

Some theorems on nonconvex optimization

Daniel Bienstock and Alexander Michalka

Columbia University

Informs 2013

1.

Cardinality constrained, convex quadratic programming.

$$\begin{aligned} \min \quad & x^T Q x + c^T x \\ \text{s.t.} \quad & A x \leq b \\ & x \geq 0, \quad \|x\|_0 \leq k \end{aligned}$$

$\|x\|_0$ = number of nonzero entries in x .

- $Q \succeq 0$
- $x \in \mathbb{R}^n$ for n possibly large
- k relatively small, e.g. $k = 100$ for $n = 10000$
- VERY hard problem – just getting good bounds is tough

2b.

Sparse vector in column space (Spielwan, Wang, Wright '12)

Given a matrix $Y \in \mathbb{R}^{n \times p}$ (n large)

$$\begin{aligned} \min \quad & \|Y - AX\|_2 \\ \text{s.t.} \quad & A \in \mathbb{R}^{n \times n}, X \in \mathbb{R}^{n \times p} \\ & X \text{ "sparse"} \end{aligned} \tag{1}$$

2b.

Sparse vector in column space (Spielwan, Wang, Wright '12)

Given a matrix $Y \in \mathbb{R}^{n \times p}$ (n large)

$$\begin{aligned} \min \quad & \|Y - AX\|_2 \\ \text{s.t.} \quad & A \in \mathbb{R}^{n \times n}, X \in \mathbb{R}^{n \times p} \\ & X \text{ "sparse"} \end{aligned} \tag{1}$$

- Both A and x are variables
- Usual “convexification” approach may not work
- Again, looks VERY hard

. AC-OPF problem in rectangular coordinates

Given a power grid, determine voltages at every node so as to minimize a convex objective

$$\begin{aligned} \min \quad & v^T A v \\ \text{s.t.} \quad & L_k \leq v^T F_k v \leq U_k, \quad k = 1, \dots, K \\ & v \in \mathbb{R}^{2n}, \quad (n = \text{number of nodes}) \end{aligned}$$

. AC-OPF problem in rectangular coordinates

Given a power grid, determine voltages at every node so as to minimize a convex objective

$$\begin{aligned} \min \quad & v^T A v \\ \text{s.t.} \quad & L_k \leq v^T F_k v \leq U_k, \quad k = 1, \dots, K \\ & v \in \mathbb{R}^{2n}, \quad (n = \text{number of nodes}) \end{aligned}$$

- voltages are complex numbers; v is the vector of voltages in rectangular coordinates (real and imaginary parts)
- $A \succeq 0$
- n could be in the tens of thousands, or more
- the F_k are very sparse (neighborhood structure for every node)
- Problem HARD when grid under distress and $L_k \approx U_k$.

Why are these problems so hard

Generic problem: $\min Q(x), \quad s.t. \quad x \in F,$

Why are these problems so hard

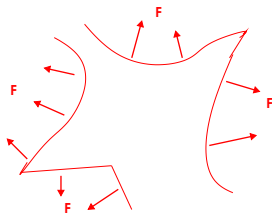
Generic problem: $\min Q(x), \quad s.t. \quad x \in F,$

- $Q(x)$ (strongly) convex, especially: positive-definite quadratic

Why are these problems so hard

Generic problem: $\min Q(x), \quad \text{s.t. } x \in F,$

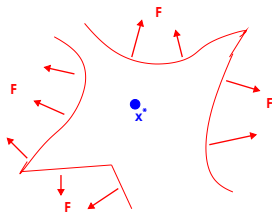
- $Q(x)$ (strongly) convex, especially: positive-definite quadratic
- F nonconvex



Why are these problems so hard

Generic problem: $\min Q(x), \quad \text{s.t. } x \in F,$

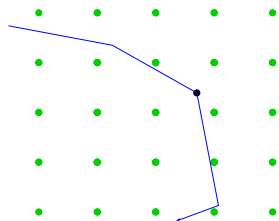
- $Q(x)$ (strongly) convex, especially: positive-definite quadratic
- F nonconvex



x^* solves $\min \left\{ Q(x), : x \in \hat{F} \right\}$ where $F \subset \hat{F}$ and \hat{F} convex

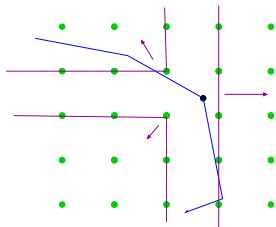
Lattice-free cuts for **linear** integer programming

Generic problem: $\min c^T x, \quad s.t. \quad Ax \leq b, \quad z \in \mathbb{Z}^n$



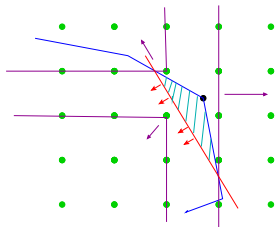
Lattice-free cuts for **linear** integer programming

Generic problem: $\min c^T x, \quad s.t. \quad Ax \leq b, \quad z \in Z^n$



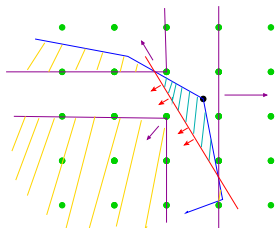
Lattice-free cuts for **linear** integer programming

