

# Netflix Movies and TV Shows Clustering

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## Introduction:

Netflix is a media distribution company. It started with DVD distribution via mail, but has evolved substantially over the course of its existence. Today, Netflix is focused on streaming video. Some of its content is licensed, and some of the content is produced in-house.

Netflix originally focused on movies, but today television shows are probably the more common format. Netflix works on a subscription model, where users get unlimited access to content with a paid subscription.



## Dataset Preview:

This dataset consists of tv shows and movies available on Netflix as of 2019. The dataset is collected from Flixable which is a third-party Netflix search engine.

### Attribute 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
- **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
- **description**: The Summary description

## Dataset summary:

The dataset contains 12 columns and 7787 rows.

There also exist some null values in our data:

Percentage of null values in director : **30.68%**

Percentage of null values in cast : **9.22%**

Percentage of null values in country : **6.51%**

Percentage of null values in date\_added : **0.13%**

Percentage of null values in rating : **0.089%**

*Since there are very few null values in date\_added and rating we will drop them.*

Finally we will do some feature engineering to create few new variables:

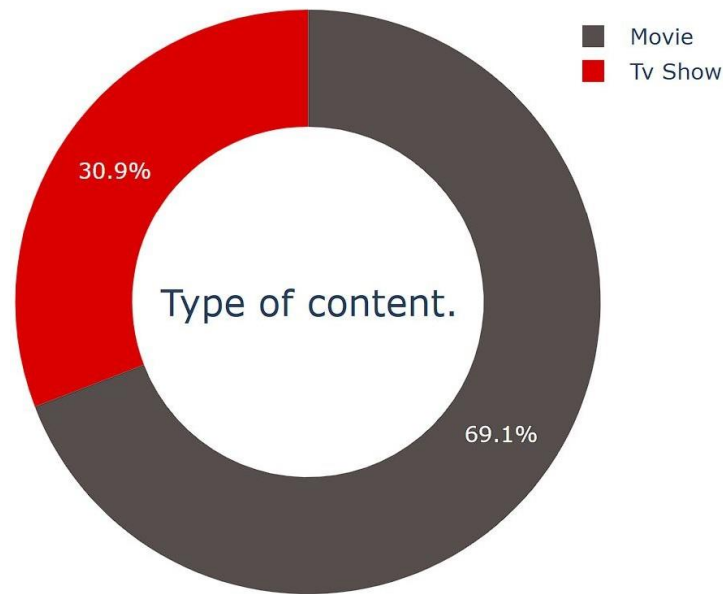
- *Compute **year\_added**, **month\_added** and **day\_added** from **date\_added** after converting it into datetime variable.*

# Exploratory Data Analysis:

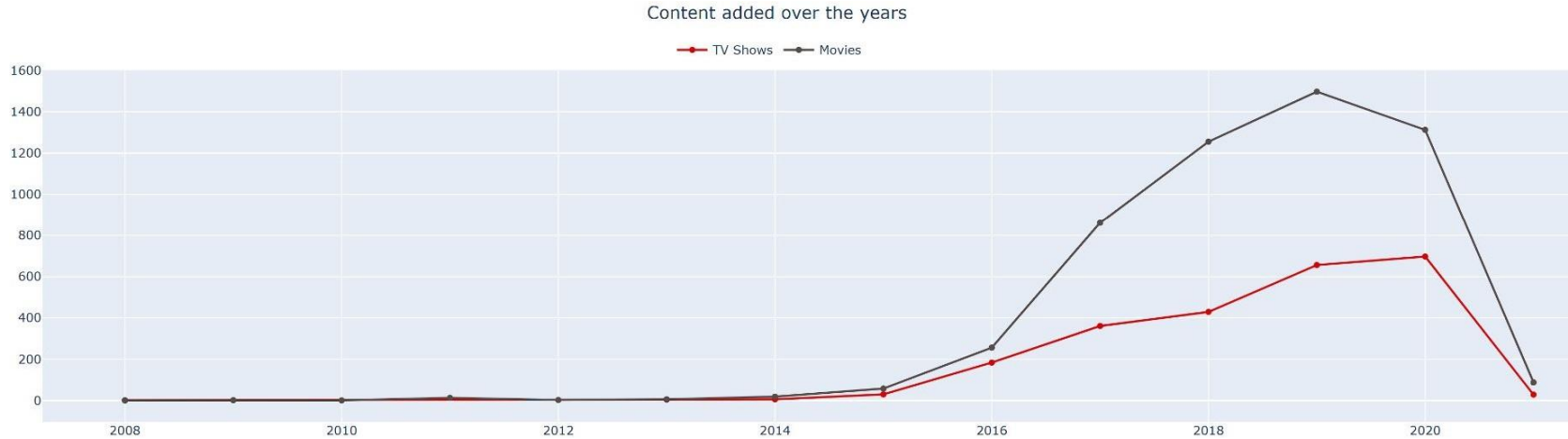
## Type:

69.1% of the content available on Netflix are movies; the remaining 30.9% are TV Shows.

Type of content watched on Netflix



## Year\_added:

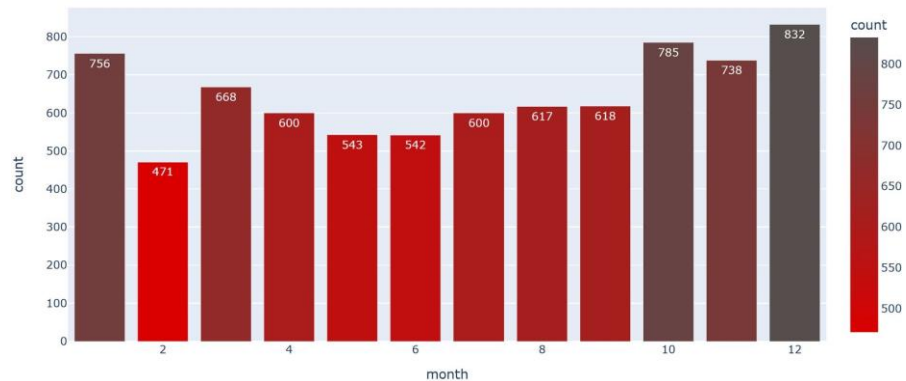


- Growth in the number of movies on Netflix is much higher than tv shows.
- From 2015 we can see a noticeable addition in the number of movies and tv shows uploaded by Netflix on its platform.
- The highest number of movies and tv shows got added in 2019 and 2020.
- The line plot shows very few movies, and tv shows got added in 2021. It is due to very little data collected from the year 2021.

