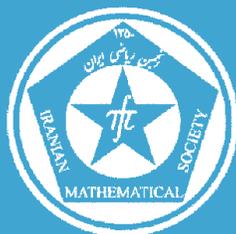


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**Title:**

**An improved infeasible interior-point method for  
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## AN IMPROVED INFEASIBLE INTERIOR-POINT METHOD FOR SYMMETRIC CONE LINEAR COMPLEMENTARITY PROBLEM

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**ABSTRACT.** We present an improved version of a full Nesterov-Todd step infeasible interior-point method for linear complementarity problem over symmetric cone (Bull. Iranian Math. Soc., 40, (2014), no. 3, 541–564). In the earlier version, each iteration consisted of one so-called feasibility step and a few -at most three - centering steps. Here, each iteration consists of only a feasibility step. Thus, the new algorithm demands less work in each iteration and admits a simple analysis of complexity bound. The complexity result coincides with the best-known iteration bound for infeasible interior-point methods.

**Keywords:** Linear complementarity problem, infeasible interior-point method, symmetric cones, polynomial complexity.

**MSC(2010):** Primary: 90C33; Secondary: 90C51.

### 1. Introduction

Consider a Euclidean Jordan algebra  $(\mathcal{J}, \circ, \langle \cdot, \cdot \rangle)$ , where  $\circ$  denotes the Jordan product and  $\mathcal{J}$  is a finite-dimensional vector space over the real field  $R$  equipped with the inner product  $\langle \cdot, \cdot \rangle$ , and let  $\mathcal{K}$  be a symmetric cone in  $\mathcal{J}$ . The monotone linear complementarity problem over symmetric cone (SCLCP) requires the computation of a vector pair  $(x, s) \in \mathcal{K} \times \mathcal{K}$  satisfying

$$(1.1) \quad s = Mx + q, \quad x \circ s = 0,$$

where  $q \in R^n$  and  $M \in R^{n \times n}$  such that  $v = Mu$  implies that  $\langle u, v \rangle \geq 0$ . Although SCLCP is not an optimization problem, it is closely related to one, because optimality conditions of several important optimization problems can be written as an SCLCP; e.g., linear optimization (LO) problem over symmetric cone (SCO). Faybusovich was the first to analyze a short-step path-following

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interior-point method (IPM) for SCLCP [2, 3]. In addition to Faybusovich's results, Rangarajan [10] proposed the first infeasible IPM (IIPM) for SCLCP. The primal-dual full-Newton step feasible IPM for LO was first analyzed by Roos et al. [12] and was later extended to infeasible version by Roos [11]. Both versions of the method were extended by Kheirfam and Mahdavi-Amiri [8] to SCLCP by using the Nesterov-Todd (NT) direction as a search direction. The obtained iteration bounds coincide with the ones derived for LO, currently being best known iteration bounds for SCLCP. Subsequently, both versions were extended by Kheirfam and Mahdavi-Amiri [6, 7] to SCLCP based on modified NT-directions using Euclidean Jordan algebra. Recently, Roos [13] proposed a new method for LO by improving the full-Newton step IIPMs so that the centering steps not be needed, whereas the above-mentioned methods require a few (at most three) centering steps in each (main) iteration. Motivated by Roos' recent work, we present a new full-NT step IIPM for SCLCP which uses only a full step in each (main) iteration. The new algorithm starts from an infeasible point, located in a small neighborhood of the central path of a perturbed SCLCP. Then, after a full-NT step the new iterate is well-centered for the new perturbed SCLCP. This kind of strategy reduces the number of iterations and the resulting complexity coincides with the best known bound, while tendering a simple analysis.

In what follows, we briefly recall some concepts, properties, and results from Euclidean Jordan algebras as needed. A comprehensive treatment of Euclidean Jordan algebra can be found in [1].

A Euclidean Jordan algebra is a triple  $(\mathcal{J}, \circ, \langle \cdot, \cdot \rangle)$ , where  $(\mathcal{J}, \langle \cdot, \cdot \rangle)$  is an  $n$ -dimensional inner product space over  $R$  and  $(x, y) \rightarrow x \circ y$  on  $\mathcal{J}$  is a bilinear mapping satisfying the following conditions for all  $x, y, z \in \mathcal{J}$ :

- (i)  $x \circ y = y \circ x$ ,
- (ii)  $x \circ (x^2 \circ y) = x^2 \circ (x \circ y)$ , where  $x^2 = x \circ x$ ,
- (iii)  $\langle x \circ y, z \rangle = \langle x, y \circ z \rangle$ ,

where the inner product  $\langle \cdot, \cdot \rangle$  is defined by  $\langle x, y \rangle := \text{tr}(x \circ y)$  for any  $x, y \in \mathcal{J}$ .

