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Review on Text Detection, Extraction and Recognition from Images and Videos

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ABSTRACT: Extraction and recognition of text present in images and videos have become a very popular research area in the last decade. Generally, text present in in image and video frames is of different size, orientation, style, etc. with complex backgrounds, noise, low resolution and contrast. These factors make the automatic text extraction and recognition in images and video frames a challenging task. Large number of techniques have been proposed by various researchers in the recent past to address the problem. This paper presents a review of various state-of-the-art techniques proposed towards different stages like detection, localization, extraction, etc., of text information processing in images and video frames. Looking at the growing popularity and the recent developments in the processing of text in images and video frames, this review imparts details of current trends and potential directions for further research activities to assist researchers.

KEYWORDS : Text Detection, Text extraction and Text Recognition.

I. INTRODUCTION

Text is born as an explicit carrier of high level semantics. This unique property makes text different from other generic visual cues, such as contour, color and texture. Therefore, detecting and recognizing texts in natural scenes have become important and vibrant research areas in computer vision. Text extraction is the task of automatically extracting structured information from unstructured and/or semi-structured machine-readable documents (text). The problem is challenging in nature due to variations in text properties and reflections. Text appearing in images is classified into three categories: document text, caption text, and scene text [6]. In contrast to caption text, scene text can have any orientation and may be distorted by the perspective projection therefore it is more difficult to detect scene text.

- **Document text**: A document image (Fig. 1) usually contains text and few graphic components. It is acquired by scanning journal, printed document, handwritten historical document, and book cover etc.
- **Caption text:** It is also known as overlay text or artificial text (Fig. 2). It is artificially superimposed on the imageat the time of editing, like subtitles and it usually describes the subject of the image content.
- Scene text: It occurs naturally as a part of the scene image and contain important semantic information such asadvertisements, names of streets, institutes, shops, road signs, traffic information, board signs, nameplates, food containers, street signs, bill boards, banners, and text on vehicle etc (Fig. 3).



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In Flanders Fields

In Handers fields the poppies blong



Fig. 1 Document Images (a) Gray-scale

Between the crosses, row on row, That mank our place; and in the sky the larks, still travely singing, thy iarce hand anis to guns before. We are the Dead, Short days ago We lived field dawn, sour surver ghore. Loved, and wore loved, and me we le In Randers fields. Jake up our guessel with the fore : Jo you from failing hairs we those

Jake up our quarrel with the fore : Jo you from failing hands we throw I he drock ; be going to have it high . Joycheck faith with up the die We chall we shap though poppies grow In Ramdero fields

John mc Gias

Punch Dec 8-1915

> (b) (b) Handwritten

procedure Nx_Delete.Brother (a (in out tree_Dr ; n: in integer) in -- saved trees: previous and following the child a_prev: tree_Dr; a_set: tree_Dr;

citize the security is a security is a security of the security is a raised into be provide the security is a raised into be provide security is a raised into be provide security is a raised security is raised security is raised.

a_sext:=a_prev.brother.brother; -- case where a_prev or a_prev.brother is not null a_prev.brother.father:=sull; a_prev.brother.brother.ereil;

m_prev.brother:=m end if;

-- if a_prev or a_prev.brother is null, we do noting when comptraint_error => null; and Nt_Deleta_Brother;

> (c) (c) Multi-color



Fig. 2 Caption text images

A. Properties of Text in Images:

Texts usually have different appearance due to changes in font, size, style, orientation, alignment, texture, color, contrast, and background. These changes will make the problem of automatic text extraction complicated and difficult. Text in images exhibit variations due to the difference in the following properties:

Size: The size of text may vary a lot.

□ Alignment: Scene text may be aligned in any direction and have geometric distortions while caption text usually aligned horizontally and sometimes may appear as non-planar text.

 \Box Color: The characters tend to have same or similar color but low contrast between text and background makes text extraction difficult.

 \Box Edge: Most caption and scene texts are designed to be easily read, hence resulting in strong edges at the boundaries of text and background.

□ **Compression:** Many images are recorded, transferred, and processed in compressed format. Thus, a faster text extraction system can be achieved if one can extract text without decompression.

Distortion: Due to changes in camera angles, some text may carry perspective distortions that affect extraction performance.



