

A new method for vehicles detection and tracking using information and image processing

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ABSTRACT

In this article, a new method of vehicles detecting and tracking is presented: A thresholding followed by a mathematical morphology treatment are used. The tracking phase uses the information about a vehicle. An original labeling is proposed in this article. It helps to reduce some artefacts that occur at the detection level. The main contribution of this article lies in the possibility of merging information of low level (detection) and high level (tracking). In other words, it is shown that many artefacts resulting from image processing (low level) can be detected, and eliminated thanks to the information contained in the labeling (high level). The proposed method has been tested on many video sequences and examples are given illustrating the merits of our approach.

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1. INTRODUCTION

The intelligent transportation system [1]-[5] are axis of great importance and topicality. It not only helps to smooth the flow of vehicles but also reduces the number of accidents. In this context, the detection and tracking of vehicles are of crucial importance, which explains the large number of works [6]-[10].

It is known that detection methods [11] based on a motion approach are of less complexity and are therefore more suitable for their use in real time. Most of these methods [12]-[16] of detection is performed by subtraction into two successive images and binarization. It appears that this approach although fast, suffers from many artifacts.

Many works have been done to reduce the problem of artefacts. In [17] the authors propose a texture mixture of Gauss in order to distinguish the background of the image, facilitating the detection of moving vehicles. This approach is better adapted within the framework of a moving camera. The work that was done in [18] exploits the principle of Mahalanobis distance in order to compute the background image allowing by the end to select moving vehicles, however the computational cost make this approach less effective in real time issue.

Wang *et al.* [19] used the notions of extended optical flux and Eigen space to differentiate static parts from moving parts. This method has a significant time cost. Chen and Zhang [20] are designed an approach regroups of the following phases: the implementation of the difference between three successive images, followed by the use of artifacts filtering, then the application of a grouping operation of the pixels and finally the exploitation of mathematical morphology and connectivity analysis process.

The detection/tracking approaches that are based on motion study are very fast, so they are exploited in real time. However, these approaches have the disadvantage of many artifacts [21], [22]. Morphological transformation is done to deal with the problems of artifacts, but remains unsuccessful. The following Figure 1 illustrates the limits of the implementation of the thresholding phase, binarization and mathematical morphology. We observe in Figure 1(c) the subdivision of objects (see circle). In this article, the proposed method uses a motion based approach for detection. The set of artifacts generated have been reduced using a procedure of tracking 'labeling'.

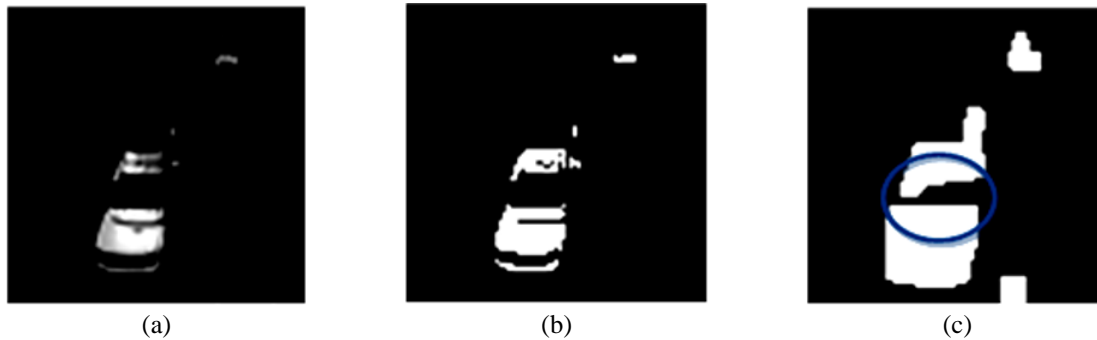


Figure 1. Image processing limits; (a) thresholding, (b) binarization, (c) mathematical morphology

2. PROPOSED METHOD

Our method proposed is subdivided into two levels: the low level (image processing) and the high level (information processing).

2.1. Image processing

The detection is achieved by the difference between two successive images, followed by a binarization. Artifacts can appear (small spots due to brightness variations). Treatment using a morphological transformation is performed. As shown in Figure 2 Each vehicle localized is registered in a Rect rectangle defined by:

x_h : is the highest pixel position of the vehicle.

x_l : is the lowest pixel position of the vehicle.

y_r : is the rightmost pixel position of the vehicle.

y_l : is the leftmost pixel position of the vehicle.

c_i, c_j : are the geometric center coordinates of the rectangle containing the vehicle.

