

# Online\_Cryptography\_Ads\_Analysis\_using\_EDA

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## 1. Defining the Question

### a) Specifying the Data Analytic Question

To create a model to identify which individuals are most likely to click on ads.

### b) Defining the Metric for Success

The study will be considered successful if we shall be able to get insight from the data using the Exploratory data analysis.

### c) Understanding the context

Advertising is a means of communication with the users of a product or service. Advertisements are messages paid for by those who send them and are intended to inform or influence people who receive them. Advertising is always present, though people may not be aware of it. In today's world, advertising uses every possible media to get its message through. It does this via television, print (newspapers, magazines, journals etc), radio, press, internet, direct selling, hoardings, mailers, contests, sponsorships, posters, clothes, events, colours, sounds, visuals and even people (endorsements). In our model, we shall try to predict the probability of a person clicking on an ad.

### d). Recording the Experimental Design

We will use exploratory data analysis, such as Univariate and Bivariate in this study to determine the relationships and differences between different variables.

### e) Data Relevance

The dataset to use for this project can be found by following this link: <https://www.bit.ly/IPAdvertisingData>

## 2. Reading the Data

```
advertising <- read.csv("https://www.bit.ly/IPAdvertisingData")
```

### 3. Checking the Data

#### a) Checking the top data

```
head(advertising)
```

```
##      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
```

#### b). Checking the bottom data

```
tail(advertising)
```

```
##      Daily.Time.Spent.on.Site Age Area.Income Daily.Internet.Usage
## 995                43.70  28    63126.96                173.01
## 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
## 995   Front-line bifurcated ability Nicholasland  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
## 995   Mayotte 2016-04-04 03:57:48      1
## 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
```

### c). Checking the Structure of the Dataset

```
str(advertising)
```

```
## '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                : chr  "2016-03-27 00:53:11" "2016-04-04 01:39:02" "2016-03-13 20:35:42"
## $ Clicked.on.Ad            : int  0 0 0 0 0 0 0 1 0 0 ...
```

### d). Checking the shape of our data

```
dim(advertising)
```

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

We have 1000 rows and 10 columns in our dataset

## 4. Tidying the Dataset

### a). Checking the Missing Values

```
colSums(is.na(advertising))
```

```
## 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
```

From the above, we can see that we do not have Missing Values in the dataset.

## b). Checking for Duplicate Values

```
duplicated_rows <- advertising[duplicated(advertising),]  
duplicated_rows
```

```
## [1] Daily.Time.Spent.on.Site Age Area.Income  
## [4] Daily.Internet.Usage Ad.Topic.Line City  
## [7] Male Country Timestamp  
## [10] Clicked.on.Ad  
## <0 rows> (or 0-length row.names)
```

We can also see that we have 0 rows containing duplicates values. This is very import for the consintency of data.

## c). Checking for Outliers

We shall use Boxplot to check for outliers in our numeric features

```
# Selecting Numeric columns
```

```
library("dplyr")
```

```
##
```

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

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

```
##
```

```
## filter, lag
```

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

```
##
```

```
## intersect, setdiff, setequal, union
```

```
# getting numeric columns using dplyr() function
```

```
numeric_col <- select_if(advertising, is.numeric)
```

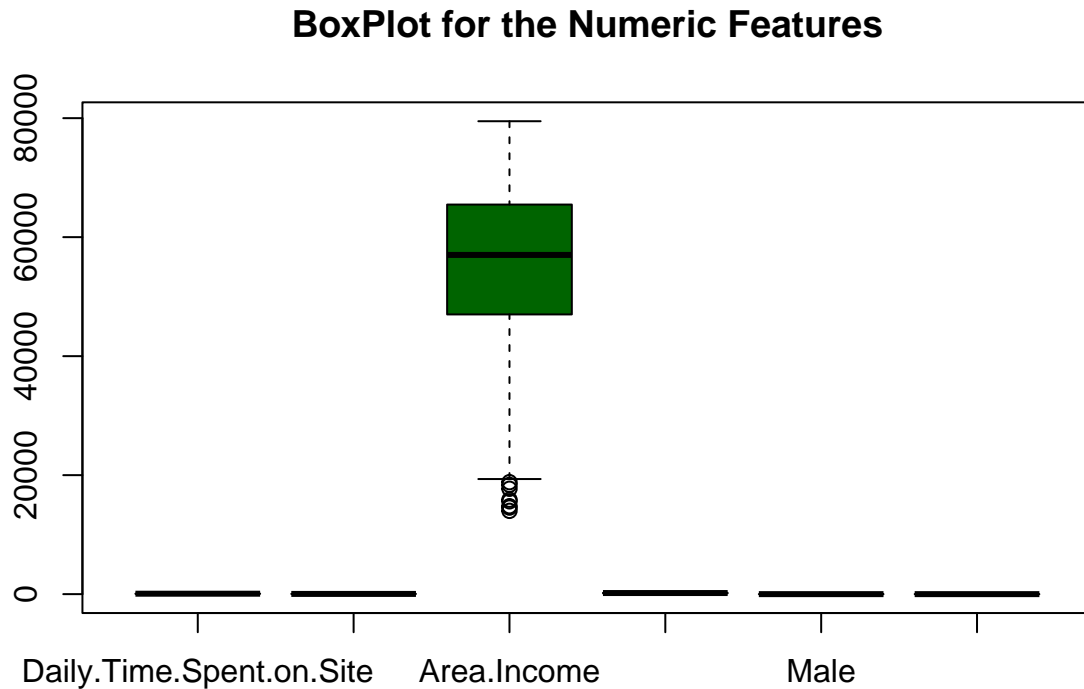
```
head(numeric_col)
```

```
## Daily.Time.Spent.on.Site Age Area.Income Daily.Internet.Usage Male  
## 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  
## 1 0  
## 2 0
```

```
## 3      0
## 4      0
## 5      0
## 6      0
```

```
# Plotting Boxplot for the Numeric columns
```

```
boxplot(numeric_col,main="BoxPlot for the Numeric Features", col="darkgreen")
```



