

SemesterArbeit

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Semester Arbeit

Description of dataset

Context

While many public datasets (on Kaggle and the like) provide Apple App Store data, there are not many counterpart datasets available for Google Play Store apps anywhere on the web. It is because, iTunes App Store page deploys a nicely indexed appendix-like structure to allow for simple and easy web scraping. On the other hand, Google Play Store uses sophisticated modern-day techniques (like dynamic page load) using JQuery making scraping more challenging. Content

Each app (row) has values for category, rating, size, and more.

Acknowledgements

All information is scraped from the Google Play Store. I downloaded this dataset from Kaggle.com

Inspiration

The Play Store apps data has enormous potential to drive app-making businesses to success. Actionable insights can be drawn for developers to work on and capture the Android market!

loading the dataset

Summary of infos for each column:

- Are they numeric?
- If numeric, what is min, max, and quartiles
- Are there any NAs?

```
googleplaystore <- read.csv("~/Studium/Data-Science-General/HS2021/EXPD/Projects/googleplaystore/googleplaystore")
playstore<-as.data.frame(googleplaystore)
summary(playstore)
```

```
##           App           Category           Rating           Reviews
## Length:10841 Length:10841 Min. : 1.000 Length:10841
## Class :character Class :character 1st Qu.: 4.000 Class :character
## Mode :character Mode :character Median : 4.300 Mode :character
## Mean : 4.193
## 3rd Qu.: 4.500
## Max. :19.000
## NA's :1474
##           Size           Installs           Type           Price
## Length:10841 Length:10841 Length:10841 Length:10841
## Class :character Class :character Class :character Class :character
## Mode :character Mode :character Mode :character Mode :character
##
##
##
## Content.Rating Genres Last.Updated Current.Ver
## Length:10841 Length:10841 Length:10841 Length:10841
## Class :character Class :character Class :character Class :character
## Mode :character Mode :character Mode :character Mode :character
##
##
##
## Android.Ver
## Length:10841
## Class :character
## Mode :character
##
##
##
##
```

The first 10 lines of dataset:

```
head(playstore,10)
```

```
##           App           Category Rating
## 1 Photo Editor & Candy Camera & Grid & ScrapBook ART_AND_DESIGN 4.1
## 2 Coloring book moana ART_AND_DESIGN 3.9
## 3 U Launcher Lite â\200" FREE Live Cool Themes, Hide Apps ART_AND_DESIGN 4.7
## 4 Sketch - Draw & Paint ART_AND_DESIGN 4.5
## 5 Pixel Draw - Number Art Coloring Book ART_AND_DESIGN 4.3
## 6 Paper flowers instructions ART_AND_DESIGN 4.4
## 7 Smoke Effect Photo Maker - Smoke Editor ART_AND_DESIGN 3.8
## 8 Infinite Painter ART_AND_DESIGN 4.1
## 9 Garden Coloring Book ART_AND_DESIGN 4.4
## 10 Kids Paint Free - Drawing Fun ART_AND_DESIGN 4.7
## Reviews Size Installs Type Price Content.Rating Genres
## 1 159 19M 10,000+ Free 0 Everyone Art & Design
## 2 967 14M 500,000+ Free 0 Everyone Art & Design;Pretend Play
## 3 87510 8.7M 5,000,000+ Free 0 Everyone Art & Design
```

## 4	215644	25M	50,000,000+	Free	0	Teen	Art & Design
## 5	967	2.8M	100,000+	Free	0	Everyone	Art & Design;Creativity
## 6	167	5.6M	50,000+	Free	0	Everyone	Art & Design
## 7	178	19M	50,000+	Free	0	Everyone	Art & Design
## 8	36815	29M	1,000,000+	Free	0	Everyone	Art & Design
## 9	13791	33M	1,000,000+	Free	0	Everyone	Art & Design
## 10	121	3.1M	10,000+	Free	0	Everyone	Art & Design;Creativity
##	Last.Updated		Current.Ver		Android.Ver		
## 1	January 7, 2018		1.0.0		4.0.3 and up		
## 2	January 15, 2018		2.0.0		4.0.3 and up		
## 3	August 1, 2018		1.2.4		4.0.3 and up		
## 4	June 8, 2018		Varies with device		4.2 and up		
## 5	June 20, 2018		1.1		4.4 and up		
## 6	March 26, 2017		1.0		2.3 and up		
## 7	April 26, 2018		1.1		4.0.3 and up		
## 8	June 14, 2018		6.1.61.1		4.2 and up		
## 9	September 20, 2017		2.9.2		3.0 and up		
## 10	July 3, 2018		2.8		4.0.3 and up		

clean up of the dataset:

Cleanups:

- removing line 10473 because it has a rating of 19. And it is most likely false data
- making reviews, size and price vectors to numeric vectors

```
#to find the NAs after getting:"NAs introduced by coercion"
#I used: which(is.na(x)) after applyin function to find
#what are the abnormalities
#the first abnormality is line 10473. I think it is faulty:
#rating is by 19 and price is everyone and size is 1,000+
#so we delete this line:
playstore[10473,]
playstore<- playstore[-10473,]

#cleaning the Reviews:
str(playstore$Reviews)
playstore$Reviews<-as.numeric(playstore$Reviews)

#cleaning the size:
str(playstore$Size)

#To replace 1.5k with 1500
#we need the following library to do all these:
#install.packages("stringr")
library("stringr")
nonDecimalVec<-stringr::str_extract(string = playstore$Size,pattern ="\\\\.([0-9]*)")
#replace NAs with empty string , so we have an easier
#job, when we use paste function later
nonDecimalVec[is.na(nonDecimalVec)]<-""
```

```

playstore$Size<-sub(pattern = "\\.[0-9]*k", "000", playstore$Size)
playstore$Size<-sub(pattern = "\\.[0-9]*M", "000000", playstore$Size)
#finally: adding the nachkommastellen back to the number
#if they had any example: 1.05k= (1+0.05)*1000:
vsel<-nonDecimalVec!=""
temp_size<-(as.numeric(playstore$Size[vsel]))
temp_size2<-rep(1, sum(vsel))
temp_size3<-as.numeric(nonDecimalVec[vsel])
temp_size4<-temp_size2+temp_size3
temp_size<-temp_size4*temp_size
playstore$Size[vsel]<-temp_size

#to replace the likes of 10k with 1000 or m with 1000000:
playstore$Size<-sub(pattern = "k", "000", playstore$Size)
playstore$Size<-sub(pattern = "M", "000000", playstore$Size)
#size also contains the string: "Varies with device"
#So we should be careful about that!
playstore$Size[playstore$Size=="Varies with device"]<-NA
playstore$Size<-as.numeric(playstore$Size)
# coercedNas<-which(is.na(playstore$Size))
# coercedNas

#cleaning the price:
str(playstore$Price)
playstore$Price<-sub(pattern = "\\$", replacement = "", x = playstore$Price)
playstore$Price<-as.numeric(playstore$Price)
#The following two lines helped me with debugging and cleaning
# coercedNas<-which(is.na(playstore$Price))
# coercedNas

#I removed the following line because we don't want to
#lose information, for example if an app doesn't have
#ratings, it could be that it was downloaded very
#little.
#playstore<-playstore[complete.cases(playstore), ]

