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ICCV
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VIRTUAL



Toward Realistic Single-View 3D Object Reconstruction with Unsupervised Learning from Multiple Images

Long-Nhat Ho

Anh Tuan Tran^{1,2}

Quynh Phung¹

Minh Hoai^{1,3}

VinAI Research¹

VinUniversity²

Stony Brook University³

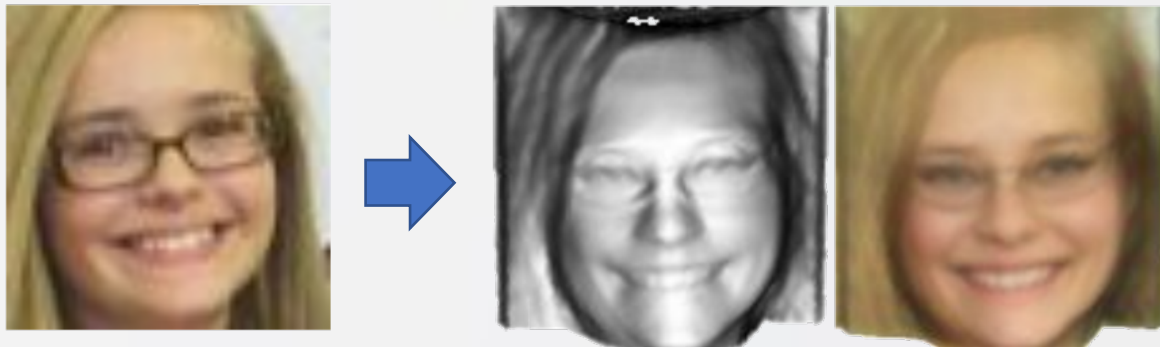




Motivation

Problem

Recover *3D structure* (shape + texture) of an object of a *known category* in a *single image*

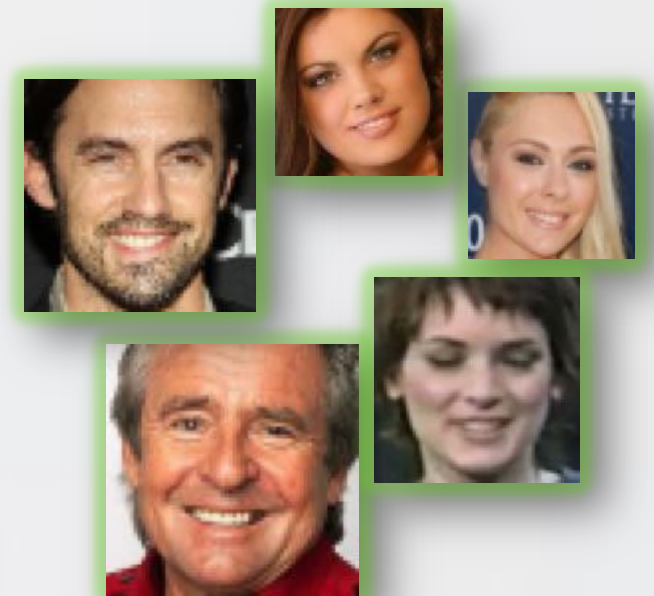
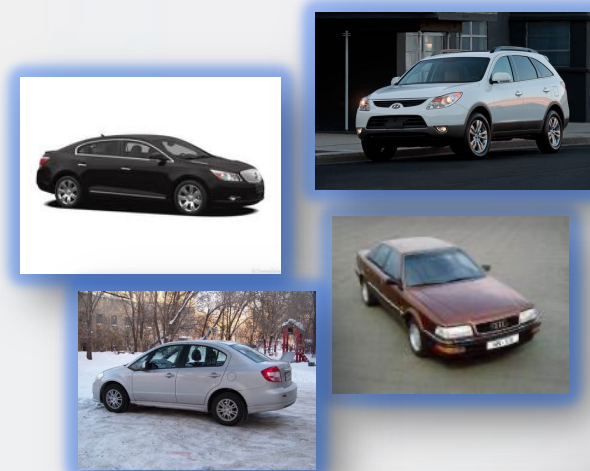


