

Learning to Wear: Details-Preserved Virtual Try-on via Disentangling Clothes and Wearer

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Introduction

Virtual Try-on

- Synthesizing a realistic image of a person wearing the given clothing



Input (Model + Target clothes)



Output

- Need to disentangle human and clothes & different types of clothes data
- Should be generalizable to various human poses and body shapes

Motivation

Inability to reflect details of the target clothing

- Detailed characteristics** of the target clothing (e.g., shape of neckline and sleeves) are not retained.
- Output often reveals characteristics of clothes in the **reference image**.

Limited understanding in 3D semantics of wearing clothes

- Parts that should **not be seen when worn** (e.g., inner side of shirt neckline) are still visible when clothing is worn.
- Overall struggle in synthesizing well-fitted images indicate a weak generalizability of the models.

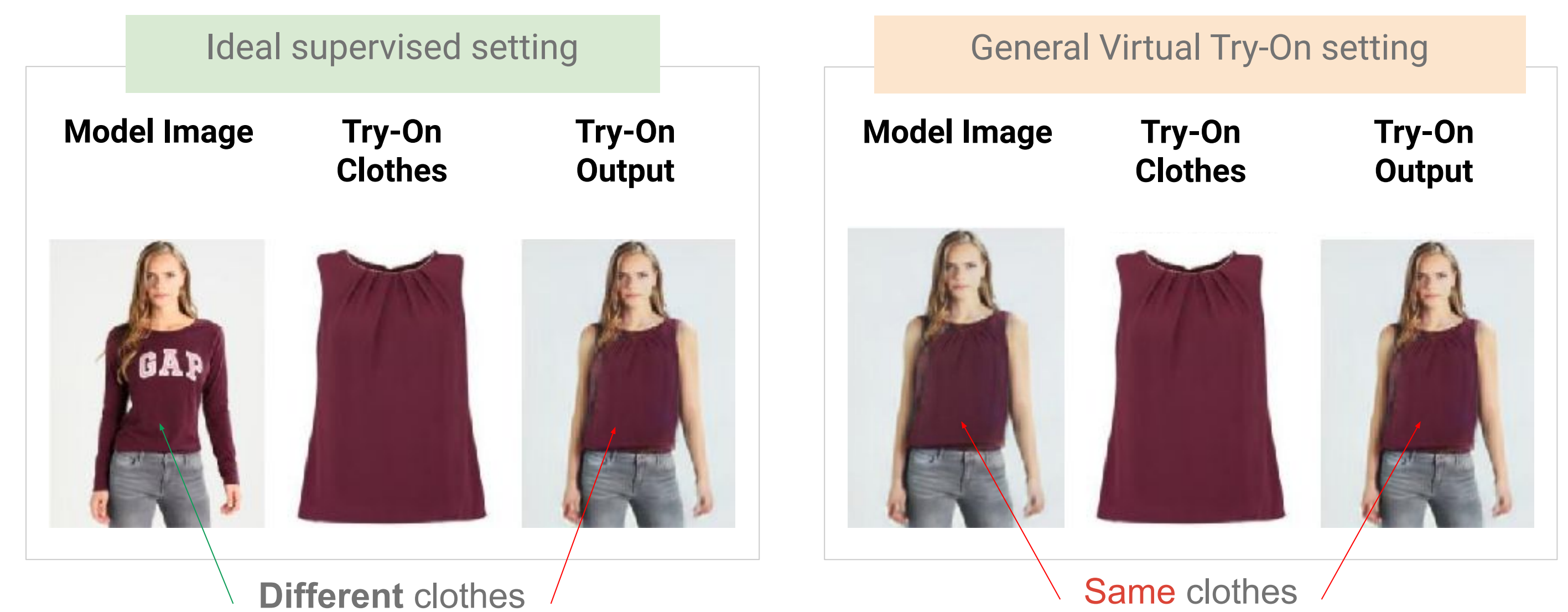
→ **Disentangle clothes & wearer and target & source clothes**



