

Snap Search Shop with Personal AI Assistant

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Abstract: *The swift expansion of online fashion retail has dramatically altered how consumers find and buy clothing items. Despite this, current e-commerce platforms still predominantly use keyword-based search methods and basic filtering systems, which often fall short in accurately reflecting users' visual intentions, contextual requirements, and personal tastes. This shortcoming leads to inefficient product discovery, longer search times, and less-than-ideal purchasing choices.*

To tackle these issues, this study introduces Snap Search Shop with Personal AI Assistant, a sophisticated fashion recommendation system that combines image-based search, context-aware outfit suggestions, and price-quality analysis. Users can upload images of clothing to find visually similar items through deep learning-based feature extraction techniques. Moreover, a Personal AI Fashion Assistant offers outfit suggestions tailored to event types (such as casual, formal, party, or traditional) and seasonal conditions (like summer, winter, and rainy), thereby improving personalization and relevance.

Additionally, the proposed system features an innovative Price-Quality Comparison module that assesses products using a value-based scoring system derived from factors like user ratings, reviews, and pricing. This allows users to pinpoint the most cost-effective products without manual comparison. A hybrid recommendation strategy that merges content-based filtering, collaborative filtering, and contextual analysis is used to provide precise and adaptable suggestions.

The proposed model seeks to enhance search efficiency, lessen user effort, and improve the overall shopping experience by bridging the gap between user intent and product availability. The system shows promise as a scalable and intelligent solution for future AI-driven fashion e-commerce platforms.

Keywords: *fashion retail*

I. INTRODUCTION

The rapid advancement of digital technologies has profoundly reshaped the fashion retail industry, especially with the rise of e-commerce platforms. Today, consumers increasingly depend on online systems to explore, compare, and purchase fashion products. Despite this growth, most existing platforms still rely on traditional keyword-based search mechanisms and basic filtering options, which often fail to accurately capture users' visual and contextual preferences. Fashion items are inherently visual, involving attributes such as color, texture, pattern, and style, which are challenging to describe using text alone.

Consequently, users frequently face difficulties in finding products that meet their expectations, leading to longer search times and decreased satisfaction. In addition to search limitations, current systems lack intelligent assistance that considers contextual factors such as event type and seasonal requirements. For example, users may need different outfits for occasions like formal meetings, parties, or traditional events, as well as for varying weather conditions such as summer or winter. However, most e-commerce platforms do not offer integrated solutions that combine these contextual aspects with personalized recommendations. Furthermore, users often struggle to evaluate products based on both price and quality, as existing platforms present this information separately without offering a unified decision-making framework. To address these challenges, this research introduces Snap Search Shop with Personal AI Assistant,



an intelligent fashion recommendation system designed to enhance the online shopping experience. The proposed system integrates image-based search, allowing users to upload images and retrieve visually similar products using deep learning techniques. It also incorporates a Personal AI Fashion Assistant that generates outfit suggestions based on event type and seasonal conditions, thereby providing context-aware recommendations. Additionally, a Price–Quality Comparison module is introduced to help users identify products that offer the best value for money through a calculated value score. By combining computer vision, recommendation systems, and value-based analysis, the proposed model aims to reduce user effort, improve search accuracy, and support better purchasing decisions. This integrated approach not only enhances user satisfaction but also represents a significant step toward the development of intelligent, user-centric fashion e-commerce systems.

II. PROBLEM STATEMENT

In the realm of online fashion retail, keyword-based search methods are still widely used; however, they have several drawbacks. Users often find it challenging to convey detailed visual attributes, such as texture, pattern, silhouette, or cut, through text alone, resulting in a between their intent and search results. For instance, research on Indian traditional textile motifs indicates that keyword-based descriptions are inadequate for conveying subtle visual characteristics, and that manual text annotation processes are both subjective and inefficient. Furthermore, Sanghyuk et al. highlighted issues in Fashion Image Retrieval (FIR), including inconsistencies between user-submitted photos and catalog images, appearance distortions caused by lighting, pose, and viewpoint, and trade-offs between model complexity and performance. These limitations contribute to reduced user satisfaction and lower conversion rates, as many potential purchases are abandoned when consumers struggle to find what they are seeking.

