

Zubov's equation for state-constrained perturbed nonlinear systems*

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Abstract: The paper gives a characterization of the uniform robust domain of attraction for a finite non-linear controlled system subject to perturbations and state constraints. We extend the Zubov approach to characterize this domain by means of the value function of a suitable infinite horizon state-constrained control problem which at the same time is a Lyapunov function for the system. We provide associated Hamilton-Jacobi-Bellman equations and prove existence and uniqueness of the solutions of these generalized Zubov equations.

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1 Introduction

The domain of attraction of a locally asymptotically stable fixed point is an important object in the analysis of the long term behavior of dynamical systems. A seminal result was made by V.I. Zubov [23], who proved that for dynamical systems induced by ordinary differential equations the domain of attraction can be represented as a sublevel set of a solution of a first order partial differential equation, the Zubov equation. Beyond characterizing the domain of attraction, the solution to Zubov's equation also yields a Lyapunov function. Zubov's idea has been used in different contexts [1, 19, 15] and was extended in various ways [3, 10, 12, 7, 11, 18]. Particularly the latter references have established a connection of Zubov's method to optimal control, as they identify the solutions of (generalized versions of) Zubov's equation as optimal value functions of suitable optimal control problems. This connection is established by the observation that Zubov's equation is a Hamilton-Jacobi-Bellman type equation and is thus naturally linked to an optimal control problem. Typically, the generalized Zubov equations derived in these references have to be

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treated in the framework of viscosity solutions, since classical solutions are in general not guaranteed to exist.

In many practical applications of ordinary differential equations the range of physically meaningful or desired states is restricted to a subset of the state space. This fact has long since been recognized in optimal control in which the treatment of state constraints is a classical yet still active research topic. A typical feature of state constraints is that they may affect the continuity of the optimal value function as well as structural properties of this function. One way to avoid discontinuities is to assume suitable controllability properties at the boundary of the state constraint set, see, e.g., [20] or [4, Section IV.5]. Unfortunately, such conditions are often not satisfied in practically relevant settings. However, when state constraints are introduced in an appropriate way, continuity of the optimal value function can be maintained even without any controllability assumptions, as demonstrated by the construction proposed in [5] for a reachability problem and extended to a more general setting of optimal control problems in [2, 14]. This approach allows to incorporate state constraints by means of appropriately chosen penalization functions into the Hamilton-Jacobi framework without losing the continuity of the optimal value function while providing an exact representation of the state constraints.

In this paper we show that the approach of incorporating state constraints from [5, 2] can be successfully carried over to the Zubov equation. We will do this for ordinary differential equations subject to deterministic perturbations, a setting first considered in [10], noting that the extension to the controlled [11] or game theoretic setting [18] is straightforward¹, see also Example 5.2, below. Following this approach, we are able to show that besides continuity all of the main features of Zubov's method can be maintained under state constraints. More precisely, we prove that for the state constrained Zubov problem the optimal value function

- is continuous, cf. Theorem 3.1 and Remark 3.2
- characterizes the domain of attraction as a sub-level set, cf. Theorem 3.1, and is a Lyapunov function for the system, cf. Remark 4.3
- is the unique viscosity solution of an appropriate Hamilton-Jacobi-Bellman equation, i.e., of a generalized Zubov equation, cf. Theorem 4.8.

Moreover, the fact that the continuity of the optimal value function is maintained implies that the approach can be used as the basis for numerical computations, which we illustrate in the last section of this paper. As usual in the Zubov literature, we will formulate our results in parallel for two variants of the Zubov equation which are linked through the Kruzhkov transform, since both variants have their specific advantages.

The paper is organized as follows. In Section 2, we specify the problem and prove useful properties of domains of attractions under state constraints. In Section 3, we provide optimal control characterizations of the uniform domain of attraction for which in Section 4 we define the related generalized Zubov equations and prove existence and uniqueness of the corresponding viscosity solutions. Finally, Section 5 concludes our paper with a numerical

¹In contrast to this, the extension to the stochastic settings in [12, 8, 7] is less straightforward and will be topic of future research.

example. Throughout the paper, in order to avoid repetitions we provide detailed proofs only for those results whose proofs significantly differ from the unconstrained case.

2 Setting of the problem

We consider a non-linear system subject to time varying perturbations given by the ordinary differential equation

$$\dot{y}(t) = f(y(t), u(t)), \quad t \geq 0, \quad (2.1)$$

where $f : \mathbb{R}^N \times U \rightarrow \mathbb{R}^N$ is the vector field, u is the perturbation input that is supposed to take values in the compact set $U \subset \mathbb{R}^m$ (with $m \geq 1$). We assume that the system has a locally robustly asymptotically stable equilibrium (for details on this assumption see (A2), below); without loss of generality we assume that this equilibrium is located in the origin.

We define the set of *admissible* input functions as

$$\mathcal{U} := \left\{ u : (0, +\infty) \longrightarrow \mathbb{R}^m \text{ measurable, } u(t) \in U \text{ a.e.} \right\}$$

and assume that the vector field f in (2.1) satisfies the following regularity assumptions (where $\|\cdot\|$ is the Euclidean norm and $B(y, R)$ the ball around y with radius $R > 0$ in \mathbb{R}^N):

(A1) (i) For each $R > 0$ there exists $L_R \geq 0$ such that:

$$\|f(x, u) - f(y, u)\| \leq L_R(\|x - y\|), \quad \forall x, y \in B(0, R), \quad \forall u \in U.$$

(ii) There exists $C_f > 0$ with $\|f(x, u)\| \leq C_f, \forall x \in \mathbb{R}^N, \forall u \in U$.

Note that (ii) may be assumed without loss of generality replacing f by $f/(1 + \|f\|)$ if necessary. Observe that this transformation only changes the speed of the trajectories but not their convergence behaviour. In particular, the domain of attraction as defined below does not change under this transformation.

Under assumption (A1), for any $u \in \mathcal{U}$ and for any $x \in \mathbb{R}^N$, there exists a unique absolutely continuous trajectory $y = y_x^u(\cdot)$ satisfying (2.1) for almost all $t \in \mathbb{R}$ and $y(0) = x$. For any $T > 0$, the set of all feasible trajectories on $[0, T]$ starting in x will be denoted as

$$S_{[0, T]}(x) := \left\{ y : [0, T] \rightarrow \mathbb{R}^N \mid \begin{array}{l} y = y_x^u(\cdot) \text{ is absolutely continuous, satisfies (2.1) for} \\ \text{some } u \in \mathcal{U} \text{ and almost all } t \in (0, T), \text{ and } y(0) = x \end{array} \right\}.$$

Again under assumption (A1), for any $T > 0$ and $x \in \mathbb{R}^N$, $S_{[0, T]}(x)$ is a compact set in $W^{1,1}(0, T)$ for the topology of $C([0, T]; \mathbb{R}^N)$. Moreover, the set-valued map $x \rightsquigarrow S_{[0, T]}(x)$ is Lipschitz continuous from \mathbb{R}^N in $C([0, T]; \mathbb{R}^N)$.

For every $t \geq 0$ and $x \in \mathbb{R}^N$, we denote by $\mathcal{R}(x, t)$ the reachable set, at time t , starting from the position x , that is

$$\mathcal{R}(x, t) := \{ y_x^u(t) \mid y_x^u(\cdot) \in S_{[0, t]}(x) \}.$$

It is clear that under assumption (A1) the set $\mathcal{R}(x, t)$ is bounded for every $x \in \mathbb{R}^N$ and every $t \geq 0$. Moreover, we will use the set $\mathcal{R}_\infty(x) := \bigcup_{t \geq 0} \mathcal{R}(x, t)$.

