

Database-Based Hand Pose Estimation

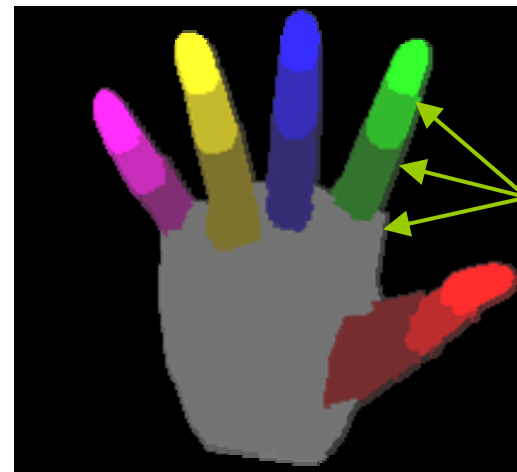
CSE 6367 – Computer Vision
Vassilis Athitsos
University of Texas at Arlington

Static Gestures (Hand Poses)

- Given a hand model, and a single image of a hand, estimate:
 - 3D hand shape (joint angles).
 - 3D hand orientation.



Input image



Joints

Articulated hand model

Static Gestures

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 - 3D hand shape (joint angles).
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Input image



Articulated hand model

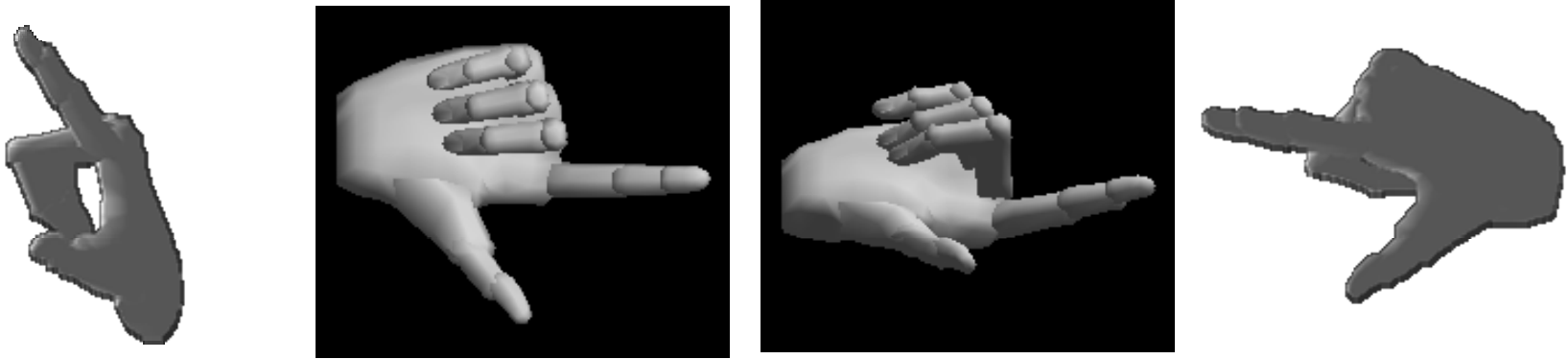
Goal: Hand Tracking Initialization



- Given the 3D hand pose in the previous frame, estimate it in the current frame.
 - Problem: no good way to automatically initialize a tracker.

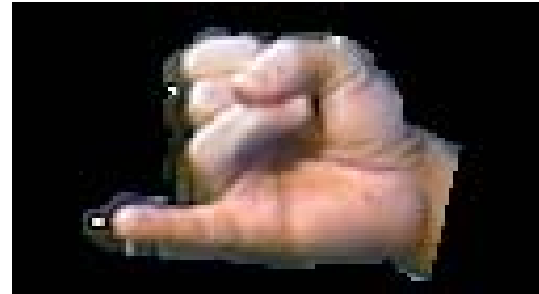
Rehg et al. (1995), Heap et al. (1996), Shimada et al. (2001),
Wu et al. (2001), Stenger et al. (2001), Lu et al. (2003), ...

Assumptions in Our Approach



- A few tens of distinct hand shapes.
 - All 3D orientations should be allowed.
 - Motivation: American Sign Language.

Assumptions in Our Approach



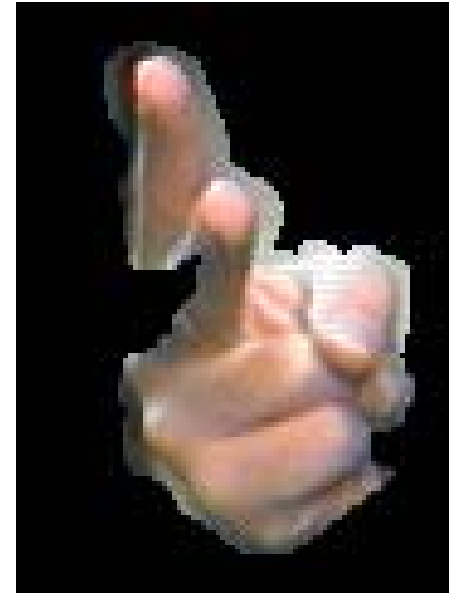
- A few tens of distinct hand shapes.
 - All 3D orientations should be allowed.
 - Motivation: American Sign Language.
- Input: single image, bounding box of hand.