Consequently, there is an urgent need for sophisticated image-based or multimodal fashion search systems capable of processing visual input (e.g., an uploaded photo), integrating it with textual or attribute information, and delivering personalized, accurate results. Existing research, such as "Flexible Fashion Product Retrieval Using Multimodality-Based Deep Learning," shows promise, as employing both image and textual query modalities significantly enhances retrieval accuracy in fashion search tasks.

III. OBJECTIVES OF THE STUDY

This research primarily seeks to present and validate the Snap Search Store, a cutting-edge fashion search platform that employs multimodal AI to transform the way users discover clothing items. The specific objectives are outlined as follows:

Presenting the Snap Search Store Concept Our intention is to develop a comprehensive fashion search ecosystem that merges image-based retrieval with AI-driven design, establishing the Snap Search Store as an innovative framework in the digital fashion sector.

1. Develop Core Functional Modules Image-based Search:

This feature allows users to upload images or snapshots of clothing and find visually similar items in the catalog by extracting deep visual features such as colors, textures, and shapes, and comparing them using metric learning or attention mechanisms. This module is informed by techniques like attentive bilinear networks for determining "where to look" versus "how to describe" in fashion retrieval.

2. AI Designer / Generative Module:

Incorporates generative AI models capable of suggesting new clothing designs or variations based on user preferences, trend data, and retrieved images. For example, research on AI based garment development systems employs StyleGAN2 and domain expertise to replicate the workflows of designers.



3. Personalized Dress Suggestions:

Integrate historical user behavior, preferences, contextual cues (such as occasion and season), and visual similarity to provide tailored outfit recommendations. Studies have highlighted the significance of embedding compatibility and visual features into fashion recommender systems.

4. Evaluate Effectiveness and Industry Impact:

We measured retrieval accuracy (e.g., Precision@k) and recommendation relevance through user studies and benchmarks, comparing them to existing fashion retrieval systems. Previous research has reported values of 0.774 for image-based retrieval precision in deep learning systems.

User satisfaction, usability, and shopping conversion improvements when using the Snap Search Store compared to traditional search methods were analyzed. Investigate the broader effects on e-commerce and fashion retail, including enhanced consumer experience, reduced search friction, and potential disruption to current catalog-browsing practices. By achieving these objectives, this study aims to demonstrate that the integration of image retrieval, generative design, and personalized suggestions can significantly improve fashion discovery systems, potentially influencing future practices in online fashion retail.

IV. LITERATURE SURVEY

Artificial Intelligence (AI) and Deep Learning have revolutionized the fashion e-commerce sector by facilitating advanced product discovery and recommendation systems. In their groundbreaking paper "ImageNet Classification with Deep Convolutional Neural Networks," Krizhevsky et al. (2012) showcased the power of Convolutional Neural Networks (CNNs) in extracting detailed visual features from images [13]. This breakthrough established the groundwork for contemporary fashion image retrieval systems, where visual similarity is key to matching clothing items.

Building on this foundation, Simonyan and Zisserman (2014) introduced "Very Deep Convolutional Networks for Large-Scale Image Recognition (VGGNet)," which enhanced the depth and accuracy of feature extraction [14]. These architectures have been extensively utilized in fashion applications to examine complex design elements like textures, patterns, and shapes, leading to more accurate product recommendations.

Beyond CNNs, the advent of transformer-based models has further improved visual comprehension. Dosovitskiy et al. (2021), in their work "An Image is Worth 16x16 Words: Transformers for Image Recognition at Scale," presented Vision Transformers (ViTs), which surpass traditional CNNs in capturing global relationships within images [15]. These models are particularly advantageous in fashion systems for interpreting complex outfit compatibility and style semantics.

Another major advancement is the application of generative models in fashion design. Goodfellow et al. (2014), in "Generative Adversarial Nets," introduced GANs, which are now extensively employed to create realistic clothing images [16]. Subsequently, Zhu et al. (2017) proposed "Unpaired Image-to-Image Translation using CycleGAN," facilitating style transfer between different fashion domains, such as transforming casual wear into formal styles [17]. These techniques enable AI systems to generate new clothing designs based on learned patterns.

