

## Starring ...

- Introduction
- Data Description
- Pre-Processing
- Natural Language Processing
- PCA: Eliminating Curse of Dimensionality
- K Means Clustering
- Hierarchical Clustering
- Interpretations and Results
- Conclusions



Have you ever wondered why certain movies get nominated for Oscars? What's so special about them?



## Introduction

- IMDB (Internet Movie Database): World's most popular and authoritative source for movie ratings and user reviews.
- OMDB: Opensource IMDB API used for collecting Top 250 movies
- Goals :
  - Use Clustering to recognize patterns among top rated movies
  - Identify what attributes such as Genre, Plot, Runtime etc contribute towards their high rating
  - Identify similarities of movies within the same cluster based on these attributes

## Data Description

#### Possible Clustering Attributes:

- Year
- Rated
- Runtime
- Director
- Writer
- Actor
- Language
- Genre

Number of Rows: 250 Number of Columns: 26 Number of Duplicated Rows: 0

## Pre-Processing

- Year Continuous to Binary
- Rated Combined less frequent categories(Unrated, TV-PG, X, GP, TV – MA) as 'Others'
- Runtime Binned in 3 categories:
  - Short : <90 minutes
  - Normal : 90 150 minutes
  - Long: >150 minutes

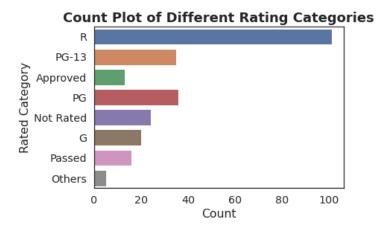
- Directors/Actors Filtered top directors/actors with 4 or more movies
- Language- 5 Most frequent
- Country 10 Most frequent
- Plot Processed using NLP
- All the categorical variables were converted into dummy variables

#### Year

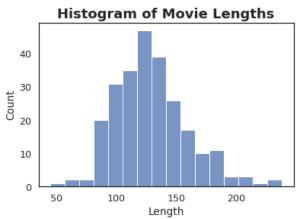
# Year Histogram Count of movies built before and after 1990 140 Binerized Years Before 1990 150 400 200 400 200

