Streaming Services Analysis

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Introduction:

Streaming platforms are on the rise and they have taken over the way the world views TV and Movies. Nowadays, more people are paying for streaming services versus cable TV- so it is safe to say that the future of TV is leaning towards streaming services. The various streaming platforms are available to consumers through apps, smart TVs, game consoles, streaming media players, and Amazon's Fire TV. Consumers have shown that they love things that are easily accessible and not expensive. Streaming platforms have made it so that the consumers get exactly that! Netflix is the most popular online streaming websites around with over 183 million paid subscribers across the globe.

Since Netflix's creation in 1997, it is safe to say that the platform has changed greatly-what started off as a DVD rental service has transformed into a streaming platform used by millions. This growth led us to wonder about how exactly Netflix has evolved since its creation and how other streaming services compare. Hulu, Disney +, and Amazon Prime have all become Netflix's main competition. Amazon Prime has about 150 million subscribers and is the fastest growing platform due to the amount of content they offer. Hulu has focused more on the shows currently on air and had only hit 30.4 million subscribers by the end of 2019. Lastly, with Disney + only launching one year ago- they already have 60 million subscribers and counting. With these diverse streaming platforms available to us at our fingertips, we want to know what people want more of: TV shows or Movies.

Extraction:

We used 2 datasets (both CSV files) from the public platform Kaggle.com. Each dataset was based on IMDB and Rotten Tomatoes ratings for TV shows and Movies on the top 4 streaming platforms: Hulu, Netflix, Amazon Prime, and Disney+. We used Pandas functions in Jupyter Notebook to load both CSV files.

Figure 1: In [1]: import pandas as pd
from sqlalchemy import create_engine In [2]: movie_file = (r"C:\UPenn\ETT_Project\Netflix_Data_Analysis\movies.csv")
 movie_df = pd.read_csv(movie_file)
 movie_df.head() Out[2]: Rotten Prime Type Directors ID Title Genres Age IMDb Netflix Hulu Disney+ Country Language Tomato Video United Christopher Action, Adventure, Sci-2010 13+ 8.8 87% 0 0 English, Japanese, Fre Inception States.United Nolan Fi,Thriller Kingdom Lana The 1 2 1999 18+ 8.7 87% 0 0 Wachowski,Lilly Action,Sci-Fi United States English Wachowski Anthony 8.5 0 2 3 Infinity 2018 13+ 84% 0 Russo,Joe Action, Adventure, Sci-Fi United States English Russo Back to Robert Adventure,Comedy,Sci-1985 7+ 8.5 96% 0 United States English the Zemeckis Future The Good, the Italy, Spain, West 1966 18+ 8.8 97% 0 4 5 0 Sergio Leone Western Italian Bad and the Ugly

Figure 2:

```
In [3]: tv_file = (r"C:\UPenn\ETL_Project\Netflix_Data_Analysis\tv_shows.csv")
    tv_df= pd.read_csv(tv_file)
    tv_df.head()
```

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	ID_Number	Title	Year	Age	IMDb	Rotten Tomatoes	Netflix	Hulu	Prime Video	Disney+
0	0	Breaking Bad	2008	18+	9.5	96%	1	0	0	0
1	1	Stranger Things	2016	16+	8.8	93%	1	0	0	0
2	2	Money Heist	2017	18+	8.4	91%	1	0	0	0
3	3	Sherlock	2010	16+	9.1	78%	1	0	0	0
4	4	Better Call Saul	2015	18+	8.7	97%	1	0	0	0

Transformation:

Our first steps in cleaning up the datasets involved figuring out which variables were not relevant. To get the average rating on each streaming platform, we dropped the columns Rotten Tomatoes, Type, Directors, Genres, Country, Language, Runtime, Age, ID and Year on both of the datasets. After cleaning the datasets and making it workable, we started merging.

