

Introduction to Non-Rigid Registration

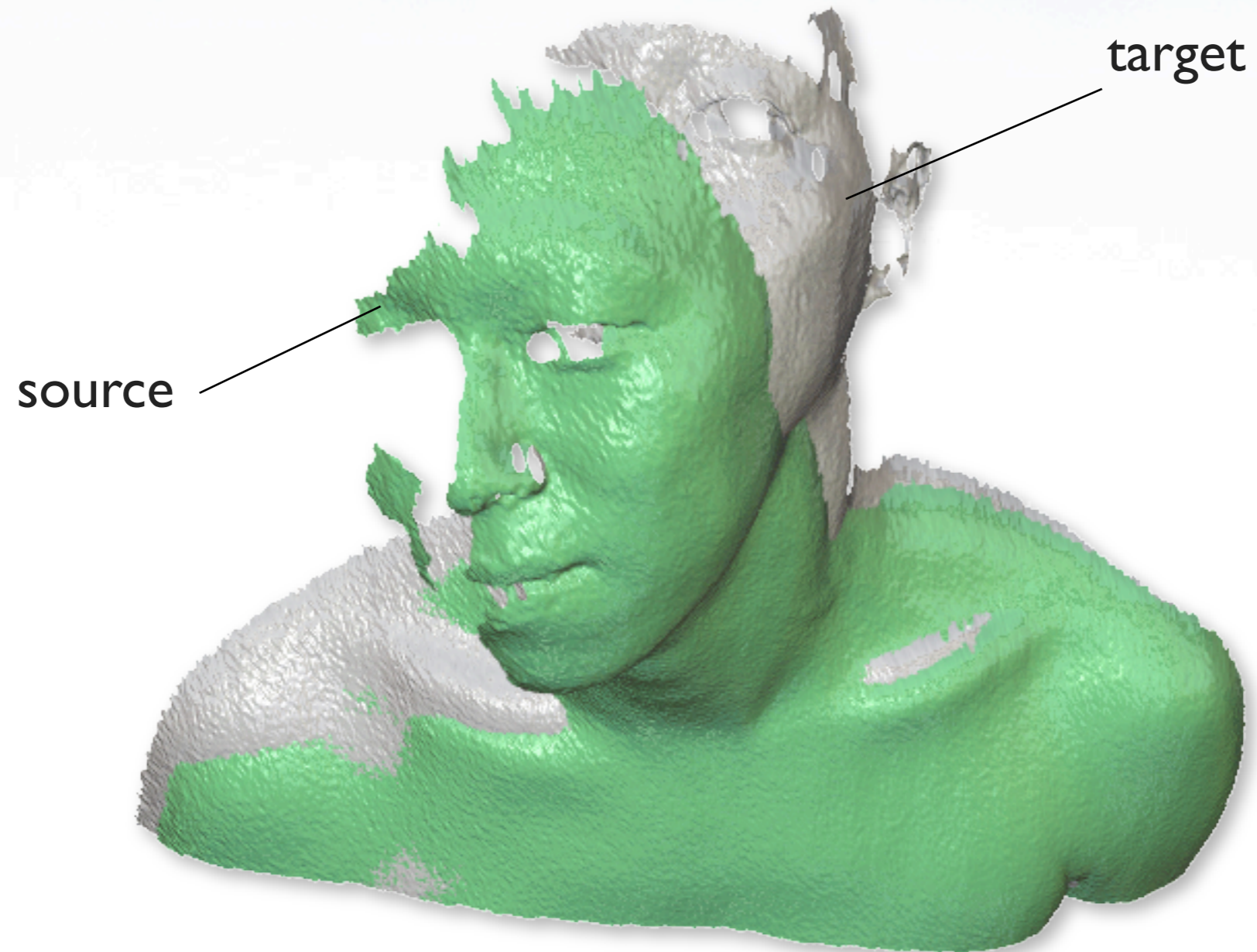
Hao Li

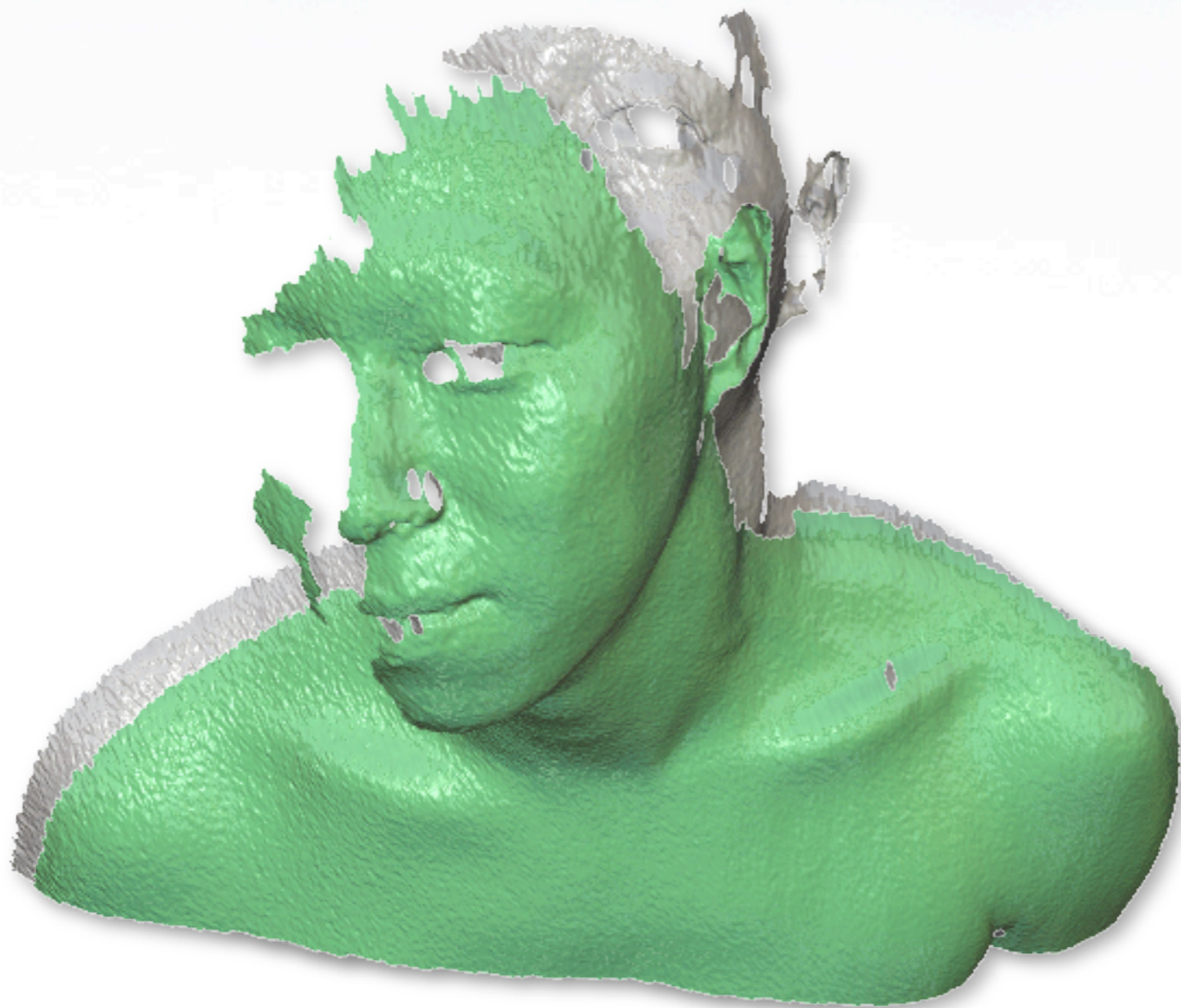


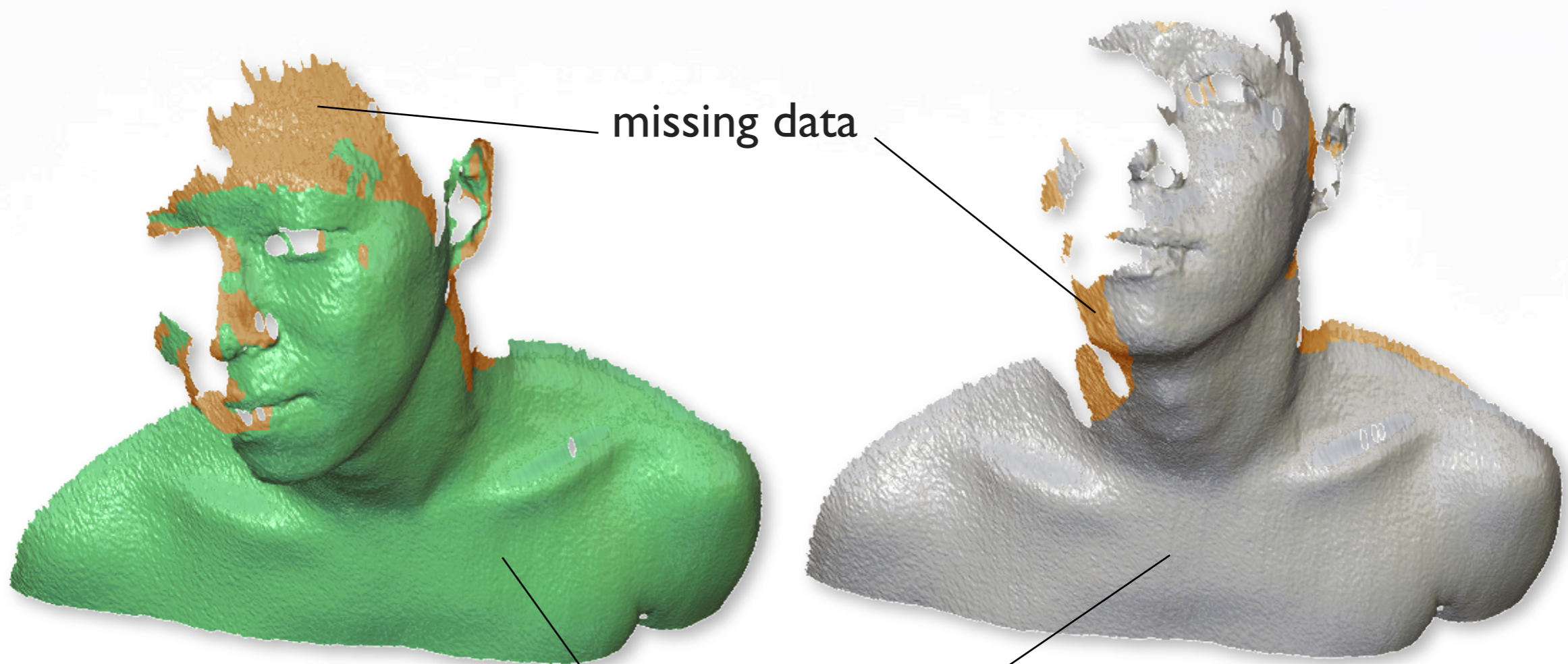
Applied Geometry Group
ETH Zurich



What is Registration?







