

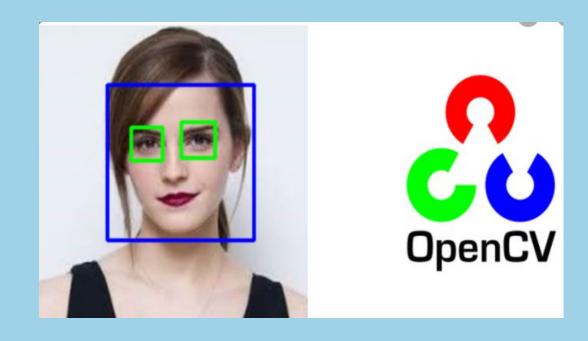
MASK DETECTION

SHIRIN KOUL

Idea: The concepts of deep
learning will
Be used detect whether a person
is wearing mask.

Methodology:

- 1. Detect face (using Open CV)
 - 2. 2. Detect Mask



TECHNOLOGY:

- 1. Language used will be python.
 - 2. Open CV for face detection.
 - 3. TensorFlow
 - 4. Keras
 - 5. MobileNet

Learning

- 1. Basics of python
 - 2. Variables
 - 3. Lists
 - 4. Tuples
 - 5. Sets
 - 6. Variables

NumPy

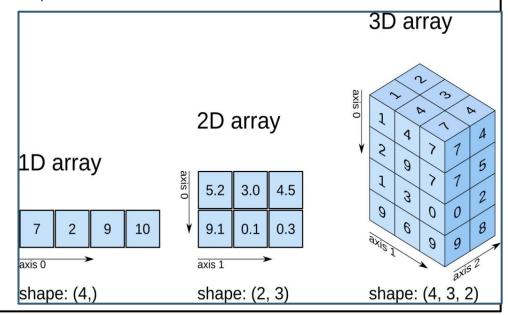
NumPy, which stands for Numerical Python, is a library consisting of multidimensional array objects and a collection of routines for processing those arrays Numpy array has the various function, methods, and variables, to ease our task of matrix computation.

Advantages of using Numpy Arrays Over Python Lists:

consumes less memory.

fast as compared to the python List.

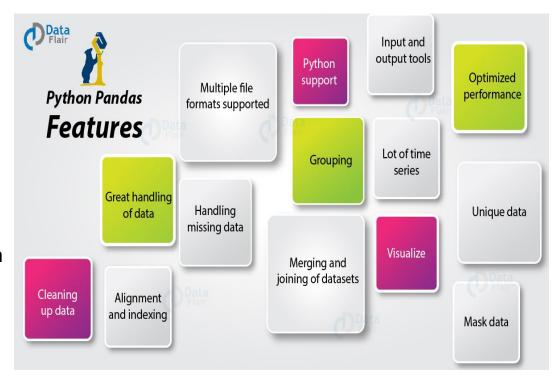
convenient to use.



Pandas

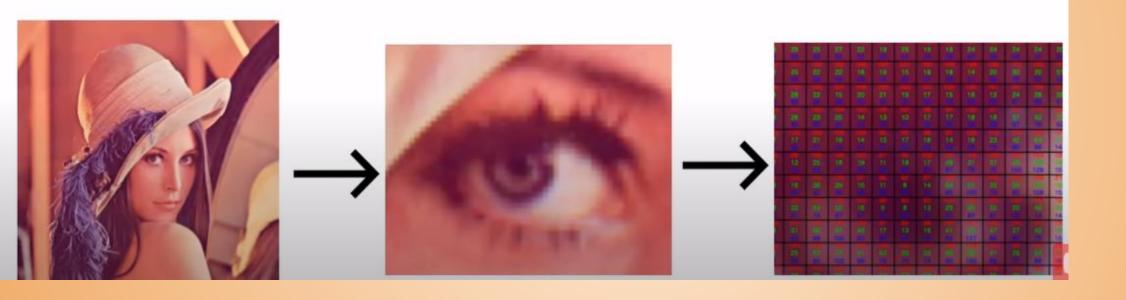
Pandas is an open-source, BSD-licensed
Python library providing high-performance,
easy-to-use data structures and data
analysis tools for the Python programming
language.

It is mainly popular for importing and analyzing data much easier. Pandas is fast and it has high-performance & productivity for users.



