





Applying MTCNN and Facenet Deep Neural Networks for face recognition to check student attendance

COMP1682 Final Year Project

Programme Title, e.g. BSc Computer Science or BSc Computing

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Facenet

Code Explaination

In my project this time, there are a total of 9 .py files containing code (the main.py file only contains the basic code, so I would like to skip it). I will explain the important parts of the code in turn below so that the lecturer can clearly understand what I have implemented in this project. I will explain in turn the important code in each file below:

managestudent.py:

```
# ========= Variables ==========
self.var fac = StringVar()
self.var cou = StringVar()
self.var year = StringVar()
self.var sem = StringVar()
self.var std id = StringVar()
self.var std name = StringVar()
self.var class = StringVar()
self.var_gender = StringVar()
self.var dob = StringVar()
self.var email = StringVar()
self.var phone = StringVar()
self.var address = StringVar()
# ====== Table Frame ========
table frame = Frame(right frame, bd=2, bg='white', relief='ridge')
table frame.place (x=5, y=80, width=615, height=335)
scroll x = ttk.Scrollbar(table frame, orient=HORIZONTAL)
scroll y = ttk.Scrollbar(table frame, orient=VERTICAL)
self.student_table = ttk.Treeview(table frame, column=(
'fac', 'cou', 'year', 'sem', 'id', 'name', 'class', 'gender', 'dob', 'email',
'phone', 'address', 'photo'),
                                 xscrollcommand=scroll x.set,
yscrollcommand=scroll y.set)
scroll x.pack(side=BOTTOM, fill=X)
scroll y.pack(side=RIGHT, fill=Y)
scroll x.config(command=self.student table.xview)
scroll y.config(command=self.student table.yview)
self.student table['show'] = 'headings'
self.student table.heading('fac', text='Faculty')
self.student table.heading('cou', text='Course')
self.student table.heading('year', text='Year')
self.student table.heading('sem', text='Semester')
self.student table.heading('id', text='Student ID')
self.student table.heading('name', text='Student Name')
self.student table.heading('class', text='Class')
self.student table.heading('gender', text='Gender')
self.student table.heading('dob', text='Birthday')
self.student table.heading('email', text='Email')
```

```
self.student table.heading('phone', text='Phone')
self.student table.heading('address', text='Address')
self.student table.heading('photo', text='Photo Sample Status')
self.student table.column('fac', width=150)
self.student table.column('cou', width=100)
self.student table.column('year', width=100)
self.student table.column('sem', width=100)
self.student table.column('id', width=100)
self.student table.column('name', width=100)
self.student table.column('class', width=100)
self.student table.column('gender', width=100)
self.student table.column('dob', width=100)
self.student table.column('email', width=100)
self.student table.column('phone', width=100)
self.student_table.column('address', width=100)
self.student table.column('photo', width=150)
self.student table.pack(fill=BOTH, expand=1)
self.student table.bind('<ButtonRelease>', self.get cur)
self.fetch data()
```

This part I declare to create a data frame in the application. This dataframe will display data pulled from a MySQL database. The data will display column by column just like in the MySQL database.

This is a short code to connect to the MySQL database, and it will get the data from MySQL to display on the system.

```
def add_data(self):
    if self.var_fac.get() == 'Select Faculty' or self.var_std_name.get() == '' or
self.var_std_id.get() == '':
        messagebox.showerror('Error','All field are
required', parent = self.root)
    else:
        try:
```

```
conn =
mysql.connector.connect(host='localhost', user="root", password='Vilong242', db=
'face recognition')
           cur=conn.cursor()
           cur.execute('insert into student
self.var fac.get(),
               self.var cou.get(),
               self.var year.get(),
               self.var sem.get(),
               self.var std id.get(),
               self.var std name.get(),
               self.var class.get(),
               self.var gender.get(),
               self.var dob.get(),
               self.var email.get(),
               self.var phone.get(),
               self.var address.get(),
               self.var radio1.get()
               ))
           conn.commit()
           self.fetch data()
           conn.close()
           messagebox.showinfo('Success', 'Student Information has been
added successfully',parent=self.root)
       except Exception as es:
           messagebox.showerror('Error',f'Due to:
{str(es)}',parent=self.root)
```

This is a code that helps add a new student information to the database, set up the required fields in the new student registration form and will display a message that the process of adding the student is successful, or meet error.

```
# ------
def update data(self):
   if self.var_fac.get() == 'Select Faculty' or self.var_std_name.get() == '' or
self.var std id.get() == '':
       messagebox.showerror('Error','All field are
required',parent=self.root)
   else:
           update = messagebox.askyesno('Update', 'Do you want to update this
student ?',parent = self.root)
           if update>0:
               conn = mysql.connector.connect(host='localhost', user="root",
password='Vilong242', db='face recognition')
               cur = conn.cursor()
               cur.execute('update student set
Faculty=%s,Course=%s,Year=%s,Semester=%s,Name=%s,Class=%s,Gender=%s,DOB=%s,Em
ail=%s,Phone=%s,Address=%s,PhotoSample=%s where Student ID=%s',(
                  self.var_fac.get(),
                   self.var cou.get(),
                   self.var year.get(),
                   self.var sem.get(),
```

```
self.var std name.get(),
                   self.var class.get(),
                   self.var gender.get(),
                   self.var dob.get(),
                   self.var email.get(),
                   self.var phone.get(),
                   self.var address.get(),
                   self.var radio1.get(),
                   self.var std id.get()
               ) )
           else :
               if not update:
           messagebox.showinfo('Success','Student Information has been
updated successfully',parent=self.root)
           conn.commit()
           self.fetch data()
           conn.close()
       except Exception as es:
           messagebox.showerror('Error', f'Due to: {str(es)}',
parent=self.root)
def delete data(self):
   if self.var std id.get() == '':
       messagebox.showerror('Error','Student ID must be
required', parent=self.root)
   else:
       try:
           delete = messagebox.askyesno('Delete', 'Do you want delete this
student ?',parent=self.root)
           if delete > 0:
               conn = mysql.connector.connect(host='localhost', user="root",
password='Vilong242',db='face recognition')
               cur = conn.cursor()
               sql = 'delete from student where Student ID=%s'
               value = (self.var std id.get(),)
               cur.execute(sql, value)
           else:
               if not delete:
                   return
           conn.commit()
           self.fetch data()
           conn.close()
           messagebox.showinfo('Delete','Student has been deleted
successfully',parent=self.root)
       except Exception as es:
           messagebox.showerror('Error', f'Due to: {str(es)}',
parent=self.root)
```

Two pieces of code I declare two functions to update and delete student data in the MySQL database. Set up the required fields in the new student registration form and will display a message when the process succeeds or encounters an error.

In the above code, I declare a search function to help teachers search for student information in the database.

```
# ====== Generate Dataset or Take Photo button
_____
def generate dataset(self):
    if self.var fac.get() == 'Select Faculty' or self.var std name.get() == '' or
self.var std id.get() == '':
       messagebox.showerror('Error','All field are
required', parent=self.root)
    else:
        try:
            conn = mysql.connector.connect(host='localhost', user="root",
password='Vilong242', db='face recognition')
            cur = conn.cursor()
            cur.execute('select * from student')
           result = cur.fetchall()
           id = 0
            for x in result:
               id += 1
            cur.execute(
            'update student set
Faculty=%s, Course=%s, Year=%s, Semester=%s, Name=%s, Class=%s, Gender=%s, DOB=%s, Em
ail=%s,Phone=%s,Address=%s,PhotoSample=%s where Student ID=%s',
                (
```

```
self.var fac.get(),
        self.var cou.get(),
        self.var year.get(),
        self.var sem.get(),
        self.var std name.get(),
        self.var class.get(),
        self.var gender.get(),
        self.var dob.get(),
        self.var email.get(),
        self.var phone.get(),
        self.var address.get(),
        self.var radio1.get(),
        self.var std id.get() == id+1
    ))
conn.commit()
name = str(self.var std name.get())
os.makedirs('dataSet/' +name+ '-' +str(id)+'')
self.fetch data()
self.reset data()
conn.close()
```

