

Introduction

In this project, exploratory data analysis (EDA) was conducted on large, structured and semi-structured datasets. With the help of EDA, important analysis about data and its relation with various parameters was studied. The different types of datasets analyses were twitter data set, New York Times ad click dataset, Real Direct dataset and live twitter data.

Problem 1: Data Acquisition

For this problem, the data source used was Twitter data using Twitter Search API, collected over a period of ten days. Twitter Streaming API is used in Problem 5. TwitteR and ROAuth were the packages that were used to carry out twitter data collection. TwitteR provided functions like setup_twitter_oauth which creates a connection to Twitter's Search API. The resulting list of retrieved tweets is stored in a data frame by using the function twListToDF(). Over a period of ten days, I created different data frames that stored that day's tweets. At the end, I combined all these data frames into one unit and stored this collective data as both cvs file and Rdata by using write.csv and save commands respectively. tf is the local data frame containing collected tweets. These are the column variables it has.

```
> colnames(tf)
[1] "text"           "favorited"      "favoriteCount"  "replyToSN"      "created"         "truncated"      "replyToSID"
[8] "id"            "replyToUID"     "statusSource"   "screenName"     "retweetCount"    "isRetweet"      "retweeted"
[15] "longitude"      "latitude"
> |
```

	text	favorited	favoriteCount	replyToSN	created	truncated	replyToSID	id	replyToUID	statusSource
1	RT @AndrewGuapo_: "Young Metro doesn't trust Don...	FALSE	0	NA	2016-02-25 23:59:59	FALSE	NA	703006559421513728	NA	<a href="
2	RT @steph93065: Trump doesn't need to change a thi...	FALSE	0	NA	2016-02-25 23:59:59	FALSE	NA	703006559329230849	NA	<a href="
3	@Thidran @Enjenovix @duckspeakeasy @IndJournal O...	FALSE	0	Thidran	2016-02-25 23:59:59	FALSE	703005778106568704	703006559169847296	176874550	<a href="
4	Debate another chance for Rubio, Cruz to try to slo...	FALSE	1	NA	2016-02-25 23:59:59	FALSE	NA	703006558309842944	NA	<a href="
5	RT @GregJaye: Hey all #SECprimary voters: Think abo...	FALSE	0	NA	2016-02-25 23:59:59	FALSE	NA	703006558083362816	NA	<a href="
6	RT @ReneSaldanajr: All of a sudden @realDonaldTrump...	FALSE	0	NA	2016-02-25 23:59:59	FALSE	NA	70300655752180737	NA	<a href="
7	Bernie's ego rivals Trump's. No doubt about it! #imWi...	FALSE	1	NA	2016-02-25 23:59:59	FALSE	NA	703006557458403328	NA	<a href="
8	Yes. When Trump mentions a candidate, the candidat...	FALSE	1	NA	2016-02-25 23:59:59	FALSE	NA	703006557328580608	NA	<a href="
9	Trump is the GOP's 2019>s Frankenstein monster. ...	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	70300655944255488	NA	<a href="
10	RT @JoeNBC: Trump Dominates South. Most distressl...	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	703006555512430595	NA	<a href="
11	#What Trump, Clinton and the 2016 Election Mean for...	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	703006554849554434	NA	<a href="
12	RT @CommonWhiteGirl: Donald Trump's cousin dragg...	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	703006554786766848	NA	<a href="
13	@andersoncooper just absolutely destroyed a Trump ...	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	703006553633239040	NA	<a href="
14	Former Mexican president says he will not pay for Do...	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	703006553138438144	NA	<a href="
15	"I'm votin' for Donald Trumpe" are fightin' words.ced	FALSE	1	NA	2016-02-25 23:59:58	FALSE	NA	703006552932745216	NA	<a href="

