



# SCAM! Transferring humans between images with Semantic Cross Attention

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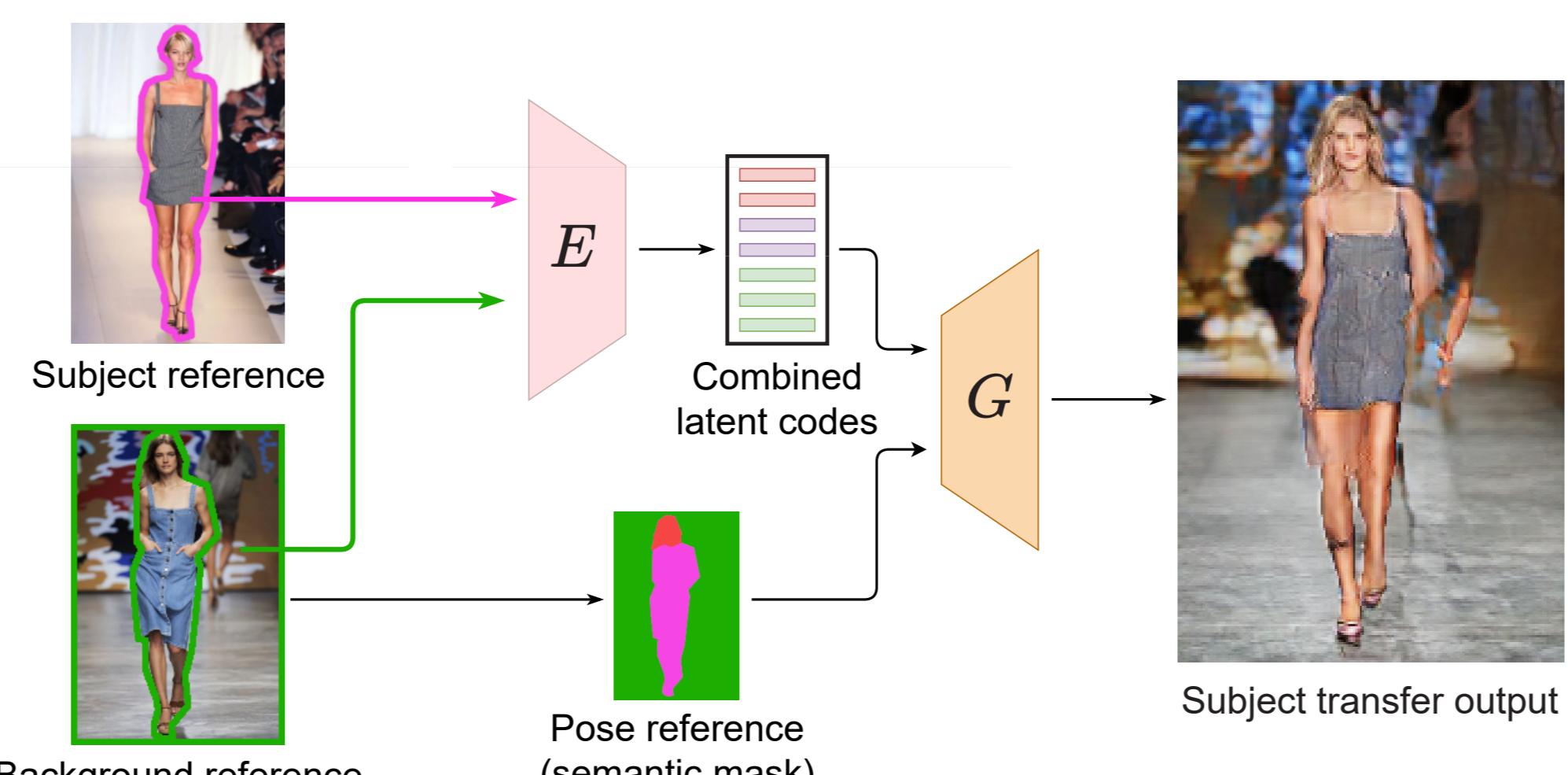
## Introduction