missing data

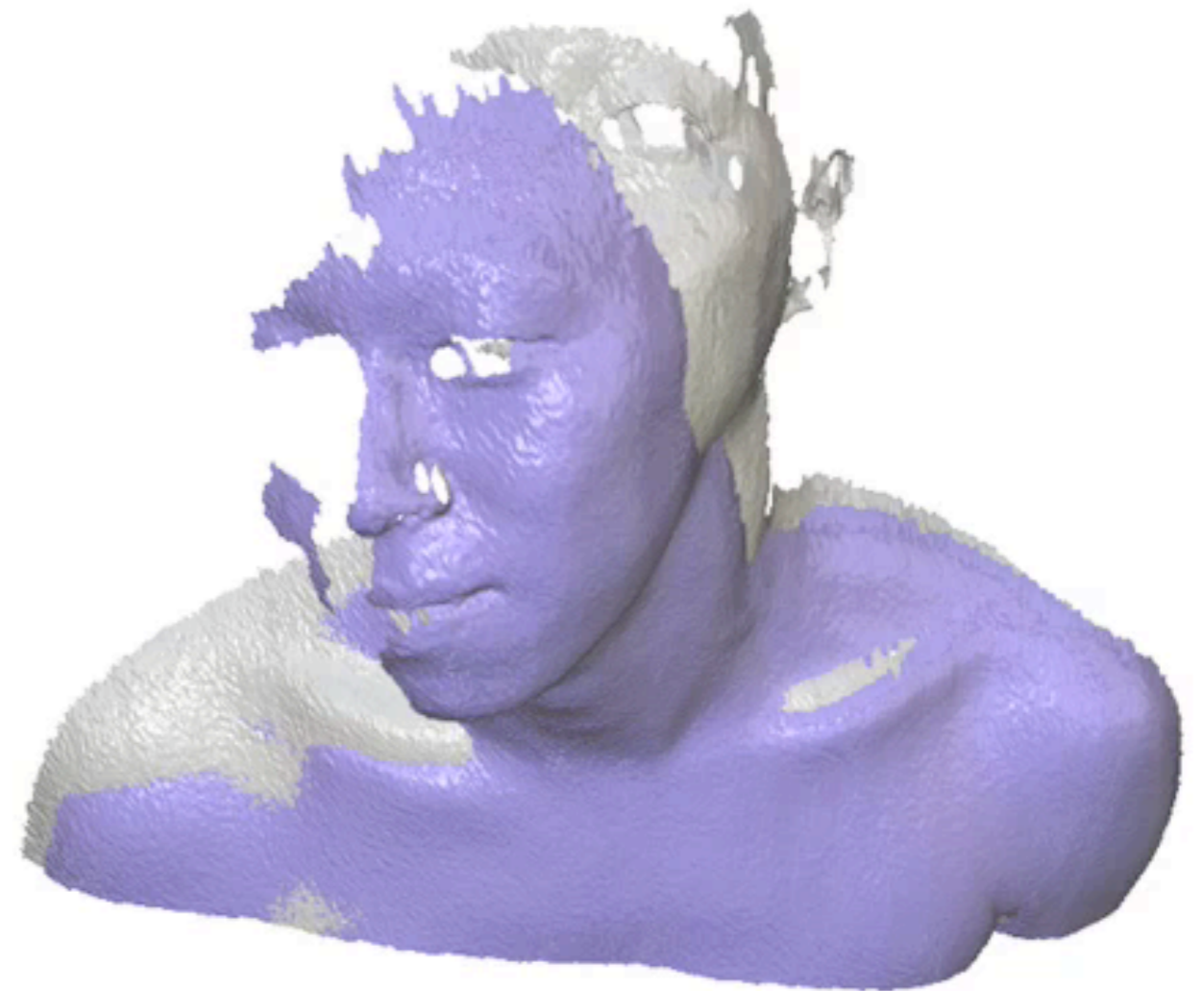
overlapping regions



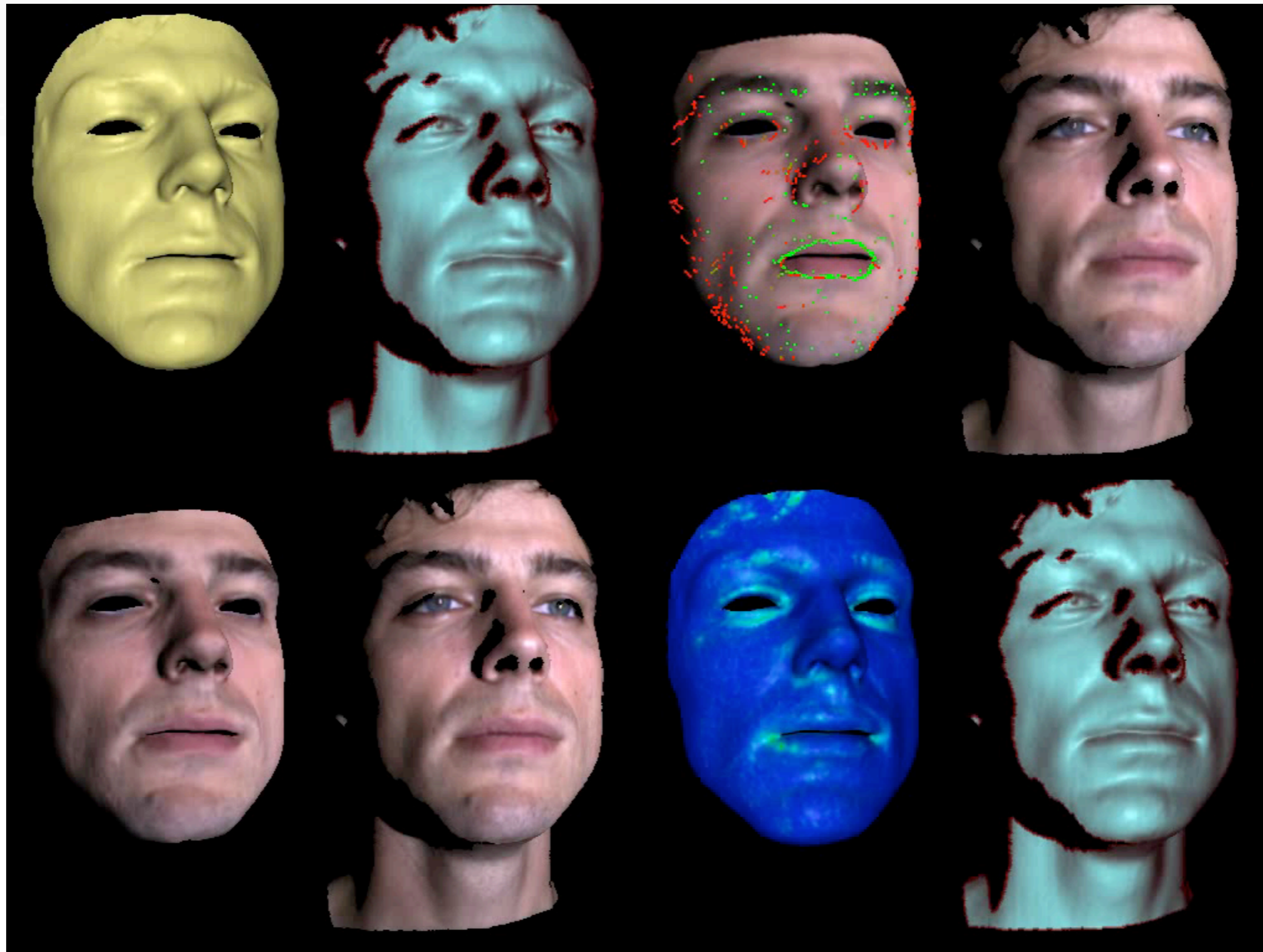
initial alignment



non-rigid registration



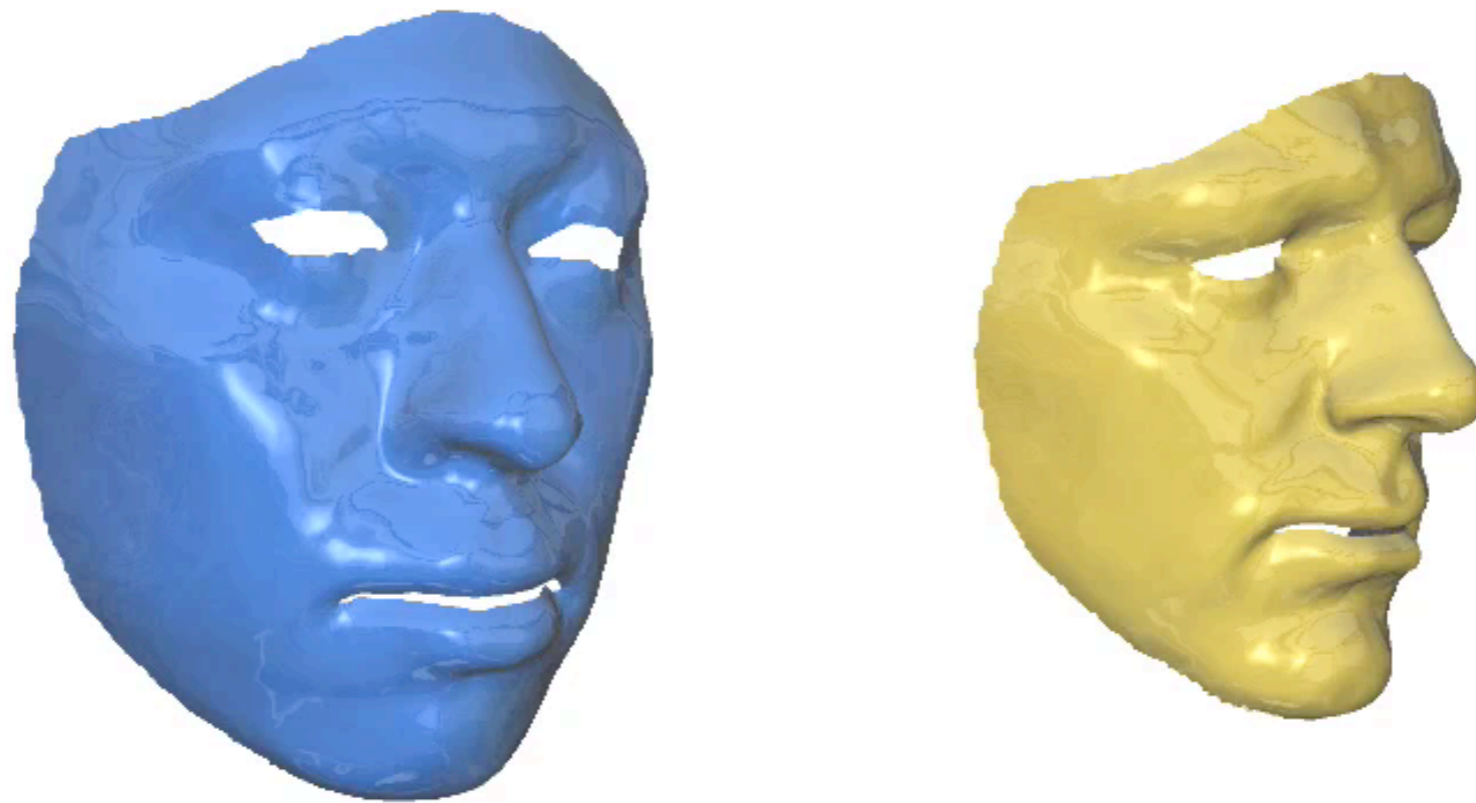
Template Fitting



Collaboration with T.Weise and L.Van Gool



Deformation Transfer



Collaboration with T.Weise and L.Van Gool



Correspondence between Shapes

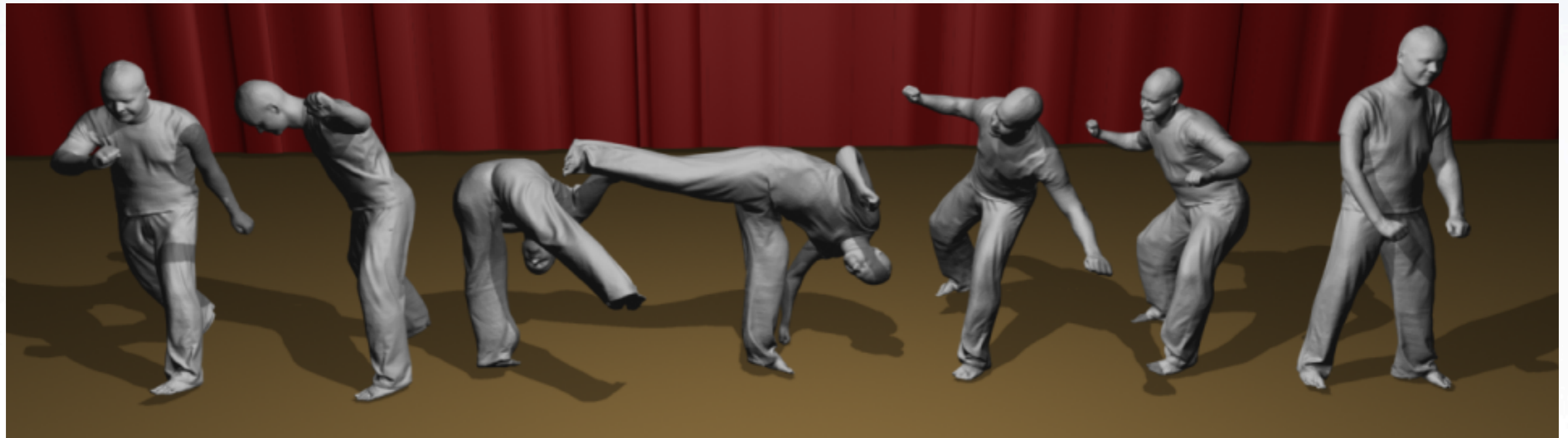


average shape warping

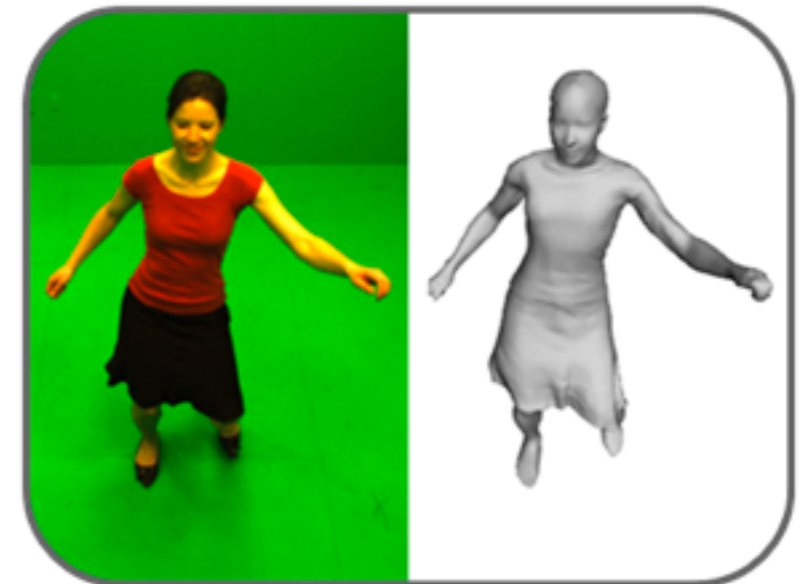
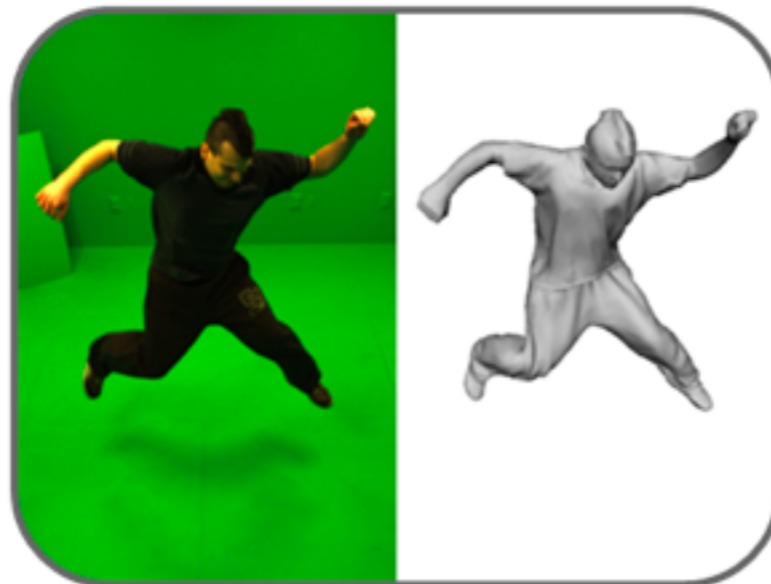
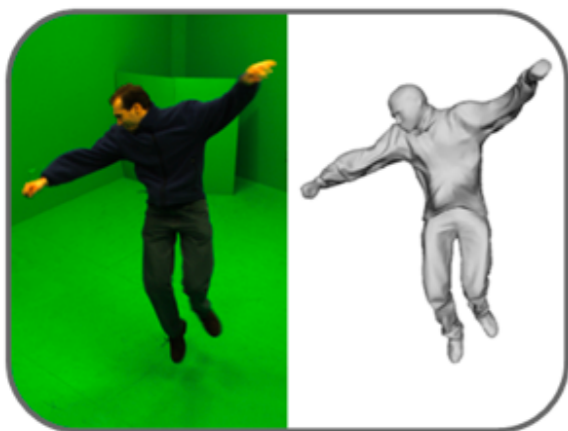


Collaboration with T.Weise and L.Van Gool

Performance Capture – More next week



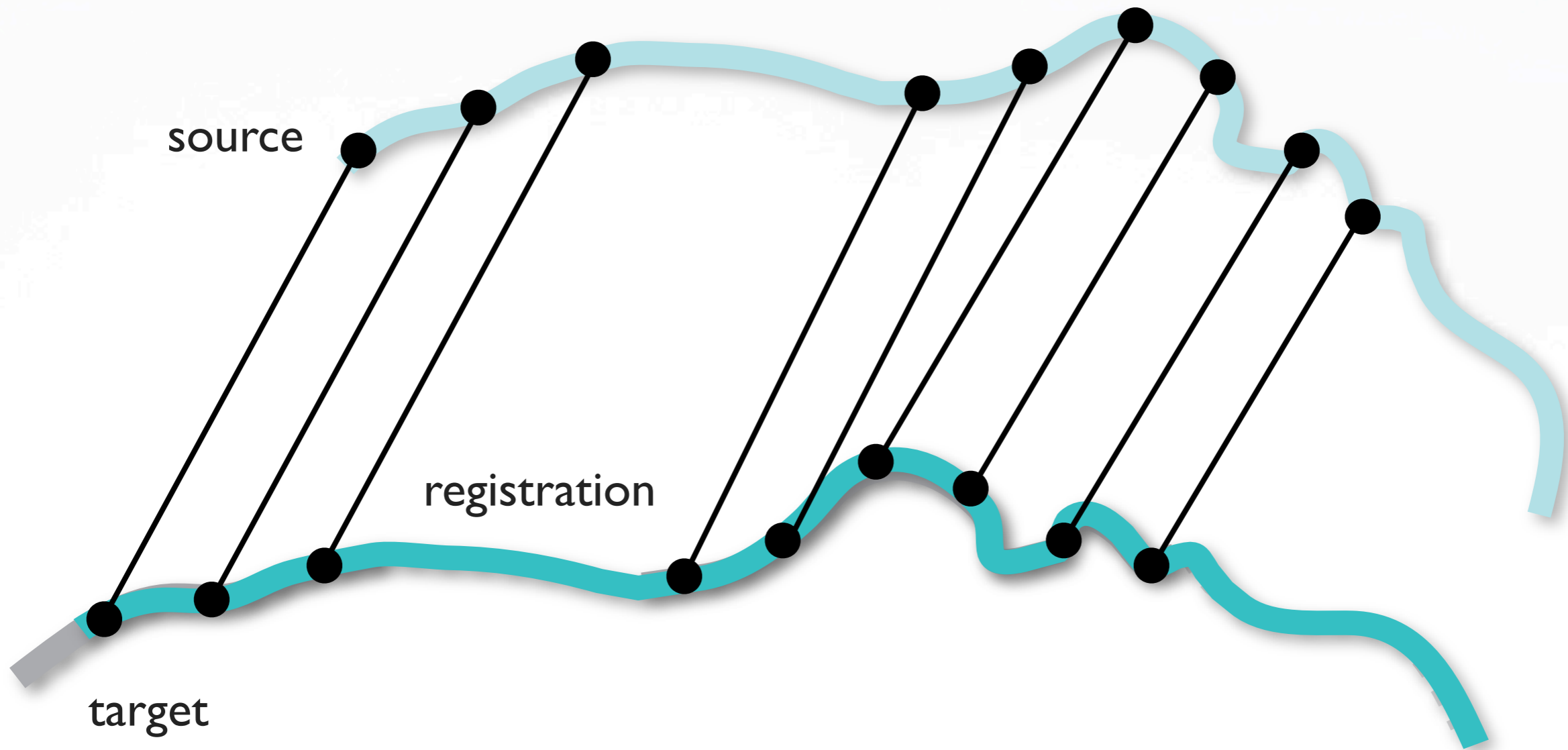
Source: [De Aguiar et al. 08]



