

In [ ]:

```

import cv2
import numpy as np

cap = cv2.VideoCapture(0)
i = 0
while(cap.isOpened()):
    _,frame = cap.read()
    roi = frame[100:350,100:350]
    roi = cv2.cvtColor(roi,cv2.COLOR_BGR2GRAY)
    #gray = cv2.GaussianBlur(gray,(7,7),0)
    #hist = cv2.equalizeHist(gray)
    #edge = cv2.Canny(gray,25,255)
    roi = cv2.resize(roi,(28,28))
    cv2.imwrite("gesture_data/train/one/"+str(i)+".jpg",roi)
    #contour,_ = cv2.findContours(edge.copy(),cv2.RETR_TREE,cv2.CHAIN_APPROX_SIMPLE)

    #cnt = max(contour, key= lambda x: cv2.contourArea(x))

    #print(len(contour))
    cv2.rectangle(frame,(100,100),(350,350),(255,0,0),3)
    cv2.imshow("frame",frame)
    #cv2.imshow("edge",edge)
    #cv2.drawContours(b_f,contour,-1,(0,255,0),3)
    #cv2.imshow("b_f",b_f)

    cv2.imshow("roi",roi)
    if cv2.waitKey(1) & 0xFF == ord("q"):
        break
    if i>2000:
        break
    i = i+1
cap.release()
cv2.destroyAllWindows()

```

```
import cv2 import numpy as np
```

```
cap = cv2.VideoCapture(0) i = 0 while(cap.isOpened()): _,frame = cap.read() roi = frame[100:350,100:350] roi =
cv2.cvtColor(roi,cv2.COLOR_BGR2GRAY)
```

```

#gray = cv2.GaussianBlur(gray,(7,7),0)
#hist = cv2.equalizeHist(gray)
#edge = cv2.Canny(gray,25,255)
roi = cv2.resize(roi,(28,28))
cv2.imwrite("gesture_data/valid/five/"+str(i)+".jpg",roi)
#contour,_ = cv2.findContours(edge.copy(),cv2.RETR_TREE,cv2.CHAIN_APPROX_SIMPLE)

#cnt = max(contour, key= lambda x: cv2.contourArea(x))

#print(len(contour))
cv2.rectangle(frame,(100,100),(350,350),(255,0,0),3)
cv2.imshow("frame",frame)
#cv2.imshow("edge",edge)
#cv2.drawContours(b_f,contour,-1,(0,255,0),3)
#cv2.imshow("b_f",b_f)

cv2.imshow("roi",roi)
if cv2.waitKey(1) & 0xFF == ord("q"):
    break
if i>200:
    break
i = i+1

cap.release() cv2.destroyAllWindows()

```

In [33]:

```

import keras
import numpy as np
import matplotlib.pyplot as plt

from keras.models import Sequential
from keras.utils import to_categorical
from keras.layers import Dense, Dropout, Flatten, Conv2D, MaxPooling2D
from keras.layers.normalization import BatchNormalization
from keras.layers.advanced_activations import LeakyReLU
import os
from os import listdir
from os.path import isdir
from numpy import asarray
import cv2
from keras.preprocessing.image import img_to_array
from numpy import expand_dims
from keras.preprocessing.image import ImageDataGenerator

from sklearn.model_selection import train_test_split
from keras.utils import to_categorical

```

In [51]:

```
s = 150
m = Sequential()
m.add(Conv2D(32, kernel_size=(3,3), activation="linear", input_shape=(s,s,1), padding="same"))
m.add(BatchNormalization(axis=-1))
m.add(LeakyReLU(alpha=0.1))
m.add(MaxPooling2D((2,2), padding="same"))
m.add(Dropout(0.3))

m.add(Conv2D(64, kernel_size=(3,3), activation="linear", padding="same"))
m.add(BatchNormalization(axis=-1))
m.add(LeakyReLU(alpha=0.1))
m.add(MaxPooling2D((2,2), padding="same"))
m.add(Dropout(0.3))

m.add(Conv2D(128, kernel_size=(3,3), activation="linear", padding="same"))
m.add(BatchNormalization(axis=-1))
m.add(LeakyReLU(alpha=0.1))
m.add(MaxPooling2D((2,2), padding="same"))
m.add(Dropout(0.4))

m.add(Flatten())
m.add(Dense(120, activation="linear"))
m.add(BatchNormalization(axis=-1))
m.add(LeakyReLU(alpha=0.1))
m.add(Dropout(0.3))

m.add(Dense(60, activation="linear"))
m.add(BatchNormalization(axis=-1))
m.add(LeakyReLU(alpha=0.1))
m.add(Dropout(0.2))

m.add(Dense(30, activation="linear"))
m.add(BatchNormalization(axis=-1))
m.add(LeakyReLU(alpha=0.1))
m.add(Dropout(0.1))

m.add(Dense(10, activation="softmax"))
```

