

## Exercises

1. Any principal submatrix of a positive semidefinite matrix is again positive semidefinite.
2. If a diagonal element of a positive semidefinite matrix is zero, so are all offdiagonal elements of this row.
3. For any given regular  $B \in \mathbb{R}^{n \times n}$  there holds  $A \in \mathcal{S}_+^n \Leftrightarrow BAB^T \in \mathcal{S}_+^n$ .
4. Given orthogonal  $P \in \mathbb{R}^{n \times k}$  ( $P^T P = I_k$ ) and  $A \in \mathcal{S}_+^n$ , assume  $A \notin \{PUP^T : U \succeq 0\}$ . Then there is a  $v \in \mathbb{R}^n$  with  $P^T v = 0$  and  $v^T A v > 0$ .
5.  $A \in \mathcal{S}_+^n \Leftrightarrow \langle A, B \rangle \geq 0 \quad \forall B \in \mathcal{S}_+^n$ .
6. (\*) Let  $P \in \mathbb{R}^{n \times k}$  be an orthogonal matrix ( $P^T P = I_k$ ), then  $\{PUP^T : U \in \mathcal{S}_+^k\}$  is a face of  $\mathcal{S}_+^n$  and any face of  $\mathcal{S}_+^n$  of dimension at least one has such a representation.
7. Semidefinite Scaling: Given regular  $B \in \mathbb{R}^{n \times n}$ , the semidefinite programs

$$\inf\{\langle C, X \rangle : \langle A_i, X \rangle = b_i, i = 1, \dots, m, X \succeq 0\}$$

and

$$\inf\{\langle BCB^T, \bar{X} \rangle : \langle BA_i B^T, \bar{X} \rangle = b_i, i = 1, \dots, m, \bar{X} \succeq 0\}$$

have the same optimal value and there is an automorphism on  $\mathcal{S}_+^n$  that bijectively maps feasible/optimal  $X$ -solutions to feasible/optimal  $\bar{X}$ -solutions.

8. (\*) Show that the ellipsope relaxation

$$\min\{\langle C, X \rangle : X_{ii} = 1, i = 1, \dots, n, X \succeq 0\}$$

and the quadratic 0-1 relaxation

$$\min\{\langle Q, \bar{X} \rangle : \bar{X}_{11} = 1, \bar{X}_{1i} = \bar{X}_{ii}, i = 2, \dots, n, \bar{X} \succeq 0\}$$

are scalings of each other.

9. (\*\*) Prove or disprove: If  $\{X \succeq 0 : \langle A_i, X \rangle = b_i, i = 1, \dots, m\}$  is bounded, there is a scaling by some regular  $B \in \mathbb{R}^n \times n$  so that for matrices in  $\{\bar{X} = BXB^T \succeq 0 : \langle B^{-T} A_i B^{-1}, \bar{X} \rangle = b_i, i = 1, \dots, m\}$  the trace is constant.
10. (\*) Let  $\hat{X} \in \mathcal{S}_+^n$  be a point in the relative interior of the feasible set of  $\inf\{\langle C, X \rangle : \langle A_i, X \rangle = b_i, i = 1, \dots, m, X \succeq 0\}$  having eigenvalue decompositions  $\hat{X} = P\Lambda P^T$  with diagonal positive definite  $\Lambda \in \mathcal{S}_+^k$ ,  $k \leq n$  and  $P^T P = I_k$ . Prove that the program  $\inf\{\langle PCP^T, U \rangle : \langle PA_i P^T, U \rangle = b_i, i = 1, \dots, m, U \succeq 0\}$  is equivalent to the first and that its dual satisfies strong duality.

11. Let  $\hat{M} \subset \mathbb{R}^{m+1}$  be some convex compact set and  $\hat{y} \in \mathbb{R}^m$ . Find a saddle point theorem that proves the existence of saddlepoints for the problem

$$\inf_{y \in \mathbb{R}^m} \sup_{(\gamma, g) \in \hat{M}, \eta \in \mathbb{R}_+^m} [\gamma + \langle g - \eta, y \rangle + \frac{1}{2} \|y - \hat{y}\|^2].$$

12. (\*) Extend the Gauss-Seidel approach for dealing with sign constraint variables to box-constraints.
13. (\*\*) Extend the spectral bundle approach for SDP to second order cone problems. Show that in this case there is no need for a residual part in the model and a  $P$  with at most three columns suffices to guarantee convergence.
14. (\*) Try to complete the proof on the bound on the number of strong nodal domains, i.e., in the notation of the sketch, show  $g^T M g \leq \lambda_k$  (this might need an idea).
15. (\*) Given a connected graph  $G = (N, E)$ , prove that the second smallest eigenvalue of the Laplace matrix of  $G$  is strictly positive (hint: use Perron-Frobenius or the Matrix Tree Theorem).
16. (\*) Given a connected graph  $G = (N, E)$ , denote the weighted Laplacian by  $L_w(G)$  and let  $\hat{a}(G) := \sup\{\lambda_2(L_w(G)) : \sum_{ij \in E} w_{ij} = |E|, w \in \mathbb{R}_+^E\}$ . Prove

$$\begin{aligned} \frac{|E|}{\hat{a}(G)} &= \text{maximize} && \sum_{i \in N} \|v_i\|^2 \\ &\text{subject to} && (\sum_{i \in N} v_i)^2 = 0, \\ &&& \|v_i - v_j\|^2 \leq 1 \quad \text{for } ij \in E, \\ &&& v_i \in \mathbb{R}^n \text{ for } i \in N. \end{aligned}$$

In proving this, make certain that strong duality holds whenever you dualize!

17. (\*) Given a tree  $G = (N, E)$ , let  $v_i, i = 1, \dots, n$  be an optimal embedding corresponding to the maximized second smallest eigenvalue of the Laplace matrix of  $G$ . Show that the origin either coincides with a  $v_i$  for some  $i \in N$  or there is an edge  $ij \in E$  with  $0 \in [v_i, v_j]$ .