

Hand Gesture Recognition by using Logical Heuristics

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Outline

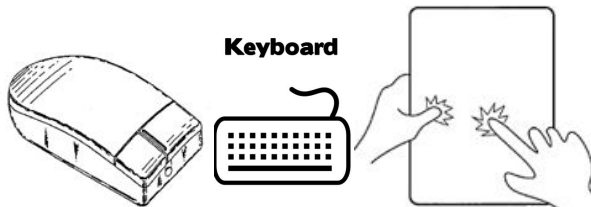
- 1 Introduction
 - Motivation
 - Existing Approaches
- 2 Our Approach
 - Overview
 - Skin Color Detection
 - Hand Gesture Recognition
- 3 Evaluation
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Motivation

Existing Computer input Interfaces

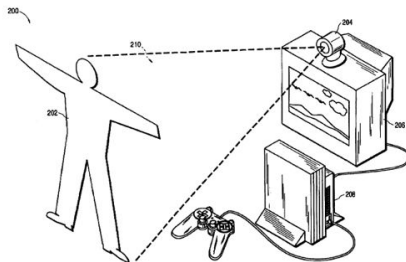


Efficient **BUT** require

- Additional equipped hardware, hence extra space
- A Physical Touch

Motivation

Emerging Category: Input by Gesture Recognition



Natural HCI **BUT** not efficient because of

- Different types of gestures in complex backgrounds
- Complicated computational procedures

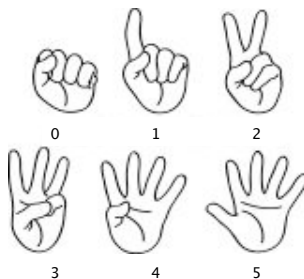
In This Paper

Aim: Natural HCI by focussing on

- Recognizing gestures in complex backgrounds
- Fast processing on standard PC and cameras

Recognize 6 different hand gestures

- No Training phase
- Simple and Fast Algorithm



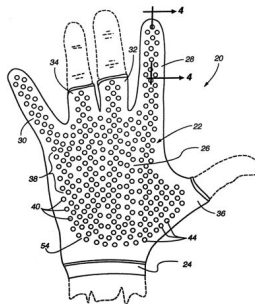
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Existing Approaches

Wearable Data Gloves (Pandit et al., 2009)

- Requires data gloves with markers to be worn to extract hand posture
 - Efficient BUT not a suitable interface
 - Complex calculations



Existing Approaches

Template Matching (Alon et al., 2005)(Stefan et al., 2008)

- Requires sets of huge training database of desired gestures to be recognized
 - Lack in response time
 - Prepare new training data for new gestures

Computer Vision Based

- 3D hand modeling
 - Uses many approximation processes
 - Complicated

Our Approach

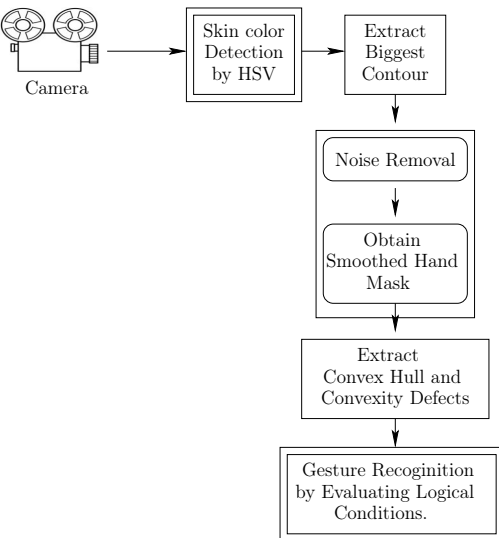
Computer Vision Based

- Suitable Interface
 - No additional wearable equipment
 - *Uses a standard digital camera*
- Computationally simple & fast
 - No training phase
 - *Uses Logical Conditions*
- Recognizes Gestures in
 - Static Conditions
 - *Background remains same (images)*
 - Dynamic
 - *Background lightening, colors etc change (videos)*

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Overview: Block Diagram



2 Major Steps

- 1 Skin Color Detection
- 2 Gesture Recognition

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Skin Color Detection

Used *HSV* (**H**ue, **S**aturation, **V**alue) color scheme

- More related to human perception (Albiol et al., 2001)
- Used classical method to detect skin pixels
 - By setting Upper & Lower bound values

$$H_{min} \leq H \leq H_{max} \begin{cases} H_{min} & 0^\circ \\ H_{max} & 20^\circ \end{cases} \quad (1)$$

$$S_{min} \leq S \leq S_{max} \begin{cases} S_{min} & 45 \\ S_{max} & 255 \end{cases} \quad (2)$$

