



Motion Capture with Linear Blend Skinning (LBS)

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About me

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学习及工作经历：

2017.09 - 现在： 东南大学自动化学院，副教授
2014.07 - 2017.08： 微软亚洲研究院，副研究员
2012.10 - 2013.03： 瑞士洛桑联邦理工大学，访问学者
2009.09 - 2014.07： 清华大学，博士
2005.09 - 2009.07： 东南大学，学士



Today

I What is Motion Capture?

II Linear Blend Skinning (LBS)

III Motion Capture with LBS

IV Advanced Techniques

V Conclusion



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What is Motion Capture?

“... is a technique of digitally recording 3D human movements for immediate or delayed analysis and playback...”



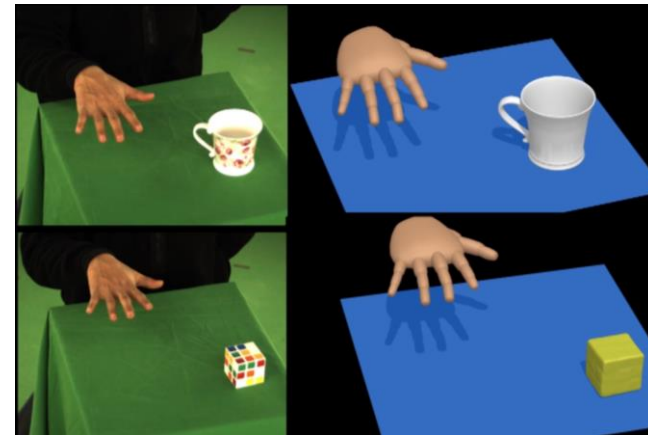
Wikipedia



Full body



Fine detail face



Complex interaction



Applications of Motion Capture



Film industry



Robotics



Biomechanics



Video games



Natural user interaction



Virtual reality



Sensor & Maker System

- Pros
 - accurate
 - real time and robust
- Cons
 - sparse tracker and low spatial resolution
 - expensive hardware
 - cumbersome to wear special suits



Electromechanical system

Hand mocap system

Marker system



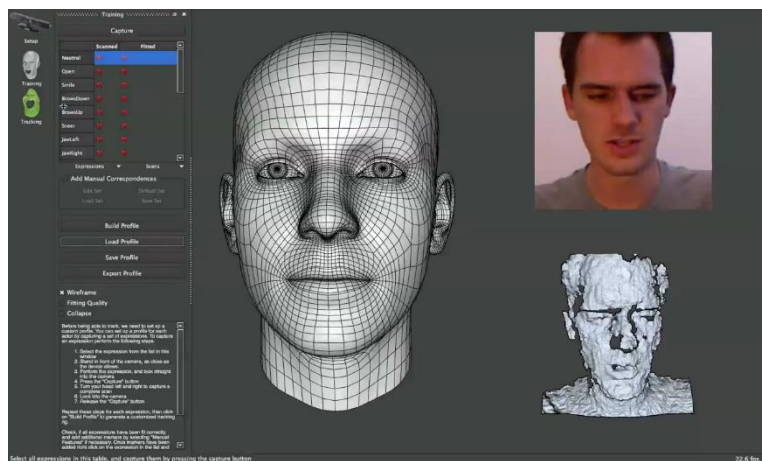
Video-based System

□ Pros

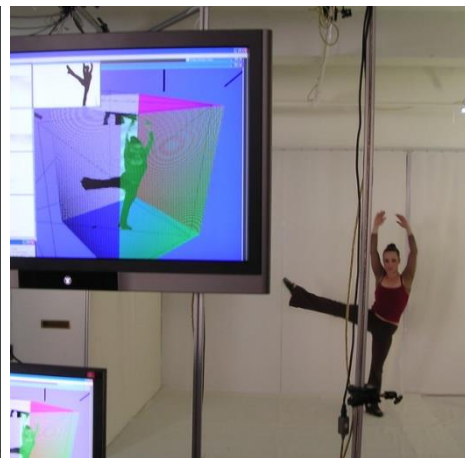
- no markers, no sensors or no special suits
- small and inexpensive hardware configurations
- high spatial resolution

□ Cons

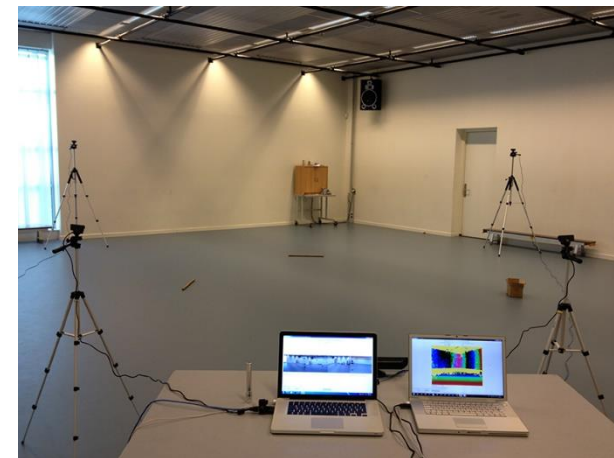
- not as accurate as marker systems
- not as stable as marker and sensor systems



Faceshift®



Organic Motion®

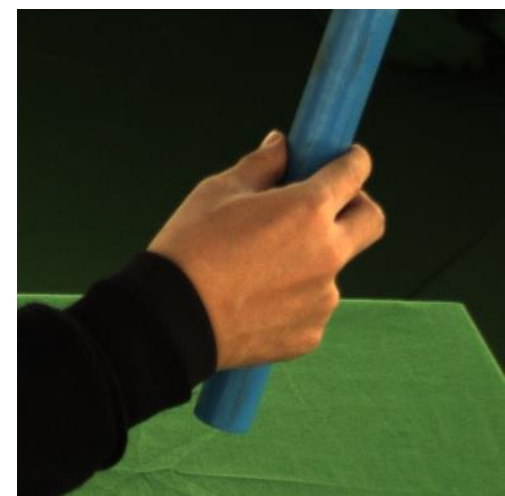


iPiSoft®



Challenges for Video-based System

- ❑ Capture high **spatial** resolution motion
 - few discernible features
 - complex background
- ❑ Difficult and nontrivial **temporal** tracking
 - significant occlusions
 - lighting changes