Table 1

text	favorited	favoriteCount	replyToSN	created	truncated	replyToSID	id	replyToUID	statusSource	screenName	retweetCount	isRetweet	retweeted
RT @AndrewGuapo_: "Young Metro doesn't trust Donald Trump" @Metroboon #Houston = Ssk, https://t.co/NAqg79HfLuz	FALSE	0	NA	2016-02-25 23:59:59	FALSE	NA	703006559421513728	NA	Twitter for Android	Michelle	886	TRUE	
RT @steph93065: Trump doesn't need to change a thing. @Hence The party is the people. NOT party leaders & donors. @jakestapper	FALSE	0	NA	2016-02-25 23:59:59	FALSE	NA	703006559329230849	NA	Twitter for Android	MARILYNLEVINeo1	146	TRUE	
@Thidran @Enjenovix @duckspeakeasy @IndJournal Only a few, but I stand against the Trump Con.	FALSE	0	Thidran	2016-02-25 23:59:59	FALSE	703005778106568704	703006559169847296	176874550	TweetDeck	TheliciaWilson	0	FALSE	
Debate another chance for Rubio, Cruz to try to slow Trump: https://t.co/WEF1dU8f #xdata! https://t.co/TqYV92wP	FALSE	1	NA	2016-02-25 23:59:59	FALSE	NA	703006558309842944	NA	TweetDeck	FOGAAtlanta	1	FALSE	
RT @GregJaye: Hey all #SECprimary voters: Think about this: The Forcast	FALSE	0	NA	2016-02-25 23:59:59	FALSE	NA	703006558083362816	NA	Twitter Web Client	lovetheanation	1	TRUE	
RT @ReneSaldanajr: All of a sudden @realDonaldTrump's "Build that Wall" takes on a whole new meaning: https://t.co/hPghNNWZ	FALSE	0	NA	2016-02-25 23:59:59	FALSE	NA	70300655752180737	NA	Twitter Web Client	KY_LAN	8	TRUE	
Bernie's ego rivals Trump's. No doubt about it! #imWinning @ProgressivePartyTest https://t.co/helULM0d	FALSE	1	NA	2016-02-25 23:59:59	FALSE	NA	703006557328580608	NA	Twitter Web Client	BJLiane	0	FALSE	
Yes. When Trump mentions a candidate, the candidate's response should be, "What the heck does that mean, exactly?" https://t.co/RNPYOCaw	FALSE	1	NA	2016-02-25 23:59:59	FALSE	NA	703006554849554434	NA	TweetDeck	SacksonTBO	0	FALSE	
Trump is the GOP's 2019>s Frankenstein monster. Now he's 7% strong enough to destroy the party. https://t.co/iecl8c1tQQ	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	70300655944255488	NA	Twitter Web Client	bobychmbr1	0	FALSE	
RT @JoeNBC: Trump Dominates South. Most distressing for conservatives is fact voters know he's most liberal & support him anyway https://t.co/yf	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	7030065512430595	NA	Twitter for iPhone	GaraganyP	98	TRUE	
#What Trump, Clinton and the 2016 Election Mean for! Stocks Hit! https://t.co/gwHJH6SE	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	703006554849554434	NA	WordPress.com	Mariahodgson5	0	FALSE	
RT @CommonWhiteGirl: Donald Trump's cousin dragged him through the mud in his obituary https://t.co/0fryhMMW	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	703006554786766848	NA	Twitter for iPhone	dafneid	5940	TRUE	
@andersoncooper just absolutely destroyed a Trump Supporter/Republican strategist in Ssk & Ssk & Ssk & Ssk. have it all Anderson, have it all	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	703006553633239040	NA	Twitter for iPhone	amodul	0	FALSE	
Former Mexican president says he will not pay for Donald Trump's "bigly" wall	FALSE	0	NA	2016-02-25 23:59:58	FALSE	NA	703006553138438144	NA	Twitter for iOS	devcode88	0	FALSE	
"I'm votin' for Donald Trump" are fightin' words! Ssk & Ssk	FALSE	1	NA	2016-02-25 23:59:58	FALSE	NA	703006552932745216	NA	Twitter for iPhone	DakotaLinda91	0	FALSE	
RT @PisaPierluigi: Bernie supporters Vs. Trump supporters https://t.co/GVND8qMcZ	FALSE	0	NA	2016-02-25 23:59:57	FALSE	NA	70300655324710400	NA	Twitter for Android	BradleyWest97	696	TRUE	
RT @Zacharycohen: Hilary can't beat trump. Bernie can. Fast.	FALSE	0	NA	2016-02-25 23:59:57	FALSE	NA	7030065535164180	NA	Twitter for iPhone	DrSveinud	2	TRUE	
@mowes So, remember that convo we had a while back about TrumpCruz, Cruz/Trump getting the nomination? What was that?	FALSE	0	increws	2016-02-25 23:59:57	FALSE	703003958382044048	703006550825444160	14489996	Twitter Web Client	LNewber	0	FALSE	
@knucklets @McGrahamClegg saw this thought of you... https://t.co/E2chd27zy	FALSE	2	knucklets	2016-02-25 23:59:57	FALSE	70284384913637120	70300655036507904	2894973208	Echofon	Chumber	0	FALSE	
RT @lennardum: It would be SUPER cool to have a president who rates women's looks on a scale of 1-10, right guys? https://t.co/0Xs4gHJL8	FALSE	0	NA	2016-02-25 23:59:57	FALSE	NA	70300654992550400	NA	Twitter for iOS	KristaBerzola	1049	TRUE	
RT @aubrynthorne: 21 Questions for Donald #Trump https://t.co/eyv97T6a	FALSE	0	NA	2016-02-25 23:59:57	FALSE	NA	703006549745274880	NA	Twitter for iPhone	Countrygown35	5	TRUE	
RT @DefendingHUSA: TODAY'S POLLS VA - TRUMP +14% GA - TRUMP +28%	FALSE	0	NA	2016-02-25 23:59:57	FALSE	NA	70300654942137344	NA	Mobile Web (Mobi)	JuliaSoufan	190	TRUE	

