

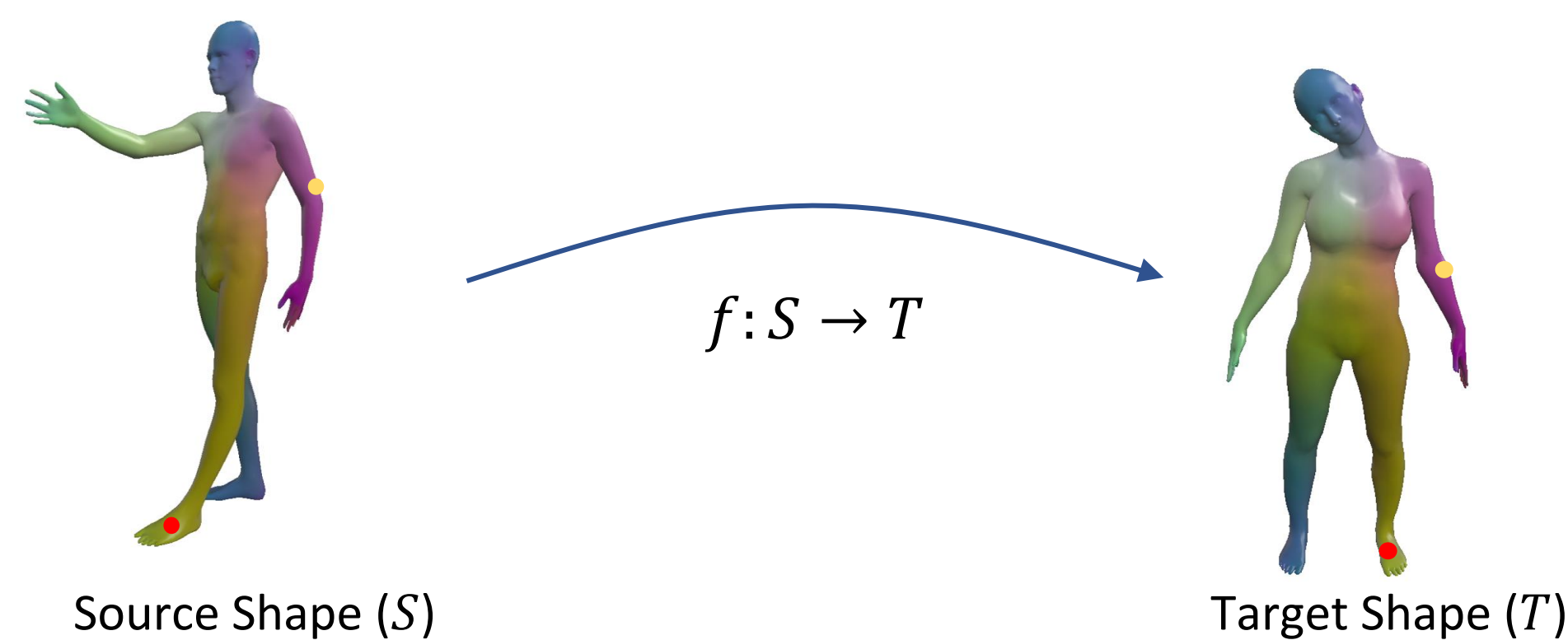
# Dense Correspondence between Human Bodies via Learning Transformation Synchronization on Graphs

