

Product Recommendation System Using AI

Anuradha J. Shinde, Student, Computer Engineering, SKN Sinhgad Institute Technology And Science, Lonavala, India, anuradhashinde.sknsits.comp@gmail.com

Varsha Y. Devkar, Student, Computer Engineering, SKN Sinhgad Institute Technology And Science, Lonavala, India, varshadevkar.sknsits.comp@gmail.com

Nikita M. Patil, Student, Computer Engineering, SKN Sinhgad Institute Technology And Science, Lonavala, India, nikitapatil.sknsits.comp@gmail.com

Vaibhavi S. Bembade, Student, Computer Engineering, SKN Sinhgad Institute Technology And Science, Lonavala, India, vaibhavibembade.sknsits.comp@gmail.com

Ravishankar C. Bhaganagare, Professor, Computer Engineering, SKN Sinhgad Institute Technology And Science, Lonavala, India, rcbhaganagare.sknsits@sinhgad.edu

Abstract- In today's rapidly evolving e-commerce landscape, offering personalized product recommendations isn't just advantageous—it's essential. Our project introduces a dynamic and responsive recommendation system aimed at enhancing the online shopping experience, making it seamless and enjoyable. Consider the typical online shopping platforms, where users are inundated with a plethora of product choices. It's easy to feel overwhelmed and spend valuable time searching for the perfect item. This is where our project steps in, revolutionizing how customers discover products. Imagine a system that not only comprehends your preferences but also evolves with each interaction. Similar to a helpful in-store sales associate who remembers your tastes, our system acts as a virtual guide, adapting to your needs and suggesting products that align with your style and requirements. Whether you're browsing through product offerings, exploring detailed information, or proceeding through the checkout process, our project simplifies each step, making the entire journey more efficient and enjoyable. Just as a trusted friend might recommend products they know you'll adore, our system leverages your past interactions and preferences to offer personalized recommendations. This project injects a personalized touch into the online shopping realm, enriching the customer experience and increasing efficiency. As users interact with the system, it learns and adjusts to their individual preferences, ensuring a more gratifying shopping experience and facilitating the discovery of products that truly resonate.

Keywords: Recommendation Engine, Deep Learning (TensorFlow/Keras), Natural Language Processing (NLP)

I. INTRODUCTION

In today's digital age, e-commerce platforms have become integral to our daily lives, offering convenience and accessibility. However, the absence of personalized recommendations often leads to a lackluster shopping experience. This project endeavors to bridge this gap by developing a recommendation system that mimics the personalized assistance one would receive in a physical store. By leveraging AngularJS, Python, and cutting-edge algorithms, we aim to create an immersive and intuitive shopping experience that caters to individual preferences.

II. LITERATURE SURVEY

A. Deepa Sharma [1] This paper explores the impact of artificial intelligence (AI) on e-commerce in India. It

highlights how AI is playing a crucial role in advancing business operations, enhancing economic growth, and improving the standard of living. The paper emphasizes the importance of AI in meeting consumer needs efficiently and quickly through e-commerce platforms. It discusses how AI helps businesses analyse consumer behaviour, adapt to changing trends, and manufacture quality products. The positive impact of AI on e-commerce growth is underscored, particularly in terms of marketing strategies and customer satisfaction. The paper also touches on the foundations of AI and its influence on e-commerce, making it more accessible and convenient for consumers, including those with limited literacy skills.

B. Qian Zhang et al [3] In this paper they discuss the role of artificial intelligence (AI) in recommender systems. It

highlights the importance of personalized recommendations in addressing information overload and improving user experiences in the context of e-services. The paper reviews various AI techniques applied in recommender systems, including fuzzy techniques, transfer learning, neural networks, deep learning, active learning, natural language processing, computer vision, and evolutionary computing. It also outlines the core components of recommender systems and their main categories: content-based, collaborative filtering-based,. The paper provides valuable insights into the current state of AI-driven recommender systems and identifies research challenges and future directions in this field.

