



Article

Fractional Transformation-Based Decentralized Robust Control of a Coupled-Tank System for Industrial Applications

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Abstract: Petrochemical and dairy industries, waste management, and paper manufacturing fall under the category of process industries where flow and liquid control are essential. Even when liquids are mixed or chemically treated in interconnected tanks, the fluid and flow should constantly be observed and controlled, especially when dealing with nonlinearity and imperfect plant models. In this study, we propose a nonlinear dynamic multiple-input multiple-output (MIMO) plant model. This model is then transformed through linearization, a technique frequently utilized in the analysis and modeling of fractional processes, and decoupling for decentralized fixed-structure H-infinity robust control design. Simulation tests based on MATLAB and SIMULINK are subsequently executed. Numerous assessments are conducted to evaluate tracking performance, external disturbance rejection, and plant parameter fluctuations to gauge the effectiveness of the proposed model. The objective of this work is to provide a framework that anticipates potential outcomes, paving the way for implementing a reliable controller synthesis for MIMO-connected tanks in real-world scenarios.

Keywords: Bernoulli principle; flow rates; fluid study; H-infinity control design; linear time invariant system; liquid levels; MATLAB mixsyn; optimization problem; system linearization



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1. Introduction

Control of linked tank systems is recognized as a crucial challenge in process industries such as pharmaceuticals, food processing, autonomous liquid dispensing, and other chemical-related sectors [1,2]. The traditional proportional, integral, and derivative (PID) controller, noted for its effectiveness in linear systems and its intuitive user-friendly control, is widely used in these industries [3–8]. However, when we have multiple-input multiple-output (MIMO) coupled-tank systems, which are nonlinear in essence, standard control methods typically fail to deliver consistent performance because of nonlinearities, parameter variations, and external disturbances. This is where the importance of a linearization process arises, a process often employed in fractional systems [9]. To address these failures, much approaches have explored nonlinear control strategies for multi-tank systems. Such approaches include controls related to nonlinear sliding modes [8], nonlinear backstepping [10], constrained prediction [11], convolution network [12], and fuzzy methods [13–15], which are some of the techniques employed to tackle the challenges posed by these nonlinear systems.

We propose a controller that represents a novel approach and unexplored in the context of two-tank systems. The controller is based on H-infinity (H_∞) theory, which enables decentralized control of MIMO systems as well as robust control of single-input single-output (SISO) systems [16–20]. A recent study has also demonstrated the potential of fractional transformation-based intelligent H_∞ systems in controlling a direct current servo motor [21]. Nonetheless, fixed-structure H_∞ controllers offer a significant advantage over traditional H_∞ synthesis methods, as our controller can be designed with specific orders and fixed structures. This allows process industries to utilize traditional PID controllers, which are linear such as the real-time controllers. Nevertheless, traditional techniques are often limited by design objectives such as response speed, disturbance rejection, and robust stability. The complex structure and control bandwidth of MIMO systems impose additional challenges for traditional H_∞ control architectures [19].

Our advanced linear robust controller tackles these challenges by providing a solution with a simpler gain structure, PID-like characteristics, sophisticated lead-lag features, and an observer-based architecture. Furthermore, a non-smooth H_∞ optimizer in the frequency domain is employed to tune the parameters of the proposed controller [22]. With their inherent robustness and linearity, these controllers are ready for direct deployment.

The objective of the present study is to regulate the liquid level in a MIMO coupled-tank system by manipulating the operation of the water pump within the tanks. To reach this objective, the MIMO nonlinear model is first linearized, and then a decoupled linear model is derived, enabling the decentralized and robust construction of our fixed-structure H_∞ controller. Subsequently, a reliable fixed-order robust synthesis tailored for the decoupled MIMO model is introduced, ensuring that the time domain performance remains uncompromised [23]. The robust control toolbox in MATLAB proves to be invaluable in the optimization of the proposed decentralized synthesis architecture. Capable of managing plant nonlinearity, external interferences, and measurement noise simultaneously, our advanced robust controller guarantees consistent liquid levels in the MIMO coupled-tank system, synthesizing apt control signals in response to target liquid levels.

The structure of the article is as follows. Section 2 delves into MIMO modeling of the linked tank system. In Section 3, we describe the decoupling model and proposed controller. Section 4 discusses our results. The conclusions of this work are in Section 5.

2. Modeling

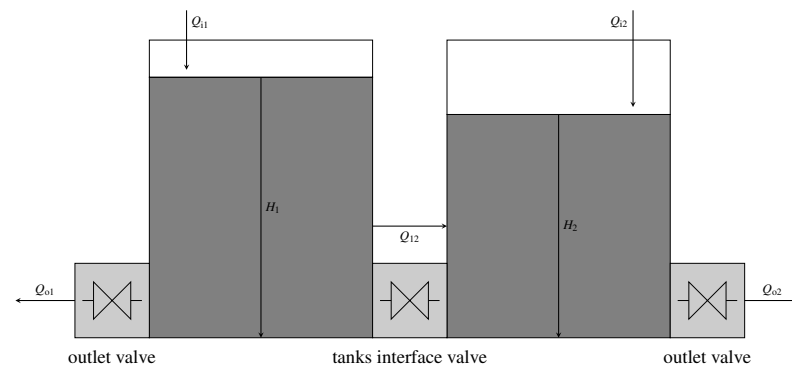
This section describes an in-depth discussion of nonlinear MIMO system.

