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NONLINEAR OPTIMIZATION — Remark on Lecture 16

On Pg.1 of slides for Lecture 16, it was claimed that, given $A \in \mathbb{S}^n$ and $b \in \mathbb{R}^n$, the following two problems are equivalent:

$$\begin{aligned} \max_{\lambda} & -b^{T}(A+\lambda I)^{\dagger}b-\lambda & \max_{\lambda,t} & -t-\lambda \\ \text{s.t.} & \begin{cases} A+\lambda I \succeq 0 \\ b \in \mathcal{R}(A+\lambda I) \end{cases} & \text{s.t.} & \begin{bmatrix} A+\lambda I & b \\ b^{T} & t \end{bmatrix} \succeq 0. \end{aligned}$$

First the constraint $\lambda \geq 0$ is missing in both of the optimization problems, because as a dual variable for the inequality constraint $x^T x \leq 1$, there is a sign constraint on λ . But for ease of notation we leave it out for now.

So why are the above problems equivalent? First, you can prove that the following two problems are equivalent:

$$\begin{array}{ll} \max_{\lambda,t} & -t - \lambda & \max_{\lambda,t} & -t - \lambda \\ \text{s.t.} & A + \lambda I \succeq 0 & \text{s.t.} & A + \lambda I \succeq 0 \\ & t = b^T (A + \lambda I)^{\dagger} b & t \ge b^T (A + \lambda I)^{\dagger} b \\ & b \in \mathcal{R}(A + \lambda I) & b \in \mathcal{R}(A + \lambda I). \end{array}$$

Therefore, it suffices to show that for any (λ, t) ,

$$\begin{bmatrix} A + \lambda I & b \\ b^T & t \end{bmatrix} \succeq 0 \iff \begin{cases} A + \lambda I \succeq 0 \\ t \ge b^T (A + \lambda I)^{\dagger} b \\ b \in \mathcal{R}(A + \lambda I) \end{cases}$$
(1)

To get an idea of how to prove the equivalence, first recall Schur's theorem and its proof:

Theorem 1 Suppose $A \in \mathbb{R}^{k \times k}$ is symmetric positive definite. Then for $B \in \mathbb{R}^{n \times k}$, $C \in \mathbb{R}^{n \times n}$,

$$\begin{bmatrix} A & B^T \\ B & C \end{bmatrix} \succeq 0 \iff C - BAB^T \succeq 0 \text{ ; also, } \begin{bmatrix} A & B^T \\ B & C \end{bmatrix} \succ 0 \iff C - BAB^T \succ 0$$

Proof. Note that if Q is an invertible matrix, for any matrix X of compatible dimension, we have

$$X \succeq 0 \iff QXQ^T \succeq 0$$
; also, $X \succeq 0 \iff QXQ^T \succeq 0$.

Since

$$\begin{bmatrix} A & B^T \\ B & C \end{bmatrix} = \begin{bmatrix} I_k & 0 \\ BA^{-1} & I_n \end{bmatrix} \begin{bmatrix} A & 0 \\ 0 & C - BA^{-1}B^T \end{bmatrix} \begin{bmatrix} I_k & A^{-1}B^T \\ 0 & I_n \end{bmatrix}$$

and $\begin{bmatrix} I_k & 0\\ BA^{-1} & I_n \end{bmatrix}$ is invertible, the theorem follows. \Box

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Now return to our constraints $A + \lambda I \succeq 0$ and $t - b^T (A + \lambda I) b \ge 0$. Let's put it on the diagonal to obtain a block diagonal matrix, and "conjugate" it as in the proof of Schur's theorem: for any $y \in \mathbb{R}^n$,

$$\begin{bmatrix} I_n & 0\\ y^T & 1 \end{bmatrix} \begin{bmatrix} A+\lambda I_n & 0\\ 0 & t-b^T(A+\lambda I_n)^{\dagger}b \end{bmatrix} \begin{bmatrix} I_n & y\\ 0 & 1 \end{bmatrix} = \begin{bmatrix} A+\lambda I & 0\\ y^T(A+\lambda I) & t-b^T(A+\lambda I)^{\dagger}b \end{bmatrix} \begin{bmatrix} I & y\\ 0 & 1 \end{bmatrix}$$
$$= \begin{bmatrix} A+\lambda I & (A+\lambda I)y\\ y^T(A+\lambda I) & t-b^T(A+\lambda I)^{\dagger}b + y^T(A+\lambda I)y \end{bmatrix}$$

With this we are ready to proof (1). If RHS holds, let $y \in \mathbb{R}^n$ satisfy $(A + \lambda I)y = b$. Since

$$(A + \lambda I)(A + \lambda I)^{\dagger}(A + \lambda I) = A + \lambda I$$

by definition of Moore-Penrose inverse, $b^T (A + \lambda I)^{\dagger} b = y^T (A + \lambda I) y$. This shows that

$$\begin{bmatrix} A + \lambda I & b \\ b^T & t \end{bmatrix} = \begin{bmatrix} A + \lambda I & (A + \lambda I)y \\ y^T (A + \lambda I) & t - b^T (A + \lambda I)^{\dagger} b + y^T (A + \lambda I)y \end{bmatrix}$$

is positive semidefinite.

Conversely, if LHS of (1) holds, we must have $b \in \mathcal{R}(A + \lambda I)$; if not, the projection b' of b on $\mathcal{R}(A + \lambda I)^{\perp} = \text{Null}(A + \lambda I)$ is non-zero, and for any $\eta > 0$,

$$\begin{bmatrix} -\eta b'^T & 1 \end{bmatrix} \begin{bmatrix} A + \lambda I & b \\ b^T & t \end{bmatrix} \begin{bmatrix} -\eta b' \\ 1 \end{bmatrix} = -2\eta \|b'\|^2 + t < 0$$

for sufficiently large η , contradicting LHS of (1). Hence $b = (A + \lambda I)y$ for some $y \in \mathbb{R}^n$. Together with the block diagonalization, the RHS of (1) holds.

The strong duality result for the primal-dual pair

$$\min x^T A x + 2b^T x \quad \text{s.t.} \quad x^T x \leq 1 \\ \max -t - \lambda \quad \text{s.t.} \quad \begin{bmatrix} A + \lambda I & b \\ b^T & t \end{bmatrix} \succeq 0, \ \lambda \geq 0$$

can be found in Appendix B of the text.

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