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# IMDB Movie Data Set Analysis: Clustering

Sergio Navia, Erik Vela, Jason Wo

# Dataset

- IMDB Movies Dataset
- <https://www.kaggle.com/datasets/harshitshankhdhar/imdb-dataset-of-top-1000-movies-and-tv-shows>

Our Dataset consisted of the top 1000 movies from IMDb. It includes information on these movies based on several different categories such as rating, runtime, box office gross, year of release and many others.



# Variables of the Dataset

- Poster Link
  - A link to a poster of the movie provided by Amazon that IMDB uses.
- Series Title
  - The title of the movie.
- Release Year
  - The year the movie was released according to IMDB.
- Certificate
  - The movie rating decided by the country the movie was originally released in.
- Runtime
  - Duration of the movie in minutes.
- Genre
  - The genre the film was categorized in.
- IMDB Rating
  - A rating of the movie on the IMDB website.
- Overview
  - A description of the movie. Sometimes a summary.

# Variables of the Dataset

- Meta Score
  - a weighted average of reviews from top critics and publications for a given movie calculated by the Metacritic website. <https://www.metacritic.com/about-us/>
- Director
  - Name of the person who directed the movie.
- Star 1, Star 2, Star 3, Star 4
  - The names of the main cast of the movie.
- Number of Votes
  - According to the IMDB, “IMDb registered users can cast a vote (from 1 to 10) on every released title in the database. Individual votes are then aggregated and summarized as a single IMDb rating, visible on the title's main page.”
  - This variable is the total amount of these votes for a particular movie.
- Gross
  - Money earned at the box office for every release of a movie. (Certain movies are released in theaters more than once.)

# Variables of the dataset

The following variables are being used in the clustering analysis:

- Released\_Year:
  - The year the movie was released.
- Runtime:
  - The duration/length of the movie (minutes).
- IMDB\_Rating
  - Rating of the movie on the IMDB site with a scale 1-10.
- Meta Score
  - A weighted average of reviews from top critics on the movie.
- Gross
  - Gross earning of the movie, or the amount of money that the movie make(in US dollar).



# Data Preparation

This Dataset initially was not usable, it contained many categorical variables, N/A terms, and missing or obviously incorrect values.

- Our first step in removing the data was to remove all N/A and non usable entries in the dataset
- Feature engineered Genre variable from single multiclass variable into several binary variables
- We then removed the units of measurement attached to some of the numerical values (e.g. minutes in runtime)
- Finally, we removed the column that contained the names of the movies and saved them as the names of the rows
- After all the data cleaning, we were left with 713 rows and 5 columns to do analysis on.
- In order to account for how large Gross can be and affecting the clustering, we standardized the dataset.

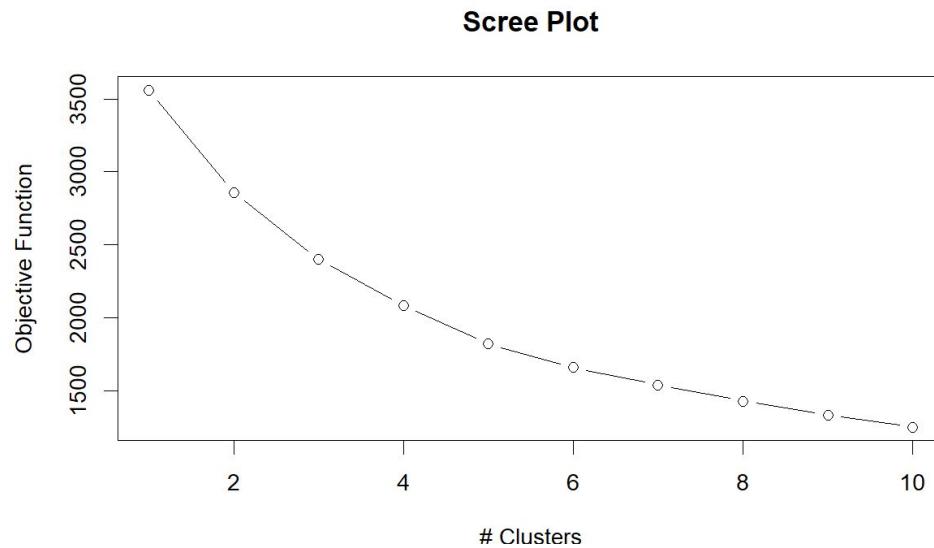
# Data Preparation

- We could have turned certificate into numerical values by giving each rating of the movie a number from 1 - n, where n is the total number of ratings a movie can be given.
- But as soon as we tried numbering the ratings, we noticed that different countries had different rating scales compared to the U.S.



