



Tencent  
AI Lab



# FFHQ-UV: Normalized Facial UV-Texture Dataset for 3D Face Reconstruction

Poster: TUE-AM-035

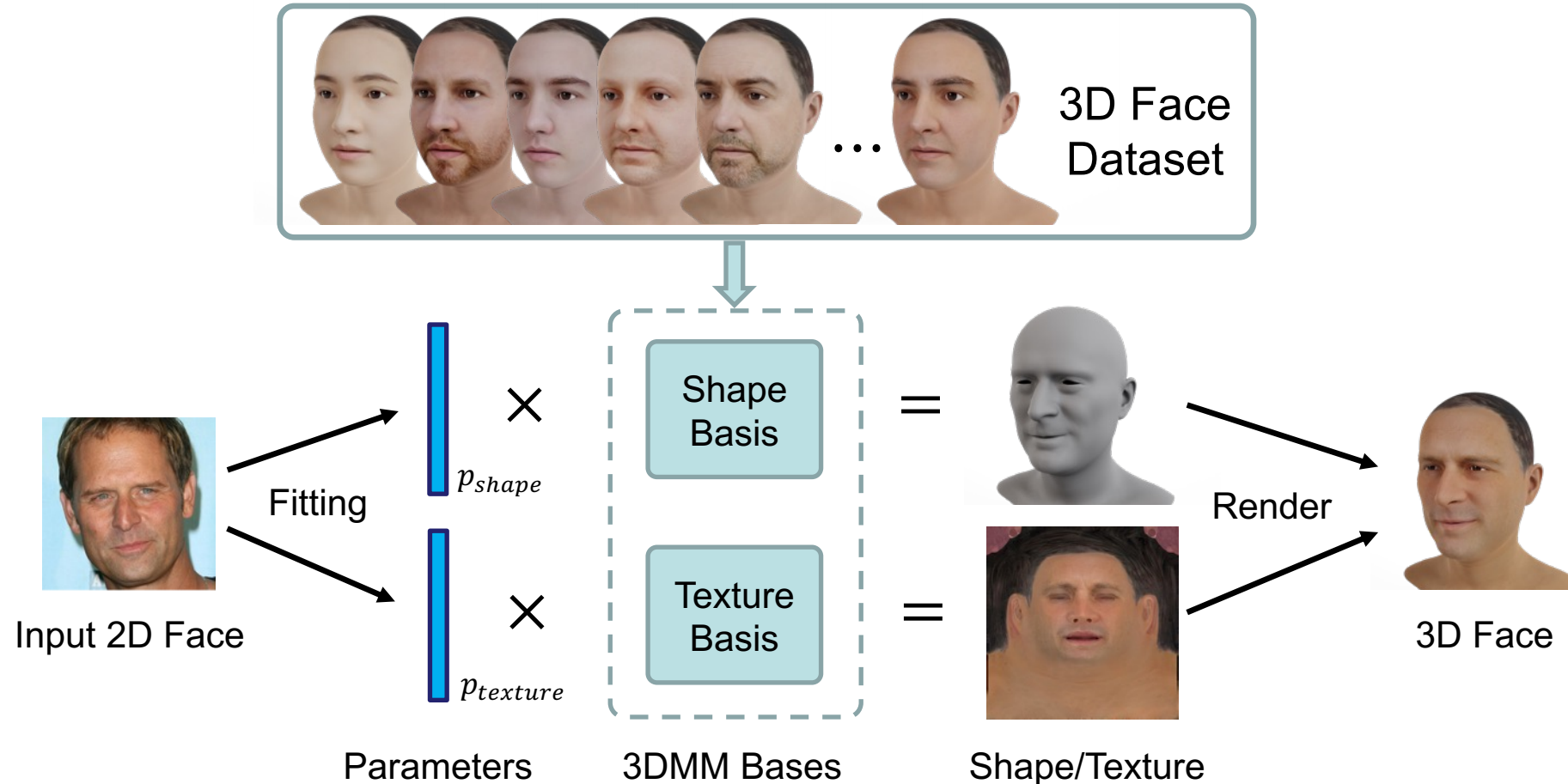
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<sup>2</sup>Tencent AI Lab

# 3D face reconstruction based on 3DMM

- The reconstruction can be achieved by parametric fitting with the 3DMM.
- 3D face dataset is critical for constructing strong 3DMM bases.



