

Quadratic convergence of Newton's method to the optimal solution of second-order cone optimization

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Quadratic convergence of Newton's method (1 of 32)

Outline

Second-order cone optimization

Convergence under strict complementarity

Failure of strict complementarity

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Standard form

We aim to solve

$$\begin{aligned} (P) & \min\{c^T x \ : \ Ax = b, \ x \in \mathcal{L}^{\bar{n}}_+\}, \\ (D) & \max\{b^T y \ : \ A^T y + s = c, \ s \in \mathcal{L}^{\bar{n}}_+\}, \end{aligned}$$

where

$$\begin{array}{l} \mathcal{L}^{\bar{n}}_{+} := \mathbb{L}^{n_{1}}_{+} \times \ldots \times \mathbb{L}^{n_{p}}_{+}, \\ \mathbb{L}^{n_{i}}_{+} := \{x^{i} := (x^{i}_{1}, \ldots, x^{i}_{n_{i}})^{T} \in \mathbb{R}^{n_{i}} : x^{i}_{1} \geq \|x^{i}_{2:n_{i}}\|\}, \quad i = 1, \ldots, p, \\ \mathbb{P} \ A \in \mathbb{R}^{m \times \bar{n}}, \ c \in \mathbb{R}^{\bar{n}}, \ b \in \mathbb{R}^{m}, \\ \mathbb{P} \ A := (A_{1}, \ldots, A_{p}), \ x := (x^{1}; \ldots; x^{p}), \ s := (s^{1}; \ldots; s^{p}), \ \text{and} \ c := (c^{1}; \ldots; c^{p}), \\ \mathbb{P} \ \bar{n} := \sum_{i=1}^{p} n_{i}. \end{array}$$

Regularity conditions

Assumption

A is assumed to be a full row rank matrix.

Assumption (Interior point condition)

There exists a primal-dual feasible (x; y; s) so that $x, s \in int(\mathcal{L}^{\overline{n}}_{+})$.

▶ As a result, the optimal set is written as the set of solutions of

$$Ax = b, \quad x \in \mathcal{L}^{\bar{n}}_+,$$
$$A^T y + s = c, \quad s \in \mathcal{L}^{\bar{n}}_+,$$
$$x \circ s = 0,$$

where $x \circ s = (x^1 \circ s^1; \ldots; x^p \circ s^p)$, and

$$x^{i} \circ s^{i} = \begin{pmatrix} (x^{i})^{T} s^{i} \\ x_{1}^{i} s_{2:n_{i}}^{i} + s_{1}^{i} x_{2:n_{i}}^{i} \end{pmatrix}, \quad \forall \ i = 1, \dots, p.$$

Interior point method

For $\mu > 0$, we solve a system of perturbed optimality conditions:

$$\begin{array}{lll} Ax & = b, & x \in \operatorname{int}(\mathcal{L}^{\overline{n}}_{+}), \\ A^{T}y + s & = c, & s \in \operatorname{int}(\mathcal{L}^{\overline{n}}_{+}), \\ x \circ s & = \mu e. \end{array}$$

where $e^{i} = (1; \mathbf{0}) \in \mathbb{R}^{n_{i}}$, and $e = (e^{1}; ...; e^{p})$.

- ▶ This system has a unique solution, the so called central solution.
- As $\mu \to 0$, the trajectory converges to a maximally complementary solution.

Let \mathcal{P}^* and \mathcal{D}^* be the sets of primal and dual optimal solutions.

Definition

An optimal solution $(x^*; y^*; s^*) \in \mathcal{P}^* \times \mathcal{D}^*$ is maximally complementary if

 $(x^*; y^*; s^*) \in \operatorname{ri}(\mathcal{P}^* \times \mathcal{D}^*).$

Illustration of the interior point method



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Quadratic convergence of Newton's method

The optimality conditions can be written as F((x; y; s)) = 0 and $x, s \in \mathcal{L}^{\bar{n}}_+$, where

$$F((x;y;s)) := \begin{pmatrix} Ax - b \\ A^Ty + s - c \\ x \circ s \end{pmatrix}.$$

