## Congratulations! You passed!



1/1 point 1. Suppose *m*=4 students have taken some class, and the class had a midterm exam and a final exam. You have collected a dataset of their scores on the two exams, which is as follows:

midterm exam	(midterm exam) <sup>2</sup>	final exam
89	7921	96
72	5184	74
94	8836	87
69	4761	78

You'd like to use polynomial regression to predict a student's final exam score from their midterm exam score. Concretely, suppose you want to fit a model of the form  $h_{\theta}(x) = \theta_0 + \theta_1 x_1 + \theta_2 x_2$ , where  $x_1$  is the midterm score and  $x_2$  is (midterm score)². Further, you plan to use both feature scaling (dividing by the "max-min", or range, of a feature) and mean normalization.

What is the normalized feature  $x_1^{(1)}$ ? (Hint: midterm = 89, final = 96 is training example 1.) Please round off your answer to two decimal places and enter in the text box below.

Answer: 0.32



2. You run gradient descent for 15 iterations

with lpha=0.3 and compute

1/1 point

 $J(\theta)$  after each iteration. You find that the

value of  $J(\theta)$  **decreases** quickly then levels

off. Based on this, which of the following conclusions seems

most plausible?

Answer:  $\alpha = 0.3$  is an effective choice of learning rate.



1/1

point

3. Suppose you have m=23 training examples with n=5 features (excluding the additional allones feature for the intercept term, which you should add). The normal equation is  $\theta=(X^TX)^{-1}X^Ty$ . For the given values of m and n, what are the dimensions of  $\theta$ , X, and y in this equation?

Answer: X=23\*6; y=23\*1;  $\theta=6*1$ 



1/1

point

4. Suppose you have a dataset with m=50 examples and n=200000 features for each example. You want to use multivariate linear regression to fit the parameters  $\theta$  to our data. Should you prefer gradient descent or the normal equation?

Answer: Gradient descent, since  $(X^TX)^{-1}$  will be very slow to compute in the normal equation.



5. Which of the following are reasons for using feature scaling?

1/1 point

Answer: It speeds up gradient descent by making it require fewer iterations to get to a good solution.