A Case Study Report on Tabular Text Data Extraction Using EasyOCR

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Case Study Overview

This case study travels through the critical process of extracting text data from images and converting it into a structured tabular format, such as Excel or CSV. Beyond just data retrieval, this conversion enables seamless analysis, manipulation, and integration into existing workflows, unlocking the full potential of the data for tasks like analysis and decision-making. The study focuses on evaluating EasyOCR, a versatile Python module designed for text extraction from images, to address these challenges. Despite EasyOCR lacking native table recognition support, its adaptability for reading both natural scene text and dense document text makes it a promising solution.

The objectives of the study include detecting tabular regions within input images, extracting text data accurately, converting it back into a tabular format, and identifying areas for performance enhancement. To do this, I used various techniques from OpenCV to prepare the image and identify text regions. Then, I used EasyOCR to recognize the text within these regions. Methodologically, the process involves preprocessing images, detecting text regions, passing the region of interest to EasyOCR for text detection, and converting the result back into a tabular form using Python scripts.

Challenges such as handling captions, resizing images, and watermarks are encountered, yet EasyOCR proves efficient, particularly for tables with empty columns. By combining EasyOCR for text extraction and a custom approach for table recognition, effective data extraction from image-based tables is achieved, addressing diverse needs effectively.

This project focuses on turning text found in images into organized tables, like those in Excel or CSV files. Extracting text from images is important because it allows us to analyze and use the information more easily. However, traditional methods struggle with converting image text into structured tables. EasyOCR, a Python tool, helps with this task, even though it doesn't directly recognize tables.

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I. Introduction:

The importance of extracting text data from images extends beyond just data retrieval; it lies in the subsequent conversion of this extracted text into a structured and easily manipulable format. Converting the extracted text into a tabular format, such as Excel or CSV, holds immense significance for enabling seamless analysis, manipulation, and integration into existing workflows. By structuring the extracted text into rows and columns, organizations can unlock the full potential of the data, facilitating tasks such as data analysis, reporting, and decision-making.

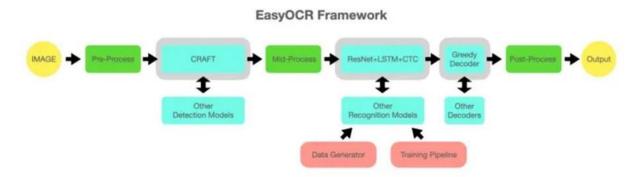


Figure 1.1: EasyOCR Framework

The purpose of this case study is to explore the efficacy of EasyOCR, a Python module specifically designed for extracting text from images, in addressing the aforementioned challenges. EasyOCR distinguishes itself by its versatility in reading both natural scene text and dense text in documents, offering a comprehensive solution for a wide range of applications. With support for over 80 languages, EasyOCR presents a promising tool for organizations seeking to streamline their text extraction processes across diverse linguistic contexts.

Objectives of the Case Study:

- 1. Tabular Region Detection: The primary objective of this case study is to detect tabular regions within given input images by performing a sequence of OpenCV operations.
- **2.** *Text Extraction:* Building upon the detection of tabular regions, this case study aims to evaluate EasyOCR's efficacy in accurately extracting text data located within these identified tabular regions.

- 3. *Tabular Conversion:* In addition to text extraction, this case study seeks to convert the extracted text data back into a tabular format accurately.
- 4. Identification of Performance Enhancements: Beyond assessing EasyOCR's current capabilities, this case study endeavors to identify potential areas for performance enhancement. By pinpointing any shortcomings or challenges encountered during the tabular region detection, text extraction, or conversion processes, the study aims to propose recommendations for refining EasyOCR's functionality and optimizing its utility in real-world applications.

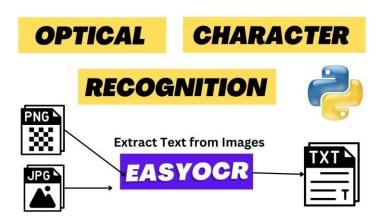


Figure 1.2: Text Detection using EasyOCR

II. Methodology: Back to Index

1. Text Extraction Process from Images:

The text extraction process from images involves several steps to accurately identify and capture textual content embedded within the visual data. This process begins with reading the input image and proceeds through various image processing techniques to enhance the visibility and clarity of the text regions. Once the text regions are identified, optical character recognition (OCR) technology is applied to extract the textual content from the image.

2. Modules or Packages Used for Text Extraction and Tabular Conversion:

• In this case study, the primary tool utilized for text extraction from images are OpenCV for extracting tabular regions and EasyOCR, a Python module specifically designed for extracting text from images. EasyOCR leverages deep learning-based models to recognize

and extract text from images with high accuracy and efficiency. Additionally, for the conversion of the extracted text data into a tabular format, standard Python libraries such as Numpy, Pandas may be employed to structure the data into rows and columns resembling a spreadsheet.

