Foundation of Financial Data Science (FE 582)

(Homework 1)

Prof. Dragos Bozdog

Student Name: **Paras Garg**

Course Section: **FE 582 A**

Problem 1 –

Explore realdirect.com thinking about how buyers and sellers would navigate, and how the website is organized. Use the datasets provided for Bronx, Brooklyn, Manhattan, Queens, and Staten Island. Do the following:

- Load in and clean the data
- Conduct exploratory data analysis in order to find out where there are outliers or missing values, decide how you will treat them, make sure the dates are formatted correctly, make sure values you think are numerical are being treated as such, etc.
- Conduct exploratory data analysis to visualize and make comparisons for residential building category classes across boroughs and across time (1, 2, 3 family homes, coops, and condos). Use histograms, box plots, scatter plots, or other visual graphs. Provide summary statistics along with your conclusions.

Analysis 1 –

```
### CLEANING ENVIRONMENT AND SETTING WORK DIRECTORY
rm(list=ls())
setwd("C:/Users/Paras Garg/Documents/R/FE Assignments/ ")
### INCLUING USEFUL PACKAGES
library('gdata')
library('ggplot2')
library('plyr')
library("doBy")

### LOADING DATASETS
bronx <- read.xls("rollingsales_bronx.xls", perl = "C:\\Perl64\\bin\\perl.exe", pattern="BOROUGH")
brooklyn <- read.xls("rollingsales_brooklyn.xls", perl = "C:\\Perl64\\bin\\perl.exe", pattern="BOROUGH")
manhattan <- read.xls("rollingsales_manhattan.xls", perl = "C:\\Perl64\\bin\\perl.exe", pattern="BOROUGH")
queens <- read.xls("rollingsales_queens.xls", perl = "C:\\Perl64\\bin\\perl.exe", pattern="BOROUGH")
staten <- read.xls("rollingsales_statenisland.xls", perl = "C:\\Perl64\\bin\\perl.exe", pattern="BOROUGH"
```

```
### DATA FORMATTING FUNCTION
format_data <- function (df) {
  df$GROSS.SQUARE.FEET.N <- as.numeric((gsub("[^[:digit:]]","", df$GROSS.SQUARE.FEET)))
  df$LAND.SQUARE.FEET.N <- as.numeric((gsub("[^[:digit:]]","", df$LAND.SQUARE.FEET)))
  df$SALE.PRICE.N <- as.numeric(gsub("[^[:digit:]]","",df$SALE.PRICE))
  df$SALE.DATE <- as.Date(df$SALE.DATE)
  df$YEAR.BUILT <- as.numeric(as.character(df$YEAR.BUILT))
  return (df)
}
```

```
### DATA CLEANING FUNCTION
clean_data <- function (df) {
  #Removing NA values
  df <- df[!is.na(df$GROSS.SQUARE.FEET.N), ]
  df <- df[!is.na(df$LAND.SQUARE.FEET.N), ]
  df <- df[!is.na(df$SALE.PRICE.N), ]
  #Removing outliners
  df$SALE.PRICE.LOG <- log(df$SALE.PRICE.N)
  df <- df[df$SALE.PRICE.LOG > 5, ]
  #Categorize sales buildings
  family_category <- grepl("FAMILY", df$BUILDING.CLASS.CATEGORY) * 1
  condos_category <- grepl("CONDOS", df$BUILDING.CLASS.CATEGORY) * 2
  coops_category <- grepl("COOPS", df$BUILDING.CLASS.CATEGORY) * 3

  category <- as.character(family_category + condos_category + coops_category)
  category[category == "1"] <- "FAMILY"
  category[category == "2"] <- "CONDOS"
  category[category == "3"] <- "COOPS"
  category[category == "0"] <- "OTHERS"

  df$BUILDING.CLASS.CATEGORY.N <- factor(category)
  return (df)
}
```

```
### DATA FRAMES FORMATTING AND CLEANING
bronxDf <- clean_data(format_data(bronx))
brooklynDf <- clean_data(format_data(brooklyn))
manhattanDf <- clean_data(format_data(manhattan))
queensDf <- clean_data(format_data(queens))
statenDf <- clean_data(format_data(staten))
```

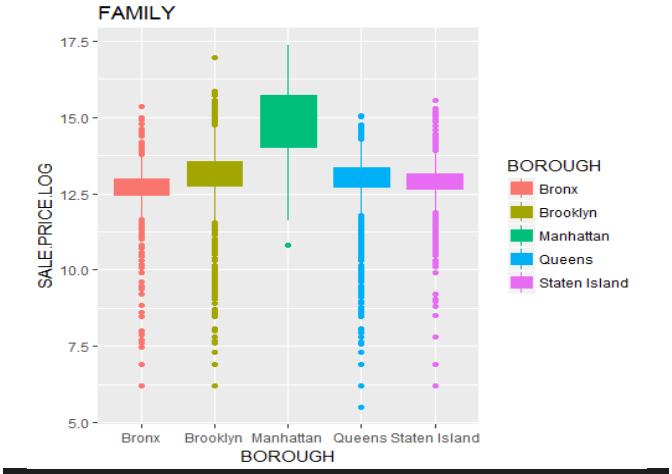
Analysis across boroughs

#Functions

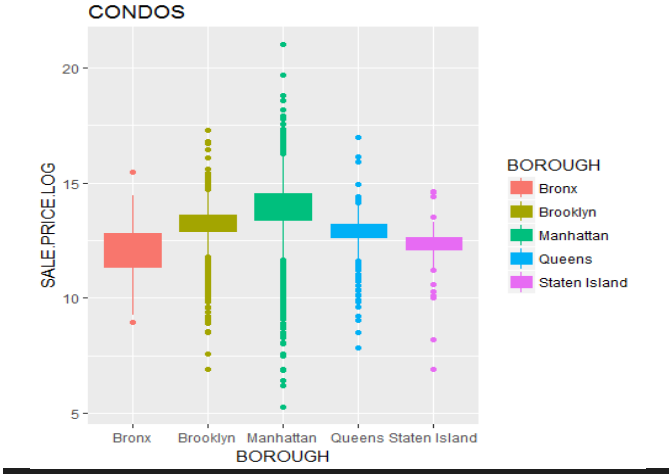
```
  # FUNCTION: Borough v/s Sale Price for particular Building Class
  borough_sp <- function (class) {
      outliers <- bronxDf$SALE.PRICE.LOG[bronxDf$BUILDING.CLASS.CATEGORY.N == class]
      borough_1 <- data.frame(BOROUGH = rep("Bronx", length(outliers)), SALE.PRICE.LOG = outliers)
      outliers <- brooklynDf$SALE.PRICE.LOG[brooklynDf$BUILDING.CLASS.CATEGORY.N == class]
      borough_2 <- data.frame(BOROUGH = rep("Brooklyn", length(outliers)), SALE.PRICE.LOG = outliers)
      outliers <- manhattanDf$SALE.PRICE.LOG[manhattanDf$BUILDING.CLASS.CATEGORY.N == class]
      borough_3 <- data.frame(BOROUGH = rep("Manhattan", length(outliers)), SALE.PRICE.LOG = outliers)
      outliers <- queensDf$SALE.PRICE.LOG[queensDf$BUILDING.CLASS.CATEGORY.N == class]
      borough_4 <- data.frame(BOROUGH = rep("Queens", length(outliers)), SALE.PRICE.LOG = outliers)
      outliers <- statenDf$SALE.PRICE.LOG[statenDf$BUILDING.CLASS.CATEGORY.N == class]
      borough_5 <- data.frame(BOROUGH = rep("Staten Island", length(outliers)), SALE.PRICE.LOG = outliers)

      finalDf <- rbind(borough_1, borough_2, borough_3, borough_4, borough_5)
      ggplot(finalDf, aes(x=BOROUGH, y=SALE.PRICE.LOG, fill=BOROUGH, colour=BOROUGH, group=BOROUGH)) +
        geom_boxplot() + ggtitle(class)
  }

  # FUNCTION: Building Class v/s Sale Price for particular Borough
  building_sp <- function (df, borough) {
    BUILDING.CLASS <- df$BUILDING.CLASS.CATEGORY.N
    SALE.PRICE.LOG <- df$SALE.PRICE.LOG

    ggplot(df, aes(x=BUILDING.CLASS, y=SALE.PRICE.LOG, fill=BUILDING.CLASS,
                colour=BUILDING.CLASS, group=BUILDING.CLASS)) + geom_boxplot() + ggtitle(borough)
  }

  # FUNCTION: Sale Price v/s Gross Square feet for particular Borough
  gross_sp <- function (df, borough) {
    SALE.PRICE.LOG <- df$SALE.PRICE.LOG
    GROSS.SQFT.LOG <- log(df$GROSS.SQUARE.FEET.N)
    BUILDING.CLASS <- df$BUILDING.CLASS.CATEGORY.N

    ggplot(df, aes(x=SALE.PRICE.LOG, y=GROSS.SQFT.LOG, fill=BUILDING.CLASS, colour=BUILDING.CLASS)) +
      geom_point() + ggtitle(borough)
  }

  # FUNCTION: Sale Price v/s Frequency for particular Borough
  sp_freq <- function (df, borough) {
    par(mfrow=c(2,2))
    hist(log(df[df$BUILDING.CLASS.CATEGORY.N=="FAMILY",]$SALE.PRICE.N), col="#8daeb4",
         main="Family Homes", xlab= "SALE.PRICE.LOG", cex.lab=0.75, cex.main=0.75, cex.axis=0.75)
    hist(log(df[df$BUILDING.CLASS.CATEGORY.N=="CONDOS",]$SALE.PRICE.N), col="#0d447a",
         main="CONDOS", xlab= "SALE.PRICE.LOG", cex.lab=0.75, cex.main=0.75, cex.axis=0.75)
    hist(log(df[df$BUILDING.CLASS.CATEGORY.N=="COOPS",]$SALE.PRICE.N), col="#ffbe4c",
         main="COOPS", xlab= "SALE.PRICE.LOG", cex.lab=0.75, cex.main=0.75, cex.axis=0.75)
    hist(log(df[df$BUILDING.CLASS.CATEGORY.N=="OTHERS",]$SALE.PRICE.N), col="#fddf5f",
         main="Others", xlab= "SALE.PRICE.LOG", cex.lab=0.75, cex.main=0.75, cex.axis=0.75)
    title(main = borough, outer = TRUE, cex.main=1.0, line=-1)
}
```

#Plots

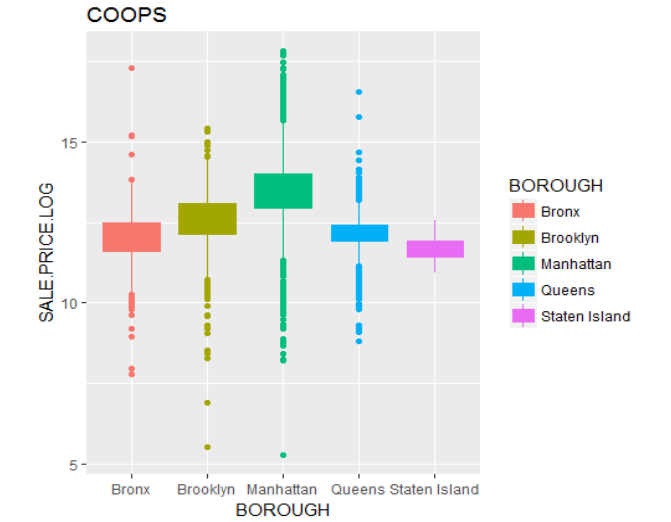
```
# PLOT: Borough v/s Sale Price for particular Building Class
borough_sp("FAMILY")
```



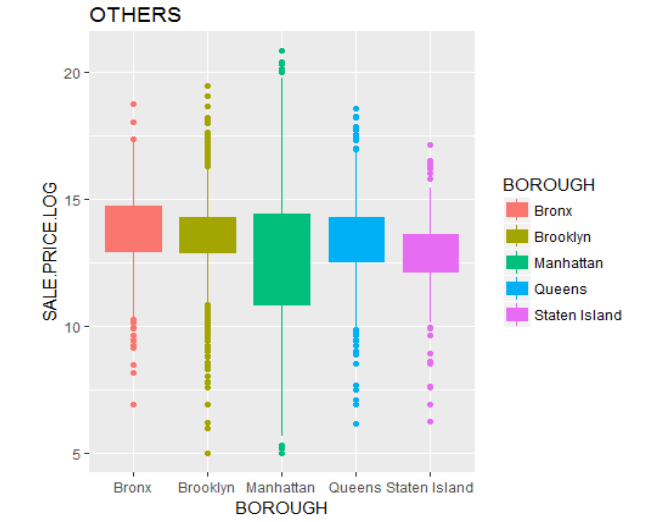

```
borough_sp("CONDOS")
```



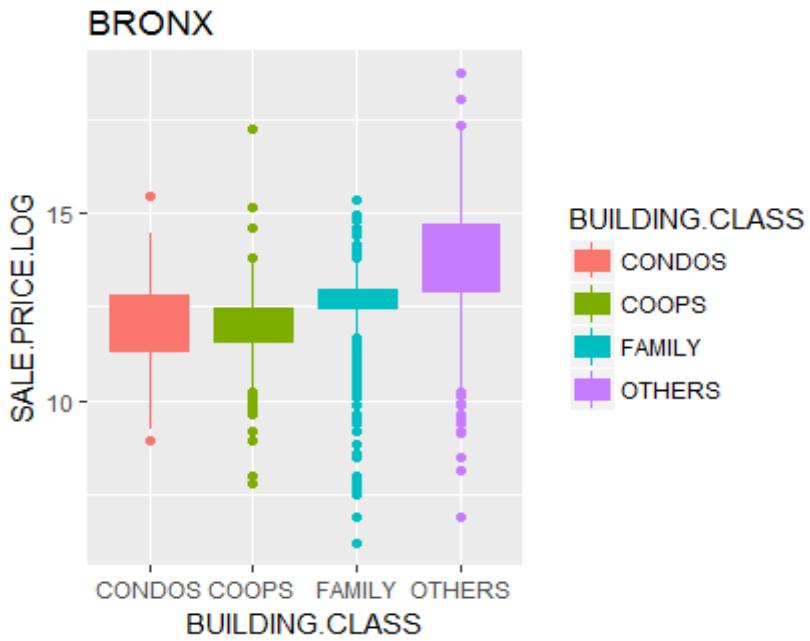

```
borough_sp("COOPS")
```



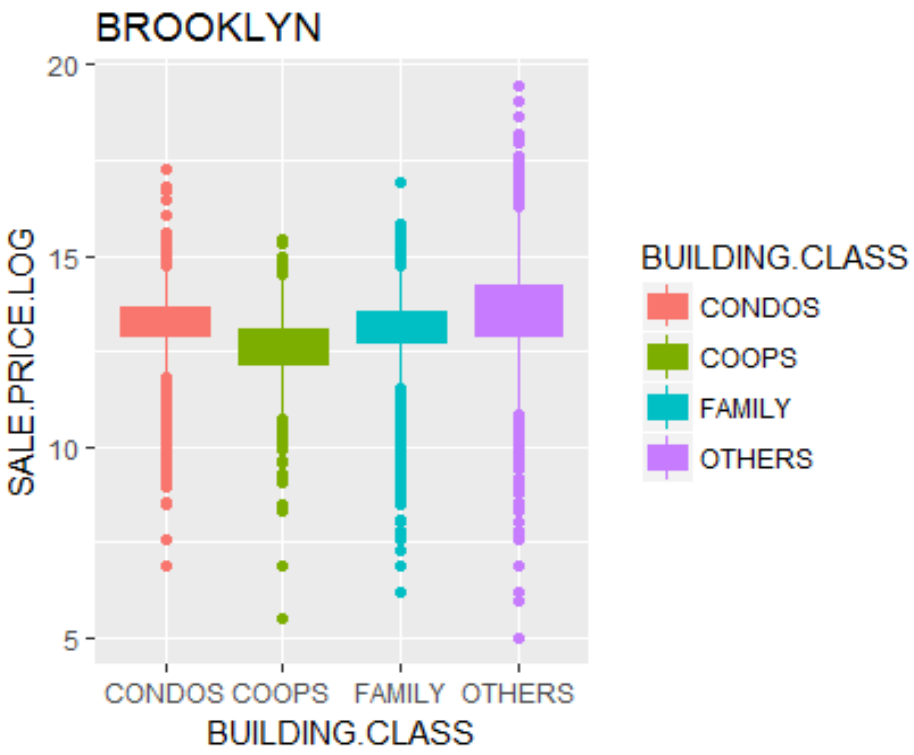

```
borough_sp("OTHERS")
```



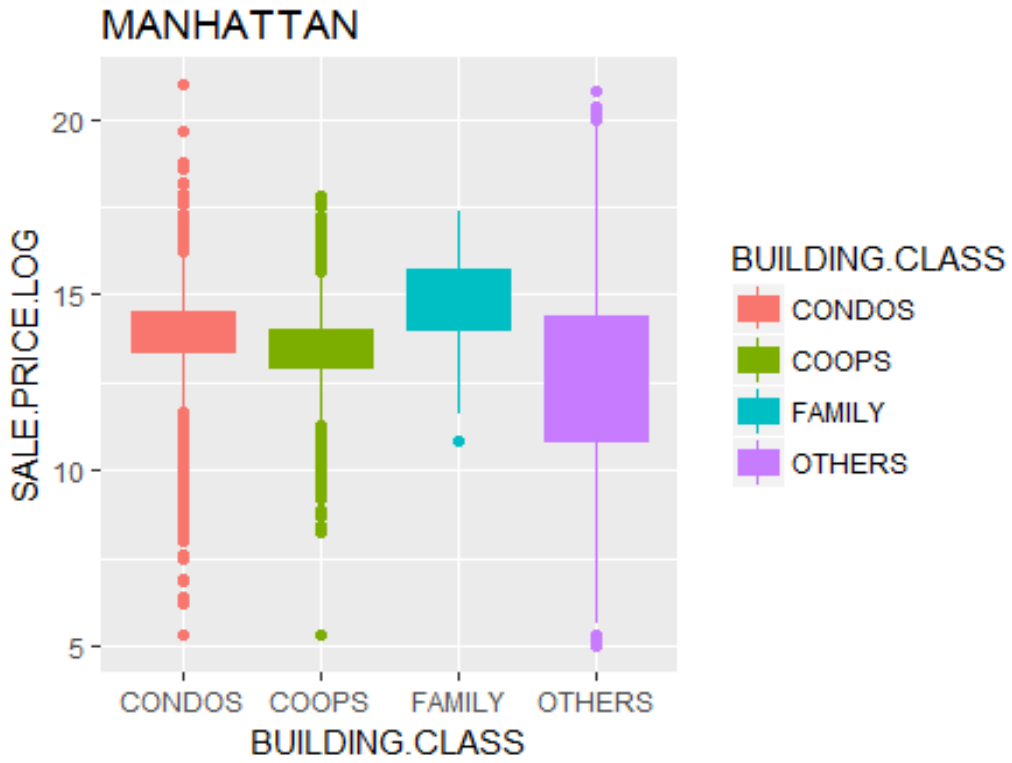

```
#PLOT: Building Class v/s Sale Price for particular Borough
building_sp(bronxDf, "BRONX")
```



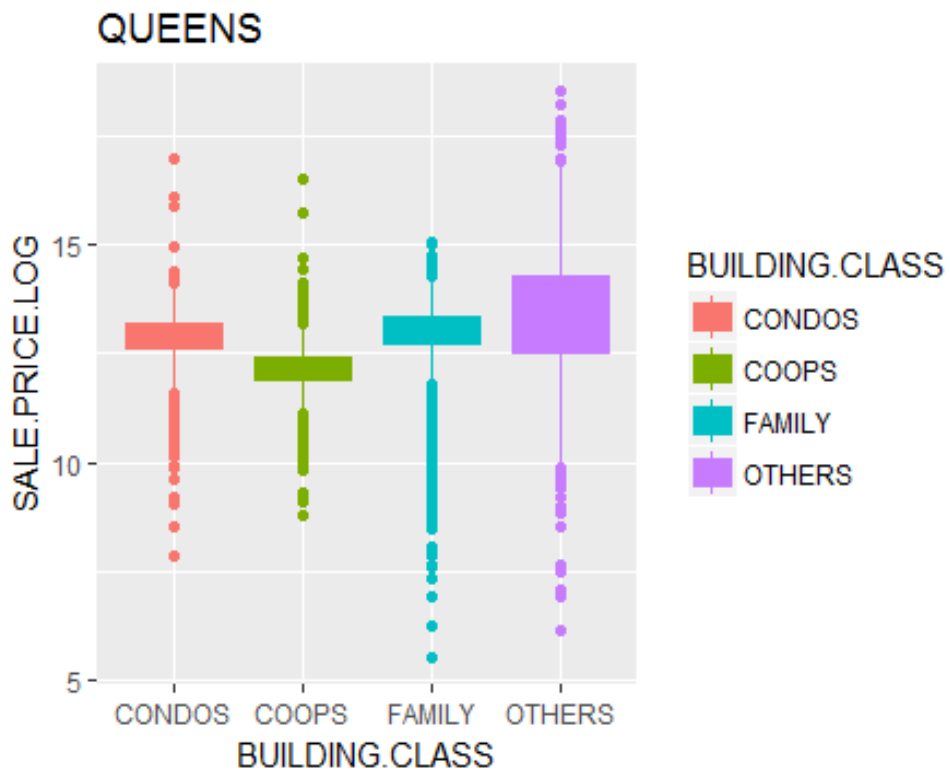
```
building_sp(brooklynDf, "BROOKLYN")
```



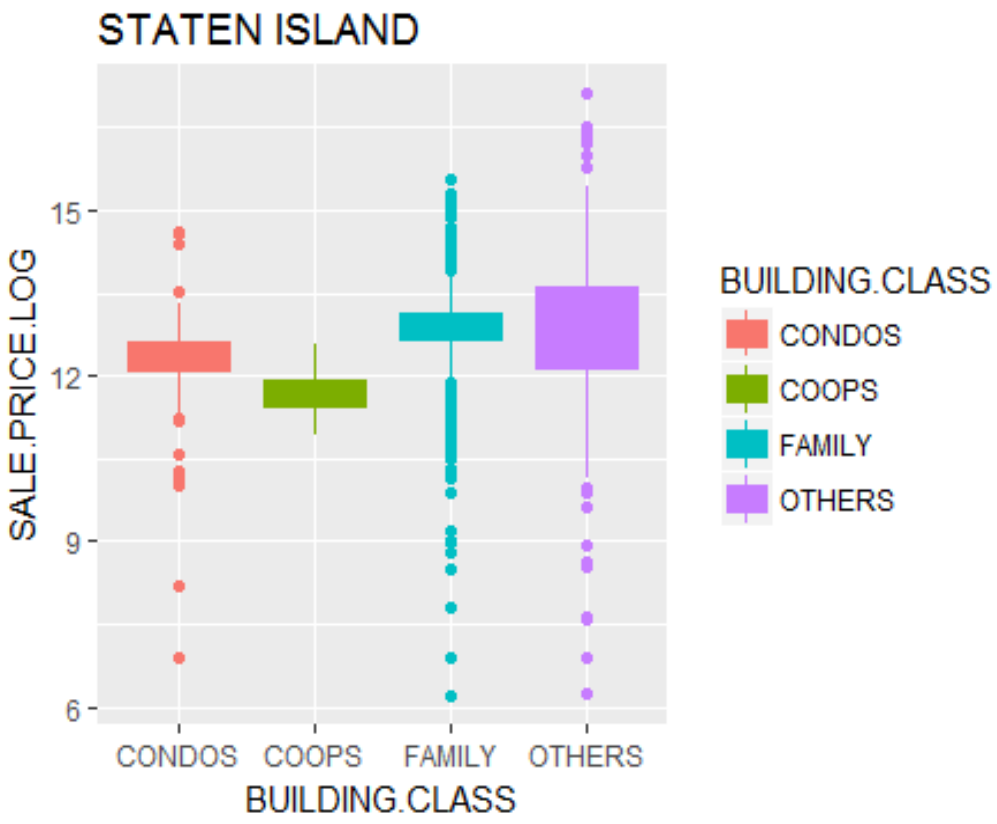

```
building_sp(manhattanDf, "MANHATTAN")
```



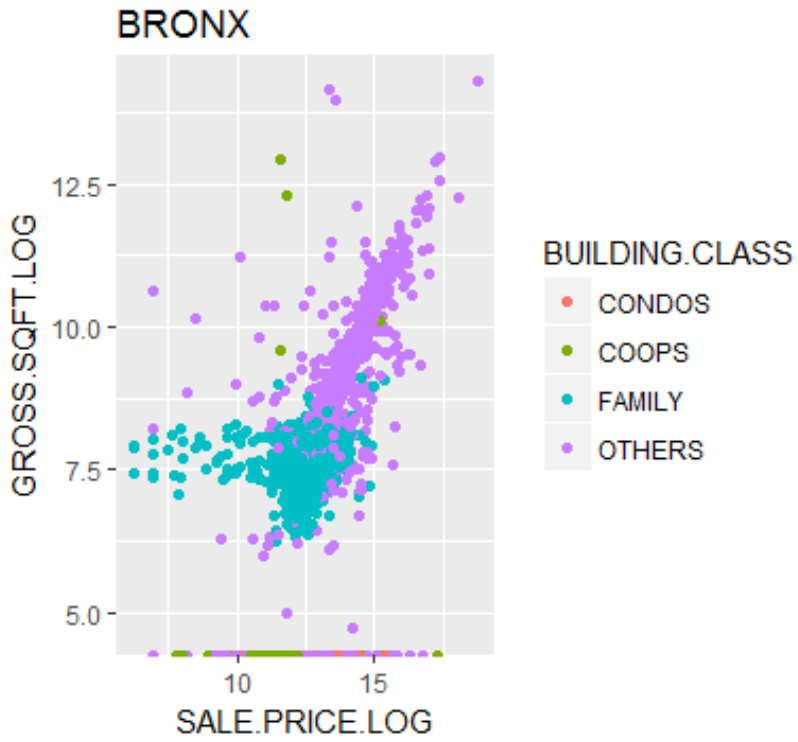
```
building_sp(queensDf, "QUEENS")
```



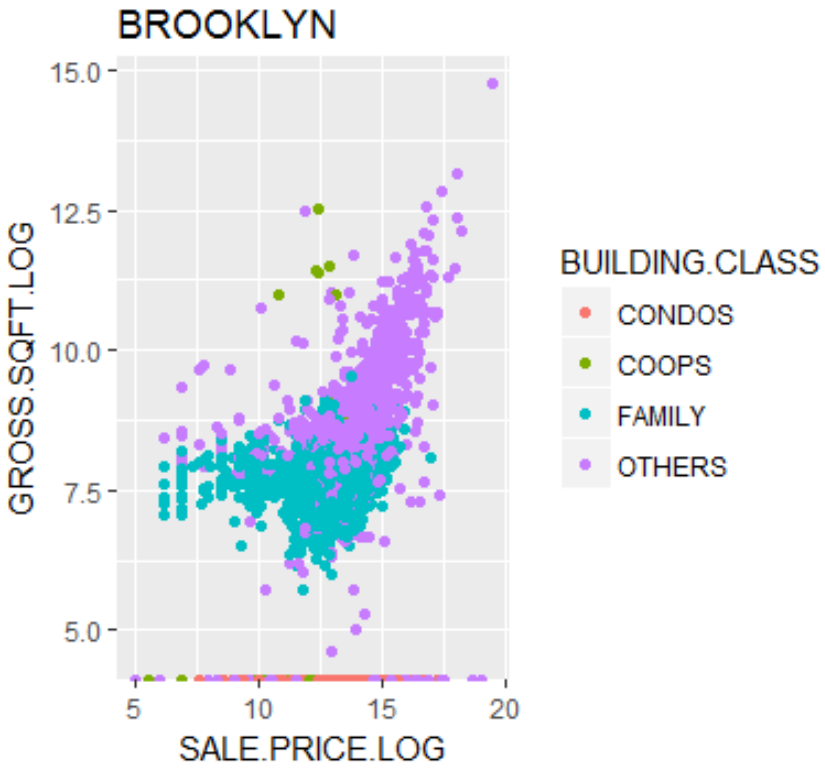

```
building_sp(statenDf, "STATEN ISLAND")
```



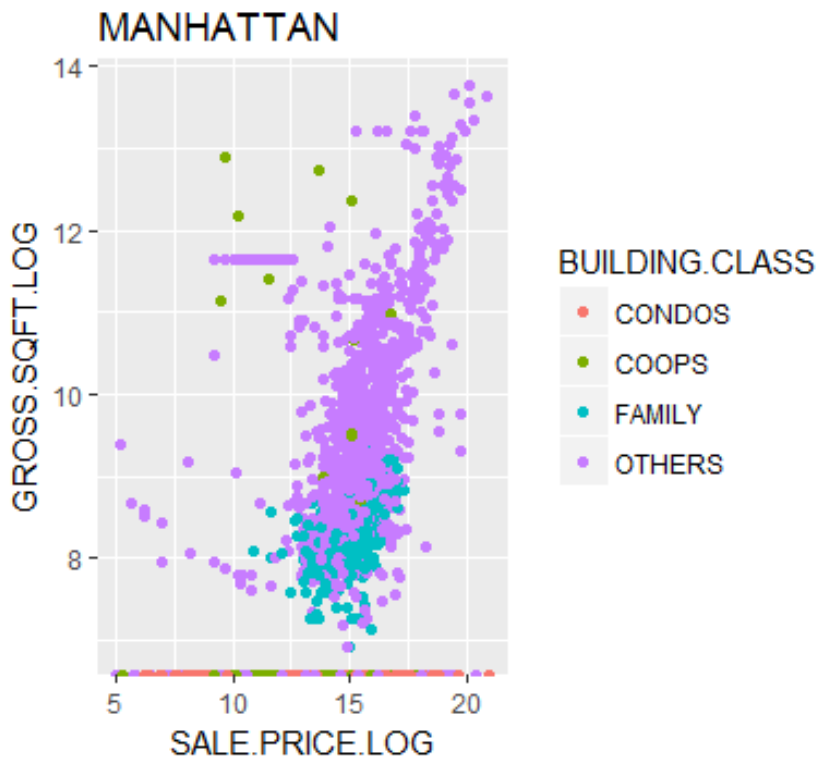
```
# PLOT: Sale Price v/s Gross Square feet for particular Borough
gross_sp(bronxDf, "BRONX")
```



