

# Signature Forgery Detection Using Convolutional Neural Network

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**Abstract** - Each person's signature may be distinctive. Signatures, on the other hand, provide a number of difficulties because two signatures created by the same individual may appear to be extremely identical. Even when two signatures are signed by the same person, several features of the signature can differ. A Convolutional Neural Network (CNN) based solution is proposed in this paper in which the model is trained on a dataset of signatures and predictions are produced as to whether a given signatures is real or forged.

**Key Words:** Convolution Neural Network, Handwritten signature, Dataset, Image Preprocessing, Data Augmentation.

## 1. INTRODUCTION

A handwritten signature is a scripted name or legal mark made by hand with the intention of permanently authenticating the writing. Because signatures are created by moving a pen across a piece of paper, movement is possibly the most crucial feature of a signature. Signature verification is critical because, unlike passwords, signatures cannot be changed or forgotten because they are unique to each individual, and thus is regarded as the most significant way of verification. Signature verification techniques and systems are separated into offline signature and online signature methods. Although small-scale data studies have received a lot of attention in recent years, most deep learning approaches still require a significant number of samples to train their system. To put it another way, most studies still require several (multiple) signature samples to complete their training process. It offers an off-line handwritten signature verification approach based on Convolutional neural networks in this work (CNN).

## 2. EXISTING SYSTEM

The existing technology makes use of digital signatures, generating one for each column and embedding it in the least significant bits of selected pixels in each associated column. The message digest 5 technique is used to generate digital signatures, and the signature is embedded in the allocated pixels using the four least significant bits replacement process. The digital signature's embedding in the targeted pixel is absolutely harmless and undetectable to the human visual system. The suggested forgery detection technique has shown promising results against a variety of forgeries put into digital photos, successfully detecting and pointing out fabricated columns.

## 3. PROPOSED SYSTEM

The handwritten signature is a behavioural biometric that is based on changing behaviour rather than any physiological aspects of the individual signature. Because a person's signature changes over time, verification and authentication are necessary. It may take a long time for the signature to be authenticated because of the flaws that must be corrected. Higher signature irregularity might sometimes contribute to a higher rate of false applications.

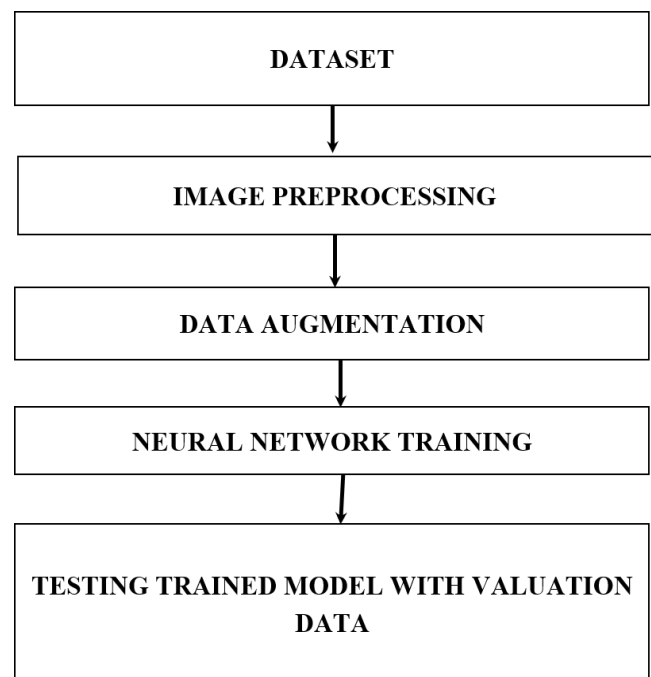


Fig 1: Flow Chart

**DATASET:** From the training phase to evaluating the performance of recognition algorithms, proper datasets are expected at all stages of object recognition research. All of the photos used in the collection were found on the internet and were found using a name search on a variety of languages' sources.

**IMAGE PREPROCESSING:** Images downloaded from the internet come in a range of formats, sizes, and quality levels. Final photos that would be used as a dataset for a deep neural network classifier were preprocessed to increase feature consistency extraction.

**DATA AUGMENTATION:** The primary goal of augmentation is to expand the dataset and introduce some distortion to the images in order to reduce over fitting during the training. Image data augmentation is a method of increasing the size of a training dataset artificially. Photographs in the dataset are being changed.

**NEURAL NETWORK TRAINING:** The goal of network training is to teach the neural network the properties that distinguish one class from the others. As a result of the increased use of augmented photographs, the networks' odds of learning the required traits have increased.

**TESTING TRAINED MODEL WITH VALUATION DATA:**

Finally, by processing the input photos in the valuation dataset, the trained network is utilized to recognize the class of given images and the results are processed.

**4. RESULT**

The proposed method effectively performed offline signature verification with increased efficiency and accuracy, as well as detecting skilled forgeries very easily. The usage of Python and its libraries, as well as a solution based on Signature forging was successfully detected using a Convolutional Neural Network (CNN). The training and testing results demonstrated that the large mass of datasets was a serious challenge. Signature authentication is simply demonstrated that as the number of datasets increases, the proportion of testing accuracy has also increased.