The Jacobian of F is given by

$$\nabla F((x;y;s)) := \begin{pmatrix} A & 0 & 0 \\ 0 & A^T & I \\ L(s) & 0 & L(x) \end{pmatrix}.$$

L(x) is a block diagonal matrix:

$$L(x) := \operatorname{diag}(L(x^{1}), \dots, L(x^{p})),$$
$$L(x^{i}) := \begin{pmatrix} x_{1}^{i} & (x_{2:n_{i}}^{i})^{T} \\ x_{2:n_{i}}^{i} & x_{1}^{i}I_{n_{i}-1} \end{pmatrix}$$

• ∇F is Lipschitz continuous with global constant $\tau_1 = 2$.

Sufficient conditions for nonsingularity

The Jacobian ∇F is nonsingular (Alizadeh and Goldfarb) at $(x^*; y^*; s^*)$ if

- $(x^*; y^*; s^*)$ is strictly complementary,
- $(x^*; y^*; s^*)$ is primal-dual nondegenerate.

Definition (Strict complementarity)

An optimal solution $(x^*; y^*; s^*) \in \mathcal{P}^* \times \mathcal{D}^*$ is strictly complementary if

 $x^* + s^* \in \operatorname{int}(\mathcal{L}^{\bar{n}}_+).$

Let $\tan(x^i, \mathbb{L}^{n_i}_+)$ be the tangent space to $\mathbb{L}^{n_i}_+$ at x^i .

Definition (Nondegeneracy-Transversality)

A primal-feasible solution x is called nondegenerate if

$$\tan(x^1, \mathbb{L}^{n_1}_+) \times \ldots \times \tan(x^p, \mathbb{L}^{n_p}_+) + \operatorname{Ker}(A) = \mathbb{R}^{\bar{n}}.$$

A dual feasible solution (y; s) is called nondegenerate if

$$\tan(s^1, \mathbb{L}^{n_1}_+) \times \ldots \times \tan(s^p, \mathbb{L}^{n_p}_+) + \mathcal{R}(A^T) = \mathbb{R}^{\bar{n}}.$$

Distance to the optimal set

Let $(\hat{x}; \hat{y}; \hat{s})$ be a primal-dual optimal solution.

The primal and dual optimal sets can be equivalently written as

$$\begin{cases} x \in \hat{x} + \operatorname{Ker}(A), \\ \hat{s}^T x = 0, \\ x \in \mathcal{L}^{\bar{n}}_+, \end{cases} \qquad \qquad \begin{cases} s \in \hat{s} + \mathcal{R}(A^T), \\ \hat{x}^T s = 0, \\ s \in \mathcal{L}^{\bar{n}}_+. \end{cases}$$

The distance between $(x(\mu); y(\mu); s(\mu))$ and the affine space in the above system:

- can be bounded by Hoffman error bound.
- ▶ θ_1 and θ_2 are Hoffman condition numbers for the primal and dual systems.

Lemma (Hölderian error bound)

Let $(x(\mu); y(\mu); s(\mu))$ be a central solution with

$$\mu \leq \hat{\mu} := \min\left\{\frac{1}{\theta_1 p}, \frac{1}{\theta_2 p}\right\}.$$

Then there exists $(x; y; s) \in \mathcal{P}^* \times \mathcal{D}^*$, $\gamma > 0$, and $\kappa > 0$ so that

 $\|x(\mu) - x\| \le \kappa(p\mu)^{\gamma}, \quad \|y(\mu) - y\| \le \kappa(p\mu)^{\gamma}, \quad \|s(\mu) - s\| \le \kappa(p\mu)^{\gamma}.$

Quadratic convergence to a strictly complementary solution

Theorem

Assume that there exists $\beta_1 > 0$ so that

$$\|\nabla F((x^*; y^*; s^*))^{-1}\| \le \beta_1.$$

Let a central solution $(x(\mu); y(\mu); s(\mu))$ with

$$\mu < \min\left\{p^{-1}\left(4\sqrt{3}\beta_1\kappa\right)^{-\frac{1}{\gamma}}, \hat{\mu}\right\}$$

be given.

From $(x(\mu); y(\mu); s(\mu))$ Newton's method is quadratically convergent to $(x^*; y^*; s^*)$.

