



# AN OVERVIEW OF CHARACTER RECOGNITION FOCUSED ON OFFLINE HANDWRITING

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## ABSTRACT

The programmed acknowledgment of written by hand content -, for example, letters, compositions or whole books - has been a center of serious examination for a very long while. Especially in this field of unconstrained penmanship acknowledgment the written work styles of different journalists were managed, serious challenges are encountered. Engagement is basic to the achievement of learning exercises, for example, composing, and can be advanced with suitable feedback. Here, we portray a learning scientific framework called Tracer, which infers behavioral engagement measures which makes perceptions of behavioral examples of understudies composing on a cloud-based application. Be that as it may, the discoveries from the present study recommend that envisioning power could be helpful to recognize diverse understudy practices when drawing nearer a composition task. It is gotten from a system neural based framework for unconstrained penmanship acknowledgment. As such it performs template-free spotting, i.e. it is a bit much for a catchphrase to show up in the preparation set. The watchword spotting is complete utilizing a change of the binarization calculation in conjunction with an intermittent neural system. This paper exhibit that the proposed frameworks beats not just an established element time contorting based approach moreover a present day watchword spotting framework.

**Keywords:** optical character recognition, keyword spotting, Line based, point based, Height based visualization, image smoothing, noise removal.

## 1. INTRODUCTION

AN OCR is a framework which can read message from a printed archive and can send it to the PC for further preparing. Robotized character acknowledgment framework has picked up prevalence for its application in Computer Vision; content based choice making and insightful content acknowledgment frameworks. The acknowledgment of the characters from a printed archive or from any manually written scripts should be portioned essentially. A proficient calculation needs several amounts of parts as a sweeping number of components lead toward countless hence cause time utilizations. Literary reports misuse is extremely agonizing and exorbitant given their measurements, which are regularly vast. In the writing, there are a few techniques for computing a diminished number of measurements for an arrangement of information. Division of the report picture (finding and separating content information) gets to be noteworthy issue. Content occurrences are to be segregated from different articles, for example, design. Lines, images, bits of commotion, and so on. Also, the content can show up in different styles, sizes and introductions. Maybe the most troublesome part of division emerges from the way that characters can be touching each other or some other graphical item. To address these division issues, we have executed a division framework in view of a various leveled, repetitive neural system structure. The objective of division is to disconnect and concentrate all content that might be available on a report picture. The principal phase of preparing uses an ANN that characterizes examples of associated pixels from the report picture. The associated pixel examples are produced by an arrangement of capacities that permit a solitary pass pixel network investigation of the archive picture with synchronous

component era. Manages bunching of content records by neural systems. For representation of content records is utilized the Vector Space (VS) model, which depicts the content archives by VS lattice  $X$ . Multidimensional space of framework  $X$  for content reports grouping requires the high computational intricacy accordingly. It is required of its diminishment. The content reports handling in the multidimensional space of network  $X$  requires high computational multifaceted nature when all is said in done; along these lines it is expected to change the content record representation from high dimensional content archive space into diminished space with lower measurement. Many-sided quality of LSI model taking into account SVD. The paper manages conceivable arrangement of the content archive bunching by projective measurement of subspaces by neural systems. There exist approaches that tackle that issue by utilizing Projective Adaptive Resonance Theory (PART) neural system. This is because of the way that OCR of machine printed reports is adequately precise and the archive word excess is sufficiently high that recovery is not hindered by incidental interpretation blunders. High word excess in the objective archives can make up for blemished translation nonetheless, this may not work if record word repetition is low as is basic in short reports, or if acknowledgment exactness is not high as is normal for written by hand records. Be that as it may, if an essayist composes reliably, the measurable structure of the best arrangements of different penmanship occasions of the same word will be comparative, and that comparability will yield a decent N-best records match. This vigor to interpretation blunders permits the techniques to function admirably even on little archives with low word repetition. The enhanced hunt capacity that we have exhibited can possibly make



manually written databases more profitable to clients. Most archive examination and acknowledgment frameworks have been tried on printed reports first because of the way that the qualities of such records are less demanding to extricate.

## 2. EXISTING WORK

In existing framework has not ready to execute as revision of the incline and to standardize the span of the content lines pictures. The cloud-based methodology can't track users' composing practices in the event that they „copy and paste“ from another editorial manager. Especially in the space of unconstrained penmanship acknowledgment, in which the composition styles of different journalists was managed, extreme troubles are experienced. Current engagement measures depend generally on information gathered by onlookers or self-reported by the members. Because of the multifaceted nature of the information caught amid the written work movement, it is trying to create a basic and significant representation and can't be perform a non-uniform inclination rectification to change the content as right position. Our prior work on disconnected from the net penmanship acknowledgment framework is ordinary.

