

Classical and Learning-based Approaches to 4D Reconstruction

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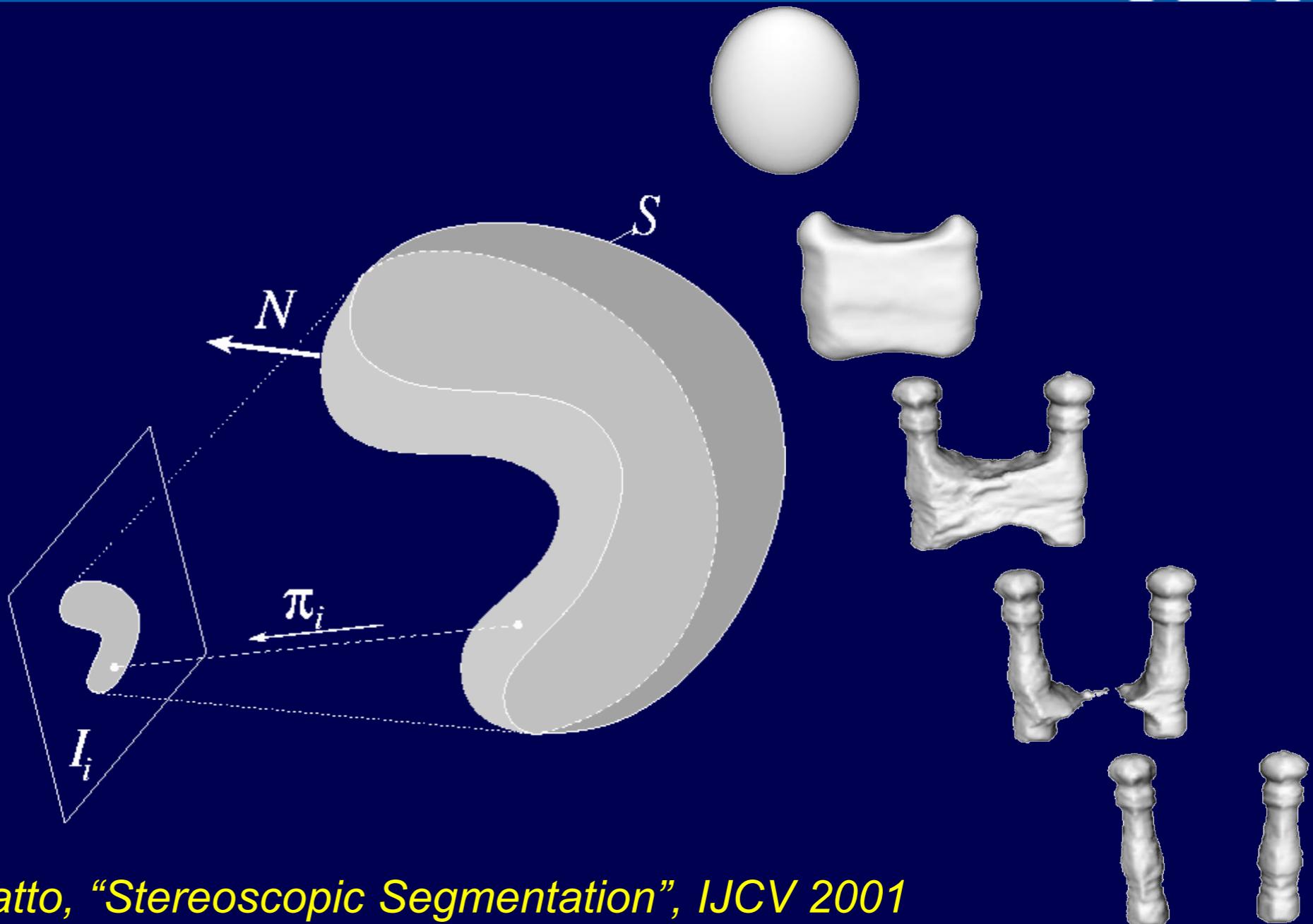
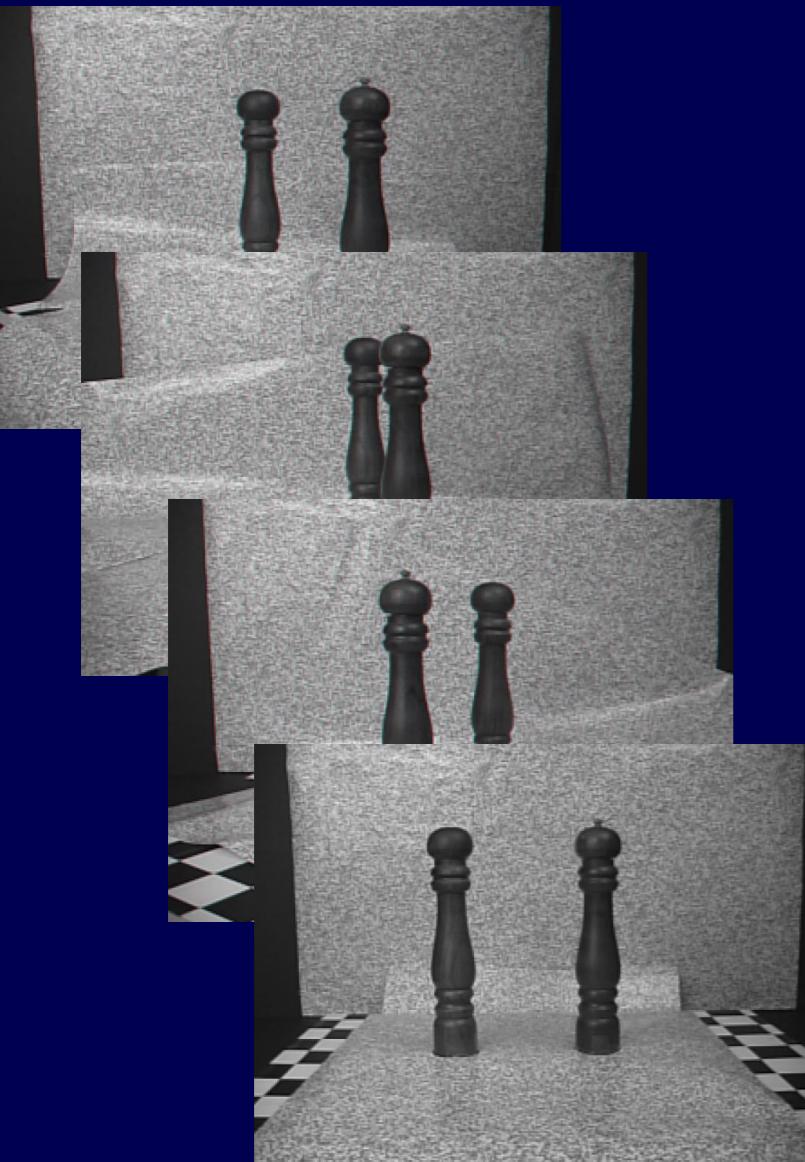
Munich Center for Machine Learning



