Linear Regression with Multiple Variables

5 questions

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)	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 $x_2^{(2)}$? (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.

-0.37

2. You run gradient descent for 15 iterations

with $\alpha=0.3$ and compute

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

value of $J(\theta)$ decreases slowly and is still

decreasing after 15 iterations. Based on this, which of the

following conclusions seems most plausible?

- Rather than use the current value of α , it'd be more promising to try a smaller value of α (say $\alpha=0.1$).
- $\alpha=0.3$ is an effective choice of learning rate.
- Rather than use the current value of α , it'd be more promising to try a larger value of α (say $\alpha=1.0$).
- 3. Suppose you have m=28 training examples with n=4 features (excluding the additional all-ones feature for the intercept term, which you should add). The normal equation is $\theta = (X^T X)^{-1} X^T y$. For the given values of m and n, what are the dimensions of θ , X, and y in this equation?
 - X is 28×5 , y is 28×5 , θ is 5×5
 - $X ext{ is } 28 \times 4$, y is 28×1 , $\theta ext{ is } 4 \times 4$
 - X is 28×5 , y is 28×1 , θ is 5×1
 - $X ext{ is } 28 \times 4$, $y ext{ is } 28 \times 1$, $\theta ext{ is } 4 \times 1$
- 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 θ to our data. Should you prefer gradient descent or the normal equation?

	Gradient descent, since it will always converge to the optimal $ heta.$
	The normal equation, since it provides an efficient way to directly find the solution.
	Gradient descent, since $(X^TX)^{-1}$ will be very slow to compute in the normal equation.
	The normal equation, since gradient descent might be unable to find the optimal $ heta.$
5. Which	of the following are reasons for using feature scaling?
	It prevents the matrix $\boldsymbol{X}^T\boldsymbol{X}$ (used in the normal equation) from being non-invertable (singular/degenerate).
	It speeds up gradient descent by making it require fewer iterations to get to a good solution.
	It speeds up solving for $\boldsymbol{\theta}$ using the normal equation.
	It is necessary to prevent gradient descent from getting stuck in local optima.

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