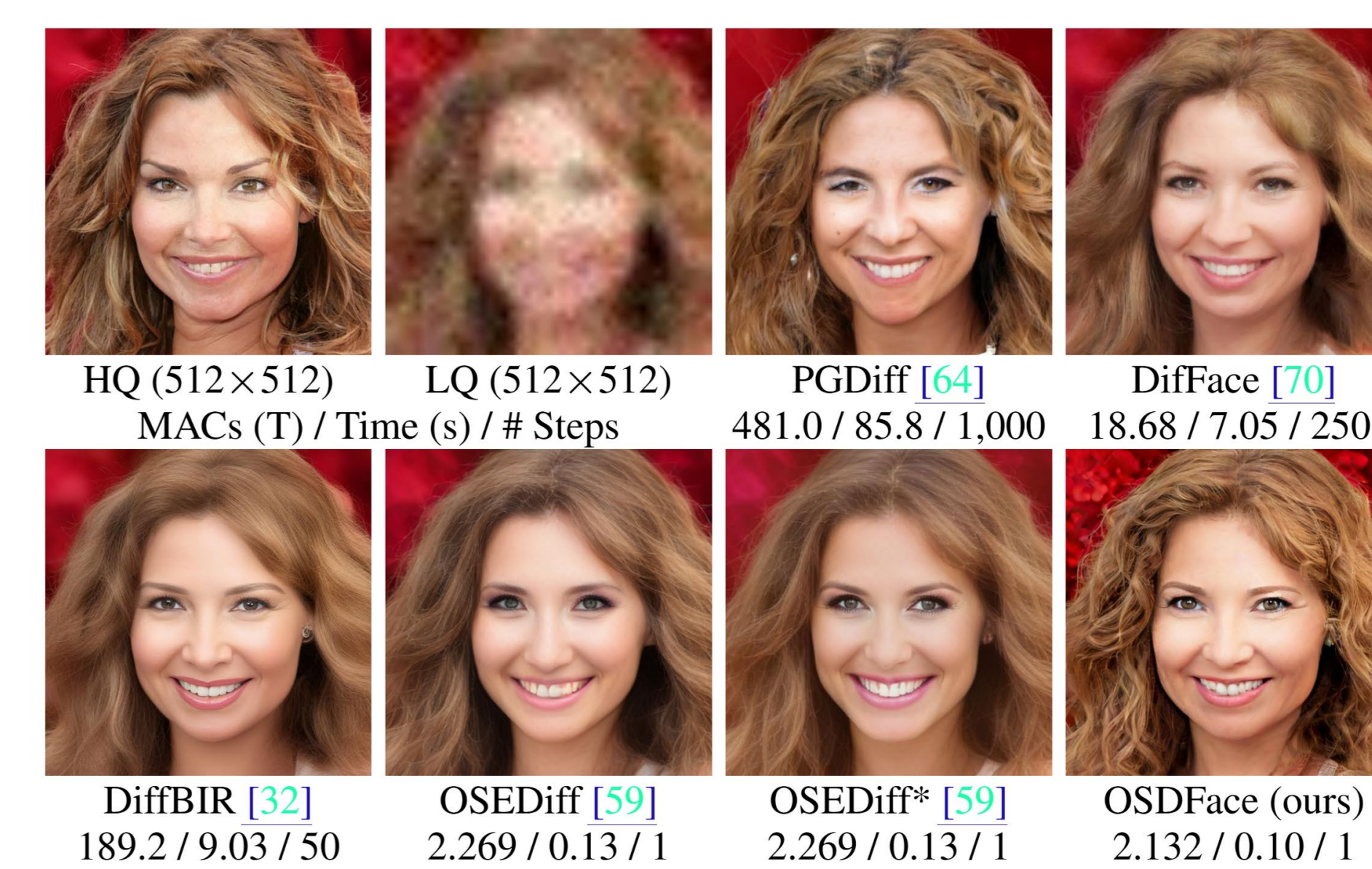
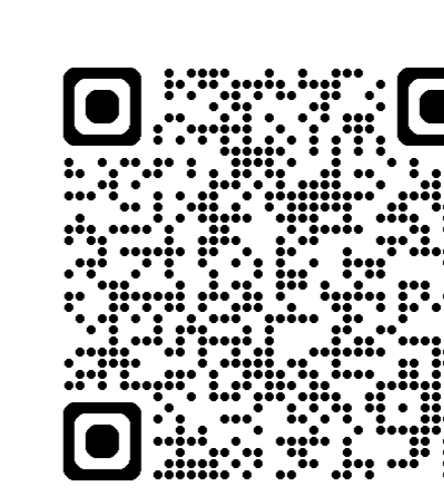


