

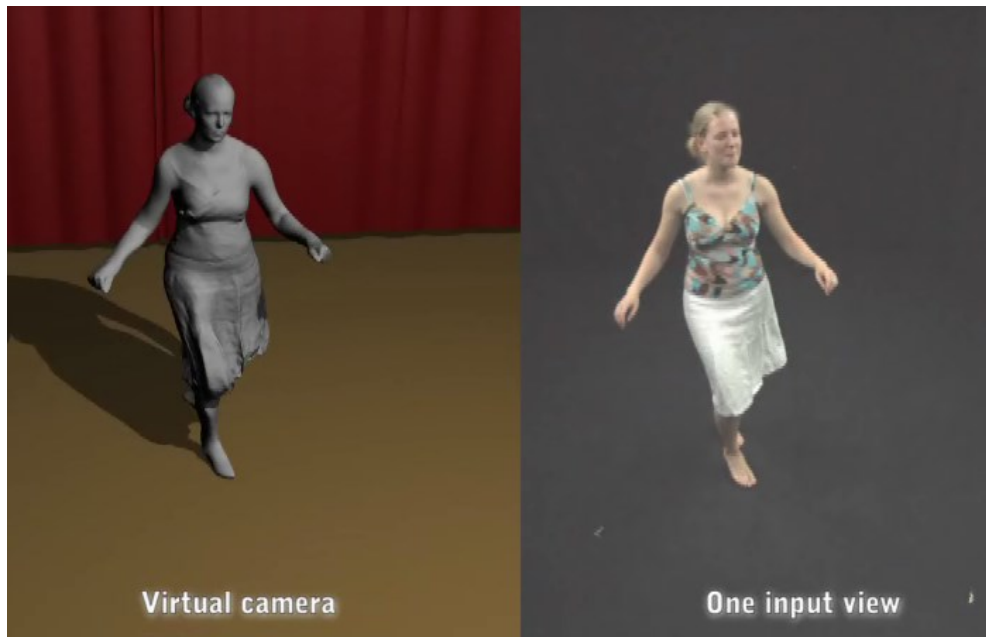
Performance Capture with Depth Cameras

Christian Theobalt

Graphics, Vision & Video



Marker-less Performance Capture



[de Aguiar et al., SIGGRAPH 2008]



[Liu et al. CVPR 2011, PAMI 2013]

- High-quality shape, motion and texture of people in general clothing
- Input: multi-view video, no markers
- Exceed capabilities of existing Mocap technology by far

High-quality Binocular Face Capture



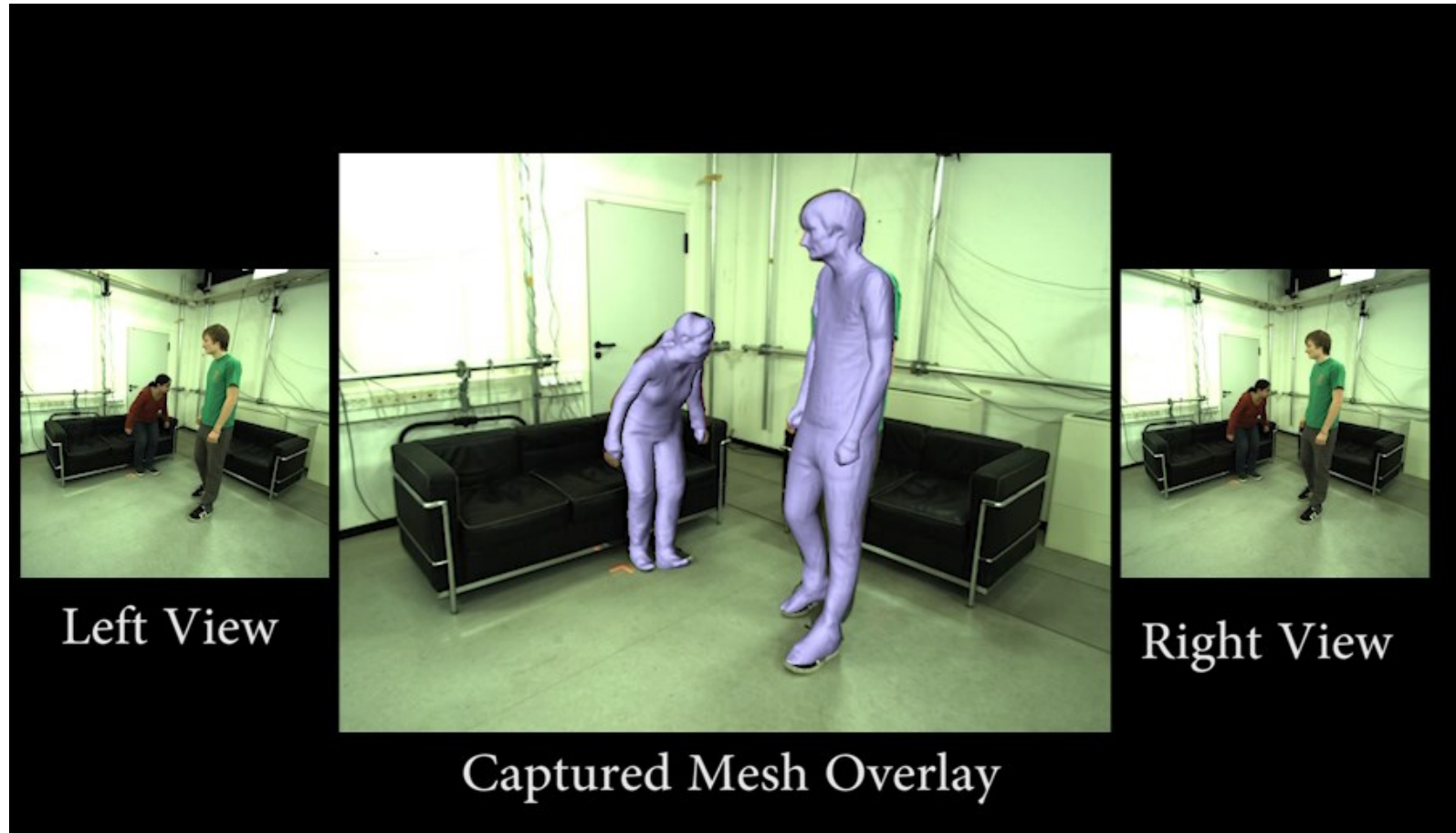
Left View



Right View

L.Valgaerts, C. Wu, A. Bruhn, H.-P. Seidel, C. Theobalt, Binocular Facial Performance Capture under Uncontrolled Illumination, [SIGGRAPH Asia 2012](#)

On-Set Performance Capture with a Stereo Camera



C. Wu, C. Stoll, L. Valgaerts, C. Theobalt, On-set Performance Capture of Multiple Actors with a Stereo Camera, [SIGGRAPH Asia 2013](#)

Space-Time Upsampling

- Space-Time Joint Upsampling



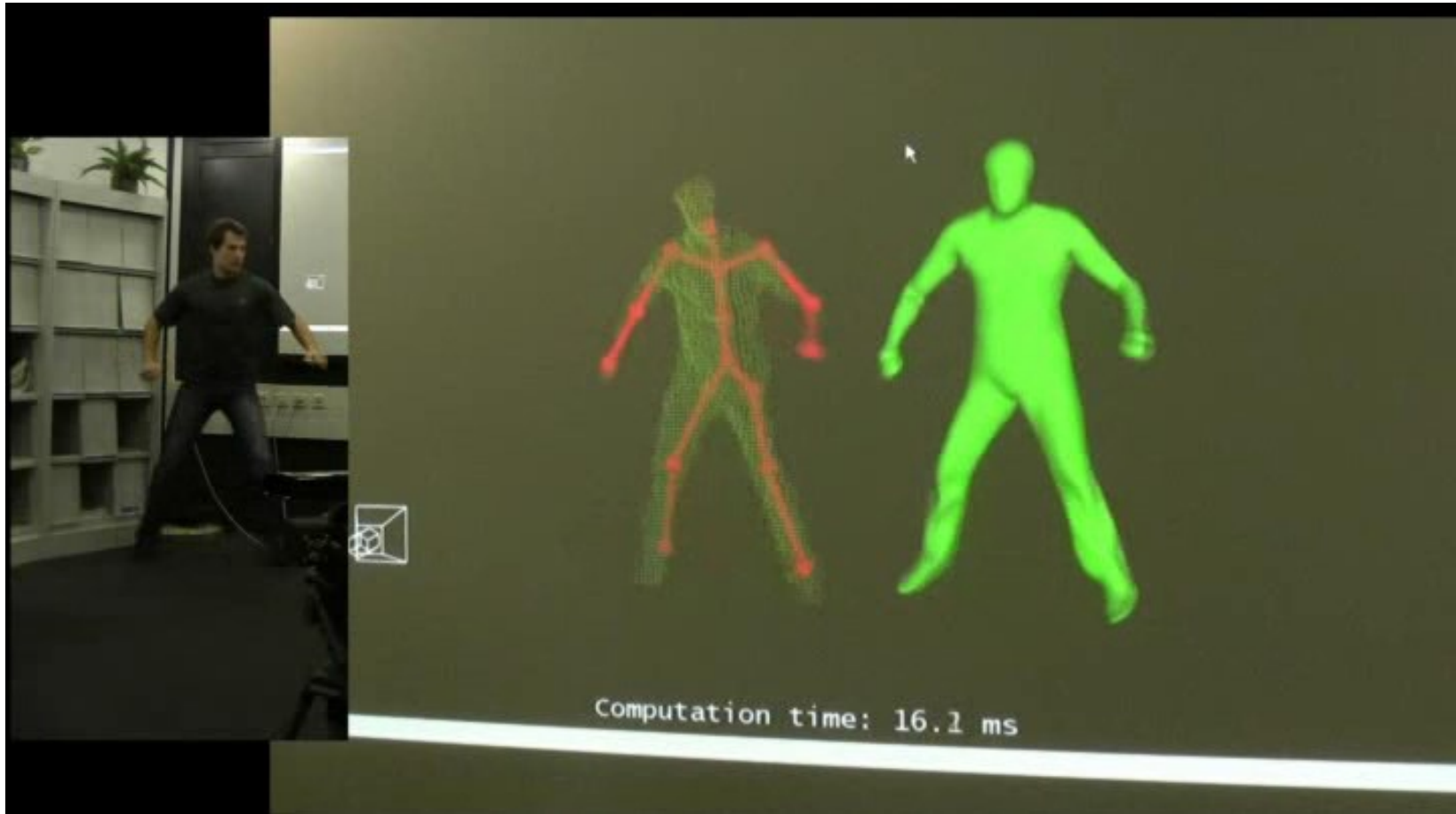
Aligned Video



Filtered Distance Map
(rendered as mesh)

C. Richardt, C. Stoll, N. Dogdson; H.-P. Seidel, C. Theobalt, *Coherent Spatiotemporal Filtering, Upsampling and Rendering of RGBZ Videos*, in [CGF \(Proc. EUROGRAPHICS\)](#), 2012.

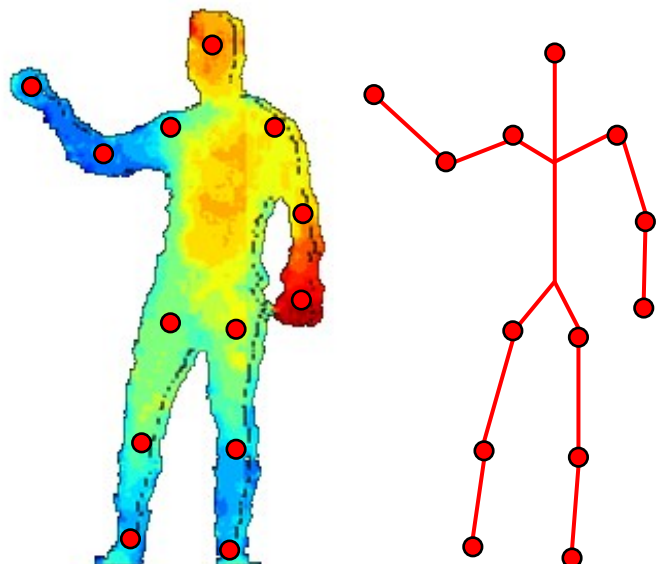
Real-time Skeletal Tracking – Single Depth Camera



A. Baak, M. Müller, G. Bharaj, H.-P. Seidel, C. Theobalt, *A Data-Driven Approach for Real-Time Full Body Pose Reconstruction from a Depth Camera*, [Proc. ICCV](#), 2011.

