

AI Assignment 1

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1.1 We have $\theta = [\theta_0, \theta_1, \theta_2]^T$ and $x_i = [1, x_{i0}, x_{i1}]^T$ for $h_\theta(x) = \theta^T x$.

So we derive for the gradient:

$$\begin{aligned}\frac{\partial L(\theta)}{\partial \theta_0} &= \frac{\partial}{\partial \theta_0} \sum_i (\theta_0 + \theta_1 x_{i0} + \theta_2 x_{i1} - y_i)^2 \\ &= \sum_i \frac{\partial}{\partial \theta_0} (\theta_0 + \theta_1 x_{i0} + \theta_2 x_{i1} - y_i)^2 \\ &= \sum_i 2(\theta_0 + \theta_1 x_{i0} + \theta_2 x_{i1} - y_i)\end{aligned}$$

Doing the same for θ_1 and θ_2 we get:

$$\begin{aligned}\frac{\partial L(\theta)}{\partial \theta_1} &= \sum_i 2(\theta_0 + \theta_1 x_{i0} + \theta_2 x_{i1} - y_i) * x_{i0} \\ \frac{\partial L(\theta)}{\partial \theta_2} &= \sum_i 2(\theta_0 + \theta_1 x_{i0} + \theta_2 x_{i1} - y_i) * x_{i1}\end{aligned}$$

To find a minimum by gradient descent we use:

$$\theta^{k+1} \leftarrow \theta^k - \eta_k \frac{\partial L(\theta)}{\partial \theta} \text{ where we have } \eta = 0.01 \text{ for all } k.$$

$$\text{Thus, } \theta^{k+1} \leftarrow \theta^k - 0.01 * \frac{\partial L(\theta)}{\partial \theta}.$$

Initial parameter $\theta^0 = [0, 0, 0]^T$, so compute for θ^1 simultaneously:

$$\begin{aligned}\theta_0^1 &= \theta_0^0 - 0.01 * \sum_i 2(\theta_0^0 + \theta_1^0 x_{i0} + \theta_2^0 x_{i1} - y_i) \\ &= 0 - 0.01 * \sum_i 2(0 + 0 * x_{i0} + 0 * x_{i1} - y_i) \\ &= 0.02 * \sum_i y_i = 0.02 * 37 = 0.74.\end{aligned}$$

$$\begin{aligned}
\theta_1^1 &= \theta_1^0 - 0.01 * \sum_i 2(\theta_0^0 + \theta_1^0 x_{i0} + \theta_2^0 x_{i1} - y_i) * x_{i0} \\
&= 0 - 0.01 * \sum_i 2(0 + 0 * x_{i0} + 0 * x_{i1} - y_i) * x_{i0} \\
&= 0.02 * \sum_i (y_i * x_{i0}) \\
&= 0.02 * (6 * 1 + 4 * 2 + 7 * 3 + 10 * 3 + 10 * 4) = 0.02 * 105 = 2.1.
\end{aligned}$$

$$\begin{aligned}
\theta_2^1 &= \theta_2^0 - 0.01 * \sum_i 2(\theta_0^0 + \theta_1^0 x_{i0} + \theta_2^0 x_{i1} - y_i) * x_{i1} \\
&= 0 - 0.01 * \sum_i 2(0 + 0 * x_{i0} + 0 * x_{i1} - y_i) * x_{i1} \\
&= 0.02 * \sum_i (y_i * x_{i1}) \\
&= 0.02 * (6 * 2 + 4 * 1 + 7 * 2 + 10 * 3 + 10 * 5) = 0.02 * 110 = 2.2.
\end{aligned}$$

So for the first iteration we have $\theta^1 = [0.74, 2.1, 2.2]^T$.