Note that, by Theorem III.2.1 of [1], the symmetric cone  $\mathcal{K}$  coincides with the set of squares  $\{x^2 : x \in \mathcal{J}\}$  of some Euclidean Jordan algebra  $\mathcal{J}$ . We assume that there exists an element  $e$  such that  $x \circ e = e \circ x = x$ , for all  $x \in \mathcal{J}$ . The rank of  $(\mathcal{J}, \circ, \langle \cdot, \cdot \rangle)$  is defined to be

$$r := \max \{ \deg(x) : x \in \mathcal{J} \},$$

where  $\deg(x)$  is the degree of  $x \in \mathcal{J}$ , given by

$$\deg(x) := \min \{ k : \{e, x, \dots, x^k\} \text{ are linearly dependent} \}.$$

For any  $x \in \mathcal{J}$ , the Lyapunov transformation  $L(x) : \mathcal{J} \rightarrow \mathcal{J}$  is defined as  $L(x)y = x \circ y$ , for all  $y \in \mathcal{J}$ . It follows from (i) and (iii) above that the Lyapunov transformation is symmetric; i.e.,  $\langle L(x)y, z \rangle = \langle y, L(x)z \rangle$  holds for

all  $y, z \in \mathcal{J}$ . Specially,  $L(x)e = x$  and  $L(x)x = x^2$ , for  $x \in \mathcal{J}$ . Using the Lyapunov transformation, the quadratic representation of  $x \in \mathcal{J}$  is defined as

$$P(x) := 2L(x)^2 - L(x^2),$$

where  $L(x)^2 = L(x)L(x)$ . For any  $x \in \mathcal{K}$ ,  $L(x)$  is positive semidefinite; i.e.,  $\langle L(x)x, x \rangle \geq 0$ , and  $\langle x, y \rangle = 0$  if and only if  $x \circ y = 0$ , for any  $x, y \in \mathcal{K}$  (Lemma 2.2 of [2]). An element  $c \in \mathcal{J}$  is idempotent if  $c \circ c = c \neq 0$ , which is also primitive if it cannot be written as a sum of two idempotents. We say that  $\{c_1, c_2, \dots, c_k\}$  is a Jordan frame, if each  $c_i$  is a primitive idempotent,  $c_i \circ c_j = 0$  ( $i \neq j$ ) and  $\sum_{i=1}^k c_i = e$ . Let  $(\mathcal{J}, \circ, \langle \cdot, \cdot \rangle)$  be a Euclidean Jordan algebra with  $\text{rank}(\mathcal{J}) = r$ . Then, for any  $x \in \mathcal{J}$ , there exist a Jordan frame  $\{c_1, c_2, \dots, c_r\}$  and real numbers  $\lambda_1(x), \lambda_2(x), \dots, \lambda_r(x)$  such that  $x = \sum_{i=1}^r \lambda_i(x)c_i$  (spectral decomposition, Theorem III.1.2 of [1]). Every  $\lambda_i(x)$  is called an eigenvalue of  $x$ . We denote the minimum and maximum eigenvalues of  $x$  by  $\lambda_{\min}(x)$  and  $\lambda_{\max}(x)$ , respectively. If two elements  $x$  and  $y$  share the same Jordan frames in the spectral decompositions, then they operator commute; i.e., they satisfy  $L(x)L(y) = L(y)L(x)$  (Theorem 27 of [14]). The trace of  $x$  is defined to be  $\text{tr}(x) := \sum_{i=1}^r \lambda_i(x)$  and the Frobenius norm of  $x$  is  $\|x\|_F := \sqrt{\langle x, x \rangle} = \sqrt{\lambda_1(x)^2 + \dots + \lambda_r(x)^2}$ . Observe that  $\|e\|_F = \sqrt{r}$ . We also see that  $x \in \mathcal{K}$  (respectively,  $x \in \text{int}\mathcal{K}$ ) if and only if  $\lambda_i(x) \geq 0$  (respectively,  $\lambda_i(x) > 0$ ), for all  $i = 1, 2, \dots, r$ . For any  $x \in \mathcal{J}$ , having the spectral decomposition  $x = \sum_{i=1}^r \lambda_i(x)c_i$ , we denote

$$x^{\frac{1}{2}} := \sum_{i=1}^r \sqrt{\lambda_i(x)}c_i, \text{ if } \lambda_i(x) \geq 0 \text{ and } x^{-1} := \sum_{i=1}^r \lambda_i(x)^{-1}c_i, \text{ if } \lambda_i(x) \neq 0.$$

