

Movie Recommendation System

Objective : Is to create a recommendation system that seeks to predict or filter preferences according to the user's choices. Recommender systems produce a list of recommendations in any of the two ways - Collaborative filtering: Collaborative filtering approaches build a model from the user's past behavior (i.e., items purchased or searched by the user) as well as similar decisions made by other users. This model is then used to predict items (or ratings for items) that users may have an interest in. Content-based filtering: Content-based filtering approaches use a series of discrete characteristics of an item in order to recommend additional items with similar properties. Content-based filtering methods are totally based on a description of the item and a profile of the user's preferences. It recommends items based on the user's past preferences.

IMPORT LIBRARY

```
import pandas as pd
```

```
import numpy as np
```

IMPORT DATASET

```
df=pd.read_csv('https://github.com/YBI-Foundation/Dataset/raw/main/Movies%20Recommendation.csv')
```

DESCRIBE DATA

```
df.head()
```

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```
df.info()
```

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```
df.shape
```

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```
df.columns
```

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GET FEATURE SELECTION

```
df_features=df[['Movie_Genre','Movie_Keywords','Movie_Tagline','Movie_Cast','Movie_Director']].fillna('')
```

```
df_features.shape
```

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```
df_features
```

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Next steps: [Generate code with df_features](#) [View recommended plots](#)

```
x = df_features['Movie_Genre'] + ' ' + df_features['Movie_Keywords'] + ' ' + df_features['Movie_Tagline'] + ' ' + df_features['Movie_Cast']
```

```
x
```

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```
x.shape
```

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GET FEATURE TEXT CONVERSION TO TOKENS

```
from sklearn.feature_extraction.text import TfidfVectorizer
```

```
tfidf = TfidfVectorizer()
```

```
x = tfidf.fit_transform(x)
```

```
x.shape
```

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```
print(x)
```

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GET SIMILARITY SCORE USING COSINE SIMILARITY (cosine_similarity computes the L2-Normalised dot product to vector)

```
from sklearn.metrics.pairwise import cosine_similarity
```

```
Similarity_Score = cosine_similarity(x)
```

```
Similarity_Score
```


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```
Similarity_Score.shape
```

 Show hidden output

GET MOVIE NAME AS INPUT FROM USER AND VALIDATE FOR CLOSEST SPELLING

```
Favourite_Movie_Name = input('Enter your favourite movie name : ')
```

 Enter your favourite movie name : avtaar


```
All_Movies_Title_List = df['Movie_Title'].tolist()
```

```
import difflib
```

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```
Movie_Recommendation = difflib.get_close_matches(Favourite_Movie_Name, All_Movies_Title_List)
```

```
print(Movie_Recommendation)
```

 ['Avatar', 'Gattaca']

```
Close_Match = Movie_Recommendation[0]
```

```
print(Close_Match)
```

 Avatar

```
Index_of_Close_Match_Movie = df[df.Movie_Title == Close_Match]['Movie_ID'].values[0]
```

```
print(Index_of_Close_Match_Movie)
```

 2692

```
Recommendation_Score = list(enumerate(Similarity_Score[Index_of_Close_Match_Movie]))
print(Recommendation_Score)
```

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```
len(Recommendation_Score)
```

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GET ALL MOVIES SORTED BASED ON RECOMMENDATION SCORE WRT FAVIOURITE MOVIES

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```
Sorted_Similar_Movies = sorted(Recommendation_Score, key = lambda x:x[1], reverse = True)
print(Sorted_Similar_Movies)
```

 [(2692, 1.0000000000000002), (3276, 0.11904275527845871), (3779, 0.10185805797079382), (62, 0.10153560702418994), (2903, 0.1006378731438

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```
print('Top 30 Movies suggested for you : \n')
```

```
i = 1

for movie in Sorted_Similar_Movies:
    index = movie[0]
    title_from_index = df[df.index==index]['Movie_Title'].values[0]
    if (i<31):
        print(i, '.',title_from_index)
        i+=1
```

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TOP 10 MOVIE RECOMMENDATION SYSTEM

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```
Movie_Name = input(' Enter your favourite movie name : ')
list_of_all_titles = df['Movie_Title'].tolist()
Find_Close_Match = difflib.get_close_matches(Movie_Name, list_of_all_titles)
Close_Match = Find_Close_Match[0]
Index_of_Movie = df[df.Movie_Title == Close_Match]['Movie_ID'].values[0]
Recommendation_Score = list(enumerate(Similarity_Score[Index_of_Movie]))
sorted_similar_movies = sorted(Recommendation_Score, key = lambda x:x[1], reverse = True)
print('Top 10 Movies suggested for you : \n')
i = 1
for movie in sorted_similar_movies:
    index = movie[0]
    title_from_index = df[df.Movie_ID==index]['Movie_Title'].values
    if (i<11):
        print(i, '.',title_from_index)
        i+=1
```

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Start coding or [generate](#) with AI.

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