

$$\min x_1^2 + 0.5x_2^2 + x_1x_2 - 2x_1 - 3x_2$$

$$x_1 + x_2 \leq 1$$

i) Sufficient?

$$f(\underline{x}) = \underline{x}^T \begin{bmatrix} 1 & 0.5 \\ 0.5 & 0.5 \end{bmatrix} \underline{x} + [-2 \ -3] \underline{x}$$

constraint is a linear ineq (convex)

convex f ? $\nabla^2 f = 2 \begin{bmatrix} 1 & 0.5 \\ 0.5 & 0.5 \end{bmatrix} \rightarrow Q \succeq 0?$ $\det(Q) = 0.25 > 0$
 $\lambda_1(Q) = 1.5 > 0$
 $\Rightarrow Q \succeq 0 \Rightarrow \int \text{convex}$

\Rightarrow KKT are necessary and sufficient.

KKT: $L = \underline{x}^T Q \underline{x} + b^T \underline{x} + \lambda(x_1 + x_2 - 1)$ $\rightarrow 2x_1 + x_2 - 2 + \lambda = 0$
 $\hookrightarrow \nabla_x L = 0 \Leftrightarrow 2Q\underline{x} + b + \lambda \mathbf{1} = 0$ (*) $x_1 + x_2 - 3 + \lambda = 0$
 $\lambda(x_1 + x_2 - 1) = 0$

cases

1) $\lambda = 0 \Rightarrow 2x_1 + x_2 - 2 = 0$ (*)

$$x_1 + x_2 - 3 = 0$$

$$\Rightarrow \begin{matrix} x_1 = -1 \\ x_2 = 4 \end{matrix} \rightarrow x_1 + x_2 = 3 > 1 \quad \textcircled{\text{xx}}$$

2) $\lambda > 0 \Rightarrow x_1 + x_2 = 1 \Rightarrow \lambda = 2 \Rightarrow \begin{matrix} x_1 = -1 \\ x_2 = 2 \end{matrix}$

Suff and necessity $\Rightarrow (x_1, x_2) = (-1, 2)$ is a minimizer.

ii) Convex cost + linear constraints
 + Slater ($x_1 = x_2 = 0, x_1 + x_2 < 1$)

\Rightarrow Strong duality.

Dual: $\mathcal{L}(\underline{x}, \lambda) = \underline{x}^T Q \underline{x} + b^T \underline{x} + \lambda(x_1 + x_2 - 1)$
 $(X = \mathbb{R}^2)$

$$\min_{\underline{x} \in X = \mathbb{R}^2} \mathcal{L}(\underline{x}, \lambda) \Leftrightarrow \nabla_{\underline{x}} \mathcal{L}(\underline{x}, \lambda) = 0$$

$$\Leftrightarrow 2Q\underline{x} + b + \lambda \mathbf{1} = 0$$

$$\underline{x}^* = -\frac{1}{2} Q^{-1} (b + \lambda \mathbf{1})$$

$$Q = \begin{bmatrix} 1 & 0.5 \\ 0.5 & 0.5 \end{bmatrix} \rightarrow Q^{-1} = \frac{1}{0.25} \begin{bmatrix} 0.5 & 0.5 \\ -0.5 & 1 \end{bmatrix} = \begin{bmatrix} 2 & -2 \\ -2 & 4 \end{bmatrix}$$

$$\underline{x}^* = \begin{bmatrix} -1 \\ 4 - \lambda \end{bmatrix}$$

Dual: $\max_{\lambda \geq 0} \mathcal{L}(\underline{x}^*, \lambda) = \max_{\lambda \geq 0} 1 + \frac{1}{2} (4 - \lambda)^2 + \lambda - 4 + 2 - 3(4 - \lambda) + \lambda(2 - \lambda)$

$$= -\frac{1}{2} \lambda^2 - 5 + 2\lambda$$

$$\max_{\lambda \geq 0} \dots \quad \underline{f}' = 0 \quad -\lambda + 2 = 0 \Rightarrow \lambda = 2$$

$$\underline{c.s.} \subseteq \Rightarrow \underline{x}^* = \begin{bmatrix} -1 \\ 4 - \lambda \end{bmatrix} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

Strong duality \Rightarrow duality gap = 0