

# GAMES Webinar

## Generative Models for Clothed Human

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01

## Compatible and Diverse Fashion Image Inpainting

# Fashion Recommendation as Inpainting

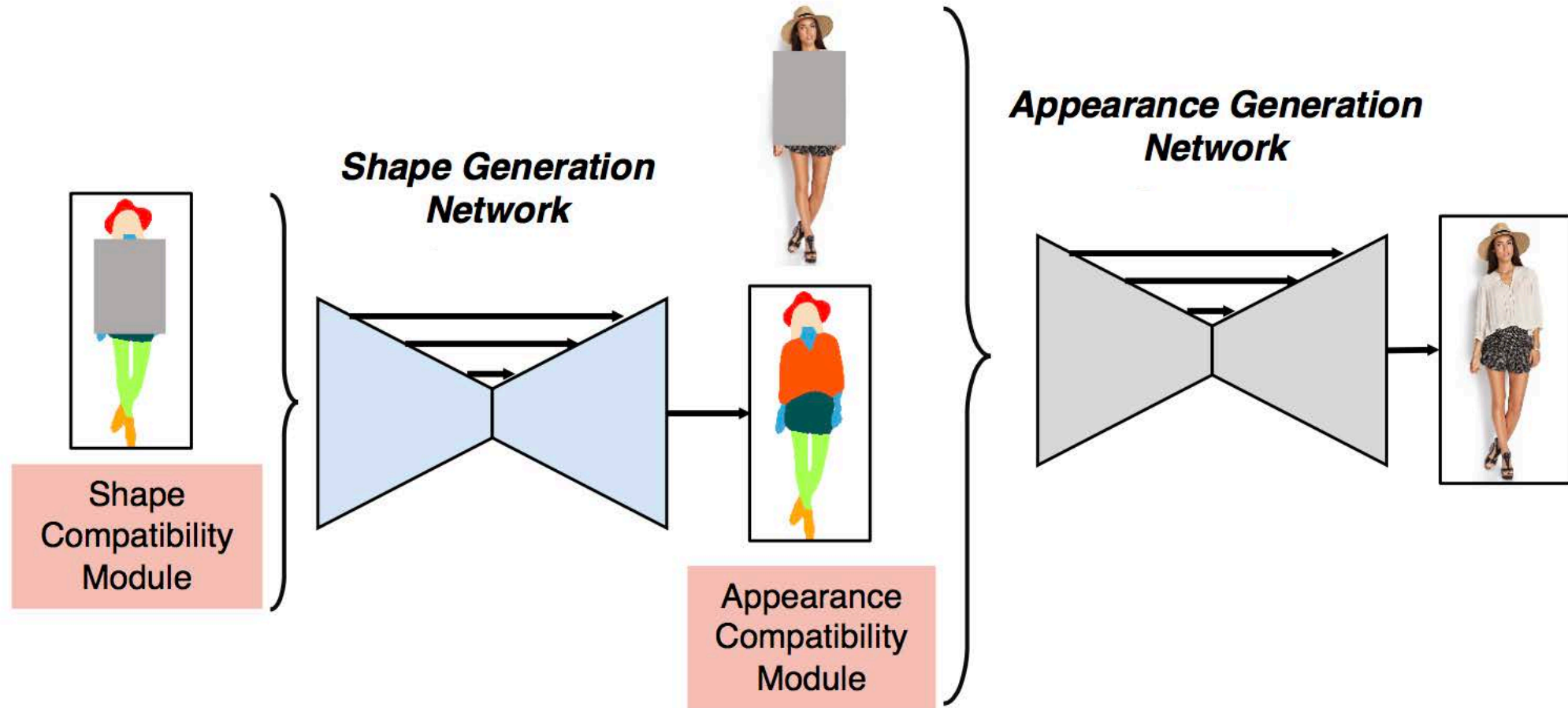


**Compatibility:** Matching style of existing garments

**Diversity:** Multiple results should be generated

**Shape and Appearance**

# Framework

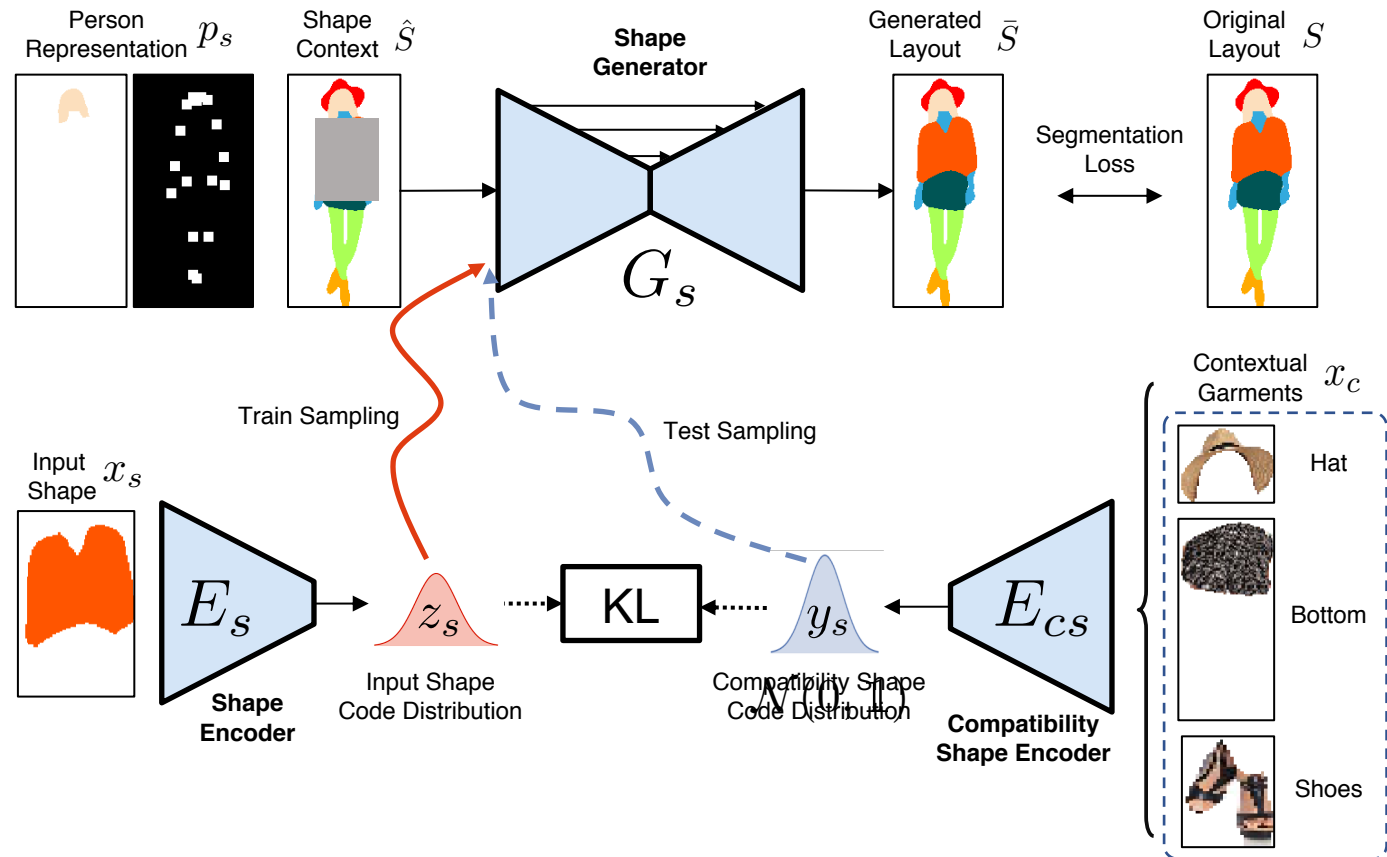


Two-stage framework that disentangles the generation of **shape** and appearance.

