cleaning.R

rmadh

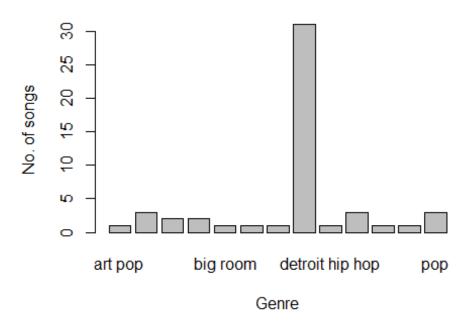
2020-02-27

```
#Top Songs Analysis
#importing dataset top10s and copying it to test data
data <- read.csv("C:/Users/rmadh/OneDrive/Desktop/Lecture Notes/MVA/Top-Songs</pre>
-Analysis-master/top10s.csv", header = TRUE)
View(data)
#Data Cleaning
#Adding column Rank which will denote rank of a song based on it popularity.
# popularity from 90 - 100 is Rank 10 and so on
for(x in 1:length(data$pop)){
  if(data[x,15] \leftarrow 100 \&\& data[x,15] \rightarrow 80){
    data[x,16] = 5
  }else if(data[x,15] < 80 && data[x,15] >= 60){
    data[x,16] = 4
  }else if(data[x,15] < 60 && data[x,15] >= 40){
    data[x, 16] = 3
  }else if(data[x,15] < 40 && data[x,15] >= 20){
    data[x,16] = 2
  }else if(data[x,15] < 20 && data[x,15] >= 0){
    data[x,16] = 1
  }
}
data$pop <- NULL</pre>
dim(data)
## [1] 603 15
#removing values with 0 BPM and duration as 0 seconds
data_clean <- data[-c(433),]</pre>
names(data_clean)[15]<- "rating"</pre>
View(data_clean)
#checking the ranges for all columns
dim(data_clean)
## [1] 602 15
```

```
library(plyr)
library(ggplot2)

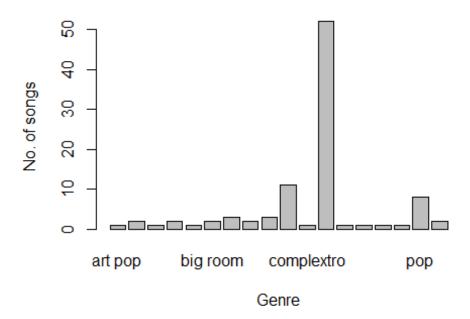
#Finding top genre for 3 years
year1 = data_clean[data_clean$year == 2010,]
gen1 = count(year1$top.genre)
barplot(gen1$freq, names.arg = gen1$x,main = 'Top Genres for 2010',xlab = 'Ge
nre',ylab = 'No. of songs')
```

Top Genres for 2010



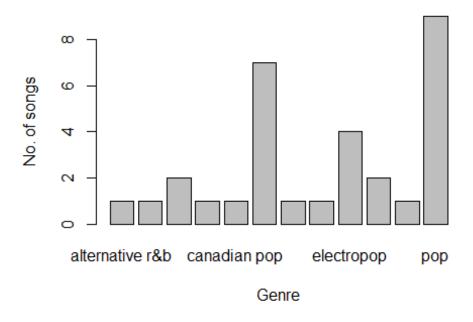
```
year2 = data_clean[data_clean$year == 2015,]
gen2 = count(year2$top.genre)
barplot(gen2$freq, names.arg = gen2$x,main = 'Top Genres for 2015',xlab = 'Ge
nre',ylab = 'No. of songs')
```

Top Genres for 2015



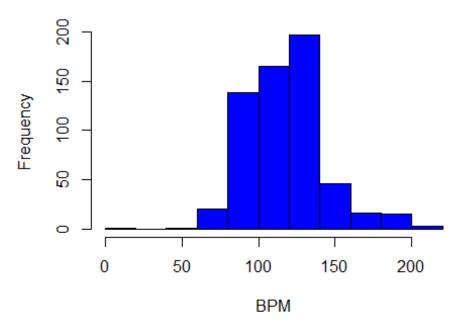
```
year3 = data_clean[data_clean$year == 2019,]
gen3 = count(year3$top.genre)
barplot(gen3$freq, names.arg = gen3$x,main = 'Top Genres for 2019',xlab = 'Ge
nre',ylab = 'No. of songs')
```

