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Under-Actuated Back-Stepping: An Introduction

Jingjing Jiang¹ and Alessandro Astolfi²

Abstract—The stabilization problem for a class of under-actuated systems is solved. This is achieved via a novel back-stepping based method that we call under-actuated back-stepping. The method is developed for linear under-actuated systems first and then extended to nonlinear systems via an example. Numerical simulations are given to demonstrate the effectiveness of the proposed under-actuated back-stepping method.

I. INTRODUCTION

Under-actuation is a technical term describing systems (usually mechanical systems) with fewer control inputs than degrees-of-freedom. In particular, the state of an under-actuated mechanical system is unable to follow arbitrary reference trajectories. As a result, when compared to the control of fully-actuated (mechanical) system, the control design for under-actuated systems is more challenging. One of the broadly used methods to solve control problems for under-actuated systems is based on the linearization technique [1], [2]. Moreover, as energy is a fundamental concept in the control of mechanical systems, “energy-shaping control” (also known as Passivity-Based Control (PBC)), first proposed by Takegaki *et al.* [3], is also a popular way to control mechanical systems [4], [5]. The main drawback of PBC is its difficult applicability, especially for under-actuated systems. The paper [6] has utilized adaptive control to deal with under-actuated systems, while the papers [7], [8] have applied Lyapunov-based control to mechanical systems. Finally, Sliding Mode Control (SMC) is another method used to control mechanical systems [9]. However, for many mechanical systems it is difficult to find a surface suitable for the application of SMC. In addition, other control techniques, such as optimal control [10] and hybrid and switching-based control [11], [12], have also been used. More recently, robust control has been exploited for the study of mechanical systems to deal with model uncertainties, nonholonomic constraints and disturbances [13], [14]. To sum up, apart from linearization based control techniques, all other methods have been mainly used to stabilize particular classes of under-actuated systems.

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The back-stepping method, a well-known control design technique developed in [15], [16], is a constructive method to design stabilizers for classes of nonlinear systems. The systems amenable to the application of back-stepping can be written in the so-called feedback form

$$\begin{aligned}\dot{x} &= f_x(x) + g_x(x)z, \\ \dot{z} &= f(x, z) + g(x)u,\end{aligned}\tag{1}$$

where $x(t) \in \mathbb{R}^n$ and $z(t) \in \mathbb{R}^m$ are the states and $u(t) \in \mathbb{R}^m$ is the input. In addition, $f_x(x) : \mathbb{R}^n \rightarrow \mathbb{R}^n$, $g_x(x) : \mathbb{R}^n \rightarrow \mathbb{R}^{n \times m}$, $f(x, z) : \mathbb{R}^{n+m} \rightarrow \mathbb{R}^{n+m}$ and $g(x, z) : \mathbb{R}^{n+m} \rightarrow \mathbb{R}^{m \times m}$ are smooth mappings. Under the assumptions that $\det(g(x)) \neq 0$ for all $x \in \mathbb{R}^n$ and u_x stabilizes the zero equilibrium of the subsystem

$$\dot{x} = f_x(x) + g_x(x)u_x(x),$$

we can always find a state feedback controller $u = u(x, z)$, through back-stepping, such that the origin of the closed-loop system is globally asymptotically stable [15], [16], [17], [18]. Note that in this “fully-actuated” case, that is the case in which the dimension of z and of u are the same, one can select any stabilizer u_x in the initial step of back-stepping.

The main contribution of the paper is as follows. The paper proposes a novel method, the under-actuated back-stepping method, which is inspired by standard back-stepping, to solve the stabilization problem for classes of under-actuated (mechanical) system. The proposed method is explained in detail for linear time-invariant under-actuated systems first and then the nonlinear extension is discussed via an example.

The rest of the paper is organized as follows. In Section II the dynamics of two classes of linear under-actuated system are given and the stabilization problem for these systems is formulated. Solutions to the stated problems are presented in Section III, in which the stability properties of the resulting closed-loop systems with the controllers developed by the proposed methodology are given. In addition, one numerical example to illustrate how under-actuated back-stepping works is provided. Section IV studies how the design procedure developed for linear systems can be extended to nonlinear systems and uses the Inertia Wheel Pendulum system to demonstrate the effectiveness of the method. Finally, conclusions and suggestions for future work are given in Section V.