The remainder of our work is organized as follows. In the next section, we introduce the perturbed problem and describe our proposed algorithm. Section 3 gives an analysis of the algorithm. In subsection 3.1, we derive an upper bound for the proximity measure after a full step. Subsection 3.2 serves to derive an upper bound for  $\omega(v)$ . In subsection 3.3, we fix the values of the parameters  $\theta$  and  $\tau$  in the algorithm. Here,  $\tau$  is a uniform upper bound for the values of the proximity measure,  $\delta(x, s; \mu)$ , occurring during the course of the algorithm, and  $\theta$  determines the progress to feasibility and optimality of the iterates. As a result, we realize the algorithm to be well-defined for the chosen values of  $\theta$  and  $\tau$ . Finally, we derive the complexity of the algorithm coinciding with the best known iteration bound for IIPMs.

## 2. Infeasible full-NT step IPM

In the case of an infeasible method, we call the pair  $(x, s)$  an  $\epsilon$ -solution of (1.1) if the Frobenius norm of the residual vector  $s - Mx - q$  does not exceed  $\epsilon$ , and also  $\langle x, s \rangle := \text{tr}(x \circ s) \leq \epsilon$ .

**2.1. The perturbed problem.** In accordance with the available results on IIPMs (e.g., see [8]), it is assumed that there exists a solution  $(x^*, s^*)$  such that

$$(2.1) \quad \|x^*\|_\infty \leq \rho_p, \max\{\|s^*\|_\infty, \|\rho_p M e + q\|_F\} \leq \rho_d,$$

where  $\rho_p$  and  $\rho_d$  are positive numbers. Furthermore, we choose arbitrarily  $(x^0, s^0) \in \text{int}\mathcal{K}$  and  $\mu^0 > 0$  such that

$$(2.2) \quad x^0 = \rho_p e, \quad s^0 = \rho_d e, \quad \text{and} \quad \mu^0 = \rho_p \rho_d,$$

as the starting point of the algorithm, where  $\rho_p$  and  $\rho_d$  are defined as in (2.1). The initial value of the residual vector is denoted to be  $r_q^0 = s^0 - Mx^0 - q$ . In general, we have  $r_q^0 \neq 0$ . However, a sequence of perturbed problems is constructed below in a such a way that the initial iterate is strictly feasible for the first perturbed problem in the sequence.

For any  $\nu$ , with  $0 < \nu \leq 1$ , we consider the perturbed problem (SCLCP $_\nu$ ), defined by

$$s - Mx - q = \nu r_q^0, \quad (x, s) \in \mathcal{K} \times \mathcal{K} \quad (\text{SCLCP}_\nu).$$

It is obvious that  $(x, s) = (x^0, s^0)$  is a strictly feasible solution of (SCLCP $_\nu$ ), when  $\nu = 1$ . This means that the perturbed problem (SCLCP $_\nu$ ) satisfies the interior point condition (IPC), for  $\nu = 1$ ; i.e., there exists  $(x^0, s^0) \in \text{int}\mathcal{K} \times \text{int}\mathcal{K}$  such that  $s^0 - Mx^0 - q = \nu r_q^0$ , which then straightforwardly leads to the following result.

**Theorem 2.1.** (Theorem 3.1 of [8]) *Let (1.1) be feasible and  $0 < \nu \leq 1$ . Then, the perturbed problem (SCLCP $_\nu$ ) satisfies the IPC.*

Let (1.1) be feasible and  $0 < \nu \leq 1$ . Then, Theorem 2.1 implies that the problem (SCLCP $_\nu$ ) satisfies the IPC, for  $0 < \nu \leq 1$ , and hence a corresponding central path exists. This means that the system

$$(2.3) \quad \begin{aligned} s - Mx - q &= \nu r_q^0, \quad (x, s) \in \mathcal{K} \times \mathcal{K}, \\ x \circ s &= \mu e, \end{aligned}$$

has a unique solution for every  $\mu > 0$ , as the  $\mu$ -center of the perturbed problem (SCLCP $_\nu$ ). The set of  $\mu$ -centers is called the central path. In what follows, the parameters  $\mu$  and  $\nu$  always satisfy the relation  $\mu = \nu \mu^0 = \nu \rho_p \rho_d$ . It is also worth noting that, according to (2.2),  $x^0 \circ s^0 = \mu^0 e$ ; hence,  $(x^0, s^0)$  is the  $\mu^0$ -center of the perturbed problem (SCLCP $_\nu$ ) for  $\nu = 1$ . Therefore, the algorithm can easily be started since by construction we have the initial starting point lying exactly on the central path of (SCLCP $_\nu$ ) for  $\nu = 1$ .

