# **Capstone Project Report**

### **Face Detection And Recognition**

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(Batch 4)

#### • Problem Statement

Face recognition such as that used in our phones, relies on two important steps. First step is face detection, which can be implemented using simple classifiers (such as Haar Cascade) or CNNs. Second part involves the recognition of the identity of the person, which further uses a CNN for this task. Implement a face recognition network using Haar Cascade for detection, followed by using VGG-19 for the task of recognition. Make sure you use a pre-trained VGG model and freeze the weights of starting layers before fine tuning the mode Face detection can be regarded as a specific case of object-class detection, which focuses on the detection of frontal human faces. Once the facial region is obtained, we can use deep learning methods such as CNNs to extract a wide range of features from images. Deep neural networks can be used to produce a bunch of numbers each of which describes a face (known as face encodings) and can be used for both facial recognition and search

#### **Prerequisites**

- Software:
  - Python 3 (Use anaconda as your python distributor as well)
- Tools:
  - Pandas
  - Numpy
  - Matplotlib
  - Seaborn
  - OpenCv
  - Deep Learning
- <u>Dataset</u>: Custom Dataset

#### Method Used

In this case we gonna use the deep learning model (CNN) to train our model on image and at end we gonna freeze the weights trained and then we use that trained weight for any process . For better accuracy we can use VGG19 model which has its own CNN layer inbuilt

#### • Implementation:

#### **1.** Load all required libraries

```
In [1]: from keras.optimizers import RMSprop
    from keras.preprocessing.image import ImageDataGenerator
    from keras.models import Sequential
    from keras.layers import Conv2D,Input,Dense,ZeroPadding2D,BatchNormalization,Activation,MaxPooling2D,Flatten,Dropout
    from keras.models import Model, load_model
    from keras.callbacks import TensorBoard,ModelCheckpoint
    from sklearn.metrics import f1_score
    from sklearn.utils import shuffle
    import imutils
    import numpy as np
    import warnings
    warnings.filterwarnings('ignore')
```

#### 2. Data Augmentation And Reading Dataset

Found 70 images belonging to 2 classes.

```
In [9]: checkpoint = ModelCheckpoint('model2-{epoch:03d}.model',monitor='val_loss',verbose=0,save_best_only=True,mode='auto')
```

#### 3. Model Building

```
In [12]: vgg19 = VGG19(weights='imagenet',include_top=False,input_shape=(128,128,3))
        for layer in vgg19.layers:
           layer.trainable = False
        model = Sequential()
        model.add(vgg19)
        model.add(Flatten())
        model.add(Dense(2,activation='sigmoid'))
        model.summary()
        Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg19/vgg19_weights_tf_dim_ordering_tf_
        1s notop, h5
        80142336/80134624 [=============] - 21s Ous/step
       Model: "sequential'
       Layer (type)
                                Output Shape
                                                      Param #
        ______
        vgg19 (Functional)
                                (None, 4, 4, 512)
                                                      20024384
        flatten (Flatten)
                                (None, 8192)
                                                      0
        dense (Dense)
                                (None, 2)
                                                      16386
                                                                                                   Activate Windo
        ______
        Total params: 20,040,770
        Trainable params: 16,386
        Non-trainable narams
```

## 4. Compiling Model and Training Weights

After Running through 20 Epochs we have reached to the point with good Accuracy now we will save the model weight or freeze for further use

```
In [18]: from keras.models import load_model
model.save("model-10.h5")
```

Now we will use the opency library and gonna use hard cascade yml file to detect face and use our model weight to predict the faces

```
In [1]: import cv2
        import numpy as np
        from keras.models import load_model
        model=load_model("model-10.h5")
        labels_dict={0:'Shikha',1:'Pavan'}
        color_dict={0:(0,0,255),1:(0,255,0)}
        size = 4
        webcam = cv2.VideoCapture(0) #Use camera 0
         # We Load the xml file
        classifier = cv2.CascadeClassifier('haarcascade_frontalface_default.xml')
        while True:
             (rval, im) = webcam.read()
            im=cv2.flip(im,1,1) #Flip to act as a mirror
             # Resize the image to speed up detection
            mini = cv2.resize(im, (im.shape[1] // size, im.shape[0] // size))
             # detect MultiScale / faces
            faces = classifier.detectMultiScale(mini)
    # Draw rectangles around each face
    for f in faces:
        (x, y, w, h) = [v * size for v in f] #Scale the shapesize backup
        #Save just the rectangle faces in SubRecFaces
        face_img = im[y:y+h, x:x+w]
        resized=cv2.resize(face_img,(128,128))
        normalized=resized/255.0
        reshaped=np.reshape(normalized,(1,128,128,3))
        reshaped = np.vstack([reshaped])
        result=model.predict(reshaped)
        #print(result)
        label=np.argmax(result,axis=1)[0]
        \label{eq:cv2.rectangle} {\sf cv2.rectangle(im,(x,y),(x+w,y+h),color\_dict[label],2)}
        {\tt cv2.rectangle(im,(x,y-40),(x+w,y),color\_dict[label],-1)}
        cv2.putText(im, labels_dict[label], (x, y-10),cv2.FONT_HERSHEY_SIMPLEX,0.8,(255,255,255),2)
    # Show the image
    cv2.imshow('LIVE',
                          im)
    key = cv2.waitKey(10)
    # if Esc key is press then break out of the loop
    if key == 27: #The Esc key
        break
# Stop video
webcam.release()
```

# **OUTPUT**

