Characterization of Domain of Fixed-time Stability under Control Input Constraints

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Abstract—In this paper, we study the effect of control input constraints on the domain of attraction of an FxTS equilibrium point. We first present a new result on FxTS, where we allow a positive term in the time derivative of the Lyapunov function. We provide analytical expressions for the domain of attraction and the settling time to the equilibrium in terms of the coefficients of the positive and negative terms that appear in the time derivative of the Lyapunov function. We show that this result serves as a robustness characterization of FxTS equilibria in the presence of an additive, vanishing disturbances. We use the new FxTS result in formulating a provably feasible quadratic program (QP) that computes control inputs that drive the trajectories of a class of nonlinear, control-affine systems to a goal set, in the presence of control input constraints.

I. INTRODUCTION

In control problems where the objective is to stabilize closed-loop trajectories to a desired set or a set of interest, control Lyapunov functions (CLFs) are commonly used to design the control input [1], [2]. Traditionally, CLFs have been used to design closed-form expressions for control inputs using Sontag’s formula [1], [3]. More recently, quadratic programs (QPs) have gained popularity for control synthesis; with this approach, the CLF conditions are formulated as inequalities that are linear in the control input [4], [5], and the control input is computed as a solution to these parametric QPs. These methods are suitable for real-time implementation as QPs can be solved very efficiently. In most of the prior work on QP-based control design, the feasibility of the underlying QP is not guaranteed particularly in the presence of input constraints.

The work in [2], [4], [5] considers the design of control laws so that reachability objectives, such as reaching the desired goal set, are achieved as time goes to infinity, i.e., asymptotically or exponentially. Much attention has been paid recently to the concepts of finite- and fixed-time stability, where the system trajectories reach an equilibrium point or a set in a finite or fixed time, as opposed to asymptotically or exponentially. Fixed-time stability (FxTS), introduced in [6], is a stronger notion than exponential stability, where the time of convergence is finite, and is uniformly bounded for all initial conditions. Research has also shown that there is also a correlation between a faster rate of convergence and better disturbance rejection properties for a dynamical system [6], [7]. A lot of work has been since done in the field of FxTS; the authors in [8], [9] discuss necessary and sufficient conditions for FxTS; [10], [11] present FxTS results from a sliding-mode perspective (see also [12]–[14] for some examples of applications of FxTS theory in control and estimation problems). Recently, the concept of fixed-time CLF (FxT-CLF) was introduced [15], which combines the notion of CLF and FxTS in a QP, but without any feasibility guarantees.

The aforementioned papers study global FxTS, which requires unbounded control authority. Since it is not possible to guarantee FxTS from arbitrary initial conditions in the presence of control input constraints, it is important to study the domain of attraction from which FxTS can be guaranteed in the presence of input bounds. To this end, in this paper, we present new Lyapunov conditions on FxTS by introducing a (possibly positive) linear term in the upper bound of the derivative of the Lyapunov function. We show that FxTS can still be guaranteed from a domain of attraction that depends upon the relative magnitude of the positive and the negative terms in the bound of the time derivative of the Lyapunov function. We compute an upper bound on the time of convergence to the equilibrium, which is also a function of the relative magnitude of the positive and negative terms. We discuss the relation between the proposed results on FxTS and the robustness of FxTS systems under additive vanishing disturbances. Besides, based on the results in [16], we use the new FxTS conditions in a QP formulation, where the control objective is to drive closed-loop trajectories to a goal set in a given fixed time, in the presence of control input constraints. The results of this paper extend and formalize the results in [15] in a QP framework, such that feasibility, as well as fixed-time convergence, can be simultaneously guaranteed from a domain of attraction that is a function of the input bounds and time of convergence. We perform numerical experiments to relate the domain of attraction with the required time of convergence and with the control input bounds.

