

A New Skin Color Based Face Detection Algorithm by Combining Three Color Model Algorithms

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Abstract: Human face recognition systems have gained a considerable attention during last decade due to its vast applications in the field of computer and advantages over previous biometric methods. There are many applications with respect to security, sensitivity and secrecy. Face detection is the most important and first step of recognition system. Human face detection suffers from various challenges due to variation regarding image conditions, size, resolution, poses and rotation. Its accurate and robust detection has been a great task for the researchers. There exist various numbers of methods and techniques for face detection but none can guarantee successful in all conditions for all kinds of faces and images. Some methods are exhibiting good results in certain conditions and others are good with different kinds of images. Face detection based on skin color is found to be more effective technique because of the properties of skin color which is unique and can be easily separated from the other objects present in the image and background.

This paper introduces a new approach to face detection systems using the skin color of a subject. This system can detect a face regardless of the background of the picture, which is an important phase for face identification. The images used in this system are color images which give additional information about the images than the gray images provide.

In face detection, the two respective classes are the "face area" and the "non-face area". This new approach to face detection is based on color tone values specially defined for skin area detection within the image frame.

Keywords: Color Space, Face detection, Skin Color.

I. Introduction

When you are talking to a person, you usually look at his face; the expression of a person's face plays a very important role when communicating with other people. Due to its uniqueness, the face is also the most important and effective characteristics for recognizing a person.

Compared with fingerprints or retinas, taking a face image from a person is very easy. Therefore face recognition has become one of the most popular applications in the field of computer vision [1]. With the recent major terrorist attacks in the civil world, there have been increasingly substantial interests in the development of intelligent surveillance cameras that can automatically detect and recognize known criminals as well as suspicious characters. Due to such uncertain times, humans are beginning to seek support from computer systems to aid in the process of identification and location of faces in everyday scenes [2].

Images of faces vary considerably depending on lightning, occlusion, pose, facial expression, and identity. Color transforms must be implemented to deal with all remaining variation in distinguishing face skin color.

In this paper a face detection system is proposed and implemented using a skin region analysis method, within the image according to YCbCr color, where in each pixel is precisely classified as being either skin or non-skin. This system will detect the face from any background. This system is implemented using the scientific language which is efficient to deal with matrices mathematic operation and image processing.

For detecting face there are various algorithms including skin color based algorithms. Color is an important feature of human faces. Using skin-color as a feature for tracking a face has several advantages. Color processing is much faster than processing other facial features. Under certain lighting conditions, color is orientation invariant. This property makes motion estimation much easier because only a translation model is needed for motion estimation. However, color is not a physical phenomenon; it is a perceptual phenomenon that is related to the spectral characteristics of electromagnetic radiation in the visible wavelengths striking the retina. Tracking human faces using color as a feature has several problems like the color representation of a face obtained by a camera which is influenced by many factors (ambient light, object movement, etc.), different cameras produce significantly different color values even for the same person under the same lighting conditions and skin color differs from person to person. In order to use color as a feature for face tracking, we have to solve these problems. It is also robust towards changes in orientation and scaling and can tolerate occlusion well. A disadvantage of the color cue is its sensitivity to illumination color changes and, especially in the case of RGB, sensitivity to illumination intensity. One way to increase tolerance towards intensity changes in images is to

transform the RGB image into a color space whose intensity and chromaticity are separate and use only chromaticity part for detection.

In this paper we have presented a comparative study of three well known skin color face localization/detection algorithms and have produced a new algorithm based on skin color classification in YCbCr color model. Our results show that we can localize the face more effectively by using the proposed algorithm.

II. Color Models for Skin Color Classification

The study on skin color classification has gained increasing attention in recent years due to the active research in content-based image representation. For instance, the ability to locate image object as a face can be exploited for image coding, editing, indexing or other user interactivity purposes.

Moreover, face localization also provides a good stepping stone in facial expression studies. It would be fair to say that the most popular algorithm to face localization is the use of color information, whereby estimating areas with skin color is often the first vital step of such strategy. Hence, skin color classification has become an important task. Much of the research in skin color based face localization and detection is based on RGB, YCbCr and HSI color spaces. In this section the color spaces are being described.

2.1 RGB Color Space

RGB is based on the Cartesian coordinate system which is an aid cube as shown in Figure 1. The cube has the RGB values in the three corners - colors like cyan, magenta and yellow and the other three corners - the black and white at the origin is at the corner farthest from the origin. The gray scale is located on the line joining black and white. They are called additive "primary" because the colors are summed to produce the desired color. Due to the high correlation between color components: red, green and blue, as each component is subject to the effect of luminance of the light intensity of the environment, so that suffers dissatisfaction on the part of many image processing applications. In practical, this model is not well suited to describe colors in terms of human interpretation [3].

