

Cryptography Course Target Audience.(EDA)

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Exploratory Data Analysis of an online Cryptography Course Advert.

1) Defining the question.

a) Specifying the question.

Who is the most likely target audience for an online cryptography course?

b) The metric of success.

Finding a specific group of people who are likely to click on the advertisements given the data available.

c) The context.

Determining which factors make a person one of the target audiences for future forecasting.

d) Experimental design.

i) Loading and reading the data

ii) Data cleaning.

iii) Exploratory analysis.

iv) Conclusion.

v) Recommendation.

e) Appropriateness of the Data Available.

The data available has been collected from previous targeted adverts by the same entrepreneur and is therefore appropriate for this study.

2) Loading and reading the data.

```
# loading libraries
#
library(data.table)
library(tibble)
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.1 --

## v ggplot2 3.3.6      v dplyr 1.0.9
## v tidyr 1.2.0        v stringr 1.4.0
## v readr 2.1.2        v forcats 0.5.1
## v purrr 0.3.4

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::between()   masks data.table::between()
## x dplyr::filter()    masks stats::filter()
## x dplyr::first()     masks data.table::first()
## x dplyr::lag()       masks stats::lag()
## x dplyr::last()      masks data.table::last()
## x purrr::transpose() masks data.table::transpose()

library(corrplot)

## corrplot 0.92 loaded

library(ggplot2)
library(GGally)

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

# loading the dataset.
#
target <- fread('http://bit.ly/IPAdvertisingData')

# checking the top of the dataset.
#
head(target)

##      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
```

```
# checking the bottom of the dataset
#
tail(target)
```

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

```
# Converting dataset into tibble
#
target1 <- tibble(target)
target1
```

```
## # A tibble: 1,000 x 10
## 'Daily Time Spen~' Age 'Area Income' 'Daily Interne~' 'Ad Topic Line' City
## <dbl> <int> <dbl> <dbl> <chr> <chr>
## 1 69.0 35 61834. 256. Cloned 5thgene~ Wrig~
## 2 80.2 31 68442. 194. Monitored nati~ West~
## 3 69.5 26 59786. 236. Organic bottom~ Davi~
## 4 74.2 29 54806. 246. Triple-buffere~ West~
## 5 68.4 35 73890. 226. Robust logisti~ Sout~
```

```
## 6          60.0    23      59762.          227. Sharable clien~ Jami~
## 7          88.9    33      53853.          208. Enhanced dedic~ Bran~
## 8          66     48      24593.          132. Reactive local~ Port~
## 9          74.5    30      68862          222. Configurable c~ West~
## 10         69.9    20      55642.          184. Mandatory homo~ Rami~
## # ... with 990 more rows, and 4 more variables: Male <int>, Country <chr>,
## #   Timestamp <dtm>, 'Clicked on Ad' <int>
```

3) Data cleaning.

```
# checking for missing values
#
colSums(is.na(target1))
```

```
## 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
```

```
# The dataset contains no missing values
```

```
# checking for duplicates
#
duplicates <- target1[duplicated(target1),]
duplicates
```

```
## # A tibble: 0 x 10
## # ... with 10 variables: Daily Time Spent on Site <dbl>, Age <int>,
## #   Area Income <dbl>, Daily Internet Usage <dbl>, Ad Topic Line <chr>,
## #   City <chr>, Male <int>, Country <chr>, Timestamp <dtm>,
## #   Clicked on Ad <int>
```

```
# handling duplicates
#
# The dataset contains no duplicated rows
```

```
# checking for and removing outliers using IQR
#

Q1 = quantile(target1$'Age')
Q3 = quantile(target1$'Age')
IQR = IQR(target1$'Age')
no_out = subset(target1, target1$'Age' > (Q1 - 1.5*IQR) & target1$'Age' < (Q3 + 1.5*IQR))
no_out
```

```
## # A tibble: 748 x 10
##   'Daily Time Spent' Age 'Area Income' 'Daily Interne~' 'Ad Topic Line' City
##   <dbl> <int>      <dbl>          <dbl> <chr>          <chr>
## 1      69.0    35      61834.          256. Cloned 5thgene~ Wrig~
## 2      80.2    31      68442.          194. Monitored nati~ West~
## 3      69.5    26      59786.          236. Organic bottom~ Davi~
## 4      74.2    29      54806.          246. Triple-buffere~ West~
## 5      60.0    23      59762.          227. Sharable clien~ Jami~
## 6      88.9    33      53853.          208. Enhanced dedic~ Bran~
## 7      66      48      24593.          132. Reactive local~ Port~
## 8      74.5    30      68862.          222. Configurable c~ West~
## 9      83.1    37      62491.          231. Team-oriented ~ East~
## 10     69.6    48      51637.          113. Centralized co~ West~
## # ... with 738 more rows, and 4 more variables: Male <int>, Country <chr>,
## #   Timestamp <dtm>, 'Clicked on Ad' <int>
```

```
# selecting rows where clicked on ad is 1
#
target2 <- no_out %>% filter(no_out$'Clicked on Ad' == 1)
target2
```

```
## # A tibble: 361 x 10
##   'Daily Time Spent' Age 'Area Income' 'Daily Interne~' 'Ad Topic Line' City
##   <dbl> <int>      <dbl>          <dbl> <chr>          <chr>
## 1      66      48      24593.          132. Reactive local~ Port~
## 2      69.6    48      51637.          113. Centralized co~ West~
## 3      63.4    23      52182.          141. Persistent dem~ New ~
## 4      55.4    37      23937.          129. Customizable m~ West~
## 5      54.7    36      31088.          118. Grass-roots so~ Jess~
## 6      41.5    52      32636.          165. Mandatory disi~ Sout~
## 7      48.5    28      38067.          134. Ameliorated cl~ West~
## 8      52.0    52      58296.          129. Monitored syst~ Sout~
## 9      70.2    34      32709.          119. Open-architect~ Palm~
## 10     55.6    23      30228.          213. Multi-layered ~ Port~
## # ... with 351 more rows, and 4 more variables: Male <int>, Country <chr>,
## #   Timestamp <dtm>, 'Clicked on Ad' <int>
```

```
# dropping redundant variables
#
drop <- c("Timestamp", "Ad Topic Line")
target_ad = target2[,!(names(target2) %in% drop)]
target_ad
```

```
## # A tibble: 361 x 8
##   'Daily Time Spent ~' Age 'Area Income' 'Daily Interne~' City Male Country
##   <dbl> <int>      <dbl>          <dbl> <chr> <int> <chr>
## 1      66      48      24593.          132. Port~      1 Austr~
## 2      69.6    48      51637.          113. West~      1 Egypt
## 3      63.4    23      52182.          141. New ~      1 Spain
## 4      55.4    37      23937.          129. West~      0 Palest~
## 5      54.7    36      31088.          118. Jess~      1 Britis~
## 6      41.5    52      32636.          165. Sout~      0 Burundi
## 7      48.5    28      38067.          134. West~      1 Tuvalu
```

```
## 8          52.0    52      58296.      129. Sout~    0 Greece
## 9          70.2    34      32709.      119. Palm~    0 Britis~
## 10         55.6    23      30228.      213. Port~    0 Senegal
## # ... with 351 more rows, and 1 more variable: 'Clicked on Ad' <int>
```

4) Exploratory Analysis.