C. Mohamed Khoali et al [3] This paper, they provide a comprehensive overview of the application of deep learning techniques in e-commerce recommendation systems. They emphasize the significance of recommendation systems in enhancing user engagement and personalization in e-commerce platforms. The paper delves into various deep learning methods, including Multi-layer Perceptron, Auto-encoder, Convolutional Neural Network (CNN), Recurrent Neural Network (RNN), Restricted Boltzmann Machine, Attentional Models, and others, highlighting their utility in handling non-linear. The authors acknowledge the benefits of deep learning in recommendation systems, such as improved modelling capabilities and automatic feature extraction, while also addressing limitations, including data requirements and model interpretability. This paper offers valuable insights into the integration of deep learning into recommendation systems, fostering advancements in the field.

D. Ransome Epie Bawack et al [4] In this paper they conduct a comprehensive study on the intersection of artificial intelligence (AI) and e-commerce. The study combines bibliometric analysis with an extensive literature review to synthesize existing research and propose future directions. The authors find that AI in ecommerce primarily revolves around recommender systems, with core research themes including sentiment analysis, trust, personalization, and optimization. China-based institutions emerge as leaders in this research area, and most research papers are published in computer science, AI, business, and management outlets. This study provides valuable insights for researchers and practitioners interested in leveraging AI to enhance ecommerce endeavours, offering guidance for future research in the field.

III. PROBLEM STATEMENT

The proliferation of choices in e-commerce often overwhelms customers, hindering their ability to find products that resonate with their preferences. Existing platforms struggle to deliver personalized recommendations, leading to reduced customer satisfaction and business performance. This project aims to develop a robust recommendation system that utilizes data-driven insights

and advanced technologies to offer tailored product suggestions, thereby enhancing the online shopping experience.

Problem Description

A. Objectives:

- 1) Develop a robust recommendation system for personalized product suggestions.
- 2) Utilize natural language processing (NLP) for understanding user queries.
- 3) Implement deep learning techniques to enhance recommendation accuracy.
- 4) Create an interactive user interface with essential features.
- 5) Integrate a chatbot for user interaction and query handling.
- 6) Continuously improve the recommendation system based on user feedback.
- 7) Monitor and maintain system performance and accuracy.
- 8) Enhance the online shopping experience through tailored recommendations.
- 9) Increase customer satisfaction and business growth through improved recommendations.

B. Components used:

- 1) Natural Language Processing (NLP)
- 2) Deep Learning (TensorFlow/Keras)
- 3) User Interface (AngularJS)
- 4) Chatbot Integration
- 5) Data Preprocessing
- 6) User Feedback Mechanism
- 7) System Monitoring and Maintenance
- 8) Product Data (JSON/CSV)

IV. SYSTEM OVERVIEW AND DESIGN

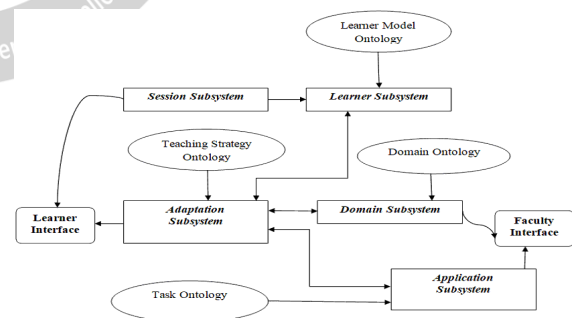


Figure 1 : Architecture Diagram

The educational system comprises several interconnected components and subsystems aimed at facilitating effective teaching and learning experiences. These include the Learner Model Ontology, which encapsulates formal representations of learning models and theories such as cognitive processes and instructional methods, and the Teaching Strategy Ontology, which encompasses knowledge about various teaching approaches and assessment methods.

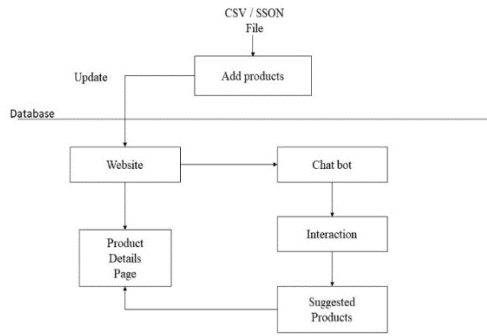


Figure 2 : DFD Level 0

In the product recommendation system, users interact with a chatbot to explore product options. When a user initiates a conversation with the chatbot, their queries are transmitted as data flows to the "Interact with Chatbot" process. The chatbot interprets these queries, potentially gathering additional information from the user to refine recommendations, and generates responses accordingly.

