

Capstone Project
23CSP-339 - Full Stack-I

Capstone Recommendation Engine

A PROJECT REPORT

Submitted by

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in partial fulfilment for the award of the degree of

Bachelor of Technology in Computer Science

IN

Big Data Analytics



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BONAFIDE CERTIFICATE

Certified that this project report "**Capstone Recommendation Engine**" is the Bonafide work of **Yashasvi** carried out the project work under my/our supervision.

SIGNATURE

SIGNATURE

Submitted for the project viva-voice examination held on

INTERNAL EXAMINER

EXTERNAL EXAMINER

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Project Title

Capstone Recommendation Engine

Project Description

The Capstone Recommendation Engine is a full-stack web application that provides personalized movie recommendations to users. The application is built using React.js for the user interface, Node.js Express for the backend API, and MongoDB as the database. A Python FastAPI service implements the recommendation logic, dynamically loading movies based on real-time data in MongoDB.

Objectives

- Develop a functional recommendation engine supporting add/view features for users and movies.
- Integrate collaborative filtering for better user recommendations.
- Demonstrate scalable full-stack architecture (React, Express, FastAPI, MongoDB).

Key Features

- Users and movies can be added live from the frontend.
- Recommendations update automatically for all users as movies/interactions are added.
- Backend APIs for CRUD operations.
- Recommendation logic handled by Python FastAPI, results sent to frontend via Express.js.

Hardware/Software Requirements

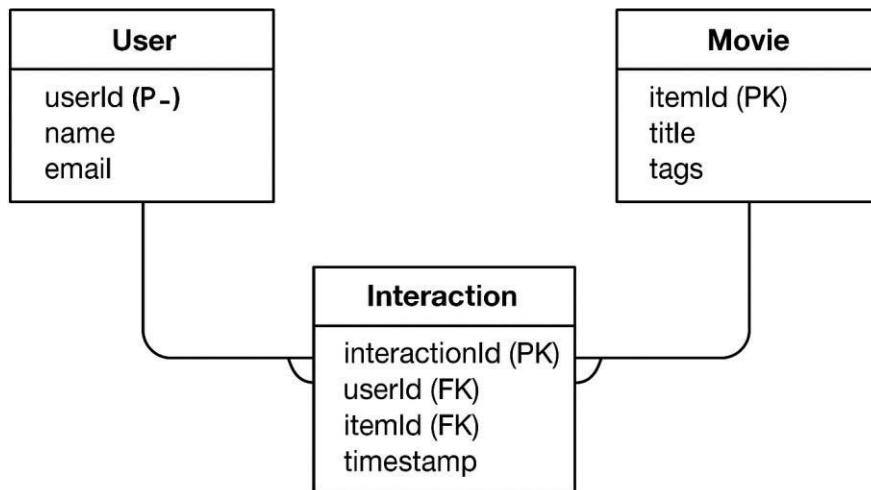
- **Hardware:** Intel i3+, 8GB RAM, 100GB HDD/SSD
- **Software:** Node.js v24.x, MongoDB v7.x, React.js, FastAPI (Python 3.14), VSCode or similar IDE, Google Chrome

ER Diagram

Entities:

- **User** – Stores user information
- **Movie** – Stores movie data with tags
- **Recommendation** – Stores movie scores

[User] ---< [Recommendation] >--- [Movie]



Database Schema

Movie Collection:

Field	Type	Description
ItemId	String	Unique movie identifier
title	String	Movie title
tags	Array	Tags associated with the movie

User Collection:

Field	Type	Description
userId	ObjectId	Unique user identifier

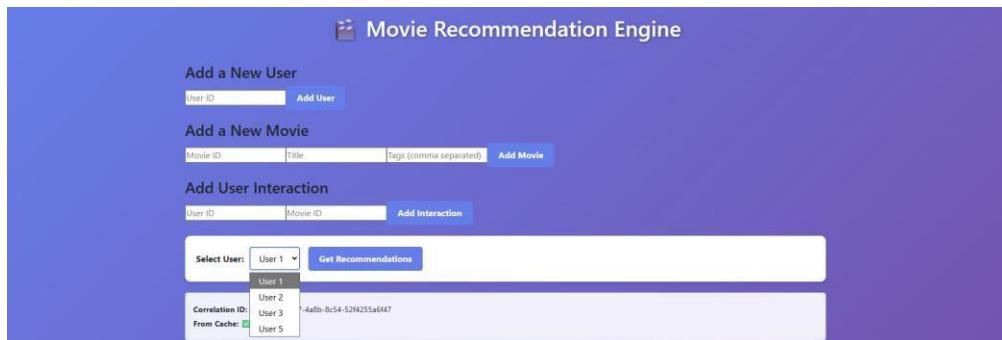
Recommendation Collection:

Field	Type	Description
userId	String	Linked user
ItemId	String	Linked movie

Front-End Screens

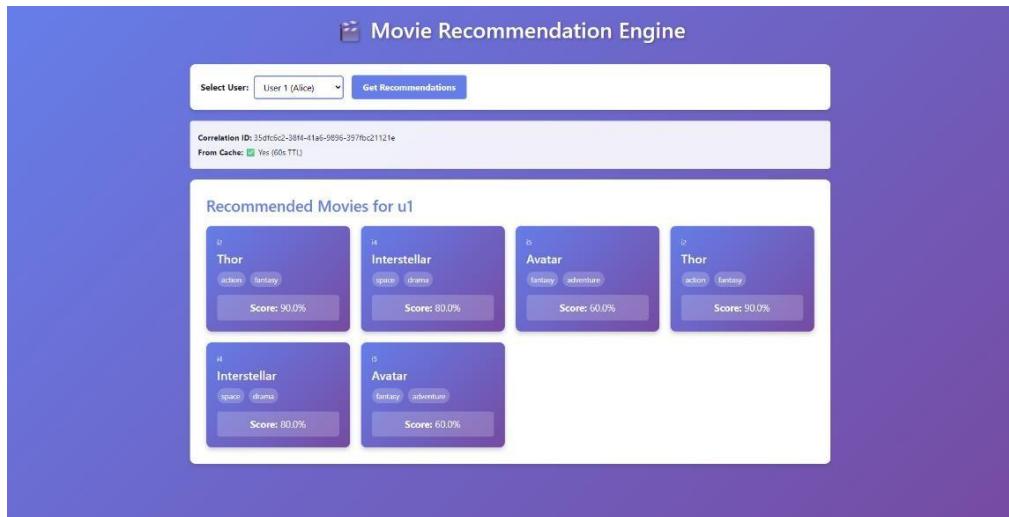
Home Screen:

- The Home Screen includes forms to add users and movies, and the dropdown to select users



Recommendation Screen:

- The recommended movies display (carousel or card list).
- Movie titles, tags, and recommendation scores visible for each recommended item.



Output Screens and Reports

Backend calculates recommendation scores and returns **JSON responses**.

C ⌂ ⌂ localhost:5000/api/recommendations?userId=12345&pageSize=5

Pretty-print □

```
{"results": [{"_id": "69024f3d27681a6a8dac78fd", "itemId": "i1", "title": "Iron Man", "tags": ["action", "superhero", "marvel"], "score": "0.48"}, {"_id": "69024f3d27681a6a8dac78fe", "itemId": "i2", "title": "Thor", "tags": ["action", "superhero", "fantasy"], "score": "0.35"}, {"_id": "69024f3d27681a6a8dac78ff", "itemId": "i3", "title": "Inception", "tags": ["sci-fi", "thriller"], "score": "0.34"}, {"_id": "69024f3d27681a6a8dac7900", "itemId": "i4", "title": "Interstellar", "tags": ["space", "drama"], "score": "0.70"}, {"_id": "69024f3d27681a6a8dac7901", "itemId": "i5", "title": "Avatar", "tags": ["fantasy", "adventure"], "score": "0.97"}], "correlationId": "24020186-97e5-4fa7-81ea-398a0cac1a32", "fromCache": false}
```

```
C:\Users\laksh>curl "http://localhost:5000/api/recommendations?userId=12345&pageSize=5"
{"results": [{"_id": "69024f3d27681a6a8dac78fd", "itemId": "i1", "title": "Iron Man", "tags": ["action", "superhero", "marvel"], "score": "0.48"}, {"_id": "69024f3d27681a6a8dac78fe", "itemId": "i2", "title": "Thor", "tags": ["action", "superhero", "fantasy"], "score": "0.35"}, {"_id": "69024f3d27681a6a8dac78ff", "itemId": "i3", "title": "Inception", "tags": ["sci-fi", "thriller"], "score": "0.34"}, {"_id": "69024f3d27681a6a8dac7900", "itemId": "i4", "title": "Interstellar", "tags": ["space", "drama"], "score": "0.70"}, {"_id": "69024f3d27681a6a8dac7901", "itemId": "i5", "title": "Avatar", "tags": ["fantasy", "adventure"], "score": "0.97"}], "correlationId": "24020186-97e5-4fa7-81ea-398a0cac1a32", "fromCache": false}
```

MongoDB stores the real time data for the recommendation engine

The image displays three separate MongoDB Compass sessions, each showing a different collection in the `recodb` database.

