

Intro to deep learning

Assignment 2

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Google Colab Link
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https://colab.research.google.com/drive/1eruRxcR9nSIYN3CAAC11AeLSVC_fKuZB?usp=sharing

Problem 1

a) calculate and show the confusion matrix

		Predicted		TP = 19	TN = 9
		1	0		
Actual	1	19	6	FP = 6	FN = 6
	0	6	9		

b) Calculate and provide the accuracy of the model

$$\text{Accuracy} = \frac{TP + TN}{TP + TN + FP + FN} = \frac{19 + 9}{19 + 9 + 6 + 6} = \frac{28}{40} = \boxed{70.00\%}$$

c) Calculate the sensitivity of the model

$$\text{Sensitivity} = \frac{TP}{TP + FN} = \frac{19}{19 + 6} = \boxed{76.00\%}$$

d) Calculate the specificity of the model

$$\text{Specificity} = \frac{TN}{TN + FP} = \frac{9}{9 + 6} = \frac{9}{15} = \boxed{60.00\%}$$

e) Calculate the F1 measure of the model

$$F1 = \frac{2 * (\text{Sensitivity} * \text{Precision})}{\text{Sensitivity} + \text{Precision}} = \frac{2(0.76 * 0.76)}{0.76 + 0.76} = \boxed{76.00\%}$$

f) Calculate the Precision

$$\text{Precision} = \frac{TP}{TP + FP} = \frac{19}{25} = \boxed{76.00\%}$$

Problem 2

a) calculate and show the confusion matrix

predicted
1 0

Actual	1	0
1	7	2
0	3	5

$$TP = 7$$

$$TN = 5$$

$$FP = 3$$

$$FN = 2$$

b) Calculate and report Accuracy, sensitivity, specificity

$$\text{Accuracy} = \frac{7+5}{17} = \boxed{70.59\%}, \text{Sensitivity} = \frac{7}{7+2} = \boxed{77.78\%}$$

$$\text{Specificity} = \frac{5}{5+3} = \boxed{62.50\%}$$

c) calculate and report the perceptron classifier
Values of w_0 , w_1 , and w_2

$$V = X_0 w_0 + X_1 w_1 + X_2 w_2, V = 0$$

Samples: $(1.5, 0)$, $(0, 3)$

$$(1.5, 0) \quad w_0 + (1.5)w_1 + (0)w_2 = 0$$

$$w_0 + 1.5w_1 = 0$$

$$w_0 = 1 \Rightarrow 1 + 1.5w_1 = 0 \Rightarrow -0.67$$

$$(0, 3) \quad w_0 + (0)w_1 + 3w_2 = 0$$

$$w_0 + 3w_2 = 0$$

$$w_0 = 1 \Rightarrow 1 + 3w_2 = 0 \Rightarrow -0.33$$

$$\boxed{w_0 = 1, w_1 = -0.67, w_2 = -0.33}$$

Problem 2 Continued

d) for the new data samples $(2, -1)$, $(-0.5, 0.5)$, calculate and report the local field (v), perception output (y), and classification label.

$$V = w_0 x_0 + w_1 x_1 + w_2 x_2$$

$$= x_0 - 0.67x_1 - 0.33x_2$$

$$= x_0 - 0.67(2) - 0.33(-0.5)$$

$$= 1 - 0.67(2) - 0.33(-0.5)$$

$$V = -0.175$$

$$y = \begin{cases} 1 & \text{if } v \geq 0 \\ 0 & \text{if } v < 0 \end{cases}$$

$$y = \text{class } 0$$

Problem 3: MNIST dataset - The MNIST dataset is divided into two sets - training and test. Each set comprises a series of images (28 x 28-pixel images of handwritten digits) and their respective labels (values from 0 - 9, representing which digit the image corresponds to).

