

Assignment5_ErenAkgunduz-1

February 18, 2024

1 Assignment 5

1.1 Eren Akgunduz

1.1.1 Deep Learning — 18 February 2024

1.1.2 [Link to notebook](#)

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

1.2 Problem 1

```
[2]: (x_train, y_train), (x_test, y_test) = mnist.load_data()
```

```
[3]: # Number of images in each training and testing set
print(x_train.shape[0], y_train.shape[0])
print(x_test.shape[0], y_test.shape[0])
```

60000 60000

10000 10000

```
[4]: # Image width and height
img_ws_hs = {x_train.shape[1], x_train.shape[2], x_test.shape[1], x_test.
    ↪shape[2]} # it's the same for all so only one value in set
img_pxs = list(img_ws_hs)[0]
print(img_ws_hs, img_pxs)
print(f"{img_pxs} x {img_pxs}")
```

{28} 28

28 x 28

```
[5]: # From provided code examples
def img_plt(images, labels):
    "Plots a figure with 10 subplots for each 0-9 digit"
    plt.figure() # figsize=(15,8)
    for i in range(1,11):
        plt.subplot(2,5,i)
        plt.imshow(images[i-1,:,:], cmap='gray')
```

```
plt.title('Label: ' + str(labels[i-1]))
plt.show()
```

```
[6]: # Shuffle the training set indices
num_train = x_train.shape[0]
train_ind = np.arange(0, num_train)
train_ind_shuffled = np.random.permutation(train_ind)
```

```
[7]: # Just to confirm
train_ind[:10]
```

```
[7]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
```

```
[8]: train_ind_shuffled[:10]
```

```
[8]: array([50966, 3978, 48962, 57185, 12814, 32445, 14080, 21062, 1251,
         48477])
```

```
[9]: x_train = x_train[train_ind_shuffled,:,:]
y_train = y_train[train_ind_shuffled]

# Select 20% of training data as validation set
x_valid = x_train[0:int(0.2 * num_train), :, :]
y_valid = y_train[0:int(0.2 * num_train)]

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

```
[10]: try:
        list({x_train.shape[0], y_train.shape[0]})[1]
    except IndexError:
        print("Number of images in training set:", list({x_train.shape[0], y_train.
↪shape[0]})[0])
        print("Number of images in validation set:", list({x_valid.shape[0],
↪y_valid.shape[0]})[0])
        print("Number of images in testing set:", list({x_test.shape[0], y_test.
↪shape[0]})[0])
    else:
        print("Oh no, seems like your data is not split properly ;) try again")
```

```
Number of images in training set: 48000
Number of images in validation set: 12000
Number of images in testing set: 10000
```

```
[11]: def plot_all(x_set, y_set, set_name):
        print(f"Selecting 10 images from {set_name} set")
        x_rnd = np.zeros((10, x_set.shape[1], x_set.shape[2]))
```

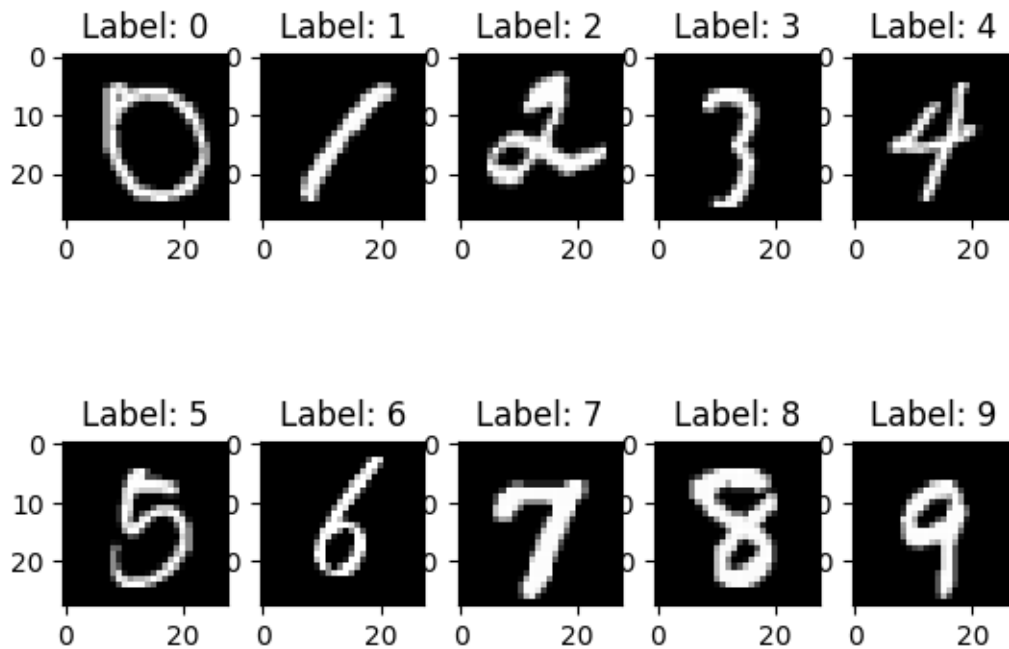
```

