



THE UNIVERSITY
of EDINBURGH



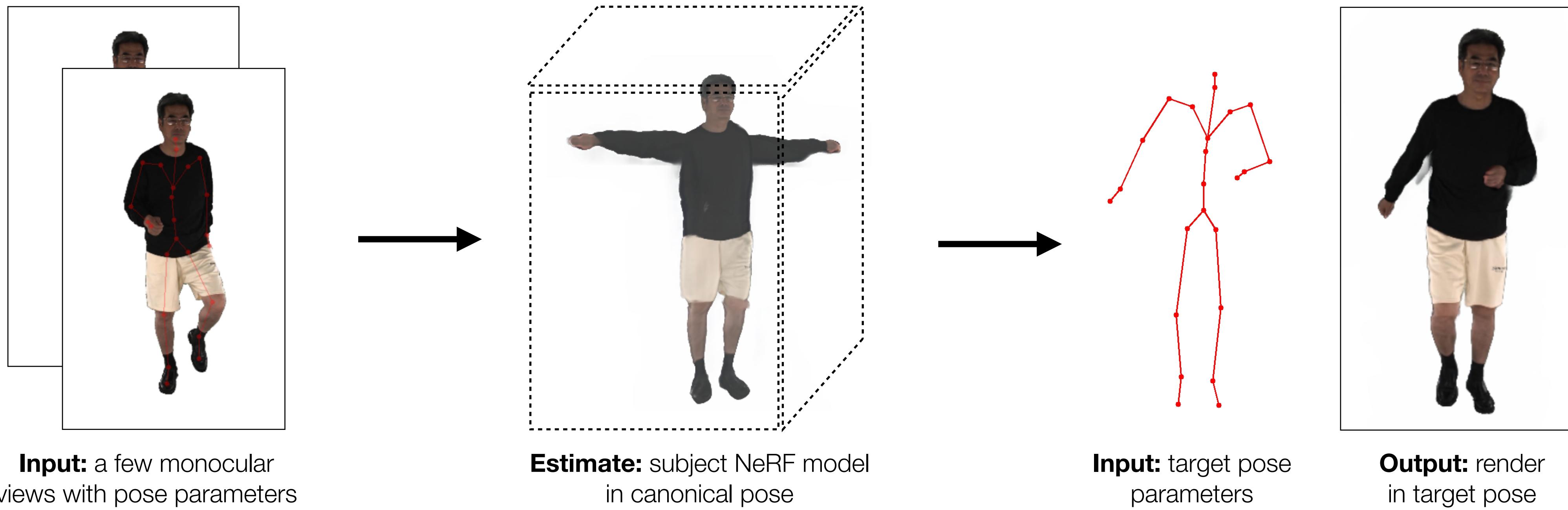
EDINBURGH CENTRE FOR
ROBOTICS

Generalized Animatable Human NeRFs from Few Views

Jakub Zadrozny, Hakan Bilen
University of Edinburgh

BMVA Symposium on 3D Vision
London | 26.02.2025

Dynamic free-viewpoint human rendering



Key applications include: augmented/virtual reality, movie production and immersive 3D communication.

Subject-specific vs. generalized



Subject-specific approaches

- Requires test-time optimization
- Needs extensive observations (typically ca. 30 frames)

+

Usually most accurate

Does not require large training set

Weng et al., *HumanNeRF*, CVPR 2022

Yu et al., *MonoHuman*, CVPR 2023

Hu et al., *GauHuman*, CVPR 2024

Generalized approaches (ours)



Uses only feed-forward passes during inference

Requires less observed views (1-3 for our model)

