

# Recommendation system

## DATASET 2

### Group 96

Data-aholic

### Members:

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# INTRODUCTION ABOUT RECOMMENDATION SYSTEM



- Helps businesses understand customer needs and preferences
- Revolutionizes the shopping experience by leveraging advanced algorithms to analyze past choices, fashion trends, and individual user preferences
- Provide users with personalized recommendations based on their unique profiles
- Saves users time

# PROBLEM STATEMENT

## MVP - MINIMUM VIABLE PRODUCT

### ADVANTAGES

- Diversity of products and users
- Data loss
- Performance and speed
- Interaction between priority components
- User interaction

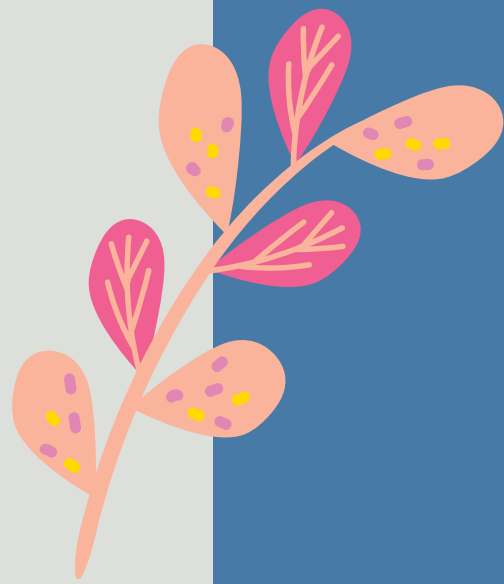
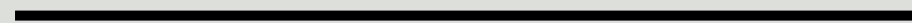
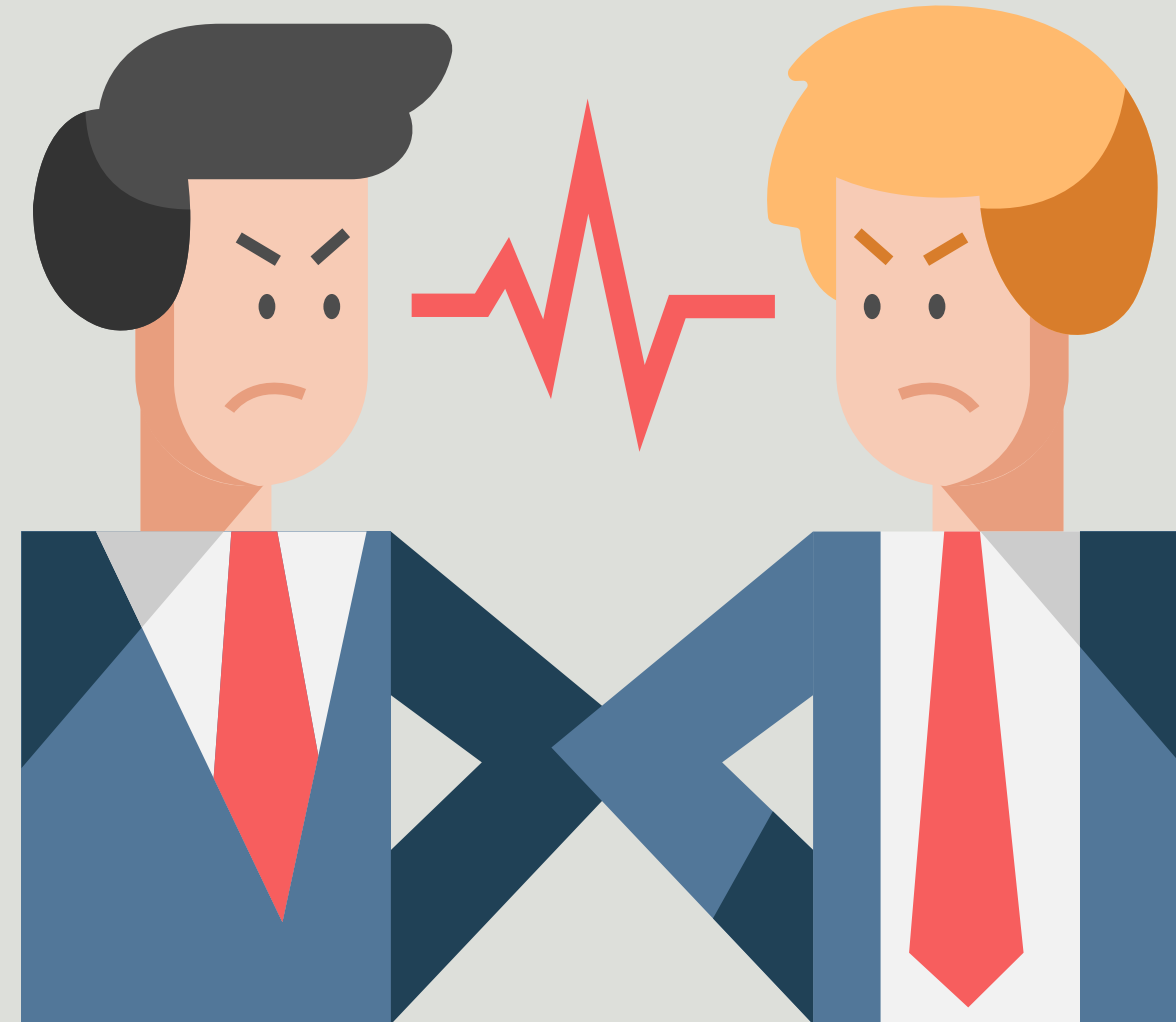