2.1. System's Description

An industrial coupled-tanks system comprises two tanks interconnected by orifice valves, with electrical valves controlling the input flow into both tanks, and orifice valves regulating the outflow from the tanks. In coupled tanks, there are typically restrictions on the maximum opening of control valves. Such systems exhibit inherent complexity and nonlinearity, presenting challenges for traditional control approaches. Figure 1 illustrates a two-tank system, and Table 1 provides descriptions of the symbols used. The control system's inputs, Q_{i1} and Q_{i2} , represent the liquid inflow rates into the tanks via the electrical valves. The control variables, Q_{o1} and Q_{o2} , indicate the liquid outflow rates through the orifice valves, while H_1 and H_2 represent the MIMO output variables. The orifice flow between the tanks, denoted as Q_{12} , facilitates the coupling between the tanks.

Table 1. Description of notations.

Notation	Definition
Q_{i1}	Flow index/control input for tank 1
Q_{i2}	Flow index/control input for tank 2
H_1	Height (liquid) in tank 1
H_2	Height (liquid) in tank 2
Q_{o1}	Outflow rate through tank 1's orifice valve
Q_{o2}	Outflow rate through tank 2's orifice valve
Q_{12}	Coupling flow rate between tanks 1 and 2 through orifice valve
A	Cross-section area of each tank

**Figure 1.** Industrial MIMO coupled tank system.

2.2. Nonlinear MIMO Mathematical Model

Next, we establish a mathematical model that captures the dynamic behavior of interconnected systems. The nonlinearity and MIMO structure of these systems typify many real-world scenarios. The equations for the dynamic flow balance in tank 1 are

$$Q_{i1} - Q_{o1} - Q_{12} = A_1 \frac{d}{dt} H_1 \quad (1)$$

and

$$Q_{12} - Q_{o2} = A_2 \frac{d}{dt} H_2, \quad (2)$$

where A_1, A_2 are the cross-section areas of tanks 1 and 2, respectively; and H_1, H_2 refer to the baseline or equilibrium liquid levels in tanks 1 and 2, respectively, that is, they represent the static or steady-state levels of the liquid in the absence of any disturbances or inputs.

Using the Bernoulli principle, the formulas stated in (1) and (2) are now converted into nonlinear dynamic equations. The Bernoulli principle, which ignores compressibility and viscosity, relates the steady flow of a flowing liquid to pressure, velocity, and elevation. The fluid flow described by the Bernoulli principle is shown as

$$\frac{P}{\rho} + \frac{v^2}{2} + gH = c,$$

where g is the gravity acceleration, H is the fluid height from the reference point, v is the fluid movement speed, P is the fluid pressure, ρ is the fluid density, and c is a constant.

Applying the Bernoulli principle to a fluid in motion, we get

$$\frac{P_1}{\rho} + \frac{v_1^2}{2} + gH_1 = \frac{P_2}{\rho} + \frac{v_2^2}{2} + gH_2. \quad (3)$$

Suppose that $v_1 = 0$, $P_1 = 0$, and $P_2 = 0$. When we solve the equation given in (3), we obtain

$$v_2 = \sqrt{2g(H_1 - H_2)}.$$

Then, the equation for coupling flow discharge is formulated as $Q_{12} = C_d a_{o12} v_2$, with C_d being the discharge coefficient and a_{o12} being the coupled orifice cross-section area between tanks 1 and 2, representing the passage that allows flow from one tank to the other. Similarly, we have that

$$Q_{o1} = C_{d2} a_{o1} \sqrt{2gH_1}, \quad Q_{o2} = C_{d3} a_{o2} \sqrt{2gH_2},$$

where a_{o1} and a_{o2} are the orifice cross-section areas for the outflow from tanks 1 and 2, respectively, which controls the liquid's discharge from the corresponding tank.

Note that the expressions given in (1) and (2) are changed to

$$Q_{i1} - C_{d2} a_{o1} \sqrt{2gH_1} - C_d a_{o12} \sqrt{2g(H_1 - H_2)} = A \frac{d}{dt} H_1$$

and

$$Q_{i2} + C_d a_{o12} \sqrt{2g(H_1 - H_2)} - C_{d3} a_{o2} \sqrt{2gH_2} = A \frac{d}{dt} H_2,$$

respectively, when the Bernoulli principle has been applied, where A is the cross-section area of tanks 1 and 2. By assuming $\alpha_1 = C_d a_{o12} \sqrt{2g}$, $\alpha_2 = C_{d2} a_{o1} \sqrt{2g}$, and $\alpha_3 = C_{d3} a_{o2} \sqrt{2g}$, our nonlinear equations are formulated as

$$Q_{i1} - \alpha_2 \sqrt{H_1} - \alpha_1 \sqrt{H_1 - H_2} = A \frac{d}{dt} H_1 \tag{4}$$

and

$$Q_{i2} + \alpha_1 \sqrt{H_1 - H_2} - \alpha_3 \sqrt{H_2} = A \frac{d}{dt} H_2. \tag{5}$$

These nonlinear equations represent the system's underlying complexity, paving the way for exploring advanced control strategies in interconnected systems.

2.3. The Perturbed Linearized MIMO Mathematical Model

Next, we aim to develop a linearized model based on perturbations from steady-state conditions, a common approach to simplifying analysis and control design in complex systems.