II. PROBLEM STATEMENT

This section formulates the stabilization problem for two classes of under-actuated linear systems. The under-actuated

back-stepping based solutions for these stabilization problems is given in Section III.

The term under-actuated is used informally to denote systems in cascaded form for which the number of input is strictly less than half the number of states. These systems arise when linearizing under-actuated mechanical systems.

Problem 1: Consider an under-actuated system, the dynamics of which are described by the equations

$$\begin{aligned}\dot{x} &= z, \\ \dot{z} &= Ax + Bu,\end{aligned}\quad (2)$$

with $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$, states $x(t) \in \mathbb{R}^n$ and $z(t) \in \mathbb{R}^n$ and input $u(t) \in \mathbb{R}^m$, with $0 < m < n$. Assume that $\text{rank}(B) = m$ hence, without loss of generality, let $B = [I_m, 0_{m \times (n-m)}]^T$. Find a controller

$$u = H_1 x + H_2 (z - Kx), \quad (3)$$

with $H_1 \in \mathbb{R}^{m \times n}$, $H_2 \in \mathbb{R}^{m \times n}$ and $K \in \mathbb{R}^{n \times n}$, such that the closed-loop system (2)-(3) is asymptotically stable.

Problem 2: Consider an under-actuated system, the dynamics of which are described by the equations

$$\begin{aligned}\dot{x} &= Fx + Gz, \\ \dot{z} &= Ax + Ez + Bu,\end{aligned}\quad (4)$$

with $F \in \mathbb{R}^{n \times n}$, $G \in \mathbb{R}^{n \times n}$, $A \in \mathbb{R}^{n \times n}$, $E \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$, states $x(t) \in \mathbb{R}^n$ and $z(t) \in \mathbb{R}^n$ and control input $u(t) \in \mathbb{R}^m$, with $0 < m < n$. Assume that $\text{rank}(B) = m$ and $\det(G) \neq 0$. Hence, without loss of generality, assume that $GB = [I_m, 0_{(n-m) \times m}]^T$. Find a controller

$$u = H_1 x + H_2 (Gz - Kx), \quad (5)$$

with $H_1 \in \mathbb{R}^{m \times n}$, $H_2 \in \mathbb{R}^{m \times n}$ and $K \in \mathbb{R}^{n \times n}$, such that the closed-loop system (4)-(5) is asymptotically stable.

It is tacitly assumed that the system (2) and (4) are controllable, hence these can be stabilized using a linear state feedback. The goal of the paper is therefore not to merely stabilize these systems, but to propose a modular design inspired by back-stepping and hence applicable, in principle, to nonlinear systems (see the example in Section IV).

III. MAIN RESULTS

In this section we give solutions to the Problems 1 and 2 stated in Section II. The corresponding under-actuated back-stepping algorithms used to find the controllers are given in Algorithms 1 and 2, respectively. Finally, an example is given at the end of this section to show how the algorithms work.

We begin with Problem 1, as the dynamics of system (2) can be regarded as a special case of the dynamics of the systems studied in Problem 2. Algorithm 1 can be used to find the controller (3) and the existence of the matrix K is proved in Lemma 1.

Algorithm 1: Solution To Problem 1

procedure UNDER-ACTUATED BACK-STEPPING

Step 1: Find a matrix $K \in \mathbb{R}^{n \times n}$ such that the conditions

$$\lambda(K) \in \mathbb{C}^-, \quad (6)$$

$$(K, B) \text{ controllable}, \quad (7)$$

$$J_1(A - K^2) = 0_{n \times n}, \quad (8)$$

where

$$J_1 = \begin{bmatrix} 0_{m \times m} & 0_{m \times (n-m)} \\ 0_{(n-m) \times m} & I_{(n-m) \times (n-m)} \end{bmatrix}, \quad (9)$$

hold.

Step 2: Set

$$H_1 = -J_2(A - K^2), \quad (10)$$

with

$$J_2 = [I_{m \times m}, 0_{(n-m) \times m}].$$

Step 3: Select H_2 such that

$$\lambda(-K + BH_2) \in \mathbb{C}^-. \quad (11)$$

end procedure

Lemma 1: Consider the system (2) with $m = 1$, yielding $B = [1, 0, 0, \dots, 0]^T$. Then there exists a matrix K such that conditions (6) to (8) in Algorithm 1 hold.

