APPLICATION OF FRACTIONAL CALCULUS IN THE DYNAMICAL ANALYSIS AND CONTROL OF MECHANICAL MANIPULATORS

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Abstract

Fractional Calculus (FC) goes back to the beginning of the theory of differential calculus. Nevertheless, the application of FC just emerged in the last two decades. In the field of dynamical systems theory some work has been carried out but the proposed models and algorithms are still in a preliminary stage of establishment. This article illustrates several applications of fractional calculus in robot manipulator path planning and control.

Mathematics Subject Classification: 26A33, 93C83, 93C85, 68T40

Key Words and Phrases: fractional calculus, robotics, modeling, dynamics

1. Introduction

Fractional calculus (FC) is a natural extension of the classical mathematics but, in spite of the work that has been done recently, the application of FC in the analysis and control of dynamical systems is still reduced. In this line of thoughts this paper addresses the application of FC in the robotic manipulating systems and is organized as follows. Section 2 studies the trajectory planning of redundant manipulators through the adoption of the pseudo-inverse of the jacobian and its influence of the fractional dynamics that occurs in the joint variables. Section 3 analyzes the spectra of several signals in industrial manipulators that exhibit vibrations due to a
non-ideal mechanical structure and to impacts of gripper. Sections 4 and 5 investigate the performance of fractional order algorithms in the position and force control of one arm in contact with a surface and two cooperating arms, respectively. Finally, Section 6 draws the main conclusions.

2. Trajectory control of redundant manipulators

A kinematically redundant manipulator is a robotic arm possessing more degrees of freedom (dof) than those required to establish an arbitrary position and orientation of the gripper. Redundant manipulators offer several potential advantages over non-redundant arms. In a workspace with obstacles, the extra degrees of freedom can be used to move around or between obstacles and thereby to manipulate in situations that otherwise would be inaccessible, see [1]–[4]. When a manipulator is redundant, it is anticipated that the inverse kinematics admits an infinite number of solutions. This implies that, for a given location of the manipulator’s gripper, it is possible to induce a self-motion of the structure without changing the location of the end effector. Several kinematic techniques for redundant manipulators control the gripper through the rates at which the joints are driven, using the pseudoinverse of the Jacobian [3], [5]. Nevertheless, these algorithms lead to a kind of chaotic motion with unpredictable arm configurations.

A kinematically redundant manipulator is a robotic arm possessing more dof than those required to establish an arbitrary position and orientation of the gripper. In Fig. 1 is depicted a planar manipulator with \( k \in \mathbb{N} \) rotational (R) joints that is redundant for \( k > 2 \). Therefore, redundant manipulators can be reconfigured to find better postures for an assigned set of task requirements but, on the other hand, have a more complex structure requiring adequate control algorithms.

We consider a manipulator with \( n \) degrees of freedom whose joint variables are denoted by \( \mathbf{q} = [q_1, q_2, ..., q_n]^T \). We assume that a class of tasks we are interested in can be described by \( m \) variables, \( \mathbf{x} = [x_1, x_2, ..., x_m]^T \) (\( m < n \)) and that the relation between \( \mathbf{q} \) and \( \mathbf{x} \) is given by:

\[
\mathbf{x} = f(\mathbf{q}),
\]

(1)

where \( f \) is a function representing the direct kinematics. Differentiating (1) with respect to time yields:

\[
\dot{\mathbf{x}} = \mathbf{J}(\mathbf{q})\dot{\mathbf{q}},
\]

(2)

where \( \dot{\mathbf{x}} \in \mathbb{R}^m \) and \( \mathbf{J}(\mathbf{q}) = \frac{\partial f(\mathbf{q})}{\partial \mathbf{q}} \in \mathbb{R}^{m \times n} \). Hence, it is possible to calculate a path \( \mathbf{q}(t) \) in terms of a prescribed trajectory \( \mathbf{x}(t) \) in the operational space. We assume that the following condition is satisfied:
max \quad \text{rank}\{\mathbf{J}(\mathbf{q})\} = m. \quad (3)

Failing to satisfy this condition usually means that the selection of manipulation variables is redundant and the number of these variables $m$ can be reduced. When condition (3) is verified, we say that the degree of redundancy of the manipulator is $n - m$. If, for some $\mathbf{q}$ we have:

$$\text{rank}\{\mathbf{J}(\mathbf{q})\} < m, \quad (4)$$

then the manipulator is in a singular state. This state is not desirable because, in this region of the trajectory, the manipulating ability is very limited.