2020

## Rating



#### Runtime



1940

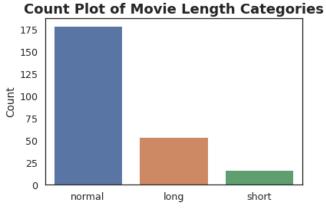
1920

1960

1980

Year

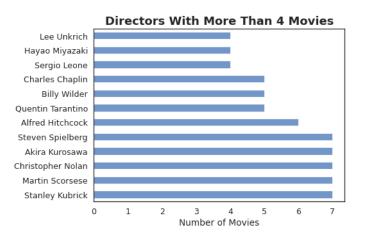
2000



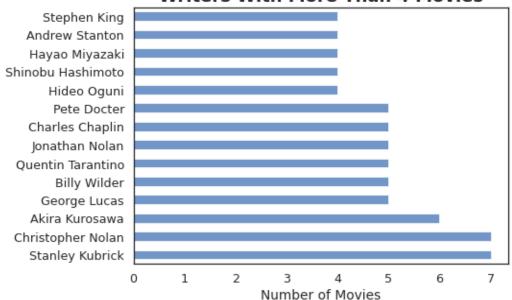
Year Category

1

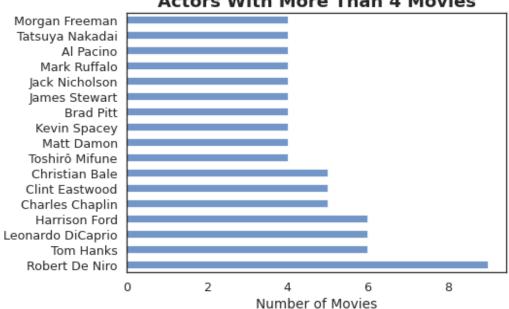
#### **Director**



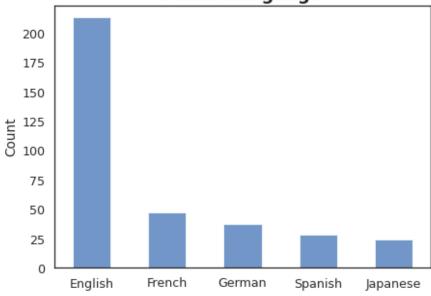
#### Writers With More Than 4 Movies



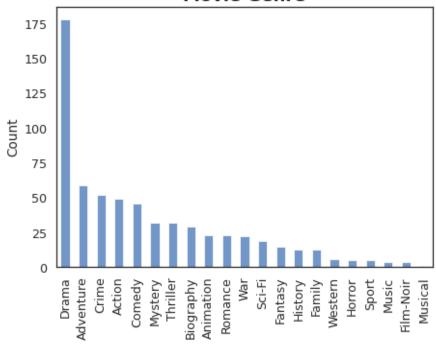
#### Actors With More Than 4 Movies







#### **Movie Genre**



## **NLP**

#### Step 1

- Movie Description Preprocessing:
  - Tokenization
  - Lemmatization
  - Removing Stop words
  - Removing Punctuations

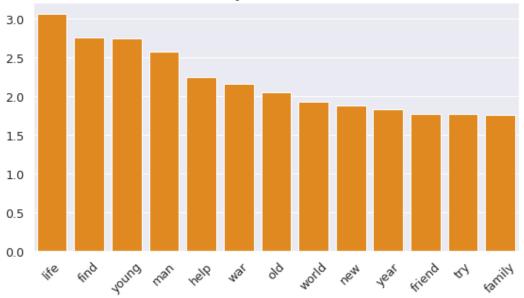
#### Step 2

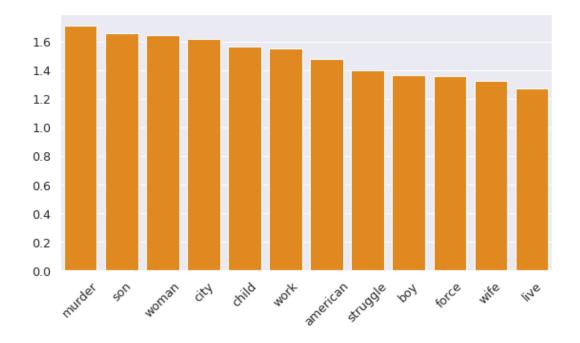
TF-IDF Vectorization

#### Step 3

• Choose 25 top words

Top 25 words

















#### **Final Dataset Attributes for PCA:**

- Year (Binary)
- Runtime
- Country
- Genre
- Actors
- Directors
- Top 25 plot words
- Rated

