





Phase-2 Submission By

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GitHub Repository: https://github.com/VIGNESH1562006/Delvering-Personalized-movie-recommendations-with-an-AI-driven-matchmaking-system.git

<u>TOPIC: Delvering Personalized movie recommendations with an AI-driven matchmaking system</u>

1.Problem Statement

Current movie recommendation systems primarily rely on user history or general popularity trends. They often miss nuanced factors like mood, preferred story arcs, niche genres, or evolving tastes over time. A more dynamic, "matchmaking" approach could bridge this gap and deliver recommendations that feel more personal and relevant.

2. Objectives of the Project

□ **Develop** an AI model that **learns user preferences** beyond simple genres or ratings.







| Design a matchmaking engine that pairs users with movies based on detailed |
|---|
| profiling (e.g., emotional tone, themes, pacing). |
| Continuously update recommendations as user interactions evolve. |
| Provide an engaging user interface that makes discovery enjoyable. |

3. Scope of the Project

In-Scope:

- Building a hybrid AI model that combines content-based and collaborative filtering for personalized recommendations.
- Creating **detailed user profiles** based on explicit (ratings, preferences) and implicit (watch time, likes) feedback.
- Integrating a matchmaking algorithm that finds the best movie-user matches based on dynamic profiling.
- Developing a **prototype web interface** (optional) for users to interact with the system and receive recommendations.
- Using public movie datasets like MovieLens or TMDB API for data.

Out-of-Scope:

- Building a full production-ready streaming platform (e.g., like Netflix).
- Collecting private, real-world user data beyond public datasets (for privacy reasons).
- Creating an extensive mobile app the focus will be on a simple prototype web version if needed.

4.Data Sources

1. MovieLens Dataset

Description:







One of the most popular datasets for movie recommendation research. Contains millions of movie ratings, user IDs, movie IDs, timestamps, and tags (like genres, moods, etc.).

· Link:

https://grouplens.org/datasets/movielens/

- Why use it?
 - o Rich user-movie interaction data. o Available in multiple sizes (small: 100K ratings, medium: 1M ratings, large: 20M+ ratings).

2. The Movie Database (TMDB) API

Description:

TMDB provides detailed movie metadata — such as genres, posters, descriptions, release dates, cast, crew, and reviews — via a free API.

- Link: https://developer.themoviedb.org/docs
- Why use it?
 - o To **enrich** your movie data with more attributes (e.g., director, popularity score, language, poster images). o Helps in building a better **content-based** recommendation engine.

3. IMDb Datasets (via IMDb Official or IMDbPY Library)

Description:

IMDb offers public downloadable datasets, including title basics, ratings, genres, actors, and crew.

Link:

https://datasets.imdbws.com/

- Why use it?
 - o Massive database of global movies and TV shows. o Good for cross-referencing additional movie features.

4. Kaggle Movie Datasets Examples:







o Netflix Movies and TV Shows Dataseto IMDb 5000 Movie Dataseto MoviesMetadata from TMDB

• Link:

https://www.kaggle.com/datasets (Search "Movies" or "Movie Recommendations")

• Why use it?

Languages: Python

- o Quickly get curated datasets for faster prototyping.
- o Some Kaggle datasets are already cleaned and combined.

5.High-Level Methodology

| | Data Collection: Use open-source datasets (e.g., MovieLens) containing user | | | | |
|-------------|---|--|--|--|--|
| | ratings, genres, tags, and movie metadata. | | | | |
| | User Profiling: Create detailed user profiles based on past interactions, | | | | |
| | preferences, and implicit feedback. | | | | |
| | Content-Based + Collaborative Filtering: Use a hybrid recommendation | | | | |
| | approach. | | | | |
| | Deep Learning: Train a neural network model (like an Autoencoder or | | | | |
| | DeepFM) to predict user-movie matches. | | | | |
| | Feedback Loop: Continuously refine recommendations based on real-time | | | | |
| | feedback (like/dislike, watch time). | | | | |
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| 6. 5 | 5.Tools and Technologies | | | | |







| Libraries: TensorFlow / PyTorch, Scikit-Learn, Surprise (for recommender systems) |
|--|
| Database: PostgreSQL or MongoDB for storing profiles and movies |
| Frontend (Optional): ReactJS or simple Flask web app |
| APIs: TMDB (The Movie Database) API for movie metadata and posters |

7.Team Members and Roles

| Name | Role | Key Responsibilities |
|----------------|--------------------------------------|---|
| Vignesh V | Project Manager + Data Scientist | Manage team, design AI models |
| Vinisha M | Machine Learning Engineer | Build and optimize recommendation engine |
| Vishalini S | Frontend Developer/Backend Developer | Create a simple web interface for users, Handledatabase, APIintegration,serversetup |
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