

# DATA INTENSIVE COMPUTING

## *Simple EDA on New York Times Data*



## Problem 2A: Simple EDA on single day data

### Step 1: Reading the data

The first step in the process of EDA is to read the data which in our case is stored in the csv file.

```
12 # Reading CSV File
13 data_nyt_day_1 = read.csv(url("http://stat.columbia.edu/~rachel/datasets/nyt1.csv"))
14 head(data_nyt_day_1)
```

The data looks something like this

	Age	Gender	Impressions	Clicks	Signed_In
1	36	0	3	0	1
2	73	1	3	0	1
3	30	0	3	0	1
4	49	1	3	0	1
5	47	1	11	0	1
6	47	0	11	1	1

### Step 2: Cleaning the data

In this step we will perform some cleaning operation to make the data suitable for EDA

```
16 # Creating Age Groups
17 data_nyt_day_1$Age_Group = cut(data_nyt_day_1$Age, c(-Inf, 17, 24, 34, 44, 54, 64, Inf),
18                               c("<18", "18-24", "25-34", "35-44", "45-54", "55-64", "65+"))
19 head(data_nyt_day_1)
20
21 # Changing the labels for Gender from 1 and 0 to Male and Female to make them more intuitive
22 data_nyt_day_1$Gender = factor(data_nyt_day_1$Gender, levels=c(0,1), labels = c("female", "male"))
23 head(data_nyt_day_1)
24
25 # Creating Summary of the data
26 summary(data_nyt_day_1)

36 # Changing the labels for impressions to make them more intuitive
37 data_nyt_day_1$Has_Impressions <- cut(data_nyt_day_1$Impressions, c(-Inf, 0, Inf), c("No", "Yes"))
38 summaryBy(Clicks~Has_Impressions, data=data_nyt_day_1, FUN=siterange)

47 # Changing the labels for User Segment to make them more intuitive
48 data_nyt_day_1$User_Segment[data_nyt_day_1$Impressions==0] <- "No_Impressions"
49 data_nyt_day_1$User_Segment[data_nyt_day_1$Impressions>0] <- "Impressions"
50 data_nyt_day_1$User_Segment[data_nyt_day_1$Clicks>0] <- "Clicks"
```

After following very simple cleaning process as described above we are able to get a beautiful data

### Grouping based on Age Groups

	Age	Gender	Impressions	Clicks	Signed_In	Age_Group
1	36	0	3	0	1	35-44
2	73	1	3	0	1	65+
3	30	0	3	0	1	25-34
4	49	1	3	0	1	45-54
5	47	1	11	0	1	45-54
6	47	0	11	1	1	45-54

### Changing labels for Gender

	Age	Gender	Impressions	Clicks	Signed_In	Age_Group
1	36	female	3	0	1	35-44
2	73	male	3	0	1	65+
3	30	female	3	0	1	25-34
4	49	male	3	0	1	45-54
5	47	male	11	0	1	45-54
6	47	female	11	1	1	45-54

### Summary Report

```
> summary(data_nyt_day_1)
```

Age	Gender	Impressions	Clicks	Signed_In	Age_Group
Min. : 0.00	female:290176	Min. : 0.000	Min. :0.00000	Min. :0.0000	<18 :150934
1st Qu.: 0.00	male :168265	1st Qu.: 3.000	1st Qu.:0.00000	1st Qu.:0.0000	18-24: 40694
Median : 31.00		Median : 5.000	Median :0.00000	Median :1.0000	25-34: 58174
Mean : 29.48		Mean : 5.007	Mean :0.09259	Mean :0.7009	35-44: 70860
3rd Qu.: 48.00		3rd Qu.: 6.000	3rd Qu.:0.00000	3rd Qu.:1.0000	45-54: 64288
Max. :108.00		Max. :20.000	Max. :4.00000	Max. :1.0000	55-64: 44738
					65+ : 28753

