

Stat 418 - Final project

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Libraray Packages

```
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 3.5.3
```

```
library(Hmisc)
```

```
## Warning: package 'Hmisc' was built under R version 3.5.3
```

```
## Loading required package: lattice
```

```
## Loading required package: survival
```

```
## Loading required package: Formula
```

```
##
```

```
## Attaching package: 'Hmisc'
```

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

```
##
```

```
##      format.pval, units
```

```
library(car)
```

```
## Warning: package 'car' was built under R version 3.5.3
```

```
## Loading required package: carData
```

```
library(xgboost)
```

```
## Warning: package 'xgboost' was built under R version 3.5.3
```

Read data

```
data <- read.csv("films.csv",na.strings = "null")
```

```
data <- data[,c(4:7,9,10)]
```

Genre - select the first one as the genre

```
for (i in 1:length(data[, 'Genre'])) {  
  a <- as.vector.factor(data[i, 'Genre'])  
  b <- unlist(strsplit(a, split=","))  
  data[i, 'Genre'] <- b[1]  
}
```

```
data$Genre[data$Genre=='Thriller'] = "Horror"
```

```
data$Genre[data$Genre=='Sport' | data$Genre=='Crime' | data$Genre=='Western'] = "Action"
```

```
data$Genre[data$Genre=="Family" | data$Genre=="Musical" | data$Genre=="Music" | data$Genre=="History"] = "Drama"
```

```
data$Genre[data$Genre=='Sci-Fi' | data$Genre=='Fantasy' | data$Genre=='War' | data$Genre=='Mystery'] = "Adventure"
```

```
data$Genre <- factor(data$Genre, levels = c("Action", "Adventure", "Animation", "Biography", "Comedy", "Drama", "Documentary", "Fantasy", "Family", "History", "Horror", "Mystery", "Musical", "Music", "Romance", "Sci-Fi", "Sport", "Thriller", "Western"))
```

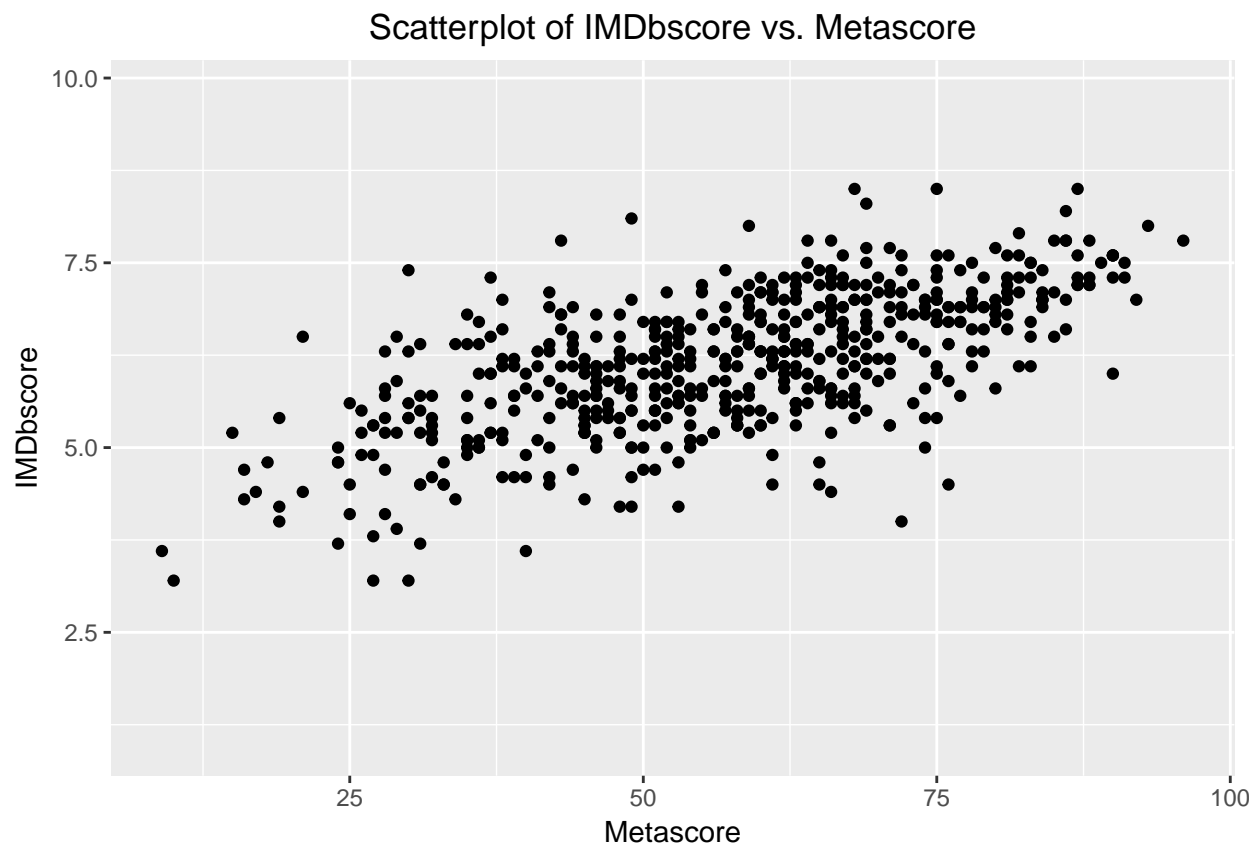
Run_time

```
Run_time <- data$Run_time
data$Run_time <- as.numeric(data$Run_time)
for (i in 1:length(data[, 'Run_time'])) {
  a <- as.vector(Run_time[i])
  b <- unlist(strsplit(a, split=" "))
  data[i, 'Run_time'] <- as.numeric(b[1])
}
```

