



Face Identificationusing Computer Vision and CNN Algorithm

Uppalapati Vasavisukanya, Dr. T. Prabhakara Rao

Department of Computer Science and Engineering, Godavari Institute of Engineering and Technology (A), JNTUK, Kakinada.

To Cite this Article

Uppalapati Vasavisukanya and Dr. T. Prabhakara Rao. Face Identificationusing Computer Vision and CNN Algorithm. International Journal for Modern Trends in Science and Technology 2022, 8(S03), pp. 162-166.<https://doi.org/10.46501/IJMTST08S0337>

Article Info

Received: 26 April 2022; Accepted: 24 May 2022; Published: 30 May 2022.

ABSTRACT

The human face is an important function for identifying a person. Even in the case of twins, everyone has their own face. Therefore, face recognition and identification are necessary to distinguish them from each other. Therefore, we propose a method that can recognize the CNN model based on deep learning. Here we collect datasets from various aspects. After preprocessing, the data is trained using the CNN algorithm. After training, you can also test the results in Open CV and upload images for face recognition.

Keywords: CNN model, openCV, deep learning, facial recognition, image recognition

1.INTRODUCTION

It's critical to be mindful of the people and systems with whom you're interacting when dealing with intelligent systems. Identity documents and keys used for centuries must be replaced in the modern day. They are readily misconstrued or fabricated. Facial recognition, fingerprinting, and other forms of personal identification can all be used to track down a specific person. The convolution layer and the CNN's convolution layer share a set of parameters. Consider both the volume of knowledge that must be memorised and the number of aspects that must be taught. As a result, the performance of the algorithm has been improved [1]. Pre-processing or feature extraction is required by other machine learning techniques. These procedures, however, are redundant in the context of image processing. Many other machine learning methods are incapable of accomplishing this. Deep learning presents its own set of difficulties. This method requires a

large amount of data to develop a depth model, limiting its utility. In view of recent advances in the disciplines of face and licence plate recognition, this article will look at the fundamentals of CNN-based face recognition technology [2].

A type of neural network is a convolutional network of neurons, which is a convolutional network of neurons. Because of the rising success of convolutional neural networks in a variety of competitions, the area of study has moved its focus to this type of network [3]. Lowering the amount of learning parameters that are utilised in the forward BP algorithm can improve the training performance by a factor of two or three. Achieving the desired effect can be accomplished through the use of convolutionally condensed neural networks. Pre-processing of the input data is kept to a bare minimum using the suggested convolutional neural network design because it is so efficient. There are three layers in a convolutional neural network, starting with

the initial input layer, followed by each successive layer, and finally the following level above. By utilising this strategy, it is feasible to produce translation and rotation, to name only a couple of readily obvious effects [5].

1.1. Research objective

1. The primary goal of this project is to use computer vision and deep learning algorithms to detect faces.
2. 1. Recognition of faces using a camera and neural network analysis is the first step.

2. RELATED WORK

Yao L S, Xu G M, Zhap F. (2020). As per his analysis Facial expression recognition utilising CNN local feature fusion was the subject of this effort. Deep neural networks (DNNs) are increasingly being used to learn discriminative representations for automatic facial expression detection in light of recent success of deep learning approaches in several disciplines. A number of contemporary deep FER systems suffer from over fitting and expression-unrelated aspects like illumination, head posture, and identification bias. A deep FER system pipeline is going to be covered in detail, with background information on the system as well as valuable applications at each stage. Analysis of FER deep learning networks, as well as training techniques for image sequences in motion, is conducted. Aside from summarising well-known indicators, data in this section summarises the competitors' performance. Survey subjects and application situations are broadened in the new version. After evaluating the remaining challenges and opportunities in this subject, we'll talk about the best ways to build robust deep FER systems in the future. **Zhang Chen. et al., (2019).** Find out about Micro expression recognition research is continuing. Facial expression recognition can be used in a variety of sectors, including mental health and social/physiological interactions. Due to advances in recent years in hardware and sensor technology, current FER systems are capable of handling more realistic application settings than only those seen in the lab. For real-world conditions, however, this technology's accuracy is only about half as high as it is in the lab. Real-world issues including lighting variance, head orientation, and subject dependency are addressed in this survey, which may not be overcome just by using the FER system. Non-visual depth sensors are another sort of non-visual sensor that offers