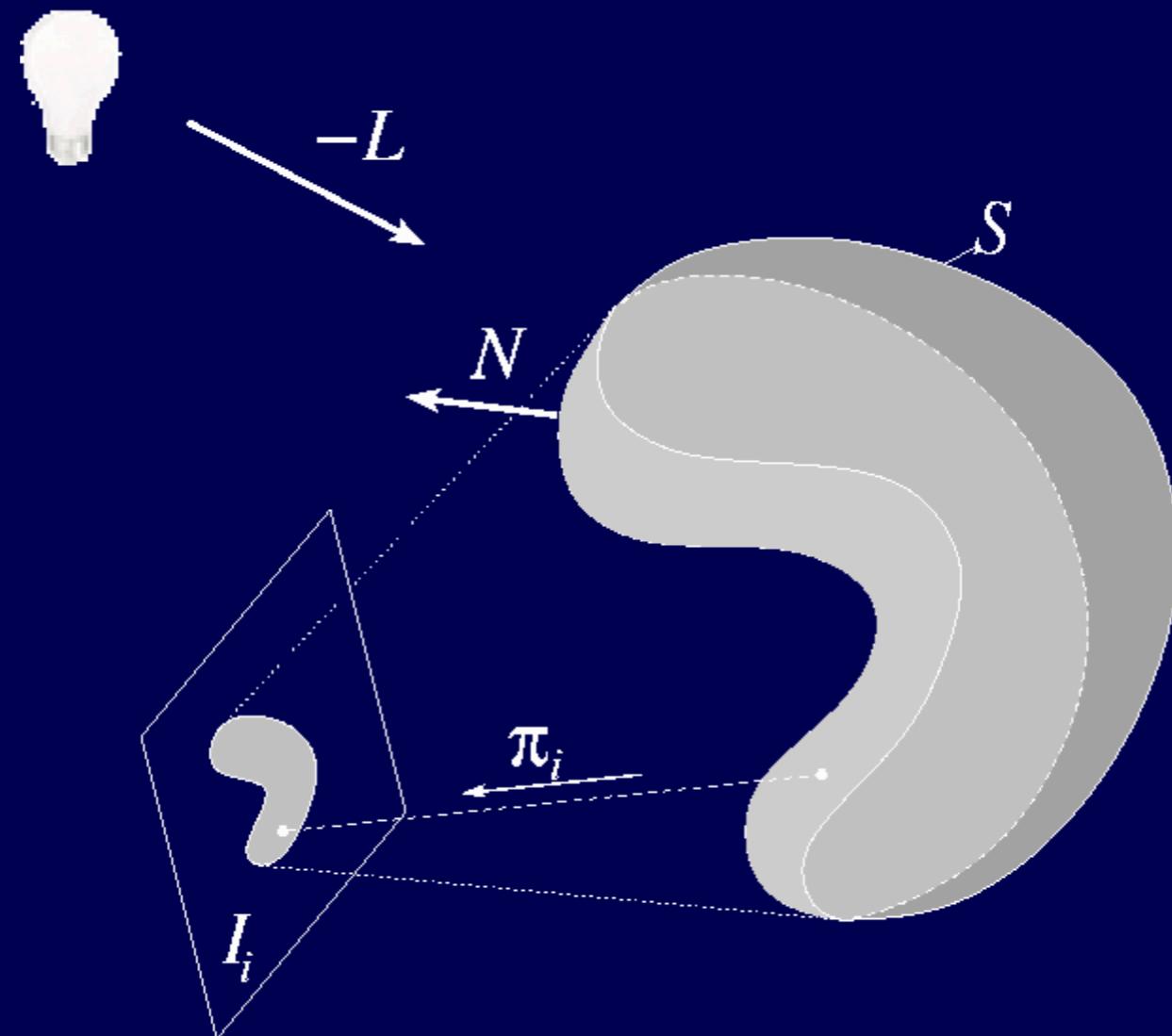
Reconstructing 3D Shape



Reconstructing 3D Shape



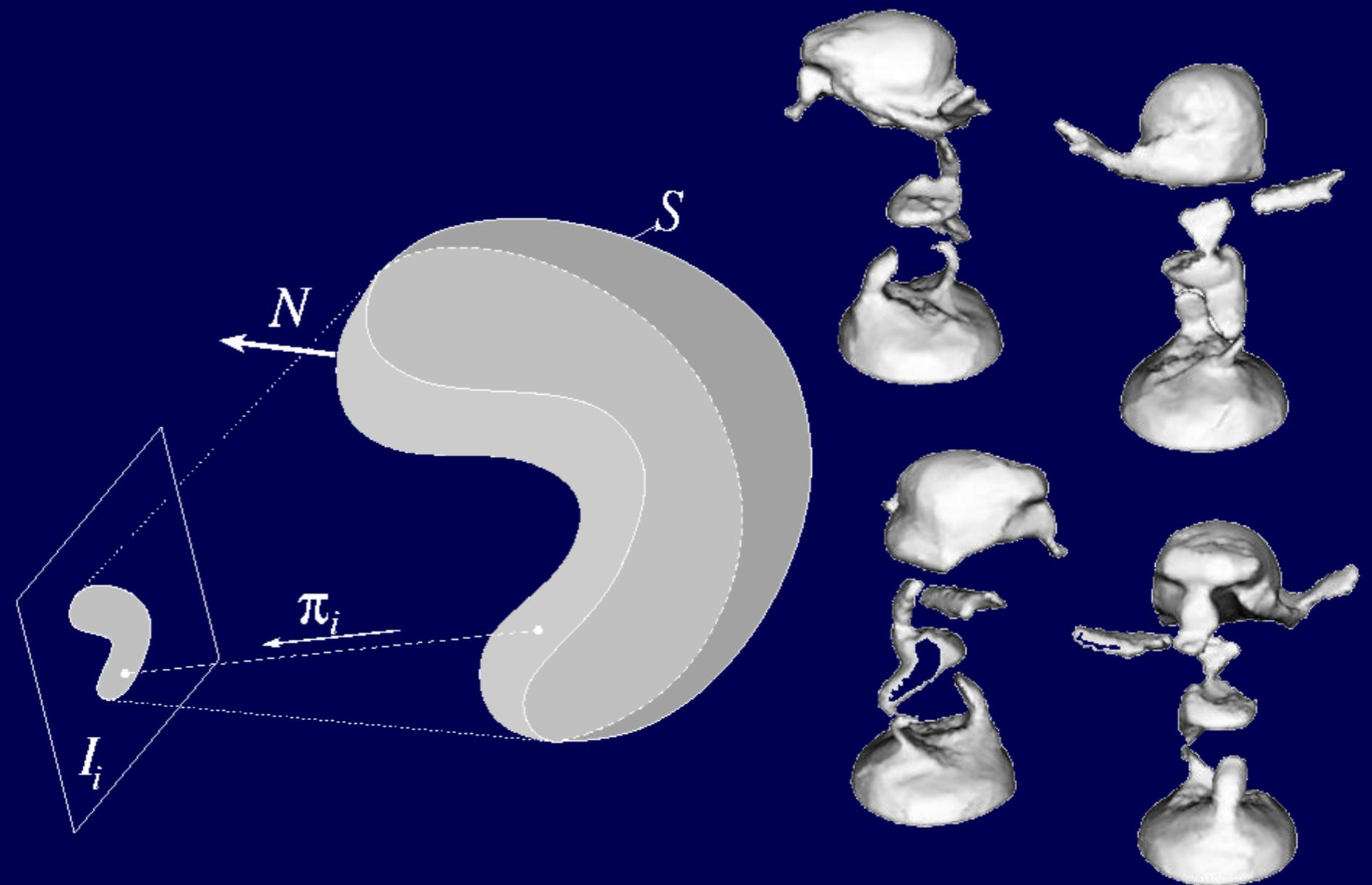
Yezzi, Soatto, "Stereoscopic Segmentation", IJCV 2001



Jin, Cremers, Yezzi, Soatto, “*Shedding Light on Stereoscopic Segmentation*”, CVPR 2004



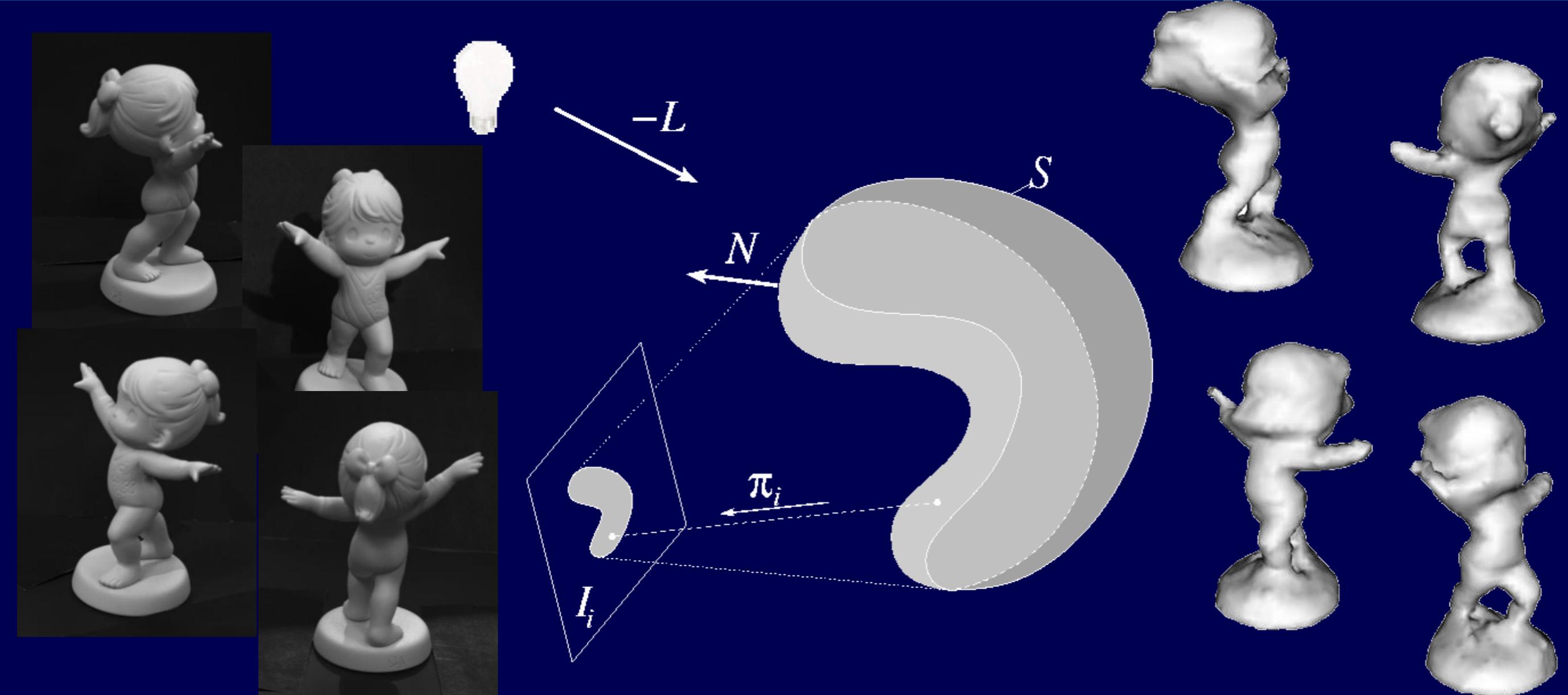
Reconstructing 3D Shape



Yezzi, Soatto, "Stereoscopic Segmentation", IJCV 2001

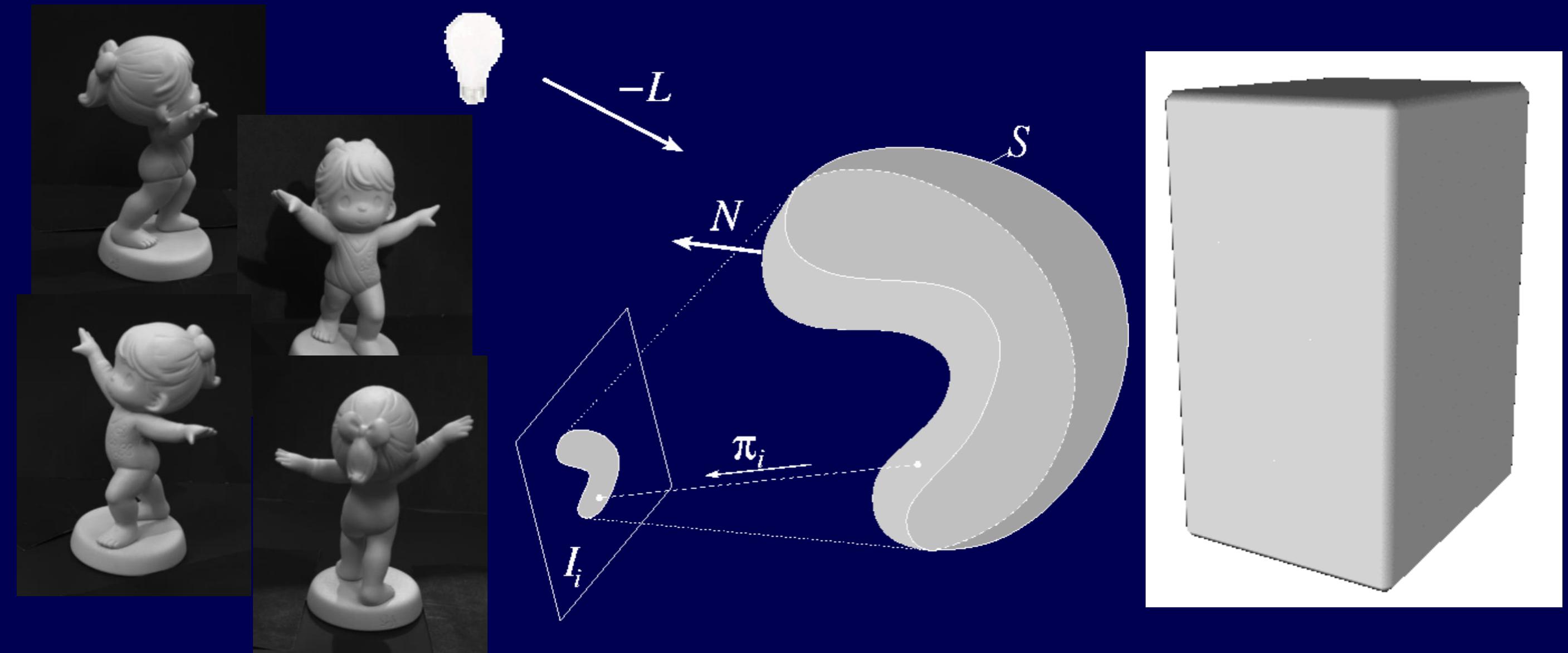


Reconstructing 3D Shape and Lighting



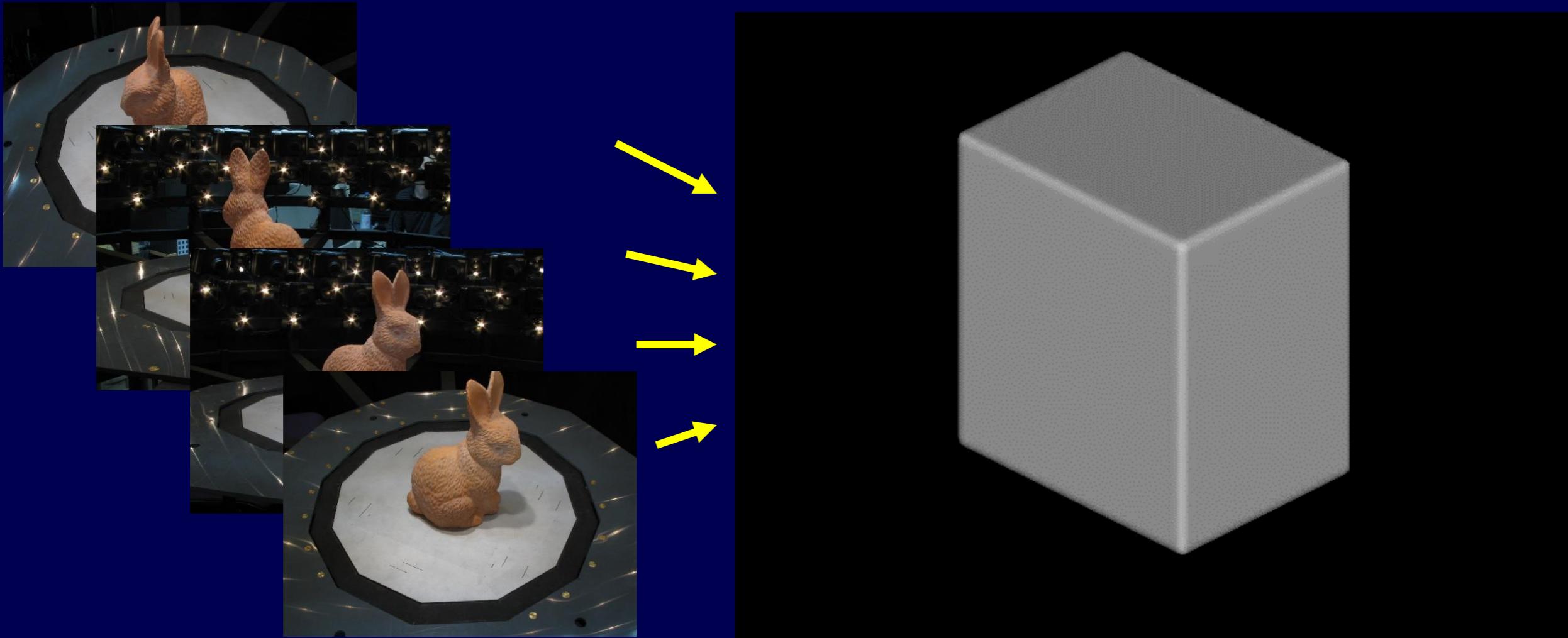
Jin, Cremers, Yezzi, Soatto, “*Shedding Light on Stereoscopic Segmentation*”, CVPR 2004