Univariate Analysis.

```
#calculating mean
#
mean(target_ad$'Daily Time Spent on Site')
```

Measures of Central Tendency.

```
## [1] 52.9731
```

```
mean(target_ad$'Age')
```

```
## [1] 39.30194
```

```
mean(target_ad$'Area Income')
```

```
## [1] 49106.58
```

```
mean(target_ad$'Daily Internet Usage')
```

```
## [1] 144.7557
```

```
#Calculating median
#
median(target_ad$'Daily Time Spent on Site')
```

```
## [1] 51.58
```

```
median(target_ad$'Age')
```

```
## [1] 39
```

```
median(target_ad$'Area Income')
```

```
## [1] 50055.33
```

```
median(target_ad$'Daily Internet Usage')
```

```
## [1] 137.43
```

```
# Calculating mode  
#  
getmode <- function(v) {  
  uniqv <- unique(v)  
  uniqv[which.max(tabulate(match(v, uniqv)))]  
}  
getmode(target_ad$'Daily Time Spent on Site')
```

```
## [1] 55.6
```

```
getmode(target_ad$'Age')
```

```
## [1] 45
```

```
getmode(target_ad$'Area Income')
```

```
## [1] 24593.33
```

```
getmode(target_ad$'Daily Internet Usage')
```

```
## [1] 120.06
```

```
getmode(target_ad$'Male')
```

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

```
# calculating minimum values  
#  
min(target_ad$'Daily Time Spent on Site')
```

Measures of Dispersion.

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

```
min(target_ad$'Age')
```

```
## [1] 20
```

```
min(target_ad$'Area Income')
```

```
## [1] 13996.5
```

```
min(target_ad$'Daily Internet Usage')
```

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

```
'-----'
```

```
## [1] "-----"
```

```
#calculating maximum values
```

```
#
```

```
max(target_ad$'Daily Time Spent on Site')
```

```
## [1] 88.97
```

```
max(target_ad$'Age')
```

```
## [1] 60
```

```
max(target_ad$'Area Income')
```

```
## [1] 77143.61
```

```
max(target_ad$'Daily Internet Usage')
```

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

```
# Get range, 1st and 3rd quantiles
```

```
#
```

```
quantile(target_ad$'Daily Time Spent on Site')
```

```
##    0%   25%   50%   75%  100%
```

```
## 32.60 43.07 51.58 61.22 88.97
```

```
quantile(target_ad$'Age')
```

```
##    0%   25%   50%   75%  100%
```

```
##   20   33   39   45   60
```

```
quantile(target_ad$'Area Income')
```

```
##          0%          25%          50%          75%          100%
```

```
## 13996.50 39799.73 50055.33 59243.46 77143.61
```

```
quantile(target_ad$'Daily Internet Usage')
```

```
##    0%   25%   50%   75%  100%
```

```
## 104.78 122.45 137.43 158.56 269.96
```



```

# variance and standard deviation
#
var(target_ad$'Daily Time Spent on Site')

## [1] 158.9318

var(target_ad$'Age')

## [1] 71.61136

var(target_ad$'Area Income')

## [1] 190658357

var(target_ad$'Daily Internet Usage')

## [1] 971.8155

'-----'

## [1] "-----"

sd(target_ad$'Daily Time Spent on Site')

## [1] 12.60682

sd(target_ad$'Age')

## [1] 8.462349

sd(target_ad$'Area Income')

## [1] 13807.91

sd(target_ad$'Daily Internet Usage')

## [1] 31.17396

# frequency of categorical variables
#
library(plyr)

## -----

## You have loaded plyr after dplyr - this is likely to cause problems.
## If you need functions from both plyr and dplyr, please load plyr first, then dplyr:
## library(plyr); library(dplyr)

```

```
## -----
```

```
##
```

```
## Attaching package: 'plyr'
```

```
## The following objects are masked from 'package:dplyr':
```

```
##
```

```
##      arrange, count, desc, failwith, id, mutate, rename, summarise,  
##      summarize
```

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

```
##
```

```
##      compact
```

```
library(dplyr)  
country_freq = count(target_ad, 'Country')  
country_freq
```

```
##           Country freq  
## 1      Afghanistan    4  
## 2         Albania    3  
## 3         Algeria    2  
## 4   American Samoa    3  
## 5         Andorra    2  
## 6         Angola    1  
## 7       Anguilla    2  
## 8  Antarctica (the territory South of 60 deg S)    1  
## 9     Antigua and Barbuda    4  
## 10        Argentina    1  
## 11        Armenia    1  
## 12        Australia    5  
## 13         Austria    1  
## 14        Bahamas    3  
## 15        Bahrain    2  
## 16       Bangladesh    2  
## 17         Belarus    2  
## 18         Belgium    1  
## 19         Belize    2  
## 20         Benin    1  
## 21         Bhutan    1  
## 22    Bosnia and Herzegovina    3  
## 23   Bouvet Island (Bouvetoya)    2  
## 24         Brazil    2  
## 25 British Indian Ocean Territory (Chagos Archipelago)    1  
## 26   British Virgin Islands    1  
## 27      Brunei Darussalam    2  
## 28       Burkina Faso    1  
## 29         Burundi    2  
## 30         Cambodia    1  
## 31         Canada    1  
## 32       Cayman Islands    2  
## 33   Central African Republic    1  
## 34         Chad    2
```

## 35	Chile	2
## 36	China	3
## 37	Christmas Island	3
## 38	Colombia	1
## 39	Comoros	1
## 40	Congo	2
## 41	Cook Islands	1
## 42	Cote d'Ivoire	2
## 43	Cuba	3
## 44	Cyprus	2
## 45	Czech Republic	2
## 46	Denmark	2
## 47	Djibouti	1
## 48	Dominica	2
## 49	Dominican Republic	1
## 50	Ecuador	1
## 51	Egypt	3
## 52	El Salvador	3
## 53	Equatorial Guinea	3
## 54	Eritrea	1
## 55	Estonia	1
## 56	Ethiopia	3
## 57	Faroe Islands	2
## 58	Fiji	3
## 59	Finland	1
## 60	France	3
## 61	French Guiana	1
## 62	French Polynesia	1
## 63	French Southern Territories	1
## 64	Gambia	1
## 65	Georgia	2
## 66	Germany	1
## 67	Greece	3
## 68	Greenland	1
## 69	Guadeloupe	1
## 70	Guam	1
## 71	Guatemala	2
## 72	Guernsey	2
## 73	Guinea	2
## 74	Guinea-Bissau	1
## 75	Guyana	3
## 76	Heard Island and McDonald Islands	2
## 77	Holy See (Vatican City State)	1
## 78	Honduras	2
## 79	Hong Kong	4
## 80	Hungary	4
## 81	Iceland	1
## 82	Indonesia	4
## 83	Iran	1
## 84	Ireland	1
## 85	Israel	2
## 86	Italy	1
## 87	Jamaica	1
## 88	Japan	2

