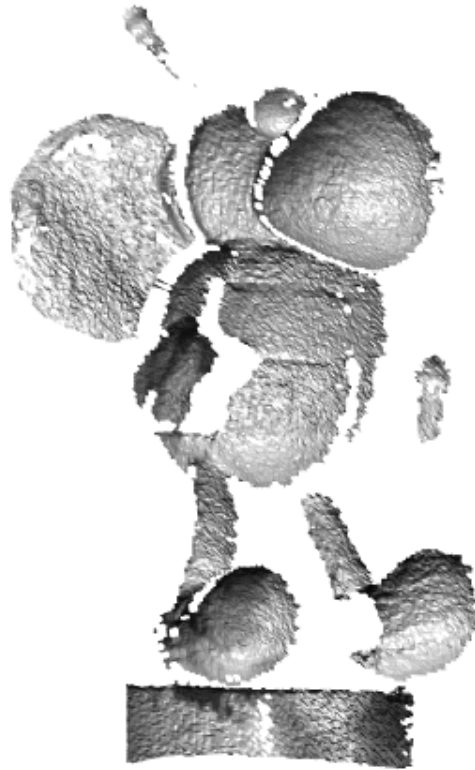


# Geometric Registration for Deformable Shapes

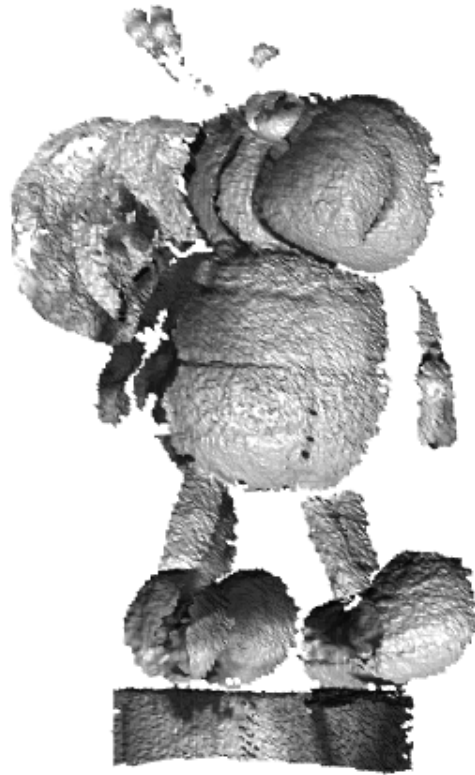
## 4.1 Dynamic Registration



# Scan Registration

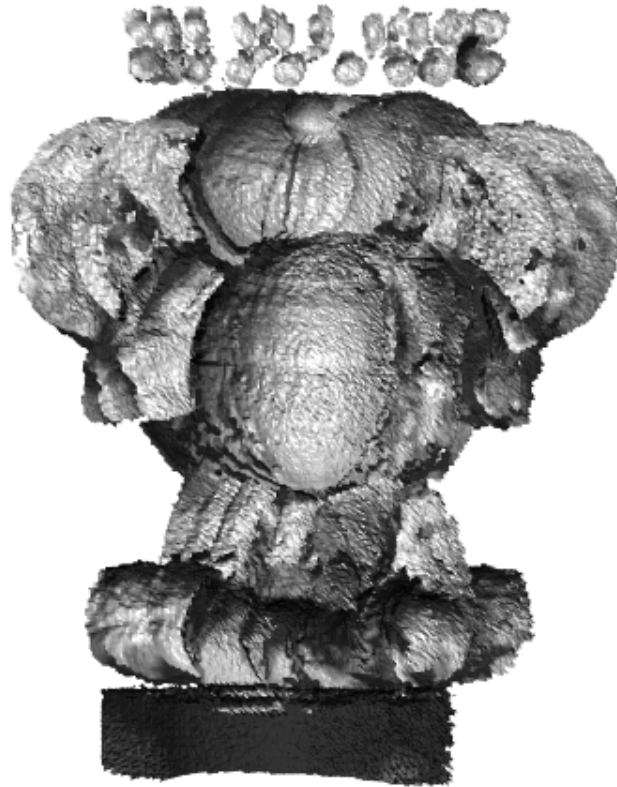


# Scan Registration



Solve for inter-frame  
motion:  $\alpha := (\mathbf{R}, \mathbf{t})$

# Scan Registration



Solve for inter-frame  
motion:  $\alpha_j := (\mathbf{R}_j, \mathbf{t}_j)$

# The Setup

---

Given:

A set of frames  $\{P_0, P_1, \dots, P_n\}$

Goal:

Recover rigid motion  $\{\alpha_1, \alpha_2, \dots, \alpha_n\}$  between adjacent frames

# The Setup

---

**Smoothly varying object motion**

**Unknown correspondence between scans**

**Fast acquisition →  
motion happens between frames**

# Insights

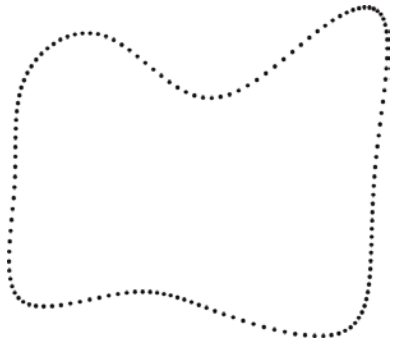
---

**Rigid registration → kinematic property of space-time surface (locally exact)**

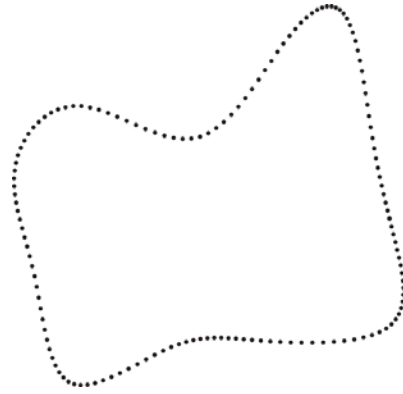
**Registration → surface normal estimation**