OPEN CV:

Introduction to Images



Binary Image



2 Levels

0 = Black | 1 = White |

1	1	1	1	1	1	1	1	1	1
1	1	1	1	1	1	1	1	1	1
1	1	1	G	0	0	0	٢	1	1
1	1	1	1	1	1	0	1	1	1
1	1	1	1	1	1	C	1	1	1
1	1	1	C	O	O	0	1	1	1
1	1	1	1	1	1	0	1	1	1
1	1	1	1	1	1	()	1	1	1
1	1	1	C	C	C	0	1	1	1
1	1	1	1	1	1	1	1	1	1

Gray Scale Image- 8 bit or 256 levels



2 Levels

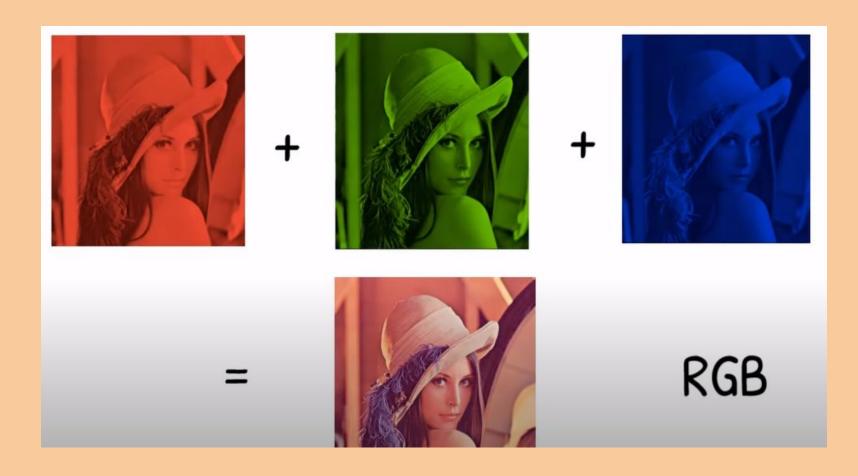


6 Levels



16 Levels

Colored Image has 3 levels. Each level is accountable for different colors amongst red blue and green which results in a colored image.

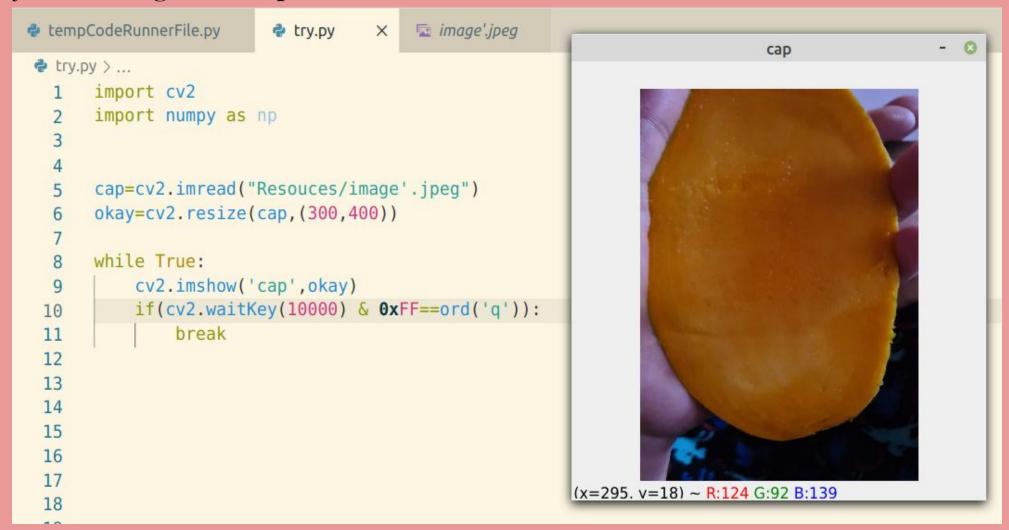


Images are stored in OpenCV as numpy nd array.

1. Viewing an image -

Its a 2 step process in OpenCV. Firstly, we read an image using opency and then we show it using the function imshow().

We can also provide a parameter in imread function to directly convert it into grayscale image or keep as it is.



