Research works

 Deep Marching Tetrahedra: a Hybrid Representation for High-Resolution 3D Shape Synthesis. Tianchang Shen, Jun Gao, Kangxue Yin, Ming-Yu Liu, Sanja Fidler NeurIPS, 2021

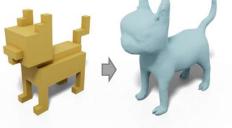


 3DStyleNet: Creating 3D Shapes with Geometric and Texture Style Variations. Kangxue Yin, Jun Gao, Maria Shugrina, Sameh Khamis, Sanja Fidler ICCV 2021 (oral)

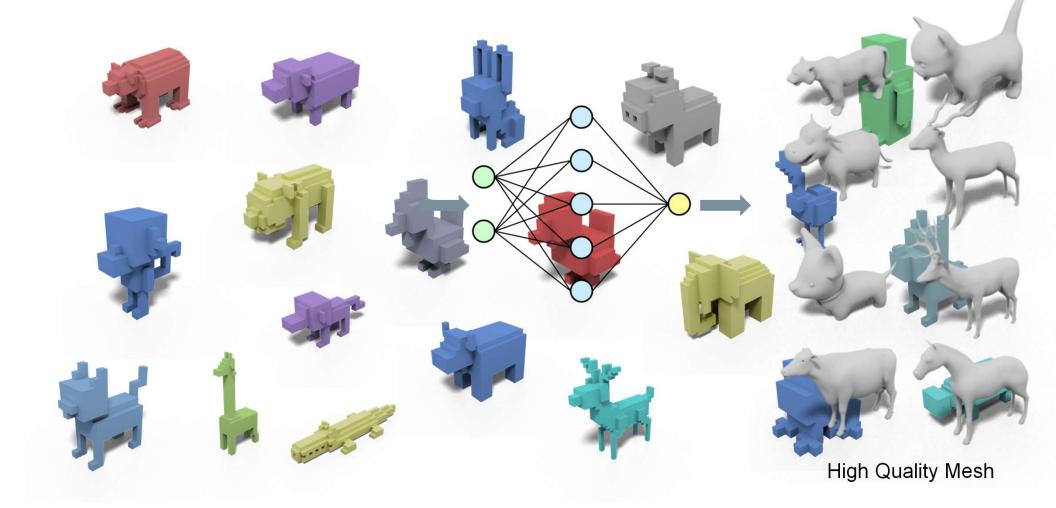


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Rough 3D Shapes -> Detailed 3D Shapes





- Discrete Representations ?
 - Limited to predefined resolution or topology

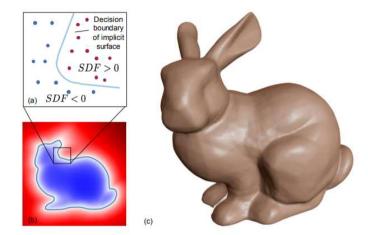


Figure from [Gao et al. 2020]

• Deep Implicit Fields (DIFs) ?

 $f_{\theta}(x, y, z) \approx s(x, y, z)$

Signed distance from (x, y, z) to closest surface $f_{\theta}(x, y, z) = 0$ defines the implicit surface



DeepSDF [Park et al. 2019]

• Deep Implicit Fields (DIFs) ?

 $f_{\theta}(x, y, z) \approx s(x, y, z)$

Pros:

- Represent arbitrary topology
- Continuous



SIREN [Sitzmann and Martel et al. 2019]

• Deep Implicit Fields (DIFs) ?

 $f_{\theta}(x, y, z) \approx s(x, y, z)$

Pros:

- Represent arbitrary topology
- Continuous

Cons:

• Regressing SDF/OF in **generative tasks** do not capture geometric details.



¹Result of [Peng et al. 2020]

• Deep Implicit Fields (DIFs) ?

 $f_{\theta}(x, y, z) \approx s(x, y, z)$

Pros:

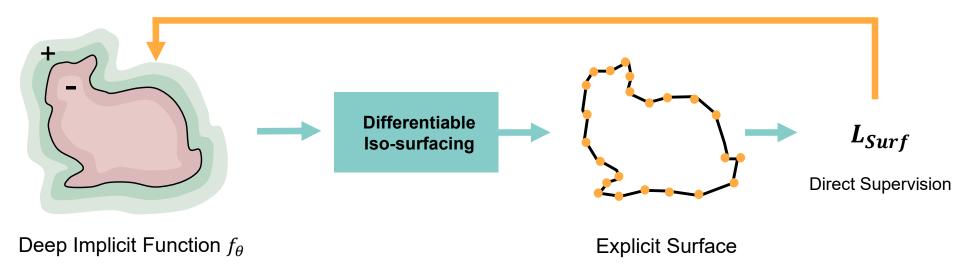
- Represent arbitrary topology
- Continuous

Cons:

- Regressing SDF/OF in **generative tasks** do not capture geometric details.
- Requires costly and lossy meshing step



