

Power flow feasibility problem

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The problem

- ① Large-scale power grids
- ② Need to know if a regime of the grid is «normal», «safe»
- ③ Ohm's law \Rightarrow quadratic equations:

$$y_i = f_i(x) = x^T A_i x + 2b_i^T x$$

y (regime) known, x is not

- ④ Need to determine if $\exists x: y = f(x)$ (means «safe»)

This problem is known as Power Flow Feasibility problem.
To solve it, the image $f(\mathbb{R}^n)$ must be examined

- ① A. Dymarsky, Convexity of a Small Ball Under Quadratic Map, arXiv:1410.1553
- ② A. Dymarsky, On the Convexity of Image of a Multidimensional Quadratic Map, arXiv:1410.2254

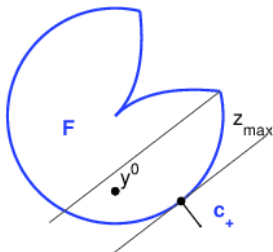
The solution

Given: the map $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$, $f_i(x) = x^T A_i x + 2b_i^T x$, $A_i^T = A_i$

Proposed algorithm for examining $F = f(\mathbb{R}^n)$:

- **Input:** $y^0 \in F$, direction c_+ : $c_+ \cdot A > 0$
- **Output:** value z_{\max} s.t. the cut $Q(c_+, z_{\max}, F)$ is convex

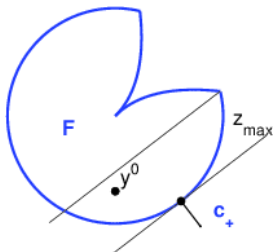
$$Q(c_+, z, F) = \{y \mid (y - y^*, c_+) \in [0, z]\} \cap F$$



The solution

The idea of the algorithm:

- 1 y^* — touching point of hyperplane c_+
- 2 Discovering boundary nonconvexities $\{F_i\}$ close to y^0
- 3 Projecting to c_+ : (F_i, c_+)
- 4 Calculating $z_{\max} = \inf_i \inf_{y \in F_i} (c_+, y)$



Boundary points of F on supporting hyperplane c :

$$\partial F_c = f(\arg \min_{x \in \mathbb{R}^n} (c, f(x))) = f(\text{Ker}(c \cdot A) - (c \cdot A)^g(c \cdot b))$$

$$\text{if } \begin{cases} c \cdot A \geq 0 \\ (c \cdot b)^T \text{Ker}(c \cdot A) = 0 \end{cases}, \text{ otherwise } \partial F_c = \emptyset$$

- ∂F_c is nonconvex $\Rightarrow F$ has nonconvexity
- ∂F_c is nonconvex $\Leftrightarrow^{(*)}$ Rank($c \cdot A$) = $n - 1$, $c \cdot A \geq 0$

Therefore,

$$z_{\max} = \inf_c \inf_{y \in \partial F_c} (c_+, \partial F_c)$$

(*) We assume Rank($c \cdot A$) < $n - 1$ to be a rare case. Condition Rank($c \cdot A$) = $n - 1$ is associated with nonconvexity of ∂F_c

Linear change of basis s.t. $\{c_+ \cdot A = I, c_+ \cdot b = 0 \Rightarrow$

- $\inf_{y \in \partial F_c} (c_+, \partial F_c) = \|(c \cdot A)^g(c \cdot b)\|^2$ for c : $\text{Rank}(c \cdot A) = n - 1$
- Adding γc_+ to c to ensure $\lambda_{\min}((c + \gamma c_+) \cdot A) = 0$
 $c_+ \cdot A = I$ by our choice of variables
- Define $z(c) = \|(c \cdot A - \lambda_{\min}(c \cdot A))^g(c \cdot b)\|^2$
- Define $c_{\text{bad}} = \{c \mid \text{Ker}(c \cdot A) \perp (c \cdot b)\}$
 $c \in c_{\text{bad}} \Leftrightarrow z(c)$ has its original meaning, useless otherwise
- Then

$$z_{\max} = \inf_{c \in c_{\text{bad}}} z(c)$$

- In general case $|c_{\text{bad}}|$ is continuum

We use gradient projection method to find

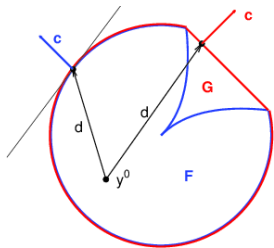
$$z_{\max} = \inf_{c \in c_{\text{bad}}} z(c)$$

