



FACE DETECTION USING COST SENSITIVE ADABOOST ALGORITHM AND SKIN COLOR SEGMENTATION. (WITH PROBABILISTIC MULTIPOSE FACE DETECTION)

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Abstract:

One can say, face detection is a mature topic in image processing as there has been a lot of work done in this area. But yet there remain some challenges that need to be addressed to use face detection for real world scenario like video surveillance. We proposed a new approach to face detection which addresses most of the challenges by using a hybrid approach which consist of combination of cost sensitive Adaboost algorithm and skin color segmentation. By applying skin color segmentation in the initial stage we eliminate the unnecessary background area and applying the Haar feature extraction on skin color segmented area reduces the time requirement drastically. With the use of extensive training set and cost sensitive Adaboost algorithm, accuracy of the face detection is increased.

Keywords- Cost Sensitive AdaBoost algorithm, multipose, Haar feature extraction.

Introduction

Face detection is a very active research topic in the field of computer vision and pattern recognition, which is widely applied in the identity authentication; man-machine interface; visual communication; virtual reality; management of public security files; content-

based retrieval and many other aspects.^[3] Therefore, it requires that the face detection system with strong adaptability to all environments. It's the key for the face detection to find an effective method to extract the common features of human face which can describe face model, namely face modelling. The common features of human face include geometry, symmetry, texture etc. As the complexity of the image increases, it can't obtain accurate test results with only single feature information, then people pay more attention to the detection method with multi-feature information. As the number of feature

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for face detection increase its time requirement also increase. The other factors such as noise, light, color change, posture and facial expressions, and many other factors makes face detection more complex. Most difficult aspect of face detection is to develop a system which can be used in real time with accuracy and timeliness. Thus we proposed this hybrid approach which addresses both the issues. This method focuses on improved AdaBoost algorithm^[1] that reduces the misclassification error of traditional AdaBoost algorithm. The improved AdaBoost algorithm is called as Cost sensitive AdaBoost algorithm. It also uses skin color segmentation, and combines the two kinds of methods, thereby making full use of the advantage of high speed of skin color detection and high detection rate and low false acceptance rate of cost sensitive AdaBoost algorithm. One more enhancement proposed is the use of multipose face images in the training set as a special category besides faces and non faces images. This gives us the different set of classifiers than the traditional Adaboost algorithm will give. Thus improves the capability of algorithm to detect multipose faces from image^[10].

Skin Color Segmentation

The process of differentiation between skin color pixel and non skin color pixel can be defined as skin color segmentation. However, there are some difficulties in robust detection

of the skin color^{[4][6]}. The appearance of the skin tone can be affected by the ambient of the light and shadows. Also there are various skin color tones such as Asians skin that has big difference with Caucasians skin type. Thus to accurately identify the skin color pixel choice of color space is a crucial part.^[8] RGB color space is not suitable as it non-uniformity, mixing of chrominance and luminance data and high correlation between them. Thus we use YCbCr color space which has the luminance and chromaticity information. The effect of irregular illumination in an image is reduced in YCbCr color space due to separation of brightness information from the chrominance and chromaticity.

The conversion from RGB color space to YCbCr color space can be done using the following equation.

$$Y = 0.299R + 0.587G + 0.114B \tag{1}$$

$$Cb = B - Y. \tag{2}$$

$$Cr = R - Y. \tag{3}$$

There are different methods for skin color segmentation. In our approach, we identify the skin color using mean and covariance of chrominance color in a set of 100 skin sample images. The mean vector and covariance matrix is obtained from the total data of 100 skin samples by appending them together. This gives us a domain of skin color pixel which can be used to identify skin color pixel.

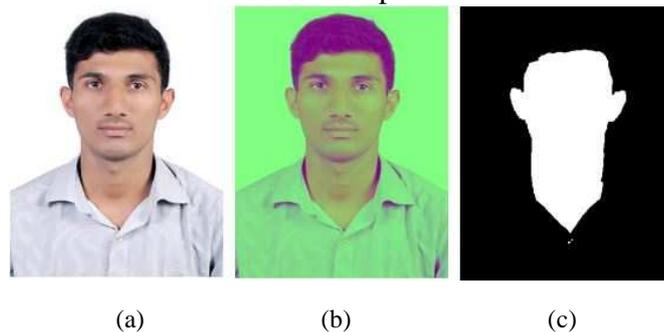


Fig. 1. (a) Original image (b) Image in YCbCr space (c) image using parametric single Gaussian model.