We can clearly see we have outliers in our Area.Income feature. We shall be removing the outlier using the Interquartile Range Method

```
outliersValue <- boxplot.stats(advertising$Area.Income)$out
```

```
advertising$Area.Income[!advertising$Area.Income %in% outliersValue]
```

```
## [1] 61833.90 68441.85 59785.94 54806.18 73889.99 59761.56 53852.85 24593.33
## [9] 68862.00 55642.32 45632.51 62491.01 51636.92 51739.63 30976.00 52182.23
## [17] 23936.86 71511.08 31087.54 23821.72 64802.33 60015.57 32635.70 61628.72
## [25] 68962.32 64828.00 38067.08 58295.82 32708.94 46179.97 51473.28 45593.93
## [33] 25583.29 30227.98 45580.92 61389.50 56770.79 76435.30 57425.87 27508.41
## [41] 57691.95 59784.18 66572.39 64929.61 57519.64 53575.48 50983.75 67058.72
## [49] 52723.34 54286.10 61526.25 58526.04 53350.11 62657.53 62722.57 67479.62
## [57] 75254.88 52336.64 56113.37 24852.90 47708.42 64654.66 71228.44 61601.05
## [65] 66281.46 73910.90 51317.33 51510.18 61005.87 32536.98 60248.97 74543.81
## [73] 75509.61 42650.32 58183.04 60465.72 57009.76 54541.56 32689.04 55605.92
```

```

## [81] 63296.87 65653.47 61652.53 30726.26 74535.94 47861.93 73600.28 58543.94
## [89] 42696.67 37334.78 71392.53 59550.05 64264.25 64147.86 25686.34 52968.22
## [97] 22473.08 64927.19 51868.85 69456.83 31947.65 51864.77 59593.56 48376.14
## [105] 56884.74 67186.54 46557.92 66541.05 33258.09 72272.90 60333.38 65229.13
## [113] 56067.38 37838.72 72683.35 56729.78 66815.54 60223.52 29727.79 49269.98
## [121] 57669.41 56791.75 63274.88 35466.80 68787.09 61227.59 56366.88 57868.44
## [129] 66618.21 73104.47 21644.91 53817.02 76368.31 67633.44 50335.46 41229.16
## [137] 42581.23 61617.98 70575.60 64122.36 52097.32 65953.76 60192.72 77460.07
## [145] 45716.48 65120.86 49995.63 71718.51 61770.34 69112.84 72524.86 36782.38
## [153] 66699.12 64287.78 56637.59 55787.58 61142.33 61625.87 73234.87 74166.24
## [161] 62669.59 57756.89 58019.64 50960.08 48246.60 28271.84 53767.12 43662.10
## [169] 62238.58 49030.03 76003.47 68094.85 64395.85 70053.27 72423.97 42995.80
## [177] 60309.58 38349.78 63115.34 31343.39 40763.13 36752.24 65044.59 53673.08
## [185] 43444.86 44248.52 62572.88 39840.55 32593.59 41629.86 43313.73 42993.48
## [193] 46004.31 49325.48 51633.34 63363.04 64045.93 73049.30 66624.60 77567.85
## [201] 53431.35 31265.75 74780.74 70410.11 37345.24 66107.84 62336.39 39132.64
## [209] 38745.29 65172.22 68519.96 54774.77 76246.96 65461.92 34127.21 35253.98
## [217] 44893.71 59621.02 20856.54 55353.41 67516.07 68737.75 76893.84 59886.58
## [225] 53441.69 41356.31 49942.66 74430.08 58633.63 72707.87 31092.93 74445.18
## [233] 49309.14 56735.14 40183.75 58348.41 72209.99 62060.11 67113.46 24030.06
## [241] 56180.93 