

CHARACTERIZATIONS OF TILT-STABLE MINIMIZERS IN
SECOND-ORDER CONE PROGRAMMING*

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Abstract. This paper is devoted to the study of tilt stability of local minimizers, which plays an important role in both theoretical and numerical aspects of optimization. This notion has been comprehensively investigated in the unconstrained framework as well as for problems of nonlinear programming with C^2 -smooth data. Available results for nonpolyhedral conic programs were obtained only under strong constraint nondegeneracy assumptions. Here we develop an approach of second-order variational analysis, which allows us to establish complete neighborhood and point-based characterizations of tilt stability for problems of second-order cone programming generated by the nonpolyhedral second-order/Lorentz/ice-cream cone. These characterizations are established under the weakest metric subregularity constraint qualification condition.

Key words. second-order variational analysis, second-order cone programming, tilt stability, normal cone mappings, graphical derivatives, C^2 -cone reducibility, metric subregularity

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1. Introduction and brief overview. This paper mainly aims at deriving efficient characterizations of tilt-stable local minimizers in problems of *second-order cone programming* (SOCP) given as follows:

$$(1.1) \quad \text{minimize } f(x) \text{ subject to } x \in \Gamma := \{x \in \mathbb{R}^n \mid g(x) \in \mathcal{Q}\},$$

where $f: \mathbb{R}^n \rightarrow \mathbb{R}$ and $g: \mathbb{R}^n \rightarrow \mathbb{R}^{1+m}$ are C^2 -smooth around the reference point \bar{x} , and where

$$(1.2) \quad \mathcal{Q} := \{(q_0, q_r) \in \mathbb{R} \times \mathbb{R}^m \mid \|q_r\| - q_0 \leq 0\}$$

is known as the *second-order/Lorentz/ice-cream cone*. Problems of this type have been well recognized in conic programming and numerous applications; see, e.g., [1, 5] for more discussions and references. Observe that, despite the imposed smoothness of f and g , the defined SOCP model is a problem of nonsmooth optimization due to the nondifferentiability at the origin of the norm function presented in (1.2).

The notion of tilt stability of local minimizers was introduced by Poliquin and Rockafellar [29] in the extended-real-valued framework of unconstrained optimization as follows. Given a function $\varphi: \mathbb{R}^n \rightarrow \overline{\mathbb{R}} := (-\infty, \infty]$ with the domain $\text{dom } \varphi := \{x \in \mathbb{R}^n \mid \varphi(x) < \infty\}$, we say that $\bar{x} \in \text{dom } \varphi$ is a *tilt-stable minimizer* of φ if there is a number $\gamma > 0$ such that the mapping

$$(1.3) \quad M_\gamma(v^*) := \arg \min \{\varphi(x) - \langle v^*, x \rangle \mid \|x - \bar{x}\| \leq \gamma\}, \quad v^* \in \mathbb{R}^n,$$

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is single-valued and Lipschitz continuous in some neighborhood of $\bar{v}^* := 0 \in \mathbb{R}^n$ with $M_\gamma(0) = \{\bar{x}\}$. The main result of [29] provides a characterization of tilt-stable local minimizers for φ via the positive-definiteness of the second-order subdifferential/generalized Hessian of φ at the reference point as introduced by the third author in [18]. The results of [29] were extended by Mordukhovich and Nghia [20, 21], who developed a new approach to tilt stability and derived *quantitative* characterization of tilt-stable local minimizers with a prescribed Lipschitz modulus of the mapping M_γ in (1.3) and then computed the exact lower bound of these moduli in terms of the second-order subdifferential and its “combined” modification in finite and infinite dimensions. The characterizations of [20, 21] were established in both *neighborhood* form involving points nearby the reference minimizer as well as via *point-based* criteria expressed entirely at the point in question.

Quite recently, Chieu, Hien, and Nghia [6] derived quantitative neighborhood characterizations of tilt stability in the abstract finite-dimensional framework of (1.3) that are expressed in the form of [21] but with replacing the dual-space combined second-order subdifferential therein by the primal-dual “subgradient graphical derivative” discussed in section 2. We refer the reader to [4, 8, 9, 19, 20, 21, 34] and the bibliographies therein for other results on tilt stability as well as closely related notions in the abstract framework of (1.3). Applications of tilt stability to multiplier criticality and convergence rates of some primal-dual algorithms in numerical optimization have been recently provided in [27].

Efficient implementations and applications of the second-order characterizations of tilt stability obtained in the abstract scheme (1.3) to structural classes of optimization problems with explicit constraints require developing adequate machinery of second-order variational analysis and subdifferential calculus under appropriate qualification conditions. First results in this direction were obtained by Mordukhovich and Rockafellar [26] for problems of nonlinear programming (NLPs) with C^2 -smooth data under the classical *linear independence constraint qualification* (LICQ) fulfilled at the given local minimizer. In this setting, as shown in [26], tilt stability is characterized by Robinson’s *strong second-order sufficient condition* (SSOSC) introduced in [31].

Since LICQ is not necessary for tilt stability in nonlinear programming (in contrast to the case of strong regularity of the associated KKT systems), several attempts have been made to relax LICQ and study tilt-stable local minimizers for NLPs with nonunique Lagrange multipliers; see [6, 11, 21, 23]. The most advanced results in this direction were obtained under the weakest *metric subregularity constraint qualification* (MSCQ). It was first used by Gfrerer and Mordukhovich [11], combined with some other weak constraint qualifications, to derive point-based sufficient conditions, necessary conditions, as well as complete characterizations of tilt-stable minimizers for NLPs. More recently, Chieu, Hien, and Nghia [6] employed MSCQ to establish a new neighborhood characterization and point-based sufficient conditions for tilt stability in NLPs with C^2 -smooth inequality constraints.

All the methods and results for tilt stability in NLPs discussed above are strongly based on the polyhedral structure of such problems. Not much is known for tilt stability of local minimizers in *nonpolyhedral* problems of conic programming. The main result of [25] yields a characterization of tilt-stable minimizers for SOCPs in the extended form of SSOSC under a *nondegeneracy* condition corresponding to LICQ in the SOCP setting. It is based on the second-order subdifferential calculations taken from Outrata and Sun [28]. Similar characterizations of tilt stability involving nondegeneracy and uniqueness of Lagrange multipliers are established in [22, 24] for C^2 -reducible problems of conic programming, where the second-order subdifferential

term is calculated in [22], based on the results of [7], for semidefinite programs entirely in terms of the program data.

The main goal of this paper is to derive verifiable sufficient conditions, necessary conditions, and complete characterizations of tilt-stable minimizers for SOCPs under an appropriate version of *MSCQ* in second-order cone programming that is introduced here. Our results include neighborhood characterizations and much more difficult *point-based* conditions for tilt stability. Furthermore, we establish point-based *quantitative* evaluations and precise formulas for computing the *exact bound* of tilt stability expressed entirely in terms of the given SOCP data. All the results obtained are the first in the literature for tilt stability in second-order cone programming without imposing nondegeneracy. In particular, they are new under the Robinson constraint qualification, which corresponds to the replacement of metric subregularity by the stronger metric regularity assumption. Our results provide new information (e.g., the calculation of the exact bound of tilt stability) even under nondegeneracy.

The rest of the paper is organized as follows. Section 2 presents some preliminaries from variational analysis and generalized differentiation systematically used in the subsequent text. In section 3 we obtain neighborhood characterizations of tilt-stable minimizers for SOCPs with neighborhood calculating the exact bound of tilt stability. In section 4 we begin developing significantly more challenging and more important point-based results on the tilt stability of local minimizers in SOCPs starting with second-order sufficient conditions for such minimizers under MSCQ. The major example presented here illustrates essential features of the obtained conditions in comparison with those known under nondegeneracy and polyhedrality. Section 5 is devoted to deriving “no-gap” point-based necessary conditions for tilt-stable minimizers in SOCPs and their complete characterizations for which an additional “2-regularity” assumption is required in some cases. In the concluding section 6 we discuss the main thrust of the paper and some open problems in this area.

Our notation is conventional in variational analysis, conic programming, and generalized differentiation; see, e.g., [4, 19, 32]. Recall that \mathbb{B} and \mathcal{S} stand, respectively, for the closed unit ball and sphere of the space in question, that $\mathbb{B}_r(x) := x + r\mathbb{B}$, and that $\mathbb{N} := \{1, 2, \dots\}$. Taking into account the structure of \mathcal{Q} in (1.2), we represent an element $q \in \mathbb{R}^{1+m}$ as $q = (q_0, q_r)$ with $q_0 \in \mathbb{R}$ and $q_r \in \mathbb{R}^m$, and also denote $\hat{q} := (-q_0, q_r)$. This notation is used, in particular, for the mapping $g(x) = (g_0(x), g_r(x))$ in the constraint system Γ from (1.1). Furthermore (cf. [5]), we have the following representations for the dual cone, the tangent cone, and the normal cone to \mathcal{Q} , respectively: $\mathcal{Q}^* = \{\hat{q} \mid q \in \mathcal{Q}\}$,

$$(1.4a) \quad T_{\mathcal{Q}}(q) = \begin{cases} \mathcal{Q} & \text{if } q = 0, \\ \mathbb{R}^{1+m} & \text{if } q \in \text{int } \mathcal{Q}, \\ \left\{ u \in \mathbb{R}^{1+m} \mid \frac{q_r}{\|q_r\|} u_r - u_0 \leq 0 \right\} & \text{if } q \in \text{bd } \mathcal{Q} \setminus \{0\}, \end{cases}$$

$$(1.4b) \quad N_{\mathcal{Q}}(q) = \begin{cases} \mathcal{Q}^* & \text{if } q = 0, \\ \{0\} & \text{if } q \in \text{int } \mathcal{Q}, \\ \left\{ \alpha \left(-1, \frac{q_r}{\|q_r\|} \right) \mid \alpha \geq 0 \right\} = \{\alpha \hat{q} \mid \alpha \geq 0\} & \text{if } q \in \text{bd } \mathcal{Q} \setminus \{0\}. \end{cases}$$

Given a (sufficiently) smooth real-valued/scalar function $\varphi : \mathbb{R}^n \rightarrow \mathbb{R}$, denote its gradient and Hessian at x by $\nabla \varphi(x)$ and $\nabla^2 \varphi(x)$, respectively. Considering further

a vector function $h: \mathbb{R}^n \rightarrow \mathbb{R}^s$ with $s > 1$, denote by $\nabla h(x)$ the Jacobian of h at x , while by $\nabla^2 h(x)$ we mean a linear mapping from \mathbb{R}^n into the space of $s \times n$ matrices defined by

$$\nabla^2 h(x)u := \lim_{t \rightarrow 0} \frac{\nabla h(x+tu) - \nabla h(x)}{t}.$$

We also employ the second-order construction

$$\nabla^2 h(x)(u, v) := (\nabla^2 h(x)u)v = \lim_{t \rightarrow 0} \frac{\nabla h(x+tu) - \nabla h(x)}{t}v.$$

2. Preliminaries from variational analysis. We first recall generalized differential constructions of variational analysis utilized in what follows. The reader can find more details in the books [19, 32] and the references therein. Starting with sets, consider a nonempty one $\Omega \subset \mathbb{R}^n$ and define the (Bouligand–Severi) *tangent cone* to Ω at $\bar{x} \in \Omega$ by

$$(2.1) \quad T_\Omega(\bar{x}) := \{v \in \mathbb{R}^n \mid \exists t_k \downarrow 0, \exists v_k \rightarrow v \text{ such that } \bar{x} + t_k v_k \in \Omega, k \in \mathbb{N}\}.$$

The (limiting, Mordukhovich) *normal cone* to Ω at $\bar{x} \in \Omega$ is given by

$$(2.2) \quad N_\Omega(\bar{x}) := \left\{ x^* \in \mathbb{R}^n \mid \exists x_k \xrightarrow{\Omega} \bar{x}, \exists x_k^* \rightarrow x^* \text{ such that } \limsup_{x \rightarrow \bar{x}} \frac{\langle x^*, x - x_k \rangle}{\|x - x_k\|} \leq 0 \right\},$$

where $x \xrightarrow{\Omega} \bar{x}$ indicates that $x \rightarrow \bar{x}$ with $x \in \Omega$. Note that both cones (2.1) and (2.2) may be nonconvex.

Considering next an extended-real-valued function $\varphi: \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$ finite at \bar{x} , the only *subdifferential* of φ at \bar{x} used in this paper is defined by

$$(2.3) \quad \partial\varphi(\bar{x}) := \left\{ x^* \in \mathbb{R}^n \mid \exists x_k \xrightarrow{\varphi} \bar{x}, \exists x_k^* \rightarrow x^* \text{ such that } \liminf_{x \rightarrow x_k} \frac{\varphi(x) - \varphi(x_k) - \langle x_k^*, x - x_k \rangle}{\|x - x_k\|} \geq 0 \right\},$$

where $x \xrightarrow{\varphi} \bar{x}$ indicates that $x \rightarrow \bar{x}$ with $\varphi(x) \rightarrow \varphi(\bar{x})$. Note that despite the nonconvexity of $\partial\varphi(\bar{x})$, the subdifferential (2.3) enjoys full calculus based on variational/extremal principles of variational analysis [19, 32]. There are close relationships between (2.3) and (2.2), namely,

$$(2.4) \quad \partial\varphi(\bar{x}) = \{x^* \in \mathbb{R}^n \mid (x^*, -1) \in N_{\text{epi } \varphi}(\bar{x}, \varphi(\bar{x}))\} \text{ and } N_\Omega(\bar{x}) = \partial\delta_\Omega(\bar{x}),$$

where $\text{epi } \varphi := \{(x, \alpha) \in \mathbb{R}^{n+1} \mid \alpha \geq \varphi(x)\}$ is the epigraph of φ , and where $\delta_\Omega(x)$ is the indicator function of Ω equal to 0 if $x \in \Omega$ and to ∞ otherwise.

Recall that a lower semicontinuous (l.s.c.) function $\varphi: \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$ is *continuously prox-regular* at \bar{x} for $\bar{x}^* \in \partial\varphi(\bar{x})$ if there exists $r > 0$ such that

$$\begin{aligned} \varphi(x) &\geq \varphi(u) + \langle x^*, x - u \rangle - (r/2)\|x - u\|^2 \\ \text{whenever } x, u &\text{ near } \bar{x} \text{ and } x^* \in \partial\varphi(u) \text{ near } \bar{x}^* \end{aligned}$$

and if the mapping $(x, x^*) \mapsto \varphi(x)$ is continuous relative to the subdifferential graph $\text{gph } \partial\varphi$ at (\bar{x}, \bar{x}^*) . This class is rather large including [32] all l.s.c. convex functions, strongly amenable functions, etc.

Considering further a set-valued mapping $F: \mathbb{R}^n \rightrightarrows \mathbb{R}^m$, recall that the *graphical derivative* of F at $(\bar{x}, \bar{v}) \in \text{gph } F$ is defined via the tangent cone (2.1) by

$$(2.5) \quad DF(\bar{x}, \bar{v})(u) := \{w \in \mathbb{R}^m \mid (u, w) \in T_{\text{gph } F}(\bar{x}, \bar{v})\}, \quad u \in \mathbb{R}^n.$$

The main *second-order* construction for extended-real-valued functions used in this paper is the following *primal-dual* one [24] known as the *subgradient graphical derivative* of $\varphi: \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$ at $\bar{x} \in \text{dom } \varphi$ for $\bar{x}^* \in \partial\varphi(\bar{x})$, which is defined via (2.3) and (2.5) by

$$(2.6) \quad D\partial\varphi(\bar{x}, \bar{x}^*)(u) := \{w \in \mathbb{R}^n \mid (u, w) \in T_{\text{gph } \partial\varphi}(\bar{x}, \bar{x}^*)\}, \quad u \in \mathbb{R}^n.$$

The generalized second-order derivative (2.6) is exploited in [6, Theorem 3.3] to derive the following neighborhood characterization of tilt-stable minimizers in the abstract framework of extended-real-valued functions. As in [20], the *exact bound of tilt stability* for φ at \bar{x} is defined by

$$\text{tilt}(\varphi, \bar{x}) = \inf_{\gamma > 0} \text{lip } M_\gamma(0),$$

where $\text{lip } M_\gamma(0)$ stands for the infimum of Lipschitz moduli of M_γ from (1.3) around the origin.

