

Parking entrance control using license plate detection and recognition

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Article Info

Article history:

Received Dec 15, 2018

Revised Jan 21, 2019

Accepted Mar 1, 2019

Keywords:

Intelligent transportation systems

License plate detection

Recognition

Segmentation

Smart parking

ABSTRACT

There is no doubt that car parking is a very challenging and interesting topic of surveillance. In the recent years, a lot of smart systems for parking lot access control were developed to control and register the car data. The aim of this paper is to use image processing methods to control the entrance of a smart parking. The steps of car plate recognition are: preprocessing, License plate detection, character extraction and recognition. In the step of preprocessing, image was enhanced and noise was reduced. After preprocessing stage, color filter was used to detect the plate region. In case of large image size DWT was used for feature extraction and decreased the time of the detection stage. In the stage of character segmentation, the image is converted from grayscale to binary according to a given threshold. Filtering the binary image after using the morphological operation method, the largest objects are determined as the segmented plate characters. Finally, the correlation method was used to recognize the segmented characters. In case of similarity, SVM was used as a good classifier. Experimental results using matlab software, view that the proposed method increase the plate detection and recognition rates. It achieved average 97.8% detection rate, 98% segmentation rate and 97% recognition rate, so it will be a good method for smart parking entrance control.

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1. INTRODUCTION

The industrialization of the world, slow paced city development, and increase in number of cars has resulted in parking problems. There is a need for a robust system to be used for searching the free parking lots. Smart Parking System (SPS), is considered to be a small version of an Intelligent Transportation Systems (ITS) [1]. This SPS using Internet of Things (IoT), to reduce the parking congestion.

In [2] authors introduced a method for parking guiding system and space analyzing. IOT begin with things which have independent connected devices. These devices should be monitored, or under controlled using connected computers [3]-[4]. We have created a smart parking guidance algorithm in [5] depending on a cost function. In order to identify the cars, image processing techniques also was introduced. To detect the illegal (unreserved) cars, they used Optical character recognition (OCR) method. OCR is used to convert the row text and documents (like text books) into electronic readable les like (pdf). In [6] a hybrid feature extraction method was introduced for facial recognition and gave average recognition rate 100%.

An automatic car parking system using FPGA based on emergency conditions was proposed to detect the drivers condition and perform specific tasks such as warn the drivers [7]. We want to use image processing techniques in order to control the parking entrance. In order to solve the problem of parking

entrance management, a lot of research was made. Automatic License Plate Detection(ALPD) is to automatically detect the license region from a given image.

In [8] an ALPD method consist of five steps was introduced for handling different hazardous rates. The previous ALPD was developed and tested on database consists of 850 images in different controlled conditions like night, indoor, blurry, day, rainy, foggy and tilted LP. This method achieve global 94% detection rate, but it can be enhanced by using a lot of filtering methods.

In [9] a new method presented to apply the LPR systems for Iranian license plates. Increasing the accuracy of the character recognition phase rate and decreasing the training rate are the main advantages of the new hybrid model. The K-NN was implemented as the first classification method; the confusion problem related to similar characters in the license plates was overcome by using the multiple SVMs classification model. The SVMs has improved the performance of the K-NN in the recognition of similar characters. The SVMs was trained and tested only for the similar characters. The proposed method gave 97% character recognition rate.

2. RESEARCH METHODS

2.1. Morphological Operation

Morphological of image preprocessing is a sequence of non-linear operations depending on the morphology or shape of the features. Morphological operation does not depend on the numerical values, but depends on the related ordering of the pixel values and so are especially effective to the preprocessing of binary images. We want to thin connectivities between large white areas extended outer region of given images and remove small thin white objects. Due to the previous needs, morphological operation [10] (erosion, then dilation) is applied on a given binary image *f*. In the erosion step: erosion image *f* by structuring element *s* output a new image *g* having values one in all pixels (*x*; *y*) of a structuring element's when *s* fits *f*. In the dilation step: dilation of a binary image *f* by a structuring element *s* output a new image *g* having values one in all pixels (*x*; *y*) of a structuring element's when *s* hits *f*. Figure 1 show an example of hitting and fitting a binary image.