62204.93 60372.64 65280.16 34309.24 59610.81 50278.89 43450.11
## [249] 25408.21 71136.49 63883.81 64902.47 66784.81 62784.85 63727.50 61608.23
## [257] 56782.18 64447.77 42042.95 67669.06 54875.95 73347.67 50199.77 50723.67
## [265] 63450.96 56694.12 70547.16 47391.95 62312.23 63100.13 73687.50 52686.47
## [273] 78119.50 57014.84 27086.40 58337.18 50216.01 53049.44 62927.96 32847.53
## [281] 32006.82 48913.07 69285.69 53700.57 52011.00 46339.25 67938.77 66348.95
## [289] 66873.90 72270.88 61610.05 76560.59 62667.51 75687.46 66744.65 67714.82
## [297] 69710.51 66269.49 60843.32 55041.60 73863.25 62378.05 63336.85 42191.61
## [305] 56194.56 61771.90 61383.79 63924.82 23975.35 70179.11 66524.80 41851.38
## [313] 61275.18 60638.38 47160.53 48537.18 53058.91 68614.98 44174.25 67050.16
## [321] 54520.14 54952.42 69476.42 54989.93 29398.61 42861.42 65883.39 65421.39
## [329] 60953.93 58476.57 66636.84 67430.96 57260.41 66359.32 57587.00 63060.55
## [337] 59998.50 74024.61 60550.66 57983.30 52736.33 46653.75 56986.73 55336.18
## [345] 42162.90 39699.13 56394.82 75044.35 53309.61 58996.12 56605.12 62475.99
## [353] 70492.60 43698.53 57737.51 31281.01 45800.48 42362.49 66691.23 56369.74
## [361] 59397.89 66025.11 68211.35 73608.99 61228.96 72325.91 44559.43 73207.15
## [369] 46722.07 45400.50 41417.27 60845.55 60812.77 64267.88 58151.87 52079.18
## [377] 26023.99 62318.38 56216.57 61806.31 51662.24 67080.94 51975.41 28019.09
## [385] 67744.56 66574.00 30487.48 74903.41 19991.72 66050.63 70449.04 64008.55
## [393] 70203.74 27262.51 49544.41 28357.27 66929.03 75524.78 66265.34 55993.68
## [401] 56379.30 31215.88 51015.11 46473.14 55479.62 68713.70 34191.23 51067.54
## [409] 46693.76 19345.36 66225.72 38609.20 37713.23 63764.28 41866.55 57846.68
## [417] 69428.73 60283.98 79332.33 53167.68 64564.07 60803.37 28387.42 58849.77
## [425] 65963.37 75180.20 61270.14 56759.48 46160.63 43870.51 50439.49 28028.74
## [433] 64238.71 65816.38 72684.44 38817.40 63976.44 37212.54 52691.79 65499.93
## [441] 63966.72 52400.88 49111.47 41232.89 52140.04 60641.09 74180.05 51869.87
## [449] 48852.58 59144.02 33951.63 58909.36 49850.52 28679.93 69869.66 48347.64
## [457] 45959.86 70005.51 51512.66 25598.75 49282.87 67240.25 42136.33 62589.84
## [465] 67384.31 25603.93 39616.00 28265.81 63879.72 70592.81 76408.19 55015.08
## [473] 51636.12 29359.20 71296.67 46422.76 52802.00 59243.46 35350.55 59677.64
## [481] 70225.60 65791.17 34191.13 51315.38 62790.96 66291.67 68030.18 43974.49
## [489] 49457.48 33987.27 28210.03 75535.14 49158.50 39809.69 65826.53 61172.07
## [497] 42898.21 68333.01 70232.95 63102.19 51847.26 63580.22 47575.44 39031.89
## [505] 70505.06 62161.26 61068.26 49090.51 62330.75 62053.37 61922.06 49525.37