THEOREM 2.1 (abstract neighborhood characterization of tilt stability). *Let $\varphi: \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$ be an l.s.c. function that is continuously prox-regular at $\bar{x} \in \text{dom } \varphi$ for $\bar{x}^* = 0$. Assume that $0 \in \partial\varphi(\bar{x})$. Then \bar{x} is a tilt-stable local minimizer of φ with modulus $\kappa > 0$ if and only if there is a constant $\eta > 0$ such that*

$$(2.7) \quad \langle u^*, u \rangle \geq \frac{1}{\kappa} \text{ whenever } u \in \mathcal{S}, u^* \in D\partial\varphi(x, x^*)(u), (x, x^*) \in \text{gph } \partial\varphi \cap \mathbb{B}_\eta(\bar{x}, 0).$$

Furthermore, the exact bound for tilt stability of φ at \bar{x} is calculated by the formula

$$(2.8) \quad \text{tilt}(\varphi, \bar{x}) = \inf_{\eta > 0} \left\{ \frac{1}{\langle u^*, u \rangle} \mid u \in \mathcal{S}, u^* \in D\partial\varphi(x, x^*)(u), (x, x^*) \in \text{gph } \partial\varphi \cap \mathbb{B}_\eta(\bar{x}, 0) \right\}.$$

In what follows we aim at employing this result to establish verifiable conditions for and characterizations of tilt stability in SOCPs by expressed entirely via the *program data*. The main attention is paid to deriving *point-based conditions* formulated exactly at the reference point. Our major assumption is the novel MSCQ that is far removed from constraint nondegeneracy being essentially weaker than the conventional Robinson constraint qualification under which the obtained results are also new. The developed techniques are rather involved and are based on second-order calculus and calculations in nonpolyhedral settings.

3. Neighborhood characterizations of tilt-stable minimizers for SOCPs. We start with neighborhood characterizations of tilt stability for SOCPs and first introduce the underlying MSCQ for the constraint system in (1.1). Recall that a mapping $Q: \mathbb{R}^n \rightrightarrows \mathbb{R}^m$ is *metrically regular* around $(\bar{x}, \bar{y}) \in \text{gph } Q$ with modulus $\sigma > 0$ if there exist neighborhoods U of \bar{x} and V of \bar{y} such that we have the estimate

$$(3.1) \quad \text{dist}(x; Q^{-1}(y)) \leq \sigma \text{dist}(y; Q(x)) \text{ for all } x \in U \text{ and } y \in V,$$

where $\text{dist}(w; \Omega)$ signifies the distance between a point and a set. The mapping Q

is *metrically subregular* at (\bar{x}, \bar{y}) with modulus σ if the distance estimate (3.1) holds with $y = \bar{y}$ therein.

DEFINITION 3.1 (MSCQ for SOCPs). *We say that the MSCQ is fulfilled for the SOCP constraint system Γ from (1.1) at a point $\bar{x} \in \Gamma$ with modulus $\sigma > 0$ if the mapping $Q(x) := g(x) - \mathcal{Q}$ is metrically subregular at $(\bar{x}, 0)$ with this modulus, i.e.,*

$$(3.2) \quad \text{dist}(x; g^{-1}(\mathcal{Q})) \leq \sigma \text{dist}(g(x); \mathcal{Q}) \quad \text{for all } x \in U.$$

To this end, it is worth emphasizing that the more restrictive metric *regularity* of the mapping $Q(\cdot)$ from Definition 3.1 around $(\bar{x}, 0)$ is equivalent to the *Robinson constraint qualification*

$$0 \in \text{int}\{g(\bar{x}) + \nabla g(\bar{x})\mathbb{R}^n - \mathcal{Q}\}.$$

Taking into account that the validity of MSCQ at $\bar{x} \in \Gamma$ yields its fulfillment at any $x \in \Gamma$ sufficiently close to \bar{x} (with the same modulus σ as can be supposed without loss of generality), we deduce from [16, Theorem 4.1], [17, Proposition 1], and the formulas in (1.4) that the tangent cone and the normal cone to Γ at x are computed by

$$(3.3) \quad \begin{aligned} T_\Gamma(x) &= \{u \in \mathbb{R}^n \mid \nabla g(x)u \in T_{\mathcal{Q}}(g(x))\} \\ &= \begin{cases} \{u \in \mathbb{R}^n \mid \nabla g(x)u \in \mathcal{Q}\} & \text{if } g(x) = 0, \\ \mathbb{R}^n & \text{if } g(x) \in \text{int } \mathcal{Q}, \\ \left\{u \in \mathbb{R}^n \mid \left\langle \frac{g_r(x)}{\|g_r(x)\|}, \nabla g_r(x)u \right\rangle - \nabla g_0(x)u \leq 0\right\} & \text{if } g(x) \in \text{bd } \mathcal{Q} \setminus \{0\}; \end{cases} \\ N_\Gamma(x) &= \begin{cases} \{\nabla g(x)^*\lambda \mid \lambda \in \mathcal{Q}^*\} & \text{if } g(x) = 0, \\ \{0\} & \text{if } g(x) \in \text{int } \mathcal{Q}, \\ \left\{\alpha \left(\nabla g_r(x)^* \frac{g_r(x)}{\|g_r(x)\|} - \nabla g_0(x)^*\right) \mid \alpha \geq 0\right\} & \text{if } g(x) \in \text{bd } \mathcal{Q} \setminus \{0\}. \end{cases} \end{aligned}$$

Given a normal vector $x^* \in N_\Gamma(x)$, consider the corresponding *multiplier set*

$$\Lambda(x, x^*) := \{\lambda \in N_{\mathcal{Q}}(g(x)) \mid \nabla g(x)^*\lambda = x^*\}$$

and define the *critical cone* to Γ at (x, x^*) by

$$\mathcal{K}_\Gamma(x, x^*) = T_\Gamma(x) \cap [x^*]^\perp.$$

Note that for every $\lambda \in \Lambda(x, x^*)$ we have $u \in \mathcal{K}_\Gamma(x, x^*)$ if and only if $u \in T_\Gamma(x)$ and $\langle \lambda, \nabla g(x)u \rangle = 0$. If $x^* = 0$, the only multiplier important for our analysis is $\lambda = 0$. Thus we introduce the set

$$\Lambda^0(x, x^*) := \begin{cases} \{0\} & \text{when } x^* = 0, \\ \Lambda(x, x^*) & \text{otherwise.} \end{cases}$$

Further, for every direction $u \in \mathbb{R}^n$ define the *directional multiplier set* by

$$\Lambda(x, x^*; u) := \arg \max_{\lambda \in \Lambda(x, x^*)} \langle (\nabla^2 \langle \lambda, g \rangle)(x) + \mathcal{H}(x, \lambda)u, u \rangle,$$

where the *curvature mapping* $\mathcal{H}: \mathbb{R}^n \times \mathbb{R}^{1+m} \rightarrow \mathbb{R}^{n \times n}$ is given by

$$(3.4) \quad \mathcal{H}(x, \lambda) := \begin{cases} \frac{-\lambda_0}{g_0(x)} (\nabla g_r(x)^* \nabla g_r(x) - \nabla g_0(x)^* \nabla g_0(x)) & \text{if } g(x) \in \text{bd } \mathcal{Q} \setminus \{0\}, \\ 0 & \text{otherwise.} \end{cases}$$

For more details regarding the above constructions, we refer to the pivotal paper [5]. In particular, it is well-known that

$$(3.5) \quad \langle \mathcal{H}(x, \lambda)u, u \rangle \geq 0$$

holds for every pair $(x, x^*) \in \text{gph } N_\Gamma$, every multiplier $\lambda \in \Lambda(x, x^*)$, and every critical direction $u \in \mathcal{K}_\Gamma(x, x^*)$ (see, e.g., [5, Proof of Theorem 29]).

Now we are ready to establish neighborhood characterizations of tilt stability for local minimizers of SOCPs and compute the exact bound of tilt stability via the given data of (1.1) and (1.2). Tilt stability of \bar{x} in (1.1) is naturally understood as the one for $\varphi(x) := f(x) + \delta_\Gamma(x)$ in the sense defined above. Then $\bar{x} \in \Gamma$ is a *stationary point* of (1.1) if $0 \in \partial\varphi(\bar{x})$, i.e.,

$$0 \in \partial f(\bar{x}) + N_\Gamma(\bar{x})$$

due to the standard first-order subdifferential sum rule [32, Exercise 10.10] valid in our setting.

THEOREM 3.2 (neighborhood characterizations of tilt stability for SOCPs). *Let $\bar{x} \in \Gamma$ be a stationary point of (1.1), and let MSCQ hold at \bar{x} with modulus σ from (3.2). Then the following three assertions are equivalent:*

- (i) *The point \bar{x} is a tilt-stable minimizer for (1.1) with modulus $\kappa > 0$.*
- (ii) *There exists a constant $\eta > 0$ such that*

$$(3.6) \quad \begin{cases} \langle (\nabla^2 f(x) + \nabla^2 \langle \lambda, g \rangle(x) + \mathcal{H}(x, \lambda))u, u \rangle \geq \frac{1}{\kappa} \\ \text{whenever } x \in \Gamma \cap \mathbb{B}_\eta(\bar{x}), \|x^*\| \leq \eta, u \in \mathcal{K}_\Gamma(x, x^* - \nabla f(x)) \cap \mathcal{S}, \\ \lambda \in \Lambda(x, x^* - \nabla f(x); u) \cap \sigma \|x^* - \nabla f(x)\| \mathbb{B}. \end{cases}$$

- (iii) *There exists a constant $\eta > 0$ such that*

$$(3.7) \quad \begin{cases} \langle (\nabla^2 f(x) + \nabla^2 \langle \lambda, g \rangle(x) + \mathcal{H}(x, \lambda))u, u \rangle \geq \frac{1}{\kappa} \\ \text{whenever } x \in \Gamma \cap \mathbb{B}_\eta(\bar{x}), \|x^*\| \leq \eta, u \in \mathcal{K}_\Gamma(x, x^* - \nabla f(x)) \cap \mathcal{S}, \\ \lambda \in \Lambda(x, x^* - \nabla f(x); u) \cap \Lambda^0(x, x^* - \nabla f(x)). \end{cases}$$

Furthermore, the exact bound for tilt stability of \bar{x} in (1.1) is calculated by

$$(3.8) \quad \text{tilt}(f + \delta_\Gamma, \bar{x}) = \inf_{\eta > 0} \sup \left\{ \frac{1}{\langle (\nabla^2 f(x) + \nabla^2 \langle \lambda, g \rangle(x) + \mathcal{H}(x, \lambda))u, u \rangle} \mid x \in \Gamma \cap \mathbb{B}_\eta(\bar{x}), \|x^*\| \leq \eta, u \in \mathcal{K}_\Gamma(x, x^* - \nabla f(x)) \cap \mathcal{S}, \lambda \in \Lambda(x, x^* - \nabla f(x); u) \cap \sigma \|x^* - \nabla f(x)\| \mathbb{B} \right\}.$$

Proof. We derive the claimed results from those in Theorem 2.1 with $\varphi(x) = f(x) + \delta_\Gamma(x)$ by using the appropriate rules of first-order and second-order generalized differential calculus. Since $f \in C^2$, it follows that $\partial(f + \delta_\Gamma)(x) = \nabla f(x) + N_\Gamma(x)$ and

$$D\partial(f + \delta_\Gamma)(x, x^*) = \nabla^2 f(x) + DN_\Gamma(x, x^* - \nabla f(x))$$

for all x sufficiently close to \bar{x} . Due to the validity of MSCQ at such x with modulus σ , we get from the second-order calculation in [15, Corollary 5.2] that

$$\begin{aligned} DN_{\Gamma}(x, x^* - \nabla f(x))(u) \\ = \left\{ \nabla^2 \langle \lambda, g \rangle(x)u + \mathcal{H}(x, \lambda)u \mid \lambda \in \Lambda(x, x^* - \nabla f(x); u) \cap \sigma \|x^* - \nabla f(x)\| \mathbb{B} \right\} \\ + N_{\mathcal{K}_{\Gamma}(x, x^* - \nabla f(x))}(u) \text{ for any } u \in \mathbb{R}^n. \end{aligned}$$

Noting that the validity of MSCQ at $x \in \Gamma$ with modulus σ ensures that this property holds at x with any modulus $\sigma' \geq \sigma$ gives us the formula

$$\begin{aligned} DN_{\Gamma}(x, x^* - \nabla f(x))(u) \\ = \left\{ \nabla^2 \langle \lambda, g \rangle(x)u + \mathcal{H}(x, \lambda)u \mid \lambda \in \Lambda(x, x^* - \nabla f(x); u) \cap \Lambda^0(x, x^* - \nabla f(x)) \right\} \\ + N_{\mathcal{K}_{\Gamma}(x, x^* - \nabla f(x))}(u) \text{ for any } u \in \mathbb{R}^n. \end{aligned}$$

Substituting the above calculations into (2.7) and (2.8) while taking into account that for every vector $\zeta \in N_{\mathcal{K}_{\Gamma}(x, x^* - \nabla f(x))}(u)$ we have $\langle \zeta, u \rangle = 0$ allows us to complete the proof of the theorem. \square

4. Point-based sufficient conditions for tilt stability in SOCP. The results of Theorem 3.2 involve *all* the points from a neighborhood of the reference stationary point \bar{x} , which makes the verification of the obtained formula (3.6) difficult. The goal of this section is to derive verifiable sufficient conditions for tilt stability in SOCPs expressed only at the point \bar{x} in question.

The cases when $g(\bar{x}) \in \text{int } \mathcal{Q}$ or $g(\bar{x}) \in \text{bd } \mathcal{Q} \setminus \{0\}$ are easy to handle, because we can equivalently rewrite the problem (1.1) as

$$(4.1) \quad \min f(x) \quad \text{subject to} \quad \phi(x) := \|g_r(x)\| - g_0(x) \leq 0.$$

In the first case this problem behaves locally like an unconstrained problem whereas in the second case this problem behaves like an C^2 -smooth nonlinear programming problem, for which point-based characterizations of tilt stability are well-known (cf. [26, 11]). For the sake of completeness we state here the corresponding results.

THEOREM 4.1. *Let $\bar{x} \in \Gamma$ be a stationary point of (1.1) with either $g(\bar{x}) \in \text{int } \mathcal{Q}$ or $g(\bar{x}) \in \text{bd } \mathcal{Q} \setminus \{0\}$. Denote*

$$\bar{\lambda} = \begin{cases} \frac{\|\nabla f(\bar{x})\|}{\|\nabla g(\bar{x})^* \hat{g}(\bar{x})\|} \hat{g}(\bar{x}) & \text{if } \nabla g(\bar{x})^* \hat{g}(\bar{x}) \neq 0 \text{ with } \hat{g}(\bar{x}) := (-g_0(\bar{x}), g_r(\bar{x})), \\ 0 & \text{else.} \end{cases}$$

If either $g(\bar{x}) \in \text{int } \mathcal{Q}$ or $\nabla g(\bar{x})^ \hat{g}(\bar{x}) \neq 0$, then \bar{x} is a tilt-stable local minimizer of (1.1) if and only if*

$$\langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \bar{\lambda}, g \rangle(\bar{x}) + \mathcal{H}(\bar{x}, \bar{\lambda}))u, u \rangle > 0 \text{ whenever } u \in \mathcal{S}, \langle \bar{\lambda}, \nabla g(\bar{x})u \rangle = 0.$$

Furthermore, the exact bound of tilt-stability is

$$\begin{aligned} \text{tilt}(f + \delta_{\Gamma}, \bar{x}) \\ = \sup \left\{ 1 / \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \bar{\lambda}, g \rangle(\bar{x}) + \mathcal{H}(\bar{x}, \bar{\lambda}))u, u \rangle \mid u \in \mathcal{S}, \langle \bar{\lambda}, \nabla g(\bar{x})u \rangle = 0 \right\}. \end{aligned}$$

If $g(\bar{x}) \in \text{bd } \mathcal{Q} \setminus \{0\}$ and $\nabla g(\bar{x})^* \hat{g}(\bar{x}) = 0$, assume that MSCQ holds at \bar{x} with modulus σ from (3.2). Then $\lambda = 0$ and \bar{x} is a tilt-stable local minimizer if the Hessian $\nabla^2 f(\bar{x})$ is positive definite. In this case we have the upper estimate

$$\text{tilt}(f + \delta_\Gamma, \bar{x}) \leq \sup \left\{ 1 / \langle \nabla^2 f(\bar{x}) u, u \rangle \mid u \in \mathcal{S} \right\}.$$

Proof. The first assertion follows from [11, Theorem 7.6] applied to (4.1) by taking into account that in this case even LICQ is fulfilled at \bar{x} . The second assertion follows from [11, Theorem 6.1] applied to (4.1). In fact, one can show $\text{dist}(g(x), \mathcal{Q}) \leq \max\{\phi(x), 0\}$ implying that MSCQ holds for the constraint $\phi(x) \leq 0$. In order to apply [11, Theorem 6.1] we must also show the so-called *bounded extreme point property* (BEPP). Given $(x, x^*) \in \text{gph } N_\Gamma$ with x close to x^* , we have to consider the extreme points of the multiplier set $\{\mu \geq 0 \mid x^* = \nabla \phi(x)\mu\}$, which is either \mathbb{R}^+ or a singleton. From this observation, BEPP follows from [14, Proposition 4.1] with $u = 0$. Thus all assumptions of [11, Theorem 6.1] are fulfilled and the statement follows. \square

In what follows we consider the remaining case $g(\bar{x}) = 0$ and we make use of the following standing assumption.