Reconstructing 3D Shape and Lighting



Jin, Cremers, Yezzi, Soatto, “*Shedding Light on Stereoscopic Segmentation*”, CVPR 2004

Provably Optimal Reconstructions



Kolev, Klodt, Brox, Cremers, Int. J. of Computer Vision '09:
Theorem: Optimal reconstructions can be computed via convex relaxation.



Super-Resolution Texture



*Goldlücke, Cremers, ICCV '09, DAGM '09**

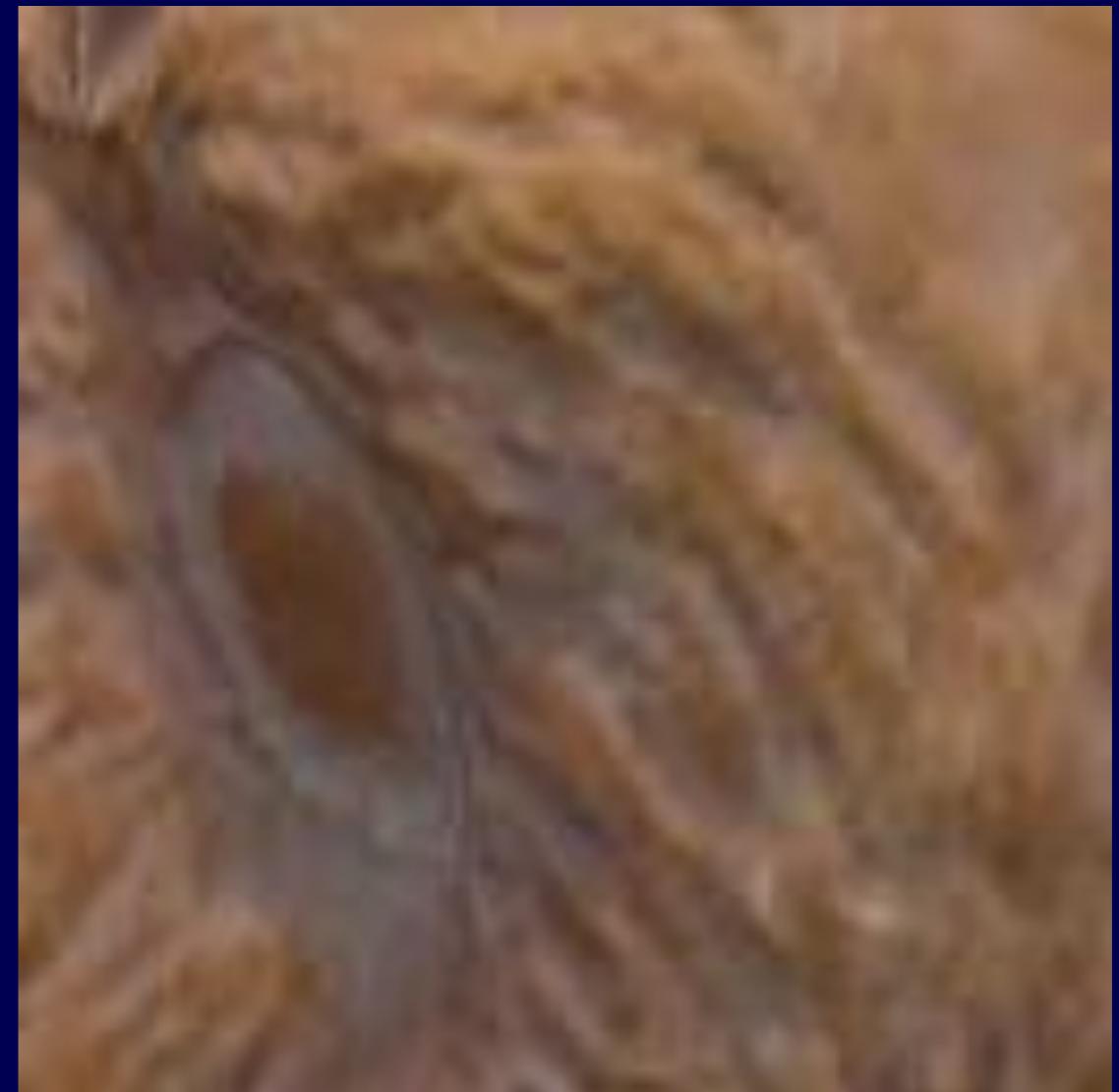
** Best Paper Award*



Super-Resolution Texture



Closeup of input image



Super-resolution texture

*Goldlücke, Cremers, ICCV '09, DAGM '09**

** Best Paper Award*

Single-View 3D Reconstruction

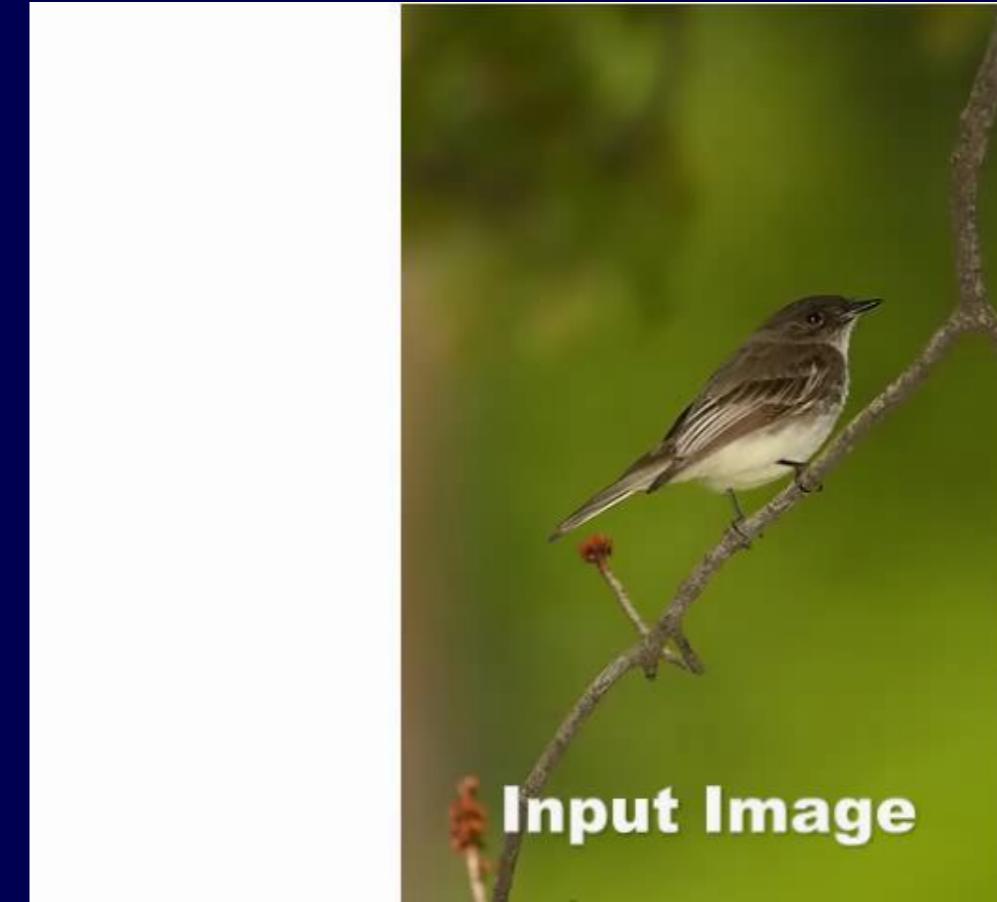


Fixed-volume silhouette-consistent minimal surface:

$$\min_S |S| \quad \text{s.t. } \text{Vol}(S) = V_0, \pi(S) = S_0$$

Toeppe et al., ACCV 2010, Oswald, Cremers, CVPR 2012*

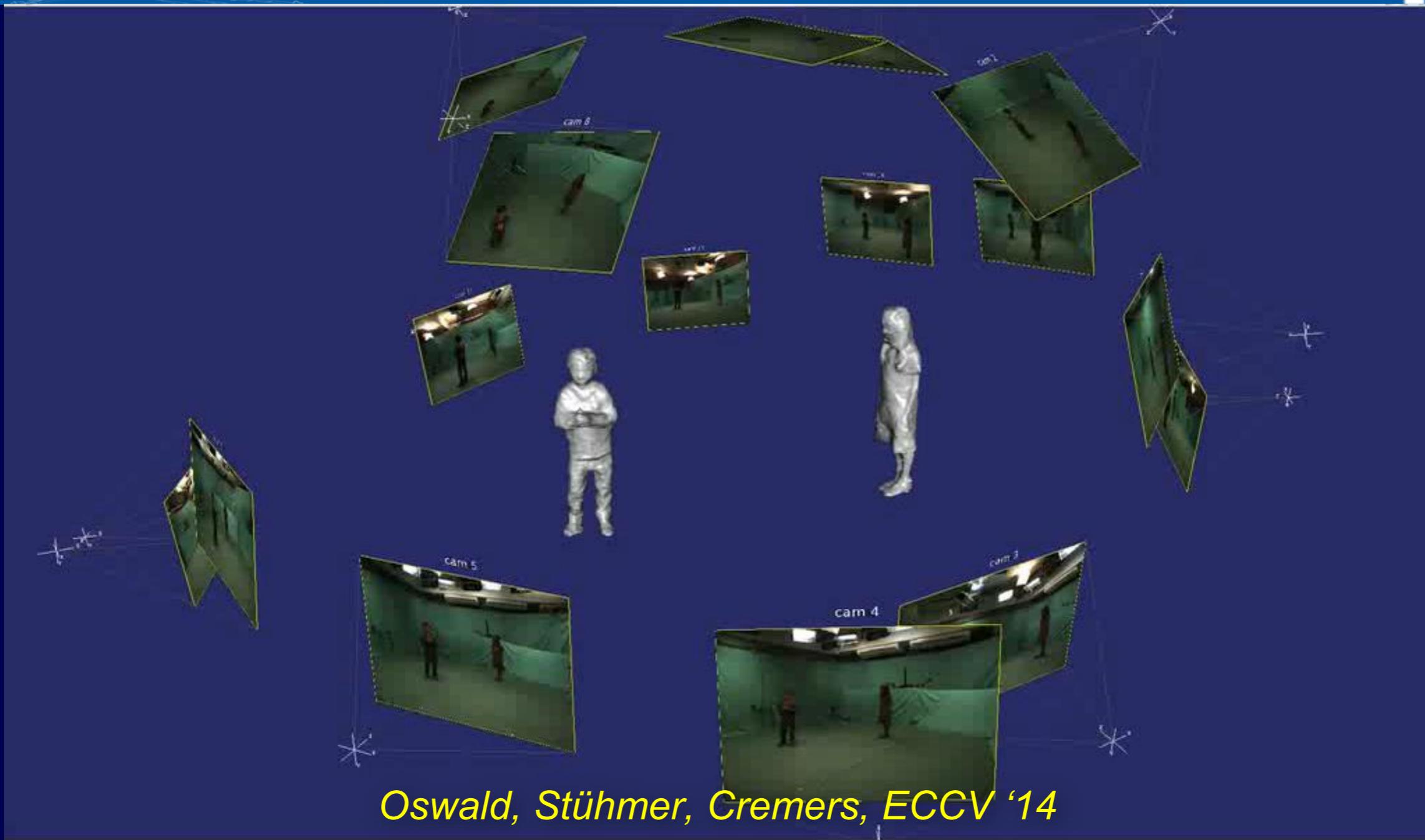
** Best Paper Honorable Mention*



Toeppe et al., ACCV 2010 , Oswald, Cremers, CVPR 2012*

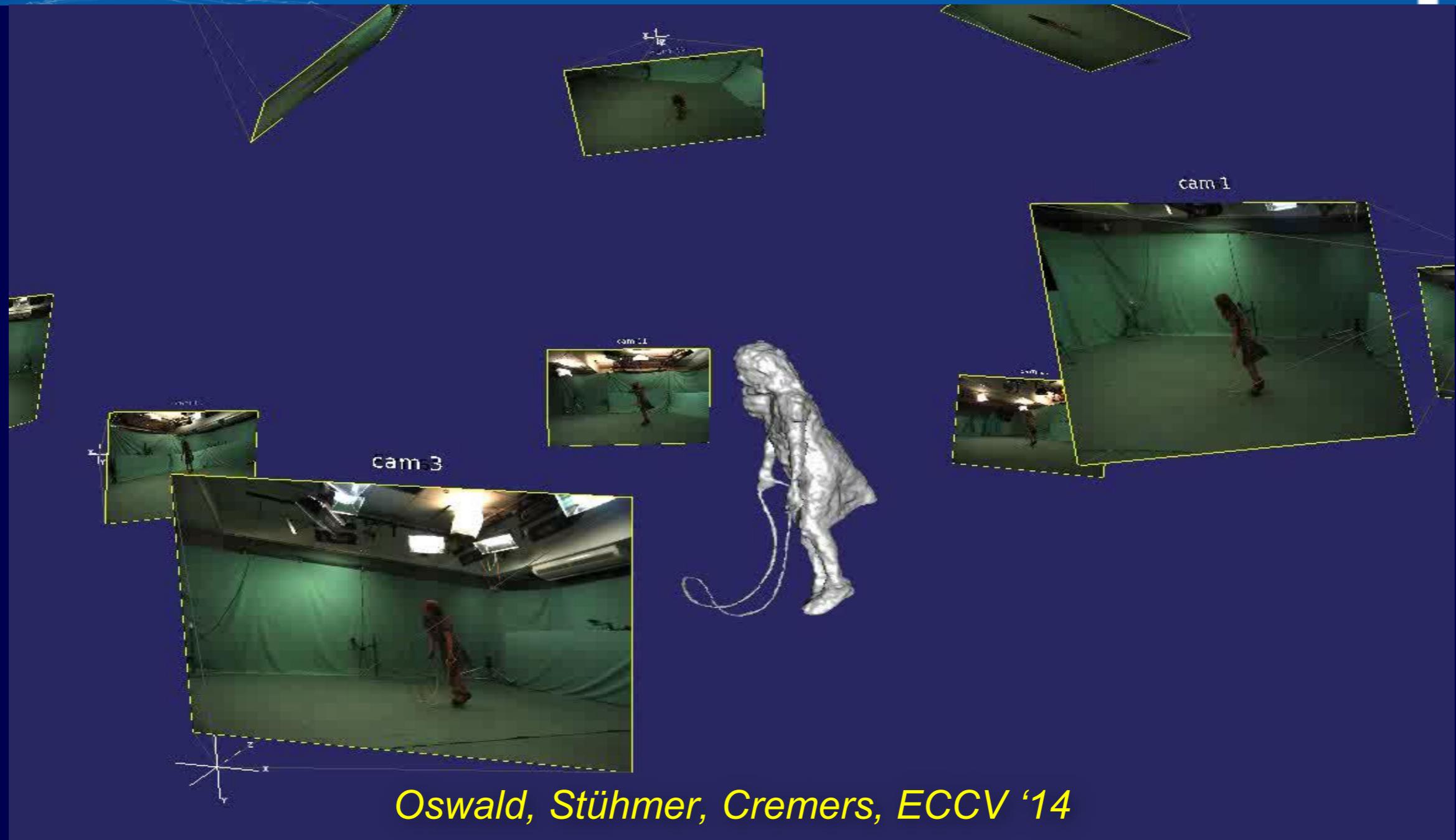
** Best Paper Honorable Mention*

4D Reconstruction from Multiview Video



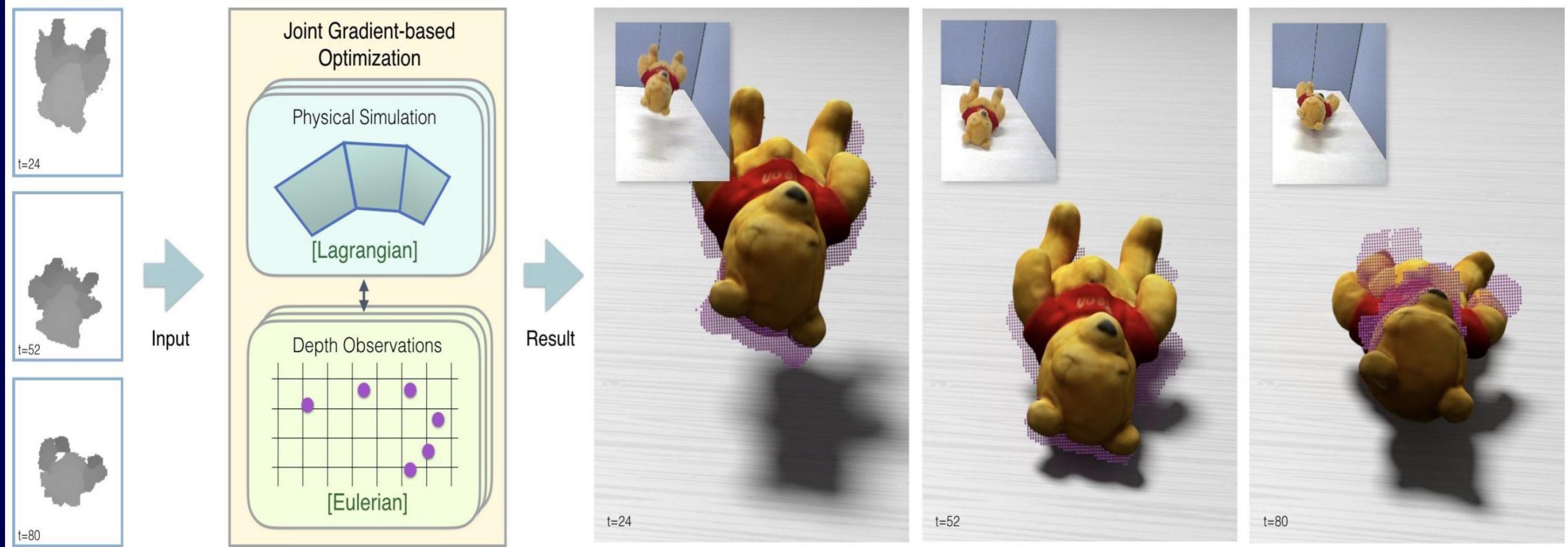
Oswald, Stühmer, Cremers, ECCV '14

4D Reconstruction from Multiview Video



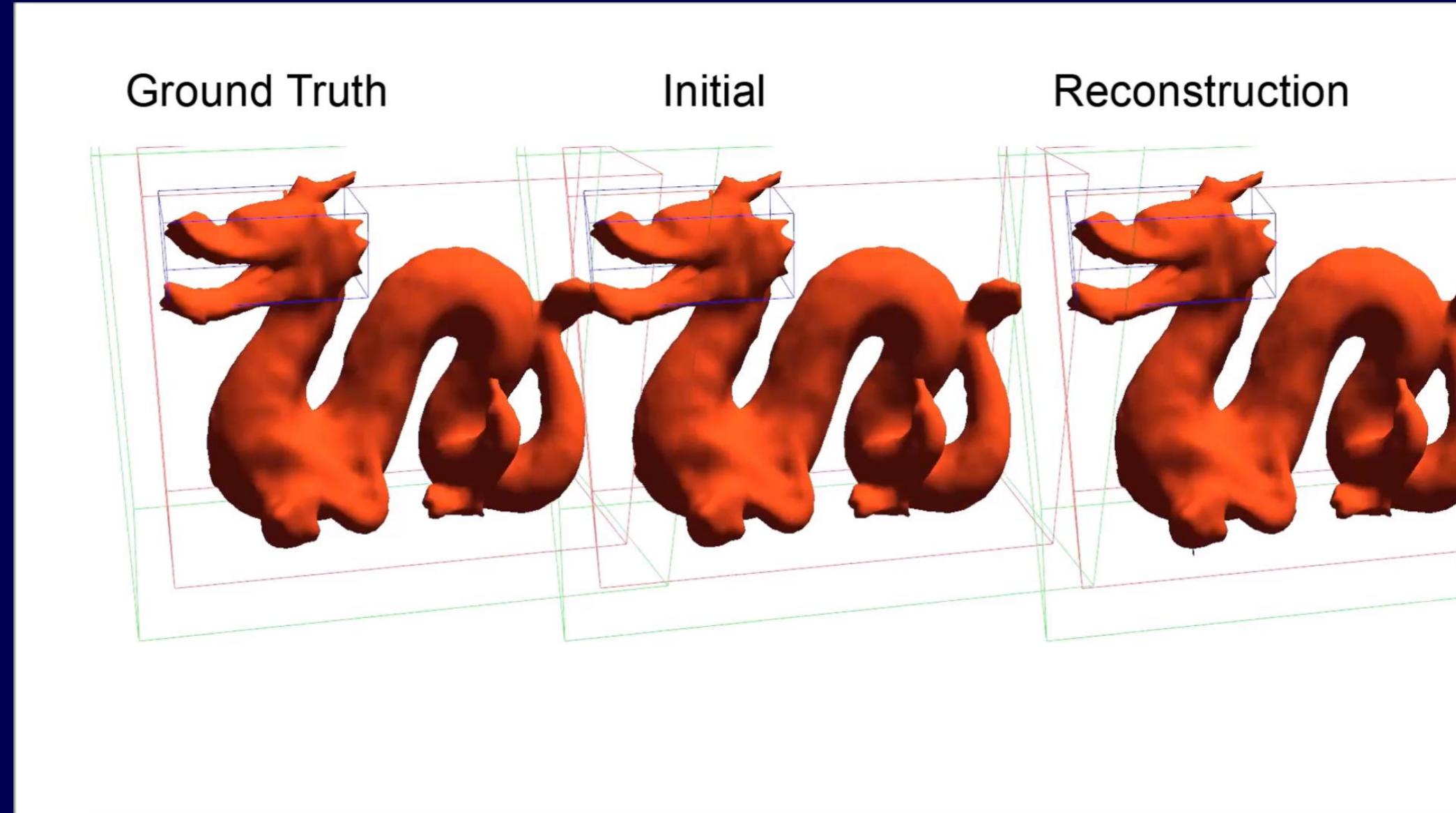
Oswald, Stühmer, Cremers, ECCV '14

Reconstructing Physical Simulations from Video



Weiss et al., CVPR 2020

Reconstructing Physical Simulations from Video



Weiss et al., CVPR 2020

Observation with an RGB-D Camera

Color



Filtered Depth



Weiss et al., CVPR 2020



Engel, Schöps, Cremers, “LSD SLAM: Large-Scale Direct Monocular SLAM”, ECCV ‘14

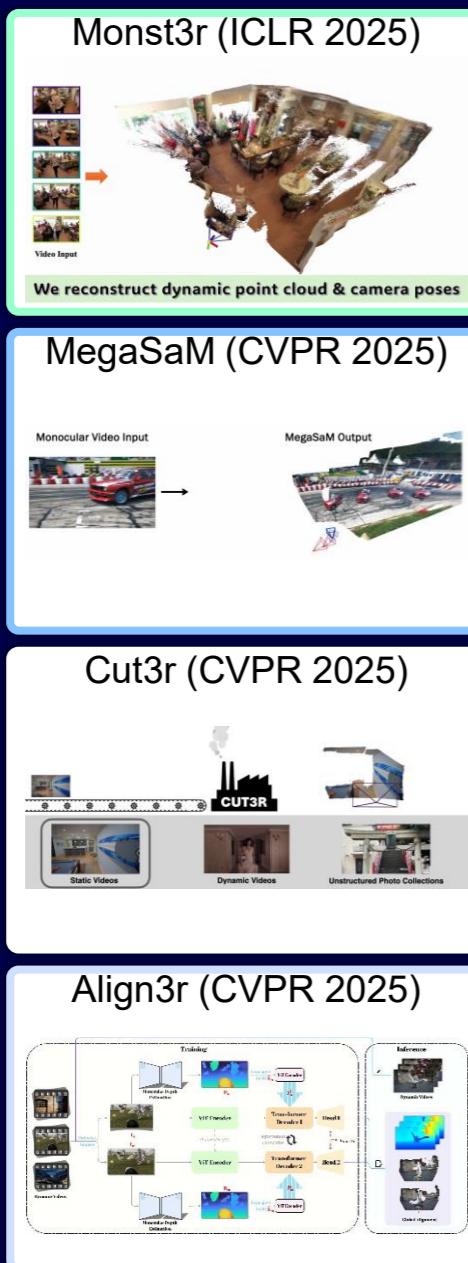
Dense Reconstructions from a Single Camera



Wimbauer et al., “MonoRec: Monocular Dense Reconstruction”, CVPR ‘21

Reconstruction from Casual Videos





Monst3r (ICLR 2025)

We reconstruct dynamic point cloud & camera poses