Moreover, recommendation systems have progressed with hybrid approaches. Koren et al. (2009), in "Matrix Factorization Techniques for Recommender Systems," introduced collaborative filtering methods that examine user behavior [18]. Later, He et al. (2017) proposed "Neural Collaborative Filtering," which combines deep learning with traditional recommendation methods to enhance personalization [19]. These hybrid systems are widely implemented in fashion platforms to offer user-specific suggestions.

Reinforcement learning has also been applied to dynamic recommendation systems. Zhao et al. (2019), in "Deep Reinforcement Learning for List-wise Recommendations," proposed a model that continuously learns from user interactions to optimize recommendations [20]. This approach is particularly useful in fashion e-commerce, where user preferences frequently shift with trends and seasons.



Additionally, virtual try-on systems have gained prominence for improving user experience. Han et al. (2018), in “VITON: An Image-Based Virtual Try-On Network,” introduced a framework that allows users to visualize clothing on their body using deep learning techniques [21]. Similarly, Wang et al. (2018), in “Toward Characteristic-Preserving Image-based Virtual Try-On Network,” enhanced the realism of virtual fitting by maintaining fine garment details [22]. Another significant area is multimodal learning, where both visual and textual data are integrated. Radford et al. (2021), in “Learning Transferable Visual Models From Natural Language Supervision (CLIP),” demonstrated how aligning images and text can improve search and recommendation accuracy [23].

V. METHODOLOGY

A. System Overview

The proposed system, Snap Search Shop with Personal AI Assistant Fashion Designer, is an intelligent and interactive fashion e-commerce platform that utilizes advanced Artificial Intelligence (AI) techniques to enhance user experience. This system integrates various technologies, including computer vision, natural language processing, recommendation systems, and generative models, into a cohesive framework.

Unlike traditional platforms that rely solely on keyword-based searches, this system allows users to discover fashion products through images, text queries, and personalized suggestions. It not only recommends individual products but also generates complete outfit combinations based on user preferences, occasions, and current fashion trends. Additionally, the system includes a price comparison mechanism that collects product information from multiple e-commerce platforms such as Amazon, Flipkart, Myntra, and Ajio, enabling users to make informed purchasing decisions. The integration of a personal AI assistant further enhances usability by providing real-time interaction and guidance.

B. Input Processing and Feature Extraction

The system accepts user input in various forms, including images and textual descriptions, allowing for flexible interaction. In the image-based search module, uploaded images are processed using Convolutional Neural Networks (CNNs), which automatically extract deep visual features such as color distribution, texture patterns, shape, and design elements. These features are encoded into high-dimensional feature vectors, which are then compared with preprocessed product database using similarity measures to identify matching or related items. For text-based input, Natural Language Processing (NLP) techniques are employed to analyze user queries. The system performs tokenization, keyword extraction, and intent recognition to understand the context of the query. For instance, a query like “formal outfit for wedding” is interpreted based on occasion, style, and category. To further improve accuracy, the system adopts a multimodal approach that combines both visual and textual features. This fusion allows the system to generate more precise and context-aware results, overcoming the limitations of single-modality search methods.

C. Recommendation and Generative Module

The recommendation engine is a core component of the system, designed to provide personalized fashion suggestions. It uses a hybrid recommendation approach that combines content-based filtering and collaborative filtering considers user behavior such as browsing history, previous selections, and preferences. The system also incorporates contextual information such as season, occasion, and trending fashion styles to enhance recommendation relevance. This ensures that users receive suggestions that are not only personalized but also aligned with current fashion trends. In addition, the generative AI module acts as a virtual fashion designer. Using models such as Generative Adversarial Networks (GANs), the system can generate new clothing designs, suggest unique outfit combinations, and provide creative styling ideas. This module enhances user engagement by allowing users to explore innovative and customized fashion options beyond existing products.