# Determining number of cluster (Scree Plot)

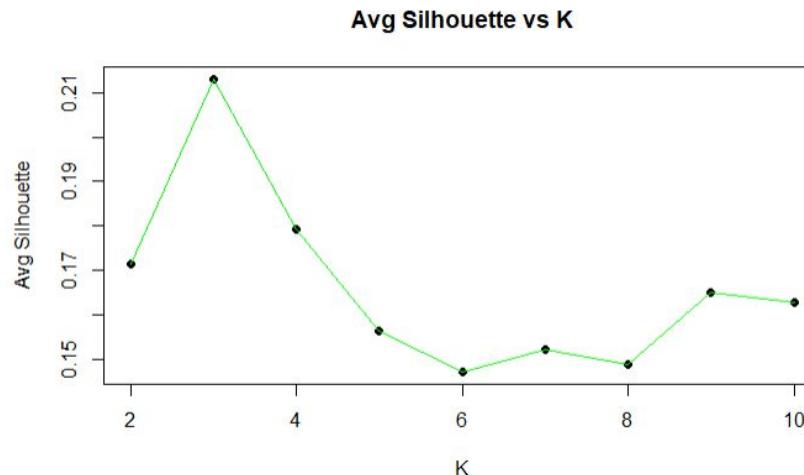
- k=2 to 6 all seems to be a pretty good option.
- k=3 seems to be the best here.
- Hard to be certain and objective
- We will use the Silhouette Statistic to help our decision.



# Determining number of cluster (Silhouette Statistic)

- The average Silhouette score measure how well the object in each cluster matches the current clustering result.
- Higher average Silhouette score indicates better performance of clustering.
- Silhouette statistic also suggest  $k=3$  to be the optimal number of cluster.

```
[1] "The average silhouette scores for K = 2 is 0.1713" "The average silhouette scores for K = 3 is 0.2131"  
[3] "The average silhouette scores for K = 4 is 0.1793" "The average silhouette scores for K = 5 is 0.1563"  
[5] "The average silhouette scores for K = 6 is 0.1471" "The average silhouette scores for K = 7 is 0.152"  
[7] "The average silhouette scores for K = 8 is 0.1488" "The average silhouette scores for K = 9 is 0.165"  
[9] "The average silhouette scores for K = 10 is 0.1628"
```