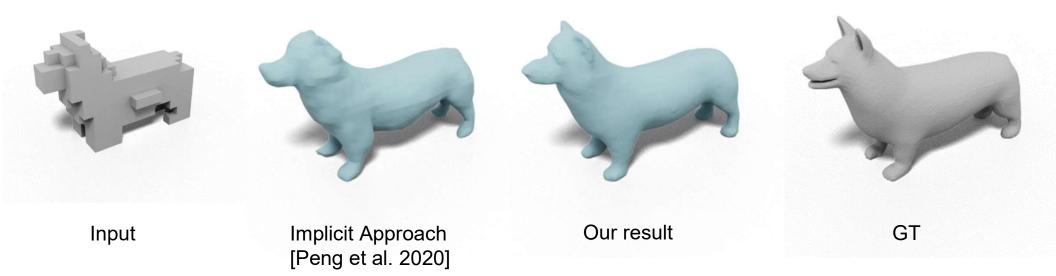
Key Idea: Differentiable Iso-surfacing



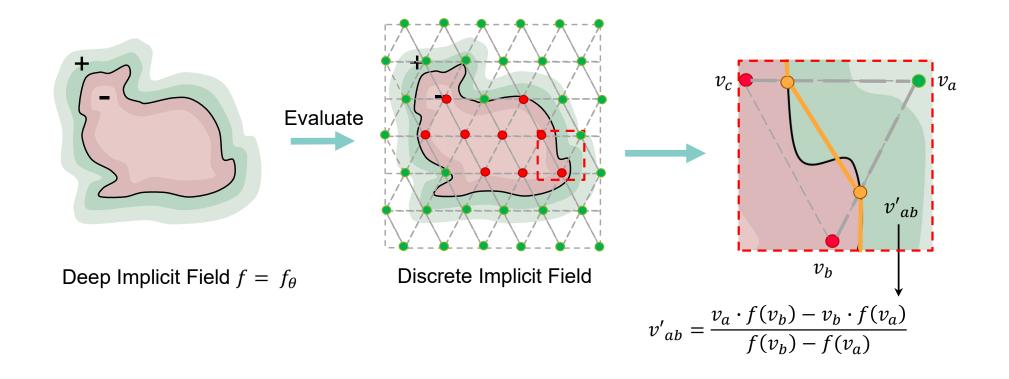
Optimizing f_{θ} for L_{Surf}

- Aware of **quantization error** from meshing
- higher quality shapes with **finer geometric details**

Key Idea: Differentiable Iso-surfacing

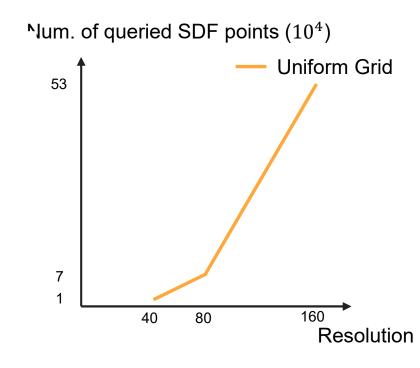


Marching Tetrahedra² (MT)



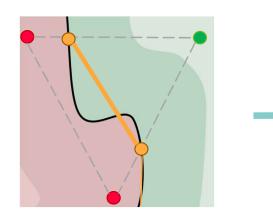
²An efficient method of triangulating equi-valued surfaces by using tetrahedral [Doi et al. 1991]

Memory and Computational Cost



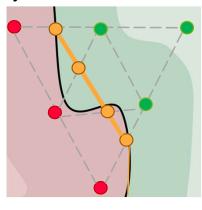
However, the computation and memory footprint grows cubically as grid resolution increases Also a limitation of previous differentiable iso-surfacing methods [Remelli et al. 2020, Liao et al. 2018]

