Converting Optically Scanned Regular or Irregular Tables to a Standardised Markup Format to be Accessible to Vision-Impaired

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Abstract

Documents use tables to communicate multidimensional information clearly, summarise and present data in an easy-to-interpret way. Tabular information in scanned PDF due to its nature without further processing is not accessible for vision-impaired people who use assistive technology such as screen readers. The lack of access to table contents limits educational and workplace opportunities for people with vision impairment. They require a complete equivalent to access table. This paper describes techniques which apply to scanned PDF document for table detection, extraction and cell segmentation to retrieve cell contents and represent them in a navigable manner to vision-impaired. The output is in mark-up format and provides navigation ability to access content of a table.

Keywords: vision–impaired; Optical Character Recognition (OCR); document layout analysis; segmentation; accessibility; markup format; regular or irregular table; navigation

1. Introduction

Tables play a very important role in storing numbers, words or items of any kind, in a definite and compact form for convenience of study, reference or calculation. The structure of a table consists of vertical and horizontal separators and their intersections create table cells. Cell positions can be extracted using image processing techniques. A table may contain different kinds of objects such as text, graphics and mathematical formula (Watanabe, Luo, & Sugie, 1995), (Tsuruoka, Takao, Tanaka, Yoshikawa, & Shinogi, 2001).

Although table cells can be extracted individually, the content cannot be accessed without OCR. The accuracy of recognized information extracted from the table cells strongly depends on performance of OCR which converts images to editable text to convey meaning. Keeping track of the order of cells during extraction and accurate information recognition from the cells are two major issues which must be considered during table information retrieval for accessibility.

Kieninger proposed a method for recognition of table structure based on robust block segmentation starting with the initial block clustering followed by post processing to correct errors and decomposition of columns into individual table cells (Kieninger, 1998).

Tupaj et al presented method to segment and analyze the document image using image processing techniques in order to detect potential table areas. They examined technical and financial tables (Tupaj, Shi, Chang, & Chang, 1996).

Yildiz et al developed a method for extracting table information by utilizing only the absolute position of text elements in a file, concretely in PDF files(Yildiz, Kaiser, & Miksch, 2005).

Ng et al applied Machine learning method and used purely surface features such as the relative locations of characters in a line and across lines. They only focused on detecting tables, columns, and rows and not on the content(Ng, Lim, & Koo, 1999).

In Seth method cell segmentation is performed via indexing of value cells by table header (Seth & Nagy, 2013).
method focused on cell segmentation and structure recognition for regular tables and CSV spread sheets.
This paper is based on the study that considers scanned PDF and focuses on regular/irregular and tables without
ruled lines but not nested tables.

2. Table Structure and Navigation Ability
Tabular data arrangement means arranging data in rows and columns. The use of tables is pervasive throughout all
communication, research and data analysis. Tables appear in print media, handwritten notes, computer software,
ar
ducial ornamentation and many other places. Tables differ significantly in variety, structure, flexibility, 
notation, representation and use. In books and technical articles, tables are typically presented apart from the main
text in numbered and captioned floating blocks. A table consists of rows and columns, the intersection of a row and a
column is a cell. The table cells may be grouped, segmented, or arranged in many different ways, and even nested
(Zielinski, 2006). To provide full access to tables, the titles of table and titles of each column or row must be
presented before reading the data in cells. It should be left up to the user to decide the most logical one-dimensional 
way to read materials. The tables are mostly inaccessible in scanned PDF documents. Additionally, there is no
guarantee for tables in untagged structured PDF documents to be represented in correct order due to lack of tags.

Table structure often contains vertical lines as column separators and horizontal lines as row separators. In an optimal
case, all horizontal and vertical lines which make table structure would be specified. Accessing each data cell
individually is based on finding the related column and row intersection point.

The proposed method in this research is cells extraction by finding (Xmin, Ymin) and (Xmax,ymax) for each cell
which represent cell bounding box. All cell segments are tagged based on positions and sent individually to OCR
software. Cell segmentation provides navigation ability by representing table cells by rows, by columns or both.

