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DEFORMATION TRANSFER OF 3D HUMAN SHAPES AND POSES ON MANIFOLDS

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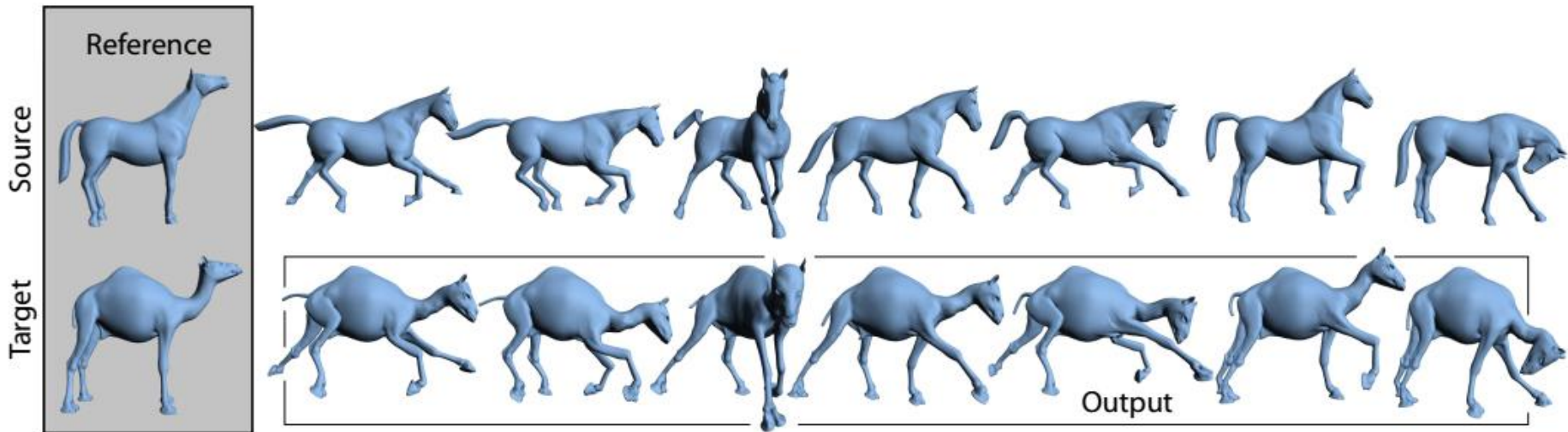
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Agenda

1. Introduction
2. Manifold Representation of 3D Human Shape
3. Deformation Transfer on Manifolds
4. Results
5. Conclusion & Future work

Introduction



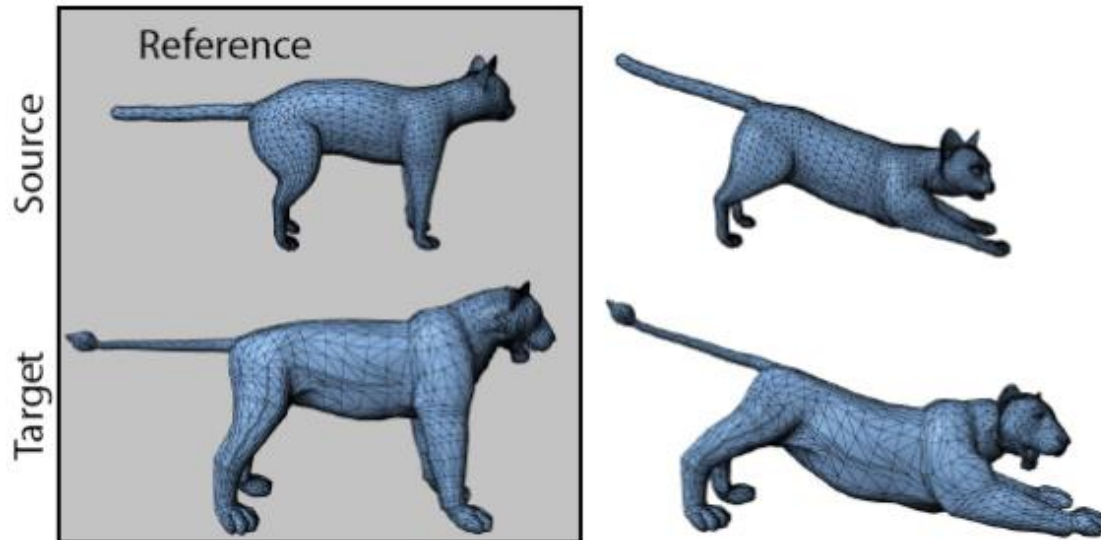
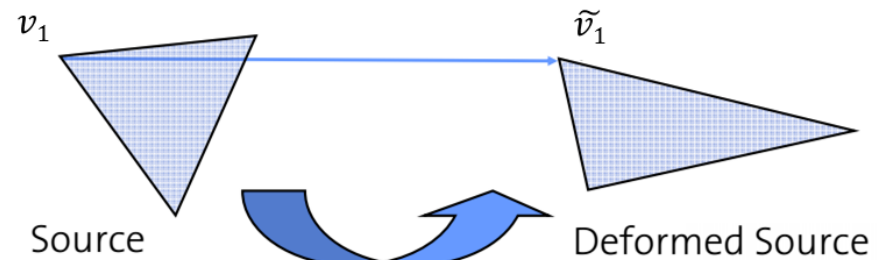
Deformation transfer copies the deformations exhibited by a source mesh onto a different target mesh.