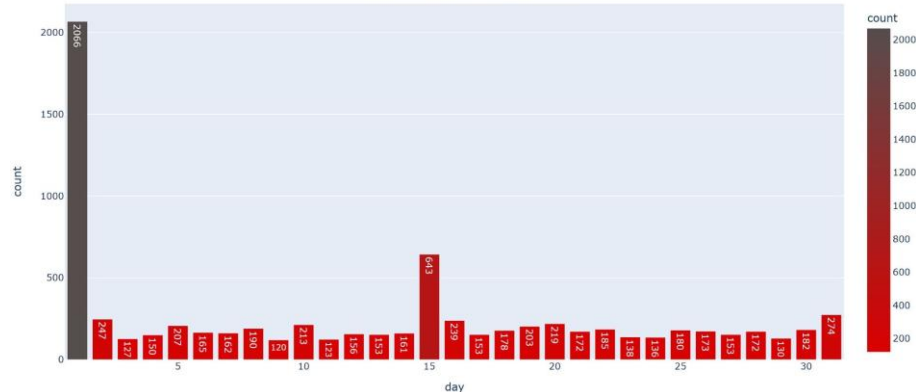
## Month added and Day added:

- Most of the content is uploaded either by year ending or beginning.
  - October, November, December, and January are months in which many shows and movies get uploaded to the platform.
  - It might be due to the winter, as in these months people may stay at home and watch shows and movies in their free time.
- 
- Most of the content is uploaded at the beginning, middle, or the end of a month.
  - Which makes 1st, 15th or 31st of a month more prominent in getting new tv shows and movies.

Month wise addition of movies and shows to the platform

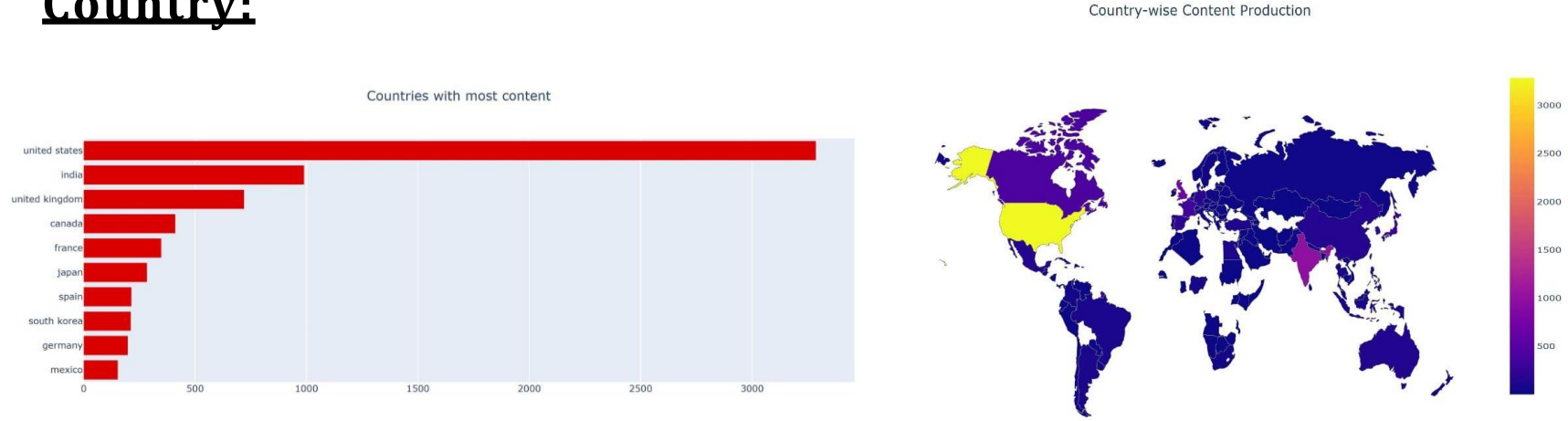


Which days are more prominent





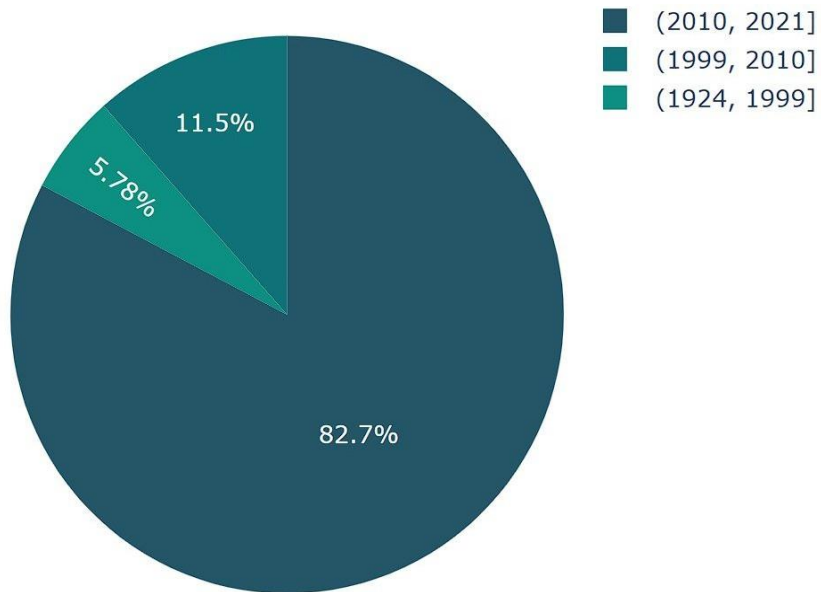
## Country:



- The majority of the content providers are in the above top-ten countries.
- Among which USA, India, and Uk create more than half of the tv shows and movies on the platform.