```

Introduction of new Kategoria Variables

```
reviewCut<-cut(playstore$Reviews,breaks = c(0,1000,10000,100000,8000000))
playstore$reviewCut<-factor(reviewCut,levels = levels(reviewCut),labels = c("0+","1000+","10k+","100k+",
playstore$reviewSuperCut<-factor(reviewCut,levels = levels(reviewCut),labels = c("0+","1000+","10k+","1

priceCut<-cut(playstore$Price,breaks = c(0,10,30,500))
playstore$priceCut<-factor(priceCut,levels = levels(priceCut),labels = c("0+","10+","30+"),ordered = T)
playstore$priceSuperCut<-factor(priceCut,levels = levels(priceCut),labels = c("0+","10+","100+"),ordered

installsfac<-factor(playstore$Installs, labels =c("0","0+","1+","5+","10+","50+","100+","500+","1,000+"
```

```

playstore$InstallsFac<-installsfac

#labels(installsfac)<-c("0","0+","1+","5+","10+","50+","100+","500+","1,000+","5,000+","10,000+","50,000+")
#playstore$InstallsFac<-installsfac

#It might make sense to convert this to numeric
#to calculate the mean. This is an ordinal category
#but the difference from category to other is not the
#same between each pair of categories:
#(500-100)=400 but (100-50)=50
Installs_<-sub(pattern = "\\+", replacement = "", x = playstore$Installs)
Installs_<-gsub(pattern = ",", replacement = "", x = Installs_)
Installs_<-as.numeric(Installs_)

installsCut<-cut(Installs_,breaks = c(0,1e+5,5e+05,1e+06,5e+06,1e+07,5e+07,1e+08,5e+08,1e+09))
insatllsCut<-factor(installsCut,levels = levels(installsCut),ordered = T)
levels(installsCut)<-c("0+","100k+","500k+","1m+","5m+","10m+","50m+","100m+","500m+")
playstore$installsCut<-insatllsCut
installsCut<-cut(Installs_,breaks = c(0,1e+2,1e+03,1e+04,1e+05,5e+05,5e+07,1e+08,5e+08,1e+09))
playstore$installsSuperCut<-factor(installsCut,levels = levels(installsCut),ordered = T)
levels(installsCut)<-c("0+","100+","1k+","10k+","50k+","50k+","50k+","50k+","50k+")

ratingCut<-cut(playstore$Rating,breaks = c(0.99,1.99,2.99,3.49,3.99,5))
ratingCut<-factor(ratingCut,levels = levels(ratingCut),labels = c("1+","2+","3+","3.5+","4+"),ordered = T)
playstore$ratingCut<-ratingCut

playstore$categoryCut<-factor(playstore$Category,levels =names(sort(table(playstore$Category),decreasing=T)),ordered = T)
tmp<-head(levels(playstore$categoryCut),10)
playstore$categoryCut<-factor(playstore$categoryCut,labels = append(tmp,rep("other",23),after = length(tmp)))

```

Interesting numbers and tables:

Means of 4 Variables:

```

mean_table<-colMeans(playstore[,c("Rating","Reviews","Size","Price")],na.rm = T)
mean_table

```

```

##          Rating      Reviews      Size      Price
## 4.191757e+00 4.441529e+05 2.211759e+07 1.027368e+00

```

mean of Installs variable:

```

Installs_mean<-mean(Installs_)
Installs_mean

```

```
## [1] 15464339
```

Quantiles of Reviews

```
quantile(playstore$Reviews,probs = c(0.25,0.5,0.75),type=7)
```

```
##      25%      50%      75%  
##      38.0  2094.0 54775.5
```

Applications with most number of Reviews:

```
##                                     App  Reviews  
## 4296                                Facebook 78143257  
## 9377                                WhatsApp Messenger 69116101  
## 5611                                Instagram 66560497  
## 6394    Messenger â\200" Text and Video Chat for Free 56644091  
## 2461                                Clash of Clans 44889695  
## 2471    Clean Master- Space Cleaner & Antivirus 42916526  
## 8387                                Subway Surfers 27721993  
## 9582                                YouTube 25639427  
## 7993 Security Master - Antivirus, VPN, AppLock, Booster 24900999  
## 2463                                Clash Royale 23132575  
## 2050                                Candy Crush Saga 22427591  
## 9009    UC Browser - Fast Download Private & Secure 17713565  
## 8183                                Snapchat 17011253  
## 67      360 Security - Free Antivirus, Booster, Cleaner 16771865  
## 6711                                My Talking Tom 14889643  
## 112      8 Ball Pool 14198028  
## 3528    DU Battery Saver - Battery Charger & Battery Life 13479633  
## 1136                                BBM - Free Calls & Messages 12843148  
## 1971 Cache Cleaner-DU Speed Booster (booster & cleaner) 12759739  
## 8979                                Twitter 11664259
```

Application with most number of installs:

```
grouped_by_app<-aggregate(Installs_~App,data=playstore,FUN = mean)  
vsel<-head(order(grouped_by_app$Installs_,decreasing=T),20)  
grouped_by_app[vsel,]
```

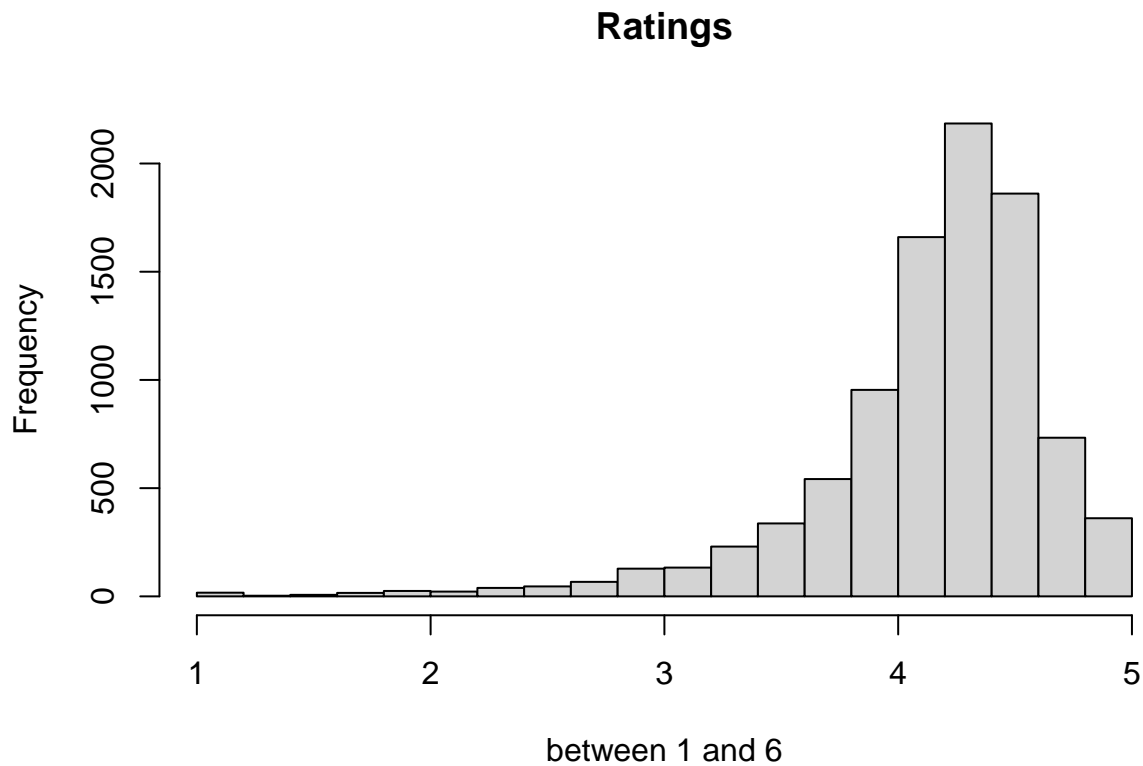
```
##                                     App  Installs_  
## 4296                                Facebook 1e+09  
## 5025                                Gmail 1e+09  
## 5083                                Google 1e+09  
## 5093    Google Chrome: Fast & Secure 1e+09  
## 5096                                Google Drive 1e+09  
## 5105                                Google News 1e+09  
## 5109                                Google Photos 1e+09  
## 5110                                Google Play Books 1e+09  
## 5111                                Google Play Games 1e+09  
## 5112    Google Play Movies & TV 1e+09  
## 5116                                Google Street View 1e+09
```