- ✗ Ill-posed problem
- ✓ Human is very good at this task via learning **3D shape prior**

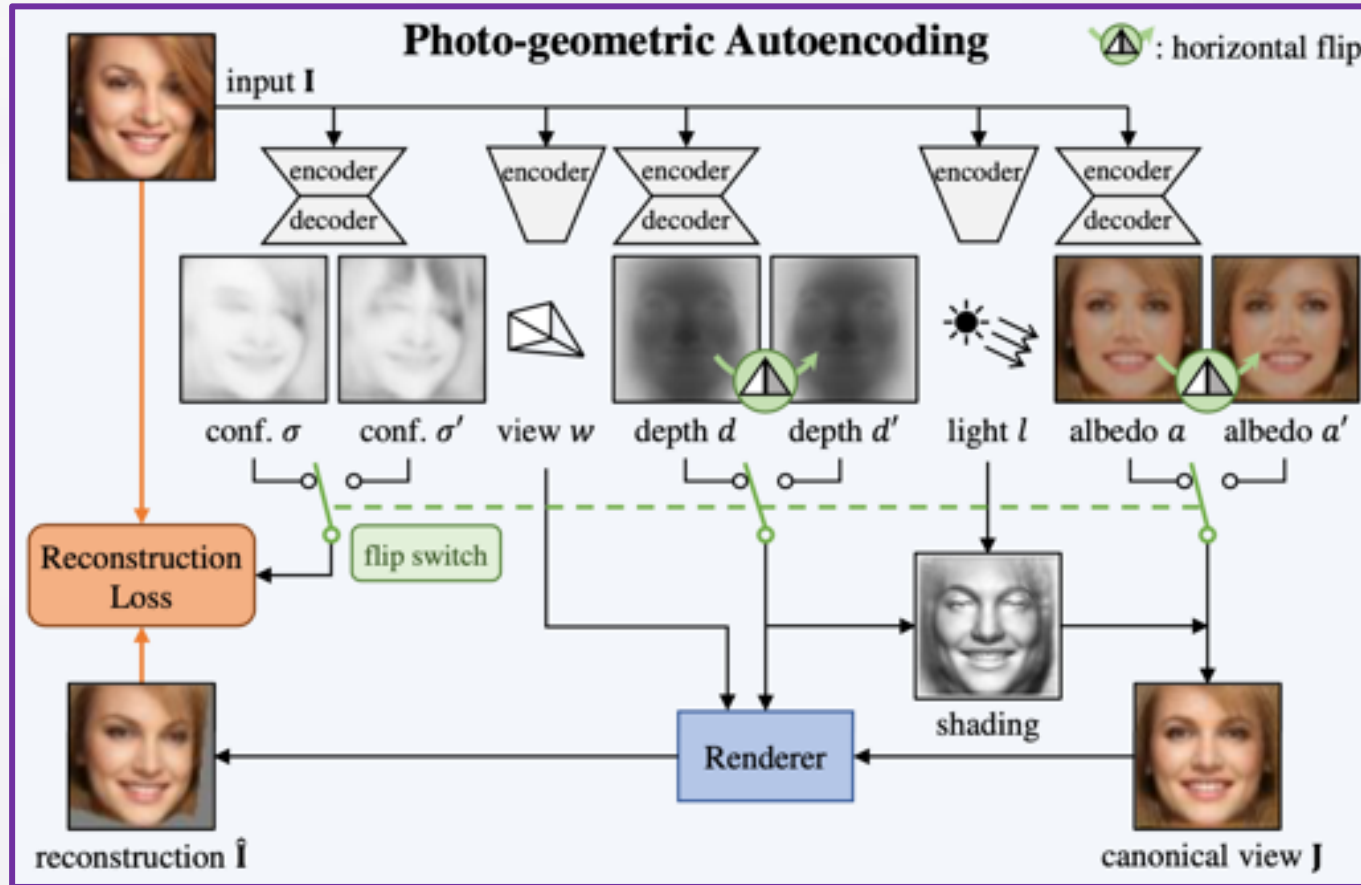
Problem

How to learn the 3D shape prior?

