

Perturbation analysis of matrix optimization

Chao Ding

Institute of Applied Mathematics

Academy of Mathematics and Systems Science, CAS

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中国科学院
CHINESE ACADEMY OF SCIENCES

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- Nonsmooth composite matrix optimizations: strong regularity, constraint nondegeneracy and beyond, *arXiv:1907.13253* (July, 2019).



Nonsmooth Composite Matrix Optimization Problem

CMatOP:

$$\begin{array}{ll} \underset{\mathbf{x} \in \mathbb{X}}{\text{minimize}} & \Phi(\mathbf{x}) \triangleq f(\mathbf{x}) + \phi \circ \lambda(g(\mathbf{x})) \\ \text{subject to} & h(\mathbf{x}) = 0, \end{array}$$

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- ★ We focus on the symmetric case just for **simplicity**;
- ★ The obtained results can be extended to **non-symmetric** cases;
- ★ This is a general model which includes many “**non-polyhedral**” OPs: **SDP**, **Eigenvalue optimization**, etc

More applications

- Fastest mixing Markov chain problem (fast load balancing of paralleled systems)
- Fastest distributed linear averaging problem
- The reduced rank approximations of transition matrices
- The low rank approximations of doubly stochastic matrices
- Low-rank approximation of matrices with linear structures
- Unsupervised learning
-

Spectral functions

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- $\phi : \mathbb{R}^n \rightarrow (-\infty, +\infty]$ is a **symmetric convex piecewise linear function**
- a **convex piecewise linear function**: a **polyhedral convex function** (Rockafellar, 1970)

Convex piecewise linear functions

Theorem (Rockafellar & Wets, 1998)

ϕ can be expressed in the form of

$$\phi(\mathbf{x}) = \phi_1(\mathbf{x}) + \phi_2(\mathbf{x}), \quad \mathbf{x} \in \mathbb{R}^n,$$

with $\phi_1 : \mathbb{R}^n \rightarrow \mathbb{R}$ and $\phi_2 : \mathbb{R}^n \rightarrow (-\infty, +\infty]$ are defined by

$$\phi_1(\mathbf{x}) := \max_{1 \leq i \leq p} \{ \langle \mathbf{a}^i, \mathbf{x} \rangle - c^i \} \quad \text{and} \quad \phi_2(\mathbf{x}) := \delta_{\text{dom } \phi}(\mathbf{x}),$$

- $\mathbf{a}^1, \dots, \mathbf{a}^p \in \mathbb{R}^n, c^1, \dots, c^p \in \mathbb{R}$ with some positive integer $p \geq 1$;
- $\text{dom } \phi$ is a polyhedral set:

$$\text{dom } \phi := \left\{ x \in \mathbb{R}^n \mid \max_{1 \leq i \leq q} \{ \langle \mathbf{b}^i, \mathbf{x} \rangle - d^i \} \leq 0 \right\}$$

- $\mathbf{b}^1, \dots, \mathbf{b}^q \in \mathbb{R}^n$ and $d^1, \dots, d^q \in \mathbb{R}$ for some positive integer $q \geq 1$.

Examples

SDP:

$$\mathbb{S}_-^n = \{X \in \mathbb{S}^n \mid \lambda_{\max}(X) \leq 0\} = \{X \in \mathbb{S}^n \mid \max_{1 \leq i \leq n} \{\langle \mathbf{e}^i, \lambda(X) \rangle\} \leq 0\}$$

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Eigenvalue optimizations:

$$s_k(X) = \sum_{i=1}^k \lambda_i(X) = \max_{1 \leq i \leq p} \{\langle \mathbf{a}^i, \lambda(X) \rangle\}$$

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- $\mathbf{a}^i \in \mathbb{R}^n$: the vector contains k ones and $n - k$ zeros

Perturbation analysis of CMatOPs

Canonically perturbed CMatOPs with parameters $(\mathbf{a}, \mathbf{b}, \mathbf{c}) \in \mathbb{X} \times \mathbb{Y} \times \mathbb{S}^n$:

$$\begin{array}{ll} \underset{\mathbf{x} \in \mathbb{X}}{\text{minimize}} & f(\mathbf{x}) - \langle \mathbf{a}, \mathbf{x} \rangle + \phi \circ \lambda(g(\mathbf{x}) + \mathbf{c}) \\ \text{subject to} & h(\mathbf{x}) + \mathbf{b} = 0 \end{array}$$

