Analysis on Skin Colour Model Using Adaptive Threshold Values for Hand Segmentation

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Abstract—The hand gesture recognition system is the hottest topic for the human-machine interaction and computer vision fields. The hand gesture recognition system is still a challenging research area in computer vision for human-computer interaction because of various device conditions, various illumination effects, and very complex background. The recognition of hand gestures used in various application areas: such as sign language recognition, man-machine interaction, human-robot interaction, and intelligent device control and many other application areas. The robust detection of hand in hand gesture recognition system has become a challenging task due to clutter background, dynamic background, and various illumination conditions in real-world conditions. Segmentation is the partitioning/separating the foreground hand region from the background region in an image. Segmentation is also pre-processing steps of the hand gesture recognition system. The recognition accuracy will increase if the hand region correctly detected. So, hand region detection is the main important step for the hand gesture recognition system.

Index Terms—Hand Detection, Hand Gesture Recognition, Colour Segmentation, Human Computer Interaction.

I. INTRODUCTION

Hand gesture segmentation is the hottest topic in man-machine interaction and computer vision. Segmentation is the process of separating the background and foreground. The objective of hand segmentation is to find out the hand region from the images region or no-hand region. There are many hands segmentation techniques such as threshold-based, edge-based, region-based, pixel-based, model-based, ANN-based, watershed-based, and clustering-based and many other segmentation methods. In all of these techniques, the skin colour threshold based method is simple and also common used method in state-of-art research. But, skin colour thresholds based method remains challenging when background contains the other objects like the human skin colour. Firstly, in this paper, YCbCr colour threshold based approach is implemented. YCbCr colour space has good characteristics such as colour independent each colour components, good colour clustering and simple to calculate. The main advantages of skin colour based segmentation are (i) it is easy and fast to implement in training and testing. (ii) It is not dependent on the distribution of neighbour colour pixels. The strength of hand segmentation for hand gesture recognition can be reducing the amount of data or the dimension of data values. In this paper, the authors are experimented with various threshold values on YCbCr colour space to detect the human skin.

The main challenges that need to solve as research issue include:

- Illumination condition is the most sensitivity for hand gesture recognition of vision based system.
- Complex backgrounds, dynamic backgrounds are also main difficult for hand gesture recognition.
- The different size of the user’s hand is another issue.
- The multiple gestures in the same background and different viewpoints.
- The several persons contains in the sense other than the real subjects.
- The occlusion problem can also occur for the hand gesture recognition problem.
- Devices are also important to capture the images because different cameras have the different output for the same images.
- In dynamic gestures, the movement of hand can degrade the quality of images due to the blurring conditions.

The motivation of our research is to find out the best threshold values of YCbCr colour space for human skin detection. YCbCr color space is the best colour model than HSV and RGB in colour clustering range and colour independent for skin segmentation [9,17,18,21]. According to the previous literature, YCbCr colour space is the best suitable colour space for human skin detection. But, the colour spaces and threshold values are not fixing.
because different taking image devices, skin tone, illumination changing, blurring condition due to the movement object, shadow and many other conditions.

The general steps of hand gesture recognition system describe the following Fig 1. In Pre-Processing step performs image resizing, image deblurring, image enhancement and so on. In hand region detection, extract the hand from the background region by using human skin colour, background subtraction method, foreground extraction and many other methods. In feature extraction step, extract the hand gesture features by using various deep learning pre-train models; supervise learning and unsupervised learning methods. The most common features are colour features, edge features, shape features, orientation features, location features and many other specific features.

The rest of the paper is described as following: Section II describes the detail explanation of the literature survey and Section III presents the ways of data collection for hand gestures. Section IV describes about the datasets. Section V explains the YCbCr colour model. The experimental results are described in section VI. The last section describes the conclusion and future work of the research.

![Fig.1. General Hand Gesture Recognition Steps](image)

**II. RELATED WROKs**

Although many researchers have solved the issues of hand segmentation from the very clutter background, the robust algorithm is still needed to reconcile in any conditions such as various illumination conditions, dynamic background, static clutter background, etc. Shaik et al [1] implemented YCbCr colour thresholds based approach. In this paper, the authors applied the threshold values between 150 & 200 for chrominance component Cr and the chrominance components Cb are between 100 & 200. But, the threshold value is not suitable to detect the hand in NUS hand pose dataset I. McBride et al (2019) proposed skin segmentation/ detection algorithms to recognize the hand gesture of the hand gesture recognition system. The authors used YCbCr skin thresholding based algorithm, RGB-H-CrCb skin thresholding based algorithm and KNN skin classification algorithm. In this proposed study, the author used chrominance component Cb and Cr between values 80 & 120 and 133 & 173. The authors tested the segmented results on the three different particular images and compare these three different algorithms. But, these threshold values for CbCr are not suitable to segment the hand in NUS Hand Pose datasets I [2].