3. Methodology
This research developed an open source application runs under Linux and contains the following modules:
1) Block segmentation
2) Line segmentation of the document
3) Caption finding
4) Table detection and extraction (regular and irregular) from scanned documents using table layout
5) Cell segmentation
6) Cell labeling to make navigable Markup format

To facilitate OCR these steps are followed by line and symbol segmentation of cell contents. Figure 1 illustrated
flowchart of system.
3.1 Block Segmentation

If the scanned PDF page contains only one column, lines in the OCR result will be presented in order. However in cases where the PDF includes more than one column, there is no guarantee that the OCR result will convey the correct line ordering. In a double column document the OCR results do not keep the reading order. Block segmentation divides the page into logical blocks, preserving the reading order.

This research utilizes morphological operations for block segmentation. By applying morphology repetitively over a whole image, specific shapes can be found, removed or modified. If a pixel is white and completely surrounded by other white pixels, then that pixel is obviously not on the image edge. The entire process actually depends on the definition of a structuring element or kernel, which defines what pixels are to be classed as neighbours for each specific morphological method. The dilate operation returns the maximum value in the neighbourhood. The erode operation returns the minimum value in the neighborhood. Eroded morphology helps to remove non-interested white spaces between text words and recognises remaining white space area as block separators.

Since PDF document may contain different logical layouts such as footer and multiple columns, block segmentation must be done both vertically and horizontally using the following steps:

- Considering the horizontal white space area;
- Running horizontal segmentation;
- Considering vertical white space area in each extracted horizontal segment; and
- Applying vertical segmentation for all extracted horizontal segments.

3.2 Line Segmentation of the Blocks

In this research, line segmentation is performed based on specifying the rectangular background (white space) as horizontal gap between text lines in the image by finding the position of all lines which meet the following conditions and assuming BL and FL are background lines (gaps) and foreground lines (text) respectively:

\[
BL_i \text{ is a horizontal gaps between text lines:}
\]

\[
If \ length(BL_i) = \text{width(block)} = W \& L_i - 1 \text{ is not similar } BL_i
\]
Assuming $BL_i$ and $BL_j$ are two consecutive horizontal gaps

$$\text{Height (FL)}_l = Y(BL_j) - Y(BL_i)$$

$FL_i$ is obtained by cropping original block image.

The line segmentation module performs based on detecting the consecutive foreground pixels which have following conditions:

Assuming $n$ background Pixels as:

$$BP_0;\cdots;BP_n$$

if $n = \text{Width(Page)}$

$$\text{for (i = 1::n)}$$

$$\text{if (j = i + 1 \& X(BP_j) = X(BP_{i+1}) \& Y(BP_j) = Y(BP_i)}$$

$Y(BP_j)$ is a segmentation position.

### 3.3 Table Extraction

The requirement of detection and identification of tables from document images is crucial to any document image analysis (Mandal, Chowdhury, Das, & Chanda, 2006). Mandal used the algorithm to observe tables that have distinct columns which implies that gaps between the fields are substantially larger than the gaps between the words in text lines.

Seth and Nagy has segmented the CSV table using only “logical layout analysis” without resorting to any appearance features or natural language understanding (Seth & Nagy, 2013). Handley described a table analysis system which reconstructs table formatting information from table images whether or not the cells are explicitly delimited. Inputs to the system are word bounding boxes and any horizontal and vertical lines that delimit cells (Handley, 2000)

Table detection in this research is based on caption finding. Bounding boxes of line segmentation results are checked for caption finding, each line satisfies in $\text{Left margin = Right margin}$ could be a caption.