Assumptions in Our Approach

input
image



skin
detection



segmented
hand

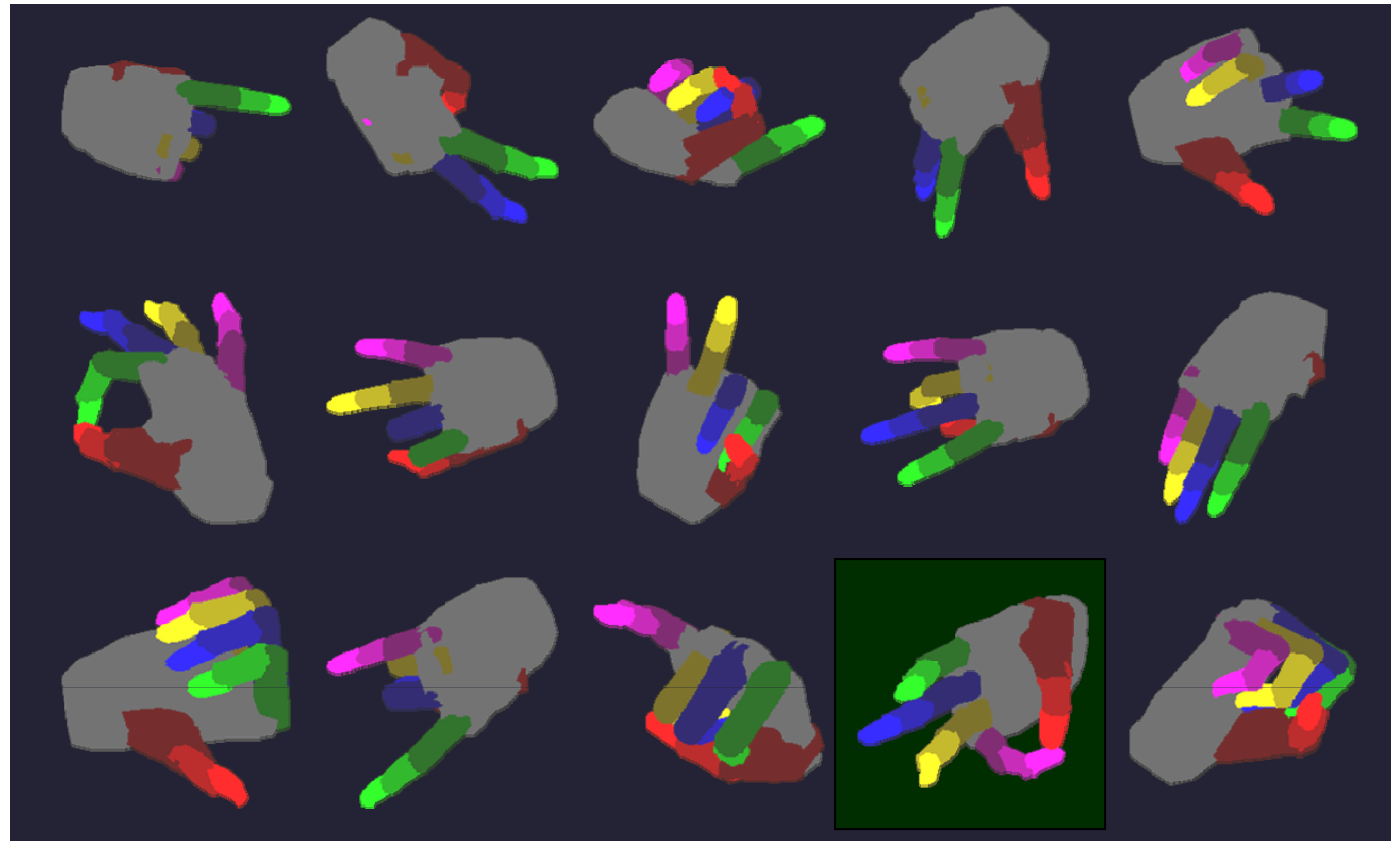
- We do not assume precise segmentation!
 - No clean contour extracted.

Approach: Database Search

- Over 100,000 computer-generated images.
 - Known hand pose.



input



Why?

- We avoid direct estimation of 3D info.
 - With a database, we only match 2D to 2D.
- We can find all plausible estimates.
 - Hand pose is often ambiguous.