## Release\_year:

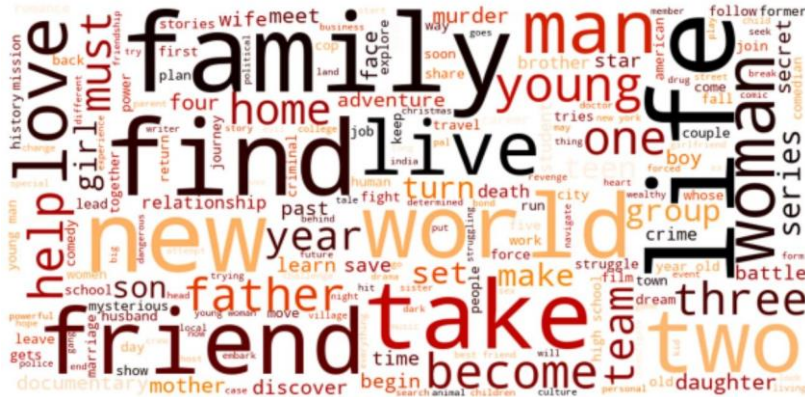
When was most content released.



- 82% of the content available was released between 2010 and 2021.
- 17.28% of the content available was released before 2010.

### Title and Description:

## Most used words in description



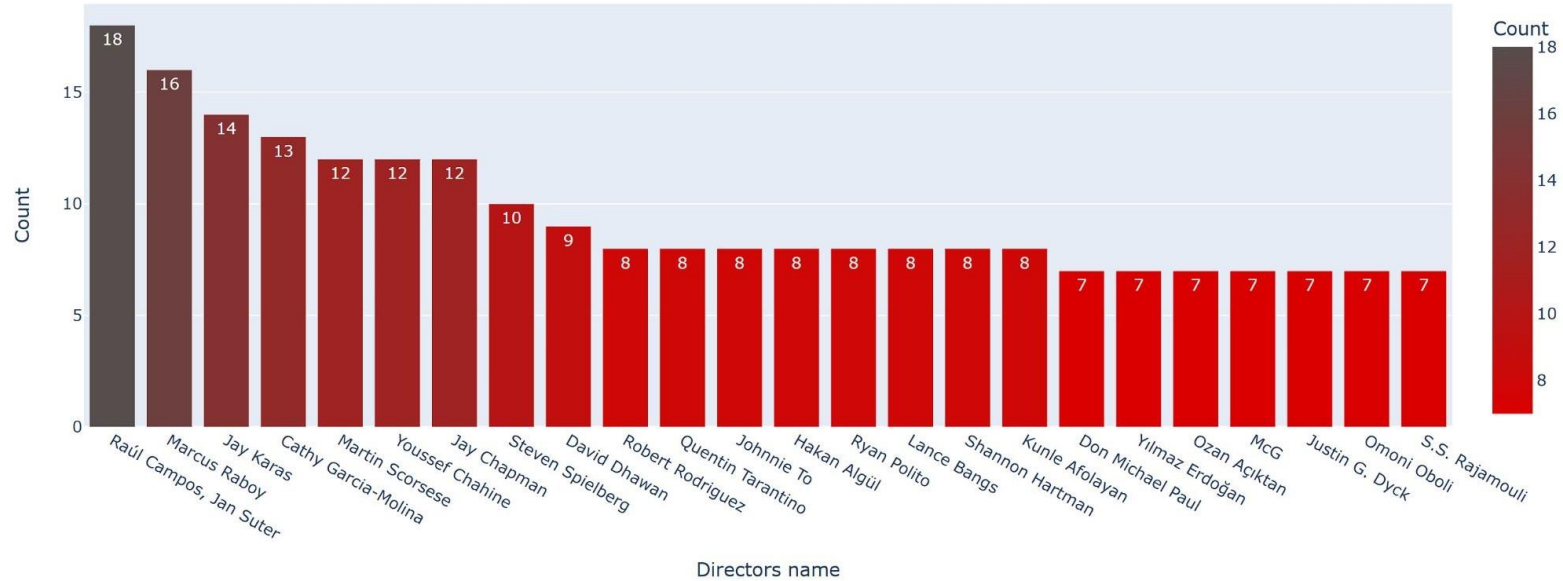
## Most used words in title



- Most repeated words in title include Christmas, Love, World, Man, and Story.
- We saw that most of the movies and tv shows got added during the winters, which tells why Christmas appeared many times in the titles.
- Most occurring words in the description of the tv shows and movies are Family, Friend, Love, Life, Woman, Man.

