

## CS 446: Machine Learning

### Homework 1

Due on Tuesday, January 23, 2018, 11:59 a.m. Central Time

1. [4 points] Intro to Machine Learning

Consider the task of classifying an image as one of a set of objects. Suppose we use a convolutional neural network to do so (you will learn what this is later in the semester).

- (a) For this setup, what is the data (often referred to as  $x^{(i)}$ )?

Your answer: The data will be the image to be classified.

- (b) For this setup, what is the label (often referred to as  $y^{(i)}$ )?

Your answer: The label will be the class which the image belongs to.

- (c) For this setup, what is the model?

Your answer: The model we used is Convolutional Neural Network.

- (d) What is the distinction between inference and learning for this task?

Your answer: Learning is associated with parameter estimation. After "learning", some parameters will be optimized. Inference is thought of as making some sort of prediction directly.

2. [8 points]  $K$ -Nearest Neighbors

*K-Nearest Neighbors* is an extension of the Nearest-Neighbor classification algorithm. Given a set of points with assigned labels, a new point is classified by considering the  $K$  points

closest to it (according to some metric) and selecting the most common label among these points. One common metric to use for KNN is the squared euclidean distance, i.e.

$$d(x^{(1)}, x^{(2)}) = \|x^{(1)} - x^{(2)}\|_2^2 \quad (1)$$

For this problem, consider the following set of points in  $\mathbb{R}^2$ , each of which is assigned with a label  $y \in \{1, 2\}$ :

$x_1$	$x_2$	$y$
1	1	2
0.4	5.2	1
-2.8	-1.1	2
3.2	1.4	1
-1.3	3.2	1
-3	3.1	2

- (a) Classify each of the following points using the Nearest Neighbor rule (i.e.  $K = 1$ ) with the squared euclidean distance metric.

Your answer:	$x_1$	$x_2$	$y$
	-2.6	6.6	1
	1.4	1.6	2
	-2.5	1.2	2

- (b) Classify each of the following points using the 3-Nearest Neighbor rule with the squared euclidean distance metric.

Your answer:	$x_1$	$x_2$	$y$
	-2.6	6.6	1
	1.4	1.6	1
	-2.5	1.2	2

- (c) Given a dataset containing  $n$  points, what is the outcome of classifying any additional point using the  $n$ -Nearest Neighbors algorithm?

Your answer: In this case  $k = n$ , the smoothing effect decreased variance which will effect overall performance negatively. It will result in underfitting.

- (d) How many parameters are *learned* when applying  $K$ -nearest neighbors?

Your answer: None. There are zero parameters being learned when applying K-nearest neighbors.