2.1.1 Algorithm of Skin Color Based Face Detection in RGB Color Space

Skin color is the most important feature of the face and is unique because of its color ingredients. Skin color pixels can be easily detected using standard color histogram which is normalized for every future change in the intensity of luminance of the division. And thus RGB vector is transformed into a vector [r, g] color standard which in turn provides a rapid means of detecting the skin. This gives the skin color region, which localizes face. As indicated early RGB suffers from the effect of luminance but is still able to allow us to eliminate some colors which are clearly outside the scope of normal skin color. After review and analysis of different levels of the RGB color space, it was found that the following rule works well in eliminating some redundant pixels that are labeled as non-face [4].

$$0.836G - 14 < B < 0.836G + 44 \Rightarrow \text{Skin} \dots\dots\dots (1)$$

$$0.79 g - 67 < B < 0.78 g + 42 \Rightarrow \text{Skin} \dots\dots\dots (2)$$

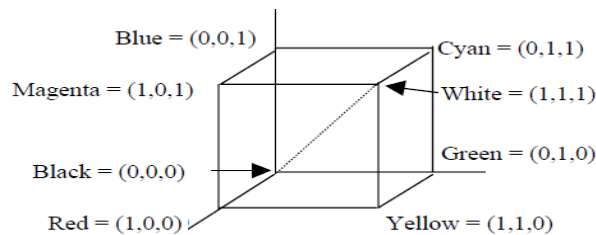


Fig. 1: RGB Color Cube

2.2 YCbCr Color Space

This color space was defined to meet the growing demand of digital processing algorithms of video information and has become widely used color space in digital videos. It has three components, two of them are the chrominance and luminance is one. This model comes in the family of television transmission color space along with YUC and YIQ and is designed for space analogue PAL and NTSC [4].

YCbCr color model has been developed to allow the transmission of color information on televisions keeping in mind that the existing television in black and white still displays images in shades of gray, has the characteristic of luminance and isolate color information, and is used in many applications such as compression.

2.2.1 Algorithm of Skin Color Based Face Detection in YCbCr Color Space

In this model, the skin color pixels belonging to the region exhibit similar Cb and Cr values as it is the skin color model based on the Cr and Cb values which can provide good coverage of human races. YCbCr signals that are created from the corresponding gamma adjusted RGB source using two defined constants KB and KR.

After gamma correction and before scaling and offsets YCbCr signals are called Y1, PB, and PR and are defined as indicated below.

$$Y' = KR * R' + (1 - KR - KB) * G' + KB * B' \dots\dots\dots (3)$$

$$PB = \frac{1}{2} * ((B' - Y') / (1 - KB)) \dots\dots\dots (4)$$

$$PR = \frac{1}{2} * ((R' - Y') / (1 - KR)) \dots\dots\dots (5)$$

Here, the prime ‘ symbols mean gamma correction is being used, thus R’, G’ and B’ nominally range from 0 to 1, with 0 representing the minimum intensity (e.g., for display of the color black) and 1 the maximum (e.g., for display of the color white). From this the resulting luminance signal (Y) value will then have a nominal range from 0 to 1. And the chrominance value (PB and PB) values will then have a nominal range from -0.5 to +0.5.

For representing signals in digital form, rounding off and scaling and offsets are added. For instance scaling and offset applied to the Y’ component per specification results in the value of 16 for black and the value of 235 for white when using an 8-bit representation. The standard has 8-bit digitized versions of CB and CR scaled to a different range of 16 to 240. As a result, rescaling by the fraction $(235-16)/(240-16) = 219/224$ is mandatory when doing color matrixing or processing in YCbCr space, leading quantization distortions when the successive processing is not executed using higher bit depths.

The form of Y’CbCr that was defined for standard-definition television use in the ITU-R BT.601 (formerly CCIR 601) standard for use with digital component video is derived from the corresponding RGB space as follows:

$$KB = 0.114$$

$$KR = 0.299$$

From the above constants and formulas, the following can be derived for ITU-R BT.601. Analog YPbPr from analogue R’G’B’ is derived as follows:

$$Y' = 0.299 * R' + 0.587 * G' + 0.114 * B' \dots\dots\dots (6)$$

$$PB = -0.168739 * R' - 0.331264 * G' + 0.5 * B' \dots\dots\dots (7)$$

$$PR = 0.5 * R' - 0.418688 * G' - 0.81312 * B' \dots\dots\dots (8)$$

Digital Y’CbCr (8 bits per sample) is derived from analogue R’G’B’ as follows:

$$Y' = 16 + (65.481 * R' + 128.553 * G' + 24.966 * B') \dots\dots\dots (9)$$

$$PB = 128 + (-37.797 * R' - 74.203 * G' + 112.0 * B') \dots\dots\dots (10)$$

$$PR = 128 + (112.0 * R' - 93.786 * G' + 18.214 * B') \dots\dots\dots (11)$$

Note that the resultant signals range from 16 to 235; the values from 0 to 15 are called foot room, while the values from 236 to 255 are called head room.