Fig. 3 Scene text images



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II. LITERATURE SURVEY ON TEXT EXTRACTION TECHNIQUES

In the past two decades, researchers have proposed numerous methods for detecting texts in natural images or videos. There are mainly three types of methods: Region based methods, texture based methods and morphological based methods. The various text extraction techniques are as follow:

A. Region based Method:

Region-based method uses the properties of the color or gray scale in the text region or their differences to the corresponding properties of the background. They are based on the fact that there is very little variation of color withintext and this color is sufficiently distinct from text's immediate background [20]. Text can be obtained by thresholding the image at intensity level in between the text color and that of its immediate background. This method is not robust to complex background. This method is further divided into two sub-approaches: connected component (CC) and edge based.

i.) Connected Component (CC) based Method:

CC-based methods use a bottom-up approach by grouping small components into successively larger components until all regions are identified in the image. A geometrical analysis is required to merge the text components using the spatial arrangement of those components so as to filter out non-text components and the boundaries of the text regions are marked. This method locate locates text quickly but fails for complex background.

ii.) Edge based Method:

Edges are a reliable feature of text regardless of color/intensity, layout, orientations, etc. Edge based method is focused on high contrast between the text and the background [5]. The three distinguishing characteristics of text embedded in images that can be used for detecting text are edge strength, density and the orientation variance. Edgebased text extraction algorithm is a general-purpose method, which can quickly and effectively localize and extract the text from both document and indoor/ outdoor images. This method is not robust for handling large size text.

B. Texture based Method

This method uses the fact that text in images have discrete textural properties that distinguish them from the background. The techniques based on Gabor filters, Wavelet, Fast fourier transform (FFT), spatial variance, SVM classifier and etc areused to detect the textual properties of the text region in the image [16]. This method is able to detect the text in the complex background. The only drawback of this method is large computational complexity in texture classification stage.

C. Morphological based Method:

Mathematical morphology is a topological and geometrical based method for image analysis. Morphological feature extraction techniques have been efficiently applied to character recognition and document analysis. It is used to extract important text contrast features from the processed images. These features are invariant against various geometrical image changes like translation, rotation, and scaling. Even after the lightning condition or text color is changed, the feature still can be maintained. This method works robustly under different image alterations.

Other approches:

Epshtein et al. [8] proposed SWT, an image operator that allows for direct extraction of character strokes from edge map. Binarization techniques, which use global, local, or adaptive thresholding, are the simplest methods for text localization. These methods are widely used for document image segmentation, as these images usually include black characters on a white background, thereby enabling successful segmentation based on thresholding. This approach has been adopted for many specific applications such as address location on postal mail, courtesy amount on checks, etc., due to its simplicity in implementation [11].



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III. METHODOLOGY OF TEXT EXTRACTION

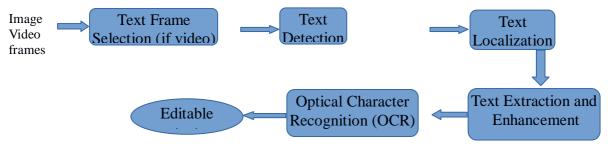


Fig. 4 Block Diagram of Text Extraction

The input image may be gray scale or color, compressed on uncompressed format. Text detection refers to the determination of the presence of text in the image while text localization is the process of determining the location of text and generating bounding boxes around it. After that, text is extracted i.e. segmented from the background. Enhancement of the extracted text is required as the text region usually has low-resolution and is prone to noise. Thereafter, the extracted text can be recognized using OCR. The block diagram of text extraction is shown in Fig. 4.

A . Applications of text extraction

- Text extraction can be used for:
- Data entry for business documents, e.g. check, passport, invoice, bank statement and receipt
- Automatic number plate rcognition
- Automatic insurance documents key information extraction
- Extracting business card information into a contact list
- More quickly make textual versions of printed documents, e.g. book scanning
- Make electronic images of printed documents searchable
- Converting handwriting in real time to control a computer
- Assistive technology for blind and visually impaired users

IV. DISCUSSION ABOUT PERFORMANCE EVALUATION

There are several difficulties related to performance evaluation in nearly all research areas in computer vision and pattern recognition (CVPR). The empirical evaluation of CVPR algorithms is a major endeavor as a means of measuring the ability of algorithms to meet a given set of requirements. Although various studies in CVPR have investigated the issue of objective performance evaluation, there has been very little focus on the problem of TIE in images and video. This section reviews the current evaluation methods used for TIE and highlights several issues in these evaluation methods.

The performance measure used for text detection, which is easier to define than for localization and extraction, is the detection rate, defined as the ratio between the number of detected text frames and all the given frames containing text. Measuring the performance of text extraction is extremely difficult and until now there has been no comparison of the different extraction methods. Instead, the performance is merely inferred from the OCR results, as the text extraction performance is closely related to the OCR output Performance evaluation of text extraction is not simple. Some of the issues related to the evaluation of text localization methods have been summarized by Antani et al. [2]

(i) Ground truth data : Unlike evaluating the automatic detection of other video events, such as video shot changes, vehicle detection, or face detection, the degree of preciseness of TIE is difficult to define. This problem is related to the



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construction of the ground truth data. The ground truth data for text localization is usually marked by bounded rectangles that include gaps between characters, words, and text lines. However, if an algorithm is very accurate and detects text at the character level, it will not include the above gaps and thus will not have a good recall rate [1].