In [52]:

```
m.summary()
```

Model: "sequential\_3"

Layer (type)	Output Shape	Param #
conv2d_7 (Conv2D)	(None, 150, 150, 32)	320
batch_normalization_13 (Batch Normalization)	(None, 150, 150, 32)	128
leaky_re_lu_13 (LeakyReLU)	(None, 150, 150, 32)	0
max_pooling2d_7 (MaxPooling2D)	(None, 75, 75, 32)	0
dropout_13 (Dropout)	(None, 75, 75, 32)	0
conv2d_8 (Conv2D)	(None, 75, 75, 64)	18496
batch_normalization_14 (Batch Normalization)	(None, 75, 75, 64)	256
leaky_re_lu_14 (LeakyReLU)	(None, 75, 75, 64)	0
max_pooling2d_8 (MaxPooling2D)	(None, 38, 38, 64)	0
dropout_14 (Dropout)	(None, 38, 38, 64)	0
conv2d_9 (Conv2D)	(None, 38, 38, 128)	73856
batch_normalization_15 (Batch Normalization)	(None, 38, 38, 128)	512
leaky_re_lu_15 (LeakyReLU)	(None, 38, 38, 128)	0
max_pooling2d_9 (MaxPooling2D)	(None, 19, 19, 128)	0
dropout_15 (Dropout)	(None, 19, 19, 128)	0
flatten_3 (Flatten)	(None, 46208)	0
dense_9 (Dense)	(None, 120)	5545080
batch_normalization_16 (Batch Normalization)	(None, 120)	480
leaky_re_lu_16 (LeakyReLU)	(None, 120)	0
dropout_16 (Dropout)	(None, 120)	0
dense_10 (Dense)	(None, 60)	7260
batch_normalization_17 (Batch Normalization)	(None, 60)	240
leaky_re_lu_17 (LeakyReLU)	(None, 60)	0
dropout_17 (Dropout)	(None, 60)	0
dense_11 (Dense)	(None, 30)	1830
batch_normalization_18 (Batch Normalization)	(None, 30)	120
leaky_re_lu_18 (LeakyReLU)	(None, 30)	0

dropout_18 (Dropout)	(None, 30)	0
dense_12 (Dense)	(None, 10)	310
=====		
Total params: 5,648,888		
Trainable params: 5,648,020		
Non-trainable params: 868		

In [53]:

```
m.compile(loss=keras.losses.categorical_crossentropy,optimizer=keras.optimizers.Adam(),metrics=metrics)
```

In [37]:

```
l = dict()
r_l = dict()
c = 0
for j in listdir("leapGestRecog/00/"):
    if not j.startswith("."):
        l[j] = c
        r_l[c] = j
        c = c+1
print(l)
```

```
{'01_palm': 0, '02_1': 1, '03_fist': 2, '04_fist_moved': 3, '05_thumb': 4,
'06_index': 5, '07_ok': 6, '08_palm_moved': 7, '09_c': 8, '10_down': 9}
```

[illegible][illegible]

[illegible]

In [40]:

In [41]:

In [42]:

[5]





In [43]:

```
Y = to_categorical(Y)
X = X.reshape((d_c,s,s,1))
X = X/255
print(X.shape)
print(Y.shape)
```

```
(20000, 150, 150, 1)
```

```
(20000, 10)
```

In [47]:

```
x_train, x_test, y_trin, y_test = train_test_split(X,Y, test_size=0.25, random_state=42)
```

