

An Introduction to a Class of Matrix Cone Programming

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Let $X \in \mathbb{R}^{m \times n}$ admit the following singular value decomposition:

$$X = \bar{U} [\Sigma(X) \ 0] \bar{V}^T = \bar{U} [\Sigma(X) \ 0] [\bar{V}_1 \ \bar{V}_2]^T = \bar{U} \Sigma(X) \bar{V}_1^T, \quad (1)$$

where $\bar{U} \in \mathcal{O}^m$, $\bar{V} \in \mathcal{O}^n$ and $\bar{V}_1 \in \mathbb{R}^{n \times m}$, $\bar{V}_2 \in \mathbb{R}^{n \times (n-m)}$ and $\bar{V} = [\bar{V}_1 \ \bar{V}_2]$. The set of such matrices (U, V) in the singular value decomposition (1) is denoted by $\mathcal{O}^{m,n}(X)$, i.e.,

$$\mathcal{O}^{m,n}(X) := \{(U, V) \in \mathbb{R}^{m \times m} \times \mathbb{R}^{n \times n} \mid X = U [\Sigma(X) \ 0] V^T\}.$$

Define the index sets a , b and c by

$$a := \{i \mid \sigma_i(X) > 0\}, \quad b := \{i \mid \sigma_i(X) = 0\} \quad \text{and} \quad c := \{m + 1, \dots, n\}. \quad (2)$$

Let $\bar{\mu}_1 > \bar{\mu}_2 > \dots > \bar{\mu}_r > 0$ be the nonzero distinct singular values of X .

Then, let

$$a_k := \{i \mid \sigma_i(X) = \bar{\mu}_k\}, \quad k = 1, \dots, r. \quad (3)$$

For any positive constant $\varepsilon > 0$, denote the closed convex cone $\mathcal{D}_n^\varepsilon$ by

$$\mathcal{D}_n^\varepsilon := \{(t, x) \in \mathbb{R} \times \mathbb{R}^n \mid \varepsilon^{-1}t \geq x_i, i = 1, \dots, n\}. \quad (4)$$

Let $\Pi_{\mathcal{D}_n^\varepsilon}(\cdot)$ be the metric projector over $\mathcal{D}_n^\varepsilon$ under the Euclidean inner product in \mathbb{R}^n . That is, for any $(t, x) \in \mathbb{R} \times \mathbb{R}^n$, $\Pi_{\mathcal{D}_n^\varepsilon}(t, x)$ is the unique optimal solution to the following convex optimization problem

$$\begin{aligned} \min \quad & \frac{1}{2}((\tau - t)^2 + \|y - x\|^2) \\ \text{s.t.} \quad & \varepsilon^{-1}\tau \geq y_i, i = 1, \dots, n. \end{aligned} \quad (5)$$

Then we have the following useful result about $\Pi_{\mathcal{D}_n^\varepsilon}(\cdot, \cdot)$.

For any $x \in \mathbb{R}^n$, let x^\downarrow be the vector of components of x being arranged in the non-increasing order $x_1^\downarrow \geq \dots \geq x_n^\downarrow$. Let $\text{sgn}(x)$ be the sign vector of x , i.e., $(\text{sgn})_i(x) = 1$ if $x_i \geq 0$ and -1 otherwise. We use “ \circ ” to denote the Hadamard product operation either for two vectors or two matrices of the same dimensions.

Proposition 1. *Assume that $\varepsilon > 0$ and $(t, x) \in \mathbb{R} \times \mathbb{R}^n$ are given. Let π be a permutation of $\{1, \dots, n\}$ such that $x^\downarrow = x_\pi$, i.e., $x_i^\downarrow = x_{\pi(i)}$, $i = 1, \dots, n$ and π^{-1} the inverse of π . For convenience, write $x_0^\downarrow = +\infty$ and $x_{n+1}^\downarrow = -\infty$. Let \bar{k} be the smallest integer $k \in \{0, 1, \dots, n\}$ such that*

$$x_{k+1}^\downarrow \leq \left(\sum_{j=1}^k x_j^\downarrow + \varepsilon t \right) / (k + \varepsilon^2) < x_k^\downarrow. \quad (6)$$

