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Stabilization of two-dimensional persistently excited linear control systems with arbitrary rate of convergence

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July 21, 2011

Abstract

We study the control system $\dot{x} = Ax + \alpha(t)bu$ where the pair (A, b) is controllable, $x \in \mathbb{R}^2$, $u \in \mathbb{R}$ is a scalar control and the unknown signal $\alpha : \mathbb{R}_+ \rightarrow [0, 1]$ is (T, μ) -persistently exciting (PE), i.e., there exists $T \geq \mu > 0$ such that, for all $t \in \mathbb{R}_+$, $\int_t^{t+T} \alpha(s)ds \geq \mu$. We are interested in the stabilization problem of this system by a linear state feedback $u = -Kx$. In this paper, we positively answer a question asked in [5] and prove the following: Assume that the class of (T, μ) -PE signals is restricted to those which are M -Lipschitzian, where $M > 0$ is a positive constant. Then, given any $C > 0$, there exists a linear state feedback $u = -Kx$ where K only depends on (A, b) and T, μ, M so that, for every M -Lipschitzian (T, μ) -PE signal, the rate of exponential decay of the time-varying system $\dot{x} = (A - \alpha(t)bK)x$ is greater than C .

1 Introduction

The aim of this paper is to continue the study of persistently excited (PE) linear control systems of [4, 5]. These systems can be written in the form

$$\dot{x} = Ax + \alpha(t)Bu, \quad (1.1)$$

where $x \in \mathbb{R}^d$, $u \in \mathbb{R}^m$ is the control and $\alpha : \mathbb{R}_+ \rightarrow [0, 1]$ is a scalar measurable signal.

In the control system (1.1), the signal α determines when the control signal u is activated. We suppose that α is not precisely known, and the only information on α we have is that it belongs to a certain class of functions \mathcal{G} .

We consider the problem of stabilization to the origin, by means of a linear state feedback $u = -Kx$, of (1.1), where the uncontrolled dynamics $\dot{x} = Ax$ can be unstable. We want this stabilization to be *uniform* with respect to the signal α i.e., K may depend on T and μ , but it should not depend on a particular signal $\alpha \in \mathcal{G}$.

The question that arises is which kind of class of functions \mathcal{G} may be useful in practice. It is obviously not suitable to choose $\mathcal{G} = L^\infty(\mathbb{R}_+, [0, 1])$ since this class contains the constant function equal to 0 and functions which are arbitrarily small in norm L^∞ or that are different from 0 only in a finite time interval, and, in these cases, the control has a small or even null influence on the behavior of the non-controlled system $\dot{x} = Ax$. We are thus looking for a class of functions that are

“active” often enough and that, at each activation, have a certain “minimum degree of activation”. For this purpose, [4, 5] consider the case where α is a persistently exciting signal (PE signal for short), that is, there exist two positive constants $T \geq \mu$ such that, for every $t \in \mathbb{R}_+$,

$$\int_t^{t+T} \alpha(s) ds \geq \mu. \quad (1.2)$$

We will refer to such PE signals as the class of (T, μ) -PE signals also denoted by $\mathcal{G}(T, \mu)$. Condition (1.2) is known as the condition of persistent excitation and appears naturally in the context of identification and adaptive control, as in [10].

Throughout this paper, we will consider only the case where the control u is scalar, and thus the matrix B is actually a column vector $b \in \mathbb{R}^d$. System (1.1) becomes then

$$\dot{x} = Ax + \alpha(t)bu \quad (1.3)$$

where $x \in \mathbb{R}^d$, $u \in \mathbb{R}$ and $\alpha \in \mathcal{G}(T, \mu)$.

One may think that, intuitively, the stabilization of a system as (1.3) is possible under reasonable hypotheses on (A, b) , due to the condition (1.2) imposed on α . Indeed, if we suppose that the linear control system defined by (A, b) is stabilizable, meaning that there exists $K^T \in \mathbb{R}^d$ such that $A - bK$ is Hurwitz, and if we suppose, in order to simplify the reasoning, that $\alpha(t) \in \{0, 1\}$, then, by choosing such a K , we can “stabilize” System (1.3) at least when $\alpha(t) = 1$, which, due to Condition (1.2) on α , happens at least for a total time μ on every time window of length T . However, we still have to deal with time intervals when $\alpha(t) = 0$. We may then think that, if we know the behavior of the uncontrolled system $\dot{x} = Ax$, we can try to compensate any destabilization due to its dynamics when $\alpha(t) = 0$ by astutely choosing K so that the stabilization provided by $A - bK$ when $\alpha(t) = 1$ eventually stabilizes the system. This intuition can also be used in the case $\alpha(t) \in [0, 1]$: one can prove (see [6]) that it is possible, for every $\rho > 0$, to choose a K that stabilizes System (1.3) for every $\alpha \in L^\infty(\mathbb{R}_+, [\rho, 1])$ uniformly with respect to α . Their argument can also be adapted, using a high-gain technique, to show that the rate of convergence can be made arbitrarily large. Then, for System (1.3) with an arbitrary $\alpha \in \mathcal{G}(T, \mu)$, by choosing ρ small enough, Equation (1.2) shows that we have $\alpha(t) \geq \rho$ at least for a total time that is uniformly lower bounded on every time window of length T ; we have then “good” time intervals where α is lower bounded by ρ and we know how to stabilize the system, and “bad” time intervals where α is smaller than ρ . We then expect that, by studying the behavior of the system in the “bad” time intervals, we can choose K in order to compensate an eventual explosive behavior of the solution in the “bad” time intervals by convergence in the “good” ones in order to stabilize the system.

This intuition was partially validated in [5], where it is shown that stabilization to the origin of System (1.3) is possible if (A, b) is a controllable pair and every eigenvalue of A has non-positive real part (see the precise statement in Theorem 2.6 below).

Let us mention that the difficult point in the argument above comes from the fact that we want to choose a feedback $u = -Kx$ with K that is independent of $\alpha \in \mathcal{G}(T, \mu)$, and thus it is the same K that acts on both “good” and “bad” intervals. The feedback gain K is chosen in such a way that the behavior on “good” intervals is known and controlled, but we know nothing a priori on the effects of a particular choice of K on “bad” intervals, and obtaining estimates in this case is nontrivial, since the usual estimates, such as Gronwall’s Lemma, are usually not fine enough to provide a useful result.

We can now ask the question of stabilization at an arbitrary rate: we want to choose a feedback $u = -Kx$ such that every solution of $\dot{x} = (A - \alpha(t)bK)x$ converges to 0 exponentially at a rate which is faster than a prescribed constant $C > 0$. In the case of a general autonomous linear control system $\dot{x} = Ax + Bu$, the question is then to study the eigenvalues of $A - BK$, and the answer is given by the Pole Shifting Theorem.

In the case $\alpha(t) \in \{0, 1\}$, one might think that stabilization at an arbitrary rate still holds true for a controllable pair (A, b) : it would suffice to choose K such that the solution of $\dot{x} = (A - bK)x$ stabilizes the system fast enough and that its convergence rate compensates the possible unstable behavior when $\alpha(t) = 0$. However, this intuition is shown to be false. In dimension $d = 2$, it was proved in [5] that there exists ρ_* so that, if $0 < \frac{\mu}{T} < \rho_*$, then the maximal rate of convergence of System (1.3) is finite, i.e., there exists a constant C such that, for every $K^T \in \mathbb{R}^2$, there exists a signal $\alpha \in \mathcal{G}(T, \mu)$ and an initial condition x_0 of norm 1 such that the solution $x(t)$ of the system $\dot{x} = (A - \alpha bK)x$ starting at x_0 satisfies

$$\limsup_{t \rightarrow +\infty} \frac{\ln \|x(t)\|}{t} \geq -C. \quad (1.4)$$

The contradiction to the intuitive idea lies in the overshoot phenomenon. The latter goes as follows. One can choose K such that the solution of $\dot{x} = (A - bK)x$ stabilizes fast enough, but the norm of the solutions of $\dot{x} = (A - bK)x$ may increase in a small time interval $[0, t]$ before exponentially decreasing with the desired convergence rate. Then, if $\alpha = 1$ on a short period of time only, it is actually the overshoot phenomenon, and not stabilization, that dominates the behavior of the solution of (1.3). By switching fast enough between $\alpha = 1$ and $\alpha = 0$ on a fixed window of time, we can repeat several times the overshoot phenomenon and still satisfy Condition (1.2). For $\frac{\mu}{T}$ small enough, one can then construct for any given K a signal $\alpha \in \mathcal{G}(T, \mu)$ such that the overshoot phenomenon dominates the stabilization provided by K . Notice that the regularity of $\alpha \in \mathcal{G}(T, \mu)$ is not an issue here since one can replace faster and faster switchings (between $\alpha = 1$ and $\alpha = 0$) by faster and faster oscillations as the norm of K increases in the above construction.

The purpose of this paper consists in addressing an open problem about the stabilization of (1.3) stated in [5] (Open Problem 5). Namely, [5] proposes to restrict the class $\mathcal{G}(T, \mu)$ of PE signals in order to recover stabilization by a linear state feedback at an arbitrary rate of convergence for System (1.3). To diminish the influence of the overshoot phenomenon, one needs to bound in some sense the possible arbitrary switches for α between the values 0 and 1. For that purpose the stabilization at an arbitrary rate of convergence for System (1.3) is conjectured for the subclass $\mathcal{D}(T, \mu, M)$ of $\mathcal{G}(T, \mu)$ of PE signals that are M -Lipschitz. The latter extra-hypothesis on PE signals clearly restrict the influence of the overshoot phenomenon. Indeed, if $\alpha(t_0) = 1$ for some t_0 , then $\alpha(t) \geq \frac{1}{2}$ for $t \in [t_0 - \frac{1}{2M}, t_0 + \frac{1}{2M}]$. Thus α cannot switch between the values 0 and 1 too fast. Therefore, we can hope to choose K in such a way that the stabilization provided by K now dominates the destabilizing effect of the overshoot phenomenon. The goal of this paper is to bring a positive answer to Open Problem 5 in the case of planar control systems (1.3), i.e., for $d = 2$.

Before presenting the plan of the paper, let us briefly describe the proof of the arbitrary rate of stabilization. We first decompose the time range into two classes of time intervals, \mathcal{J}_+ , the ‘‘good’’ intervals, where an auxiliary signal γ (obtained from α) is larger than a certain positive number, and \mathcal{J}_- , the ‘‘bad’’ intervals, where γ is small, retrieving thus the idea of ‘‘good’’ and ‘‘bad’’ intervals mentioned above. Estimations on ‘‘good’’ intervals are done by integrating the dynamics of the

control system written in polar coordinates: if we take the feedback gain K large enough, we can show that the solution rotates around the origin in “good” intervals, and the growth of the norm is estimated using the polar angle as new time. A different strategy is needed in the “bad” intervals: we resort to optimal control in order to find the “worst trajectory”, a particular solution of the system yielding the largest possible growth rate on a “bad” interval. The final part of the proof consists now in merging the two type of estimates.

The plan of the paper is the following. In Section 2, we provide the notations and definitions used throughout the paper as well as previous results on linear persistently excited control systems. We then turn in Section 3 to the core of the paper, where a precise statement of the main result is provided together with its proof.