In [50]:

```
print(x_train.shape)
print(x_test.shape)
print(y_trin.shape)
print(y_test.shape)
```

```
(15000, 150, 150, 1)
```

```
(5000, 150, 150, 1)
```

```
(15000, 10)
```

```
(5000, 10)
```

In [55]:

```
his = m.fit(x_train,y_trin,epochs=15, batch_size=128, verbose=1, validation_data=(x_test,y_
```

Train on 15000 samples, validate on 5000 samples

Epoch 1/15

15000/15000 [=====] - 73s 5ms/step - loss: 1.0262 - accuracy: 0.7437 - val\_loss: 3.7320 - val\_accuracy: 0.1188

Epoch 2/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.2112 - accuracy: 0.9815 - val\_loss: 4.6114 - val\_accuracy: 0.1016

Epoch 3/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.0859 - accuracy: 0.9937 - val\_loss: 4.0929 - val\_accuracy: 0.2766

Epoch 4/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.0468 - accuracy: 0.9971 - val\_loss: 4.0223 - val\_accuracy: 0.2406

Epoch 5/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.0297 - accuracy: 0.9983 - val\_loss: 3.0405 - val\_accuracy: 0.4010

Epoch 6/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.0235 - accuracy: 0.9980 - val\_loss: 1.0339 - val\_accuracy: 0.6382

Epoch 7/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.0186 - accuracy: 0.9988 - val\_loss: 0.0282 - val\_accuracy: 0.9952

Epoch 8/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.0126 - accuracy: 0.9993 - val\_loss: 0.0263 - val\_accuracy: 0.9962

Epoch 9/15

15000/15000 [=====] - 54s 4ms/step - loss: 0.0128 - accuracy: 0.9990 - val\_loss: 0.0084 - val\_accuracy: 0.9974

Epoch 10/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.0112 - accuracy: 0.9986 - val\_loss: 0.0890 - val\_accuracy: 0.9670

Epoch 11/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.0098 - accuracy: 0.9987 - val\_loss: 0.1349 - val\_accuracy: 0.9494

Epoch 12/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.0103 - accuracy: 0.9985 - val\_loss: 0.0212 - val\_accuracy: 0.9932

Epoch 13/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.0111 - accuracy: 0.9985 - val\_loss: 0.0020 - val\_accuracy: 0.9996

Epoch 14/15

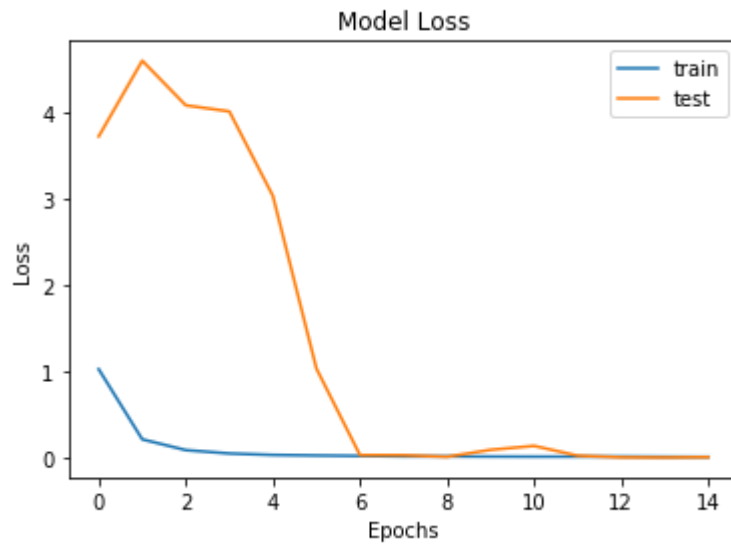
15000/15000 [=====] - 53s 4ms/step - loss: 0.0083 - accuracy: 0.9987 - val\_loss: 3.1798e-04 - val\_accuracy: 1.0000

Epoch 15/15

15000/15000 [=====] - 53s 4ms/step - loss: 0.0066 - accuracy: 0.9990 - val\_loss: 0.0044 - val\_accuracy: 0.9988