Few features



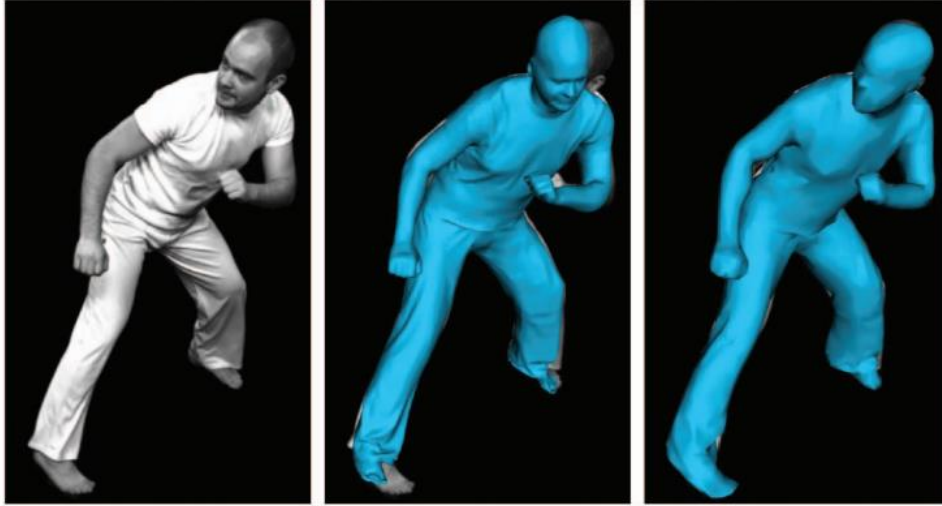
Occlusions



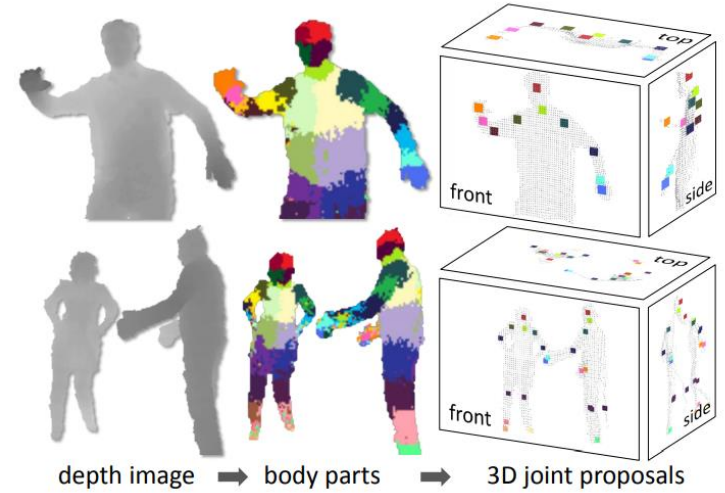
nontrivial temporal tracking



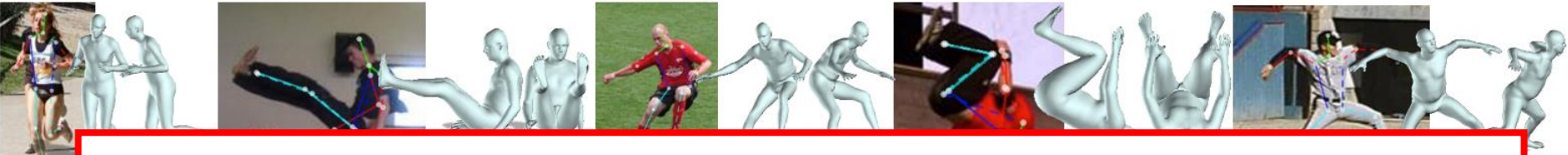
Main Solutions



Generative ✓ Method



Discriminative Method



How to perform *model deformation*?



Hybrid ✓ Method



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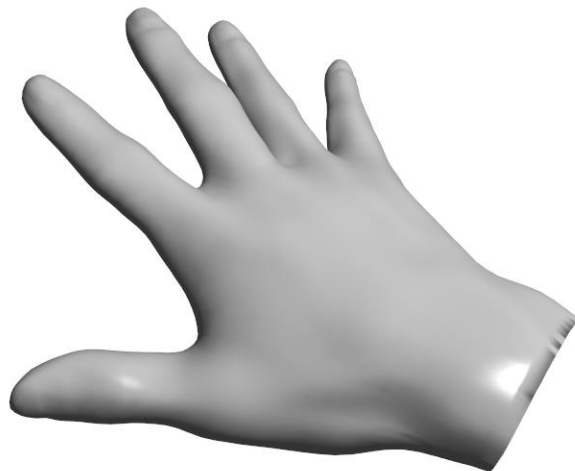
IV Advanced Techniques

V Conclusion

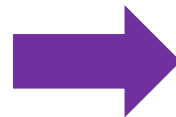


Articulate Model Deformation

- Consider the following problem
 - we want to deform the mesh model M into mesh model M'
 - implicit skeleton structure
- Articulate shape deformation is very common in real life



