



Removing Object's Shadow with Shadow Removal Percentage

Nitin M. Shivratriwar
M.E. [I.T.] P.R.M.I.T.&R Badnera
P.R.M.I.T.&R Badnera Badnera, India
nitinshivratriwar@rediffmail.com

Prof. .S .S .Kulkarni
I.T .Dept .P.R.M.I.T&R. Badnera
P.R.M.I.T.&R Badnera Badnera,India
shree2610@gmail.com

V.K.Patil
H.O.D.I.F.Poly.Badnera
Polytechnic Badnera Badnera,India
Vivekkpatil@rediffmail.com

Abstract: In this paper we introduce, a shadow detection and removal method for moving objects especially for humans and vehicles. An efficient method is presented for detecting and removing shadows from foreground figures. We assume that the foreground figures have been extracted from the input image by some background subtraction method. The homogeneity property of shadows is explored in a new way for shadow detection and image division technique is used. The process is followed by filtering, removal, boundary removal, & removal validation.

Keywords: Shadow removal, homogeneity, umbra, penumbra

I. INTRODUCTION

Shadow is a common problem that one could occur in motion estimation of daytime traffic scenes. A shadow is an area where direct light from a light source cannot reach due to obstruction by an object. A shadow occurs when an object partially or totally occludes direct light from a source of illumination. There are two types of Shadow, self Shadow & cast shadow, self shadow occurs in the portion of a object which is not illuminated by direct light. Cast Shadow is the area projected by the object in the direction of direct light. Self Shadow is again divided into two parts: Shading & Interreflection. & Cast Shadow is also again divided into two parts: Umbra & Penumbra, Umbra, is the darkest part of the shadow. In umbra, the light source is completely occluded, Penumbra, is the region in which only a portion of the light source is obscured by the occluding body. We needed shadow removal because Shadows cause tracking, segmentation or recognition algorithms to fail. Shadows have proven to be a large source of error in the detection and classification of vehicles. Real images with shadows can't be used for image synthesis.

Shadows can cause object merging, object shape distortion etc., causing error in object tracking and classification. Many shadow detection or removal is based on priori information, such as the geometry of the scene or the moving objects and the location of the light source while others are based on shadow attributes. There are three main shadow attributes, firstly, shadows or moving shadows are attached to their respective obstruction object for most of the time, secondly, transparency which is that shadow always makes the region it covers darker and lastly, homogeneity which is the ratio between pixels when illuminated and the same pixels under shadows can be roughly linear.

An image division technique is used to divide the image (between background and foreground). Image division is used because it highlights the homogeneity

property or attributes of shadow. After the image division process, adaptive thresholding process is used to remove the umbra region and projection histogram analysis is applied to the resultant image to remove the penumbra region. The global thresholding process is applied to the resultant image (after image division) in order to get the shadow's blob [2] [3].

The shadow detection method is applied to gray level images taken by a stationary camera. The Canny edge detector is used to both the foreground and background figures. After that, image subtraction is performed on the foreground edges and background edges to extract the object's edges. Object recovery process is applied to the object's edges to recover object shapes on the basis of the information in the object's edges and attributes of shadow. There are three rules or attributes of shadow. Firstly, bright foreground pixels is preserved because they are impossible to belong to shadows. Secondly, foreground pixels with attributes different from the attributes of shadow are preserved and lastly, foreground pixels nearby object edges are preserved [5].

The approach is applied for human or pedestrian and it is based on priori information which is object moment and object orientation. In order to precisely remove the unwanted shadows, presents a histogram projection method to separate each pedestrian from moving region first. Then, a coarse-to-fine approach is applied for detecting the boundaries between the pedestrian and its shadow. At the coarse stage, a moment-based method is applied to estimate the orientation of the detected pedestrian [6].

According to the orientation and silhouette features of the detected regions, a rough approximation of the exact shadow area can be detected. At the fine stage, the rough approximation of the shadow region is further refined through Gaussian shadow modeling. The major difficulty in shadow modeling is the choice of the proper model, which can reflect various appearances of shadows at different orientation and lighting.

The algorithm is also based on priori information plus a little bit of shadow attributes. The algorithm tries to locate

object-shadow boundary in order to detect the shadow. This was done by creating one or more straight lines to approximate the boundary between vehicles and their associated shadows [7].