Alternatively, digital Y’CbCr is derived from digital Rd Gd Bd (8 bits per sample) according to the following equations:

$$Y' = 16 + (65.481 * Rd + 128.553 * Gd + 24.966 * Bd) / 256 \dots\dots\dots (12)$$

$$PB = 128 + (-37.797 * Rd - 74.203 * Gd + 112.0 * B'd) / 256 \dots\dots\dots (13)$$

$$PR = 128 + (112.0 * Rd - 93.786 * Gd - 18.214 * Bd) / 256 \dots\dots\dots (14)$$

$$Y' = 16 + (0.255 * Rd + 0.5021 * Gd + 0.0975 * Bd) \dots\dots\dots (15)$$

$$PB = 128 + (-0.148 * Rd - 0.368 * Gd + 0.439 * B'd) \dots\dots\dots (16)$$

$$PR = 128 + (0.439 * Rd - 0.368 * Gd - 0.071 * Bd) \dots\dots\dots (17)$$

The above equations are used in the coding part of the proposed algorithm. Cr and Cb are used instead of PB and PR. After studying and experimenting with various thresholds, it was finally found that the best results were found by using the following decree and otherwise assume that it is NOT skin and may be removed from further consideration [5].

$$102 < Cb < 128 => \text{Skin} \dots\dots\dots (18)$$

2.3 HSI Color Space

This color model is ideal for hardware implementation. Human practice interpretation is described in terms of Hue Saturation and Intensity (Brightness). HSI model separates the intensity component of the color information-bearing and is an ideal tool for development processing algorithm based on the image description of the colors that are natural and intuitive to human health. HSI is based on the cylindrical coordinate, hue (H) is shown with a 0 angle, varying from 0 to 360, the saturation (S) corresponds to a radius ranging from 0 to 1 and finally the intensity (I) varies along the z-axis with 0 being black and white is 1. When the value of S is 0, the color is a gray value of intensity. When the value of S is 1, color is the limit of the cone tope. More saturation,

the color is white / gray / black. Changing the value of the color will change the color of red that is to 00 to green which is 1200, 2400 and blue to red black in 3600. And when I = 0, the color is black and therefore H is undefined. By varying the value of I, a color can be is darker or lighter. Shaded color comes out when the value of I is adjusted while S = 1 hold.

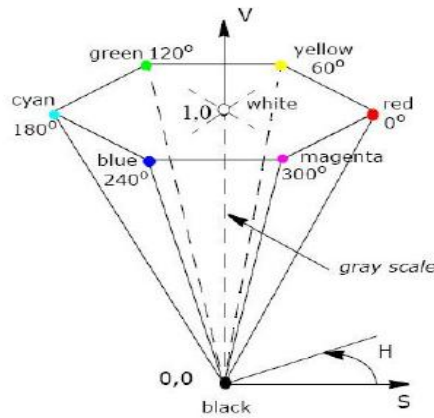


Fig. 2: Cone Model Of HSV Color Space

2.3.1 Algorithm of Skin Color Based Face Detection in HSI Color Space

The HSI color space is more intuitive and provides color information in the same manner that the humans think of colors and how artists mix colors in general. Dyed with saturation provides useful information regarding selective skin. Range for non-Hue face region is considered as shown below and it is also assumed to be the skin. HSI color model is represented using the following equation:

$$19 < H < 240 \Rightarrow \text{Not Skin} \dots\dots\dots (19)$$

HSI color space is defined by using the RGB as scale as transformation from RGB space to and HSI space [6]

$$H = 360^\circ - a / 360^\circ \text{ if } b > g \dots\dots\dots (20)$$

$$= a / 360^\circ \text{ otherwise } \dots\dots\dots (21)$$

$$\text{Where } a = \cos^{-1} 2R - G - B / (R^2 + G^2 + B^2 - RG - GB - BR) ^ {1/2} \dots\dots\dots (22)$$

$$S = 1 - [\min (R, G, B)] / I \dots\dots\dots (23)$$

$$I = (R + G + B) / 3 \dots\dots\dots (24)$$

The colors are given a set of RGB color cube. The color cube is oriented so that the value component is oriented along the axis of gray, ranging from black to white. HSV are defined solely by reference to the RGB color space, they are not absolute color spaces: to specify a color precisely requires reporting not only the HSV values, but also the characteristics of the RGB space they are based, including gamma correction in use it.

