

UE20CS312 - Data Analytics - Worksheet 1a - Part 2 - EDA with R

| ANOVA

PES University

‘SUNDEEP A, Dept. of CSE - PES1UG20CS445’

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Part I. Exploratory Data Analysis with R

Book Club Marketing Dataset

Problems

Problem 1 (1 point)

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.3.6      v purrr  0.3.4
## v tibble  3.1.8      v dplyr  1.0.9
## v tidyr   1.2.0      v stringr 1.4.0
## v readr   2.1.2      v forcats 0.5.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
cbc_df <- read_csv("C:\\Users\\HP\\Desktop\\PESU\\SEM - 5\\PES1UG20CS445\\DA\\1_b\\CharlesBookClubDataset.csv")
```

```
## New names:
## Rows: 4000 Columns: 26
## -- Column specification
## ----- Delimiter: "," chr
## (4): Name, Phone_No., Address, Job dbl (22): ...1, Seq#, ID#, Gender, M, R, F,
## FirstPurch, ChildBks, YouthBks, ...
## i Use `spec()` to retrieve the full column specification for this data. i
## Specify the column types or set `show_col_types = FALSE` to quiet this message.
## * `` -> `...1`
```

```
head(cbc_df)
```

```
## # A tibble: 6 x 26
##   ...1 `Seq#` `ID#` Gender      M      R      F FirstPurch Child~1 Youth~2 CookBks
##   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     0     1    25     1   297    14     2     22     0     1     1
## 2     1     2    29     0   128     8     2     10     0     0     0
## 3     2     3    46     1   138    22     7     56     2     1     2
## 4     3     4    47     1   228     2     1      2     0     0     0
## 5     4     5    51     1   257    10     1     10     0     0     0
```

```
## 6      5      6      60      1      145      6      2      12      0      0      0
## # ... with 15 more variables: DoItYBks <dbl>, RefBks <dbl>, ArtBks <dbl>,
## #   GeogBks <dbl>, ItalCook <dbl>, ItalAtlas <dbl>, ItalArt <dbl>,
## #   Florence <dbl>, `Related Purchase` <dbl>, Yes_Florence <dbl>,
## #   No_Florence <dbl>, Name <chr>, Phone_No. <chr>, Address <chr>, Job <chr>,
## #   and abbreviated variable names 1: ChildBks, 2: YouthBks
## # i Use `colnames()` to see all variable names
```

```
summary(cbc_df[c("M","F","R","FirstPurch","ChildBks","YouthBks","CookBks","DoItYBks","RefBks","ArtBks",
```

```
##           M           F           R       FirstPurch
## Min.      : 15.0    Min.      : 1.000    Min.      : 2.00    Min.      : 2.00
## 1st Qu.:130.0    1st Qu.: 1.000    1st Qu.: 8.00    1st Qu.:12.00
## Median :208.0    Median : 2.000    Median :12.00    Median :20.00
## Mean      :208.2    Mean      : 3.831    Mean      :13.43    Mean      :26.51
## 3rd Qu.:283.0    3rd Qu.: 6.000    3rd Qu.:16.00    3rd Qu.:36.00
## Max.      :479.0    Max.      :12.000    Max.      :36.00    Max.      :99.00
## NA's      :93      NA's      :218      NA's      :342
##      ChildBks      YouthBks      CookBks      DoItYBks
## Min.      :0.0000    Min.      :0.0000    Min.      :0.0000    Min.      :0.0000
## 1st Qu.:0.0000    1st Qu.:0.0000    1st Qu.:0.0000    1st Qu.:0.0000
## Median :0.0000    Median :0.0000    Median :0.0000    Median :0.0000
## Mean      :0.6398    Mean      :0.3048    Mean      :0.7312    Mean      :0.3508
## 3rd Qu.:1.0000    3rd Qu.:0.0000    3rd Qu.:1.0000    3rd Qu.:1.0000
## Max.      :7.0000    Max.      :5.0000    Max.      :7.0000    Max.      :5.0000
##
##      RefBks      ArtBks      GeogBks      ItalCook
## Min.      :0.0000    Min.      :0.000    Min.      :0.0000    Min.      :0.0000
## 1st Qu.:0.0000    1st Qu.:0.000    1st Qu.:0.0000    1st Qu.:0.0000
## Median :0.0000    Median :0.000    Median :0.0000    Median :0.0000
## Mean      :0.2562    Mean      :0.289    Mean      :0.3875    Mean      :0.1252
## 3rd Qu.:0.0000    3rd Qu.:0.000    3rd Qu.:1.0000    3rd Qu.:0.0000
## Max.      :4.0000    Max.      :5.000    Max.      :6.0000    Max.      :3.0000
##
##      ItalAtlas      ItalArt      Florence      Related Purchase
## Min.      :0.0000    Min.      :0.00000    Min.      :0.0000    Min.      :0.000
## 1st Qu.:0.0000    1st Qu.:0.00000    1st Qu.:0.0000    1st Qu.:0.000
## Median :0.0000    Median :0.00000    Median :0.0000    Median :0.000
## Mean      :0.0375    Mean      :0.04575    Mean      :0.0845    Mean      :0.885
## 3rd Qu.:0.0000    3rd Qu.:0.00000    3rd Qu.:0.0000    3rd Qu.:1.000
## Max.      :2.0000    Max.      :2.00000    Max.      :1.0000    Max.      :8.000
##
##      Yes_Florence      No_Florence
## Min.      :0.0000    Min.      :0.0000
## 1st Qu.:0.0000    1st Qu.:1.0000
## Median :0.0000    Median :1.0000
## Mean      :0.0845    Mean      :0.9155
## 3rd Qu.:0.0000    3rd Qu.:1.0000
## Max.      :1.0000    Max.      :1.0000
##
```

```
sprintf("The standard Deviation of M data : %f",sd(cbc_df$M,na.rm = T))
```

```
## [1] "The standard Deviation of M data : 100.926278"
```

```

sprintf("The standard Deviation of F data : %f",sd(cbc_df$F,na.rm = T))

## [1] "The standard Deviation of F data : 3.455656"
sprintf("The standard Deviation of R data : %f",sd(cbc_df$R,na.rm = T))

## [1] "The standard Deviation of R data : 8.090016"
sprintf("The standard Deviation of FirstPurch data : %f",sd(cbc_df$FirstPurch))

## [1] "The standard Deviation of FirstPurch data : 18.351380"
sprintf("The standard Deviation of ChildBks data : %f",sd(cbc_df$ChildBks))

## [1] "The standard Deviation of ChildBks data : 0.994343"
sprintf("The standard Deviation of YouthBks data : %f",sd(cbc_df$YouthBks))

## [1] "The standard Deviation of YouthBks data : 0.611940"
sprintf("The standard Deviation of CookBks data : %f",sd(cbc_df$CookBks))

## [1] "The standard Deviation of CookBks data : 1.089413"
sprintf("The standard Deviation of DoItYBks data : %f",sd(cbc_df$DoItYBks))

## [1] "The standard Deviation of DoItYBks data : 0.687999"
sprintf("The standard Deviation of RefBks data : %f",sd(cbc_df$RefBks))

## [1] "The standard Deviation of RefBks data : 0.558269"
sprintf("The standard Deviation of ArtBks data : %f",sd(cbc_df$ArtBks))

## [1] "The standard Deviation of ArtBks data : 0.600890"
sprintf("The standard Deviation of GeogBks data : %f",sd(cbc_df$GeogBks))

## [1] "The standard Deviation of GeogBks data : 0.750656"
sprintf("The standard Deviation of ItalCook data : %f",sd(cbc_df$ItalCook))

## [1] "The standard Deviation of ItalCook data : 0.385486"
sprintf("The standard Deviation of ItalAtlas data : %f",sd(cbc_df$ItalAtlas))

## [1] "The standard Deviation of ItalAtlas data : 0.214721"
sprintf("The standard Deviation of ItalArt data : %f",sd(cbc_df$ItalArt))

## [1] "The standard Deviation of ItalArt data : 0.220611"
sprintf("The standard Deviation of Florence data : %f",sd(cbc_df$Florence))

## [1] "The standard Deviation of Florence data : 0.278171"
sprintf("The standard Deviation of Related Purchase data : %f",sd(cbc_df$`Related Purchase`))

## [1] "The standard Deviation of Related Purchase data : 1.226234"
sprintf("The standard Deviation of Yes_Florence data : %f",sd(cbc_df$Yes_Florence))

## [1] "The standard Deviation of Yes_Florence data : 0.278171"

```

```
sprintf("The standard Deviation of NO_Florence data : %f",sd(cbc_df$No_Florence))
```

```
## [1] "The standard Deviation of NO_Florence data : 0.278171"
```

```
print("Count per each column : ")
```

```
## [1] "Count per each column : "
```

```
print(colSums(!is.na(cbc_df)))
```

```
##          ...1          Seq#          ID#          Gender
##          4000          4000          4000          4000
##           M           R           F       FirstPurch
##          3907          3658          3782          4000
##       ChildBks       YouthBks       CookBks       DoItYBks
##          4000          4000          4000          4000
##          RefBks          ArtBks       GeogBks       ItalCook
##          4000          4000          4000          4000
##       ItalAtlas       ItalArt       Florence Related Purchase
##          4000          4000          4000          4000
##       Yes_Florence       No_Florence       Name       Phone_No.
##          4000          4000          4000          4000
##          Address          Job
##          4000          4000
```

```
print("Number of missing Values per column :")
```

```
## [1] "Number of missing Values per column :"
```

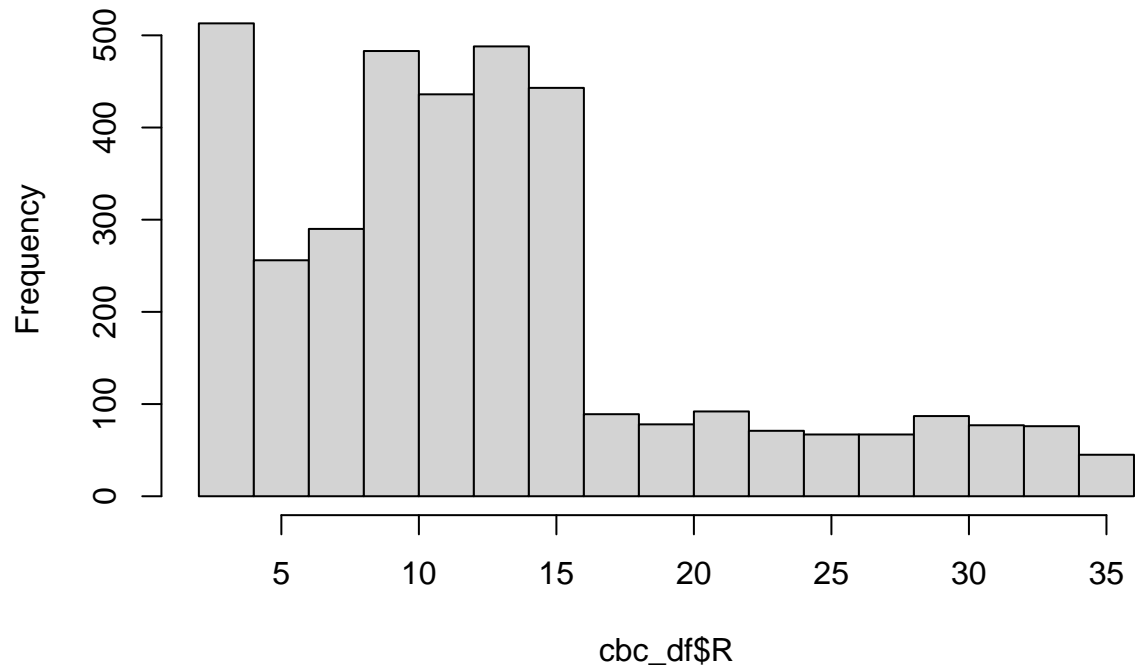
```
colSums(is.na(cbc_df))
```

```
##          ...1          Seq#          ID#          Gender
##           0           0           0           0
##           M           R           F       FirstPurch
##           93          342          218           0
##       ChildBks       YouthBks       CookBks       DoItYBks
##           0           0           0           0
##          RefBks          ArtBks       GeogBks       ItalCook
##           0           0           0           0
##       ItalAtlas       ItalArt       Florence Related Purchase
##           0           0           0           0
##       Yes_Florence       No_Florence       Name       Phone_No.
##           0           0           0           0
##          Address          Job
##           0           0
```

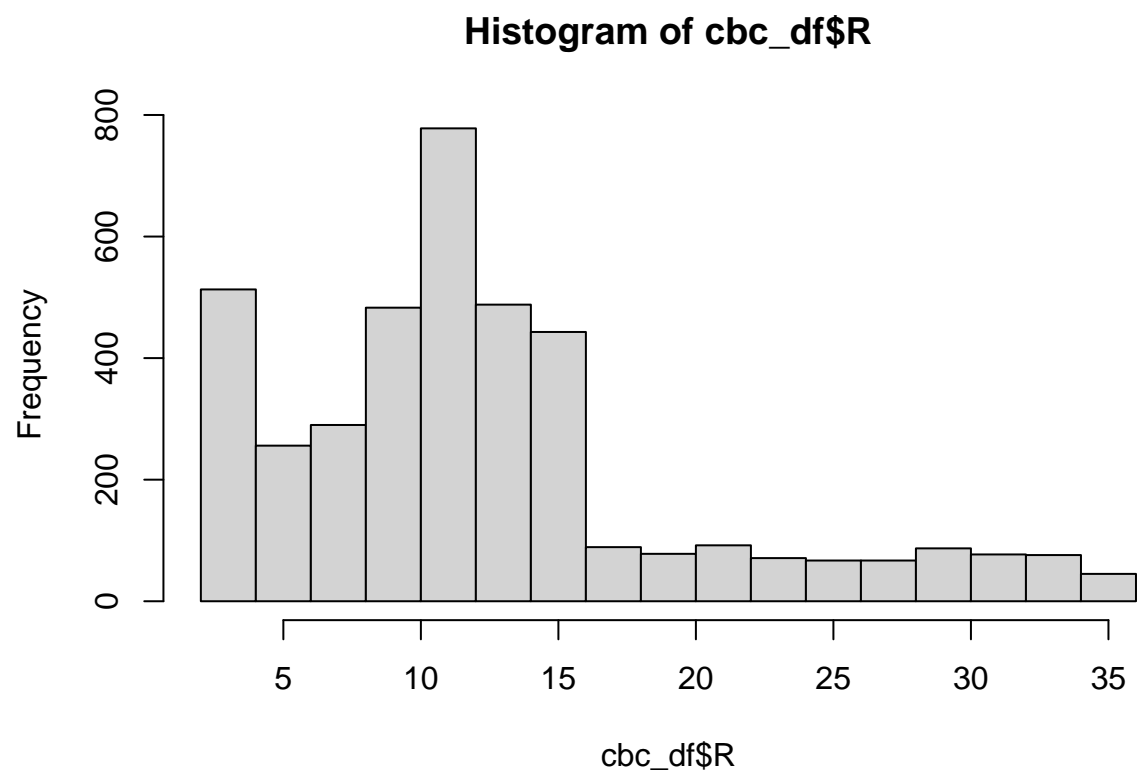
Problem 2 (2 points)

