Assignment 2

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Loading the Necessary Packages

```
library(RMySQL)
## Loading required package: DBI
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
       filter, lag
## The following objects are masked from 'package:base':
       intersect, setdiff, setequal, union
##
library(reshape)
## Attaching package: 'reshape'
## The following object is masked from 'package:dplyr':
##
##
       rename
library(reshape2)
##
## Attaching package: 'reshape2'
## The following objects are masked from 'package:reshape':
##
##
       colsplit, melt, recast
```

```
library(ggplot2)
```

Establishing the MySQL Connection

Inspecting the available tables in the database

```
dbListTables(mysqlconnection)
## [1] "movie_info" "movie_ratings"
```

Converting movie_ratings databasze into a R dataframe

Reshaping the data frame

```
# I want to keep this here, because it shows that I really did work hard on this
# and that there is always room to grow!
ratings <- movie_ratings[2:ncol(movie_ratings)]</pre>
```

```
df <- data.frame(matrix(ncol = 3, nrow = ncol(ratings)*nrow(ratings)))</pre>
x <- c("Timestamp", "movie_names", "movie_ratings")</pre>
colnames(df) <- x
movies names <- c()
individual_ratings <- c()</pre>
timestamps <- c()</pre>
for (y in 1:ncol(ratings)) {
  for (x in 1: nrow(ratings)) {
    timestamps <- c(timestamps, movie_ratings$Timestamp[x])</pre>
    movies_names <- c(movies_names, colnames(ratings)[y])</pre>
    individual_ratings <- c(individual_ratings, ratings[x,y])</pre>
  }
}
df$Timestamp <- timestamps</pre>
df$movie_names <- movies_names</pre>
df$movie_ratings <- individual_ratings</pre>
```

Reshaping the data frame (condensed version)

```
melted_movies <- melt(movie_ratings, Movie_Names <- c("Timestamp", "respondent_id"))
colnames(melted_movies) <- c("Timestamp", "respondent_id", "Movie_Name", "Rating")</pre>
```

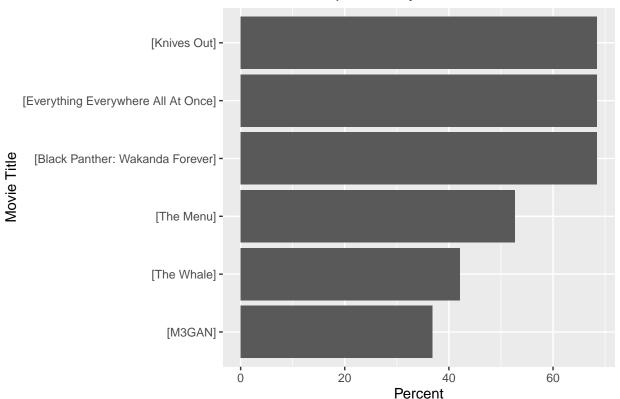
Percent of total group who watched each movie

```
percent_watched <- melted_movies %>%
  group_by(Movie_Name) %>%
  summarise(percents = (100.0 * sum(Rating != "N/A"))/n_distinct(respondent_id)) %>%
  arrange(percents)

percent_watched
```

```
## # A tibble: 6 x 2
##
     Movie_Name
                                           percents
##
     <fct>
                                              <dbl>
## 1 [M3GAN]
                                               36.8
## 2 [The Whale]
                                               42.1
## 3 [The Menu]
                                               52.6
## 4 [Black Panther: Wakanda Forever]
                                               68.4
## 5 [Everything Everywhere All At Once]
                                               68.4
## 6 [Knives Out]
                                               68.4
```

