

Artificial Intelligence–Driven Image-to-Garment Translation Using CLO3D: Bridging Generative Design and Physical Production

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Abstract—Generative artificial intelligence has transformed fashion concept development by producing detailed garment imagery from prompts or reference inputs. However, these images remain visually suggestive rather than technically complete, as they do not define seam structure, pattern geometry, fabric behavior, or construction sequence. This study addresses that limitation through an image-to-garment workflow that combines AI-generated imagery, CLO3D pattern reconstruction, fabric simulation, and manual construction. Using a case-study approach, the research traces the translation of a generated fashion image into a digital pattern set and then into a physical garment. The analysis shows that visual realism alone does not ensure manufacturability; successful translation depends on human interpretation, pattern adjustment, and simulation-based testing. CLO3D serves as the key bridge between conceptual imagery and physical production by exposing structural and material constraints that the image conceals. The findings suggest that AI can support early design exploration effectively, but technical pattern knowledge and manual production remain essential for turning generated visuals into wearable garments.

Index Terms—Artificial intelligence, CLO3D, digital pattern making, garment construction, image-to-garment translation, fashion simulation

I. INTRODUCTION

Artificial intelligence has become an increasingly important component of fashion design, particularly in generating visual concepts. Current systems can produce highly resolved, editorially styled, and materially convincing garment images with minimal prompting. This has reshaped early design stages, where mood boards and sketches once dominated,

allowing a single AI-generated image to condense silhouette, styling, and atmosphere into one immediate visual proposition [1], [5], [10], [11].

While this shift is significant, it introduces a structural limitation. Fashion design is not purely visual; it is inherently material and technical. A compelling image may suggest a direction, but it rarely provides sufficient information for actual garment production. The transition from image to object therefore remains one of the least developed aspects of AI-assisted fashion. Existing research demonstrates that garments can be reconstructed digitally through computer vision and virtual try-on systems, yet these approaches largely remain confined to simulation rather than physical production [6], [7].

This study addresses that gap by focusing on translation rather than generation. It asks whether a well-defined AI-generated garment image can be systematically converted into a physical product. This requires interpreting visual outputs as technical structures, developing and testing patterns in CLO3D, and evaluating how digital decisions translate into real materials and construction.

The study is not concerned with AI's ability to imagine garments, but with whether those imagined forms can be realized physically. By examining a complete workflow from image generation to CLO3D patterning and final construction it offers a practical and critical perspective on the capabilities and limitations of AI in contemporary fashion development.

II. LITERATURE REVIEW

A. Human AI Collaboration in Fashion

Research on human AI collaboration in fashion positions AI as a creative amplifier rather than an autonomous designer [2]. Generative tools enable rapid production of multiple design variations, allowing exploration of silhouette, proportion, and styling within a compressed timeframe [1], [4]. This has shifted AI from a back-end analytical tool to a front-end ideation partner.

However, most studies focus on conceptual design and consumer response rather than technical reconstruction. Human-AI co-creation is often evaluated in terms of novelty or preference, not manufacturability [3]. This is a critical gap, as fashion design must also satisfy structural and material requirements. As a result, AI supports creativity but does not replace the designer's interpretive role. The collaboration remains asymmetrical: AI generates possibilities, while the designer determines feasibility [1], [4].

B. Technical Systems for Garment Generation

Technical research has advanced significantly, with systems such as DressCode enabling text-to-garment generation and multimodal frameworks improving image-based editing accuracy [5], [7]. These developments show that fashion can be computationally modeled beyond surface aesthetics. Yet visual accuracy does not ensure physical feasibility. AI can convincingly render folds, seams, and layers without encoding how garments are constructed. This limitation has led to the development of physics-aware CAD systems that integrate simulation with design [6]. CLO3D is central in this context, as it allows pattern-based modeling, fit testing, and fabric simulation. In this study, it functions not as a rendering tool but as a reconstruction system bridging visual outputs and physical garments.