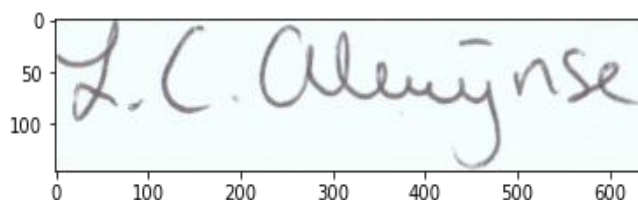


Fig 2: Original Signature

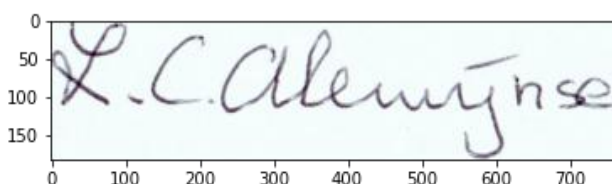


Fig 3: Forged Signature

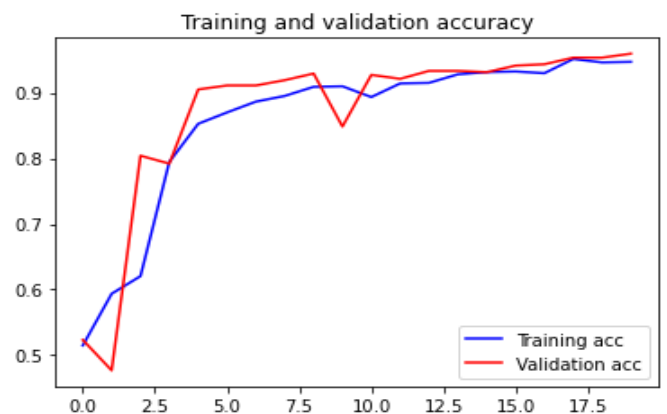


Fig 4: Training and Validation Accuracy

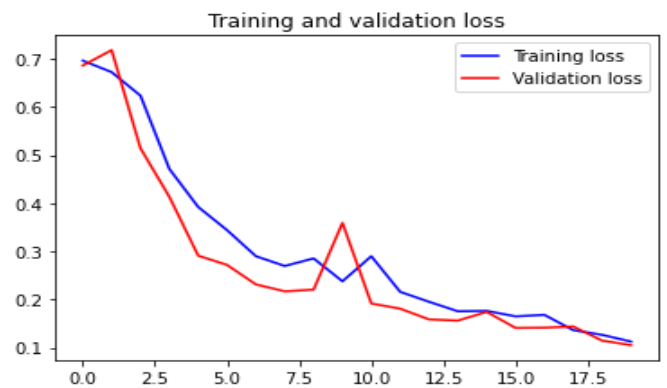


Fig 5: Training and Validation loss

Split Ratio	Training Accuracy	Validation Accuracy
6:4	98.4	95.15
7:3	98.43	95.96
8:2	97.39	95.24

**5. CONCLUSION**

Handwritten signatures are necessarily to be verified and validated in both social and legal situations. Only the intended individual's signature can be accepted. Two signatures from the same person are exceedingly unlikely to be identical. Even when two signatures are signed by the same person, the signatures might differ in various ways. With the growing digitalization of various aspects of everyday life, as well as new challenges in workplaces and agencies, effective user verification methods are essential. There is a definite need for new and improved procedures and algorithms to go along with new technology that opens up new possibilities. So, a system that can learn from signatures and predict whether the signature in issue is a fraud or not has been successfully created.

**REFERENCE**

- [1] Sarvesh Tanwar, Sumit Badotra, Medicine Gupta, Ajay Rana, "Efficient and secure multiple digital signature to prevent forgery based on ECC", International Journal of Applied Science and Engineering, vol 18, no.5, 2021.
- [2] Nura Musa Tahir, Adam N Ausat, Usam Bature, Kamal Abubakar, "Handwritten signature verification system: Using Artificial Neural Networks Approach", International Journal of Intelligent Systems and Applications (ISA), 2021.
- [3] Jivesh Poddar, Vinanti Parikh, Santhosh Kumar Bharti, "Offline Signature Recognition and Forgery Detection using Deep Learning", International Conference on Emerging Data and Industry(EDI), 2020.
- [4] Prarthana Parmer, Jahni Mehta, Saakshi Sharma, Krupa Patel, Parth Singh, "A Survey of Handwritten Signature Verification System Methodologies", JETIR Volume 6, May 2019.
- [5] Gopichand G, Shailaja G, Venkata Vinod Kumar, T. Samatha, "Digital Signature verification Using Artificial Neural Networks", International Journal of Recent Technology and engineering(IJRTE) Volume-7 Issue-5S2, January 2019.
- [6] K. Tamilarasi, Dr.S. Nithya Kalyani, D. Abirami, R. Sakthi Rekha and T. Dhanalakshmi, "Offline Signature Verification Method based on Matlab Representation", Journal on Science Engineering and Technology (JSET) Volume 5, No. 02, April 2018.
- [7] Kaushik Ravi, S.S.Shylaja, and Nypunya Devraj, "A new approach to detect paste forgeries in an image", International Conference on Image Information Processing (ICIIP), IEEE 2017.
- [8] Nasir N. Hurrah, Shabir A. Parah, Javaid A. Sheikh, "INDFORG: Industrial Forgery Detection Using Automatic Rotation Angle Detection and Correction", IEEE vol 17, issue 5, 2017.