

Emotion Detection & Classification

GROUP 07

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WORK DISTRIBUTION

1. SALEHIN RAHMAN KHAN: Result Analysis & Environment

Setup

2. MAISHA MALIHA : Dataset Analysis

3. FAIYAZ KHALED : Algorithm Analysis

4.ASHIKUL HAQUE KHAN : Image Processing, Code & Library

5.EHSANUL AMIN KHAN : Programming, Presentation

& Report Writing

OVERVIEW

- ☐ Objective : To Detect and Classify Emotions in Human Faces
- □ Data Collection : GitHub, Kaggle
- Supervised
- ☐ Image Processing
- **■** Neural Network

DATASET ANALYSIS

 We use a set of 28709 pictures of people displaying 7 emotional expression (Angry, Disgusted, Fearful, Happy, Sad, Surprised, Neutral)

 Our dataset contains two columns "Emotion" and "Pixel". Emotion column contains a numeric code ranging from 0 to 6 and Pixel column contains a string surrounded in quotes for each image.

DATASET ANALYSIS

• In input column there is data consists of 48*48 pixel grayscale images of faces and output column consists of emotions which is represented by numbers from 0 to 6.

 The dataset will be divided into two sets training and testing. Furthermore our dataset has no "nan" or "empty/null" values.

IMAGE PROCESSING

Facial Expressions –

■Quantitative Dynamics - determine the amplitude of the expression in terms of intensity levels where the levels correspond to some measures of the extent.

□ **Temporal Dynamics** - splitting the expression into three temporal phases like Onset, Apex, Offset.

IMAGE PROCESSING (Cont.)

- Image is converted to grayscale
- Gamma Correction
- Image Enhancement using Gaussian Filter for -
 - Sharpness
 - Low Aliasing
- After this normalization, the image will be fairly flat limited to noise.

IMAGE PROCESSING (Cont.)

- 1. Head Pose Identification
- 2. Face Tracking
- 3. Face Part Identification
 - i. Eye
 - ii. Eyebrow
 - iii. Mouth

ALGORITHM

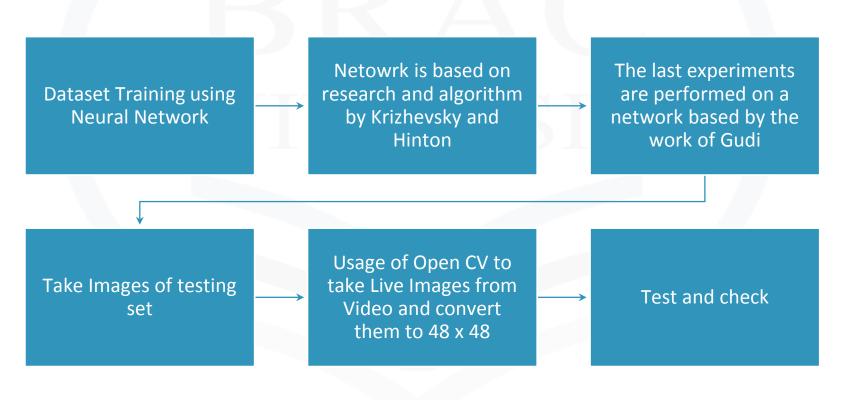
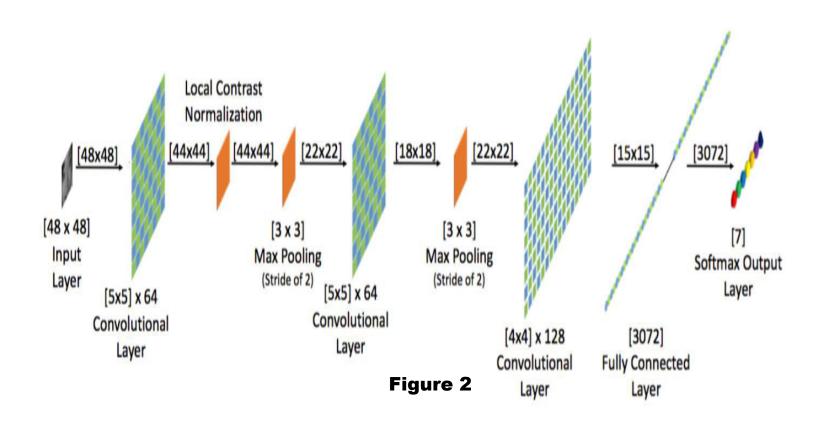


Figure 1

ALGORITHM (Cont.)

- Real time Feedback.
- •Three convolutional layers and two fully connected layers.
 - -maxpooling layers(reducing image sizes)
 - -dropout layers(reduce the chance of over fitting)
- Hyper Parameters.
 - -number of calculations remains the same.

DIAGRAM



CODE & LIBRARY

- TensorFlow,
- ☐ TFLearn,
- OpenCV,
- ☐ Python 2.7

TensorFlow WHY & HOW!

TensorFlow is an open-source software library for dataflow programming across a range of tasks. It is a symbolic math library, and also used for machine learning applications such as neural networks

Its Image Processing + learning

What Learning? Deep learning.

How? Neural Network



Neural Network

- Artificial neural networks or connectionist systems are computing systems vaguely inspired by the biological neural networks that constitute animal brains.
- RBMs (Restricted Boltzmann Machines)
- DBNs (Deep Belief Networks)
- AlexNET
- ☐ GUDI's algorithm

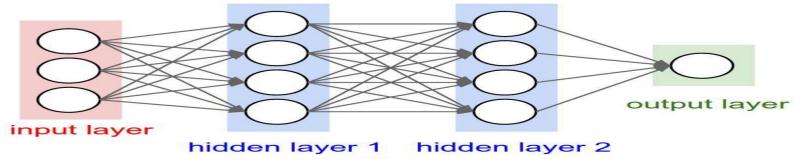


Figure 3

Open CV

- Open Computer Vision
- Used for image processing



Rapid Object Detection using a Boosted Cascade of Simple Features

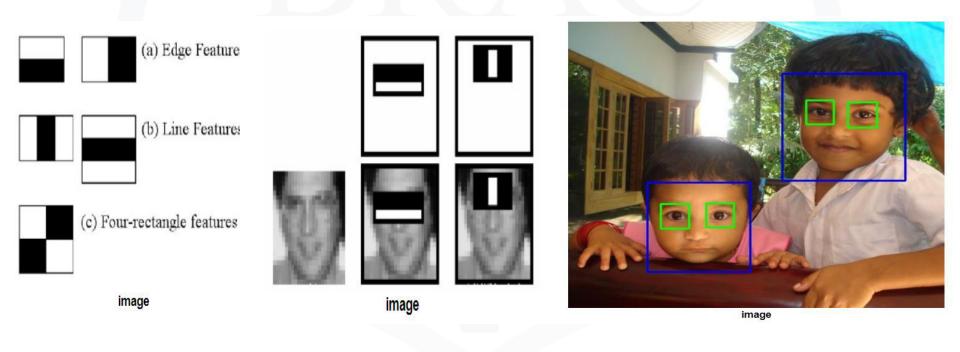
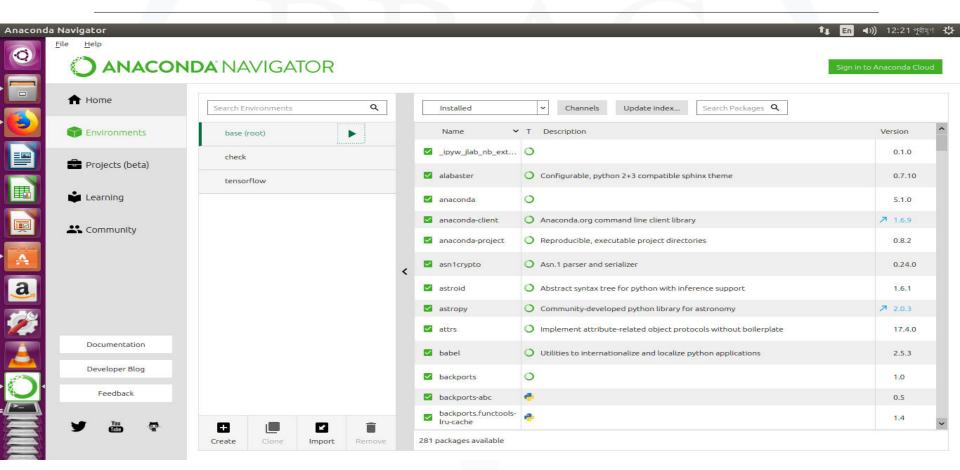