Proof: The matrices A and K can be written as

$$A = \begin{bmatrix} a_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}, \quad K = \begin{bmatrix} k_{11} & K_{12} \\ K_{21} & K_{22} \end{bmatrix}, \quad (12)$$

where $a_{11} \in \mathbb{R}$, $A_{12}^T \in \mathbb{R}^{n-1}$, $A_{21} \in \mathbb{R}^{n-1}$, $A_{22} \in \mathbb{R}^{(n-1) \times (n-1)}$, $k_{11} \in \mathbb{R}$, $K_{12}^T \in \mathbb{R}^{n-1}$, $K_{21} \in \mathbb{R}^{n-1}$ and $K_{22} \in \mathbb{R}^{(n-1) \times (n-1)}$. Since (A, B) is controllable, then (A_{22}, A_{21}) is controllable. Hence, there is at least one element of A_{21} which is nonzero. Without loss of generality, we assume that the first element of A_{21} , denoted as $a_{2,1}$, is nonzero.

Note that (7) is equivalent to controllability of (K_{22}, K_{21}) . Pick K_{22} and K_{21} to satisfy the above controllability condition. One such a choice is $K_{21} = [\alpha, 0, 0, \dots, 0]^T$, with nonzero α , and

$$K_{22} = \begin{bmatrix} \beta_{1,1} & \beta_{1,2} & \cdots & \beta_{1,n-1} \\ \beta_{2,1} & \beta_{2,2} & \cdots & \beta_{2,n-1} \\ \vdots & \vdots & \ddots & \vdots \\ \beta_{n-1,1} & \beta_{n-1,2} & \cdots & \beta_{n-1,n-1} \end{bmatrix}$$

with $\beta_{i,j} = 0$, for all $i \neq j$ and $i \neq (j+1)$, and $\beta_{i,j} \neq 0$, for all $i = j$ and $i = (j+1)$. Exploiting (8) k_{11} and K_{12} can then be calculated as

$$\begin{aligned}k_{11} &= (K_{21}^T K_{21})^{-1} K_{21}^T [A_{21} - K_{22} K_{21}], \\ K_{12} &= (K_{21}^T K_{21})^{-1} K_{21}^T [A_{22} - K_{22}^2].\end{aligned}$$

Note that if $\beta_{2,1} = 0$ then the eigenvalues of the matrix K , denoted as $\lambda_1, \lambda_2, \dots, \lambda_n$, are such that $\lambda_3 = \beta_{2,2}$, $\lambda_4 = \beta_{3,3}$, $\dots$, $\lambda_n = \beta_{n-1,n-1}$, and

$$\lambda_1 + \lambda_2 = \frac{a_{2,1}}{\alpha}, \quad \lambda_1 \lambda_2 = \frac{a_{2,1} \beta_{1,1}}{\alpha} - a_{2,2},$$

where $a_{2,2}$ denotes the element on the second row and second column of the matrix A .

Selecting

$$\begin{aligned} \beta_{1,1} &= -a_{2,2} - 1, \beta_{i,i} < 0, \quad \forall i \in \{2, 3, \dots, n-1\}, \\ \alpha &= -a_{2,1}, \beta_{i+1,i} \neq 0, \forall i \in \{2, 3, \dots, n-2\}, \end{aligned}$$

and noting that the eigenvalues are robust, *i.e.* their position is only modified by a small change in parameters, one could select $\beta_{2,1} \neq 0$ and small to satisfy (6) and (7), hence the claim. $\blacksquare$

Remark 1: Lemma 1 shows that K in Algorithm 1 exists in the case $m = 1$. It is trivial to extend the result to the case $m \geq 2$.

The stability properties for the closed-loop system (2)-(3) can then be summarized as follows.

Proposition 1: Consider the under-actuated system (2) controlled with the feedback law given by (3) with $K \in \mathbb{R}^{n \times n}$, $H_1 \in \mathbb{R}^{m \times n}$ and $H_2 \in \mathbb{R}^{m \times n}$ as in Algorithm 1. Then the closed-loop system (2)-(3) is asymptotically stable.

