

HUMAN FACE DETECTION FROM COLOR IMAGES: PRELIMINARY RESULT

Tan Shuen Chuan¹, Halim Setan² & Zulkepli Majid³

Medical Imaging Research Group

Faculty of Geoinformation Science & Engineering

Universiti Teknologi Malaysia

81310 UTM Skudai, Johore. Malaysia.

Tel: 07-5530380

{shuen_chuan@hotmail.com¹, halim@fksg.utm.my², zulkepli@fksg.utm.my³}

Abstract

Images containing faces are essential to intelligent vision-based human computer interaction where it is utilized in various applications (e.g. face recognition, face tracking, pose estimation, expression estimation, etc). This paper aims to provide preliminary investigation on human face detection from a color image. A successful and robust algorithm should be capable in identifying all facial region within a given images regardless of its position, orientation and lighting conditions. Such problems are quite challenging because the faces are non-rigid and have a high degree of variability in size, shape, color and texture. Numerous approaches have been developed and proposed to localize facial region in a single image. Our study utilized YCrCb color space method in the detection of human face from a color image with a pre-determine chrominance and luminance parameters. The result obtained illustrates the feasibility of this approach where it exhibits high and accurate segmentation rate though possess minor flaws.

Key words: Face detection, face localization, chrominance and luminance.

1.0 INTRODUCTION

Face detection is defined as: Given an arbitrary image, a solution (algorithm) developed is able to determine whether or not there are any faces in the image and, if present, return the image location and extend of each face. Facial image detection algorithms are gaining more and more attention as it can be applied in various fields. Among the applications that implementing the concept of facial area detection are facial tracking, pose estimation and face recognition [1, 2, 3, 4].

Owing to several ill-posed problems and variability found in a given input image, challenges associated with face detection can be classified to the following factors [1]:

- **Pose.** The images of a face vary due to the relative camera-face pose (frontal, 45 degree, profile, upside down), and some facial features such as an eye or the nose may become partially or wholly occluded.
- **Presence or absence of structural components.** Facial features such as beards, mustaches and glass may or may not present and there is a great deal of variability among these components including shape, color and size.
- **Facial expression.** The appearance of faces is directly affected by a person's facial expression.
- **Occlusion.** Faces may be partially occluded by other objects. In an image with a group of people, some faces may partially occlude other faces.

- **Image orientation.** Face images directly vary for different rotations about the camera's optical axis.
- **Imaging conditions.** When the image is formed, factors such as lighting (spectra, source distribution and intensity) and camera characteristics (sensor response, lenses) affect the appearance of a face.

A typical face detection framework is illustrated in Figure 1.0 [31].

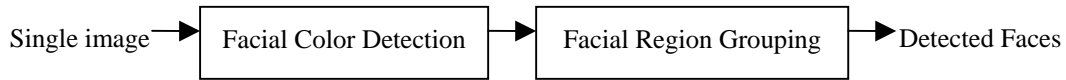


Figure 1.0: Framework of typical face detection system.

1.1 Detecting Faces in a Single Image

In this section, we review several existing techniques used to detect faces from a single intensity or color image. The single detection methods are classified into four categories; some methods clearly overlap category boundaries and are discussed at the end of this section.

1. **Knowledge-based methods.** These rule-based methods encode human knowledge of what constitutes a typical face. Usually, the rules capture the relationships between facial features. These methods are designed mainly for face localization.
2. **Feature invariant approaches.** These algorithms aim to find structural features that exist even when the pose, viewpoint, or lighting conditions vary, and then use these to locate faces. These methods are designed particularly for face localization operations.
3. **Template matching methods.** Several standard patterns of a face are stored to describe the face as a whole or the facial features separately. The correlations between an input image and the stored patterns are computed for detection. These methods have been used for both face localization and detection.
4. **Appearance-based methods.** In contrast to template matching, the models (or templates) are learned from a set of training images which should capture the representative variability of facial appearance. These learned models are then used for detection. These methods are designed mainly for face detection.