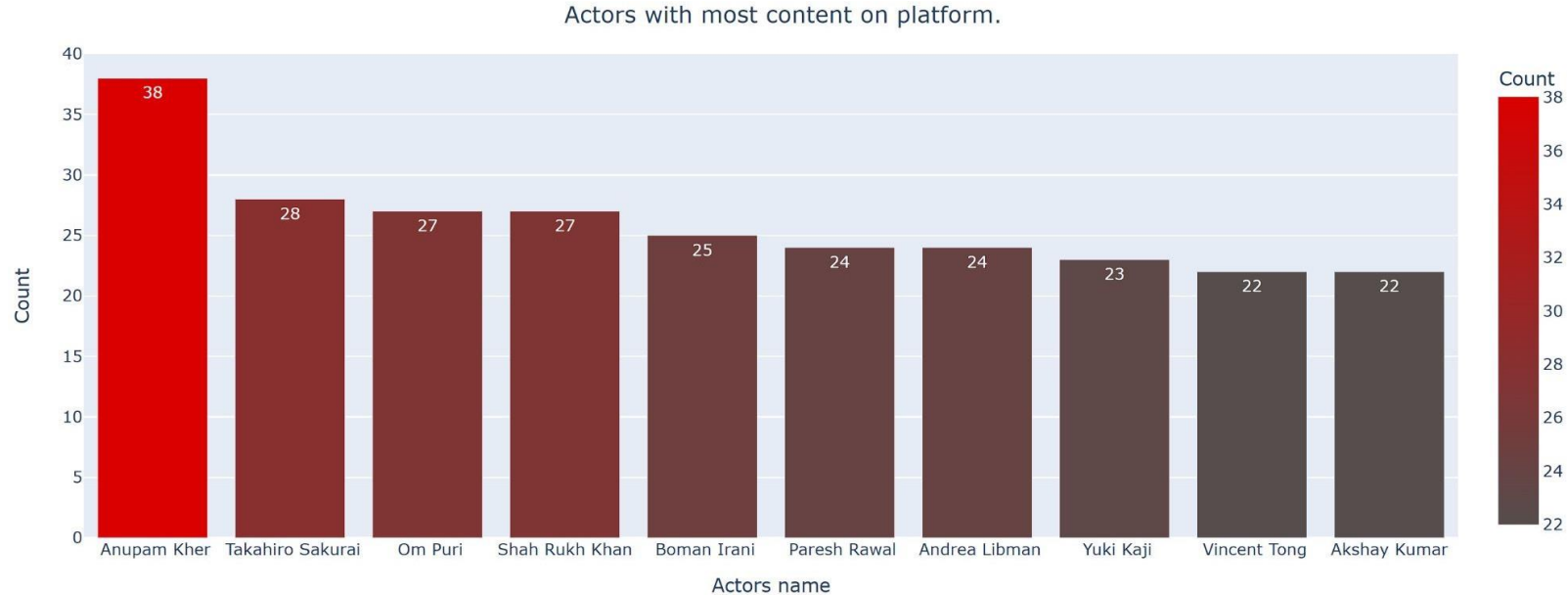
# Director:

Top 25 directors with highest number of Movies and Tv Shows.



- Raúl Campos, Jan Suter, Marcus Raboy, Jay Karas, Cathy Garcia-Molina, Jay Chapman are the top 5 directors which highest number of movies and tv shows.

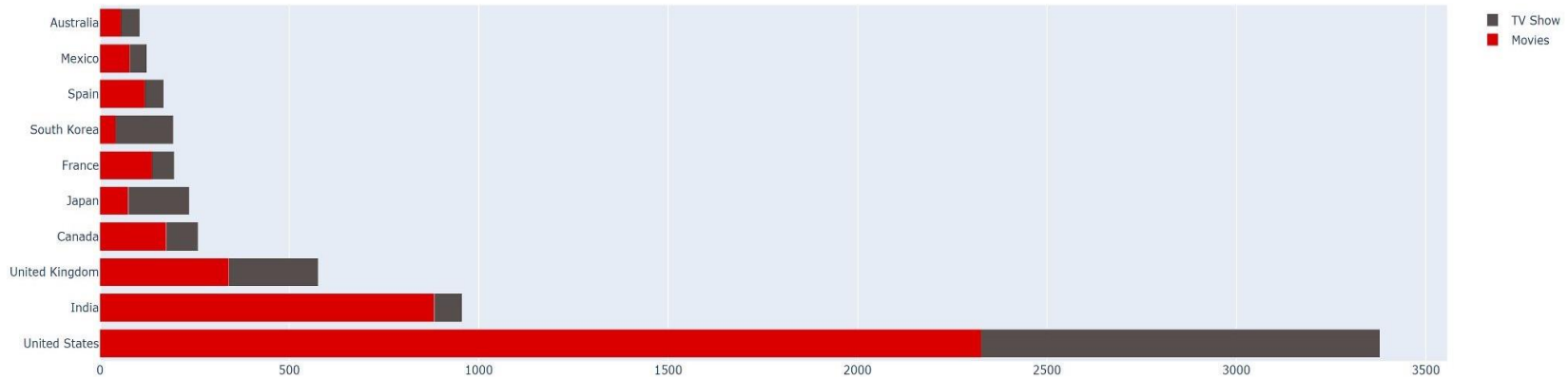
## Cast:



- Six of the actors in the top ten list with most numbers tv shows and movies are from India.
- With Anupam Kher at the top with 38 tv shows and movies in total.

