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Automated new license plate recognition in Egypt

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KEYWORDS

Egyptian license plate recognition; Plate region extraction; Segmentation; Template matching database communication **Abstract** License plate recognition (LPR) was an effective form of Automatic Vehicle Identification (AVI) systems. In this paper, a new and simple technique was presented for Egyptian vehicle's LPR system. The proposed technique consists of three major parts: Extraction of plate region, recognition of plate characters, and database communication. A video stream was one of the most important advantages of this system. The real-time was capability, and that it did not require any additional sensor input such as infrared sensors. This approach provided a good direction and performance for Automated New License Plate Recognition in Egypt.

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1. Introduction

Vehicles play very big role in transportation. In Egypt, vehicles have been increased because of population growth, cheaper, and installment of domestic car, more families have one or more cars. Therefore, control of vehicles is becoming a big problem and difficult to solve.

AVI systems are used for the purpose of effective control. LPR is a form of AVI. It is an image processing technology used to identify vehicles by only their license plates. Real time LPR plays a major role in automatic monitoring of traffic rules and maintaining law enforcement on public roads.

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Since every vehicle carries a unique license plate, no external cards, tags, or transmitters need to be recognizable, only license plate.

There are many systems of LPR, those systems can be classified as: template matching method [1,2] and learning based method [3–11]. Some of the previous algorithms have limiting conditions [12], such as fixed illumination [13], fixed type of license plate, stationary backgrounds [14], prescribed driveways [11,15], limited vehicle speeds [16,17]. This study improves many of these limitations.

2. Methodology

The proposed technique was designed for LPR system. Fig. 1 shows the algorithm of the proposed technique.

2.1. Extraction

2.1.1. Pre-image

There are three types of car detection in LPR system. The first method is sensor detection by using infrared sensors,

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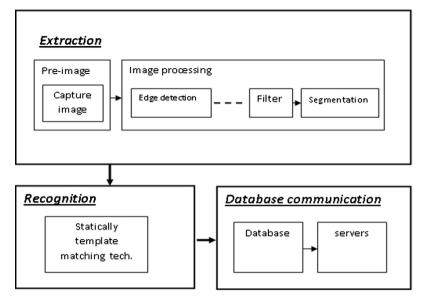


Figure 1 The proposed technique algorithm.



Figure 2 The capture of original image.



Figure 3 The gray scale image.



Figure 4 (a) Sobel edge detector, (b) dilation.



Figure 5 (a) Filled image, (b) erosion image.



Figure 6 (a) Smoothing image, (b) filtering image.

the second is image processing methods, and the third is loops techniques (magnetic loopsinductive loops).

The image processing method was used in this technique. The captured image of a car taken from 3 m by a camera. Fig. 2 shows the capture of original image.

2.1.2. Image processing

This stage was divided into two sub-stages.

• Detection stage:

The identification of objects within an image can be a very difficult task. In this stage, the capture image of the car converted to gray scale image. Fig. 3 shows the gray scale image of capture image.

The aim of this stage is to find rectangles of plate vehicles. The edge detection was applied by using Sobel edge detector. Morphological algorithm, dilation, and erosion [18–22] were used.

In dilation, every background pixel that is touching an object pixel is changed into an object pixel. Dilation makes the objects larger, and erosion is adversely of dilation. Dilation was used in vertical and horizontal lines to detect the structure of rectangle. Sobel edge detector and dilation are shown in Fig. 4.

The filling holes algorithm used to fill rectangles that result from dilation process. Smoothing the filled image result from filling holes algorithm by eroding it using erosion operation with square structure element to specify the candidate plate regions. However, there may be more than one candidate region for plate location. The filled image and erosion image are shown in Fig. 5.

Filtering and smoothing eroded image by using 2-D median filter with mask 5×5 then followed by removing unwanted objects which its counts may not form a candidate region for the plate. Fig. 6 illustrates the smoothing and filtering image.

To obtain the plate region in this phase, the other regions must be eliminated. Some criteria tests were applied to the final image, and these tests include:

1. Rectangle check

Checking that the candidate regions for plate had rectangle shape by compare white pixels count of these regions to their areas with ±5% tolerance
If count of white pixels = ±5% area of these region
This region may be a plate
Else
This region not a plate
2. Plate dimension check
The Emption plate had a fixed dimension with height equal

The Egyptian plate had a fixed dimension with height equal to 17 cm and width equal to 32 cm, so the ratio between heights to width approximately 1:2.

After rectangle check, the dimension check applied of the succeed regions of rectangle shapes, If 1.5 < width/height of the succeed region < 2.5

This region may be a plate • Else

This region not a plate

By applying the previous steps, the candidate plate regions and locations would be defined, cutting, and saving them from the gray scale image and original image to use later. Fig. 7 shows the Egyptian plate region.