```
hist(cbc_df$R)
```

Histogram of cbc_df\$R

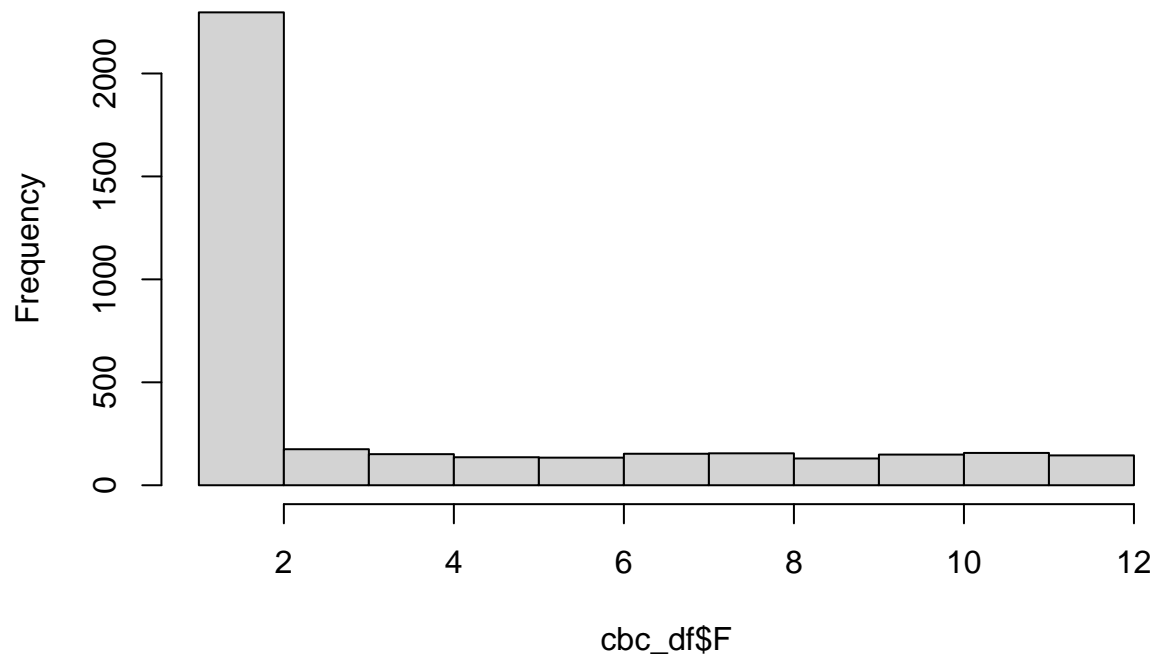


```
cbc_df$R[is.na(cbc_df$R)] <- median(cbc_df$R, na.rm = T)
hist(cbc_df$R)
```



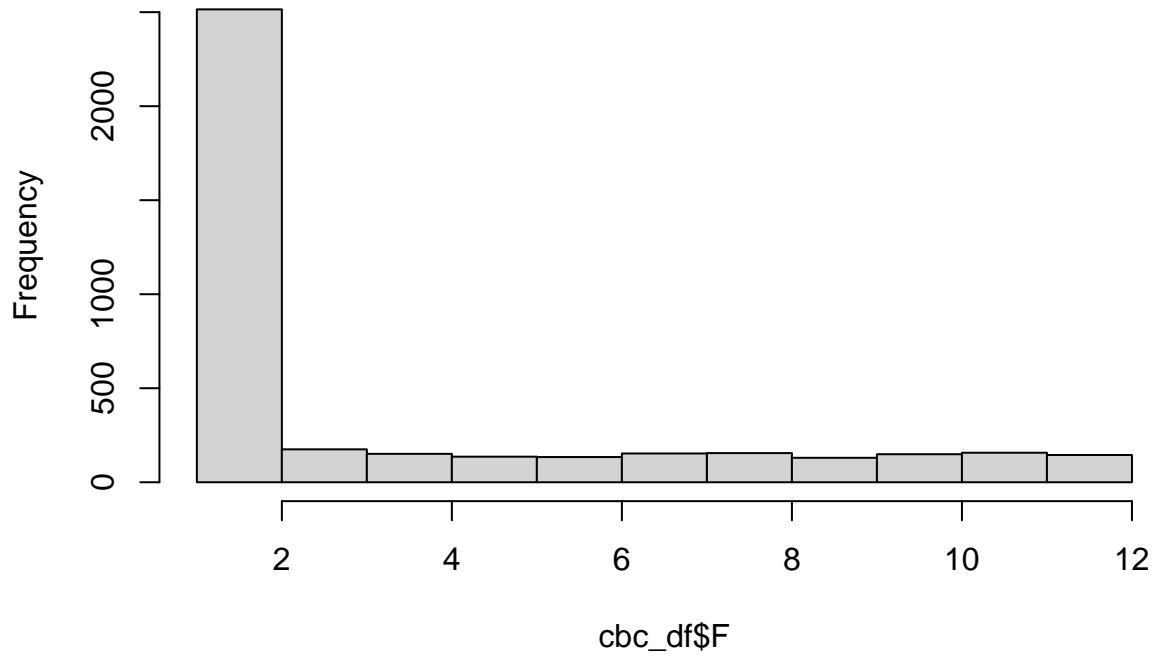
```
hist(cbc_df$F)
```

Histogram of cbc_df\$F



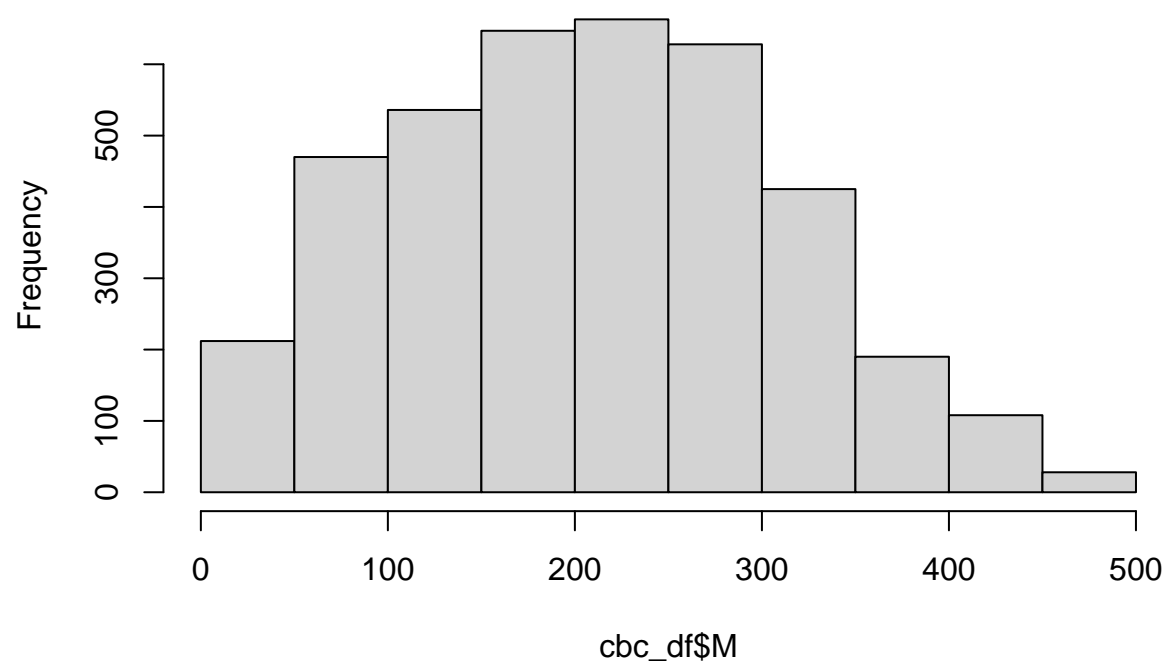
```
cbc_df$F[is.na(cbc_df$F)] <- median(cbc_df$F, na.rm = T)
hist(cbc_df$F)
```

Histogram of cbc_df\$F

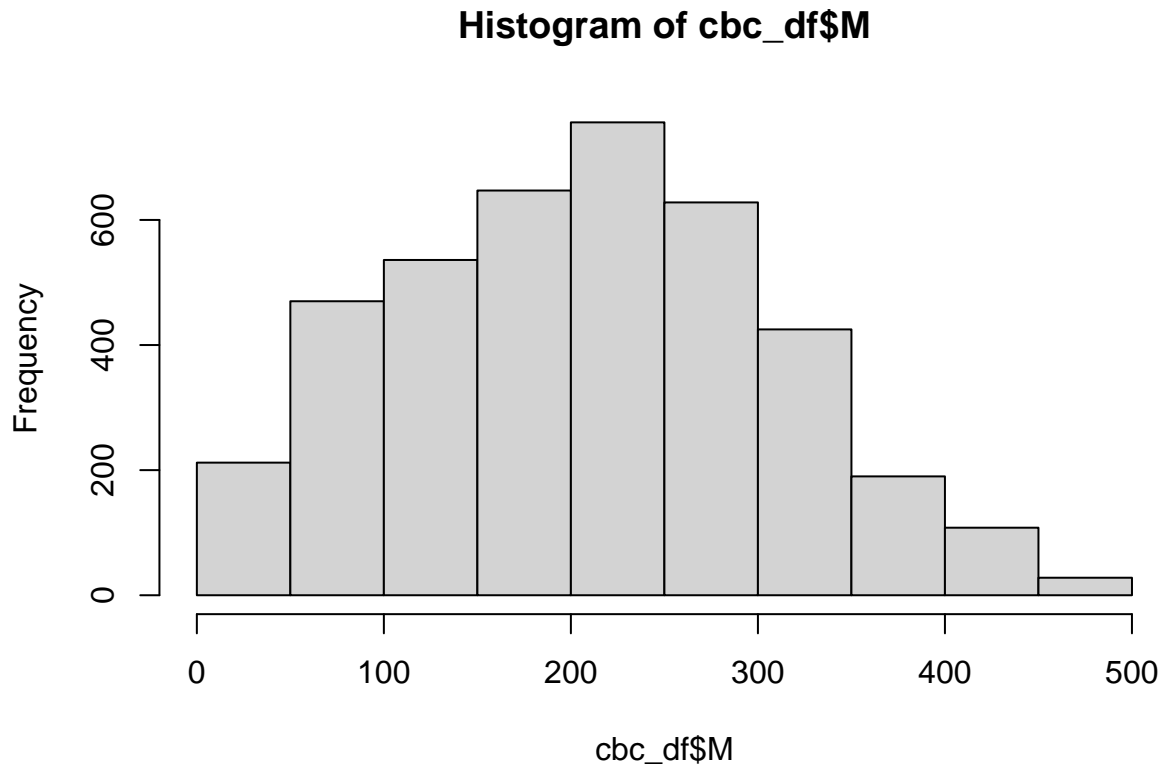


```
hist(cbc_df$M)
```


Histogram of cbc_df\$M



```
cbc_df$M[is.na(cbc_df$M)] <- mean(cbc_df$M, na.rm = T)
hist(cbc_df$M)
```



```
print("So we can observe that there are no missing values")
```

```
## [1] "So we can observe that there are no missing values"
```

```
colSums(is.na(cbc_df))
```

```
##           ...1           Seq#           ID#           Gender
##           0           0           0           0
##           M           R           F           FirstPurch
##           0           0           0           0
##           ChildBks       YouthBks       CookBks       DoItYBks
##           0           0           0           0
##           RefBks        ArtBks        GeogBks        ItalCook
##           0           0           0           0
##           ItalAtlas      ItalArt      Florence Related Purchase
##           0           0           0           0
##           Yes_Florence   No_Florence   Name           Phone_No.
##           0           0           0           0
##           Address        Job
##           0           0
```

Analysis:

Since the Histogram of R dataset is Right Skewed, So we can replace the missing values with the **median** of the R data

Similarly the Histogram of F dataset is Right Skewed, So we can replace the missing values with the **median** of the F data

Finally the Histogram of the M dataset is normally distributed, So we can replace the missing Values with **mean** of the M data

Problem 3 (2 points)

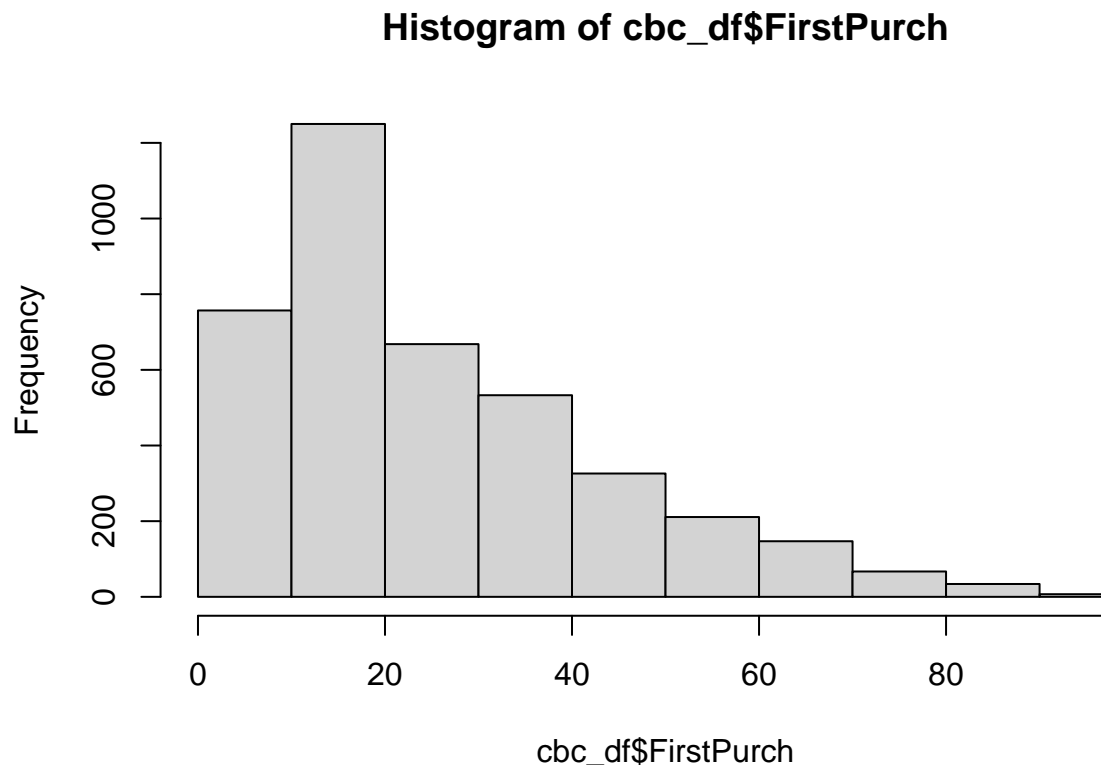
```
cbc_df$Mcode <- cut(cbc_df$M, breaks = 5)
cbc_df$Rcode <- cut(cbc_df$R, breaks = 5)
cbc_df$Fcode <- cut(cbc_df$F, breaks = 5)