additional information, such as in lighting fluctuation and position shifting situations, for example. Video or face image analysis systems can then make use of this. It is necessary to test the theoretical viability and practicality of our new expression recognition system before moving further with its development in the actual world. **XuLinlin, Zhang Shumei, Zhao Junli. (2019).** His research study of conversation was an expression recognition algorithm for concurrent convolutional neural networks. Facial expression identification is a hot topic in the field of computer vision. According to study, facial expressions communicate 55% of the information intended during nonverbal communication. Expression recognition has recently been used by the medical and advertising industries for a host of new applications. In this study, researchers used a parallel CNN framework to identify facial expressions on frontal faces. CNN focuses on two sections of the face: the eyes and the mouth. The features of the parallel models will be concatenated to generate the final feature vector. When compared to models that use the entire face as their input, this strategy provides better outcomes. In addition to AlexNet and VGG16, other CNN structures have been compared to ours to determine how we measure up. **Li Siquan, Zhang Xuanxiong. (2018).** Worked on Face recognition using convolutional neural networks is a prominent research subject at the moment. There is still a difficulty because of the wide difference in facial expression data between classes, despite recent breakthroughs in facial expression recognition. An image or video classifier is employed after features like SIFT, HOG, and LBP in the standard approach to this problem. Deep learning algorithms have been used in a recent study to provide an end-to-end framework for facial emotion recognition. In spite of the improved results, there is still a great deal of opportunity for development. To train a deep learning model, an attentional convolutional network (ACN) is used, which results in a considerable improvement over earlier models on a wide range of datasets. The crucial facial regions for each mood can be visualised using a technology we developed specifically for this purpose. Certain regions of the face appear to influence emotional responses, according to research. **ZhaiJunkui, Liu Jian. (2018).** A transfer convolutional neural network is now being investigated to determine if it can distinguish face expressions. As artificial intelligence and human-computer interfaces have improved, facial

expression recognition and analysis has become increasingly significant. There are still many unanswered questions about face expressions because of their intricacy. Consequently, it is difficult to discern face expressions. In most publications, the entire face is used as input data. Many features of the face, such as the eyes and lips, have minimal influence on a person's current emotional state. Our method was put to the test using four well-known public facial expression databases. The findings were positive. Most of the previously available approaches were outperformed by a brand-new strategy.

3. METHODOLOGY AND ALGORITHMS

The CNN model based on deep learning will be used to recognize faces in our planned study, which we believe will be successful [4 & 6]. The purpose of this project is to compile a collection of people's portraits. In order to train the model, we use the CNN approach once the data has been cleaned and pre-processed. By using OpenCV and uploading a photo for facial recognition, we can see how well our findings perform [7].

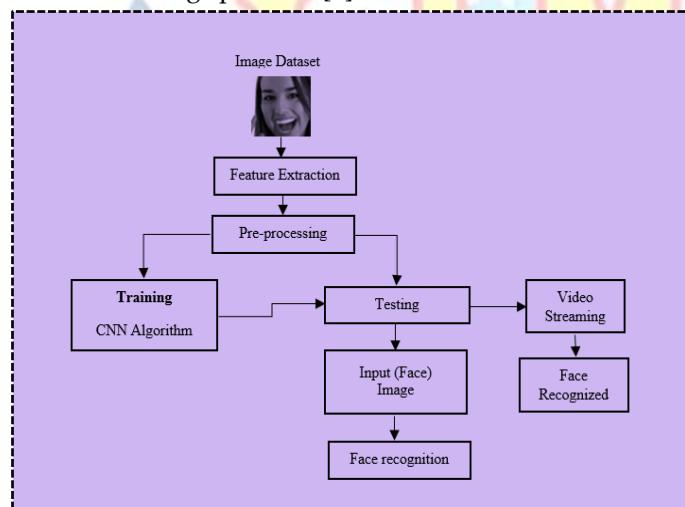


Figure.1. Block Diagram

There is a wrap-up at the end of this section, which summarizes the themes we've covered. Consider checking out the Soft Axe and Cross-Entropy tutorials if you think they'll be of use to you (and it probably will). It is not necessary to have prior understanding of these areas to succeed in this course, but it will be helpful while working with Convolutional Neural Networks.

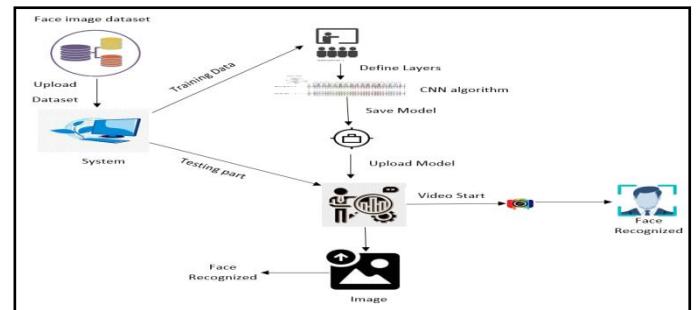


Figure.2. CNN model and system architecture

3.1. Algorithm setup of CNN

Step.1. convolutional operation

As part of our overall attack approach, we'll experiment with convolution to see how it works out for us. It is now time to talk about neural network filters, which are also known as feature detectors in some circles. Several issues must be taken into account while creating feature maps, including the properties of the maps themselves, detection layers that are used, and the presentation of discoveries [8 & 9].