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The **Karush-Kuhn-Tucker** (KKT) optimality condition for perturbed problem takes the following form:

$$\begin{cases} \mathbf{a} = \nabla f(\mathbf{x}) + h'(\mathbf{x})^* \mathbf{y} + g'(\mathbf{x})^* Y + g'(\mathbf{x})^* Z \\ \mathbf{b} = -h(\mathbf{x}) \\ \mathbf{c} \in -g(\mathbf{x}) + \partial\theta_1^*(Y) \\ \mathbf{c} \in -g(\mathbf{x}) + \partial\theta_2^*(Z) \end{cases}$$

with $\theta_1 = \phi_1 \circ \lambda$ and $\theta_2 = \phi_2 \circ \lambda$ are two spectral functions

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Strong regularity:

When the solution mapping $S_{\text{KKT}}(\mathbf{a}, \mathbf{b}, \mathbf{c})$ is Lipschitz continuous?

Why it matters?

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- **Perturbation theory**

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- **Algorithm**

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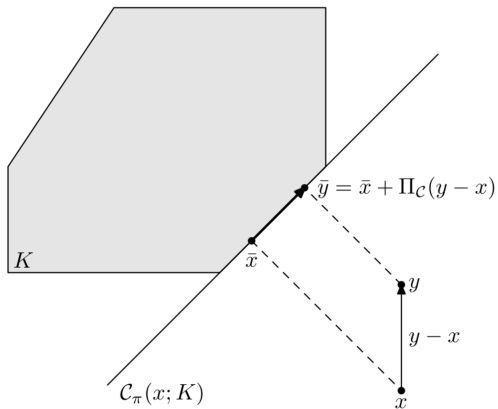
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- Tangent sets
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- The “ σ -term”: the **key** difference between NLPs (**polyhedral**) and CMatOPs (**non-polyhedral**)

The “ σ -term”: polyhedral \implies non-polyhedral



The “ σ -term”: polyhedral \implies non-polyhedral (cont'd)

Metric projection operator $\Pi_{\mathcal{K}}$:

$$\bar{A} := \Pi_{\mathcal{K}}(C) := \operatorname{argmin} \left\{ \frac{1}{2} \|Y - C\|^2 \mid Y \in \mathcal{K} \right\}$$

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If \mathcal{K} is a **polyhedral** closed convex set,

- $\Pi_{\mathcal{K}}$ is **directional differentiable** (Facchinei & Pang, 2003)¹

$$\Pi_{\mathcal{K}}(C + H) - \Pi_{\mathcal{K}}(C) = \Pi_{\mathcal{C}_{\mathcal{K}}(C)}(H) =: \Pi'_{\mathcal{K}}(C; H) \quad \forall H$$

¹F. FACCHINEI AND J. S. PANG. *Finite-Dimensional Variational Inequalities and Complementarity Problems: Volume I*, Springer-Verlag, New York, 2003.

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- $\mathcal{C}_{\mathcal{K}}(C)$ is the critical cone of \mathcal{K} at C

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If \mathcal{K} is a **non-polyhedral** closed convex set

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If \mathcal{K} is a non-polyhedral closed convex set but C^2 -cone reducible,

- $\Pi_{\mathcal{K}}$ is directional differentiable and $\Pi'_{\mathcal{K}}(C; H)$ is the unique optimal solution to (Bonnans et al., 1998)²:

$$\min \{ \|D - H\|^2 - \sigma(\bar{B}, \mathcal{T}_{\mathcal{K}}^2(\bar{A}, D)) \mid D \in \mathcal{C}_{\mathcal{K}}(C) \}$$

²J.F. BONNANS, R. COMINETTI AND A. SHAPIRO. Sensitivity analysis of optimization problems under second order regular constraints. *Mathematics of Operations Research* 23 (1998) 806–831.

