Foreground Moving Object Detection Using Background Subtraction

Chintala Amala¹, Melam Nagaraju²

¹(ECE, PSCMR College of Engineering & Technology, JNTUK Kakinada, India) ²(Information Technology, Gudlavalleru Engineering College, JNTUK Kakinada, India) Corresponding Author: Chintala Amala

Abstract: The segmentation/Detection of a moving object is one of the important step in computer vision application, such asremote sensing, medical imaging, traffic surveillance,machine/robot vision, microscopic imaging etc. In this paper "Gaussian Mixture model" for backgroundsubtraction/foreground detection has been applied which computes a foreground mask on a moving object (which is either color video frame or series of a gray scale image). The blobanalyzer has also been used to calculate statistics of identified region in a binary image. The result shows that the proposed algorithm can detect the moving object effectively while occluded by the box.

Keywords -Background Subtraction, Blob Analysis, Gaussian Mixture Model.

Date of Submission: 17-02-2019

Date of acceptance: 03-03-2019

I. INTRODUCTION

In computer vision application the real time segmentation of a moving object is an important step. The segmentation of amoving object involves "Background Subtraction". The effective of a moving object for a human being is quite easywhile for a machine/computer it is a difficult task. So it is important to develop an algorithm which is quite adaptable and can be easily implementable [1]. The Background Subtractionmethod is chosen for the real time segmentation of a moving object because it is one of the important/popular method. The detection of a moving object is obtained from the difference between the current frame & the reference frame orbackground image. There is a various application of background subtraction in computer vision which provide very important prompts [2]. For Foreground detection; current frame — Background model > Threshold. The threshold value is chosen such that it will reduce the erroneous values [3], [4].

Background modeling consists of two important majorsteps:

1) Background Initialization

2) Background Update

In background initialization, the Initial model of thebackground is calculated while in background update the model is updated such that it will adapt the various changes in the scene. There are some difficulties which arise while performing the Background Subtraction and needs to be taken into consideration:

- Gradual & Sudden illumination changes
- Motion changes due to Camera oscillation, highfrequency background, another objects(such as tree, branches ,sea etc.)
- > Shadow
- Changes in Background geometry[5]

II. RELATED WORK

The Background subtraction methods have found differentways to dealing with these type of problems [6]. When we areabout to model the background it is very important that thebackground model should be adaptable towards the variouschanges in the frames. The background model should adjustitself such that it will reduce the difficulties arising as statedabove. Due to illumination changes there is a problem forbackground modeling. Authors in reference [7] developed amethod where every pixels of the background is modeled withKalman filter such that it works well in the illuminationchanges. Gaussian mixture model is a great model used todetermine that the pixel whether it belongs to background ornot. Sometime the detection of a moving object become quitecomplex due to presence of noise in the scene. The researchershave developed many ways to reduce noise. Mostlymorphological operations are used for the reduction of noisebecause it is one of the popular method for noise reduction [1],[8].The modeling of the Background is not stationary or it is muddled

thenit becomes quite difficult to choose a perfect background. Forthis type of situation we have to use nonparametric model forbackground [9]. The non-parametric model can rapidly adjust changes in the scene which helps in the complex detection of the moving object [10]. The detection of moving object when it is seen properly in all frames of a video is easy ascompared to the detection of the object which is occluded insome frame of video. So there are different algorithms for the detection of moving object when it is occluded or not occluded[11]. This paper deals with the detection of moving object when it is occluded in some frames of a video.

This paper is organized in the following sections. Section III deals with proposed algorithm, section IV describes work flow of the proposed method, section V presents the results of the proposed method, section VI presents the summary of figures and sectionVII shows the conclusion of the algorithm. For the detection of moving object we took an example video from MATLAB 2016a.

III. PROPOSED METHOD

The proposed method consists of the following steps:

1. Averaging:

Averaging is one of the important step in computer visionapplication because it reduces noise. Basically it accumulates the image pixel that computes the average value of the pixel in the given set of images. This is a influential noise reduction technique that has power to reduce noise while simultaneously it also estimates the feature. Image averaging actually increases the Signal to Noise ratio (SNR) of an image such that it reduces the noise or high frequency image component is filtered out.

The block diagram of the proposed method is shown in figure1.



Figure 1: Block diagram of the proposed method

2. Background Subtraction using GMM

A Gaussian Mixture Model is a parametric probabilitydensity function which is a weighted sum of Gaussian component densities & given as:

$$p(X / \lambda) = \sum_{i=1}^{M} w_i g(X / \mu_i, \Sigma_i)$$
(1)

where 'X' is an n-dimensional continuous value data, W_i is the mixture of weights for i= 1 to M.

A Gaussian Mixture Model is used for modeling of thebackground is because it is one of the greatest model forbackground modeling. It models all different type of pixels[12], [13].

3. Blob Analysis

The blob analysis is one of the important step for the detection of moving object. Basically it recognize the moving object in a scene. It works on a binary image & stands for "Binary Large Object". After using this we can understand that the large connected moving pixels is our target & the smallconnected pixel is not of our interest (because these are noise)[14], [15]

IV. WORK FLOW OF THE PROPOSED METHOD

In this section of the paper, we describe the work flow of theproposed algorithm. In this section first of all an input videowhich is of 380kB in size & having 508px*362px of a frame isfetched. After this operation averaging of the fetched imagewhich reduces the noise in the given image is computed. Forthis purpose 3*3 averaging box filter has been used. After thisprocess, we applied an important method "BackgroundSubtraction" using Gaussian Mixture Model. After thisoperation a noise free foreground image has been obtained.Having noise free foreground image we have applied blobanalysis which estimates the large cluster pixels value of theimage which is of our interest. When blob analyzer estimates the large cluster pixel image then a mask is headed on thatpixels and consequently the moving object is detected.