Let  $(x, s)$  be a feasible solution of (SCLCP $_\nu$ ), and  $\mu = \nu \mu^0$ . Then, we measure proximity to the  $\mu$ -center of the perturbed problem (SCLCP $_\nu$ ) by the quantity

$$\delta(x, s; \mu) := \delta(v) := \frac{1}{2} \|v^{-1} - v\|_F, \quad \text{where } v := \frac{P(w^{-\frac{1}{2}})x}{\sqrt{\mu}} \left[ = \frac{P(w^{\frac{1}{2}})s}{\sqrt{\mu}} \right],$$

and  $w := P(x^{\frac{1}{2}})(P(x^{\frac{1}{2}})s)^{-\frac{1}{2}} [= P(s^{-\frac{1}{2}})(P(s^{\frac{1}{2}})x)^{\frac{1}{2}}]$  is called the scaling point of  $x$  and  $s$  (Lemma 3.2 of [4]). As a consequence, we have the following lemma.

**Lemma 2.2.** (Lemma 4.7 of [5]) *If  $\delta := \delta(v)$ , then*

$$\frac{1}{\rho(\delta)} \leq \lambda_{\min}(v) \leq \lambda_{\max}(v) \leq \rho(\delta), \text{ where } \rho(\delta) := \delta + \sqrt{1 + \delta^2}.$$

**2.2. An iteration of our algorithm.** Suppose that for some  $\mu \in (0, \mu^0]$  we have  $x$  and  $s$  satisfying the first equation in (2.3) for  $\nu = \frac{\mu}{\mu^0}$  and such that  $\delta(x, s; \mu) \leq \tau$ . This certainly holds at the start of the first iteration, since  $s^0 - Mx^0 - q = \nu r_q^0$ , when  $\nu = 1$  and  $\delta(x^0, s^0; \mu^0) = 0$ . We reduce  $\mu$  to  $\mu^+ := (1 - \theta)\mu$  and  $\nu$  to  $\nu^+ := (1 - \theta)\nu$  with  $\theta \in (0, 1)$ , and find new iterates  $x^+$  and  $s^+$  satisfying the first equation in (2.3), with  $\mu$  replaced by  $\mu^+$  and  $\nu$  by  $\nu^+ = \frac{\mu^+}{\mu^0}$ , and such that  $\delta(x^+, s^+; \mu^+) \leq \tau$ . Note that  $\nu^+ = (1 - \theta)\nu$ . So, the relation  $\mu = \nu\mu^0$  is maintained in every iteration.

Suppose that we have strictly feasible iterates  $x$  and  $s$  for  $(SCLCP_\nu)$ . This means that  $(x, s)$  satisfies the first equation of (2.3) with  $x \in \text{int}\mathcal{K}$  and  $s \in \text{int}\mathcal{K}$ . With  $\nu$  replaced by  $\nu^+ = (1 - \theta)\nu$ , we find displacements  $\Delta x$  and  $\Delta s$  such that

$$(2.4) \quad \begin{aligned} M\Delta x - \Delta s &= \theta\nu r_q^0, \\ s \circ \Delta x + x \circ \Delta s &= \mu e - x \circ s. \end{aligned}$$

Due to the fact that  $L(x)L(s) \neq L(s)L(x)$ , in general, the above system does not always have a unique solution. To overcome this difficulty, the second equation of the system (2.3) is replaced by the following equivalent scaled equation (see Lemma 28 of [14]):

$$P(w^{-\frac{1}{2}})x \circ P(w^{\frac{1}{2}})s = \mu e,$$

where  $w$  is the NT-scaling point of  $x$  and  $s$ . This scaling point was first proposed by Nesterov and Todd [9] for self-scaled cones and then adapted by Faybusovich [3] for symmetric cones. In this case, the system (2.4) becomes

$$(2.5) \quad \begin{aligned} M\Delta x - \Delta s &= \theta\nu r_q^0, \\ P(w^{\frac{1}{2}})s \circ P(w^{-\frac{1}{2}})\Delta x + P(w^{-\frac{1}{2}})x \circ P(w^{\frac{1}{2}})\Delta s &= \\ &= \mu e - P(w^{-\frac{1}{2}})x \circ P(w^{\frac{1}{2}})s. \end{aligned}$$

It is easily seen that  $x^+ := x + \Delta x$  and  $s^+ := s + \Delta s$  satisfy  $s - Mx - q = \nu^+ r_q^0$ . The main part of the analysis is to guarantee that  $x^+ \in \text{int}\mathcal{K}$  and  $s^+ \in \text{int}\mathcal{K}$  satisfy  $\delta(x^+, s^+; \mu^+) \leq \tau$ .