Deformation Transfer for Triangle Meshes
Robert W. Sumner and Jovan Popovic
SIGGRAPH 2004.

Introduction: Euclidean Deformation Transfer for Triangle Meshes

- Deformation is based on a per-triangle affine transformation

$$Qv_i + d = \tilde{v}_i, \quad i \in 1 \dots 4$$



Introduction: Limitations of existing solutions

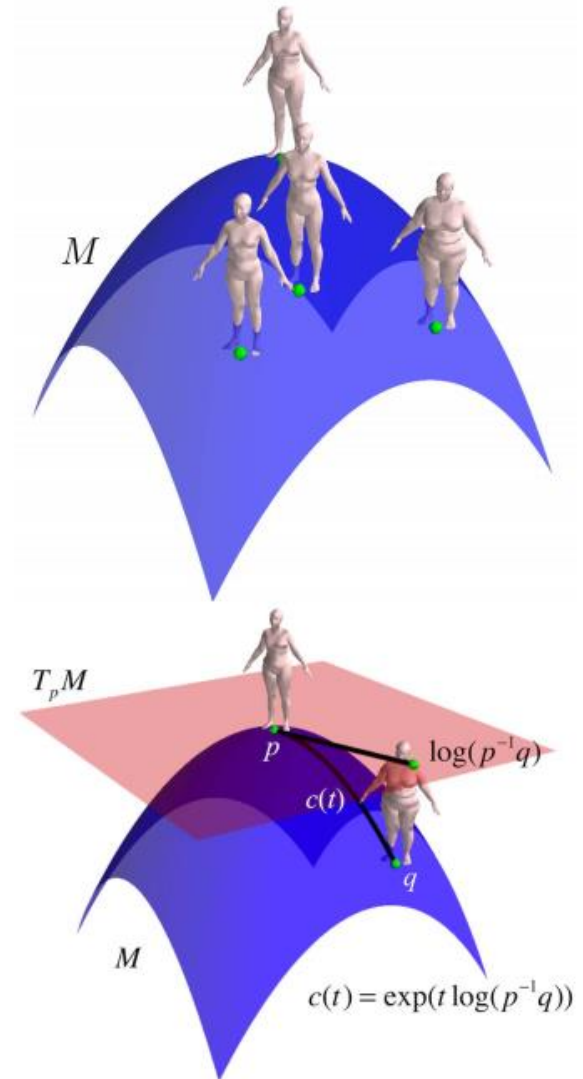
- Existing methods are based on **Euclidean representations**.
- The triangle deformation: **3x3** deformation matrix and a **3D** displacement vector.
- The **9D (redundant DoF)** of deformations is **under-constrained** as **deformations outside the plane** of the triangle are **undefined**.
 - A fourth virtual vertex defined by the cross product of two of the triangle edges is added **heuristically** .
- Deformations may have **zero** or **negative** determinant (inconsistent deformations)
 - Thus **do not exclude non-physical deformations**.

