

MOVIE

ANWAR ABU ALHUSSEIN

Required packages

HEART FAILURE CLINICAL

PROJECT LINEAR REGRESSION

```
###* Loading Packages ###* -----
```

```
suppressPackageStartupMessages(library(tidyverse))
suppressPackageStartupMessages(library(caret))
suppressPackageStartupMessages(library(ggcorrplot))
suppressPackageStartupMessages(library(Metrics))
suppressPackageStartupMessages(library(corrplot))
suppressPackageStartupMessages(library(bannerCommenter))
suppressPackageStartupMessages(library(psych))
suppressPackageStartupMessages(library(WVPlots))
suppressPackageStartupMessages(library(PerformanceAnalytics))
suppressPackageStartupMessages(library(car))
suppressPackageStartupMessages(library(dplyr))
suppressPackageStartupMessages(library(ggplot2))
suppressPackageStartupMessages(library(outliers))
suppressPackageStartupMessages(library(broom))
suppressPackageStartupMessages(library(equationomatic))
```

```
##=====
## Reading in the data == ##=====
```

```
download.file("https://raw.githubusercontent.com/ABUALHUSSEIN/test/main/data/movie.csv",
  destfile = "movie.csv")
```

```
##source(destfile) ##ls() #=====
```

```
setwd("C:/Users/WAFA/Desktop")
movie <- read.csv("movie.csv",header=TRUE)
View(movie)
```

Missing Data. We are looking for the missing values

```
is.na(movie)
```

```
##      Movie  Year Sequel Sentiment Genre Ratings Gross Budget Screens Views
##   [1,] FALSE FALSE  FALSE      FALSE FALSE      FALSE FALSE  FALSE  FALSE FALSE
##   [2,] FALSE FALSE  FALSE      FALSE FALSE      FALSE FALSE  FALSE  FALSE FALSE
##   [3,] FALSE FALSE  FALSE      FALSE FALSE      FALSE FALSE  FALSE  FALSE FALSE
```

[illegible]

[illegible]

[illegible]

[illegible]

##	[220,]	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
##	[221,]	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
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##	[223,]	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
##	[224,]	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
##	[225,]	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
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##	[227,]	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
##	[228,]	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
##	[229,]	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
##	[230,]	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	TRUE	FALSE
##	[231,]	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE
##	[232,]	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE	TRUE
##		Likes	Dislikes	Comments	Aggregate	Followers						
##	[1,]	FALSE	FALSE	FALSE		FALSE						
##	[2,]	FALSE	FALSE	FALSE		FALSE						
##	[3,]	FALSE	FALSE	FALSE		FALSE						
##	[4,]	FALSE	FALSE	FALSE		FALSE						
##	[5,]	FALSE	FALSE	FALSE		FALSE						
##	[6,]	FALSE	FALSE	FALSE		FALSE						
##	[7,]	FALSE	FALSE	FALSE		FALSE						
##	[8,]	FALSE	FALSE	FALSE		FALSE						
##	[9,]	FALSE	FALSE	FALSE		FALSE						
##	[10,]	FALSE	FALSE	FALSE		FALSE						
##	[11,]	FALSE	FALSE	FALSE		FALSE						
##	[12,]	FALSE	FALSE	FALSE		FALSE						
##	[13,]	FALSE	FALSE	FALSE		FALSE						
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##	[165,]	FALSE	FALSE	FALSE	FALSE
##	[166,]	FALSE	FALSE	FALSE	FALSE
##	[167,]	FALSE	FALSE	FALSE	FALSE
##	[168,]	FALSE	FALSE	FALSE	FALSE
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##	[172,]	FALSE	FALSE	FALSE	FALSE
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##	[175,]	FALSE	FALSE	FALSE	FALSE
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##	[192,]	FALSE	FALSE	FALSE	FALSE
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##	[194,]	FALSE	FALSE	FALSE	FALSE
##	[195,]	FALSE	FALSE	FALSE	FALSE
##	[196,]	FALSE	FALSE	FALSE	FALSE
##	[197,]	FALSE	FALSE	FALSE	FALSE
##	[198,]	FALSE	FALSE	FALSE	FALSE
##	[199,]	FALSE	FALSE	FALSE	FALSE
##	[200,]	FALSE	FALSE	FALSE	FALSE
##	[201,]	FALSE	FALSE	FALSE	FALSE
##	[202,]	FALSE	FALSE	FALSE	FALSE

```
## [203,] FALSE FALSE FALSE FALSE
## [204,] FALSE FALSE FALSE FALSE
## [205,] FALSE FALSE FALSE FALSE
## [206,] FALSE FALSE FALSE FALSE
## [207,] FALSE FALSE FALSE FALSE
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## [209,] FALSE FALSE FALSE FALSE
## [210,] FALSE FALSE FALSE FALSE
## [211,] FALSE FALSE FALSE FALSE
## [212,] FALSE FALSE FALSE FALSE
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## [222,] FALSE FALSE FALSE TRUE
## [223,] FALSE FALSE FALSE TRUE
## [224,] FALSE FALSE FALSE TRUE
## [225,] FALSE FALSE FALSE TRUE
## [226,] FALSE FALSE FALSE TRUE
## [227,] FALSE FALSE FALSE TRUE
## [228,] FALSE FALSE FALSE TRUE
## [229,] FALSE FALSE FALSE TRUE
## [230,] FALSE FALSE FALSE TRUE
## [231,] FALSE FALSE FALSE TRUE
## [232,] TRUE TRUE TRUE TRUE
```

```
sum(is.na(movie))
```

```
## [1] 59
```

Check the column names again,

```
colnames(movie)
```

```
## [1] "Movie" "Year" "Sequel"
## [4] "Sentiment" "Genre" "Ratings"
## [7] "Gross" "Budget" "Screens"
## [10] "Views" "Likes" "Dislikes"
## [13] "Comments" "Aggregate.Followers"
```

The names contain dots, I will substitute
each dot with underscore " _ "

```
var_names <- gsub("\\.", "_", colnames(movie))
var_names[14] <- "Aggregate_Followers"
var_names[14]
```

```
## [1] "Aggregate_Followers"
colnames(movie) <- var_names
colnames(movie)

## [1] "Movie"          "Year"           "Sequel"
## [4] "Sentiment"      "Genre"          "Ratings"
## [7] "Gross"          "Budget"         "Screens"
## [10] "Views"          "Likes"          "Dislikes"
## [13] "Comments"       "Aggregate_Followers"
View(movie)
```

Gives the name of columns that do not have data

```
list_na <- colnames(movie)[ apply(movie, 2, anyNA) ]
list_na
```

```
## [1] "Year"          "Sequel"        "Sentiment"
## [4] "Genre"         "Ratings"       "Gross"
## [7] "Budget"        "Screens"       "Views"
## [10] "Likes"         "Dislikes"      "Comments"
## [13] "Aggregate_Followers"
```

```
##=====
## Now we need to compute of the mean with the argument na.rm = TRUE. #This argument is compulsory
because the columns have missing data, # and this tells R to ignore them.
##=====
```

Create mean

```
average_missing <- apply(movie[,colnames(movie) %in% list_na],
                          2,
                          mean,
                          na.rm = TRUE)
average_missing
```

```
##          Year          Sequel          Sentiment          Genre
## 2.014294e+03  1.359307e+00  2.809524e+00  5.359307e+00
## Ratings      Gross      Budget      Screens
## 6.441558e+00  6.806603e+07  4.792173e+07  2.209244e+03
## Views      Likes      Dislikes      Comments
## 3.712851e+06  1.273254e+04  6.790519e+02  1.825701e+03
## Aggregate_Followers
## 3.038193e+06
```

Replace the NA Values

```
##=====
##The verb mutate from the dplyr library is useful in creating a new variable. #We don't necessarily want
to change the original column so we can create a new variable without the NA. #mutate is easy to use, # we
just choose a variable name and define how to create this variable
##=====
```

```

movie_replace <- movie %>%
  mutate(replace_mean_Year = ifelse(is.na(Year), average_missing[1], Year),
         replace_mean_Sequel = ifelse(is.na(Sequel), average_missing[2], Sequel),
         replace_mean_Sentiment = ifelse(is.na(Sentiment), average_missing[3], Sentiment),
         replace_mean_Genre = ifelse(is.na(Genre), average_missing[4], Genre),
         replace_mean_Ratings = ifelse(is.na(Ratings), average_missing[5], Ratings),
         replace_mean_Gross = ifelse(is.na(Gross), average_missing[6], Gross),
         replace_mean_Budget = ifelse(is.na(Budget), average_missing[7], Budget),
         replace_mean_Screens = ifelse(is.na(Screens), average_missing[8], Screens),
         replace_mean_Views = ifelse(is.na(Views), average_missing[9], Views),
         replace_mean_Likes = ifelse(is.na(Likes), average_missing[10], Likes),
         replace_mean_Dislikes = ifelse(is.na(Dislikes), average_missing[11], Dislikes),
         replace_mean_Comments = ifelse(is.na(Comments), average_missing[12], Comments),
         replace_mean_Aggregate_Followers = ifelse(is.na(Aggregate_Followers),
                                                    average_missing[13], Aggregate_Followers))

```

Perform the replacement

```
sum(is.na(movie_replace$Year))
```

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

```
sum(is.na(movie_replace$Sequel))
```

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

```
sum(is.na(movie_replace$Sentiment))
```

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

```
sum(is.na(movie_replace$Genre))
```

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

```
sum(is.na(movie_replace$Ratings))
```

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

```
sum(is.na(movie_replace$Gross))
```

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

```
sum(is.na(movie_replace$ Budget))
```

```
## [1] 2
```

```
sum(is.na(movie_replace$ Screens))
```

```
## [1] 11
```

```
sum(is.na(movie_replace$ Views))
```

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

```
sum(is.na(movie_replace$Likes))
```

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

```
sum(is.na(movie_replace$Dislikes))

## [1] 1

sum(is.na(movie_replace$Comments))

## [1] 1

sum(is.na(movie_replace$Aggregate_Followers))

## [1] 36

sum(is.na(movie_replace$replace_mean_Aggregate_Followers))

## [1] 0
```

My new Data.

Here I created a new frame for the data that do not contain na values

Select all columns that do not contain na values from a data by names ———

```
new_movie<- movie_replace %>%
  select(replace_mean_Year, replace_mean_Sequel,replace_mean_Sentiment,replace_mean_Genre,replace_mean_
    replace_mean_Gross,replace_mean_Budget,replace_mean_Screens,replace_mean_Views,
    replace_mean_Likes,replace_mean_Dislikes,replace_mean_Comments,replace_mean_Aggregate_Followers)
view(new_movie)
```

Table 1: new_movie

	replace_mean_Year	replace_mean_Sequel	replace_mean_Sentiment	replace_mean_Genre	replace_mean_Gross	replace_mean_Budget	replace_mean_Screens	replace_mean_Views	replace_mean_Likes	replace_mean_Dislikes	replace_mean_Comments	replace_mean_Aggregate_Followers
2014	1	0	8	6.3	9.13e+03	4000000	45.000	3280543	4632	425	636	1120000
2014	2	2	1	7.1	1.92e+05	50000000	3306.000	583289	3465	61	186	12350000
2014	1	0	1	6.2	3.07e+02	8000000	2872.000	304861	328	34	47	483000
2014	2	0	1	6.3	1.06e+08	10000000	3470.000	452917	2429	132	590	568000
2014	2	0	8	4.7	1.73e+03	5000000	2310.000	3145573	12163	610	1082	1923800
2014	1	0	3	4.6	2.90e+04	500000	2209.244	91137	112	7	1	310000
2014	1	0	8	6.1	4.26e+07	40000000	3158.000	3013011	9595	419	1020	8153000
2014	1	2	1	7.1	5.75e+02	20000000	318.000	1854103	2207	197	593	130655
2014	1	3	10	6.5	2.60e+02	8000000	2714.000	2213659	2210	419	382	125646
2014	1	0	8	6.1	4.86e+07	25000000	2253.000	5218079	11709	532	770	21697300
2014	1	4	1	7.3	3.50e+03	8800000	3555.000	3927600	13143	573	3134	24300
2014	1	0	8	5.7	1.52e+03	30000000	762.000	519327	963	94	70	386400
2014	1	0	15	5.4	8.43e+05	5000000	3185.000	1903290	28810	4382	4392	19420105
2014	1	0	8	5.2	8.59e+05	50000000	3116.000	330006	5150	707	1484	5130800
2014	3	0	3	4.4	8.30e+05	5000000	65.000	595194	85	36	39	15112
2014	1	2	8	6.6	1.18e+04	6000000	18.000	3915978	6983	247	460	253000
2014	1	0	8	6.3	7.23e+04	5000000	25.000	1391527	2479	146	182	1658900
2014	1	0	3	6.9	1.46e+07	7000000	31.000	1828235	7633	235	685	116100
2014	1	29	12	7.9	2.22e+08	65000000	3761.000	4700023	14163	538	1293	199800
2014	1	-1	3	6.6	2.16e+07	7000000	1823.000	1348142	4404	307	1033	888000
2014	1	-2	8	6.5	4.63e+07	40000000	3555.000	7977747	18690	1940	2214	2417000

[illegible]

## 4	300: Rise of an Empire	2014	2
## 5	A Haunted House 2	2014	2
## 6	A Long Way Off	2014	1
## 7	A Million Ways to Die in the West	2014	1
## 8	A Most Violent Year	2014	1
## 9	A Walk Among the Tombstones	2014	1
## 10	About Last Night	2014	1
## 11	American Sniper	2014	1
## 12	And So It Goes	2014	1
## 13	Annabelle	2014	1
## 14	Annie	2014	1
## 15	Atlas Shrugged: Who Is John Galt?	2014	3
## 16	Barefoot	2014	1
## 17	Better Living Through Chemistry	2014	1
## 18	Beyond the Lights	2014	1
## 19	Big Hero 6	2014	1
## 20	Black or White	2014	1
## 21	Blended	2014	1
## 22	Boyhood	2014	1
## 23	Brick Mansions	2014	1
## 24	Cake	2014	1
## 25	Camp X-Ray	2014	1
## 26	Cantinflas	2014	1
## 27	Captain America: The Winter Soldier	2014	2
## 28	Dawn of the Planet of the Apes	2014	2
## 29	Deliver Us from Evil	2014	1
## 30	Devil's Due	2014	1
## 31	Divergent	2014	1
## 32	Dolphin Tale 2	2014	2
## 33	Dracula Untold	2014	1
## 34	Draft Day	2014	1
## 35	Dumb and Dumber To	2014	3
## 36	Earth to Echo	2014	1
## 37	Edge of Tomorrow	2014	1
## 38	Endless Love	2014	1
## 39	Falcon Rising	2014	1
## 40	Foxcatcher	2014	1
## 41	Fury	2014	1
## 42	God Help the Girl	2014	1
## 43	God's Not Dead	2014	1
## 44	God's Pocket	2014	1
## 45	Godzilla	2014	1
## 46	Gone Girl	2014	1
## 47	Guardians of the Galaxy	2014	1
## 48	Happy Christmas	2014	1
## 49	Heaven Is for Real	2014	1
## 50	Hercules	2014	1
## 51	Horrible Bosses 2	2014	2
## 52	How to Train Your Dragon 2	2014	2
## 53	If I Stay	2014	1
## 54	In the Name of My Daughter	2014	1
## 55	Inherent Vice	2014	1
## 56	Interstellar	2014	1
## 57	Into the Storm	2014	1

## 58	Into the Woods	2014	1
## 59	It Follows	2014	1
## 60	Jack Ryan: Shadow Recruit	2014	5
## 61	Jersey Boys	2014	1
## 62	John Wick	2014	1
## 63	Kill the Messenger	2014	1
## 64	Kingsman: The Secret Service	2014	1
## 65	Kung Fu Jungle	2014	1
## 66	Left Behind	2014	1
## 67	Let's Be Cops	2014	1
## 68	Life After Beth	2014	1
## 69	Locker 13	2014	1
## 70	Lucy	2014	1
## 71	Maleficent	2014	1
## 72	Maps to the Stars	2014	1
## 73	Million Dollar Arm	2014	1
## 74	Moms' Night Out	2014	1
## 75	Mr. Peabody & Sherman	2014	1
## 76	Muppets Most Wanted	2014	3
## 77	My Old Lady	2014	1
## 78	Need for Speed	2014	1
## 79	Neighbors	2014	1
## 80	Night at the Museum: Secret of the Tomb	2014	3
## 81	Nightcrawler	2014	1
## 82	No Good Deed	2014	1
## 83	Noah	2014	1
## 84	Non-Stop	2014	1
## 85	Not Cool	2014	1
## 86	Ouija	2014	1
## 87	Paddington	2014	1
## 88	Paranormal Activity: The Marked Ones	2014	1
## 89	Penguins of Madagascar	2014	1
## 90	Planes: Fire & Rescue	2014	3
## 91	Pompeii	2014	1
## 92	Ride Along	2014	1
## 93	Rio 2	2014	2
## 94	RoboCop	2014	1
## 95	Rosewater	2014	1
## 96	Rudderless	2014	1
## 97	Sabotage	2014	1
## 98	Selma	2014	1
## 99	Serena	2014	1
## 100	Seventh Son	2014	1
## 101	Sex Tape	2014	1
## 102	Sin City: A Dame to Kill For	2014	2
## 103	Son of God	2014	1
## 104	Song One	2014	1
## 105	St. Vincent	2014	1
## 106	Taken 3	2014	3
## 107	Tammy	2014	1
## 108	Teenage Mutant Ninja Turtles	2014	1
## 109	That Awkward Moment	2014	1
## 110	The Admiral	2014	1
## 111	The Amazing Spider-Man 2	2014	2

## 112	The Babadook	2014	1
## 113	The Best of Me	2014	1
## 114	The Book of Life	2014	1
## 115	The Boxtrolls	2014	1
## 116	The Devil's Hand	2014	1
## 117	The Drop	2014	1
## 118	The Equalizer	2014	1
## 119	The Expendables 3	2014	3
## 120	The Fault in Our Stars	2014	1
## 121	The Fluffy Movie: Unity Through Laughter	2014	1
## 122	The Gambler	2014	1
## 123	The Giver	2014	1
## 124	The Good Lie	2014	1
## 125	The Hobbit: The Battle of the Five Armies	2014	6
## 126	The Homesman	2014	1
## 127	The Hundred-Foot Journey	2014	1
## 128	The Hunger Games: Mockingjay - Part 1	2014	3
## 129	The Identical	2014	1
## 130	The Imitation Game	2014	1
## 131	The Interview	2014	1
## 132	The Judge	2014	1
## 133	The Legend of Hercules	2014	1
## 134	The Lego Movie	2014	1
## 135	The Loft	2014	1
## 136	The Maze Runner	2014	1
## 137	The Monuments Men	2014	1
## 138	The November Man	2014	1
## 139	The Nut Job	2014	1
## 140	The One I Love	2014	1
## 141	The Other Woman	2014	1
## 142	The Purge: Anarchy	2014	2
## 143	The Pyramid	2014	1
## 144	The Rover	2014	1
## 145	The Salvation	2014	1
## 146	The Theory of Everything	2014	1
## 147	The Water Diviner	2014	1
## 148	The Woman in Black 2: Angel of Death	2014	2
## 149	Think Like a Man Too	2014	2
## 150	Transcendence	2014	1
## 151	Transformers: Age of Extinction	2014	4
## 152	Tusk	2014	1
## 153	Unbroken	2014	1
## 154	Veronica Mars	2014	1
## 155	When the Game Stands Tall	2014	1
## 156	Whiplash	2014	1
## 157	White God	2014	1
## 158	Wild	2014	1
## 159	Wild Tales	2014	1
## 160	Winter's Tale	2014	1
## 161	Wish I Was Here	2014	1
## 162	X-Men: Days of Future Past	2014	7
## 163	Yves Saint Laurent	2014	1
## 164	Jurassic World	2015	4
## 165	Avengers: Age of Ultron	2015	2

## 166	Furious 7	2015	7
## 167	Inside Out	2015	1
## 168	Minions	2015	2
## 169	Pitch Perfect 2	2015	2
## 170	Mission: Impossible - Rogue Nation	2015	5
## 171	Home	2015	1
## 172	Fifty Shades of Grey	2015	1
## 173	The SpongeBob Movie: Sponge Out of Water	2015	2
## 174	San Andreas	2015	1
## 175	Mad Max: Fury Road	2015	4
## 176	Straight Outta Compton	2015	1
## 177	Insurgent	2015	2
## 178	Spy	2015	1
## 179	Trainwreck	2015	1
## 180	Tomorrowland	2015	1
## 181	Get Hard	2015	1
## 182	Terminator Genisys	2015	5
## 183	Ted 2	2015	2
## 184	Paul Blart: Mall Cop 2	2015	2
## 185	Pixels	2015	1
## 186	Magic Mike XXL	2015	2
## 187	The Wedding Ringer	2015	1
## 188	Vacation	2015	5
## 189	Fantastic Four	2015	1
## 190	Poltergeist	2015	1
## 191	Jupiter Ascending	2015	1
## 192	The Age of Adaline	2015	1
## 193	Max	2015	1
## 194	The Longest Ride	2015	1
## 195	The Gift	2015	1
## 196	The Boy Next Door	2015	1
## 197	Hot Pursuit	2015	1
## 198	The DUFF	2015	1
## 199	Woman in Gold	2015	1
## 200	Entourage	2015	1
## 201	Paper Towns	2015	1
## 202	The Lazarus Effect	2015	1
## 203	Sinister 2	2015	2
## 204	Ricki and the Flash	2015	1
## 205	Project Almanac	2015	1
## 206	Hitman: Agent 47	2015	1
## 207	Dope	2015	1
## 208	Hot Tub Time Machine 2	2015	2
## 209	American Ultra	2015	1
## 210	The Gunman	2015	1
## 211	Mortdecai	2015	1
## 212	Blackhat	2015	1
## 213	Me and Earl and the Dying Girl	2015	1
## 214	The Vatican Tapes	2015	1
## 215	Maggie	2015	1
## 216	Americans	2015	1
## 217	Road Hard	2015	1
## 218	Ant-Man	2015	1
## 219	Southpaw	2015	1

## 220	Ex Machina	2015	1
## 221	The Man from U.N.C.L.E.	2015	1
## 222	Cinderella	2015	1
## 223	Chappie	2015	1
## 224	Brotherly Love	2015	1
## 225	Run All Night	2015	1
## 226	The Second Best Exotic Marigold Hotel	2015	2
## 227	Child 44	2015	1
## 228	Aloha	2015	1
## 229	Unfinished Business	2015	1
## 230	War Room	2015	1
## 231	The Gallows	2015	1
## 232		2014.29437229437	1.35930735930736

##	Sentiment	Genre	Ratings	Gross
## 1	0	8	6.3	9130
## 2	2	1	7.1	1.92e+08
## 3	0	1	6.2	30700000
## 4	0	1	6.3	1.06e+08
## 5	0	8	4.7	17300000
## 6	0	3	4.6	29000
## 7	0	8	6.1	42600000
## 8	2	1	7.1	5750000
## 9	3	10	6.5	2.6e+07
## 10	0	8	6.1	48600000
## 11	4	1	7.3	3.5e+08
## 12	0	8	5.7	15200000
## 13	0	15	5.4	84300000
## 14	0	8	5.2	85900000
## 15	0	3	4.4	830000
## 16	2	8	6.6	11800
## 17	0	8	6.3	72300
## 18	0	3	6.9	14600000
## 19	29	12	7.9	2.22e+08
## 20	-1	3	6.6	21600000
## 21	-2	8	6.5	46300000
## 22	9	3	8	25400000
## 23	0	1	5.7	20300000
## 24	2	3	6.5	1870000
## 25	3	3	7	9840
## 26	3	9	6.7	6370000
## 27	5	1	7.8	2.6e+08
## 28	-4	1	7.7	2.09e+08
## 29	0	10	6.2	30500000
## 30	0	15	4	15800000
## 31	-4	2	6.8	1.51e+08
## 32	-1	3	6.4	4.2e+07
## 33	0	1	6.3	55900000
## 34	0	3	6.8	28800000
## 35	0	8	5.8	86200000
## 36	0	2	5.8	38900000
## 37	0	1	7.9	1e+08
## 38	1	3	6.3	23400000
## 39	0	1	5.7	8690
## 40	6	9	7.1	1210000

## 41	3	1	7.6	85700000
## 42	-4	3	6.4	102000
## 43	0	3	5	60800000
## 44	0	10	6.1	104000
## 45	2	1	6.5	2.01e+08
## 46	-11	3	8.2	1.68e+08
## 47	0	1	8.1	3.33e+08
## 48	0	8	5.6	30100
## 49	-9	3	5.8	91400000
## 50	-2	1	6.1	72700000
## 51	-1	8	6.3	54400000
## 52	3	12	7.9	1.77e+08
## 53	14	3	6.8	50500000
## 54	0	3	6.1	275000
## 55	1	8	6.7	8090000
## 56	2	2	8.7	1.88e+08
## 57	0	1	5.8	47600000
## 58	-4	2	6	1.28e+08
## 59	-9	15	6.9	14700000
## 60	4	1	6.2	50500000
## 61	10	9	6.9	4.7e+07
## 62	0	1	7.2	4.3e+07
## 63	0	9	7	2450000
## 64	0	1	7.8	1.28e+08
## 65	-2	1	6.5	129000
## 66	0	1	3.1	1.4e+07
## 67	-5	8	6.5	82400000
## 68	0	8	5.7	8300
## 69	0	7	4.8	2470
## 70	1	1	6.4	1.27e+08
## 71	0	1	7	2.41e+08
## 72	0	8	6.3	348000
## 73	27	9	7.1	36400000
## 74	0	8	5.4	10400000
## 75	13	12	6.9	1.12e+08
## 76	-1	2	6.5	51200000
## 77	0	8	6.3	4010000
## 78	5	1	6.6	43600000
## 79	2	8	6.4	1.5e+08
## 80	2	2	6.3	1.14e+08
## 81	-1	10	7.9	32300000
## 82	0	10	5.6	52500000
## 83	6	1	5.9	1.01e+08
## 84	0	1	7	91400000
## 85	0	8	5.2	35700
## 86	-4	15	4.4	50800000
## 87	14	8	7.2	76100000
## 88	0	15	5	32500000
## 89	8	12	6.8	83300000
## 90	0	12	5.9	59200000
## 91	0	1	5.6	23200000
## 92	0	1	6.1	1.34e+08
## 93	4	12	6.4	1.32e+08
## 94	7	1	6.2	58600000

## 95	1	9	6.6	3090000
## 96	0	8	7.5	37400
## 97	17	1	5.7	10500000
## 98	-5	9	7.5	52100000
## 99	0	3	5.4	176000
## 100	0	1	5.5	17200000
## 101	0	8	5.1	38500000
## 102	0	10	6.6	13800000
## 103	2	9	5.5	59700000
## 104	0	3	5.8	20200
## 105	20	8	7.3	44100000
## 106	6	1	6.1	89300000
## 107	0	8	4.8	84500000
## 108	0	1	5.9	1.91e+08
## 109	0	8	6.1	2.6e+07
## 110	3	2	7.1	2590000
## 111	6	1	6.8	2.03e+08
## 112	-3	3	6.9	924000
## 113	3	3	6.6	26800000
## 114	9	12	7.3	50200000
## 115	5	12	6.8	50800000
## 116	0	7	4.8	4240
## 117	-1	10	7.1	10700000
## 118	0	1	7.2	1.02e+08
## 119	0	1	6.1	39300000
## 120	7	3	7.9	1.25e+08
## 121	0	8	7	2820000
## 122	-2	10	6.1	33600000
## 123	6	3	6.5	45100000
## 124	10	3	7.4	2720000
## 125	3	2	7.5	2.55e+08
## 126	2	3	6.6	2430000
## 127	11	8	7.3	54200000
## 128	2	2	6.8	3.37e+08
## 129	0	3	5	2820000
## 130	9	9	8.1	91100000
## 131	0	1	6.7	6110000
## 132	2	10	7.5	47100000
## 133	0	1	4.2	18800000
## 134	-2	12	7.8	2.58e+08
## 135	2	6	6.3	5980000
## 136	2	1	6.8	1.02e+08
## 137	0	3	6.1	7.8e+07
## 138	0	1	6.3	2.5e+07
## 139	0	12	5.8	64200000
## 140	2	8	7.1	512000
## 141	-1	8	6	83900000
## 142	-3	1	6.5	71500000
## 143	0	15	4.7	2750000
## 144	0	10	6.4	1110000
## 145	0	3	6.8	5000
## 146	8	9	7.8	35900000
## 147	7	3	7.2	4190000
## 148	-6	3	4.9	26500000

## 149	0	8	5.7	65200000
## 150	0	3	6.3	2.3e+07
## 151	0	1	5.8	2.45e+08
## 152	0	8	5.4	1820000
## 153	0	9	7.2	1.16e+08
## 154	10	10	6.8	3320000
## 155	0	3	6.6	30100000
## 156	2	3	8.6	13100000
## 157	11	3	6.9	280000
## 158	0	2	7.2	37900000
## 159	7	8	8.2	3080000
## 160	0	3	6.2	22500
## 161	0	8	6.7	3590000
## 162	3	1	8.1	2.34e+08
## 163	0	9	6.2	717000
## 164	1	1	7.3	6.43e+08
## 165	21	1	7.8	4.58e+08
## 166	-17	1	7.4	3.5e+08
## 167	2	12	8.6	3.45e+08
## 168	-2	1	6.6	3.25e+08
## 169	1	8	6.7	1.83e+08
## 170	3	3	7.8	1.71e+08
## 171	-2	8	6.7	1.77e+08
## 172	9	3	4.2	1.66e+08
## 173	0	8	6.1	1.62e+08
## 174	3	1	6.4	1.54e+08
## 175	-4	1	8.3	1.53e+08
## 176	5	9	8.3	1.35e+08
## 177	-6	2	6.4	1.3e+08
## 178	0	1	7.3	1.1e+08
## 179	1	12	6.7	1.05e+08
## 180	1	12	6.6	93200000
## 181	6	8	6.1	90400000
## 182	1	1	6.8	89400000
## 183	5	8	6.6	81300000
## 184	15	1	4.3	7.1e+07
## 185	10	1	5.6	7.1e+07
## 186	7	8	6.2	65900000
## 187	-8	8	6.7	64500000
## 188	13	2	6.3	54900000
## 189	20	1	4	52700000
## 190	-5	15	5	47400000
## 191	0	1	5.5	47400000
## 192	11	3	7.3	42500000
## 193	0	1	7	41900000
## 194	19	3	7.2	37400000
## 195	11	1	7.6	36100000
## 196	16	6	4.6	35400000
## 197	2	1	5	34500000
## 198	2	8	6.6	3.4e+07
## 199	19	10	7.4	33300000
## 200	6	3	7.1	32400000
## 201	6	3	6.9	31400000
## 202	7	8	5.2	25800000

## 203	-38	1	5.5	1.9e+07
## 204	25	3	6.1	23500000
## 205	7	8	6.4	22300000
## 206	10	8	5.9	16100000
## 207	2	8	7.5	16800000
## 208	9	8	5.1	12300000
## 209	21	1	6.5	10900000
## 210	3	8	5.8	10600000
## 211	-2	8	5.5	7610000
## 212	4	3	5.4	7100000
## 213	15	8	8.2	6740000
## 214	-2	15	4.5	1710000
## 215	10	3	5.6	131000
## 216	0	3	7.3	129000
## 217	14	3	6.3	106000
## 218	18	1	7.8	1.69e+08
## 219	6	1	7.7	49500000
## 220	3	3	7.7	25400000
## 221	8	6	7.6	34600000
## 222	2	8	7.1	2.01e+08
## 223	8	1	7	31600000
## 224	0	8	6.9	444000
## 225	3	8	6.6	26400000
## 226	26	1	6.6	33100000
## 227	4	4	6.4	1210000
## 228	13	15	5.5	2.1e+07
## 229	7	8	5.4	10200000
## 230	10	1	5.4	12300000
## 231	-5	15	4.4	22600000
## 232	2.80952380952381	5.35930735930736	6.44155844155844	68066033.2034632
##	Budget	Screens	Views	Likes
## 1	4e+06	45	3280543	4632
## 2	5e+07	3306	583289	3465
## 3	2.8e+07	2872	304861	328
## 4	1.1e+08	3470	452917	2429
## 5	3500000	2310	3145573	12163
## 6	5e+05	2209.2443438914	91137	112
## 7	4e+07	3158	3013011	9595
## 8	2e+07	818	1854103	2207
## 9	2.8e+07	2714	2213659	2210
## 10	12500000	2253	5218079	11709
## 11	58800000	3555	3927600	13143
## 12	3e+07	1762	519327	963
## 13	6500000	3185	19032902	38810
## 14	6.5e+07	3116	930006	5150
## 15	5e+06	65	595194	85
## 16	6e+06	18	3915978	6983
## 17	5e+06	25	1391527	2479
## 18	7e+06	31	1828235	7633
## 19	1.65e+08	3761	4700023	14163
## 20	9e+06	1823	1348142	4404
## 21	4e+07	3555	7977747	18690
## 22	4e+06	771	1671367	4572
## 23	2.8e+07	2647	2088644	6633

## 24	7e+06	482	4398243	9202
## 25	1e+06	2209.2443438914	7128	1
## 26	3e+06	382	2902492	9522
## 27	1.7e+08	3938	760262	2918
## 28	1.7e+08	3967	1735700	6772
## 29	3e+07	97	465219	1348
## 30	7e+06	2544	1844690	3728
## 31	8.5e+07	3936	463866	3400
## 32	3.6e+07	3376	384448	1230
## 33	7e+07	2209.2443438914	9149892	26427
## 34	2.7e+07	2781	522630	1248
## 35	3.5e+07	3154	3287020	7698
## 36	1.3e+07	3230	1488038	2571
## 37	1.78e+08	3490	15568277	29251
## 38	2e+07	2800	11850723	24226
## 39	4500000	2209.2443438914	735551	636
## 40	2.4e+07	66	6685088	8369
## 41	6.8e+07	3313	2276605	3946
## 42	1850000	8	1034480	6490
## 43	2e+06	780	456564	1706
## 44	1e+06	3	99427	47
## 45	1.6e+08	3952	1156609	2968
## 46	6.1e+07	3014	396010	1390
## 47	1.7e+08	4080	1313548	8567
## 48	70000	9	924347	1406
## 49	1.2e+07	2417	175017	461
## 50	1e+08	3595	9324678	15479
## 51	4.2e+07	93	1292235	5284
## 52	1.45e+08	4253	11472161	22779
## 53	1.1e+07	1272	9222933	41728
## 54	8932779.93	6	8210	6
## 55	2e+07	645	1167941	2651
## 56	1.65e+08	3561	5421705	16635
## 57	5e+07	3434	4270410	8886
## 58	5e+07	2440	817242	4391
## 59	2e+06	4	4877	6
## 60	6e+07	3387	3320754	4322
## 61	4e+07	2905	1438350	4028
## 62	2e+07	2589	4846645	14722
## 63	5e+06	374	3650720	6917
## 64	8.1e+07	3204	2767873	46023
## 65	2.5e+07	28	289922	143
## 66	1.6e+07	1825	5611593	11187
## 67	1.7e+07	1534	4450824	7315
## 68	2400000	2209.2443438914	1222921	5553
## 69	3e+05	3	30529	18
## 70	4e+07	3173	1142964	2346
## 71	1.8e+08	3948	557012	3528
## 72	1.5e+07	66	177465	595
## 73	2.5e+07	3019	1470438	4314
## 74	5e+06	1044	667852	469
## 75	1.45e+08	3934	277848	890
## 76	5e+07	3194	3037329	6696
## 77	5e+06	255	446576	659

## 78	6.6e+07	3115	1451649	7342
## 79	1.8e+07	3279	2554307	8722
## 80	1.27e+08	22	3779254	13535
## 81	8500000	2766	6082510	12522
## 82	13200000	2130	608230	895
## 83	1.25e+08	3567	13661095	41254
## 84	5e+07	3090	367551	700
## 85	6e+05	2209.2443438914	5403836	187162
## 86	5e+06	2061	11724815	30208
## 87	5.5e+07	3303	2028767	3829
## 88	5e+06	2867	105480	352
## 89	1.32e+08	32	1223790	2934
## 90	5e+07	3826	2117798	2124
## 91	1e+08	2658	355563	1568
## 92	2.5e+07	2663	8429023	27484
## 93	1.03e+08	3948	3849768	9783
## 94	1e+08	3372	3305047	11733
## 95	5e+06	371	909596	2214
## 96	5e+06	2209.2443438914	827239	3221
## 97	3.5e+07	2486	3466458	6096
## 98	2e+07	2179	3743181	16782
## 99	3e+07	59	3726728	6221
## 100	9.5e+07	2875	23360	36
## 101	4e+07	3062	2757667	3030
## 102	6.5e+07	2894	5223362	18770
## 103	2.2e+07	3260	943306	3006
## 104	6e+06	27	2426078	9230
## 105	1.3e+07	18	4176181	9463
## 106	4.8e+07	21	1544390	2975
## 107	2e+07	3465	3429055	7682
## 108	1.25e+08	3845	7908038	27312
## 109	8e+06	2809	6132551	14539
## 110	9500000	1586	134353	280
## 111	2e+08	4324	386857	4996
## 112	2e+06	147	6501107	14139
## 113	2.6e+07	8	5546710	15351
## 114	5e+07	12	5165441	17476
## 115	6e+07	3464	2545852	3964
## 116	7e+06	2209.2443438914	330363	406
## 117	12600000	809	2897407	5953
## 118	5.5e+07	3236	7075635	15858
## 119	9e+07	90	14453673	33092
## 120	1.2e+07	3173	32626778	370552
## 121	47921730.0467826	432	10747	4
## 122	2.5e+07	103	4790221	4740
## 123	2.5e+07	702	523457	2187
## 124	2e+07	461	1303646	3306
## 125	2.5e+08	3875	3554189	14152
## 126	1.6e+07	4	26528	58
## 127	2.2e+07	2023	24809	277
## 128	1.25e+08	4151	3305417	14684
## 129	1.2e+07	2209.2443438914	309610	729
## 130	1.4e+07	747	3047849	11748
## 131	4.4e+07	78	6231191	13331

