



## FACIAL EMOTIONS RECOGNITION USING CNN

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### Abstract

Nowadays, deep learning techniques know a big success in various fields including computer vision. Indeed, a convolutional neural networks (CNN) model can be trained to analyze images and identify face emotion. In this paper, we create a system that recognizes students' emotions from their faces. Our system consists of three phases: face detection using Haar Cascades, normalization and emotion recognition using CNN on FER 2013 database with seven type of expressions. Obtained results show that face emotion recognition is feasible consequently, it can help to modify their presentation according to the emotions.

**Keywords:**—facial expression, Emotion recognition, Convolutional neural networks (CNN), Deep learning, Intelligent system

### 1. INTRODUCTION

Facial expression recognition identifies emotion from face image, it is a manifestation of the activity and personality of a human. In the 20th century, the American psychologists Ekman and Friesen [2] defined six basic emotions (anger, fear, disgust, sadness, surprise and happiness), which are the same across cultures. Facial expression recognition has brought much attention in the past years due to its impact in clinical practice, social robotics and education. The purpose of this article is to implement emotion recognition in education by realizing an automatic system that analyzes students' facial expressions based on Convolutional Neural Network (CNN), which is a deep learning algorithm that are widely used in image classification. It consists of a multi-stage image processing to extract feature representations. Our system includes three phases: face detection, normalization and emotion recognition that should be one of these seven emotions: neutral, anger, fear, sadness, happiness, surprise and disgust.

### 2. RELATED WORK

Many researchers are interested in improving the learning environment with Face Emotion Recognition (FER). Tang et al. proposed a system which is able to analyze facial expressions in order to evaluate classroom teaching effect. The system is composed of five phases: data acquisition, face detection, face recognition, facial expression recognition and post-processing. The approach uses K-nearest neighbor (KNN) for classification and Uniform Local Gabor Binary Pattern Histogram Sequence (ULGBPHS) for pattern analysis. Savva et al. proposed a web application that performs an analysis of students' emotion who participating in active face-to-face

classroom instruction. The application uses webcams that are installed in classrooms to collect live recordings, then they applied machine learning algorithm on its. In Whitehill et al. proposed an approach that recognizes engagement from students' facial expressions. The approach uses Gabor features and SVM algorithm to identify engagement as students interacted with cognitive skills training software.

### 3. PROPOSED FRAMEWORK

In this section, we describe our proposed system to analyze students' facial expressions using a Convolutional Neural Network (CNN) architecture. First, the system detects the face from input image and these detected faces are cropped and normalized to a size of 48x48. Then, these face images are used as input to CNN. Finally, the output is the facial expression recognition results (anger, happiness, sadness, disgust, surprise or neutral).

#### METHODOLOGY:

A Convolutional Neural Network (CNN) is a deep artificial neural network that can identify visual patterns from input image with minimal pre-processing compared to other image classification algorithms. This means that the network learns the filters that in traditional algorithms were hand-engineered [19]. The important unit inside a CNN layer is a neuron. They are connected together, in order that the output of neurons at a layer becomes the input of neurons at the next layer. In order to compute the partial derivatives of the cost function the backpropagation algorithm is used. The term convolution refers to the use of a filter or kernel on the input image to produce a feature map. In fact, CNN model contains 3 types of layers as shown in Figure 2:



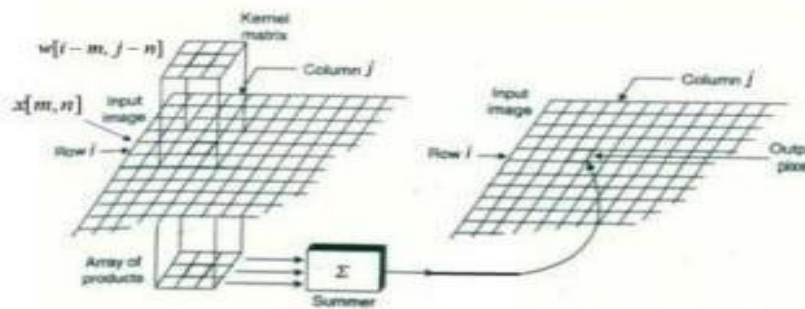
Fig. CNN architecture.