Core Issue

Lack of paired datasets

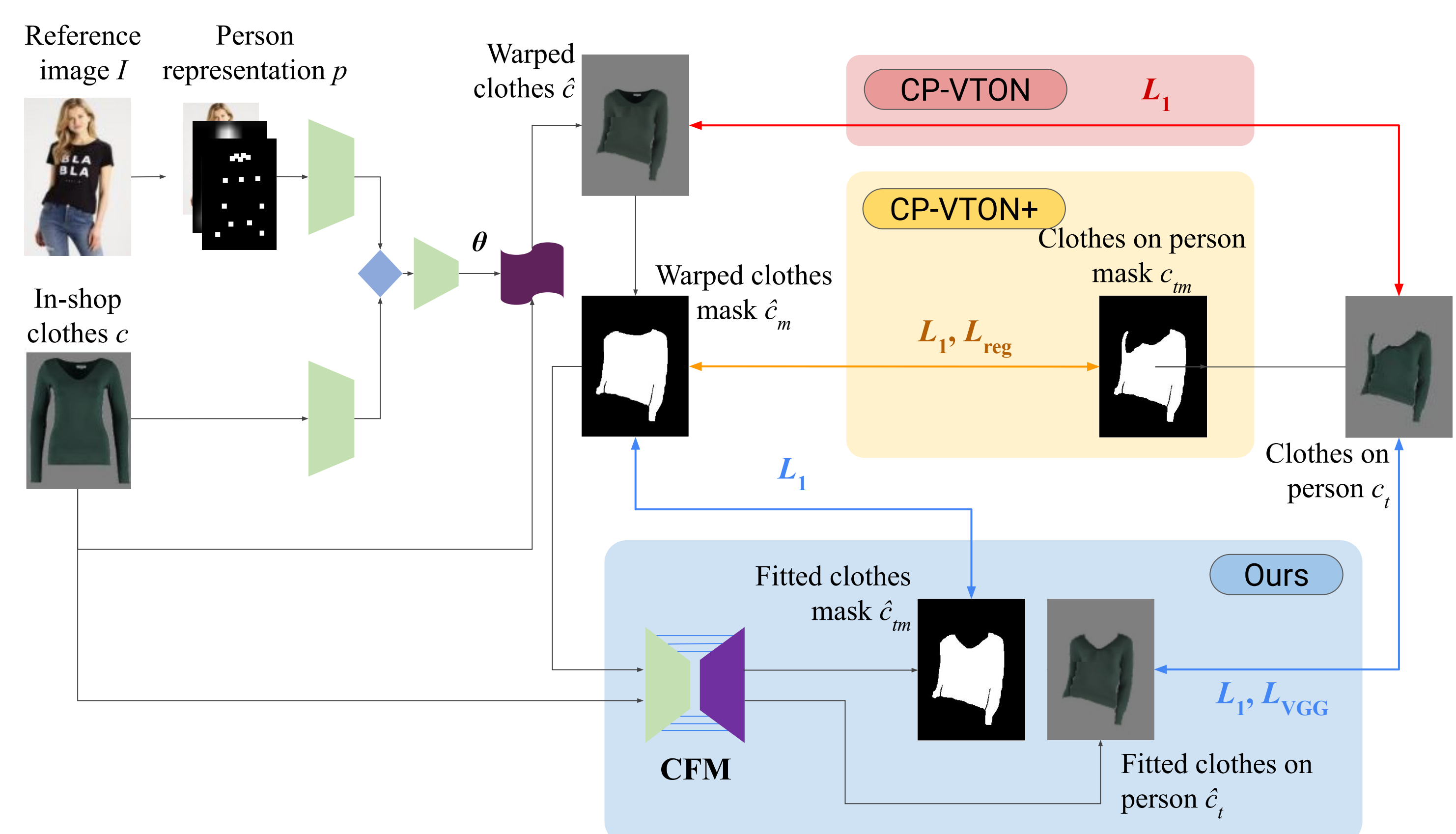
- Impossible to have two photos of a model with just different clothes on.
- Usually, models are trained to wear the **same clothes** that is already worn.



Method

Clothes Fitting Module (CFM): Learning to Wear

- Inserted between Geometric Warping Module (GWM) and Try-on Module.
- Allows GWM to perceive clothes **as the source** clothes in the reference.
- Allows CFM to perceive clothes **as the target** ground truth.
- Successfully disentangles source & target clothes and clothes & wearer**



