

Quiz-2 (CS709)

Total Marks: 10. Duration: 1hr.

Consider the following convex program¹:

$$\begin{array}{ll} \min_{x \in \mathbb{R}^n} & c^\top x, \\ \text{s.t.} & b - A^\top x \in K, \end{array}$$

where $c \in \mathbb{R}^n, b \in \mathbb{R}^m, A$ is an $n \times m$ matrix and $K \subset \mathbb{R}^m$ is a cone. Assume that the program is feasible. Given this, attempt the following:

1. Prove this statement: the above program is bounded $\Rightarrow \exists y \in \mathbb{R}^m \ni Ay + c = 0$.

[3 Marks]

2. In general, is the converse of the above statement true?

[1 Mark]

3. Fill in the blanks² in this statement: x^* is optimal for the program if and only if there exists a $y^* \in \mathbb{R}^m$ such that:

- (a) -----
- (b) $Ay^* + c = 0$
- (c) -----

Note that you need to fill each of the above blank with exactly one expression and moreover in such a fashion that once you consider the special case $K = \mathbb{R}_+^m$ (i.e., the program is a linear program³), the above conditions must turn out to be exactly the same as the KKT conditions for this case.

[6 Marks]

¹We will hence-forth refer to programs of this form as conic programs.

²Needless to say, you need to provide rigorous justification for the blanks you fill.

³ $\mathbb{R}_+^m \equiv \{x \in \mathbb{R}^m \mid x \geq 0\}$.

Quiz 2 Solutions

19/10/12

Given the following feasible conic program:

$$\textcircled{P} = \begin{array}{ll} \min & c^T x \\ \text{s.t.} & b - A^T x \in K \end{array}$$

① TST \textcircled{P} is bounded $\Rightarrow \exists y \in \mathbb{R}^m \ni Ay + c = 0$
i.e. TST \textcircled{P} is bounded $\Rightarrow c \in \mathcal{C}(A)$ (the column space of A)

If $c = 0$, there is nothing to prove, so let's assume $c \neq 0$.

Proof is by contradiction. suppose $c \notin \mathcal{C}(A)$. Then we have that $\exists v \in \mathcal{N}(A^T) \ni c^T v \neq 0$ and in particular $c^T v < 0$.
(null space of A^T) (by rank nullity theorem)

Now since \textcircled{P} is feasible, we have $\exists x_0 \in \mathbb{R}^n \ni b - A^T x_0 \in K$

$$\Rightarrow b - A^T(x_0 + \lambda v) \in K \quad \forall \lambda \geq 0$$

Consider the sequence ~~$c^T(x_0 + \lambda v)$~~ , ~~$c^T(x_0 + 2v)$~~ , ~~$c^T(x_0 + 3v)$~~ , ~~\dots~~

$$c^T(x_0 + v), c^T(x_0 + 2v), c^T(x_0 + 3v), \dots$$

~~This~~ This goes to $-\infty$ as $c^T v < 0$. Since all $x_0 + \lambda v$ are feasible, we have that $\textcircled{P} = -\infty$ i.e. \textcircled{P} is unbounded.

Hence Proved.

② The converse is not true in general.

For eg. ~~the~~ ① is unbounded whenever $K = \mathbb{R}^m$, irrespective of the relation between c & A .

$$\begin{aligned}
 \textcircled{3} \quad x^* \text{ is optimal} &\Leftrightarrow c^T x \geq c^T x^* \quad \forall \quad x \ni b - A^T x \in K \\
 &\Leftrightarrow \textcircled{P} \text{ is bounded} \& c^T x \geq c^T x^* \quad \forall \quad x \ni b - A^T x \in K \\
 &\Leftrightarrow \exists \lambda^* \in \mathbb{R}^m \ni A \lambda^* + c = 0, \quad c^T x \geq c^T x^* \quad \forall \quad x \ni b - A^T x \in K \\
 &\Leftrightarrow \exists \lambda^* \in \mathbb{R}^m \ni A \lambda^* + c = 0, \quad -\lambda^{*T} A^T x \geq -\lambda^{*T} A^T x^* \quad \forall \quad x \ni b - A^T x \in K \\
 &\Leftrightarrow \exists \lambda^* \in \mathbb{R}^m \ni A \lambda^* + c = 0, \quad \lambda^{*T} (b - A^T x) \geq \lambda^{*T} (b - A^T x^*) \quad \forall \quad x \ni b - A^T x \in K \\
 &\Leftrightarrow \exists \lambda^* \in \mathbb{R}^m \ni A \lambda^* + c = 0, \quad \min_{\substack{x \in \mathbb{R}^n \\ \text{s.t. } b - A^T x \in K}} \lambda^{*T} (b - A^T x) = \lambda^{*T} (b - A^T x^*)
 \end{aligned}$$

$$\text{Now, } \min_{\substack{x \in \mathbb{R}^n \\ \text{s.t. } b - A^T x \in K}} \lambda^{*T} (b - A^T x) = \begin{cases} 0 & \lambda^* \in K^* \\ -\infty & \lambda^* \notin K^* \end{cases}$$

$$\begin{aligned}
 \text{Hence, } x^* \text{ is optimal} &\Leftrightarrow \exists \lambda^* \in \mathbb{R}^m \ni A \lambda^* + c = 0, \quad \lambda^* \in K^*, \quad \lambda^{*T} (b - A^T x^*) = 0 \\
 \text{Necessary case } K = \mathbb{R}_+^m &\longrightarrow \begin{matrix} \downarrow & \downarrow & \downarrow \\ (\text{none}) & (\lambda^* \geq 0) & (\lambda_i^* (b - A^T x^*)_i = 0 \quad \forall i) \end{matrix} \\
 &\quad \quad \quad \underbrace{\hspace{10em}}_{\text{KKT conditions in LP case!}} \quad \text{(because } x^* \text{ is feasible)}
 \end{aligned}$$

So from now on we will call the above as KKT conditions.