These lines are located in the image by exploiting both local information e.g. statistics in intensity differences and global information e.g. principal edge directions. This method does not assume a particular lighting condition, and required no human interaction or parameter training. Experiments on practical real-world traffic video sequences demonstrate that this method is simple, robust and efficient under traffic scenes with different lighting condition.

II. MOTIVATION

Many techniques which are based on homogeneity property of shadows assume that ratio between pixels when illuminated and that are when subjected to shadow is constant. But, the ground truth data show that the ratio is highly dependent on illumination in the scene and hence shadow detection will not be effective in case if the ratio is assumed to be constant. In other words, the ratio is always changing due to different illumination for different scenes. That is why; an adaptive thresholding technique can be used to improve or to solve the problem.

Most of the shadow detection that are developed based on homogeneity property or image division technique just assume the ratio is always constant and use fix threshold value for the thresholding process (global thresholding). In addition, most of the shadow detection that use image division technique always use an additional technique after the image division process because the image division process is used to highlight the homogeneity property of shadow especially the umbra region but not penumbra. That is why; so many additional processes have proposed in order to remove the penumbra region such as multi-gradient analysis and projection histogram analysis.

III. PROPOSED SYSTEM DESIGNING & PLANNING

A. The Algorithm:

The figure 1 shows the flowchart of the removal system which is the main algorithm of the system that has been proposed and the resultant image for every process is shown with an example in figure 4. It has been assumed that the input is obtained from some background subtraction. The complete process in figure 1 is explained in the following sections.

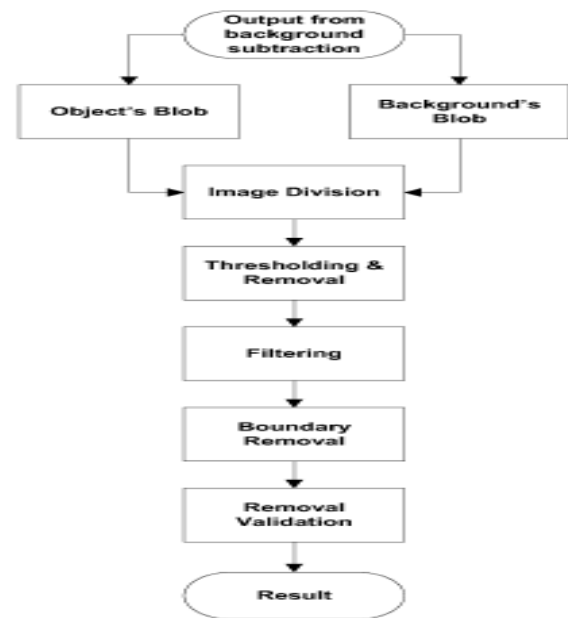


Figure 1. Overall algorithm of proposed shadow removal technic

B. Image Division:

In this process, the object's blob, $ob(x, y)$, $\{x, y \in Z^2\}$ is divided with the background's blob, $bk(x, y)$, $\{x, y \in Z^2\}$. It has been said before that the purpose of image division is to highlight the homogeneity property of shadows. Resultant image after the division process is multiplied with a constant for the purpose of increasing the signal of the resultant image. In this case, the constant value is 100 (Eq.1). The result of this process is define as $Img_Div(x, y)$.

$$Img_Div(x, y) = \frac{ob(x, y)}{bk(x, y)} \times 100, \forall x \in X, \forall y \in Y \quad (1)$$

C. Thresholding:

Thresholding is the simplest method of image segmentation. From a gray scale image, thresholding can be used to create binary images. The function of thresholding is to decide the shadow's blob in the resultant image after the image division process (Img_Div). Most of the existing approaches mark pixels lying between a certain range of value in the division image as belonging to a shadow. However, the range is highly dependent on the illumination in the scene. In addition, different scenes also produce different level at illumination. An adaptive Threshold had proposed in order to solve this type of problem. However, this technique has been tested with a few video samples and it seems that this technique does not work effectively compared to a fixed threshold. It is because the range of adaptive threshold depends on the average value of the division image and this will cause that the threshold value (range) that belongs to the shadow change randomly. Because of this problem, in this proposed technique, the range has been set according to the scene (Eq.2) and this is done by studying the histogram of the division image over a few samples.

$$Img_Th = \begin{cases} 1, t_{\min} \leq Img_Div \leq t_{\max} \\ 0, otherwise \end{cases}, \forall(x, y)$$

D. Filtering:

The function of filtering is to improve the resultant image after the thresholding process (Img_Th) and to find the biggest region which is predicted as the shadow's blob or shadow region. Filtering process consist of filling, erosion and dilation to enhance the image and labeling to predict the shadow. It is assumed that the biggest blob after the labeling process or connected component process as a shadow region.

