

Lyapunov-Stable Adaptive Stabilization of Nonlinear Systems with Matched Uncertainty

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1. INTRODUCTION

Adaptive stabilization of linear time-invariant plants with full state feedback has been considered in [1–4] using Lyapunov-based gradient update laws. Lyapunov-based adaptive stabilization has several extensions to nonlinear systems. In [5], a variation of the controller presented in [4] is shown to stabilize a class of scalar second-order nonlinear systems with partial-state-dependent uncertainty. In particular, the adaptive controller of [5] can stabilize the scalar nonlinear system $m\ddot{q}(t) + g(q(t))\dot{q}(t) + f(q(t))q(t) = bu(t)$, where the functions $f(\cdot)$ and $g(\cdot)$ are lower bounded but otherwise unknown. In the present paper, a novel full-state-feedback adaptive controller is used to stabilize n th-order nonlinear systems with bounded state-dependent uncertainty. First, we develop the controller for linear systems, then extended the result to nonlinear systems.

2. ADAPTIVE STABILIZATION FOR LINEAR SYSTEMS

We consider the single-input linear-time invariant system

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

where $A \in \mathbb{R}^{n \times n}$ and $B \in \mathbb{R}^{n \times 1}$. We make the following assumptions.

(i) The system is in companion form, where

$$A \triangleq \begin{bmatrix} -a_{n-1} & -a_{n-2} & \cdots & -a_1 & -a_0 \\ 1 & 0 & & 0 & 0 \\ 0 & 1 & & 0 & 0 \\ \vdots & & \ddots & & \vdots \\ 0 & 0 & & 1 & 0 \end{bmatrix}, \quad (2.2)$$

$$B \triangleq [b \ 0 \ \cdots \ 0]^T. \quad (2.3)$$

(ii) $b \neq 0$, and $\text{sgn}(b)$ is known.

(iii) The full state x is available for feedback.

We begin this section by providing several useful results regarding a matrix in controllable canonical form. Consider the matrix $A \in \mathbb{R}^{n \times n}$, which has the characteristic polynomial $d(s) \triangleq s^n + d_{n-1}s^{n-1} + d_{n-2}s^{n-2} + \cdots + d_1s + d_0$ and is in the companion form (2.2). The Hurwitz matrix associated with the characteristic polynomial $d(s)$ is

$$H \triangleq \begin{bmatrix} d_{n-1} & 1 & 0 & 0 & \cdots & 0 & 0 \\ d_{n-3} & d_{n-2} & d_{n-1} & 1 & \cdots & 0 & 0 \\ d_{n-5} & d_{n-4} & d_{n-3} & d_{n-2} & \cdots & 0 & 0 \\ d_{n-7} & d_{n-6} & d_{n-5} & d_{n-4} & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & d_1 & d_2 \\ 0 & 0 & 0 & 0 & \cdots & 0 & d_0 \end{bmatrix}. \quad (2.4)$$

The following result, given in [6], concerns the solution to the Lyapunov equation for a matrix in controllable canonical form.

Lemma 2.1. Consider the asymptotically stable controllable canonical form A . Let $P \in \mathbb{R}^{n \times n}$ be the positive-definite solution to the Lyapunov equation $A^T P + PA = -Q$, where $Q \in \mathbb{R}^{n \times n}$ is positive definite. Let p_1 denote the first column of P . Then p_1 satisfies $2DHDp_1 = q$, where H is given by (2.4), $D \triangleq \text{diag}(1, -1, 1, \dots)$, and

$$q \triangleq \begin{bmatrix} \sum_{1 \leq i, j \leq n, i+j=2} (-1)^{i-1} Q_{i,j} \\ \sum_{1 \leq i, j \leq n, i+j=4} (-1)^{i-2} Q_{i,j} \\ \vdots \\ \sum_{1 \leq i, j \leq n, i+j=2n} (-1)^{i-n} Q_{i,j} \end{bmatrix}. \quad (2.5)$$

Lemma 2.2. Let $g(s) \triangleq g_{n-1}s^{n-1} + g_{n-2}s^{n-2} + \cdots + g_0$ be a Hurwitz polynomial where $g_{n-1} > 0$ and define

$$A_s(k) \triangleq \begin{bmatrix} -k|b|g_{n-1} & -k|b|g_{n-2} & \cdots & -k|b|g_0 \\ 1 & 0 & & 0 \\ 0 & 1 & & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 0 & & 0 \end{bmatrix}, \quad (2.6)$$

where $k \in \mathbb{R}$. Then, there exists $k_s > 0$ such that, for all $k \geq k_s$, $A_s(k)$ is asymptotically stable and thus, there exists a positive definite $P(k)$ such that

$$A_s^T(k)P(k) + P(k)A_s(k) = -e^{-\alpha k}Q, \quad (2.7)$$

where $Q > 0$ and $\alpha \geq 0$. Furthermore, $\lim_{k \rightarrow \infty} p_1(k) = 0$ and $\lim_{k \rightarrow \infty} ke^{\alpha k} p_1(k)$ exists, where $p_1(k)$ denotes the first column of $P(k)$. If, in addition, $\alpha > 0$, then there exists $k_2 \geq k_s$ such that, for all $k \geq k_2$, $\frac{\partial P(k)}{\partial k}$ is negative definite.

Proof. Let $H(k)$ be the Hurwitz matrix associated with the characteristic polynomial of $A_s(k)$. The Hurwitz stability conditions for the characteristic polynomial of $A_s(k)$ are polynomials in k given by

$$\Lambda_1 \triangleq k|b|g_{n-1} > 0, \quad (2.8)$$

$$\Lambda_2 \triangleq \begin{vmatrix} k|b|g_{n-1} & 1 \\ k|b|g_{n-3} & k|b|g_{n-2} \end{vmatrix} > 0, \quad (2.9)$$

$$\Lambda_3 \triangleq \begin{vmatrix} k|b|g_{n-1} & 1 & 0 \\ k|b|g_{n-3} & k|b|g_{n-2} & k|b|g_{n-1} \\ k|b|g_{n-5} & k|b|g_{n-4} & k|b|g_{n-3} \end{vmatrix} > 0, \quad (2.10)$$

$$\vdots$$

$$\Lambda_n \triangleq \begin{vmatrix} \Lambda_3 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & k|b|g_2 \\ & & & k|b|g_0 \end{vmatrix} > 0. \quad (2.11)$$

For sufficiently large k , the Hurwitz conditions are satisfied since $g(s)$ is Hurwitz with positive leading coefficient. Therefore, there exists $k_s > 0$ such that, for all $k \geq k_s$, the matrix $A_s(k)$ is asymptotically stable. Then, for all $k \geq k_s$, there exists $P(k) > 0$ satisfying (2.7).