Second iteration:

$$\begin{aligned}
\theta_0^2 &= \theta_0^1 - 0.01 * \sum_i 2(\theta_0^1 + \theta_1^1 x_{i0} + \theta_2^1 x_{i1} - y_i) \\
&= 0.74 - 0.01 * \sum_i 2(0.74 + 2.1x_{i0} + 2.2x_{i1} - y_i) \\
&= 0.74 - 0.02 * 22.6 = 0.288.
\end{aligned}$$

$$\begin{aligned}
\theta_1^2 &= \theta_1^1 - 0.01 * \sum_i 2(\theta_0^1 + \theta_1^1 x_{i0} + \theta_2^1 x_{i1} - y_i) * x_{i0} \\
&= 2.1 - 0.01 * \sum_i 2(0.74 + 2.1x_{i0} + 2.2x_{i1} - y_i) * x_{i0} \\
&= 2.1 - 0.02 * \sum_i (0.74 + 2.1x_{i0} + 2.2x_{i1} - y_i) * x_{i0} = 0.6536
\end{aligned}$$

$$\begin{aligned}
\theta_2^2 &= \theta_2^1 - 0.01 * \sum_i 2(\theta_0^1 + \theta_1^1 x_{i0} + \theta_2^1 x_{i1} - y_i) * x_{i1} \\
&= \theta_2^1 - 0.02 * \sum_i (\theta_0^1 + \theta_1^1 x_{i0} + \theta_2^1 x_{i1} - y_i) * x_{i1}
\end{aligned}$$

$$= 2.2 - 0.02 * \sum_i (0.74 + 2.1x_{i0} + 2.2x_{i1} - y_i) * x_{i1} = 0.6776.$$

$$\text{Thus, } \theta^2 = [0.288, 0.6536, 0.6776]^T.$$

$$1.2 \quad \theta^* = (X^T X)^{-1} X^T y$$

$$\begin{aligned} &= \left(\begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 2 & 1 & 2 & 3 & 5 \\ 1 & 2 & 3 & 3 & 4 \end{bmatrix} * \begin{bmatrix} 1 & 1 & 2 \\ 1 & 2 & 1 \\ 1 & 3 & 2 \\ 1 & 3 & 3 \\ 1 & 4 & 5 \end{bmatrix} \right)^{-1} * \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 2 & 1 & 2 & 3 & 5 \\ 1 & 2 & 3 & 3 & 4 \end{bmatrix} * \begin{bmatrix} 6 \\ 4 \\ 7 \\ 10 \\ 10 \end{bmatrix} \\ &= \left(\begin{bmatrix} 5 & 13 & 13 \\ 13 & 39 & 43 \\ 13 & 39 & 39 \end{bmatrix} \right)^{-1} * \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 2 & 1 & 2 & 3 & 5 \\ 1 & 2 & 3 & 3 & 4 \end{bmatrix} * \begin{bmatrix} 6 \\ 4 \\ 7 \\ 10 \\ 10 \end{bmatrix} \\ &= \begin{bmatrix} 3/2 & 0 & -1/2 \\ -1/2 & -1/4 & 23/52 \\ 0 & 1/4 & -1/4 \end{bmatrix} * \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 2 & 1 & 2 & 3 & 5 \\ 1 & 2 & 3 & 3 & 4 \end{bmatrix} * \begin{bmatrix} 6 \\ 4 \\ 7 \\ 10 \\ 10 \end{bmatrix} \\ &= \begin{bmatrix} 1 & 1/2 & 0 & 0 & -1/2 \\ -20/52 & 7/52 & 17/52 & 1/13 & 1/52 \\ 1/4 & -1/4 & -1/4 & 0 & 1/4 \end{bmatrix} * \begin{bmatrix} 6 \\ 4 \\ 7 \\ 10 \\ 10 \end{bmatrix} \\ &= \begin{bmatrix} 3 \\ 23/52 \\ 5/4 \end{bmatrix} \end{aligned}$$

$$1.3 \quad \text{Analytical } \theta^* = \begin{bmatrix} 3 \\ 23/52 \\ 5/4 \end{bmatrix}.$$

Solve for $x = [1, 5, 3]^T$:

$$\theta^* * x = 233/26 = 8.96;$$

Numerical $\theta^2 = [0.288, 0.6536, 0.6776]^T$.

Solve for $x = [1, 5, 3]^T$:

$$\theta^2 * x = 5.5888;$$

- 1.4 I think the Analytical solution here works better, because for the Numerical solution we only have the result of two (2) iterations. Also, the dimensions are small so Analytical solution is better suited for this.