A vehicle has been characterized by this rectangle Rect and the matrix Mg of gray level (gray level of the pixels contained in Rect). A vehicle detected at time $t+1$ is searched in the previous image t on a search area [23], [24] defined from the vehicle's geometry center of the image $t+1$.

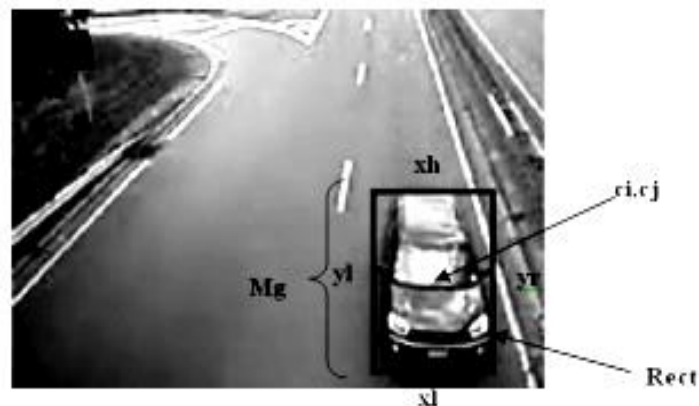


Figure 2. Parameters that represent a vehicle

Where more than one vehicle is in the search area, we must proceed with the likenhood procedure to ensure the correspondence.

Noting that in our work, the search area is defined in the images by the sizes: (-5, +5), (-10, +10) and (-20, +20) from top to bottom of the image. For this, we exploit the histograms in gray level [25] calculated from the matrix M_g of each vehicle. In Figure 3 an illustrative example is given. Assuming that we want to establish a correspondence between the vehicles of the moment $t + 1$ which is at the top of the image in Rect of Figure 3(b) and the vehicles located at time t .

After inserting the search area at time t see Figure 3(a) in discontinues lines, we found two vehicles represented by these geometric centers that are included in this area. The next phase is the modeling of the M_g gray level histograms for this vehicle and the two others belonging to the search area, these histograms are used to calculate the Vehicle Surface area.

$$Area_{Vehicle_i} = \sum Hist_{ng} V_i^2(k) \tag{1}$$

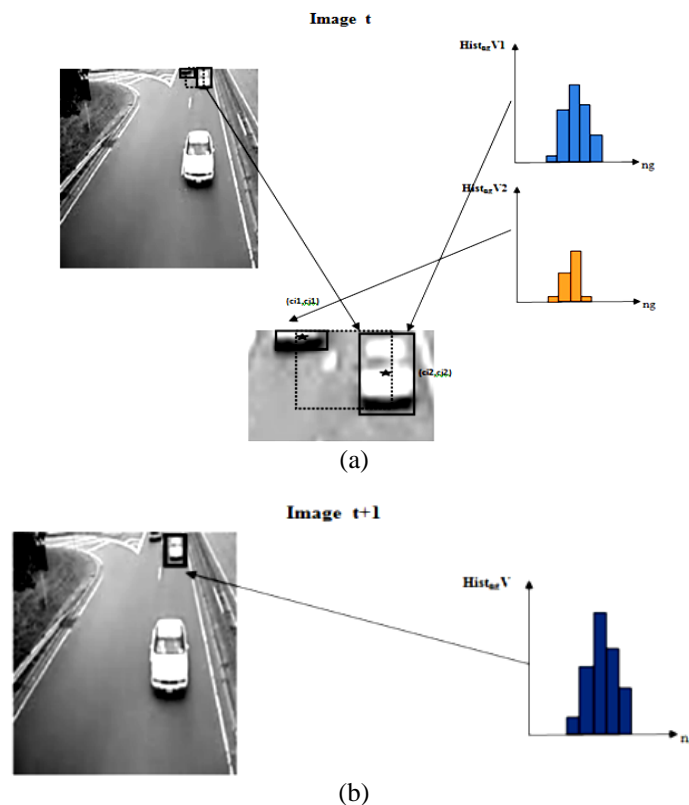


Figure 3. Illustrative example; (a) image of the moment t and the gray level histogram for the two localized vehicles, (b) image of the moment $t+1$ and the gray level histogram for the vehicle to follow

Finally, a decision is made, which is based on a comparison between the histogram surfaces of the two vehicles that are at time t and that of vehicle of time $t+1$. This decision was made by calculating the distances between these surfaces, by selecting what has the smallest distance according to the following rule:

$$Vehicle\ selected = \min\ distance\ [(Area\ VR,\ Area\ V1)\ and\ (Area\ VR,\ Area\ V2)] \tag{2}$$

- Area VR : area of a search vehicle.
- Area V1 : area of the first vehicle found.
- Area V2 : area of the second vehicle found.