**2.3. The algorithm.** A formal description of the new algorithm is given in Figure 1.

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Infeasible interior – point algorithm

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**Input :**  
accuracy parameter  $\epsilon > 0$ ;  
barraier update parameter  $\theta$ ,  $0 < \theta < 1$ ;

**begin**  
 $x := \rho_p e$ ;  $s := \rho_d e$ ;  $\nu = 1$ ;  $\mu = \rho_p \rho_d$ ;

**while**  $\max(\text{tr}(x \circ s), \|r_q\|_F) > \epsilon$  **do**  
**begin**  
 $(x, s) := (x, s) + (\Delta x, \Delta s)$ ;  
 $\mu := (1 - \theta)\mu$ ;  $\nu := (1 - \theta)\nu$ ;  
**end**

**end**

---

Figure 1 : The algorithm

### 3. Analysis of the algorithm

Let  $x$  and  $s$  denote the iterates at the start of an iteration, and assume  $\delta(x, s; \mu) \leq \tau$ .

**3.1. Upper bound for  $\delta(v^+)$ .** As established in subsection 2.2, the full-NT step generates new iterates  $x^+$  and  $s^+$  that satisfy the feasibility condition for  $(SCLCP_{\nu^+})$ , except for possibly the constraints on the cone  $\mathcal{K}$ . A crucial element in the analysis is to show that, after the full-NT step,  $\delta(x^+, s^+; \mu^+) \leq \tau$ .

Defining

$$(3.1) \quad d_x := \frac{P(w^{-\frac{1}{2}})\Delta x}{\sqrt{\mu}}, \quad d_s := \frac{P(w^{\frac{1}{2}})\Delta s}{\sqrt{\mu}},$$

the second equation of (2.5) turns to

$$d_x + d_s = v^{-1} - v.$$

In this case, using the definition of  $v$  and (3.1), we get

$$(3.2) \quad x^+ = \sqrt{\mu}P(w^{\frac{1}{2}})(v + d_x), \quad s^+ = \sqrt{\mu}P(w^{-\frac{1}{2}})(v + d_s).$$

Since  $P(w^{\frac{1}{2}})$  and its inverse  $P(w^{-\frac{1}{2}})$  are automorphisms of  $\text{int}\mathcal{K}$ ,  $x^+$  and  $s^+$  belong to  $\text{int}\mathcal{K}$  if and only if  $v + d_x$  and  $v + d_s$  belong to  $\text{int}\mathcal{K}$ . We have

$$(3.3) \quad \begin{aligned} (v + d_x) \circ (v + d_s) &= v^2 + v \circ (d_x + d_s) + d_x \circ d_s \\ &= e + d_x \circ d_s. \end{aligned}$$

**Lemma 3.1.** *The iterate  $(x^+, s^+)$  is strictly feasible if  $e + d_x \circ d_s \in \text{int}\mathcal{K}$ .*

*Proof.* The proof is similar to the proof of Lemma 4.2 in [5], and is therefore omitted.  $\square$

**Corollary 3.2.** *The iterate  $(x^+, s^+)$  is strictly feasible if  $\|\lambda(d_x \circ d_s)\|_\infty < 1$ .*

*Proof.* By Lemma 3.1, the iterate  $(x^+, s^+)$  is strictly feasible if  $e + d_x \circ d_s \in \text{int}\mathcal{K}$ . If  $\|\lambda(d_x \circ d_s)\|_\infty < 1$ . Then we have  $-1 < \lambda_i(d_x \circ d_s) < 1$ , for all  $i = 1, \dots, r$ . Therefore,

$$\lambda_i(e + d_x \circ d_s) = 1 + \lambda_i(d_x \circ d_s) > 0, \quad i = 1, \dots, r.$$

The last inequalities mean that  $e + d_x \circ d_s \in \text{int}\mathcal{K}$ , and the proof follows.  $\square$

In the sequel, we use the notation

$$(3.4) \quad \omega(v) := \frac{1}{2}(\|d_x\|_F^2 + \|d_s\|_F^2).$$

It follows, from Lemmas 2.16 and 2.12 of [5], that

$$(3.5) \quad \begin{aligned} \|\lambda(d_x \circ d_s)\|_\infty &\leq \|d_x \circ d_s\|_F \leq \frac{1}{2}\|d_x^2 + d_s^2\|_F \\ &\leq \frac{1}{2}(\|d_x\|_F^2 + \|d_s\|_F^2) = \omega(v). \end{aligned}$$