## 89	Jersey	3
## 90	Kazakhstan	1
## 91	Kenya	4
## 92	Korea	3
## 93	Kuwait	1
## 94	Lao People's Democratic Republic	2
## 95	Latvia	4
## 96	Lebanon	3
## 97	Liberia	3
## 98	Libyan Arab Jamahiriya	1
## 99	Liechtenstein	4
## 100	Lithuania	2
## 101	Luxembourg	3
## 102	Macao	1
## 103	Macedonia	1
## 104	Malawi	1
## 105	Mali	1
## 106	Malta	1
## 107	Marshall Islands	1
## 108	Martinique	2
## 109	Mauritania	1
## 110	Mayotte	3
## 111	Mexico	3
## 112	Micronesia	4
## 113	Moldova	2
## 114	Monaco	1
## 115	Mongolia	3
## 116	Montenegro	1
## 117	Montserrat	1
## 118	Morocco	1
## 119	Namibia	1
## 120	Netherlands	2
## 121	Netherlands Antilles	1
## 122	New Caledonia	2
## 123	New Zealand	2
## 124	Niger	2
## 125	Norfolk Island	1
## 126	Northern Mariana Islands	2
## 127	Pakistan	1
## 128	Palau	2
## 129	Palestinian Territory	2
## 130	Papua New Guinea	3
## 131	Paraguay	1
## 132	Peru	3
## 133	Philippines	3
## 134	Pitcairn Islands	1
## 135	Poland	3
## 136	Puerto Rico	1
## 137	Romania	1
## 138	Rwanda	2
## 139	Saint Barthelemy	2
## 140	Saint Helena	1
## 141	Saint Kitts and Nevis	1
## 142	Saint Lucia	1

## 143	Saint Martin	2
## 144	Saint Pierre and Miquelon	3
## 145	Saint Vincent and the Grenadines	3
## 146	Samoa	3
## 147	Saudi Arabia	1
## 148	Senegal	3
## 149	Serbia	2
## 150	Sierra Leone	2
## 151	Slovenia	1
## 152	Somalia	1
## 153	South Africa	6
## 154	South Georgia and the South Sandwich Islands	1
## 155	Spain	2
## 156	Suriname	1
## 157	Svalbard & Jan Mayen Islands	3
## 158	Sweden	1
## 159	Switzerland	3
## 160	Syrian Arab Republic	1
## 161	Taiwan	4
## 162	Tajikistan	2
## 163	Tanzania	1
## 164	Thailand	1
## 165	Timor-Leste	1
## 166	Tokelau	2
## 167	Tonga	1
## 168	Trinidad and Tobago	1
## 169	Turkey	6
## 170	Turkmenistan	1
## 171	Turks and Caicos Islands	3
## 172	Tuvalu	3
## 173	Uganda	4
## 174	Ukraine	1
## 175	United Arab Emirates	1
## 176	United Kingdom	1
## 177	United States Minor Outlying Islands	1
## 178	United States of America	3
## 179	United States Virgin Islands	2
## 180	Uzbekistan	1
## 181	Vanuatu	1
## 182	Venezuela	2
## 183	Vietnam	2
## 184	Western Sahara	3
## 185	Zambia	3
## 186	Zimbabwe	2

```
sorted_by_freq <- country_freq[order(-country_freq$freq),]
sorted_by_freq
```

##	Country	freq
## 153	South Africa	6
## 169	Turkey	6
## 12	Australia	5
## 1	Afghanistan	4
## 9	Antigua and Barbuda	4

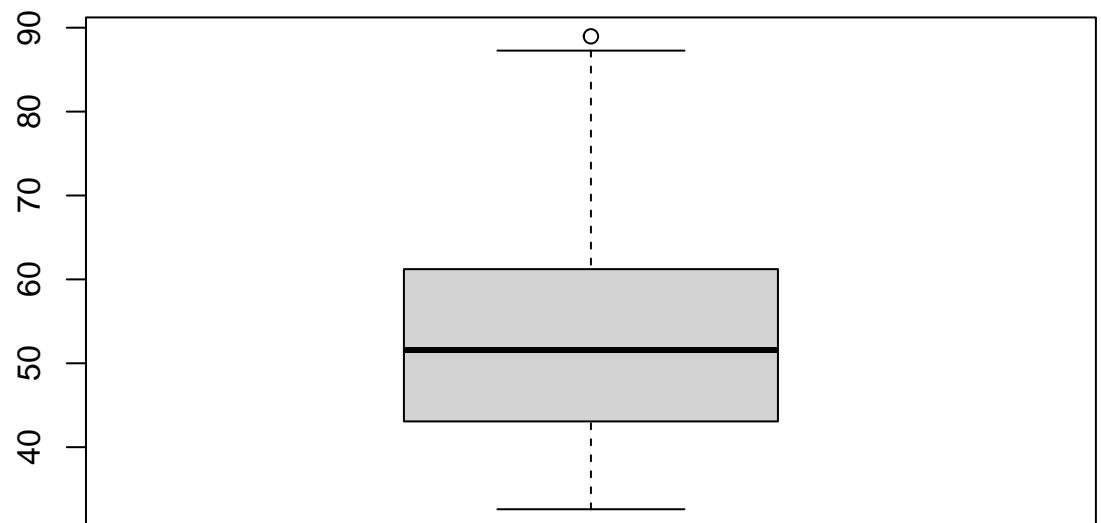
## 79	Hong Kong	4
## 80	Hungary	4
## 82	Indonesia	4
## 91	Kenya	4
## 95	Latvia	4
## 99	Liechtenstein	4
## 112	Micronesia	4
## 161	Taiwan	4
## 173	Uganda	4
## 2	Albania	3
## 4	American Samoa	3
## 14	Bahamas	3
## 22	Bosnia and Herzegovina	3
## 36	China	3
## 37	Christmas Island	3
## 43	Cuba	3
## 51	Egypt	3
## 52	El Salvador	3
## 53	Equatorial Guinea	3
## 56	Ethiopia	3
## 58	Fiji	3
## 60	France	3
## 67	Greece	3
## 75	Guyana	3
## 89	Jersey	3
## 92	Korea	3
## 96	Lebanon	3
## 97	Liberia	3
## 101	Luxembourg	3
## 110	Mayotte	3
## 111	Mexico	3
## 115	Mongolia	3
## 130	Papua New Guinea	3
## 132	Peru	3
## 133	Philippines	3
## 135	Poland	3
## 144	Saint Pierre and Miquelon	3
## 145	Saint Vincent and the Grenadines	3
## 146	Samoa	3
## 148	Senegal	3
## 157	Svalbard & Jan Mayen Islands	3
## 159	Switzerland	3
## 171	Turks and Caicos Islands	3
## 172	Tuvalu	3
## 178	United States of America	3
## 184	Western Sahara	3
## 185	Zambia	3
## 3	Algeria	2
## 5	Andorra	2
## 7	Anguilla	2
## 15	Bahrain	2
## 16	Bangladesh	2
## 17	Belarus	2
## 19	Belize	2

