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import numpy as np
import data
import matplotlib.pyplot as plt
def re lu(x):
  return np.maximum(0, x)
def softmax(x):
  exp_x_shifted = np.exp(x - np.max(x))
  exp_sums = np.sum(exp_x_shifted, axis=1).reshape((-1, 1))
  probs = exp_x_shifted / exp_sums
  return probs
def fcann2_train(X, Y_, shape=(-1, 3), max_iter=100_000, eta=0.05, lamda=0.001, print_frequency=100):
  N = len(X)
  n = len(X[0])
  C = np.max(Y_) + 1
  hidden = shape[1]
  W1 = np.random.randn(n, hidden)
  b1 = np.zeros((1, hidden))
  W2 = np.random.randn(hidden, C)
  b2 = np.zeros((1, C))
  Y_one_hot = data.class_to_onehot(Y_)
  for i in range(max_iter):
    # FORWARD
    # 1st layer
    s1 = X @ W1 + b1
    h1 = re_lu(s1)
    # second layer
    s2 = h1 @ W2 + b2
    h2 = re_lu(s2)
    # loss
    P = softmax(h2)
    if i % print_frequency == 0:
       loss = -1.0 / N * np.sum(np.log(P) * Y_one_hot)
       print(f"iteration {i} loss = {loss}")
    # BACKWARD
    dL_ds2 = P - Y_one_hot
    dL dW2 = h1.T @ dL ds2
    dL_db2 = np.sum(dL_ds2, axis=0)
    dL_ds1 = dL_ds2 @ W2.T
    dL_ds1[s1 <= 0] = 0
    dL_dW1 = dL_ds1.T @ X
    dL_db1 = np.sum(dL_ds1, axis=0)
    # ADJUSTING WEIGHTS AND BIASES
    W1 = W1 * (1 - eta * lamda) - 1 / N * eta * dL_dW1.T
    b1 -= 1 / N * eta * dL_db1
    W2 = W2 * (1 - eta * lamda) - 1 / N * eta * dL_dW2
    b2 -= 1 / N * eta * dL_db2
  return W1, b1, W2, b2
def fcann2_classify(X, W1, b1, W2, b2):
  s1 = X @ W1 + b1
  h1 = re_{lu}(s1)
  s2 = h1 @ W2 + b2
  return softmax(s2)
```

```
def decfun(W1, b1, W2, b2):
    return lambda X: fcann2_classify(X, W1, b1, W2, b2)[:, 1]

if __name__ == "__main__":
    np.random.seed(100)
    X, Y_ = data.sample_gmm_2d(6, 2, 10)
    W1, b1, W2, b2 = fcann2_train(X, Y_, (-1, 5), 10_000, 0.05, 0.001, 500)
    probs = fcann2_classify(X, W1, b1, W2, b2)
    Y = np.argmax(probs, axis=1)
    bounding_box = (np.min(X, axis=0), np.max(X, axis=0))
    data.graph_surface(decfun(W1, b1, W2, b2), bounding_box, offset=0.5)
    data.graph_data(X, Y_, Y)
    plt.show()
```