Learns a prior, in-paints unobserved details



Typically less accurate compared to subject-specific

Hu et al., *SHERF*, ICCV 2023

Li et al., *GHuNeRF*, 3DV 2024

Masuda et al., *GNH*, arxiv 2024

Ours

Our contributions

Limitations of existing work

conditioning only on a single observation or on an entire video



relying heavily on accurate body pose and parameters



struggling to model loose clothing



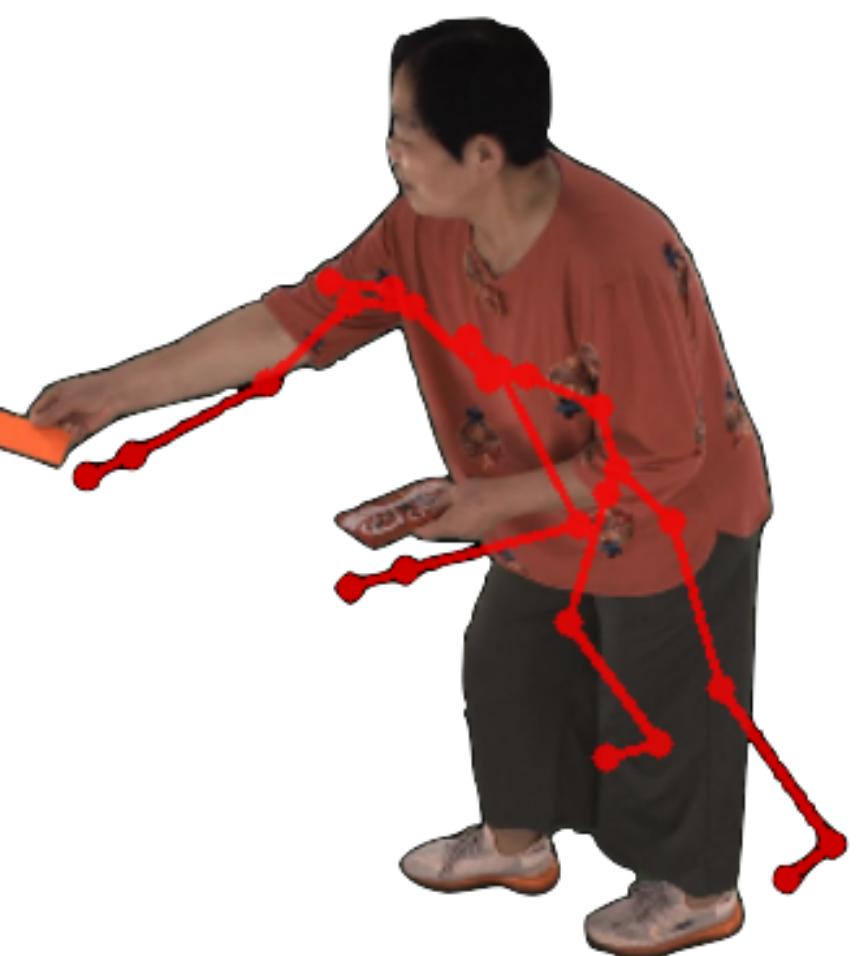
Our approach

improves rendering quality when given additional views



Estimated pose parameters

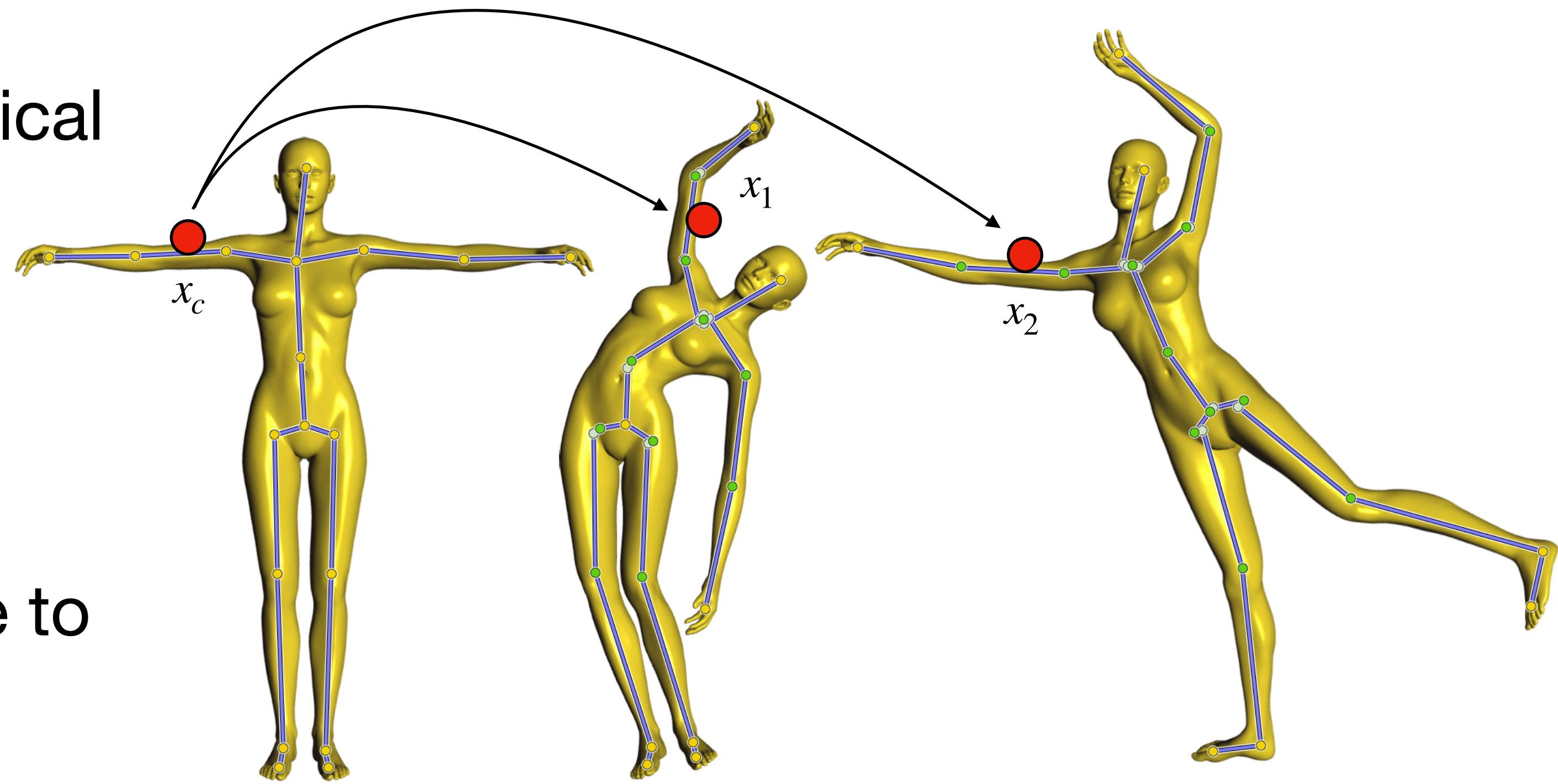
improves robustness to inaccuracies in body parameters



still a limitation, despite some improvements in architecture

Linear blend skinning

- Transforms points from the canonical pose to any given target pose
- Generally pre-defined*
- **Input:** motion/blend weights
 - softly assign points in 3D space to bones
 - *can be learned
- Can be approximately inverted