Plot

```
ggplot(data, aes(Metascore, IMDBscore)) + geom_point() + labs(title="Scatterplot of IMDBscore vs. Metascore")
```

```
## Warning: Removed 4437 rows containing missing values (geom_point).
```



Fill in Missing value

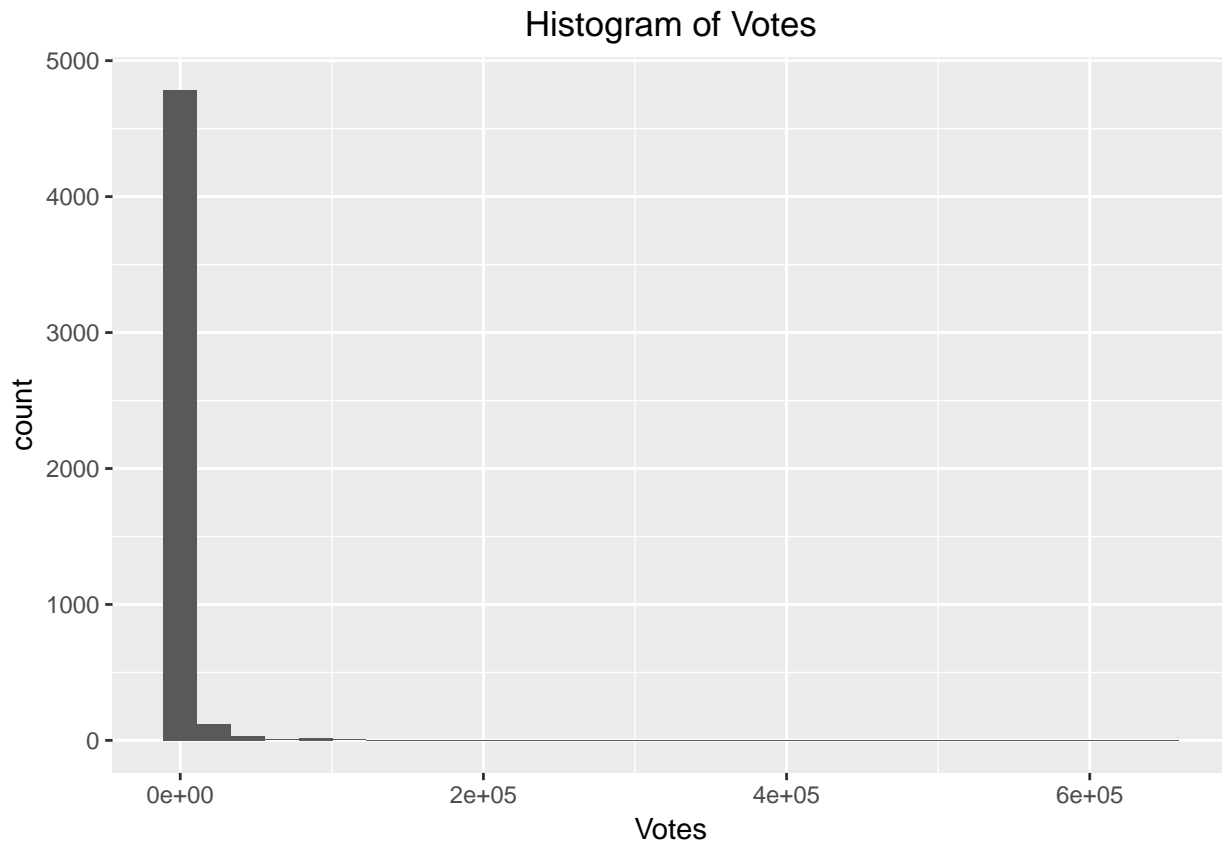
```
data$IMDBscore <- impute(data$IMDBscore, median)
data$Metascore <- impute(data$Metascore, median)
data$Votes <- impute(data$Votes, median)
data$Run_time <- impute(data$Run_time, median)
data$Genre <- impute(data$Genre, 'Drama')
```

```
data$Popularity_rank<-as.numeric(data$Popularity_rank)
data$IMDbScore<-as.numeric(data$IMDbScore)
data$Metascore<-as.numeric(data$Metascore)
data$Votes<-as.numeric(data$Votes)
data$Run_time<-as.numeric(data$Run_time)
data$Genre<- as.factor(data$Genre)
```