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Convex piecewise linear + Symmetric

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(Rockafellar & Wets, 1998): $\phi = \phi_1 + \phi_2$ with $\phi_2 = \delta_{\text{dom } \phi}$

$$\phi_1(\mathbf{x}) = \max_{1 \leq i \leq p} \{ \langle \mathbf{a}^i, \mathbf{x} \rangle - c^i \}, \quad \text{dom } \phi = \{ \mathbf{x} \in \mathbb{R}^n \mid \max_{1 \leq i \leq q} \{ \langle \mathbf{b}^i, \mathbf{x} \rangle - d^i \} \leq 0 \}$$

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Proposition

Let $\phi = \phi_1 + \phi_2 : \mathbb{R}^n \rightarrow (-\infty, \infty]$ be a given proper convex piecewise linear function. ϕ is **symmetric** over \mathbb{R}^n if and only if the functions $\phi_1 : \mathbb{R}^n \rightarrow \mathbb{R}$ and $\phi_2 : \mathbb{R}^n \rightarrow (-\infty, \infty]$ satisfy the following conditions: for any $\mathbf{x} \in \mathbb{R}^n$,

$$\phi_1(\mathbf{x}) = \max_{1 \leq i \leq p} \left\{ \max_{Q \in \mathbb{P}^n} \{ \langle Q\mathbf{a}^i, \mathbf{x} \rangle - c^i \} \right\} \quad \text{and} \quad \phi_2(\mathbf{x}) = \delta_{\text{dom } \phi}(\mathbf{x}),$$

$$\text{where } \text{dom } \phi = \left\{ \mathbf{x} \in \mathbb{R}^n \mid \max_{1 \leq i \leq q} \left\{ \max_{Q \in \mathbb{P}^n} \{ \langle Q\mathbf{b}^i, \mathbf{x} \rangle - d^i \} \right\} \leq 0 \right\}.$$

Convex piecewise linear + Symmetric (cont'd)

- For $i = 1, \dots, p$, define

$$\mathcal{D}_i := \{\mathbf{x} \in \text{dom } \phi \mid \langle \mathbf{a}^j, \mathbf{x} \rangle - c^j \leq \langle \mathbf{a}^i, \mathbf{x} \rangle - c^i \quad \forall j = 1, \dots, p\},$$

then $\text{dom } \phi = \bigcup_{i=1, \dots, p} \mathcal{D}_i$

- any $\bar{\mathbf{x}} \in \text{dom } \phi$, we have two active index sets:

$$\iota_1(\bar{\mathbf{x}}) := \{1 \leq i \leq p \mid \bar{\mathbf{x}} \in \mathcal{D}_i\}, \quad \iota_2(\bar{\mathbf{x}}) := \{1 \leq i \leq q \mid \langle \mathbf{b}^i, \bar{\mathbf{x}} \rangle - d^i = 0\}.$$

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Proposition

For any $i \in \iota_1(\bar{\mathbf{x}})$, $j \in \iota_2(\bar{\mathbf{x}})$ and $Q \in \mathbb{P}_{\bar{\mathbf{x}}}^n$ (i.e., $Q\bar{\mathbf{x}} = \bar{\mathbf{x}}$), there exist $i' \in \iota_1(\bar{\mathbf{x}})$ and $j' \in \iota_2(\bar{\mathbf{x}})$ such that $\mathbf{a}^{i'} = Q\mathbf{a}^i$ and $\mathbf{b}^{j'} = Q\mathbf{b}^j$, respectively.

Convex piecewise linear + Symmetric (cont'd)

Rockafellar & Wets, 1998, Mordukhovich & Sarabi, 2016:

- the **subgradients**:

$$\partial\phi_1(\bar{\mathbf{x}}) = \text{conv}\{\mathbf{a}^i, i \in \iota_1(\bar{\mathbf{x}})\}, \partial\phi_2(\bar{\mathbf{x}}) = \mathcal{N}_{\text{dom } \phi}(\bar{\mathbf{x}}) = \text{cone}\{\mathbf{b}^i, i \in \iota_2(\bar{\mathbf{x}})\}$$

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$$\partial\phi_1(\bar{\mathbf{x}}) = \text{conv}\{\mathbf{a}^i, i \in \iota_1(\bar{\mathbf{x}})\}, \partial\phi_2(\bar{\mathbf{x}}) = \mathcal{N}_{\text{dom } \phi}(\bar{\mathbf{x}}) = \text{cone}\{\mathbf{b}^i, i \in \iota_2(\bar{\mathbf{x}})\}$$

$$\phi_1(\mathbf{x}) = \max_{1 \leq i \leq p} \{\langle \mathbf{a}^i, \mathbf{x} \rangle - c^i\} \text{ is finite everywhere,}$$

