

Video moving target binary image processing method based on OTSU

Xiaowei Han^{1, a}, Yuan Gao^{2, b}, Yong Cao^{3, c}, Zheng Lu^{4, d}, Dun Niu^{5, e}

¹College of Information Engineering Shenyang University
Liaoning Shenyang P.R. China

²College of Information Engineering Shenyang University
Liaoning Shenyang P.R. China

³College of Information Engineering Shenyang University
Liaoning Shenyang P.R. China

⁴College of Information Engineering Shenyang University
Liaoning Shenyang P.R. China

⁵North China Regional Air Traffic Management Bureau of CAAC, Hang An Road ,Beijing Capital International Airport, Beijing, 100621, China

^aHXW69@163.com, ^b307400614@qq.com, ^c543530681@qq.com, ^d774324529@qq.com, ^eniudun1203@163.com

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Abstract. Because of some pixels of foreground image which is binarized by fixed threshold, the Gaussian mixture model can not get accurate result. So we introduce OTSU to look for the threshold which satisfy maximum-between-clusters-variance, and it will get best divisional effect. It achieves adaptive threshold image dynamic adjustment of image binarization. Then use video data to validate, the result shows that the algorithm has better result by comparing with the traditional binarization algorithm in the video image.

Introduction

There is a lot of binarization methods of image, all kinds of methods have merits and faults for themselves, the different methods will get different results for the same image. Kittler^[1] and others put forward threshold method for selecting which combine gradient value of gray with gray statistics in image. The principle of Bernsen algorithm^[2] is to get the threshold of pixel with maximum-minimum pixels in the local window. Niblack^[3] algorithm based on the mean of neighborhood pixels of each pixel and variance of neighborhood pixels of each pixel to calculate the threshold of pixels. VFCM^[4] algorithm is binarization methods which combine with the method of global thresholding of maximal variance and the FCM method. NFCM algorithm^[5] is binarization methods which combine with the method of Niblack local threshold and the FCM method.

The Gaussian mixture of pixel level can effectively describe the multimodal background and can self-adaptively change with dynamic background. It can achieve better result on motion target detection, so it has been widely paid much attention. Stauffer^[6] and Grimson use the adaptive Gaussian mixture model to patch for every pixel with Gaussian mixture distribution. And update the model by online estimation, it can reliably handle effect of illumination change and the interference of background chaos. Zhang and others^[7] judge pixels of foreground with SVM on GMM method, and it can reduce error rates of complicated environment detection. R. T. Colin and others^[8-9] establish a background model of single Gaussian. Shimada and others^[10] reduce amount of the original algorithm by removing Gaussian distribution of combined part.

Compared with other methods of threshold selection, we use OTSU method. Based on the definition of variance, it can be seen that the bigger variance between cluster constitutes the bigger difference of foreground and background in image. If intra-class variance is small, it shows that the difference of intra-class part is small. As a result, the binary threshold which makes variance between cluster biggest means get the error probability of pixel of background and foreground, so we will use

adaptive threshold of OTSU to improve the effect of binarization with the detection methods of background modeling which is Gaussian mixture model. The result shows that the algorithm have better result of the binarization that compare with the traditional binarization algorithm in the video image.

Gaussian mixture model

GMM(Gaussian mixture model) consider the color value of pixel in the background accord with Gaussian distribution, Gaussian distribution of multiple different weights simulate color changes of pixel.

The definition of model. The observations of pixel (x, y) at time t in video frame I record I_t , a series of observation $\{I_1, I_2, \dots, I_t\}$ at different times can be thought random process which independent of others, can use K Gaussian expression:

$$M_{i,t} = M(\delta_{i,t}^2, \mu_{i,t}) \quad (1)$$

Among them, $i = 1, 2, \dots, K$. The value of K is determined by factors of computing performance, usually we will take 3 to 7. The probability distribution of point (x, y) at time t is :

$$P(I_t) = \sum_{i=1}^K w_{i,t} \eta(I_t, \Sigma_{i,t}, \mu_{i,t}) \quad (2)$$

Among them, $w_{i,t}$ is weight of i-gaussian distribution at time t, it reflects proportion which gaussian distribution appears, and $\sum_{i=1}^K w_{i,t} = 1$; $\eta(I_t, \Sigma_{i,t}, \mu_{i,t})$ is probability density function whose first mean is $\mu_{i,t}$, covariance is $\Sigma_{i,t}$ at time t:

$$\eta(I_t, \Sigma_{i,t}, \mu_{i,t}) = \frac{1}{(2\pi)^{\frac{n}{2}} |\Sigma|^{\frac{1}{2}}} \exp(-\frac{1}{2}(I_t - \mu_t)^T \Sigma^{-1}(I_t - \mu_t)) \quad (3)$$

