**Environment Setup** Instructions **Multivariate Linear** Regression **Computing Parameters** Analytically **Submitting Programming Assignments** 

Reading: Lecture Slides 20 min

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TO PASS 80% or higher

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**DUE** Jul 19, 11:59 PM PDT **ATTEMPTS** 3 every 8 hours

Quiz: Linear Regression with Multiple Variables 5 questions

Octave/Matlab Tutorial

Video: Basic Operations 13 min

16 min Video: Computing on Data 13 min

Video: Moving Data Around

Video: Plotting Data 9 min

Video: Control Statements: for, while, if statement

12 min ▶ **Video:** Vectorization

13 min Review

Reading: Lecture Slides

Quiz: Octave/Matlab Tutorial

5 questions

Programming Assignment: Linear Regression

QUIZ • 10 MIN	
Linear Regression with Multiple Variable	near Regression with Multiple Variables

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 Try again 89 96

5184 74 94 **Grade** 87 View Feedback 69 **100**% 78 We keep your highest score

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  $h_ heta(x)= heta_0+ heta_1x_1+ heta_2x_2$ , 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  $x_1^{(3)}$ ? (Hint: midterm = 94, final = 87 is training example 3.) Please round off your answer to two decimal places and enter in the text box below.

Enter answer here

2. You run gradient descent for 15 iterations 1 point with lpha=0.3 and compute J( heta) after each iteration. You find that the

value of J( heta) **decreases** quickly then levels

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

igcap Rather than use the current value of lpha, it'd be more promising to try a smaller value of lpha (say lpha=0.1). igcap Rather than use the current value of lpha, it'd be more promising to try a larger value of lpha (say lpha=1.0).

 $\bigcirc \ lpha = 0.3$  is an effective choice of learning rate.

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 heta, X, and y in this equation?

igcap X is 14 imes3 , y is 14 imes1 , heta is 3 imes3

igcap X is 14 imes3 , y is 14 imes1 , heta is 3 imes1

igcap X is 14 imes 4 , y is 14 imes 1 , heta is 4 imes 1igcap X is 14 imes 4 , y is 14 imes 4 , heta is 4 imes 4

4. Suppose you have a dataset with m=50 examples and n=15 features for each example. You want to use multivariate linear regression to fit the parameters heta to our data. Should you prefer gradient descent or the

normal equation? igcup Gradient descent, since it will always converge to the optimal heta.

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

The normal equation, since it provides an efficient way to directly find the solution.

igcup 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? 1 point  $oxedsymbol{\square}$  It speeds up solving for heta using the normal equation.

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.

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1 point

1 point

1 point