CS 383 - Machine Learning

Assignment 3 - Linear Regression

1 Theory

1. (10pts) Consider the following supervised dataset:

$$X = \begin{bmatrix} -2 \\ -5 \\ -3 \\ 0 \\ -8 \\ -2 \\ 1 \\ 5 \\ -1 \\ 6 \end{bmatrix}, Y = \begin{bmatrix} 1 \\ -4 \\ 1 \\ 3 \\ 11 \\ 5 \\ 0 \\ -1 \\ -3 \\ 1 \end{bmatrix}$$

Compute the coefficients for the linear regression using least squares estimate (LSE). Show your work and remember to add a bias feature. Compute this model using **all** of the data (don't worry about separating into training and testing sets).

$$\theta = (X^T X)^{-1} (X^T Y)$$

$$(X^T X)^{-1} = \frac{1}{(10 * 169) - ((-9) * (-9))} = \frac{1}{1609} \begin{bmatrix} 169 & 9 \\ 9 & 10 \end{bmatrix} = \begin{bmatrix} \frac{169}{1609} & \frac{9}{1609} \\ \frac{9}{1609} & \frac{10}{1609} \end{bmatrix}$$

$$(X^T X)^{-1} (X^T Y) = \begin{bmatrix} \frac{169}{1609} & \frac{9}{1609} \\ \frac{9}{1609} & \frac{10}{1609} \end{bmatrix} \begin{bmatrix} 14 \\ -79 \end{bmatrix} = \begin{bmatrix} 1.028 \\ -0.400 \end{bmatrix}$$

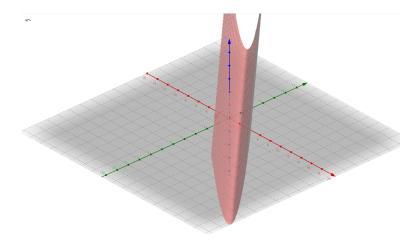
For the function $J = (x_1 + x_2 - 2)^2$, where x_1 and x_2 are a single valued variables (not vectors):

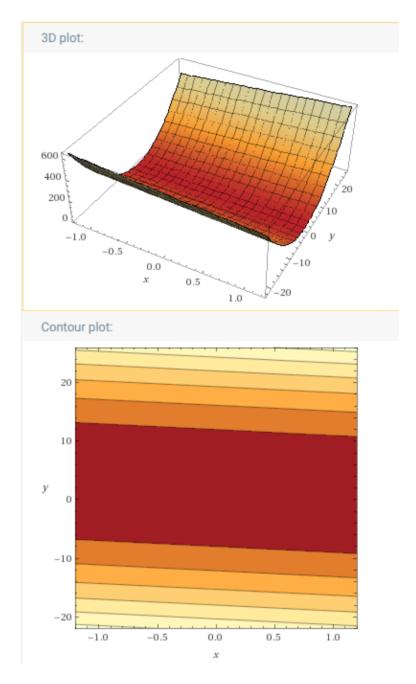
(a) What are the partial gradients, $\frac{\partial J}{\partial x_1}$ and $\frac{\partial J}{\partial x_2}$? Show work to support your answer. (4pts)

The square rule for partial derivatives state that for a function raised to the power of two, its derivative will be that function times 2. Ex: $x^2 = 2x$

$$\frac{\partial J}{\partial x_1} = 2(\mathbf{x}_1 + x_2 - 2)$$
$$\frac{\partial J}{\partial x_2} = 2(\mathbf{x}_1 + x_2 - 2)$$

(b) Create a 3D plot of x_1 vs x_2 , vs J use a software package of your choosing. (4pts)





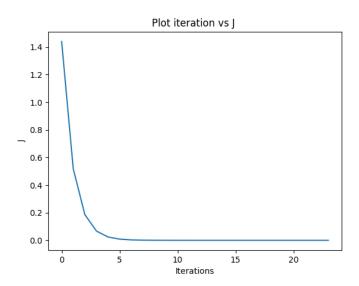
(c) Based on your plot, what are the values of x_1 and x_2 that minimize J? (2pts) in the form $[x_1, x_2]$ the values that minimize J are: [1,1] [2,0] [0,2]

2 Gradient Descent

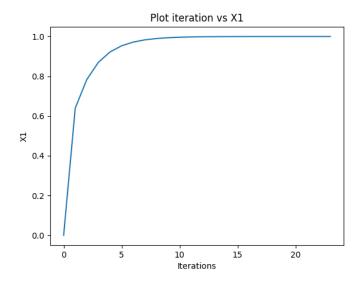
In this section we want to visualize the gradient descent process on the function $J = (x_1 + x_2 - 2)^2$. You should have already derived (pun?) the gradient of this function in the theory section. To bootstrap the process, initialize $x_1 = 0$, $x_2 = 0$ and terminate the process when the change in the J from one iteration to another is less that 2^{-32} . In addition, we'll use a learning rate of $\eta = 0.1$.

In your report you will need

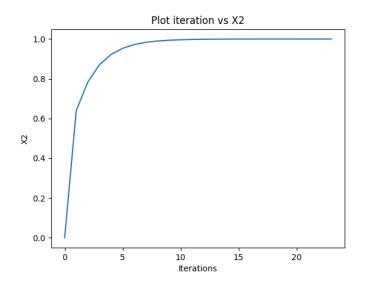
(a) Plot iteration vs J



(b) Plot iteration vs x_1



(c) Plot iteration vs x_2



3 Closed Form Linear Regression

(a) The final model in the form $y = \theta_0 + \theta_1 x_1 + \dots$

$$\mathbf{g}(\mathbf{x}) = [-142.97479460381078] + [4.6105534994738235]\mathbf{x}_1 + [0.031787731706128336]x_2$$

(b) The root mean squared error.

Root Mean Squared Error: 20.19215923804834

4 S-Folds Cross-Validation

(a) The average and standard deviation of the root mean squared error for S=2 over the 20 different seed values..

 $When S = 2 \\ Mean RMSE = 21.698287881816135 \\ Standard Deviation = 1.505806466802261$

(b) The average and standard deviation of the root mean squared error for S=4 over the 20 different seed values.

When S = 4Mean RMSE = 21.30202763614809Standard Deviation = 0.8171386807576563

(c) The average and standard deviation of the root mean squared error for S=22 over 20 different seed values.

 $\begin{aligned} \text{When S} &= 22 \\ \text{Mean RMSE} &= 21.01623896621227 \\ \text{Standard Deviation} &= 0.21324491972491713 \end{aligned}$

(d) The average and standard deviation of the root mean squared error for S=N (where N is the number of samples) over 20 different seed values. This is basically *leave-one-out* cross-validation.

 $\label{eq:When S = 44} When S = 44 \\ Mean RMSE = 21.004621083284924 \\ Standard Deviation is = 2.7519201823675253e-15$