```
1
     Example 4.4 Softmax decision boundary for the Iris dataset
 3
     @author: austin downey
 4
 5
 6
     import IPython as IP
 7
     IP.get ipython().magic('reset -sf')
8
9
     import numpy as np
10
     import matplotlib.pyplot as plt
11
     import sklearn as sk
12
13
14
    cc = plt.rcParams['axes.prop cycle'].by key()['color']
15
    plt.close('all')
16
17
18
    #%% Load your data
19
20
     # We will use the Iris data set. This dataset was created by biologist Ronald
21
     # Fisher in his 1936 paper "The use of multiple measurements in taxonomic
22
     # problems" as an example of linear discriminant analysis
23
     iris = sk.datasets.load iris()
24
25
     # for simplicity, extract some of the data sets
    X = iris['data'] # this contains the length of the petals and sepals
26
27
    Y = iris['target'] # contains what type of flower it is
28
    Y names = iris['target names'] # contains the name that aligns with the type of the
29
    feature_names = iris['feature_names'] # the names of the features
30
31
    # plot the Sepal data
32
   plt.figure(figsize=(6.5,3))
33
   plt.subplot(121)
34
   plt.grid(True)
35
   plt.scatter(X[Y==0,0],X[Y==0,1],marker='o',zorder=10)
    plt.scatter(X[Y==1,0],X[Y==1,1],marker='s',zorder=10)
36
37
    plt.scatter(X[Y==2,0],X[Y==2,1],marker='d',zorder=10)
38
    plt.xlabel(feature names[0])
39
    plt.ylabel(feature names[1])
40
41 plt.subplot(122)
42 plt.grid(True)
43 plt.scatter(X[Y==0,2],X[Y==0,3],marker='o',label=Y names[0],zorder=10)
    plt.scatter(X[Y==1,2],X[Y==1,3],marker='s',label=Y names[1],zorder=10)
44
    plt.scatter(X[Y==2,2],X[Y==2,3],marker='d',label=Y names[2],zorder=10)
45
46
   plt.xlabel(feature names[2])
47
    plt.ylabel(feature names[3])
48
    plt.legend(framealpha=1)
49
    plt.tight layout()
50
51
52
    #%% Softmax Regression
53
54
     # build the training and target set.
55
    X train = X[:, (2, 3)] # petal length, petal width
56
    y train = Y
57
58
     # build and train the softmax model
59
    softmax reg = sk.linear model.LogisticRegression(multi class="multinomial",
60
             solver="lbfgs", C=10)
61
    softmax reg.fit(X train, y train)
62
63
     # build the x values for the predictions over the entire "petal space"
64
    x grid, y grid = np.meshgrid(
65
             np.linspace(0, 7, 500),
66
             np.linspace(0, 4, 200),
```

```
67
    X new = np.vstack((x grid.reshape(-1), y grid.reshape(-1))).T # build a vector format of
68
    the mesh grid
69
70
    # predict on the vectorized format
71
    y predict = softmax reg.predict(X new)
72
    y_proba = softmax_reg.predict_proba(X_new)
73
74
75
    # convert back to meshgrid shape for plotting
76
    zz predict = y predict.reshape(x grid.shape)
77
    zz proba = y proba[:, 1].reshape(x grid.shape) # the selected column selects the
    probability that the data falls within this class.
78
79
    # plot the 2D "petal space"
80 plt.figure(figsize=(6.5, 4))
    plt.scatter(X[Y==0,2],X[Y==0,3],marker='o',label=Y names[0],zorder=10)
81
    plt.scatter(X[Y==1,2],X[Y==1,3],marker='s',label=Y names[1],zorder=10)
   plt.scatter(X[Y==2,2],X[Y==2,3],marker='d',label=Y names[2],zorder=10)
83
84 plt.contourf(x grid, y grid, zz predict, cmap='Pastel2')
85 contour = plt.contour(x grid, y grid, zz proba, [0.100,0.5,0.900], cmap=plt.cm.brg)
86 plt.clabel(contour, inline=1)
87 plt.xlabel(feature names[2])
88 plt.ylabel(feature_names[3])
   plt.legend()
89
90 plt.tight layout()
```