# PROBLEM STATEMENT

**MVP - MINIMUM VIABLE PRODUCT**

**PRESENT STATE AND COMPETITORS**

- Face competition from recommendation systems
- Minimize the impact of diversity
- Incomplete resolution of the Cold Start problem





# PROBLEM STATEMENT

**MVP - MINIMUM VIABLE PRODUCT**

## **LIMITATIONS**



- Low Performance Due to Data Loss
- Difficulty Handling Cold Start
- Slow Response Time
- Limited interpretability
- Depends on input data
- Limited understanding





# SOLUTIONS OVERVIEW

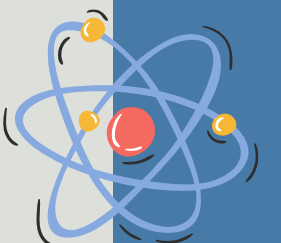
- Propose recommendation system
- Collaborative filtering assesses user behavior and preferences by comparing them to similar users, while content-based filtering focuses on the attributes of the products themselves.
- What sets our solution apart is its continuous learning and adaptability. The system automatically adjusts recommendations based on user interactions and feedback, ensuring a constantly evolving and personalized experience.



# METHODOLOGIES

## ARCHITECTURE OR MODEL STRUCTURE

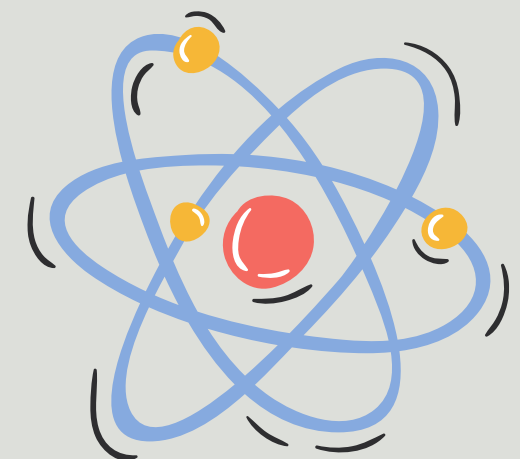
- **Convolutional Neural Networks (CNNs):**
  - **Goal:** Classify and process images of fashion products.
  - **Description:** CNN layers extract features from product images, recognizing patterns, colors, and styles.
- **Recurrent Neural Networks (RNNs):**
  - **Goal:** Process sequential data like user reviews and shopping history.
  - **Description:** RNN layers help the model understand temporal relationships and sequences in data, personalize shopping experiences.
- **Recommender System:**
  - **Goal:** Recommend products based on user shopping history and preferences.
  - **Description:** Use Collaborative Filtering to compare user shopping behavior with others, Content-Based Filtering to suggest similar products based on product features, and Real-time Recommendation to provide immediate suggestions based on user interactions.
- **Backend Server:**
  - **Goal:** Handle business logic, interact with the database, and provide APIs for frontend communication.
  - **Description:** The backend server manages frontend requests, interacts with the database for product and user information retrieval and storage.
- **Database:**
  - **Goal:** Store data related to products, users, shopping history, and reviews.
  - **Description:** Use MySQL, MongoDB to store and manage detailed information about products and users.