Many approaches for solving redundancy [5]–[7] are based on the inversion of equation (2). A solution in terms of the joint velocities is sought as:

$$\dot{\mathbf{q}} = \mathbf{J}^\#(\mathbf{q})\dot{\mathbf{x}}, \quad (5)$$

where $\mathbf{J}^\#$ is one of the generalized inverses of the $\mathbf{J}$, see [7]–[9]. It can be easily shown that a more general solution to equation (2) is given by:

$$\dot{\mathbf{q}} = \mathbf{J}^+(\mathbf{q})\dot{\mathbf{x}} + [\mathbf{I} - \mathbf{J}^+(\mathbf{q})\mathbf{J}(\mathbf{q})]\dot{\mathbf{q}}_0, \quad (6)$$

where $\mathbf{I}$ is the $n \times n$ identity matrix and $\dot{\mathbf{q}}_0$ is a $n \times 1$ arbitrary joint velocity vector and $\mathbf{J}^+$ is the pseudoinverse of the $\mathbf{J}$. The solution (6) is composed of two terms. The first term is relative to minimum norm joint velocities. The second term, the homogeneous solution, attempts to satisfy the additional constraints specified by $\dot{\mathbf{q}}_0$. Moreover, the matrix $\mathbf{I} - \mathbf{J}^+(\mathbf{q})\mathbf{J}(\mathbf{q})$ allows the projection of $\dot{\mathbf{q}}_0$ in the null space of $\mathbf{J}$. A direct consequence
is that it is possible to generate internal motions that reconfigure the manipulator structure without changing the gripper position and orientation, [7]–[9]. Another aspect revealed by the solution of (5) is that repetitive trajectories in the operational space do not lead to periodic trajectories in the joint space. This is an obstacle for the solution of many tasks because the resultant robot configurations have similarities with those of a chaotic system.

The direct kinematics and the Jacobian of a 3-link planar manipulator with rotational joints (3R–robot) has a simple recursive nature according with the expressions:

\[
\begin{bmatrix}
x \\
y
\end{bmatrix} = \begin{bmatrix}
l_1C_1 + l_2C_{12} + l_3C_{123} \\
l_1S_1 + l_2S_{12} + l_3S_{123}
\end{bmatrix},
\]

\[
J = \begin{bmatrix}
-l_1S_1 - ... - l_3S_{123} ... - l_3S_{123} \\
l_1C_1 + ... + l_3C_{123} ... + l_3C_{123}
\end{bmatrix},
\]

where \(l_i\) is the length of link \(i\), \(q_{i,k} = q_i + ... + q_k\), \(S_{i,k} = Sin(q_{i,k})\) and \(C_{i,k} = Cos(q_{i,k})\). During all the experiments it is considered \(\Delta t = 10^{-3}\) sec, \(L_{TOT} = l_1 + l_2 + l_3 = 3\) and \(l_1 = l_2 = l_3\). In the closed-loop pseudoinverse’s method the joint positions can be computed through the time integration of the velocities according with the block diagram of the inverse kinematics algorithm depicted in Fig. 2 where \(x_{ref}\) represents the vector of reference coordinates of the robot gripper in the operational space. Based on equation (8) we analyze the kinematic performances of the 3R–robot when repeating a circular motion in the operational space with frequency \(\omega_0 = 7.0\) rad sec\(^{-1}\), centre at distance \(r = [x^2 + y^2]^{1/2}\) and radius \(\rho\). Fig. 3 shows the joint positions for the inverse kinematic algorithm (5) for \(r = \{0.6, 2.0\}\) and \(\rho = \{0.3, 0.5\}\). We observe that:

- For \(r = 0.6\) occur unpredictable motions with severe variations that lead to high joint transients [10]. Moreover, we verify a low frequency signal modulation that depends on the circle being executed.

- For \(r = 2.0\) the motion is periodic with frequency identical to \(\omega_0 = 7.0\) rad sec\(^{-1}\).