2 Notations, definitions and previous results

2.1 Notations and definitions

In this paper, $\mathcal{M}_{d,m}(\mathbb{R})$ denotes the set of $d \times m$ matrices with real coefficients. When $m = d$, this set is denoted simply by $\mathcal{M}_d(\mathbb{R})$. As usual, we identify column matrices of $\mathcal{M}_{d,1}(\mathbb{R})$ with vectors of \mathbb{R}^d . The euclidean norm of an element $x \in \mathbb{R}^d$ is denoted by $\|x\|$, and the associate matrix norm of a matrix $A \in \mathcal{M}_d(\mathbb{R})$ is also denoted by $\|A\|$, whereas the symbol $|a|$ is reserved for the absolute value of a real or complex number a . The real and imaginary parts of a complex number z are denoted respectively by $\Re(z)$ and $\Im(z)$.

We shall consider control systems of the form

$$\dot{x} = Ax + \alpha(t)Bu \quad (2.1)$$

where $x \in \mathbb{R}^d$, $A \in \mathcal{M}_d(\mathbb{R})$, $u \in \mathbb{R}^m$ is the control, $B \in \mathcal{M}_{d,m}(\mathbb{R})$ and α belongs to the class of persistently exciting signals, defined below.

Definition 2.1 (PE signal and (T, μ) -signal). Let T, μ be two positive constants with $T \geq \mu$. We say that a measurable function $\alpha : \mathbb{R}_+ \rightarrow [0, 1]$ is a (T, μ) -signal if, for every $t \in \mathbb{R}_+$, one has

$$\int_t^{t+T} \alpha(s)ds \geq \mu. \quad (2.2)$$

The set of (T, μ) -signals is denoted by $\mathcal{G}(T, \mu)$. We say that a measurable function $\alpha : \mathbb{R}_+ \rightarrow [0, 1]$ is a *persistently exciting signal* (or simply *PE signal*) if it is a (T, μ) -signal for certain positive constants T and μ with $T \geq \mu$.

We shall use later a restriction of this class, namely that of Lipschitz (T, μ) -signals, which we define below.

Definition 2.2 ((T, μ, M) -signal). Let T, μ and M be positive constants with $T \geq \mu$. We say that a measurable function $\alpha : \mathbb{R}_+ \rightarrow [0, 1]$ is a (T, μ, M) -signal if it is a (T, μ) -signal and, in addition, α is globally M -Lipschitz, that is, for every $t, s \in \mathbb{R}_+$,

$$|\alpha(t) - \alpha(s)| \leq M|t - s|.$$

The set of (T, μ, M) -signals is denoted by $\mathcal{D}(T, \mu, M)$.

We can now define the object of our study.

Definition 2.3 (PE and PEL systems). Given a pair $(A, B) \in \mathcal{M}_d(\mathbb{R}) \times \mathcal{M}_{d,m}(\mathbb{R})$ and two positive constants T and μ (resp. three positive constants T , μ and M) with $T \geq \mu$, we say that the family of linear control systems

$$\dot{x} = Ax + \alpha Bu, \quad \alpha \in \mathcal{G}(T, \mu) \quad (\text{resp. } \alpha \in \mathcal{D}(T, \mu, M)) \quad (2.3)$$

is the *PE system* associated with A, B, T and μ (resp. the *PEL system* associated with A, B, T, μ and M).

The main problem we are interested in is the question of *uniform* stabilization of System (2.3) by a linear state feedback of the form $u = -Kx$ with $K \in \mathcal{M}_{m,d}(\mathbb{R})$, which makes System (2.3) take the form

$$\dot{x} = (A - \alpha(t)BK)x. \quad (2.4)$$

The problem is thus the choice of K such that the origin of the linear system (2.4) is globally asymptotically stable. With this in mind, we can introduce the following notion of stabilizer.

Definition 2.4 (Stabilizer). Let T and μ (resp. T, μ and M) be positive constants with $T \geq \mu$. We say that $K \in \mathcal{M}_{m,d}(\mathbb{R})$ is a (T, μ) -*stabilizer* (resp. (T, μ, M) -*stabilizer*) for System (2.3) if, for every $\alpha \in \mathcal{G}(T, \mu)$ (resp. $\alpha \in \mathcal{D}(T, \mu, M)$), System (2.4) is globally asymptotically stable.

We remark that K may depend on T, μ and M , but it cannot depend on the particular signal $\alpha \in \mathcal{G}(T, \mu)$ or $\alpha \in \mathcal{D}(T, \mu, M)$. We also remark that a (T, μ) -stabilizer is also a (T, μ, M) -stabilizer for every $M > 0$.

The question we are interested in is not only to stabilize a PE or PEL system, but also to stabilize it with an arbitrary rate of convergence. In order to rigorously define this idea, we introduce some concepts.

Definition 2.5. Let $(A, B) \in \mathcal{M}_d(\mathbb{R}) \times \mathcal{M}_{d,m}(\mathbb{R})$, $K \in \mathcal{M}_{m,d}(\mathbb{R})$ and $T \geq \mu > 0, M > 0$, and consider System (2.4). Fix $\alpha \in \mathcal{G}(T, \mu)$ (resp. $\alpha \in \mathcal{D}(T, \mu, M)$). We denote by $x(t; x_0)$ the solution of System (2.4) with initial condition $x(0; x_0) = x_0$.

- The *maximal Lyapunov exponent* $\lambda^+(\alpha, K)$ associated with (2.4) is defined by

$$\lambda^+(\alpha, K) = \sup_{\|x_0\|=1} \limsup_{t \rightarrow +\infty} \frac{\ln \|x(t; x_0)\|}{t}.$$

- The *rate of convergence* associated with the systems $\dot{x} = (A - \alpha(t)BK)x$, $\alpha \in \mathcal{G}(T, \mu)$ (resp. $\alpha \in \mathcal{D}(T, \mu, M)$) is defined as

$$\text{rc}_{\mathcal{G}}(T, \mu, K) = - \sup_{\alpha \in \mathcal{G}(T, \mu)} \lambda^+(\alpha, K) \quad (\text{resp. } \text{rc}_{\mathcal{D}}(T, \mu, M, K) = - \sup_{\alpha \in \mathcal{D}(T, \mu, M)} \lambda^+(\alpha, K)).$$

- The *maximal rate of convergence* associated with System (2.3) is defined as

$$\text{RC}_{\mathcal{G}}(T, \mu) = \sup_{K \in \mathcal{M}_{m,d}(\mathbb{R})} \text{rc}_{\mathcal{G}}(T, \mu, K) \quad (\text{resp. } \text{RC}_{\mathcal{D}}(T, \mu, M) = \sup_{K \in \mathcal{M}_{m,d}(\mathbb{R})} \text{rc}_{\mathcal{D}}(T, \mu, M, K)).$$

The stabilization of System (2.3) at an arbitrary rate of convergence corresponds thus to have $\text{RC}_{\mathcal{G}}(T, \mu) = +\infty$ or $\text{RC}_{\mathcal{D}}(T, \mu, M) = +\infty$.

The fact that we are interested in the maximal rate of convergence explains why we consider only the case where the pair $(A, B) \in \mathcal{M}_d(\mathbb{R}) \times \mathcal{M}_{d,m}(\mathbb{R})$ is controllable, for, when this is not the case, there are eigenvalues of $A - BK$ which do not depend on K , and then we do not have a result of stabilization at an arbitrary rate even in the case of a non-switched linear control system $\dot{x} = Ax + Bu$.

2.2 Previous results

The first stabilization problem is the case of a neutrally stable system, that is, a system in the form (2.1) such that every eigenvalue of A has non-positive real part, and those with real part zero have trivial Jordan blocks. Under such hypothesis on A , and assuming that $(A, B) \in \mathcal{M}_d(\mathbb{R}) \times \mathcal{M}_{d,m}(\mathbb{R})$ is stabilizable it is proved in [1, 4] that there exists a matrix $K \in \mathcal{M}_{m,d}(\mathbb{R})$ such that, for every $T \geq \mu > 0$, K is a (T, μ) -stabilizer for the PE system

$$\dot{x} = Ax + \alpha(t)Bu, \quad \alpha \in \mathcal{G}(T, \mu). \quad (2.5)$$

We remark that the gain K is independent of T and μ .

The next case that have been studied after the neutrally stable one has been the double integrator ([4]), which has been generalized in [5] as follows.

Theorem 2.6. *Let $(A, b) \in \mathcal{M}_d(\mathbb{R}) \times \mathbb{R}^d$ be a controllable pair and assume that the eigenvalues of A have non-positive real part. Then, for every T, μ with $T \geq \mu > 0$, there exists a (T, μ) -stabilizer for (2.5).*

In order to justify the analysis of this paper it is useful to recall briefly how the proof of Theorem 2.6 goes. To capture its main features, it is enough to consider the case of the double integrator, i.e., $A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$, $b = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$. System (2.5) is thus written as

$$\begin{cases} \dot{x}_1 = x_2, \\ \dot{x}_2 = \alpha(t)u. \end{cases} \quad (2.6)$$

For every $\nu > 0$, $K = (k_1 \ k_2)$ is a (T, μ) -stabilizer of (2.6) if and only if $(\nu^2 k_1 \ \nu k_2)$ is a $(T/\nu, \mu/\nu)$ -stabilizer of (2.6), as it can be seen by considering the equation satisfied by

$$x_\nu(t) = \begin{pmatrix} 1 & 0 \\ 0 & \nu \end{pmatrix} x(\nu t).$$

The idea of the proof is thus to construct a $(T/\nu, \mu/\nu)$ -stabilizer $K = (k_1 \ k_2)$ for (2.6) for a certain ν large enough, and then the (T, μ) -stabilizer we seek for is $(k_1/\nu^2 \ k_2/\nu)$. The construction of such a K is based on a limit process: given a family of signals $\alpha_n \in \mathcal{G}(T/\nu_n, \mu/\nu_n)$ with $\lim_{n \rightarrow +\infty} \nu_n = +\infty$, by weak- \star compactness of $L^\infty(\mathbb{R}_+, [0, 1])$ there exist a subsequence weak- \star

converging in $L^\infty(\mathbb{R}_+, [0, 1])$ to a certain limit α_* , which can be shown to satisfy $\alpha_*(t) \geq \frac{\mu}{T}$ almost everywhere. We can thus study the limit system

$$\begin{cases} \dot{x}_1 = x_2, \\ \dot{x}_2 = \alpha_*(t)u, \end{cases} \quad \alpha_*(t) \geq \frac{\mu}{T},$$

in order to obtain properties of System (2.6) by a limit process. The general idea is thus to accelerate the dynamics of (2.6) by a factor ν . This acceleration reduces the importance of the intervals where α is small, since, in the limit, $\alpha_*(t) \geq \frac{\mu}{T}$ almost everywhere, making it easier to study the behavior of the system. We thus construct a gain $K = \begin{pmatrix} k_1 & k_2 \end{pmatrix}$ for a system for which the accelerating scale ν is big enough, and finally we are able to stabilize the original system by a small gain $\begin{pmatrix} k_1/\nu^2 & k_2/\nu \end{pmatrix}$.