- Supervised
 - ✓ Require massive 3D data → hard to acquire
- **Unsupervised**
 - ✓ Observe 2D images of the same category



Previous approach – LeSym*



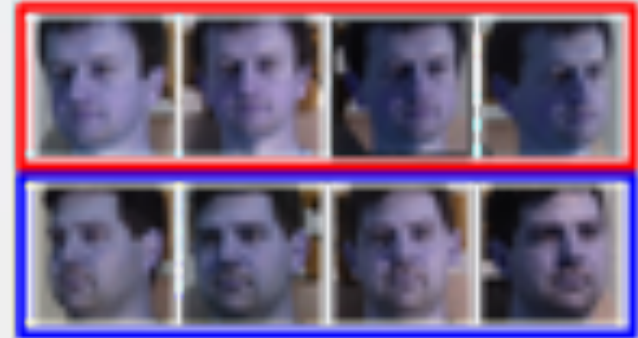
Only symmetric objects !!!

* S. Wu, C. Ruppert, and A. Vedaldi. "Unsupervised learning of probably symmetric deformable 3d objects from images in the wild". In *CVPR 2020*.

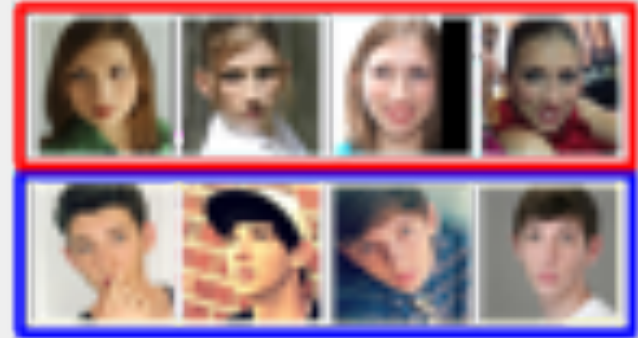
Our solution?

- Many datasets have *multiple images* for each *object instance*
 - ✓ Cover symmetric objects
- Shape consistency