**Goal:** Easily swap subjects in an image.

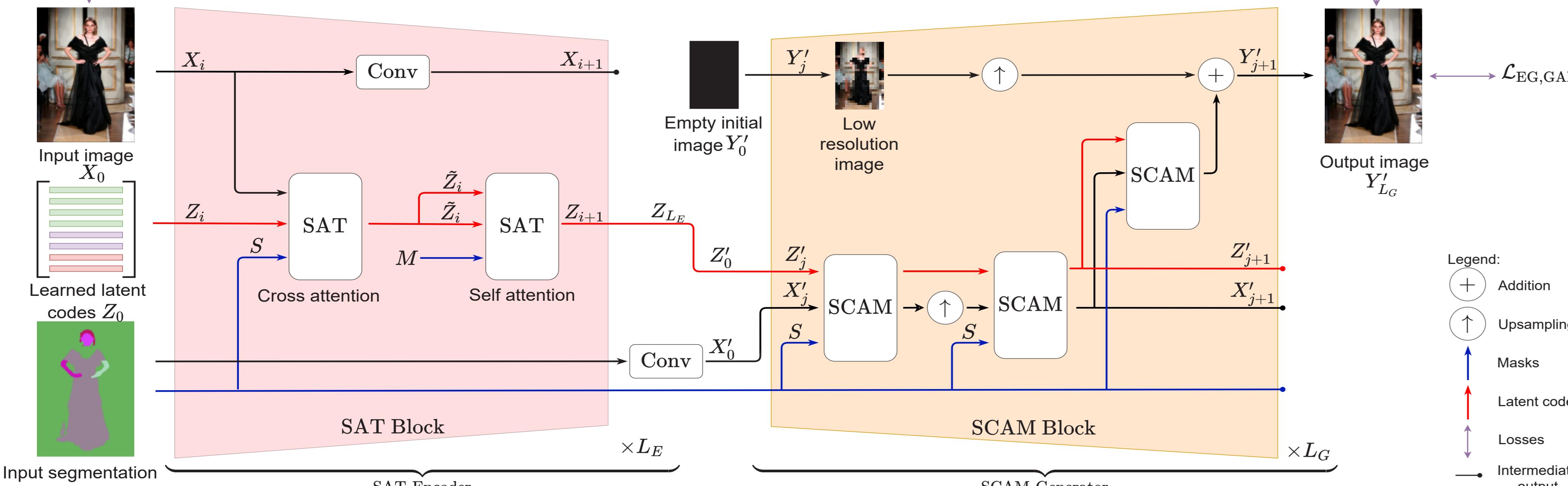
**Approach:** Separate the latent space with a *segmentation mask*.

**Prior work:** Single style code per image [ParkCVPR19],  
Single style code per segmentation region [ZhuCVPR20].

**Our method:** Multiple style codes per semantic region.



## Method



Training setup of the proposed SCAM architecture.

### Semantic Cross Attention (SCA)

**SCA:** semantically constrained attention on pixels and latents.

- Pixel and group of latents have an *assigned* semantic label.
  - Pixels or latents can only attend same semantic label latents.
- $$\text{SCA}(I_1, I_2, I_3) = \sigma\left(\frac{QK^T \odot I_3 + \tau(1 - I_3)}{\sqrt{d_m}}\right)V$$
- $$Q = W_Q I_1, K = W_K I_2 \text{ and } V = W_V I_2$$

### Semantic Cross Attention Modulation (SCAM)

**SCAM:** Multiple latents per semantic region.

- More representative power for *coarse* semantic regions.
- Discover *unsupervised semantic structures* inside the labels.