Previously, we verified that the pseudoinverse based algorithm leads to unpredictable arm configurations. In order to gain further insight into the pseudoinverse nature several distinct experiments are devised in the sequel during a time window of 300 cycles. Therefore, in a first set of experiments we calculate the Fourier transform of the 3R–robot joints velocities
for a circular repetitive motion with frequency $\omega_0 = 7.0 \text{ rad sec}^{-1}$, radius $\rho = \{0.3, 0.5\}$ and radial distances $r \in [0, L_{TOT}-\rho]$. Fig. 4 shows $|F\{q_2(t)\}|$ versus the frequency ratio $\omega_0/\omega$ and the distance $r$, where $F\{}$ represents the Fourier operator. It is verified an interesting phenomenon induced by the gripper repetitive motion $\omega_0$ because a large part of the energy is distributed along several sub-harmonics. These fractional order harmonics ($foh$) depend on $r$ and $\rho$ making a complex pattern with similarities with those revealed by chaotic systems. Joints 1 and 3 show similar velocity spectra. In the authors’ best knowledge, the $foh$ are aspects of fractional dynamics [11]–[13], but a final and assertive conclusion about a physical
interpretation is a matter still to be explored.

Figure 4: $|F(q_2(t))|$ of the 3R–robot during 300 cycles, vs $r$ and $\omega/\omega_0$, for $\rho = \{0.3, 0.5\}$, $\omega_0 = 7.0$ rad sec$^{-1}$.

3. Manipulators with vibrations

This section presents a fractional calculus perspective in the study of the robotic signals captured during an impact phase of the manipulator. In the experiment is used a steel rod flexible link. To test impacts, the link consists on a long, thin, round, flexible steel rod clamped to the end-effector of the manipulator. The robot motion is programmed in a way such that the rod moves against a rigid surface. During the motion of the manipulator the clamped rod is moved by the robot against a rigid surface. An impact occurs
and several signals are recorded with a sampling frequency of $f_s = 500$ Hz. In order to analyze the vibration and impact phenomena an acquisition system was developed [14]. The instrumentation system acquires signals from multiple sensors that capture the axis positions, mass accelerations, forces and moments and electrical currents in the motors. Afterwards, an analysis package, running off-line, reads the data recorded by the acquisition system and examines them. Due to space limitations only some of the signals are depicted. A typical time evolution of the electrical currents of robot axis motors is shown in Fig. 5 corresponding to: (i) the impact of the rod on a rigid surface, and (ii) without impact [15]. In this example, the signals present clearly a strong variation at the instant of the impact that occurs, approximately, at $t = 4$ sec. In order to study the behavior of the signal Fourier transform, a trendline can be superimposed over the spectrum based
on a power law approximation:

\[ |F\{f(t)\}| \approx c\omega^m, \quad (9) \]

where \( F\{ \} \) is the Fourier operator, \( c \in \mathbb{R}^+ \), \( \omega \) is the frequency and \( m \in \mathbb{R} \) is the slope.

**Fig. 6** shows the amplitude of the Fast Fourier Transform (FFT) of the axis 1 position signal. The trendline (9) leads to a slope \( m = -0.99 \) revealing, clearly, the integer order behavior. The others position signals were studied, revealing also an integer behavior, both under impact and no impact conditions. **Fig. 7** shows the amplitude of the FFT of the electrical current for the axis 3 motor. The spectrum was also approximated by trendlines in a frequency range larger than one decade. These trendlines (**Fig. 7**) have slopes of \( m = -1.52 \) and \( m = -1.51 \) under impact (i) and without impact (ii) conditions, respectively. The lines present a fractional order behavior in both cases. The others axis motor currents were studied, as well. Some of them, for a limited frequency range, present also fractional order behavior while others have a complicated spectrum difficult to approximate by one trendline. **Fig. 8** shows, as example, the spectrum of the \( F_z \) force. This spectrum is not so well defined in a large frequency range. All force/moments and acceleration spectra present identical behavior and, therefore, it is difficult to define accurately the behavior of the signals.

As shown in the examples, the Fourier spectrum of several signals, captured during an impact phase of the manipulator, presents a non integer be-
Figure 7: Spectrum of the axis 3 motor current.

