Assignment5_ErenAkgunduz

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

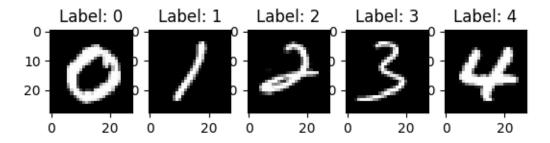
{28} 28 28 x 28

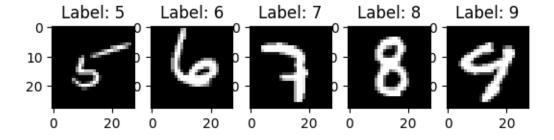
```
plt.figure() # fiqsize=(15,8)
          for i in range(1,11):
              plt.subplot(2,5,i)
              plt.imshow(images[i-1,:,:],cmap='gray')
              plt.title('Label: ' + str(labels[i-1]))
          plt.show()
[16]: # Shuffle the training set indices
      num train = x train.shape[0]
      train_ind = np.arange(0, num_train)
      train_ind_shuffled = np.random.permutation(train_ind)
[18]: # Just to confirm
      train_ind[:10]
[18]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
[19]: train_ind_shuffled[:10]
[19]: array([24551, 33152, 57043, 14668, 24996, 5256, 1884, 39580, 27605,
             31618])
[20]: 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):]
[29]: 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],__
       \rightarrowy_valid.shape[0]})[0])
          print("Number of images in testing set:", list({x_test.shape[0], y_test.
       ⇔shape[0]})[0])
          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
```

```
[30]: 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)
```

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

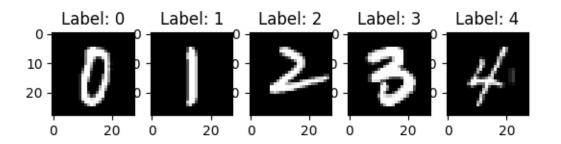
Selecting 10 images from training set

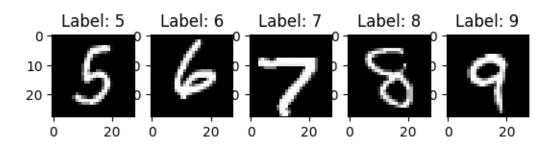




```
[32]: plot_all(x_valid, y_valid, "validation")
```

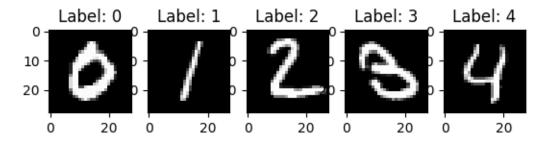
Selecting 10 images from validation set

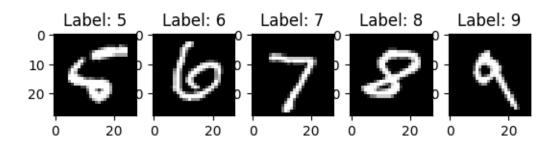




[33]: plot_all(x_test, y_test, "testing")

Selecting 10 images from testing set



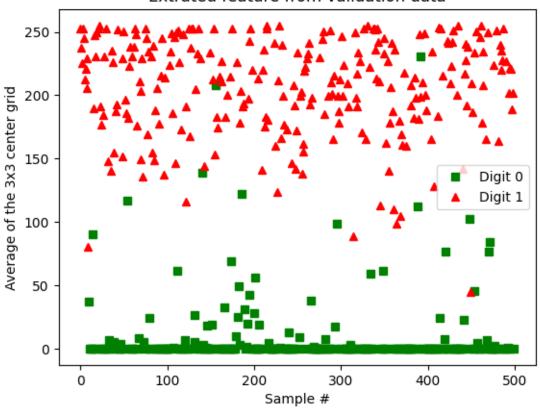


1.3 Problem 2

```
[68]: 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()
[69]: def pred_fun(features, threshold, digits):
          y_pred = np.ones(features.shape) * digits[0]
          y_pred[features > threshold] = digits[1]
          return y_pred
[70]: def acc_fun(labels_actual, labels_pred):
          acc = np.sum(labels_actual==labels_pred) / len(labels_actual) * 100
          return acc
[71]: (x_train, y_train), (x_test, y_test) = mnist.load_data()
[72]: 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)]
[73]: print(x_train_01.shape, y_train_01.shape)
     (12665, 28, 28) (12665,)
[74]: print(x_test_01.shape, y_test_01.shape)
     (2115, 28, 28) (2115,)
[75]: # 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:]
[76]: x_train_01.shape
[76]: (12165, 28, 28)
[77]: features_train = np.sum(x_train_01[:, 13:16, 13:16], axis=2)
[78]: features_train.shape
[78]: (12165, 3)
[79]: features_train = np.sum(features_train, axis=1) / 9
[80]: features_train.shape
[80]: (12165,)
[81]: features_train
[81]: array([ 0.
                         , 162.5555556,
                                           0.
                                                     , ..., 18.66666667,
               0.
                         , 198.1111111])
[84]: features_val = np.sum(x_valid_01[:, 13:16, 13:16], axis=2)
      features_val = np.sum(features_val, axis=1) / 9
[85]: features_test = np.sum(x_test_01[:, 13:16, 13:16], axis=2)
      features_test = np.sum(features_test, axis=1) / 9
[86]: feat_plt(features_val, y_valid_01, [0, 1])
```





```
[88]: # 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_valid_01, y_val_01_pred))</pre>
```

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

Training accuracy: 98.4052609946568

Validation accuracy: 97.8

Testing accuracy: 99.10165484633569

[]: