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High-order sliding modes and intelligent PID controllers: First steps toward a practical comparison

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Abstract: This communication is devoted to a "practical" comparison between high-order sliding modes and the recently introduced model-free control. The perfect knowledge of the relative degree of the output variable, which is a standard assumption for sliding modes, is assumed here. Our comparisons are based on two concrete case-studies and on numerous computer simulations. The smoothness of the input variables, the robustness with respect to noises and the straightforward extendibility of the model-free controllers to MIMO systems are highlighted.

Keywords: Nonlinear control; higher-order sliding mode control; model-free control; intelligent PIDs; online parameter estimation.

1. INTRODUCTION

The excellent robustness properties of sliding mode control with respect to perturbations and uncertainties explains its great popularity (see, e.g., Edwards & Spurgeon [1998], Perruquetti & Barbot [2002], Utkin [1992], Utkin *et al.* [1999]). *Chattering*, i.e., the high frequency commutations of the control variables, is however, as we know, a major drawback of this setting. Among the many works aimed at the attenuation of this most annoying shortcoming, let us mention *High-Order Sliding Modes* (see Levant & Alelishvili [2008]), or *HOSM*, the boundary layer approach (Slotine [1991]) and the sliding sector method (Furuta *et al.* [2000]).

There exist several classes of HOSM algorithms (see, e.g., Levant [2005], Levant & Alelishvili [2007, 2008] and the references therein) especially *nested* sliding modes, *quasi-continuous* controllers and *high order integral* sliding mode. The last category is particularly interesting to overcome the chattering effect by (artificially) raising the relative degree. An output feedback sliding mode controller can be realized by combining a sliding mode controller with a sliding mode differentiator. An interesting sliding mode controller for MIMO systems in Defoort *et al.* [2009] is based on the finite time stability of a chain of integrators from Bhat & Bernstein [2005]. We mention that sliding mode controllers had led to a huge number of exciting applications involving real systems as in Bartolini *et al.* [2000, 2003], Ghanes *et al.* [2010], Khan *et al.* [2003], Levant *et al.* [2000], Defoort *et al.* [2009], Orlov *et al.* [2003], Sira-Ramírez [2002], Spurgeon *et al.* [2003], Shkolnikov *et al.* [2000], Shkolnikov & Shtessel [2002], Shtessel & Shkolnikov [2003].

Model-free control (Fliess & Join [2008, 2009], Fliess *et al.* [2011a]) is a recently introduced approach, which does not necessitate any mathematical modeling. The unknown dynamics is approximated on a very small time interval by a very simple model which is continuously updated with the aid of online estimation techniques (Fliess & Sira-Ramírez [2003, 2008]). The loop is closed thanks to an *intelligent* PID, which provides the feedforward compensation and is easily tuned. Model-free control has already led to a number of exciting applications. [1]

Since the knowledge of the relative degree is a standard assumption in sliding mode control, [2] we deal in this article with SISO minimum phase systems of the form

$$\begin{aligned} \dot{x} &= a(x) + b(x)u, \quad x \in \mathcal{D} \subset \mathbb{R}^n, \ u \in \mathbb{R} \\ y &= s(x) \end{aligned} \tag{1}$$

where n and the smooth functions a, b and s are unknown while the relative degree r of y is constant and known:

$$y^{(r)} = f(x) + g(x)u. \tag{2}$$

With the classical assumption $g(x) \neq 0$ and the knowledge of bounds on f and g, all HOSM controllers listed above solves the problem of zeroing the output y in finite time. If the relative degree is not artificially raised, chattering appears when the output and its derivatives up to order $r-1$ are close to 0. The output y can be chosen to solve

[1] See the references in Fliess *et al.* [2011a,b].
[2] With the exception of some recent works (see, e.g., Levant [2010] and the references therein).

a stabilization or a trajectory tracking problem. We show how model-free control can be adapted to the particular situation of (1)-(2).

The knowledge of r is not a restrictive assumption for a large number of finite-dimensional systems encountered in practice: robotic manipulators, mobile robots, electric drives, for example, are modeled by a collection of nonlinear integrators, which may be simple or double. Moreover, dynamic models may be automatically generated by symbolic softwares (see, e.g., Khalil & Creusot [1997] for robotic manipulators). The identification of physical parameters involved in such models is nevertheless a difficult task. Both sliding modes as well as model-free controllers aim to bypass this last step.