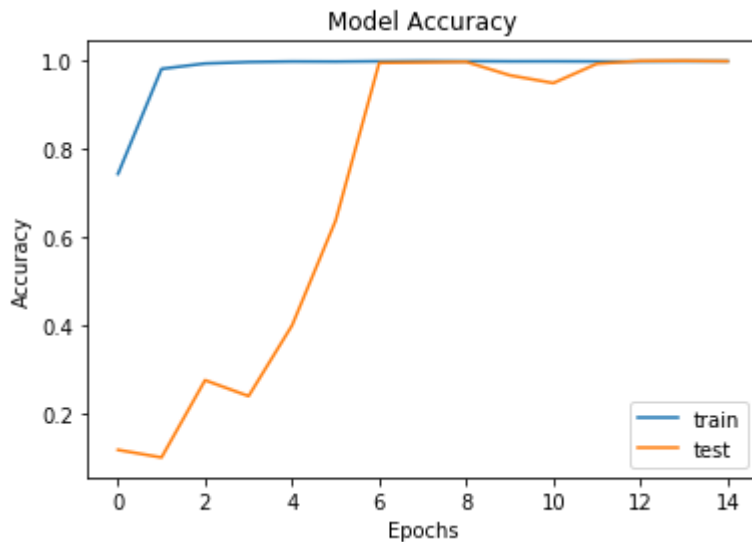
In [56]:

```
plt.plot(his.history['loss'])  
plt.plot(his.history['val_loss'])  
plt.title('Model Loss')  
plt.ylabel('Loss')  
plt.xlabel('Epochs')  
plt.legend(['train', 'test'])  
plt.show()
```



In [59]:

```
plt.plot(his.history['accuracy'])
plt.plot(his.history['val_accuracy'])
plt.title('Model Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epochs')
plt.legend(['train', 'test'])
plt.show()
```



In [62]:

```
score = m.evaluate(x_test,y_test)
print(">>>>>Test loss",score[0])
print(">>>>>Test accuracy",score[1])
```

```
5000/5000 [=====] - 6s 1ms/step
>>>>>Test loss 0.004428002888709307
>>>>>Test accuracy 0.9987999796867371
```

In [63]:

```
m.save("hand_gesture_model.h5")
```

In [102]:

```
prediction = m.predict(x_test)
#y_pred = np.argmax(prediction)
#print(prediction)
#print(y_pred)
print(np.argmax(prediction[400]), y_test[400])
```

```
3 [0. 0. 0. 1. 0. 0. 0. 0. 0.]
```

In [2]:

```
import numpy as np
import cv2
from keras import models as m
import os
from os import listdir
from os.path import isdir
from sklearn.model_selection import train_test_split
from keras.utils import to_categorical
import matplotlib.pyplot as plt
%matplotlib inline
```

In [3]:

```
m = m.load_model("hand_gesture_model.h5")
```

In [4]:

```
s = 150
l = dict()
r_l = dict()
c = 0
for j in listdir("leapGestRecog/00/"):
    if not j.startswith("."):
        l[j] = c
        r_l[c] = j
        c = c+1
print(l)
```

```
{'01_palm': 0, '02_1': 1, '03_fist': 2, '04_fist_moved': 3, '05_thumb': 4,
'06_index': 5, '07_ok': 6, '08_palm_moved': 7, '09_c': 8, '10_down': 9}
```

[illegible][illegible]

[illegible]

In [6]:

In [7]:

In [8]:

In [9]:

In [10]:

localhost:8888/notebooks/Desktop/Robotics/computer vision with python/image segmentation/hand gesture recognition using deep learni... 16/22



In [11]:

```

class_names = ["stop", "palm", "1", "fist", "fist_moved", "one", "index", "ok", "palm_moved"]
plt.figure(figsize=(15,5))
prediction = m.predict(x_test)
for i in range(1,10):
    true_label = y_test[i]
    img = x_test[i]
    img = cv2.cvtColor(x_test[i],cv2.COLOR_GRAY2RGB)
    predicted_label = np.argmax(prediction[i])
    q = index(true_label)

    plt.subplot(3,3,i)
    plt.grid(False)
    plt.xticks([])
    plt.yticks([])
    plt.imshow(img, cmap=plt.cm.binary)

    if predicted_label == q:
        color = "blue"
    else:
        color = "red"

    plt.xlabel("Predicted: {} {:.2f}% (True: {})".format(class_names[predicted_label],
                                                         100*np.max(prediction),
                                                         class_names[q]),color=color)

    print(">>>>>>q",q)
    print(">>>>>>p",predicted_label)
    print(true_label)