summary(cbc_df[c("Mcode", "Rcode", "Fcode")])
```

##	Mcode	Rcode	Fcode
##	(14.5,108] : 753	(1.97,8.8] :1059	(0.989,3.2] :2690
##	(108,201] :1112	(8.8,15.6] :1749	(3.2,5.4] : 287
##	(201,293] :1295	(15.6,22.4] : 702	(5.4,7.6] : 287
##	(293,386] : 660	(22.4,29.2] : 205	(7.6,9.8] : 285
##	(386,479] : 180	(29.2,36] : 285	(9.8,12] : 451

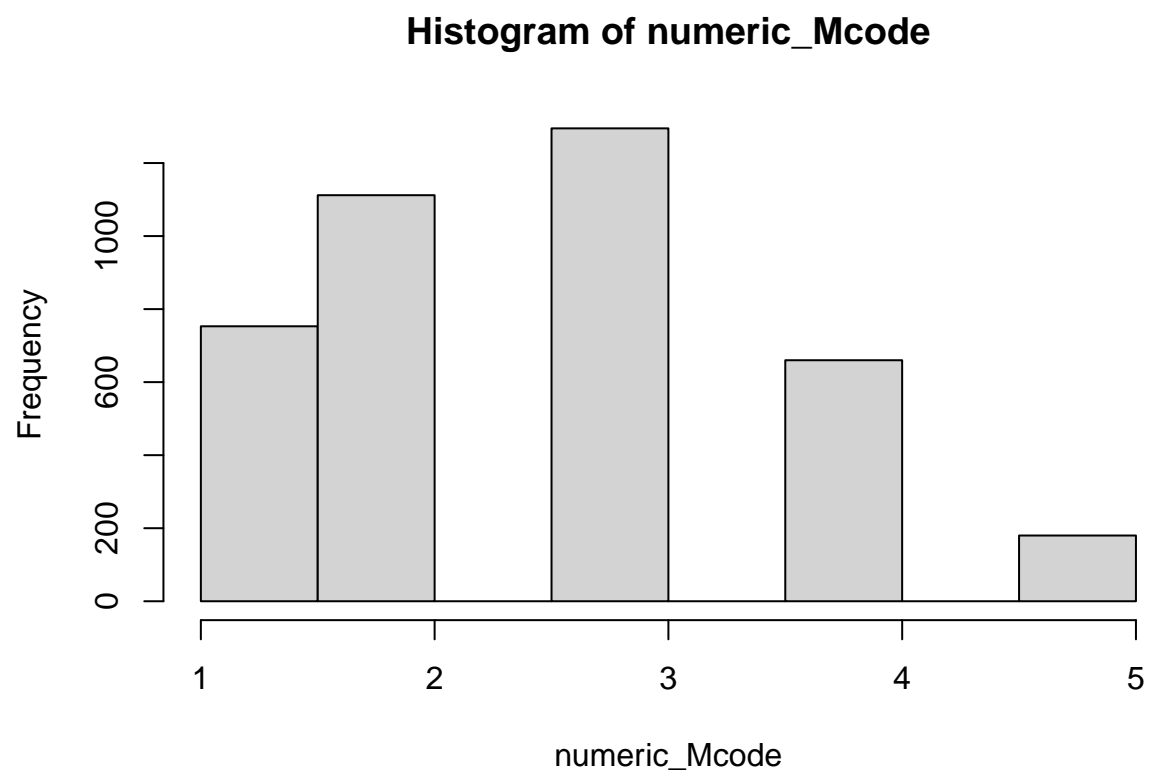
Problem 4

```
hist(cbc_df$FirstPurch)
```



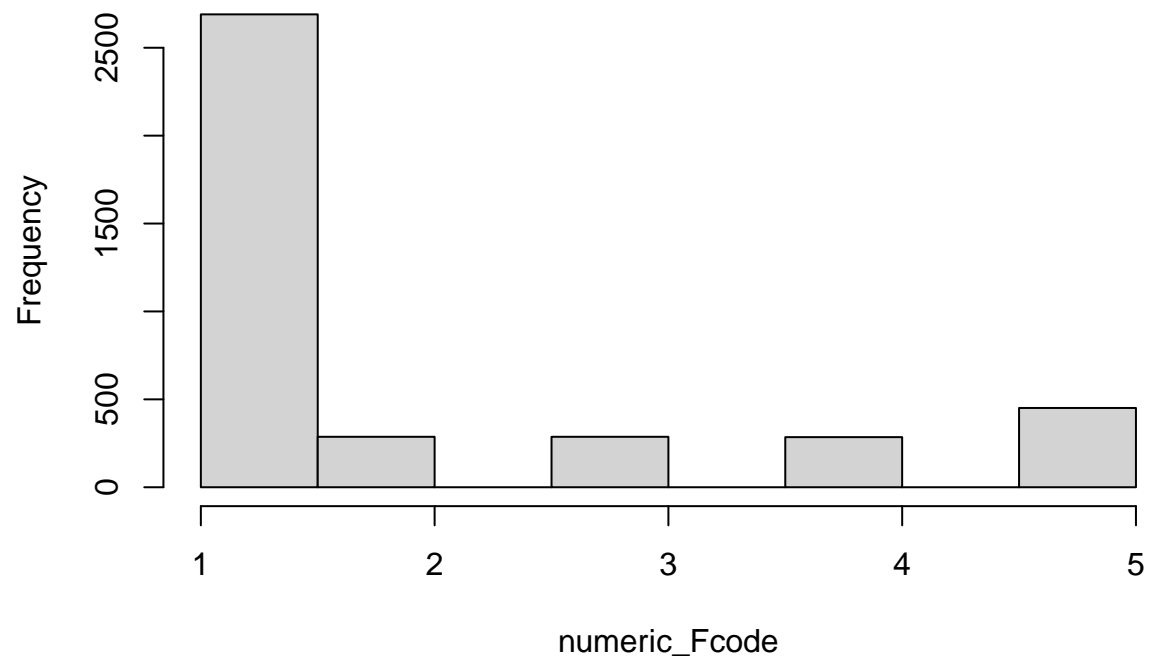
4.1 Bar Graphs (1 point)

```
numeric_Mcode=as.numeric(cbc_df$Mcode)
numeric_Rcode=as.numeric(cbc_df$Rcode)
numeric_Fcode=as.numeric(cbc_df$Fcode)
hist(numeric_Mcode)
```

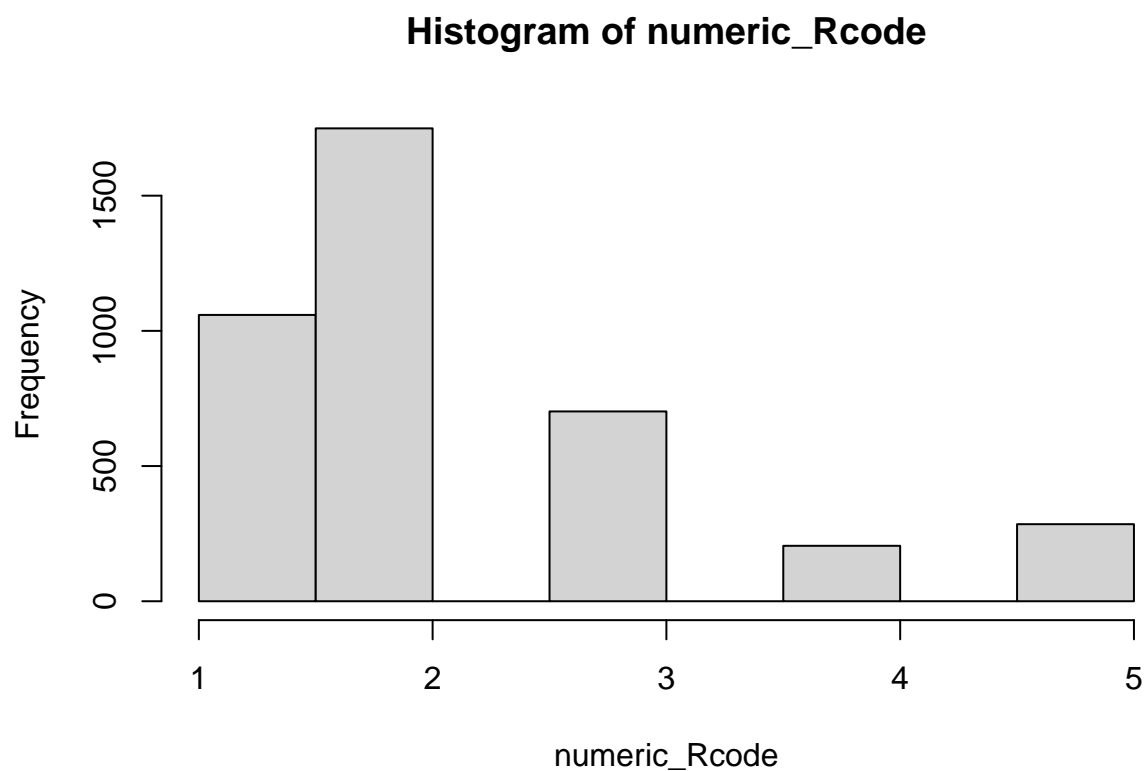


```
hist(numeric_Fcode)
```

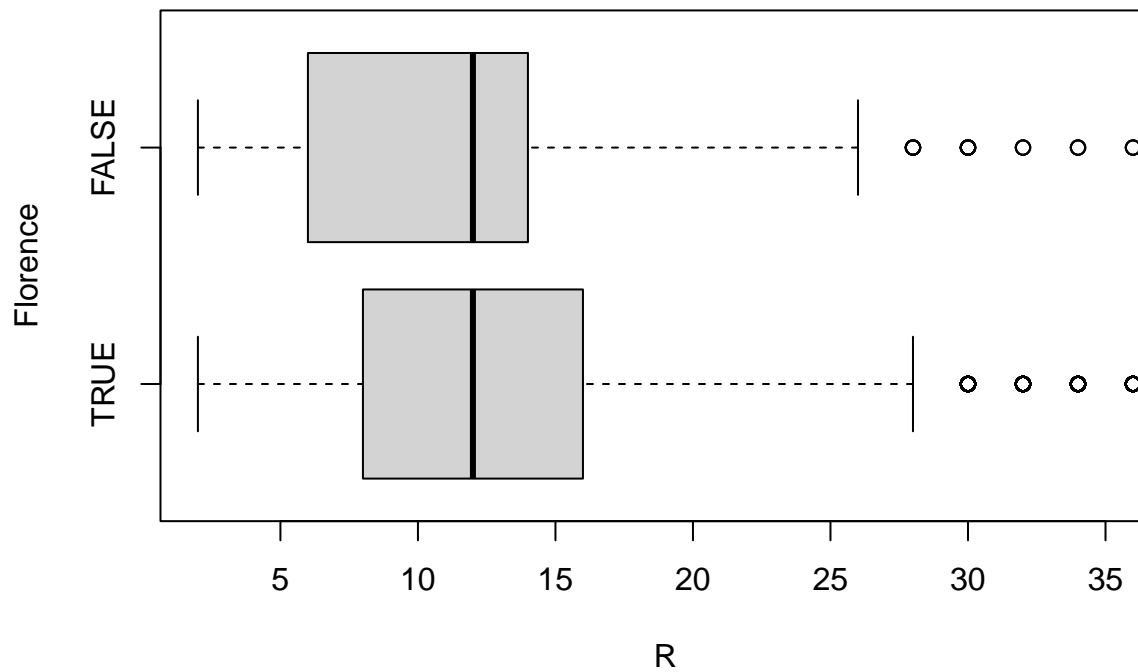
Histogram of numeric_Fcode



