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CONVEX OPTIMIZATION FOR ML & CV

Weekly Exercise 8 Exact Line Search

Exercise 1

The EVD of \mathcal{Q} is given as

$$\mathcal{Q} = U \Lambda U^T \Rightarrow \mathcal{Q}^{-1} = U \Lambda^{-1} U^T$$

where

a) U is orthogonal, $\therefore U^{-1} = U^T$ [$\because \mathcal{Q}$ symmetric]

b) $\Lambda = \text{diag}(\lambda_1, \dots, \lambda_n)$ with eigenvalues $\lambda_i > 0$.
[$\because \mathcal{Q}$ positive definite]

Now, substituting the value of \mathcal{Q} , we get

$$\frac{(x^T x)^2}{(x^T \mathcal{Q} x)(x^T \mathcal{Q}^{-1} x)} = \frac{\|x\|^4}{\langle x, U \Lambda U^T x \rangle \langle x, U \Lambda^{-1} U^T x \rangle}$$

$$= \frac{\|U^T x\|^4}{\langle U^T x, \Lambda U^T x \rangle \langle U^T x, \Lambda^{-1} U^T x \rangle}$$

[$\because \|x\| = \|Ux\|$
for orthogonal
 U .]

$$\Rightarrow \frac{(x^T x)^2}{(x^T Q x)(x^T Q^{-1} x)} = \frac{\|y\|^4}{\langle y, \Lambda y \rangle \langle y, \Lambda^{-1} y \rangle}$$

[substituting $y = U^T x$]

$$\Rightarrow = \frac{\|y\|^4}{\left(\sum_{i=1}^n \lambda_i y_i^2\right) \left(\sum_{i=1}^n \left(\frac{1}{\lambda_i}\right) y_i^2\right)}$$

$$\Rightarrow = \frac{1}{\left\{ \sum_{i=1}^n \lambda_i \left(\frac{y_i}{\|y\|}\right)^2 \right\} \left\{ \sum_{i=1}^n \left(\frac{1}{\lambda_i}\right) \left(\frac{y_i}{\|y\|}\right)^2 \right\}}$$

$$\Rightarrow = \frac{1}{\left\{ \sum_{i=1}^n \lambda_i z_i \right\} \left\{ \sum_{i=1}^n \left(\frac{1}{\lambda_i}\right) z_i \right\}}$$

[substituting $z_i = \left(\frac{y_i}{\|y\|}\right)^2$]
 $\Rightarrow z_i \geq 0, \sum_i z_i = 1$

$$\Rightarrow = \frac{1}{\gamma \left\{ \sum_{i=1}^n \left(\frac{1}{\lambda_i}\right) z_i \right\}}$$

[substituting $\gamma = \sum_{i=1}^n \lambda_i z_i$,
 $\Rightarrow \gamma \in [\lambda_1, \lambda_n]$]

Now, we can get an upper bound on $\sum_{i=1}^n \left(\frac{1}{\lambda_i}\right) z_i$, as a simple interpolation of $[\lambda_1^{-1}, \lambda_n^{-1}]$ at γ :

$$\sum_{i=1}^n \left(\frac{1}{\lambda_i}\right) z_i \leq \frac{1}{\lambda_n} + \frac{\frac{1}{\lambda_1} - \frac{1}{\lambda_n}}{\lambda_n - \lambda_1} (\lambda_n - \gamma)$$

Therefore,

$$\begin{aligned} \Rightarrow \frac{(x^T x)^2}{(x^T Q x)(x^T Q^{-1} x)} &\geq \frac{\gamma^{-1}}{\frac{1}{\lambda_n} + \frac{\frac{1}{\lambda_1} - \frac{1}{\lambda_n}}{\lambda_n - \lambda_1} (\lambda_n - \gamma)} \\ &\geq \min_{\gamma \in [\lambda_1, \lambda_n]} \frac{\gamma^{-1}}{\frac{1}{\lambda_n} + \frac{\frac{1}{\lambda_1} - \frac{1}{\lambda_n}}{\lambda_n - \lambda_1} (\lambda_n - \gamma)} \end{aligned}$$

simplification yields,

$$\Rightarrow = \lambda_1 \lambda_n \min_{\gamma} \frac{\gamma^{-1}}{(\lambda_1 + \lambda_n - \gamma)}$$

Solving for γ gives $\gamma^* = \frac{\lambda_1 + \lambda_n}{2}$. Substituting back gives the final result:

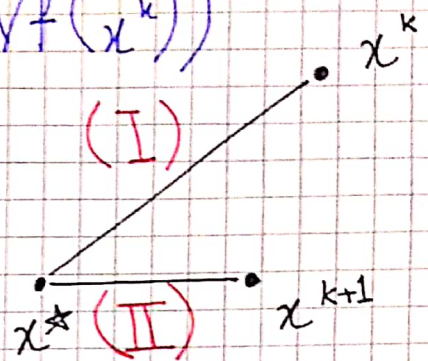
$$\therefore \frac{(x^T x)^2}{(x^T Q x)(x^T Q^{-1} x)} \geq \frac{4 \lambda_n \lambda_1}{(\lambda_n + \lambda_1)^2}$$