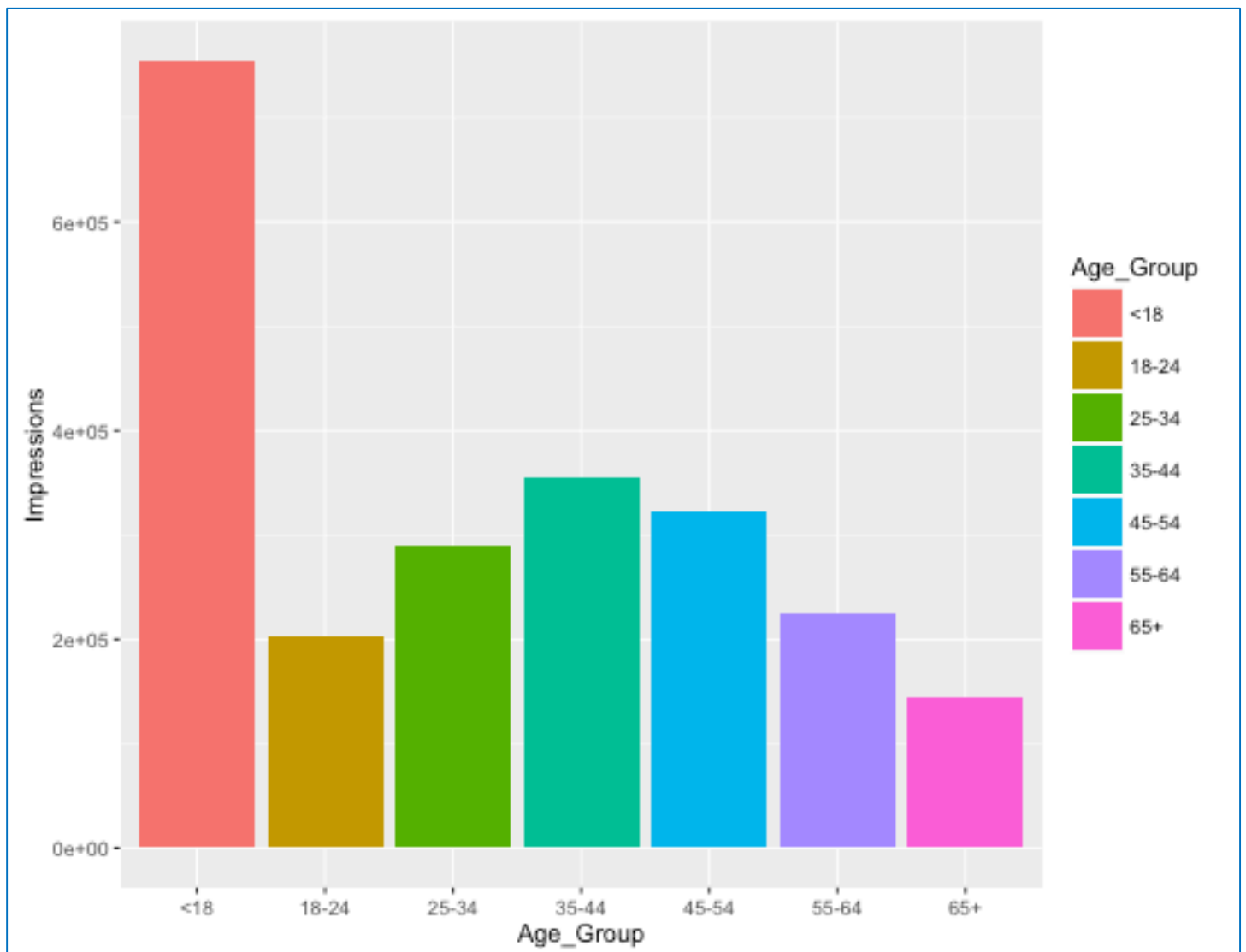
```
> summaryBy(Clicks~Has_Impressions, data=data_nyt_day_1, FUN=siterange)
```

	Has_Impressions	Clicks.FUN1	Clicks.FUN2	Clicks.FUN3	Clicks.FUN4
1	No	3066	0	0.00000000	0
2	Yes	455375	0	0.09321768	4