### A. Optical Character Recognition (OCR)

Optical character acknowledgment (Optical Character Recognition) (OCR) is the mechanical or electronic transformation of pictures of composed, physically composed or printed content into machine-encoded content. It is generally utilized as a type of information passage from printed paper information records, whether identification reports, receipts, bank explanations, mechanized receipts, business cards, mail, printouts of static-information, or any suitable documentation.

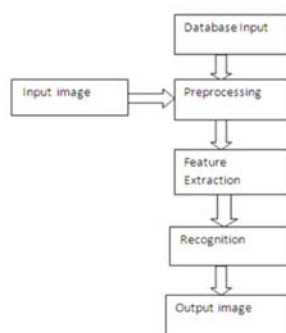


Figure-1. Optical character recognition.

### B. Character recognition

There are two essential sorts of center OCR calculation, which might create a positioned rundown of competitor characters. [Matrix coordinating includes contrasting a picture with a put away glyph on a pixel-by-pixel premise; it is otherwise called "arrangement arranging", "layout certification", or, or "picture connection". This depends on the info glyph being

effectively disconnected from whatever is left of the picture, and on the put away glyph being in a comparable textual style and at the same scale. This strategy works best with typewritten message and does not function admirably when new textual styles are experienced. This is the system the early physical photocell-based OCR actualized; rather straightforwardly. Highlight extraction breaks down glyphs into "elements" such as lines, shut circles, line bearing, and line crossing points. These are contrasted and a conceptual vector-like representation of a character, which may lessen to one or more glyph models. General methods of highlight location in PC vision are pertinent to this kind of OCR, which is regularly seen in "keen" penmanship acknowledgment and undoubtedly generally present day OCR programming. Closest neighbor classifiers, for example, the k-closest neighbor's calculation are utilized to contrast picture highlights and put away glyph includes and pick the closest match. Programming, for example, Cuneiform and Tesseract utilize a two-pass way to deal with character acknowledgment. The second pass is known as "versatile acknowledgment" and utilizes the letter shapes perceived with high certainty on the principal go to perceive better the remaining letters on the second pass. This is worthwhile for bizarre text styles or low-quality sweeps where the textual style is mutilated (e.g. obscured or faded). The OCR result can be put away in the institutionalized ALTO configuration, a devoted XML pattern kept up by the United States Library of Congress.

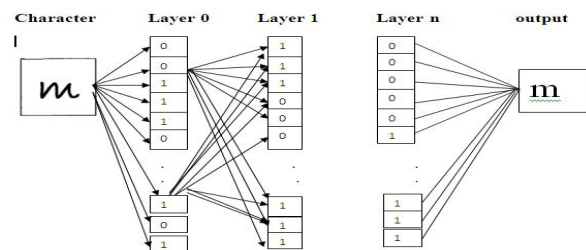


Figure-2. Character recognition.

### C. Post - processing

OCR exactness can be expanded if the yield is compelled by a vocabulary - a rundown of words that are permitted to happen in a record. This may be, for instance, every one of the words in the English dialect, or a more specialized dictionary for a particular field. This method can be hazardous if the report contains words not in the dictionary, as formal people, places or things. Tesseract utilizes its word reference to impact the character division venture, for enhanced precision. The yield stream might be a plain content stream or document of characters, however more complex OCR frameworks can safeguard the first format of the page and create, for instance, a commented on PDF that incorporates both the first picture of the page and a searchable printed representation. "Near-neighbor examination" can make utilization of co-event frequencies to right mistakes, by taking note of those specific words are regularly seen together. ". Learning of the linguistic



use of the dialect being examined can likewise figure out whether a word is prone to be a verb or a thing, for instance, permitting more prominent precision.

### 3. PROPOSED WORK

Power and consistency measures of engagement utilizing a mechanized investigation framework and assess its precision to naturally measure engagement in the connection of composing exercises. Nearby amazing from a content picture delegated having a place with the lower gauge by a Binarization are utilized to precisely assess the incline and the flat arrangement. A manufactured neural system based framework for unconstrained penmanship acknowledgment. We break down the execution of the fundamental neural systems when utilizing them as a part of an acknowledgment errand took after by watchword spotting on the delivered interpretation. Furthermore, clients expressed that the Representations were lucid with their composition movement and were valuable to offer them some assistance with reflecting on the written work process. The size standardization tries to make the framework invariant to the character size and to decrease the unfilled foundation zones. The watchword spotting as distinguishing Word-Based, Line-Based, Document-Based Keyword Spotting, Background Noise. We bring up the upsides of watchword spotting with binarization when contrasted with exemplary content line acknowledgment. The derived engagement measures by contrasting these and member self reports.