**Extension to deformable/articulated bodies**

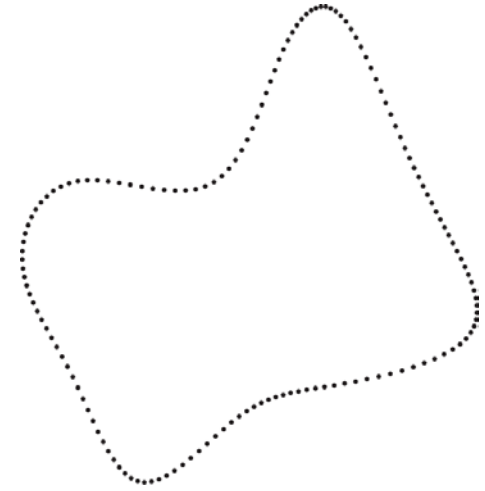
# Time Ordered Scans



$t^j$



$t^{j+1}$

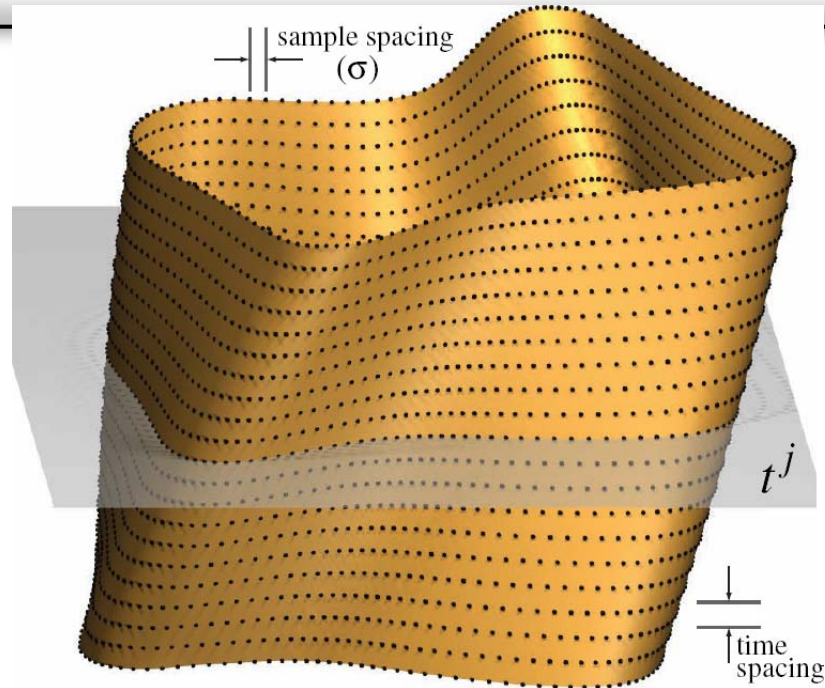
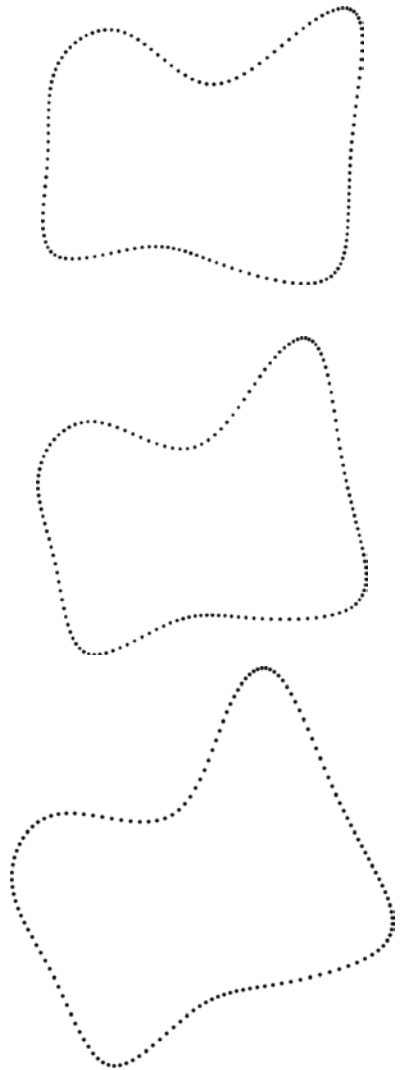


$t^{j+2}$

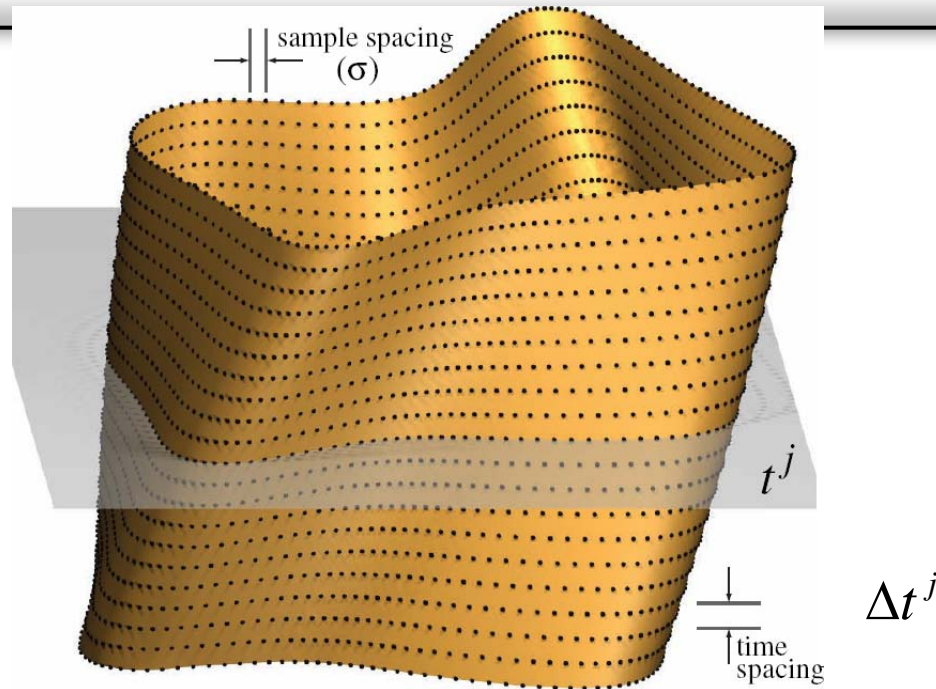
$$\tilde{P}^j \equiv \{\tilde{\mathbf{p}}_i^j\} := \{(\mathbf{p}_i^j, t^j), \mathbf{p}_i^j \in \mathbb{R}^d, t^j \in \mathbb{R}\}$$