```

## [513] 53412.32 56681.65 43299.63 47997.75 39131.53 46033.73 65856.74 54787.37  
 ## [521] 69562.46 68447.17 62772.42 78092.95 63649.04 60637.62 27241.11 42760.22  
 ## [529] 59457.52 42907.89 46132.18 46964.11 70377.23 70012.83 56457.01 67279.06  
 ## [537] 54773.99 70783.94 70510.59 64021.55 72042.85 36037.33 67526.92 55121.65  
 ## [545] 63497.62 60879.48 61467.33 70495.64 71222.40 64698.58 32252.38 55316.97  
 ## [553] 47447.89 73474.82 53549.94 58576.12 63373.70 60283.47 37345.34 34886.01  
 ## [561] 67511.86 77988.71 63001.03 61747.98 48467.68 55130.96 79484.80 67307.43  
 ## [569] 27964.60 66431.87 63551.67 40135.06 49101.67 53188.69 49742.83 63394.41  
 ## [577] 64433.99 73884.48 36424.94 28275.48 48098.86 68448.94 66429.84 41768.13  
 ## [585] 57844.96 35684.82 62792.43 51171.23 58847.07 57739.03 64631.22 50337.93  
 ## [593] 67781.31 68863.95 55901.12 64775.10 67686.16 57777.11 46868.53 40926.93  
 ## [601] 22205.74 58920.44 63006.14 24316.61 68348.99 66263.37 63493.60 56984.09  
 ## [609] 51691.55 49911.25 33502.57 65834.97 66176.97 51463.17 41059.64 61428.18  
 ## [617] 51593.46 57518.73 52656.13 52178.98 46239.14 48918.55 65227.79 55002.05  
 ## [625] 52261.73 59448.44 47314.45 55411.06 66504.16 47169.14 70889.68 55358.88  
 ## [633] 56242.70 45522.44 46931.03 55499.69 75805.12 40345.49 33239.20 68033.54  
 ## [641] 38427.66 53185.34 39723.97 43386.07 53922.43 71881.84 47139.21 68877.02  
 ## [649] 65186.58 55424.24 46500.11 58820.16 28495.21 61840.26 37908.29 69805.70  
 ## [657] 60315.19 67323.00 50055.33 43573.66 28186.65 66412.04 63965.16 58342.63  
 ## [665] 33147.19 65899.68 64188.50 58966.22 44078.24 60968.62 65620.25 65496.78  
 ## [673] 52462.04 70582.55 51816.27 23410.75 62729.40 48867.67 50971.73 67990.84  
 ## [681] 43241.19 60082.66 65180.97 67301.39 70701.31 60997.84 60805.93 50711.68  
 ## [689] 41335.84 76480.16 67132.46 52581.16 55195.61 48679.54 63109.74 44490.09  
 ## [697] 57667.99 51824.01 66198.66 73174.19 56593.80 31072.44 66773.83 72553.94  
 ## [705] 43708.88 48453.55 73413.87 58114.30 45465.25 50147.72 61004.51 53898.89  
 ## [713] 59797.64 74623.27 58677.69 62109.80 60583.02 65576.05 73882.91 50468.36  
 ## [721] 51409.45 60514.05 57195.96 52802.58 56570.06 51049.47 66629.61 70185.06  
 ## [729] 43111.41 56435.60 53223.58 57179.91 41521.28 73538.09 63664.32 61757.12  
 ## [737] 71727.51 72203.96 50671.60 47510.42 62466.10 59683.16 41097.17 39799.73  
 ## [745] 76984.21 57877.15 59047.91 72154.68 65704.79 72948.76 73941.91 57887.64  
 ## [753] 62463.70 42838.29 43778.88 71157.05 74159.69 50333.72 33293.78 38641.20  
 ## [761] 49822.78 63891.29 43881.73 48761.14 69758.31 52530.10 58363.12 60575.99  
 ## [769] 48206.04 31523.09 66187.58 69438.04 68016.90 78520.99 31998.72 56909.30  
 ## [777] 61161.29 52340.10 47338.94 50950.24 77143.61 57032.36 48554.45 39552.49  
 ## [785] 36884.23 68783.45 51119.93 44304.13 69718.19 63429.18 65756.36 77871.75  
 ## [793] 47258.59 55984.89 44275.13 25767.16 37605.11 25739.09 60188.38 67682.32  
 ## [801] 44307.18 25371.52 23942.61 50666.50 50356.06 63936.50 69874.18 50038.65  
 ## [809] 67866.95 54645.20 46780.09 67432.49 73392.28 47682.28 56735.83 51013.37  
 ## [817] 69481.85 67033.34 68717.00 59340.99 47968.32 48758.92 61230.03 54755.71  
 ## [825] 54324.73 52177.40 51163.14 66861.67 63107.88 49206.40 55942.04 33601.84  
 ## [833] 48867.36 56683.32 38260.89 54106.21 71055.22 46403.18 61690.93 26130.93  
 ## [841] 58638.75 47357.39 50086.17 51772.58 47638.30 38987.42 51363.16 35764.49  
 ## [849] 62939.50 58776.67 59106.12 50457.01 54251.78 51920.49 70324.80 52416.18  
 ## [857] 66217.31 60938.73 40243.82 60151.77 45945.88 63430.33 65882.81 64410.80  
 ## [865] 55677.12 75560.65 61067.58 72330.57 32549.95 51257.26 77220.42 52520.75  
 ## [873] 59422.47 22456.04 58443.99 50820.74 67575.12 66522.79 34903.67 43073.78  
 ## [881] 57594.70 66027.31 53012.94 61117.50 52563.22 65773.49 50506.44 66262.59  
 ## [889] 35521.88 62430.55 49597.08 42078.89 46197.59 49957.00 24078.93 53647.81  
 ## [897] 61039.13 46974.15 53042.51 48826.14 58287.86 21773.22 52252.91 27073.27  
 ## [905] 50628.31 36913.51 61009.10 53041.77 40182.84 59419.78 58235.21 68324.48  
 ## [913] 69646.35 54045.39 57806.03 53336.76 50491.45 71455.62 43241.88 58953.01  
 ## [921] 36834.04 66345.10 38645.40 60803.00 33553.90 63071.34 46737.34 55368.67  
 ## [929] 68305.91 39211.49 65956.71 40159.20 40478.83 40468.53 66980.27 34942.26  
 ## [937] 48335.20 42251.59 57330.43 75769.82 51812.71 75265.96 69868.48 72802.42

```
## [945] 39193.45 56129.89 58996.56 41547.62 59240.24 56725.47 55764.43 64235.51
## [953] 39939.39 63319.99 54725.87 69775.75 57545.56 47051.02 51600.47 68357.96
## [961] 35349.26 69784.85 50760.23 34418.09 20592.99 63528.80 44217.68 47929.83
## [969] 46024.29 51900.03 72188.90 56974.51 25682.65 41884.64 72196.29 54429.17
## [977] 58037.66 64011.26 59967.19 43155.19 51501.38 55187.85 33813.08 36497.22
## [985] 66193.81 66200.96 63126.96 71384.57 67782.17 42415.72 41920.79 29875.80
```