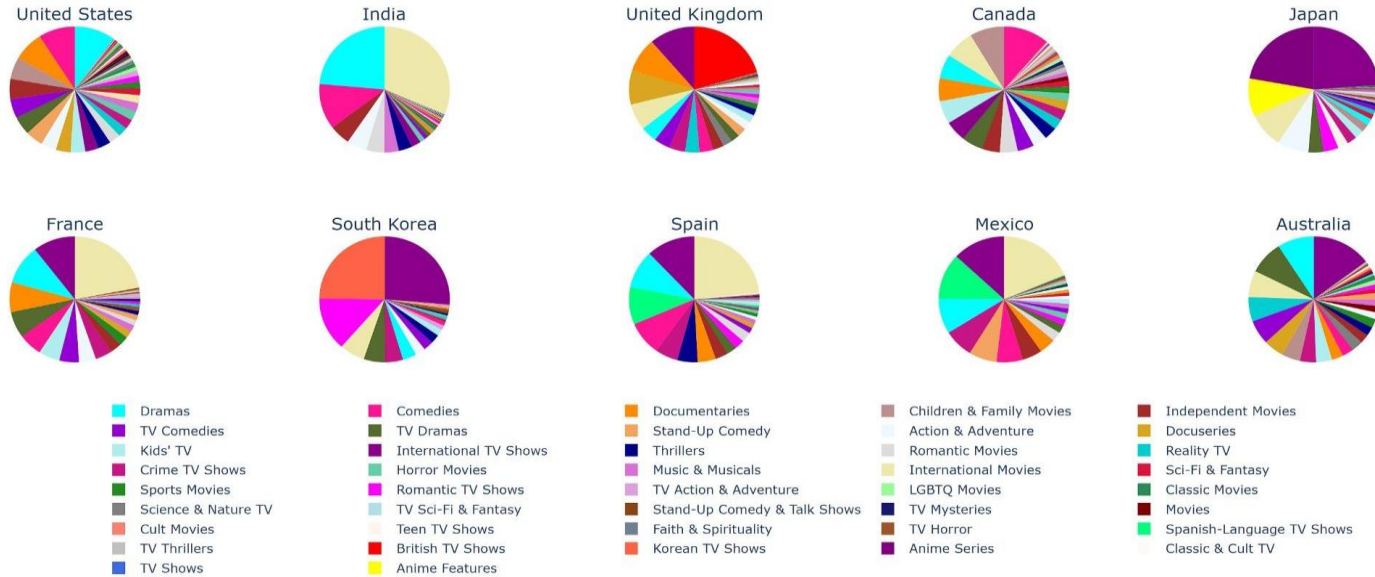
## Content vs Country:

Top ten countries and the content they provide.



- The United States is a leading producer of both types of content; this makes sense since Netflix is a US company.
- The influence of Bollywood in India explains the type of content available, and perhaps the main focus of this industry is Movies and not TV Shows.
- On the other hand, TV Shows are more frequent in South Korea, which explains the KDrama culture nowadays.

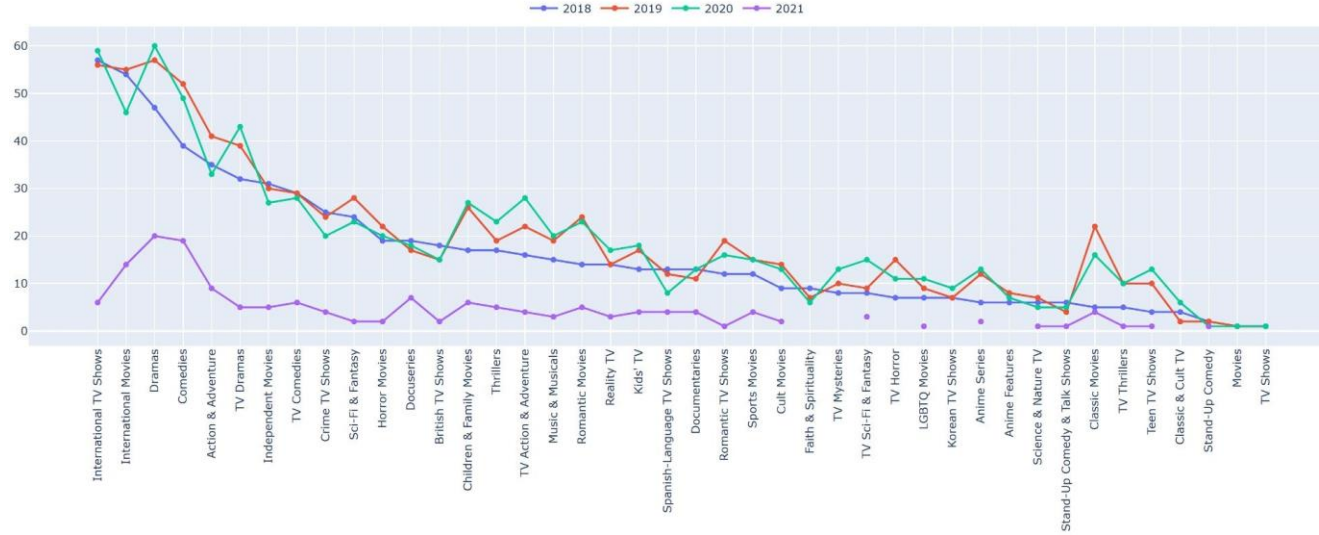
## Top ten countries and the content they provide.



- Drama, International Movies, and Comedies seem popular choices in most countries.
- British and International Tv Shows dominate in the United Kingdom.
- Regional specialties such as Anime in Japan and Korean Tv shows in South Korea are more prominent in these countries; This makes sense as anime has always been popular in Japan, and the rising k-pop culture explains the increase in Korean Tv Shows.
- It's also observed that in the countries where the regional language is not English, International Tv Shows and Movies are more in demand.