Generic problem: $\min c^T x, \quad \text{s.t.} \quad Ax \leq b, \quad z \in \mathbb{Z}^n$



Lattice-free cuts for **linear** integer programming

Generic problem: $\min c^T x, \quad \text{s.t.} \quad Ax \leq b, \quad z \in \mathbb{Z}^n$



An old trick

Don't solve

$$\min Q(x), \quad \text{over } x \in F$$

An old trick

Don't solve

$$\min Q(x), \quad \text{over } x \in F$$

Do solve

$$\min z, \quad \text{over } \text{conv} \{(x, z) : z \geq Q(x), x \in F\}$$

An old trick

Don't solve

$$\min Q(x), \quad \text{over } x \in F$$

Do solve

$$\min z, \quad \text{over } \text{conv} \{(x, z) : z \geq Q(x), x \in F\}$$

- Optimal solution at **extreme point** (x^*, z^*) of $\text{conv} \{(x, z) : z \geq Q(x), x \in F\}$
- So $x^* \in F$

Exclude-and-cut

$$\min z, \quad \text{s.t.} \quad z \geq Q(x), \quad x \in F$$

0. \hat{F} : a **convex relaxation** of $\text{conv} \{(x, z) : z \geq Q(x), x \in F\}$

Exclude-and-cut

$$\min z, \quad \text{s.t. } z \geq Q(x), \quad x \in F$$

0. \hat{F} : a **convex relaxation** of $\text{conv} \{(x, z) : z \geq Q(x), x \in F\}$
1. Let $(x^*, z^*) = \text{argmin}\{z : (x, z) \in \hat{F}\}$

Exclude-and-cut

$$\min z, \quad \text{s.t. } z \geq Q(x), \quad x \in F$$

0. \hat{F} : a **convex relaxation** of $\text{conv} \{(x, z) : z \geq Q(x), x \in F\}$
1. Let $(x^*, z^*) = \text{argmin}\{z : (x, z) \in \hat{F}\}$
2. Find an **open** set S s.t. $x^* \in S$ and $S \cap F = \emptyset$.
Examples: lattice-free sets, geometry

Exclude-and-cut

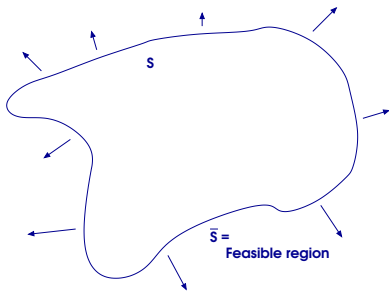
$$\min z, \quad \text{s.t. } z \geq Q(x), \quad x \in F$$

0. \hat{F} : a **convex relaxation** of $\text{conv} \{(x, z) : z \geq Q(x), x \in F\}$
1. Let $(x^*, z^*) = \text{argmin}\{z : (x, z) \in \hat{F}\}$
2. Find an **open** set S s.t. $x^* \in S$ and $S \cap F = \emptyset$.
Examples: lattice-free sets, geometry
3. Add to the formulation an inequality $\mathbf{a}z + \boldsymbol{\alpha}^T \mathbf{x} \geq \alpha_0$ valid for

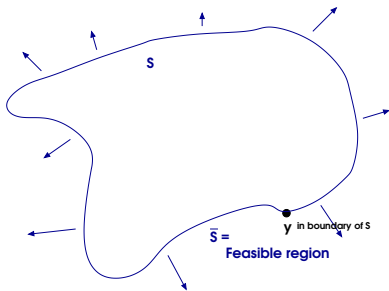
$$\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$$

but violated by (x^*, z^*) .

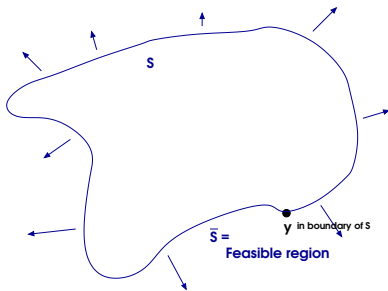
Valid **linear** inequalities for $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$.



Valid **linear** inequalities for $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$.



Valid **linear** inequalities for $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$.

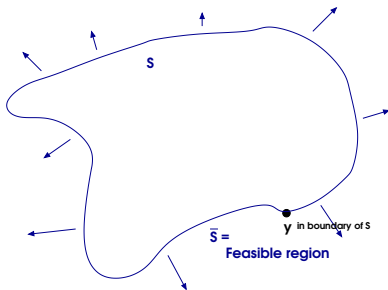


First order inequality:

$$z \geq [\nabla Q(y)]^T (x - y) + Q(y)$$

is valid EVERYWHERE

Valid **linear** inequalities for $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$.

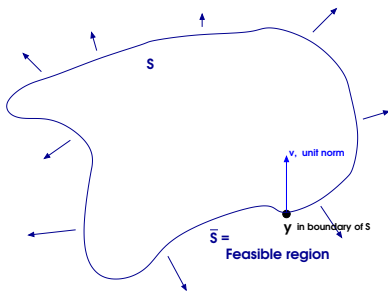


First order inequality:

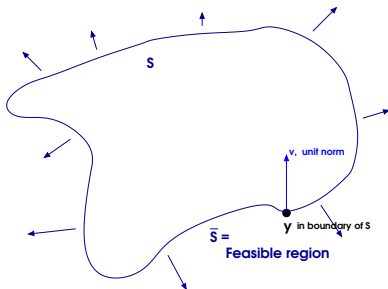
$$z \geq [\nabla Q(y)]^T (x - y) + Q(y)$$

is valid EVERYWHERE – does not cut-off any points

Valid **linear** inequalities for $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$.



