





**Student Name:** MUDABBIRA FATHIMA S

**Register Number:**510623104070

**Institution:**C.ABDUL HAKEEM COLLEGE OF

ENGINEERING AND TECHNOLOGY

**Department:**BE.COMPUTER SCIENCE ENGINEERING

Date of Submission: 08-05-2025

**Github Repository Link:** 

https://github.com/Mudabbirafathima/Movie-recommendations.git

#### 1. Problem Statement

The challenge lies in improving AI-driven personalization for movie recommendations. By analyzing behavioral patterns, ratings, and sentiment analysis, the system can make refined and engaging matches, enhancing user experience.

## 2. Project Objectives

- Develop a hybrid recommendation model combining collaborative filtering and content-based filtering.
- Increase recommendation accuracy using user sentiment analysis from reviews.
- Ensure adaptability with self-learning AI models that refine suggestions dynamically.
- Compare various machine learning approaches to optimize performance

## 3. Flowchart of the Project Workflow

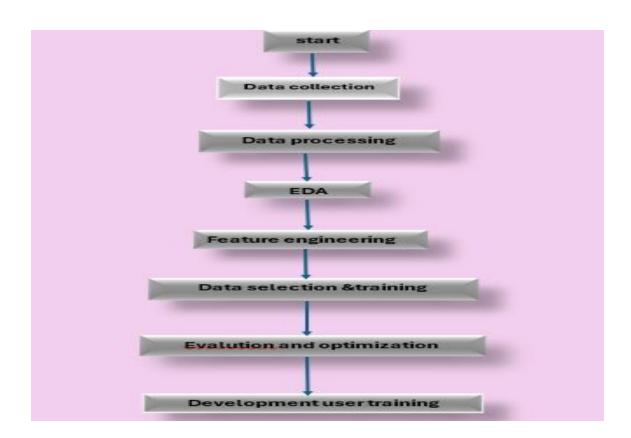
1.Start – Initiate the process.







- **2.Data Collection** Gather movie details and user preferences.
- 3.Data Processing Clean and encode relevant data features.
- **4.EDA & Insights** Analyze user-movie interactions to find patterns.
- **5.Feature Engineering** Enhance input data for improved recommendations.
- **6.Model Building** Implement and evaluate recommendation algorithms.
- 7. Deployment & Visualization Present AI-driven movie suggestions.
- **8.User Feedback Loop** Refine recommendations based on user engagement.



## 4. Data Description

*The dataset includes:* 

- 1. Movie metadata (genres, ratings, cast, director, release year).
- 2. User preferences (watch history, likes, dislikes).







- 3. Review sentiments (text analysis for emotional tone).
- 4. Streaming platform trends (popularity rankings).

## 5. Data Preprocessing

- Handling missing values (imputation).
- Encoding categorical features (genres, cast).
- Processing text for sentiment analysis (tokenization, stop-word removal).
- Feature scaling & normalization for enhanced model performance

# 6. Exploratory Data Analysis (EDA)

- -Genre-watch frequency trends to recognize popular categories.
- -Correlation mapping of ratings and user engagement.
- -User behavior segmentation through clustering.
- -Sentiment-based polarity scoring for enhanced personalization.

## 7. Feature Engineering:

- -User-movie interaction matrices for collaborative filtering.
- -TF-IDF feature extraction for analyzing movie descriptions.
- -Embedding-based similarity measures\*\* for deep personalization.
- Contextual tagging\*\* to ensure dynamic AI adaptability.







## 8. Model Building:

Comparison between approaches:

Matrix Factorization (SVD) for user-movie matchmaking.

Neural Collaborative Filtering for deep-learning personalization.

Hybrid Approach integrating multiple filtering techniques.

*Graph-based models analyzing structured connectivity insights.* 

## 9. Visualization of Results & Model Insights

Heatmapsfor similarity tracking.

- Feature importance rankings showing influential parameters.
- -Precision-recall curves assessing recommendation accuracy.
- $\hbox{-} {\it Confusion matrices for classification validation}.$

# 10. Tools and Technologies Used

- *Sufiya Firdouse*: Python The primary language for data handling and model building.
- **Roshni**: Pandas & NumPy Efficient preprocessing and feature engineering.
- *Pooja Sri*: Matplotlib & Seaborn Data visualization and exploratory data analysis (EDA).
- **Sandhiya**: Surprise & LightFM Collaborative and hybrid recommendation model development.







- Mudabbira Fathima: TensorFlow / PyTorch Deep learning models for advanced recommendations.
- *Monisha*: Flask / FastAPI Backend development and API deployment for the recommendation system.

#### 11. Team Members and Contributions

*Here's how our team:* 

#### Team Members & Responsibilities:

- Sufiya Firdouse Feature Engineering: Extracting and transforming relevant attributes for model improvement.
- Roshni EDA (Exploratory Data Analysis): Identifying patterns, trends, and insights from the dataset.
- **Pooja Sri Data Cleaning**: Handling missing values, inconsistencies, and preparing structured data for modeling.
- *Sandhiya Model Development*: Implementing collaborative filtering and hybrid recommendation models.
- *Mudabbira Fathima Advanced Model Optimization*: Exploring deep learning and fine-tuning recommendation algorithms.
- Monisha Documentation & Reporting: Structuring workflows.