



# Linear Regression with One Variable

5 questions

1  
point

1.

Consider the problem of predicting how well a student does in her second year of college/university, given how well she did in her first year.

Specifically, let  $x$  be equal to the number of "A" grades (including A-, A and A+ grades) that a student receives in their first year of college (freshmen year). We would like to predict the value of  $y$ , which we define as the number of "A" grades they get in their second year (sophomore year).

Refer to the following training set of a small sample of different students' performances (note that this training set may also be referenced in other questions in this quiz). Here each row is one training example. Recall that in linear regression, our hypothesis is  $h_{\theta}(x) = \theta_0 + \theta_1 x$ , and we use  $m$  to denote the number of training examples.

$x$	$y$
3	4
2	1
4	3
0	1

For the training set given above, what is the value of  $m$ ? In the box below, please enter your answer (which should be a number between 0 and 10).

4

1  
point

2.

For this question, assume that we are

using the training set from Q1. Recall our definition of the

cost function was  $J(\theta_0, \theta_1) = \frac{1}{2m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)})^2$ .

What is  $J(0, 1)$ ? In the box below,

please enter your answer (Simplify fractions to decimals when entering answer, and '.' as the decimal delimiter e.g., 1.5).

0.5

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1  
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3. Suppose we set  $\theta_0 = -1, \theta_1 = 0.5$ . What is  $h_{\theta}(4)$ ?

1

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1  
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4.

Let  $f$  be some function so that

$f(\theta_0, \theta_1)$  outputs a number. For this problem,

$f$  is some arbitrary/unknown smooth function (not necessarily the cost function of linear regression, so  $f$  may have local optima).

Suppose we use gradient descent to try to minimize  $f(\theta_0, \theta_1)$

as a function of  $\theta_0$  and  $\theta_1$ . Which of the

following statements are true? (Check all that apply.)

- ☐ If the first few iterations of gradient descent cause  $f(\theta_0, \theta_1)$  to **increase** rather than decrease, then the most likely cause is that we have set the learning rate  $\alpha$  to too large a value.
- ☐ No matter how  $\theta_0$  and  $\theta_1$  are initialized, so long as  $\alpha$  is sufficiently small, we can safely expect gradient descent to converge to the same solution.
- ☐ If  $\theta_0$  and  $\theta_1$  are initialized at the global minimum, then one iteration will not change their values.
- ☐ Setting the learning rate  $\alpha$  to be very small is not harmful, and can only speed up the convergence of gradient descent.

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5.

Suppose that for some linear regression problem (say, predicting housing prices as in the lecture), we have some training set, and for our training set we managed to find some  $\theta_0, \theta_1$  such that  $J(\theta_0, \theta_1) = 0$ .

Which of the statements below must then be true? (Check all that apply.)

- ☐ Our training set can be fit perfectly by a straight line,  
i.e., all of our training examples lie perfectly on some straight line.
- ☐ Gradient descent is likely to get stuck at a local minimum and fail to find the global minimum.
- ☐ For this to be true, we must have  $\theta_0 = 0$  and  $\theta_1 = 0$   
so that  $h_{\theta}(x) = 0$
- ☐ For this to be true, we must have  $y^{(i)} = 0$  for every value of  $i = 1, 2, \dots, m$ .



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