Define $\bar{y} \in \mathbb{R}^n$ and $\bar{\tau} \in \mathbb{R}_+$, respectively, by

$$\bar{y}_i := \begin{cases} \left(\sum_{j=1}^{\bar{k}} x_j^\downarrow + \varepsilon t \right) / (\bar{k} + \varepsilon^2) & \text{if } 1 \leq i \leq \bar{k}, \\ x_i^\downarrow & \text{otherwise} \end{cases}$$

and

$$\bar{\tau} := \varepsilon \bar{y}_1 = \varepsilon \left(\sum_{j=1}^{\bar{k}} x_j^\downarrow + \varepsilon t \right) / (\bar{k} + \varepsilon^2).$$

The metric projection $\Pi_{\mathcal{D}_n^\varepsilon}(t, x)$ is computed by $\Pi_{\mathcal{D}_n^\varepsilon}(t, x) = (\bar{\tau}, \bar{y}_{\pi-1})$.

For any positive constant $\varepsilon > 0$, denote the closed convex cone $\mathcal{C}_n^\varepsilon$ by

$$\mathcal{C}_n^\varepsilon := \{(t, x) \in \mathbb{R} \times \mathbb{R}^n \mid \varepsilon^{-1}t \geq \|x\|_\infty\}. \quad (7)$$

Let $\Pi_{\mathcal{C}_n^\varepsilon}(\cdot, \cdot)$ be the metric projector over $\mathcal{C}_n^\varepsilon$ under the Euclidean inner product in \mathbb{R}^n . That is, for any $(t, x) \in \mathbb{R} \times \mathbb{R}^n$, $\Pi_{\mathcal{C}_n^\varepsilon}(t, x)$ is the unique optimal solution to the following convex optimization problem

$$\begin{aligned} \min \quad & \frac{1}{2}((\tau - t)^2 + \|y - x\|^2) \\ \text{s.t.} \quad & \varepsilon^{-1}\tau \geq \|y\|_\infty. \end{aligned} \quad (8)$$

In the following discussions, we frequently drop n from $\mathcal{C}_n^\varepsilon$ when its size can be found from the context.

Assume that $\varepsilon > 0$ and $(t, x) \in \mathfrak{R} \times \mathfrak{R}^n$ are given. Let π be a permutation of $\{1, \dots, n\}$ such that $|x|^\downarrow = |x|_\pi$, i.e., $|x|_i^\downarrow = |x|_{\pi(i)}$, $i = 1, \dots, n$ and π^{-1} be the inverse of π . Let $|x|_0^\downarrow = +\infty$ and $|x|_{n+1}^\downarrow = 0$. Let $s_0 = 0$ and $s_k = \sum_{i=1}^k |x|_i^\downarrow$, $k = 1, \dots, n+1$. Let \bar{k} be the smallest integer $k \in \{0, 1, \dots, n\}$ such that

$$|x|_{\bar{k}+1}^\downarrow \leq (s_{\bar{k}} + \varepsilon t) / (\bar{k} + \varepsilon^2) < |x|_{\bar{k}}^\downarrow \quad (9)$$

or $\bar{k} = n+1$ if such an integer does not exist. Denote

$$\theta^\varepsilon(t, x) := (s_{\bar{k}} + \varepsilon t) / (\bar{k} + \varepsilon^2). \quad (10)$$

Let α, β and γ be the index sets of $|x|^\downarrow$ as

$$\alpha := \{i \mid |x|_i^\downarrow > \theta^\varepsilon(t, x)\}, \quad \beta := \{i \mid |x|_i^\downarrow = \theta^\varepsilon(t, x)\} \quad (11)$$

and

$$\gamma := \{i \mid |x|_i^\downarrow < \theta^\varepsilon(t, x)\}. \quad (12)$$

Define $\bar{y} \in \mathbb{R}^n$ and $\bar{\tau} \in \mathbb{R}_+$, respectively, by

$$\bar{y}_i := \begin{cases} \max\{\theta^\varepsilon(t, x), 0\} & \text{if } i \in \alpha, \\ |x|_i^\downarrow & \text{otherwise} \end{cases}$$

and

$$\bar{\tau} := \varepsilon \max\{\theta^\varepsilon(t, x), 0\}.$$

Proposition 2. *Assume that $\varepsilon > 0$ and $(t, x) \in \mathbb{R} \times \mathbb{R}^n$ are given.*