The result recalled in Theorem 2.6 left open, for (T, μ) given, the case where A has at least one eigenvalue with positive real part. That issue was somehow resolved by reformulating the question as a stabilization problem with arbitrary rate of convergence. Let us state the latter issue in terms of the maximal rates of convergence as the problem of determining whether $\text{RC}_g(T, \mu)$ and $\text{RC}_D(T, \mu, M)$ are finite or not. In this sense, [5] gives two results concerning the stabilization of PE systems and points out the importance played by the parameter $\frac{\mu}{T}$.

Theorem 2.7. *Let d be a positive integer. There exists $\rho^* \in (0, 1)$ such that, for every controllable pair $(A, b) \in \mathcal{M}_d(\mathbb{R}) \times \mathbb{R}^d$ and every positive T, μ satisfying $\rho^* < \frac{\mu}{T} \leq 1$, one has $\text{RC}_g(T, \mu) = +\infty$.*

This means that, at least for $\frac{\mu}{T}$ large enough, stabilization at an arbitrary rate of convergence is possible for a PE system with *any* controllable (A, b) . Nevertheless, [5] also proves that the result is false for $\frac{\mu}{T}$ small, at least in dimension 2.

Theorem 2.8. *There exists $\rho_* \in (0, 1)$ such that, for every controllable pair $(A, b) \in \mathcal{M}_2(\mathbb{R}) \times \mathbb{R}^2$ and every positive T, μ satisfying $0 < \frac{\mu}{T} < \rho_*$, one has $\text{RC}_g(T, \mu) < +\infty$.*

As recalled in the introduction, the proof of Theorem 2.8 is based on the explicit construction of fast-oscillating controls. This motivates the conjecture of the fact that $\text{RC}_D(T, \mu, M) = +\infty$. The technique used in the proof of the Theorem 2.6 could not provide any help in this case: the direct study of a limit system comes from accelerating the dynamics of the system, which would mean that the signal α would be taken in $\mathcal{D}(T/\nu, \mu/\nu, \nu M)$ for a large constant ν and thus, in the limit as ν tends to infinity, the fact that α is νM -Lipschitz would provide no additional information on a weak- \star limit function α_* .

3 Main result

The main result we want to prove concerns planar systems of the type (2.1). More specifically, we fix positive constants T, μ and M with $T \geq \mu$ and we study the PEL system

$$\dot{x} = Ax + \alpha(t)bu, \tag{3.1}$$

where $x \in \mathbb{R}^2$, (A, b) is controllable and $\alpha \in \mathcal{D}(T, \mu, M)$. Then, we get the following result.

Theorem 3.1. *Let T , μ and M be positive constants with $T \geq \mu$. Then, for the PEL System (3.1), one has $RC_{\mathcal{D}}(T, \mu, M) = +\infty$.*

The rest of the section is devoted to the proof of Theorem 3.1.

We first perform linear algebraic transformation on the control system. With no loss of generality, we can assume that (A, b) is under controllable form, i.e., $A = \begin{pmatrix} 0 & 1 \\ -d & \text{Tr}(A) \end{pmatrix}$ and $b = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$, where $\text{Tr}(A)$ is the trace of A . Moreover, if A is replaced by $A - \text{Tr}(A)\text{Id}_2$, then $RC_{\mathcal{D}}(T, \mu, M)$ is simply translated by $-\text{Tr}(A)$. It is therefore enough to prove the theorem assuming that $\text{Tr}(A) = 0$.

The system can thus be written in the form

$$\begin{cases} \dot{x}_1 = x_2, \\ \dot{x}_2 = -d + \alpha(t)u, \end{cases} \quad \alpha \in \mathcal{D}(T, \mu, M). \quad (3.2)$$

From now on, we suppose that T , μ , M , d and λ are fixed. We prove Theorem 3.1 by explicitly constructing the gain K that satisfies $\lambda^+(\alpha, K) \leq -\lambda$ for every $\alpha \in \mathcal{D}(T, \mu, M)$. To do so, we write $K = \begin{pmatrix} k_1 & k_2 \end{pmatrix}$ and thus the feedback $u = -Kx$ leads to the system

$$\dot{x} = \begin{pmatrix} 0 & 1 \\ -(d + \alpha(t)k_1) & -\alpha(t)k_2 \end{pmatrix} x.$$

The variable x_1 satisfies the scalar equation $\ddot{x}_1 + k_2\alpha(t)\dot{x}_1 + (d + k_1\alpha(t))x_1 = 0$ and we have $x_2 = \dot{x}_1$.

We remark that the signal α constant and equal to 1 is in $\mathcal{D}(T, \mu, M)$, and thus a necessary condition for K to be a (T, μ, M) -stabilizer is that the matrix

$$A - bK = \begin{pmatrix} 0 & 1 \\ -(d + k_1) & -k_2 \end{pmatrix}$$

is Hurwitz, which is the case if and only if $k_1 > -d$, $k_2 > 0$. In what follows, we restrict ourselves to search K in the form

$$K = \begin{pmatrix} k^2 & k \end{pmatrix}, \quad k \text{ positive and large.} \quad (3.3)$$

The differential equation satisfied by x_1 is thus

$$\ddot{x}_1 + k\alpha(t)\dot{x}_1 + (d + k^2\alpha(t))x_1 = 0. \quad (3.4)$$

3.1 Strategy of the proof

Let us discuss briefly the strategy that we will use to prove Theorem 3.1. We start, in Section 3.2, by making a change of variables on (3.4) that makes the systems easier to handle. The new variable y is related to x by an exponential term $e^{-\frac{k}{2} \int_0^t \alpha(s) ds + \sqrt{\frac{kM}{2}} t}$ that converges to 0 as $t \rightarrow +\infty$ (see (3.6)). The problem is then to estimate the rate of exponential growth of y (Section 3.3). We start by proving that y turns around the origin infinitely many times (Section 3.3.2). On each complete turn the exponential growth of y is estimated either by direct integration when the exciting signal is "large" (Section 3.3.4) or by optimal control where it is "small" (Section 3.3.5).

3.2 Change of variables

In order to simplify the notations, we write $h = \sqrt{2kM - 4d}$, which is well defined for $k \geq \frac{2|d|}{M}$. We consider the system in a new variable $y = (y_1 \ y_2)^T$ defined by the relations

$$\begin{cases} y_1 = x_1 e^{\frac{k}{2} \int_0^t \alpha(s) ds - \frac{h}{2} t}, \\ y_2 = \dot{y}_1 = \left(x_2 + \left(\frac{k}{2} \alpha(t) - \frac{h}{2} \right) x_1 \right) e^{\frac{k}{2} \int_0^t \alpha(s) ds - \frac{h}{2} t}, \end{cases} \quad (3.5)$$

whose choice is justified here below. The variables x and y are thus related by

$$y = e^{\frac{k}{2} \int_0^t \alpha(s) ds - \frac{h}{2} t} \begin{pmatrix} 1 & 0 \\ \frac{k}{2} \alpha(t) - \frac{h}{2} & 1 \end{pmatrix} x, \quad x = e^{-\frac{k}{2} \int_0^t \alpha(s) ds + \frac{h}{2} t} \begin{pmatrix} 1 & 0 \\ \frac{h}{2} - \frac{k}{2} \alpha(t) & 1 \end{pmatrix} y \quad (3.6)$$

and y_1 satisfies the differential equation

$$\ddot{y}_1 + h\dot{y}_1 + k^2\gamma(t)y_1 = 0 \quad (3.7)$$

with

$$\gamma(t) = \beta(t) + \frac{M - \dot{\alpha}(t)}{2k}, \quad \beta(t) = \alpha(t) \left(1 - \frac{1}{4} \alpha(t) \right). \quad (3.8)$$

The system satisfied by y is

$$\dot{y} = \begin{pmatrix} 0 & 1 \\ -k^2\gamma(t) & -h \end{pmatrix} y. \quad (3.9)$$

Since $\alpha(t) \in [0, 1]$ for every $t \in \mathbb{R}_+$, we have $\beta(t) \in [0, 3/4]$. Furthermore, since α is M -Lipschitz, β is also Lipschitz with the same Lipschitz constant, since

$$\begin{aligned} |\beta(t) - \beta(s)| &= \left| \alpha(t) - \alpha(s) - \frac{1}{4} (\alpha(t)^2 - \alpha(s)^2) \right| = |\alpha(t) - \alpha(s)| \left| 1 - \frac{\alpha(t) + \alpha(s)}{4} \right| \\ &\leq |\alpha(t) - \alpha(s)| \leq M|t - s| \end{aligned}$$

for every $t, s \in \mathbb{R}_+$. Since α satisfies the PE condition (2.2), β satisfies

$$\int_t^{t+T} \beta(s) ds \geq \frac{3}{4} \mu. \quad (3.10)$$

Since $|\dot{\alpha}(t)| \leq M$ almost everywhere on every bounded interval of \mathbb{R}_+ , γ can be bounded by

$$0 \leq \gamma(t) \leq \frac{3}{4} + \frac{M}{k}$$

almost everywhere on every bounded interval of \mathbb{R}_+ . It also satisfies the PE condition

$$\int_t^{t+T} \gamma(s) ds \geq \frac{3}{4} \mu. \quad (3.11)$$

From now on, we also suppose that

$$k \geq K_1(M) := \max \left(4M, \frac{2|d|}{M} \right), \quad (3.12)$$

so that $h \leq 2\sqrt{kM}$ and

$$0 \leq \gamma(t) \leq 1$$

for almost every $t \in \mathbb{R}_+$.

Let us discuss the change of variables (3.5). The term $e^{\frac{k}{2} \int_0^t \alpha(s) ds}$ corresponds to a classical change of variables in second-order scalar equations (see, for instance, [7]) that eliminates the term in \dot{x}_1 from (3.4), which is replaced by a new term $-\frac{1}{4}k^2\alpha(t)^2 - \frac{k}{2}\dot{\alpha}(t)$ multiplying y_1 . However, if we took only this term in the change of variables, the resulting function γ would be $\gamma(t) = \beta(t) + \frac{2d/k - \dot{\alpha}(t)}{2k}$, which may be negative at certain times t . To apply the techniques of optimal control of Section 3.3.5, it is important to manipulate a positive function γ , and that is why we introduce the term $e^{\frac{h}{2}t}$ in the change of variables.

Another important feature of this change of variables is that $x(t)$ behaves like $e^{-\frac{k}{2} \int_0^t \alpha(s) ds + \frac{h}{2}t} y(t)$. Since $h \leq 2\sqrt{kM}$ and α is persistently exciting, this exponential factor is bounded by $e^{-c_1 kt}$ for large k , for a certain $c_1 > 0$. We have now to show that the exponential growth of y is bounded by $e^{c_2 k^s t}$ for large k , for some $c_2 > 0$ and $s < 1$.

This change of variables also justifies the choice of K on the form (3.3). Equation (3.7) is a linear second-order scalar differential equation, and, in the case where its coefficients are constant, $h\dot{y}_1$ can be interpreted as a damping term and $k^2\gamma y_1$ as an oscillatory term. Such a system oscillates around the origin if $4k^2\gamma \geq h^2 = 2kM - 4d$, which is the case for k large enough. In the case where γ depends on time, the PE condition (3.11) still guarantees a certain oscillatory behavior for k large enough, which is used in order to prove Theorem 3.1.