Model update. If the current pixel values and mean of Gaussian distribution satisfy $|I_t - \mu_{i,t}| \leq D \delta_{i,t}^2$, we will consider the match is successful, and experience value of D is 2.5. If there are multiple matches, we will choose the best one. It will adjust weights of distribution when matching success:

$$w_{i,t} = \alpha N_{i,t} + (1 - \alpha) w_{i,t-1} \quad (4)$$

Among them, α is learning rate, value is between 0 and 1, the bigger α is, the fast weight update, and vice versa; $N_{i,t}$ is 1 for the distribution k of matching, the others is 0, it can increase the weight of distribution, reduce weight which does not match the distribution.

The parameters will be updated for the model which match the current pixel :

$$\mu_t = \rho I_t + (1 - \rho) \mu_{t-1} \quad (5)$$

$$\delta_t^2 = \rho (I_t - \mu_t)^T (I_t - \mu_t) + (1 - \rho) \delta_{t-1}^2 \quad (6)$$

The ρ is the other learning rate, and it's value is $\rho = \alpha \eta(\frac{I_t}{\mu_k}, \delta_k)$, η is density function of Gaussian. It remain unchanged if don't have matched distributes.

Adaptive threshold of OTSU. Because of the some pixels of foreground image which is binarization by fixed threshold in can not get accurate result. And although the overall effect can meet the requirements, but the processing of detail is not ideal. So we introduce OTSU^[11] to look for the threshold which satisfy maximum-between-clusters-variance, and it will get best divisional effect. The main idea is the gray of picture whose level is 1~m, the picture is divided two groups of

background and foreground. The percentage points of foreground is p_0 , the mean grey is l_0 ; the percentage points of foreground is p_1 , the mean grey is l_1 . the mean of all gray is :

$$l = p_0 \times l_0 + p_1 \times l_1 \quad (7)$$

Then

$$v = p_0 \times (l_0 - l)^2 + p_1 \times (l_1 - l)^2 \quad (8)$$

The t is best threshold of segmentation with traversal method which make v maximum.

The realization which combine GMM with OTSU

Implementation steps.

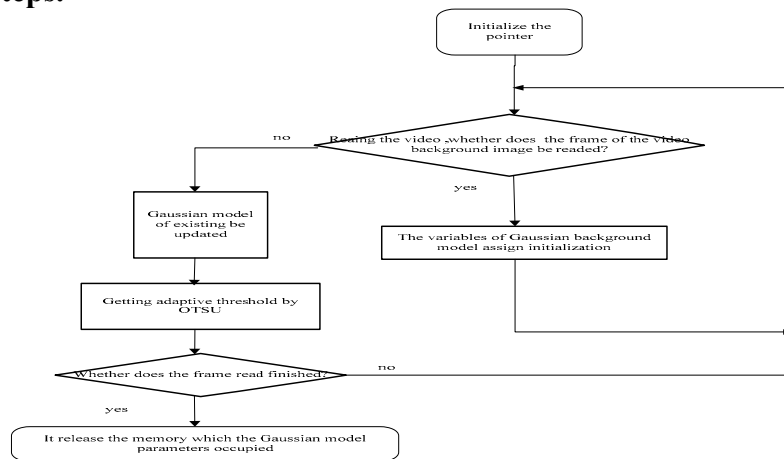


Figure 1. The implementation process

Testing result and analysis. We test program with VC++6.0 and OpenCV, selecting video of collection for test, the result of test as shown in Figure 1.