(i) The metric projection $\Pi_{\mathcal{C}^\varepsilon}(t, x)$ of (t, x) onto \mathcal{C}^ε can be computed as follows

$$\Pi_{\mathcal{C}^\varepsilon}(t, x) = (\bar{\tau}, \text{sgn}(x) \circ \bar{y}_{\pi-1}). \quad (13)$$

(ii) The mapping $\Pi_{\mathcal{C}^\varepsilon}(\cdot, \cdot)$ is piecewise linear. Denote $\delta := \sqrt{\varepsilon^2 + \bar{k}}$. For any $(\eta, h) \in \mathbb{R} \times \mathbb{R}^n$, let

$$\eta' := \begin{cases} \delta^{-1}(\varepsilon\eta + \sum_{i \in \pi^{-1}(\alpha)} \text{sgn}(x_i)h_i) & \text{if } t \geq -\varepsilon^{-1}\|x\|_1, \\ 0 & \text{otherwise.} \end{cases}$$

Let $h' = \text{sgn}(x) \circ h$. Then, the directional derivative of $\Pi_{\mathcal{C}^\varepsilon}(\cdot, \cdot)$ at (t, x) along the direction (η, h) is given by

$$(\bar{\eta}, \bar{h}) := \Pi'_{\mathcal{C}^\varepsilon}((t, x); (\eta, h)), \quad (14)$$

with

$$\left(\delta \varepsilon^{-1} \bar{\eta}, (\text{sgn}(x) \circ \bar{h})_{\pi^{-1}(\beta)} \right) = \begin{cases} \Pi_{\mathcal{D}_{|\beta|}^{\delta}}(\eta', h'_{\pi^{-1}(\beta)}) & \text{if } t > -\varepsilon^{-1} \|x\|_1, \\ \Pi_{\mathcal{C}_{|\beta|}^{\delta}}(\eta', h'_{\pi^{-1}(\beta)}) & \text{otherwise,} \end{cases}$$

$$\bar{h}_i = \bar{\eta}, \quad i \in \pi^{-1}(\alpha) \quad \text{and} \quad \bar{h}_i = h_i, \quad i \in \pi^{-1}(\gamma).$$

Note that if $\beta = \emptyset$, let $\mathcal{D}_{|\beta|}^{\delta} := \mathfrak{R}$ and $\mathcal{C}_{|\beta|}^{\delta} := \mathfrak{R}_+$.

(iii) The mapping $\Pi_{\mathcal{C}^{\varepsilon}}(\cdot, \cdot)$ is differentiable at (t, x) if and only if $t > \varepsilon \|x\|_{\infty}$, or $\varepsilon \|x\|_{\infty} > t > -\varepsilon^{-1} \|x\|_1$ but $|x|_{\bar{k}+1}^{\downarrow} < (s_k + \varepsilon t) / (\bar{k} + \varepsilon^2)$ or $t < -\varepsilon^{-1} \|x\|_1$.



For any positive constant $\varepsilon > 0$, define the matrix cone $\mathcal{M}_n^\varepsilon$ in \mathcal{S}^n as the epigraph of the convex function $\varepsilon \lambda_{\max}(\cdot)$, i.e.,

$$\mathcal{M}_n^\varepsilon := \{(t, X) \in \mathfrak{R} \times \mathcal{S}^n \mid \varepsilon^{-1}t \geq \lambda_{\max}(X)\}. \quad (15)$$

Proposition 3. *Assume that $(t, X) \in \mathfrak{R} \times \mathcal{S}^n$ is given. Let X have the eigenvalue decomposition*

$$X = \bar{P} \text{diag}(\lambda(X)) \bar{P}^T, \quad (16)$$

where $\bar{P} \in \mathcal{O}^n$. Let $\Pi_{\mathcal{M}_n^\varepsilon}(\cdot, \cdot)$ be the metric projector over $\mathcal{M}_n^\varepsilon$ under Frobenius norm in \mathcal{S}^n . Then,

$$\Pi_{\mathcal{M}_n^\varepsilon}(t, X) = (\bar{t}, \bar{P} \text{diag}(\bar{y}) \bar{P}^T) \quad \forall (t, X) \in \mathfrak{R} \times \mathcal{S}^n, \quad (17)$$

where $(\bar{t}, \bar{y}) = \Pi_{\mathcal{D}_n^\varepsilon}(t, \lambda(X)) \in \mathfrak{R} \times \mathfrak{R}^n$.