input



Building the Database

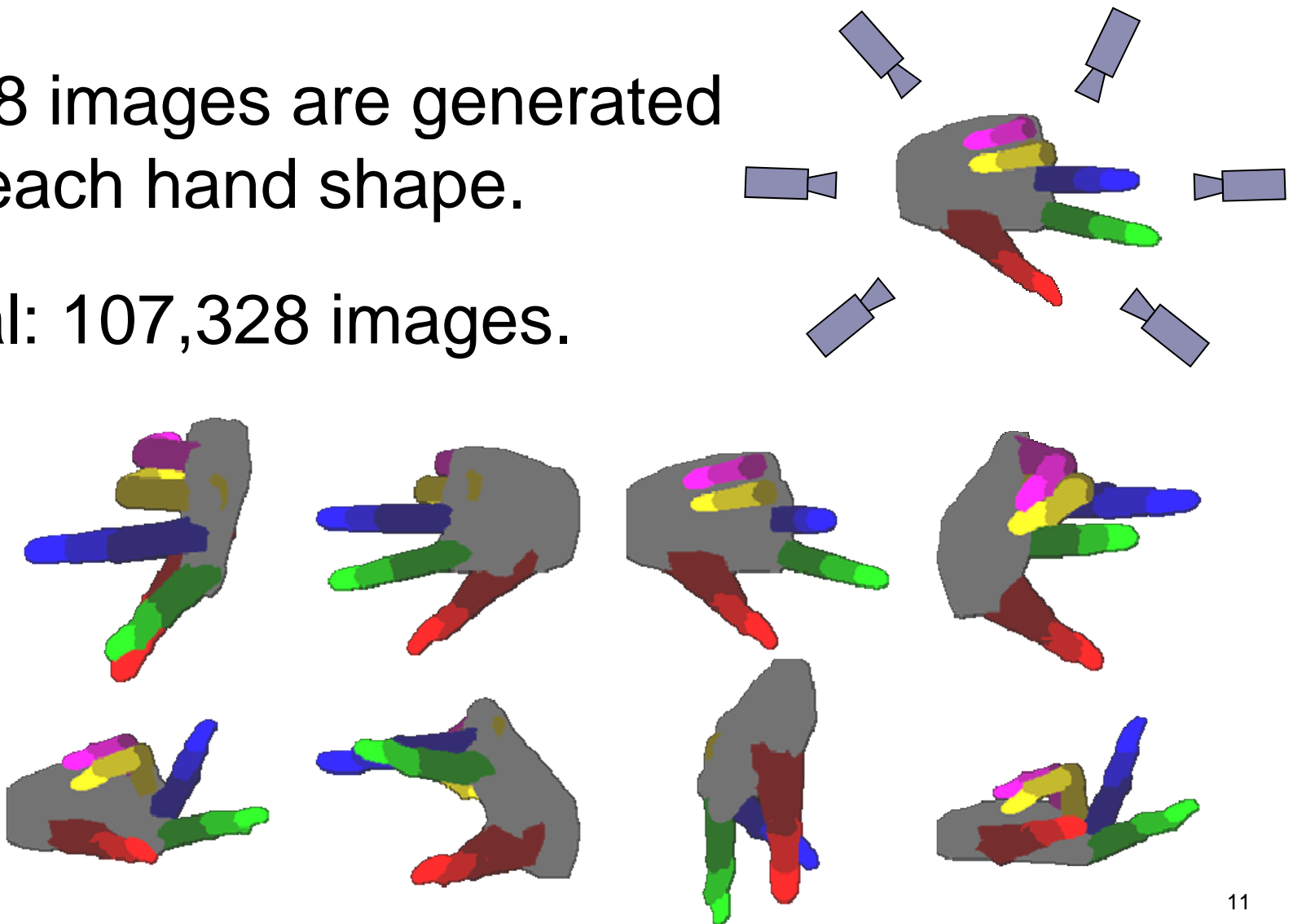
26 hand shapes



Building the Database

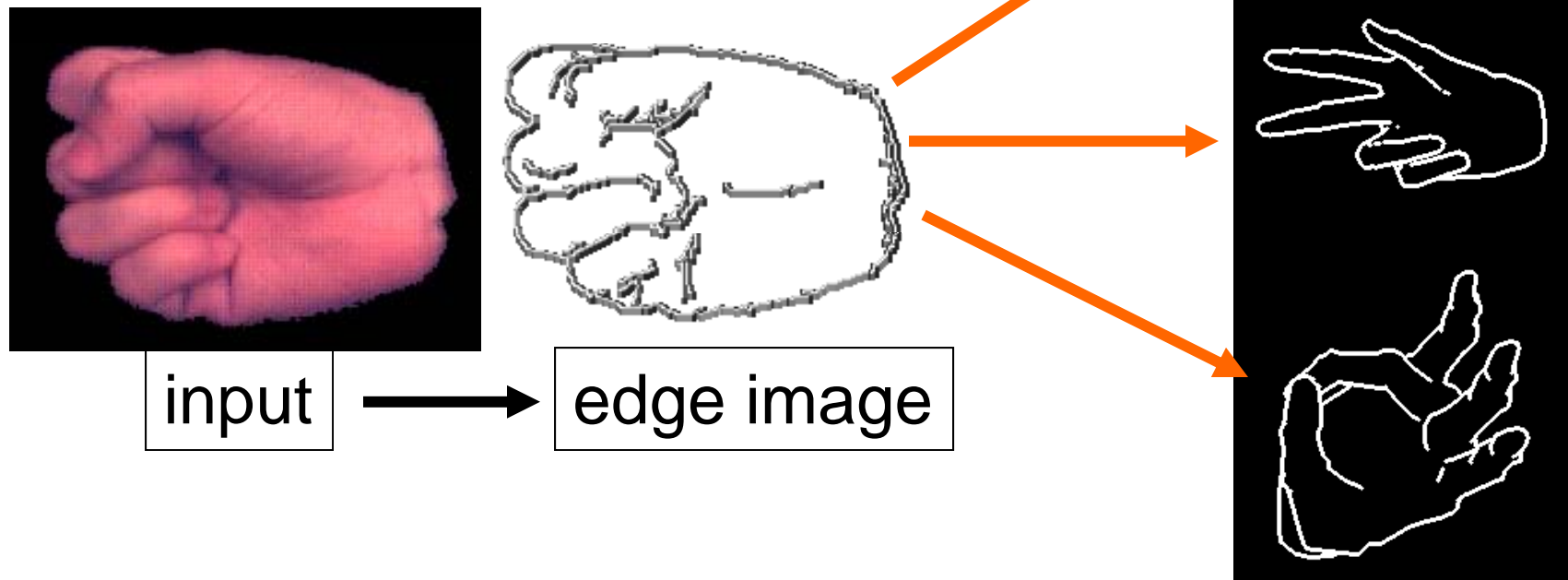
4128 images are generated
for each hand shape.

Total: 107,328 images.



Features: Edge Pixels

- We use edge images.
 - Easy to extract.
 - Stable under illumination changes.