ASSUMPTION 4.2. Let $\bar{x} \in \Gamma$ be a stationary point of (1.1) with $g(\bar{x}) = 0$, and let MSCQ hold at \bar{x} with modulus σ from (3.2).

In order to obtain the point-based conditions for tilt stability, we start with the second-order neighborhood characterizations from the previous section and consider a limiting procedure. This is the most crucial as well as demanding task. In particular, the curvature mapping \mathcal{H} has zero value at \bar{x} with $g(\bar{x}) = 0$, but we have to deal with the *discontinuous* behavior of \mathcal{H} around \bar{x} . This is yet another reason to focus on the case where $g(\bar{x}) = 0$, since \mathcal{H} is clearly continuous at the points with $g(\bar{x}) \neq 0$. The first step in computing the limits is provided by the following technical lemma.

LEMMA 4.3 (limiting procedure). Consider convergent sequences $x_k \xrightarrow{\Gamma} \bar{x}$ with $g(\bar{x}) = 0$, $x_k^* \rightarrow 0$, $\lambda^k \rightarrow \tilde{\lambda}$, and $u_k \rightarrow \tilde{u}$ as $k \rightarrow \infty$ satisfying the inclusions

$$\lambda^k \in \Lambda(x_k, x_k^* - \nabla f(x_k); u_k) \cap \sigma \|x_k^* - \nabla f(x_k)\| \mathbb{B} \quad \text{and} \quad u_k \in \mathcal{K}_\Gamma(x_k, x_k^* - \nabla f(x_k)) \cap \mathcal{S}$$

for all $k \in \mathbb{N}$ together with the condition

$$(4.2) \quad \limsup_{k \rightarrow \infty} \langle \mathcal{H}(x_k, \lambda^k) u_k, u_k \rangle < \infty.$$

Then we have the limiting relationships

$$(4.3a) \quad \tilde{\lambda} \in \Lambda(\bar{x}, -\nabla f(\bar{x})) \cap \sigma \|\nabla f(\bar{x})\| \mathbb{B} \subset \Lambda^0(\bar{x}, -\nabla f(\bar{x})),$$

$$(4.3b) \quad \tilde{u} \in \mathcal{S}, \quad \langle \tilde{\lambda}, \nabla g(\bar{x}) \tilde{u} \rangle = 0, \quad \tilde{\lambda}_0 (\|\nabla g_r(\bar{x}) \tilde{u}\|^2 - (\nabla g_0(\bar{x}) \tilde{u})^2) = 0.$$

Proof. Passing to the limit as $k \rightarrow \infty$ in the conditions

$$\nabla g(x_k)^* \lambda^k = x_k^* - \nabla f(x_k), \quad \lambda^k \in N_{\mathcal{Q}}(g(x_k)) \subset \mathcal{Q}^*, \quad \text{and} \quad \|\lambda^k\| \leq \sigma \|x_k^* - \nabla f(x_k)\|$$

verifies that $\nabla g(\bar{x})^* \tilde{\lambda} = -\nabla f(\bar{x})$, $\tilde{\lambda} \in \mathcal{Q}^*$, and $\|\tilde{\lambda}\| \leq \sigma \|\nabla f(\bar{x})\|$. This yields

$$\tilde{\lambda} \in \Lambda(\bar{x}, -\nabla f(\bar{x})) \cap \sigma \|\nabla f(\bar{x})\| \mathbb{B} \subset \Lambda^0(\bar{x}, -\nabla f(\bar{x})).$$

It follows from $\langle \lambda^k, \nabla g(x_k) u_k \rangle = 0$ as $k \in \mathbb{N}$ that $\langle \tilde{\lambda}, \nabla g(\bar{x}) \tilde{u} \rangle = 0$. Thus it remains to show that

$$(4.4) \quad \tilde{\lambda}_0 (\|\nabla g_r(\bar{x}) \tilde{u}\|^2 - (\nabla g_0(\bar{x}) \tilde{u})^2) = 0.$$

Since (4.4) is certainly true if $\tilde{\lambda}_0 = 0$, we proceed with the case where $\tilde{\lambda}_0 \neq 0$ and hence $\lambda_0^k \neq 0$ for all large $k \in \mathbb{N}$. It follows from $\tilde{\lambda} \in \mathcal{Q}^*$ that $-\tilde{\lambda}_0 \geq \|\tilde{\lambda}_r\|$. If $g(x_k) = 0$ for infinitely many k , then $\nabla g(x_k)u_k \in \mathcal{Q}$ and therefore $\nabla g(\bar{x})\tilde{u} \in \mathcal{Q}$, which in turn is equivalent to $\nabla g_0(\bar{x})\tilde{u} \geq \|\nabla g_r(\bar{x})\tilde{u}\|$. From $\langle \tilde{\lambda}, \nabla g(\bar{x})\tilde{u} \rangle = 0$ we deduce the conditions

$$-\tilde{\lambda}_0 \nabla g_0(\bar{x})\tilde{u} = \langle \tilde{\lambda}_r, \nabla g_r(\bar{x})\tilde{u} \rangle \leq \|\tilde{\lambda}_r\| \cdot \|\nabla g_r(\bar{x})\tilde{u}\| \leq -\tilde{\lambda}_0 \nabla g_0(\bar{x})\tilde{u}$$

implying that $\|\tilde{\lambda}_r\| \cdot \|\nabla g_r(\bar{x})\tilde{u}\| = -\tilde{\lambda}_0 \nabla g_0(\bar{x})\tilde{u}$. Thus we get that either $\|\tilde{\lambda}_r\| = -\tilde{\lambda}_0$ and $\|\nabla g_r(\bar{x})\tilde{u}\| = \nabla g_0(\bar{x})\tilde{u}$, or $\|\nabla g_r(\bar{x})\tilde{u}\| = \nabla g_0(\bar{x})\tilde{u} = 0$, which verifies equality (4.4) in this case. In the remaining case where $g(x_k) \neq 0$ for all but finitely many $k \in \mathbb{N}$ it follows that $g(x_k) \in \text{bd } \mathcal{Q} \setminus \{0\}$ for such k . This is due to the fact that the condition $0 \neq \lambda^k \in N_{\mathcal{Q}}(g(x_k))$ yields

$$\langle \mathcal{H}(x_k, \lambda^k)u_k, u_k \rangle = \frac{-\lambda_0^k}{g_0(x_k)} \left(\|\nabla g_r(x_k)u_k\|^2 - (\nabla g_0(x_k)u_k)^2 \right).$$

Using (4.2) together with $g_0(x_k) \rightarrow 0$ as $k \rightarrow \infty$ and (3.5) tells us that

$$\lim_{k \rightarrow \infty} \lambda_0^k (\|\nabla g_r(x_k)u_k\|^2 - (\nabla g_0(x_k)u_k)^2) = \tilde{\lambda}_0 (\|\nabla g_r(\bar{x})\tilde{u}\|^2 - (\nabla g_0(\bar{x})\tilde{u})^2) = 0,$$

which verifies (4.4) in the latter case and thus completes the proof. \square

Now we are ready to proceed with deriving point-based sufficient conditions for tilt stability in SOCP. It is instructive to split our consideration into the two cases: (a) *in-kernel* $\mathcal{K}_{\Gamma}(\bar{x}, -\nabla f(\bar{x})) \subset \ker \nabla g(\bar{x})$ and (b) *out-of-kernel* $\mathcal{K}_{\Gamma}(\bar{x}, -\nabla f(\bar{x})) \not\subset \ker \nabla g(\bar{x})$. The second case can be treated directly by passing to the limit in the conditions of Theorem 3.2 with the usage of the lemma above.

THEOREM 4.4 (sufficient condition for tilt stability in the out-of-kernel case). *In addition to the standing Assumption 4.2 suppose that there is $\bar{u} \in \mathcal{K}_{\Gamma}(\bar{x}, -\nabla f(\bar{x}))$ satisfying $\nabla g(\bar{x})\bar{u} \neq 0$. Denote*

$$(4.5) \quad \bar{\lambda} := \frac{\|\nabla f(\bar{x})\|}{\|\nabla g(\bar{x})^* \nabla \hat{g}(\bar{x})\bar{u}\|} \nabla \hat{g}(\bar{x})\bar{u} \quad \text{with} \quad \hat{g}(x) := (-g_0(x), g_r(x))$$

and assume also that for some $\kappa > 0$ we have

$$(4.6) \quad \begin{aligned} & \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \bar{\lambda}, g \rangle(\bar{x}))u, u \rangle > \frac{1}{\kappa} \quad \text{whenever} \\ & u \in \mathcal{S} \quad \text{with} \quad \langle \bar{\lambda}, \nabla g(\bar{x})u \rangle = 0, \quad \tilde{\lambda}_0 (\|\nabla g_r(\bar{x})u\|^2 - (\nabla g_0(\bar{x})u)^2) = 0. \end{aligned}$$

Then \bar{x} is a tilt-stable local minimizer for (1.1) with modulus κ . Moreover, we have the upper estimate

$$\text{tilt}(f + \delta_{\Gamma}, \bar{x}) \leq \sup \left\{ 1 / \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \bar{\lambda}, g \rangle(\bar{x}))u, u \rangle \mid u \in \mathcal{S}, \langle \bar{\lambda}, \nabla g(\bar{x})u \rangle = 0, \right. \\ \left. \tilde{\lambda}_0 (\|\nabla g_r(\bar{x})u\|^2 - (\nabla g_0(\bar{x})u)^2) = 0 \right\}.$$

Proof. Employing Theorem 3.2, it suffices to show that the second-order condition (4.6) yields (3.6). Then the claimed upper estimate of the exact bound of tilt stability follows easily. Assuming on the contrary that (3.6) fails while (4.6) holds, we find sequences $x_k \xrightarrow{\Gamma} \bar{x}$, $x_k^* \rightarrow 0$, $u_k \in \mathcal{K}_{\Gamma}(x_k, x_k^* - \nabla f(x_k)) \cap \mathcal{S}$, and $\lambda^k \in \Lambda(x_k, x_k^* - \nabla f(x_k); u_k) \cap \sigma \|x_k^* - \nabla f(x_k)\| \mathbb{B}$ satisfying

$$(4.7) \quad \langle (\nabla^2 f(x_k) + \nabla^2 \langle \lambda^k, g \rangle(x_k) + \mathcal{H}(x_k, \lambda^k))u_k, u_k \rangle < \frac{1}{\kappa} \quad \text{for all } k \in \mathbb{N},$$

which ensures, in particular, that the sequence of $\langle \mathcal{H}(x_k, \lambda^k)u_k, u_k \rangle$ is bounded. Thus passing to a subsequence if necessary tells us that u_k converges to some $\bar{u} \in \mathcal{S}$ and that λ_k converges to some $\tilde{\lambda}$ satisfying (4.3) by Lemma 4.3. Taking into account that $g(\bar{x}) = 0$ and hence $\nabla g(\bar{x})\bar{u} \in \mathcal{Q}$ and employing further

$$\Lambda(\bar{x}, -\nabla f(\bar{x})) \subset \mathcal{Q}^* \cap (\nabla g(\bar{x})\bar{u})^\perp = N_{\mathcal{Q}}(\nabla g(\bar{x})\bar{u}) \subset \{\alpha \nabla \hat{g}(\bar{x})\bar{u} \mid \alpha \geq 0\}$$

together with $\nabla g(\bar{x})^* \lambda = -\nabla f(\bar{x})$ for all $\lambda \in \Lambda(\bar{x}, -\nabla f(\bar{x}))$ ensures that $\Lambda^0(\bar{x}, -\nabla f(\bar{x})) = \{\tilde{\lambda}\}$ and hence $\tilde{\lambda} = \bar{\lambda}$. It follows from (4.7) and (3.5) that

$$\begin{aligned} \langle (\nabla^2 f(x) + \nabla^2 \langle \bar{\lambda}, g \rangle(x))\bar{u}, \bar{u} \rangle &\leq \limsup_{k \rightarrow \infty} \langle (\nabla^2 f(x_k) + \nabla^2 \langle \lambda^k, g \rangle(x_k) + \mathcal{H}(x_k, \lambda^k))u_k, u_k \rangle \\ &\leq \frac{1}{\kappa}, \end{aligned}$$

which contradicts (4.6) and thus completes the proof of the theorem. \square

Remark 4.5.

- (i) The existence of the critical direction \bar{u} with $\nabla g(\bar{x})\bar{u} \neq 0$ in fact yields the uniqueness of the multiplier. Naturally, this makes the out-of-kernel case much simpler and similar to the nondegenerate case.
- (ii) Taking into account the first remark, it is not surprising that our conditions actually coincide with the standard second-order conditions for strongly regular points of SOCP (see [5, Theorem 30]). Indeed, one can show that the conditions

$$\langle \bar{\lambda}, \nabla g(\bar{x})u \rangle = 0, \quad \bar{\lambda}_0(\|\nabla g_r(\bar{x})u\|^2 - (\nabla g_0(\bar{x})u)^2) = 0$$

actually amount to the requirement

$$u \in \text{sp } \mathcal{K}_{\Gamma}(\bar{x}, -\nabla f(\bar{x})) := \mathcal{K}_{\Gamma}(\bar{x}, -\nabla f(\bar{x})) + (-\mathcal{K}_{\Gamma}(\bar{x}, -\nabla f(\bar{x})))$$

used in [5, Theorem 30].

For our subsequent analysis in the in-kernel case we introduce the set

$$(4.8) \quad \begin{aligned} \mathcal{Z} &:= \left\{ (u, \lambda, v, w) \in \mathcal{S} \times \Lambda^0(\bar{x}, -\nabla f(\bar{x})) \times \mathcal{S} \times \mathbb{R}^n \mid \langle \lambda, \nabla g(\bar{x})u \rangle = 0, \right. \\ &\quad \left. \lambda_0(\|\nabla g_r(\bar{x})u\|^2 - (\nabla g_0(\bar{x})u)^2) = 0, \quad \nabla g(\bar{x})v = 0, \quad \lambda \in N_{\mathcal{Q}}(\nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v)) \right\}. \end{aligned}$$

Further, for every triple $(u, \lambda, v) \in \mathbb{R}^n \times \mathcal{Q}^* \times \mathbb{R}^n$ define the *infimum function*

$$(4.9) \quad \rho(u, \lambda, v) := \inf_{z \in \mathbb{R}^n} \left\{ -\lambda_0 \left(\|\nabla g_r(\bar{x})z + \nabla^2 g_r(\bar{x})(v, u)\|^2 - (\nabla g_0(\bar{x})z + \nabla^2 g_0(\bar{x})(v, u))^2 \right) \mid \right. \\ \left. \langle \lambda, \nabla g(\bar{x})z + \nabla^2 g(\bar{x})(v, u) \rangle = 0 \right\}$$

with the convention that $\rho(u, \lambda, v) := \infty$ whenever $\{z \in \mathbb{R}^n \mid \langle \lambda, \nabla g(\bar{x})z + \nabla^2 g(\bar{x})(v, u) \rangle = 0\} = \emptyset$.

Let us present another lemma before deriving the main result of this section in the in-kernel case.

LEMMA 4.6 (properties of the infimum function). *For any triple $(u, \lambda, v) \in \mathbb{R}^n \times \mathcal{Q}^* \times \mathbb{R}^n$ we get that $\rho(u, \lambda, v) \geq 0$. Furthermore, the infimum in (4.9) is attained whenever $\rho(u, \lambda, v)$ is finite.*

Proof. If $\lambda = 0$, we clearly have that $\rho(u, \lambda, v) = 0$ and that the infimum in (4.9) is attained at any $z \in \mathbb{R}^n$. Considering now the case where $\lambda \neq 0$ and $\rho(u, \lambda, v) < \infty$, pick any z satisfying $\langle \lambda, \nabla g(\bar{x})z + \nabla^2 g(\bar{x})(v, u) \rangle = 0$ and denote $\eta := \nabla g(\bar{x})z + \nabla^2 g(\bar{x})(v, u)$. It follows from $\lambda \in \mathcal{Q}^* \setminus \{0\}$ and the structure of \mathcal{Q} in (1.2) that $-\lambda_0 \geq \|\lambda_r\|$ and $-\lambda_0 > 0$. Hence

$$-\lambda_0 |\eta_0| = |\langle \lambda_r, \eta_r \rangle| \leq \|\lambda_r\| \cdot \|\eta_r\| \leq -\lambda_0 \|\eta_r\|,$$

which implies in turn the relationships

$$0 \leq -\lambda_0 (\|\eta_r\|^2 - \eta_0^2) = -\lambda_0 \left(\|\nabla g_r(\bar{x})z + \nabla^2 g_r(\bar{x})(v, u)\|^2 - (\nabla g_0(\bar{x})z + \nabla^2 g_0(\bar{x})(v, u))^2 \right),$$

and therefore $\rho(u, \lambda, v) \geq 0$. To show finally that the infimum is attained in (4.9), observe that the minimization problem therein is with a quadratic cost and one linear equality constraint. It can be equivalently reduced by some linear transformation $z = b + Ay$ to an unconstrained quadratic program of the form $\min \frac{1}{2} \langle y, By \rangle + \langle c, y \rangle$; see, e.g., [10, Chapter 10.1]. The boundedness of the cost function ensures that the matrix B is positive semidefinite. Thus, by [3, Theorem 17.1], the unconstrained optimization problem admits a solution \bar{y} , and so $\bar{z} := b + A\bar{y}$ is an optimal solution to the constrained problem under consideration. \square

The next theorem is a major result of the paper that establishes point-based sufficient conditions for tilt-stable minimizers for SOCPs in the most involved in-kernel case.