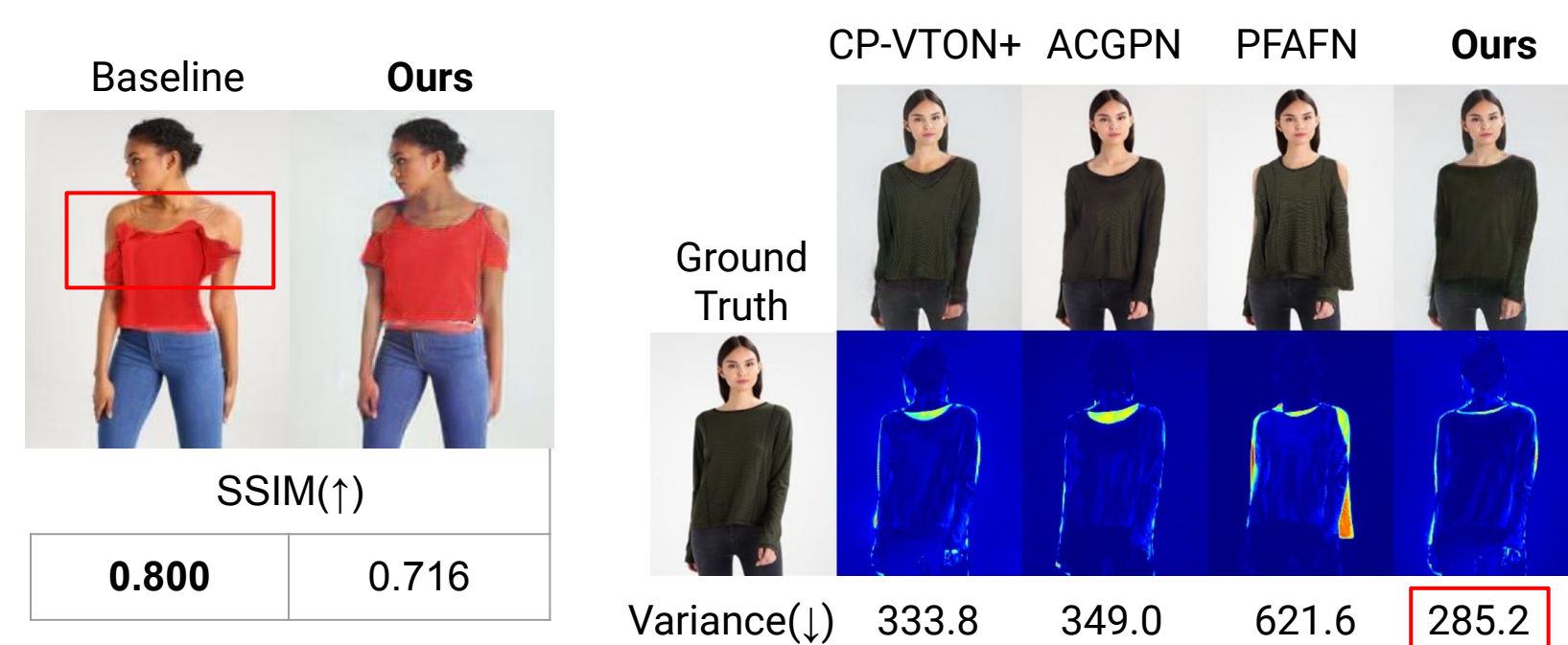
Results

- Retain properties** of target clothes, **disentangled** from the reference image
- Generalizable** to various designs, as well as body shapes and poses.



Evaluation

Flaw of existing metrics based on entire images

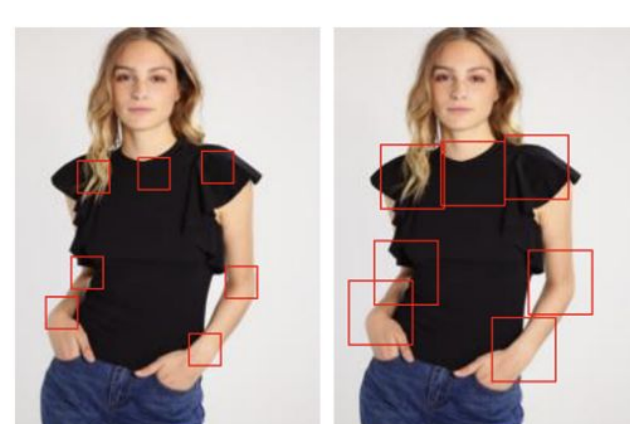


- Misalignment between qualitative and quantitative evaluation
- Considers even parts irrelevant for virtual try-on (e.g., background).

Novel approach to evaluation metrics

- Focus on areas relevant to the task (i.e., major body keypoints) for better alignment with human perception.

$$\text{Metric}_c^p(I) = \frac{1}{k} \sum_{i=1}^k \text{Metric}^{\text{all}} \left(I \left[x_i - \frac{\epsilon}{2} : x_i + \frac{\epsilon}{2}, y_i - \frac{\epsilon}{2} : y_i + \frac{\epsilon}{2} \right] \right)$$



Take-home Messages

- Previous virtual try-on models learned entangled representations that lack generalizability due to the lack of paired datasets.
- With CFM, we **disentangle** important factors of virtual try-on and detour the inherent limitation in data.
- Our **patch-based evaluation metrics** better correspond to qualitative results.