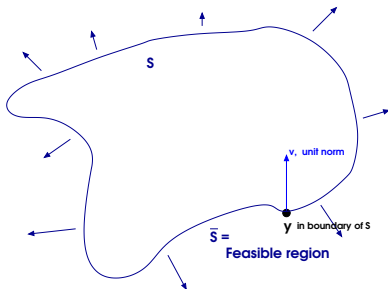
Valid **linear** inequalities for $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$.



Lifted first order inequality, for $\alpha \geq 0$:

$$z \geq \underbrace{[\nabla Q(y)]^T(x - y) + Q(y)}_{\text{first-order term} \approx Q(x)} + \underbrace{\alpha v^T(x - y)}_{\text{lifting}}$$

Valid **linear** inequalities for $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$.

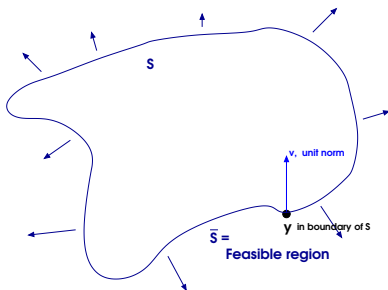


Lifted first order inequality, for $\alpha \geq 0$:

$$z \geq \underbrace{[\nabla Q(y)]^T(x - y) + Q(y)}_{\text{first-order term} \approx Q(x)} + \underbrace{\alpha v^T(x - y)}_{\text{lifting}}$$

NOT valid EVERYWHERE: $\text{RHS} > Q(x)$ for $\alpha > 0$, $v^T(x - y) > 0$ and $x \approx y$.

Valid **linear** inequalities for $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$.



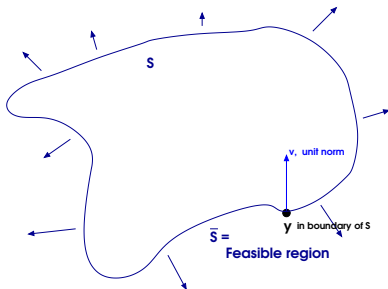
Lifted first order inequality, for $\alpha \geq 0$:

$$z \geq \underbrace{[\nabla Q(y)]^T(x - y) + Q(y)}_{\text{first-order term} \approx Q(x)} + \underbrace{\alpha v^T(x - y)}_{\text{lifting}}$$

NOT valid EVERYWHERE: $RHS > Q(x)$ for $\alpha > 0$, $v^T(x - y) > 0$ and $x \approx y$.

– want $RHS \leq Q(x)$ in \bar{S}

Valid **linear** inequalities for $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$.



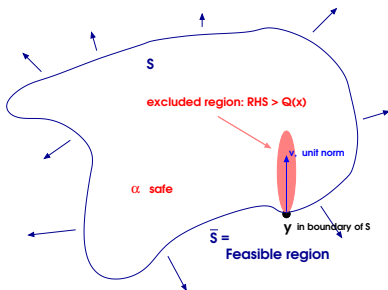
Lifted first order inequality, for $\alpha \geq 0$:

$$z \geq \underbrace{[\nabla Q(y)]^T(x - y) + Q(y)}_{\text{first-order term} \approx Q(x)} + \underbrace{\alpha v^T(x - y)}_{\text{lifting}}$$

NOT valid EVERYWHERE: $RHS > Q(x)$ for $\alpha > 0$, $v^T(x - y) > 0$ and $x \approx y$.

– want $RHS \leq Q(x)$ in \bar{S} ($\alpha = 0$ always OK)

Valid **linear** inequalities for $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$.



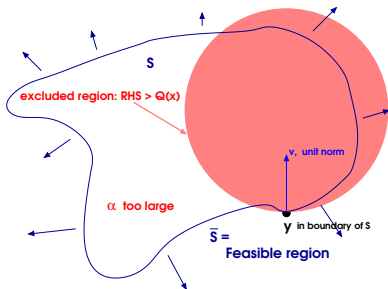
Lifted first order inequality, for $\alpha \geq 0$:

$$z \geq \underbrace{[\nabla Q(y)]^T(x - y) + Q(y)}_{\text{first-order term} \approx Q(x)} + \underbrace{\alpha v^T(x - y)}_{\text{lifting}}$$

NOT valid EVERYWHERE: $RHS > Q(x)$ for $\alpha > 0$, $v^T(x - y) > 0$ and $x \approx y$.

Want $RHS \leq Q(x)$ for $x \in \bar{S}$ ($\alpha = 0$ always OK)

Valid **linear** inequalities for $\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$.