I implemented the code to declare the generate_dataset function, the purpose of this function is to collect the face images of the students in the database. The above code helps to connect to the MySQL database, to save the collected face images by student ID. Collected images will be stored in dataSet folder.

```
if faces != [] :
                    for person in faces:
                        bounding box = person['box']
                        keypoints = person['keypoints']
                        cv2.rectangle(my frame,
                                       (bounding box[0], bounding box[1]),
                                       (bounding_box[0] + bounding_box[2],
bounding box[1] + bounding box[3]),
                                       (0, 155, 255),
                                       2)
                        cv2.circle(my frame, (keypoints['left_eye']), 2, (0,
155, 255), 2)
                        cv2.circle(my frame, (keypoints['right eye']), 2, (0,
155, 255), 2)
                        cv2.circle(my frame, (keypoints['nose']), 2, (0, 155,
255), 2)
                        cv2.circle(my frame, (keypoints['mouth left']), 2,
(0, 155, 255), 2)
                        cv2.circle(my frame, (keypoints['mouth_right']), 2,
(0, 155, 255), 2)
                        sampleNum += 1
                        file path = "dataSet/" +str(name) + '-' +str(id) +''
"/" +str(name) + "-" +str(id) + "." + str(sampleNum) + ".jpg"
                        cv2.imwrite(file path, my frame)
                    cv2.putText(my frame, str(sampleNum), (50, 50),
cv2.FONT_HERSHEY_COMPLEX, 2, (100, 200, 200), 2)
                    cv2.imshow('Take Photo', my_frame)
                if cv2.waitKey(100) & 0xFF == ord('q'):
                    break
                elif sampleNum > 100:
                    break
```

```
camera.release()
    cv2.destroyAllWindows()
    messagebox.showinfo('Result', 'Taking photo completed !!!')

except Exception as es:
    messagebox.showerror('Error', f'Due to: {str(es)}',
parent=self.root)
```

The above code declares to perform facial photography of students. I declare a variable to execute the MTCNN algorithm that I implemented in another .py file. Assign the camera variable to let the system use the webcam to collect images. Use OpenCV to represent and mark landmarks of faces detected in the image. The captured images will be saved in a folder with the student's name and id. Setup when the system captures one hundred images will stop and display a success message.

attendance.py:

```
def fetch_data(self):
    with open('attendance.csv') as file:
        df = pd.read_csv('attendance.csv')

    self.attendance_table['column'] = list(df.columns)
    self.attendance_table["show"] = 'headings'
    for column in self.attendance_table['columns']:
        self.attendance_table.heading(column,text=column)

    self.attendance_table.delete(*self.attendance_table.get_children())
    df_rows= df.to_numpy().tolist()
    for row in df_rows:
        self.attendance_table.insert("",END,values = row)
```

This is a short code to connect to the MySQL database, and it will get the data from MySQL to display on the system.

```
def search(self):
    with open('attendance.csv','r') as file:
        filereader = csv.reader(file)
        filereader_list = list(filereader)

self.attendance_table.delete(*self.attendance_table.get_children())
    word = self.name_data.get().title()
```

```
if self.name_data.get():
    for i in filereader_list:
        if word in i:
            self.attendance table.insert("", END, values=i)
```

In the above code, I declare a search function to help teachers search for student information in the database.

```
def sendemail(self):
   curr = datetime.now()
   dt = curr.strftime("%d/%m/%Y")
   subject = "Data Attendance"
   body = "Here are the data attendance"
    sender email = "universitygreenwichattendance@gmail.com"
    receiver email = "vilong242@gmail.com"
   password = 'mkbwaznnpkxulxiy'
    # Create a multipart message and set headers
   message = MIMEMultipart()
   message["From"] = sender email
   message["To"] = receiver email
   message["Subject"] = subject
   message["Bcc"] = receiver email # Recommended for mass emails
    # Add body to email
   message.attach(MIMEText(body, "plain"))
    filename = "attendance.csv" # In same directory as script
    # Open PDF file in binary mode
    with open(filename, "rb") as attachment:
        # Add file as application/octet-stream
        # Email client can usually download this automatically as attachment
       part = MIMEBase("application", "octet-stream")
       part.set payload(attachment.read())
    # Encode file in ASCII characters to send by email
    encoders.encode base64(part)
    # Add header as key/value pair to attachment part
   part.add header(
        "Content-Disposition",
        f"attachment; filename= {filename}",
    # Add attachment to message and convert message to string
   message.attach(part)
   text = message.as string()
    # Log in to server using secure context and send email
    context = ssl.create default context()
    with smtplib.SMTP SSL("smtp.gmail.com", 465, context=context) as server:
       server.login(sender email, password)
       server.sendmail(sender email, receiver email, text)
       messagebox.showinfo('Successfull', f'Attendance data on %s has been
sent'%dt)
```

In the above code, I declare a sendemail() function so that teachers can send a list of registered students to their personal email. I use python's smtplib library to setup SMTP Client to send email to another Email address on Internet via SMTP protocol. And when the user bbaasm button, the email will be automatically sent to the specified email. Besides, the system also displays a message sent successfully.

detect_face.py:

The use of this .py file is to use OpenCV to create processing functions and create classes used for the face recognition algorithm MTCNN.

```
class Network(object):
    def init (self, inputs, trainable=True):
        # The input nodes for this network
        self.inputs = inputs
        # The current list of terminal nodes
        self.terminals = []
        # Mapping from layer names to layers
        self.layers = dict(inputs)
        # If true, the resulting variables are set as trainable
        self.trainable = trainable
        self.setup()
    def setup(self):
        """Construct the network. """
        raise NotImplementedError('Must be implemented by the subclass.')
    def load(self, data path, session, ignore missing=False):
        """Load network weights.
        data path: The path to the numpy-serialized network weights
        session: The current TensorFlow session
        ignore missing: If true, serialized weights for missing layers are
ignored.
        data dict = np.load(data path, encoding='latin1',
allow pickle=True).item() #pylint: disable=no-member
        for op name in data dict:
            with tf.variable scope(op name, reuse=True):
                for param name, data in iteritems(data dict[op name]):
                    try:
                        var = tf.get variable(param name)
                        session.run(var.assign(data))
                    except ValueError:
                        if not ignore missing:
                            raise
```

```
def feed(self, *args):
        """Set the input(s) for the next operation by replacing the terminal
nodes.
        The arguments can be either layer names or the actual layers.
        assert len(args) != 0
        self.terminals = []
        for fed layer in args:
            if isinstance(fed layer, string types):
                    fed layer = self.layers[fed layer]
                except KeyError:
                    raise KeyError('Unknown layer name fed: %s' % fed layer)
            self.terminals.append(fed layer)
        return self
    def get output(self):
        """Returns the current network output."""
        return self.terminals[-1]
    def get unique name(self, prefix):
        """Returns an index-suffixed unique name for the given prefix.
        This is used for auto-generating layer names based on the type-
prefix.
        ident = sum(t.startswith(prefix) for t, in self.layers.items()) + 1
        return '%s %d' % (prefix, ident)
    def make var(self, name, shape):
        """Creates a new TensorFlow variable."""
        return tf.get variable(name, shape, trainable=self.trainable)
    def validate padding(self, padding):
        """Verifies that the padding is one of the supported ones."""
        assert padding in ('SAME', 'VALID')
```

The above section will declare the overall classes that will be used in the MTCNN algorithm, want to know more about the classes in the algorithm. Please review the report file that I submitted. There I explained in detail the classes included in the MTCNN algorithm

```
name,
             relu=True,
             padding='SAME',
             group=1,
             biased=True):
        # Verify that the padding is acceptable
        self.validate padding(padding)
        # Get the number of channels in the input
        c i = int(inp.get shape()[-1])
        # Verify that the grouping parameter is valid
        assert c i % group == 0
        assert c o % group == 0
        # Convolution for a given input and kernel
        convolve = lambda i, k: tf.nn.conv2d(i, k, [1, s h, s w, 1],
padding=padding)
        with tf.variable scope(name) as scope:
            kernel = self.make var('weights', shape=[k h, k w, c i // group,
c o])
            # This is the common-case. Convolve the input without any further
complications.
            output = convolve(inp, kernel)
            # Add the biases
            if biased:
                biases = self.make var('biases', [c o])
                output = tf.nn.bias add(output, biases)
            if relu:
                # ReLU non-linearity
                output = tf.nn.relu(output, name=scope.name)
            return output
```