Introduction: How to solve the current problems?

- We are seeking:
 - An accurate representation which **eliminates redundant DoF**.
 - Deformations to be computed in **closed-form without heuristics**.
 - **Consistent** deformations to **eliminate non-physical** deformations.

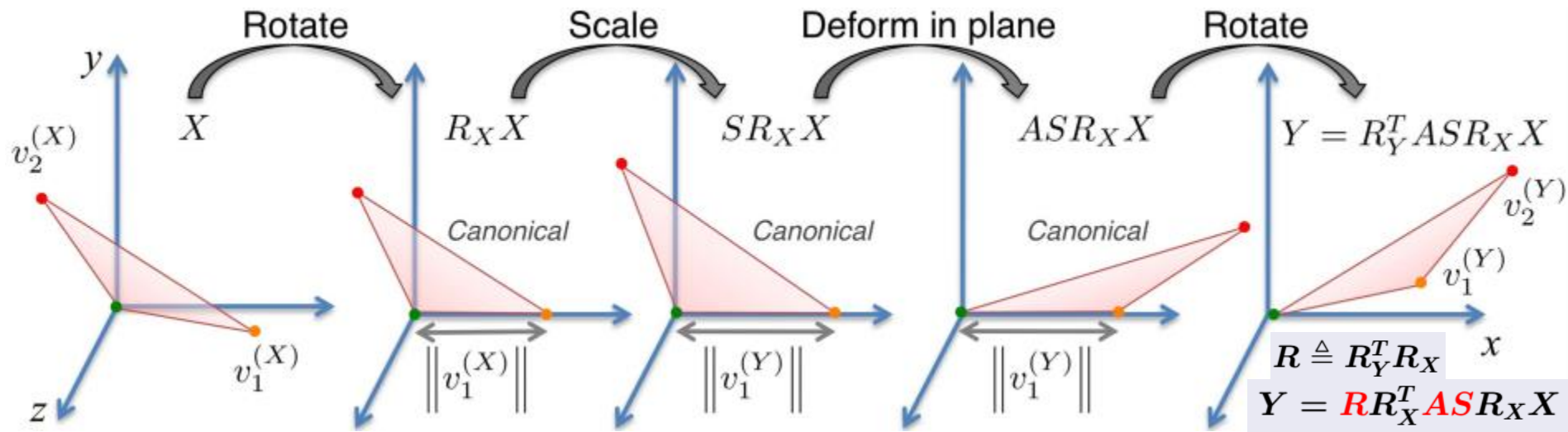
Manifold Representation of 3D Human Shape

- A shape is a point on a **non-linear manifold**, $M \triangleq G_T^N$.
 - G_T : is a **6D** Lie Group of triangle deformations
- **Advantages:**
 - Consistency (positive determinant)
 - No redundant DoF \Rightarrow less noise
 - Closed-form formulas



[Freifeld, O., Black, M. J., Lie Bodies: A Manifold Representation of 3D Human Shape](#), In *European Conf. on Computer Vision (ECCV)*, pages: 1-14, Part I, LNCS 7572, (Editors: A. Fitzgibbon et al. (Eds.)), Springer-Verlag, October 2012.

Deformation Transfer on Manifolds (Triangles deformation)



If a triangle X is **not canonical**, there is always a rotation matrix, R_x such that $R_x X$ is **canonical**.

$$S = \frac{\|v_1^{(Y)}\|}{\|v_1^{(X)}\|}, \quad A = \begin{bmatrix} 1 & U & 0 \\ 0 & V & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad U = \frac{(v_{2x}^{(Y)} - v_{2x}^{(X)})}{v_{2y}^{(X)}}, \quad V = \frac{v_{2y}^{(Y)}}{v_{2y}^{(X)}}$$

