```
In [1]: import os
        from pathlib import Path
        from keras.preprocessing import image
        import matplotlib.pyplot as plt
In [2]: p = Path(r"///Users/tanmaypriyadarshi/Desktop/pokemon")
In [3]: dirs = p.glob("*")
In [4]: image data = []
        labels = []
        image path = []
        label dict = {"Pikachu":0, "Bulbasaur":1, "Meowth":2}
        for folder dirs in dirs:
            label = str(folder dirs).split('/')[-1]
            print(label)
            cnt = 0
            for img path in folder dirs.glob("*.jpg"):
                 img = image.load img(img path, target size=(40,40))
                 img array = image.img to array(img)
                 image data.append(img array)
                 labels.append(label_dict[label])
                cnt+=1
            print(cnt)
        Pikachu
        199
        Bulbasaur
        174
```

Visualisation

Meowth 70

```
In [5]: import numpy as np
```

```
In [6]: print(len(image data))
print(len(labels))
print(labels)
443
443
0, 0, 0, 0, 0,
2, 2, 21
import random
In [7]:
random.seed(10)
```

from sklearn.utils import shuffle

In [8]:

```
In [9]: x = np.array(image data)
     y = np.array(labels)
     x,y = shuffle(x,y,random state=2)
     print(x.shape)
     print(y.shape)
     print(type(x))
     x = x/255.0
     #print(x)
     print(y)
     (443, 40, 40, 3)
     (443,)
     <class 'numpy.ndarray'>
     [1\ 2\ 0\ 0\ 2\ 0\ 1\ 0\ 1\ 0\ 1\ 1\ 1\ 1\ 0\ 1\ 2\ 0\ 1\ 2\ 2\ 1\ 2\ 0\ 1\ 0\ 0\ 2\ 0\ 0\ 1\ 0\ 0
      1 1 0 1
      2 0 2 1
      0 1 1 0
      0 0 1 0
      1 1 0 0
      \begin{smallmatrix} 0 & 0 & 1 & 2 & 2 & 0 & 1 & 0 & 2 & 1 & 2 & 2 & 1 & 0 & 0 & 1 & 2 & 2 & 2 & 1 & 2 & 1 & 0 & 1 & 2 & 0 & 2 & 0 & 1 & 0 & 1 & 0 & 0 \\ \end{smallmatrix}
     0 2 1 0
      \begin{smallmatrix} 0 & 0 & 1 & 2 & 2 & 2 & 0 & 0 & 1 & 0 & 1 & 2 & 1 & 2 & 0 & 0 & 1 & 0 & 1 & 2 & 1 & 1 & 2 & 1 & 0 & 0 & 0 & 0 & 2 & 1 & 0 & 0 & 0 \\ \end{smallmatrix}
     1 1 0 1
      1 1 1 0
      0 0 01
```

Let's draw some image

```
In [10]: def drawImg(img,label):
    name = [key for key in label_dict.keys() if label_dict[key] ==
label]
    print(name)
    plt.title(name[0])
    plt.style.use("seaborn")
    plt.axis("off")
    plt.imshow(img)
    plt.show()
```

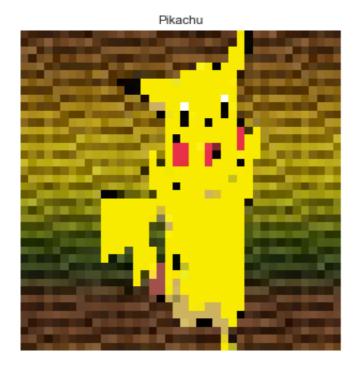
['Meowth']



['Pikachu']



['Pikachu']



['Meowth']



['Pikachu']



['Bulbasaur']



['Pikachu']

