

BCA MINOR PROJECT

FACIAL RECOGNITION SYSTEM

FINAL YEAR

Graphical User Interface :

tkinter : It is a collection of many widgets and it is used to create a GUI, using Python.

Step 1 : At first we call a function called Tk()

Step 2 : We resize our root screen -

```
root = Tk()
root.geometry ("675x500+120+120")
```

w = width
 H = Height
 x = x axis
 y = y axis

Step 3 : We fix the root screen size using min_size() and max_size().

Step 4 : We set our root wallpaper. We downloaded a external picture from internet. After downloading the image we access it using pillow module in python. It supports a range of image file formats such as PNG, JPEG, PPM, GIF, TIFF and BMP.

step 6 : We put the image as a background image of the canvas.

step 7 : After this step we set our GUI title, and we also set the icon of our GUI.

customized icon
the format is = .ico

Hons.

```
b = Button(root, text = "button text", font = ("font family",  
"font size", "font style"), fg = "font colour",  
padx = "Horizontal Gap of button text", pady = "Vertical  
Gap of button text", relief = "type of the button",  
bg = "button background colour", command = lambda:  
[fn(), fm(), fr2(), ...])
```

↳ Multiple functions

Step-9 : Place all the buttons in our GUI at specific position of the canvas.

button_place = canvas.create_window (x axis, y axis,
window = button)

Note:

For producing the clicking sound we used win sound module and create a lambda function -

clicked = lambda : Play Sound (random.choice(click list),
SND-FILENAME)

↓
Inbuilt function
of win sound module

↓
custom list
of external
clicking audio
files names

↓
Type of the choice
string which is
randomly choose
by the random.
choice()

Step 10 : After create all GUI widgets we will call the main loop function. It is used to call the all pack widgets which we added in our previous steps.

GUI COMPLETE

FOLDER STRUCTURE

- Total 3 Programs → Machine Learning
(Superset of deep learning)
1. face-sample-collects.py
 2. model-training.py
 3. Recognizer.py
- GUI Portion : 1 Program
4. GUI.py

DEEP LEARNING PORTION

1. Face sample collects : -
 - Step 1 : At first we import the necessary modules
open cv, mumpy. open cv stands for open
computer vision it is use to process image from
webcam, live detection.
 - Step-2 : Then we create the face classifier to
detect our face from the web camera frame.
we download the external extensible markup file
(xml) named 'haarcascade-frontalface-default.xml'
 - Step-3 : We create a function for extract our
face and create our own dataset by
collecting our images total 150 images.
- gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
- ↓
We convert the face image which is in Blue,
Green, Red format to GRAY scale.

We use `cv2.VideoCapture()` to stream our video by using webcam for saving total 150 face samples. We use `os` module in python, using this module at first we specified our path of single image file which is save in jpg format, then we use `cv2.imwrite()` to save the face sample at the image file location, after complete all the backend works then we use `cv2.putText()` to put the counter text in the face cropper window which is created by `cv2`, if no faces present in front of our webcam then face cropper will not open and print "face not found" in debugging console.

Note:
for (x, y, w, h) in faces:
`cropped-face = img[y:y+h, x:x+w]`

It is a logic for cropping our face using the face classifier which we have created in the previous portion in step 3.

After we done with all these things we wrap up all these things into a function named.

`fsc()`.

Then the function will call in the button of our GUI portion at the command property, with the clicked function simultaneously.

2. Model_training :

Step 1 : At first we categorize our own created dataset which contains total 150 face samples.

Training_Data, Labels = [], []

Every
images store as numpy
array using `np.asarray()`

↓
store 150
face samples
as grey scale
image [type
is unsynchronized
integer 8]

Step-2 : We labeled our face samples according their filenames which is stored in the 'faces' folder at the project directory.

Step-3 : We create our model.

`model = cv2.face.LBPHFaceRecognizer_create()`

(*) **LBPH = Local Binary Pattern Histogram.**

It is a algorithm which is used to recognize the face of a person it is known for its performance and how its able to recognize the face of a person from both front and side face.

This is inbuilt in open cv2 module.

Step-4 : After creating our model we train it using train function.

model.train(np.asarray(Training-Data), np.asarray((Labels)))

TRAINING COMPLETE

Recognizer :-

Step-1 : We create a function named face detector() which use to retake live faces from the frame of our webcam.

for (x, y, w, h) in faces:

cv2.rectangle(img, (x, y), (x+w, y+h), (0, 255, 255), 2)

roi = img[y:y+h, x:x+w]

roi = cv2.resize(roi, (200, 200))

→ This portion is used for fetching our face from live frame of our webcam.

Step-2 : After fetching our face from camera we predicted our face is authorized face or not using model.predict().

Step-3 : After predicting we create a variable result and store our prediction into this variable. Next we do

$$\text{confidence} = \text{int}(100 * (1 - (\text{result}[1]) / 300))$$

↓
This variable assign the percentage of our model accuracy which is calculated by above simple logic. After that we can typecast this to integer.

Step-4 : If the confidence variable's value is more than 83 then we use cv2.putText() to print

'It is a authorized user'.

'UNLOCKED'

If no faces found then print 'face not found'.

If the confidence variables value's is not more than 83 then print

'Fake User'.

'LOCKED'

Step-5 : We can set our own title in the live detect screen. Using cv2.imshow('Face Recognition', image)



image, face = face_detector(frame)



Face detector function is already discussed in previous steps.

Recognizer.py

⊗ Note :

After we done all these things we wrap up it into a function named *recog()*.

Model-training.py

⊗ Note :

After we done all these things we wrap up it into a function named *mt()*.

After wrapping all these things these functions will call of our GUI portion at the command property with the clicked function simultaneously.

DEEP LEARNING COMPLETE

Reset :-

os module is used to provide functions for interacting with Operating System.

```
from plyer import notification
import os

def AllFilesDelete():
    dir = 'faces/'
    for f in os.listdir(dir):
        os.remove(os.path.join(dir, f))
    print("RESETTING ALL IMAGES IN
    FACES FOLDER....")
```

plyer module is used to generate notifications.

```
notification.notify (
```

```
    title = "RESETTING FINISHED",
    message = "ALL FACE SAMPLES ARE
    DELETED",
```

This notify function is used for designing our notification which will generate when our all face samples are deleted.

```
# display time
    timeout = 2
)
```

This property is used to generate our notification after our given time. (here we set the time 2 seconds).

This AllFilesDelete() is called when we click the 'reset' button in our GUI.

RESET COMPLETE.

PROJECT COMPLETE