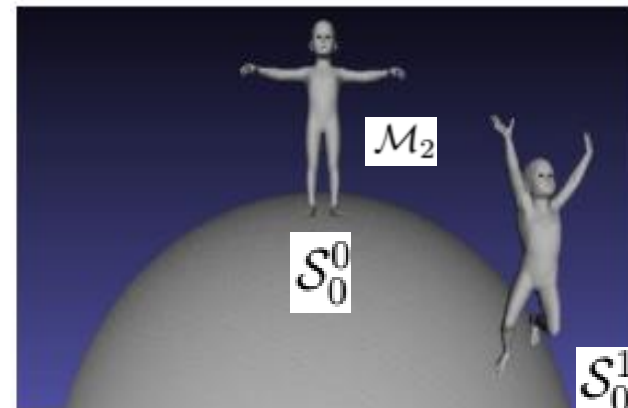
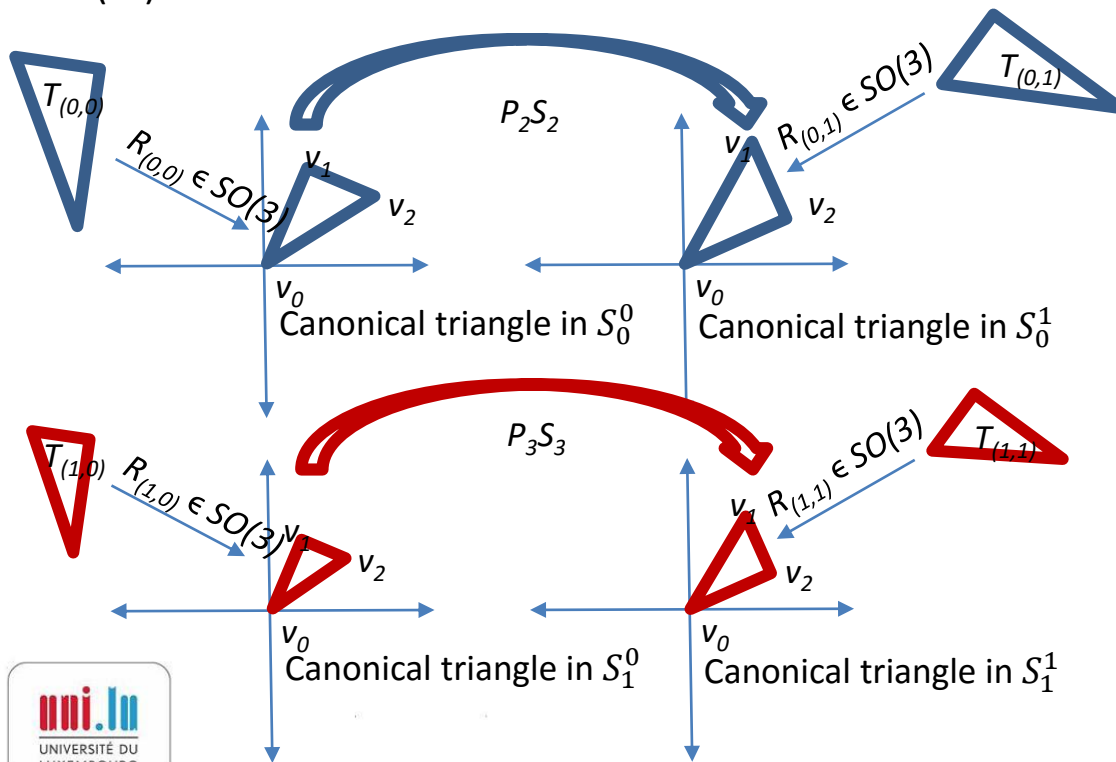
6D Lie Group of triangle deformation

Deformation Transfer on Manifolds

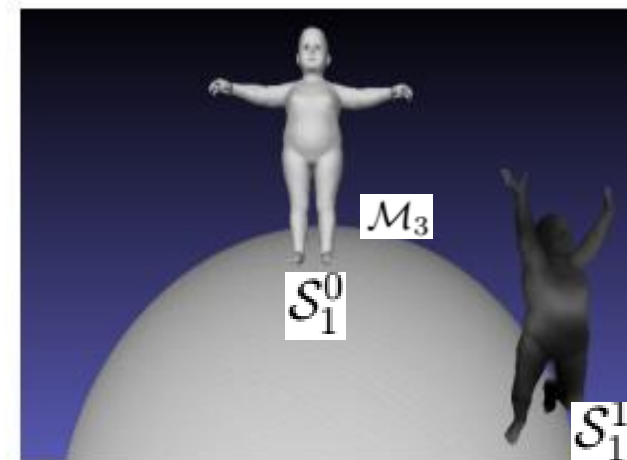
Shape (S) Rotate (R) Deform (P) Scale (S) Rotate (R)