Multi-view



Collection



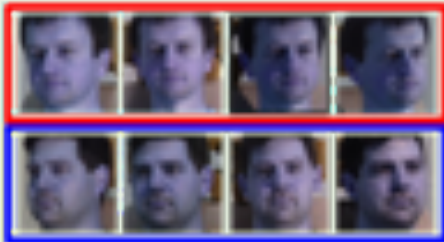
Video



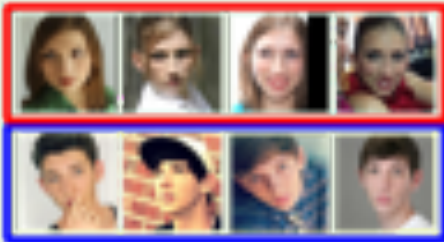
LeMul

Training

Multi-view



Collection



Video



Testing

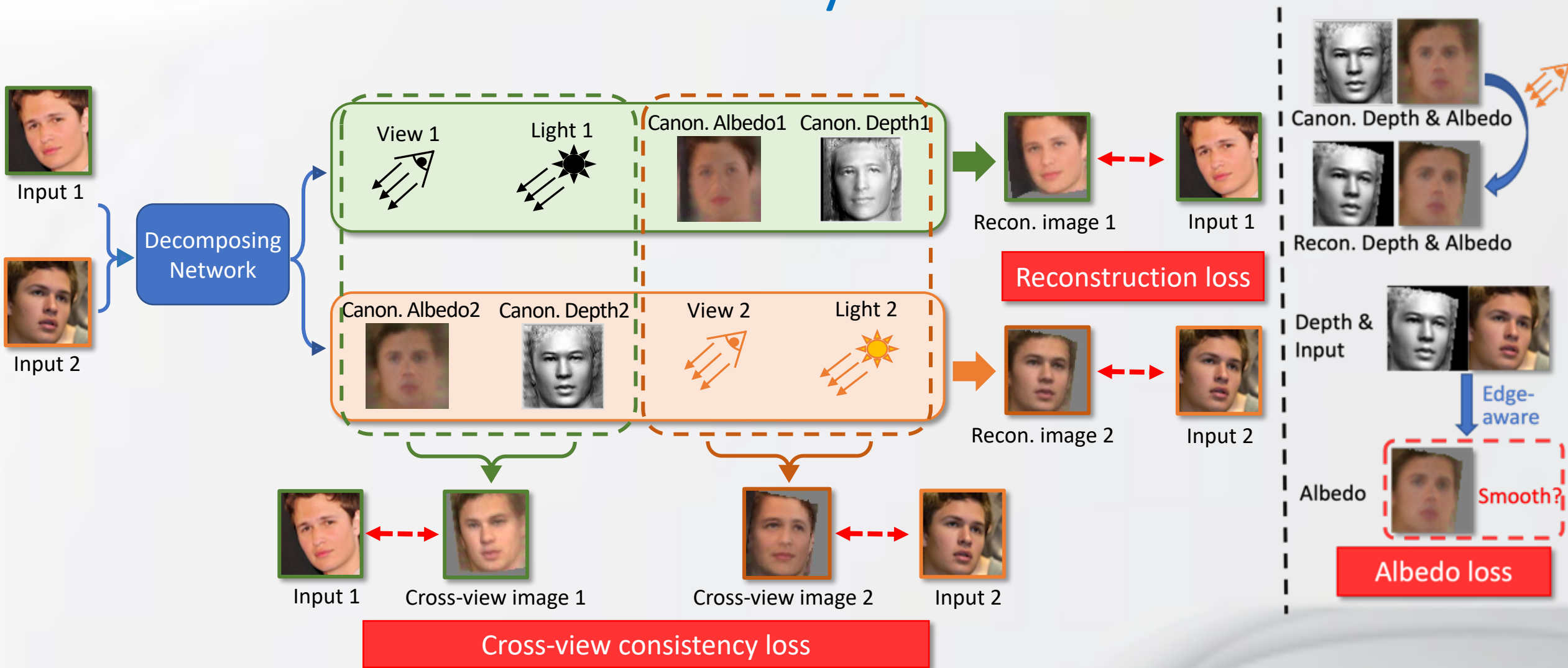


Learning from Multi-Image Datasets - *LeMul*



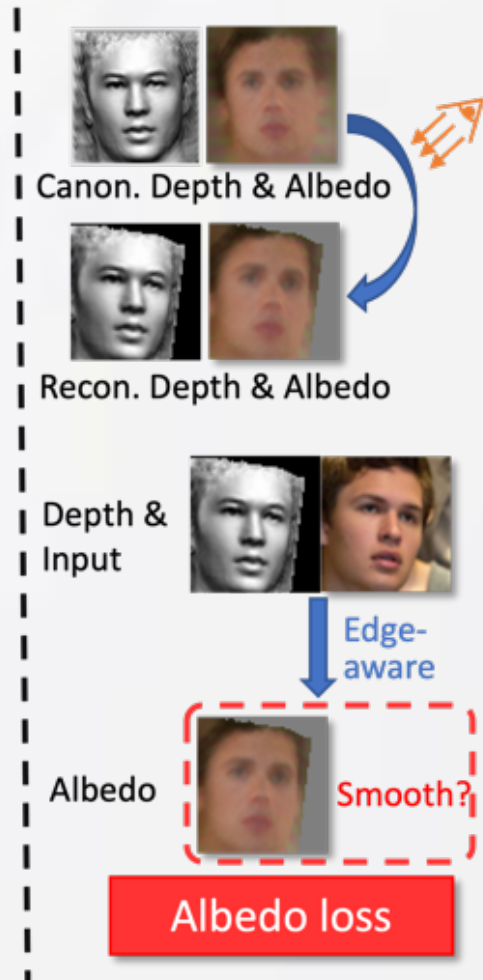
LeMul System

LeMul system



*Note that we omit the confidence maps in this figure for simplicity

LeMul system



$$\mathbb{L}^{al}(\mathbf{I}, a, d) = \frac{1}{|\Omega|} \sum_{p \in \Omega} \left\| \sum_{p_k \in \mathcal{N}(p)} w_k^c w_k^d (a(p) - a(p_k)) \right\|^2$$