# Stats about K Means Clusters

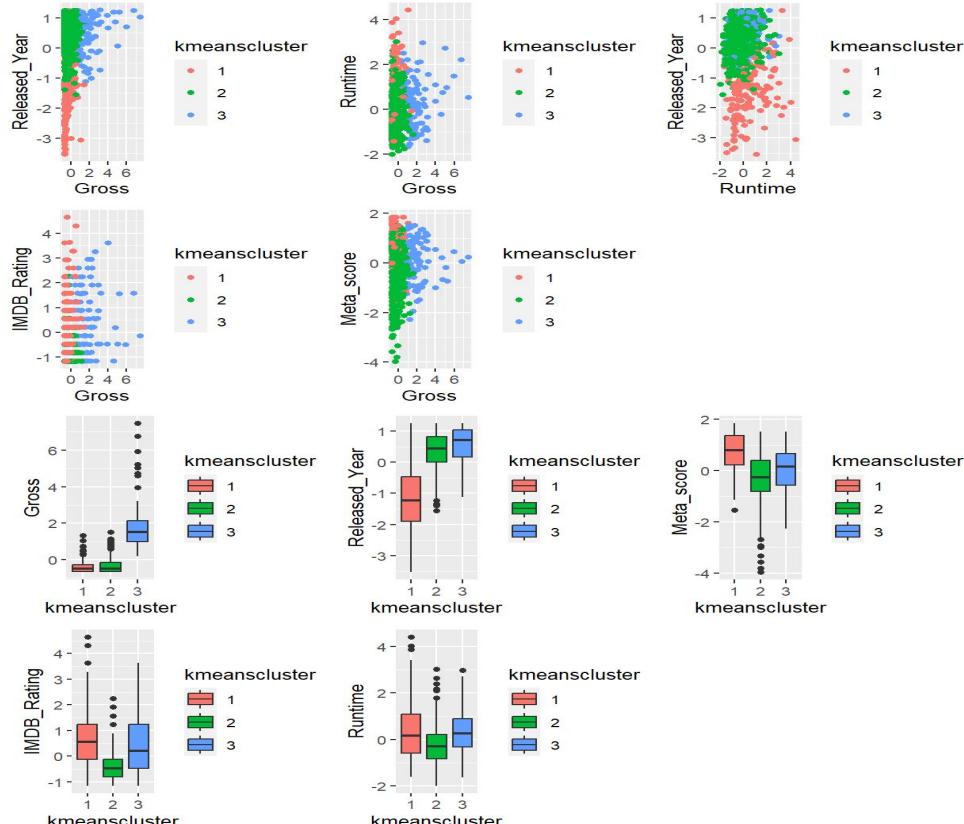
| Released_Year | Runtime       | IMDB_Rating  | Meta_score     | Gross             |
|---------------|---------------|--------------|----------------|-------------------|
| Min. :1930    | Min. : 82.0   | Min. :7.60   | Min. : 58.00   | Min. : 10177      |
| 1st Qu.:1960  | 1st Qu.:108.5 | 1st Qu.:7.90 | 1st Qu.: 80.00 | 1st Qu.: 5446437  |
| Median :1973  | Median :128.0 | Median :8.10 | Median : 87.00 | Median : 23383987 |
| Mean :1973    | Mean :133.4   | Mean :8.12   | Mean : 86.37   | Mean : 35248603   |
| 3rd Qu.:1987  | 3rd Qu.:151.5 | 3rd Qu.:8.30 | 3rd Qu.: 94.00 | 3rd Qu.: 46597035 |
| Max. :2019    | Max. :238.0   | Max. :9.30   | Max. :100.00   | Max. :232906145   |

| Released_Year | Runtime       | IMDB_Rating  | Meta_score    | Gross             |
|---------------|---------------|--------------|---------------|-------------------|
| Min. :1967    | Min. : 72.0   | Min. :7.60   | Min. :28.00   | Min. : 1305       |
| 1st Qu.:1996  | 1st Qu.:102.0 | 1st Qu.:7.70 | 1st Qu.:67.00 | 1st Qu.: 4195902  |
| Median :2004  | Median :116.0 | Median :7.80 | Median :74.00 | Median : 23345098 |
| Mean :2003    | Mean :117.5   | Mean :7.82   | Mean :73.08   | Mean : 41379821   |
| 3rd Qu.:2011  | 3rd Qu.:129.0 | 3rd Qu.:7.90 | 3rd Qu.:82.00 | 3rd Qu.: 59852210 |
| Max. :2019    | Max. :202.0   | Max. :8.60   | Max. :96.00   | Max. :251513985   |

| Released_Year | Runtime       | IMDB_Rating   | Meta_score    | Gross             |
|---------------|---------------|---------------|---------------|-------------------|
| Min. :1975    | Min. : 81.0   | Min. :7.600   | Min. :49.00   | Min. :100125643   |
| 1st Qu.:1999  | 1st Qu.:115.0 | 1st Qu.:7.800 | 1st Qu.:70.00 | 1st Qu.:191407502 |
| Median :2009  | Median :130.5 | Median :8.000 | Median :79.00 | Median :252659101 |
| Mean :2006    | Mean :131.6   | Mean :8.088   | Mean :77.81   | Mean :286723905   |
| 3rd Qu.:2015  | 3rd Qu.:147.0 | 3rd Qu.:8.300 | 3rd Qu.:85.25 | 3rd Qu.:323203039 |
| Max. :2019    | Max. :201.0   | Max. :9.000   | Max. :96.00   | Max. :936662225   |

# K-means clustering with k=3

- Cluster 1 contains movies that are
  - Older
  - Higher rating
  - Low gross income
  - Long runtime
  - 74% of this cluster is Drama
- Cluster 2 contains movies that are
  - Relatively new
  - Lower rating
  - Low gross income
  - short runtime
  - 76% of this cluster is Drama
- Cluster 3 contains movies that are
  - Newer
  - High gross income
  - 66% of this cluster is Adventure



# Movies in K Means Clusters

1

The Shawshank Redemption

The Godfather

The Godfather: Part II

12 Angry Men

Pulp Fiction

Schindler's List

Fight Club

Il buono, il brutto, il cattivo

Goodfellas

One Flew Over the Cuckoo's Nest

2

La vita è bella

Whiplash

The Intouchables

The Prestige

American History X

Léon

Capharnaüm

Kimi no na wa.