$$\begin{cases} \mathcal{M}_1 : \mathcal{S}_1^0 \ni T_{(1,0)}^k = (R_{(1,0)}^k)^T P_{\mathcal{M}_1}^k S_{\mathcal{M}_1}^k R_{(0,0)}^k T_{(0,0)}^k, \\ \mathcal{M}_2 : \mathcal{S}_0^1 \ni T_{(0,1)}^k = (R_{(0,1)}^k)^T P_{\mathcal{M}_2}^k S_{\mathcal{M}_2}^k R_{(0,0)}^k T_{(0,0)}^k, \\ \mathcal{M}_3 : \mathcal{S}_1^1 \ni T_{(1,1)}^k = (R_{(1,1)}^k)^T P_{\mathcal{M}_3}^k S_{\mathcal{M}_3}^k R_{(1,0)}^k T_{(1,0)}^k. \end{cases}$$

Manifold (M) Deformed triangle (T) Input triangle (T)



(a)



(b)

Deformation Transfer on Manifolds

$$\exists A^k \in SO(3) \quad \text{s.t.} \quad R_{(1,0)}^k = A^k R_{(0,0)}^k, \quad \forall k.$$

$$P_{\mathcal{M}_2}^k S_{\mathcal{M}_2}^k \cong P_{\mathcal{M}_3}^k S_{\mathcal{M}_3}^k, \quad \forall k.$$

$$\hat{R}_{(1,1)}^k = A^k R_{(0,1)}^k, \quad \forall k.$$

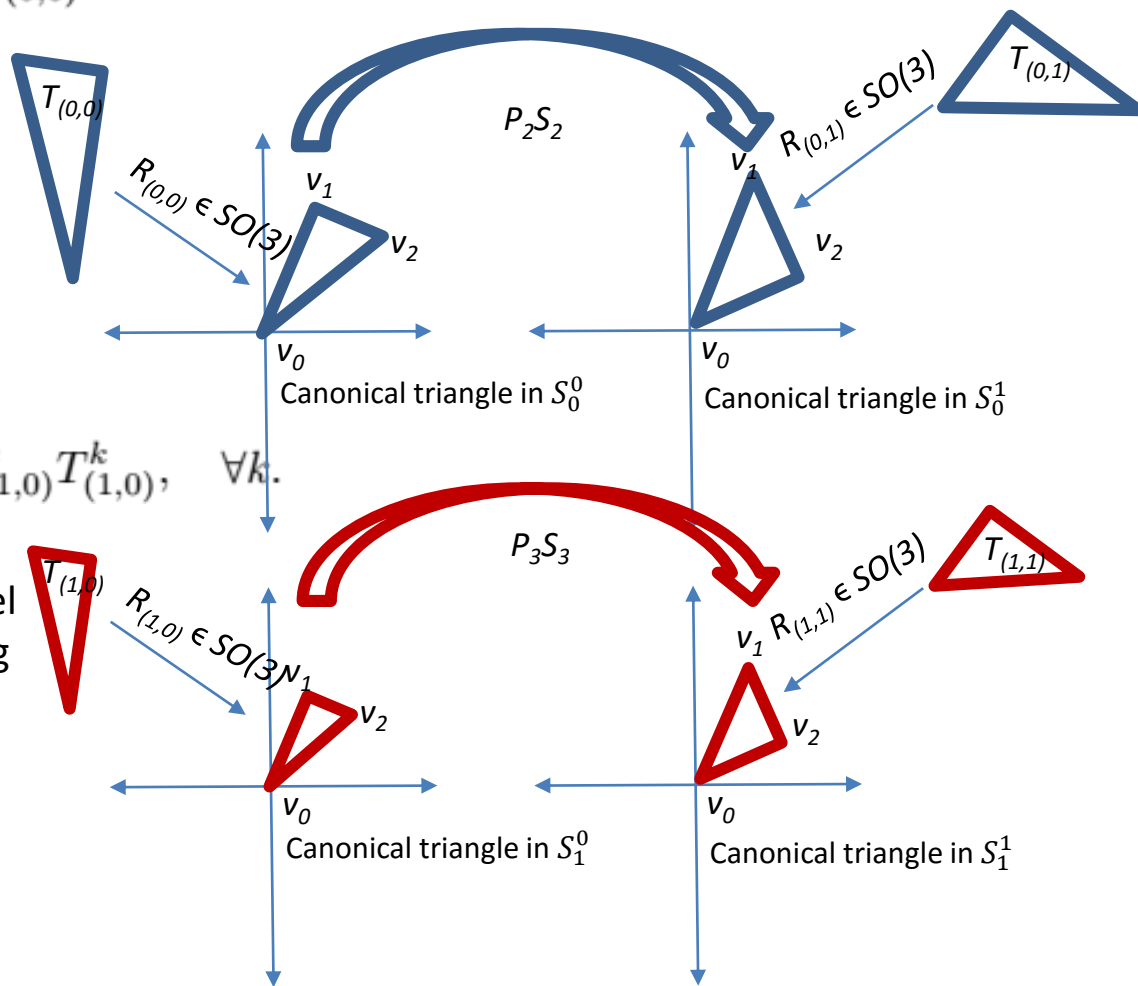
$$\hat{S}_1^1 \ni \hat{T}_{(1,1)}^k = \underbrace{(A^k R_{(0,1)}^k)^T}_{\text{Parallel Rotation}} \underbrace{P_{\mathcal{M}_2}^k S_{\mathcal{M}_2}^k}_{\text{Parallel Deformation}} \underbrace{R_{(1,0)}^k}_{\text{Parallel Scaling}} T_{(1,0)}^k, \quad \forall k.$$