# Shape Generation Network

## Image-to-image Translation (U-Net):

Person representation + Shape context -> Layout



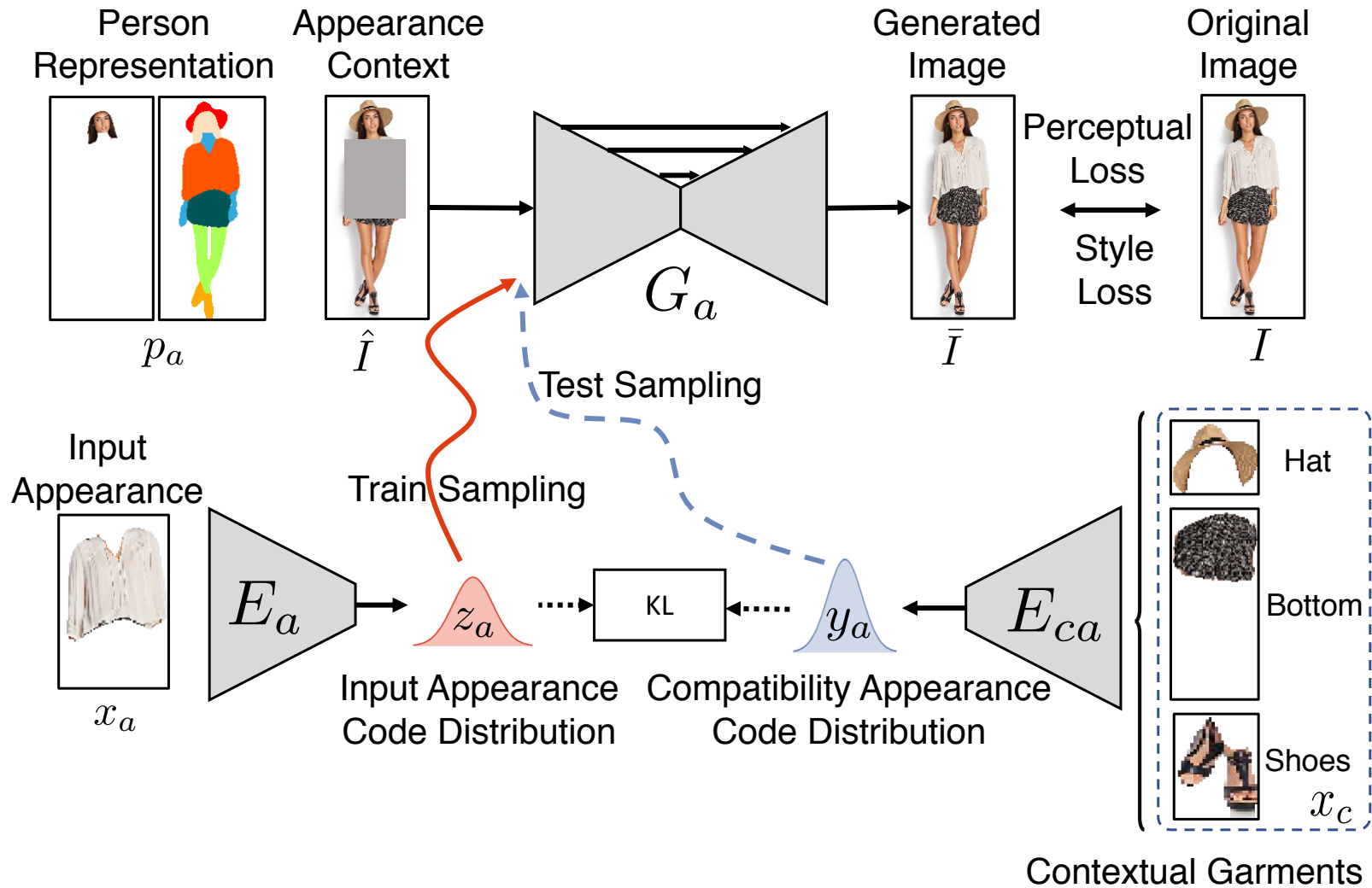
## +Variational Autoencoder (VAE):