• Segmentation:

Segmentation is one of the most important elements in automated analysis. At this step, the objects or other entities of interest are extracted from an image for recognition processing [23]. In the segmentation of plate characters, license plate is segmented into its constituent parts obtaining the characters individually.

A new Egyptian license plate has a fixed size $17 \text{ cm} \times 32 \text{ cm}$ as shown in Fig. 7. The plate region is divided into three parts, the first part is high part of plate region with 62 mm [24] that contains word of Egypt by Arabic and English, a background color of this region refers to type of car (private, taxi, etc.). The reminder region of the plate is divided vertically into two regions; right half contains plate characters, and left half contains numbers.

Each of Egyptian plate regions of gray scale image and original image was segmented into two parts with a ratio 1:2 from the height. Analyzing the first part of original image using color filter to obtain the type of car.

Then used the second part of the gray scale image. Filtering for enhancing the image and removing the noises and unwanted spots. At first outer boarder was removed then inner separator between letters and numbers. Then, median filter was applied to enhance the image, and the next step removed unwanted objects, which less than 200 pixels because there



Figure 7 The Egyptian plate.

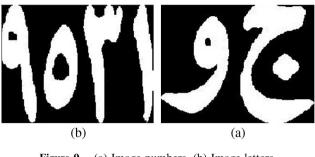


Figure 9 (a) Image numbers. (b) Image letters.

Figure 10 The individual characters and numbers.

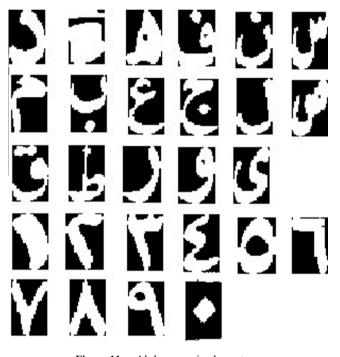


Figure 11 Alphanumeric characters.



Figure 8 (a) Image with outer border. (b) Image without border.

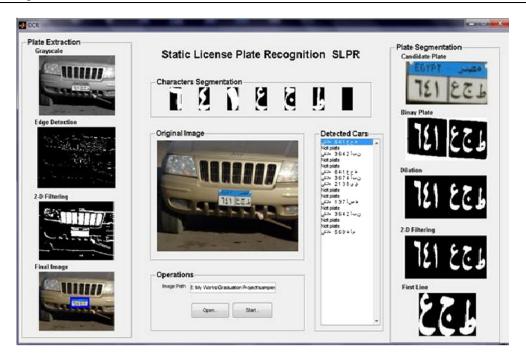


Figure 12 The characters recognition by GUI.

Table 1 Database of vehicles	refer to their colors.
Color	Refers to
Light blue	Private vehicles
Orange	Taxis
Red	Trucks
Gray	Buses
Beige	Limousines and tourists' buses
Green	Diplomatic vehicles
Yellow	Vehicles with unpaid customs
Dark blue	Police vehicles

were some letters which had a dot. Fig. 8 shows Image with and without border.

Then dilation operation was applied to the image for separating the characters from each other if the characters were close to each other. After this operation, separating the image into two parts with the same size, one part contained numbers and the other contained letters, this separation increase the performance of recognition. **Table 2**A symbol of traffic administration table.

الحرف لذكى	الحرف الأول	الار قام	احر ف	الدحاقظة
		3	.3	القاهرة
		4	3	حران
		+	3	الْجِيزَ د
		4	.3	6النوبر
	ين	4	3	الاسکترپه
	÷	4	3	الأشرقية
٤	مذ	+	3	برار سعيد
°,	di.	4	3	الدريس
من	-4 -	4	3	الاسماعيب

	City_Table			
	ID_City 👻	City_Name 🕞	IP_City -	Add New Field
	1	جيزه		
	2	المنيا		
	3	اسکندریه		
*				

Figure 13 Database of city car.

Separating image into number and letter parts is illustrated in Fig. 9.

After separation, horizontal projection was applied to find starting and end points of characters. Then, the individual characters and numbers cut from the plates are shown in Fig. 10.

2.2. Character recognition

After segmentation steps, the characters and numbers were cut into blocks with fixed size as shown in Fig. 10. These blocks were matching with previous database blocks of characters (27 alphanumeric characters – 17 alphabets and 10 numerical) with the size of 50×25 as shown in Fig. 11.

The statistical correlation method was used in matching technique. Since there were two signals as unknowns (input image and known database image) in this system, cross correlation was used.

Let $F_1(j,k)$ and $F_2(j,k)$ for $1 \le j \le J$ and $1 \le k \le K$

Represent two discrete images denoting the image to be searched and the template, respectively. The normalized cross correlation between the image pair is defined as: The graphical user interface (GUI) in MATLAB was used to build this technique. Fig. 12 shows the characters recognition by GUI.