- Top Session: interactions**

 - Documents:** Shows five documents representing user interactions. Each document includes fields like `_id`, `userId`, `itemId`, and `type`.
 - Sample Data:**
 - `_id: ObjectId('69027c93d376e7327b09d396')`, `userId: "u1"`, `itemId: "i1"`, `type: "view"`
 - `_id: ObjectId('69027c93d376e7327b09d397')`, `userId: "u1"`, `itemId: "i3"`, `type: "purchase"`
 - `_id: ObjectId('69027c93d376e7327b09d398')`, `userId: "u2"`, `itemId: "i2"`, `type: "view"`
 - `_id: ObjectId('6902ce655c36763418cd2108')`, `userId: "s" (Lakshmi)`, `itemId: "i6"`
 - `_id: ObjectId('6902d39e18ddb36964aa203c')`, `userId: "u2"`, `itemId: "i7"`

- Middle Session: items**

 - Documents:** Shows seven documents representing items. Each item has fields like `_id`, `itemId`, `title`, and `tags`.
 - Sample Data:**
 - `_id: ObjectId('69027c93d376e7327b09d38f')`, `itemId: "i1"`, `title: "Iron Man"`, `tags: Array (2)`
 - `_id: ObjectId('69027c93d376e7327b09d390')`, `itemId: "i2"`, `title: "Thor"`, `tags: Array (2)`
 - `_id: ObjectId('69027c93d376e7327b09d391')`, `itemId: "i3"`, `title: "Inception"`, `tags: Array (2)`
 - `_id: ObjectId('69027c93d376e7327b09d392')`, `itemId: "i4"`, `title: "Interstellar"`, `tags: Array (2)`
 - `_id: ObjectId('69027c93d376e7327b09d393')`, `itemId: "i5"`, `title: "Avatar"`, `tags: Array (2)`
 - `_id: ObjectId('69027c93d376e7327b09d394')`, `itemId: "i6"`, `title: "The Godfather"`, `tags: Array (2)`
 - `_id: ObjectId('69027c93d376e7327b09d395')`, `itemId: "i7"`, `title: "Pulp Fiction"`, `tags: Array (2)`

- Bottom Session: users**

 - Documents:** Shows four documents representing users. Each user has fields like `_id`, `userId`, and `name`.
 - Sample Data:**
 - `_id: ObjectId('69027c93d376e7327b09d394')`, `userId: "u1"`, `name: "Alice"`
 - `_id: ObjectId('69027c93d376e7327b09d395')`, `userId: "u2"`, `name: "Bob"`
 - `_id: ObjectId('6902d393d8b3afce547a06e')`, `userId: "u3"`
 - `_id: ObjectId('6902d6073d8b3afce547a06f')`, `userId: "u5"`

Limitation & Future Scope

Limitations

- Movie-only Recommendations:
The current system is limited to recommending movies. Other types of media are not supported yet.
- Basic Scoring Logic:
Recommendation scores are generated using simple rules or basic similarity; deep personalization based on advanced user behavior or machine learning is not yet implemented.
- No Authentication:
The platform does not require users to log in or register securely. All user management is open and does not restrict access.
- No User Profiles:
User-specific preferences and histories beyond basic interactions are not tracked or used for recommendation refinement.
- Limited User Interface:
UI is functional but lacks advanced design, accessibility features, and mobile optimization.

Future Scope

- User Authentication & Profiles:
Implement secure login and individual user account features to enable saving preferences and history.
- Machine Learning-Based Recommendations:
Upgrade the backend to use machine learning models for smarter, context-aware recommendations based on more complex user data.
- Multi-Category Expansion:
Extend the engine to recommend other content types such as books, music, or games for a richer user experience.
- Real-Time Analytics Dashboard:
Add dashboards for administrators or users to see usage statistics, popular items, and system performance metrics.
- User Ratings & Feedback:
Enable users to rate movies and provide feedback to further personalize future recommendations.
- Improved UI/UX:
Enhance visual appeal, usability, and responsiveness to deliver a professional-grade user experience.

GitHub URL

<https://github.com/LakishaJaiswal/Recommendation-Engine/tree/main>

Presentation Slides



Objective

- ▶ To create a full-stack web application that recommends movies to users based on their preferences and past interactions.
- ▶ Problem Statement:
Finding relevant movies is difficult as user choices and preferences vary widely. Personalized recommendations simplify decision-making and improve user satisfaction.

System Architecture / Workflow

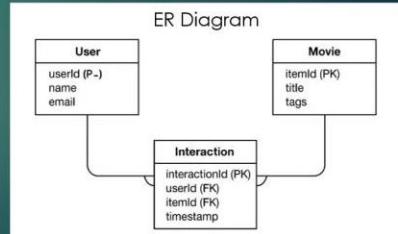
- ▶ Architecture Overview:
 - Frontend sends/receives user data, interacts with backend.
 - Backend processes requests, connects with MongoDB and Python model service.
 - Model Service dynamically generates recommendations using data from MongoDB.
- ▶ Workflow:
 - User interacts with frontend forms (add/view user/movie, get recommendations).
 - Backend APIs trigger data storage and retrieval.
 - Model service fetches movies, generates top suggestions for requested user.