Input: skeleton shape in
canonical pose

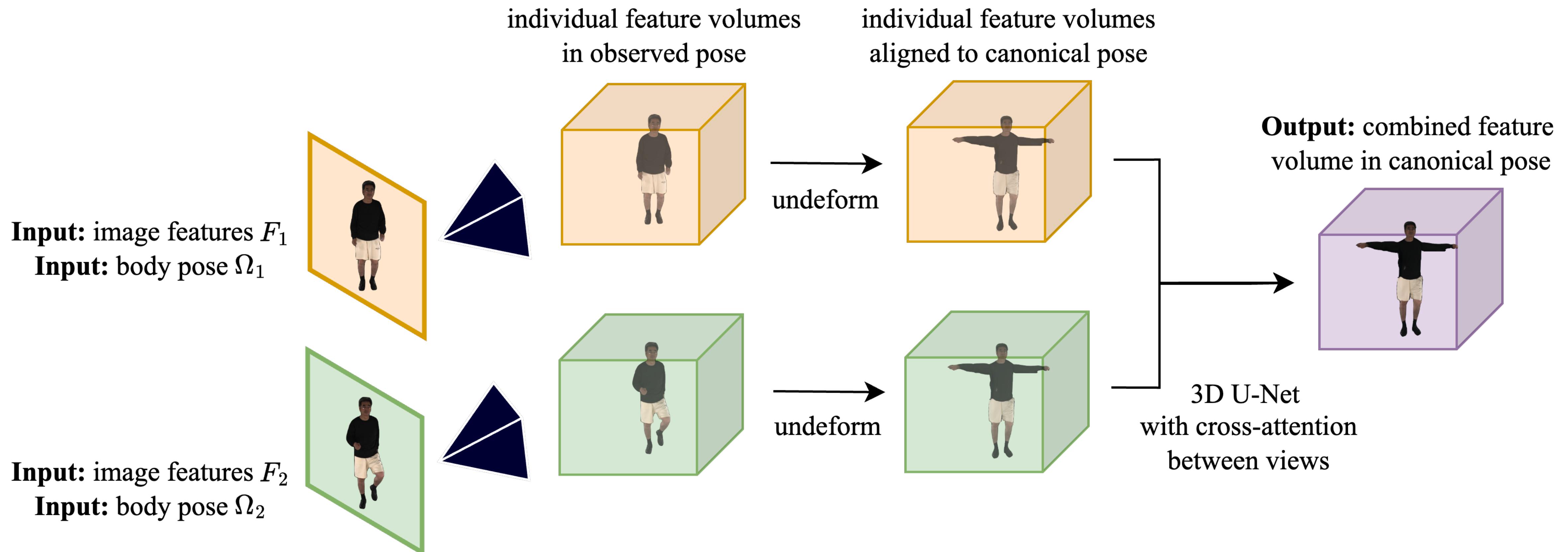
Input: joint rotations θ_1

Input: joint rotations θ_2

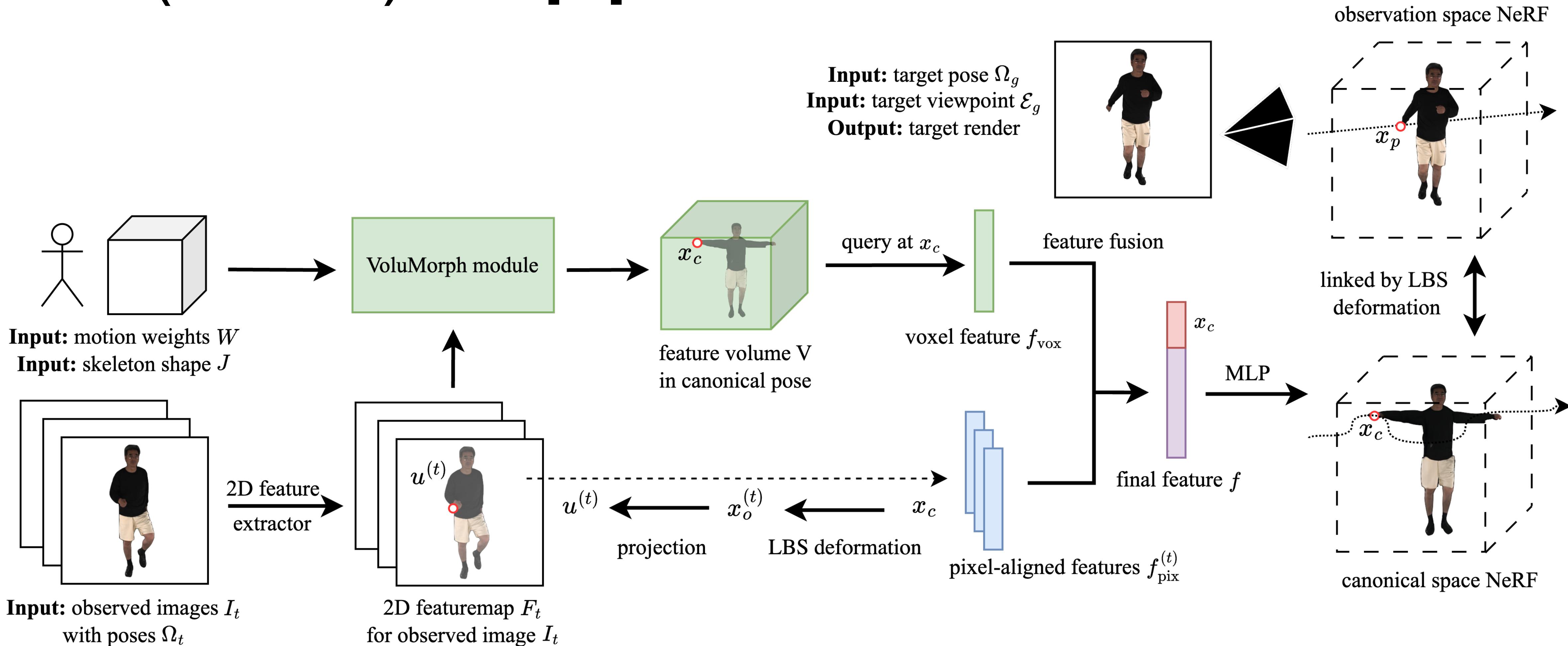
$$LBS(x_c, \theta_i) = x_i$$