Figure 8: $F_z$ force spectrum with impact.
behavior. On the other hand, the feedback fractional order systems, due to the
success in the synthesis of real noninteger differentiator and the emergence
of fractional-order controllers, have been designed and applied to control
a variety of dynamical processes [16]. Therefore the study presented here
can assist in the design of the control system to be used in eliminating or
reducing the effect of vibrations.

4. Manipulator position/force control

Raibert and Craig [19] introduced the concept of force control based on
the hybrid algorithm and, since then, several researchers developed those
ideas and proposed other schemes [20]. There are two basic methods for
force control, namely the hybrid position/force and the impedance schemes.
The first method separates the task into two orthogonal sub-spaces corre-
sponding to the force and the position controlled variables. Once established
the subspace decomposition two independent controllers are designed. The
second method [20] requires the definition of the arm mechanical impedance.
The impedance accommodates the interaction forces that can be controlled
to obtain an adequate response. The dynamical equation of a \( n \) \( \text{dof} \) robot
is:

\[
\tau = \tau = C(q, \dot{q}) + G(q) + H(q)\ddot{q} - J^T(q)F
\]

where \( \tau \) is the \( n \times 1 \) vector of actuator torques, \( q \) is the \( n \times 1 \) vector of joint
coordinates, \( H(q) \) is the \( n \times n \) inertia matrix, \( C(q, \dot{q}) \) is the \( n \times 1 \) vector of centrifugal / Coriolis terms and \( G(q) \) is the \( n \times 1 \) vector of gravitational
effects. The matrix \( J^T(q) \) is the transpose of the Jacobian and \( F \) is the force
that the load exerts in the robot gripper. For a 2\( R \)–robot the dynamics
yields:

\[
C(q, \dot{q}) = \begin{bmatrix}
-m_2 r_1 r_2 S_2 \dot{q}_2^2 - 2m_2 r_1 r_2 S_2 \dot{q}_1 \dot{q}_2 \\
m_2 r_1 r_2 S_2 \dot{q}_1^2
\end{bmatrix}
\]

(11)

\[
G(q) = \begin{bmatrix}
g(m_1 r_1 C_1 + m_2 r_1 C_1 + m_2 r_2 C_{12}) \\
gm_2 r_2 C_{12}
\end{bmatrix}
\]

(12)

\[
J^T(q) = \begin{bmatrix}
-r_1 S_1 - r_2 S_{12} & r_1 C_{11} + r_2 C_{12} \\
-r_2 S_{12} & r_2 C_{12}
\end{bmatrix}
\]

(13)

\[
H(q) = \begin{bmatrix}
(m_1 + m_2) r_1^2 - m_2 r_2^2 & m_2 r_2^2 + m_2 r_1 r_2 C_2 \\
+2m_2 r_1 r_2 C_2 + J_{1m} + J_{1g} & m_2 r_2^2 + J_2 g \\
-m_2 r_2^2 + m_2 r_1 r_2 C_2 & m_2 r_2^2 + J_2 m + J_{2g}
\end{bmatrix}
\]

(14)
where $C_{ij} = \cos(q_i + q_j)$ and $S_{ij} = \sin(q_i + q_j)$.

The numerical values adopted for the $2R$-robot \cite{21} are $m_1 = 0.5$ kg, $m_2 = 6.25$ kg, $r_1 = 1.0$ m, $r_2 = 0.8$ m, $J_{1m} = J_{2m} = 1.0$ kgm$^2$ and $J_{1g} = J_{2g} = 4.0$ kgm$^2$. The constraint plane is determined by the angle $\theta$ (Fig. 9) and the contact displacement $x_c$ of the robot gripper with the constraint surface is modeled through a linear system with a mass $M$, a damping $B$ and a stiffness $K$ with dynamics:

$$F_c = M \ddot{x}_c + B \dot{x}_c + K x_c$$

(15)

In order to study the dynamics and control of one robot we adopt the position/force hybrid control with the implementation of the integer order and fractional-order algorithms \cite{13}, \cite{16}, \cite{17}, \cite{18}, \cite{22}, \cite{23}. The system performance and robustness is analyzed in the time domain. The effect of dynamic backlash and flexibility is also investigated.

