

Official Documentation

AI Shopping Assistant

CICD / LLMOps

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AI Shopping Assistant

A full-stack, LLM-powered laptop recommendation system that combines intelligent data ingestion pipelines with an interactive chatbot workflow. The system helps users find personalised laptop recommendations based on their preferences through natural language conversations.

Project Overview

Al Shopping Assistant is a production-ready system that demonstrates advanced Al integration in e-commerce applications. The project features two distinct but interconnected workflows:

- 1. **Data Ingestion Pipeline**: An admin-triggered, automated workflow for processing and mapping laptop datasets
- 2. **Chatbot Interface**: An intelligent conversational agent that provides personalised laptop recommendations

The system is deployed on AWS Lambda and HuggingFace Spaces, integrating with Amazon S3 for storage, PostgreSQL (AIVEN) and Cloudflare D1 for data persistence, and OpenAI's GPT-40-mini models for natural language processing.

- Live Demo: Al Shopping Assistant
- Source Code: GitHub Repository
- Home Page: www.aydie.in

Tech Stack

Backend Framework

• Flask 3.0.3 - Main web framework (entry point: app.py)

Databases

- AIVEN PostgreSQL Primary storage for mapped laptop data
- Cloudflare D1 SQL Alternative serverless database option
- **SQLAIchemy 2.0.36** ORM for database abstraction
- psycopg2 2.9.10 PostgreSQL client

Cloud & Storage

- Amazon S3 Raw CSV file storage
- AWS Lambda Serverless backend deployment
- HuggingFace Spaces Application hosting
- boto3 AWS SDK



AI/LLM APIs

- OpenAl API 1.102.0 Primary language model (GPT-4o-mini)
- Google Generative Al 0.7.2 Optional Gemini integration
- LangChain 0.2.16 LLM orchestration
- LlamaIndex 0.12.38 Retrieval and indexing

Data Processing

- Pandas 2.2.2 Data manipulation
- NumPy 1.26.4 Numerical computing
- PandasSQL 0.7.3 SQL query engine for pandas
- PyArrow 17.0.0 Parquet file handling
- FastParquet 2024.5.0 High-performance parquet operations

Deployment & Production

- Gunicorn 22.0.0 WSGI server
- Mangum 0.19.0 AWS Lambda adapter

Architecture & Workflow

The system implements two distinct pipelines that share core functionality through the orchestrator.py module:

1. Data Ingestion Pipeline (Admin Workflow)

Purpose: Maintain a fresh, pre-processed laptop recommendation database

Trigger Methods:

- Web interface via ingestor.py with file upload through upload.html
- Direct function call via Orchestrator.start_internal_data_ingestion()

Process Flow:

CSV Upload → S3 Storage → Data Retrieval → LLM Mapping → Database Update

Key Components:

- 1. **File Upload** (DataIngestion.start_data_ingestion())
 - Accepts CSV files containing laptop specifications
 - Uploads to Amazon S3 bucket via AWSConnection.upload_file_to_s3()
 - Real-time progress tracking through /upload-status endpoint
- 2. Data Processing (ProductMapper.start_dataframe_product_mapping())
 - Reads structured data using parquet files for efficiency
 - Processes laptop descriptions through OpenAI GPT-4o-mini API
 - Handles batch processing with error recovery
- 3. **Specification Mapping** (ProductMapper.do_product_mapping())
 - Maps laptop features to standardized categories using OpenAI GPT models:



- GPU Intensity: Low (integrated graphics) / Medium (mid-range dedicated) / High (high-end dedicated like RTX)
 - Display Quality: Low (below Full HD) / Medium (Full HD+) / High (4K, HDR)
 - Portability: High (<1.51kg) / Medium (1.51-2.51kg) / Low (>2.51kg)
 - Multitasking: Low (8-12GB RAM) / Medium (16GB RAM) / High (32GB+ RAM)
 - Processing Speed: Low (i3, Ryzen 3) / Medium (i5, Ryzen 5) / High (i7+, Ryzen 7+)
- 4. Database Update (PostgresDataBaseUpdate.update_to_postgres_database())
 - Supports both PostgreSQL (AIVEN) and Cloudflare D1 databases
 - Drops existing table for complete refresh
 - Creates new schema based on DataFrame structure
 - Batch inserts all processed data with JSON handling for mapped columns