• $(x(\mu); y(\mu); s(\mu))$ needs to be in the convergence region of Newton's method.

Outline

Second-order cone optimization

Convergence under strict complementarity

Failure of strict complementarity

Failure of strict complementarity

Without the strict complementarity condition:

- ∇F might be singular at an optimal solution,
- Newton's method is not applicable.
- Convergence to an optimal solution is not better than linear.

We can release the dependence on the strict complementarity condition:

• We need to identify the optimal partition of the problem.

Back to the complementarity condition

The complementarity condition $x^i \circ s^i = 0$ implies:

$x^i(\epsilon)$	$s^i(\epsilon)$	Strictly complementary
Μ	$ \ \ \land$	
$\operatorname{int}(\mathbb{L}^{n_i}_+)$	$\{0\}$	Yes
$\{0\}$	$\operatorname{int}(\mathbb{L}^{n_i}_+)$	Yes
$\operatorname{bd}(\mathbb{L}^{n_i}_+)\setminus\{0\}$	$\operatorname{bd}(\mathbb{L}^{n_i}_+)\setminus\{0\}$	Yes
{0}	{0}	No
$\operatorname{bd}(\mathbb{L}^{n_i}_+)\setminus\{0\}$	$\{0\}$	No
{0}	$\operatorname{bd}(\mathbb{L}^{n_i}_+)\setminus\{0\}$	No

▶ The complementarity for linear optimization reduces to only three cases.

Optimal partition

The index set $\{1, \ldots, p\}$ is partitioned into four subsets $\mathcal{B}, \mathcal{N}, \mathcal{R}$, and $\mathcal{T} := (\mathcal{T}_1, \mathcal{T}_2, \mathcal{T}_3)$:

$$\begin{split} \mathcal{B} &:= \left\{ i \mid x_{1}^{i} > \|x_{2:n_{i}}^{i}\|, \text{ for some } x \in \mathcal{P}^{*} \right\},\\ \mathcal{N} &:= \left\{ i \mid s_{1}^{i} > \|s_{2:n_{i}}^{i}\|, \text{ for some } s \in \mathcal{D}^{*} \right\},\\ \mathcal{R} &:= \left\{ i \mid x_{1}^{i} = \|x_{2:n_{i}}^{i}\| > 0, \ s_{1}^{i} = \|s_{2:n_{i}}^{i}\| > 0, \text{ for some } (x; y; s) \in \mathcal{P}^{*} \times \mathcal{D}^{*} \right\},\\ \mathcal{T}_{1} &:= \left\{ i \mid x^{i} = s^{i} = 0, \text{ for all } (x; y; s) \in \mathcal{P}^{*} \times \mathcal{D}^{*} \right\},\\ \mathcal{T}_{2} &:= \left\{ i \mid s^{i} = 0, \text{ for all } (y; s) \in \mathcal{D}^{*}, \ x_{1}^{i} = \|x_{2:n_{i}}^{i}\| > 0, \text{ for some } x \in \mathcal{P}^{*} \right\},\\ \mathcal{T}_{3} &:= \left\{ i \mid x^{i} = 0, \text{ for all } x \in \mathcal{P}^{*}, \ s_{1}^{i} = \|s_{2:n_{i}}^{i}\| > 0, \text{ for some } (y; s) \in \mathcal{D}^{*} \right\}. \end{split}$$

- ▶ Note that $\mathcal{B}, \mathcal{N}, \mathcal{R}$, and \mathcal{T} are mutually disjoint.
- We call $(\mathcal{B}, \mathcal{N}, \mathcal{R}, \mathcal{T})$ the optimal partition.
- A solution $(x^*; y^*; s^*)$ is strictly complementary iff $\mathcal{T} = \emptyset$.

Example $(\mathcal{R}, \mathcal{T}_2 \neq \emptyset)$

- ▶ The cone in pink is *weakly inactive*, i.e., the cone constraint is active with zero Lagrange multiplier.
- ▶ This is a *nondegenerate* optimal solution



Example $(\mathcal{B}, \mathcal{R} \neq \emptyset)$

- ▶ The optimal solution is in the interior of the blue cone.
- ▶ The optimal solution is on the boundary of the pink cone.



