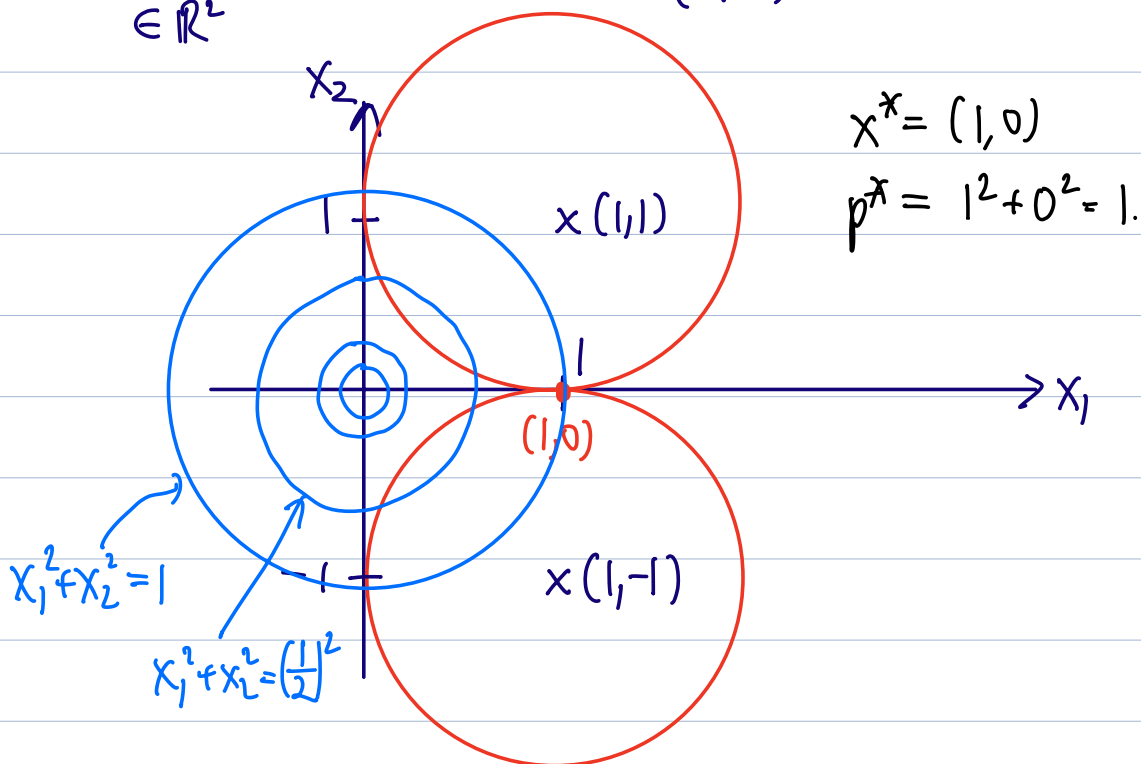


## DSA3102 Tutorial 10

### 1. BV 5.26

$$\min_{\substack{x=(x_1, x_2) \\ \in \mathbb{R}^2}} x_1^2 + x_2^2 \quad \text{s.t.}$$

$$\begin{aligned} (x_1-1)^2 + (x_2-1)^2 &\leq 1 \\ (x_1-1)^2 + (x_2+1)^2 &\leq 1 \end{aligned}$$



### b) Stationarity:

$$\begin{aligned} L(x, \lambda_1, \lambda_2) = & x_1^2 + x_2^2 + \lambda_1 [(x_1-1)^2 + (x_2-1)^2 - 1] \\ & + \lambda_2 [(x_1-1)^2 + (x_2+1)^2 - 1] \end{aligned}$$

$$\frac{\partial L}{\partial x_1} = 0 \Rightarrow 2x_1 + 2\lambda_1(x_1-1) + 2\lambda_2(x_1-1) = 0$$

$$\frac{\partial L}{\partial x_2} = 0 \Rightarrow 2x_2 + 2\lambda_1(x_2-1) + 2\lambda_2(x_2+1) = 0$$

Primal Feasibility:  $(x_1 - 1)^2 + (x_2 - 1)^2 \leq 1$   
 $(x_1 - 1)^2 + (x_2 + 1)^2 \leq 1.$

Dual Feasibility:  $\lambda_1 \geq 0, \lambda_2 \geq 0.$

Comp. Slackness:  $\lambda_1 [(x_1 - 1)^2 + (x_2 - 1)^2 - 1] = 0$   
 $\lambda_2 [(x_1 - 1)^2 + (x_2 + 1)^2 - 1] = 0$

Only primal feasible  $(x_1, x_2) = (1, 0)$ .