Trigonometry	Name:	Oeses.		
Unit 5: Trigonometric an		od:		
Real World Applic				
Part 1: You will create a collage of pictures illustrating all six trigonometric functions (sine,				
cosine, tangent, cosecant, secant, cotangent) tetc.), architecture (bridges, doorways, etc.), and				
furniture, etc.).				
Requirements: Your project must contain: 1. Pictures of the entire objects where the trigonometric properties of the entire objects where objects where the entire objects where the entire objects whe	ometric function is found			
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cosecant, secant, cotangent (no repeat picture	THE CONTRACT OF STATE			
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5. CREATIVITY!!!				
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Illustration: Your collage should be created using * white or colored poster board	g the following restrictions:			
* use scissors to cut out pictures (no tearing)				
* use glue to paste pictures (no taping)				
Grading: You will be graded according to the fo	ollowing rubric:			
Category	Points Possible	Points		
	The control of the co			
Example of Sine	1 point			
Example of Sine Example of Cosine	1 point			
Example of Cosine	1 point			
Example of Cosine Example of Tangent	1 point			
Example of Cosine Example of Tangent Example of Cosecant	1 point 1 point 1 point			
Example of Cosine Example of Tangent Example of Cosecant Example of Secant	1 point 1 point 1 point 1 point			
Example of Cosine Example of Tangent Example of Cosecant Example of Secant Example of Cotangent	1 point 1 point 1 point 1 point 1 point 1 point			
Example of Cosine Example of Tangent Example of Cosecant Example of Secant Example of Cotangent Examples of nature (at least 1)	1 point			
Example of Cosine Example of Tangent Example of Cosecant Example of Secant Example of Cotangent Examples of nature (at least 1) Examples of architecture (at least 1)	1 point 4 points (1pt for each)			
Example of Cosine Example of Tangent Example of Cosecant Example of Secant Example of Cotangent Examples of nature (at least 1) Examples of architecture (at least 1) Examples of everyday items	1 point 4 points (1pt for each)			

Figure 2.1: Input Image

3. Overview of Preprocessing Steps Applied to the Images:

Prior to text extraction, preprocessing steps are applied to the input images to enhance the quality and clarity of the textual content. These preprocessing steps include:

- **Reading the image:** The input image is read into the environment for further processing.
- *Converting it to grayscale:* The image is converted from color to grayscale to simplify the subsequent image processing operations.
- **Sharpening edges:** Edge enhancement techniques are applied to sharpen the edges of text regions, making them more distinguishable from the background.
- *Thresholding:* A thresholding operation is performed to segment the image into foreground (text) and background regions based on pixel intensity values.
- *Canny edge detection:* The Canny edge detection algorithm is applied to identify prominent edges in the image, further enhancing the visibility of text regions.
- *Closing operation on Canny:* Morphological closing operations are performed on the Canny edges to bridge small gaps and smooth the contours of text regions.
- *Contour detection:* Contours of text regions are detected using contour detection algorithms, allowing for the identification of tabular regions.
- Identifying the largest contour as the tabular region of interest (RoI): Among the detected contours, the largest contour is identified as the tabular region containing the text data.
- *Cropping the input image to the RoI:* The input image is cropped to isolate the tabular region of interest, facilitating focused text extraction and analysis.

```
image = cv2.imread('WhatsApp Image 2024-05-03 at 15.57.16_82fdd2a1.jpg')
gray_image = cv2.cvtColor(image_file, cv2.COLOR_BGR2GRAY)
kernel = np.array([
      [-1,-1,-1],
      [-1,10,-1],
      [-1,-1,-1]
])
```

```
sharpened_image = cv2.filter2D(gray_image, -1, kernel)
threshold = cv2.threshold(sharpened_image, 180, 255, cv2.THRESH_BINARY)[1]
canny = cv2.Canny(threshold, 180, 200, apertureSize = 7)
closing = cv2.morphologyEx(canny, cv2.MORPH CLOSE, kernel)
contour = cv2.findContours(closing, cv2.RETR EXTERNAL, cv2.CHAIN APPROX SIMPLE)[0]
cv2.drawContours(image_file.copy(), contour, -1, (0, 0, 255), 1)
largest_contour = None
largest bbox = None
largest_area = 0
for cnt in contour:
    area = cv2.contourArea(cnt)
   if area > largest_area:
            largest_contour = cnt
            largest bbox = cv2.boundingRect(cnt)
            largest_area = area
 imgcontour = cv2.drawContours(image_file.copy(),[largest_contour],-1,(0,0,255),1)
x, y, w, h = largest bbox
img\_with\_bbox = cv2.rectangle(imgcontour, (x, y), (x+w, y+h), (0, 255, 0), 2)
roi = image_file[y : y + h, x : x + w]
```

Table 2.1: Table Detection Python Script

- The code begins by reading an input image file named 'input_image.jpg' using the OpenCV library's **imread()** function as numpy array, storing it in the variable 'image'. Subsequently, the image is converted to grayscale using the **cvtColor()** function with **COLOR_BGR2GRAY**, resulting in a grayscale representation stored in the variable 'gray_image'.
- Next, a 3x3 kernel is defined for sharpening the edges of the image. This kernel is then applied to the grayscale image using the **filter2D()** function, resulting in a sharpened image stored in the variable 'sharpened_image'.