# Challenges in texture map recovery

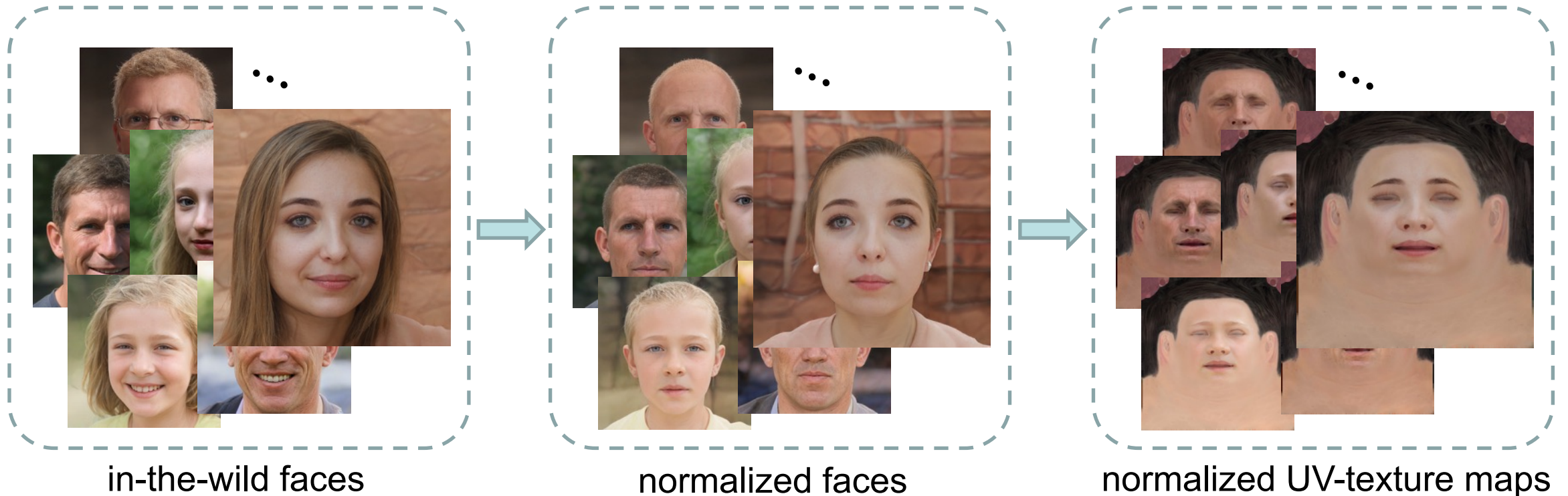
- Two key aspects:
  - *fidelity*: require higher diversity for large expressive capacities
  - *quality*: require even illumination and clean facial region
- Few existing facial texture datasets are of sufficient diversity and high-quality to satisfy both of above aspects.

Datasets	# samples	Resolution	Even illumination	Public available
LSFM*	10,000	512×512	X	X
AvatarMe*	200	6144×4096	✓	X
HiFi3DFace*	200	2048×2048	✓	X
Facescape*	847	4096×4096	✓	✓
WildUV	5638	377×595	X	X
NormAvatar	5601	256×256	✓	X
FFHQ-UV (Ours)	54,165	1024×1024	✓	✓

“\*” denotes the dataset which is captured under controlled conditions.