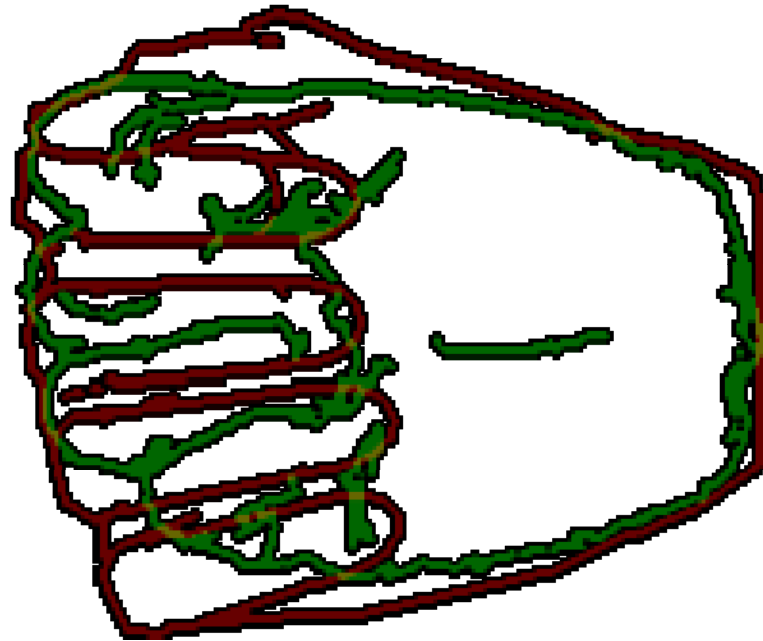
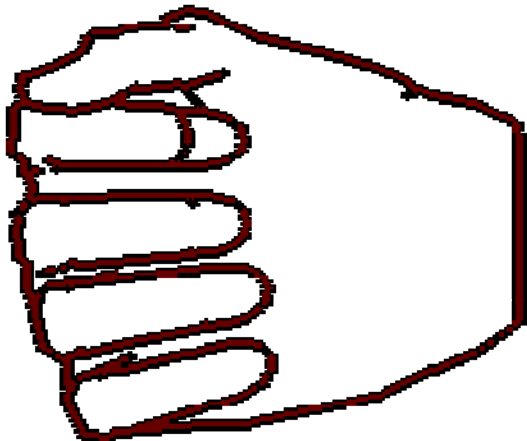


Chamfer Distance

input



model

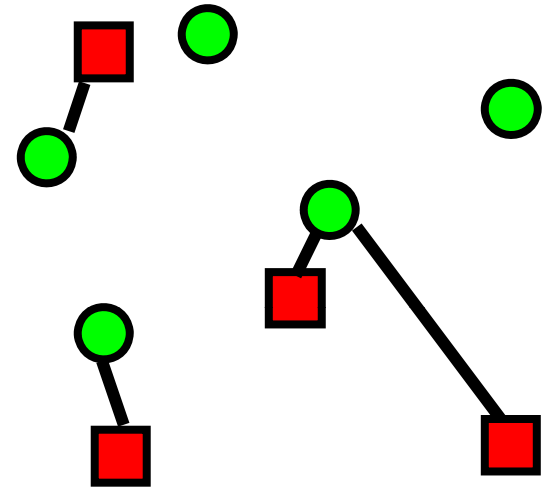


Overlaying input and model

How far apart are they?

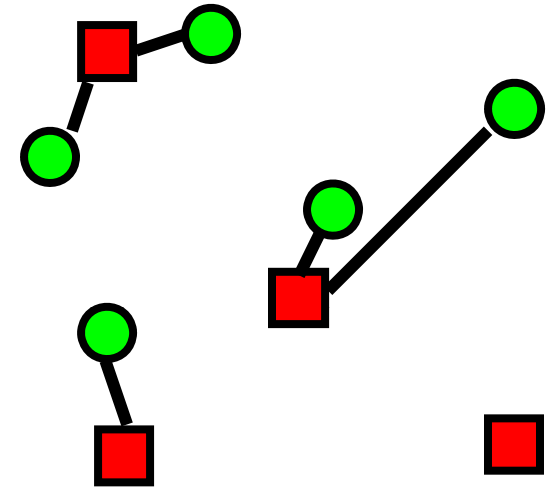
Directed Chamfer Distance

- Input: two sets of points.
 - red, green.
- $c(\text{red}, \text{green})$:
 - Average distance from each red point to nearest green point.



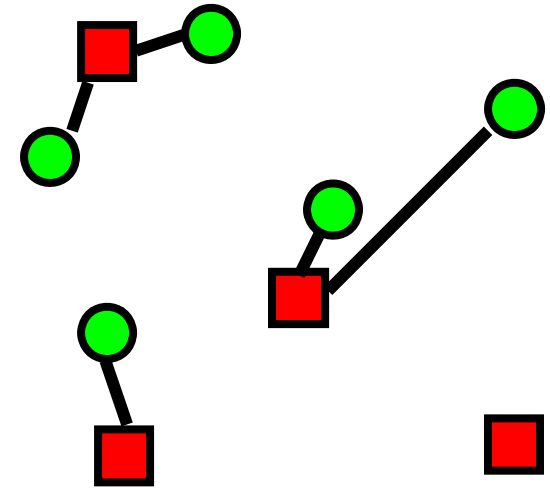
Directed Chamfer Distance

- Input: two sets of points.
 - red, green.
- $c(\text{red}, \text{green})$:
 - Average distance from each red point to nearest green point.
- $c(\text{green}, \text{red})$:
 - Average distance from each green point to nearest red point.



Chamfer Distance

- Input: two sets of points.
 - red, green.
- $c(\text{red}, \text{green})$:
 - Average distance from each red point to nearest green point.
- $c(\text{green}, \text{red})$:
 - Average distance from each green point to nearest red point.

