

1. Problem

◆ Attribute to Facial Image

blond hair
female
mouth slightly open
arched eyebrows
heavy makeup



2. Related Work and Motivation

◆ Text to image

- [StackGAN++](#), [AttnGAN](#), [MirrorGAN](#), etc.
- the input is different: text vs attribute
- can not well embed the attribute label

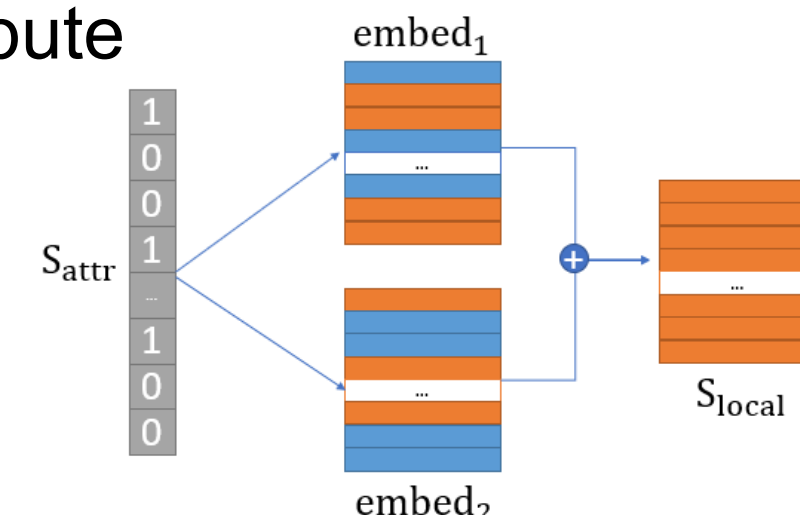
◆ Attribute to image

- [Attribute2sketch2face](#), [Lu et al.](#), [Wang et al.](#), etc.
- the generated images are always low resolution
- do not consider the relationship between different attributes

4. Details

◆ AEM: Attribute Embedding Module

- convert the input face attributes into global and local features respectively
- two path embedding
 - well reflect their meanings of the input attribute



$$S_{local} = embed_1 * S_{attr} + embed_2 * (1 - S_{attr})$$

□ self attention layer

- model the relationships between different attributes

$$f(x) = W_f * S_{local}$$

$$s_{ij} = f(x_i)^T g(x_j)$$

$$h(x) = W_h * S_{local}$$

$$S_{local'} = (S_{local'_1}, S_{local'_2}, \dots, S_{local'_N}) \in \mathbb{R}^{C \times N}$$

$$g(x) = W_g * S_{local}$$

$$\beta_{ij} = \frac{\exp(s_{ij})}{\sum_{k=1}^N \exp(s_{ik})}$$

$$S_{local'_j} = \sum_{i=1}^N \beta_{ij} h(x_i)$$

$$S_{local'} = (S_{local'_1}, S_{local'_2}, \dots, S_{local'_N}) \in \mathbb{R}^{C \times N}$$

◆ SIGM: Stacked Image Generation Module

- gradually generate faces with more details through a three-stage generator
- can generate images with high resolution

$$h_0 = F_0(z, F^{ca}(S_{global}))$$

$$h_i = F_i(h_{i-1}, F_i^{attn}(S_{local'}, h_{i-1}))$$

$$x_i = G_i(h_i)$$

◆ SCM: Similarity Constrain Module

- encode the generated images with a pretrained model: i_{local} and i_{global}
- calculate the matching degree between attribute features and image features
- the generated images can match the input attributes well