The liquids in tanks 1–2 and the flow rates, Q_{i1} and Q_{i2} namely, are assumed to be at steady-state levels. Then, the nonlinear equations adopt the forms given by

$$Q_{i1} + q_{i1} - \alpha_2 \sqrt{H_1 + h_1} - \alpha_1 \sqrt{(H_1 + h_1) - (H_2 + h_2)} = A \frac{d}{dt} (H_1 + h_1) \tag{6}$$

and

$$Q_{i2} + q_{i2} + \alpha_1 \sqrt{(H_1 + h_1) - (H_2 + h_2)} - \alpha_3 \sqrt{(H_2 + h_2)} = A \frac{d}{dt} (H_2 + h_2), \tag{7}$$

if any one of the parameters is slightly altered; where q_{i1} and q_{i2} are the perturbations in Q_{i1} and Q_{i2} , respectively; whereas h_1 and h_2 are the perturbations in the liquid levels of tanks 1 and 2, respectively, from their baseline states H_1 and H_2 . In other words, h_1 and h_2 depict how much the liquid levels deviate from their equilibrium states due to disturbances or inputs to the system.

Now, after subtracting and simplifying the nonlinear equations stated in (4) and (5) from the perturbed equations presented in (6) and (7), respectively, we obtain

$$q_{i1} - \frac{\alpha_2}{2\sqrt{H_1}} h_1 - \frac{\alpha_1 h_1}{2\sqrt{H_1 - H_2}} + \frac{\alpha_1 h_2}{2\sqrt{H_1 - H_2}} = A \frac{d}{dt} H_1$$

and

$$q_{i2} + \frac{\alpha_1 h_1}{2\sqrt{H_1 - H_2}} - \frac{\alpha_1 h_2}{2\sqrt{H_1 - H_2}} - \frac{\alpha_3 h_2}{2\sqrt{H_2}} = A \frac{d}{dt} H_2.$$

Let

$$k_1 = \frac{\alpha_1}{2\sqrt{H_1 - H_2}}, \quad k_2 = \frac{\alpha_1}{2\sqrt{H_1 - H_2}} + \frac{\alpha_2}{2\sqrt{H_1}}, \quad k_3 = \frac{\alpha_1}{2\sqrt{H_1 - H_2}} + \frac{\alpha_3}{2\sqrt{H_2}}.$$

Before delving further, it is essential to understand the term LTI, which stands for linear time-invariant, where ‘linear’ means the system obeys superposition principles, and ‘time-invariant’ denotes that the system’s behavior remains consistent over time.

Table 2 provides the system parameters and LTI values for the model presented as

$$\begin{bmatrix} \dot{h}_1 \\ \dot{h}_2 \end{bmatrix} = \begin{bmatrix} -k_2 & k_1 \\ k_1 & -k_3 \end{bmatrix} \begin{bmatrix} h_1 \\ h_2 \end{bmatrix} + \begin{bmatrix} 1/A & 0 \\ 0 & 1/A \end{bmatrix} \begin{bmatrix} q_{i1} \\ q_{i2} \end{bmatrix}, \tag{8}$$

where \dot{h}_1 and \dot{h}_2 denote the time derivatives of h_1 and h_2 , respectively. In practical terms, \dot{h}_1 describes how quickly the variation h_1 is changing over time, indicating the rate at which the liquid level in tank 1 is rising or falling. Similarly, \dot{h}_2 represents the rate of change of the variation h_2 , showing how quickly the liquid level in tank 2 is adjusting due to disturbances or inputs. The calculated values from (8) are $k_1 = 4.082249$, $k_2 = 5.344471$, $k_3 = 6.062148$.

Table 2. Extended couple tank system’s parameters and values.

Name	Expression	Value
Cross-section of tanks 1 and 2	A	4 m ²
Discharge coefficients	C_{d1}, C_{d2}, C_{d3}	1, 0.5, 0.5
Cross sections of an orifice valve	a_{o12}, a_{o1}, a_{o2}	1 cm ²
Gravitational acceleration	g	9.8 m/s ²
Sensor	Ideal	1
Liquid levels offset	H_1, H_2	0.5 m, 0.2 m
Dynamics constant 1	k_1	4.082249
Dynamics constant 2	k_2	5.344471
Dynamics constant 3	k_3	6.062148

Additionally, the system’s MIMO coupled transfer function model is provided below. From input 1 to both outputs, transfer functions are given by

$$g_{11} = \frac{0.25\sigma + 1.516}{\sigma^2 + 11.4\sigma + 15.67}, \quad g_{12} = \frac{1.021}{\sigma^2 + 11.4\sigma + 15.67},$$

where σ represents the real part of a complex variable in the Laplace domain, commonly used to analyze the dynamics in the frequency domain of control systems and transfer functions. From input 2 to both outputs, transfer functions are formulated as

$$g_{21} = \frac{1.021}{\sigma^2 + 11.4\sigma + 15.67}, \quad g_{22} = \frac{0.25\sigma + 1.334}{\sigma^2 + 11.4\sigma + 15.67}.$$

In the context of MIMO systems, the constants g_{11} , g_{12} , g_{21} , and g_{22} represent the transfer functions that define the dynamic interactions between the multiple inputs and outputs. The simulation results for the response of the MIMO LTI model are displayed in Figure 2.

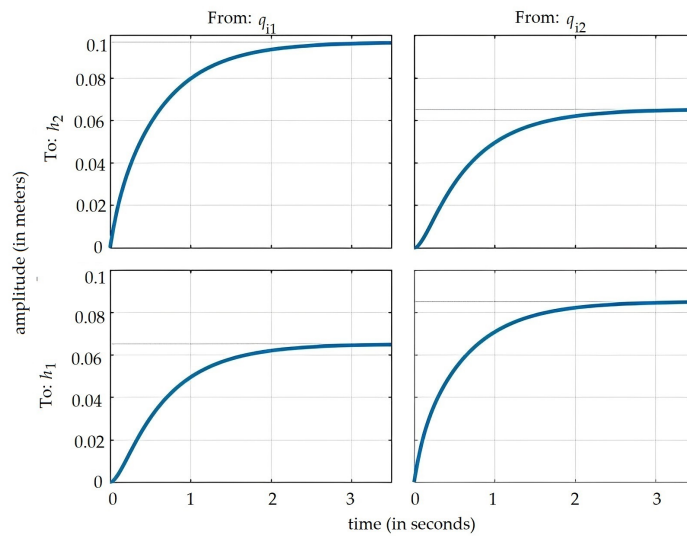


Figure 2. Coupled MIMO LTI model step response.