Assuming:

$I) j > i$

$II) L_i, L_{i+1}, \ldots, L_{j-1}$ are captions

$III) L_j$ is connected component

Then next block is table and extracted as described in block segmentation

### 3.4 Cell Segmentation

A table in a document is a rectilinear arrangement of cells where each cell contains a sequence of words. Several lines of text may compose one cell. Cells may be delimited by horizontal or vertical lines, but often this is not the case (Handley, 2000)

This research performs cell segmentation based on identification of intersections of vertical and horizontal borders as cell’s bounding box. In order to navigate through cells, the structure recognition is necessary for both regular and irregular tables. Unlike regular tables which contain fixed number of cells for all rows, number of cells in a given row of an irregular table varies.

This research investigates the tables which include horizontal and vertical line borders as separators of cells in the table. The following steps shows how to obtain the positions of the borders using image processing techniques such as applying morphology, compose and composite as the following snippet shows.

```bash
convert $1 -threshold 75% bin.png
#convert to binary image
height=$(identify -format "%h" bin.png)
width=$(identify -format "%w" bin.png)
#obtain size of image
```
convert bin.png -alpha off -channel A -transparent white -transparent black -negate +channel

-write mpr:ORG +clone -negate -morphology Erode rectangle:60x1 -mask mpr:ORG -morphology Dilate rectangle:60x1 +mask -compose Lighten -composite -negate txt:-| grep -Ev '#FFFFFF '| sed '1d;s/:.*/ /g;s/,/ /g'| awk '{ print $2 ,$1 }'| sort -b -k1n,1
awk '{ p=print $1 -p,$1+$2 }{p=$1 }{ if (NR==1) print $1 ,$1 ,$2 }'| awk 'p>1'
awk '{ print $2 }'| sort -b -k1n,1> hbox.dat

#obtain x coordinate value of intersections

convert bin.png -rotate 90 -alpha off -channel A -transparent white -transparent black -negate +channel -write mpr:ORG +clone -negate -morphology Erode rectangle:60x1 -mask mpr:ORG -morphology Dilate rectangle:60x1 +mask -compose Lighten -composite -negate -rotate 90 txt:-| grep -Ev '#FFFFFF '| sed '1d;s/:.*/ /g;s/,/ /g'| awk '{ print $1+$2 }'| sort -b -k1n,1
awk '{ p=print $1 -p,$1+$2 }{p=$1 }{ if (NR==1) print $1 ,$1 ,$2 }'| awk 'p>1'
awk '{ print $2 }'| sort -b -k1n,1> vbox.dat

#obtain y coordinate value of intersections

nv=$( cat vbox.dat |wc -l)
nh=$( cat hbox.dat |wc -l)
for (( j=1 ; j <$nh;j++)) do
for (( i=1 ; i <$nv;i++)) do
xs=$(cat vbox.dat | awk 'NR=='$i''
xe=$(cat vbox.dat | awk 'NR=='$i'+1''
ys=$(cat hbox.dat | awk 'NR=='$j''
ys=$(cat hbox.dat | awk 'NR=='$j'+1''

x=$(($xe-$xs))
y=$(($ye-$ys))

convert $1 -shave 0x0 -repage $width"x"$height"+0+0" png:-|convert png:- -crop $x"x"$y"+"$xs"+"$ys 'colum'$i'row'$j'.png
done
done

#Cell extraction using top left and bottom right coordinate values and cropping

In terms of cell segmentation the tables are divided into two categories, namely, regular and irregular tables.

3.4.1 Regular Tables

In regular tables, all vertical borders meet all horizontal borders so all intersections are the coordinate points of cell bounding boxes. It means obtaining intersections leads to access cell bounding boxes. In regular table of size $m \times n$ where $m$ is number of rows and $n$ is number of columns, each row has $n$ number of cells.

