University of Asia Pacific

Team B1-G3

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Project Title

The title of our project is "Movie Recommendation Engine".

INTRODUCTION

This data set contains information about movies. Which is collected from The Movie Database (TMDb), including user ratings and revenue.

Dataset

In this project, we are using The **TMDb** Dataset. We have collected it from <u>Kaggle</u>.. Our dataset is in two parts of a csv file. They are -

- tmdb_5000_credits.csv
- tmdb_5000_movies.csv

There are almost 5000 movies in our dataset in which we will train our model.

Data Source Summary

We (Kaggle) have removed the original version of this dataset per a DMCA takedown request from IMDB. In order to minimize the impact, we're replacing it with a similar set of films and data fields from The Movie Database (TMDb) in accordance with their terms of use. The bad news is that kernels built on the old dataset will most likely no longer work.

The good news is that:

- We can port our existing kernels over with a bit of editing. This kernel offers functions and examples for doing so. We can also find a general introduction to the new format here.
- The new dataset contains full credits for both the cast and the crew, rather than just the first three actors.

- Actors and actresses are now listed in the order they appear in the credits. It's unclear what
 ordering the original dataset used; for the movies I spot checked it didn't line up with either
 the credits order or IMDB's stars order.
- The revenues appear to be more current. For example, IMDB's figures for Avatar seem to be from 2010 and understate the film's global revenues by over \$2 billion.
- Some of the movies that we weren't able to port over (a couple of hundred) were just bad entries. For example, this IMDB entry has basically no accurate information at all. It lists Star Wars Episode VII as a documentary.

Data Source Details

Several of the new columns contain json. We can save a bit of time by porting the load data functions [from this kernel]().

Even in simple fields like runtime may not be consistent across versions. For example, the previous dataset shows the duration for Avatar's extended cut while TMDB shows the time for the original version.

There's now a separate file containing the full credits for both the cast and crew.

All fields are filled out by users so don't expect them to agree on keywords, genres, ratings, or the like.

Our existing kernels will continue to render normally until they are re-run.

If you are curious about how this dataset was prepared, the code to access TMDb's API is posted <u>here</u>.

All columns:

Data columns (total 29 columns): # Column Non-Null Count Dtype 0 budget 4845 non-null int64 1 genres 4845 non-null object 2 homepage 1719 non-null object 3 id 4845 non-null int64 4 keywords 4845 non-null object 5 original_language 4845 non-null object 6 original_title 4845 non-null object 7 overview 4842 non-null object 8 popularity 4845 non-null float64 9 production_companies 4845 non-null object 10 production_countries 4845 non-null object 11 release_date 4844 non-null object	
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10 production_countries 4845 non-null object	
11 release date ASAA non null object	
11 Telease_uate 4844 Holl-Hull Object	
12 revenue 4845 non-null int64	
13 runtime 4843 non-null float64	
<pre>14 spoken_languages 4845 non-null object</pre>	
15 status 4845 non-null object	
16 tagline 4001 non-null object	
17 title 4845 non-null object	
18 vote_average 4845 non-null float64	
19 vote_count 4845 non-null int64	
20 movie_id_x 4845 non-null int64	
21 cast_x 4845 non-null object	
22 crew_x 4845 non-null object	
23 movie_id_y 4845 non-null int64	
24 cast_y 4845 non-null object	
25 crew_y 4845 non-null object	
26 movie_id 4845 non-null int64	
27 cast 4845 non-null object	
28 crew 4845 non-null object	

Selected columns:

```
Data columns (total 7 columns):
              Non-Null Count Dtype
     Column
     movie_id 4845 non-null
 0
                            int64
                             object
     title
              4845 non-null
 1
                            object
 2
     overview 4842 non-null
              4845 non-null
 3
     genres
                            object
              4845 non-null
                            object
 4
     keywords
                             object
              4845 non-null
 5
     cast
              4845 non-null
                             object
 6
     crew
```

Lost columns:

- actor1facebook_likes
- actor2facebook_likes
- actor3facebook_likes
- aspect_ratio
- casttotalfacebook likes
- color
- content_rating
- directorfacebooklikes
- facenumberinposter
- moviefacebooklikes
- movieimdblink
- numcriticfor_reviews
- numuserfor_reviews

Algorithm

We have used K-Nearest Neighbor(KNN) for our dataset. KNN makes inference about a movie, KNN will calculate the "distance" between the target movie and every other movie in its database, then it ranks its distances and returns the top K nearest neighbor movies as the most similar movie recommendations.

Conclusion

This dataset was generated from The Movie Database API. This product uses the TMDb API but is not endorsed or certified by TMDb. Their API also provides access to data on many additional movies, actors and actresses, crew members, and TV shows.