E. Boundary Removal:

The purpose of boundary removal is to remove the shadow's boundary. The first step in this process is to get the coordinates of the boundary (object's blob). This is also called as boundary tracing process. After that, each boundary pixel and its neighbor is checked whether it is a shadow pixel or not. In this case, neighbor pixels that are only located in the horizontal, vertical, and diagonal (45 and -45 degree) of the boundary pixel with certain offset (range between neighbor pixels and boundary pixel) are checked.

Figure 2 shows the boundary pixel (grey colored pixel that is located at the centre) and the neighboring pixels (black Colored pixels that are located at the horizontal, vertical and diagonal of the boundary pixel) with an offset value equal to two. So, if the boundary pixel or its neighboring pixels are detected as shadow pixels, the boundary is assumed as a pixel from the shadow's boundary. At this stage, the boundary pixel and its neighboring pixel is removed (turn the pixel into white color).

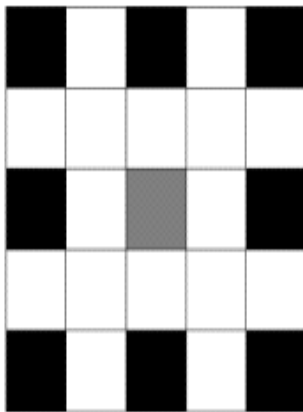


Figure 2. Neighbor pixels checked around a boundary pixel with an offset of two

F. Removal Validation:

Removal validation process consists of two sub processes which are the Percentage Checking process and the Vertical Scan process. The function of the Percentage Checking process is to check whether the removal process was correct or not. This is done by checking the percentage of area that has been removed in the removal process (Eq.3), where BR represent the percentage of area that has been removed over the area of whole object's blob. Based on the study and analysis of sample images, the shadow removal is correct if the percentage value is within a range that is dependent on a scene (Eq.4), where RV, percent_min and percent_max represent removal validation result, minimum percentage and maximum percentage (the range). This percentage range will be explained later in section IV. If the percentage value does not fall in that range, it is assumed that the removal did not work correctly.

$$BR = \frac{\text{area that has been removed}}{\text{object's blob area}} \times 100 \quad (3)$$

$$RV = \begin{cases} \text{true, } \text{percent_min} \leq BR \leq \text{percent_max} \\ \text{false, otherwise} \end{cases} \quad (4)$$

The second sub process is the Vertical Scan process which will check which part of the object's blob is predicted as a shadow region. In the Filtering process, it is assumed that the biggest blob after that labeling process is the shadow's region. However based on the study of input samples, sometimes, the second biggest blob is the correct shadow region and the biggest blob is not a shadow region.

Figure 3 shows an example where the Filtering process has done a wrong prediction and based on the analysis this biggest blob (wrong predicted shadow's region) is always located at the center of the object's blob.

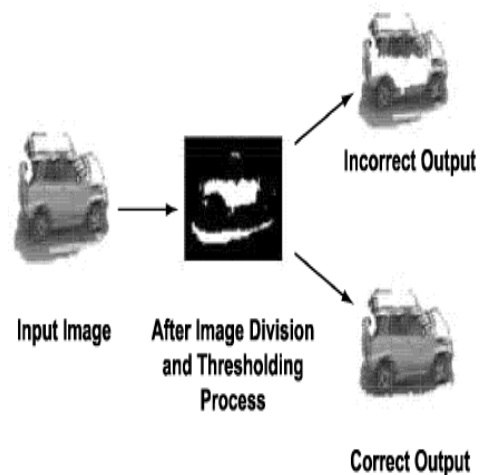


Figure 3. Vertical scan to determine correct removal

So in the vertical scan process, a vertical scan is performed though the centroid of the object's blob just to make sure that the predicted shadow's region is not located at the center of object's blob. However, the Vertical Scan process can only be applied on certain scenes. Some scenes are not suitable because it will only cause a poorer result. The complete process of shadow removal is shown in following figure