Tech Stack

- ▶ Frontend: React.js for fast, interactive UI
- ▶ Backend: Node.js with Express for scalable REST APIs
- ▶ Database: MongoDB for flexible, real-time storage
- ▶ Model Microservice: Python FastAPI for dynamic recommendation logic
- ▶ Other Tools: Axios (API requests), NodeCache (caching), UUID (unique identifiers)

Database Design

- ▶ ER Diagram:
 - Three collections: Users, Items (Movies), Interactions
 - Users and Movies are connected via Interactions collection (mapping userId and itemId).
- ▶ Database Schema:
 - Users: userId
 - Items: itemId, title, tags
 - Interactions: userId, itemId



Key Features and Working

► Key Features:

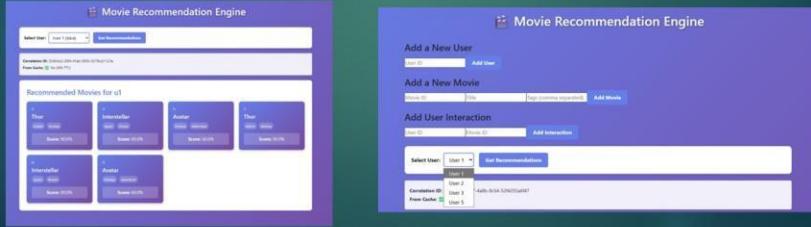
- Add new users, movies, and interactions through easy-to-use forms.
- Recommendations update instantly with each database change.
- Clean, responsive UI allows real-time testing.
- Data persists in MongoDB and is always up-to-date.

► Working:

- Model Service (Python FastAPI) fetches all movies from MongoDB.
- Randomly selects/topK movies for recommendations (demo logic).
- Scores are assigned to each recommended movie.
- Results are sent as JSON to the backend and displayed in the frontend.

Front-End Screens

- Home: User selection, add movie/user/interaction forms.
- Recommendations: Display of recommended movies as cards/carousel.
- Description:
Users can add themselves and movies, then immediately view personalized recommendations in an intuitive interface.



Limitations and Future Scope

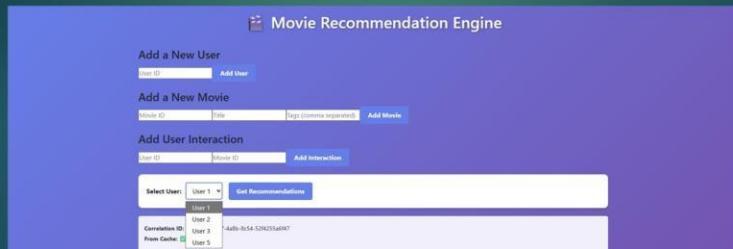
► Limitations

- Only supports movie recommendations (no other media types yet).
- Recommendation scores use basic, non-personalised logic.
- No password/authentication; open user management.
- UI and features are focused on core functionality.

► Future Scope

- Integrate user authentication and secure logins.
- Add machine learning models for smarter recommendations.
- Expand engine to support books, music, and other content.
- Add user ratings, reviews, and preference tracking.
- Develop advanced analytics dashboards for usage and results.

Outputs:



This screenshot shows the 'Movie Recommendation Engine' UI after selecting 'User 1'. It displays a list of recommended movies with their scores: Thor (Score: 90.0%), Interstellar (Score: 80.0%), Avatar (Score: 60.0%), Thor (Score: 90.0%), Interstellar (Score: 80.0%), and Avatar (Score: 60.0%). Below this, a detailed view of the recommendation process is shown in a MongoDB Compass interface, illustrating the query and pipeline stages involved in generating the results.

This screenshot shows a browser window at localhost:5000/api/recommendations?userId=12345&pageSize=5. The response is a JSON object containing an array of movie documents. To the right, two instances of the MongoDB Compass interface are shown, both displaying the same recommendation pipeline and data, which corresponds to the JSON output from the API.

```
{"results": [{"_id": "69024f3d27681a6a8dac78fd", "itemId": "i1", "title": "Iron Man", "tags": ["action", "superhero", "marvel"], "score": "0.48"}, {"_id": "69024f3d27681a6a8dac78fe", "itemId": "i2", "title": "Thor", "tags": ["action", "superhero", "fantasy"], "score": "0.35"}, {"_id": "69024f3d27681a6a8dac78ff", "itemId": "i3", "title": "Inception", "tags": ["sci-fi", "thriller"], "score": "0.34"}, {"_id": "69024f3d27681a6a8dac7900", "itemId": "i4", "title": "Interstellar", "tags": ["space", "drama"], "score": "0.70"}, {"_id": "69024f3d27681a6a8dac7901", "itemId": "i5", "title": "Avatar", "tags": ["fantasy", "adventure"], "score": "0.97"}], "correlationId": "24020186-97e5-4fa7-81ea-398a0cac1a32", "fromCache": false}
```

