

Movie Recommendation System

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Introduction



<u>Technical background of project:</u>

A Movie Recommendation System is a type of machine learning application designed to suggest movies to users based on their preferences or similarities between items (movies) or users.

Technical Concepts used:

- Cosine Similarity
- Vectorization of Movie Attributes
- Feature Extraction

Motivation:

With the abundance of movie content available today, users often struggle to decide what to watch. A recommendation system simplifies this process by filtering options based on user preferences.

Introduction



Problem Statement:

To develop an efficient and scalable movie recommendation system that provides users with personalized suggestions based on similarity metrics.

Area of application:

This system can be applied to online streaming platforms, e-commerce websites, and personal movie libraries.

Introduction



Dataset and input format:

	inde	ex	budget	genres	homepage	id	keywords	original_language	original_title	overview	popularity	•••	runtime	spoken_languages	status
	0	0	237000000	Action Adventure Fantasy Science Fiction	http://www.avatarmovie.com/	19995	culture clash future space war space colony so	en	Avatar	In the 22nd century, a paraplegic Marine is di	150.4 <u>37577</u>		162.0	[("iso_639_1": "en", "name": "English"), ["iso	Released
,	1	1	300000000	Adventure Fantasy Action	http://disney.go.com/disneypictures/pirates/	285	ocean drug abuse exotic island east india trad	en	Pirates of the Caribbean: At World's End	Captain Barbossa, long believed to be dead, ha	139.082615		169.0	[{"iso_639_1": "en", "name": "English"}]	Released
	2	2	245000000	Action Adventure Crime	http://www.sonypictures.com/mov/es/spectre/	206647	spy based on novel secret agent sequel mi6	en	Spectre	A cryptic message from Bond's past sends him o	107.376788		148.0	[{"iso_639_1": "fr", "name": "Fran\u00e7ais"},	Released

Objective



Main Objective:

To build a recommendation system that predicts movies a user might like based on cosine similarity of movie feature vectors.

Sub Objective:

- Preprocessing and feature extraction from movie datasets
- Representing movies as vectors based on attributes (e.g., genres, keywords)
- Implementing cosine similarity for recommendations

Methodology



Steps:

1. Data Preprocessing:

Cleaning the dataset, handling missing values, and encoding features (e.g., genres, directors) into vectors.

2. Vectorization:

Represent movies as vectors (e.g., [1, 0, 1, 0]).

3. Similarity Calculation:

Use cosine similarity to compute the angle between movie vectors, assigning higher similarity scores for similar movies.

4. Recommendation Engine:

For a given movie or user, sort and recommend the top N most similar movies.

Methodology

Steps:

<u>Input</u>

```
# getting the movie name from the user
movie_name = input(' Enter your favourite movie name : ')

Enter your favourite movie name : fight club
```

Similarity score

```
# getting a list of similar movies
similarity_score = list(enumerate(similarity[index_of_the_movie]))
print(similarity_score)

[(0, 0.0), (1, 0.0), (2, 0.0), (3, 0.03172085537943224), (4, 0.0),
]
[] len(similarity_score)

4803
```

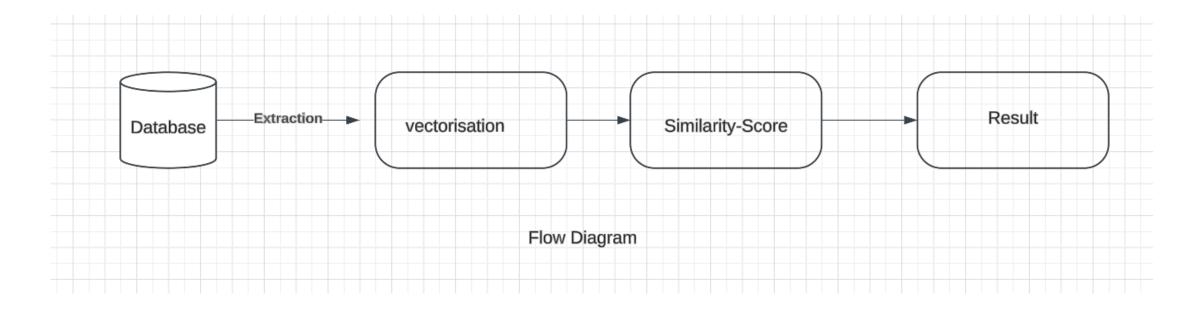


Result

Working Model



Technical Diagram



Working Model

Working Module



Steps to Describe:

- 1. Data Preprocessing:
 - \circ Clean and structure the dataset (e.g., handling missing values). \circ Convert genres, ratings, and other features into vectors.
- 2. Feature Engineering:
 - o Create a vector representation for each movie based on features (e.g., genres, average ratings).
- 3. Cosine Similarity Computation:
 - Compute cosine similarity between movies to find the closest matches.
- 4. Recommendation Algorithm:
 - Use similarity scores to recommend movies:

Working Model



Attained Deliverables

1.Preprocessed Dataset:

Cleaned, structured dataset with features such as genres, ratings, and movie descriptions transformed into a usable format for the system.

2. Vectorized Representations:

Numerical vectors for movies and user profiles created using feature extraction and encoding techniques.

3. Similarity Matrix:

A matrix containing pairwise cosine similarity scores between movies, representing how closely related they are.

4. Recommendation System:

A fully functional model that provides personalized movie recommendations based on user preferences.

Results



Tests Cases

1. Search for a movie

```
[49] # getting the movie name from the user
    movie_name = input(' Enter your favourite movie name : ')

Enter your favourite movie name : Spider-man
```

2. Sort the movie on basis of their similarity score using cosine similarity

```
[56] # sorting the movies based on their similarity score

sorted_similar_movies = sorted(similarity_score, key = lambda x:x[1], reverse = True)

print(sorted_similar_movies)

→ [(159, 1.0), (5, 0.3188331558421017), (30, 0.31791901982229703), (1559, 0.18281312771525624), (382, 0.16501718739122473), (3575, 0.16167041055132)
```

Results



Tests Cases

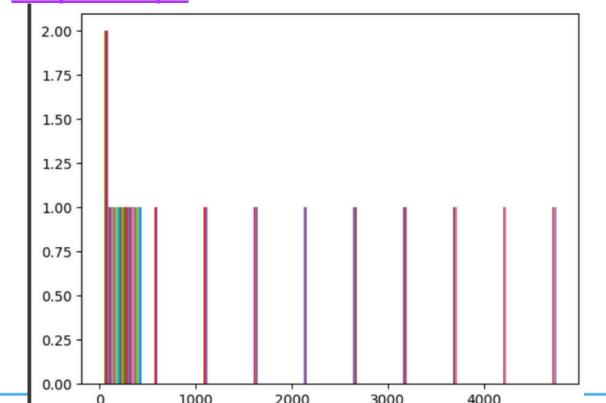
3. Result

Movies suggested for you : 1 . Spider-Man 2 . Spider-Man 3 3 . Spider-Man 2 4 . The Notebook 5 . Seabiscuit 6 . Clerks II 7 . The Ice Storm 8 . Oz: The Great and Powerful 9 . Horrible Bosses 10 . The Count of Monte Cristo

11 . In Good Company

12 Finding Nome

Output Graph



Conclusion



Justification of Objectives:

The system successfully meets the objectives by efficiently recommending movies based on user preferences and cosine similarity.

Future Scope:

1. Enhanced Recommendation

Techniques Hybrid Recommendation

Systems:

Combine collaborative filtering, content-based filtering, and knowledge graphs to improve accuracy and diversity in recommendations.

2. Multi-Platform

Deployment <u>Cross-Platform</u>

Integration:

Deploy the system on multiple platforms, including web, mobile apps, and streaming services like Netflix or YouTube.

References



https://www.kaggle.com/datasets/parasharmanas/movierecommendation-system



Thank You