Proof: To begin with define the new variable Δz as $\Delta z = z - Kx$. Using the new variable Δz the system (2) can be rewritten as

$$\begin{aligned} \dot{x} &= Kx + \Delta z, \\ \dot{\Delta z} &= (A - K^2)x - K\Delta z + Bu. \end{aligned} \quad (13)$$

Substituting the control law (3) into equations (13) yields

$$\begin{aligned} \dot{x} &= Kx + \Delta z, \\ \dot{\Delta z} &= (A - K^2 + BH_1)x + (BH_2 - K)\Delta z. \end{aligned}$$

By equations (8) and (10) one has

$$\begin{aligned} \dot{x} &= Kx + \Delta z, \\ \dot{\Delta z} &= (BH_2 - K)\Delta z. \end{aligned}$$

The condition (7) implies that there exists H_2 such that (11) holds. Hence, there exists a symmetric positive definite matrix Q such that

$$(BH_2 - K)^T Q + Q(BH_2 - K) = -I.$$

Similarly, by condition (6), there exists a symmetric positive definite matrix P such that

$$K^T P + PK = -I.$$

Let ϵ be a sufficiently small positive constant. Consider the Lyapunov function candidate

$$L(x, z) = \epsilon x^T P x + \Delta z^T Q \Delta z. \quad (14)$$

Its time derivative along the trajectories of the closed-loop system is such that

$$\begin{aligned} \dot{L} &= \epsilon(x^T K^T + \Delta z^T) P x + \epsilon x^T P (Kx + \Delta z) \\ &\quad + \Delta z^T [(BH_2 - K)^T Q + Q(BH_2 - K)] \Delta z \\ &\leq \epsilon x^T (K^T P + PK) x + \frac{\epsilon}{2} x^T x + 2\epsilon \Delta z^T P^T P \Delta z \\ &\quad + \Delta z^T [(BH_2 - K)^T Q + Q(BH_2 - K)] \Delta z \\ &\leq -\frac{\epsilon}{2} x^T x - (1 - 2\epsilon \bar{\sigma}_P^2) \Delta z^T \Delta z, \end{aligned}$$

where $\bar{\sigma}_P$ denotes the maximum singular value of the matrix P . Therefore, $\dot{L} \leq 0$ for all $0 < \epsilon < \frac{1}{2\bar{\sigma}_P^2}$. In addition,

$$\dot{L} = 0 \iff x = z = 0,$$

hence the claim. $\blacksquare$

Remark 2: If the system (2) is fully-actuated, *i.e.* $m = n$, then trivially conditions (7) and (8) hold. Therefore, in Step 1 the matrix K only needs to satisfy condition (6), consistently with standard back-stepping, *i.e.* any “stabilizing” feedback can be selected in the initial step of back-stepping.

Problem 2 is a generalization of Problem 1 and can also be solved using the under-actuated back-stepping method. The steps of the control design are detailed in Algorithm 2. The existence of the matrix K and the formal properties of the closed-loop system (4)-(5) are given in Lemma 2 and Proposition 2, respectively.

Algorithm 2: Solution To Problem 2

procedure UNDER-ACTUATED BACK-STEPPING
Step 1: Find a matrix $K \in \mathbb{R}^{n \times n}$ such that the conditions

$$\lambda(F + K) \in \mathbb{C}^-, \quad (15)$$

$$(GEG^{-1} - K, GB) \text{ controllable}, \quad (16)$$

$$J_1[GA - K(F + K) + GEG^{-1}K] = 0_{n \times n}, \quad (17)$$

where

$$J_1 = \begin{bmatrix} 0_{m \times m} & 0_{m \times (n-m)} \\ 0_{(n-m) \times m} & I_{(n-m) \times (n-m)} \end{bmatrix},$$

hold.

Step 2: Set

$$H_1 = -J_2[GA - K(F + K) + GEG^{-1}K], \quad (18)$$

with

$$J_2 = [I_{m \times m}, 0_{(n-m) \times m}].$$

Step 3: Select H_2 such that

$$\lambda(GEG^{-1} - K + GBH_2) \in \mathbb{C}^-. \quad (19)$$

end procedure

Lemma 2: Consider the system (4) with $m = 1$, yielding $B = [1, 0, 0, \dots, 0]^T$. Then there exists a matrix K such that conditions (15) to (17) hold.

