

Advanced Algorithms

Floriano Zini

Free University of Bozen-Bolzano
Faculty of Computer Science

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Lab 12 – Linear regression and gradient descent

Assignment 10



Exercise 1

Consider the problem of predicting how well students do in their second year of college/university, given how well they did in their 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).

Exercises 1 through 3 will use the training set on the right of a small sample of different students' performances. 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 given training set, what is the value of m ?

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Exercise 2

For this question, continue to assume that we are using the training set given in the previous slide and let $J(\theta_0, \theta_1)$ be the cost function as defined in the lectures. What is $J(0, 1)$?

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Exercise 3

Suppose we set $\theta_0 = -2$, $\theta_1 = 0.5$. What is $h_\theta(6)$?

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Exercise 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 θ_0 and θ_1 . Which of the following statements are true? (Check all that apply.)

- Setting the learning rate a to be very small is not harmful, and can only speed up the convergence of gradient descent.
- If θ_0 and θ_1 are initialized so that $\theta_0 = \theta_1$, then by symmetry (because we do simultaneous updates to the two parameters), after one iteration of gradient descent, we will still have $\theta_0 = \theta_1$.
- If θ_0 and θ_1 are initialized at the global minimum, then one iteration of gradient descent will not change their values.
- 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 a to too large a value.

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Exercise 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 θ_0, θ_1 such that $J(\theta_0, \theta_1) = 0$. Which of the statements below must then be true? (Check all that apply.)

- We can perfectly predict the value of y even for new examples that we have not yet seen. (e.g., we can perfectly predict prices of even new houses that we have not yet seen.)
- For this to be true, we must have $\theta_0 = 0$ and $\theta_1 = 0$ so that $h_\theta(x) = 0$
- Our training set can be fit perfectly by a straight line, i.e., all of our training examples lie perfectly on some straight line.
- This is not possible: By the definition of $J(\theta_0, \theta_1)$, it is not possible for there to exist θ_0 and θ_1 so that $J(\theta_0, \theta_1) = 0$