

Importing required libraries

```
In [2]: import numpy as np # for matrix calculations
import os # for file hierarchy handling
import matplotlib.pyplot as plt # plotting the images and data
import matplotlib.image as mpimg # reading the images
from sklearn.model_selection import train_test_split # for preprocessing
from PIL import Image # image processing
import glob # file handling and folder scanner
import cv2 # for converting images to matrix
```

Getting folder paths of each images in list

```
In [3]: listofSubNumberDict = os.listdir("/kaggle/input/practice2/imgs")
path = '/kaggle/input/practice2/imgs/'
listofSubNumberDict.sort()
print(listofSubNumberDict)
```

```
['0', '1', '2', '3', '4', '5', '6', '7', '8', '9']
```

getting data from images and display images

```
In [9]: # getting views of images in dataset
filepath = "/kaggle/input/practice2/imgs/0/1.jpg"
img = Image.open(filepath)

# get width and height
width = img.width
height = img.height

# display width and height
print("The height of the image is: ", height)
print("The width of the image is: ", width)

Image.open("/kaggle/input/practice2/imgs/0/1.jpg")
```

```
The height of the image is: 70
The width of the image is: 50
```

```
Out[9]:
```



finding number of images in each folder

```
In [11]: for i in listofSubNumberDict:
print(f'there is {len(os.listdir(f"/kaggle/input/practice2/imgs/{i}"))} of images of {i}')
```

```
there is 130 of images of 0
there is 130 of images of 1
there is 130 of images of 2
there is 130 of images of 3
there is 130 of images of 4
there is 130 of images of 5
there is 130 of images of 6
there is 130 of images of 7
there is 130 of images of 8
there is 130 of images of 9
```

converting all images of 0-9 to matrix in form of numpy array

```
In [12]: img_extension = ['.jpg']
files = []

for i in listofSubNumberDict:
    allImgPath = f'/kaggle/input/practice2/imgs/{i}/'
    files.extend(glob.glob(allImgPath + '.*' + e)) for e in img_extension]

allImg = np.asarray([cv2.imread(file) for file in files])
```

getting information of matrix converted from images

```
In [13]: print(allImg.shape)
print(allImg[0].shape)
print(allImg.dtype)
print(type(allImg))
print(len(allImg))
```

```
(1300, 70, 50, 3)
(70, 50, 3)
uint8
<class 'numpy.ndarray'>
1300
```

labelling the list of images for training

```
In [14]: counter = 0
dig = 0
labels = []
for i in range(0,130):
    labels.append(0)

while(dig<=9):
    if(counter%130==0):
        dig +=1
        labels.append(dig)
        counter+=1
labels.remove(10)
labels = np.array(labels)
```

verifying the labels

```
In [15]: print(labels)
c=0
for i in range(0,10):
    for j in labels:
        if i==j:
            c += 1
    print(f"number of {i} is {c}")
    c = 0
```

```
[0 0 0 ... 9 9 9]
number of 0 is 130
number of 1 is 130
number of 2 is 130
number of 3 is 130
number of 4 is 130
number of 5 is 130
number of 6 is 130
number of 7 is 130
number of 8 is 130
number of 9 is 130
```

scaling the matrix for getting values of each elements 0-1 that will optimise the training

```
In [16]: for i in range(0, len(allImg)):
allImg[i] = allImg[i]/255
```

Train - Test Split for finding accuracy of unidentified images

```
In [17]: X = allImg
Y = np.asarray(labels)
x_train , x_test , y_train , y_test = train_test_split(X,Y,test_size=0.2,random_state=1)
```

getting number of images splitted by train test split

```
In [18]: print (f'The X images are splitted into train set : {len(x_train)} test set : {len(x_test)}')
print (f'The Y labels are splitted into train set : {len(y_train)} test set : {len(y_test)}')
print(x_train[0])
```

```
The X images are splitted into train set : 1040 test set : 260
The Y labels are splitted into train set : 1040 test set : 260
```

```
[[[1 1 1]
[1 1 1]
[1 1 1]
...
[1 1 1]
[1 1 1]
[1 1 1]]
```

```
[[[1 1 1]
[1 1 1]
[1 1 1]
...
[1 1 1]
[1 1 1]
[1 1 1]]
```

```
[[[1 1 1]
[1 1 1]
[1 1 1]
...
[1 1 1]
[1 1 1]
[1 1 1]]
```

```
...
```

```
[[[1 1 1]
[1 1 1]
[1 1 1]
...
[1 1 1]
[1 1 1]
[1 1 1]]
```

```
[[[1 1 1]
[1 1 1]
[1 1 1]
...
[1 1 1]
[1 1 1]
[1 1 1]]
```

```
[[[1 1 1]
[1 1 1]
[1 1 1]
...
[1 1 1]
[1 1 1]
[1 1 1]]]
```

importing tensorflow for building the model

```
In [19]: import tensorflow as tf
```

implementing neural network

```
In [20]: model = tf.keras.models.Sequential()
model.add(tf.keras.layers.Flatten())
model.add(tf.keras.layers.Dense(128, activation=tf.nn.relu))
model.add(tf.keras.layers.Dense(128, activation=tf.nn.relu))
model.add(tf.keras.layers.Dense(10, activation=tf.nn.softmax))
model.compile(optimizer='adam',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])
```

feed the model with our training matrix and testing labels

```
In [28]: model.fit(x_train, y_train, epochs=10)
```

```
Epoch 1/10
33/33 [=====] - 0s 14ms/step - loss: 0.1337 - accuracy: 0.9644
Epoch 2/10
33/33 [=====] - 0s 13ms/step - loss: 0.1336 - accuracy: 0.9712
Epoch 3/10
33/33 [=====] - 0s 13ms/step - loss: 0.0909 - accuracy: 0.9846
Epoch 4/10
33/33 [=====] - 0s 13ms/step - loss: 0.0858 - accuracy: 0.9856
Epoch 5/10
33/33 [=====] - 0s 13ms/step - loss: 0.0597 - accuracy: 0.9962
Epoch 6/10
33/33 [=====] - 0s 13ms/step - loss: 0.0424 - accuracy: 0.9971
Epoch 7/10
33/33 [=====] - 0s 13ms/step - loss: 0.0364 - accuracy: 1.0000
Epoch 8/10
33/33 [=====] - 0s 13ms/step - loss: 0.0285 - accuracy: 1.0000
Epoch 9/10
33/33 [=====] - 0s 13ms/step - loss: 0.0273 - accuracy: 1.0000
Epoch 10/10
33/33 [=====] - 0s 13ms/step - loss: 0.0207 - accuracy: 1.0000
```

```
Out[28]: <keras.callbacks.History at 0x7fd40ee6d1d0>
```

finding loss and accuracy of unidentified images

```
In [29]: val_loss, val_acc = model.evaluate(x_test, y_test)
print(val_loss)
```

```
print(val_acc*100)
```

```
9/9 [=====] - 0s 4ms/step - loss: 0.9341 - accuracy: 0.7538  
0.9341462254524231  
75.38461685180664
```