

①

$$\begin{aligned}
 \text{a) } \frac{df}{dx_1} &= 2x_1 + 2 = 0 \Rightarrow x_1 = -1 \\
 \frac{df}{dx_2} &= 6x_2 + 0.5 \Rightarrow x_2 = -\frac{1}{12} \\
 &\Rightarrow x^* = \left(-1, -\frac{1}{12}\right) \Rightarrow \\
 &P^* = (-1)^2 + 3\left(-\frac{1}{12}\right)^2 + 2(-1) + 0.5\left(-\frac{1}{12}\right) \\
 &= -1.02 \\
 \frac{d^2f}{dx_1^2} &= 2 > 0 \\
 \frac{d^2f}{dx_2^2} &= 6 > 0 \quad \left\{ \begin{array}{l} \rightarrow \text{it's a minimum} \end{array} \right.
 \end{aligned}$$

$$\text{b) } \nabla f = \begin{pmatrix} 2x_1 + 2 \\ 6x_2 + 0.5 \end{pmatrix}$$

$$x_0 = (3, -1), f_0 = 17.5$$

$$\left(\nabla f = \begin{pmatrix} 8 \\ -5.5 \end{pmatrix} \right) \Rightarrow x_1 = x_0 - \alpha \nabla f = (3, -1) - (0.2)(8, 5.5) = (1.4, 0.1)$$

$$x_1 = (1.4, 0.1), f_1 = 4.48$$

$$\left(\nabla f = \begin{pmatrix} 4.8 \\ 1.1 \end{pmatrix} \right) \Rightarrow x_2 = x_1 - \alpha \nabla f = (1.4, 0.1) - (0.2)(4.8, 1.1) = (0.44, -0.12)$$

$$x_2 = (0.44, -0.12), f_2 = 1.05$$

$$\left(\nabla f = \begin{pmatrix} 2.88 \\ -0.22 \end{pmatrix} \right) \Rightarrow x_3 = x_2 - \alpha \nabla f = (0.44, -0.12) - (0.2)(2.88, -0.22) = (-0.136, -0.076)$$

$$x_3 = (-0.136, -0.076), f_3 = -0.27$$

yes as the values clearly demonstrate, the algorithm is minimizing it.

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$$\textcircled{C} \nabla f = \begin{pmatrix} 2u_1 + 2 \\ 6u_2 + 0.5 \end{pmatrix}$$

$$x_0 = (3, -1), f_0 = 17.5$$

$$\nabla f = (8, -5.5) \Rightarrow x_1 = x_0 - u \nabla f = (-1, 1.75)$$

$$x_1 = (-1, 1.75), f_1 = 9.06$$

$$\nabla f = (0, 11) \Rightarrow x_2 = x_1 - u \nabla f = (-1, -3.75)$$

$$x_2 = (-1, -3.75), f_2 = 39.31$$

$$\nabla f = (0, -22) \Rightarrow x_3 = x_2 - u \nabla f = (-1, 7.25)$$

$$x_3 = (-1, 7.25), f_3 = 160.3$$

\Rightarrow It's not minimizing. The u (step size) is not suitable, (too large).

$$\textcircled{d} \text{ The function is: } L_{\theta} = \frac{1}{N} \|y - x^T \theta\|_2^2$$

$$= \frac{1}{N} \sum (y - x^T \theta)^2$$

$$\nabla L = \frac{dL}{d\theta} = \frac{2}{N} (y - x^T \theta) (-x^T) = 0$$

$$\Rightarrow (y - x^T \theta) (-x^T) = 0 \Rightarrow -x^T y + x^T x \theta = 0$$

$$\Rightarrow \theta = (x^T x)^{-1} (x^T y)$$

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$$a) \nabla f = \begin{pmatrix} 2u_1 \\ 2u_2 \end{pmatrix}$$

$$\Delta u = -\nabla f$$

$$f(x + u \Delta x) > f(x) + \alpha u \nabla f(x)^T \Delta x \xRightarrow{\Delta x = -\nabla f} f(x - u \nabla f) > f(x) - \alpha u \|\nabla f\|_2^2$$

$$\Rightarrow f((u_1, u_2) - u(2u_1, 2u_2)) > f(u_1, u_2) - \alpha u (4u_1^2 + 4u_2^2)$$

$$\Rightarrow (1-2u)^2 u_1^2 + (1-2u)^2 u_2^2 > (1-4\alpha u) u_1^2 + (1-4\alpha u) u_2^2$$

$$\Rightarrow (1-2u)^2 (u_1^2 + u_2^2) > (1-4\alpha u) (u_1^2 + u_2^2)$$

$$\Rightarrow 1 - 4u + 4u^2 > 1 - 4\alpha u \Rightarrow 4u(u - 1 + \alpha) > 0$$

$$\Rightarrow u - 1 + \alpha > 0$$

b) $u = (0.5, 1) \Rightarrow \Delta u = -\nabla f = (1, 2)$

$$\mu = 10, \alpha = 0.5, \beta = 0.1$$

while ($\mu - 0.5 > 0$) \rightarrow derived from previous part

$$\mu = \mu(0.1)$$

iter 0 : $\mu = 10$

iter 1 : 1 \rightarrow at 2 iterations the algorithm stops

iter 2 : $\mu = 0.1$, and the final μ is 0.1

iter 3 (X)

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