Throughout the paper, we assume that the origin $x = 0$ is an equilibrium for the system (2.1) and that we have the following local uniform asymptotic stability property

- (A2) (i) $f(0, 0) = 0$;
(ii) There exists a \mathcal{KL} -function² $\bar{\beta}$ and $\bar{r} > 0$ such that

$$\|y_x^u(t)\| \leq \bar{\beta}(\|x\|, t) \quad \forall x \in B(0, \bar{r}), \forall t \geq 0, \forall u \in \mathcal{U}.$$

Note that (ii) implies the existence of $\bar{\varepsilon} > 0$ such that

$$y_x^u(t) \in B(0, \bar{r}/2) \quad \forall x \in B(0, \bar{\varepsilon}), \forall t \geq 0, \forall u \in \mathcal{U}. \quad (2.2)$$

Our goal in this paper is to characterize the robust domain of attraction of the origin under state constraints. Here, “robust” is to be understood in the sense “for all perturbations $u \in \mathcal{U}$ ”. In order to formally define this object, we consider an open set $\Omega_{ad} \subset \mathbb{R}^N$ of *admissible states*, whose complement $\Omega_{ad}^c = \mathbb{R}^N \setminus \Omega_{ad}$ defines the *state constraints* or *obstacles* we impose on the system. We assume that the state constraints do not touch the equilibrium $x = 0$, i.e., that there exists $\bar{r} > 0$ such that $B(0, \bar{r}) \subset \Omega_{ad}$. We may without loss of generality assume that this $\bar{r} > 0$ is the same as in (A2).

Using this admissible set, we define the set of admissible input functions as

$$\mathcal{U}_{ad}(x) := \{u \in \mathcal{U} \mid y_x^u(t) \in \Omega_{ad} \text{ for all } t \geq 0\}$$

and the *robust domain of attraction of $x = 0$ w.r.t. Ω_{ad}* as

$$\mathcal{D} := \{x \in \mathbb{R}^N \mid \mathcal{U}_{ad}(x) = \mathcal{U} \text{ and } \|y_x^u(t)\| \rightarrow 0 \text{ for all } u \in \mathcal{U}\}.$$

In words, \mathcal{D} contains all those initial values x for which all corresponding solutions $y_x^u(\cdot)$ stay in Ω_{ad} and converge to 0.

In order to relate the domain of attraction to a Zubov type PDE, similar to [10] we need to consider a uniform version of \mathcal{D} . In extension to the “uniform attraction rate” imposed in [10], here we also need to consider a “uniform distance” to the obstacle set Ω_{ad}^c . To this end, we define the distance between a point $x \in \mathbb{R}^N$ and a set $A \subset \mathbb{R}^N$ by $\text{dist}(x, A) := \inf_{y \in A} \|x - y\|$. Then, for $\delta \geq 0$, we define the set of δ -admissible input functions as

$$\mathcal{U}_{ad, \delta}(x) := \{u \in \mathcal{U} \mid \text{dist}(y_x^u(t), \Omega_{ad}^c) > \delta \text{ for all } t \geq 0\}.$$

Note that $\mathcal{U}_{ad, 0}(x) = \mathcal{U}_{ad}(x)$. The *robust domain of uniform attraction of $x = 0$ w.r.t. Ω_{ad}* is then defined by

$$\mathcal{D}_0 := \left\{ x \in \mathbb{R}^N \left| \begin{array}{l} \text{there exists } \delta > 0 \text{ with } \mathcal{U}_{ad, \delta}(x) = \mathcal{U} \\ \text{and there exists a function } \beta(t) \rightarrow 0 \text{ as } t \rightarrow \infty \text{ with} \\ \|\|y_x^u(t)\| \leq \beta(t) \text{ for all } t \in [0, \infty) \text{ and all } u \in \mathcal{U} \end{array} \right. \right\}.$$

²A function $\beta : \mathbb{R}^+ \times \mathbb{R}^+ \rightarrow \mathbb{R}^+$ is of class \mathcal{KL} if it is continuous, $\beta(0, t) = 0$, $r \mapsto \beta(r, t)$ is strictly increasing for each $t \geq 0$ and $t \mapsto \beta(r, t)$ is strictly decreasing to 0 for all $r > 0$.

The set \mathcal{D}_0 is a uniform version of \mathcal{D} in the sense that for each $x \in \mathcal{D}_0$ the trajectories have a positive distance δ to Ω_{ad}^c and a speed of convergence $\beta(t)$ towards 0 which both do not depend on $u \in \mathcal{U}$. Note that both δ and $\beta(t)$ are allowed to depend on x , hence we require uniformity only w.r.t. $u \in \mathcal{U}$. The relation between \mathcal{D}_0 and \mathcal{D} will be analysed below.

An important object for our analysis is the first hitting time with respect to $B(0, \bar{\varepsilon})$, that is the first time a given trajectory reaches the ball $B(0, \bar{\varepsilon})$ with $\bar{\varepsilon}$ from (2.2):

$$t(x, u) := \inf\{t \geq 0 \mid y_x^u(t) \in B(0, \bar{\varepsilon})\}$$

with the standard convention $\inf \emptyset = \infty$. By definition, for $x \in B(0, \bar{\varepsilon})$ the hitting time is zero. Moreover, from (2.2) we obtain $y_x^u(t) \in B(0, \bar{r}/2)$ for all $t \geq t(x, u)$ and because of $B(0, \bar{r}) \subset \Omega_{ad}$ this implies

$$\text{dist}(y_x^u(t), \Omega_{ad}^c) > \bar{r}/2 \quad \forall t \geq t(x, u). \quad (2.3)$$

This observation in particular implies $B(0, \bar{\varepsilon}) \subset \mathcal{D}_0$, with $\delta = \bar{r}/2$ and $\beta(t) = \bar{\beta}(\bar{\varepsilon}, t)$. The following lemma establishes further links between the sets \mathcal{D}_0 , \mathcal{D} and the hitting time $t(x, u)$.

Lemma 2.1 (i) $\mathcal{D}_0 = \{x \in \mathbb{R}^N \mid \mathcal{U}_{ad, \delta}(x) = \mathcal{U} \text{ for some } \delta > 0 \text{ and } \sup_{u \in \mathcal{U}} t(x, u) < \infty\}$.

(ii) \mathcal{D}_0 is open.

(iii) $\mathcal{D}_0 = \text{int } \mathcal{D}$.

Proof: (i) Let $x \in \mathcal{D}_0$, pick the corresponding $\delta > 0$ and $\beta(t)$ and let $T > 0$ be such that $\beta(t) < \bar{\varepsilon}$ for all $t \geq T$. Then we have $\mathcal{U}_{ad, \delta}(x) = \mathcal{U}$ and $\|y_x^u(T)\| \leq \beta(x) < \bar{\varepsilon}$ which shows $y_x^u(T) \in B(0, \bar{\varepsilon})$. Hence, $t(x, u) < T$ for all $u \in \mathcal{U}$ and thus $\sup_{u \in \mathcal{U}} t(x, u) \leq T < \infty$.

Conversely, let $x \in \mathbb{R}^N$ be such that $T := \sup_{u \in \mathcal{U}} t(x, u) < \infty$ and $\mathcal{U}_{ad, \delta}(x) = \mathcal{U}$ for some $\delta > 0$. Then, for $t \geq T$ we have $\|y_x^u(t)\| \leq \bar{\beta}(\bar{r}, t)$. Hence, for $t \geq T$ we can choose $\beta(t) := \bar{\beta}(\bar{r}, t)$. For $t \in [0, T]$, since the vector field is bounded, the solutions are uniformly bounded, say, by a constant $M > 0$. Setting $\beta(t) := M$ for $t \in [0, T]$ then yields the function $\beta(t)$ with the desired properties.