Proof: Similarly to the proof of Lemma 1, we rewrite the matrices A and K as in (12) and the matrices E and F as

$$E = \begin{bmatrix} e_{11} & E_{12} \\ E_{21} & E_{22} \end{bmatrix}, \quad F = \begin{bmatrix} f_{11} & F_{12} \\ F_{21} & F_{22} \end{bmatrix}, \quad (20)$$

where $e_{11} \in \mathbb{R}$, $E_{12}^T \in \mathbb{R}^{n-1}$, $E_{21} \in \mathbb{R}^{n-1}$, $E_{22} \in \mathbb{R}^{(n-1) \times (n-1)}$, $f_{11} \in \mathbb{R}$, $F_{12}^T \in \mathbb{R}^{n-1}$, $F_{21} \in \mathbb{R}^{n-1}$, $F_{22} \in \mathbb{R}^{(n-1) \times (n-1)}$. Since G is invertible, without loss of generality we can assume $G = I_n$. System (4) is controllable, indicating that

$$\left(\begin{bmatrix} F_{11} & F_{12} & 0_{n-1}^T \\ F_{21} & F_{22} & I_{n-1} \\ A_{21} & A_{22} & E_{22} \end{bmatrix}, \begin{bmatrix} 1 \\ 0_{n-1} \\ E_{21} \end{bmatrix} \right) \text{ is controllable,}$$

hence $(E_{21} + F_{21}) \neq 0_{n-1}$. Without loss of generality, we can assume that its first element is nonzero. Choose K_{21} and K_{22} such that $E_{21} - K_{21} = [\alpha, 0, 0, \dots, 0]^T$, with nonzero α , and

$$E_{22} - K_{22} = \begin{bmatrix} \beta_{1,1} & \beta_{1,2} & \cdots & \beta_{1,n-1} \\ \beta_{2,1} & \beta_{2,2} & \cdots & \beta_{2,n-1} \\ \vdots & \vdots & \ddots & \vdots \\ \beta_{n-1,1} & \beta_{n-1,2} & \cdots & \beta_{n-1,n-1} \end{bmatrix},$$

with $\beta_{i,j} = 0$, for all $i \neq j$ and $i \neq (j+1)$. Exploiting (17) k_{11} and K_{12} can then be calculated as

$$\begin{aligned} k_{11} &= [(K_{21} - E_{21})^T (K_{21} - E_{21})]^{-1} (K_{21} - E_{21})^T \\ &\quad \times [A_{21} - K_{21} f_{11} - K_{22} F_{21} - K_{22} K_{21} + E_{22} K_{21}], \\ K_{12} &= [(K_{21} - E_{21})^T (K_{21} - E_{21})]^{-1} (K_{21} - E_{21})^T \\ &\quad \times [A_{22} - K_{21} F_{12} - K_{22} F_{22} - K_{22}^2 + E_{22} K_{22}]. \end{aligned}$$

Let the eigenvalues of $(F + K)$ be $\lambda_1, \lambda_2, \dots, \lambda_n$. Then by choosing $\beta_{i,i} \gg 0$, for all $i \in \{2, 3, \dots, n-1\}$ we have

$$\begin{aligned} \lambda_1 + \lambda_2 &\approx \frac{e_{2,1} + f_{2,1}}{\alpha} \beta_{1,1} + e_{2,2} + f_{2,2} \\ &\quad + \frac{a_{2,1} - \sum_{i=1}^n (e_{2,i} f_{i,1})}{\alpha}, \\ \lambda_1 \lambda_2 &\approx -\beta_{1,1}^2 + (e_{2,2} + f_{2,2} - \frac{e_{2,1} + f_{2,1}}{\alpha} e_{2,2}) \beta_{1,1} \\ &\quad - \frac{a_{2,1} - \sum_{i=1}^n (e_{2,i} f_{i,1})}{\alpha} \beta_{1,1} + \frac{e_{2,1} + f_{2,1}}{\alpha} f_{1,2} e_{2,1} \\ &\quad + \frac{[a_{2,1} - \sum_{i=1}^n (e_{2,i} f_{i,1})](e_{2,2} + f_{2,2})}{\alpha} \\ &\quad + \frac{e_{2,1} + f_{2,1}}{\alpha} (\sum_{i=4}^n (e_{2,i} e_{i,2}) - a_{3,2} - e_{2,2}^2) \\ &\quad + \frac{e_{2,1} + f_{2,1}}{\alpha} e_{2,3} (e_{3,2} - \beta_{2,1}), \end{aligned}$$

where $e_{i,j}$ and $f_{i,j}$ denote the element of the j^{th} -column i^{th} -row of the matrix E and F , respectively, and $\lambda_i \approx -\beta_{i-1,i-1}$ for all $i \in \{3, 4, \dots, n\}$. Since $e_{2,1} + f_{2,1} \neq 0$, it is always possible to find α and $\beta_{1,1}$ such that $\lambda_1 < 0$ and $\lambda_2 < 0$, hence the claim. ■

