



# **Capstone Project-4**

Presentation on

# Netflix Movies and TV Shows Clustering

Presented By:

Md Shahrukh Khan

Anis Bagwan

Ankit Rai

Dayanand Kshirsagar

# Agenda



- Defining ProblemStatement
- Data Pipeline
- Data Summary
- EDA
- Applying Model Clustering
- Conclusion



# ΑI

### Problem Statement

- Netflix has become dominant company in the on-demand media industry, with 167 million paying subscribers around the world.
- We have been provided a dataset collected from Flixable which is a third-party Netflix search engine. Our job is to:

- > To perform Exploratory Data Analysis
- Understanding what type content is available in different countries
- > Is Netflix has increasingly focusing on TV rather than movies in recent years.
- > Clustering similar content by matching text-based features

# Data pipeline



- Data Pre-processing: After exploring and understanding our data, we did data cleaning by handling Null/missing values, checking for duplicate values. We further changed "date\_added" variable to its appropriate datetime format and created a new variable "year\_added" by extracting year from it.
- **EDA:** We performed exploratory analysis of data and find useful insights.
- Creating a model: After identifying useful features, we performed text cleaning- by removing stopwords, punctuation and doing stemming of words. After calculating clean text lengths, we standardize those values and applied two clustering algorithms-K-means and HAC (Hierarchal Agglomerative Clustering).

### Data summary

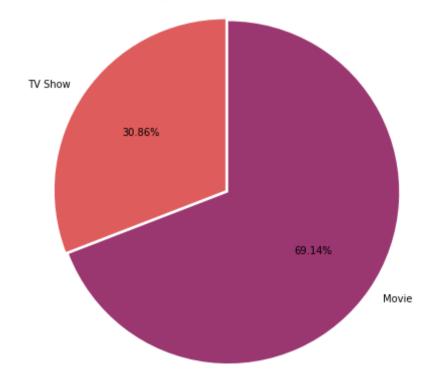


- **show\_id**: Unique ID for every Movie / Tv Show
- **type**: Shows it is a Movie or TV Show
- **title**: Gives title of the Movie / Tv Show
- **director**: Director of the Movie
- cast: Actors involved in the movie / show
- **country**: Country where the movie / show was produced
- date added: Date it was added on Netflix
- release\_year: Actual Release year of the movie / show
- rating: TV Rating of the movie / show
- **duration**: Total Duration in minutes or number of seasons
- **listed in**: Genre of Movies and TV Shows
- **description**: The Summary description of movies and TV shows

- Clearly number of Movies on Netflix outnumbered the number of TV
   Shows.
- Almost 70%
   content are movies
   while rest 30% are
   TV Shows.

Share of TV Show and Movie in dataset

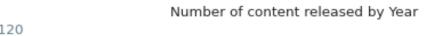


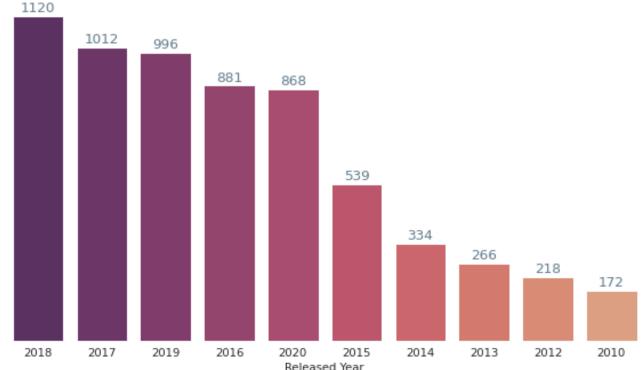




- As per the contents available on Netflix, most of them are from recent years – i.e 2018,2017 and 2019.
- The trend shows that
   as we go from year
   2010 to 2018, the
   number of contents
   on the basis of
   respective year
   release increases.

