## Learning Shape Symmetries and UV-maps for 3D Mesh Reconstruction

**Research Internship Project** 

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# Learning Shape and Color Representation

- Reconstruct 3D shape & texture from RGB
- Shape-texture encoding for recognition signature
  - View invariance and consistency



# Learning Shape and Color Representation

- Reconstruct 3D shape & texture from RGB
- Shape-texture encoding for recognition signature
  - View invariance and consistency
  - Shape-color uniqueness



# Learning Shape and Color Representation

- I. Advance state-of-the-art
  - Standard benchmarks



- II. Supermarket scenarios
  - Synthetic data pipeline



# Challenges

#### **3D** Ground Truth

- $\succ$  synthetic data
- $\succ$  implicit supervision

#### **Shape-Color Decomposition**

- network modularity
- flexible training
- $\succ$  UV mapping

#### **Consistent Representation**

 $\succ$  template deformation



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## Shape-Texture Reconstruction Overview

#### Narrow Domain



[Güler et al. 2018] (Denspose), [Rempe et al 2021] (HuMor)

# Shape-Texture Reconstruction Overview

#### Narrow Domain



[Güler et al. 2018] (Denspose), [Rempe et al 2021] (HuMor) NeRF



#### Implicit



[Mildenhall et al. 2020], [Xie et al. 2021], [Xiang et al. 2021] [Mildenhall et al. 2020] (pix2surf), [Kanazawa et al. 2018] (CMR), [Xiang et al. 2021] (MCMR)

## Mesh-Based Differential Rendering



- Single mean shape [Kanazawa et al. 2018] (CMR)
- Multiple mean shapes [Xiang et al. 2021] (MCMR)



### Multi-Class Mesh Reconstruction (MCMR)



 $\checkmark$  General representation

- ✓ Shape-texture modularity
- $\checkmark$  Consistent representation

### Multi-Class Mesh Reconstruction (MCMR)



- General representation
  Shape-texture modularity
- $\checkmark$  Consistent representation

- $\times$  Texture quality
  - layout
- $\times$  Efficient 3D representation
  - class features

## More Efficient Representation







 $f_2(U)$ 











 $\downarrow \Delta V$ 



# Symmetry-Aware Representation

- Increase 3D accuracy for symmetric objects
- Reduces extra DoF
  - minimal network changes
  - better mesh quality



### Improved Texture Mapping



• Differential Rendering Mode

## Improved Texture Mapping



- Differential Rendering Mode
  - Grid vs UV-map sampling
- Texture layout
  - Symmetry-aware cut
  - Supervise UV-map quality



# UV-map distortions



# UV-map distortions



## Minimizing UV-map distortions





## Pix2Surf distortion losses

$$L_{\text{isom}} = \sum_{pk\sim ij} \left| \left\| xyz_{ij} - xyz_{pk} \right\| - \left\| uv_{ij} - uv_{pk} \right\| \right|$$

$$L_{flip}^{1/p} = \|Relu((\widehat{\Delta_1 uv} \times \widehat{\Delta_2 uv})_Z)^p\|_1$$



# Bibliography

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### Appendix: MCMR architecture



#### Appendix: Pix2Surf architecture