Proposition 2: Consider the under-actuated system (4) with the controller (5) with $K \in \mathbb{R}^{n \times n}$, $H_1 \in \mathbb{R}^{m \times n}$ and $H_2 \in$

$\mathbb{R}^{m \times n}$ as in Algorithm 2. Then the zero equilibrium of the closed-loop system is asymptotically stable.

Proof: To begin with define the new variable

$$\Delta z = Gz - Kx. \quad (21)$$

Recall that G^{-1} exists since $\det(G) \neq 0$. Using the new variable Δz the system (4) can be rewritten as

$$\begin{aligned} \dot{x} &= (F + K)x + \Delta z, \\ \dot{\Delta z} &= [GA - K(F + K) + GEG^{-1}K]x \\ &\quad + (GEG^{-1} - K)\Delta z + GBu. \end{aligned} \quad (22)$$

Substituting the control law (5), together with equations (18) and (19) into the equation (22), yields

$$\begin{aligned} \dot{x} &= (F + K)x + \Delta z, \\ \dot{\Delta z} &= (GEG^{-1} - K + GBH_2)\Delta z. \end{aligned}$$

The condition (16) implies that we can always find a matrix H_2 such that (19) holds. Similarly to the proof of Proposition 1, there exists matrices P and Q such that

$$\begin{aligned} P &= P^T > 0, \quad Q = Q^T > 0, \\ M^T Q + QM &= -I_n, \\ (F + K)^T P + P(F + K) &= -I_n, \end{aligned}$$

where $M = GEG^{-1} - K + GBH_2$.

Again, let ϵ be a sufficiently small positive constant. Consider the Lyapunov function candidate (14), the time derivative of which along the trajectories of the closed-loop system is such that

$$\begin{aligned} \dot{L} &= \epsilon[x^T (F + K)^T + \Delta z^T]Px + \epsilon x^T P[(F + K)x + \Delta z] \\ &\quad + \Delta z^T (GEG^{-1} - K + GBH_2)^T Q \Delta z \\ &\quad + \Delta z^T Q (GEG^{-1} - K + GBH_2) \Delta z \\ &\leq \epsilon x^T [(F + K)^T P + P(F + K)]x + \frac{\epsilon}{2} x^T x \\ &\quad + 2\epsilon \Delta z^T P^T P \Delta z \\ &\quad + \Delta z^T (GEG^{-1} - K + GBH_2)^T Q \Delta z \\ &\quad + \Delta z^T Q (GEG^{-1} - K + GBH_2) \Delta z \\ &\leq -\frac{\epsilon}{2} x^T x - (1 - 2\epsilon \bar{\sigma}_P^2) \Delta z^T \Delta z \leq 0, \end{aligned}$$

where $\bar{\sigma}_P$ is the maximum singular value of the matrix P , hence the claim. ■

Example 1: Consider the two degrees-of-freedom, under-actuated linear system described by the equations

$$\begin{aligned} \dot{x} &= \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} x + \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} z, \\ \dot{z} &= \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} x + \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} z + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u, \end{aligned} \quad (23)$$

with states $x(t) \in \mathbb{R}^2$ and $z(t) \in \mathbb{R}^2$ and input $u(t) \in \mathbb{R}$.