## 5. Exploratory Data Analysis

### a) Univariate Analysis

We shall use the Graphical method to do the Univariate Analysis

#### i). Measures of Central Tendency

```
# Mean Age
```

```
mean(advertising$Age)
```

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

We can see the average age is around 36 years

```
# Median Age
```

```
median(advertising$Age)
```

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

The median age is 35 years

```
# Average Daily spent on Site
```

```
mean(advertising$Daily.Time.Spent.on.Site)
```

```
## [1] 65.0002
```

```
mean(advertising$Daily.Internet.Usage)
```

```
## [1] 180.0001
```

```
# Mode of Area Income
```

```
mode <- function(v) {
  uniqv <- unique(v)
  uniqv[which.max(tabulate(match(v, uniqv)))]
}
```

```
mode(advertising$Area.Income)
```



```
## [1] 61833.9
```

The modal Area Income is 61833

## ii). Measures of Dispersion

```
# Checking the range of the Numeric columns of the Area Income  
range(advertising$Area.Income)
```

```
## [1] 13996.5 79484.8
```

```
# Checking the maximum age  
max(advertising$Age)
```

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

```
# Checking the minimum age  
min(advertising$Age)
```

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

```
# Checking the Variance of the Area Income  
var(advertising$Area.Income)
```

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

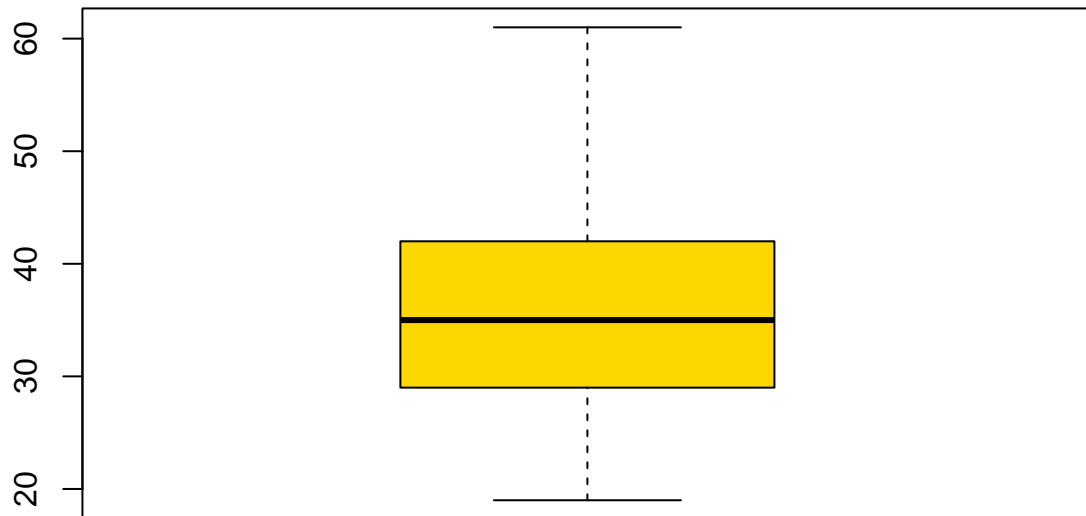
```
# Checking the Standard Deviation of the Area Income  
sd(advertising$Area.Income)
```