(ii) Let $x \in \mathcal{D}_0$ and consider the corresponding $\delta > 0$ and $\beta(\cdot)$. Let $T > 0$ be such that $\beta(t) < \bar{\varepsilon}/2$ for all $t \geq T$. By continuity of the trajectories in the initial value x there exists a neighbourhood $B(x, \varepsilon)$ of x such that $\|y_x^u(t) - y_z^u(t)\| < \max\{\delta/2, \bar{\varepsilon}/2\}$ for all $z \in B(x, \varepsilon)$ and all $t \in [0, T]$. This implies $\text{dist}(y_z^u(t), \Omega_{ad}^c) > \delta/2$ for all $z \in B(x, \varepsilon)$ and all $t \in [0, T]$ as well as $y_z^u(T) \in B(0, \bar{\varepsilon})$, hence $t(z, u) \leq T$ for all $u \in \mathcal{U}$. Together with (2.3) this implies $\mathcal{U}_{ad, \min\{\delta/2, \bar{r}/2\}}(z) = \mathcal{U}$, hence by (i) we can conclude $z \in \mathcal{D}_0$. Thus, $B(x, \varepsilon) \subset \mathcal{D}_0$ and consequently \mathcal{D}_0 is open.

(iii) Since $\mathcal{D}_0 \subseteq \mathcal{D}$ follows directly from the definition, the inclusion $\text{int } \mathcal{D}_0 \subseteq \text{int } \mathcal{D}$ is clear and by (ii) it implies $\mathcal{D}_0 \subseteq \text{int } \mathcal{D}$.

To see the converse inclusion $\text{int } \mathcal{D} \subset \mathcal{D}_0$, let $x \in (\text{int } \mathcal{D}) \setminus \mathcal{D}_0$. Since $x \notin \mathcal{D}_0$, by (i) either $\sup_{u \in \mathcal{U}} t(x, u) = \infty$ or $\mathcal{U}_{ad, \delta} \neq \mathcal{U}$ for all $\delta > 0$ must hold. If the former holds, then as in [10, Proof of Proposition 2.3(iv)] we obtain $x \in \partial \mathcal{D}$, contradicting $x \notin \text{int } \mathcal{D}$.

Hence assume $T := \sup_{u \in \mathcal{U}} t(x, u) < \infty$. Then $\mathcal{U}_{ad, \delta} \neq \mathcal{U}$ for all $\delta > 0$ must hold and thus we can conclude the existence of $u_n \in \mathcal{U}$ and $t_n > 0$ with $\text{dist}(y_x^{u_n}(t_n), \Omega_{ad}^c) \rightarrow 0$ as

$n \rightarrow \infty$. Since (2.3) and $t(x, u_n) \leq T$ imply $t_n \leq T$ and f is bounded by (A1)(ii), both t_n and $y_n := y_x^{u_n}(t_n)$ must be bounded. Hence, the Lipschitz assumption (A1)(i) yields that for any $\varepsilon > 0$ the set $\{y_\xi^{u_n}(t_n) \mid \xi \in B(x, \varepsilon)\}$ contains a ball $B(y_n, \rho)$ with $\rho > 0$ independent of n . For sufficiently large n this implies $B(y_n, \rho) \not\subset \Omega_{ad}$. This means that $u_n \notin \mathcal{U}_{ad,0}(x_n)$ for some $x_n \in B(x, \varepsilon)$ and consequently $x_n \notin \mathcal{D}$. Since $\varepsilon > 0$ was arbitrary, this implies $x \in \partial\mathcal{D}$, again contradicting $x \notin \text{int } \mathcal{D}$. Hence, $(\text{int } \mathcal{D}) \setminus \mathcal{D}_0 = \emptyset$, implying $\text{int } \mathcal{D} \subset \mathcal{D}_0$. \square

Remark 2.2 In words, (iii) implies that \mathcal{D}_0 and \mathcal{D} coincide except for sets with void interior. Moreover, \mathcal{D}_0 is the largest subset of \mathcal{D} which can be expressed as a strict sub-level set $V^{-1}((-\infty, c)) := \{x \in \mathbb{R}^N \mid V(x) < c\}$ of a continuous function $V : \mathbb{R}^N \rightarrow \mathbb{R}$, which is the core of Zubov's method. Indeed, the fact that Zubov's method naturally leads to a characterization of an open set is also the reason for choosing the set of admissible states Ω_{ad} as an open set and not — as it is more common in the control literature — as a closed set. \square

3 Characterization of the set \mathcal{D}_0

As an important step towards characterizing \mathcal{D}_0 via a Zubov type equation, in this section we introduce an optimal control problem whose optimal value function characterizes \mathcal{D}_0 . The idea used here has been developed in many papers for the case of differential systems without state constraints, see, e.g., [7, 10, 11, 12]. The main novelty here is the extension to the case when the trajectories are constrained to an open set Ω_{ad} . To define the adequate optimal control problem, we consider a running cost $g : \mathbb{R}^N \times U \rightarrow \mathbb{R}$ satisfying:

(A3) (i) g is continuous and for each $R > 0$ there exists $L_{g,R} \geq 0$ such that:

$$|g(x, u) - g(y, u)| \leq L_g \|x - y\|, \quad \forall x, y \in B(0, R), \quad \forall u \in U.$$

(ii) $g(0, u) = 0$ for any $u \in U$, and for every $c > 0$ we have:

$$g_c := \inf\{g(x, u) \mid \|x\| \geq c, u \in U\} > 0.$$

(iii) $\int_0^\infty g(y_x^u(t), u(t)) dt$ is finite if $t(x, u)$ is finite.

In the particular but common case of exponential convergence, i.e., when $\bar{\beta}$ in (A2) is of the form $\bar{\beta}(r, t) = Ce^{-\sigma t}r$ for appropriate constants $C, \sigma > 0$, any g satisfying (i) and (ii) will also satisfy (iii). In this case, any function of the form $g(x, u) = (\lambda_1 + \lambda_2 \|u\|)\|x\|$ for $\lambda_1 > 0$ and $\lambda_2 \geq 0$ will satisfy (i)–(iii). If the convergence in (A2) is slower than exponential, g must be sufficiently “flat” around 0 in order to satisfy (iii). A general construction which shows the existence of such a g whenever (A2) holds can be found, e.g., in [11, Section 3]. Note that the converse implication to (iii), i.e., that the integral value is infinite if $t(x, u) = \infty$, follows from (ii) with $c = \bar{\varepsilon}$.

Like in the other Zubov papers cited above, the cost g measures the convergence of the solutions to 0. In order to incorporate state constraints into our setting, we need another cost that indicates when a given trajectory violates the constraints in Ω_{ad} . For this, we consider a function $h : \mathbb{R}^N \rightarrow [0, +\infty]$ satisfying:

(A4) h is locally Lipschitz continuous on Ω_{ad} , $h(x) = \infty$ iff $x \notin \Omega_{ad}$, and $h(x_n) \rightarrow \infty$ for $x_n \rightarrow x \notin \Omega_{ad}$, $h(0) = 0$.

The function h can be chosen for instance as $h(x) := \frac{\|x\|}{\text{dist}(x, \Omega_{ad}^c)}$. This general expression always satisfies the requirements of (A4), however, for particular domains Ω_{ad} , simpler expressions may exist and could be preferred. In any case, the exact expression of the function h is not important as long as h satisfies (A4). The same remark holds for the function g satisfying (A3).

For $x \in \mathbb{R}^N$, we introduce the value function $V : \mathbb{R}^N \rightarrow \mathbb{R}^+ \cup \{+\infty\}$ defined by:

$$V(x) := \sup_{u \in \mathcal{U}} \sup_{t \in [0, \infty)} \left\{ \int_0^t g(y_x^u(\tau), u(\tau)) d\tau + h(y_x^u(t)) \right\}. \quad (3.4)$$

Consider also the Kruzhkov transformed optimal value function $v : \mathbb{R}^N \rightarrow [0, 1]$ given by:

$$v(x) := 1 - e^{-V(x)} = \sup_{u \in \mathcal{U}} \sup_{t \in [0, \infty)} \left\{ 1 - \exp \left(- \int_0^t g(y_x^u(\tau), u(\tau)) d\tau - h(y_x^u(t)) \right) \right\}. \quad (3.5)$$

The following theorem shows the relation between V , v and \mathcal{D}_0 .