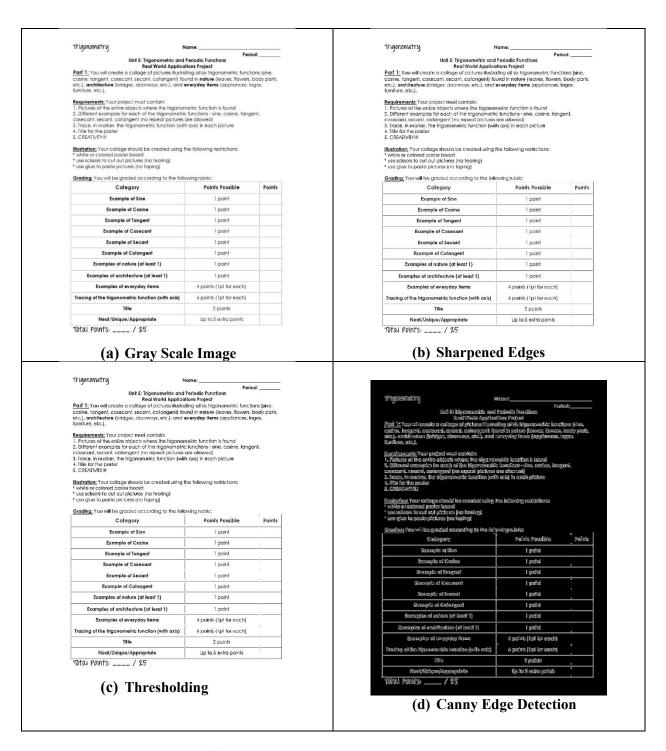


Table 2.2: Preprocessing Techniques on Input Image

The sharpened image undergoes a thresholding operation using the **threshold()** function, where pixels with intensity values greater than 180 are set to 255 (white), while others are set to 0 (black). This binary thresholded image is stored in the variable 'threshold'.

- ♣ Subsequently, the Canny edge detection algorithm is applied to the thresholded image using the Canny() function, resulting in the detection of prominent edges, which are stored in the variable 'canny'.
- A morphological closing operation is then performed on the edges detected by the Canny algorithm using the **morphologyEx()** function with **MORPH_CLOSE**. This operation fills gaps between edges and smooths the contours, producing a closed image stored in the variable 'closing'.

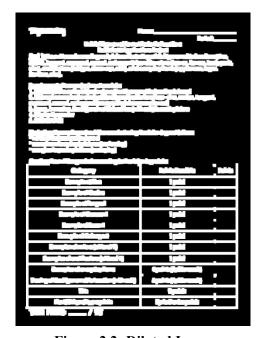


Figure 2.2: Dilated Image

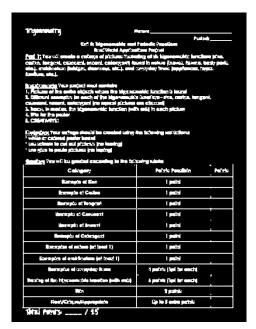


Figure 2.3: Closing Operation
On Input Image

- ♣ I chose to use a closing operation instead of dilation to specifically ignore any text that's directly above the table or attached to its top. Closing helps smooth out edges and merge nearby elements, making it easier to focus on just the main table content. This way, we can avoid extracting text that might be close to the table but isn't part of it.
- ♣ Contours are then detected in the closed image using the **findContours()** function, which returns a list of contours stored in the variable 'contour'.
- ♣ The largest contour is identified by iterating through the list of contours and comparing their areas. The bounding box coordinates (x, y, width, height) of the largest contour are extracted and stored in individual variables.

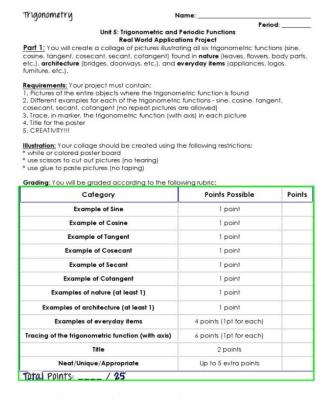


Figure 2.4: Largest Contour Detected

Finally, a rectangle is drawn around the region of interest (ROI) defined by the bounding box coordinates on a copy of the original image, highlighting the detected tabular region. The region of interest is then extracted from the original image using the bounding box coordinates and stored in the variable 'roi' for further processing.