II. MATHEMATICAL PRELIMINARIES

Notations: In the rest of the paper, $\mathbb{R}$ denotes the set of real numbers, and $\mathbb{R}_+$ denotes the set of non-negative real numbers. We use $\| \cdot \|$ to denote the Euclidean norm. We use $\partial S$ to denote the boundary of a closed set $S$ and $\text{int}(S) = S \setminus \partial S$, to denote its interior.

Next, we review the notion of fixed-time stability. Consider the nonlinear system

$$\dot{x}(t) = f(x(t)), \quad x(0) = x_0,$$  

(1)
where $x \in \mathbb{R}^n$ and $f : \mathbb{R}^n \to \mathbb{R}^n$ is continuous with $f(0) = 0$. Assume that the solution of (1) exists and is unique. The authors in [6] presented the following result for FxTS.

**Lemma 1 (6).** Suppose there exists a continuously differentiable, positive definite, radially unbounded function $V$ for the system (1) such that

$$
\dot{V}(x) \leq -aV(x)^p - bV(x)^q,
$$

(2)

for all $x \neq 0$, where $a, b > 0$, $0 < p < 1$ and $q > 1$. Then, the origin of (1) is FxTS, and the time of convergence $T$ is uniformly bounded as $T \leq \frac{a(1-p)}{b(q-1)}$.

III. MAIN RESULTS

In this section, we present a new result on FxTS. Particularly, we introduce another term in the upper bound of $V$ in (2), and allow this term to take positive values. Consider a positive definite, continuously differentiable function $V : \mathbb{R}^n \to \mathbb{R}$, such that its time derivative along the trajectories of (1) satisfies

$$
\dot{V}(x(t)) \leq -\alpha_1 V(x(t))^{\gamma_1} - \alpha_2 V(x(t))^{\gamma_2} + \delta_1 V(x(t)),
$$

(3)

for all $t \geq 0$, with $\alpha_1, \alpha_2 > 0$, $\delta_1 \in \mathbb{R}$, $\gamma_1 = 1 + \frac{1}{\mu}$, $\gamma_2 = 1 - \frac{1}{\mu}$ for some $\mu > 1$.

**New FxTS Lyapunov conditions:** Before presenting the main result, we need the following lemma.

**Lemma 2.** Let $V_0, \alpha_1, \alpha_2 > 0$, $\delta_1 \in \mathbb{R}$, $\gamma_1 = 1 + \frac{1}{\mu}$ and $\gamma_2 = 1 - \frac{1}{\mu}$, where $\mu > 1$. Define

$$
I := \int_0^T \frac{dV}{-\alpha_1 V^{\gamma_1} - \alpha_2 V^{\gamma_2} + \delta_1 V}.
$$

(4)

Then, the following holds:

(i) If $0 \leq \delta_1 < 2\sqrt{\alpha_1 \alpha_2}$, we have for all $V_0 \geq 0$

$$
I \leq \frac{\mu}{\alpha_1 k_1} \left( \frac{\pi}{2} - \tan^{-1} k_2 \right),
$$

(5)

where $k_1 = \sqrt{\frac{4\alpha_1 \alpha_2 - \delta_1^2}{4\alpha_1^2}}$ and $k_2 = -\frac{\delta_1}{\sqrt{4\alpha_1 \alpha_2 - \delta_1^2}}$.

(ii) If $\delta_1 \geq 2\sqrt{\alpha_1 \alpha_2}$ and $V_0 \leq k \frac{\delta_1 - \sqrt{\delta_1^2 - 4\alpha_1 \alpha_2}}{2\alpha_1}$ with $0 < k < 1$, we have for all $V_0 \geq 0$

$$
I \leq \frac{\mu k}{(1-k)\sqrt{\alpha_1 \alpha_2}}.
$$

(6)

Lemma 2 gives upper bounds on the integral $I$ for various cases (which will serve as the upper-bound on the fixed time of convergence as shown next). The proof is provided in Appendix I. Now we are ready to present our main result on new Lyapunov conditions for FxTS.