To implement a decentralized control strategy, we opt to split our MIMO model into two distinct SISO systems. Numerous decoupling techniques have been previously documented in scientific publications. In this particular study, we employ a linear decoupling method. We want to introduce an approach for diminishing the interdependency among the control loops by introducing additional decouplers, commonly referred to as controllers, into the multi-loop control system. With this approach, the decoupling matrix D_m is designed to minimize interactions across all loops [24]. Consequently, we present the constituent elements for a 2-input-2-output system by means of

$$D_m = \begin{bmatrix} d_{11} & d_{12} \\ d_{21} & d_{22} \end{bmatrix}, \quad (9)$$

where $d_{11} = 1$, $d_{12} = -g_{12}/g_{11}$, and $d_{21} = -g_{21}/g_{22}$, $d_{22} = 1$ are constant elements of the decoupling matrix designed to reduce or eliminate the mentioned interactions, ensuring each control loop operates more independently of the others. The constants d_{11} , d_{12} , d_{21} , d_{22} , g_{11} , g_{12} , g_{21} , and g_{22} are essential in tuning the system to achieve the desired performance while minimizing unwanted couplings.

After decoupling, a MIMO system may still suffer interactions and perform poorly. The decoupling seeks to counteract the effects of interactions caused by the cross-coupling of system variables [25]. The decoupler's layout, which was produced using (9), is shown in Figure 3.

The specific values for the decoupler parameters are $d_{11} = 1$, $d_{12} = -1.021/(0.25\sigma + 1.516)$, $d_{21} = -1.021/(0.25\sigma + 1.334)$, and $d_{22} = 1$. The values make clear that the decoupler described is not in static form, which makes it difficult to implement this decoupling matrix [26]. One strategy for solving this problem is to use static decoupling. The only advantages offered by static decoupling are the constants d_{12} and d_{21} , whose values are calculated as

$$d_{12} = \lim_{\sigma \rightarrow 0} -\frac{1.021}{0.25\sigma + 1.516} = -\frac{k_{12}}{k_{11}} \quad (10)$$

and

$$d_{21} = \lim_{\sigma \rightarrow 0} -\frac{1.021}{0.25\sigma + 1.334} = -\frac{k_{21}}{k_{22}}, \quad (11)$$

where now $k_{12} = 1.021$, $k_{11} = 1.516$, $k_{21} = 1.021$, and $k_{22} = 1.334$.

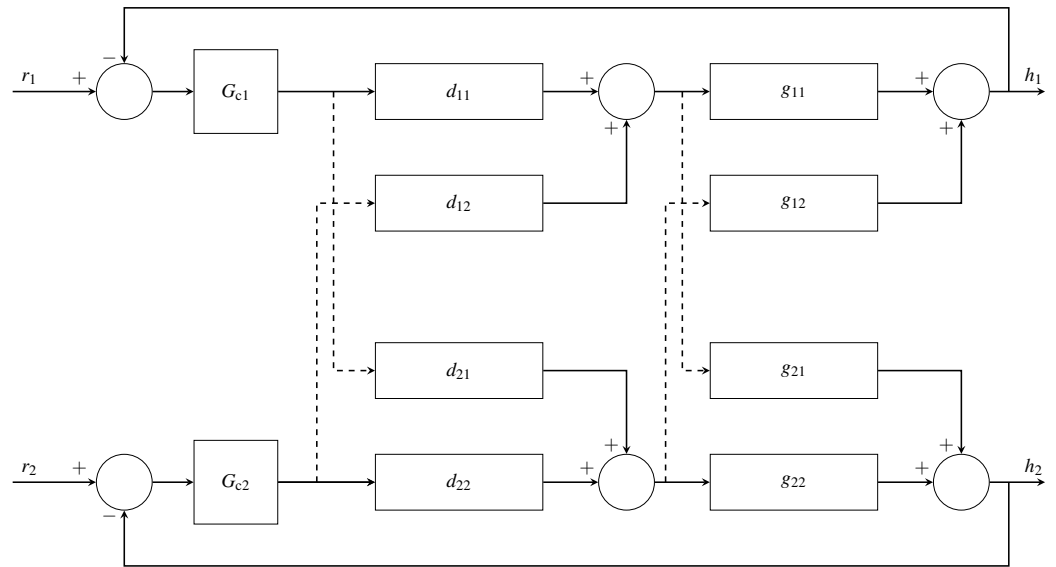


Figure 3. Decoupled layout for MIMO process, where r_1, r_2 are the desired responses of liquid height in tanks 1 and 2, whereas G_{c1} and G_{c2} are decentralized controllers for h_1 and h_2 , respectively.

Note that the formulas given in (10) and (11) can be used to determine the two separate SISO systems. Then, the expressions stated in

$$g_1 = g_{11} - \frac{k_{21}}{k_{22}} g_{12} = \frac{0.25\sigma + 0.734559}{\sigma^2 + 11.4\sigma + 15.67}$$

$$g_2 = g_{22} - \frac{k_{12}}{k_{11}} g_{21} = \frac{0.25\sigma + 0.646374}{\sigma^2 + 11.4\sigma + 15.67}$$

represent the estimated independent transfer functions. The simulation results for the response of the decoupled MIMO LTI model are shown in Figure 4. These results validate the derived linearized models, laying a foundation for exploring more nuanced control design strategies. They hint at the intriguing complexity of the system’s dynamics, subtly underscoring the potential benefits of fractal and fractional approaches in such contexts.