y_rnd = np.arange(10)
for digit in range(10):
    x_d = x_set[y_set == digit, :, :]
    x_rnd[digit, :, :] = x_d[0, :, :] # selecting first digit from the set
img_plt(x_rnd, y_rnd)

```

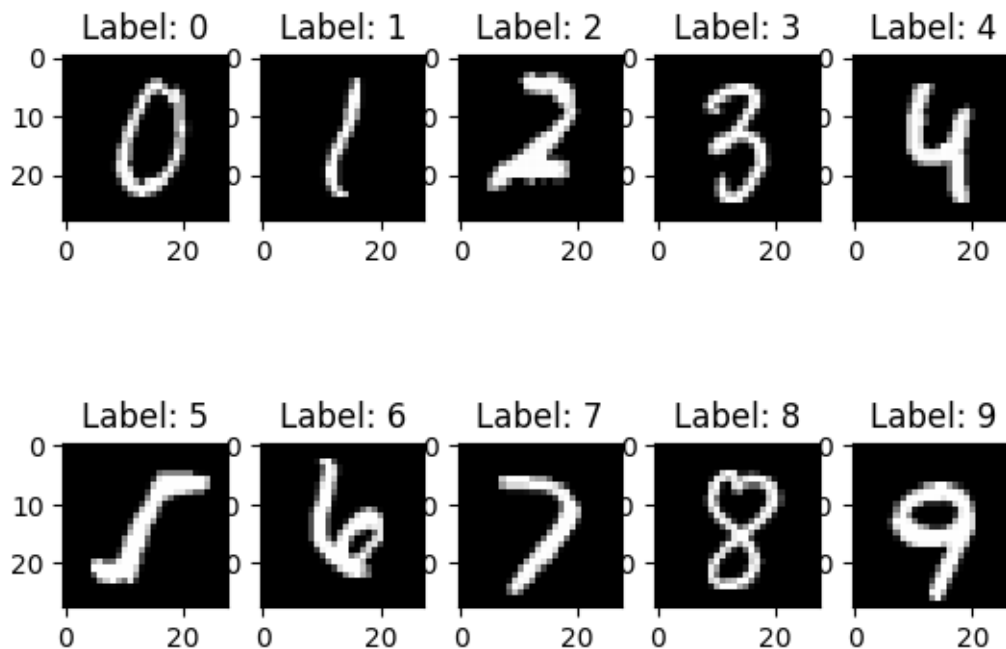
```
[12]: plot_all(x_train, y_train, "training")
```

Selecting 10 images from training set



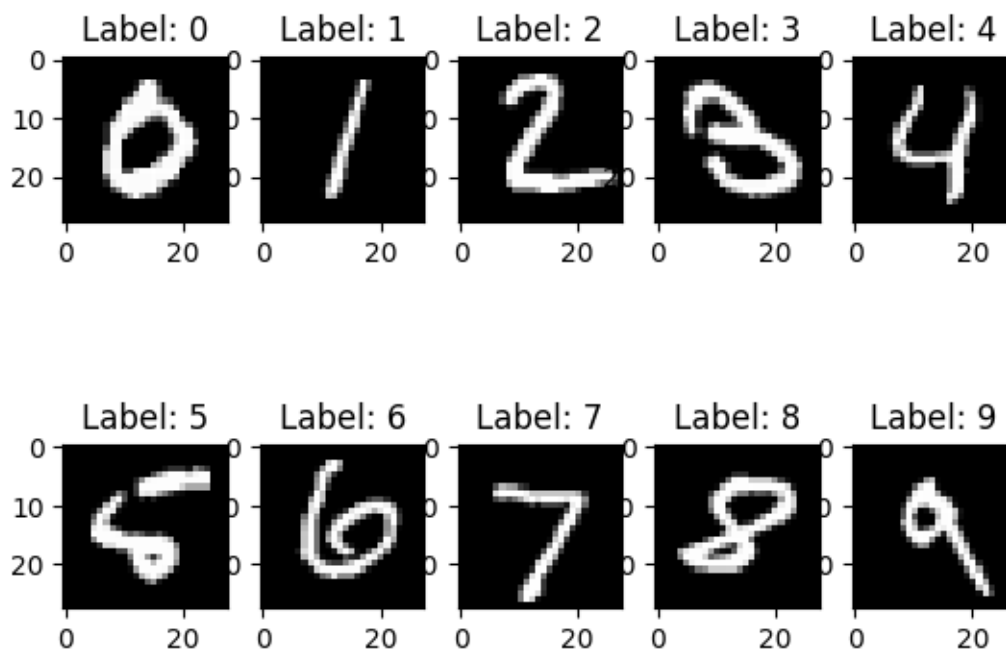
```
[13]: plot_all(x_valid, y_valid, "validation")
```

Selecting 10 images from validation set