**Corollary 3.3.** *If  $\omega(v) < 1$ , then the iterate  $(x^+, s^+)$  is strictly feasible.*

*Proof.* Due to (3.5),  $\omega(v) < 1$  implies  $\|\lambda(d_x \circ d_s)\|_\infty < 1$ . By Corollary 3.2, the proof is complete.  $\square$

Assuming  $\omega(v) < 1$ , which guarantees strict feasibility of the iterate  $(x^+, s^+)$ , we proceed by deriving an upper bound for  $\delta(x^+, s^+; \mu^+)$ . By definition, we have

$$\delta(x^+, s^+; \mu^+) := \frac{1}{2}\|v^+ - (v^+)^{-1}\|_F,$$

where  $v^+ := \frac{P((w^+)^{-\frac{1}{2}})x^+}{\sqrt{\mu^+}} \left[ = \frac{P((w^+)^{\frac{1}{2}})s^+}{\sqrt{\mu^+}} \right]$ . In what follows, we denote  $\delta(x^+, s^+; \mu^+)$  shortly by  $\delta(v^+)$ .

**Lemma 3.4.** *Let  $\omega(v) < 1$ . Then, we have*

$$\delta(v^+) \leq \frac{\theta\sqrt{r} + \omega(v)}{2\sqrt{(1-\theta)(1-\omega(v))}}.$$

*Proof.* Since  $\omega(v) < 1$ , using corollaries 3.3 and 3.2, Lemma 3.1, (3.3) and (3.2), it follows that  $v + d_x, v + d_s$  and  $(v + d_x) \circ (v + d_s)$  belong to  $\text{int}\mathcal{K}$ . Similar to the proof of lemma 3.3 of [8], we have

$$\sqrt{1-\theta}v^+ \sim [P(v + d_x)^{\frac{1}{2}}(v + d_s)]^{\frac{1}{2}}.$$

Using Theorem 4 of [15] and (3.5), we have

$$\begin{aligned}
\lambda_{\min}(v^+)^2 &= \lambda_{\min}\left(P\left(\frac{v+d_x}{\sqrt{1-\theta}}\right)^{\frac{1}{2}}\left(\frac{v+d_s}{\sqrt{1-\theta}}\right)\right) \\
&\geq \lambda_{\min}\left(\left(\frac{v+d_x}{\sqrt{1-\theta}}\right) \circ \left(\frac{v+d_s}{\sqrt{1-\theta}}\right)\right) \\
&= \frac{\lambda_{\min}(e+d_x \circ d_s)}{1-\theta} \\
&= \frac{1+\lambda_{\min}(d_x \circ d_s)}{1-\theta} \\
&\geq \frac{1-\|d_x \circ d_s\|_F}{1-\theta} \\
&\geq \frac{1-\omega(v)}{1-\theta}.
\end{aligned}$$

Hence, using Lemma 2.9 of [10], the above inequality, Lemma 30 of [14], (3.3), the triangle inequality and (3.5), we may write

$$\begin{aligned}
2\delta(v^+) &= \|v^+ - (v^+)^{-1}\|_F \\
&\leq \frac{\|(v^+)^2 - e\|_F}{\lambda_{\min}(v^+)} \\
&\leq \frac{\sqrt{1-\theta}}{\sqrt{1-\omega(v)}} \left\| P\left(\frac{v+d_x}{\sqrt{1-\theta}}\right)^{\frac{1}{2}} \left(\frac{v+d_s}{\sqrt{1-\theta}}\right) - e \right\|_F \\
&\leq \frac{\sqrt{1-\theta}}{\sqrt{1-\omega(v)}} \left\| \left(\frac{v+d_x}{\sqrt{1-\theta}}\right) \circ \left(\frac{v+d_s}{\sqrt{1-\theta}}\right) - e \right\|_F \\
&= \frac{\sqrt{1-\theta}}{\sqrt{1-\omega(v)}} \left\| \frac{e+d_x \circ d_s}{1-\theta} - e \right\|_F \\
&= \frac{\|\theta e + d_x \circ d_s\|_F}{\sqrt{(1-\theta)(1-\omega(v))}} \\
&\leq \frac{\theta\sqrt{r} + \omega(v)}{\sqrt{(1-\theta)(1-\omega(v))}}.
\end{aligned}$$

This completes the proof.  $\square$

**3.2. Upper bound for  $\omega(v)$ .** We may easily check that the system (2.5), defining the search directions  $\Delta x$  and  $\Delta s$ , can be expressed in terms of the scaled search directions  $d_x$  and  $d_s$  as follows:

$$\begin{aligned}
(3.6) \quad \bar{M}d_x - d_s &= \frac{P(w^{\frac{1}{2}})}{\sqrt{\mu}} \theta \nu r_q^0, \\
d_x + d_s &= v^{-1} - v,
\end{aligned}$$

where  $\overline{M} := P(w)^{\frac{1}{2}}MP(w)^{\frac{1}{2}}$  is positive semidefinite [8].

