

Assignment 4: Amran Ahmed

Amran Ahmed

04/03/2022

```
#Importing datasets
```

```
if (!require('ggplot2'))  
{  
  install.packages('ggplot2');  
  library(ggplot2);  
}
```

```
## Loading required package: ggplot2
```

```
## Warning in register(): Can't find generic 'scale_type' in package ggplot2 to  
## register S3 method.
```

```
#importing my data sets  
data2015 <- read.csv("2015_ontario_public_library_statistics_open_data_dec_2017rev.csv")  
  
data2016 <- read.csv("2016_ontario_public_library_statistics_open_data_2016.csv")  
  
data2017 <- read.csv("2017_ontario_public_library_statistics_open_data_july_2019_rev1.csv")
```

```
#Merged all files into one object
```

```
#organizing my column names  
common_columns <- Reduce(intersect, list(colnames(data2015), colnames(data2016), colnames(data2017)))
```

```
#combining my files data  
datacombined <- rbind(subset(data2015, select = common_columns), subset(data2016, select = common_columns), subset(data2017, select = common_columns))
```

```
#Creating a new column that represents operating revenue per active card holder (total operating Revenue  
/ number of active card holders).
```

```
#Removing the commas  
datacombined$B2.9..Total.Operating.Revenues <- gsub(",", "", datacombined$B2.9..Total.Operating.Revenues)  
  
datacombined$A1.14..No..of.Active.Library.Cardholders <- gsub(",", "", datacombined$A1.14..No..of.Active.Library.Cardholders)  
  
#Checking the class of the columns  
class(datacombined$A1.14..No..of.Active.Library.Cardholders)
```

```
## [1] "character"
```

```
class(datacombined$B2.9..Total.Operating.Revenues)
```

```
## [1] "character"
```

```
#Making NA into zero's
```

```
datacombined[is.na(datacombined)] <- 0
```

```
# Making the variables numeric form
```

```
datacombined$A1.14..No..of.Active.Library.Cardholders <- as.numeric(datacombined$A1.14..No..of.Active.L
```

```
datacombined$B2.9..Total.Operating.Revenues <- as.numeric(datacombined$B2.9..Total.Operating.Revenues)
```

```
#Removing the zeros from the specified columns
```

```
datacombined <- subset(datacombined, datacombined$A1.14..No..of.Active.Library.Cardholders > 0 & datacom
```

```
#Dividing Operating Revenue by Active Card Holder
```

```
datacombined$B2.9..Total.Operating.Revenues / datacombined$A1.14..No..of.Active.Library.Cardholders
```

```
## [1] 106.711771 80.325359 126.425102 107.902344 205.356113 390.569620
## [7] 116.555893 206.356877 96.624116 201.863696 60.553506 91.902072
## [13] 151.567500 120.850985 69.581386 302.703125 130.684037 48.255371
## [19] 205.587570 88.434641 93.102474 170.855263 67.011848 149.464939
## [25] 145.128591 80.890073 264.297450 45.161580 113.392346 105.590629
## [31] 100.743738 150.912301 442.074469 74.795663 25.116438 209.057592
## [37] 109.165008 141.558969 77.022573 87.677265 165.610394 206.853414
## [43] 190.575955 138.507433 66.164443 56.319471 58.533333 208.913655
## [49] 73.432878 38.337781 84.838095 139.061472 248.936090 149.922004
## [55] 372.751220 107.945170 160.546154 29.830667 106.120000 146.583493
## [61] 72.164487 242.569177 54.133159 117.712737 218.292514 166.266421
## [67] 178.254192 94.132222 71.230469 139.651642 43.765333 131.111215
## [73] 186.716667 118.690000 114.532086 125.894146 43.364384 232.547310
## [79] 68.359786 214.441613 141.654289 190.952808 60.398788 113.891471
## [85] 53.790301 66.024507 71.891044 163.539935 165.087435 159.114120
## [91] 168.025806 30.294493 55.976809 121.879570 116.004558 325.496552
## [97] 112.869367 151.764017 19.989474 62.441333 121.178670 113.222993
## [103] 152.007663 93.012220 197.268702 81.249531 118.888818 155.329964
## [109] 107.408293 147.301955 193.036535 200.387682 280.572838 116.365539
## [115] 77.300657 73.615176 129.243243 128.644120 224.900000 161.950769
## [121] 299.026247 79.290918 161.306644 144.363296 162.176887 151.593859
## [127] 101.012994 636.750000 141.286070 102.973529 135.933264 105.002165
## [133] 142.982210 111.894260 169.877575 132.281601 62.386788 127.629481
## [139] 107.454187 73.904298 189.556004 74.110340 229.047244 442.788732
## [145] 122.375424 139.162454 147.106562 146.491761 20.816279 295.536913
## [151] 95.578261 4257.600000 98.504057 269.068650 199.868293 91.864660
## [157] 160.507799 48.998602 1179.750000 62.255385 191.002304 93.796243
## [163] 115.550000 177.463335 70.635271 477.369048 180.375575 86.581167
## [169] 78.870183 124.845395 40.912360 262.714286 60.081798 125.776124
## [175] 93.354260 244.145329 295.983063 73.059490 109.930769 22.886947
## [181] 135.978166 120.881826 48.928561 155.039422 1322.322581 158.856919
## [187] 53.400114 63.539545 47.332021 203.698938 100.560897 125.539207
## [193] 153.055556 349.527778 99.350877 208.381356 215.163069 173.937219
## [199] 157.455276 76.551477 155.513249 129.043111 215.108025 86.157506
```