2. Video Capturing - Open CV is used in capturing video using webcam or can play any video which can also be edited using Opencv functions.

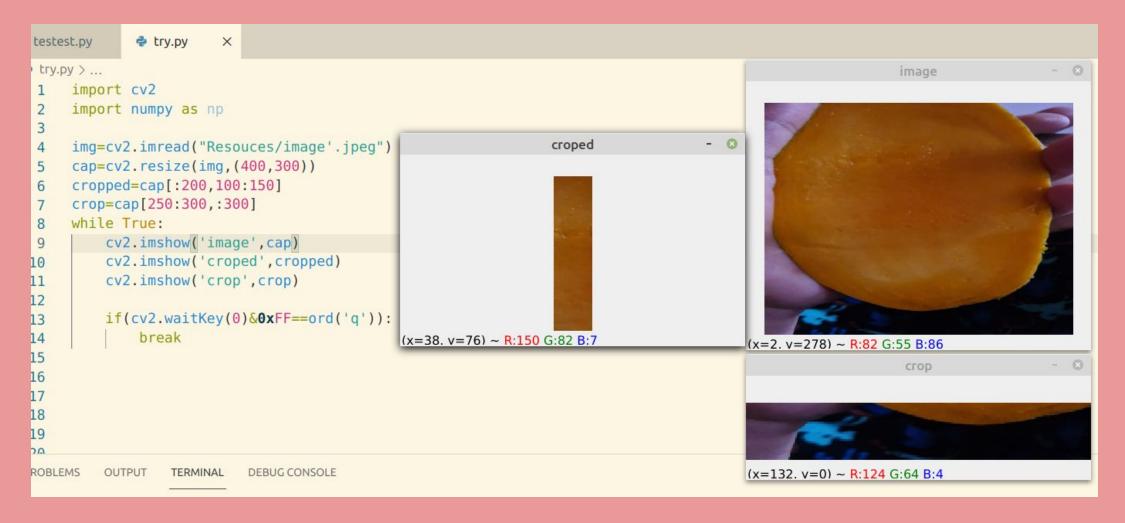
Video is basically a collection of images viewed really fast hence we use loop to display that.

3. Image Editing -Image can be transformed into different kind of image like colored into grey scale and vice versa or into Canny image or blurred image just by using a simple function cvtColor() where it takes the name of the image that needs to be converted as a parameter with the conversion it has to get.

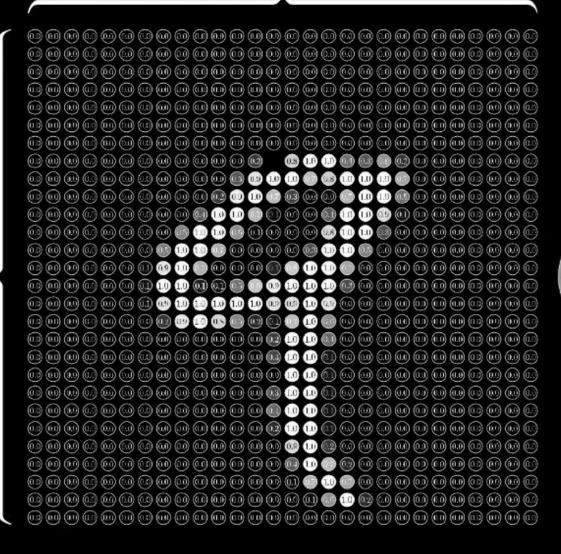
4. Resizing image- Function resize which takes the image and the final dimension as parameters in order to resize an image.



