

Text string detection from natural scenes english language essay

[Linguistics](#), [English](#)



Mr. Rampurkar Vyankatesh Vijaykumar., VPCOE Baramati. Mrs. Gyankamal J. Chhajed. Asst. Prof., VPCOE Baramati.

Abstract

In this paper we have presented different methods to find strings of characters from natural scene images. We have presented different techniques like extraction of character string regions from scenery images based on siblings and thickness of characters, efficient binarization and enhancement technique followed by a suitable connected component analysis procedure, text string detection from natural scenes by structure-based partition and grouping, and a robust algorithm for text detection in images. It is assumed that characters have closed contours, and a character string consists of characters which lie on a straight line in most cases. Therefore, by extracting closed contours and searching neighbors of them, character string regions can be extracted; Image binarization successfully processes natural scene images having shadows, non-uniform illumination, low contrast and large signal-dependent noise. Connected component analysis is used to define the final binary images that mainly consist of

text regions. One technique chooses the candidate text characters

from connected components by gradient feature and color feature. Color-based partition performs better than gradientbased

partition, but it takes more time to detect text strings on each color layer. The text line grouping is able to extract text strings with arbitrary orientations. The combination of color-based partition and adjacent character grouping (CA) gives the best performance.

Key terms

Connected Components Analysis, Adjacent Character grouping, Image partition, Text line grouping, Text string detection, Color reduction technique, Edge detection.

1. Introduction

Text information in natural scene images serves as important clues for many image-based applications such as scene understanding, content-based image retrieval, assistive navigation, and automatic geocoding. However, locating text from a complex background with multiple colors is a challenging task. A lot of objects on which characters are written exist in our living environment. We humans get much information from these texts. It is expected that robots act in our living environment and support us in the future. If robots can read text on objects such as packages and signs, robots can get information from

them, and they can use it in their activation and support for us. Owing to the progress of OCR, computers have been able to read text in images. However, images have many non-character textures, and they make it difficult to read text by OCR. To cope with that problem, we need to extract character string regions from images. Indexing images or videos requires information about their content. This content is often strongly related to the textual information appearing in them, which can be divided into two groups: Text appearing accidentally in an image that usually does not represent anything important related to the content of the image. Such texts are referred to as scene text. Text produced separately from the image is in general a very good key to understand the image which is called artificial text. In contrast to scene text, artificial text is not only an important source of information but also a significant entity for indexing and retrieval purposes. Natural scene images contain text information which is often required to be automatically recognized and processed. Localization of text and simplification of the background in images is the main objective of automatic text detection approaches. However, text localization in complex images is an intricate process due to the often bad quality of images, different backgrounds or different fonts, colors, sizes of texts appearing in them. In order to be successfully recognizable by an OCR system, an image having text must fulfill certain requirements, like a monochrome text and background where the background-to-text contrast should be high. This paper strives toward methodologies that aid automatic detection, segmentation and recognition of visual text entities in complex natural scene images. The algorithms of text extraction from images can be broadly classified under

three types. They are gradient feature based, color segmentation based, and texture analysis based. The gradient feature based algorithm is based on the idea that pixels which have high gradient are the candidates of characters since edges exist between a character and background. In this paper, Chucai Yi and YingLi Tian propose a new framework to extract text strings with multiple sizes and colors, and arbitrary orientations from scene images with a complex and cluttered background [3]. The proposed framework consists of two main steps: a) image partition to find text character candidates based on gradient feature and color uniformity. In this step, Chucai Yi and YingLi Tian propose two methods to partition scene images into binary maps of non overlapped connected components: gradient-based method and color-based method b) Character candidate grouping to detect text strings based on joint structural features of text characters in each text string such as character sizes, distances between two neighboring characters, and character alignment. In this step, Chucai Yi and YingLi Tian propose two methods of structural analysis of text strings: adjacent character grouping method and text line grouping method. Figure 1: Examples of text in natural scene images [3]

2. Related Work

One of the characteristics of common characters in real images is that most of them are capable of producing closed contour when edge extraction process is applied. So Tomohiro Nishino[1] takes an approach to detect closed contours from images. Moreover, it is assumed that a character string consists of characters which lie on a straight line. From these