The comparison between sliding modes and model-free control is kept here at a practical level for the following two reasons:

- The mathematical backgrounds are different:
  - · differential inclusions for sliding modes,
  - · various algebraic tools for model-free control.
- In the presence of unavoidable noise corruptions the concepts of finite-time and asymptotic stabilities boil both down to "practical" stability.

Our paper is organized as follows. After a short review of model-free control in Section 2, Section 3 defines, via online algebraic estimation techniques, a new controller which takes into account the knowledge of the relative degree. Two nonlinear case-studies are analyzed in Section 4, where numerous computer simulations are displayed. Some concluding remarks are provided in Section 5.

2. A SHORT REVIEW OF MODEL-FREE CONTROL

2.1 Generalities

Replace the unknown, or at least poorly known, system equation by the "ultra-local" model [3]

$$y^{(\nu)}(t) = F(t) + \alpha u(t) \tag{3}$$

which is continuously updated, where

- the constant coefficient α is chosen by the practitioner, such that αu and $y^{(\nu)}$ are of the same order of magnitude;
- the time-varying function $F(t)$, which is estimated thanks to the knowledge of u and y, subsumes the structural properties;
- the order ν of derivation is always ≤ 2.

The loop may be closed, if $\nu = 1$ for example, via an *intelligent* PI controller, or i-PI,

$$u = -\frac{F}{\alpha} + \frac{\dot{y}^\star}{\alpha} + K_P e + K_I \int e \tag{4}$$

where $y^\star$ is the output reference trajectory, $e = y - y^\star$ is the tracking error, K_P, K_I are the usual PI gains.

2.2 Identification and estimation issues

Consider

$$L\left(\frac{d}{dt}\right) z = \phi + \alpha u$$

where

- ϕ and α are unknown constants,
- $L(\frac{d}{dt}) \in \mathbb{R}[\frac{d}{dt}]$ is a linear differential operator with constant coefficients,
- $\frac{d^\nu z}{dt^\nu} = y$ for some $\nu \geq 0$.

We thus approximate an unknown function like F in Equation (3) by a piecewise constant one. The above equation reads in the operational domain:

$$L(s)Z = \frac{\phi}{s} + \alpha U + I(s) \tag{5}$$

where $I \in \mathbb{R}[s]$ is a polynomial associated to the initial conditions. For $N \geq 1$ sufficiently large, $\frac{d^N I}{ds^N} \equiv 0$. Multiplying both sides of equation (5) by $\frac{d^N}{ds^N}$ permits to get rid of the initial conditions.

$$\frac{d^N}{ds^N} L(s)Z = \frac{(-1)^N N!}{s^{N+1}}\phi + \alpha \frac{d^N U}{ds^N} \tag{6}$$

Rewriting Equation (6) with $N_1 \neq N$ yields the *linear identifiability* (Fliess & Sira-Ramírez [2003, 2008]) of ϕ and α. Multiplying both sides of Equations (6) by s^{-M}, where $M \geq 0$ is sufficiently large, permits to get rid of positive powers of s, i.e., of derivatives with respect to time. The corresponding formulae in the time domain are easily deduced thanks to the correspondence between $\frac{d}{ds}$ and the multiplication by $-t$ in the time domain. Note that the above computations together with the sampling provides a linear discrete filter which is easily implemented for real time applications.

Remark 1. See also Michel [2010] for another approach of the estimation of α.

3. SYNTHESIZING THE CONTROLLER WITH THE KNOWLEDGE OF THE RELATIVE DEGREE

Rewrite Equation (2) as follows:

$$y^{(r)}(t) + \sum_{i=-m}^{r-1} a_i y^{(i)}(t) = h(x,t) + g(x)u(t) \tag{7}$$

where $h(x,t) = f(x) + \sum_{i=-m}^{r-1} a_i y^{(i)}$, $m \in \mathbb{N}$, $y^{(-m)} = \underbrace{\int \cdots \int}_{m} y$. The left hand side of Equation (7) is a Hurwitz polynomial, which is related to an obvious stabilization procedure.