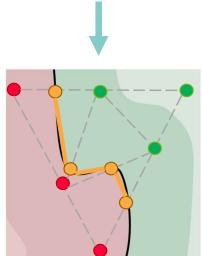
Volume Subdivision



Bad approximation of local surface

Only subdivide surface tets

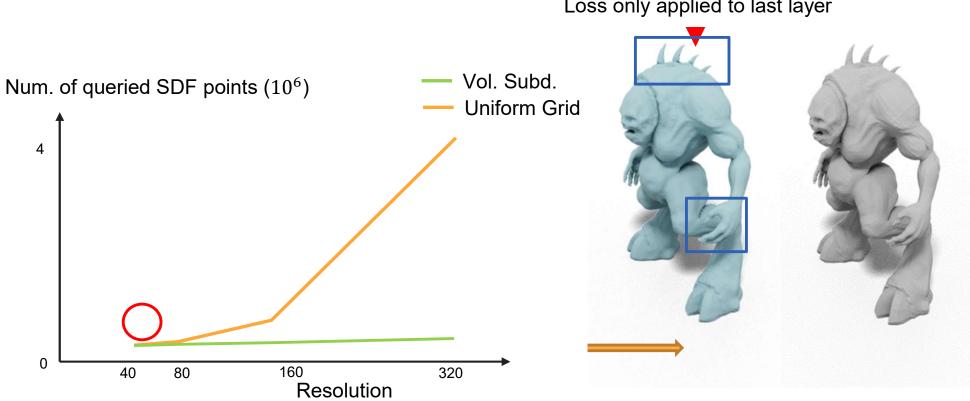




Local updates to SDFs and positions

Volume Subdivision

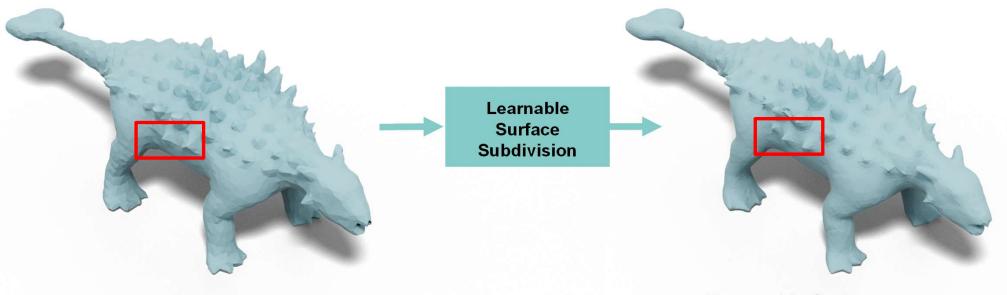
Automatically learns the subdivision hierarchy