$\lambda_1 \geq 0, \lambda_2 \geq 0$ ,  $2 - 2\lambda_2 = 0$ ,  $-2\lambda_1 + 2\lambda_2 = 0$   
 $\lambda_1 = \lambda_2$

$\lambda_1 > 0, \lambda_2 = \lambda_1 > 0$   
 $\lambda_1 = \lambda_2 = 0$   $2 = 0.$

The KKT conditions do not give a solution for  $x_1, x_2, \lambda_1, \lambda_2$ .

c)  $g(\lambda_1, \lambda_2) = \inf_{x_1, x_2} L(x_1, x_2, \lambda_1, \lambda_2)$

$= \inf_{x_1, x_2} \left\{ x_1^2 + x_2^2 + \lambda_1 [(x_1 - 1)^2 + (x_2 - 1)^2 - 1] \right.$   
 $\left. + \lambda_2 [(x_1 - 1)^2 + (x_2 + 1)^2 - 1] \right\}$

$$= \inf_{x_1, x_2} (1 + \lambda_1 + \lambda_2) x_1^2 + (1 + \lambda_1 + \lambda_2) x_2^2 - 2(\lambda_1 + \lambda_2) x_1 - 2(\lambda_1 - \lambda_2) x_2 + \lambda_1 + \lambda_2$$

$L$  reaches its minimum when

$$x_1 = \frac{\lambda_1 + \lambda_2}{1 + \lambda_1 + \lambda_2}, \quad x_2 = \frac{\lambda_1 - \lambda_2}{1 + \lambda_1 + \lambda_2}.$$

$$g(\lambda_1, \lambda_2) = \begin{cases} -\frac{(\lambda_1 + \lambda_2)^2 - (\lambda_1 - \lambda_2)^2}{1 + \lambda_1 + \lambda_2} + \lambda_1 + \lambda_2 & 1 + \lambda_1 + \lambda_2 \geq 0 \\ -\infty & \text{else.} \end{cases}$$

Lagrange dual problem:

$$\max_{\lambda_1, \lambda_2 \geq 0} -\frac{(\lambda_1 + \lambda_2)^2 - (\lambda_1 - \lambda_2)^2}{1 + \lambda_1 + \lambda_2} + \lambda_1 + \lambda_2 \quad \text{s.t. } 1 + \lambda_1 + \lambda_2 \geq 0$$

$$\max_{\lambda_1, \lambda_2 \geq 0} \frac{\lambda_1 + \lambda_2 - (\lambda_1 - \lambda_2)^2}{1 + \lambda_1 + \lambda_2}$$

Since  $g(\lambda_1, \lambda_2)$  is symmetric ( $g(\lambda_1, \lambda_2) = g(\lambda_2, \lambda_1)$ ) the optimum, if it exists, must be  $\lambda_1 = \lambda_2$ .

$$\max_{\lambda_1 \geq 0} g(\lambda_1, \lambda_2) = \frac{2\lambda_1}{1 + 2\lambda_1}$$

$d^* \neq 1$  achieved when  $\lambda_1 \rightarrow +\infty$ .

Strong duality holds!  $p^* = d^* = 1$ .

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3.  $\min_{x \in \mathbb{R}^2} f(x) \quad \text{s.t.} \quad \underline{5x_1 + x_2 \geq 4}$   
explicit constraint

$$f(x) = \begin{cases} 10x_1 + 3x_2 & , \quad \underline{(x_1, x_2) \in \{0, 1\}^2} \\ +\infty & \text{else. implicit constraint} \end{cases}$$

b)  $L(x, \lambda) = 10x_1 + 3x_2 + \lambda(4 - 5x_1 - x_2)$

Dual function:

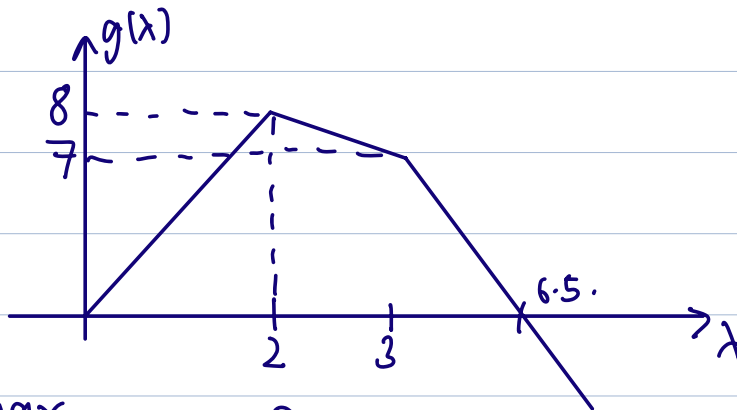
$$g(\lambda) = \inf_{(x_1, x_2) \in \{0, 1\}^2} \{10x_1 + 3x_2 + \lambda(4 - 5x_1 - x_2)\}$$

$$= \inf_{(x_1, x_2) \in \{0, 1\}^2} \{4\lambda + (10 - 5\lambda)x_1 + (3 - \lambda)x_2\}$$