## Year added Vs Type:

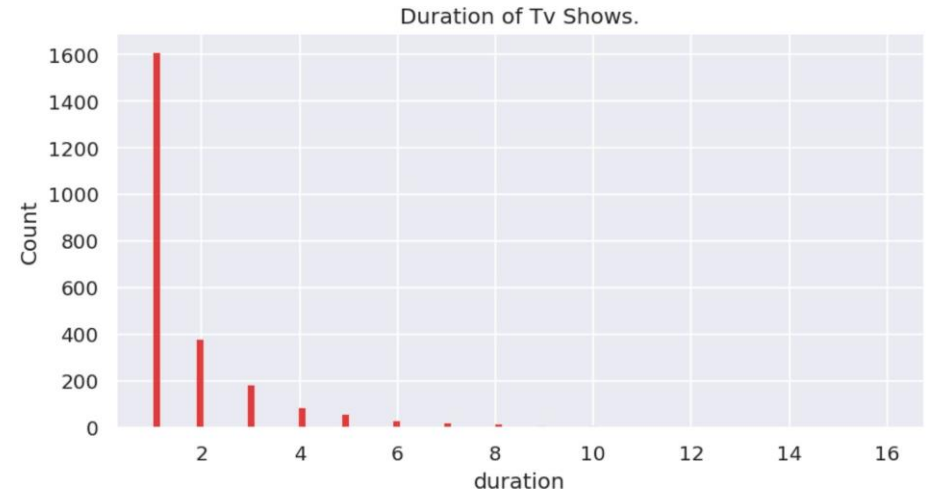
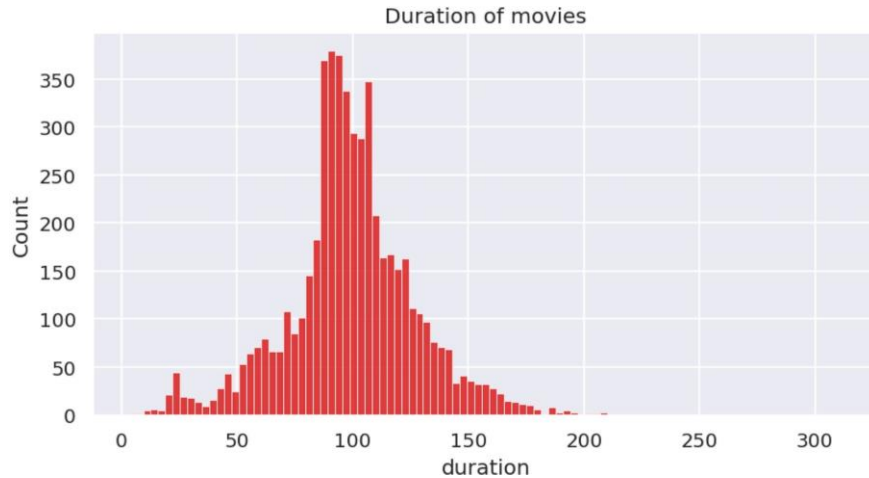
Most added Genres in recent years.



- From the EDA we did above, we saw that there are more Movies than Tv Shows on Netflix, which might be enough to assume that Netflix focuses more on Movies than Tv Shows. But the data proves this assumption wrong.
- The above line plot shows that Netflix has been adding many International Tv Shows in the recent years compared to Movies.
- From this observation, we can say that Netflix might be shifting slowly towards Tv Shows.

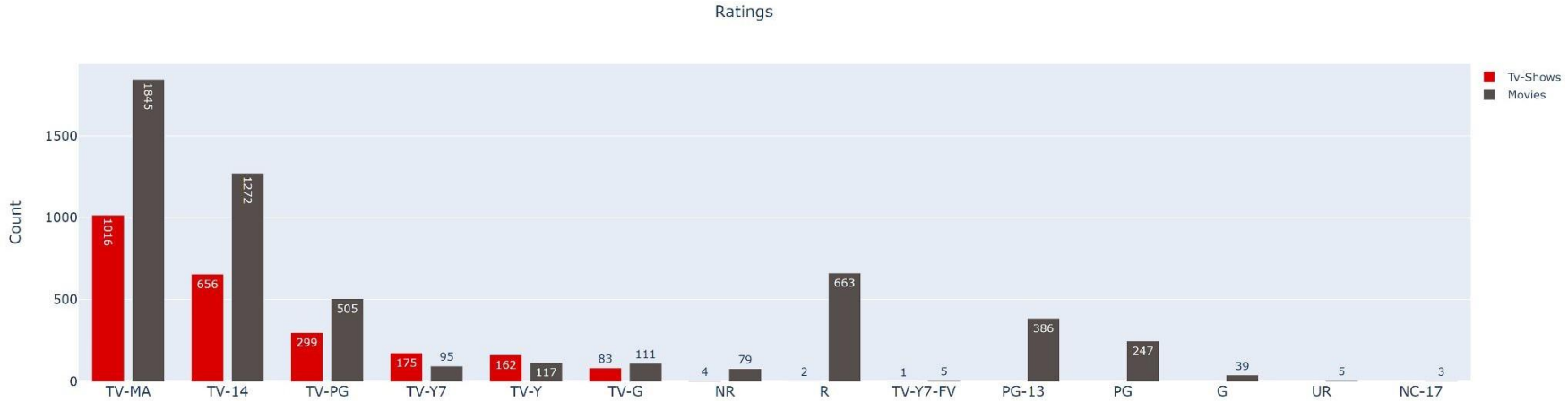


## Duration:



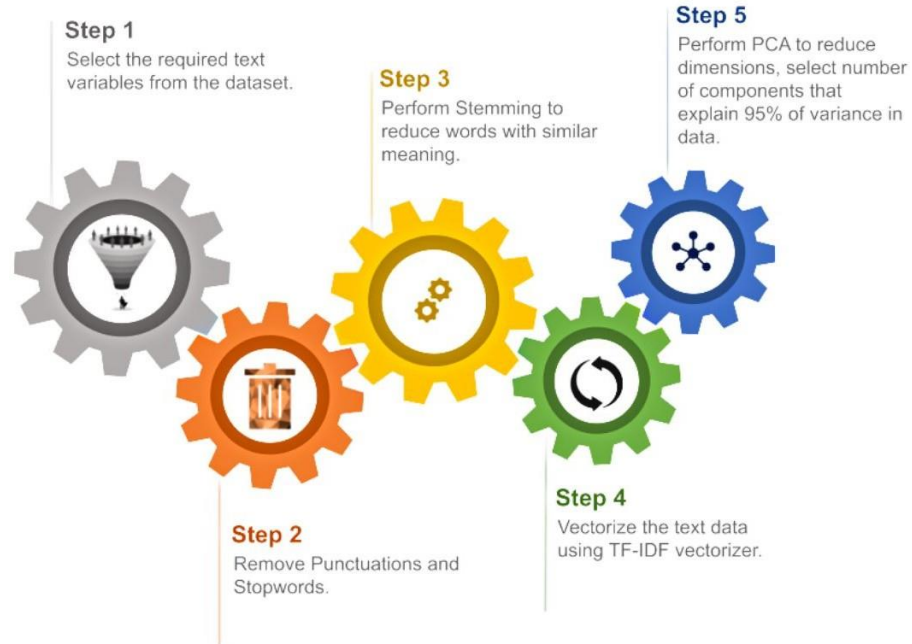
- Most of the Tv Shows last for 1 or 2 seasons, it is rare for a show to have more than 5 seasons.
- Most of the movies last for 90 to 120 minutes.