**Theorem 1.** Let $V : \mathbb{R}^n \to \mathbb{R}$ be a continuously differentiable, positive definite, proper function, satisfying

$$
\dot{V}(x) \leq -\alpha_1 V(x)^{\gamma_1} - \alpha_2 V(x)^{\gamma_2} + \delta_1 V(x),
$$

(7)

for all $x \in \mathbb{R}^n \setminus \{0\}$ along the trajectories of (1) with $\alpha_1, \alpha_2 > 0$, $\delta_1 \in \mathbb{R}$, $\gamma_1 = 1 + \frac{1}{\mu}$, $\gamma_2 = 1 - \frac{1}{\mu}$ for some $\mu > 1$. Then, there exists a neighborhood $D \subseteq \mathbb{R}^n$ of the origin such that for all $x(0) \in D$, the trajectories of (1) satisfy $x(t) \in D$ for all $t \geq 0$, and reach the origin within a fixed time $T$, where

$$
D = \left\{ \begin{array}{ll}
\mathbb{R}^n; & r > 1, \\
\{ x \mid V(x) \leq k^\mu \left( \frac{\delta_1 - \sqrt{\delta_1^2 - 4\alpha_1 \alpha_2}}{2\alpha_1} \right)^\mu \}; & r \geq 1,
\end{array} \right.
$$

(8)

$$
T \leq \frac{\mu}{\alpha_1 k_1 \left( \frac{\pi}{2} - \tan^{-1} k_2 \right)},
$$

(9)

where $r := \frac{\delta_1}{2\sqrt{\alpha_1 \alpha_2}}$, $0 < k < 1$, $k_1 = \sqrt{\frac{4\alpha_1 \alpha_2 - \delta_1^2}{4\alpha_1^2}}$ and $k_2 = -\frac{\delta_1}{\sqrt{4\alpha_1 \alpha_2 - \delta_1^2}}$.

Proof. Note that the domain of attraction $D$ and the time of convergence $T$ are functions of the ratio $r := \frac{\delta_1}{2\sqrt{\alpha_1 \alpha_2}}$. The three cases, namely, $r \leq 0$, $0 \leq r < 1$ and $r \geq 1$ are studied separately.

For $r \leq 0$, one can recover the right-hand side of (2) from (7), and it follows from Lemma 1 that $D = \mathbb{R}^n$, and from part (i) in Lemma 2 with $r \leq 0$, it follows that $T \leq 2\sqrt{\alpha_1 \alpha_2} / \alpha_1$. Next, consider the case when $0 \leq r < 1$. First it is shown that there exists $D \subseteq \mathbb{R}^n$ containing the origin such that $V(x) < 0$ for all $x \in D \setminus \{0\}$, so that any sub-level set of the function $V$ contained in $D$ is forward-invariant. The right-hand side of (7) can be re-arranged so that (7) reads

$$
\dot{V}(x) \leq V(x) (-\alpha_1 V(x)^{\gamma_1-1} - \alpha_2 V(x)^{\gamma_2-1} + \delta_1).
$$

Note that $V(x) > 0$ for all $x \neq 0$. Thus, for $\dot{V}(x)$ to take negative values for all $x \neq 0$, it is needed that

$$
\min_{x \neq 0} (-\alpha_1 V(x)^{\gamma_1-1} - \alpha_2 V(x)^{\gamma_2-1} + \delta_1) < 0
$$

$$
\iff \delta_1 < \min_{x \neq 0} (\alpha_1 V(x)^{\gamma_1-1} + \alpha_2 V(x)^{\gamma_2-1})
$$

$$
\iff \delta_1 < \min_{x \neq 0} (\alpha_1 V(x)^{\frac{\gamma_1}{2}} + \alpha_2 V(x)^{-\frac{\gamma_2}{2}}).
$$