Category	Points Possible	Points
Example of Sine	1 point	
Example of Cosine	1 point	
Example of Tangent	1 point	
Example of Cosecant	1 point	
Example of Secant	1 point	
Example of Cotangent	1 point	
Examples of nature (at least 1)	1 point	
Examples of architecture (at least 1)	1 point	1
Examples of everyday items	4 points (1pt for each)	
Tracing of the trigonometric function (with axis)	6 points (1pt for each)	
Title	2 points	
Neat/Unique/Appropriate	Up to 5 extra points	

Figure 2.5: Input Image Cropped to Region of Interest

4. Passing RoI to EasyOCR for Text Detection:

After extracting RoI by preprocessing the given input image, we can simply pass the RoI to EasyOCR module for detecting only text present in the tabular region.

```
import easyocr as oc
reader = oc.Reader(['en'])
result = reader.readtext(roi)
```

Table 2.3: Passing RoI to EasyOCR Module

This code snippet imports the EasyOCR library for text recognition on a designated region of interest (ROI) extracted from an image. An OCR reader object is instantiated using the 'Reader()' function from the EasyOCR library, configured to recognize text in the English language ('en'). This object, named 'reader', is then utilized to invoke the 'readtext()' method, passing the ROI extracted from the image as its parameter. The 'readtext()' method processes the text within the specified ROI using optical character recognition techniques and returns the result, which typically comprises the detected text and its corresponding bounding box coordinates. This result is captured in the variable 'result' for subsequent analysis or further processing.

Detected Test Results: [([[145, 8], [228, 8], [228, 35], [145, 35]], 'Category', 0.9775009221672206), ([[433, 11], [555, 11], [555, 31], [433, 31]], 'Points Possible', 0.9985647332397556), ([[625, 11], [677, 11], [677, 31], [625, 31]], 'Points', 0.9948524298369781), ([[129, 51], [247, 51], [247, 51], [247, 69], [129, 69]], 'Example of Sine', 0.870297427798347), (([1476, 46], [523, 46], [523, 70], [476, 70]], 'Point', 0.9625031258880472), (([[115, 81], [258, 81], [258, 106], [115, 106]], 'Example of Cosine', 0.9817100291745914), (([481, 87], [521, 87], [521, 105], [481, 105]], 'Point', 0.9293741400607151), (([114, 117], [262, 117], [262, 141], [114, 141]], 'Example of Tangent', 0.7430791363523259), (([476, 116], [523, 116], [523, 140], [476, 140]], 'Point', 0.9696398408886909), (([109, 155], [269, 155], [269, 175], [1 09, 175]], 'Example of Cosecant', 0.962936743247506), (([481, 157], [521, 157], [521, 173], [481, 173]], 'Point', 0.9638947131924763), ([[117, 191], [259, 191], [259, 211], [117, 211]], 'Example of Secant', 0.8150827687667433), (([479, 191], [521, 191], [521, 211], [479, 211]], 'point', 0.9634583272227314), (([105, 227], [273, 227], [273, 247], [105, 247]], 'Example of Cotangent', 0.8167816188413394), (([486, 246]], 'Point', 0.8965821886855843), (([77, 261], [299, 261], [299, 281], [77, 281]], 'Examples of nature (at least 1)', 0.7540243659569185), (([479, 261], [521, 261], [521, 281], [479, 281]], 'point', 0.953802130288594), (([55, 29], [321, 299], [321, 319], [55, 319]], 'Examples of architecture (at least 1)', 0.8848066390658598), (([476, 296], [523, 296], [5

Figure 2.6: List of Tuples held by result variable

The list contains collection of tuples which encompass three values in it. They are as follows,

```
([[145,\,8],\,[228,\,8],\,[228,\,35],\,[145,\,35]],\,'Category',\,0.9775009221672206) - 0^{th}\,Index\,\,of\,\,List\,\,([[145,\,8],\,[228,\,8],\,[228,\,35],\,[228,\,35]),\,[228,\,35])
```

1st **Index of Tuple:** [[145, 8], [228, 8], [228, 35], [145, 35]] – It specifies the bounding box information around the text.

2nd Index of Tuple: 'Category' – It specifies the actual detected text content from RoI Image

3rd Index of Tuple: 0.9775009221672206 – This is the confidence value of the detected text information.

Category	Points Possible	Points	
Example of Sine	1 point		
Example of Cosine	1 point		
Example of Tangent	1 point		
Example of Cosecant	1 point		
Example of Secant	1 point		
Example of Cotangent	1 point		
Examples of nature (at least 1)	1 point		
Examples of architecture (at least 1)	1 point		
Examples of everyday items	4 points (1pt for each)		
Tracing of the trigonometric function (with axis)	6 points (1pt for each)		
Title	2 points		
Neat/Unique/Appropriate	Up to 5 extra points		

Figure 2.7: Detected Text in RoI After Passing it to EasyOCR Module

5. Converting Back to Tabular Form:

The process of converting back to tabular representation involves several steps aimed at structuring data into a format called a table. It begins with

