A PROJECT REPORT ON

On

PRUS: Product Recommendation System Based on User Specification and Customer Reviews

Submitted in the partial fulfilment of requirements to

CS – 363 – Term Paper

By

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Goal of the Project

The research paper titled "PRUS: Product Recommender System Based on User Specifications and Customers Reviews" introduces a new and improved way to recommend products to users by understanding their specific needs more accurately. Most existing recommendation systems only focus on the overall sentiment of product reviews or highlight only the positive aspects. However, this approach can overlook important negative feedback that may affect a user's decision. To solve this problem, the proposed system, called PRUS, takes into account both positive and negative sentiments found in customer reviews.

One of the main features of PRUS is that it allows users to **specify exactly which product features they care about**, such as camera quality, battery life, or screen resolution. Instead of looking at the entire review as one opinion, the system breaks the reviews down into **individual sentences** and identifies opinions related to each feature. It then performs **aspect-level sentiment analysis**, which means it understands the user's opinion on each feature separately, rather than as a whole.

To rank the products effectively, the authors introduce a method called **RANK-ify**, which assigns scores to each product based on how often and how positively or negatively each feature is mentioned. This scoring method also gives users the flexibility to decide whether they want to prioritize positive feedback, negative feedback, or both equally.

Overall, the goal of the paper is to create a more **personalized, accurate,** and **useful product recommendation system** that helps users make better decisions by understanding detailed customer opinions, not just average ratings or general reviews.

Mechanism to be Followed

Here's the **step-by-step mechanism** of how the PRUS system works, explained in simple words:

Step 1: Collect Product Reviews

The system starts by collecting a large number of **customer reviews** from platforms like Amazon, focusing on products such as mobile phones.

Step 2: Clean and Prepare the Reviews

The reviews are **cleaned** by removing useless data like short or incomplete reviews, stopwords, punctuation, and numbers. Words are also simplified using techniques like **lemmatization** so that similar words are treated as the same.

Step 3: Break Reviews into Sentences

Each review is **split into sentences** to find specific opinions about different features (like battery, camera, etc.) rather than looking at the whole review.

Step 4: Extract Features and Sentiments

From each sentence, the system **identifies the product feature** being discussed (e.g., "battery") and the **sentiment** (positive, negative, or neutral) expressed about that feature using tools like **TextBlob**.

Step 5: Match Features with User Query

The user gives a **search query** specifying the features they care about (e.g., "good camera, long battery life"). The system matches these requested features with the features found in the reviews.

Step 6: Assign Scores to Features

Each product is assigned **scores** for each feature based on how many times that feature is mentioned positively or negatively. A formula is used to give more or less importance to positive or negative feedback, depending on what the user prefers.

Step 7: Rank Products (RANK-ify Algorithm)

The **RANK-ify algorithm** calculates a total score for each product by combining all feature scores. Products are then **ranked** from best to worst based on how well they match the user's needs.

Step 8: Display Recommended List

Finally, the system shows a **ranked list of recommended products** that best fit the user's specified features and preferences, considering both good and bad reviews.

Tools, and Algorithms needed for Implementation

Frontend (React.js)

Tool / Library	Purpose
React.js	Frontend JavaScript library to build UI
Axios or Fetch API	To send HTTP requests to backend
React Router	For page navigation (e.g., /products , /home)
Tailwind CSS / Bootstrap	UI styling and responsiveness
Vite / Create React App	For creating and managing the React app

Backend (Python)

Tool / Library	Purpose
FastAPI	REST API backend (highly recommended for modern APIs)
Flask (alt.)	Lightweight web framework (also suitable for REST APIs)
TextBlob / NLTK / spaCy	For sentiment and feature extraction in reviews
Pandas / NumPy	Data processing, managing review datasets
Uvicorn	ASGI server to run FastAPI
Pydantic	Request validation and type-checking in FastAPI
scikit-learn (optional)	For ML-based models if needed for ranking

Database (if needed)

Tool	Purpose
MongoDB	If you prefer NoSQL-style document storage

M Development Tools

Tool	Purpose
VS Code	Recommended IDE for both React and Python
Postman	To test API endpoints during backend development
Git + GitHub	Version control and code hosting
Node.js + npm	Required to run React.js and install packages
Python 3.9+	For backend development

Deployment Tools (Later Stage)

Tool	Purpose
Vercel / Netlify	To deploy React frontend
Render / Railway / Heroku	To deploy FastAPI/Flask backend