## 132	5e+07	3003	1325872	4829
## 133	7e+07	2104	797229	1606
## 134	6e+07	3775	439159	1847
## 135	1.4e+07	1841	381071	238
## 136	3.4e+07	173	3156436	18124
## 137	7e+07	3083	10649	21
## 138	1.5e+07	2776	1935432	3089
## 139	4.2e+07	3427	2269032	3046
## 140	1e+05	20	1223891	1309
## 141	4e+07	3205	8479994	35071
## 142	9e+06	2805	1142295	3895
## 143	6500000	685	4092871	8781
## 144	12250000	5	702	1
## 145	11712310.56	2	253631	170
## 146	1.5e+07	15	3226251	18240
## 147	22500000	320	3281842	4968
## 148	1e+06	2602	698	16
## 149	2.4e+07	2225	2063089	5858
## 150	1e+08	3455	14141585	36646
## 151	2.1e+08	4233	170909	791
## 152	2800000	449	473100	670
## 153	6.5e+07	3131	6637551	19833
## 154	6e+06	291	476747	2079
## 155	1.5e+07	794	280566	477
## 156	3300000	42	7750223	17541
## 157	2487820.27	2	166612	571
## 158	1.5e+07	1061	5976092	9343
## 159	3300000	4	697105	1023
## 160	6e+07	2965	719976	1312
## 161	6e+06	68	865690	1375
## 162	2e+08	3996	2285	9
## 163	1.2e+07	2	550734	894
## 164	1.5e+08	4274	9143740	34746
## 165	2.5e+08	4276	10366624	31552
## 166	1.9e+08	4004	59056	330
## 167	1.75e+08	3946	1438926	4632
## 168	7.4e+07	4301	1341909	1607
## 169	2.9e+07	3473	9214467	39824
## 170	1.5e+08	3956	8748596	20352
## 171	1.35e+08	3708	10341783	24413
## 172	4e+07	3646	31859569	49900
## 173	7.4e+07	3641	5536822	29411
## 174	1.1e+08	3777	12632836	36508
## 175	1.5e+08	3702	2732371	13030
## 176	2.8e+07	2757	848970	12607
## 177	1.1e+08	3875	2834800	5664
## 178	6.5e+07	3711	6649290	20750
## 179	3.5e+07	3158	9511288	19903
## 180	1.9e+08	3972	999867	4212
## 181	4e+07	3175	10078326	26565
## 182	1.55e+08	3758	84870	265
## 183	6.8e+07	3442	6711914	29903
## 184	3e+07	90	2389347	8809
## 185	8.8e+07	3723	5340100	26134

## 186	14800000	3355	5128288	18475
## 187	2.3e+07	3003	4826940	10521
## 188	3.1e+07	3411	2554327	10062
## 189	1.2e+08	3995	7560211	24168
## 190	3.5e+07	3240	3651828	13998
## 191	1.76e+08	3181	3920842	10535
## 192	2.5e+07	2991	11036701	50002
## 193	2e+07	2855	2947239	19201
## 194	3.4e+07	3366	2393017	13291
## 195	5e+06	1648	2513544	6970
## 196	4e+06	2602	5588384	15144
## 197	3.5e+07	3003	6964819	26601
## 198	8500000	2575	12792898	56001
## 199	1.1e+07	258	6495	82
## 200	3e+07	3108	1841293	5879
## 201	1.2e+07	3031	2854910	23254
## 202	3300000	2666	4442147	10605
## 203	1e+07	2766	11037833	36874
## 204	1.8e+07	1603	638374	4018
## 205	1.2e+07	2893	6646785	16041
## 206	3.5e+07	3261	13154873	46684
## 207	7e+05	2002	50444	238
## 208	1.4e+07	2880	11496	194
## 209	2.8e+07	2778	2996539	1874
## 210	4e+07	2816	3098749	4311
## 211	6e+07	2648	3850758	13363
## 212	7e+07	2567	2409338	6923
## 213	8e+06	34	4032265	18398
## 214	8495000	420	1364537	3824
## 215	4500000	79	5085068	14359
## 216	4e+06	14	63724	115
## 217	1500000	22	44963	109
## 218	1.3e+08	3856	9597644	32558
## 219	3e+07	2772	11476882	40496
## 220	1.5e+07	1255	419470	2218
## 221	7.5e+07	3638	5216680	20010
## 222	9.5e+07	3845	10164908	22726
## 223	4.9e+07	3201	7384182	23597
## 224	1900000	200	890619	6352
## 225	5e+07	3171	5671767	10073
## 226	1e+07	1573	831044	2427
## 227	5e+07	66	3701061	9325
## 228	3.7e+07	2815	7119456	18803
## 229	3.5e+07	2777	3450614	6823
## 230	3e+06	2209.2443438914	66872	400
## 231	1e+05	2720	659772	2841
## 232	47921730.0467826	2209.2443438914	3712851.29004329	12732.5367965368
##	Dislikes	Comments	Aggregate_Followers	
## 1	425	636	1120000	
## 2	61	186	12350000	
## 3	34	47	483000	
## 4	132	590	568000	
## 5	610	1082	1923800	
## 6	7	1	310000	

## 7	419	1020	8153000
## 8	197	593	130655
## 9	419	382	125646
## 10	532	770	21697300
## 11	573	3134	24300
## 12	94	70	386400
## 13	4382	4392	19420105
## 14	707	1484	5130800
## 15	36	39	15112
## 16	247	460	253000
## 17	146	182	1658900
## 18	235	685	116100
## 19	538	1293	199800
## 20	307	1033	888000
## 21	1940	2214	2417000
## 22	207	741	105000
## 23	255	1235	3209000
## 24	454	1150	4769100
## 25	0	0	2182
## 26	558	2296	3038193.44897959
## 27	66	837	8030000
## 28	187	889	114000
## 29	72	162	744600
## 30	581	729	9536
## 31	152	987	1030000
## 32	129	228	276750
## 33	1342	5278	395500
## 34	153	227	147000
## 35	446	1122	17064000
## 36	553	643	88586
## 37	1730	6439	5610000
## 38	1343	2577	21500
## 39	98	92	1060000
## 40	467	1580	13720000
## 41	331	1286	1888000
## 42	181	374	66600
## 43	413	890	412000
## 44	10	12	3038193.44897959
## 45	112	547	1870000
## 46	58	342	20640000
## 47	269	1285	2750000
## 48	107	132	5887700
## 49	34	133	3038193.44897959
## 50	1130	3925	9414000
## 51	124	362	1650000
## 52	862	2863	671000
## 53	924	3609	1800000
## 54	0	0	58900
## 55	82	797	3038193.44897959
## 56	751	4316	1865000
## 57	569	3058	301000
## 58	112	346	4720000
## 59	1	1	3038193.44897959
## 60	347	1105	147000

## 61	133	543	27323
## 62	405	2732	3038193.44897959
## 63	234	1119	1045200
## 64	944	6946	5407000
## 65	17	9	3038193.44897959
## 66	2111	7595	116800
## 67	546	909	2356000
## 68	193	335	1463000
## 69	4	2	18100
## 70	167	311	3038193.44897959
## 71	135	464	5633
## 72	39	71	644000
## 73	168	511	130000
## 74	38	44	919000
## 75	45	88	8839043
## 76	564	1061	9850000
## 77	50	45	3038193.44897959
## 78	533	2305	2594000
## 79	298	693	14240000
## 80	362	1221	6480000
## 81	543	2170	3038193.44897959
## 82	118	387	1550500
## 83	3812	18077	1810000
## 84	19	37	648786
## 85	3145	24919	2720000
## 86	2150	4926	130000
## 87	500	665	124000
## 88	45	85	11444
## 89	123	226	47200
## 90	485	626	3603000
## 91	106	267	370000
## 92	977	2195	21586000
## 93	704	1151	154400
## 94	1077	4319	250000
## 95	186	632	3038193.44897959
## 96	89	432	217000
## 97	411	977	3678000
## 98	565	4973	9224
## 99	405	1074	3038193.44897959
## 100	5	5	804300
## 101	418	251	4521000
## 102	627	2796	8620000
## 103	325	1401	3185900
## 104	184	373	33500
## 105	310	885	269849
## 106	136	380	216000
## 107	675	727	727000
## 108	3439	8533	1260000
## 109	653	846	11783000
## 110	43	308	3038193.44897959
## 111	233	864	4520000
## 112	890	2928	3038193.44897959
## 113	535	1271	48231
## 114	871	3229	10364000

## 115	378	554	1280000
## 116	52	92	3038193.44897959
## 117	153	569	3038193.44897959
## 118	574	1966	1800000
## 119	1336	5005	1480000
## 120	4752	38363	4240000
## 121	1	1	818000
## 122	454	773	2740000
## 123	149	565	184100
## 124	211	564	759800
## 125	262	1782	2613000
## 126	1	8	25748
## 127	12	52	20700
## 128	332	1176	31030000
## 129	97	113	275873
## 130	253	1237	3038193.44897959
## 131	1265	3430	6619435
## 132	104	378	5563500
## 133	630	456	1174806
## 134	105	289	4690000
## 135	27	43	14586
## 136	213	2753	4734000
## 137	1	0	3849
## 138	155	567	148000
## 139	546	554	3038193.44897959
## 140	148	239	226000
## 141	2233	3479	2731000
## 142	193	882	49424
## 143	794	2046	420000
## 144	0	1	9842
## 145	11	58	3038193.44897959
## 146	261	2104	8204
## 147	445	2099	1899400
## 148	1	9	3086
## 149	336	346	24388000
## 150	1459	6811	5987
## 151	362	230	2814900
## 152	104	380	180100
## 153	815	3403	269000
## 154	166	232	2536000
## 155	56	82	585000
## 156	631	2760	858000
## 157	36	70	3038193.44897959
## 158	649	1333	781200
## 159	39	429	3038193.44897959
## 160	76	189	1810000
## 161	79	67	1818778
## 162	0	1	10280000
## 163	44	77	168700
## 164	1074	5107	6180000
## 165	989	3843	10070000
## 166	8	39	11890000
## 167	262	496	232000
## 168	764	48	250000

## 169	998	1987	7336000
## 170	649	1842	6605000
## 171	1675	3426	5070000
## 172	13960	9119	946000
## 173	1840	1281	184000
## 174	2210	7559	9737600
## 175	497	1774	768700
## 176	237	1560	55618
## 177	5746	66	4240000
## 178	750	1666	265000
## 179	2581	2955	2014000
## 180	66	250	1198000
## 181	1418	2395	2939000
## 182	13	63	3877901
## 183	984	1767	10988000
## 184	935	892	1618000
## 185	2007	3717	2466000
## 186	858	1579	8392000
## 187	478	755	2284000
## 188	464	871	2347000
## 189	3524	7139	881000
## 190	969	2205	1066
## 191	565	1668	7460000
## 192	1005	3525	776000
## 193	625	1842	2113
## 194	369	584	324925
## 195	270	1105	1655987
## 196	913	1499	4599000
## 197	1111	1293	1630000
## 198	2083	4102	6714000
## 199	3	7	675000
## 200	314	634	788000
## 201	459	1087	4184000
## 202	691	2739	2007000
## 203	1885	4360	5699
## 204	130	269	1887
## 205	955	2787	253499
## 206	3565	8578	209000
## 207	3	28	986572
## 208	9	31	1891977
## 209	32	189	155000
## 210	341	881	1520000
## 211	453	1276	3045000
## 212	340	714	1334000
## 213	302	1298	2208
## 214	689	772	3841
## 215	600	2468	3744000
## 216	28	14	129000
## 217	27	42	1188000
## 218	2672	8359	3038193.44897959
## 219	1383	4435	3038193.44897959
## 220	46	239	3038193.44897959
## 221	500	2300	3038193.44897959
## 222	4245	5262	3038193.44897959

```
## 223          786          3481    3038193.44897959
## 224          293          700    3038193.44897959
## 225          480         1712    3038193.44897959
## 226           99          247    3038193.44897959
## 227          641         1859    3038193.44897959
## 228         1128         2290    3038193.44897959
## 229          325          409    3038193.44897959
## 230           67          201    3038193.44897959
## 231          431          606    3038193.44897959
## 232 679.051948051948 1825.7012987013    3038193.44897959
```

```
sum(is.na(new_movie))
```

```
## [1] 0
```

```
#txt <- "Explore the data set" #banner(txt, centre = TRUE, bandChar = "-") ##-----
- ## Explore the data set - ##-----
```

```
#txt <- "Data contents" #banner(txt, centre = TRUE, bandChar = "=") ##=====
## Data contents == ##=====
```

```
Hmisc::contents(new_movie)
```

```
##
## Data frame:new_movie 232 observations and 14 variables    Maximum # NAs:0
##
##
##           Storage
## Movie      character
## Year       character
## Sequel     character
## Sentiment  character
## Genre      character
## Ratings   character
## Gross      character
## Budget    character
## Screens   character
## Views     character
## Likes     character
## Dislikes  character
## Comments  character
## Aggregate_Followers character
```

```
psych::describe(new_movie, skew = TRUE,
                 IQR = TRUE)
```

```
##           vars    n   mean    sd median trimmed   mad min max range
## Movie*      1 232 116.50 67.12  116.5  116.50 85.99   1 232   231
## Year*       2 232   1.59  0.91   1.0   1.49  0.00   1  3    2
## Sequel*     3 232   1.55  1.29   1.0   1.22  0.00   1  8    7
## Sentiment*  4 232  17.65  9.98  12.0  17.23  7.41   1 37   36
## Genre*      5 232   5.67  4.08   6.0   5.52  7.41   1 12   11
## Ratings*    6 232  25.08  9.86  26.0  25.30 10.38   1 46   45
## Gross*      7 232 107.39 62.69 106.5 107.18 80.80   1 216  215
## Budget*     8 232  55.42 29.24  59.5  56.28 31.13   1 105  104
## Screens*    9 232  98.78 58.97  95.5  98.04 79.32   1 201  200
## Views*     10 232 116.50 67.12  116.5 116.50 85.99   1 232  231
```

```
## Likes*          11 232 114.40 66.07 114.5 114.49 84.51 1 228 227
## Dislikes*       12 232 99.07 59.41 100.5 98.99 74.13 1 204 203
## Comments*       13 232 104.75 63.62 104.5 104.52 81.54 1 214 213
## Aggregate_Followers* 14 232 96.62 50.89 106.0 96.78 58.56 1 191 190
##               skew kurtosis se IQR
## Movie*         0.00 -1.22 4.41 115.50
## Year*          0.89 -1.20 0.06 2.00
## Sequel*        2.68 7.47 0.08 0.00
## Sentiment*     0.53 -0.92 0.66 12.25
## Genre*         0.27 -1.45 0.27 10.00
## Ratings*      -0.20 -0.42 0.65 14.00
## Gross*         0.02 -1.22 4.12 108.25
## Budget*       -0.28 -1.11 1.92 47.50
## Screens*       0.08 -1.28 3.87 106.25
## Views*         0.00 -1.22 4.41 115.50
## Likes*        -0.01 -1.21 4.34 113.50
## Dislikes*      0.00 -1.17 3.90 99.50
## Comments*      0.02 -1.22 4.18 109.50
## Aggregate_Followers* -0.06 -0.92 3.34 78.50
```

```
glimpse(new_movie)
```

```
## Rows: 232
## Columns: 14
## $ Movie      <chr> "13 Sins", "22 Jump Street", "3 Days to Kill", "30~
## $ Year       <chr> "2014", "2014", "2014", "2014", "2014", "2014", "2~
## $ Sequel     <chr> "1", "2", "1", "2", "2", "1", "1", "1", "1", "1", ~
## $ Sentiment  <chr> "0", "2", "0", "0", "0", "0", "0", "2", "3", "0", ~
## $ Genre      <chr> "8", "1", "1", "1", "8", "3", "8", "1", "10", "8", ~
## $ Ratings    <chr> "6.3", "7.1", "6.2", "6.3", "4.7", "4.6", "6.1", "~
## $ Gross      <chr> "9130", "1.92e+08", "30700000", "1.06e+08", "17300~
## $ Budget     <chr> "4e+06", "5e+07", "2.8e+07", "1.1e+08", "3500000",~
## $ Screens    <chr> "45", "3306", "2872", "3470", "2310", "2209.244343~
## $ Views      <chr> "3280543", "583289", "304861", "452917", "3145573"~
## $ Likes      <chr> "4632", "3465", "328", "2429", "12163", "112", "95~
## $ Dislikes   <chr> "425", "61", "34", "132", "610", "7", "419", "197"~
## $ Comments   <chr> "636", "186", "47", "590", "1082", "1", "1020", "5~
## $ Aggregate_Followers <chr> "1120000", "12350000", "483000", "568000", "192380~
```

```
head(new_movie, 10)
```

```
##               Movie Year Sequel Sentiment Genre Ratings
## 1              13 Sins 2014      1         0      8      6.3
## 2              22 Jump Street 2014      2         2      1      7.1
## 3              3 Days to Kill 2014      1         0      1      6.2
## 4      300: Rise of an Empire 2014      2         0      1      6.3
## 5              A Haunted House 2 2014      2         0      8      4.7
## 6              A Long Way Off 2014      1         0      3      4.6
## 7  A Million Ways to Die in the West 2014      1         0      8      6.1
## 8              A Most Violent Year 2014      1         2      1      7.1
## 9      A Walk Among the Tombstones 2014      1         3     10      6.5
## 10             About Last Night 2014      1         0      8      6.1
##      Gross   Budget      Screens   Views Likes Dislikes Comments
## 1      9130   4e+06         45 3280543 4632      425      636
## 2 1.92e+08 5e+07        3306 583289 3465      61      186
```

```
## 3 30700000 2.8e+07 2872 304861 328 34 47
## 4 1.06e+08 1.1e+08 3470 452917 2429 132 590
## 5 17300000 3500000 2310 3145573 12163 610 1082
## 6 29000 5e+05 2209.2443438914 91137 112 7 1
## 7 42600000 4e+07 3158 3013011 9595 419 1020
## 8 5750000 2e+07 818 1854103 2207 197 593
## 9 2.6e+07 2.8e+07 2714 2213659 2210 419 382
## 10 48600000 12500000 2253 5218079 11709 532 770
## Aggregate_Followers
## 1 1120000
## 2 12350000
## 3 483000
## 4 568000
## 5 1923800
## 6 310000
## 7 8153000
## 8 130655
## 9 125646
## 10 21697300
```

```
tail(new_movie)
```

```
## Movie Year Sequel Sentiment
## 227 Child 44 2015 1 4
## 228 Aloha 2015 1 13
## 229 Unfinished Business 2015 1 7
## 230 War Room 2015 1 10
## 231 The Gallows 2015 1 -5
## 232 2014.29437229437 1.35930735930736 2.80952380952381
## Genre Ratings Gross Budget
## 227 4 6.4 1210000 5e+07
## 228 15 5.5 2.1e+07 3.7e+07
## 229 8 5.4 10200000 3.5e+07
## 230 1 5.4 12300000 3e+06
## 231 15 4.4 22600000 1e+05
## 232 5.35930735930736 6.44155844155844 68066033.2034632 47921730.0467826
## Screens Views Likes Dislikes
## 227 66 3701061 9325 641
## 228 2815 7119456 18803 1128
## 229 2777 3450614 6823 325
## 230 2209.2443438914 66872 400 67
## 231 2720 659772 2841 431
## 232 2209.2443438914 3712851.29004329 12732.5367965368 679.051948051948
## Comments Aggregate_Followers
## 227 1859 3038193.44897959
## 228 2290 3038193.44897959
## 229 409 3038193.44897959
## 230 201 3038193.44897959
## 231 606 3038193.44897959
## 232 1825.7012987013 3038193.44897959
```

```
car::brief(new_movie)
```

```
## 232 x 14 data.frame (227 rows and 10 columns omitted)
## Movie Year . . . Comments Aggregate_Followers
```

```
##           [c]           [c]           [c]           [c]
## 1    13 Sins      2014           636           1120000
## 2    22 Jump Street 2014           186           12350000
## 3     3 Days to Kill 2014           47           483000
## . . .
## 231 The Gallows    2015           606           3038193.44897959
## 232                2014.29437229437 1825.7012987013 3038193.44897959
```

```
str(new_movie)
```

```
## 'data.frame':   232 obs. of  14 variables:
## $ Movie      : chr  "13 Sins" "22 Jump Street" "3 Days to Kill" "300: Rise of an Empire" ..
## $ Year       : chr  "2014" "2014" "2014" "2014" ...
## $ Sequel     : chr  "1" "2" "1" "2" ...
## $ Sentiment  : chr  "0" "2" "0" "0" ...
## $ Genre      : chr  "8" "1" "1" "1" ...
## $ Ratings    : chr  "6.3" "7.1" "6.2" "6.3" ...
## $ Gross      : chr  "9130" "1.92e+08" "30700000" "1.06e+08" ...
## $ Budget     : chr  "4e+06" "5e+07" "2.8e+07" "1.1e+08" ...
## $ Screens    : chr  "45" "3306" "2872" "3470" ...
## $ Views      : chr  "3280543" "583289" "304861" "452917" ...
## $ Likes      : chr  "4632" "3465" "328" "2429" ...
## $ Dislikes   : chr  "425" "61" "34" "132" ...
## $ Comments   : chr  "636" "186" "47" "590" ...
## $ Aggregate_Followers: chr  "1120000" "12350000" "483000" "568000" ...
```

We have 232 obs. of 14 variables

```
##=====
## EDA: Exploratory Data Analysis == ##=====
movie_data <- new_movie[,6:ncol(new_movie)]
```

Check the column names again,

```
colnames(movie_data)
```

```
## [1] "Ratings"      "Gross"        "Budget"
## [4] "Screens"      "Views"        "Likes"
## [7] "Dislikes"     "Comments"     "Aggregate_Followers"
```

```
#txt <- "Get the summary statistics of the variables" #banner(txt, centre = TRUE, bandChar = ")
##***** ## Get the summary statistics
of the variables ##*****
```

```
summary(movie_data)
```

```
## Ratings      Gross      Budget      Screens
## Length:232    Length:232    Length:232    Length:232
## Class :character Class :character Class :character Class :character
## Mode :character Mode :character Mode :character Mode :character
## Views      Likes      Dislikes      Comments
## Length:232    Length:232    Length:232    Length:232
## Class :character Class :character Class :character Class :character
## Mode :character Mode :character Mode :character Mode :character
```

```
## Aggregate_Followers
## Length:232
## Class :character
## Mode :character
```

```
Hmisc::describe(movie_data)
```

```
## movie_data
##
## 9 Variables      232 Observations
## -----
## Ratings
##      n missing distinct
##    232      0      46
##
## lowest : 3.1 4    4.2 4.3 4.4, highest: 8.1 8.2 8.3 8.6 8.7
## -----
## Gross
##      n missing distinct
##    232      0      216
##
## lowest : 1.01e+08 1.02e+08 1.05e+08 1.06e+08 1.12e+08
## highest: 9130      91400000 924000    93200000 9840
## -----
## Budget
##      n missing distinct
##    232      0      105
##
## lowest : 1.03e+08 1.1e+07  1.1e+08  1.25e+08 1.27e+08
## highest: 8e+06   9.5e+07  9500000  9e+06    9e+07
## -----
## Screens
##      n missing distinct
##    232      0      201
##
## lowest : 103 1044 1061 12    1255, highest: 818 9    90 93 97
## -----
## Views
##      n missing distinct
##    232      0      232
##
## lowest : 10078326 10164908 10341783 1034480 10366624
## highest: 943306  9511288  9597644  99427    999867
## -----
## Likes
##      n missing distinct
##    232      0      228
##
## lowest : 1      10062 10073 1023 10521, highest: 9463 9522 9595 963 9783
## -----
## Dislikes
##      n missing distinct
##    232      0      204
##
## lowest : 0      1      10    1005 104 , highest: 98    984 989 99    998
```

```
## -----
## Comments
##      n  missing distinct
##    232      0      214
##
## lowest : 0      1      1020 1033 1061, highest: 909  9119 92   977  987
## -----
## Aggregate_Followers
##      n  missing distinct
##    232      0      191
##
## lowest : 10070000 10280000 1030000  10364000 1045200
## highest: 9536      9737600  9842      9850000  986572
## -----
```

```
psych::describe(movie_data, skew = TRUE,
                  IQR = TRUE)
```

```
##              vars  n   mean    sd median trimmed  mad min max range
## Ratings*          1 232  25.08  9.86   26.0   25.30 10.38   1  46   45
## Gross*            2 232 107.39 62.69  106.5   107.18 80.80   1 216  215
## Budget*           3 232  55.42 29.24   59.5    56.28 31.13   1 105  104
## Screens*          4 232  98.78 58.97   95.5    98.04 79.32   1 201  200
## Views*            5 232 116.50 67.12  116.5   116.50 85.99   1 232  231
## Likes*            6 232 114.40 66.07  114.5   114.49 84.51   1 228  227
## Dislikes*         7 232  99.07 59.41  100.5    98.99 74.13   1 204  203
## Comments*         8 232 104.75 63.62  104.5   104.52 81.54   1 214  213
## Aggregate_Followers* 9 232  96.62 50.89  106.0    96.78 58.56   1 191  190
##              skew kurtosis  se    IQR
## Ratings*      -0.20    -0.42 0.65  14.00
## Gross*         0.02    -1.22 4.12 108.25
## Budget*       -0.28    -1.11 1.92  47.50
## Screens*       0.08    -1.28 3.87 106.25
## Views*        0.00    -1.22 4.41 115.50
## Likes*       -0.01    -1.21 4.34 113.50
## Dislikes*     0.00    -1.17 3.90  99.50
## Comments*     0.02    -1.22 4.18 109.50
## Aggregate_Followers* -0.06    -0.92 3.34  78.50
```

```
#txt <- "The Dependent Variables in this study is Ratings" #banner(txt, centre = TRUE, bandChar = ":")
```

```
##:::##### ## The Dependent Variables in this study is Ratings "re-
place_mean_Ratings" :: ##:::#####
```

We are going to explore the distribution of this variable

```
movie_replace
```

```
##              Movie Year Sequel Sentiment Genre
## 1              13 Sins 2014      1         0    8
## 2              22 Jump Street 2014      2         2    1
## 3              3 Days to Kill 2014      1         0    1
## 4              300: Rise of an Empire 2014      2         0    1
## 5              A Haunted House 2 2014      2         0    8
## 6              A Long Way Off 2014      1         0    3
```


## 7	A Million Ways to Die in the West	2014	1	0	8
## 8	A Most Violent Year	2014	1	2	1
## 9	A Walk Among the Tombstones	2014	1	3	10
## 10	About Last Night	2014	1	0	8
## 11	American Sniper	2014	1	4	1
## 12	And So It Goes	2014	1	0	8
## 13	Annabelle	2014	1	0	15
## 14	Annie	2014	1	0	8
## 15	Atlas Shrugged: Who Is John Galt?	2014	3	0	3
## 16	Barefoot	2014	1	2	8
## 17	Better Living Through Chemistry	2014	1	0	8
## 18	Beyond the Lights	2014	1	0	3
## 19	Big Hero 6	2014	1	29	12
## 20	Black or White	2014	1	-1	3
## 21	Blended	2014	1	-2	8
## 22	Boyhood	2014	1	9	3
## 23	Brick Mansions	2014	1	0	1
## 24	Cake	2014	1	2	3
## 25	Camp X-Ray	2014	1	3	3
## 26	Cantinflas	2014	1	3	9
## 27	Captain America: The Winter Soldier	2014	2	5	1
## 28	Dawn of the Planet of the Apes	2014	2	-4	1
## 29	Deliver Us from Evil	2014	1	0	10
## 30	Devil's Due	2014	1	0	15
## 31	Divergent	2014	1	-4	2
## 32	Dolphin Tale 2	2014	2	-1	3
## 33	Dracula Untold	2014	1	0	1
## 34	Draft Day	2014	1	0	3
## 35	Dumb and Dumber To	2014	3	0	8
## 36	Earth to Echo	2014	1	0	2
## 37	Edge of Tomorrow	2014	1	0	1
## 38	Endless Love	2014	1	1	3
## 39	Falcon Rising	2014	1	0	1
## 40	Foxcatcher	2014	1	6	9
## 41	Fury	2014	1	3	1
## 42	God Help the Girl	2014	1	-4	3
## 43	God's Not Dead	2014	1	0	3
## 44	God's Pocket	2014	1	0	10
## 45	Godzilla	2014	1	2	1
## 46	Gone Girl	2014	1	-11	3
## 47	Guardians of the Galaxy	2014	1	0	1
## 48	Happy Christmas	2014	1	0	8
## 49	Heaven Is for Real	2014	1	-9	3
## 50	Hercules	2014	1	-2	1
## 51	Horrible Bosses 2	2014	2	-1	8
## 52	How to Train Your Dragon 2	2014	2	3	12
## 53	If I Stay	2014	1	14	3
## 54	In the Name of My Daughter	2014	1	0	3
## 55	Inherent Vice	2014	1	1	8
## 56	Interstellar	2014	1	2	2
## 57	Into the Storm	2014	1	0	1
## 58	Into the Woods	2014	1	-4	2
## 59	It Follows	2014	1	-9	15
## 60	Jack Ryan: Shadow Recruit	2014	5	4	1

## 61	Jersey Boys	2014	1	10	9
## 62	John Wick	2014	1	0	1
## 63	Kill the Messenger	2014	1	0	9
## 64	Kingsman: The Secret Service	2014	1	0	1
## 65	Kung Fu Jungle	2014	1	-2	1
## 66	Left Behind	2014	1	0	1
## 67	Let's Be Cops	2014	1	-5	8
## 68	Life After Beth	2014	1	0	8
## 69	Locker 13	2014	1	0	7
## 70	Lucy	2014	1	1	1
## 71	Maleficent	2014	1	0	1
## 72	Maps to the Stars	2014	1	0	8
## 73	Million Dollar Arm	2014	1	27	9
## 74	Moms' Night Out	2014	1	0	8
## 75	Mr. Peabody & Sherman	2014	1	13	12
## 76	Muppets Most Wanted	2014	3	-1	2
## 77	My Old Lady	2014	1	0	8
## 78	Need for Speed	2014	1	5	1
## 79	Neighbors	2014	1	2	8
## 80	Night at the Museum: Secret of the Tomb	2014	3	2	2
## 81	Nightcrawler	2014	1	-1	10
## 82	No Good Deed	2014	1	0	10
## 83	Noah	2014	1	6	1
## 84	Non-Stop	2014	1	0	1
## 85	Not Cool	2014	1	0	8
## 86	Ouija	2014	1	-4	15
## 87	Paddington	2014	1	14	8
## 88	Paranormal Activity: The Marked Ones	2014	1	0	15
## 89	Penguins of Madagascar	2014	1	8	12
## 90	Planes: Fire & Rescue	2014	3	0	12
## 91	Pompeii	2014	1	0	1
## 92	Ride Along	2014	1	0	1
## 93	Rio 2	2014	2	4	12
## 94	RoboCop	2014	1	7	1
## 95	Rosewater	2014	1	1	9
## 96	Rudderless	2014	1	0	8
## 97	Sabotage	2014	1	17	1
## 98	Selma	2014	1	-5	9
## 99	Serena	2014	1	0	3
## 100	Seventh Son	2014	1	0	1
## 101	Sex Tape	2014	1	0	8
## 102	Sin City: A Dame to Kill For	2014	2	0	10
## 103	Son of God	2014	1	2	9
## 104	Song One	2014	1	0	3
## 105	St. Vincent	2014	1	20	8
## 106	Taken 3	2014	3	6	1
## 107	Tammy	2014	1	0	8
## 108	Teenage Mutant Ninja Turtles	2014	1	0	1
## 109	That Awkward Moment	2014	1	0	8
## 110	The Admiral	2014	1	3	2
## 111	The Amazing Spider-Man 2	2014	2	6	1
## 112	The Babadook	2014	1	-3	3
## 113	The Best of Me	2014	1	3	3
## 114	The Book of Life	2014	1	9	12

## 115	The Boxtrolls	2014	1	5	12
## 116	The Devil's Hand	2014	1	0	7
## 117	The Drop	2014	1	-1	10
## 118	The Equalizer	2014	1	0	1
## 119	The Expendables 3	2014	3	0	1
## 120	The Fault in Our Stars	2014	1	7	3
## 121	The Fluffy Movie: Unity Through Laughter	2014	1	0	8
## 122	The Gambler	2014	1	-2	10
## 123	The Giver	2014	1	6	3
## 124	The Good Lie	2014	1	10	3
## 125	The Hobbit: The Battle of the Five Armies	2014	6	3	2
## 126	The Homesman	2014	1	2	3
## 127	The Hundred-Foot Journey	2014	1	11	8
## 128	The Hunger Games: Mockingjay - Part 1	2014	3	2	2
## 129	The Identical	2014	1	0	3
## 130	The Imitation Game	2014	1	9	9
## 131	The Interview	2014	1	0	1
## 132	The Judge	2014	1	2	10
## 133	The Legend of Hercules	2014	1	0	1
## 134	The Lego Movie	2014	1	-2	12
## 135	The Loft	2014	1	2	6
## 136	The Maze Runner	2014	1	2	1
## 137	The Monuments Men	2014	1	0	3
## 138	The November Man	2014	1	0	1
## 139	The Nut Job	2014	1	0	12
## 140	The One I Love	2014	1	2	8
## 141	The Other Woman	2014	1	-1	8
## 142	The Purge: Anarchy	2014	2	-3	1
## 143	The Pyramid	2014	1	0	15
## 144	The Rover	2014	1	0	10
## 145	The Salvation	2014	1	0	3
## 146	The Theory of Everything	2014	1	8	9
## 147	The Water Diviner	2014	1	7	3
## 148	The Woman in Black 2: Angel of Death	2014	2	-6	3
## 149	Think Like a Man Too	2014	2	0	8
## 150	Transcendence	2014	1	0	3
## 151	Transformers: Age of Extinction	2014	4	0	1
## 152	Tusk	2014	1	0	8
## 153	Unbroken	2014	1	0	9
## 154	Veronica Mars	2014	1	10	10
## 155	When the Game Stands Tall	2014	1	0	3
## 156	Whiplash	2014	1	2	3
## 157	White God	2014	1	11	3
## 158	Wild	2014	1	0	2
## 159	Wild Tales	2014	1	7	8
## 160	Winter's Tale	2014	1	0	3
## 161	Wish I Was Here	2014	1	0	8
## 162	X-Men: Days of Future Past	2014	7	3	1
## 163	Yves Saint Laurent	2014	1	0	9
## 164	Jurassic World	2015	4	1	1
## 165	Avengers: Age of Ultron	2015	2	21	1
## 166	Furious 7	2015	7	-17	1
## 167	Inside Out	2015	1	2	12
## 168	Minions	2015	2	-2	1

## 169	Pitch Perfect 2	2015	2	1	8
## 170	Mission: Impossible - Rogue Nation	2015	5	3	3
## 171	Home	2015	1	-2	8
## 172	Fifty Shades of Grey	2015	1	9	3
## 173	The SpongeBob Movie: Sponge Out of Water	2015	2	0	8
## 174	San Andreas	2015	1	3	1
## 175	Mad Max: Fury Road	2015	4	-4	1
## 176	Straight Outta Compton	2015	1	5	9
## 177	Insurgent	2015	2	-6	2
## 178	Spy	2015	1	0	1
## 179	Trainwreck	2015	1	1	12
## 180	Tomorrowland	2015	1	1	12
## 181	Get Hard	2015	1	6	8
## 182	Terminator Genisys	2015	5	1	1
## 183	Ted 2	2015	2	5	8
## 184	Paul Blart: Mall Cop 2	2015	2	15	1
## 185	Pixels	2015	1	10	1
## 186	Magic Mike XXL	2015	2	7	8
## 187	The Wedding Ringer	2015	1	-8	8
## 188	Vacation	2015	5	13	2
## 189	Fantastic Four	2015	1	20	1
## 190	Poltergeist	2015	1	-5	15
## 191	Jupiter Ascending	2015	1	0	1
## 192	The Age of Adaline	2015	1	11	3
## 193	Max	2015	1	0	1
## 194	The Longest Ride	2015	1	19	3
## 195	The Gift	2015	1	11	1
## 196	The Boy Next Door	2015	1	16	6
## 197	Hot Pursuit	2015	1	2	1
## 198	The DUFF	2015	1	2	8
## 199	Woman in Gold	2015	1	19	10
## 200	Entourage	2015	1	6	3
## 201	Paper Towns	2015	1	6	3
## 202	The Lazarus Effect	2015	1	7	8
## 203	Sinister 2	2015	2	-38	1
## 204	Ricki and the Flash	2015	1	25	3
## 205	Project Almanac	2015	1	7	8
## 206	Hitman: Agent 47	2015	1	10	8
## 207	Dope	2015	1	2	8
## 208	Hot Tub Time Machine 2	2015	2	9	8
## 209	American Ultra	2015	1	21	1
## 210	The Gunman	2015	1	3	8
## 211	Mortdecai	2015	1	-2	8
## 212	Blackhat	2015	1	4	3
## 213	Me and Earl and the Dying Girl	2015	1	15	8
## 214	The Vatican Tapes	2015	1	-2	15
## 215	Maggie	2015	1	10	3
## 216	Americons	2015	1	0	3
## 217	Road Hard	2015	1	14	3
## 218	Ant-Man	2015	1	18	1
## 219	Southpaw	2015	1	6	1
## 220	Ex Machina	2015	1	3	3
## 221	The Man from U.N.C.L.E.	2015	1	8	6
## 222	Cinderella	2015	1	2	8