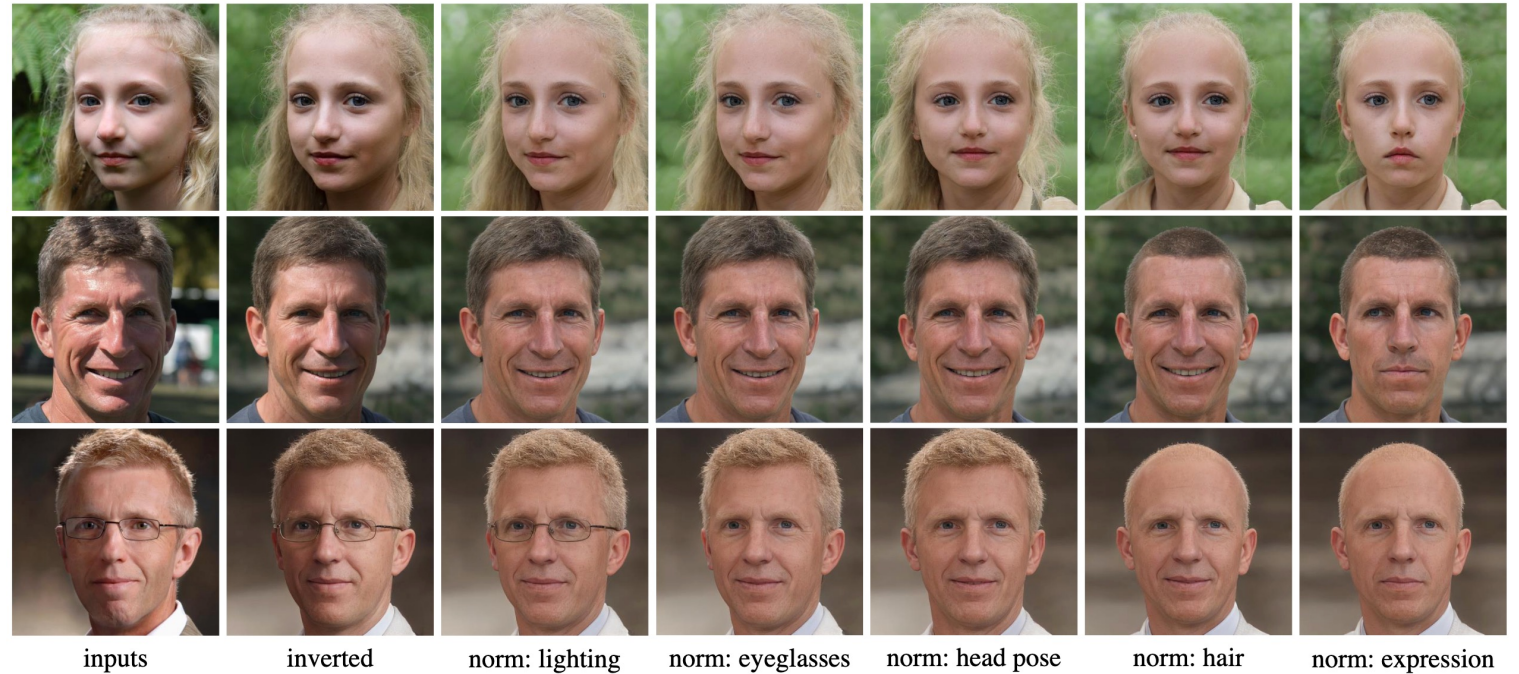
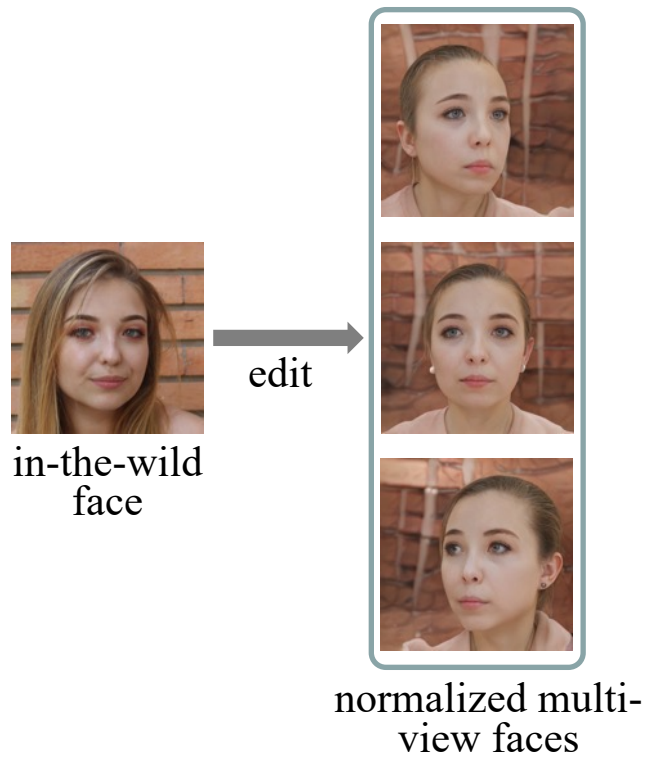
# Our contribution

