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Math 408 - Homework 9

4.1.3 $f(x_1, x_2) = x_1^2 + x_2^2 + x_1 x_2 - x_1$

$\frac{\partial f}{\partial x_1} = 2x_1 + x_2 - 1$

$\frac{\partial f}{\partial x_2} = 2x_2 + x_1$

$x^{(0)} = (0, 0)$

 t_k -line minimization

$\nabla f = \begin{pmatrix} 2x_1 + x_2 - 1 \\ 2x_2 + x_1 \end{pmatrix}$

$Hf = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix} \succ 0$

$\phi_0(t) = f\left(\begin{pmatrix} 0 \\ 0 \end{pmatrix} + t d^{(0)}\right) = f(t d^{(0)})$

a) $d^{(0)} = -\nabla f(x^{(0)}) = -\begin{pmatrix} -1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

$\phi_0(t) = f\left(\begin{pmatrix} t \\ 0 \end{pmatrix}\right) = t^2 - t \quad \phi_0'(t) = 2t - 1 = 0 \Rightarrow t = 1/2$

$\phi_0''(t) = 2 > 0 \Rightarrow t = 1/2$ is the minimizer of $\phi_0(t)$

$x^{(1)} = x^{(0)} + t d^{(0)} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1/2 \\ 0 \end{pmatrix} //$

b) $p^{(0)} = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \quad p^{(0)^{-1}} = \begin{bmatrix} 1/2 & 0 \\ 0 & 1/2 \end{bmatrix}$

$d^{(0)} = -\begin{bmatrix} 1/2 & 0 \\ 0 & 1/2 \end{bmatrix} \begin{pmatrix} -1 \\ 0 \end{pmatrix} = \begin{pmatrix} 1/2 \\ 0 \end{pmatrix}$

$x^{(1)} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 1/2 \\ 0 \end{pmatrix} = \begin{pmatrix} 1/4 \\ 0 \end{pmatrix} //$

c) eigenvalues of $Hf = -3, -1 \quad \gamma_0 = \max\{0, 1+3\} = 4$

$p^{(0)} = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} + 4 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 6 & 1 \\ 1 & 6 \end{bmatrix}$

$d^{(0)} = -\frac{1}{35} \begin{bmatrix} 6 & -1 \\ -1 & 6 \end{bmatrix} \begin{pmatrix} -1 \\ 0 \end{pmatrix} = \frac{1}{35} \begin{pmatrix} 6 \\ -1 \end{pmatrix} = \begin{pmatrix} 6/35 \\ -1/35 \end{pmatrix}$

$x^{(1)} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} + \frac{1}{2} \begin{pmatrix} 6/35 \\ -1/35 \end{pmatrix} = \begin{pmatrix} 3/35 \\ -1/70 \end{pmatrix} //$

4.1.4 If $d^{(k)} = -\nabla f(x^{(k)})$ then $-\nabla f(x^{(k)})^T d^{(k)} = -\|\nabla f(x^{(k)})\|^2$

 $\Rightarrow$ we can take $\beta = 1$ in Thm 4.1.1 (a)

$d^{(k)} = -\nabla f(x^{(k)}) = -\begin{pmatrix} 2x_1^{(k)} + x_2^{(k)} - 1 \\ 2x_2^{(k)} + x_1^{(k)} \end{pmatrix} \quad \text{Suppose } x^{(k)} = \begin{pmatrix} x_1^{(k)} \\ x_2^{(k)} \end{pmatrix}$

converges to x^*

$\|d^{(k)}\| = \sqrt{(2x_1^{(k)} + x_2^{(k)} - 1)^2 + (2x_2^{(k)} + x_1^{(k)})^2}$

is bounded for every converging subsequence of $x^{(k)}$

(2)

$\therefore$ By Thm 4.1.1, $f(x^{(k+1)}) \leq f(x^{(k)}) \quad \forall k \in \{0, 1, 2, \dots\}$ with $\nabla f(x^{(k)}) \neq 0$.