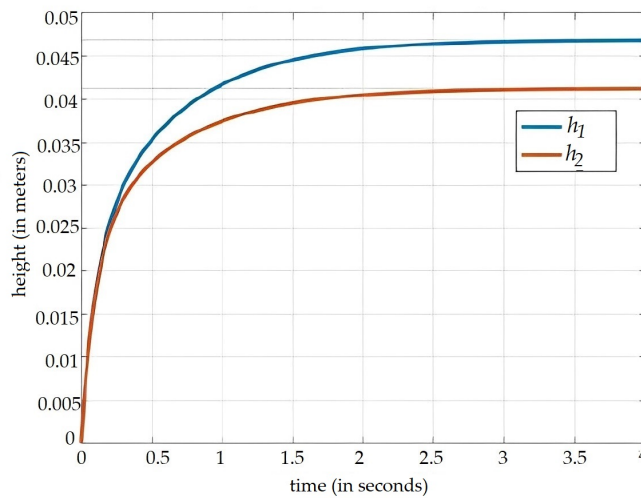


Figure 4. Decoupled MIMO LTI model step response.

3. Control Design for the Decoupled MIMO System

In this section, we shift our attention towards the construction of control mechanisms for our modeled decoupled MIMO system, which subtly leads to a connection with the multifaceted nature of fractional systems.

3.1. Context

With our decoupled MIMO LTI perturbed model, our primary challenge is to develop a controller capable of producing a suitable transient response and of following the reference input accurately. In this study, we construct and compare three different types of controllers: our H_∞ control system, the traditional H_∞ , and the PID.

The inflow rates in both tanks serve as control variables, while the liquid levels in each tank represent the measured variables. Performance comparisons among the controllers are made based on their ability to track a reference signal, along with evaluating transient characteristics such as settling time, rising time, steady-state accuracy, and overshoot.

3.2. Traditional PID Design

The traditional PID controllers are widely utilized in the process industries because they are straightforward and have a low cost. For those dynamic models with inherently nonlinear behavior, the PID controllers can occasionally lose some of their effectiveness. Additionally, they do not provide the requisite transient responsiveness and reference tracking parameters. Traditional methods are used to determine the PID parameters. The traditional PID controller is designed utilizing the Zeigler Nicholas pole-zero placement method. The decoupled LTI models of the connected tank system are employed to develop the traditional controllers.

The equations for the PID controller in the frequency and time domains are stated as

$$U(s) = \left(K_p + K_i \frac{1}{s} + K_d s \right) E(s)$$

and

$$u(t) = K_p e(t) + K_i \int_0^t e(\tau) d\tau + K_d \frac{d}{dt} e(t),$$

where $E(t)$ is the Laplace transform of the error in the time domain. This error is the difference between the reference value (or set-point) and the system output. In the frequency domain (s -domain), this error is denoted by $E(s)$. Also, $e(t)$ is the error in the time domain, that is, the instantaneous value of the difference between the desired value and the output at a given time t . The PID controller aims to minimize this error by adjusting the system output. Three parameters or gains of the PID controller are the proportional component (K_p) that acts on the present error, the integral component (K_i) that acts on the accumulated past errors, and the derivative component (K_d) that anticipates future errors based on their rate of change.

3.3. Traditional H_∞ Controller Design

In this design, both close-loop frequency transfer functions, that is, the complementary sensitivity function, $T(s)$ namely, and the sensitivity function, $S(s)$ say, are shaped using complicated weights or filters. Note that $S(s)$ is the function that transfers frequency between the error signal $E(s)$ and the reference input $R(s)$. In contrast, the complementary sensitivity function $T(s)$ represents the transfer between the reference input $R(s)$ and the output $Y(s)$.

The filters or complex weights have an impact on the shape of closed-loop frequency responses in a traditional H_∞ controller's architecture. Thus, the norm of H_∞ , that is, the peak of the objective function, is decreased by choosing appropriate filters or complex weights for $S(s)$ and $T(s)$.

The structural complexity and control bandwidth of this traditional H_∞ controller is primarily constrained through this H_∞ norm. For the purpose of optimizing the controller parameters, the constrained objective function equations are provided as

$$\|w_S(s)S(s)\|_\infty \leq 1 \quad (12)$$

and

$$\|w_T(s)T(s)\|_\infty \leq 1,$$

where $w_S(s)$ and $w_T(s)$ are weighting filters used to shape the frequency responses of the sensitivity $S(s)$ and complementary sensitivity $T(s)$, respectively, in the proposed H_∞ controller design.

The sensitivity $S(s)$ should ideally be as low as feasible over the specified bandwidth range. The bounds on peak specifications, which provide a margin of robustness, are mentioned in (12). In addition to being reliable, the optimum H_∞ controller outperforms the traditional PID controller. The `mixsyn` MATLAB tool was employed to implement our H_∞ controller.

3.4. Design of Fixed-Structure H_∞ Synthesis

The traditional H_∞ controller's order is derived from the combination of the shaping weights and the order of the plant. However, this design complexity in the traditional H_∞ controller can limit its practical applications. To address these limitations, this study proposes a decentralized fixed-structure H_∞ synthesis.

The robust toolbox of MATLAB is utilized to build the fixed-structure H_∞ synthesis. First, the system is represented in the standard form as shown in Figure 5.