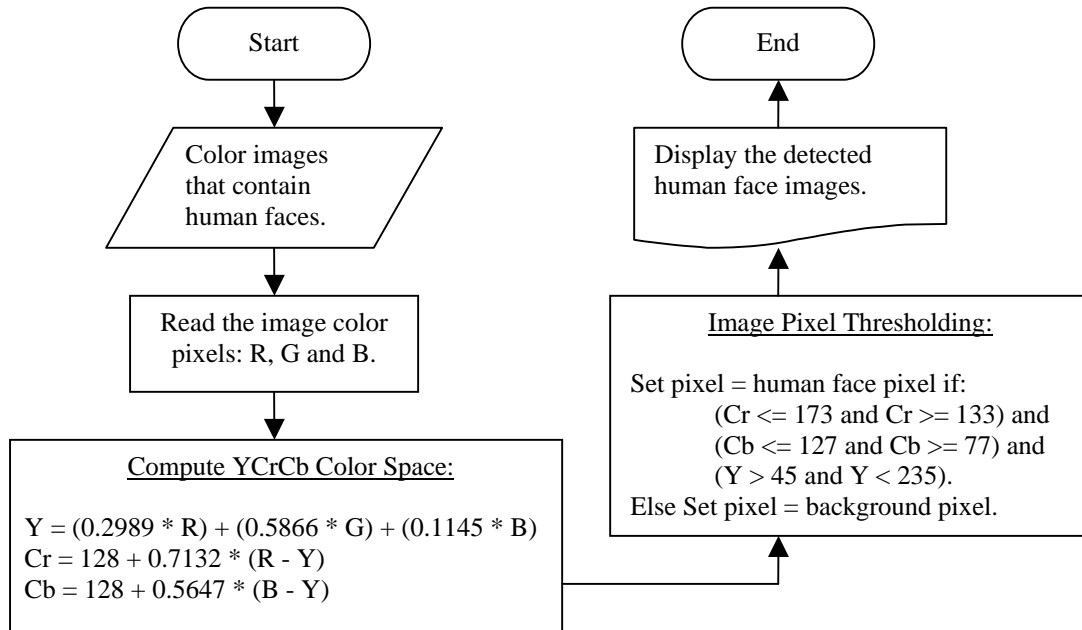
Table 1 summarized the algorithms and representative works for face detection in a single image within these four categories.

2.0 METHODOLOGY

In this study, we have chosen skin color space which belongs to Feature invariant category to be used in the detection of human faces in a given color images. Generally, our system constitutes of three main steps: reading the images pixel value in RGB, YCrCb color space computation [50] and images' pixel thresholding [15, 30] according to a pre-determine parameters. Figure 2.1 illustrates the framework of our human face detection system.

Table 1: Categorization of Methods for Face Detection in a Single Image

Approach	Representative Works
Knowledge-based	Multi-resolution rule-based method [33]
Feature invariant <ul style="list-style-type: none"> - Facial Features - Texture - Skin Color - Multiple Features 	Grouping of edges [34, 35] Space Gray-level Dependence matrix (SGLD) of face pattern [36] Mixture of Gaussian [37, 38] Integration of skin color, size and shape [27, 39]
Template matching <ul style="list-style-type: none"> - Predefined face templates - Deformable Templates 	Shape template [16, 40] Active Shape Model (ASM) [41]
Appearance-based method <ul style="list-style-type: none"> - Eigen-face - Distribution-based - Neural Network - Support Vector Machine (SVM) - Naïve Bayes Classifier - Hidden Markov Model (HMM) - Information-Theoretical Approach 	Eigenvector decomposition and clustering [6, 42] Gaussian distribution and multi-layer perceptron schemes [43] Ensemble of neural networks and arbitration schemes [3, 44] SVM with polynomial kernel [5, 45] Joint statistics of local appearance and position [46] Higher order statistics with HMM [47] Kullback relative information [48, 49]

**Figure 2.1: Framework of Human Face Detection.**

2.1 Human Face Detection Algorithm

This section explains the implementation of the human face detection using YCrCb color space:

1. Read a given color image's pixel value (R,G and B) that contains human face from left to right and from top to bottom.

2. According to [50], for each set of pixels (R, G and B) read, compute the YCrCb color space value corresponding with them by using formula (2.1).

$$\begin{aligned} Y &= (0.2989 * R) + (0.5866 * G) + (0.1145 * B) \\ Cr &= 128 + 0.7132 * (R - Y) \\ Cb &= 128 + 0.5647 * (B - Y) \end{aligned} \quad (2.1)$$

3. Perform image pixels thresholding with a pre-determine threshold value [15, 30] for Y, Cr and Cb respectively.

$$\begin{aligned} \text{Set pixel} &= \text{human face pixel if:} \\ & \quad (Cr \leq 173 \text{ and } Cr \geq 133) \text{ and} \\ & \quad (Cb \leq 127 \text{ and } Cb \geq 77) \text{ and} \\ & \quad (Y > 45 \text{ and } Y < 235). \\ \text{Otherwise Set pixel} &= \text{background pixel.} \end{aligned} \quad (2.2)$$

4. Display the detected human face images for result observations.

3.0 EXPERIMENTAL RESULTS

In this section, we have conducted several experiments on color images that consist of human face. Their dimension is fixed at 1920 x 480 pixels.

Figure 3.1(a) and Figure 3.2 (a) are the sample color images. Figure 3.1(b) and Figure 3.2 (b) illustrate the human face detection results obtained from these color images where only CrCb color space is utilized. The results derived where YCrCb color space is used to localize human face are shown in Figure 3.1 (c) and Figure 3.2 (c) respectively.



Figure 3.1(a) – Input color image.



Figure 3.2(a) – Input color image.



Figure 3.1(b) – Detected human face using only CrCb color space.



Figure 3.2(b) – Detected human face using only CrCb color space.



Figure 3.1(c) – Detected human face using YCrCb color space.



Figure 3.2(c) – Detected human face using YCrCb color space.

4.0 CONCLUSION

From Figure 3.1 (b) and Figure 3.2 (b), it can be concluded that noise occur (where pixels possess similarity color space with human face color space are classified as face pixels) when only CrCb color space is applied. However, with the use of luminance component, Y, the face detection scheme is improved (Figure 3.1c and Figure 3.2c) where most of the detected pixels are belonging to the face color space. In order to better classify the facial area, a comprehensive technique should be introduced.

5.0 RECOMMENDATION OF FUTURE WORKS

From the output results produced in Section 3.0, our face detection system still have several aspects that can be improved; these include:

- Utilization of artificial intelligence for the sake of enhancement in face segmentation accuracy.
- A mechanism should be introduced to generalize an appropriate parameters range (Y, Cr and Cb) to be used when deal with different circumstances (for example: the different between the skin fairness of Asian and African).

ACKNOWLEDGEMENTS

The authors would like to thank the Ministry of Science, Technology and Environment (MOSTI) for funding this project (Vote: 74537). Moreover, we would like to thank Dr. Albert Chong from the University of Otago (New Zealand) for providing us with consecutive guidance and assistance. We also like to appreciate several researchers and academicians whose publications manage to assist this research.

REFERENCES

1. Ming-Hsuan Yang, David J. Kriegman and Narendra Ahuja (2002). *Detecting Faces in Images: A Survey*. IEEE Transactions on Pattern Analysis and Machine Intelligence. Vol. 24, No. 1, pp. 34-58
2. S. Palanivel, B.S. Venkatesh and B. Yegnanarayana (2002). *Real Time Face Authentication System using Auto-associative Neural Network Models*. Indian Institute of Technology Madras, India.
3. Linlin Huang, Akinobu Shimizu, Yoshishiro Higihara and Hidefumi Kobatake (2002). *Robust Face Detection using a Modified Radial Basis Function Network*. IEICE Transaction, Inf. & Syst., Vol. E8.
4. B. Menser and F. Muller (1999). *Face Detection in Color Images using Principal Components Analysis*. Aachen University of Technology, Germany.
5. Elena Casiraghi, Raffaella Lanzarotti and Giuseppe Lipori (2003). *A Face Detection System based on Color and Support Vector Machines*. University Studies of Milan, Itali.
6. K. Sandeep and A. N. Rajagopalan (2003). *Human Face Detection in Cluttered Color Images using Skin Color and Edge Information*.
7. C. Carcia, G. Zikos and G. Tziritas (1999). *Face Detection in Color Images using Wavelet Packet Analysis*. Institute of Computer Science, Foundation for Research and Technology-Hellas, Greece.
8. Ming-Hsuan Yang and Narendra Ahuja (1998). *Detecting Human Faces in Color Images*. Beckman Institute and Department of Electrical and Computer Engineering, Univesity of Illinois, Urbana.
9. Michael J. Jones and James M. Rehg (1998). *Statistical Color Models with Application to Skin Detection*. Cambridge Research Laboratory, Compaq Computer Corporation, One Cambridge Center, Cambridge.
10. Siddharth Joshi and Gaurav Srivastava (2003). *EE368: Digital Image Processing – Face Detection (Project Report)*.
11. Vladimir Vezhnevets (2002). *Face and Facial Feature Tracking for Natural Human-Computer Interface*. Graphics & Media Lab., Dept. of Applied Mathematics and Computer Science of Moscow State University, Moscow, Rusia.
12. Prem Kuchi, Prasad Gabbur, P. Subbanna Bhat and Sumam David (2002). *Human Face Detection and Tracking using Skin Color Modeling and Connected Component Operators*. Dept. of E & C Engg., Karnataka Regional Engineering College, Surathkal, Karnataka.
13. Emiliano Acosta, Luis Torres, Alberto Albiol and Edward Delp (2002). *An Automatic Face Detection and Recognition System for Video Indexing Applications*. Polytechnic University of Catalonia, Spain; Polytechnic University of Valencia, Spain and Purdue University West Lafayette, USA.
14. Alberto Aibioli, Luis Torrest and Edward J. Delp (2001). *Optimum Color Spaces for Skin Color Detection*. Polytechnic University of Catalonia, Spain; Polytechnic University of Valencia, Spain and Purdue University, USA.
15. Alberto Aibioli, Luis Torrest, Charles A. Bouman and Edward J. Delp (2001). *A Simple and Efficient Face Detection Algorithm for Video Database Applications*. Polytechnic University of Catalonia, Spain; Polytechnic University of Valencia, Spain and Purdue University, USA.
16. Diedrick Marius, Sumita Pennathur and Klint Rose (2003). *Face Detection using Color Thresholding and Eigenimage Template Matching*.