2.2. Information processing “Labeling”

For each vehicle detected at time t , it is attributed a label defined by a doublet (a_t, c_t) and (a_{t+1}, c_{t+1}) at time $t+1$. The detection system scans the image from top to bottom and from left to right. In the image t ,

the first vehicle detected $a_t=1$, the second vehicle detected will have $a_t=2$. So the number a_t gives the order of appearance of the vehicle in the image t . The digit c_t indicates the correspondence, the vehicle of label (a_{t+1}, c_{t+1}) appeared at the a_{t+1} time in the image $t+1$ and corresponds to the vehicle which appeared at the c_{t+1} time in the previous picture t .

In Figure 4(a), we have two images t , which represent an empty road and a road has three vehicles carrying the labels (1,1), (2,2) and (3,3). While in Figure 4(b), we can see two images $t+1$, which represent a road to a vehicle labeled by (1,0) and another to three vehicles labeled by (1,1), (2, 2) and (3,3). In the images $t+1$, the label (1,0) indicates that there is a vehicle that has just appeared as a new object and that this vehicle does not appear in the image t , while the pairs (1,1), (2,2) and (3,3) mean that there are three vehicles in the image $t+1$ and that it is vehicles looks like the same vehicles labeled by (1,1), (2,2) and (3,3).



Figure 4. Principle of labels; (a) images t , (b) image $t+1$

2.3. Discussion and ascertainment

In the previous section, we discussed the basic principle of labeling. Now we want to develop this principle. A simple analysis shows that for a given image:

- Max (a_t) is the total number of vehicles detected at time t .
- If at time $t+1$, we observe that one of the vehicles carries a label $c_{t+1}=0$, then the number of vehicles at time $t+1$ becomes Max (a_t)+1.
- If max (a_{t+1})=max(a_t)-1, then there is a disappearance of a vehicle in the image $t+1$.

Since the low level of processing, which is particularly based on thresholding and mathematical morphology, generates a lot of errors, these causes: the duplication of objects, the bad appearance, and consequently reflects on the rules described above. For example, if the same vehicle of the instant t will be subdivided into two objects at the instant $t+1$, in this case we can arrive in situations where these two objects will generate a problem of increase of vehicles number at time $t+1$.

Another very important example is that one can have a situation where a vehicle already appeared, will be detected as a new object. In order to rise to these problems, a correction approach that is based on rules that can be summarized by the following points:

- The phenomenon of appearance is only for a new vehicle that is in the area of appearance (the area farthest from the camera).
- On the other hand, the phenomenon of disappearance is defined if a vehicle of moment t appears in the zone of disappearance (the zone closest to the camera) and that this object does not appear at time $t+1$.

In this article, we considered the simple case of a one-way road (from top to bottom or from right to left). We performed during an appearance of a vehicle the value 0 to c . The addition of these rules allows a clear improvement for our detection/tracking method. Our approach is to apply a correction procedure, it is given by the following algorithm.

Begin

- a. Read of image.
- b. An enlargement of the search area built around vehicles with labelling errors ($c_i=0$ and this vehicle does not belong to the area of appearance).
- c. Repeat the operation of the terrain geometry and the mapping.
- d. Assign labels for vehicles with labelling errors.
- e. Group objects that have the same label and (if they exist) in the same object.
- f. Test the labels of the vehicles again and redo the abovementioned spots for the case of incorrect "label error" detections.
- g. Stop as soon as you get a good detection.

End

3. EVALUATION OF OUR METHOD AND RESULTS

If we try to make an evaluation of our approach compared to the existing works, we can notice that this approach can be applied in real time compared to the different methods which are based on the artificial intelligence [26], [27]. In addition, our method is based on a principle of fusion between the aspect of image processing and information processing 'labels' which gives an advantage over conventional methods based on the aspect of image processing [28], [29].