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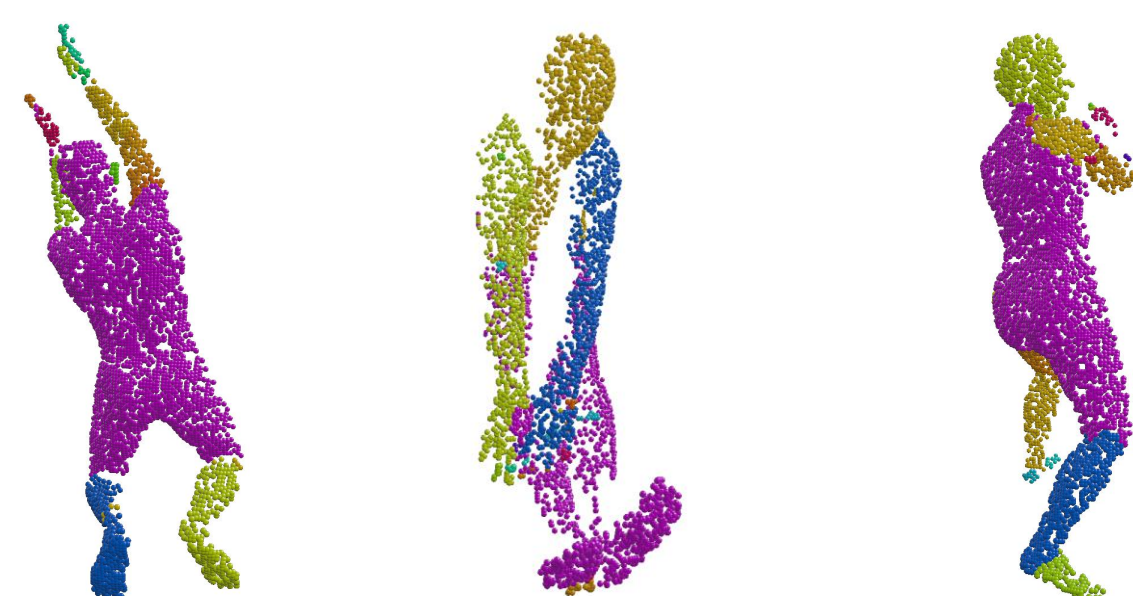
## Dense Correspondence

Dense Correspondence aims at computing point-wise mapping function  $f$ , that maps any point on the source shape  $S$  to a point on the target shape  $T$  (points marked as red dots). We focus on point cloud inputs, each generated by a depth camera from a single view-point. This also generalizes to mesh inputs.



## Motivation – Local Rigidity

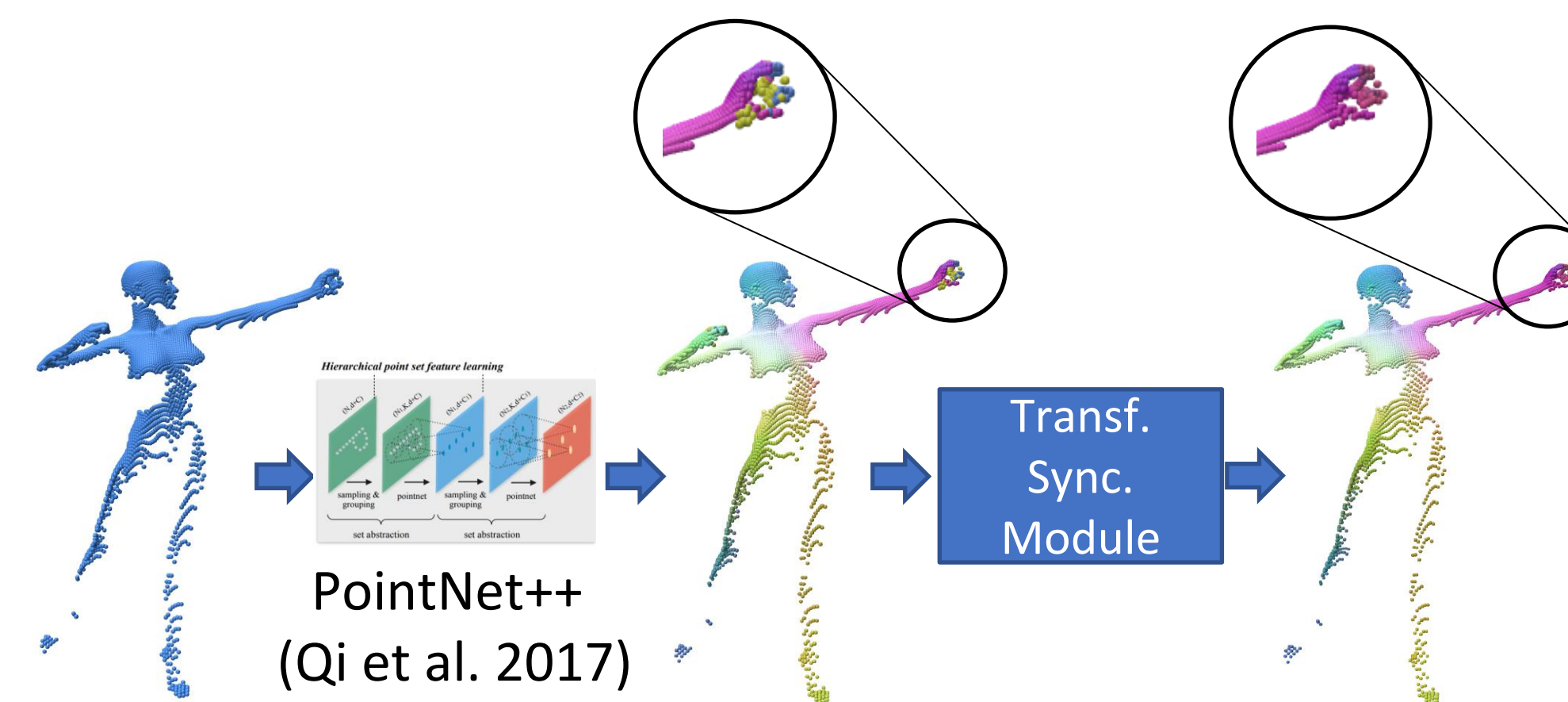
Our key assumption is that **human Bodies behaves, to good approximation, as an articulated rigid body.** This motivates us to rectifies incorrect correspondences via enforcing local rigidity. Specifically, we implement this enforcement procedure as a differentiable neural network module.