Declare the Convolution Layer in the MTCNN algorithm

```
@layer
def prelu(self, inp, name):
    with tf.variable_scope(name):
        i = int(inp.get_shape()[-1])
        alpha = self.make_var('alpha', shape=(i,))
        output = tf.nn.relu(inp) + tf.multiply(alpha, -tf.nn.relu(-inp))
```

Declare the PreLu Layer in the MTCNN algorithm

Declare the Maxpool Layer in the MTCNN algorithm

```
@layer
def fc(self, inp, num out, name, relu=True):
    with tf.variable scope(name):
        input shape = inp.get shape()
        if input shape.ndims == 4:
            # The input is spatial. Vectorize it first.
            dim = 1
            for d in input shape[1:].as list():
                dim *= int(d)
            feed in = tf.reshape(inp, [-1, dim])
        else:
            feed in, dim = (inp, input shape[-1].value)
        weights = self.make var('weights', shape=[dim, num out])
        biases = self.make var('biases', [num out])
        op = tf.nn.relu layer if relu else tf.nn.xw plus b
        fc = op(feed in, weights, biases, name=name)
        return fc
```

Declare the Fully Connected Layer in the MTCNN algorithm

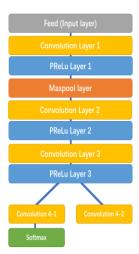
```
@layer
def softmax(self, target, axis, name=None):
    max axis = tf.reduce max(target, axis, keepdims=True)
```

```
target_exp = tf.exp(target-max_axis)
normalize = tf.reduce_sum(target_exp, axis, keepdims=True)
softmax = tf.div(target_exp, normalize, name)
return softmax
```

Declare the Softmax Layer in the MTCNN algorithm

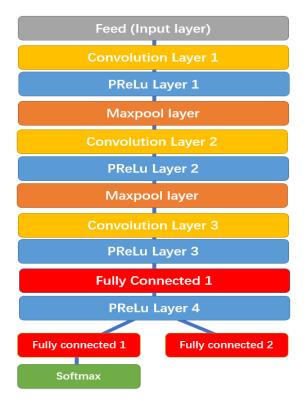
Next, after declaring the classes used in the algorithm, I will in turn arrange the classes that have just formed into the 3 main networks of MTCNN that are: P -Net, R-Net and O-Net

Above is the arrangement of the P-Net class in order:



```
class RNet(Network):
   def setup(self):
        (self.feed('data') #pylint: disable=no-value-for-parameter, no-member
             .conv(3, 3, 28, 1, 1, padding='VALID', relu=False, name='conv1')
             .prelu(name='prelu1')
             .max pool(3, 3, 2, 2, name='pool1')
             .conv(3, 3, 48, 1, 1, padding='VALID', relu=False, name='conv2')
             .prelu(name='prelu2')
             .max pool(3, 3, 2, 2, padding='VALID', name='pool2')
             .conv(2, 2, 64, 1, 1, padding='VALID', relu=False, name='conv3')
             .prelu(name='prelu3')
             .fc(128, relu=False, name='conv4')
             .prelu(name='prelu4')
             .fc(2, relu=False, name='conv5-1')
             .softmax(1,name='prob1'))
        (self.feed('prelu4') #pylint: disable=no-value-for-parameter
             .fc(4, relu=False, name='conv5-2'))
```

Above is the arrangement of the R-Net class in order:



```
class ONet(Network):
   def setup(self):
        (self.feed('data') #pylint: disable=no-value-for-parameter, no-member
             .conv(3, 3, 32, 1, 1, padding='VALID', relu=False, name='conv1')
             .prelu(name='prelu1')
             .max pool(3, 3, 2, 2, name='pool1')
             .conv(3, 3, 64, 1, 1, padding='VALID', relu=False, name='conv2')
             .prelu(name='prelu2')
             .max pool(3, 3, 2, 2, padding='VALID', name='pool2')
             .conv(3, 3, 64, 1, 1, padding='VALID', relu=False, name='conv3')
             .prelu(name='prelu3')
             .max pool(2, 2, 2, 2, name='pool3')
             .conv(2, 2, 128, 1, 1, padding='VALID', relu=False,
name='conv4')
             .prelu(name='prelu4')
             .fc(256, relu=False, name='conv5')
             .prelu(name='prelu5')
             .fc(2, relu=False, name='conv6-1')
             .softmax(1, name='prob1'))
        (self.feed('prelu5') #pylint: disable=no-value-for-parameter
             .fc(4, relu=False, name='conv6-2'))
        (self.feed('prelu5') #pylint: disable=no-value-for-parameter
             .fc(10, relu=False, name='conv6-3'))
```

And finally the O-Net class:



```
def create mtcnn(sess, model path):
    if not model path:
        model path, = os.path.split(os.path.realpath( file ))
    with tf.variable scope('pnet'):
        data = tf.placeholder(tf.float32, (None, None, None, 3), 'input')
        pnet = PNet({'data':data})
        pnet.load(os.path.join(model path, 'det1.npy'), sess)
    with tf.variable scope('rnet'):
        data = tf.placeholder(tf.float32, (None, 24, 24, 3), 'input')
        rnet = RNet({'data':data})
        rnet.load(os.path.join(model path, 'det2.npy'), sess)
    with tf.variable scope('onet'):
        data = tf.placeholder(tf.float32, (None, 48, 48, 3), 'input')
        onet = ONet({'data':data})
        onet.load(os.path.join(model path, 'det3.npy'), sess)
   pnet fun = lambda img : sess.run(('pnet/conv4-2/BiasAdd:0',
'pnet/prob1:0'), feed dict={'pnet/input:0':img})
    rnet fun = lambda img : sess.run(('rnet/conv5-2/conv5-2:0',
'rnet/prob1:0'), feed dict={'rnet/input:0':img})
    onet fun = lambda img : sess.run(('onet/conv6-2/conv6-2:0', 'onet/conv6-
3/conv6-3:0', 'onet/prob1:0'), feed dict={'onet/input:0':imq})
    return pnet fun, rnet fun, onet fun
```

Finally, set up a complete MTCNN algorithm

Next, after creating a complete MTCNN algorithm. The next thing, I need to declare a detect_face function to apply the MTCNN algorithm I just code in to be able to recognize faces.

```
def detect_face(img, minsize, pnet, rnet, onet, threshold, factor):
    """Detects faces in an image, and returns bounding boxes and points for
them.
    img: input image
    minsize: minimum faces' size
    pnet, rnet, onet: caffemodel
```

```
threshold: threshold=[th1, th2, th3], th1-3 are three steps's threshold
    factor: the factor used to create a scaling pyramid of face sizes to
detect in the image.
    11 11 11
    factor count=0
    total boxes=np.empty((0,9))
    points=np.empty(0)
    h=img.shape[0]
    w=img.shape[1]
    minl=np.amin([h, w])
   m=12.0/minsize
    minl=minl*m
    # create scale pyramid
    scales=[]
    while minl>=12:
        scales += [m*np.power(factor, factor count)]
        minl = minl*factor
        factor count += 1
```