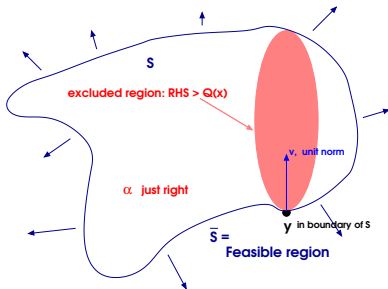
Lifted first order inequality, for $\alpha \geq 0$:

$$z \geq \underbrace{[\nabla Q(y)]^T(x - y) + Q(y)}_{\text{first-order term} \approx Q(x)} + \underbrace{\alpha v^T(x - y)}_{\text{lifting}}$$

NOT valid EVERYWHERE: $RHS > Q(x)$ for $\alpha > 0$, $v^T(x - y) > 0$ and $x \approx y$.

Want $RHS \leq Q(x)$ for $x \in \bar{S}$ ($\alpha = 0$ always OK)

Valid **linear** inequalities for $\mathcal{F} = \{ (x, z) : x \in \bar{S}, z \geq Q(x) \}$.



Lifted first order inequality, for $\alpha \geq 0$:

$$z \geq \underbrace{[\nabla Q(y)]^T(x - y) + Q(y)}_{\text{first-order term} \approx Q(x)} + \underbrace{\alpha v^T(x - y)}_{\text{lifting}}$$

NOT valid EVERYWHERE: $RHS > Q(x)$ for $\alpha > 0$, $v^T(x - y) > 0$ and $x \approx y$.

Want $RHS \leq Q(x)$ for $x \in \bar{S}$ ($\alpha = 0$ always OK)

Valid **linear** inequalities for $\mathcal{F} \doteq \{(x, z) \in \mathbb{R}^n \times \mathbb{R} : x \in \bar{S}, z \geq Q(x)\}$.

Valid **linear** inequalities for $\mathcal{F} \doteq \{(x, z) \in \mathbb{R}^n \times \mathbb{R} : x \in \bar{S}, z \geq Q(x)\}$.

Given $y \in \partial S$, let

$$\alpha^* \doteq \sup \{ \alpha \geq 0 : Q(x) \geq [\nabla Q(y)]^T(x-y) + Q(y) + \alpha v^T(x-y) \}$$

valid for \mathcal{F} .

Valid **linear** inequalities for $\mathcal{F} \doteq \{(x, z) \in \mathbb{R}^n \times \mathbb{R} : x \in \bar{S}, z \geq Q(x)\}$.

Given $y \in \partial S$, let

$$\alpha^* \doteq \mathbf{sup} \{ \alpha \geq \mathbf{0} : Q(x) \geq [\nabla Q(y)]^T(x-y) + Q(y) + \alpha v^T(x-y) \}$$

valid for \mathcal{F} . Note: $\alpha^* = \alpha^*(v, y)$

Valid **linear** inequalities for $\mathcal{F} \doteq \{ (x, z) \in \mathbb{R}^n \times \mathbb{R} : x \in \bar{S}, z \geq Q(x) \}$.

Given $y \in \partial S$, let

$$\alpha^* \doteq \sup \{ \alpha \geq 0 : Q(x) \geq [\nabla Q(y)]^T(x-y) + Q(y) + \alpha v^T(x-y) \}$$

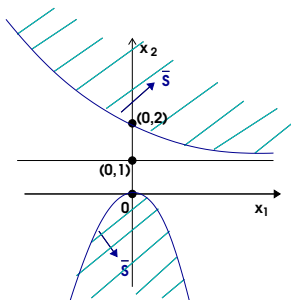
valid for \mathcal{F} . Note: $\alpha^* = \alpha^*(v, y)$

Theorem. If Q is convex and differentiable, then $\text{conv}(\mathcal{F})$ is given by

$$\begin{aligned} Q(x) &\geq [\nabla Q(y)]^T(x-y) + Q(y) && \forall y \\ Q(x) &\geq [\nabla Q(y)]^T(x-y) + Q(y) + \alpha^* v^T(x-y) \\ &&& \forall v \text{ and } y \in \partial S. \end{aligned}$$

(abridged)

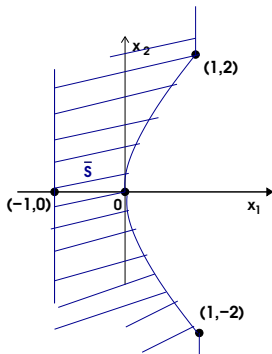
$$S = \{x \in \mathbb{R}^2 : -x_1^2 \leq x_2 \leq 1 + e^{-x_1}\}, \quad Q(x) = x_2 + e^{-x_2} - 1.$$



With $v = (0, 1)^T$, the lifted first-order inequality at $(0, 0)$ is $z \geq \alpha^* x_2$
 $\Rightarrow \alpha^* = e^{-1}$.

$$S = \{x \in \mathbb{R}^2 : x_1 \geq 1\} \cup \{x \in \mathbb{R}^2 : 0 \leq x_1 \leq 1 \mid |x_2| \leq (2x_1 - x_1^2)^{1/2} + x_1\},$$

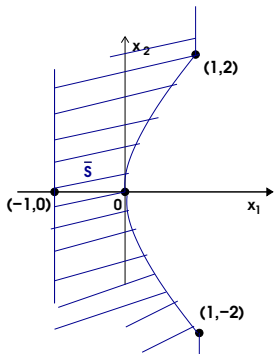
$$Q(x) = \|x\|^2$$



If $v = (1, 0)^T$, lifted first-order inequality at 0 is $z \geq \alpha^* x_1 \Rightarrow \alpha^* = 2$.