![Figure 9: The 2R–robot and the constraint surface.](image)

The structure of the position/force hybrid control algorithm is depicted in Fig. 10. The diagonal $n \times n$ selection matrix $S$ has elements equal to one (zero) in the position (force) controlled directions and $I$ is the $n \times n$ identity matrix. In this paper the $y_c$, $(x_c)$ cartesian coordinate is position (force) controlled, yielding:

$$S = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad J_c(q) = \begin{bmatrix} -l_1 C_{\theta 11} - l_2 C_{\theta 12} & -l_2 C_{\theta 12} \\ l_1 S_{\theta 11} + l_2 S_{\theta 12} & +l_2 S_{\theta 12} \end{bmatrix}$$

(16)

We now analyze the system performance both for ideal transmissions and robots with dynamic phenomena at the joints, such as backlash and flexibility. Moreover, we compare the response of $FO$ and the PD: $C_P(s) = K_p$.
+ K_d s and PI: \( C_F(s) = K_p + K_i \ s^{-1} \) controllers, in the position and force loops. Both algorithms were tuned by trial and error having in mind getting a similar performance in the two cases. The resulting parameters were \( \{K_P, \ \alpha_P\} \equiv \{10^5, \ 1/2\} \), \( \{K_F, \ \alpha_F\} \equiv \{10^3, \ 1/5\} \) and \( PD - PI: \) \( \{K_p, K_i\} \equiv \{10^3, \ 10^2\} \) for the position and force loops, respectively. Moreover, it is adopted the operating point \( \{x, y\} \equiv \{1, \ 1\} \), a constraint surface with parameters \( \{\theta, M, B, K\} \equiv \{\pi/2, \ 10^3, \ 1.0, \ 10^2\} \) and a controller sampling frequency \( f_c = 1 \ kHz \). In order to study the system dynamics we apply, separately, rectangular pulses, at the position and force references, that is, we perturb the references with \( \{y_{cd}, F_{cd}\} = \{10^{-1}, \ 0\} \) and \( \{y_{cd}, F_{cd}\} = \{0, \ 10^{-1}\} \). \textbf{Fig. 11} depicts the time response of the 2\( R \)-robot under the action of the \( FO \) and the \( PD - PI \) controllers for ideal transmissions at the joints.

In a second phase (\textbf{Fig. 12}) we analyze the response of robots with dynamic backlash at the joints. For the \( i^{th} \) joint \( (i = 1, 2) \), with gear clearance \( h_i \), the backlash reveals impact phenomena between the inertias, which obey the principle of conservation of momentum and the Newton law:

\[
\dot{q}_i = \frac{\dot{q}_i (J_{ii} - \varepsilon J_{im}) + \dot{q}_{im} J_{im} (1 + \varepsilon)}{J_{ii} + J_{im}}, \tag{17}
\]

\[
\dot{q}_{im} = \frac{\dot{q}_i J_i (1 + \varepsilon) + \dot{q}_{im} (J_{im} - \varepsilon J_{ii})}{J_{ii} + J_{im}}, \tag{18}
\]
where $0 \leq \varepsilon \leq 1$ is a constant that defines the type of impact ($\varepsilon = 0$ inelastic impact, $\varepsilon = 1$ elastic impact) and $\dot{q}_i$ and $\dot{q}_{im}$ ($\ddot{q}_i$ and $\ddot{q}_{im}$) are the velocities of the $i^{th}$ joint and motor before (after) the collision, respectively. The parameter $J_{ii}$ ($J_{im}$) stands for the link (motor) inertias of joint $i$. In the simulations is adopted $h_i = 1.8 \times 10^{-4}$ rad and $\varepsilon_i = 0.8$.

In a third phase (Fig. 13) it is studied the case of compliant joints, where the dynamic model corresponds to (10) augmented by the equations:

$$
T = J_m \ddot{q}_m + B_m \dot{q}_m + K_m (q_m - q),
$$

(19)

$$
K_m (q_m - q) = J(q) \ddot{q} + C(q, \dot{q}) + G(q),
$$

(20)

where $J_m$, $B_m$ and $K_m$ are the $n \times n$ diagonal matrices of the motor and transmission inertias, damping and stiffness, respectively. In the simulations we adopt $K_{mi} = 2.0 \times 10^6$ Nm rad$^{-1}$ and $B_{mi} = 10^4$ Nms rad$^{-1}$ ($i = 1,2$).
Figure 12: Time response for 2R–robot with dynamic backlash at the joints under the action of the FO and PD–PI controllers for pulses perturbations $\delta y_{cd} = 10^{-1}$ m and $\delta Fy_{cd} = 10^{-1}$ Nm at the robot position and force references, respectively.

