

3D-Aware Multi-Class Image-to-Image Translation with NeRFs

Senmao Li¹ Joost van de Weijer² Yaxing Wang^{1*} Fahad Shahbaz Khan^{3,4} Meiqin Liu⁵ Jian Yang¹ ¹VCIP,CS,Nankai University, ²Universitat Auto noma de Barcelona, ³Mohamed bin Zayed University of AI, ⁴Linkoping University, ⁵Beijing Jiaotong University

> Paper ID 81 Code: <u>https://github.com/sen-mao/3di2i-translation</u>

Problems



- **No prior works** investigate 3D-aware GANs for 3D consistent multi-class image-to-image (3D-aware I2I) translation.
- Delta 2D-I2I translation methods applied to 3D-I2I translation tasks result in three main challenges (1. underestimating viewpoint changes, 2.

identity change, 3. a geometrically unrealistic ear) when changing the viewpoint.





• We decouple the learning process into multi-class 3D-aware generation (step1) and 3D-aware I2I translation (step2).





step1: (1) training an unconditional 3D-aware generative model on datasets (i.e., StyleNeRF) and (2) partially initializing the multi-class 3D-aware generative model (i.e., multi-class StyleNeRF).





step1: (1) training an unconditional 3D-aware generative model on datasets (i.e., StyleNeRF) and (2) partially initializing the multi-class 3D-aware generative model (i.e., multi-class StyleNeRF).





step2: 3D-aware I2I translation architecture adapted from the trained multi-class StyleNeRF (**step1**). This initialization inherits the capacity of being sensitive of view information.





step2: 3D-aware I2I translation architecture adapted from the trained multi-class StyleNeRF (step1). This initialization inherits the capacity of being sensitive of view information.



• The generated images of step1 (top) and step2 (bottom), which show

that we correctly align the outputs of both the NeRF mode F and the

adaptor A.





• several techniques for **step2**: relative regularization loss and hierarchical representation constrain



relative regularization loss



hierarchical representation constrain

Inference time



inference: the 3D image (e.g. female) is fed into the trained encoder E, and through the adaptor A and generator G, it is eventually translated into other categories of 3D image (e.g. male).



Ablation study



multi-class StyleNeRF (step1) training from scratch (top) causes artifact and mode collapse.





Multi-class StyleNeRF (from scratch)



Muti-class StyleNeRF (Ours, initialize by StyleNeRF)

Ablation study



Both using a single mapping network (left) and using two mapping networks without concatenating (right) fails to generate satisfactory results.



Ablation study



Comparison with baselines.* denotes that we used the results provided by StarGANv2. + means that we used the pre-trained networks provided

by authors.

Dataset		CelebA-HQ		AFHQ	
Method		TC↓	FID↓	TC↓	FID↓
*MUNIT		30.240	31.4	28.497	41.5
*DRIT		35.452	52.1	25.341	95.6
*MSGAN		31.641	33.1	34.236	61.4
StarGANv2		10.250	13.6	3.025	16.1
Ours (3D)		3.743	22.3	2.067	15.3

	TC↓	(unc)FID↓	TC↓	(unc)FID↓
†Liu et al. [35]	13.315	17.8	3.462	20.0
StarGANv2	10.250	12.2	3.025	9.9
†Kunhee et al. [24]	10.462	6.7	3.241	10.0
Ours (3D)	3.743	18.7	2.067	11.4

Impact of several components in the performance on AFHQ. Ini.: initialization method for multi-class StyleNeRF, Ada.: Unet-like adaptor, Hrc.:

Hierarchical representation constrain, Rrl.: Relative regularization loss.

Ini.	Ada.	Hrc.	Rrl.	TC↓	FID↓
Y	Ν	Ν	Ν	2.612	23.8
Y	Y	Ν	Ν	2.324	23.1
Y	Y	Y	Ν	2.204	16.1
Y	Y	Y	Y	2.067	15.3

Results



Our approach produces consistent results across viewpoints (up and bottom, left). User study (bottom, right).



Output (female)





Results



• More results of 3D-aware I2I translation of **female into male (top)** and **male into female (bottom)** on Celeba-HQ 1024×1024







We are the first to explore 3D-aware multi-class I2I translation, which allows generating 3D consistent videos.

We decouple 3D-aware I2I translation into two steps. Step1: we propose a multi-class StyleNeRF. To train this multi-class StyleNeRF effectively, we provide a new training strategy. Step2: we propose a 3D-aware I2I translation architecture.

 To further address the view-inconsistency problem of 3D-aware I2I translation, we propose several techniques: (1) a unet-like adaptor, (2) a hierarchical representation constraint and (3) a relative regularization loss.