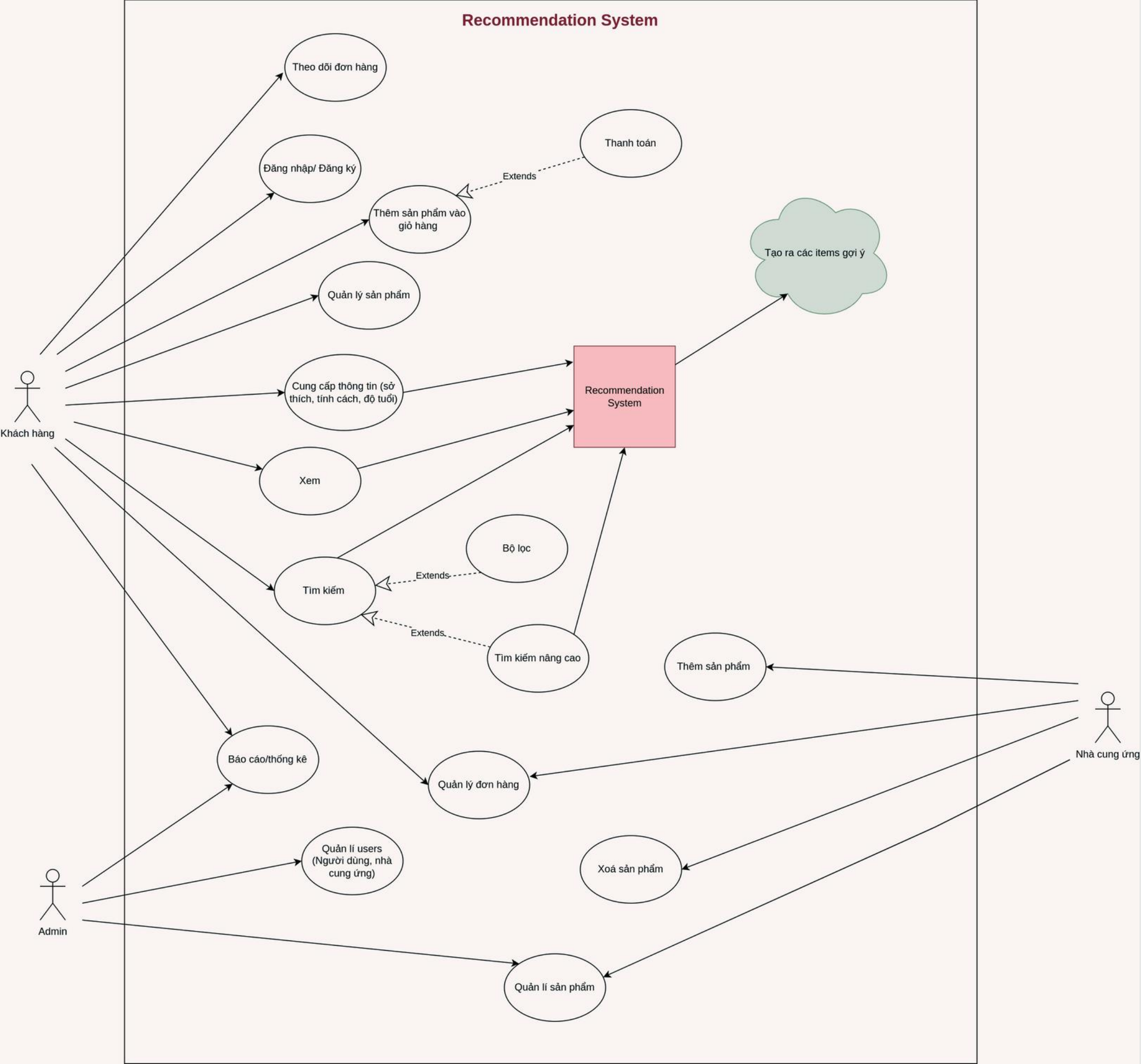
# METHODOLOGIES

## TECHNOLOGIES PLANNED FOR USE

- **Machine Learning Framework:** Utilize machine learning frameworks such as TensorFlow or PyTorch to build and train AI models.
- **Frontend Framework:** Employ popular frontend frameworks Reactjs to develop an interactive and user-friendly user interface.
- **Database:** Use a suitable database management system like MySQL or MongoDB for data storage and retrieval.
- **APIs and Protocols:** Implement RESTful API for seamless interaction between the frontend and backend.





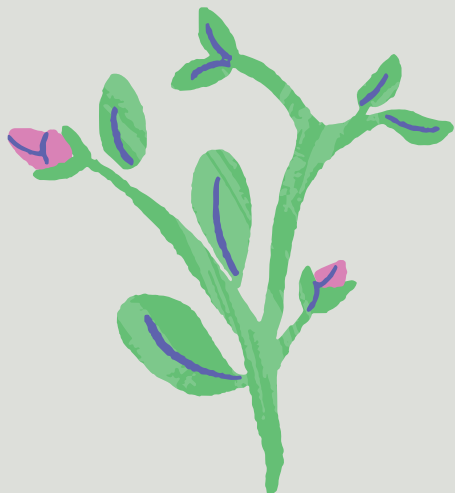


# USE CASES



# PERFORMANCE METRICS

- **Conversion Rate**
  - $\text{Number of transactions} / \text{Number of website visits} * 100.$
- **Retention Rate**
  - $(\text{Number of customers at the end of the period} - \text{Number of new customers acquired during the period}) / \text{Number of customers at the beginning of the period} * 100.$
- **Shopping Cart Abandonment Rate**
  - $(\text{Number of abandoned carts} / \text{Number of initiated transactions}) * 100.$
- **Website Traffic and Unique Visitors**
  - Number of visits, Number of unique visitors.
- **User Engagement**
  - Average session duration, average number of pages per session, click-through rate.





# TIMELINE AND ROADMAP

Phase	Mission	Task	Objectives	Timeline
1	Define Strategy and Research	Proposal	Implement ideas in the template	1 week
2	UI Design, Design and Development Planning	UI Design	UI design	5 days
3	Code, Report	Frontend, Backend, AI Model, Visualization	Build model AI combine with web e-commerce	2 weeks



