

Capstone Project-4

NETFLIX MOVIES AND TV SHOWS CLUSTERING

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ΑI

Point for Discussion

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- Dataset Processing
- Data Encoding
- □ EDA
- Implementation

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K Means Clustering

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Problem Statement

This dataset consists of ty shows and movies available on Netflix as of 2019. In 2018, they released an interesting report which shows that the number of TV shows on Netflix has nearly tripled since 2010. The streaming service's number of movies has decreased by more than 2,000 titles since 2010, while its number of TV shows has nearly tripled. It will be interesting to explore what all other insights can be obtained from the same dataset.

Understanding what type content is available in different countries

Is Netflix has increasingly focusing on TV rather than movies in recent years.

Introduction



We all watch a lot of TV shows. Many online streaming services offer a large number of TV shows, which are at our disposal to watch, at the price of a subscription cost. The major online streaming services across the world are Netflix, Prime Video, Hulu, and Disney+. Netflix is a popular entertainment service used by people around the world.



Dataset

The dataset is collected from Flixable which is a third-party Netflix search engine. Integrating this dataset with other external datasets such as IMDB ratings, rotten tomatoes can also provide many interesting findings. It includes over 6,234 entries and 12 columns.



Dataset Variable Information

show_id: Unique ID for every Movie / Tv Show

type: Identifier - A Movie or TV Show

title: Title of the Movie / Tv Show

director: Director of the Movie



Dataset Variable Information

cast: Actors involved in the movie / show

country: Country where the movie / show was produced

date_added : Date it was added on Netflix



Dataset Variable Information

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: Genere

description: The Summary description



Dataset Inspection & Processing

After data inspection we found that there are 6,234 entries and 12 columns to work with for this EDA. Few columns that contain null values, "director," "cast," "country," "date_added," "rating."

NaN Values Processing

The easiest way to get rid of them would be to delete the rows with the missing data for missing values. However, this wouldn't be beneficial to our EDA since it is a loss of information. Since "director," "cast," and "country" contain the majority of null values, we chose to treat each missing value is unavailable. The other two label "date_added" and "rating" contain an insignificant portion of the data, so it drops from the dataset. Finally, we can see that there are no more missing values in the data frame.



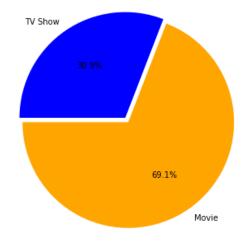
Data Encoding

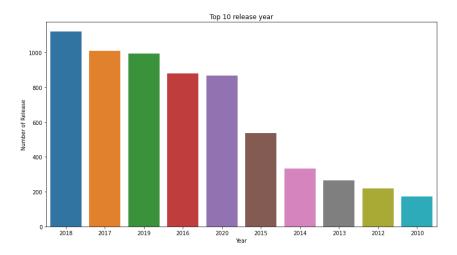
Label Encoding refers to converting the labels into a numeric form so as to convert them into the machine-readable form. Machine learning algorithms can then decide in a better way how those labels must be operated. It is an important pre-processing step for the structured dataset in supervised learning. LabelEncoder can be used to normalize labels. It can also be used to transform non-numerical labels (as long as they are hashable and comparable) to numerical labels. Here we convert type, country, target_ages, rating, listed_in into numerical values using labelencoder.