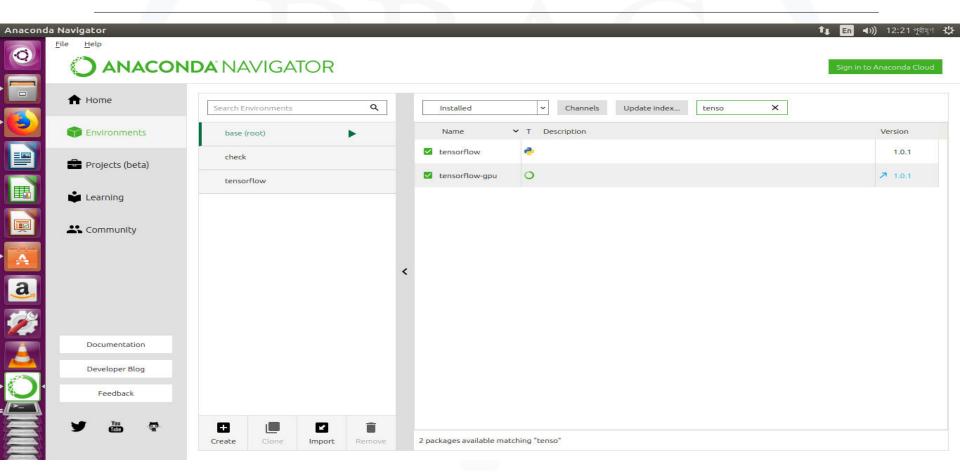
Image 2

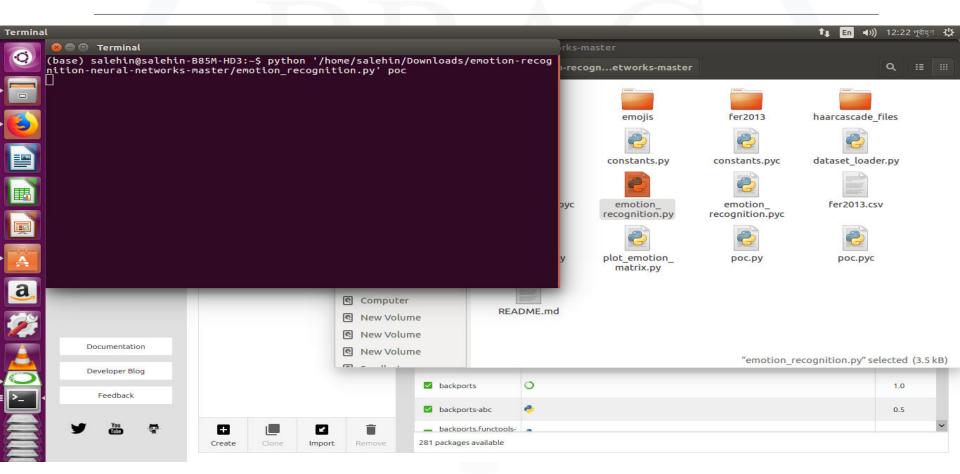
Image 1

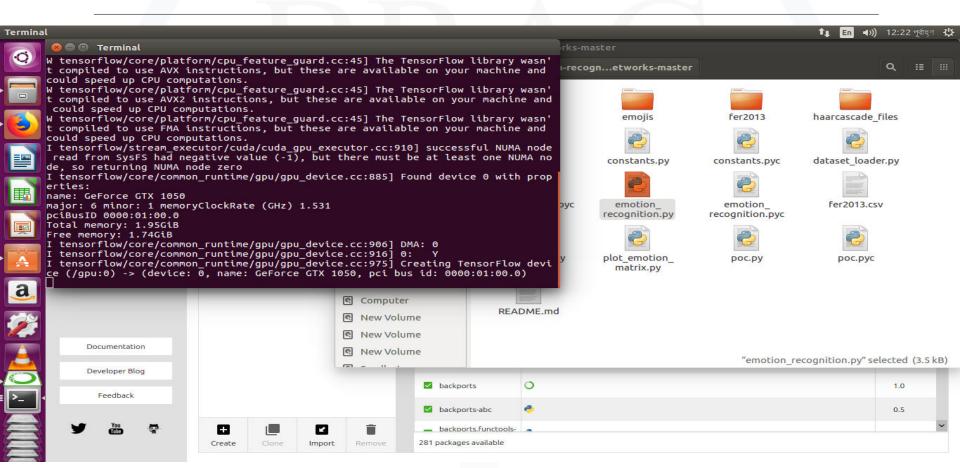
Image 3

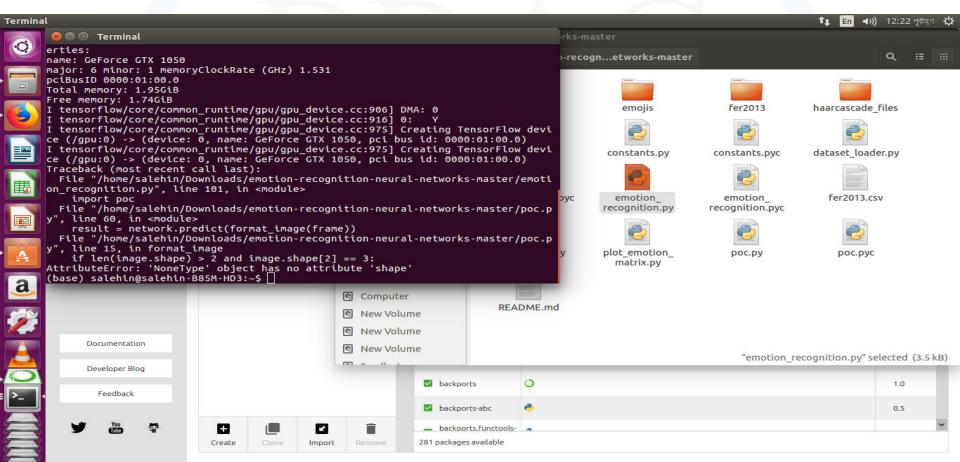
CODES











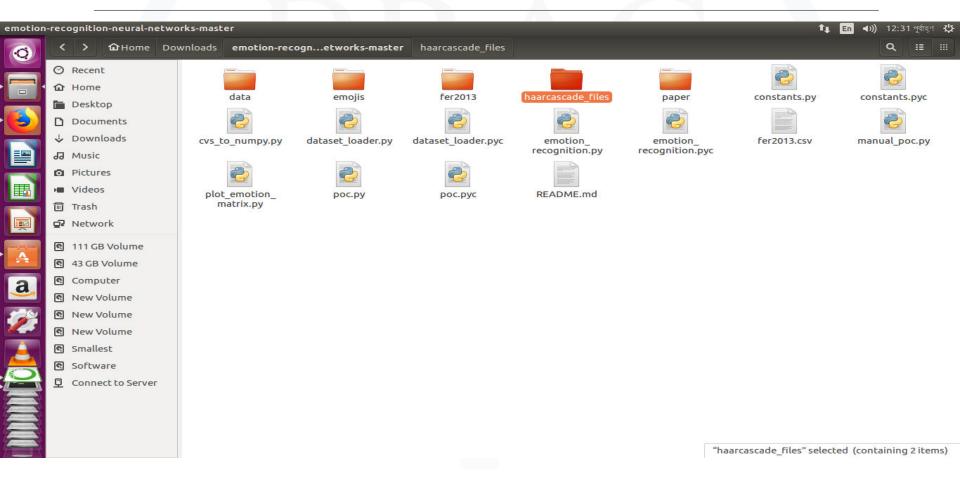
```
emotion_recognition.py (~/Downloads/emotion-recognition-neural-networks-master) - gedit
     Open ▼ IFI
    rom __future__ import division, absolute import
    import re
    import numpy as np
    from dataset loader import DatasetLoader
    import tflearn
    from tflearn.layers.core import input_data, dropout, fully_connected, flatten
   from tflearn.layers.conv import conv 2d, max pool 2d, avg pool 2d
    from tflearn.layers.merge ops import merge
    from tflearn.layers.normalization import local response normalization
    from tflearn.layers.estimator import regression
    from constants import *
    from os.path import isfile, join
    import random
    import sys
    class EmotionRecognition:
     def __init__(self):
       self.dataset = DatasetLoader()
     def build network(self):
        # Smaller 'AlexNet'
        # https://github.com/tflearn/tflearn/blob/master/examples/images/alexnet.py
       print('[+] Building CNN')
        self.network = input_data(shape = [None, SIZE_FACE, SIZE_FACE, 1])
        self.network = conv_2d(self.network, 64, 5, activation = 'relu')
        #self.network = local_response_normalization(self.network)
        self.network = max_pool_2d(self.network, 3, strides = 2)
        self.network = conv_2d(self.network, 64, 5, activation = 'relu')
        self.network = max_pool_2d(self.network, 3, strides = 2)
        self.network = conv_2d(self.network, 128, 4, activation = 'relu')