```
hist(numeric_Rcode)
```

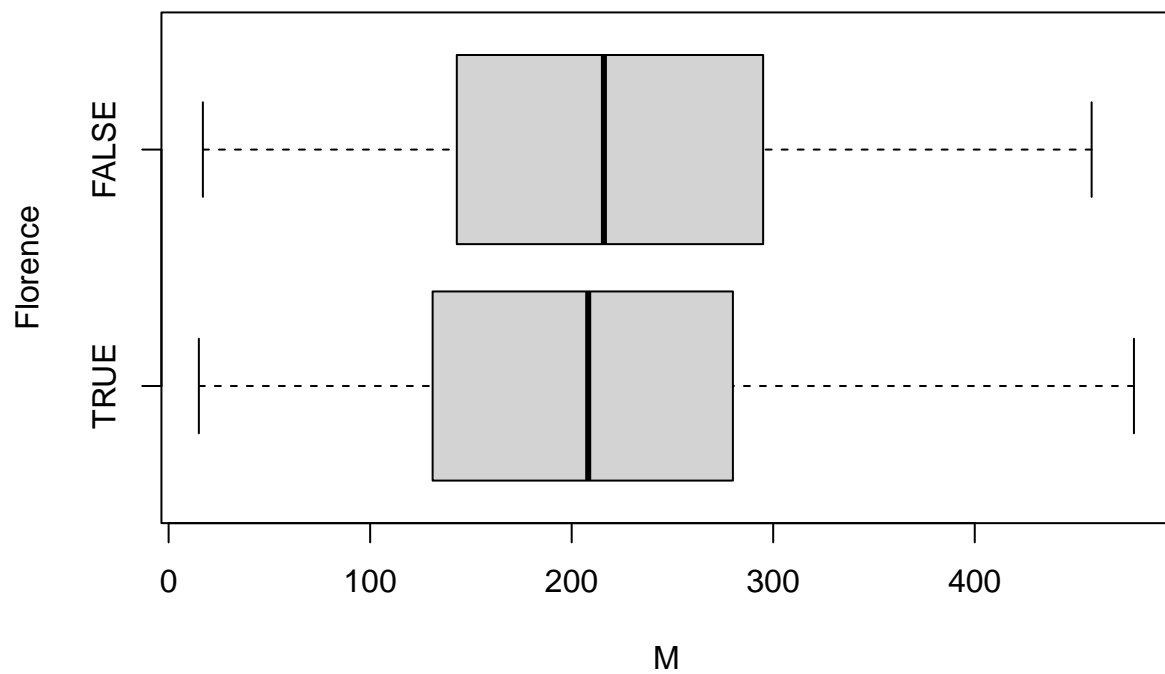


```
cbc_df$Florence=factor(cbc_df$Florence, labels=c("TRUE","FALSE"))
F_R_T=cbc_df$R[which(cbc_df$Florence=="TRUE")]
F_R_F=cbc_df$R[which(cbc_df$Florence=="FALSE")]
boxplot(F_R_T,F_R_F,horizontal = TRUE,xlab="R",ylab="Florence",names = c("TRUE","FALSE"))
```

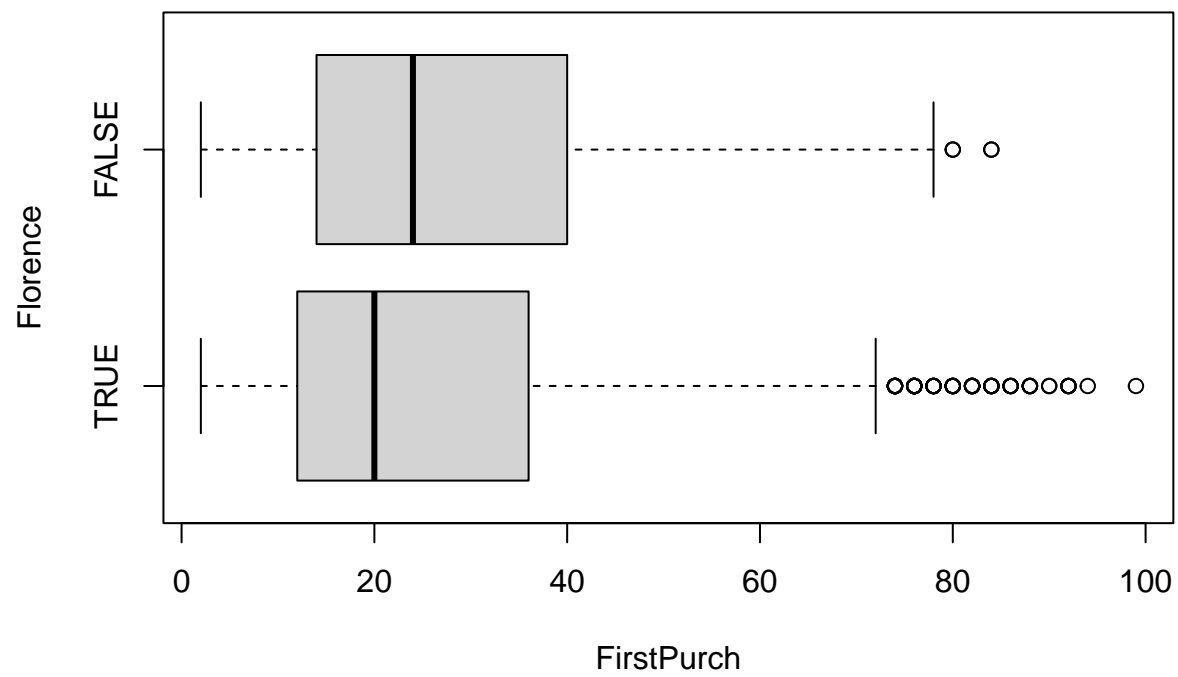


4.2 Box Plot (1 point)

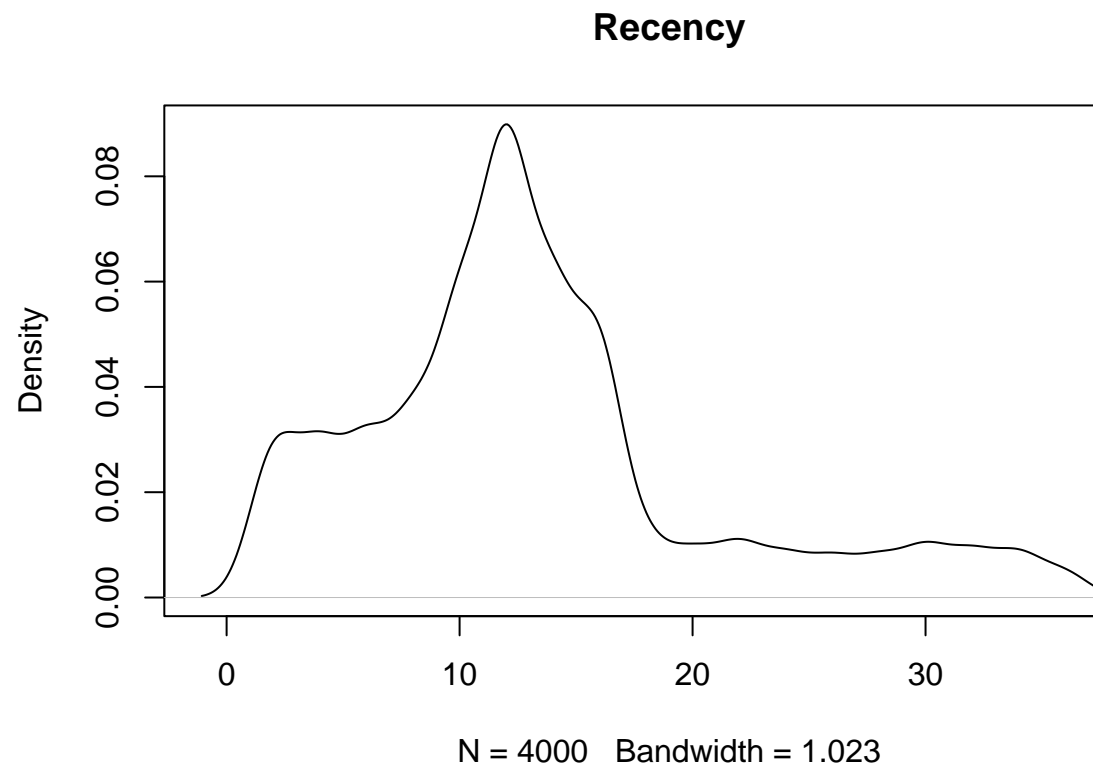
```
F_M_T=cbc_df$M[which(cbc_df$Florence=="TRUE")]
F_M_F=cbc_df$M[which(cbc_df$Florence=="FALSE")]
boxplot(F_M_T,F_M_F,horizontal = TRUE,xlab="M",ylab="Florence",names = c("TRUE","FALSE"))
```



```
F_First_T=cbc_df$FirstPurch[which(cbc_df$Florence=="TRUE")]
F_First_F=cbc_df$FirstPurch[which(cbc_df$Florence=="FALSE")]
boxplot(F_First_T,F_First_F,horizontal = TRUE,xlab="FirstPurch",ylab="Florence",names = c("TRUE", "FALSE"))
```

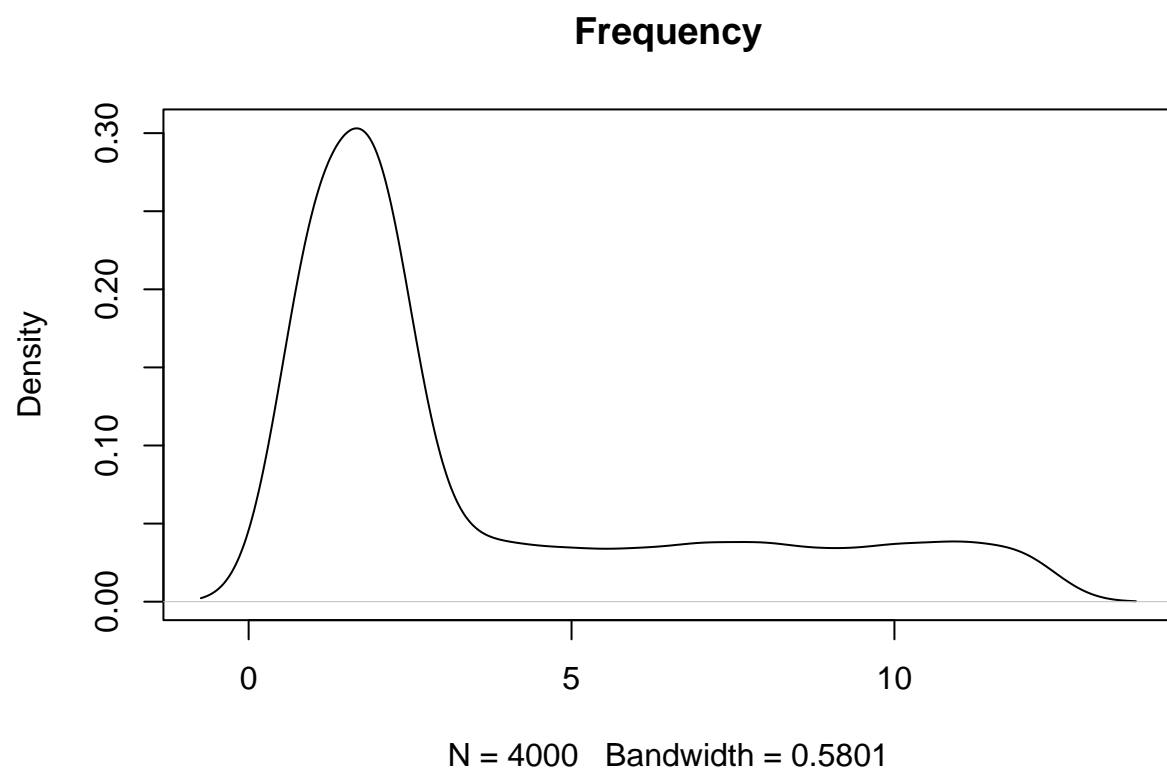



```
plot(density(cbc_df$R),main="Recency")
```

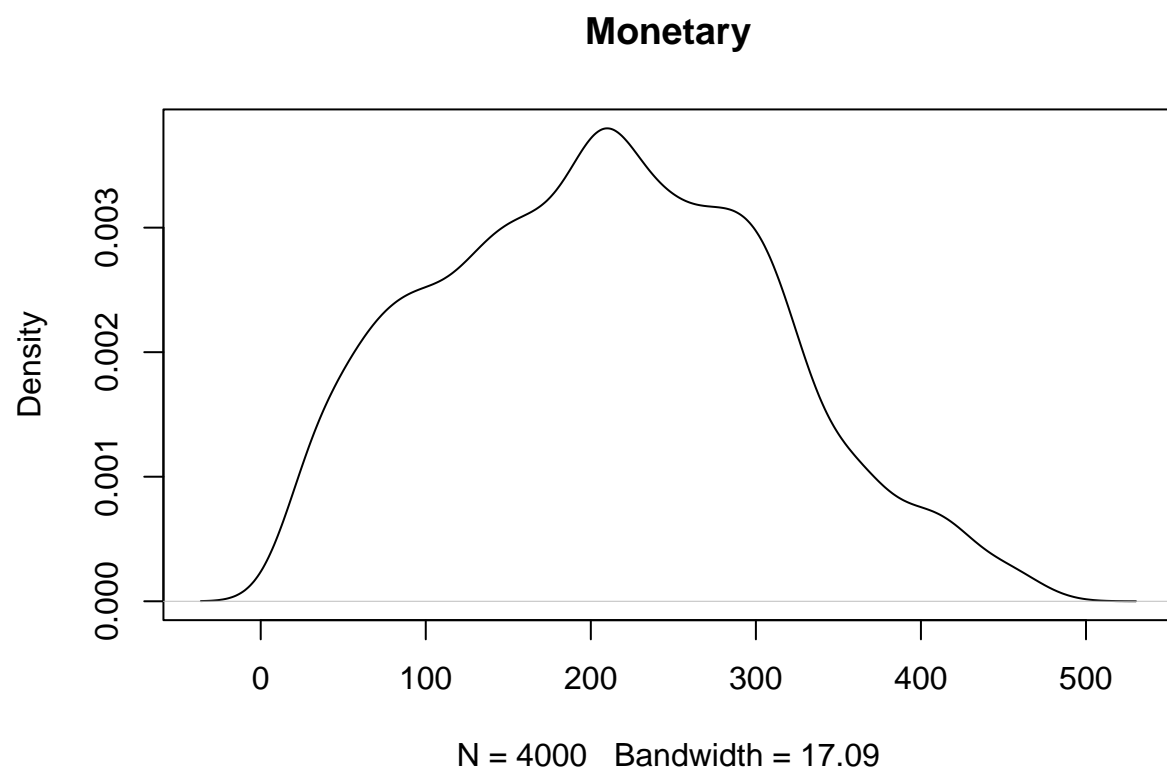


4.3 Density Plot (1 point)

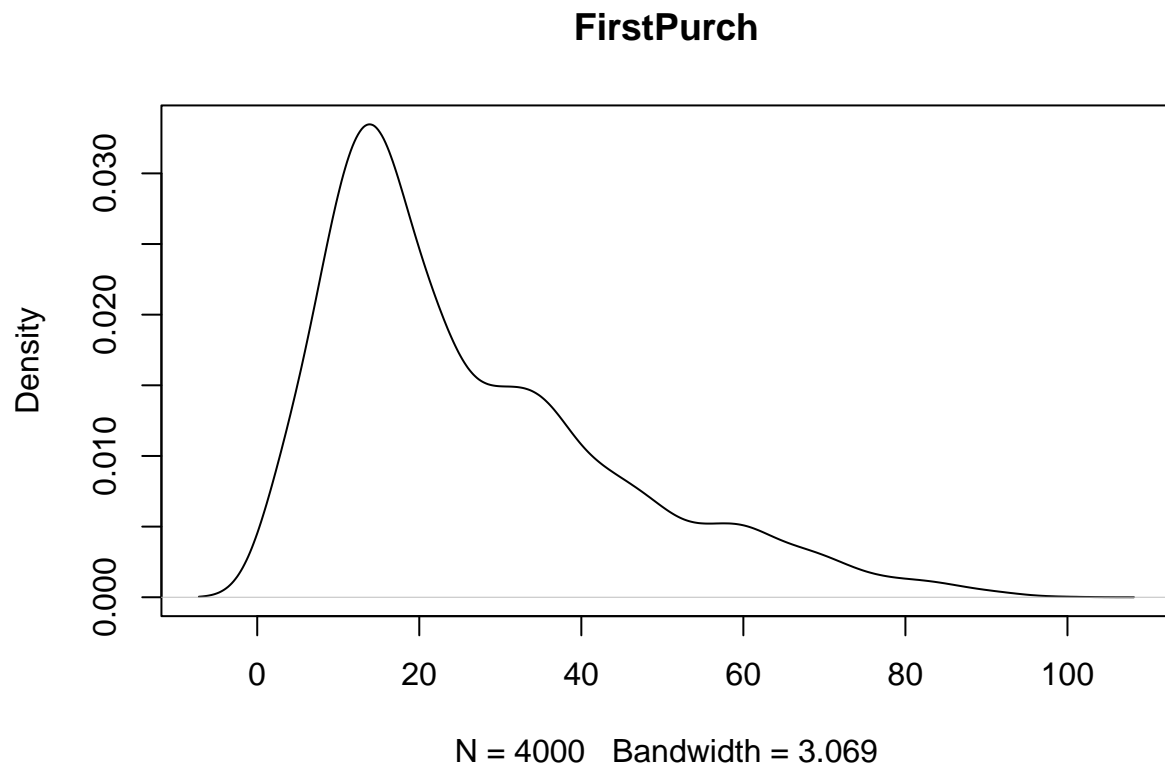
```
plot(density(cbc_df$F),main="Frequency")
```



