

6.046 Problem Set 8

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Problem 1

(A) We want to find:

$$\begin{aligned}
\frac{\partial f_{A,b}}{\partial x_1} &= \frac{\partial}{\partial x_1} \frac{1}{2} x^T (A^T A) x - \frac{\partial}{\partial x_1} b^T A x && \text{Apply product rule} \\
\frac{\partial f_{A,b}}{\partial x_1} &= \frac{1}{2} \left(\frac{\partial}{\partial x_1} x^T \right) (A^T A) x + \frac{1}{2} x^T (A^T A) \left(\frac{\partial}{\partial x_1} x \right) - b^T A \left(\frac{\partial}{\partial x_1} x \right) && \text{Eliminate the empty terms} \\
\frac{\partial f_{A,b}}{\partial x_1} &= \frac{1}{2} \begin{pmatrix} 1 & 0 \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & 8 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & 8 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} - \begin{pmatrix} 1 & 3 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ -1 & 2 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \\
\frac{\partial f_{A,b}}{\partial x_1} &= x_1 + x_1 + 2 \\
\frac{\partial f_{A,b}}{\partial x_1} &= 2x_1 + 2
\end{aligned}$$

And then, for the second one:

$$\begin{aligned}
\frac{\partial f_{A,b}}{\partial x_2} &= \frac{\partial}{\partial x_2} \frac{1}{2} x^T (A^T A) x - \frac{\partial}{\partial x_2} b^T A x && \text{Apply product rule} \\
\frac{\partial f_{A,b}}{\partial x_2} &= \frac{1}{2} \left(\frac{\partial}{\partial x_2} x^T \right) (A^T A) x + \frac{1}{2} x^T (A^T A) \left(\frac{\partial}{\partial x_2} x \right) - b^T A \left(\frac{\partial}{\partial x_2} x \right) && \text{Eliminate the empty terms} \\
\frac{\partial f_{A,b}}{\partial x_2} &= \frac{1}{2} \begin{pmatrix} 0 & 1 \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & 8 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & 8 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} - \begin{pmatrix} 1 & 3 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ -1 & 2 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \\
\frac{\partial f_{A,b}}{\partial x_2} &= 4x_2 + 4x_2 - 8 \\
\frac{\partial f_{A,b}}{\partial x_2} &= 8x_2 - 8
\end{aligned}$$

(B) To be a critical point, we have to show that both the derivatives are equal to 0. If we set $2x_1 + 2 = 0$, we see that $x_1 = -1$ is the only solution, and if we set $8x_2 - 8 = 0$, we see that $x_2 = 1$ is the only solution. So, we see that any critical point must have $x_2 = 1$ and $x_1 = -1$.

(C) The first step here is to find the Hessian. The Hessian here can be given by $\begin{bmatrix} \frac{\partial f_{A,b}}{\partial x_1 x_1} & \frac{\partial f_{A,b}}{\partial x_1 x_2} \\ \frac{\partial f_{A,b}}{\partial x_2 x_1} & \frac{\partial f_{A,b}}{\partial x_2 x_2} \end{bmatrix}$

which is numerically $\begin{pmatrix} 2 & 0 \\ 0 & 8 \end{pmatrix}$. Thus we need to find the largest α such that $x^T \begin{pmatrix} 2 & 0 \\ 0 & 8 \end{pmatrix} x \geq \alpha$ is true for all x . We are told that x is a unit norm matrix. It is fairly easy to see that since the hessian is diagonal with one value larger than the other, we can simply take the unit norm as a vector with just a single one to try to minimize this quantity. Using $x = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$, we get that $f_{A,b}$ is α strongly convex where $\alpha = 2$. By a similar argument, we can see that it is β smooth where $\beta = 8$.

(D) We know the formula for convergence given α strong convexity and β smoothness is $O(\frac{\beta}{\alpha} \log \frac{f(x_0) - f(x^*)}{\epsilon})$. In this case, since α, β are constants, and we know the optimal solution from linear algebra $x^* = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$, and thus $f_{A,b}(x^*) = -5$, we get that this gradient descent will converge in $O(\log \frac{f(x_0) + 5}{\epsilon})$.

Problem 2

Sorry I been swamped recently and struggling w some issues :(