## 23	Bouvet Island (Bouvetoya)	2
## 24	Brazil	2
## 27	Brunei Darussalam	2
## 29	Burundi	2
## 32	Cayman Islands	2
## 34	Chad	2
## 35	Chile	2
## 40	Congo	2
## 42	Cote d'Ivoire	2
## 44	Cyprus	2
## 45	Czech Republic	2
## 46	Denmark	2
## 48	Dominica	2
## 57	Faroe Islands	2
## 65	Georgia	2
## 71	Guatemala	2
## 72	Guernsey	2
## 73	Guinea	2
## 76	Heard Island and McDonald Islands	2
## 78	Honduras	2
## 85	Israel	2
## 88	Japan	2
## 94	Lao People's Democratic Republic	2
## 100	Lithuania	2
## 108	Martinique	2
## 113	Moldova	2
## 120	Netherlands	2
## 122	New Caledonia	2
## 123	New Zealand	2
## 124	Niger	2
## 126	Northern Mariana Islands	2
## 128	Palau	2
## 129	Palestinian Territory	2
## 138	Rwanda	2
## 139	Saint Barthelemy	2
## 143	Saint Martin	2
## 149	Serbia	2
## 150	Sierra Leone	2
## 155	Spain	2
## 162	Tajikistan	2
## 166	Tokelau	2
## 179	United States Virgin Islands	2
## 182	Venezuela	2
## 183	Vietnam	2
## 186	Zimbabwe	2
## 6	Angola	1
## 8	Antarctica (the territory South of 60 deg S)	1
## 10	Argentina	1
## 11	Armenia	1
## 13	Austria	1
## 18	Belgium	1
## 20	Benin	1
## 21	Bhutan	1
## 25	British Indian Ocean Territory (Chagos Archipelago)	1

## 26	British Virgin Islands	1
## 28	Burkina Faso	1
## 30	Cambodia	1
## 31	Canada	1
## 33	Central African Republic	1
## 38	Colombia	1
## 39	Comoros	1
## 41	Cook Islands	1
## 47	Djibouti	1
## 49	Dominican Republic	1
## 50	Ecuador	1
## 54	Eritrea	1
## 55	Estonia	1
## 59	Finland	1
## 61	French Guiana	1
## 62	French Polynesia	1
## 63	French Southern Territories	1
## 64	Gambia	1
## 66	Germany	1
## 68	Greenland	1
## 69	Guadeloupe	1
## 70	Guam	1
## 74	Guinea-Bissau	1
## 77	Holy See (Vatican City State)	1
## 81	Iceland	1
## 83	Iran	1
## 84	Ireland	1
## 86	Italy	1
## 87	Jamaica	1
## 90	Kazakhstan	1
## 93	Kuwait	1
## 98	Libyan Arab Jamahiriya	1
## 102	Macao	1
## 103	Macedonia	1
## 104	Malawi	1
## 105	Mali	1
## 106	Malta	1
## 107	Marshall Islands	1
## 109	Mauritania	1
## 114	Monaco	1
## 116	Montenegro	1
## 117	Montserrat	1
## 118	Morocco	1
## 119	Namibia	1
## 121	Netherlands Antilles	1
## 125	Norfolk Island	1
## 127	Pakistan	1
## 131	Paraguay	1
## 134	Pitcairn Islands	1
## 136	Puerto Rico	1
## 137	Romania	1
## 140	Saint Helena	1
## 141	Saint Kitts and Nevis	1
## 142	Saint Lucia	1

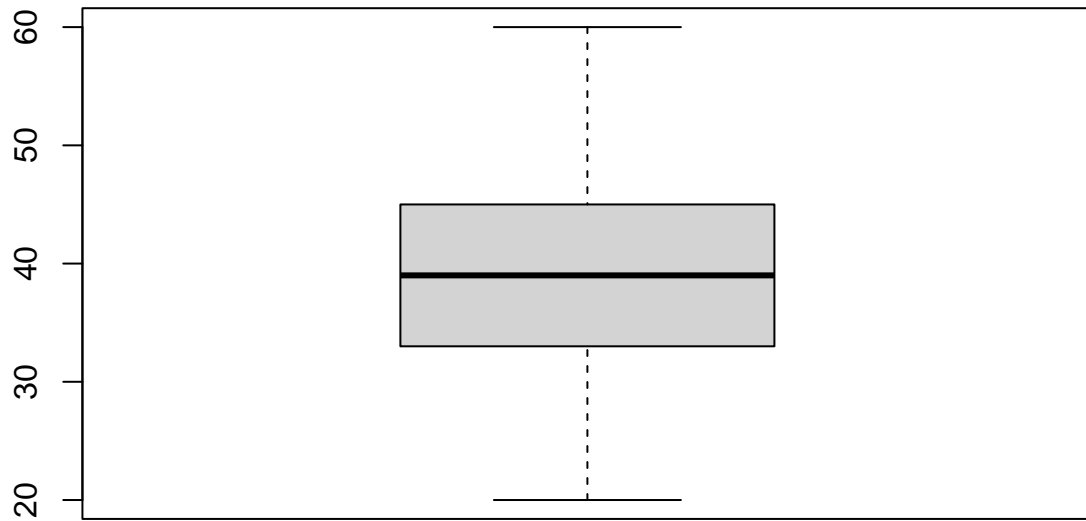

```
## 147          Saudi Arabia 1
## 151          Slovenia 1
## 152          Somalia 1
## 154    South Georgia and the South Sandwich Islands 1
## 156          Suriname 1
## 158          Sweden 1
## 160    Syrian Arab Republic 1
## 163          Tanzania 1
## 164          Thailand 1
## 165    Timor-Leste 1
## 167          Tonga 1
## 168    Trinidad and Tobago 1
## 170    Turkmenistan 1
## 174          Ukraine 1
## 175    United Arab Emirates 1
## 176    United Kingdom 1
## 177    United States Minor Outlying Islands 1
## 180          Uzbekistan 1
## 181          Vanuatu 1
```

```
#summary of above using boxplots
#
boxplot(target_ad$'Daily Time Spent on Site')
```