Apply the techniques sketched in Section 2.2 in order to obtain estimates of piecewise constant approximations F and α of h and g in Equation (7). [4] Then

$$\boxed{u = -\frac{F}{\alpha}}$$

defines a stabilizing controller.

4. NUMERICAL SIMULATIONS

The two systems below are *flat* (Fliess *et al.* [1995], Lévine [2009], Sira-Ramírez *et al.* [2004]). We know moreover

[3] See in the same proceedings Fliess *et al.* [2011b] and the references therein for more details.

[4] A lack of space prevents us from reproducing detailed calculations, which show that a rough estimate of g is sufficient.

that flatness-based control provides a most efficient control strategy. Let us therefore emphasize again that our only aim here is a comparison between sliding modes and model-free control.

4.1 A kinematic car model

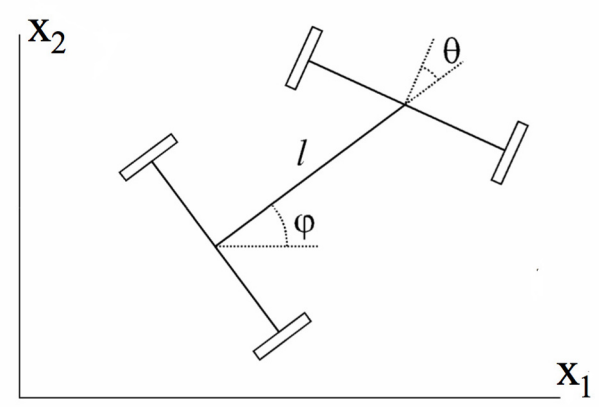


Fig. 1. A car in Cartesian coordinates

The kinematic car model, which is borrowed from Levant & Alelishvili [2008], is depicted in Figure 1. It satisfies the following equations:

$$\dot{x}_1 = v\cos\varphi \quad \dot{x}_2 = v\sin\varphi,$$
$$\dot{\varphi} = \frac{v}{l}\tan\theta \quad \dot{\theta} = u$$

The variables x_1 and x_2 represent the Cartesian coordinates of the rear-axle middle point, [5] φ is the orientation angle, while v is the longitudinal velocity. The parameter l measures the distance between the two axles and θ is the steering angle. The control task is to drive the car toward a reference trajectory of the form $x_2 = \bar{g}(x_1)$ where x_1 and x_2, i.e. $\bar{g}(x_1)$, are measured. Let $y = s(x_1, x_2) = x_2 - \bar{g}(x_1)$, $v = 10m/s$, $l = 5m$, $\bar{g}(x_1) = 10\sin(0.05x_1) + 5$. Initial conditions are set to 0. Levant & Alelishvili [2008] utilizes the following controller:

$$u = -\frac{z_2 + 2(|z_1| + |z_0|^{2/3})^{-1/2}(z_1 + |z_0|^{2/3}\mathrm{sign}(z_0))}{|z_2| + 2(|z_1| + |z_0|^{2/3})^{1/2}}.$$

The variables z_0, z_1 and z_2 are respectively the filtered values of the noisy output y and of its derivatives of first and second orders. They are the outputs of the following second order differentiator:

$$\dot{z}_0 = \nu_0, \quad \nu_0 = -14.7361|z_0 - \sigma|^{2/3}\mathrm{sign}(z_0 - \sigma) + z_1,$$
$$\dot{z}_1 = \nu_1, \quad \nu_1 = -30|z_1 - \nu_0|^{1/2}\mathrm{sign}(z_1 - \nu_0) + z_2$$
$$\dot{z}_2 = -440\ \mathrm{sign}(z_2 - \nu_1).$$

As indicated in Levant & Alelishvili [2008], a step sampling of 0.0001 second is used to simulate the sliding mode control using the Euler method.

Set, for implementing our model-free control, $m = 0$ and $r = 3$ in Equation (7). The parameters a_0, a_1 and a_2 of Equation (7) are chosen according to a classical third order dynamics $(s - a)(s^2 + 2\xi\omega_n s + \omega_n^2) = 0$.