#### **Final Dataset Dimension:**

• 250 rows and 95 columns





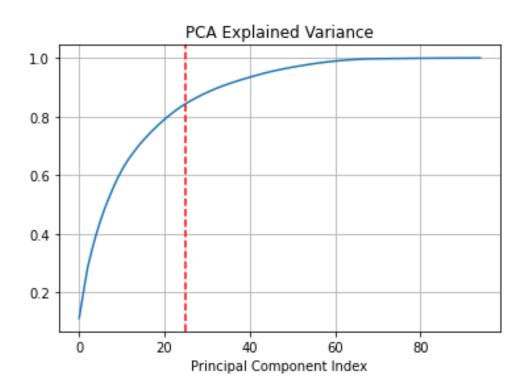
#### **PCA**

#### Purpose:

• Reducing dimension of the data

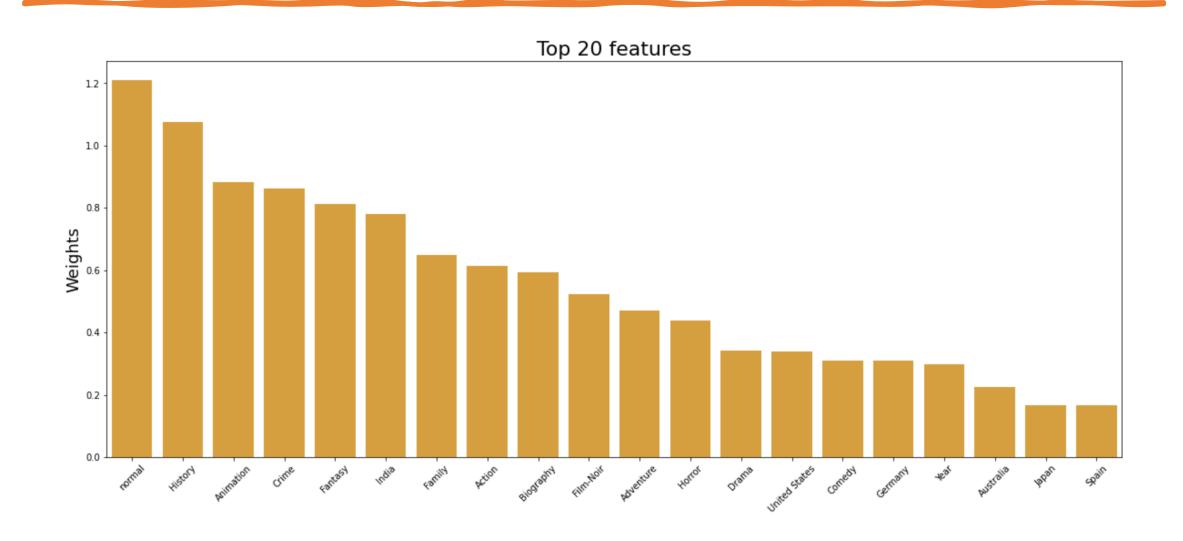
#### **Cutoff Point:**

• 80% of explained variance – 22 PCs



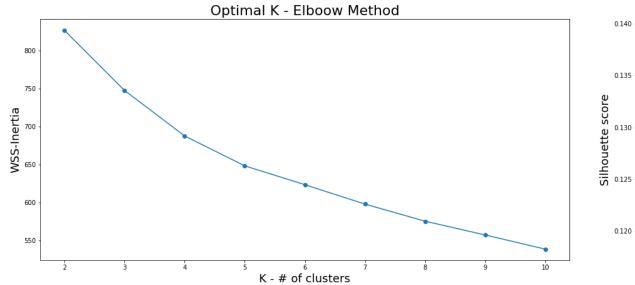
PC	16	17	18	19	20	21	22	23	24	25
EVR	0.72	0.73	0.75	0.76	0.78	0.79	0.8	0.81	0.82	0.83

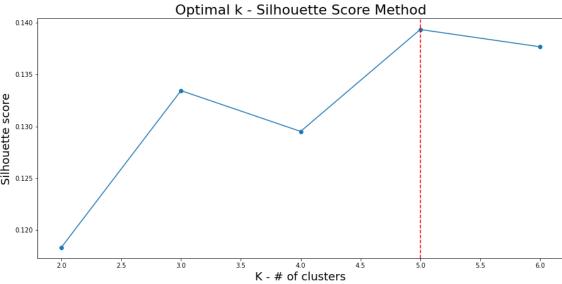
### First 20 Principal Component (PC1-20):