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Let $\psi(\mathbf{x}) := \max_{1 \leq i \leq q} \{\langle \mathbf{b}^i, \mathbf{x} \rangle - d^i\}$. Then, $\text{dom } \phi = \{\mathbf{x} \in \mathbb{R}^n \mid \psi(\mathbf{x}) \leq 0\}$

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Tangent sets

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For $\theta_1 = \phi_1 \circ \lambda$:

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- **Tangent set** of epigraph:

$$\mathcal{T}_{\text{epi } \theta_1}(\bar{X}, \theta(\bar{X})) = \text{epi } \theta'_1(\bar{X}; \cdot) := \{(H, y) \in \mathbb{S}^n \times \mathbb{R} \mid \theta'_1(\bar{X}; H) \leq y\}$$

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- The **lineality space**:

$$\mathcal{T}_{\theta_1}^{\text{lin}}(\bar{X}) := \{H \in \mathbb{S}^n \mid \theta'_1(\bar{X}; H) = -\theta'_1(\bar{X}; -H)\}$$

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Proposition

$H \in \mathcal{T}_{\theta_1}^{\text{lin}}(\bar{X})$ if and only if $\langle \mathbf{z}, \lambda'(\bar{X}; H) \rangle$ is a constant for any $\mathbf{z} \in \partial\phi_1(\lambda(\bar{X}))$, i.e.,

$$\langle \lambda'(\bar{X}; H), \mathbf{a}^i - \mathbf{a}^j \rangle = 0 \quad \forall i, j \in \iota_1(\lambda(\bar{X})).$$

Tangent sets (cont'd)

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- **Tangent set** of \mathcal{K} :

$$\begin{aligned}\mathcal{T}_{\mathcal{K}}(\bar{X}) &= \{H \in \mathbb{S}^n \mid \zeta'(\bar{X}; H) \leq 0\} \\ &= \{H \in \mathbb{S}^n \mid \langle \mathbf{b}^i, \lambda'(\bar{X}; H) \rangle \leq 0 \quad \forall i \in \iota_2(\lambda(\bar{X}))\}\end{aligned}$$

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Proposition

$H \in \text{lin}(\mathcal{T}_{\mathcal{K}}(\bar{X}))$ if and only if $\langle \mathbf{b}^i, \lambda'(\bar{X}; H) \rangle = 0$ for any $i \in \iota_2(\lambda(\bar{X}))$.

Tangent sets: SDP

$$\mathbb{S}_-^n = \{X \in \mathbb{S}^n \mid \lambda_{\max}(X) \leq 0\} = \{X \in \mathbb{S}^n \mid \max_{1 \leq i \leq n} \{\langle \mathbf{e}^i, \lambda(X) \rangle\} \leq 0\}$$

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$$\bar{X} = \bar{V} \begin{bmatrix} 0_\alpha & \cdots & 0 \\ \vdots & 0_\beta & \vdots \\ 0 & \cdots & \Lambda_\gamma(\bar{X}) \end{bmatrix} \bar{V}^T, \quad \iota_2(\lambda(\bar{X})) = \alpha \cup \beta = \bar{\gamma}$$

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$$\begin{aligned} \mathcal{T}_{\mathbb{S}_-^n}(\bar{X}) &= \{H \in \mathbb{S}^n \mid \langle \mathbf{e}^i, \lambda'(\bar{X}; H) \rangle \leq 0 \quad \forall i \in \iota_2(\lambda(\bar{X}))\} \\ &= \{H \in \mathbb{S}^n \mid \bar{V}_{\bar{\gamma}}^T H \bar{V}_{\bar{\gamma}} \preceq 0\} \end{aligned}$$

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Critical cone

For $\theta_1 = \phi_1 \circ \lambda$:

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- Let $\bar{Y} \in \partial\theta_1(\bar{X})$. Denote $A = \bar{X} + \bar{Y}$.
- **Critical cone:**

$$\begin{aligned}\mathcal{C}(A; \partial\theta_1(\bar{X})) &:= \{H \in \mathbb{S}^n \mid \theta'_1(\bar{X}; H) \leq \langle \bar{Y}, H \rangle\} \\ &= \{H \in \mathbb{S}^n \mid \theta'_1(\bar{X}; H) = \langle \bar{Y}, H \rangle\}\end{aligned}$$

