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Convex Relaxation Methods for Nonconvex
Polynomial Optimization Problems

By André A. Keller

Université des Sciences et Technologies de Lille - France



EUROPMENT



Contents

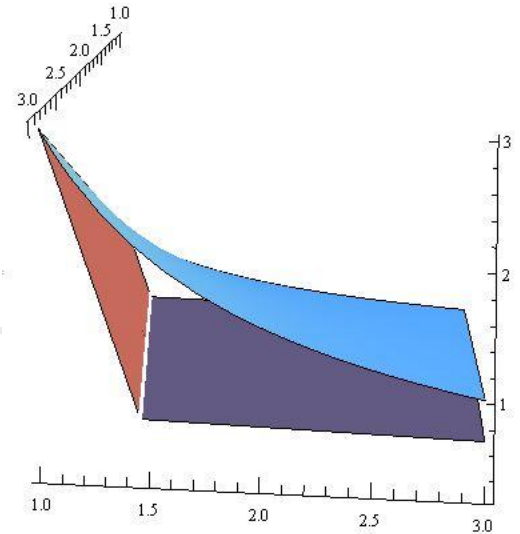
- Introduction to POPs and relaxation types
- **1.** Convexification techniques
- **2.** Reformulation-Linearization Techniques
- **3.** Semidefinite relaxations
- Conclusion on trigonometric polynomials



Introduction

Polynomial optimization problems (POPs),
types of convex relaxations

FACTORABLE RELAXATION



$$\begin{cases} z \geq \frac{x}{y} \text{ s.t.} \\ x \in [x^L, x^U], y \in [y^L, y^U] \end{cases}$$



$$\begin{cases} z \geq (xy^U - yx^L + x^L y^U) / (y^U)^2, \\ z \geq (xy^L - yx^U + x^U y^L) / (y^L)^2, \\ \text{s.t. } x \in [x^L, x^U], y \in [y^L, y^U]. \end{cases}$$



$$\begin{cases} zy \geq x \text{ s.t.} \\ z \in \left[\frac{x^L y^L}{y^U}, \frac{x^U y^U}{y^L} \right] \\ x \in [x^L, x^U], y \in [y^L, y^U] \end{cases}$$

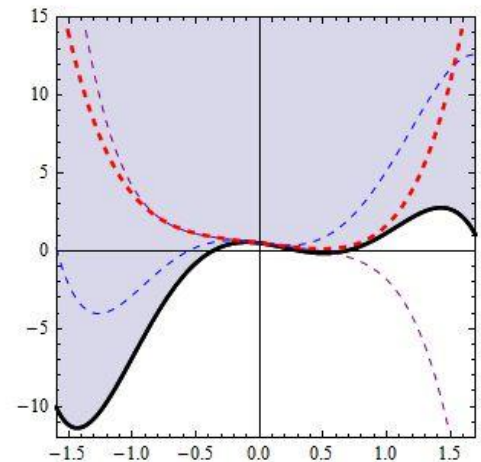
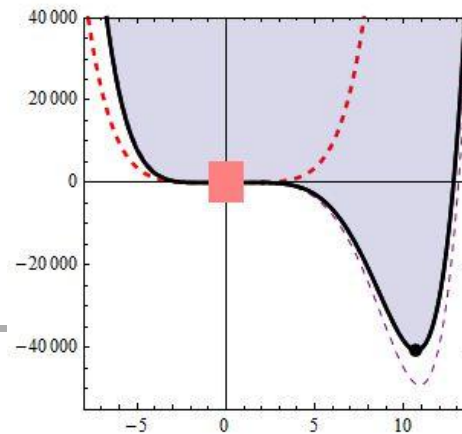


$$\begin{cases} zy - (z - z^L)(y - y^U) \geq x, \\ zy - (z - z^U)(y - y^L) \geq x, \\ \text{s.t. } x \in [x^L, x^U], y \in [y^L, y^U]. \end{cases}$$