Table 1: Time response for a pulse $\delta y_{cd}$ at the robot reference.

<table>
<thead>
<tr>
<th>No</th>
<th>$C(s)$</th>
<th>FO%</th>
<th>$\epsilon_x$[m]</th>
<th>$T_p$[s]</th>
<th>$T_s$[s]</th>
</tr>
</thead>
<tbody>
<tr>
<td>ideal</td>
<td>PD – PI</td>
<td>23.48</td>
<td>$99.0 \times 10^{-4}$</td>
<td>$0.122 \times 10^{-2}$</td>
<td>$13.0 \times 10^{-2}$</td>
</tr>
<tr>
<td></td>
<td>FO</td>
<td>18.98</td>
<td>$79 \times 10^{-3}$</td>
<td>$3.36 \times 10^{-2}$</td>
<td>$18.0 \times 10^{-2}$</td>
</tr>
<tr>
<td>backlash</td>
<td>PD – PI</td>
<td>0.37</td>
<td>$2.1 \times 10^{-3}$</td>
<td>$38.0 \times 10^{-2}$</td>
<td>$8.0 \times 10^{-2}$</td>
</tr>
<tr>
<td></td>
<td>FO</td>
<td>0.36</td>
<td>$1.4 \times 10^{-4}$</td>
<td>$3.0 \times 10^{-2}$</td>
<td>$11.8 \times 10^{-2}$</td>
</tr>
<tr>
<td>flexible</td>
<td>PD – PI</td>
<td>2.28</td>
<td>$3.9 \times 10^{-3}$</td>
<td>$40.3 \times 10^{-2}$</td>
<td>$15.0 \times 10^{-1}$</td>
</tr>
<tr>
<td></td>
<td>FO</td>
<td>1.8</td>
<td>$1.4 \times 10^{-3}$</td>
<td>$30.2 \times 10^{-2}$</td>
<td>$30.4 \times 10^{-1}$</td>
</tr>
</tbody>
</table>
Table 2: Time response for a pulse $\delta F_{ycd}$ at the robot reference.

<table>
<thead>
<tr>
<th>No</th>
<th>C(s)</th>
<th>$PO%$</th>
<th>$e_{ss}[m]$</th>
<th>$T_p[s]$</th>
<th>$T_s[s]$</th>
</tr>
</thead>
<tbody>
<tr>
<td>ideal</td>
<td>$PD - PI$</td>
<td>22.0</td>
<td>$1.3 \times 10^{-3}$</td>
<td>$8.0 \times 10^{-3}$</td>
<td>$9.1 \times 10^{-2}$</td>
</tr>
<tr>
<td></td>
<td>$FO$</td>
<td>29.50</td>
<td>$1.3 \times 10^{-3}$</td>
<td>$8.9 \times 10^{-3}$</td>
<td>$9.3 \times 10^{-2}$</td>
</tr>
<tr>
<td>backlash</td>
<td>$PD - PI$</td>
<td>5.98</td>
<td>$9.9 \times 10^{-3}$</td>
<td>$4.0 \times 10^{-1}$</td>
<td>$4.0 \times 10^{-2}$</td>
</tr>
<tr>
<td></td>
<td>$FO$</td>
<td>8.6</td>
<td>$9.9 \times 10^{-3}$</td>
<td>$7.9 \times 10^{-2}$</td>
<td>$4.0 \times 10^{-2}$</td>
</tr>
<tr>
<td>flexible</td>
<td>$PD - PI$</td>
<td>3.2</td>
<td>$9.9 \times 10^{-2}$</td>
<td>$6.0 \times 10^{-2}$</td>
<td>$6.0 \times 10^{-1}$</td>
</tr>
<tr>
<td></td>
<td>$FO$</td>
<td>1.8</td>
<td>$9.9 \times 10^{-3}$</td>
<td>$4.0 \times 10^{-1}$</td>
<td>$4.5 \times 10^{-1}$</td>
</tr>
</tbody>
</table>

Figure 13: Time response for $2R$-robot with flexibility at the joints under the action of the $FO$ and $PD - PI$ controllers for pulses perturbations $\delta y_{cd} = 10^{-1}$ m and $\delta F_{ycd} = 10^{-1}$ Nm at the robot position and force references, respectively.