THEOREM 4.7 (sufficient conditions for tilt stability in the in-kernel case). *In addition to the standing Assumption 4.2 suppose that*

$$(4.10) \quad \nabla g(\bar{x})u = 0 \text{ for all } u \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x})).$$

Assume also that there is a number $\kappa > 0$ such that the following two conditions are satisfied:

(a) *For every $u \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x})) \cap \mathcal{S}$ and every $\lambda \in \Lambda(\bar{x}, -\nabla f(\bar{x}); u)$ we have*

$$\langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle > \frac{1}{\kappa}.$$

(b) *For every quadruple $(u, \lambda, v, w) \in \mathcal{Z}$ we have*

$$(4.11) \quad \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle + \frac{\rho(u, \lambda, v)}{\nabla g_0(\bar{x})w + \frac{1}{2} \nabla^2 g_0(\bar{x})(v, v)} > \frac{1}{\kappa} \\ \text{whenever } \nabla g(\bar{x})w + \frac{1}{2} \nabla^2 g(\bar{x})(v, v) \neq 0 \quad \text{and}$$

$$(4.12) \quad \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle > \frac{1}{\kappa} \quad \text{whenever } \rho(u, \lambda, v) = 0.$$

Then \bar{x} is a tilt-stable local minimizer for (1.1) with modulus κ . Moreover, we have the upper estimate

$$(4.13) \quad \text{tilt}(f + \delta_\Gamma, \bar{x}) \leq \frac{1}{\min\{\chi_1, \chi_2, \chi_3\}},$$

where the numbers χ_i , $i = 1, 2, 3$, are calculated by

$$\begin{aligned}\chi_1 &:= \inf \left\{ \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle \mid u \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x})) \cap \mathcal{S}, \right. \\ &\quad \left. \lambda \in \Lambda(\bar{x}, -\nabla f(\bar{x}); u) \right\}, \\ \chi_2 &:= \inf \left\{ \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle \mid \exists v, w \text{ with } (u, \lambda, v, w) \in \mathcal{Z}, \rho(u, \lambda, v) = 0 \right\}, \\ \chi_3 &:= \inf \left\{ \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle + \frac{\rho(u, \lambda, v)}{\nabla g_0(\bar{x})w + \frac{1}{2} \nabla^2 g_0(\bar{x})(v, v)} \mid \right. \\ &\quad \left. (u, \lambda, v, w) \in \mathcal{Z}, \nabla g(\bar{x})w + \frac{1}{2} \nabla^2 g(\bar{x})(v, v) \neq 0 \right\}\end{aligned}$$

with the set \mathcal{Z} defined in (4.8).

Proof. Let us show that the fulfillment of both conditions (a) and (b) implies that the tilt stability characterization (3.6) of Theorem 3.2 holds. This verifies all the conclusions of the theorem including the upper bound estimate (4.13), which follows from (3.8) and the proof below.

Arguing by contradiction, suppose that (3.6) fails, i.e., there are sequences $x_k \xrightarrow{\Gamma} \bar{x}$, $x_k^* \rightarrow 0$, $u_k \in \mathcal{K}_\Gamma(x_k, x_k^* - \nabla f(x_k)) \cap \mathcal{S}$, and $\lambda^k \in \Lambda(x_k, x_k^* - \nabla f(x_k); u_k) \cap \sigma \|x_k^* - \nabla f(x_k)\| \mathbb{B}$ satisfying

$$(4.14) \quad \langle (\nabla^2 f(x_k) + \nabla^2 \langle \lambda^k, g \rangle(x_k) + \mathcal{H}(x_k, \lambda^k))u_k, u_k \rangle < \frac{1}{\kappa} \text{ for all } k \in \mathbb{N},$$

and then show that either condition (a) or condition (b) is violated. Passing to a subsequence if necessary, we get that the sequences u_k and λ^k converge to some \tilde{u} and $\tilde{\lambda}$, respectively. Since (4.14) yields the boundedness of $\langle \mathcal{H}(x_k, \lambda^k)u_k, u_k \rangle$, it follows from Lemma 4.3 that

$$\tilde{u} \in \mathcal{S}, \tilde{\lambda} \in \Lambda^0(\bar{x}, -\nabla f(\bar{x})), \langle \tilde{\lambda}, \nabla g(\bar{x})\tilde{u} \rangle = 0, \tilde{\lambda}_0(\|\nabla g_r(\bar{x})\tilde{u}\|^2 - (\nabla g_0(\bar{x})\tilde{u})^2) = 0.$$

Let us split the subsequent proof into the following two cases:

Case I: We have that $x_k = \bar{x}$ for infinitely many k . Suppose without loss of generality that it holds for all $k \in \mathbb{N}$. Then $\nabla g(\bar{x})u_k \in \mathcal{Q}$ whenever $k \in \mathbb{N}$, and thus $\nabla g(\bar{x})\tilde{u} \in \mathcal{Q}$ yielding $\tilde{u} \in T_\Gamma(\bar{x})$ by (3.3) due to the imposed MSCQ. Together with $\langle \tilde{\lambda}, \nabla g(\bar{x})\tilde{u} \rangle = 0$ it tells us that $\tilde{u} \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$. Next we show that $\tilde{\lambda} \in \Lambda(\bar{x}, -\nabla f(\bar{x}); \tilde{u})$. Assuming on the contrary that $\tilde{\lambda} \notin \Lambda(\bar{x}, -\nabla f(\bar{x}); \tilde{u})$, we find $\mu \in \Lambda(\bar{x}, -\nabla f(\bar{x}))$ satisfying $\langle \mu - \tilde{\lambda}, \nabla^2 g(\bar{x})(\tilde{u}, \tilde{u}) \rangle > 0$, and therefore

$$(4.15) \quad \langle \mu - \tilde{\lambda}, \nabla^2 g(\bar{x})(u_k, u_k) \rangle > 0 \text{ for all large } k.$$

Consider the three possibilities here: $\tilde{\lambda} = 0$, $\tilde{\lambda} \in \text{int } \mathcal{Q}^*$, and $\tilde{\lambda} \in \text{bd } \mathcal{Q}^* \setminus \{0\}$. If $\tilde{\lambda} = 0$, take any sequence $t_k \downarrow 0$ and get by $\tilde{u} \in \mathcal{Q}$ the equalities

$$\text{dist}(g(\bar{x} + t_k \tilde{u}); \mathcal{Q}) = \text{dist}(g(\bar{x}) + t_k \nabla g(\bar{x})\tilde{u} + O(t_k^2); \mathcal{Q}) = O(t_k^2) \text{ as } k \rightarrow \infty,$$

which allow us, being combined with MSCQ, to find a bounded sequence of z_k satisfying $g(\bar{x} + t_k \tilde{u} + t_k^2 z_k) \in \mathcal{Q}$. Using it together with $\mu \in \mathcal{Q}^*$ and $\nabla g(\bar{x})^* \mu = -\nabla f(\bar{x}) = 0$

gives us

$$\begin{aligned} 0 \geq \langle \mu, g(\bar{x} + t_k \tilde{u} + t_k^2 z_k) \rangle &= \left\langle \mu, g(\bar{x}) + \nabla g(\bar{x})(t_k \tilde{u} + t_k^2 z_k) + \frac{t_k^2}{2} \nabla^2 g(\bar{x})(\tilde{u}, \tilde{u}) + o(t_k^2) \right\rangle \\ &= \left\langle \mu, \frac{t_k^2}{2} \nabla^2 g(\bar{x})(\tilde{u}, \tilde{u}) + o(t_k^2) \right\rangle. \end{aligned}$$

Dividing the above inequality by $\frac{t_k^2}{2}$ and passing to the limit as $k \rightarrow \infty$, we conclude that

$$\langle \mu, \nabla^2 g(\bar{x})(\tilde{u}, \tilde{u}) \rangle \leq 0 = \langle \tilde{\lambda}, \nabla^2 g(\bar{x})(\tilde{u}, \tilde{u}) \rangle$$

and thus arrive at a contradiction in the case where $\tilde{\lambda} = 0$.

Assuming now that $\tilde{\lambda} \in \text{int } \mathcal{Q}^*$, we get $\lambda^k + \alpha(\mu - \tilde{\lambda}) \in \text{int } \mathcal{Q}^*$ for all $\alpha > 0$ sufficiently small and all $k \in \mathbb{N}$ sufficiently large. Together with $\nabla g(\bar{x})^*(\mu - \tilde{\lambda}) = -\nabla f(\bar{x}) - (-\nabla f(\bar{x})) = 0$ it brings us to

$$\lambda^k + \alpha(\mu - \tilde{\lambda}) \in \Lambda(\bar{x}, x_k^* - \nabla f(\bar{x})) = \Lambda(x_k, x_k^* - \nabla f(x_k)) \text{ and}$$

$$\langle \lambda^k + \alpha(\mu - \tilde{\lambda}), \nabla^2 g(x_k)(u_k, u_k) \rangle > \langle \lambda^k, \nabla^2 g(x_k)(u_k, u_k) \rangle,$$

which contradicts the condition $\lambda^k \in \Lambda(x_k, x_k^* - \nabla f(x_k); u_k)$ that follows from the negation of (3.6).

Finally, we consider the remaining possibility where $\tilde{\lambda} \in \text{bd } \mathcal{Q}^* \setminus \{0\}$. Then it follows from the structure of \mathcal{Q}^* that $-\tilde{\lambda}_0 = \|\tilde{\lambda}_r\| > 0$, $\mu_0 + \|\mu_r\| \leq 0$, and

$$\begin{aligned} \mu_0 - \tilde{\lambda}_0 + \left\langle \frac{\tilde{\lambda}_r}{\|\tilde{\lambda}_r\|}, \mu_r - \tilde{\lambda}_r \right\rangle &= \mu_0 - \tilde{\lambda}_0 + \left\langle \frac{\tilde{\lambda}_r}{\|\tilde{\lambda}_r\|}, \mu_r \right\rangle - \|\tilde{\lambda}_r\| \\ &\leq \mu_0 - \tilde{\lambda}_0 + \frac{\|\tilde{\lambda}_r\|}{\|\tilde{\lambda}_r\|} \|\mu_r\| - \|\tilde{\lambda}_r\| \\ &= \mu_0 - \tilde{\lambda}_0 + \|\mu_r\| - \|\tilde{\lambda}_r\| \leq 0 \end{aligned}$$

with the usage of the Cauchy–Schwarz inequality. Since the latter holds as equality if and only if the two involved vectors are collinear, we have $\mu_0 - \tilde{\lambda}_0 + \langle \frac{\tilde{\lambda}_r}{\|\tilde{\lambda}_r\|}, \mu_r - \tilde{\lambda}_r \rangle = 0$ if and only if $\mu = \alpha \tilde{\lambda}$ for some number $\alpha \in \mathbb{R}$. This implies by employing $\nabla g(\bar{x})^* \mu = \nabla g(\bar{x})^* \tilde{\lambda} = -\nabla f(\bar{x})$ that $(\alpha - 1) \nabla g(\bar{x})^* \tilde{\lambda} = 0$ and consequently that $-\nabla f(\bar{x}) = \nabla g(\bar{x})^* \tilde{\lambda} = 0$ due to $\mu \neq \tilde{\lambda}$. The latter contradicts the condition $\tilde{\lambda} = 0$ by $\tilde{\lambda} \in \Lambda^0(\bar{x}, -\nabla f(\bar{x}))$. Thus we get that

$$\mu_0 - \tilde{\lambda}_0 + \left\langle \frac{\tilde{\lambda}_r}{\|\tilde{\lambda}_r\|}, \mu_r - \tilde{\lambda}_r \right\rangle < 0,$$

which ensures that $\mu_0 - \tilde{\lambda}_0 + \langle \frac{\lambda_r^k}{\|\lambda_r^k\|}, \mu_r - \tilde{\lambda}_r \rangle < 0$ for all k sufficiently large and hence $\zeta_0^k + \|\zeta_r^k\| < 0$ with $\zeta^k := \lambda^k + \alpha_k(\mu - \tilde{\lambda})$, where $\alpha_k > 0$ is chosen to be sufficiently small. Taking into account that $\zeta^k \in \mathcal{Q}^*$ and combining it with $\nabla g(\bar{x})^* \zeta^k = \nabla g(\bar{x})^* \lambda^k = x_k^* - \nabla f(\bar{x})$ tell us that $\zeta^k \in \Lambda(\bar{x}, x_k^* - \nabla f(\bar{x}))$ for all $k \in \mathbb{N}$. It shows together with (4.15) that

$$\langle \zeta^k, \nabla^2 g(\bar{x})(u_k, u_k) \rangle > \langle \lambda^k, \nabla^2 g(\bar{x})(u_k, u_k) \rangle, \quad k \in \mathbb{N},$$

which contradicts the inclusion $\lambda^k \in \Lambda(\bar{x}, x_k^* - \nabla f(\bar{x}); u_k)$ and thus finishes the proof in Case I.

Case II: We have that $x_k \neq \bar{x}$ for all but finitely many k . Suppose without loss of generality that it holds for all $k \in \mathbb{N}$. By passing to a subsequence if necessary, we

get that the sequence $(x_k - \bar{x})/t_k$ converges to some $\bar{v} \in \mathcal{S}$, where $t_k := \|x_k - \bar{x}\|$. Using $g(x_k) \in \mathcal{Q}$, $\langle \lambda^k, g(x_k) \rangle = 0$ for all k , and the closedness of the cone \mathcal{Q} leads us to the relationships

$$\lim_{k \rightarrow \infty} \frac{g(x_k)}{t_k} = \lim_{k \rightarrow \infty} \frac{g(x_k) - g(\bar{x})}{t_k} = \nabla g(\bar{x})\bar{v} \in \mathcal{Q}$$

$$\text{and } 0 = \lim_{k \rightarrow \infty} \frac{\langle \lambda^k, g(x_k) - g(\bar{x}) \rangle}{t_k} = \langle \tilde{\lambda}, \nabla g(\bar{x})\bar{v} \rangle,$$

which imply that $\bar{v} \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$ and consequently that $\nabla g(\bar{x})\bar{v} = 0$ by the assumptions of the theorem. Denoting $w_k := t_k^{-2}(x_k - \bar{x})$ yields

$$(4.16) \quad g(x_k) = g(\bar{x}) + t_k^2 \left(\nabla g(\bar{x})w_k + \frac{1}{2} \nabla^2 g(\bar{x})(\bar{v}, \bar{v}) \right) + o(t_k^2)$$

$$= t_k^2 \left(\nabla g(\bar{x})w_k + \frac{1}{2} \nabla^2 g(\bar{x})(\bar{v}, \bar{v}) \right) + o(t_k^2).$$

To proceed further, we split our analysis into the three subcases:

$$\limsup_{k \rightarrow \infty} \|g(x_k)\|/t_k^2 = \infty, \quad \limsup_{k \rightarrow \infty} \|g(x_k)\|/t_k^2 = 0, \quad \text{and } 0 < \limsup_{k \rightarrow \infty} \|g(x_k)\|/t_k^2 < \infty,$$

which correspond to the following steps in the proof.

Step 1: Assume that $\limsup_{k \rightarrow \infty} \|g(x_k)\|/t_k^2 = \infty$. Then deduce from (4.16) that $\limsup_{k \rightarrow \infty} \|\nabla g(\bar{x})w_k\| = \infty$. Passing to a subsequence if necessary, we get that $\lim_{k \rightarrow \infty} \|\nabla g(\bar{x})w_k\| = \infty$ and that the sequence of $\nabla g(\bar{x})w_k/\|\nabla g(\bar{x})w_k\|$ converges to some element $\nabla g(\bar{x})w \in \mathcal{S}$. It follows from (4.16) that

$$\nabla g(\bar{x})w = \lim_{k \rightarrow \infty} \frac{g(x_k)}{t_k^2 \|\nabla g(\bar{x})w_k\|} \in \mathcal{Q} \quad \text{and} \quad \langle \tilde{\lambda}, \nabla g(\bar{x})w \rangle = \lim_{k \rightarrow \infty} \frac{\langle \lambda^k, g(x_k) \rangle}{t_k^2 \|\nabla g(\bar{x})w_k\|} = 0,$$

which yields $w \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$. Together with $\nabla g(\bar{x})w \neq 0$ it contradicts the imposed assumption.