- We create a normalized UV-texture dataset
  - derive from in-the-wild face images, cheap and diverse
  - even illuminations, neutral expressions, and clean facial regions



# Dataset creation pipeline

- StyleGAN-based facial image editing
  - generate multi-view normalized faces from single-view in-the-wild faces

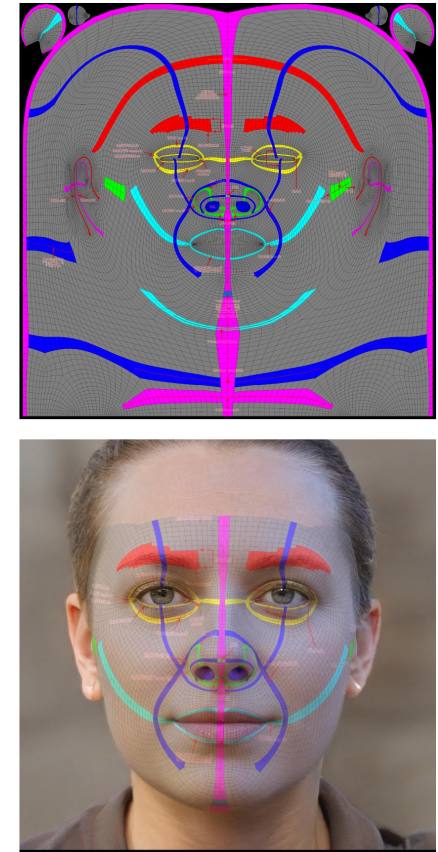
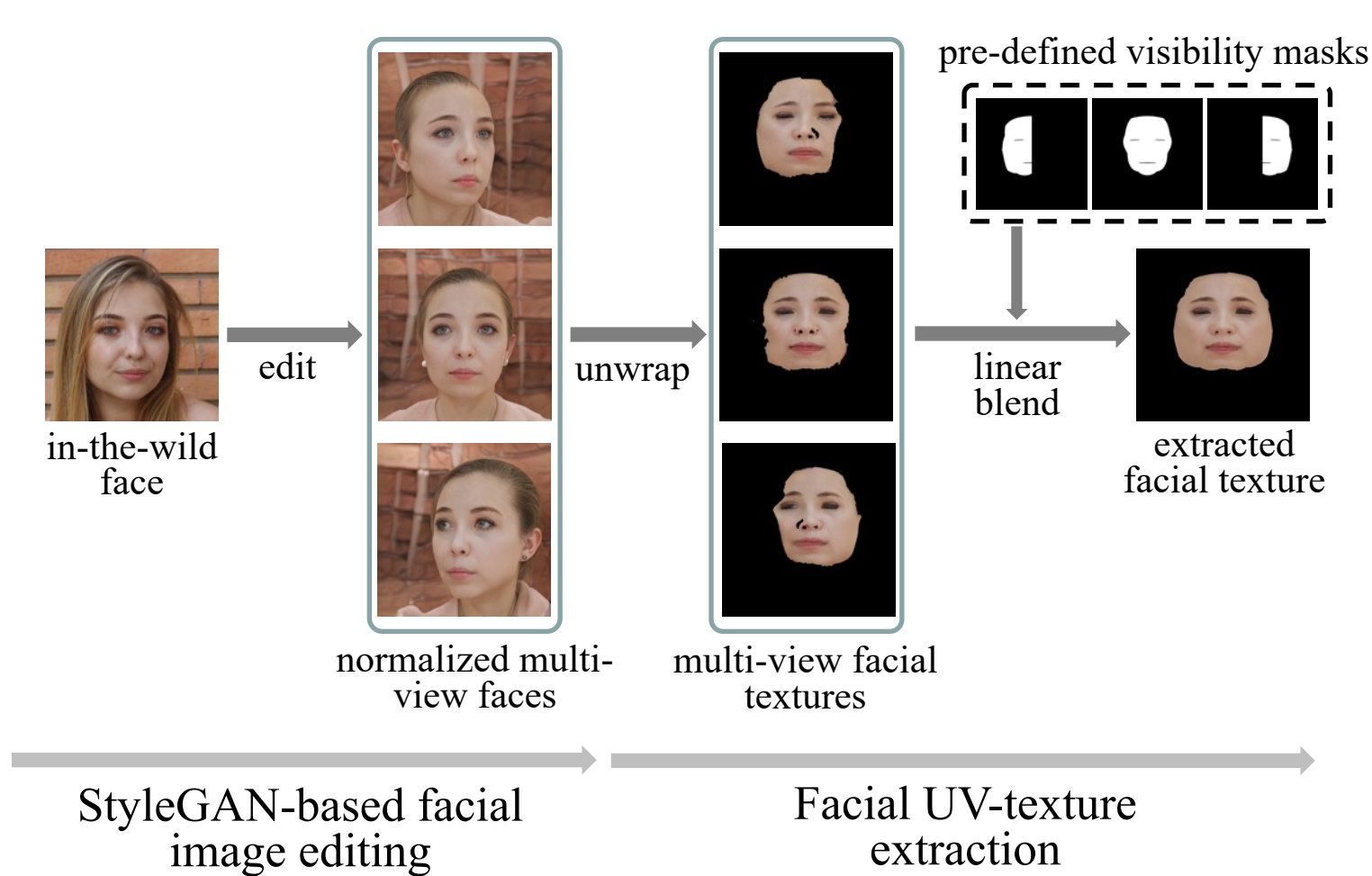