This is the cvs file with tweets, stored on disk.

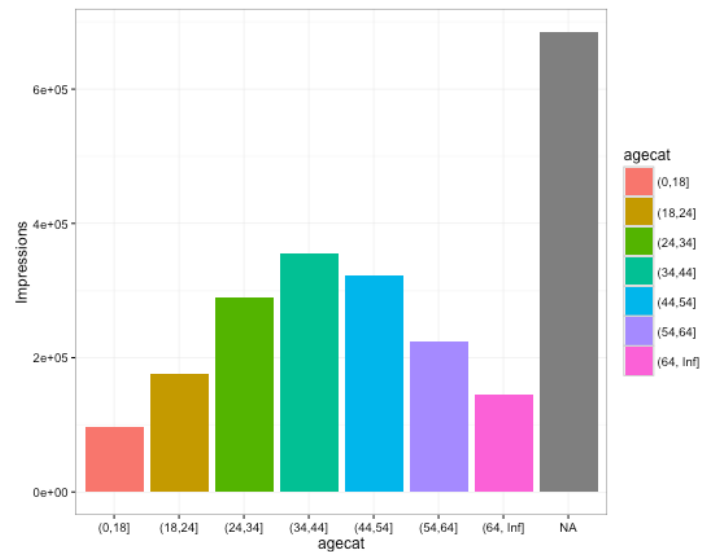
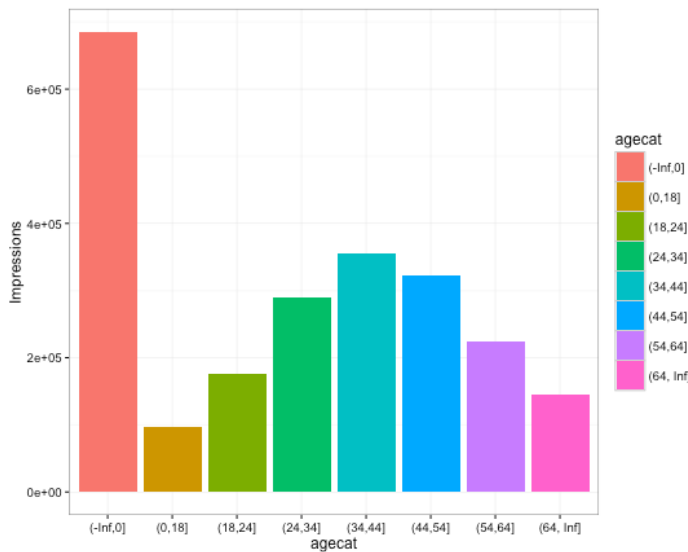
Problem 2: Simple EDA

In this problem, we were provided with the New York Times dataset which had the information about the ad clicking behavior of readers with attributes such as age, gender, impressions, number of clicks, signed in or not.

