

ISSN 2063-5346



HANDWRITTEN CHARACTER RECOGNITION USING ARTIFICIAL NEURAL NETWORKS

Jagbeer Singh, Soumya Kohli, Tanu Sharma, Muskan Goyal,
Nisha Singh

Article History: Received: 01.02.2023

Revised: 07.03.2023

Accepted: 10.04.2023

Abstract

The written character recognition (HCR) methodology still includes a big variety of applications. Across any union government like an Asian nation, reading communicating addresses in many states contains distinct languages. One vital application of HCR within the automatic industry altogether developed nations is the verification of order of payment amounts and signatures. The documents' optical character recognition is compared to written documents by an individual. Characters from many sorts of files, as well as image and word document files, square measure translated victimization this OCR. This study article's primary goal is to supply an answer for numerous handwriting recognition ways, as well as bit input via a mobile screen and picture files. Artificial neural networks applied math techniques, and different techniques square measure utilized by popular methodologies to handle issues that don't seem to be linearly dissociative. This analysis study uses a spread of comparison and recognition techniques to spot handwriting characters in image documents. in addition, the study contrasts the analytical way to support vector machine (SVM) heuristic network methodology with applied math, templet similarity, well-organized pattern identification, and comprehensive ways. The technique, depiction, and style of the Written Character Recognition System, as a system assessing and developing outcomes, also are coated within the article. The objective is to point out how well neural networks recognize handwriting characters.

Keywords— Support vector machine; written character process; artificial intelligence; neural network

Meerut Institute of Engineering and Technology, Meerut

DOI:10.31838/ecb/2023.12.s1-B.197

Introduction

This document is a template. Written numerals and symbol recognition became more and more vital in the current digital environment because of their sensible applications in every type of regular activity. It is incontestable the point that, lately, numerous identification systems are evolved or are planned to be used in numerous areas wherever big categorization potency is required. Structures that recognize handwriting letters, characters, and digits assist folks in many complicated tasks that might be long and expensive. a wonderful instance of the usage of automated process structures in financial institutions to method finance cheques.

In recent years, handwritten symbol identification has been among the foremost interesting and difficult analysis domain in picture process and pattern identification. It contributes a major part in the development of automation processes and improves the confluence between man and machine in the very type of application. many studies have been conducted to develop new techniques and ways of reducing time intervals whereas up recognition accuracy,

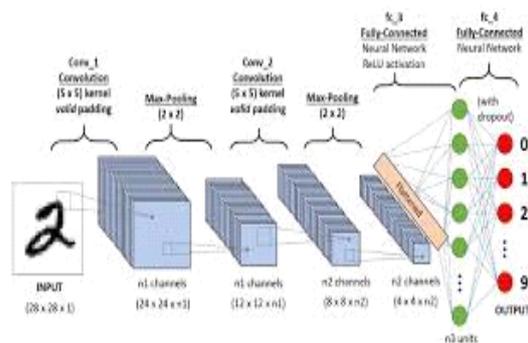


Fig. 1. Convolutional Neural Network Architecture

When folks browse Handwriting characters, letters, words, or digits, their sensory system is heavily concerned. When reading handwriting, it seems easy, however, it's not as straightforward as folks believe. though everything is finished unconsciously, a

personality will add up to what they see supported by what their brains are schooled in human might not realize how troublesome it's to resolve handwriting issues. The visual pattern recognition

The challenge is simply apparent once developing an automatic data processing system to browse handwriting. the synthetic neural networks approach is widely considered the foremost effective methodology for designing systematic models for recognizing characters which are handwritten. In a simplified type, expert systems help in replicating how the brain of a human being works once reading handwriting. Furthermore, reading handwriting is long and tedious, particularly once folks square measure needed to scan many written reports written by completely varying folks. A neural layered network is the best suitable planned structure thanks to its capacity to procure that means from complicated information and discover tendencies in the information that square measure troublesome to identify victimization in different human techniques or humans. the first goal of the given research paper is to make a representation that may be used to browse Written numerals, symbols, and words from pictures victimization the Convolution Neural Network notion. the subsequent segments can offer a summary of the connected task, conceptual context, design, procedure, exploratory results, and summary.