Total Number of cells $m \times n$

$m=$Number of rows

$n=$ Number of columns

for( $i=1$ to $m$): row($i$)has $n$ cells

$X_{i}, Y_{i}$ is top left coordinate point

$X_{i+1}, Y_{i}$ is top right coordinate point

$X_{i}, Y_{i+1}$ is bottom left coordinate point

$X_{i+1}, Y_{i+1}$ is bottom right coordinate point

Figure 2 illustrates an image of a regular table, its vertical borders, horizontal borders, points of intersection (right)
and the table of cell bounding boxes (left)

![Table of Cell Bounding Boxes](image)

**Figure 2.** A regular Table, Its Vertical Borders, Horizontal Borders, Points of Intersection (Right) and the Table of Cell Bounding Boxes (Left)

### 3.4.2 Irregular Tables

There is no particular relationship between number of rows and number of columns in irregular tables. Therefore, there is no guarantee that every vertical border meets all horizontal borders. All points of intersection must be checked to make sure whether it is a coordinate point of a cell bounding. The point \((X_t, Y_t)\) is considered as the coordinate of the top left of a cell bounding box if and only if there are two intersection points such that \((X_f, Y_f)\) is the top right coordinate value and \((X_b, Y_b)\) is the bottom left coordinate values which meet the following conditions:

\[
\begin{align*}
(X_f > X_t) & \quad \text{&} \quad (Y_t = Y_f) \\
(X_b = X_t) & \quad \text{&} \quad (Y_b > Y_t)
\end{align*}
\]

As a result \((X_b, Y_b)\) becomes the bottom right coordinate value of the cell which is extracted and labelled as Cell(i) using the top left and bottom right coordinate values of the cell boundary box.

Figure 3 shows an irregular table with vertical and horizontal intersections and extracted cells (not in order).

![Irregular Table](image)

**Figure 3.** Irregular Table
Figure 4 shows different steps for cell segmentation including sample table (top left) horizontal borders extraction (middle left) vertical borders extraction (bottom left) and segmented cells (right).

![Figure 4](image)

**Figure 4.** Sample Table (Top Left) Horizontal Borders Extraction (Middle Left) Vertical Borders Extraction (Bottom Left) and Segmented Cells (Right)

### 3.5 Cell Labelling

Each cell is represented by two coordinate values of top left and bottom right corners.

- **Top left corner** = $TL$
- **Bottom right corner** = $BR$

$cell_i$ and $cell_j$ locate in same column if $xTL_i = xTL_j \& xBR_i = xBR_j$

$cell_i$ and $cell_j$ locate in same row if $yTL_i = yTL_j \& yBR_i = yBR_j$

Assuming $cell_i$ and $cell_j$ : if $xTL_i = xTL_j \& xBR_i < xBR_j$

Then they locate in same column but table is irregular and $cell_i$ may also locates in another column. $cell_i$ and $cell_j$ are located in same row if $yTL_i = yTL_j \& yBR_i = yBR_j$

Assuming $cell_i$ and $cell_j$ if $yTL_i = yTL_j \& yBR_i < yBR_j$

Then they locate in same row but table is irregular and $cell_j$ may also locates in another row.

### 3.6 Markup Presentation and Navigation Ability

In marks-up format table is presented with the `<table></table>` tag.

Table row is defined with the `<tr>` tag.

First row indicates table header and is defined with the `<th></th>` tag.

Table cell is defined with the `<td></td>` tag.

Number of rows in table is number of `<tr>` tag between `<table></table>`

Number cells in each rows is number of `<td>` between `<tr></tr>`

Marks-up format supports navigation ability. Navigation through the table means finding by Row, by Column or by Cell.
cell$_{ij}$ indicates cell(j) which is located row(i). Thus to navigate through it needs to extract the part of marks-up file located between $i^{th}$ of <tr></tr> then in extracted part of find the part located between $j^{th}$ of <td></td>

The following snippet shows how to navigate through

- The specific row$_i$

```bash
START_ROW=location-of-$i^{th}$<tr>=$(cat marks-up.html|grep -bo $(cat merks-up.html|grep -m$i <tr> |tail –n1))
END_ROW=location-of-$i^{th}$</tr>=$(cat marks-up.html|grep -bo $(cat merks-up.html|grep -m$i </tr> |tail –n1))
LENTGH=$[END_ROW-START_ROW]
File=$( cat marks-up.html)
ROW=$(File:$START_ROW:$LENTGH)
```