3 Idiots

Oldeuboi

3

The Dark Knight

The Lord of the Rings: The Return of the King

Inception

The Lord of the Rings: The Fellowship of the Ring

Forrest Gump

The Lord of the Rings: The Two Towers

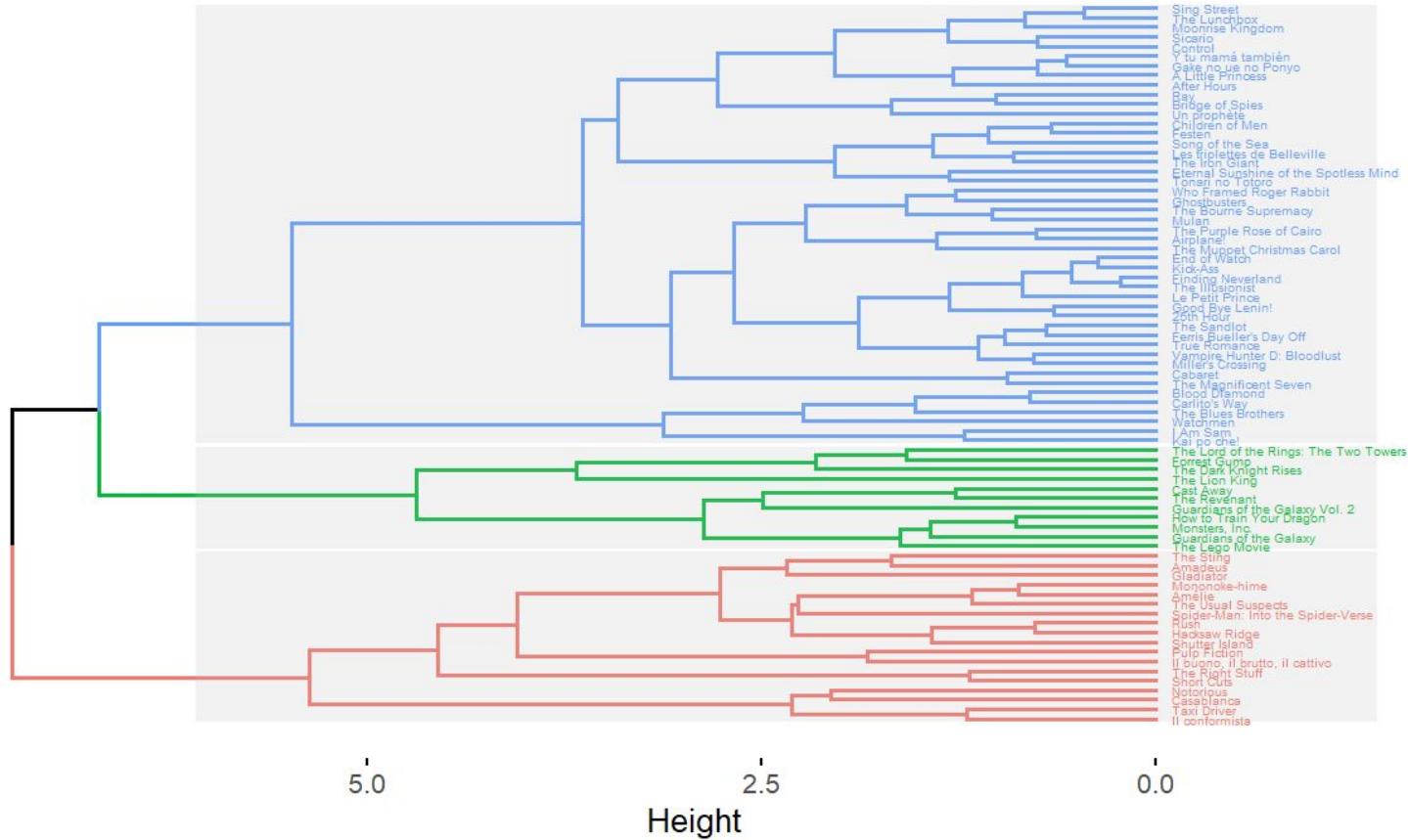
The Matrix

Star Wars: Episode V - The Empire Strikes Back

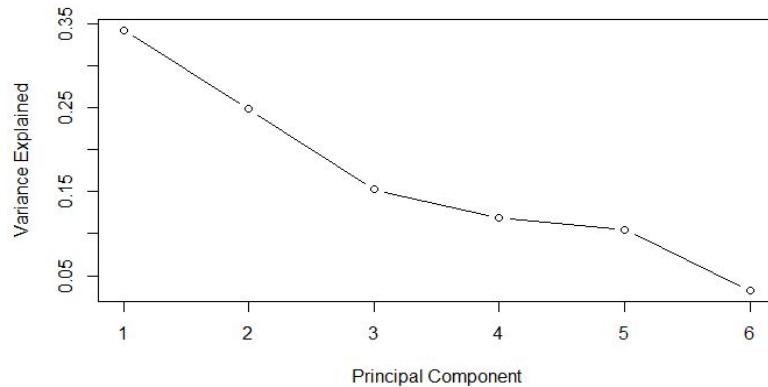
Interstellar

Saving Private Ryan

## Cluster Dendrogram



# Principal Component Analysis



| VarianceExplained<br><dbl> | SummationOfVarExplained<br><dbl> |
|----------------------------|----------------------------------|
| 0.34296273                 | 0.3429627                        |
| 0.24874222                 | 0.5917049                        |
| 0.15284221                 | 0.7445472                        |
| 0.11866335                 | 0.8632105                        |
| 0.10469427                 | 0.9679048                        |
| 0.03209523                 | 1.0000000                        |

Scree plot demonstrates that only 3 or 4 principal components are needed to explain a relatively large proportion of the variance in the dataset based on the 5 numerical variables

Standard deviations (1, ..., p=5):

```
[1] 1.2395715 1.1568047 0.9382903 0.7972365 0.7805708
```

Rotation (n x k) = (5 x 5):

|               | PC1        | PC2        | PC3         | PC4         | PC5        |
|---------------|------------|------------|-------------|-------------|------------|
| Released_Year | 0.4880631  | 0.4417639  | -0.25208172 | -0.62759340 | 0.3304852  |