2 LITERATURE REVIEW

Handwritten recognition systems were initial developed in the Fifties. The goal of written character recognition is to simulate human reading skills so the pc will edit and skim the add a similar manner that humans do with documents. In recent years, the handwriting recognition technology has been in all the foremost interesting and difficult analysis areas within the domain of picture process and pattern identification. It

considerably contributes to the innovation of computerization processes and upgrades the confluence between human beings and computers in a very variety of applications.

Handwritten character recognition may be a branch of computing that has laptop vision and pattern recognition. A laptop that performs handwriting recognition acquires and detects symbols in hard copies, images, devices with touch sensors on screen, and different origins and converts them into computer understandable type. It is used in optical character recognition yet as a lot of advanced intelligent character recognition systems. The majority of those structures use artificial intelligence mechanisms like neural networks. After each of the symbol's square measure extracted, an identification engine is employed to spot the analogous laptop symbol.

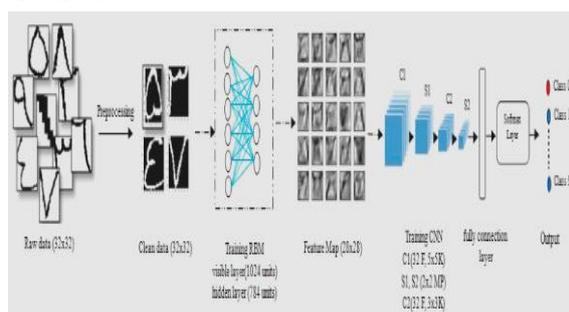


Fig. 2. Handwritten image analysis

There square measure presently many completely different recognition techniques out there. An initial image coaching set is employed to coach neural network recognizers. The character identifications square measure then made by the trained network. every neural network learns the properties that distinguish coaching pictures in a very unique way. It then searches for similar properties within the target image so as to spot it. though neural networks square measure fast to line up, they'll be inaccurate if they learn properties that square measure irrelevant to the target information.

In recent years, written character identification has been one in all the foremost interesting and difficult analysis domains picture process and pattern

identification. It acts as a crucial part in the upgradation of automation processes and improves the communication between man and machine in a very type of applications. many studies have been conducted to develop new techniques and ways for reducing time intervals and up recognition accuracy.

1. Written character recognition offline Offline character recognition is once typewritten or written characters are scanned then reborn into a binary or grayscale image. The binary image is employed within the feature extraction and recognition technique. Offline character recognition may well be a tough task as a result of no temporal order knowledge for character strokes is provided. As a result, offline character recognition is thought to be a a lot of difficult task than its on-line counterpart.

2. Written character recognition on-line on-line written character recognition is additionally referred to as period of time character recognition. Writing and recognition are dispensed at the same time during this case. The user will write a character on any sensory area wherever sensors will observe pen movements, and also the characters are recognised based mostly on these pen movements. as a result of temporal order knowledge is out there, on-line character recognition is much easier than offline character recognition.

This analysis is focused to provide solutions to the subsequent problems:

- Discuss the various methods and approaches employed in Written symbol identification?
- However will the presentation of Written identification structures be refined using artificial (ANN)?

The given research paper is meant for college students and teachers. UN agency wishes to transform written notes and research papers into computerized format. In spite of the exaggerated use of automation in pedagogy establishments, writing by hand remains a vicinity of students' and teachers'

everyday lives. Students take written records while attending lectures and whereas study from numerous origins.

Few folks use the notes to jot down their thoughts, plans, and ideas. Similarly, lecturers might have written records that they desire to speak to students. As a consequence, the aim of this paper is to motivate students and teachers to form a structure that will enable them to transform their written tasks into digital tasks which will be kept and commune digitally.

3 PROJECTED METHODOLOGY

a. General design

This segment considers the look and design of the projected written symbol recognition structure make use of the artificial neural network method. The projected structure includes of input preliminary processing, Convolutional Neural Network, and output segments as given in the below figure.