→ Detecting the number of columns from the resultant list of tuples. This step is crucial as it determines how the data will be organized horizontally.

```
def detect_columns(result, start = 0):
    cols = []
    col_ref = []
    x1 = []
    y1 = []
    for tup in range(len(result)):
        x, y = result[tup][0][0]
        x1.append(x)
        y1.append(y)
```

```
beg = y1[start] - 10
end = y1[start] + 10
for i in range(start, len(result)):
    # st.write(i, y1[i])
    if (y1[i] >= beg) and (y1[i] <= end):
        cols.append(result[i][1])
        col_ref.append(x1[i])
    return (x1, cols, col_ref)

x1, cols, col_ref = detect_columns(result)</pre>
```

Table 2.4: Detecting Number of Columns from result variable

Once the number of columns is determined, the next step involves appending the detected row information to the corresponding columns. Each row of data is assigned to its respective column based on the arrangement determined earlier. This ensures that the data is properly aligned within the table structure.

Table 2.5: Appending Remaining Detected Text Information to Respective Columns

Figure 2.8: (a) Initial Empty Table

- (b) List of Columns
- (c) Column References for Appending Rows

Finally, after all the row information has been appended to the columns, the resultant list is converted into a Pandas DataFrame. Pandas is a powerful Python library commonly used for data manipulation and analysis, and converting the list to a DataFrame allows for further processing, analysis, and visualization of the tabular data. By structuring the data into a tabular representation, it becomes easier to interpret, analyze, and derive insights from the dataset.

```
df = pd.DataFrame(table).transpose()

df = df.rename(columns = {i:cols[i] for i in range(len(cols))})

df.to_excel("results.xlsx")

print(df)
```

Table 2.6: Converting List to DataFrame and Loding them to Excel/ CSV Files

	Category	Points Possible	Points
0	Example of Sine	point	None
1	Example of Cosine	point	None
2	Example of Tangent	point	None
3	Example of Cosecant	point	None
4	Example of Secant	point	None
5	Example of Cotangent	point	None
6	Examples of nature (at least 1)	point	None
7	Examples of architecture (at least 1)	point	None
8	Examples of everyday items	points (Ipt for each)	None
9	Title	points (Ipt for each)	None
10	Neat/Unique / Appropriate	2 points	None
11	25	Up to 5 extra points	None

Figure 2.9: DataFrame Outcome

	Α	В	С	D
1		Category	Points Possible	Points
2	0	Example of Sine	point	
3	1	Example of Cosine	point	
4	2	Example of Tangent	point	
5	3	Example of Cosecant	point	
6	4	Example of Secant	point	
7	5	Example of Cotangent	point	
8	6	Examples of nature (at least 1)	point	
9	7	Examples of architecture (at least 1)	point	
10	8	Examples of everyday items	points (Ipt for each)	
11	9	Title	points (Ipt for each)	
12	10	Neat/Unique / Appropriate	2 points	
13	11	25	Up to 5 extra points	
1/				

Figure 2.10: DataFrame Loaded in Excel File

III. Challenges: Back to Index

Here are some of the challenges that I faced while doing this study,

• I came to know that EasyOCR does not offer native support for table recognition. For tasks specifically involving the recognition of tables within images, a more suitable solution would be *PaddleOCR's PP-Structure* model. This model is specifically designed to recognize and extract structured information from tables present in images. So, I found it difficult to translate detected texts information into tabular formats like Excel, CSV, TSV files. I just overcame this challenge by analyzing the outcome returned by **result** variable

After analyzing the output of **result** variable, I just highlighted tuples with different colors based on the pattern that I identified. The tuples highlighted with different color in the below image implies different meaning.

- Green Color Indicates Number of columns of Input Image
- Yellow Color Indicates Individual row information of Column 1
- Unhighlighted Text Indicates row information of Column 2

```
([[143, 7], [230, 7], [230, 35], [143, 35]], 'Category', 0.7159639437158294),
([[432, 8], [556, 8], [556, 32], [432, 32]], 'Points Possible', 0.8149897168967521),
([[625, 11], [677, 11], [677, 31], [625, 31]], 'Points', 0.9109005020581996),
([[129, 51], [247, 51], [247, 69], [129, 69]], 'Example of Sine', 0.8702974277998347),
([[480, 50], [521, 50], [521, 69], [480, 69]], 'point', 0.793118081793997),
([[115, 81], [258, 81], [258, 106], [115, 106]], 'Example of Cosine', 0.9817100291745914),
([[481, 87], [521, 87], [521, 105], [481, 105]], 'point', 0.9293741400607151),
([[114, 117], [262, 117], [262, 141], [114, 141]], 'Example of Tangent', 0.7430791363523259),
([[476, 116], [523, 116], [523, 140], [476, 140]], 'point', 0.9696398408886909),
([[108, 154], [269, 154], [269, 175], [108, 175]], 'Example of Cosecant', 0.8046274847951407).
```

Figure 3.1: A Snapshot of output returned by result variable

I found out the columns just by taking into account only (x1, y1) coordinates which bounds the detected text region. That is, $result[i][0][0] \Rightarrow 143$, 7 here the value of i ranges from 0 to n, where n is the number of texts detected and stored as tuple information.