Now, we consider the asymptotic properties of $p_1(k)$. For all $k \geq k_s$, the inverse of the Hurwitz matrix exists and can be expressed as $H^{-1}(k) = \frac{1}{\det(H(k))} \hat{H}(k)$, where

$$\hat{H}(k) \triangleq \begin{bmatrix} [H(k)]_{1,1} & -[H(k)]_{2,1} & \cdots & (-1)^{n+1} [H(k)]_{n,1} \\ -[H(k)]_{1,2} & [H(k)]_{2,2} & & \\ \vdots & & \ddots & \vdots \\ (-1)^{n+1} [H(k)]_{1,n} & \cdots & & [H(k)]_{n,n} \end{bmatrix}. \quad (2.12)$$

where $[H(k)]_{i,j}$ is the (i,j) th minor of $H(k)$. The determinant of $H(k)$ is a degree n polynomial in k , while $[H(k)]_{i,j}$ is a polynomial in k of degree not exceeding $n-1$. Therefore, $\lim_{k \rightarrow \infty} H^{-1}(k) = 0$, and $\lim_{k \rightarrow \infty} kH^{-1}(k)$ exists. Using Lemma 2.1 we obtain $\lim_{k \rightarrow \infty} p_1(k) = \lim_{k \rightarrow \infty} \frac{1}{2} D^{-1} H^{-1}(k) D^{-1} e^{-\alpha k} q = 0$ and $\lim_{k \rightarrow \infty} k e^{\alpha k} p_1(k) = \lim_{k \rightarrow \infty} \frac{1}{2} D^{-1} k H^{-1}(k) D^{-1} q$ exists, where q is determined from Q using (2.5).

Next, we show that for k sufficiently large, $\frac{\partial P(k)}{\partial k} < 0$. Taking the partial derivative of (2.7) with respect to k yields $-A_s^T(k) \frac{\partial P(k)}{\partial k} - \frac{\partial P(k)}{\partial k} A_s(k) = \hat{Q}(k)$ where

$$\hat{Q}(k) \triangleq -\alpha e^{-\alpha k} Q + \delta e_1^T P(k) + P(k) e_1 \delta^T, \quad (2.13)$$

$e_1 \triangleq [1 \ 0 \ \cdots \ 0]^T \in \mathbb{R}^n$, and $\delta \triangleq -|b| [g_{n-1} \ \cdots \ g_0]^T \in \mathbb{R}^n$. Define $\Omega(k) \triangleq \sqrt{\frac{\alpha}{2}} e^{-\frac{\alpha}{2} k} Q^{\frac{1}{2}} - \sqrt{\frac{2}{\alpha}} e^{\frac{\alpha}{2} k} Q^{-\frac{1}{2}} P(k) e_1 \delta^T$, and since $0 \leq \Omega^T(k) \Omega(k)$, it follows that $\delta e_1^T P(k) + P(k) e_1 \delta^T \leq \frac{\alpha}{2} e^{-\alpha k} Q + \frac{2}{\alpha} e^{\alpha k} \delta e_1^T P(k) Q^{-1} P(k) e_1 \delta^T$. Combining this with (2.13) yields

$$\hat{Q}(k) \leq -e^{-\alpha k} \left[\frac{\alpha}{2} Q - \frac{2}{\alpha} e^{2\alpha k} \delta p_1^T(k) Q^{-1} p_1(k) \delta^T \right]. \quad (2.14)$$

Since $\lim_{k \rightarrow \infty} k e^{\alpha k} p_1(k)$ exists, it follows that $\lim_{k \rightarrow \infty} e^{\alpha k} p_1(k) = 0$, and thus $\lim_{k \rightarrow \infty} \frac{2}{\alpha} e^{2\alpha k} \delta p_1^T(k) Q^{-1} p_1(k) \delta^T = 0$. Let $k_2 \geq k_s$ be such that, for all $k \geq k_2$, $\frac{2}{\alpha} e^{2\alpha k} \delta p_1^T(k) Q^{-1} p_1(k) \delta^T < \frac{\alpha}{2} Q$, and thus $\hat{Q}(k) < 0$. Then it follows that, for all $k \geq k_2$, $-A_s^T(k) \frac{\partial P(k)}{\partial k} - \frac{\partial P(k)}{\partial k} A_s(k) < 0$. Since $A_s(k)$ is asymptotically stable, $\frac{\partial P(k)}{\partial k} < 0$. \square

Now, we present a Lyapunov proof of a high-gain stabilizing controller for the linear system (2.1)-(2.3).

Lemma 2.3. Consider the linear system (2.1)-(2.3). Let $g(s)$ be the Hurwitz polynomial

$$g(s) \triangleq g_{n-1} s^{n-1} + g_{n-2} s^{n-2} + \cdots + g_0, \quad (2.15)$$

where $g_{n-1} > 0$. Define $G \triangleq [g_{n-1} \ g_{n-2} \ \cdots \ g_0]$, and consider the feedback

$$u(t) = -\text{sgn}(b) k G x(t), \quad (2.16)$$

where $k \in \mathbb{R}$. Then there exists $k_s > 0$ such that, for all $k \geq k_s$, the origin of the closed-loop system is asymptotically stable.

