

Consider the primal program (page 89). Define

$$(B, t) = (I_n, 1), \quad Y = \begin{bmatrix} 0 & 0 \\ 0 & I_n \end{bmatrix}.$$

This is a feasible solution for the primal because $P \in \mathcal{B}_n$.

The value of the primal is equal to 1. The value of the dual is equal to $\frac{1}{n} \sum_{i=1}^M \lambda_i = 1$, too, because

$$\sum \lambda_i \stackrel{(i)}{=} \sum \lambda_i x_i^T x_i = \text{Tr}(\sum \lambda_i x_i x_i^T) \stackrel{(ii)}{=} \text{Tr}(I) = n.$$

In particular,

" \Rightarrow " Application of strong duality:

$P \in \mathcal{B}_n$: \checkmark

Let y_1^*, \dots, y_M^* be an optimal solution of the dual program. After reordering we may assume

$$y_1^* > 0, \dots, y_M^* > 0, \quad y_{M+1}^* = \dots = y_N^* = 0.$$

By assumption, $(B, t) = (I_n, 1)$, $Y = \begin{bmatrix} 0 & 0 \\ 0 & I_n \end{bmatrix}$ is an optimal solution of the primal. By complementary slackness (Theorem III.5.2) we know $s_i = 0$, $i = 1, \dots, M$.

$$\text{Hence } \left\langle \begin{bmatrix} 1 & -x_i^T \\ -x_i^T & x_i x_i^T \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & I_n \end{bmatrix} \right\rangle = 1 \text{ for } i \in [M].$$

This shows $\|x_i\| = 1$, condition (a).

Let $Z^* = \sum z_{ij}^* E_{ij}$ be an optimal solution for the dual. Then (Theorem 2.6): $I_n Z^* = \alpha I_n$ for $\alpha > 0$, and $(\det Z^*)^{1/n} = \frac{1}{n}$. Hence, $Z^* = \frac{1}{n} I_n$.

Let Y^* be optimal for the primal. Then

$$Y^* \left(- \sum_{i,j} z_{ij}^* E_{i+n, j+n} + \sum_i y_i^* \begin{bmatrix} 1 & -x_i^T \\ -x_i & x_i x_i^T \end{bmatrix} \right) = 0$$

$$\Leftrightarrow \sum_i y_i^* \begin{bmatrix} 1 & -x_i^T \\ -x_i & x_i x_i^T \end{bmatrix} = \begin{bmatrix} * & 0 \\ 0 & \frac{1}{n} I_n \end{bmatrix}. \quad \left(\begin{array}{l} \text{because} \\ \begin{bmatrix} 0 & 0 \\ 0 & I_n \end{bmatrix} \begin{bmatrix} A & B \\ c & D \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ c & D \end{bmatrix} \\ = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \end{array} \right)$$

Define $\lambda_i = n y_i^*$, then (b), (c) follow from the equality above. \square

Corollary 8 Let $P \subseteq \mathbb{R}^n$ be a polytope with $\dim P = n$.

(i) If $E_{\text{out}}(P) = B_n$, then $\frac{1}{n} B_n \subseteq P \subseteq B_n$.

(ii) If $E_{\text{in}}(P) = B_n$, then $B_n \subseteq P \subseteq n B_n$.

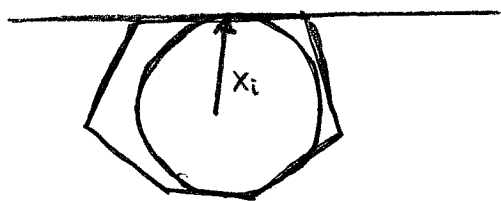
Both inclusions are best possible \rightarrow Exercises.

Proof (i) follows from (ii) by polarity.

(ii) Apply Theorem 7: We have

$$\sum_{i=1}^m \lambda_i x_i = 0, \quad \sum_{i=1}^m \lambda_i x_i x_i^T = I_n, \quad \lambda_i > 0.$$

Consider the supporting hyperplane through $x_i \in B_n \cap \partial P$.



It is orthogonal to x_i .

Hence,

$$B_n \subseteq P \subseteq Q = \{x \in \mathbb{R}^n : x_i^T x \leq 1, i \in [m]\}.$$

If $x \in Q$, then $x^T x_i \in [-\|x\|, 1]$. Hence,

$$\begin{aligned} 0 &\leq \sum_{i=1}^m \lambda_i (1 - x^T x_i) (\|x\| + x^T x_i) \\ &= \|x\| \sum_{i=1}^m \lambda_i + (1 - \|x\|) \sum_{i=1}^m \lambda_i x^T x_i - \sum_{i=1}^m \lambda_i (x^T x_i)^2 \\ &= \|x\| m + 0 - \|x\|^2 \end{aligned}$$

$$\Rightarrow \|x\| \leq m.$$

because $\sum \lambda_i = \sum \lambda_i x_i^T x_i$

$$\begin{aligned} &= \text{Tr}(\sum \lambda_i x_i x_i^T) \\ &= \text{Tr}(I_n) = m. \end{aligned}$$

□