

Movie Recommendation System

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Project Details.

This is a movie recommendation project made for **CodeClause**. Functions have been made instead of modelling for the predictions. The data set has been taken from *Kaggle*.

Loading the data and packages

```
library(tidyverse)
library(ggplot2)

data<- read.csv("C:/Users/91835/Desktop/CodeClause_data/movie_ratings.csv")
```

Exploring and cleaning the Data

Finding the mean of metascore and replacing *NA* with the mean value The mean is 74.27941, rounding off to 74

```
head(data)
```

```
##   ID          movie year imdb metascore  votes
## 1  0      Gladiator 2000  8.5         67 1187559
## 2  1      Memento  2000  8.5         80 1014199
## 3  2       Snatch  2000  8.3         74  707162
## 4  3 Requiem for a Dream 2000 8.3         68 688394
## 5  4         X-Men  2000  7.4         64  526411
## 6  5      Cast Away 2000  7.8         73  465125
```

```
dim(data)
```

```
## [1] 900  6
```

```
summary(data)
```

```
##           ID          movie          year          imdb
##  Min.      : 0.0   Length:900      Min.      :2000   Min.      :4.100
##  1st Qu.: 424.8   Class :character  1st Qu.:2004   1st Qu.:6.700
```

```
## Median : 874.5    Mode :character    Median :2008    Median :7.300
## Mean   : 874.5                    Mean   :2008    Mean   :7.239
## 3rd Qu.:1324.2                    3rd Qu.:2013    3rd Qu.:7.800
## Max.   :1749.0                    Max.   :2017    Max.   :9.900
##
##      metascore      votes
## Min.   : 61.00    Min.   : 93428
## 1st Qu.: 67.00    1st Qu.: 184223
## Median : 73.00    Median : 246739
## Mean   : 74.28    Mean   : 317636
## 3rd Qu.: 81.00    3rd Qu.: 382583
## Max.   :100.00    Max.   :2020298
## NA's   :424
```

```
work_data<- data %>%
  replace_na(list(metascore=74))
```

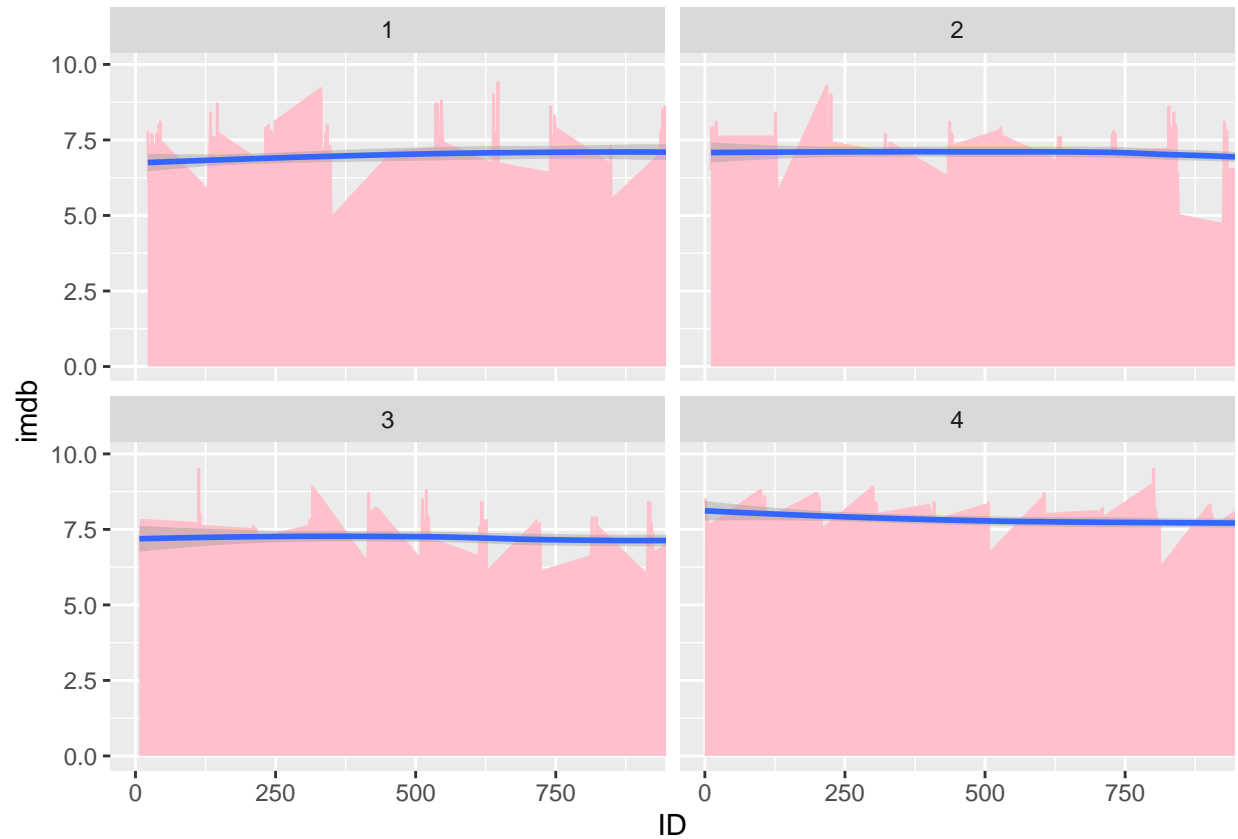
Binning the votes into 4 equal bins, 4 =Higher votes category, 1=Lower votes category.

```
work_data<-work_data %>%
  mutate(vote_category=ntile(votes,4))
```