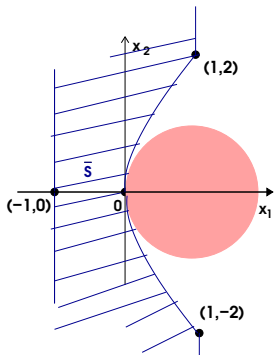
$$S = \{x \in \mathbb{R}^2 : x_1 \geq 1\} \cup \{x \in \mathbb{R}^2 : 0 \leq x_1 \leq 1 \mid |x_2| \leq (2x_1 - x_1^2)^{1/2} + x_1\},$$

$$Q(x) = \|x\|^2$$



If $v = (1, 0)^T$, lifted first-order inequality at 0 is $z \geq \alpha^* x_1 \Rightarrow \alpha^* = 2$.
 Because for $\alpha^* = 2R$, $Q(x) \leq \alpha^* x_1$ iff $|x_2| \leq (2Rx_1 - x_1^2)^{1/2}$

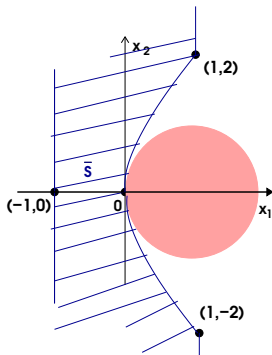
Suppose $S = \{x \in \mathbb{R}^2 : x_1 \geq 1\} \cup \{x \in \mathbb{R}^2 : 0 \leq x_1 \leq 1 \text{ and } |x_2| \leq (2x_1 - x_1^2)^{1/2} + x_1\}$, and $Q(x) = \|x\|^2$



With $v = (1, 0)^T$, the lifted first-order inequality at $(0, 0)$ is $z \geq \alpha^* x_1 \Rightarrow \alpha^* = 2$. **Why?**

Because for $\alpha^* = 2R$, $Q(x) \leq \alpha^* x_1$ iff $|x_2| \leq (2Rx_1 - x_1^2)^{1/2}$

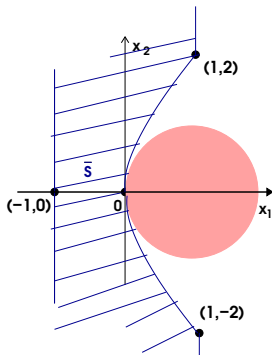
Suppose $S = \{x \in \mathbb{R}^2 : x_1 \geq 1\} \cup \{x \in \mathbb{R}^2 : 0 \leq x_1 \leq 1 \text{ and } |x_2| \leq (2x_1 - x_1^2)^{1/2} + x_1\}$, and $Q(x) = \|x\|^2$



With $v = (1, 0)^T$, the lifted first-order inequality at $(0, 0)$ is $z \geq \alpha^* x_1 \Rightarrow \alpha^* = 2$. **Why?**

Because for $\alpha^* = 2R$, $Q(x) \leq \alpha^* x_1$ iff $|x_2| \leq (2Rx_1 - x_1^2)^{1/2} \leq (2x_1 - x_1^2)^{1/2} + x_1$: true for $R \leq 1$.

Suppose $S = \{x \in \mathbb{R}^2 : x_1 \geq 1\} \cup \{x \in \mathbb{R}^2 : 0 \leq x_1 \leq 1 \text{ and } |x_2| \leq (2x_1 - x_1^2)^{1/2} + x_1\}$, and $Q(x) = \|x\|^2$



With $v = (1, 0)^T$, the lifted first-order inequality at $(0, 0)$ is $z \geq \alpha^* x_1 \Rightarrow \alpha^* = 2$. **Why?**

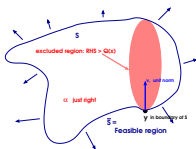
Because for $\alpha^* = 2R$, $Q(x) \leq \alpha^* x_1$ iff $|x_2| \leq (2Rx_1 - x_1^2)^{1/2} \leq (2x_1 - x_1^2)^{1/2} + x_1$: true for $R \leq 1$.

But **fails to hold** for $R > 1$ and $x_1 \approx 0$!

Lifted first-order inequality at $y \in \partial S$, in the direction of v : $Q(x) \geq [\nabla Q(y)]^T(x - y) + Q(y) + \alpha^* v^T(x - y)$



Lifted first-order inequality at $y \in \partial S$, in the direction of v : $Q(x) \geq [\nabla Q(y)]^T(x - y) + Q(y) + \alpha^* v^T(x - y)$



Theorem. If

- $Q(x)$ grows faster than linearly in every direction, and
- There is a ball with interior in the infeasible region, but containing y at its boundary

then the quantity α^* is a “max” and not just a “sup”, i.e. the lifted inequality is tight at some point other than y

Quadratics in action

Lifted first-order inequalities for $\mathcal{F} = \{(x, z) : x \in \bar{\mathcal{S}}, z \geq \|x\|^2\}$.