| Runtime       | -0.3010805 | 0.5112006  | 0.65381430  | 0.10421520  | 0.4579198  |
| IMDB_Rating   | -0.5993749 | 0.2606866  | -0.02676844 | -0.54696280 | -0.5224054 |
| Meta_score    | -0.5584336 | -0.1774663 | -0.53837548 | -0.02427944 | 0.6051611  |
| Gross         | -0.0068283 | 0.6663834  | -0.46735101 | 0.54360641  | -0.2048453 |

- Fairly New
- Low Ratings
- Shorter Runtime
- High Grossing
- Long Runtime
- Newer
- Long Runtime
- Low Grossing
- Low Rating
- Older

# Top 10 Movies by PC Value

1

|    |                                |
|----|--------------------------------|
| 1  | The Butterfly Effect           |
| 2  | Seven Pounds                   |
| 3  | I Am Sam                       |
| 4  | Flipped                        |
| 5  | Jeux d'enfants                 |
| 6  | Saw                            |
| 7  | Kai po chel                    |
| 8  | Fear and Loathing in Las Vegas |
| 9  | Gifted                         |
| 10 | Tropa de Elite                 |



2

|    |   |
|----|---|
| 1  | Avengers: Endgame                             |
| 2  | Star Wars: Episode VII - The Force Awakens    |
| 3  | Avengers: Infinity War                        |
| 4  | Avatar  |
| 5  | Titanic                                       |
| 6  | The Dark Knight                               |
| 7  | The Avengers                                  |
| 8  | The Lord of the Rings: The Return of the King |
| 9  | The Dark Knight Rises                         |
| 10 | Rogue One                                     |