According to Algorithm 2 the first step is to find a matrix K satisfying conditions (15), (16) and (17). One such a choice is given by

$$K = \begin{bmatrix} 0 & 4 \\ -\frac{1}{2} & -2 \end{bmatrix},$$

yielding

$$H_1 = [-1.5, -2].$$

Then H_2 has to be chosen such that (19) holds. One such a choice is given by

$$H_2 = [-3, -10],$$

yielding the overall controller

$$\begin{aligned} u &= H_1 x + H_2 \left(\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} z - Kx \right) \\ &= [-6.5, -10]x + [-3, -10]z. \end{aligned} \quad (24)$$

IV. TOWARDS NONLINEAR UNDER-ACTUATED BACK-STEPPING

In this section we discuss how the under-actuated back-stepping method can be used to design stabilizers for nonlinear under-actuated systems. The dynamics of a class of under-actuated mechanical systems can be described by the equations

$$\begin{aligned} \dot{x} &= z, \\ \dot{z} &= A(x)x + Ez + Bu, \end{aligned} \quad (25)$$

where the states $x(t) \in \mathbb{R}^n$ and $z(t) \in \mathbb{R}^n$ denotes the position and velocity vectors, respectively, while the input $u(t) \in \mathbb{R}^m$ represents the force or torque vector. The mapping $A(x) : \mathbb{R}^n \rightarrow \mathbb{R}^{n \times n}$ is smooth and $B \in \mathbb{R}^{n \times m}$ and $E \in \mathbb{R}^{n \times n}$ are constant matrices. Assume that $\text{rank}(B) = m$ hence, without loss of generality, let $B = [I_m, 0_{(n-m) \times m}^T]^T$. To find a control law $u(x, z)$ such that the zero equilibrium of the closed-loop system is asymptotically stable we can exploit the under-actuated back-stepping method developed.

The first step is to find the mappings $K : \mathbb{R}^n \rightarrow \mathbb{R}^n$ and $H_2 : \mathbb{R}^m \rightarrow \mathbb{R}^n$ such that the following properties hold.

P1) $\frac{\partial(BH_2(x) - K(x))}{\partial x} = 0_{n \times n}.$

P2) The point $x = 0$ and $\Delta z = 0$ are globally asymptotically stable equilibrium of the subsystems

$$\dot{x} = K(x),$$

and

$$\dot{\Delta z} = \left[E - \frac{\partial K(x)}{\partial x} + BH_2(x) \right] \Delta z,$$

uniformly in x , respectively.

P3) $J_1(A(x) + EK(x) - \frac{\partial K(x)}{\partial x}K(x)) = 0_{n \times n}$, where J_1 is defined as in (9).

The second step of the method requires, in this case, to calculate $H_1 : \mathbb{R}^m \rightarrow \mathbb{R}^n$ as

$$H_1(x) = -J_2(A(x) + EK(x) - \frac{\partial K(x)}{\partial x}K(x)),$$

where J_2 is as given in (18).

Finally, globally asymptotic stability of the zero equilibrium of the system (25) in closed-loop with the control law

$$u = H_2(x)(z - K(x)) + H_1(x)$$

can be proved by using a Lyapunov function candidate of the form

$$L(x, z) = (z - K(x)x)^T Q(x)(z - K(x)x) + \epsilon x^T P(x)x,$$

with $\epsilon > 0$ and sufficiently small and $P : \mathbb{R}^n \rightarrow \mathbb{R}^{n \times n}$ and $Q : \mathbb{R}^n \rightarrow \mathbb{R}^{n \times n}$ positive definite.

The following example is used to illustrate how under-actuated back-stepping can be applied to nonlinear systems.

Example 2: Consider an inertia wheel pendulum, the dynamics of which are described by the equations [1]

$$\begin{aligned} \dot{x} &= z, \\ \dot{z} &= \begin{bmatrix} u, \\ (b_2 - b_1)M \sin x_1 \end{bmatrix}, \end{aligned} \quad (26)$$

where

$$\begin{aligned} M &= \frac{(m_1 l_1 + m_2 l_2)g}{I_1 + m_1 l_1^2 + m_2 l_2^2}, \quad b_1 = \frac{1}{I_1 + m_1 l_1^2 + m_2 l_2^2}, \\ b_2 &= \frac{1}{I_2} + \frac{1}{I_1 + m_1 l_1^2 + m_2 l_2^2}, \quad u = M \sin x_1 - b_1 \tau \end{aligned}$$

$x = [q_1, b_2 q_1 + b_1 q_2]^T$ and z are states of the system, while u is the control input. q_1 and q_2 represent the pendulum angle and the wheel angle, respectively. Note that I_1 and I_2 denote the moment of inertia of the pendulum around its center of mass and the moment of inertia of the wheel (plus actuator's rotor), respectively; m_1 and m_2 represent the mass of the rod and the mass of the inertia wheel plus the mass of the actuator, respectively; l_1 and l_2 describes the distance to the center of mass of the rod and the length of the pendulum, respectively. In addition, τ denotes the torque generated by the actuator acting between the wheel and the pendulum, and g is the gravity constant. Suppose that the parameter values are given as: $b_1 = 1.5/(kg \times m^2)$, $b_2 = 668.17/(kg \times m^2)$, $M = 74.0/s^2$.