assumptions, characterstring should be found from regions where closed contoursare arranged with regularly. Assuming that charactersincluded in a character string are aligned horizontally, string regions can be extracted by detecting horizontallyaligned closed contours. Tomohiro Nishino explainshow to detect the horizontally aligned closed contour. First, a circumscribed rectangle of a closed contouris calculated. Next, the rectangle is slid to right by somepixels as much as the width of the rectangle. If the rectangleincludes the center of a circumscribed rectangle ofanother closed contour, these two closed contours are assumedto be aligned horizontally and to be included inthe same character string. Closed contours which are isolatedare assumed not to be characters. Circumscribedrectangles of each character string are assumed to bestring regions. By this process, string regions of horizontallyaligned closed contours are extracted. Both closedand unclosed contours which lie left or right of characterstring regions are extracted, and each thickness iscalculated. These contours are added to the characterstring region if they have the similar thickness to that ofcharacters in that region. Basilios Gatos[2] produces gray level image and invertedgray level image. Then, calculate the two correspondingbinary images using an adaptive binarizationand image enhancement technique. In the sequel, theproposed technique involves a decision function that indicateswhich image between binary images contains textinformation. An efficient algorithm which can automatically detect, localize and extract horizontally aligned text in images(and digital videos) with complex backgrounds is presentedby by Julinda Gllavata, Ralph Ewerth and BerndFreisleben[4]. The proposed approach is based on theapplication of a color reduction technique,

a method foredge detection, and the localization of text regions usingprojection profile analyses and geometrical properties. 2Figure 2: The flowchart of the framework [3]

3. Programmer's design

Fig. 2 depicts the flowchart of the framework. Theproposed framework consists of two main steps, givenhere. Step 1) Image partition to find text character candidatesbased on gradient feature and color uniformity. In this step, Chucai Yi and YingLi Tian propose twomethods to partition scene images into binary maps ofnonoverlapped connected components: gradient-basedmethod and color-based method. A post processing isthen performed to remove the connected componentswhich are not text characters by size, aspect ratio, andthe number of inner holes. Step 2) Character candidate grouping to detect textstrings based on joint structural features of text charactersin each text string such as character sizes, distancesbetween two neighboring characters, and character alignment. In this step, Chucai Yi and YingLi Tian propoasetwo methods of structural analysis of text strings: adjacentcharacter grouping method and text line groupingmethod.

4. Image Partition

To extract text information from a complex background, image partition is first performed to group togetherpixels that belong to the same text character, obtaininga binary map of candidate character components. Based on local gradient features and uniform colors oftext characters, we design a gradient-based partition algorithmand a color-based partition algorithm, respectively. Figure 3: We compare results of four morphological

operators with result of our gradient-based partition.[3] Figure 4: Some examples of color-based partition, where the left column contains original images and other columns contain the corresponding dominant color layers. [3]

5. Connected Components Grouping

The image partition creates a set of connected components S from an input image, including both text characters and unwanted noises. Observing that text information appears as one or more text strings in most natural scene images, we perform heuristic grouping and structural analysis of text strings to distinguish connected components representing text characters from those representing noises. Assuming that a text string has at least three characters in alignment, we develop two methods to locate regions containing text strings: adjacent character grouping and text line grouping, respectively. In both algorithms, a connected component C is described by four metrics: height, width, area, and centroid. In addition, we use D to represent the distance between the centroids of two neighboring characters.

6. Adjacent Character Grouping

Text strings in natural scene images usually appear in alignment, each text character in a text string must possess character siblings at adjacent positions. The structure features among sibling characters can be used to determine whether the connected components belong to text characters or unexpected noises. Here, five constraints are defined to decide whether two connected components are siblings of each other. 1) Considering the capital and lowercase characters, the height ratio falls between $1/T_1$ and T_1 . 2) The

distance between two connected components should not be greater than T_2 times the width of the wider one. 3) For text strings aligned approximately horizontally, the difference between Y-coordinates of the connected component centroids should not be greater than T_3 times the height of the higher one. 4) Two adjacent characters usually appear in the same font size, thus their area ratio should be greater than $1/T_4$ and less than T_4 . 5) If the connected components are obtained from gradient based partition, the color difference between them should be lower than a predefined threshold T_5 . In their system Chucai Yi and YingLi Tian set $T_1 = 2$, $T_2 = 3$, $T_3 = 0.5$, $T_4 = 406$.

1. Mathematical Model

For two connected components C and C' they can be grouped together as sibling components if the above five constraints are satisfied. When C and C' are grouped together, their sibling sets will be updated according to their relative locations. That is, when C is located on the left of C' , C' will be added into the right-sibling set of C , which is simultaneously added into the left-sibling set of C' . The reverse operation will be applied when C is located on the right of C' . To create sibling groups corresponding to complete text strings, Chucai Yi and YingLi Tian merge together any two sibling groups $SG(C_1)$ and $SG(C_2)$ when their intersection contains no less than two connected components. At this point, each sibling group can be considered as a fragment of a text string. Repeat the merge process until no sibling groups can be merged together.