Pikachu

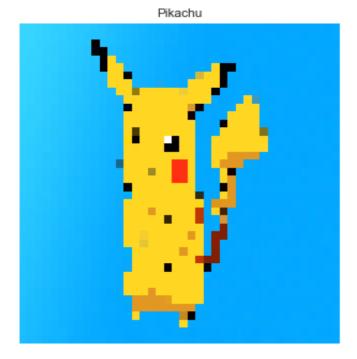


['Bulbasaur']

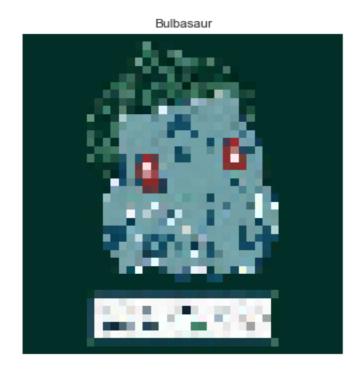
Bulbasaur



['Pikachu']



['Bulbasaur']



['Bulbasaur']

Bulbasaur



['Bulbasaur']



['Bulbasaur']

Bulbasaur



['Pikachu']

Pikachu

['Bulbasaur']

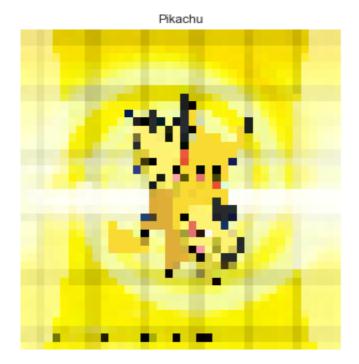
Bulbasaur



['Meowth']

Meowth

['Pikachu']

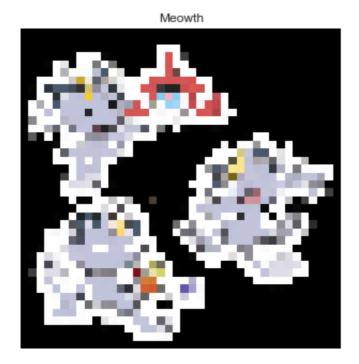


['Bulbasaur']

Bulbasaur



['Meowth']



training and testing dataset.....