M



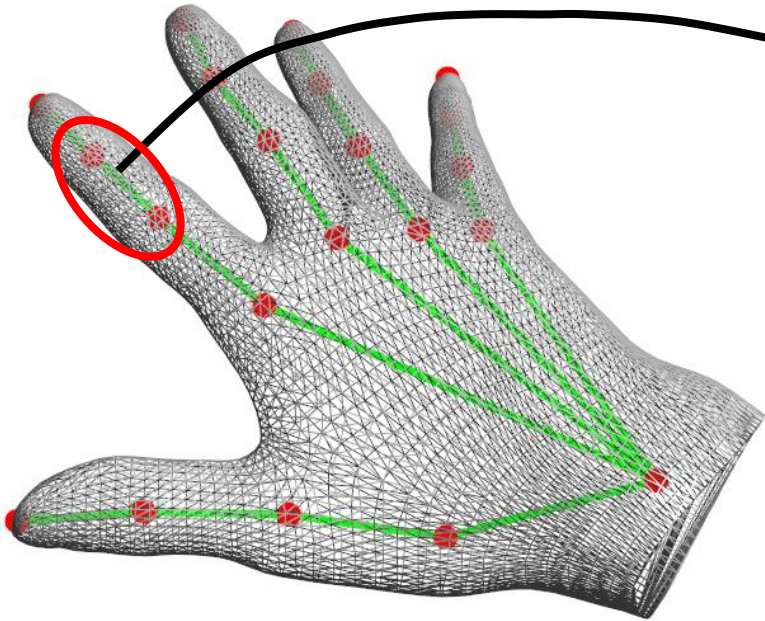
M'



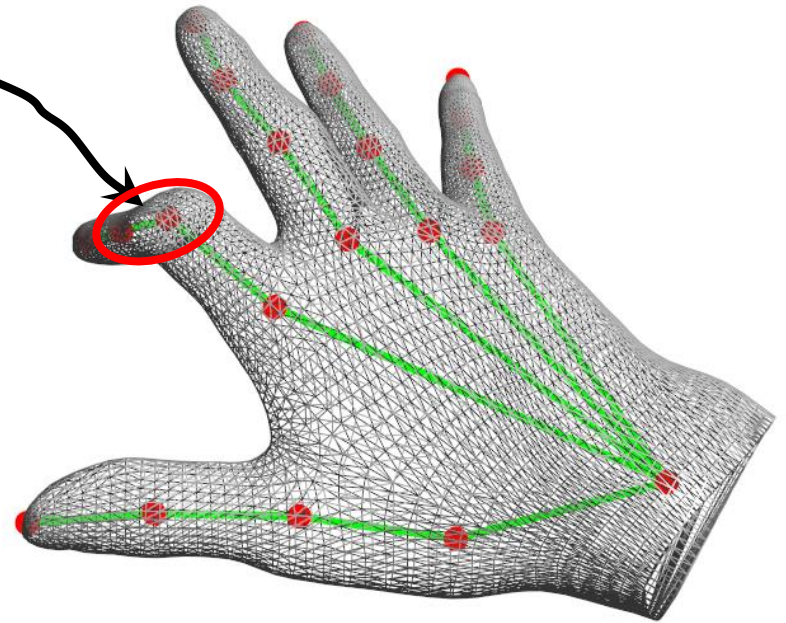
Basic Concepts

- Skinning transformations

T



Rest pose M

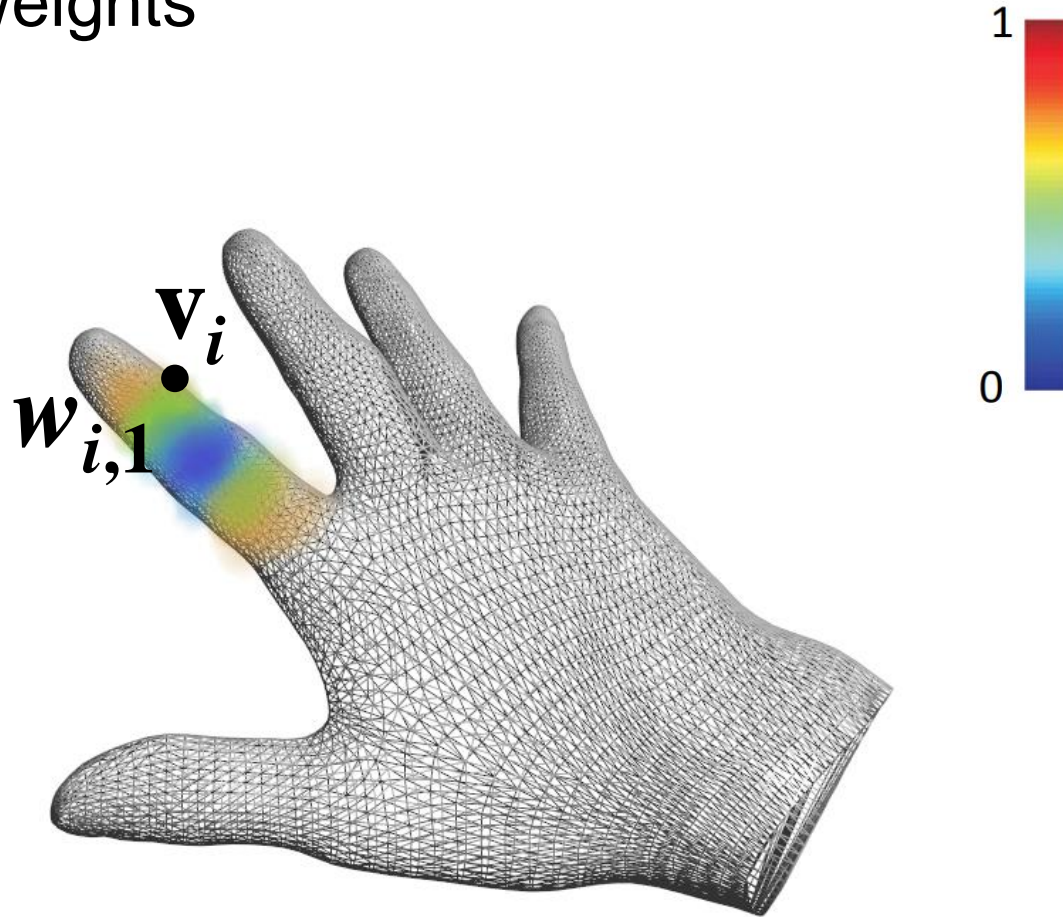


Target pose M'