3.3 Properties of the system in the new variables

3.3.1 Polar coordinates

We now wish to study System (3.9) and the corresponding differential equation (3.7). To do so, we first write this system in polar coordinates in the plan (y_1, \dot{y}_1) : we define the variables $r \in \mathbb{R}_+$ and $\theta \in \mathbb{R}$ (or $\theta \in \mathbb{R}/2\pi\mathbb{Z}$, depending on the context) by the relations

$$\begin{aligned} y_1 &= r \cos \theta, \\ \dot{y}_1 &= r \sin \theta, \end{aligned}$$

which leads to the equations

$$\dot{\theta} = -\sin^2 \theta - k^2\gamma(t)\cos^2 \theta - h \sin \theta \cos \theta, \quad (3.13a)$$

$$\dot{r} = r \sin \theta \cos \theta (1 - k^2\gamma(t)) - hr \sin^2 \theta. \quad (3.13b)$$

For nonzero solutions we can write (3.13b) as

$$\frac{d}{dt} \ln r = \sin \theta \cos \theta (1 - k^2\gamma(t)) - h \sin^2 \theta. \quad (3.13c)$$

3.3.2 Rotations around the origin

Let us consider Equation (3.13a). If $\sin \theta \cos \theta \geq 0$, then $\dot{\theta} \leq 0$, with the strict inequality being true for all times except when $\sin \theta = 0$ and $\gamma = 0$. The following lemma shows that in the general case it is still possible, for k large enough, to guarantee that y keeps on turning clockwise around the origin, even if, at certain points, it may go counterclockwise for a short period of time.

Lemma 3.2. *There exists $K_2(T, \mu, M)$ such that, for $k > K_2(T, \mu, M)$, the solution θ of (3.13a) satisfies $\lim_{t \rightarrow +\infty} \theta(t) = -\infty$.*

Proof. We start by fixing $t \in \mathbb{R}_+$ and the interval $I = [t, t + T]$. Equation (3.10) shows that there exists $t_\star \in I$ such that $\beta(t_\star) \geq \frac{3\mu}{4T}$. Since β is M -Lipschitz, we have $\beta(s) \geq \frac{\mu}{2T}$ if $|s - t_\star| \leq \frac{\mu}{4MT}$, and thus, since $\gamma(s) \geq \beta(s)$, we have $\gamma(s) \geq \frac{\mu}{2T}$ for $|s - t_\star| \leq \frac{\mu}{4MT}$. If we take

$$k \geq \max \left\{ 1, \left(\frac{\mu}{2MT^2} \right)^4 \right\}, \quad (3.14)$$

we have $\frac{\mu}{4MTk^{1/4}} \leq \frac{\mu}{4MT}$ and $\frac{\mu}{4MTk^{1/4}} \leq T/2$, which implies that at least one of the intervals $\left[t_\star - \frac{\mu}{4MTk^{1/4}}, t_\star \right]$ and $\left[t_\star, t_\star + \frac{\mu}{4MTk^{1/4}} \right]$ is included in I ; let us denote this interval by $J = [s_0, s_1]$, so that $s_1 - s_0 = \frac{\mu}{4MTk^{1/4}}$ and $\gamma(s) \geq \frac{\mu}{2T}$ for $s \in J$.

If $s \in J$, one can estimate $\dot{\theta}$ in (3.13a) by

$$-\dot{\theta}(s) \geq \sin^2 \theta(s) + \frac{\mu k^2}{2T} \cos^2 \theta(s) + h \sin \theta(s) \cos \theta(s) = (\sin \theta(s) \cos \theta(s)) \begin{pmatrix} 1 & \frac{h}{2} \\ \frac{h}{2} & \frac{\mu k^2}{2T} \end{pmatrix} \begin{pmatrix} \sin \theta(s) \\ \cos \theta(s) \end{pmatrix}.$$

In particular, if

$$k > \frac{2MT}{\mu}, \quad (3.15)$$

then the matrix $\begin{pmatrix} 1 & \frac{h}{2} \\ \frac{h}{2} & \frac{\mu k^2}{2T} \end{pmatrix}$ is positive definite and thus $\dot{\theta}(s) < 0$ for every $s \in J$. Therefore θ is strictly decreasing on J and is a bijection between J and its image $\theta(J)$. One can write Equation (3.13a) on J as

$$\frac{\dot{\theta}}{\sin^2 \theta + k^2 \gamma \cos^2 \theta + h \sin \theta \cos \theta} = -1 \quad (3.16)$$

and, by integrating from s_0 to s_1 and using the relation

$$\int_{-\pi/2}^{\pi/2} \frac{d\theta}{\sin^2 \theta + a \cos^2 \theta + b \sin \theta \cos \theta} = \frac{2\pi}{\sqrt{4a - b^2}}, \quad a > 0, b^2 < 4a,$$

which can be computed directly by the change of variables $\hat{t} = \tan \theta$, we obtain

$$\begin{aligned}
\frac{\mu}{4MTk^{1/4}} = s_1 - s_0 &= - \int_{s_0}^{s_1} \frac{\dot{\theta}(s)}{\sin^2 \theta(s) + k^2 \gamma(s) \cos^2 \theta(s) + h \sin \theta(s) \cos \theta(s)} ds \\
&\leq \int_{\theta(s_1)}^{\theta(s_0)} \frac{d\theta}{\sin^2 \theta + \frac{k^2 \mu}{2T} \cos^2 \theta + h \sin \theta \cos \theta} \\
&\leq \int_{\theta(s_1)}^{\theta(s_1) + \pi(N+1)} \frac{d\theta}{\sin^2 \theta + \frac{k^2 \mu}{2T} \cos^2 \theta + h \sin \theta \cos \theta} \\
&= \frac{2\pi(N+1)}{\sqrt{\frac{2k^2 \mu}{T} - h^2}} \leq \frac{2\pi(N+1)}{\sqrt{\frac{2\mu}{T} k^2 - 4Mk}},
\end{aligned} \tag{3.17}$$

where N is the number of rotations of angle π during the interval J , i.e., $N = \left\lfloor \frac{\theta(s_0) - \theta(s_1)}{\pi} \right\rfloor$. Therefore

$$\theta(s_0) - \theta(s_1) \geq \pi N \geq k^{3/4} \frac{\mu}{8MT} \sqrt{\frac{2\mu}{T} - \frac{4M}{k}} - \pi. \tag{3.18}$$

On the other hand, one can estimate $\dot{\theta}$ in (3.13a) for every $s \in I$ by $\dot{\theta}(s) \leq h$, so that

$$\theta(s_0) - \theta(t) \leq h(s_0 - t), \quad \theta(t+T) - \theta(s_1) \leq h(t+T - s_1). \tag{3.19}$$

Thus, by (3.18) and (3.19), we obtain

$$\theta(t+T) - \theta(t) \leq 2\sqrt{kMT} - k^{3/4} \frac{\mu}{8MT} \sqrt{\frac{2\mu}{T} - \frac{4M}{k}} + \pi.$$

The expression on the right-hand side tends to $-\infty$ as $k \rightarrow +\infty$ and the parameters T , μ and M are fixed. Hence, there exists $K_*(T, \mu, M)$ such that, if

$$k \geq K_*(T, \mu, M), \tag{3.20}$$

then $2\sqrt{kMT} - k^{3/4} \frac{\mu}{8MT} \sqrt{\frac{2\mu}{T} - \frac{4M}{k}} + \pi \leq -2\pi$ and thus $\theta(t+T) - \theta(t) \leq -2\pi$. We group conditions (3.14), (3.15) and (3.20) in a single one by setting

$$K_2(T, \mu, M) = \max \left\{ 1, \left(\frac{\mu}{2MT^2} \right)^4, \frac{2MT}{\mu}, K_*(T, \mu, M) \right\}$$

and asking that $k > K_2(T, \mu, M)$. Under this condition, the solution completes at least one entire clockwise rotation by the end of the interval $[t, t+T]$. This result being true for every $t \in \mathbb{R}_+$, the proof is completed. \blacksquare

3.3.3 Decomposition of the time in intervals \mathcal{J}_+ and \mathcal{J}_-

Using Lemma 3.2, we can decompose \mathbb{R}_+ in a sequence of intervals (depending on α) on which the solution rotates by an angle π around the origin. More precisely, we define the sequence $(t_n)_{n \in \mathbb{N}}$ by induction as

$$t_0 = \inf\{t \geq 0 \mid \frac{\theta(t)}{\pi} \in \mathbb{Z}\}, \quad t_n = \inf\{t \geq t_{n-1} \mid \theta(t) = \theta(t_{n-1}) - \pi\}, \quad n \geq 1, \tag{3.21}$$

and the continuity of θ and Lemma 3.2 show that this sequence is well defined. We also define the sequence of intervals $(I_n)_{n \in \mathbb{N}}$ by $I_n = [t_{n-1}, t_n]$ for $n \geq 1$ and $I_0 = [0, t_0]$.

Let us show a first result about the behavior of θ on these intervals.

Lemma 3.3. *Let $n \geq 1$. Then, for every $t \in I_n = [t_{n-1}, t_n]$, one has*

$$\theta(t_n) \leq \theta(t) \leq \theta(t_{n-1}). \quad (3.22)$$

Proof. The first inequality in (3.22) is a consequence of the definition of t_n : if there was $t \in I_n$ with $\theta(t) < \theta(t_n)$, then, by the continuity of θ , there would be $s \in [t_{n-1}, t]$ such that $\theta(s) = \theta(t_n) = \theta(t_{n-1}) - \pi$, leading to a contradiction.

The second inequality in (3.22) can also be proved by contradiction. We suppose that there exists $t \in I_n$ such that $\theta(t) > \theta(t_{n-1})$. Then, by continuity of θ , there exists $s_0, s_1 \in [t_{n-1}, t]$ such that $\theta(s_0) = \theta(t_{n-1})$, $\theta(s_1) > \theta(t_{n-1})$ and $\theta(s) \in [\theta(t_{n-1}), \theta(t_{n-1}) + \pi/2]$ for every $s \in [s_0, s_1]$. But we have that $\theta(t_{n-1}) = 0 \pmod{\pi}$, so that $\sin \vartheta \cos \vartheta \geq 0$ for $\vartheta \in [\theta(t_{n-1}), \theta(t_{n-1}) + \pi/2]$, and thus, by (3.13a), $\dot{\theta}(s) \leq 0$ for almost every $s \in [s_0, s_1]$, which contradicts the fact that $\theta(s_0) < \theta(s_1)$. ■

We now split the intervals of the sequence $(I_n)_{n \in \mathbb{N}^*}$ into two classes, \mathcal{J}_+ and \mathcal{J}_- , according to the behavior of β on these intervals. We define

$$\mathcal{J}_+ = \{I_n \mid n \in \mathbb{N}^* \mid \exists t \in I_n, \beta(t) \geq 2/\sqrt{k}\}, \quad \mathcal{J}_- = \{I_n \mid n \in \mathbb{N}^* \mid \forall t \in I_n, \beta(t) < 2/\sqrt{k}\}.$$

3.3.4 Estimations on intervals \mathcal{J}_+

We start by studying the intervals in the class \mathcal{J}_+ . We first claim that, for k large enough, we have $\gamma(t) \geq 1/\sqrt{k}$ for almost every $t \in I$ and every $I \in \mathcal{J}_+$.