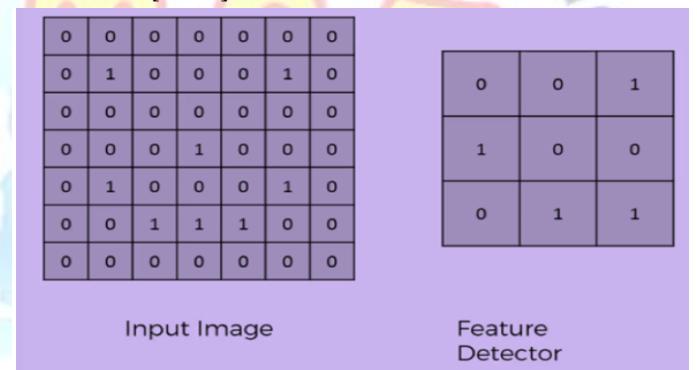


Figure.3. Convolution operation model

Step.1.(b). Relu Layer

The Rectified Linear Unit, often known as the ReLU, will be employed in the second half of this stage. In this session, we will look at Convolutional Neural Networks and linearity in more detail. Although you may not require a comprehensive understanding of CNN, it is never a bad idea to brush up on your knowledge [10].

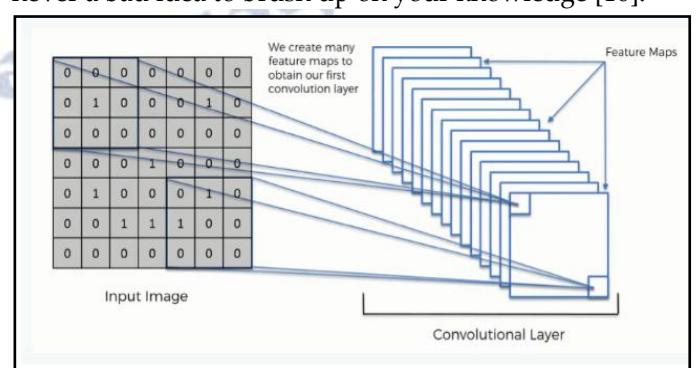


Figure.4. Convolutional model

Step.2. Pooling Layer

During this section, you will learn about the notion of pooling and how it works in more detail. In this particular instance, we're primarily interested with increasing the efficiency of the pooling process [9 & 11]. We'll talk about the concept of mean pooling in this session (or total pooling). With the help of an interactive visual presentation tool, we'll summaries the most important ideas and explain how the concept is put into action in practice.

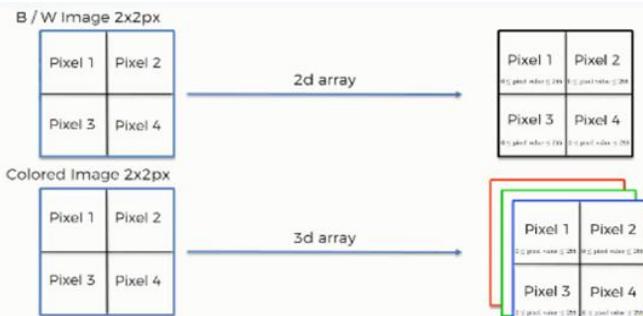


Figure.5. Convolutional neural networks are used to scan images

Step.3. Flattening

With Convolutional Neural Networks, this article will demonstrate how to transition from pooled to flattened layers when working with the network structure [12].

Step.4. Full Connection

In this section, we'll synthesize everything we've studied thus far. You can better understand Convolutional Neural Networks and the "neurons" that are formed as a result if you have this information [14].

4. RESULT AND DISCUSSION

4.1. OUTPUT SCREEN SHOTS WITH DESCRIPTION

Home: Face-recognition challenge



Figure.6. Face recognition by image

Using a webcam to recognise faces

1. In order to avoid any confusion, please read this before posting any comments or questions.
2. After reading the blog post, do not attempt to run the code! You must be familiar with the code's functionality in order to run and troubleshoot it.
3. OpenCV v2 is recommended.
4. You'll need a functional webcam for this script to work [13].
5. Your questions may have already been addressed in the other comments and queries.



Figure.6. Face recognition by image using webcam

4.2. TEST CASES

Table.1. Face recognition test result

Input	Output	Result
Input image	Model recognize the face successfully	Success

Table.2. Building a test case model

S.N O	Test cases	I/O	Normal O/T	Real O/T	P/F
1	Read the dataset.	Dataset path.	Dataset need to read successfully.	Dataset fetched successfully.	P
2	Performing pre-processing on the dataset	Pre-processing part takes place	Pre-processing should be performed on dataset	Pre-processing successfully completed.	P
3	Model Building	Model Building for the clean data	Need to create model using required algorithms	Model Created Successfully.	P
6	Face recognition	Video streaming Or image upload	Face recognized successfully	Model recognize the face image successfully	P

5. CONCLUSION

A result of this research, it became possible to develop face-recognition technology. As a result of our research, we were able to build image- and video-based techniques as a result of the CNN algorithm. Face-recognition algorithms were trained on this dataset before being put to the test on real-world pictures and videos using the results [15].