OSDFace: One-Step Diffusion Model for Face Restoration



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Contribution

- **OSDFace: One-Step Diffusion model for FACE restoration.** The **First** attempt to utilize one-step diffusion for restoring faces.
- **VRE: Visual Representation Embedder.** Using low-quality **VQ dictionary**, VRE captures rich prior from LQ images for a deeper understanding of visual content.
- **Realistic Face Alignment.** Facial identity loss for identity consistency and GAN loss for distribution alignment.
- **SOTA performance on Face Restoration** Lower complexity, Smaller model size, **80ms** for 512x512 face.

Methods

Architecture

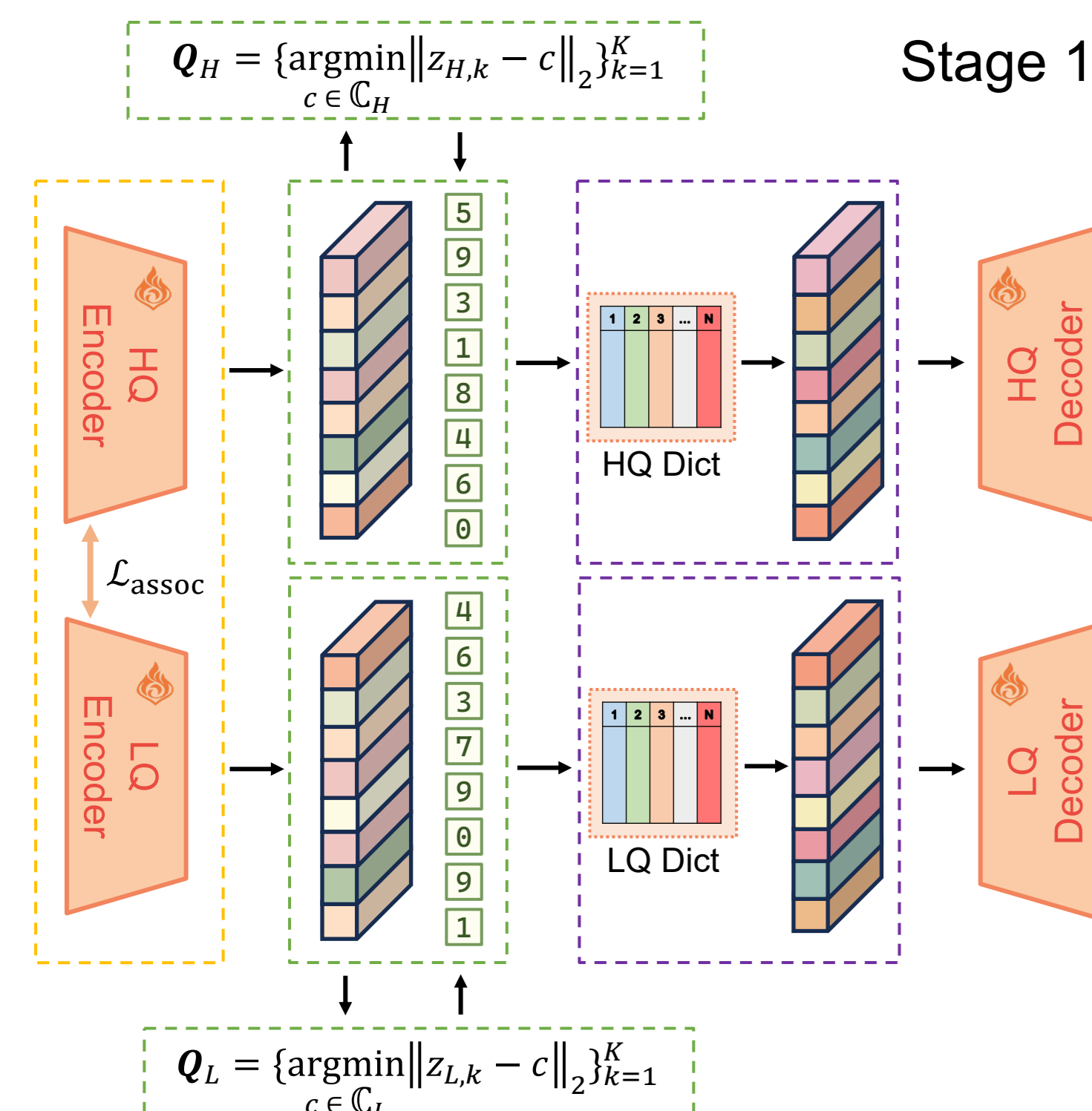


Fig.1. Illustration of Stage 1 Arch.

Visual Representation Embedder (VRE)

- The architecture consists of the **VRE** and **VAE decoder**;
- Two VQ dictionaries corresponding to the **HQ** and **LQ image categories** and train VQVAE using self-reconstruction.
- Align the categories between LQ and HQ faces.
- Enhancing the diagonal correlation within VQ dictionaries