## Ratings:



- TV-MA tops the charts, indicating that mature content is more popular on Netflix.
- This popularity is followed by TV-14 and TV-PG, which are Shows focused on Teens and Older kids.
- Very few titles with a rating NC-17 exist. It can be understood since this type of content is purely for the audience above 17.

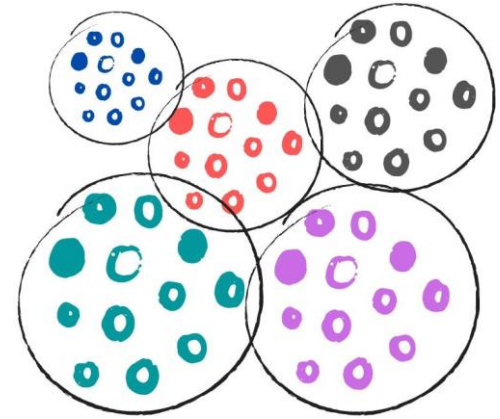
# Data Preprocessing.



# Creating Clusters:

## What is clustering?

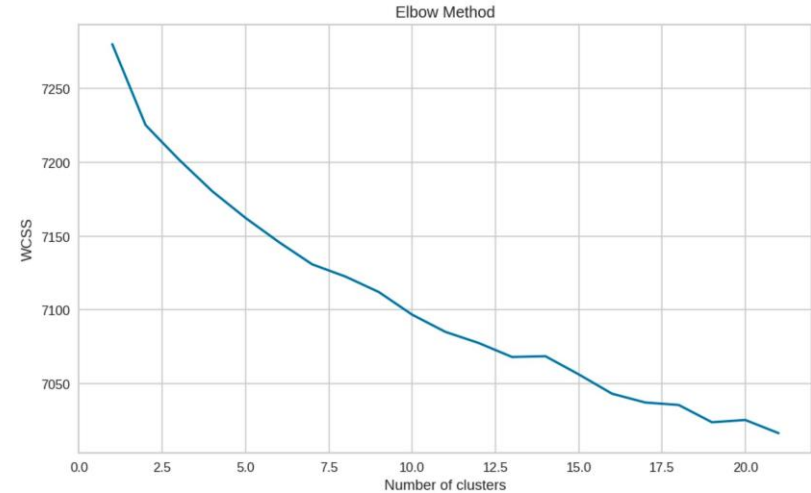
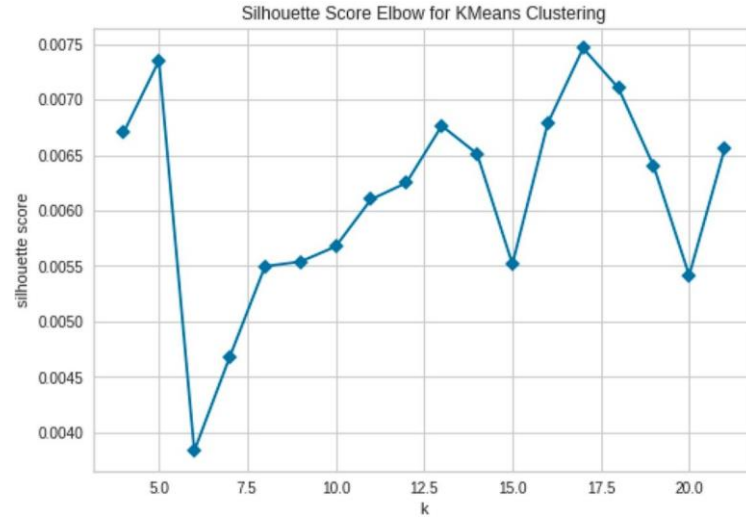
**Clustering** is the task of dividing the data points into a number of groups such that data points in the same groups are more similar to other data points in the same group and dissimilar to the data points in other groups. It is basically a collection of objects on the basis of similarity and dissimilarity between them.