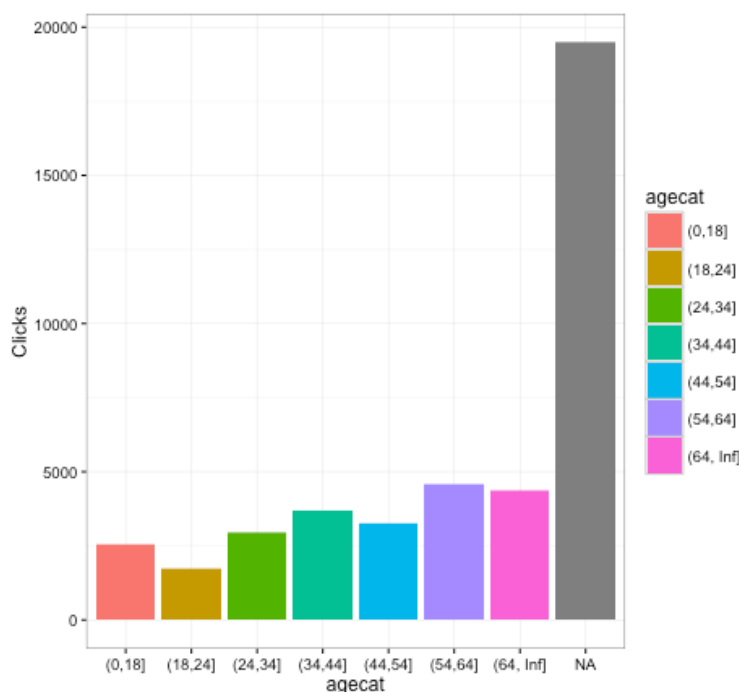
First part of the problem is to categorize the age variable into intervals of age.

```
data1$agecat <- cut(data1$Age, c(-Inf, 0, 18, 24, 34, 44, 54, 64, Inf))
```

Next, plotting impressions and click-through-rates for this interval of age. Here, upon noticing a high value for $(-\text{Inf}, 0]$, I realized that these users were not logged in. So setting the signed in value for these as NA and generated the bar plot again.

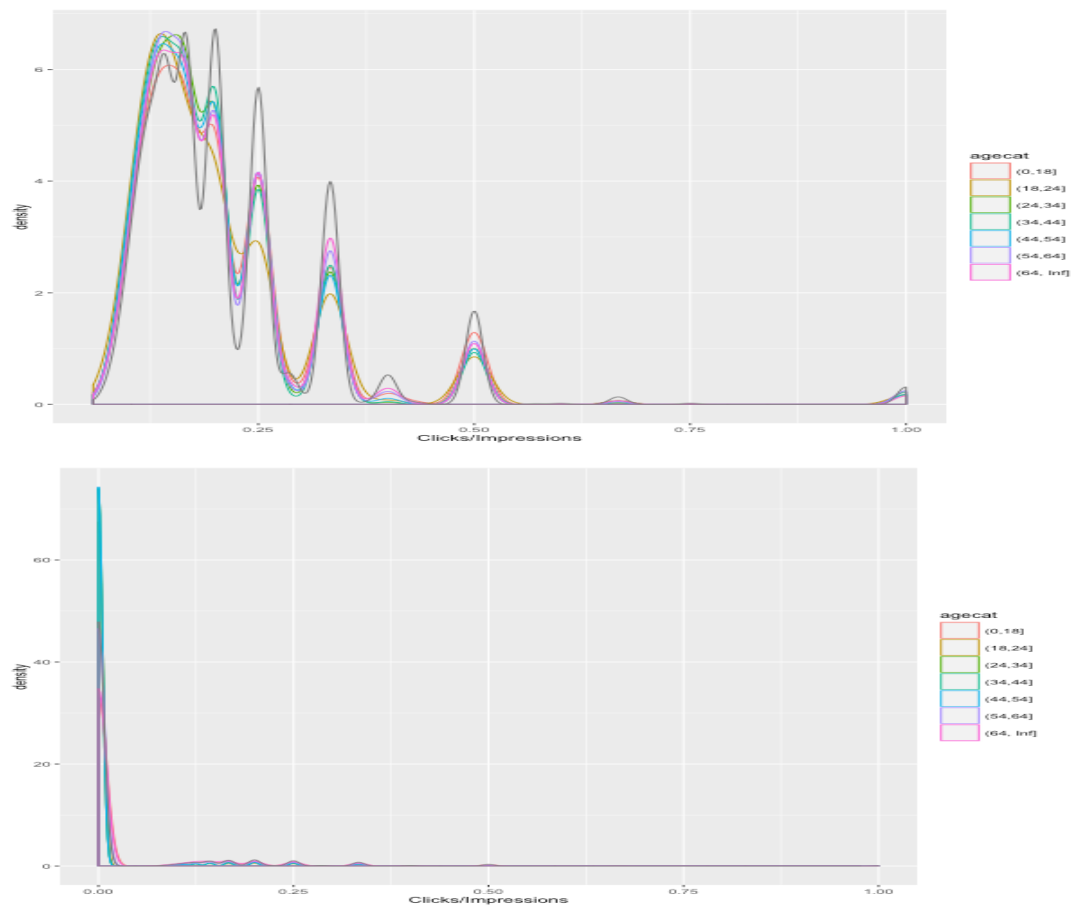


As is seen from the figure, the bar plot shows the number of impressions for the age intervals which is highest for age group $(34, 44]$.