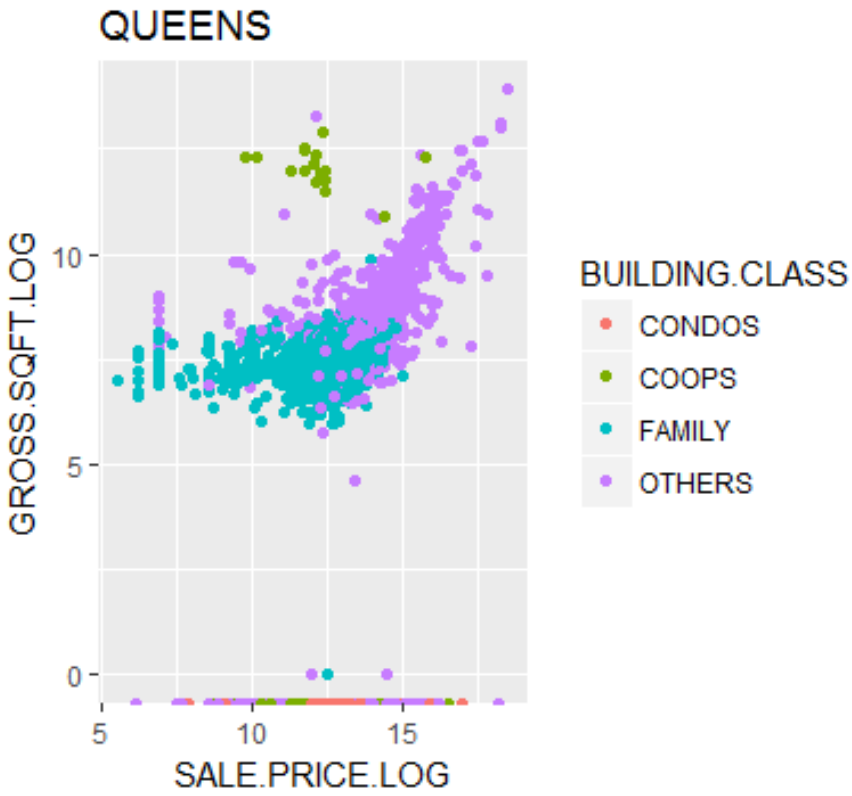

```
gross_sp(brooklynDf, "BROOKLYN")
```



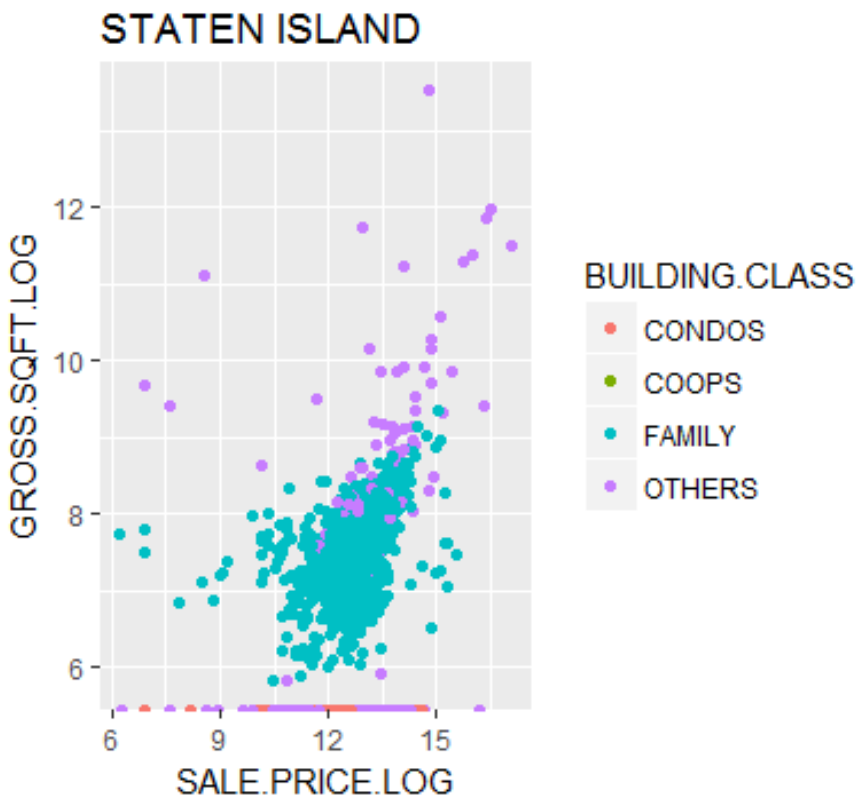
```
gross_sp(manhattanDf, "MANHATTAN")
```




```
gross_sp(queensDf, "QUEENS")
```

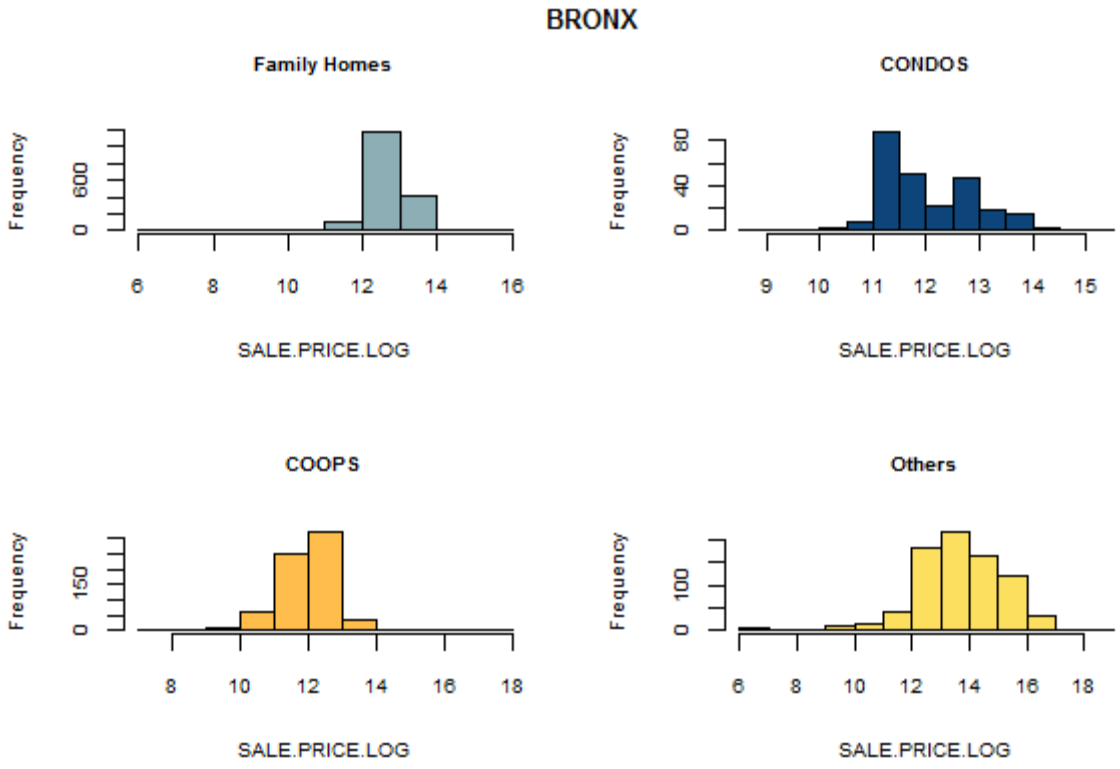


```
gross_sp(statenDf, "STATEN ISLAND")
```



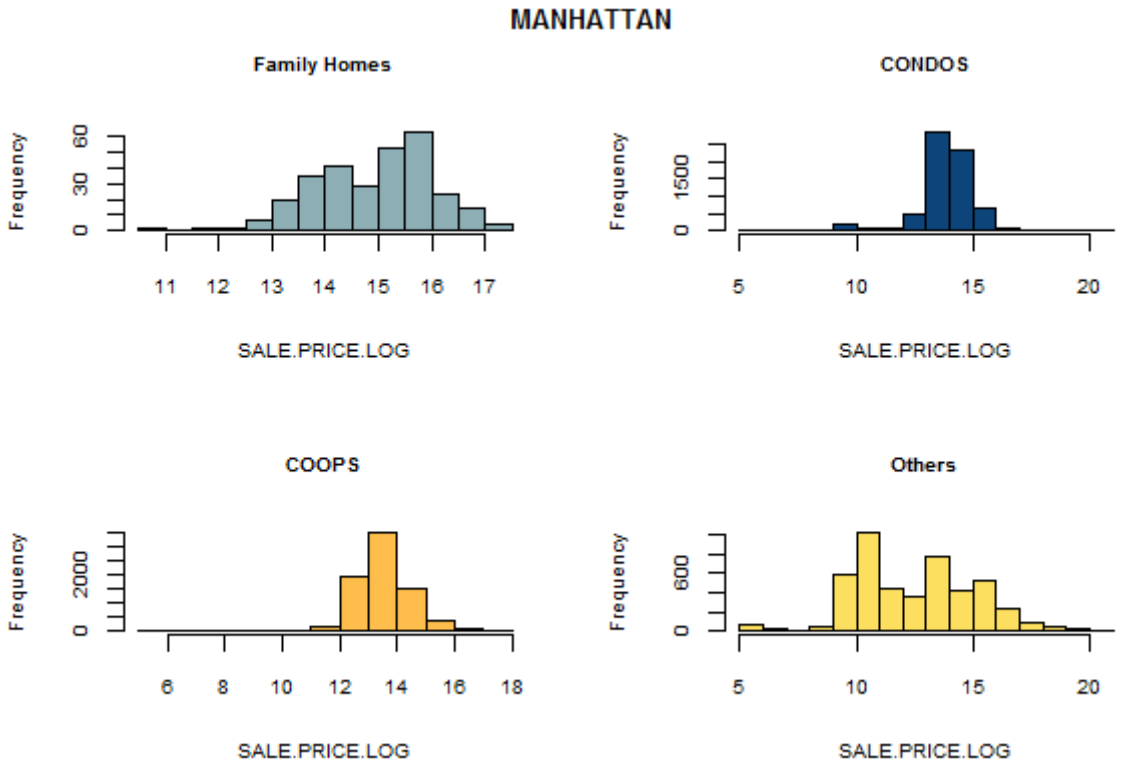

```
# PLOT: Sale Price v/s Frequency for particular Borough
```

```
sp_freq(bronxDf, "BRONX")
```

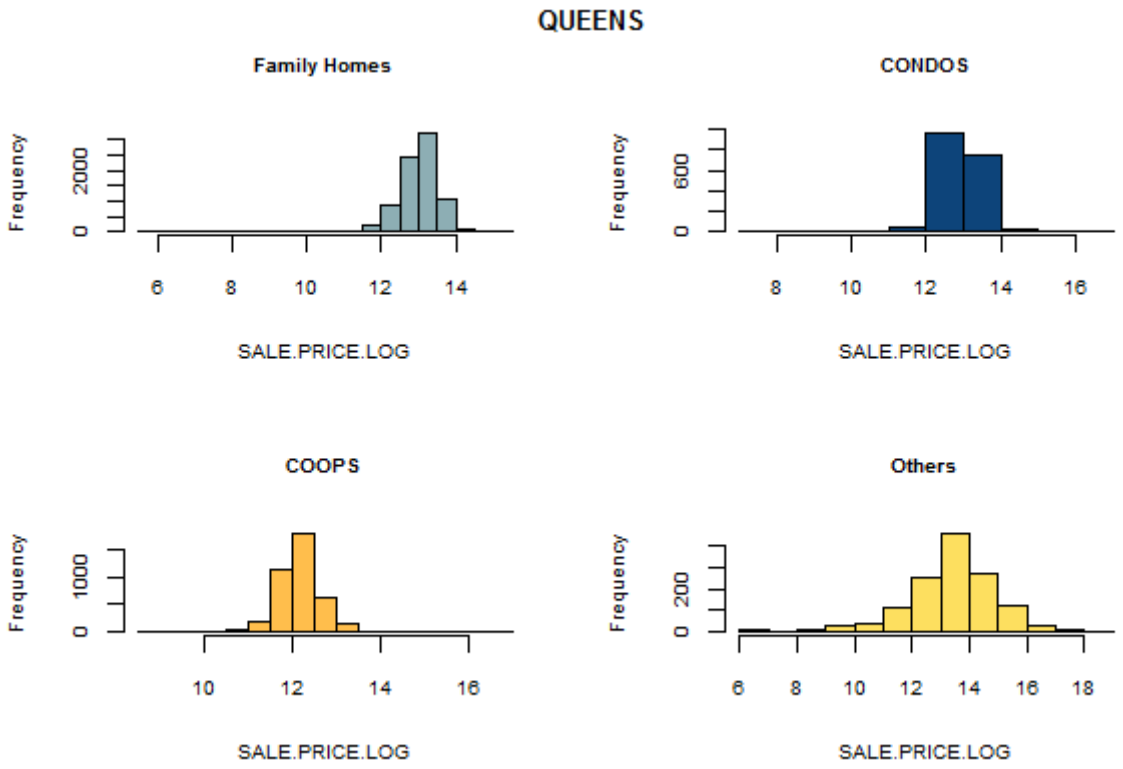


```
sp_freq(brooklynDf, "BROOKLYN")
```

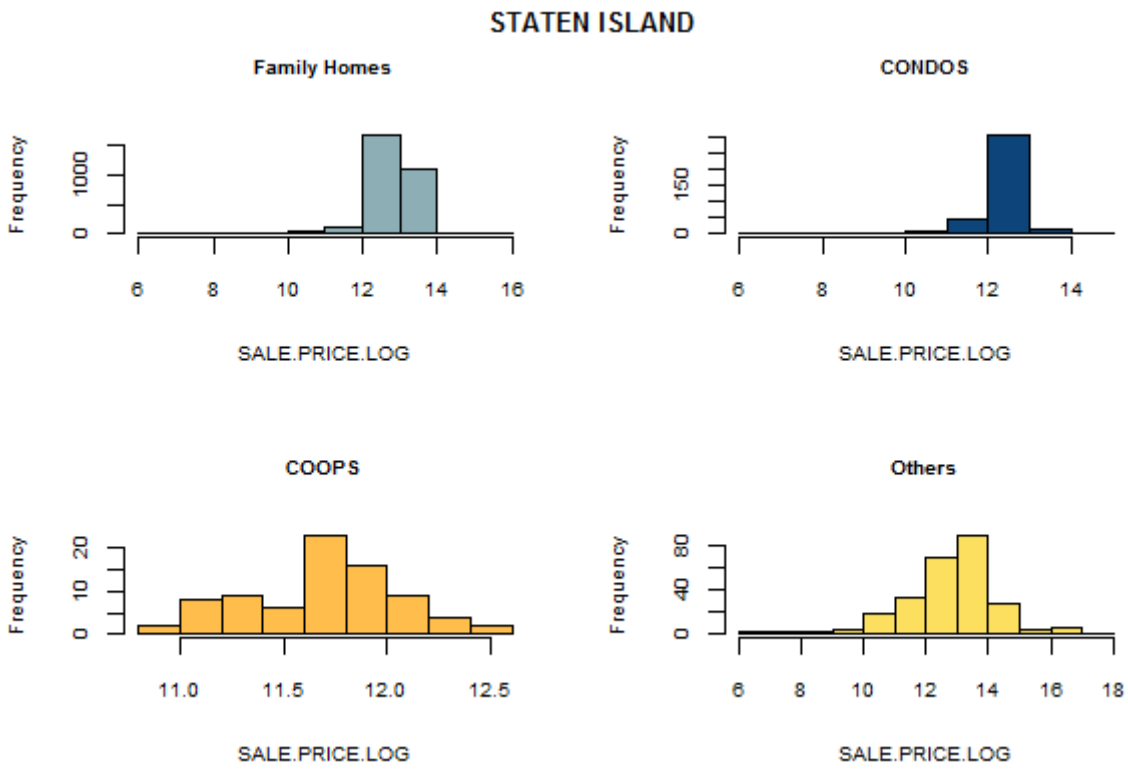

```
sp_freq(manhattanDf, "MANHATTAN")
```



```
sp_freq(queensDf, "QUEENS")
```




```
sp_freq(statenDf, "STATEN ISLAND")
```



#Summary

```
# SUMMARY: Sale prices and gross square feet across boroughs and building classes
summary_stats <- function(df, borough) {
   summaryBy(data = df, SALE.PRICE.N + GROSS.SQUARE.FEET.N ~ BUILDING.CLASS.CATEGORY.N,
             FUN = c(length, mean, median),
             fun.names = c("Total no.", "Mean", "Median"),
             var.names = c(borough))
 }
```

```
summary_stats(bronxDf, "Bronx")
```

```
BUILDING.CLASS.CATEGORY.N SALE.PRICE.N.Total no. GROSS.SQUARE.FEET.N.Total no. SALE.PRICE.N.Mean GROSS.SQUARE.FEET.N.Mean
                   CONDOS                   257                           257          266826.9                   0.000
                    COOPS                   686                           686          263969.6                1003.746
                   FAMILY                  1778                          1778          369892.4                2283.850
                   OTHERS                   794                           794         2455387.0               23878.543
SALE.PRICE.N.Median GROSS.SQUARE.FEET.N.Median
          128500.0                         0.0
          167500.0                         0.0
          360000.0                      2112.0
          799221.5                      4887.5
```

```
summary_stats(brooklynDf, "Brooklyn")
```

```
BUILDING.CLASS.CATEGORY.N SALE.PRICE.N.Total no. GROSS.SQUARE.FEET.N.Total no. SALE.PRICE.N.Mean GROSS.SQUARE.FEET.N.Mean
                   CONDOS                  2997                          2997          715748.7                   0.0000
                    COOPS                  2516                          2516          394294.0                 272.5008
                   FAMILY                  6404                          6404          656383.1                2406.7775
                   OTHERS                  2345                          2345         2136167.2                9213.2678
SALE.PRICE.N.Median GROSS.SQUARE.FEET.N.Median
             570220                           0
             290000                           0
             540000                        2264
             800000                        3878
```

```
summary_stats(manhattanDf, "Manhattan")
```

```
BUILDING.CLASS.CATEGORY.N SALE.PRICE.N.Total no. GROSS.SQUARE.FEET.N.Total no. SALE.PRICE.N.Mean GROSS.SQUARE.FEET.N.Mean
                   CONDOS                  6795                          6795            2045327                   0.0000
                    COOPS                  7621                          7621            1174558                 193.7747
                   FAMILY                   292                           292            5020158                4110.6027
                   OTHERS                  4692                          4692            5607579               37895.2543
SALE.PRICE.N.Median GROSS.SQUARE.FEET.N.Median
            1078000                           0
             640000                           0
            3562500                        3600
             318000                        2747
```

```
summary_stats(queensDf, "Queens")
```

```
BUILDING.CLASS.CATEGORY.N SALE.PRICE.N.Total no. GROSS.SQUARE.FEET.N.Total no. SALE.PRICE.N.Mean GROSS.SQUARE.FEET.N.Mean
                   CONDOS                  1799                          1799          481906.0                   0.0000
                    COOPS                  3979                          3979          222739.2                 797.5838
                   FAMILY                  8146                          8146          494445.8                1853.7823
                   OTHERS                  1347                          1347         2095031.2               10010.7342
SALE.PRICE.N.Median GROSS.SQUARE.FEET.N.Median
             400000                           0
             190000                           0
             465000                        1686
             750000                        3344
```

```
summary_stats(statenDf, "Staten Island")
```

```
BUILDING.CLASS.CATEGORY.N SALE.PRICE.N.Total no. GROSS.SQUARE.FEET.N.Total no. SALE.PRICE.N.Mean GROSS.SQUARE.FEET.N.Mean
                   CONDOS                   383                           383          260863.2                   0.000
                    COOPS                    79                            79          125509.5                   0.000
                   FAMILY                  2974                          2974          438998.6                1929.299
                   OTHERS                   257                           257          953589.7                8858.459
SALE.PRICE.N.Median GROSS.SQUARE.FEET.N.Median
             247500                           0
             117500                           0
             400000                        1782
             435000                           0
```

Analysis across time

```
# DAY AND MONTH FETCHER FUNCTION
day_month <- function (df) {
  df$DAY <- format(df$SALE.DATE, "%A")
  df$DAY <- factor(df$DAY, levels = c("Monday", "Tuesday", "Wednesday", "Thursday", "Friday"))

  df$MONTH <- format(df$SALE.DATE, "%B")
  df$MONTH <- factor(df$MONTH, levels = c("January", "February", "March", "April", "May", "June",
                                          "July", "August", "September", "October", "November",
                                          "December"))
  return(df)
}

#Building data frames
bronxTime <- day_month(bronxDf)
brooklynTime <- day_month(brooklynDf)
manhattanTime <- day_month(manhattanDf)
queensTime <- day_month(queensDf)
statenTime <- day_month(statenDf)
```

#Functions

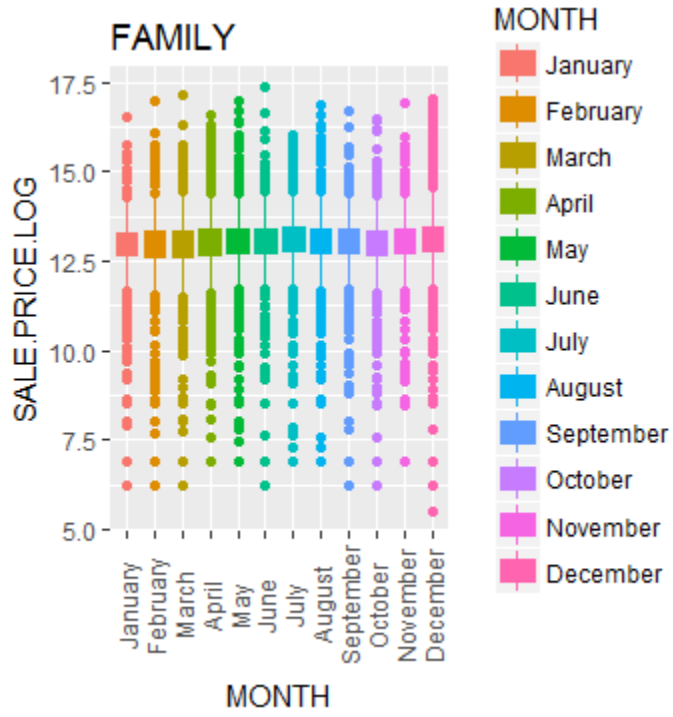
```
#FUNCTION: Total sale of Building Class in Months
monthly_sale <- function (class) {
  filter <- bronxTime$SALE.PRICE.LOG[bronxTime$BUILDING.CLASS.CATEGORY.N == class]
  month <- bronxTime$MONTH[bronxTime$BUILDING.CLASS.CATEGORY.N == class]
  borough_1 <- data.frame(BOROUGH = rep("Bronx", length(filter)), SALE.PRICE.LOG=filter, MONTH=month)
  filter <- brooklynTime$SALE.PRICE.LOG[brooklynTime$BUILDING.CLASS.CATEGORY.N == class]
  month <- brooklynTime$MONTH[brooklynTime$BUILDING.CLASS.CATEGORY.N == class]
  borough_2 <- data.frame(BOROUGH = rep("Brooklyn", length(filter)), SALE.PRICE.LOG=filter, MONTH=month)

  filter <- manhattanTime$SALE.PRICE.LOG[manhattanTime$BUILDING.CLASS.CATEGORY.N == class]
  month <- manhattanTime$MONTH[manhattanTime$BUILDING.CLASS.CATEGORY.N == class]
  borough_3 <- data.frame(BOROUGH = rep("Manhattan", length(filter)), SALE.PRICE.LOG=filter, MONTH=month)

  filter <- queensTime$SALE.PRICE.LOG[queensTime$BUILDING.CLASS.CATEGORY.N == class]
  month <- queensTime$MONTH[queensTime$BUILDING.CLASS.CATEGORY.N == class]
  borough_4 <- data.frame(BOROUGH = rep("Queens", length(filter)), SALE.PRICE.LOG=filter, MONTH=month)

  filter <- statenTime$SALE.PRICE.LOG[statenTime$BUILDING.CLASS.CATEGORY.N == class]
  month <- statenTime$MONTH[statenTime$BUILDING.CLASS.CATEGORY.N == class]
borough_5 <- data.frame(BOROUGH=rep("Staten Island", length(filter)), SALE.PRICE.LOG=filter,MONTH=month)

  finalDf <- rbind(borough_1, borough_2, borough_3, borough_4, borough_5)
  ggplot(finalDf, aes(x=MONTH, y=SALE.PRICE.LOG, fill=MONTH, colour=MONTH, group=MONTH)) +
   theme(axis.text.x=element_text(angle = 90, vjust = 0.5)) +
    geom_boxplot() + ggtitle(class)
}

#FUNCTION: Number of sale of Building Class by Month for particular Borough
building_sale_month <- function (df, class, borough) {
  SALE.PRICE.LOG <- df$SALE.PRICE.LOG[df$BUILDING.CLASS.CATEGORY.N == class]
  ggplot(df, aes(x=MONTH, y=SALE.PRICE.LOG, fill=MONTH, colour=MONTH, group=MONTH)) +
    geom_boxplot() + theme(axis.text.x=element_text(angle = 90, vjust = 0.5)) +
    xlab(borough) + ggtitle(class)
}
```

```r
#FUNCTION: Sale Price v/s Gross Square feet for particular Borough
gross_sp_time <- function (df, borough) {
  SALE.PRICE.LOG <- df$SALE.PRICE.LOG
  GROSS.SQFT.LOG <- log(df$GROSS.SQUARE.FEET.N)
  MONTH <- df$MONTH

  ggplot(df, aes(x=SALE.PRICE.LOG, y=GROSS.SQFT.LOG, fill=MONTH, colour=MONTH, group=MONTH)) +
    geom_point() + ggtitle(borough)
}

#FUNCTION: Sale Price frequency quarter for particular Borough
sp_freq_time <- function (df, borough) {
  par(mfrow=c(2,2))
  hist(log(df[(df$MONTH=="January" | df$MONTH=="Fabruary" | df$MONTH=="March"),]$SALE.PRICE.N),
       col="#8daeb4", main="QUATER 1", xlab= "SALE.PRICE.LOG",
       cex.lab=0.75, cex.main=0.75, cex.axis=0.75)
  hist(log(df[(df$MONTH=="April" | df$MONTH=="May" | df$MONTH=="June"),]$SALE.PRICE.N),
       col="#8daeb4", main="QUATER 2", xlab= "SALE.PRICE.LOG",
      cex.lab=0.75, cex.main=0.75, cex.axis=0.75)
  hist(log(df[(df$MONTH=="July" | df$MONTH=="August" | df$MONTH=="September"),]$SALE.PRICE.N),
       col="#8daeb4", main="QUATER 3", xlab= "SALE.PRICE.LOG",
       cex.lab=0.75, cex.main=0.75, cex.axis=0.75)
  hist(log(df[(df$MONTH=="October" | df$MONTH=="November" | df$MONTH=="December"),]$SALE.PRICE.N),
       col="#8daeb4", main="QUATER 4", xlab= "SALE.PRICE.LOG",
       cex.lab=0.75, cex.main=0.75, cex.axis=0.75)
  title(main = borough, outer = TRUE, cex.main=1.0, line=-1)
}
```

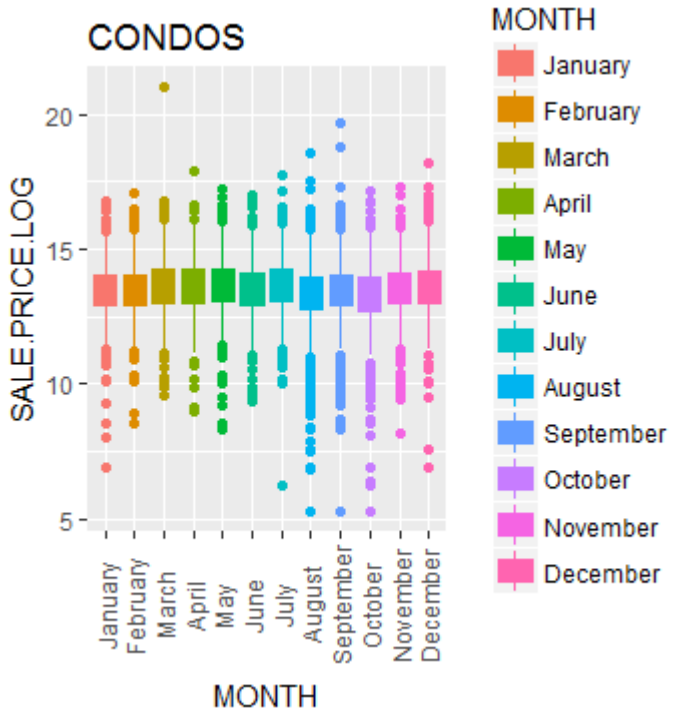
#Plot

#PLOT: Total sale of Building Class in Months

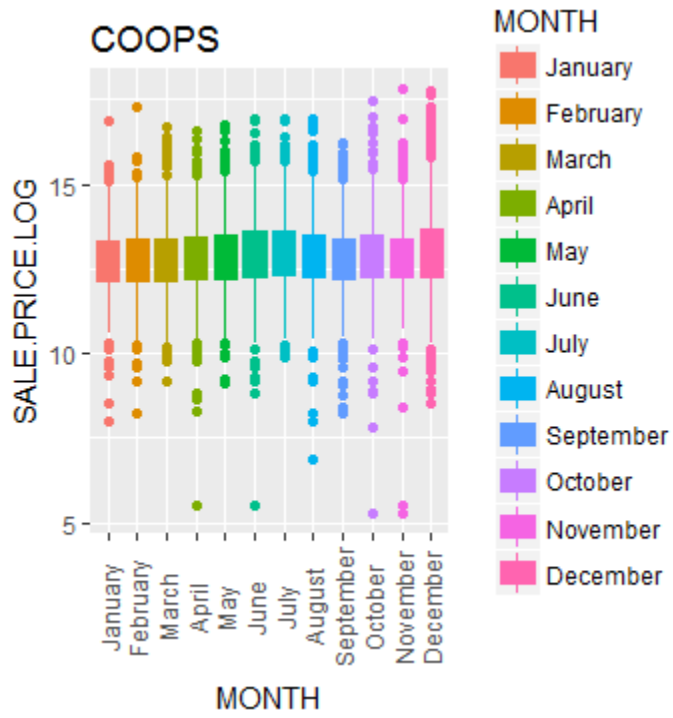
```r
monthly_sale("FAMILY")
```



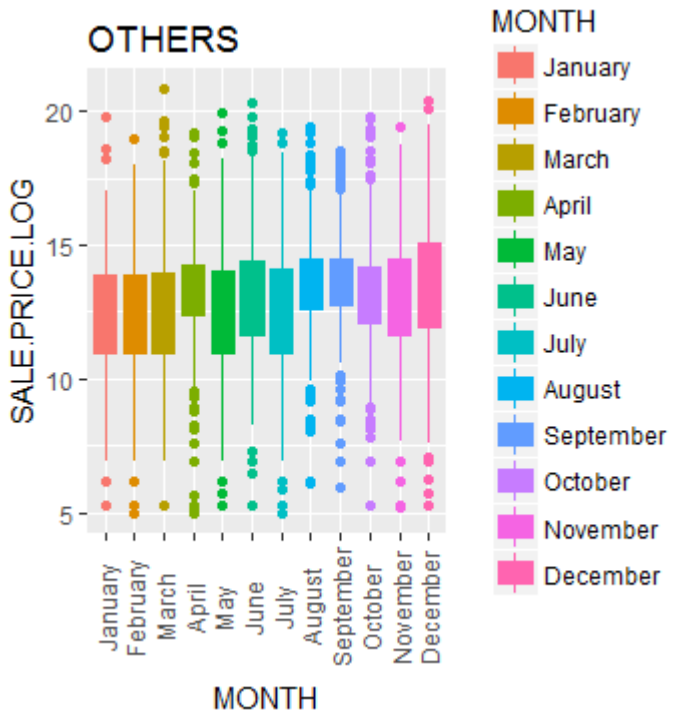

```r
monthly_sale("CONDOS")
```



```
monthly_sale("COOPS")
```