StyleGAN-based facial  
image editing



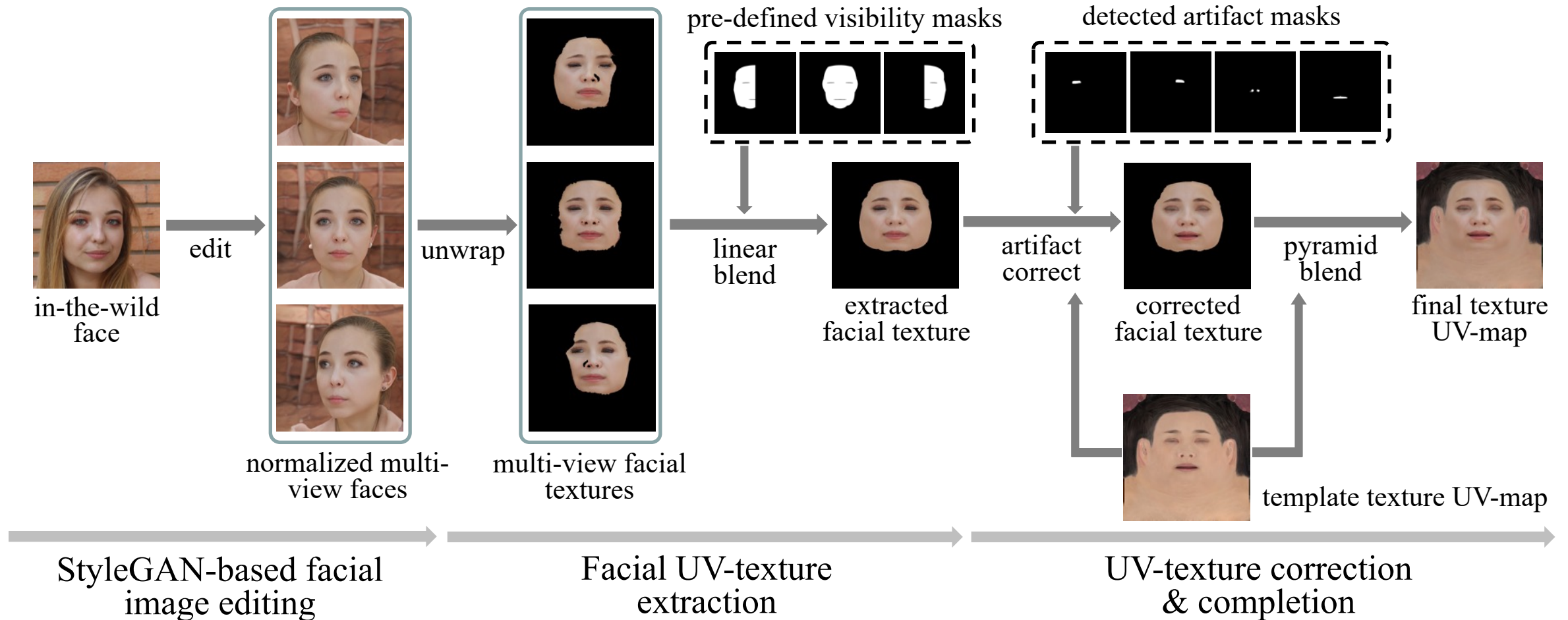
# Dataset creation pipeline

- Facial UV-texture extraction
  - extract multi-view facial textures and blend them together



# Dataset creation pipeline

- UV-texture correction & completion
  - correct artifacts and generate complete UV-texture map



# FFHQ-UV dataset

- Based on FFHQ dataset, which contains diverse in-the-wild faces.
- Over 50,000 UV-texture maps at  $1024 \times 1024$  resolution.
- High-quality and evenly illuminated, ready for realistic renderings.





# Is the FFHQ-UV diverse?

- ID diversity of face datasets
  - FFHQ-UV inherits the most ID variations in FFHQ (over 90%).
  - FFHQ-UV has a higher ID standard deviation value compared to Facescape, indicating that FFHQ-UV is more diverse.

Datasets	FFHQ	FFHQ-Inv	FFHQ-Norm	FFHQ-UV*	Facescape*
ID std.	100%	91.86%	90.06%	90.01%	84.24%

“\*” indicates that ID features are extracted from rendered face images.

# Does the FFHQ-UV evenly illuminated?

- We present Brightness Symmetry Error (BS Error) to measure the illumination of UV maps.

$$BS\ Error = \|\mathcal{B}_\alpha(T^Y) - \mathcal{F}_h(\mathcal{B}_\alpha(T^Y))\|_1$$

$T^Y$ : the  $Y$  channel of  $T$  in YUV space;

$\mathcal{B}_\alpha$ : the Gaussian blurring operation with the kernel size of  $\alpha$ ;

$\mathcal{F}_h$ : the horizontal flip operation

- The StyleGAN-based editing step in our pipeline facilitates even illumination.
- FFHQ-UV is competitive with Facescape (under controlled conditions).