#### A. Handwritten document pre- processing

To begin with, the associated segments of the archive picture are separated and investigated so as to take out a few outlines like deletions, over-burden or underlined zones which one knows, from the earlier, that they don't describe the penmanship. At that point, the staying associated parts are portioned into graphemes. This group does not allude to a particular penmanship portrayal and might be confounding. The graphemes are really rudimentary examples of the penmanship that are delivered by a division calculation in view of the investigation of the minima on the upper form. The connection of two (separately three) neighboring graphemes (grapheme  $i$  and grapheme  $i+1$ ) gives what we call bigrams (individually trigrams) of graphemes.

**Step 1:** Calculate  $b_i$ , which are the normal  $b$  estimations of the crossing point of distinguished line  $i$  and the associated part's bouncing box ( $i=1..n$ )

**Step 2:** Exclude from the method the keep going line  $n$  if the accompanying condition is not fulfilled:

$$a=a_i \quad \sum_{b=b_i-(b_i-b_{i-1})^{10}}^{a_i} F(a,b) / \sum_{a=a_i}^{a_i} F(a,b) > 0.01$$

**Step 3:** For each line  $i$ ,  $i=1..n-1$ , we characterize zones  $Z_i$  as per the accompanying requirement:

$$b_i + \frac{b_{i+1}-b_i}{2} < b < b_{i+1}$$

At that point, we process the skeleton of the associated segment, recognize all intersection focuses and expel them from the skeleton in the event that they lie inside zone  $Z_i$ . On the off chance that no intersection point exists, in the division zone  $Z_i$  we uproot all skeleton focuses on the focal point of the zone.

**Step 4:** For each zone  $z_i$  banner with id 1 the skeleton parts that converge with line  $i$ . Every other part are hailed with id 2. At last, in every zone  $z_i$  detachment of the underlying associated segment into various portions is proficient by relegating to a pixel the id of the nearest skeleton pixel

#### B. Noise removable

Commotion in records is characterized in light of the criteria in the event that it is subject to the hidden substance or free of the fundamental substance. Stray imprints, peripheral clamor, ink blobs and salt and pepper uproar are self-governing of size; area of the focal substance moreover surface of the watched spot example is free of the hidden substance. Obscure, pixel move or seep through on other hand is indigent commotion, as they show themselves contrastingly relying upon the substance. Such substance subordinate clamor is nearly more hard to display, scientifically nonlinear and regularly multiplicative. Clamor can likewise be arranged taking into account its consistency in properties such as periodicity of event in the report, its shape, position and dark qualities

#### C. Types of noise

##### a) Gaussian noise

Gaussian clamor is measurable commotion that has a likelihood thickness capacity of the typical circulation (otherwise called Gaussian appropriation). At the end of the day, the qualities that the commotion can tackle are Gaussian disseminated. It is most normally utilized as added substance repetitive sound yield added substance white Gaussian commotion (AWGN).

##### b) Poisson noise

Poisson commotion has a likelihood thickness capacity of a Poisson circulation

##### c) Salt and pepper noise

It speaks to itself as haphazardly happening white and dark pixels. A viable clamor diminishment strategy for this kind of commotion includes the utilization of a middle channel. Salt and pepper clamor creeps into pictures in circumstances where brisk homeless people, for example, defective exchanging, happen. The picture after contortion from salt and pepper commotion resembles the picture connected



#### D. Median filtering

Middle separating is one sort of smoothing strategy. The execution is not that vastly improved than Gaussian obscure for elevated amounts of commotion, though, for dot clamor And salt and pepper clamor (imprudent commotion), it is especially successful Median channel is the one of the strategy to perform this commotion decrease by performing neighborhood averaging, which can stifle segregated out of reach commotion yet the symptom is it obscures sudden changes like sharp edges. The middle channel is a compelling strategy that can stifle disengaged commotion without obscuring sharp edges. In Median Filtering, all the pixel qualities are initially sorted into numerical request and afterward supplanted with the center pixel esteem.