Nguyen et al (2018) introduce the new hand segmentation by combining the Mask R-CNN and the tracking. The original Mask R-CNN can’t detect the hand when the blur condition due to the motion. The author’s proposed method produced a better result than the original Mask R-CNN [3].

Zhang et al (2018) implemented novel hand segmentation by combining saliency detection and skin colour detection. The implemented method is more robust than state-of-art research in complex background [4]. Rahmat and his partners (2016) proposed skin colour segmentation by combining three different colour spaces such as Normalized RGB, HSV and YCbCr colour space. The skin colour detection accuracy is more robust than the single colour spaces. The recognition accuracy was obtained 91.05% on two different datasets (ECU and HGR) [5].

Thakur and his friends (2011) study that novel skin colour detection model by combining RGB, HS and CbCr and named as RGB-HS-CbCr to detect the human face. The proposed model can successfully detect the human skin regions and human non-skin regions [6]. Yingxin et al (2016) presents a robust hand gesture recognition method via convolutional neural network. In this paper, the authors firstly pre-process on the hand gesture input by using canny edge detection and then extracted features by using CNN. The accuracy of the experimental results on original the input image is lower than the accuracy of the edge extracted images. Canny edge detection can be used to eliminate the illumination variation on the original input image [15].

Phaung et al (2003) implemented skin segmentation algorithm by combining colour and edge information. In this paper, the authors tested five different colour spaces such as YCbCr, HSV, RGB, CIE XYZ and CIE Lab [8]. Phaung et al (2005) also have written the next paper for human skin colour segmentation on four different colour spaces such as CIE-Lab, YCbCr, HSV and RGB colour spaces. [7].

Rungruangbaiyok et al (2015) proposed skin colour threshold-based segmentation methods with the optimized threshold values on various colour models. In this paper, the ensemble Threshold Segmentation method (ETS) with the HSV colour space model is the most accurate method than other colour space threshold based
methods. The threshold values of all colour spaces are between 0 and 1. The accuracy was obtained 98.51 % on NUS Hand Pose Dataset I. [14].

Wang et al (2012) implemented hand segmentation methods by combining skin colour and background information. The performance of the proposed system is better than the single skin colour model [12]. Xu et al (2017) proposed the hand region identification algorithm by using YC\textsubscript{g}Cr colour space model. YC\textsubscript{g}Cr colour space is based on YCbCr colour space and also similar [11]. Patidar et al (2013) proposed that hand segmentation and tracking by using mix colour model such as (YCbCr+HSV) and (YCbCr+RGB) [13].

Tabassum et al (2010) implemented statistical skin detection algorithms for sub-Continental Human Images. In this paper, the authors implemented with three different colour models (HSV, YUV, YUV and YIQ). According to the experimental result, the HSV colour space model is the best performance models than other two colour space models to Indian sub-continental skin colour recognition [16]. Deng et al (2011) proposed real time hand gesture recognition by using cost matrix and context based matching approach. In this paper, the authors used YCbCr skin colour model to detecting the human hand. In the experimental results, the precision rates are between 70% and 90 % [17].

Ahlvers et al (2005) implemented face detection and head tracking with morphological hole mapping without using model. In this paper, the authors used three different colour modes (RGB skin cluster, YCbCr skin cluster and HSV skin cluster model) [18]. Lee et al. (2018) implemented static hand gesture recognition using wristband-based contour features. Wristband detection and skin color detection used to detect the hand and watershed segmentation and also used region merging techniques to remove the overlap condition. Finally, Minimum cosine distance used to recognize the static hand gestures. In this paper, the classification accuracy is up to 99.31% for 29 gestures. The general issue of this paper is difficult to detect the wristband when the background contains dark color and dark shadows within the gesturing hand.[19]

Sahoo et al. (2018) used grey world algorithm to remove illumination variation from the input hand gesture images and discrete wavelet transform and fisher ration (F-ration) used to extract features. Finally, the hand gestures recognized based on Support Vector Machine Algorithm. This paper is experimented on three different datasets such as MUDB dataset, Jochen-Triesch dataset and Self-Constructed dataset. The mean accuracy was obtained 98.64 %, 95.42 % and 99.08 % respectively. The issues of this paper can be misclassifying gestures vocabulary when rotation noise above 15 degree. [20].