Theorem 1. Assume that $(t, X) \in \mathbb{R} \times \mathbb{R}^{m \times n}$ is given. Let X have the singular value decomposition (1). Let $\Pi_{\mathcal{K}^\varepsilon}(\cdot, \cdot)$ be the metric projector over \mathcal{K}^ε under Frobenius norm in $\mathbb{R}^{m \times n}$. For any $(t, X) \in \mathbb{R} \times \mathbb{R}^{m \times n}$, we have

$$\Pi_{\mathcal{K}^\varepsilon}(t, X) = \left(\bar{t}, \bar{U} \begin{bmatrix} \text{diag}(\bar{y}) & 0 \end{bmatrix} \bar{V}^T \right), \quad (18)$$

where

$$(\bar{t}, \bar{y}) = \Pi_{\mathcal{C}^\varepsilon}(t, \sigma(X)) \in \mathbb{R} \times \mathbb{R}^m.$$



Define $S : \mathfrak{R}^{m \times m} \rightarrow \mathcal{S}^m$ and $T : \mathfrak{R}^{m \times m} \rightarrow \mathfrak{R}^{m \times m}$ as follows

$$S(Z) := \frac{1}{2}(Z + Z^T) \quad \text{and} \quad T(Z) := \frac{1}{2}(Z - Z^T).$$

For any given $(t, X) \in \mathfrak{R} \times \mathfrak{R}^{m \times n}$, Let X have the singular value decomposition (1). For convenience, write $\sigma_0(X) = +\infty$ and $\sigma_{n+1}(X) = 0$. Let $s_0 = 0$ and $s_k = \sum_{i=1}^k \sigma_i(X)$, $k = 1, \dots, n+1$. Let \bar{k} be the smallest integer $k \in \{0, 1, \dots, n\}$ such that

$$\sigma_{k+1}(X) \leq (s_k + \varepsilon t) / (k + \varepsilon^2) < \sigma_k(X) \tag{19}$$

or $\bar{k} = n + 1$ if such an integer does not exist.

Denote by

$$\theta(t, \sigma(X)) := (s_{\bar{k}} + \varepsilon t) / (\bar{k} + \varepsilon^2). \quad (20)$$

Let α , β and γ be the index sets of $\sigma(X)$, which are defined by (11) and (12).

Let $\delta := \sqrt{1 + \bar{k}}$. Define a linear operator $\rho : \Re \times \Re^{m \times n} \rightarrow \Re$ as follows

$$\rho(\eta, H) := \begin{cases} \delta^{-1}(\text{Tr}(S(\bar{U}_\alpha^T H (\bar{V}_1)_\alpha)) + \eta) & \text{if } t \geq -\|X\|_* , \\ 0 & \text{otherwise.} \end{cases} \quad (21)$$

Denote by

$$\left(g_0(t, \sigma(X)), g(t, \sigma(X)) \right) := \Pi_C(t, \sigma(X)).$$

Define $\Omega_1 \in \mathbb{R}^{m \times m}$, $\Omega_2 \in \mathbb{R}^{m \times m}$ and $\Omega_3 \in \mathbb{R}^{m \times (n-m)}$ (depend on X) as follows, for any $i, j \in \{1, \dots, m\}$,

$$(\Omega_1)_{ij} := \begin{cases} \frac{g_i(t, \sigma(X)) - g_j(t, \sigma(X))}{\sigma_i(X) - \sigma_j(X)} & \text{if } \sigma_i(X) \neq \sigma_j(X), \\ 0 & \text{otherwise,} \end{cases} \quad (22)$$