## 223					Chappie 2015	1	8	1
## 224					Brotherly Love 2015	1	0	8
## 225					Run All Night 2015	1	3	8
## 226					The Second Best Exotic Marigold Hotel 2015	2	26	1
## 227					Child 44 2015	1	4	4
## 228					Aloha 2015	1	13	15
## 229					Unfinished Business 2015	1	7	8
## 230					War Room 2015	1	10	1
## 231					The Gallows 2015	1	-5	15
## 232					NA	NA	NA	NA
##	Ratings	Gross	Budget	Screens	Views	Likes	Dislikes	Comments
## 1	6.3	9130	4000000	45	3280543	4632	425	636
## 2	7.1	192000000	50000000	3306	583289	3465	61	186
## 3	6.2	30700000	28000000	2872	304861	328	34	47
## 4	6.3	106000000	110000000	3470	452917	2429	132	590
## 5	4.7	17300000	3500000	2310	3145573	12163	610	1082
## 6	4.6	29000	500000	NA	91137	112	7	1
## 7	6.1	42600000	40000000	3158	3013011	9595	419	1020
## 8	7.1	5750000	20000000	818	1854103	2207	197	593
## 9	6.5	26000000	28000000	2714	2213659	2210	419	382
## 10	6.1	48600000	12500000	2253	5218079	11709	532	770
## 11	7.3	350000000	58800000	3555	3927600	13143	573	3134
## 12	5.7	15200000	30000000	1762	519327	963	94	70
## 13	5.4	84300000	6500000	3185	19032902	38810	4382	4392
## 14	5.2	85900000	65000000	3116	930006	5150	707	1484
## 15	4.4	830000	5000000	65	595194	85	36	39
## 16	6.6	11800	6000000	18	3915978	6983	247	460
## 17	6.3	72300	5000000	25	1391527	2479	146	182
## 18	6.9	14600000	7000000	31	1828235	7633	235	685
## 19	7.9	222000000	165000000	3761	4700023	14163	538	1293
## 20	6.6	21600000	9000000	1823	1348142	4404	307	1033
## 21	6.5	46300000	40000000	3555	7977747	18690	1940	2214
## 22	8.0	25400000	4000000	771	1671367	4572	207	741
## 23	5.7	20300000	28000000	2647	2088644	6633	255	1235
## 24	6.5	1870000	7000000	482	4398243	9202	454	1150
## 25	7.0	9840	1000000	NA	7128	1	0	0
## 26	6.7	6370000	3000000	382	2902492	9522	558	2296
## 27	7.8	260000000	170000000	3938	760262	2918	66	837
## 28	7.7	209000000	170000000	3967	1735700	6772	187	889
## 29	6.2	30500000	30000000	97	465219	1348	72	162
## 30	4.0	15800000	7000000	2544	1844690	3728	581	729
## 31	6.8	151000000	85000000	3936	463866	3400	152	987
## 32	6.4	42000000	36000000	3376	384448	1230	129	228
## 33	6.3	55900000	70000000	NA	9149892	26427	1342	5278
## 34	6.8	28800000	27000000	2781	522630	1248	153	227
## 35	5.8	86200000	35000000	3154	3287020	7698	446	1122
## 36	5.8	38900000	13000000	3230	1488038	2571	553	643
## 37	7.9	100000000	178000000	3490	15568277	29251	1730	6439
## 38	6.3	23400000	20000000	2800	11850723	24226	1343	2577
## 39	5.7	8690	4500000	NA	735551	636	98	92
## 40	7.1	1210000	24000000	66	6685088	8369	467	1580
## 41	7.6	85700000	68000000	3313	2276605	3946	331	1286
## 42	6.4	102000	1850000	8	1034480	6490	181	374
## 43	5.0	60800000	2000000	780	456564	1706	413	890

## 44	6.1	104000	1000000	3	99427	47	10	12
## 45	6.5	201000000	160000000	3952	1156609	2968	112	547
## 46	8.2	168000000	61000000	3014	396010	1390	58	342
## 47	8.1	333000000	170000000	4080	1313548	8567	269	1285
## 48	5.6	30100	70000	9	924347	1406	107	132
## 49	5.8	91400000	12000000	2417	175017	461	34	133
## 50	6.1	72700000	100000000	3595	9324678	15479	1130	3925
## 51	6.3	54400000	42000000	93	1292235	5284	124	362
## 52	7.9	177000000	145000000	4253	11472161	22779	862	2863
## 53	6.8	50500000	11000000	1272	9222933	41728	924	3609
## 54	6.1	275000	8932780	6	8210	6	0	0
## 55	6.7	8090000	20000000	645	1167941	2651	82	797
## 56	8.7	188000000	165000000	3561	5421705	16635	751	4316
## 57	5.8	47600000	50000000	3434	4270410	8886	569	3058
## 58	6.0	128000000	50000000	2440	817242	4391	112	346
## 59	6.9	14700000	2000000	4	4877	6	1	1
## 60	6.2	50500000	60000000	3387	3320754	4322	347	1105
## 61	6.9	47000000	40000000	2905	1438350	4028	133	543
## 62	7.2	43000000	20000000	2589	4846645	14722	405	2732
## 63	7.0	2450000	5000000	374	3650720	6917	234	1119
## 64	7.8	128000000	81000000	3204	2767873	46023	944	6946
## 65	6.5	129000	25000000	28	289922	143	17	9
## 66	3.1	14000000	16000000	1825	5611593	11187	2111	7595
## 67	6.5	82400000	17000000	1534	4450824	7315	546	909
## 68	5.7	8300	2400000	NA	1222921	5553	193	335
## 69	4.8	2470	300000	3	30529	18	4	2
## 70	6.4	127000000	40000000	3173	1142964	2346	167	311
## 71	7.0	241000000	180000000	3948	557012	3528	135	464
## 72	6.3	348000	15000000	66	177465	595	39	71
## 73	7.1	36400000	25000000	3019	1470438	4314	168	511
## 74	5.4	10400000	5000000	1044	667852	469	38	44
## 75	6.9	112000000	145000000	3934	277848	890	45	88
## 76	6.5	51200000	50000000	3194	3037329	6696	564	1061
## 77	6.3	4010000	5000000	255	446576	659	50	45
## 78	6.6	43600000	66000000	3115	1451649	7342	533	2305
## 79	6.4	150000000	18000000	3279	2554307	8722	298	693
## 80	6.3	114000000	127000000	22	3779254	13535	362	1221
## 81	7.9	32300000	8500000	2766	6082510	12522	543	2170
## 82	5.6	52500000	13200000	2130	608230	895	118	387
## 83	5.9	101000000	125000000	3567	13661095	41254	3812	18077
## 84	7.0	91400000	50000000	3090	367551	700	19	37
## 85	5.2	35700	600000	NA	5403836	187162	3145	24919
## 86	4.4	50800000	5000000	2061	11724815	30208	2150	4926
## 87	7.2	76100000	55000000	3303	2028767	3829	500	665
## 88	5.0	32500000	5000000	2867	105480	352	45	85
## 89	6.8	83300000	132000000	32	1223790	2934	123	226
## 90	5.9	59200000	50000000	3826	2117798	2124	485	626
## 91	5.6	23200000	100000000	2658	355563	1568	106	267
## 92	6.1	134000000	25000000	2663	8429023	27484	977	2195
## 93	6.4	132000000	103000000	3948	3849768	9783	704	1151
## 94	6.2	58600000	100000000	3372	3305047	11733	1077	4319
## 95	6.6	3090000	5000000	371	909596	2214	186	632
## 96	7.5	37400	5000000	NA	827239	3221	89	432
## 97	5.7	10500000	35000000	2486	3466458	6096	411	977

## 98	7.5	52100000	20000000	2179	3743181	16782	565	4973
## 99	5.4	176000	30000000	59	3726728	6221	405	1074
## 100	5.5	17200000	95000000	2875	23360	36	5	5
## 101	5.1	38500000	40000000	3062	2757667	3030	418	251
## 102	6.6	13800000	65000000	2894	5223362	18770	627	2796
## 103	5.5	59700000	22000000	3260	943306	3006	325	1401
## 104	5.8	20200	6000000	27	2426078	9230	184	373
## 105	7.3	44100000	13000000	18	4176181	9463	310	885
## 106	6.1	89300000	48000000	21	1544390	2975	136	380
## 107	4.8	84500000	20000000	3465	3429055	7682	675	727
## 108	5.9	191000000	125000000	3845	7908038	27312	3439	8533
## 109	6.1	26000000	8000000	2809	6132551	14539	653	846
## 110	7.1	2590000	9500000	1586	134353	280	43	308
## 111	6.8	203000000	200000000	4324	386857	4996	233	864
## 112	6.9	924000	2000000	147	6501107	14139	890	2928
## 113	6.6	26800000	26000000	8	5546710	15351	535	1271
## 114	7.3	50200000	50000000	12	5165441	17476	871	3229
## 115	6.8	50800000	60000000	3464	2545852	3964	378	554
## 116	4.8	4240	7000000	NA	330363	406	52	92
## 117	7.1	10700000	12600000	809	2897407	5953	153	569
## 118	7.2	102000000	55000000	3236	7075635	15858	574	1966
## 119	6.1	39300000	90000000	90	14453673	33092	1336	5005
## 120	7.9	125000000	12000000	3173	32626778	370552	4752	38363
## 121	7.0	2820000	NA	432	10747	4	1	1
## 122	6.1	33600000	25000000	103	4790221	4740	454	773
## 123	6.5	45100000	25000000	702	523457	2187	149	565
## 124	7.4	2720000	20000000	461	1303646	3306	211	564
## 125	7.5	255000000	250000000	3875	3554189	14152	262	1782
## 126	6.6	2430000	16000000	4	26528	58	1	8
## 127	7.3	54200000	22000000	2023	24809	277	12	52
## 128	6.8	337000000	125000000	4151	3305417	14684	332	1176
## 129	5.0	2820000	12000000	NA	309610	729	97	113
## 130	8.1	91100000	14000000	747	3047849	11748	253	1237
## 131	6.7	6110000	44000000	78	6231191	13331	1265	3430
## 132	7.5	47100000	50000000	3003	1325872	4829	104	378
## 133	4.2	18800000	70000000	2104	797229	1606	630	456
## 134	7.8	258000000	60000000	3775	439159	1847	105	289
## 135	6.3	5980000	14000000	1841	381071	238	27	43
## 136	6.8	102000000	34000000	173	3156436	18124	213	2753
## 137	6.1	78000000	70000000	3083	10649	21	1	0
## 138	6.3	25000000	15000000	2776	1935432	3089	155	567
## 139	5.8	64200000	42000000	3427	2269032	3046	546	554
## 140	7.1	512000	100000	20	1223891	1309	148	239
## 141	6.0	83900000	40000000	3205	8479994	35071	2233	3479
## 142	6.5	71500000	9000000	2805	1142295	3895	193	882
## 143	4.7	2750000	6500000	685	4092871	8781	794	2046
## 144	6.4	1110000	12250000	5	702	1	0	1
## 145	6.8	5000	11712311	2	253631	170	11	58
## 146	7.8	35900000	15000000	15	3226251	18240	261	2104
## 147	7.2	4190000	22500000	320	3281842	4968	445	2099
## 148	4.9	26500000	1000000	2602	698	16	1	9
## 149	5.7	65200000	24000000	2225	2063089	5858	336	346
## 150	6.3	23000000	100000000	3455	14141585	36646	1459	6811
## 151	5.8	245000000	210000000	4233	170909	791	362	230

## 152	5.4	1820000	2800000	449	473100	670	104	380
## 153	7.2	116000000	65000000	3131	6637551	19833	815	3403
## 154	6.8	3320000	6000000	291	476747	2079	166	232
## 155	6.6	30100000	15000000	794	280566	477	56	82
## 156	8.6	13100000	3300000	42	7750223	17541	631	2760
## 157	6.9	280000	2487820	2	166612	571	36	70
## 158	7.2	37900000	15000000	1061	5976092	9343	649	1333
## 159	8.2	3080000	3300000	4	697105	1023	39	429
## 160	6.2	22500	60000000	2965	719976	1312	76	189
## 161	6.7	3590000	6000000	68	865690	1375	79	67
## 162	8.1	234000000	200000000	3996	2285	9	0	1
## 163	6.2	717000	12000000	2	550734	894	44	77
## 164	7.3	643000000	150000000	4274	9143740	34746	1074	5107
## 165	7.8	458000000	250000000	4276	10366624	31552	989	3843
## 166	7.4	350000000	190000000	4004	59056	330	8	39
## 167	8.6	345000000	175000000	3946	1438926	4632	262	496
## 168	6.6	325000000	74000000	4301	1341909	1607	764	48
## 169	6.7	183000000	29000000	3473	9214467	39824	998	1987
## 170	7.8	171000000	150000000	3956	8748596	20352	649	1842
## 171	6.7	177000000	135000000	3708	10341783	24413	1675	3426
## 172	4.2	166000000	40000000	3646	31859569	49900	13960	9119
## 173	6.1	162000000	74000000	3641	5536822	29411	1840	1281
## 174	6.4	154000000	110000000	3777	12632836	36508	2210	7559
## 175	8.3	153000000	150000000	3702	2732371	13030	497	1774
## 176	8.3	135000000	28000000	2757	848970	12607	237	1560
## 177	6.4	130000000	110000000	3875	2834800	5664	5746	66
## 178	7.3	110000000	65000000	3711	6649290	20750	750	1666
## 179	6.7	105000000	35000000	3158	9511288	19903	2581	2955
## 180	6.6	93200000	190000000	3972	999867	4212	66	250
## 181	6.1	90400000	40000000	3175	10078326	26565	1418	2395
## 182	6.8	89400000	155000000	3758	84870	265	13	63
## 183	6.6	81300000	68000000	3442	6711914	29903	984	1767
## 184	4.3	71000000	30000000	90	2389347	8809	935	892
## 185	5.6	71000000	88000000	3723	5340100	26134	2007	3717
## 186	6.2	65900000	14800000	3355	5128288	18475	858	1579
## 187	6.7	64500000	23000000	3003	4826940	10521	478	755
## 188	6.3	54900000	31000000	3411	2554327	10062	464	871
## 189	4.0	52700000	120000000	3995	7560211	24168	3524	7139
## 190	5.0	47400000	35000000	3240	3651828	13998	969	2205
## 191	5.5	47400000	176000000	3181	3920842	10535	565	1668
## 192	7.3	42500000	25000000	2991	11036701	50002	1005	3525
## 193	7.0	41900000	20000000	2855	2947239	19201	625	1842
## 194	7.2	37400000	34000000	3366	2393017	13291	369	584
## 195	7.6	36100000	5000000	1648	2513544	6970	270	1105
## 196	4.6	35400000	4000000	2602	5588384	15144	913	1499
## 197	5.0	34500000	35000000	3003	6964819	26601	1111	1293
## 198	6.6	34000000	8500000	2575	12792898	56001	2083	4102
## 199	7.4	33300000	11000000	258	6495	82	3	7
## 200	7.1	32400000	30000000	3108	1841293	5879	314	634
## 201	6.9	31400000	12000000	3031	2854910	23254	459	1087
## 202	5.2	25800000	3300000	2666	4442147	10605	691	2739
## 203	5.5	19000000	10000000	2766	11037833	36874	1885	4360
## 204	6.1	23500000	18000000	1603	638374	4018	130	269
## 205	6.4	22300000	12000000	2893	6646785	16041	955	2787

## 206	5.9	16100000	35000000	3261	13154873	46684	3565	8578
## 207	7.5	16800000	700000	2002	50444	238	3	28
## 208	5.1	12300000	14000000	2880	11496	194	9	31
## 209	6.5	10900000	28000000	2778	2996539	1874	32	189
## 210	5.8	10600000	40000000	2816	3098749	4311	341	881
## 211	5.5	7610000	60000000	2648	3850758	13363	453	1276
## 212	5.4	7100000	70000000	2567	2409338	6923	340	714
## 213	8.2	6740000	8000000	34	4032265	18398	302	1298
## 214	4.5	1710000	8495000	420	1364537	3824	689	772
## 215	5.6	131000	4500000	79	5085068	14359	600	2468
## 216	7.3	129000	4000000	14	63724	115	28	14
## 217	6.3	106000	1500000	22	44963	109	27	42
## 218	7.8	169000000	130000000	3856	9597644	32558	2672	8359
## 219	7.7	49500000	30000000	2772	11476882	40496	1383	4435
## 220	7.7	25400000	15000000	1255	419470	2218	46	239
## 221	7.6	34600000	75000000	3638	5216680	20010	500	2300
## 222	7.1	201000000	95000000	3845	10164908	22726	4245	5262
## 223	7.0	31600000	49000000	3201	7384182	23597	786	3481
## 224	6.9	444000	1900000	200	890619	6352	293	700
## 225	6.6	26400000	50000000	3171	5671767	10073	480	1712
## 226	6.6	33100000	10000000	1573	831044	2427	99	247
## 227	6.4	1210000	50000000	66	3701061	9325	641	1859
## 228	5.5	21000000	37000000	2815	7119456	18803	1128	2290
## 229	5.4	10200000	35000000	2777	3450614	6823	325	409
## 230	5.4	12300000	3000000	NA	66872	400	67	201
## 231	4.4	22600000	100000	2720	659772	2841	431	606
## 232	NA	NA	NA	NA	NA	NA	NA	NA
##	Aggregate_Followers replace_mean_Year replace_mean_Sequel							
## 1		1120000		2014.000		1.000000		
## 2		12350000		2014.000		2.000000		
## 3		483000		2014.000		1.000000		
## 4		568000		2014.000		2.000000		
## 5		1923800		2014.000		2.000000		
## 6		310000		2014.000		1.000000		
## 7		8153000		2014.000		1.000000		
## 8		130655		2014.000		1.000000		
## 9		125646		2014.000		1.000000		
## 10		21697300		2014.000		1.000000		
## 11		24300		2014.000		1.000000		
## 12		386400		2014.000		1.000000		
## 13		19420105		2014.000		1.000000		
## 14		5130800		2014.000		1.000000		
## 15		15112		2014.000		3.000000		
## 16		253000		2014.000		1.000000		
## 17		1658900		2014.000		1.000000		
## 18		116100		2014.000		1.000000		
## 19		199800		2014.000		1.000000		
## 20		888000		2014.000		1.000000		
## 21		2417000		2014.000		1.000000		
## 22		105000		2014.000		1.000000		
## 23		3209000		2014.000		1.000000		
## 24		4769100		2014.000		1.000000		
## 25		2182		2014.000		1.000000		
## 26		NA		2014.000		1.000000		

## 27	8030000	2014.000	2.000000
## 28	114000	2014.000	2.000000
## 29	744600	2014.000	1.000000
## 30	9536	2014.000	1.000000
## 31	1030000	2014.000	1.000000
## 32	276750	2014.000	2.000000
## 33	395500	2014.000	1.000000
## 34	147000	2014.000	1.000000
## 35	17064000	2014.000	3.000000
## 36	88586	2014.000	1.000000
## 37	5610000	2014.000	1.000000
## 38	21500	2014.000	1.000000
## 39	1060000	2014.000	1.000000
## 40	13720000	2014.000	1.000000
## 41	1888000	2014.000	1.000000
## 42	66600	2014.000	1.000000
## 43	412000	2014.000	1.000000
## 44	NA	2014.000	1.000000
## 45	1870000	2014.000	1.000000
## 46	20640000	2014.000	1.000000
## 47	2750000	2014.000	1.000000
## 48	5887700	2014.000	1.000000
## 49	NA	2014.000	1.000000
## 50	9414000	2014.000	1.000000
## 51	1650000	2014.000	2.000000
## 52	671000	2014.000	2.000000
## 53	1800000	2014.000	1.000000
## 54	58900	2014.000	1.000000
## 55	NA	2014.000	1.000000
## 56	1865000	2014.000	1.000000
## 57	301000	2014.000	1.000000
## 58	4720000	2014.000	1.000000
## 59	NA	2014.000	1.000000
## 60	147000	2014.000	5.000000
## 61	27323	2014.000	1.000000
## 62	NA	2014.000	1.000000
## 63	1045200	2014.000	1.000000
## 64	5407000	2014.000	1.000000
## 65	NA	2014.000	1.000000
## 66	116800	2014.000	1.000000
## 67	2356000	2014.000	1.000000
## 68	1463000	2014.000	1.000000
## 69	18100	2014.000	1.000000
## 70	NA	2014.000	1.000000
## 71	5633	2014.000	1.000000
## 72	644000	2014.000	1.000000
## 73	130000	2014.000	1.000000
## 74	919000	2014.000	1.000000
## 75	8839043	2014.000	1.000000
## 76	9850000	2014.000	3.000000
## 77	NA	2014.000	1.000000
## 78	2594000	2014.000	1.000000
## 79	14240000	2014.000	1.000000
## 80	6480000	2014.000	3.000000

## 81	NA	2014.000	1.000000
## 82	1550500	2014.000	1.000000
## 83	1810000	2014.000	1.000000
## 84	648786	2014.000	1.000000
## 85	2720000	2014.000	1.000000
## 86	130000	2014.000	1.000000
## 87	124000	2014.000	1.000000
## 88	11444	2014.000	1.000000
## 89	47200	2014.000	1.000000
## 90	3603000	2014.000	3.000000
## 91	370000	2014.000	1.000000
## 92	21586000	2014.000	1.000000
## 93	154400	2014.000	2.000000
## 94	250000	2014.000	1.000000
## 95	NA	2014.000	1.000000
## 96	217000	2014.000	1.000000
## 97	3678000	2014.000	1.000000
## 98	9224	2014.000	1.000000
## 99	NA	2014.000	1.000000
## 100	804300	2014.000	1.000000
## 101	4521000	2014.000	1.000000
## 102	8620000	2014.000	2.000000
## 103	3185900	2014.000	1.000000
## 104	33500	2014.000	1.000000
## 105	269849	2014.000	1.000000
## 106	216000	2014.000	3.000000
## 107	727000	2014.000	1.000000
## 108	1260000	2014.000	1.000000
## 109	11783000	2014.000	1.000000
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## 111	4520000	2014.000	2.000000
## 112	NA	2014.000	1.000000
## 113	48231	2014.000	1.000000
## 114	10364000	2014.000	1.000000
## 115	1280000	2014.000	1.000000
## 116	NA	2014.000	1.000000
## 117	NA	2014.000	1.000000
## 118	1800000	2014.000	1.000000
## 119	1480000	2014.000	3.000000
## 120	4240000	2014.000	1.000000
## 121	818000	2014.000	1.000000
## 122	2740000	2014.000	1.000000
## 123	184100	2014.000	1.000000
## 124	759800	2014.000	1.000000
## 125	2613000	2014.000	6.000000
## 126	25748	2014.000	1.000000
## 127	20700	2014.000	1.000000
## 128	31030000	2014.000	3.000000
## 129	275873	2014.000	1.000000
## 130	NA	2014.000	1.000000
## 131	6619435	2014.000	1.000000
## 132	5563500	2014.000	1.000000
## 133	1174806	2014.000	1.000000
## 134	4690000	2014.000	1.000000

## 135	14586	2014.000	1.000000
## 136	4734000	2014.000	1.000000
## 137	3849	2014.000	1.000000
## 138	148000	2014.000	1.000000
## 139	NA	2014.000	1.000000
## 140	226000	2014.000	1.000000
## 141	2731000	2014.000	1.000000
## 142	49424	2014.000	2.000000
## 143	420000	2014.000	1.000000
## 144	9842	2014.000	1.000000
## 145	NA	2014.000	1.000000
## 146	8204	2014.000	1.000000
## 147	1899400	2014.000	1.000000
## 148	3086	2014.000	2.000000
## 149	24388000	2014.000	2.000000
## 150	5987	2014.000	1.000000
## 151	2814900	2014.000	4.000000
## 152	180100	2014.000	1.000000
## 153	269000	2014.000	1.000000
## 154	2536000	2014.000	1.000000
## 155	585000	2014.000	1.000000
## 156	858000	2014.000	1.000000
## 157	NA	2014.000	1.000000
## 158	781200	2014.000	1.000000
## 159	NA	2014.000	1.000000
## 160	1810000	2014.000	1.000000
## 161	1818778	2014.000	1.000000
## 162	10280000	2014.000	7.000000
## 163	168700	2014.000	1.000000
## 164	6180000	2015.000	4.000000
## 165	10070000	2015.000	2.000000
## 166	11890000	2015.000	7.000000
## 167	232000	2015.000	1.000000
## 168	250000	2015.000	2.000000
## 169	7336000	2015.000	2.000000
## 170	6605000	2015.000	5.000000
## 171	5070000	2015.000	1.000000
## 172	946000	2015.000	1.000000
## 173	184000	2015.000	2.000000
## 174	9737600	2015.000	1.000000
## 175	768700	2015.000	4.000000
## 176	55618	2015.000	1.000000
## 177	4240000	2015.000	2.000000
## 178	265000	2015.000	1.000000
## 179	2014000	2015.000	1.000000
## 180	1198000	2015.000	1.000000
## 181	2939000	2015.000	1.000000
## 182	3877901	2015.000	5.000000
## 183	10988000	2015.000	2.000000
## 184	1618000	2015.000	2.000000
## 185	2466000	2015.000	1.000000
## 186	8392000	2015.000	2.000000
## 187	2284000	2015.000	1.000000
## 188	2347000	2015.000	5.000000

## 189	881000	2015.000	1.000000
## 190	1066	2015.000	1.000000
## 191	7460000	2015.000	1.000000
## 192	776000	2015.000	1.000000
## 193	2113	2015.000	1.000000
## 194	324925	2015.000	1.000000
## 195	1655987	2015.000	1.000000
## 196	4599000	2015.000	1.000000
## 197	1630000	2015.000	1.000000
## 198	6714000	2015.000	1.000000
## 199	675000	2015.000	1.000000
## 200	788000	2015.000	1.000000
## 201	4184000	2015.000	1.000000
## 202	2007000	2015.000	1.000000
## 203	5699	2015.000	2.000000
## 204	1887	2015.000	1.000000
## 205	253499	2015.000	1.000000
## 206	209000	2015.000	1.000000
## 207	986572	2015.000	1.000000
## 208	1891977	2015.000	2.000000
## 209	155000	2015.000	1.000000
## 210	1520000	2015.000	1.000000
## 211	3045000	2015.000	1.000000
## 212	1334000	2015.000	1.000000
## 213	2208	2015.000	1.000000
## 214	3841	2015.000	1.000000
## 215	3744000	2015.000	1.000000
## 216	129000	2015.000	1.000000
## 217	1188000	2015.000	1.000000
## 218	NA	2015.000	1.000000
## 219	NA	2015.000	1.000000
## 220	NA	2015.000	1.000000
## 221	NA	2015.000	1.000000
## 222	NA	2015.000	1.000000
## 223	NA	2015.000	1.000000
## 224	NA	2015.000	1.000000
## 225	NA	2015.000	1.000000
## 226	NA	2015.000	2.000000
## 227	NA	2015.000	1.000000
## 228	NA	2015.000	1.000000
## 229	NA	2015.000	1.000000
## 230	NA	2015.000	1.000000
## 231	NA	2015.000	1.000000
## 232	NA	2014.294	1.359307
##	replace_mean_Sentiment	replace_mean_Genre	replace_mean_Ratings
## 1	0.000000	8.000000	6.300000
## 2	2.000000	1.000000	7.100000
## 3	0.000000	1.000000	6.200000
## 4	0.000000	1.000000	6.300000
## 5	0.000000	8.000000	4.700000
## 6	0.000000	3.000000	4.600000
## 7	0.000000	8.000000	6.100000
## 8	2.000000	1.000000	7.100000
## 9	3.000000	10.000000	6.500000

## 10	0.000000	8.000000	6.100000
## 11	4.000000	1.000000	7.300000
## 12	0.000000	8.000000	5.700000
## 13	0.000000	15.000000	5.400000
## 14	0.000000	8.000000	5.200000
## 15	0.000000	3.000000	4.400000
## 16	2.000000	8.000000	6.600000
## 17	0.000000	8.000000	6.300000
## 18	0.000000	3.000000	6.900000
## 19	29.000000	12.000000	7.900000
## 20	-1.000000	3.000000	6.600000
## 21	-2.000000	8.000000	6.500000
## 22	9.000000	3.000000	8.000000
## 23	0.000000	1.000000	5.700000
## 24	2.000000	3.000000	6.500000
## 25	3.000000	3.000000	7.000000
## 26	3.000000	9.000000	6.700000
## 27	5.000000	1.000000	7.800000
## 28	-4.000000	1.000000	7.700000
## 29	0.000000	10.000000	6.200000
## 30	0.000000	15.000000	4.000000
## 31	-4.000000	2.000000	6.800000
## 32	-1.000000	3.000000	6.400000
## 33	0.000000	1.000000	6.300000
## 34	0.000000	3.000000	6.800000
## 35	0.000000	8.000000	5.800000
## 36	0.000000	2.000000	5.800000
## 37	0.000000	1.000000	7.900000
## 38	1.000000	3.000000	6.300000
## 39	0.000000	1.000000	5.700000
## 40	6.000000	9.000000	7.100000
## 41	3.000000	1.000000	7.600000
## 42	-4.000000	3.000000	6.400000
## 43	0.000000	3.000000	5.000000
## 44	0.000000	10.000000	6.100000
## 45	2.000000	1.000000	6.500000
## 46	-11.000000	3.000000	8.200000
## 47	0.000000	1.000000	8.100000
## 48	0.000000	8.000000	5.600000
## 49	-9.000000	3.000000	5.800000
## 50	-2.000000	1.000000	6.100000
## 51	-1.000000	8.000000	6.300000
## 52	3.000000	12.000000	7.900000
## 53	14.000000	3.000000	6.800000
## 54	0.000000	3.000000	6.100000
## 55	1.000000	8.000000	6.700000
## 56	2.000000	2.000000	8.700000
## 57	0.000000	1.000000	5.800000
## 58	-4.000000	2.000000	6.000000
## 59	-9.000000	15.000000	6.900000
## 60	4.000000	1.000000	6.200000
## 61	10.000000	9.000000	6.900000
## 62	0.000000	1.000000	7.200000
## 63	0.000000	9.000000	7.000000

## 64	0.000000	1.000000	7.800000
## 65	-2.000000	1.000000	6.500000
## 66	0.000000	1.000000	3.100000
## 67	-5.000000	8.000000	6.500000
## 68	0.000000	8.000000	5.700000
## 69	0.000000	7.000000	4.800000
## 70	1.000000	1.000000	6.400000
## 71	0.000000	1.000000	7.000000
## 72	0.000000	8.000000	6.300000
## 73	27.000000	9.000000	7.100000
## 74	0.000000	8.000000	5.400000
## 75	13.000000	12.000000	6.900000
## 76	-1.000000	2.000000	6.500000
## 77	0.000000	8.000000	6.300000
## 78	5.000000	1.000000	6.600000
## 79	2.000000	8.000000	6.400000
## 80	2.000000	2.000000	6.300000
## 81	-1.000000	10.000000	7.900000
## 82	0.000000	10.000000	5.600000
## 83	6.000000	1.000000	5.900000
## 84	0.000000	1.000000	7.000000
## 85	0.000000	8.000000	5.200000
## 86	-4.000000	15.000000	4.400000
## 87	14.000000	8.000000	7.200000
## 88	0.000000	15.000000	5.000000
## 89	8.000000	12.000000	6.800000
## 90	0.000000	12.000000	5.900000
## 91	0.000000	1.000000	5.600000
## 92	0.000000	1.000000	6.100000
## 93	4.000000	12.000000	6.400000
## 94	7.000000	1.000000	6.200000
## 95	1.000000	9.000000	6.600000
## 96	0.000000	8.000000	7.500000
## 97	17.000000	1.000000	5.700000
## 98	-5.000000	9.000000	7.500000
## 99	0.000000	3.000000	5.400000
## 100	0.000000	1.000000	5.500000
## 101	0.000000	8.000000	5.100000
## 102	0.000000	10.000000	6.600000
## 103	2.000000	9.000000	5.500000
## 104	0.000000	3.000000	5.800000
## 105	20.000000	8.000000	7.300000
## 106	6.000000	1.000000	6.100000
## 107	0.000000	8.000000	4.800000
## 108	0.000000	1.000000	5.900000
## 109	0.000000	8.000000	6.100000
## 110	3.000000	2.000000	7.100000
## 111	6.000000	1.000000	6.800000
## 112	-3.000000	3.000000	6.900000
## 113	3.000000	3.000000	6.600000
## 114	9.000000	12.000000	7.300000
## 115	5.000000	12.000000	6.800000
## 116	0.000000	7.000000	4.800000
## 117	-1.000000	10.000000	7.100000

## 118	0.000000	1.000000	7.200000
## 119	0.000000	1.000000	6.100000
## 120	7.000000	3.000000	7.900000
## 121	0.000000	8.000000	7.000000
## 122	-2.000000	10.000000	6.100000
## 123	6.000000	3.000000	6.500000
## 124	10.000000	3.000000	7.400000
## 125	3.000000	2.000000	7.500000
## 126	2.000000	3.000000	6.600000
## 127	11.000000	8.000000	7.300000
## 128	2.000000	2.000000	6.800000
## 129	0.000000	3.000000	5.000000
## 130	9.000000	9.000000	8.100000
## 131	0.000000	1.000000	6.700000
## 132	2.000000	10.000000	7.500000
## 133	0.000000	1.000000	4.200000
## 134	-2.000000	12.000000	7.800000
## 135	2.000000	6.000000	6.300000
## 136	2.000000	1.000000	6.800000
## 137	0.000000	3.000000	6.100000
## 138	0.000000	1.000000	6.300000
## 139	0.000000	12.000000	5.800000
## 140	2.000000	8.000000	7.100000
## 141	-1.000000	8.000000	6.000000
## 142	-3.000000	1.000000	6.500000
## 143	0.000000	15.000000	4.700000
## 144	0.000000	10.000000	6.400000
## 145	0.000000	3.000000	6.800000
## 146	8.000000	9.000000	7.800000
## 147	7.000000	3.000000	7.200000
## 148	-6.000000	3.000000	4.900000
## 149	0.000000	8.000000	5.700000
## 150	0.000000	3.000000	6.300000
## 151	0.000000	1.000000	5.800000
## 152	0.000000	8.000000	5.400000
## 153	0.000000	9.000000	7.200000
## 154	10.000000	10.000000	6.800000
## 155	0.000000	3.000000	6.600000
## 156	2.000000	3.000000	8.600000
## 157	11.000000	3.000000	6.900000
## 158	0.000000	2.000000	7.200000
## 159	7.000000	8.000000	8.200000
## 160	0.000000	3.000000	6.200000
## 161	0.000000	8.000000	6.700000
## 162	3.000000	1.000000	8.100000
## 163	0.000000	9.000000	6.200000
## 164	1.000000	1.000000	7.300000
## 165	21.000000	1.000000	7.800000
## 166	-17.000000	1.000000	7.400000
## 167	2.000000	12.000000	8.600000
## 168	-2.000000	1.000000	6.600000
## 169	1.000000	8.000000	6.700000
## 170	3.000000	3.000000	7.800000
## 171	-2.000000	8.000000	6.700000

## 172	9.000000	3.000000	4.200000
## 173	0.000000	8.000000	6.100000
## 174	3.000000	1.000000	6.400000
## 175	-4.000000	1.000000	8.300000
## 176	5.000000	9.000000	8.300000
## 177	-6.000000	2.000000	6.400000
## 178	0.000000	1.000000	7.300000
## 179	1.000000	12.000000	6.700000
## 180	1.000000	12.000000	6.600000
## 181	6.000000	8.000000	6.100000
## 182	1.000000	1.000000	6.800000
## 183	5.000000	8.000000	6.600000
## 184	15.000000	1.000000	4.300000
## 185	10.000000	1.000000	5.600000
## 186	7.000000	8.000000	6.200000
## 187	-8.000000	8.000000	6.700000
## 188	13.000000	2.000000	6.300000
## 189	20.000000	1.000000	4.000000
## 190	-5.000000	15.000000	5.000000
## 191	0.000000	1.000000	5.500000
## 192	11.000000	3.000000	7.300000
## 193	0.000000	1.000000	7.000000
## 194	19.000000	3.000000	7.200000
## 195	11.000000	1.000000	7.600000
## 196	16.000000	6.000000	4.600000
## 197	2.000000	1.000000	5.000000
## 198	2.000000	8.000000	6.600000
## 199	19.000000	10.000000	7.400000
## 200	6.000000	3.000000	7.100000
## 201	6.000000	3.000000	6.900000
## 202	7.000000	8.000000	5.200000
## 203	-38.000000	1.000000	5.500000
## 204	25.000000	3.000000	6.100000
## 205	7.000000	8.000000	6.400000
## 206	10.000000	8.000000	5.900000
## 207	2.000000	8.000000	7.500000
## 208	9.000000	8.000000	5.100000
## 209	21.000000	1.000000	6.500000
## 210	3.000000	8.000000	5.800000
## 211	-2.000000	8.000000	5.500000
## 212	4.000000	3.000000	5.400000
## 213	15.000000	8.000000	8.200000
## 214	-2.000000	15.000000	4.500000
## 215	10.000000	3.000000	5.600000
## 216	0.000000	3.000000	7.300000
## 217	14.000000	3.000000	6.300000
## 218	18.000000	1.000000	7.800000
## 219	6.000000	1.000000	7.700000
## 220	3.000000	3.000000	7.700000
## 221	8.000000	6.000000	7.600000
## 222	2.000000	8.000000	7.100000
## 223	8.000000	1.000000	7.000000
## 224	0.000000	8.000000	6.900000
## 225	3.000000	8.000000	6.600000