Identification of the optimal partition

We define the following condition numbers:

$$\begin{split} \sigma_{\mathcal{B}} &:= \min_{i \in \mathcal{B}} \max_{x \in \mathcal{P}^*} \{ x_1^i - \| x_{2:n_i}^i \| \}, \\ \sigma_{\mathcal{N}} &:= \min_{i \in \mathcal{N}} \max_{(y,s) \in \mathcal{D}^*} \{ s_1^i - \| s_{2:n_i}^i \| \}, \\ \sigma_1 &:= \min\{\sigma_{\mathcal{B}}, \sigma_{\mathcal{N}} \}, \\ \sigma_2 &:= \min_{i \in \mathcal{R}} \max_{(x;y;s) \in \mathcal{P}^* \times \mathcal{D}^*} \{ x_1^i + s_1^i - \| x_{2:n_i}^i + s_{2:n_i}^i \| \}, \\ \sigma_3 &:= \max_{(x;y;s) \in \mathcal{P}^* \times \mathcal{D}^*} \{ \| (x;y;s) \| \}. \end{split}$$

Theorem

Let $(x(\mu), y(\mu), s(\mu))$ be a central solution with

$$\mu < \tilde{\mu} := \min\bigg\{\frac{\sigma_1^2}{2p^2}, \frac{\sigma_1\sigma_2}{4p^2}, \ \frac{1}{p}\bigg(\frac{1}{4\kappa}\min\bigg\{\frac{\sigma_1}{2p}, \frac{\sigma_2}{2p}\bigg\}\bigg)^{\frac{1}{\gamma}}, \ \hat{\mu}\bigg\}.$$

The the optimal partition $(\mathcal{B}, \mathcal{N}, \mathcal{R}, \mathcal{T})$ can be identified from $(x(\mu), y(\mu), s(\mu))$.

Quadratic convergence to the unique optimal solution

We prove quadratic convergence of Newton's method to the unique optimal solution.

- We need the optimal partition $(\mathcal{B}, \mathcal{N}, \mathcal{R}, \mathcal{T})$ to be known.
- We need $(\mathcal{T}_1, \mathcal{T}_2, \mathcal{T}_3)$ to be correctly identified.

Assumption

It is assumed that $\mu < \tilde{\mu}$ allows for a complete identification of $(\mathcal{T}_1, \mathcal{T}_2, \mathcal{T}_3)$.

▶ The optimal partition is used to reformulate the dual problem.

Existence of \mathcal{R}

Assume that the primal and dual nondegeneracy conditions hold.

Lemma

Let $(x^*; y^*; s^*)$ be the unique optimal solution. Then $\mathcal{R} = \emptyset$ implies $\mathcal{T} = \emptyset$.

As a consequence, if $\mathcal{R} = \emptyset$, then

- The unique optimal solution can be obtained by solving two linear systems of equations.
- ▶ The primal and dual problems are easy to solve.

In the sequel, we assume that $\mathcal{R} \neq \emptyset$.

Problem reduction

Assume that $\mathcal{T}_1, \mathcal{T}_3 \neq \emptyset$.

• If we drop $c^i - A_i^T y \in \mathbb{L}_+^{n_i}$ for $i \in \mathcal{T}_1 \cup \mathcal{T}_3$, then we get

$$(\mathbf{D}'_{\text{SOCO}}) \quad \max\left\{b^T y: A_i^T y + s^i = c^i, \ s^i \in \mathbb{L}^{n_i}_+, \ i \in \{1, \dots, p\} \setminus \{\mathcal{T}_1 \cup \mathcal{T}_3\}\right\},\$$

and its dual is written as

$$\begin{aligned} (\mathbf{P}'_{\text{SOCO}}) & \min \Big\{ \sum_{i \in \{1, \dots, p\} \setminus \{\mathcal{T}_1 \cup \mathcal{T}_3\}} (c^i)^T x^i : \\ & \sum_{i \in \{1, \dots, p\} \setminus \{\mathcal{T}_1 \cup \mathcal{T}_3\}} A_i x^i = b, \ x^i \in \mathbb{L}^{n_i}_+, \ i \in \{1, \dots, p\} \setminus \{\mathcal{T}_1 \cup \mathcal{T}_3\} \Big\}. \end{aligned}$$

Let $(\bar{x}; \bar{y}; \bar{s})$ be an optimal solution of (P'_{SOCO}) and (D'_{SOCO}) .