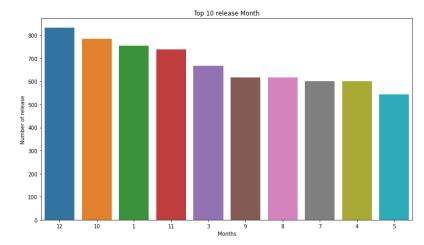




Percentage of Netflix Titles that are either Movies or TV Shows



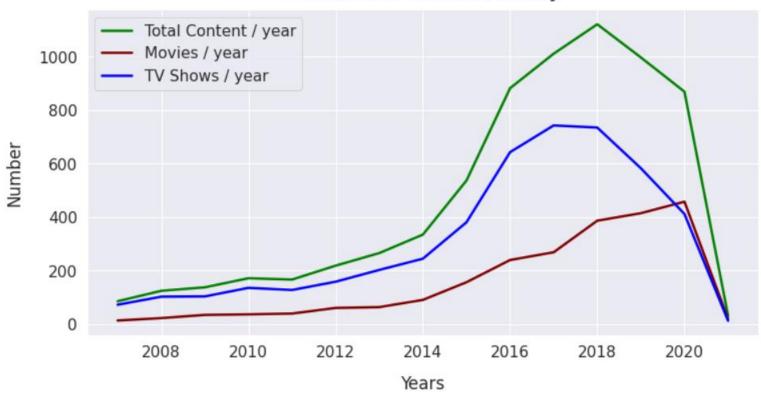






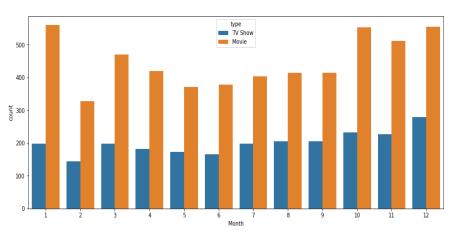


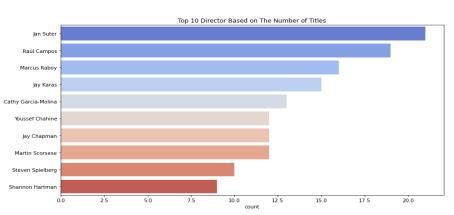
Production Growth Yearly

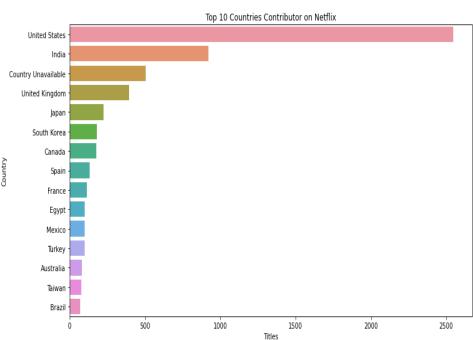














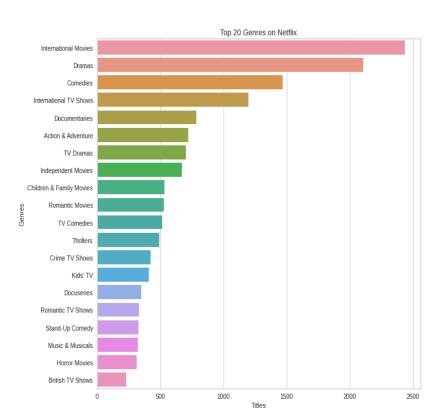


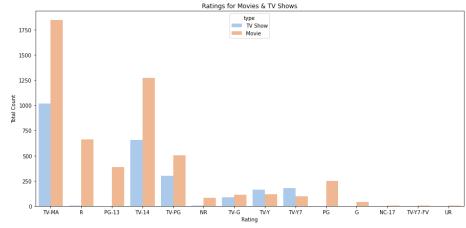
It seems like words like "Love", "Man", "World", "Story", "Christmas" are very common in titles.