Distribution of number of content released per year



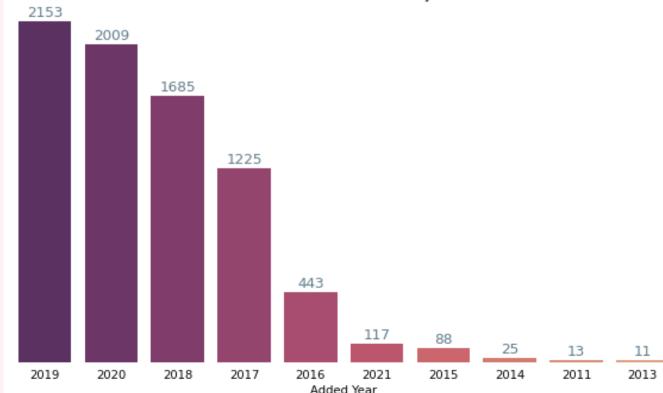




- Clearly, we can see the number of contents on the platform is increasing year on year.
- There is drastic increase in 2016-2017, this is probably Netflix was launched in India this year. and to attract Indian viewers it started adding Indian contents as well.

Distribution of content added on Netflix by year



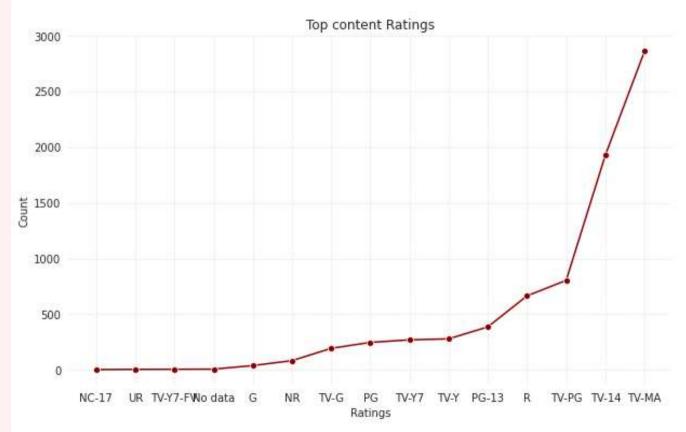






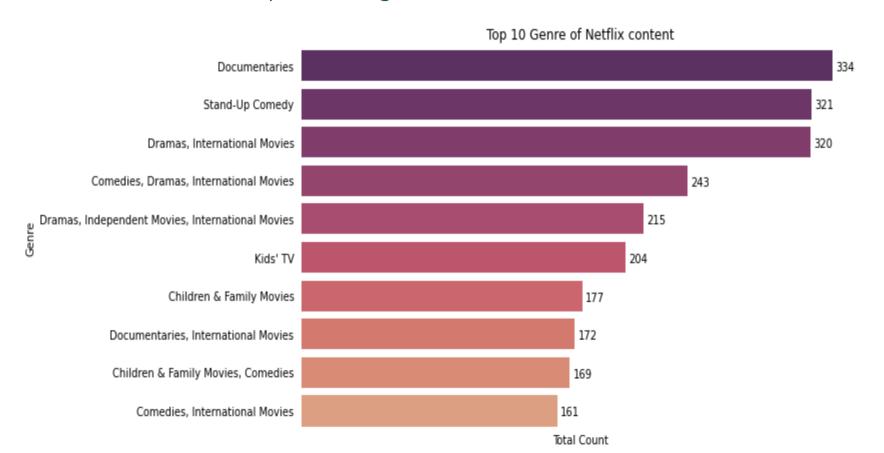
Distribution of ratings of content added on Netflix

- The plot shows with a rating of TV-MA are in the clear majority. This is followed by TV-14.
- These top two ratings have way too much numbers of contents compared to other ratings.