GitHub & References

- ▶ GitHub Repository: [LakishaJaiswal/Recommendation-Engine](https://github.com/LakishaJaiswal/Recommendation-Engine)
- ▶ Documentation:
 - ▶ MongoDB: <https://mongodb.com/docs>
 - ▶ Node.js: <https://nodejs.org/en/docs>
 - ▶ React.js: <https://react.dev>
 - ▶ FastAPI: <https://fastapi.tiangolo.com>
 - ▶ ByteXL Capstone Guidelines

THANK YOU

Project Code

Backend

- **server.js** (main Express server, MongoDB + APIs)

javascript

```
const express = require('express');
const cors = require('cors');
const axios = require('axios');
const { MongoClient } = require('mongodb');
const NodeCache = require('node-cache');
const { v4: uuidv4 } = require('uuid');
```

```
const app = express();
const cache = new NodeCache({ stdTTL: 60 });

app.use(cors({
  origin: ['http://localhost:3000'],
  credentials: true
}));
app.use(express.json());

const MONGO_URL = 'mongodb://localhost:27017';
const DB_NAME = 'recodb';
const MODEL_URL = 'http://localhost:8000';

app.get('/health', (req, res) => {
  res.json({ status: 'Backend running on port 5000' });
});

// Add User
app.post('/api/adduser', async (req, res) => {
  const { userId } = req.body;
  try {
    const client = new MongoClient(MONGO_URL);
    await client.connect();
    const db = client.db(DB_NAME);
    await db.collection('users').updateOne({ userId }, { $set: { userId } }, { upsert: true });
    await client.close();
    res.json({ status: 'User added!', userId });
  } catch (err) {
    res.status(500).json({ error: err.message });
  }
});

// Get All Users
app.get('/api/users', async (req, res) => {
  try {
    const client = new MongoClient(MONGO_URL);
    await client.connect();
  }
```

```
const db = client.db(DB_NAME);
const users = await db.collection('users').find({ }).toArray();
await client.close();
res.json(users.map(u => u.userId));
} catch (err) {
  res.status(500).json({ error: err.message });
}
});
```

// Add Movie

```
app.post('/api/addmovie', async (req, res) => {
  const { itemId, title, tags } = req.body;
  try {
    const client = new MongoClient(MONGO_URL);
    await client.connect();
    const db = client.db(DB_NAME);
    await db.collection('items').insertOne({ itemId, title, tags });
    await client.close();
    res.json({ status: 'Movie added!', itemId });
  } catch (err) {
    res.status(500).json({ error: err.message });
  }
});
```

// Add Interaction

```
app.post('/api/interact', async (req, res) => {
  const { userId, itemId } = req.body;
  try {
    const client = new MongoClient(MONGO_URL);
    await client.connect();
    const db = client.db(DB_NAME);
    await db.collection('interactions').insertOne({ userId, itemId });
    await client.close();
    res.json({ status: 'Interaction saved!', userId, itemId });
  } catch (err) {
    res.status(500).json({ error: err.message });
  }
});
```

```

});

// Get Recommendations

app.get('/api/recommendations', async (req, res) => {
  const userId = req.query.userId;
  const cacheKey = `reco_${userId}`;
  const correlationId = uuidv4();

  const cached = cache.get(cacheKey);
  if (cached) {
    return res.json({
      results: cached,
      fromCache: true,
      correlationId: correlationId
    });
  }

  try {
    const modelRes = await axios.get(`${MODEL_URL}/recommend`, {
      params: { userId, topK: 5 }
    });

    const itemIds = modelRes.data.recommendations.map(r => r.itemId);
    const client = new MongoClient(MONGO_URL);
    await client.connect();
    const db = client.db(DB_NAME);
    const results = await db.collection('items').find({ itemId: { $in: itemIds } })
      .toArray();
    await client.close();

    const movieMap = {};
    results.forEach(movie => { movieMap[movie.itemId] = movie; });

    const uniqueResults = [];
    const seenIds = new Set();
    for (const rec of modelRes.data.recommendations) {
      if (!seenIds.has(rec.itemId) && movieMap[rec.itemId]) {
        uniqueResults.push(movieMap[rec.itemId]);
      }
    }
  }
}