(ii) **Performance measure :** After determining the ground truth data, a decision has to be made on which measures to use in the matching process between localized results and ground truth data. Normally, the **recall and precision rates** are used. Additionally, a method is also needed for comparing the ground truth data and the algorithm output: pixel-by-pixel, character-by-character, or rectangle-by-rectangle comparison.

(iii) Application dependence : The aim of each text localization system can differ. Some applications require that all the text in the input image must be located, while others only focus on extracting important text. In addition, the performance also depends on the weights assigned to false alarm or false dismissal.

(iv) Public database : Although many researchers seek to compare their methods with others, there are no domainspecific or general comprehensive databases of images or videos containing text. Therefore, researchers use their own databases for evaluating the performance of the algorithms. Further, since many algorithms include specific assumptions and are usually optimized on a particular database, it is hard to conduct a comprehensive objective comparison.

Table 1 shows performance of various approaches in text detection and extraction.

Author, year	Technique Used	Images	Parameters	Remarks
Yao et al.[17],	CC and Support Vector	Complex	PR=64%	Pixels of each character
2007	Machine (SVM)	background images	RR=60%	assumed to have similar
				color.
Lai et al. [13],	Edge detection and K-means	Signboard Images		Efficient for uneven
2008	clustering			illumination.
Zhang et al.	Discrete Wavelet Transform	Background	DR= 94.5%,	Text character Color
[20], 2008	(DWT), k-means clustering,	images	FAR= 13.6%	independent.
	morphology Operations	wit		
		h different		
		languages,		
		fon		
		ts and sizes		
Song et al. [16],	Histogram Projection and	Chinese text	PR=77.05%	K=3 gives best
2008	color based K-means		RR=75.63%	performance.
	clustering			1
	_			
Dinh et al. [7],	Edge detection and Histogram	Signboard Texts		Low complexity
2008	Projection			algorithm.
Fan et al.[9],	Stroke features and connected	Caption text	PR=95.2%	Color information is not
2009	component	images	RR= 94.5%	fully used.
		Democratic		La la man dan t
Audithan et	Haar DWT, Morphological	Document images	DR =94.8 %	Independent of
al.[5], 2009	Dilation operator, logical			contrast.
	AND operator, Dynamic			
	thresholding			

Table 1 Various Text Extraction Techniques



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Anoualet al.[3], 2010	Edge detection, texture features, connected component analysis	Complex background images	PR=95% RR=89%	Robust and effective.
Kumar et al [12], 2010	CC Analysis	ICDAR 2003 scene images	PR=90% RR=89%	Capable of Multilingual Text extraction.
Hassanzadeh et al.[10], 2011	Morphological operator, Decision classifier	Logo detection in document images	PR=95.6% Accuracy=86.9 %	A novel and fast method for logo detection.
Zaravi et al. [18], 2011	DWT, Dynamic thresholding, Region of Interest (ROI)	Colored book and journal cover sheets	DR=91.20%	Robust to noise.
Zhang et al. [19], 2012	Edge Enhancement and CC	Web images and caption text images	DR=92.4%	Not sensitive to various types of background noises.
Seeri et al. [15], 2012	Median filter, Sobel edge detectoreconnected component labeling, order static filter.	Kannada text images	PR=84.21% RR=83.16% Accuracy = 75.77%	Fails to extract very small characters.
Azadboni et al. [6], 2012	FFT Domain Filtering , SVM Classification, K- means clustering	Scene text images	DR= 98.10%	Text characters having uniform colour.
Anupama et al.[4], 2013	Morphology operators, Histogram Projection (X and Y histogram)	Handwritten Telugu document images.	DR=98.54%, Accuracy =98.29%	Fail in case of touching characters and over- lapping lines.
Raj et al. [14], 2014	CC based	Natural Scene Images (Devanagari text)	PR= 72.8%, RR=74.2 %	Fails for small slanted/curved text.

V.CONCLUSION

This paper provided a comprehensive survey of text information extraction in images and video. As a result a textimage-analysis is needed to enable a text information extraction system to be used for any type of image, including scanned document images, real scene images through a video camera, caption text images and videoimages. The above study shows that, most of the methods fall into any one of the above techniqus and also that there is a limitation in each technique to give a better detection rate with fewer false alarms without any constraints for text region extraction in different type of images. But still, to need a completely robust and generalized technique for text segmentation, it is difficult to provide appropriate input to the optical character recognition (OCR) system. So, a combined method has been proposed for automatic text content extraction from different images and video which is independent of the various characteristics of text.



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