Facial Emotion Recognition Using CNN

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Abstract. Now a day technology grow rapidly and it affect life of people in many area one of these emerging technology is Artificial intelligence. Our project is entitled on facial emotion recognition (FER) using CNN. Now a day frauds become increase every where. Deep learning handle these issue by processing big data. So the main Goal of our project set out to recognise emotion of human being using convolutional neural network. The data used in these project is from one of biggest public dataset repository Kaggle. First we preprocess data and avoiding data imbalance by appling data augmentation. Secondly we build four layer Convolutional neural network architecture with two fully connected layer. We used ReLU activation function for each layer and finally we apply Softmax algorithms on the last layer of the model. Despite the good performance of these project, there still seems to be great room for improvement. https://github.com/R-hab/cvproject.git

Keywords: Deep Learning, Convolutional Neural Network, FER

Introduction

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Our life world is drastically influenced by technology [1]. Technology has advanced into many areas of our lives, making information more accessible, improving communication, fraud detection, authentication. Deep learning is key solution for every walk of our life challenges. Facial expressions are one of the main information channels in interpersonal communication. Therefore, it is natural that facial emotion research has gained a lot of attention over the past decade with applications in perceptual and cognitive sciences.

Over the last few years, there is an increase in the desire to use deep learning tools to understand human behavior patterns. One of the most trendy challenges in this field is facial expression recognition. Facial emotions are important factors in human communication that help to understand the intentions of others. In general, people infer the emotional state of other people, such as happyness, sadness and anger, using facial expressions.

The main goal of these project is to recognise human facial emotion using convolutional neural network, we used fer2013 public dataset from kaggle repository, we concern on preprocess the data for better performance. But these dataset still needs bit more preprocessing, because the dataset set contain noisy image and also occculusion usually(hand) as well as glass, beside these it contain 043 imbalanced data which affect model performance.

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2 Related works

Earlier works on emotion recognition, rely on the traditional two-step machine out learning approach, where in the first step, some features are extracted from the one images, and in the second step, a classifier (such as neural network, SVM, or one random forest) is used to detect the emotions. Some of the popular hand-crafted one features are used for facial expression recognition. Then, a classifier would assign out the best emotion to the image. These approaches seemed to work fine on simpler 052 datasets. but with the advent of more challenging datasets (which havemore intra-class variation), they started to show their limitation.

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Despite the notable success of traditional facial recognition methods through 055 the extracted of handcrafted features, over the past decade researchers have directed to the deep learning approach due to its high automatic recognition capacity. In this context, we will present some recent studies in FER, which show proposed methods of deep learning in order to obtain better detection. Train 050 and test on several static or sequential databases.

Nowadays Researchers are working to solve complex problems using most progressive and rapidly growing computer vision applications. First challenge in facial emotion recognition of facial expressions, held during the IEEE conference on Face and Gesture Recognition 2011[2].

Facial communication studies have been carried out for years. But for any experiment, there was still room for progress. That is why this topic is convenient. The key objective of the researchers is to enhance the precision of a basic data collection FER2013.

Lopes et al. Studied the impact of data pre-processing before the training the network in order to have a better emotion classification. Data augmentation, rotation correction, cropping, down sampling with 32x32 pixels and intensity normalisation are the steps that were applied before CNN, which consist of two convolution-pooling layers ending with two fully connected with 256 and 7 neurons. The best weight gained at the training stage are used at the test stage. This experience was evaluated in three accessible databases: CK+, JAFFE, BU-075 3DFE. Researchers shows that combining all of these pre-processing steps is 076 more effective than applying them separately[3].

Ekman et al. defined seven basic emotions, irrespective of culture in which a human grows with the seven expressions anger, feared, happy, sad, contempt, disgust, and surprise[4]. Most of the existing approaches are based on engineered features (e.g. HOG, LBPH, and Gabor) where the classifier's hyperparameters are tuned to give best recognition accuracy's across a single database, or a small collection of similar databases[5].

Facial emotion recognition have application on many areas like. Robotics. Medicine, Deriving assistance system, Lie detection and so on, Recently it develops in area of neuroscience and cognitive science, further more it attract researchers attention in computer vision. And day to day researcher gets great accuracy using deep learning.

The efficiency of CNN-based image recognition methods has also been continuously improved and has gradually replaced the traditional facial expression 089

image recognition methods. In the 2012 Image Net Object-Oriented Recognition opportunity Challenge, some scholars used the CNN model Alex Net to win the championship. opposite the 2014 ILSVRC competition, the CNN model Google Net architecture won opposite place in the classification[6].

Recently, researchers have made extraordinary accomplishment in facial expression detection[7], which led to improvements in neuroscience and cognitive science that drive the advancement of research, in the field of facial expression.

All of the above works achieve significant improvements over the traditional $_{098}$ works on emotion recognition, but they are using the raw data as the input. In $_{099}$ this work, we try to modify the data, such as we only use the vital data from $_{100}$ the image.

Generally, As we observe from many researcher in computer vision, they are interested in image processing. specially facial emotion is still open for scholar's. This is because Image pose difficulty i.e a person's different(side) position make it difficult to predict emotion of person. However researcher day to day gets improved accuracy.

