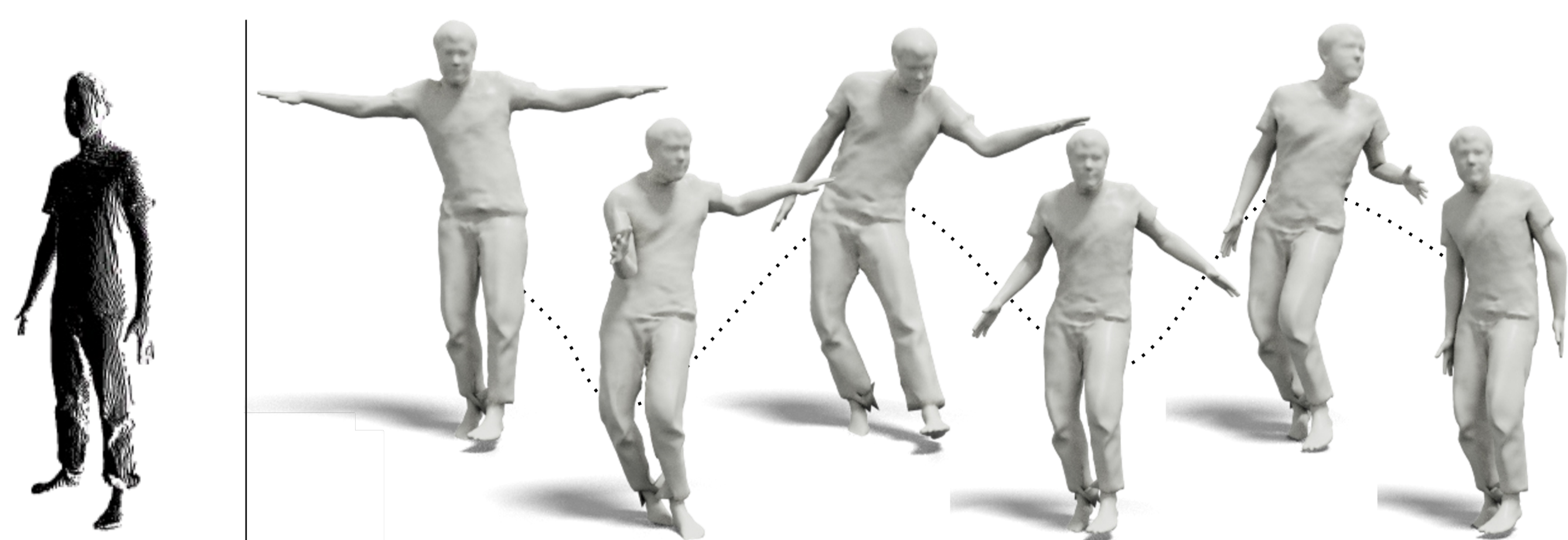


Task

Modelling clothed human from **monocular depth** sequences, can be used for shape completion, texture transfer and re-animation.



Motivation

Mesh Representation

- existing coherent connectivity
- not flexible in resolution & topology

Point Representation

- flexible
- per-frame Poisson Reconstruction
- incoherent mesh after reconstruction