► **Relaxation**

EXAMPLE 1

• Nonconvex POP



$$\underset{x \in \mathbb{R}}{\text{minimize}} \quad p(x) \equiv \frac{1}{2}x - 4x^2 + 7x^3 + \frac{1}{2}x^4 - 2x^5 + \frac{3}{20}x^6$$

subject to:

$$-8 \leq x \leq 14$$

► 2 constraints

$$\underset{x \in \mathbb{R}, y \in \mathbb{R}^5}{\text{minimize}} \quad q(x, y) \equiv \frac{1}{2}y_1 - 4y_2 + 7y_3 + \frac{1}{2}y_4 - 2y_5 + \frac{3}{20}x^6$$

subject to:

$$y_1 - x = 0,$$

$$y_2 - y_1x = 0,$$

$$y_3 - y_2x = 0,$$

$$y_4 - y_3x = 0,$$

$$y_5 - y_4x = 0,$$

$$-8 \leq x \leq 14,$$

► 14 constraints

$$(-8, 64, -512, 4096, -32768)^T \leq \mathbf{y} \leq (14, 196, 2744, 38416, -32768)^T$$

**Reformulated
convex POP**

• Global minimum

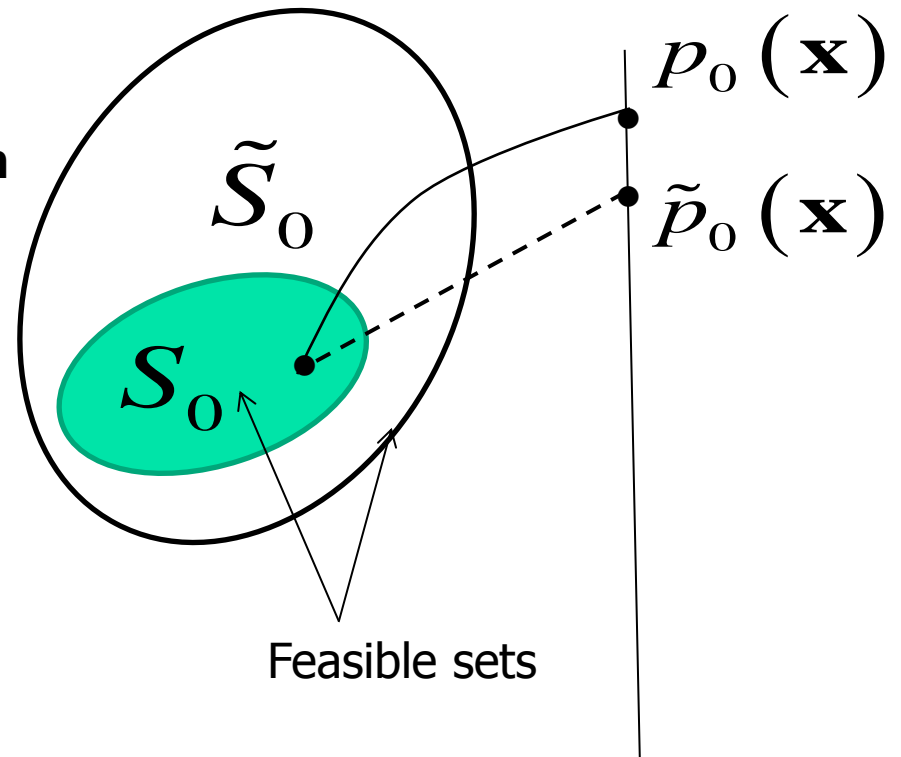
$$\hat{x} = 10.7074$$

$$q(\hat{x}, \hat{\mathbf{y}}) = p(\hat{x}) = -40,740$$

LAGRANGIAN RELAXATION

- Inequality constrained optimization

$$\left[\begin{array}{l} \mathcal{P}_0 : \underset{\mathbf{x} \in \mathbb{R}^n}{\text{minimize}} \quad p_0(\mathbf{x}) \\ \text{subject to:} \\ p_j(\mathbf{x}) \leq 0, \quad j = 1, \dots, m \end{array} \right.$$



- Lagrange relaxation

$$\tilde{\mathcal{P}}_0 : \underset{\mathbf{x} \in \mathbb{R}^n}{\text{minimize}} \quad L(\mathbf{x}, \mathbf{y}) = p_0(\mathbf{x}) + \sum_{i=1}^m y_i p_i(\mathbf{x})$$

