

(Article)

# Sliding-Mode Control of Horizontal Coupled-Tank Systems Using Variable-Speed Pump

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# ABSTRACT

In many chemical and food processing industries, liquid is pumped into and held in interconnected linked tanks. Nevertheless, due to the high non-linearity and complexity of that system, level and flow regulation among these tanks is a non-trivial task. The current research focuses on the regulation of the liquid level in two horizontally linked tanks. The three most common sliding-mode control (SMC) algorithms that can be found in the literature—proportional-derivative SMC (PD-SMC), proportional-integral-derivative SMC (PID-SMC), and dynamic SMC—are compared. The impact of sensor noise on the functioning of the controller and the chattering phenomenon is highlighted in particular. Simulations are done for the proposed control algorithms. Using the MATLAB optimization toolbox, control parameters are chosen to improve the designed performance indices. Experiments are carried out on a designed test setup to investigate the effect of sensor noise. Results showed that PD-SMC has a superior performance over the other two algorithms, PID-SMC and dynamic-SMC both in simulation and experimental results.

Keywords: Horizontal coupled-tank, SMC, Dynamic-SMC, variable-speed control pump

## **1. Introduction**

Several industrial sectors, including petrochemicals, drugs, and paper industries, utilize liquid-level control in multi-tank systems. However, because of its complexity and strong nonlinearity, a connected tank system poses a challenging control problem. To maintain a target liquid level irrespective of system uncertainty and external disturbances, an accurate model and an effective control strategy are crucial. Because of the PID controller's simple structure, many tuning techniques of its parameters have been adopted, [1], [2]. However, typical PID controller tuning methods fail to provide convenient behavior to coupled-tank systems. Furthermore, most PID algorithms were developed utilizing lower-order linearized process models [3-6]. The lower-order linearization causes further parametric uncertainty. There have been rare attempts to build higherorder models of PID controllers, [7], [8]. Moreover, a robustness measure should be considered throughout the design process in order to avoid model-uncertainty-caused problems. A comparison was carried out between PID and fuzzy control in [9]. For the nonlinear quadruple tank system, a controller based on Artificial Neural Network (ANN) was developed in [10]. Adaptive and backstepping algorithms were also applied in the coupled tank system [11-13]. Backstepping control based on observers, lag controller, and MPC are proposed in [14], [15], and [16] respectively. In [17], a robust nonlinear approach for the control of liquid levels in a quadruple tank system (QTS) is developed based on the design of an integrator backstepping super-twisting controller. Decentralized algorithms were designed to regulate the coupled tank system level [18] and [19]. In [20] and [21], a fractional-order proportional integral (FOPI) control for coupled tank systems was built. Integral-order PD (IOPD), integer-order PI (IOPI), cascaded FOPD, and FOPI control algorithms for the coupled tank system were compared in [22]. For nonlinear process control, fractional and integral order controllers were assessed and compared based on various models [23]. For managing the level of liquid in spherical connected tanks, a fuzzy FOPID algorithm was proposed in [24]. To address the level control in a multi-input multi-output coupled tank system, a FOID controller was devised [25]. Moreover, TS (Takage-Sugeno) fuzzy controller and fuzzy knowledge-based decoupled control are proposed in [26] and [27] respectively.

SMC offers various appealing characteristics, including good disturbance rejection, improved transient performance, and faster response. SMC laws are fundamentally more robust in

case of uncertainties [28], [29]. The design and analysis of variable structure systems (VSS) with sliding modes were investigated in [30] and [31]. A fuzzy SMC with a nonlinear sliding surface was presented in [32], with a fuzzy logic controller utilized to improve the chattering phenomenon. An adaptive fuzzy SMC was presented in [33]. For better smoothness of the switching signal in the coupled tanks system, a neuro-fuzzy-SMC with a nonlinear sliding surface was designed [34]. Two controllers, backstepping PI-SMC, and PI-SMC were investigated for a quadruple tank in [35]. A SMC of static type for the coupled tanks problem was developed in [36]. Two distinct dynamic SMC algorithms were also developed to eliminate the chattering problem [36]. Feedback linearization in conjunction with the SMC algorithm was used in a quadruple tank system [37]. A SMC of the second order was developed in [38] and [39]. An observer-based control for a fourconnected tank system employing higher-order SMC was presented in [40]. To minimize the disturbance effect on the connected tank system, an adaptive feed-forward second-order SMC was proposed [41]. A chattering-free SMC was developed in [42]. To enhance the tracking behavior of connected tanks under varied uncertainty, an SMC with a variable boundary layer was investigated in [43]. Fuzzy FOSMC was offered as a robust, chatter-free technique for connected tanks [44]. SMC was proposed for the MIMO quadruple tank with time delay compensation in [45].