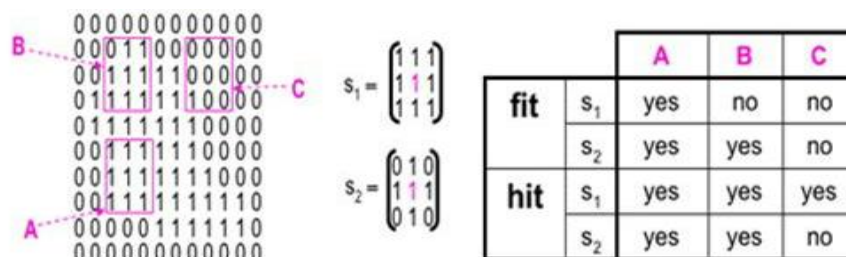


Figure 1. Example of hitting and fitting a binary image

2.2. Discrete Wavelet Transform

Wavelet transforms are mathematical functions used to convert data into a lot of frequency components and each component is studied with a resolution corresponding to its scale. Wavelets were presented independently in the field of mathematics [11], electrical engineering and quantum physics. In the last decades, many new wavelet applications were appeared like image compression, prediction of earthquakes, turbulence, radar and human vision. The mother wavelet decomposition function for an image is defined as:

$$\Psi_{u,v}(t) = \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} f(x,y) \exp^{-j\pi \frac{(ux+vy)}{N}} \tag{1}$$

Where,

$\exp^{-j\pi \frac{(ux+vy)}{N}}$ is a Kernel function, $f(x,y)$ is a 2D image, and N is count of pixels.

The Wavelet transform is a useful computational tool for applications of signal and image processing [12]. DWT is used in a wide range in pattern recognition area [13], [14].

Figure 2 view an example of DWT one level decomposition. LL1 describes approximate information, LH1 describes Horizontal information, HL1 describes vertical edge information, and finally the diagonal information is presented by HH1.

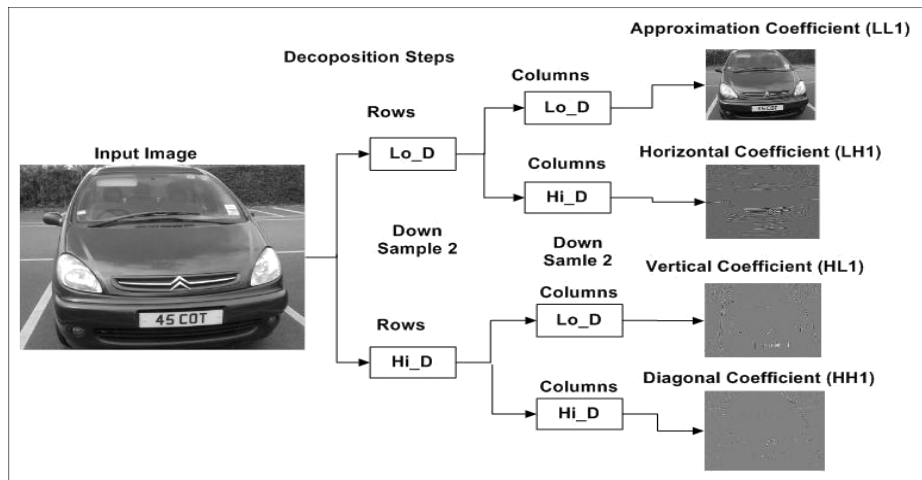


Figure 2. DWT coefficients

3. PROPOSED METHOD

This paper presents a hybrid framework for robust license plate detection and recognition system. Figure 3 view the general steps of ALPR system.

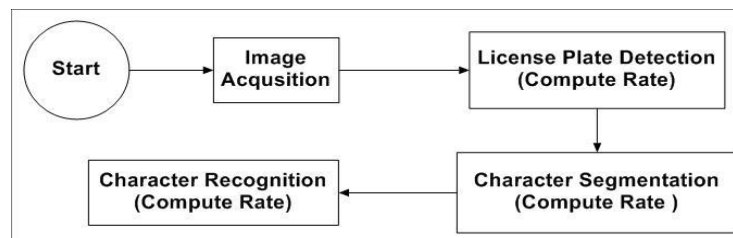


Figure 3. General steps of ALPR system

3.1. Image Acquisition

Figure 4 show how to get the car image. A digital camera is installed in the entrance of the parking, to collect the car image. Once the image is captured, we directly go to the next step (Plate Detection).