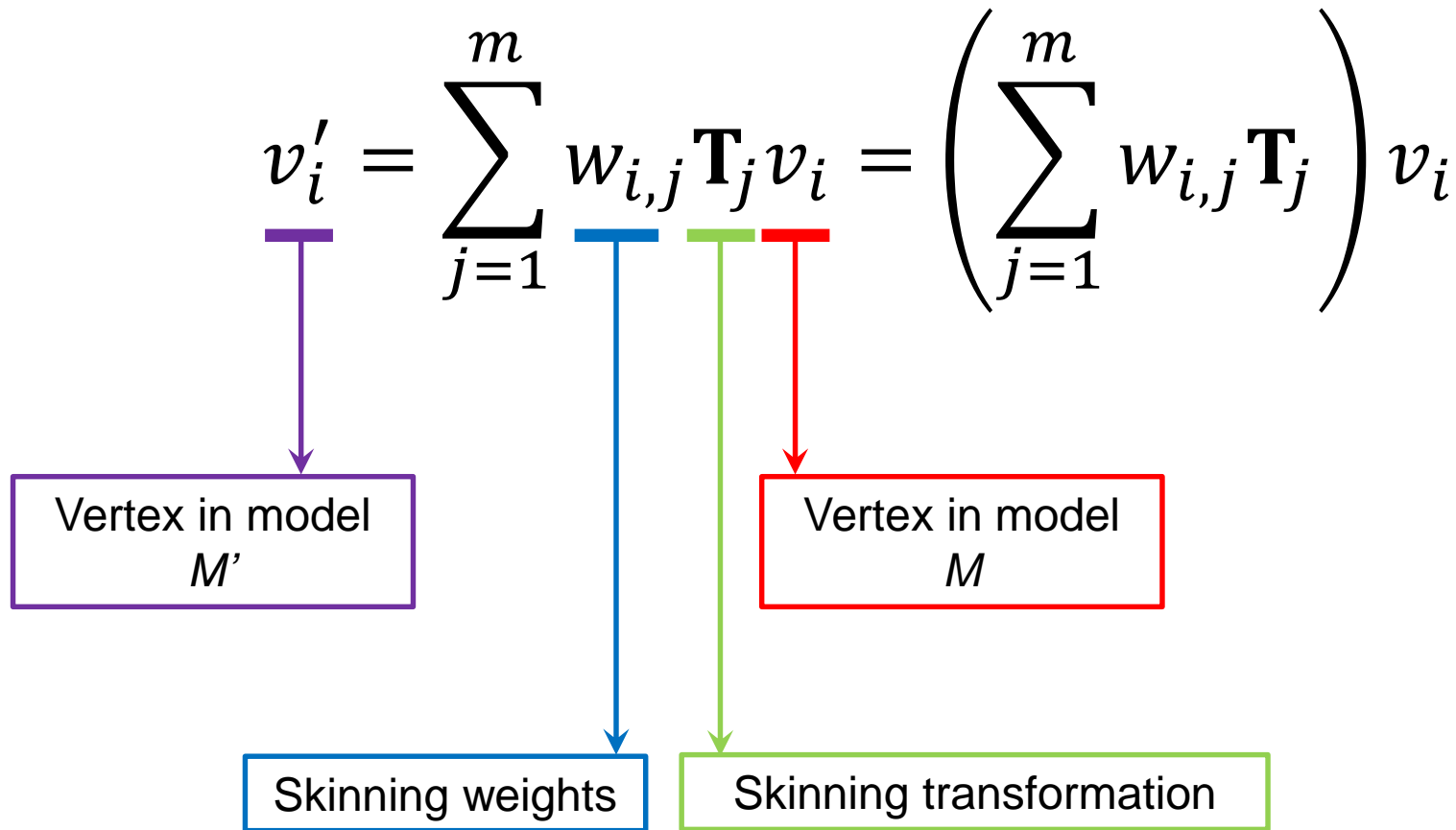
Basic Concepts

□ Skinning weights





Linear Blend Skinning (LBS)





Linear Blend Skinning (LBS)

$$\mathbf{v}'_i = \sum_{j=1}^m w_{i,j} \mathbf{T}_j \mathbf{v}_i = \left(\sum_{j=1}^m w_{i,j} \mathbf{T}_j \right) \mathbf{v}_i$$

- How to compute the Skinning weights $w_{i,j}$?
 - “Automatic Rigging and Animation of 3D Characters”, SIGGRAPH 2007
<http://www.mit.edu/~ibaran/autorig/pinocchio.html>
 - Key Idea
 - heat equilibrium to find the weights

$$-\Delta w_j + H w_j = H p_j$$





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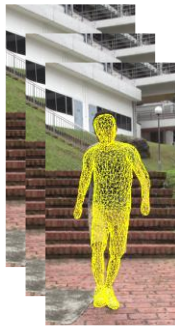
V Conclusion



Motion Capture with LBS

Search optimal skinning transformations T to drive the articulate mesh model with $lbs(T)$ to best match the observed image data O

$$\min_T E(lbs(T), O)$$



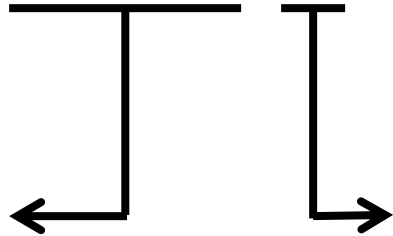
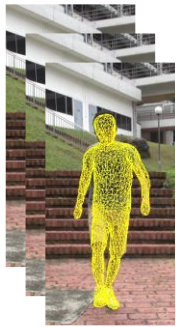
- How to perform the skinning transformations T to drive the articulate mesh model M ?
- How to measure the inconsistency between the deformed model and observed image data?
- How to find an optimal skinning transformations?