Proof. The system (2.1)-(2.3) with the feedback (2.16) can be written as $\dot{x}(t) = [A_s(k) + \Delta] x(t)$, where $A_s(k)$ is given by (2.6), $\Delta \triangleq [\delta \ 0_{n \times (n-1)}]^T \in \mathbb{R}^{n \times n}$, and $\delta \triangleq [-a_{n-1} \ \cdots \ -a_0]^T \in \mathbb{R}^n$. Lemma 2.2 implies that there exists $k_1 > 0$ such that, for all $k \geq k_1$, $A_s(k)$ is asymptotically stable. For all $k \geq k_1$, let $P(k) > 0$ be the solution to the Lyapunov equation $A_s^T(k) P(k) + P(k) A_s(k) = -(Q + I)$, where $Q > 0$. Furthermore, let $p_1(k)$ denote the first column of $P(k)$. Next, consider the Lyapunov candidate $V(x) = x^T P(k) x$, where $k \geq k_1$. Taking the derivative along the closed-loop trajectory yields

$$\begin{aligned} \dot{V}(x) &= -x^T (Q + I) x \\ &\quad + x^T [\Delta^T P(k) + P(k) \Delta] x. \end{aligned} \quad (2.17)$$

Since $0 \leq \left(\sqrt{2} P(k) \Delta - \frac{1}{\sqrt{2}} I \right)^T \left(\sqrt{2} P(k) \Delta - \frac{1}{\sqrt{2}} I \right)$, it follows that $\Delta^T P(k) + P(k) \Delta \leq \frac{1}{2} I + 2 \Delta^T P^2(k) \Delta$. Combining this with (2.17) yields $\dot{V}(x) \leq -x^T Q x - \frac{1}{2} x^T x + 2x^T [\delta p_1^T(k) p_1(k) \delta^T] x$. Since Lemma 2.2 implies $p_1(k) \rightarrow 0$ as $k \rightarrow \infty$, let $k_s \geq k_1 > 0$ be such that, for all $k \geq k_s$, $\delta p_1^T(k) p_1(k) \delta^T \leq \frac{1}{4} I$. Therefore, for all $k \geq k_s$, $\dot{V}(x) \leq -x^T Q x < 0$ for $x \neq 0$, and the origin is asymptotically stable. \square

Now, we present a Lyapunov-stable adaptive stabilization algorithm for linear systems.

Theorem 2.1. Consider the linear system (2.1)-(2.3). Let $g(s)$ be the Hurwitz polynomial (2.15) where $g_{n-1} > 0$. Define $G \triangleq [g_{n-1} \ g_{n-2} \ \cdots \ g_0]$, and consider the adaptive feedback controller

$$u(t) = -\text{sgn}(b) k(t) G x(t), \quad (2.18)$$

$$\dot{k}(t) = e^{-\alpha k(t)} x^T(t) R x(t), \quad (2.19)$$

where R is positive definite and $\alpha > 0$. Then, there exists $k_s > 0$, such that for all $k_e \geq k_s$, the equilibrium solution $(0, k_e)$ of the closed-loop system (2.1)-(2.3) and (2.18)-(2.19) is Lyapunov stable. Furthermore, for all initial conditions $x(0)$ and $k(0)$, $k_\infty \triangleq \lim_{t \rightarrow \infty} k(t)$ exists and $\lim_{t \rightarrow \infty} x(t) = 0$.

Proof. The dynamics (2.1)-(2.3) with the feedback (2.18) can be expressed as

$$\dot{x}(t) = \tilde{A}(k(t)) x(t), \quad (2.20)$$

where $\tilde{A}(k(t)) \triangleq A - k(t) \text{sgn}(b) B G$. Lemma 2.3 implies that there exists k_s such that for all $k \geq k_s$, $\tilde{A}(k)$ is asymptotically stable. Let $k_e \geq k_s$, define $A_e \triangleq \tilde{A}(k_e)$, and

define $\tilde{k}(t) \triangleq k_e - k(t)$ so that (2.19)-(2.20) can be written as $\dot{x} = A_e x + \text{sgn}(b)\tilde{k}BGx$. Since A_e is asymptotically stable, there exists $P_e > 0$ such that $A_e^T P_e + P_e A_e = -Q$, where $Q > 0$. Next, consider the Lyapunov candidate $V(x, \tilde{k}) \triangleq x^T P_e x + \tilde{k}^2$, where $V : \mathbb{R}^n \times \mathcal{D} \rightarrow [0, \infty)$ and the domain \mathcal{D} will be determined. The derivative of $V(x, \tilde{k})$ along a closed-loop trajectory is

$$\begin{aligned} \dot{V}(x, \tilde{k}) &= \tilde{k}x^T [\text{sgn}(b)G^T B^T P_e + \text{sgn}(b)P_e BG] x \\ &\quad - x^T Qx - 2\tilde{k}e^{-\alpha k} x^T R x. \end{aligned} \quad (2.21)$$

First, consider the case $\tilde{k} \geq 0$, then $\dot{V}(x, \tilde{k}) \leq -x^T Qx + \tilde{k}\sigma_1 x^T x$, where $\sigma_1 \triangleq \lambda_{\max}(\text{sgn}(b)G^T B^T P_e + \text{sgn}(b)P_e BG)$. If $\sigma_1 \leq 0$, then $\dot{V}(x, \tilde{k}) \leq -x^T Qx$. If $\sigma_1 > 0$, then let $0 < \varepsilon < \frac{\lambda_{\min}(Q)}{\sigma_1}$. Thus, for all \tilde{k} such that $0 \leq \tilde{k} \leq \frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_1}$, $\dot{V}(x, \tilde{k}) \leq -\varepsilon x^T x$.

Now, consider the case $\tilde{k} \leq 0$, then $\dot{V}(x, \tilde{k}) \leq -x^T Qx + \tilde{k}\sigma_2 x^T x$, where $\sigma_2 \triangleq \lambda_{\min}(\text{sgn}(b)G^T B^T P_e + \text{sgn}(b)P_e BG - 2R)$. If $\sigma_2 \geq 0$, then $\dot{V}(x, \tilde{k}) \leq -x^T Qx$. If $\sigma_2 < 0$, then, for all \tilde{k} such that $\frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_2} \leq \tilde{k} \leq 0$, $\dot{V}(x, \tilde{k}) \leq -\varepsilon x^T x$. If $\sigma_1 \leq 0$, then $\sigma_2 < 0$, so define the domain

$$\mathcal{D} \triangleq \begin{cases} \{\tilde{k} \in \mathbb{R} : -\infty < \tilde{k} \leq \frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_1}\}, & \sigma_1 > 0, \sigma_2 \geq 0 \\ \{\tilde{k} \in \mathbb{R} : \frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_2} \leq \tilde{k} \leq \frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_1}\}, & \sigma_1 > 0, \sigma_2 < 0 \\ \{\tilde{k} \in \mathbb{R} : \frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_2} \leq \tilde{k} < \infty\}, & \sigma_1 \leq 0, \sigma_2 < 0 \end{cases} \quad (2.22)$$

Thus, for all $x \in \mathbb{R}^n$ and all $\tilde{k} \in \mathcal{D}$, $\dot{V}(x, \tilde{k}) \leq -\varepsilon x^T x$ and the solution $(0, k_e)$ is Lyapunov stable.