Top Genres for 2019

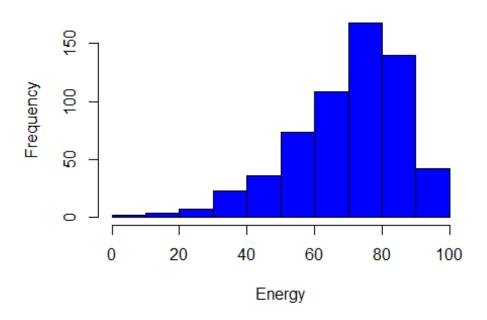


#Histogram view of audio properties
hist(data_clean\$bpm, breaks=12,col="blue",xlab="BPM")

Histogram of data_clean\$bpm

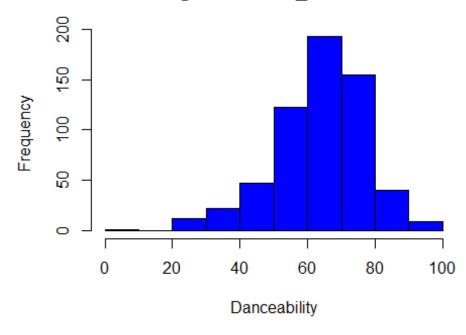


Histogram of data_clean\$nrgy

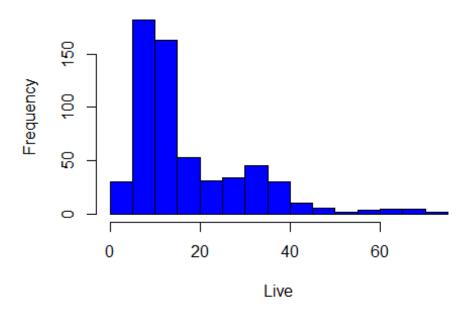


hist(data_clean\$dnce, breaks=12,col="blue",xlab="Danceability")

Histogram of data_clean\$dnce

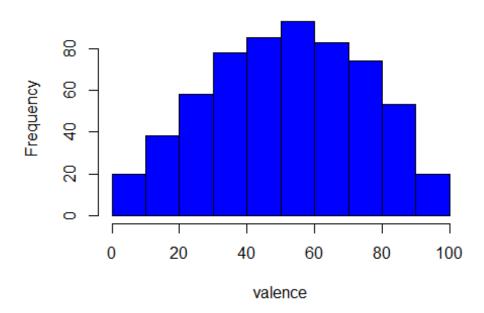


Histogram of data_clean\$live

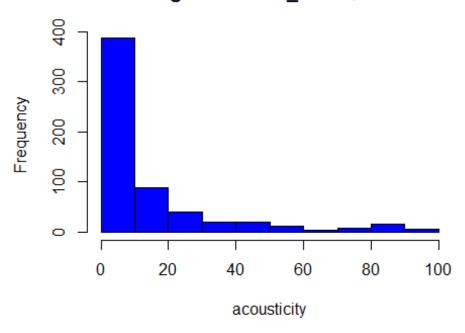


hist(data_clean\$val, breaks=12,col="blue",xlab="valence")

