Segmentation of handwritten Gurmukhi text into lines

Ashu Kumar
Student, M. Tech., Yadwindra College of Engineering, Talwandi Sabo, Punjab, India

Simpel Rani Jindal
Assistant Professor, Yadwindra College of Engineering, Talwandi Sabo, Punjab, India

ABSTRACT
Text line segmentation is an essential pre-processing stage for handwriting recognition in many Optical Character Recognition (OCR) systems. It is an important step because inaccurately segmented text lines will cause errors in the recognition stage. Text line segmentation of the handwritten documents is still one of the most complicated problems in developing a reliable OCR. The nature of handwriting makes the process of text line segmentation very challenging. Text characteristics 3can vary in font, size, shape, style, orientation, alignment, texture, color, contrast and background information. These variations turn the process of word detection complex and difficult [2]. In the case of handwritten documents, differently from machine printed, the complexity of the problem even increases. Since handwritten text can vary greatly depending on the user skill, disposition and even cultural background. A new technique to segment a handwritten document into distinct lines of text is presented. The proposed method is robust to handle line fluctuation.

Keywords
OCR, Line Segmentation, Piece-wise separating lines

1. INTRODUCTION
A lot of research work has been investigated for character recognition of Gurmukhi script. For an optical character recognition (OCR) system, segmentation phase is an important phase and accuracy of any OCR heavily depends upon segmentation phase. Incorrect segmentation leads to incorrect recognition. Segmentation phase include line, word and character segmentation. Before word and character segmentation, line segmentation is performed to find the number of lines and boundaries of each line in any input document image. Incorrect line segmentation may result in decrease in recognition accuracy.

For segmentation of lines from handwritten text, survey papers are available [1,2]. Considerable amount of work has been carried out to segment lines of handwritten Roman script and there are varied and some well developed techniques [3-7]. But very little work has been carried out for Indic scripts like Devnagri, Bengali, Gurmukhi etc. Only a few papers are available for segmentation of handwritten Indic scripts [8-11].

The simplest and most widely used method to segment the lines is to use the inter-line gap in horizontal projection as line boundaries. This method does not work well on skewed, fluctuating or proximate images. Here, we are modifying the method to segment text lines based on histogram projection. Figures 1and 2 shows three kinds of sample documents on which the line segmentation is performed. The rest of the paper is organized as follows.

Section 2 describes problems associated with line segmentation. Section 3 describes the method to be proposed. Experiments and results are discussed in section 4 which is followed by conclusion in section 5.

2. SEGMENTATION CHALLENGES
When dealing with handwritten text, line segmentation has to solve some obstacles that are uncommon in modern printed text. Among the most predominant are:

- Skewed lines: lines of text in general are not straight.

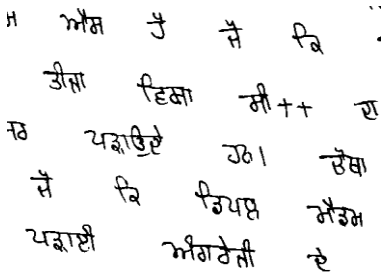

Fig 1: Skewed lines

- Fluctuating lines:

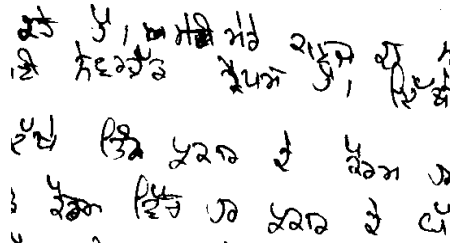

Fig 2: Fluctuating lines

- Line proximity: Small gaps between neighbouring text lines will cause touching and overlapping of

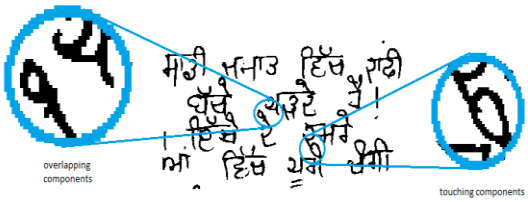

Fig 3: Line proximity

components, usually words or letters, between lines and irregularity in geometrical properties of the line, such as line width, height, distance in between words and lines, leftmost position etc.

