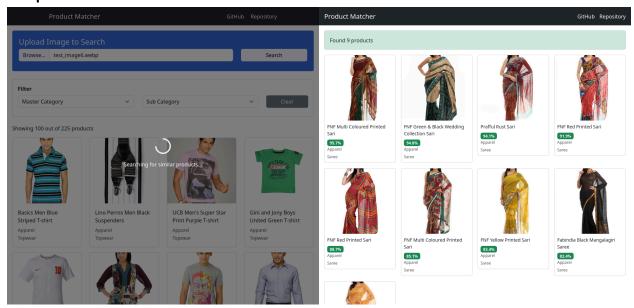
# PROJECT REPORT Submission by: ANIK HALDER 24MCA0251

<u>Visual Product Matcher Build</u> is a web-based fashion product catalogue and search application built with Flask, FastAPI, Bootstrap 5 and MongoDB Atlas. The application employs a custom built, fine-tuned and hosted hybrid machine learning architecture combining ResNet50 and OpenAI-CLIP models, outperforming traditional standalone CLIP implementations by more than 10% with 90% visual similarity and 95% semantic accuracy.

Link to deployed Application: <a href="https://visual-product-matcher-1hkn.onrender.com">https://visual-product-matcher-1hkn.onrender.com</a>
Link to Github Repository: <a href="https://github.com/Anikrage/Visual Product Matcher">https://github.com/Anikrage/Visual Product Matcher</a>

## **Sample Screenshot:**



### **Model Architecture:**

Unlike conventional single-model approaches, the custom built model uses:

- 1. Three-stage ranking algorithm that first identifies the visual matches using ResNet50 and CLIP image embeddings using cosine similarity.
- 2. Followed by a semantic filtering using CLIP text embeddings to eliminate gender/category confusions.
- 3. and then finally combining the scores to create a hybrid score and ranking the results. This approach addresses the common failures in simpler systems that often mismatch men's and women's products.

#### **Full Stack Implementation:**

The frontend uses Flask templating with Bootstrap 5 and vanilla javascript for responsive UI, while the backend FastAPI service handles the ML inference. The database is hosted online with MongoDB Atlas that stores product metadata, pre-computed embeddings, links to product image hosted via fast CDN using Cloudinary, with indexed queries for fast filtering.

## **Optimizations:**

The application follows a separation of concern approach, Flask for presentation, FastAPI for ML inference and MongoDB for persistence, enabling independent scaling. Further optimizations are done via:

- 1. pre-computed embeddings eliminating real-time inference
- 2. Numpy vectorized similarity calculations
- 3. heap-based selection for O(n log k) complexity
- 4. lazy loading with async pagination
- 5. CDN-cached resources

## **Cloud Deployment:**

- 1. The custom built model is hosted via HuggingFace Spaces that provides an API for connectivity and inference
- 2. The frontend and backend is hosted via Render with automatic configuration for continuous integration and deployment.

This approach results in a production system that demonstrates complete software engineering from data preprocessing and model training to API deployment, frontend integration and cloud deployment, achieving 1-2 second response times.