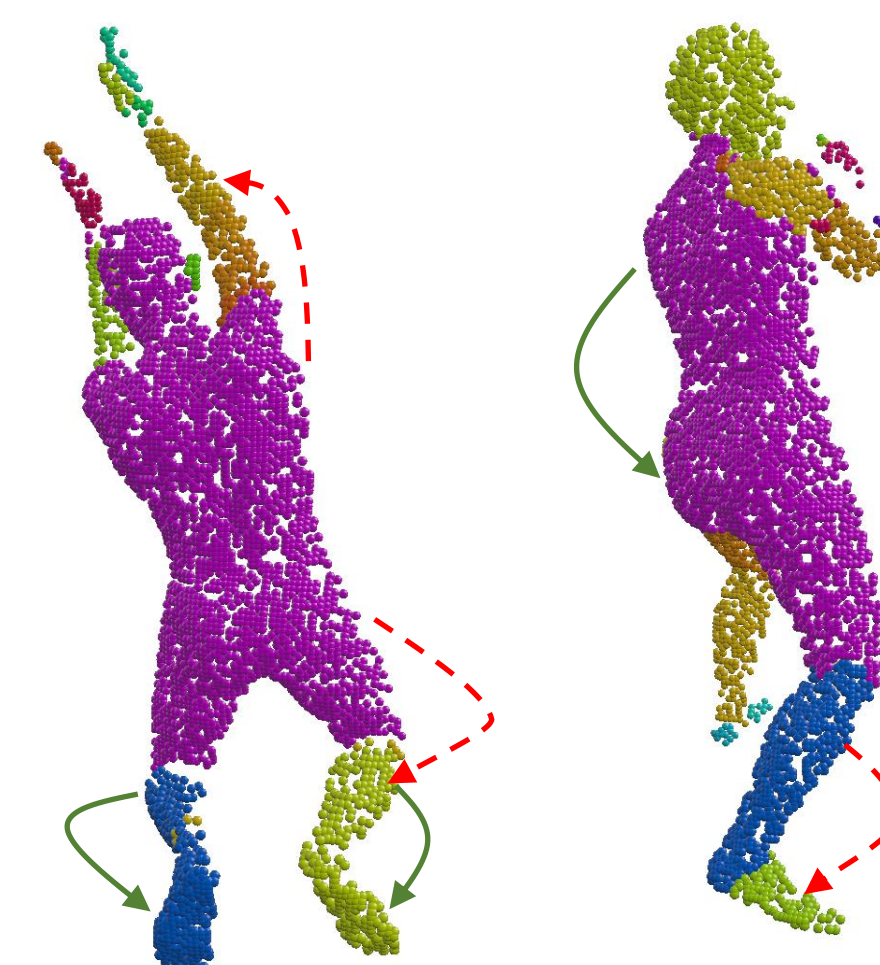
(Approximately) Rigid Body Parts

## Dense Correspondence Framework

Our network consists of two modules. The first module extracts point-wise feature vectors using the standard PointNet++ backbone. The second module (Transformation Synchronization Module) learns to enforce local rigidity of human body by synchronizing point-wise rigid transformation.