Histogram of data_clean\$val

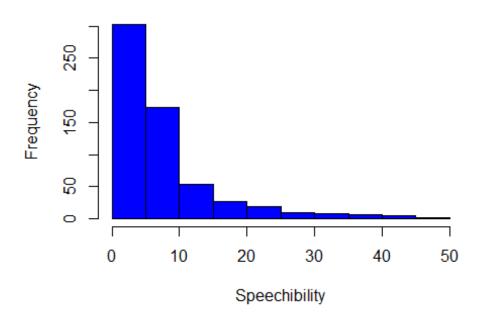


Histogram of data_clean\$acous

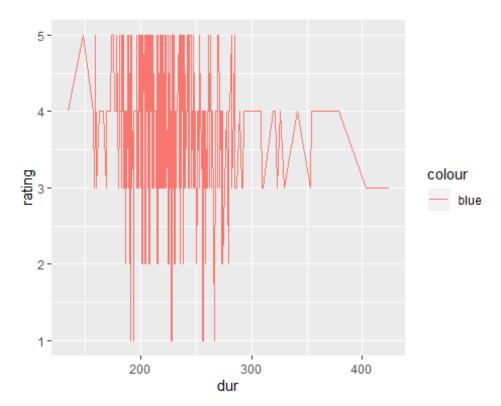


hist(data_clean\$spch, breaks=12,col="blue",xlab="Speechibility")