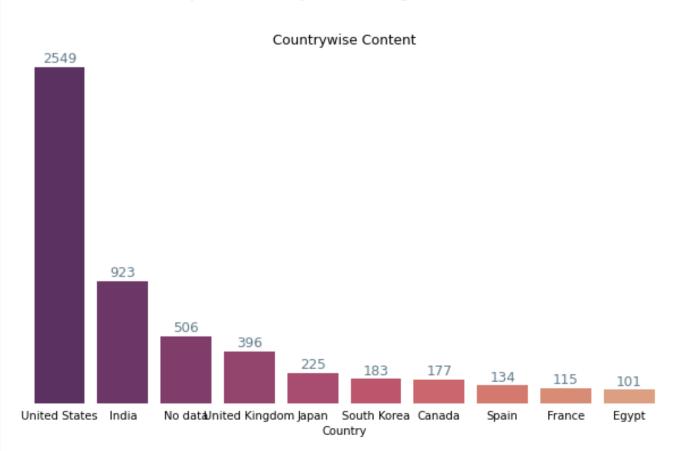
Top content genre available in Netflix





Top content producing countries

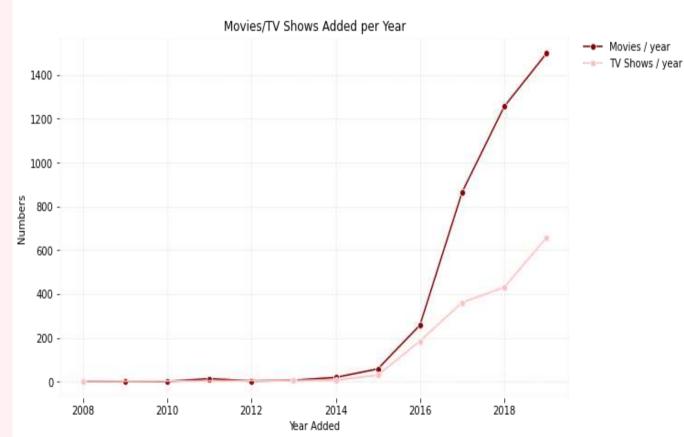
- The United States account for the majority of the content created on Netflix, numbering 2549 titles.
- India is the second largest with 923 titles.





Content added on Netflix per year

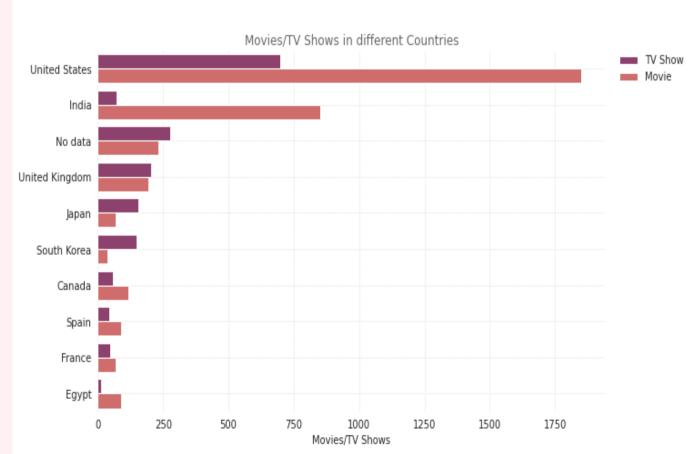
- As can be seen in this plot, both TV shows and Movies content numbers increased drastically after 2016.
- Also the number of movies added were much higher compared to TV shows number.



- The United States
   account for the majority
   of the content created
   on Netflix.
- India is the second largest.
- In both countries, number of movies outnumbers the number of shows.



Distribution of Movies/TV Shows produced in various countries



### Data Pre-processing



We cannot go straight from raw text to fitting a machine learning model. We must clean text first, which means splitting it into words and handling punctuation.

For clustering we choose "description" and "Listed\_in" variables. Before clustering we need to pre-process the data. So that we filtered data with following steps:

Text Example: "After an awful accident, a couple admitted to a grisly hospital are separated and must find each other to escape — before death finds them."