2. Chatbot Workflow (User Interaction)

Purpose: Provide personalised laptop recommendations through conversational Al

Conversation States:

- normal Standard conversation flow
- awaiting_feedback Waiting for user satisfaction confirmation
- awaiting_rating Collecting user rating (1-5 scale)
- ended Conversation terminated

Process Flow:

User Input → Moderation → Profile Building → Intent Confirmation → Recommendation → Feedback

Key Functions:

- 1. Input Processing (/chat endpoint in app.py)
 - Greeting detection via get_chat_completion() with OpenAI GPT-4o-mini
 - Content moderation using moderation_check() with omni-moderation-latest
 - JSON filtering through filter_json_from_response()
- 2. **Profile Building** (Orchestrator.get_chat_completion())
 - Extracts user preferences through guided conversation
 - Uses system instruction prompts from SystemInstruction.system_instruction
 - Builds a comprehensive user profile dictionary: python

```
{ 'GPU intensity': 'low/medium/high',
  'Display quality': 'low/medium/high',
  'Portability': 'low/medium/high',
  'Multitasking': 'low/medium/high',
  'Processing speed': 'low/medium/high',
```



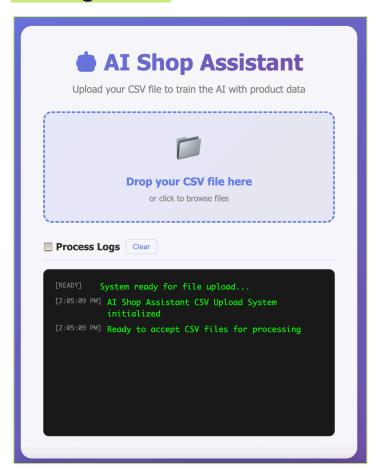
```
'Budget': 'numerical_value'
}
```

- 3. **Intent Confirmation** (Orchestrator.intent_confirmation_check())
 - Validates profile completeness using IntentConfirmation.intent_confirmation prompt
 - Ensures all 6 required attributes are captured
 - Returns JSON response with confirmation status
- **4. Recommendation Engine** (ProductRecommendation.recommend_product())
 - Queries database via LoadFromDatabase.fetch_query_engine_data()
 - Filters by budget using QueryEngine.filter_budget()
 - Calculates match scores via QueryEngine.filter_by_user_score() and ProductMapper.map_the_score()
 - Returns top 3 recommendations with detailed specifications

5. Response Generation

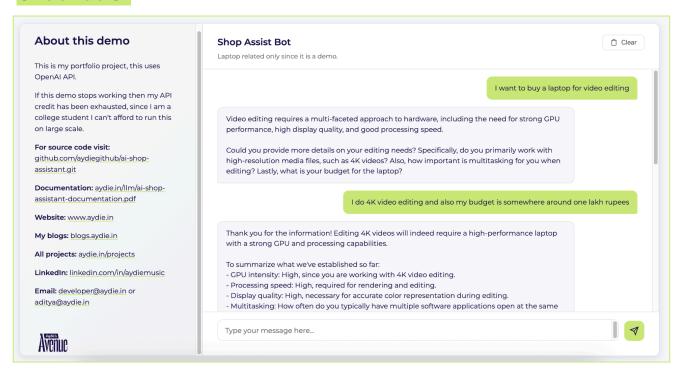
- Formats recommendations using ProductRecommender.system_message prompt
- Handles satisfaction feedback through /feedback endpoint
- Routes to human agent via route_to_human_agent() if user is unsatisfied
- Collects ratings through /rate endpoint for continuous improvement