Lemma 3.4. *There exists $K_3(M)$ such that, for $k > K_3(M)$ and for every $I \in \mathcal{J}_+$, one has $\beta(t) \geq 1/\sqrt{k}$ for every $t \in I$ and $\gamma(t) \geq 1/\sqrt{k}$ for almost every $t \in I$.*

Proof. We fix an interval $I = [t_{n-1}, t_n] \in \mathcal{J}_+$ and we note $t_* \in I$ an element of I such that $\beta(t_*) \geq 2/\sqrt{k}$. Since β is M -Lipschitz, for every t such that $|t - t_*| \leq \frac{1}{M\sqrt{k}}$, we have $1/\sqrt{k} \leq \beta(t) \leq 3/\sqrt{k}$. In particular, since $\gamma(t) \geq \beta(t)$ on \mathbb{R}_+ , we have $\gamma(t) \geq 1/\sqrt{k}$ for $|t - t_*| \leq \frac{1}{M\sqrt{k}}$.

The idea is to show that, for k large enough, $I \subset \left[t_* - \frac{1}{M\sqrt{k}}, t_* + \frac{1}{M\sqrt{k}} \right]$. This is done by proving that, for k large enough, the number of rotations of angle π around the origin done on each of the intervals $\left[t_* - \frac{1}{M\sqrt{k}}, t_* \right]$ and $\left[t_*, t_* + \frac{1}{M\sqrt{k}} \right]$ is larger than 1.

We take $s_0, s_1 \in \left[t_* - \frac{1}{M\sqrt{k}}, t_* + \frac{1}{M\sqrt{k}} \right]$, $s_0 < s_1$. For every $s \in [s_0, s_1]$, we have

$$\begin{aligned} -\dot{\theta}(s) &\geq \sin^2 \theta(s) + k^{3/2} \cos^2 \theta(s) + h \sin \theta(s) \cos \theta(s) = \\ &= (\sin \theta(s) \quad \cos \theta(s)) \begin{pmatrix} 1 & \frac{h}{2} \\ \frac{h}{2} & k^{3/2} \end{pmatrix} \begin{pmatrix} \sin \theta(s) \\ \cos \theta(s) \end{pmatrix}, \end{aligned}$$

and the matrix $\begin{pmatrix} 1 & \frac{h}{2} \\ \frac{h}{2} & k^{3/2} \end{pmatrix}$ is positive definite if

$$k > M^2. \quad (3.23)$$

We take k satisfying (3.23). We can thus write Equation (3.13a) on $[s_0, s_1]$ as (3.16), and by integrating as in (3.17), we obtain

$$s_1 - s_0 \leq \int_{\theta(s_1)}^{\theta(s_1) + \pi(N(s_0, s_1) + 1)} \frac{d\theta}{\sin^2 \theta + k^{3/2} \cos^2 \theta + h \sin \theta \cos \theta} \leq \frac{\pi(N(s_0, s_1) + 1)}{k^{3/4} \sqrt{1 - \frac{M}{k^{1/2}}}},$$

where $N(s_0, s_1) = \left\lfloor \frac{\theta(s_0) - \theta(s_1)}{\pi} \right\rfloor$ is the number of rotations of angle π around the origin done by the solution between s_0 and s_1 . Hence

$$N(s_0, s_1) \geq k^{3/4} \frac{(s_1 - s_0)}{\pi} \sqrt{1 - \frac{M}{k^{1/2}}} - 1,$$

and, in particular,

$$N\left(t_\star - \frac{1}{M\sqrt{k}}, t_\star\right) \geq \frac{k^{1/4}}{M\pi} \sqrt{1 - \frac{M}{k^{1/2}}} - 1, \quad N\left(t_\star, t_\star + \frac{1}{M\sqrt{k}}\right) \geq \frac{k^{1/4}}{M\pi} \sqrt{1 - \frac{M}{k^{1/2}}} - 1.$$

For M fixed, we have $\frac{k^{1/4}}{M\pi} \sqrt{1 - \frac{M}{k^{1/2}}} - 1 \xrightarrow[k \rightarrow +\infty]{} +\infty$, and thus there exists $K_\star(M)$ such that, for

$$k > K_\star(M), \tag{3.24}$$

one has $\frac{k^{1/4}}{M\pi} \sqrt{1 - \frac{M}{k^{1/2}}} - 1 > 1$. Therefore $N\left(t_\star - \frac{1}{M\sqrt{k}}, t_\star\right) > 1$ and $N\left(t_\star, t_\star + \frac{1}{M\sqrt{k}}\right) > 1$, and then

$$\theta(t_\star) - \theta\left(t_\star + \frac{1}{M\sqrt{k}}\right) > \pi, \quad \theta\left(t_\star - \frac{1}{M\sqrt{k}}\right) - \theta(t_\star) > \pi.$$

By definition of I and thanks to Lemma 3.3,

$$t_\star - \frac{1}{M\sqrt{k}} < t_{n-1}, \quad t_\star + \frac{1}{M\sqrt{k}} > t_n,$$

and then $I \subset \left[t_\star - \frac{1}{M\sqrt{k}}, t_\star + \frac{1}{M\sqrt{k}}\right]$. According to (3.23) and (3.24) the lemma is proved by setting $K_3(M) = \max\{M^2, K_\star(M)\}$. \blacksquare

By using the above result, we can estimate the divergence rate of the solution over an interval $I \in \mathcal{J}_+$.

Lemma 3.5. *There exists $K_4(M)$ such that, for $k > K_4(M)$ and for every $I = [t_{n-1}, t_n] \in \mathcal{J}_+$, the solution of (3.13c) satisfies*

$$r(t_n) \leq r(t_{n-1}) e^{4Mk^{1/2}(t_n - t_{n-1})}. \tag{3.25}$$

Proof. We start by taking

$$k > K_3(M), \tag{3.26}$$

so that we can apply Lemma 3.4 and obtain that, for almost every $t \in I$, $\beta(t), \gamma(t) \geq 1/\sqrt{k}$ and

$$\begin{aligned} -\dot{\theta}(t) &\geq \sin^2 \theta(t) + k^{3/2} \cos^2 \theta(t) + h \sin \theta(t) \cos \theta(t) = \\ &= (\sin \theta(t) \quad \cos \theta(t)) \begin{pmatrix} 1 & \frac{h}{2} \\ \frac{h}{2} & k^{3/2} \end{pmatrix} \begin{pmatrix} \sin \theta(t) \\ \cos \theta(t) \end{pmatrix} > 0. \end{aligned}$$

Hence θ is a continuous bijection between $I = [t_{n-1}, t_n]$ and its image $[\theta(t_n), \theta(t_{n-1})]$. We note by τ the inverse of θ , defined on $[\theta(t_n), \theta(t_{n-1})]$; τ thus satisfies

$$\frac{d\tau}{d\vartheta}(\vartheta) = \frac{1}{\dot{\theta}(\tau(\vartheta))} = -\frac{1}{\sin^2 \vartheta + k^2 \gamma(\tau(\vartheta)) \cos^2 \vartheta + h \sin \vartheta \cos \vartheta}. \quad (3.27)$$

Writing $\rho = r \circ \tau$ and using Equations (3.13c) and (3.27), we have

$$\frac{d}{d\vartheta} \ln \rho = -\frac{\sin \vartheta \cos \vartheta (1 - k^2 \gamma \circ \tau(\vartheta)) - h \sin^2 \vartheta}{\sin^2 \vartheta + k^2 \gamma \circ \tau(\vartheta) \cos^2 \vartheta + h \sin \vartheta \cos \vartheta}.$$

We can integrate this expression from $\theta(t_n)$ to $\theta(t_{n-1}) = \theta(t_n) + \pi$, obtaining

$$\ln \frac{r(t_n)}{r(t_{n-1})} = \int_{\theta(t_n)}^{\theta(t_n)+\pi} F(\vartheta, \gamma \circ \tau(\vartheta)) d\vartheta,$$

with $F(\vartheta, \gamma) = \frac{\sin \vartheta \cos \vartheta (1 - k^2 \gamma) - h \sin^2 \vartheta}{\sin^2 \vartheta + k^2 \gamma \cos^2 \vartheta + h \sin \vartheta \cos \vartheta}$. If $\gamma_0 \geq 1/\sqrt{k}$ is constant, then

$$\int_{\theta(t_n)}^{\theta(t_n)+\pi} F(\vartheta, \gamma_0) d\vartheta \leq 0, \quad (3.28)$$

since F is π -periodic on ϑ and, using the change of variables $\hat{t} = \tan \vartheta$,

$$\int_{-\pi/2}^{\pi/2} F(\vartheta, \gamma_0) d\vartheta = \int_{-\infty}^{+\infty} \frac{(1 - k^2 \gamma_0) \hat{t} - h \hat{t}^2}{(\hat{t}^2 + h \hat{t} + k^2 \gamma_0)(\hat{t}^2 + 1)} d\hat{t} \leq \int_{-\infty}^{+\infty} \frac{(1 - k^2 \gamma_0) \hat{t}}{(a_0 \hat{t}^2 + b_0)(\hat{t}^2 + 1)} d\hat{t} = 0,$$

where $a_0 = \frac{k^2 \gamma_0 - h^2/4}{k^2 \gamma_0 + h^2/4}$ and $b_0 = \frac{k^2 \gamma_0}{2} - \frac{h^2}{8}$ are positive because $\gamma_0 \geq 1/\sqrt{k}$ and thanks to (3.23).

By (3.28), we have

$$\ln \frac{r(t_n)}{r(t_{n-1})} \leq \int_{\theta(t_n)}^{\theta(t_n)+\pi} [F(\vartheta, \gamma \circ \tau(\vartheta)) - F(\vartheta, \gamma_0)] d\vartheta. \quad (3.29)$$

We compute

$$\frac{\partial F}{\partial \gamma}(\vartheta, \gamma) = -\frac{k^2 \sin \vartheta \cos \vartheta}{(\sin^2 \vartheta + k^2 \gamma \cos^2 \vartheta + h \sin \vartheta \cos \vartheta)^2},$$

and thus, for $t \in I$,

$$\left| \frac{\partial F}{\partial \gamma}(\vartheta, \gamma(t)) \right| \leq \frac{k^2 |\sin \vartheta| |\cos \vartheta|}{(\sin^2 \vartheta + k^{3/2} \cos^2 \vartheta + h \sin \vartheta \cos \vartheta)^2}.$$

We now take $\gamma_0 = \beta(t_{n-1})$ in (3.29), obtaining

$$\ln \frac{r(t_n)}{r(t_{n-1})} \leq \int_{\theta(t_n)}^{\theta(t_n)+\pi} \frac{k^2 |\sin \vartheta| |\cos \vartheta|}{(\sin^2 \vartheta + k^{3/2} \cos^2 \vartheta + h \sin \vartheta \cos \vartheta)^2} |\gamma \circ \tau(\vartheta) - \beta(t_{n-1})| d\vartheta. \quad (3.30)$$

For almost every $t \in I$, one can estimate

$$|\gamma(t) - \beta(t_{n-1})| \leq |\beta(t) - \beta(t_{n-1})| + \left| \frac{\dot{\alpha}(t)}{2k} \right| \leq M(t_n - t_{n-1}) + \frac{M}{2k}. \quad (3.31)$$