Motion Capture with LBS

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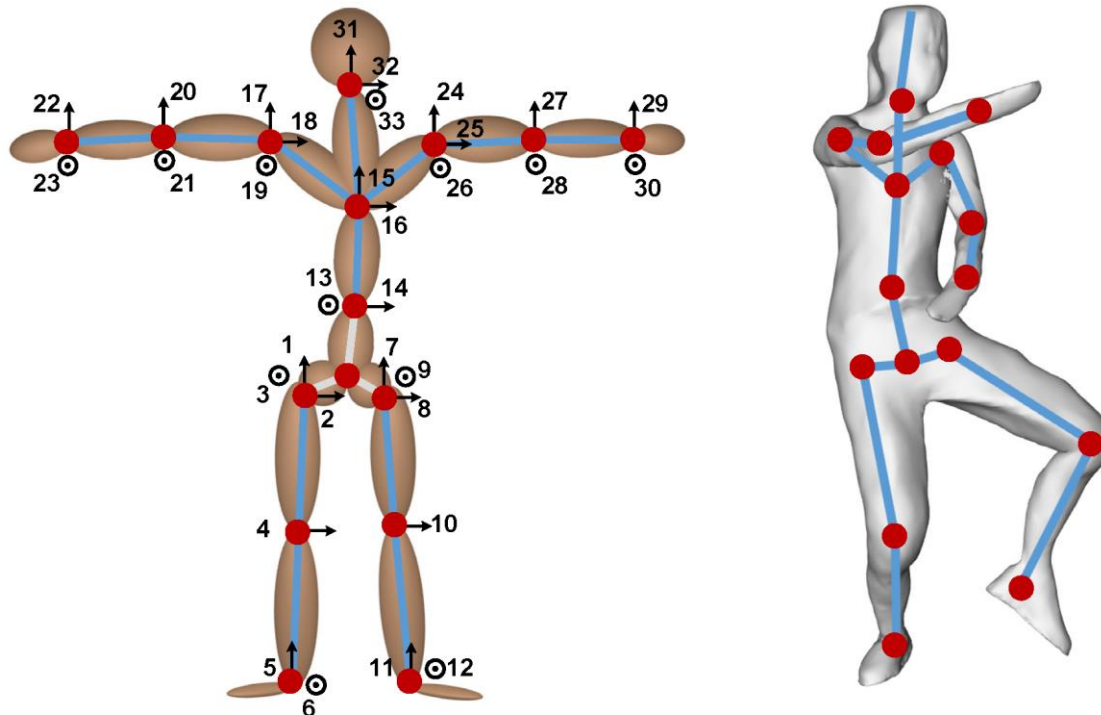


- How to perform the skinning transformations T to drive the articulate mesh model M ?
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Skinning Transformations for Articulate Pose

- Articulate pose is represented as **tree structure**
- N skinning transformations: (T_1, T_2, \dots, T_n)
 - Each T_i joint has several Degree of Freedom (DoFs)
 - All children will be influenced by their parents





Skinning Transformations for Articulate Pose

- Skinning transformations for rigid body movement
 - Rotation + translation, *a.k.a*, **screw displacement**
 - A screw is an 6-dimension ordered pair

$$S = (\mathbf{S}, \mathbf{V})$$

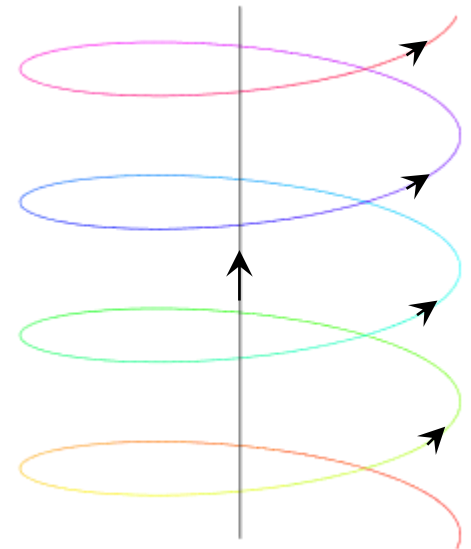
- **Twist** of rigid body (actually, twist is a screw)

$$T = (\vec{\omega}, \mathbf{v} + \mathbf{d} \times \vec{\omega})$$

$\vec{\omega}$ is the angular velocity vector

\mathbf{d} is the translation vector

\mathbf{v} is the velocity of the origin of the moving frame





Skinning Transformations for Articulate Pose

- **Revolute joints** in the articulate pose
 - Joint position is \mathbf{P} , one of rotation axis is $\vec{\omega}$

$$\xi = (\mathbf{P} \times \vec{\omega}, \vec{\omega}) = (\mathbf{v}, \vec{\omega})$$

- It can also be represented it as a matrix

$$\hat{\xi} = \begin{bmatrix} 0 & -\omega_z & \omega_y & v_1 \\ \omega_z & 0 & -\omega_x & v_2 \\ -\omega_y & \omega_x & 0 & v_3 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

- Homogenous transformation \mathbf{T}^h_i is

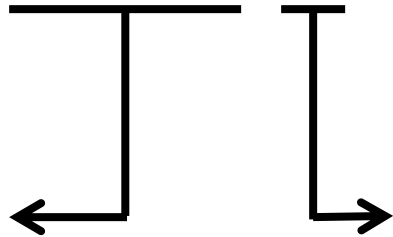
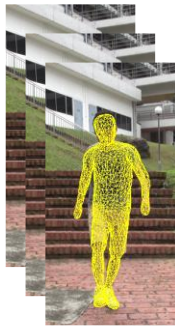
$$\mathbf{T}^h_i = e^{\theta_0 \hat{\xi}_0} \cdot \prod_{j \in \text{Parent}(i), j \neq 0} e^{\theta_j \hat{\xi}_j}$$



Motion Capture with LBS

Search optimal skinning transformations T to drive the articulate mesh model with $lbs(T)$ to best match the observed image data O

$$\min_T E(lbs(T), O)$$

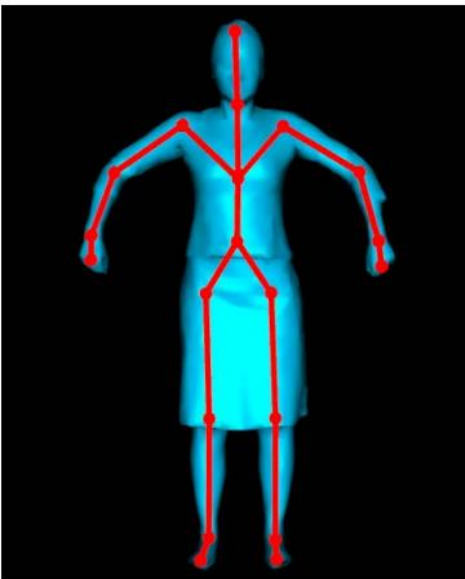


- How to perform the skinning transformations T to drive the mesh model M ?
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Analysis-by-synthesis

