

P1 $\min_{y \in \mathbb{R}^n} \frac{1}{2} \|x - y\|_2^2$ subject to $Ay \leq b$

(a) Lagrangian $L(y, \lambda) = \frac{1}{2} \|x - y\|_2^2 + \lambda^T (Ay - b)$

Dual function $g(\lambda) = \inf_{y \in \mathbb{R}^n} L(y, \lambda)$
 $= \inf_{y \in \mathbb{R}^n} \left(\frac{1}{2} \|x - y\|_2^2 + \lambda^T (Ay - b) \right) \rightarrow \text{convex over } y$
 $= \frac{1}{2} \|x - (x - A^T \lambda)\|_2^2 + \lambda^T (A(x - A^T \lambda) - b)$
 $= \frac{1}{2} \lambda^T A A^T \lambda + \lambda^T A x - \lambda^T A A^T \lambda - \lambda^T b$
 $= -\frac{1}{2} \lambda^T A A^T \lambda + (Ax - b)^T \lambda$

(b) As the primal problem is feasible by assumption, the strong duality holds.

And, the problem is convex, thus KKT condition is necessary & sufficient condition for optimality. Let y^*, λ^* be the optimal points for primal, dual problem, resp.

KKT conditions: $\begin{cases} Ay^* - b \leq 0 & \text{(PF)} \\ \lambda^* \geq 0 & \text{(DF)} \\ (\lambda^*)^T (Ay^* - b) = 0 & \text{(CS)} \\ \nabla_y L(y, \lambda^*)|_{y=y^*} = 0 & \text{(LS)} \end{cases}$

(LS) $\rightarrow -x + y^* + A^T \lambda^* = 0 \Rightarrow y^* = x - A^T \lambda^* = x - A^T (A A^T)^+ (Ax - b)$

$d^* = \max_{\lambda \geq 0} g(\lambda) = \min_{\lambda \geq 0} -g(\lambda)$

$-\nabla_{\lambda} g(\lambda) = A A^T \lambda + b - Ax = 0 \Rightarrow \lambda^* = (A A^T)^+ (Ax - b)$

P2 - (c).

$-g(\lambda) = \frac{1}{2} \lambda^T A A^T \lambda + (b - Ax)^T \lambda$

$-\nabla_{\lambda} g(\lambda) = A A^T \lambda + b - Ax$

Let $\lambda_1, \lambda_2 \geq 0$. $\|(-\nabla_{\lambda} g(\lambda_1)) - (-\nabla_{\lambda} g(\lambda_2))\|_2$
 $= \|(A A^T \lambda_1 + b - Ax) - (A A^T \lambda_2 + b - Ax)\|_2$
 $= \|A A^T (\lambda_1 - \lambda_2)\|_2 \leq \|A A^T\|_F \|\lambda_1 - \lambda_2\|_2$

Hence $-g(\lambda)$ has a Lipschitz continuous gradient with the Lipschitz constant $L = \|A A^T\|_F$. Thus I used $\sigma_k = 1/L = \|A A^T\|_F^{-1}$ for the step size.

P3 - (c)

$f_0(x) = \frac{1}{2} x^T H x + c^T x$. $\nabla f_0(x) = Hx + c$

$\|\nabla f_0(x) - \nabla f_0(y)\| = \|H(x - y)\|_2 \leq \|H\|_F \|x - y\|_2 \quad \forall x, y$

Hence f_0 has a Lipschitz continuous gradient with the Lipschitz constant $L = \|H\|_F$. Thus I used $\sigma_k = 1/L = \|H\|_F^{-1}$ for the step size.