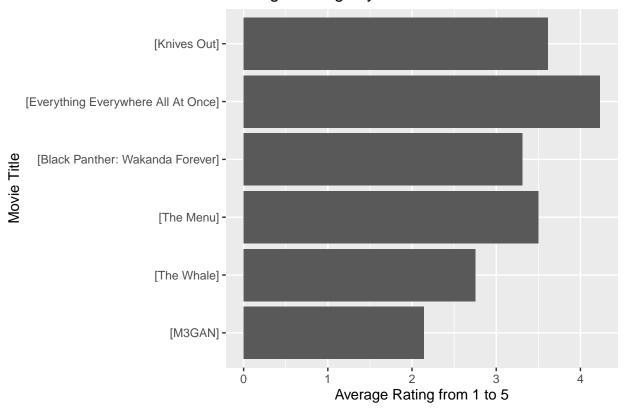
Percent of People Surveyed who had seen Each Mo



Average Rating

```
Avg_Ratings <- melted_movies %>%
  group_by(Movie_Name) %>%
  filter(Rating != "N/A") %>%
  summarise(Avgs = sum(as.double(Rating))/sum(Rating != "N/A"))
Avg_Ratings
```

Average Ratings by Movie Titles



Converting movie_info into a R dataframe

```
#for reference "movie_info" is a quick table I put together and added to my
#MySQL so that I could work with some outside characteristics
result = dbSendQuery(mysqlconnection, "select * from movie_info")
movie_info <- fetch(result)</pre>
```

Joining movie_ratings and movie_info for analysis

```
Avg_Ratings_and_Info <- merge(x = Avg_Ratings, y = movie_info, by = "Movie_Name",
                                 all.x = TRUE)
Avg_Ratings_and_Info
##
                              Movie_Name
                                                   Genre Release_Day
                                             Avgs
## 1
        [Black Panther: Wakanda Forever] 3.307692 Action
                                                                   11
## 2 [Everything Everywhere All At Once] 4.230769 Comedy
                                                                   25
                            [Knives Out] 3.615385 Mystery
                                                                   27
## 4
                                 [M3GAN] 2.142857 Horror
                                                                    6
## 5
                              [The Menu] 3.500000 Horror
                                                                   18
## 6
                             [The Whale] 2.750000
                                                  Drama
    Release_Month Release_Year Box_Office
                                              Budget Runtime IMDb_Rating
##
## 1
                11
                           2022 842200000 250000000
                                                         161
                                                                     7.0
## 2
                3
                          2022 106000000 25000000
                                                         139
                                                                     8.0
## 3
               11
                           2019 311900000 40000000
                                                        130
                                                                    7.9
## 4
                           2023 150500000 12000000
                                                        102
                                                                     6.4
                1
## 5
                           2022
                                 79300000 30000000
                                                        106
               11
                                                                    7.2
## 6
                9
                           2022
                                 15900000 3000000
                                                        117
                                                                     8.0
# IMDb does ratings out of 10, whereas we did rankings out of 5
Avg_Ratings_and_Info$Scaled_IMDb_Rating <- Avg_Ratings_and_Info$IMDb_Rating/2.0
Percent_Watched_and_Info <- merge(x = percent_watched, y = movie_info, by = "Movie_Name",
                                 all.x = TRUE)
```

Difference from my survey's average ratings compare to IMDb's ratings

```
Avg_Diffs <- Avg_Ratings_and_Info %>%
summarise(Movie_Name, Difference = Avgs - Scaled_IMDb_Rating) %>%
arrange(desc(Difference))

Avg_Diffs

## Movie_Name Difference
## 1 [Everything Everywhere All At Once] 0.2307692
## 2 [The Menu] -0.1000000
## 3 [Black Panther: Wakanda Forever] -0.1923077
## 4 [Knives Out] -0.3346154
## 5 [M3GAN] -1.0571429
```

```
\# M3GAN and The Whale having the largest difference from the IMDb Rating makes \# sense as they were the least watched
```

[The Whale] -1.2500000

6

```
# Seems like the folks who answered my survey are more critical than the IMDb
# average

ggplot(Avg_Diffs, aes(x = Difference, y = Movie_Name, fill = Difference)) +
   geom_bar(stat = "identity") +
   labs(title = "The Difference between the Average Ratings for my Survey Compared with IMDb's Ratings",
        y = "Movie Name")
```

The Difference between the Average Ratings for my