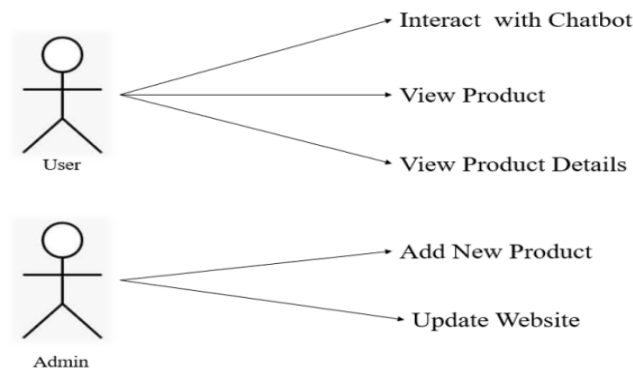


Figure 3 : DFD Level 1

A level 1 Data Flow Diagram (DFD) for a Product Recommendation System typically shows the main processes and data flows involved in the system. It might illustrate processes like user input, recommendation generation, and feedback collection. Data flows could include user preferences, product information, and recommendation outputs. This diagram provides a high-level overview of how data moves through the system, focusing on the interactions between users and the recommendation engine.

V. METHODOLOGY

A. Data Collection:

-Describe the sources of your product data, whether it's from JSON, CSV files, or APIs.

-Explain the preprocessing steps you took to clean and prepare the data for analysis. This might include removing duplicates, handling missing values, and standardizing formats.

B. Recommendation Model:

-Detail the choice between deep learning and matrix factorization techniques for your recommendation engine.

-If you chose deep learning, describe the architecture of your neural network (e.g., embeddings, layers).

-If you chose matrix factorization, explain the method used (e.g., Singular Value Decomposition, Alternating Least Squares).

-Discuss how you trained the model and any hyperparameters you tuned.

C. Natural Language Processing:

-Explain how you applied NLP techniques to analyze user queries and product descriptions.

-Describe any text preprocessing steps, such as tokenization, stop-word removal, and stemming or lemmatization.

-Detail the NLP models or algorithms you used, such as sentiment analysis, named entity recognition, or topic modeling.

D. User Interface:

-Provide an overview of the AngularJS-based interface you developed.

-Describe the features of the interface, including product listing, details, shopping cart, and checkout.

-Include screenshots or mock-ups to illustrate the user experience.

E. Chatbot Integration:

-Explain how you integrated a chatbot for user interactions and queries.

-Describe the functionality of the chatbot, including its ability to understand and respond to user queries.

-Detail the technologies or frameworks used to build the chatbot (e.g., Dialogflow, Rasa).

F. User Feedback:

-Discuss how you gathered user feedback to improve the recommendation system.

-Explain the methods used to collect feedback, such as surveys, interviews, or user analytics.

-Describe how you incorporated user feedback into the recommendation system, whether through retraining the model or adjusting parameters.

G. Monitoring and Maintenance:

-Outline your approach to monitoring and maintaining the system for performance and accuracy.

-Describe the metrics you used to evaluate system performance, such as precision, recall, and user engagement.

-Explain how you addressed any issues or challenges that arose during monitoring and maintenance.

VI. REQUIREMENT SPECIFICATION

1. User Requirements: Users expect a user-friendly interface and personalized recommendations.
2. Functional Requirements: Ensure smooth addition and updating of products, proficient chatbot responses, and accurate recommendations.

3. Non-Functional Requirements: Maintain minimum accuracy levels, instantaneous response times, scalability, and security.
4. Technical Requirements: Utilize AngularJS, Python, TensorFlow, Keras, and NLTK library.
5. Security and Privacy Requirements: Uphold strict data security and privacy standards.
6. Regulatory and Compliance Requirements: Comply with data protection laws and industry standards.
7. Budget and Resource Requirements: Allocate adequate resources for successful project completion.
8. Timeline and Milestone Requirements: Adhere to a defined timeline with key milestones.