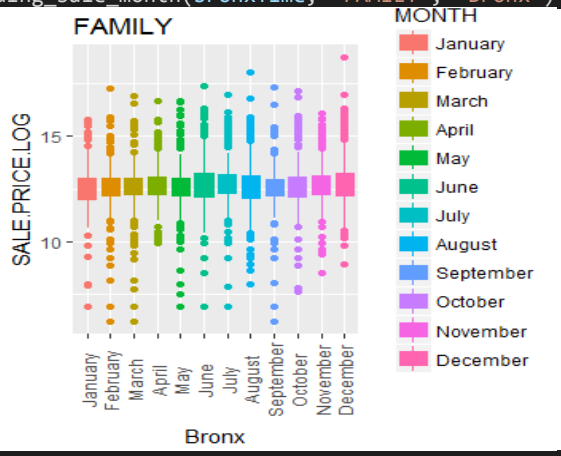



```
monthly_sale("OTHERS")
```

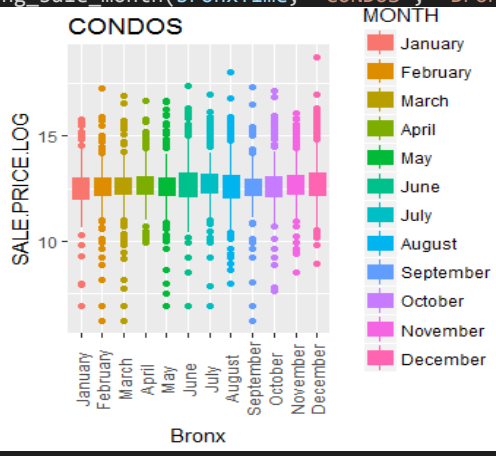



```
#PLOT: Number of sale of Building Class by Month for particular Borough
```

```
building_sale_month(bronxTime, "FAMILY", "Bronx")
```

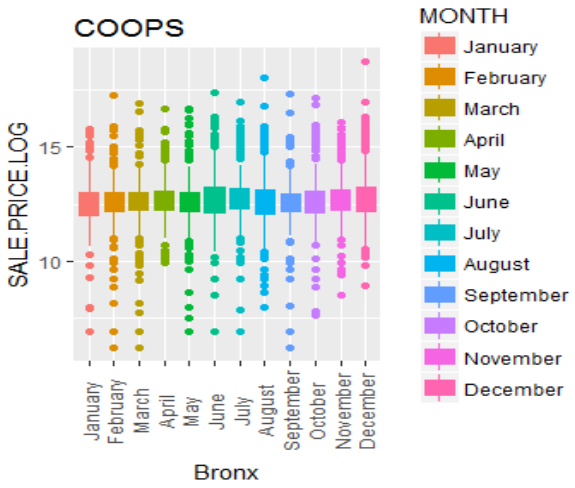


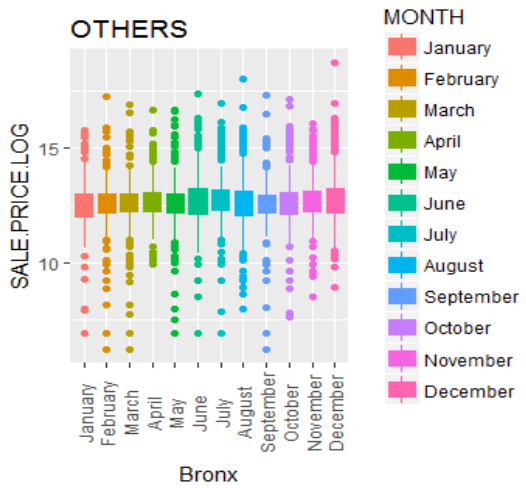

```
building_sale_month(bronxTime, "CONDOS", "Bronx")
```



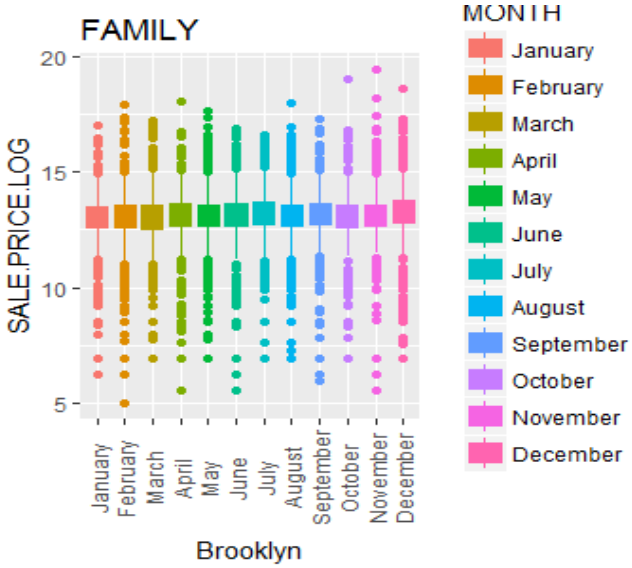

```
building_sale_month(bronxTime, "COOPS", "Bronx")
```

```
building_sale_month(bronxTime, "OTHERS", "Bronx")
```

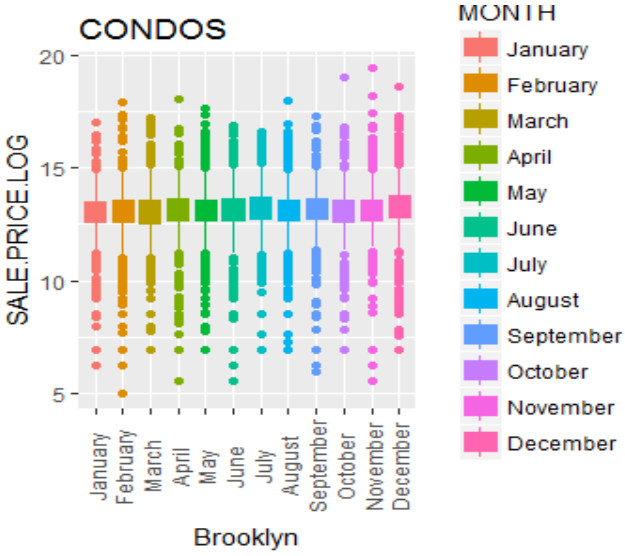




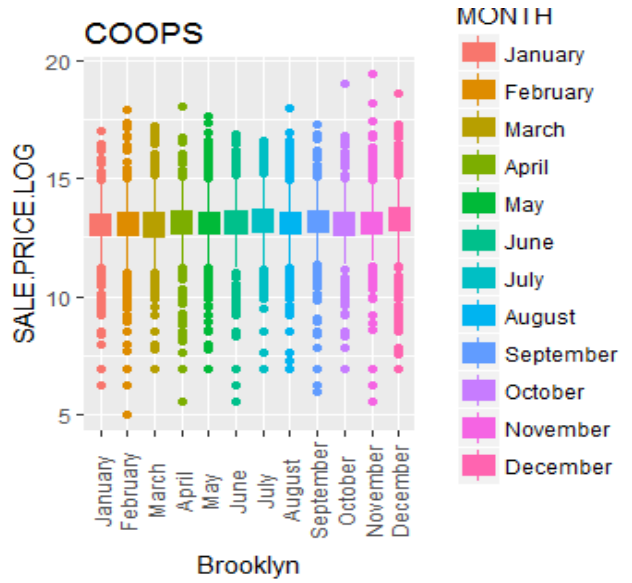

```
building_sale_month(brooklynTime, "FAMILY",
"Brooklyn")
```



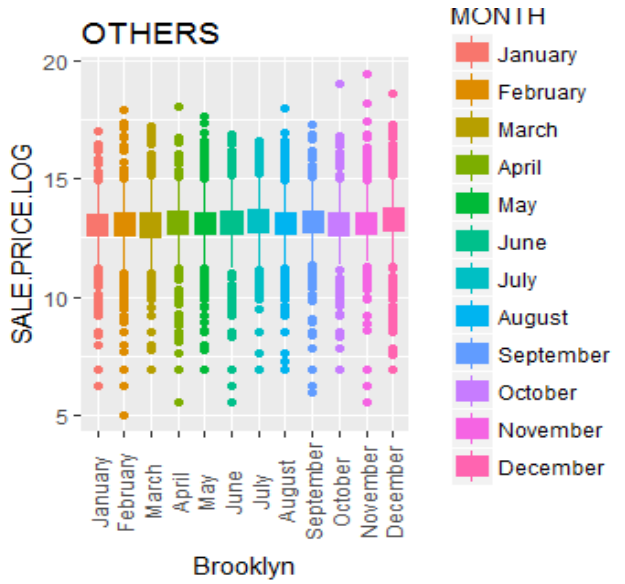

```
building_sale_month(brooklynTime, "CONDOS",
"Brooklyn")
```



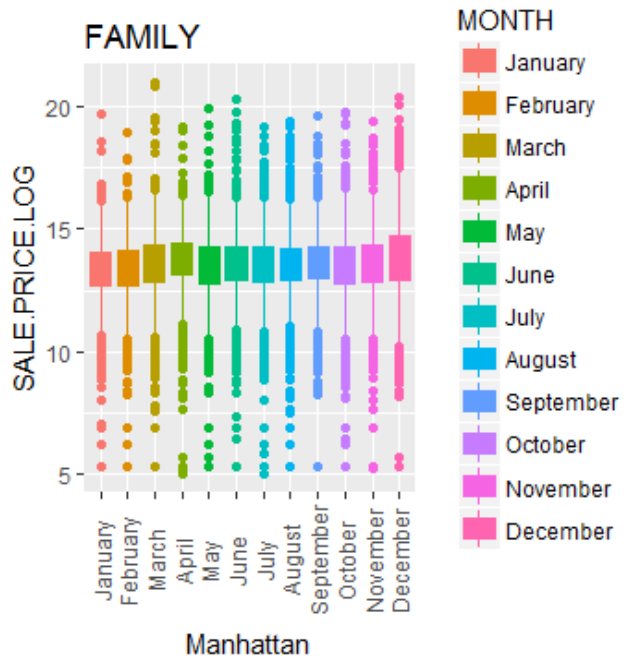

```
building_sale_month(brooklynTime, "COOPS",
"Brooklyn")
```



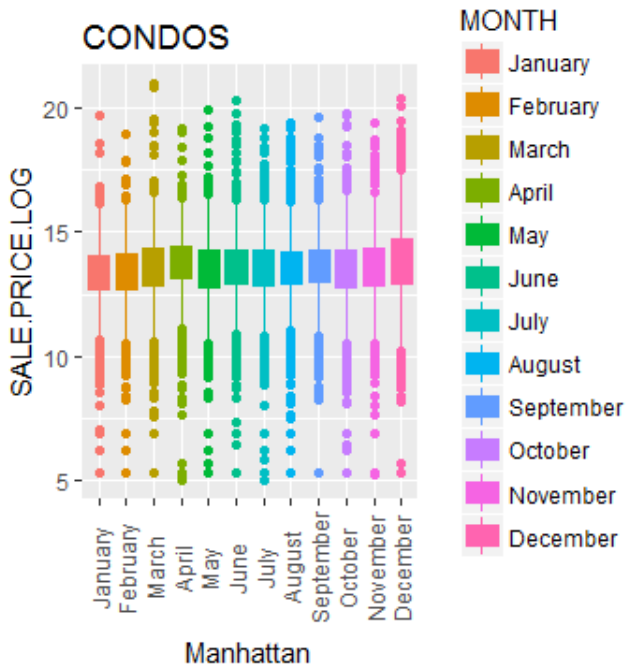

```
building_sale_month(brooklynTime, "OTHERS",
"Brooklyn")
```



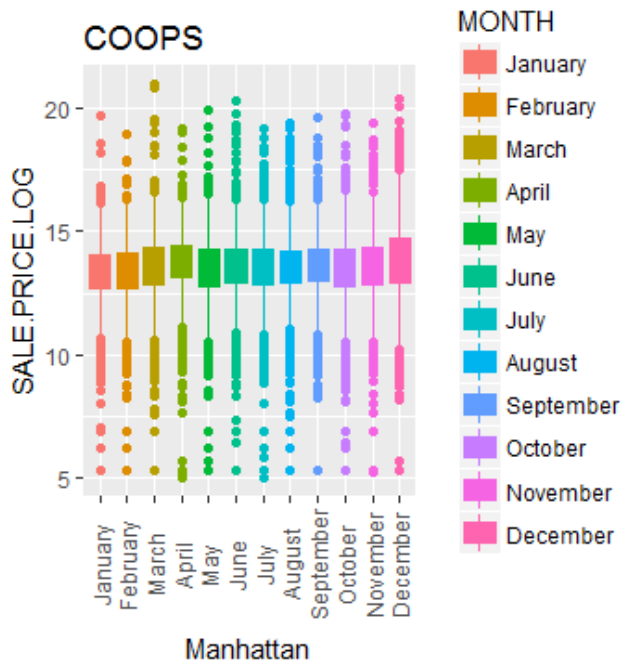
```
building_sale_month(manhattanTime, "FAMILY",
"Manhattan")
```



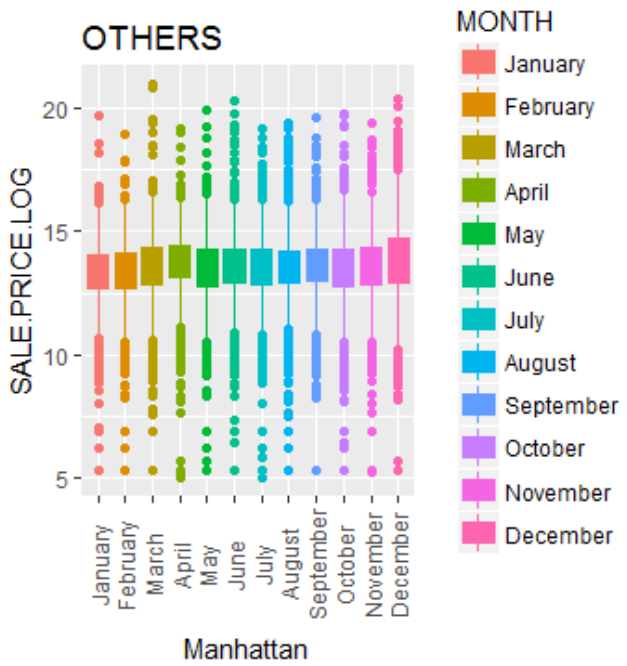

```
building_sale_month(manhattanTime, "CONDOS",
"Manhattan")
```



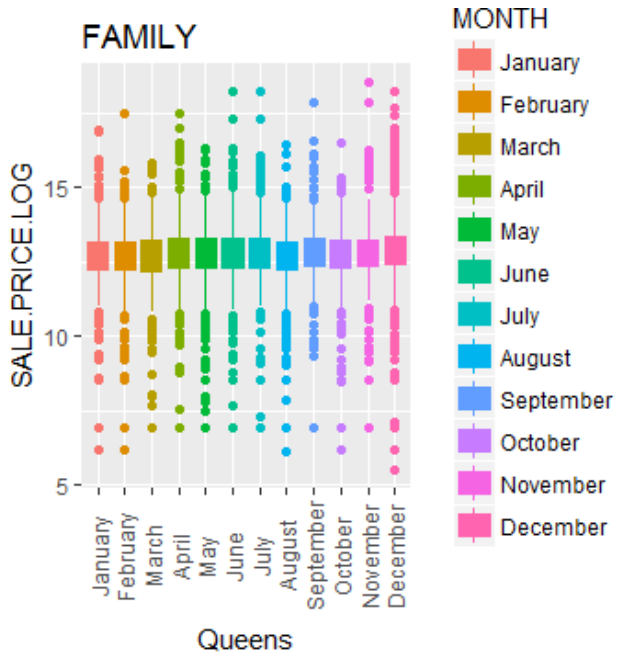

```
building_sale_month(manhattanTime, "COOPS",
"Manhattan")
```



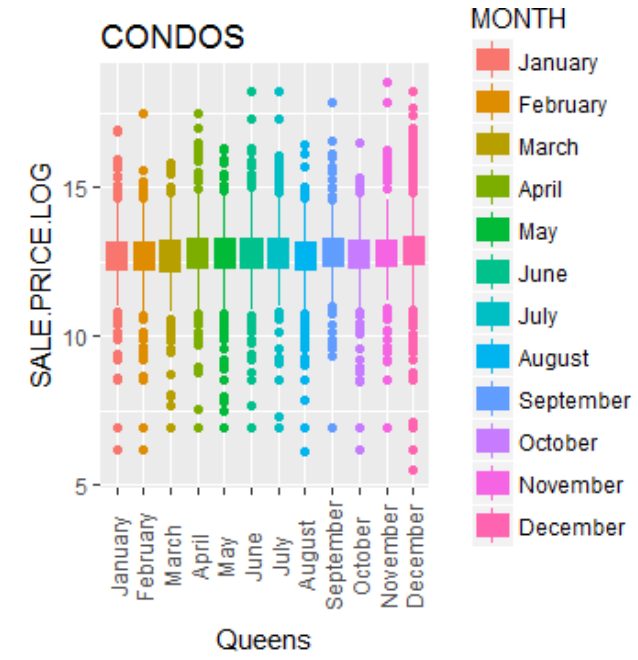

```
building_sale_month(manhattanTime, "OTHERS",
"Manhattan")
```



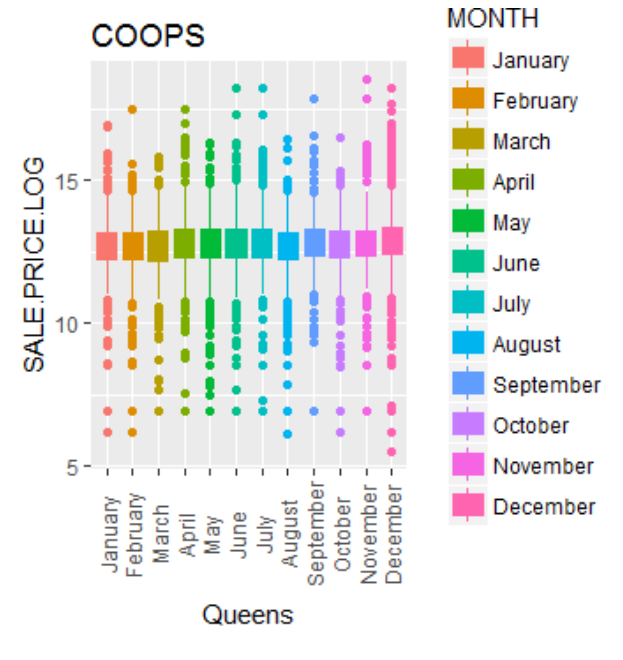
```
building_sale_month(queensTime, "FAMILY", "Queens")
```



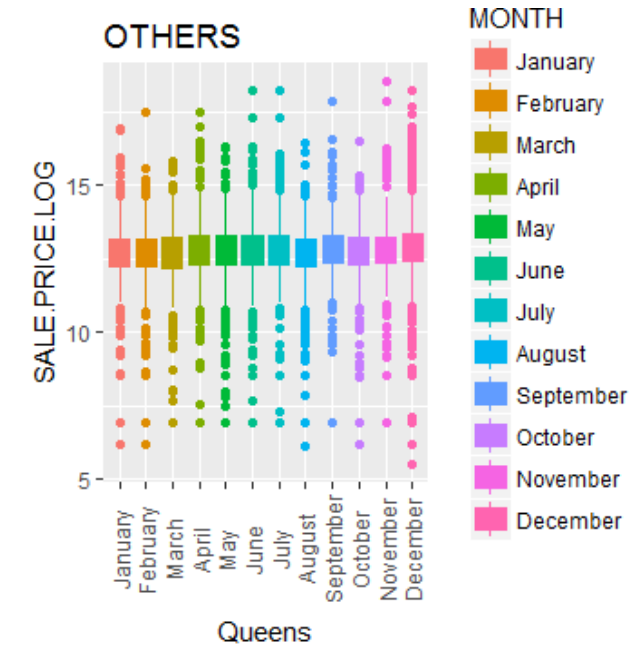

```
building_sale_month(queensTime, "CONDOS", "Queens")
```



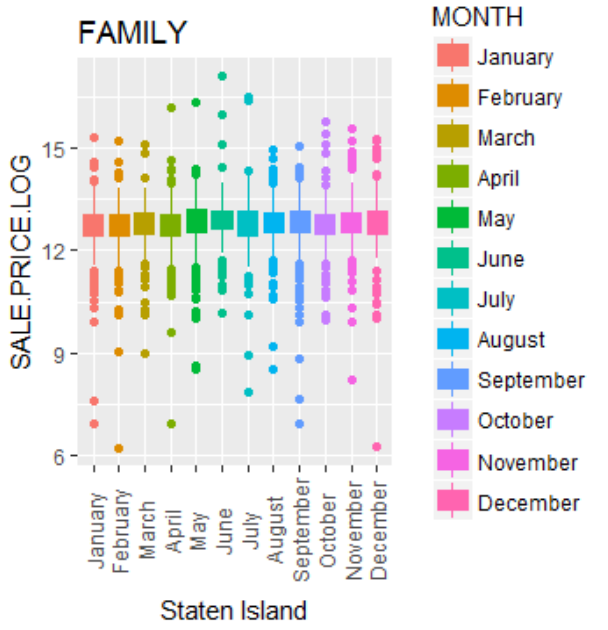

```
building_sale_month(queensTime, "COOPS", "Queens")
```



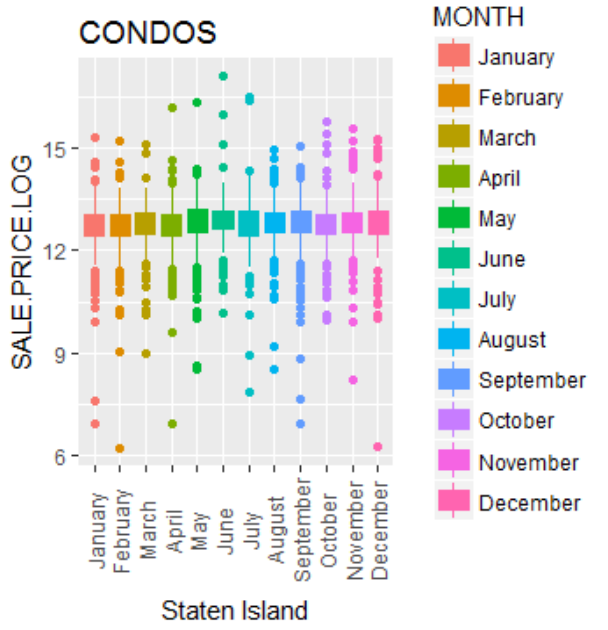

```
building_sale_month(queensTime, "OTHERS", "Queens")
```



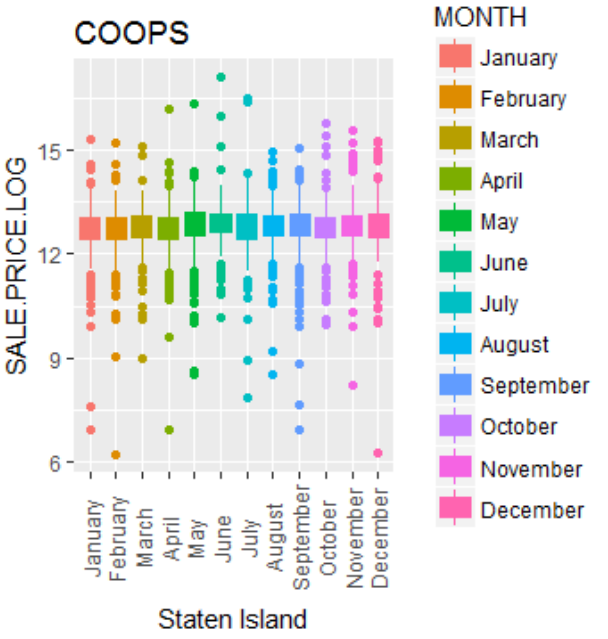
```
building_sale_month(statenTime, "FAMILY", "Staten Island")
```



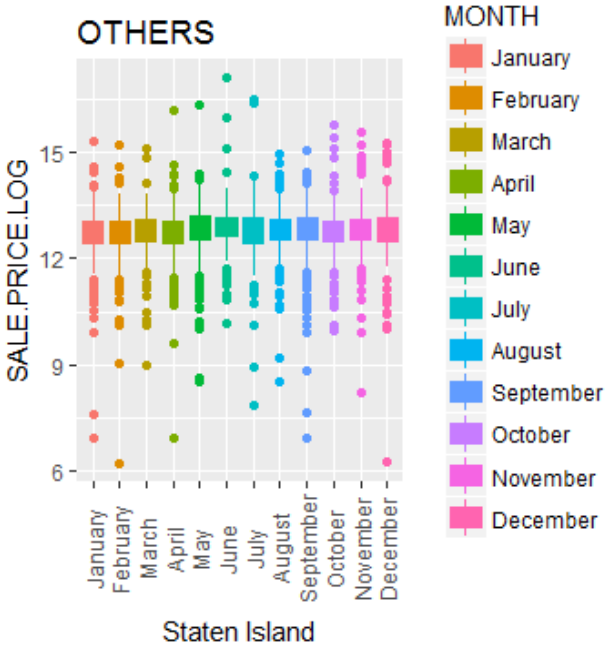

```
building_sale_month(statenTime, "CONDOS", "Staten Island")
```



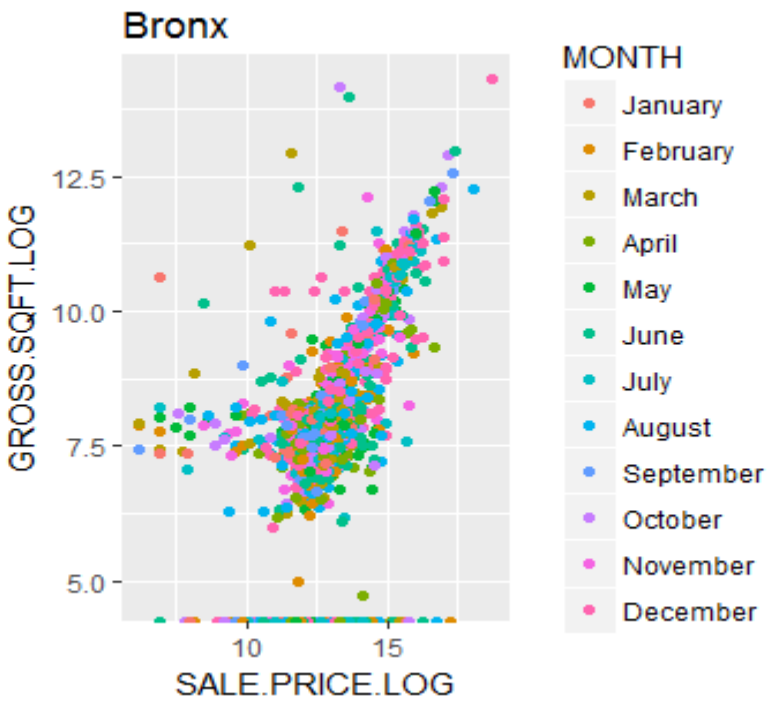

```
building_sale_month(statenTime, "COOPS", "Staten Island")
```



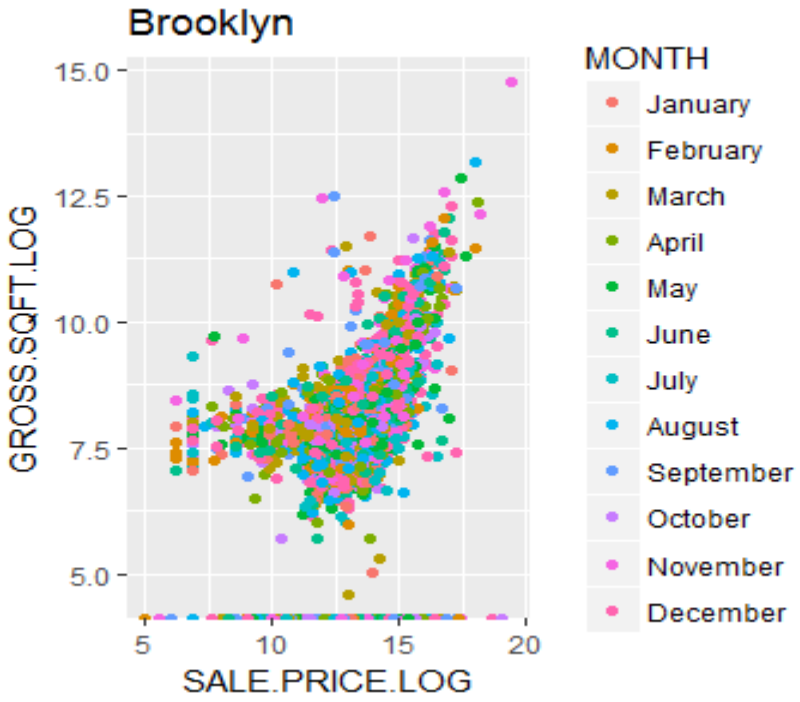

```
building_sale_month(statenTime, "OTHERS", "Staten Island")
```