Skin Color Detection

- Left Side (Colored)
 - Original Image
 - Houses color is similar to skin color

- Right Side (Black & White)
 - **Black**: non skin color
 - **White** dots: skin color



Skin Color Detection



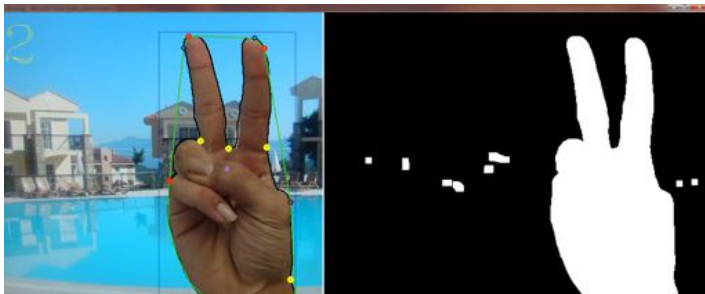
Naive & Robust BUT

- Classifies noisy objects as skin

Reduce Noise

- Assumption:
 - Biggest connected white dots area is of Human Hand

Noise Reduction & Hand Detection



Reduce Noise

- ① Take median of **H** & **S** values for each pixel of biggest contour (the hand)
- ② Compare color values of extracted skin objects whose:
 - $area < 20\%$ of the area of biggest contour
- ③ Smoothed Skin Pixel Binary Mask of a hand is obtained

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Hand Gesture Recognition

6 Gesture Types

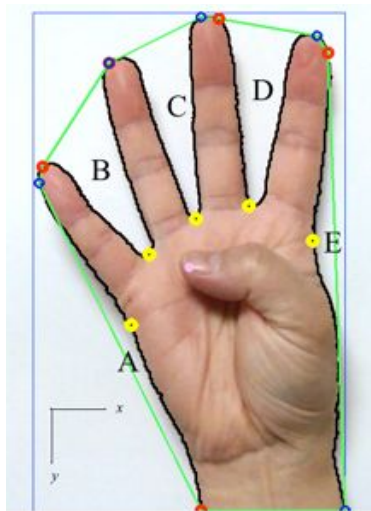
- Counting Number of fingers in a Hand (*zero - five*)
 - *zero* gesture of a closed hand

Gestures Recognition by evaluating Logical Conditions derived using

- Convex Hull
- Convexity Defects

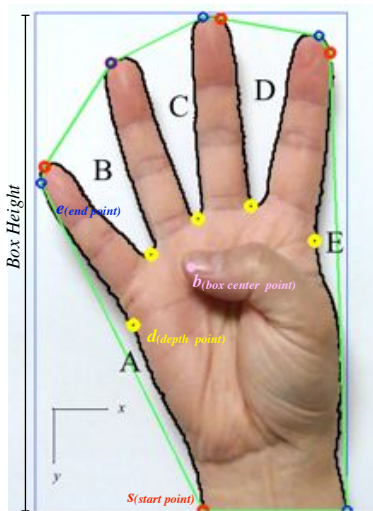
Convex Hull & Convexity Defects

- Convex Hull
 - Outer green line around the hand
- Convexity Defects
 - Holes in difference b/w convex hull & hand
 - Five convexity defects in picture
 - **A** through **E** bounded by black & green outlines



Information from Convexity Defects

- For each Convexity defect
 - Start Point s_x
 - Depth Point d_y
 - Box Center Point b_y
 - Length of Defect l_d
 - Box Height



Heuristics towards Gestures Recognition

For each Convexity defect c_d

$$\text{count} = \begin{cases} 1 & \text{if } (s_y < b_y \text{ or } d_y < b_y) \text{ and } (s_y < d_y) \text{ and } l_d > \frac{\text{box height}}{n} \\ 0 & \text{otherwise} \end{cases} \quad (3)$$

$$l_d = \sqrt{(s_x - d_x)^2 + (s_y - d_y)^2} \quad (4)$$

$$\text{num of fingers} = \sum_{c_d \in C_D} \text{count} \quad (5)$$

- In Equation (3)
 - **First condition:** Checks if convexity defect is of a straight finger
 - **Third condition:** To filter momentary convexity defects caused due to hand rotation