Where:

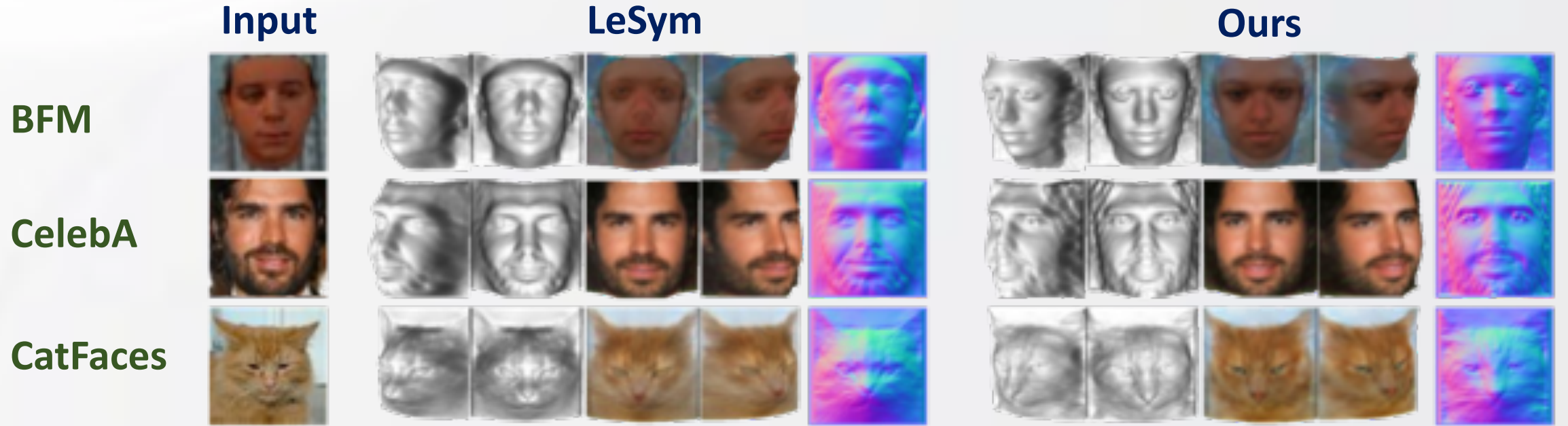
$\mathcal{N}(p)$: the neighbors of a pixel