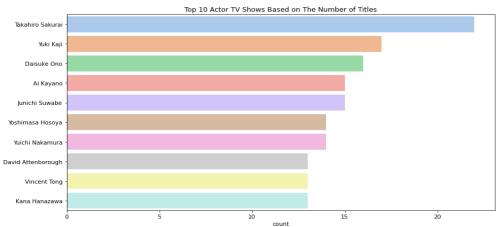






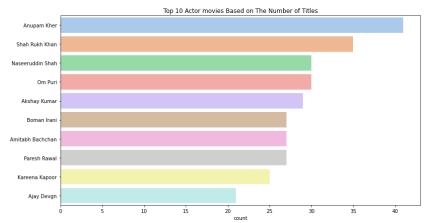


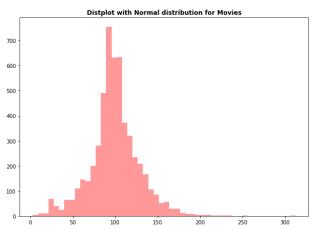


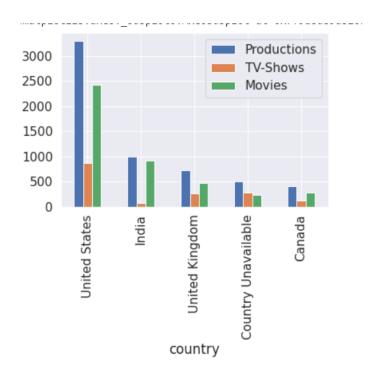






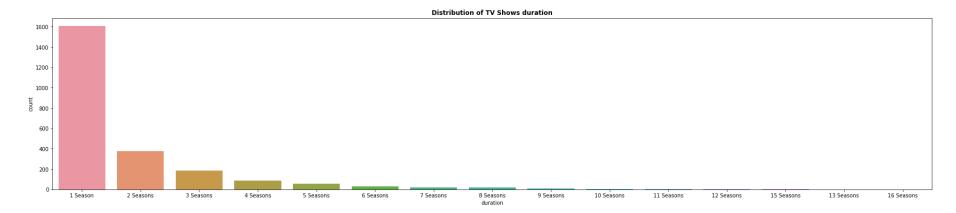








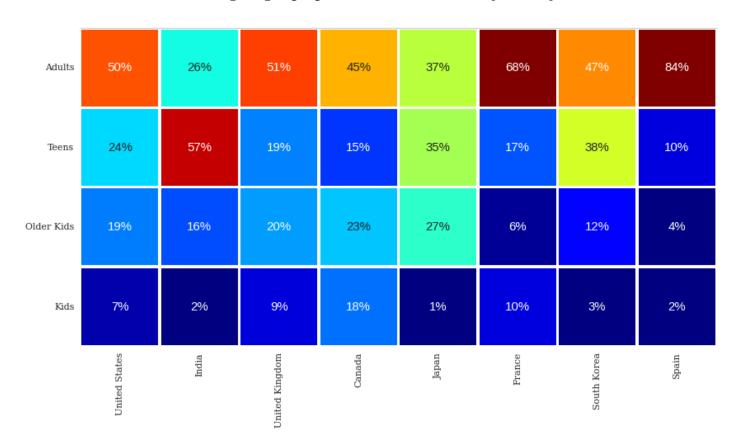






EDA (Correlation Heatmap)

Target ages proportion of total content by country





TV-MA

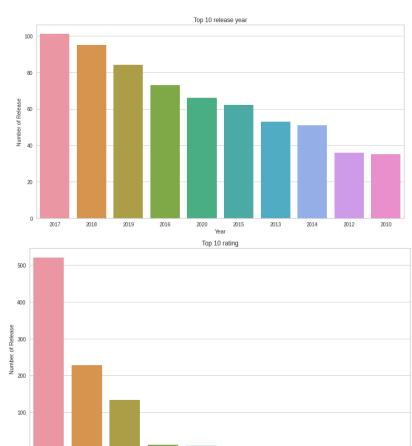
TV-PG

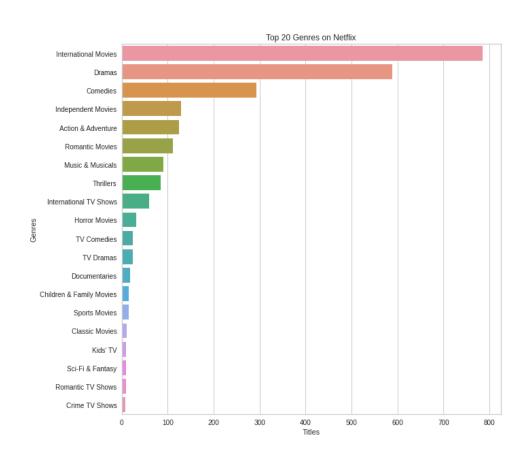
TV-Y7

Year

PG-13



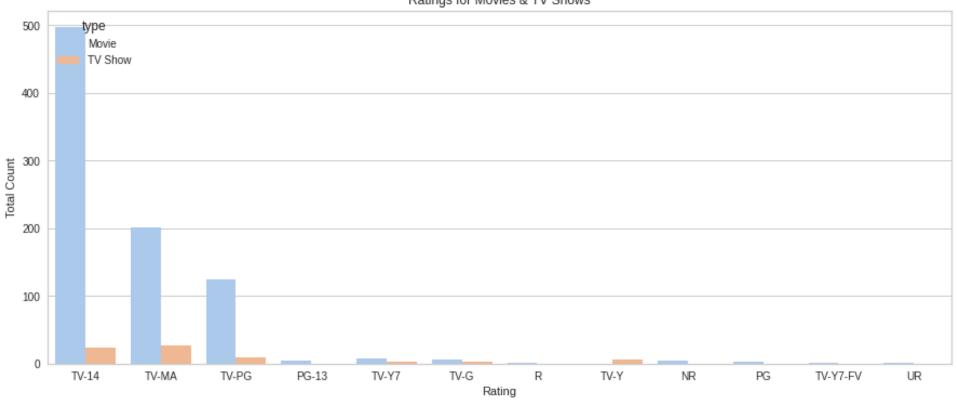




EDA(India)