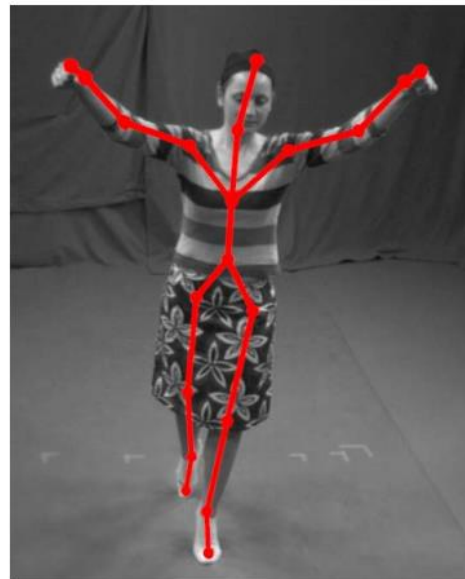
- Deform the mesh model M (**synthesize**)
 - Articulation pose is driven by $\theta=(\theta_0, \theta_1, \dots, \theta_n)$
- Compare the differences between the **projected synthesis** and **real** captured **image** data (**analysis**)



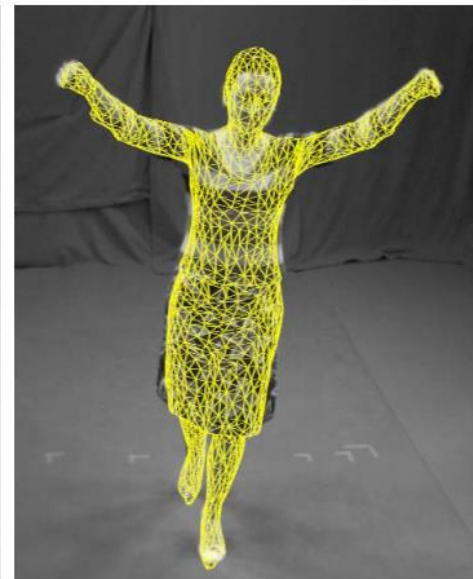
(a) Mesh model



(b) Segmented images



(c) Estimated skeleton

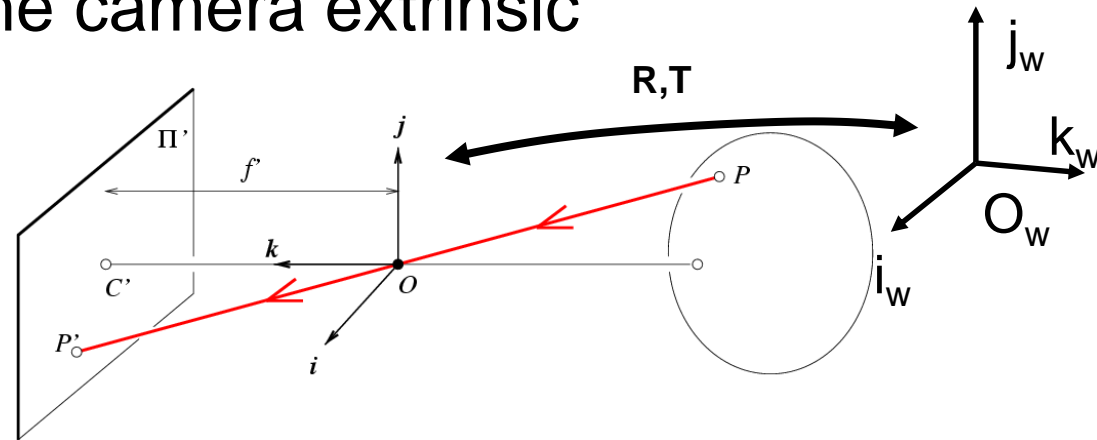


(d) Deformed surface



Camera Projection Equation

- Suppose the camera parameters are K, R, t
 K is the camera intrinsic
 R, t is the camera extrinsic



- Camera projection equation for 3D point x and its corresponding 2D point X (homogeneous)

$$sX = K(Rx + t)$$

s is the scale factor



3D-2D Consistency

- For reducing the scale factor s , we use X to cross the two sides of projection equation

$$X \times sX = X \times K(Rx + T)$$

↓
0

$$[X]_{\times} K(Rx + t) = 0$$



or

$$[X]_{\times} KRx = -[X]_{\times} Kt$$

↓
3D point

Here, $[X]_{\times} = \begin{bmatrix} 0 & -1 & Y \\ 1 & 0 & -X \\ -Y & X & 0 \end{bmatrix}$



3D-2D Consistency

- Take **silhouette** as example

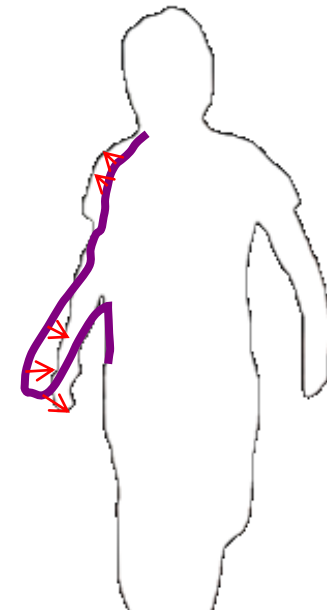


Image
contour

Projected
contour



ur
ng

How do we know the 3D vertex position in the projected contour?

