Stabilization of Inverted Cart-Pendulum System Using PI$^\lambda D^\mu$ Controller: A Frequency-Domain Approach

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This paper focuses on the angular stabilization of inverted cart-pendulum system using PI$^\lambda D^\mu$ controller. The tuning of PI$^\lambda D^\mu$ controller is formulated as a nonlinear optimization problem, in which the objective function is composed of five design conditions in frequency domain. Particle swarm optimization technique has been used for optimizing PI$^\lambda D^\mu$ parameters. Also a PID controller has been designed based on same specifications, and a comparative study has been carried out. All the responses have been calculated using FOMCON toolbox of Matlab/Simulink.

1. Introduction

In recent years, Fractional-order PID (PI$^\lambda D^\mu$) controller has become a good generalization of standard PID controller using fractional calculus. Compared to PID controller, the tuning of PI$^\lambda D^\mu$ is more complex and remains a challenging problem. As the PID controller is one of the most commonly used controllers in industries, and in recent years fractional calculus [1–5] has been used to present the more generalized form of PID controller, that is, PI$^\lambda D^\mu$ controller.

It is well known that inverted cart-pendulum [6] is an example of underactuated, nonminimum phase and highly unstable system. Because of this, it is very difficult to design a controller for a system like inverted pendulum. The design becomes more difficult because of the physical constraints on track length, applied voltage, and the pendulum angle.

Podlubny has proposed a generalization of the PID controller as PI$^\lambda D^\mu$ controller. He also demonstrated that the fractional order PID controller has better response than classical PID controller [1, 7]. Also, many valuable studies have been done for fractional order controllers and their implementations [8–14]. Tuning of the PI$^\lambda D^\mu$ controller using the frequency-domain approaches is studied in many papers. For example, [15] propose a method based on optimization strategies. Tuning of H∞ controllers for fractional single-input single-output (SISO) system was suggested in [12]. In [16], tuning method for PI$^\lambda D^\mu$ based on Ziegler-Nichols and Astrom-Hagglund tuning rules is presented.

In this study, a PI$^\lambda D^\mu$ controller has been designed for an unstable system (inverted cart-pendulum system) using frequency-domain approach, and particle swarm optimization (PSO) has been used for obtaining the PI$^\lambda D^\mu$ parameters ($K_p$, $K_i$, $K_d$, $\lambda$, $\mu$).

This work is organized as follows: Section 2 covers inverted cart-pendulum system, Section 3 covers FOPID controller, Section 4 presents particle swarm optimization, Section 5 presents FOPID controller design, Section 6 presents simulation results, and the conclusion is provided in Section 7.


2. Inverted Cart-Pendulum System

The inverted pendulum system [6] has two degrees of freedom of motions as shown in Figure 1. The first is linear motion of the cart in the X-axis and is given by

$$\dot{\theta} = mL \times \left( \sigma \left\{ F - bx \right\} \cos \theta - mL (\dot{\theta})^2 \cos \theta (m + M) \sin \theta \right)^{-1},$$

(1)

where $\sigma = mL^2(M + m \cos^2 \theta) + J(M + m)$.

The second is rotation of the pendulum about the X-Y plane and is described by

$$\ddot{x} = 1 \times \left( \sigma \left\{ (J + mL^2) \left\{ F - bx - mL \dot{\theta}^2 \sin \theta \right\} + mL^2 g \cos \theta \sin \theta \right\} \right)^{-1}.$$  

(2)

Now linearising (1) and (2) for small angle of $\theta$ from the equilibrium point and neglecting the friction coefficient $b$ (which is very small compared to other parameters), the following equations are obtained:

$$\dot{\theta} = mL \times \left( \sigma' \left\{ F + (m + M) g \theta \right\} \right)^{-1},$$

(3)

where $\sigma' = mL^2 + J(M + m)$

$$\ddot{x} = 1 \times \left( \sigma' \left\{ (J + mL^2) F + mL^2 g \theta \right\} \right)^{-1}.$$  

(4)

Now taking the Laplace Transform of (3) and (4), the following transfer function is obtained:

$$G_1(s) = \frac{\Theta(s)}{F(s)} = \frac{\frac{mLs^2}{s^2 ((J + mL^2) + M^2 \theta^2) \frac{s^2 - mgL (M + m)}}}{s^2 - 6.807)}.$$  

(5)

From Table 1 putting all values in (5),

$$G_1(s) = \frac{0.2638 s^2}{s^2 (s^2 - 6.807)} \approx \frac{0.2638}{s^2 - 6.807}.$$  

(6)

Now the DC motor used to convert control voltage $U$ to force $F$ is represented by only a gain block of gain = 15 for simplicity. Then $G_1(s)$ become

$$G_1(s) = \frac{3.957}{s^2 - 6.807}.$$  

(7)

3. PI^D^μ Controller

The differential equation of the PI^D^μ controller is described by [16, 17]

$$u(t) = K_p e(t) + K_i \int \frac{e(t)}{s^\lambda} + K_d \frac{e(t)}{s^\mu}$$  

(8)