# USER INTERFACE



- Link figma: <https://www.figma.com/file/VakLhldrKctbPeKRAtYADV/Untitled?type=design&node-id=1%3A164&mode=design&t=SzunJLviS9b9GC1V-1>





# **LIMITATIONS AND FUTURE ENHANCEMENTS**

## **Limitations:**

- Limited interpretability
- Depends on input data
- Limited understanding

## **Potential future enhancements:**

- Improvements we aim to implement in the future involve incorporating Long-Short Term Memory (LSTM) for predicting market trends. This entails analyzing historical data, market conditions, and seasonal factors to provide predictions for the next 2-3 months. This enhancement will play a crucial role for sales agents to apply informed strategies and make insightful decisions.
- Additionally, we plan to apply eXplainable AI (XAI) to elucidate why the model makes specific predictions over short periods (1 day, 7 days) or longer durations (1 month, 2 months, etc.). This transparency will enhance user trust and understanding, fostering a more reliable and interpretable AI system.

# CONCLUSION

## **MVP Objectives:**

- Analyze user preferences based on past purchases, browsing history, and explicit feedback.
- Generate relevant and personalized fashion recommendations to streamline decision-making.
- Improve user engagement and retention by offering an enjoyable and tailored shopping experience.
- Address challenges such as data loss, Cold Start problems, and slow response times.

*In summary, the MVP for the recommendation fashion system is designed to be a fundamental yet effective solution, focusing on user-centric improvements to revolutionize the online fashion shopping experience.*