Figure 4. Image acquisition

3.2. Plate Detection

Plate area extraction or detection is the most important step because the license plate is in a small region and may be anywhere in the given image. It can be defined by its features. The proposed detection method depends on the License Plate color. The goal of the detection step is to mark an area with highest probability of having the plate. The detection method depends on two License Colors(White and Yellow backgrounds). The given RGB Image is converted to index image using `rgb2ind()` function, then use two color filter. White range is $(R > 0.5; G > 0.3; B > 0.81)$ and Yellow range is $(R > 0.3; G > 0.3; B < 0.31)$. Now we shall remove the noise objects using `bwareaopen()` function to remove objects smaller than 200. Using the morphological operation for both erosion and dilation. Remove objects smaller than 3500 pixel, then the area with the biggest size is determined as the plate area. Figure 5 view example of detected plates with two colors yellow and white backgrounds.

The detected area is converted from `rgb` to `gray` to be used in the stage of character segmentation due to the following (2).

$$Y = 0.2989 * R + 0.5870 * G + 0.1140 * B \tag{2}$$



Figure 5. Detected plates

In case of white or yellow foreground car, the areas with the smallest size are extracted. the detected areas are not considered to be the correct plate unless the segmentation stage is applied.

In this case, the segmentation stage is used to classify the detected areas. The area that contains characters is considered as the correct plate area. Figure 6 view example of detected plates with two colors yellow and white.

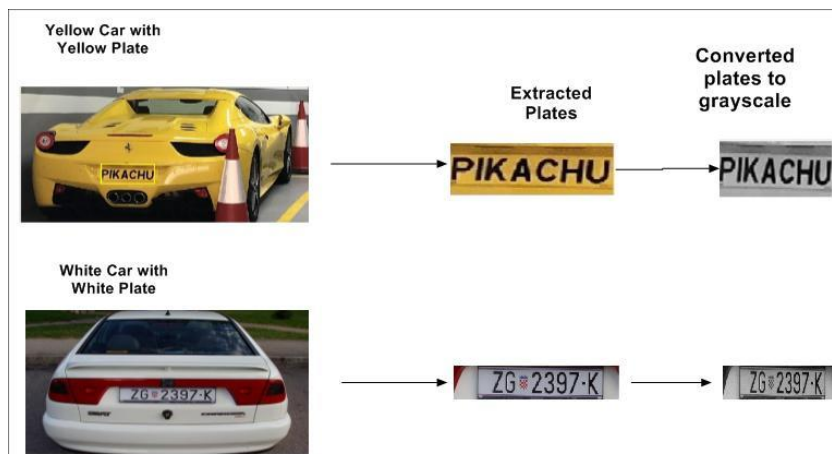


Figure 6. Detected plates in yellow and white foreground

3.3. Character Segmentation

In this stage we have the extracted gray plate and want to segment the characters. Figure 8 view the Steps of the Segmentation stage. The given plate image is converted from grayscale to BW (black and white), based on a threshold corresponding to the following.

$$Y_{ij} = \begin{cases} 1 & \text{if } x_{ij} > \theta, \\ 0 & \text{Othewise.} \end{cases}$$

Using two filters steps:

- Remove the smallest objects in the binary image, and leave the largest 6 objects.
- Remove objects which are not in the same vertical range.

Figure 7 view object to be removed, because it is not in the same vertical range. Finally, the segmented character is assigned to the final step (Character Recognition).



Figure 7. Removed objects

The following algorithm is a pseudo code defines how to remove objects, which are not characters.

Algorithm 1: Pseudo Code for Removing non Characters

Require: Objects, Obj_num, Obj_rowrange
Ensure: New Objects
 1: **for** i=1 : Obj_num **do**
 2: Get Objects i height and width
 3: **if** Objects i_width > Objects i_height **then**
 4: **if** Objects i_rowrange != Obj_rowrange **then**
 5: Remove Objects_i from NewObjects
 6: **endif**
 7: **endif**
 8: **endfor**

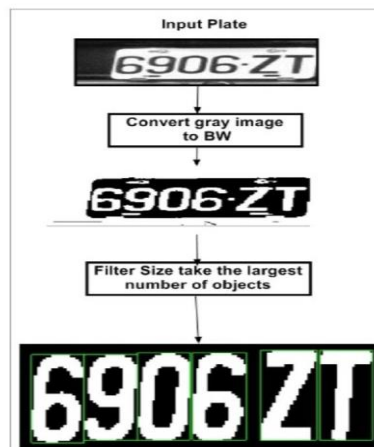


Figure 8. Segmentation process

3.4. Character Recognition

In this step, we have the segmented characters and want to recognize it. The following (3) was used for computing correlation rate between two images A and B.

$$r = \frac{\sum_i \sum_j (A_{ij} - \bar{A})(B_{ij} - \bar{B})}{\sqrt{(\sum_i \sum_j (A_{ij} - \bar{A})^2 + \sum_i \sum_j (B_{ij} - \bar{B})^2)}} \tag{3}$$