Real-time Skeleton Tracking from Depth

Discriminative approaches

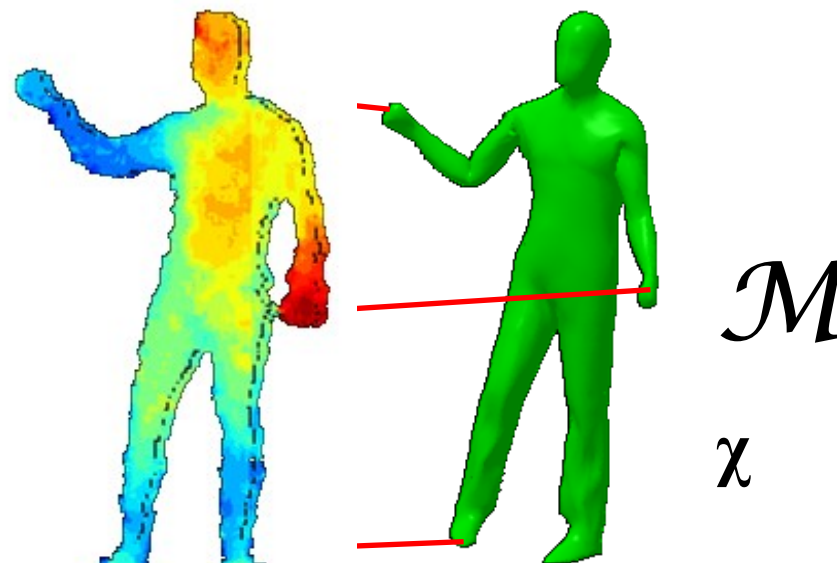


Shotton et al., *CVPR*, 2011

Taylor et al., *CVPR*, 2012

....

Generative approaches



Bleiweiss et al., *SIGGRAPH ASIA Sketches*, 2009

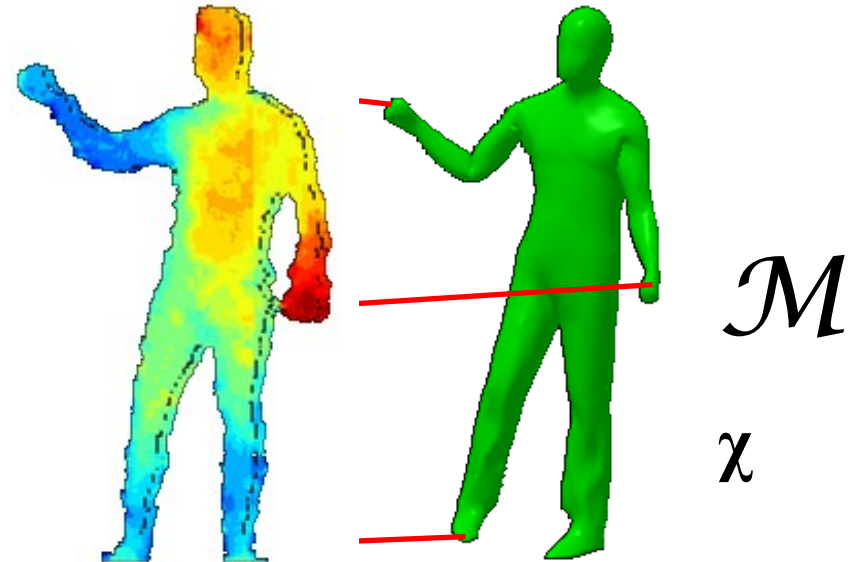
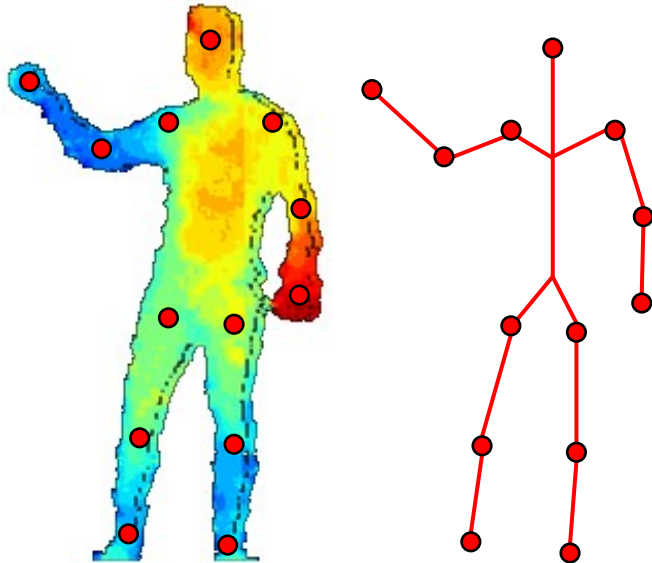
Knoop et al., *Robotics and Autonomous Systems*, 2009

Ganapathi et al. *CVPR* 2010, *ECCV* 2012...