Substitute $s = V(x)^{\frac{1}{2}}$ to denote $\alpha_1 V(x)^{\frac{\gamma_1}{2}} + \alpha_2 V(x)^{-\frac{\gamma_2}{2}} = \alpha_1 s + \alpha_2 s^{-1}$. Then, the function $p : \mathbb{R}_+ \to \mathbb{R}$ defined as

$$
p(s) := \alpha_1 s + \alpha_2 s^{-1}
$$

is a strictly convex function since $\frac{d^2 p}{ds^2} = 2\alpha_2 s^{-3} > 0$ for all $k > 0$ and has a unique minimizer. The derivative of $p$ reads $\frac{dp}{ds} = \alpha_1 - \frac{\alpha_2}{s^2}$, which has a unique root in $\mathbb{R}_+$ at $s = \sqrt{\alpha_2 / \alpha_1}$. Thus the minimum is attained for $s = \left( \frac{\alpha_2}{\alpha_1} \right)^{\frac{1}{2}}$. Define $V^* := \left( \frac{\alpha_2}{\alpha_1} \right)^{\frac{1}{2}}$ and $\delta^* := \alpha_1 V^* \frac{1}{2} + \alpha_2 V^* \frac{1}{-2}$ = $2\sqrt{\alpha_1 \alpha_2}$ so that $\alpha_1 V(x)^{\frac{\gamma_1}{2}} + \alpha_2 V(x)^{-\frac{\gamma_2}{2}} \geq \delta^*$ for all $x \in \mathbb{R}^n$. Thus, for $r < 1$, it holds that $\delta_1 < \delta^* \leq \alpha_1 V(x)^{\frac{1}{2}} + \alpha_2 V(x)^{-\frac{1}{2}}$ for all $x$, and so, $\dot{V}(x) < 0$ for all $x \in \mathbb{R}^n \setminus \{0\}$. Since $D$ is defined as the largest sub-level set of $V$ such that $\dot{V}(x)$ takes negative values for $x \in D \setminus \{0\}$, it holds that in the case when $0 \leq r < 1$, $D = \mathbb{R}^n$.

1 Only the non-negative root is of interest, since $s = V(x)^{\frac{1}{2}} \geq 0$. 

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Finally, for the case when \( r \geq 1 \), it holds that
\[
\delta_1 = 2(\alpha_1 \alpha_2)^{\frac{3}{2}}
\]
and
\[
\delta_1 < 2(\alpha_1 \alpha_2)^{\frac{3}{2}}
\]
(see Figure 1). It can be easily verified that if \( \delta_1 < 2\sqrt{\alpha_1 \alpha_2} \), then the domain of attraction is the entire \( \mathbb{R}^n \), and for a given \( \alpha_1, \alpha_2 \), as \( \delta_1 \) increases, the domain of attraction shrinks.

Robustness perspective: In comparison to Lemma 1, Theorem 1 allows a positive term \( \delta_1V \) in the upper bound of the time derivative of the Lyapunov function. This term also captures the robustness against a class of Lipschitz continuous, or vanishing, additive disturbances in the system dynamics, as shown in the following result. Consider the system
\[
\dot{x} = f(x) + \psi(x),
\]
where \( f, \psi : \mathbb{R}^n \to \mathbb{R}^n \), \( f(0) = 0 \) and there exists \( L > 0 \) such that for all \( x \in \mathbb{R}^n \), \( \|\psi(x)\| \leq L\|x\| \).

Corollary 1. Let the origin for the nominal system \( \dot{x} = f(x) \) be FxTS and assume that there exists a Lyapunov function \( V : \mathbb{R}^n \to \mathbb{R} \) satisfying the conditions of Lemma 1. Assume that there exist \( k_1, k_2 > 0 \) such that \( V(x) \geq k_1\|x\|^2 \) and \( \|\delta \| \leq k_2\|x\| \) for all \( x \in \mathbb{R}^n \). Then, the origin of the perturbed system (12) is FxTS.