Theorem 3.1 Assume (A1)–(A2) and let g and h satisfy (A3) and (A4), respectively. Then,

- (i) $\mathcal{D}_0 = \{x \in \mathbb{R}^N \mid V(x) < \infty\} = \{x \in \mathbb{R}^N \mid v(x) < 1\}$.
- (ii) V is continuous on \mathcal{D}_0 with $V(x_n) \rightarrow \infty$ if $x_n \rightarrow x \notin \mathcal{D}_0$ or $\|x_n\| \rightarrow \infty$.
- (iii) v is continuous on \mathbb{R}^N .

Proof: (i) First, by definition of v , we obtain immediately the equality between the two sets $\{x \in \mathbb{R}^N \mid V(x) < \infty\}$ and $\{x \in \mathbb{R}^N \mid v(x) < 1\}$. It remains to show the first identity.

Let $x \in \mathcal{D}_0$. For proving $V(x) < \infty$, using [10, Proposition 3.1(i)] it is sufficient to show

$$\sup_{y \in \mathcal{R}(x, t), t \in [0, \infty)} h(y) < \infty,$$

because the arguments in [10, Proof of Proposition 3.1(i)] already prove that the supremum over the integral in the definition of V is finite.

Since $\|y_x^u(t)\| \leq \beta(t)$ for any $u \in \mathcal{U}$, the reachable set $\mathcal{R}_\infty(x)$ is bounded, hence $\overline{\mathcal{R}_\infty(x)}$ is compact. Moreover, since $\mathcal{U} = \mathcal{U}_{ad, \delta}(x)$ for some $\delta > 0$, $\overline{\mathcal{R}_\infty(x)} \subset \Omega_{ad}$ follows and since h is continuous on Ω_{ad} it will attain a (finite) maximum on $\overline{\mathcal{R}_\infty(x)}$ which shows the claim.

Let $x \notin \mathcal{D}_0$. Then either the existence of $\beta(t)$ or the existence of δ in the definition of \mathcal{D}_0 is not satisfied. In the first case, we can proceed as in [10, Proof of Proposition 3.1(i)] in order to conclude $V(x) = \infty$. In the second case, the existence of β again implies that $\mathcal{R}_\infty(x)$ is bounded. Moreover, the non-existence of δ implies the existence of controls u_n and times t_n with $\text{dist}(y_x^{u_n}(t_n), \Omega_{ad}^c) \rightarrow 0$. Since $x_n := y_x^{u_n}(t_n)$ lies in the bounded set $\mathcal{R}_\infty(x)$, we can find a subsequence x_{n_k} converging to some $x_0 \notin \Omega_{ad}$. This implies $h(x_{n_k}) \rightarrow \infty$ and since $V(x) \geq h(x_{n_k})$ we obtain $V(x) = \infty$.

(ii) In order to prove continuity, let $x, y \in \mathcal{D}_0$. Then we obtain

$$\begin{aligned} |V(x_1) - V(x_2)| &\leq \sup_{u \in \mathcal{U}} \sup_{t \in [0, \infty)} \left(\int_0^t |g(y_{x_1}^u(\tau), u(\tau)) - g(y_{x_2}^u(\tau), u(\tau))| d\tau \right. \\ &\quad \left. + |h(y_{x_1}^u(t)) - h(y_{x_2}^u(t))| \right) \\ &\leq \sup_{u \in \mathcal{U}} \int_0^\infty |g(y_{x_1}^u(\tau), u(\tau)) - g(y_{x_2}^u(\tau), u(\tau))| d\tau \end{aligned} \quad (3.6)$$

$$+ \sup_{u \in \mathcal{U}} \sup_{t \in [0, \infty)} |h(y_{x_1}^u(t)) - h(y_{x_2}^u(t))|. \quad (3.7)$$

For (3.6), we can follow the proof of [10, Proposition 3.1(iii)] in order to see that this term vanishes as $x_2 \rightarrow x_1$. In order to show the same for (3.7), let $u \in \mathcal{U}$ be arbitrary. Then, as $x_1 \in \mathcal{D}_0$, the convergence $h(y_{x_1}^u(t)) \rightarrow 0$ holds as $t \rightarrow \infty$, and due to the uniformity of this convergence in u , the continuous dependence of the trajectories on the initial value and the local asymptotic stability, this convergence is uniform for all $u \in \mathcal{U}$ and all initial values near x_1 . This means that we can find a neighborhood $B(x_1, \delta)$ and a function $\gamma(t) \rightarrow 0$ such that $|h(y_{x_2}^u(t))| \leq \gamma(t)$ holds for all $x_2 \in B(x_1, \delta)$. This implies that the supremum over t in (3.7) is attained on a finite interval $[0, T]$. On a compact time interval, however, the map $x \mapsto h(y_x^u(t))$ is continuous in x uniformly in u and t , implying that (3.7) also tends to 0 as $x_2 \rightarrow x_1$. This shows the desired continuity.

For the second assertion, first consider $x_n \rightarrow x \notin \mathcal{D}_0$. Since $x \notin \mathcal{D}_0$, either the existence of $\beta(t)$ or the existence of δ must fail for this point. In the first case, $V(x_n) \rightarrow \infty$ follows by the same arguments as in the proof of [10, Proposition 3.1(iv)]. In the second case, as in the proof of Lemma 2.1(iii) we find a bounded sequence of times $t_n \in [0, \infty)$ and a sequence of functions $u_n \in \mathcal{U}$ with $\text{dist}(y_{x_n}^{u_n}(t_n), \Omega_{ad}^c) \rightarrow 0$ as $n \rightarrow \infty$. Since t_n is bounded and $x_n \rightarrow x$ we can conclude $\text{dist}(y_{x_n}^{u_n}(t_n), \Omega_{ad}^c) \rightarrow 0$ and thus $V(x_n) \geq h(y_{x_n}^{u_n}(t_n)) \rightarrow \infty$. In case $\|x_n\| \rightarrow \infty$ the assertion follows similar to [10, Proof of Proposition 3.1(iv)].

(iii) Follows immediately from (i) and (ii) using the formula $v(x) = 1 - e^{-V(x)}$. \square

Remark 3.2 Both the original V and the transformed v have advantages and disadvantages. The original function V has an easier definition and is thus easier to handle in proofs. However, it is unbounded and only defined on the subset \mathcal{D}_0 which makes it inconvenient for constructing numerical approximations and turns the determination of \mathcal{D}_0 via V into a free boundary value problem. In contrast to this, the function v is defined, bounded and continuous on the whole \mathbb{R}^N . Thus, it is much better suited for numerical computations. However, this advantage comes at the expense of a more complicated definition. \square

4 Zubov's equation

Theorem 3.1 paves the way for characterizing the robust domain of uniform attraction \mathcal{D}_0 under state constraints as a sub-level set of solutions to suitable partial differential equations. For this, we need to work within the framework of viscosity solutions which we recall in the next definition. Let $\mathcal{F} : \mathbb{R}^N \times \mathbb{R} \times \mathbb{R}^N \rightarrow \mathbb{R}$ be a continuous function, and consider the general PDE on an open set $\mathcal{O} \subseteq \mathbb{R}^N$

$$\mathcal{F}(x, v(x), Dv(x)) = 0 \quad x \in \mathcal{O}. \quad (4.8)$$

Definition 4.1 (i) A lower semi-continuous (l.s.c.) function $v : \mathcal{O} \rightarrow \mathbb{R}$ is a viscosity super-solution of (4.8) if for any $\phi \in C^1(\mathcal{O})$ and any $x \in \operatorname{argmin}_{\mathcal{O}}(v - \phi)$, we have $\mathcal{F}(x, v(x), D\phi(x)) \geq 0$.

(ii) An upper semi-continuous (u.s.c.) function $v : \mathcal{O} \rightarrow \mathbb{R}$ is a viscosity sub-solution of (4.8) if for any $\phi \in C^1(\mathcal{O})$ and $x \in \operatorname{argmax}_{\mathcal{O}}(v - \phi)$, we have $\mathcal{F}(x, v(x), D\phi(x)) \leq 0$.