17. Douglas Chai and Abdesselam Bouzerdoum (1999). *A Bayesian Approach to Skin Color Classification in YCrCb Color Space*. Edith Cowan University, Australia.
18. G. Gomez, M. Sanchez and L. Enrique Sucar (2002). *On Selecting an Appropriate Color Space for Skin Detection*. Dept. of Computing, ITESM-Morelos, Mexico.
19. Jae Y. Lee and Suk I. Yoo (2002). *An Elliptical Boundary Model for Skin Color Detection*. School of Computer Science and Engineering, Seoul National University, Korea.
20. Vladimir Vezhnevets, Vassili Sazonov and Alla Andreeva (2003). *A Survey on Pixel-based Skin Color Detection Techniques*. Faculty of Computational Mathematics and Cybernetics, Moscow State University, Moscow.
21. Gines Garcia Mateos and Cristina Vicente Chicote (2000). *A Unified Approach to Face Detection, Segmentation and Location Using HIT Maps*. University of Murcia, University of Cartagena, Spain.
22. Jean-Christophe Terrillon and Shigeru Akamatsu (1998). *Comparative Performance of Different Chrominance Spaces for Color Segmentation and Detection of Human Faces in Complex Scene Images*. ATR Human Information Processing Laboratories, Japan.
23. Richard P.Schumeyer and Kenneth E.Barner (1997). *A Color-Based Classifier for Region Identification in Video*. Dept. of Electrical and Computer Engineering, University of Delaware, USA.
24. Vinay P.Kumar and Tomaso Poggio (1999). *Learning-Based Approach to Real Time Tracking and Analysis of Faces*. Massachusetts Institute of Technology Artificial Intelligence Laboratory, USA.
25. Dongyun Kim, Jin-Hyung Lee and Joong-Ho Won (2002). *EE368: Final Project Report – Face Detection using Skin Color Model and Maximal Rejection Classification*. Dept. of Electrical Engineering, Stanford University, UK.
26. Douglas Chai and King N.Ngan (1999). *Face Segmentation using Skin-Color Map in Videophone Applications*. IEEE Transactions on Circuits and System for Video Technology, Vol. 9, No. 4, pp. 551-564.
27. Jiaping Fan, David K. Y. Yau, Ahmed K.Elmagarmid and Walide GIaref (2001). *Automatic Image Segmentation by Integrating Color-Edge Extraction and Seeded Region Growing*. IEEE Transaction on Image Processing, Vol. 10, No. 10, pp. 1454-1466.
28. Ioannis Kompatsiaris, George Mantzaras and Michael G.Strintzis (2000). *Spatiotemporal Segmentation and Tracking of Objects in Color Image Sequences*. Information Processing Laboratory, Electrical and Computer Engineering Department, Greece.
29. I. Grinias, Y. Mavrikakis and G. Tziritas (2001). *Region Growing Color Image Segmentation Applied to Face Detection*. Dept. of Computer Science, University of Crete, Greece.
30. Douglas Chai and King N. Ngan (1997). *Locating Facial Region of a Head-and-Shoulders Color Image*. Dept. of Electrical & Electronic Engineering, University of Western Australia, Australia.
31. Alberto Albiol, Luis Torres & Edward J.Delp (2000). *An Unsupervised Color Image Segmentation Algorithm for Face Detection Applications*. Polytechnic University of Valencia, Polytechnic University of Catalonia, Spain.
32. Bernd Heisele, Thomas Serre, Massimiliano Pontil and Tomaso Poggio (2001). *Component-base Face Detection*. Center for Biological and Computational Learning, USA; Dept. pf Information Engineering, University of Siena, Italy.
33. G. Yang and T. S. Huang (1994). *Human Face Detection in Complex Background*. Pattern Recognition, Vol. 27, No. 1, pp. 53-56.
34. K. C. Yow and R. Cipolla (1997). *Feature-based Human Face Detection*. Image and Vision Computing, Vol. 15, No. 9, pp. 713-735.
35. T. K. Leung, M. C. Burl and P. Perona (1995). *Finding Faces in Cluttered Scenes using Random Labeled Graph Matching*. Proc. Fifth IEEE Int'l Conf. Computer Vision, pp. 637-644.
36. Y.Dai and Y.Nakano (1996). *Face-texture Model Baded on SGLD and Its Application in Face Detection in a Color Scene*. Pattern Recognition, Vol.29, No.6, pp. 1007-1017.
37. S.McKenna, S. Gong and Y. Raja (1998). *Modeling Facial Color and Identity with Gaussian Mixtures*. Pattern Recognition, Vol.31, No.12, pp. 1883-1892.