5. Cropping an Image



28

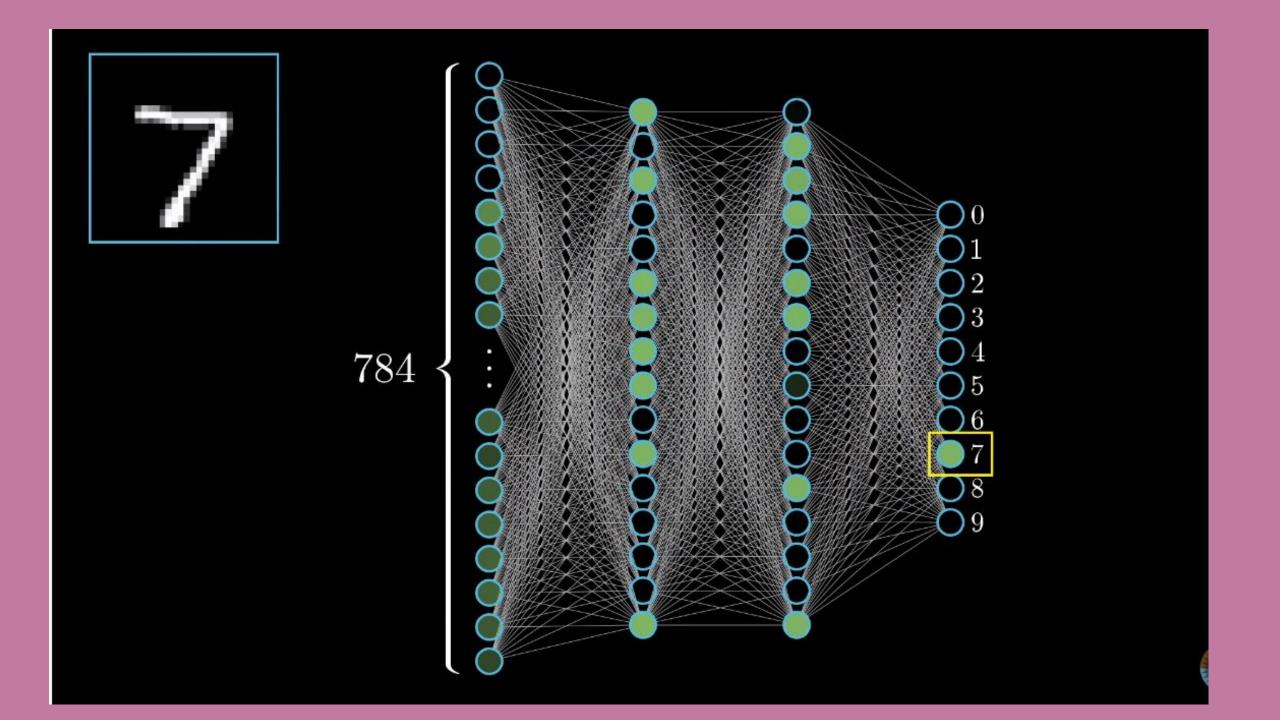


28

 $28 \times 28 = 784$

0.58

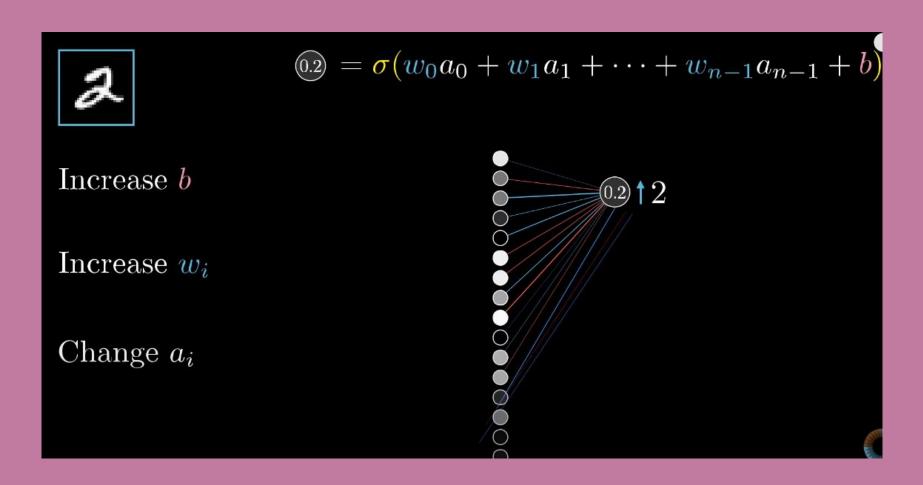
"Activation"



Cost function - The square of the difference between the output answer and the actual answer.

The aim is to reduce this cost function.

Back Propagation- Core algorithm behind how neural networks learn.