Loss only applied to last layer

Surface Subdivision

End to end trainable Preserve sharp features



Triangular Mesh

Parametric Surface

Reconstruction Loss Produces Mean Shape



Input

Prediction

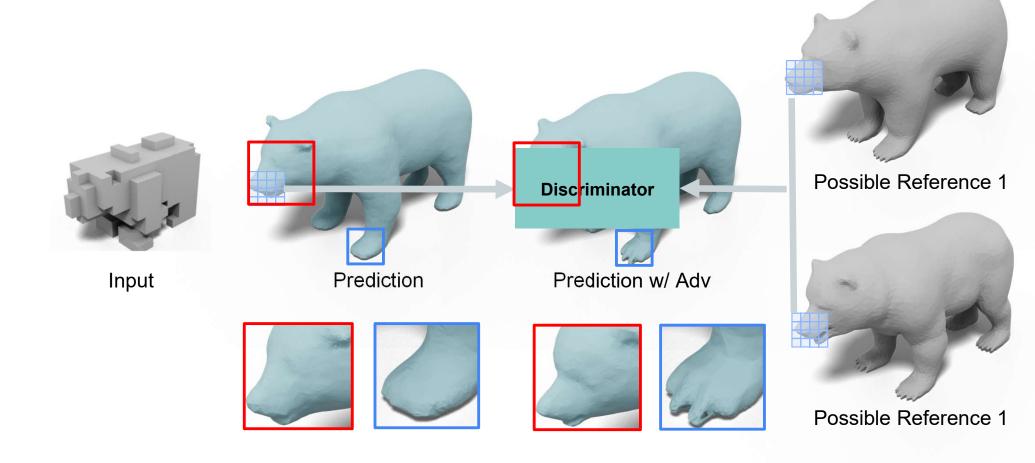


Possible Reference 1



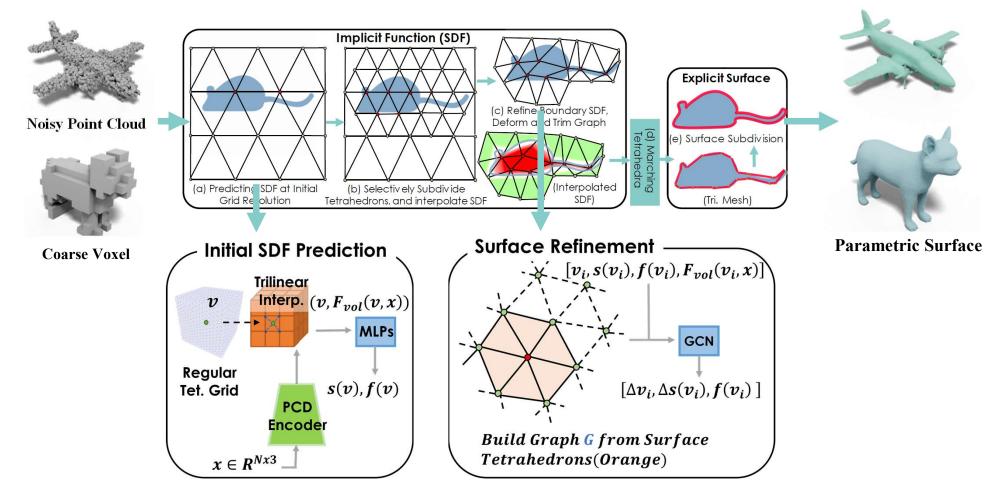
Possible Reference 1

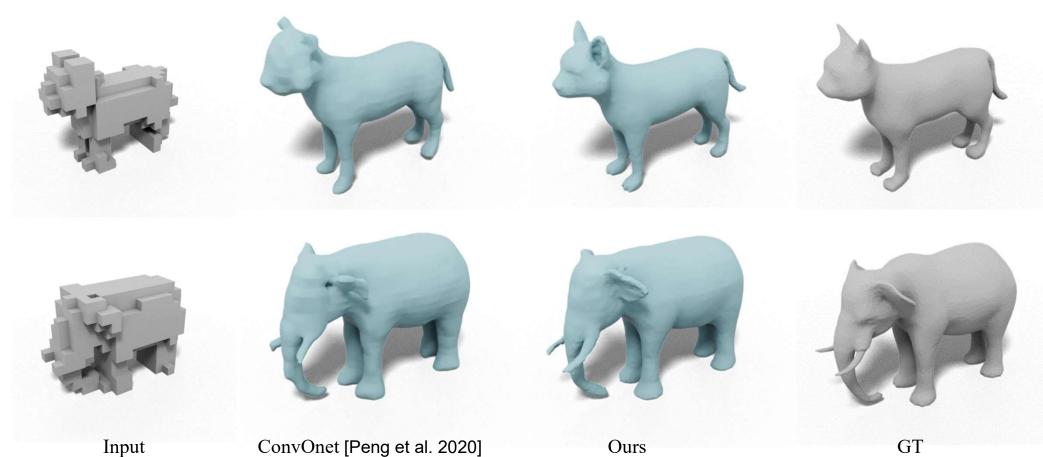
Reconstruction Loss Produces Mean Shape



Deep Marching Tetrahedra

Refer to our paper for more details

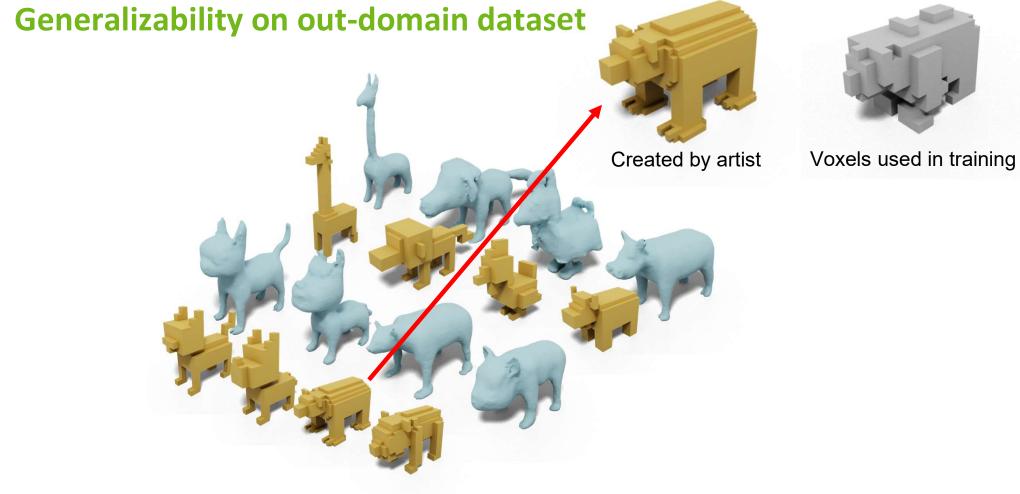




Input

ConvOnet [Peng et al. 2020]

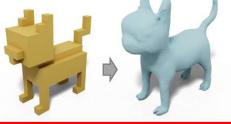
Ours



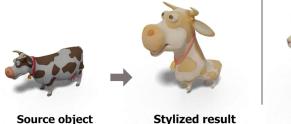
Minecraft Shapes

Research works

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Target style

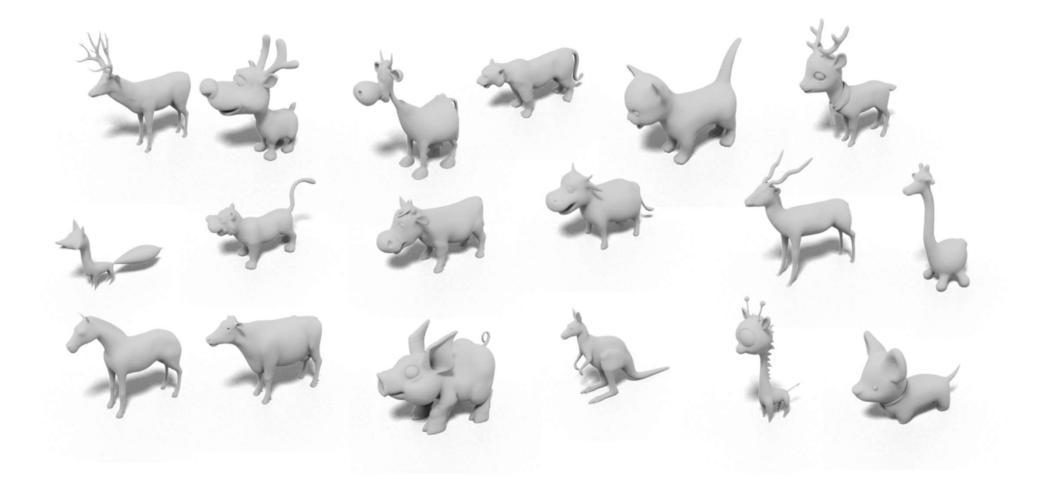
Neural Style Transfer for Images





Photos from [Gatys et al. 2016]

