

Intersection Cuts for Mixed-Integer Signomial Sets

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1 Introduction

Intersection cuts [3, 10] have been tackled in recent years to tighten the polyhedral outer approximation of certain non-convex sets such as lattice set [1], bi-level set [5], outer product set [2] and quadratic set [9].

In this work, we study the intersection cuts for *signomial* set and *mixed-integer-signomial* set. Signomial Programs (SPs) and Mixed-Integer Signomial Programs (MISPs) [6, 8] contain signomial terms, the intersection cuts can tighten their Linear Programming (LP) relaxations.

Given an n -dimensional multi-index $\alpha = (\alpha_1, \alpha_2, \dots, \alpha_n) \in \mathbf{R}^n$, a signomial term is defined as $x^\alpha := \prod_{i=1}^n x_i^{\alpha_i}$. We assume the variables being positive, thus allowing the multi-index α to have negative components. We define the signomial set as follows :

$$\mathcal{S}_s := \{(x, y) \in \mathbf{R}_+^n \times \mathbf{R} : y(\cdot) x^\alpha\},$$

where \cdot can be $\leq, =$ or \geq .

We define the mixed-integer counterpart of the signomial set. Let $[\underline{x}, \bar{x}]$ be the box constraint on x such that $\underline{x} > 0$ implicitly imposes the positiveness of x . W.l.o.g., we assume the first $n_1 \leq n$ entries of x to be integer, and let $\mathcal{D} = (\mathbf{Z}_+^{n_1} \times \mathbf{R}_+^{n-n_1}) \cap [\underline{x}, \bar{x}]$ be the bounded mixed-integer domain. Similarly, the mixed-integer-signomial set is defined as :

$$\mathcal{S}_{mis} := \{(x, y) \in \mathcal{D} \times \mathbf{R} : y(\cdot) x^\alpha\}.$$

For a non-convex set \mathcal{S} , a convex set \mathcal{C} is \mathcal{S} -free if $\text{int } \mathcal{C} \cap \mathcal{S} = \emptyset$. The intersection cut framework requires two ingredients to generate a valid inequality to separate \bar{z} :

- i) a translated simplicial conic relaxation \mathcal{R} with vertex \bar{z} such that $\mathcal{S} \subset \mathcal{R}$ and $\bar{z} \notin \mathcal{S}$;
- ii) a \mathcal{S} -free set \mathcal{C} such that $\bar{z} \in \text{int } \mathcal{C}$.

The spatial-branch-and-bound (sBB) algorithm [7] is a general technique to find the global solution of non-convex MINLPs. The second ingredient of intersection cuts, i.e., recession conic relaxation, can be retrieved from the LP-based sBB algorithm. If two \mathcal{S} -free sets $\mathcal{C}_1, \mathcal{C}_2$ are such that $\mathcal{C}_1 \subset \mathcal{C}_2$, the intersection cut derived from \mathcal{C}_2 dominates the one derived from \mathcal{C}_1 [4]. Therefore, the maximal \mathcal{S} -free sets are ideal and crucial to generate strong intersection cuts.

The set \mathcal{S} has a *difference-of-concave* (DC) representation if it can be expressed as

$$\mathcal{S} := \{z \in \mathbf{R}^p : f_1(z) - f_2(z) \leq 0\},$$

where f_1 and f_2 are concave functions.

[10] proposed the *linearization* technique to derive \mathcal{S} -free sets in DC-representations. The linearization technique constructs the following \mathcal{S} -free set :

$$\mathcal{C} := \{z \in \mathbf{R}^p : f_1(z) - (f_2(\tilde{z}) + \nabla f_2(\tilde{z}) \cdot (z - \tilde{z})) \geq 0\},$$

where $f_2(\tilde{z}) + \nabla f_2(\tilde{z}) \cdot (z - \tilde{z})$ is the linearization (over-estimator) of f_2 at \tilde{z} .

Indeed, a non-convex set \mathcal{S} may have various DC-representations. An *ideal* DC-representation should yield maximal \mathcal{S} -free sets using the linearization technique.

2 Our results

We show that the signomial set can be transformed into the following *normalized* DC-representation :

$$\mathcal{S}_s = \{(u, v) \in \mathbf{R}_+^h \times \mathbf{R}_+^l : u^\beta - v^\gamma \leq 0\}, \quad (1)$$

where $\beta \in \mathbf{R}_+^h$ and $\gamma \in \mathbf{R}_+^l$ are h - and l - dimensional multi-indices such that $\max\{\|\beta\|_1, \|\gamma\|_1\} = 1$. Because u^β, v^γ are concave, the representation yields a \mathcal{S}_s -free set using the linearization technique.

For any non-convex set \mathcal{S} , we give a sufficient condition to guarantee the maximality of a \mathcal{S} -free set ; then, we show that the normalized DC-representation is *ideal* in the sense that it yields the maximal \mathcal{S}_s -free set in the non-negative orthant. Some non-ideal DC-representations of \mathcal{S}_s are given as well.

We propose a method to enlarge the \mathcal{S}_s -free set into the mixed-integer-signomial-free (\mathcal{S}_{mis} -free) set, where both nonlinear constraints and bounded/integer variables are featured. A byproduct is a strong convex relaxation of the following mixed-integer power cone :

$$\{(x, t) \in \mathcal{D} \times \mathbf{R} : x^\alpha \geq t\}, \quad (2)$$

where $\alpha \in \mathbf{R}_+^p$, $\|\alpha\|_1 \leq 1$, and $\mathcal{D} \subset \mathbf{R}_+^p$ is a bounded mixed-integer domain. The relaxation is stronger than the continuous relaxation $\{(x, t) \in \mathbf{R}_+^p \times \mathbf{R} : x^\alpha \geq t\}$.

The intersection cut separation problem is reduced to univariate root-finding problems. Such root-finding problems can be solved numerically in quadratic convergence rate by Newton-like algorithms.

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