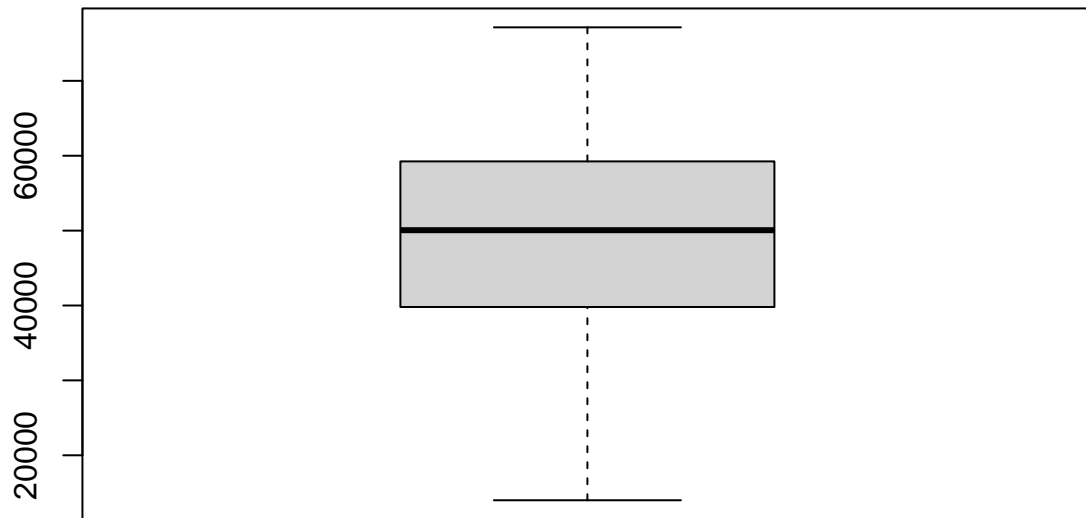


Visualization.

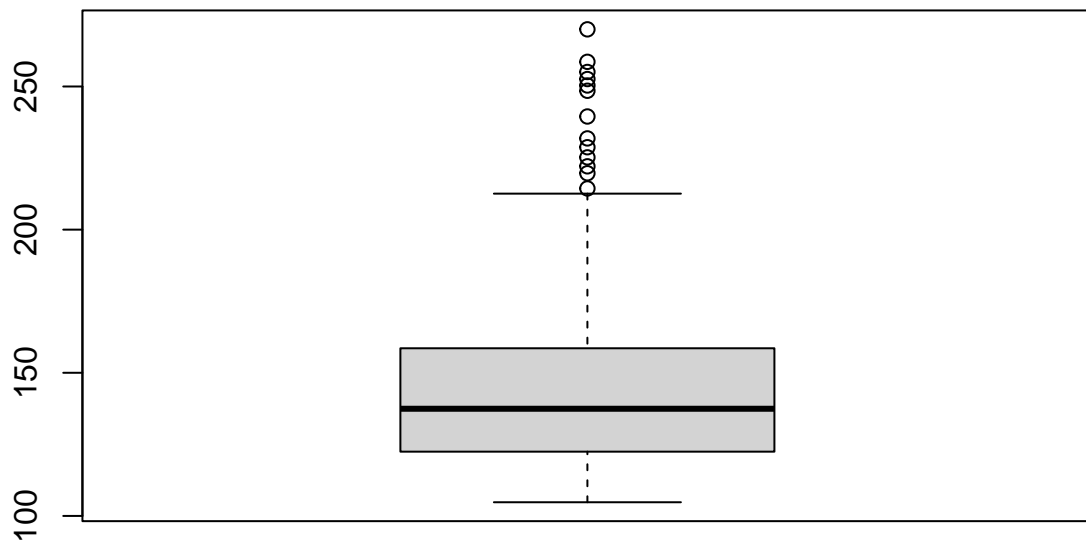
```
boxplot(target_ad$'Age')
```



```
boxplot(target_ad$'Area Income')
```

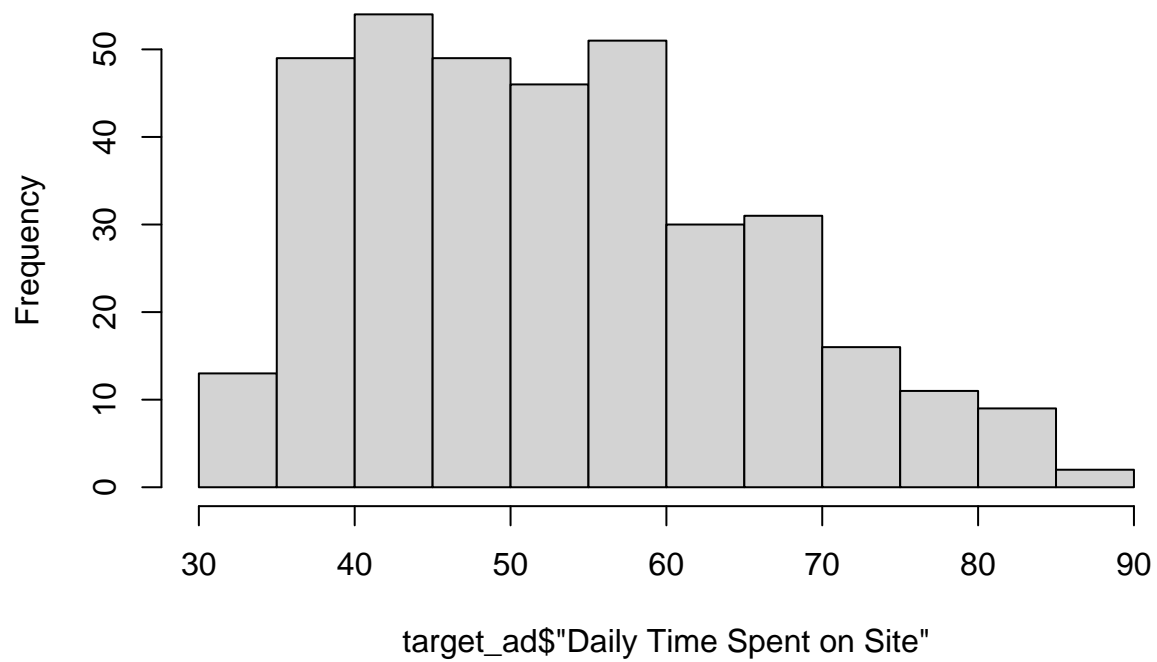


```
boxplot(target_ad$'Daily Internet Usage')
```