#### E. Methods for feature extraction

Line - based and Height based procedure is one the manually written script recognized method which utilizes the element, In this system longest line keep running of dark pixels on the lines of a content word will be any longer than that of transcribed script . This is so on the grounds that the characters in a word are by and large associated.

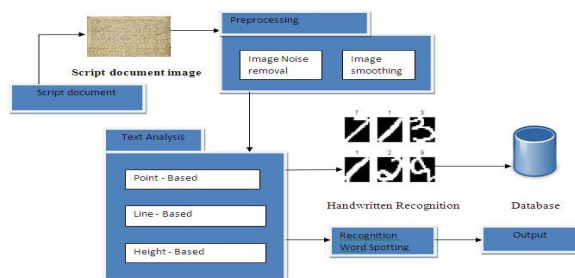


Figure-3.

#### System framework

#### 4. IMPLEMENTATION

##### A. Modules

- Script document/image
- Text Analysis
- Image preprocessing
- Intensity Based Algorithm
- Word spotting / ASCII Output

##### B. Modules description

##### A) Script document/Image

Catchphrase spotting alludes to the procedure of recovering all examples of a given word from a report. We concentrate on written by hand records, for example, letters, updates, or original copies (E.g. Resource

Document). The written by hand report has been Scanned and changed over into pictures, without deciphering the records to information.

##### B) Text analysis

The arrangement is removed by a sliding window that have to move from the left to the directly over the picture. The catchphrase spotting as distinguishing Point-Based, Line-Based, Height-Based, Keyword Spotting. Neural system to penmanship acknowledgment comprises of two sections. The principal phase is a preprocessing stage, done by the neural system. It maps each position of an input sequence to a vector, indicating the probability of each character possibly being written at that position.

##### C) Image preprocessing

Gaussian, on the grounds that it is powerless to commotion present on crude natural picture information, so in any case, the crude picture is convolved with a Gaussian channel. Somewhat obscured variant of the first which is not influenced by a solitary uproarious pixel to any noteworthy degree. After culmination of numerous cycle of preprocess all records are divided into individual content lines. From every line, a succession of highlight vectors is separated, which is then submitted to the neural system.

##### D) Intensity Based Algorithm (IBA)

In this algorithm, a series is defined as a group of events represented by a line. These events are grouped based on the duration between neighboring events. Each line is associated with a weight that indicates the intensity of the line. We define a hashmap, where each entry contains a time threshold and a corresponding weight value. For example, (0.5h, 0.8) indicates that the time threshold is 0.5h and its corresponding weight is 0.8. If the duration between neighboring events is less than the smallest time threshold, we assign that corresponding weight to the series to which that segment be-longs. Based on our experiences with writing activities in learning situations is shown.

##### E) Word spotting / Output

Dynamic Time Warping (DTW) based keyword spotting system, while the second one is a recently proposed learning-based keyword spotting system using Hidden Markov Models (HMM). A state-of-the-art Hand Written Recognition system to transcribe the text. Hand Written Recognition(HWR) has provide the language information can have a positive effect on the recognition rate.

#### 5. EXPERIMENTS AND RESULTS

Using handwritten images which extracted from the training set we evaluated the performance of our method. For the test data 2000 pairs of connected characters were used and to constructing rules another 1000 pairs were used and determine the thresholds. 89.2% of the patterns were correctly segmented by using this method. The comparison with algorithms was shown in



the table. The segmentation rates of the sum of the correct rate, error rate, and rejection rate were re-calculated to total 100%. The quality of the segmented images seems to

be improved in the case of touching digits that share a stroke.

Algorithm	Correct rate	Error rate	Rejection rate	database
Intensity Based algorithm(IBA)	80.80%	19.20%	-	Original data, 120 images
Line-based Visualization	88.15%	7.15%	4.70%	200 images
Point- based Visualization	88.06%	3.94%	8.00%	250 images

## 6. PERFORMANCE ANALYSIS

Dataset comprises of aggregate of 1550 number of tests of capitalized English letters in order, lowercase English letter sets and English numerals. Three arrangements of capitalized English letters in order are tried, three arrangements of lowercase English letters in

order are tried and the Pictures of various configurations are considered to upgrade the flexibility of the venture. Pictures from various journalists are utilized. In this manner slight difference in composing styles is likewise considered.