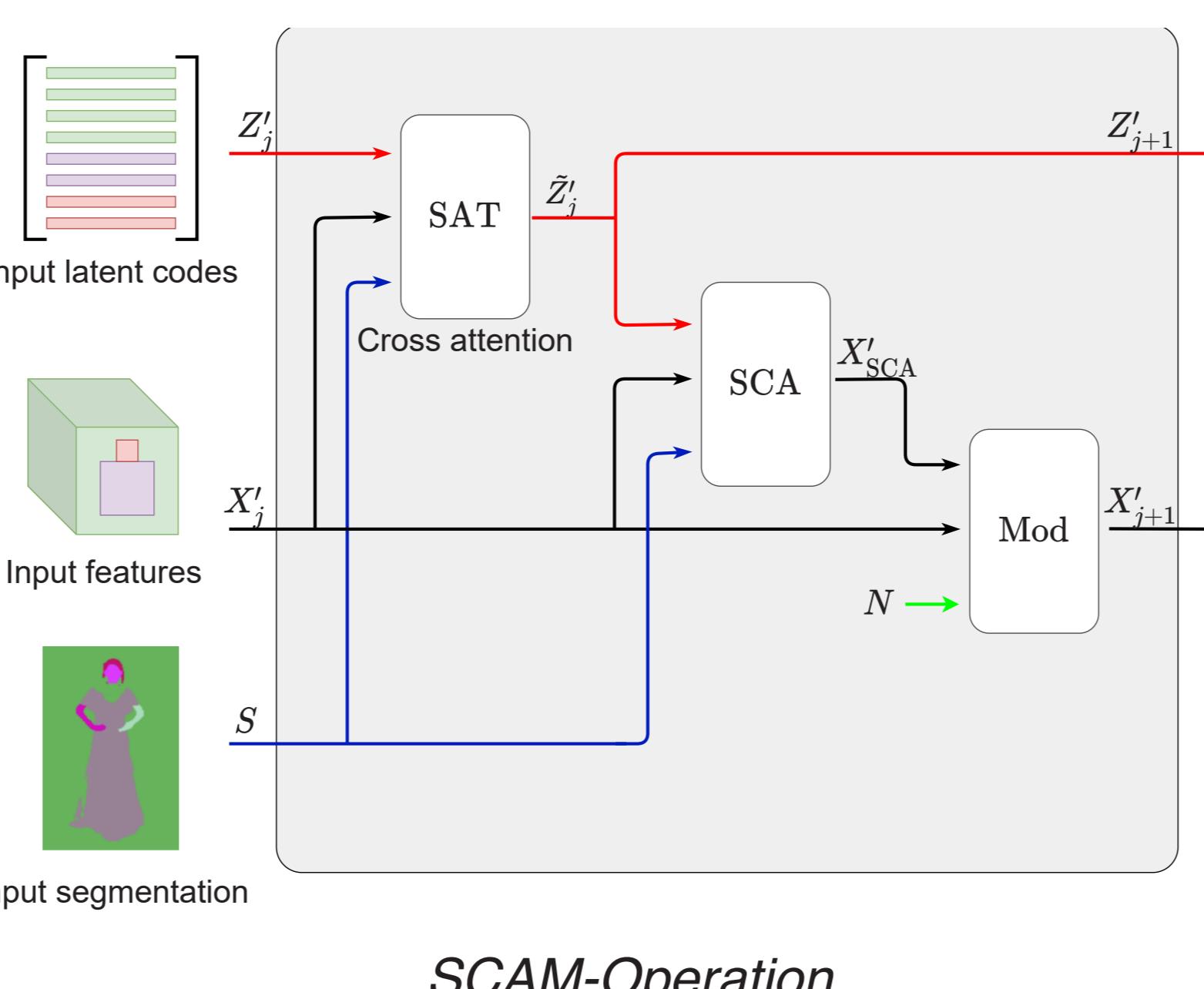
### Training setup

- Hinge GAN loss + L1 loss + Perceptual Loss.
- PatchGAN discriminator.