It will first create a pyramid to scale the entire input image

```
# first stage
for scale in scales:
    hs=int(np.ceil(h*scale))
    ws=int(np.ceil(w*scale))
    im_data = imresample(img, (hs, ws))
    im_data = (im_data-127.5)*0.0078125
    img_x = np.expand_dims(im_data, 0)
    img_y = np.transpose(img_x, (0,2,1,3))
    out = pnet(img_y)
    out0 = np.transpose(out[0], (0,2,1,3))
    out1 = np.transpose(out[1], (0,2,1,3))

    boxes, _ = generateBoundingBox(out1[0,:,:,1].copy(),
out0[0,:,:,:].copy(), scale, threshold[0])

# inter-scale nms
```

```
pick = nms(boxes.copy(), 0.5, 'Union')
                     if boxes.size>0 and pick.size>0:
                               boxes = boxes[pick,:]
                               total boxes = np.append(total boxes, boxes, axis=0)
          numbox = total boxes.shape[0]
          if numbox>0:
                    pick = nms(total boxes.copy(), 0.7, 'Union')
                     total boxes = total boxes[pick,:]
                    regw = total boxes[:,2]-total boxes[:,0]
                     regh = total boxes[:,3]-total boxes[:,1]
                    qq1 = total boxes[:,0]+total boxes[:,5]*regw
                     qq2 = total boxes[:,1]+total boxes[:,6]*regh
                     gg3 = total boxes[:,2]+total boxes[:,7]*regw
                     qq4 = total boxes[:,3]+total boxes[:,8]*regh
                     total boxes = np.transpose(np.vstack([qq1, qq2, qq3, qq4,
total boxes[:,4]]))
                     total boxes = rerec(total boxes.copy())
                     total boxes[:,0:4] = np.fix(total boxes[:,0:4]).astype(np.int32)
                    dy, edy, dx, edx, y, ey, x, ex, tmpw, tmph = pad(total boxes.copy(),
w, h)
          numbox = total boxes.shape[0]
          if numbox>0:
                     # second stage
                    temping = np.zeros((24,24,3,numbox))
                     for k in range(0, numbox):
                               tmp = np.zeros((int(tmph[k]),int(tmpw[k]),3))
                               tmp[dy[k]-1:edy[k],dx[k]-1:edx[k],:] = img[y[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]-1:ey[k],x[k]
1:ex[k],:]
                               if tmp.shape[0]>0 and tmp.shape[1]>0 or tmp.shape[0]==0 and
tmp.shape[1] == 0:
                                         tempimg[:,:,:,k] = imresample(tmp, (24, 24))
                               else:
                                         return np.empty()
                     temping = (temping-127.5)*0.0078125
                     tempimg1 = np.transpose(tempimg, (3,1,0,2))
                     out = rnet(tempimg1)
```

```
out0 = np.transpose(out[0])
                     out1 = np.transpose(out[1])
                     score = out1[1,:]
                     ipass = np.where(score>threshold[1])
                      total boxes = np.hstack([total boxes[ipass[0],0:4].copy(),
np.expand dims(score[ipass].copy(),1)])
                     mv = out0[:,ipass[0]]
                     if total boxes.shape[0]>0:
                                pick = nms(total boxes, 0.7, 'Union')
                                total boxes = total boxes[pick,:]
                                total boxes = bbreg(total boxes.copy(), np.transpose(mv[:,pick]))
                                total boxes = rerec(total boxes.copy())
          numbox = total boxes.shape[0]
          if numbox>0:
                      # third stage
                     total boxes = np.fix(total boxes).astype(np.int32)
                      dy, edy, dx, edx, y, ey, x, ex, tmpw, tmph = pad(total boxes.copy(),
w, h)
                      temping = np.zeros((48,48,3,numbox))
                      for k in range(0, numbox):
                                tmp = np.zeros((int(tmph[k]),int(tmpw[k]),3))
                                tmp[dy[k]-1:edy[k], dx[k]-1:edx[k], :] = img[y[k]-1:ey[k], x[k]-1:ey[k], x[k]-1:ey[k
1:ex[k],:]
                                if tmp.shape[0]>0 and tmp.shape[1]>0 or tmp.shape[0]==0 and
tmp.shape[1]==0:
                                           tempimg[:,:,:,k] = imresample(tmp, (48, 48))
                                else:
                                           return np.empty()
                      tempimg = (tempimg-127.5)*0.0078125
                      tempimg1 = np.transpose(tempimg, (3,1,0,2))
                      out = onet(tempimg1)
                     out0 = np.transpose(out[0])
                     out1 = np.transpose(out[1])
                     out2 = np.transpose(out[2])
                     score = out2[1,:]
                     points = out1
                      ipass = np.where(score>threshold[2])
```

```
points = points[:,ipass[0]]
        total boxes = np.hstack([total boxes[ipass[0],0:4].copy(),
np.expand dims(score[ipass].copy(),1)])
       mv = out0[:,ipass[0]]
        w = total boxes[:,2]-total boxes[:,0]+1
        h = total boxes[:,3]-total boxes[:,1]+1
        points[0:5,:] = np.tile(w, (5, 1))*points[0:5,:] +
np.tile(total boxes[:,0],(5, 1))-1
        points[5:10,:] = np.tile(h, (5, 1))*points[5:10,:] +
np.tile(total boxes[:,1],(5, 1))-1
        if total boxes.shape[0]>0:
            total boxes = bbreg(total boxes.copy(), np.transpose(mv))
            pick = nms(total boxes.copy(), 0.7, 'Min')
            total boxes = total boxes[pick,:]
            points = points[:,pick]
    return total boxes, points
```

Then, set up the system to scan the input image, apply the MTCNN algorithm just declared above and return the results as the location coordinates and landmarks of the face.

```
# function [boundingbox] = bbreg(boundingbox,reg)

def bbreg(boundingbox,reg):
    """Calibrate bounding boxes"""
    if reg.shape[1]==1:
        reg = np.reshape(reg, (reg.shape[2], reg.shape[3]))

w = boundingbox[:,2]-boundingbox[:,0]+1
    h = boundingbox[:,3]-boundingbox[:,1]+1
    b1 = boundingbox[:,0]+reg[:,0]*w
    b2 = boundingbox[:,1]+reg[:,1]*h
    b3 = boundingbox[:,2]+reg[:,2]*w
    b4 = boundingbox[:,3]+reg[:,3]*h
    boundingbox[:,0:4] = np.transpose(np.vstack([b1, b2, b3, b4]))
    return boundingbox
```

```
def generateBoundingBox(imap, reg, scale, t):
    """Use heatmap to generate bounding boxes"""
    stride=2
    cellsize=12
    imap = np.transpose(imap)
    dx1 = np.transpose(reg[:,:,0])
    dy1 = np.transpose(reg[:,:,1])
    dx2 = np.transpose(reg[:,:,2])
    dy2 = np.transpose(reg[:,:,3])
    y, x = np.where(imap >= t)
    if y.shape[0] == 1:
        dx1 = np.flipud(dx1)
        dy1 = np.flipud(dy1)
        dx2 = np.flipud(dx2)
        dy2 = np.flipud(dy2)
    score = imap[(y,x)]
    reg = np.transpose(np.vstack([ dx1[(y,x)], dy1[(y,x)], dx2[(y,x)],
dy2[(y,x)]]))
    if req.size==0:
        reg = np.empty((0,3))
   bb = np.transpose(np.vstack([y,x]))
    q1 = np.fix((stride*bb+1)/scale)
    q2 = np.fix((stride*bb+cellsize-1+1)/scale)
   boundingbox = np.hstack([q1, q2, np.expand dims(score,1), reg])
    return boundingbox, reg
```

Finally, we will declare 2 more functions to help draw boxes on the screen to mark faces on the input image

facenet.py:

Next, will come to the file facenet.py. The goal of this file is to create a Facenet algorithm to apply to face recognition techniques

```
def triplet_loss(anchor, positive, negative, alpha):
    """Calculate the triplet loss according to the FaceNet paper
    Args:
    anchor: the embeddings for the anchor images.
    positive: the embeddings for the positive images.
```

```
negative: the embeddings for the negative images.
    Returns:
     the triplet loss according to the FaceNet paper as a float tensor.
    with tf.variable scope('triplet loss'):
        pos dist = tf.reduce sum(tf.square(tf.subtract(anchor, positive)),
1)
        neg dist = tf.reduce sum(tf.square(tf.subtract(anchor, negative)),
1)
        basic loss = tf.add(tf.subtract(pos dist, neg dist), alpha)
        loss = tf.reduce mean(tf.maximum(basic loss, 0.0), 0)
    return loss
def center loss(features, label, alfa, nrof classes):
    nrof features = features.get shape()[1]
    centers = tf.get_variable('centers', [nrof_classes, nrof_features],
dtype=tf.float32,
                              initializer=tf.constant initializer(0),
trainable=False)
    label = tf.reshape(label, [-1])
    centers batch = tf.gather(centers, label)
   diff = (1 - alfa) * (centers batch - features)
    centers = tf.scatter sub(centers, label, diff)
   with tf.control dependencies([centers]):
        loss = tf.reduce_mean(tf.square(features - centers_batch))
    return loss, centers
```