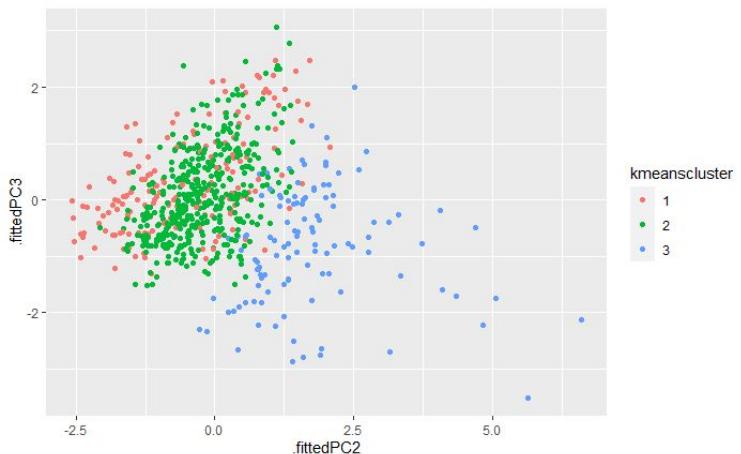
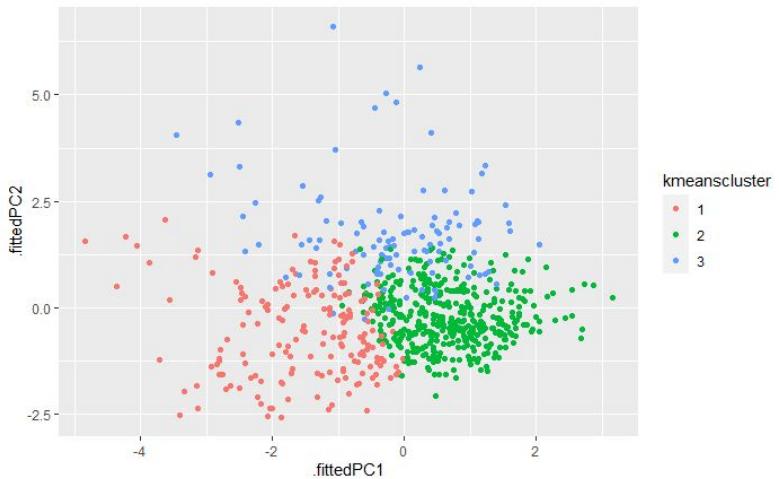
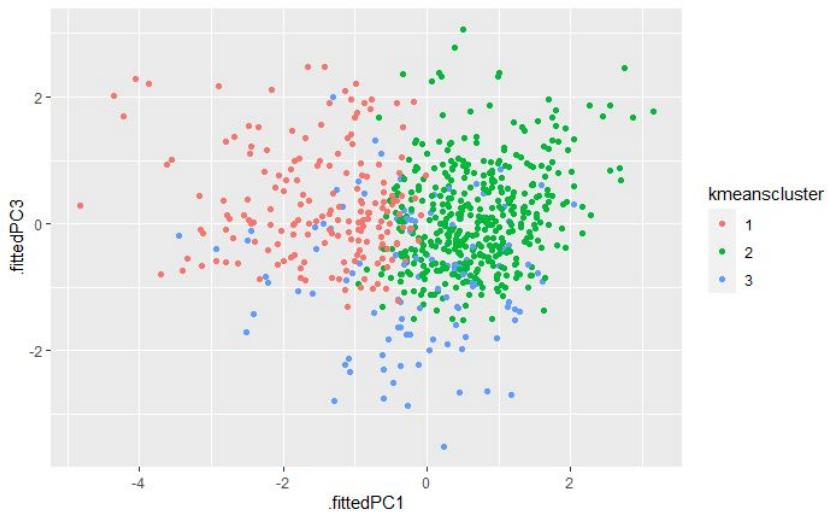


3

|    |                                   |
|----|-----------------------------------|
| 1  | Bound by Honor                    |
| 2  | Kal Ho Naa Ho                     |
| 3  | Doctor Zhivago                    |
| 4  | Lagaan: Once Upon a Time in India |
| 5  | I Am Sam                          |
| 6  | Kelly's Heroes                    |
| 7  | Veer-Zaara                        |
| 8  | Malcolm X                         |
| 9  | My Name Is Khan                   |
| 10 | Dogville                          |



# K Means Clusters compared to Principal Components



# Conclusion



- ❖ Data standardization was extremely important in this dataset, given the differing magnitudes of values based on variables/units
- ❖ By feature engineering the “Genre” variable so that it was more accessible, it dominated the clustering process
- ❖ Clustering based on only numerical values proved more effective, given that our K-Means clusters were well separated based on PCA