Source: [Vlasic et al. 08]



Ingredients for Registration



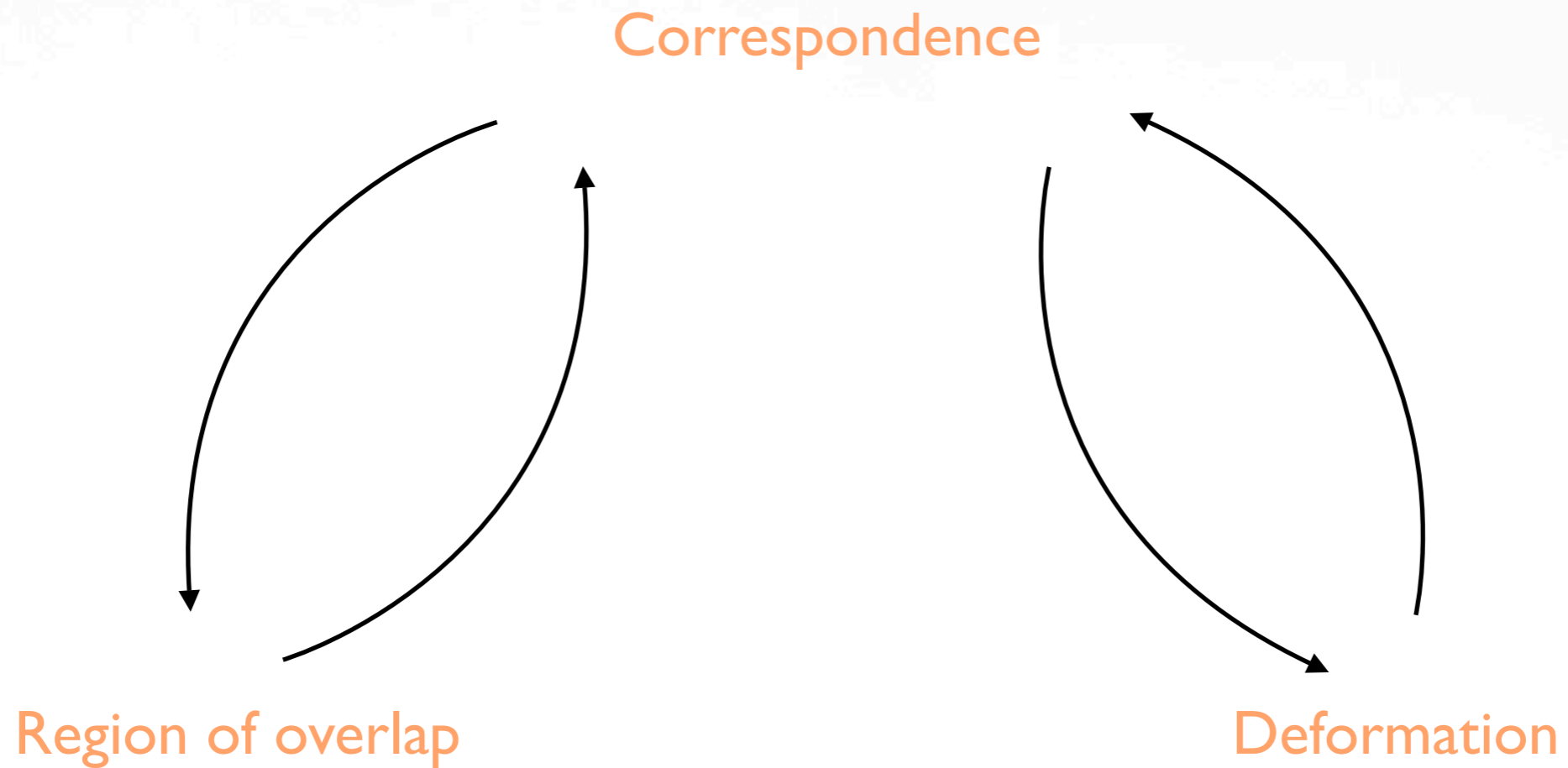
Partial Overlap

Correspondence

Deformation



Three Interdependent Challenges



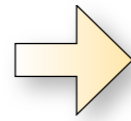
Rigid Alignment



Problem



$\mathcal{M}_1 \dots \mathcal{M}_n$



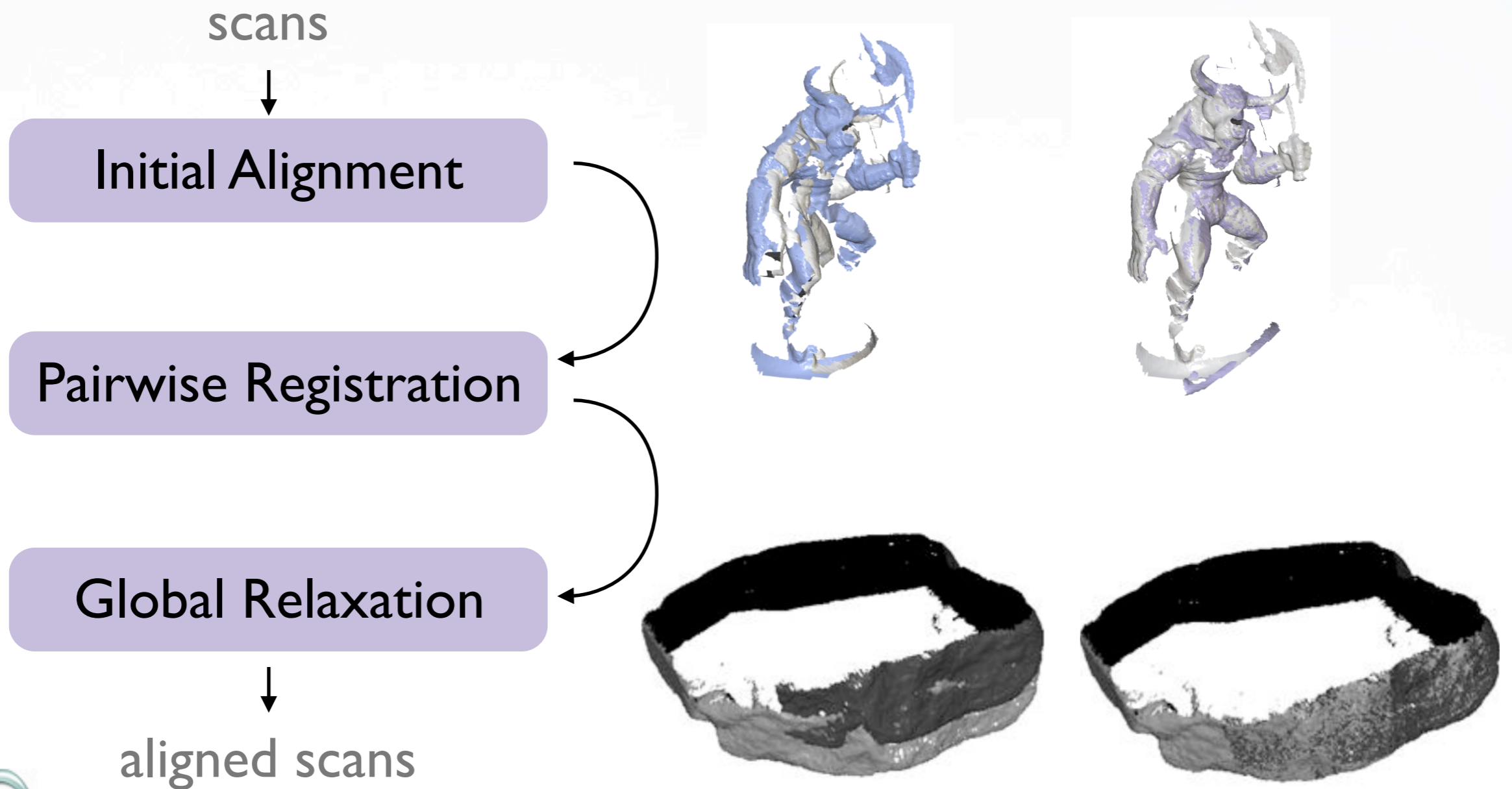
$(R_1, \mathbf{t}_1) \dots (R_n, \mathbf{t}_n)$



$\tilde{\mathcal{M}}_1 \dots \tilde{\mathcal{M}}_n$

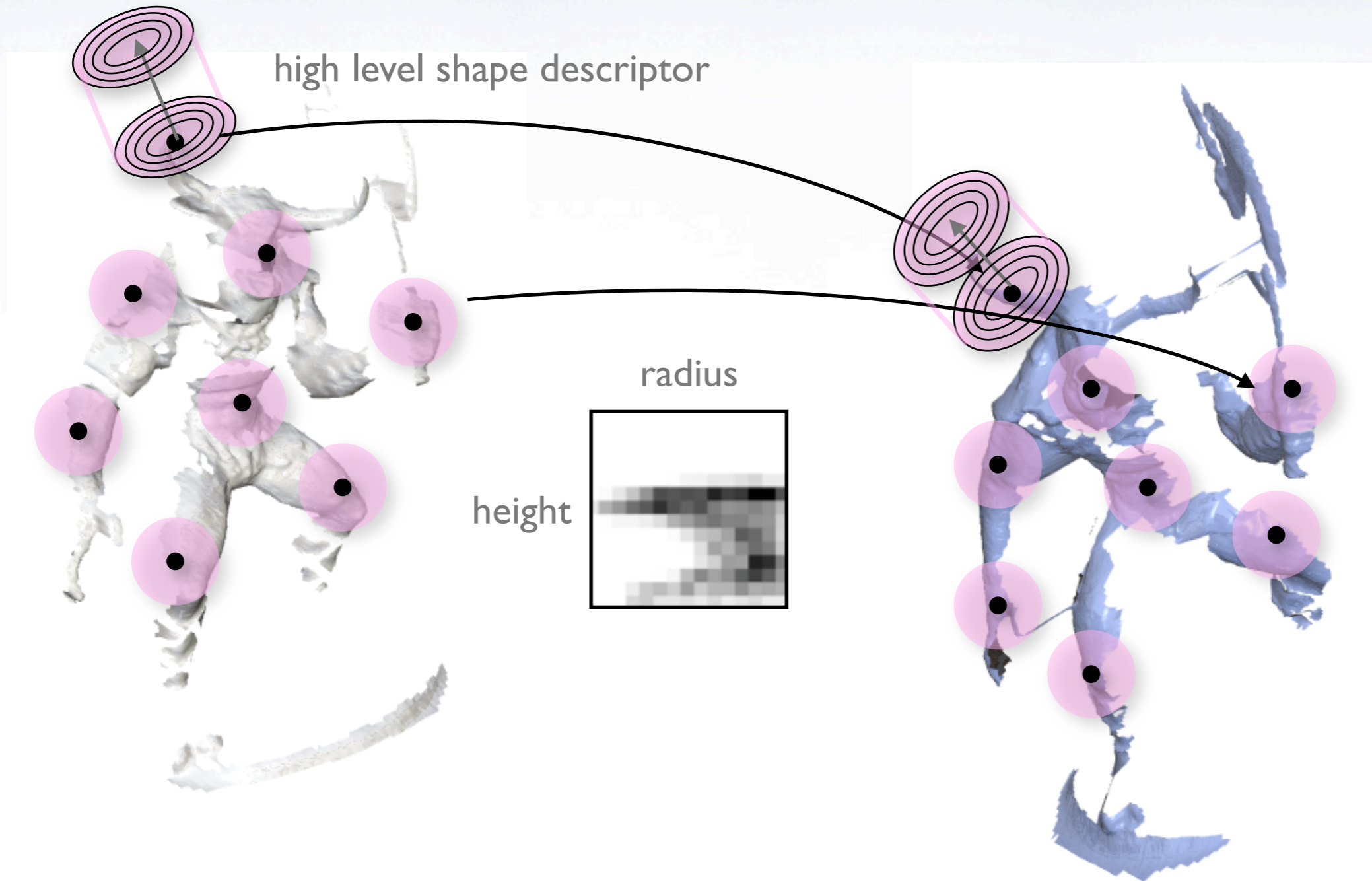


Registration Pipeline



source: [Pulli et al. 99]

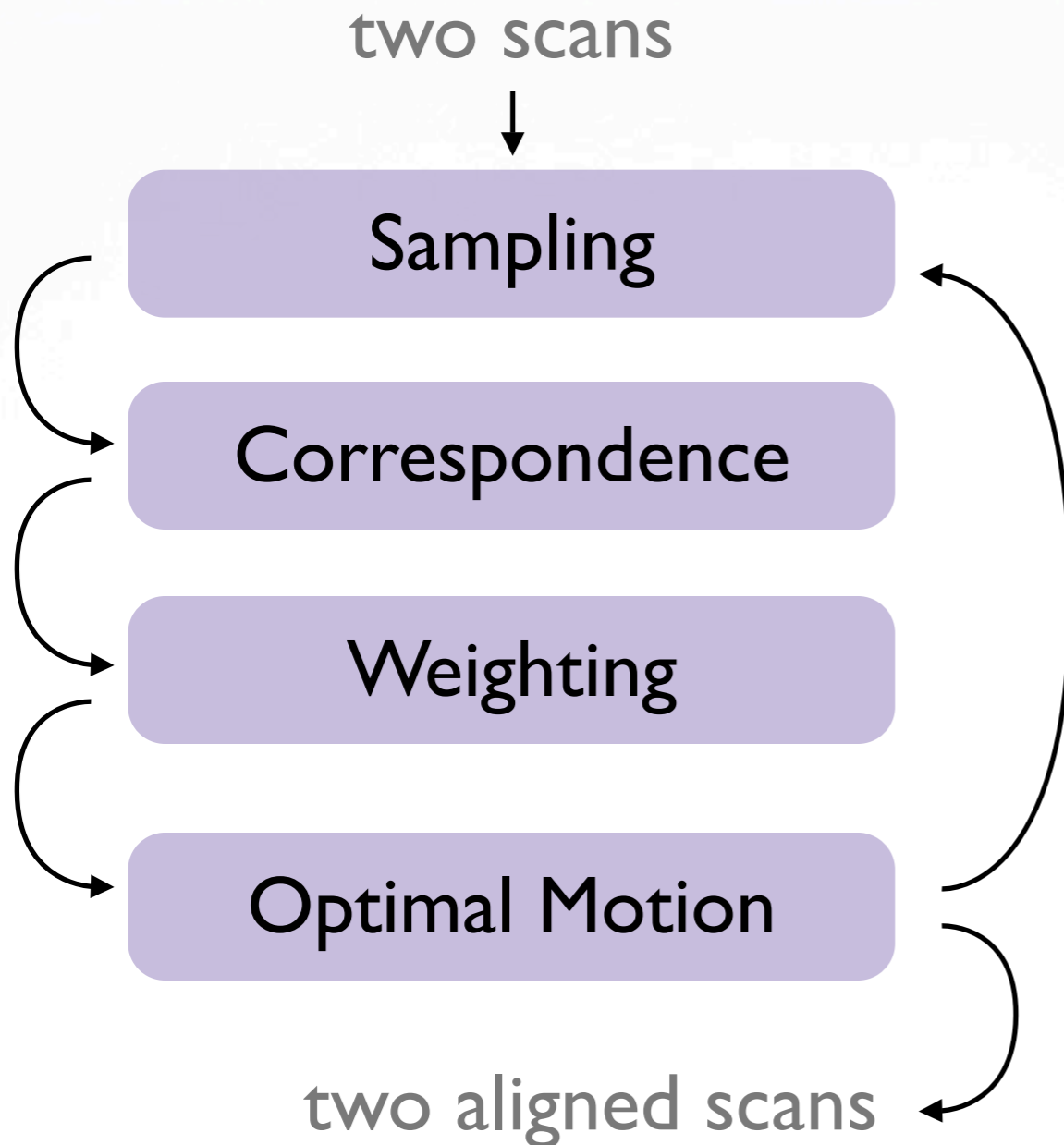
Initial Alignment



- Motion invariant shape descriptor (Spin images, harmonic shape context)
- Correspondence search (Brute force, branch and bound, RANSAC)