## [205]	111.798650	137.029165	83.757526	271.973373	96.293919	230.657192
## [211]	68.673167	249.577116	25.453875	133.143695	83.829708	55.412323
## [217]	138.738182	109.374663	154.602072	193.412121	77.367898	126.260054
## [223]	294.850000	187.962062	121.331210	169.295276	101.979915	184.816131
## [229]	87.557119	109.332440	171.580488	169.365714	95.352941	621.223301
## [235]	177.781726	181.870098	85.320038	117.683612	1081.437500	106.307306
## [241]	113.452632	225.987217	104.783251	56.549446	1010.100000	912.164420
## [247]	192.880700	93.512876	87.209448	99.183201	105.277567	86.707576
## [253]	108.796974	93.597409	239.621145	109.925806	101.503557	245.880658
## [259]	27.154462	151.568850	177.370803	49.259098	248.051338	287.072648
## [265]	100.034858	183.819643	32.808791	116.006873	232.264000	141.130268
## [271]	174.866946	196.249933	37.959049	119.340000	135.561240	212.089208
## [277]	146.284965	208.610315	187.005554	86.421166	58.151111	54.942104
## [283]	43.410322	72.697457	197.913194	113.671145	614.114286	127.108168
## [289]	53.076273	51.223958	89.750439	164.829244	168.901627	148.225271
## [295]	217.830913	151.823116	265.767750	147.704429	131.689416	83.639151
## [301]	138.853073	80.411833	196.549407	377.240000	83.630827	92.450459
## [307]	78.809917	199.661299	95.277778	162.045303	83.206790	81.787500
## [313]	135.124173	98.051471	216.810714	586.732143	115.932687	221.263345
## [319]	102.843046	214.842759	53.806763	89.299492	28.004000	182.279786
## [325]	50.039066	259.888889	135.764750	48.018204	205.514177	200.756598
## [331]	105.021980	123.686833	133.992000	73.392500	163.112805	116.187643
## [337]	72.284694	253.860947	112.222963	102.602520	99.083024	97.852610
## [343]	156.695189	444.642386	66.494212	25.197674	233.045990	129.779652
## [349]	134.689309	232.883929	84.479201	156.692297	207.333286	228.142764
## [355]	143.990816	65.431511	57.322581	83.619048	174.814061	73.038916
## [361]	36.277068	96.523906	140.580856	291.067114	185.946024	336.339623
## [367]	113.294710	166.687023	275.126946	318.360000	91.645935	80.718406
## [373]	296.895613	88.645374	118.084290	214.765321	166.889215	186.930057
## [379]	99.478099	69.686347	134.825337	44.859416	140.289000	129.373333
## [385]	115.343750	107.737101	161.141764	32.035616	466.326613	69.992063
## [391]	226.216560	146.275000	240.986677	64.435583	207.751066	42.083333
## [397]	65.534375	79.842620	182.844572	139.790085	146.664530	169.351304
## [403]	157.682699	59.957070	118.595745	62.581674	144.013514	113.585394
## [409]	158.340032	9.736538	100.373563	157.291416	116.810038	124.558719
## [415]	95.076810	174.857143	133.649235	124.276651	164.702032	114.195579
## [421]	103.920101	200.678554	196.891057	286.379410	148.510870	79.027455
## [427]	85.332100	138.363636	138.984428	522.661538	134.119186	295.813411
## [433]	101.749405	165.031206	203.385906	150.618721	143.849691	74.384312
## [439]	693.285714	115.541353	90.964441	116.740211	105.257225	142.395769
## [445]	93.044118	189.479848	127.301763	96.807504	122.716972	109.359667
## [451]	75.463169	193.047076	47.086935	230.057534	409.656357	134.686656
## [457]	177.882774	170.043134	149.843342	19.147727	184.000000	134.905952
## [463]	4157.600000	101.586399	137.619647	170.625227	91.048375	123.074873
## [469]	43.422650	864.325000	74.058115	192.220513	69.334776	152.556164
## [475]	178.159029	76.729255	441.102273	216.144653	96.663254	81.306673
## [481]	128.165674	38.924612	112.295082	62.069897	116.887646	81.251429
## [487]	224.124561	260.394962	67.699717	94.439216	251.762027	138.191731
## [493]	223.481640	65.423697	133.316589	1280.483871	164.152834	53.059600
## [499]	78.359744	45.550986	266.137626	96.047710	89.129401	66.666667
## [505]	84.030670	137.709677	192.936745	151.615144	146.735934	87.171642
## [511]	156.899200	137.925990	223.049634	75.607409	160.369184	173.914738
## [517]	174.171081	203.607407	105.892094	230.283806	63.203339	230.473744
## [523]	58.152985	145.438370	88.358576	53.030612	123.018622	109.118412