## Transformation Sync. Module



Good →  
Bad →

### Message Passing

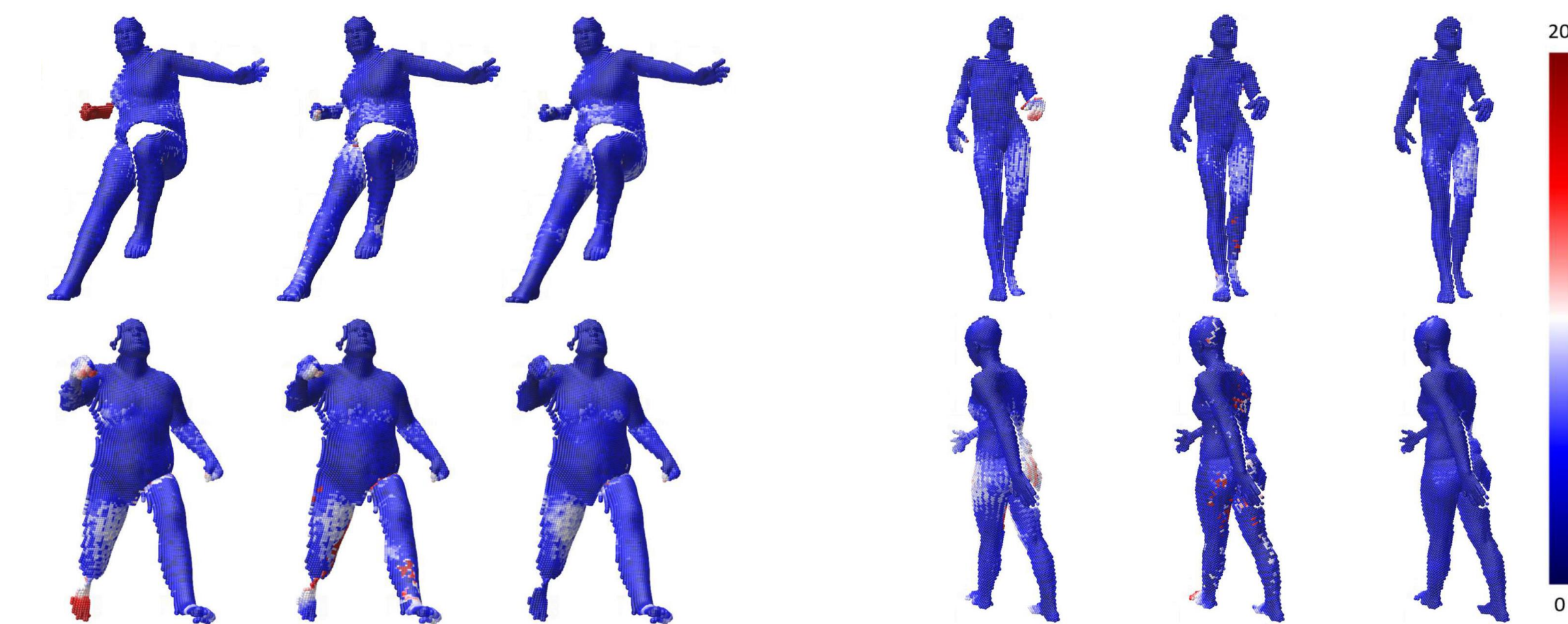
$$w_i = \text{msg\_passing}(d_{N_i})$$

$$R_i, t_i = \sum_{j \in N_i} w_j (R_j, t_j) / \sum_{j \in N_i} w_j$$

Given initial correspondences, we compute point-wise local rigid transformations to a template shape. Enforcing local rigidity is casted as propagating rigid transformation within rigid body parts, which is naturally implemented as a message passing module. The module learns the weighting parameters under an iterative reweighting scheme to guide the transformation propagation, i.e. by suppressing bad edges and encouraging good edges.