V. **RESULTS**

In this section the detection result of a moving objectoccluded by the box have been shown. The proposed algorithm is implemented in MATLAB 2016a. The input video is of 380kB in size and having frame rate 15 frame/sec.



Figure 2: Input Image

Figure 4: Detected ball before occlusion



Figure 3: Foreground Image (Noise Free)

Figure 5: Detected ball after occlusion

Each frame is of 508px*362 px. In this simulation the no.of initial video frame for training the background model istaken as 1000, the threshold is chosen as 0.7, number of Gaussian modes in mixture model

is 5 & variance for initializing the new Gaussian mode is taken as 1000*1000 & the minimum blob area is taken as 100.

VI. **Summary Of Figures**

I have inserted four figures of simulation result. Figure 2 is the 21st frame of the input video which is fetched in MATLAB 2016a successfully. Figure 3describes the foregroundimage of the frame which is obtained after backgroundsubtraction using Gaussian mixture model. Since my objectiveis to detect the moving object so my concentration is on onlythe moving ball not on the box so I used blob analyzer whichworks on binary image and keep the cluster of moving pixel of the image and then using bounding box a rectangular box isdrawn over moving object. Figure 4shows the detection of moving object before occlusion while Figure 5shows detection of moving object after occlusion.

VII. **CONCLUSION**

From the above results it is concluded that the backgroundsubtraction is one of the important step for the detection of moving object. The result also shows that the backgroundsubtraction gives satisfactory result for the detection of themoving object while it is occluded.

References

- [1]. Stauffer, Chris, and W. Eric L. Grimson. "Adaptive background mixturemodels for real-time tracking," Computer Vision and PatternRecognition, IEEE Computer Society Conference on. Vol. 2. IEEE, 1999.
- Olugboja, Adedeji, and Zenghui Wang. "Detection of Moving Objectsusing Foreground Detector and improved morphological [2]. filter,"Information Science and Control Engineering (ICISCE), 3rd InternationalConference on IEEE, 2016.
- [3]. Vahora, Safvan, Narendra Chauhan, and NileshPrajapati. "A robustmethod for moving object detection using modified statistical meanmethod," International Journal of Advanced Information Technology 2.1:65. 2012
- Bobick, Aaron F., and James W. Davis. "The recognition of humanmovement using temporal templates," IEEE Transactions on patternanalysis and machine intelligence 23.3: 257-267. 2001 [4].
- Rajan, P., and Dr S. Prakash. "Moving Foreground Object Detection and Background Subtraction Using Adaptive-K GMM: A [5]. Survey, "IJARCSMS, ISSN: 2321 7782, 2014.
- Haritaoglu, Ismail, David Harwood, and Larry S. Davis. "W4: real-timesurveillance of people and their activities," IEEE [6]. Transactions on patternanalysis and machine intelligence 22.8: 809-830. 2000.
- Ridder, Christof, Olaf Munkelt, and Harald Kirchner. "Adaptivebackground estimation and foreground detection using kalman-[7]. filtering,"Proceedings of International Conference on recent Advances inMechatronics, 1995.
- [8]. YumnahHasan, Muhammad UmairArif, Amad Asif and RanaHammadRaza, "Comparative analysis of vehicle detection in urban trafficenvironment using Haar cascaded classifiers and blob statistics." FutureTechnologies Conference (FTC). IEEE, 2016.
- Piccardi, Massimo. "Background subtraction techniques-a review,"Systems, man and cybernetics, 2004 IEEE international [9]. conference on.Vol. 4. IEEE, 2004.
- Elgammal, Ahmed, David Harwood, and Larry Davis. "Non-parametricmodel for background subtraction," European conference on [10]. computervision. Springer Berlin Heidelberg, 2000.
- Iraei, Iman, and Karim Faez. "Object tracking with occlusion handlingusing mean shift, Kalman filter and Edge Histogram," [11]. PatternRecognition and Image Analysis (IPRIA), 2015 2nd InternationalConference on. IEEE, 2015.
- [12]. H. Zhou, H. Kong, J.M Alvarer, D. Creghton and S. Nahavandi, "Fastroad detection and tracking in aerial videos," Intelligent VehiclesSymposium Proceedings, IEEE, 2014.
- Rother, Carsten, Vladimir Kolmogorov, and Andrew Blake. "Grabcut:Interactive foreground extraction using iterated graph cuts," [13]. ACMtransactions on graphics (TOG). Vol. 23. No. 3. ACM, 2004
- Hailing Zhou, Lyndon Llewellyn, Lei Wei, Doug Creghton and SaedNahavandi, "Marine object detection using background [14]. modelling andblob analysis," Systems, Man, and Cybernetics (SMC), InternationalConference on. IEEE, 2015. Chantakamo, Anchisa, and MahasakKetcham. "The multi vehiclerecognition using hybrid blob analysis and feature-based,"
- [15]. InformationTechnology and Electrical Engineering (ICITEE), 7th InternationalConference on. IEEE, 2015.

Chintala Amala" Foreground Moving Object Detection Using Background Subtraction" International Journal Of Engineering Science Invention (Ijesi), Vol. 08, No. 02, 2019, Pp 59-62