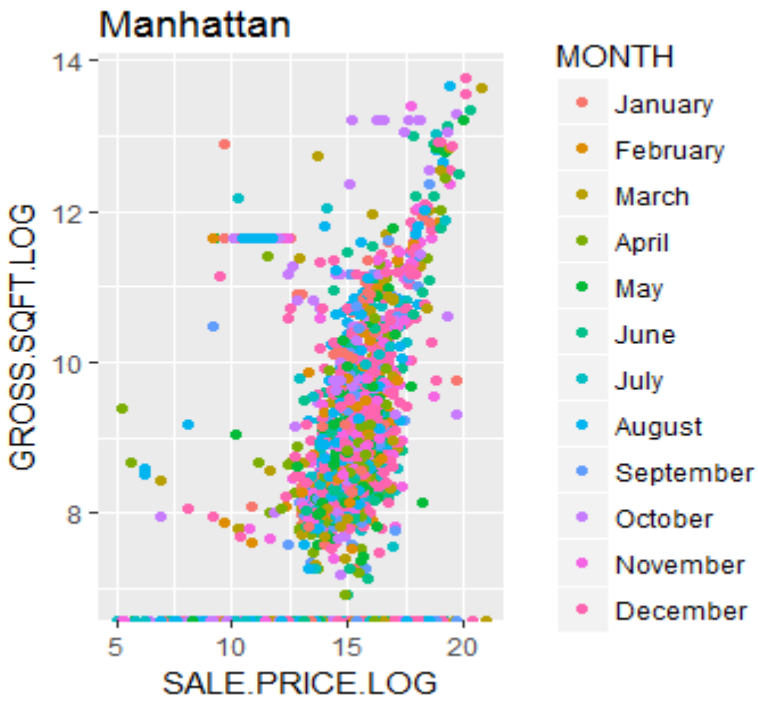
```
# PLOT: Sale Price v/s Gross Square feet for particular Borough
gross_sp_time(bronxTime, "Bronx")
```



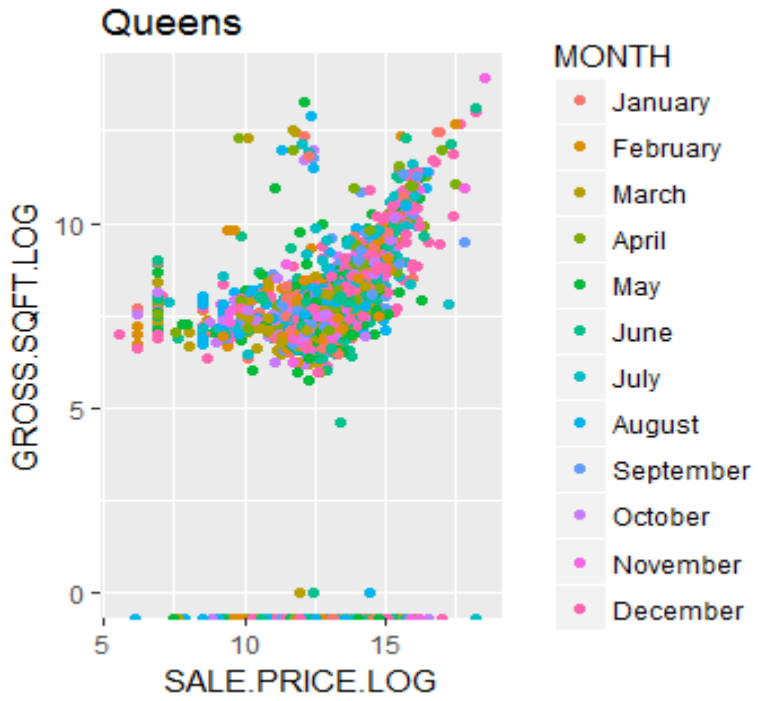

```
gross_sp_time(brooklynTime, "Brooklyn")
```



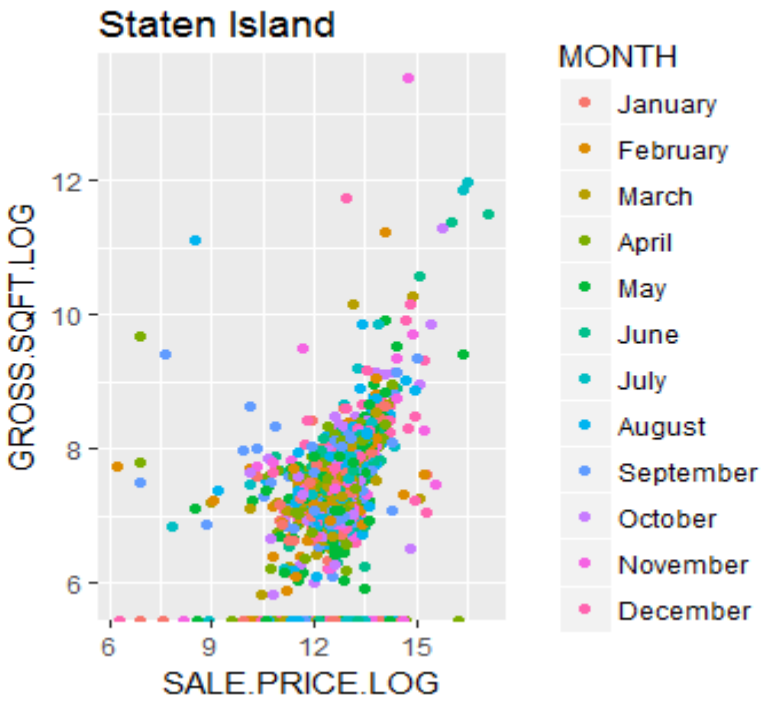
```
gross_sp_time(manhattanTime, "Manhattan")
```



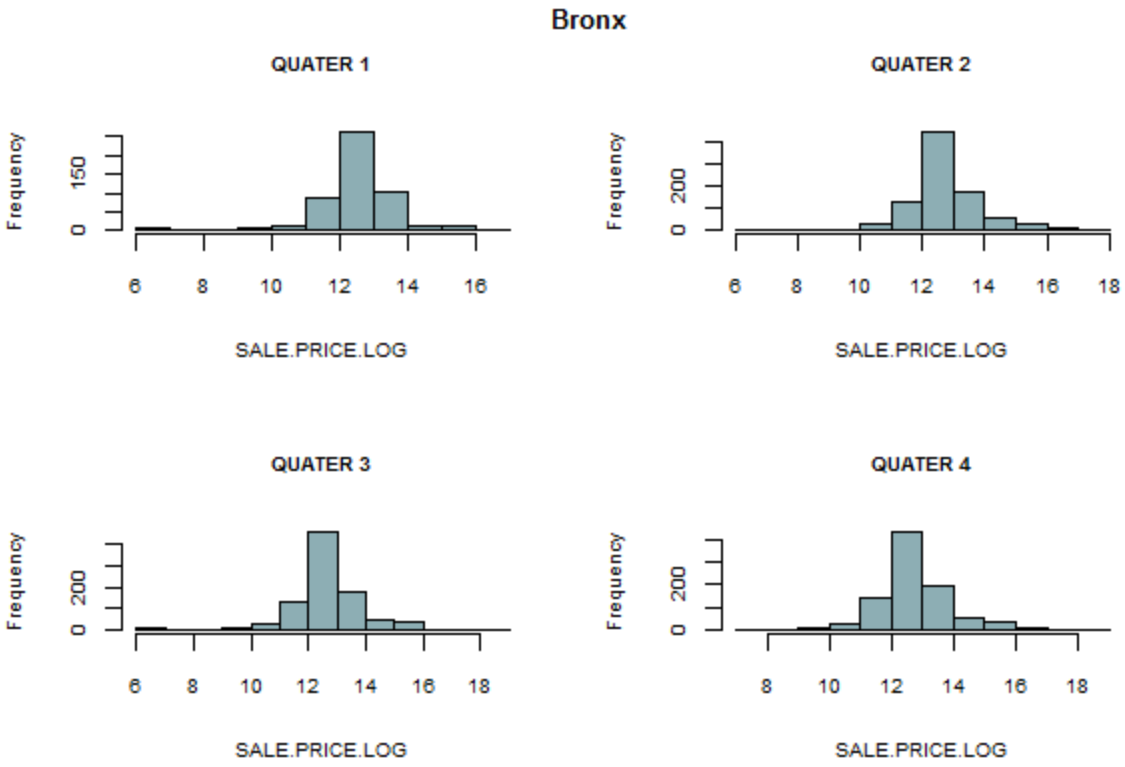

```
gross_sp_time(queensTime, "Queens")
```



```
gross_sp_time(statenTime, "Staten Island")
```



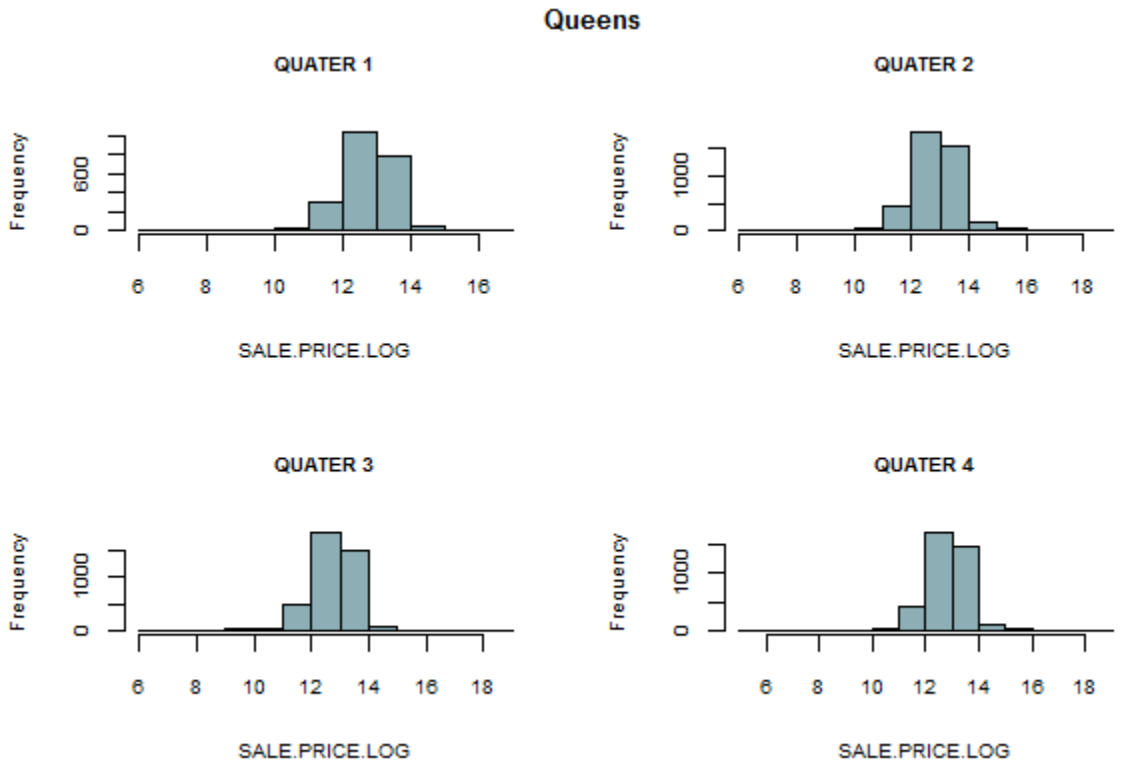

```
#PLOT: Sale Price frequency per quarter for particular Borough
sp_freq_time(bronxTime, "Bronx")
```



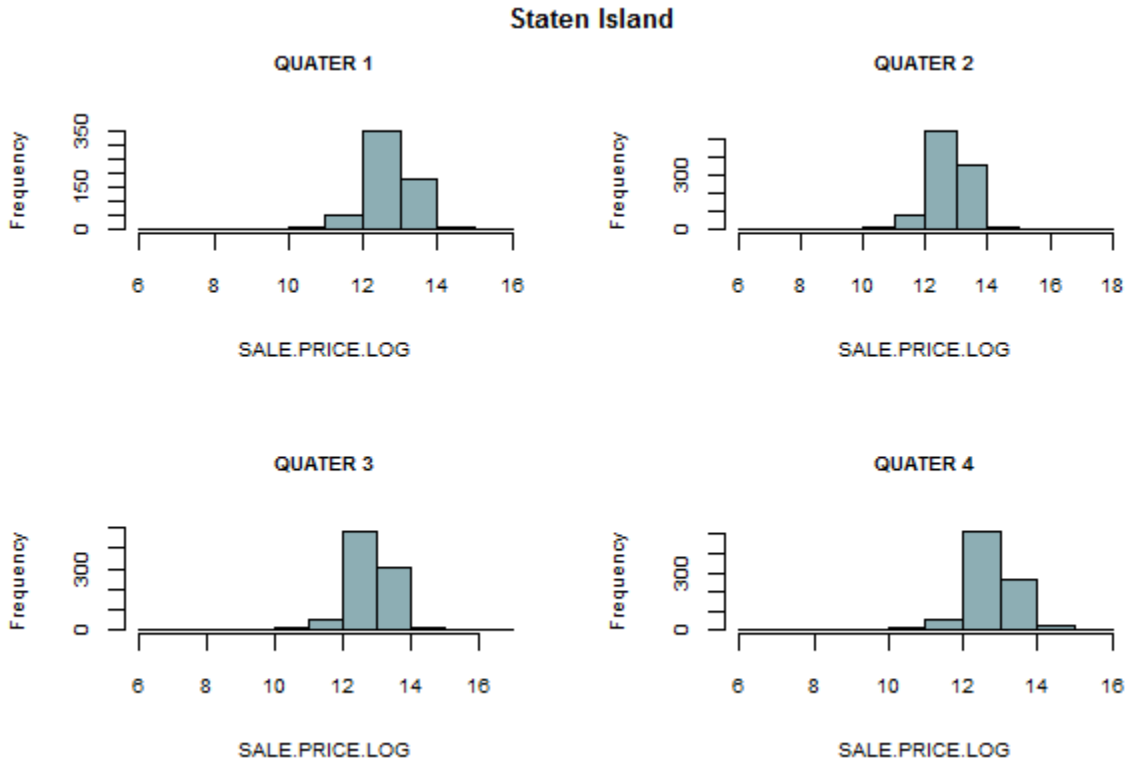
```
sp_freq_time(brooklynTime, "Brooklyn")
```


```
sp_freq_time(manhattanTime, "Manhattan")
```

```
sp_freq_time(queensTime, "Queens")
```




```
sp_freq_time(statenTime, "Staten Island")
```



#Summary

```
# SUMMARY: Sale prices and gross square feet across boroughs and time
summary_time_stats <- function(df, borough) {
  summaryBy(data = df, SALE.PRICE.N + GROSS.SQUARE.FEET.N ~ MONTH,
            FUN = c(length, mean, median),
            fun.names = c("Total no.", "Mean", "Median"),
            var.names = c(borough))
}
```

```
summary_time_stats(bronxTime, "Bronx")
```

MONTH	SALE.PRICE.N.Total no.	GROSS.SQUARE.FEET.N.Total no.	SALE.PRICE.N.Mean	GROSS.SQUARE.FEET.N.Mean	SALE.PRICE.N.Median	GROSS.SQUARE.FEET.N.Median
January	241	241	468661.7	3721.502	325000.0	1680.0
February	271	271	609003.9	2987.613	335000.0	1818.0
March	270	270	648749.2	5715.515	335000.0	1871.0
April	286	286	584264.4	3235.500	352500.0	1793.5
May	277	277	727687.5	4995.884	350000.0	1802.0
June	325	325	918962.9	10689.606	355100.0	1785.0
July	310	310	879052.8	6229.094	365000.0	1855.5
August	342	342	913196.5	5112.588	329671.0	1616.5
September	268	268	592973.7	4250.377	343808.5	1674.0
October	276	276	808232.1	11694.848	351961.0	1478.5
November	280	280	693206.0	6260.996	364291.5	2011.0
December	369	369	1559662.9	13214.870	360000.0	1998.0

```
summary_time_stats(brooklynTime, "Brooklyn")
```

MONTH	SALE.PRICE.N.Total no.	GROSS.SQUARE.FEET.N.Total no.	SALE.PRICE.N.Mean	GROSS.SQUARE.FEET.N.Mean	SALE.PRICE.N.Median	GROSS.SQUARE.FEET.N.Median
January	1084	1084	690759.3	2056.320	500000.0	1310.5
February	961	961	861892.4	2579.260	493500.0	1530.0
March	1109	1109	776480.0	2258.821	480299.0	1600.0
April	1143	1143	818587.8	2173.326	525000.0	1512.0
May	1343	1343	825509.3	2495.027	525000.0	1668.0
June	1379	1379	799055.6	2279.107	535000.0	1404.0
July	1213	1213	846016.2	1894.715	575201.0	1224.0
August	1514	1514	774728.2	2404.673	526735.5	1366.5
September	1038	1038	830116.5	2383.630	541109.0	1351.0
October	944	944	910898.5	2159.425	499999.5	1511.0
November	1044	1044	1187286.6	5981.781	530250.0	1646.0
December	1490	1490	1077395.2	3232.554	585000.0	1920.0

```
summary_time_stats(manhattanTime, "Manhattan")
```

MONTH	SALE.PRICE.N.Total no.	GROSS.SQUARE.FEET.N.Total no.	SALE.PRICE.N.Mean	GROSS.SQUARE.FEET.N.Mean
January	1256	1256	1717979	10815.642
February	1165	1165	1785443	14651.658
March	1283	1283	4253838	11335.376
April	1404	1404	2077174	1526.558
May	1869	1869	1944212	9152.575
June	1888	1888	2848222	5984.824
July	1928	1928	1803474	11346.335
August	2254	2254	2040651	6973.436
September	1179	1179	2401805	4182.036
October	1425	1425	2670320	11310.987
November	1381	1381	2620176	13020.017
December	2368	2368	4506253	11871.874

SALE.PRICE.N.Median	GROSS.SQUARE.FEET.N.Median
650000	0
660000	0
750000	0
857000	0
740000	0
775000	0
756500	0
763750	0
763687	0
718500	0
740000	0
880000	0

```
summary_time_stats(queensTime, "Queens")
```

MONTH	SALE.PRICE.N.Total no.	GROSS.SQUARE.FEET.N.Total no.	SALE.PRICE.N.Mean	GROSS.SQUARE.FEET.N.Mean
January	1145	1145	478277.4	2098.355
February	1087	1087	478506.1	1866.179
March	1114	1114	440126.2	2022.285
April	1248	1248	581658.5	2262.099
May	1456	1456	509883.8	1854.923
June	1364	1364	618452.9	2232.905
July	1320	1320	615280.0	1908.843
August	1612	1612	439368.4	1674.073
September	1121	1121	570180.1	1827.181
October	1175	1175	462252.1	1554.170
November	1110	1110	661170.6	2524.911
December	1519	1519	853460.8	3034.942

SALE.PRICE.N.Median	GROSS.SQUARE.FEET.N.Median
360000	1203
354900	1230
353750	1232
380000	1296
385000	1170
390000	1244
380000	1088
350000	1120
400000	1296
379000	1248
381500	1202
410000	1364

```
summary_time_stats(statenTime, "Staten Island")
```

MONTH	SALE.PRICE.N.Total no.	GROSS.SQUARE.FEET.N.Total no.	SALE.PRICE.N.Mean	GROSS.SQUARE.FEET.N.Mean
January	308	308	404432.1	1571.409
February	310	310	395660.7	1852.584
March	289	289	403977.7	1834.149
April	321	321	428267.2	1679.112
May	379	379	451718.9	1818.456
June	327	327	549079.0	2372.239
July	213	213	551156.7	3068.606
August	372	372	431252.7	1937.720
September	299	299	432454.5	1862.341
October	304	304	437521.6	2010.372
November	234	234	476889.4	4830.500
December	337	337	464574.8	2223.466

SALE.PRICE.N.Median	GROSS.SQUARE.FEET.N.Median
361500	1570.0
375500	1564.0
360000	1584.0
370000	1600.0
399000	1660.0
409000	1670.0
380000	1449.0
390000	1623.0
376292	1670.0
368500	1600.5
385000	1663.5
380000	1540.0

Problem 2 –

The datasets provided nyt1.csv, nyt2.csv, and nyt3.csv represents three (simulated) days of ads shown and clicks recorded on the New York Times homepage. Each row represents a single user. There are 5 columns: age, gender (0=female, 1=male), number impressions, number clicks, and logged-in. Use R to handle this data. Perform some exploratory data analysis:

- Create a new variable, age_group, that categorizes users as "<20", "20-29", "30-39", "40-49", "50-59", "60-69", and "70+".
- For each day:
    - Plot the distribution of numbers of impressions and click-through-rate (CTR = #clicks/#impressions) for these age categories
    - Define a new variable to segment or categorize users based on their click behavior.
    - Explore the data and make visual and quantitative comparisons across user segments/demographics (<20-year-old males versus <20-year-old females or logged-in versus not, for example).
- Extend your analysis across days. Visualize some metrics and distributions over time.

Analysis 2 –

```
# CLEAR WORK DIRECTORY
rm(list=ls())
setwd("C:/Users/Paras Garg/Documents/R/FE Assignments")

# USING PACKAGES
library('doBy')
library('ggplot2')
library('plyr')

### LOADING DATASETS
day_1 <- data.frame(read.csv("nyt1.csv"))
day_2 <- data.frame(read.csv("nyt2.csv"))
day_3 <- data.frame(read.csv("nyt3.csv"))
```

Functions

```
# FUNCTION: Create new variable 'AGE_GROUP'
age_var <- function (df) {
  df$Age_Group[df$Age < 20] <- '<20'
  df$Age_Group[df$Age >= 20 & df$Age < 30] <- '20-29'
  df$Age_Group[df$Age >= 30 & df$Age < 40] <- '30-39'
  df$Age_Group[df$Age >= 40 & df$Age < 50] <- '40-49'
  df$Age_Group[df$Age >= 50 & df$Age < 60] <- '50-59'
  df$Age_Group[df$Age >= 60 & df$Age < 70] <- '60-69'
  df$Age_Group[df$Age >= 70] <- '70+'
  return (df)
}

# FUNCTION: Data preparation by categorizing values
data_prep <- function (df) {
  df$Gender[df$Gender == 1] <- "Male"
  df$Gender[df$Gender == 0] <- "Female"
  df$Signed_In[df$Signed_In == 1] <- "Signed In"
  df$Signed_In[df$Signed_In == 0] <- "Not Signed In"
  return(df)
}
```

```
# FUNCTION: Create new variable according to clicks
click_var <- function (df) {
  df$Clicks_Cat <- cut(df$Clicks, c(-Inf,0,1,2,3,4,Inf), include.lowest = FALSE)
  return (df)
}
```

Data preparation

```
### Data Preparation
day_1 <- data_prep(age_var(day_1))
day_2 <- data_prep(age_var(day_2))
day_3 <- data_prep(age_var(day_3))
```

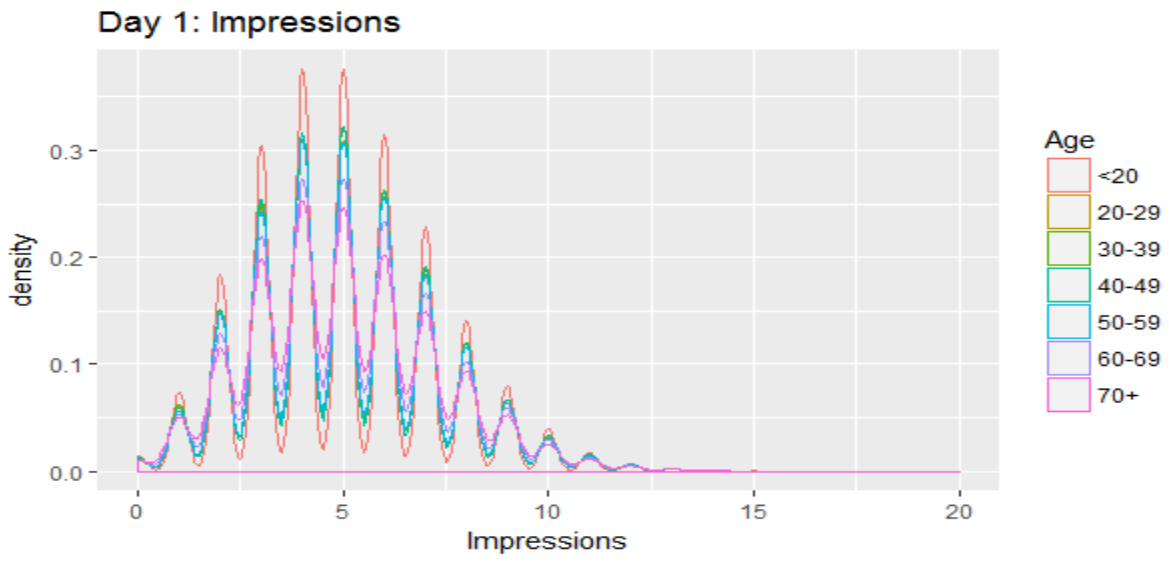
Impressions and CTR distribution for each day

#Functions

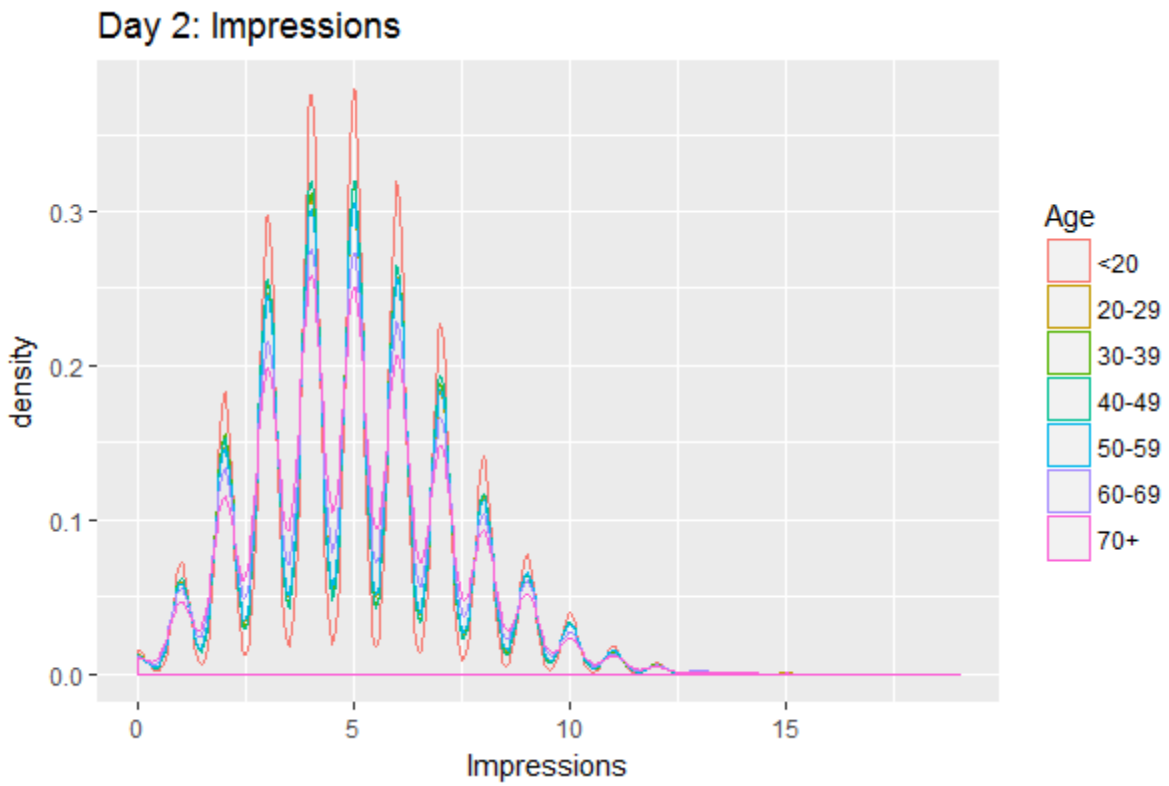
```
# FUNCTION: Plots for impressions
imp <- function(df, day) {
  ggplot(df, aes(x=Impressions, color=Age_Group)) + geom_density() +
    xlab("Impressions") + labs(color="Age") + ggtitle(paste(day, "Impressions", sep=": "))
}

# FUNCTION: Plots for CTR
ctr <- function(df, day, ref) {
  if (ref == "Imp") {
    ref <- "(Impressions > 0)"
  } else if (ref == "Clks") {
    ref <- "(Clicks > 0)"
  }
  ggplot(df, aes(x=Clicks/Impressions, color=Age_Group)) + geom_density() +
    xlab("Click-through-rate (CTR)") + labs(color="Age") +
    ggtitle(paste(day, "CTR", ref, sep=" : "))
}
```