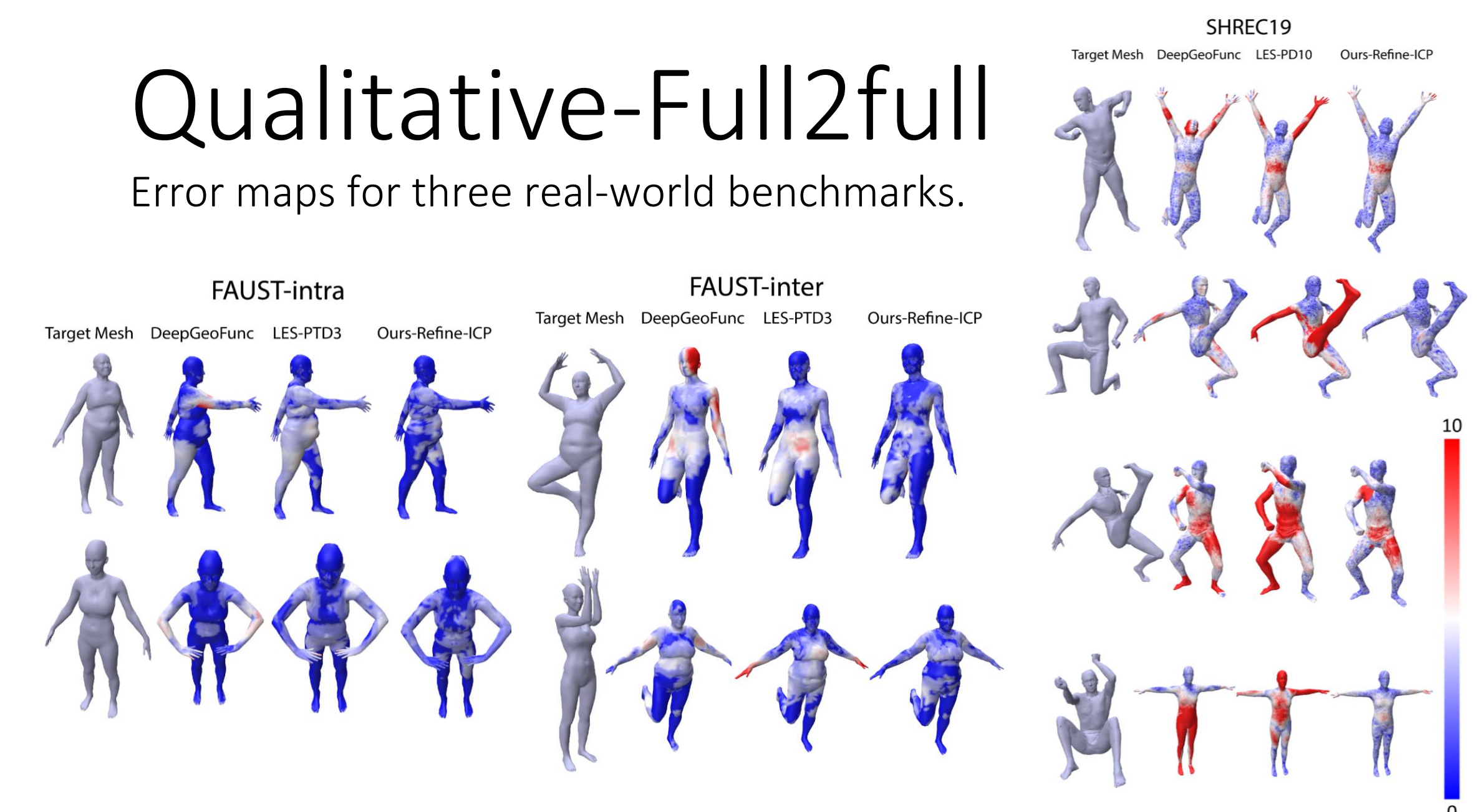
## Qualitative-Partial2Full.

Left to Right: Error maps of Baseline I, Baseline II, Ours approach



## Qualitative-Full2full

Error maps for three real-world benchmarks.



## Quantitative Results

Full-to-Full	Faust Inter-subject	Faust Intra-Subject	Shrec19
Baseline I	2.08cm	1.97cm	6.69cm
Baseline II	2.05cm	2.32cm	5.02cm
Our Approach	1.72cm	1.42cm	3.50cm

Partial-to-Full	Surreal (Synthetic)	Faust	Shrec19
Baseline I	2.02cm	1.98cm	5.48cm
Baseline II	1.93cm	1.95cm	6.15cm
Our Approach	1.71cm	1.90cm	4.81cm

GitHub