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Proposition

$H \in \mathcal{C}(A; \partial\theta_1(\bar{X}))$ if and only if $H \in \mathbb{S}^n$ satisfies for any $i, j \in \eta_1(\bar{\mathbf{x}}, \bar{\mathbf{y}})$,

$$\langle \text{diag}(\bar{U}^T H \bar{U}), \mathbf{a}^i \rangle = \langle \text{diag}(\bar{U}^T H \bar{U}), \mathbf{a}^j \rangle = \max_{i \in \iota_1(\bar{\mathbf{x}})} \langle \lambda'(\bar{X}; H), \mathbf{a}^i \rangle,$$

where the index set $\eta_1(\bar{\mathbf{x}}, \bar{\mathbf{y}}) \subseteq \iota_1(\bar{\mathbf{x}})$:

$$\eta_1(\bar{\mathbf{x}}, \bar{\mathbf{y}}) := \left\{ i \in \iota_1(\bar{\mathbf{x}}) \mid \sum_{i \in \iota_1(\bar{\mathbf{x}})} u^i \mathbf{a}^i = \bar{\mathbf{y}}, \sum_{i \in \iota_1(\bar{\mathbf{x}})} u^i = 1, 0 < u^i \leq 1 \right\}$$

with $\bar{\mathbf{x}} := \lambda(\bar{X})$ and $\bar{\mathbf{y}} := \lambda(\bar{Y})$.

Critical cone (cont'd)

For $\theta_2 = \phi_2 \circ \lambda$:

Critical cone (cont'd)

For $\theta_2 = \phi_2 \circ \lambda$:

- Let $\bar{Z} \in \mathcal{N}_{\mathcal{K}}(\bar{X})$. Denote $B = \bar{X} + \bar{Z}$.
- **Critical cone:**

$$\mathcal{C}(B; \mathcal{N}_{\mathcal{K}}(\bar{X})) := \mathcal{T}_{\mathcal{K}}(\bar{X}) \cap \bar{Z}^\perp = \{H \in \mathbb{S}^n \mid \zeta'(\bar{X}; H) \leq 0, \langle \bar{Z}, H \rangle = 0\}$$

Proposition

$H \in \mathcal{C}(B; \mathcal{N}_{\mathcal{K}}(\bar{X}))$ if and only if $H \in \mathbb{S}^n$ satisfies for any $i \in \eta_2(\bar{\mathbf{x}}, \bar{\mathbf{z}})$,

$$0 = \langle \text{diag}(\bar{V}^T H \bar{V}), \mathbf{b}^i \rangle = \max_{i \in \iota_2(\bar{\mathbf{x}})} \langle \lambda'(\bar{X}; H), \mathbf{b}^i \rangle,$$

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with $\bar{\mathbf{x}} := \lambda(\bar{X})$ and $\bar{\mathbf{z}} := \lambda(\bar{Z})$.

Critical cone: SDP

- $\mathbb{S}_-^n = \{X \in \mathbb{S}^n \mid \lambda_{\max}(X) \leq 0\} = \{X \in \mathbb{S}^n \mid \max_{1 \leq i \leq n} \{\langle e^i, \lambda(X) \rangle\} \leq 0\}$
- $\bar{Z} \in \mathcal{N}_{\mathbb{S}_-^n}(\bar{X})$

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$$\bar{X} + \bar{Z} = \bar{V} \begin{bmatrix} \Lambda_\alpha(\bar{Z}) & \cdots & 0 \\ \vdots & 0_\beta & \vdots \\ 0 & \cdots & \Lambda_\gamma(\bar{X}) \end{bmatrix} \bar{V}^T, \quad \left\{ \begin{array}{l} \iota_2(\bar{\mathbf{x}}) = \alpha \cup \beta = \bar{\gamma} \end{array} \right.$$

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$H \in \mathcal{C}(B; \mathcal{N}_{\mathbb{S}_-^n}(\bar{X}))$ if and only if for any $i \in \eta_2(\bar{\mathbf{x}}, \bar{\mathbf{z}})$,

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The “ σ -term”