```
plot(density(cbc_df$M),main="Monetary")
```



```
plot(density(cbc_df$FirstPurch),main="FirstPurch")
```



Part II. ANOVA

Problems

Problem 1 (2 points)

```
library(ggpubr)
library(dplyr)
library(ggplot2)
library(ggpubr)
library(broom)
library(car)

## Loading required package: carData
##
## Attaching package: 'car'
## The following object is masked from 'package:dplyr':
##
##   recode
## The following object is masked from 'package:purrr':
##
##   some
```

```
data <- read.csv('C:\\Users\\HP\\Desktop\\PESU\\SEM - 5\\PES1UG20CS445\\DA\\1_b\\Scenario 1.csv')
anova_1 <- aov(No.of.items ~ POI, data = data)
summary(anova_1)
```

```
##              Df Sum Sq Mean Sq F value Pr(>F)
## POI           4     127    31.75   1.025  0.393
## Residuals    995    30827    30.98
```

Analysis:

- 1.. Scully can use one-way ANOVA Test, because we are analyzing a single independent variable.
- 2.. aov() is a function used by Scully for ANOVA in R programming
- 3..NULL HYPOTHESIS : It states that the means are the same.

ALTERNATE HYPOTHESIS : It states that the means of every group is different.

Output: Since the P-Value is greater than alpha(5%).We accept the NULL Hypothesis, which means that there is no difference between the means .

Problem 2 (3 points)

```
data=read.csv('C:\\Users\\HP\\Desktop\\PESU\\SEM - 5\\PES1UG20CS445\\DA\\1_b\\Scenario 2.csv')
anova_2=aov(No.of.items ~ POI*Priority,data=data)
summary(anova_2)
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## POI           4      317    79.29    2.880  0.0218 *
## Priority       4      690   172.53    6.268 5.57e-05 ***
## POI:Priority   16      347    21.66    0.787  0.7019
## Residuals     975   26839    27.53
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Analysis:

- 1.. We choose two-way ANOVA because there are two independent variables
- 2.. aov() function is used in R programming to find the AVOVA
- 3.. NULL HYPOTHESIS: the means are the same
- ALTERNATE HYPOTHESIS : the means are different
- output : Since the P-value is very less then the alpha(5%) value, we reject the NULL Hypothesis.
- 4..Global mean is not considered