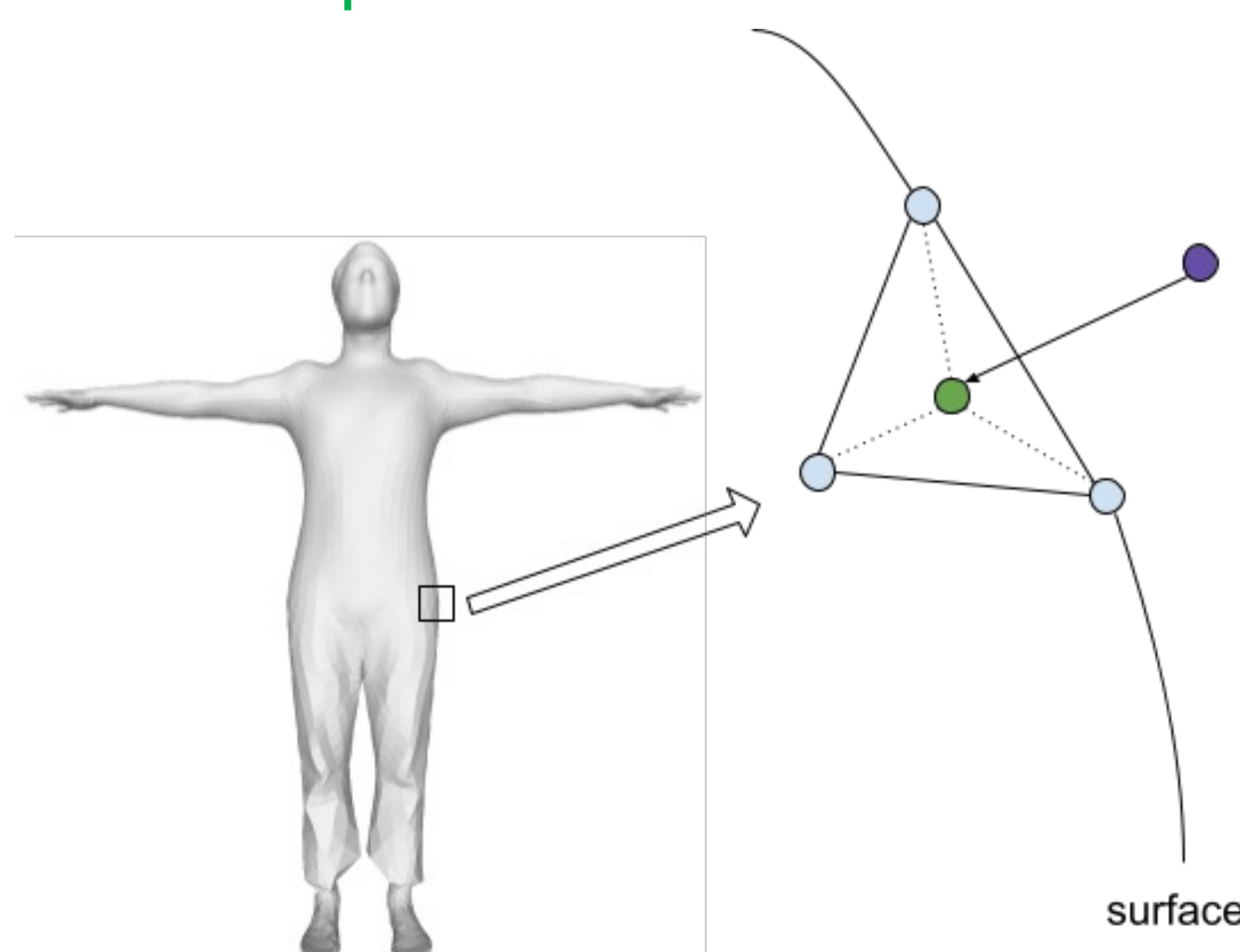
Neural Implicit Representation

- continuous
- per-frame Marching Cube
- incoherent mesh after extraction

Combine Neural Implicit and Mesh Representation

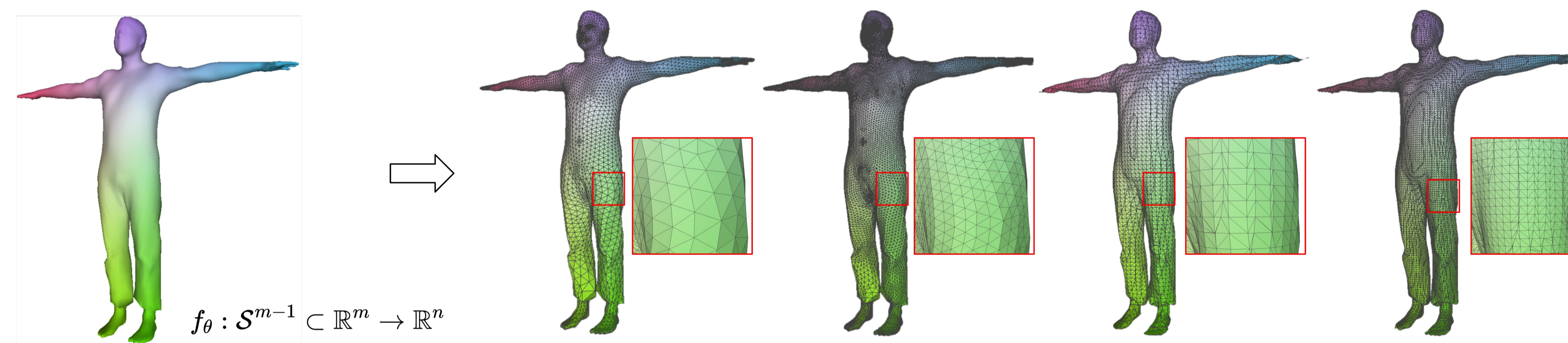
- continuous
- coherent
- no per-frame mesh extraction

