

OSEMN Assignment -Best Buy customers rate iPhones with larger memory more favorably

Overview

Best Buy Co., Inc. is an American multinational consumer electronics corporation. people can have good electronic goods with reasonable prices from best buy using it's website or by visiting stores. it is headquartered in Richfield, Minnesota, a Minneapolis suburb. It operates in the United States, Puerto Rico, Mexico, Canada and China.[5] The company was founded by Richard M. Schulze and Gary Smoliak in 1966 as an audio specialty store. In 1983, it was renamed and rebranded with more emphasis placed on consumer electronics.

Best Buy's subsidiaries include CinemaNow, Geek Squad, Magnolia Audio Video, Pacific Sales, and Cowboomb. Best Buy operates under the Best Buy, Best Buy Mobile, Geek Squad, Magnolia Audio Video and Pacific Sales brands in the United States; the Best Buy, Geek Squad, Best Buy Mobile brands in Canada; Best Buy Mobile and Five Star in China; and Best Buy, Best Buy Express, and Geek Squad in Mexico.[5] Best Buy sells cellular phones with phones from Verizon Wireless, AT and T Mobility, Sprint PCS, Boost Mobile, and T-Mobile USA,[6] in regular stores and standalone Best Buy Mobile stores in shopping malls.

Best Buy was named "Company of the Year" by Forbes magazine in 2004,[7] "Specialty Retailer of the Decade" by Discount Store News in 2001,[8] ranked in the Top 10 of "America's Most Generous Corporations" by Forbes in 2005 (based on 2004 giving),[9] and made Fortune magazine's List of Most Admired Companies in 2006 source:wikipedia Best Buy Mobile stores in shopping malls.

This paper will check and confirm that whether iPhones which have larger memory have higher ratings than low memory iPhones or not. ratings of high phones with lower memory such as 16 Gb and 32Gb as well as higher memory phones such as 64 and 128GB. I started our work with collecting data about iPhones such as their memory and ratings. Next step would be to take average of ratings memorywise and compare them.

Gathering Data

To gather the data needed for this comparison, I utilized the Best Buy review API . which returns JSON (JavaScript Object Notation) which has three data elements that I am interested in for this report. The ID ,SKU and Ratings.

ID works as an identifier assigned to a particular product. The SKU attribute provides the SKU identifier for the product of a review .Data type of SKU is The rating attribute provides an average of all the ratings submitted for a product by reviewers. The ratings that can be submitted using a range from zero to five where five is the best rating that can be given. Ratings may be returned using decimals (example 3.5).data type of ratings is float.

The Reviews API provides access to individual reviews that have been submitted on BESTBUY.COM for the products we sell. You can use it to retrieve information such as the author of the review, the date that the review was submitted, the reviewer's rating (0-5 where 5 is the best) of a product and the reviewer's comments about the product. Along with real-time calls archives are available for reviews data. These are updated daily. Refer to archives for documentation on our archives.

The following shows the code needed to pull the data from best buy review api

```
library(jsonlite)

##
## Attaching package: 'jsonlite'
##
## The following object is masked from 'package:utils':
##
##      View

library (plyr)

url <- 'http://api.remix.bestbuy.com/' # the Best Buy API
path <- 'v1/reviews' # Reviews path
search <- '(sku=1722036|sku=172463|sku=1751846|sku=1755302|sku=1752845|sku=1761054|sku=1755302)'
format <- '?format=json&' # Results will be in json format
key <- 'apiKey=ztk9ycxpg6rcmu9wfbq7gqe9&' # here is my key
# get your own at:
# https://developer.bestbuy.com/documentation/reviews
# Select 'Get API Key' in top right hand corner
fields <- 'show=id,sku,rating&' # I only want these fields
pageSize <- 'pageSize=100' # 100 results per page is the max allowed

# Paste0 allows me to concatenate all of the above into one long URL
URLall <- paste0(url, path, search, format, key, fields, pageSize)

#Here is what it looks like all together
# The above URL will create a page in json format
# I can read the json and convert it to a list using the jsonlite package
l = jsonlite::fromJSON(URLall)
# If I look at l, I see that
# My search yielded 7291 total records over 73 total pages
# But the the API only returned 100 records (100 is the max per page)

head(l)
```

```

## $from
## [1] 1
##
## $to
## [1] 100
##
## $total
## [1] 7291
##
## $currentPage
## [1] 1
##
## $totalPages
## [1] 73
##
## $queryTime
## [1] "0.037"

# to get all 7291 records, I will need to access pages 1-73
# if you look at the documentation for the api:
# https://developer.bestbuy.com/documentation#pagination-pagination
# you will see that you can specify the page you want

pageText <- '&page=' # i will be the page number

# r allows many types of loops - including a for loop

for (i in 1:73)
{
  URLpage <-paste0(URLall, pageText, i)

  l = jsonlite::fromJSON(URLpage)

  # The ldply function in the plyr package allows us
  # to convert our list (l) to a dataframe (df)
  dfnew <- ldply (l, data.frame)

  # the first 9 rows are unneeded header information
  # the first 10 columns are unneeded header information
  # we will drop both the unneeded columns and rows
  dfnew<-dfnew[-c(1:9),-c(1:10)]

  # we will write each page out to a text file
  # each page will be appended to the bestbuy.csv file
  write.table(dfnew, file="bestbuy.csv", col.names=F, row.names=F, append=T, sep=",")
  # ignore the warning messages