Real-time Skeleton Tracking from Depth

Discriminative approaches

Generative approaches

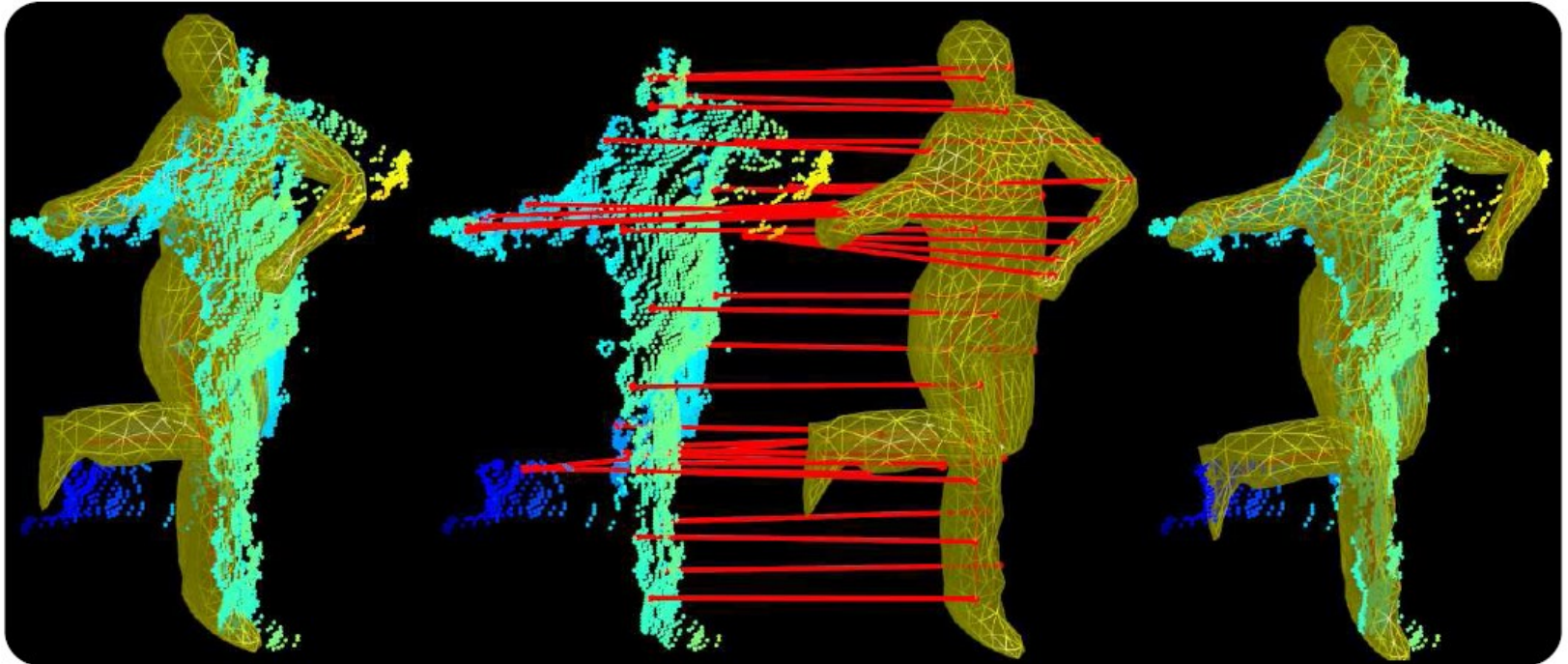


Baak et al.,/CCV 2011

Helten et al., 3DV 2013

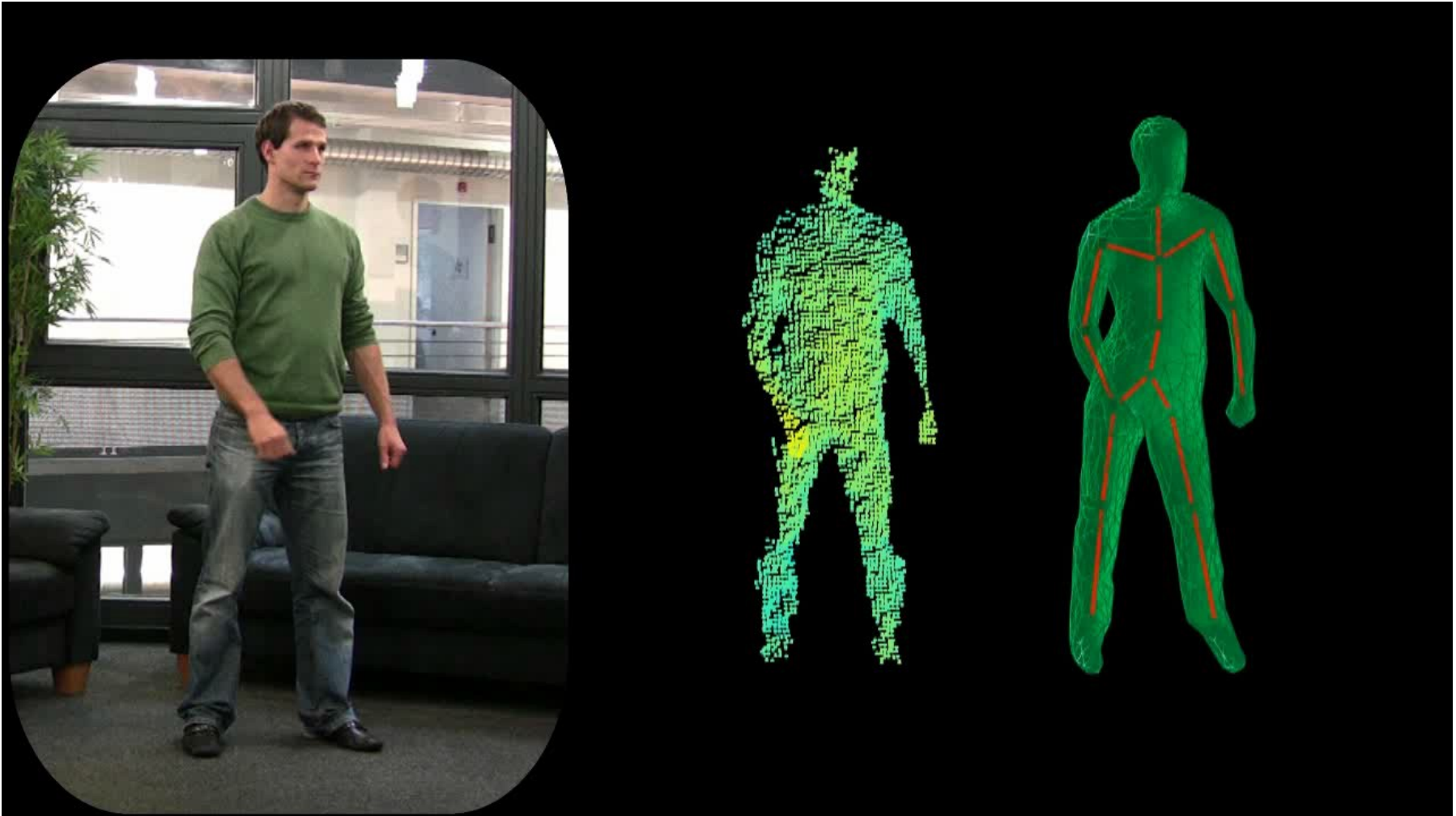
Wei et al., SIGGRAPH Asia 2012

Solver I: Generative Pose Optimization

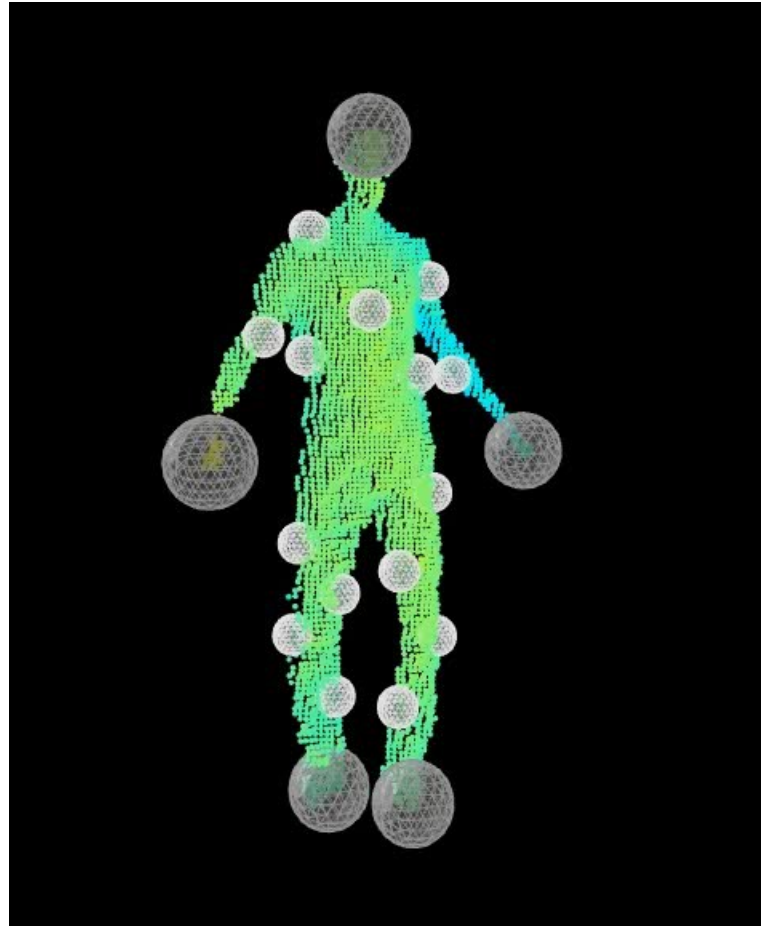


Correspondence hypothesis between model and point cloud →
optimize pose parameters that minimize distance

Solver I: Local Pose Minima

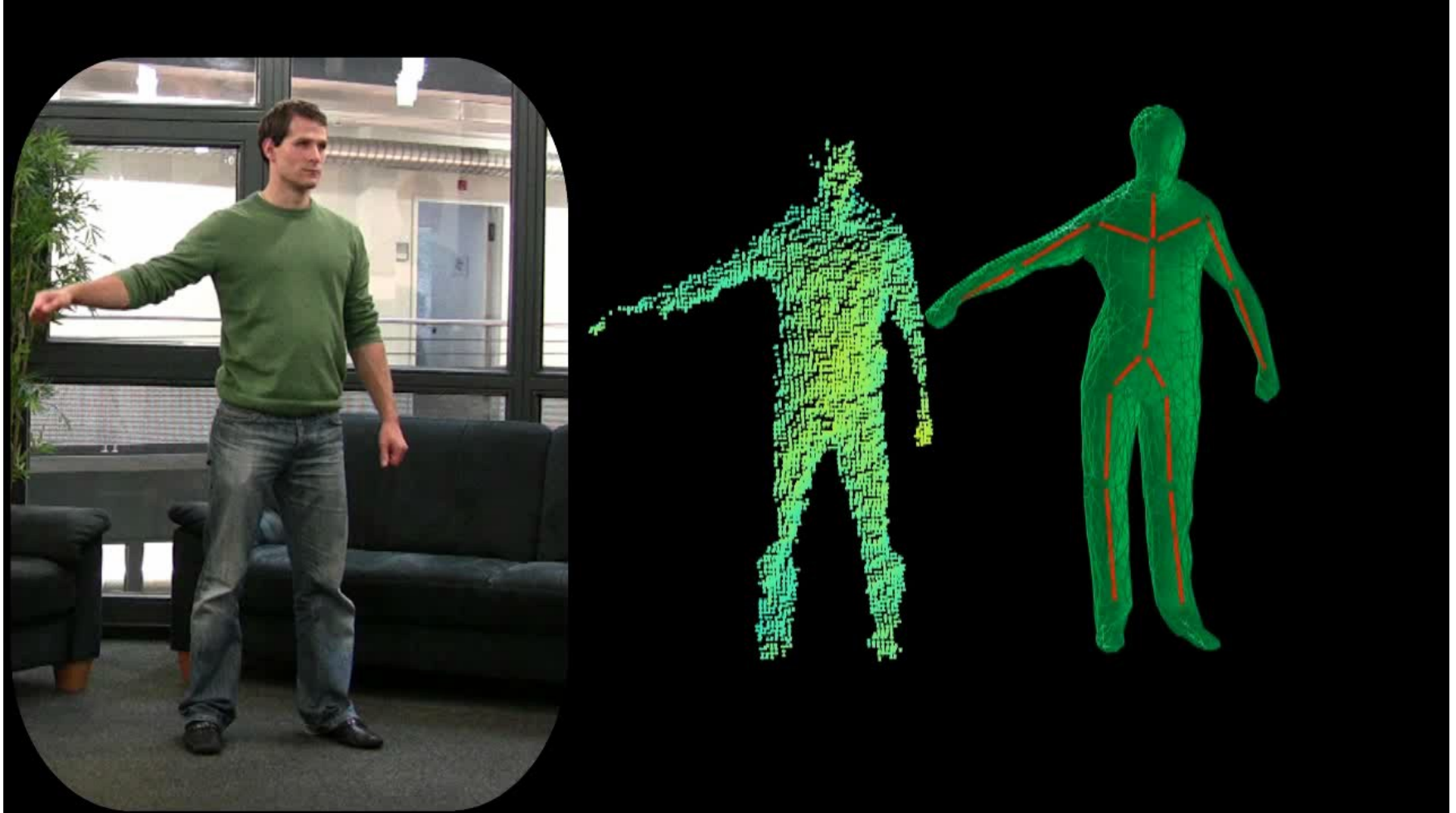


Solver II: Geodesic Extrema



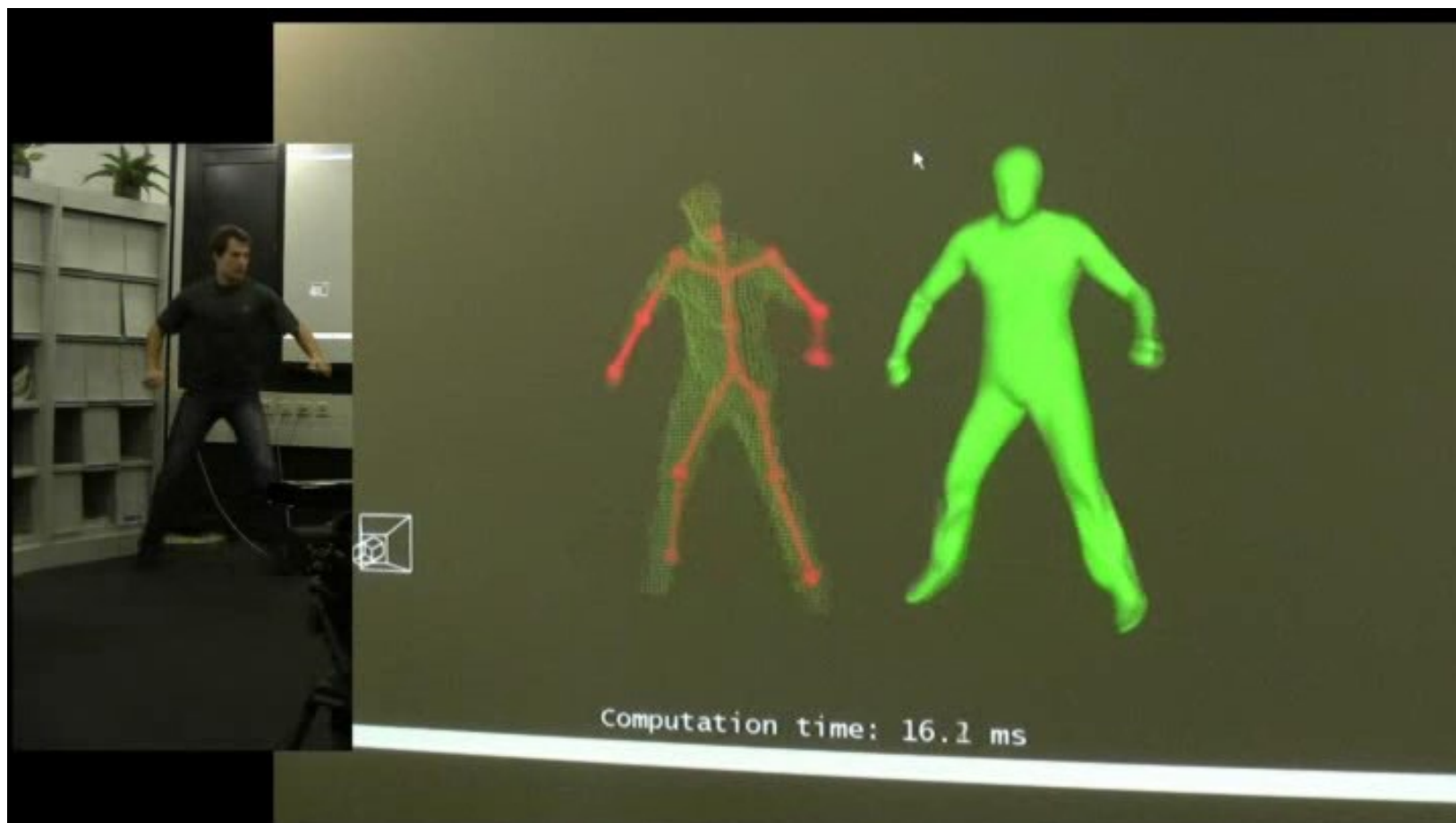
Label 5 end-effectors + temporal smoothness