Visualizing data

We observe that movies which are in vote category 4 tend to have a higher imdb rating.

```
ggplot(work_data, aes(x=ID, y=imdb))+
  geom_path(color="pink")+
  geom_area(fill="pink")+
  geom_smooth()+
  coord_cartesian(xlim = c(0,900))+
  facet_wrap(~vote_category)
```



Recommendations

Top rated movie functions

This function tells the top rated movies from the list. The criteria that it uses is : imdb rating ≥ 8.5 8.5 Metascore ≥ 85 Vote category 4 (Higher Votes)

```
topRated<- function(){
for(i in 1:nrow(work_data))
{
  if(work_data$imdb[i]>=8.5 && work_data$metascore[i]>=85 && work_data$vote_category[i]==4)
  {
    print(work_data[i, "movie"])
  }
}
}
#Calling topRated function
topRated()
```

```
## [1] "The Lord of the Rings: The Fellowship of the Ring"
## [1] "Sen to Chihiro no kamikakushi"
## [1] "The Lord of the Rings: The Two Towers"
## [1] "The Pianist"
## [1] "The Lord of the Rings: The Return of the King"
```

```
## [1] "The Departed"
## [1] "Whiplash"
```

Recommendations by ID

This function will take in an argument which will be a numeric. The argument signifies the ID for whom the recommendations are to be generated. The way it works is that it reads the movie that has been seen by the user. It considers its metascore and imdb ratings and recommends movies in that same range. The range has been set as: imdb to imdb+0.5 metascore to metascore+7 The function will consider the movie present at the ID as that user's preference and generate recommendations in that range only.

```
recommend<- function(x)
{
  for(i in 1:nrow(work_data))
  {
    if(work_data$imdb[i]>=(work_data$imdb[x]) && work_data$imdb[i]<(work_data$imdb[x]+0.5))
    {
      if(work_data$metascore[i]>=work_data$metascore[x] && work_data$metascore[i]<(work_data$metascore[x]+7))
      {
        print(work_data[i, "movie"])
      }
    }
  }
}

#Calling the function
recommend(2)
```

```
## [1] "Memento"
## [1] "The Pianist"
## [1] "The Departed"
```