$$(\Omega_2)_{ij} := \begin{cases} \frac{g_i(t, \sigma(X)) + g_j(t, \sigma(X))}{\sigma_i(X) + \sigma_j(X)} & \text{if } \sigma_i(X) + \sigma_j(X) \neq 0, \\ 0 & \text{otherwise} \end{cases} \quad (23)$$

and for any $i \in \{1, \dots, m\}$ and $j \in \{1, \dots, n-m\}$

$$(\Omega_3)_{ij} := \begin{cases} \frac{g_i(t, \sigma(X))}{\sigma_i(X)} & \text{if } \sigma_i(X) \neq 0, \\ 0 & \text{if } \sigma_i(X) = 0, \end{cases} \quad (24)$$

Theorem 2. Assume that $(t, X) \in \mathbb{R} \times \mathbb{R}^{m \times n}$ is given. Let X have the singular value decomposition (1). Then, the metric projector over the matrix cone \mathcal{K} , $\Pi_{\mathcal{K}}(\cdot, \cdot)$ is directionally differentiable at (t, X) . For any given direction $(\eta, H) \in \mathbb{R} \times \mathbb{R}^{m \times n}$, the directional derivative $\Pi'_{\mathcal{K}}((t, X); (\eta, H))$ is given by

(i) if $t > \|X\|_2$, then $\Pi'_{\mathcal{K}}((t, X); (\eta, H)) = (\eta, H)$.

(ii) if $\|X\|_2 \geq t > -\|X\|_*$, then $\Pi'_{\mathcal{K}}((t, X); (\eta, H)) = (\bar{\eta}, \bar{H})$ with



$$\begin{aligned} \bar{\eta} &= \delta^{-1} \psi_0^\delta(\eta, H), \\ \bar{H} &= \bar{U} \begin{bmatrix} \delta^{-1} \psi_0^\delta(\eta, H) I_{\alpha\alpha} & 0 & (\Omega_1)_{\alpha\gamma} \circ S(A_{\alpha\gamma}) \\ 0 & \Psi^\delta(\eta, H) & S(A_{\beta\gamma}) \\ (\Omega_1)_{\gamma\alpha} \circ S(A_{\gamma\alpha}) & S(A_{\gamma\beta}) & S(A_{\gamma\gamma}) \end{bmatrix} \bar{V}_1^T \\ &\quad + \bar{U} \begin{bmatrix} (\Omega_2)_{aa} \circ T(A_{aa}) & (\Omega_2)_{ab} \circ T(A_{ab}) \\ (\Omega_2)_{ba} \circ T(A_{ba}) & T(A_{bb}) \end{bmatrix} \bar{V}_1^T + \bar{U} \begin{bmatrix} (\Omega_3)_{ac} \circ B_{ac} \\ B_{bc} \end{bmatrix} \bar{V}_2^T \end{aligned}$$

where $A := \bar{U}^T H \bar{V}_1$, $B := \bar{U}^T H \bar{V}_2$ and $(\psi_0^\delta(\eta, H), \Psi^\delta(\eta, H)) \in \mathfrak{R} \times \mathfrak{R}^{|\beta| \times |\beta|}$ is given by

$$\left(\psi_0^\delta(\eta, H), \Psi^\delta(\eta, H) \right) := \Pi_{\mathcal{M}_{|\beta|}^\delta}(\rho(\eta, H), S(\bar{U}_\beta^T H (\bar{V}_1)_\beta)). \quad (25)$$

(iii) if $t = -\|X\|_*$, then $\Pi'_{\mathcal{K}}((t, X); (\eta, H)) = (\bar{\eta}, \bar{H})$ with

$$\begin{aligned}\bar{\eta} &= \delta^{-1} \psi_0^\delta(\eta, H), \\ \bar{H} &= \bar{U} \begin{bmatrix} \bar{\eta} I_{\alpha\alpha} & 0 \\ 0 & \Psi_1^\delta(\eta, H) \end{bmatrix} \bar{V}_1^T \\ &\quad + \bar{U} [\Omega_2 \circ T(A)] \bar{V}_1^T + \bar{U} \begin{bmatrix} (\Omega_3)_{ac} \circ B_{ac} \\ \Psi_2^\delta(\eta, H) \end{bmatrix} \bar{V}_2^T\end{aligned}$$