Proof. The time derivative of \( V \) along the system trajectories of (12) reads
\[
\dot{V} = \frac{\partial V}{\partial x}f(x) + \frac{\partial V}{\partial x}\psi(x) \leq -aV^p - bV^q + k_2L\|x\|^2
\]
\[
\leq -aV^p - bV^q + \frac{k_2L}{k_1}V.
\]
Hence, using Theorem 1, we obtain the origin of (12) is FxTS for all \( x(0) \in D \), where \( D \) is a neighborhood of the origin. As per the conditions of Theorem 1, \( D = \mathbb{R}^n \) or \( D \subset \mathbb{R}^n \), depending upon whether the ratio \( \frac{k_2L}{2k_1\sqrt{ab}} \) is less or greater than 1, respectively.

IV. CONTROL SYNTHESIS UNDER INPUT CONSTRAINTS

In this section, we use the Lyapunov condition (7) in conjunction with Theorem 1 in a QP formulation to compute a control input so that the closed-loop trajectories reach a desired goal set within a fixed time. Consider the system:
\[
\dot{x}(t) = f(x(t)) + g(x(t))u(x(t)),
\]
where $x \in \mathbb{R}^n$ is the state vector, $f : \mathbb{R}^n \to \mathbb{R}^n$ and $g : \mathbb{R}^n \to \mathbb{R}^{n \times m}$ are continuous functions, and $u \in U \subset \mathbb{R}^m$ is the control input where $U$ denotes the control input constraint set. In addition, consider a goal set to be reached in a user-defined fixed time $T$, defined as $S_G := \{ x \mid h_G(x) \leq 0 \}$, where $h_G : \mathbb{R}^n \to \mathbb{R}$ is a continuously differentiable function. Consider the QP

$$
\begin{align}
\min_{v \in \mathbb{R}^m, \delta \in \mathbb{R}} & \frac{1}{2} z^T H z + F^T z \\
\text{s.t.} & \quad A_n v \leq b_u, \\
& \quad L_y h_G(x) + L_y h_G(x) v \leq \delta_1 h_G(x) - \alpha_1 \max(0, h_G(x))^{\gamma_1} - \alpha_2 \max(0, h_G(x))^{\gamma_2},
\end{align}
$$

where $z := [v^T \delta_1]^T \in \mathbb{R}^{m+1}$, $H := \text{diag}(p_{u_1}, \ldots, p_{u_m}, p_1)$ is a diagonal matrix consisting of positive weights $p_{u_1}, p_1 > 0$, $F := [0_m^T, q_1]$ with $q_1 > 0$ and $0_m \in \mathbb{R}^m$ a vector consisting of zeros. The parameters $\alpha_1, \alpha_2, \gamma_1, \gamma_2$ are fixed, and are chosen as $\alpha_1 = \alpha_2 = \frac{H_f}{2\mu}$, $\gamma_1 = 1 + \frac{1}{\mu}$ and $\gamma_2 = 1 - \frac{1}{\mu}$ with $\mu > 1$. Constraint (14b) encodes the control input constraints $u \in U = \{ v \mid A_n v \leq b_u \}$, while (14c) encodes the FxT-CLF condition. Below, we show that the QP (14) is feasible, and under certain conditions, the control input defined as the solution of (14) lead to FxTS convergence of the closed-loop trajectories. Let the solution of (14) be denoted as $z^*(\cdot) = [v^*(\cdot)^T \delta_1^*(\cdot)]^T$.

**Lemma 3.** If the set $U$ is non-empty, then the QP (14) is feasible for all $x \notin S_G$.

**Proof.** Choose any $\bar{v} \in U$, and since $U$ is non-empty, there exists such $\bar{v}$. For $x \notin S_G$, we have that $h_G(x) > 0$ by definition, and thus $h_G(x) \neq 0$. Define

$$
\bar{\delta}_1 = L_y h_G(x) + L_y h_G(x) \bar{v} + \alpha_1 h_G(x)^{\gamma_1} + \alpha_2 h_G(x)^{\gamma_2},
$$

so that (14c) is satisfied. Thus, the couple $(\bar{v}, \bar{\delta}_1)$ satisfies the constraints of the QP (14) and hence, the QP (14) is feasible, for all $x \notin S_G$.