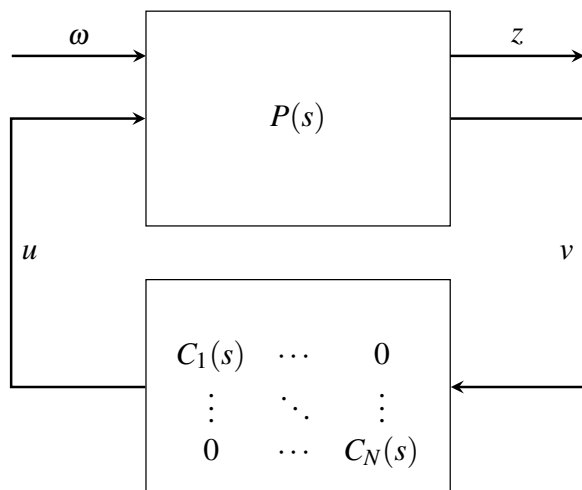


Figure 5. Structured H_∞ synthesis in standard representation, where C , P , u , w , and z are earlier defined, while v is the error signal for the SISO system, v is the input to the controller. We use v as a general representation in this figure and this is defined in as $v(t) = r(t) - y(t) = e(t)$, with $r(t)$ being the reference value, $y(t)$ being the true output, and $e(t)$ being the error.

Figure 5 illustrates the standard representation of the structured H_∞ synthesis, which can be divided into two main sections:

- Block $P(s)$: represents the non-tunable LTI components of the MIMO system.
- Block $C(s)$: contains the tunable components diagonally as $\text{diag}(C_1(s), \dots, C_N(s))$. Each of these components has a specific structure and is an LTI control element.

The suitable weights w are taken into account to obtain the targeted sensitivity and complementary sensitivity. Then, the control parameters are improved in the frequency domain using non-smooth H_∞ optimization [19]. These fixed-structure linear optimal and robust controllers are more practical from an industrial standpoint.

The standard form presented in Figure 5 can be used to configure SISO or MIMO systems, according to the linear robust control [27]. In any control system block diagrams, the block $C(s)$ is where all the control elements are arranged and the remaining parts are arranged in the block $P(s)$. In the standard representation, the weight w is composed of disturbances, commands, and external noise. Consider the partition formulated as

$$\begin{bmatrix} z \\ y \end{bmatrix} = P \begin{bmatrix} w \\ u \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} \\ p_{21} & p_{22} \end{bmatrix} \begin{bmatrix} w \\ u \end{bmatrix},$$

where z is composed of error signals, while the closed loop optimal objective function from w to z is given by a linear rational transformation stated as

$$T_{wz} = F_{LTI}(P, C) = p_{11} + p_{12}C(I_2 - p_{22}C)^{-1}p_{21},$$

where T_{wz} denotes the closed-loop transfer function from w to z ; $F_{LTI}(P, C)$ describes the relationship between the process and controller; and I_2 is the identity matrix of order 2.

Note that the fixed-order linear structure controller is formulated as

$$C_j(s) = K_p + \frac{K_i}{s} + \frac{K_d s}{T_f s + 1}, \quad j \in \{1, \dots, M\},$$

where M is the total number of distinct specifications or scenarios being considered in the controller design. The suggested controller parameters are optimized via non-smooth H_∞ optimization, where, as mentioned, K_p is the proportional gain, K_i is the integral gain, T_f is the first order derivative filter, and K_d is the derivative gain coefficient.

Critical design specifications, including disturbance rejection, control bandwidth, and high stability margins, are encapsulated as

$$\|w_j(s)S_j(s)\|_\infty \leq 1, \quad \|w_j(s)T_j(s)\|_\infty \leq 1, \quad j \in \{1, \dots, M\}.$$

The H_∞ norm stated in $H(s) = F_{LTI}(P(s), \text{diag}(C_1(s), \dots, C_N(s)))$ is minimized to satisfy the desired conditions.

The optimization process initiates with randomly generated parameter values and iteratively refines until optimal values are achieved. The results and comparisons are discussed in Section 4, which also presents the optimized controller parameters. The optimized parameter values of the proposed controller are provided in Table 3.

Table 3. Optimized parameters of decentralized fixed-structure H_∞ controller.

Symbols	Description	Values for h_1 Tracking	Values for h_2 Tracking
K_p	Proportional gain	6730	4210
K_i	Integral gain	3230	3660
K_d	Derivative gain	−9750	−8580
T_f	1st order filter coefficient	1.64	2.44

4. Performance Analysis and Discussion

This section evaluates the proposed decentralized control approach for the MIMO-coupled tank system against traditional H_∞ and PID controllers, focusing on transient response and command tracking. Precise tracking of desired signals remains a prominent challenge in industries.

4.1. Context

Despite the affordability and simplicity of traditional PID controllers making them a staple in process industries, they sometimes falter, especially when faced with nonlinear system dynamics. In addition, traditional H_∞ controllers ensure robust performance, but they are marred by their intricate structure, making them less practical for some applications. The proposed fixed-order H_∞ controller strikes a balance, offering dependable performance with a simpler implementation. A distinctive benefit of this synthesis is the flexibility it offers in design. The controller effectively shapes the sensitivity $S(s)$ and complementary sensitivity $T(s)$, uninfluenced by complex higher-order weights, contrasting the more intricate standard H_∞ controllers.

