# Efficient Model-based Tracking of the Articulated Motion of Hands

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#### PRORI EM STATEMENT

Track the 3D position, orientation and full articulation (26 DoFs) of a human hand that possibly manipulates an object, given a sequence of either multiview or RGB-D frames of the scene.

#### **MOTIVATION**

The markerless tracking of hand articulations is a challenging problem with diverse applications such as H.C.I., understanding human grasping, robot learning by demonstration, etc.

### MAIN IDEA

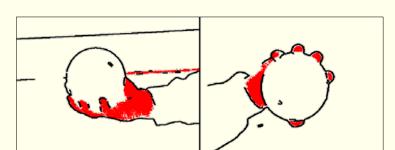
Jointly consider the observed scene: extract full-image features and produce full hypotheses about it. Compare hypotheses and observed features in parallel [4]. Use these scores to drive an iterative optimization process using Particle Swarm Optimization (PSO) [5].

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#### Input



- Frames are acquired by a multi-camera setup or a Kinect.
- Edge (black) and skin color (red) cues are extracted for the multiview case.



• For the case of Kinect, skin color and depth cues, along with the **temporal continuity** assumption are used to **segment** the hand.



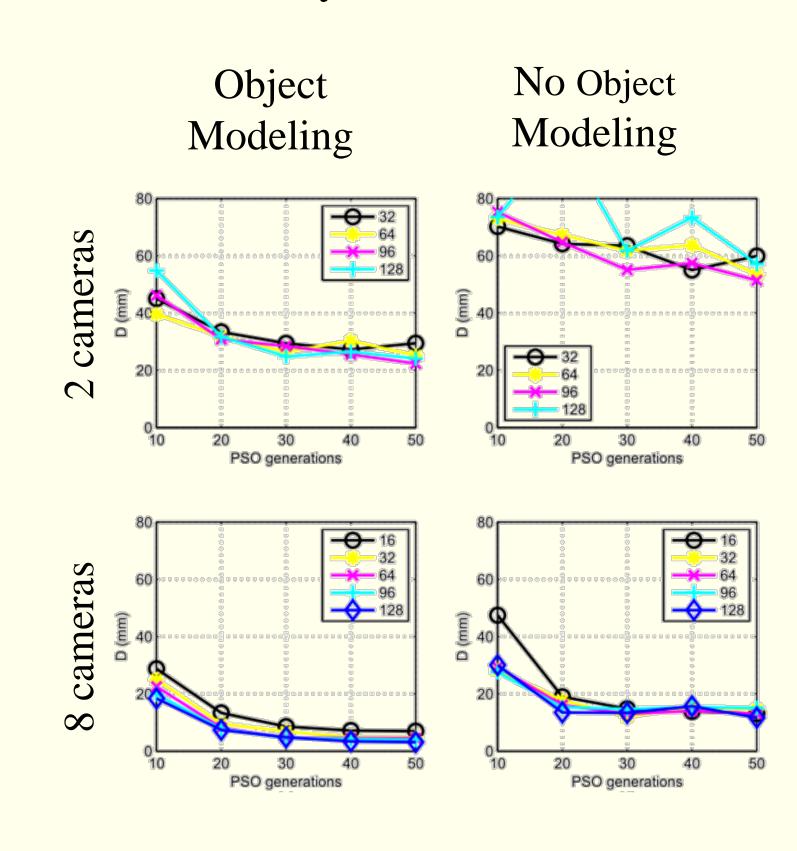
#### Fit model to data

- Employ a parametric hand model [1].
  □ Comprised of 15 cylinders and 22 spheres.
  □ 26 DoFs: 6D global pose, 20 kinematics angles.
  + For the hand-object case, add a parametric object (9 or 10 DoFs).
- From a full configuration (all 26 DoFs of the hand model plus potentially the object DoFs), a **skin occupancy map**, an **edge map** and a **depth map** can be **synthesized** by means of rendering.
- These maps are used to quantify the discrepancy between observation and hypothesis (objective function).
- The objective function also **penalizes physically implausible** configurations (hand-hand and hand-object collision checking).
- A variant of the **PSO** method [5] searches in the model parameter space for the best scoring configuration.
  - ☐ Efficient evaluation of multiple hypotheses on the GPU [4].
- Candidate poses for the **next frame** are obtained by **perturbing** the solution of the **previous frame**.

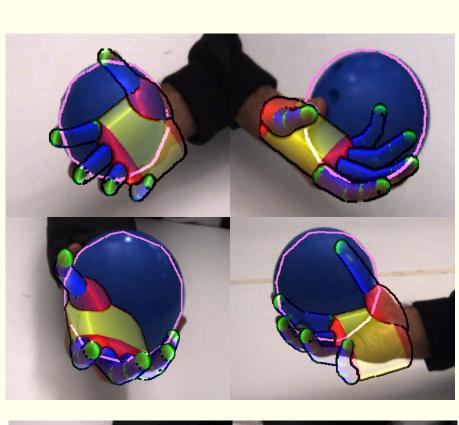
# EXPERIMENTAL RESULTS

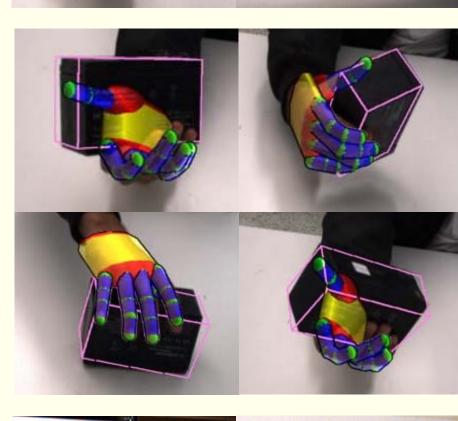
Input from a Multi-view System

Quantitative evaluation on synthetic data



64 particles and 40 generations for 4 views yield 2fps on a modern PC







Input from Kinect

Single-view depth image Hand in isolation



64 particles and 30 generations yield
15fps on a modern PC

# STRENGTHS OF THE APPROACH

rendering

- Occlusions serve as visual cues through modeling.
- Joint optimization: no simplifying assumptions over the problem structure, simultaneous consideration of all parameters.
- Careful design and exploitation of parallelism in a GPU implementation [4] lead to a computationally efficient system that accepts input of multiple modalities [1-3].
- Minimally invasive markerless approach.

# KEY REFERENCES

- 1. Oikonomidis, I., Kyriazis, N., Argyros, A. A. "Markerless and Efficient 26-DOF Hand Pose Recovery". *ACCV*, 2010.
- 2. Oikonomidis, I., Kyriazis, N., Argyros, A. A. "Full DOF Tracking of a Hand Interacting with an Object by Modeling Occlusions and Physical Constraints". *ICCV*, 2011.
- 3. Oikonomidis, I., Kyriazis, N., Argyros, A. A. "Efficient Model-based 3D Tracking of Hand Articulations using Kinect". *BMVC*, 2011.
- 4. Kyriazis, N., Oikonomidis, I., Argyros., A. A. "A GPU-powered Computational Framework for Efficient 3D Model-based Vision". *Technical Report TR420, ICS-FORTH*, 2011.
- 5. Kennedy, J., Eberhart, R. "Particle swarm optimization". *International Conference on Neural Networks*, 1995.



reduction

AND Operation

generation