Separation problem. Given $(x^*, z^*) \in \mathbb{R}^n \times \mathbb{R}$, find a lifted-first order inequality maximally violated by (x^*, z^*) (if any)

Quadratics in action

Lifted first-order inequalities for $\mathcal{F} = \{(x, z) : x \in \bar{\mathcal{S}}, z \geq \|x\|^2\}$.

Separation problem. Given $(x^*, z^*) \in \mathbb{R}^n \times \mathbb{R}$, find a lifted-first order inequality maximally violated by (x^*, z^*) (if any)

Theorem: We can separate in polynomial time when:

Quadratics in action

Lifted first-order inequalities for $\mathcal{F} = \{(x, z) : x \in \bar{S}, z \geq \|x\|^2\}$.

Separation problem. Given $(x^*, z^*) \in \mathbb{R}^n \times \mathbb{R}$, find a lifted-first order inequality maximally violated by (x^*, z^*) (if any)

Theorem: We can separate in polynomial time when:

- \bar{S} (or S) is a union of polyhedra

Quadratics in action

Lifted first-order inequalities for $\mathcal{F} = \{(x, z) : x \in \bar{S}, z \geq \|x\|^2\}$.

Separation problem. Given $(x^*, z^*) \in \mathbb{R}^n \times \mathbb{R}$, find a lifted-first order inequality maximally violated by (x^*, z^*) (if any)

Theorem: We can separate in polynomial time when:

- \bar{S} (or S) is a union of polyhedra
- S is a convex ellipsoid or paraboloid (many cases)

Special case: complement of an ellipsoid

Lifted first-order inequalities for

$$\mathcal{F} = \{(x, z) : x^T A x - 2b^T x + c \geq 0, z \geq \|x\|^2\}. \quad \text{Here, } A \succ 0.$$

Special case: complement of an ellipsoid

Lifted first-order inequalities for

$$\mathcal{F} = \{(x, z) : x^T A x - 2b^T x + c \geq 0, z \geq \|x\|^2\}. \quad \text{Here, } A \succ 0.$$

Let $\lambda =$ **largest** eigenvalue of A . Then:

Special case: complement of an ellipsoid

Lifted first-order inequalities for

$$\mathcal{F} = \{(x, z) : x^T A x - 2b^T x + c \geq 0, z \geq \|x\|^2\}. \quad \text{Here, } A \succ 0.$$

Let $\lambda = \text{largest}$ eigenvalue of A . Then:

Theorem. The **strongest** lifted first-order inequality at $\bar{x} \in \mathbb{R}^n$ is:

$$z \geq 2[(I - \lambda^{-1}A)\bar{x} + \lambda^{-1}b]^T(x - \bar{x}) + \bar{x}(I - \lambda^{-1}A)\bar{x} + 2\lambda^{-1}b^T\bar{x} - \lambda^{-1}c$$

Special case: complement of an ellipsoid

Lifted first-order inequalities for

$$\mathcal{F} = \{(x, z) : x^T A x - 2b^T x + c \geq 0, z \geq \|x\|^2\}. \quad \text{Here, } A \succ 0.$$

Let $\lambda =$ **largest** eigenvalue of A . Then:

Theorem. The **strongest** lifted first-order inequality at $\bar{x} \in \mathbb{R}^n$ is:

$$z \geq 2[(I - \lambda^{-1}A)\bar{x} + \lambda^{-1}b]^T(x - \bar{x}) + \bar{x}(I - \lambda^{-1}A)\bar{x} + 2\lambda^{-1}b^T\bar{x} - \lambda^{-1}c$$

The right-hand side is the **first-order** (tangent), at \bar{x} , for the convex quadratic

$$x(I - \lambda^{-1}A)x + 2\lambda^{-1}b^T x - \lambda^{-1}c.$$

Special case: complement of an ellipsoid

Lifted first-order inequalities for

$$\mathcal{F} = \{(x, z) : x^T A x - 2b^T x + c \geq 0, z \geq \|x\|^2\}. \quad \text{Here, } A \succ 0.$$

Let $\lambda = \mathbf{largest}$ eigenvalue of A . Then:

Theorem. The **strongest** lifted first-order inequality at $\bar{x} \in \mathbb{R}^n$ is:

$$z \geq 2[(I - \lambda^{-1}A)\bar{x} + \lambda^{-1}b]^T(x - \bar{x}) + \bar{x}(I - \lambda^{-1}A)\bar{x} + 2\lambda^{-1}b^T\bar{x} - \lambda^{-1}c$$

The right-hand side is the **first-order** (tangent), at \bar{x} , for the convex quadratic

$$x(I - \lambda^{-1}A)x + 2\lambda^{-1}b^T x - \lambda^{-1}c.$$

Corollary:

$$\text{conv}(\mathcal{F}) = \{(x, z) : z \geq x(I - \lambda^{-1}A)x + 2\lambda^{-1}b^T x - \lambda^{-1}c, z \geq \|x\|^2\}.$$

Obtained by Modaresi and Vielma (2013)

But ... Exclude-and-cut, again

$$\min z, \quad \text{s.t.} \quad z \geq Q(x), \quad x \in F$$

0. \hat{F} : a **convex relaxation** of $\text{conv} \{(x, z) : z \geq Q(x), x \in F\}$