# 1. Remove Punctuation

Text after removing punctuations:

After an awful accident a couple admitted to a grisly hospital are separated and must find each other to escape — before

death finds them

# 2. Remove Stop-words

Text after removing stopwords:
 awful accident couple admitted grisly hospital separated must find escape — death finds

#### 3. Stemming

Text after removing stopwords:

aw accid coupl

admit grisli hospit

separ must find

escap — death find

# 4. Length of processed text

Calculate the length of text we got from first three steps to do clustering The **K-Means** algorithm searches for a predetermined number of clusters within an unlabelled multidimensional dataset

The **Elbow Method** is one of the most popular methods to determine this optimal value of k number of clusters.

To determine the optimal number of clusters, we have to select the value of k at the "elbow" i.e. the point after which the distortion/inertia start decreasing in a linear fashion.

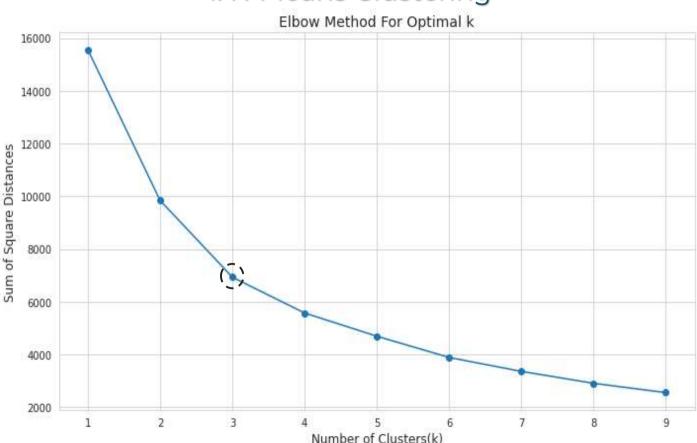
Thus from this chart we need to check, which would be the best number of clusters from 2,3,4,5, and 7.

We found elbow formation at k=3.

### **Applying Model**



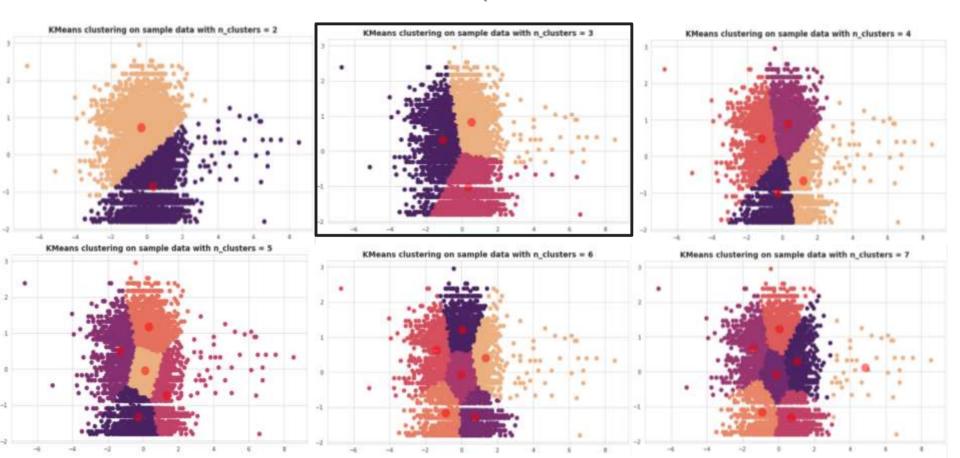
#### 1. K-Means Clustering



### **Applying Model**



### Different clusters to check optimum number of clusters



#### Silhouette Score for K-Means



#### Let's see the Silhouette scores of 2,3,4,5,6,7 and 8 number of clusters

- Silhouette score is used to evaluate the quality of clusters created using clustering algorithms such as K-Means in terms of how well samples are clustered with other samples that are similar to each other.
- The Silhouette score is calculated for each sample of different clusters.

