# Carnegie Mellon University in Qatar

AI for Medicine

15-182/282 - Spring 2022

# Assignment 3

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Due on: March 29, 2022 by midnight

## **Instructions:**

- $\bullet$  This assignment has a maximum score of 100 points for both, 15-182 and 15-282 students.
- $\bullet\,$  You should submit your solution through Gradescope.

Question	Points	Score
Learning a Linear Regression Model	50	
Malignant or Benign Tumour	50	
Total:	100	

### Problem 1: Learning a Linear Regression Model (50 Points)

As an example of supervised learning, consider the four points (1, 2), (2, 1), (3, 4), and (4, 3). These points can be thought of as a training set, wherein each point (x, y) includes a feature x and an associated label y. For instance, the point (1, 2) assumes x = 1 and y = 2; the other points can be interpreted similarly.

Suppose we want to "learn" the **hypothesis function**  $h_{\theta}(x) = \theta_0 + \theta_1 x$  that best fits the points of the training set. A natural interpretation of "best fits" is that a **cost function**, say, the **Mean Squared Error** (MSE) of the value of  $h_{\theta}(x)$  compared with the value of y over all the given points in the training set is minimized.

Answer the following questions after you show all your work.

12pts

- (a) What will be the value of MSE if:
  - i.  $\theta_0 = 0$  and  $\theta_1 = 1$
  - ii.  $\theta_0 = 1$  and  $\theta_1 = 0$
  - iii.  $\theta_0 = 1$  and  $\theta_1 = 3$
  - iv.  $\theta_0 = 1 \text{ and } \theta_1 = 3/5$

18pts

- (b) Starting off with  $\theta_0 = 0$  and  $\theta_1 = 0$ , what will be the values of  $\theta_0$  and  $\theta_1$  after 3 rounds of running gradient descent, assuming:
  - i. The learning rate  $\alpha = 0.5$
  - ii. The learning rate  $\alpha = 0.2$
  - iii. The learning rate  $\alpha = 0.7$

8pts

(c) This part is bonus for 15-182 and mandatory for 15-282: Solve this problem (i.e., finding the optimal  $\theta_0$  and  $\theta_1$  that best fit the given training set) through a system of equations rather than gradient descent. What are the optimal solutions for  $\theta_0$  and  $\theta_1$ ? (Hint: think of taking the derivatives of the cost function with respect to  $\theta_0$  then with respect to  $\theta_1$ ). Show all your work.

12pts

(d) This part is only for 15-282: Write a Python program that solves this problem using gradient descent. Does your program produce the same optimal solution for  $\theta_0$  and  $\theta_1$  as calculated in part c?

Assignment continues on the next page(s)

### Problem 2: Malignant or Benign Tumour (50 Points)

Let us consider training a model to recognize whether a tumour is malignant or benign in a given input (e.g., an MRI represented in a specific format). To this end, suppose you are given a training set that consists of pairs (x, y), where x is a vector of 0's and 1's, with each component  $x_i$  in the vector corresponding to the presence  $(x_i = 1)$  or absence  $(x_i = 0)$  of a particular feature in the input. The value of y is +1 if the input is known to have a malignant tumour and -1 if it is known to have a benign one.

In this problem, we will assume that there are five features, namely,  $x_1$ ,  $x_2$ ,  $x_3$ ,  $x_4$ , and  $x_5$  that dictate whether a tumour is malignant or benign in any given input. The table below shows a training set with six examples and their corresponding classes.

	<b>X</b> <sub>1</sub>	X <sub>2</sub>	<b>X</b> <sub>3</sub>	X <sub>4</sub>	<b>X</b> <sub>5</sub>	у
Example <b>1</b>	1	1	0	1	1	+1
Example 2	0	0	1	1	0	-1
Example 3	0	1	1	0	0	+1
Example <b>4</b>	1	0	0	1	0	-1
Example <b>5</b>	1	0	1	0	1	+1
Example <b>6</b>	1	0	1	1	0	-1

Figure 1: Six examples, each represented as a vector of 1's and 0's, where 1 or 0 indicates the presence or absence of a certain feature  $x_i$ , respectively. The label y indicates whether the example has a malignant (y = +1) or benign (y = -1) tumour.

Suppose we want to "learn" the **hypothesis function**  $h_{\theta}(x) = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_3 + \theta_4 x_4 + \theta_5 x_5$  that best fits the examples of the training set. For this sake, let us consider using Mean Squared Error (MSE) as a **cost function** and transforming the problem into a classification problem via using thresholding (Note: As we will see later in class when covering logistic regression, this is not necessarily a good way to do classification). In particular, if  $h_{\theta}(x) > 0$ , the tumour of the given input will be considered malignant, and if  $h_{\theta}(x) < 0$ , the tumour will be deemed benign (the special case where  $h_{\theta}(x) = 0$  will be assumed "wrong").

25pts

(a) Assuming a learning rate  $\alpha = 0.5$ , what will be the values of  $\theta_0$ ,  $\theta_1$ ,  $\theta_2$ ,  $\theta_3$ ,  $\theta_4$ , and  $\theta_5$  after one round of running gradient descent, if we start off with  $\theta_0 = 0$ ,  $\theta_1 = 0$ ,  $\theta_2 = 0$ ,  $\theta_3 = 0$ ,  $\theta_4 = 0$ , and  $\theta_5 = 0$ ? Do not only provide the answer but show all your work.

8pts

(b) Based on the above defined classification approach and the  $\theta$ s learnt from one round of running gradient descent, how many of the given examples will be classified correctly? Do not only provide the answer but show your work.

7pts

(c) Based on the above defined classification approach and the  $\theta$ s learnt from one round of running gradient descent, what will be the tumour type of this new unknown example  $[x_1 = 0, x_2 = 1, x_3 = 0, x_4 = 0, x_5 = 1]$ ?

10pts

(d) This part is only for 15-282: Write a Python program that implements this problem, assuming the training set given in Figure 1. You can stop your training phase after 10 iterations of gradient descent. Your program should be able to classify any new example (but of course, not necessarily correctly).