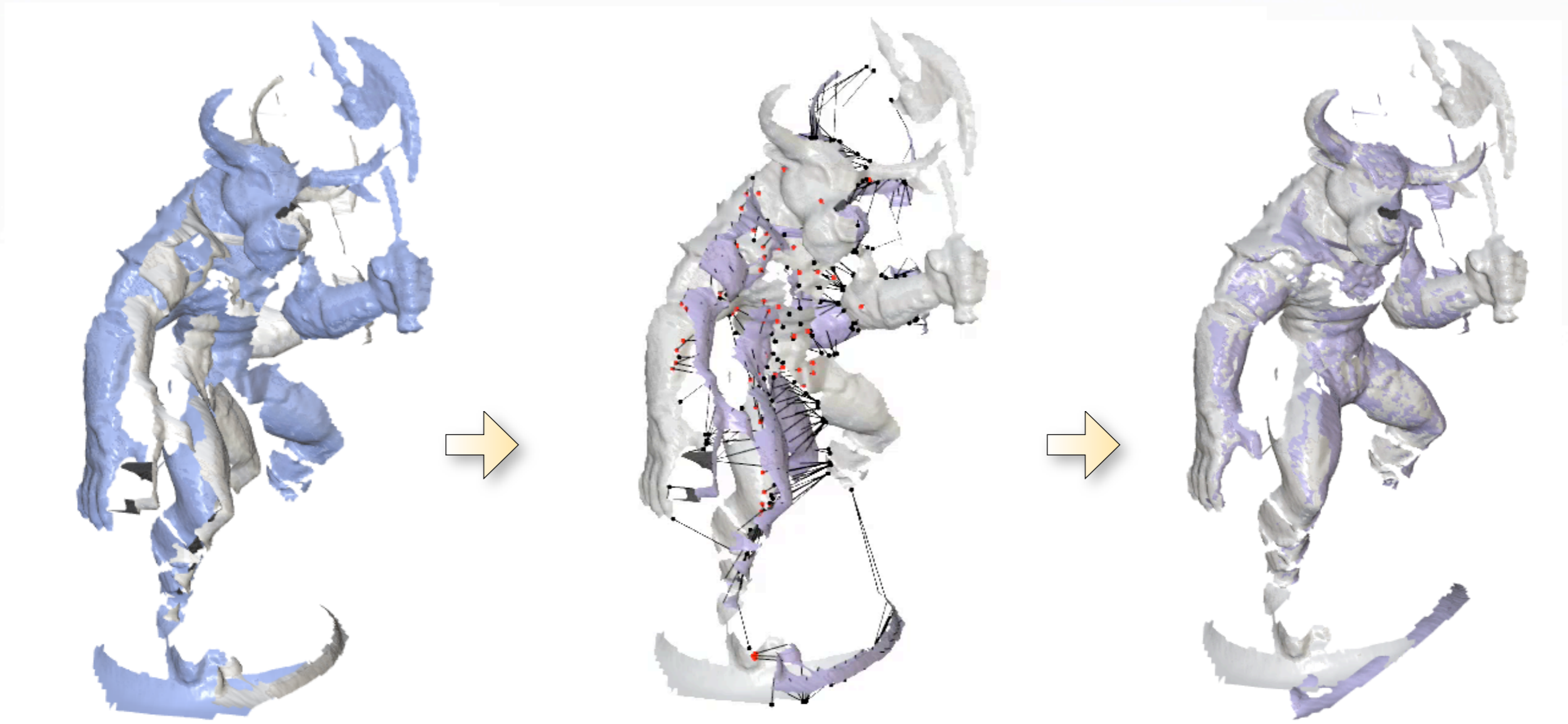
Iterative Closest Point (ICP)



- Uniform, Normal Importance, etc...
- Closest point via *kd* tree
- Correspondence compatibility heuristics
- Different optimization techniques

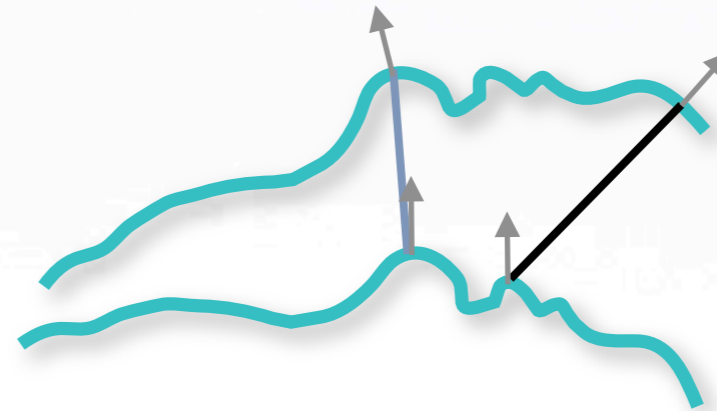


ICP Optimization

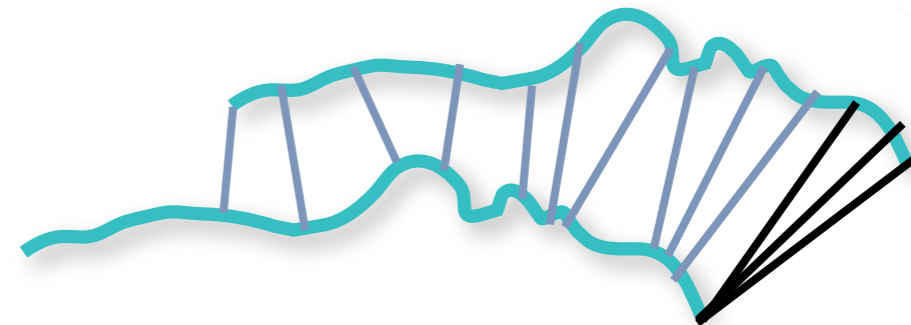


Effective Weighting Schemes

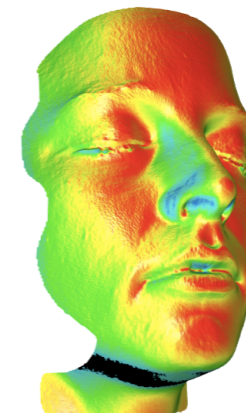
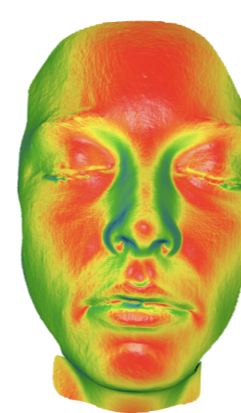
- Normal compatible correspondences



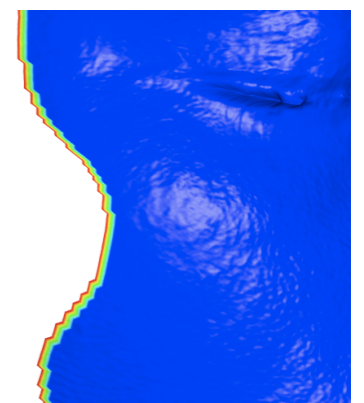
- Prune correspondences to boundaries and points that are too far



- Known surface confidence weights



camera orientation

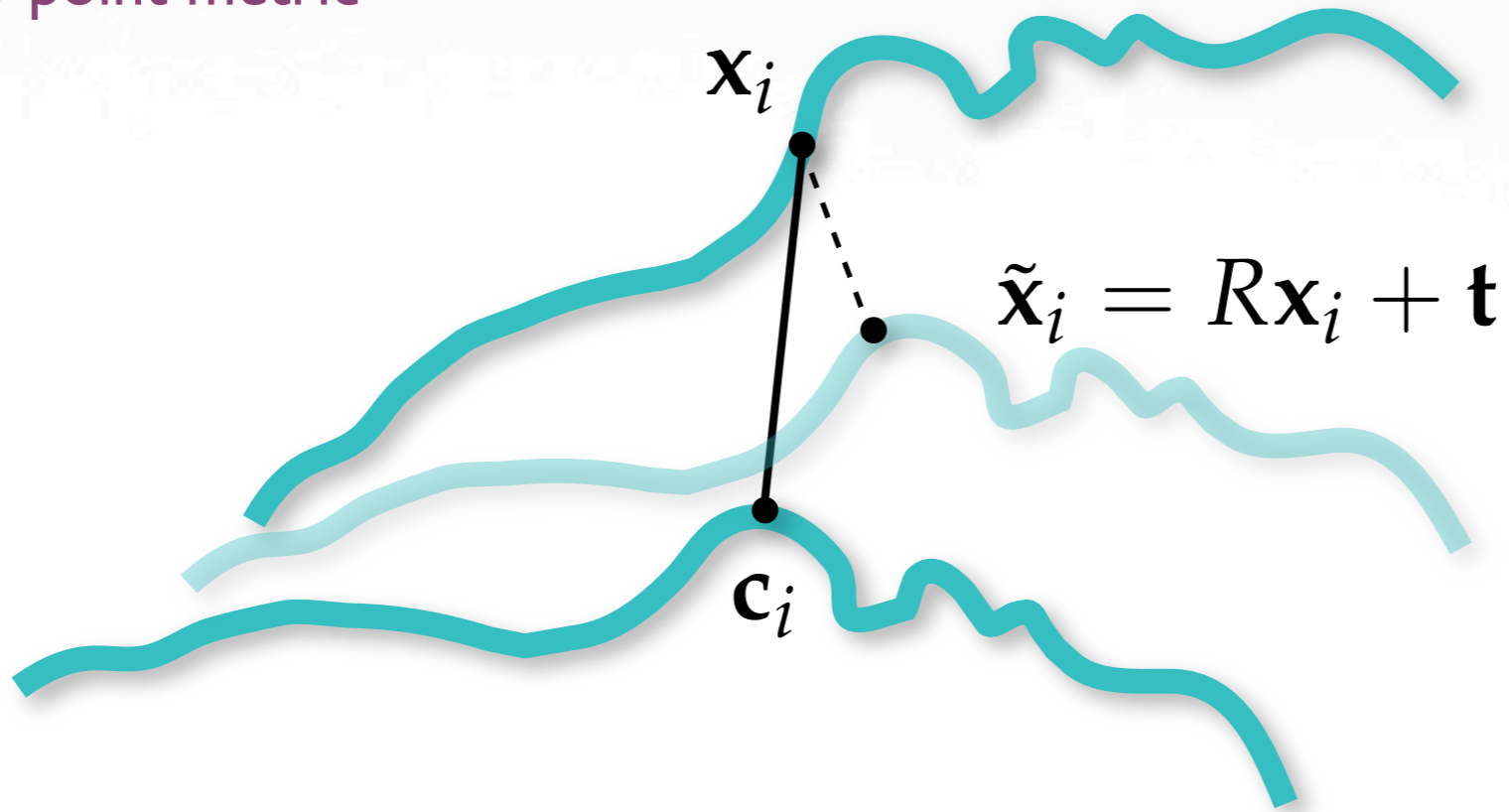


feathering



Optimization

Point-to-point metric



$$E_{\text{point}} = \sum_i \|(R x_i + t - c_i)\|^2$$

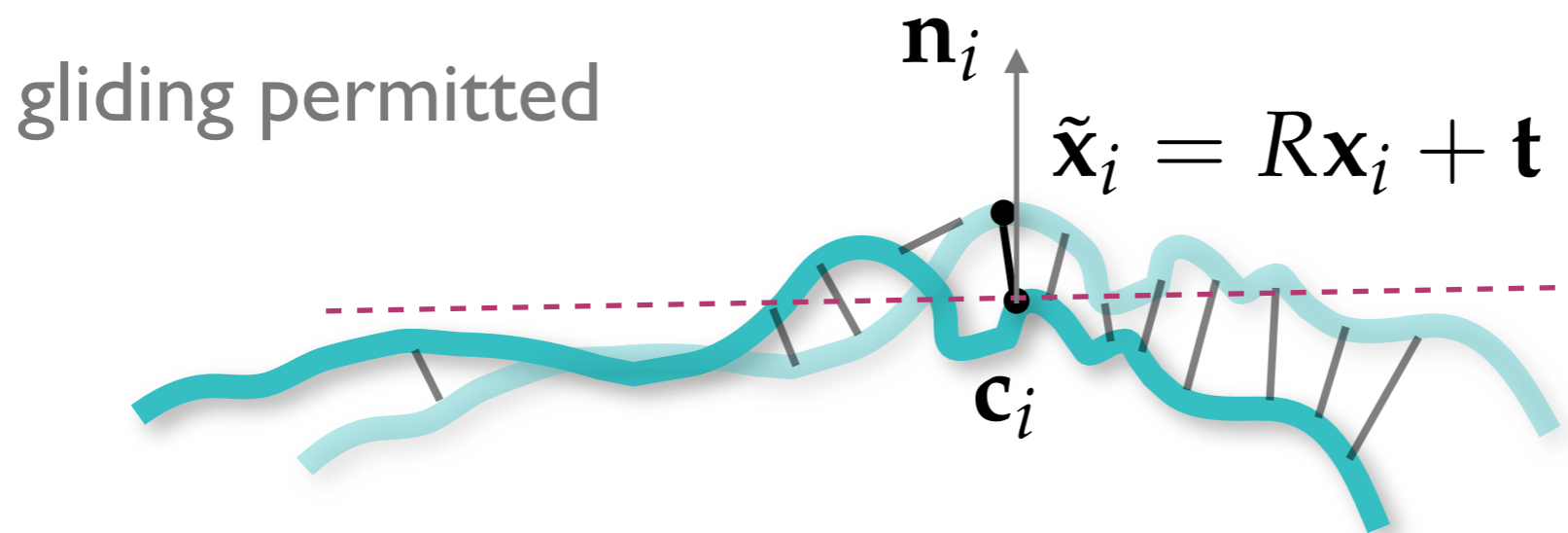
- Closed form solution via quaternions [Horn 87]



Optimization

[Chen and Medioni '91]

Point-to-plane metric



$$E_{\text{plane}} = \sum_i \|\mathbf{n}_i^t (R\mathbf{x}_i + \mathbf{t} - \mathbf{c}_i)\|^2$$