Recognized Gestures of Different Forms

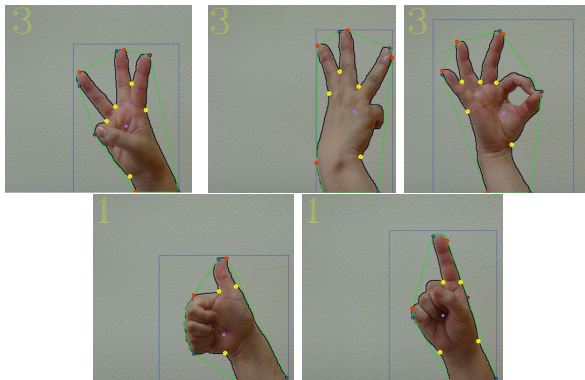


Figure: Different Forms of Three & One Counts

Recognized Gestures in Different Orientation

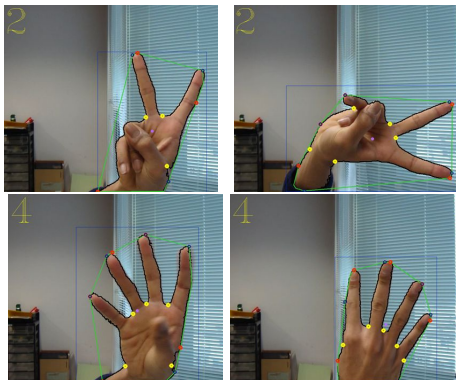


Figure: Counts of Two & Four in Different Orientation in Complex background

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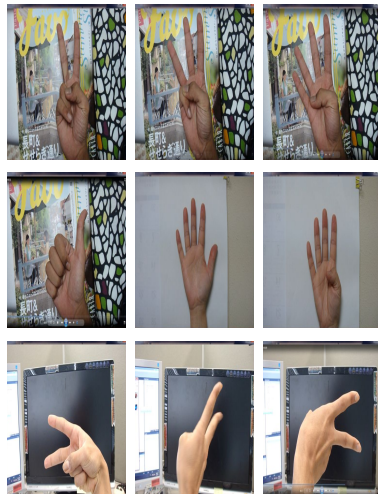
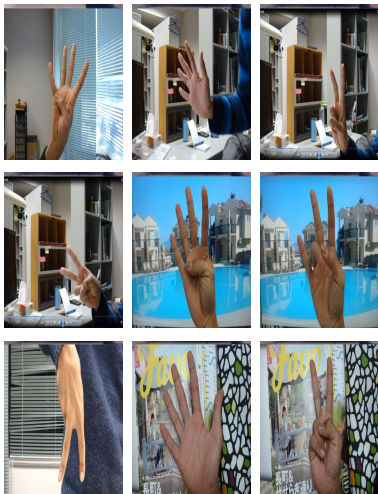
Data

- Development Data D_d
 - To Design Heuristics
- Evaluation Data D_e
 - To Evaluate Performance

Data	Total num. of Images	Images in Complex Backgrounds
D_d	60	20
D_e	90	60
Total	150	80

- Front Facing Camera
 - 8 MegaPixels Sony Digital
 - 1280x720 Resolution
 - 29.970 FPS
- Each image consists of a gesture from one hand

Data Samples



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Results on Development & Evaluation Data

G	Development D_d			Evaluation D_e			$D_d + D_e$		
	#I	P	R	#I	P	R	#I	P	R
0	10	0.70	0.70	15	0.64	0.47	25	0.67	0.56
1	10	0.75	0.60	15	0.55	0.40	25	0.63	0.48
2	10	0.78	0.70	15	0.70	0.47	25	0.74	0.56
3	10	0.89	0.80	15	0.73	0.53	25	0.80	0.64
4	10	1.00	0.90	15	0.75	0.60	25	0.86	0.72
5	10	1.00	0.90	15	0.82	0.60	25	0.90	0.72
Total	60	0.85	0.77	90	0.70	0.51	150	0.77	0.61

G=Gesture Type, I=Num of Images

$$P = \frac{\text{Num. of correctly detected}}{\text{Total num. detected}}, \quad R = \frac{\text{Num. of correctly detected}}{\text{Total num of corrects}}$$

Discussion

G	Development D_d			Evaluation D_e			$D_d + D_e$		
	#I	P	R	#I	P	R	#I	P	R
0	10	0.70	0.70	15	0.64	0.47	25	0.67	0.56
1	10	0.75	0.60	15	0.55	0.40	25	0.63	0.48
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Total	60	0.85	0.77	90	0.70	0.51	150	0.77	0.61