# Space-time Surface

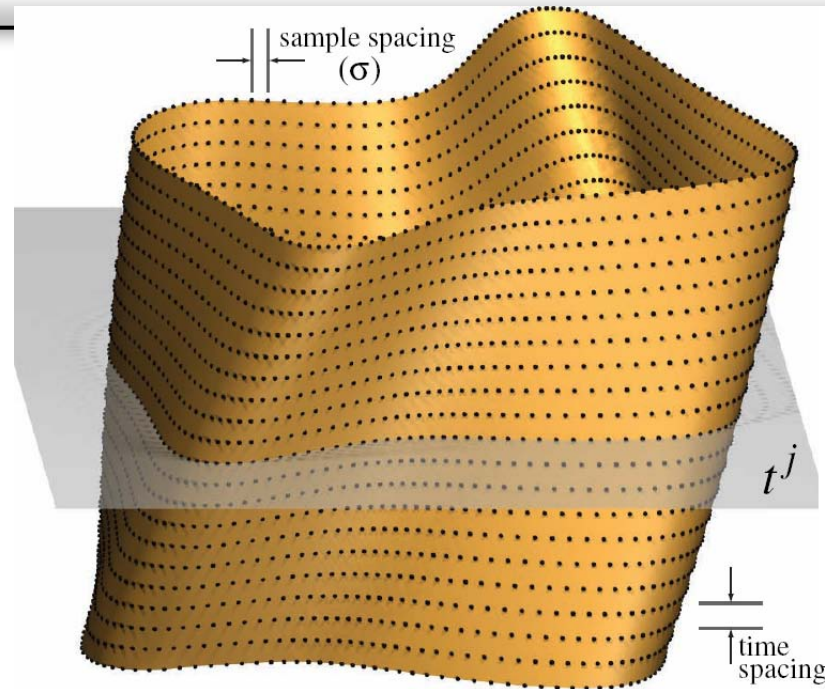


# Space-time Surface



$$\tilde{\mathbf{p}}_i^j \rightarrow \tilde{\alpha}_j(\tilde{\mathbf{p}}_i^j) = \left( \mathbf{R}_j \mathbf{p}_i^j + \mathbf{t}_j, \boxed{t^j + \Delta t^j} \right)$$

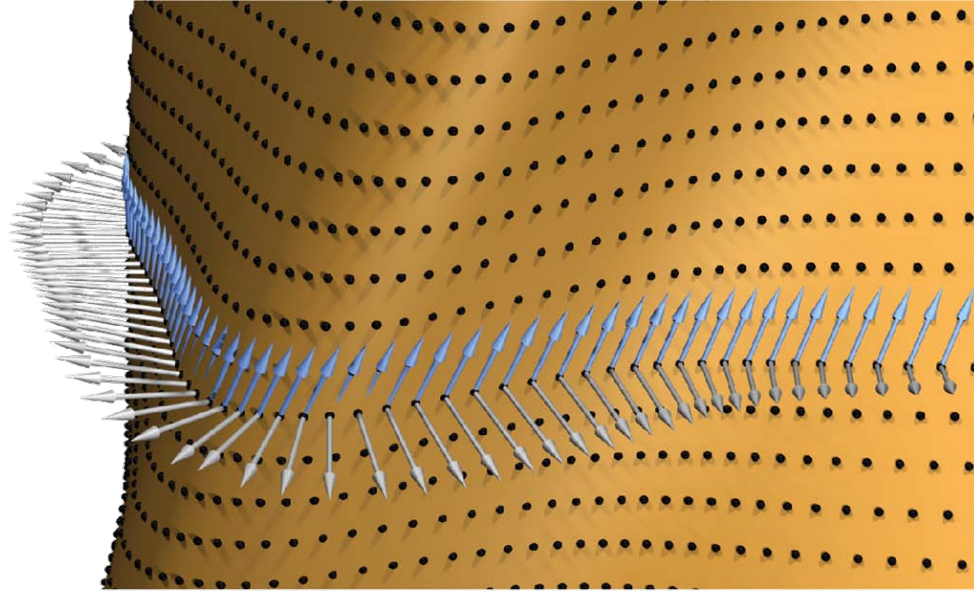
# Space-time Surface



$$\tilde{\mathbf{p}}_i^j \rightarrow \tilde{\alpha}_j(\tilde{\mathbf{p}}_i^j) = \left( \mathbf{R}_j \mathbf{p}_i^j + \mathbf{t}_j, t^j + \Delta t^j \right)$$

$$\tilde{\alpha}_j = \operatorname{argmin} \sum_{i=1}^{|P^j|} d^2(\tilde{\alpha}_j(\tilde{\mathbf{p}}_i^j), S)$$