Enable diversity at test time

## +Compatibility Encoder:

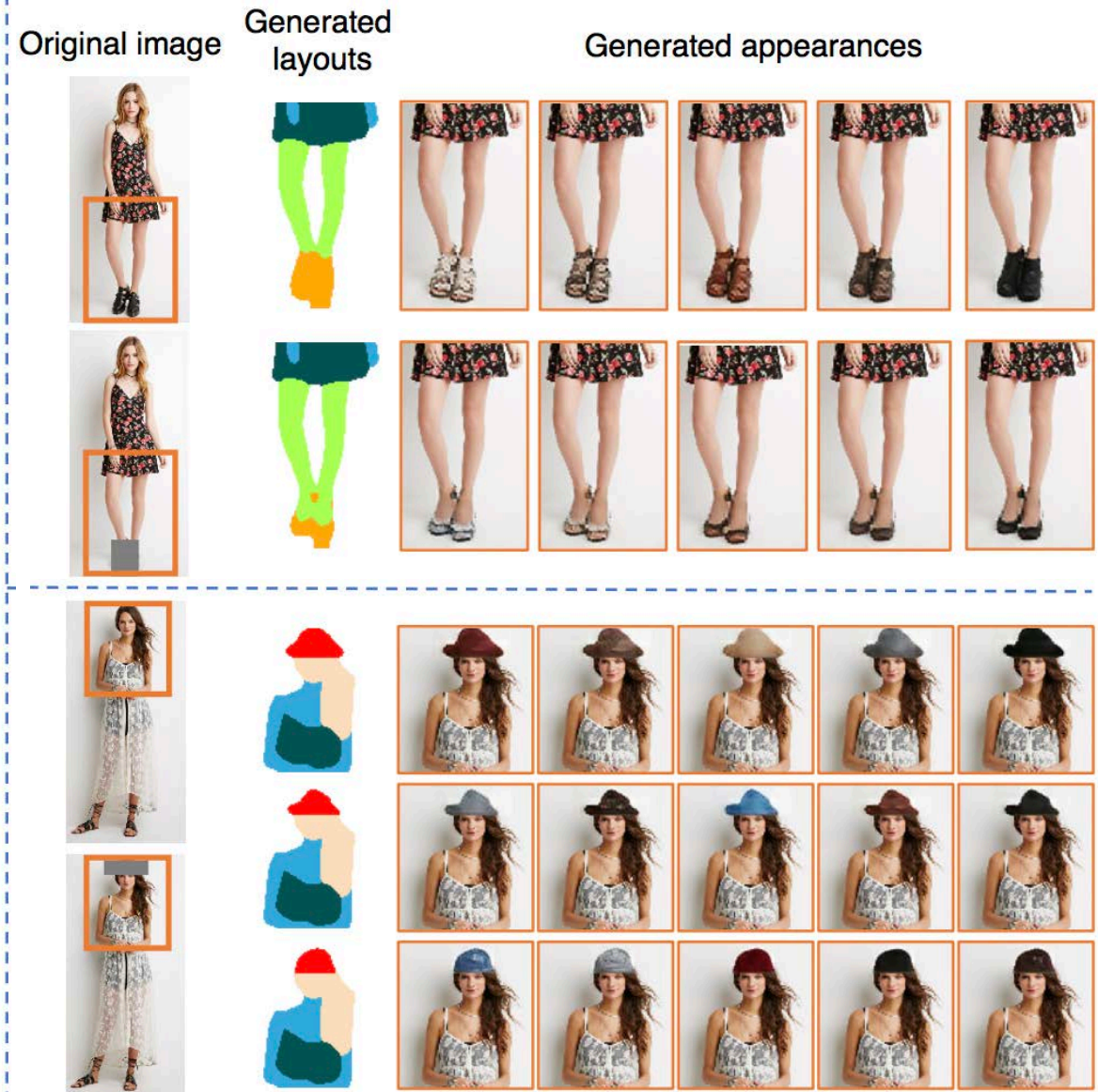
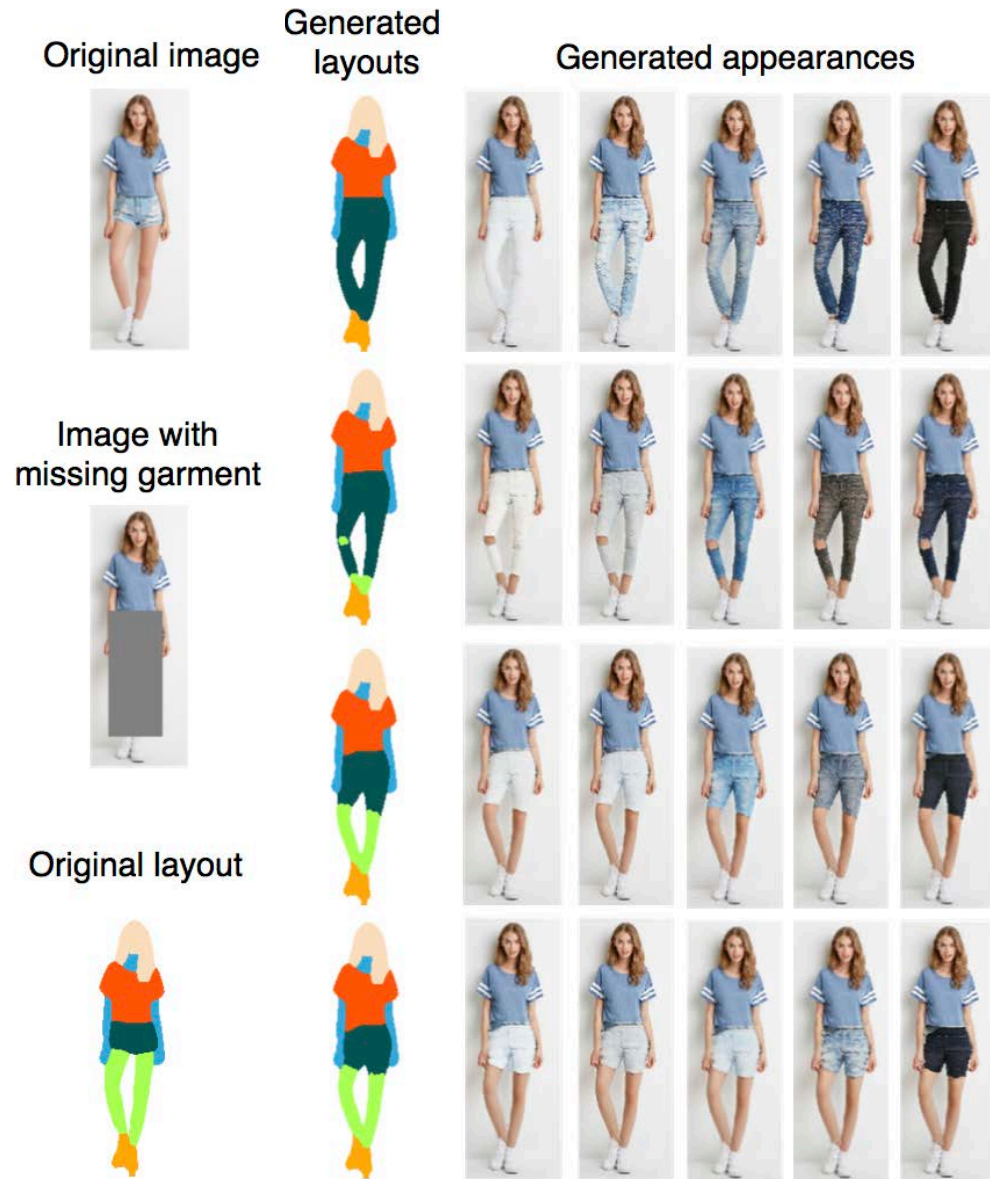
Inject compatibility (co-occurrence) using contextual garments

# Appearance Generation Network





# Inpainting Results



# Comparison with Other Methods

Original image and input



FiNet



FiNet w/o two-stage



FiNet w/o two-stage w/o comp



