

Movie Recommendation System Using NLP & TMDB API

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01 **Dataset Source**

Dataset Source

- 1 Dataset Link
 - Dataset:\https://www.kaggle.com/datasets/t mdb/tmdb-movie-metadata
- 2 Files used
 - tmdb_5000_credits.csv
 - tmdb_5000_movies.csv





Data Preprocessing



Overview

 Initially found 3 missing overviews — manually created and updated them.



Selected columns

 genres, movie_id, keywords, overview, title, cast, crew



Data Extraction

- From cast, extracted top 3 actor names.
- From crew, selected only the director name.



Final Columns

- Combined all relevant text into a single column called tags.
- Final columns used: movie_id, title, tags



Preprocessing

 Applied spaCy model (en_core_web_sm) for preprocessing and lemmatization. 03

Feature Engineering



Feature Engineering

- 1 Tag Weighting
 - Attempted custom weighting for tags and overview, but it didn't improve results, so discarded.
- 2 Comparison
 - Compared CountVectorizer and TF-IDF:
 - CountVectorizer gave better similarity results in this case.

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Recommendation Logic

Recommendation Logic



Similarity Calculation

Used cosine similarity on vectorized data.



Movie Retrieval

- For each selected movie:
- Retrieved top 5 similar movies using similarity scores.

05 TMDB API Integration



Details Fetched

Used TMDB API to fetch details:

- Poster
- Title
- Genres
- Rating
- Overview
- Runtime
- Release Date

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Streamlit Web App

Streamlit Web App



Top Picks You Might Enjoy: ⇔

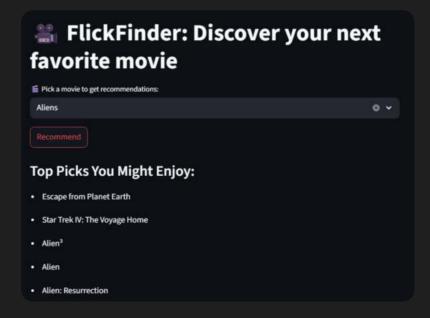
- · Escape from Planet Earth
- · Star Trek IV: The Voyage Home
- Alien³
- Alien
- · Alien: Resurrection

07 **Known Issue**



Known Issue

- 1 API Reliability
 - TMDB API sometimes fails.
- 2 Fallback Mechanism
 - Fallback: Shows only the recommended movie titles without extra details.



Code Snippets

Preprocessing and Lemmatization

API Inegration

```
# Fetch Movie Details from TMDB
def get_movie_details(title):
       search_url = f
"https://api.themoviedb.org/3/search/movie?api key={TMDB API KEY}
        res = requests.get(search_url).json()
        if res['results']:
           movie = res['results'][0]
           movie_id = movie['id']
           detail_url = f"https://api.themoviedb.org/3/movie/{movie_id
}?api_key={TMDB_API_KEY}*
           details = requests.get(detail_url).json()
           poster = f"https://image.tmdb.org/t/p/original{movie.get(
                'title': movie.get('original_title', 'N/A'),
                'poster': poster,
                'overview': details.get('overview', 'N/A'),
                'rating': details.get('vote_average', 'N/A'),
                'genres': ", ".join([g['name'] for g in details.get(
'genres', [])]),
                'release_date': details.get('release_date', 'N/A'),
                'runtime': details.get('runtime', 0)
    except Exception as e:
```

Vector Conversion and Recommendation

```
# Initialize CountVectorizer Vectorizer
cv=CountVectorizer(max features=5000)
# Applying it on 'tags' column
vectors_countvector = cv.fit_transform(new_df[
'tags']).toarray()
#calculating similarity score
similarity countvector=cosine similarity(
vectors countvector)
# Function to display recommended movies.
def recommed(movie):
    movi index=new df[new df['title']==movie
    distance=similarity countvector[movi index]
    movies list=sorted(list(enumerate(distance
)),reverse=True,key=lambda x:x[1])[1:6]
    for i in movies list:
        print(new df.iloc[i[0]].title)
recommed('Avatar')
# Aliens
# Independence Day
# Titan A.E.
# Small Soldiers
```

Thank You