**Convolution Layer:** is the first layer to extract features from an input image. The primary purpose of Convolution in case of a ConvNet is to extract features from the input image.

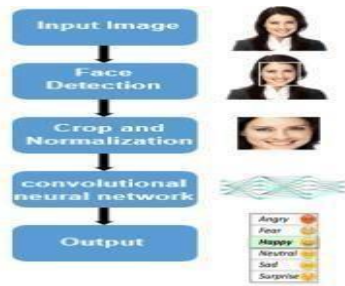
Convolution preserves the spatial relationship between pixels by learning image features using small squares of input data. It performs a dot product between two matrices, where one is the image and the other is a kernel. The convolution formula is represented in Equation 1:

$$net(t, f) = (x * w)[t, f] = \sum_m \sum_n x[m, n] w[t - m, f - n] \quad (1)$$

Where  $net(t, f)$  is the output in the next layer,  $x$  is the input image,  $w$  is the filter matrix and  $*$  is the convolution operation. Figure 3, shows how the convolution works.



**STRUCTUREOFFER:**



**CNNIMPLEMENTATION**

We used OpenCV library [16] to capture live frames from webcam and to detect students' faces based on Haar Cascades method [14] as shown in Figure 8. Haar Cascades uses the Adaboost learning algorithm invented by Freund et al. [15], who won the 2003 Gödel Prize for their work. The Adaboost learning algorithm chose a few number of significant features from a large set in order to provide effective result of classifiers. We built a Convolutional Neural Network model using TensorFlow [18] Keras [17] high level API.

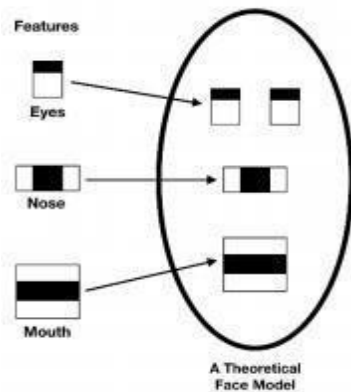


FIG: Face Detection using Haar Cascades

**4. EXPERIMENTAL RESULTS**

We trained our Convolutional Neural Network model using FER 2013 database which includes seven emotions (happiness, anger, sadness, disgust, neutral, fear and surprise). The detected face images are resized to 48x48 pixels, and converted to grayscale images then were used for inputs to the CNN model. Thus, 9 youthful master's students from our faculty participated in the experiment, among them there were two wearing glasses. The Figure 11 shows the emotions' results of 9 students. The predicted emotion label are represented with red text, and the red bar represents the probability of the emotion.

We achieved an accuracy rate of 70% at the the 106 epochs. To evaluate the efficiency and the quality of our proposed method we calculated confusion matrix, precision, recall and F1-score as shown in Figure 12 and in Figure 13, respectively. Our model is very good for predicting happy and surprised faces. However it predicts quite poorly feared faces because it confuses them with sad faces. In Keras, we used ImageDataGenerator class to perform image augmentation as shown in Figure 9. This class allowed us to transform the training images by rotation, shifts, shear, zoom and flip. The configuration used is :rotation\_range=10,width\_shift\_range=0.1, zoom\_range=0.1,height\_shift\_range=0.1 and horizontal\_flip=True.



*Original    Transformed*  
Fig. Image augmentation using Keras.

## 5. CONCLUSION AND FUTURE WORK

In this paper, we presented a Convolutional Neural Network model for students' facial expression recognition. The proposed model includes 4 convolutional layers, 4 maxpooling and 2 fully connected layers. The system recognizes faces from students' input images using Haar-like detector and classifies them into seven facial expressions: surprise, fear, disgust, sad, happy, angry and neutral. The proposed model achieved an accuracy rate of 70% on FER 2013 database. Our facial expression recognition system can help to recognize comprehension towards his presentation. Thus, in our future work we will focus on applying Convolutional Neural Network model on 3D students' face image in order to extract their emotions.

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