Data Program Computation Results

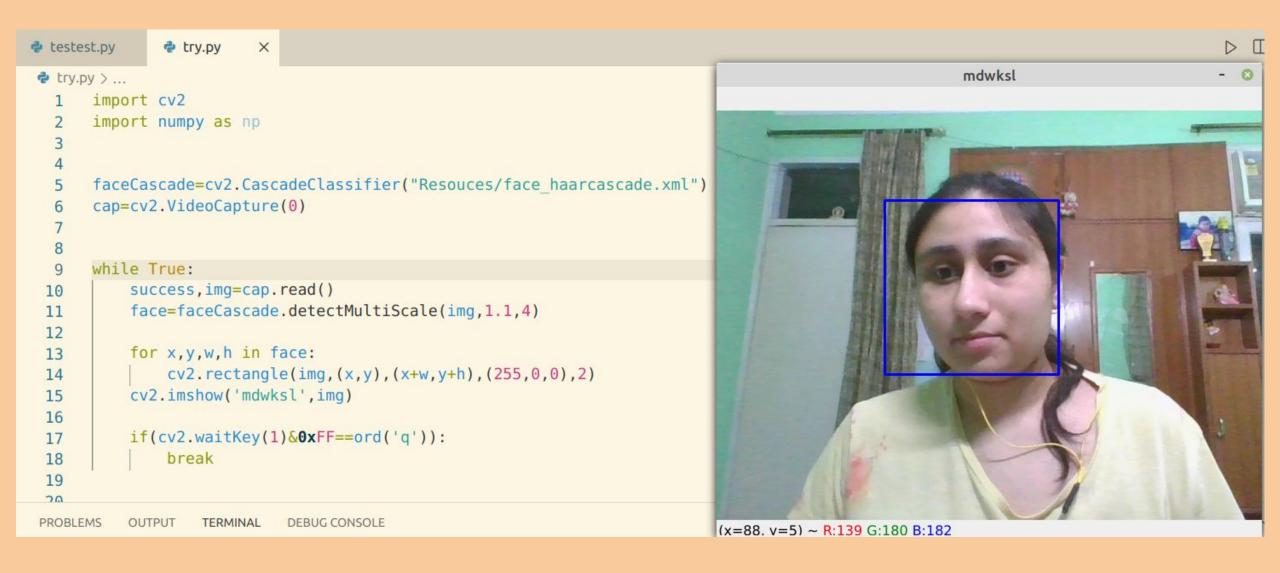




IDEA

- 1. To check whether there is some face present.
- 2. If Yes
- 3. Pass the face cropped and resized to our trained model network.
- 4. If No-Repeat
- 5. Face to be detected using dnn module.

FACE DETECTION



```
testest.py
              try.py
                         X
try.py > ...
      import cv2
      import numpy as np
       faceCascade=cv2.CascadeClassifier("Resouces/face haarcascade.xml")
       cap=cv2.VideoCapture(0)
  8
      while True:
  9
           success,img=cap.read()
 10
 11
           face=faceCascade.detectMultiScale(img,1.1,4)
 12
           for x,y,w,h in face:
 13
               cv2.rectangle(img,(x,y),(x+w,y+h),(255,0,0),2)
 14
           cv2.imshow('mdwksl',img)
 15
 16
          if(cv2.waitKey(1)&0xFF==ord('q')):
 17
               break
 18
 19
 20
PROBLEMS
                            DEBUG CONSOLE
          OUTPUT
                  TERMINAL
```



Training Model

The model was trained using data set from kaggle. 80% of the images were used to train the data and 20% were used for its testing.

There were 2 labels in data set: With Mask and Without Mask.

The images and labels were converted to numpy array and binary respectively.

Then a model was trained. A function was used to augment the images slightly which resulted in slightly different kinds of images while training of data which results in better accuracy.

After all the the model was finally saved and used.

WHY DNN WODULE OF OPEN CV?

- 1. Better accuracy in real time.
- 2. Introduced in Open CV in 2017

What is DNN module in OpenCV?