The feasibility of (14) is guaranteed because of the presence of the slack term $\delta_1 V$. Note that in the absence of such a term, (14) might be infeasible due to the presence of the control input constraints. This is one of the main advantages of using the condition (7) as compared to (2). The following result is adapted from [16] that shows FxTS of the set $S_G$ under the control input $u(x) = v^*(x)$.

**Theorem 2 (16),** The closed-loop trajectories under the effect of the control input defined as $u(x) = v^*(x)$ reach the set $S_G$ within a fixed time $\bar{T}$ for all $x(0) \in D$, where:

(i) $D = \mathbb{R}^n$ and $\bar{T} = T$ if $\max \delta_1(x) \leq 0$;

(ii) $D = \mathbb{R}^n$ and $\bar{T} \leq \frac{\sup x \mu}{\alpha_1 k_1(x)} \left( \frac{T}{2} - \tan^{-1} k_2(x) \right)$, where

$$
k_1(x) = \sqrt{\frac{4\alpha_1 \alpha_2 - \delta_1(x)^2}{4\alpha_1}} \quad \text{and} \quad k_2(x) = \frac{-\delta_1(x)}{\sqrt{4\alpha_1 \alpha_2 - \delta_1(x)^2}} \quad \text{if} \quad \max \delta_1(x) < 2\sqrt{\alpha_1 \alpha_2};
$$

(iii) $D = \{ z \mid V(z) \leq \inf_x \left( \frac{\delta_1(x) - \sqrt{\delta_1(x)^2 - 4\alpha_1 \alpha_2}}{2\alpha_1} \right)^\mu \}$ and

$$
\bar{T} \leq \frac{\mu}{k_1 \sqrt{\delta_1 \alpha_2}} \quad \text{if} \quad \max \delta_1(x) > 2\sqrt{\alpha_1 \alpha_2}.
$$

From the expression of domain of attraction $D$ in (8), it can be observed that the domain $D$ shrinks as the ratio $r$ increases. In particular, for the case when $\alpha_1 = \alpha_2 = \alpha$ and $r \geq 1$, the domain of attraction is given as $D = \{ x \mid V(z) \leq (r - \sqrt{r^2 - 1})^\mu \}$ where $r = \frac{\delta_1}{\alpha_2}$. Under the same conditions, the domain of attraction for the closed-loop system (13) under $u = v^*$ is given as $D = \{ x \mid V(z) \leq (r_M - \sqrt{r_M^2 - 1})^\mu \}$, which is a function of $r_M := \sup \frac{\delta_1(x)}{\alpha_2}$.

Figure 2 plots the boundary of the set $D$ for various values of $r_M$. Thus, if $\delta_1$ is small relative to $\alpha_1, \alpha_2$, then the domain of attraction is large for fixed-time convergence, i.e., the slack term corresponding to $\delta_1$ in QP (14) characterizes the trade-off between the domain of attraction and time of convergence for given control input bounds. Intuitively, for a given control input constraint set, a larger value of $T$ results into smaller values of $\alpha_1, \alpha_2$, which would result in satisfaction of (14c) with smaller value of $\delta_1$. Conversely, for a given $T$ (and thus, for a given pair $\alpha_1, \alpha_2$), a larger control authority would result into satisfaction of (14c) with smaller $\delta_1$. This relation between the domain of attraction, the input constraints, and the time of convergence is studied in more detail in [16], where KKT conditions are used to compute the closed-form expression for the solution of the QP (14), and it is shown that increasing the control bounds or the require time of convergence results in smaller values of $r_M$. In this paper, we verify this relation via numerical simulations.