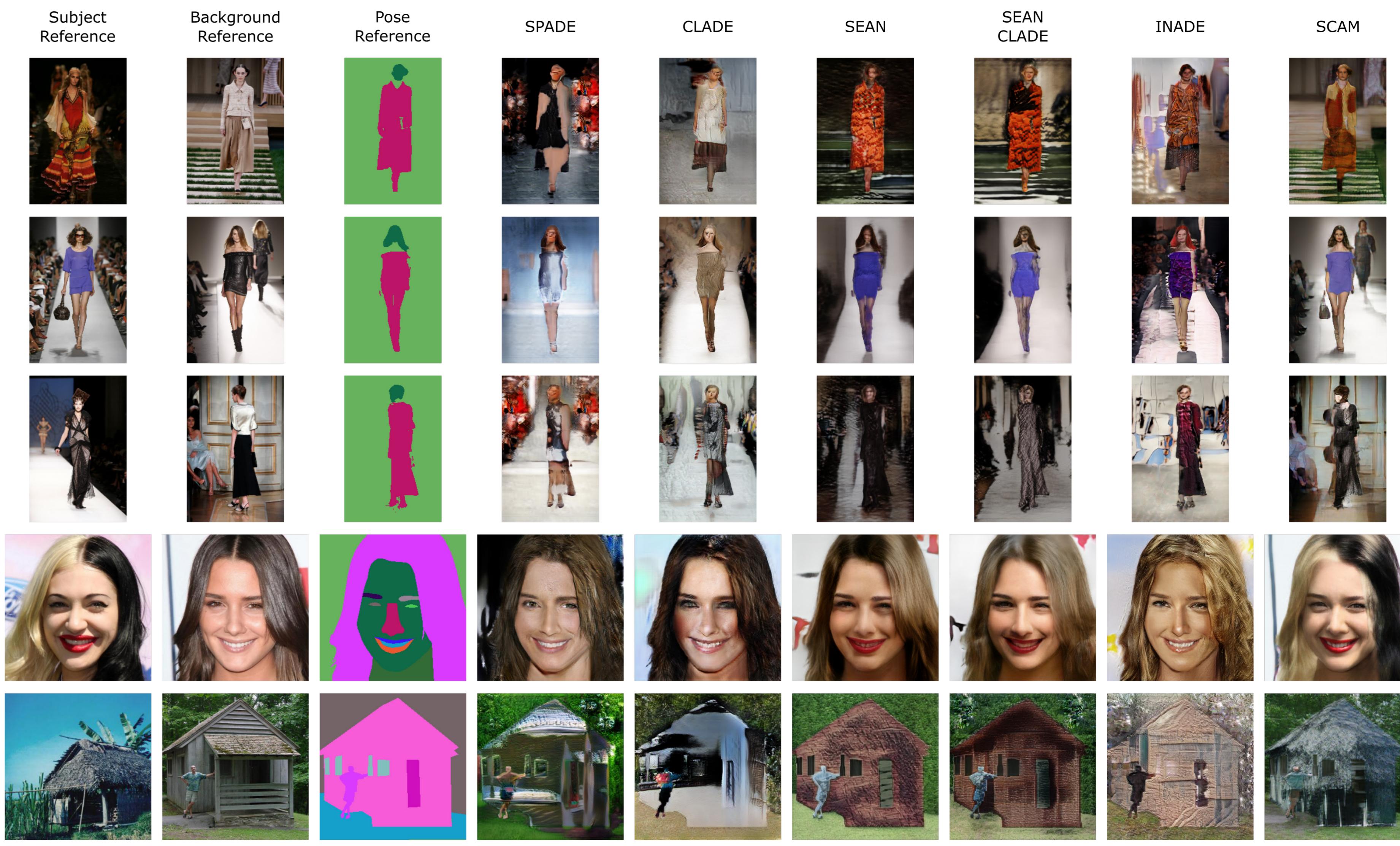
### Semantic Attention Transformer (SAT)

**SAT Operation:** *transformer-like* operation with SCA.

- SAT encode/decode data in a *semantically constrained* fashion.



## Subject Transfer Results



## Unsupervised Semantic Knowledge



## Quantitative Results

Method	iDesigner				CelebAMask-HQ				ADE20K			
	PSNR ↑ R-FID ↓	S-FID ↓	REIDSim ↑	REIDAcc ↑	PSNR ↑ R-FID ↓	S-FID ↓	PSNR ↑ R-FID ↓	PSNR ↑ R-FID ↓	PSNR ↑ R-FID ↓	S-FID ↓	PSNR ↑ R-FID ↓	PSNR ↑ R-FID ↓
SPADE [ParkCVPR19]	10.4	66.7	67.5	0.67	0.26	10.9	38.2	38.3	10.7	59.7		
CLADE [TanTPAMI21]	11.3	45.4	46.1	0.68	0.29	10.8	41.8	42.0	10.4	53.7		
SEAN-CLADE [TanTPAMI21]	15.3	48.4	56.1	0.75	0.31	16.2	19.8	24.3	14.0	38.7		
INADE [TanCVPR21]	12.0	33.0	33.9	0.72	0.34	12.24	22.7	23.4	11.3	48.6		
SEAN [ZhuCVPR20]	14.9	53.5	58.7	0.74	0.30	16.2	18.9	22.8	14.6	47.6		
SCAM (Ours)	<b>21.4</b>	<b>13.2</b>	<b>26.9</b>	<b>0.81</b>	<b>0.56</b>	<b>21.9</b>	<b>15.5</b>	<b>19.8</b>	<b>20.0</b>	<b>27.5</b>		