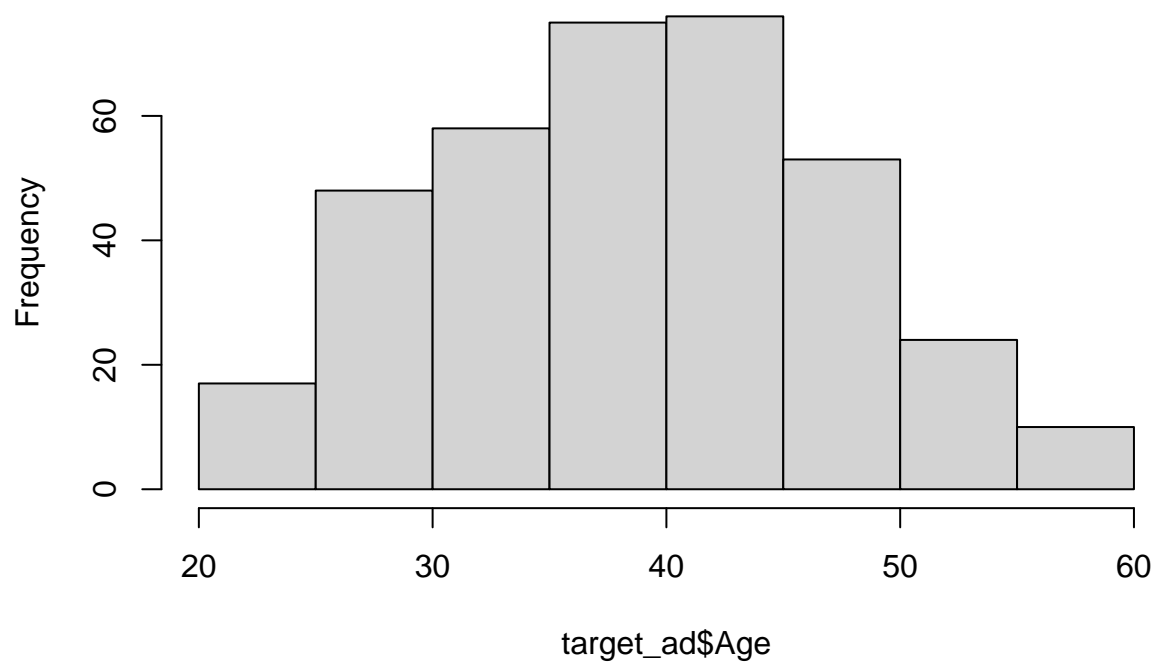
```
# visualisation of numeric variables using a histogam  
#  
hist(target_ad$'Daily Time Spent on Site')
```

Histogram of target_ad\$"Daily Time Spent on Site"



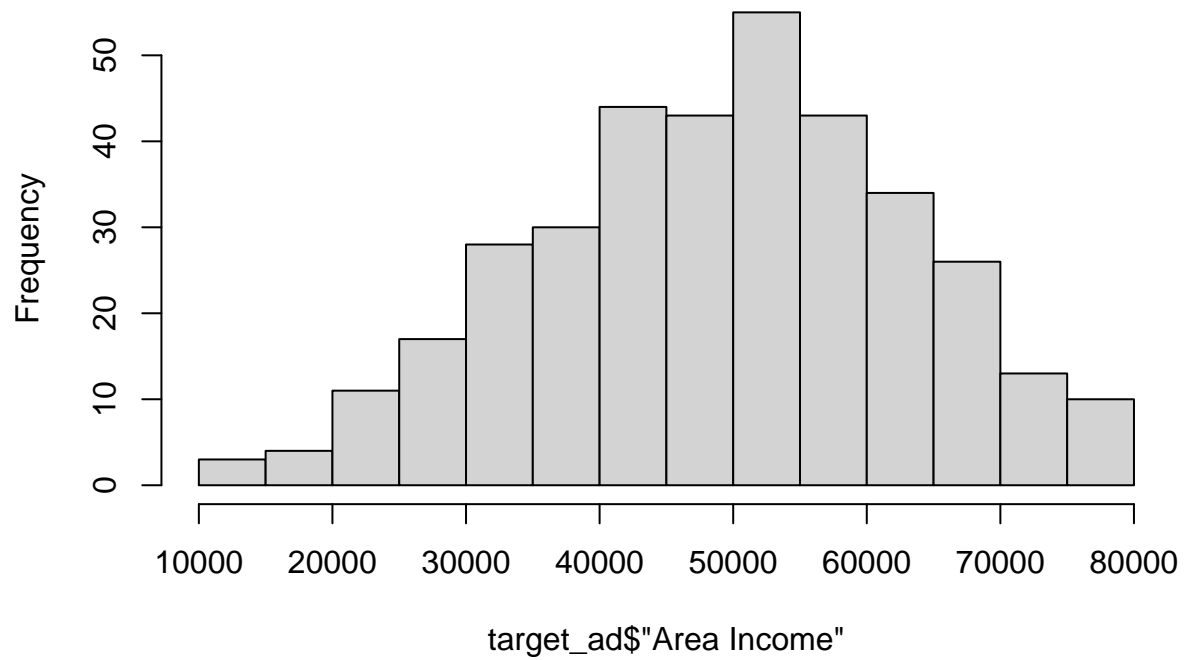
```
hist(target_ad$'Age')
```

Histogram of target_ad\$Age



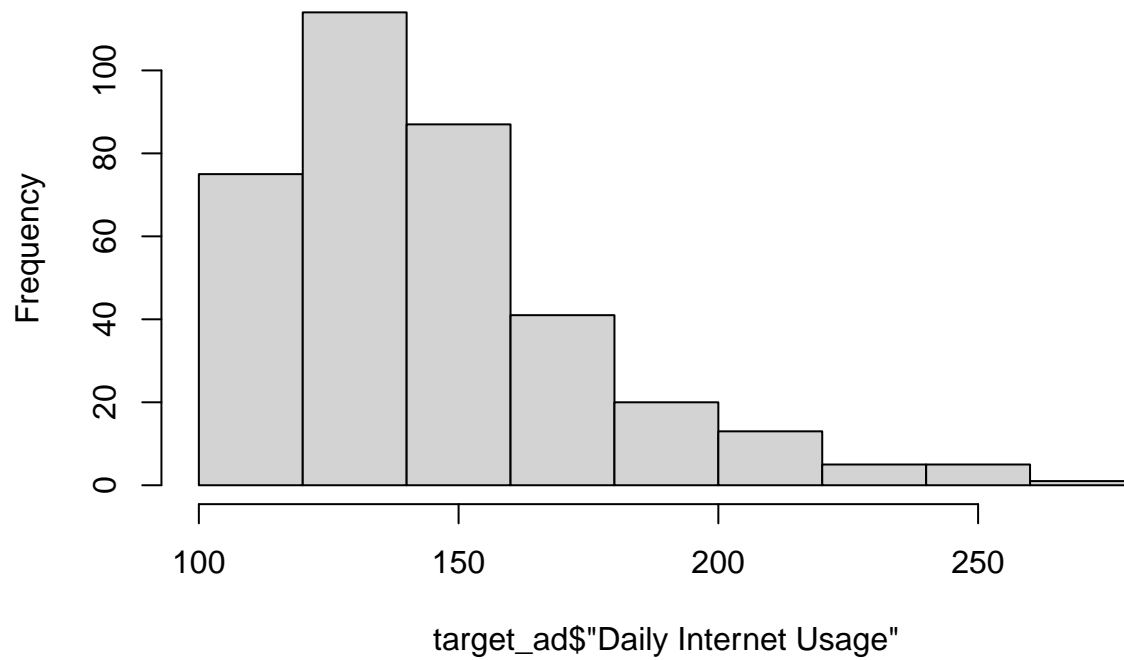
```
hist(target_ad$'Area Income')
```

Histogram of target_ad\$"Area Income"