Define a Neural Field on the surface!

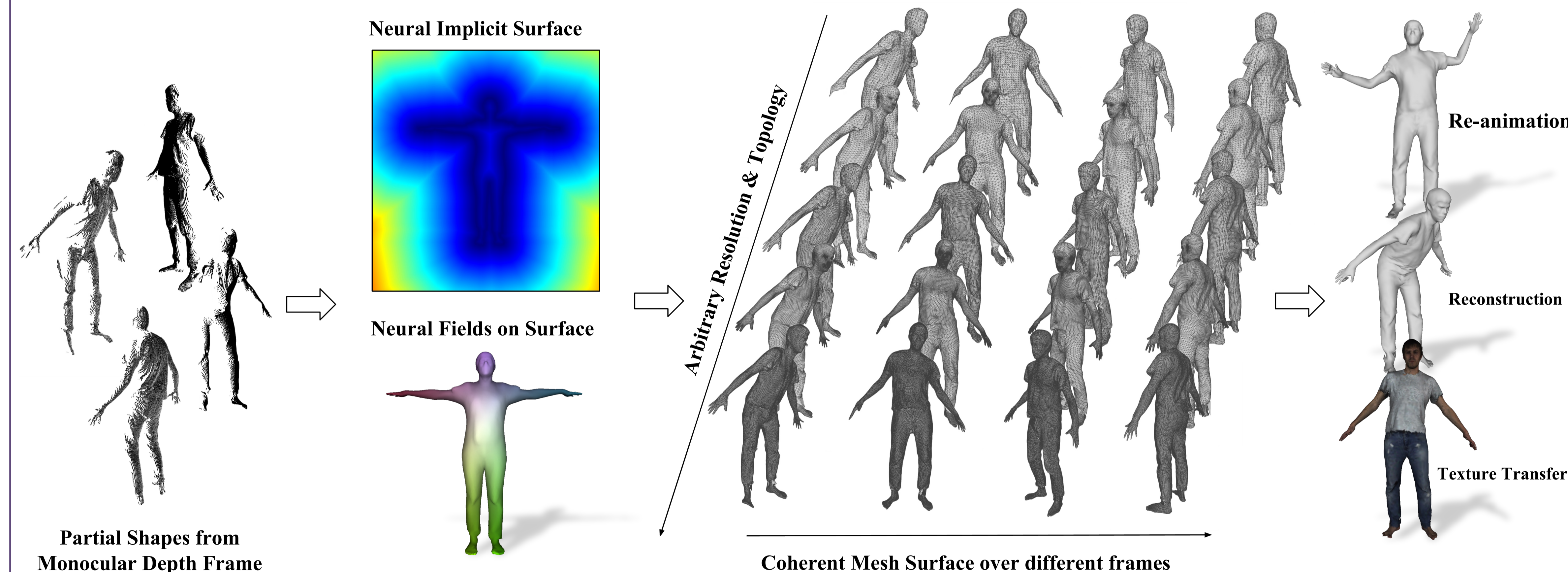


Neural Surface Fields (NSF)

We define a neural field solely on the surface, which utilizes the mesh surface **coherency** and **connectivity**. NSF generalizes to the surface in **arbitrary resolution** and **topology** without retraining.