MegaSaM (CVPR 2025)

Monocular Video Input → MegaSaM Output

Cut3r (CVPR 2025)

Static Videos, Dynamic Videos, Unstructured Photo Collections

Align3r (CVPR 2025)

Training and Inference flowchart showing video inputs, encoders, decoders, and losses.

Impressive results

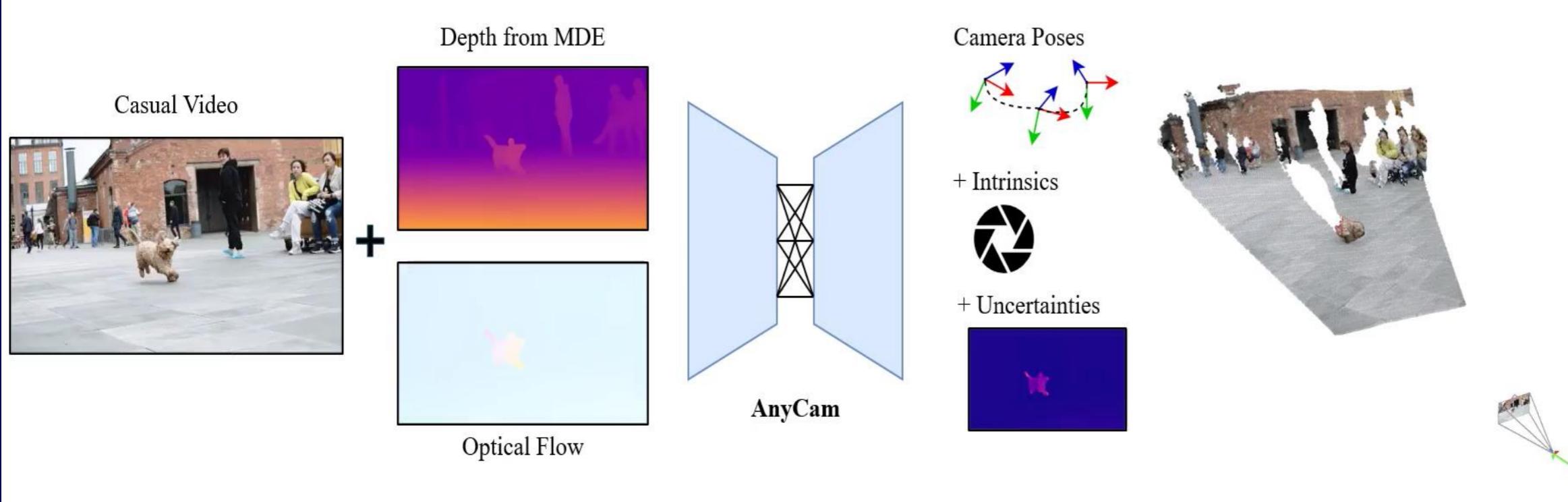
but

Supervised training



- ⚠ Expensive data collection
- ⚠ Limited datasets
- ⚠ Dataset biases
- ⚠ Sim-to-real gap

AnyCam: Reconstruction from Casual Videos



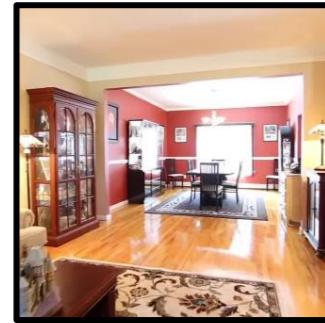
Felix Wimbauer^{1,2,3} Weirong Chen^{1,2,3} Dominik Muhle^{1,2} Christian Rupprecht³ Daniel Cremers^{1,2}

¹Technical University of Munich ²MCML ³University of Oxford

Wimbauer et al., “AnyCam: Learning to Recover Camera Poses and Intrinsics from Casual Videos”, CVPR ‘25

AnyCam is self-supervised on casual videos:

RealEstate10K



WalkingTours



OpenDV



YouTube VOS



EpicKitchens

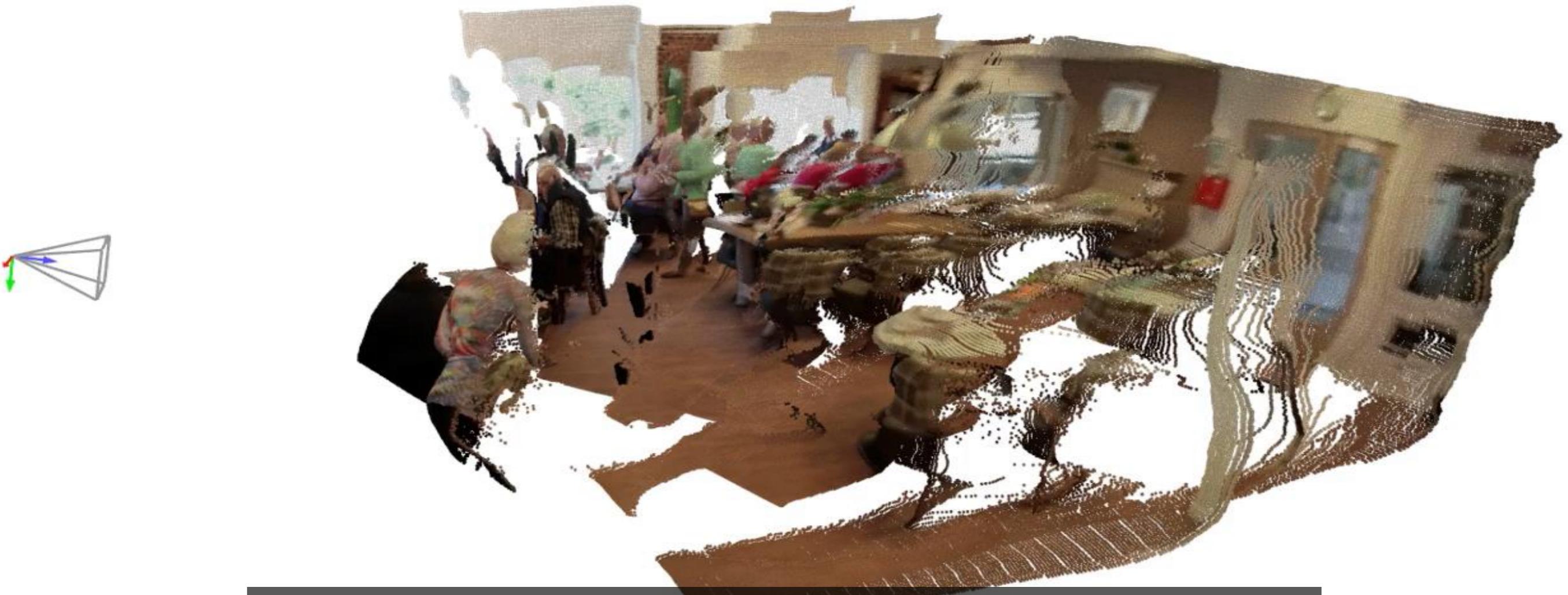


From YouTube

No ground truth data

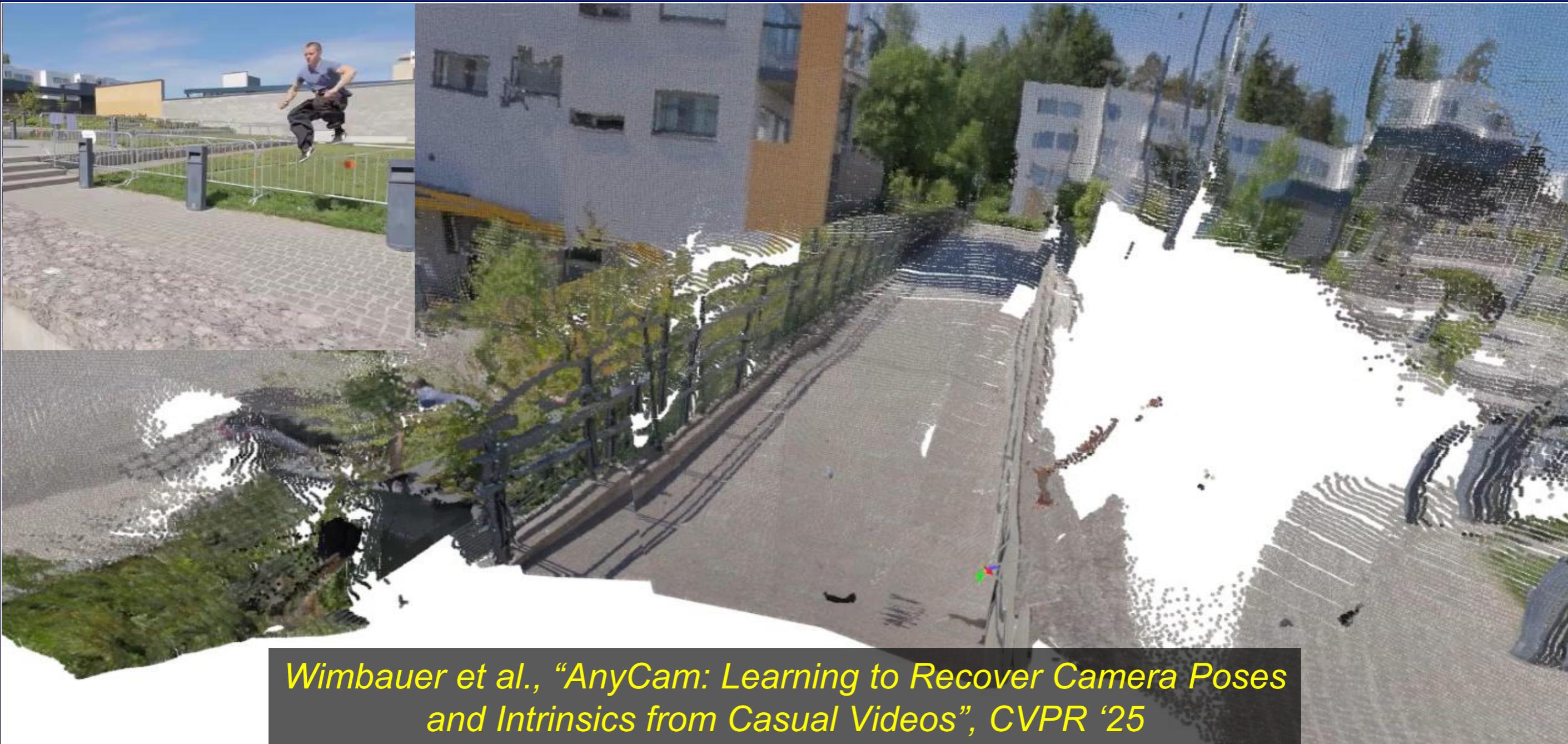
Wimbauer et al., “AnyCam: Learning to Recover Camera Poses and Intrinsics from Casual Videos”, CVPR ‘25