Solver II: (Discriminative) Database Lookup



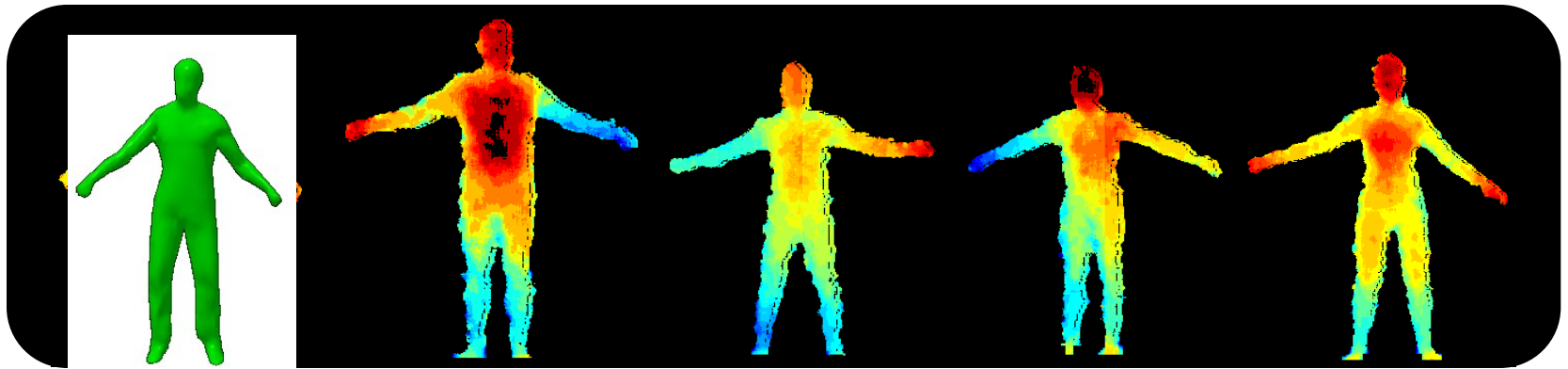
Kd-tree: Find closest pose from 23000 in a database

Fusion: Hypothesis Selection



A. Baak, M. Müller, G. Bharaj, H.-P. Seidel, C. Theobalt, *A Data-Driven Approach for Real-Time Full Body Pose Reconstruction from a Depth Camera*, [Proc. ICCV](#), 2011.

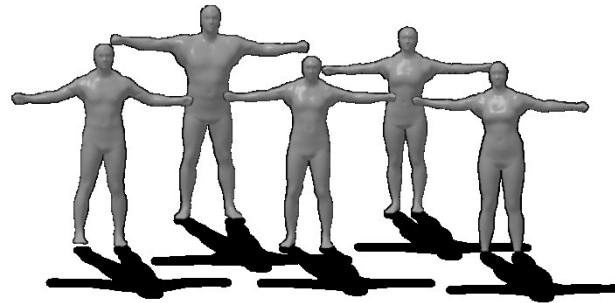
Personalization of Body Model



- Fast fitting of parametric human body model
- Personalized Hybrid Depth Tracker

T. Helten, A. Baak, G. Bharaj, M. Müller, H.-P. Seidel, C. Theobalt, *Personalization and Evaluation of a Real-time Depth-based Full Body Tracker*, [Proc. of 3DV](#), 2013.