Cost Sensitive AdaBoost Algorithm

We use cost-sensitive AdaBoost algorithm that is based on the AdaBoost algorithm, followed by short CS-Boost algorithm, the specific process is as follows:^{[5][7]}

1) Given the training images $(x_1, y_1) \dots, (x_n, y_n)$, x is the image of the sample, $y = \{-1, 1\}$. Respectively non-face and face samples,

WeakLearn indicates Weak learning algorithm, T is the number of iterations, Constant cost $C > 1$;^[7]

2) The definition of the cost of misclassified positive and negative samples: $C_+ = \sqrt{C}$, $C_- = 1/\sqrt{C}$.

3) Initialized the weight of positive and negative samples:

$$w_{i,j} = \begin{cases} C_+ / (C_+ N_+ + C_- N_-) & \text{if } y_i = +1 \\ C_- / (C_+ N_+ + C_- N_-) & \text{if } y_i = -1 \end{cases} \quad (4)$$

Where, N_+ plus or minus the number of samples;

4) Let $t=1 \dots T$

a) Normalized weight

$$W_{i,j} = w_{i,j} / (\sum w_{i,j}) \text{ for } j=1 \text{ upto } n.$$

b) Used weight W_t and training samples, called weak classifier learning algorithm WeakLearn, get a weak classifier h_t .

c) Calculated Error rate of h_t :

$$\epsilon_t = \sum_{i=1}^n P_i |h_t(x_i) - y_i|, \text{ let } \alpha = \log(1 - \epsilon_t) / \epsilon_t \quad (5)$$

d) Let

$$W_{i+1,j} = \begin{cases} w_{i,j} \exp(-\alpha / C_s) & \text{if } y_i h_t(x_i) = +1 \\ w_{i,j} \exp(C_s \alpha) & \text{if } y_i h_t(x_i) = -1 \end{cases}$$

$$\text{where } C_s = \begin{cases} C_+ & \text{if } y_i = +1 \\ C_- & \text{if } y_i = -1 \end{cases} \quad (6)$$

Output the final strong classifier:

$$H(X) = \text{sign} [\sum_{t=1}^T \alpha_t h_t(x)] \quad (7)$$

Cost sensitive AdaBoost algorithm takes into account the cost of different types of classification errors, and tries to minimize the cost of wrong classification. As compared to AdaBoost algorithm, Cost sensitive AdaBoost algorithm taken into account the cost of the misclassification at two levels. First, initialize the weight of the sample according to misclassification costs. Another, consider cost

impact while the weight updating. The learning process will be more focused on the positive samples with wrong classification. This could get a higher detection rate and improve detection speed.

Cost Sensitive AdaBoost Based Face Detection Algorithm

This algorithm is based on the basic framework proposed by Viola and Jones.

A. Haar feature space

To give a face classifier speed as quickly as possible, complex features should not be used. Viola and Jones propose a class of simple rectangular features.^{[1][2]} The flexibility of these features is far less than other adjustable classifier. It's more sensitive for edges and other simple image structure, and relatively rough, only the horizontal and vertical directions are available. However, these features provide a rectangular image of the very rich characterization, which supports effective learning. Also in the calculation of high efficiency to make up their lack of flexibility. Characteristic value of the rectangular graph is black part of the black pixel values subtracting the white part of the black pixel value, one box indicates the rectangular area of the detection. In such an area, these four characteristics of the size and position can be arbitrarily chosen, assuming that the rectangular area the size of 24 x 24, then all of the rectangular characteristics of the data will be very large, more than 70,000. In order to perform fast computation of rectangular features, the algorithm uses a graph called the integral image representation. Position (x, y) points on the map with

this location and at the top of the left and all the pixels, expressed as

$$(x,y)=\sum_{x'=x_1,y'=y_1}^x i(x',y') \tag{8}$$

B. Feature selection and classifier construction

In the AdaBoost learning algorithm, each training sample has an initial weight, representative importance of current sample. According to each candidate weak classifier on the training set the classification results and the weight of the sample, you can calculate the weighted classification error rate of the current weak classifier in the training set. Each one selected a minimum misclassification rate of weak classifier; the minimum classification error rate of the value of cut-off point is the weak classifier threshold. Then according to this weak classifier obtained the results of the classification to update the weights of training samples, reducing the weight of the sample correctly classified, increasing incorrectly classified sample weight. The purpose is to make learning algorithm to concentrate on the sample of incorrectly classified. A detailed training process is as follows:

- 1) Given the training images $(x_1, y_1) \dots, (x_n, y_n)$, x is the image of the sample, $y = \{-1, 1\}$. Respectively non-face and face samples.
- 2) Initialization face and non-face images weight $w_{i,j} = 1/2m, 1/2n$, where $i = 0, 1$; m and n respectively for non-face and face the number of samples.
- 3) Let $t = 1, \dots, T$
 - a) Normalized weight,

$$W_{i,j} = W_{i,j} / \sum_{j=1}^n W_{i,j} \tag{9}$$

- b) For each feature f_j , trained a weak classifier h_j , calculated its weighted classification error

$$\epsilon_j = \sum_j w_{i,j} |h_j(x_i) - y_i| \tag{10}$$

- c) Select a weak classifier h_t of the smallest error rate ϵ_t , calculated

$$\beta = \epsilon_t / (1 - \epsilon_t) \tag{11}$$

- d) Update weights

$$W_{t+1,i} = W_{t,i} \beta^{1 - e_i} \tag{12}$$

where $\epsilon_t = 1$ misclassification,
 $\epsilon_t = 0$ correct classification

- 4) The final strong classifier:

$$H(x) = \text{sign} \left[\sum_{t=1}^T \alpha_t h_t(x) \right], \alpha_t = \log(1 / \beta_t) \tag{13}$$

The cascade structure of classifier Cascade structure could quickly filter out most of inhuman face region using the fewest possible terms, leaving the region is more likely to face sub-windows, for more complex calculations.

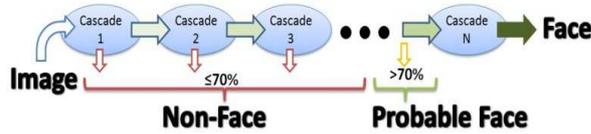


Fig.2. Cascade of classifier.

The training process of cascade structure is driven by some detected targets, beginning set the detection target expected to be achieved, namely, single-layer classifier to achieve the detection rate and false positive rate, the overall false positive rate of classifier.^[9] The specific process is as follows:

1) User-defined values of F : the maximum acceptable single-layer classifier misjudgement rate, and d values: namely, single-layer classifier minimum acceptable detection rate.

The user to select the classifier the final false positive rate F_{wp} ;

P : Single-layer classifier for the training sample set of human face.

4) N : Single-layer classifier for the training sample set of non-face.

5) $F_0=1.0, D_0=1.0, i=0$;

6) If $F_i > F_{wp}$

a) 1. $i=i+1, n_i=0 ; F_i= F_{i-1} ;$

b) $F_i > f * F_{i-1}$
 $n_i= n_i+1$

Pairs of P and N , with the AdaBoost algorithm for training a classifier with n_i characteristic: Using the current Multi-layer classifier to classify the validation sets, get the current multi-layer classifier D_i and F_i . Reduction

threshold of the current first i -layer classifier, until the current multi-layer classifier detection rate atleast $d * D_{i-1}$ (This action will affect F_i of the current classifier)

c) $N=\Phi$

d) If $F_i > F_{wp}$, Updated the non-face sets N with the current classification misclassification of multi-layer classifier non-face images.

Architecture of face detection system using cost sensitive AdaBoost algorithm and skin color segmentation.

The system consists of two phases:

- a) Training phase, and
- b) Detection phase.

In training phase we train cost sensitive AdaBoost algorithm to create cascade of classifier. This cascade of classifier will be used in detection phase. The following figure shows the flow of the system in detection phase. The block diagram of the system is shown below:

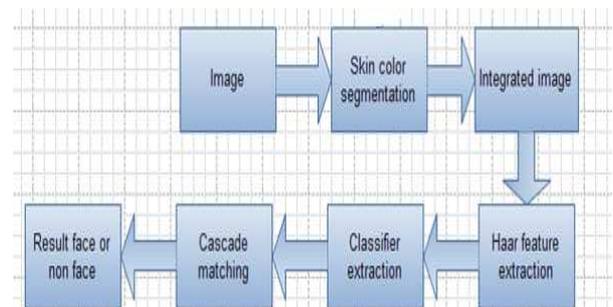


Fig. 3. Architecture diagram of face detection using Cost sensitive Adaboost algorithm and skin color segmentation.

Expected Result

This algorithm should be able to detect the faces from the images including the crucial multipose faces accurately with minimum false rate and as fast as possible.

Conclusion

We have proposed a hybrid approach for face detection which is a combination of cost sensitive Adaboost algorithm and Skin Color Segmentation. This approach aims to increase accuracy and to decrease time requirement of face detection system. It also has a new feature as probabilistic multipose face detection which

is an improvement over traditional face detection system particularly focusing on detecting faces in different poses.

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