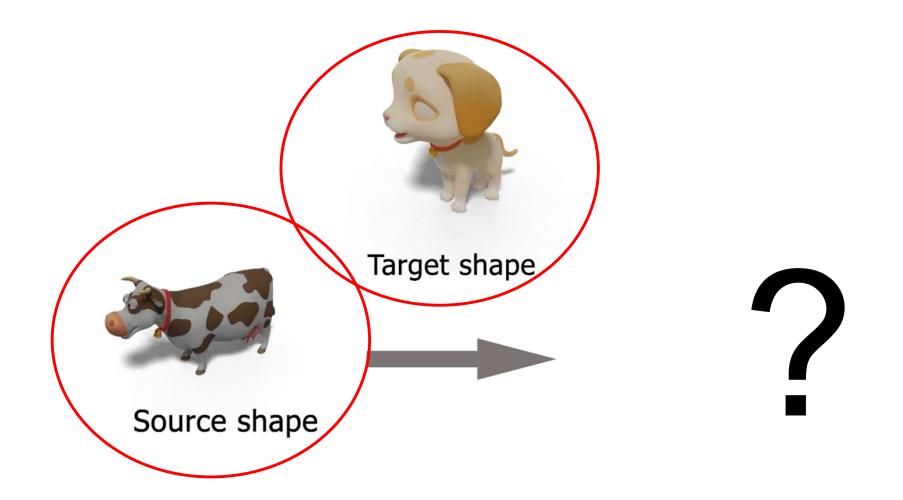




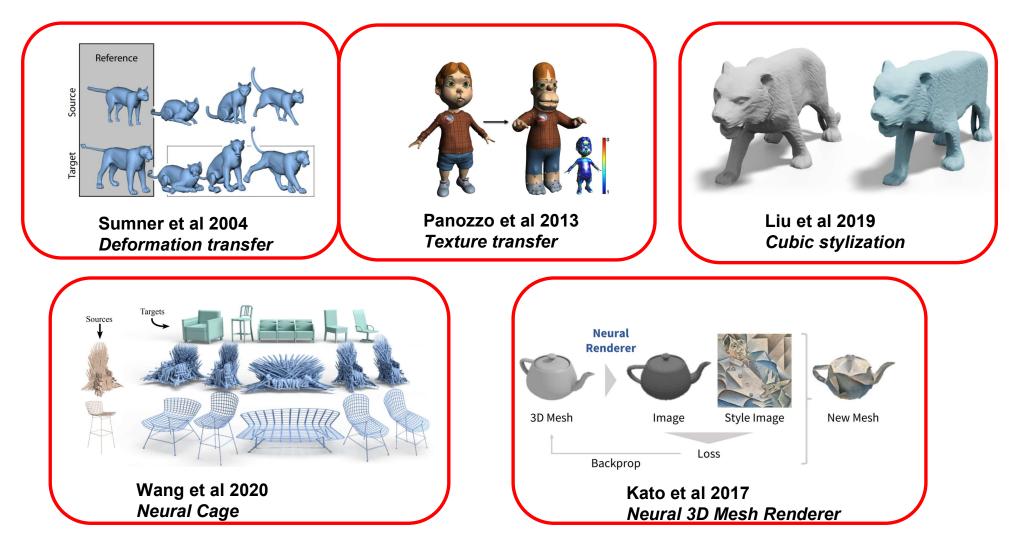




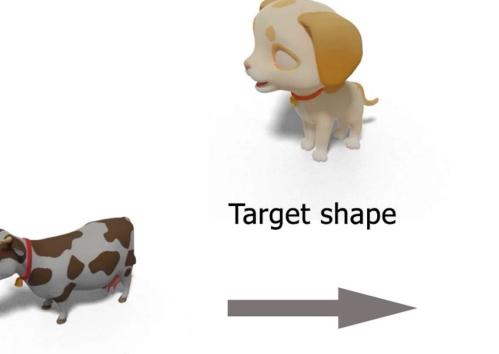
Our 3D Style Transfer



Prior Works



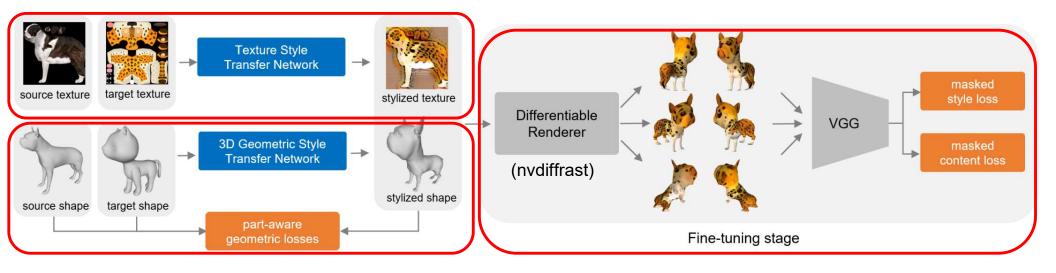
Our 3D Style Transfer



Source shape

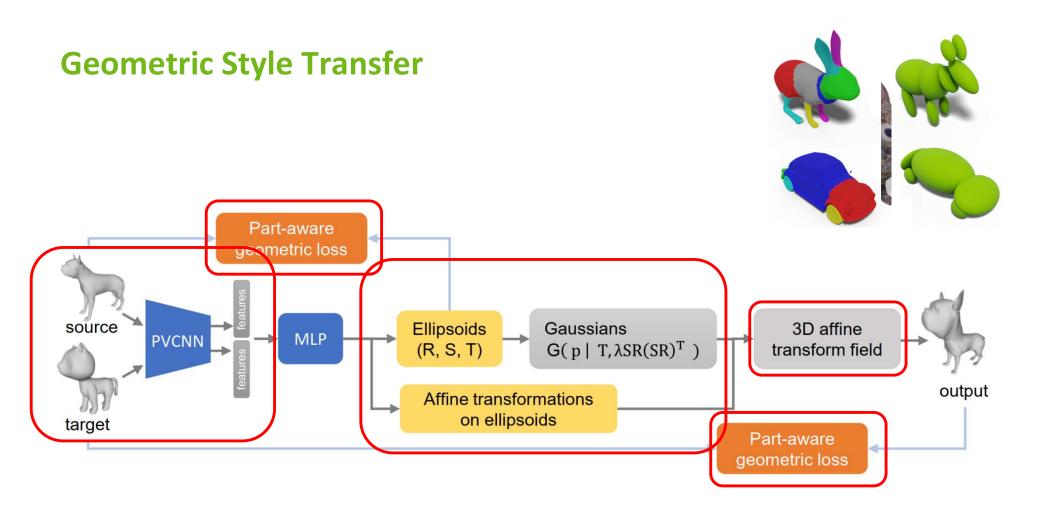
Stylized result

Our 3D Style Transfer

