

# GRADIENT DESCENT EXERCISE

a)

Let  $\Phi: \mathbb{R}^2 \rightarrow \mathbb{R}^6$  be the function that transforms the vector

$$x^t = (x_1 \ x_2) \text{ into } \Phi(x)^t = (1 \ x_1 \ x_2 \ x_1 x_2 \ x_1^2 \ x_2^2).$$

Now, let  $X$  be the matrix with the transformed samples in its rows.

$$X = \begin{bmatrix} \Phi(x_1) \\ \vdots \\ \Phi(x_n) \end{bmatrix}$$

The matrix is now set up for regression.

$$\hat{\beta} = \underset{\beta}{\operatorname{argmin}} \underbrace{\sum_{i=1}^n (\Phi(x_i) \beta - y_i)^2}_{J(\beta)}$$

Applying Gradient Descent yields the following:

$$\beta^{(k+1)} = \beta^{(k)} - \eta \nabla J(\beta^{(k)}).$$

Now Computing the gradient w.r.t.  $\beta$  of  $L_2(f_{\beta}(x_i), y_i)$ .

$$\nabla_{\beta} L_2(f_{\beta}(x_i), y_i) = 2(\Phi(x_i) \beta - y_i) \begin{bmatrix} 1 \\ x_i^{(1)} \\ x_i^{(2)} \\ x_i^{(1)} x_i^{(2)} \\ (x_i^{(1)})^2 \\ (x_i^{(2)})^2 \end{bmatrix}$$

where  $x_i^{(j)}$  indicates feature  $j$  of sample  $i$  from the data-set.

b)

$$\beta^{(0)} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\eta = 0.1$$

$$R^{(1)} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$n = 1 \text{ to } (n-2)$$

$$\begin{bmatrix} 1 \\ 4 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0.4 \\ 1.6 \\ 0.4 \\ 1 \end{bmatrix}$$

$$\rightarrow \beta = \begin{bmatrix} 0 \\ 0 \end{bmatrix} - 0.1 \cdot \begin{bmatrix} 1 \\ 16 \\ 1 \end{bmatrix} \begin{bmatrix} 1.6 \\ 6.4 \\ 0.4 \end{bmatrix}$$

$$\rightarrow \beta^{(2)} = \begin{bmatrix} 0.4 \\ 1.6 \\ 0.4 \\ 1.6 \\ 6.4 \\ 0.4 \end{bmatrix} - 0.1 \cdot 2 \left( \underbrace{0.4 + 1.6 \cdot 2 + 0.4 \cdot 8 + 1.6 \cdot 16 + 6.4 \cdot 4 + 0.4 \cdot 64 + 14}_{87.6} \right) \begin{bmatrix} 1 \\ 2 \\ 8 \\ 16 \\ 4 \\ 64 \end{bmatrix}$$

$$\beta^{(2)} = \begin{bmatrix} 0.4 \\ 1.6 \\ 0.4 \\ 1.6 \\ 6.4 \\ 0.4 \end{bmatrix} - 19.52 \begin{bmatrix} 1 \\ 2 \\ 8 \\ 16 \\ 4 \\ 64 \end{bmatrix} = \begin{bmatrix} -19.12 \\ -37.44 \\ -155.76 \\ -310.72 \\ -71.68 \\ -1248.88 \end{bmatrix}$$

$$c) \quad \beta^{(1)} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} - 0.2 \sum_{i=1}^2 (\Phi(x_i) \beta^{(0)} - y_i) \vec{\Phi}(x_i)$$

$$\beta^{(1)} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} - 0.2 \left( \begin{bmatrix} -2 \\ -8 \\ -2 \\ -8 \\ -32 \\ -2 \end{bmatrix} + \begin{bmatrix} 14 \\ 28 \\ 112 \\ 224 \\ 56 \\ 896 \end{bmatrix} \right)$$

$$= \begin{bmatrix} -2.4 \\ -4 \\ -22 \\ -43.2 \\ -4.8 \\ -178.8 \end{bmatrix}$$

$$\begin{bmatrix} -2.4 \end{bmatrix}$$

$\rho$

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\beta^{(2)} = \begin{bmatrix} -4 \\ -22 \\ -43.2 \\ -4.8 \\ -178.8 \end{bmatrix} - 0.2 \left( \begin{bmatrix} -2.4 & -4 & -4.8 & -1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix} + \begin{bmatrix} -2.4 & -12 & -44 & - \end{bmatrix} \right)$$

-12.2

$$\begin{aligned} & \left( 259.2 - 43.2 - 715.2 + 1 \right) \begin{bmatrix} 1 \\ 3 \\ 2 \\ 6 \\ 9 \\ 4 \end{bmatrix} \\ & \quad \quad \quad 1075 \end{aligned}$$

$$\beta^{(2)} = \begin{bmatrix} -2.4 \\ -4 \\ -22 \\ -43.2 \\ -4.8 \\ -178.8 \end{bmatrix} + \begin{bmatrix} 2.44 \\ 2.44 \\ 0 \\ 0 \\ 2.44 \\ 0 \end{bmatrix} * \begin{bmatrix} 215 \\ 645 \\ 430 \\ 1290 \\ 1935 \\ 860 \end{bmatrix}$$

$$\beta^{(2)} = \begin{bmatrix} 215.04 \\ 643.44 \\ 408 \\ 1246.8 \\ 1932.64 \\ 681.2 \end{bmatrix}$$