```
[14]: plot_all(x_test, y_test, "testing")
```

Selecting 10 images from testing set



1.3 Problem 2

```
[15]: def feat_plt(feature, labels, digits):  
    plt.figure()  
    sample_nums = np.arange(feature.shape[0])  
    plt.plot(sample_nums[labels==digits[0]], feature[labels==digits[0]], 'gs',  
             sample_nums[labels==digits[1]], feature[labels==digits[1]], 'r^')  
    plt.legend([f'Digit {digits[0]}', f'Digit {digits[1]}'])  
    plt.xlabel('Sample #')  
    plt.ylabel('Average of the 3x3 center grid')  
    plt.title('Extrated feature from validation data')  
    plt.show()
```

```
[16]: def pred_fun(features, threshold, digits):  
    y_pred = np.ones(features.shape) * digits[0]  
    y_pred[features > threshold] = digits[1]  
    return y_pred
```

```
[17]: def acc_fun(labels_actual, labels_pred):  
    acc = np.sum(labels_actual==labels_pred) / len(labels_actual) * 100  
    return acc
```

```
[18]: (x_train, y_train), (x_test, y_test) = mnist.load_data()
```

```
[19]: x_train_01 = x_train[np.logical_or(y_train==0, y_train==1), :, :]  
    y_train_01 = y_train[np.logical_or(y_train==0, y_train==1)]  
    x_test_01 = x_test[np.logical_or(y_test==0, y_test==1), :, :]  
    y_test_01 = y_test[np.logical_or(y_test==0, y_test==1)]
```

```
[20]: print(x_train_01.shape, y_train_01.shape)
```

```
(12665, 28, 28) (12665,)
```

```
[21]: print(x_test_01.shape, y_test_01.shape)
```

```
(2115, 28, 28) (2115,)
```

```
[22]: # Shuffle the training set indices  
    num_train_01 = x_train_01.shape[0]  
    train01_ind = np.arange(0, num_train_01)  
    train01_ind_shuffled = np.random.permutation(train01_ind)  
  
    x_train_01 = x_train_01[train01_ind_shuffled, :, :]  
    y_train_01 = y_train_01[train01_ind_shuffled]  
  
    # Select 500 images for validation set  
    x_valid_01 = x_train_01[0:500, :, :]  
    y_valid_01 = y_train_01[0:500]
```

```
# The rest of the training set
x_train_01 = x_train_01[500:, :, :]
y_train_01 = y_train_01[500:]
```

