

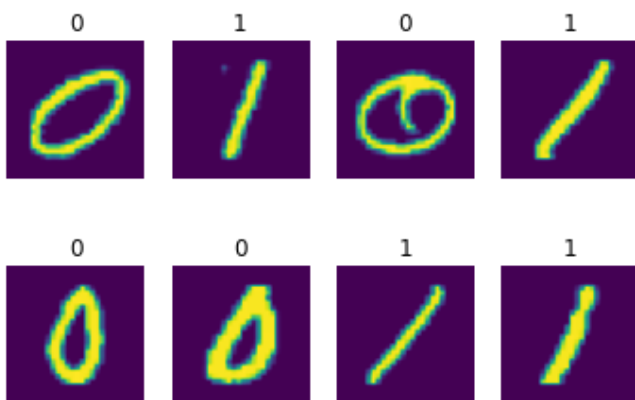
```
import numpy as np
from matplotlib import pyplot as plt
```

```
raw_img_data = np.load('data.npy')
label_data = np.load('label.npy')
```

```
normalized_img_data = raw_img_data / 255.0
normalized_imgshow_array = normalized_img_data.reshape(14780, 28, 28)
```

```
index = np.random.randint(0, len(normalized_imgshow_array), size=8)
i = 0

fig, axs = plt.subplots(2, 4)
for r in range(2):
    for c in range(4):
        axs[r, c].imshow(normalized_imgshow_array[index[i]].reshape(28, 28))
        axs[r, c].set_title(str(label_data[index[i]]))
        axs[r, c].axis('off')
        i+=1
```



```
relabelled_data = np.where(label_data == 1, -1, label_data)
relabelled_data = np.where(relabelled_data == 0, 1, relabelled_data)
```

```
(unique, counts) = np.unique(relabeled_data, return_counts=True)
frequencies_relabeled = np.asarray((unique, counts)).T
```

```
(unique, counts) = np.unique(label_data, return_counts=True)
frequencies_original = np.asarray((unique, counts)).T
```

```
print("original frequency counts")
print(frequencies_original)
print("after relabeling (1 --> -1, 0 --> 1)")
print(frequencies_relabeled)
```

```
original frequency counts
[[ 0 6903]
 [ 1 7877]]
after relabeling (1 --> -1, 0 --> 1)
[[ -1 7877]
 [ 1 6903]]
```

```
train_num_samples = int(0.8 * len(relabeled_data))
test_num_samples = len(relabeled_data) - train_num_samples
print("80-20 Test-Train Split = " + str(train_num_samples) + "-" + str(test_num_samples))
```

```
train_indices = np.random.randint(0, len(relabeled_data), size=train_num_samples)
test_indices = np.random.randint(0, len(relabeled_data), size=test_num_samples)
```

```
80-20 Test-Train Split = 11824-2956
```

```
X_train, X_test = normalized_img_data[train_indices], normalized_img_data[test_indices]
Y_train, Y_test = relabeled_data[train_indices], relabeled_data[test_indices]
```

```
print("Size of Train vs Test = " + str(len(X_train)) + "-" + str(len(X_test)))
```

```
Size of Train vs Test = 11824-2956
```

```
mu, sigma = 0, 1
d = 28*28
beta_0 = np.random.normal(mu, sigma, 1)[0]
beta_1 = np.random.normal(mu, sigma, d)
```

```
def loss_function(beta_0, beta_1, X, Y):
    """
    :param beta_0, beta_1: Logistic Regression coefficients
    :return: Loss over Training set (X and Y are globals)
    """
    loss_function_sum = 0.0
    m = len(X)
    for i in range(m):
        loss_function_sum += np.log(1 + np.exp(-Y[i] * (beta_0 + np.sum(beta_1 @ X[i, :]))))
    loss_function_result = loss_function_sum / m
    return loss_function_result
```

```
def compute_gradients(beta_0, beta_1, X, Y):
    """
    :return: Gradient of beta_0, Gradient of beta_1
    """
    d_beta_0_sum = 0.0
    d_beta_1_sum = 0.0
    m = len(X)
    for i in range(m):
        exponent_term = np.exp(-Y[i] * (beta_0 + beta_1.T @ X[i, :]))
        d_beta_0_sum += Y[i] * exponent_term / (1 + exponent_term)
        d_beta_1_sum += Y[i] * X[i] * exponent_term / (1 + exponent_term)

    d_beta_0 = (-1.0 / m) * d_beta_0_sum
    d_beta_1 = (-1.0 / m) * d_beta_1_sum
    return d_beta_0, d_beta_1
```

```
def evaluate_accuracy(beta_0, beta_1, X, Y):
    correct = 0
    for i in range(len(X)):
        predicted_label_prob = 1.0 / (1.0 + np.exp(Y[i] * (beta_0 + beta_1.T @ X[i])))
        prediction = 1 if predicted_label_prob > 0.5 else -1
        label = Y[i]
        is_equal = np.array_equal(prediction, label)
        if is_equal:
            correct += 1

    return (correct/len(X))
```

```

num_iterations = 50
learning_rate = 0.05

training_loss_array, training_accuracy_array = [], []
test_loss_array, test_accuracy_array = [], []
for iter in range(num_iterations):

    loss = loss_function(beta_0, beta_1, X_train, Y_train)
    d_beta_0, d_beta_1 = compute_gradients(beta_0, beta_1, X_train, Y_train)

    beta_0 = beta_0 - learning_rate * d_beta_0
    beta_1 = beta_1 - learning_rate * d_beta_1

    train_accuracy = evaluate_accuracy(beta_0, beta_1, X_train, Y_train)
    test_accuracy = evaluate_accuracy(beta_0, beta_1, X_test, Y_test)
    test_loss = loss_function(beta_0, beta_1, X_test, Y_test)

    training_loss_array.append(loss)
    training_accuracy_array.append(train_accuracy)
    test_loss_array.append(test_loss)
    test_accuracy_array.append(test_accuracy)

    if(iter % 5 == 0):
        print("[{: 2d}] Accuracy on test set: {:.4f}".format(iter, test_accuracy))

```

```

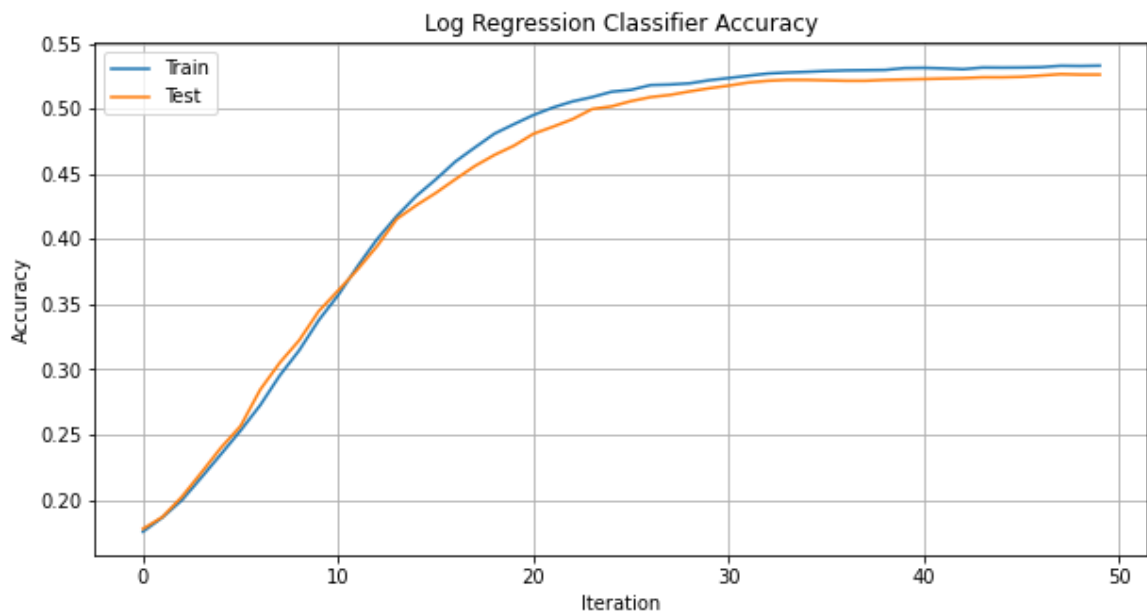
[ 0] Accuracy on test set: 0.1779
[ 5] Accuracy on test set: 0.2568
[10] Accuracy on test set: 0.3606
[15] Accuracy on test set: 0.4354
[20] Accuracy on test set: 0.4807
[25] Accuracy on test set: 0.5058
[30] Accuracy on test set: 0.5176
[35] Accuracy on test set: 0.5217
[40] Accuracy on test set: 0.5227
[45] Accuracy on test set: 0.5244

```

```

plt.rcParams['figure.figsize']=(10, 5)
plt.plot(training_accuracy_array)
plt.plot(test_accuracy_array)
plt.title('Log Regression Classifier Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Iteration')
plt.legend(['Train', 'Test'], loc='upper left')
plt.grid()
plt.show()

```



```
plt.plot(training_loss_array)
plt.plot(test_loss_array)
plt.title('Log Regression Classifier Loss')
plt.ylabel('Loss')
plt.xlabel('Iteration')
plt.legend(['Train', 'Test'], loc='upper left')
plt.grid()
plt.show()
```

