CS 210 – Introduction to Data Science Movie Data Analysis Project

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Motivation

The motivation behind this project is to gain a deeper understanding of personal movie-watching preferences, explore patterns in the data, and utilize web scraping techniques to enrich the dataset with additional features. By exporting information from the Letterboxd platform, the goal is to uncover insights into movie genres, themes, top actors, and directors, providing a comprehensive overview of movie-watching habits. This project intended to developed more to built a user-friendly search engine matching with the personal taste of watching movies.

Data Source

The movie dataset used in this project is sourced from a personal collection and is stored in a CSV file ('watched.csv'). The data includes information about watched movies, such as the title, date of viewing, and a link to the corresponding Letterboxd page. This link is used to enhance the dataset, meaning that language, top 5 actor, top 5 genres, top 5 themes, director etc. of that movie was not available in downloaded watched.csv document and added by web scraping. We can say that it was a home-made API thanks to the Letterboxd for not guaranteed use of its own API:)

Data Analysis

Techniques Used

- Web scraping with BeautifulSoup for extracting information from Letterboxd.
- Multi-threading using ThreadPoolExecutor to enhance data extraction efficiency.
- Descriptive statistics, bar plots, and histograms for exploratory data analysis.

Stages of Analysis

- 1. Data Extraction: Utilized web scraping techniques to extract information from Letterboxd, including top actors, directors, genres, themes, language, and statistics for each movie.
- 2. Data Cleaning: Handled missing values and applied specific cleaning steps, such as filling missing language values with the mode.

- 3. Exploratory Data Analysis (EDA): Conducted analysis on the enriched dataset to explore distribution patterns of genres, themes, top actors, and directors. Examined the time difference between movie release years and watch dates.
- 4. Visualization: Generated plots to visually represent key findings.

Findings

What I Learned About Myself

- Movie Preferences: Identified top genres, themes, and favorite actors/directors, providing insights into personal movie-watching preferences. Note that my favorite director is not Nolan, it is Tarkovsky who I watched all his movies, but as he has less movies than Nolan, Nolan seems to be the winner:(
- Temporal Patterns: Explored trends in the time difference between movie release years and watch dates, understanding how much I follow contemporary film releases. This was different than my assumptions of myself, so it was good to see.

Limitations and Future Work

Limitations

- Data Availability: The analysis heavily relies on the availability and accuracy of data from Letterboxd. Incomplete or inaccurate information on the platform may affect the results.
- Biases: Movies released before my birth year are excluded to avoid biases, but this may limit the analysis.

Future Work

- Improved Data Collection: Explore options to enhance data collection, such as integrating with more movie databases or using APIs for richer information.
- Advanced Analysis: Implement advanced analysis techniques, including sentiment analysis on reviews, collaborative filtering for movie recommendations, or applying machine learning for predictive modeling.
- Make the user entered searched movie, in a more proper way to search. There may be another key components that is related with my movie habits and we do not consider. So, it would be good to think on them and explore more interesting relations. Also, this search might be useful for me to pick the upcoming movie that I would watch from the movie pool. Considering that I am really an undecisive person in selection, it would be really helpful for me:)

Future Plans

I plan to continually update and refine the project, incorporating additional features and insights. Future iterations may include the integration of more data sources, implementing user reviews sentiment analysis, and enhancing the visualizations for a more interactive experience.

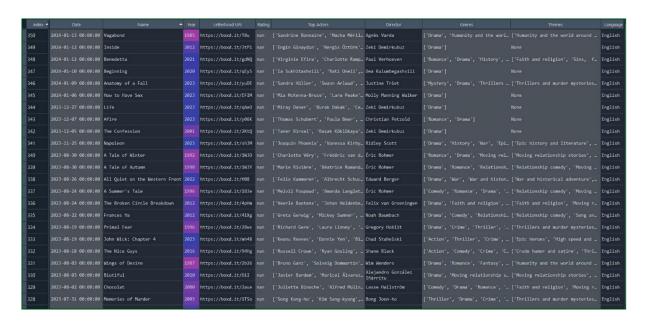
Feel free to contribute, provide feedback, or suggest improvements to make this project more comprehensive and insightful.