#### K - Means

- Find Optimal K
  - Elbow Method 4 or 5
  - Silhouette Score 5

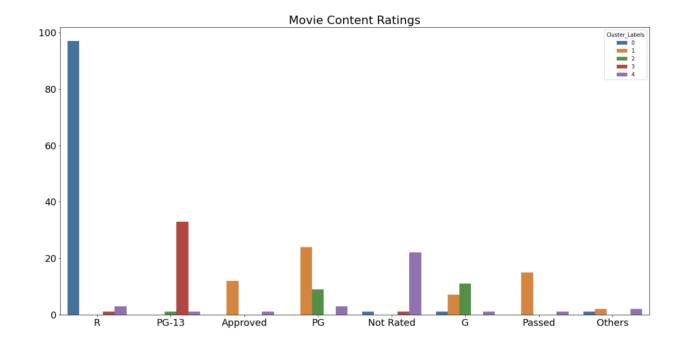


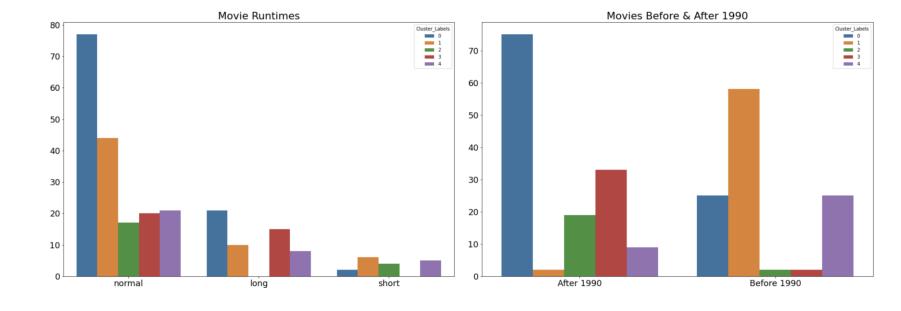


## K-Means

- Mini Batch K-Means
  - K= 5 clusters
- Cluster Sizes:

0	1	2	3	4
100	60	21	35	34

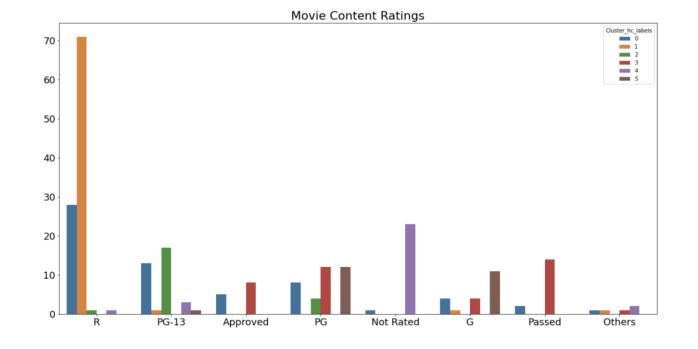


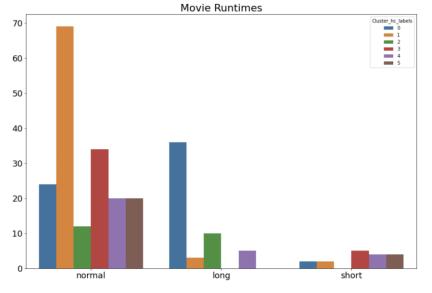


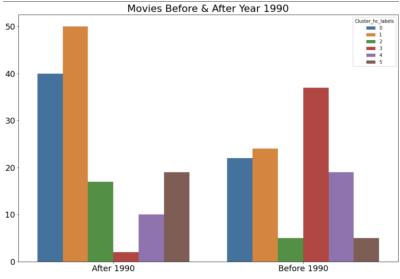
## Hierarchical -Agglomerative

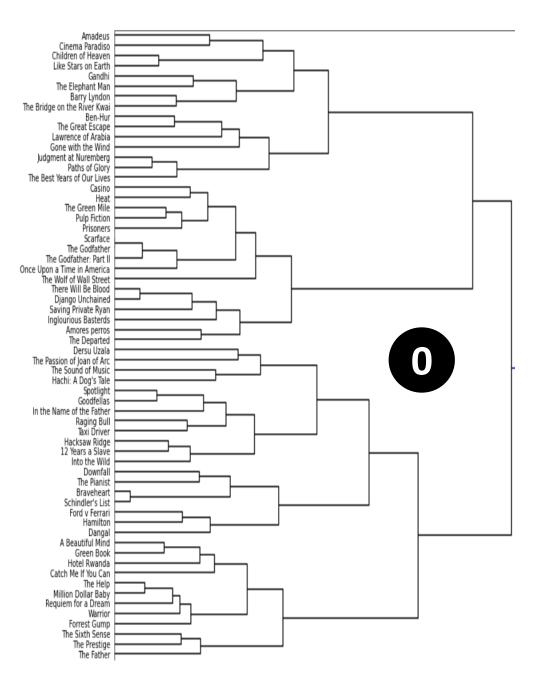
- Dendrogram
  - 6 clusters
  - Linkage : Ward
  - Affinity : Euclidean