VII. RESULTS

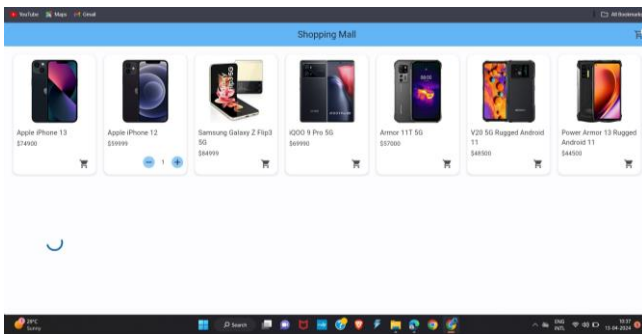


Figure 4 : User Interface of Shopping Mall

The user interface of our shopping mall recommendation system not only needs to be visually appealing but also intuitive and user-friendly, providing a seamless and engaging shopping experience. From browsing product recommendations to making purchases, every interaction within the UI should be designed with the user's needs and preferences in mind. This includes considerations such as ease of navigation, efficient search functionality, personalized recommendations, and smooth checkout processes.

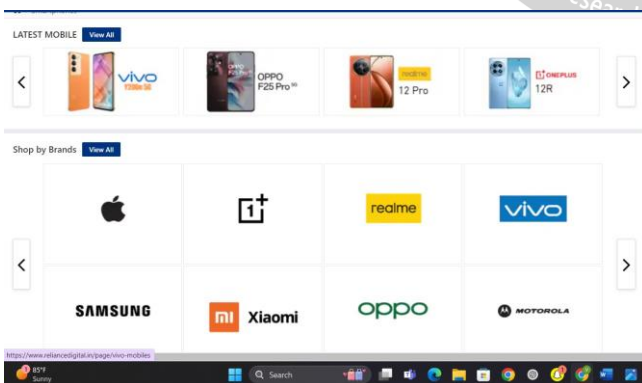


Figure 5 : Shop By Brand UI

This project aims to revolutionize the way users interact with mobile recommendation systems by introducing a specialized UI focused on brand preferences. Traditionally, recommendation systems have relied on generic algorithms to suggest products based on user behavior or item similarities. However, such approaches often overlook the significant role that brands play in shaping consumer

choices. By integrating a "Shop by Brands" UI, users can now explore products curated specifically from their favorite brands, fostering a deeper connection and engagement with the platform.

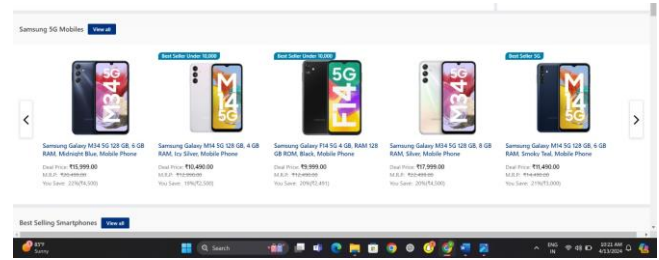


Figure 6 : Recommended Product List

In the realm of mobile applications, recommendation systems play a vital role in helping users navigate through a plethora of options. At the heart of these systems lies the user interface (UI) for presenting recommended products. This project focuses on optimizing the UI for recommended products within a mobile recommendation system, aiming to enhance user satisfaction, engagement, and conversion rates. By combining user-centered design principles with iterative testing, we strive to create an intuitive interface that delivers personalized recommendations effectively.

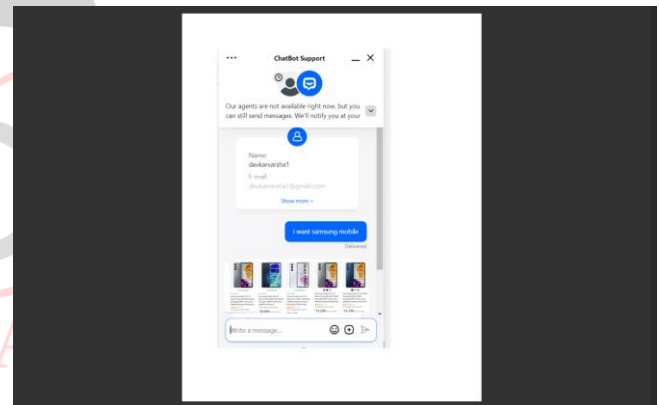


Figure 7 : Chatbot For System

"In today's mobile-centric world, navigating the vast array of available apps can be overwhelming. To address this, integrating a chatbot UI into mobile recommendation systems offers a solution. By leveraging natural language processing and machine learning, chatbots provide users with a conversational interface to receive personalized app suggestions, enhancing user experience and facilitating informed decision-making."