```

```

        uniqueResults.push({ ...movieMap[rec.itemId], score:
parseFloat(rec.score) });

        seenIds.add(rec.itemId);

    }

}

cache.set(cacheKey, uniqueResults, 60);
res.json({
    results: uniqueResults,
    fromCache: false,
    correlationId: correlationId
});
} catch (err) {
    res.status(500).json({ error: 'Recommendation error', details: err.message });
}
});

app.listen(5000, () => {
    console.log('✓ Backend running on port 5000');
});

```

- **db/seed.js** (for demo data)

```

const { MongoClient } = require('mongodb');

async function seed() {
    const client = new MongoClient('mongodb://localhost:27017');
    await client.connect();
    const db = client.db('recodb');

    // Clear existing
    await db.collection('items').deleteMany({});
    await db.collection('users').deleteMany({});

```

```

await db.collection('interactions').deleteMany({ });

// Insert movies
await db.collection('items').insertMany([
  { itemId: 'i1', title: 'Iron Man', tags: ['action', 'superhero'] },
  { itemId: 'i2', title: 'Thor', tags: ['action', 'fantasy'] },
  { itemId: 'i3', title: 'Inception', tags: ['sci-fi', 'thriller'] },
  { itemId: 'i4', title: 'Interstellar', tags: ['space', 'drama'] },
  { itemId: 'i5', title: 'Avatar', tags: ['fantasy', 'adventure'] }
]);
// Insert users
await db.collection('users').insertMany([
  { userId: 'u1', name: 'Alice' },
  { userId: 'u2', name: 'Bob' }
]);
// Insert interactions
await db.collection('interactions').insertMany([
  { userId: 'u1', itemId: 'i1', type: 'view' },
  { userId: 'u1', itemId: 'i3', type: 'purchase' },
  { userId: 'u2', itemId: 'i2', type: 'view' }
]);
console.log('✓ Database seeded successfully');
await client.close();

```

seed().catch(console.error);

Modelservice

app.py (Python FastAPI microservice, dynamic recommendations)

```

from fastapi import FastAPI, Query
from fastapi.middleware.cors import CORSMiddleware
from pymongo import MongoClient
import random

app = FastAPI()
app.add_middleware(

```

```
CORSMiddleware, allow_origins=["*"], allow_credentials=True,  
allow_methods=["*"], allow_headers=["*"])
```

```
MONGO_URL = "mongodb://localhost:27017"
```

```
DB_NAME = "recodb"
```

```
@app.get("/health")  
def health():  
    return { "status": "Model service running" }
```

```
@app.get("/recommend")  
def recommend(userId: str = Query(...), topK: int = Query(5)):  
    client = MongoClient(MONGO_URL)  
    db = client[DB_NAME]  
    all_movies = list(db.items.find({ }))  
    client.close()
```

```
if len(all_movies) == 0:  
    return { "recommendations": [] }
```

```
sampled = random.sample(all_movies, min(topK, len(all_movies)))  
recs = [  
    { "itemId": movie["itemId"], "score": round(random.uniform(0.7, 1.0), 2),  
    "title": movie["title"] }  
    for movie in sampled  
]  
return { "recommendations": recs }
```