Now we have to see which of (i) – (iii) is possible:

– The function f is coercive since $Hf > 0$ (Ex 3.3.5 on p 35)

$\therefore f \rightarrow \infty$ as $\|x\| \rightarrow \infty$.

– The function f is also convex since $Hf > 0$.

– If $\nabla f = 0 \Rightarrow \begin{pmatrix} 2x_1 + x_2 - 1 \\ 2x_2 + x_1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow \begin{aligned} x_1 &= -2x_2 \\ -4x_2 + x_2 - 1 &= 0 \Rightarrow -3x_2 = 1 \end{aligned}$

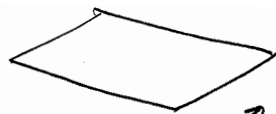
$\Rightarrow (x_1, x_2) = \begin{pmatrix} 2/3 \\ -1/3 \end{pmatrix} \quad \therefore f$ has only one critical pt with $Hf > 0$

$\Rightarrow \begin{pmatrix} 2/3 \\ -1/3 \end{pmatrix}$ is the global minimizer of f with $f\left(\begin{pmatrix} 2/3 \\ -1/3 \end{pmatrix}\right) = -1/3$

$\therefore$ (ii) is not possible ($f(x^{(k)}) \not\rightarrow -\infty$ as $k \rightarrow \infty$)

$\exists$ only one critical point $\therefore$ (i) or (iii) is possible.

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graph is a plane in $\mathbb{R}^3$ - no critical points.

4.1.6 $f(x_1, x_2) = x_1 + x_2$ $x^{(0)} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ $\nabla f = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ $d^{(k)} = \begin{pmatrix} -1 \\ -1 \end{pmatrix}$

In Thm 4.1.1 (a) is true by choosing $\beta = 1$ and (b) also true since $d^{(k)}$ is a constant vector

Choose t_k by Armijo's rule and get $x^{(k+1)} = x^{(k)} + t_k d^{(k)}$

$$= \begin{pmatrix} x_1^{(k)} \\ x_2^{(k)} \end{pmatrix} + t_k \begin{pmatrix} -1 \\ -1 \end{pmatrix} = \begin{pmatrix} x_1^{(k)} - t_k \\ x_2^{(k)} - t_k \end{pmatrix}$$

$$\therefore f(x^{(k+1)}) = x_1^{(k)} + x_2^{(k)} - 2t_k < x_1^{(k)} + x_2^{(k)} = f(x^{(k)})$$

(i) is never true in Thm 4.1.1 since $\nabla f(x^{(k)}) = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ always. To look at (ii) & (iii) we need to understand t_k

Armijo's rule

$$\phi_k(t) = f(x^{(k)} + t d^{(k)}) = f\left(\begin{pmatrix} x_1^{(k)} \\ x_2^{(k)} \end{pmatrix} + t \begin{pmatrix} -1 \\ -1 \end{pmatrix}\right) = f\left(\begin{pmatrix} x_1^{(k)} - t \\ x_2^{(k)} - t \end{pmatrix}\right)$$

$$= x_1^{(k)} + x_2^{(k)} - 2t$$

$$\phi_k(0) = x_1^{(k)} + x_2^{(k)}$$

$$\phi_k'(t) = -2 < 0 \quad \forall t$$

In particular, $\phi_k'(0) = -2$

How to choose t_k ? $t_k = \text{largest } t \in \{1, \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \dots\}$

$$\text{s.t. } \frac{\phi_k(t) - \phi_k(0)}{t} \leq \cdot 1 \phi_k'(0) = \cdot 1 (-2) = -\cdot 2$$

$$\Rightarrow \frac{x_1^{(k)} + x_2^{(k)} - 2t - (x_1^{(k)} + x_2^{(k)})}{t} = -2 < -\cdot 2 \quad \text{always.}$$

$\therefore t_k = 1$ always.