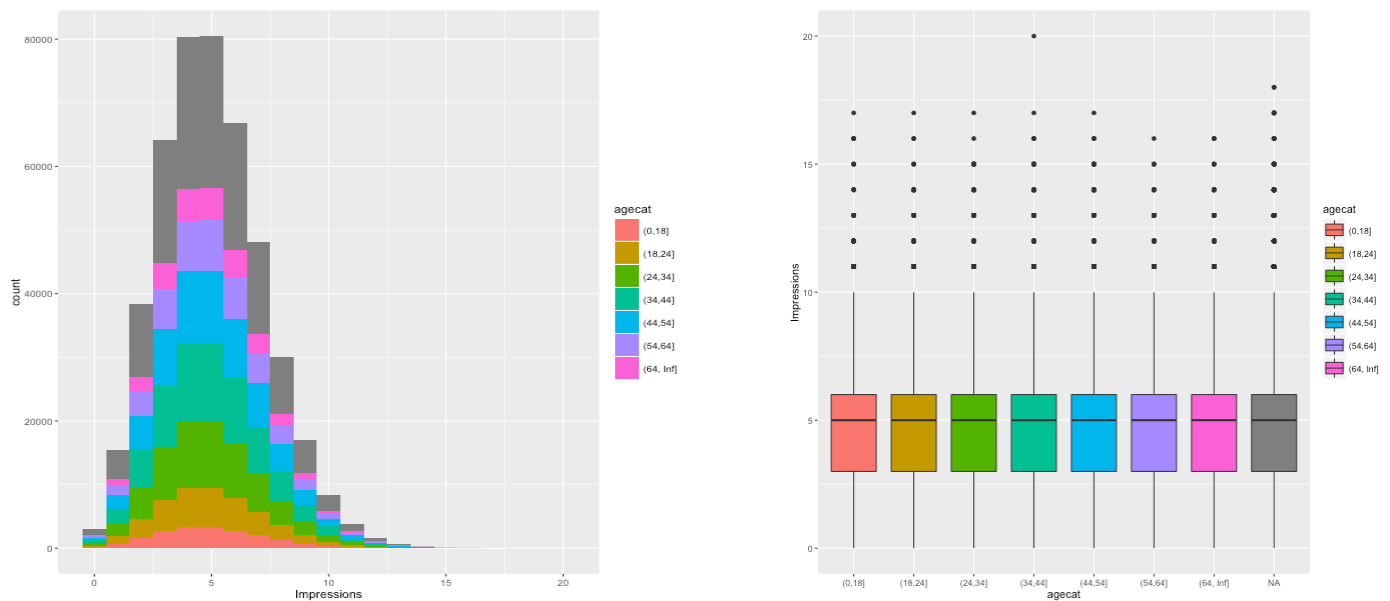


This figure shows the number of clicks for each age group. It can be inferred that people from age group $(54, 64]$ appeared to click on the ads the most, followed by $(64, \text{Inf}]$ (not counting the unsigned users). This shows that elderly people are more interested in (or easily distracted by) online advertisements. Similarly age group $(18, 24]$ seemed to click on ads the least which is probably because most of the people in this age group are students or young working professionals who might not spend much time clicking on ads.

Following shows the density distribution of CTR with respect to age. First figure is for Clicks>0 and second figure is for Impressions>0.



Following are a histogram and box plot that show the relation between CTR and age group.