Face recognition systems, including architectures, strategies, algorithms, methodologies, databases for training and testing pictures, and performance measurements for face recognition systems, are all explored in depth in this study. The fact that different studies have been undertaken in an attempt to tackle the challenges associated with the aforementioned technique, as well as some advantages and limits, should not be overlooked, it is necessary to emphasise.

5.1. FUTURE SCOPES

Our model will be more accurate and predictive if we have a larger amount of data to work with. This method can improve a computer's ability to recognise facial emotions by increasing the amount of data it receives [8]. In the future, face recognition algorithms will make heavy use of the many hidden layers of hybrid Wavelet-CNN technology that have been discovered so far. It is possible to employ a deep learning algorithm or a Neuro system to select the optimal CNN algorithm parameter values for the greatest face detection system performance, in a manner similar to how people reason about their decisions [9 & 11]. While Neuro systems are not perfect, they are a great alternative for addressing the interpretability and accuracy requirements of system modelling.

Conflict of interest statement

Authors declare that they do not have any conflict of interest.

REFERENCES

- [1] Yao L S, Xu G M, Zhap F. Facial Expression Recognition Based on CNN Local Feature Fusion. *Laser and Optoelectronics Progress*, 2020, 57(03): 032501.
- [2] ZouJiancheng, Deng Hao. An automatic facial expression recognition method based on convolutional neural network. *Journal of North China University of Technology*, 2019, 31 (5): 51-56.
- [3] Liu Sijia, Chen Zhikun, Wang Fubin, et al. Multi-angle face recognition based on convolutional neural network. *Journal of North China University of Technology (Natural Science Edition)*, 2019, 41 (4): 103-108.
- [4] Li Huihui. Research on facial expression recognition based on cognitive machine learning. Guangzhou: South China University of Technology, 2019.
- [5] Zhang Chen. Research on some key technologies of facial micro-expression recognition. 2019.
- [6] Li Yong, Lin Xiaozhu, Jiang Mengying. Facial expression recognition based on cross-connection LeNet-5 network. *Journal of Automation*, 2018, 44 (1): 176-182.
- [7] Li Siquan, Zhang Xuanxiong. Research on Facial Expression Recognition Based on Convolutional Neural Networks. *Journal of Software*, 2018, v.17; No.183 (01): 32-35.
- [8] HouYuqingyang, QuanJicheng, Wang Hongwei. Overview of the development of deep learning. *Ship Electronic Engineering*, 2017, 4: 5-9.
- [9] Xue Liang, Huang Meichuan. Emotional analysis of lyrics in Chinese popular music Big data analysis method based on new media music terminal. *Music Culture Industry*, 2017, (4): 77-81.
- [10] Xie S, Hu H. Facial expression recognition with FRR-CNN. *Electronics Letters*, 2017, 53 (4): 235-237.
- [11] Mao Xu, Wei Cheng, QianZhao et al. Facial expression recognition based on transfer learning from deep convolutional networks. 2015 11th Int. Conf. Nat. Comput. 2015: pp.702 - 708
- [12] Liu Jianwei, Liu Yuan, LuoXionglin. Progress in Deep Learning Research. *Application Research of Computers*, 2014, 31 (7): 1921-1942.
- [13] ShenHuijun. Research and implementation of face recognition image preprocessing method. *Science and Technology and Innovation*, 2014 (18): 119-120.
- [14] GuoYanhong. Research on collaborative filtering algorithm and application of recommendation system. Dalian: Dalian University of Technology, 2008: 1-41.
- [15] Liu Mingqi, Ni Guoqiang, Chen Xiaomei. Research on Pretreatment Algorithm of Dorsal Vein Image. *Optics Technology*, 2007, 33: 255-256.
- [16] Parvathi, D. S. L., Leelavathi, N., Ravikumar, J. M. S. V., & Sujatha, B. (2020, July) Emotion Analysis Using Deep Learning. In 2020 International Conference on Electronics and Sustainable Communication Systems (ICESC) (pp. 593-598). IEEE.
- [17] Kumar, J. R., Sujatha, B., & Leelavathi, N. (2021, February). Automatic Vehicle Number Plate Recognition System Using Machine Learning. In IOP Conference Series: Materials Science and Engineering (Vol. 1074, No. 1, p. 012012). IOP Publishing."