Methods	Facescape*	w/o editing	FFHQ-UV (Ours)
BS Error	6.984	11.385	7.293

“\*” denotes the dataset which is captured under controlled conditions.

# 3D face reconstruction with FFHQ-UV

- GAN-based texture decoder
  - we train a GAN-based texture decoder using FFHQ-UV
  - the goal of our UV-texture recovery is to find the latent code  $z$

$$\tilde{T} = G_{tex}(z)$$

- Three-stage reconstruction algorithm
  1. Linear 3DMM initialization
    - initialize parameters using a trained Deep3D, except texture latent code
  2. Texture latent code  $z$  optimization
    - find a latent code  $z$  of the texture decoder, while fixing other parameters
  3. Joint parameter optimization
    - jointly optimize all the parameters

# Quantitative results on the RELAY benchmark

- Our method outperforms state-of-the-art single-image reconstruction approaches.

Methods	nose	mouth	forehead	cheek	all
3DDFA-v2	1.903	1.597	2.477	1.757	1.926
GANFit	1.928	1.812	2.402	1.329	1.868
MGCNet	1.771	1.417	2.268	1.639	1.774
Deep3D	1.719	1.368	2.015	1.528	1.657
stage 1	1.557	1.661	1.940	1.014	1.543
PCA tex basis	1.904	1.419	1.773	0.982	1.520
w/o multi-view	1.780	1.419	1.711	0.980	1.473
w/ Facescape (scratch)	1.731	1.653	1.711	1.207	1.576
w/ Facescape (finetune)	1.570	1.576	1.581	1.074	1.450
Ours	1.681	1.339	1.631	0.943	<b>1.399</b>



# Quantitative results on the RELAY benchmark

- Our method outperforms baseline variants: initialization in stage 1, using PCA texture basis, and using dataset without multi-view faces.

Methods	nose	mouth	forehead	cheek	all
3DDFA-v2	1.903	1.597	2.477	1.757	1.926
GANFit	1.928	1.812	2.402	1.329	1.868
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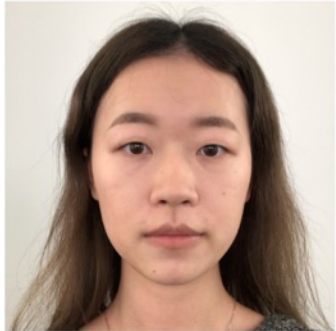
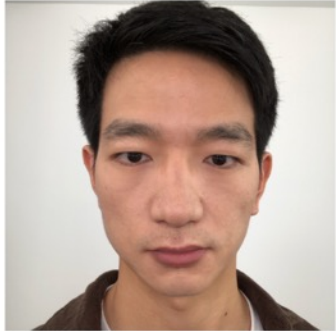
# Quantitative results on the RELAY benchmark

- Using the proposed FFHQ-UV dataset outperforms using texture UV-maps from Facescape dataset.

Methods	nose	mouth	forehead	cheek	all
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# Visual results

- Comparison with the GANFit and AvatarMe methods.
  - our shapes are more faithful to input faces
  - our texture maps are more evenly illuminated and of higher quality



Input image

GANFit  
/AvatarMe

Ours

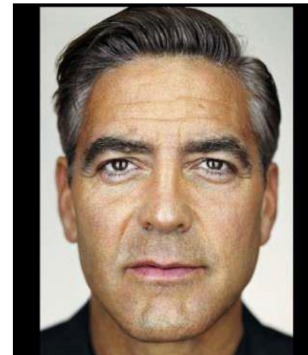
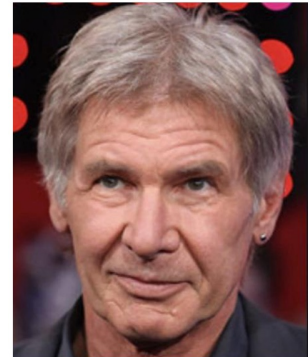
GANFit