w_k^c : the intensity weighting term

w_k^d : the depth weighting term

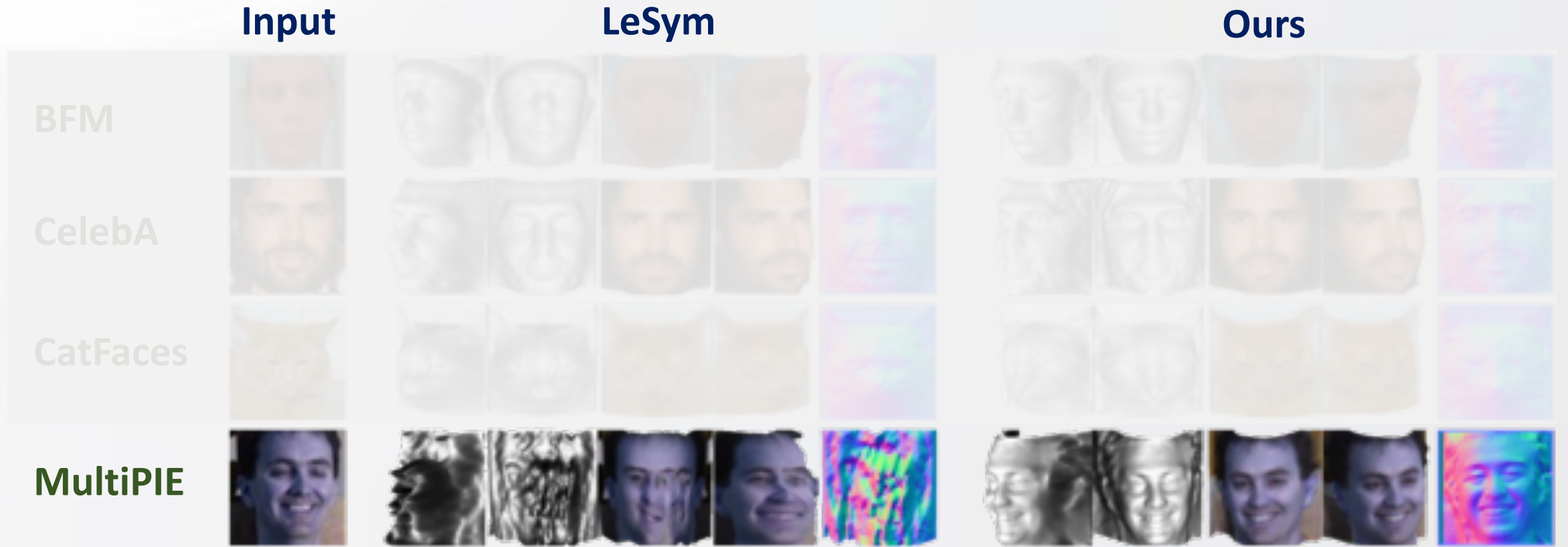
Results

Qualitative results



single-image, symmetric objects

Qualitative results



multi-view dataset

Qualitative results

Input

LeSym

Ours

BFM

CelebA

CatFaces

MultiPIE