## [529]	158.923952	111.838542	78.125573	113.805699	270.912500	255.192972
## [535]	121.519681	162.088583	137.705728	195.461378	89.408884	114.046343
## [541]	159.588133	593.800000	246.333333	292.804878	185.198203	160.261521
## [547]	89.806686	120.307890	677.187500	95.847805	93.442105	221.231405
## [553]	81.214634	51.825437	365.100000	1052.010568	177.901402	91.813808
## [559]	93.590402	117.865823	72.752242	82.139062	44.808549	87.092185
## [565]	271.376147	109.959693	95.608604	212.653696	26.044614	188.750097
## [571]	213.636265	47.746483	231.975458	290.123388	91.553419	216.863338
## [577]	17.253304	286.306773	162.830769	150.695260	183.080172	172.289461
## [583]	32.841280	100.031847	137.789383	241.442399	154.069370	205.199539
## [589]	92.669331	72.349930	41.411483	57.905740	147.301176	72.879230
## [595]	231.926108	126.158545	553.566667	114.348463	58.555800	48.644886
## [601]	97.700018	177.556331	164.131100	180.623306	226.969834	129.835449
## [607]	260.837578	148.967373	147.203262	126.491379	146.008367	68.963396
## [613]	168.412752	287.240000	96.804461	71.125000	141.610032	210.544492
## [619]	74.755140	185.595974	88.664234	117.094851	137.382782	106.081897
## [625]	186.201014	583.576271	117.276252	189.196262	98.186605	300.909091
## [631]	72.199715	67.218750	136.510000	205.123245	57.668299	337.460000
## [637]	139.610759	110.832685	218.222236	173.294430	97.085516	131.096085
## [643]	202.929412	129.428571	168.715385	39.456300	71.229743	237.794872
## [649]	142.750590	118.807251	94.024484	113.643527	175.613411	327.074902
## [655]	69.075391	44.448485	233.395324	159.924072	125.740174	161.147771
## [661]	81.884254	128.724431	253.833910	155.329345	154.228523	71.166588
## [667]	64.573333	176.015571	105.982667	38.346498	128.811523	143.008837
## [673]	225.102881	170.014822	321.813636	140.858289	170.250000	292.956000
## [679]	518.360000	94.907474	85.590932	277.360745	46.475170	135.648046
## [685]	205.437873	170.519692	181.907965	111.657658	73.721190	137.387108
## [691]	46.432000	105.141647	130.960526	229.419355	233.955357	149.391443
## [697]	50.380822	492.458065	178.781928	237.101029	224.061633	253.416888
## [703]	77.129073	138.549180	41.062559	96.261000	73.706846	136.610962
## [709]	205.958214	149.478702	289.360596	134.408469	15.019654	105.355060
## [715]	68.587836	150.066667	53.264821	163.551067	12.163900	90.066895
## [721]	145.425400	136.043713	151.904851	77.446612	219.691698	132.553883
## [727]	148.826694	169.881215	115.317392	115.733650	177.771451	198.706762
## [733]	299.477464	155.066759	73.318410	80.340000	131.185185	148.334468
## [739]	568.815385	110.683924	361.048295	108.484406	167.611132	239.144231
## [745]	164.642202	143.730141	101.923039	656.200000	113.078947	95.976986
## [751]	111.794029	141.408719	181.098044	97.994879	187.317375	135.205521
## [757]	62.882829	125.607035	105.585182	58.332764	206.018664	129.484808
## [763]	216.541985	310.838150	120.503350	175.899056	199.228814	170.815126
## [769]	20.425000	221.856250	137.958549	363.366667	119.410596	146.377203
## [775]	179.718157	101.392118	122.427732	89.920536	987.900000	67.361732
## [781]	177.329923	73.641243	118.143460	170.039611	70.405863	691.450000
## [787]	212.252415	71.799893	88.810971	131.438623	39.938731	143.362832
## [793]	73.194576	110.531700	57.713889	258.484000	228.285714	79.150142
## [799]	169.627451	164.751532	148.472875	183.011408	205.635223	142.124551
## [805]	1045.138889	171.491278	52.292974	87.179679	41.982225	308.953055
## [811]	112.399616	170.846065	92.326667	70.681077	147.903226	268.932865
## [817]	145.972341	144.311111	104.326707	174.264676	141.177373	260.028008
## [823]	117.360363	145.224606	330.792866	159.497802	228.426667	134.503925
## [829]	222.083333	93.147802	321.666473	44.161616	144.823141	103.170672
## [835]	56.006699	148.565879	113.493169	173.913754	98.413793	79.340213
## [841]	151.140940	290.887500	296.216783	185.728163	182.244898	110.251532
## [847]	179.219979	89.788765	109.472652	144.981848	454.973333	407.453846

```
## [853] 464.993750 173.005482 294.556461 97.839046 92.950076 804.097561
## [859] 110.724331 91.041026 276.912910 84.336585 65.770387 385.120000
## [865] 1041.451910 215.318996 57.221992 77.358232 118.772529 79.319520
## [871] 119.400000 51.282986 78.241710 355.076923 112.742852 95.836355
## [877] 215.525000 30.582349 186.745799 264.530883 184.664495 234.570502
## [883] 300.387232 124.319209 193.668218 41.169604 265.450909 140.544000
## [889] 170.715209 226.761376 188.490511 31.184795 117.875000 107.906382
## [895] 236.494644 165.234729 213.374567 91.651453 70.506980 44.395122
## [901] 82.305238 192.078910 70.546171 294.758824 134.061499 559.666667
## [907] 149.595066 57.929555 44.165644 101.029023 186.662175 150.462099
## [913] 191.888845 235.069203 177.057416 283.810078 147.106479 149.008870
## [919] 148.262987 125.347220 61.601161 111.863551 893.780000 100.827612
## [925] 61.522876 102.530401 219.009448 84.821053 189.225933
```

```
#duplicating my dataframe
```

```
newdatacombined <- datacombined
```

```
#Creating new variable column for Operating Revenue per Active Card Holder
```

```
newdatacombined$newcolumn <- newdatacombined$B2.9..Total.Operating.Revenues / newdatacombined$A1.14..No
```