AnyCam: Reconstruction from Casual Videos



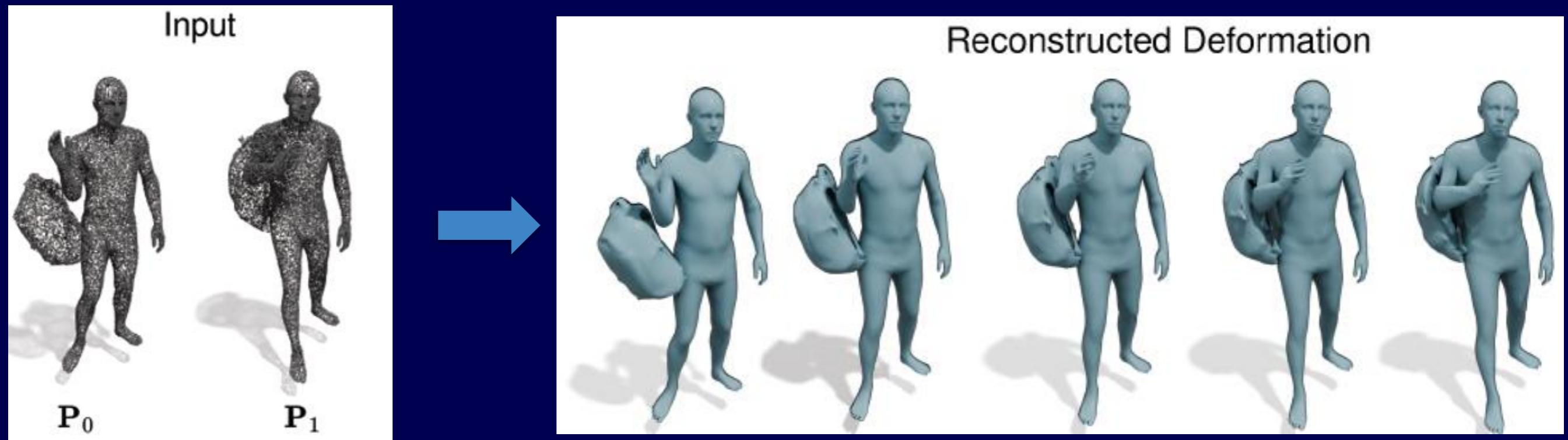
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AnyCam: Reconstruction from Casual Videos



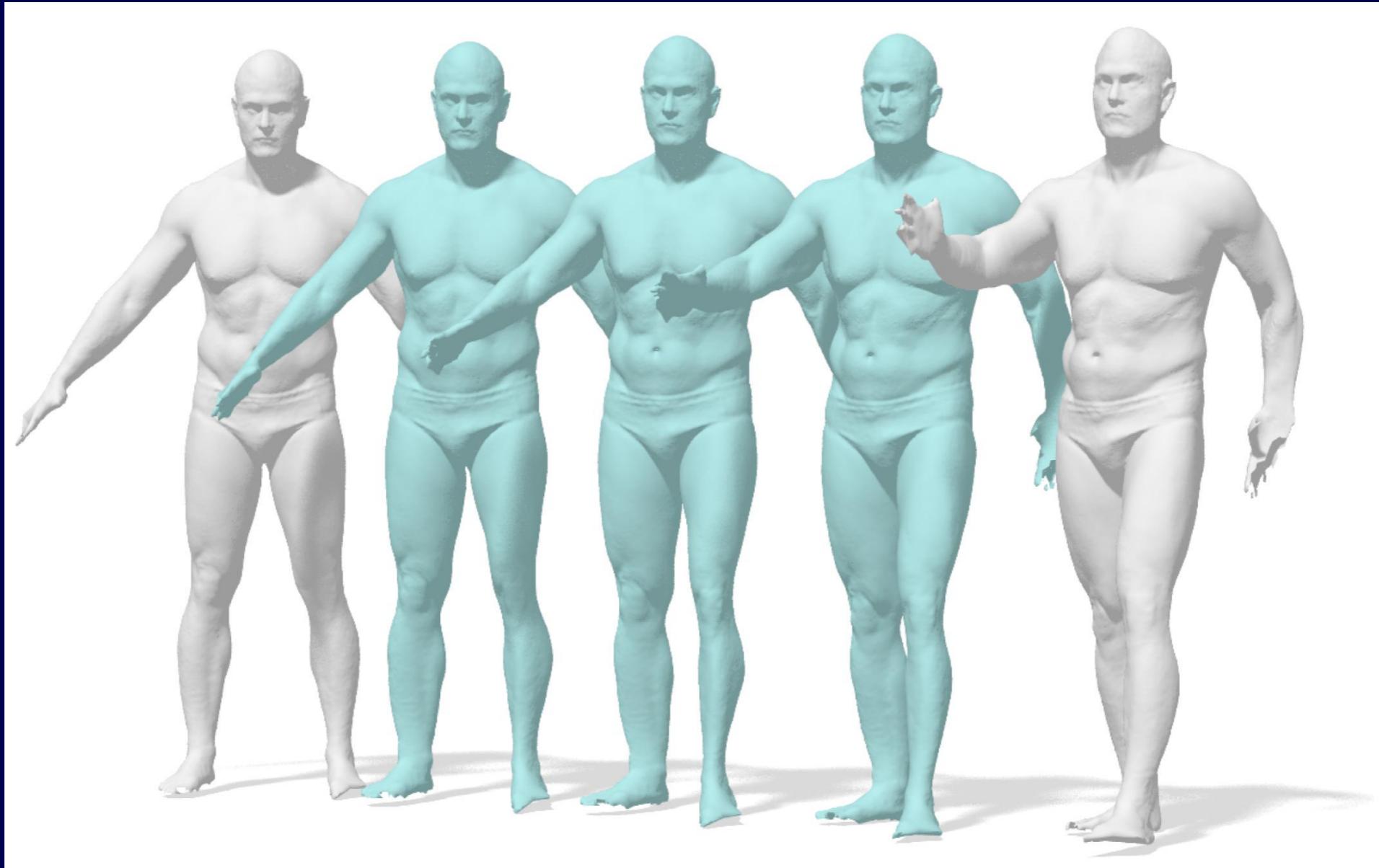
Wimbauer et al., “AnyCam: Learning to Recover Camera Poses and Intrinsics from Casual Videos”, CVPR ‘25

4D Reconstruction from Sparse Observations



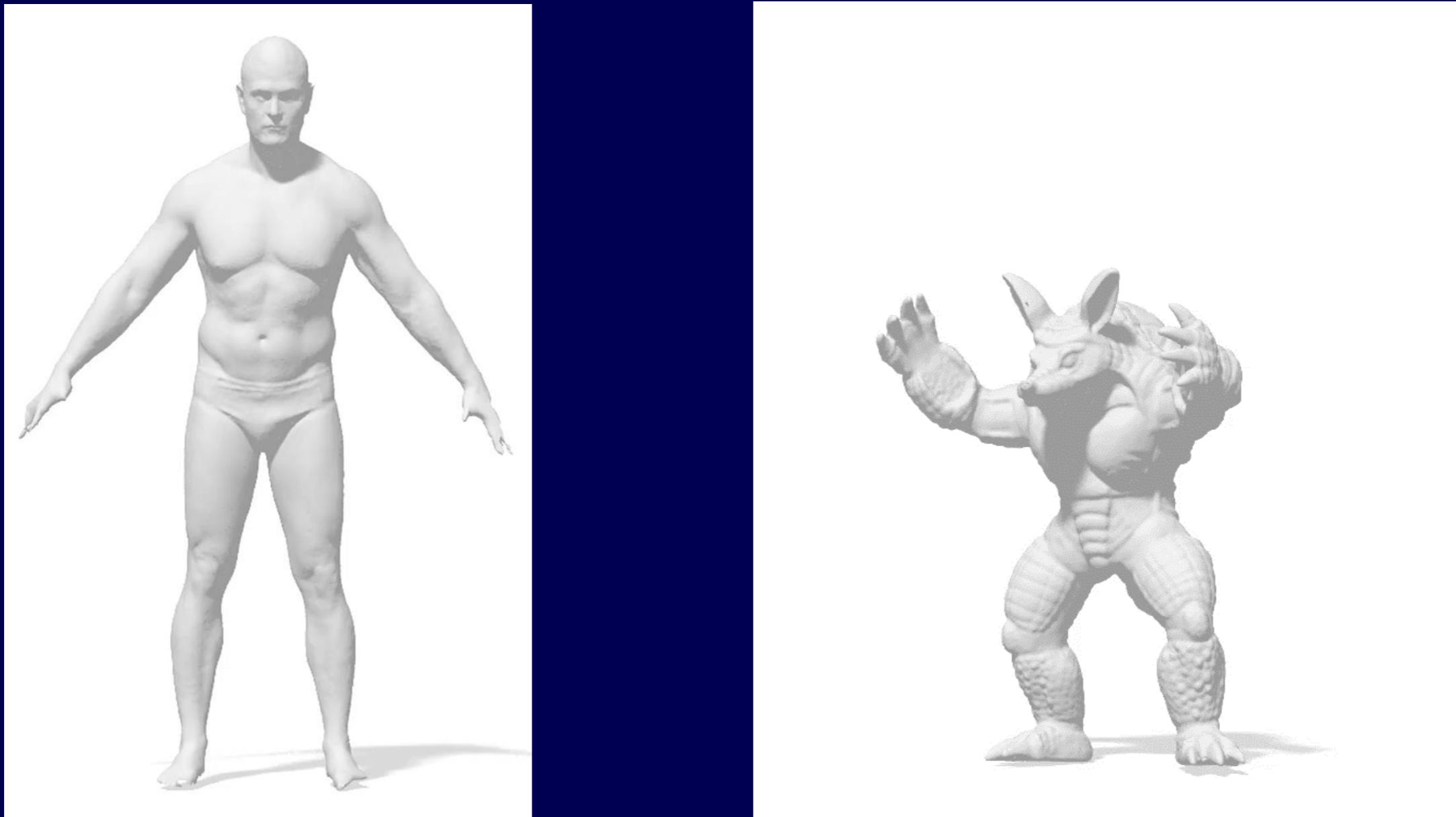
Sang et al., “4Deform: Neural Surface Deformation for Robust Shape Interpolation”, CVPR ‘25

