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.

Enter answer here	
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1 point

2.

You run gradient descent for 15 iterations Linear Regression with Multiple Variables

Quiz, 5 questions $\,$ 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 α
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 $\alpha = 1.0$).

1 point

3.

Suppose you have m=23 training examples with n=5 features (excluding the additional all-ones 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 θ , X, and y in this equation?

$$X$$
 is 23×5 , y is 23×1 , θ is 5×5

$$X$$
 is 23 × 6, y is 23 × 6, θ is 6 × 6

$$X ext{ is } 23 \times 6$$
, $y ext{ is } 23 \times 1$, $\theta ext{ is } 6 \times 1$

$$X$$
 is 23×5 , y is 23×1 , θ is 5×1

1 point

4.

The normal equation, since gradient descent might be unable to find the optimal θ. The normal equation, since it provides an efficient way to directly find the solution. Gradient descent, since (X ^T X) ⁻¹ will be very slow to compute in the normal equation. Gradient descent, since it will always converge to the optimal θ. 1 point 5. Which of the following are reasons for using feature scaling? It speeds up solving for θ using the normal equation. It prevents the matrix X ^T 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 is necessary to prevent gradient descent from getting stuck in local optima. I, George Wolberg, understand that submitting work that isn't my own may result in permanent failure of this course or deactivation of my Coursera account. Learn more about Coursera's Honor Code	Linear Regress	se you have a dataset with $m=50$ examples and $n=200000$ features for 100000 features for 100000 features for 100000 from 100000 features for 1000000 features for 10000000 features for 100000000 features for 1000000000 features for 1000000000000 features for $1000000000000000000000000000000000000$
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