Simulation with a Brownian noise. Consider first that the measurements of x_1 and x_2 are altered by an additive Brownian noise [6] a realization of which is shown in the Figure 2. According to Figure 3 model-free control provides better tracking performances. Moreover, the control inputs depicted in the Figures 4 and 5 show that the model-free control input is smoother. Softer steering angle is also provided by the model-free controller (see Figures 6 and 7).

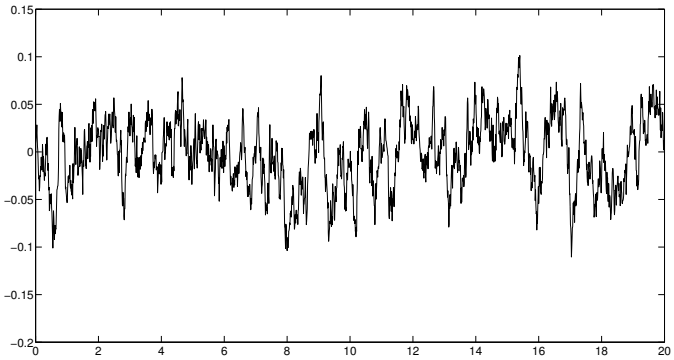


Fig. 2. Brownian noise realization used in the simulations

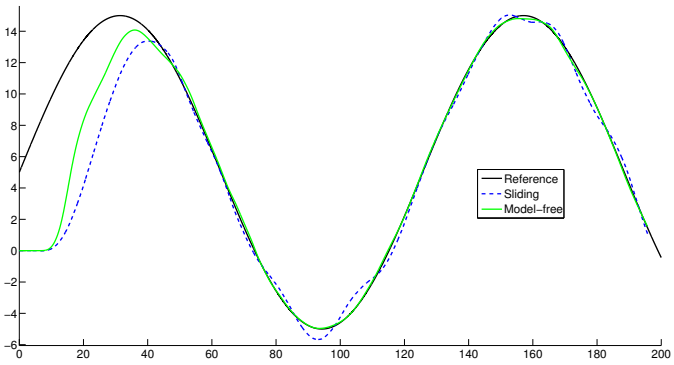


Fig. 3. Output tracking for both sliding mode and i-PID controllers

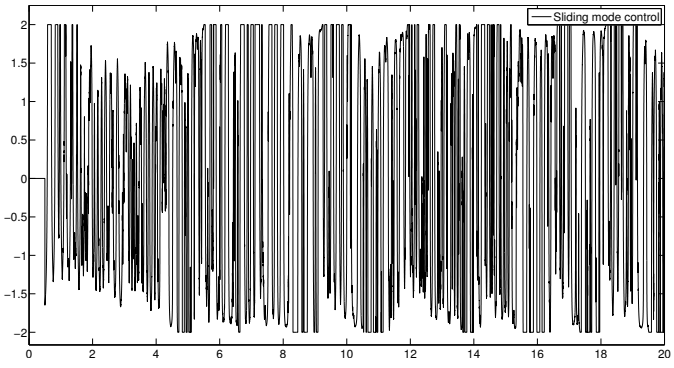


Fig. 4. Sliding mode control input

[5] Remember that this middle point is a *flat output* (Fliess *et al.* [1995], Lévine [2009], Sira-Ramírez *et al.* [2004]).

[6] A Brownian noise is the integral of a Gaussian white noise. In the simulation, we used a low-pass filtered band-limited Gaussian white noise.