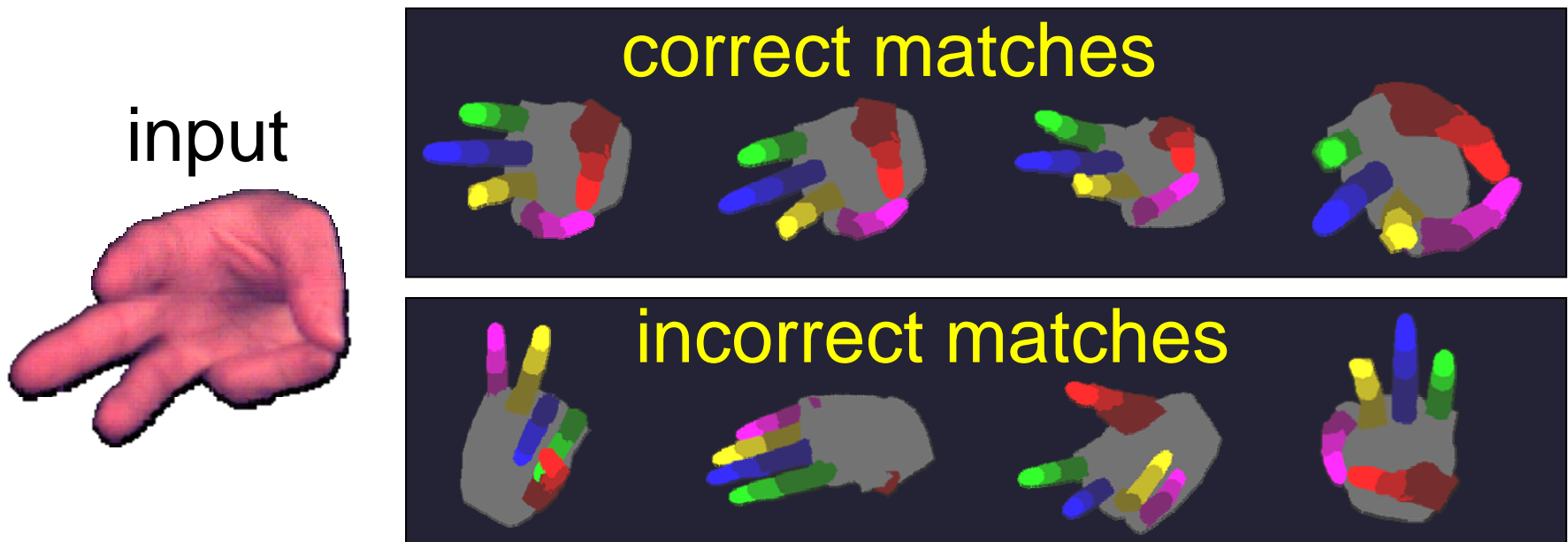


Chamfer distance:

$$C(\text{red}, \text{green}) = c(\text{red}, \text{green}) + c(\text{green}, \text{red})$$

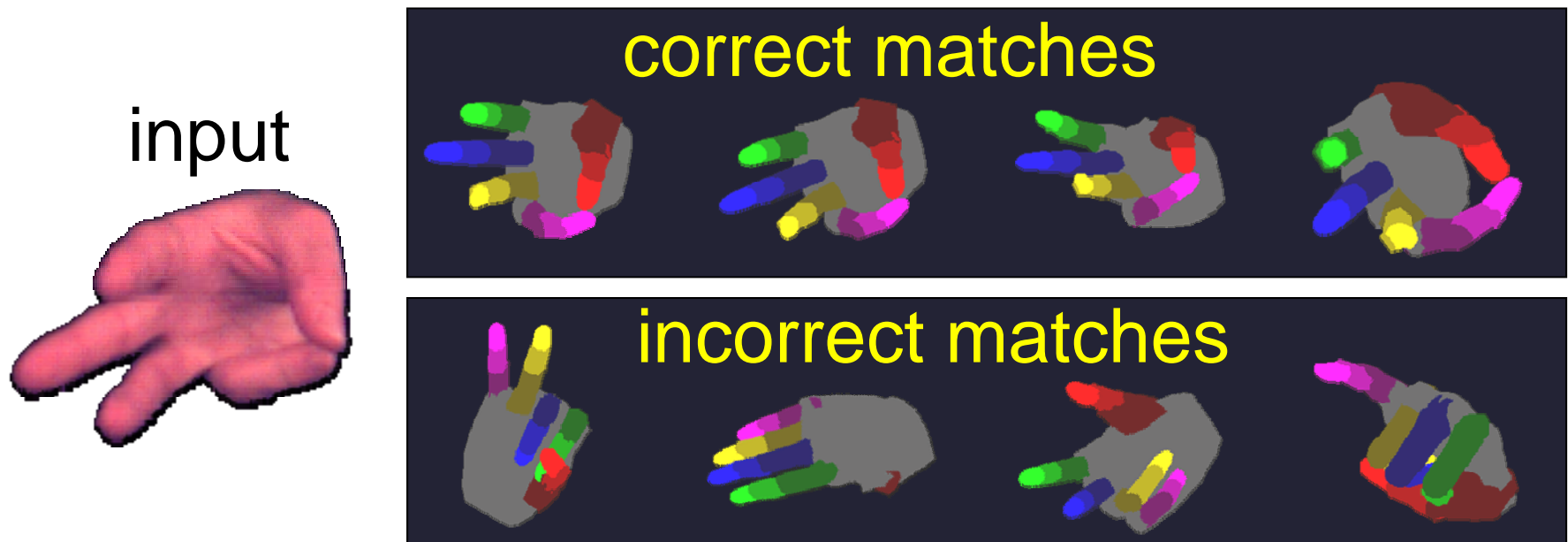
Evaluating Retrieval Accuracy

- A database image is a *correct match* for the input if:
 - the hand shapes are the same,
 - 3D hand orientations differ by at most 30 degrees.