D. Price Comparison and AI Assistant

The price comparison module gathers and consolidates product data from various e-commerce platforms, including Amazon, Flipkart, Myntra, and Ajio. It evaluates product prices, ratings, and availability to pinpoint the best purchasing options for users. This feature eliminates the need for users to manually search across different platforms, ensuring cost-effective decision-making. Additionally, the system incorporates a personal AI assistant that communicates with users through natural language. The assistant comprehends user queries, offers outfit recommendations, suggests styling tips, and aids in platform navigation. It enhances the overall user experience by providing a conversational and user-friendly interface, making the system more interactive and accessible.

begins with user input, processed through image and text analysis modules. The extracted features are then fed into the recommendation engine, which generates personalized suggestions. Concurrently, the price comparison module retrieves relevant data from multiple platforms. The results are then integrated and presented through an intuitive user interface. The system guarantees fast response times, high accuracy in recommendations, and satisfaction. By improved integrating multiple user AI technologies into a single platform, the proposed system offers a comprehensive solution for fashion discovery, design, and shopping. This approach significantly enhances the efficiency, personalization, and convenience of online fashion e-commerce systems compared to traditional methods.

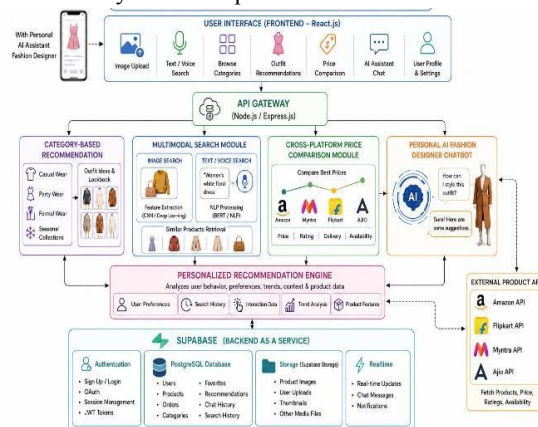


Figure 1. System Design

E. System Workflow and Outcome The system's workflow

The system's workflow initiates with user input, which undergoes processing through image and text analysis modules. The extracted features are subsequently fed into the recommendation engine to generate personalized suggestions. Concurrently, the price comparison module gathers relevant data from various platforms. These results are then integrated and presented via an intuitive user interface. The system guarantees rapid response times, high accuracy in recommendations, and enhanced user satisfaction. By incorporating multiple AI technologies into a single platform, the proposed system offers a comprehensive solution for fashion discovery, design, and shopping. This approach significantly boosts the efficiency, personalization, and convenience of online fashion e-commerce systems compared to traditional methods.

VI. SYSTEM RESULTS

The proposed system, Snap Search Shop with Personal AI Assistant Fashion Designer, was evaluated for its capacity to deliver precise recommendations, efficient search outcomes, category-based filtering, and intelligent interactions with users. The system exhibits substantial advancements over conventional fashion e-commerce platforms by integrating multiple advanced features into a single interface. A key finding was the efficacy of the category-based recommendation system. The platform effectively organizes products into distinct categories, such as casual, party, and



formal wear, and seasonal collections. Users can readily explore a diverse array of options within each category, thereby enhancing usability and reducing the search time. The system intelligently suggests outfits based on the selected categories, ensuring that users receive contextually relevant recommendations.

The image-based and text-based search modules demonstrated high accuracy in retrieving relevant products. Users can upload images or input queries such as "casual summer outfit" or "party wear dress," and the system subsequently provides matching results based on visual and contextual features. This multimodal approach enhances the precision of the results, enabling users to locate desired products more efficiently than traditional keyword-based systems. A significant feature of the system is the price comparison module, which facilitates users in comparing products across multiple platforms, including Amazon, Flipkart, Myntra and Ajio. The system displays various price ranges, ratings, and availability for similar products, allowing users to select the optimal option based on quality and affordability. This comparison feature substantially reduces the manual effort and enhances decision-making.