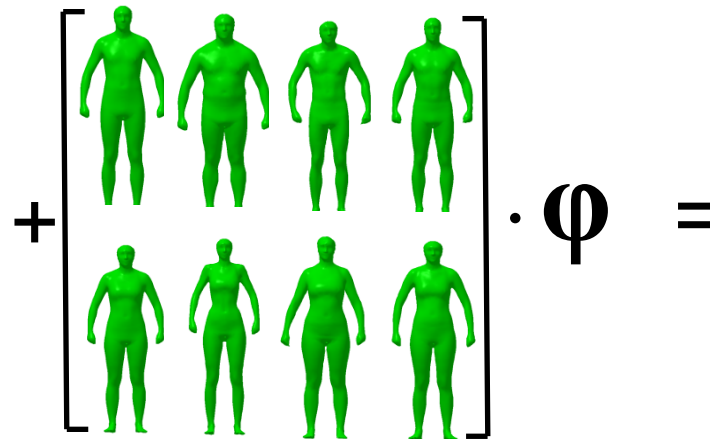
Shape-parameterized Model



Pre-recorded
laser scans

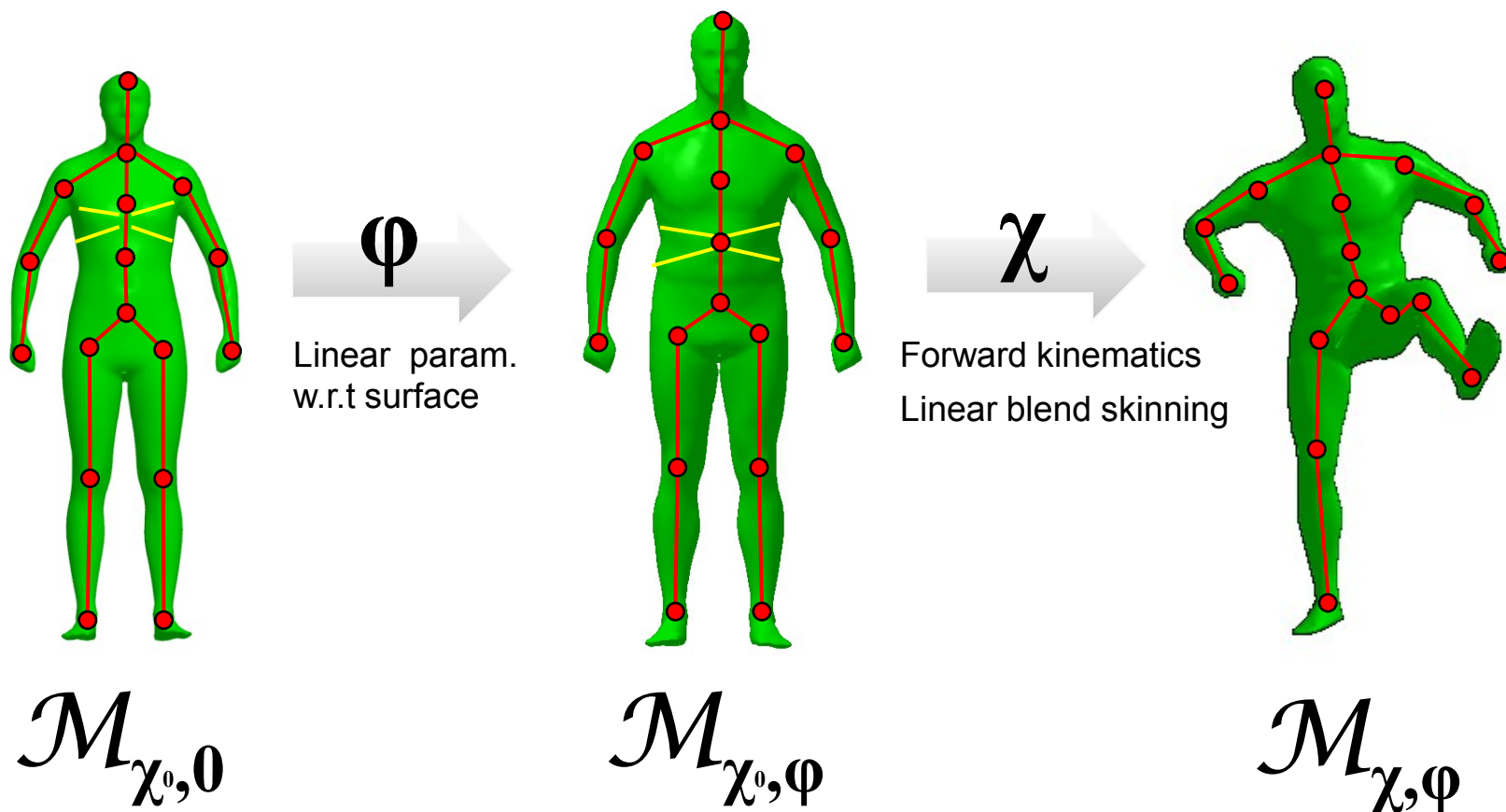


PCA



$$\mathcal{M}_{\chi_0,0} + \Phi \cdot \phi = \mathcal{M}_{\chi_0,\phi}$$

Pose Parameterization



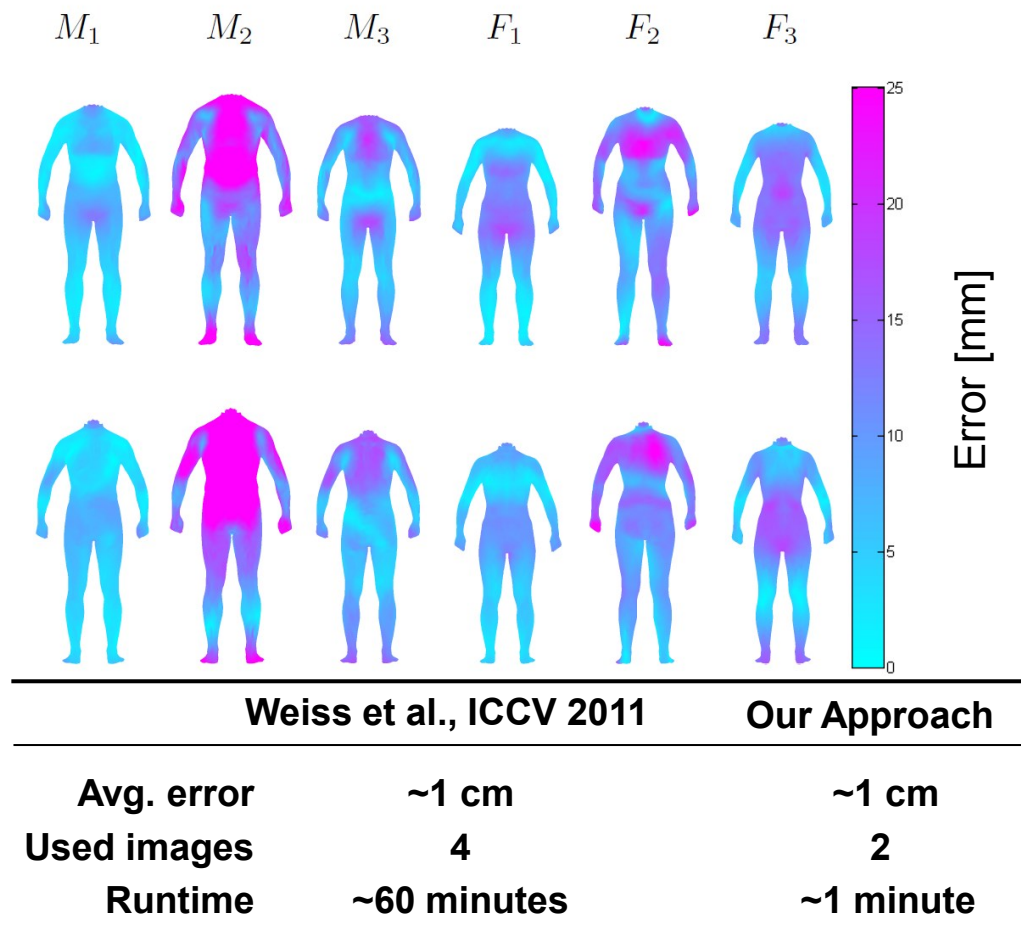
Shape Estimation

Personalization and Evaluation of a
Real-time Depth-based Full Body Tracker

T. Helten, A. Baak, G. Bharaj, M. Müller,
H.-P. Seidel, and C. Theobalt

3DV 2013

Shape Estimation Results



Personalized Depth Tracker

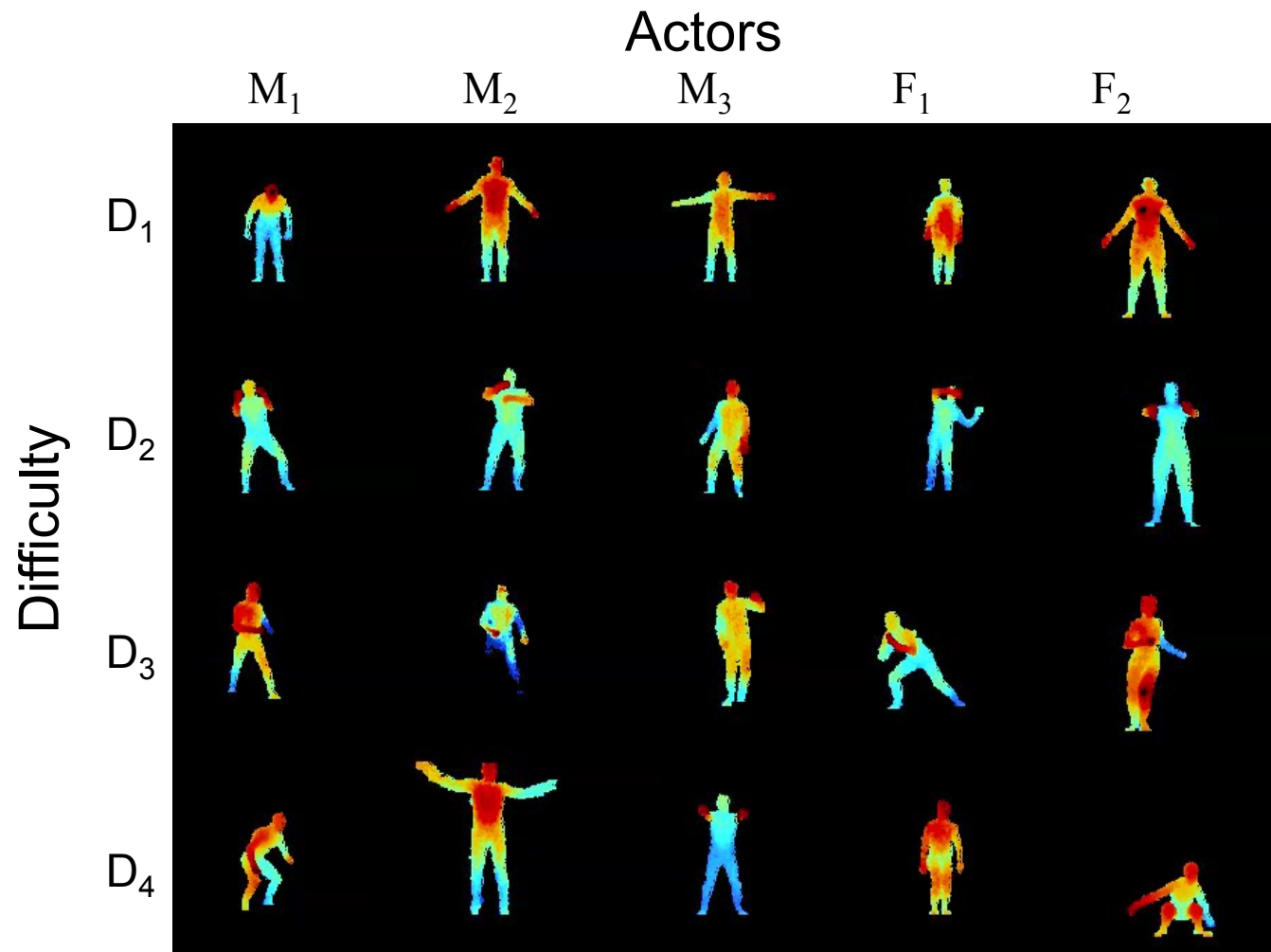
Personalization and Evaluation of a
Real-time Depth-based Full Body Tracker

T. Helten, A. Baak, G. Bharaj, M. Müller,
H.-P. Seidel, and C. Theobalt

3DV 2013

T. Helten, A. Baak, G. Bharaj, M. Müller, H.-P. Seidel, C. Theobalt, *Personalization and Evaluation of a Real-time Depth-based Full Body Tracker*, [Proc. of 3DV](#), 2013.

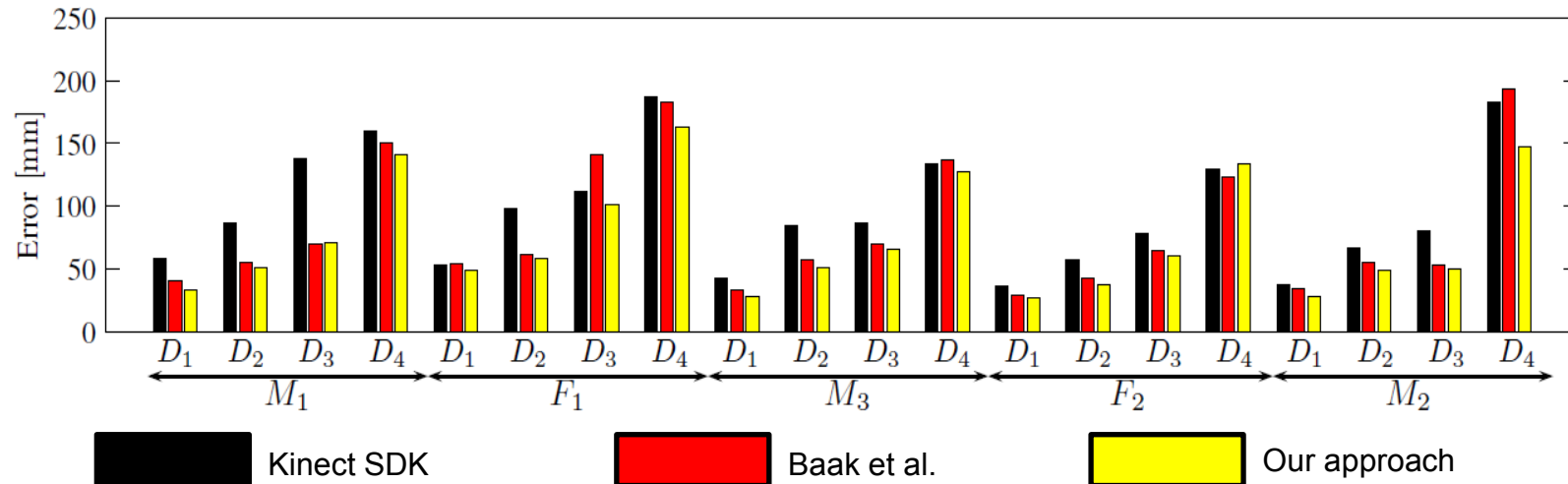
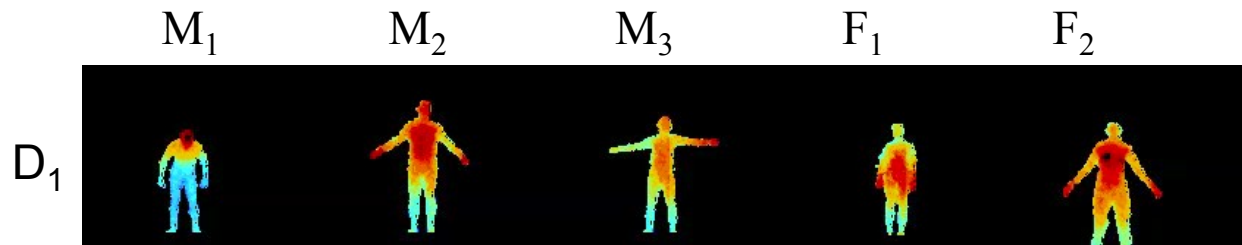
Evaluation Dataset



- Calibrated Depth data • Ground-truth marker positions & marker positions

Evaluation Dataset

Actors



- Calibrated Depth data & marker positions
- Ground-truth marker positions

Results

Real-time Tracking with One Depth Camera and Inertial Sensors

Thomas Helten* Meinard Müller' Hans-Peter Seidel* Christian Theobalt*

*Saarland University and MPI Informatik

'International Audio Laboratories Erlangen

ICCV 2013, Sydney, Australia

T. Helten, M. Müller, and H.-P. Seidel, and C. Theobalt, *Real-time Body Tracking with One Depth Camera and Inertial Sensors*, [ICCV 2013](#).

Results – Evaluation Dataset

Real-time Tracking with Two Depth Cameras and Inertial Sensors

Meinard Müller' Hans-Peter Seidel* Christian Theobalt*
'Saarland University and MPI Informatik
'International Audio Laboratories Erlangen

ICCV 2013, Sydney, Australia

Real-time Tracking with One Depth Camera and Inertial Sensors

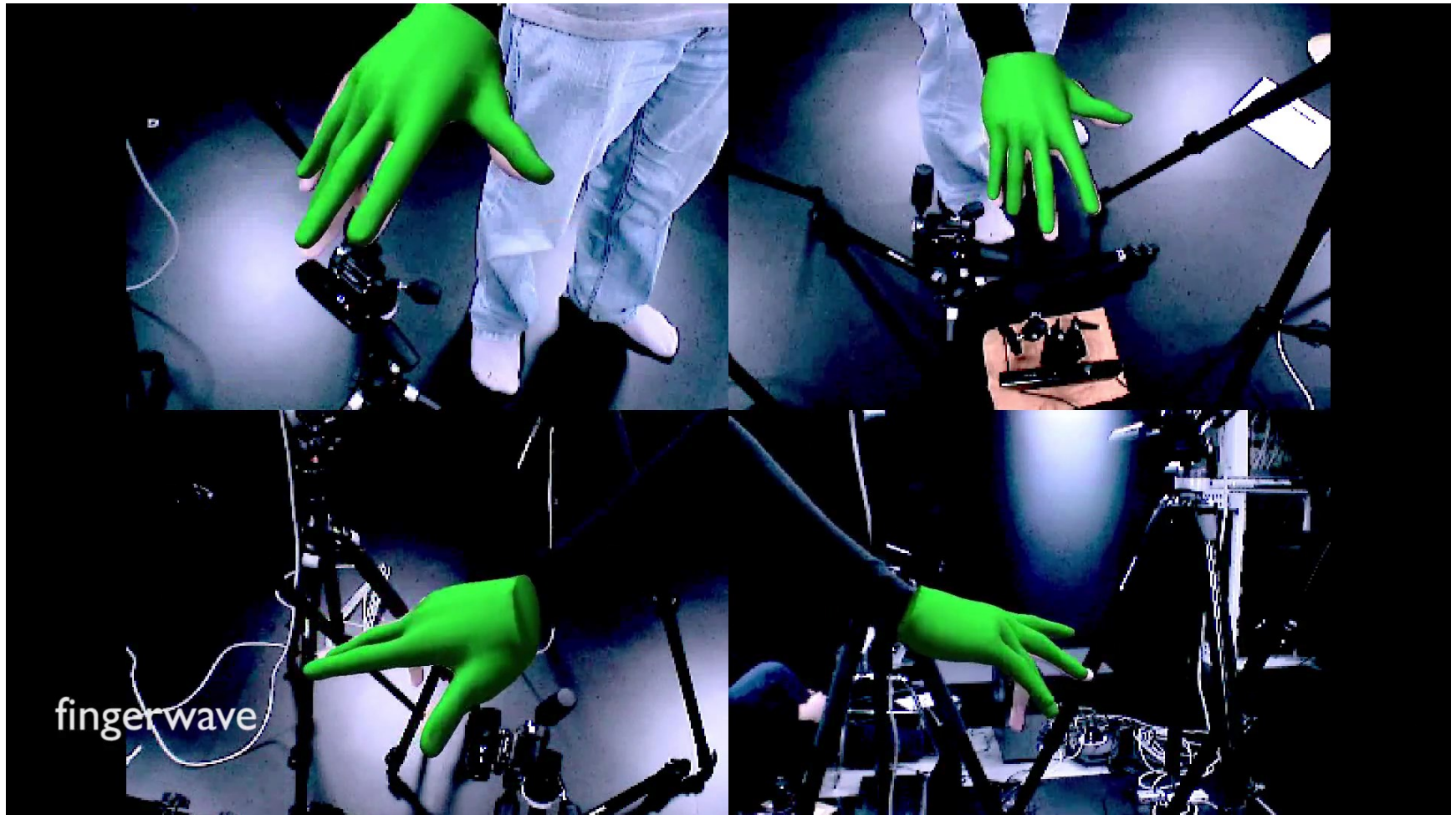
Thomas Helten* Meinard Müller' Hans-Peter Seidel* Christian Theobalt*
'Saarland University and MPI Informatik
'International Audio Laboratories Erlangen

