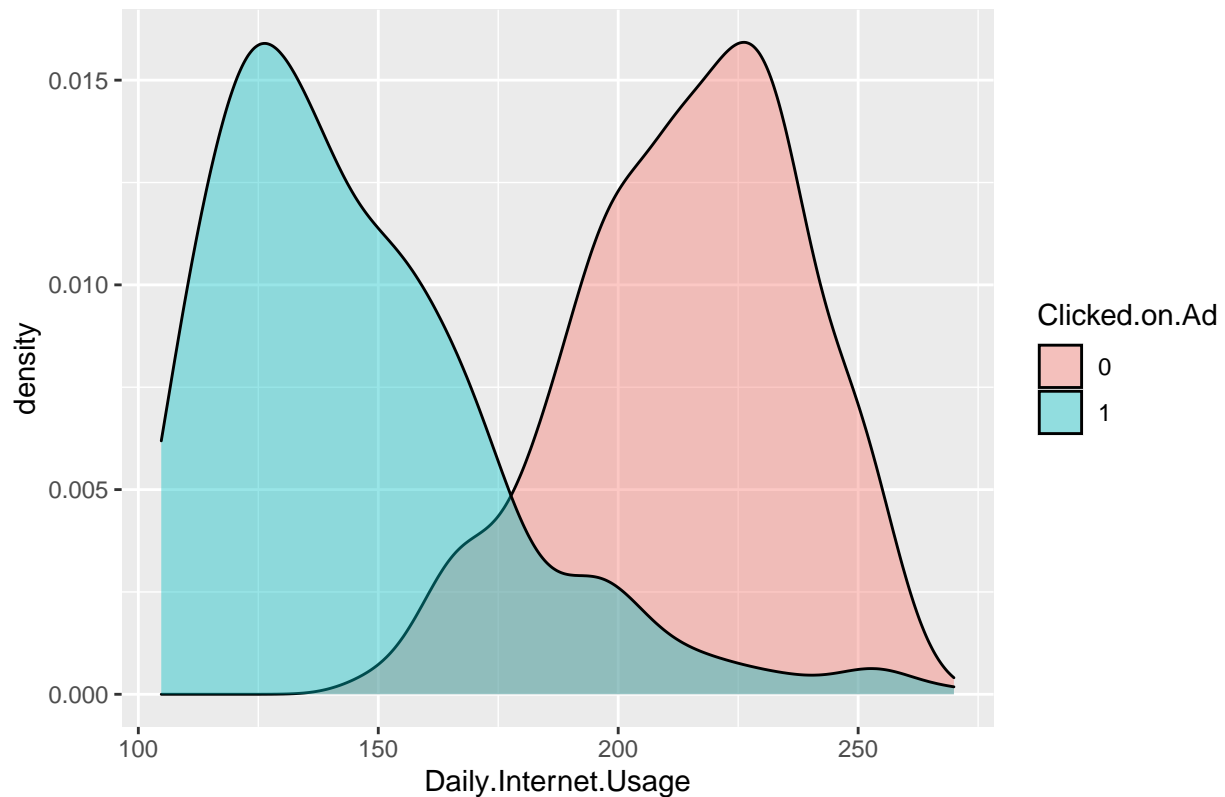
Relationship between time spent on site and chances of clicking on an ad



People who spend less time on the site are likely to click on an ad as compared to those who spend alot of time on the site. .

```
#Internet Usage and it's relationship to clicking an ad
ggplot(df_clean,
  aes(x = Daily.Internet.Usage,
      fill = Clicked.on.Ad)) +
  geom_density(alpha = 0.4) +
  labs(title = "Relationship between time spent on site and chances of clicking on an ad")
```

Relationship between time spent on site and chances of clicking on an ad



It seems the longer people spend on the internet, the likelier they are to click on the ads.

Conclusion

- i) People who have a daily internet usage of less than 175 are more likely to click on an ad
- ii) People who spend less than 70mins on the site are likely to click on ad
- iii) People above 40 are more likely to click on an ad
- iv) People with an income of less than 60000 are most likely to click on an ad

Challenging the solution

- i) It would be great to do some hypothesis testing on the conclusions made from Exploratory Data Analysis, this way we could ascertain the chances of specific person clicking on an ad or not.
- ii) Also, it would be necessary to create a predictive model and perform some feature importance selection to choose which variables are most important to use when deciding who will click on an ad or not when using the website.