- The specific column$_j$

```bash
START_CELL=location-of-$j^{th}$<td>=$(echo $ROW|grep -bo $(echo $ROW| grep -m$j <td> |tail –n1))
END_CELL=location-of-$j^{th}$</td>=$(echo $ROW|grep -bo $(echo $ROW| grep -m$j </td> |tail –n1))
LENTGH=$[END_CELL-START_CELL]
CELL=$(ROW:$START_ROW:$LENTGH)
```

### 3.7 Symbol Segmentation

Symbol Segmentation module processes binary image of cell and converts content into a sequence of images. Each output image contains at least one unconnected black segment on a transparent background. It finds first non-transparent pixel in input image, extracts segment at this coordinate value and removes that segment from image. This task is repeated until all non-transparent pixels are processed. In order to improve performance of segmentation, a distance morphological function is used to create larger space between symbols.

In this module, the segmented lines of a cell obtained in previous step are converted into a series of isolated symbols and sent to OCR considering their orders.

### 3.8 Table in Spread Sheet

Tables in spread sheet usually are presented without border lines. This research treat them such as table with border lines. Applying line segmentation horizontally and vertically makes boarder lines then cell segmentation is performed as stated in section 3.4. This method can be used for information retrieval and navigate through ‘table of contents” The following snippet is used for making border lines as it is shown in Figure 5.

```bash
in="$1"
height=$(convert "$in" -format '%h' info:)
width=$(convert "$in" -format '%w' info:)
#obtain table dimensions
convert "$in" -threshold 60% -fuzz 1% -trim +repage -scale 50% bin.mpc
#convert to binary image
convert bin.mpc -morphology erode diamond:2 bbin.mpc
#remove non-interested white spaces by eroding
max=$(convert bbin.mpc txt:-|sed 's/.* #/ /g;1d;s/ / /g'|grep 'FFFFFF'| awk '{ print $1, $2 }'| sort -b -k2n,2| awk '{ a[$2 ]++} END { for (i in a) print i,a[i]} ' |sort -b -k2n,2| awk 'END { print $2 }')
```
#find maximum height of horizontal white spaces

```bash
x=$(($ convert bbin.mpc txt:-| sed 's/::*/ /g;1d;s,/;/g'|grep 'FFFFFF'| awk '{ print $1,$2 }') | sort -b -k2n,2| awk '{ a[$2 ]+=} END { for (i in a) print i,a[i] }' | awk '{max=' $2<5' | sort -b -k1n,1| awk '{print $1}' | awk '{p=print $1,p}{p=$1 }{ if (NR ==1) print $1 ,$1 }' | awk '$1-$2 !=1 ' | awk '{print $1 }')

nox=${#x[@]}
```

#find locations of extended horizontal white spaces

```bash
for (( e=0;e<$nox ;e++ )) do
  l=$(($e+1))
  if [[ $e -ne $(($nox-1)) ]]; then
    convert bin.mpc -crop "0x""($((${x[$l]}-${x[$e]}}))"+0""${x[$e]}" hor$l.png
  else
    convert bin.mpc -crop "0x""($height-${x[$e]})"+0""${x[$e]}" hor$l.png
  fi
done

convert $(ls hor*.png) -background red -splice 1x1+0+0 -append row.png
```

#insert horizontal border lines in proper locations

```bash
max=$(convert bbin.mpc txt:-|sed 's/::*/ /g;1d;s,/;/g'|grep 'FFFFFF'| awk '{ print $2,$1 }') | sort -b -k2n,2| awk '{ a[$2 ]+=} END { for (i in a) print i,a[i] }' | awk '{max=' $2<5' | sort -b -k1n,1| awk '{print $1}' | awk '{p=print $1,p}{p=$1 }{ if (NR ==1) print $1 ,$1 }' | awk '$1-$2 !=1 ' | awk '{print $1 }')
```

```bash
#find maximum width of vertical white spaces