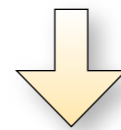


Optimization

$$E_{\text{plane}} = \sum_i \|\mathbf{n}_i^t (R\mathbf{x}_i + \mathbf{t} - \mathbf{c}_i)\|^2 \quad \sin(\theta) \approx \theta, \cos(\theta) \approx 1$$



$$E_{\text{plane}} = \sum_i \|\mathbf{n}_i^t (\mathbf{x}_i - \mathbf{c}_i) + \mathbf{r}^t (\mathbf{x}_i \times \mathbf{n}_i) + \mathbf{n}_i^t \mathbf{t}\|^2$$



$$\begin{bmatrix} -(\mathbf{x}_1 \times \mathbf{n}_1)^t & \mathbf{n}_1^t \\ -(\mathbf{x}_2 \times \mathbf{n}_2)^t & \mathbf{n}_2^t \\ \vdots & \vdots \end{bmatrix} \begin{bmatrix} \mathbf{r} \\ \mathbf{t} \end{bmatrix} = \begin{bmatrix} -(\mathbf{x}_1 - \mathbf{c}_1)^t \mathbf{n}_1 \\ -(\mathbf{x}_2 - \mathbf{c}_2)^t \mathbf{n}_2 \\ \vdots \end{bmatrix}$$

overdetermined linear system



In Practice

$$E_{\text{tot}} = E_{\text{plane}} + \lambda E_{\text{point}} \quad \lambda \approx 0.1$$

$$E_{\text{plane}} = \sum_i \|\mathbf{n}_i^t (R\mathbf{x}_i + \mathbf{t} - \mathbf{c}_i)\|^2$$

$$E_{\text{point}} = \sum_i \|(R\mathbf{x}_i + \mathbf{t} - \mathbf{c}_i)\|^2$$

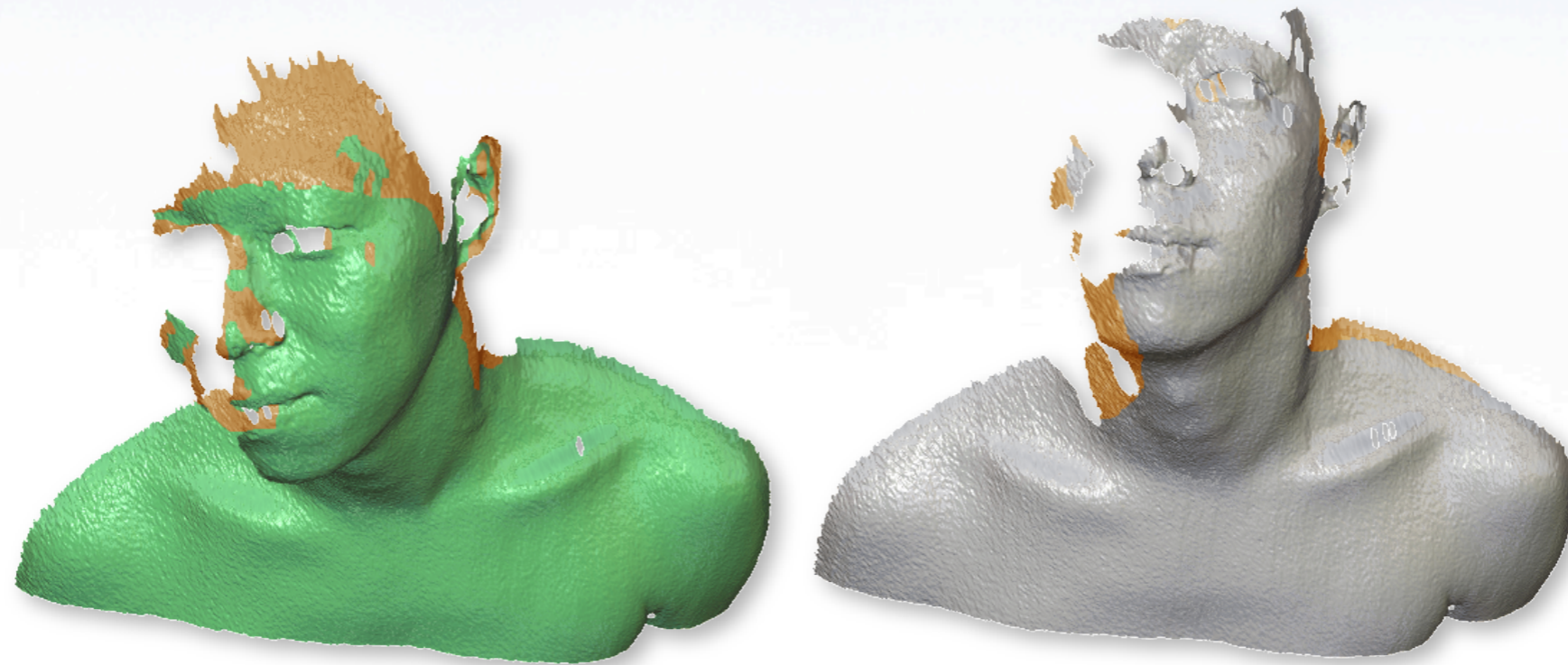
- Eplane alone can exhibit oscillations
- Favor Eplane when convergence is slow



Non-Rigid Registration



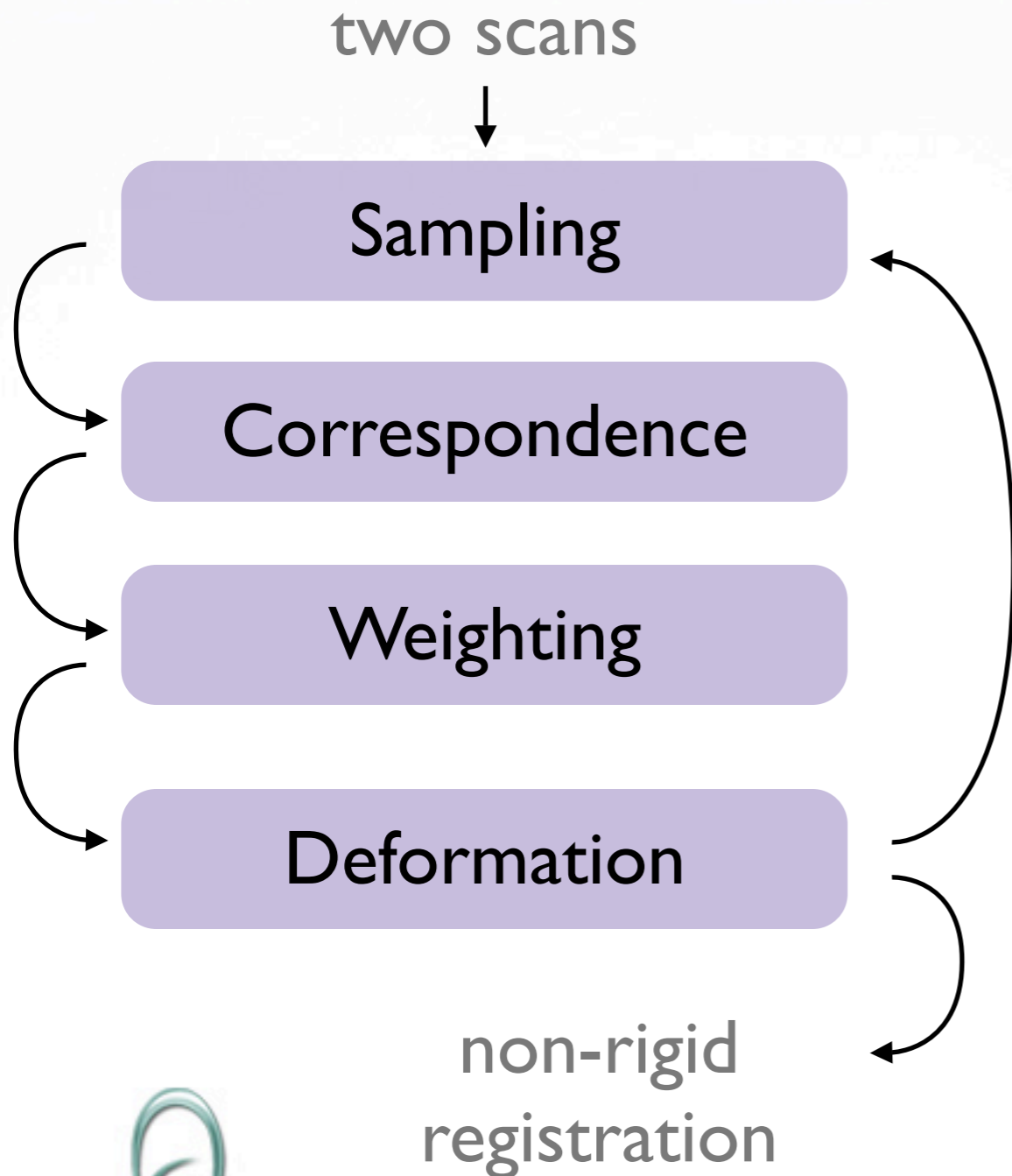
Problem



- Partial overlap unknown
- Correspondences not given
- No prior about deformation
- Interdependence of problems



Decoupled Approach



Correspondence must be robust w.r.t. underlying deformation

In general: Non-linear problem

$$E_{\text{tot}} = \alpha_{\text{fit}} E_{\text{fit}} + \alpha_{\text{reg}} E_{\text{reg}}$$



Non-Linear Optimization

$$E_{\text{tot}} = \alpha_{\text{fit}} E_{\text{fit}} + \alpha_{\text{reg}} E_{\text{reg}} \quad E_{\text{tot}} = \|\mathbf{f}(\mathbf{x})\|^2$$

1st order Taylor

$$\|\mathbf{f}(\mathbf{x}^{k+1})\|^2 \approx \|\mathbf{f}(\mathbf{x}^k) + J_{\mathbf{f}}(\mathbf{x}^{k+1} - \mathbf{x}^k)\|^2$$

$$\|\mathbf{f}(\mathbf{x}^{k+1})\|^2 \approx \|\mathbf{f}(\mathbf{x}^k) + J_{\mathbf{f}}\Delta\mathbf{x}^k\|^2$$

Gauss-Newton

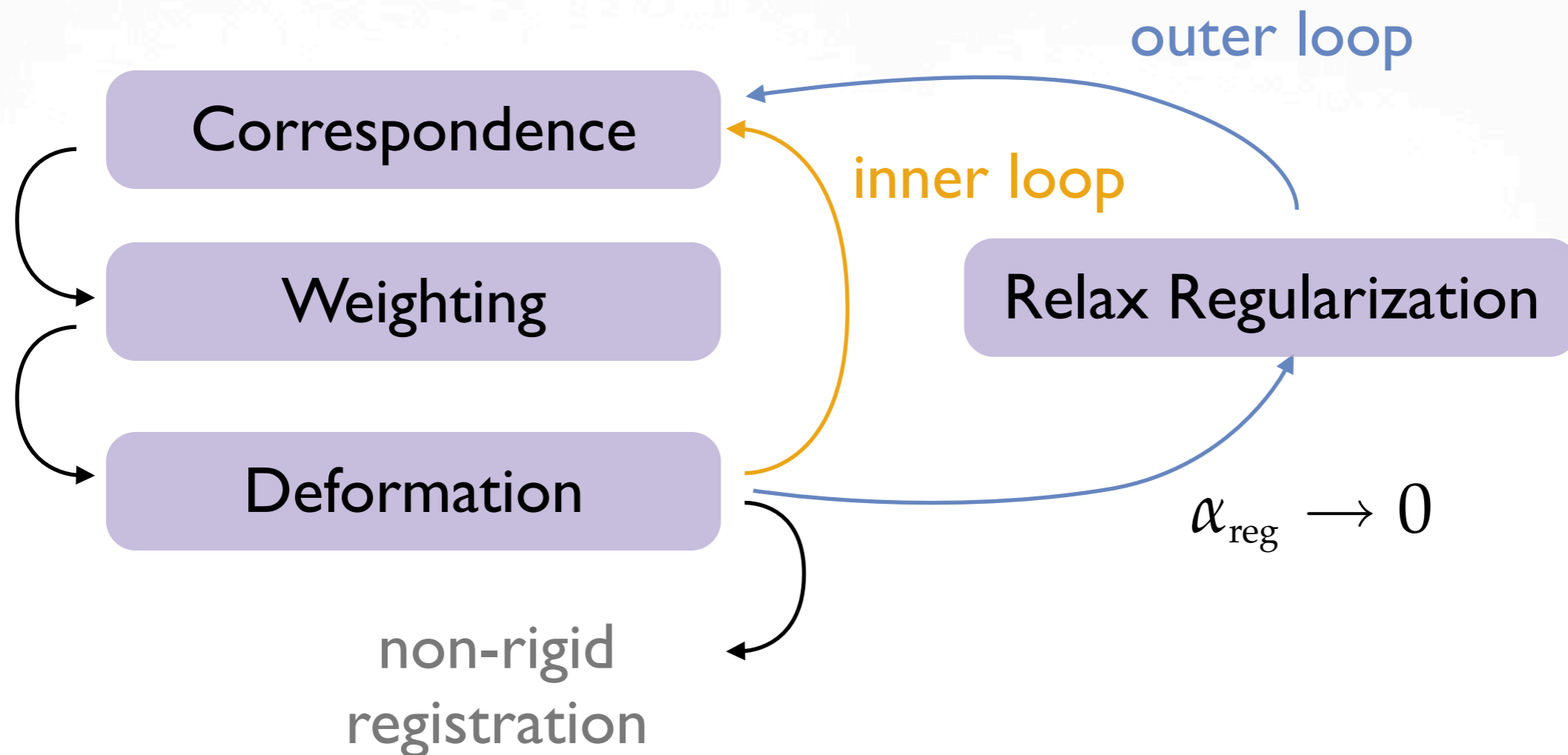
$$\Delta\mathbf{x}_{\text{min}}^k = \arg \min_{\Delta\mathbf{x}^k} E_{\text{tot}}$$

$$J_{\mathbf{f}}^t J_{\mathbf{f}} \Delta\mathbf{x}_{\text{min}}^k = -J_{\mathbf{f}}^t \mathbf{f}(\mathbf{x}^k)$$

- Use direct solver with sparse Cholesky factorization
- Extension to Levenberg-Marquardt
- Other techniques: Quasi-Newton, ...



Scheduled Regularization



- Energy landscape smoothing (prevents local minima)
- Other technique: multi-resolution



Deformation Models



Linear Regularization Methods

- Smooth Displacement

$$E_{\text{displ}} = \sum_{(\mathbf{x}_i, \mathbf{x}_j) \in \text{edges}} \|\mathbf{d}_i - \mathbf{d}_j\|^2 / \|\mathbf{x}_i - \mathbf{x}_j\|^2$$

- Smooth Affine Transforms

$$E_{\text{affine}} = \sum_{(\mathbf{x}_i, \mathbf{x}_j) \in \text{edges}} \|T_i - T_j\|_F^2, \quad T \in \mathbb{R}^{4 \times 4}$$

- Linear Variational Techniques [Botsch and Sorkine 07]

$$E_{\text{memb}} = \int_{\Omega} \|\mathbf{x}_u\|^2 + \|\mathbf{x}_v\|^2 \mathrm{d}u\mathrm{d}v$$

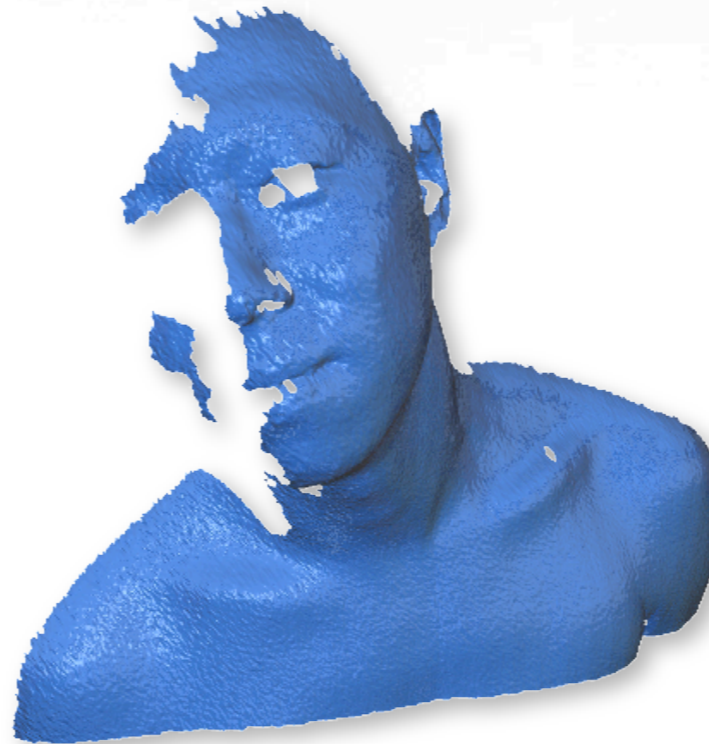
$$E_{\text{plate}} = \int_{\Omega} \|\mathbf{x}_{uu}\|^2 + 2\|\mathbf{x}_{uv}\| + \|\mathbf{x}_{vv}\|^2 \mathrm{d}u\mathrm{d}v$$