        self.network = dropout(self.network, 0.3)
        self.network = fully connected(self.network, 3072, activation = 'relu')
        self.network = fully connected(self.network, len(EMOTIONS), activation = 'softmax')
        self.network = regression(self.network,
         optimizer = 'momentum',
         loss = 'categorical crossentropy')
        self.model = tflearn.DNN(
         self.network.
          checkpoint path = SAVE DIRECTORY + '/emotion recognition'.
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```

```
dataset_loader.py (~/Downloads/emotion-recognition-neural-networks-master) - gedit
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     Open ▼ F1
    rom os.path import join
    import numpy as np
    from constants import *
    import cv2
    class DatasetLoader(object):
     def init (self):
      def load from save(self):
        self. images
                          = np.load(join(SAVE DIRECTORY, SAVE DATASET IMAGES FILENAME))
        self._labels
                          = np.load(join(SAVE_DIRECTORY, SAVE_DATASET_LABELS_FILENAME))
        self._images_test = np.load(join(SAVE_DIRECTORY, SAVE_DATASET_IMAGES_TEST_FILENAME))
        self. labels test = np.load(join(SAVE DIRECTORY, SAVE DATASET LABELS TEST FILENAME))
                        = self._images.reshape([-1, SIZE_FACE, SIZE_FACE, 1])
        self._images_test = self._images.reshape([-1, SIZE_FACE, SIZE_FACE, 1])
                      = self._labels.reshape([-1, len(EMOTIONS)])
        self._labels_test = self._labels.reshape([-1, len(EMOTIONS)])
      @property
      def images(self):
       return self._images
      @property
     def labels(self):
       return self. labels
      @property