The time responses (Tables 1 and 2), namely the percent overshoot $PO\%$, the steady-state error $e_{ss}$, the peak time $T_p$, and the settling time $T_s$, reveal that, although tuned for similar performances in the first case, the $FO$ is superior to the $PD - PI$ in the cases with dynamical phenomena at the robot joints.

It is clear that the $FO$ demonstrates better performance for joints having nonlinearities.
5. Control of two cooperating manipulators

Two robots carrying a common object are a logical alternative for the case in which a single robot is not able to handle the load. The choice of a robotic mechanism depends on the task or the type of work to be performed and, consequently, is determined by the position of the robots and by their dimensions and structure. In general, the selection is done through experience and intuition; nevertheless, it is important to measure the manipulation capability of the robotic system [24] that can be useful in the robot operation. In this perspective it was proposed the concept of kinematic manipulability [25] and its generalization by including the dynamics [26] or, alters natively, the statistical evaluation of manipulation. Other related aspects such as the coordination of two robots handling objects, collision avoidance and free path planning have been also investigated [27]. With two cooperative robots the resulting interaction forces have to be accommodated and consequently, in addition to position feedback, force control is also required to accomplish adequate performances [28], [29]. We consider two 2\(R\) cooperating manipulators with identical dimensions. The contact of the robot gripper with the load is modeled through a linear system with a mass \(M\), a damping \(B\) and a stiffness \(K\) (Fig. 14).

![Figure 14: The 2\(R\) dual arm robot and the contact between the robot gripper and the object.](image)

The controller architecture (Fig. 15) is inspired on the impedance and compliance schemes. Therefore, we establish a cascade of force and position algorithms as internal an external feedback loops, respectively, where \(x_d\) and \(F_d\) are the payload desired position coordinates and contact forces.

We analyze the system performance both for robots ideal transmissions and robots with dynamic phenomena at the joints, such as backlash and flexibility. Moreover, we compare the response of \(FO\) and classical algorithms...
namely $PD: C_P(s) = K_p (1 + T_d s)$ and $PI: C_F(s) = K_F [1 + (T_i s)^{-1}]$, in the position and force loops, respectively. Both algorithms were tuned by trial and error having in mind getting a similar performance in the two cases. The resulting parameters were $FO: \{K_p, \alpha_P\} \equiv \{10^4, 1/2\}, \{K_F, \alpha_F\} \equiv \{2, 1/5\}$ and $PD – PI: \{K_p, K_d\} \equiv \{10^4, 10^2\}, \{K_p, K_i\} \equiv \{10, 10^4\}$ for the position and force loops, respectively. Moreover, it is adopted the operating point, the center of the object $A \equiv \{x, y\} \equiv \{0, 1\}$ and a object surface with parameters $\{\theta, M, B_j, K_j\} \equiv \{0, 10, 10^2\}$. In order to study the system dynamics we apply, separately, small amplitude rectangular pulses, at the position and force references. Therefore, we perturb the references with $x_d = 10^{-3}$, $y_d = 10^{-3}$, $F_{xd} = 1.0$, $F_{yd} = 1.0$ and we analyze the system performance in the time domain. To evaluate the performance of the proposed algorithms we compare the response for robots with dynamical phenomena at the joints. In all experiments the controller sampling frequency is $f_c = 10$ kHz for the operating point $A$ of the object and a contact force of each gripper of $\{F_{xj}, F_{yj}\} \equiv \{0.5, 5\}$ Nm for the $j$th ($j = 1, 2$) robot. Fig. 16 depicts the time response of the robot $A$, under the action of the $FO$ and the $PD – PI$ algorithms, for robots with ideal transmissions at the joints.

In Figs. 17 and 18 we analyze the response of robots with dynamic backlash and dynamic flexibility at the joints.