## 226	26.000000	1.000000	6.600000
## 227	4.000000	4.000000	6.400000
## 228	13.000000	15.000000	5.500000
## 229	7.000000	8.000000	5.400000
## 230	10.000000	1.000000	5.400000
## 231	-5.000000	15.000000	4.400000
## 232	2.809524	5.359307	6.441558
##	replace_mean_Gross	replace_mean_Budget	replace_mean_Screens
## 1	9130	4000000	45.000
## 2	192000000	50000000	3306.000
## 3	30700000	28000000	2872.000
## 4	106000000	110000000	3470.000
## 5	17300000	3500000	2310.000
## 6	29000	500000	2209.244
## 7	42600000	40000000	3158.000
## 8	5750000	20000000	818.000
## 9	26000000	28000000	2714.000
## 10	48600000	12500000	2253.000
## 11	350000000	58800000	3555.000
## 12	15200000	30000000	1762.000
## 13	84300000	6500000	3185.000
## 14	85900000	65000000	3116.000
## 15	830000	5000000	65.000
## 16	11800	6000000	18.000
## 17	72300	5000000	25.000
## 18	14600000	7000000	31.000
## 19	222000000	165000000	3761.000
## 20	21600000	9000000	1823.000
## 21	46300000	40000000	3555.000
## 22	25400000	4000000	771.000
## 23	20300000	28000000	2647.000
## 24	1870000	7000000	482.000
## 25	9840	1000000	2209.244
## 26	6370000	3000000	382.000
## 27	260000000	170000000	3938.000
## 28	209000000	170000000	3967.000
## 29	30500000	30000000	97.000
## 30	15800000	7000000	2544.000
## 31	151000000	85000000	3936.000
## 32	42000000	36000000	3376.000
## 33	55900000	70000000	2209.244
## 34	28800000	27000000	2781.000
## 35	86200000	35000000	3154.000
## 36	38900000	13000000	3230.000
## 37	100000000	178000000	3490.000
## 38	23400000	20000000	2800.000
## 39	8690	4500000	2209.244
## 40	1210000	24000000	66.000
## 41	85700000	68000000	3313.000
## 42	102000	1850000	8.000
## 43	60800000	2000000	780.000
## 44	104000	1000000	3.000
## 45	201000000	160000000	3952.000
## 46	168000000	61000000	3014.000

## 47	333000000	170000000	4080.000
## 48	30100	70000	9.000
## 49	91400000	12000000	2417.000
## 50	72700000	100000000	3595.000
## 51	54400000	42000000	93.000
## 52	177000000	145000000	4253.000
## 53	50500000	11000000	1272.000
## 54	275000	8932780	6.000
## 55	8090000	20000000	645.000
## 56	188000000	165000000	3561.000
## 57	47600000	50000000	3434.000
## 58	128000000	50000000	2440.000
## 59	14700000	2000000	4.000
## 60	50500000	60000000	3387.000
## 61	47000000	40000000	2905.000
## 62	43000000	20000000	2589.000
## 63	2450000	5000000	374.000
## 64	128000000	81000000	3204.000
## 65	129000	25000000	28.000
## 66	14000000	16000000	1825.000
## 67	82400000	17000000	1534.000
## 68	8300	2400000	2209.244
## 69	2470	300000	3.000
## 70	127000000	40000000	3173.000
## 71	241000000	180000000	3948.000
## 72	348000	15000000	66.000
## 73	36400000	25000000	3019.000
## 74	10400000	5000000	1044.000
## 75	112000000	145000000	3934.000
## 76	51200000	50000000	3194.000
## 77	4010000	5000000	255.000
## 78	43600000	66000000	3115.000
## 79	150000000	18000000	3279.000
## 80	114000000	127000000	22.000
## 81	32300000	8500000	2766.000
## 82	52500000	13200000	2130.000
## 83	101000000	125000000	3567.000
## 84	91400000	50000000	3090.000
## 85	35700	600000	2209.244
## 86	50800000	5000000	2061.000
## 87	76100000	55000000	3303.000
## 88	32500000	5000000	2867.000
## 89	83300000	132000000	32.000
## 90	59200000	50000000	3826.000
## 91	23200000	100000000	2658.000
## 92	134000000	25000000	2663.000
## 93	132000000	103000000	3948.000
## 94	58600000	100000000	3372.000
## 95	3090000	5000000	371.000
## 96	37400	5000000	2209.244
## 97	10500000	35000000	2486.000
## 98	52100000	20000000	2179.000
## 99	176000	30000000	59.000
## 100	17200000	95000000	2875.000

## 101	38500000	40000000	3062.000
## 102	13800000	65000000	2894.000
## 103	59700000	22000000	3260.000
## 104	20200	6000000	27.000
## 105	44100000	13000000	18.000
## 106	89300000	48000000	21.000
## 107	84500000	20000000	3465.000
## 108	191000000	125000000	3845.000
## 109	26000000	8000000	2809.000
## 110	2590000	9500000	1586.000
## 111	203000000	200000000	4324.000
## 112	924000	2000000	147.000
## 113	26800000	26000000	8.000
## 114	50200000	50000000	12.000
## 115	50800000	60000000	3464.000
## 116	4240	7000000	2209.244
## 117	10700000	12600000	809.000
## 118	102000000	55000000	3236.000
## 119	39300000	90000000	90.000
## 120	125000000	12000000	3173.000
## 121	2820000	47921730	432.000
## 122	33600000	25000000	103.000
## 123	45100000	25000000	702.000
## 124	2720000	20000000	461.000
## 125	255000000	250000000	3875.000
## 126	2430000	16000000	4.000
## 127	54200000	22000000	2023.000
## 128	337000000	125000000	4151.000
## 129	2820000	12000000	2209.244
## 130	91100000	14000000	747.000
## 131	6110000	44000000	78.000
## 132	47100000	50000000	3003.000
## 133	18800000	70000000	2104.000
## 134	258000000	60000000	3775.000
## 135	5980000	14000000	1841.000
## 136	102000000	34000000	173.000
## 137	78000000	70000000	3083.000
## 138	25000000	15000000	2776.000
## 139	64200000	42000000	3427.000
## 140	512000	100000	20.000
## 141	83900000	40000000	3205.000
## 142	71500000	9000000	2805.000
## 143	2750000	6500000	685.000
## 144	1110000	12250000	5.000
## 145	5000	11712311	2.000
## 146	35900000	15000000	15.000
## 147	4190000	22500000	320.000
## 148	26500000	1000000	2602.000
## 149	65200000	24000000	2225.000
## 150	23000000	100000000	3455.000
## 151	245000000	210000000	4233.000
## 152	1820000	2800000	449.000
## 153	116000000	65000000	3131.000
## 154	3320000	6000000	291.000

## 155	30100000	15000000	794.000
## 156	13100000	3300000	42.000
## 157	280000	2487820	2.000
## 158	37900000	15000000	1061.000
## 159	3080000	3300000	4.000
## 160	22500	60000000	2965.000
## 161	3590000	6000000	68.000
## 162	234000000	200000000	3996.000
## 163	717000	12000000	2.000
## 164	643000000	150000000	4274.000
## 165	458000000	250000000	4276.000
## 166	350000000	190000000	4004.000
## 167	345000000	175000000	3946.000
## 168	325000000	74000000	4301.000
## 169	183000000	29000000	3473.000
## 170	171000000	150000000	3956.000
## 171	177000000	135000000	3708.000
## 172	166000000	40000000	3646.000
## 173	162000000	74000000	3641.000
## 174	154000000	110000000	3777.000
## 175	153000000	150000000	3702.000
## 176	135000000	28000000	2757.000
## 177	130000000	110000000	3875.000
## 178	110000000	65000000	3711.000
## 179	105000000	35000000	3158.000
## 180	93200000	190000000	3972.000
## 181	90400000	40000000	3175.000
## 182	89400000	155000000	3758.000
## 183	81300000	68000000	3442.000
## 184	71000000	30000000	90.000
## 185	71000000	88000000	3723.000
## 186	65900000	14800000	3355.000
## 187	64500000	23000000	3003.000
## 188	54900000	31000000	3411.000
## 189	52700000	120000000	3995.000
## 190	47400000	35000000	3240.000
## 191	47400000	176000000	3181.000
## 192	42500000	25000000	2991.000
## 193	41900000	20000000	2855.000
## 194	37400000	34000000	3366.000
## 195	36100000	5000000	1648.000
## 196	35400000	4000000	2602.000
## 197	34500000	35000000	3003.000
## 198	34000000	8500000	2575.000
## 199	33300000	11000000	258.000
## 200	32400000	30000000	3108.000
## 201	31400000	12000000	3031.000
## 202	25800000	3300000	2666.000
## 203	19000000	10000000	2766.000
## 204	23500000	18000000	1603.000
## 205	22300000	12000000	2893.000
## 206	16100000	35000000	3261.000
## 207	16800000	700000	2002.000
## 208	12300000	14000000	2880.000

## 209	10900000	28000000	2778.000
## 210	10600000	40000000	2816.000
## 211	7610000	60000000	2648.000
## 212	7100000	70000000	2567.000
## 213	6740000	8000000	34.000
## 214	1710000	8495000	420.000
## 215	131000	4500000	79.000
## 216	129000	4000000	14.000
## 217	106000	1500000	22.000
## 218	169000000	130000000	3856.000
## 219	49500000	30000000	2772.000
## 220	25400000	15000000	1255.000
## 221	34600000	75000000	3638.000
## 222	201000000	95000000	3845.000
## 223	31600000	49000000	3201.000
## 224	444000	1900000	200.000
## 225	26400000	50000000	3171.000
## 226	33100000	10000000	1573.000
## 227	1210000	50000000	66.000
## 228	21000000	37000000	2815.000
## 229	10200000	35000000	2777.000
## 230	12300000	3000000	2209.244
## 231	22600000	100000	2720.000
## 232	68066033	47921730	2209.244
##	replace_mean_Views	replace_mean_Likes	replace_mean_Dislikes
## 1	3280543	4632.00	425.0000
## 2	583289	3465.00	61.0000
## 3	304861	328.00	34.0000
## 4	452917	2429.00	132.0000
## 5	3145573	12163.00	610.0000
## 6	91137	112.00	7.0000
## 7	3013011	9595.00	419.0000
## 8	1854103	2207.00	197.0000
## 9	2213659	2210.00	419.0000
## 10	5218079	11709.00	532.0000
## 11	3927600	13143.00	573.0000
## 12	519327	963.00	94.0000
## 13	19032902	38810.00	4382.0000
## 14	930006	5150.00	707.0000
## 15	595194	85.00	36.0000
## 16	3915978	6983.00	247.0000
## 17	1391527	2479.00	146.0000
## 18	1828235	7633.00	235.0000
## 19	4700023	14163.00	538.0000
## 20	1348142	4404.00	307.0000
## 21	7977747	18690.00	1940.0000
## 22	1671367	4572.00	207.0000
## 23	2088644	6633.00	255.0000
## 24	4398243	9202.00	454.0000
## 25	7128	1.00	0.0000
## 26	2902492	9522.00	558.0000
## 27	760262	2918.00	66.0000
## 28	1735700	6772.00	187.0000
## 29	465219	1348.00	72.0000

## 30	1844690	3728.00	581.0000
## 31	463866	3400.00	152.0000
## 32	384448	1230.00	129.0000
## 33	9149892	26427.00	1342.0000
## 34	522630	1248.00	153.0000
## 35	3287020	7698.00	446.0000
## 36	1488038	2571.00	553.0000
## 37	15568277	29251.00	1730.0000
## 38	11850723	24226.00	1343.0000
## 39	735551	636.00	98.0000
## 40	6685088	8369.00	467.0000
## 41	2276605	3946.00	331.0000
## 42	1034480	6490.00	181.0000
## 43	456564	1706.00	413.0000
## 44	99427	47.00	10.0000
## 45	1156609	2968.00	112.0000
## 46	396010	1390.00	58.0000
## 47	1313548	8567.00	269.0000
## 48	924347	1406.00	107.0000
## 49	175017	461.00	34.0000
## 50	9324678	15479.00	1130.0000
## 51	1292235	5284.00	124.0000
## 52	11472161	22779.00	862.0000
## 53	9222933	41728.00	924.0000
## 54	8210	6.00	0.0000
## 55	1167941	2651.00	82.0000
## 56	5421705	16635.00	751.0000
## 57	4270410	8886.00	569.0000
## 58	817242	4391.00	112.0000
## 59	4877	6.00	1.0000
## 60	3320754	4322.00	347.0000
## 61	1438350	4028.00	133.0000
## 62	4846645	14722.00	405.0000
## 63	3650720	6917.00	234.0000
## 64	2767873	46023.00	944.0000
## 65	289922	143.00	17.0000
## 66	5611593	11187.00	2111.0000
## 67	4450824	7315.00	546.0000
## 68	1222921	5553.00	193.0000
## 69	30529	18.00	4.0000
## 70	1142964	2346.00	167.0000
## 71	557012	3528.00	135.0000
## 72	177465	595.00	39.0000
## 73	1470438	4314.00	168.0000
## 74	667852	469.00	38.0000
## 75	277848	890.00	45.0000
## 76	3037329	6696.00	564.0000
## 77	446576	659.00	50.0000
## 78	1451649	7342.00	533.0000
## 79	2554307	8722.00	298.0000
## 80	3779254	13535.00	362.0000
## 81	6082510	12522.00	543.0000
## 82	608230	895.00	118.0000
## 83	13661095	41254.00	3812.0000

## 84	367551	700.00	19.0000
## 85	5403836	187162.00	3145.0000
## 86	11724815	30208.00	2150.0000
## 87	2028767	3829.00	500.0000
## 88	105480	352.00	45.0000
## 89	1223790	2934.00	123.0000
## 90	2117798	2124.00	485.0000
## 91	355563	1568.00	106.0000
## 92	8429023	27484.00	977.0000
## 93	3849768	9783.00	704.0000
## 94	3305047	11733.00	1077.0000
## 95	909596	2214.00	186.0000
## 96	827239	3221.00	89.0000
## 97	3466458	6096.00	411.0000
## 98	3743181	16782.00	565.0000
## 99	3726728	6221.00	405.0000
## 100	23360	36.00	5.0000
## 101	2757667	3030.00	418.0000
## 102	5223362	18770.00	627.0000
## 103	943306	3006.00	325.0000
## 104	2426078	9230.00	184.0000
## 105	4176181	9463.00	310.0000
## 106	1544390	2975.00	136.0000
## 107	3429055	7682.00	675.0000
## 108	7908038	27312.00	3439.0000
## 109	6132551	14539.00	653.0000
## 110	134353	280.00	43.0000
## 111	386857	4996.00	233.0000
## 112	6501107	14139.00	890.0000
## 113	5546710	15351.00	535.0000
## 114	5165441	17476.00	871.0000
## 115	2545852	3964.00	378.0000
## 116	330363	406.00	52.0000
## 117	2897407	5953.00	153.0000
## 118	7075635	15858.00	574.0000
## 119	14453673	33092.00	1336.0000
## 120	32626778	370552.00	4752.0000
## 121	10747	4.00	1.0000
## 122	4790221	4740.00	454.0000
## 123	523457	2187.00	149.0000
## 124	1303646	3306.00	211.0000
## 125	3554189	14152.00	262.0000
## 126	26528	58.00	1.0000
## 127	24809	277.00	12.0000
## 128	3305417	14684.00	332.0000
## 129	309610	729.00	97.0000
## 130	3047849	11748.00	253.0000
## 131	6231191	13331.00	1265.0000
## 132	1325872	4829.00	104.0000
## 133	797229	1606.00	630.0000
## 134	439159	1847.00	105.0000
## 135	381071	238.00	27.0000
## 136	3156436	18124.00	213.0000
## 137	10649	21.00	1.0000

## 138	1935432	3089.00	155.0000
## 139	2269032	3046.00	546.0000
## 140	1223891	1309.00	148.0000
## 141	8479994	35071.00	2233.0000
## 142	1142295	3895.00	193.0000
## 143	4092871	8781.00	794.0000
## 144	702	1.00	0.0000
## 145	253631	170.00	11.0000
## 146	3226251	18240.00	261.0000
## 147	3281842	4968.00	445.0000
## 148	698	16.00	1.0000
## 149	2063089	5858.00	336.0000
## 150	14141585	36646.00	1459.0000
## 151	170909	791.00	362.0000
## 152	473100	670.00	104.0000
## 153	6637551	19833.00	815.0000
## 154	476747	2079.00	166.0000
## 155	280566	477.00	56.0000
## 156	7750223	17541.00	631.0000
## 157	166612	571.00	36.0000
## 158	5976092	9343.00	649.0000
## 159	697105	1023.00	39.0000
## 160	719976	1312.00	76.0000
## 161	865690	1375.00	79.0000
## 162	2285	9.00	0.0000
## 163	550734	894.00	44.0000
## 164	9143740	34746.00	1074.0000
## 165	10366624	31552.00	989.0000
## 166	59056	330.00	8.0000
## 167	1438926	4632.00	262.0000
## 168	1341909	1607.00	764.0000
## 169	9214467	39824.00	998.0000
## 170	8748596	20352.00	649.0000
## 171	10341783	24413.00	1675.0000
## 172	31859569	49900.00	13960.0000
## 173	5536822	29411.00	1840.0000
## 174	12632836	36508.00	2210.0000
## 175	2732371	13030.00	497.0000
## 176	848970	12607.00	237.0000
## 177	2834800	5664.00	5746.0000
## 178	6649290	20750.00	750.0000
## 179	9511288	19903.00	2581.0000
## 180	999867	4212.00	66.0000
## 181	10078326	26565.00	1418.0000
## 182	84870	265.00	13.0000
## 183	6711914	29903.00	984.0000
## 184	2389347	8809.00	935.0000
## 185	5340100	26134.00	2007.0000
## 186	5128288	18475.00	858.0000
## 187	4826940	10521.00	478.0000
## 188	2554327	10062.00	464.0000
## 189	7560211	24168.00	3524.0000
## 190	3651828	13998.00	969.0000
## 191	3920842	10535.00	565.0000

## 192	11036701	50002.00	1005.0000
## 193	2947239	19201.00	625.0000
## 194	2393017	13291.00	369.0000
## 195	2513544	6970.00	270.0000
## 196	5588384	15144.00	913.0000
## 197	6964819	26601.00	1111.0000
## 198	12792898	56001.00	2083.0000
## 199	6495	82.00	3.0000
## 200	1841293	5879.00	314.0000
## 201	2854910	23254.00	459.0000
## 202	4442147	10605.00	691.0000
## 203	11037833	36874.00	1885.0000
## 204	638374	4018.00	130.0000
## 205	6646785	16041.00	955.0000
## 206	13154873	46684.00	3565.0000
## 207	50444	238.00	3.0000
## 208	11496	194.00	9.0000
## 209	2996539	1874.00	32.0000
## 210	3098749	4311.00	341.0000
## 211	3850758	13363.00	453.0000
## 212	2409338	6923.00	340.0000
## 213	4032265	18398.00	302.0000
## 214	1364537	3824.00	689.0000
## 215	5085068	14359.00	600.0000
## 216	63724	115.00	28.0000
## 217	44963	109.00	27.0000
## 218	9597644	32558.00	2672.0000
## 219	11476882	40496.00	1383.0000
## 220	419470	2218.00	46.0000
## 221	5216680	20010.00	500.0000
## 222	10164908	22726.00	4245.0000
## 223	7384182	23597.00	786.0000
## 224	890619	6352.00	293.0000
## 225	5671767	10073.00	480.0000
## 226	831044	2427.00	99.0000
## 227	3701061	9325.00	641.0000
## 228	7119456	18803.00	1128.0000
## 229	3450614	6823.00	325.0000
## 230	66872	400.00	67.0000
## 231	659772	2841.00	431.0000
## 232	3712851	12732.54	679.0519
##	replace_mean_Comments	replace_mean_Aggregate_Followers	
## 1	636.000		1120000
## 2	186.000		12350000
## 3	47.000		483000
## 4	590.000		568000
## 5	1082.000		1923800
## 6	1.000		310000
## 7	1020.000		8153000
## 8	593.000		130655
## 9	382.000		125646
## 10	770.000		21697300
## 11	3134.000		24300
## 12	70.000		386400

## 13	4392.000	19420105
## 14	1484.000	5130800
## 15	39.000	15112
## 16	460.000	253000
## 17	182.000	1658900
## 18	685.000	116100
## 19	1293.000	199800
## 20	1033.000	888000
## 21	2214.000	2417000
## 22	741.000	105000
## 23	1235.000	3209000
## 24	1150.000	4769100
## 25	0.000	2182
## 26	2296.000	3038193
## 27	837.000	8030000
## 28	889.000	114000
## 29	162.000	744600
## 30	729.000	9536
## 31	987.000	1030000
## 32	228.000	276750
## 33	5278.000	395500
## 34	227.000	147000
## 35	1122.000	17064000
## 36	643.000	88586
## 37	6439.000	5610000
## 38	2577.000	21500
## 39	92.000	1060000
## 40	1580.000	13720000
## 41	1286.000	1888000
## 42	374.000	66600
## 43	890.000	412000
## 44	12.000	3038193
## 45	547.000	1870000
## 46	342.000	20640000
## 47	1285.000	2750000
## 48	132.000	5887700
## 49	133.000	3038193
## 50	3925.000	9414000
## 51	362.000	1650000
## 52	2863.000	671000
## 53	3609.000	1800000
## 54	0.000	58900
## 55	797.000	3038193
## 56	4316.000	1865000
## 57	3058.000	301000
## 58	346.000	4720000
## 59	1.000	3038193
## 60	1105.000	147000
## 61	543.000	27323
## 62	2732.000	3038193
## 63	1119.000	1045200
## 64	6946.000	5407000
## 65	9.000	3038193
## 66	7595.000	116800

## 67	909.000	2356000
## 68	335.000	1463000
## 69	2.000	18100
## 70	311.000	3038193
## 71	464.000	5633
## 72	71.000	644000
## 73	511.000	130000
## 74	44.000	919000
## 75	88.000	8839043
## 76	1061.000	9850000
## 77	45.000	3038193
## 78	2305.000	2594000
## 79	693.000	14240000
## 80	1221.000	6480000
## 81	2170.000	3038193
## 82	387.000	1550500
## 83	18077.000	1810000
## 84	37.000	648786
## 85	24919.000	2720000
## 86	4926.000	130000
## 87	665.000	124000
## 88	85.000	11444
## 89	226.000	47200
## 90	626.000	3603000
## 91	267.000	370000
## 92	2195.000	21586000
## 93	1151.000	154400
## 94	4319.000	250000
## 95	632.000	3038193
## 96	432.000	217000
## 97	977.000	3678000
## 98	4973.000	9224
## 99	1074.000	3038193
## 100	5.000	804300
## 101	251.000	4521000
## 102	2796.000	8620000
## 103	1401.000	3185900
## 104	373.000	33500
## 105	885.000	269849
## 106	380.000	216000
## 107	727.000	727000
## 108	8533.000	1260000
## 109	846.000	11783000
## 110	308.000	3038193
## 111	864.000	4520000
## 112	2928.000	3038193
## 113	1271.000	48231
## 114	3229.000	10364000
## 115	554.000	1280000
## 116	92.000	3038193
## 117	569.000	3038193
## 118	1966.000	1800000
## 119	5005.000	1480000
## 120	38363.000	4240000

## 121	1.000	818000
## 122	773.000	2740000
## 123	565.000	184100
## 124	564.000	759800
## 125	1782.000	2613000
## 126	8.000	25748
## 127	52.000	20700
## 128	1176.000	31030000
## 129	113.000	275873
## 130	1237.000	3038193
## 131	3430.000	6619435
## 132	378.000	5563500
## 133	456.000	1174806
## 134	289.000	4690000
## 135	43.000	14586
## 136	2753.000	4734000
## 137	0.000	3849
## 138	567.000	148000
## 139	554.000	3038193
## 140	239.000	226000
## 141	3479.000	2731000
## 142	882.000	49424
## 143	2046.000	420000
## 144	1.000	9842
## 145	58.000	3038193
## 146	2104.000	8204
## 147	2099.000	1899400
## 148	9.000	3086
## 149	346.000	24388000
## 150	6811.000	5987
## 151	230.000	2814900
## 152	380.000	180100
## 153	3403.000	269000
## 154	232.000	2536000
## 155	82.000	585000
## 156	2760.000	858000
## 157	70.000	3038193
## 158	1333.000	781200
## 159	429.000	3038193
## 160	189.000	1810000
## 161	67.000	1818778
## 162	1.000	10280000
## 163	77.000	168700
## 164	5107.000	6180000
## 165	3843.000	10070000
## 166	39.000	11890000
## 167	496.000	232000
## 168	48.000	250000
## 169	1987.000	7336000
## 170	1842.000	6605000
## 171	3426.000	5070000
## 172	9119.000	946000
## 173	1281.000	184000
## 174	7559.000	9737600

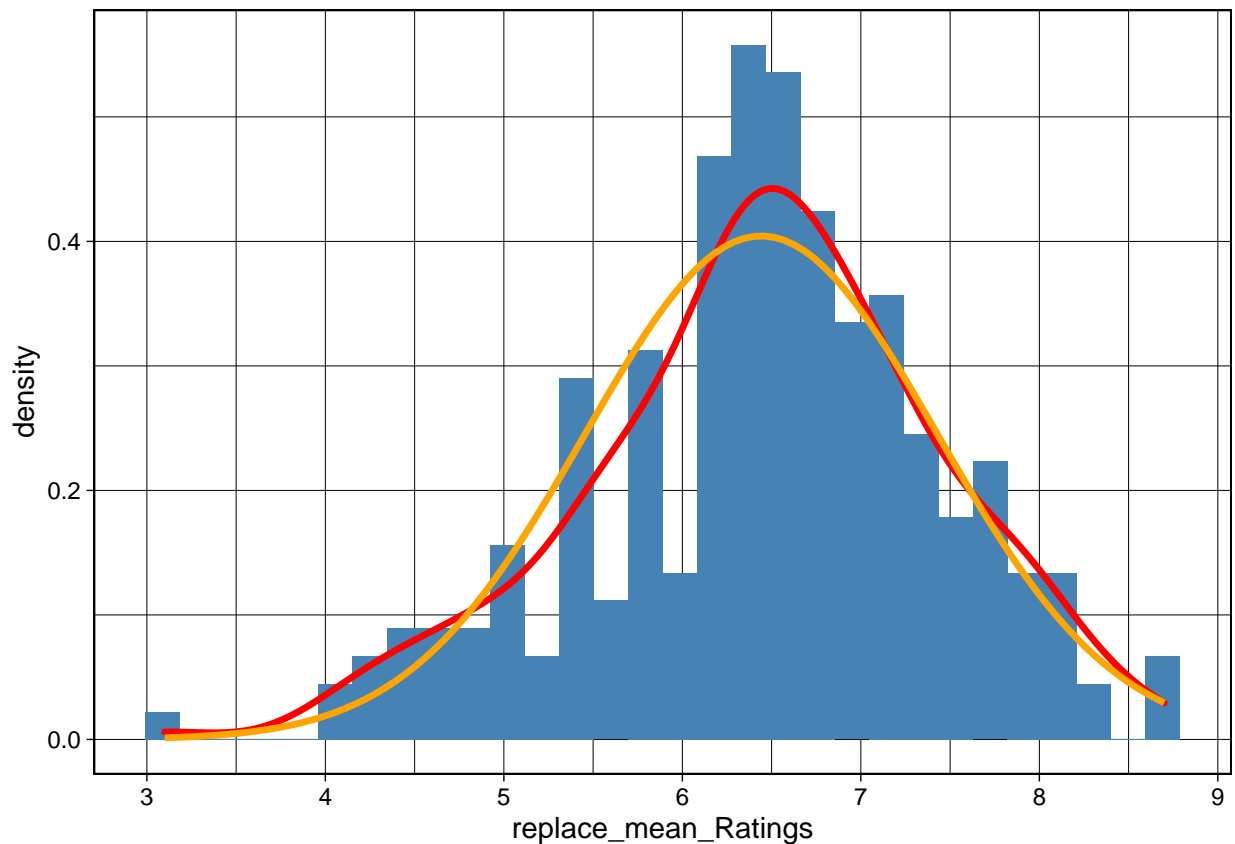
## 175	1774.000	768700
## 176	1560.000	55618
## 177	66.000	4240000
## 178	1666.000	265000
## 179	2955.000	2014000
## 180	250.000	1198000
## 181	2395.000	2939000
## 182	63.000	3877901
## 183	1767.000	10988000
## 184	892.000	1618000
## 185	3717.000	2466000
## 186	1579.000	8392000
## 187	755.000	2284000
## 188	871.000	2347000
## 189	7139.000	881000
## 190	2205.000	1066
## 191	1668.000	7460000
## 192	3525.000	776000
## 193	1842.000	2113
## 194	584.000	324925
## 195	1105.000	1655987
## 196	1499.000	4599000
## 197	1293.000	1630000
## 198	4102.000	6714000
## 199	7.000	675000
## 200	634.000	788000
## 201	1087.000	4184000
## 202	2739.000	2007000
## 203	4360.000	5699
## 204	269.000	1887
## 205	2787.000	253499
## 206	8578.000	209000
## 207	28.000	986572
## 208	31.000	1891977
## 209	189.000	155000
## 210	881.000	1520000
## 211	1276.000	3045000
## 212	714.000	1334000
## 213	1298.000	2208
## 214	772.000	3841
## 215	2468.000	3744000
## 216	14.000	129000
## 217	42.000	1188000
## 218	8359.000	3038193
## 219	4435.000	3038193
## 220	239.000	3038193
## 221	2300.000	3038193
## 222	5262.000	3038193
## 223	3481.000	3038193
## 224	700.000	3038193
## 225	1712.000	3038193
## 226	247.000	3038193
## 227	1859.000	3038193
## 228	2290.000	3038193

```
## 229          409.000          3038193
## 230          201.000          3038193
## 231          606.000          3038193
## 232         1825.701          3038193

View(movie_replace)
movie_data <- movie_replace[,19:ncol(movie_replace)]
View(movie_data)
colnames(movie_data)

## [1] "replace_mean_Ratings"          "replace_mean_Gross"
## [3] "replace_mean_Budget"          "replace_mean_Screens"
## [5] "replace_mean_Views"           "replace_mean_Likes"
## [7] "replace_mean_Dislikes"        "replace_mean_Comments"
## [9] "replace_mean_Aggregate_Followers"

ggplot(movie_data, aes(x = replace_mean_Ratings)) +
  geom_histogram(aes(y = ..density..),
    fill = "steelblue") +
  geom_density(color = "red", lwd = 1.2) +
  stat_function(fun = dnorm, args = list(mean = mean(movie_data$replace_mean_Ratings),
    sd = sd(movie_data$replace_mean_Ratings)),
    color = "orange", lwd = 1.2) +
  theme_linedraw()
```

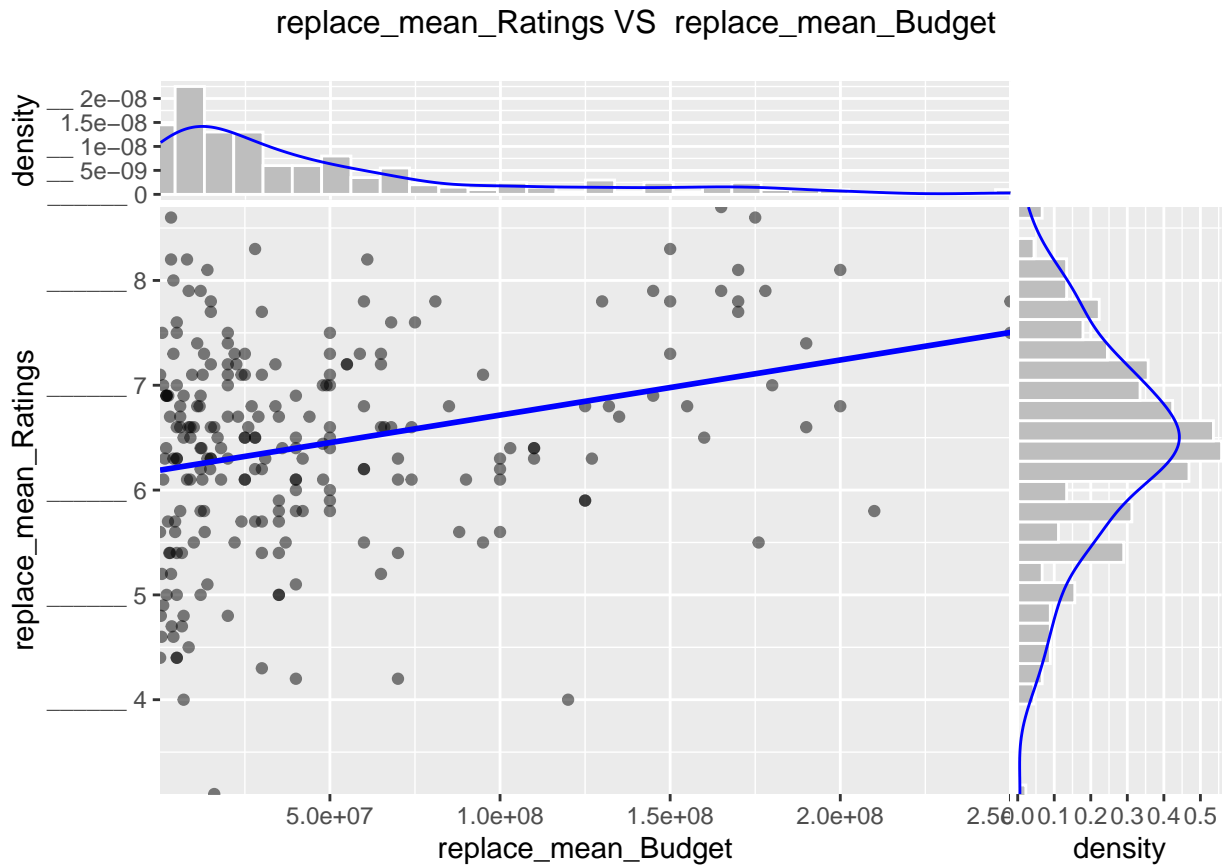


Joint Graphs

```
library(WVPlots)
```

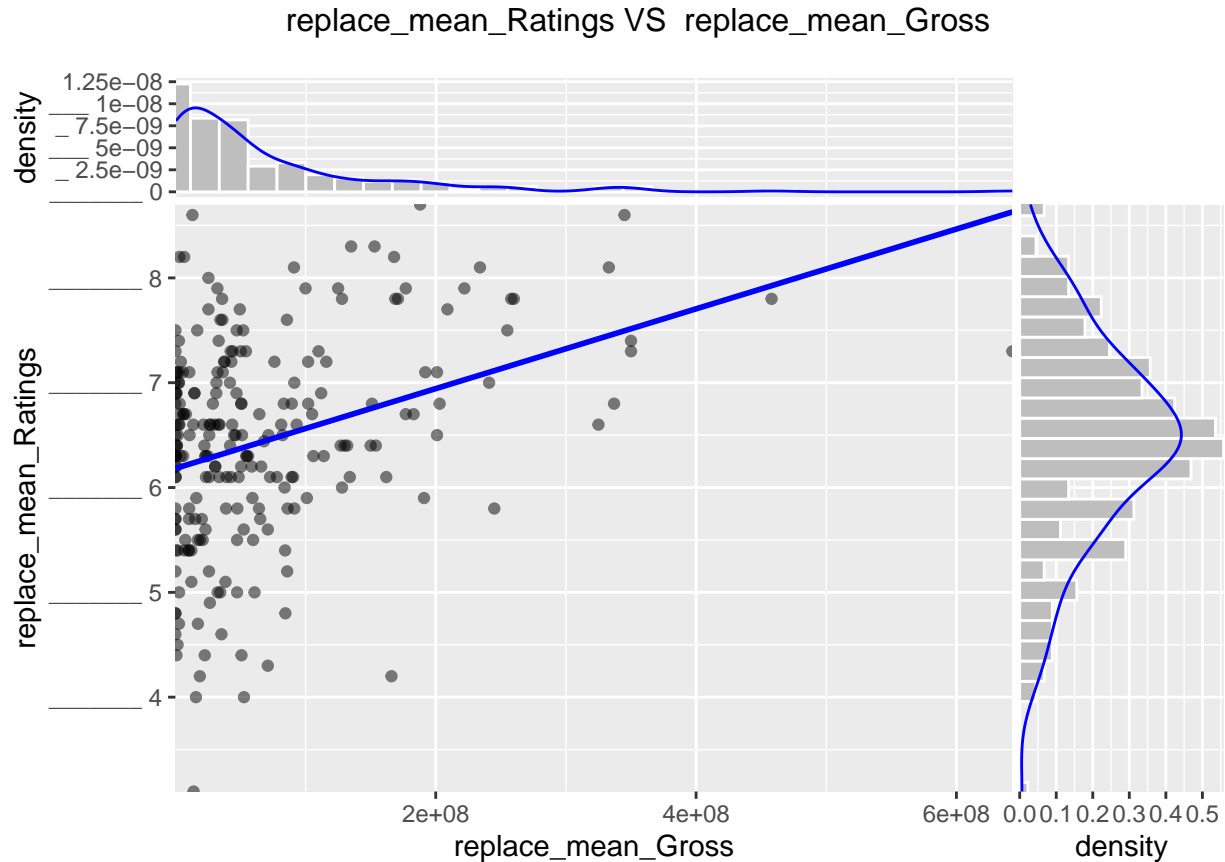
```
#replace_mean_Ratings VS replace_mean_Budget
```

```
ScatterHist(movie_data, title = "replace_mean_Ratings VS replace_mean_Budget",  
            xvar = "replace_mean_Budget", yvar = "replace_mean_Ratings",  
            smoothmethod = "lm")
```



```
## There is a positive linear relationship between the variables(replace_mean_Ratings & re-  
place_mean_Budget) #replace_mean_Ratings VS replace_mean_Gross
```

