TASK:12

MOVIE RECOMMENDATION ENGINE

CO4, CO5 S3

PROBLEM STATEMENT: In today's digital world, users are overwhelmed by the vast number of movies available on streaming platforms. Selecting a movie that aligns with user preferences has become a challenge. Hence, there is a need for an intelligent system that can recommend movies to users based on their interests, ratings, and behavior.

AIM: To design and develop a Movie Recommendation Engine using machine learning techniques that suggests movies to users based on their preferences and similarity to other users or movies.

OBJECTIVE:

- To collect and preprocess movie and rating data.
- To implement a recommendation model using Collaborative Filtering or Content-Based Filtering.
- To predict and recommend top-rated movies for a specific user.
- To evaluate the accuracy and performance of the recommendation model.

DESCRIPTION: A **Movie Recommendation System** suggests movies to users based on various techniques such as:

- Content-Based Filtering: Recommends movies similar to those a user liked, based on features like genre, director, or actors
- Collaborative Filtering: Uses the preferences of multiple users to recommend movies to a particular user based on shared interests.
- **Hybrid Systems:** Combine both approaches for better accuracy.

In this project, we use a **Collaborative Filtering approach** with the MovieLens dataset to generate movie recommendations. The system learns from user ratings to find patterns and suggest new movies that similar users have enjoyed.

ALGORITHM:

- 1. Import required libraries and dataset.
- 2. Preprocess the dataset (handle missing values, merge data if needed).
- 3. Create a user-item rating matrix.
- 4. Calculate similarity between users or movies using cosine similarity.
- 5. For a given user, find the most similar users or movies.
- 6. Predict ratings for unseen movies based on similarities.
- 7. Recommend top-N movies with the highest predicted ratings.

PROGRAM:

```
Movie Recommendation Engine using Collaborative
Filtering import pandas as pd
from sklearn.metrics.pairwise import cosine similarity
movies = pd.read csv("movies.csv") # columns: movieId, title, genres
ratings = pd.read csv("ratings.csv") # columns: userId, movieId, rating
data = pd.merge(ratings, movies, on='movieId')
user movie matrix = data.pivot table(index='userId',
columns='title', values='rating')
user movie matrix.fillna(0, inplace=True)
user similarity = cosine similarity(user movie matrix)
user similarity df = pd.DataFrame(user similarity,
                    index=user movie matrix.index,
                    columns=user movie matrix.index)
def get recommendations(user id, num recommendations=5):
   similar users = user similarity df[user id].sort values(ascending=False)
[1:6].index
  similar users ratings = user movie matrix.loc[similar users]
    mean ratings = similar users ratings.mean(axis=0)
    watched movies = user movie matrix.loc[user id]
  watched movies = watched movies [watched movies > 0].index
  mean_ratings = mean_ratings.drop(watched_movies,
  errors='ignore')
```

```
return
mean_ratings.sort_values(ascending=False).head(num_recommendations)
print("Top 5 Movie Recommendations for User 1:")
print(get_recommendations(1))

OUTPUT:
```

PS C:\Users\T PAVAN\OneDrive\Desktop\New folder> & "C:/Users/T PAVAN/AppData/Local/Programs/Python/Python3 14/python.exe" "c:/Users/T PAVAN/OneDrive/Desktop/New folder/vtu25749.py"

CONCLUSION:

The **Movie Recommendation Engine** successfully recommends movies based on user preferences using a collaborative filtering approach. By analyzing similarities between users' ratings, the system predicts and suggests relevant movies effectively. This project demonstrates the application of machine learning in real-world personalization systems, such as Netflix, Amazon Prime, and IMDb.