

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

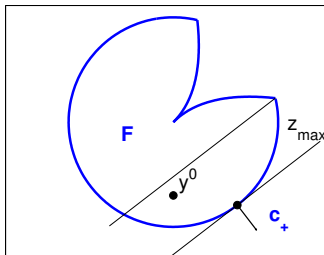
- ① A. Dymarsky, Convexity of a Small Ball Under Quadratic Map, arXiv:1410.1553
- ② A. Dymarsky, On the Convexity of Image of a Multidimensional Quadratic Map, arXiv:1410.2254

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 $F = f(\mathbb{R}^n)$:

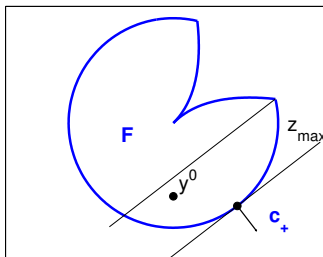
- Input: $y^0 \in F$, direction c_+ : $c_+ \cdot A > 0$
- Output: value z_{\max} s.t. the cut $Q(c_+, z_{\max}, F)$ is convex

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



The solution

- 1 Discovering boundary nonconvexities $\{F_i\}$ close to y^0
- 2 Projecting to c_+ : (F_i, c_+)
- 3 Calculating $z_{\max} = \min_{y \in F_i} (c_+, y)$



Boundary points of F on supporting hyperplane c :

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

$$\begin{cases} c \cdot A \geq 0 \\ (c \cdot b)^T \text{Ker}(c \cdot A) = 0 \end{cases}$$

or $\partial F_c = \emptyset$

- ∂F_c is nonconvex $\Rightarrow F$ has nonconvexity
- ∂F_c is nonconvex $\Leftrightarrow \text{Rg}(c \cdot A) < n, c \cdot A \geq 0$

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

We assume $\text{Rg}(c \cdot A) < n - 1$ to be a rare case

Linear change of basis s.t. $\begin{cases} c_+ \cdot A = I \\ c_+ \cdot b = 0 \end{cases} \Rightarrow$

- $\inf_{y \in \partial F_c} (c_+, \partial F_c) = \|(c \cdot A)^g (c \cdot b)\|^2$ for $c: \text{Rg}(c \cdot A) = n - 1$
- Define $z(c) = \|(c \cdot A - \lambda_{\min}(c \cdot A))^g (c \cdot b)\|^2$
- Define $c_{\text{bad}} = \{c \mid \text{Ker}(c \cdot A) \perp (c \cdot b)\}$
- Then $\boxed{z_{\max} = \inf_{c \in c_{\text{bad}}} z(c)}.$

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

We use gradient projection method to find

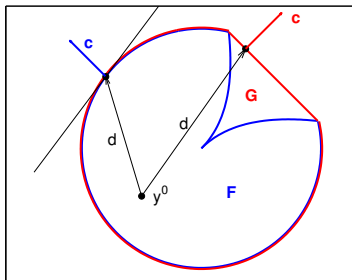
$$z_{\max} = \inf_{c \in c_{\text{bad}}} z(c)$$

- 1 Start point: obtained via *nonconvexity certificate*
- 2 Gradient $\frac{dz}{dc}$ is calculated explicitly
- 3 Normal vector n for c_{bad}
- 4 Projection onto c_{bad} is done by adjusting λ in $c' + \lambda n$

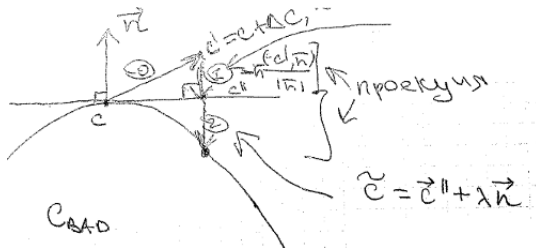
The solution

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
- \Rightarrow Obtained start point for gradient descent



The solution



Gradient step + projection

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

The solution

Projection

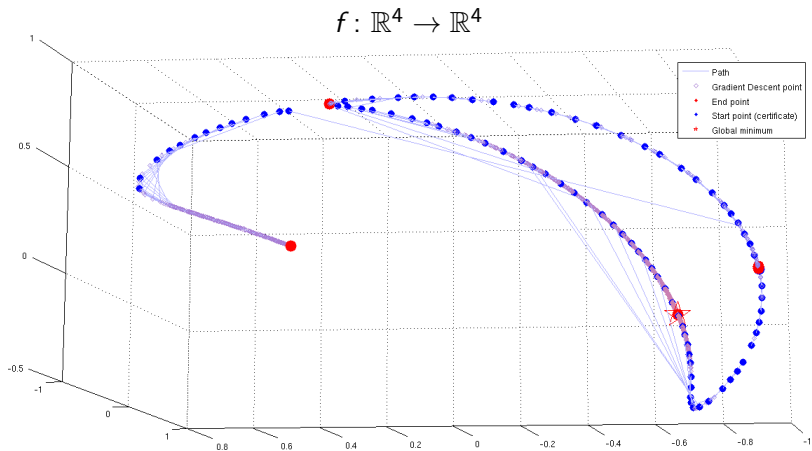
Given: point $c' \notin c_{\text{bad}}$, normal vector n

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 $m(\lambda) = (c(\lambda) \cdot b)^T x_0(\lambda)$, $x_0(\lambda) \in \text{Ker}(c(\lambda) \cdot A)$

Projection is done by finding root of $m(\lambda)$

Numerical experiment



- 1 Algorithm for examining F was proposed
- 2 Cases when $|c_{\text{bad}}| = \infty$ and $|c_{\text{bad}}| < \infty$ were considered
- 3 Algorithm for $|c_{\text{bad}}| = \infty$ was implemented
- 4 Algorithm was tested on a number of maps f

Thank you!
Questions?