- ➊ **Input:** start point obtained via *nonconvexity certificate*

$$c^0 \in c_{\text{bad}}$$

- ➋ Gradient $\frac{dz}{dc}$ is calculated explicitly
- ➌ Normal vector n for c_{bad}
- ➍ Projection of c' onto c_{bad} is done by adjusting λ in $c' + \lambda n$
- ➎ Repeat until $\frac{dz}{dc} \parallel n$
- ➏ **Output:** minimal value $z_{\max} = z(c^*)$

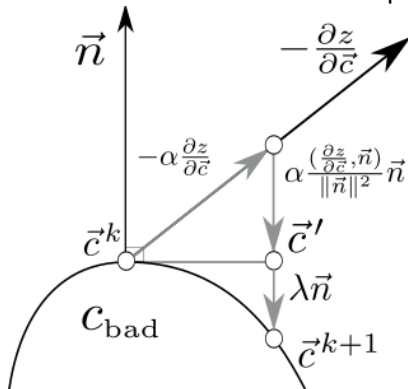
Nonconvexity certificate:



- 1 **Input:** y^0
- 2 Generating random directions d
- 3 Find $t: y^0 + td \in \partial \text{conv } F$
- 4 $y^0 + td \in F?$
- 5 If not, obtain c via dual problem
- 6 **Output:** «nonconvex» $c \in \mathcal{C}_{\text{bad}}$

\Rightarrow Obtained start point for gradient descent

Gradient step + projection



- 1 **Input:** current point $c = c^k \in c_{\text{bad}}$
- 2 Calculate $n(c)$, $\frac{dz}{dc}$,
 $c' = c - \alpha \left(\hat{1} - \frac{(\cdot, n)}{(n, n)} n \right) \frac{dz}{dc}$
- 3 Project $c' \notin c_{\text{bad}}$ onto c_{bad}
- 4 **Output:** $c^{k+1} \leftarrow$ result of projection

Projection

Given: point $c' \notin c_{\text{bad}}$, normal vector n

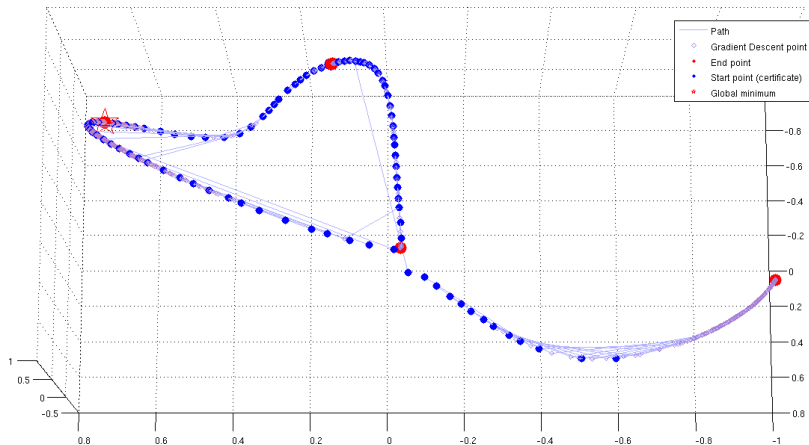
Find: point $c \in c_{\text{bad}}$ close to c'

- ① $c(\lambda) = c' + \lambda n$
- ② For some λ , $c(\lambda) \in c_{\text{bad}}$
- ③ $c(\lambda) \in c_{\text{bad}} \Leftrightarrow \text{Ker}(c(\lambda) \cdot A) \perp c(\lambda) \cdot b$
- ④ $m(\lambda) = (c(\lambda) \cdot b)^T x_0(\lambda)$, $x_0(\lambda) \in \text{Ker}(c(\lambda) \cdot A)$
- ⑤ Projection is done by finding root of $m(\lambda)$
- ⑥ Bisection method is used

Numerical experiment

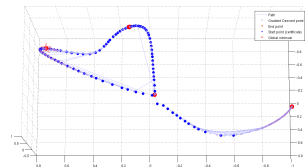
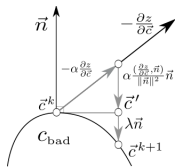
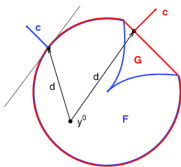
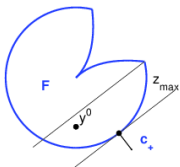
$$f: \mathbb{R}^4 \rightarrow \mathbb{R}^4$$

- 4 local minima
- Global minimum found



Algorithm for examining the set of «safe» regimes was proposed:

- ① Can determine if the whole region is «safe» at one run
- ② Cuts convex parts of the image F
- ③ General case when the set of nonconvexities is a continuum was considered
- ④ Algorithm was tested on a number of maps f



Thank you!
Questions?