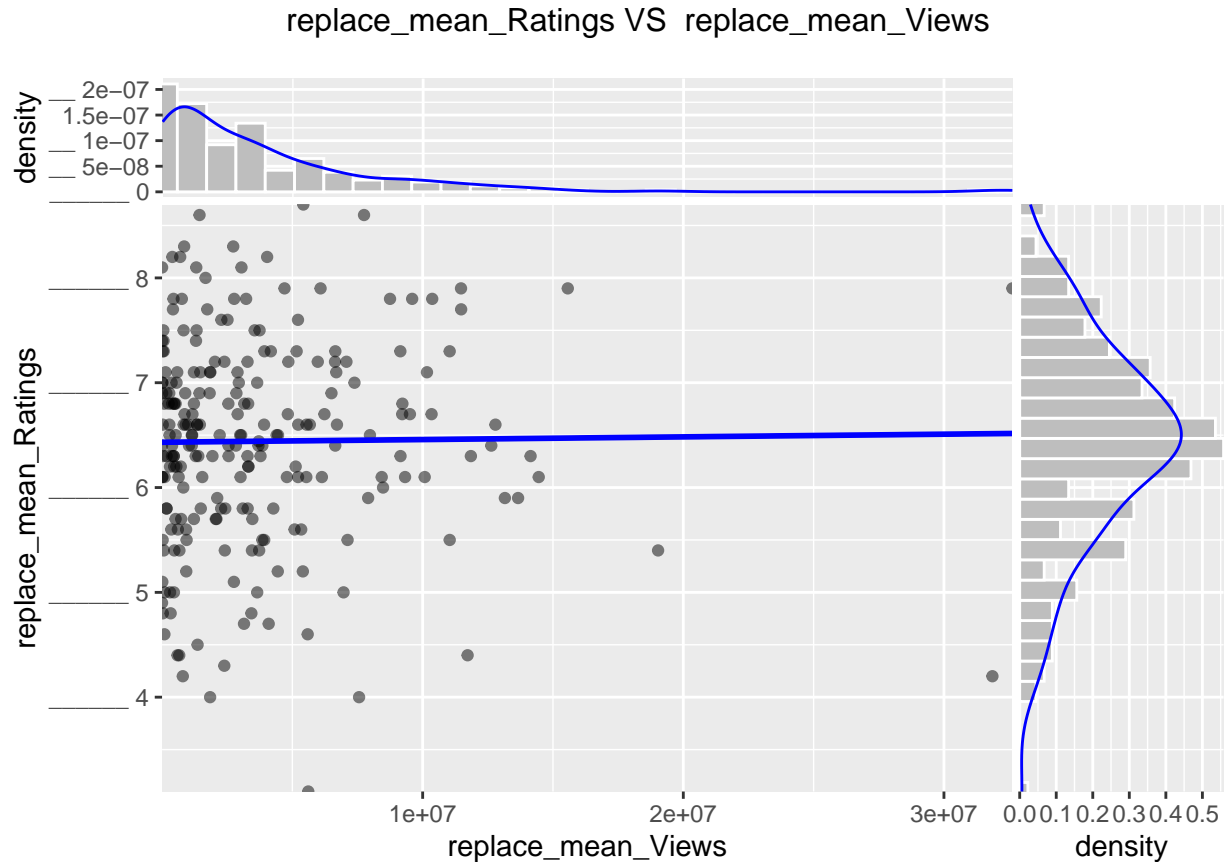
```
ScatterHist(movie_data, title = "replace_mean_Ratings VS replace_mean_Gross",  
            xvar = "replace_mean_Gross", yvar = "replace_mean_Ratings",  
            smoothmethod = "lm")
```

There is a positive linear relationship between the variables (replace_mean_Ratings & replace_mean_Gross)

replace_mean_Ratings VS replace_mean_Views

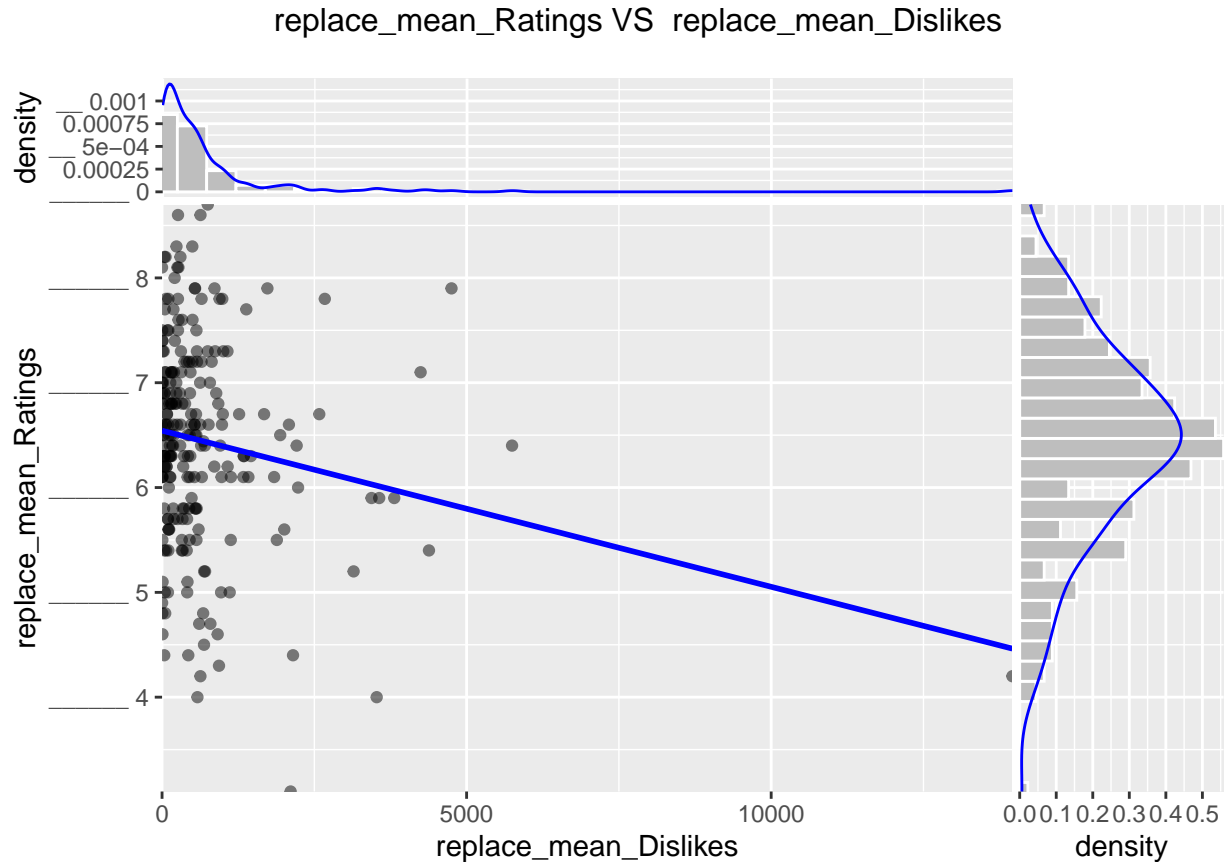
```
ScatterHist(movie_data, title = "replace_mean_Ratings VS replace_mean_Views",
             xvar = "replace_mean_Views", yvar = "replace_mean_Ratings",
             smoothmethod = "lm")
```



There is a very weak relationship between the variables(replace__mean__Ratings & replace__mean__Views)

replace__mean__Ratings VS replace__mean__Dislikes

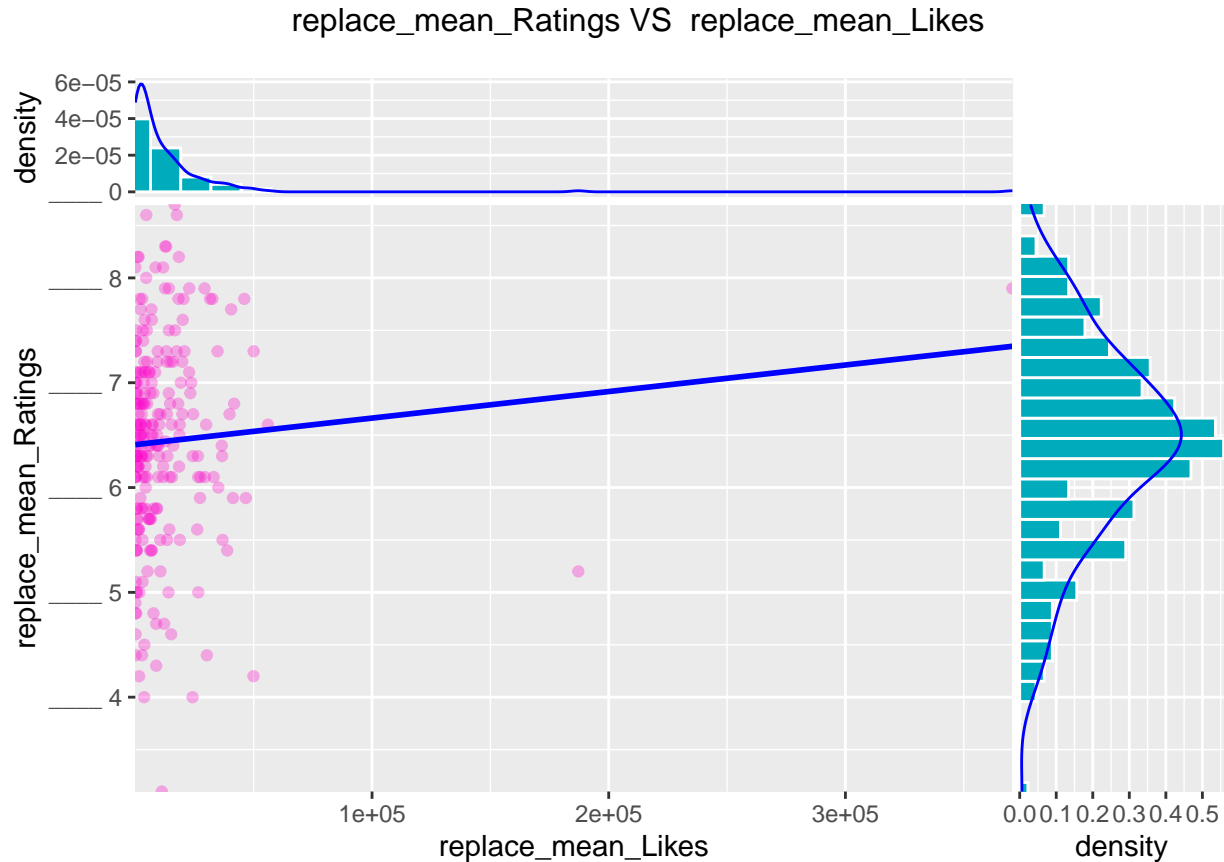
```
ScatterHist(movie_data, title = "replace_mean_Ratings VS replace_mean_Dislikes",
             xvar = "replace_mean_Dislikes", yvar = "replace_mean_Ratings",
             smoothmethod = "lm")
```



There is a negative linear relationship between the variables(replace_mean_Ratings & replace_mean_Dislikes)

replace_mean_Ratings VS replace_mean_Likes

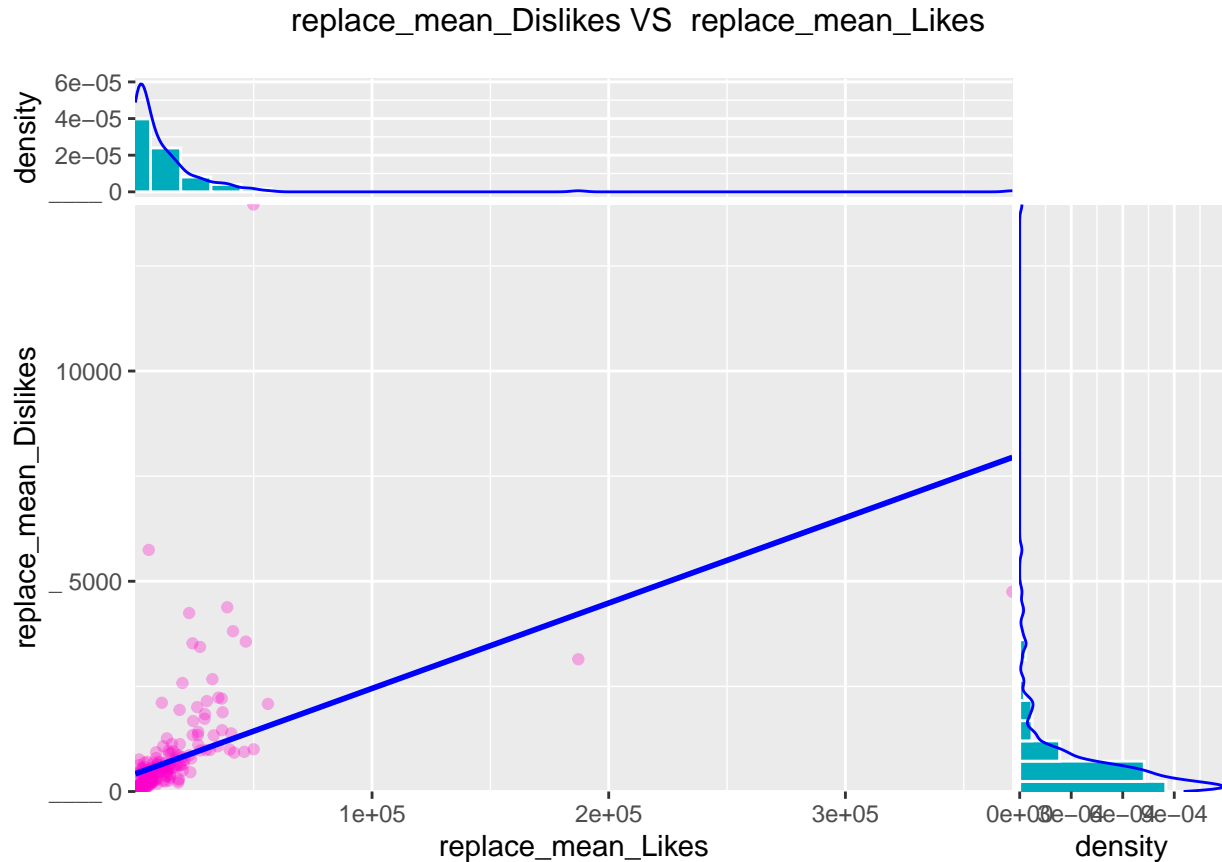
```
ScatterHist(movie_data, title = "replace_mean_Ratings VS replace_mean_Likes",
             xvar = "replace_mean_Likes", yvar = "replace_mean_Ratings",
             smoothmethod = "lm", hist_color = "#00ACBB",
             point_alpha = 0.3,
             point_color = "#FF00CC")
```



There is a positive linear relationship between the variables (replace_mean_Ratings & replace_mean_Likes)

replace_mean_Dislikes VS replace_mean_Likes

```
ScatterHist(movie_data, title = "replace_mean_Dislikes VS replace_mean_Likes",
             xvar = "replace_mean_Likes", yvar = "replace_mean_Dislikes",
             smoothmethod = "lm", hist_color = "#00ACBB",
             point_alpha = 0.3,
             point_color = "#FF00CC")
```



There is a positive linear relationship between the variables (replace_mean_Dislikes & replace_mean_Likes)

```
## ***** ## Study the correlation
## *****
```

```
psych::lowerCor(x = movie_data)
```

```
##          rp_R rp_G rp_B rp_S rp_V rp_L rp_D
## replace_mean_Ratings      1.00
## replace_mean_Gross      0.34  1.00
## replace_mean_Budget      0.29  0.72  1.00
## replace_mean_Screens      0.06  0.58  0.59  1.00
## replace_mean_Views      0.01  0.18  0.11  0.25  1.00
## replace_mean_Likes      0.07  0.11  0.01  0.16  0.68  1.00
## replace_mean_Dislikes    -0.19  0.16  0.10  0.26  0.78  0.47  1.00
## replace_mean_Comments     0.02  0.13  0.09  0.19  0.71  0.92  0.58
## replace_mean_Aggregate_Followers 0.07  0.29  0.16  0.19  0.15  0.08  0.05
##          rp_C r_A_
## replace_mean_Ratings
## replace_mean_Gross
## replace_mean_Budget
## replace_mean_Screens
## replace_mean_Views
## replace_mean_Likes
```

```
## replace_mean_Dislikes
## replace_mean_Comments          1.00
## replace_mean_Aggregate_Followers 0.03  1.00
```

```
psych::corr.test(movie_data)$p
```

```
##                                replace_mean_Ratings replace_mean_Gross
## replace_mean_Ratings          0.000000e+00      2.422017e-06
## replace_mean_Gross            8.970434e-08      0.000000e+00
## replace_mean_Budget           8.307646e-06      3.321590e-38
## replace_mean_Screens          3.970636e-01      3.158471e-22
## replace_mean_Views            8.591898e-01      7.083120e-03
## replace_mean_Likes            2.627507e-01      9.332830e-02
## replace_mean_Dislikes         4.173496e-03      1.376441e-02
## replace_mean_Comments         8.122354e-01      5.538454e-02
## replace_mean_Aggregate_Followers 2.672329e-01      6.782357e-06
##                                replace_mean_Budget replace_mean_Screens
## replace_mean_Ratings          2.076912e-04      1.000000e+00
## replace_mean_Gross            1.129341e-36      9.159566e-21
## replace_mean_Budget           0.000000e+00      2.920777e-21
## replace_mean_Screens          9.421861e-23      0.000000e+00
## replace_mean_Views            8.169808e-02      9.645549e-05
## replace_mean_Likes            8.593522e-01      1.572106e-02
## replace_mean_Dislikes         1.414848e-01      4.480573e-05
## replace_mean_Comments         1.694576e-01      3.445917e-03
## replace_mean_Aggregate_Followers 1.323881e-02      3.296656e-03
##                                replace_mean_Views replace_mean_Likes
## replace_mean_Ratings          1.000000e+00      1.000000e+00
## replace_mean_Gross            1.345793e-01      1.000000e+00
## replace_mean_Budget           1.000000e+00      1.000000e+00
## replace_mean_Screens          2.218476e-03      2.515370e-01
## replace_mean_Views            0.000000e+00      5.631184e-31
## replace_mean_Likes            1.759745e-32      0.000000e+00
## replace_mean_Dislikes         5.934931e-48      3.434477e-14
## replace_mean_Comments         5.821249e-37      4.665160e-94
## replace_mean_Aggregate_Followers 2.356921e-02      2.384126e-01
##                                replace_mean_Dislikes replace_mean_Comments
## replace_mean_Ratings          8.346991e-02      1.000000e+00
## replace_mean_Gross            2.382985e-01      7.753836e-01
## replace_mean_Budget           1.000000e+00      1.000000e+00
## replace_mean_Screens          1.075337e-03      7.252643e-02
## replace_mean_Views            2.077226e-46      1.921012e-35
## replace_mean_Likes            9.616535e-13      1.679457e-92
## replace_mean_Dislikes         0.000000e+00      8.971585e-21
## replace_mean_Comments         2.990528e-22      0.000000e+00
## replace_mean_Aggregate_Followers 4.395099e-01      6.103311e-01
##                                replace_mean_Aggregate_Followers
## replace_mean_Ratings          1.0000000000
## replace_mean_Gross            0.0001763413
## replace_mean_Budget           0.2382985034
## replace_mean_Screens          0.0725264331
## replace_mean_Views            0.3535381264
## replace_mean_Likes            1.0000000000
## replace_mean_Dislikes         1.0000000000
## replace_mean_Comments         1.0000000000
```

```
## replace_mean_Aggregate_Followers
```

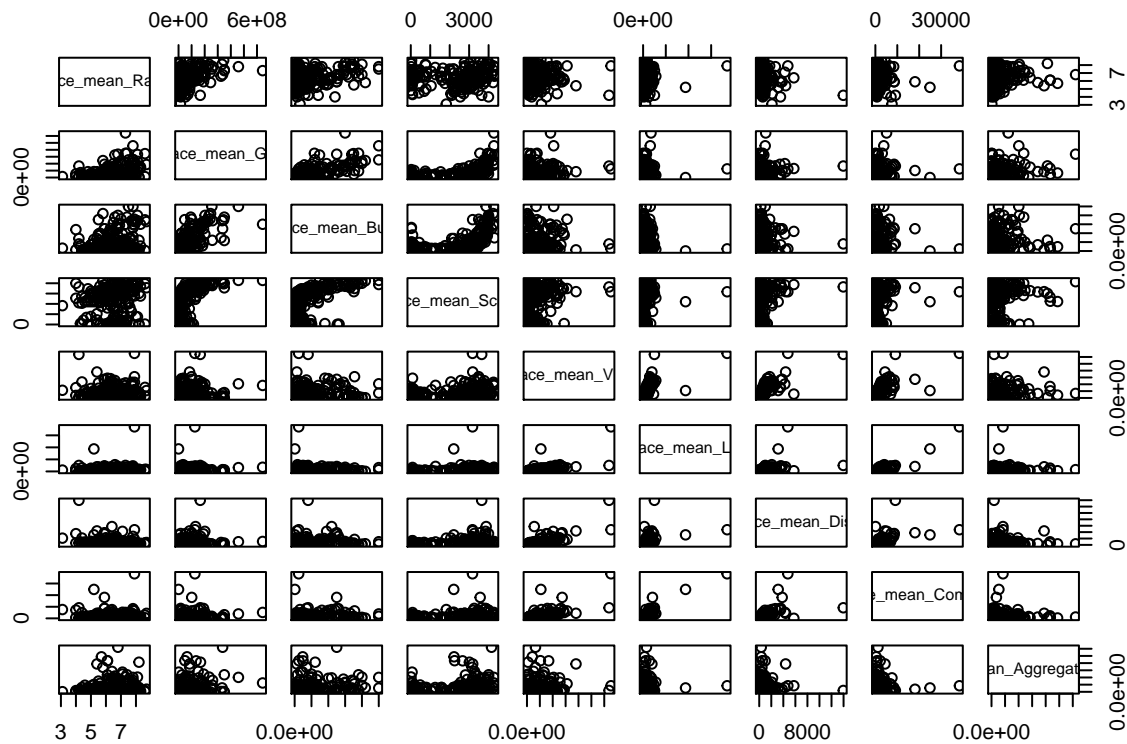
```
0.0000000000
```

Plotting the ScatterPlotMatrix

First look at the help, and the arguments

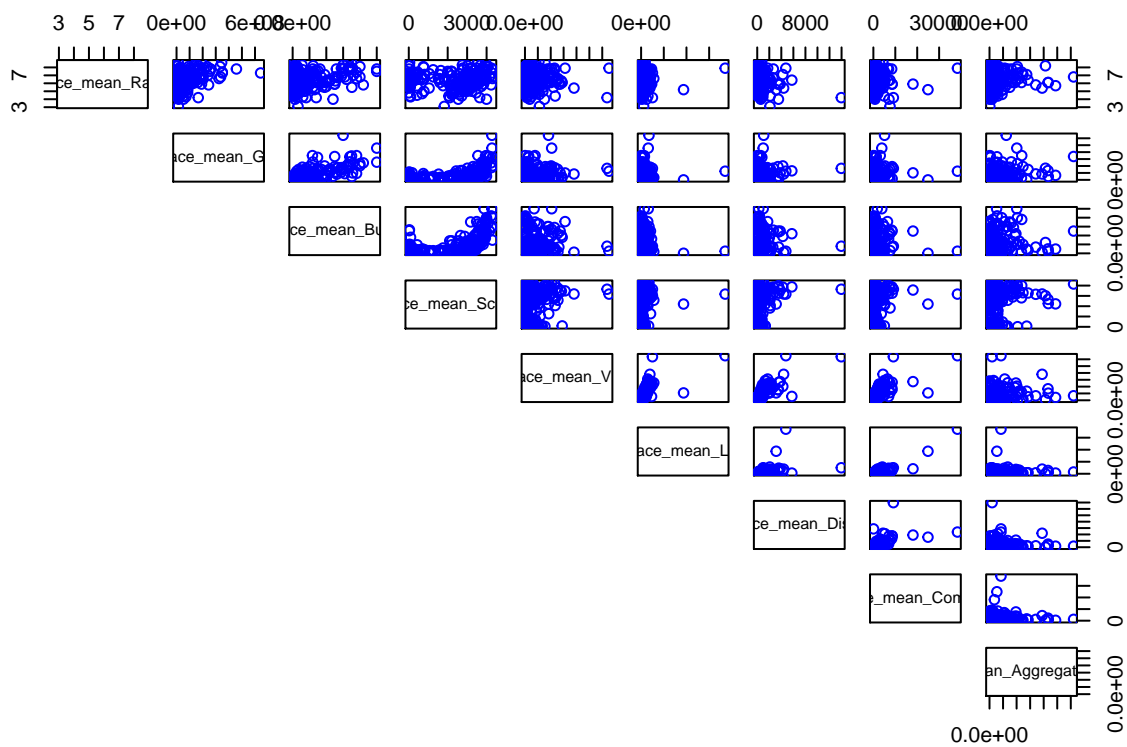
I am going to tweak the knobs a little

```
pairs(movie_data)
```



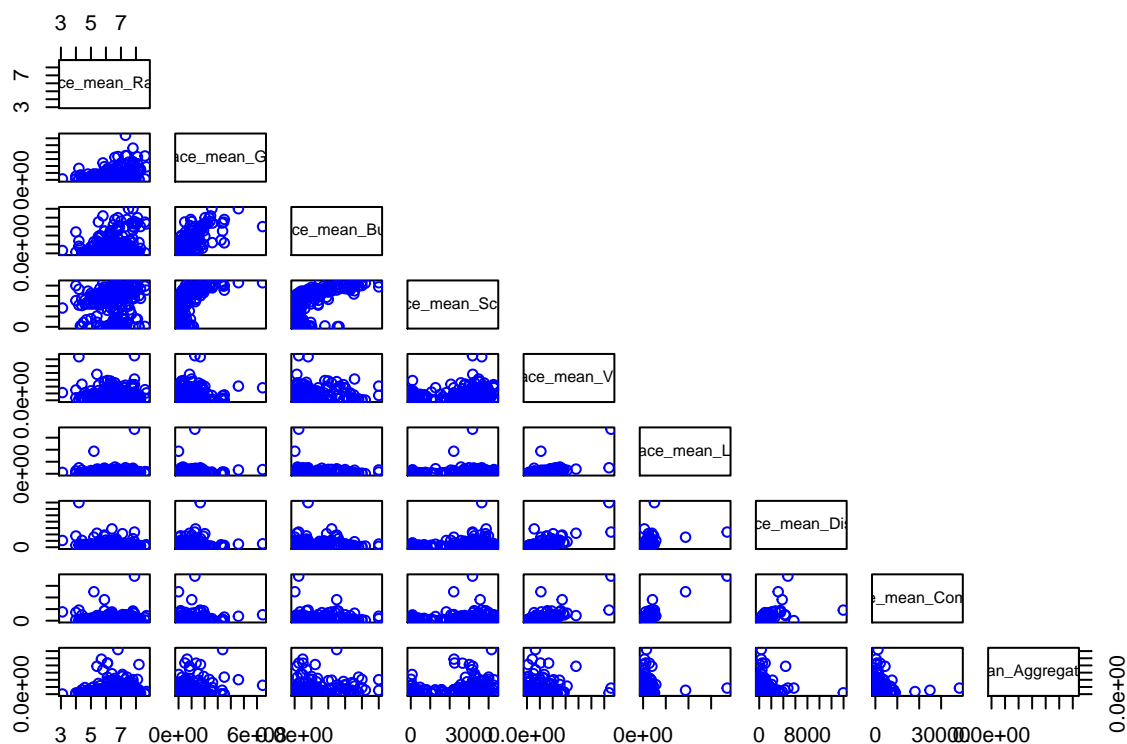
```
#change the color and get help matrix
```

```
pairs(movie_data, lower.panel = NULL, col= "blue")
```



Or if you want only the lower part matrix

```
pairs(movie_data, upper.panel = NULL, col= "blue")
```

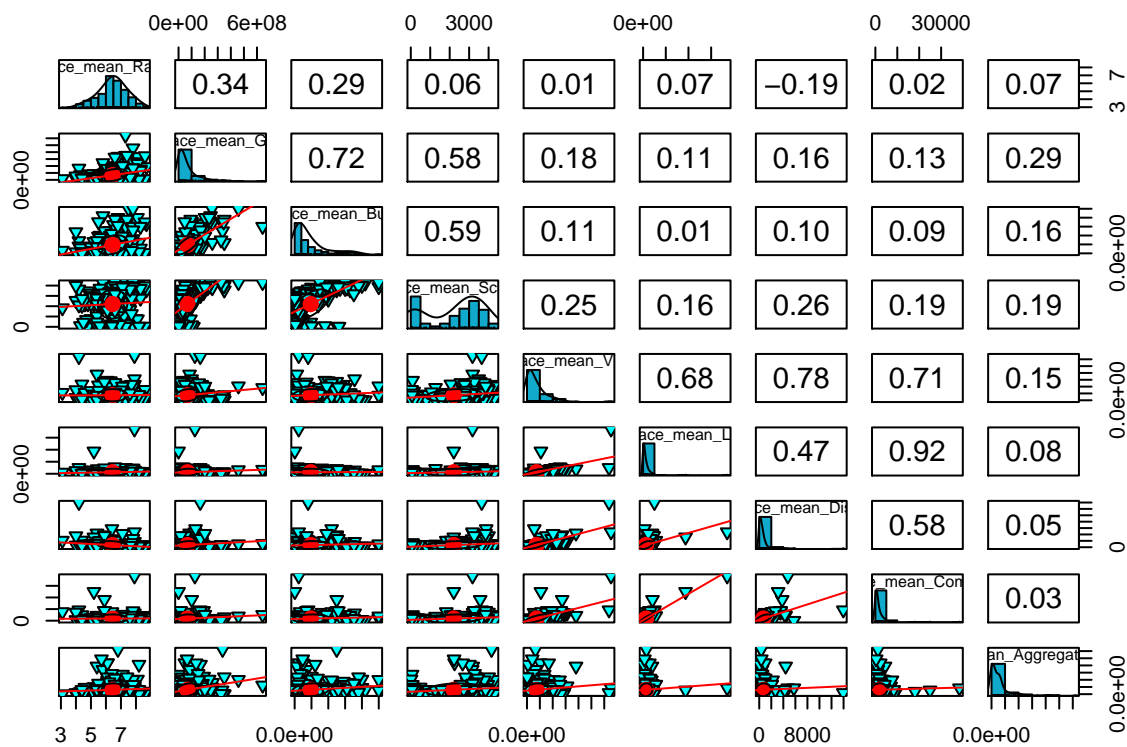



Check the documentation for more information

?pairs

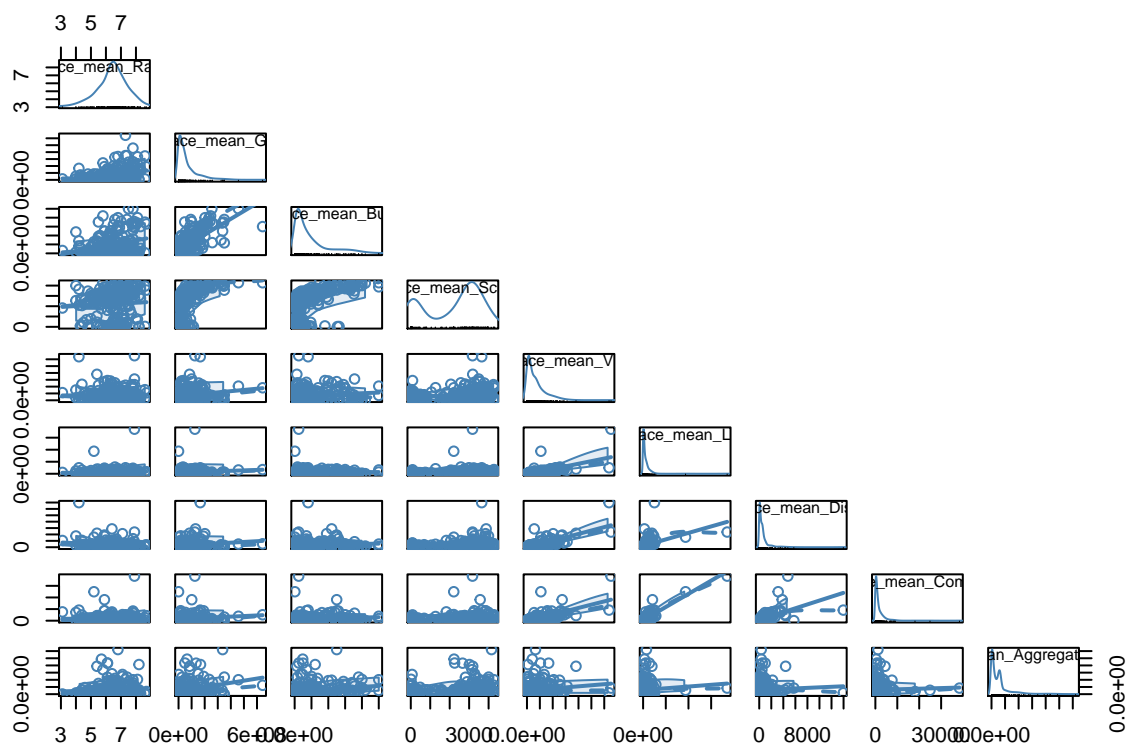
Scatter Matrix with psych package

```
library(psych)
pairs.panels(movie_data,
  method = "pearson", # Correlation method
  hist.col = "#11AACC",
  density = TRUE,
  cex.cor = 1.3,
  col = "red",
  lm = TRUE,
  pch = 25, # point character
  bg = "cyan") # background
```



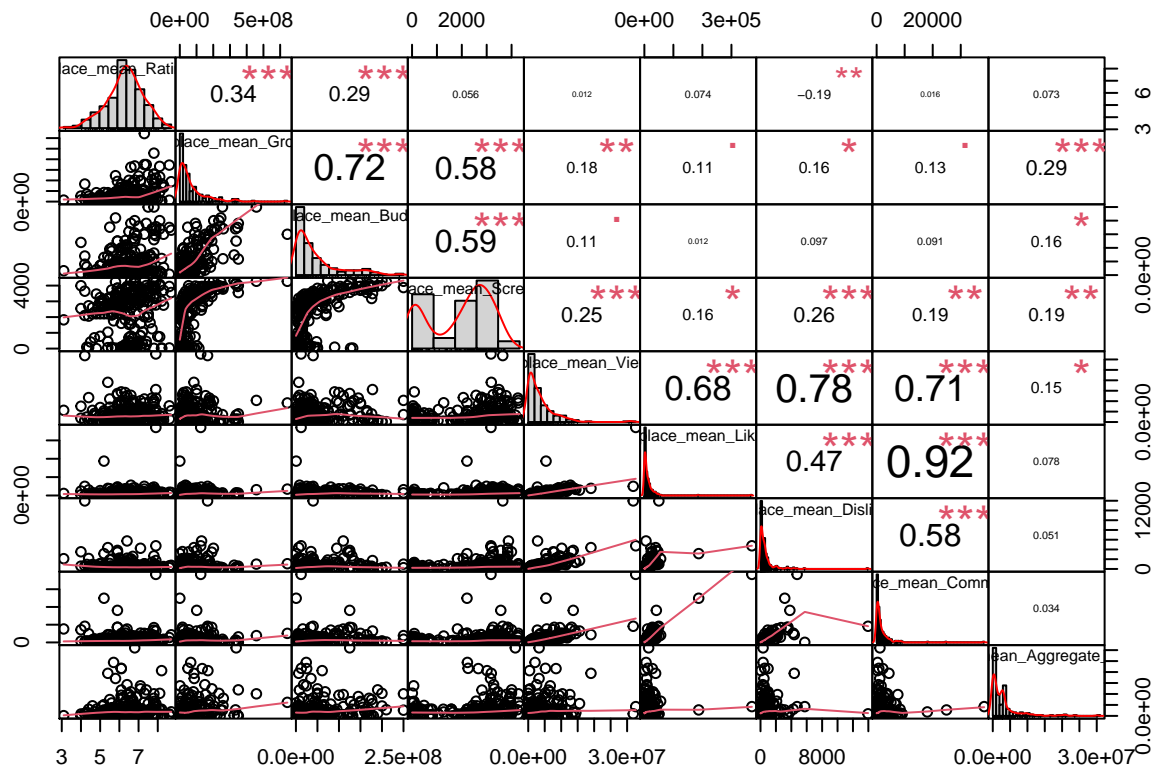
Scatter Matrix with car package

```
car::scatterplotMatrix(movie_data,
  col = "steelblue",
  pch = 21,
  upper.panel = NULL)
```