C. Legal and Ethical Considerations

Generative AI raises questions of authorship, originality, and ownership, particularly when outputs derive from existing visual datasets [9]. These concerns are significant in fashion, where design identity and authorship carry commercial value.

Although not the primary focus here, the issue remains relevant. Translating AI-generated images into

physical garments shifts them from conceptual visuals to material products, increasing the need for transparency and human oversight in the design process [9].

D. Fragmentation of Research

A key limitation in current literature is fragmentation. Research often isolates generation, simulation, and visualization, while real fashion workflows integrate all stages [10]. Image generation ultimately connects to pattern development, material selection, and construction, yet most studies stop at the visual stage. This study addresses that gap by linking AI image generation with physical garment production, using CLO3D as the intermediary between visual representation and material execution.

III. METHODOLOGY

A. Research Design

The study adopts a practice-based case study methodology. A single AI-generated fashion image was selected as the basis for reconstruction, then analyzed structurally, rebuilt in CLO3D as a digital pattern set, simulated for fit and drape, and finally constructed manually. This sequence enables direct comparison between the AI image, digital reconstruction, and physical garment.

The approach is translational rather than speculative. Instead of examining AI's ability to generate novel forms, the study tests whether a visual output can be interpreted as a manufacturing problem, requiring a shift from appearance to structure.

B. Image Interpretation and Structural Reading

The first stage involved treating the AI-generated image as a design document. Visible garment components, layering, silhouette boundaries, and surface effects were identified. The selected image features a strapless top with a central twist and a layered skirt combining opaque and sheer sections. While visually coherent, it lacks construction details. The designer therefore inferred structure. The top was interpreted as a shaped bandeau with central gathering, while the skirt was reconstructed as layered curved panels. This interpretive step is essential, as AI images do not define seams, closures, or internal support; they must be read as structural suggestions rather than explicit instructions.

C. Pattern Development in CLO3D

Following interpretation, CLO3D was used to develop the pattern structure. Starting from base blocks, asymmetry, curvature, and layered segmentation were introduced to match the AI design. CLO3D enabled testing of structural behavior, particularly how the top maintains form and how layered panels interact around the body [6].

This stage revealed the gap between visual resemblance and structural equivalence. Effects such as transparency and diagonal layering required multiple pattern pieces and careful seam planning. CLO3D thus functioned not only as a simulation tool but as a reconstruction system, exposing the technical work behind visual translation.

D. Fabric Simulation and Technical Validation

Fabric simulation was conducted to evaluate material behavior. The design relies on contrast between structured and sheer fabrics, making parameters such as weight, stiffness, and drape critical. CLO3D allowed these properties to be tested before physical production [6].

Simulation also served as validation. It assessed silhouette stability, layering behavior, and structural support, identifying potential issues before fabrication. In this sense, CLO3D acted as a diagnostic environment, revealing constraints that the AI image does not represent.

E. Physical Construction

The finalized patterns were then used for manual garment construction. This stage tested the feasibility of the digital reconstruction under real production conditions, including fabric handling, stitching, and assembly.

The garment was successfully constructed, though minor adjustments were required in proportion and stabilization. These modifications highlight that translation is not automatic but negotiated through material conditions. The outcome confirms that AI-generated images can be realized physically, provided they are supported by structured interpretation, simulation, and technical execution.