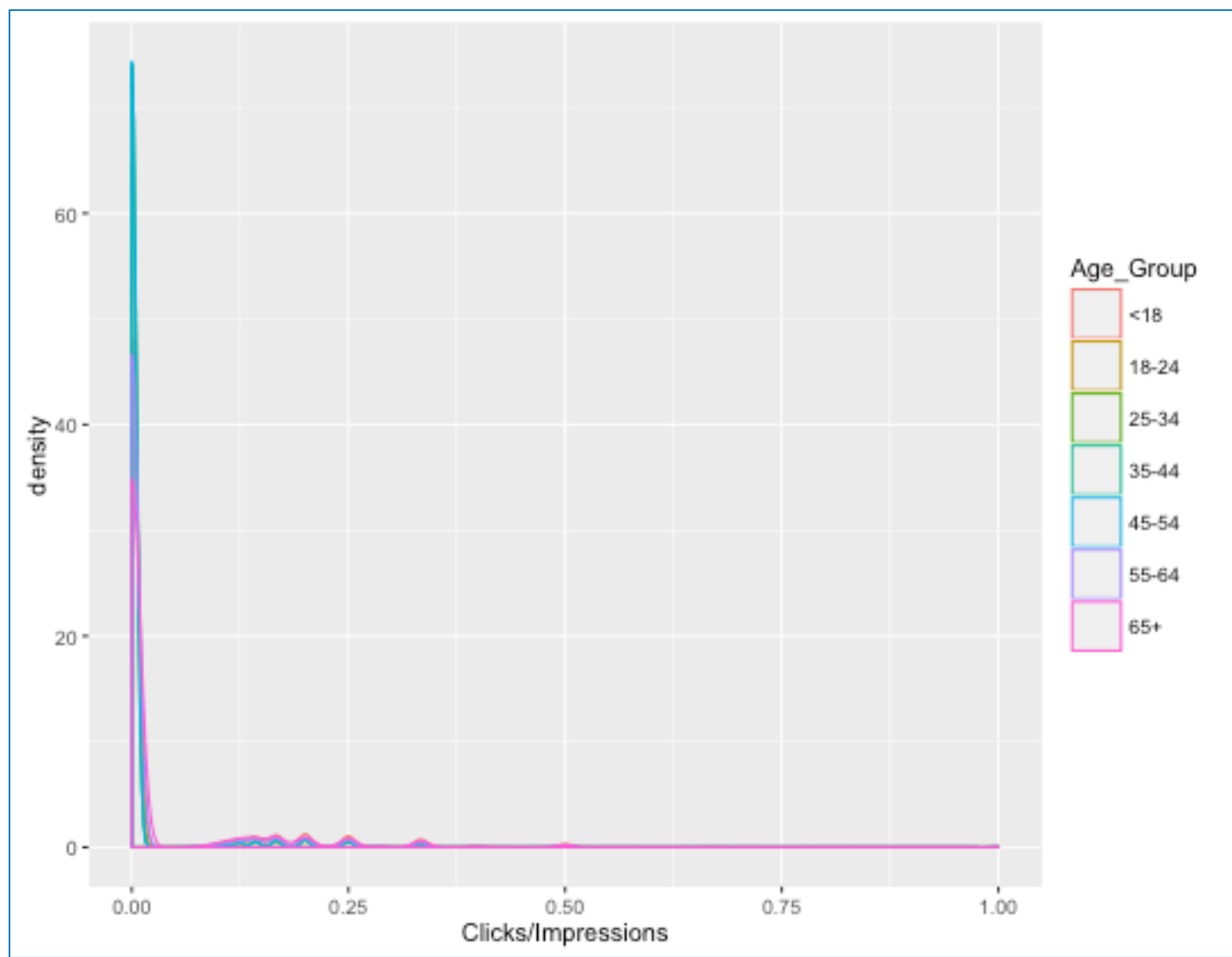
### Step 3: Generating Plots and capture meaningful information

In the next step we have to capture meaningful information from the data that we cleaned. Graphs are very intuitive way of depicting such information in a very elegant way.