The time responses (Tables 3-6), namely the percent overshoot $PO\%$, the steady-state error $e_{ss}$, the peak time $T_p$ and the settling time $T_s$ reveal that, although tuned for almost similar performances in the first case, the $FO$ is superior to the $PD – PI$ algorithms in the cases of robots with joint dynamic phenomena.
Figure 16: The time response of robots with ideal joints under the action of the FO and the PD – PI algorithms for a pulse perturbation at the robot A position reference $\delta y_d = 10^{-3}$ m and a payload $M = 1.0$ kg, $B_i = 1.0$ Ns/m and $K_i = 10^3$ N/m.

| Table 3: Time response for a pulse $\delta y_d$ at the robot A reference. |
|------------------------|--------|--------|--------|--------|
| $\text{No}$ | $C(s)$ | $PO\%$ | $\epsilon_{ss}[m]$ | $T_p[s]$ | $T_s[s]$ |
| ideal | PD – PI | 39.0 | $5.0 \times 10^{-3}$ | $1.1 \times 10^{-2}$ | $25.0 \times 10^{-2}$ |
| | FO | 43.0 | $0.9 \times 10^{-3}$ | $1.6 \times 10^{-2}$ | $15.0 \times 10^{-2}$ |
| backlash | PD – PI | 0.2 | $2.7 \times 10^{-2}$ | $37.0 \times 10^{-2}$ | $5.0 \times 10^{-1}$ |
| | FO | 0.2 | $3.5 \times 10^{-3}$ | $4.0 \times 10^{-2}$ | $4.0 \times 10^{-2}$ |
| flexible | PD – PI | 0.3 | $64.0 \times 10^{-2}$ | $38.0 \times 10^{-2}$ | $45.0 \times 10^{-2}$ |
| | FO | 0.3 | $50.0 \times 10^{-3}$ | $25.0 \times 10^{-2}$ | $19.0 \times 10^{-2}$ |

| Table 4: Time response for a pulse $\delta F_{y_d}$ at the robot A reference. |
|------------------------|--------|--------|--------|--------|
| $\text{No}$ | $C(s)$ | $PO\%$ | $\epsilon_{ss}[m]$ | $T_p[s]$ | $T_s[s]$ |
| ideal | PD – PI | 400.0 | $9.8 \times 10^{-1}$ | $1.1 \times 10^{-2}$ | $2.0 \times 10^{-1}$ |
| | FO | 115.0 | $77.0 \times 10^{-3}$ | $25.0 \times 10^{-2}$ | $5.0 \times 10^{-1}$ |
| backlash | PD – PI | 400.0 | $9.8 \times 10^{-1}$ | $1.1 \times 10^{-2}$ | $2.0 \times 10^{-1}$ |
| | FO | 100.0 | $77.0 \times 10^{-3}$ | $20.0 \times 10^{-2}$ | $4.0 \times 10^{-1}$ |
| flexible | PD – PI | 100.0 | $9.8 \times 10^{-1}$ | $1.1 \times 10^{-2}$ | $1.0 \times 10^{-1}$ |
| | FO | 100.0 | $77.0 \times 10^{-3}$ | $20.0 \times 10^{-2}$ | $4.0 \times 10^{-1}$ |
Figure 17: The time response of robots with joints having backlash under the action of the \( FO \) and the \( PD - PI \) algorithms for a pulse perturbation at the robot A position reference \( \delta y_d = 10^{-3} \) m and a payload \( M = 1 \) kg, \( B_i = 1 \) Ns/m and \( K_i = 10^3 \) N/m.

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4. Conclusions

The recent progress in the area of nonlinear dynamics and chaos promises a large area of application of the theory of $FC$. In the area of dynamical system modeling and control preliminary work has been proposed but many areas are still to be explored. In this paper several aspects of mechanical manipulator analysis and control were addressed, namely, path planning of redundant manipulators, vibration analysis in mechanical systems, and position/force control both on single and dual robotic arms involving several types on nonlinear dynamical phenomena. The results demonstrate that $FC$ constitutes a mathematical tool to be considered for the development of robotic systems with superior performances.

Figure 18: Time response of robots with joints having flexibility under the action of the $FO$ and the $PD – PI$ algorithms, for a pulse perturbation at the robot A position reference $\delta y_d = 10^{-3}$ m and a payload $M = 1$ kg, $B_i = 1$ Ns/m and $K_i = 10^3$ N/m.
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Received: December 20, 2007

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