```
[23]: x_train_01.shape
```

```
[23]: (12165, 28, 28)
```

```
[24]: features_train = np.sum(x_train_01[:, 13:16, 13:16], axis=2)
```

```
[25]: features_train.shape
```

```
[25]: (12165, 3)
```

```
[26]: features_train = np.sum(features_train, axis=1) / 9
```

```
[27]: features_train.shape
```

```
[27]: (12165,)
```

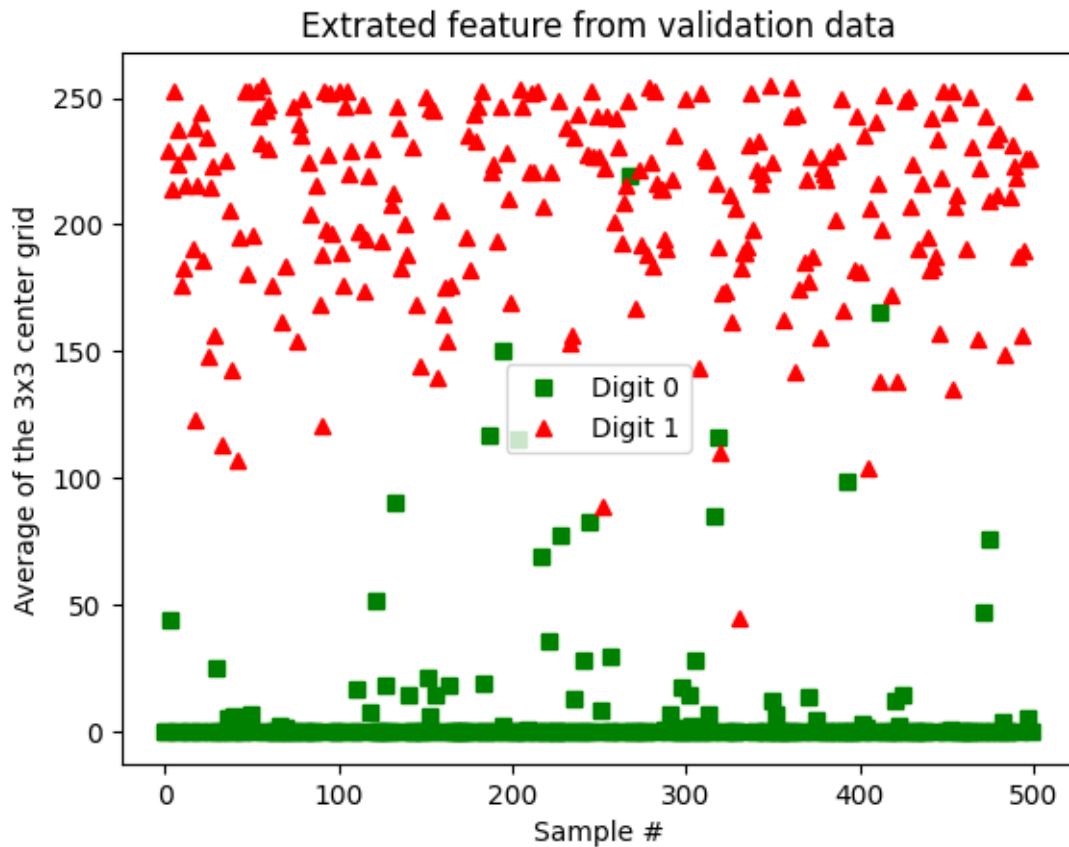
```
[28]: features_train
```

```
[28]: array([ 0.          ,  0.          , 212.66666667, ..., 197.77777778,
         0.          ,  0.          ])
```

```
[29]: features_val = np.sum(x_valid_01[:, 13:16, 13:16], axis=2)
      features_val = np.sum(features_val, axis=1) / 9
```

```
[30]: features_test = np.sum(x_test_01[:, 13:16, 13:16], axis=2)
      features_test = np.sum(features_test, axis=1) / 9
```

```
[31]: feat_plt(features_val, y_valid_01, [0, 1])
```



[32]: *# Time to guess a threshold and calculate the accuracies*

```

y_train_01_pred = np.zeros(features_train.shape)
y_val_01_pred = np.zeros(features_val.shape)
y_test_01_pred = np.zeros(features_test.shape)

threshold = 100

y_train_01_pred[features_train > threshold] = 1
y_train_01_pred[features_train <= threshold] = 0

y_val_01_pred[features_val > threshold] = 1
y_val_01_pred[features_val <= threshold] = 0

y_test_01_pred[features_test > threshold] = 1
y_test_01_pred[features_test <= threshold] = 0

print("Training accuracy:", acc_fun(y_train_01, y_train_01_pred))
print("Validation accuracy:", acc_fun(y_val_01, y_val_01_pred))

```

```
print("Testing accuracy:", acc_fun(y_test_01, y_test_01_pred))
```

Training accuracy: 98.38060008220305

Validation accuracy: 98.4

Testing accuracy: 99.10165484633569