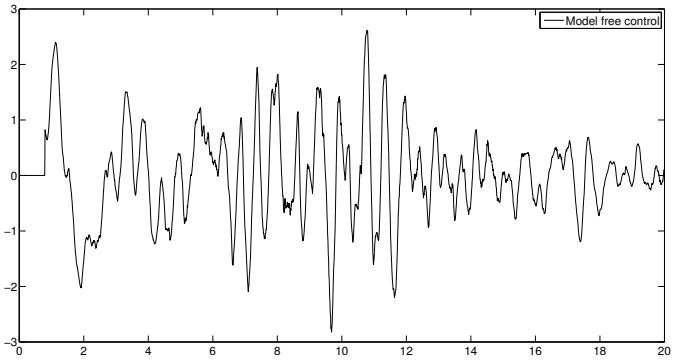


Fig. 5. Model-free control input

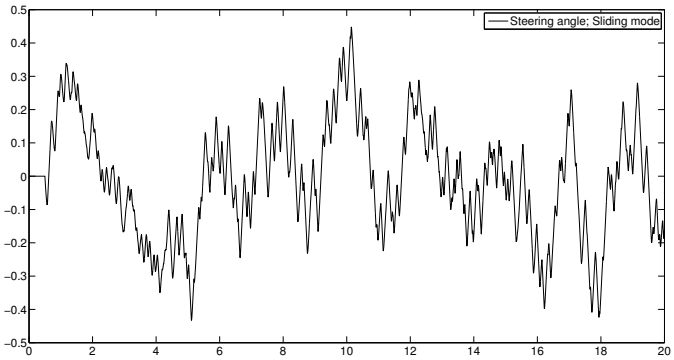


Fig. 6. Steering angle θ corresponding to the sliding mode controller

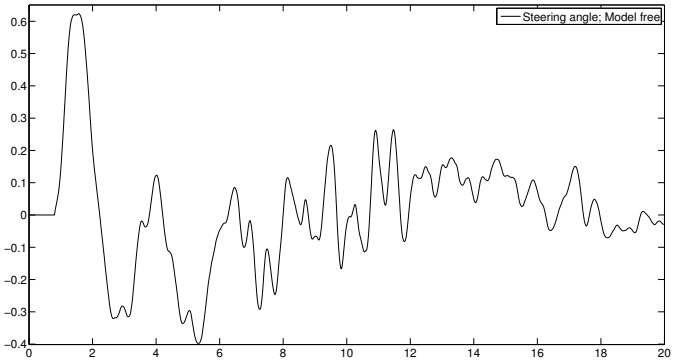


Fig. 7. Steering angle θ corresponding to the model-free controller

Simulation with a Gaussian white noise. Simulations with a zero-mean additive (band-limited) white noise of variance of 0.2, are depicted in the Figures 8, 9 and 10. They also show that the model-free controller performances are better: we achieve a better tracking without any chattering.

Remark 2. The chattering in Figure (4) can be eliminated by raising the relative degree. It would have complicated both the controller as well as the differentiator design. A well known solution is to consider $\dot{u}$ as a virtual input.

Remark 3. We took care to tune the parameters such that the control magnitudes of Figures 9 and 10 are similar. Compare also Figures 4 and 5.