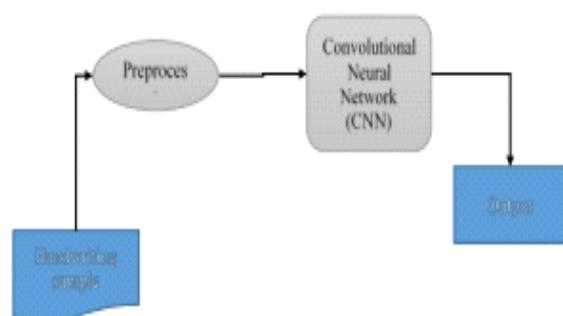


Fig. 3. Pre-processing and CNN

The explanation of the design is provided below:

- **Convolutional Neural Network(CNN)**

The structure uses the Convolutional Neural Network (CNN), a category of deep Artificial Neural Networks(ANN) are basically used for symbol analysis from given images and pictures. The image depicts the associated fundamental design of CNN which will be used in the Optical Character Recognition structure. The mentioned design depicts different kinds of layers, with the primary layer being the first input layer in which we take input and also

the last layer is the layer for giving the output. The second layer after the first layer is termed as the Convolutional layer which is followed by pooling layers and Totally - connected layers. The outline of the CNN design is shown below in the figure:

1) Input layer: The first layer is the input layer which is employed for giving the input structure with the written pictures. The layers are often colored images with RGB values or grayscale. It will have the following dimension $W \times H \times D$, looking on the given input image. The W and H stands for the breadth and height of the given image, whereas D is short form for the Depth of the image.

2) Convolutional layer: The Convolutional layer is the main building and basic constructing block of a Convolutional Neural Networks(CNN). The vast amount of the computing task that is needed to acknowledge symbols from the given input is finished during these layers. The layer is composed of a collection of understandable filters(or Kernels) referred as the frameworks of the Convolutional layers. The size of the filters is usually smaller than the actual image.

3) Pooling layer: Within the CNN style, the Pooling layers have situated in the middle of Convolution layers. They are guilty of step by step shrinking the dimensions of engaged on computations within the network. They facilitate streamlining the calculation's foundation. To do this, they lower the by combining the outputs of the outputs of the neuron teams. They run their own businesses. Therefore, the system is ready to provide the specified results.

4) Totally connected layer: The last stage of CNN uses totally connected neural networks. Simply put, they are a fully connected artificial neural network. The network-related weights are computed throughout the training session. Convolution/pooling output is shipped to a completely connected neural network calculates the highly similar symbols which best describes the depiction when the operation. As a results of this, the

relation between the depiction's details route and its category. The outcomes of network association weights are increased by the convolution/pooling procedure path. Next, associate activation perform is applied to the result.

• Modules Breakdown

HCR are a number of the detailed analysis and vital branches of machine learning and pc observation. We now put forward our working however the letters are categorized from writing in picture layout during indenture. The mentioned design depicts different kinds of layers, and the introduced representation as shown in the image two has four levels for the purpose of categorization and observation: Pre-processing, Segmentation & Chopping, Feature derivation and Classification & Acknowledgement. We now put forward our working however the letters are categorized from writing in picture layout during indenture

Input image: throughout the image acquisition stage, a written input image is non-inheritable. during this instance, the image ought to be during a given format, such PNG or JPEG.

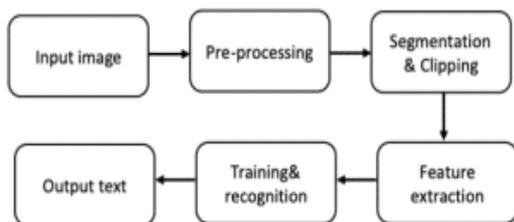


Fig. 4. Proposed System Architecture

b.1. Pre-Processing: Pre-processing gets a picture as input to carry out cleansing work. It efficiently improves the picture by removing any ambiguity. moreover, pictures is also needed to be in binary formats ie in zero one format or in the grayscale that are completed in this level. when the digitized image has undergone pre-processing to scale back noise, it's examined for skewing. Preprocessing is critical to forming knowledge that optical character recognition

systems will simply read. the first pre-processing goal is to eliminate the scene from the realm of intrigue within the picture, boost it, and build a robust distinction among the background and foreground.

