DSC 255: Machine learning

## Solutions to HW 1

- 1. Casting an image into vector form. A  $10 \times 10$  greyscale image has 100 coordinates with 1 pixel per coordinate. Thus the corresponding vector has dimension d = 100.
- 2. The length of a vector. Say  $x \in \mathbb{R}^d$  where  $x_i = 1$  for i = 1, ..., d. Then by our Euclidean distance formula

$$||x|| = \sqrt{\sum_{i=1}^{d} x_i^2} = \sqrt{\sum_{i=1}^{d} 1} = \sqrt{d}$$

- 3. Euclidean distance.  $\sqrt{8}$ .
- 4. Accuracy of a random classifier.
  - (a) If there are four labels, then no matter what the correct label is, a random classifier has exactly a 25% chance of choosing it. Therefore it has an error rate of 75%.
  - (b) The best constant classifier is the one that always returns label A. It is wrong whenever the label isn't A, which occurs 50% of the time. Thus the classifier that always returns label A has error rate 50%.
- 5. We can work out the distances from the query to all the points.

Training point	Distance to query	label
(2,2)	$\sqrt{8.5}$	star
(2,4)	$\sqrt{2.5}$	square
(2,6)	$\sqrt{4.5}$	star
(4,2)	$\sqrt{6.5}$	square
(4,4)	$\sqrt{0.5}$	star
(4,6)	$\sqrt{2.5}$	square
(6,2)	$\sqrt{12.5}$	square
(6,4)	$\sqrt{6.5}$	square
(6,6)	$\sqrt{8.5}$	star

- (a) The closest point to the query is (4,4). So the point will be classified as star.
- (b) The 3 closest points to the query are (4,4), (2,4), and (4,6). So the point will be classified as square.
- (c) The 4 closest points to the query are (4,4), (2,4), (4,6), and (2,6). These are split 50/50 between star and square. The next closest point is a tie between (4,2) and (6,4). However, since both of these have the same label (square), the 5-NN classifier will label the query square no matter how it breaks ties.

- 6. In 4-fold cross-validation, we evenly divide our data set into 4 subsets. We hold out one subset and train on the rest. In our case, this means each time we train we will do so with 7,500 data points.
- 7. For 1-NN, the LOOCV procedure will misclassify the two right points. Thus the LOOCV error for 1-NN will be 50%.

For 3-NN, the LOOCV procedure will always label the test point +. Thus the LOOCV error for 3-NN will be 25%.

- 8. Nearest neighbor on MNIST.
  - (a) Training image #4711 was test point 100's nearest neighbor, and had class 4, meaning this test point was classified correctly. See Figure 1.

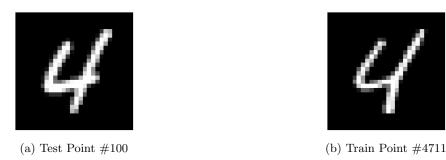


Figure 1: Test Point #100 and its Nearest Neighbor

(b) Here is some code for computing the confusion matrix:

```
confusion = np.zeros((10,10))
for i in range(0,1000):
    confusion[test_labels[i],test_predictions[i]] += 1
```

The confusion matrix is as follows. Digit 1 was misclassified the least often and digit 9 was misclassified the most often.

```
0
   100
                      0
                                  0
                                       0
                                  1
             1
0
             91
                      4
                          0
                              0
                                  1
                                       0
0
         0
                97
                     0
                          0
                                       3
1
                0
                     98
                         0
                                       1
0
         0
                 0
                      1
                          99
                                       0
0
             0
                      0
    4
         0
                 1
                          0
                                  0
                              94
                                       1
2
                      0
             1
                 1
                          1
                              1
                                  92
                                      1
```

(c) Here is code for computing the average of each digit and displaying it.

```
for digit in range(0,10):
    I = (train_labels==digit) # Find indices of training points with label 'digit'
    avg = np.mean(train_data[I,:],axis=0) # Compute the mean of these points
    show_digit(avg) # Display the mean
```

The following figures contain the mean image for every digit.