#Plots

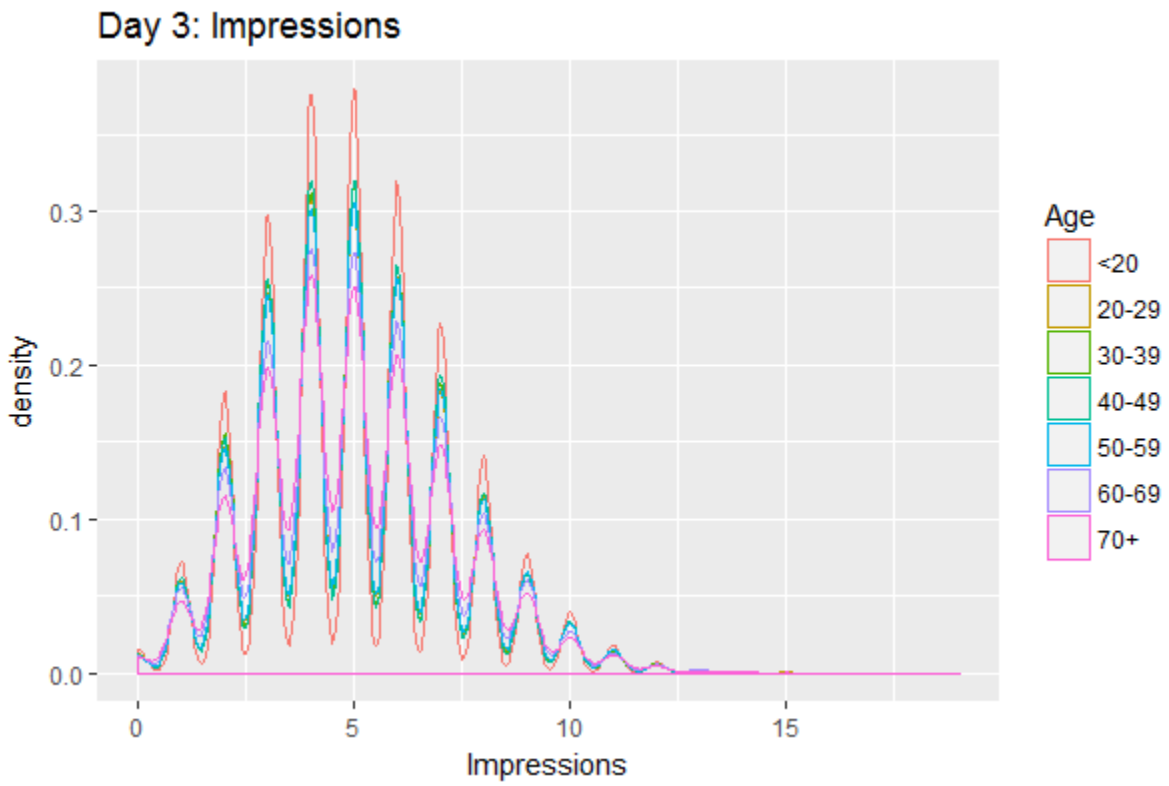
```
imp(day_1, "Day 1")
```



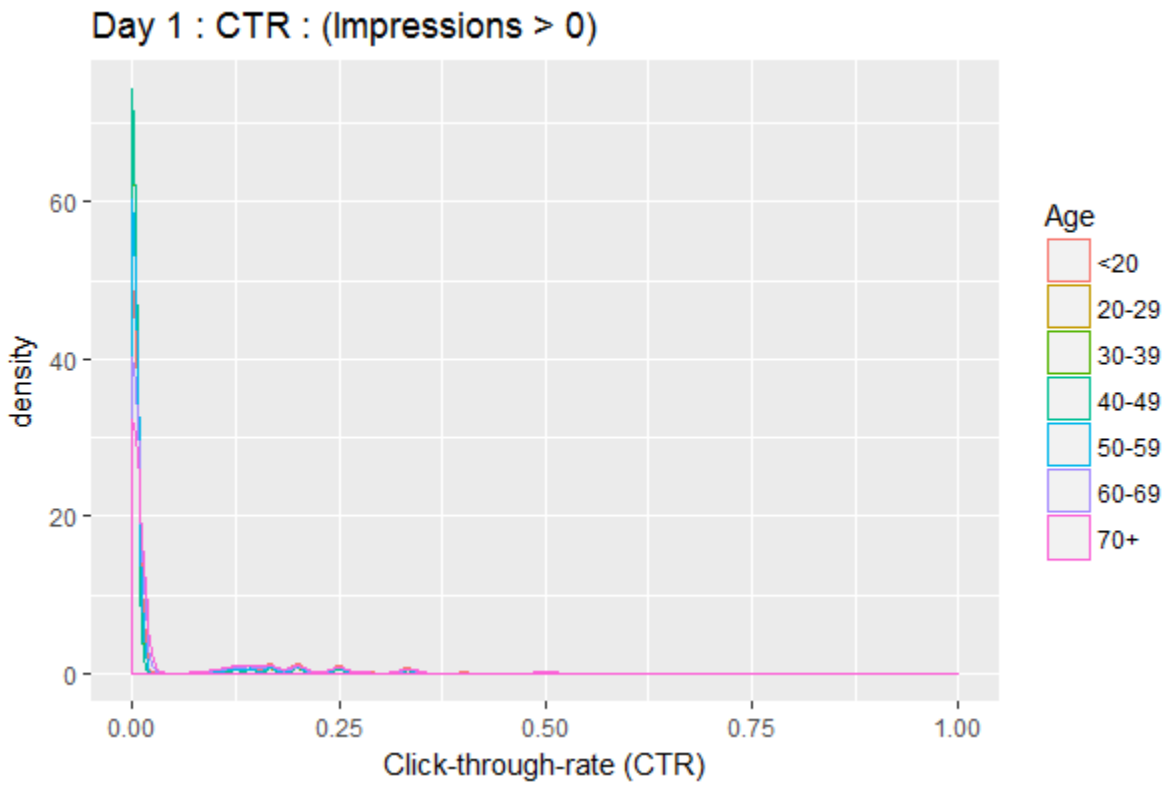
```
imp(day_2, "Day 2")
```



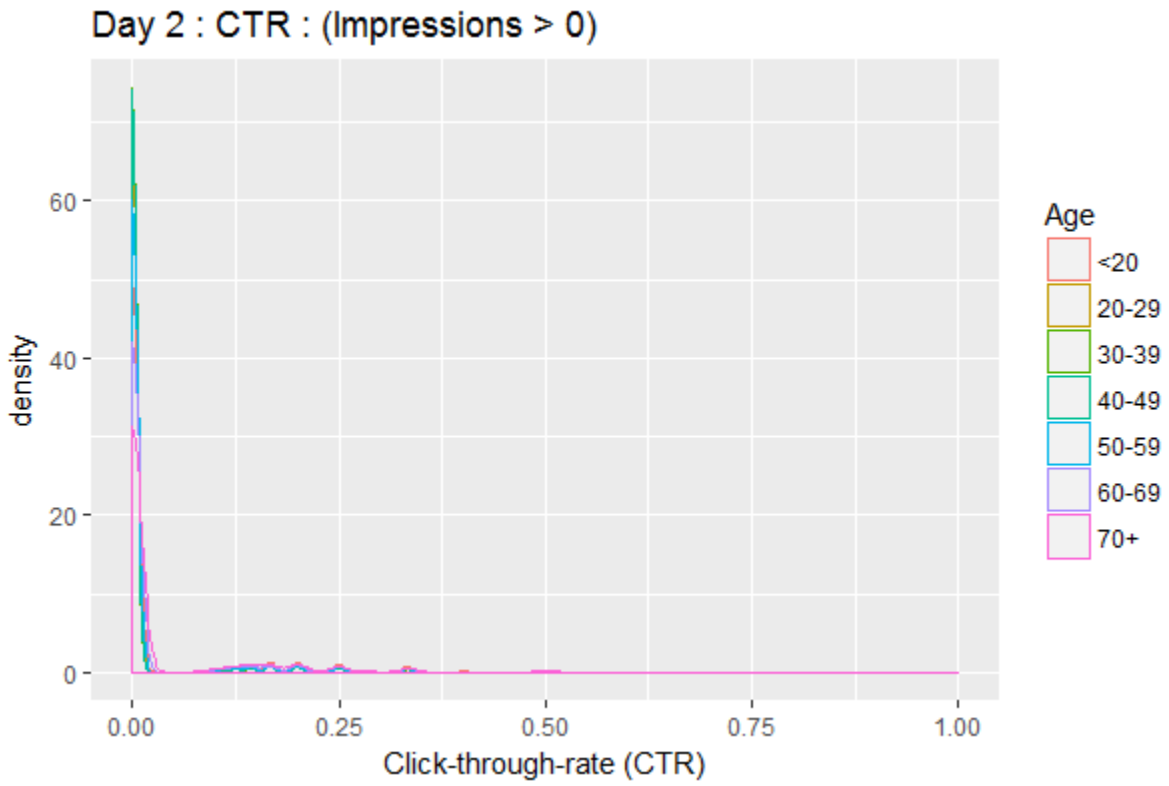

```
imp(day_2, "Day 3")
```



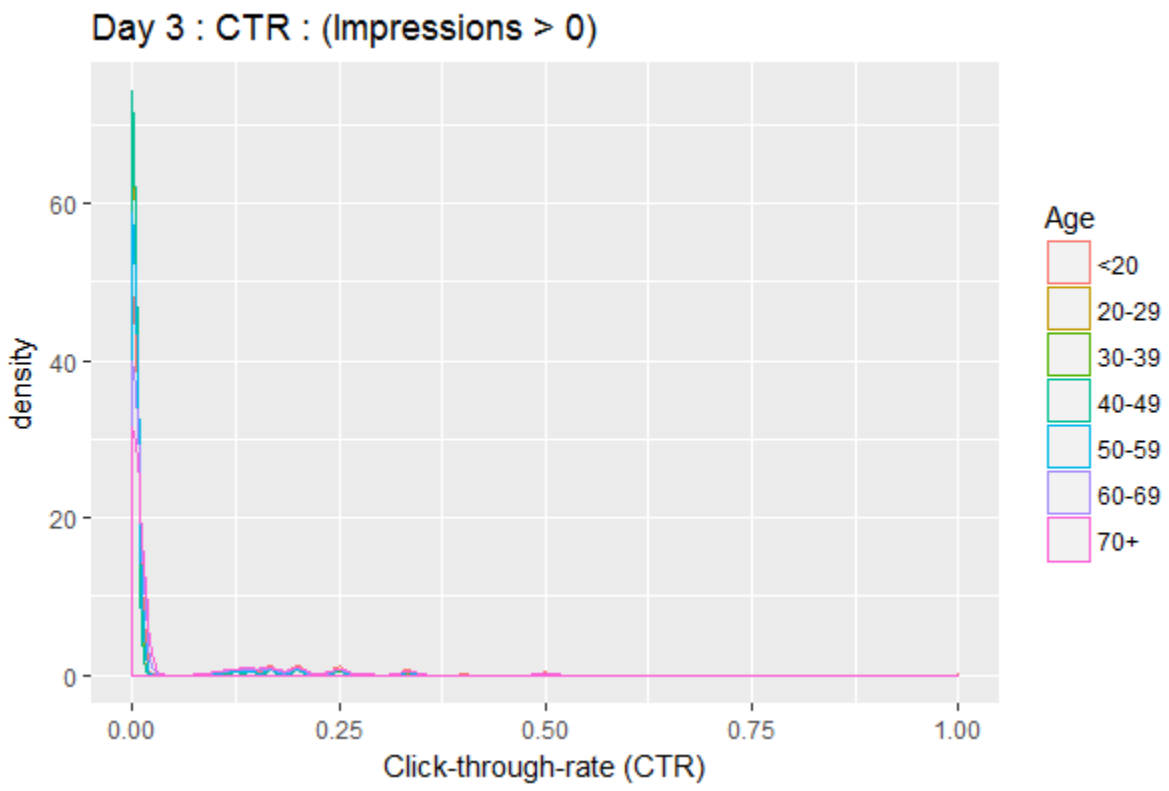
```
ctr(subset(day_1, Impressions > 0), "Day 1", "Imp")
```



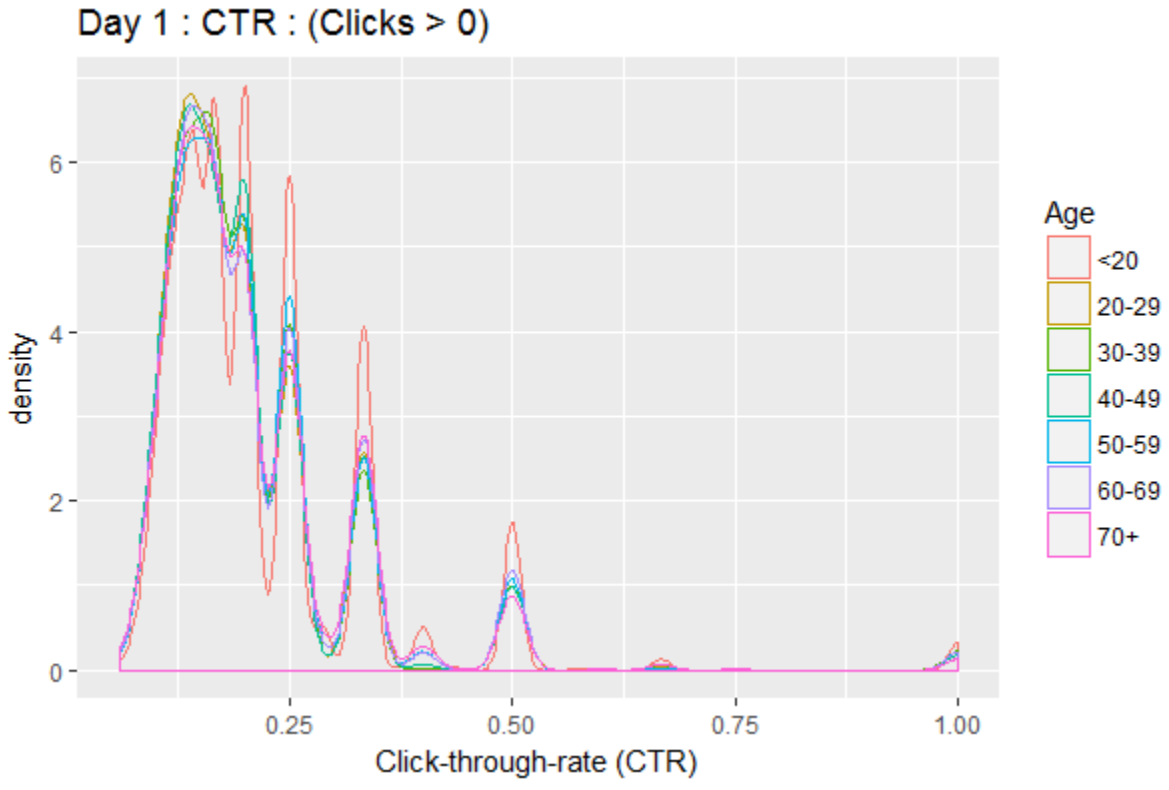

```
ctr(subset(day_2, Impressions > 0), "Day 2", "Imp")
```



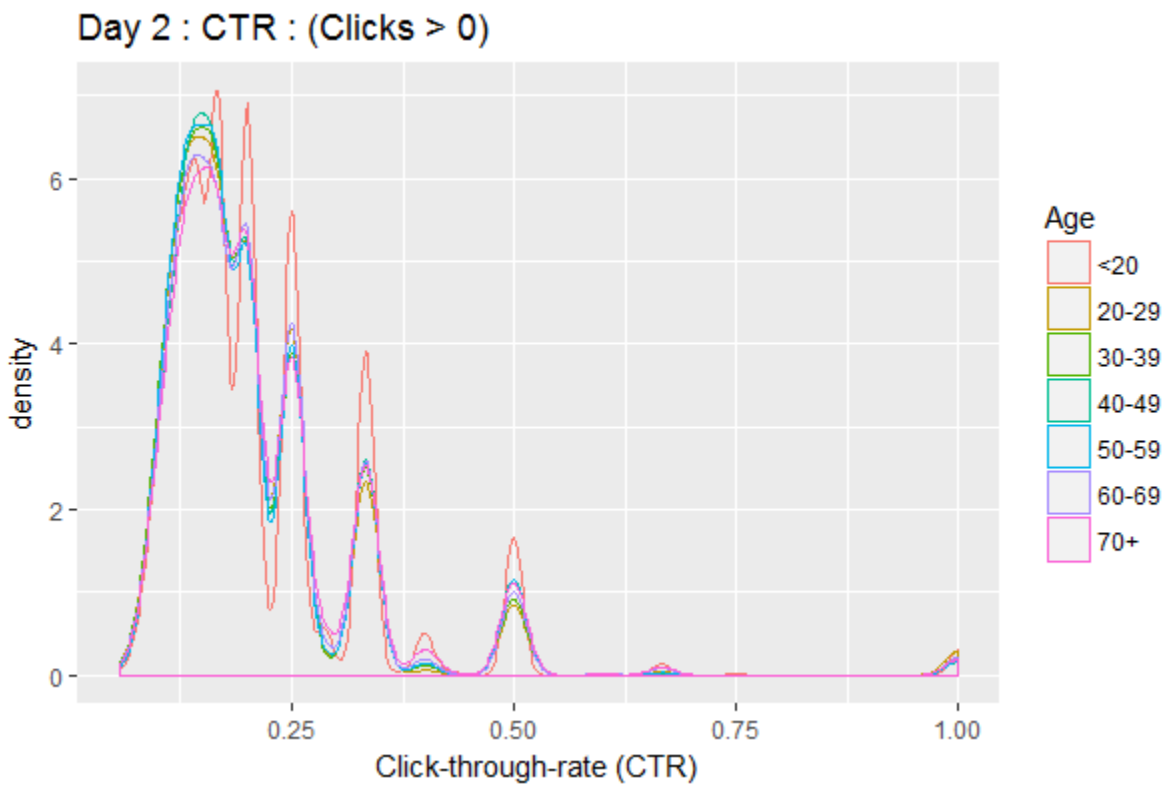
```
ctr(subset(day_3, Impressions > 0), "Day 3", "Imp")
```



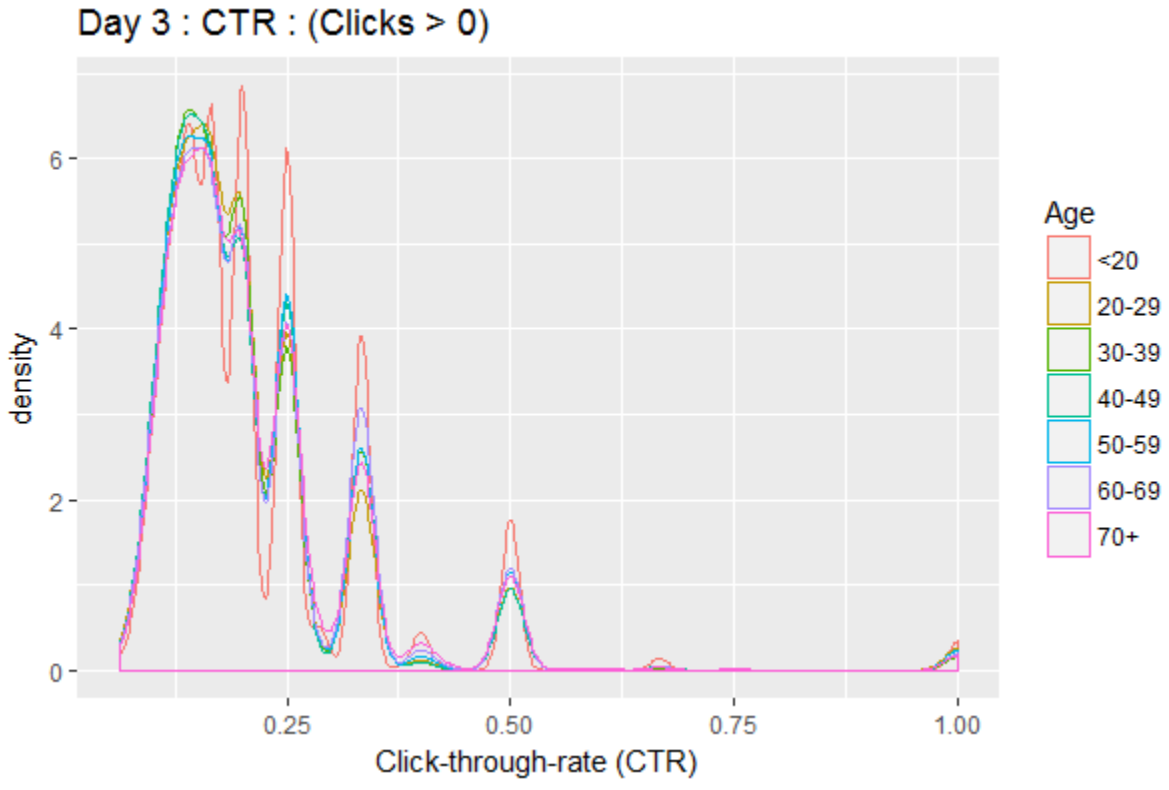

```
ctr(subset(day_1, Clicks > 0), "Day 1", "Clks")
```



```
ctr(subset(day_2, Clicks > 0), "Day 2", "Clks")
```




```
ctr(subset(day_3, Clicks > 0), "Day 3", "Clks")
```



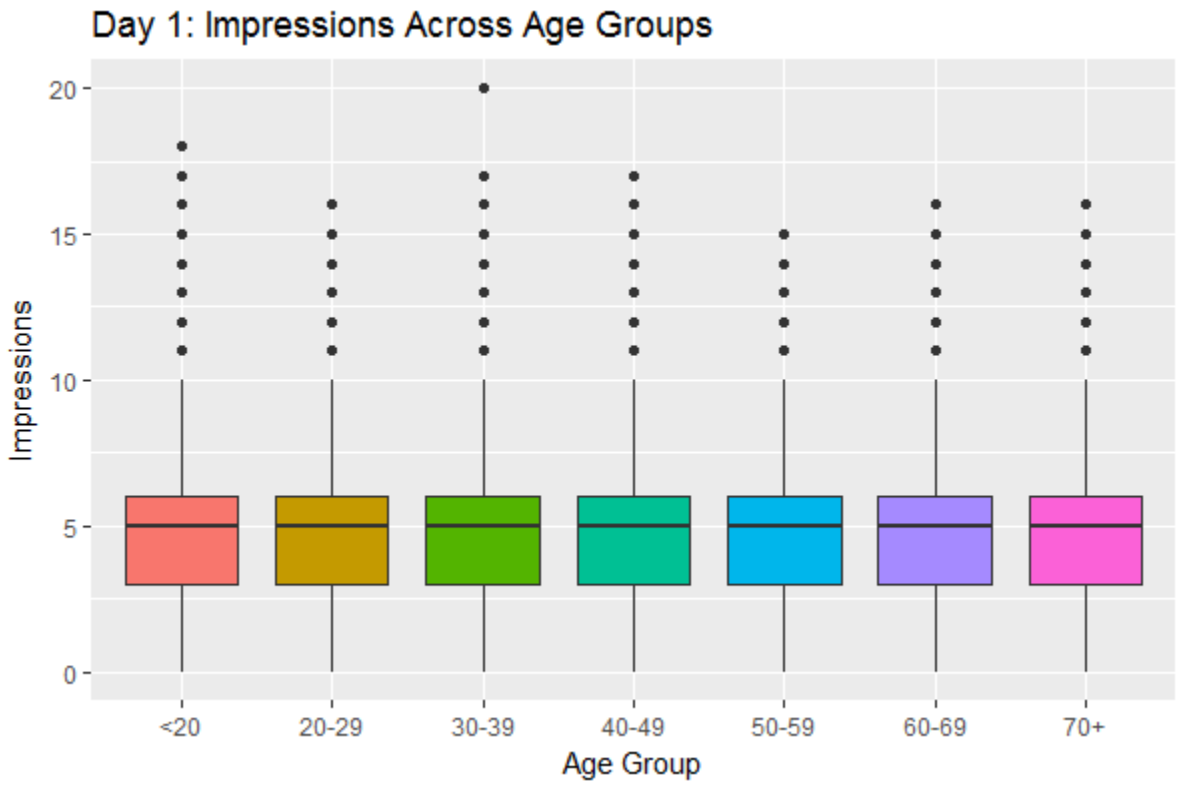
Impressions across Age Groups for each day

#Functions

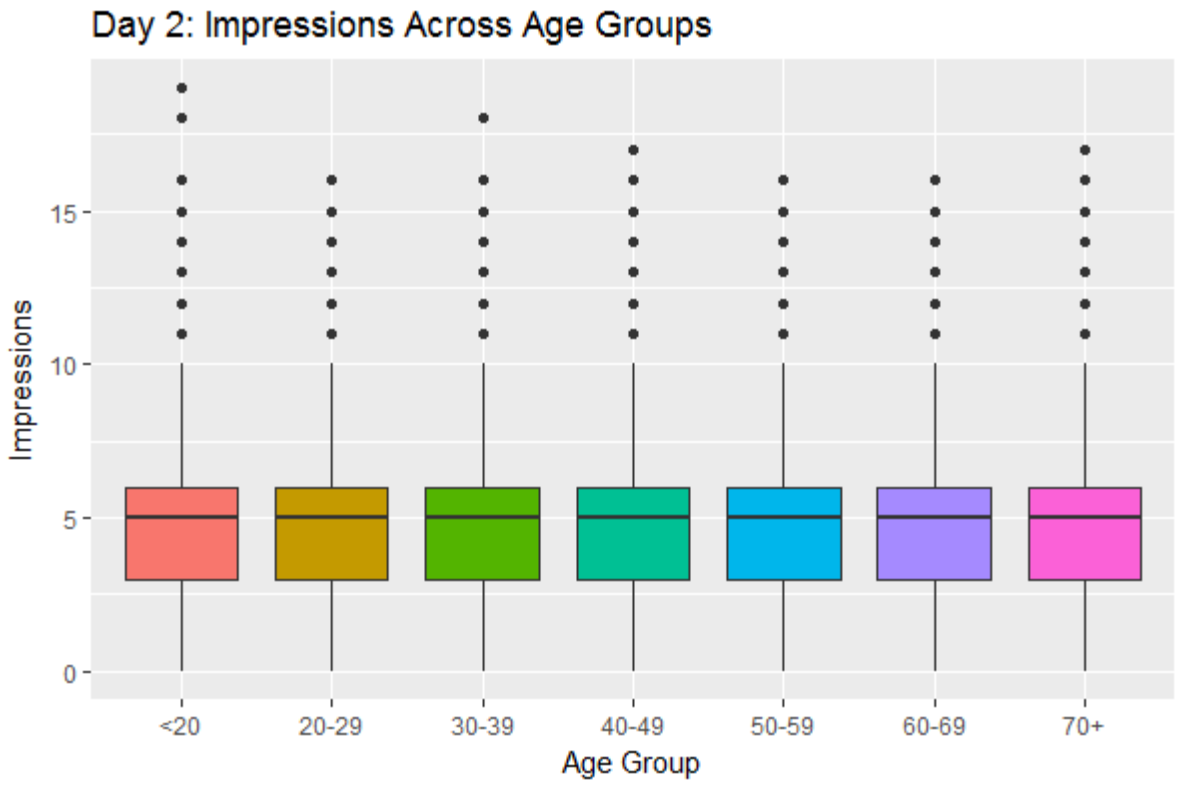
```
age_imp_boxplot <- function (df, day) {
  ggplot(df, aes(x = Age_Group, y = Impressions, fill=Age_Group)) +
    geom_boxplot() +
    xlab("Age Group") + guides(fill=FALSE) +
    ggtitle(paste(day, "Impressions Across Age Groups", sep=": "))
}

age_imp_hist <- function (df, day) {
  ggplot(df, aes(x = Impressions, fill = Age_Group)) +
    geom_histogram(bins = 15) + ggtitle(paste(day, "Impressions Across Age Groups", sep=": "))
}
```