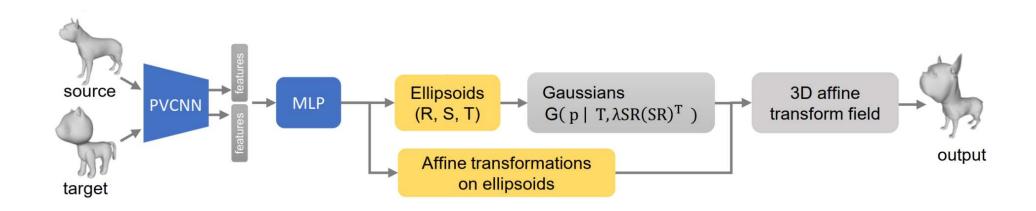


Stage 1

Stage 2

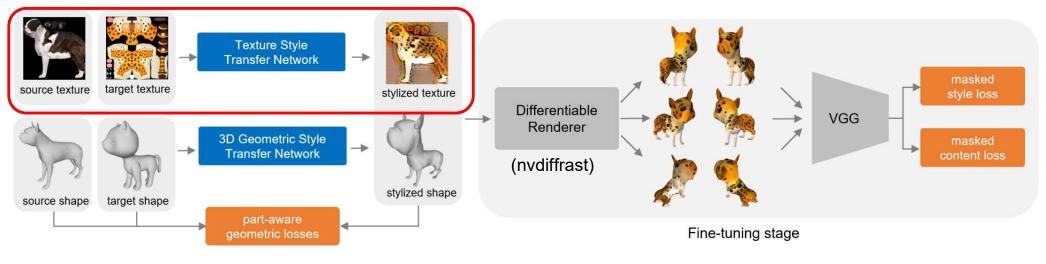


Geometric Style Transfer



Texture Style Transfer

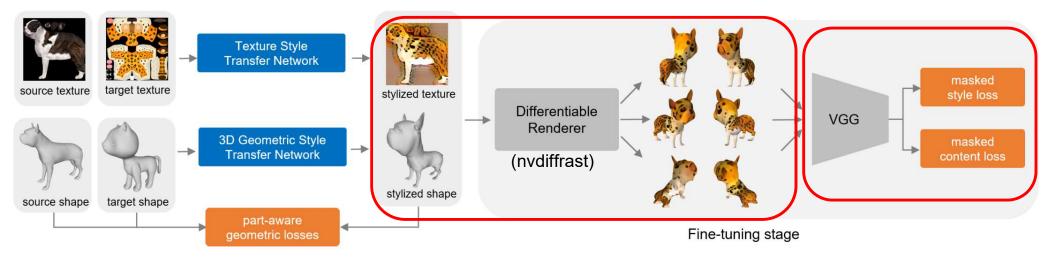




Stage 1

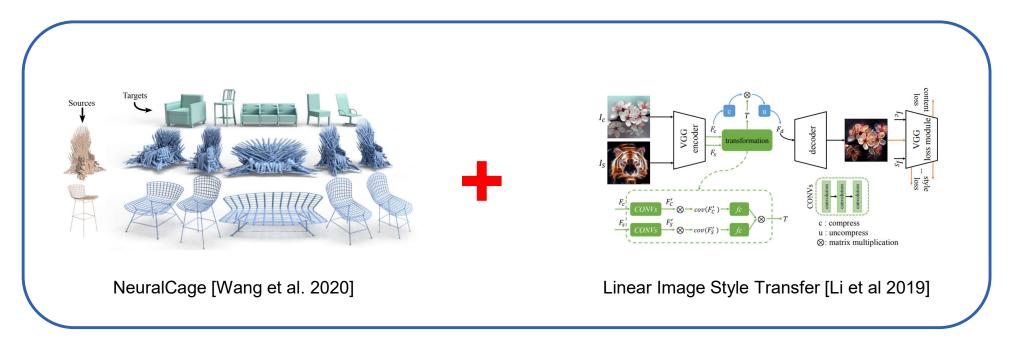
Stage 2

Joint Geometry and Texture Finetuning



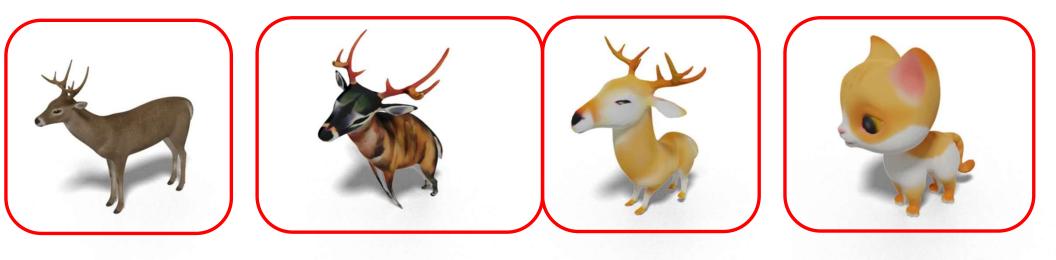
- 20 steps