First, I declare the function triplet loss, center loss of the Facenet algorithm. It will return a loss value and a centers value. These are the two decisive values to calculate the distance between the two vectors after being extracted

```
def create input pipeline (input queue, image size,
nrof preprocess threads, batch size placeholder):
    images and labels list = []
    for in range(nrof preprocess threads):
        filenames, label, control = input queue.dequeue()
        images = []
        for filename in tf.unstack(filenames):
            file contents = tf.read file(filename)
            image = tf.image.decode image(file contents, 3)
            image = tf.cond(get control flag(control[0], RANDOM ROTATE),
                            lambda: tf.py func(random rotate image,
[image], tf.uint8),
                            lambda: tf.identity(image))
            image = tf.cond(get control flag(control[0], RANDOM CROP),
                            lambda: tf.random crop(image, image size +
(3,)),
                            lambda:
tf.image.resize image with crop or pad(image, image size[0],
image size[1]))
            image = tf.cond(get control flag(control[0], RANDOM FLIP),
tf.image.random flip left right(image),
                            lambda: tf.identity(image))
            image = tf.cond(get control flag(control[0],
FIXED STANDARDIZATION),
                            lambda: (tf.cast(image, tf.float32) - 127.5) /
128.0,
                            lambda:
tf.image.per image standardization(image))
            image = tf.cond(get control flag(control[0], FLIP),
                            lambda: tf.image.flip left right(image),
                            lambda: tf.identity(image))
            # pylint: disable=no-member
            image.set shape(image size + (3,))
            images.append(image)
        images and labels list.append([images, label])
    image batch, label batch = tf.train.batch join(
```

```
images_and_labels_list, batch_size=batch_size_placeholder,
    shapes=[image_size + (3,), ()], enqueue_many=True,
    capacity=4 * nrof_preprocess_threads * 100,
    allow_smaller_final_batch=True)

return image batch, label batch
```

Next, I declare and set the path for the input images (specifically here are the face images that have been detected at the MTCNN algorithm). After the system calculates and processes complex expressions with the help of tensorflow. The result will return the vector and label values of the input images

```
def get control flag(control, field):
    return tf.equal(tf.mod(tf.floor div(control, field), 2), 1)
def add loss summaries(total loss):
    """Add summaries for losses.
    Generates moving average for all losses and associated summaries for
    visualizing the performance of the network.
   Args:
     total loss: Total loss from loss().
    Returns:
      loss averages op: op for generating moving averages of losses.
    # Compute the moving average of all individual losses and the total
loss.
    loss averages = tf.train.ExponentialMovingAverage(0.9, name='avg')
    losses = tf.get collection('losses')
    loss averages op = loss averages.apply(losses + [total loss])
    # Attach a scalar summmary to all individual losses and the total
loss; do the
    # same for the averaged version of the losses.
   for 1 in losses + [total loss]:
```

```
# Name each loss as '(raw)' and name the moving average version of
the loss

# as the original loss name.

tf.summary.scalar(l.op.name + ' (raw)', l)

tf.summary.scalar(l.op.name, loss_averages.average(l))

return loss averages op
```

In the above code, I proceed to calculate the average value of the loss values declared above. This value is required to train data

```
def train(total loss, global step, optimizer, learning rate,
moving_average_decay, update_gradient_vars,
          log histograms=True):
    # Generate moving averages of all losses and associated summaries.
    loss averages op = add loss summaries(total_loss)
    # Compute gradients.
    with tf.control dependencies([loss averages op]):
        if optimizer == 'ADAGRAD':
            opt = tf.train.AdagradOptimizer(learning rate)
        elif optimizer == 'ADADELTA':
            opt = tf.train.AdadeltaOptimizer(learning rate, rho=0.9,
epsilon=1e-6)
        elif optimizer == 'ADAM':
            opt = tf.train.AdamOptimizer(learning rate, beta1=0.9,
beta2=0.999, epsilon=0.1)
        elif optimizer == 'RMSPROP':
            opt = tf.train.RMSPropOptimizer(learning rate, decay=0.9,
momentum=0.9, epsilon=1.0)
        elif optimizer == 'MOM':
            opt = tf.train.MomentumOptimizer(learning rate, 0.9,
use nesterov=True)
        else:
            raise ValueError('Invalid optimization algorithm')
        grads = opt.compute gradients(total loss, update gradient vars)
```

```
# Apply gradients.
    apply gradient op = opt.apply gradients(grads,
global step=global step)
    # Add histograms for trainable variables.
    if log histograms:
        for var in tf.trainable variables():
            tf.summary.histogram(var.op.name, var)
    # Add histograms for gradients.
    if log histograms:
        for grad, var in grads:
            if grad is not None:
                tf.summary.histogram(var.op.name + '/gradients', grad)
    # Track the moving averages of all trainable variables.
   variable averages = tf.train.ExponentialMovingAverage(
        moving average decay, global step)
   variables averages op =
variable_averages.apply(tf.trainable_variables())
    with tf.control dependencies([apply gradient op,
variables averages op]):
        train op = tf.no op(name='train')
    return train op
```

Then, I proceed to declare a data train function so that the system can learn the existing face data in the system with the help of tensorflow.

```
def prewhiten(x):
    mean = np.mean(x)
    std = np.std(x)
    std_adj = np.maximum(std, 1.0 / np.sqrt(x.size))
    y = np.multiply(np.subtract(x, mean), 1 / std_adj)
    return y
```

```
def load data(image paths, do random crop, do random flip, image size,
do prewhiten=True):
    nrof samples = len(image paths)
    images = np.zeros((nrof samples, image size, image size, 3))
    for i in range(nrof samples):
        img = misc.imread(image paths[i])
        if img.ndim == 2:
            img = to rgb(img)
        if do prewhiten:
            img = prewhiten(img)
        img = crop(img, do_random_crop, image size)
        img = flip(img, do_random flip)
        images[i, :, :, :] = img
    return images
def get label batch(label data, batch size, batch index):
    nrof examples = np.size(label data, 0)
    j = batch index * batch size % nrof examples
    if j + batch size <= nrof examples:</pre>
        batch = label data[j:j + batch size]
    else:
        x1 = label data[j:nrof examples]
        x2 = label data[0:nrof examples - j]
        batch = np.vstack([x1, x2])
    batch int = batch.astype(np.int64)
    return batch int
def get batch(image data, batch size, batch index):
    nrof_examples = np.size(image data, 0)
    j = batch index * batch size % nrof examples
    if j + batch size <= nrof examples:</pre>
```

```
batch = image data[j:j + batch size, :, :, :]
    else:
        x1 = image data[j:nrof examples, :, :, :]
        x2 = image data[0:nrof examples - j, :, :, :]
        batch = np.vstack([x1, x2])
   batch float = batch.astype(np.float32)
    return batch float
def get triplet batch(triplets, batch index, batch size):
    ax, px, nx = triplets
   a = get batch(ax, int(batch size / 3), batch index)
   p = get batch(px, int(batch size / 3), batch index)
   n = get batch(nx, int(batch size / 3), batch index)
   batch = np.vstack([a, p, n])
    return batch
def get learning rate from file(filename, epoch):
   with open(filename, 'r') as f:
        for line in f.readlines():
            line = line.split('#', 1)[0]
            if line:
                par = line.strip().split(':')
                e = int(par[0])
                if par[1] == '-':
                    lr = -1
                else:
                    lr = float(par[1])
                if e <= epoch:</pre>
                    learning rate = lr
                else:
                    return learning_rate
```