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

## iii) Univariate Graphical

We shall be using the box plot to display our age data

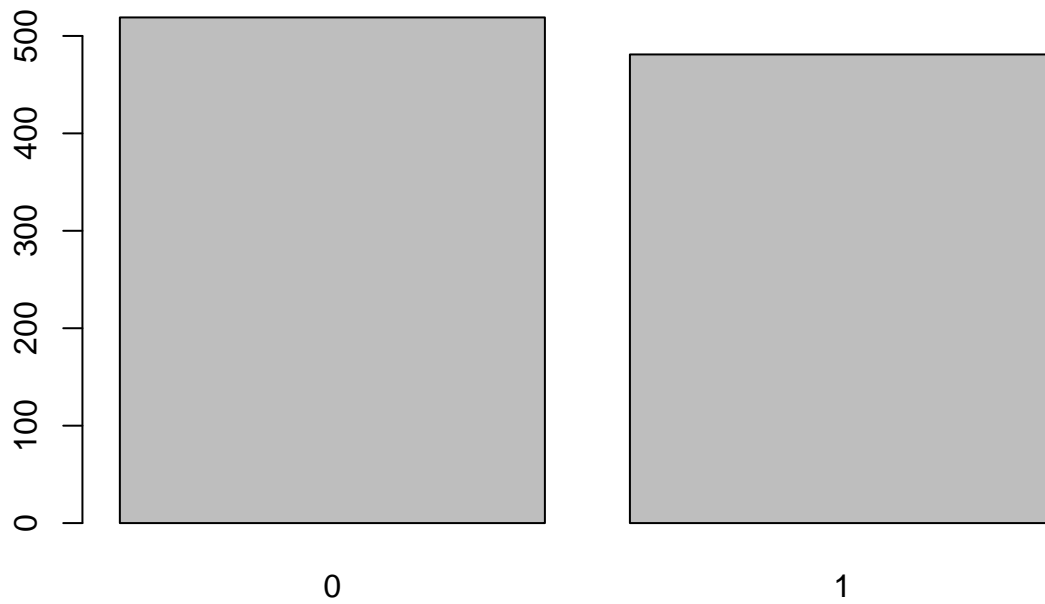
```
# Boxplot for age  
boxplot(advertising$Age,col="gold")
```



We shall use bar graph to show the gender(Male) feature of our data

```
gender_frequency <- table(advertising$Male)
barplot(gender_frequency, main = "Bargraph for Gender")
```

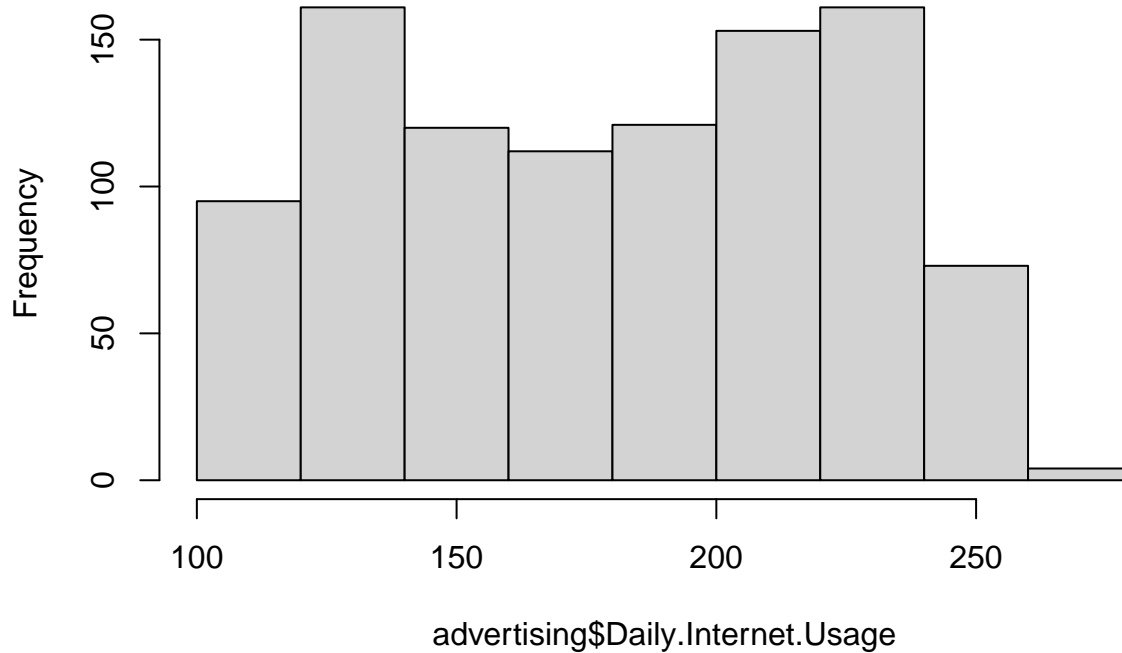
### Bargraph for Gender



We have slightly more females than males as clearly shown on the bargraph. The 0 implies Female while 1 implies male

```
# Histogram to show the Daily Internet Usage  
hist(advertising$Daily.Internet.Usage, main = "Histogram for Daily Internet Usage")
```