## 5120	Google+	1e+09
## 5238	Hangouts	1e+09
## 5611	Instagram	1e+09
## 6264	Maps - Navigate & Explore	1e+09
## 6394	Messenger â\200" Text and Video Chat for Free	1e+09
## 8130	Skype - free IM & video calls	1e+09
## 8387	Subway Surfers	1e+09
## 9377	WhatsApp Messenger	1e+09
## 9582	YouTube	1e+09

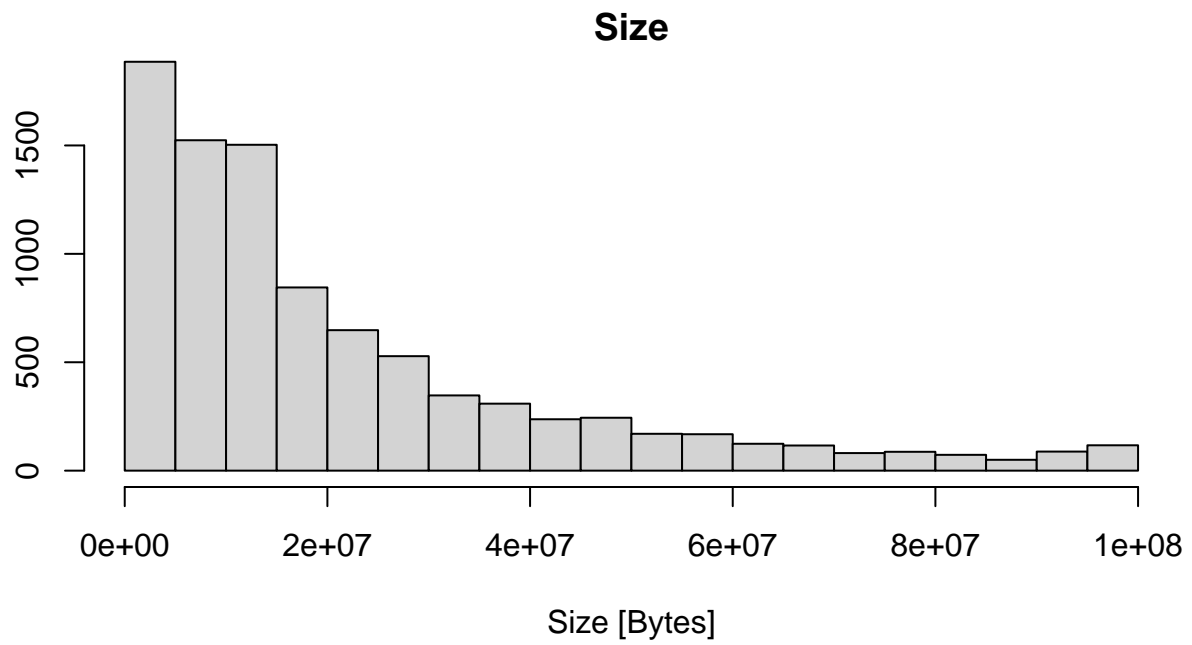
some univariate plots

Below we see a Distribution of Ratings and Size.

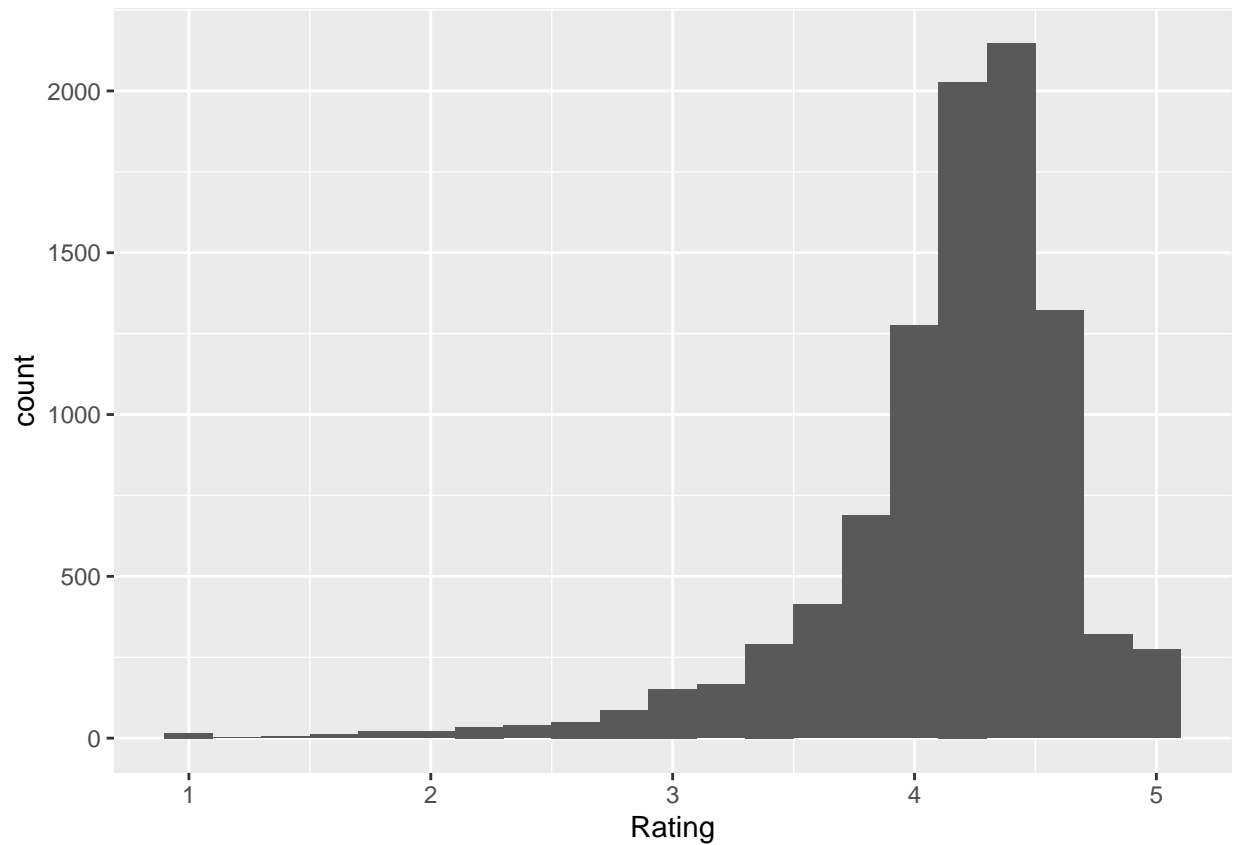
```
library(ggplot2)
hist((playstore$Rating[!is.na(playstore$Rating)]),xlab = "between 1 and 6",main = "Ratings",breaks = 20)
```



```
par(mar=c(10,3,1,1))
hist(playstore$Size,xlab = "Size [Bytes]",main = "Size")
```