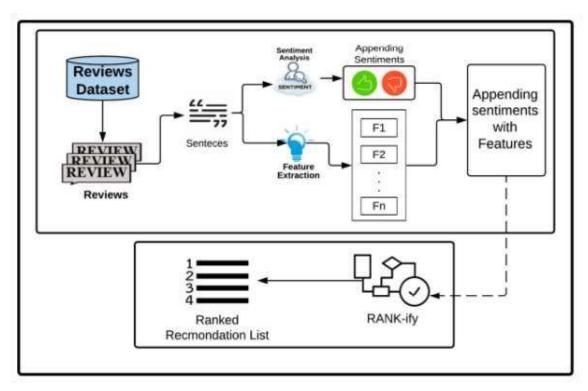
Algorithm 1 RANK-Ify Algorithm

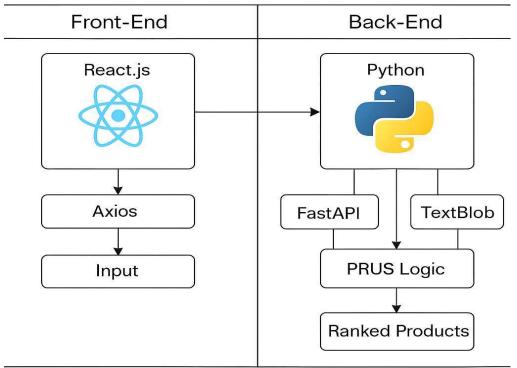
```
1: k = Number of Products in Dataset

 L<sub>p</sub> =List of Products in Dataset

 3: Q =Features in User Query
 4: for p ∈ L<sub>p</sub> do
      for U_f \in Q do
 5:
         if U_f \in p then
 6:
            FS(U_f) = FeaSco(U_f)
 7:
         end if
 8:
      end for
 9:
      RS(p) = \sum_{U_f \in Q} FS(U_f)
10:
11: end for
12: Sort Lp w.r.t. RS
13: Return top-k from L_p
```

Architecture Diagram



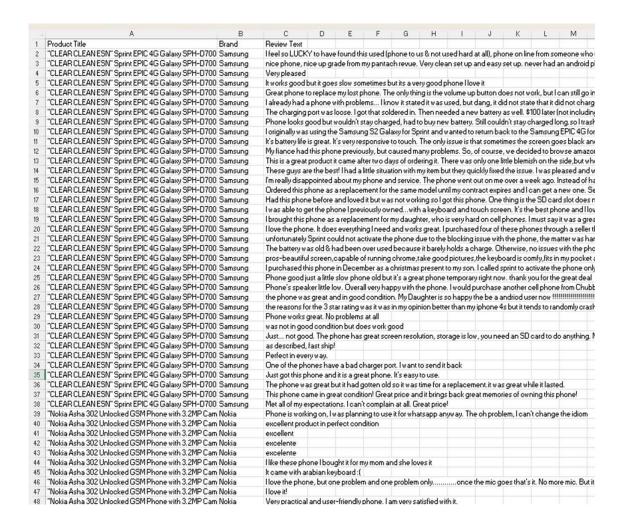


Architecture Overview

The architecture of the proposed product recommendation system is divided into two main components: the **frontend** developed using **React.js**, and the **backend** built with **Python** using frameworks like **FastAPI**. On the frontend, users interact with a clean and responsive interface built in React. They enter specific product preferences, such as "long battery life" or "high-resolution screen," through input fields or forms. Once the user submits their preferences, the React application uses a library called **Axios** to send an HTTP POST request to the backend API. This request contains the user-specified features and optionally assigned weights that indicate the importance of positive or negative sentiments. The React app is also responsible for displaying the final list of recommended products returned by the backend in a user-friendly format.

On the backend side, **FastAPI** receives the incoming request and forwards the data to the **core PRUS logic**, which is implemented in Python. The backend system uses **TextBlob** (or similar NLP tools) to process a large number of customer reviews, breaking them into sentences, extracting product features, and assigning sentiment polarity (positive or negative) to each feature. This information is then fed into the **RANK-ify algorithm**, which calculates a ranking score for each product based on how well it matches the user's query, considering both the frequency and sentiment strength of each feature. The products are sorted accordingly, and a top-N list is created. This ranked list is then sent back through the FastAPI response to the React frontend, where it is rendered for the user. This architecture ensures a modular, scalable, and highly personalized product recommendation experience based on actual customer feedback.

Data Set Description



- Name: Amazon Unlocked Mobile Review Dataset
- **Total Entries (Rows):** 413,840
- Total Columns (Features):3
- Source: Kaggle Data set
- Data set Link:

https://www.kaggle.com/datasets/PromptCloudHQ/amazon-reviewsunlocked-mobile-phones/data

Functional Requirements

These define what the system should do:

1. User Input of Preferences

Users should be able to enter specific product features (e.g., "battery life", "camera") via a web form.

2. Submit Query to Backend

The system should send the user preferences and sentiment weights (positive/negative) to the backend API using Axios.

3. Sentiment Analysis of Reviews

The backend should process reviews, perform sentiment analysis on each sentence, and identify feature-level sentiment using NLP tools like TextBlob.

4. Feature Matching and Scoring

The backend should match user-specified features with review data and calculate ranking scores using the RANK-ify algorithm.

5. Generate Ranked Product List

Based on user input and sentiment analysis, the system should return a list of ranked products.

6. Display Results on Frontend

The ranked list of recommended products should be displayed in a clear and user-friendly way (e.g., cards, list, table).

7. Handle Errors and Invalid Input

The system should validate input and gracefully handle missing data, API failures, or incorrect user queries.

Non-Functional Requirements

These define how the system should behave:

1. Performance

The system should return ranked results within a few seconds of receiving the user input.

2. Scalability

The backend should be designed to handle large datasets (e.g., thousands of reviews) and many users simultaneously.

3. Security

Input should be sanitized to prevent injection attacks. Secure communication (e.g., HTTPS) should be used in production.

4. Usability

The frontend should be intuitive, responsive (mobile-friendly), and easy to use for non-technical users.

5. Maintainability

The codebase should be modular and well-documented to allow future updates or improvements to the algorithm or UI.

6. Reliability

The system should remain available and correctly function even if one part (e.g., review sentiment) encounters errors.

7. Portability

The application should be deployable across various environments (local, cloud platforms like Vercel/Render).