I figured out, for columns, the value of y1 seems to be varying slightly whereas x1 varies greatly. The value of y1 which changes slightly can be adjusted by maintaining some threshold value. So, whatever the values that are ranging between this threshold can be appended to the **columns** list. Once the actual columns of the table are appended to **columns** list, I just appended the corresponding value of x1 (whose value changes greatly) to **column_ref** list. This is just to place the remaining data items to respective columns.

Now we have Columns and Column References for appending data items that corresponds to respective column. By measuring the length of columns, I just appended that many empty lists to the table list. Similar to the above case, for rows, the value of x1 seems to be varying slightly whereas y1 varies greatly. The value of x1 which changes slightly can be adjusted by maintaining some threshold value. So, whatever the values that are ranging between this threshold can be

appended to the corresponding empty lists which we created earlier within table list with reference to column_ref list.

• The next challenge that I faced is, whenever I resize an image or preprocess it before passing it to the EasyOCR model, the model lacks its ability to detect the correct text which resides in the given input image. The wrongly detected texts are highlighted in the Figure 3.4

Points Possible	Points
1 point	
4 points (1pt for each)	
6 points (1pt for each)	
2 points	
Up to 5 extra points	
	1 point 4 points (1pt for each) 6 points (1pt for each) 2 points

Calegory	Points Possible	Points
Example of Sine	1 point	1
Example of Cosine	1 point	
Example of Tangent	1 point	
Example of Cosecant	1 point	
Example of Secant	1 point	
Example of Colangent	1 point	
Examples of nature (at least 1)	1 point	
Examples of architecture (at least 1)	1 point	Ì
Examples of everyday items	4 points (1pt for each)	
Tracing of the trigonometric function (with axis)	δ points (Tpt for each)	
Title	2 points	
Neat/Unique/Appropriate	Up to 5 extra points	

Total Points: ____ / 25

Figure 3.2: Before Preprocessing

Figure 3.3: After Preprocessing

Detected Test Results: [([[141, 3], [232, 3], [232, 40], [141, 40]], 'Category ,', 0.9228015098251716), ([[430, 4], [556, 4], [556, 36], [430, 36]], 'Points Possible ', 0.5910789239018812), ([[125, 42], [248, 42], [248, 71], [125, 71]], "'Exa mple of Sine", 0.5827103059987683), ([[465, 42], [523, 42], [523, 71], [465, 71]], 'Tpoint', 0.5132014665955953), ([[113, 75], [261, 75], [261, 112], [113, 112]], 'Example of Cosine', 0.7083658303662448), ([[464, 78], [522, 78], [522, 108], [464, 108]], 'ipoli', 0.41084367936867405), ([[112, 116], [262, 116], [262, 140], [112, 140]], 'Example ot Tanigent', 0.64 4011376378862), ([[466, 116], [522, 116], [522, 140], [466, 140]], Ipoinf, 0.29725171867162), ([[106, 150], [270, 150], [270, 178], [106, 178]], "Example of Cosecant'", 0.8935830221264257), ([[114, 188], [260, 188], [260, 212], [114, 212]], 'Example of Secant', 0.581145718038942), ([[466, 188], [522, 188], [522, 214], [466, 214]], 'Lpoint', 0.9299289758012137), ([[103, 218], [274, 218], [274, 248], [103, 248]], 'Example oi Cotangenf', 0.7973716506747129), ([[466, 220], [522, 22 0], [522, 248], [466, 248]], Ipoini , 0.40269861700833043), ([[74, 254], [306, 254], [306, 286], [74, 286]], 'Exampies o i_nafure (af jogsf f);', 0.10818344332200303), ([[478, 254], [524, 254], [524, 286], [478, 286]], <mark>'paini -'</mark>, 0.3686596907 134466), ([[52, 292], [322, 292], [322, 324], [52, 324]], 'Examples of archiecture (atleasi W', 0.19552054484353737), ([[466, 294], [524, 294], [524, 326], [466, 326]], 'Jpoin _', 0.440099256753988), ([[82, 332], [294, 332], [294, 356], [82, 356]], "Examples of everyday items'", 0.5135205667166407), ([[402, 332], [576, 332], [576, 358], [402, 358]], '#4 points (Tpt for eachl', 0.42892416517349746), ([[13, 363], [360, 363], [360, 393], [13, 393]], 'Tracing oi lhe igonomelric funct ion (wih &xis)', 0.2598808575501405), ([[410, 364], [576, 364], [576, 394], [410, 394]], 'U points ipi for eachy', 0.272 1005127198858), ([[171, 405], [207, 405], [207, 425], [171, 425]], "Title'", 0.39498669619739274), ([[456, 404], [530, 40 4], [530, 428], [456, 428]], '12 points}', 0.64431310363352), ([[90, 436], [286, 436], [286, 462], [90, 462]], 'NeatJUniq ue/ Appropiate .', 0.4951969285837001), ([[418, 436], [568, 436], [568, 462], [418, 462]], '2P fo 5 extra points', 0.2299 2901222990567), ([[10, 470], [134, 470], [134, 493], [10, 493]], 'Total Points', 0.9263859212815321), ([[202, 470], [256, 470], [256, 493], [202, 493]], 1i051' 0.03198374489602229), ([[622.5430712064641, 9.377528171635387], [677.5817300484] 72, 4.472037318251267], [678.4569287935359, 32.62247182836461], [623.418269951528, 36.52796268174873]], 'Points', 0.91803 04640198443), ([[463.6027607015021, 151.34524533285406], [520.4883107813158, 145.32419886156282], [523.3972392984979, 175 .65475466714594], [466.5116892186842, 181.67580113843718]], 'Ipoini-', 0.2905128447667485)]