CASIA



Qualitative results

Input

LeSym

Ours

BFM

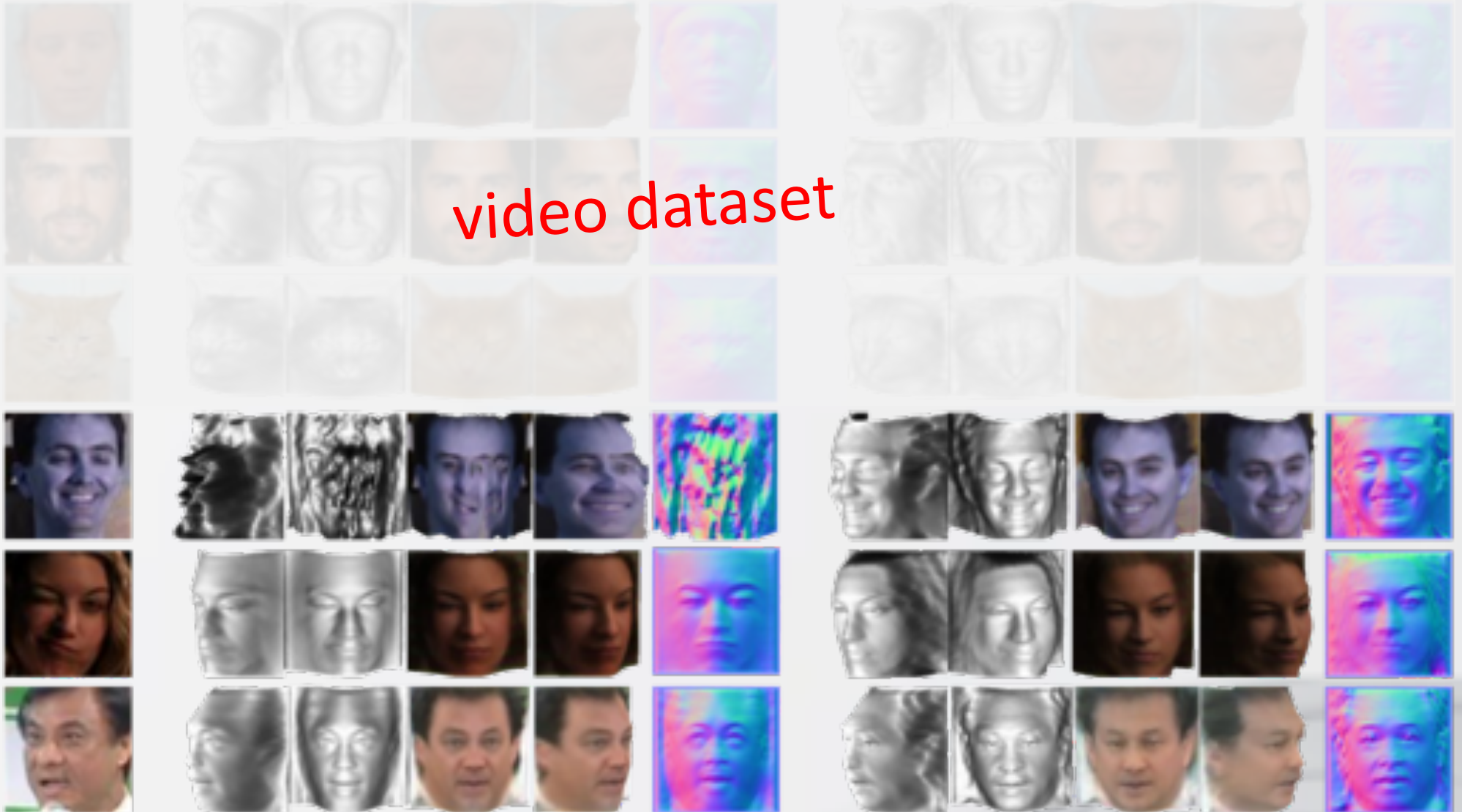
CelebA

CatFaces

MultiPIE

CASIA

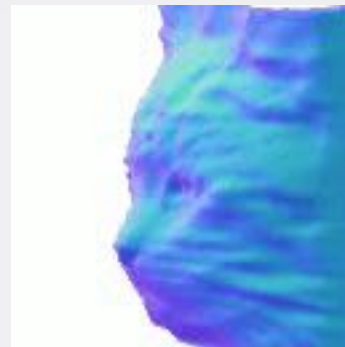
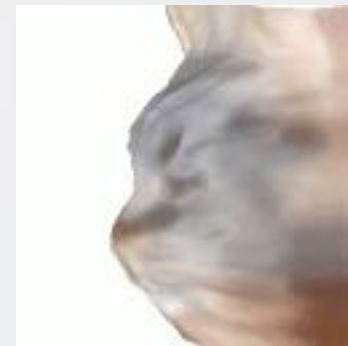
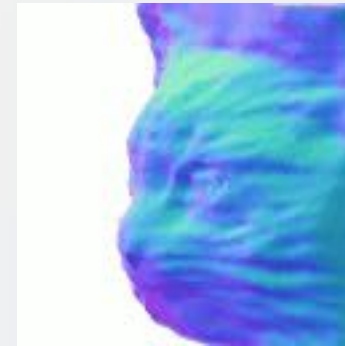
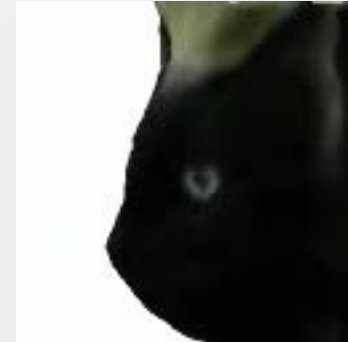
YTF