```bash
x=$(($ convert bbin.mpc txt:-| sed 's/::*/ /g;1d;s,/;/g'|grep 'FFFFFF'| awk '{ print $2,$1 }') | sort -b -k2n,2| awk '{ a[$2 ]+=} END { for (i in a) print i,a[i] }' | awk '{max=' $2<5' | sort -b -k1n,1| awk '{print $1}' | awk '{p=print $1,p}{p=$1 }{ if (NR ==1) print $1 ,$1 }' | awk '$1-$2 !=1 ' | awk '{print $1 }')
```

```bash
nox=${#x[@]}
```

#find locations of extended vertical white spaces

```bash
for (( e=0;e<$nox ;e++ )) do
  l=$(($e+1))
  if [[ $e -ne $(($nox-1)) ]]; then
    convert bin.mpc -crop $((${x[$l]}-${x[$e]})))"x0+"${x[$e]}"+0" ver$l.png
  else
    convert bin.mpc -crop $((width-${x[$e]})))"x0+"${x[$e]}"+0" ver$l.png
  fi
done
```

```bash
convert $(ls ver*.png) -background red -splice 1x1+0+0 +append column.png
```

#insert vertical border lines in proper locations
| Code  | Title                                                                 | Year |
|-------|------------------------------------------------------------------------|------|
| 123384| Interdisciplinary studies in the long nineteenth century                | 2005 |
| 34647 | 3CMedia                                                                | 20   |
| 39853 | Research                                                               | 0102 |
| 200090| A Contracorriente                                                      | 2002 |
| 32180 | A St A: Advances in Statistical Analysis                               | 0104 |
| 30533 | AA files: annals of the Architectural Association School of Architecture| 1201 |
| 36072 | AI E: International Transactions                                      | 12   |
| 15571 | AACN Advanced Critical Care                                            | 1103 |
| 44322 | AANA Journal                                                           | 1103 |
| 44323 | AAON Journal                                                           | 1110 |
| 14674 | AAP PharmSciTech                                                       | 1115 |
| 19052 | Abacus: a journal of accounting, finance and business studies          | 1501 |
| 32508 | Abhandlungen aus dem Mathematischen Seminar der Universitaet Hamburg  | 0101 |
| 6715  | Aboriginal History                                                     | 2005 |
| 35596 | About Performance                                                      | 1901 |
| 1    | Abstract and Applied Analysis                                           | 0101 |
| 15574 | Academic Emergency Medicine                                            | 1103 |
| 44325 | Academic Journal of Second Military Medical University                 | 1117 |
| 19961 | Academic Leadership                                                    | 1301 |

**Figure 5.** Table in Spread Sheet (Top), Inserting Horizontal Line Border Lines (Middle), Inserting Vertical Lines Border (Bottom)
4. Conclusion

This study focused on detection and extraction tables using document layout analysis. Furthermore, cell segmentation is performed to facilitate OCR. The markup format for the table is created by cell labelling based on bounding boxes. This output preserves the navigation ability which is required for information retrieval from tables. Cell segmentation was performed based on finding boundary boxes using horizontal and vertical border lines intersection recognition. Tables without border lines such as csv files, spared sheets or table of contents are investigated by finding extended vertical and horizontal white spaces and inserting border lines in interested location of white spaces. Then cell segmentation is performed as before.

Segmented cells are labelled based on location and sent to symbol segmentation module and OCR to be converted to plain text. The accuracy of final retrieved information from table cells severely depends on segmented cells.

This research did not involve to OCR algorithm and used TESSERACT and OCRopus two open sources OCR under Linux. For testing stage developed application applied for 25 double column scientific scanned article. 23 out of 25 (92%) regular tables were recognized and extracted. 18 out of 23 (78%) were segmented completely. TESSERACT performance showed 13 out of 18 (72%) converted to text successfully, however OCRopus accuracy was 10 of 18 (0.55). Text from each cell is tagged to make marks-up format which supports navigation and searching ability without any error.

Obtained result proved segmented cells must be preprocessed before sending to OCR. Further development will investigate segmented cells preprocessing before OCR and extracting cells from nested tables using a recursive algorithm.

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