Problem 3 (1 point)

```
HSD=TukeyHSD(anova_2)
HSD
```

```
##      Tukey multiple comparisons of means
##      95% family-wise confidence level
##
## Fit: aov(formula = No.of.items ~ POI * Priority, data = data)
##
## $POI
##              diff              lwr              upr              p adj
## Fredo-Apollonia -1.16541489 -3.527882256 1.1970525 0.6610398
## Micheal-Apollonia 0.97971301 -2.046113871 4.0055399 0.9025238
```

```

## Sonny-Apollonia -0.07537018 -2.350096344 2.1993560 0.9999848
## Vito-Apollonia -0.74221825 -3.264671502 1.7802350 0.9293713
## Micheal-Fredo 2.14512791 -0.129588606 4.4198444 0.0753886
## Sonny-Fredo 1.09004471 -0.002240047 2.1823295 0.0507698
## Vito-Fredo 0.42319665 -1.120006939 1.9664002 0.9446478
## Sonny-Micheal -1.05508319 -3.238535117 1.1283687 0.6784431
## Vito-Micheal -1.72193126 -4.162393022 0.7185305 0.3029361
## Vito-Sonny -0.66684807 -2.072054240 0.7383581 0.6932090
##
## $Priority
## diff lwr upr p adj
## high-critical -3.4375139 -5.6024538 -1.2725741 0.0001534
## low-critical -1.9212087 -3.2006497 -0.6417676 0.0004230
## med-critical -1.5518320 -2.9759071 -0.1277569 0.0247330
## None-critical -1.7809723 -4.1230935 0.5611489 0.2303819
## low-high 1.5163053 -0.4634213 3.4960319 0.2237743
## med-high 1.8856819 -0.1904534 3.9618172 0.0954896
## None-high 1.6565416 -1.1305500 4.4436332 0.4820853
## med-low 0.3693766 -0.7532733 1.4920265 0.8972243
## None-low 0.1402363 -2.0318328 2.3123055 0.9997831
## None-med -0.2291403 -2.4894290 2.0311484 0.9987096
##
## $`POI:Priority`
## diff lwr upr
## Fredo:critical-Apollonia:critical -1.8571428571 -10.257712 6.5434260
## Micheal:critical-Apollonia:critical -1.0333333333 -12.688650 10.6219830
## Sonny:critical-Apollonia:critical -0.9251700680 -9.020161 7.1698207
## Vito:critical-Apollonia:critical -1.1333333333 -10.092846 7.8261796
## Apollonia:high-Apollonia:critical -8.8333333333 -29.623681 11.9570139
## Fredo:high-Apollonia:critical -4.9583333333 -13.743859 3.8271919
## Micheal:high-Apollonia:critical -4.8333333333 -20.549359 10.8826920
## Sonny:high-Apollonia:critical -4.4487179487 -13.166400 4.2689644
## Vito:high-Apollonia:critical -4.5000000000 -15.612908 6.6129081
## Apollonia:low-Apollonia:critical -3.8333333333 -12.792846 5.1261796
## Fredo:low-Apollonia:critical -3.7655367232 -11.820852 4.2897789
## Micheal:low-Apollonia:critical -0.9242424242 -9.789276 7.9407914
## Sonny:low-Apollonia:critical -2.7088353414 -10.660959 5.2432886
## Vito:low-Apollonia:critical -3.6770833333 -11.895190 4.5410230
## Apollonia:med-Apollonia:critical 0.0952380952 -9.296883 9.4873596
## Fredo:med-Apollonia:critical -3.6236559140 -11.853116 4.6058040
## Micheal:med-Apollonia:critical -0.8333333333 -10.131059 8.4643926
## Sonny:med-Apollonia:critical -2.5613333333 -10.605728 5.4830613
## Vito:med-Apollonia:critical -3.3787878788 -11.921348 5.1637725
## Apollonia:None-Apollonia:critical -8.3333333333 -24.049359 7.3826920
## Fredo:None-Apollonia:critical -3.2878787879 -13.056670 6.4809127
## Micheal:None-Apollonia:critical -5.5000000000 -19.110477 8.1104772
## Sonny:None-Apollonia:critical -2.5933333333 -11.343646 6.1569792
## Vito:None-Apollonia:critical -0.6904761905 -11.399142 10.0181900
## Micheal:critical-Fredo:critical 0.8238095238 -8.282190 9.9298092
## Sonny:critical-Fredo:critical 0.9319727891 -2.617915 4.4818611
## Vito:critical-Fredo:critical 0.7238095238 -4.505502 5.9531215
## Apollonia:high-Fredo:critical -6.9761904762 -26.452108 12.4997273
## Fredo:high-Fredo:critical -3.1011904762 -8.026461 1.8240796
## Micheal:high-Fredo:critical -2.9761904762 -16.906958 10.9545770

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## Sonny:high-Fredo:critical	-2.5915750916	-7.394784	2.2116339
## Vito:high-Fredo:critical	-2.6428571429	-11.043426	5.7577117
## Apollonia:low-Fredo:critical	-1.9761904762	-7.205502	3.2531215
## Fredo:low-Fredo:critical	-1.9083938660	-5.366853	1.5500656
## Micheal:low-Fredo:critical	0.9329004329	-4.132833	5.9986341
## Sonny:low-Fredo:critical	-0.8516924842	-4.062472	2.3590873
## Vito:low-Fredo:critical	-1.8199404762	-5.642257	2.0023761
## Apollonia:med-Fredo:critical	1.9523809524	-3.987718	7.8924802
## Fredo:med-Fredo:critical	-1.7665130568	-5.613179	2.0801534
## Micheal:med-Fredo:critical	1.0238095238	-4.765883	6.8135021
## Sonny:med-Fredo:critical	-0.7041904762	-4.137136	2.7287552
## Vito:med-Fredo:critical	-1.5216450216	-5.999163	2.9558732
## Apollonia:None-Fredo:critical	-6.4761904762	-20.406958	7.4545770
## Fredo:None-Fredo:critical	-1.4307359307	-7.950101	5.0886291
## Micheal:None-Fredo:critical	-3.6428571429	-15.145810	7.8600955
## Sonny:None-Fredo:critical	-0.7361904762	-5.598371	4.1259903
## Vito:None-Fredo:critical	1.1666666667	-6.691346	9.0246793
## Sonny:critical-Micheal:critical	0.1081632653	-8.716719	8.9330456
## Vito:critical-Micheal:critical	-0.1000000000	-9.724061	9.5240607
## Apollonia:high-Micheal:critical	-7.8000000000	-28.885261	13.2852605
## Fredo:high-Micheal:critical	-3.9250000000	-13.387300	5.5373003
## Micheal:high-Micheal:critical	-3.8000000000	-19.904134	12.3041338
## Sonny:high-Micheal:critical	-3.4153846154	-12.814728	5.9839589
## Vito:high-Micheal:critical	-3.4666666667	-15.121983	8.1886496
## Apollonia:low-Micheal:critical	-2.8000000000	-12.424061	6.8240607
## Fredo:low-Micheal:critical	-2.7322033898	-11.520706	6.0562995
## Micheal:low-Micheal:critical	0.1090909091	-9.427077	9.6452588
## Sonny:low-Micheal:critical	-1.6755020080	-10.369520	7.0185159
## Vito:low-Micheal:critical	-2.6437500000	-11.581700	6.2941997
## Apollonia:med-Micheal:critical	1.1285714286	-8.899470	11.1566131
## Fredo:med-Micheal:critical	-2.5903225806	-11.538713	6.3580674
## Micheal:med-Micheal:critical	0.2000000000	-9.739687	10.1396871
## Sonny:med-Micheal:critical	-1.5280000000	-10.306494	7.2504940
## Vito:med-Micheal:critical	-2.3454545455	-11.582608	6.8916993
## Apollonia:None-Micheal:critical	-7.3000000000	-23.404134	8.8041338
## Fredo:None-Micheal:critical	-2.2545454545	-12.636210	8.1271192
## Micheal:None-Micheal:critical	-4.4666666667	-18.523507	9.5901737
## Sonny:None-Micheal:critical	-1.5600000000	-10.989615	7.8696152
## Vito:None-Micheal:critical	0.3428571429	-10.927689	11.6134030
## Vito:critical-Sonny:critical	-0.2081632653	-4.930982	4.5146559
## Apollonia:high-Sonny:critical	-7.9081632653	-27.254240	11.4379136
## Fredo:high-Sonny:critical	-4.0331632653	-8.416952	0.3506254
## Micheal:high-Sonny:critical	-3.9081632653	-17.656821	9.8404949
## Sonny:high-Sonny:critical	-3.5235478807	-7.769738	0.7226427
## Vito:high-Sonny:critical	-3.5748299320	-11.669821	4.5201608
## Apollonia:low-Sonny:critical	-2.9081632653	-7.630982	1.8146559
## Fredo:low-Sonny:critical	-2.8403666551	-5.471005	-0.2097281
## Micheal:low-Sonny:critical	0.0009276438	-4.540105	4.5419600
## Sonny:low-Sonny:critical	-1.7836652733	-4.078970	0.5116396
## Vito:low-Sonny:critical	-2.7519132653	-5.845361	0.3415348
## Apollonia:med-Sonny:critical	1.0204081633	-4.479055	6.5198714
## Fredo:med-Sonny:critical	-2.6984858460	-5.821971	0.4249994
## Micheal:med-Sonny:critical	0.0918367347	-5.244816	5.4284890
## Sonny:med-Sonny:critical	-1.6361632653	-4.233168	0.9608414

## Vito:med-Sonny:critical	-2.4536178108	-6.327568	1.4203320
## Apollonia:None-Sonny:critical	-7.4081632653	-21.156821	6.3404949
## Fredo:None-Sonny:critical	-2.3627087199	-8.483284	3.7578662
## Micheal:None-Sonny:critical	-4.5748299320	-15.856551	6.7068915
## Sonny:None-Sonny:critical	-1.6681632653	-5.980949	2.6446223
## Vito:None-Sonny:critical	0.2346938776	-7.295756	7.7651441
## Apollonia:high-Vito:critical	-7.7000000000	-27.423455	12.0234552
## Fredo:high-Vito:critical	-3.8250000000	-9.652658	2.0026582
## Micheal:high-Vito:critical	-3.7000000000	-17.974789	10.5747889
## Sonny:high-Vito:critical	-3.3153846154	-9.040254	2.4094849
## Vito:high-Vito:critical	-3.3666666667	-12.326180	5.5928462
## Apollonia:low-Vito:critical	-2.7000000000	-8.786790	3.3867904
## Fredo:low-Vito:critical	-2.6322033898	-7.286691	2.0222843
## Micheal:low-Vito:critical	0.2090909091	-5.737755	6.1559364
## Sonny:low-Vito:critical	-1.5755020080	-6.049026	2.8980225
## Vito:low-Vito:critical	-2.5437500000	-7.474614	2.3871138
## Apollonia:med-Vito:critical	1.2285714286	-5.478745	7.9358878
## Fredo:med-Vito:critical	-2.4903225806	-7.440086	2.4594407
## Micheal:med-Vito:critical	0.3000000000	-6.274485	6.8744851
## Sonny:med-Vito:critical	-1.4280000000	-6.063561	3.2075615
## Vito:med-Vito:critical	-2.2454545455	-7.699947	3.2090375
## Apollonia:None-Vito:critical	-7.2000000000	-21.474789	7.0747889
## Fredo:None-Vito:critical	-2.1545454545	-9.379878	5.0707873
## Micheal:None-Vito:critical	-4.3666666667	-16.283933	7.5505996
## Sonny:None-Vito:critical	-1.4600000000	-7.234436	4.3144364
## Vito:None-Vito:critical	0.4428571429	-8.010052	8.8957665
## Fredo:high-Apollonia:high	3.8750000000	-15.770032	23.5200316
## Micheal:high-Apollonia:high	4.0000000000	-19.574038	27.5740380
## Sonny:high-Apollonia:high	4.3846153846	-15.230170	23.9994006
## Vito:high-Apollonia:high	4.3333333333	-16.457014	25.1236806
## Apollonia:low-Apollonia:high	5.0000000000	-14.723455	24.7234552
## Fredo:low-Apollonia:high	5.0677966102	-14.261713	24.3973058
## Micheal:low-Apollonia:high	7.9090909091	-11.771627	27.5898084
## Sonny:low-Apollonia:high	6.1244979920	-13.162236	25.4112315
## Vito:low-Apollonia:high	5.1562500000	-14.241664	24.5541645
## Apollonia:med-Apollonia:high	8.9285714286	-10.995127	28.8522699
## Fredo:med-Apollonia:high	5.2096774194	-14.193050	24.6124047
## Micheal:med-Apollonia:high	8.0000000000	-11.879374	27.8793743
## Sonny:med-Apollonia:high	6.2720000000	-13.052961	25.5969605
## Vito:med-Apollonia:high	5.4545454545	-14.083038	24.9921285
## Apollonia:None-Apollonia:high	0.5000000000	-23.074038	24.0740380
## Fredo:None-Apollonia:high	5.5454545455	-14.558553	25.6494617
## Micheal:None-Apollonia:high	3.3333333333	-18.892483	25.5591495
## Sonny:None-Apollonia:high	6.2400000000	-13.389309	25.8693093
## Vito:None-Apollonia:high	8.1428571429	-12.434250	28.7199645
## Micheal:high-Fredo:high	0.1250000000	-14.041234	14.2912338
## Sonny:high-Fredo:high	0.5096153846	-4.938936	5.9581668
## Vito:high-Fredo:high	0.4583333333	-8.327192	9.2438586
## Apollonia:low-Fredo:high	1.1250000000	-4.702658	6.9526582
## Fredo:low-Fredo:high	1.1927966102	-3.117289	5.5028822
## Micheal:low-Fredo:high	4.0340909091	-1.647243	9.7154247
## Sonny:low-Fredo:high	2.2494979920	-1.864503	6.3634989
## Vito:low-Fredo:high	1.2812500000	-3.325918	5.8884183
## Apollonia:med-Fredo:high	5.0535714286	-1.419502	11.5266445

## Fredo:med-Fredo:high	1.3346774194	-3.292713	5.9620675
## Micheal:med-Fredo:high	4.1250000000	