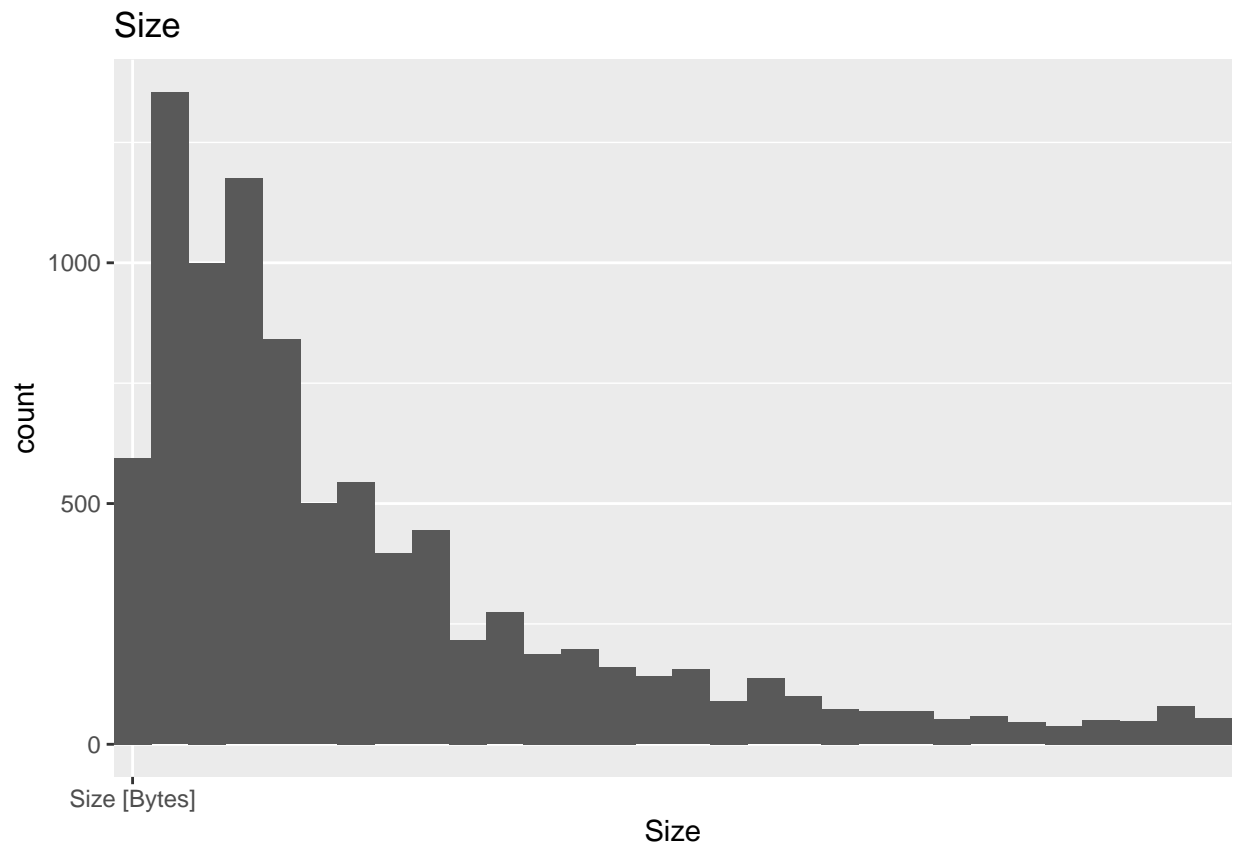
```
#ggplot versions;  
ggplot(data = playstore[!is.na(playstore$Rating),], aes(x=Rating)) +  
  geom_histogram(binwidth = 0.2)
```

```
ggplot(data = playstore,aes(x=Size))+geom_histogram()+ggtitle("Size")+xlim("Size [Bytes]")
```