Lemma

 $(\bar{x}; \bar{y}; \bar{s})$ is primal-dual nondegenerate.

Problem reduction

It follows from the optimality conditions that

$$\begin{array}{ll} (x^*)^i = \bar{x}^i, & i \in \mathcal{B} \cup \mathcal{R} \cup \mathcal{T}_2, \\ (x^*)^i = 0, & i \in \mathcal{N} \cup \mathcal{T}_1 \cup \mathcal{T}_3, \\ y^* = \bar{y}, \\ (s^*)^i = c^i - A_i^T \bar{y}, & i \in \mathcal{N} \cup \mathcal{R} \cup \mathcal{T}_3, \\ (s^*)^i = 0, & i \in \mathcal{T}_1 \cup \mathcal{T}_2. \end{array}$$

Thus, if we remove the columns of \mathcal{T}_1 and \mathcal{T}_3

• we can recover the unique optimal solutions of (P) and (D).

Primal reformulation

Let $\nu^i \in \mathbb{R}^{n_i}$ for $i \in \mathcal{B} \cup \mathcal{R} \cup \mathcal{T}_2$.

The unique optimal solution \bar{x} can be obtained by solving

$$\begin{aligned} (\mathbf{P}_{\mathrm{NLO}}) & \min & \sum_{i \in \mathcal{B} \cup \mathcal{R} \cup \mathcal{T}_2} (c_i)^T \nu^i \\ \text{s.t.} & \sum_{i \in \mathcal{B} \cup \mathcal{R} \cup \mathcal{T}_2} A_i \nu^i = b, \\ & (\nu^i)^T R_i \nu^i = 0, \qquad i \in \mathcal{R} \cup \mathcal{T}_2, \\ & \nu \in \mathcal{V}, \end{aligned}$$

where

$$\mathcal{V} := \left\{ \nu \mid \nu_1^i > 0, \ i \in \mathcal{R} \cup \mathcal{T}_2, \ \nu^i \in \operatorname{int}(\mathbb{L}^{n_i}_+), \ i \in \mathcal{B} \right\}.$$

▶ (P_{NLO}) has a unique globally optimal solution.

Dual reformulation

Let $w \in \mathbb{R}^m$, $z^i \in \mathbb{R}^{n_i}$ for $i \in \mathcal{R} \cup \mathcal{N}$.

The unique optimal solution $(\bar{y}; \bar{s})$ is the globally optimal solution of

$$\begin{array}{ll} (\mathrm{D}_{\mathrm{NLO}}) & \min & -b^T w \\ & \mathrm{s.t.} & A_i^T w = c^i, & i \in \mathcal{B} \cup \mathcal{T}_2, \\ & A_i^T w + z^i = c^i, & i \in \mathcal{R} \cup \mathcal{N}, \\ & (z^i)^T R_i z^i = 0, & i \in \mathcal{R}, \\ & z \in \mathcal{W}, \end{array}$$

where

$$\mathcal{W} := \left\{ z \mid z_1^i > 0, \ i \in \mathcal{R}, \ z^i \in \operatorname{int}(\mathbb{L}^{n_i}_+), \ i \in \mathcal{N} \right\}.$$

• (D_{NLO}) has a unique globally optimal solution.

First-order optimality conditions

Let $u^i \in \mathbb{R}^{n_i}$ for $i \in \mathcal{B} \cup \mathcal{T}_2 \cup \mathcal{R} \cup \mathcal{N}$ and $v \in \mathbb{R}^{|\mathcal{R}|}$.