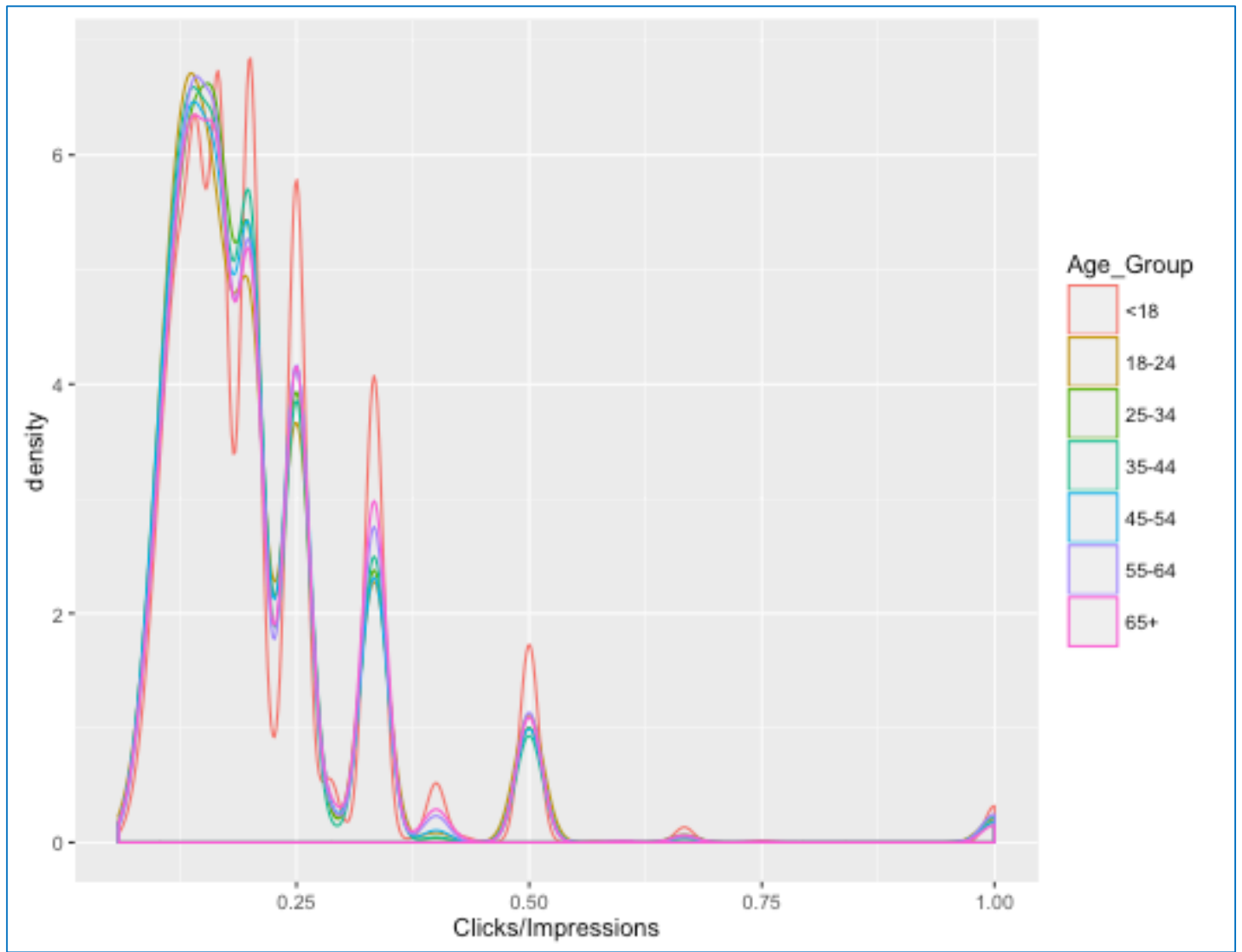
#### Graph on Age Groups vs No. of Impressions



## Graph on Click thru Rate



## Graph on Click thru Rate



## Problem 2B: Simple EDA on monthly data

### Step 1:

Now to do analysis on monthly data we will do grouping of counts of clicks, Impressions, Males and Females based on Age Groups.

This summarization will help us do EDA on the monthly data

```
3 # Creating a main collection class for capturing monthly data
4 # We are grouping counts of clicks, Impressions, Males and Females based on Age Groups
5 # This summarization will help us do EDA on the monthly data
6 main_collection <- aggregate(cbind(data_nyt_day_1$Clicks,data_nyt_day_1$Impressions,
7                                   (data_nyt_day_1$Gender == "male"),
8                                   (data_nyt_day_1$Gender == "female")
9                                   )~Age_Group,data=data_nyt_day_1,sum,na.rm=TRUE)
10 colnames(main_collection) <- c("Age_Group", "Clicks", "Impressions", "Num_of_Males", "Num_of_Females")
11 main_collection$Day <- "1"
12 head(main_collection)
```

After aggregating required data, we will get something like this

	Age_Group	Clicks	Impressions	Num_of_Males	Num_of_Females	Day
1	<18	21545	754722	9470	141464	1
2	18-24	2167	203585	21721	18973	1
3	25-34	2937	290511	30958	27216	1
4	35-44	3662	355824	37676	33184	1
5	45-54	3232	322109	34007	30281	1
6	55-64	4556	224688	23988	20750	1