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- Let $\bar{Y} \in \partial\theta_1(\bar{X})$. Denote $A = \bar{X} + \bar{Y}$ and $H \in \mathcal{C}(A; \partial\theta_1(\bar{X}))$.
- θ_1 is (parabolic) second-order directionally differentiable:

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Moreover,

$$F^*(\bar{Y}) = 2 \sum_{l=1}^r \langle \Lambda(\bar{Y})_{\alpha^l \alpha^l}, \bar{U}_{\alpha^l}^T H(\bar{X} - \bar{v}^l I)^\dagger H \bar{U}_{\alpha^l} \rangle$$

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- θ_1 is (parabolic) second-order directionally differentiable:

$$F(W) := \theta_1''(\bar{X}; H, W) = \phi_1''(\lambda(\bar{X}); \lambda'(\bar{X}; H), \lambda''(\bar{X}; H, W))$$

The σ -term of $\theta_1 \triangleq$ the conjugate function $F^*(\bar{Y})$

Moreover,

$$F^*(\bar{Y}) = 2 \sum_{l=1}^r \langle \Lambda(\bar{Y})_{\alpha^l \alpha^l}, \bar{U}_{\alpha^l}^T H(\bar{X} - \bar{v}^l I)^\dagger H \bar{U}_{\alpha^l} \rangle := \Upsilon_{\bar{X}}^1(\bar{Y}, H)$$

$$\Upsilon_{\bar{X}}^1(\bar{Y}, H) = -2 \sum_{1 \leq l < l' \leq r} \sum_{i \in \alpha^l} \sum_{j \in \alpha^{l'}} \frac{\lambda_i(\bar{Y}) - \lambda_j(\bar{Y})}{\lambda_i(\bar{X}) - \lambda_j(\bar{X})} (\bar{U}_{\alpha^l}^T H \bar{U}_{\alpha^{l'}})^2_{ij}$$

The “ σ -term” (cont’d)

For $\theta_2 = \phi_2 \circ \lambda = \delta_{\mathcal{K}}$ with

$$\mathcal{K} = \{X \in \mathbb{S}^n \mid \lambda(X) \in \text{dom } \phi\} = \{X \in \mathbb{S}^n \mid \zeta(X) \leq 0\},$$

where $\zeta = \psi \circ \lambda$

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Let $\bar{Z} \in \mathcal{N}_{\mathcal{K}}(\bar{X})$. Denote $B = \bar{X} + \bar{Z}$ and $H \in \mathcal{C}(B, \mathcal{N}_{\mathcal{K}}(\bar{X}))$

the “ σ -term” of $\mathcal{K} \triangleq$ the support function of $\mathcal{T}_{\mathcal{K}}^2(\bar{X}, H)$

$$\delta_{\mathcal{T}_{\mathcal{K}}^2(\bar{X}, H)}^*(\bar{Z}) = 2 \sum_{l=1}^r \langle \Lambda(\bar{Z})_{\alpha^l \alpha^l}, \bar{V}_{\alpha^l}^T H (\bar{X} - \bar{v}^l I)^\dagger H \bar{V}_{\alpha^l} \rangle$$

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The “ σ -term”: SDP

- $\mathbb{S}_-^n = \{X \in \mathbb{S}^n \mid \lambda_{\max}(X) \leq 0\}$
- $\bar{Z} \in \mathcal{N}_{\mathbb{S}_-^n}(\bar{X})$, $B = \bar{X} + \bar{Z}$, $H \in \mathcal{C}(B, \mathcal{N}_{\mathcal{K}}(\bar{X}))$

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$$\bar{X} + \bar{Z} = \bar{V} \begin{bmatrix} \Lambda_\alpha(\bar{Z}) & \cdots & 0 \\ \vdots & 0_\beta & 0 \\ 0 & \cdots & \Lambda_\gamma(\bar{X}) \end{bmatrix} \bar{V}^T$$

The “ σ -term” of \mathbb{S}_-^n :

$$\Upsilon_{\bar{X}}^2(\bar{Z}, H) = 2 \sum_{i \in \gamma, j \in \alpha} \frac{\lambda_j(\bar{Z})}{\lambda_i(\bar{X})} (\tilde{H})_{ij}^2,$$

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where $\tilde{H} = \bar{V}^T H \bar{V}$.