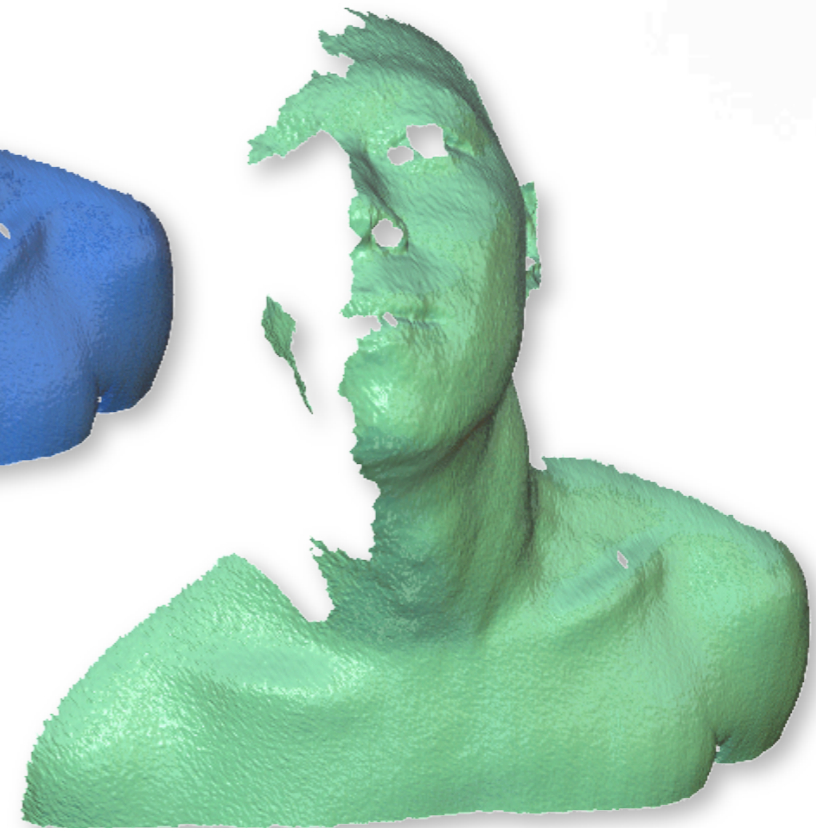
Embedded Deformation

[Sumner et al. '07]

- Efficiency
- Generality
- Natural Deformations
- Detail Preservation



Maximize local rigidity



Embedded Deformation

[Sumner et al. '07]

source

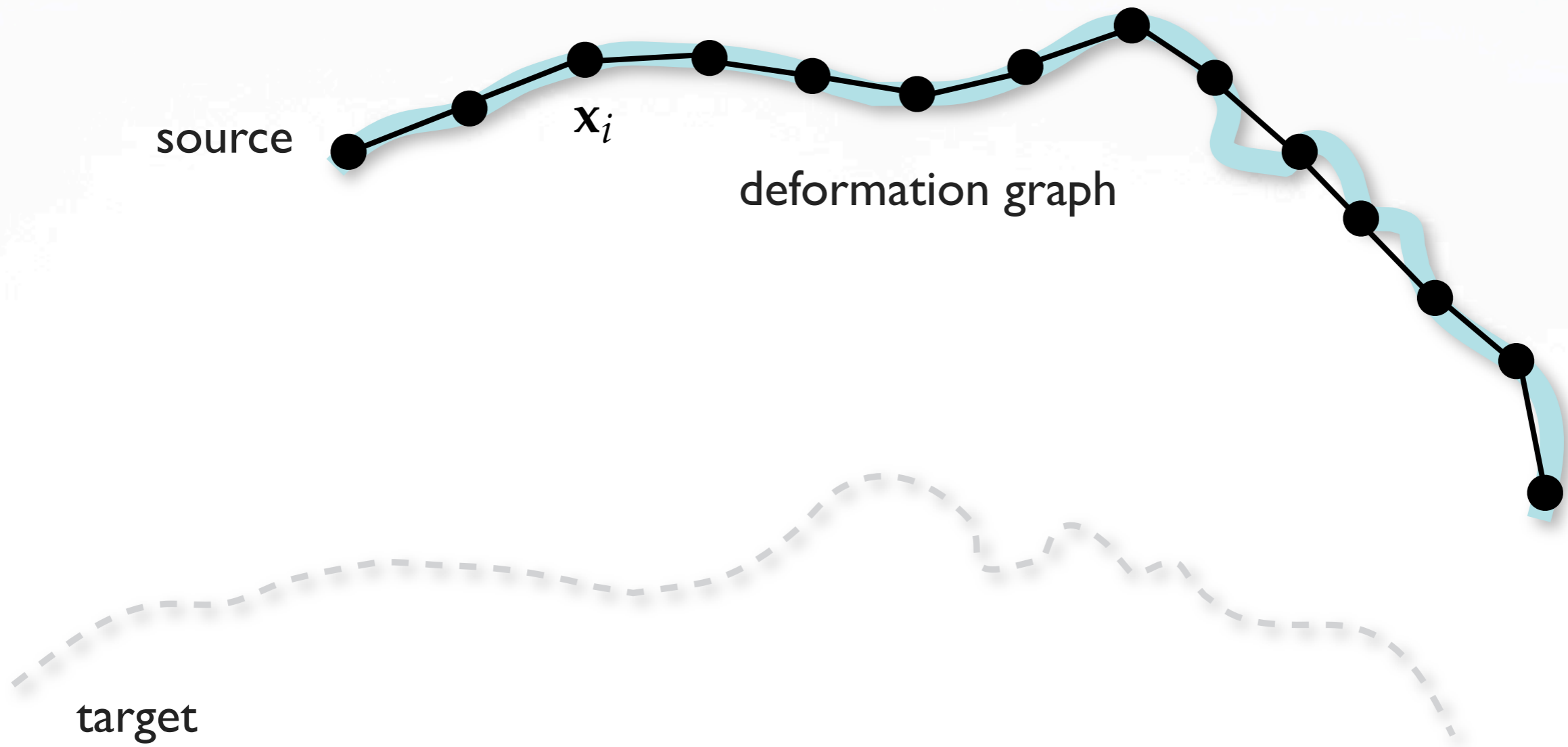


target



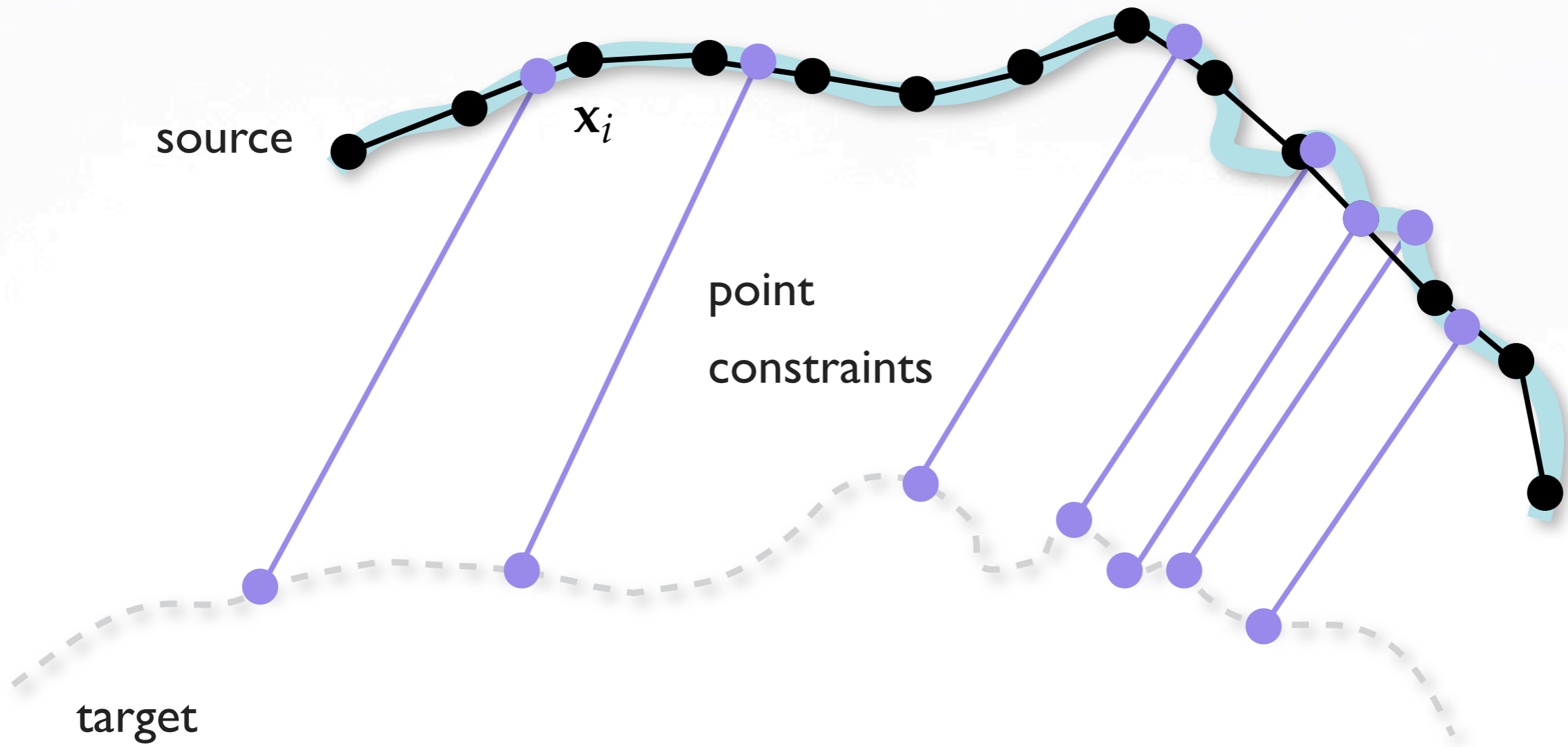
Embedded Deformation

[Sumner et al. '07]



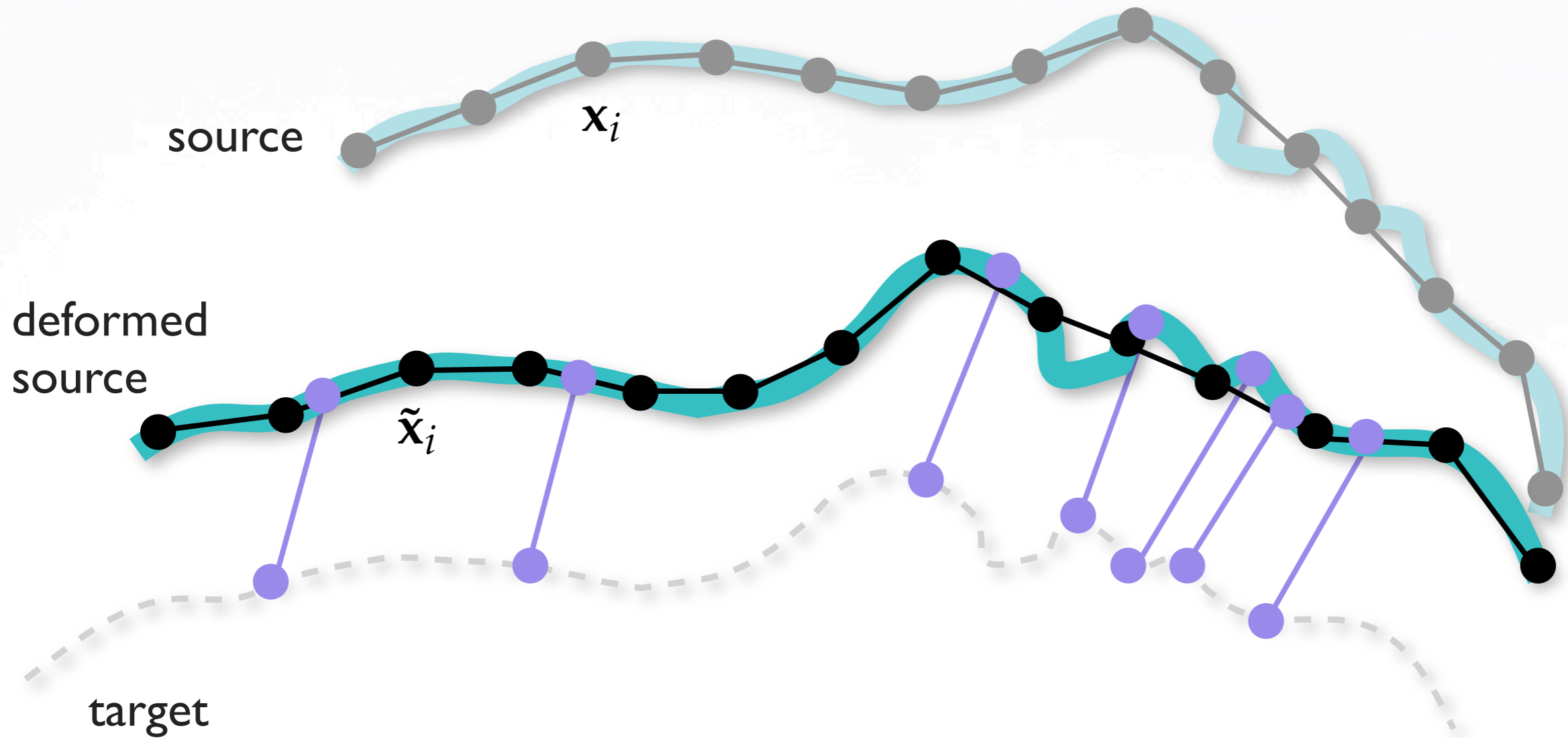
Embedded Deformation

[Sumner et al. '07]



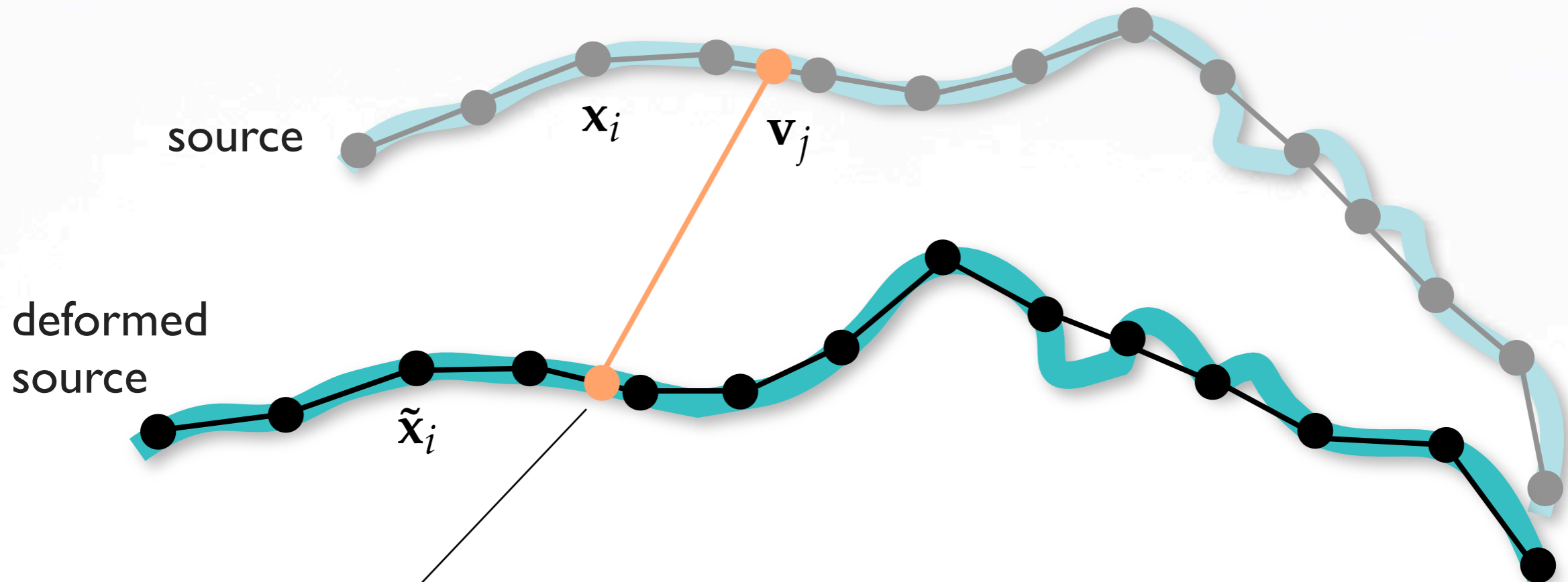
Embedded Deformation

[Sumner et al. '07]



Embedded Deformation

[Sumner et al. '07]