## Histogram for Daily Internet Usage



### b). Bivariate Analysis

#### i). Covariance

```
cov(numeric_col)
```

```
##           Daily.Time.Spent.on.Site      Age  Area.Income
## Daily.Time.Spent.on.Site      251.3370949 -4.617415e+01  6.613081e+04
## Age                          -46.1741459  7.718611e+01 -2.152093e+04
## Area.Income                  66130.8109082 -2.152093e+04  1.799524e+08
## Daily.Internet.Usage          360.9918827 -1.416348e+02  1.987625e+05
## Male                         -0.1501864  -9.242142e-02  8.867509e+00
## Clicked.on.Ad                -5.9331431  2.164665e+00 -3.195989e+03
##           Daily.Internet.Usage      Male Clicked.on.Ad
## Daily.Time.Spent.on.Site      3.609919e+02 -0.15018639 -5.933143e+00
## Age                          -1.416348e+02 -0.09242142  2.164665e+00
## Area.Income                  1.987625e+05  8.86750903 -3.195989e+03
## Daily.Internet.Usage          1.927415e+03  0.61476667 -1.727409e+01
## Male                          6.147667e-01  0.24988889 -9.509510e-03
## Clicked.on.Ad                -1.727409e+01 -0.00950951  2.502503e-01
```

## ii). Correlation

```
cor(numeric_col)
```

```
##           Daily.Time.Spent.on.Site      Age  Area.Income
## Daily.Time.Spent.on.Site      1.00000000 -0.33151334  0.310954413
## Age                          -0.33151334  1.00000000 -0.182604955
## Area.Income                  0.31095441 -0.18260496  1.000000000
## Daily.Internet.Usage         0.51865848 -0.36720856  0.337495533
## Male                        -0.01895085 -0.02104406  0.001322359
## Clicked.on.Ad               -0.74811656  0.49253127 -0.476254628
##           Daily.Internet.Usage      Male Clicked.on.Ad
## Daily.Time.Spent.on.Site      0.51865848 -0.018950855 -0.74811656
## Age                          -0.36720856 -0.021044064  0.49253127
## Area.Income                  0.33749553  0.001322359 -0.47625463
## Daily.Internet.Usage         1.00000000  0.028012326 -0.78653918
## Male                        0.02801233  1.000000000 -0.03802747
## Clicked.on.Ad              -0.78653918 -0.038027466  1.00000000
```

We shall now visualize the correlation matrix

```
# Loading the corrplot
```

```
library(corrplot)
```

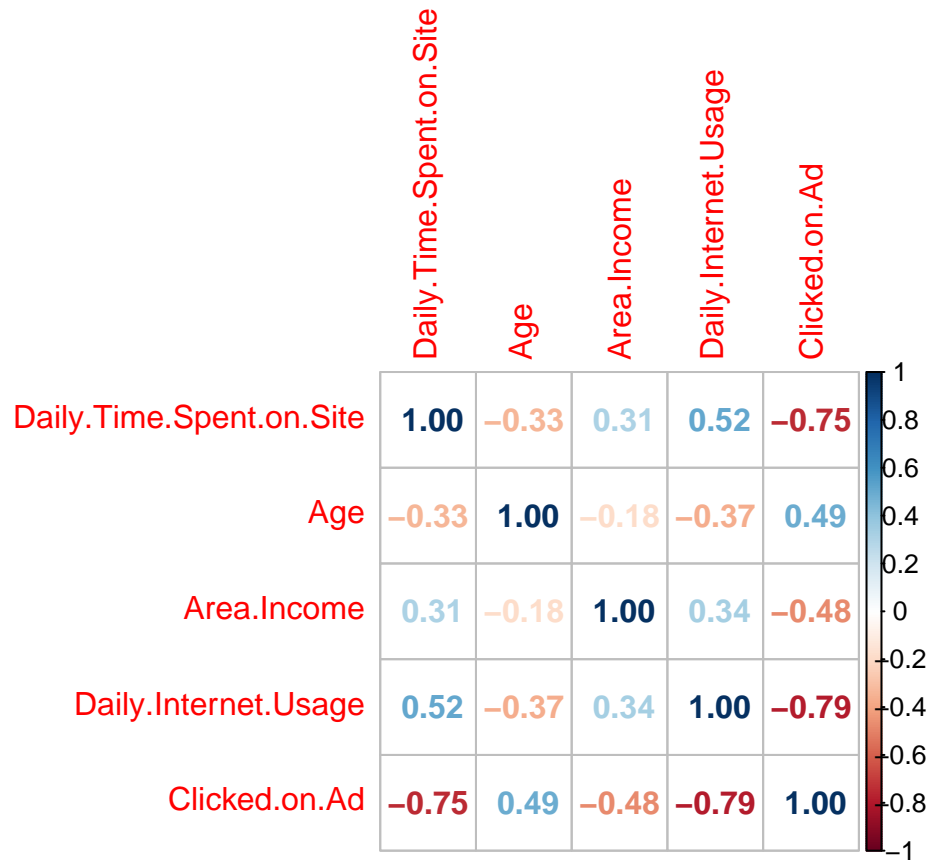
```
## corrplot 0.92 loaded
```

```
# Dropping the Male from numeric column
```

```
num_col <- subset(numeric_col, select = -c(Male))
```

```
corr_matrix <- cor(num_col)
```

```
corrplot(corr_matrix, method='number')
```