$$LBS^{-1}(x_i, \theta_i) \approx x_c$$

The VoluMorph module



Our (almost) full pipeline



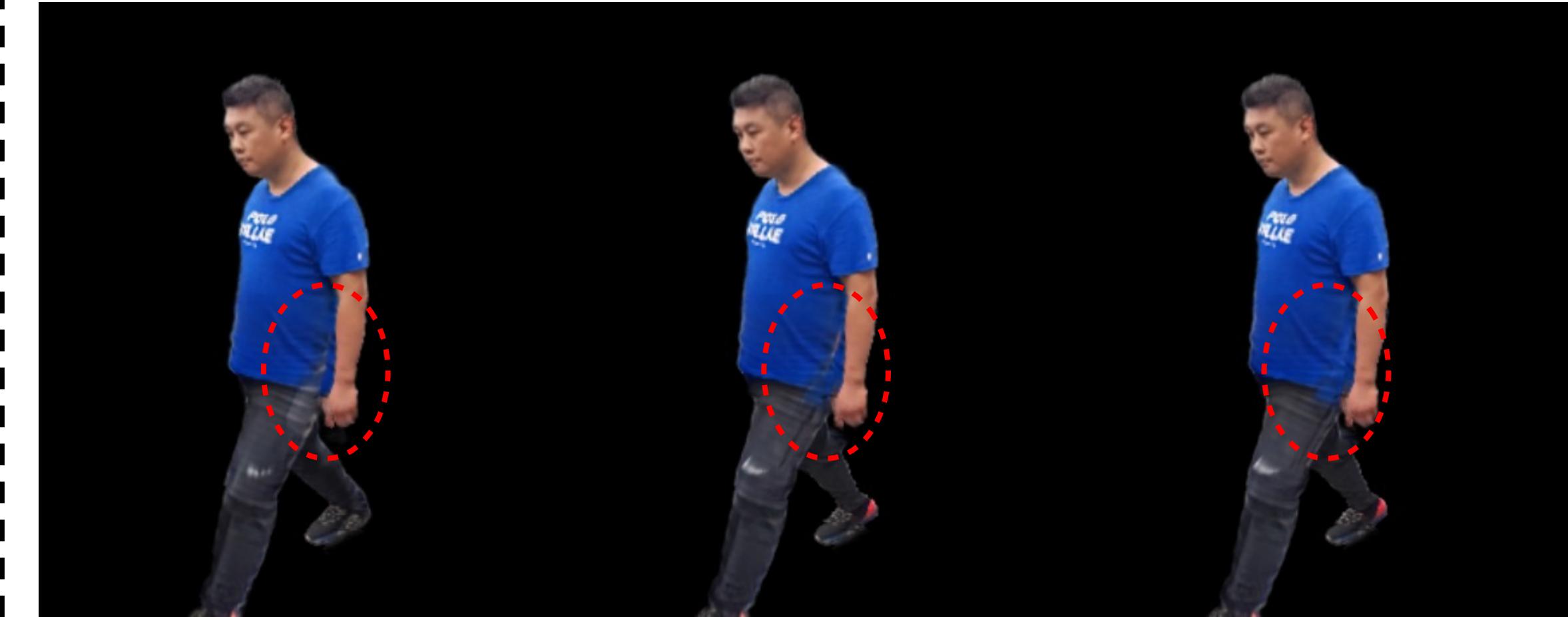
Qualitative results on HuMMan dataset

Observed 1



Observed 2