Transformation for Vote

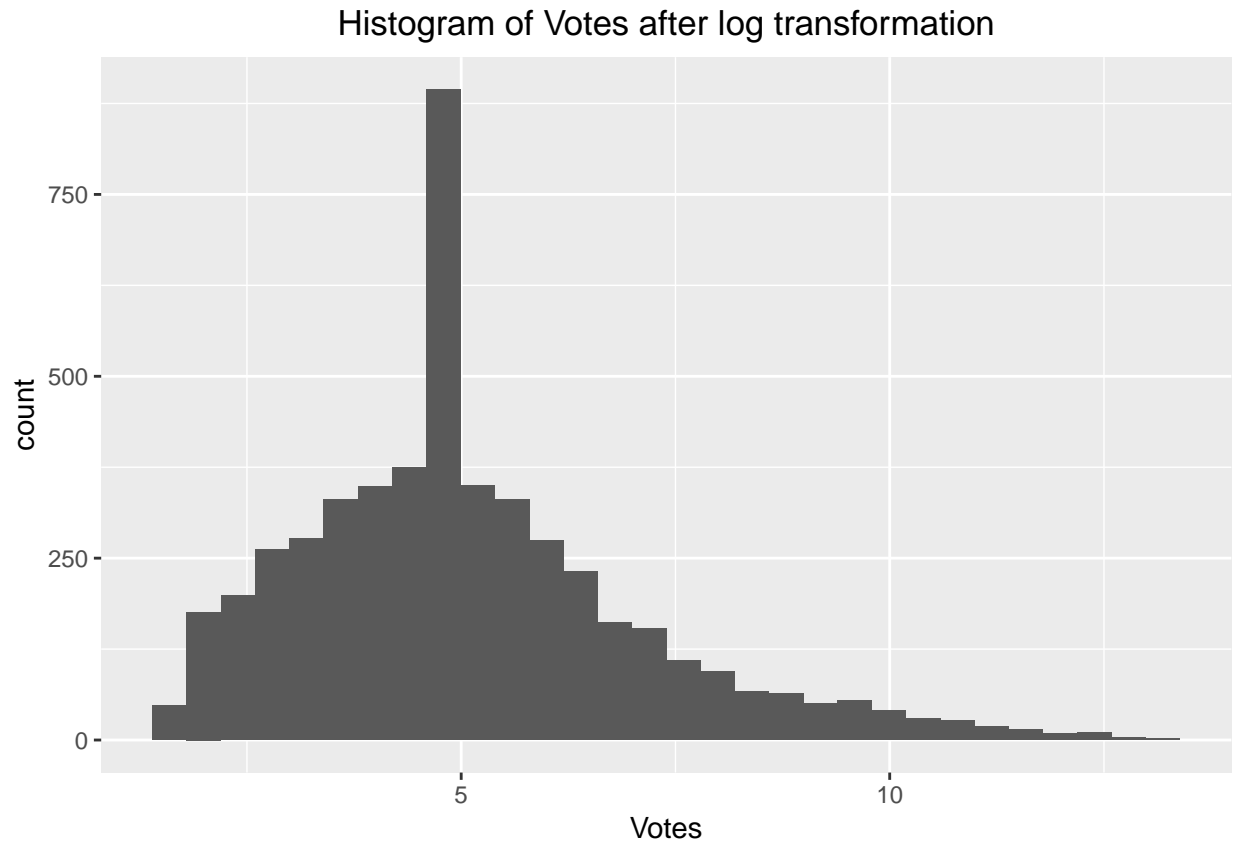
```
ggplot(data,aes(Votes))+geom_histogram()+ labs(title="Histogram of Votes")+ theme(plot.title = element_
```