(iii) A function $v \in C(\mathcal{O})$ is a viscosity solution of (4.8) if it is sub- and super-solution. \square

The first step to proving that V and v satisfy Zubov-type equations is to establish suitable dynamic programming principles for these functions. The arguments for establishing these principles are standard (see, e.g., [4]) and the corresponding proofs are thus omitted.

For the formulation of these principles, we define the set

$$\Omega_0 := \{x \in \mathcal{D}_0 \mid V(x) > h(x)\}.$$

Since \mathcal{D}_0 is open and both V and h are continuous functions on \mathcal{D}_0 , the set Ω_0 is open. Moreover, the definition of V implies $V(x) \geq h(x)$, hence for $x \in \mathcal{D}_0 \setminus \Omega_0$ the equation $V(x) = h(x)$ holds.

Moreover, for any open set $O \subset \mathbb{R}^N$ define the time

$$t_0(x, u, O) := \inf\{t \geq 0 \mid y_x^u(t) \notin O\}.$$

Then, the following dynamic programming principle holds:

Proposition 4.2 Abbreviate

$$G(x, t, u) := \int_0^t g(y_x^u(\tau), u(\tau)) d\tau \quad (4.9)$$

Then, under (A1), the following assertions are satisfied:

(i) For all $x \in \mathcal{D}_0$ and all $T \geq 0$ we have:

$$V(x) = \sup_{u \in \mathcal{U}} \left\{ G(x, \min\{T, t_0(x, u, \Omega_0)\}, u) + V(y_x^u(\min\{T, t_0(x, u, \Omega_0)\})) \right\}. \quad (4.10)$$

(ii) For all $x \in \mathbb{R}^N$ and all $T \geq 0$, we have:

$$V(x) = \sup_{u \in \mathcal{U}} \max \left(G(x, T, u) + V(y_x^u(T)), \sup_{t \in [0, T]} \{G(x, t, u) + h(y_x^u(t))\} \right). \quad (4.11)$$

(iii) For all $x \in \mathcal{D}_0$ and all $T \geq 0$ we have:

$$v(x) = \sup_{u \in \mathcal{U}} \left\{ 1 + (v(y_x^u(\min\{T, t_0(x, u, \Omega_0)\})) - 1)G(x, \min\{T, t_0(x, u, \Omega_0)\}, u) \right\}. \quad (4.12)$$

(iv) For all $x \in \mathbb{R}^N$, for all $T \geq 0$, we have:

$$v(x) = \sup_{u \in \mathcal{U}} \max \left(1 + (v(y_x^u(T)) - 1)G(x, T, u), \sup_{t \in [0, T]} \left\{ 1 - e^{-G(x, t, u) - h(y_x^u(t))} \right\} \right). \quad (4.13)$$

From (4.10), standard viscosity solution arguments yield that for all $x \in \Omega_0$ the function V satisfies the Hamilton-Jacobi-Bellman equation

$$\inf_{u \in U} \{-DV(x)f(x, u) - g(x, u)\} = 0 \quad (4.14)$$

in the viscosity sense. Likewise, (4.12) implies that the transformed function $v = 1 - e^{-V}$ is a viscosity solution of

$$\inf_{u \in U} \{-Dv(x)f(x, u) - g(x, u)(1 - v(x))\} = 0 \quad (4.15)$$

for all $x \in \Omega_0$. Moreover, for all $x \in \mathcal{D}_0$ and all $t \geq 0$, the dynamic programming principle (4.11) implies

$$V(x) \geq \sup_{u \in \mathcal{U}} \left\{ \int_0^t g(y_x^u(\tau), u(\tau)) d\tau + V(y_x^u(t)) \right\} \quad (4.16)$$

which yields that V is a viscosity super-solution of (4.14) on the whole set \mathcal{D}_0 . Similarly, (4.13) implies that v is a viscosity super-solution of (4.15) on the whole \mathbb{R}^N .

Remark 4.3 Inequality (4.16) implies

$$V(y_x^u(t)) \leq V(x) - \int_0^t g(y_x^u(\tau), u(\tau)) d\tau < V(x)$$

for all $u \in \mathcal{U}$ and all $x \neq 0$. This implies that the function V is a Lyapunov function for the system for all perturbation inputs $u \in \mathcal{U}$. Thus, the modifications we have made in order to incorporate the obstacles Ω_{ad}^c have not affected the Lyapunov function property of Zubov's original approach. \square

The following theorem introduces the two central equations of this paper, i.e., the two variants of the Zubov equation with state constraints. Its proof follows again by standard viscosity solution arguments from the respective dynamic programming principles.

Theorem 4.4 Assume (A1)–(A2) and let g and h be two functions satisfying (A3) and (A4) respectively. Then the value function V is a viscosity solution of the HJB equation

$$\min \left(\inf_{u \in U} \{-DV(x)f(x, u) - g(x, u)\}, V(x) - h(x) \right) = 0 \quad \text{for } x \in \mathcal{D}_0. \quad (4.17)$$

Likewise, the function v is viscosity solution of the equation

$$\min \left(\inf_{u \in U} \{-Dv(x)f(x, u) - g(x, u)(1 - v(x))\}, v(x) - e^{-h(x)} \right) = 0 \quad \text{for } x \in \mathbb{R}^N. \quad (4.18)$$

While the fact that V and v solve (4.17) and (4.18), respectively, follows by standard viscosity solution techniques, establishing uniqueness of the solutions to these equations requires more arguments which we provide next.

In order to incorporate the state constraints into our analysis, we start with a localized version of [10, Proposition 3.5].

Proposition 4.5 (i) Let W be a l.s.c. super-solution of (4.14) on an open set $O \subset \mathbb{R}^N$. Then W satisfies a super-optimality principle, that is for any $x \in O$, we have

$$W(x) \geq \sup_{u \in \mathcal{U}} \sup_{t \in [0, t_0(x, u, O)]} \left\{ \int_0^t g(y_x^u(\tau), u(\tau)) d\tau + W(y_x^u(t)) \right\}. \quad (4.19)$$

(ii) Let U be a u.s.c. sub-solution of (4.14) on an open set $O \subset \mathbb{R}^N$ and let $\tilde{U} : O \rightarrow \mathbb{R}$ be a continuous function with $U \leq \tilde{U}$. Then U satisfies a sub-optimality principle, that is for any $x \in O$ we have

$$U(x) \leq \sup_{u \in \mathcal{U}} \inf_{t \in [0, t_0(x, u, O)]} \left\{ \int_0^t g(y_x^u(\tau), a(\tau)) d\tau + \tilde{U}(y_x^u(t)) \right\}. \quad (4.20)$$

(iii) Let w be a l.s.c. super-solution of (4.15) on an open set $O \subset \mathbb{R}^N$. Then w satisfies a super-optimality principle, that is for any $x \in O$, we have

$$w(x) \geq \sup_{u \in \mathcal{U}} \sup_{t \in [0, t_0(x, u, O)]} \{1 + (w(y_x^u(t)) - 1)G(x, t, u)\}. \quad (4.21)$$

with G from (4.9).

(iv) Let \tilde{u} be a u.s.c. sub-solution of (4.15) on an open set $O \subset \mathbb{R}^N$. Then \tilde{u} satisfies a sub-optimality principle, that is for any $x \in O$, we have

$$\tilde{u}(x) \leq \sup_{u \in \mathcal{U}} \sup_{t \in [0, t_0(x, u, O)]} \{1 + (\tilde{u}(y_x^u(t)) - 1)G(x, t, u)\}. \quad (4.22)$$

with G from (4.9).

The proof of this proposition is omitted as it follows along the same lines as similar results in the literature: The link between the notion of sub-solution (resp. super-solution) and sub-optimality (super-optimality) has been analysed in several papers, we refer to [21, 22] or [4, Chapter III] for a proof using viscosity techniques. Another proof based on non smooth analysis is given in [13].

The main consequence of the above proposition are the following comparison principle for the equations (4.17) and (4.18).