#Plots

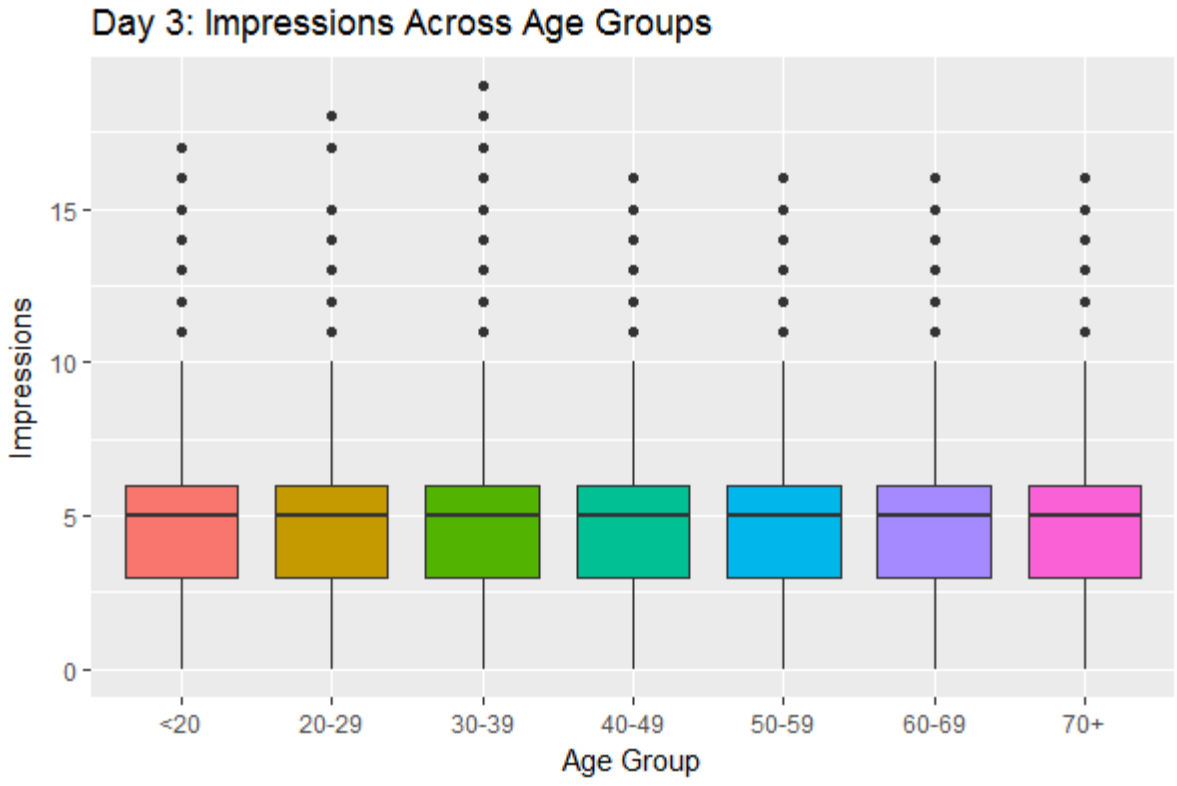
```
age_imp_boxplot(day_1, "Day 1")
```



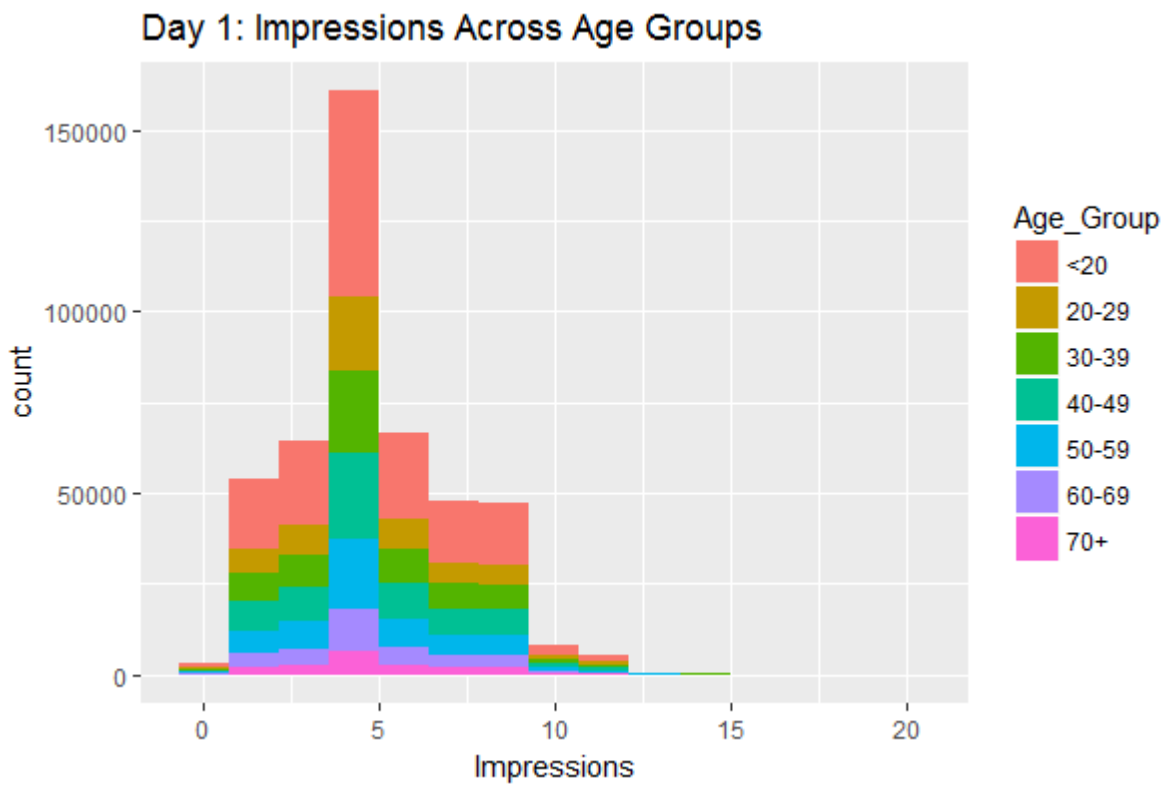
```
age_imp_boxplot(day_2, "Day 2")
```



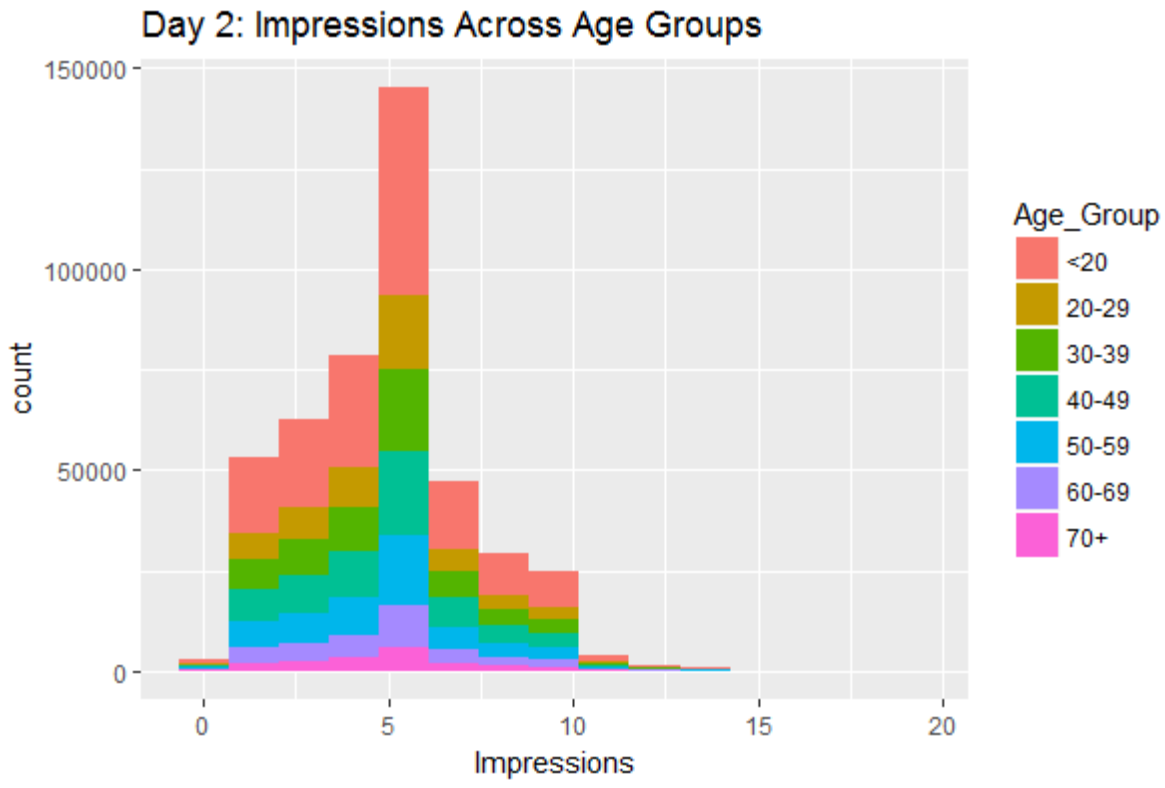

```
age_imp_boxplot(day_3, "Day 3")
```



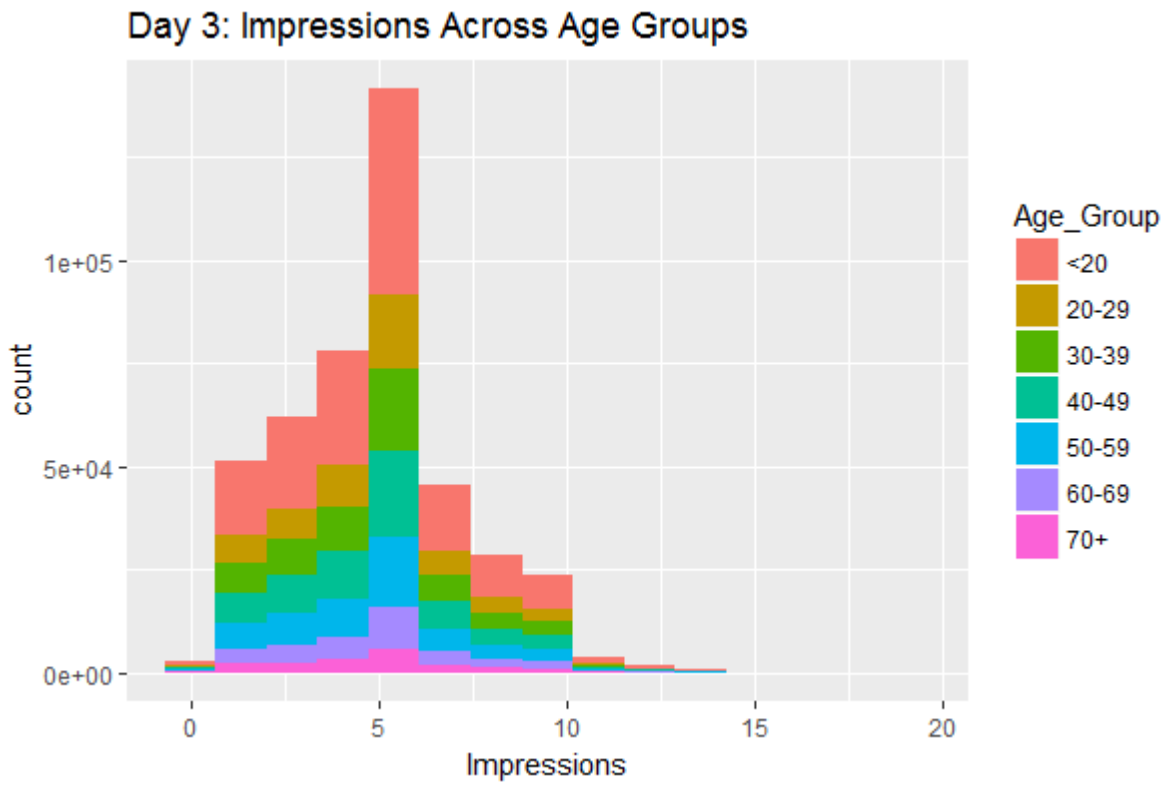
```
age_imp_hist(day_1, "Day 1")
```




```
age_imp_hist(day_2, "Day 2")
```



```
age_imp_hist(day_3, "Day 3")
```



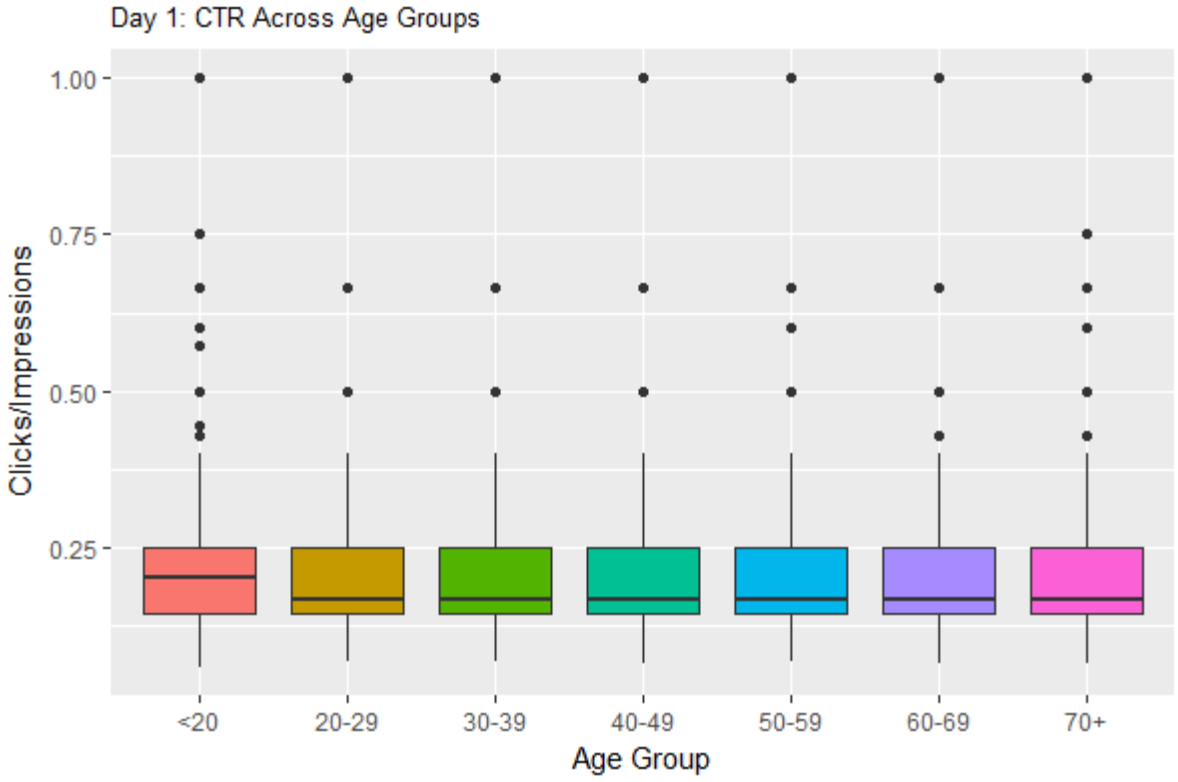
CTR across Age Groups for each day

#Functions

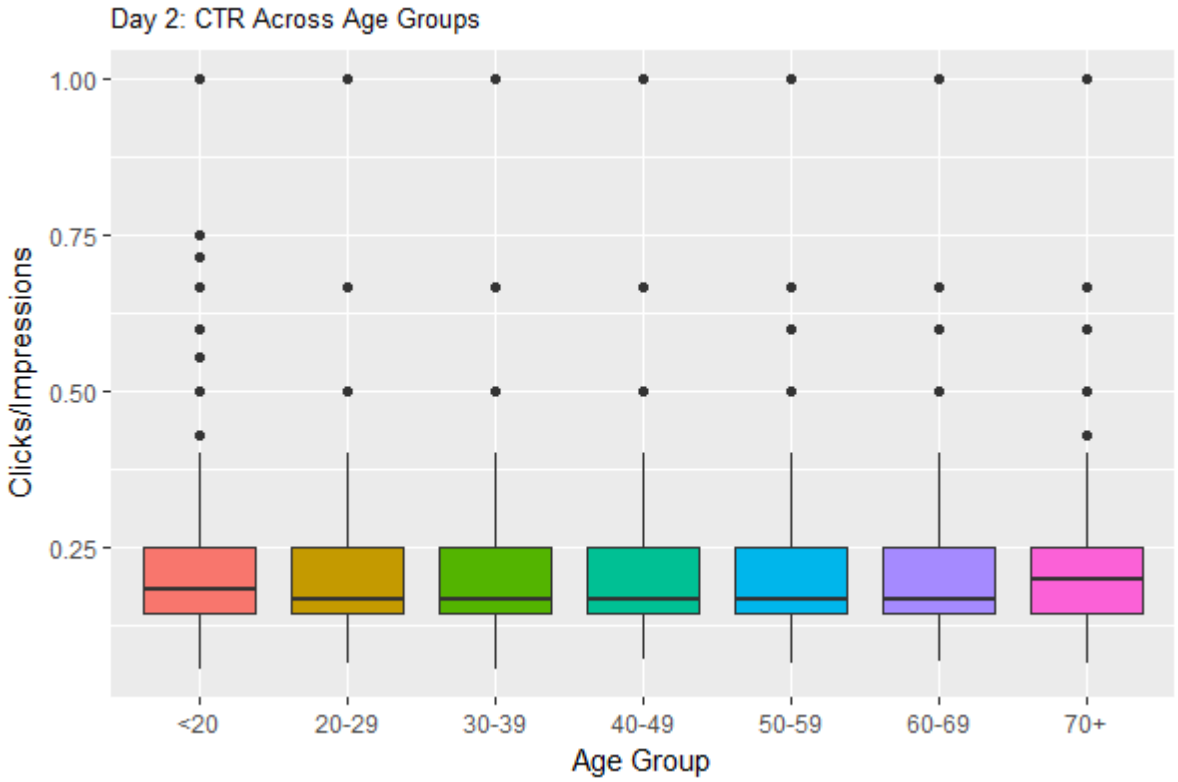
```
# Click Through Rate across Age Group
age_ctr_boxplot <- function (df, day) {
  ggplot(df, aes(x = Age_Group, y = Clicks/Impressions, fill=Age_Group)) +
    geom_boxplot() +
    xlab("Age Group") + guides(fill=FALSE) +
    ggtitle(paste(day, "CTR Across Age Groups", sep=": "))
}

age_ctr_hist <- function (df, day) {
  ggplot(df, aes(x = Clicks/Impressions, fill = Age_Group)) +
    geom_histogram(bins = 15) + ggtitle(paste(day, "CTR Across Age Groups", sep=": "))
}
```

#Plots

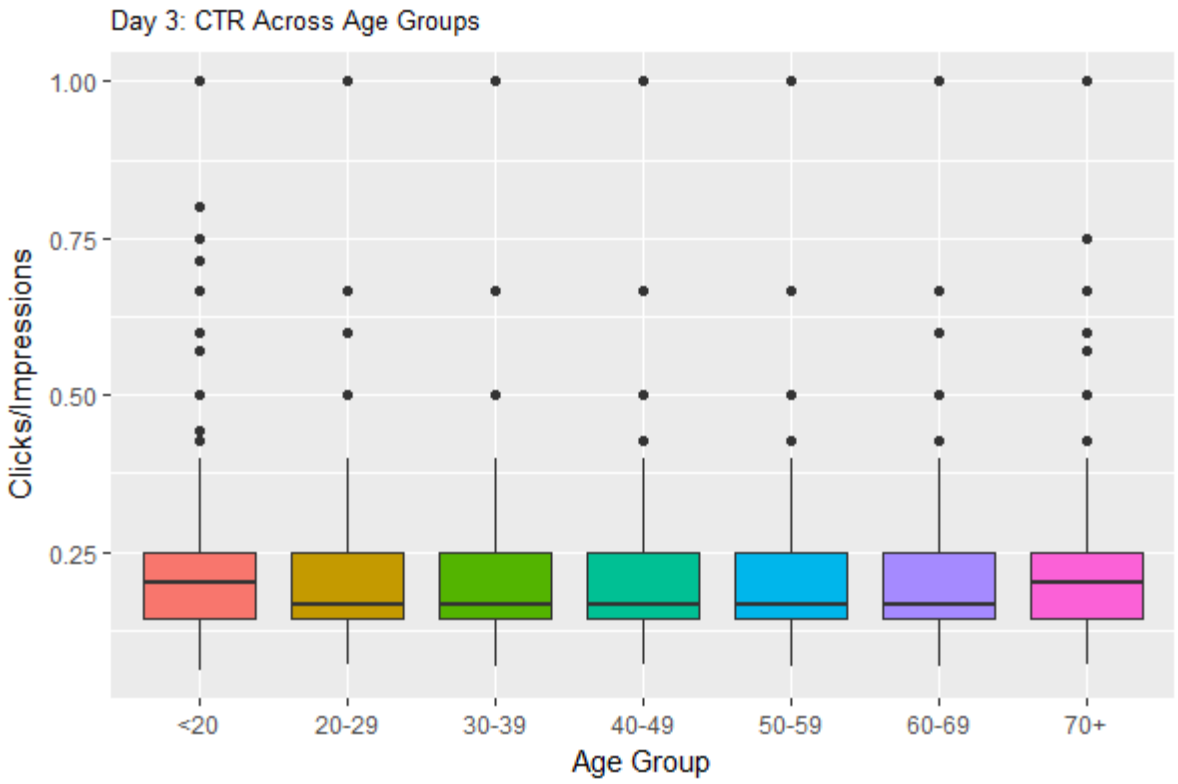
```
age_ctr_boxplot(subset(day_1, Clicks > 0 & Impressions > 0), "Day 1")
```



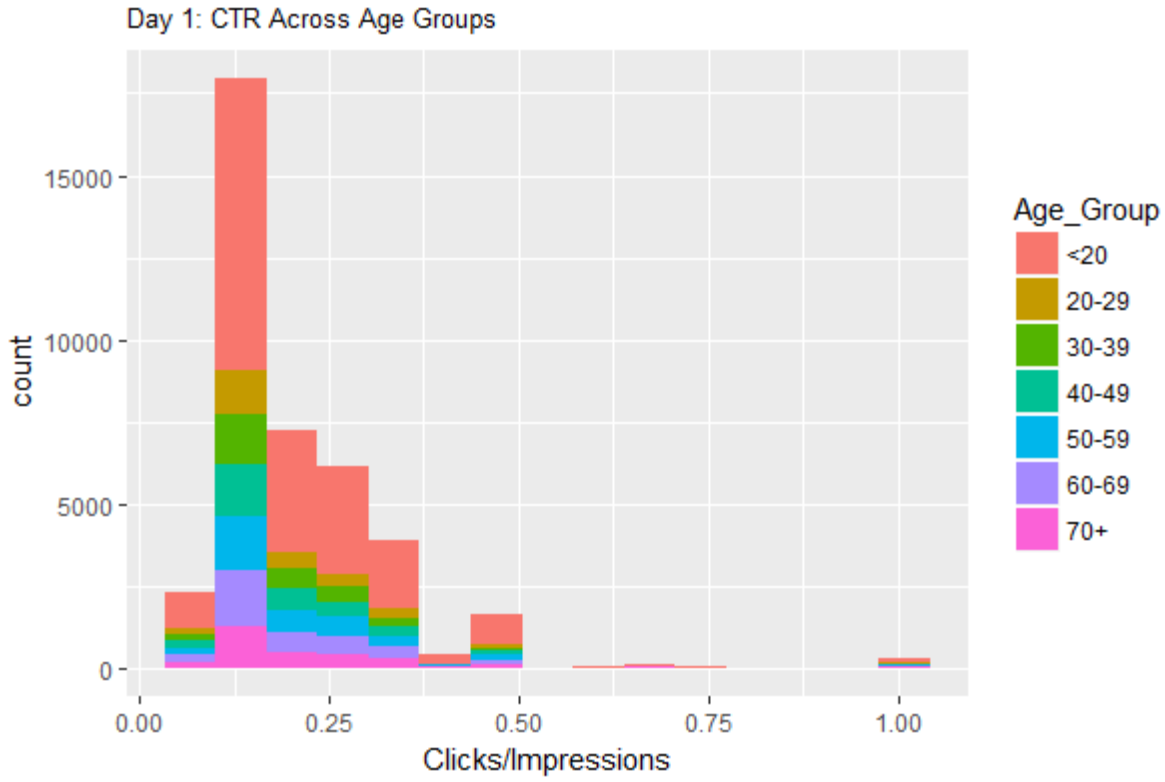

```
age_ctr_boxplot(subset(day_2, Clicks > 0 & Impressions > 0), "Day 2")
```



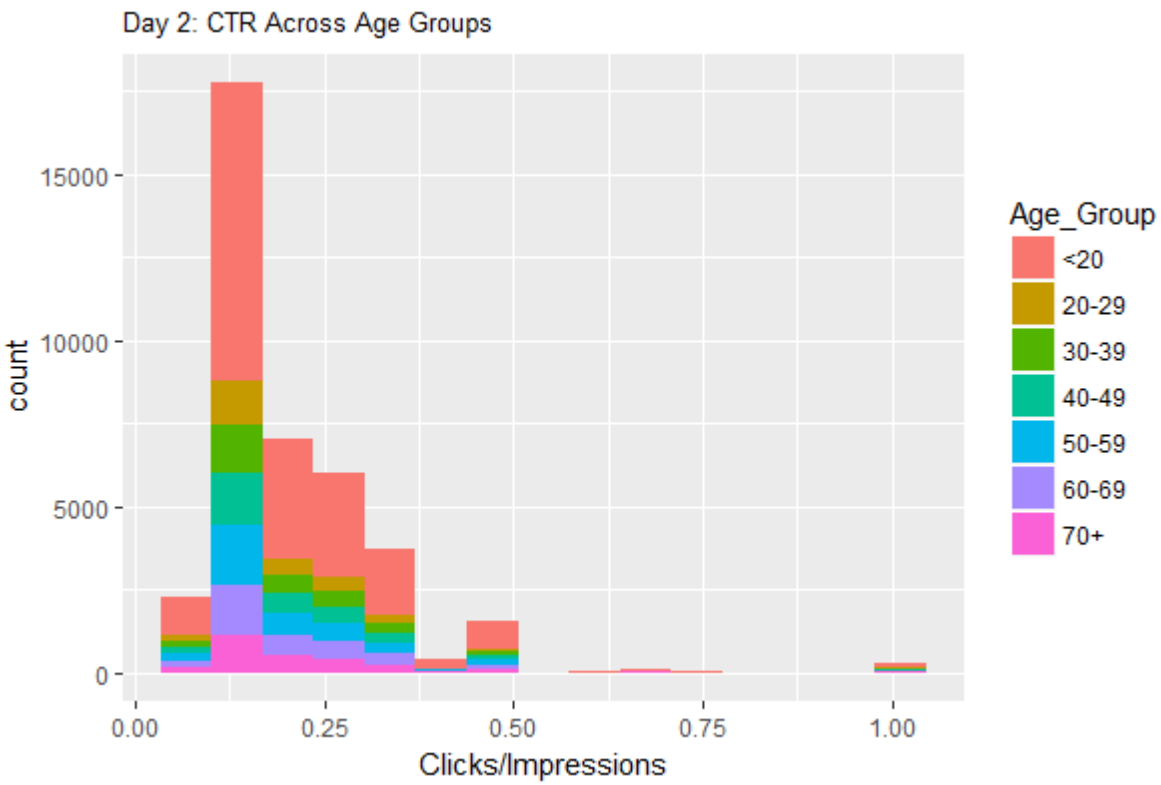
```
age_ctr_boxplot(subset(day_3, Clicks > 0 & Impressions > 0), "Day 3")
```



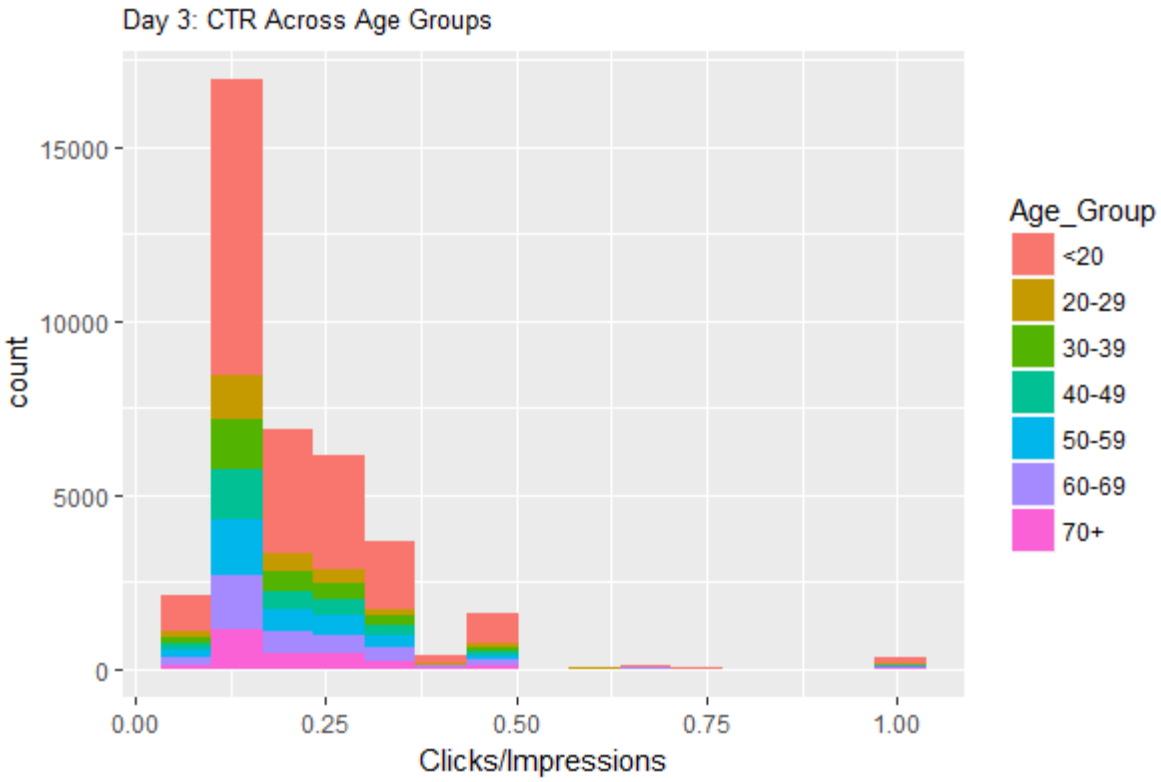

```
age_ctr_hist(subset(day_1, Clicks > 0 & Impressions > 0), "Day 1")
```



```
age_ctr_hist(subset(day_2, Clicks > 0 & Impressions > 0), "Day 2")
```




```
age_ctr_hist(subset(day_3, Clicks > 0 & Impressions > 0), "Day 3")
```



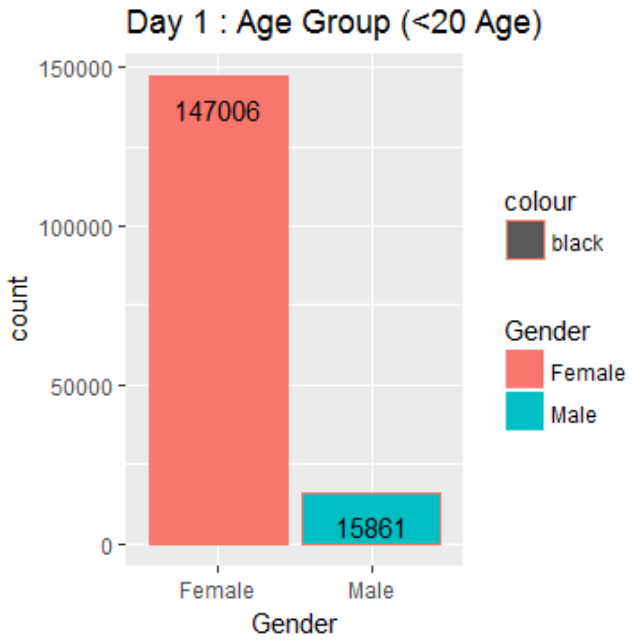
Comparisons across user segments/demographics

#Functions

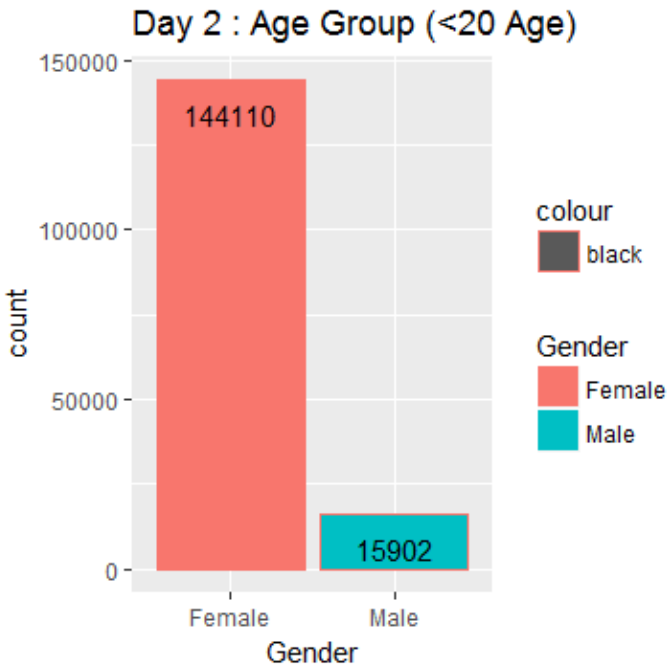
```
# Male v/s Female
male_vs_female <- function (df, age_grp, day) {
  df <- subset(df, Age_Group == age_grp)

  ggplot(df, aes(x = Gender, y = ..count..)) +
    geom_bar(aes(colour = "black", fill = Gender)) +
    geom_text(stat='count', aes(label=..count..),vjust=2) +
    xlab("Gender") +
    ggtitle(paste(day, "Age Group (<20 Age)", sep = " : "))
}

# SignedIn v/s Not SignedIn
loggedin_vs_not <- function (df, day) {
  ggplot(df, aes(x = Signed_In, y = ..count../1000)) +
    geom_bar(aes(colour = "black", fill = Signed_In)) +
    geom_text(stat='count', aes(label=..count..), vjust=2) +
    ylab("Total (in thousands)") +
    xlab("Day 1: Logged In Stats") +
    ggtitle(paste(day, "Signed In Status", sep = " : "))
}

# Clicked v/s not clicked
clicked_vs_not <- function (df, day) {
  df$Have_Clicked[df$Clicks > 0] <- "Clicked"
  df$Have_Clicked[df$Clicks == 0] <- "Not Clicked"

  ggplot(df, aes(x = Have_Clicked, y = ..count../1000)) +
    geom_bar(aes(colour = "black", fill = Have_Clicked)) +
    geom_text(stat='count', aes(label=..count..),vjust=1.5) +
    ylab("Total (in thousands)") +
    xlab("Clicks") +
    ggtitle(paste(day, "Clicks Status", sep = " : "))
}
```