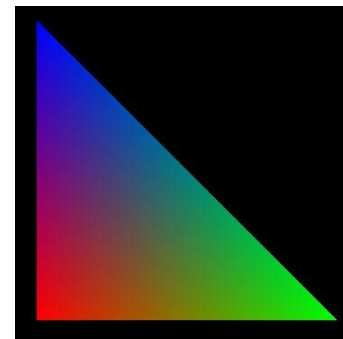


Rendering to Find 3D Position

- ❑ Two pass rendering
- ❑ **First pass**: encode the index of mesh triangles into the vertex color for each mesh face with (R, G, B) $\rightarrow 2^{24}-1=8388607$
- ❑ **Second pass**: encode the $(1,0,0)$, $(0,1,0)$, $(0,0,1)$ three colors for each mesh face to find the barycentric coordinate




1st pass: color with face index



2nd pass: color for barycentric



More about 3D-2D Consistency

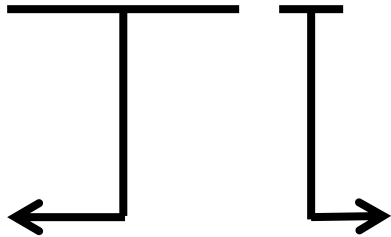
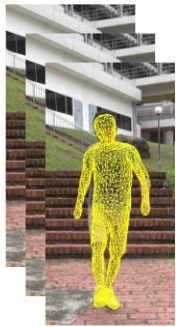
- ❑ **Silhouette** and **contour matching** is successful for model driven motion capture
 - Motion Capture Using Joint Skeleton Tracking and Surface Estimation, CVPR2009
 - Markerless Motion Capture of Interacting Characters Using Multi-view Image Segmentation, CVPR2011
- ❑ Obtain the silhouette or mask is a very challenge topic  **Deep learning!!! e.g., Mask-rcnn**
- ❑ Model Rendering with texture
 - SIFT feature matching
 - Color consistency
 - Edge consistency



Motion Capture with LBS

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$$\min_T E(lbs(T), O)$$



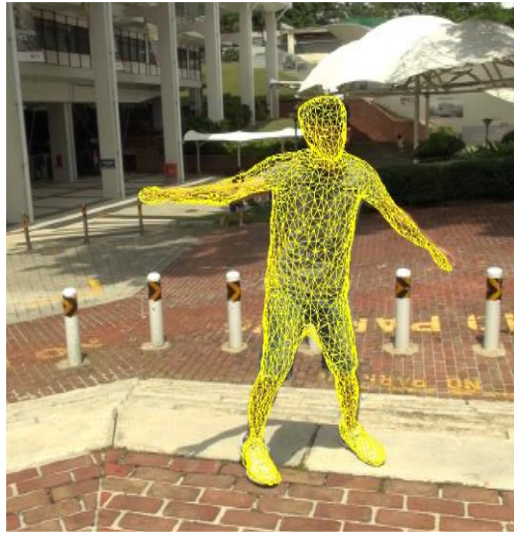
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Global Optimization

□ Sampling-based optimization

Interacting simulated annealing (ISA)



Deformed mesh



Observed image

<https://www.yangangwang.com/papers/wang-omm-2017-04>



Local Optimization

- Consider the LBS deformation of articulated shape

$$v'_i = \left(\sum_{j=1}^m w_{i,j} \mathbf{T}_j \right) v_i$$

v_i : original vertex; v'_i : deformed vertex

- It is also noted that

$$\mathbf{T}^h_i = e^{\theta_0 \hat{\xi}_0} \cdot \prod_{j \in \text{Parent}(i), j \neq 0} e^{\theta_j \hat{\xi}_j}$$



Local Optimization

- With the Taylor series at v_i and performing the linearization, we could have

$$v'_i = \left(I + \theta_0 \hat{\xi}_0 + \sum_{j=1}^m \bar{w}_j \theta_j \hat{\xi}_j \right) v_i$$

Where, $\bar{w}_j = \sum_{k \in \text{children}(j)} w_k$

- The upper equation can also be represented as

$$v'_i = v_i + J\chi \quad \rightarrow \quad \text{(B-8) in appendix B of my Ph.D. Thesis}$$

Where, χ has all the articulation parameters

---- From appendix B in my Ph.D. Thesis



Local Optimization

- Our goal is to find some χ to perform the 3D-2D matching
- Since we have

$$[X]_{\times}KRx = -[X]_{\times}Kt$$

- And

$$v'_i = v_i + J\chi$$

- Solve $\mathbf{Ax} = \mathbf{b}$ is easy and fast



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Advanced Techniques

- We have discussed about the articulated mesh deformation
- How about non-rigid?