α BB METHOD

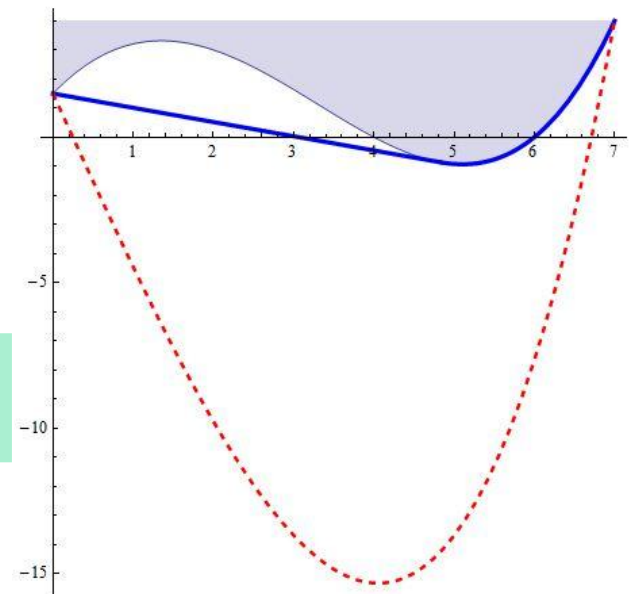
- Let a 4th degree polynomial function

$$p(x) = 1.5 + 2.8821x - 1.2774x^2 + 0.0955x^3 + 0.0051x^4$$

where $x \in [0, 7]$.

- Define $L(x) = p(x) + \alpha(x - x^L)(x - x^U)$

► make $L(x)$ convex by overpowering the nonconvexities of $p(x)$

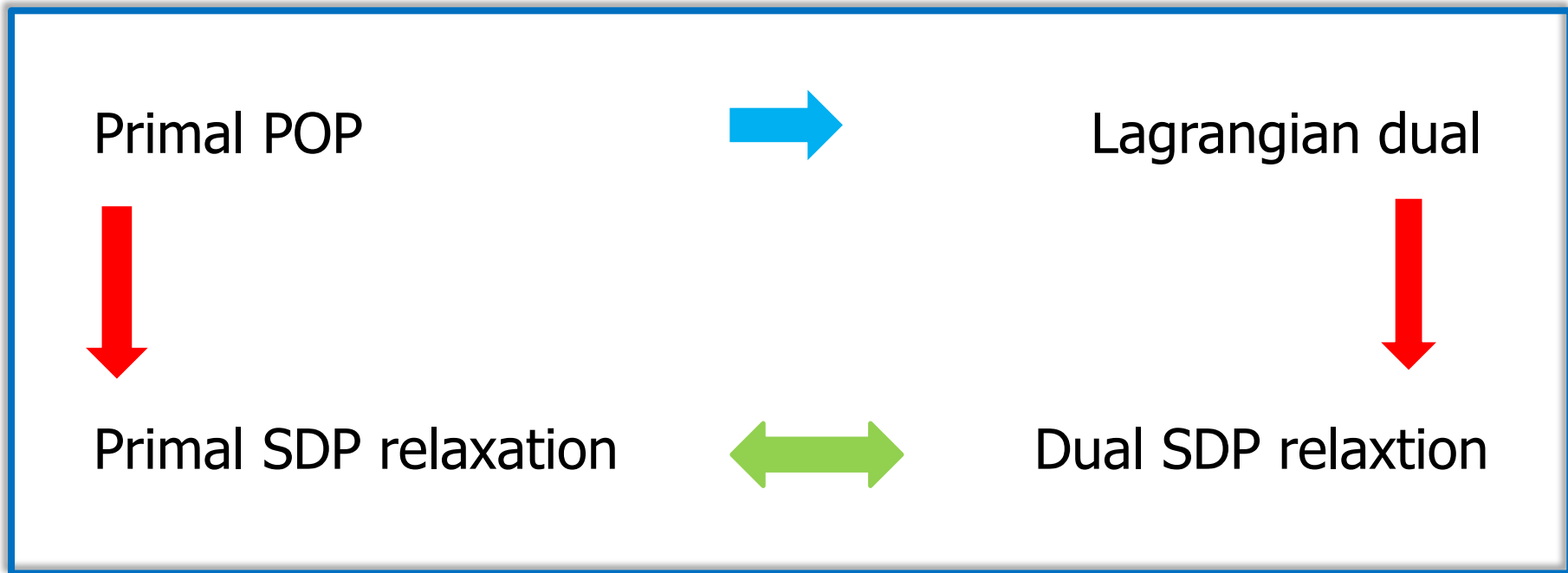




Conclusion

- Lagrange and SDP relaxations;
- polynomial/bilinear matrix inequality (PMI/BMI);
- trigonometric convex underestimators