Where both A and B are the mean of A and B respectively. In case of similarity, which cannot be recognized using the correlation method. For example (D, O and 0), (B and 8) and (2 and z), in this case further training is needed. The way is to use Support vector machine (SVM), which is defined as a learning method used for data classification. SVM considered as a statistical classification algorithm and was proposed by Cortes and Vapnik [15]. SVM becomes popular because it gave high rate in handwritten digit classification. SVM was used for image classification in [16]. This algorithm search for the maximum separating hyperplane, which is defined by the hyperplane with the maximum distance between the training tuples.

4. RESULTS AND DISCUSSION

In this experiment, we test the accuracy and efficiency of the system. The white color plates images are taken from AOLP database [17] cited in paper [18]. The yellow color plates are collected from the internet. The car images database has different conditions, like brightness, different font size, shadow, low contrast, different car colors, different font type and different images size. Table 1 view the experimental results of the detection step. In case of white background the correct detection rate was 97.4%, and in case of Yellow plate the correct rate was 97.8%.

Table 1. Experimental Plate Detection Result

LP Type	Number of Images	Success Number	Success Rate	Average Image Size(Pixel)
White-Black	117	114	97.4	640*480
Yellow-Black	46	45	97.8	1024*768
Total	163	159	97.5

Table 2 show the comparison between the proposed detection method and other reported papers. The reported papers used the same database, and we add the database of yellow color plates. It is clear that the proposed method outperform the reported method in both correct detection rate and execution time per each image.

Table 2. Comparison with the Reported Articles

No	Reference	Correct Rate	Average Execution Time(Sec.)	System Specification
1	Azam et al [19]	86.15	0.45	C: Intel Core 2 Duo, 2.2 GHz S: 120X40 (non-tilted), 80X55 (tilted)
2	Ghahnavieh et al [20]	95.5	0.0968	C: Core 2 Quad, 2.67 GHz S: 640X480 to 2000X1500
3	Tadic et al [21]	94.4	-	-
4	Panahi et al [22]	96.3	0.18	C: Intel Core i5, 2.2 GHz S: 16X80 to 25X1508
5	Proposed Method	97.5	0.123	C: Intel Core i3, 2.4 GHz S: 640*480 to 1024*768

In case of large image size (for example: 2000 * 1950 pixel), DWT was used before detection stage. DWT generate 4 coefficients (approximation, horizontal, vertical, diagonal). The approximation coefficient was used as the test image, then the detection step was applied. Using DWT decrease the time of the detection stage from 0.4892 seconds to 0.167, approximately 67% decrease time rate.

Applying character segmentation stage on the detected plates, the experimental result gave 98% correct segmentation rate.

For recognition, pattern with maximum correlation is used as the recognized pattern. In case of similarity, the MultiSvmTrain() function was used to models a given training set with a corresponding group vector and classifies a given test set using MultiSvmClassify() function according to a one versus all relation.

Each character in training set has 3 images. The different Kernel functions test results showed that the best SVMs function for this study is the Gaussian Radial Basis Function (RBF). The RBF kernel is defined as:

$$k_{RBF}(x, x') = \exp[-\gamma \|x - x'\|^2] \quad (4)$$

where γ is a scaling parameter that sets the spread of the kernel and its default value is 1. The characters which were considered as similar characters (0, D, O), (B, 6, 8), (S, 5) and (2, Z), SVM was used to build the model for each group. If the correlation method gave any characters of them, the MultiSvmClassify() will be used to classify the given characters using its group model. SVM gave average 97% recognition rate, unlike the correlation method gave 94% correct recognition rate.

5. CONCLUSION

In this paper we present a hybrid method to improve on the current techniques of car plate detection and recognition. The proposed method detects the plate number in two background colors (White and Yellow). The detection rate was increased to 97.8%. In case of big image size the DWT was used in the preprocessing stage, which decreased the time consumed in the detection stage with 67%. In the segmentation stage, the plate was converted from grayscale to black and white. A defined filter was used to remove objects which are not characters. Objects are not in the same row range, or the width is larger than height. Finally the correlation method was used to recognize the segmented characters. If the character is defined to be (D, 0, O, B, 8, Z, 2 ...) which has similarity, in this case SVM was used for classification. SVM gave 97% recognition rate and correlation method gave 94% recognition rate, which means that using SVM is essentially for increasing recognition rate.

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