$\therefore$ If you look at the sequence $\{x^{(k)}\}$

$$x^{(k+1)} = x^{(k)} + 1 \begin{pmatrix} -1 \\ -1 \end{pmatrix} = x^{(k)} - \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$\therefore \{x^{(k)}\}$ does not converge & no subsequence of it converges too. $\therefore$ (iii) does not need to be considered

As $k \rightarrow \infty$ $f(x^{(k)}) \rightarrow -\infty$ $\therefore$ (ii) is true.

The sequence $\{x^{(k)}\}$ moves far like

$$\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ -1 \end{pmatrix}, \begin{pmatrix} -2 \\ -2 \end{pmatrix}, \begin{pmatrix} -3 \\ -3 \end{pmatrix}, \dots$$

(4)

4.1.7 $f(x) = \frac{1}{2}x^T A x + b^T x$ A $n \times n$ symm matrix, $b \in \mathbb{R}^n$

$$H_f = A$$

(a) It is true that f is convex $\Leftrightarrow A \succeq 0$.

(b) Suppose $A \succ 0 \Rightarrow f$ convex. Then by Thm 3.5.2 if a $x^* \in \mathbb{R}^n$ satisfies $\nabla f(x^*) = 0$ then x^* is a global opt of f . (remember A is symm.)

$$f(x) = \frac{1}{2} \sum_{i,j=1}^n a_{ij} x_i x_j + \sum b_i x_i$$

$$\frac{\partial f}{\partial x_i} = \frac{1}{2} \sum_{j=1}^n 2a_{ij} x_j + b_i$$

$$\nabla f(x) = \begin{pmatrix} \sum a_{1j} x_j + b_1 \\ \sum a_{2j} x_j + b_2 \\ \vdots \\ \sum a_{nj} x_j + b_n \end{pmatrix}$$

$$\nabla f = 0 \Rightarrow \sum_{j=1}^n a_{ij} x_j = -b_i \Rightarrow \sum a_{ij} x_j = -b_i \quad \forall i$$

$$\Rightarrow Ax = -b \Rightarrow x = A^{-1}(-b) = -A^{-1}b$$

(A^{-1} exists since $A \succ 0$) & x unique

$\therefore \exists$ only one global min and it is $x = -A^{-1}b //$

(c) $A \succ 0$, $x^{(0)} = 0$ $d^{(k)} = -H_f(x^{(k)})^{-1} \nabla f(x^{(k)}) = -A^{-1} \nabla f(x^{(k)})$

$$\nabla f(x^{(k)}) = \frac{1}{2} A x^{(k)} + b$$

$$\therefore d^{(k)} = -A^{-1} \left(\frac{1}{2} A x^{(k)} + b \right) = -\frac{1}{2} x^{(k)} - A^{-1}b = -\frac{1}{2} x^{(k)} - A^{-1}b$$

$$\phi_k(t) = f(x^{(k)} + t d^{(k)}) \Rightarrow \phi_0(t) = f(t d^{(0)})$$

$$= f(t(-A^{-1}b)) = f(-t A^{-1}b) = \frac{1}{2} (-t A^{-1}b)^T A (-t A^{-1}b) + b(-t A^{-1}b)$$

$$= \frac{t^2}{2} \underbrace{(b^T (A^{-1})^T A A^{-1} b)}_I - t b^T A^{-1} b \quad \left(\begin{array}{l} \text{recall } A \succ 0 \Rightarrow A^{-1} \succ 0 \\ \Rightarrow A^{-1} \text{ symm} \end{array} \right)$$

$$= \frac{t^2}{2} (b^T A^{-1} b) - t b^T A^{-1} b = \underbrace{b^T A^{-1} b}_{>0} \left(\frac{t^2}{2} - t \right)$$

$$\phi'_k(t) = \frac{2t}{2} - 1 = 0 \Rightarrow t - 1 = 0 \Rightarrow t = 1. //$$

$$\therefore x^{(1)} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} + 1 \left(-\frac{1}{2} \begin{pmatrix} 0 \\ 0 \end{pmatrix} - A^{-1}b \right) = -A^{-1}b //$$