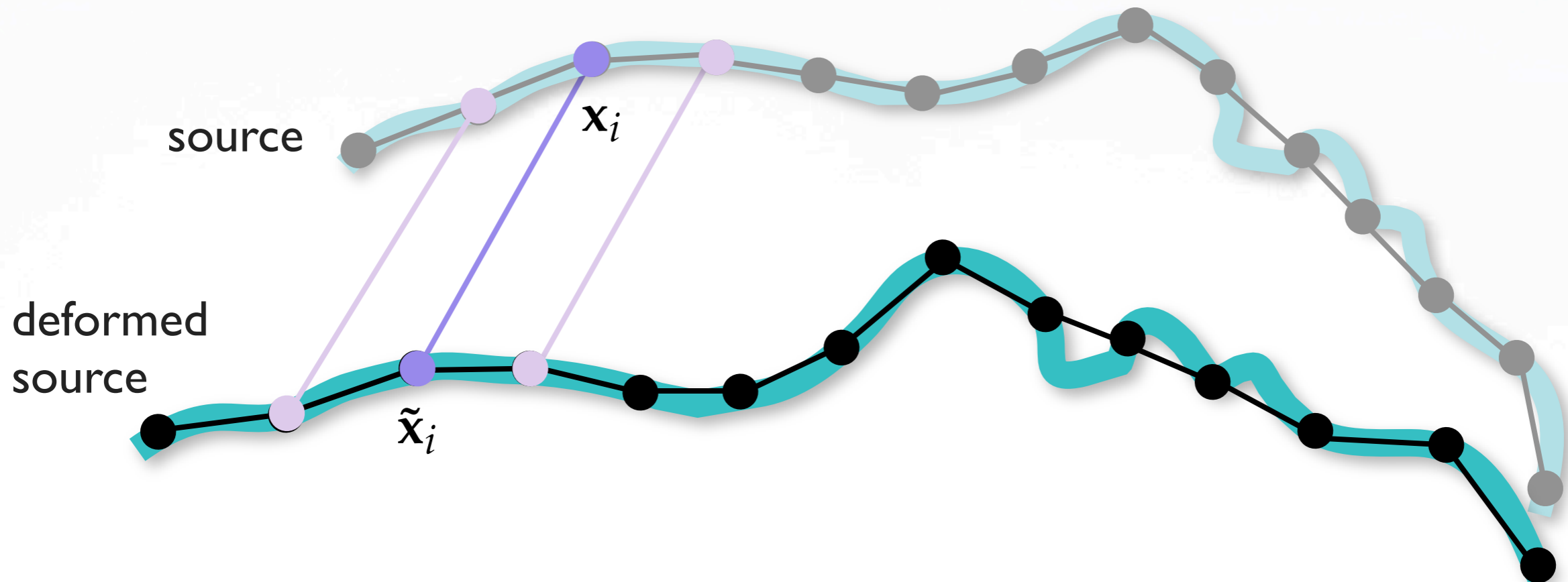
$$\tilde{\mathbf{v}}_j = \Phi_{\text{affine}}(\mathbf{v}_j) = \sum_{i=1}^n w_i(\mathbf{v}_j) [A_i(\mathbf{v}_j - \mathbf{x}_i) + \mathbf{x}_i + \mathbf{b}_i]$$

E_{rigid} measures deviation from rigid motion



Embedded Deformation

[Sumner et al. '07]



E_{smooth} regularizes the deformation locally

E_{rigid} measures deviation from rigid motion

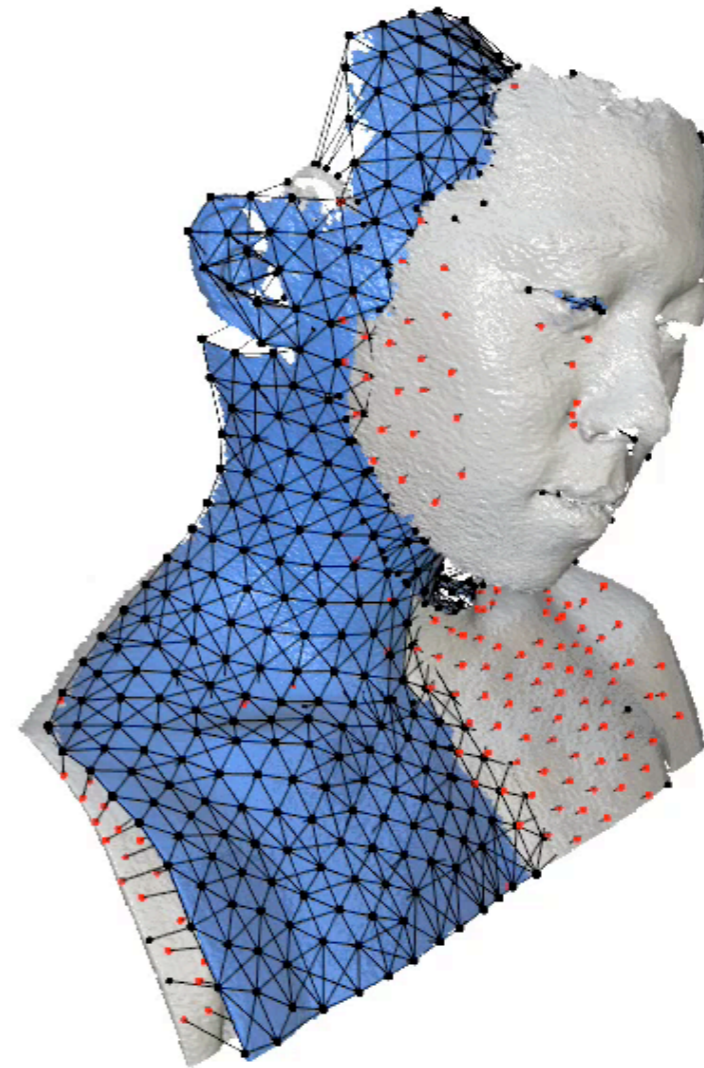
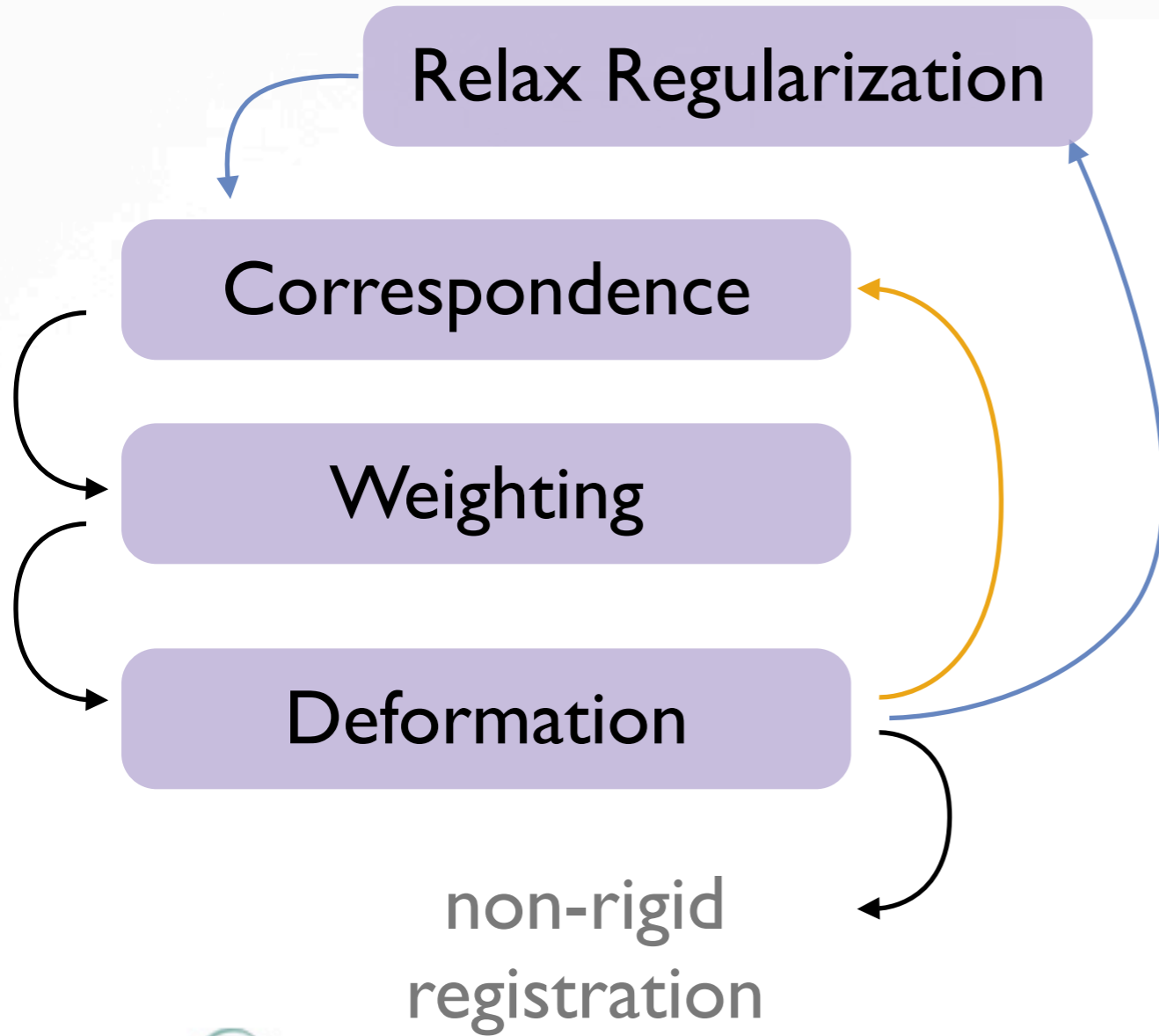


Correspondence Search



Non-Rigid ICP

$$\alpha_{\text{smooth}} \rightarrow 0 \quad \alpha_{\text{rigid}} \rightarrow 0$$

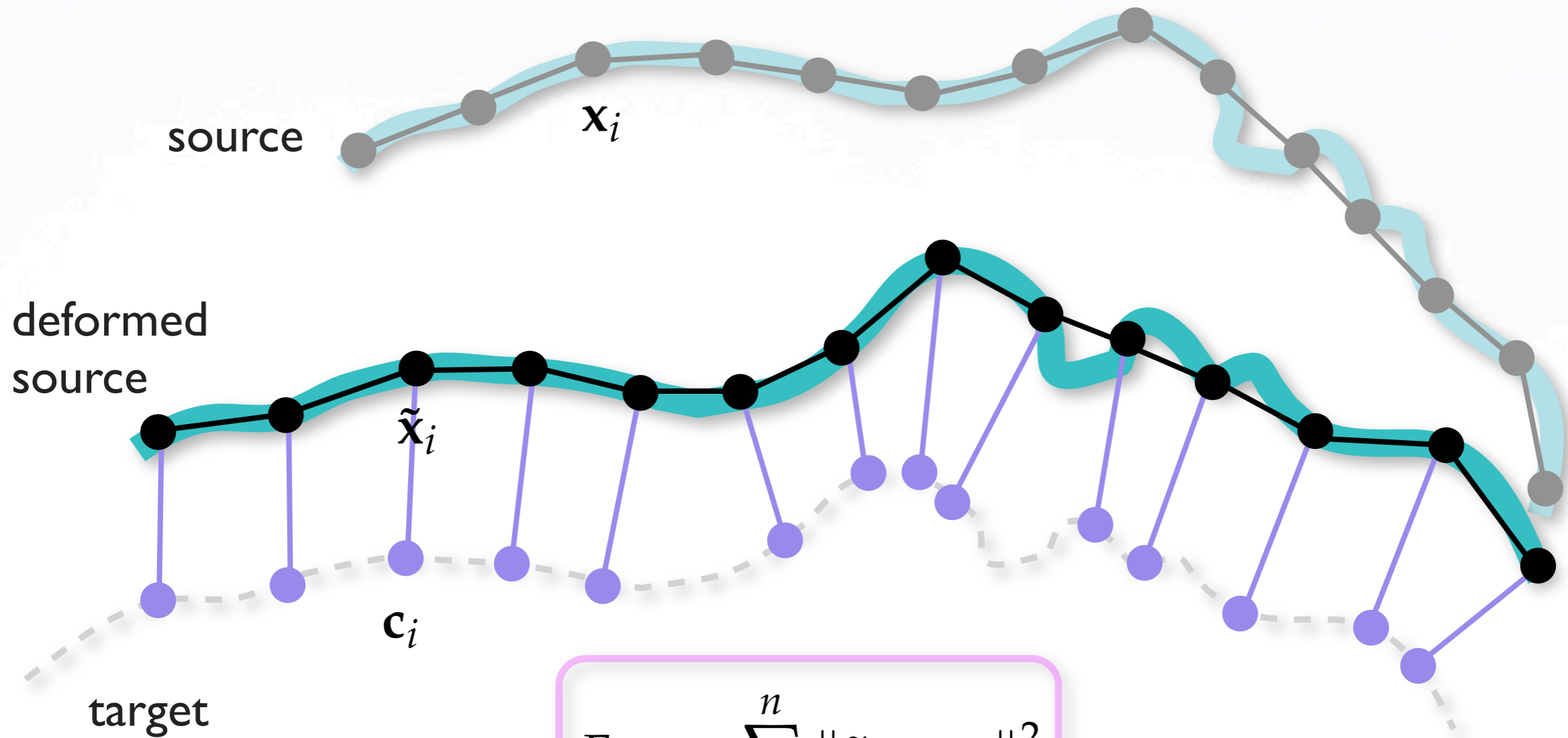


- Example with Embedded Deformation Model



Coupled Optimization

[Li et al. '08]



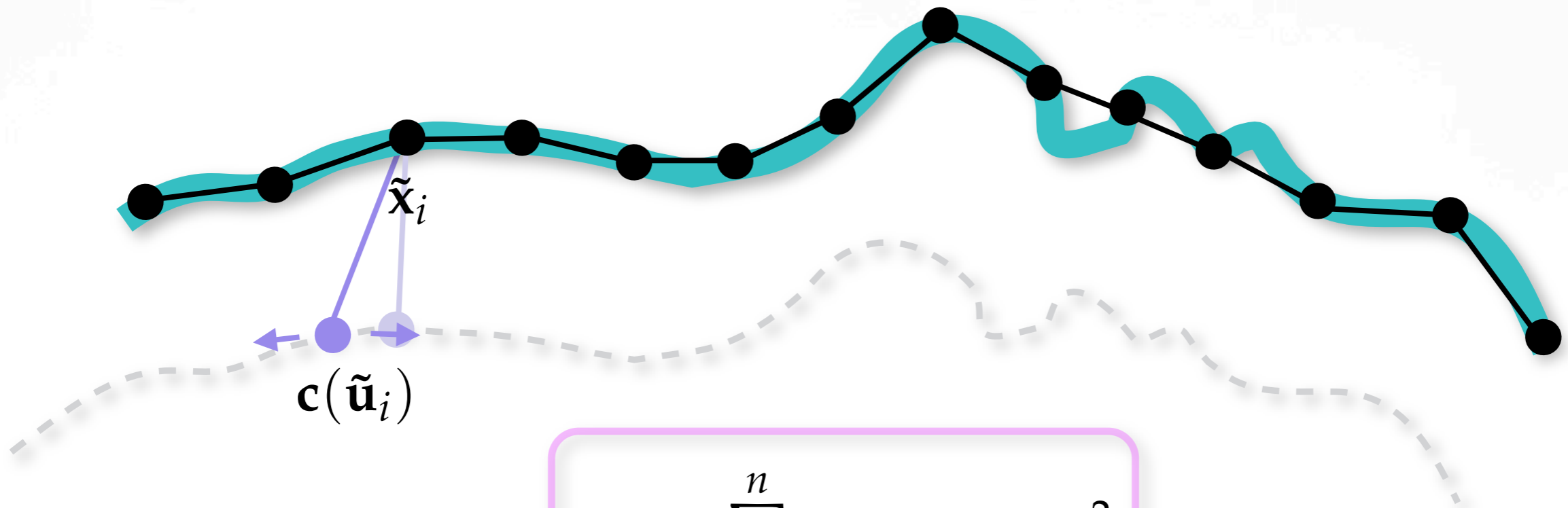
$$E_{\text{fit}} = \sum_{i=1}^n \|\tilde{x}_i - c_i\|_2^2$$



Coupled Optimization

[Li et al. '08]

$\mathbf{c}(\tilde{\mathbf{u}}_i)$
|
 $\tilde{\mathbf{u}}_i = (\tilde{u}_i, \tilde{v}_i)$ optimization variable