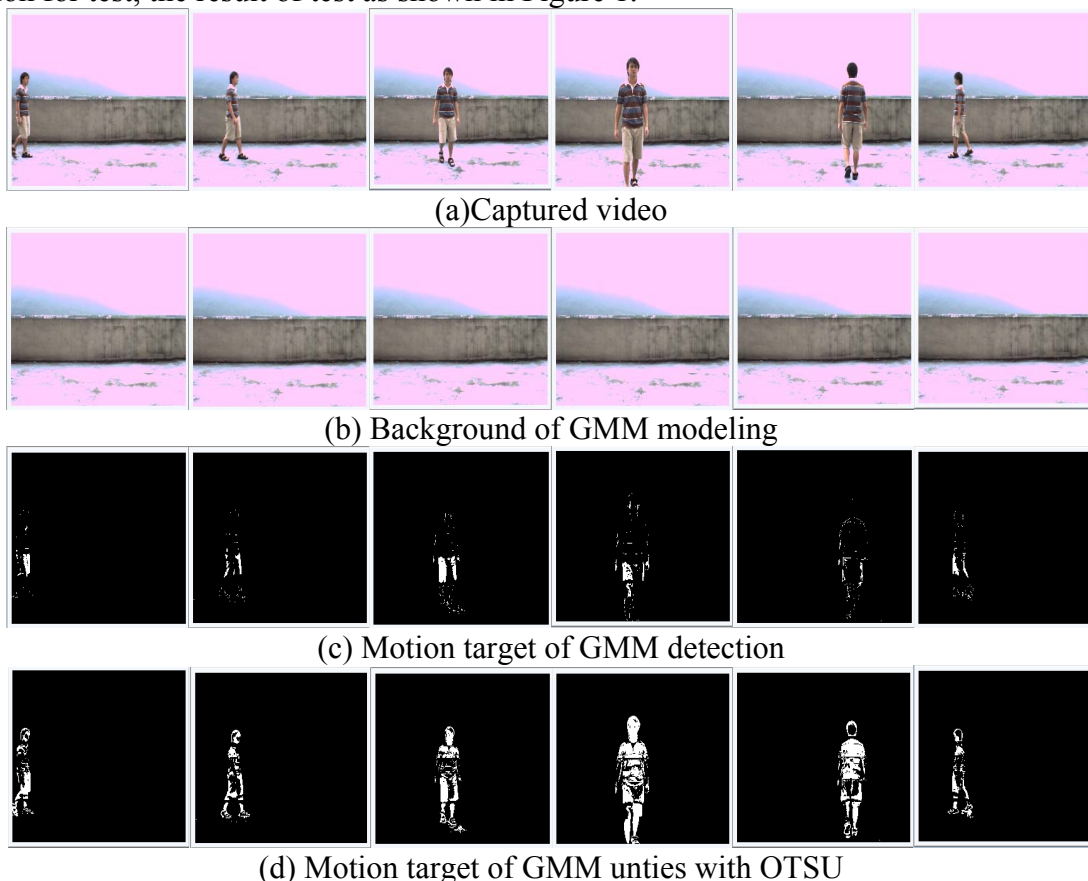


Figure 2 Motion target comparison picture of GMM unities with OTSU

The captures is video of motionless gather, and the main target of detection is moving object. Figure 1(b) and Figure 1(c) show that GMM background modeling have a good sensitive and produce less disturbing in motion object region. It was shown that the overall effect can meet the requirements from Figure 1(c) ,but the processing of detail is not ideal.

Summary

This paper propose a algorithm of moving target detection which combine with Gaussian Mixture Modeling and OTSU under studying moving object detection algorithm on readitional static camera . Because of some pixels of foreground image which is binarization by fixed threshold can not get accurate result. And although the overall effect can meet the requirements ,but the processing of detail is not ideal. So we introduce OTSU to look for the threshold which satisfy maximum-between-clusters-variance, and it will get best divisional effect. The paper combined with the GMM and OTSU by technology of OpenCV. The experimental results show that moving target detection of Gaussian Mixture Model have a good effect on adaptive threshold of OTSU. It is a beginning for the detective target processing by adaptive threshold of OTSU. We will use the method this paper work out in future research work about motion target detection of Gaussian Mixture Modeling.

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