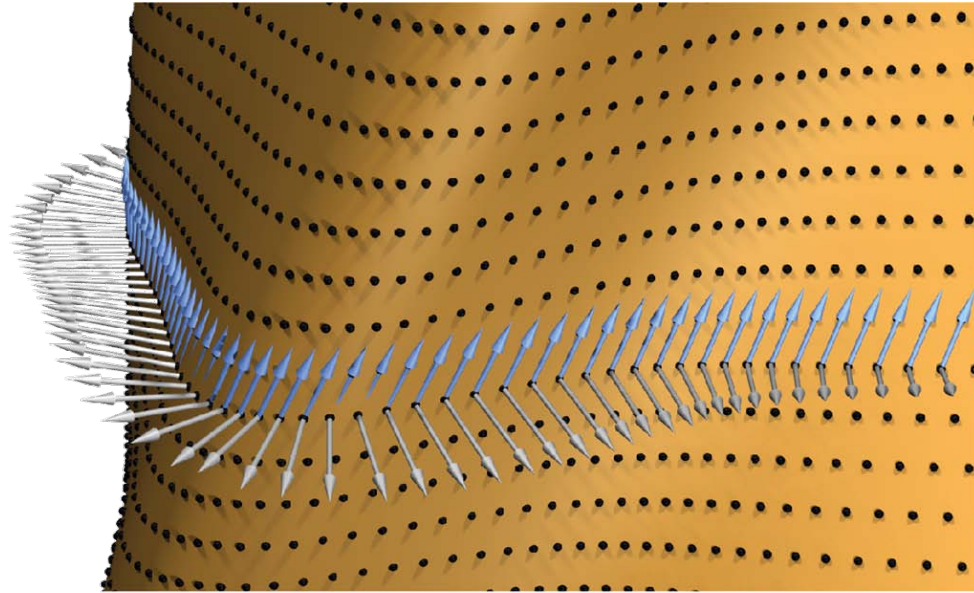
# Spacetime Velocity Vectors



Tangential point movement  $\rightarrow$  velocity vectors orthogonal to surface normals

$$\widetilde{\alpha}_j = \operatorname{argmin} \sum_{i=1}^{|P^j|} d^2(\widetilde{\alpha}_j(\widetilde{\mathbf{p}}_i^j), S)$$

# Spacetime Velocity Vectors



Tangential point movement  $\rightarrow$  velocity vectors orthogonal to surface normals

$$v(\tilde{p}_i) \cdot n(\tilde{p}_i) = 0$$

# Final Steps

(rigid) velocity vectors  $\rightarrow \tilde{\mathbf{v}}(\tilde{\mathbf{p}}_i^j) = (\mathbf{c}_j \times \mathbf{p}_i^j + \bar{\mathbf{c}}_j, 1)$

$$\min_{\mathbf{c}_j, \bar{\mathbf{c}}_j} \sum_{i=1}^{|P^j|} w_i^j \left[ (\mathbf{c}_j \times \mathbf{p}_i^j + \bar{\mathbf{c}}_j, 1) \cdot \tilde{\mathbf{n}}_i^j \right]^2$$

# Final Steps

(rigid) velocity vectors !  $\tilde{\mathbf{v}}(\tilde{\mathbf{p}}_i^j) = (\mathbf{c}_j \times \mathbf{p}_i^j + \bar{\mathbf{c}}_j, 1)$

$$\min_{\mathbf{c}_j, \bar{\mathbf{c}}_j} \sum_{i=1}^{|P^j|} w_i^j \left[ (\mathbf{c}_j \times \mathbf{p}_i^j + \bar{\mathbf{c}}_j, 1) \cdot \tilde{\mathbf{n}}_i^j \right]^2$$

$$\mathbf{A}\mathbf{x} + \mathbf{b} = 0$$

$$\mathbf{A} = \sum_{i=1}^{|P^j|} w_i^j \begin{bmatrix} \tilde{\mathbf{n}}_i^j \\ \mathbf{p}_i^j \times \tilde{\mathbf{n}}_i^j \end{bmatrix} \begin{bmatrix} \tilde{\mathbf{n}}_i^j{}^T & (\mathbf{p}_i^j \times \tilde{\mathbf{n}}_i^j)^T \end{bmatrix}$$

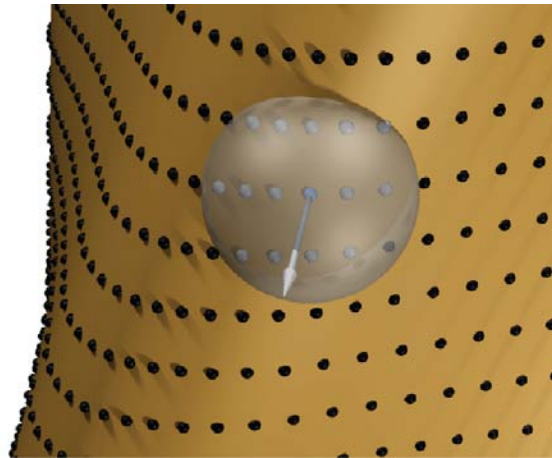
$$\mathbf{b} = \sum_{i=1}^{|P^j|} w_i^j n_i^j \begin{bmatrix} \tilde{\mathbf{n}}_i^j \\ \mathbf{p}_i^j \times \tilde{\mathbf{n}}_i^j \end{bmatrix} \quad \mathbf{x} = \begin{bmatrix} \bar{\mathbf{c}}_j \\ \mathbf{c}_j \end{bmatrix}$$

# Registration Algorithm

1. Compute time coordinate spacing ( $\sigma$ ), and form space-time surface.
2. Compute space time neighborhood using ANN, and locally estimate space-time surface normals.
3. Solve linear system to estimate  $(c_j, \bar{c}_j)$ .
4. Convert velocity vectors to rotation matrix + translation vector using Plücker coordinates and quarternions.