3. PROPOSED METHOD

There exist several methods for text line segmentation which are roughly categorized as follows. Smearing methods [1,2]: short white runs are filled with black pixels intending to form large bodies of black pixels, which will be considered as text line areas. Smearing methods can't deal well with touching and overlapping components. Horizontal projections [1,2]: a vector containing the sums of each image line is created. The local minima of that vector are assumed to be the projection of white areas in between lines, and the image is segmented accordingly. Horizontal projections can't deal well with skewed, curved and fluctuating lines. Hough transform [1,2] considers any image to compose of straight lines. It creates an angle, offset plane in which the local maxima are assumed to correlate with text lines. Hough transform has trouble detecting curved text lines. Other methods have also been proposed such as: repulsive attractive networks, stochastic methods and text line structure enhancing [1,2]. Due to many challenges in text line segmentation, although many methods have been proposed, the problem still remains open.

The method of horizontal projection of the whole text is suitable for segmentation of the text with straight lines and with large gap in lines. This method cannot segment handwritten document because it contains touching lines, overlapping lines or fluctuating lines. For example, see the Fig 1,2 & 3.

So to segment this type of text, Here, we are modifying the method to segment text lines based on histogram projection and this technique is called piece-wise projection.

At first, we divide the text into vertical chunks of width W. Width of the last chunk may differ from W. Computation of W is discussed later. Next, we compute piece-wise separating lines (PSL) from each of these chunks [12]. We compute the horizontal projection of each chunk. The projection profiles of the chunks of the image are shown in Fig 4.

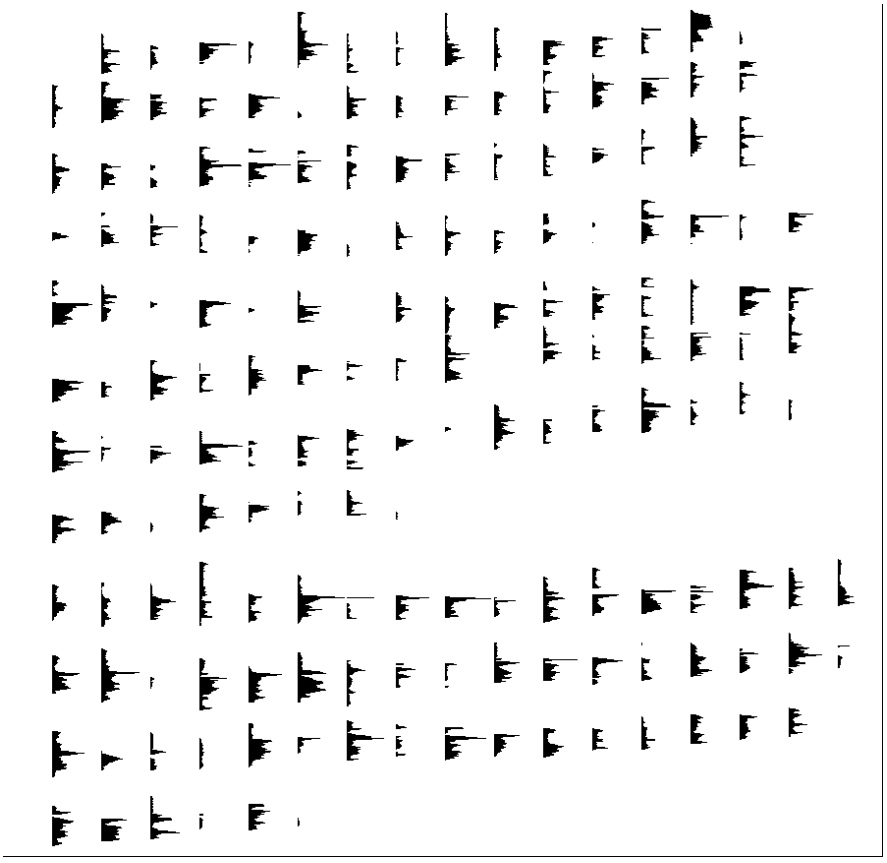

Fig 4: Projection profile of each chunk

The row where this HP is zero is a PSL. We may get a few consecutive rows whose HP is zero. Then the first row of such consecutive rows is the PSL. The PSLs of different chunks of a text are shown in Fig. 5 by horizontal lines.

All these PSLs may not be useful for line segmentation. We choose some potential PSLs as follows. For this, we compute the estimate height of word. If the distance between any two consecutive PSLs of a stripe is less than word height, we remove the upper PSL of these two PSLs. PSLs obtained after this removal are the *potential PSLs*. The potential PSLs obtained from the PSLs of Fig 5 are shown in Fig 6.