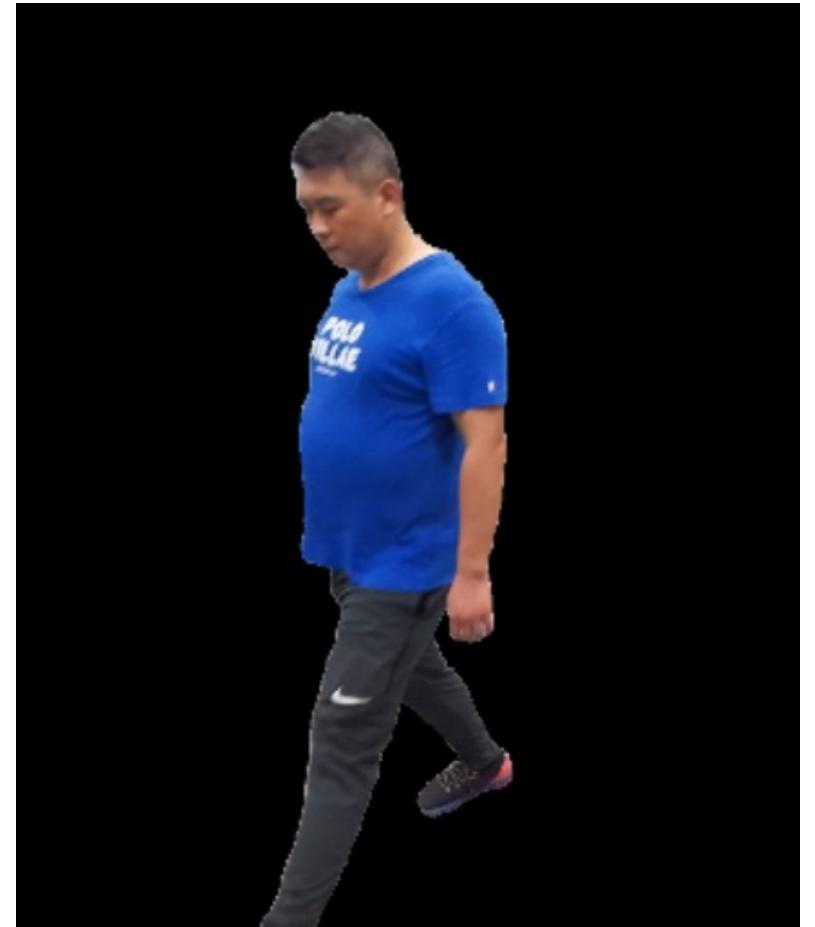
Ours (1st view)



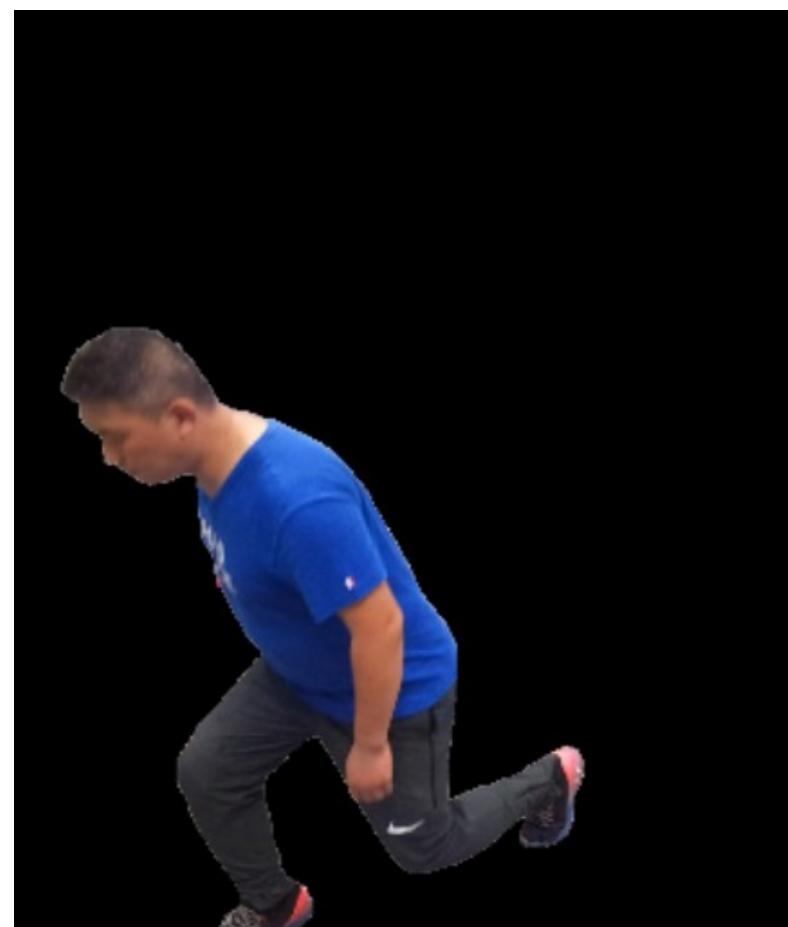
Ours (2 views)

Ours (3 views)

Ground truth



Observed 3



SHERF (1st view)



SHERF (2nd view)