From the above Correlation Matrix, we can see there is strong inverse correlation between Daily Internet Usage and Clicked on Ad.

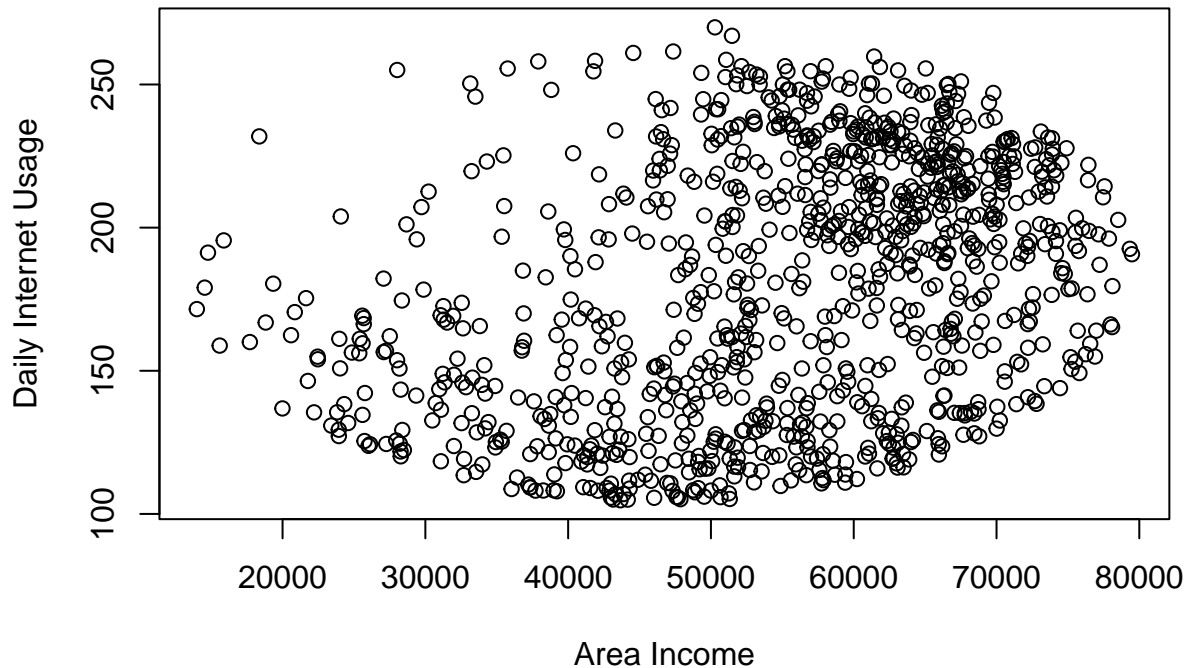
### iii). Scatterplot

We shall use the scatterplot to show the relationship between Area Income and Daily Internet Usage

```
area_income <- advertising$Area.Income
daily_usage <- advertising$Daily.Internet.Usage

plot(area_income,daily_usage,xlab="Area Income", ylab="Daily Internet Usage",main="Scatterplot for Area
```

### Scatterplot for Area Income and Daily Internet usage



## 6. Conclusion and Recommendation

From our Exploratory Data Analysis, we can easily see the relation between the various features of our dataset. We have seen that older people are more likely to click on the ads this is because there is a positive correlation between age and click on ad. Hence the entrepreneur should focus more on older people. Also we have seen that with increase on Daily Internet Usage, there is a huge reduction click on ads, this implies the entrepreneur should focus more on individual with low daily internet usage.

## 7. Follow up questions

### a) Did we have the right data?

Yes, the dataset available for this analysis was relevant to the research problem.

### b) Do we need other data to answer the research question?

No, the dataset provided had relevant information for the research question.