Declare the above functions so that the system can calculate the learning rate. This value is a required value to be able to train data

```
class ImageClass():
    "Stores the paths to images for a given class"
    def init (self, name, image paths):
       self.name = name
        self.image paths = image_paths
    def str (self):
        return self.name + ', ' + str(len(self.image paths)) + ' images'
    def len (self):
        return len(self.image paths)
def load model(model, input map=None):
    # Check if the model is a model directory (containing a metagraph and
a checkpoint file)
    # or if it is a protobuf file with a frozen graph
   model exp = os.path.expanduser(model)
    if (os.path.isfile(model exp)):
        print('Model filename: %s' % model exp)
        with gfile.FastGFile(model exp, 'rb') as f:
            graph def = tf.GraphDef()
            graph def.ParseFromString(f.read())
            tf.import graph def(graph def, input map=input map, name='')
    else:
        print('Model directory: %s' % model exp)
        meta file, ckpt file = get model filenames(model exp)
        print('Metagraph file: %s' % meta file)
       print('Checkpoint file: %s' % ckpt file)
        saver = tf.train.import_meta_graph(os.path.join(model_exp,
meta file), input map=input map)
        saver.restore(tf.get default session(), os.path.join(model exp,
ckpt file))
```

```
def get model filenames(model dir):
   files = os.listdir(model dir)
   meta files = [s for s in files if s.endswith('.meta')]
   if len(meta files) == 0:
        raise ValueError('No meta file found in the model directory (%s)'
% model dir)
   elif len(meta files) > 1:
        raise ValueError('There should not be more than one meta file in
the model directory (%s)' % model dir)
   meta file = meta files[0]
   ckpt = tf.train.get checkpoint state(model dir)
   if ckpt and ckpt.model checkpoint path:
        ckpt file = os.path.basename(ckpt.model checkpoint path)
       return meta file, ckpt file
   meta files = [s for s in files if '.ckpt' in s]
   \max \text{ step} = -1
   for f in files:
       step_str = re.match(r'(^model-[\w\-]+.ckpt-(\d+))', f)
        if step_str is not None and len(step_str.groups()) >= 2:
            step = int(step str.groups()[1])
            if step > max step:
                max step = step
                ckpt file = step str.groups()[0]
   return meta file, ckpt file
```

After the system has finished training the data, the newly trained memorized information will be stored in a model. The above function helps to save the model to the storage directory.

```
def distance(embeddings1, embeddings2, distance_metric=0):
    if distance_metric == 0:
        # Euclidian distance
        diff = np.subtract(embeddings1, embeddings2)
        dist = np.sum(np.square(diff), 1)
    elif distance_metric == 1:
```

```
# Distance based on cosine similarity
dot = np.sum(np.multiply(embeddings1, embeddings2), axis=1)
norm = np.linalg.norm(embeddings1, axis=1) *
np.linalg.norm(embeddings2, axis=1)
similarity = dot / norm
dist = np.arccos(similarity) / math.pi
else:
    raise 'Undefined distance metric %d' % distance_metric
return dist
```

Next, I will declare a function that calculates and returns the value of the distance between the vector and the trained vectors. The smaller the distance between the vectors, the greater the similarity.

```
def calculate roc(thresholds, embeddings1, embeddings2, actual issame,
nrof folds=10, distance_metric=0,
                  subtract mean=False):
   assert (embeddings1.shape[0] == embeddings2.shape[0])
    assert (embeddings1.shape[1] == embeddings2.shape[1])
    nrof pairs = min(len(actual issame), embeddings1.shape[0])
   nrof thresholds = len(thresholds)
    k fold = KFold(n splits=nrof folds, shuffle=False)
    tprs = np.zeros((nrof folds, nrof thresholds))
    fprs = np.zeros((nrof folds, nrof thresholds))
    accuracy = np.zeros((nrof folds))
    indices = np.arange(nrof pairs)
    for fold idx, (train set, test set) in
enumerate(k fold.split(indices)):
        if subtract mean:
            mean = np.mean(np.concatenate([embeddings1[train set],
embeddings2[train_set]]), axis=0)
        else:
           mean = 0.0
```

```
dist = distance(embeddings1 - mean, embeddings2 - mean,
distance metric)
        # Find the best threshold for the fold
        acc train = np.zeros((nrof thresholds))
        for threshold idx, threshold in enumerate(thresholds):
            _, _, acc_train[threshold_idx] = calculate accuracy(threshold,
dist[train set], actual issame[train set])
        best threshold index = np.argmax(acc train)
        for threshold idx, threshold in enumerate(thresholds):
            tprs[fold idx, threshold idx], fprs[fold idx, threshold idx],
= calculate accuracy(threshold,
dist[test set],
actual issame[
test set])
        _, _, accuracy[fold_idx] =
calculate accuracy(thresholds[best threshold index], dist[test set],
actual issame[test_set])
        tpr = np.mean(tprs, 0)
        fpr = np.mean(fprs, 0)
    return tpr, fpr, accuracy
def calculate accuracy(threshold, dist, actual issame):
    predict_issame = np.less(dist, threshold)
    tp = np.sum(np.logical and(predict issame, actual issame))
    fp = np.sum(np.logical and(predict issame,
np.logical not(actual issame)))
    tn = np.sum(np.logical and(np.logical not(predict issame),
np.logical not(actual issame)))
    fn = np.sum(np.logical and(np.logical not(predict issame),
actual issame))
```

```
tpr = 0 if (tp + fn == 0) else float(tp) / float(tp + fn)
    fpr = 0 if (fp + tn == 0) else float(fp) / float(fp + tn)
    acc = float(tp + tn) / dist.size
   return tpr, fpr, acc
def calculate val(thresholds, embeddings1, embeddings2, actual issame,
far target, nrof folds=10, distance metric=0,
                  subtract mean=False):
    assert (embeddings1.shape[0] == embeddings2.shape[0])
    assert (embeddings1.shape[1] == embeddings2.shape[1])
   nrof_pairs = min(len(actual_issame), embeddings1.shape[0])
   nrof thresholds = len(thresholds)
   k fold = KFold(n splits=nrof folds, shuffle=False)
   val = np.zeros(nrof folds)
   far = np.zeros(nrof folds)
    indices = np.arange(nrof pairs)
    for fold_idx, (train_set, test_set) in
enumerate(k fold.split(indices)):
        if subtract mean:
            mean = np.mean(np.concatenate([embeddings1[train set],
embeddings2[train set]]), axis=0)
        else:
           mean = 0.0
        dist = distance(embeddings1 - mean, embeddings2 - mean,
distance metric)
        # Find the threshold that gives FAR = far target
        far train = np.zeros(nrof thresholds)
        for threshold idx, threshold in enumerate(thresholds):
            , far train[threshold idx] = calculate val far(threshold,
dist[train set], actual issame[train set])
        if np.max(far train) >= far target:
            f = interpolate.interp1d(far train, thresholds,
kind='slinear')
```

```
threshold = f(far target)
        else:
            threshold = 0.0
        val[fold idx], far[fold idx] = calculate val far(threshold,
dist[test set], actual issame[test set])
   val mean = np.mean(val)
   far mean = np.mean(far)
   val std = np.std(val)
    return val mean, val std, far mean
def calculate val far(threshold, dist, actual issame):
   predict issame = np.less(dist, threshold)
    true accept = np.sum(np.logical and(predict issame, actual issame))
    false accept = np.sum(np.logical and(predict issame,
np.logical not(actual issame)))
    n same = np.sum(actual issame)
    n diff = np.sum(np.logical not(actual issame))
   val = float(true accept) / float(n same)
    far = float(false accept) / float(n diff)
   return val, far
def list variables(filename):
    reader = training.NewCheckpointReader(filename)
   variable map = reader.get variable to shape map()
   names = sorted(variable map.keys())
    return names
def write arguments to file(args, filename):
   with open(filename, 'w') as f:
        for key, value in iteritems(vars(args)):
            f.write('%s: %s\n' % (key, str(value)))
```

Based on the distance value that the system has calculated, I have implemented functions to help calculate the percentage similarity of the input image compared to the face images that have been trained through the segments above code.