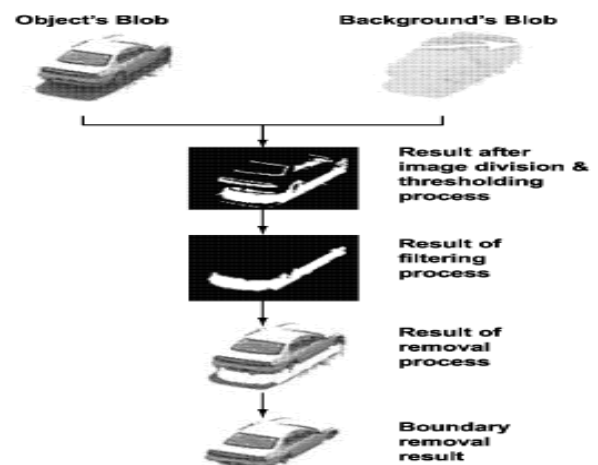


Figure 4. Shadow removal process

In figure 5, three results of the shadow detection and removal are presented. In the first scene, the input is shown in figure 5a and the output is shown figure 5b. For this scene, the

Vertical Scan process is disabled because the object orientation is not suitable application of process. This is also true for the third scene (figure 5e and 5f). In the first scene, the range of shadow pixel has been set from 20 to 50 and the percentage value for the Percentage Checking process (see Removal Validation process) is 8% to 45%. For the second scene (figure 5c and 5d), the Vertical Scan process is applied to prevent errors that have been mentioned before, in figure 4.

The shadow pixel range is set from 60 to 98 and percentage value for the Percentage Checking process is 8%-33%. For the third scene, shadow pixel range is from 45 to 75 and the percentage ranges are similar to the second scene. Table 1 show the accuracy or the percentage of correct results based on the scenes that have been shown in figure 5. In each scene, the vehicle is monitored and analyzed for a period of time and the overall success rate is calculated. This is calculated by the percentage of result from every video sample, $PV_i \{i=1..N\}$ (by getting the number of frames that have the correct result over the number of frames) and then, to get the average percentage of correct result from video samples in the same scene (PSS). Eq. 5 and Eq. 6 are the formulas that are applied in this analysis where the PV, and PSS represent the percentage of correct result from a video samples, number of video samples in a scene and percentage of correct removal in a scene.

$$PV = \frac{\text{no. of frames with correct result}}{\text{no. of frames in video sample}} \quad (5)$$

$$PSS = \frac{\sum_{i=1}^N PV_i}{N} \quad (6)$$

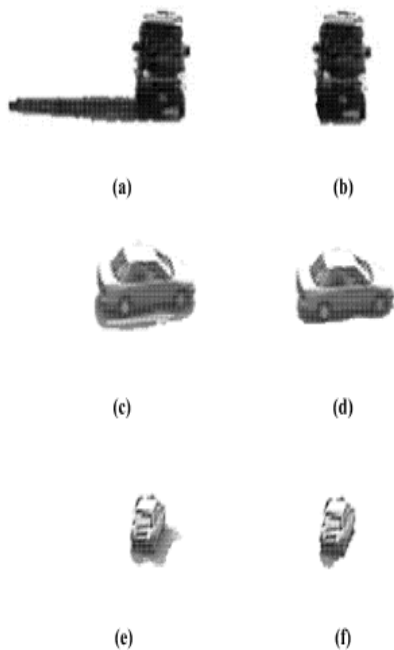


Figure 5. Shadow removal results for several scenes (before a, c, e and after b, d, f).

Table: 1 Scene and Percentage of correct removal

TABLE I SCENES AND PERCENTAGE OF CORRECT REMOVAL	
SCENE	Percentage of Correct Removal (%)
1 st	100.00
2 nd	91.39
3 rd	69.18

IV. CONCLUSION

A proposed method to detect attached shadow has been presented. The, target sequence is a daytime traffic scene involving different types of orientation and scenes. Figure 5 shows the result with different types of object size, orientation and scenes.