Frontend

- **src/App.js** (React main file, dynamic UI)

```
import React, { useState, useEffect } from 'react';  
import axios from 'axios';  
import './App.css';
```

```
function App() {  
    const [userId, setUserId] = useState("");  
    const [userList, setUserList] = useState([]);
```

```
const [recommendations, setRecommendations] = useState([]);
const [loading, setLoading] = useState(false);
const [fromCache, setFromCache] = useState(false);
const [correlationId, setCorrelationId] = useState("");

const [newMovie, setNewMovie] = useState({ itemId: "", title: "", tags: "" });
const [newUser, setNewUser] = useState({ userId: "" });
const [newInteraction, setNewInteraction] = useState({ userId: "", itemId: "" });

// Fetch users from users collection
const fetchUsers = async () => {
  try {
    const res = await axios.get('http://localhost:5000/api/users');
    setUserList(res.data);
    if (res.data.length > 0 && userId === "") {
      setId(res.data[0]);
    }
  } catch {
    setUserList([]);
  }
};

// Fetch recommendations
const fetchRecommendations = async () => {
  if (!userId) return;
  setLoading(true);
  try {
    const response = await axios.get('http://localhost:5000/api/recommendations', {
      params: { userId }
    });
    setRecommendations(response.data.results);
    setFromCache(response.data.fromCache);
    setCorrelationId(response.data.correlationId);
  } catch (error) {
    console.error('Error fetching recommendations:', error);
    alert('Failed to fetch recommendations');
  }
};
```

```
 setLoading(false);
};

useEffect(() => {
  fetchUsers();
}, []);

useEffect(() => {
  fetchRecommendations();
  // eslint-disable-next-line
}, [userId]);

return (
<div className="App">
  <header className="App-header">
    <h1>🎬 Movie Recommendation Engine</h1>
  </header>

  <div className="container">

    {/* Add a New User */}
    <form
      style={{ marginBottom: 20, marginTop: 20 }}
      onSubmit={async (e) => {
        e.preventDefault();
        await axios.post('http://localhost:5000/api/adduser', {
          userId: newUser.userId,
        });
        alert('User added!');
        setNewUser({ userId: '' });
        fetchUsers();
      }}
    >
      <h3>Add a New User</h3>
      <input
        placeholder="User ID"
        value={newUser.userId}
      >
    
```

```

onChange={e => setNewUser({ userId: e.target.value })}
/>
<button type="submit">Add User</button>
</form>

/* Add a New Movie */
<form
  style={{ marginBottom: 20 }}
  onSubmit={async (e) => {
    e.preventDefault();
    await axios.post('http://localhost:5000/api/addmovie', {
      itemId: newMovie.itemId,
      title: newMovie.title,
      tags: newMovie.tags.split(',').map(tag => tag.trim()),
    });
    alert('Movie added!');
    setNewMovie({ itemId: "", title: "", tags: "" });
  }}
>
<h3>Add a New Movie</h3>
<input
  placeholder="Movie ID"
  value={newMovie.itemId}
  onChange={e => setNewMovie({ ...newMovie, itemId: e.target.value })}
/>
<input
  placeholder="Title"
  value={newMovie.title}
  onChange={e => setNewMovie({ ...newMovie, title: e.target.value })}
/>
<input
  placeholder="Tags (comma separated)"
  value={newMovie.tags}
  onChange={e => setNewMovie({ ...newMovie, tags: e.target.value })}
/>
<button type="submit">Add Movie</button>
</form>

```

```
/* Add User Interaction */
<form
  style={{ marginBottom: 20 }}
  onSubmit={async (e) => {
    e.preventDefault();

    await axios.post('http://localhost:5000/api/interact', {
      userId: newInteraction.userId,
      itemId: newInteraction.itemId,
    });

    alert('User interaction saved!');
    setNewInteraction({ userId: '', itemId: '' });
    fetchUsers(); // update available users
  }}
>
  <h3>Add User Interaction</h3>
  <input
    placeholder="User ID"
    value={newInteraction.userId}
    onChange={e => setNewInteraction({ ...newInteraction, userId: e.target.value })}>
  />
  <input
    placeholder="Movie ID"
    value={newInteraction.itemId}
    onChange={e => setNewInteraction({ ...newInteraction, itemId: e.target.value })}>
  />
  <button type="submit">Add Interaction</button>
</form>

<div className="control-panel">
  <div className="user-selector">
    <label htmlFor="userId">Select User:</label>
    <select
      id="userId"
      value={userId}
      onChange={e => setUserId(e.target.value)}>
```

```

>
  {userList.map(uid => (
    <option key={uid} value={uid}>
      {uid.startsWith('u') ? `User ${uid.substring(1)}` : uid}
    </option>
  )))
</select>
</div>
<button onClick={fetchRecommendations} disabled={loading}>
  {loading ? 'Loading...' : 'Get Recommendations'}
</button>
</div>

{correlationId && (
  <div className="info-box">
    <p><strong>Correlation ID:</strong> {correlationId}</p>
    <p><strong>From Cache:</strong> {fromCache ? '✓ Yes (60s TTL)' : '✗ Fresh from DB'}</p>
  </div>
)
}

<div className="recommendations-container">
  <h2>Recommended Movies for {userId}</h2>
  {loading ? (
    <p>Loading recommendations...</p>
  ) : recommendations.length > 0 ? (
    <div className="carousel">
      {[...new Map(recommendations.map(m => [m.itemId, m])).values()].map((movie) => (
        <div key={movie.itemId} className="movie-card">
          <div className="movie-id">{movie.itemId}</div>
          <h3>{movie.title}</h3>
          <div className="tags">
            {movie.tags.map(tag => (
              <span key={tag} className="tag">{tag}</span>
            )))
          </div>
          <div className="score">

```

```
        <strong>Score:</strong> {(parseFloat(movie.score) *  
        100).toFixed(1)}%  
    </div>  
    </div>  
))}  
</div>  
) : (  
    <p>No recommendations found</p>  
)}  
</div>  
</div>  
</div>  
);  
}
```

```
export default App;
```