Figure 3:

In [4]:	#Transform Movie DataFrame movie_cols= ["Title", "IMDb", "Netflix", "Hulu", "Prime Video", "Disney+"] movie_transformed= movie_df[movie_cols].copy() #Rename column headers movie_transformed = movie_transformed.rename(columns={"Title": "movietitle_id",						x_id",
Out[4]:	movie_transformed.head() 4]: movietitle id movieimdb id movienetflix id moviehulu id movieprime id moviedisney id						
	\vdash	movietitle_id	movieimab_ia	movienetnix_iu	movienuiu_iu	movieprime_iu	movieuisney_iu
	0	Inception	8.8	1	0	0	0
	1	The Matrix	8.7	1	0	0	0
	2	Avengers: Infinity War	8.5	1	0	0	0
3 Back to the Future 8.5 1 0 0					0	0	
	4 The Good, the Bad and the Ugly 8.8 1 0 1 0						

Figure 4: In [5]: #Transform TV DataFrame tv_cols= ["Title", "IMDb", "Netflix", "Hulu", "Prime Video", "Disney+"]
tv_transformed= tv_df[tv_cols].copy() #Rename column headers tv_transformed = tv_transformed.rename(columns={"Title": "tvtitle_id", "IMDb" : "tvimdb_id",
"Netflix": "tvnetflix_id", "Hulu": "tvhulu id", "Prime Video": "tvprime_id", "Disney+": "tvdisney id"}) tv_transformed.head() Out[5]: tvimdb_id tvnetflix_id tvhulu_id tvprime_id tvdisney_id tvtitle_id 0 Breaking Bad 9.5 0 0 0 Stranger Things 8.8 0 0 0 1 0 0 0 Money Heist 3 1 0 0 0 Sherlock 9.1 0 0 0 Better Call Saul 8.7

Loading of data to database:

Once data was transformed, we created a database connection to PostgreSQL. In Postgres we did our analysis by averaging out all the ratings that each television show and movie received from IMBd. We then were able to see based off of the average ratings which performs better on which streaming service and overall which form of media is preferred.

Figure 5:

```
21 lines (19 sloc) 818 Bytes
     select * from movie
  2 select movieimdb_id, movienetflix_id, from movie
  4 select Avg(movieimdb_id) from movie where movienetflix_id = 1;
  5 --Netflix Movie IMDb Rating: 6.25--
  6 select Avg(movieimdb_id) from movie where moviehulu_id = 1;
     --Hulu Movie IMDb Rating: 6.14--
  8 select Avg(movieimdb_id) from movie where movieprime_id = 1;
     -- Prime Movie IMDb Rating: 5.77--
     select Avg(movieimdb_id) from movie where moviedisney_id = 1;
     --Disney+ Movie IMDb Rating: 6.44--
 12 select * from tv;
 13
 14    select Avg(tvimdb_id) from tv where tvnetflix_id = 1;
 15 --Netflix Show IMDb Rating: 7.16--
 select Avg(tvimdb_id) from tv where tvhulu_id = 1;
 17 ——Hulu Show IMDb Rating: 7.06—
 18  select Avg(tvimdb_id) from tv where tvprime_id = 1;
 19 -- Prime Show IMDb Rating: 7.18--
 20 select Avg(tvimdb_id) from tv where tvdisney_id = 1;
     --Disney+ Show IMDb Rating: 6.92--
```

Conclusion:

According to Time Magazine, streaming platforms do not release complete viewership data to the public. For example, Netflix only releases statistics on their top ten movies and TV shows. Our way around that was to examine the ratings that viewers gave to the movies and shows available on these various platforms. We chose to compare the ratings that are listed on IMBD.com to see what was most liked on these streaming platforms: TV shows or movies.

We used these datasets so we could identify the average rating IMBD users gave movies and TV shows. The final output helped us to recognize which streaming site had higher ratings for both movies and TV shows on each platform. Based on the output, we learned that on average, TV shows had higher ratings compared to movies. This leads us to believe that the consumer has a preference for TV shows, or that viewers tend to 'like' (give higher ratings) to TV shows moreso than movies. With these findings, we also predict that streaming platforms are in fact focusing more on providing TV show content versus movies.

Data Sources:

https://www.kaggle.com/ruchi798/tv-shows-on-netflix-prime-video-hulu-and-disney?select=tv_shows.csv

https://www.kaggle.com/ruchi798/movies-on-netflix-prime-video-hulu-and-disney?select=MoviesOnStreamingPlatforms_updated.csv

References:

https://time.com/5697802/most-popular-shows-movies-netflix/

 $\underline{https://www.investopedia.com/articles/markets/051215/who-are-netflixs-main-competitors-nflx.asp}$