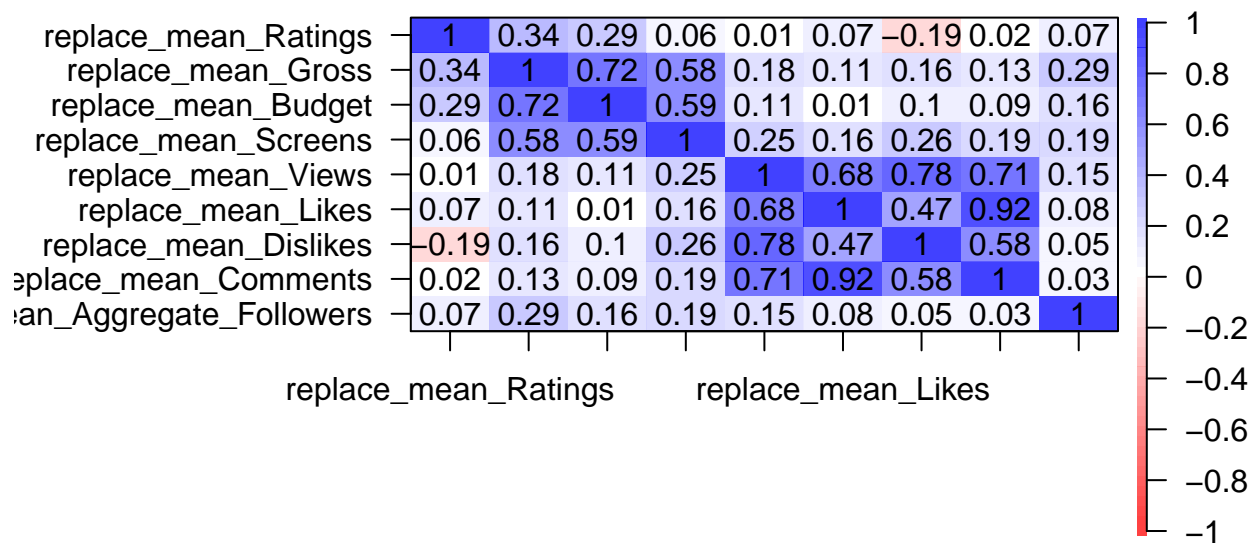


Lastly

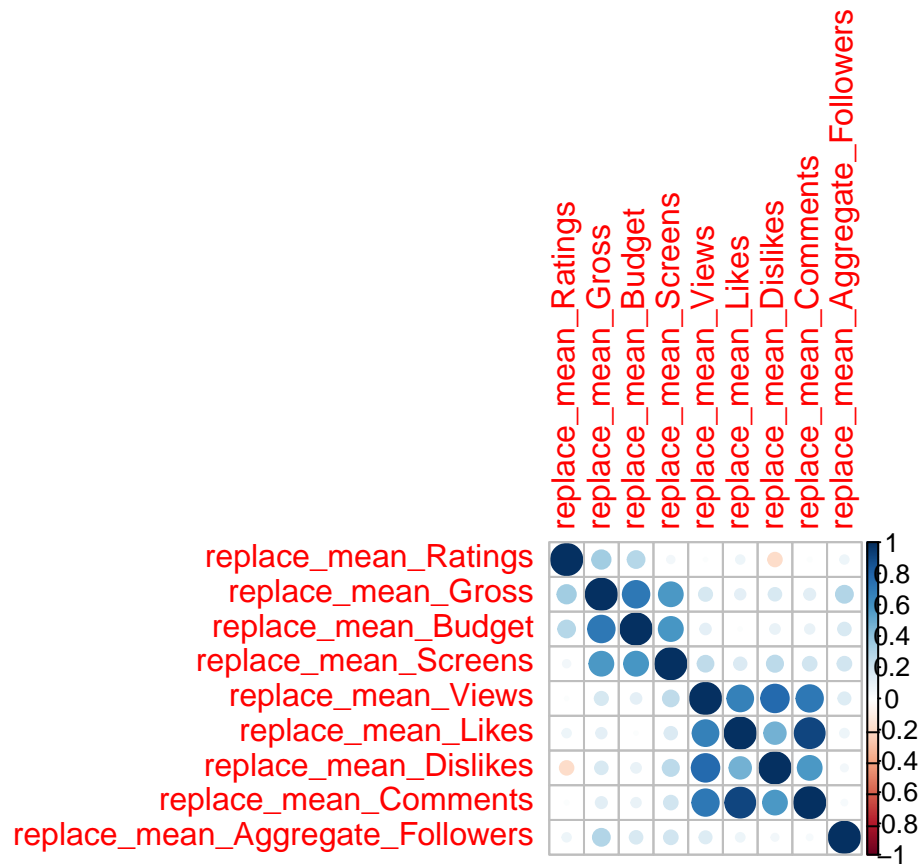
```
library(PerformanceAnalytics)
chart.Correlation(movie_data,
  histogram=TRUE,
  pch=19,
  col = "blue")
```



```
## ***** ## correlation plot matrices
## *****
correl <- cor(movie_data)
psych::cor.plot(correl)
```

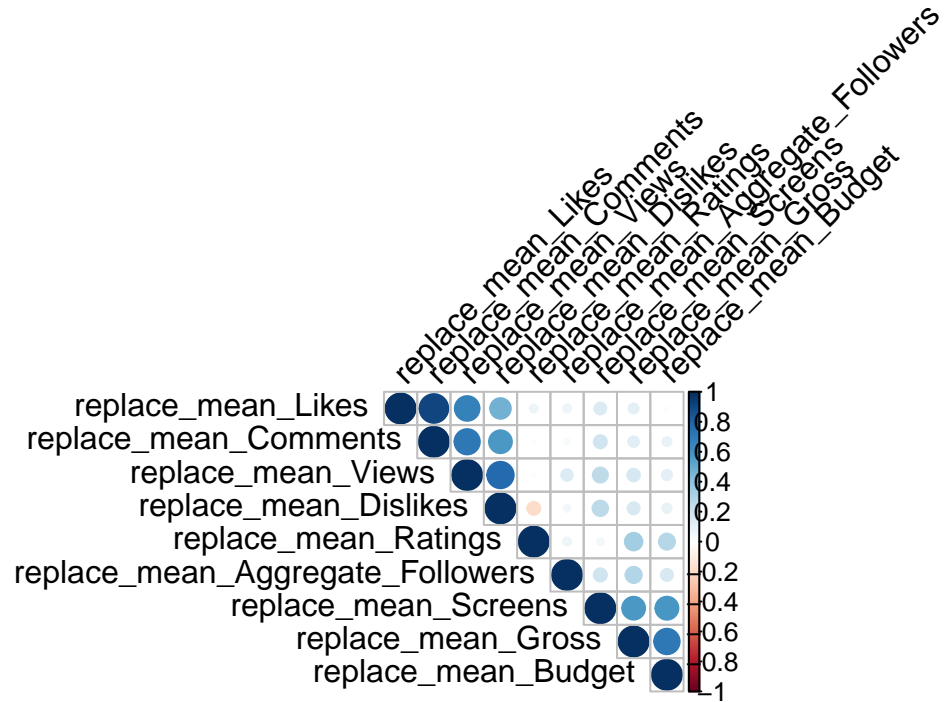


```
corrplot::corrplot(correl)
```



Tweak the knobs

```
corrplot(correl, type = "upper",
  order = "hclust",
  tl.col = "black",
  tl.srt = 45)
```

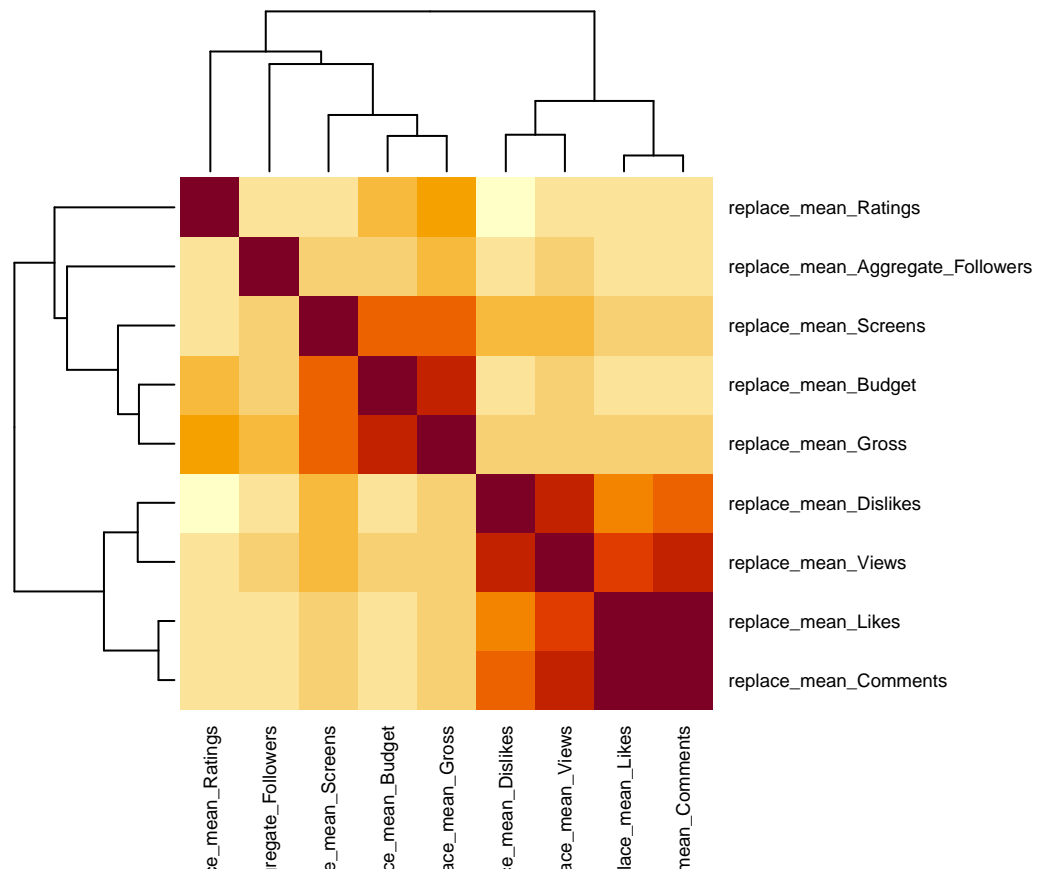


check

```
?corrplot
```

Heatmap

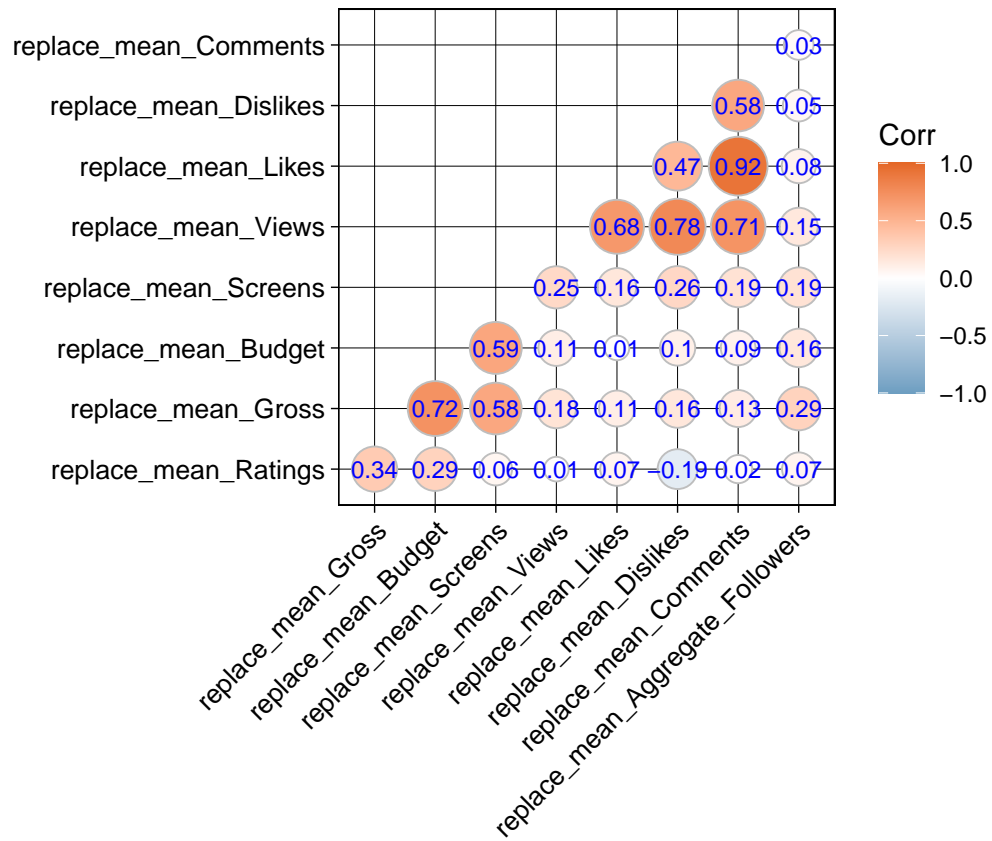
```
heatmap(correl, symm = TRUE,
        cexRow = 0.7,
        cexCol = 0.7)
```



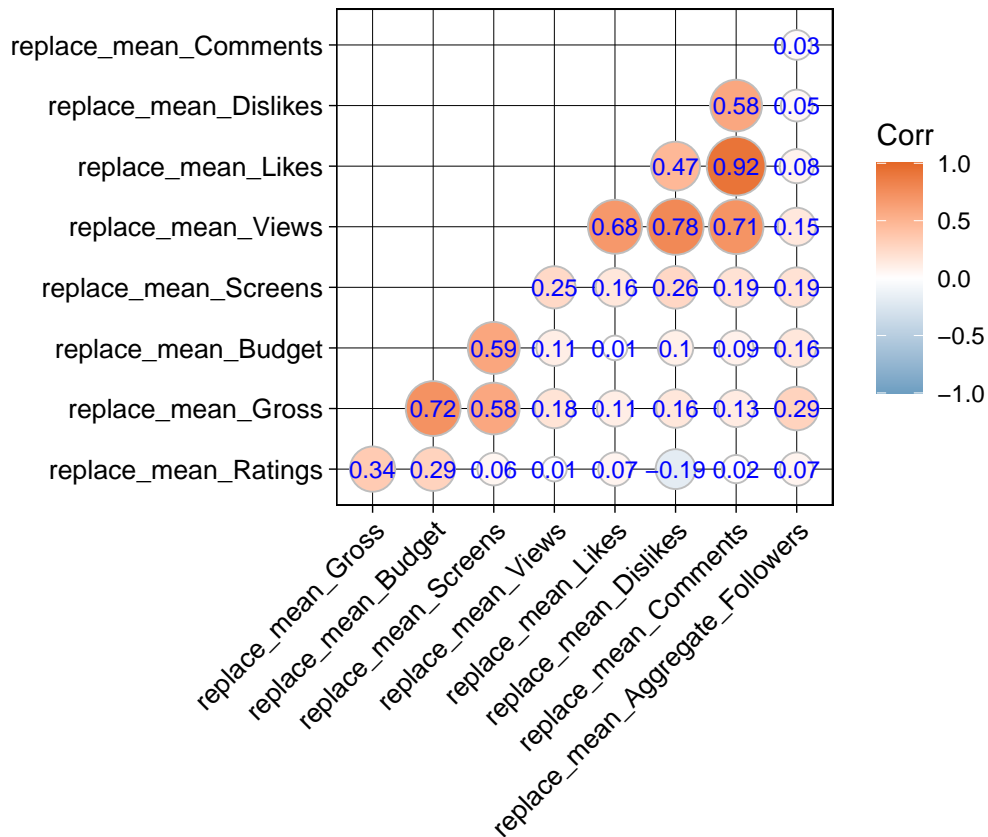
ggcorrplot

```
p <- ggcorrplot::ggcorrplot(correl,
  method = "circle",
  type = "lower",
  ggtheme = ggplot2::theme_linedraw,
  lab_col = "blue",
  lab_size = 3,
  tl.cex = 10,
  lab = TRUE,
  pch.cex = 10,
  colors = c("#6D9EC1", "white", "#E46726"))
```

p



```
p + guides(scale = "none")
```



----- ## Fitting Multiple Regression - ## -----

```
model <- lm(replace_mean_Ratings ~ replace_mean_Gross+ replace_mean_Budget+replace_mean_Screens+
  replace_mean_Views+ replace_mean_Likes+replace_mean_Dislikes+replace_mean_Comments+
  replace_mean_Aggregate_Followers, data = movie_data)
```

model

##

Call:

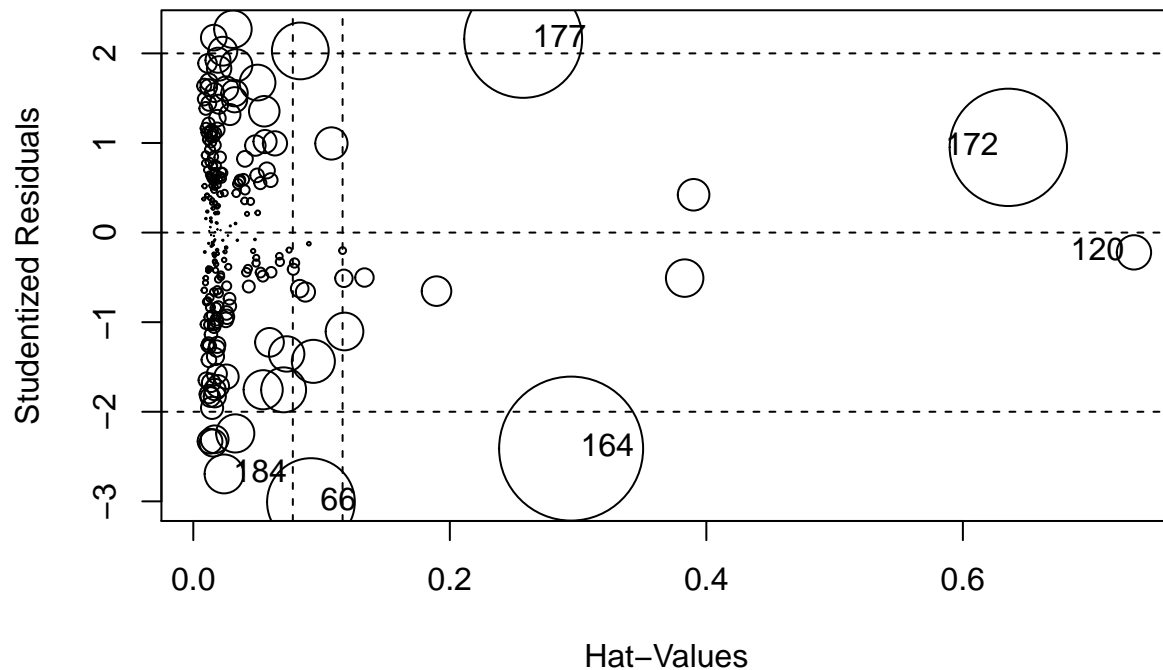
```
## lm(formula = replace_mean_Ratings ~ replace_mean_Gross + replace_mean_Budget +
##   replace_mean_Screens + replace_mean_Views + replace_mean_Likes +
##   replace_mean_Dislikes + replace_mean_Comments + replace_mean_Aggregate_Followers,
##   data = movie_data)
```

##

Coefficients:

```
##           (Intercept)           replace_mean_Gross
##           6.352e+00           4.102e-09
##   replace_mean_Budget   replace_mean_Screens
##           3.275e-09           -1.476e-04
##   replace_mean_Views   replace_mean_Likes
##           6.565e-08           8.255e-06
##   replace_mean_Dislikes   replace_mean_Comments
##           -3.554e-04           -4.943e-05
## replace_mean_Aggregate_Followers
##           -1.243e-08
```

```
influencePlot(model)
```



```
##      StudRes      Hat      CookD
## 66 -3.0078772 0.09171501 0.09797153
## 120 -0.2211173 0.73339231 0.01500800
## 164 -2.4113009 0.29456210 0.26405971
## 172 0.9512353 0.63542922 0.17530894
## 177 2.1615251 0.25718316 0.17682573
## 184 -2.6939684 0.02397499 0.01926732
```

```
##----- ## Checking lm objects - ##-----
```

```
----- # Show the components of lm object
```

```
str(model)
```

```
## List of 12
## $ coefficients : Named num [1:9] 6.35 4.10e-09 3.27e-09 -1.48e-04 6.57e-08 ...
## $ residuals : Named num [1:232] -0.1156 0.4021 0.0519 -0.3016 -1.4062 ...
## $ effects : Named num [1:232] -98.115 -5.131 0.904 3.03 -0.113 ...
## $ rank : int 9
## $ fitted.values: Named num [1:232] 6.42 6.7 6.15 6.6 6.11 ...
## $ assign : int [1:9] 0 1 2 3 4 5 6 7 8
## $ qr :List of 5
```

```

## ..$ qr      : num [1:232, 1:9] -15.2315 0.0657 0.0657 0.0657 0.0657 ...
## .. ..- attr(*, "dimnames")=List of 2
## .. .. ..$ : chr [1:232] "1" "2" "3" "4" ...
## .. .. ..$ : chr [1:9] "(Intercept)" "replace_mean_Gross" "replace_mean_Budget" "replace_mean_Screens"
## .. ..- attr(*, "assign")= int [1:9] 0 1 2 3 4 5 6 7 8
## ..$ qraux: num [1:9] 1.07 1.1 1.01 1.04 1.01 ...
## ..$ pivot: int [1:9] 1 2 3 4 5 6 7 8 9
## ..$ tol    : num 1e-07
## ..$ rank   : int 9
## ..- attr(*, "class")= chr "qr"
## $ df.residual : int 223
## $ xlevels      : Named list()
## $ call         : language lm(formula = replace_mean_Ratings ~ replace_mean_Gross + replace_mean_Bud
## $ terms        :Classes 'terms', 'formula' language replace_mean_Ratings ~ replace_mean_Gross + rep
## .. ..- attr(*, "variables")= language list(replace_mean_Ratings, replace_mean_Gross, replace_mean_B
## .. ..- attr(*, "factors")= int [1:9, 1:8] 0 1 0 0 0 0 0 0 0 ...
## .. .. ..- attr(*, "dimnames")=List of 2
## .. .. .. ..$ : chr [1:9] "replace_mean_Ratings" "replace_mean_Gross" "replace_mean_Budget" "replace
## .. .. .. ..$ : chr [1:8] "replace_mean_Gross" "replace_mean_Budget" "replace_mean_Screens" "replace
## .. ..- attr(*, "term.labels")= chr [1:8] "replace_mean_Gross" "replace_mean_Budget" "replace_mean_S
## .. ..- attr(*, "order")= int [1:8] 1 1 1 1 1 1 1 1
## .. ..- attr(*, "intercept")= int 1
## .. ..- attr(*, "response")= int 1
## .. ..- attr(*, ".Environment")=<environment: R_GlobalEnv>
## .. ..- attr(*, "predvars")= language list(replace_mean_Ratings, replace_mean_Gross, replace_mean_B
## .. ..- attr(*, "dataClasses")= Named chr [1:9] "numeric" "numeric" "numeric" "numeric" ...
## .. .. ..- attr(*, "names")= chr [1:9] "replace_mean_Ratings" "replace_mean_Gross" "replace_mean_Bu
## $ model        :'data.frame': 232 obs. of 9 variables:
## ..$ replace_mean_Ratings : num [1:232] 6.3 7.1 6.2 6.3 4.7 4.6 6.1 7.1 6.5 6.1 ...
## ..$ replace_mean_Gross   : num [1:232] 9.13e+03 1.92e+08 3.07e+07 1.06e+08 1.73e+07 2.9
## ..$ replace_mean_Budget  : num [1:232] 4.00e+06 5.00e+07 2.80e+07 1.10e+08 3.50e+06 5.0
## ..$ replace_mean_Screens : num [1:232] 45 3306 2872 3470 2310 ...
## ..$ replace_mean_Views   : num [1:232] 3280543 583289 304861 452917 3145573 ...
## ..$ replace_mean_Likes   : num [1:232] 4632 3465 328 2429 12163 ...
## ..$ replace_mean_Dislikes : num [1:232] 425 61 34 132 610 7 419 197 419 532 ...
## ..$ replace_mean_Comments : num [1:232] 636 186 47 590 1082 ...
## ..$ replace_mean_Aggregate_Followers: num [1:232] 1120000 12350000 483000 568000 1923800 ...
## ..- attr(*, "terms")=Classes 'terms', 'formula' language replace_mean_Ratings ~ replace_mean_Gross
## .. .. ..- attr(*, "variables")= language list(replace_mean_Ratings, replace_mean_Gross, replace_me
## .. .. ..- attr(*, "factors")= int [1:9, 1:8] 0 1 0 0 0 0 0 0 0 ...
## .. .. .. ..- attr(*, "dimnames")=List of 2
## .. .. .. .. ..$ : chr [1:9] "replace_mean_Ratings" "replace_mean_Gross" "replace_mean_Budget" "rep
## .. .. .. .. ..$ : chr [1:8] "replace_mean_Gross" "replace_mean_Budget" "replace_mean_Screens" "rep
## .. .. ..- attr(*, "term.labels")= chr [1:8] "replace_mean_Gross" "replace_mean_Budget" "replace_me
## .. .. ..- attr(*, "order")= int [1:8] 1 1 1 1 1 1 1 1
## .. .. ..- attr(*, "intercept")= int 1
## .. .. ..- attr(*, "response")= int 1
## .. .. ..- attr(*, ".Environment")=<environment: R_GlobalEnv>
## .. .. ..- attr(*, "predvars")= language list(replace_mean_Ratings, replace_mean_Gross, replace_mean
## .. .. ..- attr(*, "dataClasses")= Named chr [1:9] "numeric" "numeric" "numeric" "numeric" ...
## .. .. .. ..- attr(*, "names")= chr [1:9] "replace_mean_Ratings" "replace_mean_Gross" "replace_mean
## - attr(*, "class")= chr "lm"

```

```

class(model)

## [1] "lm"

typeof(model)

## [1] "list"

length(model)

## [1] 12

names(model)

## [1] "coefficients" "residuals"      "effects"      "rank"
## [5] "fitted.values" "assign"          "qr"           "df.residual"
## [9] "xlevels"      "call"           "terms"        "model"

```

Summary Function

```

summary(model)

##
## Call:
## lm(formula = replace_mean_Ratings ~ replace_mean_Gross + replace_mean_Budget +
##      replace_mean_Screens + replace_mean_Views + replace_mean_Likes +
##      replace_mean_Dislikes + replace_mean_Comments + replace_mean_Aggregate_Followers,
##      data = movie_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.42601 -0.50733  0.00213  0.53345  1.90750
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    6.352e+00  1.117e-01  56.858  < 2e-16 ***
## replace_mean_Gross    4.102e-09  9.972e-10   4.113  5.49e-05 ***
## replace_mean_Budget    3.275e-09  1.654e-09   1.980  0.04888 *
## replace_mean_Screens   -1.476e-04  5.265e-05  -2.803  0.00551 **
## replace_mean_Views     6.565e-08  2.456e-08   2.674  0.00806 **
## replace_mean_Likes     8.255e-06  5.471e-06   1.509  0.13277
## replace_mean_Dislikes  -3.554e-04  7.817e-05  -4.547  8.94e-06 ***
## replace_mean_Comments  -4.943e-05  4.576e-05  -1.080  0.28116
## replace_mean_Aggregate_Followers -1.243e-08  1.351e-08  -0.920  0.35851
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.8614 on 223 degrees of freedom
## Multiple R-squared:  0.2641, Adjusted R-squared:  0.2377
## F-statistic:    10 on 8 and 223 DF,  p-value: 6.858e-12

```

Coefficients Function

```

model$coefficients

```

```
##                (Intercept)                replace_mean_Gross
##                6.351917e+00                4.101794e-09
##                replace_mean_Budget          replace_mean_Screens
##                3.274910e-09                -1.475589e-04
##                replace_mean_Views           replace_mean_Likes
##                6.565145e-08                8.254940e-06
##                replace_mean_Dislikes        replace_mean_Comments
##                -3.554223e-04                -4.943290e-05
## replace_mean_Aggregate_Followers
##                -1.242762e-08
```

Fitted function

```
model$fitted
```

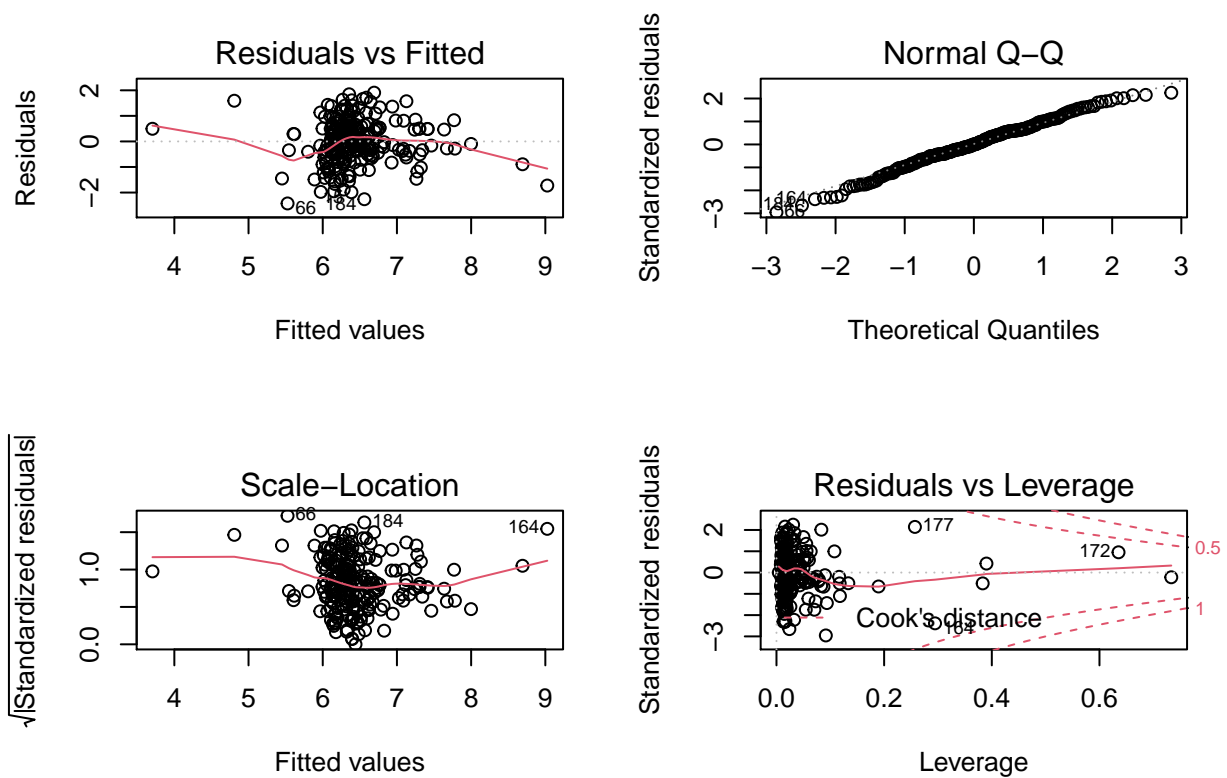
```
##          1          2          3          4          5          6          7          8
## 6.415610 6.697918 6.148062 6.601564 6.106193 6.028198 6.168008 6.359285
##          9         10         11         12         13         14         15         16
## 6.143993 6.202188 7.463003 6.252885 5.803006 6.172500 6.386971 6.570020
##         17         18         19         20         21         22         23         24
## 6.395214 6.494361 7.415767 6.154638 5.997289 6.391396 6.136607 6.398623
##         25         26         27         28         29         30         31         32
## 6.029688 6.251078 7.303404 7.238588 6.559774 5.973485 6.611769 6.118764
##         33         34         35         36         37         38         39         40
## 6.560516 6.125296 6.208010 5.966916 7.090697 6.473246 6.041684 6.519117
##         41         42         43         44         45         46         47         48
## 6.414630 6.395056 6.340912 6.320186 7.127557 6.539492 7.636159 6.305507
##         49         50         51         52         53         54         55         56
## 6.368349 6.474446 6.744861 7.410192 6.828164 6.381270 6.347684 7.127769
##         57         58         59         60         61         62         63         64
## 6.200761 6.654979 6.380380 6.329679 6.300267 6.334727 6.468455 6.484992
##         65         66         67         68         69         70         71         72
## 6.406157 5.526012 6.603536 6.056605 6.352874 6.517553 7.342074 6.383917
##         73         74         75         76         77         78         79         80
## 6.183176 6.277514 6.601079 6.133733 6.324117 6.107544 6.464840 7.322481
##         81         82         83         84         85         86         87         88
## 6.268766 6.263169 5.615677 6.447873 5.544391 6.282380 6.309471 6.068041
##         89         90         91         92         93         94         95         96
## 7.170251 6.102397 6.363173 6.646722 6.672575 6.136643 6.269107 6.067665
##         97         98         99        100        101        102        103        104
## 6.180593 6.247109 6.503405 6.299163 6.177905 6.224032 6.050183 6.518881
##        105        106        107        108        109        110        111        112
## 6.767730 6.928448 6.256379 6.062226 6.172554 6.102490 7.086462 6.385266
##        113        114        115        116        117        118        119        120
## 6.783104 6.605192 6.267857 6.013118 6.436791 6.744785 7.276007 7.998561
##        121        122        123        124        125        126        127        128
## 6.446846 6.676400 6.484438 6.361108 7.781289 6.414842 6.344593 7.307527
##        129        130        131        132        133        134        135        136
## 6.059643 6.769459 6.327271 6.267855 6.152350 6.983827 6.165718 6.942330
##        137        138        139        140        141        142        143        144
## 6.446646 6.161567 6.162020 6.375324 6.200786 6.255102 6.236031 6.395732
##        145        146        147        148        149        150        151        152
## 6.363519 6.711586 6.366511 6.079280 6.113825 6.639536 7.262706 6.280902
```

```
##      153      154      155      156      157      158      159      160
## 6.716840 6.288721 6.398473 6.692504 6.322557 6.563132 6.356152 6.110241
##      161      162      163      164      165      166      167      168
## 6.390449 7.249492 6.415857 9.026086 8.692719 7.673017 7.770060 7.117034
##      169      170      171      172      173      174      175      176
## 7.074630 7.299371 7.025682 3.705130 6.608188 6.637077 6.937519 6.588301
##      177      178      179      180      181      182      183      184
## 4.808252 6.724003 6.131516 6.820041 6.607260 6.623534 6.514045 6.561171
##      185      186      187      188      189      190      191      192
## 6.020573 6.177524 6.416834 6.188921 5.451059 6.084756 6.621738 6.763012
##      193      194      195      196      197      198      199      200
## 6.206773 6.222751 6.324577 6.162415 6.362724 6.414850 6.477763 6.141128
##      201      202      203      204      205      206      207      208
## 6.183281 6.048400 6.197925 6.286273 6.144238 5.606685 6.118272 5.997360
##      209      210      211      212      213      214      215      216
## 6.267960 6.166252 6.290084 6.274108 6.645815 6.162831 6.426126 6.356366
##      217      218      219      220      221      222      223      224
## 6.331433 6.400079 6.583394 6.299967 6.381138 5.968432 6.360047 6.264855
##      225      226      227      228      229      230      231      232
## 6.318563 6.277765 6.473365 6.214593 6.207979 6.022385 5.889451 6.441558
```

```
extract_eq(model, use_cof=TRUE)
```

$\text{replace_mean_Ratings} = \alpha + \beta_1(\text{replace_mean_Gross}) + \beta_2(\text{replace_mean_Budget}) + \beta_3(\text{replace_mean_Screens}) + \beta_4(\text{replace_mean_Ratings})$

```
par(mfrow = c(2, 2))
plot(model)
```



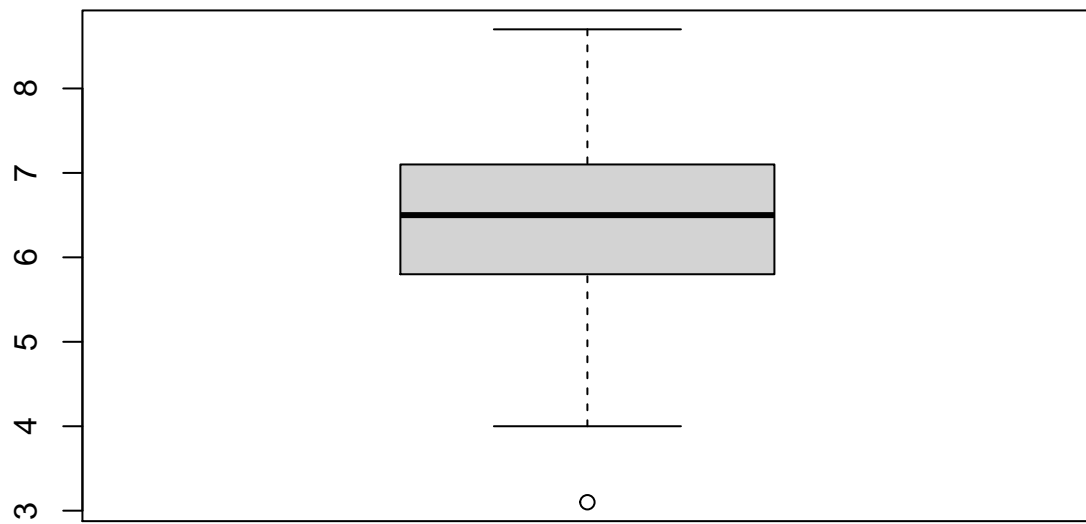
```
vif(model)
```

```
##           replace_mean_Gross           replace_mean_Budget
##           2.435939           2.487005
##           replace_mean_Screens           replace_mean_Views
##           1.760608           3.803236
##           replace_mean_Likes           replace_mean_Dislikes
##           7.709133           2.930766
##           replace_mean_Comments replace_mean_Aggregate_Followers
##           8.275607           1.144607
```