III. DATA COLLATION FOR HAND GESTURES

There are basically three ways to collect the raw data for input of hand gesture recognition system.

- The first one is applied colour glove or data glove to collect the raw hand data called glove based approach. In data glove approach, the main drawbacks are expensive because sensor node, heavily, naturalness by using data glove. The weakness of glove user has to wear the glove every time. But, the colour gloves are inexpensive; no sensors are embedded in or outside the gloves and robust method for hand gesture recognition system.
- The second way used one or more camera to collect raw input hand gesture called vision based approach. The strength of this approach is natural and more convenient for communication. But, easily affected by complex background.
- The last approach is hybrid approach to collect raw hand data by combining the above two methods.

IV. DATASETS

NUS Hand Pose Dataset I consists of 10 classes of posture, 24 samples per class with uniform simplest background condition.

NUS Hand Pose Dataset II [4] is a dataset for hand gesture recognition with complex background condition. NUS hand pose dataset II are defined 160 gesture with human noise, the hand with 2000 hand images, and 2000 background images. In this dataset, each gesture has 5 times with 40 different subjects. The age ranges of 40 subjects are 22 to 56 years. The class types of gestures are alphabet character a to j. The posture images take form 40 subjects and the size of images (160*120). All the images are RGB colour space and the format of the images are .jpeg.

The sample frames of these two datasets as shows in Fig. 2.
YCbCr colour space is depending on the threshold values. In our experiment, we applied the threshold values (i) chrominance components Cr>133 && Cr<173 and Cb>77 && Cb<127 [10], (ii) chrominance components Cr>133 && Cr<173 and Cb>80 and Cb<120 [2], (iii) luminance component Y*80 && chrominance component Cr>135 && Cr<180 and chrominance component Cb>85 && Cb<135 [9]. The values ranges of YCbCr are 0 to 255. The different threshold values of YCbCr had been tested on NUS Hand Pose Dataset I with uniform background. In Fig3, shows original RGB images, YCbCr converted images, segmented black and white images with different threshold values of each hand gestures symbols with simplest background. The block and white images transformed based on user-defined threshold values by using equation 2. According to the experimental result, all the gesture symbols in NUS Hand Pose Dataset I can correctly segment the hand region.

The first column is Original image and, the second column is YCbCr Conversion Image and the next columns are hand Segmentation Image with different thresholds by using YCbCr colour space. The same threshold values of YCbCr are also tested on NUS Hand Pose Dataset II with a complex background. The result shows in Fig4. In the following Fig 4 show tested results with different threshold values on each symbols a to j with complex background conditions. This figure results show some of the complex background image are not suitable, mis-detected the hand from the background images. According to the experimental results, the colour based threshold values (77<Cb<127 && 133<Cr<173) of YCbCr are more suitable for a simple background. But, the result may be miss detection the human hand when the backgrounds contain objects like human skin colour. In complex background condition, only YCbCr colour space can’t correctly separate the human hand from the background.

In the previous literature, the researchers used in different colour spaces and various threshold values for human skin colour detection. Skin colour thresholding based method is very popular, easy to implement, and most often used to detect/segment the human skin. The threshold value for skin colour is difficult to set uniformly because of different types of skin colour tone, various illumination conditions. According to the start-of-art research, the different threshold values of YCbCr, RGB, Normalized RGB, HSV colour spaces have been used for human skin segmentation. In this paper, only YCbCr colour space has been used to segment the human hand as shown in the following Table 1.

VI. EXPERIMENTAL SECTION

YCbCr colour space is depending on the threshold values. In our experiment, we applied the threshold values (i) chrominance components Cr>133 && Cr<173 and Cb>77 && Cb<127 [10], (ii) chrominance components Cr>133 && Cr<173 and Cb>80 and Cb<120 [2], (iii) luminance component Y*80 && chrominance component Cr>135 && Cr<180 and chrominance component Cb>85 && Cb<135 [9]. The values ranges of YCbCr are 0 to 255. The different threshold values of YCbCr had been tested on NUS Hand Pose Dataset I with uniform background. In Fig3, shows original RGB images, YCbCr converted images, segmented black and white images with different threshold values of each hand gestures symbols with simplest background. The black and white images transformed based on user-defined threshold values by using equation 2. According to the experimental result, all the gesture symbols in NUS Hand Pose Dataset I can correctly segment the hand region.