Step 2: Assume that $\limsup_{k \rightarrow \infty} \|g(x_k)\|/t_k^2 = 0$. In this setting we deduce from (4.16) that

$$0 = \lim_{k \rightarrow \infty} \frac{g(x_k)}{t_k^2} = \lim_{k \rightarrow \infty} \nabla g(\bar{x})w_k + \frac{1}{2} \nabla^2 g(\bar{x})(\bar{v}, \bar{v}).$$

Consequently, there is a vector $\tilde{w} \in \mathbb{R}^n$ such that $\nabla g(\bar{x})\tilde{w} + \frac{1}{2} \nabla^2 g(\bar{x})(\bar{v}, \bar{v}) = 0$, and hence the inclusion $(\tilde{u}, \tilde{\lambda}, \bar{v}, \tilde{w}) \in \mathcal{Z}$ holds. Next we claim that

$$(4.17) \quad \lim_{k \rightarrow \infty} \lambda_0^k \frac{\|\nabla g_r(x_k)u_k\|^2 - (\nabla g_0(x_k)u_k)^2}{t_k^2} = 0.$$

Indeed, assuming $g(x_k) = 0$ yields $\nabla g(x_k)u_k \in \mathcal{Q}$, $\lambda^k \in \mathcal{Q}^*$, and $\langle \lambda^k, \nabla g(x_k)u_k \rangle = 0$ while implying in turn the condition $\lambda_0^k(\|\nabla g_r(x_k)u_k\|^2 - (\nabla g_0(x_k)u_k)^2) = 0$. In the case where $g(x_k) \neq 0$ for infinitely many k , we deduce from (4.14) that the corresponding sequence of $\langle \mathcal{H}(x_k, \lambda^k)u_k, u_k \rangle$ must be bounded. Then taking into account the definition of $\mathcal{H}(x_k, \lambda^k)$, (3.5), and the convergence $g_0(x_k)/t_k^2 \rightarrow 0$ as $k \rightarrow \infty$ verifies the validity of (4.17).

Our next claim is that condition (4.17) ensures that $\rho(\tilde{u}, \tilde{\lambda}, \bar{v}) = 0$. This is certainly true if $\tilde{\lambda} = 0$. In the case where $\tilde{\lambda} \neq 0$ we deduce from $\tilde{\lambda}_0 \leq -\|\tilde{\lambda}\|$ that $\tilde{\lambda}_0 < 0$ and therefore

$$0 = \lim_{k \rightarrow \infty} \frac{\|\nabla g_r(x_k)u_k\|^2 - (\nabla g_0(x_k)u_k)^2}{t_k^2} = \lim_{k \rightarrow \infty} \left\{ \left\| \nabla g_r(\bar{x}) \frac{u_k}{t_k} + \nabla^2 g_r(\bar{x})(\bar{v}, \tilde{u}) \right\|^2 - \left(\nabla g_0(\bar{x}) \frac{u_k}{t_k} + \nabla^2 g_0(\bar{x})(\bar{v}, \tilde{u}) \right)^2 \right\}.$$

Furthermore, we have the equalities

$$0 = \lim_{k \rightarrow \infty} \frac{\langle \lambda^k, \nabla g(x_k)u_k \rangle}{t_k} = \lim_{k \rightarrow \infty} \langle \tilde{\lambda}, \nabla g(\bar{x}) \frac{u_k}{t_k} + \nabla^2 g(\bar{x})(\bar{v}, \tilde{u}) \rangle.$$

Supposing that the sequence of $\nabla g(\bar{x}) \frac{u_k}{t_k}$ is unbounded allows us to find a subsequence of $\nabla g(\bar{x}) \frac{u_k}{t_k} / \|\nabla g(\bar{x}) \frac{u_k}{t_k}\|$, which converges to some $\nabla g(\bar{x})z \in \mathcal{S}$ satisfying

$$\|\nabla g_r(\bar{x})z\|^2 - (\nabla g_0(\bar{x})z)^2 = 0 \text{ and } \langle \tilde{\lambda}, \nabla g(\bar{x})z \rangle = 0.$$

It follows that, depending on the sign of $\nabla g_0(\bar{x})z$, either $z \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$ or $-z \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$, which contradicts the assumption above. Thus the sequence of $\nabla g(\bar{x}) \frac{u_k}{t_k}$ is bounded, and so its subsequence converges to some element $\nabla g(\bar{x})z$ satisfying

$$(4.18) \quad \|\nabla g_r(\bar{x})z + \nabla^2 g_r(\bar{x})(\bar{v}, \tilde{u})\|^2 - (\nabla g_0(\bar{x})z + \nabla^2 g_0(\bar{x})(\bar{v}, \tilde{u}))^2 = 0,$$

and $\langle \tilde{\lambda}, \nabla g(\bar{x})z + \nabla^2 g(\bar{x})(\bar{v}, \tilde{u}) \rangle = 0$.

Employing now Lemma 4.6 shows that $\rho(\tilde{u}, \tilde{\lambda}, \bar{v})$ is nonnegative, and therefore we get that $\rho(\tilde{u}, \tilde{\lambda}, \bar{v}) = 0$. It follows finally from (4.14) and (3.5) that the inequalities

$$\begin{aligned} \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \tilde{\lambda}, g \rangle(\bar{x}))\tilde{u}, \tilde{u} \rangle &\leq \limsup_{k \rightarrow \infty} \langle (\nabla^2 f(x_k) + \nabla^2 \langle \lambda^k, g \rangle(x_k) + \mathcal{H}(x_k, \lambda^k))u_k, u_k \rangle \\ &\leq \frac{1}{\kappa} \end{aligned}$$

hold and thus contradict (4.12) in this setting.

Step 3: Assume that $0 < \limsup_{k \rightarrow \infty} \|g(x_k)\|/t_k^2 < \infty$. Passing to a subsequence if necessary allows us to use that $g(x_k)/t_k^2$ converges to some vector $q \in \mathcal{Q} \setminus \{0\}$. Then we deduce from (4.16) that $\nabla g(\bar{x})w_k$ converges to such a vector $\nabla g(\bar{x})\tilde{w}$ that $q = \nabla g(\bar{x})\tilde{w} + \frac{1}{2}\nabla^2 g(\bar{x})(\bar{v}, \bar{v})$. Since $\langle \lambda^k, g(x_k) \rangle = 0$ for all k , it follows that $\langle \tilde{\lambda}, q \rangle = 0$ and therefore $(\tilde{u}, \tilde{\lambda}, \bar{v}, \tilde{w}) \in \mathcal{Z}$. Then (4.14) together with the nonnegativity of $\langle \mathcal{H}(x_k, \lambda^k)u_k, u_k \rangle$ tells us that the sequence of $\langle \mathcal{H}(x_k, \lambda^k)u_k, u_k \rangle$ is bounded. Let us verify that

$$\limsup_{k \rightarrow \infty} \langle \mathcal{H}(x_k, \lambda^k)u_k, u_k \rangle \geq \frac{\rho(\tilde{u}, \tilde{\lambda}, \bar{v})}{q_0}.$$

Indeed, this claim holds trivially if $\tilde{\lambda} = 0$ by the nonnegativity of $\langle \mathcal{H}(x_k, \lambda^k)u_k, u_k \rangle$ and $\rho(\tilde{u}, 0, \bar{v}) = 0$. Assuming now that $\tilde{\lambda} \neq 0$ yields $\tilde{\lambda}_0 < 0$. Taking into account (4.14), (3.5), as well as the convergence $g_0(x_k)/t_k^2 \rightarrow q_0 > 0$ and $\lambda_0^k \rightarrow \tilde{\lambda}_0 < 0$ tells us that the nonnegative sequence of

$$\begin{aligned} \frac{g_0(x_k)}{-\lambda_0^k t_k^2} \langle \mathcal{H}(x_k, \lambda^k)u_k, u_k \rangle &= \frac{\|\nabla g_r(x_k)u_k\|^2 - (\nabla g_0(x_k)u_k)^2}{t_k^2} \\ &= \left\| \nabla g_r(\bar{x}) \frac{u_k}{t_k} + \nabla^2 g_r(\bar{x})(\bar{v}, \tilde{u}) + \frac{o(t_k)}{t_k} \right\|^2 - \left(\nabla g_0(\bar{x}) \frac{u_k}{t_k} + \nabla^2 g_0(\bar{x})(\bar{v}, \tilde{u}) + \frac{o(t_k)}{t_k} \right)^2 \end{aligned}$$

is bounded. The sequence of $\nabla g(x_k) \frac{u_k}{t_k}$ must also be bounded since otherwise a subsequence of

$$\frac{\nabla g(x_k)u_k/t_k}{\|\nabla g(x_k)u_k/t_k\|} = \frac{\nabla g(\bar{x})u_k/t_k + \nabla^2 g(\bar{x})(\bar{v}, \bar{u}) + o(t_k)/t_k}{\|\nabla g(x_k)u_k/t_k\|}$$

converges to some element $\nabla g(\bar{x})w \in \mathcal{S}$ satisfying

$$\begin{aligned} \|\nabla g_r(\bar{x})w\|^2 - (\nabla g_0(\bar{x})w)^2 &= \lim_{k \rightarrow \infty} \frac{g_0(x_k)}{-\lambda_0^k t_k^2 \|\nabla g(x_k)u_k/t_k\|^2} \langle \mathcal{H}(x_k, \lambda^k)u_k, u_k \rangle = 0 \text{ and} \\ \langle \tilde{\lambda}, \nabla g(\bar{x})w \rangle &= \lim_{k \rightarrow \infty} \frac{\langle \lambda^k, \nabla g(x_k)u_k/t_k \rangle}{\|\nabla g(x_k)u_k/t_k\|} = 0. \end{aligned}$$

Thus, depending on the sign of $\nabla g_0(\bar{x})w$, we have that either $w \in K_\Gamma(\bar{x}, -\nabla f(\bar{x}))$ or $-w \in K_\Gamma(\bar{x}, -\nabla f(\bar{x}))$ with $\nabla g(\bar{x})w \neq 0$, which contradicts the imposed assumption. Hence the sequence of

$$\nabla g(x_k) \frac{u_k}{t_k} = \nabla g(\bar{x}) \frac{u_k}{t_k} + \nabla^2 g(\bar{x})(\bar{v}, \bar{u}) + \frac{o(t_k)}{t_k}$$

is bounded, and so is the one of $\nabla g(\bar{x})u_k/t_k$. Passing to a subsequence if necessary tells us that $\nabla g(\bar{x})u_k/t_k$ converges to some $\nabla g(\bar{x})z$. Thus it follows that

$$\langle \tilde{\lambda}, \nabla g(\bar{x})z + \nabla^2 g(\bar{x})(\bar{v}, \bar{u}) \rangle = \lim_{k \rightarrow \infty} \left\langle \lambda^k, \nabla g(x_k) \frac{u_k}{t_k} \right\rangle = 0.$$

Remembering the definition of ρ in (4.9), we arrive at

$$\begin{aligned} &\limsup_{k \rightarrow \infty} \langle \mathcal{H}(x_k, \lambda^k)u_k, u_k \rangle \\ &= \limsup_{k \rightarrow \infty} \frac{-\lambda_0^k}{g_0(x_k)/t_k^2} \frac{\|\nabla g_r(x_k)u_k\|^2 - (\nabla g_0(x_k)u_k)^2}{t_k^2} \\ &\geq \frac{-\tilde{\lambda}_0}{q_0} \left(\|\nabla g_r(\bar{x})z + \nabla^2 g_r(\bar{x})(\bar{v}, \bar{u})\|^2 - (\nabla g_0(\bar{x})z + \nabla^2 g_0(\bar{x})(\bar{v}, \bar{u}))^2 \right) \\ &\geq \frac{\rho(\tilde{u}, \tilde{\lambda}, \bar{v})}{q_0}. \end{aligned}$$

Combining it with (4.14) and $q_0 = \nabla g_0(\bar{x})\tilde{w} + \frac{1}{2}\nabla^2 g_0(\bar{x})(\bar{v}, \bar{v})$ yields

$$\begin{aligned} &\langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))\tilde{u}, \tilde{u} \rangle + \frac{\rho(\tilde{u}, \tilde{\lambda}, \bar{v})}{\nabla g_0(\bar{x})\tilde{w} + \frac{1}{2}\nabla^2 g_0(\bar{x})(\bar{v}, \bar{v})} \\ &\leq \limsup_{k \rightarrow \infty} \langle (\nabla^2 f(x_k) + \nabla^2 \langle \lambda^k, g \rangle(x_k) + \mathcal{H}(x_k, \lambda^k))u_k, u_k \rangle \leq \frac{1}{\kappa}. \end{aligned}$$

This contradicts (4.11) and thus completes the proof of the theorem. \square

Remark 4.8. Let us briefly comment on the obtained results.

- (i) Condition (a) essentially corresponds to the case when in the limiting process we have $x_k = \bar{x}$ and hence there is no impact of the curvature term \mathcal{H} . The obtained second-order condition itself resembles the standard one from [5, Theorem 30]. The significant difference clearly is that our condition is also valid in the degenerate case and hence it does not rely on the uniqueness of the multiplier. Note also that the necessity of condition (a), with the strict inequality replaced by the nonstrict, is very straightforward; see the beginning of the proof of Theorem 5.6 below.

- (ii) Condition (b), on the other hand, corresponds to the case when $x_k \neq \bar{x}$. Naturally, this is the most difficult case to handle. It is very interesting that, although the curvature term \mathcal{H} is zero at \bar{x} , its limiting behavior around \bar{x} results in the nonnegative term represented by the so-called infimum function $\rho(u, \lambda, v)$. In other words, this means that the curvature of the second-order cone has a positive impact on the tilt stability property. Moreover, this is yet another difference compared to the standard results from [5].

The obtained sufficient conditions for tilt stability of local minimizers in SOCPs are rather involved and are hard to verify. Now we derive simplified ones, which imply the sufficient conditions in both Theorems 4.4 and 4.7. The simplified conditions for tilt stability in SOCPs derived in the next theorem are formulated in the same way in the in-kernel and out-of-kernel cases and resemble those established in [11, Theorem 6.1] for NLPs.

THEOREM 4.9 (simplified sufficient conditions for tilt stability in SOCPs). *In addition to the standing Assumption 4.2 suppose that, given a number $\kappa > 0$, we have the inequality*

$$\langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x})) u, u \rangle > \frac{1}{\kappa}$$

that is valid for all the multipliers

$$\lambda \in \tilde{\Lambda} := \bigcup_{v \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x})) \cap \mathcal{S}} \Lambda(\bar{x}, -\nabla f(\bar{x}); v)$$

and for all the vectors $u \in \mathcal{S}$ satisfying

$$\langle \lambda, \nabla g(\bar{x}) u \rangle = 0 \text{ and } \lambda_0 (\|\nabla g_r(\bar{x}) u\|^2 - (\nabla g_0(\bar{x}) u)^2) = 0.$$

Then \bar{x} is a tilt-stable local minimizer for (1.1) with the prescribed modulus κ .

Proof. Let us verify that the conditions imposed in the theorem ensure the fulfillment of the sufficient conditions for tilt stability in both Theorems 4.4 and 4.7. First, consider Theorem 4.4 and let \bar{u} denote the critical direction with $\nabla g(\bar{x}) \bar{u} \neq 0$ and $\bar{\lambda}$ the corresponding unique multiplier in $\Lambda^0(\bar{x}, -\nabla f(\bar{x}))$ given by (4.5). If $\nabla f(\bar{x}) \neq 0$, we have $\Lambda(\bar{x}, -\nabla f(\bar{x})) = \Lambda^0(\bar{x}, -\nabla f(\bar{x})) = \{\bar{\lambda}\}$ and hence $\bar{\lambda} \in \Lambda(\bar{x}, -\nabla f(\bar{x}); \bar{u})$ follows.

On the other hand, if $\nabla f(\bar{x}) = 0$ we obtain $\bar{\lambda} = 0$ and thus we have to show that $0 \in \Lambda(\bar{x}, 0; \bar{u})$. Consider any multiplier $\lambda \in \Lambda(\bar{x}, 0)$ and take an arbitrary sequence $t_k \downarrow 0$. Since $g(\bar{x}) = 0$ and $\nabla g(\bar{x}) \bar{u} \in \mathcal{Q}$, we have

$$\text{dist}(g(\bar{x} + t_k \bar{u}); \mathcal{Q}) = \text{dist}(t_k \nabla g(\bar{x}) \bar{u} + O(t_k^2); \mathcal{Q}) = O(t_k^2)$$

and consequently there exists a bounded sequence of w_k satisfying $g(\bar{x} + t_k \bar{u} + t_k^2 w_k) \in \mathcal{Q}$. Taking into account that $\lambda \in \mathcal{Q}^*$ and $\nabla g(\bar{x})^* \lambda = 0$, we get

$$\begin{aligned} 0 &\geq \limsup_{k \rightarrow \infty} \frac{\langle \lambda, g(\bar{x} + t_k \bar{u} + t_k^2 w_k) \rangle}{t_k^2} \\ &= \limsup_{k \rightarrow \infty} \frac{\left\langle \lambda, \nabla g(\bar{x})(t_k \bar{u} + t_k^2 w_k) + \frac{t_k^2}{2} \nabla^2 g(\bar{x})(\bar{u}, \bar{u}) \right\rangle}{t_k^2} \\ &= \frac{1}{2} \nabla^2 \langle \lambda, g \rangle(\bar{x})(\bar{u}, \bar{u}). \end{aligned}$$

This yields the inclusion $0 \in \Lambda(\bar{x}, 0; \bar{u})$ and thus shows that condition (4.6) of Theorem 4.4 holds.