Deformed Triangle

Parallel Rotation

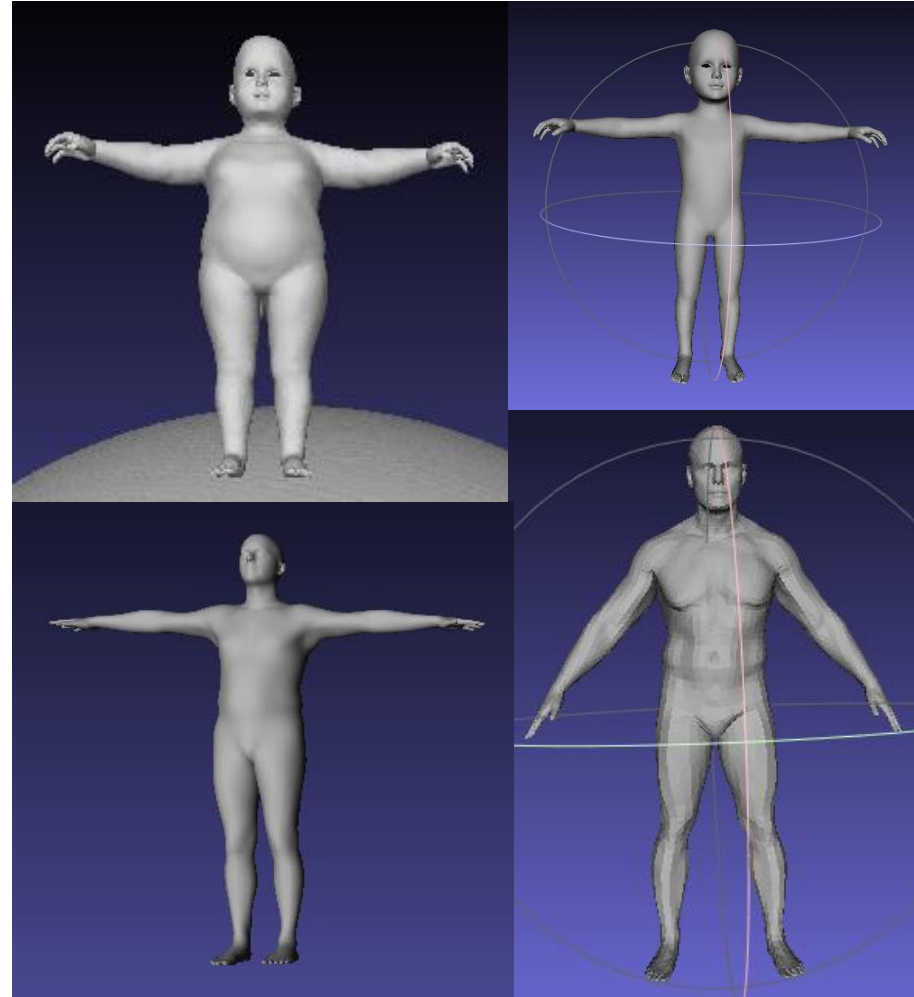
Parallel Deformation

Parallel Scaling



Results: 3D Human Datasets

- Input:
 - Rest-poses for the template and the target model to be transformed.
 - New poses taken by the input.
- SHREC (32, 2 kids x 16 poses)
 - The registered models were directly used
- FAUST (300, 10 persons x 30 poses)
 - The registered models were directly used.
- SMPL based generated models.