↑ implicit constraint

$$g(\lambda) = \begin{cases} 4\lambda & \lambda \in [0, 2] \\ 10 - \lambda & \lambda \in [2, 3] \\ 13 - 2\lambda & \lambda \in [3, \infty) \end{cases}$$

c) Primal:  $10x_1 + 3x_2 \quad (x_1, x_2) \in \{0, 1\}^2$   
 $(1, 0) \quad p^* = 10, \quad 5x_1 + x_2 \geq 4$



$$d^* = \max_{\lambda \geq 0} g(\lambda) = 8$$

$$\text{Duality gap} = p^* - d^* = 10 - 8 = 2.$$

## 2. BV 5.27

$$\min_x \|Ax - b\|_2^2 \quad \text{s.t.} \quad Gx = h. \quad \begin{matrix} \swarrow p \geq 1 \text{ ineq.} \\ G \in \mathbb{R}^{p \times n} \end{matrix}$$

a) Lagrangian:  $L(x, v) = \|Ax - b\|_2^2 + v^T(Gx - h)$

$$g(v) = \inf_{x \in \mathbb{R}^n} \left\{ \|Ax - b\|_2^2 + v^T(Gx - h) \right\}$$

$$\parallel$$

$$x^T A^T A x - 2x^T A^T b$$

$$\nabla_x L(x, v) = 2A^T A x - 2A^T b + G^T v = 0.$$

$$x^* = \frac{1}{2} (A^T A)^{-1} (2A^T b - G^T v^*)$$

$$g(v) = -\frac{1}{4} (G^T v - 2A^T b)^T (A^T A)^{-1} (G^T v - 2A^T b) - v^T h + b^T b.$$

b) KKT conditions:

Stationarity:  $2A^T A x^* - 2A^T b + G^T v^* = 0.$

$$\underbrace{2A^T(Ax^* - b)}_h + \underbrace{G^T v^*}_P = 0 \quad \downarrow \text{h.} \quad \text{--- (S)}$$

Primal Feasibility:  $Gx^* = h. \quad \uparrow \text{p.} \quad \text{--- (PF)}$

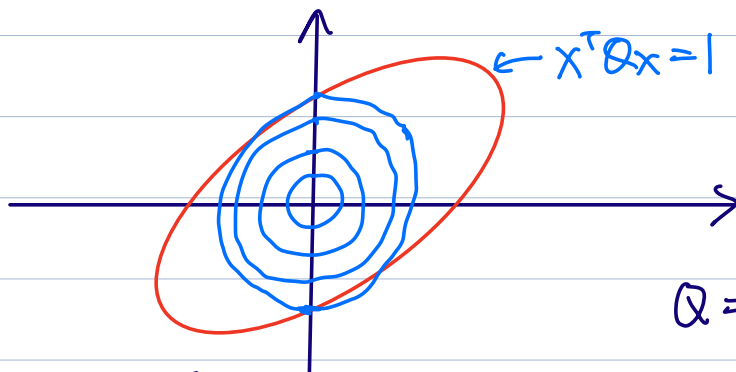
(S)  $x^* = (A^T A)^{-1} (A^T b - \frac{1}{2} G^T v^*).$

Plugging this in (PF) yields

$$G(A^T A)^{-1} (A^T b - \frac{1}{2} G^T v^*) = h$$

$$v^* = -2(G(A^T A)^{-1} G^T)^{-1} (h - G(A^T A)^{-1} A^T b).$$

5(c)  $\min_x f(x) \quad \text{s.t.} \quad h(x) = 0 \quad x^T Q x = 1$   
 $f(x) = \underline{\|x\|^2} \quad \text{s.t.} \quad x^T Q x - 1 = h(x). \quad Q \in S_{++}^n$



$$Q = \sum_{i=1}^n \lambda_i q_i q_i^T$$

$$L(x, v) = x^T x - v(x^T Q x - 1)$$

$$\nabla_x L(x, v) = 2x - v(2Qx) = 0$$

$$\Leftrightarrow (I - vQ)x = 0$$

$$\Leftrightarrow vQx = x. \quad (5)$$

$x$  is an eigenvector of  $vQ$ , with eigenvalue 1.

$$vQ = \sum_i (v\lambda_i) q_i q_i^T$$

$$(PF) \quad \underline{x^T Q x = 1.} \quad vQx = x$$

To ensure that  $x$  is an e-vector of  $vQ$  with e-value 1, we set  $v > 0$  s.t.  $v\lambda_i = 1$  for some  $i = 1, \dots, n$ .

There are  $n$  possible choices for  $v$  ( $\frac{1}{\lambda_1}, \frac{1}{\lambda_2}, \dots, \frac{1}{\lambda_n}$ ).

$$PF \quad x^T \left( \frac{1}{v} x \right) = 1 \Rightarrow v = \|x\|^2$$

Want to minimize  $\|x\|^2$ , choose  $v$  to be the smallest possible out of the  $n$  choices.  $v = \frac{1}{\lambda_1}$

Choose  $x$  to be the eigenvector corresponding to the largest eigenvalue ( $\lambda$ ) of  $Q$ .