# Motion Capture Databases Discussion of Open Problems

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### mocap.cs.cmu.edu

6 hours of motion capture data

asf/amc format Collected with the Helen Hayes Marker set (41 markers) Corresponding video



### mocap.cs.cmu.edu

- 42% locomotion: 5% jumping, 3% running, and 33% walking
- 29% common scenarios: cleaning, waiting, gestures
- 16% physical activities: basketball, boxing, dance, exercise, golf, martial arts
- 7% interacting with the environment: rough terrain, playground equipment
- 6% two subjects interacting

Unique performances: pratfalls from a clown, break dancing, and several modern dance performances



### What was it used for?

As of May 2005, over 40 technical papers published in graphics, vision, activity recognition, biomechanics, databases using mocap.cs.cmu.edu (with no affiliation to CMU). Most common area was activity recognition

In May 2005, averaging 2000 accesses/month by unique IPs (after an attempt to cull for web crawlers)

### What is missing?

Almost everything there is a "performance" and doesn't necessarily reflect "natural" human activities

Few behaviors involve interaction (with people or with objects other than the floor)

Hand and face motion was not recorded

Degrees of freedom in back/shoulder were inadequate

Video not calibrated or carefully synchronized Methods for access were very simple

# Kitchen Capture

Database of kitchen activities recorded using many different sensors:

- Motion capture with improved skeleton, hopefully hands
- Cameras (in room and on body)
- Microphones (in room and on body)
- Accelerometers
- Occasional use of other sensors: ground forces, object interaction forces, EMG, eyetracker

Long Captures: full meals cooked and consumed



# Why the Kitchen?

#### Complex manipulation and mobility tasks

- Requires an understanding of object manipulation and interaction
- Record forces and hand motion

#### Long capture of natural activities

• Mitigate effect of laboratory capture

# Why the Kitchen?

Food preparation, consumption, and nutrition central to assistive technology applications

- Aging in place
- Obesity
- Accidents in the home
- Job coaching

## What data are we recording?

Training data: labels/ground truth Hand-labeled semi-automatically Tests: unlabelled data

Record long captures—full meals (cooking and consumption) Record more modalities than will likely be necessary and record at higher quality to provide a gold standard

Synchronize data

Tightly within a modality As tightly as possible across modalities Via button switch light flashes/noise once/session



# Sensors And Data





## **Technical Challenges**

Synchronization Long recording sessions Occlusions

Are we recording the right modalities with sufficient resolution in time and space? Your input is welcome: <u>ikh@cs.cmu.edu</u>

## **Open Problems**

Dimensionality reduction Data representation Poses or motion segments Data compression Segmentation Fine or coarse grain Classification and clustering Retrieval Full state Partial state (performance animation) Generative models Filling of gaps due to occlusions (single marker or higher level) Generation of human motion http://www.cs.cmu.edu/~christos/TALKS/SIGGRAPH-07-tutorial/