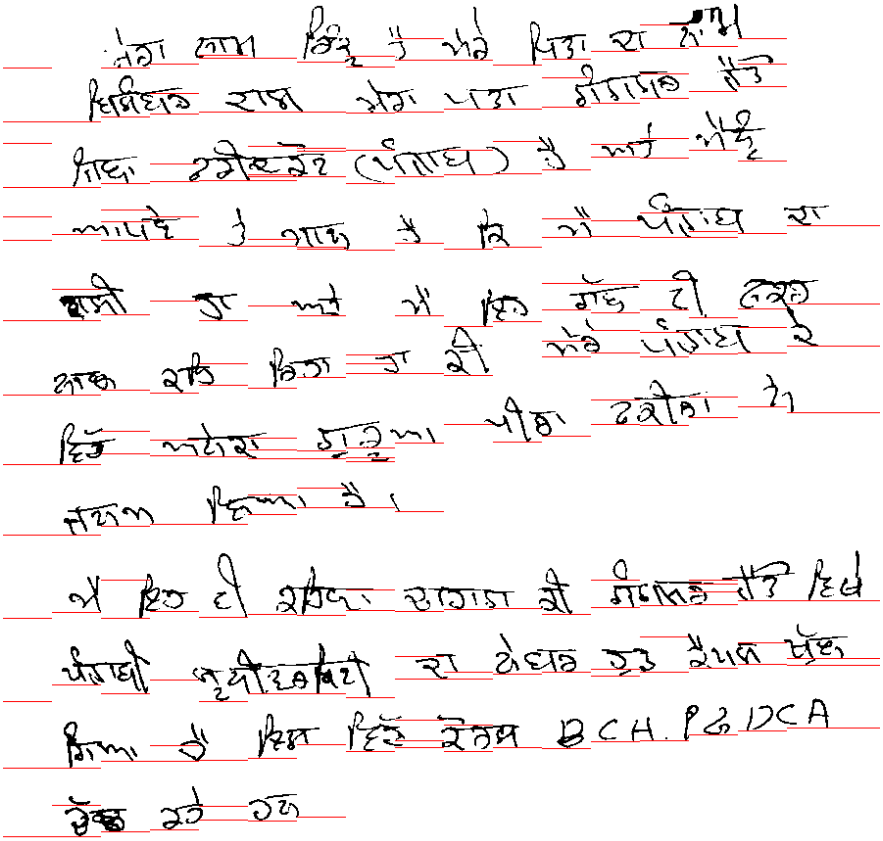

Fig 5: PSLs of each chunk

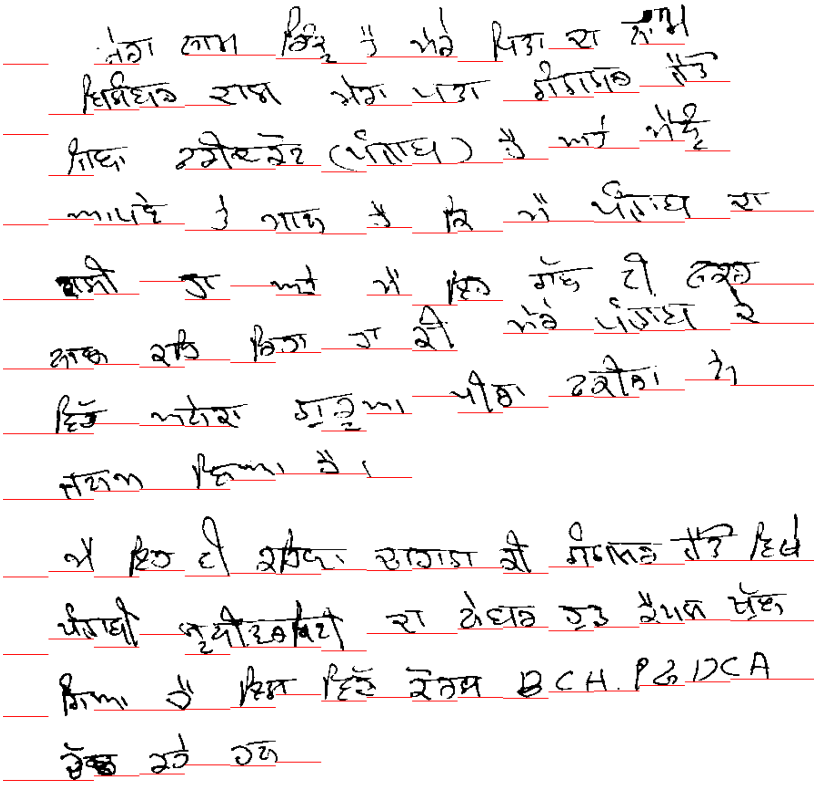

Fig 6: Potential PSLs of each chunk

We stored the y-coordinates of each potential PSL in an array for future use. By proper joining of these potential PSLs, we get individual text lines. It may be noted that sometimes because of overlapping or touching of one

component of the upper line with a component of the lower line, we may not get PSLs in some regions. Also, because of some modified characters of Gurmukhi (e.g. *adhak*, *chandrabindu*) we find some extra PSLs in a chunk. We take care of them during PSL joining, as explained next.