Working with outdoor images as input is very challenging as the illumination not only changes slowly as the daytime progresses but may change rapidly due to the changing weather conditions as well as due to the passing objects. So, the proposed technique especially the Removal Validation process will assist the removal process in order to detect whether there are shadows present in an image.

The proposed method improves object detection, which is a very critical task in video surveillance and vision-based traffic monitoring systems.

V. ACKNOWLEDGMENT

We wish to thank Prof. S.S.Kulkarni for their gentle support in developing this shadow removal system. We also desire to thank Dr.A.S.Alvi and Prof. S.A.Dhopte for their interest in the present work and for useful discussions.

VI. REFERENCES

- [1]. Mohamed Ibrahim M, Anupama R, "Scene adaptive Shadow detection algorithm." Transactions on Engineering Computing and Technology, V2, December, pp 88-91, 2004.
- [2]. Alessandro Bevilacqua, "Effective shadow detection in Traffic Monitoring Applications." Journal of WSCG, VII, pp 57-64, 2003.
- [3]. A. Bevilacqua, M. Roffilli, "Robust denoising and moving shadows detection in traffic scenes." Proc. Of IEEE Computer Society Conference on Computer Vision and Pattern Recognition, Kauai Marriott, Hawaii, pp 1-4, 2001.
- [4]. Falah E. Alsaqr, Yuan Baozong, "Moving Shadows Detection in Video Sequences." Proc. of 7th ICSP 04, V2, pp 1306-1309, 2004.
- [5]. J.M. Wang, Y.C. Chung, C. L. Chang, S. W. Chen, "Shadow Detection and Removal for Traffic Images." Proc. of IEEE international Conference on Networking, Sensing and Control, Taipei, Taiwan, VI, pp 649-654, 2004.
- [6]. Chia-Jung Chang, Wen-Fong Hu, Jun-Wei Hsieh, Yung-Sheng Chen, "Shadow Elimination for Effective Moving Object Detection with Gaussian Models." Proc. of 16th International Conference on Pattern Recognition (ICPR '02), V2, pp 540-543, 2002.
- [7]. Angie W. K. So, Kwan-Yee K. Wong, Ronald H. Y. Chung, and Francis Y. L. Chin. "Shadow Detection for Vehicles By

- Locating The Object-Shadow Boundary Proc. of IASTED Conference on Signal and Image Processing (SIP 2005), pp 315-319, 2005.
- [8]. A Leone, C. Distanti, F. Buccolieri: "A texture-based approach for shadow detection" Proceeding of IEEE Conference on Advanced Video and Signal Based Surveillance, 371 - 376, 2005. 17
- [9]. A. Leone, C. Distanti, N. Ancona, E. Stella, P. Siciliano: "Texture analysis for shadow removing in video-surveillance systems" Proceeding of IEEE International Conference on Systems, Man and Cybernetics, 6325 - 6330 vol.7, 2004.
- [10]. E. Salvador, A. Cavallaro, T. Ebrahimi: "Shadow identification and classification using invariant color models" Proceeding of IEEE International Conference on Acoustics, Speech, and Signal Processing, (ICASSP '01), 1545 - 1548 vol.3, 2001.
- [11]. C. X. Jiang, M. O. Ward: "Shadow Segmentation and Classification in a Constrained Environment" CVGIP: Image Understanding Volume 59, Issue 2, 213-225, 1994.
- [12]. K. Siala, M. Chakchouk, F. Chaieb, O. Besbes: "Moving shadow detection with support vector domain description in the color ratios space" Proceedings of the 17th International Conference On Pattern Recognition (ICPR 2004), 384 - 387 Vol.4, 2004.
- [13]. R. Cucchiara, C. Grana, M. Piccardi, A. Prati: "Detecting moving objects, ghosts, and shadows in video streams" IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI), Vol. 25(10), 1337 - 1342, 2003.
- [14]. Dong Xu, Xuelong Li, Zhengkai Liu and Yuan Yuan: "Cast shadow detection in video segmentation" Pattern Recognition Letters Vol. 26(1), 91-99, 2005.