4. Particle Swarm Optimization Method

The PSO method [18] is a wide category of swarm intelligence methods for solving the optimization problems. This technique inspired by bird flocks, fish schools, and animal herds constitute representative examples of natural systems where aggregated behaviors are met, producing impressive, collision-free, and synchronized moves. The PSO is population-based stochastic optimization which was developed by Kennedy and Eberhart in 1995. The major advantage of the PSO over other stochastic optimization methods is its simplicity. It is a population-based search algorithm.
The basic idea of the PSO is the mathematical modeling and simulation of the solution searching activities in multi-dimensional search space where optimal solution exists. In a PSO system, particles fly around in a multi-dimensional search space. During flight, each particle adjusts its position according to its own experience and the experience of the neighboring particles. Update velocity and position of particle are depended upon inertia, cognitive, and social factors. Each particle has a memory, and hence it is capable of remembering the best previous position in the search space ever visited by it. The position corresponding to the best fitness is known as $p_{best}$, and the overall best out of all the particles in the population is called $g_{best}$. The modified velocity and position of each particle can be calculated using the current velocity and position as follows.

Velocity update equation is given by

$$V_i^{k+1} = W \times V_i^k + C_1 \times \text{rand}() \times (p_{best,i} - X_i^k) + C_2 \times \text{rand}() \times (g_{best,i} - X_i^k).$$

Position update equation is given by

$$X_i^{k+1} = X_i^k + V_i^{k+1},$$

where $k$ = number of iterations, $V$ = velocity of particle, $W$ = inertia weight factor = $W_{max} - [(W_{max} - W_{min})/k] \times$ iteration number, $C_1$, $C_2$ = cognitive and social acceleration factors respectively, rand() = random numbers uniformly distributed in the range $(0, 1)$, $x$ = position of particle, and $i$ = $ith$ particle.

As shown in Figure 3, the PSO algorithm consists of just three steps, which are repeated until some stopping condition is met.

1. Evaluate the fitness value of each particle.
2. Update individual and global best fitness values and positions.
3. Update velocity and position of each particle.

The first two steps are fairly trivial. Fitness evaluation is conducted by supplying the candidate solution to the objective function. Individual and global best fitness values and positions are updated by comparing the newly evaluated fitness values against the previous individual and global best fitness values and replacing the best fitness values and positions as necessary.

5. PI\textsuperscript{\(D^\mu\)} Controller Design

The concept of frequency domain design of PI\textsuperscript{\(D^\mu\)} controllers was first proposed at [8]. If $G(s)$ is the transfer function of the process, then the objective is to find out a controller $C(s)$, so that the open loop system $G(s)C(s)$ meets the following design specifications.

5.1. Phase Margin Specification. Consider

$$\text{Arg}(G(j\omega_{gc})C(j\omega_{gc})) = -\pi + \phi_m,$$

where $\phi_m$ is the desired phase margin and $\omega_{gc}$ is the desired gain crossover frequency.

5.2. Gain Crossover Frequency Specification. Consider

$$|G(j\omega_{gc})C(j\omega_{gc})| = 1.$$ 

5.3. Robustness against System's Gain Variation. Consider

$$\left(\frac{d}{d\omega} \text{Arg}[G(j\omega)C(j\omega)]\right)_{\omega = \omega_{gc}} = 0.$$

5.4. Complementary Sensitivity Specification. Consider

$$|T(j\omega)|_{dB} \leq AdB \quad \forall \omega \geq \omega_1 \text{ (rad/s)}$$

where $A$ is the specified magnitude of the complementary sensitivity function or noise attenuation for frequencies for all $\omega \geq \omega_1$ (rad/s).
5.5. Sensitivity Specification. Consider

\[
|S(j\omega)|_{dB} = \left| \frac{1}{1 + G(j\omega)C(j\omega)} \right|_{dB} \leq BdB \quad \forall \omega \leq \omega_s \text{ (rad/s)}
\]  \hspace{1cm} (16)

where \( B \) is the specified magnitude of the sensitivity function or load disturbance rejection for frequencies for all \( \omega \leq \omega_s \) (rad/s). Now five specifications (12)–(16) have been solved using PSO to obtain the PI\textsuperscript{D\textsuperscript{2}} controller parameters, and also a conventional PID controller has been designed using the same specifications.