where $\psi_0^\delta(\eta, H) \in \mathfrak{R}$, $\Psi_1^\delta(\eta, H) \in \mathfrak{R}^{|\beta| \times |\beta|}$ and $\Psi_2^\delta(\eta, H) \in \mathfrak{R}^{|\beta| \times (n-m)}$ are given by

$$\begin{aligned} & \left(\psi_0^\delta(\eta, H), \left[\Psi_1^\delta(\eta, H) \quad \Psi_2^\delta(\eta, H) \right] \right) \\ := & \Pi_{\mathcal{K}^\delta_{|\beta|, (n-|\alpha|)}} \left(\rho(\eta, H), \left[\bar{U}_\beta^T H V_\beta \quad \bar{U}_\beta^T H \bar{V}_2 \right] \right). \end{aligned}$$

(iv) if $t < -\|X\|_*$, then

$$\Pi'_{\mathcal{K}}((t, X); (\eta, H)) = (0, 0).$$

Moreover, $\Pi_{\mathcal{K}}(\cdot, \cdot)$ is strongly B-differentiable at (t, X) , i.e., for any $(\eta, H) \in \mathfrak{R} \times \mathfrak{R}^{m \times n}$ and $(\eta, H) \rightarrow (0, 0)$, we have

$$\Pi_{\mathcal{K}}(t + \eta, X + H) - \Pi_{\mathcal{K}}(t, X) - \Pi'_{\mathcal{K}}((t, X); (\eta, H)) = O(\|(\eta, H)\|^2). \quad (26)$$



Theorem 3. *The mapping $\Pi_{\mathcal{K}}(\cdot, \cdot)$ is F -differentiable at $(t, X) \in \mathfrak{R} \times \mathfrak{R}^{m \times n}$ if and only if (t, X) satisfies one of the following conditions:*

(i) $t > \|X\|_2$;

(ii) $\|X\|_2 > t > -\|X\|_*$ but $\sigma_{\bar{k}}(X) < \theta(t, \sigma(X))$, where \bar{k} and $\theta(t, \sigma(X))$ are given by (9) and (10), respectively;

(iii) $t < -\|X\|_*$.

Denote $\delta := \sqrt{1 + \bar{k}}$. Let $\rho : \mathfrak{R} \times \mathfrak{R}^{m \times n} \rightarrow \mathfrak{R}$ be the linear operator defined by (21). Then for any $(\eta, H) \in \mathfrak{R} \times \mathfrak{R}^{m \times n}$, $\Pi'_{\mathcal{K}}(t, X)(\eta, H) = (\bar{\eta}, \bar{H})$ with



$$\bar{\eta} = \delta^{-1} \rho(\eta, H)$$

and

$$\begin{aligned} \bar{H} = & \bar{U} \begin{bmatrix} \delta^{-1} \rho(\eta, H) I_{\alpha\alpha} & (\Omega_1)_{\alpha\gamma} \circ S(A_{\alpha\gamma}) \\ (\Omega_1)_{\gamma\alpha} \circ S(A_{\gamma\alpha}) & S(A_{\gamma\gamma}) \end{bmatrix} \bar{V}_1^T \\ & + \bar{U} \begin{bmatrix} (\Omega_2)_{aa} \circ T(A_{aa}) & (\Omega_2)_{ab} \circ T(A_{ab}) \\ (\Omega_2)_{ba} \circ T(A_{ba}) & T(A_{bb}) \end{bmatrix} \bar{V}_1^T + \bar{U} \begin{bmatrix} (\Omega_3)_{ac} \circ B_{ac} \\ B_{bc} \end{bmatrix} \bar{V}_2^T \end{aligned}$$

where $A := \bar{U}^T H \bar{V}_1$ and $B := \bar{U}^T H \bar{V}_2$.

Theorem 4. $\Pi_{\mathcal{K}}(\cdot, \cdot)$ is strongly G -semismooth at any $(t, X) \in \mathbb{R} \times \mathbb{R}^{m \times n}$.