**Lemma 3.5.** *Let  $a \in R^n$  and  $\overline{M} \in R^{n \times n}$  be positive semidefinite. Then, the solution  $(u, v)$  of the linear system*

$$(3.7) \quad \begin{aligned} \overline{M}u - v &= 0, \\ u + v &= a, \end{aligned}$$

satisfies

$$\|u\|_F^2 + \|v\|_F^2 \leq \|a\|_F^2.$$

*Proof.* We have

$$\|u\|_F^2 + \|v\|_F^2 = \|u + v\|_F^2 - 2\langle u, v \rangle \leq \|u + v\|_F^2 = \|a\|_F^2.$$

The inequality follows from  $\overline{M}$  being positive semidefinite to complete the proof.  $\square$

**Lemma 3.6.** *Let  $a$  and  $b$  be two  $n$ -dimensional vectors, and let  $\overline{M} \in R^{n \times n}$  be positive semidefinite. Then, the solution  $(u, v)$  of the linear system*

$$(3.8) \quad \begin{aligned} \overline{M}u - v &= b, \\ u + v &= a, \end{aligned}$$

satisfies

$$\|u\|_F^2 + \|v\|_F^2 \leq 4\|a\|_F^2 + 6\|b\|_F^2.$$

*Proof.* It is easily seen that the system (3.8) can be written as

$$(3.9) \quad \begin{aligned} \overline{M}u - \tilde{v} &= 0, \\ u + \tilde{v} &= a + b, \end{aligned}$$

where  $\tilde{v} = v + b$ . Applying Lemma 3.5 to (3.9), it follows that

$$\|u\|_F^2 + \|\tilde{v}\|_F^2 \leq \|a + b\|_F^2 = \|a\|_F^2 + \|b\|_F^2 + 2\langle a, b \rangle \leq 2(\|a\|_F^2 + \|b\|_F^2).$$

Therefore, using the above inequality, we get

$$\begin{aligned} \|u\|_F^2 + \|v\|_F^2 &= \|u\|_F^2 + \|\tilde{v} - b\|_F^2 \\ &\leq 2\|u\|_F^2 + 2(\|\tilde{v}\|_F^2 + \|b\|_F^2) \leq 4\|a\|_F^2 + 6\|b\|_F^2. \end{aligned}$$

This completes the proof.  $\square$

Comparing the system (3.8) with the system (3.6) and considering  $(u, v) = (d_x, d_s)$ ,  $b = \frac{P(w^{\frac{1}{2}})}{\sqrt{\mu}}\theta\nu r_q^0$  and  $a = v^{-1} - v$  in (3.8), by using Lemma 3.6 and

$$\begin{aligned} \|P(w^{\frac{1}{2}})r_q^0\|_F^2 &= \langle P(w^{\frac{1}{2}})r_q^0, P(w^{\frac{1}{2}})r_q^0 \rangle = \langle P(w)r_q^0, r_q^0 \rangle \\ &= \langle P(w)r_q^0, 2\rho_d e \rangle - \langle P(w)r_q^0, 2\rho_d e - r_q^0 \rangle \leq \langle P(w)r_q^0, 2\rho_d e \rangle \\ &= \langle P(w)(2\rho_d e), 2\rho_d e \rangle - \langle P(w)(2\rho_d e - r_q^0), 2\rho_d e \rangle \\ &\leq \langle P(w)(2\rho_d e), 2\rho_d e \rangle = 4\rho_d^2 \langle P(w)e, e \rangle = 4\rho_d^2 \text{tr}(w^2), \end{aligned}$$

where the inequalities hold since  $x^0 = \rho_p e$ ,  $s^0 = \rho_d e$  and  $\|\rho_p M e + q\|_F \leq \rho_d$  imply  $2\rho_d e \succeq r_q^0 \succeq 0$ , we get

$$\begin{aligned}\omega(v) &\leq 2\|v^{-1} - v\|_F^2 + \frac{3\theta^2\nu^2}{\mu}\|P(w^{\frac{1}{2}})r_q^0\|_F^2 \\ &\leq 8\delta^2 + \frac{12\theta^2\nu^2\rho_d^2}{\mu}\text{tr}(w^2).\end{aligned}$$