Implementation



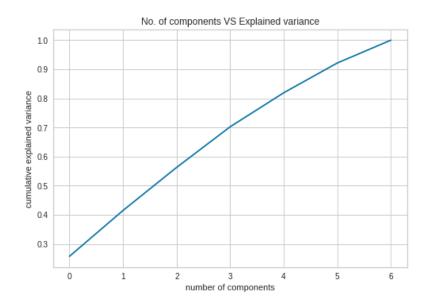
Now Lets implement 3 models on our dataset:

- 1. Principal Component Analysis
- 2. K Means Clustering
- 3. Hierarchical Clustering



1. Principal Component Analysis (PCA)

PCA is fundamentally a dimensionality reduction algorithm, but it can also be useful as a tool for visualization, for noise filtering, for feature extraction and engineering, and much more.





2. K Means Clustering

K-Means Clustering is an Unsupervised Learning algorithm, which groups the unlabeled dataset into different clusters. Here K defines the number of pre-defined clusters that need to be created in the process.

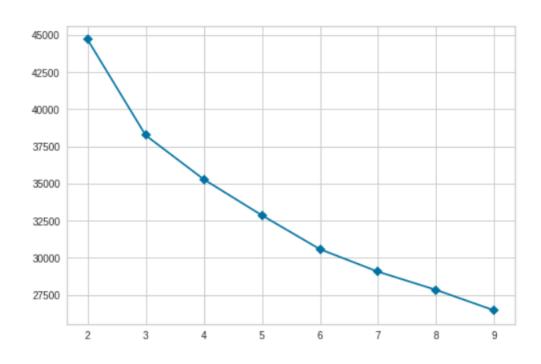
Find out how many clusters are used?

K-Elbow Method

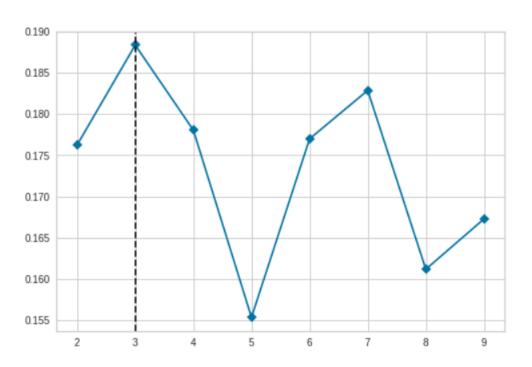
Silhouette Score

K-Elbow Method











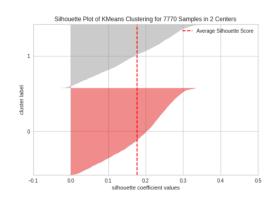
Silhouette Score

The silhouette plot displays a measure of how close each point in one cluster is to points in the neighboring clusters and thus provides a way to assess parameters like number of clusters visually. This measure has a range of [-1, 1].

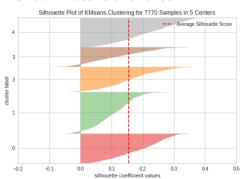
Silhouette coefficients (as these values are referred to as) near +1 indicate that the sample is far away from the neighboring clusters. A value of 0 indicates that the sample is on or very close to the decision boundary between two neighboring clusters and negative values indicate that those samples might have been assigned to the wrong cluster.



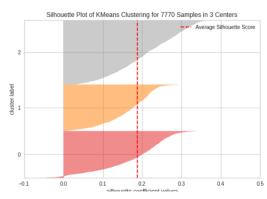
Silhouette Score



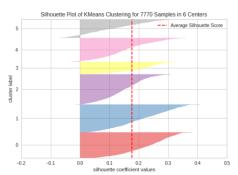
For n_clusters = 2, silhouette score is 0.17705579846281397



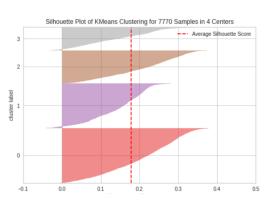
For n_clusters = 5, silhouette score is 0.1553479962910114



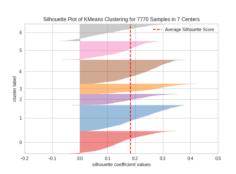
For n_clusters = 3, silhouette score is 0.1883670258470756



