## MNIST (Modified National Institute of Standards and Technology database)

In this notebook, we will create a neural network to recognize handwritten digits from the famous MNIST dataset.

We will experiment with two different networks for this task. The first one will be a multi-layer perceptron (MLP), which is a standard type of feedforward neural network with fully-connected layers of weights, and the second will be a convolutional neural network (CNN), which takes advantage of the inherently two-dimensional spatial geometry of the input images.

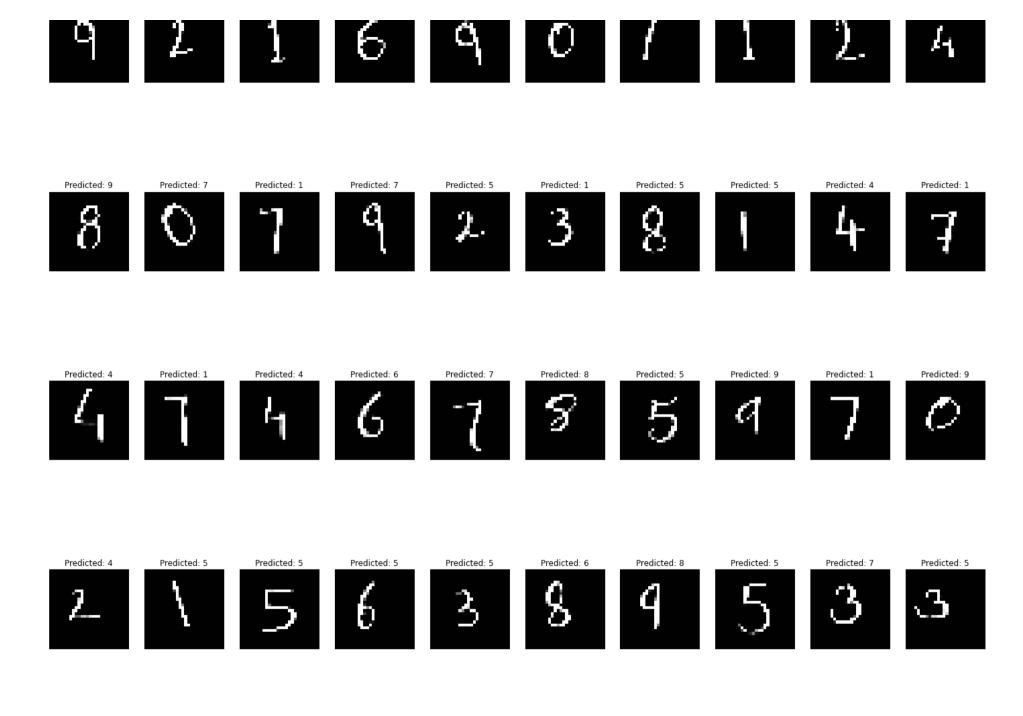
the MNIST dataset and printing a short description of its contents.

```
In [8]: import os
        import pickle
        import cv2
        import numpy as np
        import matplotlib.pyplot as plt
        import glob
        from PIL import Image
        # Mnist Class
        class Mnist():
            imq size = 784
            img_dim = (1, 28, 28)
            weights_file_name = 'sample_weight.pkl'
            def init (self):
                self.network = None
            def init network(self):
                with open(self.weights_file_name, 'rb') as f:
```

```
self.network = pickle.load(f)
def sigmoid(self, a):
    return 1 / (1 + np.exp(-a))
def softmax(self, a):
    c = np.max(a)
   exp_a = np.exp(a - c)
   sum_exp_a = np.sum(exp_a)
   y = exp_a / sum_exp_a
   return y
def predict(self, x):
   w1, w2, w3 = self.network['W1'], self.network['W2'], self.network['W3']
    b1, b2, b3 = self.network['b1'], self.network['b2'], self.network['b3']
   a1 = np.dot(x, w1) + b1
   z1 = self.sigmoid(a1)
   a2 = np.dot(z1, w2) + b2
   z2 = self.sigmoid(a2)
   a3 = np.dot(z2, w3) + b3
   \# z3 = sigmoid(a3)
   y = self.softmax(a3)
   return y
```

```
In [9]: #Creating Mnist class object
         mnist=Mnist()
         path = glob.glob("*.jpg")
         fnl image=[]
         predicted nums=[]
         image2display=[]
         for img in path:
             n=cv2.imread(img)
             # Replace this with the path to your image
             n=cv2.resize(n, (28, 28))
             predicted nums.append(img[:1])
             image2display.append(n)
             img array = np.asarray(n)
             image = cv2.cvtColor(img_array , cv2.COLOR_BGR2GRAY) #(28, 28)
             image = image / 255
         # get the center of the camera image in a square shape
             image = image.reshape(1, 784)
             fnl image.append(image)
In [10]: mnist.init network()
         plt.figure(figsize=(25,25)) # specifying the overall grid size
         v pred=[]
         for i in range(50):
             plt.subplot(5,10,i+1)
         # Create the array of the right shape to feed into the keras model
         # The 'length' or number of images you can put into the array is
         # determined by the first position in the shape tuple, in this case 1.
             plt.imshow(image2display[i])
             plt.axis('off')
             y=mnist.predict(fnl image[i])
             predicted num=np.argmax(y)
             y_pred.append(predicted_num)
             certainity=np.max(y)
             plt.title('Predicted: {}'.format(predicted_num))
         plt.show()
```

Predicted: 9 Predicted: 4 Predicted: 5 Predicted: 4 Predicted: 9 Predicted: 9 Predicted: 5 Predicted: 5 Predicted: 5 Predicted: 5 Predicted: 4



Predicted: 9 Predicted: 5 Predicted: 6 Predicted: 4 Predicted: 9 Predicted: 4 Predicted: 4 Predicted: 4 Predicted: 4 Predicted: 8 Predicted: 5 Predicted: 4 Predicted: 4 Predicted: 4 Predicted: 4 Predicted: 5 Predicted: 4 Predicted: 5 Predicted: 4 Predicted: 4 Predicted: 5 Predicted: 5 Predicted: 4 Predicted: 5 Predicted: 5 Predicted: 4 Predicted: 5 Predi

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