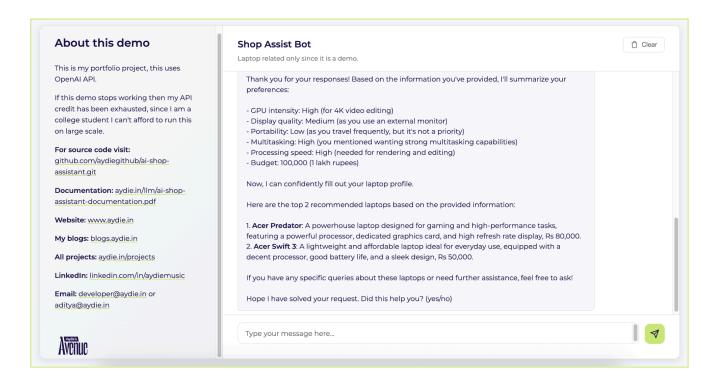
Data Ingestor UI





ChatBot UI

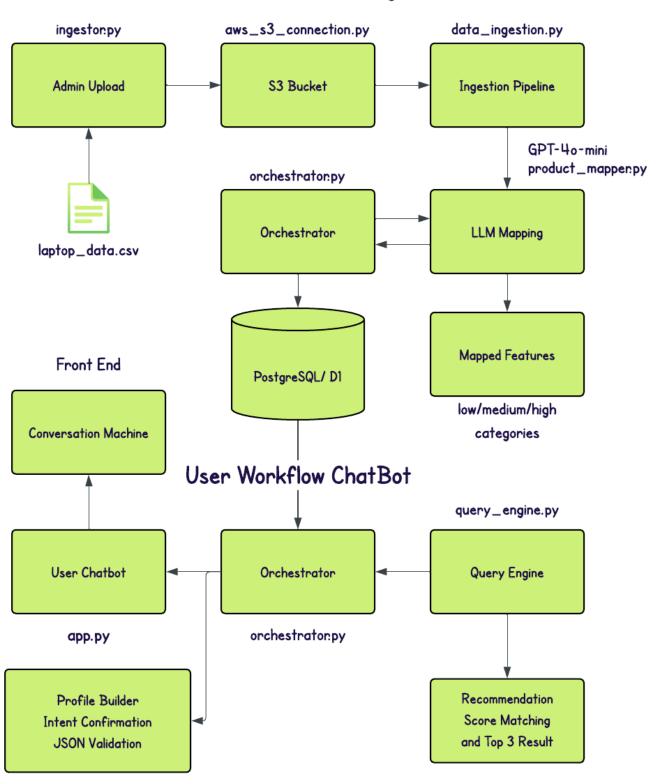






Data Flow Diagram

Admin Workflow Data Ingestion





Setup Instructions

Prerequisites

- Python 3.8+
- PostgreSQL database (AIVEN recommended) or Cloudflare D1
- AWS S3 bucket with appropriate permissions
- OpenAl API key with GPT-4o-mini access

Installation

1. Clone the repository

```
git clone https://github.com/aydiegithub/ai-shopping-assistant.git
```

2. Install dependencies

```
pip install -r requirements.txt
```

3. Environment Configuration

Create a .env file with the following variables:

```
# OpenAI Configuration
OPENAI API KEY=your openai api key
# AWS Configuration
AWS ACCESS KEY ID=your aws access key
AWS SECRET ACCESS KEY=your aws secret key
S3 BUCKET NAME=your s3 bucket name
AWS DEFAULT REGION=your region
# PostgreSQL Configuration (AIVEN)
POSTGRES HOST=your postgres host
POSTGRES PORT=your port number
POSTGRES DB NAME=your database name
POSTGRES USER=your postgres user
POSTGRES PASSWORD=your postgres password
POSTGRES TABLE NAME=table name
# Cloudflare D1 Configuration (Optional)
CLOUDFLARE ACCOUNT ID=your cloudflare account id
D1 SQL DATABASE ID=your d1 database id
CLOUDFLARE API TOKEN=your cloudflare api token
D1 SQL DATABASE NAME=your d1 database name
```



4. Database Setup

5. Run the application

```
python app.py  # UI for end user

python ingestor.py  # UI for admin for data ingestion
gunicorn -w 4 -b 0.0.0.0:8000 app:app  # Production server
```

Deployment

AWS Lambda Deployment:

The application includes the Mangum WSGI adapter for seamless Lambda deployment:

```
from mangum import Mangum
handler = Mangum(app)
```

HuggingFace Spaces:

Configure app.py with the appropriate port settings:

```
port = int(os.environ.get("PORT", 7860))
app.run(host='0.0.0.0', port=port, debug=True)
```

Usage

Data Ingestion (Admin)

1. Web Interface:

bash python ingestor.py # Navigate to http://localhost:5000 # Upload CSV files through the interface # Monitor progress via real-time status updates