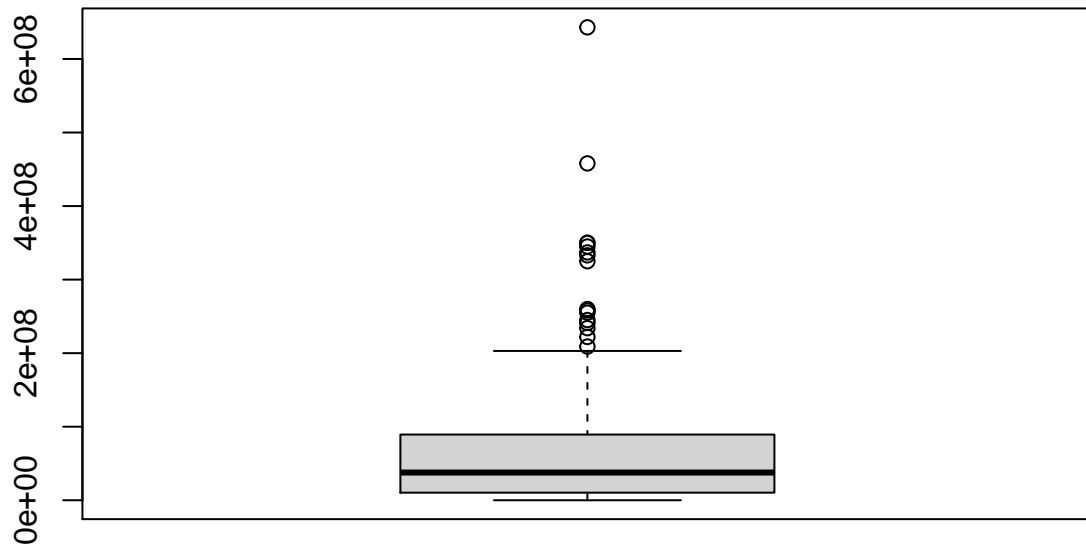
We will improve our result by removing the outlier

```
##create a boxplot to identify outliers
```

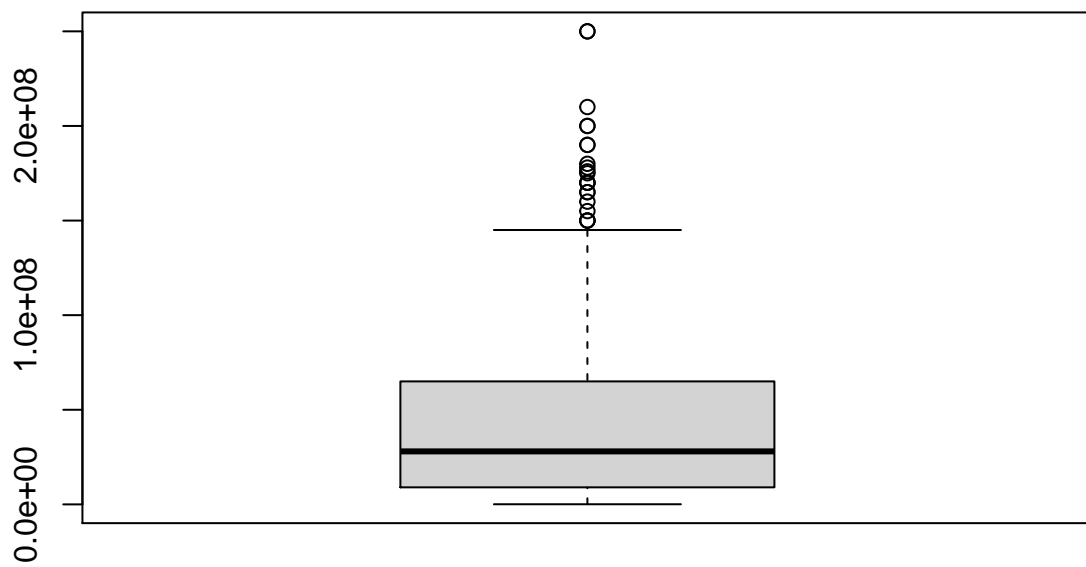
```
boxplot(movie_data$replace_mean_Ratings)
```

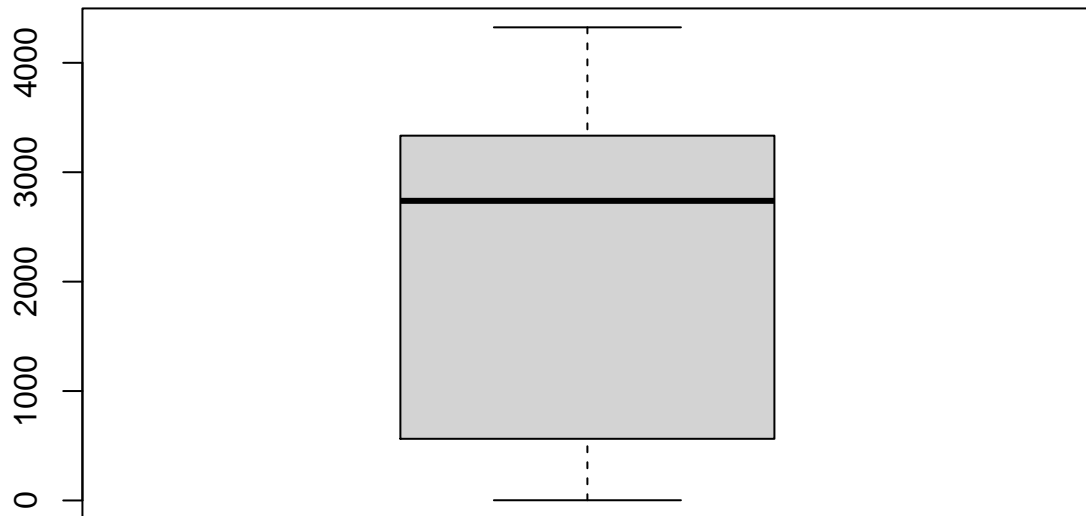
```
boxplot(movie_data$replace_mean_Gross)
```



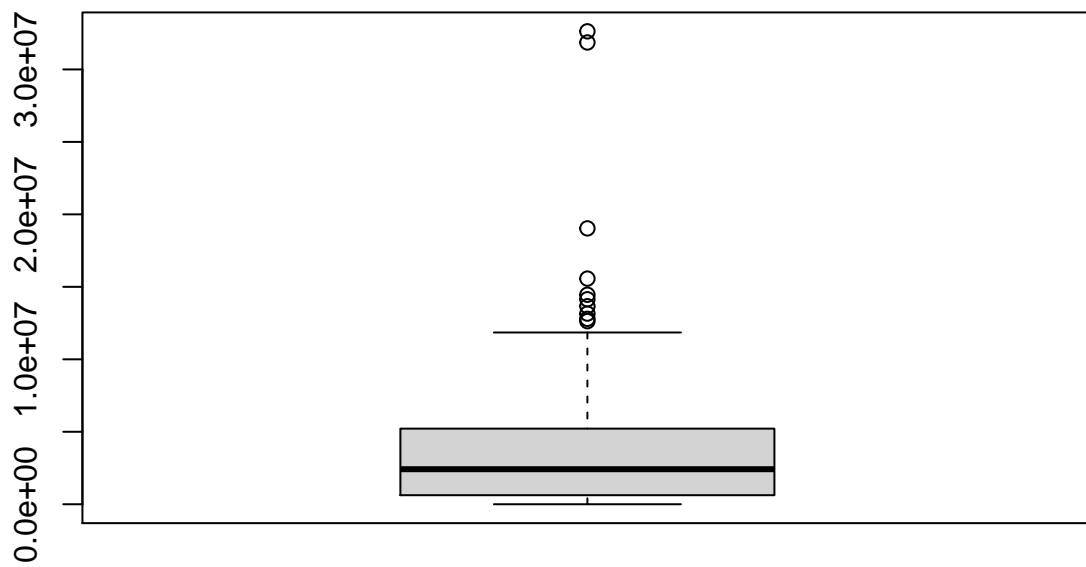
```
boxplot(movie_data$replace_mean_Budget)
```



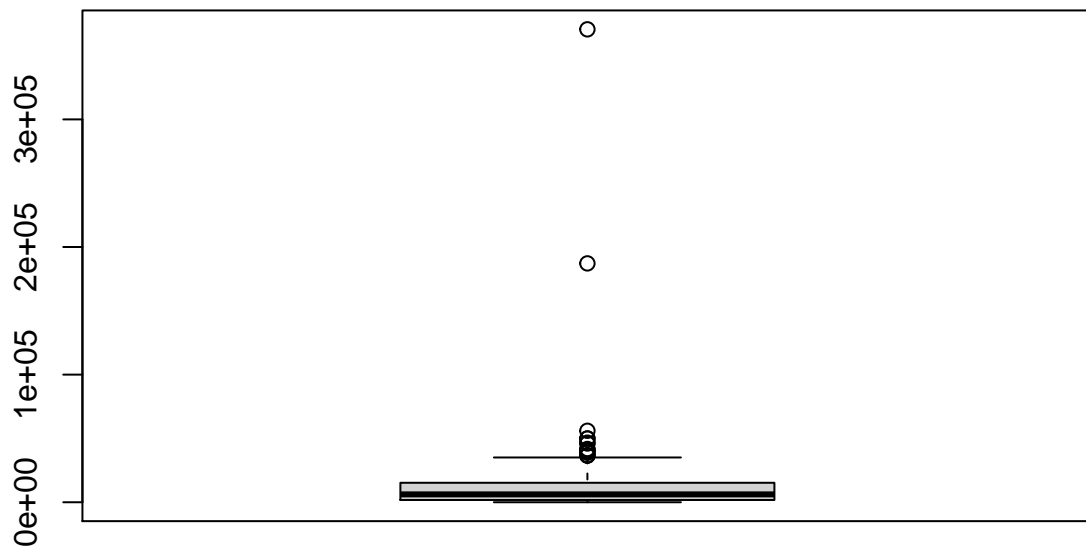
```
boxplot(movie_data$replace_mean_Screens)
```



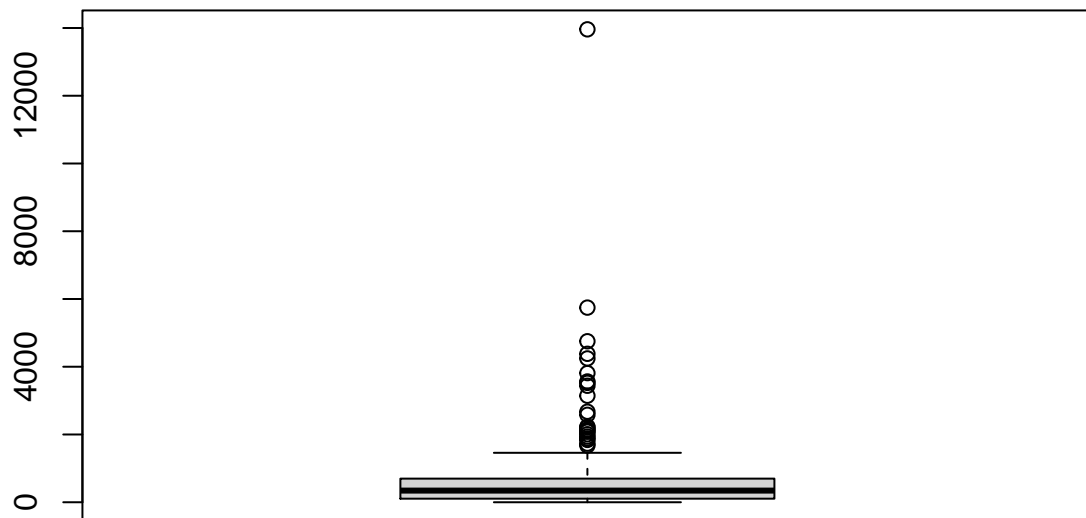
```
boxplot(movie_data$replace_mean_Views)
```



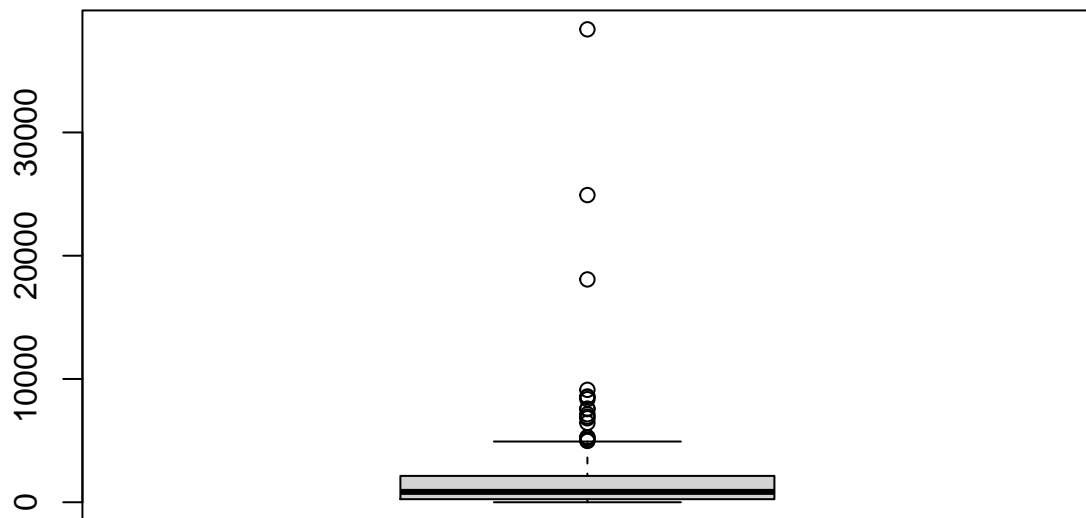
```
boxplot(movie_data$replace_mean_Likes)
```



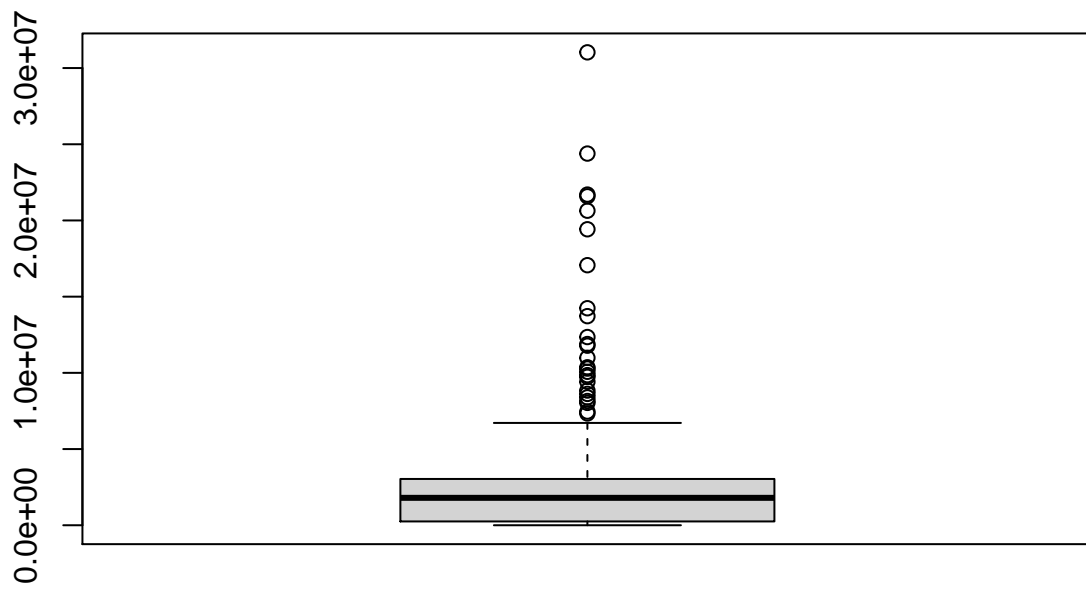
```
boxplot(movie_data$replace_mean_Dislikes)
```



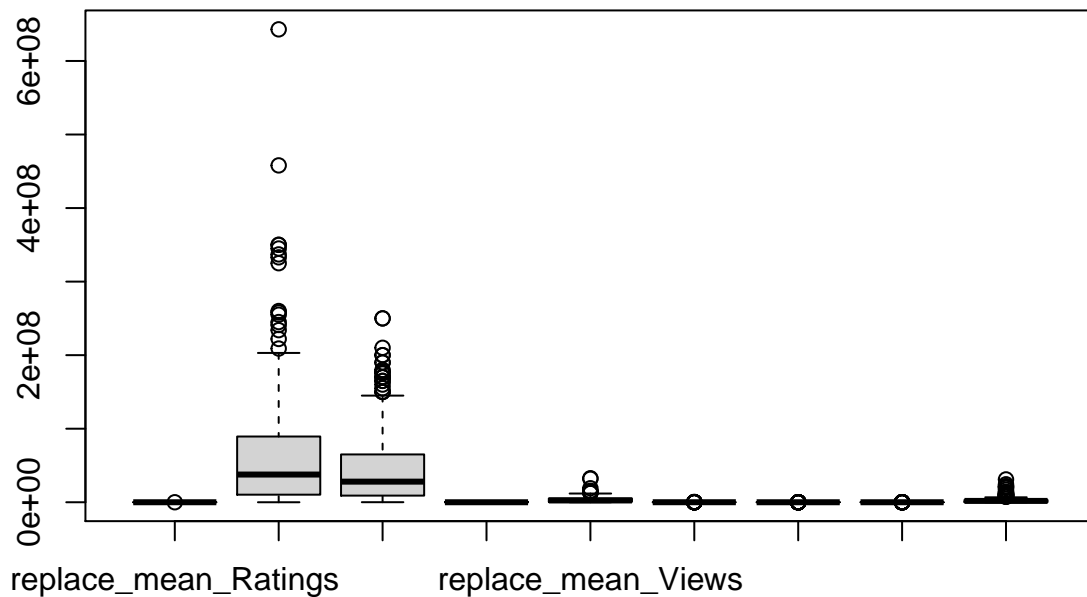
```
boxplot(movie_data$replace_mean_Comments)
```



```
boxplot(movie_data$replace_mean_Aggregate_Followers)
```

```
boxplot(movie_data)
boxplot(movie_data)$out
```

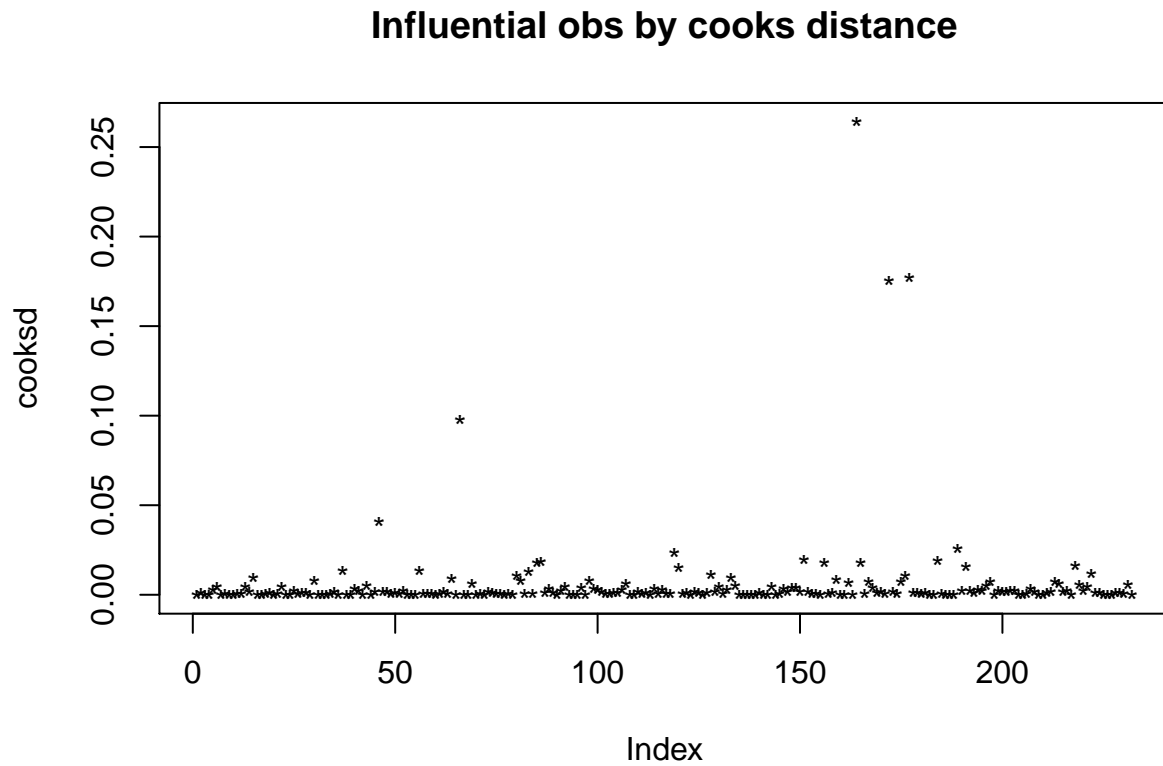


```
## [1] 3.1 350000000.0 222000000.0 260000000.0 209000000.0 333000000.0
## [7] 241000000.0 255000000.0 337000000.0 258000000.0 245000000.0 234000000.0
## [13] 643000000.0 458000000.0 350000000.0 345000000.0 325000000.0 165000000.0
## [19] 170000000.0 170000000.0 178000000.0 160000000.0 170000000.0 165000000.0
## [25] 180000000.0 200000000.0 250000000.0 210000000.0 200000000.0 150000000.0
## [31] 250000000.0 190000000.0 175000000.0 150000000.0 150000000.0 190000000.0
## [37] 155000000.0 176000000.0 19032902.0 15568277.0 13661095.0 14453673.0
## [43] 32626778.0 14141585.0 31859569.0 12632836.0 12792898.0 13154873.0
## [49] 38810.0 41728.0 46023.0 41254.0 187162.0 370552.0
## [55] 36646.0 39824.0 49900.0 36508.0 50002.0 56001.0
## [61] 36874.0 46684.0 40496.0 4382.0 1940.0 1730.0
## [67] 2111.0 3812.0 3145.0 2150.0 3439.0 4752.0
## [73] 2233.0 1675.0 13960.0 1840.0 2210.0 5746.0
## [79] 2581.0 2007.0 3524.0 2083.0 1885.0 3565.0
## [85] 2672.0 4245.0 5278.0 6439.0 6946.0 7595.0
## [91] 18077.0 24919.0 4973.0 8533.0 5005.0 38363.0
## [97] 6811.0 5107.0 9119.0 7559.0 7139.0 8578.0
## [103] 8359.0 5262.0 12350000.0 8153000.0 21697300.0 19420105.0
## [109] 8030000.0 17064000.0 13720000.0 20640000.0 9414000.0 8839043.0
## [115] 9850000.0 14240000.0 21586000.0 8620000.0 11783000.0 10364000.0
## [121] 31030000.0 24388000.0 10280000.0 10070000.0 11890000.0 7336000.0
## [127] 9737600.0 10988000.0 8392000.0 7460000.0
```

###How to Remove Outliers from Multiple Columns in R

```
cooksdc<-cooks.distance((model))
```

```
plot(cooksd,pch="*",cex=1,main="Influential obs by cooks distance")
```



```
cooksd
##          1          2          3          4          5          6
## 3.199881e-05 1.061405e-03 5.559492e-06 2.510955e-04 3.063783e-03 4.357536e-03
##          7          8          9         10         11         12
## 1.210919e-05 8.086688e-04 2.080758e-04 1.702180e-04 5.932023e-04 3.979801e-04
##         13         14         15         16         17         18
## 4.321734e-03 2.003389e-03 9.430120e-03 2.641235e-06 2.080530e-05 4.138818e-04
##         19         20         21         22         23         24
## 1.362723e-03 2.663981e-04 7.652797e-04 4.382884e-03 2.828426e-04 2.232967e-05
##         25         26         27         28         29         30
## 2.059391e-03 4.604448e-04 1.445641e-03 1.188209e-03 3.484621e-04 7.802453e-03
##         31         32         33         34         35         36
## 1.052228e-04 2.100214e-04 2.565811e-04 8.923383e-04 1.512163e-03 8.043333e-05
##         37         38         39         40         41         42
## 1.326012e-02 2.307838e-04 2.109440e-04 3.260227e-03 2.089443e-03 6.373026e-08
##         43         44         45         46         47         48
## 5.125479e-03 1.246199e-04 1.763317e-03 4.098960e-02 1.874689e-03 1.506946e-03
##         49         50         51         52         53         54
## 8.047916e-04 9.452611e-04 6.028368e-04 2.452272e-03 3.396126e-06 2.059163e-04
##         55         56         57         58         59         60
## 2.049725e-04 1.325300e-02 5.403106e-04 7.295958e-04 7.198905e-04 3.662672e-05
##         61         62         63         64         65         66
## 6.079021e-04 1.562084e-03 5.945978e-04 8.886200e-03 2.459761e-05 9.797153e-02
```

##	67	68	69	70	71	72
##	1.981886e-05	2.271515e-04	6.248938e-03	3.239033e-05	8.145974e-04	1.759739e-05
##	73	74	75	76	77	78
##	1.793060e-03	1.088856e-03	5.805818e-04	4.410799e-04	1.244508e-06	5.632846e-04
##	79	80	81	82	83	84
##	3.344251e-05	1.050861e-02	7.975237e-03	6.946140e-04	1.269043e-02	5.818775e-04
##	85	86	87	88	89	90
##	1.787468e-02	1.854010e-02	1.405631e-03	3.318075e-03	1.413009e-03	1.018163e-04
##	91	92	93	94	95	96
##	2.376351e-03	4.697717e-03	1.927801e-04	1.839860e-05	2.171721e-04	3.834666e-03
##	97	98	99	100	101	102
##	3.354370e-04	8.072584e-03	3.429251e-03	2.648879e-03	1.977100e-03	5.415673e-04
##	103	104	105	106	107	108
##	8.784224e-04	1.409200e-03	9.533434e-04	2.748543e-03	6.330677e-03	3.438133e-04
##	109	110	111	112	113	114
##	2.904724e-05	1.504729e-03	6.651857e-04	9.196852e-04	1.307158e-04	3.158387e-03
##	115	116	117	118	119	120
##	5.694482e-04	2.738409e-03	6.379014e-04	6.044638e-04	2.358869e-02	1.500800e-02
##	121	122	123	124	125	126
##	8.890892e-04	1.016521e-03	4.625472e-07	1.966664e-03	1.102509e-03	9.199738e-05
##	127	128	129	130	131	132
##	1.474201e-03	1.113599e-02	2.032217e-03	4.553796e-03	7.450113e-04	2.814528e-03
##	133	134	135	136	137	138
##	9.815489e-03	5.327287e-03	2.715033e-05	8.188774e-05	2.419848e-04	4.036642e-05
##	139	140	141	142	143	144
##	2.380219e-04	1.255565e-03	1.184534e-04	1.731816e-04	4.369847e-03	4.812187e-08
##	145	146	147	148	149	150
##	4.918238e-04	3.463467e-03	1.804750e-03	3.727347e-03	3.864774e-03	1.582902e-03
##	151	152	153	154	155	156
##	1.951475e-02	1.515652e-03	5.767143e-04	5.560868e-04	6.820157e-05	1.790988e-02
##	157	158	159	160	161	162
##	8.581471e-04	9.556858e-04	8.392391e-03	2.325468e-05	2.180515e-04	6.794319e-03
##	163	164	165	166	167	168
##	1.178358e-04	2.640597e-01	1.808081e-02	8.665188e-04	7.486733e-03	3.948200e-03
##	169	170	171	172	173	174
##	1.227517e-03	1.590776e-03	4.580027e-04	1.753089e-01	1.826302e-03	4.569832e-04
##	175	176	177	178	179	180
##	7.486957e-03	1.058590e-02	1.768257e-01	1.131626e-03	1.158710e-03	5.589147e-04
##	181	182	183	184	185	186
##	1.056423e-03	2.127828e-04	3.898227e-05	1.926732e-02	5.933617e-04	1.681899e-06
##	187	188	189	190	191	192
##	1.782805e-04	2.608060e-05	2.568686e-02	2.271504e-03	1.595054e-02	2.377370e-03
##	193	194	195	196	197	198
##	1.263146e-03	2.762880e-03	2.120557e-03	4.967029e-03	7.576990e-03	2.859548e-04
##	199	200	201	202	203	204
##	2.058958e-03	1.844200e-03	1.682653e-03	2.051455e-03	2.183725e-03	4.747076e-05
##	205	206	207	208	209	210
##	1.963914e-04	6.302017e-04	3.515945e-03	2.026057e-03	1.281909e-04	2.402708e-04
##	211	212	213	214	215	216
##	1.405737e-03	1.918368e-03	7.580368e-03	6.233217e-03	1.941239e-03	2.283365e-03
##	217	218	219	220	221	222
##	2.453915e-06	1.628784e-02	5.651963e-03	2.539792e-03	4.582412e-03	1.188407e-02
##	223	224	225	226	227	228
##	1.133999e-03	9.379196e-04	2.061179e-04	1.291130e-04	1.948335e-05	1.049369e-03

```
##          229          230          231          232
## 1.258997e-03 7.626064e-04 5.807478e-03 9.438684e-35

influential<-cooksds[(cooksds>(3*mean(cooksds,na.rm=TRUE)))]

names_of_influential<-names(influential)
names_of_influential

## [1] "46" "66" "86" "119" "151" "164" "172" "177" "184" "189"

outliers <- movie_data[names_of_influential,]
movie_data_without_outliers <-movie_data %>% anti_join(outliers)
movie_data_without_outliers
```

##	replace_mean_Ratings	replace_mean_Gross	replace_mean_Budget
## 1	6.300000	9130	4000000
## 2	7.100000	192000000	50000000
## 3	6.200000	30700000	28000000
## 4	6.300000	106000000	110000000
## 5	4.700000	17300000	3500000
## 6	4.600000	29000	500000
## 7	6.100000	42600000	40000000
## 8	7.100000	5750000	20000000
## 9	6.500000	26000000	28000000
## 10	6.100000	48600000	12500000
## 11	7.300000	350000000	58800000
## 12	5.700000	15200000	30000000
## 13	5.400000	84300000	6500000
## 14	5.200000	85900000	65000000
## 15	4.400000	830000	5000000
## 16	6.600000	11800	6000000
## 17	6.300000	72300	5000000
## 18	6.900000	14600000	7000000
## 19	7.900000	222000000	165000000
## 20	6.600000	21600000	9000000
## 21	6.500000	46300000	40000000
## 22	8.000000	25400000	4000000
## 23	5.700000	20300000	28000000
## 24	6.500000	1870000	7000000
## 25	7.000000	9840	1000000
## 26	6.700000	6370000	3000000
## 27	7.800000	260000000	170000000
## 28	7.700000	209000000	170000000
## 29	6.200000	30500000	30000000
## 30	4.000000	15800000	7000000
## 31	6.800000	151000000	85000000
## 32	6.400000	42000000	36000000
## 33	6.300000	55900000	70000000
## 34	6.800000	28800000	27000000
## 35	5.800000	86200000	35000000
## 36	5.800000	38900000	13000000
## 37	7.900000	100000000	178000000
## 38	6.300000	23400000	20000000
## 39	5.700000	8690	4500000
## 40	7.100000	1210000	24000000

## 41	7.600000	85700000	68000000
## 42	6.400000	102000	1850000
## 43	5.000000	60800000	2000000
## 44	6.100000	104000	1000000
## 45	6.500000	201000000	160000000
## 46	8.100000	333000000	170000000
## 47	5.600000	30100	70000
## 48	5.800000	91400000	12000000
## 49	6.100000	72700000	100000000
## 50	6.300000	54400000	42000000
## 51	7.900000	177000000	145000000
## 52	6.800000	50500000	11000000
## 53	6.100000	275000	8932780
## 54	6.700000	8090000	20000000
## 55	8.700000	188000000	165000000
## 56	5.800000	47600000	50000000
## 57	6.000000	128000000	50000000
## 58	6.900000	14700000	2000000
## 59	6.200000	50500000	60000000
## 60	6.900000	47000000	40000000
## 61	7.200000	43000000	20000000
## 62	7.000000	2450000	5000000
## 63	7.800000	128000000	81000000
## 64	6.500000	129000	25000000
## 65	6.500000	82400000	17000000
## 66	5.700000	8300	2400000
## 67	4.800000	2470	300000
## 68	6.400000	127000000	40000000
## 69	7.000000	241000000	180000000
## 70	6.300000	348000	15000000
## 71	7.100000	36400000	25000000
## 72	5.400000	10400000	5000000
## 73	6.900000	112000000	145000000
## 74	6.500000	51200000	50000000
## 75	6.300000	4010000	5000000
## 76	6.600000	43600000	66000000
## 77	6.400000	150000000	18000000
## 78	6.300000	114000000	127000000
## 79	7.900000	32300000	8500000
## 80	5.600000	52500000	13200000
## 81	5.900000	101000000	125000000
## 82	7.000000	91400000	50000000
## 83	5.200000	35700	600000
## 84	7.200000	76100000	55000000
## 85	5.000000	32500000	5000000
## 86	6.800000	83300000	132000000
## 87	5.900000	59200000	50000000
## 88	5.600000	23200000	100000000
## 89	6.100000	134000000	25000000
## 90	6.400000	132000000	103000000
## 91	6.200000	58600000	100000000
## 92	6.600000	3090000	5000000
## 93	7.500000	37400	5000000
## 94	5.700000	10500000	35000000

## 95	7.500000	52100000	20000000
## 96	5.400000	176000	30000000
## 97	5.500000	17200000	95000000
## 98	5.100000	38500000	40000000
## 99	6.600000	13800000	65000000
## 100	5.500000	59700000	22000000
## 101	5.800000	20200	6000000
## 102	7.300000	44100000	13000000
## 103	6.100000	89300000	48000000
## 104	4.800000	84500000	20000000
## 105	5.900000	191000000	125000000
## 106	6.100000	26000000	8000000
## 107	7.100000	2590000	9500000
## 108	6.800000	203000000	200000000
## 109	6.900000	924000	2000000
## 110	6.600000	26800000	26000000
## 111	7.300000	50200000	50000000
## 112	6.800000	50800000	60000000
## 113	4.800000	4240	7000000
## 114	7.100000	10700000	12600000
## 115	7.200000	102000000	55000000
## 116	7.900000	125000000	12000000
## 117	7.000000	2820000	47921730
## 118	6.100000	33600000	25000000
## 119	6.500000	45100000	25000000
## 120	7.400000	2720000	20000000
## 121	7.500000	255000000	250000000
## 122	6.600000	2430000	16000000
## 123	7.300000	54200000	22000000
## 124	6.800000	337000000	125000000
## 125	5.000000	2820000	12000000
## 126	8.100000	91100000	14000000
## 127	6.700000	6110000	44000000
## 128	7.500000	47100000	50000000
## 129	4.200000	18800000	70000000
## 130	7.800000	258000000	60000000
## 131	6.300000	5980000	14000000
## 132	6.800000	102000000	34000000
## 133	6.100000	78000000	70000000
## 134	6.300000	25000000	15000000
## 135	5.800000	64200000	42000000
## 136	7.100000	512000	100000
## 137	6.000000	83900000	40000000
## 138	6.500000	71500000	9000000
## 139	4.700000	2750000	6500000
## 140	6.400000	1110000	12250000
## 141	6.800000	5000	11712311
## 142	7.800000	35900000	15000000
## 143	7.200000	4190000	22500000
## 144	4.900000	26500000	1000000
## 145	5.700000	65200000	24000000
## 146	6.300000	23000000	100000000
## 147	5.400000	1820000	2800000
## 148	7.200000	116000000	65000000

## 149	6.800000	3320000	6000000
## 150	6.600000	30100000	15000000
## 151	8.600000	13100000	3300000
## 152	6.900000	280000	2487820
## 153	7.200000	37900000	15000000
## 154	8.200000	3080000	3300000
## 155	6.200000	22500	60000000
## 156	6.700000	3590000	6000000
## 157	8.100000	234000000	200000000
## 158	6.200000	717000	12000000
## 159	7.800000	458000000	250000000
## 160	7.400000	350000000	190000000
## 161	8.600000	345000000	175000000
## 162	6.600000	325000000	74000000
## 163	6.700000	183000000	29000000
## 164	7.800000	171000000	150000000
## 165	6.700000	177000000	135000000
## 166	6.100000	162000000	74000000
## 167	6.400000	154000000	110000000
## 168	8.300000	153000000	150000000
## 169	8.300000	135000000	28000000
## 170	7.300000	110000000	65000000
## 171	6.700000	105000000	35000000
## 172	6.600000	93200000	190000000
## 173	6.100000	90400000	40000000
## 174	6.800000	89400000	155000000
## 175	6.600000	81300000	68000000
## 176	5.600000	71000000	88000000
## 177	6.200000	65900000	14800000
## 178	6.700000	64500000	23000000
## 179	6.300000	54900000	31000000
## 180	5.000000	47400000	35000000
## 181	5.500000	47400000	176000000
## 182	7.300000	42500000	25000000
## 183	7.000000	41900000	20000000
## 184	7.200000	37400000	34000000
## 185	7.600000	36100000	5000000
## 186	4.600000	35400000	4000000
## 187	5.000000	34500000	35000000
## 188	6.600000	34000000	8500000
## 189	7.400000	33300000	11000000
## 190	7.100000	32400000	30000000
## 191	6.900000	31400000	12000000
## 192	5.200000	25800000	3300000
## 193	5.500000	19000000	10000000
## 194	6.100000	23500000	18000000
## 195	6.400000	22300000	12000000
## 196	5.900000	16100000	35000000
## 197	7.500000	16800000	700000
## 198	5.100000	12300000	14000000
## 199	6.500000	10900000	28000000
## 200	5.800000	10600000	40000000
## 201	5.500000	7610000	60000000
## 202	5.400000	7100000	70000000