Pix2Pix + noise



BicycleGAN



VUNET

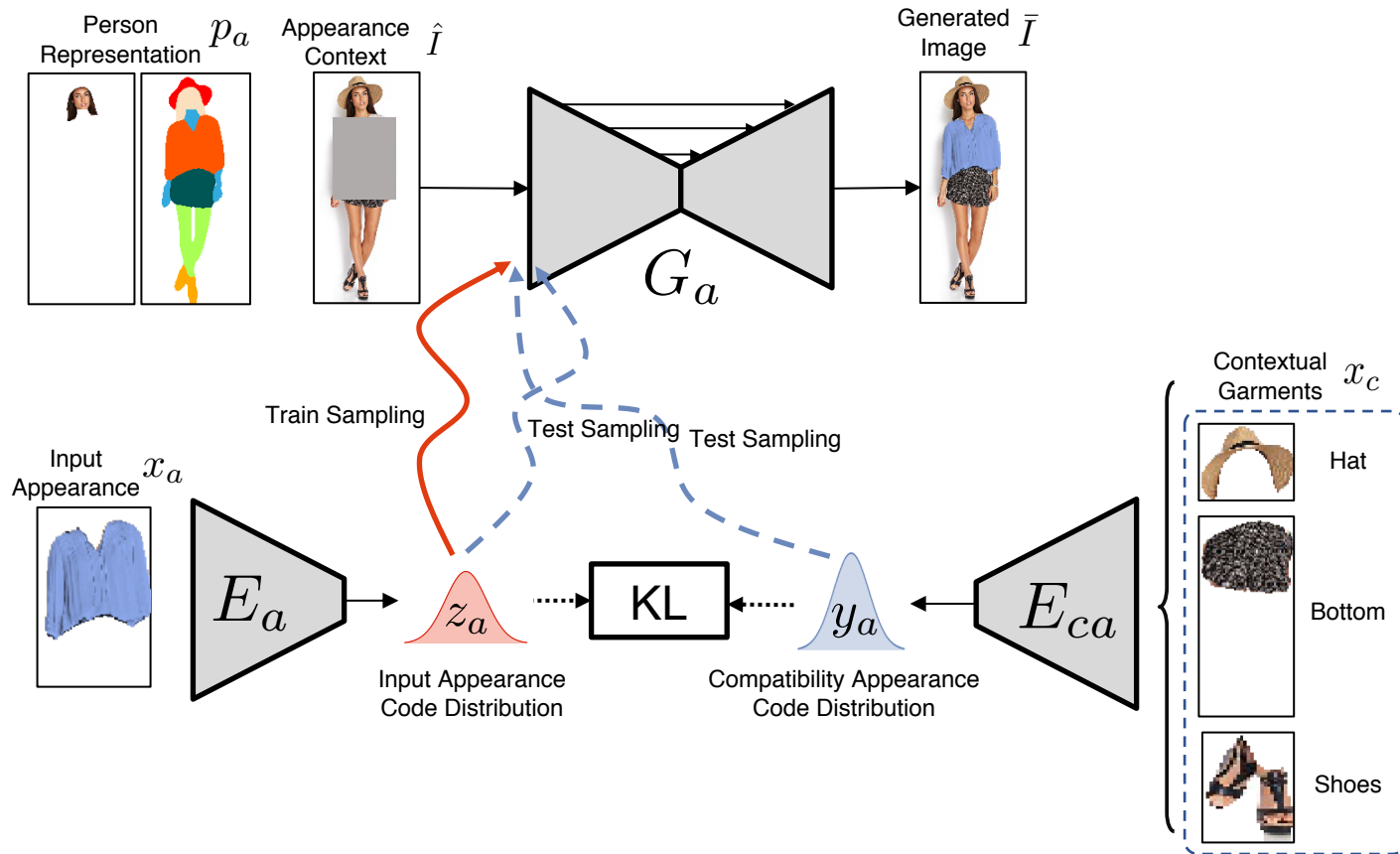


ClothNet





# Virtual Try-on Applications



**Disable compatibility encoder.**

**Test sampling from shape/appearance encoder.**

**Input a target clothing for reconstruction.**

# Virtual Try-on Results

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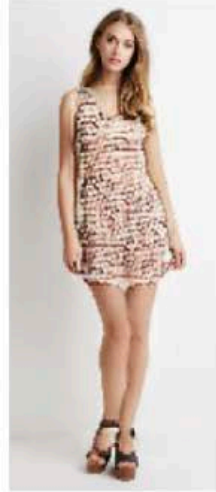
Reference