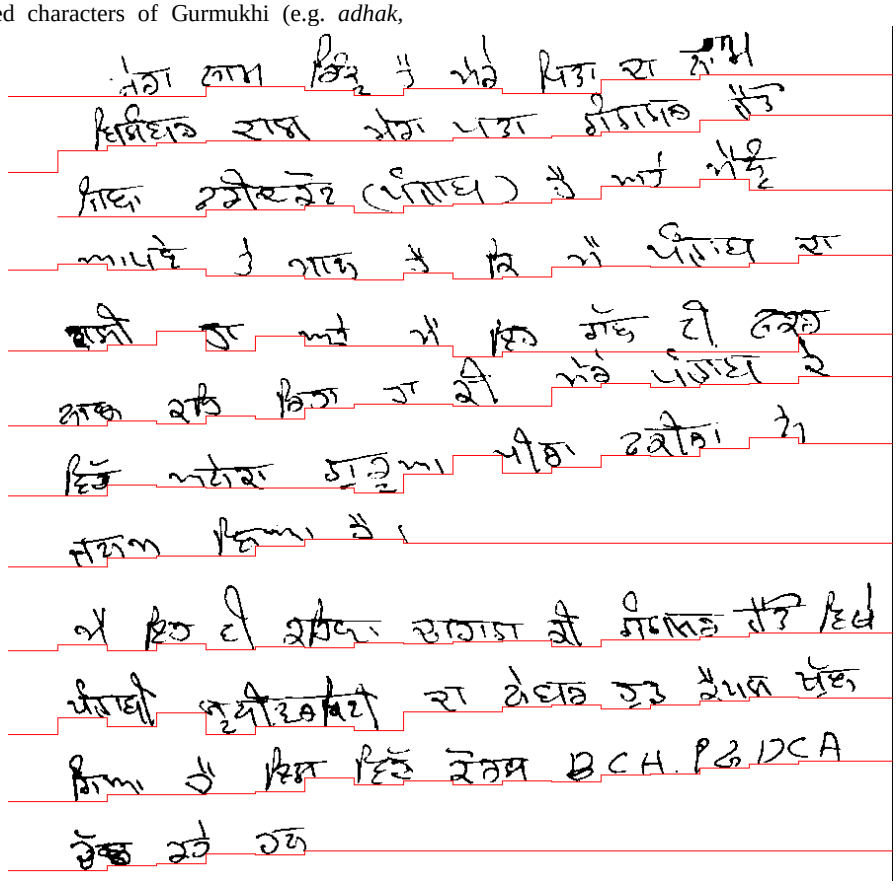

Fig 7: Segmented Text

To join a PSL of the *i*th chunk to a PSL of *(i* + 1*)*th chunk, we search for the closest PSL on upper as well as lower side, then check for the closest PSL and distance of PSL from K_i should be less than 72% of height of word. The distance between joining PSLs should be near half of height of word. If distance is larger than height of word then there is no need to join. The height of word is probably 40-50. So the distance can be upto 3/4th of height of word. By experimenting, we have reached the conclusion of 72%.