In the Figure 5 that follows, four types of video sequences: two in road (the traffic flow from top to bottom of image), the third one inside a tunnel and the last one is in road (the traffic flow from right to left of image). We can observe a clear improvement of the detection/tracking procedure and this because of the implementation of the correction procedure. In these examples, we have tried to correct several errors, among these corrections the grouping of objects that carry the same labels and the deletion of objects that have labels of $c_i=0$, and which are outside the area of appearance. The modification of the objects labels which carry labels $c_i=0$ and which are outside the zone of appearance.



Figure 5. Detection/tracking results; (a) without the correction phase, (b) with the correction phase (*continue*)

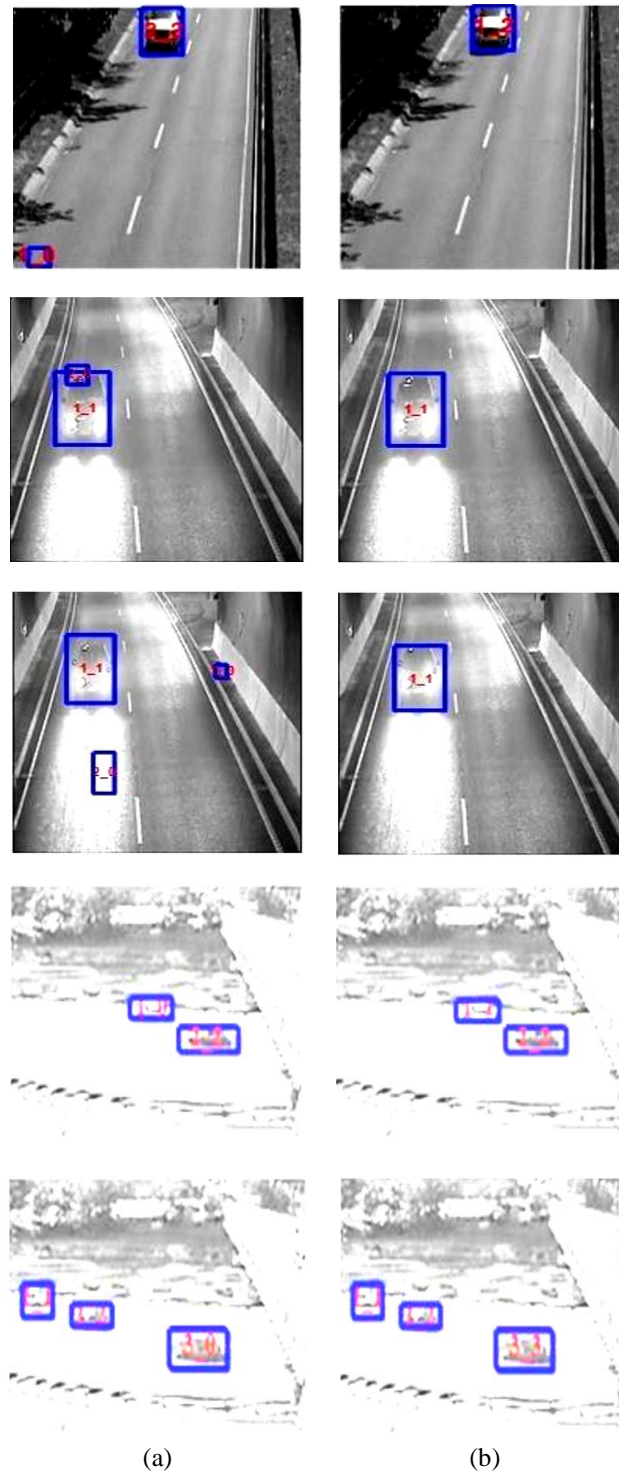


Figure 5. Detection/tracking results; (a) without the correction phase, (b) with the correction phase

4. CONCLUSION

In this article, a method based on two levels of processing, namely image processing and information processing, has been presented to detect and track vehicles in video sequences. Our contribution is to add a correction loop by exploiting the level of information processing which is based mainly on the principle of labeling, the latter showed its effectiveness in order to enhance the results of the vehicles detection especially in the case of the subdivision of objects, and in the presence of artifacts. The perspective work is concentrated around the traffic road modelization, using a transfert matrix and the labels management.

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