- **src/index.js**

```
import 'bootstrap/dist/css/bootstrap.min.css';  
import React from 'react';  
import ReactDOM from 'react-dom/client';  
import './index.css';  
import App from './App';  
import reportWebVitals from './reportWebVitals';  
  
const root = ReactDOM.createRoot(document.getElementById('root'));  
root.render(  
    <App />,  
    document.getElementById('root')  
)
```

```
<React.StrictMode>
  <App />
</React.StrictMode>
);
reportWebVitals();
```

- **src/App.css**

```
* {
  margin: 0;
  padding: 0;
  box-sizing: border-box;
}

body {
  font-family: 'Segoe UI', Tahoma, Geneva, Verdana, sans-serif;
  background: linear-gradient(135deg, #667eea 0%, #764ba2 100%);
  min-height: 100vh;
}

.App {
  min-height: 100vh;
  padding: 20px;
}

.App-header {
  text-align: center;
  color: white;
  margin-bottom: 40px;
}

.App-header h1 {
  font-size: 2.5em;
  text-shadow: 2px 2px 4px rgba(0,0,0,0.3);
}

.container {
```

```
max-width: 1200px;
margin: 0 auto;
}

.control-panel {
background: white;
padding: 20px;
border-radius: 10px;
margin-bottom: 20px;
display: flex;
gap: 15px;
align-items: center;
box-shadow: 0 4px 6px rgba(0,0,0,0.1);
}

.user-selector {
display: flex;
gap: 10px;
align-items: center;
}

.user-selector label {
font-weight: bold;
}

.user-selector select {
padding: 8px 12px;
border: 2px solid #667eea;
border-radius: 5px;
font-size: 1em;
}

button {
padding: 10px 20px;
background: #667eea;
color: white;
border: none;
```

```
border-radius: 5px;  
cursor: pointer;  
font-size: 1em;  
font-weight: bold;  
transition: background 0.3s;  
}  
  
}
```

```
button:hover:not(:disabled) {  
background: #764ba2;  
}
```

```
button:disabled {  
opacity: 0.5;  
cursor: not-allowed;  
}
```

```
.info-box {  
background: rgba(255,255,255,0.9);  
padding: 15px;  
border-radius: 5px;  
margin-bottom: 20px;  
font-size: 0.9em;  
}
```

```
.info-box p {  
margin: 5px 0;  
}
```

```
.recommendations-container {  
background: white;  
padding: 30px;  
border-radius: 10px;  
box-shadow: 0 4px 6px rgba(0,0,0,0.1);  
}
```

```
.recommendations-container h2 {  
color: #667eea;
```

```
margin-bottom: 20px;  
font-size: 1.8em;  
}  
  
.carousel {  
display: grid;  
grid-template-columns: repeat(auto-fit, minmax(250px, 1fr));  
gap: 20px;  
}  
  
.movie-card {  
background: linear-gradient(135deg, #667eea 0%, #764ba2 100%);  
color: white;  
padding: 20px;  
border-radius: 10px;  
transition: transform 0.3s, box-shadow 0.3s;  
box-shadow: 0 4px 8px rgba(0,0,0,0.2);  
}  
  
.movie-card:hover {  
transform: translateY(-10px);  
box-shadow: 0 8px 16px rgba(0,0,0,0.3);  
}  
  
.movie-id {  
font-size: 0.8em;  
opacity: 0.8;  
margin-bottom: 8px;  
}  
  
.movie-card h3 {  
font-size: 1.4em;  
margin-bottom: 10px;  
}  
  
.tags {  
display: flex;
```

```
flex-wrap: wrap;  
gap: 8px;  
margin-bottom: 15px;  
}  
  
.tag {  
background: rgba(255,255,255,0.2);  
padding: 4px 8px;  
border-radius: 20px;  
font-size: 0.85em;  
}  
  
.score {  
background: rgba(255,255,255,0.2);  
padding: 10px;  
border-radius: 5px;  
text-align: center;  
font-size: 1.1em;  
}
```

Folder Structure

```
recommendation_engine/  
|  
|   └── backend/  
|       |   └── server.js  
|       |   └── package.json  
|       |   └── package-lock.json  
|       └── db/  
|           └── seed.js  
|  
|   └── modelservice/
```

```
|   ├── app.py
|   └── requirements.txt
|
└── frontend/
    ├── public/
    |   └── index.html
    ├── src/
    |   ├── App.js
    |   ├── index.js
    |   └── App.css
    ├── package.json
    └── package-lock.json
└── README.md
```

References

- MongoDB Documentation – <https://www.mongodb.com/docs>
- Node.js Documentation – <https://nodejs.org/en/docs/>
- React.js Documentation – <https://react.dev/>
- FastAPI Documentation - <https://fastapi.tiangolo.com/>

Bibliography

1. ByteXL Full Stack Development Capstone Project Guidelines
2. Various online tutorials on Node.js, MongoDB, and React.js