AvatarMe

Ours

# Visual results

- Comparison with the Normalized Avatar method.
  - our results better resemble the input faces



Input image

Normalized Avatar

Ours

Input image

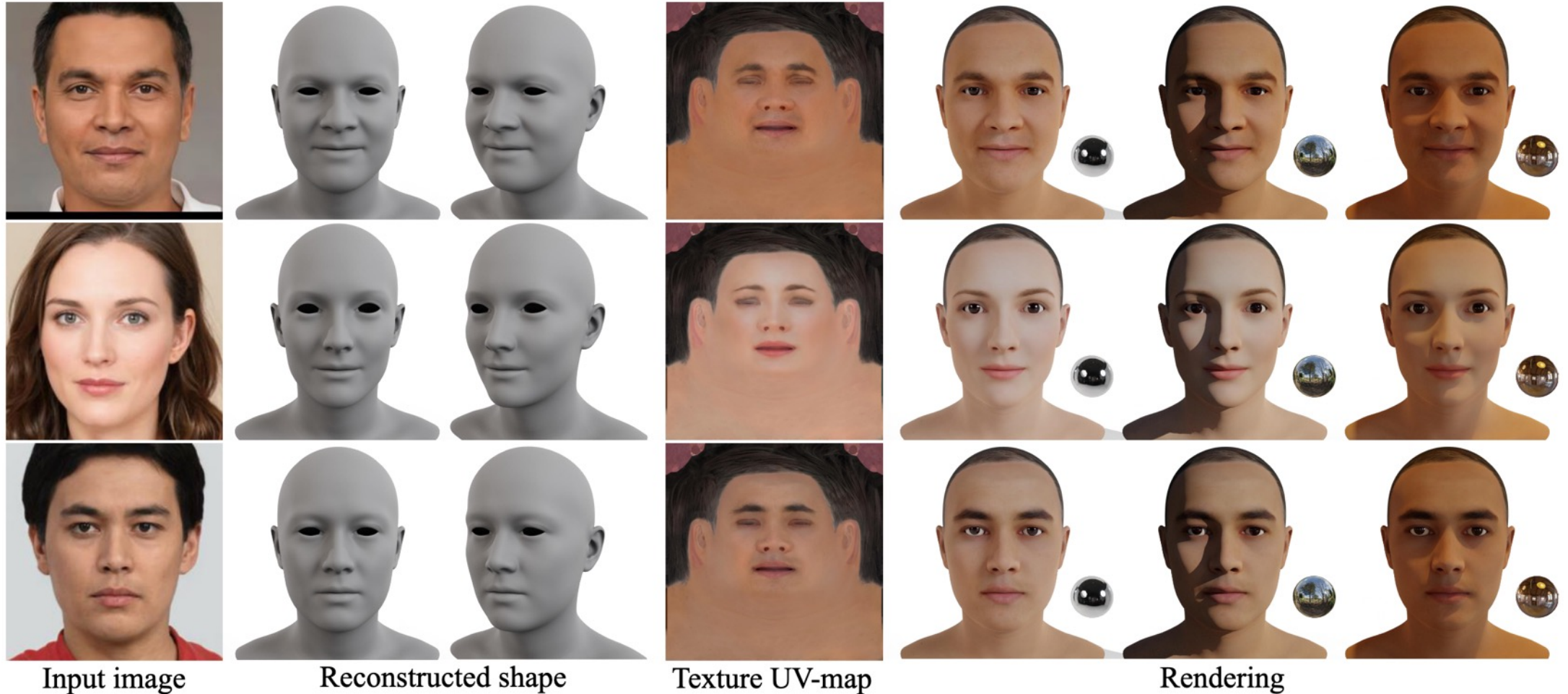
Normalized Avatar

Ours



# Visual results

- Examples of our reconstructed results.
  - our textures can be rendered with different lighting conditions



# Thanks



Code Link: <https://github.com/csbhr/FFHQ-UV>