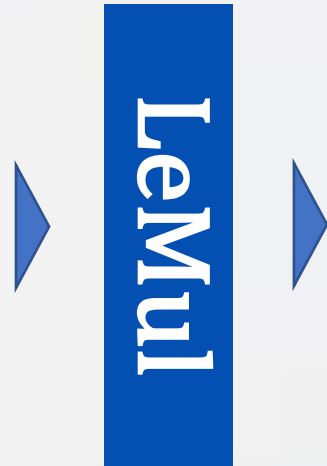
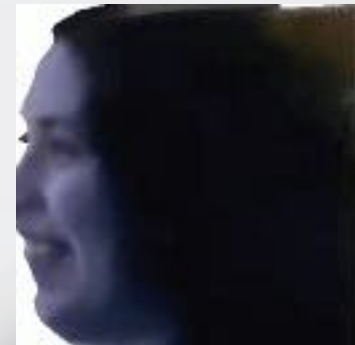
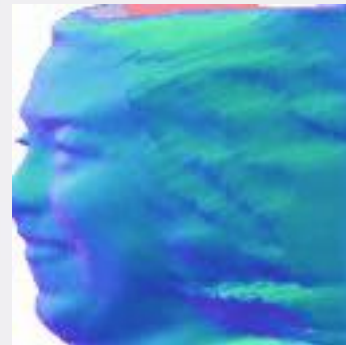
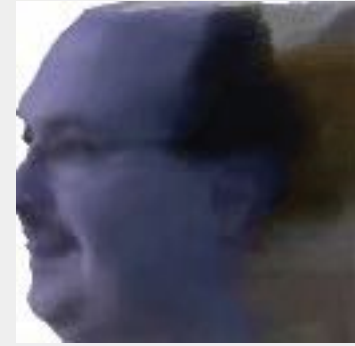
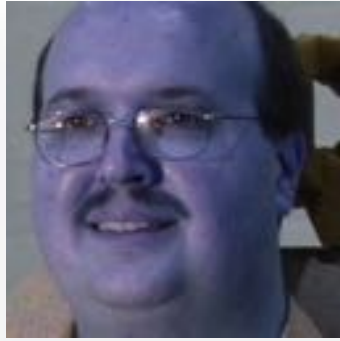
Cat Faces (single + symmetric)



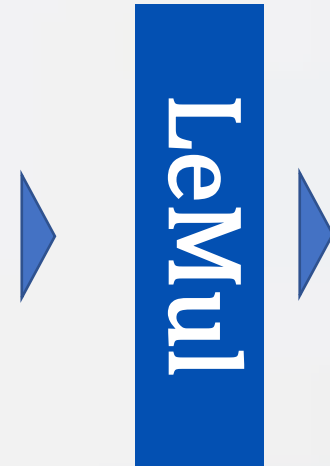
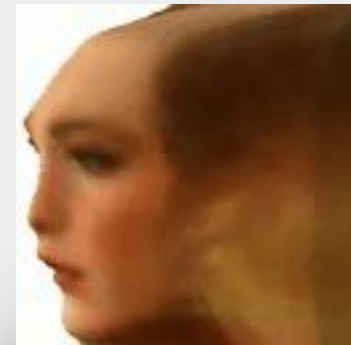
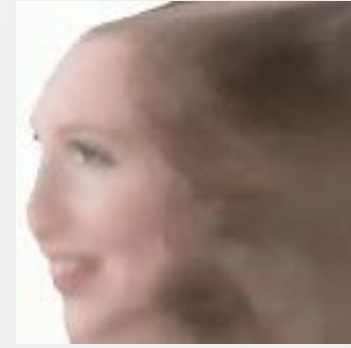
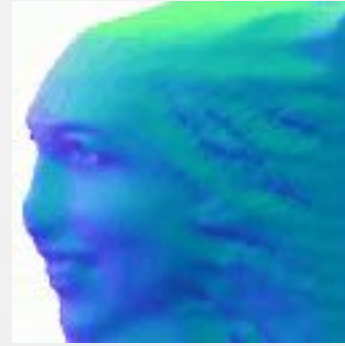
LeMUI



Multi-PIE (multi-view)



CASIA-WebFace (image collection)



CASIA-WebFace (image collection)

In-the-wild

Input

LeSym (CelebA)

LeMul (CASIA)



Quantitative results

- ✓ Better surface reconstruction on BFM
- ✓ Better voted via user surveys on all datasets

No	Baseline	SIDE($\times 10^{-2}$) \downarrow	MAD(deg.) \downarrow
(1)	Supervised	0.410 ± 0.103	10.78 ± 1.01
(2)	Const. null depth	2.723 ± 0.371	43.34 ± 2.25
(3)	Average G.T. depth	1.990 ± 0.556	23.26 ± 2.85
(4)	LeSym	0.793 ± 0.140	16.51 ± 1.56
(5)	LeMul (proposed)	0.834 ± 0.169	15.49 ± 1.50

BFM results comparison with baselines.



THANK YOU

<https://github.com/VinAIResearch/LeMul>