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01 Introduction

- Challenge: accurately visualizing products and ensuring customer satisfaction.
- Solved by virtual try-on solutions, which offer users an immersive and realistic preview of how clothing items look on them before making a purchase.
- Our Minimum Viable Product (MVP) addresses the growing demand for a tailored and engaging online shopping experience, focusing on two major sportswear brands — Nike and Adidas.
- Platform used: web-based, where users can select among products, and virtually try on selected product.
- This enhances users' confidence in online purchases and minimizing the uncertainty associated with size, fit, and style, revolutionizing the way users interact with and choose sportswear from these iconic brands.

Problem Statement

- Address lies in the inherent limitations of online apparel shopping, particularly in the sportswear sector represented by Nike and Adidas.
- Pain points or inefficiencies:



- Trying on clothes in physical stores: time-consuming, inconvenient, and sometimes unhygienic. It requires the user to travel to the store, wait in line, and try on multiple clothes to find the right fit and style.
- Frustrating online shopping experience for customers who cannot try on the clothes before purchasing them. Customers often grapple with uncertainty regarding sizing, fit, and overall aesthetic, leading to an increase in return rates and operational costs for both consumers and retailers.

O2 Problem Statement

- Present online retail landscape lacks a comprehensive solution specifically tailored to
 Nike and Adidas sportswear, and while there are generic virtual try-on platforms
 available, their lack of specialization results in suboptimal user experiences. Competitors in
 the virtual try-on space offer broader solutions but lack the specificity needed to address
 the unique challenges associated with Nike and Adidas sportswear.
- The MVP seeks to fill this void, providing a targeted and immersive virtual try-on experience that not only addresses current inefficiencies but also distinguishes itself from competitors by catering specifically to the distinct needs of sportswear enthusiasts.

03 Solution Overview - Task Analysis

The task involves a process known as image-to-image translation, which is a type of image synthesis in computer vision and graphics and can be achieved using deep learning techniques, specifically Generative Adversarial Networks (GANs).

Our solution - Virtual Fitting Room - involves 3 processes as follow:

- 1. **Data Preparation:** Collect a dataset of images of clothes and people wearing different types of clothes from Adidas and Nike.
- 2. **Clothes Segmentation**: Use U-2-NET model to identify and extract the clothes from the image that contains the target clothing, in the form of a binary mask. This step is skipped if the image only shows the clothes.
- 3. **Clothes Transfer:** Feed the image of person X (who will try on clothes) and the binary mask Y of the clothes into the trained generator network (C-VTON). The generator should output an image of person X wearing the clothes Y (Or the given image of the clothes, instead of the clothes on person y).

O3 Solution Overview - Task Analysis

The user will communicate with this virtual fitting room via a website that is built by Flask. Users will upload images of themselves to the website and pick clothes they want to try on. The website will then return a new image with the user wearing the new clothes.



Convenient way for people to try on clothes

travel to physical stores 24/7 availability

Helps customers make more informed decisions about size and fit, decreases return rates and associated costs for both consumers and retailers.

04 Methodologies - Overview

Supposing we have:

- **Input X**: image of person we want to fit the clothes on.
- Target Y: either image of only the targeted clothes, or image if model that is wearing the targeted clothes.

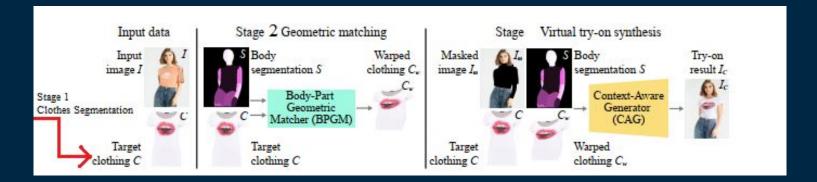
We want to retrieve the clothes from **Y** and apply that on input **X** to create a simulation of the targeted clothes on input **X**.







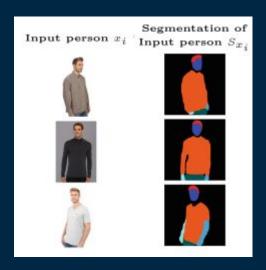
04 Methodologies - Architecture



INPUT DATA:

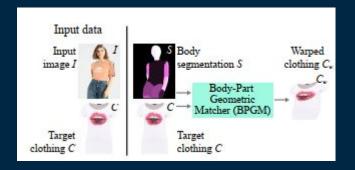
- **Input Image I:** The full image body of the user
- Target clothing C: Which can happen in 2 cases:
 - The target clothing is an image of the clothes.
 - The target clothing is worn by a human model.