- For n\_clusters = **2**, silhouette score is **0.3551**415129079424
- For n\_clusters = **3**, silhouette score is **0.3558**9698124108055
- For n\_clusters = 4, silhouette score is 0.3266968907071311
- For n\_clusters = **5**, silhouette score is **0.3355**8430881056384
- For n\_clusters = **6**, silhouette score is **0.3557**380959007992
- For n\_clusters = **7**, silhouette score is **0.3548**817152999796
- For n\_clusters = **8**, silhouette score is **0.3522**803075712804

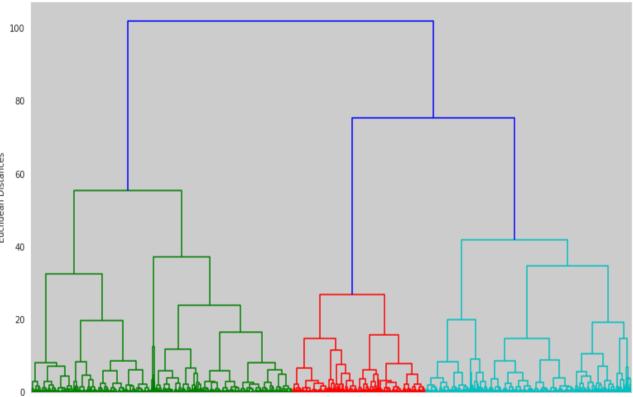
- Hierarchical
   Agglomerative clustering
   starts with treating each
   observation as an
   individual cluster, and then
   iteratively merges clusters
   until all the data points are
   merged into a single
   cluster.
- Dendrograms are used to represent hierarchical clustering results.
- The number of appropriate clusters will be the number of vertical lines which are being intersected by the line drawn using the threshold in this case 3.

# **Applying Model**



#### 2. Hierarchical Agglomerative Clustering

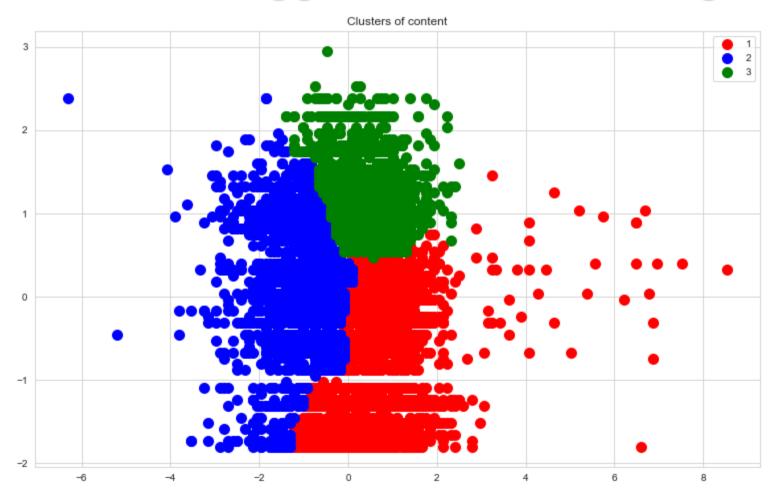




Content

# Hierarchical Agglomerative Clustering





### Conclusion



- We started this project with the intention to obtain some useful insights related to the type of Netflix content. For this, we performed exploratory data analysis on our data after cleaning and making it easy to analyze. This analysis helped us to understand the trend.
- ✓ We found that most of the content on Netflix are of TV-MA and TV-14 rating.
- ✓ USA and India are two countries producing the maximum number of content.
- ✓ Documentaries and Stand up are top genre in terms of number of contents they have on platform. Further we found number of movies on Netflix outnumbers TV-shows.
- Our next job was to make an unsupervised clustering model. For this, we processed our text by removing unuseful characters like - stopwords, punctuation and did stemming. After getting the length for each text feature we rescaled them for generalization and started applying algorithms.
- We first used K-means clustering. In order to find appropriate cluster number we used elbow method and finally got the best silhouette score of around 0.35. Next, we applied Hierarchal Agglomerative Clustering for which we made dendrogram. We also obtained silhouette score of around 0.32.



# Thank You