Finally, we turn our attention to Theorem 4.7 and verify that both conditions (a) and (b) are fulfilled. The validity of (a) follows from the fact that for every $u \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$ and every $\lambda \in \Lambda(\bar{x}, -\nabla f(\bar{x}); u)$ we have $\langle \lambda, \nabla g(\bar{x})u \rangle = 0$ and $\lambda_0(\|\nabla g_r(\bar{x})u\|^2 - (\nabla g_0(\bar{x})u)^2) = 0$. To verify (b), consider an arbitrary quadruple $(u, \lambda, v, w) \in \mathcal{Z}$ and observe that for any $\mu \in \Lambda(\bar{x}, -\nabla f(\bar{x}))$ we get

$$\begin{aligned} \left\langle \mu - \lambda, \frac{1}{2}\nabla^2 g(\bar{x})(v, v) \right\rangle &= \left\langle \mu - \lambda, \nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v) \right\rangle \\ &= \left\langle \mu, \nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v) \right\rangle \leq 0, \end{aligned}$$

which yields $\lambda \in \Lambda(\bar{x}, -\nabla f(\bar{x}); v)$. Then (b) follows by taking into account that $\rho(u, \lambda, v) \geq 0$ by Lemma 4.6 and $\nabla g_0(\bar{x})w + \frac{1}{2}\nabla^2 g_0(\bar{x})(v, v) > 0$ due to $0 \neq \nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v) \in \mathcal{Q}$. \square

It is important to observe that the *nondegeneracy* condition, which means that $\nabla g(\bar{x})$ has full rank, implies that the multiplier set $\Lambda(\bar{x}, -\nabla f(\bar{x}))$ is a *singleton* and that the term $\rho(u, \lambda, v)$ *vanishes*. Furthermore, in this case the sufficient conditions obtained in Theorems 4.4, 4.7, and 4.9 are *equivalent*. The next example demonstrates that all these phenomena *fail* for programs that exhibit *degeneracy*.

Example 4.10 (tilt stability under degeneracy). Consider the following program of type (1.1):

$$\begin{aligned} \text{minimize } f(x) &:= \frac{1}{4} \left(3x_1^2 + \frac{7 - \sqrt{15}}{\sqrt{15}}(x_2^2 + x_1x_2) \right) - x_3 \quad \text{with } x = (x_1, x_2, x_3) \in \mathbb{R}^3 \\ \text{subject to } g(x) &= (g_1(x), g_2(x), g_3(x)) \in \mathcal{Q}, \end{aligned}$$

where the functions $g_i(x)$, $i = 1, 2, 3$, are defined by

$$\begin{aligned} g_1(x) &:= \frac{1}{2}(x_1^2 + x_2^2 + x_1x_2), \quad g_2(x) := \frac{1}{4} \left(\frac{1}{2}x_1^2 + x_2^2 + x_1x_2 \right), \\ g_3(x) &:= \frac{1}{4}(x_1^2 + x_2^2 + x_1x_2) + x_3. \end{aligned}$$

It is not hard to check by direct calculations that $\bar{x} := (0, 0, 0)$ is a stationary point of the SOCP under consideration. We intend to show that Theorem 4.9 cannot detect that \bar{x} is a tilt-stable local minimizer of this problem while the more complicated Theorem 4.7 can.

Let us first check that MSCQ is fulfilled for the constraint system $g(x) \in \mathcal{Q}$ at \bar{x} represented by

$$\mathcal{Q} = \{q = (q_0, q_r) \in \mathbb{R} \times \mathbb{R}^2 \mid h(q) := \|q_r\| - q_0 \leq 0\}.$$

For any x_1, x_2 denote $\alpha := g_1(x_1, x_2, 0)$ and $\beta := g_2(x_1, x_2, 0)$ and observe that $h(g(x_1, x_2, \hat{x}_3)) = 0$ for $\hat{x}_3 := -\alpha/2 \pm \sqrt{\alpha^2 - \beta^2}$. We have furthermore that $x = (x_1, x_2, x_3) \in g^{-1}(\mathcal{Q})$ if and only if

$$-\frac{\alpha}{2} - \sqrt{\alpha^2 - \beta^2} \leq x_3 \leq -\frac{\alpha}{2} + \sqrt{\alpha^2 - \beta^2} > 0.$$

Consider now $x_3 := -\alpha/2 + \sqrt{\alpha^2 - \beta^2} + \gamma$ with some $\gamma > 0$ and note that $g_3(x_1, x_2, x_3) = \sqrt{\alpha^2 - \beta^2} + \gamma$. Taking into account that $\sqrt{3}\alpha \leq 2\sqrt{\alpha^2 - \beta^2}$ and $\beta \leq \alpha/2$ and

denoting by $L > 0$ a Lipschitz constant of h , we get the distance estimate

$$\begin{aligned} \text{dist}(x; g^{-1}(\mathcal{Q})) &\leq \gamma = \frac{2}{\sqrt{3}} \left(\sqrt{\left(\alpha + \frac{\sqrt{3}}{2}\gamma \right)^2} - \alpha \right) \\ &\leq \frac{2}{\sqrt{3}} \left(\sqrt{\alpha^2 + 2\sqrt{\alpha^2 - \beta^2}\gamma + \gamma^2} - \alpha \right) \\ &= \frac{2}{\sqrt{3}} \left(\sqrt{\beta^2 + (\sqrt{\alpha^2 - \beta^2} + \gamma)^2} - \alpha \right) \\ &= \frac{2}{\sqrt{3}} |h(g(x))| \leq \frac{2}{\sqrt{3}} L \text{dist}(g(x); \mathcal{Q}). \end{aligned}$$

The case where $x_3 := -\alpha/2 - \sqrt{\alpha^2 - \beta^2} + \gamma$ with some $\gamma < 0$ can be treated similarly, and hence the claimed metric subregularity is verified.

To proceed further, calculate the needed values $\nabla f(\bar{x}) = (0, 0, -1)$,

$$\begin{aligned} \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x})) &= \ker \nabla g(\bar{x}) = \mathbb{R}^2 \times \{0\}, \quad \text{and } \Lambda(\bar{x}, -\nabla f(\bar{x})) \\ &= \{(a, b, 1) \mid a \leq -\sqrt{b^2 + 1}\}. \end{aligned}$$

Thus, given a direction $(t, s, 0) \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$, we get that $(a, b, 1) \in \Lambda(\bar{x}, -\nabla f(\bar{x}); (t, s, 0))$ if and only if the pair (a, b) is a solution to the optimization problem:

$$\text{maximize } a(t^2 + s^2 + ts) + \frac{b}{2} \left(\frac{t^2}{2} + s^2 + ts \right) + \frac{1}{2}(t^2 + s^2 + ts) \text{ subject to } a \leq -\sqrt{b^2 + 1}.$$

Taking into account that $t^2 + s^2 + ts > 0$, we can reduce the latter problem to the unconstrained maximization with respect to b only:

$$\text{maximize } -\sqrt{b^2 + 1}(t^2 + s^2 + ts) + \frac{b}{2} \left(\frac{t^2}{2} + s^2 + ts \right) + \frac{1}{2}(t^2 + s^2 + ts),$$

which can be solved explicitly. Denoting $A := t^2 + s^2 + ts$ and $B := t^2/2 + s^2 + ts$, the optimal value of b and consequently of a are calculated by

$$(4.19) \quad b = \frac{B}{\sqrt{4A^2 - B^2}} \quad \text{and} \quad a = -\frac{2A}{\sqrt{4A^2 - B^2}}.$$

Note that the numbers A , B , and $A - B$ are nonnegative, and hence it follows that $b \geq 0$. Furthermore, for any multiplier $\lambda = (a, b, 1)$ and any direction $u = (p, q, 0) \in \ker \nabla g(\bar{x})$ we have

$$(4.20) \quad \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x})) u, u \rangle = \left(a + \frac{b}{4} + 2 \right) p^2 + \left(a + \frac{b}{2} + \frac{7}{2\sqrt{15}} \right) (q^2 + pq).$$

Let us check that the number in (4.20) is nonnegative for all $\lambda \in \Lambda(\bar{x}, -\nabla f(\bar{x}); v)$ with some $v = (t, s, 0) \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x})) \cap \mathcal{S}$. Indeed, note that (4.19) yields $(a+b/4+2) > 0$ and $(a+b/4+2) > (a+b/2+7/2\sqrt{15})$ by $B^2 \leq A^2$. Since $t^2 + s^2 = 1$, we get that $A = \mu B$ for some $\mu \in [1, 2]$, and so

$$\sqrt{4A^2 - B^2} \left(a + \frac{b}{2} + \frac{7}{2\sqrt{15}} \right) = \left(\frac{7}{2\sqrt{15}} \sqrt{4\mu^2 - 1} + \frac{1}{2} - 2\mu \right) B \geq 0$$

for all $\mu \in [8/11, 2] \supset [1, 2]$. The equality is attained therein if and only if $\mu = 2$, which gives us the multiplier $\tilde{\lambda} := (-4/\sqrt{15}, 1/\sqrt{15}, 1)$ and the corresponding directions equal either $v = (\pm 1, 0, 0)$ or $v = (\pm\sqrt{2}/2, \mp\sqrt{2}/2, 0)$. This ensures therefore that for every multiplier $\lambda \neq \tilde{\lambda} = (-4/\sqrt{15}, 1/\sqrt{15}, 1)$ the number in (4.20) is strictly positive whenever direction $u = (p, q, 0)$ with $p^2 + q^2 = 1$ is taken. Indeed, it follows from the estimate

$$\begin{aligned} \left(a + \frac{b}{4} + 2\right)p^2 + \left(a + \frac{b}{2} + \frac{7}{2\sqrt{15}}\right)(q^2 + pq) &\geq \left(a + \frac{b}{2} + \frac{7}{2\sqrt{15}}\right)(p^2 + q^2 + pq) \\ &\geq \frac{1}{2} \left(a + \frac{b}{2} + \frac{7}{2\sqrt{15}}\right). \end{aligned}$$

Considering now the noted multiplier $\tilde{\lambda} = (a, b, 1) = (-4/\sqrt{15}, 1/\sqrt{15}, 1)$, we have that $(a + b/2 + 7/2\sqrt{15}) = 0$, and so the number in (4.20) equals $(a + b/4 + 2)p^2 \geq 0$. Choosing the directions $u = (0, \pm 1, 0)$ gives us the equality

$$\langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \tilde{\lambda}, g \rangle(\bar{x}))u, u \rangle = 0,$$

which shows that the tilt stability of \bar{x} cannot be detected by Theorem 4.9.

On the other hand, with the same $\tilde{\lambda}$ and $u = (0, \pm 1, 0)$ we get $\rho(u, \tilde{\lambda}, v) > 0$ for all the directions $v = (k, l, 0)$ such that $k + 2l \neq 0$ and $k^2 + l^2 = 1$. Indeed, using the values $\langle \nabla^2 g(\bar{x})u, v \rangle = \pm(k/2 + l)(1, 1/2, 1/2)$ and $\nabla g(\bar{x})z = (0, 0, z_3)$ for any $z = (z_1, z_2, z_3)$ and taking into account that $\langle \tilde{\lambda}, \nabla g(\bar{x})z + \langle \nabla^2 g(\bar{x})u, v \rangle \rangle = 0$ tells us that $z_3 = \pm(7 - \sqrt{15})/2\sqrt{15}(k/2 + l)$. Consequently, it yields

$$\nabla g(\bar{x})z + \langle \nabla^2 g(\bar{x})u, v \rangle = \pm \left(\frac{k}{2} + l\right) \left(1, \frac{1}{2}, \frac{7}{2\sqrt{15}}\right)$$

and $49/60(k/2 + l)^2 + (k/2 + l)^2/4 - (k/2 + l)^2 > 0$ if $k/2 + l \neq 0$. This verifies that $\rho(u, \tilde{\lambda}, v) > 0$.

It follows from the proof of Theorem 4.9 that $\lambda \in \Lambda(\bar{x}, -\nabla f(\bar{x}); v)$ for any quadruple $(u, \lambda, v, w) \in \mathcal{Z}$. Since it is the case for the chosen vectors $\tilde{\lambda}$, $u = (0, \pm 1, 0)$, and v equal to either $(\pm 1, 0, 0)$ or $v = (\pm\sqrt{2}/2, \mp\sqrt{2}/2, 0)$, with some $w \in \mathbb{R}^n$, we get $\rho(u, \tilde{\lambda}, v) > 0$ and thus deduce from Theorem 4.7 in the in-kernel case that \bar{x} is a tilt-stable minimizer for the SOCP under consideration.

5. Point-based necessary conditions and criteria for tilt stability. This section is mainly devoted to deriving point-based necessary conditions for tilt-stable minimizers of SOCPs. The obtained results are complementary to the sufficient conditions for such minimizers given in section 4 and, being unified with the latter, allow us to establish *complete point-based characterizations* of tilt stability in second-order cone programming.

As in section 4, it makes sense to consider separately the in-kernel and out-of-kernel cases. We start with the easier out-of-kernel case in which the following *no-gap* necessary condition holds while being different from the sufficient one in Theorem 4.4 only by the nonstrict inequality sign.

THEOREM 5.1 (necessary condition for tilt-stable minimizers of SOCPs in the out-of-kernel case). *Let \bar{x} be a tilt-stable local minimizer of (1.1) with modulus κ under the standing Assumption 4.2. Suppose in addition that there is a vector $\bar{u} \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$ satisfying $\nabla g(\bar{x})\bar{u} \neq 0$ and take the multiplier $\tilde{\lambda}$ defined in (4.5).*

Then for every direction $u \in \mathcal{S}$ satisfying $\langle \bar{\lambda}, \nabla g(\bar{x})u \rangle = 0$ and $\bar{\lambda}_0(\|\nabla g_r(\bar{x})u\|^2 - (\nabla g_0(\bar{x})u)^2) = 0$ we have the condition

$$(5.1) \quad \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \bar{\lambda}, g \rangle(\bar{x}))u, u \rangle \geq \frac{1}{\kappa}.$$

Furthermore, the exact bound of tilt stability of (1.1) at \bar{x} is lower estimated by

$$\text{tilt}(f + \delta_\Gamma, \bar{x}) \geq \sup \left\{ 1 / \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \bar{\lambda}, g \rangle(\bar{x}))u, u \rangle \mid u \in \mathcal{S}, \langle \bar{\lambda}, \nabla g(\bar{x})u \rangle = 0, \right. \\ \left. \bar{\lambda}_0(\|\nabla g_r(\bar{x})u\|^2 - (\nabla g_0(\bar{x})u)^2) = 0 \right\}.$$

Proof. We proceed with the verification of the necessary condition (5.1) while observing that the exact bound estimate follows directly from the proof below. Assuming on the contrary that condition (5.1) fails for the tilt-stable minimizer \bar{x} , find $\tilde{u} \in \mathcal{S}$ satisfying

$$\langle \bar{\lambda}, \nabla g(\bar{x})\tilde{u} \rangle = 0, \bar{\lambda}_0(\|\nabla g_r(\bar{x})\tilde{u}\|^2 - (\nabla g_0(\bar{x})\tilde{u})^2) = 0, \\ \text{and } \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \bar{\lambda}, g \rangle(\bar{x}))\tilde{u}, \tilde{u} \rangle < \frac{1}{\kappa}$$

and then show that \bar{x} is not a tilt-stable local minimizer of (3.2) with modulus κ by using the neighborhood characterization of this property taken from Theorem 3.2. We proceed with considering the following two possible settings.