Based on the analysis given in Section IV, the first step is to find $K(x) = [k_1(x), k_2(x)]^T$, where $k_1(x) : \mathbb{R}^2 \rightarrow \mathbb{R}$ and $k_2(x) : \mathbb{R}^2 \rightarrow \mathbb{R}$, and $H_2(x)$ such that P1) to P3) hold. One such a choice for $K(x)$ and $\tilde{u}(x, z) \triangleq H(x)(z - K(x))$ is given by

$$\begin{aligned} k_2(x) &= \frac{1}{2} \ln \frac{1 - \frac{4}{\pi} x_1 - \tanh(x_2)}{1 + \frac{4}{\pi} x_1 - \tanh(x_2)}, \\ k_1(x) &= -\frac{\pi}{2(e^{x_2} + e^{-x_2})^2} \ln \frac{1 - \frac{4}{\pi} x_1 - \tanh(x_2)}{1 + \frac{4}{\pi} x_1 - \tanh(x_2)} \\ &\quad - \frac{\pi}{4} \left[1 - \left(\frac{4}{\pi} x_1 + \tanh(x_2) \right)^2 \right] M \sin x_1, \end{aligned} \quad (27)$$

Fig. 1: Time histories of the pendulum angle q_1 and the torque generated by the actuator τ for the inertia wheel pendulum with the controller (31).

and

$$\begin{aligned} \tilde{u}(x, z) = & \frac{\partial k_2}{\partial x_1} \Delta_2 + \dot{\Delta}_1^* - \gamma_4(\Delta_1 - \Delta_1^*) + \frac{\partial k_1}{x_1} \Delta_1 \\ & + \frac{\partial k_1}{x_2} \Delta_2, \end{aligned} \quad (28)$$

where $\Delta_1 = z_1 - k_1(x)$, $\Delta_2 = z_2 - k_2(x)$, $\gamma_4 > 0$ and

$$\Delta_1^* = -\frac{\frac{\partial k_2}{\partial x_2}}{\frac{\partial k_2}{\partial x_1}} \Delta_2 - \gamma_3 \Delta_2, \quad (29)$$

with $\gamma_3 > 0$, respectively. Note that $\frac{\partial k_2}{\partial x_1} < 0$, hence Δ_1^* is well-defined.

The second step is to calculate $H_1(x)$ as

$$H_1(x) = \frac{\partial k_1}{\partial x_1} k_1(x) + \frac{\partial k_1}{\partial x_2} k_2(x), \quad (30)$$

with k_1 as given in (27).

Finally, the control input u to the system (26) is calculated as

$$u = H_1(x) + \tilde{u}(x, z), \quad (31)$$

where $H_1(x)$ and $\tilde{u}(x, z)$ are calculated in (30) and (28), respectively.

Simulation results are given in Fig. 1 which shows that the controller (31) based on the proposed under-actuated back-stepping method is effective in stabilizing the zero equilibrium of the inertia wheel pendulum. Furthermore, the speed of convergence can be changed by choosing different values of γ_3 and γ_4 .

As illustrated by the examples, the key step of the under-actuated back-stepping method is the first step of the Algorithms 1 and 2. This restricts the selection of the initial

stabilizing controller: such a restriction is not present in standard back-stepping.

V. CONCLUSIONS

This paper proposes a novel control method, the under-actuated back-stepping method, to solve the stabilization problem for classes of under-actuated systems. The technique is studied in details for linear systems and its extension to nonlinear under-actuated systems is briefly discussed. Two numerical examples, one linear example given in Section III and the other nonlinear one given in Section IV, show the effectiveness of the proposed method and its main features. In future we will focus on the application of the under-actuated back-stepping method on classes to nonlinear systems.

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