```

```

}

# Now we can read in the complete set of page data
iphone.data <-read.csv("bestbuy.csv", header = F)

# Since we set col.names=F above, we need to name our columns
iphone.data<-rename(iphone.data,c(V1="ID", V2="SKU", V3 = "Rating"))
# now, our data is ready to analyse

```

Scrubbing the Data

to scrub the data means get useful data which is important to calculate the result for the question asked above. here we need the group of higher and lower memory iPhone and their average ratings. so we take ID, SKU ID's, ratings for iPhone 16GB, 32GB, 64GB and 128GB and take the average of ratings

```

h1_iphonedata<-iphone.data[grepl("7640008",iphone.data$SKU),]
h2_iphonedata<-iphone.data[grepl("7640035",iphone.data$SKU),]
h3_iphonedata<-iphone.data[grepl("1752872",iphone.data$SKU),]
h4_iphonedata<-iphone.data[grepl("1755715",iphone.data$SKU),]
mean_h1_iphonedata<-mean(h1_iphonedata$Rating,)
mean_h2_iphonedata<-mean(h2_iphonedata$Rating,)
mean_h3_iphonedata<-mean(h3_iphonedata$Rating,)
mean_h4_iphonedata<-mean(h4_iphonedata$Rating,)

```

Exploring the data

In this section, we will take a deeper look into our Useful data. Our data is presented in a tabular format with three columns. The first column gives the ID, the second column gives the desired SKU, and the third column gives rating. Let us begin by determining the data type of Data using the class function.

```

class(h1_iphonedata)
## [1] "data.frame"

class(h2_iphonedata)
## [1] "data.frame"

class(h3_iphonedata)
## [1] "data.frame"

class(h4_iphonedata)
## [1] "data.frame"

```

Now let us go ahead and examine the structure of useful Data using the str function. This function is used to the structure of an R object.

```
str(h1_iphonedata)

## 'data.frame': 6186 obs. of 3 variables:
## $ ID : int 49096232 47146465 47102302 47094567 45324375 45302510 44572353 44411807 4
## $ SKU : int 7640008 7640008 7640008 7640008 7640008 7640008 7640008 7640008 7
## $ Rating: int 5 5 5 5 5 5 5 5 5 ...

str(h2_iphonedata)

## 'data.frame': 24181 obs. of 3 variables:
## $ ID : int 50562784 49721452 49707095 48364027 48033306 48018326 47673724 47451633 4
## $ SKU : int 7640035 7640035 7640035 7640035 7640035 7640035 7640035 7640035 7
## $ Rating: int 5 5 4 5 5 3 5 5 5 ...

str(h3_iphonedata)

## 'data.frame': 2417 obs. of 3 variables:
## $ ID : int 41336160 39407850 41336160 39407850 38310323 38094006 37718426 35989100 3
## $ SKU : int 1752872 1752872 1752872 1752872 1752872 1752872 1752872 1752872 1
## $ Rating: int 5 5 5 5 5 4 4 5 5 ...

str(h4_iphonedata)

## 'data.frame': 11732 obs. of 3 variables:
## $ ID : int 16780390 16779462 16779383 16778428 16780390 16779462 16779383 16778428 4
## $ SKU : int 1755715 1755715 1755715 1755715 1755715 1755715 1755715 1755715 1
## $ Rating: int 4 5 5 5 4 5 5 5 5 ...
```