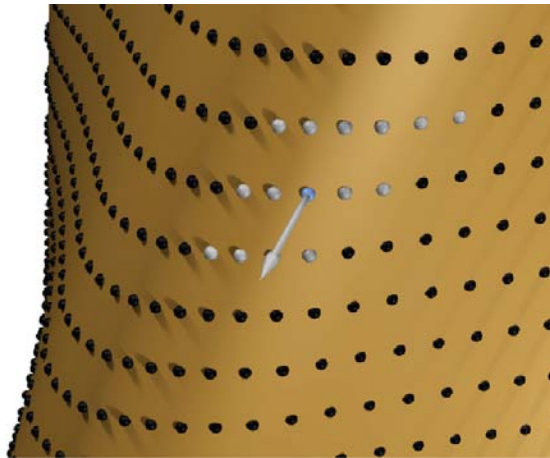


# Normal Estimation: PCA Based



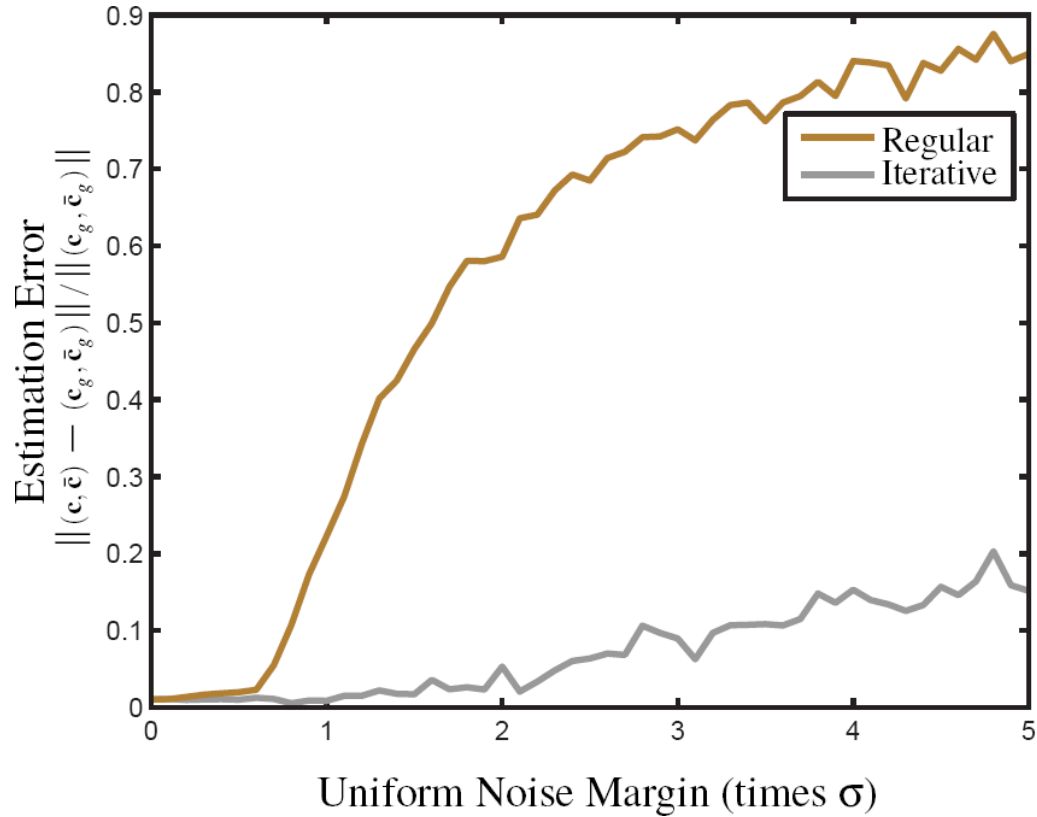
Plane fitting using PCA using chosen neighborhood points.

# Normal Estimation: Iterative Refinement



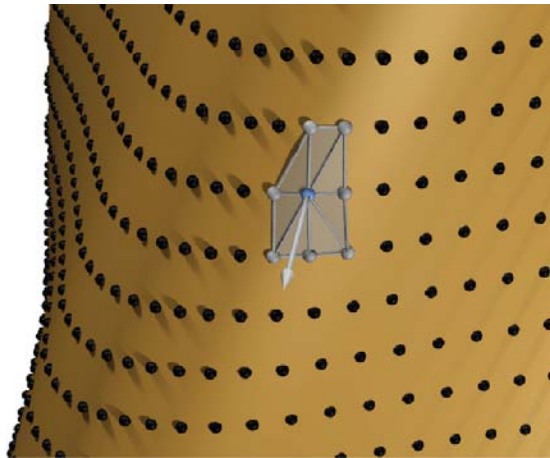
Update neighborhood with current velocity estimate.

# Normal Refinement: Effect of Noise



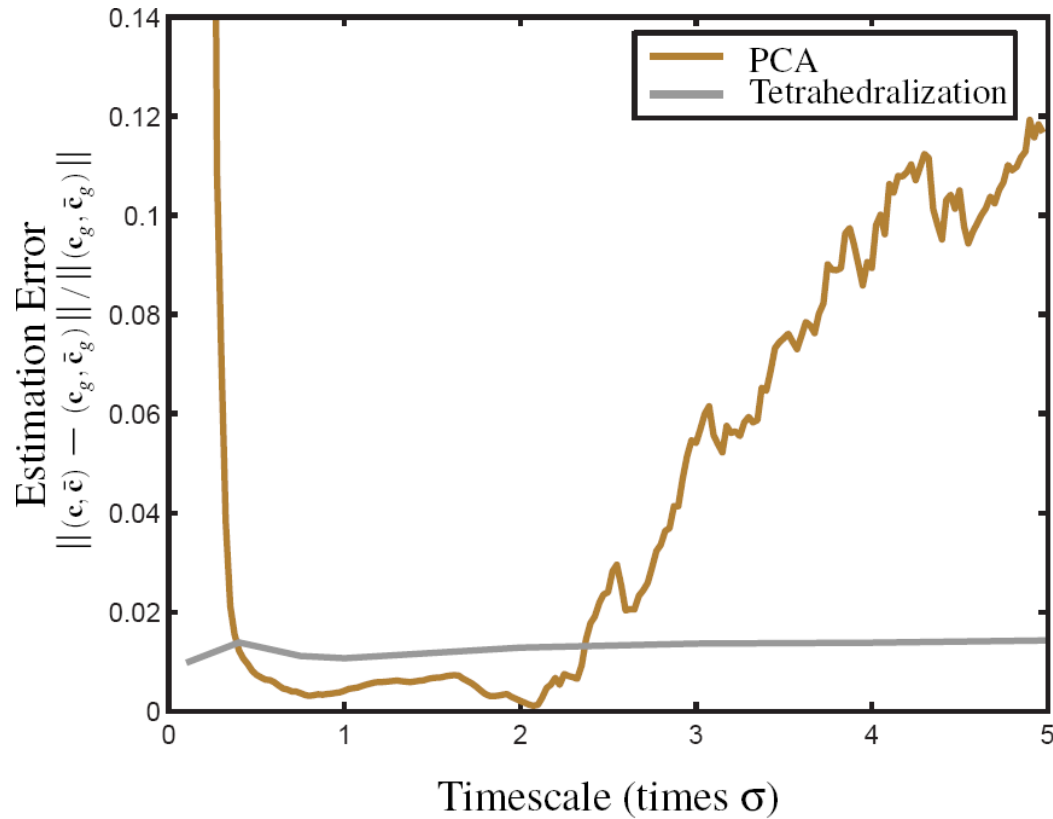
Stable, but more expensive.