**Numerical experiments:** We consider the following system:

$$
\begin{align*}
\dot{x}_1 &= x_2 + x_1 (x_1^2 + x_2^2 - 1) + x_1 u, \\
\dot{x}_2 &= -x_1 + \zeta(x_2) (x_1^2 + x_2^2 - 1) + x_2 u,
\end{align*}
$$

where $x = [x_1, x_2]^T \in \mathbb{R}^2, u \in \mathbb{R}, \zeta(z) = (0.8 + 0.2e^{-100|z|}) \tanh(z)$ and $S_G = \{ x \mid \|x\| \leq 1 \}$. Note that in the absence of the control input, the trajectories diverge.
away from $S_G$, i.e., the set $S_G$ is unstable for the open-loop system. We define $h_G(x) = \|x\|^2 - 1$. We impose control input bounds of the form $\|u\| \leq u_{\text{max}}$, where $u_{\text{max}} > 0$. The initial conditions are chosen as $x(0) = [3.33, 1.33]^T$.

We choose $p_{u_1}, p_{u_2} = 1, \mu = 2$ for the numerical simulations. First, we studied the effect of the control input bound on the maximum value of $\delta_1$. We fixed $T = 1, p_1 = 100, q_1 = 1000$, and varied $u_{\text{max}}$. Figure 3 plots the maximum value of $\max_x \delta_1(x)$ for various values of $u_{\text{max}} \in [16, 25]$.

It can be observed that $\delta_1$ decreases as the control authority of the system increases. This verifies the intuition that the domain of attraction expands as the control authority increases.

Figure 4 plots the norm of the control input with time for various values of $u_{\text{max}}$. The value of $u_{\text{max}}$ increases from 16 to 25 from blue to red. It can be observed that in every case, the system trajectories do utilize the maximum available control authority at the beginning of the simulation, while the control input decreases to zero as the system trajectories approach the goal set.

Next, we fix $u_{\text{max}} = 16, p_1 = 100, q_1 = 1000$ and vary the required time of convergence $T$ between 1 and 10. Figure 5 shows the variation of $\max_x \delta_1(x)$ as a function of the convergence time $T$. As $T$ increases (or equivalently, $\alpha_1, \alpha_2$ decrease), the maximum value of $\delta_1(x)$ decreases. This implies that for a larger time of convergence, there is a larger domain of attraction starting from which convergence can be achieved in the given time.

These (numerical) relations indicate that for a required domain of attraction $D$, one can choose the parameters $u_{\text{max}}$ and $T$ so that the presented QP in (14) guarantees FxTS for any initial condition in $D$. Feasibility of the QP (14) guarantees that for this choice of parameters, a control input exists and renders the goal set FxTS within the chosen time $T$. Conversely, for a given input bound and required time of convergence, it is possible to find the largest domain of attraction by computing the maximum value of $\delta_1$.

**V. CONCLUSION**

We proposed a new result on FxTS by allowing a positive linear term to appear in the time derivative of the Lyapunov function. We characterized the domain of attraction, as well as the upper bound on the time of convergence for fixed-time stability as a function of the coefficients of the positive and the negative terms in the upper bound of the time derivative of the Lyapunov function. We then used the new FxTS result in a QP formulation and showed that the feasibility of the QP is guaranteed due to the presence of the slack term that corresponds to the newly added linear term in our FxTS result. For the QP-based control design technique, we numerically established a relation of the maximum value of this slack term, which characterizes the domain of attraction for fixed-time convergence, with the control input bound, and with the required time of convergence. It is thus shown that with an appropriate choice of the required time of convergence and control input bounds, the presented result can guarantee FxTS from the desired domain of attraction.

In the future, we would like to study multi-objective problems involving both safety and convergence requirements and find the relations between the largest domain of attraction for fixed-time convergence and the largest subset of the safe set that can be rendered forward invariant, parametrized by the control input bounds and the time of convergence.

