

Convex Relaxations for Cubic Polynomials Problems

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Introduction

We consider a problem in the form :

$$\begin{array}{ll} \min_x & c'x \\ \text{subject to} & g_s(x) \leq 0, \quad s \in S \\ & 0 \leq x^l \leq x \leq x^u < \infty \end{array}$$

where $g(x)$ is a polynomial function of degree up to 3:

$$g_s(x) = \sum_{i \in I} a_i x^i, \quad s \in S$$

with a_i a real coefficient and

$x^i = \prod_{k=1}^n x_k^{i_k}$, $i_k \in \mathbf{N}_0$ and $\sum_{k=1}^n i_k \leq 3$.

Introduction

- ▶ If all $g_s(x)$ are convex functions then problem is convex, so any local optimum solution is also global.
- ▶ In general functions $g_s(x)$ are not convex.

Approach:

For nonconvex $g_s(x)$ replace it by convex approximation.

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talk about branch and bound

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Convex under approximation

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Picture of polyhedral under approximation

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Figure of convex relaxation

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(the convex should in this case be better than polyhedral)

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Convex Relaxations

We consider for simplicity a single constraint

$$g(x) \leq 0$$

We want to build a convex relaxation for the set

$$S = \{x \mid g(x) \leq 0\}$$

that is a set S^C such that

- ▶ S^C is convex.
- ▶ $S \subseteq S^C$

Ideally we would like to have that if $S^{C'}$ is another convex set such that $S \subseteq S^{C'}$ then $S^C \subseteq S^{C'}$, that is S^C is the convex hull of S .

Convex Relaxations

Unfortunately this approach not attainable in general.

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(Mention Sahinidis and Bao here?)

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What can be done?

build relaxations for individual terms for which usually a tight convex relaxation is known.

Quadratic Case

We consider for simplicity a single constraint

$$g(x) \leq 0$$

If there are no terms of order 3, then we can write

$$g(x) = a_0 + \sum_{i \neq j} b_{ij} x_i x_j + \sum_{i=1}^n c_i x_i^2$$

Quadratic Case

In this case we have to deal with terms of the form $\alpha x \cdot y$ and βx^2 .

Consider $x \cdot y$ with $x \in [x^l, x^u]$ and $y \in [y^l, y^u]$. To build a relaxation of this term we introduce a new variable z and the two linear constraints

$$\begin{aligned} z &\geq y^l \cdot x + x^l \cdot y - x^l \cdot y^l \\ z &\geq y^u \cdot x + x^u \cdot y - x^u \cdot y^u \end{aligned}$$

In the case of a $-x \cdot y$ term the relaxation is

$$\begin{aligned} z &\leq y^l \cdot x + x^u \cdot y - x^u \cdot y^l \\ z &\leq y^u \cdot x + x^l \cdot y - x^l \cdot y^u \end{aligned}$$

Quadratic Case

Previous relaxations are known as McCormic convex and concave envelopes of $x \cdot y$ in $[x^l, x^u] \times [y^l, y^u]$.

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picture of relaxations of $x \cdot y$

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For the terms x^2 we have that this term is convex, so we can leave it unchanged a possible relaxation.

However if we require the relaxation to be polyhedral we can use:

$$z \geq (x^i)^2 + 2x^i \cdot (x - x^i) = -(x^i)^2 + 2x^i \cdot x, \quad i = 1, \dots, N_i$$

as an polyhedral approximation.

We note that each of the previous inequalities is simply a subgradient inequality for x^2 at a point x^i .

For $-x^2$ the best approximation we can have is the secant approximation

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formula for secant approximation

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Beyond the Quadratic Case

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mention factorable programming

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mention in some cases might be better direct approach

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Convex Relaxations for Cubic Problems

From the Factorable Programming approach we can easily get the following relaxations for terms of order 3:

To relax the terms $x \cdot y \cdot z$ we introduce the additional variables w_{xyz} , w_{xy} , w_{xz} , w_{yz} and the constraints :

$$\begin{aligned}
 w_{xyz} &\geq z^l \cdot w_{xy} + y^l \cdot w_{xz} + x^l \cdot w_{yz} \\
 &\quad - y^l \cdot z^l \cdot x - x^l \cdot z^l \cdot y - y^l \cdot z^l \cdot z + x^l \cdot y^l \cdot z^l \\
 w_{xyz} &\geq z^l \cdot w_{xy} + y^u \cdot w_{xz} + x^u \cdot w_{yz} \\
 &\quad - y^u \cdot z^l \cdot x - x^u \cdot z^l \cdot y - y^u \cdot z^l \cdot z + x^u \cdot y^u \cdot z^l \\
 w_{xyz} &\geq z^u \cdot w_{xy} + y^l \cdot w_{xz} + x^u \cdot w_{yz} \\
 &\quad - y^l \cdot z^u \cdot x - x^u \cdot z^u \cdot y - y^l \cdot z^u \cdot z + x^u \cdot y^l \cdot z^u \\
 w_{xyz} &\geq z^u \cdot w_{xy} + y^u \cdot w_{xz} + x^l \cdot w_{yz} \\
 &\quad - y^u \cdot z^u \cdot x - x^l \cdot z^u \cdot y - y^u \cdot z^u \cdot z + x^l \cdot y^u \cdot z^u \\
 z_{\gamma_1 \gamma_2} &\geq \gamma_2^l \cdot \gamma_1 + \gamma_1^l \cdot \gamma_2 - \gamma_1^l \cdot \gamma_2^l \\
 z_{\gamma_1 \gamma_2} &\geq \gamma_2^u \cdot \gamma_1 + \gamma_1^u \cdot \gamma_2 - \gamma_1^u \cdot \gamma_2^u
 \end{aligned}$$

Relaxations of Cubic Terms – Trilinear Terms