2.3. Database communication

2.3.1. Database

Database was built by using Microsoft access database. It depended on the majority information of a car such that: type of a car, detect a car city, and faults cost.

2.3.1.1. Type of a car. Plate color which recognizes types of the cars. There are many color plates in Table 1.

2.3.1.2. Detect a car city. Traffic administration in Egypt designed rules to identify each car in Egypt. These rules depended on the number of digits of characters and numerical numbers to detect a car city. A symbol of a traffic administration table which used to detect the city (government) of the car in Table 2 and database table in Fig. 13.

$$R(m,n) = \frac{\sum_{j}\sum_{k}F_{1}(j,k)F_{2}(j-m+(M+1)/2,k-n+(N+1)/2)}{\left[\sum_{j}\sum_{k}|F_{1}(j,k)|^{2}\right]^{1/2}\left[\sum_{j}\sum_{k}|F_{2}(j-m+(M+1)/2,k-n+(N+1)/2)|^{2}\right]^{1/2}}$$

	The fa	ul ts	the cost					
	الاشارية	كىر	1 <i>5</i> 0p					
	بورد انشاه السور	ائىر غە ائە		15	qū			
	اكان من الاستمنيص	نقل اعدار کلت	1 <i>5</i> 0p					
			(a	ı)		·		
	Fault_Table							
	ID_Fault 🔻	Cost	٠	Type_Fault	٠	Add New Field		
	1	150		کس اشارہ				
	2	150		عدد رکاب اکثر				
	3							
*	(New)							

(b)

Fault_Car				
ID_Car 👻	ID_Fault 👻	Date_Fault 👻	ID_City	Ψ.
1	1	12/1/2010		4
2	1	5/7/2009		1
3	2	3/2/2010		3
1	2	12/5/2010		2
4	1	12/1/2010		1
3	1	12/6/2010		5

Figure 14 Faults toll and two database tables.

Table 3Zero normal form table.

Car_num	City	Color	Туре	Fault	Date fault	IP	Cost	City	Color	Туре	Fault	Date fault	IP city	Cost fault
						city	fault	Cairo	Blue	Private	Beak sign	12/1/2010		150p
و - ج 246	Cairo	Blue	Private	Beak sign	12/1/2010		150p							

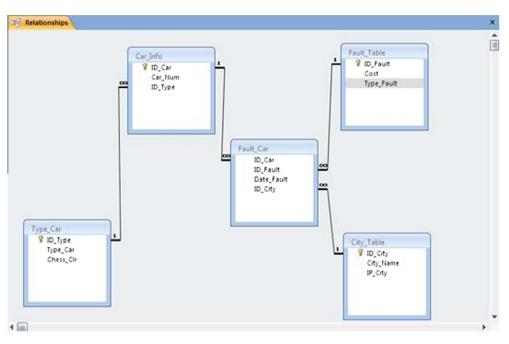


Figure 15 The relation among tables.

🛃 de						T	🚺 ITSDO) X
	I.T	S Data	base Trai	nsmitter					I.T.S D	atabase R	eceiver		
	الميارة	فطكيه	تاريخ المدلط	يع تعالق	مل _ن لنغاط		Wai	ting for N	ew Faluit	s			
1						î		السيارة	الطكيه	دا زيغ المدانلة	نوع المغزاناتة	مكاني المغالبة	1.
3	1												-
4						_							
5													
5						1	_						
8						_							=
0													
10	1					_	_						
- 18						-	-						
12													
13													
14							_						
													-
	Wisio	e Cone wen S	levare. Iran	iemilie New Faul	8e				3.0	rl D	0.1		

Figure 16 Database transmit and receive.

2.3.1.3. Collect information about the faults. There are a lot of faults for careless drivers which cause big problems. So, traffic administration put a toll to restrict of careless driver. Fig. 14 shows the faults toll and two database tables.

2.3.1.4. Assemble database. All the previous information collected in the zero normal form table which shown in Table 3.

The construction and relationships among database facilities tables are illustrated in Fig. 15.

2.3.2. Network and servers

The User Datagram Protocol (UDP) was used to send and receive data through wireless network among the servers of the cities. By using UDP, computer applications and datagram can send messages to other hosts on an Internet Protocol (IP). Fig. 16 shown the database transmitted and received by using UDP and GUI.

3. Experimental results

The proposed technique had been experimented to measure the performance and accuracy of the system. The system was tested by 100 patterns under several conditions. The characters and numbers were obvious and clear in new plate that played a good condition for identification. The identification was 91%. The failed identification came from mostly motion blurred or overlapped by other Vehicle's body or slant and dirty in the plates.

4. Conclusions

This system was designed by GUI and UDP under Matlab software for recognizing a new license plate in Egypt. A video stream at a real time was advantage of this technique. The performance and accuracy were excellent (91%). Database was built by traffic administration rules, so that the system could be applied as practical application.

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