2. Programmatic Access:

```
python from src.backend.orchestrator import Orchestrator orch = Orchestrator() orch.start_internal_data_ingestion( local_file_path="src/database/laptop_data.csv", s3_file_name="laptop_data.csv" )
```

Chatbot Interaction (End User)

- 1. Start Conversation: Send greeting to /chat endpoint
- 2. **Profile Building**: Answer guided questions about laptop needs
- 3. Receive Recommendations: Get personalized top 3 suggestions
- 4. Provide Feedback: Rate assistance (1-5) and optionally escalate to human agent



Example Conversation Flow

```
POST /chat
  "message": "Hi, I need help finding a laptop",
  "messages": [],
 "state": "normal"
}
Response:
  "message": "Hello there! I am here to help you. I am your personal laptop
assistant. What kind of laptop are you looking for?",
  "messages": [...],
  "state": "normal"
}
POST /chat
  "message": "I'm a video editor working with 4K content in After Effects",
  "messages": [...],
 "state": "normal"
}
Response:
  "message": "Great! Working with After Effects and 4K content requires
high GPU intensity and processing power. Do you need portability for travel
work?",
 "messages": [...],
 "state": "normal"
}
POST /chat
  "message": "Mostly stationary work, budget around 150000 INR",
  "messages": [...],
 "state": "normal"
Response:
   "message": "{'GPU intensity': 'high', 'Display quality': 'high',
'Portability': 'low', 'Multitasking': 'high', 'Processing speed': 'high',
'Budget': '150000'}\n\nBased on
                                     your requirements for 4K video
editing:\n\n1. **ASUS ROG Strix G15**: High-performance gaming laptop with
```



```
RTX 3070, 16GB RAM, AMD Ryzen 7, 15.6\" 144Hz display - Rs. 145,000\n2.

**MSI Creator 15**: Content creation focused with RTX 3060, 32GB RAM, Intel
i7, 15.6\" 4K display - Rs. 149,999\n3. **HP OMEN 15**: Gaming laptop with
RTX 3060, 16GB RAM, AMD Ryzen 7, 15.6\" 165Hz display - Rs. 139,000\n\nHope
I have solved your request. Did this help you? (yes/no)",

"messages": [...],

"state": "awaiting_feedback"
}
```

Dependencies

Core Web Framework

flask==3.0.3 python-dotenv==1.0.1

Database & ORM

SQLAlchemy==2.0.36 sqlalchemy-cloudflare-d1==0.1.0 psycopg2==2.9.10

Data Handling

pandas==2.2.2 numpy==1.26.4 pandasql==0.7.3 pyarrow==17.0.0 fastparquet==2024.5.0

LLM/AI APIs

openai==1.102.0 google-generativeai==0.7.2 llama-index==0.12.38 langchain==0.2.16 langchain-community==0.2.16 tenacity

Cloud & AWS/Cloudflare Libraries

boto3>=1.26.0,<2 cloudflare>=3.0.0,<4 mangum==0.19.0



HTTP Requests & Production

requests==2.32.3 gunicorn==22.0.0

Data Validation

pydantic>=2.7.4,<3 pydantic-core>=2.16.3,<3

Development Tools

black==24.8.0 pytest==8.3.3

API Endpoints

Chatbot Endpoints (app.py)

- GET / Main chatbot interface
 - Serves index.html template
 - Returns: HTML chatbot interface
- POST /chat Main conversation endpoint

```
Body: {"message": "user_input", "messages": [], "state": "normal"}
Returns: {"message": "bot_response", "messages": [...], "state":
"normal|awaiting_feedback"}
```

- Functions: Greeting detection, moderation, profile building, intent confirmation
- POST /feedback User satisfaction feedback

```
-Body: {"message": "yes/no", "messages": [...]}
```

- Returns: Rating prompt or human agent escalation
- Functions: Satisfaction assessment, routing logic
- POST /rate User rating submission (1-5 scale)

```
-Body: {"message": "1-5", "messages": [...]}
- Returns: {"message": "Thank you for your feedback! Chat ended.",
"state": "ended"}
```



Data Ingestion Endpoints (ingestor.py)