Exercise 2

For the line-search algorithm:

$$\tau^k = \operatorname{argmin}_{\tau} f(x^k - \tau \nabla f(x^k))$$

$$(i) \quad \tau^k = \frac{\|\nabla f(x^k)\|^2}{\|\nabla f(x^k)\|_a^2}$$



Also:

$$\nabla f(x^k) = \mathcal{A}x^k - b$$

$$(ii) \quad \Rightarrow \quad = \mathcal{A}(x^k - x^*)$$

$$\left[\begin{array}{l} \because \text{for } x^k = x^* \\ \Rightarrow \nabla f(x^*) = \mathcal{A}x^* - b = 0 \\ \Rightarrow b = \mathcal{A}x^* \end{array} \right]$$

We can now get

$$\therefore \|x^k - x^*\|_a^2 = \langle x^k - x^*, \mathcal{A}(x^k - x^*) \rangle$$

$$= \langle x^k - x^*, \mathcal{A} \mathcal{A}^{-1} \mathcal{A}(x^k - x^*) \rangle$$

[multiplying by $I = \mathcal{A} \mathcal{A}^{-1}$]

$$= \langle \mathcal{A}^T(x^k - x^*), \mathcal{A}^{-1} \mathcal{A}(x^k - x^*) \rangle$$

$$= \langle \mathcal{A}(x^k - x^*), \mathcal{A}^{-1} \mathcal{A}(x^k - x^*) \rangle$$

[$\because \mathcal{A}$ symmetric]

$$= \|\mathcal{A}(x^k - x^*)\|_{\mathcal{A}^{-1}}^2$$

$$(I) \quad \therefore \|x^k - x^*\|_a^2 = \|\nabla f(x^k)\|_{\mathcal{A}^{-1}}^2 \quad \text{--- (iii) [from (ii)]}$$

Now,

$$\therefore \|x^k - x^*\|_a^2 - \|x^{k+1} - x^*\|_a^2 = \left\{ \|x^k - x^*\|_a^2 - \|x^k - \tau_k \nabla f(x^k) - x^*\|_a^2 \right\}$$

\Rightarrow

$$= \left\{ \left(\|x^k\|_a^2 - 2 \langle x^k, x^* \rangle_a + \|x^*\|_a^2 \right) - \left(\|x^k - \tau_k \nabla f(x^k)\|_a^2 + \|x^*\|_a^2 - 2 \langle x^k - \tau_k \nabla f(x^k), x^* \rangle_a \right) \right\}$$

\Rightarrow

$$= \|x^k\|_a^2 - \|x^k - \tau_k \nabla f(x^k)\|_a^2 - 2 \langle \nabla f(x^k), x^* \rangle_a$$

\Rightarrow

$$= \left\{ 2 \tau_k \langle x^k, \nabla f(x^k) \rangle_a - 2 \tau_k \langle \nabla f(x^k), x^* \rangle_a - \tau_k^2 \|\nabla f(x^k)\|_a^2 \right\}$$

\Rightarrow

$$= 2 \tau_k \langle \nabla f(x^k), x^k - x^* \rangle_a - \tau_k^2 \|\nabla f(x^k)\|_a^2$$

\Rightarrow

$$= \frac{2 \|\nabla f(x^k)\|_a^4}{\|\nabla f(x^k)\|_a^2} - \frac{\|\nabla f(x^k)\|_a^4}{\|\nabla f(x^k)\|_a^2} \quad \left[\text{from (i)} \right]$$

\Rightarrow

$$= \frac{\|\nabla f(x^k)\|_a^4}{\|\nabla f(x^k)\|_a^2} \quad \text{--- (iv)}$$

(5)

Finally,

$$\begin{aligned} \therefore \|x^{k+1} - x^*\|_Q^2 &= \|x^k - x^*\|_Q^2 - \left(\|x^k - x^*\|_Q^2 - \|x^{k+1} - x^*\|_Q^2 \right) \\ \Rightarrow &= \|x^k - x^*\|_Q^2 - \frac{\|\nabla f(x^k)\|_Q^4}{\|\nabla f(x^k)\|_Q^2} \quad [\text{from (iv)}] \end{aligned}$$

$$\Rightarrow = \|\nabla f(x^k)\|_{Q^{-1}}^2 - \frac{\|\nabla f(x^k)\|_Q^4}{\|\nabla f(x^k)\|_Q^2} \quad [\text{from (iii)}]$$

$$\Rightarrow = \|\nabla f(x^k)\|_{Q^{-1}}^2 \left(1 - \frac{\|\nabla f(x^k)\|_Q^4}{\|\nabla f(x^k)\|_Q^2 \cdot \|\nabla f(x^k)\|_Q^2} \right)$$

$$\Rightarrow = \|\nabla f(x^k)\|_{Q^{-1}}^2 \left(1 - \frac{(z^T z)^2}{(z^T Q z)(z^T Q^{-1} z)} \right)$$

[substituting $z = \nabla f(x^k)$]

$$\Rightarrow \leq \|\nabla f(x^k)\|_{Q^{-1}}^2 \left(1 - \frac{4\lambda_1 \lambda_n}{(\lambda_1 + \lambda_n)^2} \right)$$

[exercise 1]

$$\Rightarrow = \left(\frac{\lambda_1 - \lambda_n}{\lambda_1 + \lambda_n} \right)^2 \|\nabla f(x^k)\|_{Q^{-1}}^2$$

$$(II) \therefore \|x^{k+1} - x^*\|_Q^2 \leq \left(\frac{\lambda_1 - \lambda_n}{\lambda_1 + \lambda_n} \right)^2 \|x^k - x^*\|_Q^2 \quad [\text{from (iii)}]$$