CMatOP:

$$\begin{array}{ll} \underset{\mathbf{x} \in \mathbb{X}}{\text{minimize}} & f(\mathbf{x}) + \theta_1(g(\mathbf{x})) \\ \text{subject to} & h(\mathbf{x}) = 0, \\ & g(\mathbf{x}) \in \mathcal{K} \end{array}$$

Robinson CQ

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$$\begin{aligned} & \underset{\mathbf{x} \in \mathbb{X}}{\text{minimize}} && f(\mathbf{x}) + \theta_1(g(\mathbf{x})) \\ & \text{subject to} && h(\mathbf{x}) = 0, \\ & && g(\mathbf{x}) \in \mathcal{K} \end{aligned}$$

Proposition

Let $\bar{\mathbf{x}} \in \mathbb{X}$ be a feasible point of the CMatOP. We say that the **Robinson CQ (RCQ)** holds at $\bar{\mathbf{x}}$ if

$$\begin{bmatrix} h'(\bar{\mathbf{x}}) \\ g'(\bar{\mathbf{x}}) \end{bmatrix} \mathbb{X} + \begin{bmatrix} \{0\} \\ \mathcal{T}_{\mathcal{K}}(g(\bar{\mathbf{x}})) \end{bmatrix} = \begin{bmatrix} \mathbb{Y} \\ \mathbb{S}^n \end{bmatrix}.$$

Thus, the set of Lagrange multipliers $\mathcal{M}(\bar{\mathbf{x}})$ is a non-empty, convex, bounded and compact subset if and only if the **RCQ** holds at $\bar{\mathbf{x}}$.

Second-order optimality conditions

Critical cone of CMatOP:

$$\mathcal{C}(\bar{\mathbf{x}}) := \{\mathbf{d} \in \mathbb{X} \mid h'(\bar{\mathbf{x}})\mathbf{d} = 0, g'(\bar{\mathbf{x}})\mathbf{d} \in \mathcal{C}(A; \partial\theta_1(g(\bar{\mathbf{x}}))), g'(\bar{\mathbf{x}})\mathbf{d} \in \mathcal{C}(B; \mathcal{N}_{\mathcal{K}}(g(\bar{\mathbf{x}})))\}$$

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Theorem (“no gap” second-order optimality conditions)

Suppose that $\bar{\mathbf{x}} \in \mathbb{X}$ is a locally optimal solution of the CMatOP and the RCQ holds. Then, the following inequality holds: for any $\mathbf{d} \in \mathcal{C}(\bar{\mathbf{x}})$,

$$\sup_{(\mathbf{y}, Y, Z) \in \mathcal{M}(\bar{\mathbf{x}})} \left\{ \langle \mathbf{d}, \mathcal{L}''_{\mathbf{xx}}(\bar{\mathbf{x}}, \mathbf{y}, Y, Z)\mathbf{d} \rangle - \Upsilon_{g(\bar{\mathbf{x}})}^1(Y, g'(\bar{\mathbf{x}})\mathbf{d}) - \Upsilon_{g(\bar{\mathbf{x}})}^2(Z, g'(\bar{\mathbf{x}})\mathbf{d}) \right\} \geq 0.$$

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Conversely, let $\bar{\mathbf{x}}$ be a feasible point of the CMatOP such that $\mathcal{M}(\bar{\mathbf{x}})$ is nonempty. Suppose that the RCQ holds at $\bar{\mathbf{x}}$. Then the following condition: for any $\mathbf{d} \in \mathcal{C}(\bar{\mathbf{x}}) \setminus \{0\}$,

$$\sup_{(\mathbf{y}, Y, Z) \in \mathcal{M}(\bar{\mathbf{x}})} \left\{ \langle \mathbf{d}, \mathcal{L}''_{\mathbf{xx}}(\bar{\mathbf{x}}, \mathbf{y}, Y, Z)\mathbf{d} \rangle - \Upsilon_{g(\bar{\mathbf{x}})}^1(Y, g'(\bar{\mathbf{x}})\mathbf{d}) - \Upsilon_{g(\bar{\mathbf{x}})}^2(Z, g'(\bar{\mathbf{x}})\mathbf{d}) \right\} > 0$$