-2.210332	10.4603323
## Sonny:med-Fredo:high	2.3970000000	-1.892640	6.6866401
## Vito:med-Fredo:high	1.5795454545	-3.584181	6.7432714
## Apollonia:None-Fredo:high	-3.3750000000	-17.541234	10.7912338
## Fredo:None-Fredo:high	1.6704545455	-5.337970	8.6788792
## Micheal:None-Fredo:high	-0.5416666667	-12.328686	11.2453523
## Sonny:None-Fredo:high	2.3650000000	-3.135609	7.8656089
## Vito:None-Fredo:high	4.2678571429	-4.000411	12.5361253
## Sonny:high-Micheal:high	0.3846153846	-13.739644	14.5088750
## Vito:high-Micheal:high	0.3333333333	-15.382692	16.0493586
## Apollonia:low-Micheal:high	1.0000000000	-13.274789	15.2747889
## Fredo:low-Micheal:high	1.0677966102	-12.657539	14.7931322
## Micheal:low-Micheal:high	3.9090909091	-10.306589	18.1247707
## Sonny:low-Micheal:high	2.1244979920	-11.540530	15.7895264
## Vito:low-Micheal:high	1.1562500000	-12.665255	14.9777549
## Apollonia:med-Micheal:high	4.9285714286	-9.621641	19.4787835
## Fredo:med-Micheal:high	1.2096774194	-12.618581	15.0379360
## Micheal:med-Micheal:high	4.0000000000	-10.489459	18.4894594
## Sonny:med-Micheal:high	2.2720000000	-11.446929	15.9909289
## Vito:med-Micheal:high	1.4545454545	-12.562304	15.4713947
## Apollonia:None-Micheal:high	-3.5000000000	-22.748121	15.7481214
## Fredo:None-Micheal:high	1.5454545455	-13.250694	16.3416032
## Micheal:None-Micheal:high	-0.6666666667	-18.237717	16.9043838
## Sonny:None-Micheal:high	2.2400000000	-11.904423	16.3844228
## Vito:None-Micheal:high	4.1428571429	-11.289973	19.5756876
## Vito:high-Sonny:high	-0.0512820513	-8.768964	8.6664003
## Apollonia:low-Sonny:high	0.6153846154	-5.109485	6.3402541
## Fredo:low-Sonny:high	0.6831812256	-3.486875	4.8532375
## Micheal:low-Sonny:high	3.5244755245	-2.051373	9.1003239
## Sonny:low-Sonny:high	1.7398826074	-2.227174	5.7069388
## Vito:low-Sonny:high	0.7716346154	-3.704807	5.2480764
## Apollonia:med-Sonny:high	4.5439560440	-1.836734	10.9246462
## Fredo:med-Sonny:high	0.8250620347	-3.672189	5.3223134
## Micheal:med-Sonny:high	3.6153846154	-2.625526	9.8562954
## Sonny:med-Sonny:high	1.8873846154	-2.261536	6.0363054
## Vito:med-Sonny:high	1.0699300699	-3.977505	6.1173648
## Apollonia:None-Sonny:high	-3.8846153846	-18.008875	10.2396443
## Fredo:None-Sonny:high	1.1608391608	-5.762350	8.0840283
## Micheal:None-Sonny:high	-1.0512820513	-12.787821	10.6852570
## Sonny:None-Sonny:high	1.8553846154	-3.536204	7.2469734
## Vito:None-Sonny:high	3.7582417582	-4.437903	11.9543866
## Apollonia:low-Vito:high	0.6666666667	-8.292846	9.6261796
## Fredo:low-Vito:high	0.7344632768	-7.320852	8.7897789
## Micheal:low-Vito:high	3.5757575758	-5.289276	12.4407914
## Sonny:low-Vito:high	1.7911646586	-6.160959	9.7432886
## Vito:low-Vito:high	0.8229166667	-7.395190	9.0410230
## Apollonia:med-Vito:high	4.5952380952	-4.796883	13.9873596
## Fredo:med-Vito:high	0.8763440860	-7.353116	9.1058040
## Micheal:med-Vito:high	3.6666666667	-5.631059	12.9643926
## Sonny:med-Vito:high	1.9386666667	-6.105728	9.9830613
## Vito:med-Vito:high	1.1212121212	-7.421348	9.6637725
## Apollonia:None-Vito:high	-3.8333333333	-19.549359	11.8826920
## Fredo:None-Vito:high	1.2121212121	-8.556670	10.9809127

## Micheal:None-Vito:high	-1.0000000000	-14.610477	12.6104772
## Sonny:None-Vito:high	1.9066666667	-6.843646	10.6569792
## Vito:None-Vito:high	3.8095238095	-6.899142	14.5181900
## Fredo:low-Apollonia:low	0.0677966102	-4.586691	4.7222843
## Micheal:low-Apollonia:low	2.9090909091	-3.037755	8.8559364
## Sonny:low-Apollonia:low	1.1244979920	-3.349026	5.5980225
## Vito:low-Apollonia:low	0.1562500000	-4.774614	5.0871138
## Apollonia:med-Apollonia:low	3.9285714286	-2.778745	10.6358878
## Fredo:med-Apollonia:low	0.2096774194	-4.740086	5.1594407
## Micheal:med-Apollonia:low	3.0000000000	-3.574485	9.5744851
## Sonny:med-Apollonia:low	1.2720000000	-3.363561	5.9075615
## Vito:med-Apollonia:low	0.4545454545	-4.999947	5.9090375
## Apollonia:None-Apollonia:low	-4.5000000000	-18.774789	9.7747889
## Fredo:None-Apollonia:low	0.5454545455	-6.679878	7.7707873
## Micheal:None-Apollonia:low	-1.6666666667	-13.583933	10.2505996
## Sonny:None-Apollonia:low	1.2400000000	-4.534436	7.0144364
## Vito:None-Apollonia:low	3.1428571429	-5.310052	11.5957665
## Micheal:low-Fredo:low	2.8412942989	-1.628629	7.3112171
## Sonny:low-Fredo:low	1.0567013818	-1.094498	3.2079010
## Vito:low-Fredo:low	0.0884533898	-2.899632	3.0765391
## Apollonia:med-Fredo:low	3.8607748184	-1.580120	9.3016692
## Fredo:med-Fredo:low	0.1418808092	-2.877290	3.1610521
## Micheal:med-Fredo:low	2.9322033898	-2.344073	8.2084798
## Sonny:med-Fredo:low	1.2042033898	-1.266357	3.6747638
## Vito:med-Fredo:low	0.3867488444	-3.403597	4.1770948
## Apollonia:None-Fredo:low	-4.5677966102	-18.293132	9.1575389
## Fredo:None-Fredo:low	0.4776579353	-5.590346	6.5456619
## Micheal:None-Fredo:low	-1.7344632768	-12.987750	9.5188239
## Sonny:None-Fredo:low	1.1722033898	-3.065644	5.4100512
## Vito:None-Fredo:low	3.0750605327	-4.412724	10.5628449
## Sonny:low-Micheal:low	-1.7845929171	-6.065758	2.4965722
## Vito:low-Micheal:low	-2.7528409091	-7.509875	2.0041930
## Apollonia:med-Micheal:low	1.0194805195	-5.561101	7.6000615
## Fredo:med-Micheal:low	-2.6994134897	-7.476035	2.0772078
## Micheal:med-Micheal:low	0.0909090909	-6.354229	6.5360474
## Sonny:med-Micheal:low	-1.6370909091	-6.087303	2.8131208
## Vito:med-Micheal:low	-2.4545454545	-7.752417	2.8433256
## Apollonia:None-Micheal:low	-7.4090909091	-21.624771	6.8065889
## Fredo:None-Micheal:low	-2.3636363636	-9.471476	4.7442035
## Micheal:None-Micheal:low	-4.5757575758	-16.422157	7.2706422
## Sonny:None-Micheal:low	-1.6690909091	-7.295819	3.9576372
## Vito:None-Micheal:low	0.2337662338	-8.118936	8.5864682
## Vito:low-Sonny:low	-0.9682479920	-3.665806	1.7293104
## Apollonia:med-Sonny:low	2.8040734366	-2.482844	8.0909909
## Fredo:med-Sonny:low	-0.9148205726	-3.646772	1.8171312
## Micheal:med-Sonny:low	1.8755020080	-3.241847	6.9928507
## Sonny:med-Sonny:low	0.1475020080	-1.962435	2.2574389
## Vito:med-Sonny:low	-0.6699525374	-4.235747	2.8958415
## Apollonia:None-Sonny:low	-5.6244979920	-19.289526	8.0405304
## Fredo:None-Sonny:low	-0.5790434465	-6.509375	5.3512885
## Micheal:None-Sonny:low	-2.7911646586	-13.970818	8.3884882
## Sonny:None-Sonny:low	0.1155020080	-3.922755	4.1537590
## Vito:None-Sonny:low	2.0183591509	-5.358299	9.3950170
## Apollonia:med-Vito:low	3.7723214286	-1.906806	9.4514493

## Fredo:med-Vito:low	0.0534274194	-3.376523	3.4833774
## Micheal:med-Vito:low	2.8437500000	-2.677867	8.3653670
## Sonny:med-Vito:low	1.1157500000	-1.842768	4.0742683
## Vito:med-Vito:low	0.2982954545	-3.826736	4.4233269
## Apollonia:None-Vito:low	-4.6562500000	-18.477755	9.1652549
## Fredo:None-Vito:low	0.3892045455	-5.893298	6.6717068
## Micheal:None-Vito:low	-1.8229166667	-13.193301	9.5474674
## Sonny:None-Vito:low	1.0837500000	-3.455910	5.6234104
## Vito:None-Vito:low	2.9866071429	-4.676035	10.6492488
## Fredo:med-Apollonia:med	-3.7188940092	-9.414439	1.9766509
## Micheal:med-Apollonia:med	-0.9285714286	-8.081398	6.2242552
## Sonny:med-Apollonia:med	-2.6565714286	-8.081284	2.7681411
## Vito:med-Apollonia:med	-3.4740259740	-9.613290	2.6652378
## Apollonia:None-Apollonia:med	-8.4285714286	-22.978784	6.1216407
## Fredo:None-Apollonia:med	-3.3831168831	-11.138406	4.3721722
## Micheal:None-Apollonia:med	-5.5952380952	-17.841066	6.6505902
## Sonny:None-Apollonia:med	-2.6885714286	-9.113771	3.7366283
## Vito:None-Apollonia:med	-0.7857142857	-9.695863	8.1244345
## Micheal:med-Fredo:med	2.7903225806	-2.748178	8.3288236
## Sonny:med-Fredo:med	1.0623225806	-1.927589	4.0522339
## Vito:med-Fredo:med	0.2448680352	-3.902736	4.3924725
## Apollonia:None-Fredo:med	-4.7096774194	-18.537936	9.1185812
## Fredo:None-Fredo:med	0.3357771261	-5.961569	6.6331236
## Micheal:None-Fredo:med	-1.8763440860	-13.254937	9.5022487
## Sonny:None-Fredo:med	1.0303225806	-3.529859	5.5905041
## Vito:None-Fredo:med	2.9331797235	-4.741637	10.6079967
## Sonny:med-Micheal:med	-1.7280000000	-6.987588	3.5315881
## Vito:med-Micheal:med	-2.5454545455	-8.539311	3.4484023
## Apollonia:None-Micheal:med	-7.5000000000	-21.989459	6.9894594
## Fredo:None-Micheal:med	-2.4545454545	-10.095244	5.1861528
## Micheal:None-Micheal:med	-4.6666666667	-16.840248	7.5069142
## Sonny:None-Micheal:med	-1.7600000000	-8.046410	4.5264101
## Vito:None-Micheal:med	0.1428571429	-8.667734	8.9534481
## Vito:med-Sonny:med	-0.8174545455	-4.584535	2.9496261
## Apollonia:None-Sonny:med	-5.7720000000	-19.490929	7.9469289
## Fredo:None-Sonny:med	-0.7265454545	-6.780044	5.3269533
## Micheal:None-Sonny:med	-2.9386666667	-14.184139	8.3068056
## Sonny:None-Sonny:med	-0.0320000000	-4.249052	4.1850521
## Vito:None-Sonny:med	1.8708571429	-5.605177	9.3468915
## Apollonia:None-Vito:med	-4.9545454545	-18.971395	9.0623038
## Fredo:None-Vito:med	0.0909090909	-6.610427	6.7922448
## Micheal:None-Vito:med	-2.1212121212	-13.728266	9.4858418
## Sonny:None-Vito:med	0.7854545455	-4.318131	5.8890398
## Vito:None-Vito:med	2.6883116883	-5.321316	10.6979398
## Fredo:None-Apollonia:None	5.0454545455	-9.750694	19.8416032
## Micheal:None-Apollonia:None	2.8333333333	-14.737717	20.4043838
## Sonny:None-Apollonia:None	5.7400000000	-8.404423	19.8844228
## Vito:None-Apollonia:None	7.6428571429	-7.789973	23.0756876
## Micheal:None-Fredo:None	-2.2121212121	-14.749172	10.3249299
## Sonny:None-Fredo:None	0.6945454545	-6.269687	7.6587778
## Vito:None-Fredo:None	2.5974025974	-6.708944	11.9037495
## Sonny:None-Micheal:None	2.9066666667	-8.854130	14.6674631
## Vito:None-Micheal:None	4.8095238095	-8.472942	18.0919895
## Vito:None-Sonny:None	1.9028571429	-6.327986	10.1337001

##	p adj
## Fredo:critical-Apollonia:critical	1.0000000
## Micheal:critical-Apollonia:critical	1.0000000
## Sonny:critical-Apollonia:critical	1.0000000
## Vito:critical-Apollonia:critical	1.0000000
## Apollonia:high-Apollonia:critical	0.9981321
## Fredo:high-Apollonia:critical	0.9338284
## Micheal:high-Apollonia:critical	0.9999924
## Sonny:high-Apollonia:critical	0.9780080
## Vito:high-Apollonia:critical	0.9991044
## Apollonia:low-Apollonia:critical	0.9979278
## Fredo:low-Apollonia:critical	0.9927551
## Micheal:low-Apollonia:critical	1.0000000
## Sonny:low-Apollonia:critical	0.9999507
## Vito:low-Apollonia:critical	0.9960368
## Apollonia:med-Apollonia:critical	1.0000000
## Fredo:med-Apollonia:critical	0.9968462
## Micheal:med-Apollonia:critical	1.0000000
## Sonny:med-Apollonia:critical	0.9999855
## Vito:med-Apollonia:critical	0.9993823
## Apollonia:None-Apollonia:critical	0.9657925
## Fredo:None-Apollonia:critical	0.9999602
## Micheal:None-Apollonia:critical	0.9991328
## Sonny:None-Apollonia:critical	0.9999963
## Vito:None-Apollonia:critical	1.0000000