Suppose first that $\nabla f(\bar{x}) \neq 0$, and so $\nabla g(\bar{x})^* \bar{\lambda} = -\nabla f(\bar{x}) \neq 0$ with $\bar{\lambda} \neq 0$. This yields

$$\nabla g(\bar{x})\bar{u} \in \text{bd } \mathcal{Q} \setminus \{0\}, \|\nabla g_r(\bar{x})\bar{u}\| = \nabla g_0(\bar{x})\bar{u} > 0, \text{ and } -\bar{\lambda}_0 = \|\bar{\lambda}_r\| > 0.$$

Therefore $|\nabla g_0(\bar{x})\tilde{u}| = \|\nabla g_r(\bar{x})\tilde{u}\|$ and, depending on the sign of $\nabla g_0(\bar{x})\tilde{u}$, we have that either $\tilde{u} \in \mathcal{K}_\Gamma(x, -\nabla f(\bar{x}))$ or $-\tilde{u} \in \mathcal{K}_\Gamma(x, -\nabla f(\bar{x}))$. Since the left-hand side of (5.1) is quadratic in u , assume without loss of generality that $\tilde{u} \in \mathcal{K}_\Gamma(x, -\nabla f(\bar{x}))$. Using the same arguments as in the proof of Theorem 4.4 leads us to the equalities

$$\Lambda^0(\bar{x}, -\nabla f(\bar{x})) = \Lambda(\bar{x}, -\nabla f(\bar{x})) = \{\bar{\lambda}\},$$

which imply that $\bar{\lambda} \in \Lambda(\bar{x}, -\nabla f(\bar{x}); \tilde{u})$ and show that (3.6) is violated with $u = \tilde{u}$, $x = \bar{x}$, $x^* = 0$, and $\lambda = \bar{\lambda}$ for every $\eta > 0$. This verifies that \bar{x} is not a tilt-stable local minimizer of (1.1) with modulus κ .

Now suppose that $\nabla f(\bar{x}) = 0$, and hence $\bar{\lambda} = 0$ by (4.5). Choosing an arbitrary sequence $t_k \downarrow 0$ and using MSCQ gives us $x_k \in \Gamma$ satisfying the estimate

$$\|x_k - (\bar{x} + t_k \bar{u})\| \leq \sigma \text{dist}(g(\bar{x} + t_k \bar{u}); \mathcal{Q}) = o(t_k) \text{ for all large } k \in \mathbb{N}.$$

If $g(x_k) \in \text{int } \mathcal{Q}$ holds for infinitely many k , then for these k we have $\tilde{u} \in \mathcal{K}_\Gamma(x_k, 0) = \mathbb{R}^n$ and $\langle (\nabla^2 f(x_k) + \nabla^2 \langle 0, g \rangle(x_k))\tilde{u}, \tilde{u} \rangle < 1/\kappa$ when k is sufficiently large. Thus the neighborhood characterization (3.6) for the tilt stability of \bar{x} with modulus κ fails whenever $\eta > 0$ with $u = \tilde{u}$, $x = x_k$, $x^* = \nabla f(x_k)$, $\lambda = 0$, and large k . Suppose further that $g(x_k) \in \text{bd } \mathcal{Q}$ for all but finitely many k . Since

$$g(x_k) = g(\bar{x}) + \nabla g(\bar{x})(x_k - \bar{x}) + o(\|x_k - \bar{x}\|) = t_k \nabla g(\bar{x})\bar{u} + o(t_k),$$

it follows that $g_0(x_k) = \|g_r(x_k)\| > 0$ for all large k . Next we consider the real-valued function $h(x) := \|g_r(x)\| - g_0(x)$ and observe that the condition

$$\nabla h(x_k) = \frac{g_r(x_k)^*}{\|g_r(x_k)\|} \nabla g_r(x_k) - \nabla g_0(x_k) \neq 0$$

yields the existence of x'_k which is arbitrary close to x_k and such that $h(x'_k) < 0$, i.e., $g(x'_k) \in \text{int } Q$. Thus we are in the same position as just before, which gives us a contradiction with (3.6).

The remaining case is where $\nabla h(x_k) = 0$. By the assumed MSCQ we get from (3.3) that $T_\Gamma(x_k) = \{u \in \mathbb{R}^n \mid \nabla g(x_k)u \in T_Q(g(x_k))\}$, which together with

$$T_Q(g(x_k)) = \left\{ v \in \mathbb{R}^n \mid \left\langle \frac{g_r(x_k)}{\|g_r(x_k)\|}, v_r \right\rangle - v_0 \leq 0 \right\} \supset \left(-1, \frac{g_r(x_k)}{\|g_r(x_k)\|} \right)^\perp$$

ensures the following relationships:

$$\begin{aligned} T_\Gamma(x_k) &\supset \left\{ u \in \mathbb{R}^n \mid \left\langle \frac{g_r(x_k)}{\|g_r(x_k)\|}, \nabla g_r(x_k)u \right\rangle - \nabla g_0(x_k)u = 0 \right\} \\ &= \{u \in \mathbb{R}^n \mid \nabla h(x_k)u = 0\} = \mathbb{R}^n. \end{aligned}$$

Thus $\tilde{u} \in \mathcal{K}_\Gamma(x_k, 0)$, and for every $\eta > 0$ condition (3.6) is violated again with $u = \tilde{u}$, $x = x_k$, $x^* = \nabla f(x_k)$, and $\lambda = 0$ when k is chosen sufficiently large. This completes the proof of the theorem. \square

Combining the results of Theorems 4.4 and 5.1, we arrive at the following effective point-based characterization of tilt stability for SOCPs in the out-of-kernel case.

THEOREM 5.2 (point-based characterization of tilt-stable minimizers for SOCPs in the out-of-kernel case). *In the setting of standing Assumption 4.2, assume that there exists $\bar{u} \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$ such that $\nabla g(\bar{x})\bar{u} \neq 0$. Then \bar{x} is a tilt-stable local minimizer for (1.1) with some modulus $\kappa > 0$ if and only if for every direction $u \in \mathcal{S}$ satisfying $\langle \bar{\lambda}, \nabla g(\bar{x})u \rangle = 0$ and $\bar{\lambda}_0(\|\nabla g_r(\bar{x})u\|^2 - (\nabla g_0(\bar{x})u)^2) = 0$ we have*

$$\langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \bar{\lambda}, g \rangle(\bar{x}))u, u \rangle > 0,$$

where the multiplier $\bar{\lambda}$ is defined in (4.5). Furthermore, the exact bound for tilt stability of (1.1) at \bar{x} is precisely calculated by the formula

$$\begin{aligned} \text{tilt}(f + \delta_\Gamma, \bar{x}) &= \sup \left\{ 1 / \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \bar{\lambda}, g \rangle(\bar{x}))u, u \rangle \mid u \in \mathcal{S}, \langle \bar{\lambda}, \nabla g(\bar{x})u \rangle = 0, \right. \\ &\quad \left. \bar{\lambda}_0(\|\nabla g_r(\bar{x})u\|^2 - (\nabla g_0(\bar{x})u)^2) = 0 \right\}. \end{aligned}$$

Proof. The major observation here is that the modulus of tilt stability is not specified in the formulation of the theorem, which concerns therefore tilt-stable local minimizers of (1.1) with *some* modulus $\kappa > 0$. Thus the claimed characterization and exact bound formula are derived from combining the corresponding results of Theorems 4.4 and 5.1. \square

Remark 5.3. Just as in section 4, the uniqueness of the multiplier in the out-of-kernel case leads to the fact that the obtained conditions coincide with the standard second-order conditions from [5, Theorem 30]; see also Remark 4.5.

Prior to examining the in-kernel case in what follows, we present the following technical lemma that concerns a certain error bound for the second-order cone constraint in (1.1).

LEMMA 5.4 (error bound for second-order cone constraint systems). *Define $h: \mathbb{R}^n \rightarrow \mathbb{R}$ by $h(x) := \|g_r(x)\| - g_0(x)$. Given any $\delta > 0$, there is a number $\varepsilon > 0$ and a neighborhood U of \bar{x} such that for every $x_0 \in U$ with $|h(x_0)| < \varepsilon \|g_r(x_0)\|$ and $\|\nabla h(x_0)\| \geq \delta$ we can find \tilde{x} satisfying*

$$h(\tilde{x}) = 0 \text{ and } \|\tilde{x} - x_0\| \leq 2 \frac{|h(x_0)|}{\delta}.$$

Proof. Let $L > 0$ be a common Lipschitz constant for the mappings $g(x)$ and $h(x)$ on $\bar{x} + \mathbb{B}$. We can see that h is C^2 -smooth around every x with $g_r(x) \neq 0$ and its derivatives calculated by

$$\begin{aligned} \nabla h(x) &= \frac{g_r(x)^*}{\|g_r(x)\|} \nabla g_r(x) - \nabla g_0(x), \\ \nabla^2 h(x) &= \frac{g_r(x)^*}{\|g_r(x)\|} \nabla^2 g_r(x) - \nabla^2 g_0(x) + \frac{1}{\|g_r(x)\|} \nabla g_r(x)^* \left(I - \frac{g_r(x) g_r(x)^*}{\|g_r(x)\|^2} \right) \nabla g_r(x), \end{aligned}$$

where I is the identity matrix. Thus we can find constants $A, B > 0$ such that

$$\|\nabla^2 h(x)\| \leq A + \frac{B}{\|g_r(x)\|} \text{ for all } x \in \bar{x} + \mathbb{B} \text{ with } g_r(x) \neq 0.$$

Take an arbitrary number $\delta > 0$ and then choose $\gamma > 0$ and $0 < \varepsilon \leq \delta/4L$ so small that

$$\frac{A|h(x)| + 2B\varepsilon}{\delta^2} \leq \frac{1}{2} \text{ and } \gamma + 2 \frac{|h(x)|}{\delta} \leq 1 \text{ for all } x \in \mathbb{B}_\gamma(\bar{x}).$$

Pick $x_0 \in \mathbb{B}_\gamma(\bar{x})$ such that $|h(x_0)| < \varepsilon \|g_r(x_0)\|$ and $\|\nabla h(x_0)\| \geq \delta$ and denote $\rho := 2|h(x_0)|/\delta$. Then there exists x_1 satisfying $h(x_0) + \nabla h(x_0)(x_1 - x_0) = 0$ and $\|x_1 - x_0\| \leq |h(x_0)|/\delta$. Furthermore, for all $x \in \mathbb{B}_\rho(x_0) \subset \bar{x} + \mathbb{B}$ we get the inequalities

$$\begin{aligned} \|g_r(x)\| &\geq \|g_r(x_0)\| - L\|x - x_0\| \geq \|g_r(x_0)\| - L\rho = \|g_r(x_0)\| - 2L \frac{|h(x_0)|}{\delta} \\ &\geq \|g_r(x_0)\| \left(1 - 2L \frac{\varepsilon}{\delta} \right) \geq \frac{\|g_r(x_0)\|}{2}, \end{aligned}$$

which bring us to the estimate

$$\|\nabla^2 h(x)\| \leq A + \frac{2B}{\|g_r(x_0)\|}.$$

It implies that $\nabla h(\cdot)$ is Lipschitz continuous on $\mathbb{B}_\rho(x_0)$ with constant $A + 2B/\|g_r(x_0)\|$ and that

$$\tau := \frac{1}{\delta} \left(A + \frac{2B}{\|g_r(x_0)\|} \right) \|x_1 - x_0\| \leq \left(A + \frac{2B}{\|g_r(x_0)\|} \right) \frac{|h(x_0)|}{\delta^2} \leq \frac{A|h(x_0)| + 2B\varepsilon}{\delta^2} \leq \frac{1}{2}.$$

Denoting $\hat{\tau} := \tau^{-1}(1 - \sqrt{1 - 2\tau}) \leq 2$, we apply Robinson's stability result from [30, Theorem 2] to get the existence of $\tilde{x} \in x_0 + \hat{\tau}|h(x_0)|/\delta\mathbb{B} \subset \mathbb{B}_\rho(x_0)$ with $h(\tilde{x}) = 0$. This completes the proof. \square

To proceed further with deriving necessary conditions for and then characterizations of tilt-stable minimizers in the in-kernel case of SOCPs, we invoke the notion

of 2-regularity, which was initiated by Tret'yakov [33] in the case of zero Jacobian and then was strongly developed by Avakov [2], whose definition is mainly reflected in what follows.

DEFINITION 5.5 (2-regularity of mappings). *Let $g : \mathbb{R}^n \rightarrow \mathbb{R}^s$ be twice Fréchet differentiable at $\bar{x} \in \mathbb{R}^n$. We say that g is 2-regular at the point \bar{x} in the direction $v \in \mathbb{R}^n$ if for any $p \in \mathbb{R}^s$ the system*

$$\nabla g(\bar{x})z + \nabla^2 g(\bar{x})(v, w) = p, \quad \nabla g(\bar{x})w = 0$$

admits a solution $(z, w) \in \mathbb{R}^n \times \mathbb{R}^n$.

The next theorem provides two necessary conditions for tilt-stable minimizers in SOCPs with a prescribed modulus in the in-kernel setting. Both of them are parallel to the sufficient conditions of Theorem 4.7 with the replacement of the strict inequality therein by the nonstrict one. Note that only the second condition below requires the assumption on 2-regularity in addition to the standing MSCQ. In order to better understand the structure of the following results, we refer the reader to Remark 4.8.

THEOREM 5.6 (necessary condition for tilt-stable minimizers of SOCPs in the in-kernel case). *Let \bar{x} be a tilt-stable local minimizer for (1.1) with modulus κ under the standing Assumption 4.2, and assume that for every critical direction $u \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$ we have $\nabla g(\bar{x})u = 0$. Suppose further that for every $(u, \lambda, v, w) \in \mathcal{Z}$ fulfilling $\nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v) = 0$ the mapping $g(\cdot)$ is 2-regular in the direction v . Then the following conditions are satisfied:*

(a) *For every $u \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x})) \cap \mathcal{S}$ and every $\lambda \in \Lambda(\bar{x}, -\nabla f(\bar{x}); u)$ we have*

$$\langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle \geq \frac{1}{\kappa}.$$

(b) *For every $(u, \lambda, v, w) \in \mathcal{Z}$ we have*

$$(5.2) \quad \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle + \frac{\rho(u, \lambda, v)}{\nabla g_0(\bar{x})w + \frac{1}{2}\nabla^2 g_0(\bar{x})(v, v)} \geq \frac{1}{\kappa}$$

whenever $\nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v) \neq 0$ and

$$(5.3) \quad \langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle \geq \frac{1}{\kappa} \quad \text{whenever } \rho(u, \lambda, v) = 0.$$

Furthermore, the lower estimate of the exact bound of tilt stability is given by

$$\text{tilt}(f + \delta_\Gamma, \bar{x}) \geq \frac{1}{\min\{\chi_1, \chi_2, \chi_3\}},$$

where the numbers χ_i , $i = 1, 2, 3$, are taken from Theorem 4.7.

Proof. Note first that the claimed lower estimate of the tilt stability bound follows from the proofs of the necessity of conditions (a) and (b). To verify the necessity of (a) for tilt stability, we just pass to the limit as $\eta \downarrow 0$ in the neighborhood characterization (ii) of Theorem 3.2. To prove the necessity of (b), pick a quadruple $(u, \lambda, v, w) \in \mathcal{Z}$ and get for any $t > 0$ that

$$\begin{aligned} & \text{dist}(g(\bar{x} + tv + t^2w); \mathcal{Q}) \\ &= \text{dist}\left(g(\bar{x}) + t\nabla g(\bar{x})v + t^2\left(\nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v)\right) + o(t^2); \mathcal{Q}\right) \\ &= o(t^2). \end{aligned}$$

We split the subsequent proof into the following two cases.

Case I: $\lambda = 0$. In this case $\nabla f(\bar{x}) = 0$, and by the assumed MSCQ for every $t > 0$ we find some $x_t \in \Gamma$ satisfying $\|x_t - (\bar{x} + tv + t^2w)\| = o(t^2)$. If there is a sequence $t_k \downarrow 0$ such that $g(x_{t_k}) \neq 0$, then it follows from the proof of Theorem 5.1 that $\langle \nabla^2 f(\bar{x})u, u \rangle \geq 1/\kappa$. This verifies (5.3) since $\rho(u, 0, v) = 0$. Suppose now that $g(x_t) = 0$ for all small $t > 0$. By

$$g(x_t) = g(\bar{x} + tv + t^2w + o(t^2)) = t^2 \left(\nabla g(\bar{x})w + \frac{1}{2} \nabla^2 g(\bar{x})(v, v) \right) + o(t^2)$$

we have $\nabla g(\bar{x})w + \frac{1}{2} \nabla^2 g(\bar{x})(v, v) = 0$, and thus $g(\cdot)$ is 2-regular at \bar{x} in the direction v . Fixing any $q \in \text{int } \mathcal{Q}$ and applying [13, Proposition 2(c)] ensures the existence of $\beta > 0$ such that

$$\text{dist}(x_t; g^{-1}(t^3q)) \leq \frac{\beta}{\|x_t - \bar{x}\|} \|t^3q - g(x_t)\| = O(t^2) \quad \text{for small } t > 0.$$

Thus we find points x'_t for which $\|x'_t - x_t\| = O(t^2)$ and $g(x'_t) = t^3q \in \text{int } \mathcal{Q}$. This tells us that $\mathcal{K}_\Gamma(x'_t, 0) = \mathbb{R}^n$. Using further Theorem 3.2 with $x = x'_t$, $x^* = \nabla f(x'_t)$, and $\lambda = 0$ yields $\langle \nabla^2 f(x'_t)u, u \rangle \geq \frac{1}{\kappa}$ for all small $t > 0$. To get finally (5.3) in this case, we pass to the limit as $t \downarrow 0$ with taking into account the equality $\rho(u, 0, v) = 0$.