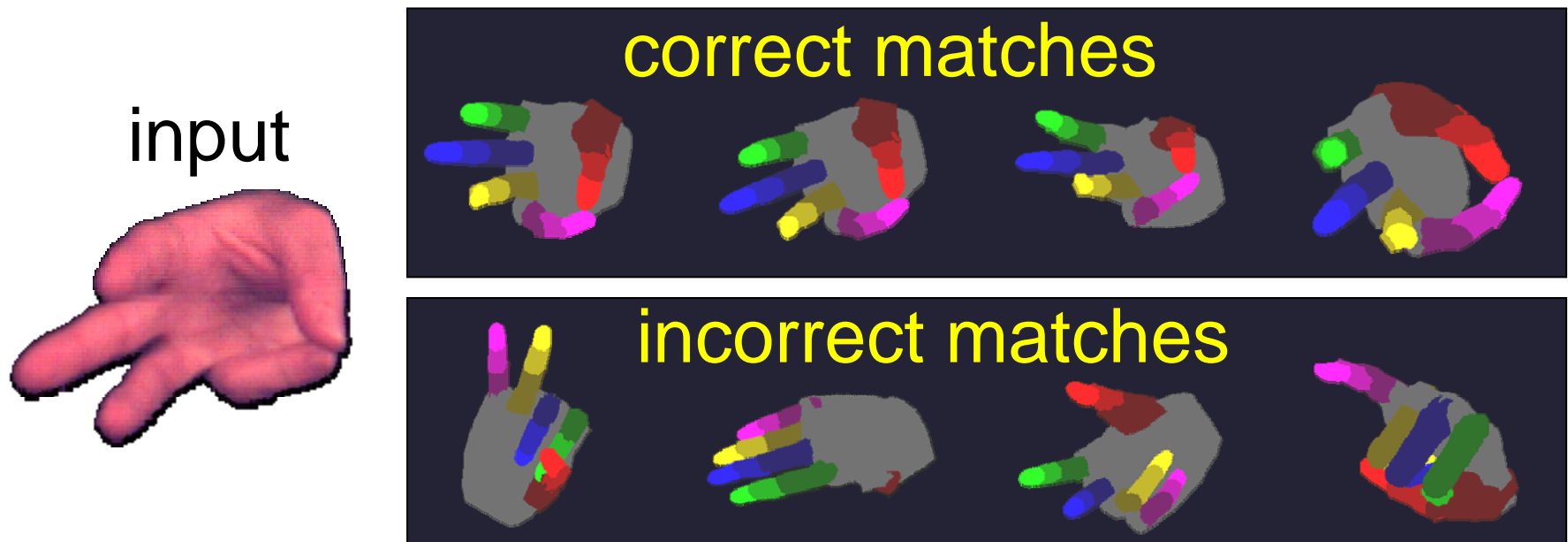
Evaluating Retrieval Accuracy

- An input image has 25-35 correct matches among the 107,328 database images.
 - Ground truth for input images is estimated by humans.



Evaluating Retrieval Accuracy

- Retrieval accuracy measure: what is the rank of the highest ranking correct match?

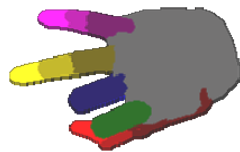


Evaluating Retrieval Accuracy

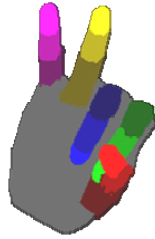
input



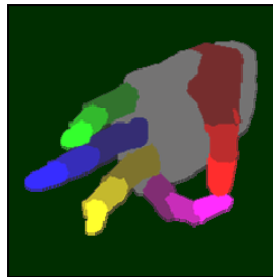
rank 1



rank 2



rank 3



rank 4



rank 5



rank 6

...

...



highest ranking
correct match

Results on 703 Real Hand Images

Rank of highest ranking correct match	Percentage of test images
1	15%
1-10	40%
1-100	73%

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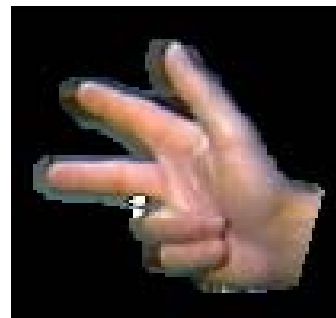
- Results are better on “nicer” images:
 - Dark background.
 - Frontal view.
 - For half the images, top match was correct.