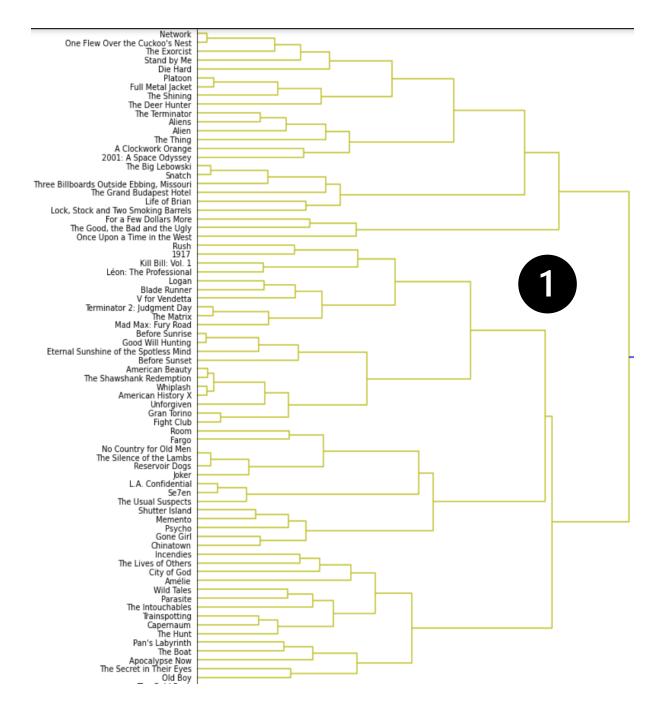
0	1	2	3	4	5
62	74	22	39	29	24

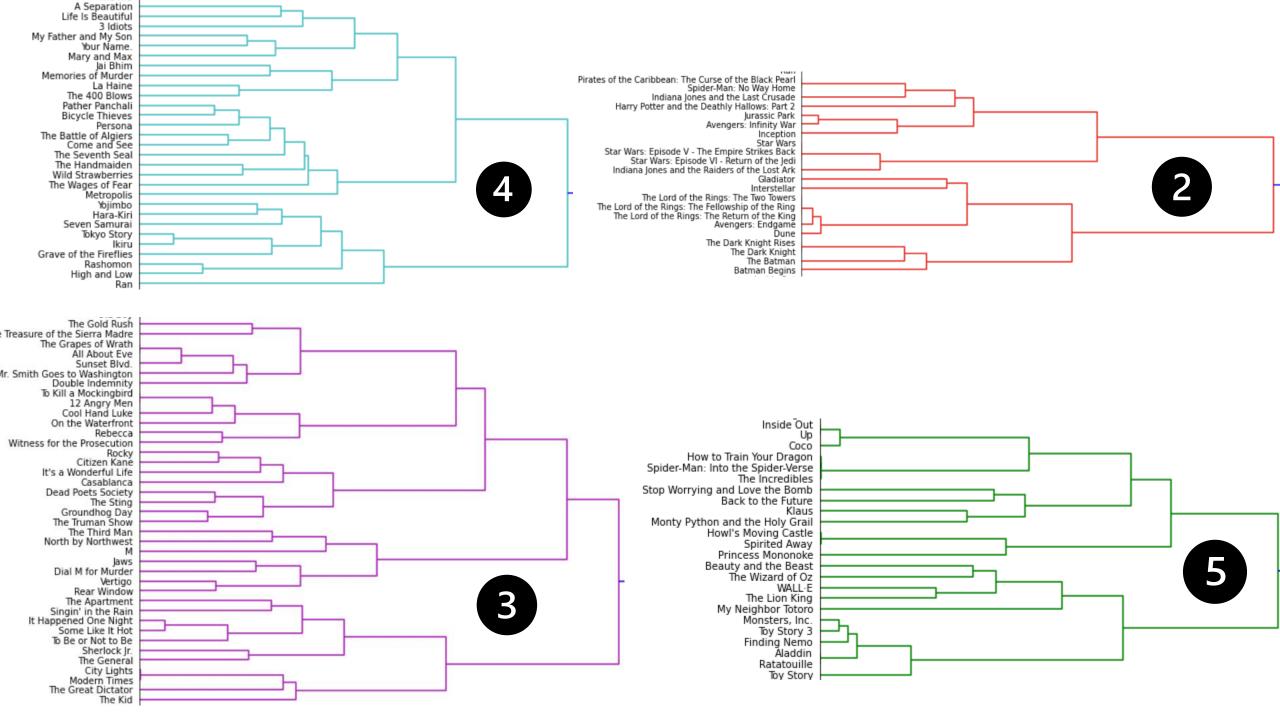


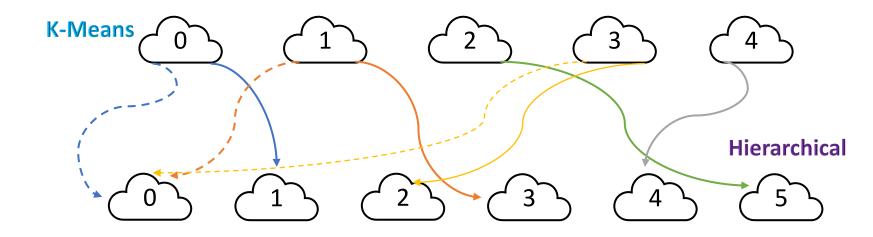












## Cluster Comparison

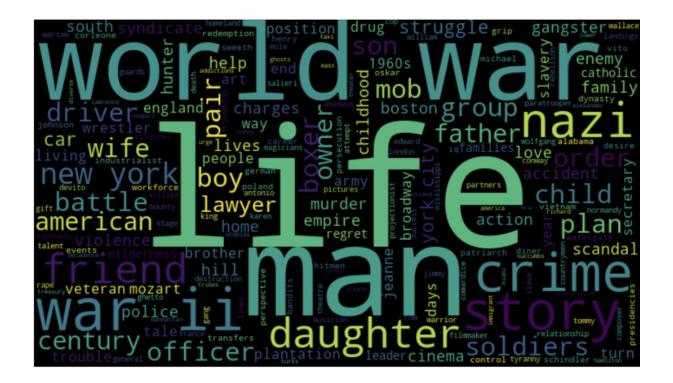
## Hierarchical

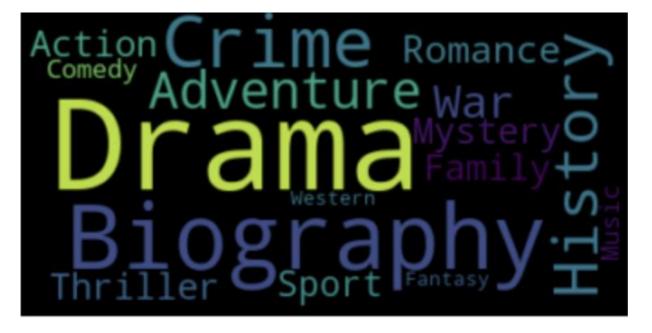
		0	1	2	3	4	5	All
	0	30	69	0	0	1	0	100
<u> </u>	1	12	2	4	39	0	3	60
XU XU	2	0	0	0	0	0	21	21
<u> </u>	3	14	1	18	0	2	0	35
	4	6	2	0	0	26	0	34
	All	62	74	22	39	29	24	250