Volume-Preserving Shape Interpolation



Eisenberger, Laehner, Cremers, SGP 2019

Volume-Preserving Shape Interpolation



Eisenberger, Laehner, Cremers, SGP 2019

4D Reconstruction from Sparse Observations



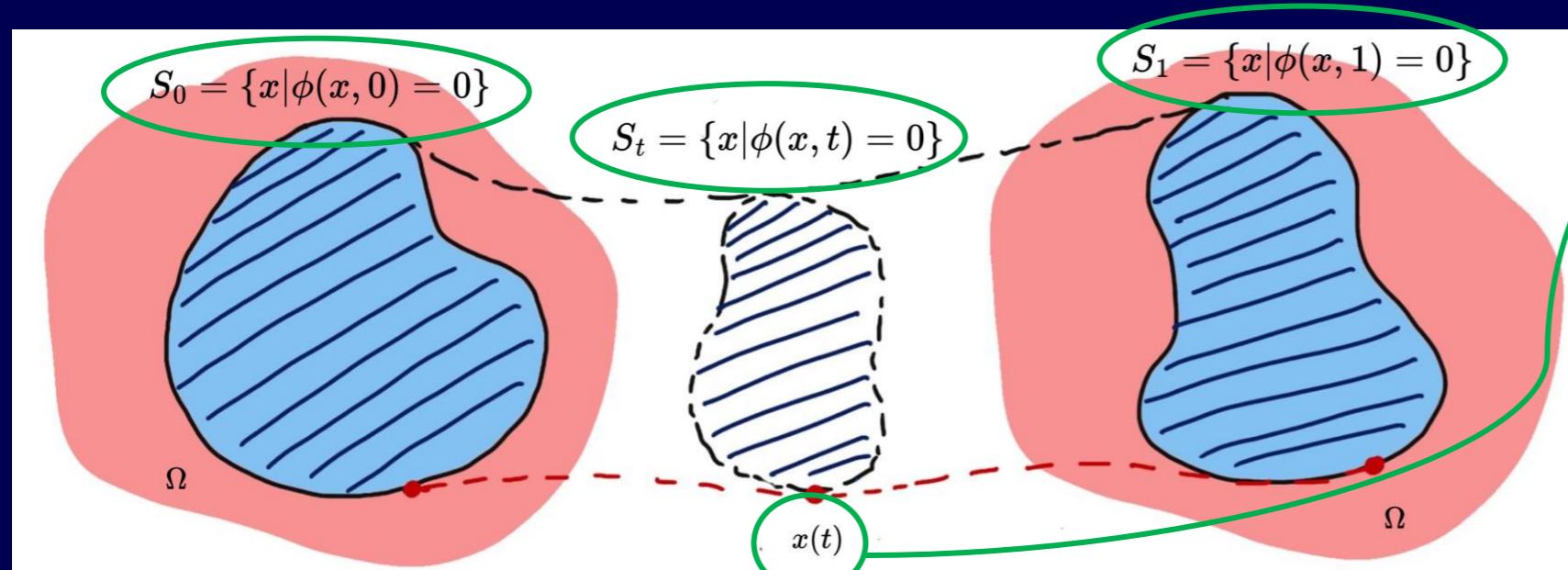
keyframe point clouds

keyframe kinect data

generated intermediate frames

Sang et al., “4Deform: Neural Surface Deformation for Robust Shape Interpolation”, CVPR ‘25

4D Reconstruction from Sparse Observations



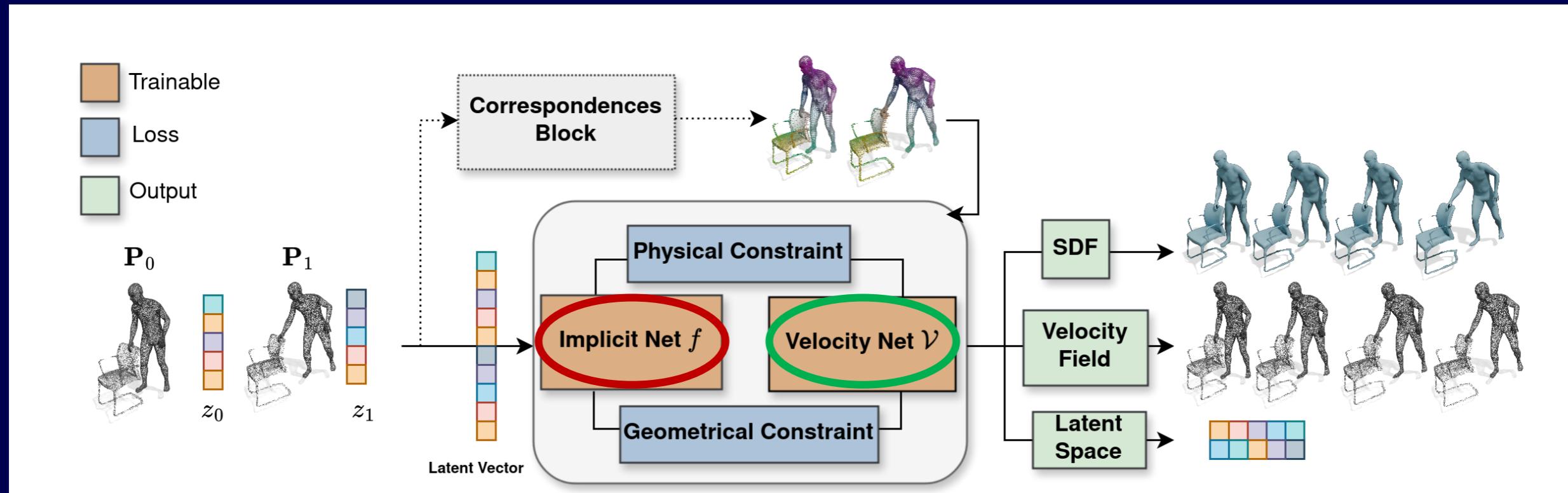
$$\phi(x(t), t) = 0$$

$$\frac{d}{dt} \phi(x, t) = \boxed{\partial_t \phi} + \boxed{V^\top} \boxed{\nabla \phi} = 0$$

$$\uparrow \quad \frac{d}{dt} x(t)$$

Sang et al., “4Deform: Neural Surface Deformation for Robust Shape Interpolation”, CVPR ‘25

4D Reconstruction from Sparse Observations



Geometrical Constraints

- Normal deformation constraint
- Level set equation constraint
- Matching loss

Physical Constraints

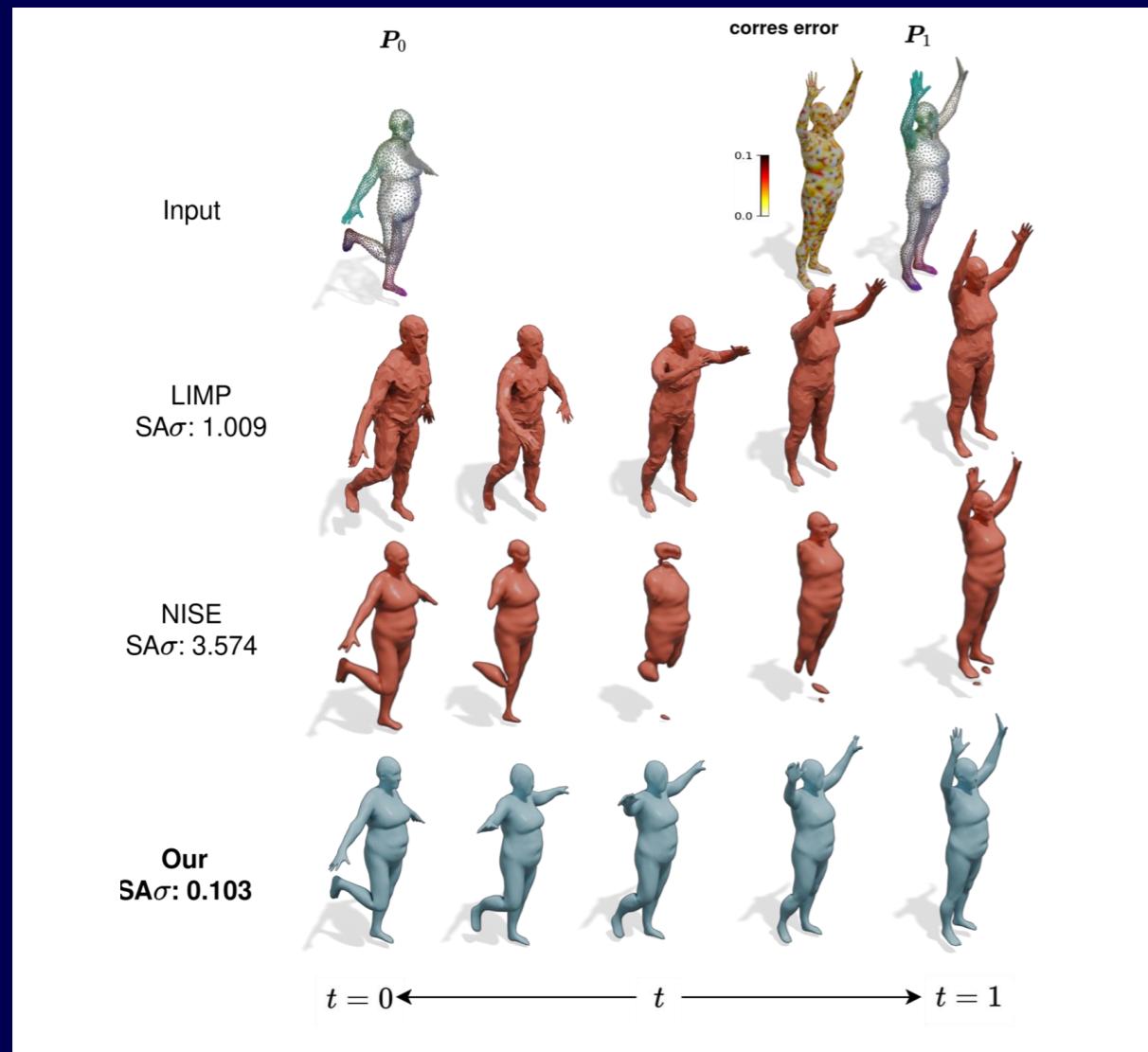
- Spatial smoothness velocity
- Volume preserving deformation
- Stretching constraint
- Distortion constraint

Sang et al., “4Deform: Neural Surface Deformation for Robust Shape Interpolation”, CVPR ‘25