Figure 3.4: EasyOCR fails to detect text accurately after preprocessing

IV. Limitations:

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This is not applicable for table with too many empty cells and for blurred images, but it
outperforms in case an entire column is empty as shown in the figure below. It is prone
to None value

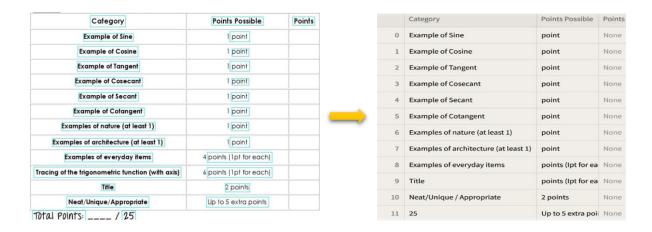


Figure 4.1: Input and Outcome

• If a null value is present in between any cells in a particular column then the data that has to be placed below the null value will be **shifted up to Null value's position** and the place for null value will be preserved at the end of the table.

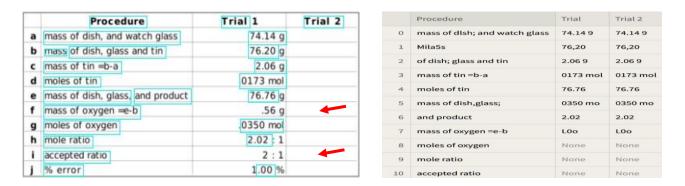


Figure 4.2: Left Image: RoI, Right Image: Tabular Representation with None values at end of Table

- **Preprocessing**/ **Resizing** original image comparably affects the EasyOCR's model ability to produce a correct outcome.
- As mentioned in the EasyOCR documentation, the detected text will not follow natural
 human reading so depending on the context the columns may interchange in the output
 by preserving its data items in the corresponding columns.

• Watermarks/ background information present in an image may affect the positioning of data items in the table slightly.

S.No	Parameter	Specifications	Result	Test Method
1.	Mashed Potato	Negative	Negative	FSSAI Manual
2.	Sweet Potato	Negative	Negative	FSSAI Manual
3.	Other Starch	Negative	Negative	FSSAI Manual
4.	Rancid Stuff (Old Ghee)	Negative	Negative	FSSAI Manual
5.	Synthetic Colouring Matter	Negative	Negative	FSSAI Manual
6.	Vegetable Oil & Fat	Negative	Negative	FSSAI Manual
7.	Test For Vanaspati	Negative	Negative	FSSAI Manual
8	Curcumin	Negative	Negative	FSSAI Manual
9.	Dalda	Negative	Negative	FSSAI Manual
10.	Lead, mg/kg	Max. 2.5	Not Detected	FSSAI Manual
11.	Arsenic, mg/kg	Max. 1.1	Not Detected	FSSAI Manual
12.	Mercury, mg/kg	Max. 1.0	Not Detected	FSSAI Manual
13.	Cadmium, mg/kg	Max. 1.5	Not Detected	FSSAI Manual

Figure 4.3: RoI with Watermark/ Background Information

	SNo	Parameter	Specifications	Result	Test Method
0	Mashed Potato	Mashed Potato	Negative	Negative	FSSAI Manua
1	Sweet Potato	Sweet Potato	Negative	Negative	FSSAI Manua
2	Other Starch	Other Starch	Negative	Negative	FSSAI Manua
3	Rancid Stuff (Old Ghee)	Rancid Stuff (Old Ghee)	Negative	Negative	FSSAI Manua
4	Synthetic Colouring Matter	Synthetic Colouring Matter	Negative	Negative	FSSAI Manua
5	Vegetable Qil & Fat	Vegetable Qil & Fat	Negative	Negative	FSSAI Manua
6	Test For Vanaspati	Test For Vanaspati	Negative	Negative	FSSAI Manua
7	Curcumin	Curcumin	Negative	Negative	FSSAI Manua
8	Dalda	Tabb	Negative	Negative	FSSAI Manua
9	10.	Dalda	Max 2.5	Not Detect	FSSAI Manua

Figure 4.4: Labs in the previous image detected as Tabb and placed in between actual data items

V. <u>Problem Identified</u>:

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♣ In some cases, the region of interest was detected along with the caption information of the table as shown in the Figure 4.3. Since here I'm considering the text data that is present in first few tuples as the columns for a given input image.

