

# Movie Recommendation Engine - Beginner Guide

## Introduction

We're building a simple system that recommends movies similar to a movie the user likes.

Like how Netflix shows 'Because you watched X...'

## Libraries Used

Libraries Used:

- pandas: To read and work with CSV files (data handling)
- sklearn.metrics.pairwise.cosine\_similarity: To find similarity between movies

## Step-by-Step Code Explanation

Step 1: Load the CSV Files

-----

```
ratings = pd.read_csv("ratings.csv")
```

```
movies = pd.read_csv("movies.csv")
```

Step 2: Merge the Two Files

-----

```
data = pd.merge(ratings, movies, on='movieId')
```

Step 3: Create a User-Movie Rating Table

-----

```
user_movie_matrix = data.pivot_table(index='userId', columns='title', values='rating')
```

Step 4: Fill Missing Ratings with 0

-----

```
user_movie_matrix.fillna(0, inplace=True)
```

Step 5: Calculate Similarity Between Movies

# Movie Recommendation Engine - Beginner Guide

```
-----  
from sklearn.metrics.pairwise import cosine_similarity  
similarity_matrix = cosine_similarity(user_movie_matrix.T)
```

## Step 6: Convert Similarity Matrix to DataFrame

```
-----  
movie_similarity_df = pd.DataFrame(similarity_matrix, index=user_movie_matrix.columns,  
columns=user_movie_matrix.columns)
```

## Step 7: Create the Recommendation Function

```
-----  
def recommend_movies(movie_title, top_n=5):  
    if movie_title not in movie_similarity_df.columns:  
        return f"Movie '{movie_title}' not found in dataset."  
    similarity_scores = movie_similarity_df[movie_title].sort_values(ascending=False)  
    return similarity_scores[1:top_n+1].index.tolist()
```

## Step 8: Run It!

```
-----  
movie_to_search = "Toy Story (1995)"  
recommendations = recommend_movies(movie_to_search)  
print(f"Recommendations for '{movie_to_search}':")  
for i, rec in enumerate(recommendations, 1):  
    print(f"{i}. {rec}")
```

## Sample Output

Sample Output:

```
-----  
Recommendations for 'Toy Story (1995)':
```

# Movie Recommendation Engine - Beginner Guide

1. Bug's Life, A (1998)
2. Aladdin (1992)
3. Beauty and the Beast (1991)
4. Lion King, The (1994)
5. Hercules (1997)

## Conclusion

Congratulations! You've built a simple content-based movie recommendation engine using user rating data.

You can improve this project by:

- Adding genre-based filtering
- Using collaborative filtering with surprise or LightFM
- Building a web app interface using Streamlit

Best of luck with your project!