ICCV 2013, Sydney, Australia

6 sequences of varying difficulty

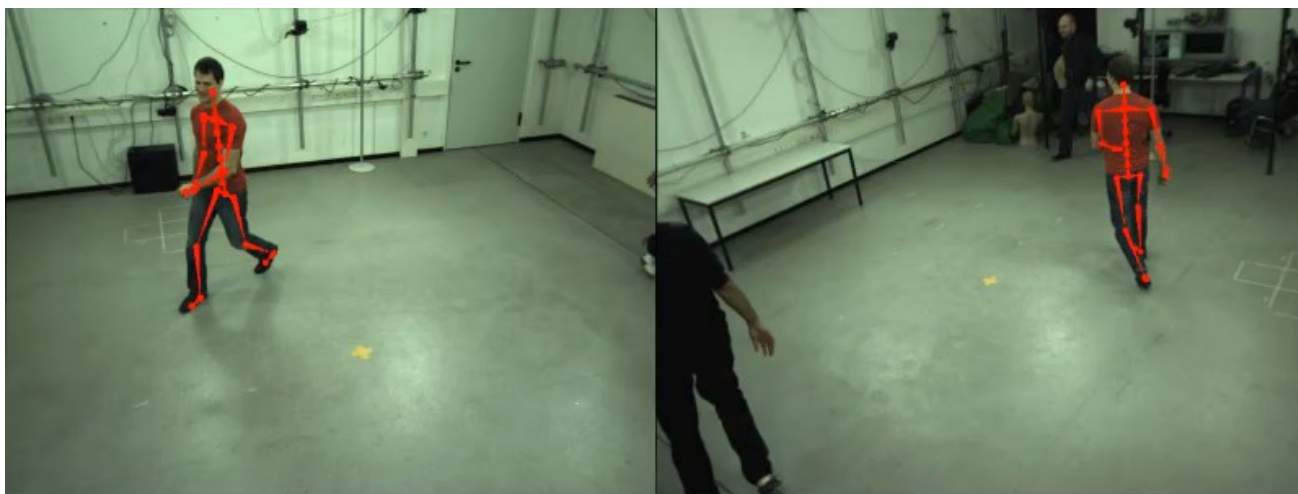
- Kinect
- XSENS IMUs
- Ground truth: 38 markers

Real-time Articulated Hand Tracking

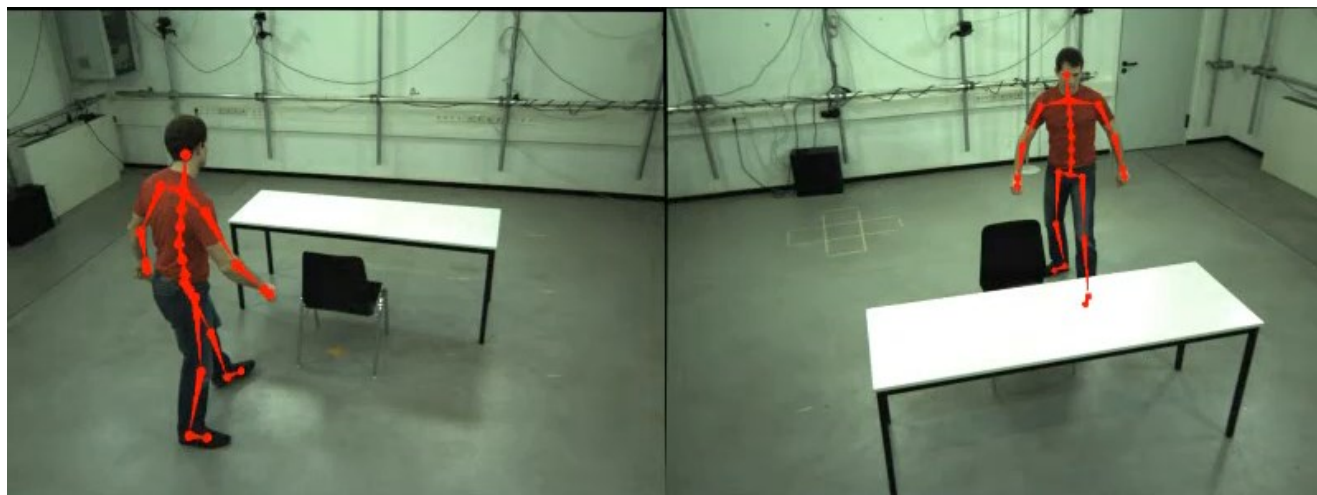


S. Sridhar, A. Oulasvirta, and C. Theobalt, *Interactive Markerless Articulated Hand Motion Tracking using RGB and Depth Data*, ICCV 2013

Real-time Full-Body Motion Capture



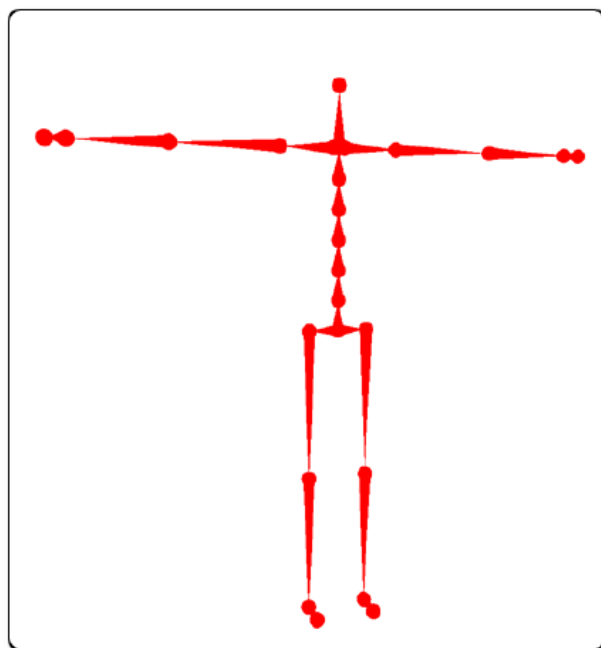
Occlusions,
fast motions,
dynamic background...



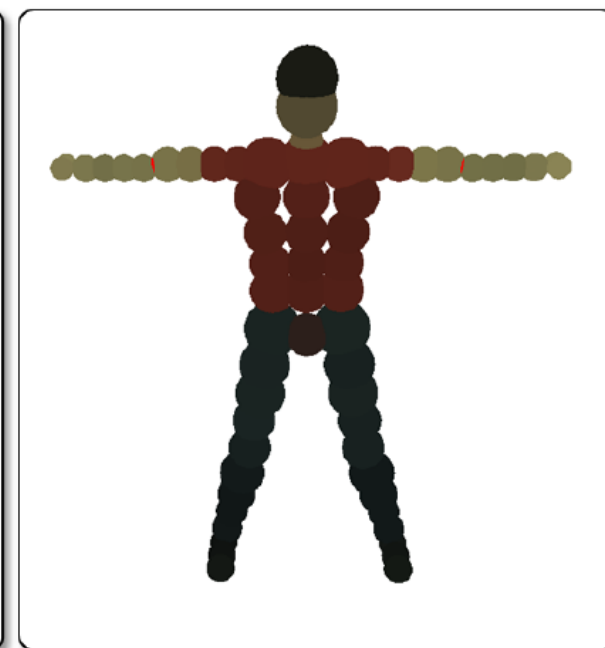
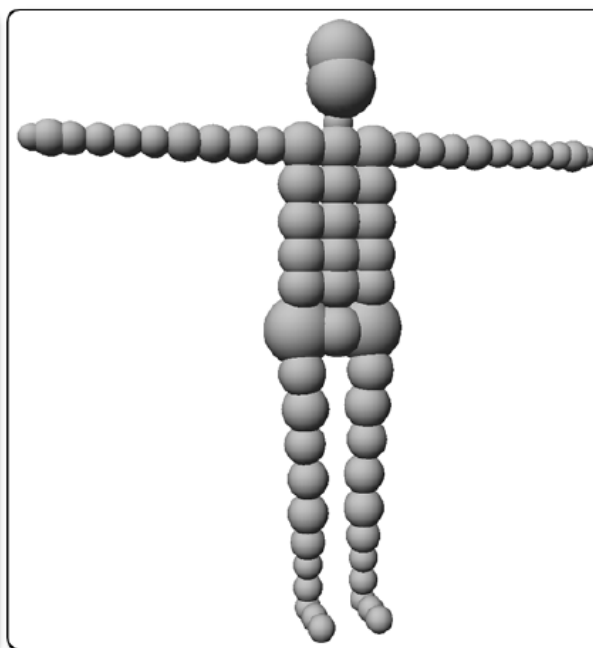
C. Stoll, N. Hasler, J. Gall, H.-P. Seidel, C. Theobalt, *Fast Articulated Motion Tracking using a Sums of Gaussians Body Model*, [Proc. ICCV](#), 2011.