>> Biographies & Historical Movies



- **Rated** R, PG-13
- Runtime Long (More than 150 minutes)
- Directors Martin Scorsese, Quentin Tarantino, Steven Spielberg
- Actors Robert De Niro, Al Pacino, Tom Hanks, Leonardo Di Caprio
- Movies The Godfather: I &II, Casino, The Wolf of Wall Street, Catch me If you Can, Gandhi, Dangal

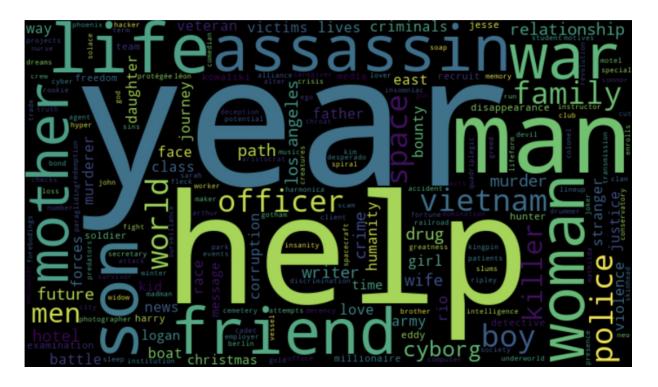


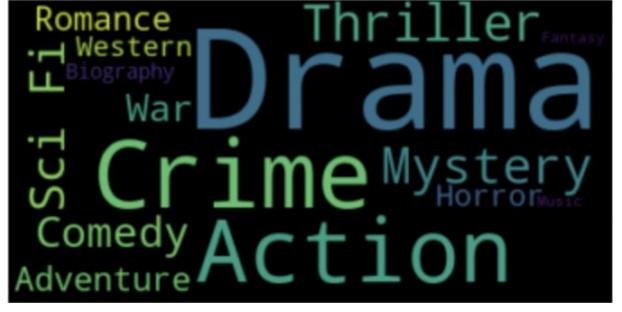


>> Rated R, Crime Movies



- **Rated** R
- Runtime Normal (90-150 minutes)
- **Directors** Stanley Kubric, Sergio Leone
- Actors Kevin Spacey, Clint Eastwood, Morgan Freeman
- Movies Se7en, The Terminator, A Clockwork Orange, The Matrix

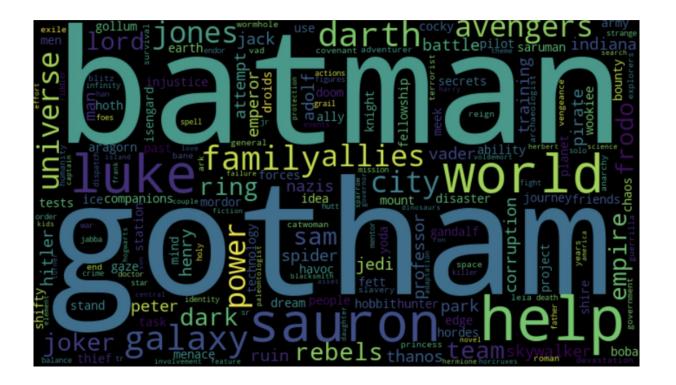






>> Superhero/Action Movies

- **Year** After 1990
- **Rated** PG -13, PG
- Directors Christopher Nolan, Steven Spielberg
- Actors Harrison Ford, Christian Bale
- Movies Batman Trilogy, Interstellar, Avengers-Endgame & Infinity War, Star Wars, Indiana Jones