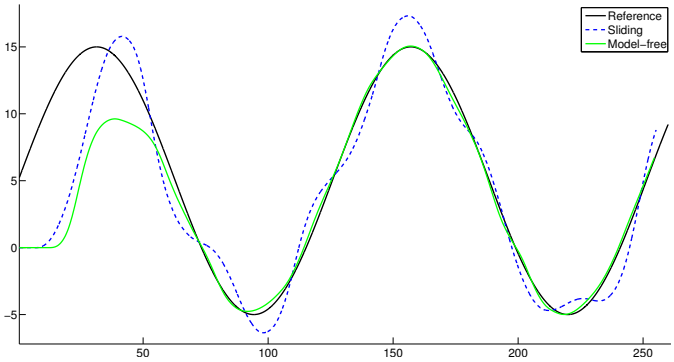


Fig. 8. Output tracking for both sliding mode and i-PID controllers

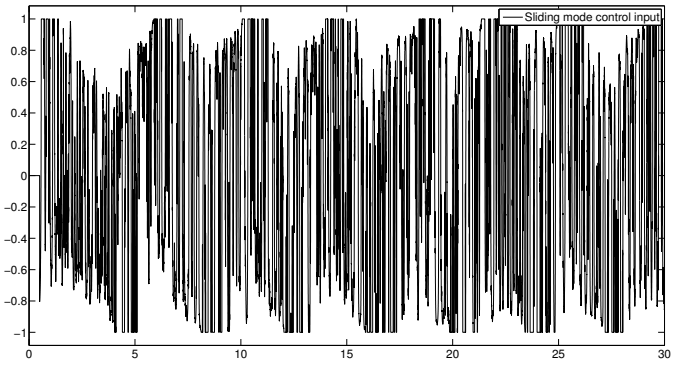


Fig. 9. Sliding mode control input

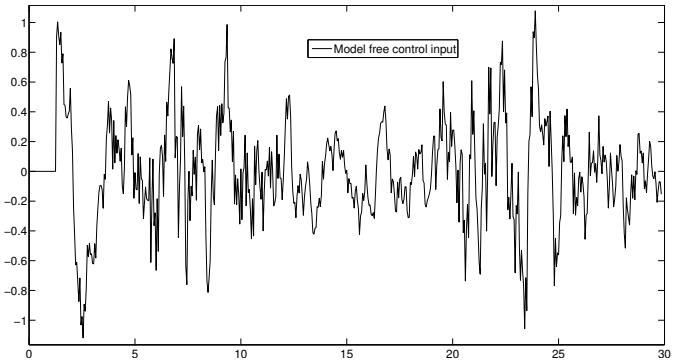


Fig. 10. Model-free control

4.2 A three axis robot

Figure (11) depicts a robotic manipulator with 3 arms and 3 input torques. It satisfies the equation

$$\ddot{q} = M(q)^{-1}\left(G(q) + H(\dot{q}, q) + u\right)$$

where $q = [q_1, q_2, q_3]^T$, $M(q)$ is the symmetric inertia matrix, $u = [u_1, u_2, u_3]^T$ are the control torques. We force the robot's end effector to draw a circle in a horizontal plane with constant angular velocity. It yields the reference trajectories $q_1^\star = 2\pi t$ while $q_2^\star$ and $q_3^\star$ are constant. Take for this purpose the model-free controller $\ddot{e} + K_d\dot{e} + K_P e = F(t) + \alpha u$, where $m = 0$ and $r = 2$, α is a 3×3 matrix, and $e = q - q^\star$. Figures 12, 13, 14, 15 and 16 illustrates the effectiveness of the straightforward controller $u = \alpha^{-1}F$.

Remark 4. Extending model-free control to some MIMO systems is straightforward. It does not seem to be the case for sliding modes.

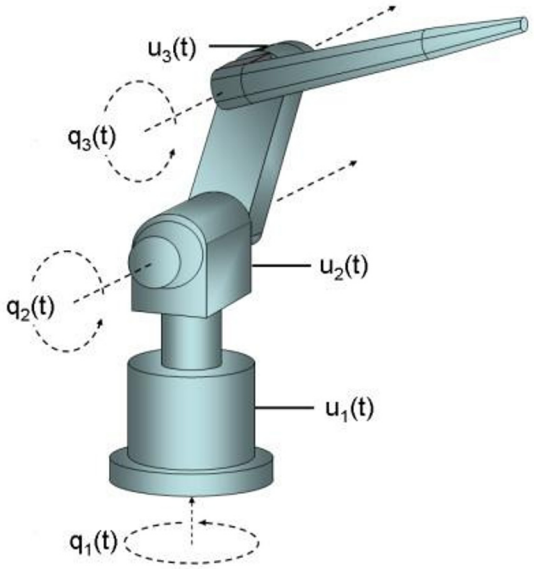


Fig. 11. Manutec R3 robot

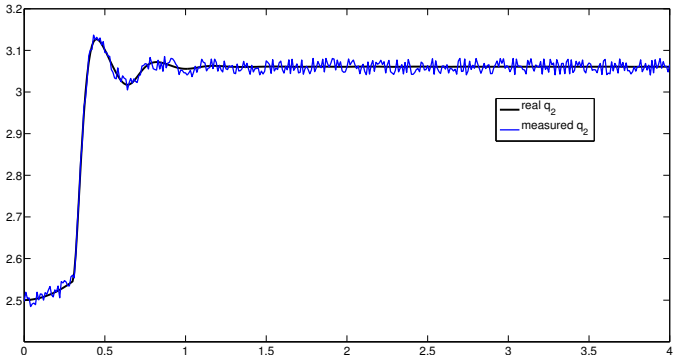


Fig. 12. Real and measured variable q_2.