Fig no. 1 AI generated garment

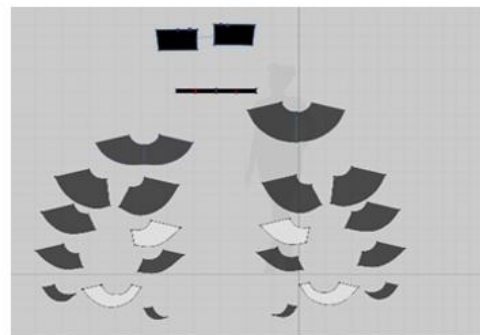


Fig no. 2 Pattern development in CLO3D



Fig no. 3 Final look in CLO3D

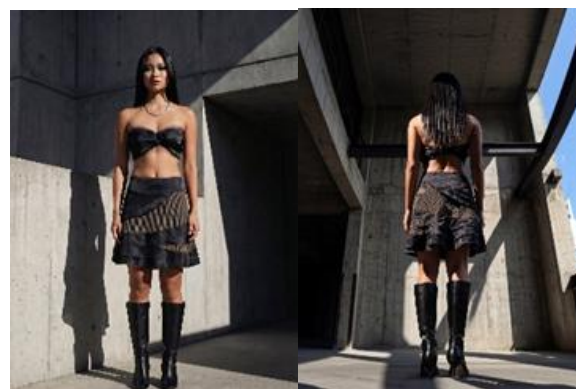


Fig no. 4 Photoshoot of physical constructed garment

IV. RESULTS AND DISCUSSION

A. From AI Image to Constructible Form

The case study demonstrates that a well-generated AI fashion image can be translated into a physical garment when approached through technical reconstruction rather than direct replication. The AI output functions as a visual template, defining silhouette, layering, and surface qualities, but not construction. CLO3D and manual pattern development are required to convert this visual intent into a manufacturable form.

This clarifies AI's role in fashion design: it initiates the design problem but does not resolve it. The designer remains responsible for translating the image into pattern logic, maintaining structural balance, and preserving the visual identity of the original output [1], [4].

B. The Role of CLO3D as a Translation Layer

CLO3D serves as the central mechanism in the workflow, enabling the garment to be modeled as a system of pattern components rather than a flat image. Its simulation capabilities allow iterative adjustments, ensuring that proportion, layering, and fit are resolved before production [6].

This is critical because AI-generated designs often fail at the structural level. CLO3D converts visual ambiguity into measurable geometry, introducing technical control into the process. Rather than accelerating blindly, it enables informed decision-making through iterative testing [6].

C. Material Differences Between Digital and Physical Garments

A key finding is the difference between digital appearance and physical behavior. While AI images present smooth and controlled forms, real fabrics respond to gravity, tension, and construction constraints. Elements such as layering and transparency behave differently once materialized.

This highlights that AI realism is primarily visual, not physical. CLO3D reduces this gap through simulation, but only physical construction fully reveals material behavior. The final garment must therefore be evaluated as a material object rather than a visual approximation [5], [6].

D. Human Expertise as the Critical Mediator

Human expertise remains essential throughout the process. AI cannot determine construction logic, material constraints, or fabrication methods. These decisions rely on pattern knowledge and practical experience [1], [4].

The workflow expands the designer's role, positioning them as an interpreter of AI outputs and a mediator between visual concept and physical execution. AI generates possibilities, but feasibility depends on human judgment.

E. Significance of the Workflow

The study demonstrates that AI-generated imagery can extend beyond visualization into pre-production when supported by structured reconstruction. By integrating AI generation, CLO3D simulation, and manual construction, the workflow connects digital creativity with material execution.

At the same time, the process remains inherently hybrid. AI provides speed and variation, CLO3D ensures structural validation, and manual production grounds the outcome in reality. The final garment is therefore the result of coordinated human-machine collaboration rather than automated generation [6], [10].

V. CONCLUSION

This research demonstrates that AI-generated fashion imagery can be translated into physical garments through a hybrid workflow combining visual interpretation, CLO3D reconstruction, fabric simulation, and manual production. The study establishes that the image is not the endpoint of design but the starting point of a technical process completed through pattern logic and material execution.

CLO3D is central to this process, acting as the bridge between visual form and manufacturable structure. It allows designers to test feasibility prior to production, reducing errors and clarifying how visual elements must be re-engineered for the body [6]. The manual construction stage then validates the simulation, confirming that garments are ultimately shaped by material behavior as much as by design intent.

The findings suggest that AI can support fashion production effectively, but only when its outputs are treated as generative inputs rather than final designs. The future of image-to-garment translation lies in

integrated workflows where AI expands creative possibilities while human expertise ensures structural and material viability [1], [4]. In this context, AI's value is not in replacing design processes, but in accelerating them while maintaining technical accountability.

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