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\[ I(b,w) = \begin{cases} 1; & (Cb(77,127) \ & \& Cr(133,173)) \\ 0; & \text{otherwise} \end{cases} \quad (2) \]

Where \( Y, \ Cb, \ Cr = [0;255] \) and 0=black and 1=white;

Table 1. Different Threshold Values for Human Skin Colour Detection

<table>
<thead>
<tr>
<th>References</th>
<th>Thresholds Values of Colour Channel</th>
<th>Application Domain</th>
<th>Colour Space</th>
<th>Datasets</th>
</tr>
</thead>
<tbody>
<tr>
<td>Shaik et al (2015)</td>
<td>( (100 &lt; Cb &lt; 150) &amp; &amp; (150 &lt; Cr &lt; 200) )</td>
<td>Human Skin Colour Segmentation</td>
<td>CbCr</td>
<td>Internet Source</td>
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<tr>
<td>McBride et al (2019)</td>
<td>( (80 &lt; Cb &lt; 120) &amp; &amp; (133 &lt; Cr &lt; 173) )</td>
<td>Human Skin Colour Segmentation</td>
<td>CbCr</td>
<td>Pratheepan Dataset</td>
</tr>
<tr>
<td>Rahmat et al (2004)</td>
<td>( \text{Normalized R/Normalized G} &gt; 1.185 &amp; \ 0.2 &lt; S &lt; 0.6 &amp; 0 &lt; H &lt; 25 \text{ or } 335 &lt; H &lt; 360 ) ( (77 &lt; Cb &lt; 127) &amp; &amp; (133 &lt; Cr &lt; 173) )</td>
<td>Human Skin Colour Segmentation</td>
<td>Combining Normalized RGB, HSV, CbCr</td>
<td>ECU, HGR Datasets</td>
</tr>
<tr>
<td>KuKharev et al (2004)</td>
<td>( (Y &gt; 80) &amp; &amp; (85 &lt; Cb &lt; 135) &amp; &amp; (135 &lt; Cr &lt; 180) )</td>
<td>Face Recognition</td>
<td>YCbCr</td>
<td>-</td>
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<td>Chai et al (1999)</td>
<td>( (77 &lt; Cb &lt; 127) &amp; &amp; (133 &lt; Cr &lt; 173) )</td>
<td>Face Segmentation</td>
<td>CbCr</td>
<td>-</td>
</tr>
<tr>
<td>Deng et al (2011)</td>
<td>( (45 &lt; Y &lt; 180) &amp; &amp; (126 &lt; Cb &lt; 143) &amp; &amp; (122 &lt; Cr &lt; 130) )</td>
<td>Hand Segmentation</td>
<td>YCbCr</td>
<td>images from CCD Camera</td>
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<tr>
<td>Ahlvers et al (2005)</td>
<td>( Y &gt; 80 &amp; &amp; (85 &lt; Cb &lt; 135) &amp; &amp; (135 &lt; Cr &lt; 180) )</td>
<td>Face Segmentation</td>
<td>YCbCr</td>
<td>Real Time Web Cam Images</td>
</tr>
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</table>

VII. CONCLUSION AND FUTURE WORK

Hand Gesture Recognition is one of the requirement as the language between man and machine. Hand Segmentation is the most important step of hand gesture recognition system. There are many segmentation methods. Among them, colour based segmentation is fast processing and aslo good feature to detect the human hand. If the accurately segment the hand from the background, then can correctly recognize the gesture. In this study, analyze YCbCr color space hand segmentation algorithm on two different NUS hand pose dataset I with the simplest background and NUS hand pose dataset II with complex background and human noise background. Colour is the most important reason that lead to recognize and discriminate information. The threshold based skin colour detection algorithms mainly impact on illumination, various taking image devices, objects like human skin colour and ethnicity. The current issue on YCbCr threshold-based segmentation cannot correctly recognize the hand when the background contains like human skin. In the future, we will consider mix-colour space model instead of using only YCbCr colour space model and then we will study attention on hand segmentation with deep learning.

DECLARATION

All author(s) have disclosed no conflicts of interest. The project was self-funded.

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Fig. 4. Hand Detection using YCbCr colour model with a uniform background.
### Fig. 5: Hand Detection using YCbCr colour mode with complex background

<table>
<thead>
<tr>
<th>Input Image</th>
<th>YCbCr Color Image</th>
<th>Segmented Image with threshold values (77&lt;Cb&lt;127)&amp;(133&lt;Cr&lt;173)</th>
<th>Segmented Image with threshold values (80&lt;Cb&lt;120)&amp;(133&lt;Cr&lt;173)</th>
<th>Segmented Image with threshold values (Y&gt;80) &amp; (85&lt;Cb&lt;135) &amp; (135&lt;Cr&lt;180)</th>
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<td><img src="image8" alt="Segmented Image" /></td>
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