Proposition 4.6 (i) Let $W : \Omega \rightarrow \mathbb{R}$, $\Omega \subset \mathbb{R}^N$ open with $\mathcal{D}_0 \subseteq \Omega$, be a l.s.c. super-solution of (4.17). Then $W(x) \geq V(x)$ holds for all $x \in \mathcal{D}_0$ with V from (3.4).

If, moreover, $W(x_n) \rightarrow \infty$ holds whenever $x_n \rightarrow x \in \partial\Omega$, then $\Omega = \mathcal{D}_0$ follows.

(ii) Let $U : \Omega \rightarrow \mathbb{R}$, $\mathcal{D}_0 \subseteq \Omega$, be a continuous sub-solution of (4.17) with $U(0) \leq 0$. Then $U(x) \leq V(x)$ holds for all $x \in \mathcal{D}_0$ with V from (3.4).

Proof: (i) We first show the inequality $W(x) \geq V(x)$ on \mathcal{D}_0 .

Let $x \in \mathcal{D}_0$, fix some $\varepsilon > 0$ and pick u_ε and t_ε such that the supremum on the right hand side of (3.4) is attained up to this ε . This implies

$$V(x) \leq \int_0^{t_\varepsilon} g(y_x^{u_\varepsilon}(\tau), u_\varepsilon(\tau)) d\tau + h(y_x^{u_\varepsilon}(t_\varepsilon)) + \varepsilon.$$

Since W is a super-solution of (4.17) on Ω , it is also a super-solution of (4.14) on Ω . Hence, (4.20) holds with $O = \Omega$. Since any solution starting in \mathcal{D}_0 stays in \mathcal{D}_0 for all future times, we can use (4.19) in order to obtain the inequality

$$W(x) \geq \int_0^{t_\varepsilon} g(y_x^{u_\varepsilon}(\tau), u_\varepsilon(\tau)) d\tau + W(y_x^{u_\varepsilon}(t_\varepsilon)).$$

Combining the two inequalities and using that W being a super-solution of (4.17) implies $W(x) \geq h(x)$ we obtain

$$W(x) - V(x) \geq W(y_x^{u_\varepsilon}(t_\varepsilon)) - h(y_x^{u_\varepsilon}(t_\varepsilon)) - \varepsilon \geq -\varepsilon$$

which shows the claim since $\varepsilon > 0$ was arbitrary.

In order to show the identity $\Omega = \mathcal{D}_0$ we again use that W is a super-solution of (4.14) on Ω and thus satisfies (4.19) with $O = \Omega$. We show that this property implies that for each $x \notin \mathcal{D}_0$ the value $W(x)$ cannot be finite. To this end, note that for every $u \in \mathcal{U}$

$$W(x) \geq \int_0^t g(y_x^u(\tau), u(\tau)) d\tau + W(y_x^u(t))$$

holds for all $t \in [0, t_0(x, u, \Omega)]$. Now, we distinguish three cases.

(a) $t_0(x, u, \Omega) = \infty$ and $y_x^u(t) \notin \mathcal{D}_0$ for all $t \geq 0$. In this case the integral grows unboundedly and $W(x) = \infty$ follows.

(b) $t_0(x, u, \Omega) = \infty$ and $y_x^u(t) \in \mathcal{D}_0$ for some $t \geq 0$. In this case by continuity of the solution there exist times t_n with $x_n := y_x^u(t_n) \in \mathcal{D}_0$ and $\text{dist}(x_n, \partial\mathcal{D}_0) \rightarrow 0$ as $n \rightarrow \infty$. Then, the inequality $W \geq V$ on \mathcal{D}_0 and the fact that $V(x_n) \rightarrow \infty$ for $\text{dist}(x_n, \partial\mathcal{D}_0) \rightarrow 0$ implies $W(x_n) \rightarrow \infty$ for $n \rightarrow \infty$ and hence again $W(x) = \infty$ follows.

(c) $t_0(x, u_\varepsilon, \Omega)$ is finite. Then, for a sequence $(t_n)_n$ such that $t_n \rightarrow t_0(x, u, \Omega_1)$, the positions $x_n := y_x^u(t_n)$ satisfy $\text{dist}(x_n, \partial\Omega) \rightarrow 0$. This implies that $W(y_x^u(t_n)) \rightarrow \infty$ and hence $W(x) = \infty$.

Summarizing, in all cases we obtain $W(x) = \infty$, which shows that W cannot attain finite values outside \mathcal{D}_0 and thus $\Omega = \mathcal{D}_0$ must hold.

(ii) Consider the set

$$\Omega_1 := \{x \in \Omega \mid U(x) > h(x)\},$$

which is open since both U and h are continuous. For $x \notin \Omega_1$ we have

$$U(x) \leq h(x) \leq V(x)$$

which proves the desired inequality. In order to prove the inequality for $x \in \Omega_1$, observe that since U is a sub-solution of (4.17), it is a sub-solution of (4.14) on Ω_1 , hence (4.20) holds with $\tilde{U} = U$ on $O = \Omega_1$. Hence, for each $\varepsilon > 0$ we can find a control u_ε satisfying

$$U(x) \leq \int_0^t g(y_x^{u_\varepsilon}(\tau), u_\varepsilon(\tau)) d\tau + U(y_x^{u_\varepsilon}(t)) + \varepsilon$$

for all $t \in [0, t_0]$ with $t_0 := t_0(x, u_\varepsilon, \Omega_1)$. Now, from (4.16) we get

$$V(x) \geq \int_0^t g(y_x^{u_\varepsilon}(\tau), u_\varepsilon(\tau)) d\tau + V(y_x^{u_\varepsilon}(t))$$

for all $t \geq 0$. Hence, for all $t \in [0, t_0]$ we get

$$U(x) - V(x) \leq U(y_x^{u_\varepsilon}(t)) - V(y_x^{u_\varepsilon}(t)) + \varepsilon. \quad (4.23)$$

Now, in case t_0 is finite, by definition of Ω_1 we get

$$U(y_x^{u_\varepsilon}(t_0)) - V(y_x^{u_\varepsilon}(t_0)) = h(y_x^{u_\varepsilon}(t_0)) - V(y_x^{u_\varepsilon}(t_0)) \leq 0.$$

In case $t_0 = \infty$, since $x \in \mathcal{D}_0$ we get $y_x^{u_\varepsilon}(t) \rightarrow 0$, when t tends to $+\infty$, and thus by continuity of U and V for each sufficiently large t we get

$$U(y_x^{u_\varepsilon}(t)) - V(y_x^{u_\varepsilon}(t)) \leq U(0) - V(0) + \varepsilon \leq \varepsilon.$$

Hence, in both cases we get $U(x) - V(x) \leq 2\varepsilon$ which shows the claim. \square

The next proposition yields the analogous comparison result for the scaled optimal value function v .

Proposition 4.7 (i) Let $w : \mathbb{R}^N \rightarrow \mathbb{R}$ be a continuous super-solution of (4.18) satisfying $w(0) \geq 0$. Then $w(x) \geq v(x)$ holds for all $x \in \mathbb{R}^N$ with v from (3.5).

(ii) Let $u : \Omega \rightarrow \mathbb{R}$, $\mathcal{D}_0 \subseteq \Omega$, be a continuous sub-solution of (4.18) with $u(0) \leq 0$. Then $u(x) \leq v(x)$ holds for all $x \in \mathbb{R}^N$ with v from (3.5).

Proof: The proof follows with the same arguments as that of Proposition 4.5 using (4.21) and (4.22) in place of (4.19) and (4.20), cf. also [10, Propositions 3.6 and 3.7]. \square

As a direct consequence of Propositions 4.6 and 4.7, we get the following uniqueness result.

Theorem 4.8 Assume (A1)–(A2) and let g and h be two functions satisfying (A3) and (A4) respectively. Then the following assertions hold.

(i) Let $\Omega \subset \mathbb{R}^N$, $\mathcal{D}_0 \subseteq \Omega$ be a set on which a continuous viscosity solution W of (4.17) exists satisfying $W(x_n) \rightarrow \infty$ for each sequence $x_n \in \Omega$ with $x_n \rightarrow x \in \partial\Omega$. Then $\Omega = \mathcal{D}_0$.