#Plots

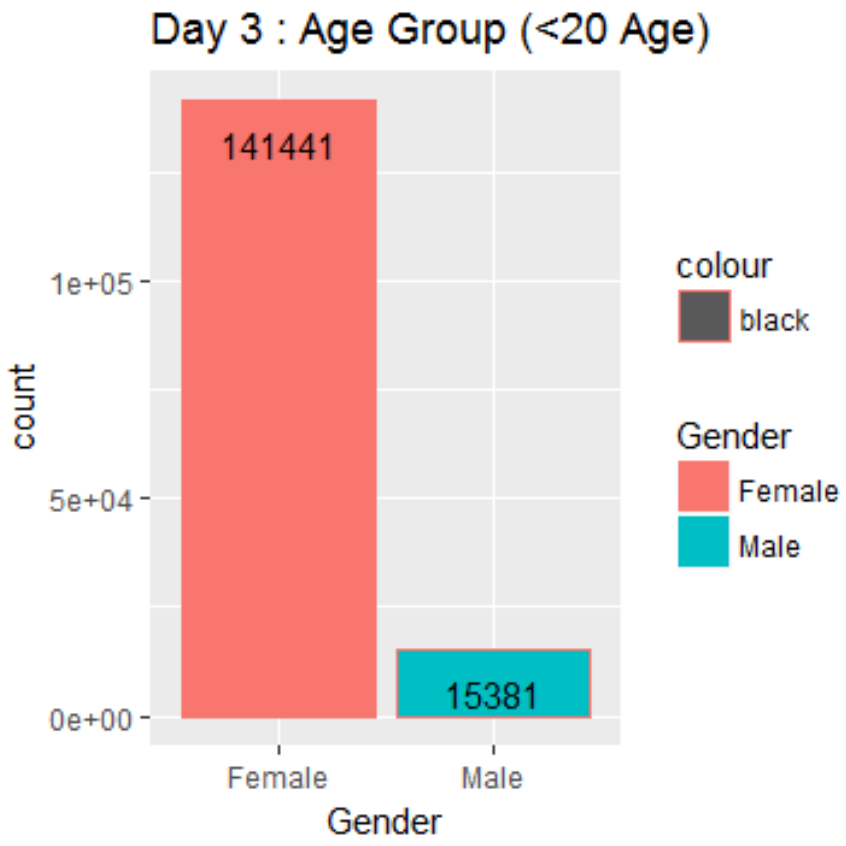
```
male_vs_female(day_1, "<20", "Day 1")
```



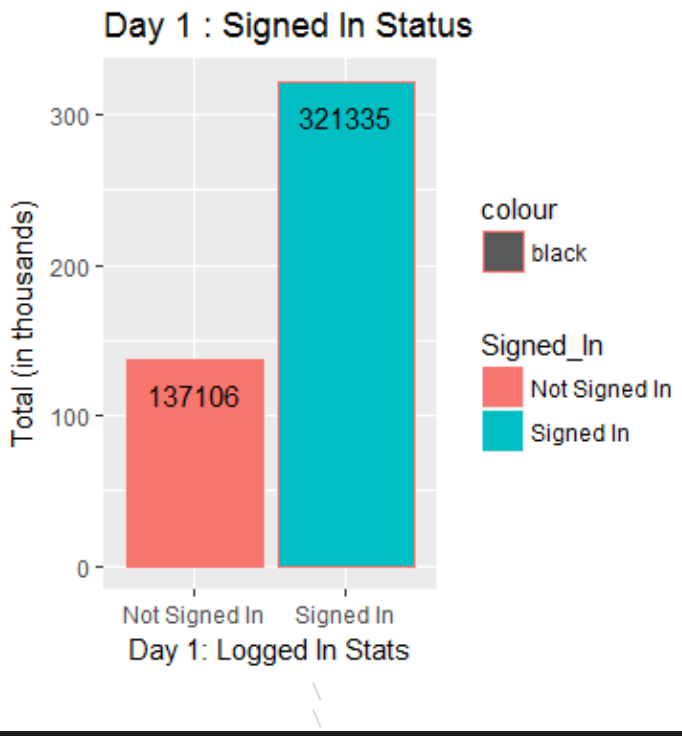

```
male_vs_female(day_2, "<20", "Day 2")
```



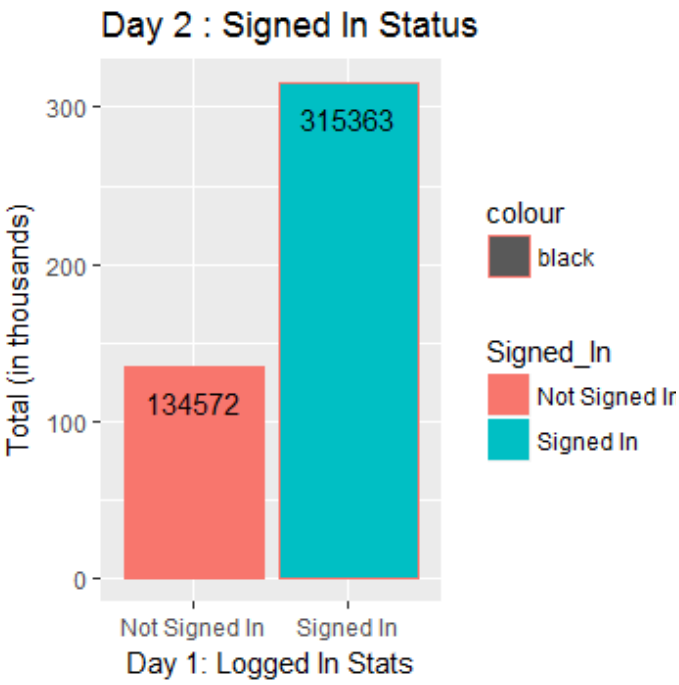

```
male_vs_female(day_3, "<20", "Day3")
```



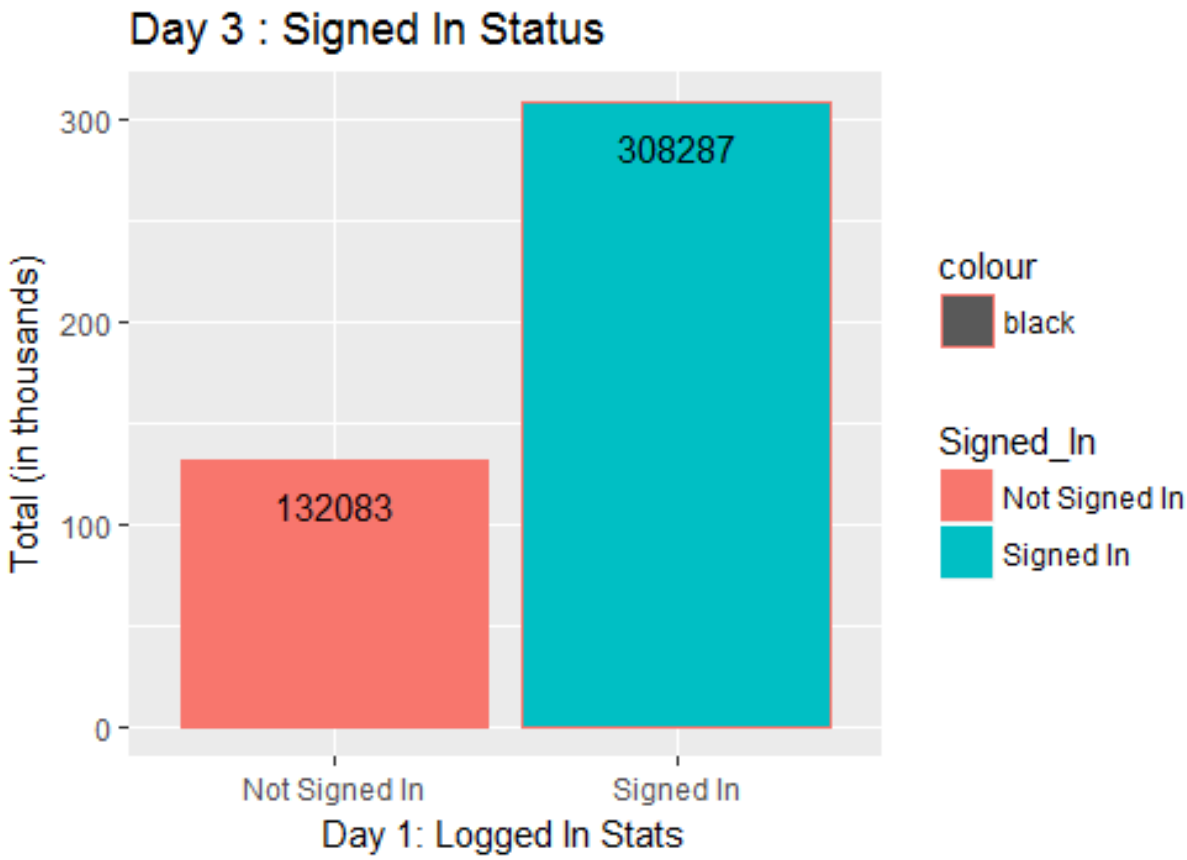
```
loggedin_vs_not(day_1, "Day 1")
```



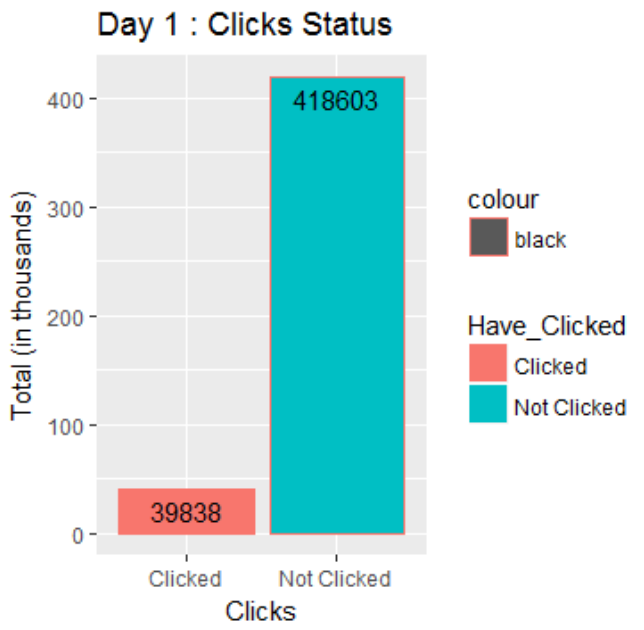

```
loggedin_vs_not(day_2, "Day 2")
```



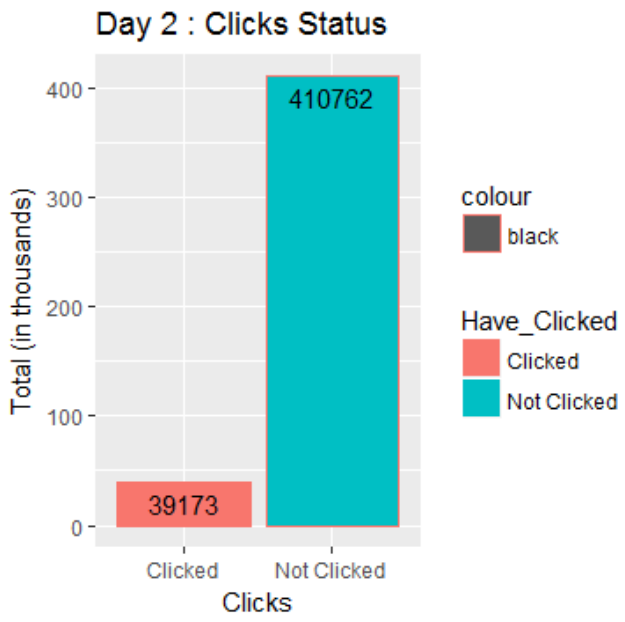

```
loggedin_vs_not(day_3, "Day 3")
```



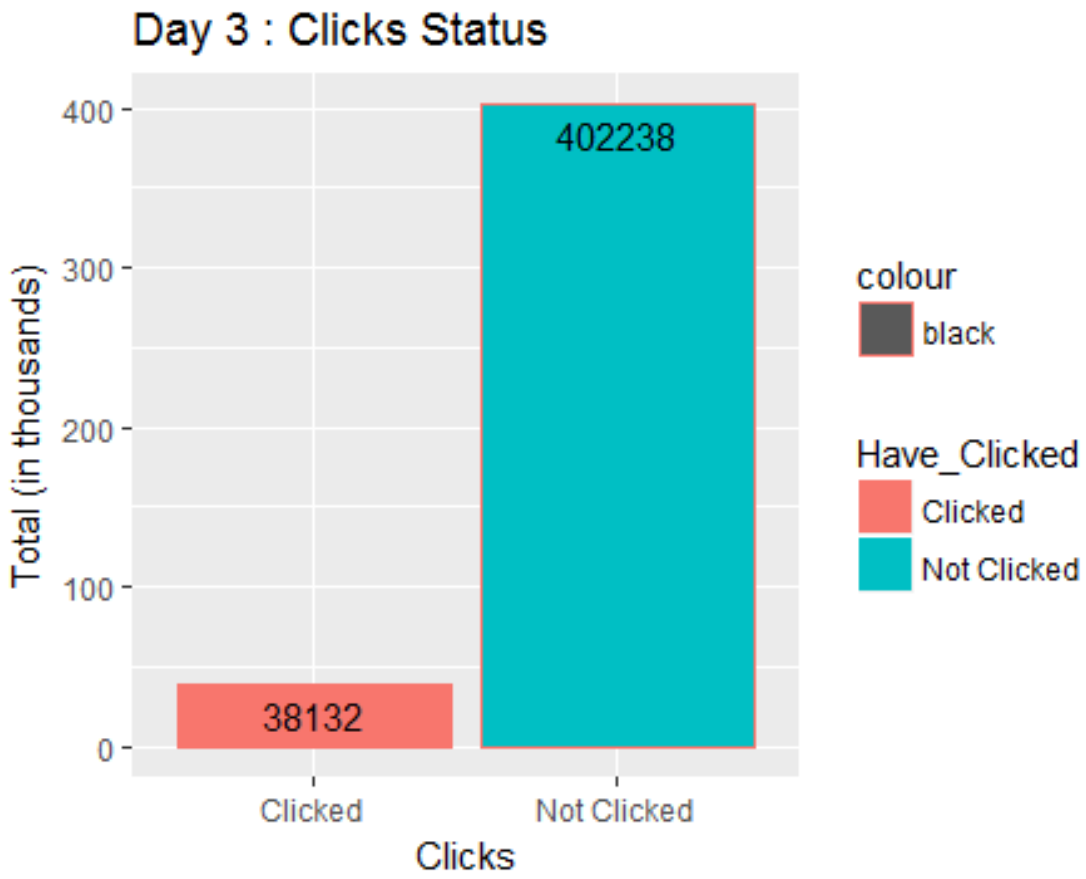
```
clicked_vs_not(day_1, "Day 1")
```




```
clicked_vs_not(day_2, "Day 2")
```



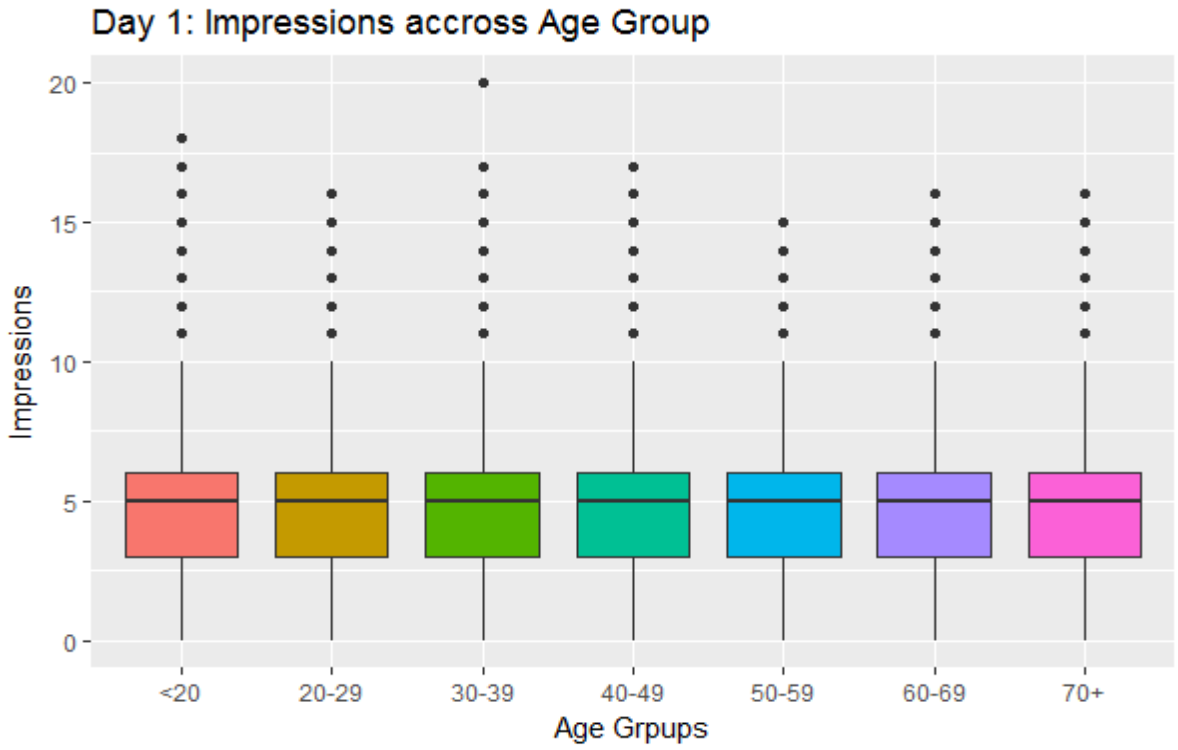

```
clicked_vs_not(day_3, "Day 3")
```



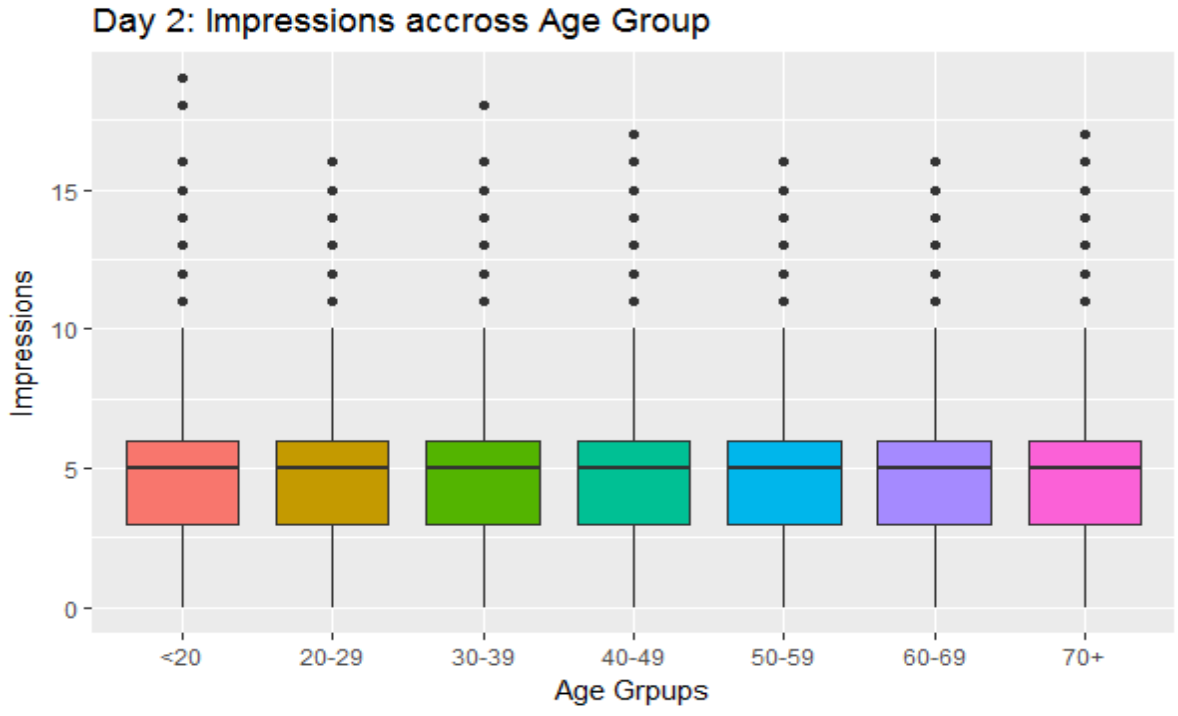
Metrics and Distribution Analysis

#IMPRESSION ACROSS AGE GROUPS
#Plots

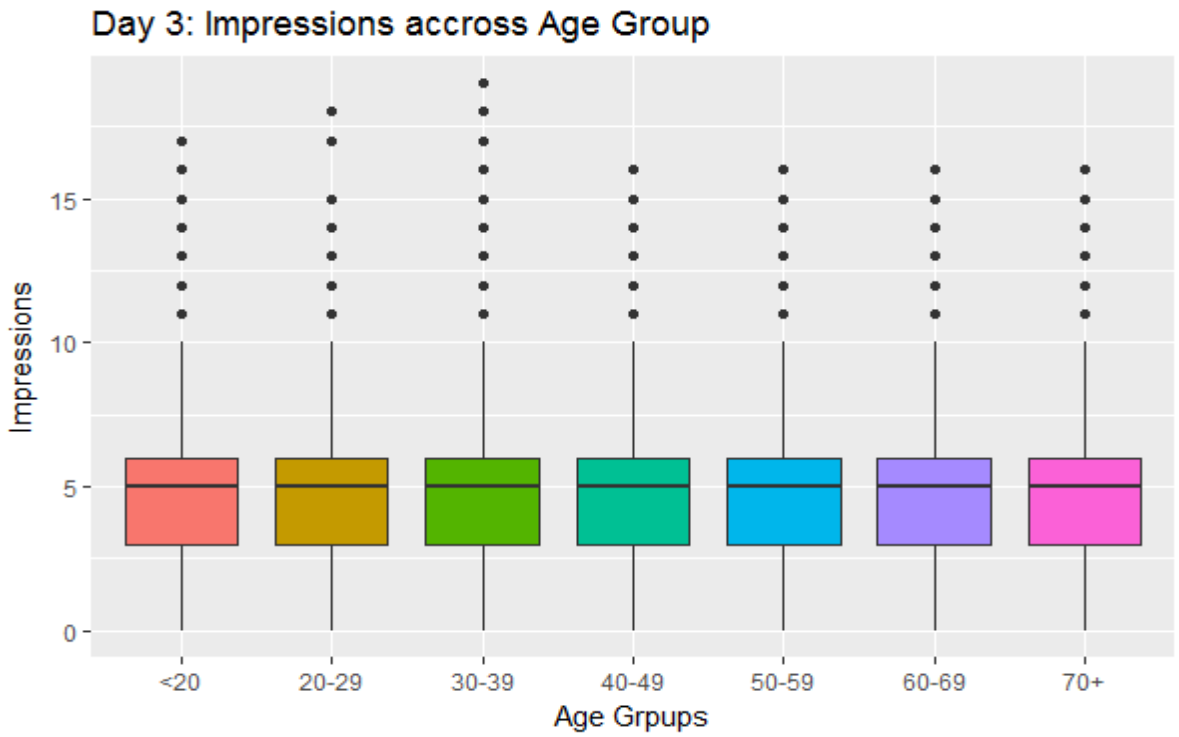
```
ggplot(day_1,aes(x=Age_Group, y=Impressions, fill=Age_Group)) + geom_boxplot() +
  xlab("Age Grpups") + guides(fill=FALSE) + ggtitle("Day 1: Impressions accross Age Group")
```




```
ggplot(day_2,aes(x=Age_Group, y=Impressions, fill=Age_Group)) + geom_boxplot() +
  xlab("Age Grpups") + guides(fill=FALSE) + ggtitle("Day 2: Impressions accross Age Group")
```

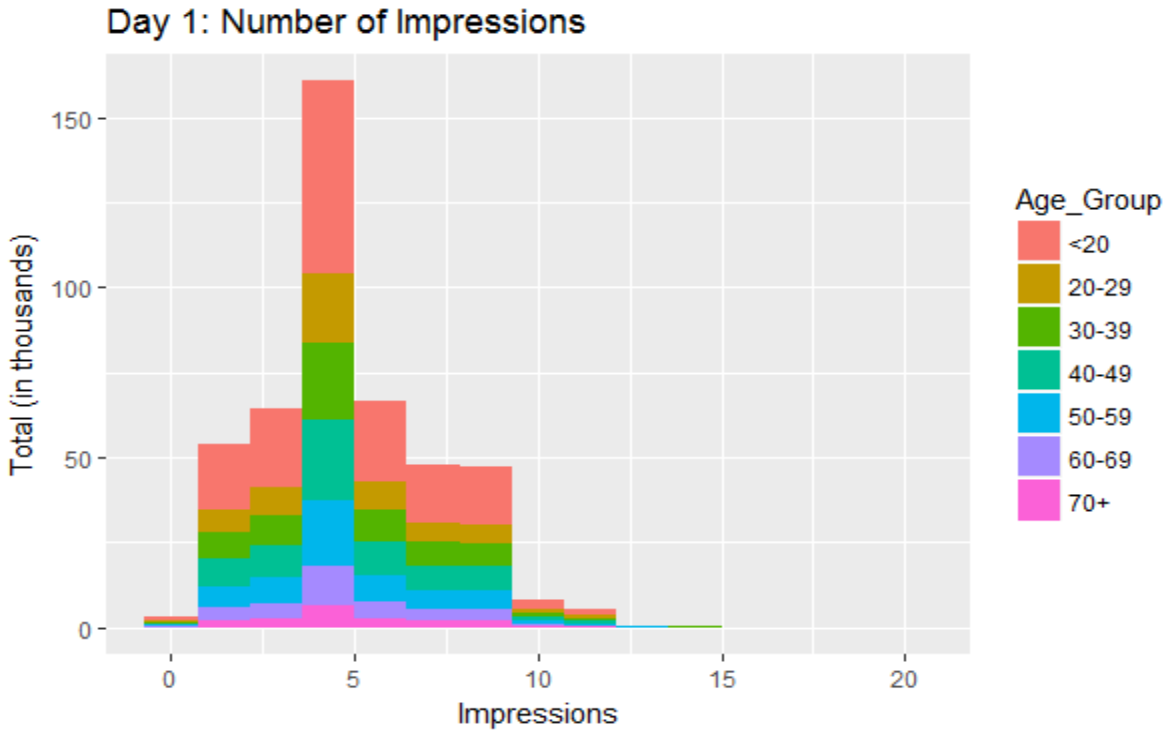


```
ggplot(day_3,aes(x=Age_Group, y=Impressions, fill=Age_Group)) + geom_boxplot() + xlab("Age Grpups") +
  guides(fill=FALSE) + ggtitle("Day 3: Impressions accross Age Group")
```