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```

ggplot(data=data1, aes(x=agecat, y=Clicks, fill=agecat)) + geom_bar(stat="identity") + theme_bw()
data1$agecat[data1$Signed_In == 0] = NA
data1$Gender[data1$Signed_In == 0] = NA
summary(data1)
ggplot(data=data1, aes(x=agecat, y=Clicks, fill=agecat)) + geom_bar(stat="identity") + theme_bw()
data2 = na.omit(subset(data1, Impressions>0)) %>% group_by(agecat) %>% summarise(Impressions = sum(Impressions), Clicks = sum(Clicks))
ggplot(data=data2, aes(x=agecat, y=Clicks/Impressions, fill=agecat)) + geom_bar(stat="identity") + theme_bw()
ggplot(subset(data1, Impressions>0), aes(x=Clicks/Impressions, colour=agecat)) + geom_density()
ggplot(subset(data1, Clicks>0), aes(x=Clicks/Impressions, colour=agecat)) + geom_density()
#First graph histogram impressions with age
ggplot(data1, aes(x=Impressions, fill=agecat)) + geom_histogram(binwidth=1)
#Second graph boxplot
ggplot(data1, aes(x=agecat, y=Impressions, fill=agecat)) + geom_boxplot()

```

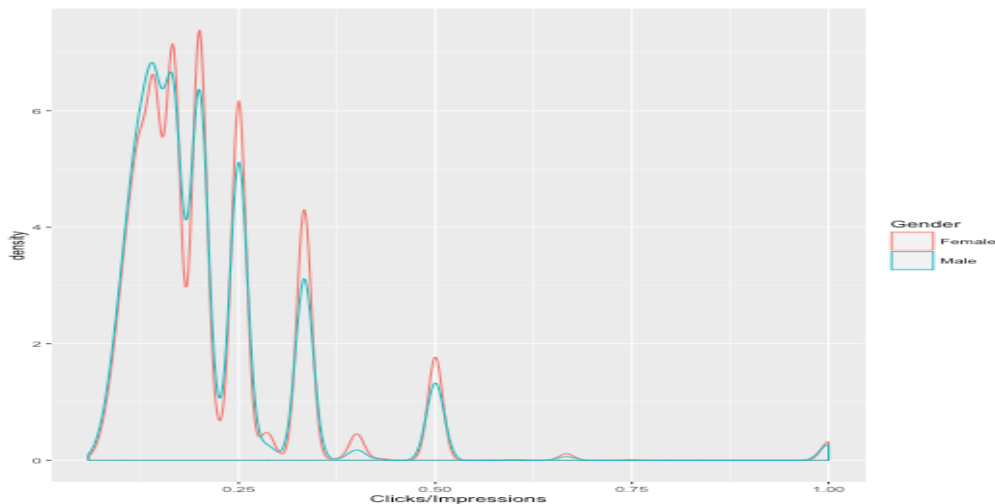
Next, a new variable “clickcat” was defined to differentiate between users based on their click behavior which would hold three values: No Impressions, Only Impressions(No Click) and Clicks.

```

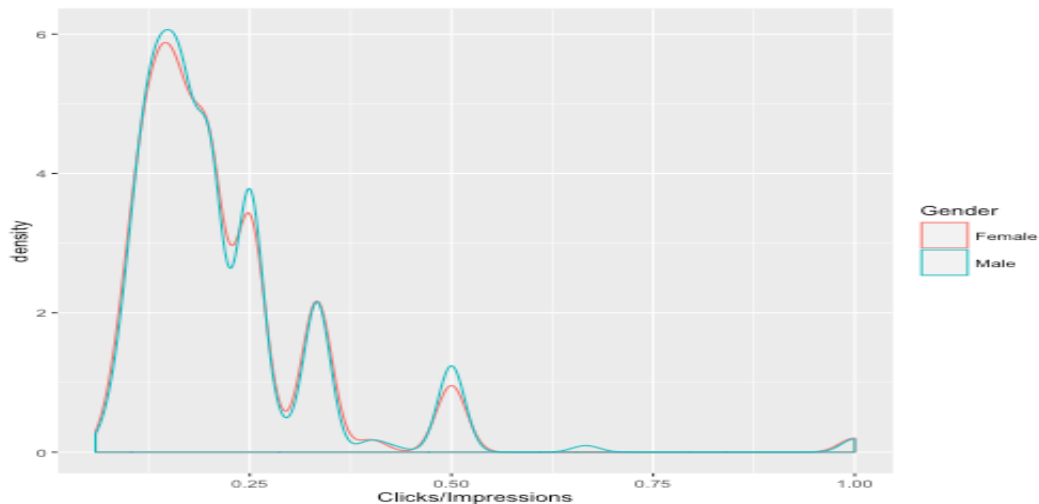
data1$clickcat[data1$Impressions==0] <- "NoImps"
data1$clickcat[data1$Impressions >0] <- "Imps"
data1$clickcat[data1$Clicks >0] <- "Clicks"