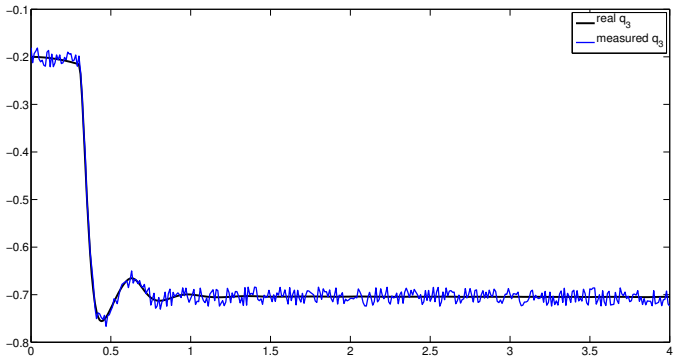


Fig. 13. Real and measured variable q_3.

5. CONCLUSION

Our two case-studies lead to the following observations, which need further theoretical and practical investigations in order to be fully confirmed:

- The model-free controllers yield smooth control variable, without any chattering.

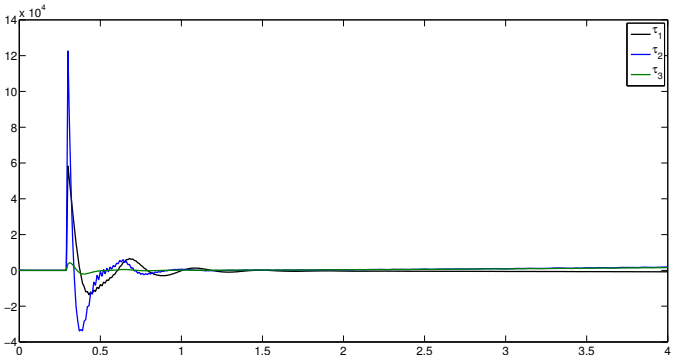


Fig. 14. Control inputs.

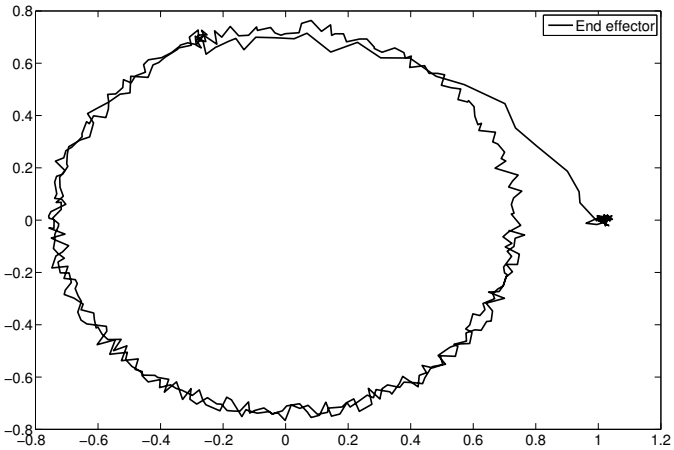


Fig. 15. Measured robot's end effector position (circular trajectory).

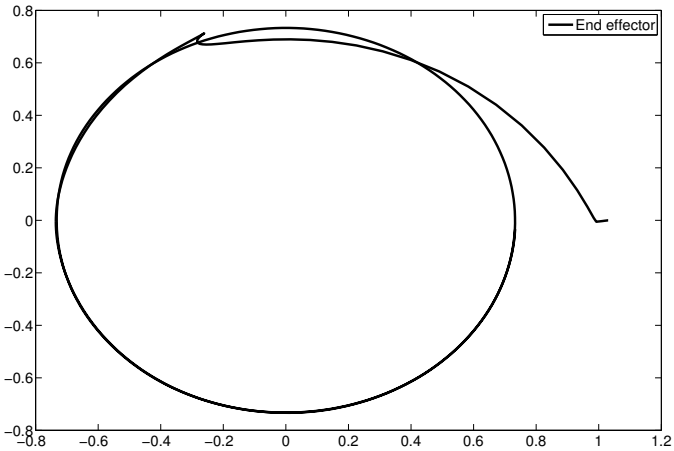


Fig. 16. Real (noise-free) robot's end effector position (circular trajectory).

- The implementation of model-free control only necessitates a standard discrete linear filter, whereas a rather complex derivation of noisy signals is used for sliding modes, where, moreover, the calculation of fractional powers is quite demanding.
- The choice of the model-free controller parameters is obvious.
- The model-free controller seem to be more robust with respect to corrupting noises.
- It is straightforward to extend model-free control to some MIMO systems.

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