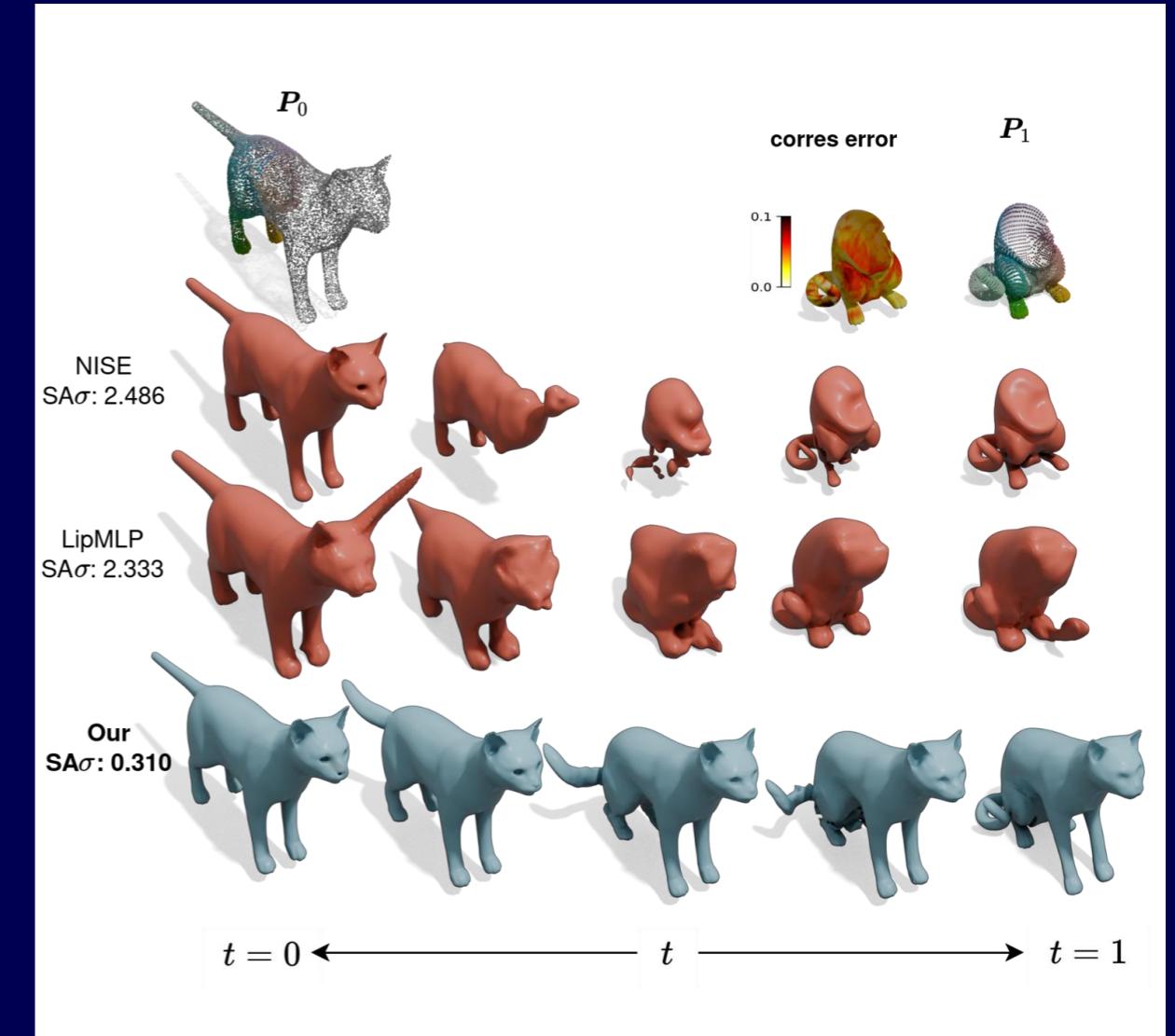
4D Reconstruction from Sparse Observations



Large deformation



Partial shape deformation



Sang et al., “4Deform: Neural Surface Deformation for Robust Shape Interpolation”, CVPR ‘25

2D input images:



Sang et al., “TwoSquared: 4D Generation from 2D Image Pairs”, arxiv ‘25

2D input images:



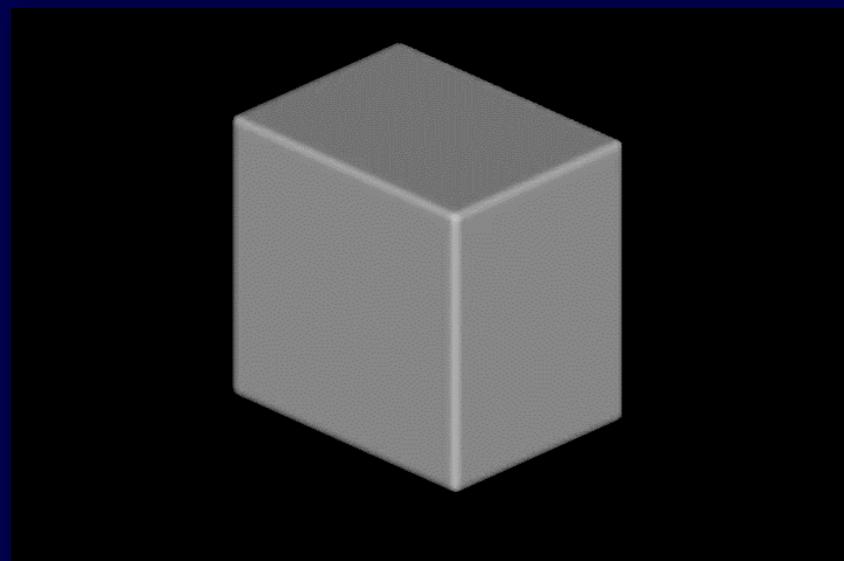
Sang et al., “TwoSquared: 4D Generation from 2D Image Pairs”, arxiv ‘25



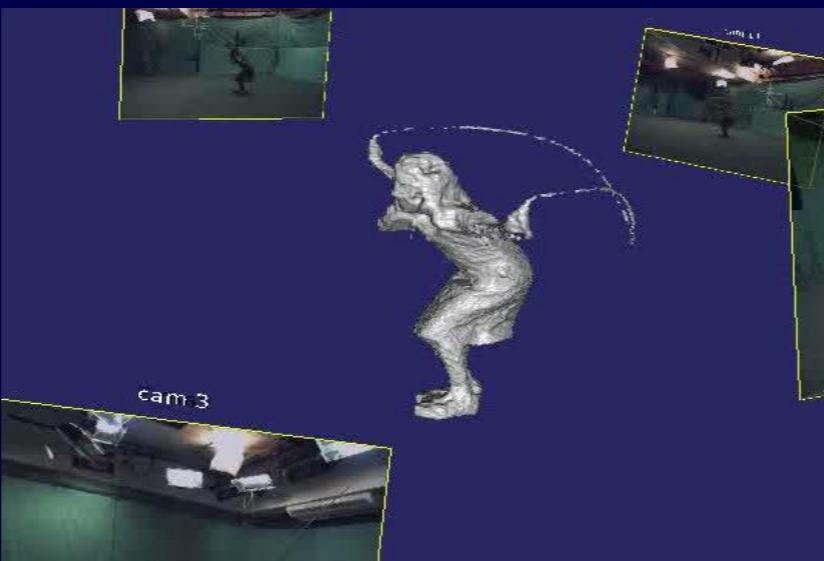
Sang et al., “TwoSquared: 4D Generation from 2D Image Pairs”, arxiv ‘25



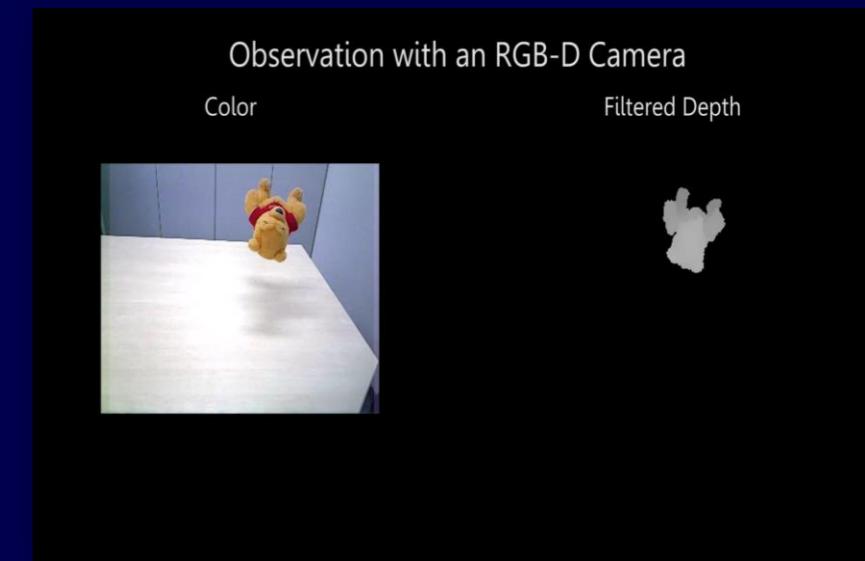
Summary



Convex multi-view 3D reconstruction



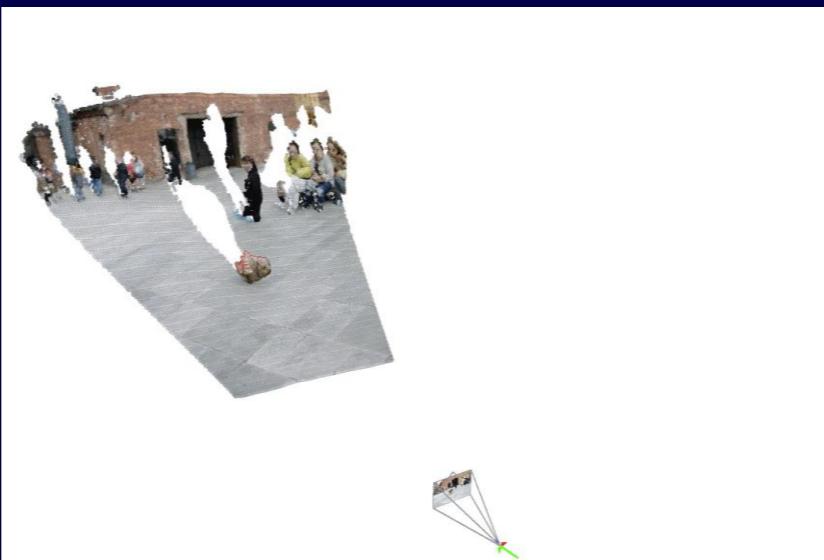
Convex multi-view 4D Reconstruction



Physical simulations from video



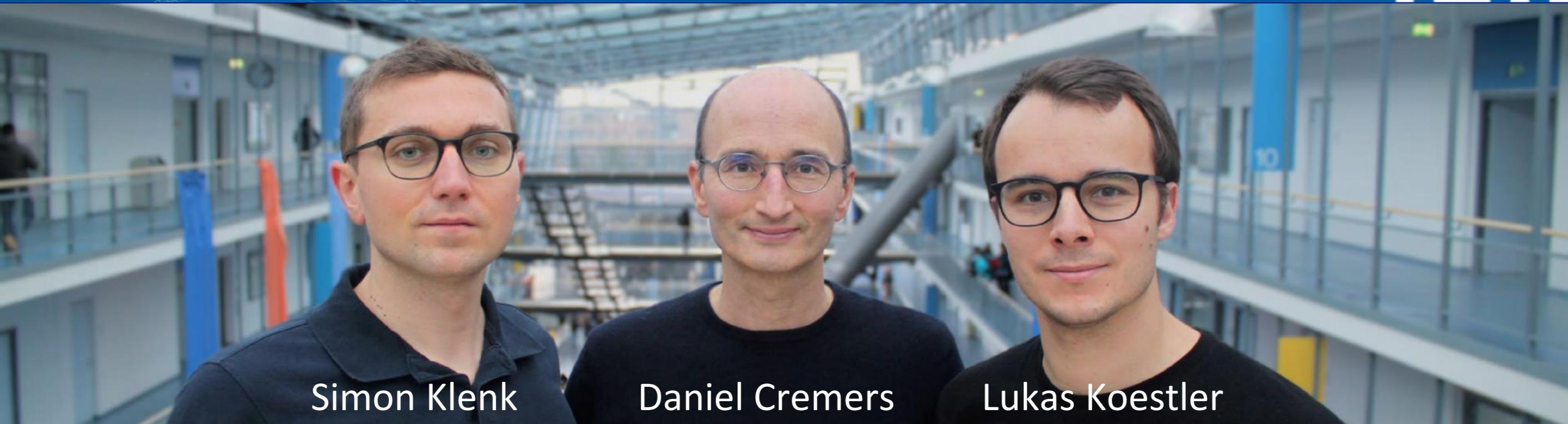
MonoRec for dense reconstruction



AnyCam for dynamic reconstruction



4Deform & TwoSquared



Simon Klenk

Daniel Cremers

Lukas Koestler

Interested in joining
as PhD intern or full time?