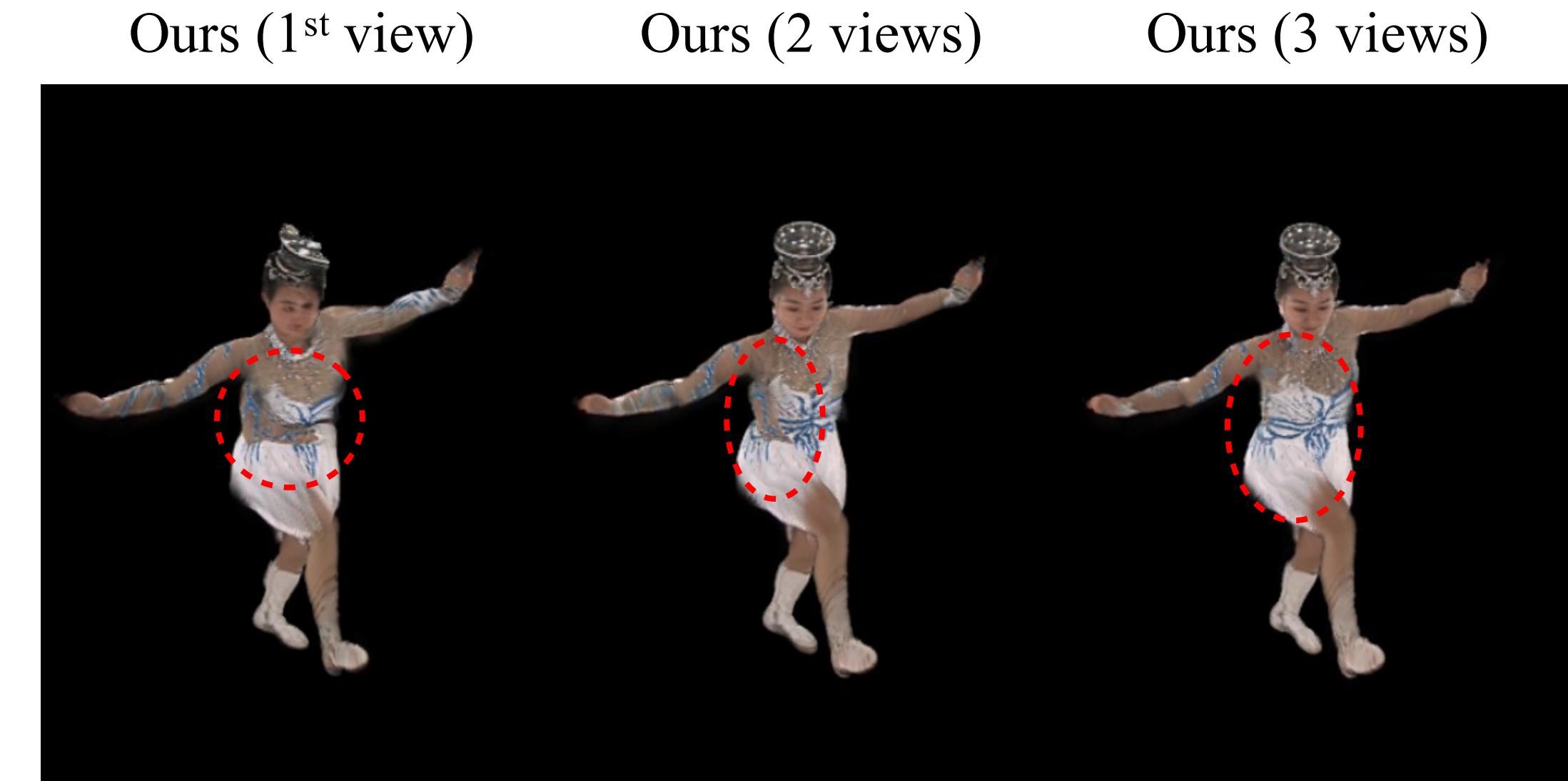
SHERF (3rd view)