YOUKU

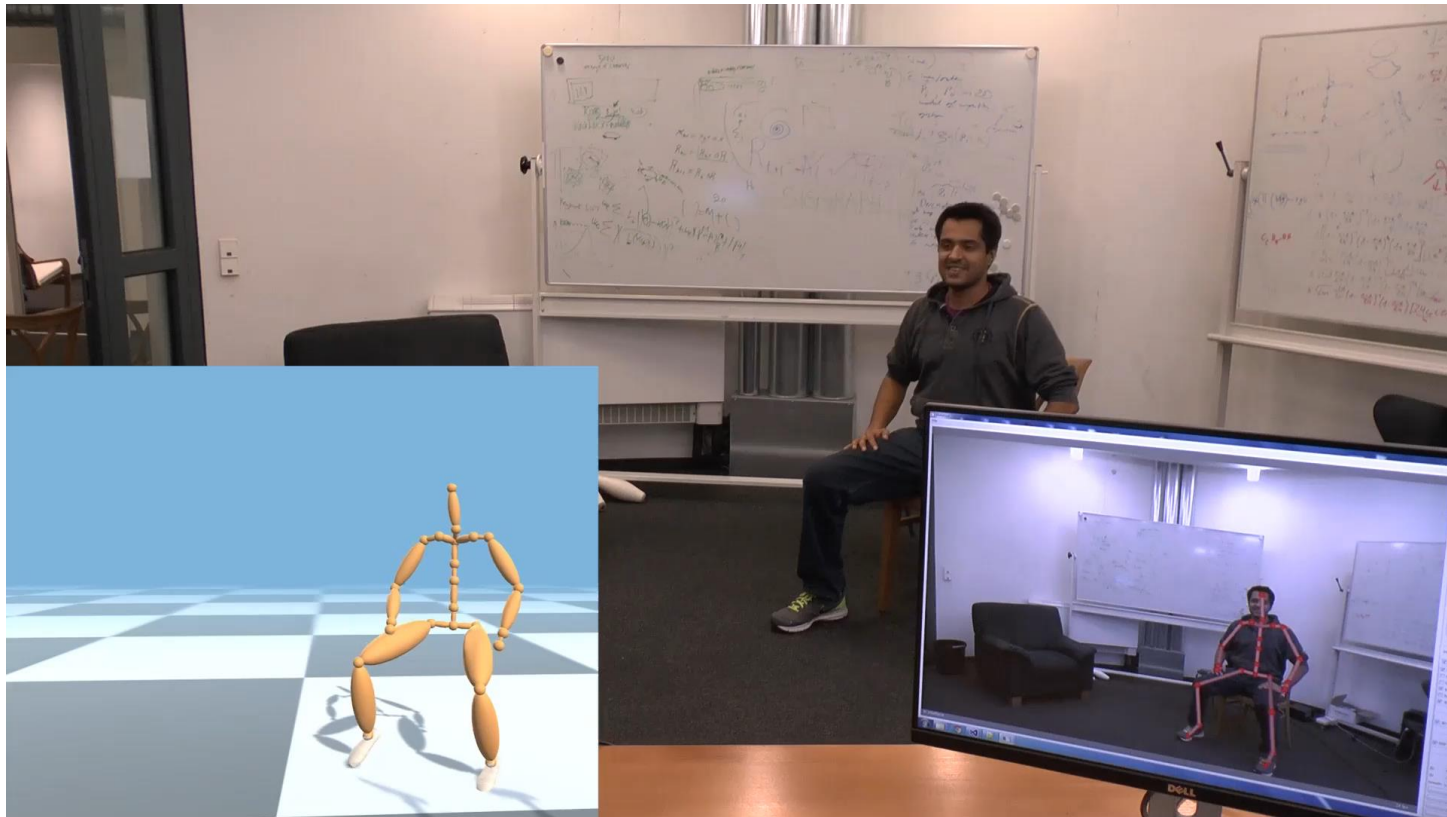
"Girl"
500 frames, 30 fps

- Robust Non-Rigid Motion Tracking and Surface Reconstruction Using L0 Regularization, *IEEE Transactions on Visualization and Computer Graphics (TVCG)*, 2018



Advanced Techniques

- How about motion capture without LBS?



- VNect: Real-time 3D Human Pose Estimation with a Single RGB Camera, SIGGRAPH 2017

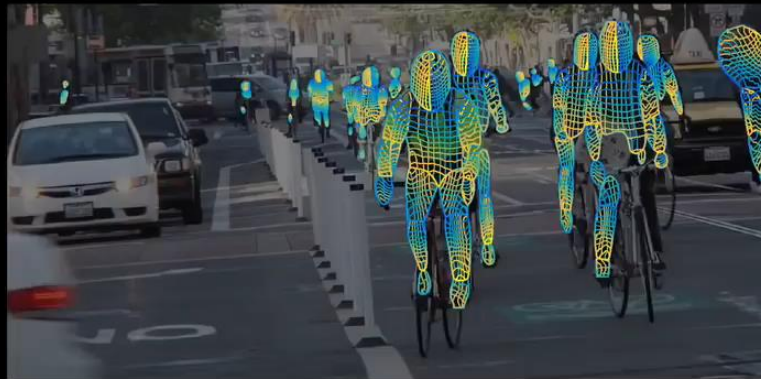


Advanced Techniques

- How about motion capture without LBS?

DensePose:

Dense Human Pose Estimation In The Wild



Rıza Alp Güler *
INRIA, CentraleSupélec

Natalia Neverova
Facebook AI Research

Iasonas Kokkinos
Facebook AI Research

* Rıza Alp Güler was with Facebook AI Research during this work.

- Dense human pose estimation aims at mapping all human pixels of an RGB image to the 3D surface of the human, CVPR2018



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Conclusion

- ❑ LBS is a very basic and widely used technique for articulated motion capture
- ❑ LBS suffers from the candy-wrapper effect

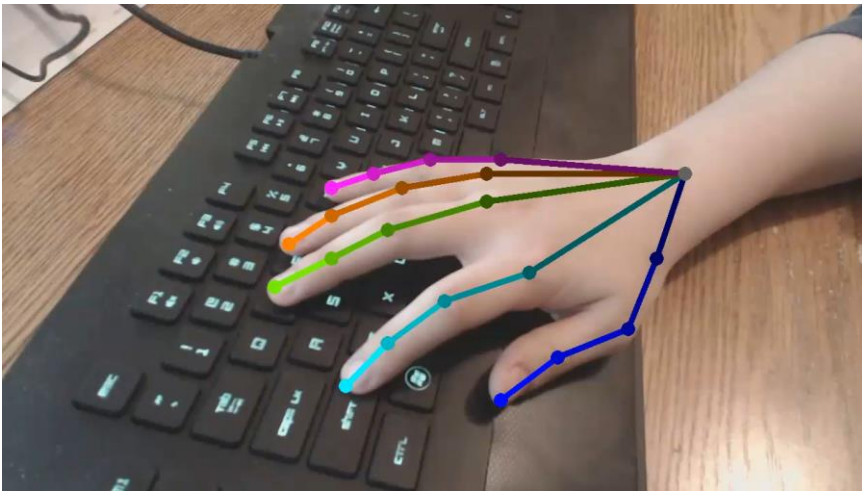
LBS





Easter Egg

- Currently, I am focusing on real-time motion capture for **hand** gesture



- The dataset (***OneHand10K***) will be released in my personal website soon! 😊
 - More than **10K** with 21 2D joints and **Mask**

- Mask-pose Cascaded CNN for 2D Hand Pose Estimation from Single Color Image, *IEEE Transactions on Circuits and Systems for Video Technology (TCSVT)*, 2018. **Accepted**



Any Questions?