3 Dataset

In these project we used Public dataset from kaggel FER-2013. The data consists 116 of 48x48 pixel grayscale images of faces. The faces have been automatically 117 registered so that the face is more or less centred and occupies about the same 118 amount of space in each image.

The task is to categorize each face based on the emotion shown in the facial expression into one of seven categories (0=Angry, 1=Disgust, 2=Fear, 3=Happy, 4=Sad, 5=Surprise, 6=Neutral). The training set consists of 28,709 examples and the public test set consists of 3,589 examples.

dataset problems and our solution for the challenge. The first problem is oc- 125 clusion, when the image or part of image is covered by hand or glass. this kind 126 of image must be removed from dataset because the model will not recognise the 127 emotion. The second problem, is data imbalance so data class must be balanced 128 or not to be exaggerated variation. 129

The other problem is contrast variation these can be too dark or too light ¹³⁰ and this can be affect CNN model performance. These is solved by changing the ¹³¹ focus of image to face. The other one is Intra-class variation, which means some ¹³² image in dataset are not human face as there are drawings and animated face. ¹³³ so we have checked the presence of these kinda image.

happiness surprise









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Image-1 sample data from dataset

Methods

For these project we used deep learning, CNN architecture to extract feature's ¹⁵³ of image to classify Emotion of human. A CNN is a Deep Learning algorithm which takes an input image, assigns importance, learnable weights and biases to 155 various aspects/objects in the image and is able to differentiate between images. ¹⁵⁶

The proposed Architecture we have used is CNN architecture Model, the Public data sets we have used contains 7 class facial emotion. First we load the data set and display some sample images. In order to examine the effects of the 159 size of the training set, the data set was divided into different sized training 160 and testing sets. Increasing the size of the training set has positive effect on the 161 result up to a certain level. So we split the data 80 by 20 for training and testing 162 respectively.

We have done preprocessing opration after we load the data. since the dataset. 164 contain noise, some light and also color variation it needs preprocessing step 165 inorder for better performance for our CNN model.

Based on our knowledge with classifying facial expressions regardless of skin ¹⁶⁷ color, facial expression recognition using CNN must also be conducted independently of the color information of the input image. So in preprocessing step we 169 have normalized the image in order to avoid feature difference between image 170 and produce improved image of face.

After we normalize, we convert images to one channel or to grav scale color. ¹⁷² This is because of colored image are difficult to process. Then we reshape image dimension and avoid unnecessary portion of image. This reduce using much ¹⁷⁴ memory size and increase running speed. And also We will use Haar classifier ¹⁷⁵ which trained by Haar-like small features and it will used texture descriptor. ¹⁷⁶ and its main features are linear, edge, canter, and diagonal finally, we build our 177 own CNN model alongside feature extraction and pooling to down sample the 178 feature of image.

4.1 Architecture

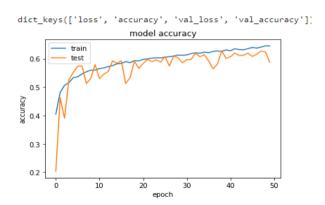
This project is performed using CNN. Feature extraction part consists of four 104 convolutional layers, each four layer are followed by rectified linear unit (ReLU) 185 activation function, and after the first layer, we use Batch Normalization. They 186 are then followed by a dropout layer and two fully-connected layers, each two 187 followed by a ReLU activation, then, another dropout layer, which ends with 100 another fully connected layer. From that, we extract feature vectors, which we concatenate, and then send to a similar network consisting of a dropout layer and two fully-connected layers, each two followed by a ReLU activation, then, 191 another dropout layer, which ends with another fully connected layer.

This model is then trained by optimizing a loss function using RMSprop 193 optimizer. The loss function in this work is categorical cross entropy loss. Adding 194 dropout enables us to train our models from scratch even on very small datasets. 195

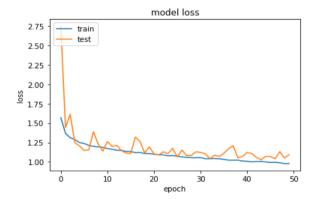
Experiment Result

Intially we used 48x48 input image and then we resize it. For the FER2013 204 dataset, we used 28709 images for training, 3589 for validation, and 3589 for 205 testing. Each image we randomly assigned to the train, validation, or test set. 206 We were able to achieve an accuracy rate of around 65 percent. The train and 207 validation loss and accuracy are reported in Figure below.

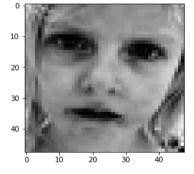
Some Predicted Results:

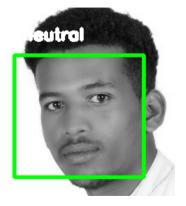


ECCV-16 submission



actual label is:sadness predicted label is:sadness







6 Conclusion

This project is about facial emotion recognition. We use CNN model to develop the project. As we observed from this project the dataset we used has noisy which hinder or decrease model performance. so it needs much preprocessing stage. As of now we didn't cover all issue, for the future we planned to use pytorch deep learning frame work and improve performance of recognizing facial expression of human, because this area need attention now a day to overcome fraud everywhere.

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