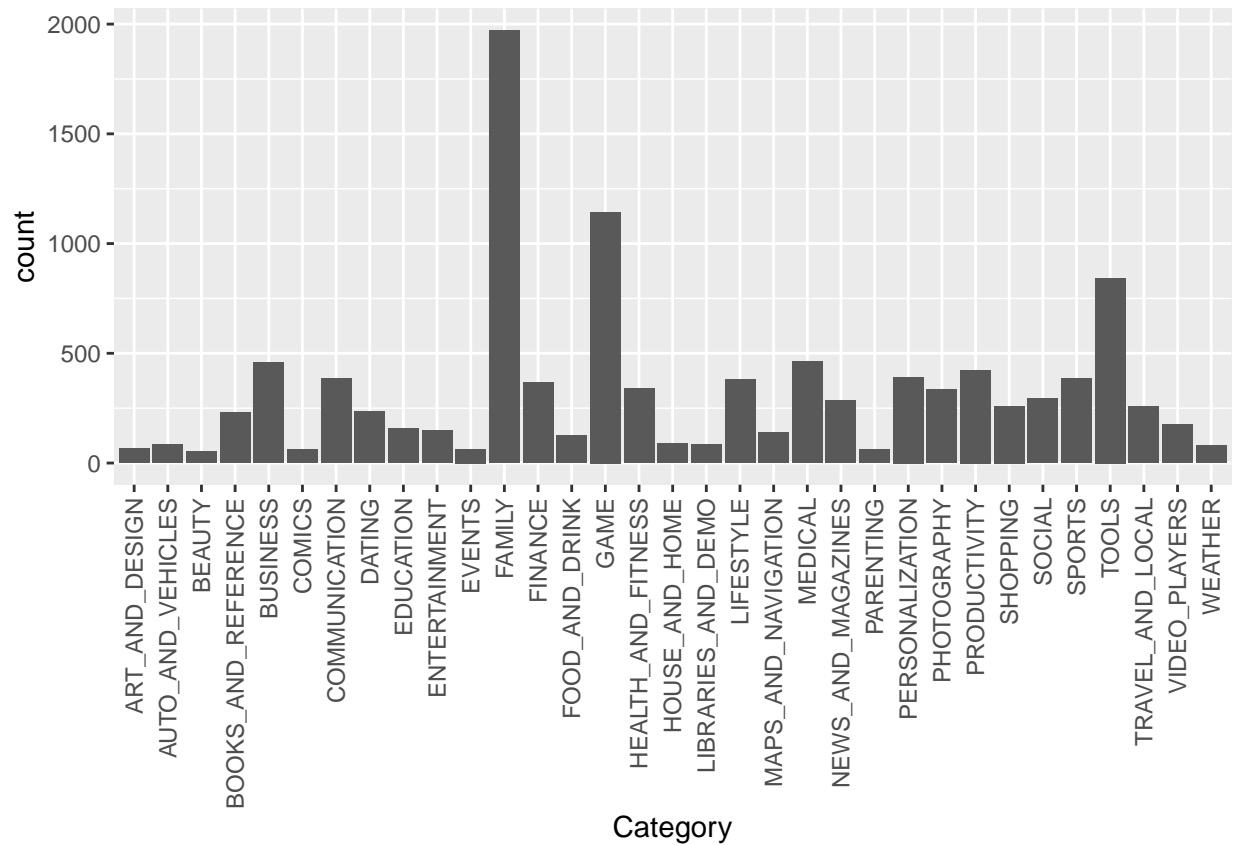
```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```

```
## Warning: Removed 1695 rows containing non-finite values (stat_bin).
```



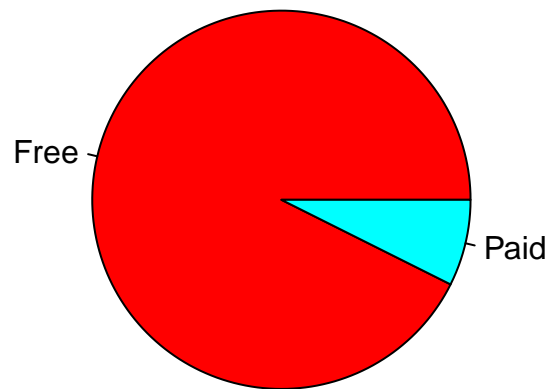
The most frequent category:

```
ggplot(playstore,aes(x=Category))+geom_bar()+  
  guides(x = guide_axis(angle = 90))
```



Are paid apps or free apps the most frequent?

```
pie(table(playstore$Type),col = rainbow(2))
```



some bivariate plots

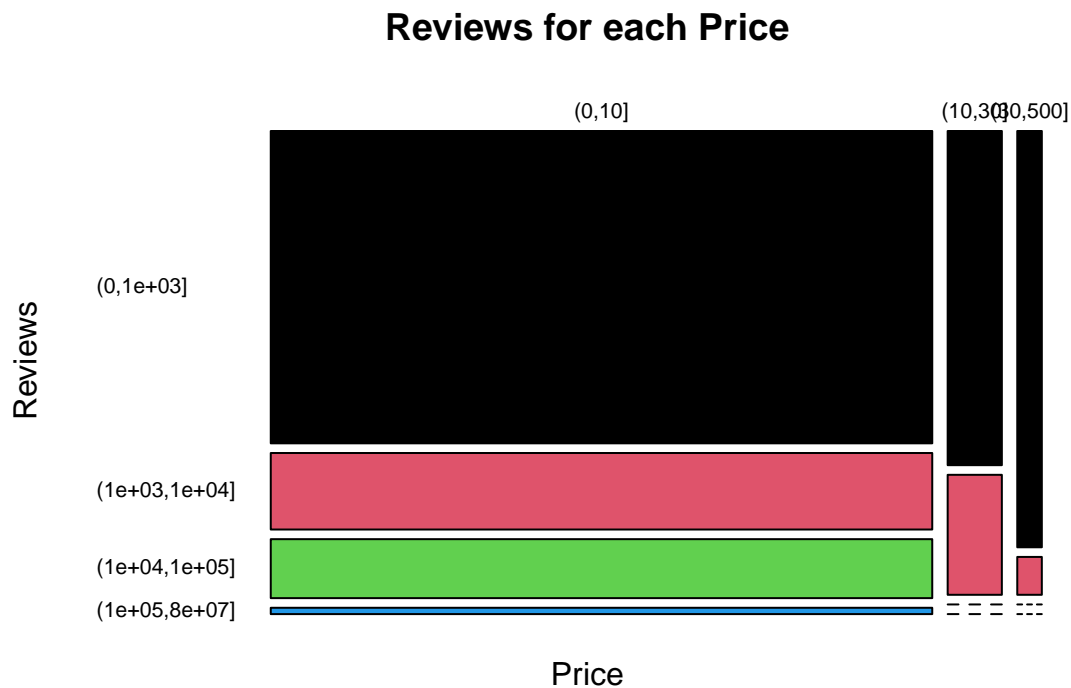
Review Vs Price

Which price and review range contains the most apps?

```
library(vcd)
```

```
## Loading required package: grid
```

```
mosaicplot(table(priceCut,reviewCut),ylab = "Reviews",xlab = "Price",main="Reviews for each Price",col=
```



The exact percentages for the above Mosaic Plot:

```
prop.table(table(priceCut,reviewCut))
```

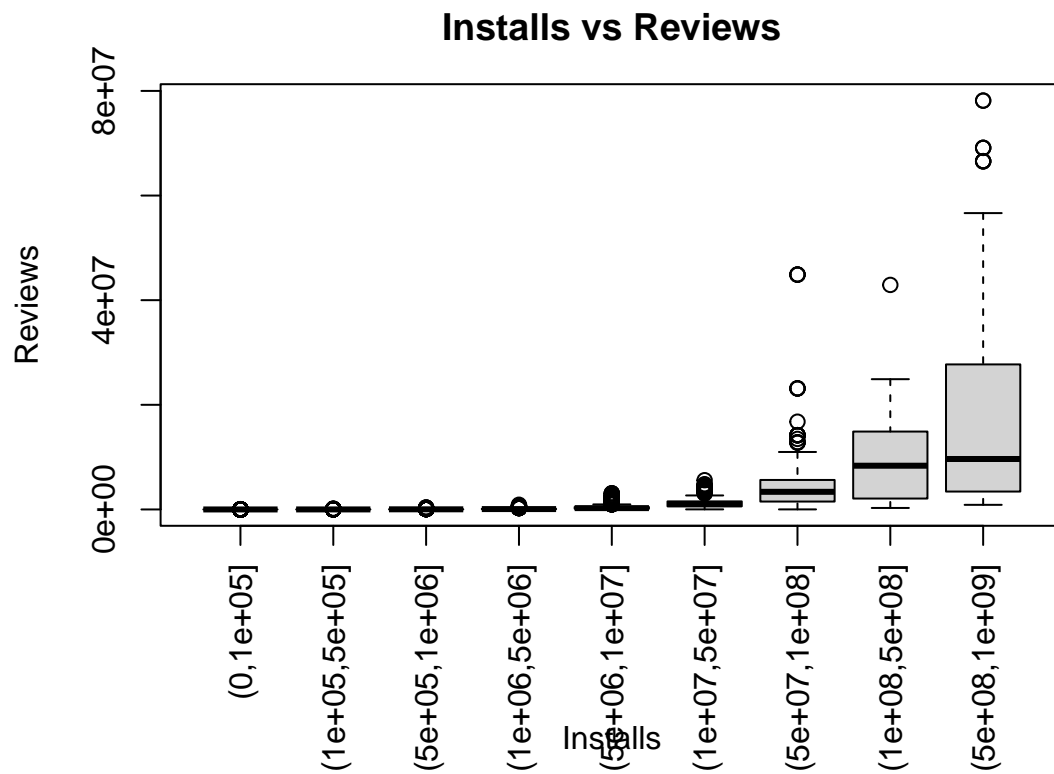
```
##           reviewCut
## priceCut  (0,1e+03] (1e+03,1e+04] (1e+04,1e+05] (1e+05,8e+07]
##   (0,10]  0.614640884  0.150552486  0.116022099  0.012430939
##   (10,30] 0.053867403  0.019337017  0.000000000  0.000000000
##   (30,500] 0.030386740  0.002762431  0.000000000  0.000000000
```

Reviews Vs Number of Installs

As number of installs grows, the median of number of reviews increases as well:

```
par(xpd=T,mar=c(8,4,3,5))

boxplot(Reviews~installsCut,data = playstore,las=3,xlab = "",main="Installs vs Reviews")
mtext(text = "Installs",side = 1,line = 5)
```



```
par(cex=1)
cor(playstore$Reviews,Installs_,method = "spearman")
```

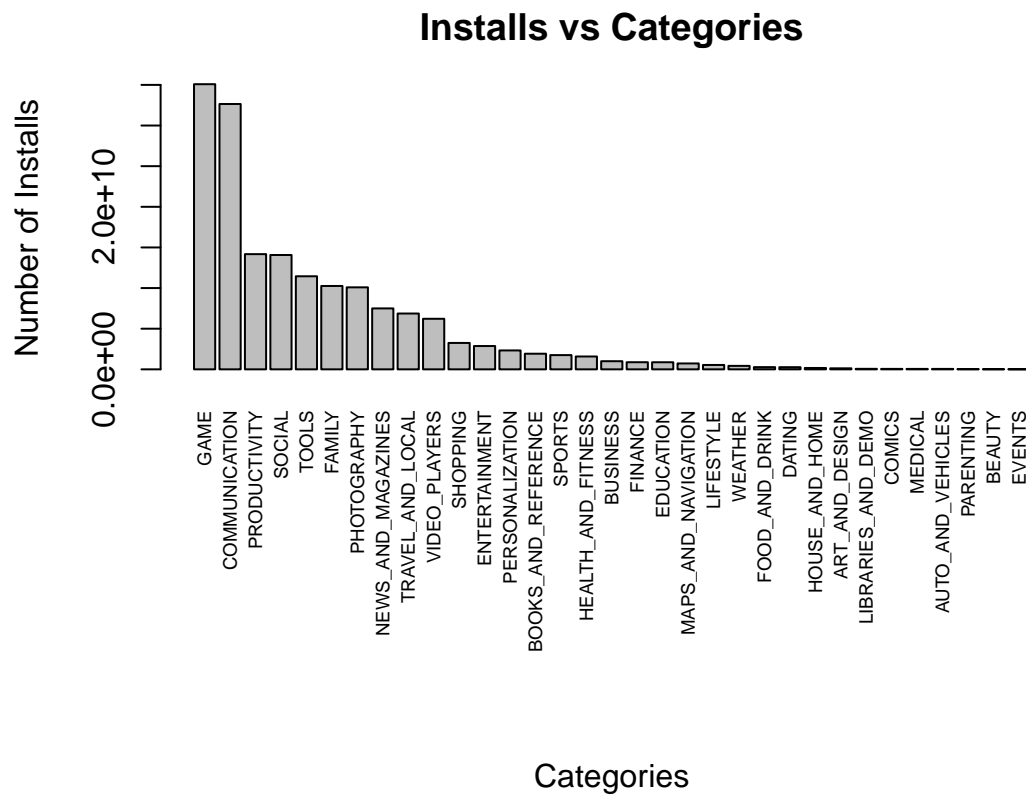
```
## [1] 0.9712189
```

Installs per Category

Category with most number of installs:

```
par(xpd=T,mar=c(12,4,3,5))
install_cat_table<-aggregate(Installs_~Category,data = playstore,FUN=sum)
vsel<-order(install_cat_table[,2],decreasing=T)
tab.agg<-install_cat_table[,2]
names(tab.agg)<-install_cat_table[,1]