Text string in scene images can be described by corresponding adjacent character groups. To extract a region containing a text string, Chucai Yi and YingLi Tian calculate rectangle covering all of the connected components in the corresponding adjacent character group.

Advantages:-

1) The structure features among sibling characters can be used to determine whether the connected components belong to text characters or unexpected noises. 2) Character grouping is performed to combine the candidate text characters into text strings which contain at least three character members in alignment. 4 Figure 5: Two detected adjacent character groups marked in red and green [3]

7. Text Line Grouping

In order to locate text strings with arbitrary orientations, Chucai Yi and YingLi Tian develop text line grouping method. To group together the connected components which correspond to text characters in the same string which is probably non horizontal, Chucai Yi and YingLi Tian use centroid as the descriptor of each connected component. In order to locate text strings with arbitrary orientations, Chucai Yi and YingLi Tian develop text line grouping method. To group together the connected components which correspond to text characters in the same string which is probably non horizontal, Chucai Yi and YingLi Tian use centroid as the descriptor of each connected component.

7. 1. Mathematical Model Given a set of connected component centroids, groups of collinear character centroids are computed, as shown below. $M = \{m \mid C \in S \ \& \ m = \text{centroid}(C)\}$ $L = \{G \mid G \text{ is subset of } M, |G| \geq 3, \text{ they are character centroids \& are collinear}\}$ Where M denotes the set of centroids of all of the connected components obtained from image partition, and L denotes the set of text lines which are composed of text character centroids in alignment. Chucai Yi and YingLi Tian design an efficient algorithm to extract

regions containing text strings. At first, they remove the centroids from the set M if areas of their corresponding connected components are smaller than the predefined threshold T_s . Then, three points m_i, m_j, m_k are randomly selected from the set to form two line segments. They calculate the length difference, and incline angle difference between line segments $m_i m_j$ and $m_j m_k$ as shown $d = D(m_i, m_j) / D(m_j, m_k) = \frac{|x_i - x_j|}{|x_j - x_k|}$, if $|x_i - x_j| < \frac{\sqrt{2}}{2} |x_j - x_k|$, if $|x_i - x_j| > \frac{\sqrt{2}}{2} |x_j - x_k|$. The three centroids are approximately collinear if $1/T_6 < d < T_6$ and $1/T_7 < \theta < T_7$. Here $T_6 = 2, T_7 = \sqrt{12}$. Thus, they compose a preliminary fitted line $l_u = m_i, m_j, m_k, u = \text{index of fitted line}$. Other collinear centroids along l_u can be added into the end position to form a complete text string increasingly. For now, each text string is described by a fitted line. The location and size of the region containing a text string is defined by the connected components whose centroids are cascaded in the corresponding fitted line.

Advantages:-

1) In order to locate text strings with arbitrary orientations, Chucai Yi and YingLi Tian develop text line grouping method. 2) To group together the connected components which correspond to text characters in the same string which is probably non horizontal, they use centroid as the descriptor of each connected component.

8. Image binarization and connected component analysis based method

Proposed methodology is presented in Figure 7. Starting from the scene image, Basilios Gatos produces Figure 6: Resulting fitted lines from

centroids cascading. Red line corresponds to text region while cyan lines are false positives to be removed [3] gray level image and inverted gray level image. Then, calculate the two corresponding binary images using an adaptive binarization and image enhancement technique. In the sequel, the proposed technique involves a decision function that indicates which image between binary images contains text information. In first fig. the original binary image is selected while in second fig. the inverted binary image is selected. Finally, a procedure that detects connected components of text areas is applied. 8. 1. Data independence and Data Flow architecture Data Flow architecture is represented in figure 8. 8. 2. Turing Machine state transition diagram is represented in figure 9.

9. Conclusion

Due to the unpredictable text appearances and complex backgrounds, text detection in natural scene images is still an unsolved problem. In this paper we have presented methods to find strings of characters from natural scene images. We have presented techniques like extraction of character string regions from scenery images based on siblings and thickness of characters, efficient binarization and enhancement technique followed by a suitable connected component analysis procedure, text string detection from natural scenes by structure based partition and grouping, and a robust algorithm. Figure 7: a) Flowchart of the proposed method for text detection in natural scene images (original binary image is selected). b) Flowchart of the proposed method for text detection in natural scene images (inverted binary image is selected). [2] Figure 8: Data Flow architecture for text detection in

images. Also, we have presented an approach to detect, localize, and extract texts appearing in grayscale or color images as well as locate text strings with arbitrary orientations. ⁶Figure 9: state transition diagram Our future work will focus on developing learning-based methods for text extraction from complex backgrounds and text normalization for OCR recognition.