Computer offers some poor cousins of these perceptual spaces which can also be found in the software interface, such as HSV. They are easy mathematical transformations RGB, and they seem to be perception systems because they use the color lightness / saturation value terminology. But take a look around, do not be fooled. Perceptual dimensions of color are poorly scaled by the color specifications that are provided in these and other systems. For example, saturation and brightness are combined, so that saturation scale may also contain a wide range of light (for example, it can evolve from white to green, which is a combination of both brightness and saturation). Similarly, the color and lightness are confounded if, for example, a saturated blue and yellow can be saturated designated as the same "lightness", but have large differences in the perceived lightness. These defects make the systems difficult to use to control the appearance of a color in a systematic approach manner. If more tweaking is needed to get the desired effect, the system offers little benefit more struggling with raw RGB specifications [7].

III. Proposed Face Detection System

In the proposed algorithm, it is assumed that by combining the detected region from the entire three algorithms, skin region is extracted. After combining the three algorithms and extracting skin color, following this face is detected by edges using edge detection which draws boundary across the face region. It is something like if one algorithm detects the skin color for an image and others two algorithms fails and gives the false result; even then the face is extracted by using the combination algorithm. This supposition of combining algorithm is based on the basic idea of Venn diagram from set theory. Suppose the result from RGB color is 'A', results from YCbCr color space is 'B' and the result from HIS color space is 'C' and if any of the result contains a skin image then the union of the three will surely be a skin image. This can be understood by the following example:: let, A = {a, b, c, d}, B = {b, e} and C = {b, f}. If "a" = "skin image" then skin image is

present only in A and the next two are failed in extracting the skin color then the union of the three $U = (A \cup B \cup C) = \{a, b, c, d, e, f\}$ hence this algorithm can detect skin color or else if we take 'e' or 'f' as skin image then also the union (U) will have skin image in it also this is applicable if all the three (A, B, C) has skin image. Hence in all sense the skin region are going to be detected.

After extracting the skin region, facial features such as eyes, nose and mouth are extracted. The image obtained after applying skin color information is given to binarization i.e., it is transformed to gray-scale image and then to a binary image by applying suitable threshold. This is done to eradicate the hue and saturation values and consider only the luminance part. This luminance part is then transformed to binary image with some threshold because the features we want to consider further for face extraction are darker than the background colors. Opening and closing operations are performed to filter out noise. The morphological close operation is nothing but a dilation followed by erosion, and the morphological open operation is erosion followed by dilation, these operations are performed to remove holes.

IV. Experimental Results

In this section, images resulted from the proposed algorithm are shown with example (Input-1). The example has six images as outputs: segmented image, applying RGB to segmented image, image after edge detection, binarization, noise removed by performing morphological operation, output image (Face detected).

Segmentation: Segmentation is the main part of the algorithm, gives the minute description of the pixel. So the suitable operations can be performed. The operation performed in the proposed algorithm is based on the three color spaces used.

By Applying RGB: Convert image resulted from segmentation to color image.

Edge detection: This is done in order to trace the edges and separate the regions containing face with that of non-face.

Binarization: It converts the image into binary image.

Morphological operation: As explained earlier, it performs open and close operation to remove noise.

Output image: Shows the outcome of the algorithm



Input Image-1



Image Segmented



Applying RGB to Segmented Image-1

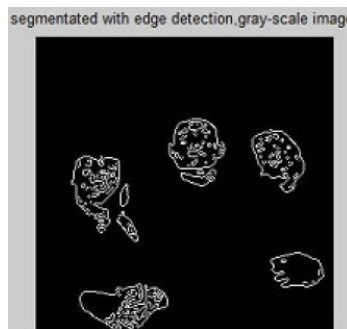
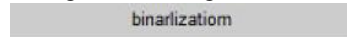


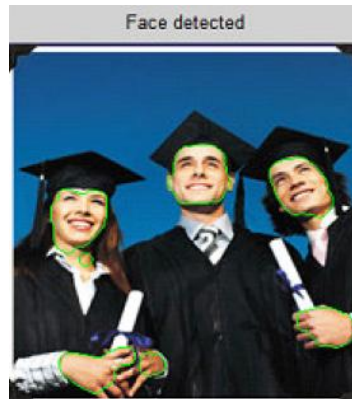
Image-1 after Edge Detection



Binarlization



Noise Removed by Performing Morphological Operation



Output of Image-1 (Face Detected)

V. Conclusions

From the work carried out on this thesis, it is concluded that it is very difficult and still a challenging task to build up an automatic face detection method that works effectively in all situations whether the image consists of different face poses, bad image conditions or images affected by illumination. Researchers are still struggling hard to find out the perfect solution to this complicated problem. But they exist, different face detection methods to overcome different problems of face detection. For instance, to overcome the problem of lighting (illumination) RGB color space is not used; instead, HSV or YCrCb are used. Some methods or techniques show good results for particular problems or applications, while others show good results for different applications. So it may not be insane to say that researchers still have a lot of work to do under face detection.

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