(ii) The function V from (3.4) is the unique continuous viscosity solution of (4.17) on $\Omega = \mathcal{D}_0$ with $V(0) = 0$.

(iii) The function v from (3.5) is the unique continuous viscosity solution of (4.18) on \mathbb{R}^N with $v(0) = 0$.

Proof. (i) In order to prove the inclusion $\mathcal{D}_0 \subseteq \Omega$ we proceed by contradiction. Assuming that this inclusion does not hold, we can find $x \in \partial\Omega \cap \mathcal{D}_0$. Let $x_n \rightarrow x$ be a sequence in Ω . Since W is a viscosity solution of (4.17) on $\mathcal{D}_0 \cap \Omega$, by Proposition 4.6(ii) we obtain $W(x_n) \leq V(x_n) \rightarrow V(x) < \infty$ for $n \rightarrow \infty$, implying $W(x_n) \not\rightarrow \infty$. This contradicts $x \in \partial\Omega$.

The assertion (ii) follows immediately from Proposition 4.6, while (iii) is an immediate consequence of Proposition 4.7. \square

5 Numerical simulations

Theorem 4.8 in conjunction with Theorem 3.1 shows that by computing a solution to the Hamilton-Jacobi equations (4.17) or (4.18) we can compute the uniform domain of attraction \mathcal{D}_0 . Due to the non-linearity of the state system, it is in general hopeless to get an analytical expression of the value function V or of v . However, the equations can be approximately solved numerically, at least in low space dimensions. To this end, it is advantageous to consider an approximation of the Kruzhkov transformed optimal value function v via (4.18). This is because v is bounded and because the Zubov equation that characterizes v can be solved on the whole space \mathbb{R}^N and thus in particular on a large bounded closed domain \mathcal{K} that contains \mathcal{D}_0 .

Consider a simple grid \mathcal{G} with edges x_i and a uniform time step $h > 0$. For simplicity, we assume that 0 is a node of the grid \mathcal{G} . An approximation of v can be obtained by using a first order Semi-Lagrangian scheme (see [4, Appendix A]) leading to the problem of solving the following equation.

$$v_h(x_i) = \max \left\{ \max_{u \in \mathcal{U}} (1 - hg(x_i, u))v_h(x_i + hf(x_i, u)) + hg(x_i, u), e^{h(x_i)} \right\}, \quad (5.1)$$

where v_h is piecewise affine on each simplex of the grid \mathcal{G} , and satisfies $v_h(0) = 0$. The Semi-Lagrangian scheme is taken here just an example of possible approximation scheme. Other numerical schemes can also be considered like finite difference, Markov Chain methods and in order to improve the accuracy of the method the approximations $hg(x_i, u)$ (rectangle rule) and $x_i + hf(x_i, u)$ (forward Euler approximation) can be replaced by more accurate quadrature rules and ODE solvers, respectively. In the numerical example, below, we used a rectangular quadrature rule and the forward Euler with step size $h/50$ for this purpose.

Unfortunately, (5.1) has a singularity in 0 since $g(0, u)$ vanishes. This fact was already pointed out in [9] for the case without obstacle. As a consequence of this singularity, the usual fixed-point arguments do not apply and the convergence is not guaranteed. Following the ideas introduced in [9], we use a regularization of (4.18) by introducing an approximation of g :

$$g_\varepsilon(x, u) := \max\{g(x, u), \varepsilon\},$$

where $\varepsilon > 0$ is a small parameter. Then the regularized Zubov equation that is considered is

$$\min \left(\inf_{u \in \mathcal{U}} \{-Dv^\varepsilon(x)f(x, u) - g(x, u) + g_\varepsilon(x, u)v^\varepsilon(x)\}, v^\varepsilon(x) - e^{-h(x)} \right) = 0 \quad \text{for } x \in \mathbb{R}^N. \quad (5.2)$$

A good feature of this regularization is that v_ε converges to v when ε tends to 0. An even better feature is that for all sufficiently small $\varepsilon > 0$ the sub-level characterization $\mathcal{D}_0 = \{x \in \mathbb{R}^N \mid v^\varepsilon(x) < 1\}$ still holds; hence in terms of computing \mathcal{D}_0 the regularization (5.2) does not introduce any errors. The arguments used in [9] for proving this property can be carried over in a straightforward way to the setting in this paper. The discretization of (5.2) leads to the equation:

$$v_h^\varepsilon(x_i) = \max \left\{ \max_{u \in \mathcal{U}} (1 - hg_\varepsilon(x_i, u))v_h^\varepsilon(x_i + hf(x_i, u)) + hg_\varepsilon(x_i, u), e^{h(x_i)} \right\}, \quad (5.3)$$

with again v_h^ε being a piecewise affine function on each simplex of the grid \mathcal{G} and $v_h^\varepsilon(0) = 0$. Now a straightforward application of fixed-point arguments allow to conclude the existence and uniqueness of the solution v_h^ε of (5.3).

In order to illustrate this numerical procedure we provide two examples.

Example 5.1 The first example is a system considered in [17, Section C.2] whose unperturbed version was presented in [15]. Its dynamics is given by the two-dimensional system

$$\begin{aligned}\dot{x}_1 &= -x_1 + x_2 \\ \dot{x}_2 &= -\frac{1}{10}x_1 - 2x_2 - x_1^2 + \left(u + \frac{1}{10}\right)x_1^3\end{aligned}\tag{5.4}$$

In order to meet the global boundedness requirement (A1)(ii), as in [17] we perform the transformation

$$\tilde{f}(x, u) = \frac{5f(x, u)}{\sqrt{25 + \|f(x, u)\|^2}},$$

which does not change the domain of attraction. Using the perturbation range $u \in [-0.03, 0.03]$ the origin is a locally asymptotically stable equilibrium. In fact, it is even exponentially stable, hence the running cost $g(x, u) = 10^{-3}\|x\|_2$ satisfies all requirements in (A3). Using this function, we get the numerical approximation for v depicted in Figure 5.1.

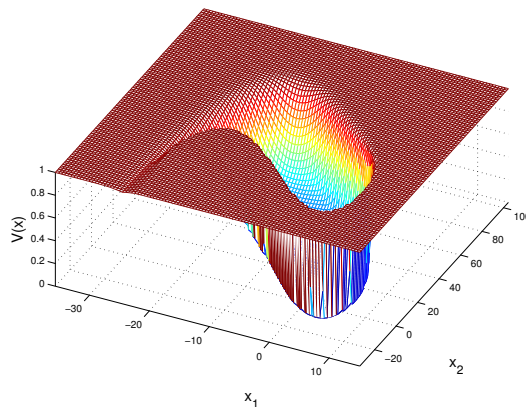


Figure 5.1: Solution of Zubov's equation without obstacles

In the next computation we introduce a ball with radius $r = 2$ around the point $y = (-12, 47.5)^T$ as an obstacle. Since we are solving the Kruzhkov transformed Zubov equation (4.18) we directly specify the function $1 - e^{-h(x)}$ as

$$1 - e^{-h(x)} = \min \{1, \max\{0, 1 + sr^2 - s\|x - y\|^2\}\}$$

with s being a tuning parameter defining the slope of the function around the obstacle. For our computation we use $s = 0.1$. With this choice, one easily checks that this function satisfies (A4). The numerical solution of (4.18) is shown in Figure 5.2(left), an enlargement of the area around the obstacle is given in Figure 5.2(right). In this enlargement, the continuity of the solution around the obstacle is clearly visible.