But ... Exclude-and-cut, again

$$\min z, \quad \text{s.t. } z \geq Q(x), \quad x \in F$$

0. \hat{F} : a **convex relaxation** of $\text{conv} \{(x, z) : z \geq Q(x), x \in F\}$
1. Let $(x^*, z^*) = \text{argmin}\{z : (x, z) \in \hat{F}\}$

But ... Exclude-and-cut, again

$$\min z, \quad \text{s.t. } z \geq Q(x), \quad x \in F$$

0. \hat{F} : a **convex relaxation** of $\text{conv} \{(x, z) : z \geq Q(x), x \in F\}$
1. Let $(x^*, z^*) = \text{argmin}\{z : (x, z) \in \hat{F}\}$
2. Find an **open** set S s.t. $x^* \in S$ and $S \cap F = \emptyset$.
Examples: lattice-free sets, geometry

But ... Exclude-and-cut, again

$$\min z, \quad \text{s.t.} \quad z \geq Q(x), \quad x \in F$$

0. \hat{F} : a **convex relaxation** of $\text{conv} \{(x, z) : z \geq Q(x), x \in F\}$
1. Let $(x^*, z^*) = \text{argmin}\{z : (x, z) \in \hat{F}\}$
2. Find an **open** set S s.t. $x^* \in S$ and $S \cap F = \emptyset$.
Examples: lattice-free sets, geometry
3. Add to the formulation an inequality $\mathbf{a}z + \boldsymbol{\alpha}^T \mathbf{x} \geq \alpha_0$ valid for

$$\{(x, z) : x \in \bar{S}, z \geq Q(x)\}$$

but violated by (x^*, z^*) .

A classical problem: the trust-region subproblem

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad x \in \mathbb{R}^n \end{aligned}$$

A classical problem: the trust-region subproblem

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad x \in \mathbb{R}^n \end{aligned}$$

- A a **general** quadratic

A classical problem: the trust-region subproblem

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad x \in \mathbb{R}^n \end{aligned}$$

- A a **general** quadratic
- Polynomial-time solvable!

A classical problem: the trust-region subproblem

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad x \in \mathbb{R}^n \end{aligned}$$

- A a **general** quadratic
- Polynomial-time solvable! e.g. S-Lemma

A classical problem: the trust-region subproblem

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad x \in \mathbb{R}^n \end{aligned}$$

- A a **general** quadratic
- Polynomial-time solvable! e.g. S-Lemma

Sturm and Zhang (2000): two extensions are polynomially solvable:

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad \|x - x^0\|^2 \leq r \end{aligned}$$

(one additional ball inequality), and

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad c^T x \leq c^0 \end{aligned}$$

(one added linear inequality).

A classical problem: the trust-region subproblem

Ye and Zhang (2003): two **parallel** linear inequalities are added:

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad d^0 \leq c^T x \leq c^0 \end{aligned}$$

A classical problem: the trust-region subproblem

Ye and Zhang (2003): two **parallel** linear inequalities are added:

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad d^0 \leq c^T x \leq c^0 \end{aligned}$$

→ Adding a system $Ax \leq b$ makes the problem NP-hard

A classical problem: the trust-region subproblem

Ye and Zhang (2003): two **parallel** linear inequalities are added:

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad d^0 \leq c^T x \leq c^0 \end{aligned}$$

→ Adding a system $Ax \leq b$ makes the problem NP-hard

Anstreicher and Burer (2012): Ye-Zhang case formulated as convex program

A classical problem: the trust-region subproblem

Ye and Zhang (2003): two **parallel** linear inequalities are added:

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad d^0 \leq c^T x \leq c^0 \end{aligned}$$

→ Adding a system $Ax \leq b$ makes the problem NP-hard

Anstreicher and Burer (2012): Ye-Zhang case formulated as convex program

Burer and Yang (2013)

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1 \\ & a_i^T x \leq b_i \quad i = 1, \dots, m \end{aligned}$$

poly-time solvable if no two linear constraints intersect within unit ball

A classical problem: the trust-region subproblem

Ye and Zhang (2003): two **parallel** linear inequalities are added:

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1, \quad d^0 \leq c^T x \leq c^0 \end{aligned}$$

→ Adding a system $Ax \leq b$ makes the problem NP-hard

Anstreicher and Burer (2012): Ye-Zhang case formulated as convex program

Burer and Yang (2013)

$$\begin{aligned} \min \quad & x^T A x + b^T x + c \\ \text{s.t.} \quad & \|x\|^2 \leq 1 \\ & a_i^T x \leq b_i \quad i = 1, \dots, m \end{aligned}$$

poly-time solvable if no two linear constraints intersect within unit ball

$$\forall i \neq j, \quad \{x : a_i^T x = b_i\} \cap \{x : a_j^T x = b_j\} \cap \{x : \|x\|^2 \leq 1\} = \emptyset$$

A generalization

$$\begin{aligned} \text{(TLIN):} \quad & \min \quad x^T A x + b^T x + c \\ & \text{s.t.} \quad \|x\|^2 \leq 1 \\ & \quad \quad a_i^T x \leq b_i \quad i = 1, \dots, m \\ & \quad \quad x \in \mathbb{R}^n. \end{aligned}$$