With the release of OpenCV 3.3 the **deep neural network** (dnn.) library has been substantially overhauled, allowing us to load pre-trained networks via the Caffe, TensorFlow, and Torch/PyTorch frameworks and then use them to classify input images. 21-Aug-2017

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```
import cv2
                                                                                          mdwksl
     from imutils.video.videostream import VideoStream
     import numpy as np
     from tensorflow.keras.models import load model
     # from tensorflow import *
     from tensorflow.python.keras.applications.mobilenet v2 in
     from tensorflow.python.keras.preprocessing.image import
     import imutils
 9
10
11
12
     # faceCascade=cv2.CascadeClassifier("Resouces/face haarca
     prototxtFile=r"/media/shirin/DATA/PROJECT PYTHON MASK DE1
13
14
     weightsFile=r"/media/shirin/DATA/PROJECT PYTHON MASK DET/
     faceNet=cv2.dnn.readNet(prototxtFile, weightsFile)
15
16
     maskModel=load model("/media/shirin/DATA/PROJECT PYTHON
17
18
19
     vs=VideoStream(src=0).start()
20
     # width as 500
21
22
     while True:
23
                                                               (x=637. v=97) ~ R:164 G:178 B:173
         img=vs.read()
24
         # img=imutils.resize(img, width=500)
25
26
         blob=cv2.dnn.blobFromImage(img, 1.0, (224, 224), (104.0, 177.0, 123.0))
         faceNet.setInput(blob)
27
28
29
         det=faceNet.forward()
         h=img.shape[0]
30
31
         w=img.shape[1]
32
33
         faces=[]
         position=[]
34
35
         prediction=[]
```

```
faces=[]
position=[]
prediction=[]
for i in range(0 , det.shape[2]):
    prob=det[0,0,i,2]
    # print("printitng probability")
    # print(prob)
    if(prob>0.5):
        box=det[0,0,i,3:7]
        box[0]*=w
        box[1]*=h
        box[2]*=w
        box[3]*=h
        (startX, startY, endX, endY)=box.astype("int")
        (startX, startY) = (max(0, startX), max(0, startY))
        (endX, endY) = (min(w-1, endX), min(h-1, endY))
        # print(startX, startY, endX, endY)
        face=img[startY:endY,startX:endX]
        face=cv2.cvtColor(face,cv2.COLOR BGR2RGB)
        face=cv2.resize(face,(224,224))
        face=img to array(face)
        face=preprocess input(face)
        faces.append(face)
        position.append((startX, startY, endX, endY))
if(len(faces)>0):
    faces=np.array(faces,dtype="float32")
    prediction=maskModel.predict(faces, batch size=32)
```

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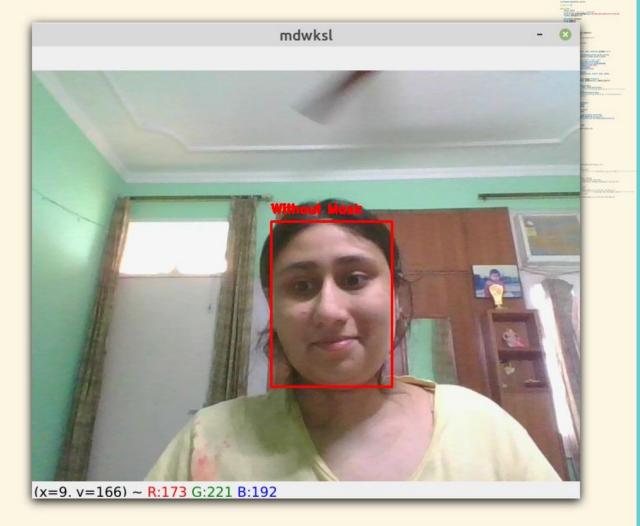
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61 62

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65 66



```
if(len(faces)>0):
   faces=np.array(faces,dtype="float32")
   prediction=maskModel.predict(faces, batch size=32)
for i in range(0,len(position)):
   (startX, startY, endX, endY)=position[i]
   # print("-----")
   # print(position[i])
   (WithMask, WithoutMask)=prediction[i]
   # print("**********printitng predicttion of i**********")
   # print(prediction[i])
   label=""
   colour=()
   if(WithMask>WithoutMask):
       label="With Mask"
       colour=(0,255,0)
   else:
       label="Without Mask"
       colour=(0,0,255)
   cv2.putText(img, label, (startX, startY-10),
       cv2.FONT HERSHEY DUPLEX, 0.5, colour, 2)
   cv2.rectangle(img,(startX,startY),(endX,endY),colour,2)
cv2.imshow('mdwksl',img)
if(cv2.waitKey(1)&0xFF==ord('g')):
   break
```