INTRODUCTION

#The objective of this project is to understand how Ontario Public Libraries can be more successful? In this analysis we will be looking at how to improve Ontario public libraries and the metric for this will mostly be revenue as a cue for success. The data set that will be used is between 2015-2017. In this analysis we will gather the data for three different insights which can be used to improve the success of Ontario public libraries. Success Criteria for this analysis includes a library being more successful when it has: Higher revenue, and more cardholders. The measured data set or columns that will be looked at include: revenue, and the number of active library cardholders. The categorical data set or columns that will be looked at include: cities or the town. The assumptions in the analysis include: Libraries with more visitors have higher revenues. Having a higher revenue makes a library more successful. To analyze this data, the analysis will gain insight from data visualized looking at tables, graphs, mean distribution, median distribution, and the quantile points of the data.

#INSIGHT 1: Creating a table of two columns that has the data for city names and newcolumn (total revenue per active cardholder). The insight we will gain from this data will indicate which cities have the highest operating revenue per cardholder.

```
#Making newcolumn and City/Town into one object
```

```
twodatacombined<- dplyr::select (newdatacombined,newcolumn,A1.10.City.Town)
```

```
#Putting the new object in ascending order by values in (newcolumn)
```

```
twodatacombined <- twodatacombined[order(twodatacombined$newcolumn),]
```

```
#Looking at the last values in the object, this shows the highest values in (newcolumn) which shows th  
tail(twodatacombined)
```

```
##      newcolumn A1.10.City.Town
## 305  1081.438  Seine River FN
## 206  1179.750      Gogama
## 623  1280.484  Garden Village
## 243  1322.323  Garden Village
```

```
## 579 4157.600 Britt
## 199 4257.600 Britt
```

#INSIGHT 2: looking at the average and overall distribution, with plot of revenue vs cardholders. Seeing higher total revenue with higher total cardholder. Libraries with more cardholders have higher total operating revenue numbers and so the insight gained here is that we need to invest more in understanding how these libraries are able to attract more cardholders.

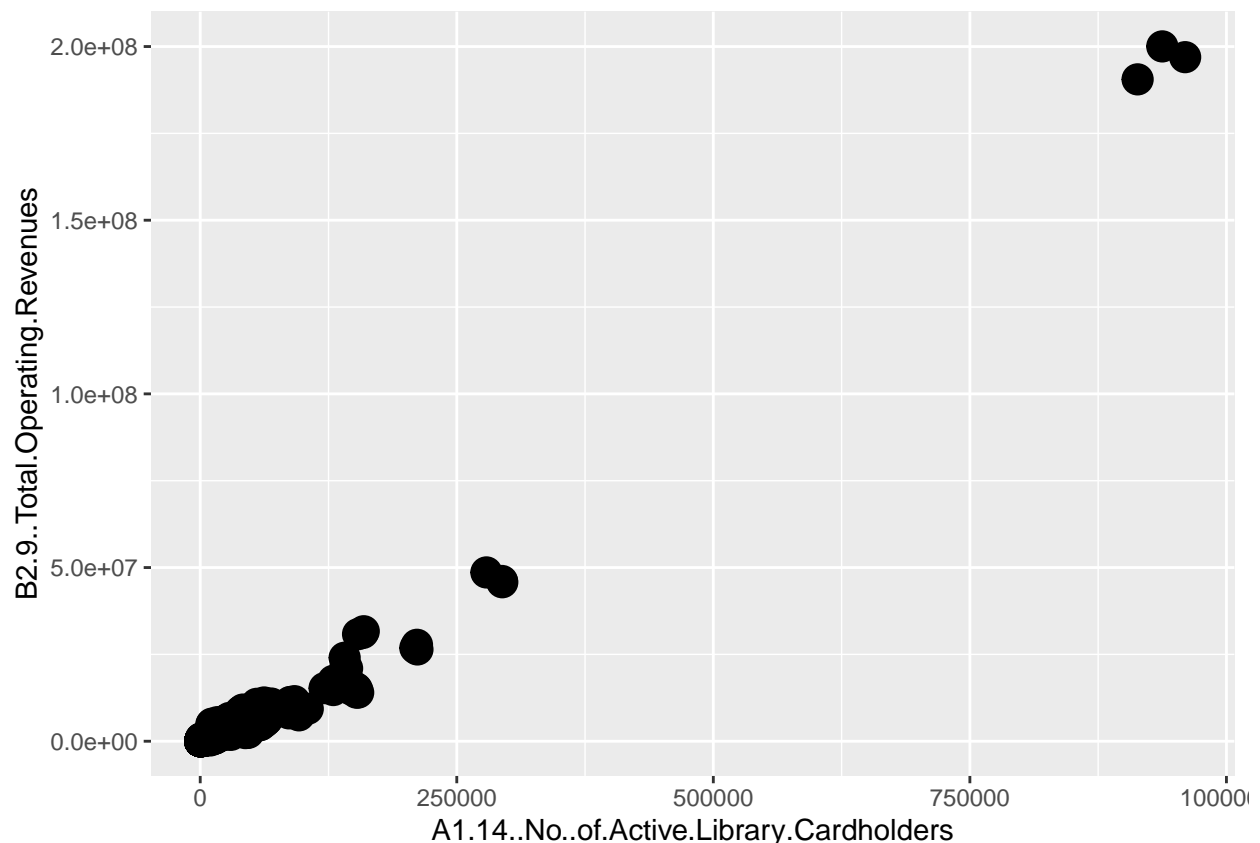
```
#average for newcolumn, total revenue per cardholder
mean(newdatacombined$newcolumn, na.rm = TRUE)
```

```
## [1] 172.1813
```

```
median(newdatacombined$newcolumn, na.rm = TRUE)
```

```
## [1] 134.906
```

```
#plot showing the distribution of operating revenue vs number of active cardholders
ggplot(data = newdatacombined, mapping = aes(x = A1.14..No..of.Active.Library.Cardholders, y = B2.9..Total.Operating.Revenues))
```



#INSIGHT 3: Looking at the quantile points of the newcolumn to understand the distribution of the newcolumn values. This data will show us the quantile points which demonstrate the five values for the minimum, the 1st quantile, the median, then mean 3rd quantile, and the maximum for every level of operating revenue per active cardholder.