Let's try to understand what my functions does.

Enhance_dataset.py

This Python code focuses on extracting and enriching movie data from the Letterboxd platform. It begins by loading a movie dataset into a Pandas DataFrame, converting date columns, and then defines functions for web scraping various details from Letterboxd, such as top actors, director, language, genres, themes, and ratings. The use of a ThreadPoolExecutor enables parallelized extraction for efficiency. After dynamically adding columns with extracted information to the DataFrame, the code performs data cleaning by filling missing language values with the mode. Finally, the processed DataFrame is printed to the console, and the results are exported to an Excel file named 'extracted1.xlsx'. The code showcases an effective approach to automate the extraction and analysis of movie-related information from Letterboxd.

You may see some parts of the output created by this code in here.



Note that you can access xlsx files in the folder as well.

List_to_be_compared.py

Working mechanism is quite similar to the **Enhance_dataset.py.** Please see the explanation above. This function creates a searching movie pool that I may ask for correlation with my watched data, indicating my movie taste. You can see some parts of the output below, as well as the .xlsx file.

Date	Name	Tags	Letterboxd URI	escripti	Top Actors	Director	Genres	Themes	Language
1	Poor Things		https://boxd.it/tNWU		['Emma Stone', 'Mark Ruffalo', 'Will	Yorgos Lanthimos	['Romance', 'Science Fic	['Humanity and th	English
2	World War III	2022	https://boxd.it/BZBU		['Mohsen Tanabande', 'Mahsa Hejazi',	Houman Seyyedi	['Drama']		English
3	The Boy and the Heron		https://boxd.it/ipeM		['Soma Santoki', 'Masaki Suda', 'Ko	Hayao Miyazaki	['Animation', 'Adventure		English
4	About Dry Grasses		https://boxd.it/oNbO		['Deniz Celiloğlu', 'Merve Dizdar',	Nuri Bilge Ceylan	['Drama']	None	English
5	All of Us Strangers		https://boxd.it/Bz3C	nan	['Andrew Scott', 'Paul Mescal', 'Jam	Andrew Haigh	['Romance', 'Drama', 'Fantasy']		English
6	The Promised Land		https://boxd.it/B1hA		['Mads Mikkelsen', 'Amanda Collin',	Nikolaj Arcel	['History', 'Drama']	None	English
7	Saltburn		https://boxd.it/z4eg		['Barry Keoghan', 'Jacob Elordi', 'R	Emerald Fennell	['Comedy', 'Drama', 'Thr…		English
8	Past Lives		https://boxd.it/oNB8	nan	['Greta Lee', 'Teo Yoo', 'John Magar	Celine Song	['Drama', 'Romance', 'Mo		English
9	Heroic		https://boxd.it/AafQ	nan	['Santiago Sandoval', 'Fernando Cuau	David Zonana	['Drama', 'Thriller']	None	English
10	The Teachers' Lounge		https://boxd.it/BINY			İlker Çatak	['Drama']		English
11	Infinity Pool		https://boxd.it/oPSK		['Alexander Skarsgård', 'Mia Goth',	Brandon Cronenberg			English
12	Fools	2022	https://boxd.it/lenu		['Dorota Kolak', 'Łukasz Simlat', 'T	Tomasz Wasilewski	['Drama']	None	English
13	Green Border		https://boxd.it/GqGq	nan	['Jalal Altawil', 'Maja Ostaszewska'	Agnieszka Holland			English
14	Fingernails		https://boxd.it/tJ82		['Jessie Buckley', 'Riz Ahmed', 'Jer	Christos Nikou	['Science Fiction', 'Rom	['Relationship co	English
15	Monster		https://boxd.it/DJEM		['Sakura Ando', 'Eita Nagayama', 'So	Hirokazu Kore-eda	['Thriller', 'Drama', 'M		English
16	Critical Zone		https://boxd.it/HnYy		['Amir Pousti', 'Shirin Abedinirad',	Ali Ahmadzadeh			English
17	The Holdovers		https://boxd.it/vHza		['Paul Giamatti', 'Dominic Sessa', "	Alexander Payne	['Comedy', 'Drama', 'Und	['Underdogs and c	English
18	May December		https://boxd.it/vEE2		['Natalie Portman', 'Julianne Moore'	Todd Haynes	['Drama', 'Comedy', 'Mov	['Moving relation	English
19	Fair Play		https://boxd.it/yirc		['Phoebe Dynevor', 'Alden Ehrenreich	Chloe Domont	['Thriller', 'Drama', 'I		English
20	The Delinquents		https://boxd.it/qLyA		['Daniel Elías', 'Esteban Bigliardi'	Rodrigo Moreno	['Comedy', 'Drama']	None	English
21	Inside the Yellow Cocoon Shell		https://boxd.it/Gama	nan	['Le Phong Vu', 'Nguyen Thinh', 'Ngu	Pham Thien An	['Drama']		English
22	Dream Scenario		https://boxd.it/v2h0		['Nicolas Cage', 'Julianne Nicholson	Kristoffer Borgli	['Comedy', 'Fantasy', 'R	['Relationship co	English
23	Subtraction	2022	https://boxd.it/C69o		['Navid Mohammadzadeh', 'Taraneh Ali	Mani Haghighi	['Drama', 'Thriller']		English