Next, we show that $k(t)$ converges. The dynamics (2.1)-(2.3) with the feedback (2.18) can be expressed as

$$\dot{x}(t) = [A_s(k) + \Delta] x(t), \quad (2.23)$$

where $A_s(k)$ is given by (2.6), $\Delta \triangleq [\delta \quad 0_{n \times (n-1)}]^T \in \mathbb{R}^{n \times n}$, and $\delta \triangleq [-a_{n-1} \quad \cdots \quad -a_0]^T \in \mathbb{R}^n$. Lemma 2.2 implies that there exists $k_s > 0$ such that, for all constant $k \geq k_s$, $A_s(k)$ is asymptotically stable. For $k \geq k_s$, define $V_0(x, k) \triangleq x^T P(k)x$, where $P(k) > 0$ satisfies the Lyapunov equation $A_s^T(k)P(k) + P(k)A_s(k) = -e^{\alpha k}R$, and $\alpha > 0$. Taking the derivative of $V_0(x, k)$ along a trajectory of (2.19) and (2.23) yields

$$\begin{aligned} \dot{V}_0(x, k) &= -e^{-\alpha k} x^T R x + \dot{k} x^T \frac{\partial P(k)}{\partial k} x \\ &\quad + x^T [\Delta^T P(k) + P(k)\Delta] x \end{aligned} \quad (2.24)$$

Define $\Omega(k) \triangleq \sqrt{\frac{1}{2}}e^{-\frac{\alpha}{2}k}R^{\frac{1}{2}} - \sqrt{2}e^{\frac{\alpha}{2}k}R^{-\frac{1}{2}}P(k)\Delta$, and since $0 \leq \Omega^T(k)\Omega(k)$, it follows that $\Delta^T P(k) + P(k)\Delta \leq \frac{1}{2}e^{-\alpha k}R + 2e^{\alpha k}\Delta^T P(k)R^{-1}P(k)\Delta$. Combining this with (2.24) yields

$$\begin{aligned} \dot{V}_0(x, k) &\leq -e^{-\alpha k} x^T \left[\frac{1}{2}R - 2e^{2\alpha k}\delta p_1^T(k)R^{-1}p_1(k)\delta^T \right] x \\ &\quad + \dot{k} x^T \frac{\partial P(k)}{\partial k} x. \end{aligned} \quad (2.25)$$

Lemma 2.2 implies that $\lim_{k \rightarrow \infty} ke^{\alpha k}p_1(k)$ exists, and thus $\lim_{k \rightarrow \infty} e^{\alpha k}p_1(k) = 0$. It follows that $\lim_{k \rightarrow \infty} 2e^{2\alpha k}\delta p_1^T(k)R^{-1}p_1(k)\delta^T = 0$, and there exists $k_1 \geq k_s$ such that, for all $k \geq k_1$, $2e^{2\alpha k}\delta p_1^T(k)R^{-1}p_1(k)\delta^T \leq \frac{1}{4}R$. Then, for all $k \geq k_1$, $\dot{V}_0(x, k) \leq -\frac{1}{4}e^{-\alpha k}x^T R x + \dot{k}x^T \frac{\partial P(k)}{\partial k}x$. Lemma 2.2 also implies that there exists $k_2 \geq k_s$ such that, for all $k \geq k_2$, $\frac{\partial P(k)}{\partial k} < 0$. Thus, for all $k \geq k_3 \triangleq \max(k_1, k_2)$, $\dot{V}_0(x, k) \leq -\frac{1}{4}e^{-\alpha k}x^T R x = -\frac{1}{4}\dot{k}$, which implies

$$\dot{V}_0(x, k)dt \leq -\frac{1}{4}dk. \quad (2.26)$$

To prove that $\lim_{t \rightarrow \infty} k(t)$ exists, suppose that $k(t)$ diverges to infinity in either finite or infinite time. Then there exists $t_3 > 0$ such that $k(t_3) = k_3$. Since $k(t)$ does not escape at t_3 , it follows from (2.19) that $x(t)$ does not escape at time t_3 . Let $t > t_3$ be such that $k(\cdot)$ exists on $[t_3, t]$. Integrating (2.26) from t_3 to t and from k_3 to $k(t)$ and solving for $k(t)$ yields $k(t) \leq k(t_3) + 4V_0(x(t_3), k_3) - 4V_0(x(t), k(t)) \leq k(t_3) + 4V_0(x(t_3), k_3)$. Hence $k(t)$ is bounded on $[0, \infty)$, and thus $k(t)$ does not diverge to infinity. Since $k(t)$ is non-decreasing, $k_\infty \triangleq \lim_{t \rightarrow \infty} k(t)$ exists.

Next, we show that $x(t)$ is bounded. Define the function $V_1(x) = x^T x$. Taking the derivative of $V_1(x)$ along a trajectory of (2.19)-(2.20) yields $\dot{V}_1(x, k) = x^T [\tilde{A}^T(k) + \tilde{A}(k)]x$. Since $k(t)$ converges, there exist $\eta > 0$ such that $\dot{V}_1(x, k) \leq \eta x^T R x = \eta e^{\alpha k}\dot{k}$, which implies $\dot{V}_1(x, k)dt \leq \eta e^{\alpha k}dk$. Integrating from 0 to t and from $k(0)$ to $k(t)$ and solving for $V_1(x(t))$ yields $V_1(x(t)) \leq \frac{\eta}{\alpha}e^{\alpha k(t)} + V_1(x(0)) - \frac{\eta}{\alpha}e^{\alpha k(0)}$. Since $k(t)$ is bounded, we conclude that $V_1(x(t))$ is bounded. Thus, $x(t)$ is bounded.