We take k satisfying (3.12), which means that $0 \leq \gamma(t) \leq 1$ for almost every $t \in \mathbb{R}_+$, and thus, by integrating (3.16) from t_{n-1} to t_n , we obtain

$$\begin{aligned} t_n - t_{n-1} &= - \int_{t_{n-1}}^{t_n} \frac{\dot{\theta}(s)}{\sin^2 \theta(s) + k^2 \gamma(s) \cos^2 \theta(s) + h \sin \theta(s) \cos \theta(s)} ds \geq \\ &\geq \int_{\theta(t_n)}^{\theta(t_{n-1})+\pi} \frac{d\theta}{\sin^2 \theta + k^2 \cos^2 \theta + h \sin \theta \cos \theta} \geq \frac{\pi}{k}, \end{aligned}$$

from where we get $\frac{1}{k} \leq \frac{t_n - t_{n-1}}{\pi}$ and thus (3.31) becomes $|\gamma(t) - \beta(t_{n-1})| \leq M \left(1 + \frac{1}{2\pi}\right) (t_n - t_{n-1}) < 2M(t_n - t_{n-1})$. We use this estimate in (3.30), which leads to

$$\ln \frac{r(t_n)}{r(t_{n-1})} \leq 2k^2 M(t_n - t_{n-1}) \int_{\theta(t_n)}^{\theta(t_{n-1})+\pi} \frac{|\sin \vartheta| |\cos \vartheta|}{(\sin^2 \vartheta + k^{3/2} \cos^2 \vartheta + h \sin \vartheta \cos \vartheta)^2} d\vartheta. \quad (3.32)$$

Notice that, for any $a > 0$ and b satisfying $b^2 < 4a$,

$$\int_{-\pi/2}^{\pi/2} \frac{|\sin \vartheta| |\cos \vartheta|}{(\sin^2 \vartheta + a \cos^2 \vartheta + b \sin \vartheta \cos \vartheta)^2} d\vartheta = \frac{1}{A} + \frac{B}{A^{3/2}} \arctan(B/\sqrt{A}) \leq \frac{1}{A} \left(1 + \frac{\pi}{2} C\right),$$

with $A = a - b^2/4 > 0$, $B = b/2$ and $C = B/\sqrt{A} = \frac{b}{\sqrt{4a-b^2}}$. Applying this to (3.32) we have

$$\ln \frac{r(t_n)}{r(t_{n-1})} \leq \frac{2k^{1/2} M(t_n - t_{n-1})}{1 - \frac{M}{k^{1/2}}} \left(1 + \frac{\pi}{2} \sqrt{\frac{kM}{2k^{3/2} - 2kM}}\right)$$

and, since $\frac{1}{1 - \frac{M}{k^{1/2}}} \left(1 + \frac{\pi}{2} \sqrt{\frac{kM}{2k^{3/2} - 2kM}}\right) \xrightarrow[k \rightarrow +\infty]{} 1$, there exists $K_*(M)$ such that, if

$$k \geq K_*(M), \quad (3.33)$$

then $\frac{1}{1 - \frac{M}{k^{1/2}}} \left(1 + \frac{\pi}{2} \sqrt{\frac{kM}{2k^{3/2} - 2kM}}\right) \leq 2$, and thus $\ln \frac{r(t_n)}{r(t_{n-1})} \leq 4k^{1/2} M(t_n - t_{n-1})$. We collect (3.12), (3.26) and (3.33) by setting $K_4(M) = \max\{K_1(M), K_3(M), K_*(M)\}$ and requiring that $k > K_4(M)$. Under this hypothesis, we obtain $r(t_n) \leq r(t_{n-1}) e^{4Mk^{1/2}(t_n - t_{n-1})}$, as required. \blacksquare

3.3.5 Estimations on intervals \mathcal{J}_-

We wish here to obtain a result analogous to Lemma 3.5 for the intervals of the class \mathcal{J}_- . We start by characterizing the duration of these intervals and the behavior of γ on them.

Lemma 3.6. *There exists $K_5(T, \mu, M)$ such that, if $k > K_5(T, \mu, M)$, then for every $I = [t_{n-1}, t_n] \in \mathcal{J}_-$ one has $\gamma(t) \leq 3/\sqrt{k}$ for almost every $t \in I$ and*

$$\frac{\pi}{1 + h + 3k^{3/2}} \leq t_n - t_{n-1} < T.$$

Proof. We fix $I = [t_{n-1}, t_n] \in \mathcal{J}_-$. If

$$k \geq M^2, \quad (3.34)$$

then $0 \leq \gamma(t) - \beta(t) \leq \frac{M}{k} \leq \frac{1}{\sqrt{k}}$, and thus $\gamma(t) \leq 3/\sqrt{k}$ almost everywhere on I . In addition, if

$$k > \left(\frac{8T}{3\mu} \right)^2, \quad (3.35)$$

we have $\beta(t) < 2/\sqrt{k} < \frac{3\mu}{4T}$, and thus, by the persistence of excitation (3.10) of β , we obtain that $t_n - t_{n-1} < T$. Furthermore, (3.13a) implies that $-\dot{\theta} \leq 1 + 3k^{3/2} + h$ almost everywhere on I , and then by integrating on I we get $t_n - t_{n-1} \geq \frac{\pi}{1+h+3k^{3/2}}$. The lemma is proved by taking

$$K_5(T, \mu, M) = \max \left\{ M^2, \left(\frac{8T}{3\mu} \right)^2 \right\},$$

which collects (3.34) and (3.35). ■

We suppose from now on that $k > K_5(T, \mu, M)$. We define the class

$$\mathcal{D}(T, \mu, M, k) = \left\{ \alpha \left(1 - \frac{1}{4}\alpha \right) + \frac{M - \alpha}{2k} \mid \alpha \in \mathcal{D}(T, \mu, M) \right\}$$

where γ belongs. We fix $I = [t_{n-1}, t_n] \in \mathcal{J}_-$ and we remark that, if $\gamma \in \mathcal{D}(T, \mu, M, k)$, then, for every $t_0 \in \mathbb{R}_+$, the function $t \mapsto \gamma(t + t_0)$ is also in $\mathcal{D}(T, \mu, M, k)$. Up to a translation in time, we can then suppose $I = [0, \tau]$ with $\tau = t_n - t_{n-1} \in \left[\frac{\pi}{1+h+3k^{3/2}}, T \right)$. The solution $r(\tau)$ of (3.13c) at time τ can be written as $r(\tau) = r(0)e^{\Lambda\tau}$ for a certain constant Λ . Our goal is to estimate Λ uniformly with respect to γ , i.e., to estimate the maximal value of $\frac{1}{\tau} \ln \frac{\|y(\tau)\|}{\|y(0)\|}$ over all $\gamma \in \mathcal{D}(T, \mu, M, k)$ with $\|\gamma\|_{L^\infty(0, \tau)} \leq 3/\sqrt{k}$, where y is a solution of (3.9) with both $y(0)$ and $y(\tau)$ in the axis y_1 .

By homogeneity reasons we can choose $y_1(0) = -1$. Thus, by enlarging the class where we take γ , Λ is upper-bounded by the solution of the problem

$$\left\{ \begin{array}{l} \text{Find } \sup \frac{1}{\tau} \ln \|y(\tau)\| \text{ with} \\ \tau \in \left[\frac{\pi}{1+h+3k^{3/2}}, T \right], \quad \gamma \in L^\infty([0, \tau], [0, 1]), \\ \dot{y} = \begin{pmatrix} 0 & 1 \\ -3k^{3/2}\gamma(t) & -h \end{pmatrix} y, \quad y(0) = \begin{pmatrix} -1 \\ 0 \end{pmatrix}, \quad y(\tau) \in \left\{ \begin{pmatrix} \xi \\ 0 \end{pmatrix} \mid \xi \in \mathbb{R}_+ \right\}. \end{array} \right. \quad (3.36)$$

The discussion above can be summarized by the following result.

Lemma 3.7. *Let $\Lambda(T, M, k)$ be the solution of Problem (3.36) and take $K_5(T, \mu, M)$ as in Lemma 3.6. If $k > K_5(T, \mu, M)$, then, for every $\gamma \in \mathcal{D}(T, \mu, M, k)$ and for every $I = [t_{n-1}, t_n] \in \mathcal{J}_-$, we have*

$$r(t_n) \leq r(t_{n-1}) e^{\Lambda(T, M, k)(t_n - t_{n-1})}. \quad (3.37)$$

We can now focus on the problem of solving the maximization problem (3.36). We start by proving that the sup is attained.

Lemma 3.8. *Let $k > K_5(T, \mu, M)$ where K_5 is defined as in Lemma 3.6 and we note $\Lambda(T, M, k)$ the solution of Problem (3.36). Then there exist $\tau_\star \in \left[\frac{\pi}{1+h+3k^{3/2}}, T\right]$ and $\gamma_\star \in L^\infty([0, \tau_\star], [0, 1])$ such that, if y_\star denotes the solution of*

$$\dot{y}_\star = \begin{pmatrix} 0 & 1 \\ -3k^{3/2}\gamma_\star(t) & -h \end{pmatrix} y_\star, \quad y_\star(0) = \begin{pmatrix} -1 \\ 0 \end{pmatrix},$$

then $y_\star(\tau) \in \left\{ \begin{pmatrix} \xi \\ 0 \end{pmatrix} \mid \xi \in \mathbb{R}_+ \right\}$ and $\frac{1}{\tau_\star} \ln \|y_\star(\tau_\star)\| = \Lambda(T, M, k)$.

Proof. We start by taking a sequence $(\tau_n, \gamma_n)_{n \in \mathbb{N}}$ with $\tau_n \in \left[\frac{\pi}{1+h+3k^{3/2}}, T\right]$ and $\gamma_n \in L^\infty([0, \tau_n], [0, 1])$ such that, denoting by y_n the solution of

$$\begin{cases} \dot{y}_n = \begin{pmatrix} 0 & 1 \\ -3k^{3/2}\gamma_n(t) & -h \end{pmatrix} y_n, \\ y_n(0) = \begin{pmatrix} -1 \\ 0 \end{pmatrix}, \end{cases} \quad (3.38)$$

we have $\lim_{n \rightarrow +\infty} \frac{1}{\tau_n} \ln \|y_n(\tau_n)\| = \Lambda(T, M, k)$. Up to extending γ_n by 0 outside $[0, \tau_n]$, we can suppose that $\gamma_n \in L^\infty(I, [0, 1])$ where $I = [0, T]$ and thus, by weak- \star compactness of this space and by the compactness of $\left[\frac{\pi}{1+h+3k^{3/2}}, T\right]$, we can find a subsequence of $(\gamma_n)_{n \in \mathbb{N}}$ converging weak- \star to a certain function $\gamma_\star \in L^\infty(I, [0, 1])$ and such that the corresponding subsequence of $(\tau_n)_{n \in \mathbb{N}}$ converges to $\tau_\star \in \left[\frac{\pi}{1+h+3k^{3/2}}, T\right]$. To simplify the notation, we still write $(\gamma_n)_{n \in \mathbb{N}}$ and $(\tau_n)_{n \in \mathbb{N}}$ for these subsequences.

We denote by y_\star the solution of

$$\begin{cases} \dot{y}_\star = \begin{pmatrix} 0 & 1 \\ -3k^{3/2}\gamma_\star(t) & -h \end{pmatrix} y_\star, \\ y_\star(0) = \begin{pmatrix} -1 \\ 0 \end{pmatrix}. \end{cases} \quad (3.39)$$

By considering the solutions y_n of (3.38) to be defined on $[0, T]$ and up to extracting a subsequence, we have $\lim_{n \rightarrow +\infty} y_n = y_\star$ uniformly on $[0, T]$, as it follows from Gronwall's lemma (see [4, Proposition 21] for details).