## 203	8.200000	6740000	8000000
## 204	4.500000	1710000	8495000
## 205	5.600000	131000	4500000
## 206	7.300000	129000	4000000
## 207	6.300000	106000	1500000
## 208	7.800000	169000000	130000000
## 209	7.700000	49500000	30000000
## 210	7.700000	25400000	15000000
## 211	7.600000	34600000	75000000
## 212	7.100000	201000000	95000000
## 213	7.000000	31600000	49000000
## 214	6.900000	444000	1900000
## 215	6.600000	26400000	50000000
## 216	6.600000	33100000	10000000
## 217	6.400000	1210000	50000000
## 218	5.500000	21000000	37000000
## 219	5.400000	10200000	35000000
## 220	5.400000	12300000	3000000
## 221	4.400000	22600000	100000
## 222	6.441558	68066033	47921730
##	replace_mean_Screens	replace_mean_Views	replace_mean_Likes
## 1	45.000	3280543	4632.00
## 2	3306.000	583289	3465.00
## 3	2872.000	304861	328.00
## 4	3470.000	452917	2429.00
## 5	2310.000	3145573	12163.00
## 6	2209.244	91137	112.00
## 7	3158.000	3013011	9595.00
## 8	818.000	1854103	2207.00
## 9	2714.000	2213659	2210.00
## 10	2253.000	5218079	11709.00
## 11	3555.000	3927600	13143.00
## 12	1762.000	519327	963.00
## 13	3185.000	19032902	38810.00
## 14	3116.000	930006	5150.00
## 15	65.000	595194	85.00
## 16	18.000	3915978	6983.00
## 17	25.000	1391527	2479.00
## 18	31.000	1828235	7633.00
## 19	3761.000	4700023	14163.00
## 20	1823.000	1348142	4404.00
## 21	3555.000	7977747	18690.00
## 22	771.000	1671367	4572.00
## 23	2647.000	2088644	6633.00
## 24	482.000	4398243	9202.00
## 25	2209.244	7128	1.00
## 26	382.000	2902492	9522.00
## 27	3938.000	760262	2918.00
## 28	3967.000	1735700	6772.00
## 29	97.000	465219	1348.00
## 30	2544.000	1844690	3728.00
## 31	3936.000	463866	3400.00
## 32	3376.000	384448	1230.00
## 33	2209.244	9149892	26427.00

## 34	2781.000	522630	1248.00
## 35	3154.000	3287020	7698.00
## 36	3230.000	1488038	2571.00
## 37	3490.000	15568277	29251.00
## 38	2800.000	11850723	24226.00
## 39	2209.244	735551	636.00
## 40	66.000	6685088	8369.00
## 41	3313.000	2276605	3946.00
## 42	8.000	1034480	6490.00
## 43	780.000	456564	1706.00
## 44	3.000	99427	47.00
## 45	3952.000	1156609	2968.00
## 46	4080.000	1313548	8567.00
## 47	9.000	924347	1406.00
## 48	2417.000	175017	461.00
## 49	3595.000	9324678	15479.00
## 50	93.000	1292235	5284.00
## 51	4253.000	11472161	22779.00
## 52	1272.000	9222933	41728.00
## 53	6.000	8210	6.00
## 54	645.000	1167941	2651.00
## 55	3561.000	5421705	16635.00
## 56	3434.000	4270410	8886.00
## 57	2440.000	817242	4391.00
## 58	4.000	4877	6.00
## 59	3387.000	3320754	4322.00
## 60	2905.000	1438350	4028.00
## 61	2589.000	4846645	14722.00
## 62	374.000	3650720	6917.00
## 63	3204.000	2767873	46023.00
## 64	28.000	289922	143.00
## 65	1534.000	4450824	7315.00
## 66	2209.244	1222921	5553.00
## 67	3.000	30529	18.00
## 68	3173.000	1142964	2346.00
## 69	3948.000	557012	3528.00
## 70	66.000	177465	595.00
## 71	3019.000	1470438	4314.00
## 72	1044.000	667852	469.00
## 73	3934.000	277848	890.00
## 74	3194.000	3037329	6696.00
## 75	255.000	446576	659.00
## 76	3115.000	1451649	7342.00
## 77	3279.000	2554307	8722.00
## 78	22.000	3779254	13535.00
## 79	2766.000	6082510	12522.00
## 80	2130.000	608230	895.00
## 81	3567.000	13661095	41254.00
## 82	3090.000	367551	700.00
## 83	2209.244	5403836	187162.00
## 84	3303.000	2028767	3829.00
## 85	2867.000	105480	352.00
## 86	32.000	1223790	2934.00
## 87	3826.000	2117798	2124.00

## 88	2658.000	355563	1568.00
## 89	2663.000	8429023	27484.00
## 90	3948.000	3849768	9783.00
## 91	3372.000	3305047	11733.00
## 92	371.000	909596	2214.00
## 93	2209.244	827239	3221.00
## 94	2486.000	3466458	6096.00
## 95	2179.000	3743181	16782.00
## 96	59.000	3726728	6221.00
## 97	2875.000	23360	36.00
## 98	3062.000	2757667	3030.00
## 99	2894.000	5223362	18770.00
## 100	3260.000	943306	3006.00
## 101	27.000	2426078	9230.00
## 102	18.000	4176181	9463.00
## 103	21.000	1544390	2975.00
## 104	3465.000	3429055	7682.00
## 105	3845.000	7908038	27312.00
## 106	2809.000	6132551	14539.00
## 107	1586.000	134353	280.00
## 108	4324.000	386857	4996.00
## 109	147.000	6501107	14139.00
## 110	8.000	5546710	15351.00
## 111	12.000	5165441	17476.00
## 112	3464.000	2545852	3964.00
## 113	2209.244	330363	406.00
## 114	809.000	2897407	5953.00
## 115	3236.000	7075635	15858.00
## 116	3173.000	32626778	370552.00
## 117	432.000	10747	4.00
## 118	103.000	4790221	4740.00
## 119	702.000	523457	2187.00
## 120	461.000	1303646	3306.00
## 121	3875.000	3554189	14152.00
## 122	4.000	26528	58.00
## 123	2023.000	24809	277.00
## 124	4151.000	3305417	14684.00
## 125	2209.244	309610	729.00
## 126	747.000	3047849	11748.00
## 127	78.000	6231191	13331.00
## 128	3003.000	1325872	4829.00
## 129	2104.000	797229	1606.00
## 130	3775.000	439159	1847.00
## 131	1841.000	381071	238.00
## 132	173.000	3156436	18124.00
## 133	3083.000	10649	21.00
## 134	2776.000	1935432	3089.00
## 135	3427.000	2269032	3046.00
## 136	20.000	1223891	1309.00
## 137	3205.000	8479994	35071.00
## 138	2805.000	1142295	3895.00
## 139	685.000	4092871	8781.00
## 140	5.000	702	1.00
## 141	2.000	253631	170.00

## 142	15.000	3226251	18240.00
## 143	320.000	3281842	4968.00
## 144	2602.000	698	16.00
## 145	2225.000	2063089	5858.00
## 146	3455.000	14141585	36646.00
## 147	449.000	473100	670.00
## 148	3131.000	6637551	19833.00
## 149	291.000	476747	2079.00
## 150	794.000	280566	477.00
## 151	42.000	7750223	17541.00
## 152	2.000	166612	571.00
## 153	1061.000	5976092	9343.00
## 154	4.000	697105	1023.00
## 155	2965.000	719976	1312.00
## 156	68.000	865690	1375.00
## 157	3996.000	2285	9.00
## 158	2.000	550734	894.00
## 159	4276.000	10366624	31552.00
## 160	4004.000	59056	330.00
## 161	3946.000	1438926	4632.00
## 162	4301.000	1341909	1607.00
## 163	3473.000	9214467	39824.00
## 164	3956.000	8748596	20352.00
## 165	3708.000	10341783	24413.00
## 166	3641.000	5536822	29411.00
## 167	3777.000	12632836	36508.00
## 168	3702.000	2732371	13030.00
## 169	2757.000	848970	12607.00
## 170	3711.000	6649290	20750.00
## 171	3158.000	9511288	19903.00
## 172	3972.000	999867	4212.00
## 173	3175.000	10078326	26565.00
## 174	3758.000	84870	265.00
## 175	3442.000	6711914	29903.00
## 176	3723.000	5340100	26134.00
## 177	3355.000	5128288	18475.00
## 178	3003.000	4826940	10521.00
## 179	3411.000	2554327	10062.00
## 180	3240.000	3651828	13998.00
## 181	3181.000	3920842	10535.00
## 182	2991.000	11036701	50002.00
## 183	2855.000	2947239	19201.00
## 184	3366.000	2393017	13291.00
## 185	1648.000	2513544	6970.00
## 186	2602.000	5588384	15144.00
## 187	3003.000	6964819	26601.00
## 188	2575.000	12792898	56001.00
## 189	258.000	6495	82.00
## 190	3108.000	1841293	5879.00
## 191	3031.000	2854910	23254.00
## 192	2666.000	4442147	10605.00
## 193	2766.000	11037833	36874.00
## 194	1603.000	638374	4018.00
## 195	2893.000	6646785	16041.00

## 196	3261.000	13154873	46684.00
## 197	2002.000	50444	238.00
## 198	2880.000	11496	194.00
## 199	2778.000	2996539	1874.00
## 200	2816.000	3098749	4311.00
## 201	2648.000	3850758	13363.00
## 202	2567.000	2409338	6923.00
## 203	34.000	4032265	18398.00
## 204	420.000	1364537	3824.00
## 205	79.000	5085068	14359.00
## 206	14.000	63724	115.00
## 207	22.000	44963	109.00
## 208	3856.000	9597644	32558.00
## 209	2772.000	11476882	40496.00
## 210	1255.000	419470	2218.00
## 211	3638.000	5216680	20010.00
## 212	3845.000	10164908	22726.00
## 213	3201.000	7384182	23597.00
## 214	200.000	890619	6352.00
## 215	3171.000	5671767	10073.00
## 216	1573.000	831044	2427.00
## 217	66.000	3701061	9325.00
## 218	2815.000	7119456	18803.00
## 219	2777.000	3450614	6823.00
## 220	2209.244	66872	400.00
## 221	2720.000	659772	2841.00
## 222	2209.244	3712851	12732.54
##	replace_mean_Dislikes	replace_mean_Comments	
## 1	425.0000	636.000	
## 2	61.0000	186.000	
## 3	34.0000	47.000	
## 4	132.0000	590.000	
## 5	610.0000	1082.000	
## 6	7.0000	1.000	
## 7	419.0000	1020.000	
## 8	197.0000	593.000	
## 9	419.0000	382.000	
## 10	532.0000	770.000	
## 11	573.0000	3134.000	
## 12	94.0000	70.000	
## 13	4382.0000	4392.000	
## 14	707.0000	1484.000	
## 15	36.0000	39.000	
## 16	247.0000	460.000	
## 17	146.0000	182.000	
## 18	235.0000	685.000	
## 19	538.0000	1293.000	
## 20	307.0000	1033.000	
## 21	1940.0000	2214.000	
## 22	207.0000	741.000	
## 23	255.0000	1235.000	
## 24	454.0000	1150.000	
## 25	0.0000	0.000	
## 26	558.0000	2296.000	

## 27	66.0000	837.000
## 28	187.0000	889.000
## 29	72.0000	162.000
## 30	581.0000	729.000
## 31	152.0000	987.000
## 32	129.0000	228.000
## 33	1342.0000	5278.000
## 34	153.0000	227.000
## 35	446.0000	1122.000
## 36	553.0000	643.000
## 37	1730.0000	6439.000
## 38	1343.0000	2577.000
## 39	98.0000	92.000
## 40	467.0000	1580.000
## 41	331.0000	1286.000
## 42	181.0000	374.000
## 43	413.0000	890.000
## 44	10.0000	12.000
## 45	112.0000	547.000
## 46	269.0000	1285.000
## 47	107.0000	132.000
## 48	34.0000	133.000
## 49	1130.0000	3925.000
## 50	124.0000	362.000
## 51	862.0000	2863.000
## 52	924.0000	3609.000
## 53	0.0000	0.000
## 54	82.0000	797.000
## 55	751.0000	4316.000
## 56	569.0000	3058.000
## 57	112.0000	346.000
## 58	1.0000	1.000
## 59	347.0000	1105.000
## 60	133.0000	543.000
## 61	405.0000	2732.000
## 62	234.0000	1119.000
## 63	944.0000	6946.000
## 64	17.0000	9.000
## 65	546.0000	909.000
## 66	193.0000	335.000
## 67	4.0000	2.000
## 68	167.0000	311.000
## 69	135.0000	464.000
## 70	39.0000	71.000
## 71	168.0000	511.000
## 72	38.0000	44.000
## 73	45.0000	88.000
## 74	564.0000	1061.000
## 75	50.0000	45.000
## 76	533.0000	2305.000
## 77	298.0000	693.000
## 78	362.0000	1221.000
## 79	543.0000	2170.000
## 80	118.0000	387.000

## 81	3812.0000	18077.000
## 82	19.0000	37.000
## 83	3145.0000	24919.000
## 84	500.0000	665.000
## 85	45.0000	85.000
## 86	123.0000	226.000
## 87	485.0000	626.000
## 88	106.0000	267.000
## 89	977.0000	2195.000
## 90	704.0000	1151.000
## 91	1077.0000	4319.000
## 92	186.0000	632.000
## 93	89.0000	432.000
## 94	411.0000	977.000
## 95	565.0000	4973.000
## 96	405.0000	1074.000
## 97	5.0000	5.000
## 98	418.0000	251.000
## 99	627.0000	2796.000
## 100	325.0000	1401.000
## 101	184.0000	373.000
## 102	310.0000	885.000
## 103	136.0000	380.000
## 104	675.0000	727.000
## 105	3439.0000	8533.000
## 106	653.0000	846.000
## 107	43.0000	308.000
## 108	233.0000	864.000
## 109	890.0000	2928.000
## 110	535.0000	1271.000
## 111	871.0000	3229.000
## 112	378.0000	554.000
## 113	52.0000	92.000
## 114	153.0000	569.000
## 115	574.0000	1966.000
## 116	4752.0000	38363.000
## 117	1.0000	1.000
## 118	454.0000	773.000
## 119	149.0000	565.000
## 120	211.0000	564.000
## 121	262.0000	1782.000
## 122	1.0000	8.000
## 123	12.0000	52.000
## 124	332.0000	1176.000
## 125	97.0000	113.000
## 126	253.0000	1237.000
## 127	1265.0000	3430.000
## 128	104.0000	378.000
## 129	630.0000	456.000
## 130	105.0000	289.000
## 131	27.0000	43.000
## 132	213.0000	2753.000
## 133	1.0000	0.000
## 134	155.0000	567.000

## 135	546.0000	554.000
## 136	148.0000	239.000
## 137	2233.0000	3479.000
## 138	193.0000	882.000
## 139	794.0000	2046.000
## 140	0.0000	1.000
## 141	11.0000	58.000
## 142	261.0000	2104.000
## 143	445.0000	2099.000
## 144	1.0000	9.000
## 145	336.0000	346.000
## 146	1459.0000	6811.000
## 147	104.0000	380.000
## 148	815.0000	3403.000
## 149	166.0000	232.000
## 150	56.0000	82.000
## 151	631.0000	2760.000
## 152	36.0000	70.000
## 153	649.0000	1333.000
## 154	39.0000	429.000
## 155	76.0000	189.000
## 156	79.0000	67.000
## 157	0.0000	1.000
## 158	44.0000	77.000
## 159	989.0000	3843.000
## 160	8.0000	39.000
## 161	262.0000	496.000
## 162	764.0000	48.000
## 163	998.0000	1987.000
## 164	649.0000	1842.000
## 165	1675.0000	3426.000
## 166	1840.0000	1281.000
## 167	2210.0000	7559.000
## 168	497.0000	1774.000
## 169	237.0000	1560.000
## 170	750.0000	1666.000
## 171	2581.0000	2955.000
## 172	66.0000	250.000
## 173	1418.0000	2395.000
## 174	13.0000	63.000
## 175	984.0000	1767.000
## 176	2007.0000	3717.000
## 177	858.0000	1579.000
## 178	478.0000	755.000
## 179	464.0000	871.000
## 180	969.0000	2205.000
## 181	565.0000	1668.000
## 182	1005.0000	3525.000
## 183	625.0000	1842.000
## 184	369.0000	584.000
## 185	270.0000	1105.000
## 186	913.0000	1499.000
## 187	1111.0000	1293.000
## 188	2083.0000	4102.000

## 189	3.0000	7.000
## 190	314.0000	634.000
## 191	459.0000	1087.000
## 192	691.0000	2739.000
## 193	1885.0000	4360.000
## 194	130.0000	269.000
## 195	955.0000	2787.000
## 196	3565.0000	8578.000
## 197	3.0000	28.000
## 198	9.0000	31.000
## 199	32.0000	189.000
## 200	341.0000	881.000
## 201	453.0000	1276.000
## 202	340.0000	714.000
## 203	302.0000	1298.000
## 204	689.0000	772.000
## 205	600.0000	2468.000
## 206	28.0000	14.000
## 207	27.0000	42.000
## 208	2672.0000	8359.000
## 209	1383.0000	4435.000
## 210	46.0000	239.000
## 211	500.0000	2300.000
## 212	4245.0000	5262.000
## 213	786.0000	3481.000
## 214	293.0000	700.000
## 215	480.0000	1712.000
## 216	99.0000	247.000
## 217	641.0000	1859.000
## 218	1128.0000	2290.000
## 219	325.0000	409.000
## 220	67.0000	201.000
## 221	431.0000	606.000
## 222	679.0519	1825.701
##	replace_mean_Aggregate_Followers	
## 1	1120000	
## 2	12350000	
## 3	483000	
## 4	568000	
## 5	1923800	
## 6	310000	
## 7	8153000	
## 8	130655	
## 9	125646	
## 10	21697300	
## 11	24300	
## 12	386400	
## 13	19420105	
## 14	5130800	
## 15	15112	
## 16	253000	
## 17	1658900	
## 18	116100	
## 19	199800	

## 20	888000
## 21	2417000
## 22	105000
## 23	3209000
## 24	4769100
## 25	2182
## 26	3038193
## 27	8030000
## 28	114000
## 29	744600
## 30	9536
## 31	1030000
## 32	276750
## 33	395500
## 34	147000
## 35	17064000
## 36	88586
## 37	5610000
## 38	21500
## 39	1060000
## 40	13720000
## 41	1888000
## 42	66600
## 43	412000
## 44	3038193
## 45	1870000
## 46	2750000
## 47	5887700
## 48	3038193
## 49	9414000
## 50	1650000
## 51	671000
## 52	1800000
## 53	58900
## 54	3038193
## 55	1865000
## 56	301000
## 57	4720000
## 58	3038193
## 59	147000
## 60	27323
## 61	3038193
## 62	1045200
## 63	5407000
## 64	3038193
## 65	2356000
## 66	1463000
## 67	18100
## 68	3038193
## 69	5633
## 70	644000
## 71	130000
## 72	919000
## 73	8839043

## 74	9850000
## 75	3038193
## 76	2594000
## 77	14240000
## 78	6480000
## 79	3038193
## 80	1550500
## 81	1810000
## 82	648786
## 83	2720000
## 84	124000
## 85	11444
## 86	47200
## 87	3603000
## 88	370000
## 89	21586000
## 90	154400
## 91	250000
## 92	3038193
## 93	217000
## 94	3678000
## 95	9224
## 96	3038193
## 97	804300
## 98	4521000
## 99	8620000
## 100	3185900
## 101	33500
## 102	269849
## 103	216000
## 104	727000
## 105	1260000
## 106	11783000
## 107	3038193
## 108	4520000
## 109	3038193
## 110	48231
## 111	10364000
## 112	1280000
## 113	3038193
## 114	3038193
## 115	1800000
## 116	4240000
## 117	818000
## 118	2740000
## 119	184100
## 120	759800
## 121	2613000
## 122	25748
## 123	20700
## 124	31030000
## 125	275873
## 126	3038193
## 127	6619435

## 128	5563500
## 129	1174806
## 130	4690000
## 131	14586
## 132	4734000
## 133	3849
## 134	148000
## 135	3038193
## 136	226000
## 137	2731000
## 138	49424
## 139	420000
## 140	9842
## 141	3038193
## 142	8204
## 143	1899400
## 144	3086
## 145	24388000
## 146	5987
## 147	180100
## 148	269000
## 149	2536000
## 150	585000
## 151	858000
## 152	3038193
## 153	781200
## 154	3038193
## 155	1810000
## 156	1818778
## 157	10280000
## 158	168700
## 159	10070000
## 160	11890000
## 161	232000
## 162	250000
## 163	7336000
## 164	6605000
## 165	5070000
## 166	184000
## 167	9737600
## 168	768700
## 169	55618
## 170	265000
## 171	2014000
## 172	1198000
## 173	2939000
## 174	3877901
## 175	10988000
## 176	2466000
## 177	8392000
## 178	2284000
## 179	2347000
## 180	1066
## 181	7460000

```
## 182          776000
## 183          2113
## 184          324925
## 185          1655987
## 186          4599000
## 187          1630000
## 188          6714000
## 189          675000
## 190          788000
## 191          4184000
## 192          2007000
## 193           5699
## 194           1887
## 195          253499
## 196          209000
## 197          986572
## 198          1891977
## 199          155000
## 200          1520000
## 201          3045000
## 202          1334000
## 203           2208
## 204           3841
## 205          3744000
## 206          129000
## 207          1188000
## 208          3038193
## 209          3038193
## 210          3038193
## 211          3038193
## 212          3038193
## 213          3038193
## 214          3038193
## 215          3038193
## 216          3038193
## 217          3038193
## 218          3038193
## 219          3038193
## 220          3038193
## 221          3038193
## 222          3038193
```

Other Ways of Removing Outliers

```
outliers <- function(x) {

  Q1 <- quantile(x, probs=.25)
  Q3 <- quantile(x, probs=.75)
  iqr = Q3-Q1

  upper_limit = Q3 + (iqr*1.5)
  lower_limit = Q1 - (iqr*1.5)
```

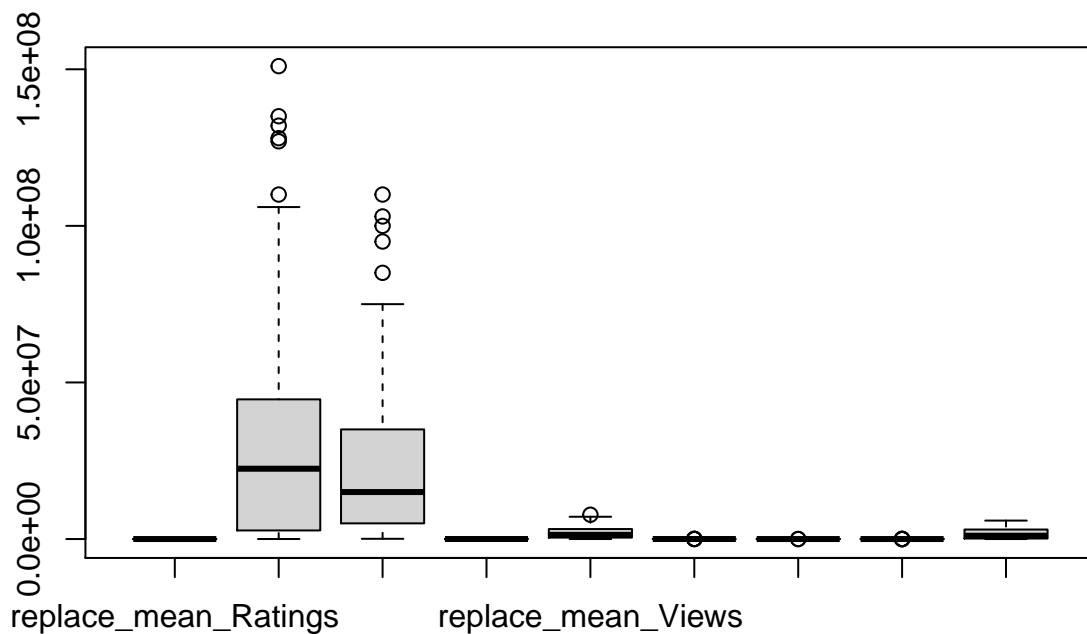
```

  x > upper_limit | x < lower_limit
}

remove_outliers <- function(movie_data, cols = names(movie_data)) {
  for (col in cols) {
    movie_data <- movie_data[!outliers(movie_data[[col]]),]
  }
  movie_data
}

movie_data_without_outliers <- remove_outliers(movie_data)
boxplot(movie_data_without_outliers)

```



```

mode2 <- lm(replace_mean_Ratings ~ replace_mean_Gross+replace_mean_Budget+
  replace_mean_Screens+replace_mean_Views+replace_mean_Likes+
  replace_mean_Dislikes+replace_mean_Comments+
  replace_mean_Aggregate_Followers, data = movie_data_without_outliers)
mode2

##
## Call:
## lm(formula = replace_mean_Ratings ~ replace_mean_Gross + replace_mean_Budget +
##   replace_mean_Screens + replace_mean_Views + replace_mean_Likes +
##   replace_mean_Dislikes + replace_mean_Comments + replace_mean_Aggregate_Followers,
##   data = movie_data_without_outliers)
##

```

```
## Coefficients:
##              (Intercept)              replace_mean_Gross
##              6.406e+00              6.083e-09
##      replace_mean_Budget      replace_mean_Screens
##              8.219e-10              -1.053e-04
##      replace_mean_Views      replace_mean_Likes
##              2.056e-07              4.247e-05
##      replace_mean_Dislikes      replace_mean_Comments
##              -3.518e-03              4.347e-04
## replace_mean_Aggregate_Followers
##              -5.468e-08

summary(mode2)

##
## Call:
## lm(formula = replace_mean_Ratings ~ replace_mean_Gross + replace_mean_Budget +
##      replace_mean_Screens + replace_mean_Views + replace_mean_Likes +
##      replace_mean_Dislikes + replace_mean_Comments + replace_mean_Aggregate_Followers,
##      data = movie_data_without_outliers)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.0240 -0.4576  0.0238  0.5994  1.7028
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      6.406e+00  1.278e-01  50.122 < 2e-16 ***
## replace_mean_Gross      6.083e-09  2.323e-09   2.618  0.00982 **
## replace_mean_Budget      8.219e-10  3.307e-09   0.249  0.80409
## replace_mean_Screens     -1.053e-04  5.989e-05  -1.758  0.08099 .
## replace_mean_Views      2.056e-07  6.869e-08   2.993  0.00327 **
## replace_mean_Likes      4.247e-05  2.166e-05   1.961  0.05183 .
## replace_mean_Dislikes     -3.518e-03  4.219e-04  -8.338 6.61e-14 ***
## replace_mean_Comments      4.347e-04  1.564e-04   2.779  0.00621 **
## replace_mean_Aggregate_Followers -5.468e-08  4.276e-08  -1.279  0.20308
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7503 on 139 degrees of freedom
## Multiple R-squared:  0.4087, Adjusted R-squared:  0.3747
## F-statistic: 12.01 on 8 and 139 DF,  p-value: 6.118e-13

##mode2
```

Coefficients Function & Fitted function

```
mode2$coefficients

##              (Intercept)              replace_mean_Gross
##              6.406250e+00              6.083238e-09
##      replace_mean_Budget      replace_mean_Screens
##              8.218750e-10              -1.052669e-04
##      replace_mean_Views      replace_mean_Likes
```

```
##                2.056063e-07                4.247493e-05
##      replace_mean_Dislikes      replace_mean_Comments
##                -3.517640e-03                4.346602e-04
## replace_mean_Aggregate_Followers
##                -5.467840e-08
```

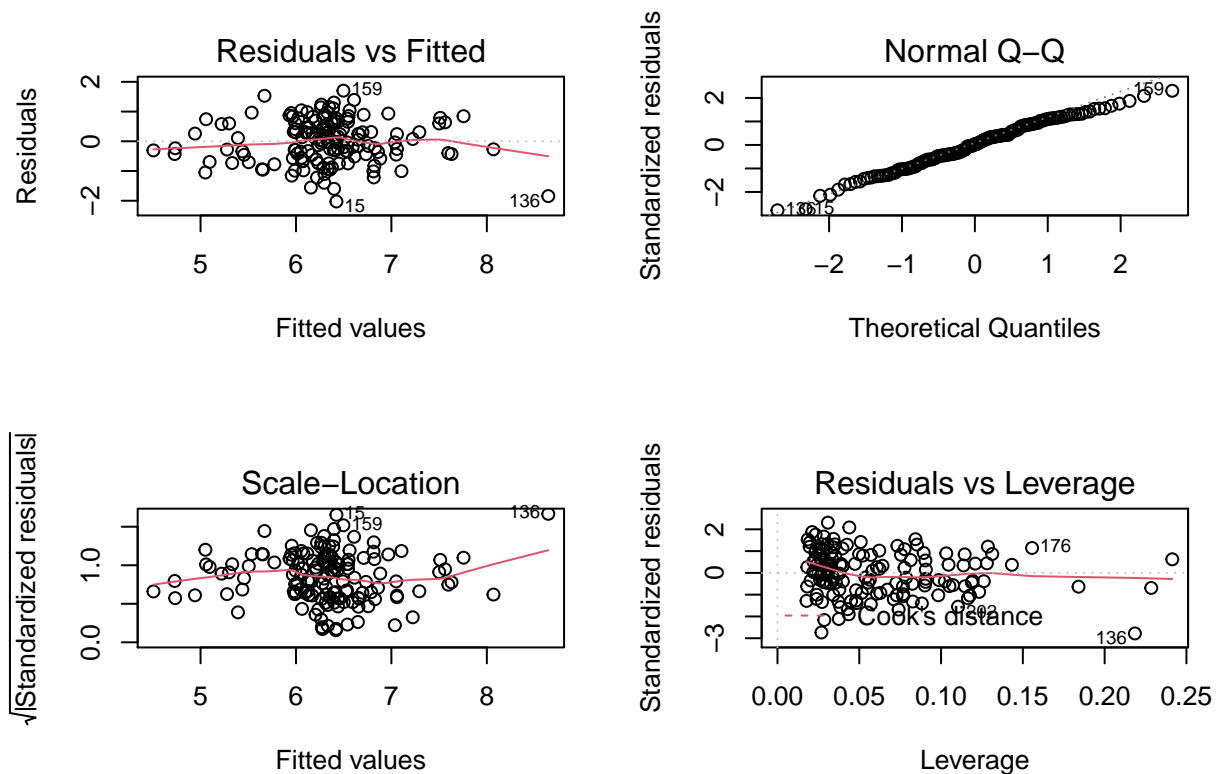
```
mode2$fitted
```

```
##      1      3      4      5      6      8      9     12
## 5.996307 6.264724 6.733561 5.653923 6.156633 6.404149 5.536023 6.164211
##      14     15     16     17     18     20     22     23
## 4.941691 6.424042 6.828363 6.274396 6.662411 6.137923 6.608920 6.449629
##      24     25     26     29     30     31     32     34
## 6.309884 6.175960 6.277484 6.535573 5.050542 7.058145 6.097436 6.023786
##      36     39     41     42     43     44     48     49
## 5.058101 5.992989 6.561813 6.418135 5.773516 6.233743 6.014372 6.545342
##      51     54     55     58     59     60     61     62
## 6.882976 6.413355 6.648594 6.822019 6.328947 6.524212 6.652740 7.630271
##      63     65     67     68     69     70     72     73
## 7.036410 6.268306 6.331475 6.049728 6.399046 6.593941 6.333968 6.440033
##      74     77     78     81     82     84     87     88
## 6.356167 6.205274 5.993289 6.968040 6.343674 6.697152 5.669903 6.220929
##      90     91     93     95     96     97     99    100
## 5.299405 6.212429 6.100798 6.125461 6.347761 5.986654 6.332298 6.233256
##     101     103     104     105     106     107     110     113
## 5.438220 6.057485 6.812373 7.223350 7.105583 5.505138 6.118870 7.050157
##     115     116     117     118     121     122     123     124
## 5.932905 5.955586 6.788112 7.596657 6.371884 6.395747 6.539160 6.260547
##     126     127     129     130     132     133     135     136
## 6.440230 6.537213 5.988150 7.500545 6.389730 4.506650 6.271713 8.643786
##     137     138     139     140     142     143     144     145
## 6.613089 6.500705 5.220515 6.285496 6.655391 5.644049 6.422627 6.295457
##     146     147     148     152     154     155     156     157
## 8.069440 6.545474 6.295421 6.287577 5.965321 6.402715 7.755896 6.205964
##     158     159     160     161     163     176     178     184
## 6.416749 6.497216 6.063183 6.314037 6.440939 7.511661 7.058097 4.729022
##     187     188     190     193     194     195     196     197
## 6.462584 5.977274 5.277473 6.400573 6.301998 6.709325 5.331711 5.455449
##     199     200     201     202     204     205     207     208
## 6.551103 6.057074 6.491866 6.299153 6.356702 6.133383 6.266433 6.078380
##     209     210     211     212     213     214     215     216
## 6.859954 6.127716 6.377081 6.067589 7.562500 4.734097 6.815290 6.327373
##     217     220     221     224     225     226     227     228
## 6.278007 6.397386 7.292690 5.949846 6.757688 6.317184 5.991899 5.391892
##     229     230     231     232
## 6.072432 5.967279 5.094989 6.170112
```

```
extract_eq(mode2,use_cof=TRUE)
```

$\text{replace_mean_Ratings} = \alpha + \beta_1(\text{replace_mean_Gross}) + \beta_2(\text{replace_mean_Budget}) + \beta_3(\text{replace_mean_Screens}) + \beta_4(\text{replace_mean_Dislikes}) + \beta_5(\text{replace_mean_Comments}) + \beta_6(\text{replace_mean_Aggregate_Followers})$

```
par(mfrow = c(2, 2))
plot(mode2)
```

We can see here, that the coefficient is different between our new model and our previous model

Also, after removing the outlier from our dataset the residual standard error is decreasing, while the R^2 is increasing

It means that our new model fits better to our data compared to the previous model

VIF for new model

```
vif(mode2)
```

```
##           replace_mean_Gross           replace_mean_Budget
##           1.645659           1.637972
##           replace_mean_Screens           replace_mean_Views
##           1.654996           4.454480
##           replace_mean_Likes           replace_mean_Dislikes
##           4.124335           3.212534
##           replace_mean_Comments replace_mean_Aggregate_Followers
##           3.497897           1.051134
```

```
glance(mode2)
```

```
## # A tibble: 1 x 12
##   r.squared adj.r.squared sigma statistic p.value    df logLik   AIC   BIC
##   <dbl>      <dbl> <dbl>    <dbl>   <dbl> <dbl> <dbl> <dbl> <dbl>
## 1    0.409      0.375 0.750     12.0 6.12e-13     8  -163.  346.  376.
```

```
## # ... with 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>
```

```
tidy(mode2)
```

```
## # A tibble: 9 x 5
```

##	term	estimate	std.error	statistic	p.value
##	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	(Intercept)	6.41e+ 0	0.128	50.1	7.11e-91
## 2	replace_mean_Gross	6.08e- 9	0.00000000232	2.62	9.82e- 3
## 3	replace_mean_Budget	8.22e-10	0.00000000331	0.249	8.04e- 1
## 4	replace_mean_Screens	-1.05e- 4	0.0000599	-1.76	8.10e- 2
## 5	replace_mean_Views	2.06e- 7	0.0000000687	2.99	3.27e- 3
## 6	replace_mean_Likes	4.25e- 5	0.0000217	1.96	5.18e- 2
## 7	replace_mean_Dislikes	-3.52e- 3	0.000422	-8.34	6.61e-14
## 8	replace_mean_Comments	4.35e- 4	0.000156	2.78	6.21e- 3
## 9	replace_mean_Aggregate_Followers	-5.47e- 8	0.0000000428	-1.28	2.03e- 1

vif<10 for all variables the model good.

```
#####  
### SPLITTING THE DATA ### TRAINING AND TEST SETS ###  
#####  
#####  
###*
```

Let's say we want to split the data in 80:20 for train :test dataset

```
set.seed(232)  
ind <- createDataPartition(movie_data$ replace_mean_Ratings,  
                           p = 0.8, times = 1, list = FALSE)  
length(ind)
```

```
## [1] 187
```

```
train_set <- movie_data[ind, ]  
test_set <- movie_data[-ind, ]  
nrow(train_set); nrow(test_set)
```

```
## [1] 187
```

```
## [1] 45
```

Training the model

```
lm_fit <- lm(replace_mean_Ratings ~ . , data = train_set)  
broom::tidy(lm_fit)
```

```
## # A tibble: 5 x 5
```

##	term	estimate	std.error	statistic	p.value
##	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	(Intercept)	6.28e+0	0.126	49.7	2.73e-106
## 2	replace_mean_Gross	3.72e-9	0.00000000105	3.54	5.15e- 4
## 3	replace_mean_Budget	4.47e-9	0.00000000185	2.41	1.69e- 2
## 4	replace_mean_Screens	-1.26e-4	0.0000593	-2.13	3.43e- 2
## 5	replace_mean_Views	6.20e-8	0.0000000270	2.29	2.30e- 2

```
## 6 replace_mean_Likes          1.20e-5 0.00000581      2.06 4.08e- 2
## 7 replace_mean_Dislikes      -3.19e-4 0.0000819      -3.90 1.38e- 4
## 8 replace_mean_Comments      -8.59e-5 0.0000490      -1.75 8.14e- 2
## 9 replace_mean_Aggregate_Followers -8.22e-9 0.0000000163    -0.504 6.15e- 1
```

```
broom::glance(lm_fit)
```

```
## # A tibble: 1 x 12
##   r.squared adj.r.squared sigma statistic p.value    df logLik   AIC   BIC
##   <dbl>      <dbl> <dbl>      <dbl>    <dbl> <dbl> <dbl> <dbl> <dbl>
## 1    0.281        0.249 0.878        8.72 5.12e-10     8  -236.  493.  525.
## # ... with 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>
```

We've built a machine learning model and trained it on `train_set`

```
###* *** Prediction *** # -----
```

```
pred <- predict(object = lm_fit, newdata = test_set, type = "response")
```

```
head(pred)
```

```
##          3          4          6          13          17          18
## 6.160153 6.687476 6.004852 5.930496 6.339215 6.431399
```

```
**** Model Evaluation *** # -----
```

```
actual <- test_set$replace_mean_Ratings
mae <- Metrics::mae(actual = actual, predicted = pred)
mse <- Metrics::mse(actual = actual, predicted = pred)
rmse <- Metrics::rmse(actual = actual, predicted = pred)
```

These common metrics are used to evaluate the model.

Table of results

```
knitr::kable(cbind(mae, mse, rmse))
```

mae	mse	rmse
0.660215	0.6630907	0.8143038