To formulate the objective function (also known as fitness function), (12) and (13) can be combined as
Inverted cart and pendulum system

\[ G(j\omega_{gc}) C(j\omega_{gc}) = e^{j(-\pi + \phi_m)} \]

\[ = G(j\omega_{gc}) C(j\omega_{gc}) = \cos(-\pi + \phi_m) + j \sin(-\pi + \phi_m) \]

\[ \Rightarrow \text{Re} \{ G(j\omega_{gc}) C(j\omega_{gc}) \} + j \text{Im} \{ G(j\omega_{gc}) C(j\omega_{gc}) \} \]

\[ = - \cos(\phi_m) - j \sin(\phi_m) \]

\[ \Rightarrow \left[ \text{Re} \{ G(j\omega_{gc}) C(j\omega_{gc}) \} + \cos(\phi_m) \right] + j \left[ \text{Im} \{ G(j\omega_{gc}) C(j\omega_{gc}) \} + \sin(\phi_m) \right] = 0. \]

From (18), let

\[ J_1 = \text{Re} \{ G(j\omega_{gc}) C(j\omega_{gc}) \} + \cos(\phi_m), \]

\[ J_2 = \text{Im} \{ G(j\omega_{gc}) C(j\omega_{gc}) \} + \sin(\phi_m). \]

From (14), (15), and (16), let

\[ J_3 = \left( \frac{d}{d\omega} \left( \text{Arg} [G(j\omega) C(j\omega)] \right) \right)_{\omega = \omega_{gc}} \]

\[ J_4 = \left\{ \left| T(j\omega_i) \right| - 10^{A/20} \right\}, \]

\[ J_5 = \left\{ \left| S(j\omega_i) \right| - 10^{B/20} \right\}. \]

Now, the objective function \( J \) which is minimized to satisfy specifications (12)–(16) is given as

\[ J = w_1 \times |J_1| + w_2 \times |J_2| + w_3 \times |J_3| + w_4 \times |J_4| + w_5 \times |J_5|, \]

where \( |J_1|, |J_2|, |J_3|, |J_4|, \) and \( |J_5| \) are the magnitudes of \( J_1, J_2, J_3, J_4, \) and \( J_5, \) respectively. \( w_1, w_2, w_3, w_4, \) and \( w_5 \) are the corresponding weights assigned to \( J_1, J_2, J_3, J_4, \) and \( J_5. \)

6. Simulation Results

In Figure 4, the block diagram of whole system is given. The angle of the rod is controlled via the \( \text{PI}^\lambda \text{D}^\mu \) controller where the aim is to hold the rod at the upright position. Some parameters for PSO program are given in Table 2.

The values of \( J_1, J_2, J_3, J_4, \) and \( J_5 \) are given in Table 4, and the overall value of objective function \( J \) is less in case of \( \text{PI}^\lambda \text{D}^\mu \) controller as compared to PID controller. In (23), the values of all weights are considered as \( w_1 = w_2 = w_3 = w_4 = w_5 = 1. \)

Here design specifications have been considered as \( \omega_{gc} = 20 \text{ rad/sec}, \phi_m = 60^\circ, \omega_1 = 0.01 \text{ rad/sec}, \omega_2 = 100 \text{ rad/sec}, \) and \( A = B = -20 \text{ dB}. \) Using the objective function of (23) and after running the PSO program with parameters given in Table 2, the \( \text{PI}^\lambda \text{D}^\mu \) parameters are given in Table 3, and by taking \( \lambda = \mu = 1, \) PID controller has been designed using the same conditions as used for \( \text{PI}^\lambda \text{D}^\mu \) controller. The PID parameters are given in Table 3. These are the values for particular case and may vary for running the PSO many times, but the results are almost same for all cases.

The initial angle has been taken as 0.1 rad. The simulation results are shown in Figures 5, 6, and 7 for cart angle, control voltage, and error, respectively, using both \( \text{PI}^\lambda \text{D}^\mu \) and PID controllers. The Bode diagrams of open loop system are
Table 4: Objective function values.

<table>
<thead>
<tr>
<th>Controller</th>
<th>$J_1$</th>
<th>$J_2$</th>
<th>$J_3$</th>
<th>$J_4$</th>
<th>$J_5$</th>
<th>$J$</th>
</tr>
</thead>
<tbody>
<tr>
<td>PI$^\lambda$D$^\mu$</td>
<td>0.0099</td>
<td>$-0.0058$</td>
<td>0.0014</td>
<td>0.0302</td>
<td>$-0.2778 \times 10^{-5}$</td>
<td>0.0473</td>
</tr>
<tr>
<td>PID</td>
<td>$-3.6750 \times 10^{-7}$</td>
<td>5.1383 $\times 10^{-6}$</td>
<td>0.0225</td>
<td>0.0814</td>
<td>$-0.0995$</td>
<td>0.2035</td>
</tr>
</tbody>
</table>

7. Conclusion

In this work, an attempt has been made to test the effectiveness of PI$^\lambda$D$^\mu$ controller for a highly unstable system like inverted cart pendulum. Two different controllers (PI$^\lambda$D$^\mu$ and PID controllers) using frequency domain approach based on the same specifications have been designed. It is shown in Figures 5–7 that PI$^\lambda$D$^\mu$ controller performs better than conventional PID controller. In Bode diagrams as shown in Figures 8 and 9, it is clear that gain margin is better for PI$^\lambda$D$^\mu$ controller than that for conventional PID controller.

It is also seen that the use of particle swarm optimization technique for calculating controller parameters is very simple and provides good convergence towards optimal values.

In this work, only cart angle has been stabilized at its upright position, but on the other hand, cart position stabilization along with cart angle using PI$^\lambda$D$^\mu$ controller and real-time implementation might be the subject of future work.

References