- P & R on D_e are worse than D_d
 - Gesture Detection in D_e is more difficult because of more complex backgrounds

Data	Total num. of Images	Images in Complex Backgrounds
D_d	60	20
D_e	90	60

Discussion

G	Development D_d			Evaluation D_e			$D_d + D_e$		
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5	10	1.00	0.90	15	0.82	0.60	25	0.90	0.72
Total	60	0.85	0.77	90	0.70	0.51	150	0.77	0.61

- P & R increases towards higher finger counts
 - Hand spread is much wider
 - Therefore, logical evaluation and hand detection are relatively accurate

Overall Performance

G	Development D_d			Evaluation D_e			$D_d + D_e$		
	#I	P	R	#I	P	R	#I	P	R
0	10	0.70	0.70	15	0.64	0.47	25	0.67	0.56
1	10	0.75	0.60	15	0.55	0.40	25	0.63	0.48
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5	10	1.00	0.90	15	0.82	0.60	25	0.90	0.72
Total	60	0.85	0.77	90	0.70	0.51	150	0.77	0.61

- 77% Precision at 61% Recall from
 - 150 Gestures, 53% in complex/colorful backgrounds
- **Computational Cost**
 - OpenCV Library, Windows CPU 2 GB RAM
 - Usage < 20%
 - 1 min HD video clip with 15 gestures took
 - 1 min 2 secs

Recognition in Different Conditions

- Colorful Background



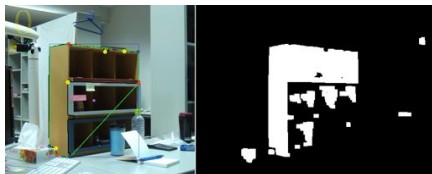
- Poor Lightening



Sample Outputs

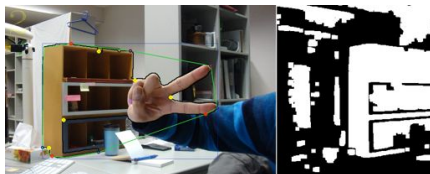
Difficulties with Skin Color & Hand Detection

- Large Skin Color Objects



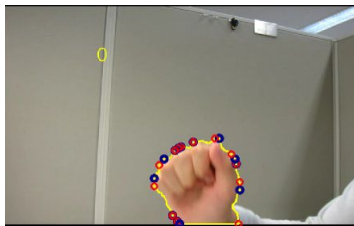
- Incorrect hand detection

- Combined contour of wooden box and hand is extracted

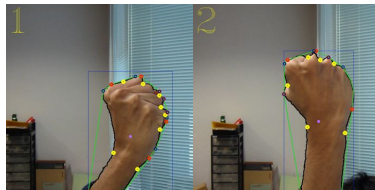


Difficulties with Logical Conditions

- Gesture *zero* (closed hand) count produces no convexity defects



- Incorrect gesture recognition of count zero
 - Inclusion of arm causing formation of convexity defects



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Conclusion

- Proposed heuristics derived from convex hull and convexity defects to recognize hand gesture.
 - Computationally fast and simple
 - Recognizes 6 hand gestures in complex backgrounds
 - Orientation & Rotation Free
- **Future**
 - Better Hand Detection Technique to bring robustness to system
 - Derive robust logical conditions towards more complex gestures
 - Full Body Posters (Running, Walking)

Finish

Thank you for your attention.

References



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Jonathan Alon, Vassilis Athitsos, Quan Yuan, and Stan Sclaroff: Simultaneous localization and recognition of dynamic hand gestures. In Proceedings of IEEE Workshop on Motion and Video Computing, vol.2, pp.254-260 (Jan, 2005).



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Pandit, A. Dand, D. Mehta, Sabesan, S. Daftery, A.: "A Simple Wearable Hand Gesture Recognition Device Using iMEMS," Soft Computing and Pattern Recognition, 2009. SOCPAR '09. International Conference of , vol., no., pp.592-597, 4-7 (Dec. 2009)

Backup

- Focus on gestures from standard camera
 - Laptop
 - Smartphone etc
- Focus on robust heuristically define the gestures
- Object of interest
 - Hand usually takes up the largest area
- Delivers high precision in plain backgrounds
 - Needs work on complex backgrounds
- Manually annotated the gestures