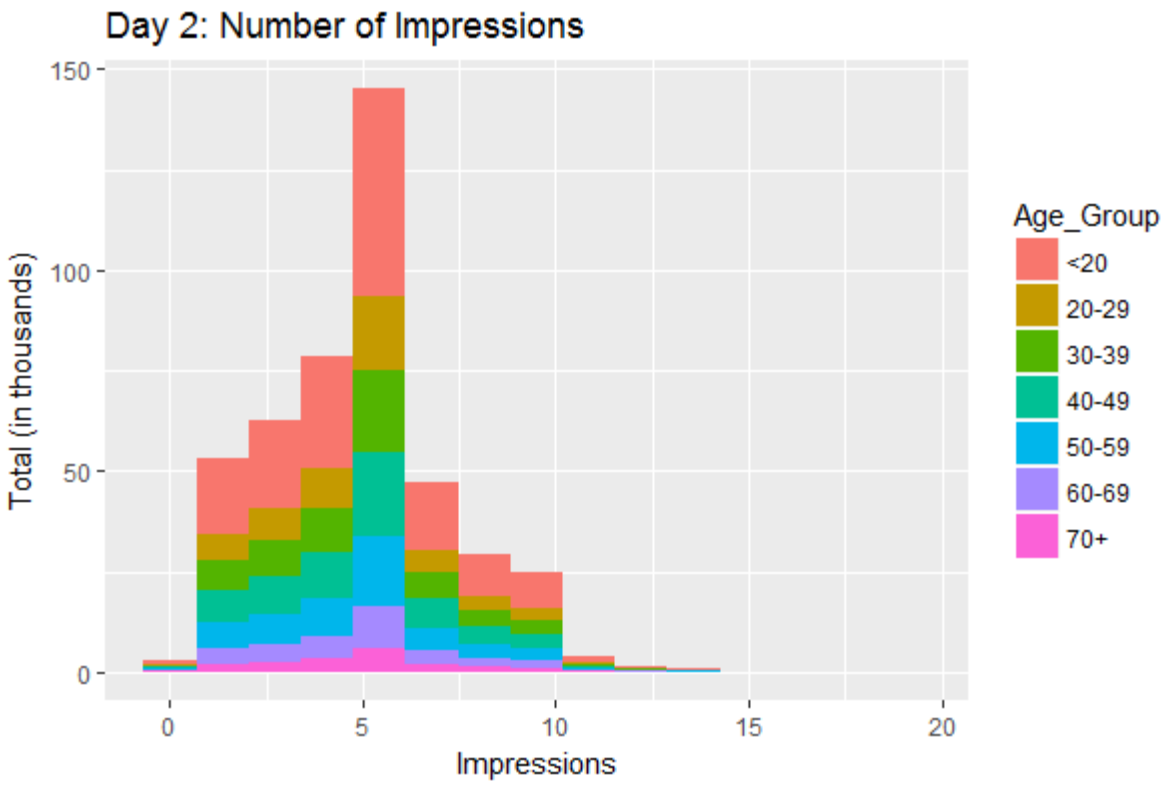


#NUMBER OF IMPRESSIONS ACROSS AGE GROUPS
#Plots

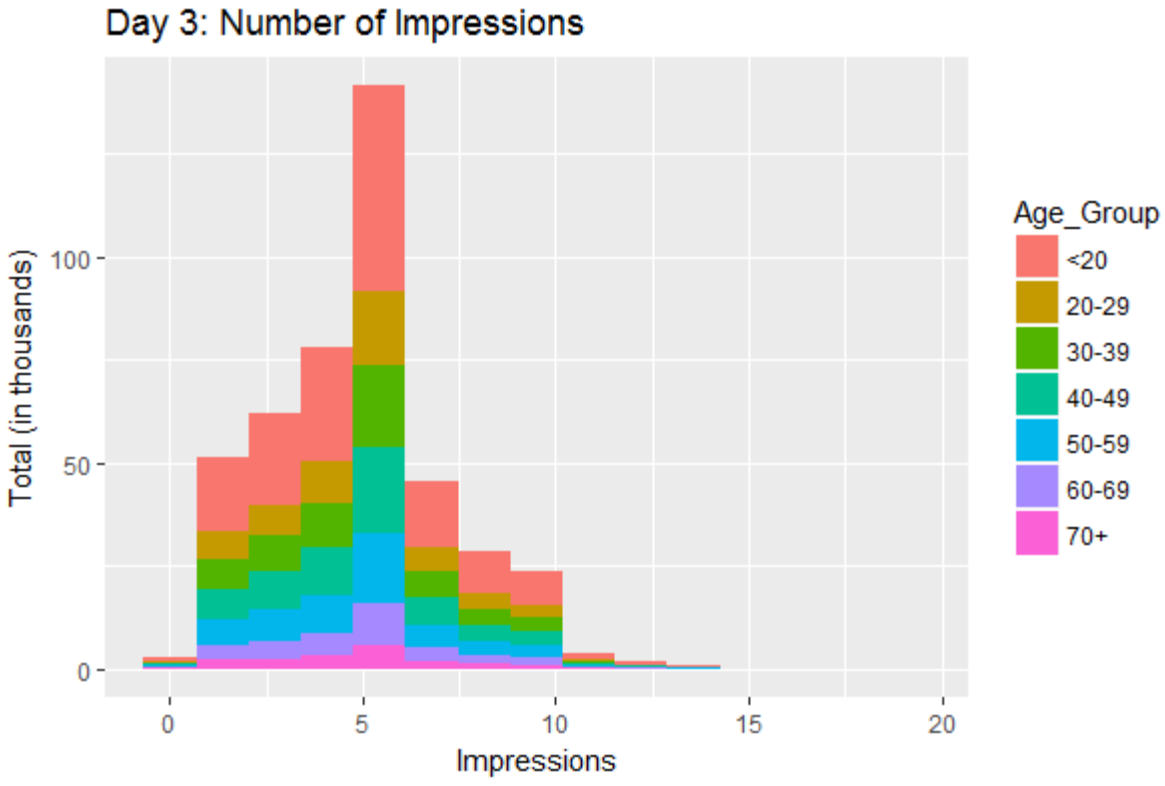
```
ggplot(day_1, aes(x=Impressions, y = ..count../1000, fill=Age_Group)) +
  ylab("Total (in thousands)") + geom_histogram(bins = 15) + ggtitle("Day 1: Number of Impressions")
```



```
ggplot(day_2, aes(x=Impressions, y = ..count../1000, fill=Age_Group)) +
  ylab("Total (in thousands)") + geom_histogram(bins = 15) + ggtitle("Day 2: Number of Impressions")
```

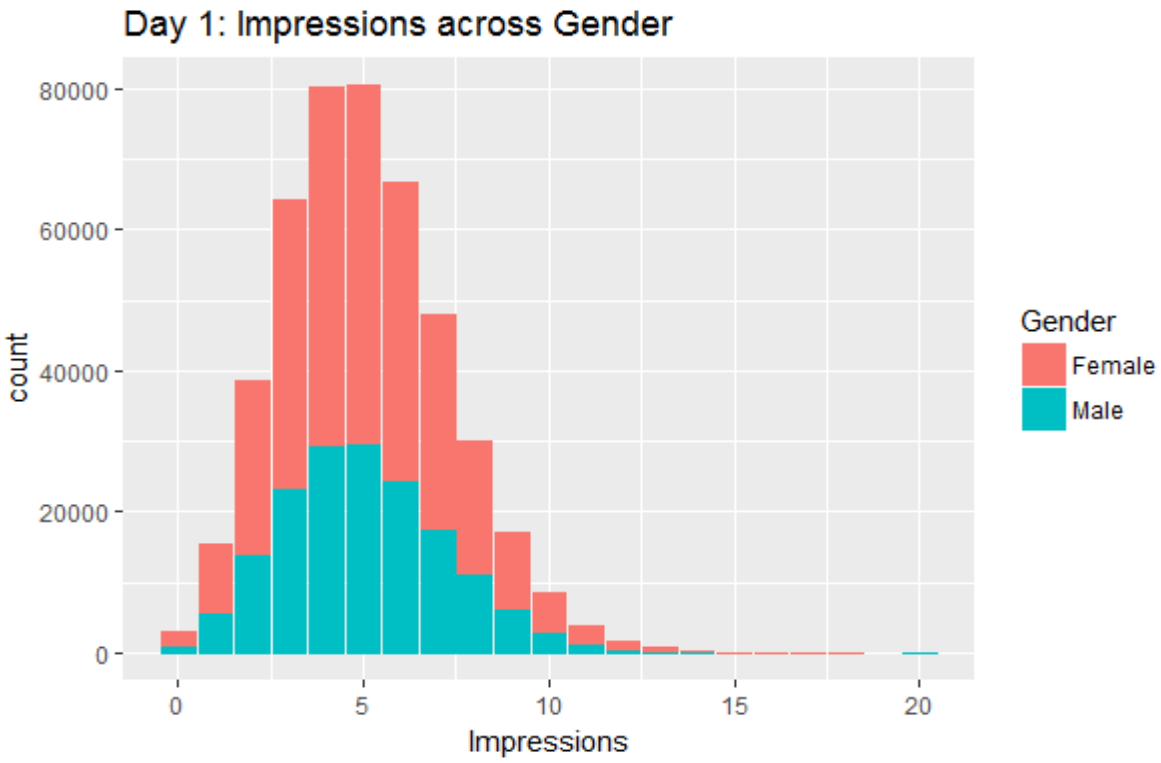



```
ggplot(day_3, aes(x=Impressions, y = ..count../1000, fill=Age_Group)) +
  ylab("Total (in thousands)") + geom_histogram(bins = 15) + ggtitle("Day 3: Number of Impressions")
```

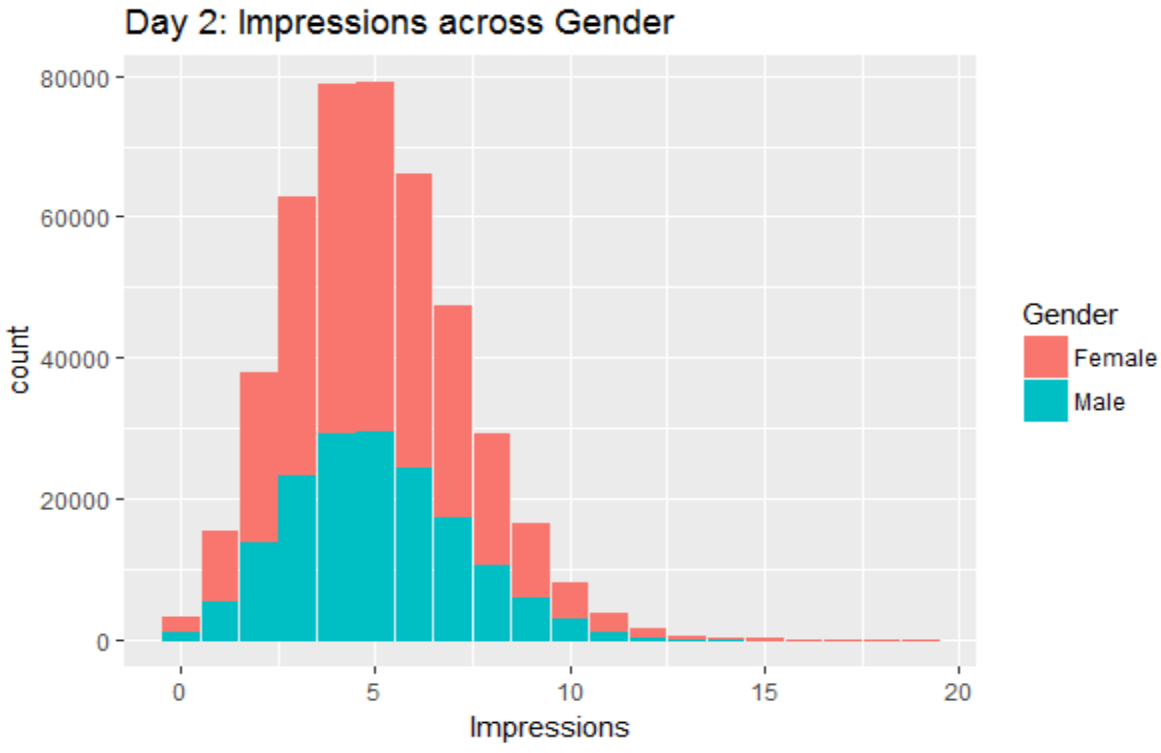


#NUMBER OF IMPRESSIONS ACROSS GENDER
#Plots

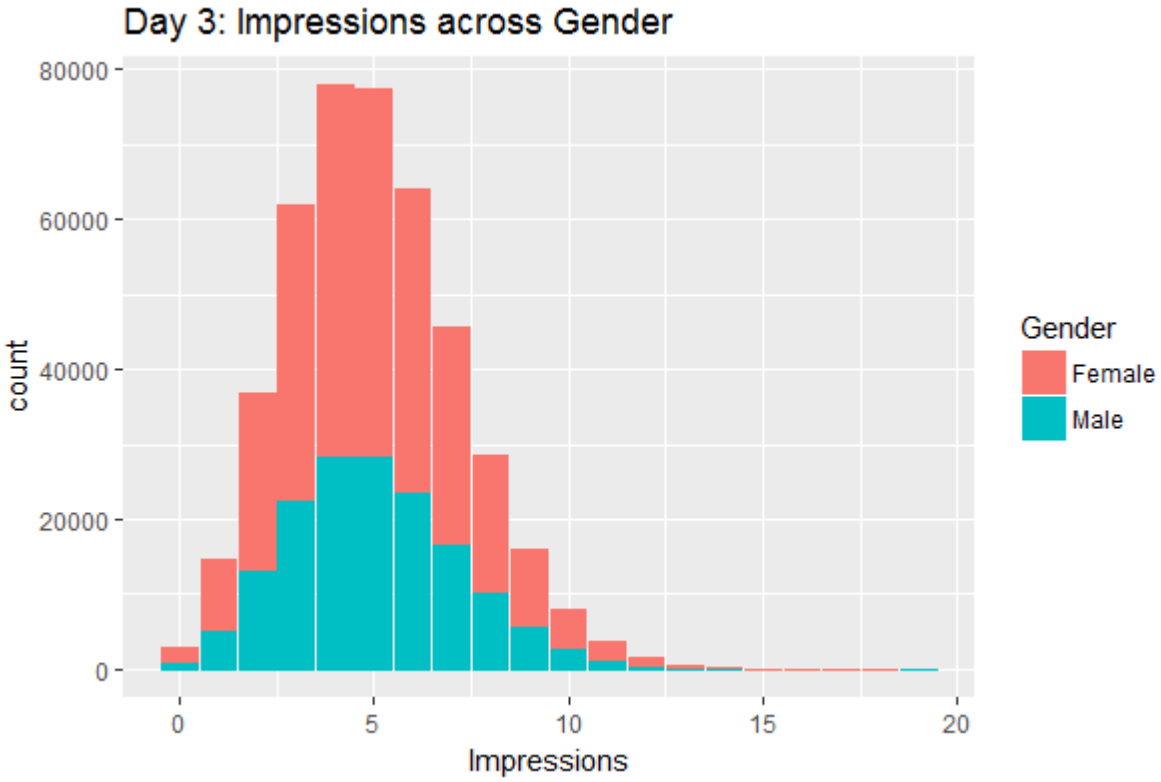
```
ggplot(day_1, aes(x=Impressions,  fill=Gender, colour=Gender)) +
  geom_bar() + xlab("Impressions") + ggtitle("Day 1: Impressions across Gender")
```




```
ggplot(day_2, aes(x=Impressions,  fill=Gender, colour=Gender)) +
  geom_bar() + xlab("Impressions") + ggtitle("Day 2: Impressions across Gender")
```

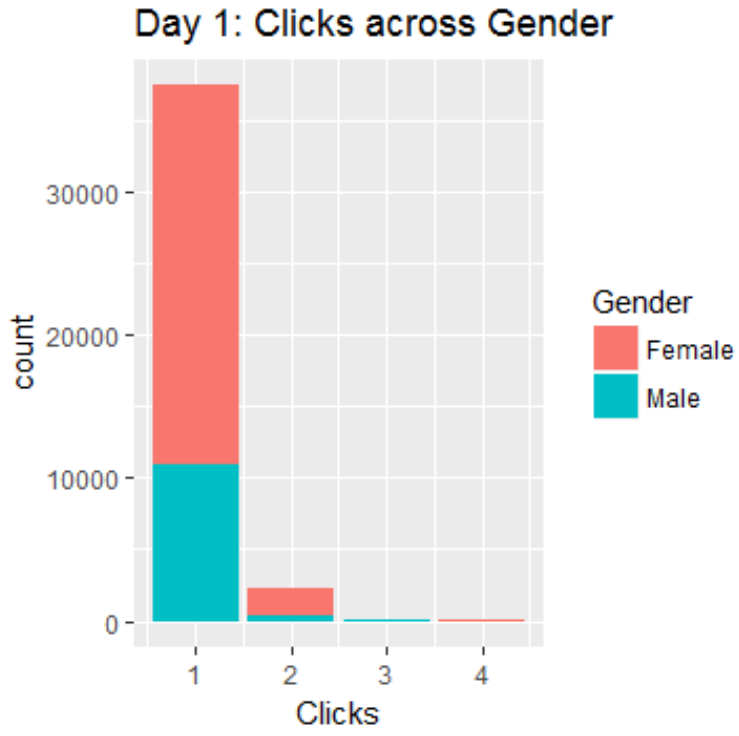


```
ggplot(day_3, aes(x=Impressions,  fill=Gender, colour=Gender)) +
  geom_bar() + xlab("Impressions") + ggtitle("Day 3: Impressions across Gender")
```

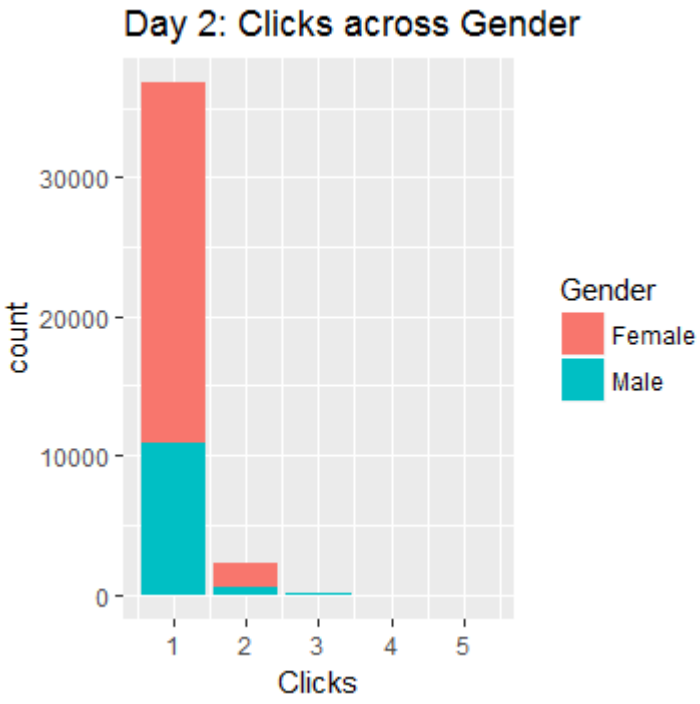


#DISTRIBUTION OF CLICKS ACROSS GENDER
#Plots

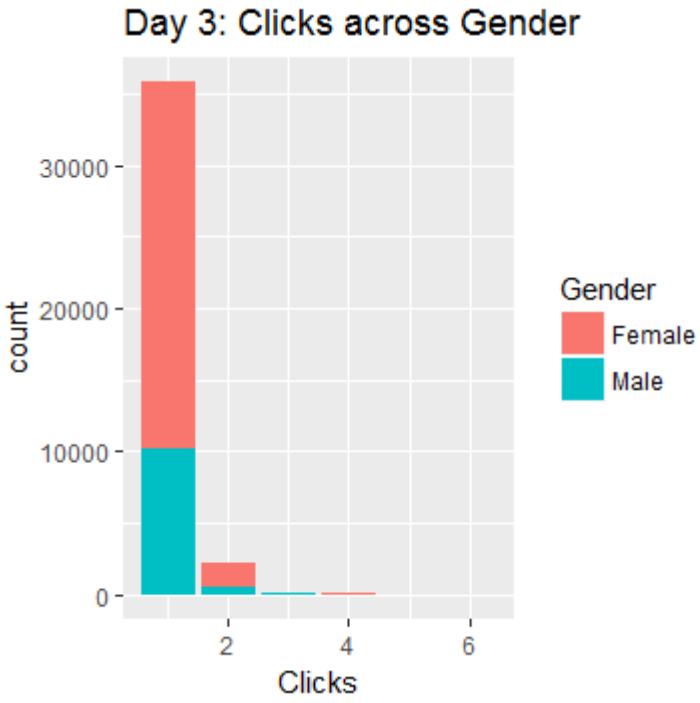
```
ggplot(subset(day_1, Impressions > 0 & Clicks > 0), aes(x=Clicks, fill=Gender)) + geom_bar() +
  ggtitle("Day 1: Clicks across Gender")
```



```
ggplot(subset(day_2, Impressions > 0 & Clicks > 0), aes(x=Clicks, fill=Gender)) + geom_bar() +
  ggtitle("Day 2: Clicks across Gender")
```

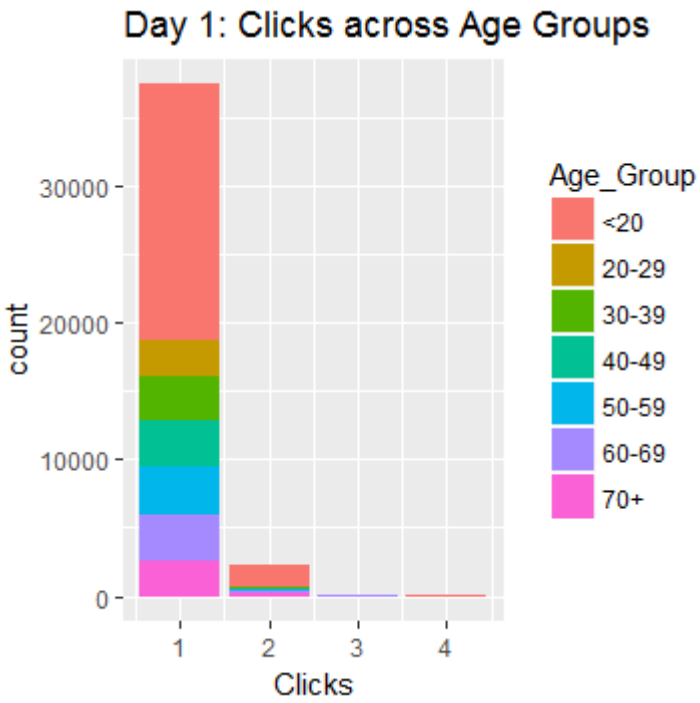



```
ggplot(subset(day_3, Impressions > 0 & Clicks > 0), aes(x=Clicks, fill=Gender)) + geom_bar() +
  ggtitle("Day 3: Clicks across Gender")
```

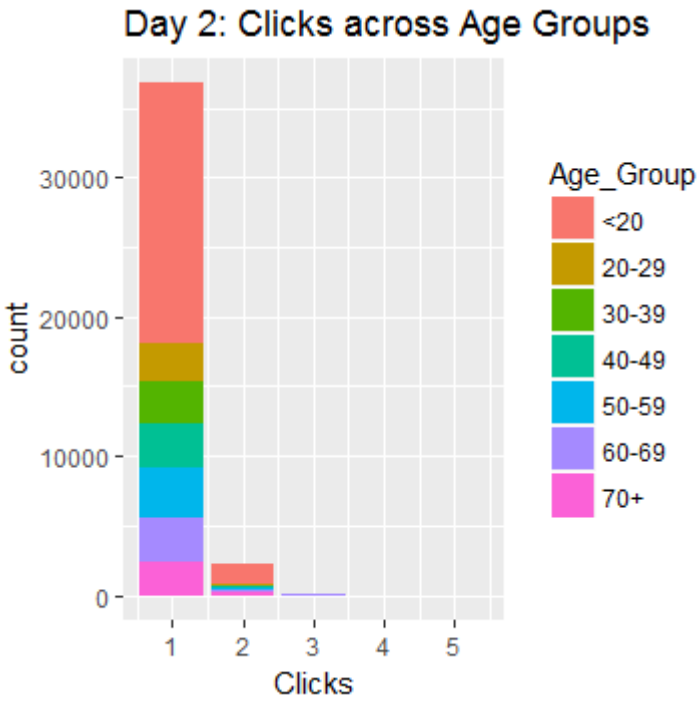


#DISTRIBUTION OF CLICKS ACROSS AGE GROUPS
#Plots

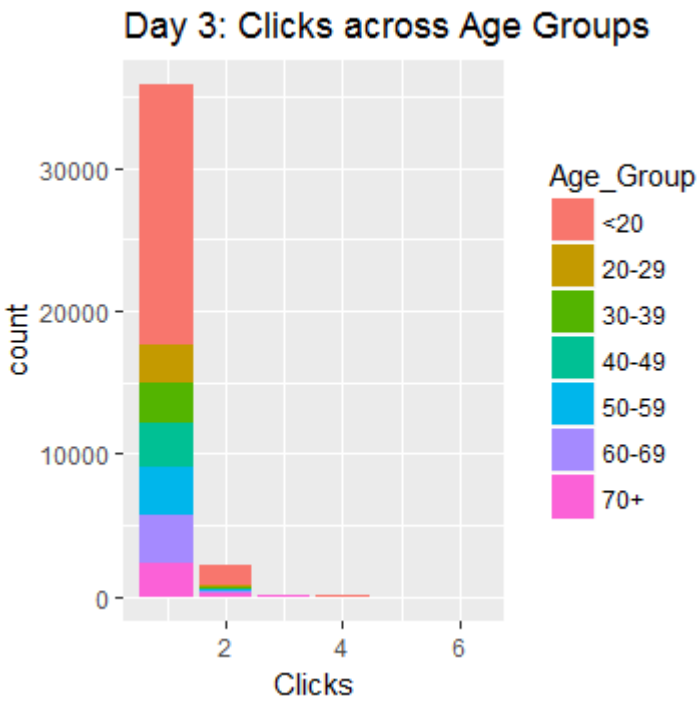
```
ggplot(subset(day_1, Impressions > 0 & Clicks > 0), aes(x=Clicks, fill=Age_Group)) + geom_bar() +
  ggtitle("Day 1: Clicks across Age Groups")
```




```
ggplot(subset(day_2, Impressions > 0 & Clicks > 0), aes(x=Clicks, fill=Age_Group)) + geom_bar() +
  ggtitle("Day 2: Clicks across Age Groups")
```

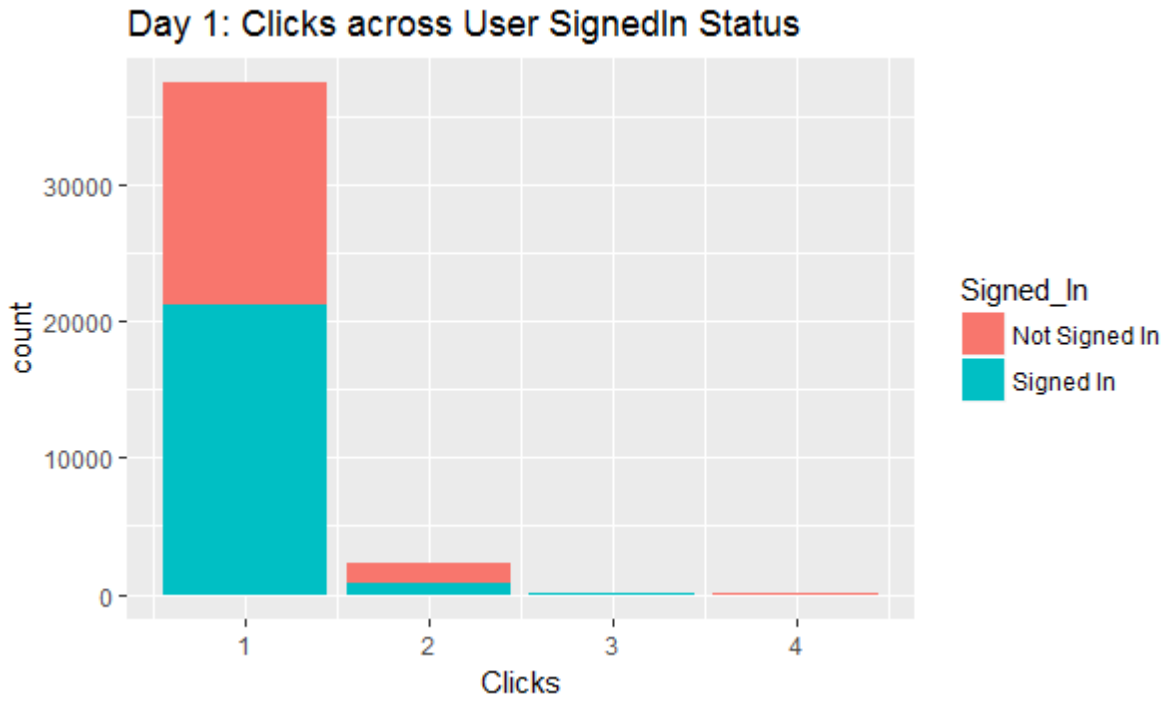


```
ggplot(subset(day_3, Impressions > 0 & Clicks > 0), aes(x=Clicks, fill=Age_Group)) + geom_bar() +
  ggtitle("Day 3: Clicks across Age Groups")
```

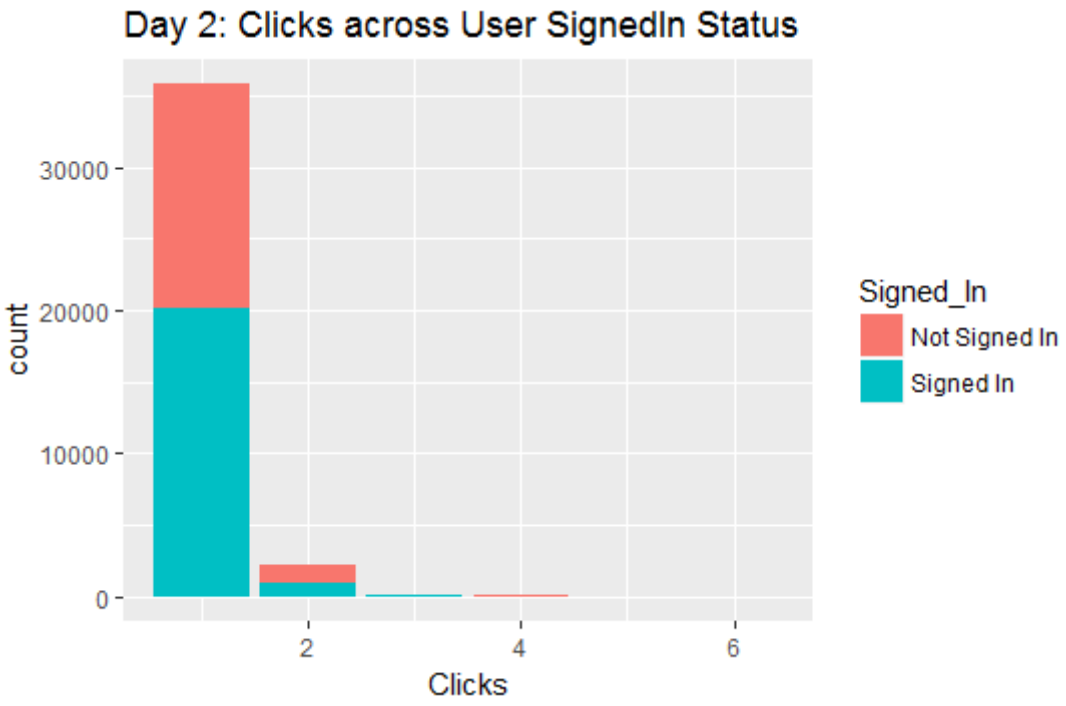


#DISTRIBUTION OF CLICKS ACROSS SIGNED AND UNSIGNED USERS
#Plots

```
ggplot(subset(day_1, Impressions > 0 & Clicks > 0), aes(x=Clicks, fill=Signed_In)) + geom_bar() +
  ggtitle("Day 1: Clicks across User SignedIn Status")
```



```
ggplot(subset(day_3, Impressions > 0 & Clicks > 0), aes(x=Clicks, fill=Signed_In)) + geom_bar() +
  ggtitle("Day 2: Clicks across User SignedIn Status")
```




```
ggplot(subset(day_3, Impressions > 0 & Clicks > 0), aes(x=Clicks, fill=Signed_In)) + geom_bar() +
  ggtitle("Day 3: Clicks across User SignedIn Status")
```