For n_clusters = 6, silhouette score is 0.1772754775006096



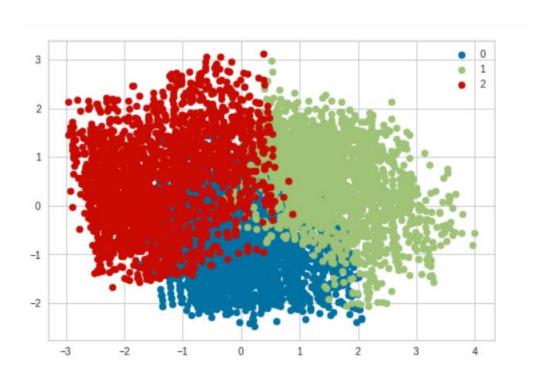
For n_clusters = 4, silhouette score is 0.17907236997528836



For n_clusters = 7, silhouette score is 0.18256493863062262

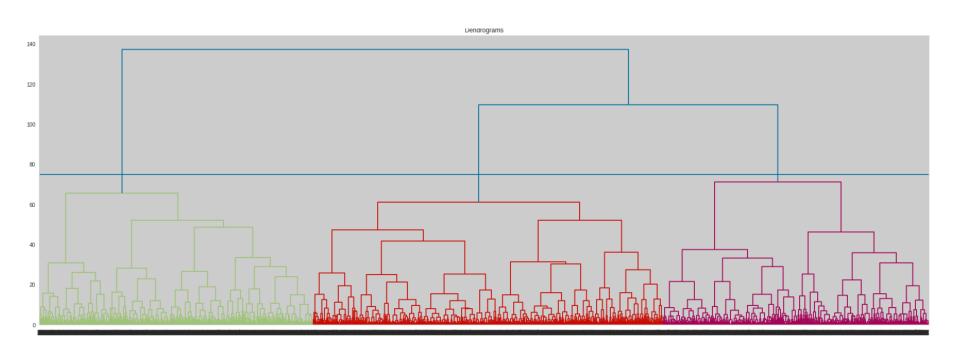


K Means Clustering



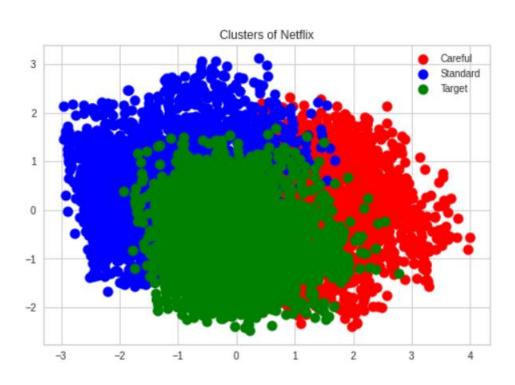
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Hierarchical clustering



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Agglomerative Clustering





Conclusion

We have drawn many interesting inferences from the dataset Netflix titles; here's a summary of the few of them:

The most content type on Netflix is movies. It appears that Netflix has focused more attention on increasing Movie content than TV Shows. Movies have increased much more dramatically than TV shows

There are about 70% movies and 30% TV shows on Netflix.

Most films were released in the years 2018, 2019, and 2020.

The number of releases have significantly increased after 2015 and have dropped in 2021 because of Covid 19.

The months of October, November, December and January had the largest number of films and television series released.



Conclusion

More of the content is released in holiday season - October, November, December and January.

The United States has the highest number of content on Netflix by a huge margin followed by India.

Raul Campos and Jan Sulter collectively have directed the most content on Netflix.

Anupam Kher has acted in the highest number of films on Netflix. Drama is the most popular genre followed by comedy.

International movies are the top most genre in netflix which is fllowed by standup comedy and Drams.

Most of the movies have duration of between 50 to 150

Highest number of tv_shows consisting of single season



Conclusion

Using correlation heat map we see that in India mostly teens watching netflix so question arises that what content teens watched.

TV-MA has the highest number of ratings for tv shows i,e adult ratings

In India teens mostly watched international movies.

Principal Component analysis (PCA)reduced the number of components as 7 with approximately 99% of variance.

For K Means clustering to find out number of k we used elbow and silhouette score method.

Using both the methods we found k=3 is optimal value of clustering.

Using Hierarchical clustering method again we find out that k=3 is optimal value of clustering.



Thank You