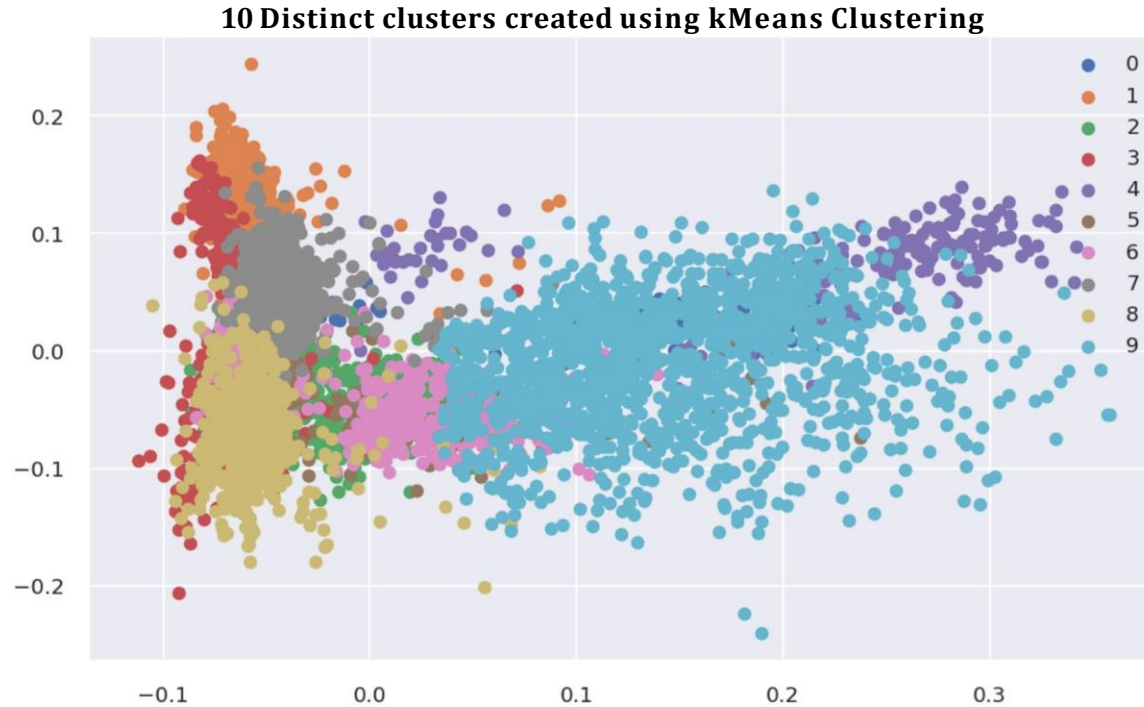
## How to cluster similar data?

To create clusters we will use the K-Means Clustering; which is an iterative process in which the dataset is grouped into k number of predefined non-overlapping clusters or subgroups, making the inner points of the cluster as similar as possible while trying to keep the clusters at distinct space it allocates the data points to a cluster so that the sum of the squared distance between the clusters centroid and the data point is at a minimum.

## Determining optimal value for k:



- Using the Silhouette Score and Elbow Method we select the optimal number of clusters to be 10.



- The numbers 0 to 9 represent 10-distinct clusters formed by K-means clustering.
- Each cluster contains data points similar to those in the same groups but varies from other groups.

## Data represented by each cluster:

**Cluster 0:** Documentaries.

**Cluster 1:** Family and Children Movies.

**Cluster 2:** Musical Movies and Documentaries.

**Cluster 3:** Stand Up Comedy and Comedy Shows.

**Cluster 4:** Korean and Romantic Tv Shows.

**Cluster 5:** Science, Nature, Reality, Crime Tv Shows and Docuseries.

**Cluster 6:** International Movies.

**Cluster 7:** Anime Series and Tv Shows.

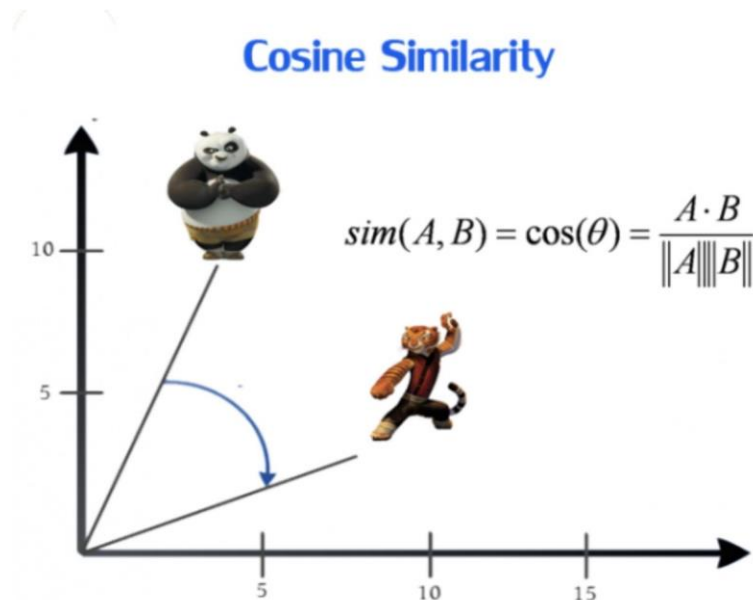
**Cluster 8:** International Tv Shows.

**Cluster 9:** Action, Adventure and Independent Movies.



# Getting Recommendations:

We obtained recommendations for Movies and Tv- Shows using Cosine similarity.



```
# Lets try getting recommendations for Movies.
print(*recommendations('Bad Boys'), sep='\n')
```

```
Bad Boys II
GoldenEye
Tortilla Soup
Martin Lawrence Live: Runteldat
War on Everyone
Madam Secretary
Slow West
Operation Odessa
Tremors 5: Bloodline
Act of Valor
```

```
# Lets try getting recommendations for Tv-Shows.
print(*recommendations('13 Reasons Why'), sep='\n')
```

```
13 Reasons Why: Beyond the Reasons
The Staircase
Unsolved Mysteries
Mind Game
The Mist
Twice Upon A Time
We Are the Wave
Re:Mind
Super Dark Times
Cam
```



## Conclusions:

- It was interesting to find that majority of the content available on Netflix is Movies.
- But in the recent years it has been focusing more on Tv-Shows.
- Most of these contents are released either in the year ending or the beginning.
- United States and India are among the top 5 countries that produce all of the available content on the platform.
- Also 6 of the actors among the top ten actors with maximum content are from India.
- TV-MA tops the charts, indicating that mature content is more popular on Netflix.
- $k=10$  was found to be an optimal value for clusters using which we grouped our data into 10 distinct clusters.
- Using the given data a simple recommender system was created using cosine\_similarity and recommendations for Movies and Tv Shows were obtained..

## Future Scope:

- Integrating this dataset with other external datasets such as IMDB ratings, rotten tomatoes can also provide many interesting findings.
- More time could be given into building a better recommender system, which later can be deployed on web for usage.



***Thank you.***