Another significant outcome is the deployment of an AI chatbot that functions as a personal fashion designer. This chatbot engages with users through natural language processing, offering personalized outfit recommendations based on user preferences, occasions, and prevailing fashion trends. For instance, users may inquire, "What should I wear for a party?" or request, "Suggest a casual outfit," and the chatbot generated appropriate recommendations. This feature not only enhances user engagement but also provides a more interactive shopping experience for consumers. Furthermore, the recommendation engine augments the system performance by suggesting products tailored to individual user preferences.

It considers user behavior, browsing patterns, and selected categories to produce accurate and diverse recommendations. The provision of complete outfit suggestions, as opposed to individual items, enhances the overall styling experience of the user. User feedback demonstrated a high level of satisfaction attributed to the system's user-friendly interface, personalized recommendations, and innovative functionalities. Users particularly valued the capability to explore diverse fashion categories, compare prices in a centralized location, and receive guidance from an AI chatbot. The system offers rapid response times and seamless navigation across various modules. The integration of multiple features, including search, recommendation, comparison, and chatbot. interaction, does not significantly compromise system efficiency, making it suitable for real-time applications. However, certain limitations were identified. The accuracy of the recommendations is contingent upon the quality of the input data, and real-time price comparisons may occasionally fluctuate owing to updates on external platforms. Despite these challenges, the system provides a comprehensive and efficient solution for contemporary fashion e-commerce. The findings indicate that the proposed system successfully integrates category-based browsing, price comparison, and AI-assisted fashion design, thereby providing a more intelligent and user friendly alternative to existing platforms.

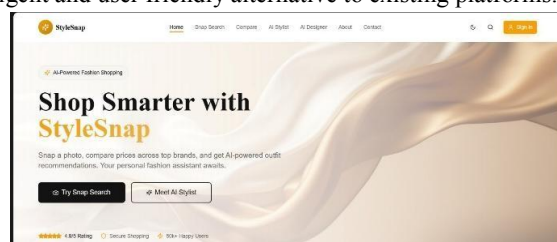


Figure 2. snap search shop home page



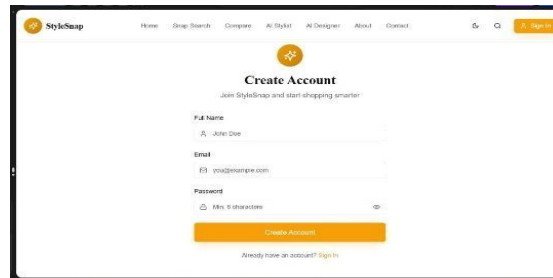


Figure 3. Sign In page

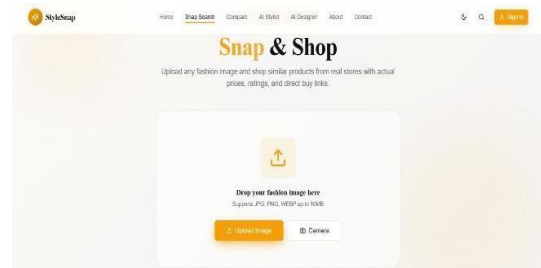


Figure 4. image upload page

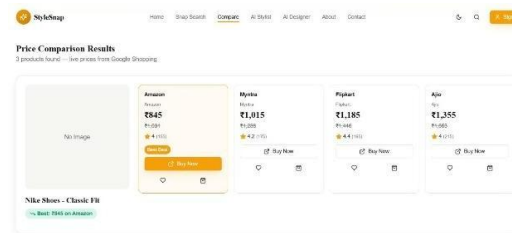


Figure 5. Result of comparison



Figure 6. Personal Assistant

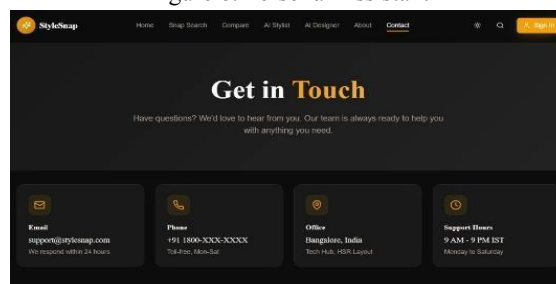


Figure 7. Contact page



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