Qualitative results on DNA-Rendering dataset

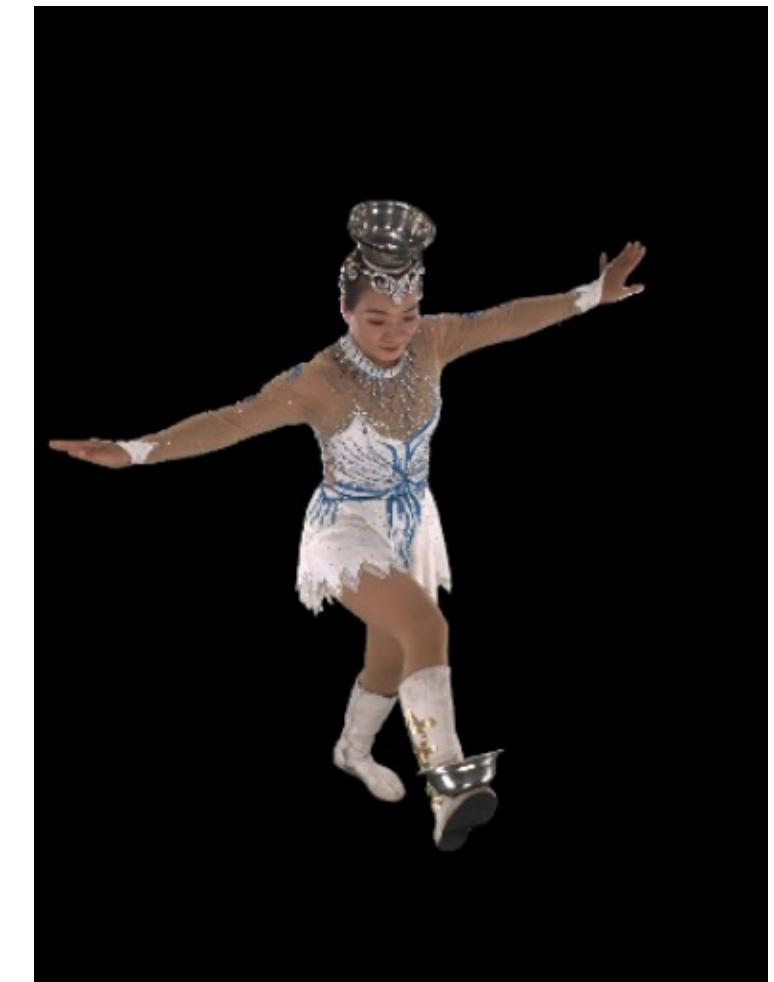
Observed 1



Observed 2



Ground truth



Observed 3

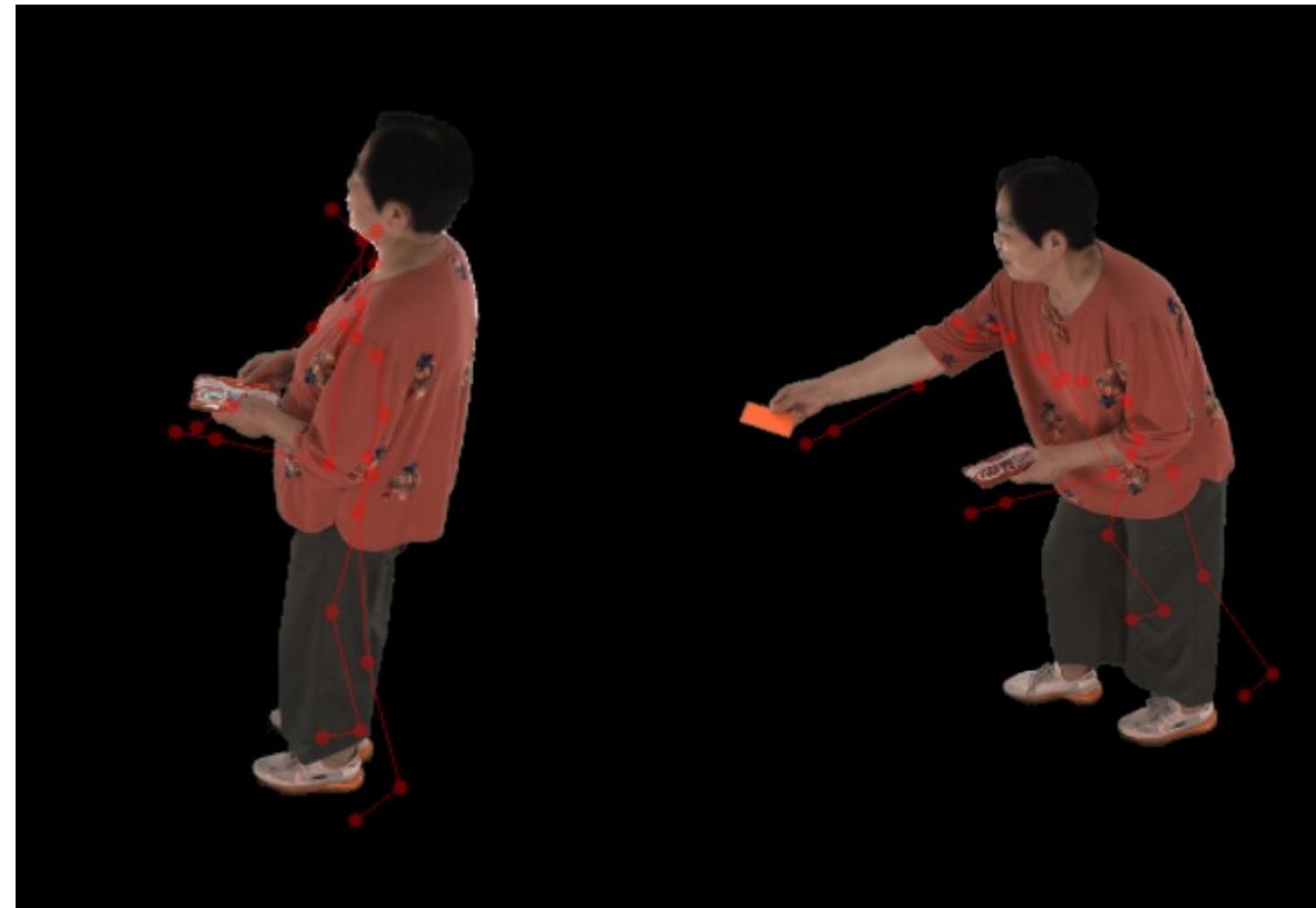


SHERF (1st view)



Qualitative results using estimated body parameters

Observed 1



Observed 2

Ours (1st view)



Ours (2 views)

Ours (3 views)