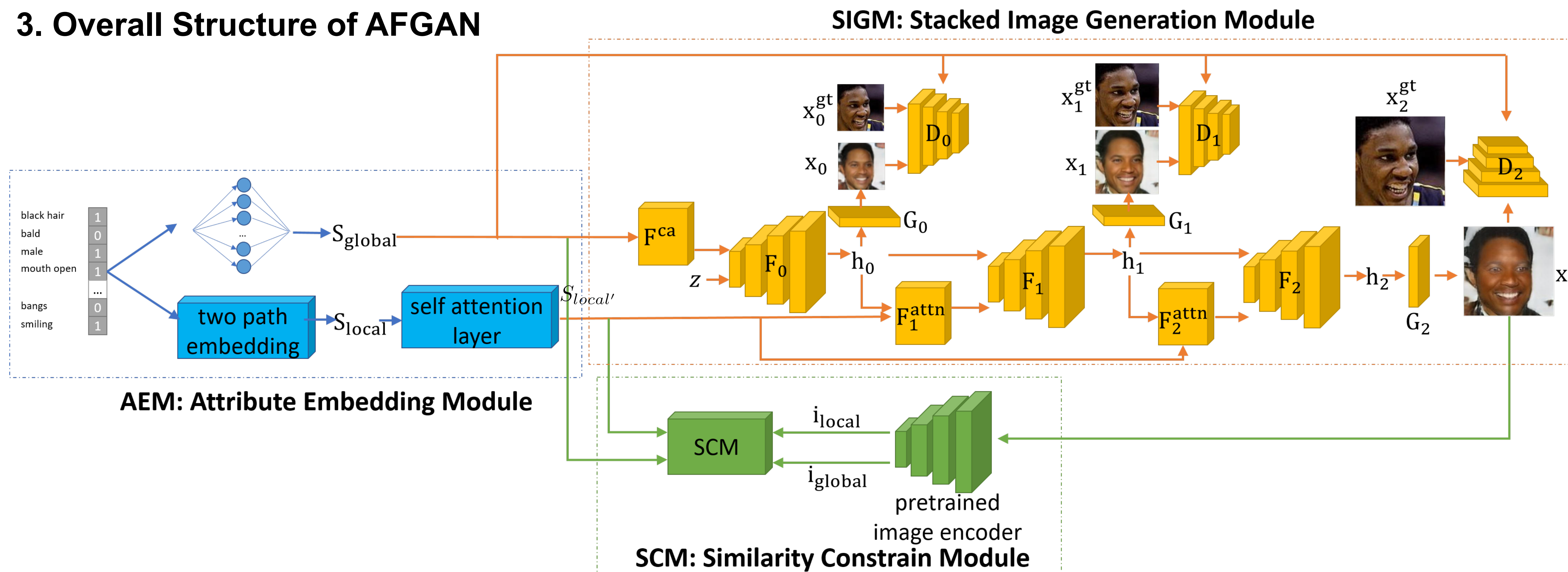
$$s = S_{local'}^T i_{local} \quad \bar{s}_{ij} = \frac{\exp(s_{ij})}{\sum_{k=1}^N \exp(s_{kj})} \quad \alpha_{ij} = \frac{\exp(\gamma_1 \bar{s}_{ij})}{\sum_{k=1}^2 89 \exp(\gamma_1 \bar{s}_{ik})}$$

$$c_i = \sum_{j=1}^{289} \alpha_{ij} i_{local_j} \quad R(c_i, S_{local'}) = \frac{c_i^T S_{local'}}{\|c_i\| \|S_{local'}\|}$$

$$R^{local}(Q, D) = \log \left(\sum_{i=1}^N \exp(\gamma_2 R(c_i, S_{local'})) \right)^{\frac{1}{\gamma_2}}$$

$$R^{global}(Q, D) = \frac{i_{global}^T S_{global}}{\|i_{global}\| \|S_{global}\|}$$

3. Overall Structure of AFGAN



5. Objective Function

◆ Generator

➢ Overall

$$\mathcal{L} = \mathcal{L}_G + \mathcal{L}_{SCM}$$

➢ In SIGM

$$\mathcal{L}_G = \sum_{i=0}^2 \mathcal{L}_{G_i}$$

$$\mathcal{L}_{G_i} = -\frac{1}{2} \mathbb{E}_{x_i \sim p_{G_i}} [\log(D_i(x_i))] - \frac{1}{2} \mathbb{E}_{x_i \sim p_{G_i}} [\log(D_i(x_i, S_{global}))]$$