3D Sum of Gaussians Body Model

- Skeleton + Gaussians with color model



58 DoFs



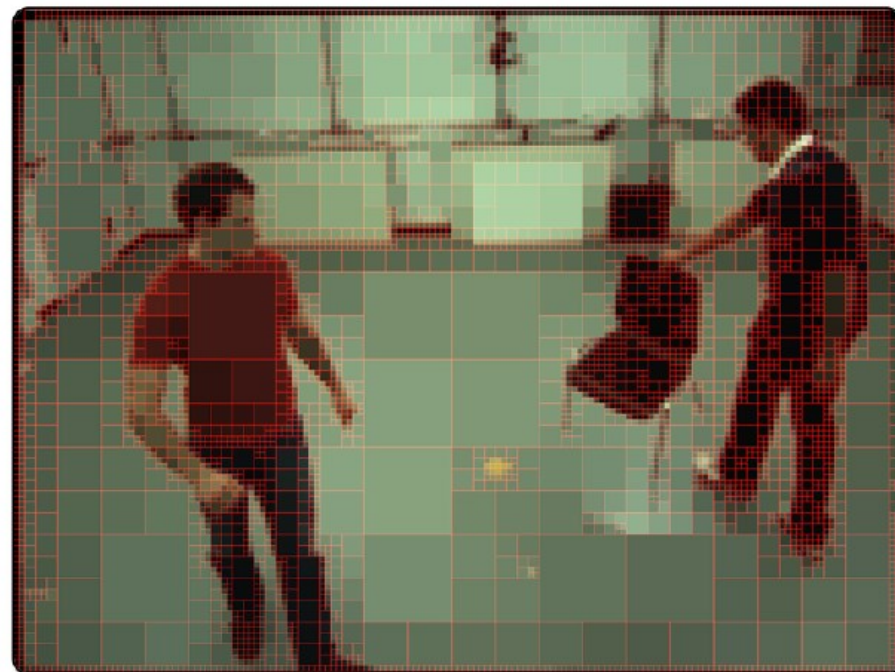
63 Gaussians

2D Sum of Gaussians Body Model

- Each video frame converted to 2D SoG



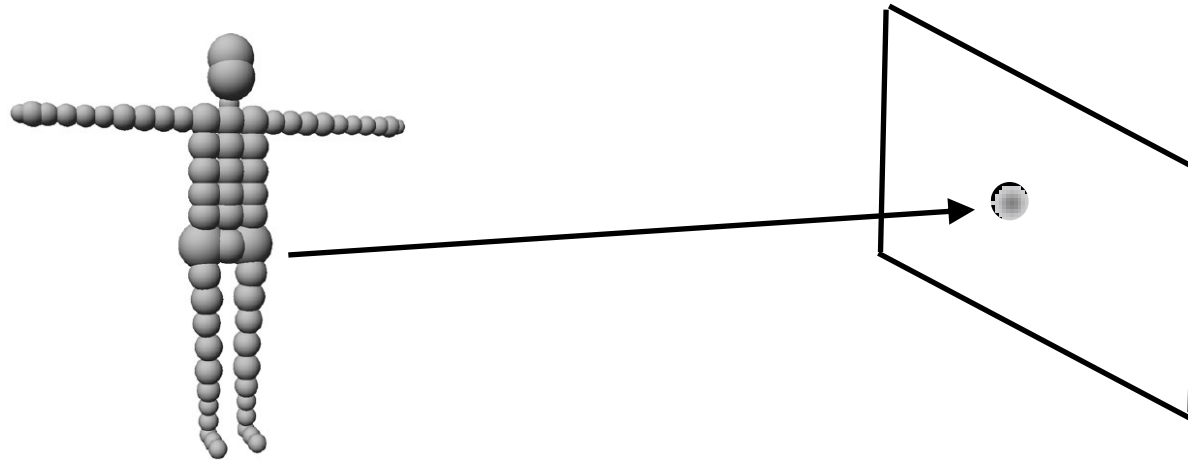
Video frame



Quad tree –
each region one 2D Gaussian

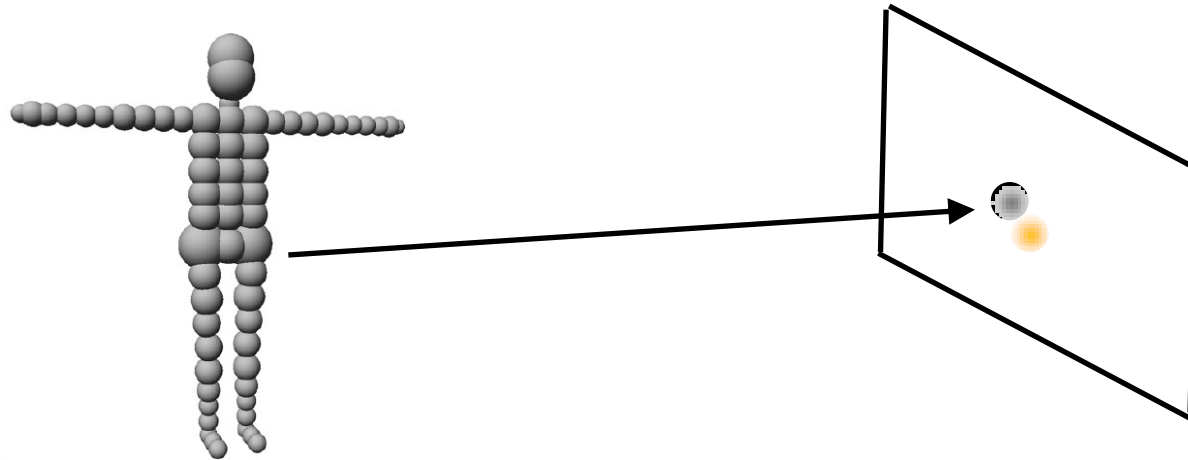
3D-2D Similarity Function

- Projection of 3D Gaussians into cameras



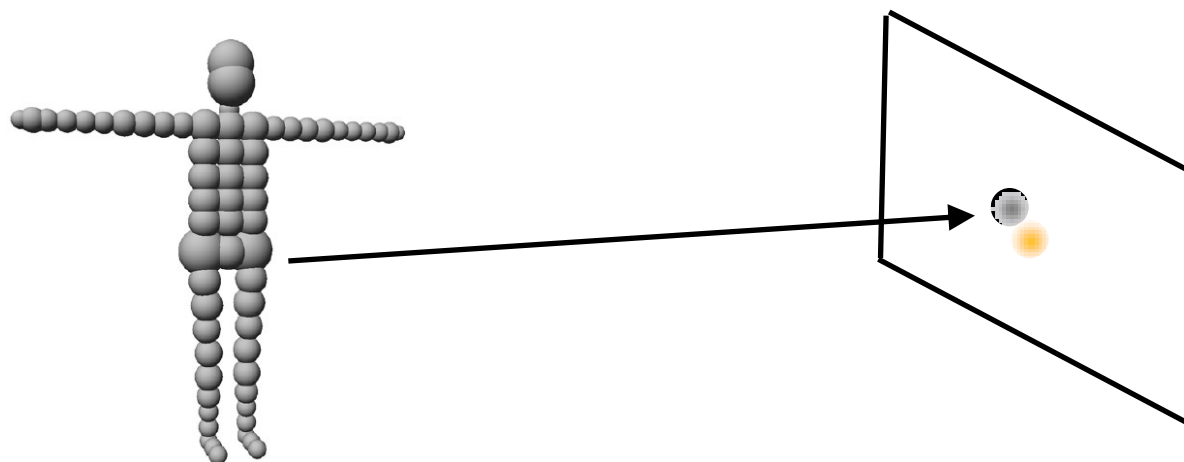
3D-2D Similarity Function

- Projection of 3D Gaussians into cameras



3D-2D Similarity Function

- Projection of 3D Gaussians into cameras



- 2D-2D SoG similarity in one camera view

$$E_{ij} = \underbrace{d(\mathbf{c}_i, \mathbf{c}_j)}_{\text{Color similarity}} \int_{\Omega} \underbrace{\mathcal{B}_i(x)}_{\text{Projected model Gaussian}} \underbrace{\mathcal{B}_j(x)}_{\text{Image Gaussian}} dx$$

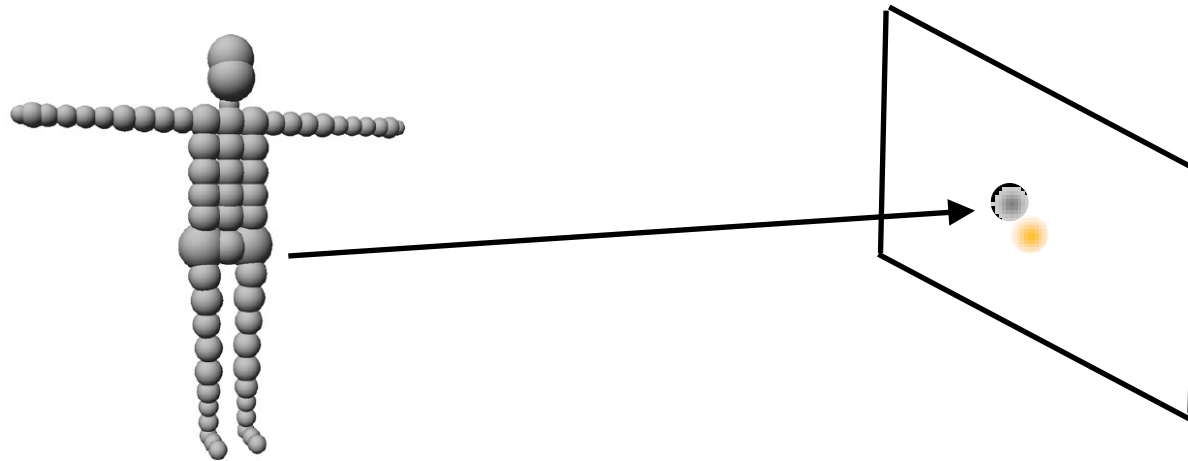
Color similarity

Projected model
Gaussian

Image Gaussian

3D-2D Similarity Function

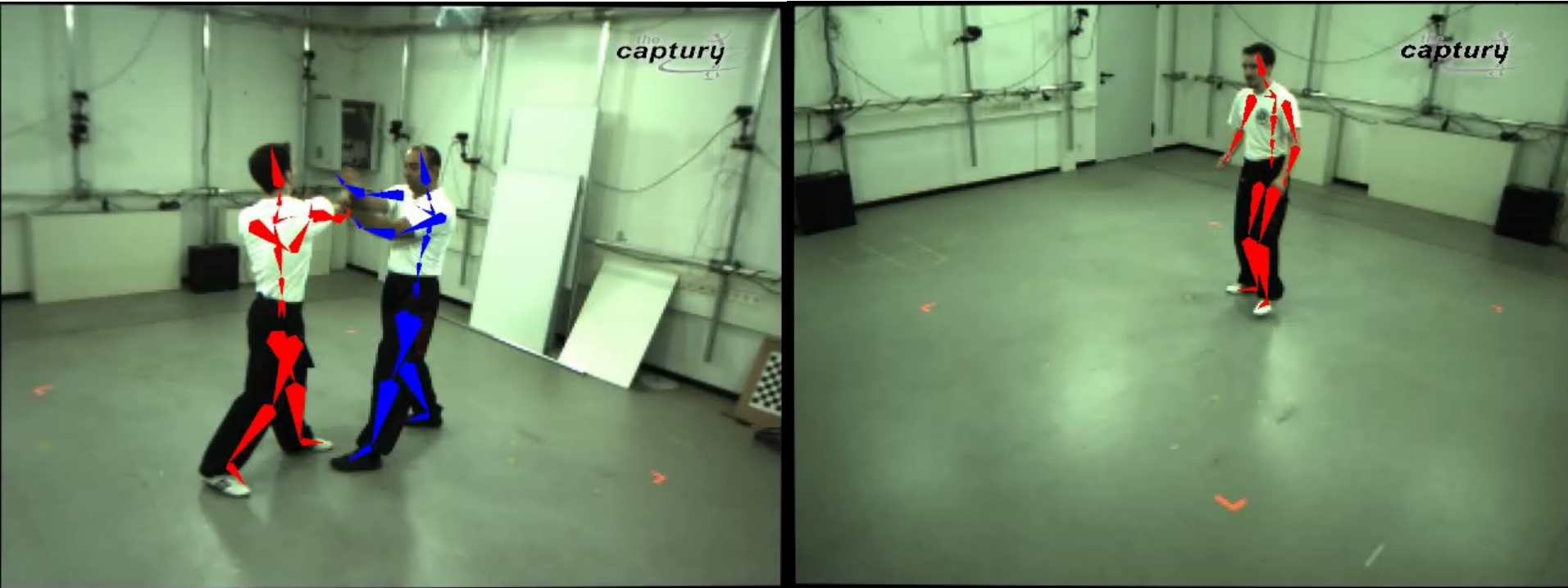
- Projection of 3D Gaussians into cameras