b.2. Segmentation: When the given pictures are pre-processed, every symbol are detached employing a segmentation procedure. The given symbols are then kept in a series of pictures. Then borders in every symbol picture are removed if the border is out there. Further, each character is level up to a given dimension. Pre-processing is critical to forming knowledge that optical character recognition systems will simply read. the first pre-processing goal is to eliminate the scene from the realm of intrigue within the picture, boost it, and build a robust distinction among the background and foreground.

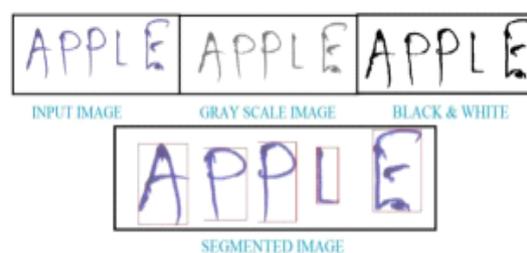


Fig. 5. Segmented Image

b.3 Feature Extraction: During this stage, the image's options are retrieved and outlined in accordance with the character's height, the number of horizontal lines, the character's widths, and alternative characteristics. disk shape, pixel count, placement of assorted elements, and quantity of vertically headed arcs are many examples.

b.4 Classification and Recognition: During this section, the Artificial Neural Network(ANN) is employed for the categorization and the identification of the symbols from the image. The foremost Artificial Neural Networks(ANN) which are then utilized by the Optical Symbol Analysis structures are known to be the Multiplayer Perception (MLP) and the Korhonen's Self Organizing Maps.

Output text: The resultant output of the system gives the desired text of the image.



Fig. 6. Neural Network Processing

• Experimental Results And Discussion

Under Windows seven software package, MATLAB R2015a was used to implement the CNN formula. The implemented computer code was developed employing a 4GB RAM and Intel Core i7-2640 electronic equipment. Table one presents the check findings. The table has 3 columns, the primary of that lists the amount of coaching. The last column displays the perfection earned by the Convolutional Neural Network approach for properly known photos; the column after the first one lists the quantity of trial structures.

The experiment shows that the typical perfection will increase because the variety of coaching pictures grows. This is because a lot of coaching pictures lead to a lot of correct info concerning the coaching parameters, that successively improves classification accuracy throughout the testing section. One verifies the exactitude earned.

Test Results for Handwriting Character Recognition with NIST Dataset.

No. of Training Images	No. of Testing Images	Average Accuracy (%)
200	200	65.32%
300	200	74.43%
500	200	80.84%
600	200	85.21%
800	200	87.65%
1000	200	92.91%

Fig. 7. Test Results for Handwritten Character Recognition

4 OUTCOME ANALYSES

The goal of HCR exploitation Neural Networks is to spot written characters.

employing a neural network, the "Handwritten Character Recognition System" is placed into follow. during this approach, 1st{the initial} image is first converted to grayscale, when that it's metameric and reborn to black and white. The system displays the final result when pre-processing and segmentation operations. Because of the utilization of artificial character recognition and neural networks for character detection, written character recognition systems perform and observe characters way more accurately than the present commonplace approach. The model is capable of detecting symbols of more than twenty four languages.

Fig. 8. Optical Character Recognition

The Neural Network are acknowledged to be capable of providing smart recognition rate at this as compare to other strategies. Because of the utilization of artificial character recognition and neural networks for character detection, written character recognition systems perform and observe characters way more accurately than the present commonplace approach.

Within the field of image process, written character recognition technology offers image definition, pre-processing, segmentation, and recognition capabilities whereas nonetheless maintaining a high level of accuracy

5 CONCLUSION

The ultimate main aim of this study was to form a system that may help and encourage the classification and identification of the handwritten characters and digits. Character and digit identification is very crucial in this digitised world, notably in organisations that subsume written documents that has to be analysed exploitation pc systems. Handwriting classification and recognition systems assist organisations and people in finishing advanced tasks. the present system processed and skim handwriting characters and digits exploitation neural networks. Convolution Neural Networks (CNN) were employed in the system, beside coaching knowledge, to permit for simple recognition of characters and digits.