# Normal Estimation: Local Triangulation



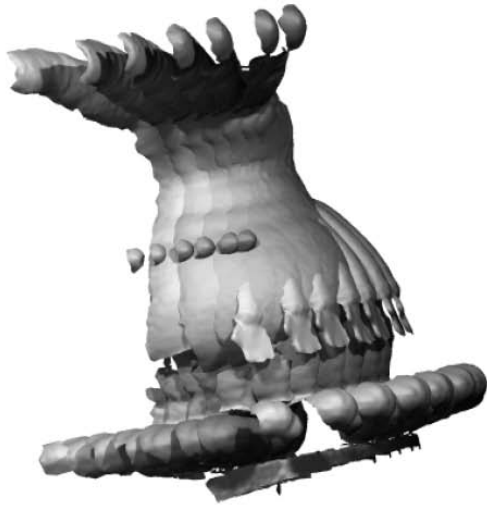
Perform local surface triangulation (tetrahedralization).

# Normal Estimation



Stable, but more expensive.

# Comparison with ICP

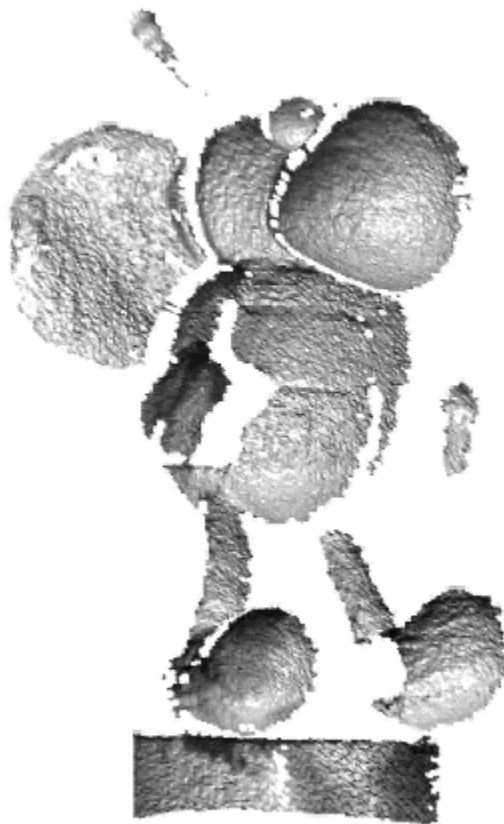


ICP point-plane

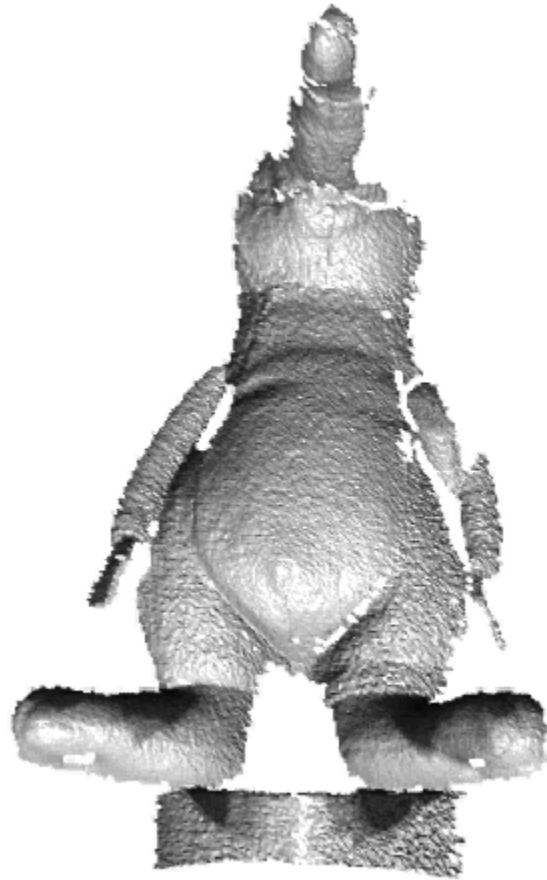


Dynamic registration

# Rigid: Bee Sequence (2,200 frames)

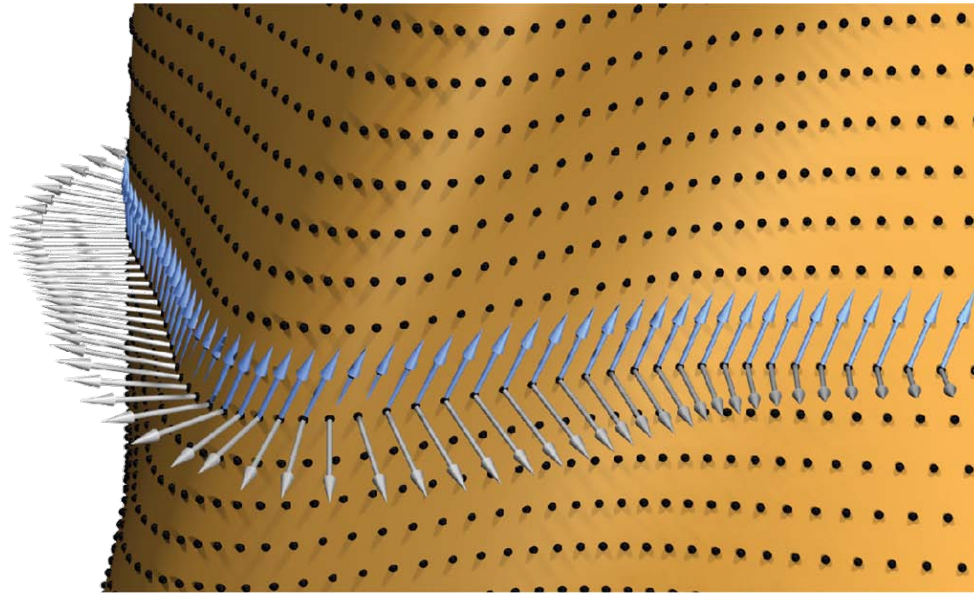


# Rigid: Coati Sequence (2,200 frames)



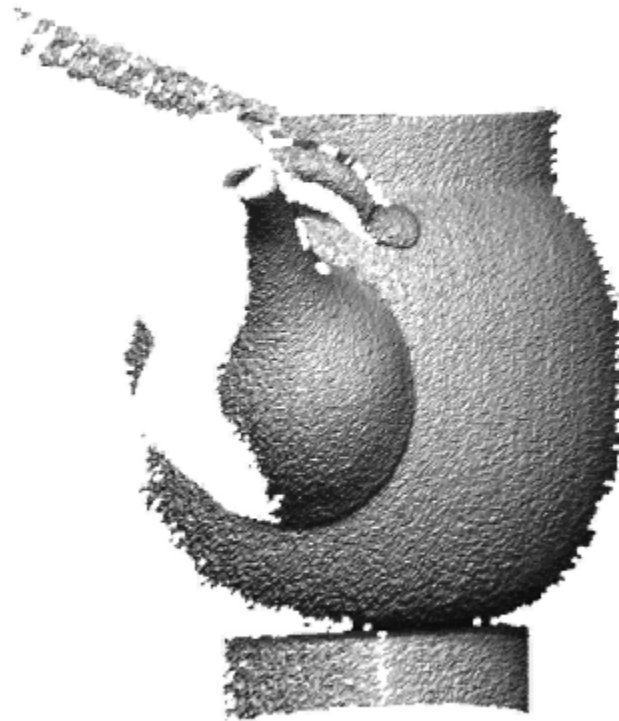


# Handling Large Number of Frames



# Rigid/Deformable: Teapot Sequence

(2,200 frames)



# Deformable Bodies

$$\min_{\mathbf{c}_j, \bar{\mathbf{c}}_j} \sum_{i=1}^{|P^j|} w_i^j \left[ (\mathbf{c}_j \times \mathbf{p}_i^j + \bar{\mathbf{c}}_j, \mathbf{1}) \cdot \tilde{\mathbf{n}}_i^j \right]^2$$