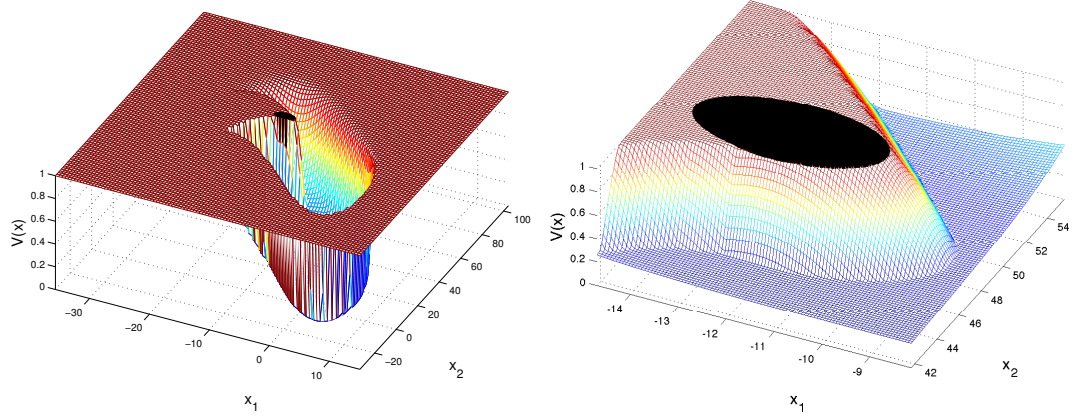


Figure 5.2: Solution of Zubov's equation with state constraints, whole solution (left) and detail around obstacle (right)

The numerical computations were performed using the time step $h = 0.2$ and piecewise bilinear interpolation on an adaptively generated [16] rectangular finite element grid with about 30000 vertices. The resulting discretized equation (5.3) was solved using the Gauss-Seidel type increasing coordinate iteration described in [16, Section 3], the maximization was performed by discretizing U with 3 equidistant perturbation values. \square

Example 5.2 Our second example demonstrates that our results also apply to the minimization problem in which Zubov's equation delivers a control Lyapunov function [11]. We consider the three dimensional nonholonomic integrator studied by Brockett [6], given by

$$\begin{aligned}\dot{x}_1 &= u_1 \\ \dot{x}_2 &= u_2 \\ \dot{x}_3 &= x_2 u_1.\end{aligned}$$

We use the stage cost $g(x, u) = \|x\|_2^2/10$. The obstacle is given by the cylinder

$$\{x = (x_1, x_2, x_3)^T \in \mathbb{R}^3 \mid (x_1 - y_1)^2 + (x_2 - y_2)^2 \leq r^2\}$$

with center point $y = (-0.4, 0)$ and radius $r = 0.1$. The obstacle function is specified similar to the one from Example 5.1 by

$$1 - e^{-h(x)} = \min \{1, \max \{0, 1 + sr^2 - s((x_1 - y_1)^2 + (x_2 - y_2)^2)\}\}$$

with slope parameter $s = 20$.

The numerical scheme introduced at the beginning of this section is easily adapted to the minimization task by replacing “ $\max_{u \in U}$ ” in (5.1) and (5.3) by “ $\min_{u \in U}$ ”. The following computations were performed with the same algorithms as in Example 5.1 with time step $h = 0.01$, adaptively generated rectangular finite elements covering $\Omega = [-1, 1]^3$ with about 200000 vertices and a discrete set of control values discretizing the set $U = [-1, 1]^2$ with 9 equidistant values.

Figure 5.3 shows different level sets of the resulting function $v : \Omega \rightarrow \mathbb{R}$, on the left without obstacle and on the right with obstacle. Figure 5.4 shows the graph of V on the plane $\{(x_1, x_2, 0) \mid (x_1, x_2) \in [-1, 1]^2\}$, in which the shape of the obstacle function $1 - e^{-h(x)}$ is clearly visible.

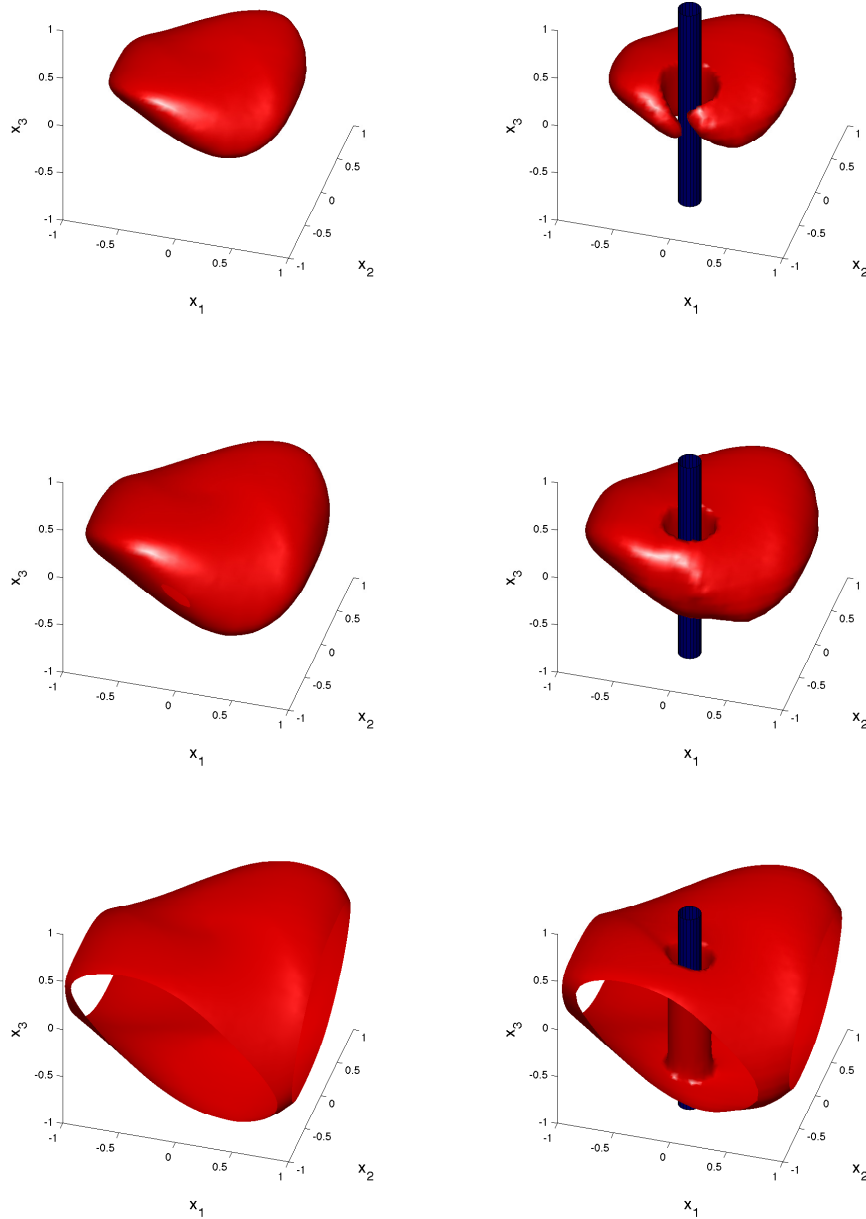


Figure 5.3: Level sets $\{x \in [-1, 1]^3 \mid v(x) = \ell\}$ without obstacle (left) and with obstacle (right) for $\ell = 0.18, 0.3, 0.5$ (top to bottom)

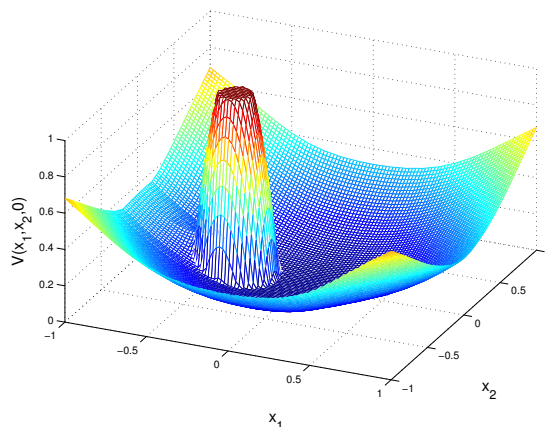


Figure 5.4: Graph of v over the plane $\{(x_1, x_2, 0) \mid (x_1, x_2) \in [-1, 1]^2\}$

□

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