Case II: $\lambda \neq 0$. Since $\lambda \in \mathcal{Q}^*$, it follows that $-\lambda_0 \geq \|\lambda_r\|$, and consequently $-\lambda_0 > 0$. We have furthermore that $|\nabla g_0(\bar{x})u| = \|\nabla g_r(\bar{x})u\|$ and $\langle \lambda, \nabla g(\bar{x})u \rangle = 0$, which implies that either $u \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$ or $-u \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$. This gives us $\nabla g(\bar{x})u = 0$ by the assumption of the theorem. We split the subsequent analysis in this case into the following two steps.

Step 1: $\nabla g(\bar{x})w + \frac{1}{2} \nabla^2 g(\bar{x})(v, v) = 0$. Then $g(\bar{x} + tv + t^2w) = o(t^2)$, and it follows from [13, Proposition 2(c)] due to the imposed 2-regularity of $g(\cdot)$ at \bar{x} in the direction v that there is $\beta > 0$ for which

$$\text{dist}(\bar{x} + tv + t^2w; g^{-1}(0)) \leq \frac{\beta}{\|tv + t^2w\|} \|g(\bar{x} + tv + t^2w)\| = o(t) \quad \text{for small } t > 0.$$

This allows us to find for such t some vectors $x_t \in \mathbb{R}^n$ for which $g(x_t) = 0$ and $\|x_t - (\bar{x} + tv + t^2w)\| = o(t)$. If $\rho(u, \lambda, v) \neq 0$, then there is nothing to prove, and hence we suppose that $\rho(u, \lambda, v) = 0$. Employing Lemma 4.6 in this case gives us a vector $z \in \mathbb{R}^n$ satisfying

$$\begin{aligned} & -\lambda_0 \left(\|\nabla g_r(\bar{x})z + \nabla^2 g_r(\bar{x})(v, u)\|^2 - (\nabla g_0(\bar{x})z + \nabla^2 g_0(\bar{x})(v, u))^2 \right) \\ &= \langle \lambda, \nabla g(\bar{x})z + \nabla^2 g(\bar{x})(v, u) \rangle = 0, \end{aligned}$$

which implies in turn the equalities

$$\begin{aligned} \nabla g(x_t)(u + tz) &= (\nabla g(\bar{x}) + t\nabla^2 g(\bar{x})v + o(t))(u + tz) \\ &= \nabla g(\bar{x})u + t(\nabla g(\bar{x})z + \nabla^2 g(\bar{x})(v, u)) + o(t) = o(t). \end{aligned}$$

It follows from [13, Proposition 2(d)] that there is a number $\beta' > 0$ such that for all $t > 0$ sufficiently small we get vectors $u_t \in \mathbb{R}^n$ satisfying

$$\nabla g(x_t)u_t = 0 \quad \text{and} \quad \|u_t - (u + tz)\| \leq \beta' \frac{\|\nabla g(x_t)(u + tz)\|}{\|x_t - \bar{x}\|} = \frac{o(t)}{t},$$

which tells us that $u_t \in \mathcal{K}_\Gamma(x_t, \nabla g(x_t)^*\lambda)$. Applying [13, Proposition 2(d)] again ensures that the Jacobian matrix $\nabla g(x_t)$ is of full rank, and therefore

$$\Lambda(x_t, \nabla g(x_t)^*\lambda) = \Lambda(x_t, \nabla g(x_t)^*\lambda; u_t) = \{\lambda\}.$$

Employing the neighborhood condition (3.7) from Theorem 3.2 with $x = x_t$, $x^* = \nabla f(x_t) + \nabla g(x_t)^* \lambda$, and $u = u_t/\|u_t\|$ and then taking into account that $\mathcal{H}(x_t, \lambda) = 0$ due to $g(x_t) = 0$, we arrive at

$$\left\langle (\nabla^2 f(x_t) + \nabla^2 \langle \lambda, g \rangle(x_t)) \frac{u_t}{\|u_t\|}, \frac{u_t}{\|u_t\|} \right\rangle \geq \frac{1}{\kappa} \text{ for all small } t > 0.$$

Passing there to the limit as $t \downarrow 0$ verifies (5.3) in this setting.

Step 2: $\nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v) \neq 0$. Remembering that $\lambda \neq 0$ and $\lambda \in N_{\mathcal{Q}}(\nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v))$ in this case gives us the conditions

$$\lambda \in \text{bd } \mathcal{Q}^*, \quad \nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v) \in \text{bd } \mathcal{Q}, \quad \text{and} \quad \hat{\lambda} = \alpha(\nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v))$$

for some $\alpha > 0$. It implies, in particular, that $-\lambda_0 = \|\lambda_r\| > 0$ and $\nabla g_0(\bar{x})w + \frac{1}{2}\nabla^2 g_0(\bar{x})(v, v) = \|\nabla g_r(\bar{x})w + \frac{1}{2}\nabla^2 g_r(\bar{x})(v, v)\| > 0$. Note also that $\nabla g(\bar{x})^* \lambda \neq 0$, since otherwise $\nabla f(\bar{x}) = 0$ and consequently $\lambda = 0$. For the function $h(x) = \|g_r(x)\| - g_0(x)$ we have the representations

$$\begin{aligned} h(\bar{x} + tv + t^2 w) &= \|g_r(\bar{x} + tv + t^2 w)\| - g_0(\bar{x} + tv + t^2 w) \\ &= \|g_r(\bar{x}) + t\nabla g_r(\bar{x})v + t^2(\nabla g_r(\bar{x})w + \frac{1}{2}\nabla^2 g_r(\bar{x})(v, v))\| \\ &\quad - \left(g_0(\bar{x}) + t\nabla g_0(\bar{x})v + t^2(\nabla g_0(\bar{x})w + \frac{1}{2}\nabla^2 g_0(\bar{x})(v, v)) \right) + o(t^2) \\ &= o(t^2), \\ \nabla h(\bar{x} + tv + t^2 w) &= \frac{g_r(\bar{x} + tv + t^2 w)^*}{\|g_r(\bar{x} + tv + t^2 w)\|} \nabla g_r(\bar{x} + tv + t^2 w) - \nabla g_0(\bar{x} + tv + t^2 w) \\ &= \frac{(t^2(\nabla g_r(\bar{x})w + \frac{1}{2}\nabla g_r(\bar{x})(v, v)) + o(t^2))^*}{\|t^2(\nabla g_r(\bar{x})w + \frac{1}{2}\nabla g_r(\bar{x})(v, v)) + o(t^2)\|} (\nabla g_r(\bar{x}) + O(t)) \\ &\quad - (\nabla g_0(\bar{x}) + O(t)) \\ &= \frac{(\lambda_r + o(t^2)/t^2)^*}{\|\lambda_r\| + o(t^2)/t^2} (\nabla g_r(\bar{x}) + O(t)) - (\nabla g_0(\bar{x}) + O(t)) \\ &= \frac{\lambda^*}{\|\lambda_r\|} \nabla g(\bar{x}) + \frac{o(t^2)}{t^2}, \end{aligned}$$

which yield, in particular, the relationships

$$\begin{aligned} \lim_{t \downarrow 0} \|\nabla h(\bar{x} + tv + t^2 w)\| &= \frac{\|\lambda^* \nabla g(\bar{x})\|}{\|\lambda_r\|} \\ &= \frac{\|\nabla f(\bar{x})\|}{\|\lambda_r\|} > 0 \quad \text{and} \quad \lim_{t \downarrow 0} \|\nabla h(x_t)\| \\ &= \frac{\|\lambda^* \nabla g(\bar{x})\|}{\|\lambda_r\|} > 0. \end{aligned}$$

Using now Lemma 5.4 gives us vectors x_t satisfying $h(x_t) = 0$ and $\|x_t - (\bar{x} + tv + t^2 w)\| = o(t^2)$ for small $t > 0$. Defining for such t the multipliers $\lambda^t := \alpha \hat{g}(x_t)/t^2 \in N_{\mathcal{Q}}(g(x_t))$ and employing

$$g(x_t) = g(\bar{x} + tv + t^2 w) + o(t^2) = t^2 \left(\nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v) \right) + o(t^2)$$

ensures that $\lim_{t \downarrow 0} \lambda^t = \lambda$. Since for $\rho(u, \lambda, v) = \infty$ condition (5.2) certainly holds, we consider the case where $\rho(u, \lambda, v)$ is finite. Then Lemma 4.6 gives us $z \in \mathbb{R}^n$ satisfying the equalities

$$\begin{aligned} & \langle \lambda, \nabla g(\bar{x})z + \nabla^2 g(\bar{x})(v, u) \rangle \\ &= 0, \quad \rho(u, \lambda, v) \\ &= -\lambda_0 \left(\|\nabla g_r(\bar{x})z + \nabla^2 g_r(\bar{x})(v, u)\|^2 - (\nabla g_0(\bar{x})z + \nabla^2 g_0(\bar{x})(v, u))^2 \right). \end{aligned}$$

Remembering that $g_0(x_t) = \|g_r(x_t)\|$ leads us to the expressions

$$\begin{aligned} \nabla h(x_t)(u + tz) &= \frac{g_r(x_t)^*}{\|g_r(x_t)\|} \nabla g_r(x_t)(u + tz) - \nabla g_0(x_t)(u + tz) \\ &= \left\langle \frac{\widehat{g}(x_t)}{\|g_r(x_t)\|}, \nabla g(x_t)(u + tz) \right\rangle \\ &= \left\langle \frac{\widehat{g}(x_t)}{\|g_r(x_t)\|}, \nabla g(\bar{x})u + t(\nabla g(\bar{x})z + \nabla^2 g(\bar{x})(v, u)) + o(t) \right\rangle \\ &= t \left\langle \frac{\lambda^t}{\|\lambda_r^t\|}, \nabla g(\bar{x})z + \nabla^2 g(\bar{x})(v, u) \right\rangle + o(t) \\ &= t \left\langle \frac{\lambda^t}{\|\lambda_r^t\|} - \frac{\lambda}{\|\lambda_r\|}, \nabla g(\bar{x})z + \nabla^2 g(\bar{x})(v, u) \right\rangle + o(t) = o(t). \end{aligned}$$

In this way we get, whenever $t > 0$ is sufficiently small, that

$$\begin{aligned} \nabla h(x_t)u_t &= \frac{1}{\|\lambda_r^t\|} \langle \lambda^t, \nabla g(x_t)u_t \rangle = 0 \quad \text{for } u_t := u + tz - \frac{\nabla h(x_t)(u + tz)}{\|\nabla h(x_t)\|^2} \nabla h(x_t)^* \\ &= u + tz + o(t). \end{aligned}$$

It follows that $u_t \in \mathcal{K}_\Gamma(x_t, x_t^* - \nabla f(x_t))$, where $x_t^* := \nabla f(x_t) + \nabla g(x_t)^* \lambda^t \rightarrow 0$ as $t \rightarrow 0$. Since $N_Q(g(x_t)) = \{\alpha \widehat{g}(x_t) \mid \alpha \geq 0\}$ and $x_t^* - \nabla f(x_t) \neq 0$, we easily see that $\Lambda(x_t, x_t^* - \nabla f(x_t)) = \{\lambda^t\}$. Therefore, the neighborhood condition (3.7) from Theorem 3.2 implies that

$$(5.4) \quad \left\langle \left(\nabla^2 f(x_t) + \nabla^2 \langle \lambda^t, g \rangle(x_t) + \mathcal{H}(x_t, \lambda^t) \right) \frac{u_t}{\|u_t\|}, \frac{u_t}{\|u_t\|} \right\rangle \geq \frac{1}{\kappa} \quad \text{for small } t > 0.$$

Using the construction of the curvature function \mathcal{H} in (3.4) together with the limiting relations $\lambda_0^t \rightarrow \lambda_0$ and $g_0(x_t)/t^2 \rightarrow \nabla g_0(\bar{x})w + \frac{1}{2}\nabla^2 g_0(\bar{x})(v, v)$ as $t \downarrow 0$, we get

$$\begin{aligned} \lim_{t \downarrow 0} \langle \mathcal{H}(x_t, \lambda^t)u_t, u_t \rangle &= \lim_{t \downarrow 0} \frac{-\lambda_0^t}{g_0(x_t)} \left(\|\nabla g_r(x_t)u_t\|^2 - (\nabla g_0(x_t)u_t)^2 \right) \\ &= \lim_{t \downarrow 0} \frac{-\lambda_0^t}{g_0(x_t)} \left(\|\nabla g_r(\bar{x})u + t(\nabla g_r(\bar{x})z + \nabla^2 g_r(\bar{x})(v, u)) + o(t)\|^2 \right. \\ &\quad \left. - (\nabla g_0(\bar{x})u + t(\nabla g_0(\bar{x})z + \nabla^2 g_0(\bar{x})(v, u)) + o(t))^2 \right) \\ &= \lim_{t \downarrow 0} \frac{-\lambda_0^t}{g_0(x_t)/t^2} \left(\|\nabla g_r(\bar{x})z + \nabla^2 g_r(\bar{x})(v, u) + \frac{o(t)}{t}\|^2 \right. \\ &\quad \left. - (\nabla g_0(\bar{x})z + \nabla^2 g_0(\bar{x})(v, u) + \frac{o(t)}{t})^2 \right) \\ &= \frac{\rho(u, \lambda, v)}{\nabla g_0(\bar{x})w + \frac{1}{2}\nabla^2 g_0(\bar{x})(v, v)}. \end{aligned}$$

Finally, we arrive at (5.2) by passing to the limit in (5.4) as $t \downarrow 0$. \square

We conclude this section by deriving the following verifiable characterization of tilt-stable minimizers for SOCPs in the in-kernel case.

THEOREM 5.7 (point-based characterization of tilt-stable minimizers for SOCPs in the in-kernel case). *In addition to the standing Assumption 4.2 suppose that $\nabla g(\bar{x})u = 0$ for every critical direction $u \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x}))$ and that for every quadruple $(u, \lambda, v, w) \in \mathcal{Z}$ with $\nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v) = 0$ the mapping $g(\cdot)$ is 2-regular in the direction v . Then \bar{x} is a tilt-stable local minimizer for problem (1.1) with some modulus $\kappa > 0$ if and only if the following conditions hold simultaneously:*

- (a) *For every $u \in \mathcal{K}_\Gamma(\bar{x}, -\nabla f(\bar{x})) \cap \mathcal{S}$ and every $\lambda \in \Lambda(\bar{x}, -\nabla f(\bar{x}); u)$ we get*

$$\langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle > 0.$$

- (b) *For every quadruple $(u, \lambda, v, w) \in \mathcal{Z}$ we get*

$$\langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle + \frac{\rho(u, \lambda, v)}{\nabla g_0(\bar{x})w + \frac{1}{2}\nabla^2 g_0(\bar{x})(v, v)} > 0$$

whenever $\nabla g(\bar{x})w + \frac{1}{2}\nabla^2 g(\bar{x})(v, v) \neq 0$ and

$$\langle (\nabla^2 f(\bar{x}) + \nabla^2 \langle \lambda, g \rangle(\bar{x}))u, u \rangle > 0 \text{ whenever } \rho(u, \lambda, v) = 0.$$

Furthermore, the exact bound of tilt-stability of \bar{x} in (1.1) is calculated by the formula

$$\text{tilt}(f + \delta_\Gamma, \bar{x}) = \frac{1}{\min\{\chi_1, \chi_2, \chi_3\}},$$

where the numbers χ_i , $i = 1, 2, 3$, are taken from Theorem 4.7.

Proof. We derive the claimed statements from the results of Theorems 4.7 and 5.6 by taking into account that the modulus of tilt stability is not specified in this theorem. \square

6. Concluding remarks. The results of this paper and the presented proofs show that the situation with deriving pointwise conditions (sufficient, necessary, and characterizations) for tilt-stable minimizers in nonpolyhedral cone programming without nondegeneracy (and thus with nonunique Lagrange multiplies) is a highly challenging and dramatically more involved task in comparison with polyhedral and/or nondegenerate settings. Our attention to this issue is motivated not only by the theoretical interest but also by the increasing importance of tilt stability for the design and justification of primal-dual numerical algorithms of optimization and obtaining their convergence rates.

Based on advanced recent developments in second-order variational analysis, we are able to establish here comprehensive qualitative and quantitative results on tilt stability in SOCPs generated by a single second-order cone under merely the MSCQ that is far removed not only from nondegeneracy but also from the conventional Robinson constraint qualification in conic programming. However, it remains to study SOCPs described by products of second-order cones and to investigate other remarkable classes of conic programs. Note that the recent results of [12] allow us to calculate the main generalized differential construction of second-order variational analysis implemented here, the so-called subgradient graphical derivative, for the normal cone mappings generated by constraint sets in C^2 -reducible conic programs. This makes it possible to proceed with deriving neighborhood characterizations of tilt-stable minimizers in such programs based on the abstract criterion in [6] and the technique

developed above. On the other hand, establishing point-based conditions for tilt stability required by applications is a big issue for our future research.

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