align_mtcnn.py :

```
def align mtcnn(input dir,
                  output dir,
                  image size=182,
                  margin=44,
                  random order=None,
                  gpu memory fraction=1.0,
                  detect multiple faces=False):
    sleep(random.random())
    output dir = os.path.expanduser(output dir)
    if not os.path.exists(output dir):
        os.makedirs(output dir)
    # Store some git revision info in a text file in the log directory
    src path, = os.path.split(os.path.realpath( file ))
    store_revision_info(src_path, output dir, ' '.join(sys.argv))
    dataset = get dataset(input dir)
   print('Creating networks and loading parameters')
   with tf.Graph().as default():
        gpu options =
tf.GPUOptions(per process gpu memory fraction=gpu memory fraction)
        sess = tf.Session(config=tf.ConfigProto(qpu options=qpu options,
log device placement=False))
        with sess.as default():
            pnet, rnet, onet = create mtcnn(sess, None)
   minsize = 20 # minimum size of face
    threshold = [0.6, 0.8, 0.8] # three steps's threshold
    factor = 0.709 # scale factor
    # Add a random key to the filename to allow alignment using multiple
processes
    random key = np.random.randint(0, high=99999)
   bounding boxes filename = os.path.join(output dir,
'bounding boxes %05d.txt' % random key)
    with open (bounding boxes filename, "w") as text file:
        nrof images total = 0
        nrof successfully aligned = 0
        if random order:
            random.shuffle(dataset)
        for cls in dataset:
            output class dir = os.path.join(output dir, cls.name)
            if not os.path.exists(output class dir):
                os.makedirs(output class dir)
                if random order:
                    random.shuffle(cls.image paths)
            for image path in cls.image paths:
                nrof_images_total += 1
                filename =
os.path.splitext(os.path.split(image path)[1])[0]
                output filename = os.path.join(output class dir, filename
+ '.png')
                print(image path)
```

```
if not os.path.exists(output filename):
                    try:
                         img = misc.imread(image path)
                    except (IOError, ValueError, IndexError) as e:
                        errorMessage = '{}: {}'.format(image path, e)
                        print(errorMessage)
                    else:
                         if img.ndim < 2:
                             print('Unable to align "%s"' % image path)
                             text file.write('%s\n' % (output filename))
                             continue
                         if img.ndim == 2:
                             img = to rgb(img)
                         img = img[:, :, 0:3]
                        bounding_boxes, _ = detect_face(img, minsize,
pnet, rnet, onet, threshold, factor)
                        nrof faces = bounding boxes.shape[0]
                        if nrof faces > 0:
                             det = bounding boxes[:, 0:4]
                             det arr = []
                             img size = np.asarray(img.shape)[0:2]
                             if nrof faces > 1:
                                 if detect multiple faces:
                                    for i in range(nrof faces):
                                         det arr.append(np.squeeze(det[i]))
                                 else:
                                     bounding box size = (det[:, 2] -
det[:, 0]) * (det[:, 3] - det[:, 1])
                                     img center = img size / 2
                                     offsets = np.vstack([(det[:, 0] +
det[:, 2]) / 2 - img center[1],
                                                          (det[:, 1] +
det[:, 3]) / 2 - img center[0]])
                                     offset dist squared =
np.sum(np.power(offsets, 2.0), 0)
                                     index = np.argmax(
                                        bounding box size -
offset dist squared * 2.0) # some extra weight on the centering
                                     det arr.append(det[index, :])
                             else:
                                 det arr.append(np.squeeze(det))
                             for i, det in enumerate(det arr):
                                 det = np.squeeze(det)
                                 bb = np.zeros(4, dtype=np.int32)
                                 bb[0] = np.maximum(det[0] - margin / 2, 0)
                                 bb[1] = np.maximum(det[1] - margin / 2, 0)
                                 bb[2] = np.minimum(det[2] + margin / 2,
img size[1])
                                 bb[3] = np.minimum(det[3] + margin / 2,
img size[0])
                                 cropped = img[bb[1]:bb[3], bb[0]:bb[2], :]
                                 scaled = misc.imresize(cropped,
(image size, image size), interp='bilinear')
                                 nrof successfully aligned += 1
                                 filename base, file extension =
```

This is the function that executes the MTCNN algorithm that has been initialized in the detect_face.py file. When executing this function, the system will proceed to use the MTCNN algorithm and the detect_face function I declared above to detect the face coordinates in the input image. I have set thresholds for face detection to be 0.6, 0.8, 0.8 respectively. With such a threshold ratio, the system will certainly not be able to miss any faces in the image.

Next, the system will proceed to cut the faces in the input and output images for easy data training, the coordinates of the detected faces will be saved in a .txt file in the face_align.. folder.

face_contrib.py:

```
class Face:
    def __init__(self):
        self.name = None
        self.bounding_box = None
        self.image = None
        self.container_image = None
        self.embedding = None
```

First, I create a class Face to hold variables that store face information

```
class Recognition:
    def __init__(self, facenet_model_checkpoint, classifier_model):
        self.detect = Detection()
        self.encoder = Encoder(facenet_model_checkpoint)
        self.identifier = Identifier(classifier_model)
```

```
def add identity(self, image, person name):
        faces = self.detect.find faces(image)
        if len(faces) == 1:
            face = faces[0]
            face.name = person name
            face.embedding = self.encoder.generate embedding(face)
            return faces
    def identify(self, image):
        faces = self.detect.find faces(image)
        for i, face in enumerate(faces):
            if debug:
                cv2.imshow("Face: " + str(i), face.image)
            face.embedding = self.encoder.generate embedding(face)
            face.name, face.prob = self.identifier.identify(face)
        return faces
class Identifier:
    def init (self, classifier model):
        with open(classifier model, 'rb') as infile:
            self.model, self.class names = pickle.load(infile)
    def identify(self, face):
        if face.embedding is not None:
            predictions = self.model.predict proba([face.embedding])
            best class indices = np.argmax(predictions, axis=1)
            return self.class names[best class indices[0]],
predictions[0][best class indices[0]]
```

Next, I declare the Recognition and Identifier classes to assign identifiers to the recognized faces, if that face is the face of an existing face in the model, that face's identifier is the corresponding label. response.

```
class Encoder:
    def init (self, facenet model checkpoint):
       self.sess = tf.Session()
       with self.sess.as default():
            load model(facenet model checkpoint)
    def generate embedding(self, face):
        # Get input and output tensors
        images placeholder =
tf.get default graph().get tensor by name("input:0")
        embeddings =
tf.get default graph().get tensor by name("embeddings:0")
       phase train placeholder =
tf.get default graph().get tensor by name("phase train:0")
       prewhiten face = prewhiten(face.image)
        # Run forward pass to calculate embeddings
        feed dict = {images placeholder: [prewhiten face],
phase train placeholder: False}
        return self.sess.run(embeddings, feed dict=feed dict)[0]
```

Above is the code to convert the images directly from the camera into embedding vectors so that we can calculate the distance between these vectors and the vectors already in the model.

```
class Detection:
    # face detection parameters
    minsize = 20  # minimum size of face
    threshold = [0.65, 0.7, 0.7]  # three steps's threshold
    factor = 0.709  # scale factor

def __init__(self, face_crop_size=160, face_crop_margin=32):
    self.pnet, self.rnet, self.onet = self._setup_mtcnn()
    self.face_crop_size = face_crop_size
    self.face_crop_margin = face_crop_margin
```

```
def setup mtcnn(self):
        with tf.Graph().as default():
            gpu options =
tf.GPUOptions(per process gpu memory fraction=gpu memory fraction)
            sess = tf.Session(config=tf.ConfigProto(gpu options=gpu options,
log device placement=False))
            with sess.as default():
                return create mtcnn(sess, None)
    def find faces(self, image):
        faces = []
        bounding boxes, = detect face(image, self.minsize,
                                                           self.pnet,
self.rnet, self.onet,
                                                           self.threshold,
self.factor)
        for bb in bounding boxes:
            face = Face()
            face.container image = image
            face.bounding box = np.zeros(4, dtype=np.int32)
            img size = np.asarray(image.shape)[0:2]
            face.bounding box[0] = np.maximum(bb[0] - self.face_crop_margin / 
2, 0)
            face.bounding box[1] = np.maximum(bb[1] - self.face crop margin /
2, 0)
            face.bounding box[2] = np.minimum(bb[2] + self.face crop margin /
2, img_size[1])
            face.bounding box[3] = np.minimum(bb[3] + self.face crop margin /
2, img size[0])
            cropped = image[face.bounding_box[1]:face.bounding_box[3],
face.bounding box[0]:face.bounding box[2], :]
            face.image = misc.imresize(cropped, (self.face crop size,
self.face crop size), interp='bilinear')
            faces.append(face)
```

