Linear Regression with Multiple Variables

Quiz, 5 questions

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) ²	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 = 72, final = 74 is training example 2.) Please round off your answer to two decimal places and enter in the text box below.

.59

1 point

2

You run gradient descent for 15 iterations

with lpha=0.3 and compute J(heta) after each

iteration. You find that the value of $J(\theta)$ increases over

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

most plausible?

lpha=0.3 is an effective choice of learning rate.
Rather than use the current value of $lpha$, it'd be more promising to try a smaller value of $lpha$ (say $lpha=0.1$).

Rather than use the current value of α , it'd be more promising to try a larger value of α (say $\alpha=1.0$).

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3.

Suppose you have $m=14$ training examples with $n=3$ features (excluding the additional all-ones feature for the intercept
term, which you should add). The normal equation is $ heta=(X^TX)^{-1}X^Ty$. For the given values of m and n , what are the
dimensions of θ , X , and y in this equation?

- $igg(X ext{ is } 14 imes 4$, y is 14 imes 4 , heta is 4 imes 4
- X is 14×4 , y is 14×1 , θ is 4×1
- X is 14×3 , y is 14×1 , θ is 3×3
- igcap X is 14 imes 3, y is 14 imes 1, heta is 3 imes 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 θ to our data. Should you prefer gradient descent or the normal equation?

- 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.
- 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.

1 point

5.

Which of the following are reasons for using feature scaling?

- It prevents the matrix X^TX (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 is necessary to prevent the normal equation from getting stuck in local optima.
- It speeds up gradient descent by making each iteration of gradient descent less expensive to compute.

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