VIII. CONCLUSION

In addition to its technical capabilities, our experience with Python in developing the product recommendation system has highlighted its broader advantages and implications. Python's extensive ecosystem of libraries and frameworks not only facilitated the implementation of complex algorithms but also encouraged collaboration and knowledge sharing within our development team. Its clear and concise syntax enabled rapid prototyping and iterative

development, allowing us to quickly adapt to changing requirements and user feedback.

Moreover, Python's popularity and community support have been invaluable assets throughout the project lifecycle. Access to a vast array of resources, including documentation, tutorials, and online forums, empowered us to overcome challenges and explore innovative solutions effectively. The active Python community also provided opportunities for networking, learning, and staying abreast of industry trends and best practices, enriching our development experience and fostering continuous improvement.

Beyond its technical merits, Python's versatility and cross-disciplinary applicability have profound implications for the future of recommendation systems and data-driven decision-making. By harnessing Python's capabilities, we've not only built a powerful recommendation engine but also laid the foundation for integrating our system with other domains such as e-commerce, content personalization, and targeted advertising. This interdisciplinary approach opens up new avenues for leveraging data to enhance user experiences, drive business growth, and address societal challenges.

IX. ACKNOWLEDGMENT

This project aims to revolutionize the online shopping experience by providing personalized recommendations and enhancing user engagement. By leveraging advanced technologies and user feedback, we strive to create a dynamic platform that fosters customer satisfaction and drives business growth.

X. REFERENCES

- [1] DEEPA SHARMA," IMPACT OF AI ON E-COMMERCE", Artificial Intelligence In Finance: Trends And Applications (pp.1-28), 2021, <http://dx.doi.org/10.1201/9781003129639-1>.
- [2] Qian Zhang, Jie Lu and Yaochu Jin," Artificial intelligence in recommender systems", Complex & Intelligent Systems (2021) 7:439-457 <https://doi.org/10.1007/s40747-020-00212-w>.
- [3] Mohamed KHOALI, Abdelhak Tali and Yassin Laaziz," A survey of Artificial Intelligence based ECommerce Recommendation System", Springer international Publishing, 2021, http://dx.doi.org/10.1007/978-3-030-53440-0_12.
- [4] Ransome Epie Bawack, Samuel Fosso Wamba, Kevin Daniel André Carillo and Shahriar Akter," Artificial intelligence in E-Commerce: a bibliometric study and literature review", Electronic Markets (2022) 32:297-338.
- [5] Jatin Sharma, Kartikay Sharma, Kaustubh Garg and Avinash Kumar Sharma," Product Recommendation System a Comprehensive Review", IOP Conf. Series: Materials Science and Engineering 1022, 2021.
- [6] S. Zhang, L. Yao, A. Sun, and Y. Tay (2018). Deep Learning based Recommender System: A Survey and New Perspectives. ACM Computing Surveys, Vol.1, No.1
- [7] G. Adomavicius and A. Tuzhilin (2005). Toward the Next Generation of Recommender Systems: A Survey of the State-of-the-Art and Possible Extensions. IEEE Transactions on Knowledge and data engineering, vol. 17, no.6, pp. 734 - 749
- [8] I. Goodfellow, Y. Bengio, and A. Courville. (2016). Deep Learning. Chapter 6. 9. G.K. Dziugaite and D.M. Roy. (2015). Neural network matrix factorization. arXiv preprint arXiv:1511.06443.
- [9] Hu, Yong, Yifan Zhang, and Jun Wang. "Collaborative filtering with aspect-based opinion mining: A tensor factorization approach." In Proceedings of the 19th ACM SIGKDD International Conference on Knowledge Discovery and Data Mining, pp. 971-979. 2013. (This paper introduces a novel approach that combines collaborative filtering with aspect-based opinion mining to improve recommendation quality by capturing user preferences at a finer granularity.)