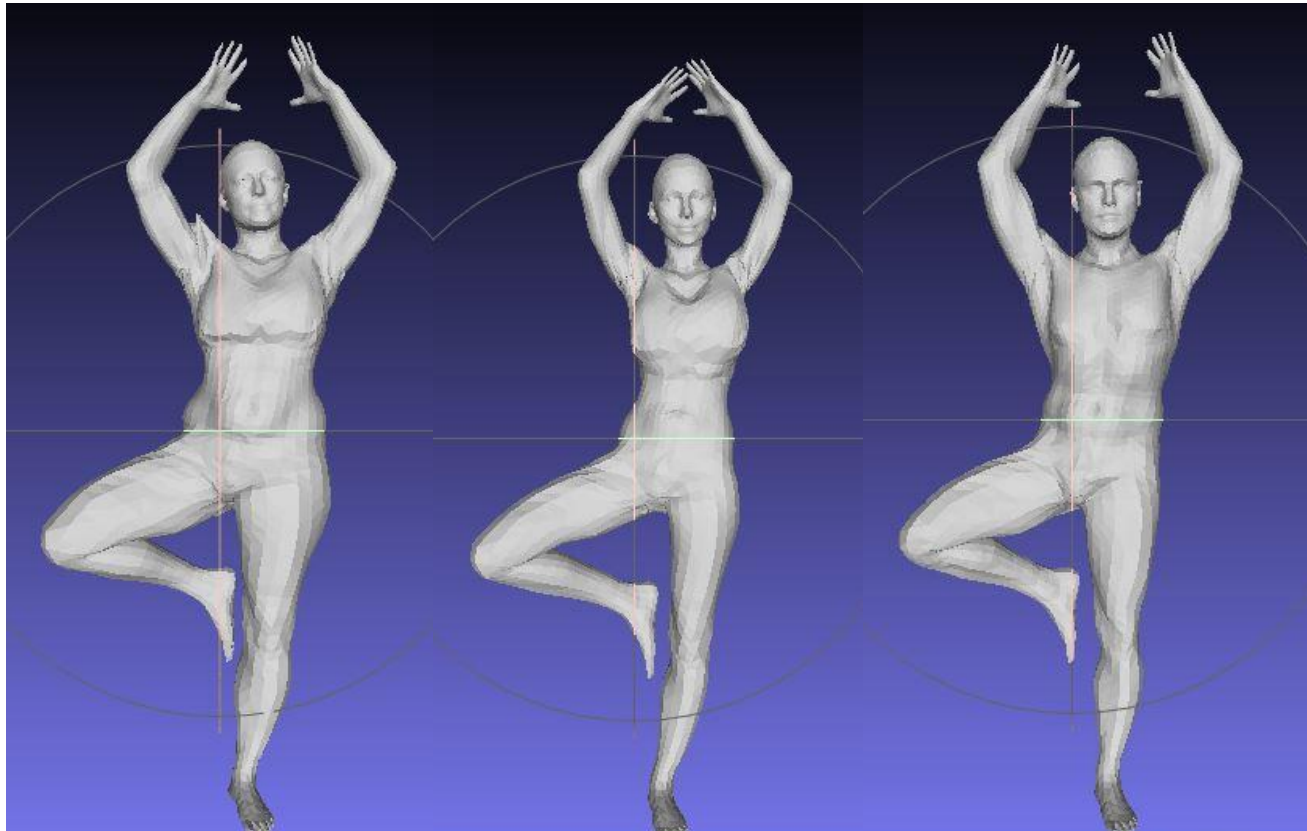


Results: SHREC



1st row: Input poses, 1st col: Input rest-poses
2nd row: our method
3rd row: Popovic & Sumner 2004

Results: Deformation Transfer applied to different rest-poses



Output generated using Popovic & Sumner 2004

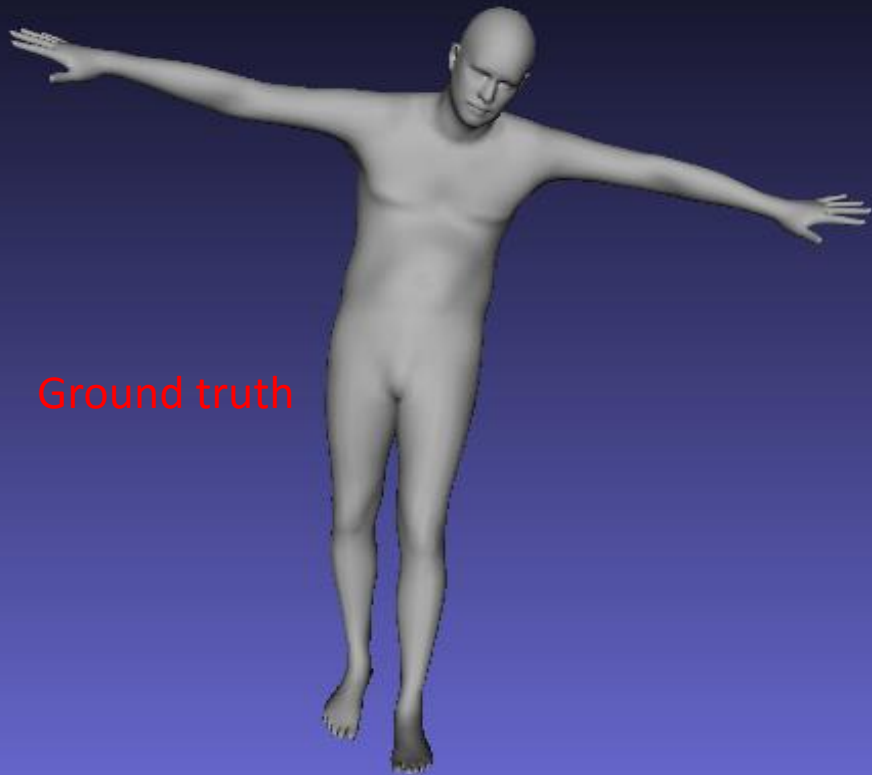
- **Any** pose can be taken as a **starting (rest, template)** pose as long as it is available for the input and target models.
- **The closer** the rest-poses are, the more **accurate** the output is.

Results: FAUST



1st row: Input poses
1st col: Input rest-poses
2nd row: our method
3rd row: Popovic & Sumner 2004

Results: SMPL



Ground truth



Computed

Conclusion & Future Work

- A novel deformation transfer technique to copy deformations on manifolds was proposed. Its advantages:
 - It uses minimal required DoF (**eliminates redundant DoF**).
 - Consistent deformations to **eliminate non-physical** deformations.
 - Deformations to be computed in **closed-form** without heuristics.
 - A **more accurate representation** than the traditional Euclidean representation.
 - **Less computationally expensive** (Parallel computations on triangles and no lifting up to the tangent space is required).
- Extend Manifold representation to broader applications like 3D Shape and Facial descriptors, Mesh encoding, Mesh editing, ... etc.

Thanks 😊

Questions ?

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