Finally, here is the code to help execute the calculation function applying the triplet loss function declared in the facenet.py file. After the calculation is complete, the distance between the two vectors will be returned and compared with the set thresholds of 0.6,0.7, and 0.7. If this threshold is exceeded, the label value assigned to that vector will be returned.

train.py:

```
def train(data dir,
        model,
        classifier filename,
        use split dataset=None,
        batch size=1000,
        image size=160,
        seed=\overline{123}
        min nrof images per class=20,
        nrof train images per class=10):
    with tf.Graph().as default():
        with tf.Session() as sess:
            np.random.seed(seed=seed)
            if use split dataset:
                dataset tmp = get dataset(data dir)
                train set, test set = split dataset(dataset tmp,
min nrof images per class,
nrof train images per class)
                dataset = train set
            else:
                dataset = get dataset(data dir)
            # Check that there are at least one training image per class
            for cls in dataset:
                assert (len(cls.image paths) > 0)
                messagebox.showerror('Error', 'There must be at least one
image for each class in the dataset')
            paths, labels = get image paths and labels(dataset)
            messagebox.showinfo("Trainning Process",f'Number of people: %d'%
len (dataset))
            messagebox.showinfo("Trainning Process", f'Number of images: %d'
% len(paths))
            # Load the model
            load model(model)
            # Get input and output tensors
            images placeholder =
tf.get default graph().get tensor by name("input:0")
            embeddings =
```

```
tf.get default graph().get tensor by name("embeddings:0")
            phase train placeholder =
tf.get default graph().get tensor by name("phase train:0")
            embedding size = embeddings.get shape()[1]
            # Run forward pass to calculate embeddings
            messagebox.showinfo('Training Process', 'Calculating features for
images')
            nrof images = len(paths)
            nrof batches per epoch = int(math.ceil(1.0 * nrof images /
batch size))
            emb array = np.zeros((nrof images, embedding size))
            for i in range (nrof batches per epoch):
                    start index = i * batch size
                    end_index = min((i + 1) * batch_size, nrof_images)
                    paths batch = paths[start index:end index]
                    images = load data(paths batch, False, False, image size)
                    feed dict = {images placeholder: images,
phase train placeholder: False}
                    emb array[start index:end index, :] =
sess.run(embeddings, feed dict=feed dict)
            classifier filename exp = os.path.expanduser(classifier filename)
            # ============== Train classifier ================
           messagebox.showinfo('Training Conduct', 'Training classifier')
           model = SVC(kernel='linear', probability=True)
           model.fit(emb array, labels)
            # Create a list of class names
           class_names = [cls.name.replace('_', ' ') for cls in dataset]
            # Saving classifier model
            with open(classifier filename exp, 'wb') as outfile:
               pickle.dump((model, class names), outfile)
           messagebox.showinfo('Trainning Finish','Data has been trained
successfully !!!')
if name == ' main ':
    align mtcnn('dataSet', 'face align')
    train('face align/', 'models/20180402-114759.pb',
'models/your model.pkl')
```

The file train.py is the file that executes the data train functions that have been initialized in the system. When conducting data training, the system will proceed to take the face images detected by MTCNN, process them and then pass them to the Facenet algorithm for processing.

Set condition is that there must be more than two faces for the system to start training. When conducting the training process, the system will display a message about how many people and how many pictures. Then the system loads the model to use for storage. After the

training is complete, the system will display a success message and save the data containing the vectors that have been labeled with the face into the model.

detection.py :

```
def attendance(i, n, c, f):
    with open('attendance.csv', 'r+', newline='\n') as file:
        DataList = file.readlines()
        namelist = []
        for data in DataList:
            ent = data.split(',')
            namelist.append(ent[0])
        if ((i not in namelist) and (n not in namelist) and (c not in namelist) and (f not in namelist)):
            curr = datetime.now()
            dt = curr.strftime("%d/%m/%Y")
            h = curr.strftime('%H:%M:%S')
            file.writelines(f'\n{i},{n},{c},{f},{dt},{dt}),
```

In the file detection.py, I initialized a attendance function to remember the information of the recognized face. The recognized faces will be automatically stored information and recognized date and time in the file attendance.csv.

```
user="root", password='Vilong242',
                                                   db='face recognition')
                    cur = conn.cursor()
                    cur.execute("select Student ID from student where
Student ID=" + str(id))
                    i = cur.fetchone()
                    i = '+'.join(i)
                    cur.execute("select Name from student where Student_ID="
+ str(id))
                   n = cur.fetchone()
                    n = '+'.join(n)
                    cur.execute("select Class from student where Student ID="
+ str(id))
                    c = cur.fetchone()
                    c = '+'.join(c)
                    cur.execute("select Faculty from student where
Student ID=" + str(id))
                    f = cur.fetchone()
                    f = '+'.join(f)
                    cv2.putText(frame, f'Student ID:{i}', (face bb[0],
face bb[3] + 20), cv2.FONT HERSHEY COMPLEX, 0.6, colors[idx], thickness=2,
lineType=1)
                    cv2.putText(frame, f'Name:{n}', (face bb[0], face bb[3] +
45), cv2.FONT HERSHEY COMPLEX, 0.6, colors[idx], thickness=2, lineType=1)
                    cv2.putText(frame, f'Class:{c}', (face bb[0], face bb[3]
+ 70), cv2.FONT HERSHEY COMPLEX, 0.6, colors[idx], thickness=2, lineType=1)
                    cv2.putText(frame, f'Faculty:{f}', (face bb[0],
face bb[3] + 95), cv2.FONT HERSHEY COMPLEX, 0.6, colors[idx], thickness=2,
lineType=1)
                    cv2.putText(frame, '{:.02f}'.format(face.prob * 100),
(face bb[0], face bb[3] + 120), cv2.FONT HERSHEY SIMPLEX, 0.6, colors[idx],
thickness=1, lineType=2)
```

The above code executes the display of the basic information of the face stored in the database when the system recognizes it. The information stored in the database will be retrieved and displayed on the identification screen. If the face is not recognized, the system will return the result Unknown People.

```
def run(model checkpoint, classifier, output file=None):
    frame interval = 3 # Number of frames after which to run face detection
    fps display interval = 5 # seconds
    frame rate = 0
    frame count = 0
    camera = cv2.VideoCapture(0)
    ret, frame = camera.read()
    width = frame.shape[1]
    height = frame.shape[0]
    if output file is not None:
        video format = cv2.VideoWriter fourcc(*'XVID')
        out = cv2.VideoWriter(output file, video format, 20, (width, height))
    face recognition = Recognition(model checkpoint, classifier)
    start time = time.time()
    colors = np.random.uniform(0, 255, size=(1, 3))
    while True:
        # Capture frame-by-frame
        ret, frame = camera.read()
        if (frame count % frame interval) == 0:
            faces = face recognition.identify(frame)
```

```
for i in range(len(colors), len(faces)):
                colors = np.append(colors, np.random.uniform(150, 255,
size=(1, 3)), axis=0)
            # Check our current fps
            end time = time.time()
            if (end time - start time) > fps display interval:
                frame rate = int(frame count / (end time - start time))
                start time = time.time()
                frame count = 0
        add overlays(frame, faces, frame rate, colors)
        frame count += 1
        cv2.imshow('Welcome to school', frame)
        if output file is not None:
            out.write(frame)
        if cv2.waitKey(1) & 0xFF == ord('q'):
           break
    # When everything is done, release the capture
    if output file is not None:
       out.release()
    camera.release()
    cv2.destroyAllWindows()
if name == ' main ':
    run('models', 'models/your_model.pkl')
```

The above code executes the loading of data in the model to launch the declared calculation functions. The end purpose is to return the result as the value of that face.