➢ In SCM

$$\mathcal{L}_{SCM} = \mathcal{L}_1^{local} + \mathcal{L}_2^{local} + \mathcal{L}_1^{global} + \mathcal{L}_2^{global}$$

$$\mathcal{L}_1^{local} = -\sum_{i=1}^M \log P^{local}(D_i|Q_i)$$

$$\mathcal{L}_2^{local} = -\sum_{i=1}^M \log P^{local}(Q_i|D_i)$$

$$\mathcal{L}_1^{global} = -\sum_{i=1}^M \log P^{global}(D_i|Q_i)$$

$$\mathcal{L}_2^{global} = -\sum_{i=1}^M \log P^{global}(Q_i|D_i)$$

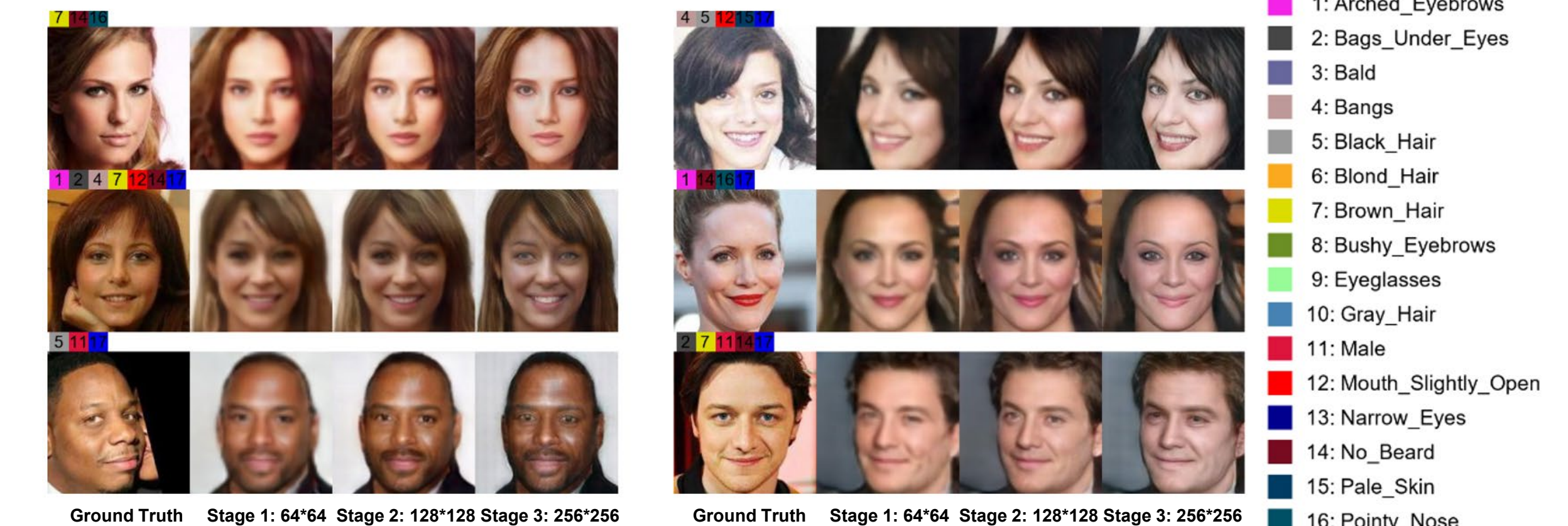
◆ Discriminator

$$\mathcal{L}_D = \sum_{i=0}^2 \mathcal{L}_{D_i}$$

$$\mathcal{L}_{D_i} = -\frac{1}{2} \mathbb{E}_{x_i^{gt} \sim p_{data_i}} [\log D_i(x_i^{gt})] - \frac{1}{2} \mathbb{E}_{x_i \sim p_{G_i}} [\log(1 - D_i(x_i))] - \frac{1}{2} \mathbb{E}_{x_i^{gt} \sim p_{data_i}} [\log D_i(x_i^{gt}, S_{global})] - \frac{1}{2} \mathbb{E}_{x_i \sim p_{G_i}} [\log(1 - D_i(x_i, S_{global}))]$$

6. Experimental Results

◆ The generated face images of three stages in SIGM



◆ The comparison of generated images with other methods

□ Qualitative results



□ Quantitative results

	BRISQUE↓	IS↑	FID↓	MS-SSIM↓
AttnGAN	62.843	5.124	40.254	0.398
Wang et al.	—	2.2	43.8	—
AFGAN(ours)	35.979	5.853	36.607	0.347

Setting	Classification accuracy
AttnGAN	0.902
AFGAN w/o AEM	0.924
AFGAN w/o SCM	0.940
AFGAN(ours)	0.955