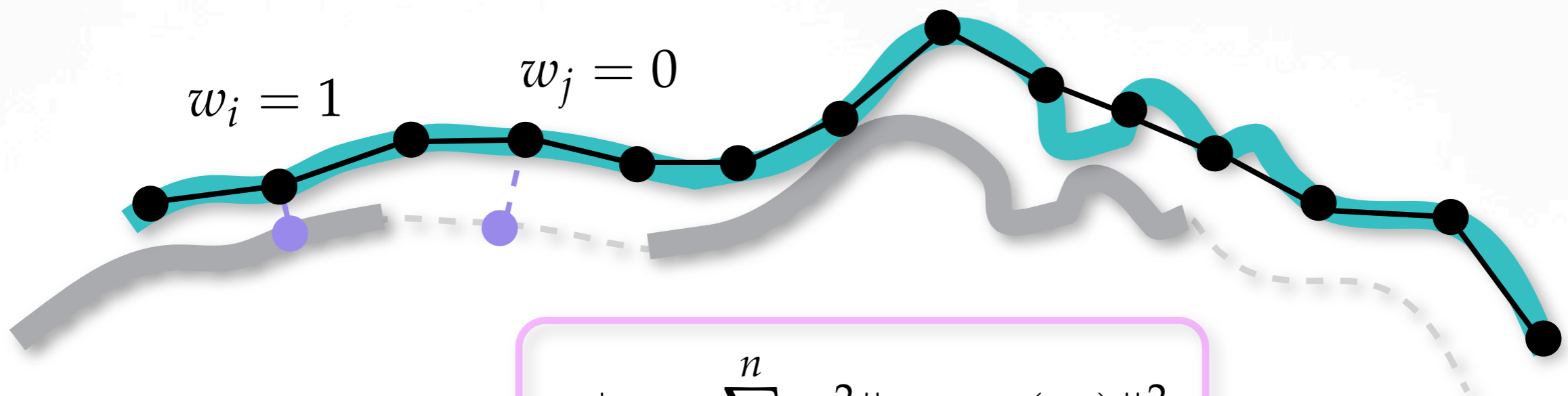


$$E_{\text{fit}} = \sum_{i=1}^n \|\tilde{\mathbf{x}}_i - \mathbf{c}(\tilde{\mathbf{u}}_i)\|_2^2$$



Coupled Optimization

[Li et al. '08]



$$E_{\text{fit}}^* = \sum_{i=0}^n w_i^2 \|\tilde{\mathbf{x}}_i - \tilde{\mathbf{c}}(\mathbf{u}_i)\|_2^2$$



Coupled Optimization

[Li et al. '08]

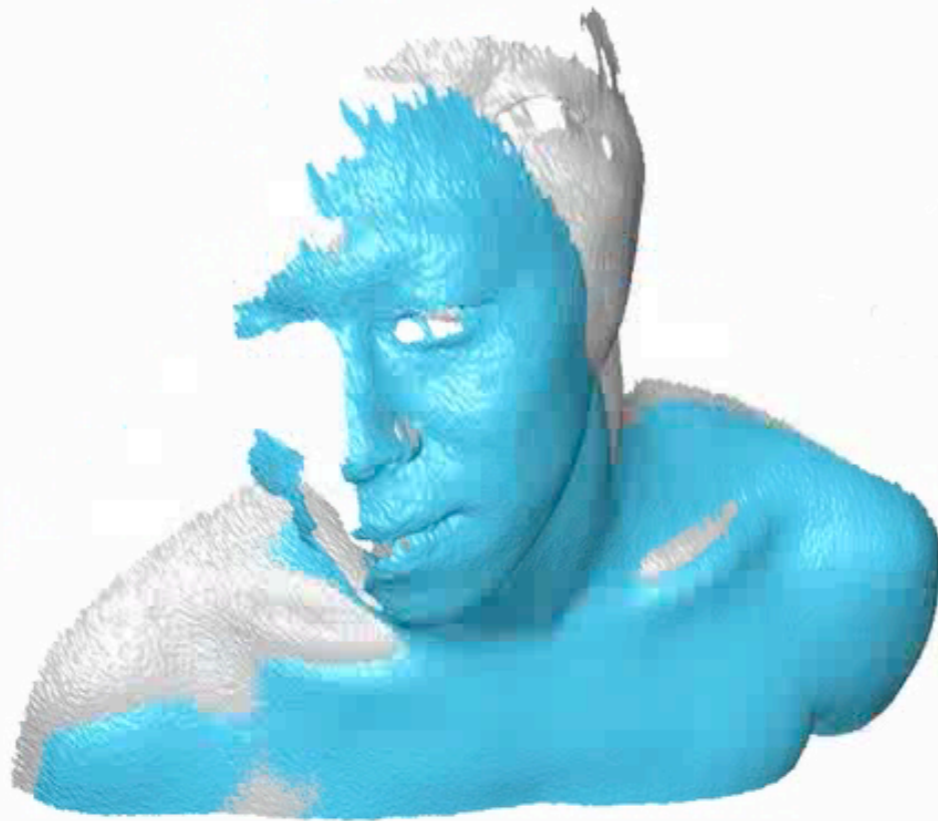
$$E = \alpha_{\text{rigid}} E_{\text{rigid}} + \alpha_{\text{smooth}} E_{\text{smooth}} + \alpha_{\text{fit}} E_{\text{fit}}^* + \alpha_{\text{conf}} E_{\text{conf}}$$

- Minimize deformation energy
- Minimize alignment error
- Maximize regions of overlap

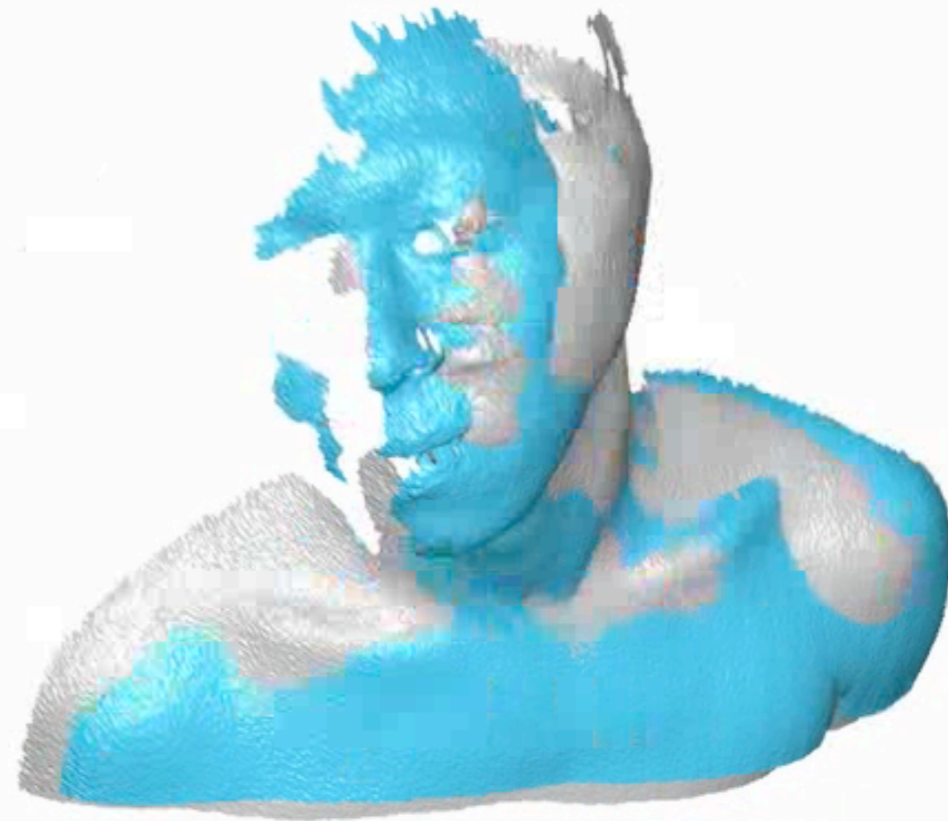


Optimization

219 iterations 2 min 19 s



Initial Alignment



Registration



Energy Term Visualization



E_{conf}



E_{fit}



E_{smooth}



E_{rigid}



In Progress



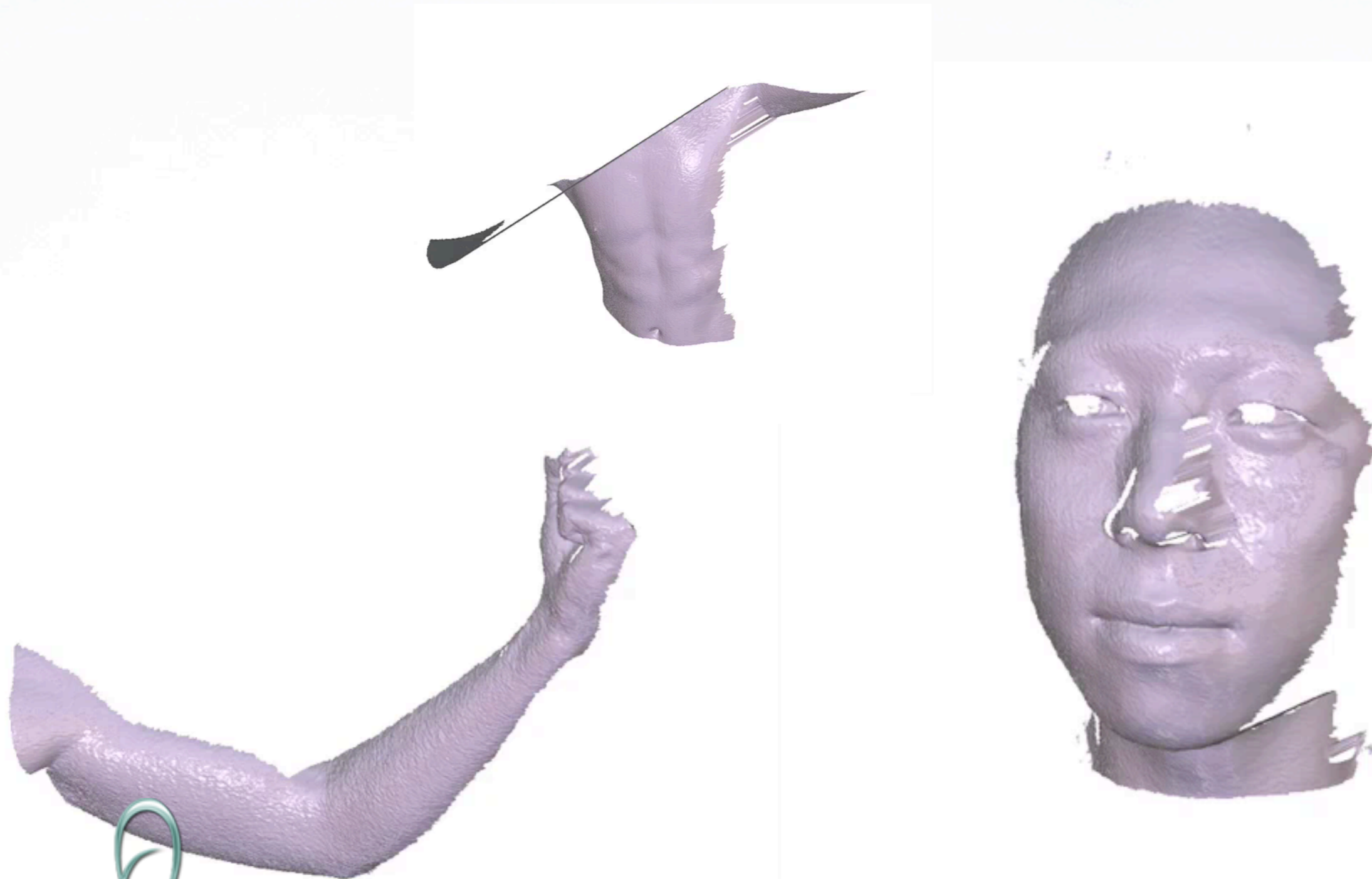
Limitations of Template-based Methods



- Requires template with all details
- Still sparse motion capture
- No distinction of transient and persistent data



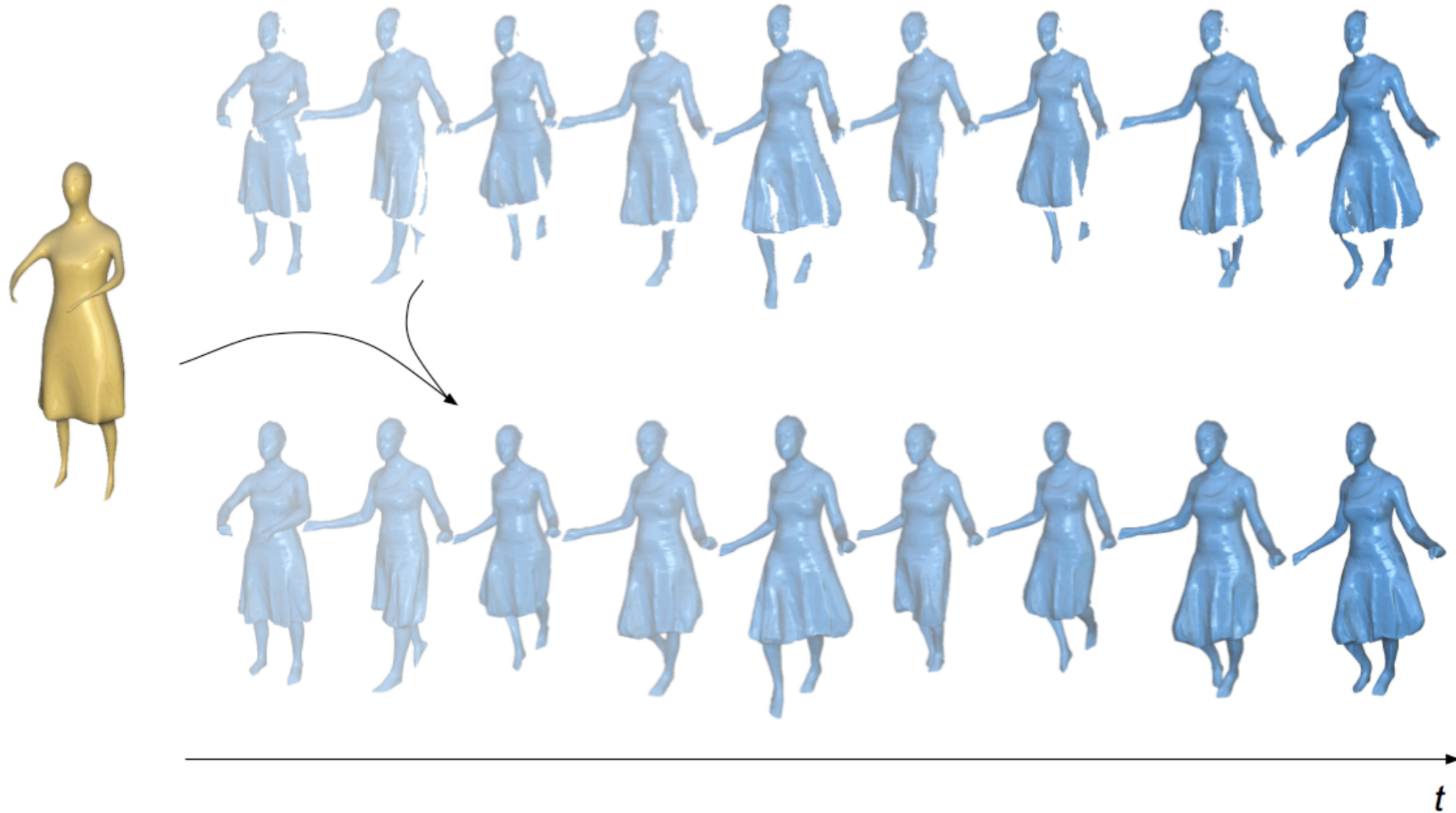
Capturing Deformable Shapes



Data provided with T.Weise and L.Van Gool



Geometry and Motion Reconstruction



data provided by Stanford and MPI Saarbrücken



input data



template fitting



data provided by Stanford and MPI Saarbrücken



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