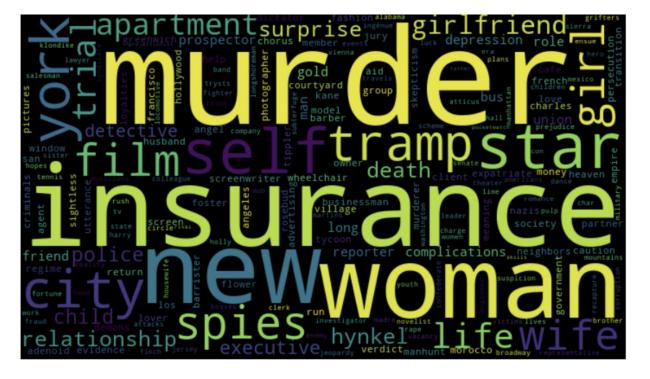




>> Retro Movies



- **Year** Before 1990
- Rated –PG, Passed, Approved
- **Directors** –Alfred Hitchcock, Billy Wilder, Charles Chaplin
- Actors Charles Chaplin, James Stewart
- Movies Vertigo, The Apartment, Modern Times, Rear Window

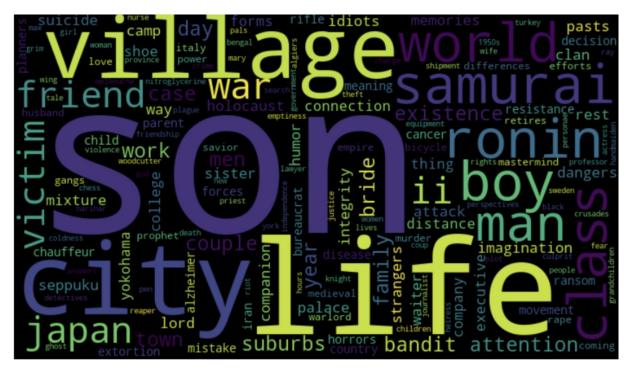


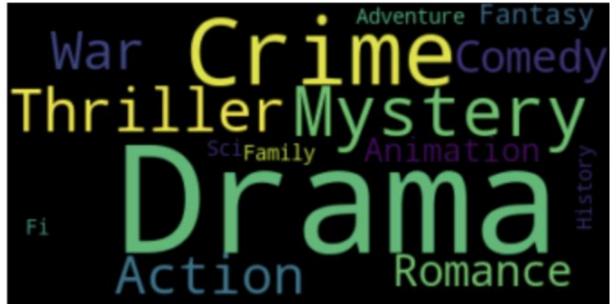




>> Movies from around the world

- Rated –Not Rated
- Country Japan, France, India, Australia, Italy
- Directors Akira Kurosawa
- Actors Toshiro Mifune, Tatsuya Nakadai
- Movies Seven Samurai, 3 Idiots, Yojimbo, Life is Beautiful, A Separation, Ran

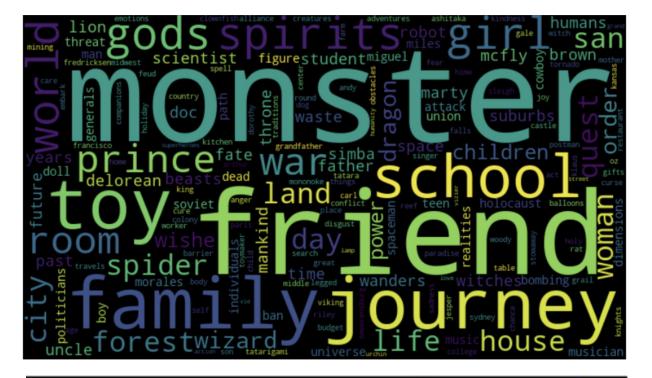




>> Animated/Kids/Family Movies



- Year Mostly after 1990
- Rated –PG, G
- **Directors** Hayao Miyazaki, Lee Unkrich
- Actors Tom Hanks
- Movies Toy Story 1 &3, Spirited Away, Finding Nemo, Aladdin, Wall-E, The Incredibles, Beauty and the Beast





## Conclusions

- Hierarchical Clustering : More Defined Clusters
- Dominating Genre: Crime, Drama, Action & Biography Movies
- Superhero Action & Sci-Fi Movies: More popular after 1990.
   Thanks to better VFX, DC & Marvel.
- Retro Movies : Popular for Comedy & Romance Dramas.
- International Movies: Dominantly from Japan
- Kids & Family Movies: Mostly Anime with Adventurous Plots