Facial Emotion Recognition

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Problem Statement:

Detect different types of emotions displayed by humans

What is emotion detection?

- Process of identifying human emotion from facial_expressions as well as from verbal expressions. This is both something that humans do automatically but computational methodologies have also been developed.
- The task of emotion recognition often involves the analysis of human expressions in multimodal forms such as texts, audio, or video.
- Different emotion types are detected through the integration of information from facial expressions, body movement and gestures, and speech.
- The technology is said to contribute in the emergence of the so-called emotional or emotive Internet.

Importance of Emotion Detection

- 90% of communication can be non verbal according to researchers.
- However, technology is struggling to keep up. It is difficult to analyse the various intonations and intentions.
- Many countries are speculating the adaption of emotion recognizers in key areas of a city like airports, railway stations, markets, etc.
- This is to automatically analyse emotions and the threats displayed
- This will result in reduction of anti-social activities and maintain harmony.

Importance of Emotion Detection

- Global IT companies like Google and Facebook have invested in startups dedicated to detection of emotion and human physiological signals.
- The emotion recognition solution market is worth USD 6.7 billion in 2016 and estimated to reach USD 67.1 billion by 2021

How to detect emotions?

There are various approaches used to detect emotions based on the available data, namely:

- Knowledge-based techniques- utilize domain knowledge and the semantic and syntactic characteristics of language in order to detect certain emotion types.
- Statistical-based techniques- commonly involve Machine Learning techniques eg. SVM, Naive Bayes; Deep-learning algos like CNN etc.

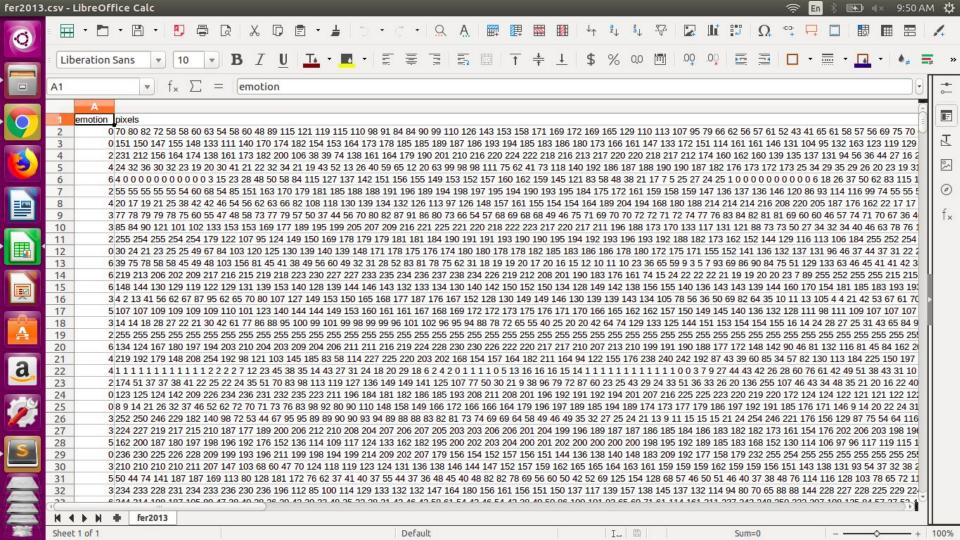
Technology Stack

Programming Language used:
Python

- NumPy
- Pandas
- OpenCV
- Keras
- Theano

Data Set

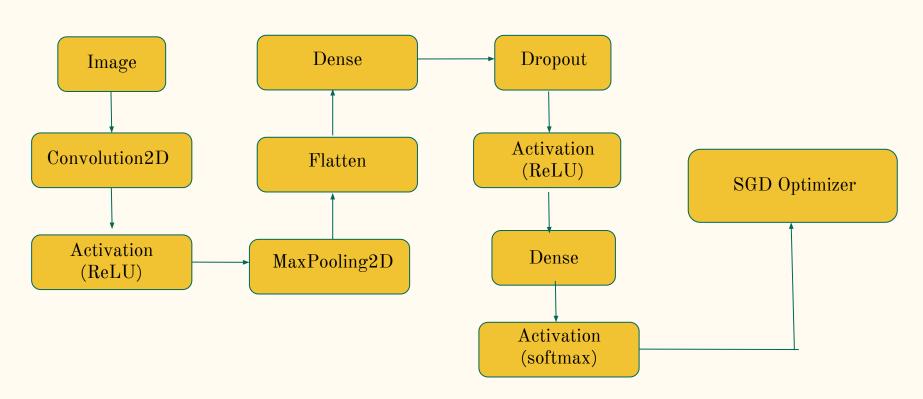
- The data used is an open source dataset from a Kaggle competition
- It involves image pixel values stored in a csv file
- Images are 48 x 48 pixels
- It has labels for 7 different emotions mainly displayed by humans
- These are 0=Angry, 1=Disgust, 2=Fear, 3=Happy, 4=Sad, 5=Surprise,
 6=Neutral
- The data set consists of 28,709 examples



Preprocessing

- The only preprocessing operation performed in "normalization".
- All discreet pixel values are normalized to [0, 255].