In particular, $y_\star(\tau_\star) \in \left\{ \begin{pmatrix} \xi & 0 \end{pmatrix}^\top \mid \xi \in \mathbb{R}_+ \right\}$. Moreover, $\frac{1}{\tau_\star} \ln \|y_\star(\tau_\star)\| = \lim_{n \rightarrow +\infty} \frac{1}{\tau_n} \ln \|y_n(\tau_n)\| = \Lambda(T, M, k)$, which completes the proof. \blacksquare

Since the supremum in Problem (3.36) is attained, the Pontryagin Maximum Principle (PMP for short) can be used to characterize the trajectory y_\star reaching the sup of (3.36). For a formulation of the PMP with boundary conditions as those used here, see, for instance, [3, Theorem 7.3].

Lemma 3.9. *Let τ_* , γ_* and y_* be as in the statement of Lemma 3.8. Then $\gamma_*(t)$ is piecewise constant with values in $\{0, 1\}$. Moreover, there exist $s_1, s_2 \in (0, \tau_*)$ with $s_1 \leq s_2$ such that $\gamma_*(t) = 1$ if $t \in [0, s_1) \cup (s_2, \tau_*]$ and $\gamma_*(t) = 0$ if $t \in (s_1, s_2)$. The trajectory y_* is contained in the quadrant $Q_2 = \{(y_1, y_2) \mid y_1 \leq 0, y_2 \geq 0\}$ during the interval $[0, s_1]$ and in the quadrant $Q_1 = \{(y_1, y_2) \mid y_1 \geq 0, y_2 \geq 0\}$ during $[s_2, \tau_*]$.*

Proof. The adjoint vector $p = (p_1, p_2)$ whose existence is guaranteed by the PMP satisfies

$$\begin{cases} \dot{p}_1(t) = 3k^{3/2}\gamma_*(t)p_2(t), \\ \dot{p}_2(t) = hp_2(t) - p_1(t). \end{cases} \quad (3.40)$$

The Hamiltonian is given by $p \begin{pmatrix} 0 & 1 \\ -3k^{3/2}\gamma & -h \end{pmatrix} y = p_1 y_{2*} - 3k^{3/2}\omega p_2 y_{1*} - hp_2 y_{2*}$, and so the maximization condition provided by the PMP writes

$$\gamma_*(t)p_2(t)y_{1*}(t) = \min_{\omega \in [0, 1]} \omega p_2(t)y_{1*}(t). \quad (3.41)$$

Define $\Phi(t) = p_2(t)y_{1*}(t)$ so that (for almost every t such that $\Phi(t) \neq 0$)

$$\gamma_*(t) = \begin{cases} 0 & \text{if } \Phi(t) > 0, \\ 1 & \text{if } \Phi(t) < 0. \end{cases} \quad (3.42)$$

We remark that Φ is absolutely continuous and

$$\dot{\Phi}(t) = hp_2(t)y_{1*}(t) - p_1(t)y_{1*}(t) + p_2(t)y_{2*}(t).$$

Hence $\dot{\Phi}$ is absolutely continuous as well, which shows that Φ is of class \mathcal{C}^1 . Define $Z = \{t \in [0, \tau_*] \mid \Phi(t) = 0\}$. For every connected component $\hat{Z} \subset (0, \tau_*)$ there exists $t \in \hat{Z}$ such that $\dot{\Phi}(t) = 0$. Hence, Φ and $\dot{\Phi}$ both vanish at every density point of Z . Since p never vanishes and

$$\Phi(t) = \begin{pmatrix} p_1(t) & p_2(t) \end{pmatrix} \begin{pmatrix} 0 \\ y_{1*}(t) \end{pmatrix}, \quad \dot{\Phi}(t) = \begin{pmatrix} p_1(t) & p_2(t) \end{pmatrix} \begin{pmatrix} -y_{1*}(t) \\ y_{2*}(t) + hy_{1*}(t) \end{pmatrix},$$

we can conclude that Z does not have density points. Therefore, the zeros of Φ are isolated and $\gamma_*(t)$ is piecewise constant with values in $\{0, 1\}$ (up to a modification on a set of measure zero).

In order to conclude the proof of the lemma we adapt the techniques developed in [2] for the analysis of time-optimal two-dimensional control problems. We start by defining the matrices

$$F = \begin{pmatrix} 0 & 1 \\ 0 & -h \end{pmatrix}, \quad G = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix},$$

so that $\begin{pmatrix} 0 & 1 \\ -3k^{3/2}\gamma & -h \end{pmatrix} y = Fy - 3k^{3/2}\gamma Gy$ and

$$\Phi(t) = p(t)Gy_*(t), \quad \dot{\Phi}(t) = p(t)[G, F]y_*(t),$$

where $[G, F] = GF - FG$ is the commutator of the matrices G and F . We define the functions

$$\Delta_A(y) = \det(Fy, Gy) = y_1 y_2, \quad \Delta_B(y) = \det(Gy, [G, F]y) = y_1^2.$$

Outside $\Delta_A^{-1}(0)$ the vectors Fy and Gy are linearly independent. Hence, for every $y \in \mathbb{R}^2 \setminus \Delta_A^{-1}(0)$ there exist $f_S(y), g_S(y) \in \mathbb{R}$ such that $[G, F]y = f_S(y)Fy + g_S(y)Gy$. We have $\Delta_B(y) = f_S(y) \det(Gy, Fy) = -f_S(y)\Delta_A(y)$, which shows that

$$f_S(y) = -\frac{\Delta_B(y)}{\Delta_A(y)} = -\frac{y_1}{y_2}.$$

We now want to characterize the times at which γ_* switches between 0 and 1. We take an open time interval J during which y_* is outside the axes and we assume that γ_* switches at $t_* \in J$. Equation (3.42) and the continuity of Φ show that $\Phi(t_*) = 0$. The discussion above also shows that

$$\dot{\Phi}(t_*) = p(t_*)[G, F]y_*(t_*) = f_S(y_*(t_*))p(t_*)Fy_*(t_*) \neq 0. \quad (3.43)$$

By the PMP, the Hamiltonian

$$t \mapsto p(t)Fy_*(t) - 3k^{3/2}\gamma_*(t)p(t)Gy_*(t) \quad (3.44)$$

is constant almost everywhere and equal to $\frac{\lambda_0}{\tau_*^2} \ln \|y_*(\tau_*)\|$ for some $\lambda_0 \geq 0$. We deduce that $p(t_*)Fy_*(t_*) > 0$ and that the signs of $\dot{\Phi}(t_*)$ and $f_S(y_*(t_*))$ coincide. Hence, at most one switch may happen on J , from 1 to 0 if the trajectory lies in Q_1 and from 0 to 1 if it lies in Q_2 .

Let us focus on what happens on the axes. Starting from $y_*(0) = (-1, 0)^T$, the choice of $\gamma_*(t) = 0$ cannot maximize the cost. Hence the trajectory enters in Q_2 and $\gamma_*(t) = 1$ in a right-neighborhood of 0. Moreover, it exits Q_2 through the y_2 -axis. Since both vector fields corresponding to $\gamma = 0$ and $\gamma = 1$ are transversal to the positive semi-axis y_2 and point towards Q_1 , there exists a unique s_* such that $y_*(s_*)$ is in the y_2 -axis.

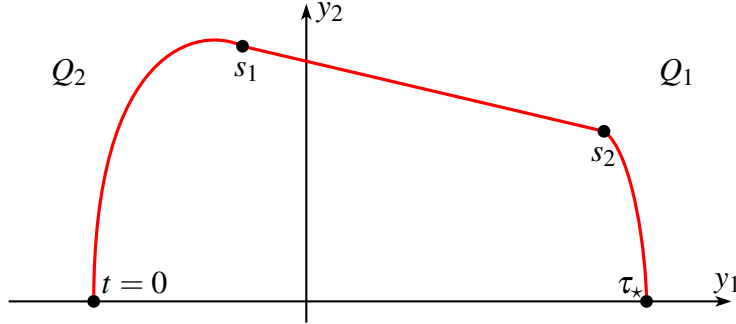


Figure 3.1: Representation of the solution y_* . As stated in Lemma 3.9, y_* is a solution of (3.39) with $\gamma_*(t) = 1$ on $[0, s_1)$, $\gamma_*(t) = 0$ on (s_1, s_2) and $\gamma_*(t) = 1$ on $(s_2, \tau_*]$. The solution y_* lies on Q_2 on $[0, s_1]$ and on Q_1 on $[s_2, \tau_*]$.

Finally, we remark that the trajectories of the vector field corresponding to $\gamma = 0$ never reach the y_1 -axis in finite time unless they start on it. Therefore, either γ_* is identically equal to 1 or it switches twice, once from 1 to 0 in Q_2 and then from 0 to 1 in Q_1 (see Figure 3.1). ■

Lemma 3.9 reduces the optimization problem (3.36) into a maximization over the two scalar parameters s_1 and s_2 . A bound on the maximal value of such problem is given by the following lemma.

Lemma 3.10. *Let $K_5(T, \mu, M)$ be as in Lemma 3.6. There exists $K_6(M)$ such that, if $k > K_5(T, \mu, M)$ and $k > K_6(M)$, then*

$$\mathbf{\Lambda}(T, M, k) \leq \sqrt{3}k^{3/4}. \quad (3.45)$$

Proof. We assume that $k > K_5(T, \mu, T)$ and we take τ_* , γ_* and y_* as in Lemma 3.8. We then have $\mathbf{\Lambda}(T, M, k) = \frac{1}{\tau_*} \ln \|y_*(\tau_*)\|$. Let s_1 and s_2 be as in Lemma 3.9. Then, in the interval $[0, s_1]$, we have $\gamma_*(t) = 1$ and thus y_* satisfies

$$\dot{y}_* = \begin{pmatrix} 0 & 1 \\ -3k^{3/2} & -h \end{pmatrix} y_*, \quad y_*(0) = \begin{pmatrix} -1 \\ 0 \end{pmatrix}. \quad (3.46)$$

We now take

$$k > \frac{M^2}{36} \quad (3.47)$$

so that $3k^{3/2} > h^2/4$ and we can thus define the positive quantity $\omega = \sqrt{3k^{3/2} - h^2/4}$. A direct computation shows that the solution of (3.46) is

$$y_{1*}(t) = -e^{-\frac{h}{2}t} \left(\cos \omega t + \frac{h}{2\omega} \sin \omega t \right), \quad (3.48a)$$

$$y_{2*}(t) = \left(\omega + \frac{h^2}{4\omega} \right) e^{-\frac{h}{2}t} \sin \omega t. \quad (3.48b)$$

In the interval $[s_1, s_2]$, we have $\gamma_*(t) = 0$ and then y_* satisfies

$$\dot{y}_* = \begin{pmatrix} 0 & 1 \\ 0 & -h \end{pmatrix} y_*,$$

which yields the solution

$$y_{1*}(t) = \frac{1}{h} \left(1 - e^{-h(t-s_1)} \right) y_{2*}(s_2) + y_{1*}(s_1), \quad (3.49a)$$

$$y_{2*}(t) = e^{-h(t-s_1)} y_{2*}(s_1). \quad (3.49b)$$

Finally, in the interval $[s_2, \tau_*]$, we have $\gamma_*(t) = 1$ and thus the differential equation satisfied by y_* is the same as in (3.46), but we now consider the boundary condition $y_*(\tau_*) = \begin{pmatrix} \xi \\ 0 \end{pmatrix}$ with $\xi > 0$.