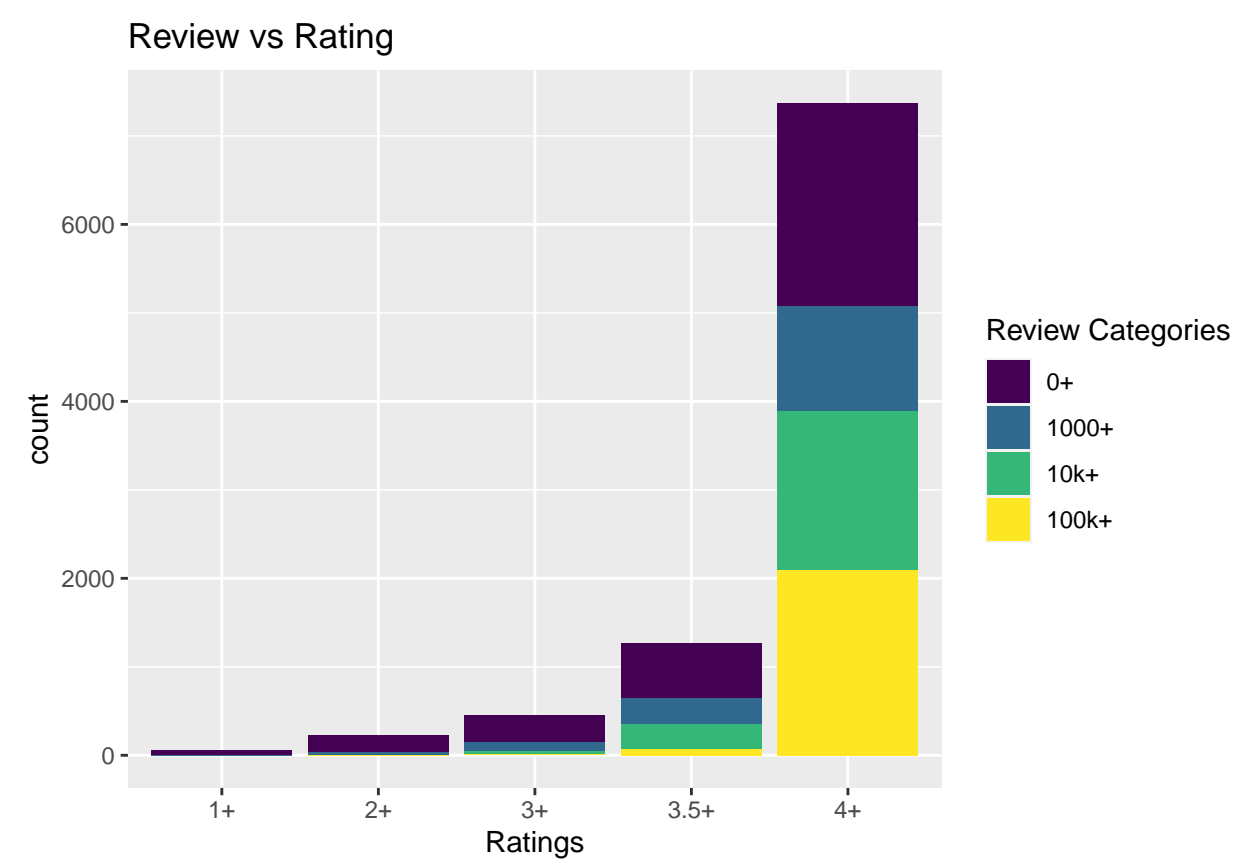
df.bar<-barplot(tab.agg[vsel],las=3,cex.names = 0.6,ylab = "Number of Installs",main="Installs vs Category")
par(cex=1)
mtext(text = "Categories",side = 1,line = 10)
```



Reviews Vs Rating

What is the distribution of reviews for each rating?

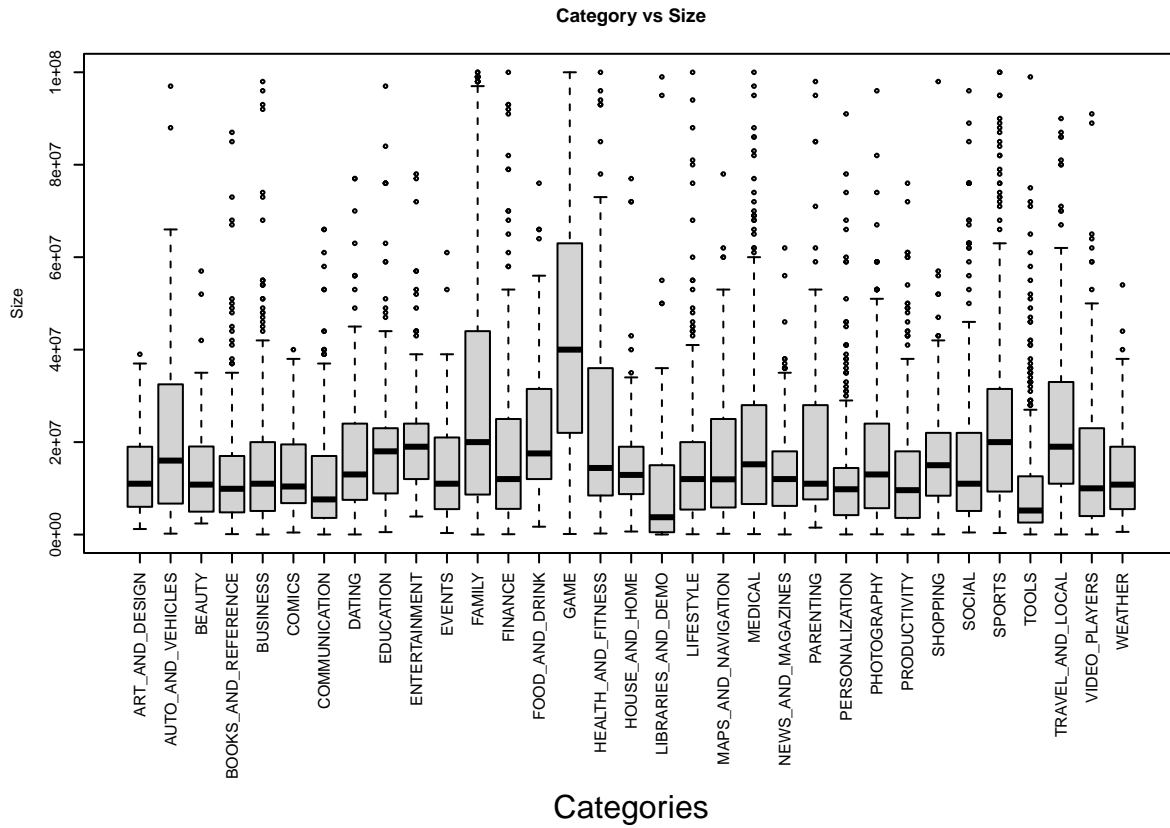
```
ggplot(subset(playstore, !is.na(ratingCut)), aes(x=ratingCut, fill=reviewCut)) +
  geom_bar(position = "stack") + ggtitle("Review vs Rating") +
  xlab("Ratings") + guides(fill=guide_legend(title="Review Categories"))
```



Category vs Size

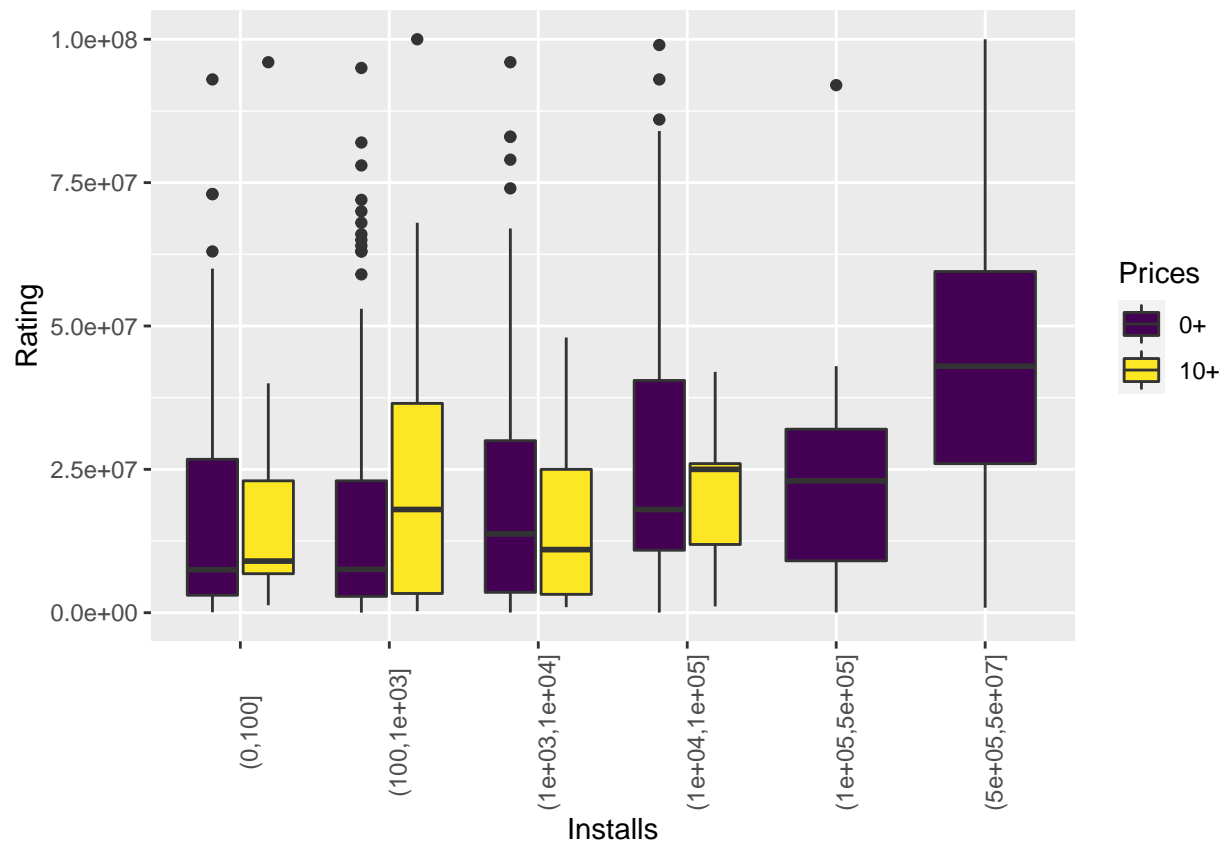
The most size instensive category:

```
par(xpd=NA,mar=c(15,5,4,3),cex=0.5)
boxplot(Size~Category,data=playstore,las=3,ylab = "Size",xlab = "",main="Category vs Size")
par(cex=1)
mtext(text = "Categories",side = 1,line = 13)
```

some Multivariate plots

```
playstore_complete=playstore[complete.cases(playstore),]
ggplot(playstore_complete,aes(x=installsSuperCut,y=Size,fill=priceSuperCut))+geom_boxplot()+xlab("Installs")
```



```
library(ggmosaic)
```

```
## Warning: package 'ggmosaic' was built under R version 4.1.2
```

```
##
```

```
## Attaching package: 'ggmosaic'
```

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

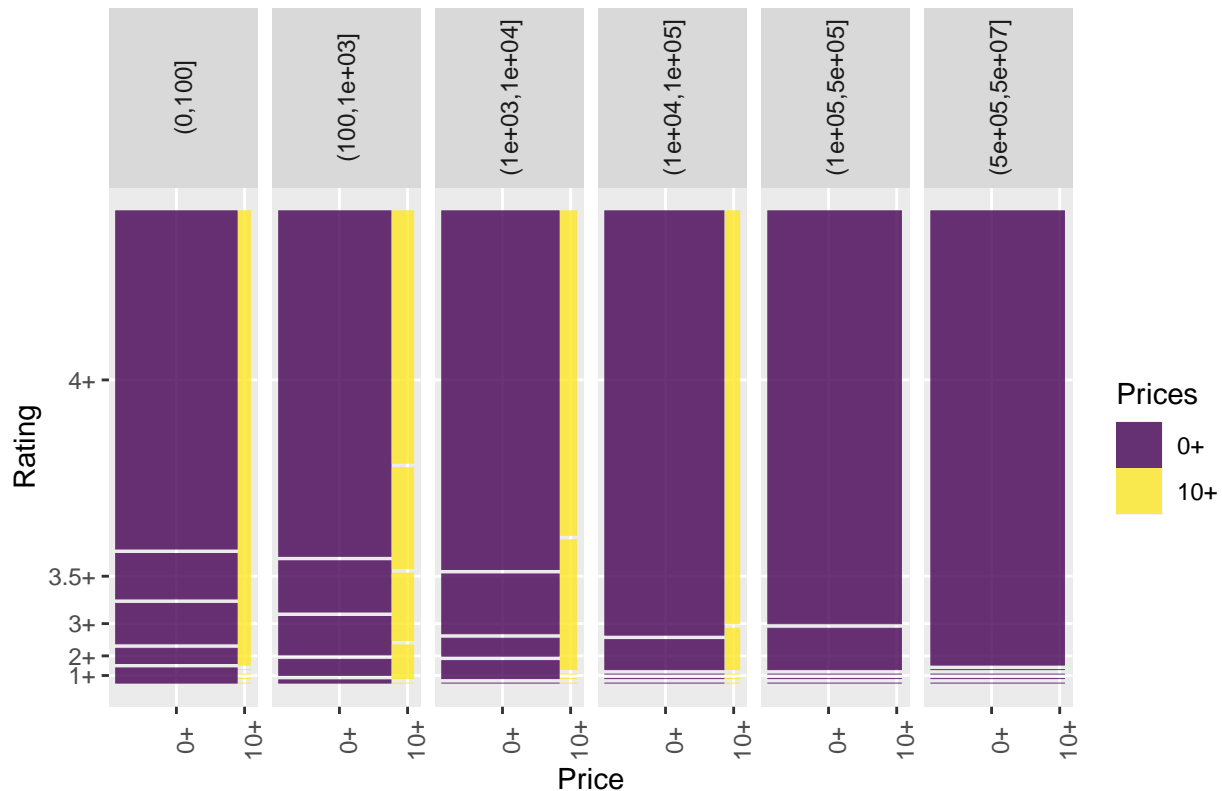
```
##
```

```
## mosaic, spine
```

```
# ggplot(playstore_complete)+geom_mosaic(aes(x=product(ratingCut,priceCut,installsCut),fill=installsCut,
```

```
ggplot(playstore_complete)+geom_mosaic(aes(x=product(ratingCut,priceSuperCut),fill=priceSuperCut))+face
```

Price Vs Rating Vs installs



Inferences and Conclusions

In this notebook we did Exploratory Data Analysis on Google Play Store Apps dataset. We drew interesting inferences from this dataset:

-
- Based on the dataset we can infer that most of the Apps in play store belongs to Family and Gaming categories followed by Tools, Medical and Business.
 - Also based on the type metric it seems that only 7% of apps are paid and around 93% of apps are free to install.
 - From the ratings, it appears that people are tend to give ratings in the range of 3 to 5. And the more reviews an app has, the better is its rating.
 - From the sizes, it appears that most apps tend to be below 20 MB in size. And the most size intensive apps are mostly games.
 - Based on the popularity, the apps in Gaming category were installed most number of time followed by Communication, Productivity and Social.
 - Based on the number of installs, the facebook app instilled most number of time followed by gmail.
 - Based on the reviews, Facebook, WhatsApp and instagram has most no of Reviews in google platy store.

- Most (61%) of the apps are $[0,10)$ dollars and have $[0,1000)$ Review
- There is also a big correlation between number of installs and number of reviews for each app. The more an app is installed, the more reviews it has.