## Step 2:

In the next step we will iterate over monthly data and perform the same steps as we performed previously and collect the data into the main collection

```
14 # Loop for capturing monthly data.
15 # This function will fetch data for each day, aggregate required fields and merge into the main collection
16 x <- 2
17 repeat {
18   print(paste("Collecting New York Times data for Day ", x, sep=""))
19   data_nyt_day_n <- read.csv(url(paste("http://stat.columbia.edu/~rachel/datasets/nyt", x, ".csv", sep = "")))
20   data_nyt_day_n$Age_Group <- cut(data_nyt_day_n$Age, c(-Inf, 17, 24, 34, 44, 54, 64, Inf),
21                                   c("<18", "18-24", "25-34", "35-44", "45-54", "55-64", "65+"))
22   data_nyt_day_n$Gender = factor(data_nyt_day_n$Gender, levels=c(0,1), labels = c("female", "male"))
23   temp_collection <- aggregate(cbind(data_nyt_day_n$Clicks,data_nyt_day_n$Impressions,
24                                       (data_nyt_day_n$Gender == "male"),
25                                       (data_nyt_day_n$Gender == "female"))
26                               ~Age_Group,data=data_nyt_day_n,sum,na.rm=TRUE)
27   colnames(temp_collection) <- c("Age_Group", "Clicks", "Impressions", "Num_of_Males", "Num_of_Females")
28   temp_collection$Day <- x
29   main_collection <- rbind(main_collection, temp_collection)
30   rm(data_nyt_day_n)
31   x = x+1
32   if (x == 32){
33     break
34   }
```

## Step 3:

Next we will group data based on different age groups to analyze them more deeply

```
37 # Extracting data for different age groups
38 # This will help us deeply analyse data for different age groups
39 Age_Group_1 <- main_collection[main_collection$Age_Group=="<18",]
40 Age_Group_2 <- main_collection[main_collection$Age_Group=="18-24",]
41 Age_Group_3 <- main_collection[main_collection$Age_Group=="25-34",]
42 Age_Group_4 <- main_collection[main_collection$Age_Group=="35-44",]
43 Age_Group_5 <- main_collection[main_collection$Age_Group=="45-54",]
44 Age_Group_6 <- main_collection[main_collection$Age_Group=="55-64",]
45 Age_Group_7 <- main_collection[main_collection$Age_Group=="65+",]
```



The aggregated monthly data for users falling under Age Group 35-44 will look like this

> Age\_Group\_4

	Age_Group	Clicks	Impressions	Num_of_Males	Num_of_Females	Day
4	35-44	3662	355824	37676	33184	1
11	35-44	3519	343978	37610	31364	2
18	35-44	3408	336558	35704	31495	3
25	35-44	3453	339483	36465	31471	4
32	35-44	2808	282338	30267	26331	5
39	35-44	6014	591235	61066	57315	6
46	35-44	3468	345754	35353	33931	7
53	35-44	3582	356379	38146	33145	8
60	35-44	3592	352064	36697	33731	9
67	35-44	3513	346599	36591	32745	10
74	35-44	3642	368141	38729	35026	11
81	35-44	3008	303404	31880	28832	12
88	35-44	5942	605061	61580	59487	13
95	35-44	3436	338404	36621	31156	14
102	35-44	2553	264407	28051	24862	15
109	35-44	2807	272316	28320	26061	16
116	35-44	2684	269045	28907	24923	17
123	35-44	2748	275143	28591	26317	18
130	35-44	2574	255684	27315	23844	19
137	35-44	4403	448103	47842	41546	20
144	35-44	2685	276618	30303	25222	21
151	35-44	2620	266372	27605	25795	22
158	35-44	2546	260451	27709	24314	23
165	35-44	2830	273689	29148	25793	24
172	35-44	2444	258459	27293	24609	25
179	35-44	2144	219616	23928	19975	26
186	35-44	4545	451888	47503	42690	27
193	35-44	2746	269317	27888	25968	28
200	35-44	2841	279782	29506	26455	29
207	35-44	2723	274267	28360	26573	30
214	35-44	3439	338410	35297	32502	31

## Step 4: Analyzing monthly data by plotting graphs

Next we will plot some beautiful graphs which will give us intuition about behavior change in different Age Groups.

For example, for the users falling under Age Group 35-44 we see that the maximum number of clicks were on Day 6<sup>th</sup> and 13<sup>th</sup>