CNN architecture



CNN architecture

- Convolution2D: Convolution operator for filtering windows of two-dimensional inputs.
 - No. of convolution filters: 32, No. of rows/columns in convolution kernel: 3
- Activation (ReLU): ReLU is the most used activation function in convolutional neural networks. f(z) = 0 when z < 0 and f(z) = z when z >= 0.
- MaxPooling2D: We have a 2x2 filter that we'll run over our input. For each of the regions, we will take max of that region and create an output matrix where each element is the max of a region in the original input.
- Flatten: Flattens the input. Does not affect the batch size. From m×n to (mn)×1
- Dense: dense layer with 128 outputs.

CNN architecture (contd.)

- Dropout: randomly drops 'p' fraction (in this case 0.4) of i/p units at each update during training time, which helps prevent overfitting.
- Activation(softmax): function will calculate the probabilities of each target class over all possible target classes. Later the calculated probabilities will be helpful for determining the target class for the given inputs. The main advantage of using Softmax is the output probabilities range. The range will be 0 to 1.
- Stochastic Gradient Descent Optimizer

Haar Features

- Haar-like features are used in any kind of object detection.
- These were used by Viola and Jones for the first real time face detector.
- A Haar-like feature considers adjacent rectangular regions at a specific location in a detection window, sums up the pixel intensities in each region and calculates the difference between these sums.
- This difference is then used to categorize subsections of an image
- In detection phase, a window is slided across the image. For each subsection, haar-like features are calculated.
- Difference is compared to a learned threshold that separated objects from non objects.

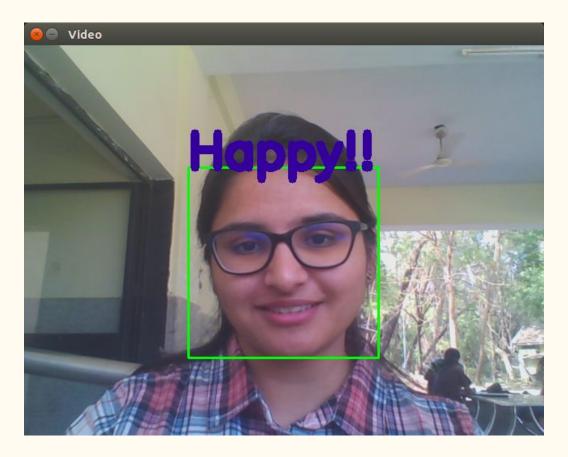
Face Detection using Haar Cascade

- We first need to detect a face. For this, we are using haar cascade.
- The haarcascade_frontalface_default.xml is a haar cascade designed by OpenCV to detect the frontal face.
- It is a machine learning based approach where a cascade function is trained from a lot of positive and negative images. It is then used to detect objects in other images.
- OpenCV comes with its own trainer and detector.
- We load the classifier and give an input which is in grayscale form. The location of the detected faces is returned in the form of a rectangle with coordinates x, y, w, h.

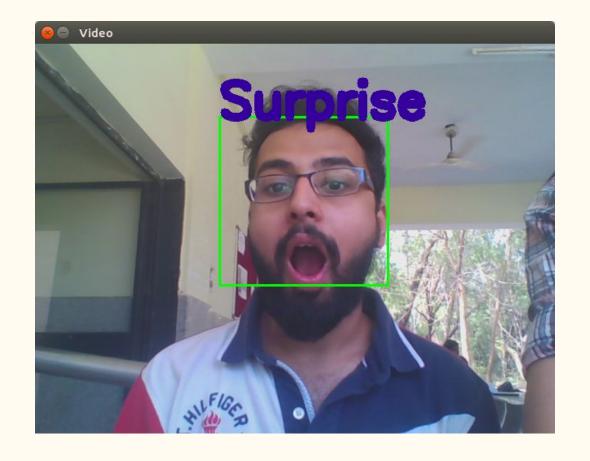
Working model

The following are the steps:

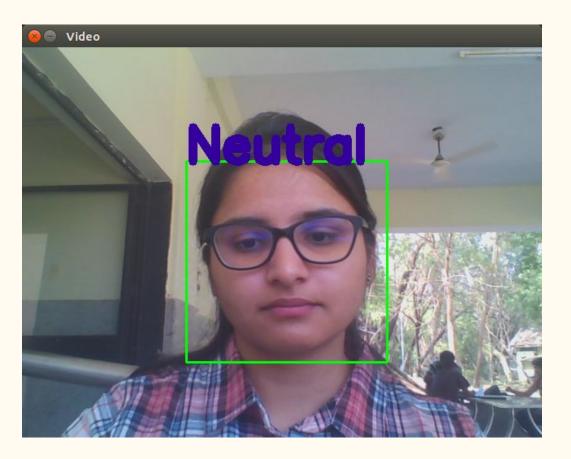
- Load the fer2013 dataset and normalize it.
- The neural network is built and trained with batch size 150 and number of epochs as 15.
- The model architecture is stored and the weights are also stored.
- OpenCV's VideoCapture function is used to first capture video frames from web camera.
- These frames are first used to detect faces.
- Detected faces are then given to the model to predict the output.



Classified as happy



Classified as Surprise



Classified as neutral

References

- Real time CNN for emotion and gender classification by Arriaga, Ploger, Valdenegro
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- https://www.kaggle.com/c/challenges-in-representation-learning-facial-expreses-in-representation-expreses-in-representation-expreses-in-representation-expreses-in-representation-expreses-in-representation-expreses-in-representation-expreses-in-re
- https://en.wikipedia.org/wiki/Emotion_recognition