Now, we show that $\lim_{t \rightarrow \infty} x(t) = 0$. Since $\lim_{t \rightarrow \infty} k(t)$ exists, $\tilde{A}(k(t))$ is bounded. The dynamics (2.20) implies $\|\dot{x}(t)\| \leq \|\tilde{A}(k(t))\| \|x(t)\|$ and since $\tilde{A}(k(t))$ and $x(t)$ are bounded, it follows that $\dot{x}(t)$ is bounded. Thus, $\frac{d}{dt}[e^{-\alpha k(t)}x^T(t)R x(t)] = e^{-\alpha k(t)}(-\alpha \dot{k}(t)x^T(t)R x(t) + 2x^T(t)R \dot{x}(t))$ is bounded. Since the derivative of $\dot{k}(t) = e^{-\alpha k(t)}x^T(t)R x(t)$ is bounded, $\dot{k}(t)$ is uniformly continuous. Since $\dot{k}(t)$ is uniformly continuous and $\lim_{t \rightarrow \infty} k(t)$ exists, Barabatal's lemma implies that $\lim_{t \rightarrow \infty} e^{-\alpha k(t)}x^T(t)R x(t) = 0$. Thus, $\lim_{t \rightarrow \infty} x(t) = 0$. \square

3. ADAPTIVE STABILIZATION FOR NONLINEAR SYSTEMS

In this section, we consider adaptive stabilization for the n th order nonlinear system

$$\begin{aligned} \dot{q}^{(n)}(t) &+ m_{n-1}(q^{(n-1)}, \dots, \dot{q}, q)q^{(n-1)}(t) \\ &+ m_{n-2}(q^{(n-1)}, \dots, \dot{q}, q)q^{(n-2)}(t) \\ &+ \cdots + m_1(q^{(n-1)}, \dots, \dot{q}, q)\dot{q}(t) \\ &+ m_0(q^{(n-1)}, \dots, \dot{q}, q)q(t) = bu(t), \end{aligned} \quad (3.1)$$

where, for $i = 0, \dots, n-1$, $m_i : \mathbb{R}^n \rightarrow \mathbb{R}$ and $b \in \mathbb{R}$. We make the following assumptions.

- (i) The functions m_0, \dots, m_{n-1} are locally Lipschitz.
- (ii) The functions m_0, \dots, m_{n-1} are bounded. That is, for $i = 0, \dots, n-1$, there exists $\mu > 0$ such that, for all $q^{(n-1)}, \dots, \dot{q}, q \in \mathbb{R}$, $|m_i(q^{(n-1)}, \dots, \dot{q}, q)| \leq \mu$. The bound μ is unknown.
- (iii) $b \neq 0$, and $\text{sgn}(b)$ is known.
- (iv) The full state $q, \dot{q}, \dots, q^{(n-1)}$ is available for feedback.

The system (3.1) can be written in the state-dependent controllable canonical form

$$\dot{x}(t) = A(x(t))x(t) + Bu(t), \quad (3.2)$$

where

$$A(x) \triangleq \begin{bmatrix} -m_{n-1}(x) & -m_{n-2}(x) & \cdots & -m_0(x) \\ 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 \end{bmatrix}, \quad (3.3)$$

$$B^T \triangleq [b \ 0 \ \cdots \ 0], \quad (3.4)$$

$$x^T \triangleq [q^{(n-1)} \ q^{(n-2)} \ \cdots \ \dot{q} \ q]. \quad (3.5)$$

We now present a nonlinear extension of Lemma 2.3.

Lemma 3.1. Consider the system (3.2)-(3.4). Let $g(s)$ be the Hurwitz polynomial

$$g(s) \triangleq g_{n-1}s^{n-1} + g_{n-2}s^{n-2} + \cdots + g_0, \quad (3.6)$$

where $g_{n-1} > 0$. Define $G \triangleq [g_{n-1} \ g_{n-2} \ \cdots \ g_0]$, and consider the feedback

$$u(t) = -\text{sgn}(b)kGx(t), \quad (3.7)$$

where $k \in \mathbb{R}$. Then there exists $k_s > 0$ such that, for all $k \geq k_s$, the origin of the closed-loop system is globally asymptotically stable.

Proof. The system (3.2)-(3.4) with the feedback (3.7) can be written as

$$\dot{x}(t) = [A_s(k) + \Delta(x(t))]x(t), \quad (3.8)$$

where where $A_s(k)$ is given by (2.6), $\Delta(x) \triangleq [\delta(x) \ 0_{n \times (n-1)}]^T \in \mathbb{R}^{n \times n}$, and $\delta(x) \triangleq [-m_{n-1}(x) \ \cdots \ -m_0(x)]^T \in \mathbb{R}^n$. Lemma 2.2 implies that there exists $k_1 > 0$ such that, for all $k \geq k_1$, the matrix $A_s(k)$ is asymptotically stable. For all $k \geq k_1$, let $P(k) > 0$ be the solution to the Lyapunov equation $A_s^T(k)P(k) + P(k)A_s(k) = -(Q + I)$ where $Q > 0$. Furthermore, let $p_1(k)$ denote the first column of $P(k)$. Lemma 2.2 also provides the asymptotic property $\lim_{k \rightarrow \infty} p_1(k) = 0$. Now, consider the Lyapunov candidate $V(x) \triangleq x^T P(k)x$, where $k \geq k_1$. Taking the derivative along a closed-loop trajectory yields

$$\dot{V}(x) = -x^T(Q + I)x + x^T[\Delta^T(x)P(k) + P(k)\Delta(x)]x. \quad (3.9)$$

Since $0 \leq \Omega(x)^T \Omega(x)$ where $\Omega(x) \triangleq \sqrt{2}P(k)\Delta(x) - \frac{1}{\sqrt{2}}I$, it follows that $\Delta^T(x)P(k) + P(k)\Delta(x) \leq \frac{1}{2}I + 2\Delta^T(x)P^2(k)\Delta(x)$. Combining this with (3.9) yields

$$\dot{V}(x) \leq -x^T Q x - \frac{1}{2}x^T x + 2x^T [\delta(x)p_1^T(k)p_1(k)\delta^T(x)]x. \quad (3.10)$$

Since $p_1(k) \rightarrow 0$ as $k \rightarrow \infty$, let $k_s \geq k_1$ be such that, for all $k \geq k_s$, $p_1^T(k)p_1(k) \leq \left(\frac{1}{4\delta^T \delta}\right)I$, where $\delta^T \triangleq [\mu \ \cdots \ \mu]$. Therefore, for all $k \geq k_s$, $\dot{V}(x) \leq -x^T Q x$. Hence, for all $k \geq k_s$, the origin is globally asymptotically stable. \square

Now we present the main result of this paper, namely, Lyapunov-stable adaptive stabilization of a class of nonlinear systems.