     def images_test(self):
       return self._images_test
      @property
      def labels test(self):
       return self._labels_test
      @property
     def num examples(self):
       return self. num examples
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```



```
poc.py (~/Downloads/emotion-recognition-neural-networks-master) - gedit
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     Open ▼ J+1
    # Proof-of-concept
    import cv2
    import sys
    from constants import *
    from emotion recognition import EmotionRecognition
    import numpy as np
    cascade classifier = cv2.CascadeClassifier(CASC PATH)
    def brighten(data,b):
         datab = data * b
        return datab
    def format image(image):
      if len(image.shape) > 2 and image.shape[2] == 3:
        image = cv2.cvtColor(image, cv2.COLOR BGR2GRAY)
        image = cv2.imdecode(image, cv2.CV_LOAD_IMAGE_GRAYSCALE)
      faces = cascade_classifier.detectMultiScale(
         scaleFactor = 1.3,
         minNeighbors = 5
      # None is we don't found an image
     if not len(faces) > 0:
       return None
      max_area_face = faces[0]
      for face in faces:
        if face[2] * face[3] > max area face[2] * max area face[3]:
          max_area_face = face
      # Chop image to face
      face = max area face
      image = image[face[1]:(face[1] + face[2]), face[0]:(face[0] + face[3])]
      # Resize image to network size
        image = cv2.resize(image, (SIZE_FACE, SIZE_FACE), interpolation = cv2.INTER_CUBIC) / 255.
      except Exception:
        print("[+] Problem during resize")
        return None
      # cv2.imshow("Lol". image)
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```