If it exists, we join the right co-ordinate of PSL_i with the left co-ordinate of the PSL in the *(i* +1*)*th chunk. If it does not exist, we extend the PSL_i horizontally in the right direction until it reaches the right boundary of the *(i* + 1*)*th chunk or intersects a black pixel of any component in the *(i* + 1*)*th chunk. If the extended part intersects the black pixel of a component of the *(i* + 1*)*th chunk, we decide the "belongingness" of the component in the upper line or lower line. Based on the belongingness of this component, we extend this line in such a way that the component falls in its actual line[12]. Belongingness of a component is decided as follows.

We compute the distances from the intersecting point to the topmost and bottommost point of the component. Let d_1 be the top distance and d_2 the bottom distance and *word_height* is estimated to be 40 for A4 size paper having 18-20 text lines written. If $d_1<d_2$ and $d_1<(word\_height/2)$ then the component belongs to the lower line. If $d_2 \leq d_1$ and $d_2<(word\_height/2)$ then the component belongs to the upper line. If $d_1 > (word\_height/2)$ and $d2>(word\_heigh/2)$ then we assume the component touches another component of the lower line [12]. If the component belongs to the upper-line (lower-line) then the line is extended following the contour of the lower part (upper part) of the component so that the component can be included in the upper line (lower line) as shown in Fig 8.

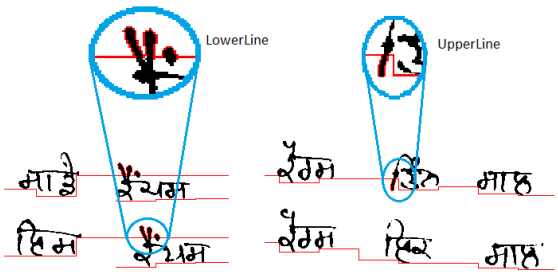

Fig 8: Upperline and Lowerline contours

To follow the contour, we are testing the 8-connectivity (8 neighbouring points). To test the connectivity, we have numbered the pixel to 0 and its neighbouring pixels from 0 to 8 depending upon the type of contour, whether it is upperline or lowerline contour. For upperline (lowerline) contour, we numbered the pixels in clockwise (anti-clockwise) direction as shown in Fig 9:

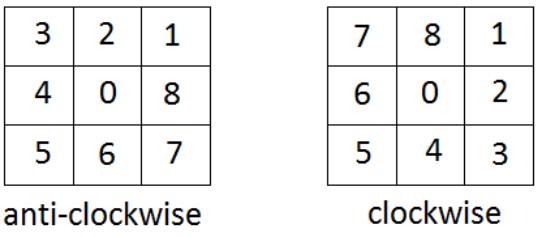

Fig 9: Anti-clockwise contour for lowerline and clockwise contour for upperline

The method for both contours is same. The method is that if the pixel at position 1 is black and at position 2, the pixel is white, then pixel at position 2 will become the current pixel. Similarly, if 2 is black and 3 is white then current pixel will be 3 and so on. Sometimes, the values are repeated and program stucks. At that po int, increment current pixel's x coordinate by 2. The line segmented text is shown in Fig 7.

We have estimated the width size of chunk say W as 70. It is because if we take W = 50, it will make very small chunks and the chances of intersecting with the component are more and if we take W=100, number of chunks will be less and it can protect various lines to segment in case of the document in which text lines are very close to each other.

4. EXPERIMENTS AND RESULTS

The experiments are performed on various handwritten text images in Gurmukhi Script. The images with high skewness, less line gap, more gap in words etc. are considered. For experiments, we considered only single column document pages. By viewing the results on the computer's display, we calculate line segmentation accuracy manually. Accuracy of line extraction algorithm is measured according to the following rule. Distributions of experimental results of line segmentation module are given in Table 1. The accuracy of segmentation of lines for six handwritten text images in different handwritings is given in Table 1.

Table 1: Line Segmentation Accuracy

Text Images	Number of lines correctly segmented/ Total number of lines	Accuracy
Doc 1	12/12	100%
Doc 2	18/20	90%
Doc 3	11/16	69%
Doc 4	18/21	86%
Doc 5	2/19	11%
Doc 6	6/20	30%

The segmented image of Doc 1 is shown in Fig 7. Doc 5 and Doc 6 have very low segmentation accuracy. It is because the lines are very close to each other. The words of lines are overlapping highly. We tried to use contour tracing to accurately segment the intersecting components as shown in Fig 8. This is the example from Doc 2. This method can be applied to other Indian scripts too. But its limitation is that it is size dependant. In future we plan to use different sized text.

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