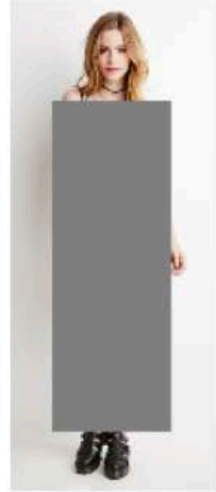
Input



Transfer



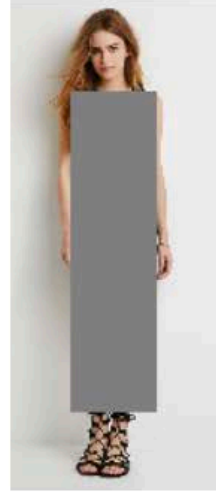
Input



Transfer



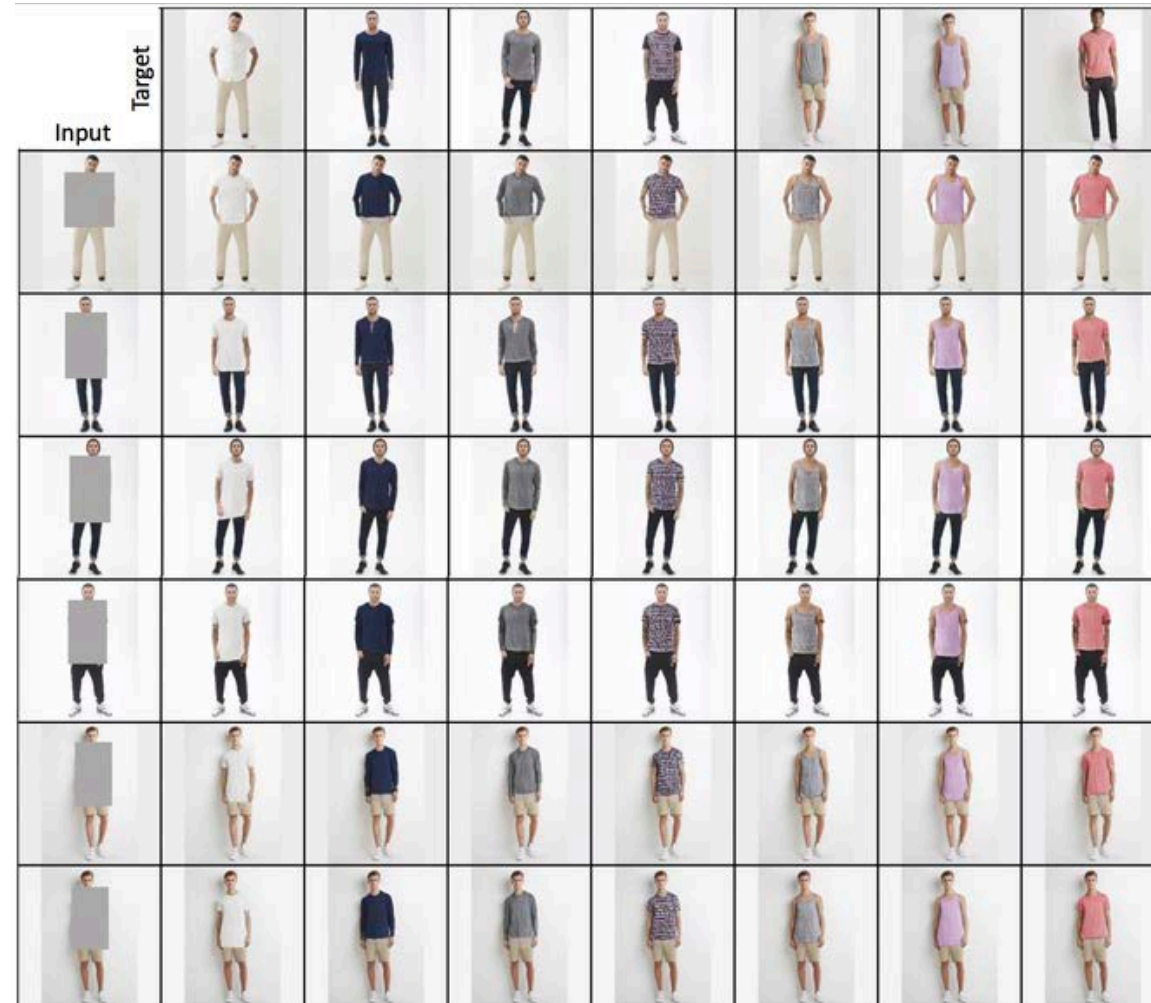
Input



Transfer



# Virtual Try-on Results

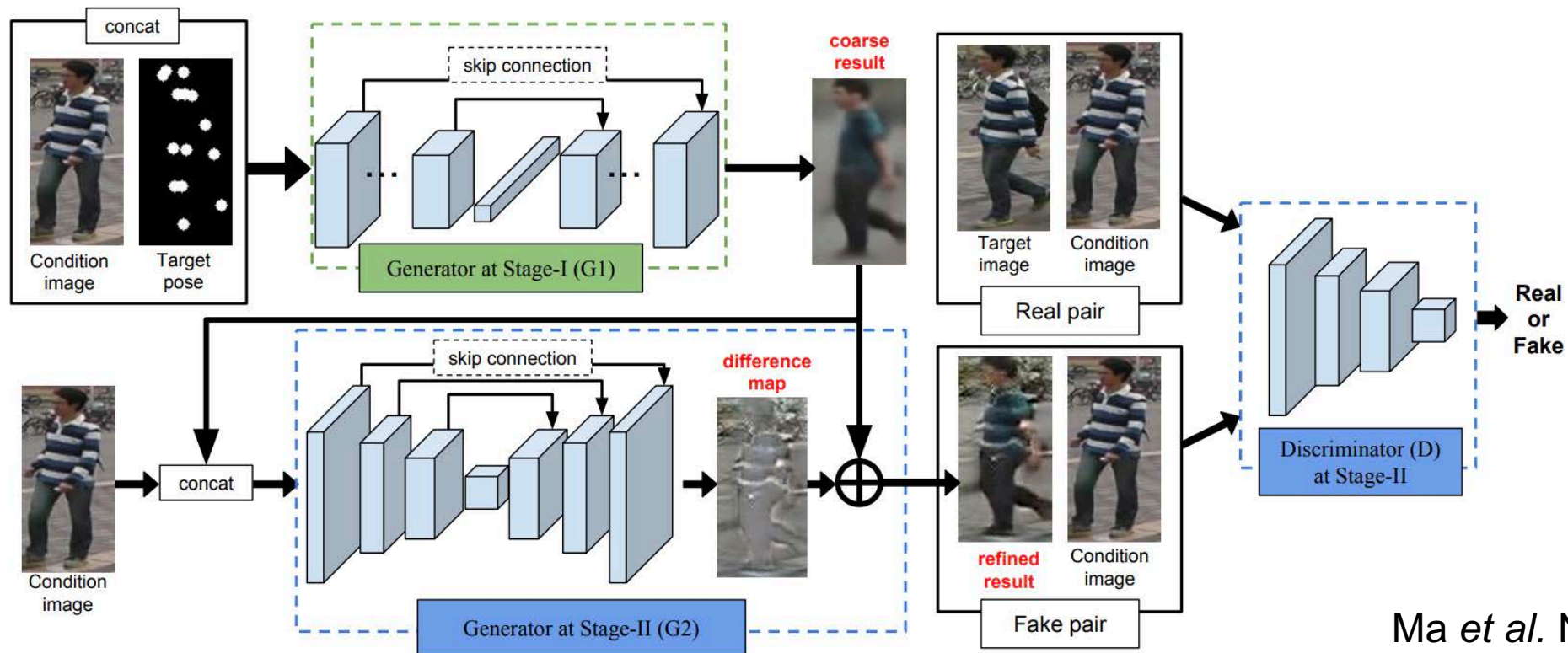


02

## ClothFlow: A Flow-Based Model for Clothed Person Generation

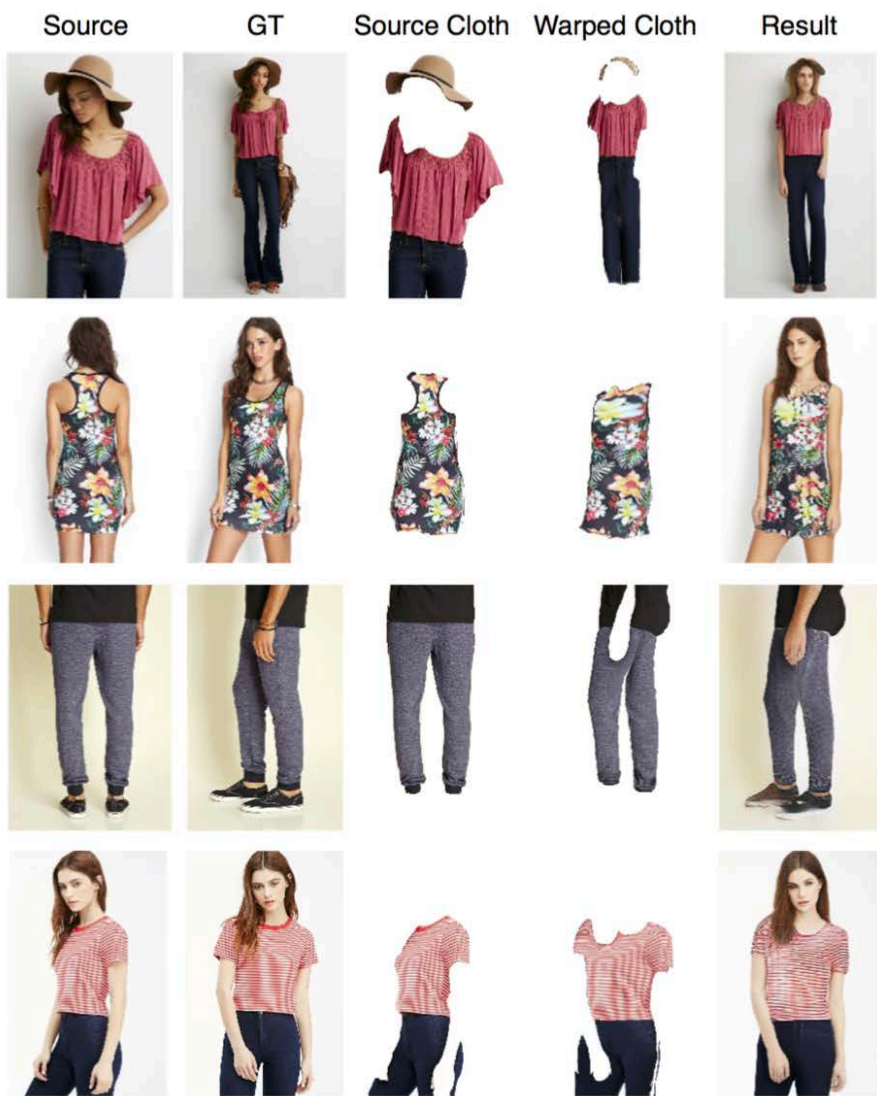


# Pose-guided Person Image Generation

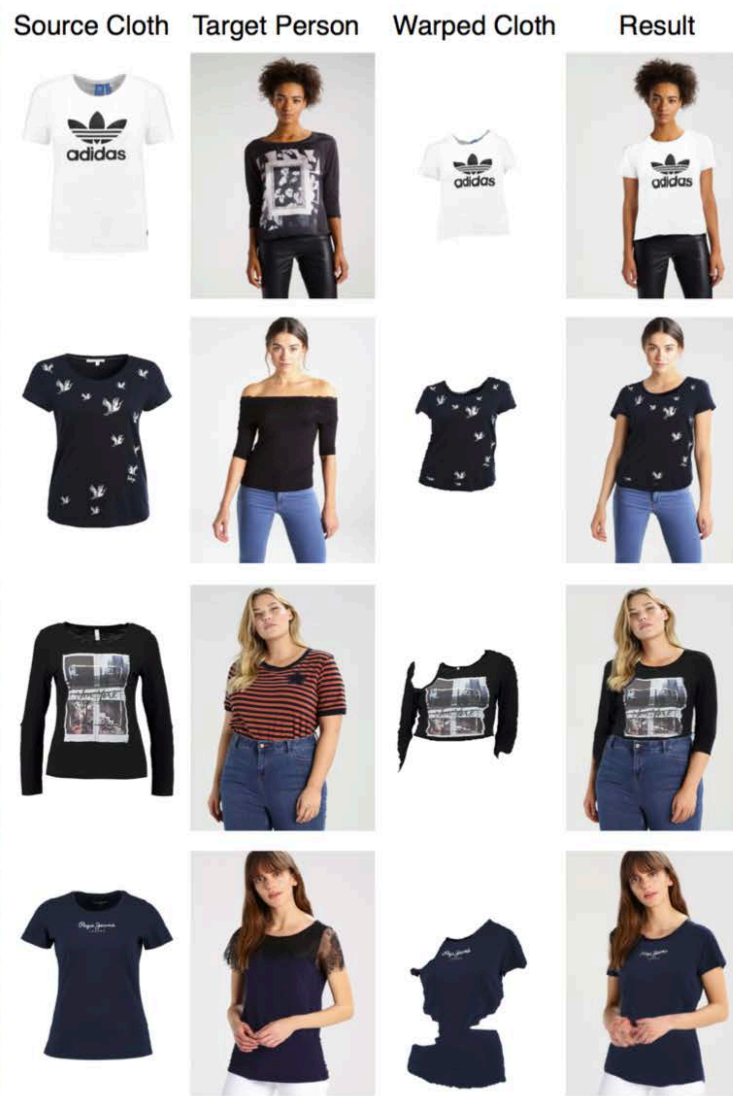