Analysis performed for uppercase English alphabets:  
Total number of characters trained=130  
Total number of characters tested=520  
Total number of characters recognized correctly=474  
Total accuracy=91.15%

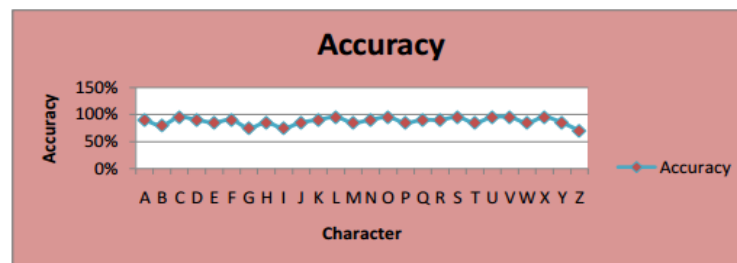


Fig 7.1: Analysis representing accuracy levels for uppercase alphabets

Analysis performed for lowercase English alphabets:  
Total number of characters trained=130  
Total number of characters tested=520  
Total number of characters recognized correctly=471  
Total accuracy=90.57%



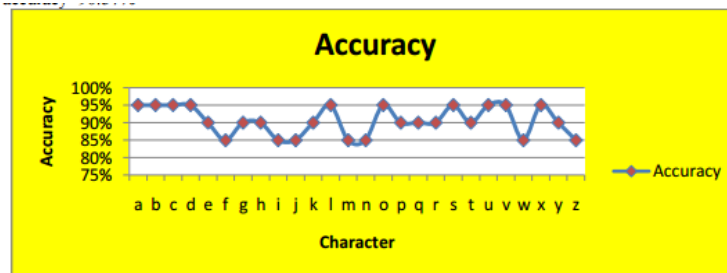


Fig 7.2: Analysis representing accuracy levels for lowercase alphabets

Analysis performed for digits written in English:  
 Total number of characters trained=50  
 Total number of characters tested=200  
 Total number of characters recognized correctly=182  
 Total accuracy=91%

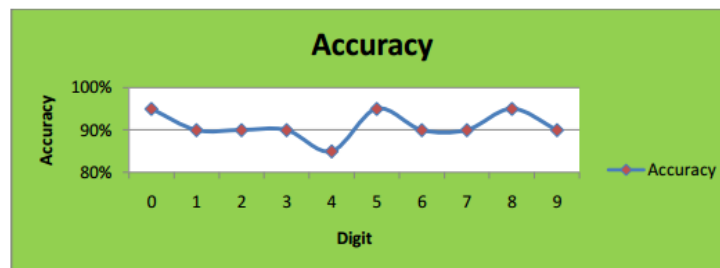


Fig 7.3: Analysis representing accuracy levels for digits

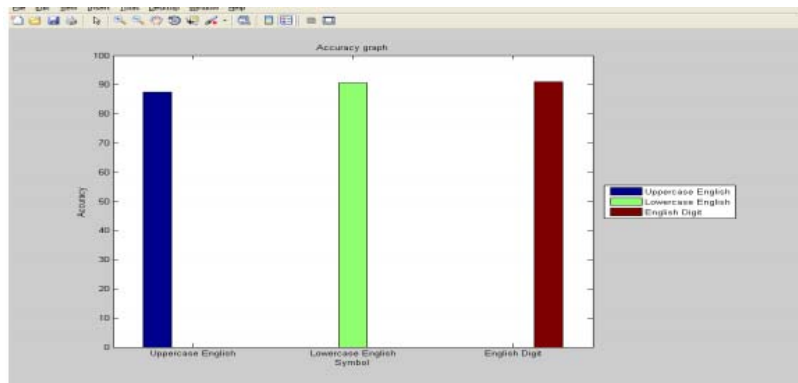


Fig 7.4 : Graph showing total accuracy for uppercase alphabets, lowercase alphabets and digits.

$$\text{Accuracy (Recognition rate)} = \frac{\text{Number of characters recognized correctly}}{\text{Total number of characters}}$$

Analysis performed for digits written in English:  
 Total number of characters trained=50  
 Total accuracy=91%

## 7. CONCLUSIONS AND FUTURE WORK

In this paper we displayed a novel methodology utilizing bidirectional long-transient neural systems in mix with a changed form of the Intensity Based calculation. In our proposed work, we introduce three types of method, line - based, height - based and point - based visualizations. This system has several advantages. First, it is a line based approach and does not need any word segmentation. Secondly, although the system needs to be

trained, it does not require bounding boxes around characters or words as often needed in the literature. The only requirement is a transcription of the text lines in the training set. Finally, being derived from a general neural network based handwritten text recognition system, any arbitrary string can be searched for, not just the words appearing in the Training set.

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