Theorem 3.1. Consider the nonlinear system (3.2)-(3.4). Let $g(s)$ be the Hurwitz polynomial (3.6), where $g_{n-1} > 0$. Define $G \triangleq [g_{n-1} \ g_{n-2} \ \cdots \ g_0]$, and consider the adaptive feedback controller

$$u(t) = -\text{sgn}(b)k(t)Gx(t), \quad (3.11)$$

$$\dot{k}(t) = e^{-\alpha k(t)}x^T(t)Rx(t), \quad (3.12)$$

where R is positive definite and $\alpha > 0$. Then, there exists $k_s > 0$, such that for all $k_e \geq k_s$, the equilibrium solution $(0, k_e)$ of the closed-loop system (3.2)-(3.4) and (3.11)-(3.12) is Lyapunov stable. Furthermore, for all initial conditions $x(0)$ and $k(0)$, $k_\infty \triangleq \lim_{t \rightarrow \infty} k(t)$ exists and $\lim_{t \rightarrow \infty} x(t) = 0$.

Proof. The dynamics (3.2)-(3.4) with the feedback (3.11) can be written as

$$\dot{x}(t) = [A_s(k) + \Delta(x(t))]x(t), \quad (3.13)$$

where $A_s(k)$ is given by (2.6), $\Delta(x) \triangleq [\delta(x) \ 0_{n \times (n-1)}]^T \in \mathbb{R}^{n \times n}$, and $\delta(x) \triangleq [-m_{n-1}(x) \ \cdots \ -m_0(x)]^T \in \mathbb{R}^n$. Lemma 2.2 implies that there exists k_1 such that for all $k \geq k_1$, $A_s(k)$ is asymptotically stable. For all $k \geq k_1$, let $P(k) > 0$ be the solution to the Lyapunov equation $A_s^T(k)P(k) + P(k)A_s(k) = -(Q + I)$, where $Q > 0$. Furthermore, let $p_1(k)$ denote the first column of $P(k)$. Lemma 2.2 also provides the asymptotic property $\lim_{k \rightarrow \infty} p_1(k) = 0$. Since $p_1(k) \rightarrow 0$ as $k \rightarrow \infty$, let $k_s \geq k_1$ be such that, for all $k \geq k_s$, $p_1^T(k)p_1(k) \leq \left(\frac{1}{4\delta^T \delta}\right)I$, where $\delta^T \triangleq [\mu \ \cdots \ \mu]$. Let $k_e \geq k_s$, define $A_e \triangleq A_s(k_e)$, and define $\tilde{k}(t) \triangleq k_e - k(t)$ so that (3.12)-(3.13) can be written as

$$\dot{x} = A_e x + \text{sgn}(b)\tilde{k}BGx + \Delta(x)x. \quad (3.14)$$

Since A_e is asymptotically stable, then there exists $P_e > 0$ such that $A_e^T P_e + P_e A_e = -(Q + I)$, where $Q > 0$. Let p_e denote the first column of P_e . Next, consider the Lyapunov

candidate $V(x, \tilde{k}) \triangleq x^\top P_e x + \tilde{k}^2$, where $V : \mathbb{R}^n \times \mathcal{D} \rightarrow [0, \infty)$ and the domain \mathcal{D} will be determined. The derivative of $V(x, \tilde{k})$ along a closed-loop trajectory is

$$\begin{aligned} \dot{V}(x, \tilde{k}) &= x^\top [A_e^\top P_e + P_e A_e] x - 2\tilde{k}\dot{k} \\ &\quad + \text{sgn}(b)\tilde{k}x^\top G^\top B^\top P_e x + \text{sgn}(b)\tilde{k}x^\top P_e B G x \\ &\quad + x^\top [\Delta^\top(x)P_e + P_e \Delta(x)] \\ &\leq -x^\top (Q + I) x - 2\tilde{k}e^{-\alpha k} x^\top R x \\ &\quad + \tilde{k}x^\top [\text{sgn}(b)G^\top B^\top P_e + \text{sgn}(b)P_e B G] x \\ &\quad + x^\top \left[\frac{1}{2}I + 2\Delta^\top(x)P_e \Delta(x) \right] x \\ &\leq -x^\top (Q + I) x - 2\tilde{k}e^{-\alpha k} x^\top R x \\ &\quad + \tilde{k}x^\top [\text{sgn}(b)G^\top B^\top P_e + \text{sgn}(b)P_e B G] x \\ &\quad + \frac{1}{2}x^\top x + 2x^\top [\delta(x)p_e^\top p_e \delta^\top(x)] x \\ &\leq \tilde{k}x^\top [\text{sgn}(b)G^\top B^\top P_e + \text{sgn}(b)P_e B G] x \\ &\quad - x^\top Q x - 2\tilde{k}e^{-\alpha k} x^\top R x. \end{aligned} \quad (3.15)$$

First, consider the case $\tilde{k} \geq 0$, then $\dot{V}(x, \tilde{k}) \leq -x^\top Q x + \tilde{k}\sigma_1 x^\top x$, where $\sigma_1 \triangleq \lambda_{\max}(\text{sgn}(b)G^\top B^\top P_e + \text{sgn}(b)P_e B G)$. If $\sigma_1 \leq 0$, then $\dot{V}(x, \tilde{k}) \leq -x^\top Q x$. If $\sigma_1 > 0$, then let $0 < \varepsilon < \lambda_{\min}(Q)$. Thus, for all \tilde{k} such that $0 \leq \tilde{k} \leq \frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_1}$, $\dot{V}(x, \tilde{k}) \leq -\varepsilon x^\top x$.

Now, consider the case $\tilde{k} \leq 0$, then $\dot{V}(x, \tilde{k}) \leq -x^\top Q x + \tilde{k}\sigma_2 x^\top x$, where $\sigma_2 \triangleq \lambda_{\min}(\text{sgn}(b)G^\top B^\top P_e + \text{sgn}(b)P_e B G - 2R)$. If $\sigma_2 \geq 0$, then $\dot{V}(x, \tilde{k}) \leq -x^\top Q x$. If $\sigma_2 < 0$, then, for all \tilde{k} such that $\frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_2} \leq \tilde{k} \leq 0$, $\dot{V}(x, \tilde{k}) \leq -\varepsilon x^\top x$. If $\sigma_1 \leq 0$, then $\sigma_2 < 0$, so define the domain

$$\mathcal{D} \triangleq \begin{cases} \{\tilde{k} \in \mathbb{R} : -\infty < \tilde{k} \leq \frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_1}\}, & \sigma_1 > 0, \sigma_2 \geq 0 \\ \{\tilde{k} \in \mathbb{R} : \frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_2} \leq \tilde{k} \leq \frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_1}\}, & \sigma_1 > 0, \sigma_2 < 0 \\ \{\tilde{k} \in \mathbb{R} : \frac{\lambda_{\min}(Q) - \varepsilon}{\sigma_2} \leq \tilde{k} < \infty\}, & \sigma_1 \leq 0, \sigma_2 < 0 \end{cases} \quad (3.16)$$

Thus, for all $x \in \mathbb{R}^n$ and all $\tilde{k} \in \mathcal{D}$, $\dot{V}(x, \tilde{k}) \leq -\varepsilon x^\top x$ and the solution $(0, k_e)$ is Lyapunov stable.