- 2D-2D SoG similarity – closed form solution

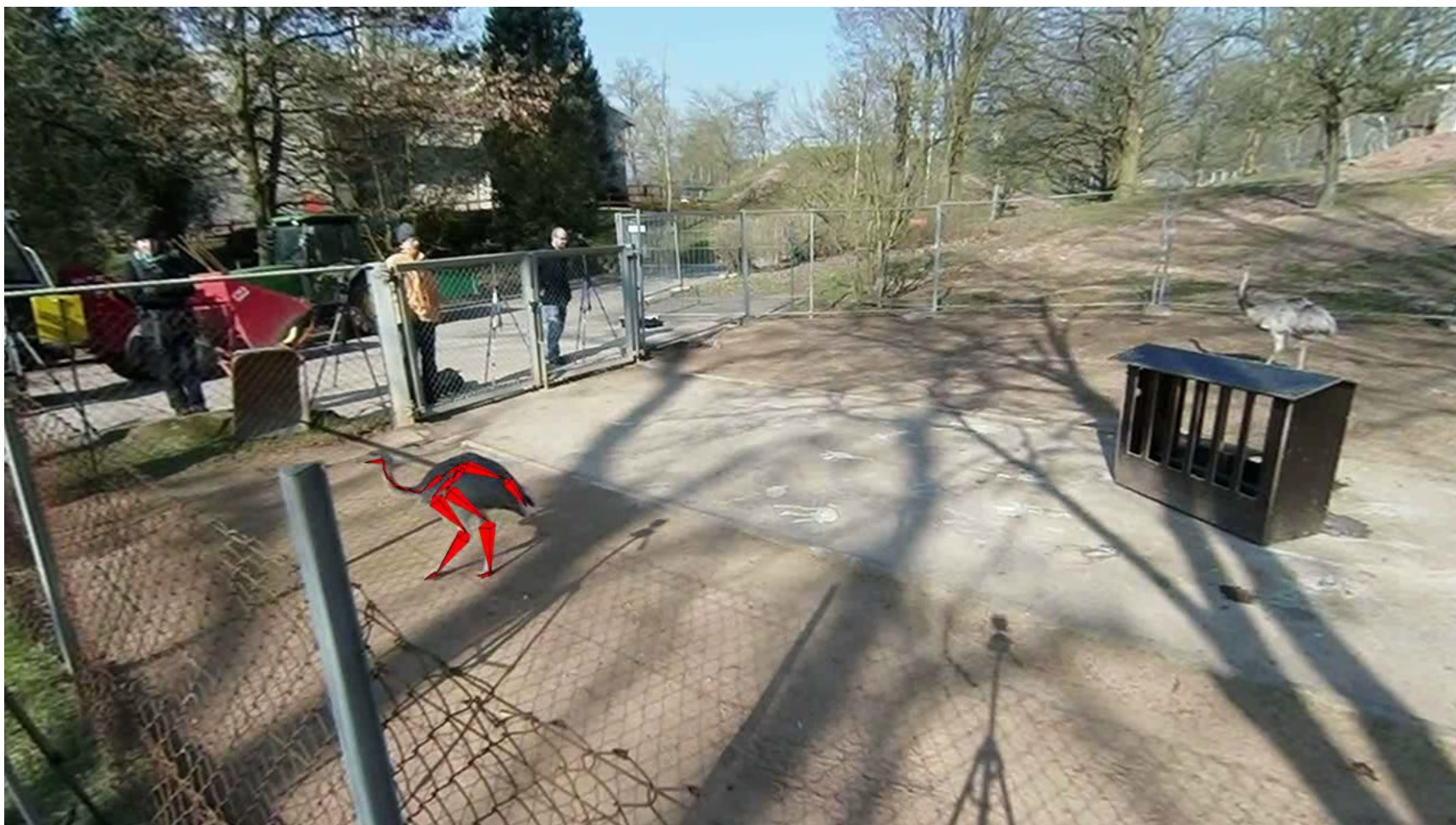
$$\begin{aligned}
 E_{ij} &= d(\mathbf{c}_i, \mathbf{c}_j) \int_{\Omega} \mathcal{B}_i(x) \mathcal{B}_j(x) dx \\
 &= d(\mathbf{c}_i, \mathbf{c}_j) 2\pi \frac{\sigma_i^2 \sigma_j^2}{\sigma_i^2 + \sigma_j^2} \exp \left(-\frac{\|\mu_i - \mu_j\|^2}{\sigma_i^2 + \sigma_j^2} \right)
 \end{aligned}$$

More Results



Min nr. of cameras: 5
Pose estimation > 30 fps

Space-time Reconstruction



8 GoPros, 30 fps, not frame synchronized

[Elhayek et al., CVPR 2012]

www.thecaptury.com



Real-time Hand-Tracking: Setup

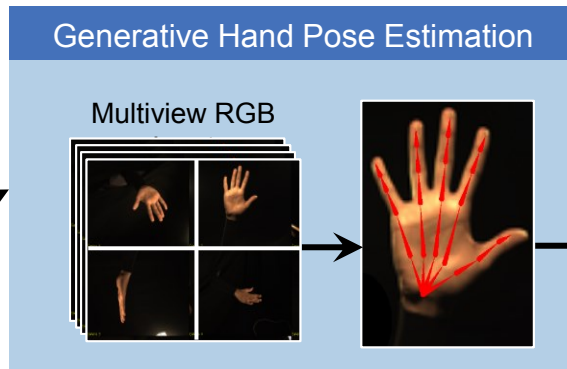
5 video cameras

1 Creative TOF Camera



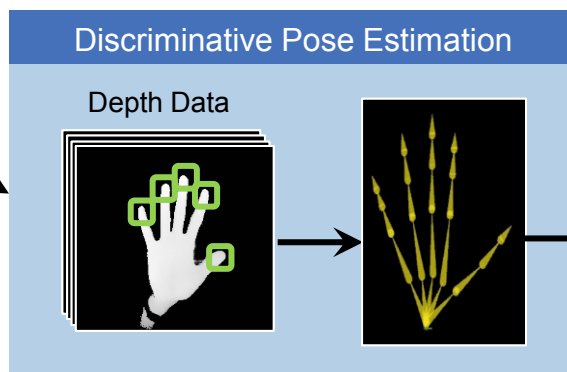
Fusion of two Tracking Strategies

1



$$\theta_0^t = \theta^{t-1} + \alpha(\theta^{t-1} - \theta^{t-2}) \quad \theta_G$$

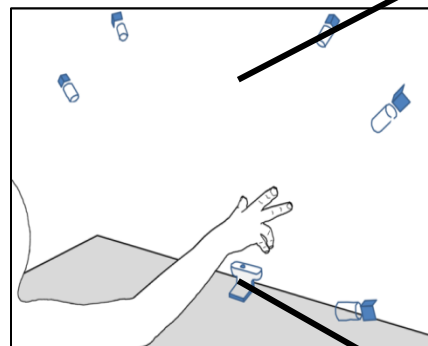
2



$$\theta_0^t = \tilde{\theta}_D^t \quad \theta_D$$

Pose Candidate Fusion

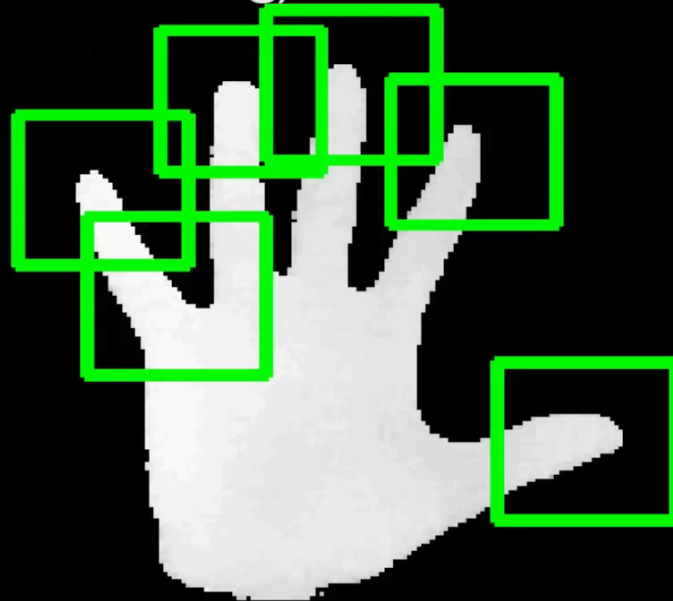
3



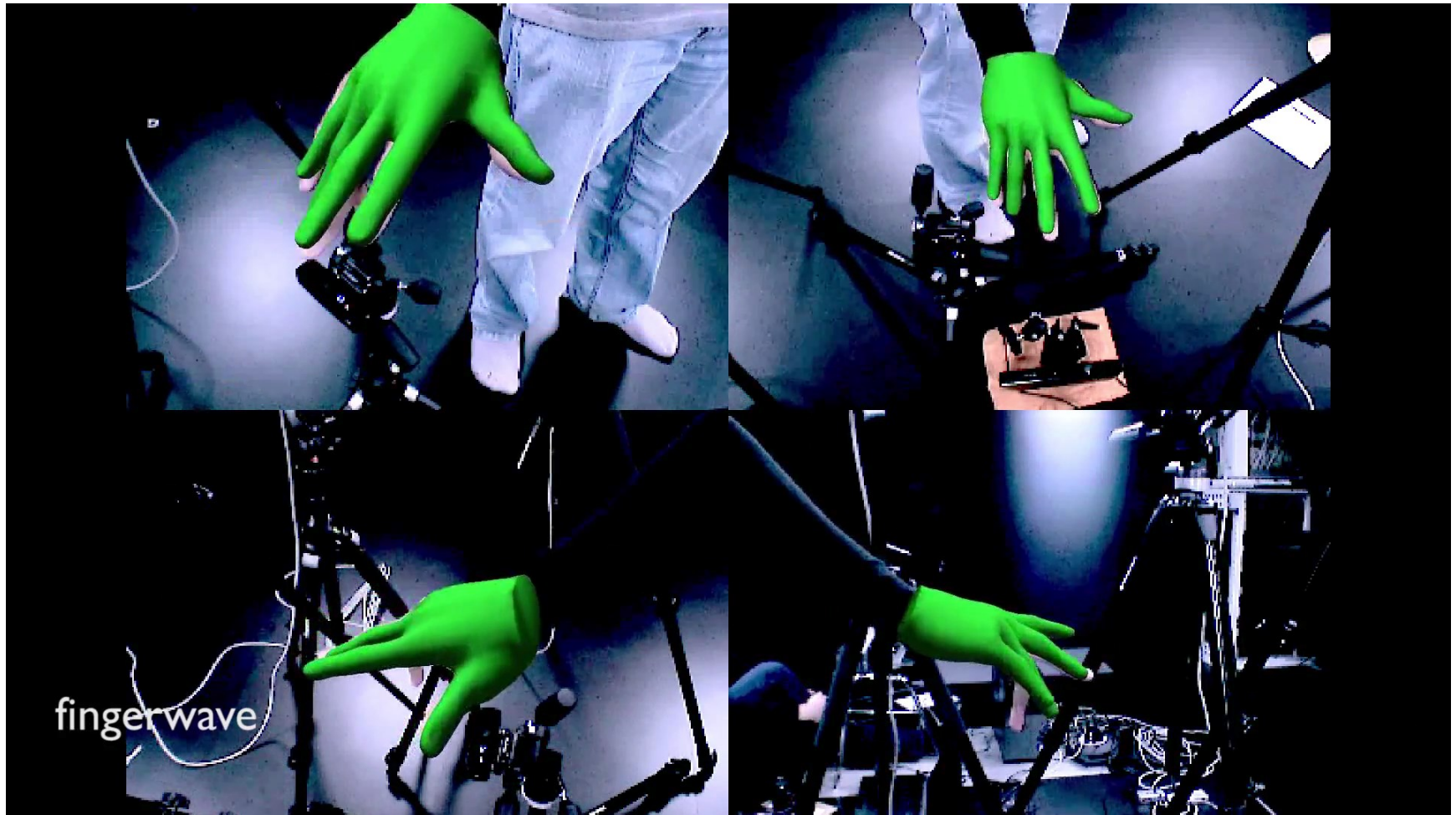
Camera Setup

Discriminative Pose Estimation

Our Approach (Before Processing)

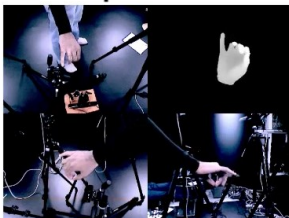


Real-time Hand Tracking



> 30 fps

Evaluation Dataset: Dexter 1

abdadd 	fingercount 	fingerwave 
flexex I 	pinch 	random 
tigergrasp 	7 Sequences 5 RGB Cameras (only 3 shown) 1 ToF Depth Camera	

- Accuracy on annotated finger tip position: ~ 13 mm

Thank You !

gvv.mpi-inf.mpg.de



www.thecaptury.com

Thanks to:



ERC Starting Grant
CapReal