```

```

>>>>>>q 0
>>>>>>p 0
[1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
>>>>>>q 3
>>>>>>p 3
[0. 0. 0. 1. 0. 0. 0. 0. 0. 0.]
>>>>>>q 5
>>>>>>p 5
[0. 0. 0. 0. 0. 1. 0. 0. 0. 0.]
>>>>>>q 9
>>>>>>p 9
[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]
>>>>>>q 9
>>>>>>p 9
[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]
>>>>>>q 5
>>>>>>p 5
[0. 0. 0. 0. 0. 1. 0. 0. 0. 0.]
>>>>>>q 5
>>>>>>p 5
[0. 0. 0. 0. 0. 1. 0. 0. 0. 0.]
>>>>>>q 8
>>>>>>p 8
[0. 0. 0. 0. 0. 0. 0. 0. 1. 0.]
>>>>>>q 0
>>>>>>p 0
[1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]

```



Predicted: stop 100% (True: stop)



Predicted: c 100% (True: c)



Predicted: one 100% (True: one)



Predicted: fist 100% (True: fist)



Predicted: c 100% (True: c)



Predicted: palm\_moved 100% (True: palm\_moved)



Predicted: one 100% (True: one)



Predicted: one 100% (True: one)



Predicted: stop 100% (True: stop)

In [12]:

```

class_names = ["down", "palm", "1", "fist", "fist_moved", "thumb", "index", "ok", "palm_mov
plt.figure(figsize=(15,5))
prediction = m.predict(x_test)
for i in range(1,20):
    true_label = y_test[i]
    img = x_test[i]
    img = cv2.cvtColor(x_test[i],cv2.COLOR_GRAY2RGB)
    predicted_label = np.argmax(prediction[i])
    q = index(true_label)

    plt.subplot(4,5,i)
    plt.grid(False)
    plt.xticks([])
    plt.yticks([])
    plt.imshow(img, cmap=plt.cm.binary)

    if predicted_label == q:
        color = "blue"
    else:
        color = "red"

    plt.xlabel("Predicted: {} {:.2f}% (True: {})".format(class_names[predicted_label],
                                                         100*np.max(prediction),
                                                         class_names[q]),color=color)

    print(">>>>>>q",q)
    print(">>>>>>p",predicted_label)
    print(true_label)

```

```

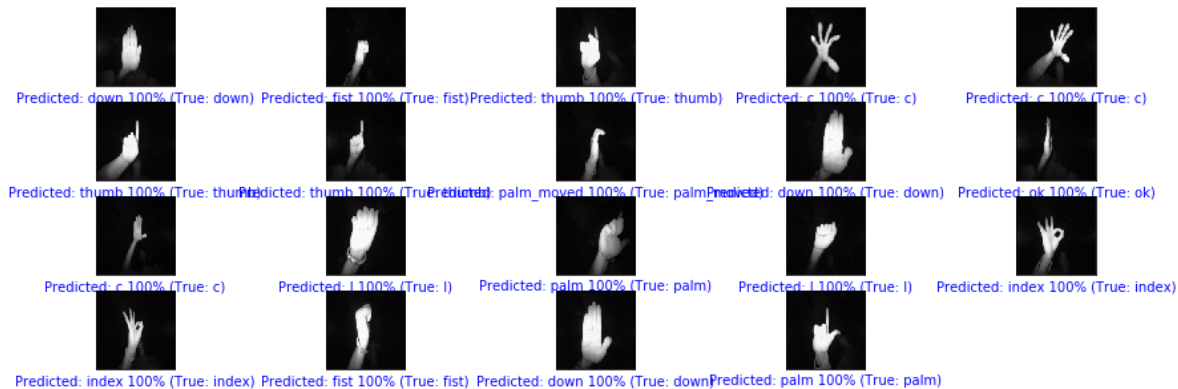
>>>>>>q 0
>>>>>>p 0
[1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
>>>>>>q 3
>>>>>>p 3
[0. 0. 0. 1. 0. 0. 0. 0. 0. 0.]
>>>>>>q 5
>>>>>>p 5
[0. 0. 0. 0. 0. 1. 0. 0. 0. 0.]
>>>>>>q 9
>>>>>>p 9
[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]
>>>>>>q 9
>>>>>>p 9
[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]
>>>>>>q 5
>>>>>>p 5
[0. 0. 0. 0. 0. 1. 0. 0. 0. 0.]
>>>>>>q 5
>>>>>>p 5
[0. 0. 0. 0. 0. 1. 0. 0. 0. 0.]
>>>>>>q 8
>>>>>>p 8
[0. 0. 0. 0. 0. 0. 0. 0. 1. 0.]
>>>>>>q 0
>>>>>>p 0
[1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
>>>>>>q 7
>>>>>>p 7
[0. 0. 0. 0. 0. 0. 1. 0. 0. 0.]