Read_extracted.py

This Python code conducts an exploratory data analysis (EDA) on a movie dataset extracted from Letterboxd. It utilizes Pandas, Matplotlib, and Seaborn for data manipulation and visualization. The EDA starts by displaying the first few rows, general information, descriptive statistics, data types, and missing values of the DataFrame. Subsequently, it delves into analyzing the top actors and directors by visualizing their occurrence counts in bar plots. The distribution of movie genres and themes is explored similarly, showcasing the most frequent genres and themes through bar plots. Finally, the analysis examines the time difference between the release year and the watched year for movies released after 2015, shedding light on potential biases in movie preferences based on release years. The code effectively combines data exploration and visualization techniques to gain insights into the user's movie-watching patterns and preferences. You may see the outputs on "EDA and Data Visualization Outputs" folder.

All of them may be interpreted easily but it is good to explain plot_time_differences a bit. It is created to show how much I am following the trends of the new releases of the year. The bar plot illustrates the average time difference between movie release years and the years they were watched, focusing on films released after 2015. Positive bars indicate the user tends to watch movies a certain number of years after their release. Small error bars suggest consistent behavior for movies released in a given year, while larger ones imply more variability. The analysis helps uncover trends and potential biases in the user's movie preferences based on release years. See figure 1 to clearly understand.

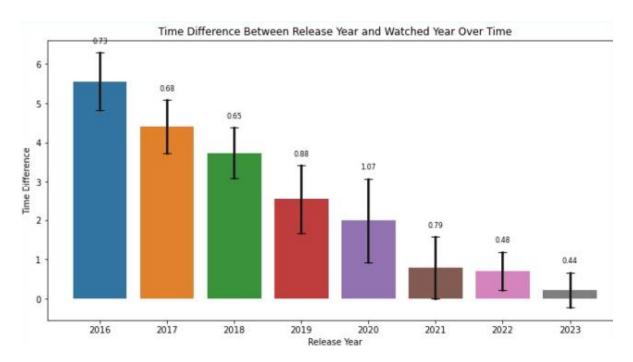


Figure 1. Time Difference Between Release Year and Watched Year Over Time

Please refer to the visuals which have names in the form of plot_top_NUMBER_VARIABLE.png in the outputs folder.

Explore_data.py

In this set of codes, I'm working on a content-based movie recommendation system. I start by preprocessing the dataset, converting string representations of lists, exploding columns with multiple values, and handling missing data. I then identify the top actors, directors, genres, and themes based on movie counts. To create a numerical representation, I manually assign scores to these top values and apply these mappings to encode the dataset. After that, for a user-entered movie, I calculate similarity scores by summing the encoded values for actors, director, genre, and theme. The dataset is then sorted based on these similarity scores, and I display the top similar movies. This process allows me to offer personalized movie recommendations based on the user's watched movies and preferences.

User may search for a movie from the pool and these codes calculates correlated movies from the pool that I have already watched and display similars. See sample search.



As I explained above, this program is open to develop more and more. Anyways, hope you enjoyed so far!