- 10 seconds on RTX 2080Ti

Baseline



Geometry

Texture image

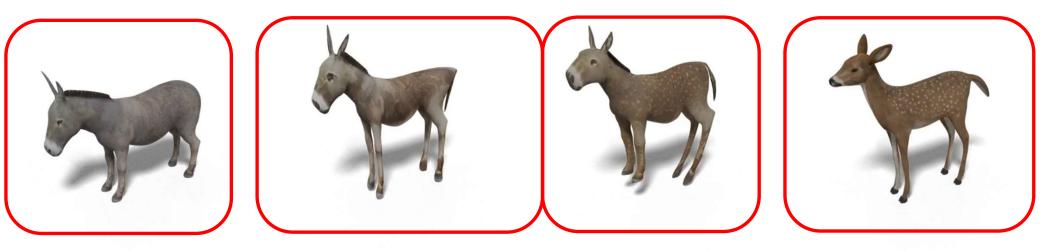


Source shape

NeuralCage [Wang et al. 2020] + Linear Style Transfer[Li et al 2019]

Our result

Target shape

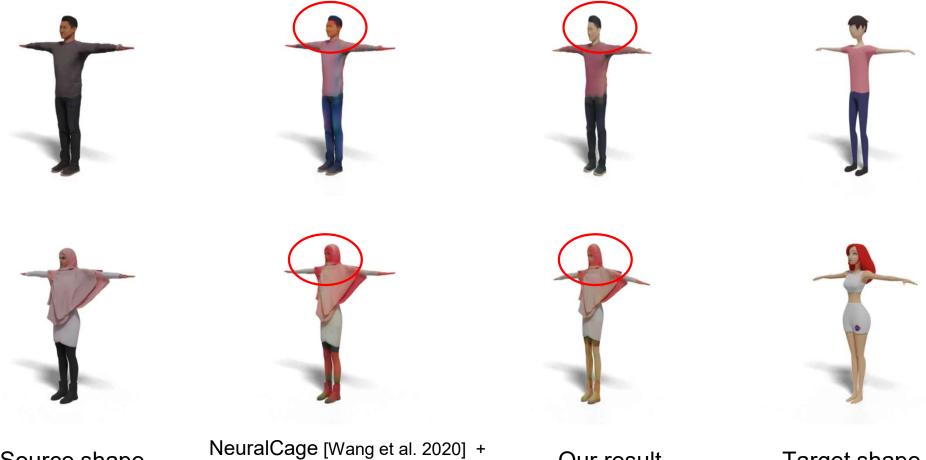


Source shape

NeuralCage [Wang et al. 2020] + Linear Style Transfer[Li et al 2019]