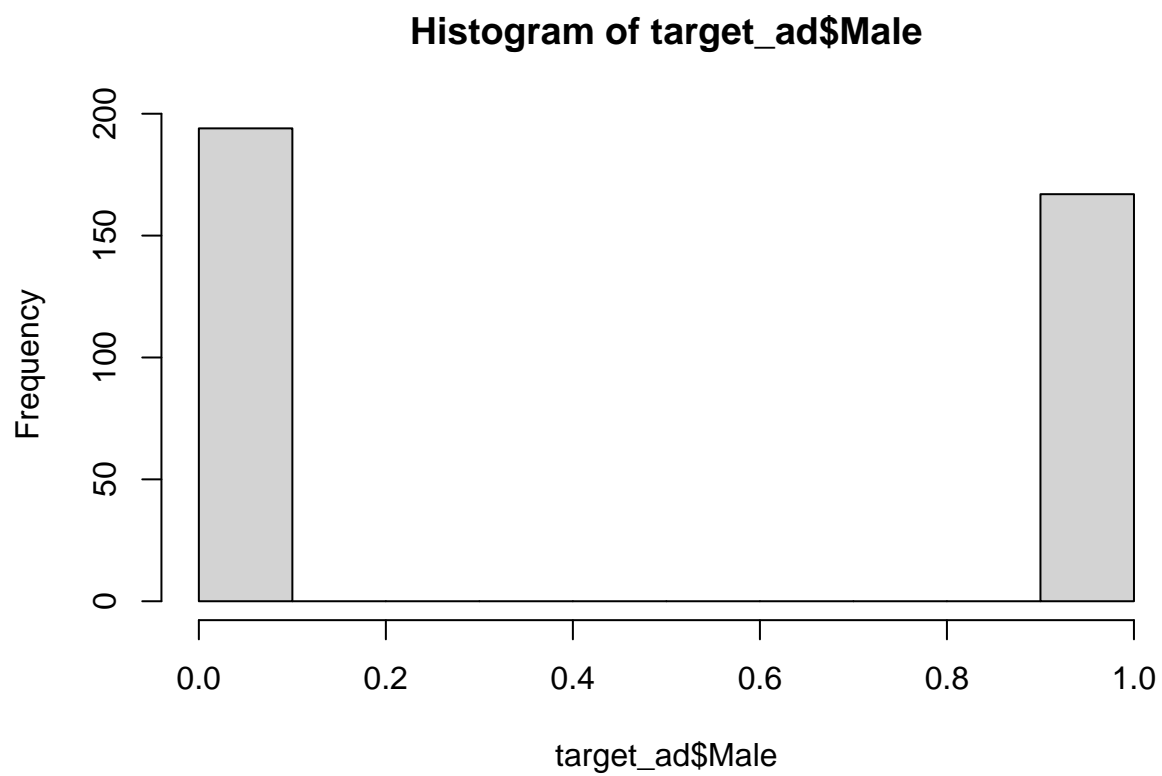


```
hist(target_ad$'Daily Internet Usage')
```

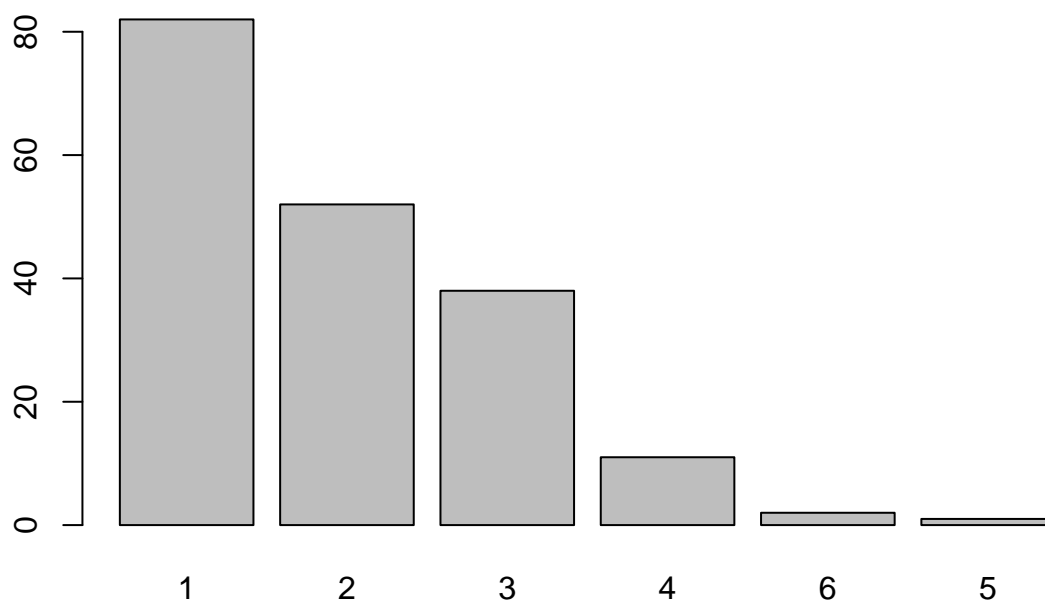
Histogram of target_ad\$"Daily Internet Usage"



```
hist(target_ad$'Male')
```

```
# Barplot of countries.  
#  
x <- table(country_freq$freq)  
barplot(x[order(x, decreasing=TRUE)])
```



Bivariate Analysis #### Covariance and Correlation.

Covariance of numeric variables

#

```
drop <- c("Country", "City", "Location", "Ad Topic Line", "Clicked on Ad")
target4 = target_ad[!(names(target_ad) %in% drop)]
target4
```

A tibble: 361 x 5

	'Daily Time Spent on Site'	Age	'Area Income'	'Daily Internet Usage'	Male
	<dbl>	<int>	<dbl>	<dbl>	<int>
## 1	66	48	24593.	132.	1
## 2	69.6	48	51637.	113.	1
## 3	63.4	23	52182.	141.	1
## 4	55.4	37	23937.	129.	0
## 5	54.7	36	31088.	118.	1
## 6	41.5	52	32636.	165.	0
## 7	48.5	28	38067.	134.	1
## 8	52.0	52	58296.	129.	0
## 9	70.2	34	32709.	119.	0
## 10	55.6	23	30228.	213.	0