This yields the solution

$$y_{1*}(t) = \xi e^{-\frac{h}{2}(t-\tau_*)} \left(\cos \omega(t-\tau_*) + \frac{h}{2\omega} \sin \omega(t-\tau_*) \right), \quad (3.50a)$$

$$y_{2*}(t) = -\xi \left(\omega + \frac{h^2}{4\omega} \right) e^{-\frac{h}{2}t} \sin \omega(t-\tau_*). \quad (3.50b)$$

We have

$$\mathbf{\Lambda}(T, M, k) = \frac{1}{\tau_*} \ln \xi. \quad (3.51)$$

To simplify the notation, we write $\sigma = s_2 - s_1$. Using the parametrization of the solution given above and imposing that the curves given in (3.49) and (3.50) coincide at the point s_2 , we get

$$\xi e^{\frac{h}{2}(\tau_* - s_2)} \sin \omega(\tau_* - s_2) = \frac{e^{-h\sigma}}{\omega + \frac{h^2}{4\omega}} y_{2*}(s_1), \quad (3.52a)$$

$$\xi e^{\frac{h}{2}(\tau_* - s_2)} \cos \omega(\tau_* - s_2) = y_{1*}(s_1) + y_{2*}(s_1) \left[\frac{1}{h} \left(1 - e^{-h\sigma} \right) + \frac{he^{-h\sigma}}{2\omega^2 + h^2/4} \right]. \quad (3.52b)$$

We can thus express ξ in terms of s_1 , σ and τ_* , and rewrite (3.51) as

$$\mathbf{\Lambda}(T, M, k) = \frac{-h(\tau_* - s_2) + \ln \left[\left(y_{1*}(s_1) + y_{2*}(s_1) \left[\frac{1}{h} (1 - e^{-h\sigma}) + \frac{he^{-h\sigma}}{2\omega^2 + h^2/4} \right] \right)^2 + \left(\frac{e^{-h\sigma} y_{2*}(s_1)}{\omega + \frac{h^2}{4\omega}} \right)^2 \right]}{2[s_1 + \sigma + (\tau_* - s_2)]}. \quad (3.53)$$

To give a bound on $\mathbf{\Lambda}(T, M, k)$, we first use that $-h(\tau_* - s_2) \leq 0$ and $\tau_* - s_2 \geq 0$. By the expression (3.48b) of y_{2*} in $[0, s_1]$, we get the estimate

$$\frac{e^{-h\sigma} y_{2*}(s_1)}{\omega + \frac{h^2}{4\omega}} \leq \sin \omega s_1.$$

We also have that $y_{1*}(s_2) \geq 0$ and $y_{2*}(s_2) \geq 0$, and then (3.50) implies that $\sin \omega(\tau_* - s_2) \geq 0$ and $\cos \omega(\tau_* - s_2) \geq 0$. Equation (3.52b) implies that

$$y_{1*}(s_1) + y_{2*}(s_1) \left[\frac{1}{h} \left(1 - e^{-h\sigma} \right) + \frac{he^{-h\sigma}}{2\omega^2 + h^2/4} \right] \geq 0.$$

Recall, moreover, that Lemma 3.9 implies that $y_{1*}(s_1) \leq 0$. We also have $\frac{1}{h}(1 - e^{-h\sigma}) \leq \sigma$ and, by (3.48b), we obtain

$$y_{2*}(s_1) \frac{he^{-h\sigma}}{2\omega^2 + h^2/4} \leq \frac{h}{2\omega} \sin \omega s_1.$$

We bound $y_{2*}(s_1)$ from above by $\left(\omega + \frac{h^2}{4\omega} \right) \sin \omega s_1$ and, by combining all the previous estimates, we obtain

$$\mathbf{\Lambda}(T, M, k) \leq \frac{\ln(\sin^2 \omega s_1) + \ln \left[1 + \left(\sigma \left(\omega + \frac{h^2}{4\omega} \right) + \frac{K}{2\omega} \right)^2 \right]}{2(s_1 + \sigma)}.$$

We may also suppose that $k > \frac{M^2}{9}$, which in particular implies (3.47), and, in this case, we have $\frac{h}{2\omega} \leq 1$ and $\omega + \frac{h^2}{4\omega} \leq 2\omega$, which finally yields

$$\mathbf{\Lambda}(T, M, k) \leq \frac{\ln(\sin^2 \omega s_1) + \ln \left[1 + (2\omega\sigma + 1)^2 \right]}{2(s_1 + \sigma)}.$$

We now define $s' = \omega s_1$, $\sigma' = \omega\sigma$, and then we have

$$\mathbf{\Lambda}(T, M, k) \leq \omega \frac{\ln(\sin^2 s') + \ln \left[1 + (2\sigma' + 1)^2 \right]}{2(s' + \sigma')}.$$

A direct computation shows that the function

$$(s', \sigma') \mapsto \frac{\ln(\sin^2 s') + \ln \left[1 + (2\sigma' + 1)^2 \right]}{2(s' + \sigma')}$$

is upper bounded over $(\mathbb{R}_+^*)^2$ by 1, and, by bounding ω by $\sqrt{3}k^{3/4}$, we obtain the desired estimate (3.45) under the hypothesis $k > \max(K_5(T, \mu, M), K_6(M))$ with $K_6(M) = M^2/9$. ■

By combining this result with Lemma 3.7, we obtain the desired estimate on the growth of y .

Corollary 3.11. *Let $K_5(T, \mu, M)$ be as in Lemma 3.6 and $K_6(M)$ as in Lemma 3.10. If $k > K_5(T, \mu, M)$ and $k > K_6(M)$, then, for every $\gamma \in \mathcal{D}(T, \mu, M, k)$ and $I = [t_{n-1}, t_n] \in \mathcal{J}_-$, the solution r of (3.13b) satisfies*

$$r(t_n) \leq r(t_{n-1})e^{\sqrt{3}k^{3/4}(t_n - t_{n-1})}.$$

3.3.6 Estimate of y

Now that we estimated the growth of y on intervals of the classes \mathcal{J}_+ and \mathcal{J}_- , we only have to join these results in order to estimate the growth of y over an interval $[0, t]$.

Lemma 3.12. *There exists $K_7(T, \mu, M)$ such that, for $k > K_7(T, \mu, M)$, there exists a constant C depending only on T, M and k such that, for every signal $\alpha \in \mathcal{D}(T, \mu, M)$, every solution y of (3.9) and every $t \in \mathbb{R}_+$, we have*

$$\|y(t)\| \leq C \|y(0)\| e^{2k^{3/4}t}. \quad (3.54)$$

Proof. We suppose that $k > K_i$ for $i = 1, \dots, 6$ in order to be in measure to apply all the previous results. Let us fix $\alpha \in \mathcal{D}(T, \mu, M)$ and $t \in \mathbb{R}_+$.

Since the sequence $(t_n)_{n \in \mathbb{N}}$ defined in (3.21) tends monotonically to $+\infty$ as $n \rightarrow +\infty$, then there exists $N \in \mathbb{N}$ such that $t \in [t_{N-1}, t_N)$ (with the convention that $t_{-1} = 0$). We can use Lemma 3.5 and Corollary 3.11 to estimate the growth of y in each interval I_n , $n = 1, \dots, N-1$. The length of the two intervals $I_0 = [0, t_0]$ and $[t_{N-1}, t]$ is bounded by T , since, as proved in Lemma 3.2, $\theta(s+T) - \theta(s) \leq -2\pi$ for every $s \in \mathbb{R}_+$. By Equation (3.13c), we have $\frac{d}{dt} \ln r \leq k^2 + h + 1$, and then

$$r(t_0) \leq r(0)e^{T(k^2+h+1)}, \quad r(t) \leq r(t_{N-1})e^{T(k^2+h+1)}.$$

We now combine these two results with (3.25) and (3.45), which yields

$$r(t) \leq e^{2T(k^2+h+1)} r(0) \left(\prod_{\substack{n=1 \\ I_n \in \mathcal{J}_+}}^{N-1} e^{4Mk^{1/2}(t_n - t_{n-1})} \right) \left(\prod_{\substack{n=1 \\ I_n \in \mathcal{J}_-}}^{N-1} e^{\sqrt{3}k^{3/4}(t_n - t_{n-1})} \right) \leq Cr(0)e^{\sqrt{3}k^{3/4}t + 4Mk^{1/2}t}$$

with $C = e^{2T(k^2+h+1)}$, which depends only on T, k and M (through h). It suffices to take k large enough, and more precisely $k \geq \left(\frac{4M}{2-\sqrt{3}} \right)^4$, in order to obtain (3.54). We then take K_7 as the maximum among K_i , $i = 1, \dots, 6$, and $\left(\frac{4M}{2-\sqrt{3}} \right)^4$ and the proof is concluded. ■

3.4 Proof of Theorem 3.1

By combining (3.54) and the relation (3.6) between x and y , we can prove Theorem 3.1 .

Proof of Theorem 3.1. Let λ be a real constant. We take $k > K_7(T, \mu, M)$ and we consider the feedback gain $K = (k^2 \quad k)$. By (3.6), for every $t \in \mathbb{R}_+$, we have

$$\|x(t)\| \leq e^{-\frac{k}{2} \int_0^t \alpha(s) ds + \frac{h}{2} t} \left(1 + \frac{h}{2} + \frac{k}{2}\right) \|y(t)\|, \quad \|y(t)\| \leq e^{\frac{k}{2} \int_0^t \alpha(s) ds - \frac{h}{2} t} \left(1 + \frac{h}{2} + \frac{k}{2}\right) \|x(t)\|,$$

and then, in particular, $\|y(0)\| \leq \left(1 + \frac{h}{2} + \frac{k}{2}\right) \|x(0)\|$. Thus, by combining these with (3.54), we obtain that $\|x(t)\| \leq C' \|x(0)\| e^{-\frac{k}{2} \int_0^t \alpha(s) ds + \frac{h}{2} t + 2k^{3/4} t}$ where C' is a constant depending only on k , M and T . We now use $\int_0^t \alpha(s) ds \geq \frac{\mu}{T} t - \mu$ to obtain $\|x(t)\| \leq \bar{C} \|x(0)\| e^{(-\frac{k}{2} \frac{\mu}{T} + \frac{h}{2} + 2k^{3/4}) t}$ for a new constant \bar{C} , which depends on k , M , T and μ . There exists $K(T, \mu, M, \lambda)$ such that, for $k > K(T, \mu, M, \lambda)$, we have $-\frac{k}{2} \frac{\mu}{T} + \frac{h}{2} + 2k^{3/4} \leq -\lambda$ and then $\|x(t)\| \leq \bar{C} \|x(0)\| e^{-\lambda t}$. This concludes the proof, since $\limsup_{t \rightarrow +\infty} \frac{\ln \|x(t)\|}{t} \leq -\lambda$. ■

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