As a result, supported the coaching knowledge keep within the system's info, it absolutely was straightforward to distinguish and recognise different Handwritten characters and digits. Image Acquisition stage, digitisation, pre - processing , segmentation, feature extraction, classification and recognition were all the different phases of handwriting identification . Unit testing, Integration testing, GUI testing, and Validation checking are all various type of testing that are needed for testing the system. The ultimate system reached the desired detailed correctness and precision, identification and acceptance needs. The present study's findings are often applied to character recognition in alternative languages.

6 REFERENCES

- [1] Wilkinson, T., & Brun, A. (2016). Semantic and verbatim word spotting using deep neural networks. In 2016 15th International Conference on Frontiers in Handwriting Recognition (pp. 307-312). IEEE. Williams, A. Convolutional Neural Networks in Python: Introduction to Convolutional Neural Networks. Scotts Valley, CA: Createspace Independent Publishing Platform.
- [2] Wu, C., Fan, W., He, Y., Sun, J., & Naoi, S. (2014). Handwritten character recognition by alternately trained relaxation convolutional neural network. In 2014 14th International Conference on Frontiers in Handwriting Recognition (pp. 291-296). IEEE.
- [3] Marinai, S. (2008). Machine Learning in Document Analysis and Recognition. Berlin, Germany: Springer Science & Business Media. [24] Nair, R. R., Sankaran, N., Kota, B. U., Tulyakov, S., Setlur, S., & Govindaraju, V. (2018, April). Knowledge transfer using Neural network based approach for handwritten text recognition. In 2018 13th IAPR International Workshop on Document Analysis Systems (pp. 441-446). IEEE.
- [4] Pal, A., & Singh, D. (2010). Handwritten English character recognition using neural network (NN). International Journal of Computer Science and Communication, (IJCS) 1(2), 141-144.
- [5] Perez, C. (2019). Machine Learning and Neural Networks (ML & NN). Examples with MATLAB. Chicago: Independently Published.
- [6] Rautaray, S. S., & Agrawal, A. (2015). Vision based hand gesture recognition for human computer interaction: a survey. Artificial intelligence review, 43(1)
- [7] F.S. Hardwick; Gioulis, A.; R.N. Akram; K. Markantonakis. Block-Chain-based electronic voting: a method for electronic voting with voter privacy and decentralization. In the Proceedings of the 2018 IEEE International Conference on the Internet of Things, Green Computing and Communications, Cyber , and IEEE Smart Data, which took place in Halifax, Newfoundland, Canada, from July 30 to August 3, 2018.
- [8] Follow My Vote, Vote, F.M. The Future's Secure Mobile Voting Platform-

20. Online availability [https:// follow-myvote.com/viewed](https://follow-myvote.com/viewed) on July 26, 2021).
- [9] O.O., Okediran; AA. Sijuade; Wahab, Secure Electronic Democratic Utilizing a Half and half Cryptosystem and Steganography. *J. Adv. Math. Comput. Sci.* 2019, 34, 1–26.
- [10] U. Jafar; A State of the Art Survey and Future Directions for Block-Chain-Based Electronic Voting Systems, by M.J.A. Aziz. In the Proceedings of the 2020 International Conference on Advances in Cyber Security, Penang, Malaysia, December 8 to 9,
- [11] G.G. Dagher; PB, Marella; M. Milojkovic; Mohler, J. Broncovote: Using Ethereum's Block-Chain, a safe voting system. From January 22-24, 2018, Funchal, Portugal, hosted fourth international conference on information systems security & privacy.
- [12] T.U., Sree; N. Yerukala; A.N. Tentu; A secret sharing scheme employing identity-based signatures by A.A. Rao Published in the Proceedings of the 2019 IEEE International Conference on Electrical, Computer, and Communication Technologies, which took place in Tamil Nadu, India, from February 20-22, 2019.