## Micheal:critical-Fredo:critical	1.0000000
## Sonny:critical-Fredo:critical	0.9999997
## Vito:critical-Fredo:critical	1.0000000
## Apollonia:high-Fredo:critical	0.9998810
## Fredo:high-Fredo:critical	0.8249665
## Micheal:high-Fredo:critical	1.0000000
## Sonny:high-Fredo:critical	0.9585717
## Vito:high-Fredo:critical	0.9999884
## Apollonia:low-Fredo:critical	0.9997057
## Fredo:low-Fredo:critical	0.9474043
## Micheal:low-Fredo:critical	1.0000000
## Sonny:low-Fredo:critical	0.9999996
## Vito:low-Fredo:critical	0.9907540
## Apollonia:med-Fredo:critical	0.9999741
## Fredo:med-Fredo:critical	0.9943061
## Micheal:med-Fredo:critical	1.0000000
## Sonny:med-Fredo:critical	1.0000000
## Vito:med-Fredo:critical	0.9999527
## Apollonia:None-Fredo:critical	0.9932746
## Fredo:None-Fredo:critical	1.0000000
## Micheal:None-Fredo:critical	0.9999869
## Sonny:None-Fredo:critical	1.0000000
## Vito:None-Fredo:critical	1.0000000
## Sonny:critical-Micheal:critical	1.0000000
## Vito:critical-Micheal:critical	1.0000000
## Apollonia:high-Micheal:critical	0.9997940
## Fredo:high-Micheal:critical	0.9987003
## Micheal:high-Micheal:critical	1.0000000
## Sonny:high-Micheal:critical	0.9998479

## Vito:high-Micheal:critical	0.9999960
## Apollonia:low-Micheal:critical	0.9999974
## Fredo:low-Micheal:critical	0.9999907
## Micheal:low-Micheal:critical	1.0000000
## Sonny:low-Micheal:critical	1.0000000
## Vito:low-Micheal:critical	0.9999964
## Apollonia:med-Micheal:critical	1.0000000
## Fredo:med-Micheal:critical	0.9999976
## Micheal:med-Micheal:critical	1.0000000
## Sonny:med-Micheal:critical	1.0000000
## Vito:med-Micheal:critical	0.9999998
## Apollonia:None-Micheal:critical	0.9952426
## Fredo:None-Micheal:critical	1.0000000
## Micheal:None-Micheal:critical	0.9999861
## Sonny:None-Micheal:critical	1.0000000
## Vito:None-Micheal:critical	1.0000000
## Vito:critical-Sonny:critical	1.0000000
## Apollonia:high-Sonny:critical	0.9989627
## Fredo:high-Sonny:critical	0.1229204
## Micheal:high-Sonny:critical	0.9999983
## Sonny:high-Sonny:critical	0.2831871
## Vito:high-Sonny:critical	0.9967116
## Apollonia:low-Sonny:critical	0.8536035
## Fredo:low-Sonny:critical	0.0178781
## Micheal:low-Sonny:critical	1.0000000
## Sonny:low-Sonny:critical	0.4181876
## Vito:low-Sonny:critical	0.1665849
## Apollonia:med-Sonny:critical	1.0000000
## Fredo:med-Sonny:critical	0.2115316
## Micheal:med-Sonny:critical	1.0000000
## Sonny:med-Sonny:critical	0.8241631
## Vito:med-Sonny:critical	0.8168239
## Apollonia:None-Sonny:critical	0.9591701
## Fredo:None-Sonny:critical	0.9995826
## Micheal:None-Sonny:critical	0.9990844
## Sonny:None-Sonny:critical	0.9995689
## Vito:None-Sonny:critical	1.0000000
## Apollonia:high-Vito:critical	0.9994990
## Fredo:high-Vito:critical	0.7622721
## Micheal:high-Vito:critical	0.9999997
## Sonny:high-Vito:critical	0.9149740
## Vito:high-Vito:critical	0.9997322
## Apollonia:low-Vito:critical	0.9964949
## Fredo:low-Vito:critical	0.9324808
## Micheal:low-Vito:critical	1.0000000
## Sonny:low-Vito:critical	0.9999113
## Vito:low-Vito:critical	0.9750053
## Apollonia:med-Vito:critical	1.0000000
## Fredo:med-Vito:critical	0.9814541
## Micheal:med-Vito:critical	1.0000000
## Sonny:med-Vito:critical	0.9999922
## Vito:med-Vito:critical	0.9988431
## Apollonia:None-Vito:critical	0.9808806
## Fredo:None-Vito:critical	0.9999958

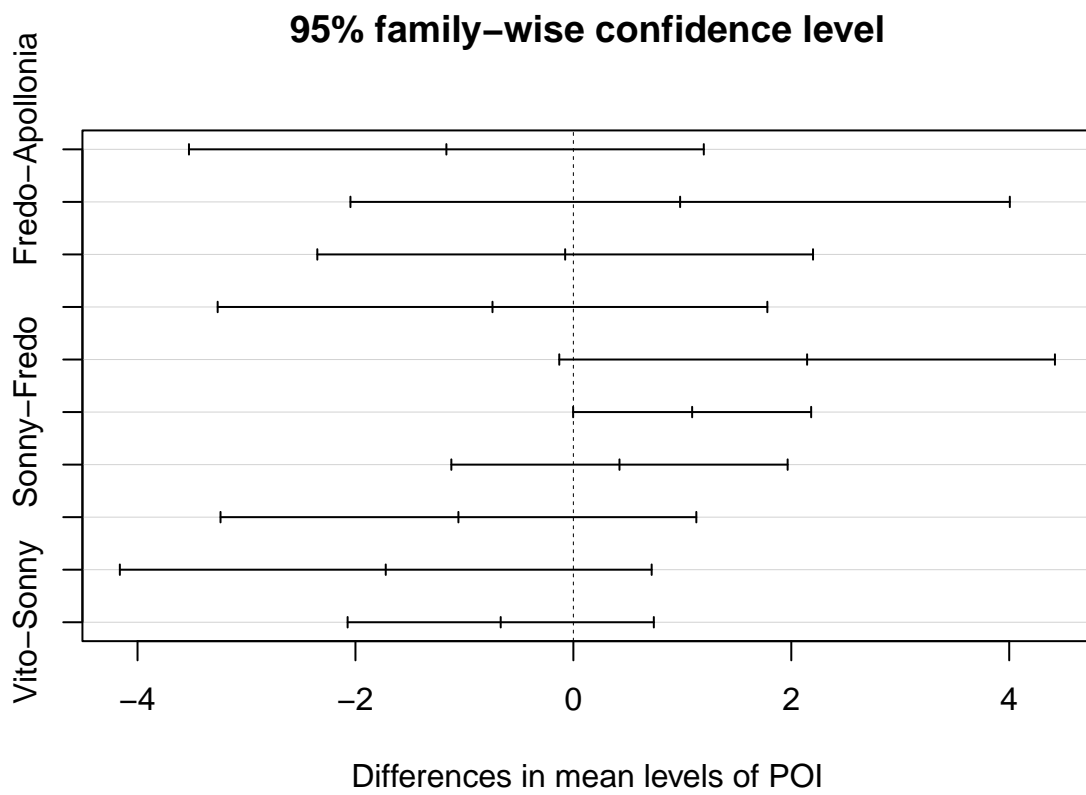
## Micheal:None-Vito:critical	0.9998246
## Sonny:None-Vito:critical	0.9999998
## Vito:None-Vito:critical	1.0000000
## Fredo:high-Apollonia:high	1.0000000
## Micheal:high-Apollonia:high	1.0000000
## Sonny:high-Apollonia:high	1.0000000
## Vito:high-Apollonia:high	1.0000000
## Apollonia:low-Apollonia:high	0.9999998
## Fredo:low-Apollonia:high	0.9999997
## Micheal:low-Apollonia:high	0.9992050
## Sonny:low-Apollonia:high	0.9999862
## Vito:low-Apollonia:high	0.9999996
## Apollonia:med-Apollonia:high	0.9959481
## Fredo:med-Apollonia:high	0.9999995
## Micheal:med-Apollonia:high	0.9991875
## Sonny:med-Apollonia:high	0.9999794
## Vito:med-Apollonia:high	0.9999988
## Apollonia:None-Apollonia:high	1.0000000
## Fredo:None-Apollonia:high	0.9999991
## Micheal:None-Apollonia:high	1.0000000
## Sonny:None-Apollonia:high	0.9999860
## Vito:None-Apollonia:high	0.9993773
## Micheal:high-Fredo:high	1.0000000
## Sonny:high-Fredo:high	1.0000000
## Vito:high-Fredo:high	1.0000000
## Apollonia:low-Fredo:high	1.0000000
## Fredo:low-Fredo:high	0.9999990
## Micheal:low-Fredo:high	0.6138533
## Sonny:low-Fredo:high	0.9522111
## Vito:low-Fredo:high	0.9999989
## Apollonia:med-Fredo:high	0.4081536
## Fredo:med-Fredo:high	0.9999978
## Micheal:med-Fredo:high	0.7753085
## Sonny:med-Fredo:high	0.9401375
## Vito:med-Fredo:high	0.9999932
## Apollonia:None-Fredo:high	1.0000000
## Fredo:None-Fredo:high	1.0000000
## Micheal:None-Fredo:high	1.0000000
## Sonny:None-Fredo:high	0.9977719
## Vito:None-Fredo:high	0.9748409
## Sonny:high-Micheal:high	1.0000000
## Vito:high-Micheal:high	1.0000000
## Apollonia:low-Micheal:high	1.0000000
## Fredo:low-Micheal:high	1.0000000
## Micheal:low-Micheal:high	0.9999991
## Sonny:low-Micheal:high	1.0000000
## Vito:low-Micheal:high	1.0000000
## Apollonia:med-Micheal:high	0.9999554
## Fredo:med-Micheal:high	1.0000000
## Micheal:med-Micheal:high	0.9999991
## Sonny:med-Micheal:high	1.0000000
## Vito:med-Micheal:high	1.0000000
## Apollonia:None-Micheal:high	1.0000000
## Fredo:None-Micheal:high	1.0000000

## Micheal:None-Micheal:high	1.0000000
## Sonny:None-Micheal:high	1.0000000
## Vito:None-Micheal:high	0.9999995
## Vito:high-Sonny:high	1.0000000
## Apollonia:low-Sonny:high	1.0000000
## Fredo:low-Sonny:high	1.0000000
## Micheal:low-Sonny:high	0.8196226
## Sonny:low-Sonny:high	0.9970216
## Vito:low-Sonny:high	1.0000000
## Apollonia:med-Sonny:high	0.6077425
## Fredo:med-Sonny:high	1.0000000
## Micheal:med-Sonny:high	0.9147176
## Sonny:med-Sonny:high	0.9950023
## Vito:med-Sonny:high	1.0000000
## Apollonia:None-Sonny:high	0.9999991
## Fredo:None-Sonny:high	1.0000000
## Micheal:None-Sonny:high	1.0000000
## Sonny:None-Sonny:high	0.9999409
## Vito:None-Sonny:high	0.9944231
## Apollonia:low-Vito:high	1.0000000
## Fredo:low-Vito:high	1.0000000
## Micheal:low-Vito:high	0.9991576
## Sonny:low-Vito:high	1.0000000
## Vito:low-Vito:high	1.0000000
## Apollonia:med-Vito:high	0.9868774
## Fredo:med-Vito:high	1.0000000
## Micheal:med-Vito:high	0.9994107
## Sonny:med-Vito:high	0.9999999
## Vito:med-Vito:high	1.0000000
## Apollonia:None-Vito:high	0.9999999
## Fredo:None-Vito:high	1.0000000
## Micheal:None-Vito:high	1.0000000
## Sonny:None-Vito:high	1.0000000
## Vito:None-Vito:high	0.9998943
## Fredo:low-Apollonia:low	1.0000000
## Micheal:low-Apollonia:low	0.9869055
## Sonny:low-Apollonia:low	0.9999999
## Vito:low-Apollonia:low	1.0000000
## Apollonia:med-Apollonia:low	0.9054736
## Fredo:med-Apollonia:low	1.0000000
## Micheal:med-Apollonia:low	0.9947852
## Sonny:med-Apollonia:low	0.9999992
## Vito:med-Apollonia:low	1.0000000
## Apollonia:None-Apollonia:low	0.9999880
## Fredo:None-Apollonia:low	1.0000000
## Micheal:None-Apollonia:low	1.0000000
## Sonny:None-Apollonia:low	1.0000000
## Vito:None-Apollonia:low	0.9997756
## Micheal:low-Fredo:low	0.8117201
## Sonny:low-Fredo:low	0.9862030
## Vito:low-Fredo:low	1.0000000
## Apollonia:med-Fredo:low	0.6152487
## Fredo:med-Fredo:low	1.0000000
## Micheal:med-Fredo:low	0.9433994

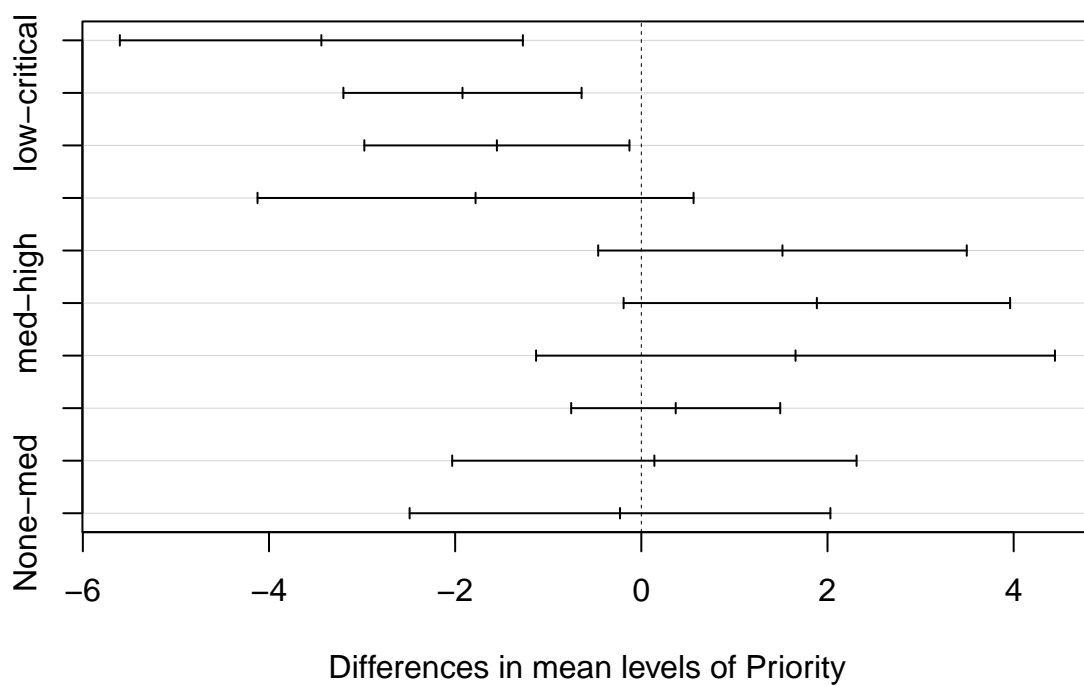
## Sonny:med-Fredo:low	0.9874904
## Vito:med-Fredo:low	1.0000000
## Apollonia:None-Fredo:low	0.9999675
## Fredo:None-Fredo:low	1.0000000
## Micheal:None-Fredo:low	1.0000000
## Sonny:None-Fredo:low	0.9999990
## Vito:None-Fredo:low	0.9988854
## Sonny:low-Micheal:low	0.9985991
## Vito:low-Micheal:low	0.9155719
## Apollonia:med-Micheal:low	1.0000000
## Fredo:med-Micheal:low	0.9329413
## Micheal:med-Micheal:low	1.0000000
## Sonny:med-Micheal:low	0.9998125
## Vito:med-Micheal:low	0.9935768
## Apollonia:None-Micheal:low	0.9718573
## Fredo:None-Micheal:low	0.9999680
## Micheal:None-Micheal:low	0.9995786
## Sonny:None-Micheal:low	0.9999962
## Vito:None-Micheal:low	1.0000000
## Vito:low-Sonny:low	0.9998767
## Apollonia:med-Sonny:low	0.9656949
## Fredo:med-Sonny:low	0.9999637
## Micheal:med-Sonny:low	0.9998239
## Sonny:med-Sonny:low	1.0000000
## Vito:med-Sonny:low	1.0000000
## Apollonia:None-Sonny:low	0.9988462
## Fredo:None-Sonny:low	1.0000000
## Micheal:None-Sonny:low	0.9999999
## Sonny:None-Sonny:low	1.0000000
## Vito:None-Sonny:low	0.9999992
## Apollonia:med-Vito:low	0.7420503
## Fredo:med-Vito:low	1.0000000
## Micheal:med-Vito:low	0.9754895
## Sonny:med-Vito:low	0.9997155
## Vito:med-Vito:low	1.0000000
## Apollonia:None-Vito:low	0.9999595
## Fredo:None-Vito:low	1.0000000
## Micheal:None-Vito:low	1.0000000
## Sonny:None-Vito:low	1.0000000
## Vito:None-Vito:low	0.9995121
## Fredo:med-Apollonia:med	0.7707768
## Micheal:med-Apollonia:med	1.0000000
## Sonny:med-Apollonia:med	0.9867237
## Vito:med-Apollonia:med	0.9320639
## Apollonia:None-Apollonia:med	0.9147582
## Fredo:None-Apollonia:med	0.9972447
## Micheal:None-Apollonia:med	0.9946896
## Sonny:None-Apollonia:med	0.9985154
## Vito:None-Apollonia:med	1.0000000
## Micheal:med-Fredo:med	0.9811455
## Sonny:med-Fredo:med	0.9998966
## Vito:med-Fredo:med	1.0000000
## Apollonia:None-Fredo:med	0.9999508
## Fredo:None-Fredo:med	1.0000000

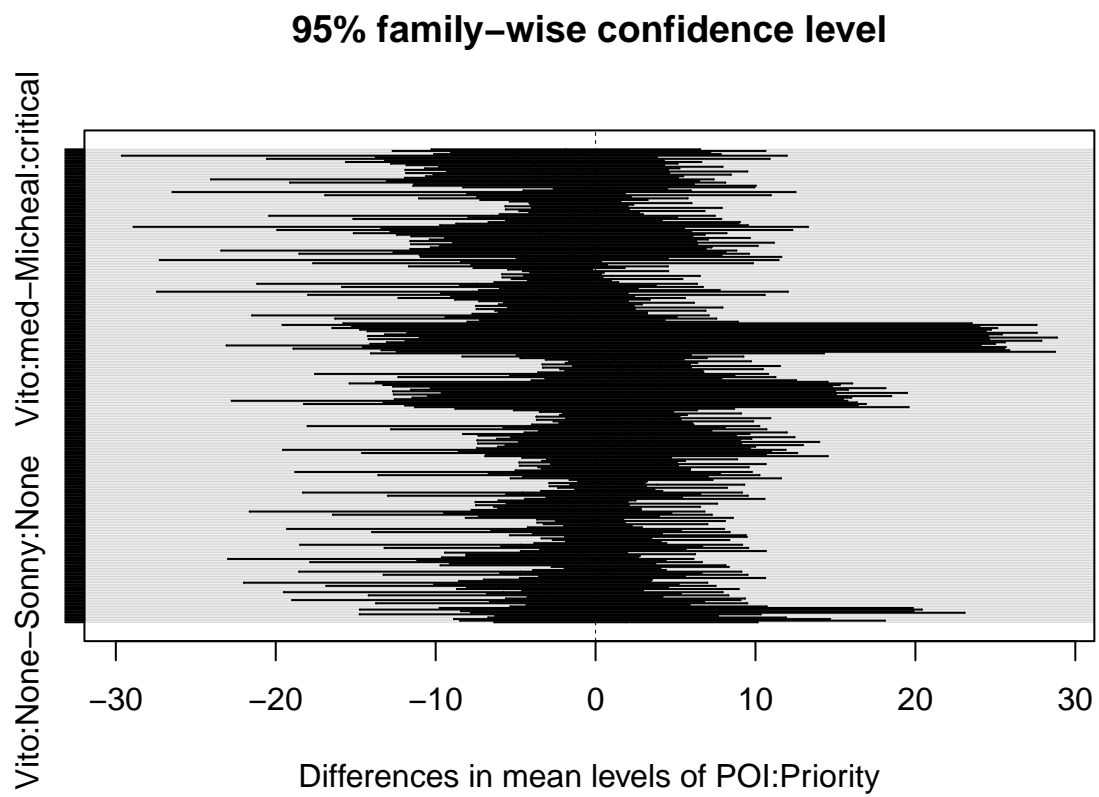
## Micheal:None-Fredo:med	1.0000000
## Sonny:None-Fredo:med	1.0000000
## Vito:None-Fredo:med	0.9996457
## Sonny:med-Micheal:med	0.9999743
## Vito:med-Micheal:med	0.9981452
## Apollonia:None-Micheal:med	0.9740085
## Fredo:None-Micheal:med	0.9999829
## Micheal:None-Micheal:med	0.9996276
## Sonny:None-Micheal:med	0.9999988
## Vito:None-Micheal:med	1.0000000
## Vito:med-Sonny:med	1.0000000
## Apollonia:None-Sonny:med	0.9983877
## Fredo:None-Sonny:med	1.0000000
## Micheal:None-Sonny:med	0.9999997
## Sonny:None-Sonny:med	1.0000000
## Vito:None-Sonny:med	0.9999999
## Apollonia:None-Vito:med	0.9999054
## Fredo:None-Vito:med	1.0000000
## Micheal:None-Vito:med	1.0000000
## Sonny:None-Vito:med	1.0000000
## Vito:None-Vito:med	0.9999621
## Fredo:None-Apollonia:None	0.9999497
## Micheal:None-Apollonia:None	1.0000000
## Sonny:None-Apollonia:None	0.9990735
## Vito:None-Apollonia:None	0.9847246
## Micheal:None-Fredo:None	1.0000000
## Sonny:None-Fredo:None	1.0000000
## Vito:None-Fredo:None	0.9999988
## Sonny:None-Micheal:None	0.9999999
## Vito:None-Micheal:None	0.9998567
## Vito:None-Sonny:None	1.0000000

```
plot(HSD)
```



95% family-wise confidence level





Analysis:

This test compares a set of means for the given category of data we observe that the pairs- high-critical,low-critical and med-critical are insignificant(i.e. have a p-value < 0.05)