

Chapter VIII Copositive Programming

Until now we made the impression that convex optimization is easy ($\hat{=}$ polynomial time solvable). This is (basically) true if it comes to optimizing over the cones $\mathbb{R}_{\geq 0}^n$, \mathcal{L}^{n+1} , $\mathcal{S}_{\geq 0}^n$, \mathcal{O}^n .

But: Conic programming can be NP-hard.

§ 1 The completely positive and the copositive cone

Def. 1 $\mathcal{CP}_n = \text{cone} \{ xx^T : x \in \mathbb{R}_{\geq 0}^n \}$ is called the cone of completely positive matrices

Prop. 2 \mathcal{CP}_n is a proper convex cone.

Proof:

- \mathcal{CP}_n is convex cone: ✓
- \mathcal{CP}_n is pointed: clear because $\mathcal{CP}_n \subseteq \mathcal{S}_{\geq 0}^n \cap \mathbb{R}_{\geq 0}^{n(n+1)/2}$.
- $\text{int } \mathcal{CP}_n \neq \emptyset$: clear because $i(e_i + e_j)(e_i + e_j)^T$, $1 \leq i < j \leq n$, are linearly independent and lie in \mathcal{CP}_n .

- $C_{\mathbb{R}_n}$ is closed: Let $(A_i)_{i \in \mathbb{N}}$ be a sequence with $A_i \in C_{\mathbb{R}_n}$ and $\lim_{i \rightarrow \infty} A_i = A$.

By the theorem of Carathéodory there are $N = \frac{n(n+1)}{2}$ vectors $x_j^i \in \mathbb{R}_{\geq 0}^n, j=1, \dots, N$, so that $A_i = \sum_{j=1}^N x_j^i (x_j^i)^T$.

Claim: $\{x_j^i : i \in \mathbb{N}, j=1, \dots, N\}$ is bounded.

Proof: For suppose not. Then there is a $j \in \{1, \dots, N\}$ so that $\{x_j^i : i \in \mathbb{N}\}$ is not bounded. Hence there is a coordinate $k \in \{1, \dots, n\}$ so that $\{(x_j^i)_k : i \in \mathbb{N}\}$ is not bounded. Since $x_j^i \in \mathbb{R}_{\geq 0}^n$, $(A_i)_{kk} = (x_j^i)_k^2 \rightarrow \infty$ for $i \rightarrow \infty$. So the limit A cannot exist. Contradiction.

Because of the claim the sets $\{x_j^i : i \in \mathbb{N}\}$, for each j , are contained in a compact subset of $\mathbb{R}_{\geq 0}^n$. Hence there exists a converging subsequence of $(x_j^i)_{i \in \mathbb{N}}$ converging to $x_j \in \mathbb{R}_{\geq 0}^n$. Now we find a common converging subsequence for all j and get $A = \sum_{j=1}^N x_j x_j^T$. □

Def. 3 $\text{COP}_n = \{ A \in S_{\geq 0} : x^T A x \geq 0 \text{ for all } x \in \mathbb{R}_{\geq 0}^n \}$.

is called the cone of copositive matrices.

Clear: $S_{\geq 0}^n + \mathbb{R}_{\geq 0}^{n(n+1)/2} \subseteq \text{COP}_n$.

(but in general \subsetneq).

Proposition 4 $(\text{CP}_n)^* = \text{COP}_n$.

Proof " \subseteq ": Let $A \in (\text{CP}_n)^*$. Then for all $x \in \mathbb{R}_{\geq 0}^n$

$$0 \leq \langle A, xx^T \rangle = x^T A x \Rightarrow A \in \text{COP}_n.$$

" \supseteq ": Let $A \in \text{COP}_n$ and let $B \in \text{CP}_n$. Then

$$B = \sum_{i=1}^k \lambda_i x_i x_i^T, \quad \lambda_i \geq 0, \quad x_i \in \mathbb{R}_{\geq 0}^n \text{ and}$$

$$\langle A, B \rangle = \sum_{i=1}^k \lambda_i \langle A, x_i x_i^T \rangle = \sum_{i=1}^k \underbrace{\lambda_i}_{\geq 0} \underbrace{x_i^T A x_i}_{\geq 0} \geq 0. \quad \square$$

Def. 5 Given $C, A_1, \dots, A_m \in S^n$, $b_1, \dots, b_m \in \mathbb{R}$. This defines a copositive program

$$(P) \quad p^* = \sup_{X \in \text{CP}_n} \langle C, X \rangle \quad (D) \quad d^* = \inf_{Y \in \mathbb{R}^m} \sum_{j=1}^m y_j A_j - C$$

$$\langle A_j, X \rangle = b_j, \quad j \in [m]$$

$$\sum_{j=1}^m y_j A_j - C \in \text{COP}_n$$