Second-order optimality conditions

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is necessary and sufficient for the **quadratic growth condition** at the point $\bar{\mathbf{x}}$: for any $\mathbf{x} \in N$ such that $h(\mathbf{x}) = 0$ and $g(\mathbf{x}) \in \mathcal{K}$,

$$f(\mathbf{x}) + \phi_1 \circ \lambda(g(\mathbf{x})) \geq f(\bar{\mathbf{x}}) + \phi_1 \circ \lambda(g(\bar{\mathbf{x}})) + \rho \|\mathbf{x} - \bar{\mathbf{x}}\|^2.$$

Strong second-order sufficient condition

Definition

Let $\bar{\mathbf{x}} \in \mathbb{X}$ be a stationary point of the CMatOP. We say the **strong second-order sufficient condition** holds at $\bar{\mathbf{x}}$ if for any $\mathbf{d} \in \hat{\mathcal{C}}(\bar{\mathbf{x}}) \setminus \{0\}$,

$$\sup_{(\mathbf{y}, Y, Z) \in \mathcal{M}(\bar{\mathbf{x}})} \left\{ \langle \mathbf{d}, \mathcal{L}''_{\mathbf{xx}}(\bar{\mathbf{x}}, \mathbf{y}, Y, Z) \mathbf{d} \rangle - \Upsilon_{g(\bar{\mathbf{x}})}^1(Y, g'(\bar{\mathbf{x}}) \mathbf{d}) - \Upsilon_{g(\bar{\mathbf{x}})}^2(Z, g'(\bar{\mathbf{x}}) \mathbf{d}) \right\} > 0$$

with

$$\hat{\mathcal{C}}(\bar{\mathbf{x}}) := \bigcap_{(\mathbf{y}, Y, Z) \in \mathcal{M}(\bar{\mathbf{x}})} \text{app}(\mathbf{y}, Y, Z),$$

where for any $(\mathbf{y}, Y, Z) \in \mathcal{M}(\bar{\mathbf{x}})$, the set $\text{app}(\mathbf{y}, Y, Z)$ is given by

$$\text{app}(\bar{\mathbf{y}}, \bar{Y}, \bar{Z}) := \{ \mathbf{d} \in \mathbb{X} \mid h'(\bar{\mathbf{x}}) \mathbf{d} = 0, g'(\bar{\mathbf{x}}) \mathbf{d} \in \text{aff}(\mathcal{C}(A; \partial\theta_1(g(\bar{\mathbf{x}}))), g'(\bar{\mathbf{x}}) \mathbf{d} \in \text{aff}(\mathcal{C}(B; \mathcal{N}_{\mathcal{K}}(g(\bar{\mathbf{x}})))) \}.$$

Constraint nondegeneracy (LICQ)

The **constraint nondegeneracy** for the CMatOP is defined as follows

$$\begin{bmatrix} h'(\bar{\mathbf{x}}) \\ g'(\bar{\mathbf{x}}) \\ g'(\bar{\mathbf{x}}) \end{bmatrix} \mathbb{X} + \begin{bmatrix} \{0\} \\ \mathcal{T}_{\theta_1}^{\text{lin}}(g(\bar{\mathbf{x}})) \\ \text{lin}(\mathcal{T}_{\mathcal{K}}(g(\bar{\mathbf{x}}))) \end{bmatrix} = \begin{bmatrix} \mathbb{Y} \\ \mathbb{S}^n \\ \mathbb{S}^n \end{bmatrix}.$$

Strong regularity of CMatOPs

Theorem

Let $\bar{\mathbf{x}} \in \mathbb{X}$ be a stationary point of CMatOP with multipliers $(\bar{\mathbf{y}}, \bar{Y}, \bar{Z})$:

(i) the strong second order sufficient condition and constraint nondegeneracy hold at $\bar{\mathbf{x}}$;

(ii) every element in $\partial F(\bar{\mathbf{x}}, \bar{\mathbf{y}}, \bar{Y}, \bar{Z})$ is nonsingular;

(iii) $(\bar{\mathbf{x}}, \bar{\mathbf{y}}, \bar{Y}, \bar{Z})$ is a strongly regular solution of the KKT system.

It holds that (i) \implies (ii) \implies (iii).

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(iii) \implies (i) can be established for particular CMatOPs:

- NLSDP (**Sun, MOR 2006**)
- CMatOPs with the sum of k -largest eigenvalues, etc (in our work)

Thank you!