Histogram of data_clean\$spch



```
#Line chart for popularity and Duration
ggplot(data_clean) +geom_line(aes(x = dur, y = rating, color = "blue"))
```



```
# T-Test on dataset columns Duration and rating
t.test(data_clean$dur,data_clean$rating, var.equal = TRUE, paired=FALSE)
##
   Two Sample t-test
##
##
## data: data_clean$dur and data_clean$rating
## t = 158.71, df = 1202, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 218.0532 223.5116
## sample estimates:
## mean of x mean of y
## 224.611296
                3.828904
#Comparing relation between two top genre from 2010 to 2019.
star5 = data_clean[which(data_clean$rating==5),]
with(star5,t.test(dnce[top.genre=="dance pop"],dnce[top.genre=="pop"],var.equ
al=TRUE))
##
##
   Two Sample t-test
##
```

```
## data: dnce[top.genre == "dance pop"] and dnce[top.genre == "pop"]
## t = -1.0029, df = 40, p-value = 0.3219
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -13.676389
                4,604961
## sample estimates:
## mean of x mean of y
## 67.03571 71.57143
with(star5,t.test(nrgy[top.genre=="dance pop"],nrgy[top.genre=="pop"],var.equ
al=TRUE))
##
   Two Sample t-test
##
## data: nrgy[top.genre == "dance pop"] and nrgy[top.genre == "pop"]
## t = 1.7587, df = 40, p-value = 0.08629
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -1.433565 20.647851
## sample estimates:
## mean of x mean of y
## 66.67857 57.07143
with(star5, t.test(bpm[top.genre=="dance pop"], bpm[top.genre=="pop"], var.equal
=TRUE))
##
## Two Sample t-test
## data: bpm[top.genre == "dance pop"] and bpm[top.genre == "pop"]
## t = 2.1881, df = 40, p-value = 0.03456
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
     1.147886 28.923542
## sample estimates:
## mean of x mean of v
## 119.3929 104.3571
with(star5, t.test(val[top.genre=="dance pop"], val[top.genre=="pop"], var.equal
=TRUE))
##
  Two Sample t-test
## data: val[top.genre == "dance pop"] and val[top.genre == "pop"]
## t = -1.4541, df = 40, p-value = 0.1537
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -27.825938
                 4.540224
## sample estimates:
```

```
## mean of x mean of y
## 48.78571 60.42857
#-----PCA-----
#Splitting the rating column in 2 groups as we need 2 levels for t test
#and var test (f test) calculation, so rating 1 has ratings in range 1 to 3
#and rating 5 has ratings in range from 4 to 5.
#A new column v16 stores this new rating value which is used for above mentio
ned tests
for(y in 1:length(data clean$rating)){
 if(data clean[y,15] \rightarrow= 1 & data clean[y,15] <= 3){
   data_clean[y, 16] = 1
 }else{
   data_clean[y, 16] = 5
 }
}
View(data clean)
#We are selecting audio properties to check if any correlation
#exist between them and does that affect the rating energy, danceability, val
ence, acoustics
#and speechability is observed.
aud_prop_cor = cor(data_clean[c(7,8,11,13,14)])
              nrgy
                          dnce
                                     val
                                               acous
                                                            spch
## nrgy 1.0000000 0.16685024 0.4102908 -0.5625564 0.10711812
## dnce 0.1668502 1.00000000 0.5049296 -0.2413363 -0.02922118
## val
         0.4102908 0.50492963 1.0000000 -0.2486811 0.12284677
## acous -0.5625564 -0.24133632 -0.2486811 1.0000000 0.00246410
## spch
         0.1071181 -0.02922118 0.1228468 0.0024641 1.00000000
# Correlation is low but danceability and valence are closely related
# Calculating PCA for the cleaned data
data_pca = prcomp(aud_prop_cor,scale. = TRUE)
data pca
## Standard deviations (1, .., p=5):
## [1] 1.4439153 1.0176814 1.0011165 0.7365874 0.5784789
##
## Rotation (n x k) = (5 \times 5):
                                      PC3
                                                 PC4
                                                             PC5
                PC1
                           PC2
## nrgy -0.53106816 0.3018103 -0.3408606 -0.3818033 -0.60408400
## dnce -0.43372652 -0.5131816 0.3929811 0.4823965 -0.40172805
## val -0.52681796 -0.1571937 0.3907000 -0.5388521 0.50472255
## acous 0.49239464 -0.1382874 0.5100046 -0.5094188 -0.46777338
## spch -0.09928882 0.7757074 0.5626977 0.2676767 -0.01184546
summary(data pca)
```

```
## Importance of components:
##
                           PC1
                                 PC2
                                        PC3
                                               PC4
                                                       PC5
                         1.444 1.0177 1.0011 0.7366 0.57848
## Standard deviation
## Proportion of Variance 0.417 0.2071 0.2004 0.1085 0.06693
## Cumulative Proportion 0.417 0.6241 0.8246 0.9331 1.00000
data_pca$x
               PC1
                            PC2
                                                     PC4
##
                                        PC3
                                                                  PC5
      -1.168320396 -0.435362099 -0.039151263 -1.271456683 -0.2412041809
## 1
## 2
      -1.317131044 1.378217388 1.385378953 -0.136793273 -1.1308962645
      ## 3
      -1.602879141 -0.305790987 -0.636065986 -0.552231502 -0.2164379874
## 4
## 5
      -0.445434523 -0.041214120 -1.082152129 0.039428584 -0.4127259666
data_pca1 = cbind(data.frame(data_clean$V16),data_pca$x)
data_pca1
##
                                          PC2
      data_clean.V16
                              PC1
                                                       PC3
                                                                   PC4
## 1
                   5 -1.168320396 -0.435362099 -0.039151263 -1.271456683
## 2
                   5 -1.317131044 1.378217388 1.385378953 -0.136793273
## 3
                   5 -1.432924832 0.285364744 0.704262305 -0.036255619
## 4
                   5 -1.602879141 -0.305790987 -0.636065986 -0.552231502
## 5
                   5 -0.445434523 -0.041214120 -1.082152129 0.039428584
##
                PC5
## 1
      -0.2412041809
## 2
      -1.1308962645
## 3
      -0.3415971627
## 4
      -0.2164379874
## 5
      -0.4127259666
var.test(PC3~data clean$V16,data=data pca1)
##
## F test to compare two variances
##
## data: PC3 by data_clean$V16
## F = 1.022, num df = 146, denom df = 454, p-value = 0.8534
## alternative hypothesis: true ratio of variances is not equal to 1
## 95 percent confidence interval:
## 0.7915999 1.3436023
## sample estimates:
## ratio of variances
##
            1.021978
```

```
#t.test(PC1~data_clean$V16, data=data_pca)
#t.test(PC2~data_clean$V16, data=data_pca)
t.test(PC3~data_clean$V16, data=data_pca1)

##
## Welch Two Sample t-test
##
## data: PC3 by data_clean$V16
## t = -0.065215, df = 245.03, p-value = 0.9481
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1945103 0.1820429
## sample estimates:
## mean in group 1 mean in group 5
## -0.004711502 0.001522178
```