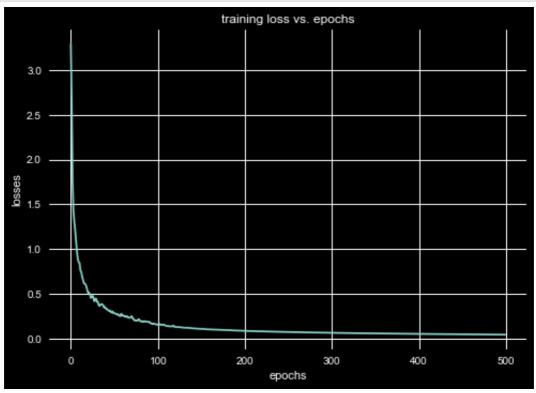
```
In [18]: print(x.shape)
         print(y.shape)
         print(xtest.shape)
         print(ytest.shape)
         (354, 40, 40, 3)
          (354,)
          (89, 40, 40, 3)
          (89,)
In [19]: | \mathbf{def} \operatorname{softmax}(z) :
              ex_pa = np.exp(z)
              ans = ex pa/np.sum(ex pa,axis=1,keepdims=True)
              return ans
In [20]: class NeuralNetwork:
              def init (self,input size,layer,output size):
                  np.random.seed(0)
                  model = \{\}
                  model['w1'] = np.random.randn(input size, layer[0])
                  model['b1'] = np.zeros((1,layer[0]))
                  model['w2'] = np.random.randn(layer[0], layer[1])
                  model['b2']= np.zeros((1,layer[1]))
                  model['w3'] = np.random.randn(layer[1],output size)
                  model['b3']= np.zeros((1,output_size))
                  self.model = model
              def forward(self,x):
                  w1,w2,w3 = self.model['w1'],self.model['w2'],self.model['w3
          1
                  b1,b2,b3 = self.model['b1'],self.model['b2'],self.model['b3
          ']
                  z1 = np.dot(x,w1) + b1
                  a1 = np.tanh(z1)
                  z2 = np.dot(a1,w2) + b2
                  a2 = np.tanh(z2)
                  z3 = np.dot(a2,w3) + b3
                  y = softmax(z3)
                  self.activation outputs = (a1,a2,y )
                  return y_
```

```
def backward(self,x,y,learning rate=0.001):
       w1,w2,w3 = self.model['w1'],self.model['w2'],self.model['w3
']
       b1,b2,b3 = self.model['b1'],self.model['b2'],self.model['b3
']
       m = x.shape[0]
       a1,a2,y = self.activation outputs
       delta3 = y_ - y
       dw3 = np.dot(a2.T, delta3)
       db3 = np.sum(delta3,axis=0)/float(m)
       delta2 = (1-np.square(a2))*np.dot(delta3,w3.T)
       dw2 = np.dot(a1.T,delta2)
       db2 = np.sum(delta2,axis=0)/float(m)
       delta1 = (1-np.square(a1))*np.dot(delta2,w2.T)
       dw1 = np.dot(x.T, delta1)
       db1 = np.sum(delta1,axis=0)/float(m)
       # update the model parameter using gradient descent...
       self.model['w1'] -= learning_rate*dw1
       self.model['b1'] -= learning rate*db1
       self.model['w2'] -= learning rate*dw2
       self.model['b2'] -= learning rate*db2
       self.model['w3'] -= learning_rate*dw3
       self.model['b3'] -= learning_rate*db3
   def predict(self,x):
       y_out = self.forward(x)
       return np.argmax(y out,axis=1)
   def summary(self):
       w1,w2,w3 = self.model['w1'],self.model['w2'],self.model['w3
1
       a1,a2,y_ = self.activation_outputs
       print("w1",w1.shape)
       print("a1",a1.shape)
       print("w2",w2.shape)
       print("a2",a2.shape)
```

print("w3",w3.shape)

```
print("y_",y_.shape)
In [21]: def loss(y hot,p):
              l = -np.mean(y hot*np.log(p))
             return 1
         def one hot(y,depth):
             m = y.shape[0]
             y oht = np.zeros((m,depth))
             print(y oht.shape)
             y_oht[np.arange(m),y]=1
             return y_oht
In [29]: def train(x,y,model,epochs,learning rate,logs=True):
             training losses = []
             classes = len(np.unique(y))
             Y 	ext{ OHT = one hot(y,classes)}
              for ix in range(epochs):
                  y_{-} = model.forward(x)
                  l = loss(Y OHT, y)
                  model.backward(x,Y_OHT,learning_rate)
                  training losses.append(1)
                  if logs and ix%50==0:
                      print("Epoch %d loss %.4f"%(ix,1))
             return training losses
In [30]: model = NeuralNetwork(input_size=4800,layer=[100,50],output_size=3)
In [31]: print(x.shape)
         (354, 4800)
In [32]: x = x.reshape(x.shape[0],-1)
         print(x.shape)
         xtest = xtest.reshape(xtest.shape[0],-1)
         print(xtest.shape)
         (354, 4800)
         (89, 4800)
```

```
In [34]: plt.style.use("dark_background")
    plt.title("training loss vs. epochs")
    plt.xlabel("epochs")
    plt.ylabel("losses")
    plt.plot(losses)
    plt.show()
```



finding the accuracy....

```
In [35]: def getAccuracy(x,y,model):
    outputs = model.predict(x)
    acc = np.sum(outputs==y)/y.shape[0]
    return acc
```

```
In [36]: print("Train Acc is %.4f"%getAccuracy(x,y,model))
    print("Test Acc is %.4f"%getAccuracy(xtest,ytest,model))

Train Acc is 0.9633
    Test Acc is 0.6629
In []:
```