4.2. Results and Discussion

The optimal parameters for the decentralized control of MIMO-coupled tank systems were depicted in Table 3, and the corresponding expressions for these controllers are provided as

$$\begin{aligned}
 A_{h_1} &= \begin{bmatrix} -0.01 & 0 & 0 \\ 2.28600 & -1.8680 & -1.3870 \\ 0 & 4 & 0 \end{bmatrix}, & B_{h_1} &= \begin{bmatrix} 4.091 \\ 0 \\ 0 \end{bmatrix}, \\
 C_{h_1} &= [894100 \quad -7.2990 \quad -54210], & D_{h_1} &= [0], \\
 A_{h_2} &= \begin{bmatrix} -0.01 & 0 & 0 \\ 2.52200 & -2.0590 & -1.3440 \\ 0 & 4 & 0 \end{bmatrix}, & B_{h_2} &= \begin{bmatrix} 4.087 \\ 0 \\ 0 \end{bmatrix}, \\
 C_{h_2} &= [987400 \quad -8.0580 \quad -52610], & D_{h_2} &= [0].
 \end{aligned}$$

Before proceeding, it is important to clarify a particular aspect of Table 3, namely the negative values of the derivative gain, K_d . These values are not errors but a direct result of the optimization process in the fixed-structure H_∞ controller design.

After determining the optimized parameters, the resulting controllers are represented in Table 3. These controllers provide the state-space structure of traditional H_∞ controllers for decentralized tracking of h_1 and h_2 , respectively. This specific structure is pivotal for the effectiveness of decentralized control and directly contributes to the optimized performance we observe.

While uncommon and negative derivative gains can be part of a balanced solution minimizing an error criterion, such values might pose practical implementation challenges due to the potential phase addition and destabilizing effects. These negative K_d values could be problematic in a real-world application. Furthermore, fine-tuning or constraints in the optimization process may be introduced to ensure non-negative derivative gains. Nevertheless, this might deviate the system's performance from the optimal results obtained for the controllers presented here. These controllers provide the state-space structure of traditional H_∞ controllers for decentralized tracking of h_1 and h_2 , respectively.

The comparison in terms of transient definition and commanded tracking among three controllers is shown in Figures 6 and 7. Table 4 presents a comparison based on transient parameters such as steady-state accuracy, percentage overshoot, rising time, and settling time. It can be seen from Table 4 that both H_∞ controllers have no overshoot in the transient response, while the traditional PID controller exhibits a significant overshoot in the response.

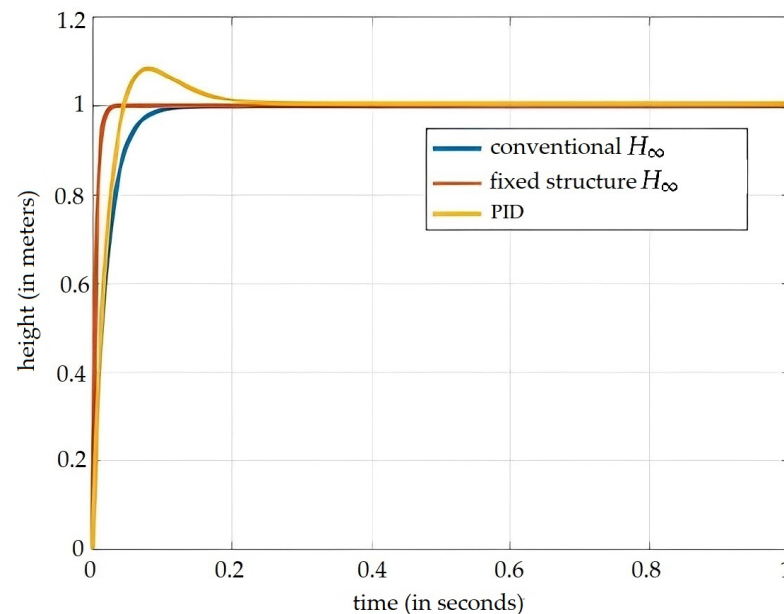


Figure 6. Step response for the decentralized control of liquid level h_1 .

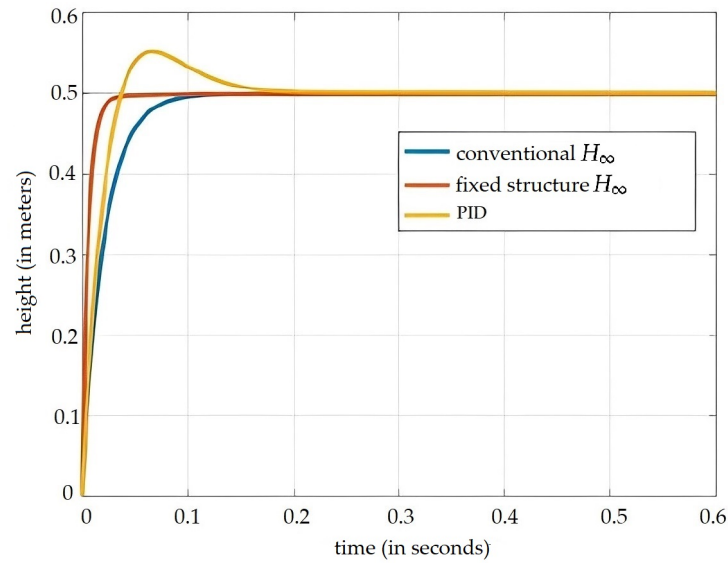


Figure 7. Step response for the decentralized control of liquid level h_2 .

Table 4. Controllers performance evaluation for h_2 .

