# **CSE3086– NoSQL Databases**

# J Component - Project Report

# **Review I**

# Book Recommendation using hybrid recommendation system

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M. Tech CSE Integrated with Business Analytics

Submitted to

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# **Worklet details**

Programme	M. Tech with Specialization	
Course Name / Code	NoSQL Databases CSE3086	
Slot	D1	
Faculty Name	Dr. A. Bhuvaneswari	
Digital Assignment		
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# **Team Members(s) Contributions – Tentatively planned for implementation:**

Worklet Tasks	Contributor's Names
Dataset Collection	Dazzle A J
Preprocessing	Dazzle A J
Architecture/ Model/ Flow diagram	Dazzle A J & Shashank Singh
ML Model Development & Deployment	Dazzle A J
App Development & Deployment	Shashank Singh
Database Integration	Shashank Singh
Model integration with App	Shashank Singh
Technical Report writing	Amal G
Presentation preparation	Amal G

## **ABSTRACT**

This project focuses on developing a hybrid recommendation model that integrates BERT (Bidirectional Encoder Representations from Transformers) with collaborative filtering to enhance book recommendations within a mobile application. By leveraging Firebase database for efficient data storage and retrieval, the system aims to provide personalized book suggestions tailored to individual user preferences and reading habits. The BERT model excels in understanding the contextual relationships within book descriptions, genres, and user reviews, allowing it to extract meaningful semantic features that enrich the recommendation process. In parallel, collaborative filtering analyzes user interactions, such as ratings and reading history, to identify patterns and similarities among users, further refining the recommendation accuracy. The implementation involves collecting user interaction data, training the BERT model on a curated dataset of bookrelated content, and applying collaborative filtering techniques to develop comprehensive user profiles. The hybrid recommendation engine will synthesize insights from both methodologies to generate real-time, personalized book recommendations displayed through a user-friendly mobile application. Expected outcomes include significantly improved recommendation relevance and accuracy, leading to enhanced user engagement and satisfaction. By combining advanced machine learning techniques with a robust backend infrastructure, this project represents a significant advancement in the realm of intelligent recommendation systems for books, ultimately aiming to enrich the reading experience and foster a deeper connection between users and literature.

#### 1. Introduction:

In the digital age, the vast availability of books across various genres presents both opportunities and challenges for readers seeking personalized recommendations. Traditional recommendation systems often struggle to provide relevant suggestions due to their reliance on simplistic algorithms that overlook the nuanced preferences of individual users. This project addresses the problem of delivering accurate and meaningful book recommendations by developing a hybrid recommendation model that combines the strengths of BERT (Bidirectional Encoder Representations from Transformers) and collaborative filtering. The primary objective is to create a system that not only understands the contextual relationships within book descriptions and user reviews but also leverages user interaction data to enhance recommendation accuracy.

One of the key challenges faced in this project is effectively integrating the two methodologies—BERT's advanced natural language processing capabilities and collaborative filtering's reliance on user behavior. This requires careful consideration of data preprocessing, model training, and the design of the recommendation engine to ensure that both components work harmoniously. Additionally, ensuring the system's scalability and responsiveness while managing a growing database of books and user interactions is crucial for maintaining a seamless user experience.

By utilizing a Firebase database for data storage and retrieval, the project aims to provide real-time recommendations that adapt to users' evolving preferences. Ultimately, this project seeks to transform the way readers discover books, fostering a deeper connection between users and literature while overcoming the limitations of traditional recommendation systems.

# 2. Dataset and Database specific Tool to be used (Details):

The dataset consists of a collection of books, formatted as a CSV file, which includes various attributes for each book. Each entry in the dataset is characterized by the following columns:

- bookID: A unique identifier for each book.
- title: The title of the book.
- authors: The names of the authors, which may include multiple authors for some titles.
- average rating: The average rating of the book, likely sourced from user reviews.
- isbn: The International Standard Book Number (ISBN) for the book, which is a unique identifier for books.
- isbn13: The 13-digit version of the ISBN.

- language\_code: The language in which the book is written, represented by a code (e.g., "eng" for English).
- num\_pages: The total number of pages in the book.
- ratings\_count: The total number of ratings the book has received.
- text\_reviews\_count: The number of text reviews submitted by readers.
- publication date: The date the book was published.
- publisher: The name of the publisher responsible for publishing the book.