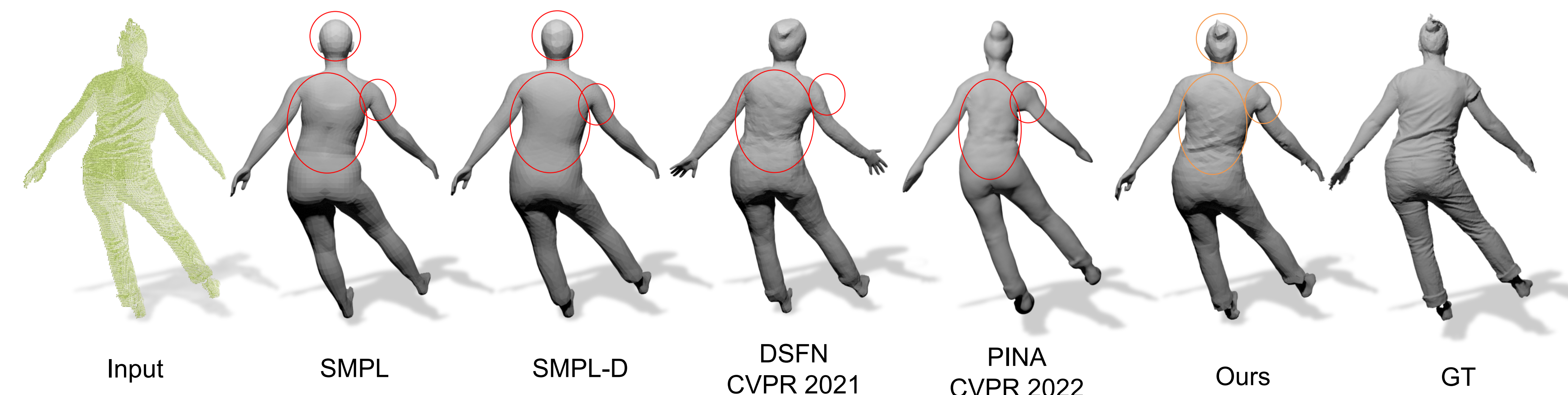


NSF for Human Modelling

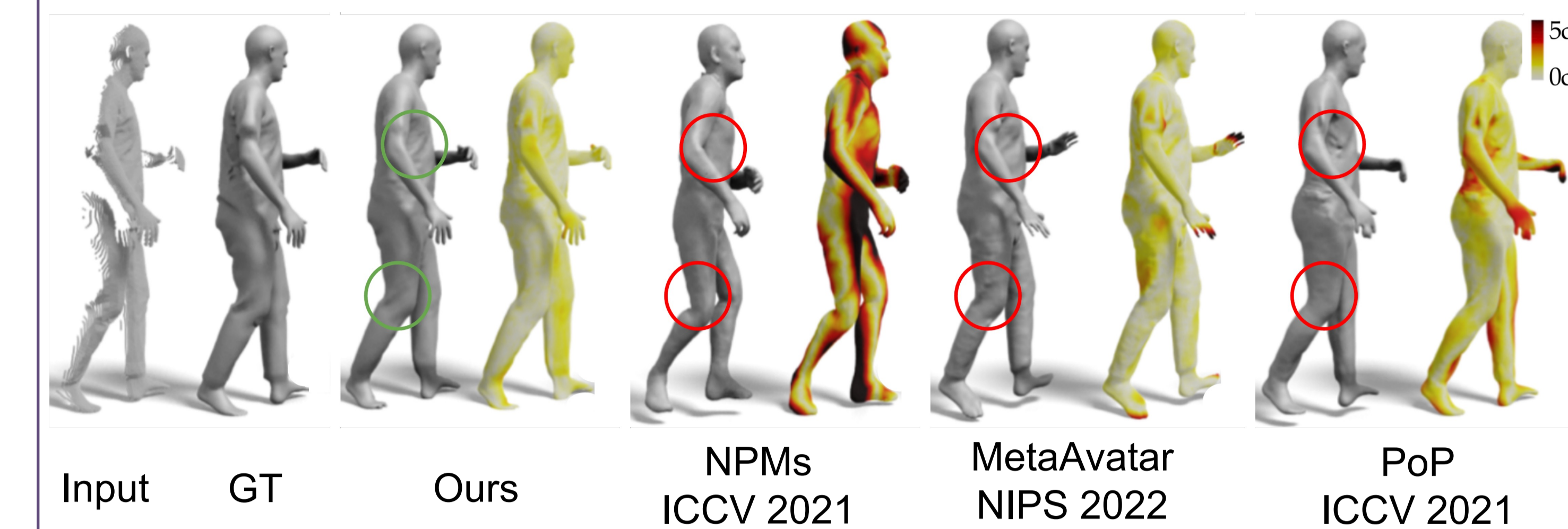


Experimental Results

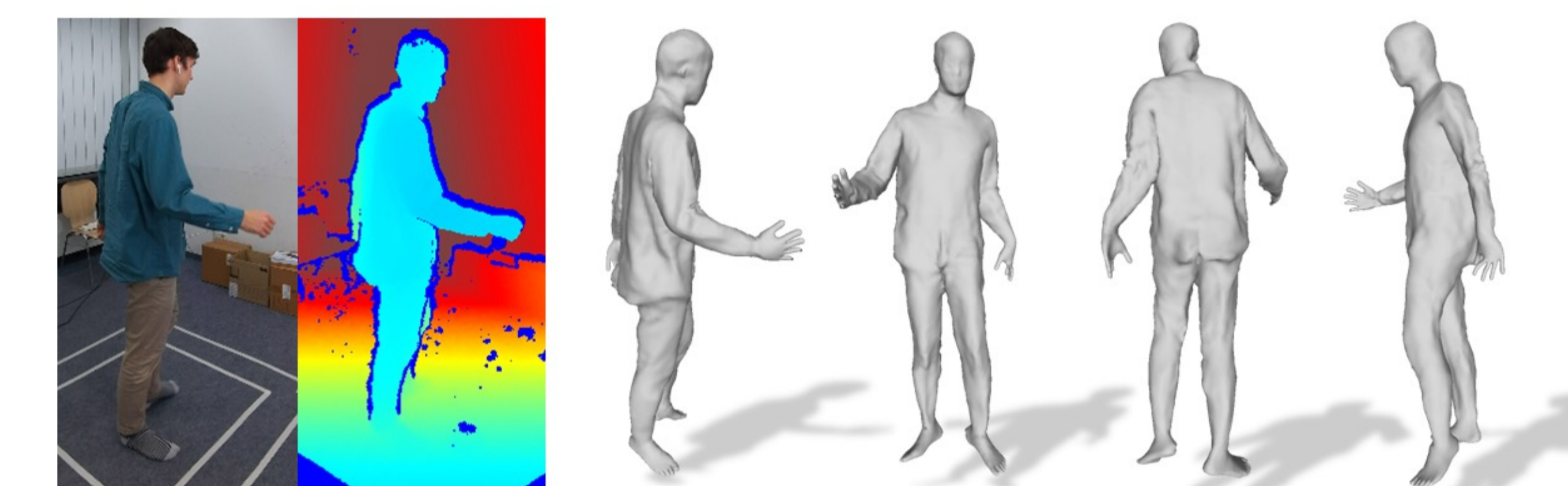
Results on BuFF dataset:



Results On CAPE dataset:



Results on DSFN dataset:



Key Takeaways

- **40x**: NSF is 40x faster than point-based representation, which needs per frame Poisson Reconstruction. (0.039s vs 1.6s)
- **180x**: NSF is 180x faster than implicit representation, which needs per frame Marching Cube surface extraction. (0.039s vs 7s)
- **97.4%**: 97.4% memory compared to volume-based feature representation at resolution of 128³.
- **86.0%**: 86.0% memory compared to triplane-based feature representation at resolution of 128³.

Our NSF achieves state-of-the-art accuracy while being more **efficient** and **faster coherent** mesh surface reconstruction!

Summary

- NSF, a novel hybrid surface representation which is
 - efficient, no surface extraction per frame
 - flexible, to arbitrary resolution & topology
 - coherent, across all frames

Code and results available: yuxuan-xue.com/nsf