Controller	Rise Time (s)	Settling Time (s)	Over-Shoot (%)	Steady-State Error	Controller Order
Traditional H_∞	0.0448	0.0801	0	0%	Third
Fixed-structure H_∞	0.0131	0.0249	0	0%	Second
PID	0.0268	0.1451	10.3289	0.1033	Second

Compared to the traditional PID controller, the traditional H_∞ controller delivers robust performance but with a somewhat slower rise time. In contrast, the proposed fixed-structure H_∞ controller shows robust performance across all tested parameters. The response comparison for step disturbance for all three controllers is shown in Figure 8. From this figure, note that the proposed fixed-structure H_∞ synthesis provides fast and robust step disturbance rejection as compared to other traditional controllers. The proposed fixed-structure H_∞ controller also gives a robust performance in the presence of parameter uncertainty. The performance of the proposed controller is evaluated for a wide range of parameter values and the results are shown in Figure 9.

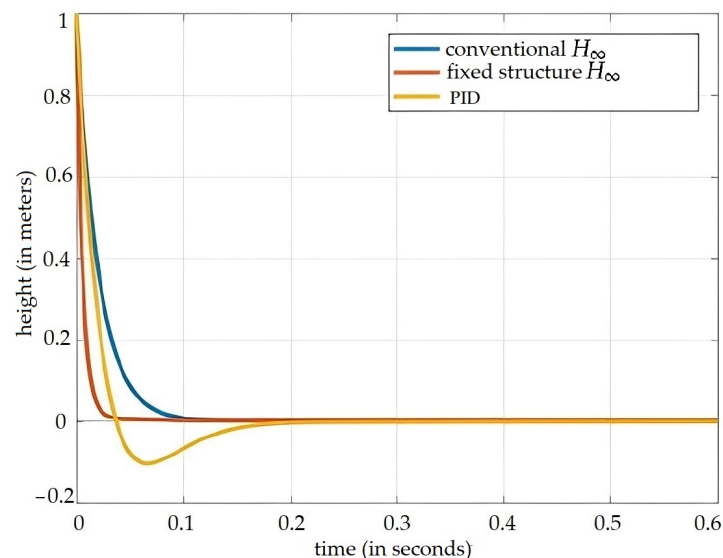


Figure 8. Comparison of controllers for step disturbance response in liquid level h_2 .

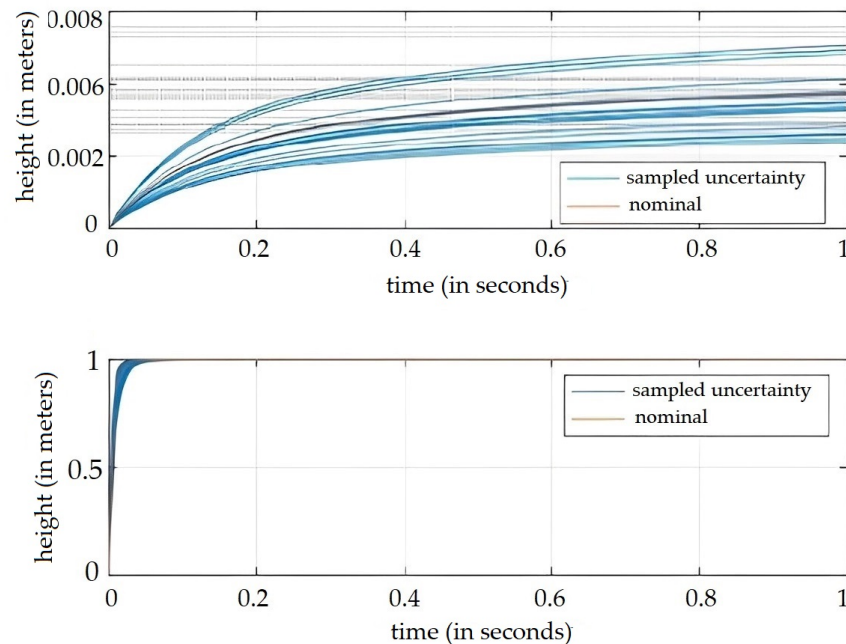


Figure 9. Performance of the proposed controller for **(top)** open loop step response with parameters uncertainty in liquid level h_2 and **(bottom)** step input tracking.

5. Conclusions

In this research, we proposed a decentralized intelligent robust controller for a MIMO coupled-tank system. This system has been a longstanding challenge in process industries, and numerous control strategies have been previously employed to manage it, such as nonlinear sliding mode control, nonlinear backstepping control, predictive constrained control, convolution network control, and fuzzy control. However, as demonstrated in Table 4, our proposed controller, based on H_∞ theory, has shown to outperform traditional H_∞ and PID controllers in terms of transient requirements.

The fixed order and fixed structure of the proposed controller are additional advantages. These characteristics highlight the pertinence of our study in the context of fractal and fractional systems, where the complexity often resides in the intricate structure and variable order of the controllers. In contrast to traditional H_∞ synthesis, where the order increases when dealing with complex filter transfer functions, our proposed controller maintains a fixed-order, an advantage previously demonstrated in the control of a direct current servo motor. The proposed controller is decentralized robust linear and utilizes intelligent non-smooth H_∞ optimization for its parameters. Its straightforward gains or optimized PID-based structure makes it practical and easy to use. Despite the complex structure and control bandwidth challenges associated with MIMO systems, our proposed linear robust controller has shown, through comparative simulation results, to be robust against parameter changes.

Future work should consider addressing the issue of actuator saturation, which is critical in process industries. A potential solution could be the incorporation of a strategy related to anti-integral windup in the design of a decentralized fixed-structure H_∞ controller. To ensure stability in the presence of actuator saturation, additional simulation results can be obtained prior to real-time implementation, ensuring no risk of systemic instability or damage. Furthermore, while our study provided promising simulation results, future research should also focus on real-world verification to further substantiate the efficacy of the proposed controller.

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