DSC 190 - Homework 02

Due: Wednesday, April 13

Write your solutions to the following problems by either typing them up or handwriting them on another piece of paper. Unless otherwise noted by the problem's instructions, show your work or provide some justification for your answer. Homeworks are due via Gradescope at 11:59 PM.

Problem 1.

Your friend really dislikes Gaussian basis functions, and says "linear basis functions are all you need". You don't agree – you say that linear basis functions won't give you the sort of nice, smooth, non-linear decision boundaries you saw in the lecture on radial basis function networks, no matter how many linear basis functions are used. In fact, you claim that if linear basis functions are used, then the decision boundary in the original space will remain linear.

More formally, let k be the number of linear basis functions used. Any linear basis function takes the form $\varphi(\vec{x}; \vec{a}) = \vec{a} \cdot \vec{x}$, where \vec{a} is a vector of parameters. So define $\varphi_i(\vec{x}; \vec{a}^{(i)}) = \vec{a}^{(i)} \cdot \vec{x}$ to be the ith linear basis function, where $\vec{a}^{(i)}$ is the ith vector in a set of k fixed vectors, $\{\vec{a}^{(1)}, \ldots, \vec{a}^{(k)}\}$.

Show that the prediction function $H(\vec{x}) = w_0 + w_1 \varphi_1(\vec{x}) + \ldots + w_k \varphi_k(\vec{x})$ is in fact linear in terms of the coordinates of \vec{x} . That is, if $\vec{x} = (x_1, \ldots, x_d)^T$, then H can be written as

$$H(\vec{x}) = w_0 + c_1 x_1 + \dots c_d x_d,$$

where c_0, \ldots, c_d are constants with respect to \vec{x} (the fact that H can be written in this way implies that the decision boundary is linear in the original space).

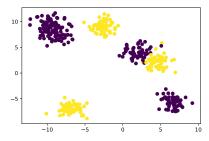
Hint: In particular, show that

$$c_j = \sum_{i=1}^k w_i \vec{a}_j^{(i)}$$

by plugging in the definition of φ_i to the definition of H above, expanding dot products, and grouping the terms involving each x_i . Here, recall that the notation $\vec{a}_j^{(i)}$ is referring to the jth entry of the vector $\vec{a}^{(i)}$; that is, it is a scalar.

Problem 2.

The CSV file located at https://f000.backblazeb2.com/file/jeldridge-data/001-blobs/data.csv contains a small data set of points in \mathbb{R}^2 along with labels in $\{-1,1\}$. The first two columns of the CSV are the two features x_1 and x_2 , and the third column is the label. When plotted, the data looks like the below:



The purple points belong to the negative class, while the yellow points belong to the positive class.

Train a simple radial basis function network on this data using Gaussian RBFs of the form seen in lecture:

$$\varphi_i(\vec{x}; \vec{\mu}, \sigma) = e^{-\|\vec{x} - \vec{\mu}\|^2 / \sigma^2}.$$

Use 6 centers placed at

$$\vec{\mu}^{(1)} = (-10, 10)^T$$

$$\vec{\mu}^{(2)} = (-3, 10)^T$$

$$\vec{\mu}^{(3)} = (-2, 5)^T$$

$$\vec{\mu}^{(4)} = (-5, 5)^T$$

$$\vec{\mu}^{(5)} = (-6, -5)^T$$

$$\vec{\mu}^{(6)} = (-7, -5)^T$$

and $\sigma = 4$. Learn the weights (w_0, \dots, w_6) of the RBF network by minimizing mean square loss in feature space (remember that we saw how to do this using numpy in Discussion 01).

This is a coding problem, but it is not autograded. Instead, you'll use your code to answer the questions below. For each part, include the code used to produce the answer. We'll grade your answers based on correctness, and we'll assign partial credit based on your code. You may complete this question in any language of your choosing, though we suggest Python since it is what is used in lecture and discussion.

- a) What is the weight vector \vec{w} that was learned by minimizing the mean square loss? Show the code used to train the RBF network, from reading in the data all the way to printing \vec{w} .
- b) What is the output of the radial basis function network's prediction function, H, at the point $\vec{x} = (-2,8)^T$? Your answer should be in the form of a real number don't report the sign of the prediction, but rather the number itself).

Show your code for computing $H(\vec{x})$.

c) What is the accuracy of this classifier on the training set?