This dataset encompasses a diverse range of books, including popular series like "Harry Potter" and works by renowned authors such as J.K. Rowling, Douglas Adams, and Bill Bryson. The information contained within this dataset is valuable for developing recommendation systems, as it provides insights into user preferences through ratings and reviews, as well as essential bibliographic details that can enhance the contextual understanding of the books.

We are using Google Firebase for our database storage.

## 3. Algorithms / Techniques description:

- **BERT:** It is a state-of-the-art natural language processing model developed by Google that utilizes a transformer architecture to understand the context of words in a sentence by considering both the left and right context simultaneously. This allows BERT to capture the nuances of language and perform well on various language understanding benchmarks. BERT can be fine-tuned on specific datasets, enabling it to adapt to different domains, including book recommendations, by analyzing textual data such as book descriptions and user reviews.
- Collaborative Filtering Technique: On the other hand, it is a popular recommendation technique that relies on user behavior and preferences to suggest items. It operates on the principle that users who agreed in the past will likely agree in the future. Collaborative filtering can be divided into two main types: user-based and item-based. User-based collaborative filtering identifies similarities between users based on their past interactions, while item-based collaborative filtering focuses on the relationships between items based on user ratings. This technique effectively captures the collective preferences of users, allowing for personalized recommendations

## • Pseudocode:

## Step 1: Initialize the system

- > Initialize Firebase database connection
- Load book dataset from CSV

Load user interaction data (ratings, reviews)

## Step 2: Preprocess data

- > For each book in dataset:
  - Clean and tokenize book descriptions.
  - Store cleaned data for BERT processing.
- For each user interaction:
  - o Create user-item interaction matrix.

#### Step 3: Train BERT model

- > Define BERT model architecture.
- > For each book description in cleaned data:
  - Train BERT model on book descriptions to obtain embeddings.

#### **Step 4: Train Collaborative Filtering model**

- > Define collaborative filtering model (user-based or item-based).
- > Train collaborative filtering model on user-item interaction matrix.

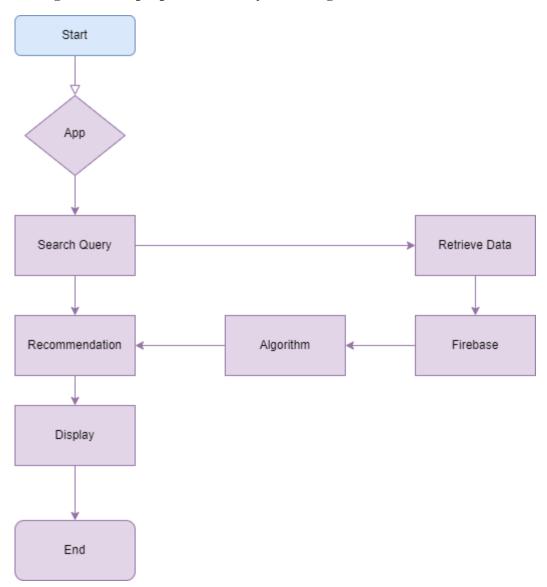
### **Step 5: Generate recommendations**

- Function getRecommendations(user\_id):
- // Get user profile from collaborative filtering
  - user\_profile = collaborative\_filtering\_model.predict(user\_id)
- // Get book embeddings from BERT
  - book\_embeddings = BERT\_model.get\_embeddings(book\_descriptions)
- // Combine collaborative filtering results with BERT embeddings
  - recommendations = []
  - For each book in book\_embeddings:
    - score = calculate\_combined\_score(user\_profile, book)
    - recommendations.append((book.title, score))
- // Sort recommendations by score
  - Sort recommendations by score in descending order
- // Return top N recommendations
  - return recommendations[:N]

#### Step 6: Display recommendations in mobile app

Display getRecommendations(current\_user\_id) in mobile app interface

4. Block Diagram of the proposed work / system design:



5. GitHub Repository Link (where your j comp project work can be seen for assessment)

https://github.com/AJDazzle/BookWise-Book-Recommendation-App-

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