04 Methodologies - Architecture - Stage 1 (Clothes Segmentation)



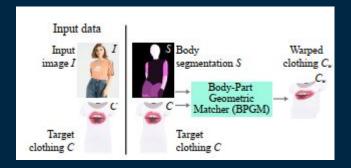
- Optional stage
- Solves the case of target clothing C being an image of clothes worn by a human model.
- The segmentation model identifies and **extracts clothes** from the given image, producing a binary mask.
- Segmentation of clothes is achieved by using the U-2-NET model.

Methodologies - Architecture - Stage 2 (Geometric Matching)



- A simplified geometric matching module named body-part geometric matcher (BPGM) is used.
- It warps the target clothing item so that it matches the geometry of the person's body in the input image.
- This involves adjusting the shape and orientation of the clothing item to look like it's being worn by the person.
- This module can solve the problem of challenging poses and arm configurations.

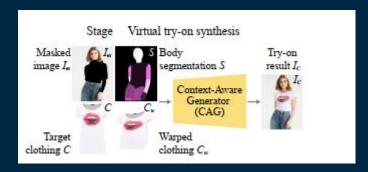
Methodologies - Architecture - Stage 2 (Geometric Matching)



STEPS:

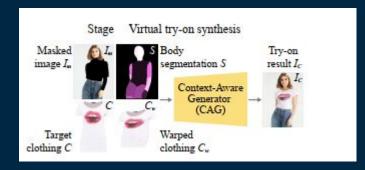
- DensePose model is applied to generate the body segmentations S from input image I.
- The target clothing **C**, along with **S**, is fed into 2 distinct encoders to generate 2 corresponding feature representations.
- These feature representations go through a few preprocessing before being organized to compute a correlation matrix.
- The correlation matrix is used to define a 2n^2-vector that represents the warped clothing Cw.

Methodologies - Architecture - Stage 3 (Visual Try-on Synthesis)



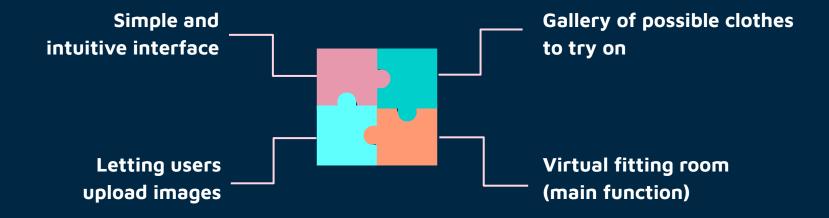
- Involves a context-aware generator (CAG).
- Takes the warped clothing item from Stage 2 and integrates it with the original person image to create a final image.

Methodologies - Architecture - Stage 3 (Visual Try-on Synthesis)



STEPS:

- Define Image Context (IC) as a channel-wise concatenation of S, C, Cw and masked input image Im.
- Context-Aware Normalization (CAN): a normalization layer used in the CAG module. It adjusts the features of the images based on the local context and the group information. This helps to preserve the details and textures of the clothes and the body parts.
- CAG consists of a sequence of ResNet blocks augmented with CANs added before every convolutional layer. These CAN layers efficiently utilize the IC and feed the generator with critical contextual information.









Frechet Inception Distance	Speed	User satisfaction
A lower FID indicates that the generated samples are more similar to the real data.	How fast the virtual try-on system can process the input and output images.	How happy the users are with the virtual try-on system and its features.

07 Roadmap Market Research and 01 Frontend Development 05 User Needs Analysis 02 **Scope Definition** 06 Testing Design UI 03 07 Launch 04 **Backend Development** 80 Feedback Loop

- **Al solution for virtual try-on**: Try2Buy presents a system that can generate realistic images of people wearing different clothes, given an input image of a person and a binary mask of the clothes.
- **Key components**: The system consists of image-to-image translation, clothes segmentation, clothes transfer, data preparation and evaluation techniques.
- **Benefits and impact**: The system can enhance customer experience, boost sales and marketing, and promote sustainability and innovation in the fashion industry.