- a) Use mnist function in keras.datasets to split the MNIST dataset into the training and testing sets. Print the following: The number of images in each training and testing set, and the image width and height.
- b) Write a function (with images of ten digits and labels as the input) that plots a figure with 10 subplots for each 0-9 digits. Each subplot has the number of the handwritten digit in the title.
- c) Create a loop to call the plot function in (b) with images from each set to create three figures. Note: the code has to select the images randomly. Include all the 10 digits in each figure. Show the results of your code.
- d) In machine learning, we usually divide the training set into two sets of training and validation sets to adjust a machine learning model parameters. In your code, randomly select 20% of the training images and their corresponding labels and name them as x_valid and y_valid, respectively. Name the remaining training images and their labels as x_train and y_train, respectively. Print the number of images in each training and validation set. Note: that there are no overlaps between the two sets.

```
[ ] from keras.datasets import mnist
import matplotlib.pyplot as plt
import numpy as np

# PART A

# Split the MNIST dataset into training and testing sets
(x_train, y_train), (x_test, y_test) = mnist.load_data()

# Print out the training set
print("There are", x_train.shape[0], "Images in the MNIST training set where each has a width of:", x_train.shape[1], "and a height of:", x_train.shape[2])

# Print out the testing set
print("There are", x_test.shape[0], "Images in the MNIST testing set where each has a width of:", x_test.shape[1], "and a height of:", x_test.shape[2])

There are 60000 Images in the MNIST training set where each has a width of: 28 and a height of: 28
There are 10000 Images in the MNIST testing set where each has a width of: 28 and a height of: 28
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[ ] # PART B

def plot_figure(images, labels, digit_order):
    digits = digit_order

    #x_train_i = x_train_from_zero_to_nine[10,:,:]

    fig, axs = plt.subplots(nrows=2, ncols=5)

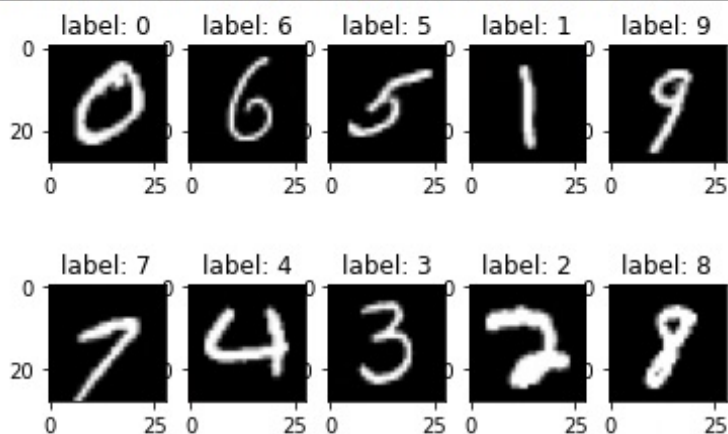
    axs = axs.ravel()

    for i in range(0, 10):
        x_train_from_zero_to_nine = x_train[y_train==int(digits[i]),:,:]
        axs[i].imshow(x_train_from_zero_to_nine[i], cmap='gray')
        axs[i].set_title('label: ' + str(digits[i]))

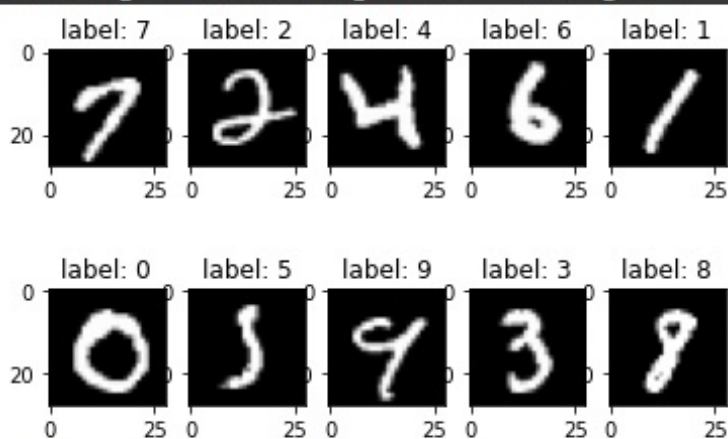
    plt.show()

# PART C
for i in range(0, 3):
    print("Selecting 10 random images from training set")
    digits = np.arange(0, 10)
    np.random.shuffle(digits)
    plot_figure(x_train, y_train, digits)
```

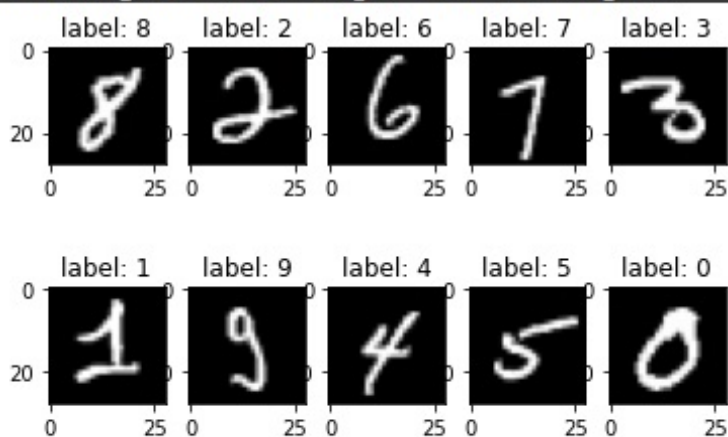
➔ Selecting 10 random images from training set



Selecting 10 random images from training set



Selecting 10 random images from training set



```
[ ] # PART D
num_train_img = x_train.shape[0]
train_ind=np.arange(0, num_train_img)
train_ind_S = np.random.permutation(train_ind)

x_train = x_train[train_ind_S,:,:]
y_train = y_train[train_ind_S]

# Selecting 20% of the training data
x_valid = x_train[0:int(0.2*num_train_img),:,:]
y_valid = y_train[0:int(0.2*num_train_img)]

print("x_valid:", x_valid.shape[0])
print("y_valid:", y_valid.shape[0])

print()

# The rest of the training set
x_train = x_train[int(0.2*num_train_img):,:,:]
y_train = y_train[int(0.2*num_train_img):]

print("x_train:", x_train.shape[0])
print("y_train:", y_train.shape[0])
```

```
x_valid: 12000
y_valid: 12000
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x_train: 48000
y_train: 48000
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