... with 351 more rows

```
cov(target4)
```

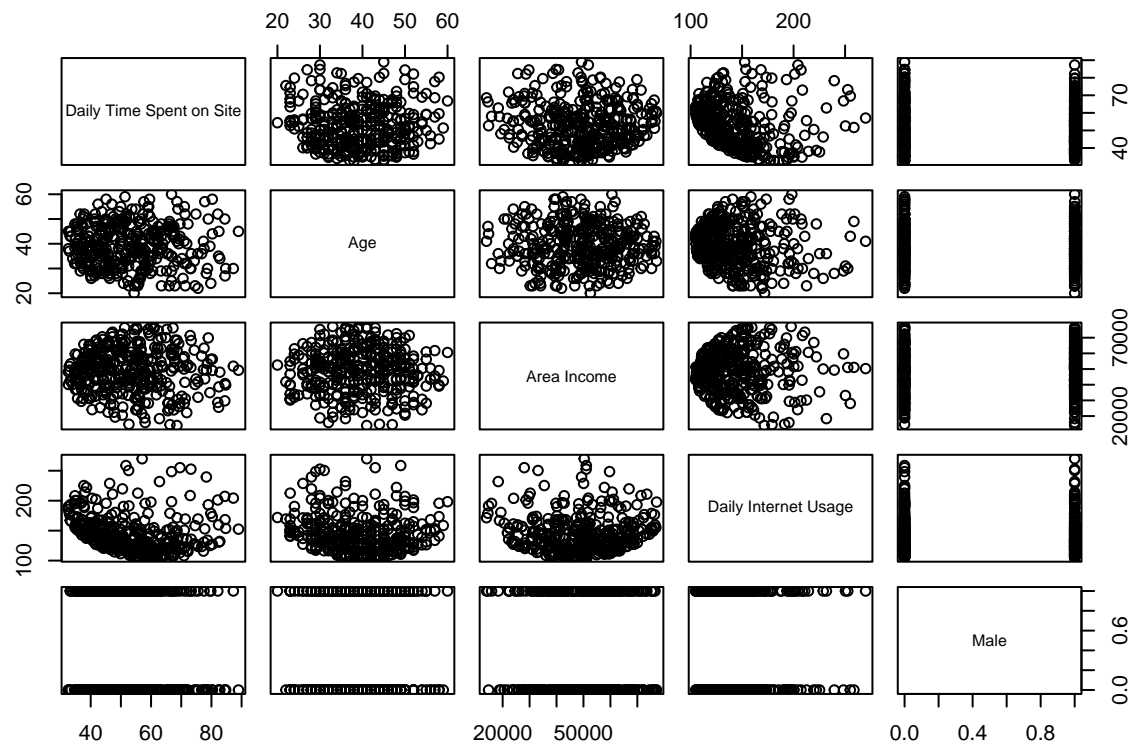
	Daily Time Spent on Site	Age	Area Income
--	--------------------------	-----	-------------

```
## Daily Time Spent on Site      158.9317865 -1.329912e+00 -7.350829e+03
## Age                          -1.3299116  7.161136e+01 -2.411858e+03
## Area Income                  -7350.8286871 -2.411858e+03  1.906584e+08
## Daily Internet Usage         -68.1159257 -3.085422e+01 -7.959702e+03
## Male                         -0.7673003 -2.339951e-02  3.556316e+02
##                               Daily Internet Usage      Male
## Daily Time Spent on Site     -6.811593e+01 -0.767300323
## Age                          -3.085422e+01 -0.023399508
## Area Income                  -7.959702e+03 355.631570714
## Daily Internet Usage         9.718155e+02 -0.003606494
## Male                         -3.606494e-03  0.249292090
```

```
#correlation matrix
#
correlation = cor(target4)
correlation
```

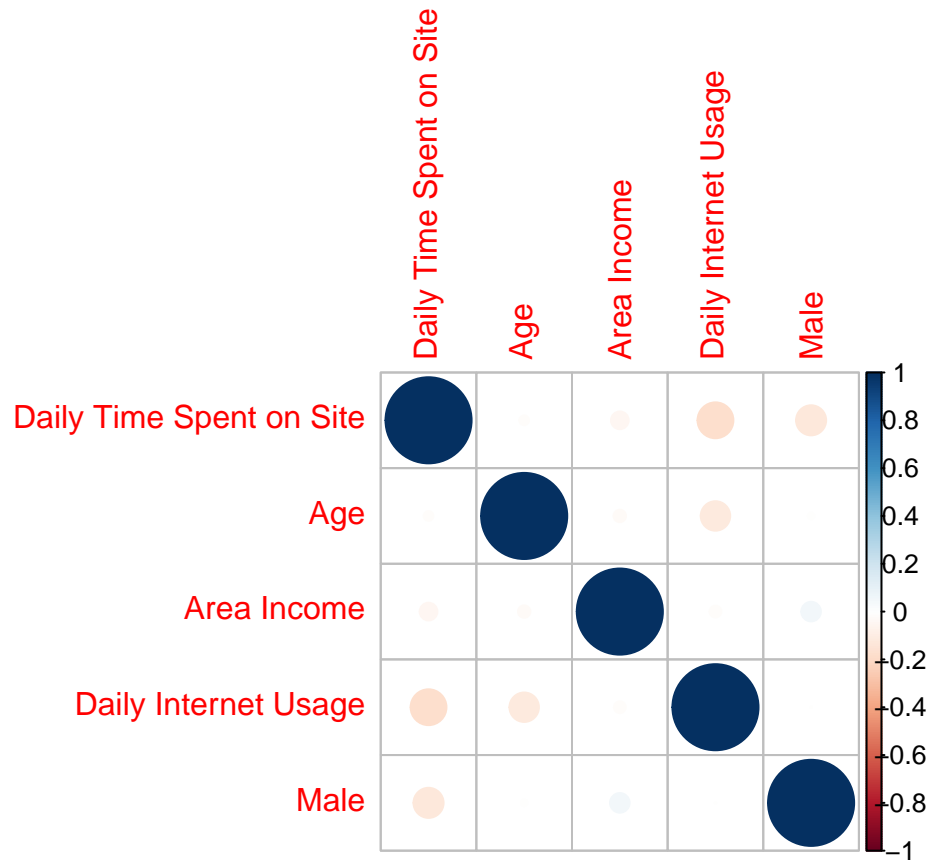
```
##                               Daily Time Spent on Site      Age Area Income
## Daily Time Spent on Site      1.00000000 -0.012465980 -0.04222824
## Age                          -0.01246598  1.000000000 -0.02064110
## Area Income                  -0.04222824 -0.020641098  1.00000000
## Daily Internet Usage         -0.17332107 -0.116958473 -0.01849171
## Male                         -0.12190057 -0.005538109  0.05158437
##                               Daily Internet Usage      Male
## Daily Time Spent on Site     -0.173321069 -0.121900574
## Age                          -0.116958473 -0.005538109
## Area Income                  -0.018491706  0.051584371
## Daily Internet Usage         1.000000000 -0.000231707
## Male                         -0.000231707  1.000000000
```

```
# With a scatterplot matrix
#
plot(target4)
```



Visualization.

```
# visualization of the correlation matrix
#
corrplot(correlation)
```



5) Conclusion

Give the results of this analysis the most likely group of people to click on the targeted ads are:

- a) Those aged between 35-45
- b) Who spend between 50 minutes to an hour per day on the site
- c) From South Africa, Turkey, Australia. though not limited to this countries only.

6) Recommendation.

More data is needed for further analysis. However, what is currently analyzed could still be useful to launch a pilot advertising program and gauge how well it works and reaches the target audience.