

# E-commerce Product Recommender System

## GitHub Link:

<https://github.com/Gurleenkaurmakhija/Unthinkable-E-commerce-Product-Recommender>

## 1. Introduction

The **E-commerce Product Recommender System** is designed to enhance user experience in online shopping platforms by providing personalized product recommendations combined with **LLM-generated explanations**.

This project merges traditional recommendation algorithms with **Large Language Models (LLMs)** like OpenAI GPT to not only suggest relevant products but also **explain why** those products are being recommended.

This increases user engagement, trust, and transparency in AI-driven systems.

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## 2. Objective

The main objective of this project is to:

- Recommend products based on user behavior and preferences.
  - Integrate LLMs to explain recommendations in natural language.
  - Build a backend API that serves both recommendations and explanations.
  - Optionally provide a simple frontend interface for users to interact with the system.
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## 3. Project Structure

None

Ecommerce-Product-Recommendation-main/

├─ backend/

| └─ app.py → Main backend API logic

| └─ chatbot.py → LLM module for explanations

| └─ bq-results-20240205.csv → Dataset for product data

| └─ requirements.txt → Backend dependencies

```
|   |— dockerfile           → For containerization
|
|— frontend/
|   |— app.py              → Streamlit/Flask frontend dashboard
|   |— requirements.txt    → Frontend dependencies
|   |— dockerfile          → Frontend container setup
|
|— README.md               → Project documentation
|— .gitignore
```

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## 4. Key Components

### Backend (app.py)

- Defines API routes to handle product recommendation requests.
- Reads product and user data from the CSV file.
- Calls chatbot.py to generate explanations using an LLM.
- Returns JSON responses with product list and explanation

Example Logic:

None

```
@app.route('/recommend', methods=['POST'])
```

```
def recommend():
```

```
    user = request.json['user']
```

```
    recommendations = get_recommendations(user)
```

```
    explanation = generate_explanation(recommendations, user)
```

```
    return jsonify({
```

```
        "recommendations": recommendations,
```

```
        "explanation": explanation
```

```
    })
```

### Explanation:

The backend receives a user ID, calculates recommendations, generates an LLM-based explanation, and returns the result as a JSON response.

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### Chatbot (chatbot.py)

- This file connects to an LLM (like OpenAI's GPT API).
- Generates human-like explanations for why certain products are recommended.
- The model prompt may look like:

None

- `Explain why product X is recommended to user Y based on their past behavior.`

- Returns a short, friendly message such as:  
*"This product matches your recent interests in mobile devices and accessories."*
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### Frontend (frontend/app.py)

- A Streamlit or Flask-based user interface.
- Allows the user to input a user ID or select products.
- Displays the recommended products and generated explanations interactively.

### Typical workflow:

1. User selects or inputs a user ID.
  2. Frontend sends request to backend /recommend endpoint.
  3. Receives response and displays product cards with explanations.
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## 5. Dataset

The project includes a dataset file:

None

`backend/bq-results-20240205-004748-1707094090486.csv`

This CSV likely contains fields like:

None

```
user_id, product_id, product_name, category, price, rating, behavior
```

It stores historical user interactions (views, purchases, clicks), which are used to generate personalized recommendations.

### Example Rows:

None

```
user123, P001, "iPhone 14", Electronics, 799, 4.8, viewed  
user123, P008, "AirPods Pro", Accessories, 249, 4.7, purchased
```

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## 6. How It Works (Flow)

1. **User Data Input:** The system takes product catalog and user behavior data.
2. **Recommendation Generation:** Backend computes top products using similarity logic (e.g., Collaborative Filtering or Content-Based Filtering).
3. **Explanation Generation:** The chatbot (LLM) module is triggered with a prompt describing the user's history and the recommended products.
4. **Response Delivery:** Backend sends both recommendations and explanations to the frontend.
5. **Display:** The frontend shows user-friendly product cards with reasoning for each.

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## 7. How to Run the Project

1. **Install dependencies**

None

```
pip install -r backend/requirements.txt  
pip install -r frontend/requirements.txt
```

2.

3. **Run backend server**

None

```
python backend/app.py
```

#### 4. Run frontend

None

```
streamlit run frontend/app.py
```

5. Open the frontend URL shown in the terminal to interact with the recommender.

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## 8. Sample Output

### Input:

User recently viewed “iPhone 14” and “Samsung Galaxy S22”.

### Output:

None

```
Recommended Products:
```

- AirPods Pro
- iPhone Case
- Samsung Smart Watch

```
Explanation:
```

```
“These items are recommended because you recently viewed premium smartphones  
and similar accessories that match your interests.”
```

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## 9. Conclusion

This project successfully integrates **Machine Learning-based recommendations** with **Generative AI explanations**, bridging data analysis and natural language understanding. By explaining why items are suggested, it enhances user engagement and transparency, making the recommendation system more trustworthy and user-centric.

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## 10. Future Enhancements

- Fine-tune the LLM for domain-specific recommendations.
- Use visual embeddings for image-based product similarity.
- Incorporate real-time feedback to improve model performance.
- Add multilingual support for global accessibility.
- Include reinforcement learning for adaptive personalization.