# Our Results



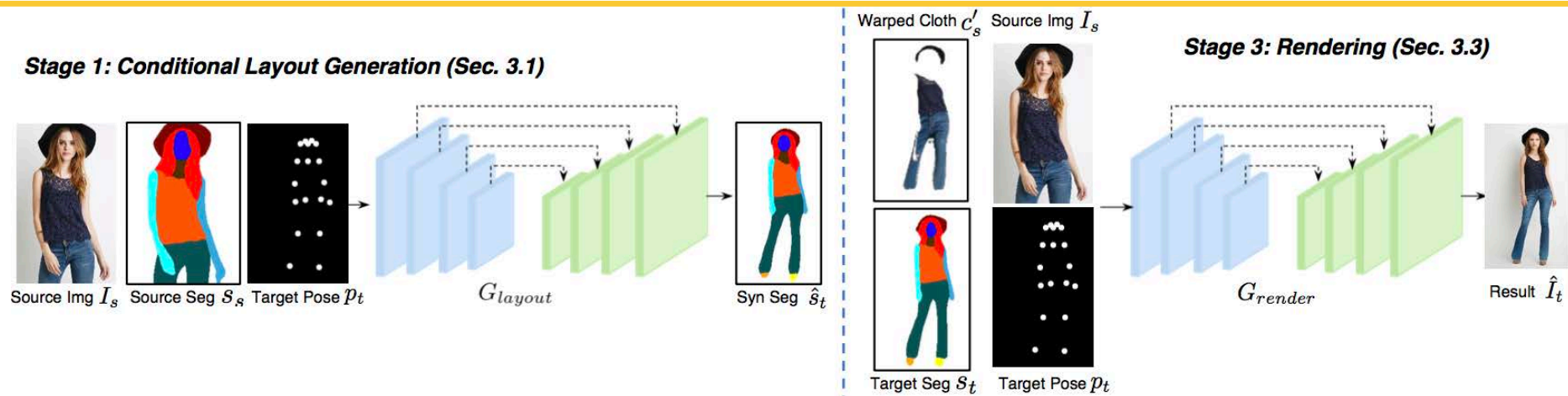
Pose guided person generation



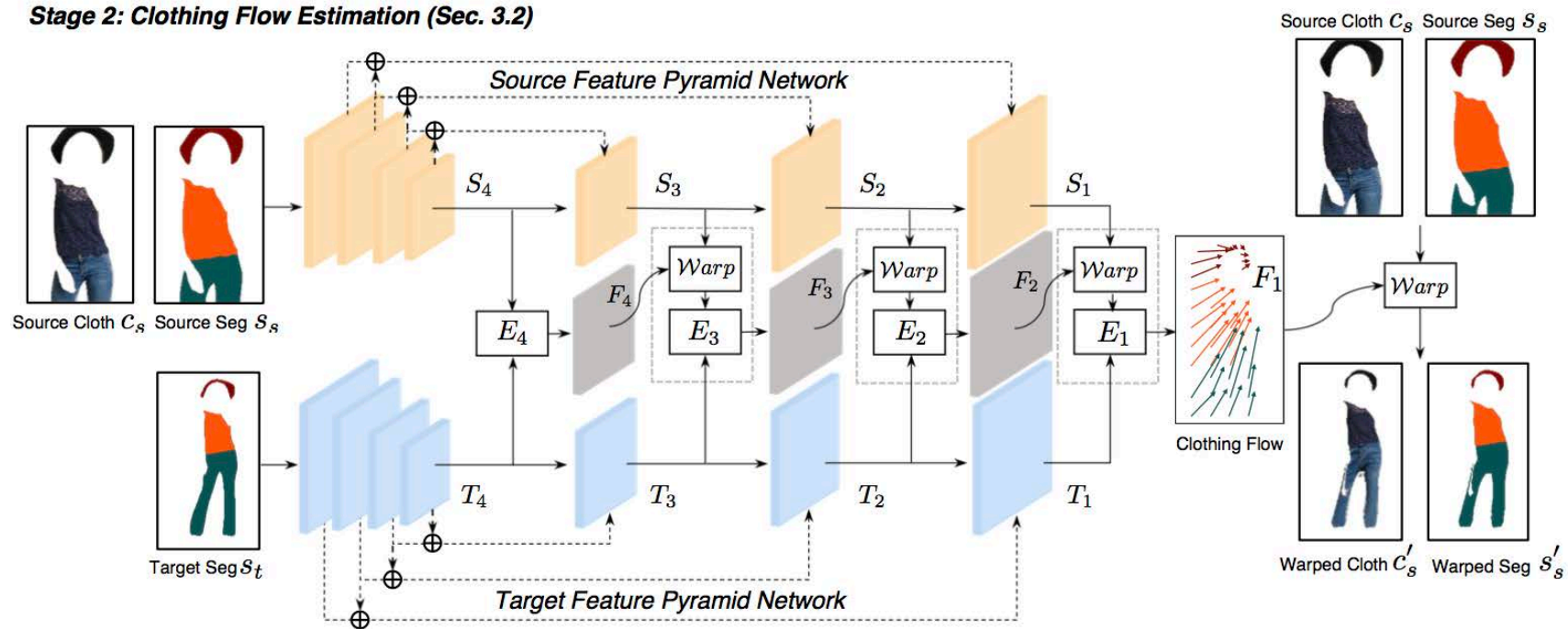
Virtual try-on

- Warp clothing regions by estimating **appearance flow** between two images.
- Appearance flow: a **2D dense flow field** specifying which pixels in the source image could be **redirected** to reconstruct the target image.