Our result

Target shape



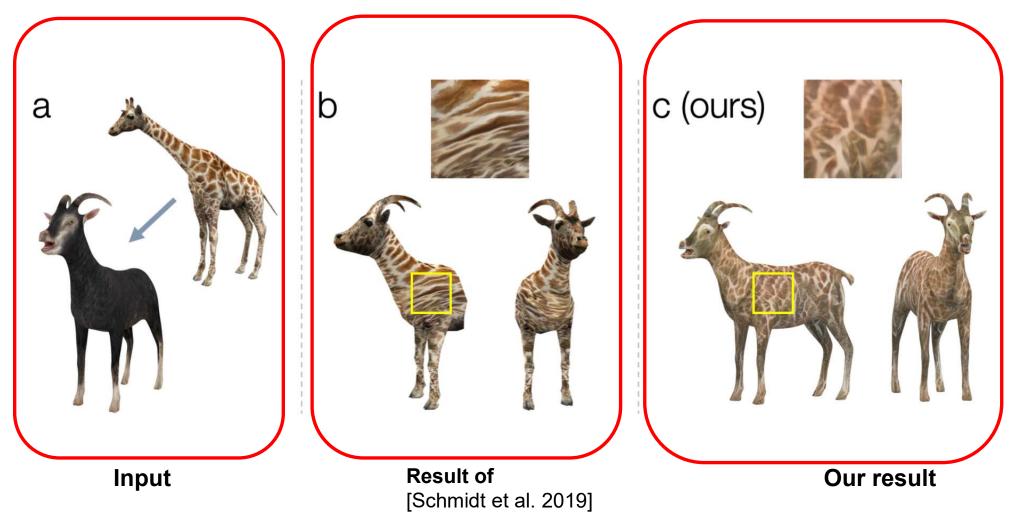
Source shape

Linear Style Transfer[Li et al 2019]

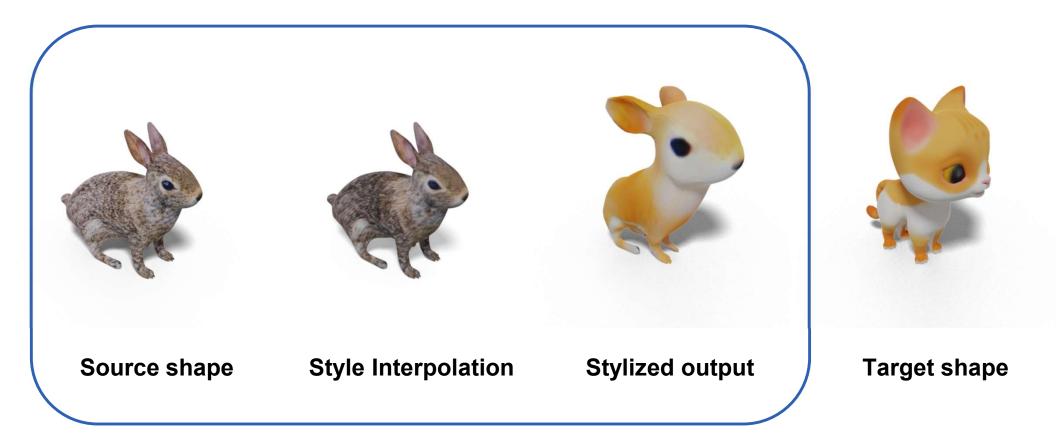
Our result

Target shape

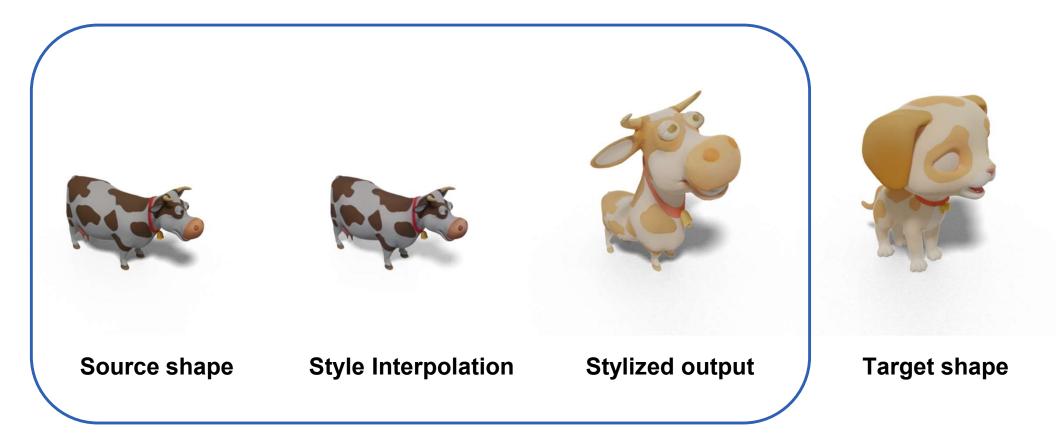
Texture Transfer for Mesh



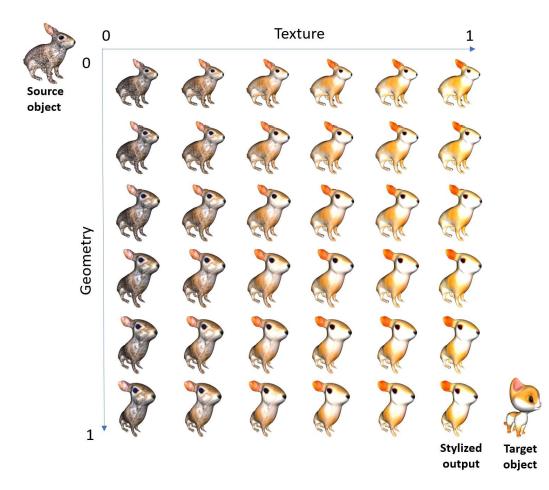
Linear Style Interpolation in Real-time



Linear Style Interpolation in Real-time



Linear Style Interpolation in Real-time



Thank you!

targets sources

https://nv-tlabs.github.io/3DStyleNet/