Architecture

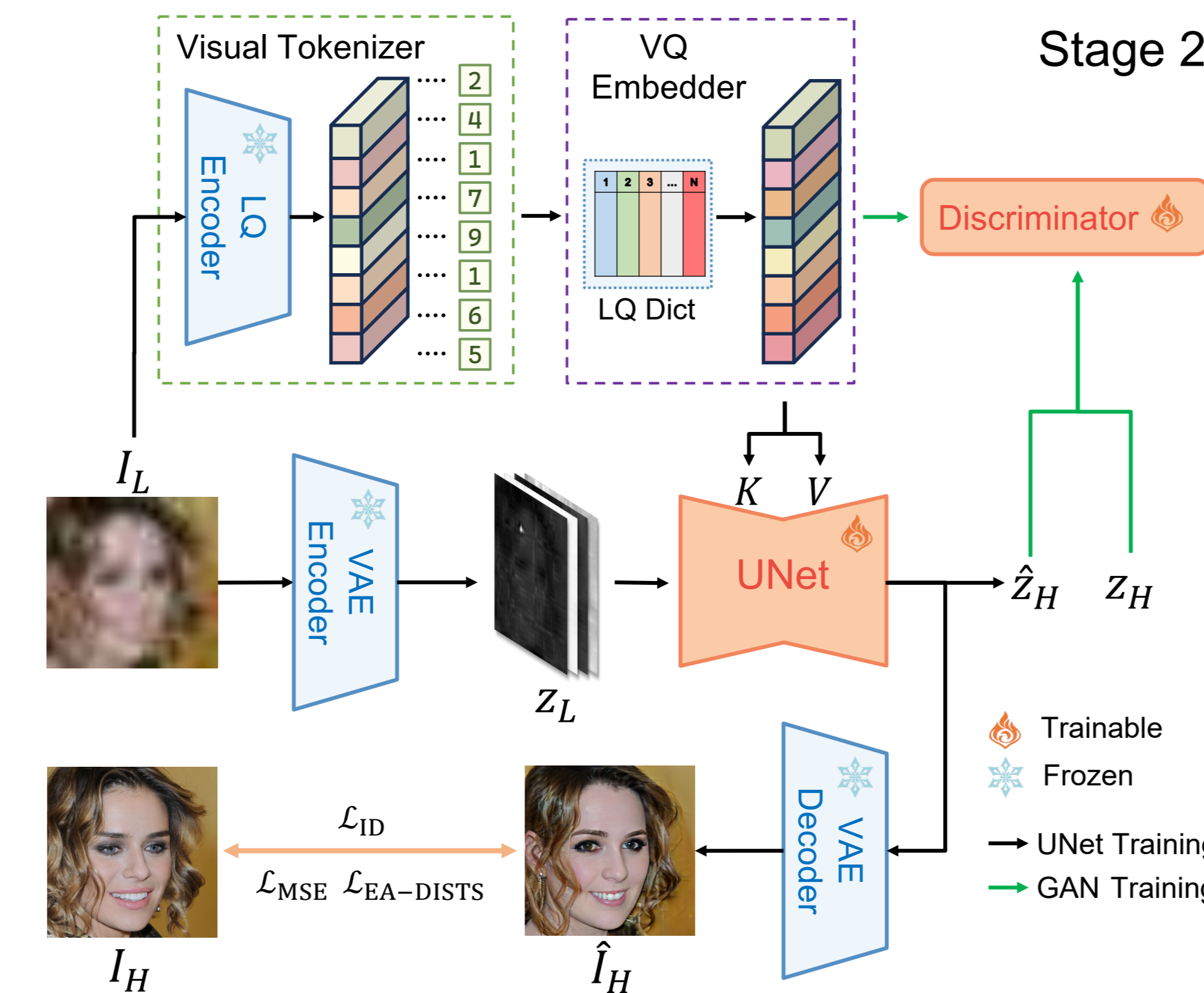


Fig.2. Illustration of Stage 2 Arch.

Realistic Face Alignment

Loss Functions:

- Pixel Reconstruction
- Facial Identity
- Perceptual
- GAN

Total Arch:

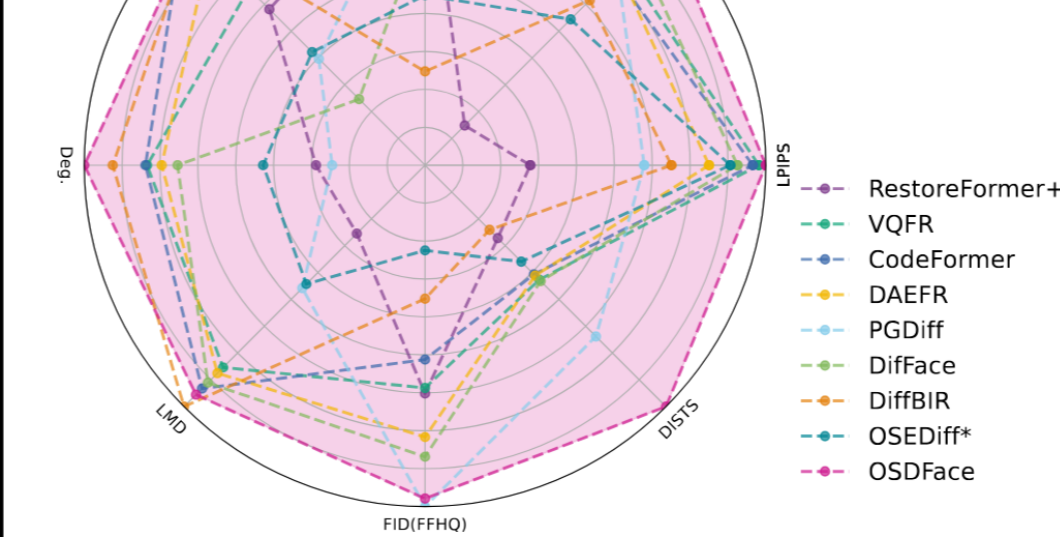
- Stable Diffusion
- LoRA

Advantages:

- More harmonious
- More realistic
- More ID consistent

Comparisons

Radar Chart



Real-World Datasets

Table 1. Synthetic Dataset – CelebA-Test from DAEFR degradation pipeline.