# ClothFlow Framework

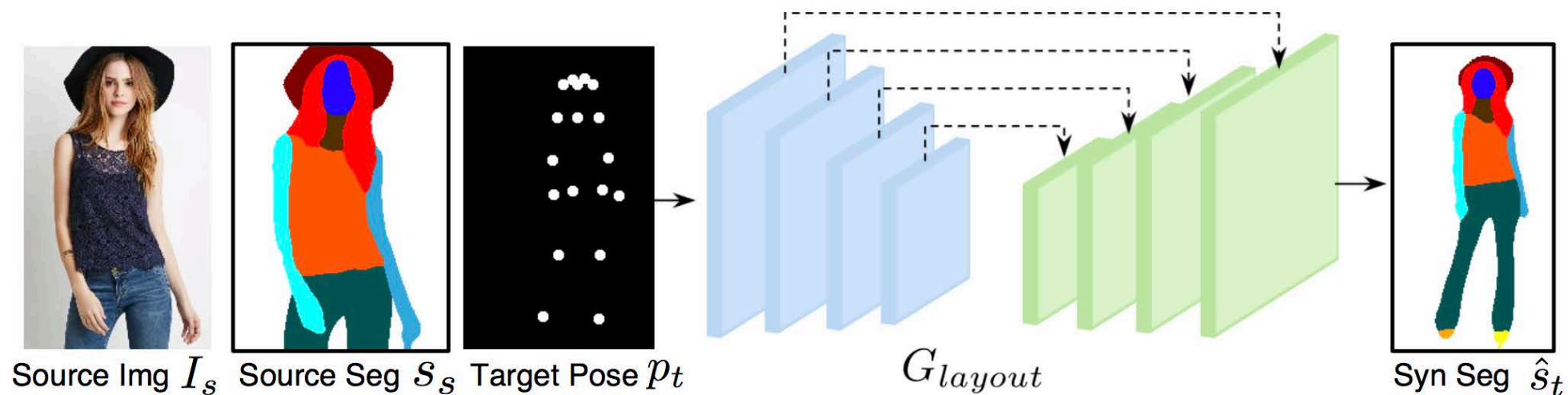


**Stage 2: Clothing Flow Estimation (Sec. 3.2)**



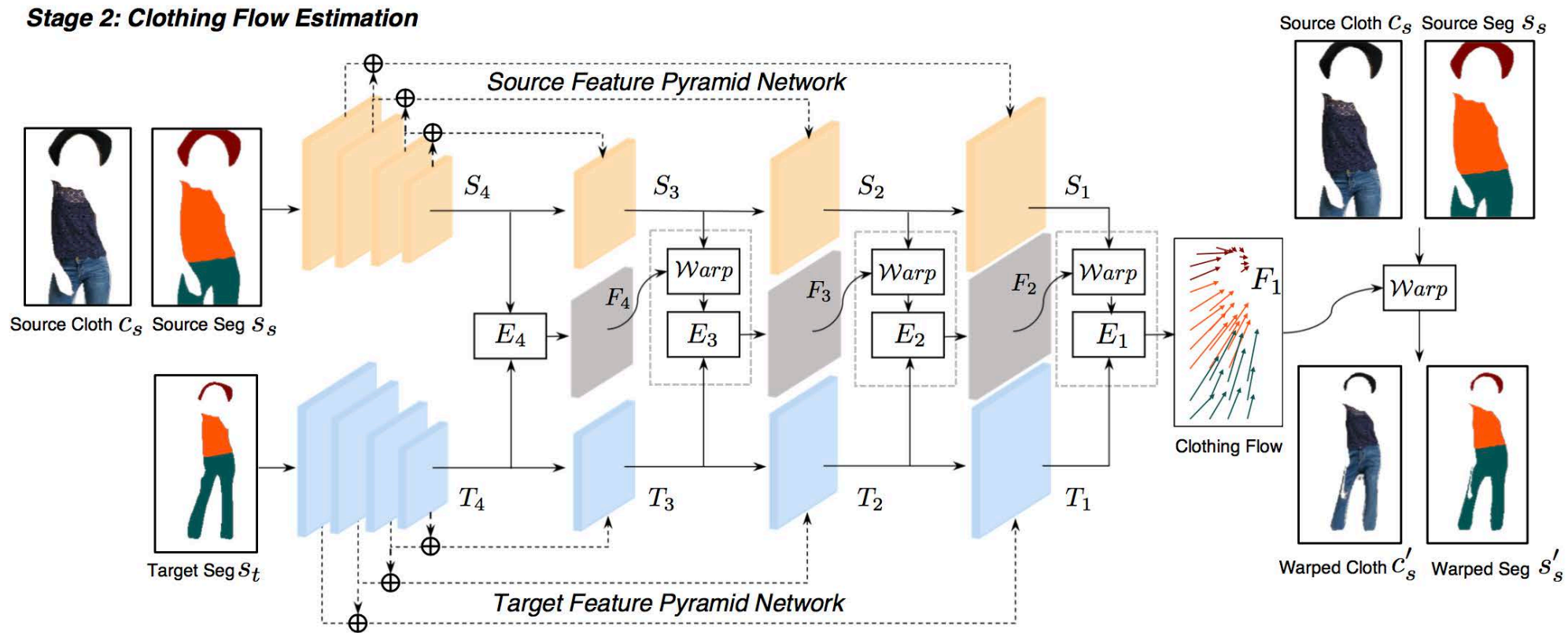
# Hallucinate Target Segmentation Map

## Stage 1: Conditional Layout Generation



- Extract **poses** and **segmentation maps** with off-the-shelf models.
- U-Net predicts the **target segmentation map** with segmentation loss.
- Provide **structural constraints** for generating appearance.

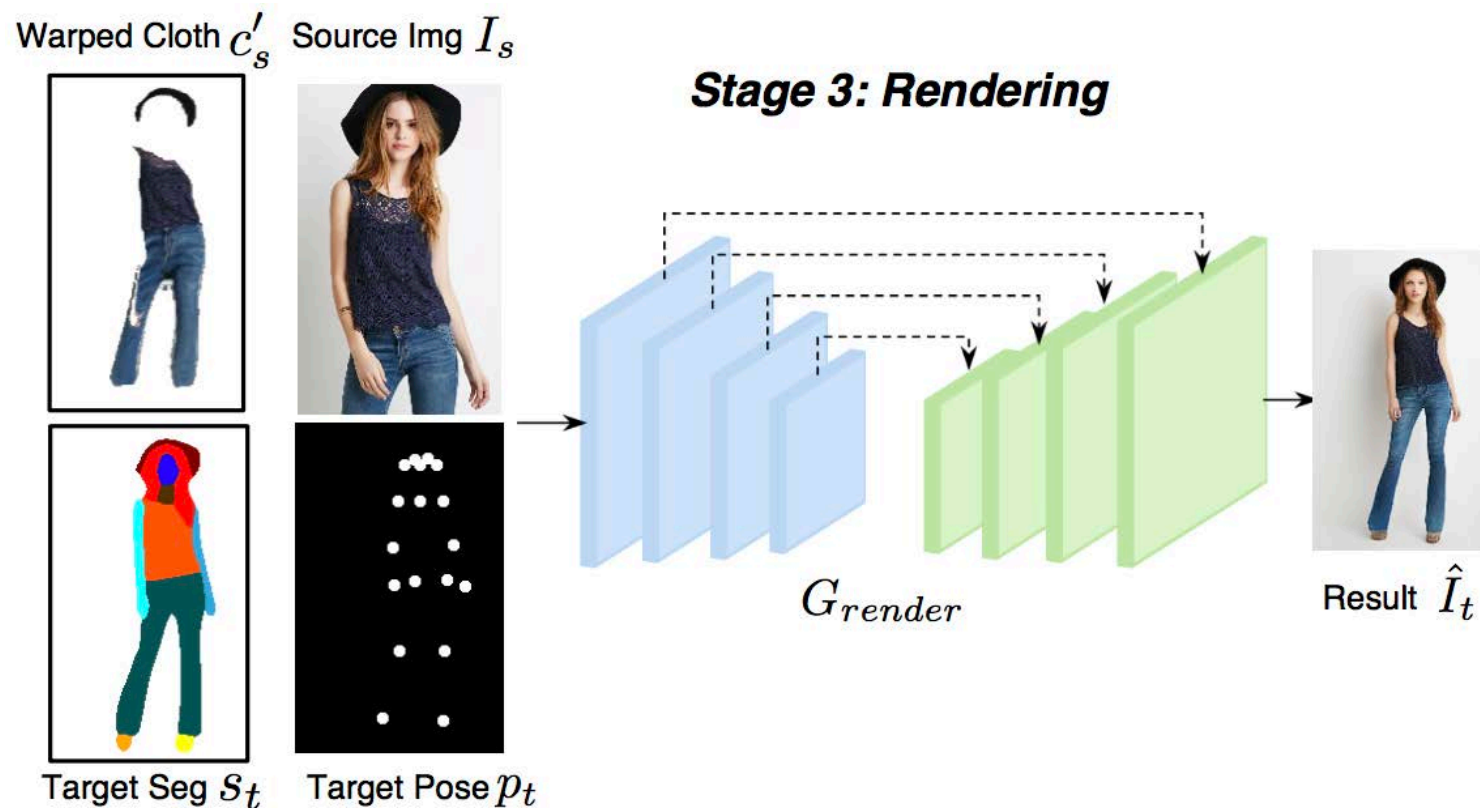
# Estimate Flow between Source and Target



- Borrow idea from **optical flow** estimation.
- **Dual feature pyramid** networks to encode source and target information.
- Gradually estimate the appearance flow in a **cascaded** fashion.
- Warped clothing: **perceptual loss**; Warped segmentation: **L1 loss**.



# Render the Final Result



- U-Net taking all guidance to synthesize the final result.
- Train: use GT target segmentation. Testing: Synthesized segmentation from stage 1.



# Pose-guided Person Generation Results



Quantitative comparison

Methods	Deform	Dense	SSIM	IS
PG <sup>2</sup> [26]	✗	✗	0.762	3.09
DSC [34]	✓	✗	0.761	3.35
VUNET [7]	✗	✗	0.786	3.09
BodyROI7 [27]	✗	✗	0.614	3.23
DPT [28]	✗	✓	0.785	3.61
Soft-Gated [4]	✓	✗	0.793	3.31
CBI [10]	✓	✓	<b>0.835</b>	2.92
w/o Layout	✓	✗	0.758	3.63
w/o Flow	✗	✗	0.757	3.71
w/o Flow + TPS	✓	✗	0.758	3.74
w/o Cascade	✓	✗	0.759	3.74
w/o Style	✓	✗	0.756	3.56
ClothFlow	✓	✗	0.760	3.75
ClothFlowDense	✓	✓	0.771	<b>3.88</b>

Pair-wise user preference

DSC [34]	VUNET [7]	DPT [28]	CBI [10]	Soft-Gated [4]
80.9%	63.4%	90.2%	69.7%	57.9%