38. J. Yang and A. Waibel (1996). *A Real-Time Face Tracker*. Proc. Third Workshop Applications of Computer Vision, pp. 142-147.
39. R. Kjeldsen and J. Kender (1996). *Finding Skin in Color Images*. Proc. Second Int'l Conf. Automatic Face and Gesture Recognition, pp. 321-317.
40. I. Craw, D. Tock and A. Bennett (1992). *Finding Face Features*. Proc. Second European Conf. Computer Vision, pp. 92-96.
41. A. Lanitis, C.J. Taylor and T.F. Cootes (1995). *An Automatic Face Identification System using Flexible Appearance Models*. Image and Vision Computing. Vol. 13, No. 5, pp. 393-401.
42. M. Turk and A. Pentland (1991). *Eigenfaces for Recognition*. J. Cognitive Neuroscience, Vol. 3, No. 1, pp. 71-86.
43. K.-K. Sung and T. Poggio (1998) *Example-based Learning for View-based Human Face Detection*. IEEE Trans. Pattern Analysis and Machine Intelligence, Vol. 20, No. 1, pp. 39-51.
44. H. Rowley, S. Bahuja and T. Kanade (1998). *Neural Network-based Face Detection*. IEEE Trans. Pattern Analysis and Machine Intelligence, Vol. 20, No. 1, pp. 23-38.
45. E. Osuna, R. Freund and F. Girosi (1997). *Training Support Vector Machines: An Application to Face Detection*. Proc. IEEE Conf. Computer Vision and Pattern Recognition, pp. 130-136.
46. H. Scheiderman and T. Kanade (1998). *Probabilistic Modeling of Local Appearance and Spatial Relationships for Object Recognition*. Proc. IEEE Conf. Computer Vision and Pattern Recognition, pp. 45-51.
47. A. Rajagopalan, K. Kumar, J. Karlekar, R. Manivasakan, M. Patil, U. Desai, P. Poonacha and S. Chaudhuri (1998). *Finding Faces in Photographs*. Proc. Sixth IEEE Int'l Conf. Computer Vision, pp. 640-645.
48. M.S. Lew (1996) *Information Theoretic View-based and Modular Face Detection*. Proc. Second Int'l Conf. Automatic Face and Gesture Recognition, pp. 198-203.
49. A.J. Colmenarez and T.S. Huang (1997). *Face Detection with Information-based Maximum Discrimination*. Proc. IEEE Conf. Computer Vision and Pattern Recognition, pp. 782-787.
50. The Unofficial DVD Specifications Guide - RGB to YCrCb Color Conversion.
<http://www.dvd-replica.com/DVD/coloryuv.php>. 12 Jun 2004.