Next, we show that, $k(t)$ converges. Lemma 2.2 implies that there exists $k_s > 0$ such that, for all constant $k \geq k_s$, $A_s(k)$ is asymptotically stable. For $k \geq k_s$, define $V_0(x, k) \triangleq x^\top P(k)x$, where $P(k) > 0$ satisfies the Lyapunov equation $A_s^\top(k)P(k) + P(k)A_s(k) = -e^{\alpha k}R$ and $\alpha > 0$. Taking the derivative of $V_0(x, k)$ along a trajectory of (3.12) and (3.13) yields

$$\begin{aligned} \dot{V}_0(x, k) &= -e^{-\alpha k} x^\top R x + \dot{k} x^\top \frac{\partial P(k)}{\partial k} x \\ &\quad + x^\top [\Delta^\top(x)P(k) + P(k)\Delta(x)] x \end{aligned} \quad (3.17)$$

Define $\Omega(x, k) \triangleq \sqrt{\frac{1}{2}}e^{-\frac{\alpha}{2}k} R^{\frac{1}{2}} - \sqrt{2}e^{\frac{\alpha}{2}k} R^{-\frac{1}{2}} P(k)\Delta(x)$, and since $0 \leq \Omega^\top(k)\Omega(k)$, it follows that $\Delta^\top(x)P(k) + P(k)\Delta(x) \leq \frac{1}{2}e^{-\alpha k}R + 2e^{\alpha k}\Delta^\top(x)P(k)R^{-1}P(k)\Delta(x)$.

Combining this with (3.17) yields

$$\begin{aligned} \dot{V}_0(x, k) &\leq -\frac{1}{2}e^{-\alpha k} x^\top R x + \dot{k} x^\top \frac{\partial P(k)}{\partial k} x \\ &\quad + 2e^{\alpha k} x^\top \Delta^\top(x)P(k)R^{-1}P(k)\Delta(x)x. \end{aligned} \quad (3.18)$$

Lemma 2.2 implies that $\lim_{k \rightarrow \infty} k e^{\alpha k} p_1(k)$ exists, and thus $\lim_{k \rightarrow \infty} e^{\alpha k} p_1(k) = 0$. Since $\delta(x)$ is bounded for all $x \in \mathbb{R}$, it follows that $\lim_{k \rightarrow \infty} 2e^{2\alpha k} \delta(x) p_1^\top(k) R^{-1} p_1(k) \delta^\top(x) = 0$, and there exists $k_1 \geq k_s$ such that, for all $k \geq k_1$, $2e^{2\alpha k} \delta(x) p_1^\top(k) R^{-1} p_1(k) \delta^\top(x) \leq \frac{1}{4}R$. Then, for all $k \geq k_1$, $\dot{V}_0(x, k) \leq -\frac{1}{4}e^{-\alpha k} x^\top R x + \dot{k} x^\top \frac{\partial P(k)}{\partial k} x$. Lemma 2.2 also implies that there exists $k_2 \geq k_s$ such that, for all $k \geq k_2$, $\frac{\partial P(k)}{\partial k} < 0$. Thus, for all $k \geq k_3 \triangleq \max(k_1, k_2)$, $\dot{V}_0(x, k) \leq -\frac{1}{4}e^{-\alpha k} x^\top R x = -\frac{1}{4}\dot{k}$, which implies

$$\dot{V}_0(x, k) dt \leq -\frac{1}{4} dk. \quad (3.19)$$

To prove that $\lim_{t \rightarrow \infty} k(t)$ exists, suppose that $k(t)$ diverges to infinity in either finite or infinite time. Then there exists $t_3 > 0$ such that $k(t_3) = k_3$. Since $k(t)$ does not escape at t_3 , it follows from (3.12) that $x(t)$ does not escape at time t_3 . Let $t > t_3$ be such that $k(\cdot)$ exists on $[t_3, t]$. Integrating (3.19) from t_3 to t and from k_3 to $k(t)$ and solving for $k(t)$ yields $k(t) \leq k(t_3) + 4V_0(x(t_3), k_3) - 4V_0(x(t), k(t)) \leq k(t_3) + 4V_0(x(t_3), k_3)$. Hence $k(t)$ is bounded on $[0, \infty)$, and thus $k(t)$ does not diverge to infinity. Since $k(t)$ is non-decreasing, $k_\infty \triangleq \lim_{t \rightarrow \infty} k(t)$ exists.

Next, we show that $x(t)$ is bounded. Define the function $V_1(x) = x^\top x$. Taking the derivative of $V_1(x)$ along a trajectory of (3.12)-(3.13) yields $\dot{V}_1(x, k) = x^\top [A_s^\top(k) + A_s(k)] x + x^\top [\Delta^\top(x) + \Delta(x)] x$. Since $k(t)$ converges and $\Delta(x)$ is bounded, there exist $\eta > 0$ such that $\dot{V}_1(x, k) \leq \eta x^\top R x = \eta e^{\alpha k} \dot{k}$, which implies $\dot{V}_1(x, k) dt \leq \eta e^{\alpha k} dk$. Integrating from 0 to t and from $k(0)$ to $k(t)$ and solving for $V_1(x(t))$ yields $V_1(x(t)) \leq \frac{\eta}{\alpha} e^{\alpha k(t)} + V_1(x(0)) - \frac{\eta}{\alpha} e^{\alpha k(0)}$. Since $k(t)$ is bounded, we conclude that $V_1(x(t))$ is bounded. Thus, $x(t)$ is bounded.