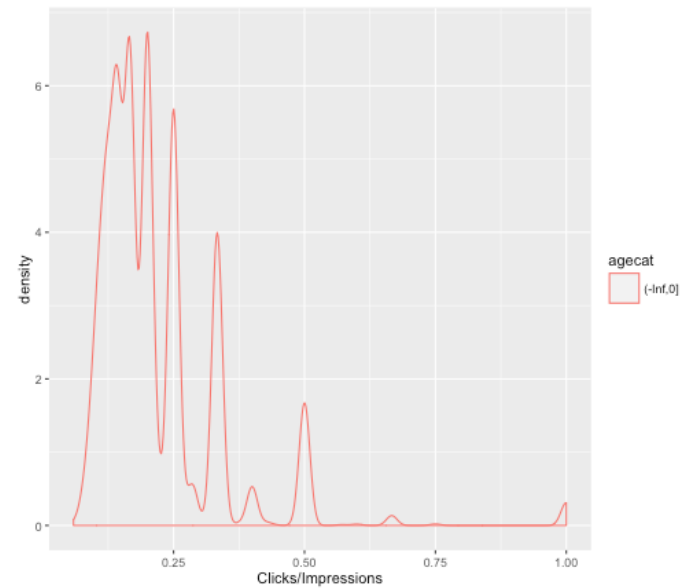
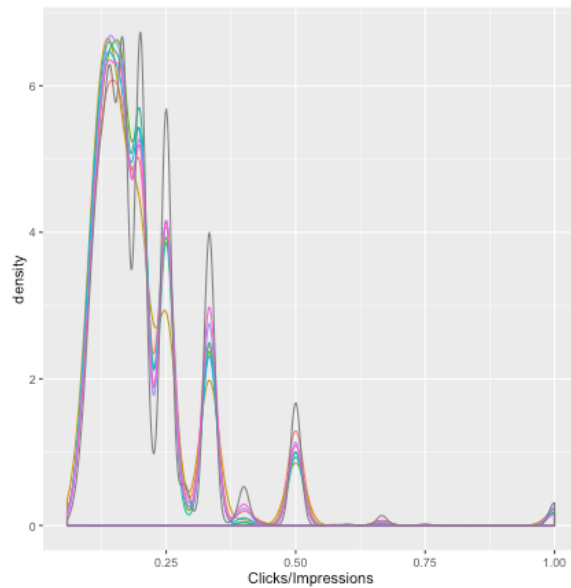
```

Next part of the problem was to explore the data and make visual and quantitative comparisons across user segments. Here are some observations. This following graph shows a normal density distribution graph for click through rate patterns of males and females of all age groups

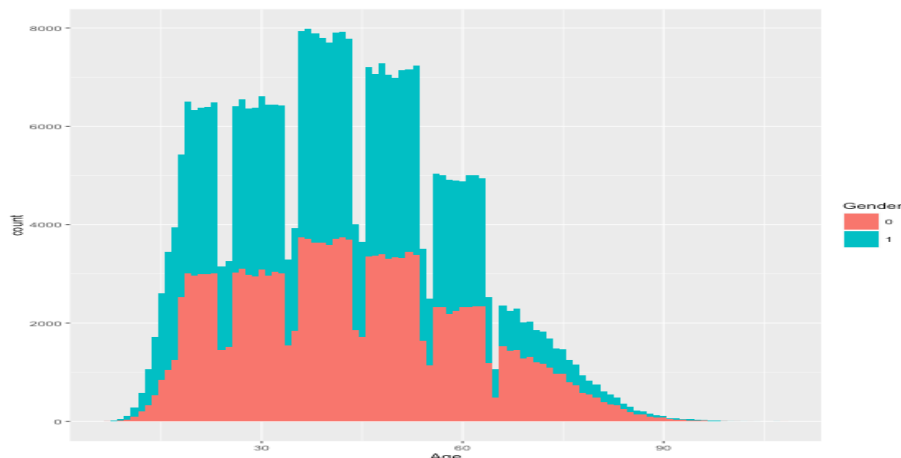


The following graph compares the Click Through Rate patterns of less than 18 year old male & females.





In the above graphs, figure 1 shows the density distribution for logged in users and figure 2 shows density distribution for not logged in users.



Here, there is a normal distribution of age with males usually twice the number of females except for >65 where difference starts to reduce.

Now looking at some statistics.

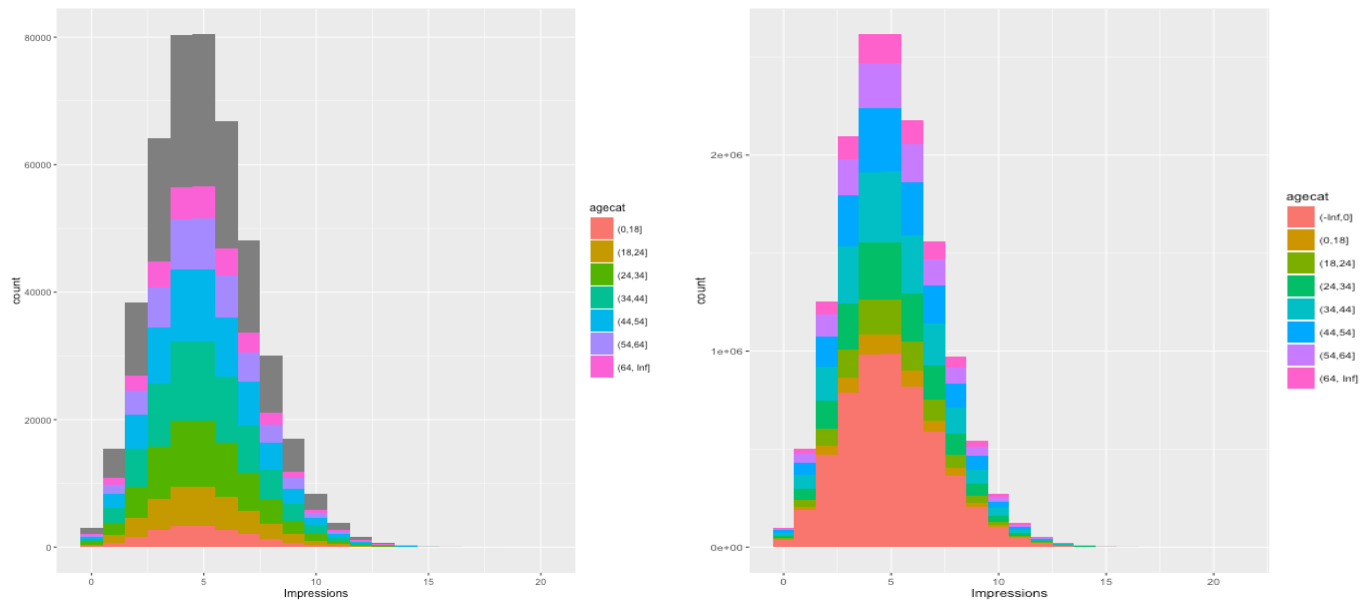
```
> s <- function(x){c(length = length(x),min = min(x),mean = mean(x), max = max(x), median = median(x))}
> summaryBy(Age~agecat, data =data1, FUN=s)
```

	agecat	Age.length	Age.min	Age.mean	Age.max	Age.median