The first-order optimality conditions for (D_{NLO}) are given by

$$\begin{cases} -\sum_{i\in\mathcal{B}\cup\mathcal{T}_2\cup\mathcal{R}\cup\mathcal{N}}A_iu^i &= b,\\ -u^i - 2v_iR_iz^i &= 0, \quad i\in\mathcal{R},\\ -u^i &= 0, \quad i\in\mathcal{N},\\ A_i^Tw &= c^i, \quad i\in\mathcal{B}\cup\mathcal{T}_2,\\ A_i^Tw + z^i &= c^i, \quad i\in\mathcal{R}\cup\mathcal{N},\\ (z^i)^TR_iz^i &= 0, \quad i\in\mathcal{R},\\ z\in\mathcal{W}. \end{cases}$$

 It bears a striking resemblance to the optimality conditions of second-cone program.

Constraint qualification

Let $(\bar{w}; \bar{z})$ be the unique globally optimal solution of (D_{NLO}) .

Lemma

Under the dual nondegeneracy condition, the Jacobian of equality constraints at $(\bar{w}; \bar{z})$ has full row rank.

▶ There exist unique Lagrange multipliers.

Nonsingularity of the Jacobian

The first-order optimality conditions can be written as G((w; z; u; v)) = 0 and $z \in \mathcal{W}$, where

$$G((w; z; u; v)) := \begin{pmatrix} -\sum_{i \in \mathcal{B} \cup \mathcal{T}_2 \cup \mathcal{R} \cup \mathcal{N}} A_i u^i - b \\ -u^i - 2v_i R_i z^i \\ -u^i \\ A_i^T w - c^i \\ A_i^T w + z^i - c^i \\ (z^i)^T R_i z^i \end{pmatrix}.$$

Let $(\bar{u}; \bar{v})$ be the unique Lagrange multipliers associated with $(\bar{w}; \bar{z})$.

Lemma

Under the primal and dual nondegeneracy conditions, the Jacobian ∇G is nonsingular at $(\bar{w}; \bar{z}; \bar{u}; \bar{v})$.

 The primal nondegeneracy leads to a second-order condition at the globally optimal solution.

Quadratic convergence

Let ϵ be the convergence region of Newton's method.

Theorem

Assume that the primal and dual nondegeneracy conditions hold. Let

$$\mu < \min\left\{p^{-1}\left(4\sqrt{2}\beta_2\kappa\left(\sqrt{3} + \frac{2p\sqrt{|\mathcal{R}|}}{\sigma_2}\left(1 + \frac{2\sigma_3}{\sigma_2}\right)\right)\right)^{-\frac{1}{\gamma}}, \ \tilde{\mu}\right\},\$$

in which β_2 denotes an upper bound for $\|\nabla G((\bar{w}; \bar{z}; \bar{u}; \bar{v}))^{-1}\|$. Then Newton's method converges to $(\bar{x}; \bar{y}; \bar{s})$ with quadratic rate.

Discussion

To establish quadratic convergence:

▶ If strict complementarity holds,

$$\mu < \min\left\{p^{-1}\left(4\sqrt{3}\beta_1\kappa\right)^{-\frac{1}{\gamma}}, \hat{\mu}\right\}.$$

▶ If strict complementarity fails,

$$\mu < \min\left\{p^{-1}\left(4\sqrt{2}\beta_2\kappa\left(\sqrt{3} + \frac{2p\sqrt{|\mathcal{R}|}}{\sigma_2}\left(1 + \frac{2\sigma_3}{\sigma_2}\right)\right)\right)^{-\frac{1}{\gamma}}, \ \tilde{\mu}\right\}.$$

- ▶ Quadratic convergence is harder to achieve when strict complementarity fails.
- μ has to be small enough so that the optimal partition can be identified.

Conclusions:

Under the primal and dual nondegeneracy conditions:

- We proved quadratic convergence of Newton's method to the strict complementarity solution.
- We proved quadratic convergence of Newton's method to the maximally complementary solution.

Future directions:

- Strong second-order conditions to release the assumption on the identification of $(\mathcal{T}_1, \mathcal{T}_2, \mathcal{T}_3)$.
- ▶ Quadratic convergence to the unique optimal solution of semidefinite optimization using the optimal partition.

If you would like to see more about parametric second-order cone optimization, come to the other session tomorrow.

Thank you for your attention Any questions?