Examples

initial
image



segmented
hand



edge
image



correct
match



rank: 1

Examples

initial
image



segmented
hand



edge
image



correct
match



rank: 644

Examples

initial
image



segmented
hand



edge
image



incorrect
match



rank: 1

Examples

initial
image



segmented
hand



edge
image



correct
match



rank: 1

Examples

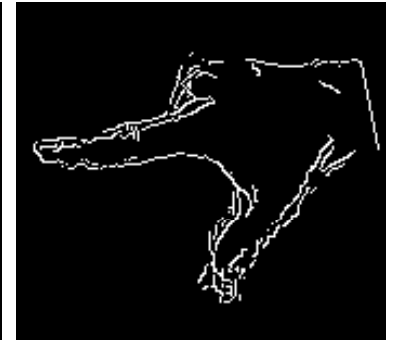
initial
image



segmented
hand



edge
image



correct
match



rank: 33

Examples

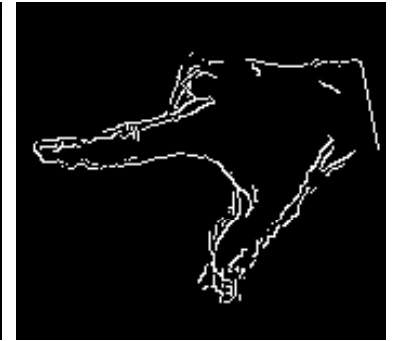
initial
image



segmented
hand



edge
image



incorrect
match



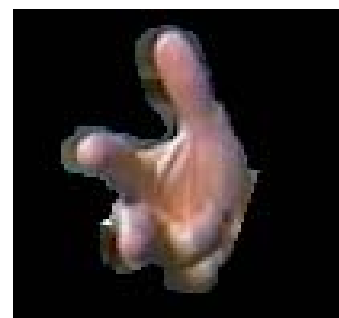
rank: 1

Examples

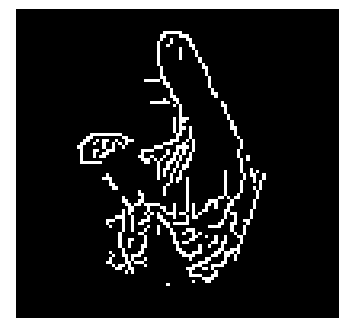
“hard”
case



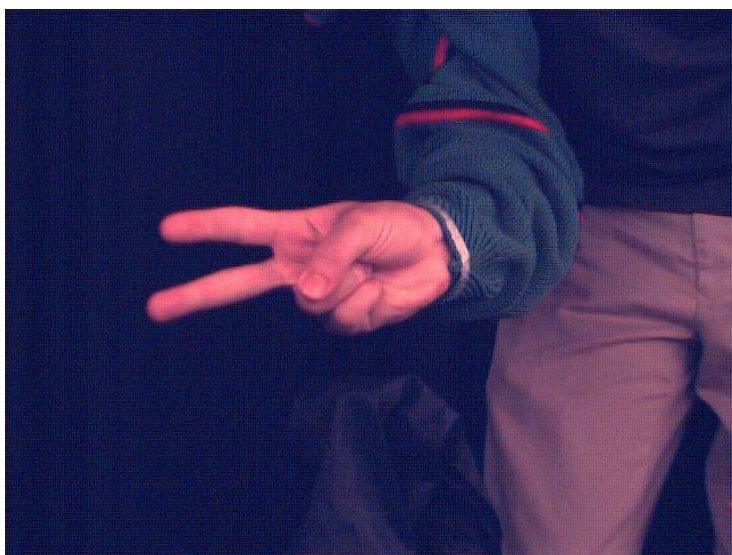
segmented
hand



edge
image



“easy”
case



segmented
hand



edge
image

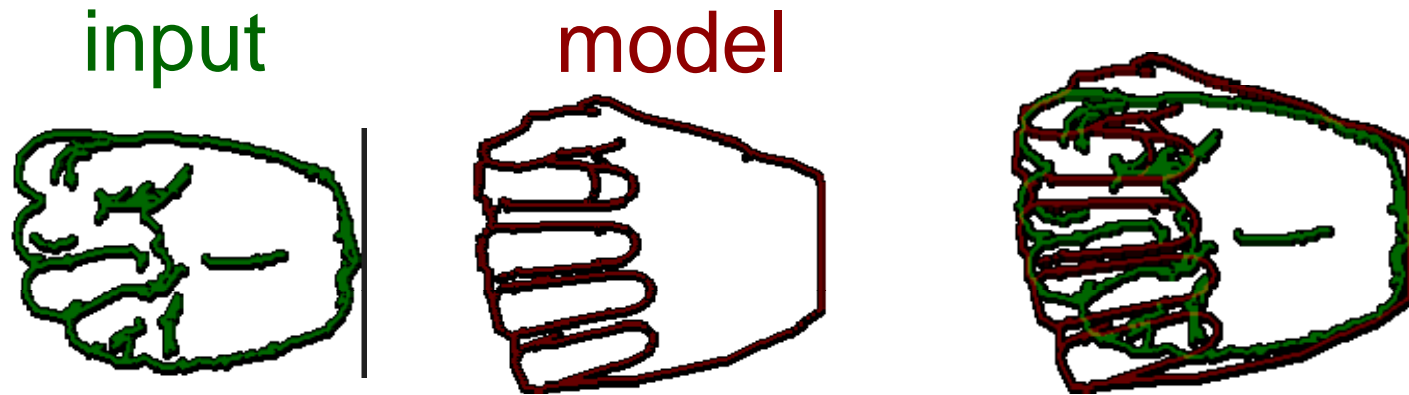


Research Directions

- More accurate similarity measures.
- Better tolerance to segmentation errors.
 - Clutter.
 - Incorrect scale and translation.
- Verifying top matches.
- Registration.



Efficiency of the Chamfer Distance



- Computing chamfer distances is slow.
 - For images with d edge pixels, $O(d \log d)$ time.
 - Comparing input to entire database takes over 4 minutes.
 - Must measure 107,328 distances.