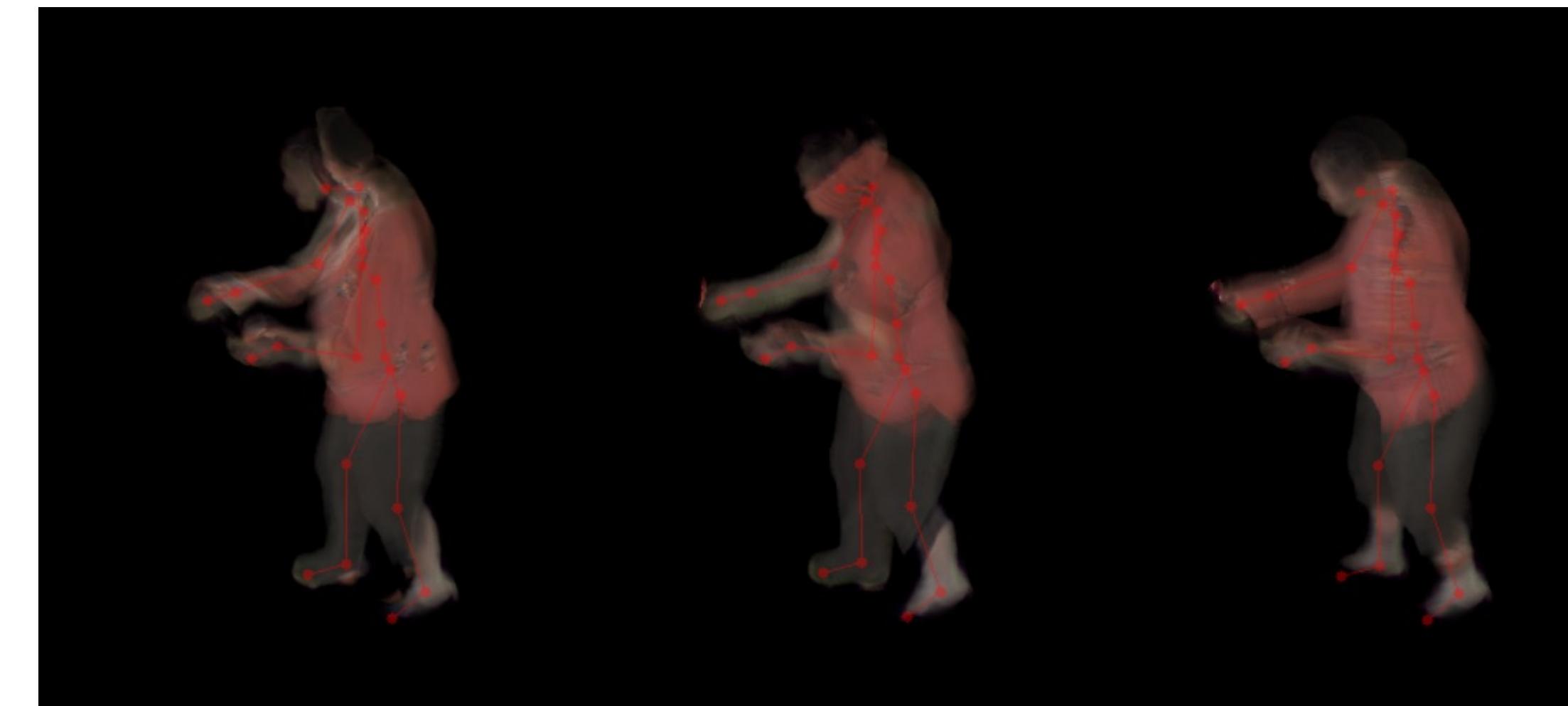
Ground truth



Observed 3



SHERF (1st view)



SHERF (2nd view)

Quantitative results

Dataset	Motion seqs. (train – test)	Actors	Cameras
HuMMan	339 (317 – 22)	153	10
DNA-Rendering	436 (372 – 64)	136	48

Method	HuMMan		DNA-Rendering			
	<i>accurate body params</i>		<i>accurate body params</i>		<i>estimated body params</i>	
	PSNR↑	LPIPS*↓	PSNR↑	LPIPS*↓	PSNR↑	LPIPS*↓
Ours (1 view)	26.62	34.24	27.83	40.27	27.43	44.12
Ours (2 views)	27.31	30.90	28.31	37.80	26.87	48.26
Ours (3 views)	27.56	29.59	28.59	36.68	27.04	47.22
Ours (4 views)	27.56	29.53	28.58	36.82	27.06	47.10
SHERF	26.87	45.03	28.49	48.22	26.93	61.97
GHuNeRF	23.88	66.27	27.80	70.84	27.19	75.13

- We evaluate rendering in **novel poses**
- Improvement on the perceptual metric, close results on PSNR (note: PSNR favours oversmoothed results over slight misalignments)
- Accurate body parameters: quality improves from 1 observation to 3, then saturates
- Estimated body parameters: best quality achieved using 1 observation

Limitations and future work

- **Limitations:**
 - our approach still struggles with estimated pose parameters
 - we do not model loose clothing accurately
- **Future work:**
 - further improve robustness to estimated body parameters and investigate the use of estimated camera poses
 - address modelling loose clothing and interactions with objects
 - improve efficiency (Gaussian splatting?)
- **Thank you!**