LAGRANGE AND SDP RELAXATIONS



SINE FUNCTION: CONVEX ENVELOPE

► Convex envelope

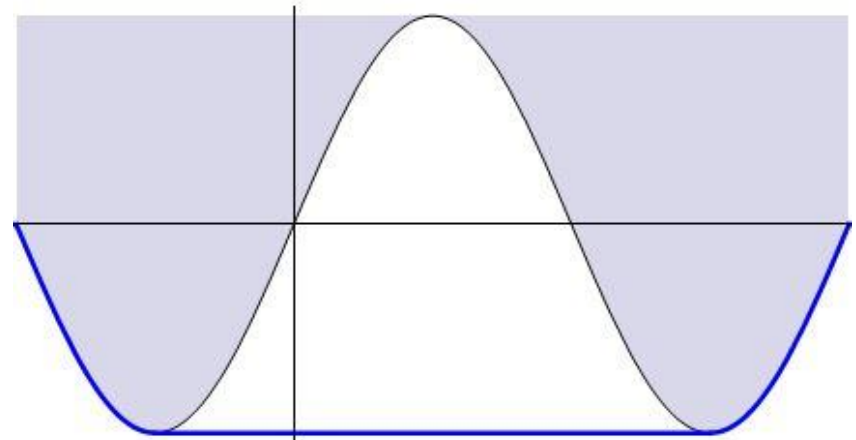
$$p_{env}(x) = \begin{cases} \sin(x), & x \in [x^L, x_{min}^1] \\ -1, & x \in [x_{min}^1, x_{min}^2] \\ \sin(x), & x \in [x_{min}^2, x^U] \end{cases}$$

2 minima which value is -1 at

and $x = 3\pi / 2$

$$x = -\pi / 2$$

$$p(x) = \sin(x) \text{ s.t. } x \in [x^L, x^U]$$



**2 global minima
which value is -1 at:**

$$x \in \{-\pi / 2, 3\pi / 2\}$$

SINE FUNCTION: TRIGONOMETRIC CONVEX UNDERESTIMATOR

$$p(x) = \sin(x + 1.5), x \in [0, 10.984]$$

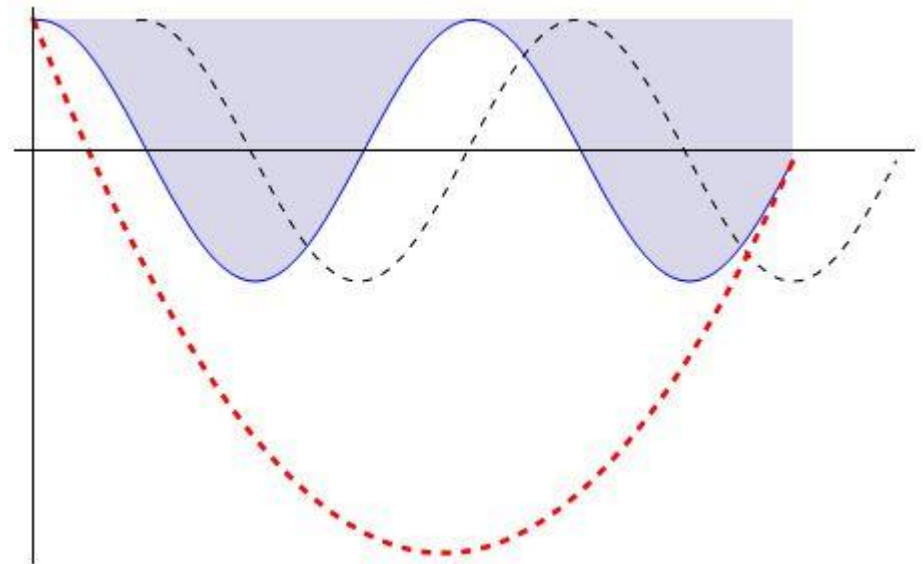
- **Origin:** Caratzoulas & Floudas (2005)

- The **functional form** for the function $p(x) = \alpha \sin(x + s)$ with $x \in [x^L, x^U]$ is

$$\phi(x) = -a \sin(k(x - x_s)) + b$$

- For **this example** we obtain

$$\phi(x) = -24.91 \sin(0.0979(x + 10.11)) + 21.83$$



GENERAL TRIGONOMETRIC FUNCTION: CONVEX TRIGO. UNDERESTIMATOR

- **Example** adapted from Caratzoulas & Floudas (2005)

- The **functional form** is

$$p(x) = \sin(x) + \sin\left(\frac{10x}{3}\right) + \log_e(x) - 0.84x$$

where $x \in [1.5, 12.484]$

- The figure shows the **translated** function $p(x + 1.5)$ on $[0, 10.984]$ and the convex trigonometric underestimator.