The summary of the dataset displays the minimum, first quarter, median, mean, third quarter, and maximum value of each variable. The summary is displayed as followed

```
summary(h1_iphonedata)

##           ID           SKU           Rating
## Min.   :37230684   Min.   :7640008   Min.   :3.000
## 1st Qu.:38316781   1st Qu.:7640008   1st Qu.:5.000
## Median :39404405   Median :7640008   Median :5.000
## Mean   :40092620   Mean   :7640008   Mean   :4.856
## 3rd Qu.:41173868   3rd Qu.:7640008   3rd Qu.:5.000
## Max.   :49096232   Max.   :7640008   Max.   :5.000

summary(h2_iphonedata)
```

```
##          ID          SKU          Rating
##  Min.    :36906235  Min.    :7640035  Min.    :1.000
## 1st Qu.:38724484  1st Qu.:7640035  1st Qu.:5.000
## Median :42183487  Median :7640035  Median :5.000
## Mean    :41534137  Mean    :7640035  Mean    :4.831
## 3rd Qu.:43876564  3rd Qu.:7640035  3rd Qu.:5.000
## Max.    :50562784  Max.    :7640035  Max.    :5.000
```

```
summary(h3_iphonedata)
```

```
##          ID          SKU          Rating
##  Min.    :18315609  Min.    :1752872  Min.    :3.000
## 1st Qu.:20468079  1st Qu.:1752872  1st Qu.:5.000
## Median :22251119  Median :1752872  Median :5.000
## Mean    :25423821  Mean    :1752872  Mean    :4.745
## 3rd Qu.:32568988  3rd Qu.:1752872  3rd Qu.:5.000
## Max.    :41336160  Max.    :1752872  Max.    :5.000
```

```
summary(h4_iphonedata)
```

```
##          ID          SKU          Rating
##  Min.    :16778428  Min.    :1755715  Min.    :1.000
## 1st Qu.:20548244  1st Qu.:1755715  1st Qu.:4.000
## Median :24163676  Median :1755715  Median :5.000
## Mean    :28244691  Mean    :1755715  Mean    :4.618
## 3rd Qu.:35348344  3rd Qu.:1755715  3rd Qu.:5.000
## Max.    :47363454  Max.    :1755715  Max.    :5.000
```

Result This section will display the result after analyzing the data

```
mean_h1_iphonedata
```

```
## [1] 4.855965
```

```
mean_h2_iphonedata
```

```
## [1] 4.831479
```

```
mean_h3_iphonedata
```

```
## [1] 4.745139
```

```
mean_h4_iphonedata
```

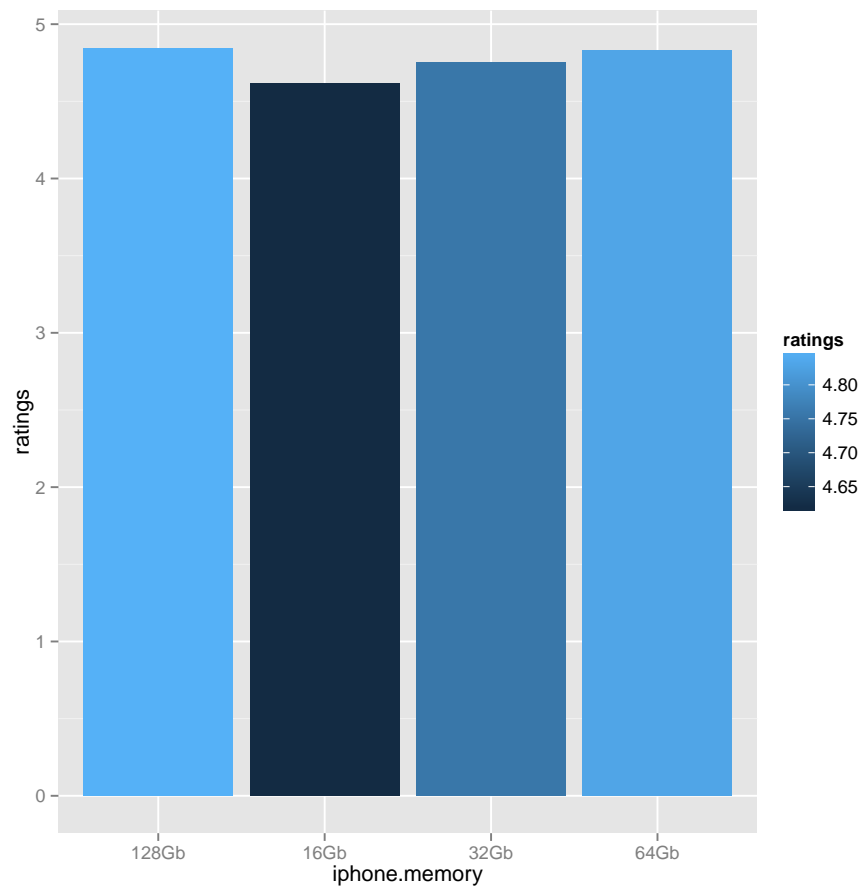
```
## [1] 4.617542
```

```
iphone.memory<-c("128Gb","64Gb","32Gb","16Gb")
ratings=c(4.845714,4.827815,4.75431,4.618387)
dfg=data.frame(iphone.memory,ratings)
dfg

##  iphone.memory ratings
## 1      128Gb 4.845714
## 2      64Gb 4.827815
## 3      32Gb 4.754310
## 4      16Gb 4.618387
```

Creating graph

```
library(ggplot2)
g<-ggplot(dfg,aes(x=iphone.memory,y=ratings,fill=ratings))+geom_bar(stat="identity")
g
```



As from the graph it is clear that average rating 16Gb is 4.65 and as memory is increasing from 32gb to 128gb average rating is also increasing (4.70, 4.75 and 4.80). Hence it is clear that best buy customers rate iPhone with the larger memory more favorably.