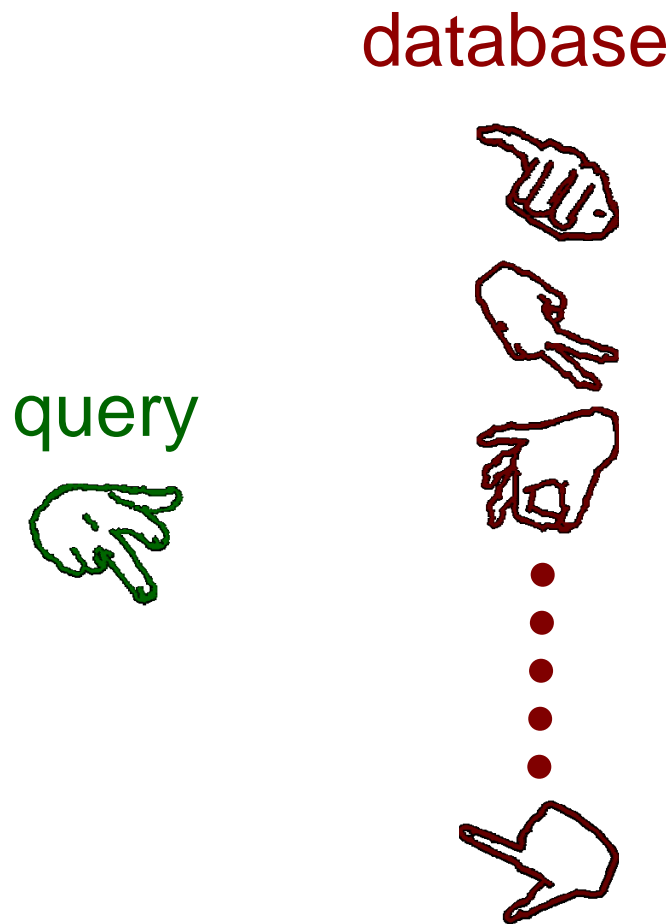
The Nearest Neighbor Problem

database

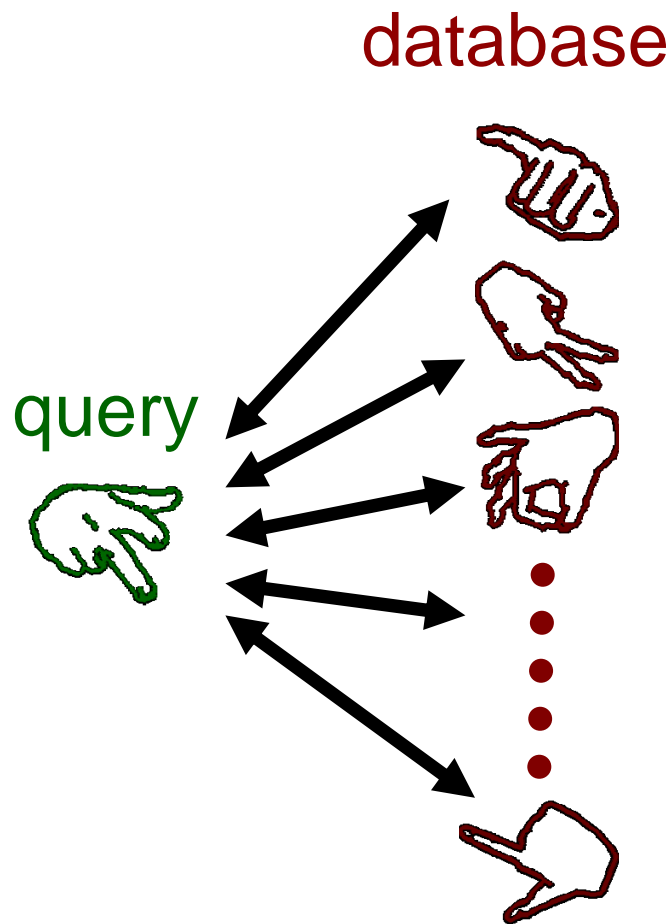


The Nearest Neighbor Problem

- Goal:
 - find the k nearest neighbors of query q .

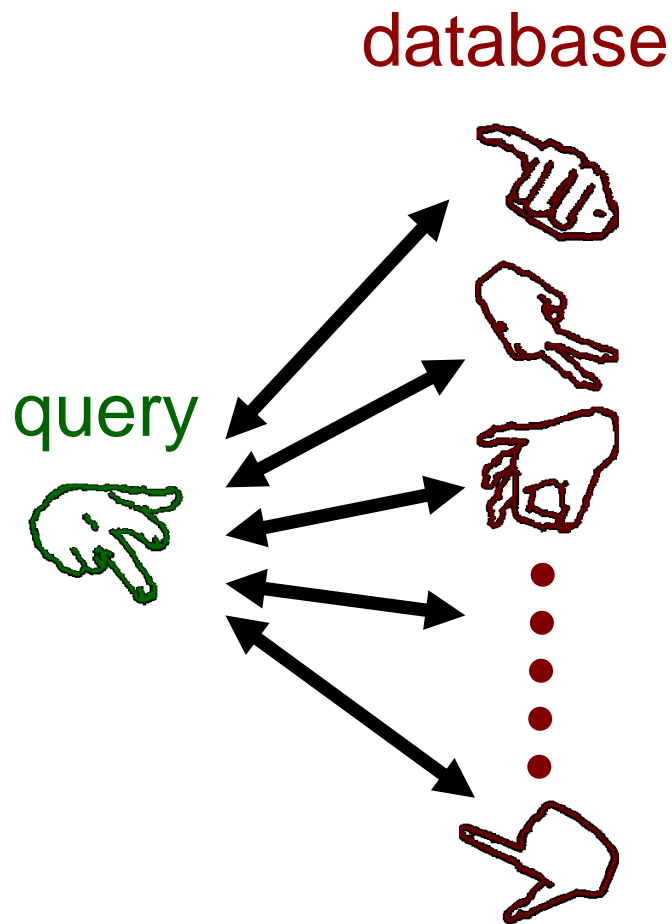


The Nearest Neighbor Problem



- Goal:
 - find the k nearest neighbors of query q .
- Brute force time is linear to:
 - n (size of database).
 - time it takes to measure a *single distance*.

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