Now, using the inequality  $\text{tr}(w^2) \leq \frac{\text{tr}(x^2)}{\mu\lambda_{\min}(v)^2}$  (Lemma 4.5 of [5]), Lemma 2.2,  $\text{tr}(x) \leq r\rho_p(2 + \rho(\delta)^2)$  (Lemma 3.7 of [8]) and  $\mu = \nu\rho_p\rho_d$ , we obtain:

$$\begin{aligned}\omega(v) &\leq 8\delta^2 + \frac{12\theta^2\nu^2\rho_d^2}{\mu} \times \frac{\text{tr}(x^2)}{\mu\lambda_{\min}(v)^2} \leq 8\delta^2 + \frac{12\theta^2\rho(\delta)^2}{\rho_p^2}\text{tr}(x)^2 \\ (3.10) \quad &\leq 8\delta^2 + 12\theta^2r^2\rho(\delta)^2(2 + \rho(\delta)^2)^2.\end{aligned}$$

**3.3. Values for  $\theta$  and  $\tau$ .** Our aim is to find a positive number  $\tau$  such that if  $\delta := \delta(v) \leq \tau$ , then  $\delta(v^+) \leq \tau$ . By Lemma 3.4, this holds if  $\omega(v) < 1$  and

$$(3.11) \quad \frac{\theta\sqrt{r} + \omega(v)}{2\sqrt{(1-\theta)(1-\omega(v))}} \leq \tau.$$

Assuming  $\delta(v) \leq \tau$ , we therefore need to find  $\tau$  such that the above inequality holds, with  $\theta$  as large as possible. We choose

$$(3.12) \quad \theta = \frac{1}{46r}, \quad \tau = \frac{1}{16}.$$

Using  $\delta \leq \tau$ , it follows from (3.10), with the right-hand side of (3.10) being monotonically increasing with respect to  $\delta$ , that

$$\begin{aligned}\omega(v) &\leq 8\tau^2 + 12\theta^2r^2\rho(\tau)^2(2 + \rho(\tau)^2)^2 \\ &= 8\left(\frac{1}{16}\right)^2 + 12\left(\frac{1}{46r}\right)^2r^2\rho\left(\frac{1}{16}\right)^2\left(2 + \rho\left(\frac{1}{16}\right)^2\right)^2 \\ &= 0.0943 < 1.\end{aligned}$$

The above inequality means, using Corollary 3.3, that the iterate  $(x^+, s^+)$  is strictly feasible. From Lemma 3.4, it follows that

$$\begin{aligned}\delta(v^+) &\leq \frac{\theta\sqrt{r} + \omega(v)}{2\sqrt{(1-\theta)(1-\omega(v))}} \leq \frac{\frac{1}{46\sqrt{r}} + 0.0943}{2\sqrt{(1-\frac{1}{46r})(1-0.0943)}} \\ &= 0.0617 < \frac{1}{16} = \tau.\end{aligned}$$

This implies that (3.11) holds. Therefore, the algorithm is well-defined in the sense that the property  $\delta(x, s; \mu) \leq \tau$  is maintained in all iterations.

**3.4. Complexity.** We have found that if at the start of an iteration the iterate satisfies  $\delta(x, s; \mu) \leq \tau$  and  $\tau$  and  $\theta$  are defined as in (3.12), then after the full-NT step, the iterate is strictly feasible and satisfies  $\delta(x^+, s^+; \mu^+) \leq \tau$ . This establishes that the algorithm is well-defined.

In each main iteration, both the barrier parameter  $\mu$  and the norm of the residual vector are reduced by the factor  $1 - \theta$ . Hence, the total number of main iterations is bounded above by

$$\frac{1}{\theta} \log \frac{\max\{\text{tr}(x^0 \circ s^0), \|r_q^0\|_F\}}{\epsilon}.$$

Thus, we next state the main result of our work.

**Theorem 3.7.** *If (1.1) has a solution  $(x^*, s^*)$  such that  $\|x^*\|_\infty \leq \rho_p$  and  $\|s^*\|_\infty \leq \rho_d$ , then after at most*

$$46r \log \frac{\max\{\text{tr}(x^0 \circ s^0), \|r_q^0\|_F\}}{\epsilon}$$

*iterations the algorithm finds an  $\epsilon$ -solution of (1.1).*

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