Type	Methods	Wider-Test				LFW-Test				WebPhoto-Test							
		C-IQA↑	M-IQA↑	MUSIQ↑	NIQE↓	FID↓	C-IQA↑	M-IQA↑	MUSIQ↑	NIQE↓	FID↓	C-IQA↑	M-IQA↑	MUSIQ↑	NIQE↓	FID↓	
Non-Diffusion	RestoreFormer++ [56]	0.4535	0.2301	72.3612	3.9524	70.5083	8.8019	57.3723	72.7880								
	VQFR [14]	0.3390	0.2131	73.2969	4.8374	62.2595	5.7297	58.1044	24.9093								
	CodeFormer [71]	0.3412	0.2151	75.9432	4.5157	62.1972	5.3819	62.0280	26.8595								
	DAEFR [49]	0.3580	0.2146	74.9853	4.4161	62.8184	5.6332	62.0341	26.0458								
Multi-Step Diffusion	PGDiff [61] (s=1,000)	0.3866	0.1949	69.5676	4.0010	71.5996	7.3109	44.6258	26.3694								
	DiFace [67] (s=250)	0.3469	0.2126	66.7451	4.6381	63.4511	5.4759	49.8075	22.2370								
	DiffBIR [30] (s=50)	0.3740	0.2340	75.6360	6.2801	61.0238	5.1042	71.7767	32.5109								
One-Step Diffusion	OSEDiff [57] (s=1)	0.4708	0.3165	51.2566	6.6968	72.7112	10.9948	89.8160	68.3579								
	OSEDiff* [57] (s=1)	0.3496	0.2200	69.9807	5.3280	67.4026	7.4082	81.3624	37.1309								
	OSDFace (ours, s=1)	0.3365	0.1773	75.6398	3.8840	60.0708	5.2867	45.4150	17.0617								

Experiments

Tab.1. Ablation Study of VRE

Methods	C-IQA↑	M-IQA↑	MUSIQ↑	NIQE↓	FID↓
Learnable	0.6714	0.5243	74.5986	4.0190	41.7842
DAPE [58]	0.6863	0.5622	74.8107	4.3801	45.9380
VRE (w/o assoc)	0.6791	0.5180	75.2328	3.8228	44.3360
VRE (w/ assoc)	0.6946	0.5356	75.2911	3.8793	41.9502

Tab.2. Ablation Study of Losses

\mathcal{L}_{ID}	$\mathcal{L}_{EA-DISTS}$	\mathcal{L}_G	$\mathcal{L}_{EA-LPIPS}$	C-IQA↑	M-IQA↑	MUSIQ↑	NIQE↓	FID↓
✓	✓	✓	✓	0.6724	0.5243	74.5986	4.0190	41.7842
✓	✓	✓	✓	0.6710	0.5387	74.1060	4.3223	55.1016
✓	✓	✓	✓	0.6674	0.5470	75.2021	4.1484	46.1684
✓	✓	✓	✓	0.6590	0.5081	75.2336	3.9857	45.7834
✓	✓	✓	✓	0.6946	0.5356	75.2911	3.8793	41.9502

Tab.3. Complexity comparison

Methods	PGDiff [61]	DiFace [67]	DiffBIR [30]	OSEDiff [57]	OSDFace
Step	1,000	250	50	1	1
Time (s)	85.81	7.05	9.03	0.13	0.10
Param (M)	176.4	175.4	3,042	1,302	978.4
MACs (G)	480,997	18,682	189,208	2,269	2,132

Visual Comparisons