### REFERENCES


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Next, we consider the case when $\delta_1 > 2\sqrt{\alpha_1 \alpha_2}$. In this case, the roots of $\gamma(m) = 0$ are real. Let $a \leq b$ be such that $\alpha_1 m^2 - \delta_1 m + \alpha_2 = \alpha_1(m-a)(m-b)$. This substitution allows us to factorize the denominator to evaluate the integral $I$. Note that since $ab = \alpha_2 > 0$ and $a + b = \delta_1$, we have $0 < a \leq b$. Since $V_0^\frac{1}{2} \leq k \frac{1}{\sqrt{2\alpha_1 \alpha_2}} = ka$ where $k < 1$, we have that $-\alpha_1 V_0^\frac{1}{2} - \alpha_2 V_0^{\frac{3}{2}} + \delta_1 V_0 \leq 0$ for all $V_0 \leq V_0$, i.e., the denominator $\delta_1 V_0 - \alpha_1 V_0^{\frac{3}{2}} + \alpha_2 V_0^{\frac{3}{2}}$ does not vanish for $V_0 \in [0, V_0]$. Thus, we obtain that

$$I = \mu \int_{V_0^\frac{1}{2}}^{0} \frac{dm}{(-\alpha_1 m^2 - \alpha_2 + \delta_1 m)} = -\frac{\mu}{\alpha_1} \int_{V_0^\frac{1}{2}}^{0} \frac{dm}{(m-a)(m-b)} = -\frac{\mu}{\alpha_1(a-b)} \left( \int_{V_0^\frac{1}{2}}^{0} \frac{dm}{m-a} - \int_{V_0^\frac{1}{2}}^{0} \frac{dm}{m-b} \right).$$

Evaluating the integrals, we obtain

$$I = \frac{-\mu}{\alpha_1(a-b)} \left( \log \left( \frac{a}{V_0^\frac{1}{2} - a} \right) - \log \left( \frac{b}{V_0^\frac{1}{2} - b} \right) \right) = \frac{\mu}{\alpha_1(a-b)} \left( \log \left( \frac{b}{a} \right) + \log \left( \frac{V_0^\frac{1}{2} - a}{V_0^\frac{1}{2} - b} \right) \right) \leq \frac{\mu}{\alpha_1(b-a)} \left( \log \left( \frac{b - ka}{a(1-k)} \right) - \log \left( \frac{b}{a} \right) \right).$$

It can be easily shown that the above upper-bound on $I$ decreases monotonically as the ratio $\frac{b}{\alpha_1(a-b)}$ increases from $1$ to $\infty$. Thus, the maximum value of this upper-bound is achieved in the limit when $\frac{b}{\alpha_1(a-b)} = 1$. Note also that $\lim_{\delta_1 \to \infty} \frac{\delta_1}{\sqrt{\alpha_1 \alpha_2}} \to 1$ implies that $b \to a$ and in the limit, we have $a = b = \frac{\delta_1}{\sqrt{\alpha_1 \alpha_2}}$. Thus, the maximum value of the upper-bound can be computed by taking limit $b \to a$ as follows:

$$I \leq \frac{\mu}{\alpha_1(b-a)} \left( \log \left( \frac{b - ka}{a(1-k)} \right) - \log \left( \frac{b}{a} \right) \right) \leq \frac{\mu}{\alpha_1(b-a)} \left( \log \left( \frac{b - ka}{a(1-k)} \right) - \log \left( \frac{b}{a} \right) \right) = \frac{\mu}{\alpha_1(b-a)} \left( \log \left( \frac{b - ka}{a(1-k)} \right) - \log \left( \frac{b}{a} \right) \right) = \frac{\mu k}{\alpha_1(b-a)} \left( \frac{b - ka}{a(1-k)} \right) - \log \left( \frac{b}{a} \right).$$

which completes the proof. 

\[ \square \]