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.1 --
```

```
## v tibble  3.1.6      v dplyr    1.0.7
## v tidyr   1.1.4      v stringr 1.4.0
## v readr   2.1.1      v forcats 0.5.1
## v purrr   0.3.4
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
library(ggplot2)
```

```
#this shows the quantile points
quantile(twodatacombined$newcolumn, probs=c(0, 0.25, 0.5, 0.75, 1))
```

```
##           0%           25%           50%           75%           100%
##    9.736538   91.041026  134.905952  187.317375 4257.600000
```

#Recommended Actions With the data gained from this report we've seen specific insights on total revenue per active cardholder. The government is advised to do further investigation on the top performing (highest revenue per cardholder) libraries logistical functioning information and their data to better understand their success and how to replicate that for other less performing libraries to improve their success..

The insights gained from looking at the median and average operating revenue per number cardholder is the knowledge of how well a typical library is expected to perform on average operating revenue per number and the median tells the exact middle value for which any other value is either below or above that number. This insight into the data allows the government to further visualize how the data is distributed and how high or low a library's operating revenue per number value is in comparison to other libraries.

The government should also perform more focused analysis on the libraries in the lowest and highest ends of the quantile points. This will provide the government with further information on exactly how to classify the lower end performing libraries, further research can look into the availability of resources in these areas including electronic resources which could be contributing to their decreased success.

```
#Fixing PDF knitting issue with this package install.packages("tinytex") tinytex::install_tinytex()
```

```
#Appendix of Lab 3 Source Code
```

```
if (!require('ggplot2'))
{
  install.packages('ggplot2');
  library(ggplot2);
}
```

```
#importing my data sets
```

```
data2015 <- read.csv("2015_ontario_public_library_statistics_open_data_dec_2017rev.csv")
```

```
data2016 <- read.csv("2016_ontario_public_library_statistics_open_data_2016.csv")
```

```
data2017 <- read.csv("2017_ontario_public_library_statistics_open_data_july_2019_rev1.csv")
```

```

#organizing my column names
common_columns <- Reduce(intersect, list(colnames(data2015),colnames(data2016),colnames(data2017)))

#combining my files data
datacombined <- rbind(subset(data2015, select = common_columns), subset(data2016, select = common_columns))

#Removing the commas
datacombined$B2.9..Total.Operating.Revenues <- gsub(",", "", datacombined$B2.9..Total.Operating.Revenues)

datacombined$A1.14..No..of.Active.Library.Cardholders <- gsub(",", "", datacombined$A1.14..No..of.Active.Library.Cardholders)

#Checking the class of the columns
class(datacombined$A1.14..No..of.Active.Library.Cardholders)

class(datacombined$B2.9..Total.Operating.Revenues)

#Making NA into zero's
datacombined[is.na(datacombined)] <- 0

# Making the variables numeric form
datacombined$A1.14..No..of.Active.Library.Cardholders <- as.numeric(datacombined$A1.14..No..of.Active.Library.Cardholders)

datacombined$B2.9..Total.Operating.Revenues <- as.numeric(datacombined$B2.9..Total.Operating.Revenues)

class(datacombined$A1.14..No..of.Active.Library.Cardholders)

class(datacombined$B2.9..Total.Operating.Revenues)

#Removing the zeros from the specified columns
datacombined <- subset(datacombined, datacombined$A1.14..No..of.Active.Library.Cardholders > 0 & datacombined$B2.9..Total.Operating.Revenues > 0)

#Dividing Operating Revenue by Active Card Holder
datacombined$B2.9..Total.Operating.Revenues / datacombined$A1.14..No..of.Active.Library.Cardholders

#duplicating my dataframe
newdatacombined <- datacombined

#Creating new variable column for Operating Revenue per Active Card Holder
newdatacombined$newcolumn <- newdatacombined$B2.9..Total.Operating.Revenues / newdatacombined$A1.14..No..of.Active.Library.Cardholders

```