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



```
d <- data
d$Votes = log(d$Votes+1)
ggplot(d,aes(Votes))+geom_histogram() + labs(title="Histogram of Votes after log transformation")+ theme
```

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



Split data

```
#set.seed(2019)
#index <- sample(1:5000,3500,replace = FALSE)
#data <- data[index,]
#test_data<- data[-index,]
```

Model - Linear Regression

```
reg <- lm(IMDbScore ~ Popularity_rank + Metascore + log(Votes) + Run_time + Genre, data = data)
summary(reg)
```

```
##
## Call:
## lm(formula = IMDbScore ~ Popularity_rank + Metascore + log(Votes) +
##     Run_time + Genre, data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.8620 -0.6727  0.0934  0.7895  4.0798
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   3.366e+00  2.230e-01  15.092  < 2e-16 ***
```

```
## Popularity_rank 3.750e-05 1.662e-05 2.256 0.024116 *
## Metascore      2.762e-02 3.212e-03 8.601 < 2e-16 ***
## log(Votes)     -5.127e-02 1.248e-02 -4.108 4.06e-05 ***
## Run_time       1.067e-02 8.888e-04 12.000 < 2e-16 ***
## GenreAdventure 2.290e-01 9.411e-02 2.433 0.014997 *
## GenreAnimation 3.995e-01 1.131e-01 3.531 0.000418 ***
## GenreBiography 9.251e-01 1.271e-01 7.277 3.94e-13 ***
## GenreComedy    1.959e-01 6.008e-02 3.261 0.001116 **
## GenreDrama     5.958e-01 5.696e-02 10.459 < 2e-16 ***
## GenreHorror    -4.002e-01 7.173e-02 -5.580 2.54e-08 ***
## GenreRomance   4.255e-01 1.390e-01 3.061 0.002220 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.303 on 4988 degrees of freedom
## Multiple R-squared:  0.1209, Adjusted R-squared:  0.119
## F-statistic: 62.36 on 11 and 4988 DF, p-value: < 2.2e-16
sprintf("The RMSE of the model is %f",sqrt(sum(residuals(reg)^2)/reg$df.residual))

## [1] "The RMSE of the model is 1.303423"
```

Model - Xgboost

```
data1<-data
data1$Run_time<- as.numeric(data1$Run_time)
data1$Genre <- as.numeric(data1$Genre)
x<-as.matrix(data1[,c(1,3:6)])
y<-as.matrix(data1$IMDbscore)
xgb<-xgboost(data = x, label = y, max.depth = 6,eta = 0.3, nthread = 2, verbose=2, nround = 10)

## [1] train-rmse:4.182815
## [2] train-rmse:3.061179
## [3] train-rmse:2.313911
## [4] train-rmse:1.829632
## [5] train-rmse:1.530007
## [6] train-rmse:1.348144
## [7] train-rmse:1.247435
## [8] train-rmse:1.187237
## [9] train-rmse:1.154672
## [10] train-rmse:1.136153

summary(xgb)

##           Length Class           Mode
## handle           1 xgb.Booster.handle externalptr
## raw            32406 -none-         raw
## niter           1 -none-         numeric
## evaluation_log    2 data.table      list
## call            16 -none-         call
## params           4 -none-         list
## callbacks         2 -none-         list
## feature_names     5 -none-         character
## nfeatures         1 -none-         numeric
```