```
cvs_to_numpy.py (~/Downloads/emotion-recognition-neural-networks-master/data) - gedit
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     Open ▼ Fi
    def emotion to vec(x):
        d = np.zeros(len(EMOTIONS))
        d[x] = 1.0
        return d
    def flip image(image):
        return cv2.flip(image, 1)
    def data to image(data):
        #print data
        data_image = np.fromstring(str(data), dtype = np.uint8, sep = ' ').reshape((SIZE_FACE, SIZE_FACE))
        data_image = Image.fromarray(data_image).convert('RGB')
        data_image = np.array(data_image)[:, :, ::-1].copy()
        data_image = format_image(data_image)
        return data image
    FILE PATH = 'fer2013.csv'
   data = pd.read csv(FILE PATH)
    labels = []
    images = []
    index = 1
    total = data.shape[0]
    for index, row in data.iterrows():
        emotion = emotion to vec(row['emotion'])
        image = data_to_image(row['pixels'])
        if image is not None:
            labels.append(emotion)
            images.append(image)
            #labels.append(emotion)
            #images.append(flip_image(image))
        else:
            print "Error"
        index += 1
        print "Progreso: {}/{} {:.2f}%".format(index, total, index * 100.0 / total)
    print "Total: " + str(len(images))
    np.save('data_kike.npy', images)
    np.save('labels kike.npy', labels)
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```



EXPECTED RESULTS

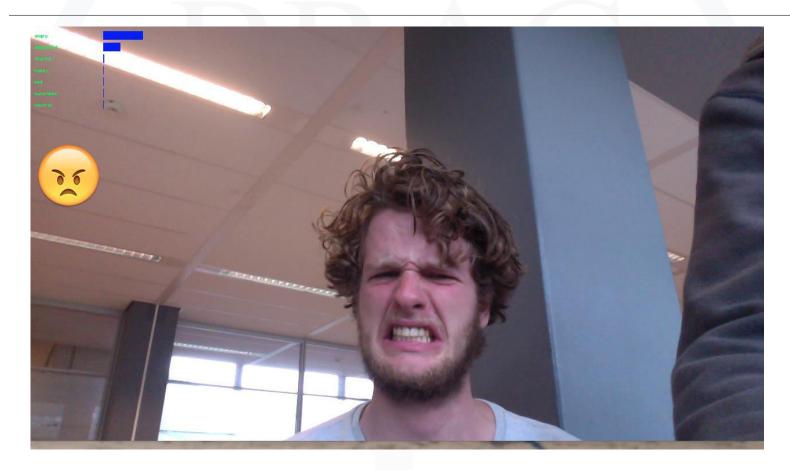


Image 4

CONCLUSION

- Emotion is abstract
- ☐ Face expression is not always the emotion
- ☐ Real Time
- Real Challenges

REFERENCES

- Kaggle. Challenges in representation learning:Facial expression recognition challenge, 2013.
- A. Krizhevsky and G. Hinton. Learning multiple layers of features from tiny images, 2009.
- OpenSourceComputerVision. Face detection using haar cascades. URL
 http://docs.opencv.org/master/d7/d8b/tutorial_py_face_detection.html.