♣ To overcome this, I just measured the length of columns detected. If it is less than 2 then that will be stored in the 'header' variable.

```
if len(cols) < 2:
   header = cols[0]
   counter += 1
   x1, cols, col_ref = detect_columns(result,
   counter)</pre>
```

Table 5.1: Python Script for Checking Caption/Header Information

```
def detect_columns(result, start = 0):
```

Figure 5.1: By default, the 'start' is set to zero in function definition

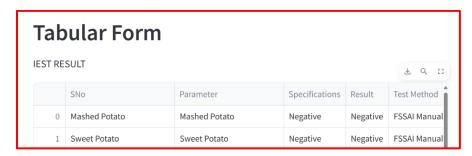
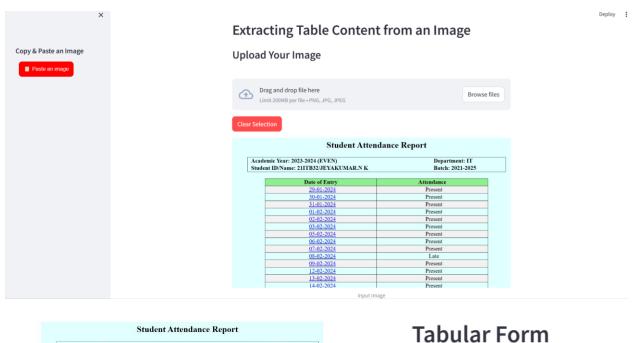


Figure 5.2: Detected Caption Information Displayed Separately Outside the Table

VI. Conclusion: Back to Index

In conclusion, while EasyOCR may not be suitable for tables with excessive empty cells, it excels when entire columns are empty. Preprocessing or resizing images may impact its accuracy slightly, but overall, EasyOCR reliably preserves data integrity, even if column order may vary. Despite potential challenges like watermarks, EasyOCR remains a valuable tool for efficient and accurate table data extraction. By combining EasyOCR for general text extraction and the approach that I followed for specialized table recognition, table image-based data extraction can be achieved, serving to diverse needs effectively.

VII. Related Works: Back to Index



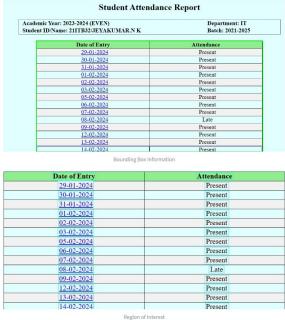




Figure 6.1: Deployed as a Web Application Using Streamlit

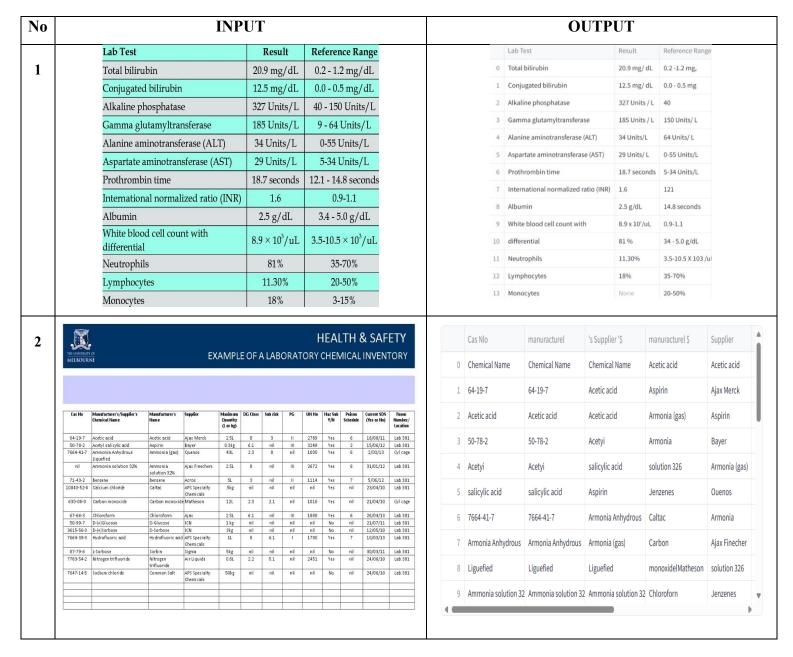


Table 6.1 Passing Different Inputs to the application

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VIII. References:

- https://www.jaided.ai/easyocr/tutorial/
- https://www.analyticsvidhya.com/blog/2021/06/text-detection-from-images-using-easyocr-hands-on-guide/
- https://towardsdatascience.com/pre-processing-in-ocr-fc231c6035a7