1	(-Inf,0]	137106	0	0.00000	0	0
2	(0,18]	19252	7	16.03350	18	16
3	(18,24]	35270	19	21.26904	24	21
4	(24,34]	58174	25	29.50335	34	30
5	(34,44]	70860	35	39.49468	44	39
6	(44,54]	64288	45	49.49258	54	49
7	(54,64]	44738	55	59.49819	64	60
8	(64, Inf]	28753	65	72.98870	108	72

```
> summaryBy(Gender+Signed_In+Impressions+Clicks~agecat,data =data1)
```

	agecat	Gender.mean	Signed_In.mean	Impressions.mean	Clicks.mean
1	(-Inf,0]	0.0000000	0	4.999657	0.14207985
2	(0,18]	0.6421151	1	4.998961	0.13105132
3	(18,24]	0.5338531	1	5.006635	0.04845478
4	(24,34]	0.5321621	1	4.993829	0.05048647
5	(34,44]	0.5316963	1	5.021507	0.05167937
6	(44,54]	0.5289790	1	5.010406	0.05027377
7	(54,64]	0.5361885	1	5.022308	0.10183736
8	(64, Inf]	0.3632664	1	5.012347	0.15128856

Last part of the problem, was to extend the findings and calculation analysis from one day to a month. As we can see from the graphs below, the data is normally distributed over the month too.



The following graph shows the density distribution over 30 days. It gives a click through rate pattern of users with respect to age over a period of 30 days.