# Virtual Try-on: Replace the source image by a clothing image



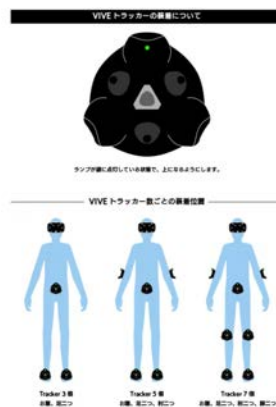
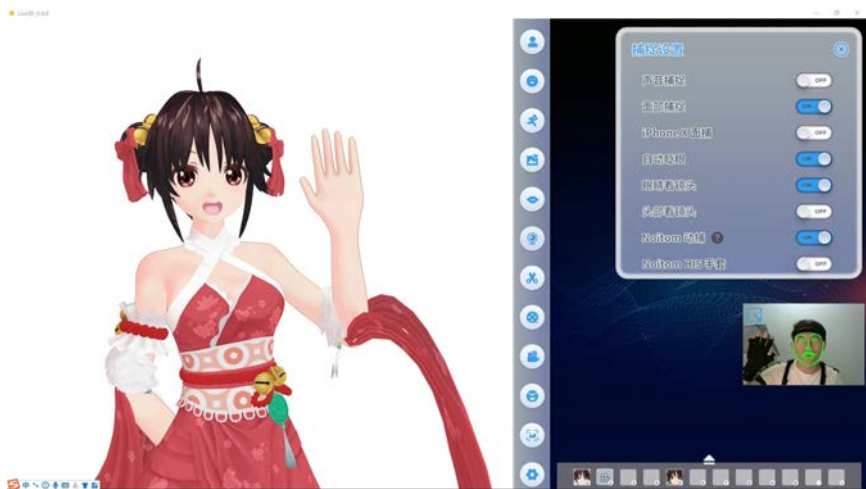
03

## Content Creation at Huya



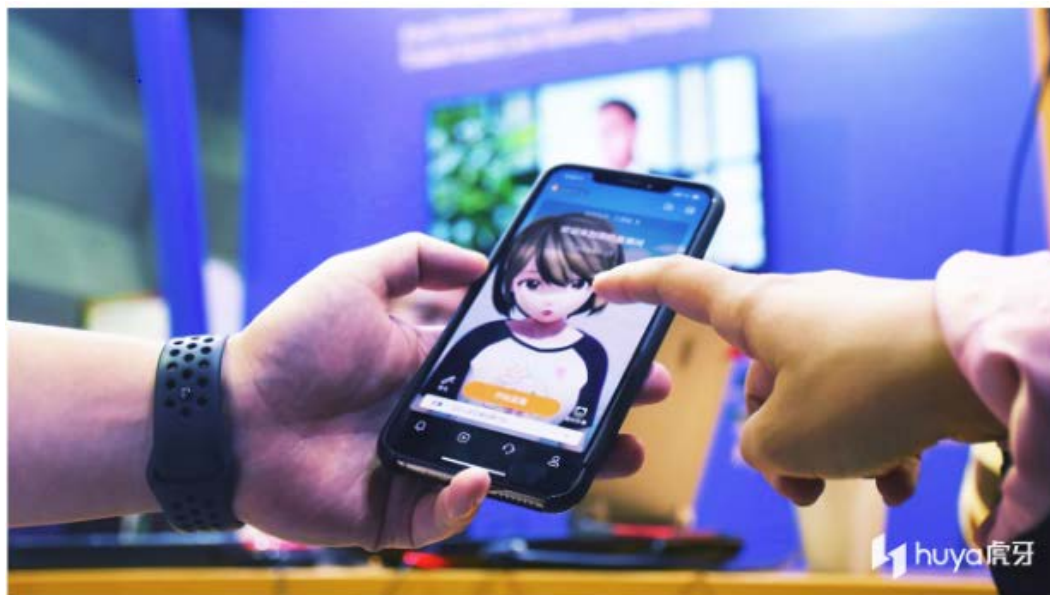


# 虚拟直播风声水起



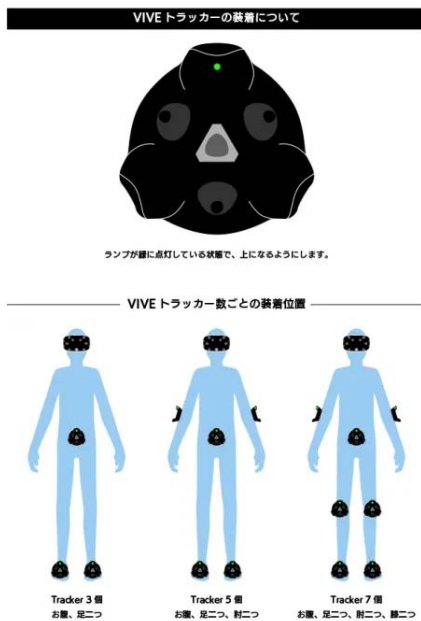


# 虎牙的虚拟直播





# 存在问题



依赖动捕设备



虚拟形象自由度低



限制小主播，缺少新玩法

# 低成本方案：基于单张图片的人脸驱动虚拟直播

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第一步：上传单张图片



第二步：单目摄像头直播



第三步：动起来



## 更多结果



该算法有两种模式：

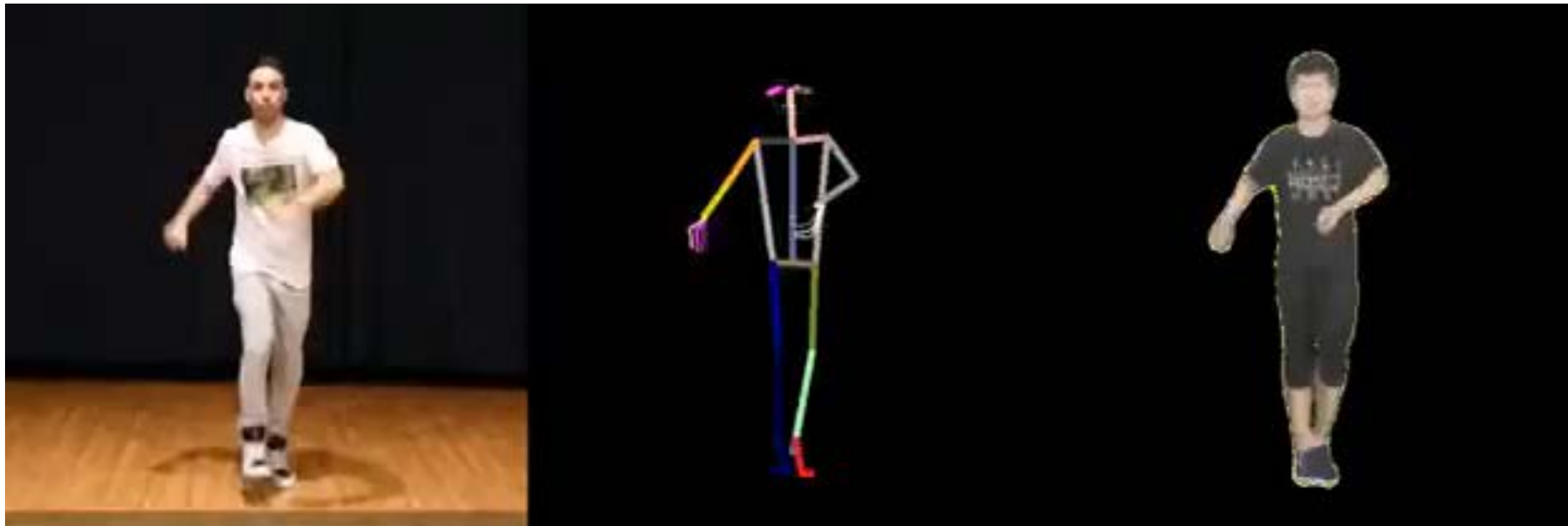
左：无法很好的保持身份，头可以自由转动

右：头不能自由转动，但是身份保持得很好



# 肢体驱动

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联系方式：[hanxintong@huya.com](mailto:hanxintong@huya.com)



**THANK YOU**