- $P = \{x : a_i^T x \leq b_i \quad i = 1, \dots, m\}$

A generalization

$$\begin{aligned} \text{(TLIN):} \quad & \min \quad x^T A x + b^T x + c \\ & \text{s.t.} \quad \|x\|^2 \leq 1 \\ & \quad \quad a_i^T x \leq b_i \quad i = 1, \dots, m \\ & \quad \quad x \in \mathbb{R}^n. \end{aligned}$$

- $P = \{x : a_i^T x \leq b_i \quad i = 1, \dots, m\}$
- F^* = the number of **faces** of P that intersect the unit ball

A generalization

$$\begin{aligned} \text{(TLIN):} \quad & \min \quad x^T A x + b^T x + c \\ & \text{s.t.} \quad \|x\|^2 \leq 1 \\ & \quad \quad a_i^T x \leq b_i \quad i = 1, \dots, m \\ & \quad \quad x \in \mathbb{R}^n. \end{aligned}$$

- $P = \{x : a_i^T x \leq b_i \quad i = 1, \dots, m\}$
- F^* = the number of **faces** of P that intersect the unit ball
- Ye-Zhang (or Anstreicher-Burer) case: $F^* = 3$.
- Burer-Yang case: $F^* = m + 1$

A generalization

$$\begin{aligned} \text{(TLIN):} \quad & \min \quad x^T A x + b^T x + c \\ & \text{s.t.} \quad \|x\|^2 \leq 1 \\ & \quad \quad a_i^T x \leq b_i \quad i = 1, \dots, m \\ & \quad \quad x \in \mathbb{R}^n. \end{aligned}$$

- $P = \{x : a_i^T x \leq b_i \quad i = 1, \dots, m\}$
- F^* = the number of **faces** of P that intersect the unit ball
- Ye-Zhang (or Anstreicher-Burer) case: $F^* = 3$.
- Burer-Yang case: $F^* = m + 1$

Theorem: Problem **TLIN** can be solved in time polynomial in the problem size and F^* .

A stronger generalization

$$\begin{aligned} \text{TGEN}(S,K): \quad & \min \quad x^T A x + b^T x + c \\ & \text{s.t.} \quad \|x - x^k\|^2 \leq f_k \quad k \in S \\ & \quad \quad \|x - y^k\|^2 \geq g_k \quad k \in K \\ & \quad \quad a_i^T x \leq b_i \quad i = 1, \dots, m \end{aligned}$$

A stronger generalization

$$\begin{aligned} \mathbf{TGEN(S,K):} \quad & \min \quad x^T A x + b^T x + c \\ & \text{s.t.} \quad \|x - x^k\|^2 \leq f_k \quad k \in S \\ & \quad \quad \|x - y^k\|^2 \geq g_k \quad k \in K \\ & \quad \quad a_i^T x \leq b_i \quad i = 1, \dots, m \end{aligned}$$

$$P = \{x : a_i^T x \leq b_i \quad i = 1, \dots, m\}.$$

Theorem:

1. For every **fixed** $|S| \geq 1, |K| \geq 0$, problem **TGEN(S,K)** can be solved in time polynomial in the problem size and F^* .

F^* = number of **faces** of P intersecting $\bigcap_{k \in S} \{x : \|x - x^k\|^2 \leq f_k\}$.

A stronger generalization

$$\begin{aligned} \mathbf{TGEN(S,K):} \quad & \min \quad x^T A x + b^T x + c \\ & \text{s.t.} \quad \|x - x^k\|^2 \leq f_k \quad k \in S \\ & \quad \quad \|x - y^k\|^2 \geq g_k \quad k \in K \\ & \quad \quad a_i^T x \leq b_i \quad i = 1, \dots, m \end{aligned}$$

$$P = \{x : a_i^T x \leq b_i \quad i = 1, \dots, m\}.$$

Theorem:

1. For every **fixed** $|S| \geq 1, |K| \geq 0$, problem **TGEN(S,K)** can be solved in time polynomial in the problem size and F^* .

$$F^* = \text{number of faces of } P \text{ intersecting } \bigcap_{k \in S} \{x : \|x - x^k\|^2 \leq f_k\}.$$

2. For every **fixed** $|K| \geq 0$, and $m \geq 0$, problem **TGEN(\emptyset, K)** polytime solvable.

A stronger generalization

$$\begin{aligned} \mathbf{TGEN}(S,K): \quad & \min \quad x^T A x + b^T x + c \\ & \text{s.t.} \quad \|x - x^k\|^2 \leq f_k \quad k \in S \\ & \quad \quad \|x - y^k\|^2 \geq g_k \quad k \in K \\ & \quad \quad a_i^T x \leq b_i \quad i = 1, \dots, m \end{aligned}$$

$$P = \{x : a_i^T x \leq b_i \quad i = 1, \dots, m\}.$$

Theorem:

1. For every **fixed** $|S| \geq 1, |K| \geq 0$, problem **TGEN(S,K)** can be solved in time polynomial in the problem size and F^* .

$$F^* = \text{number of faces of } P \text{ intersecting } \bigcap_{k \in S} \{x : \|x - x^k\|^2 \leq f_k\}.$$

2. For every **fixed** $|K| \geq 0$, and $m \geq 0$, problem **TGEN(\emptyset, K)** polytime solvable.

(SODA 2014)