```

```

>>>>>>>q 9
>>>>>>>p 9
[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]
>>>>>>>q 2
>>>>>>>p 2
[0. 0. 1. 0. 0. 0. 0. 0. 0. 0.]
>>>>>>>q 1
>>>>>>>p 1
[0. 1. 0. 0. 0. 0. 0. 0. 0. 0.]
>>>>>>>q 2
>>>>>>>p 2
[0. 0. 1. 0. 0. 0. 0. 0. 0. 0.]
>>>>>>>q 6
>>>>>>>p 6
[0. 0. 0. 0. 0. 0. 1. 0. 0. 0.]
>>>>>>>q 6
>>>>>>>p 6
[0. 0. 0. 0. 0. 0. 1. 0. 0. 0.]
>>>>>>>q 3
>>>>>>>p 3
[0. 0. 0. 1. 0. 0. 0. 0. 0. 0.]
>>>>>>>q 0
>>>>>>>p 0
[1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
>>>>>>>q 1
>>>>>>>p 1
[0. 1. 0. 0. 0. 0. 0. 0. 0. 0.]

```



In [13]:

```

from sklearn.metrics import confusion_matrix, classification_report
y_pred = m.predict(x_test)
print(classification_report(np.argmax(y_test,axis=1),np.argmax(y_pred,axis=1)))
print(confusion_matrix(np.argmax(y_test,axis=1),np.argmax(y_pred,axis=1)))
#print(np.argmax(y_pred,axis=1))
#print(np.argmax(y_test,axis=1))

```

	precision	recall	f1-score	support
0	1.00	0.99	1.00	518
1	1.00	1.00	1.00	511
2	1.00	1.00	1.00	503
3	1.00	1.00	1.00	508
4	1.00	1.00	1.00	466
5	0.99	1.00	0.99	490
6	1.00	1.00	1.00	499
7	1.00	1.00	1.00	502
8	1.00	1.00	1.00	518
9	1.00	1.00	1.00	485
accuracy			1.00	5000
macro avg	1.00	1.00	1.00	5000
weighted avg	1.00	1.00	1.00	5000

```

[[ 513  0  0  1  0  4  0  0  0  0]
 [  0 510  0  0  0  1  0  0  0  0]
 [  0  0 503  0  0  0  0  0  0  0]
 [  0  0  0 508  0  0  0  0  0  0]
 [  0  0  0  0 466  0  0  0  0  0]
 [  0  0  0  0  0 490  0  0  0  0]
 [  0  0  0  0  0  0 499  0  0  0]
 [  0  0  0  0  0  0  0 502  0  0]
 [  0  0  0  0  0  0  0  0 518  0]
 [  0  0  0  0  0  0  0  0  0 485]]

```

## Python Programming illustrating numpy.full method

```
import numpy as np
```

```
a = np.full([2, 2], 67, dtype = int) print("\nMatrix a : \n", a)
```

```
c = np.full([3, 3], 10.1) print("\nMatrix c : \n", c)
```

```
o/p Matrix a : [[67 67] [67 67]]
```

```
Matrix c : [[10.1 10.1 10.1] [10.1 10.1 10.1] [10.1 10.1 10.1]]
```

```
ori_path = "gesture_data/" r = 0 im_data = [] label = [] for i in listdir(ori_path):
```