- GET / Upload interface
 - Serves upload.html template
 - Returns: File upload interface with progress tracking
- POST /upload-csv CSV file upload and processing
 - Body: multipart/form-data with file
 - Returns: { "message": "Upload started", "status": "processing"}
 - Functions: File validation, S3 upload, LLM mapping, database update
- GET /upload-status Real-time upload progress

```
- Returns: {"status": "uploading|processing|completed|error", "progress": 0-100, "message": "...", "logs": [...]}
```

• GET /logs - Recent processing logs

```
- Returns: {"logs": [{"timestamp": "HH:MM:SS", "message": "...", "level": "info|error"}]}
```

Core Functions Reference

Orchestrator (src/backend/orchestrator.py)

- initialise_conversation() Sets up system instruction for new chats
- get_chat_completion(input_messages, json_format=False) OpenAl API wrapper with retry logic
- moderation_check(input_message) Content safety validation using omni-moderation-latest
- intent_confirmation_check(input_message) Profile completeness validation
- start_product_recommendation(input_message) Triggers recommendation pipeline
- set_user_profile(message) Extracts user preferences from conversation
- route_to_human_agent(input_message) Escalation logic for unsatisfied users

Product Mapper (src/backend/product_mapper.py)

- do_product_mapping(laptop_description) Maps a single laptop description to categories using GPT-4o-mini
- start_dataframe_product_mapping(df) Batch processes entire DataFrame with parquet optimisation
- map_the_score(mapped_column, user_profile) Calculates compatibility scores



Query Engine (src/backend/query_engine.py)

- filter_budget(data, criteria) Filters laptops by budget constraints
- filter_by_user_score(data, user_profile) Ranks laptops by compatibility score

Product Recommendation (src/backend/product_recommender.py)

- recommend_product(user_profile) End-to-end recommendation pipeline
- calculate_score(mapping_column, user_profile) Score calculation wrapper

Contributing

- 1. Fork the repository
- 2. Create a feature branch (git checkout -b feature/AmazingFeature)
- 3. Commit your changes (git commit -m 'Add some AmazingFeature')
- 4. Push to the branch (git push origin feature/AmazingFeature)
- 5. Open a Pull Request

Development Guidelines

- Follow PEP 8 style guidelines (use black==24.8.0 for formatting)
- Add comprehensive logging using the project's logging framework
- Write unit tests for new functionality using pytest==8.3.3
- Update documentation for API changes
- Ensure all environment variables are documented
- Test both ingestion and chatbot workflows before submitting



Project Structure

```
ai-shopping-assistant/
— app.py
                          # Main Flask chatbot application
ingestor.py
                           # Admin data ingestion interface
                          # Python dependencies
requirements.txt
- .env
                          # Environment variables (not tracked)
.gitignore
                          # Git ignore rules
- LICENSE
                          # Apache 2.0 License
README.md
                          # Project documentation
Dockerfile
                          # Container configuration
setup.py
                          # Package setup
diagnostics.py
                          # System diagnostics
templates.py
                          # Template utilities
— test.py
                          # Test utilities
tester.ipynb
                          # Jupyter notebook for testing
- src/
  backend/
   ├─ orchestrator.py  # Central coordination logic
     product recommender.py # Recommendation engine
                         # Database querying and filtering
     — query_engine.py
     └─ prompts.py
                          # LLM prompt templates
    - constants/
      ____init__.py  # Configuration constants and env var
   — database/
     - aws s3 connection.py # S3 storage operations
     - aiven posgresql update.py # PostgreSQL operations
      - cloudflare connection.py # Cloudflare D1 database ops
     - load from database.py # Data retrieval operations
      - laptop data.csv # Sample dataset
      laptop data mapped.parquet # Pre-processed mapped data
    — frontend/
      - static/
      │ ├─ main.js
                            # Frontend JavaScript
        └─ style.css
                            # Frontend styling
      └─ templates/
         ├─ logo/
                           # Brand assets
          index.html
                            # Chatbot interface

    □ upload.html  # Admin upload interface

   logging/
     └─ init .py
                            # Logging configuration
    - utils/
      └─ init .py
                            # Utility functions
```



License

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Author & Contact

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This project demonstrates advanced AI integration patterns in production environments, showcasing best practices for LLM-powered applications, data pipeline automation, and scalable cloud deployment. The system serves as a comprehensive example of modern AI-driven e-commerce solutions with real-world deployment considerations.