**Cluster points, and solve smaller systems.**

**Propagate solutions with regularization.**

# Deformable: Hand (100 frames)

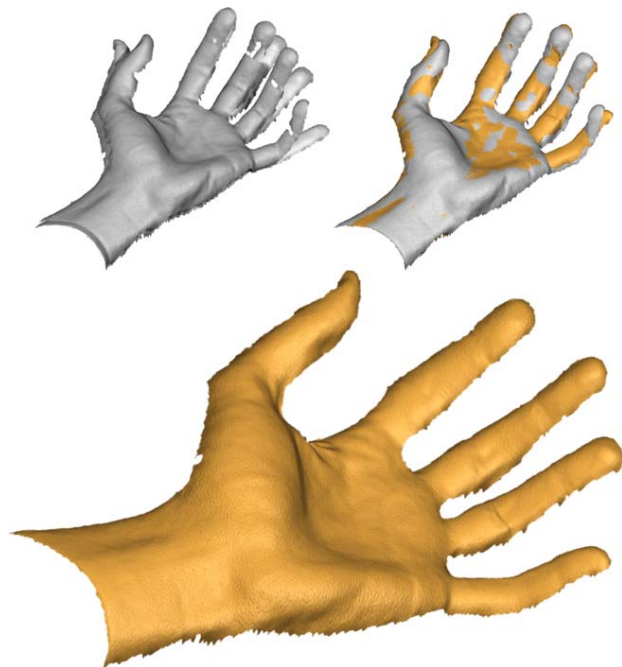


input frames

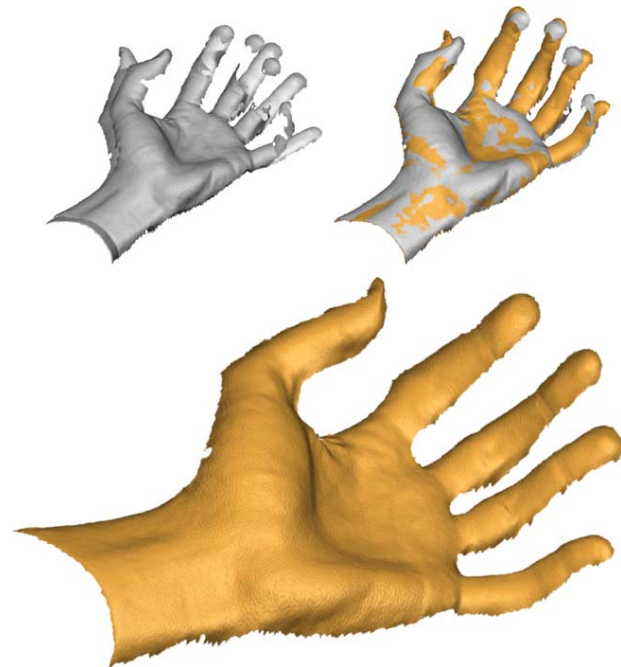


registered result

# Deformable: Hand (100 frames)



scan #1 : scan #50

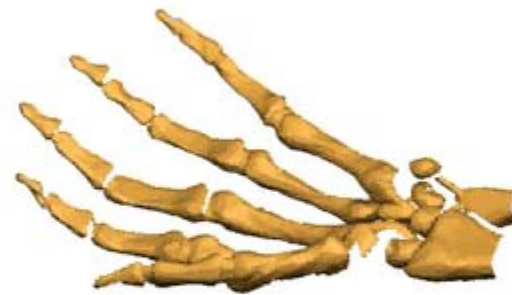


scan #1 : scan #100

# Deformation + scanner motion: Skeleton (100 frames)

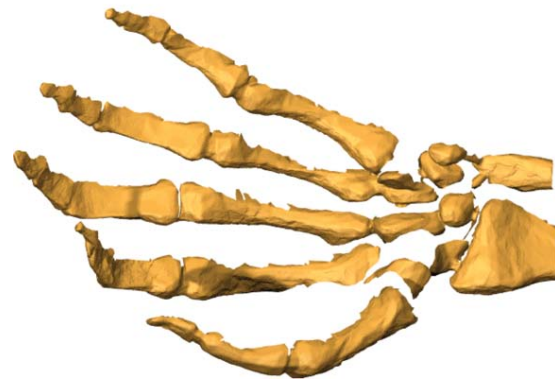
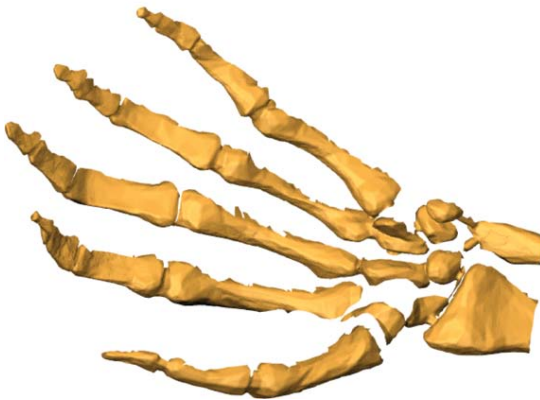
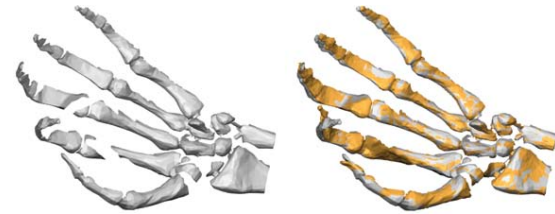
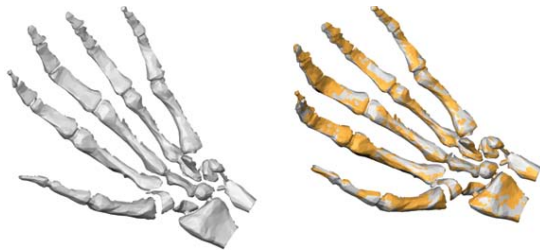


input frames



registered result

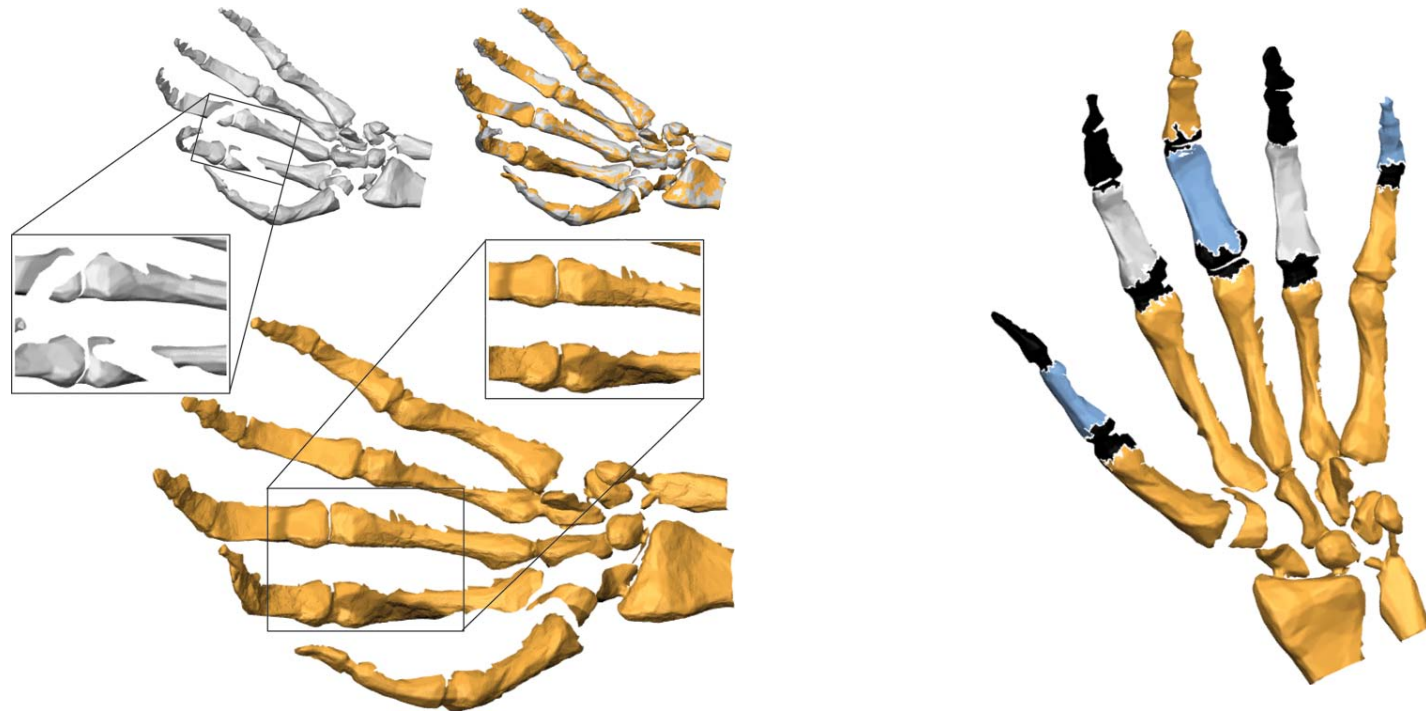
# Deformation + scanner motion: Skeleton (100 frames)



scan #1 : scan #50

scan #1 : scan #100

# Deformation + scanner motion: Skeleton (100 frames)



rigid components



# Performance (on 2.4GHz Athlon Dual Core, 2GB RAM)

Model	# scans	# points/scan (in 1000s)	Time (mins)
bunny (simulated)	314	33.8	13
bee	2,200	20.7	51
coati	2,200	28.1	71
teapot (rigid)	2,200	27.2	68
skeleton (simulated)	100	55.9	11
hand	100	40.1	17

# Conclusion

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**Simple algorithm using kinematic properties of space-time surface.**

**Easy modification for deformable bodies.**

**Suitable for use with fast scanners.**

# Limitations

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**Need more scans, dense scans, ...**

**Sampling condition → time and space**



thank you