Next, we show that $\lim_{t \rightarrow \infty} x(t) = 0$. The dynamics (3.13) implies

$$\|\dot{x}(t)\| \leq (\|A_s(k)\| + \|\Delta(x(t))\|) \|x(t)\|. \quad (3.20)$$

Since $\lim_{t \rightarrow \infty} k(t)$ exists, $A_s(k)$ is bounded. Furthermore, for all $x \in \mathbb{R}^n$, $\Delta(x(t))$ is bounded. Since $\tilde{A}(k(t))$, $\Delta(x(t))$, and $x(t)$ are bounded, it follows from (3.20) that $\dot{x}(t)$ is bounded. Thus, $\frac{d}{dt} [e^{-\alpha k(t)} x^\top(t) R x(t)] = e^{-\alpha k(t)} (-\alpha \dot{k}(t) x^\top(t) R x(t) + 2x^\top(t) R \dot{x}(t))$ is bounded. Since the derivative of $\dot{k}(t) = e^{-\alpha k(t)} x^\top(t) R x(t)$ is bounded, $\dot{k}(t)$ is uniformly continuous. Since $\dot{k}(t)$ is uniformly continuous and $\lim_{t \rightarrow \infty} k(t)$ exists, Barablat's lemma implies that $\lim_{t \rightarrow \infty} e^{-\alpha k(t)} x^\top(t) R x(t) = 0$. Thus, $\lim_{t \rightarrow \infty} x(t) = 0$. \square

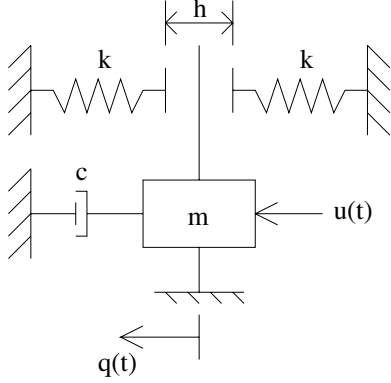


Fig. 1. Nonlinear spring-mass-damper with Coulomb friction and a dead zone in the spring stiffness.

4. NONLINEAR SPRING-MASS-DAMPER EXAMPLE

In this section, we consider Lyapunov-stable adaptive stabilization of the nonlinear spring-mass-damper

$$m\ddot{q}(t) + \hat{c}(\dot{q}(t)) + \hat{k}(q(t)) = u(t), \quad (4.1)$$

where

$$\hat{c}(\dot{q}(t)) \triangleq \begin{cases} (c + \frac{d}{\delta}) \dot{q}(t), & |\dot{q}(t)| < \delta, \\ c\dot{q}(t) + \text{sgn}(\dot{q}(t))d, & |\dot{q}(t)| \geq \delta, \end{cases} \quad (4.2)$$

$$\hat{k}(q(t)) \triangleq \begin{cases} k_0(q(t) + \frac{h}{2}), & q(t) \leq -\frac{h}{2}, \\ 0, & |q(t)| < \frac{h}{2}, \\ k_0(q(t) - \frac{h}{2}), & q(t) \geq \frac{h}{2}, \end{cases} \quad (4.3)$$

and $\delta, c, d, h, k_0 > 0$. The function $\hat{c}(\cdot)$ is a continuous approximation of Coulomb friction and satisfies assumption (i). The function $\hat{k}(\cdot)$ is a linear spring with a deadzone. This nonlinear system is shown in Figure 1. Note that the uncontrolled system has a continuum of equilibria, and the origin of the system is semistable, but not asymptotically stable. (For the definition of semistability, see [7]). For this example, the mass $m = 3$ kg, the viscous friction $c = 2$ kg/s, the Coulomb friction $d = 20$ N, the spring stiffness $k_0 = 2$ kg/s², the deadzone gap $h = 10$ m, and $\delta = 0.1$ m/s.

Note that the system (4.1) satisfies assumptions (i)-(iv) and the adaptive controller presented in Theorem 3.1 can be used to stabilize the origin. This controller is given by

$$u(t) = -k(t) \begin{bmatrix} g_1 & g_0 \end{bmatrix} \begin{bmatrix} \dot{q}(t) \\ q(t) \end{bmatrix}, \quad (4.4)$$

$$k(t) = e^{-\alpha k(t)} \begin{bmatrix} \dot{q}(t) \\ q(t) \end{bmatrix}^T R \begin{bmatrix} \dot{q}(t) \\ q(t) \end{bmatrix}, \quad (4.5)$$

where $g_0 > 0$, $g_1 > 0$, $k'(0) \geq 0$, $\alpha > 0$, and R is positive definite. We choose the controller parameters $g_0 = 11$, $g_1 = 7$, $\alpha = 0.1$, and $R = I$. The system (4.1) with the adaptive controller (4.4)-(4.5) is simulated with the initial conditions $k(0) = 0$, $q(0) = -25$ m, and $\dot{q}(0) = 10$ m/s. The time histories of the position $q(t)$ and velocity $\dot{q}(t)$ for the open-loop and closed-loop systems are shown in Figure 2. The the equilibrium of the open-loop system is

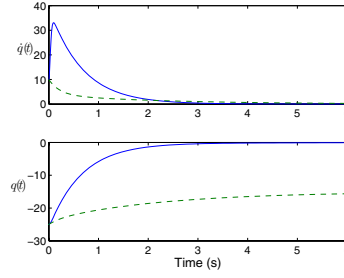


Fig. 2. The open-loop (dashed) and closed-loop (solid) time histories for the position and velocity of the mass. The open-loop system is semistable, and the adaptive controller stabilizes the origin.

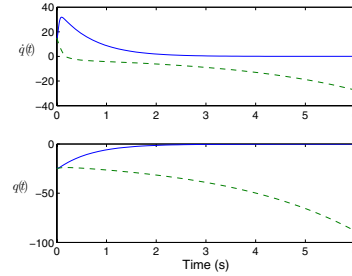


Fig. 3. The open-loop (dashed) and closed-loop (solid) time histories for the position and velocity of the mass. The open-loop system is unstable, and the adaptive controller stabilizes the origin.

semistable, and, while the velocity converges to zero, the position converges to approximately -15.6 m. The adaptive controller stabilizes the equilibrium so that both the velocity and position converge to zero. The adaptive parameter k converges to approximately 42.3.

Next we apply Theorem 3.1 to the unstable system $m\ddot{q}(t) + \hat{c}(\dot{q}(t)) - \hat{k}(q(t)) = u(t)$, which is a modification of (4.1) in which the sign of the stiffness term is negative. This system is simulated with the adaptive controller (4.4)-(4.5) connected in feedback. The initial conditions are $k(0) = 0$, $q(0) = -25$ m, and $\dot{q}(0) = 15$ m/s. Figure 3 shows the time histories of the position and velocity for the open-loop and closed-loop system. The open-loop system is unstable, and the adaptive controller stabilizes the origin.

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