Sensor signals suffer from noise in industrial settings. This could be due to a variety of factors, for example, lengthy connection of cables, vicinity to different electrical equipment, etc. The controller's ability to control these noises is significant in such instances. Numerous studies claimed that the SMC model is a robust fast controller capable of dealing with uncertain nonlinear systems. Nonetheless, the sensor noise effect on the chattering behavior of SMC hasn't been addressed.

The current research conducted an examination to compare three controllers: PID-SMC, PD-SMC, and dynamic-SMC. Simulink-MATLAB was used to compare the performance of the three algorithms. The current research focuses in particular on the examination of the sensor noise effect on those SMC algorithm's behavior, specifically the chattering problem. The study is structured in six sections: Section 2 presents the concept of the connected tank. Section 3 proposes the design of the control algorithms. Section 4 discusses and displays the simulation results. Section 5 presents the experimental work and finally, Section 6 presents the conclusion.

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# 2. Model of the coupled tank system

The governing equations for the horizontally connected tanks seen in Figure 1 are as follows:

$$C\frac{dh_1}{dt} = q - q_1, C\frac{dh_2}{dt} = q_1 - q_2,$$
(1)

where q is the inlet rate of flow  $(m^3/s)$ ,  $q_1$  is the rate of flow between tanks 1 and 2 in  $(m^3/s)$ ,  $h_1$  is the level of  $1^{st}$  tank in (m),  $h_2$  is the level of  $2^{nd}$  tank (m),  $q_2$  is a rate of flow out of  $2^{nd}$  tank in  $(m^3/s)$  and C is the cross-sectional area of  $1^{st}$  and  $2^{nd}$  tank in  $(m^2)$ . The rates of flow  $q_1$  and  $q_2$  are expressed as follows,

$$q_{1} = c_{12}\sqrt{2g(h_{1} - h_{2})} \quad \text{for } h_{1} > h_{2},$$

$$q_{2} = c_{2}\sqrt{2gh_{2}} \quad \text{for } h_{2} > 0,$$
(2)
(3)

where  $c_{12}$ , and  $c_2$  are the coupling and outlet orifices areas (m<sup>2</sup>), respectively, and g in (m<sup>2</sup>/s) is the gravitational acceleration. The flow q into the 1<sup>st</sup> tank is positive in the connected tanks system since the pump only pumps water into the tank. Consequently, the inflow rate will be

$$q \ge 0 \tag{4}$$



Figure 1. Horizontal connected tanks system

At equilibrium, the derivatives must be zero for a desired constant liquid level, i.e.,

$$\dot{h}_1 = \dot{h}_2 = 0,$$
 (5)

therefore,

$$-\frac{c_{12}}{C}\sqrt{2g|h_1 - h_2|}\operatorname{sgn}(h_1 - h_2) + \frac{Q}{C} = 0, \frac{c_{12}}{C}\sqrt{2g|h_1 - h_2|}\operatorname{sgn}(h_1 - h_2) - \frac{c_2}{C}\sqrt{2gh_2} = 0 \quad (6)$$

where Q denotes the inflow rate at equilibrium. To justify the constraint applied to the inlet rate of flow in equation (4),  $sgn(h_1 - h_2)$  should be positive.

Considering  $z_1 = h_2 > 0$ ,  $z_2 = h_1 - h_2 > 0$ ,  $\mathbf{Z} = \begin{bmatrix} Z_1 \\ Z_2 \end{bmatrix}$ , u = q(t)Also let  $a_1 = \frac{c_2 \sqrt{2g}}{C}$  &  $a_2 = \frac{c_{12} \sqrt{2g}}{C}$ 

As a result, the dynamic model can be stated as

$$\dot{z}_1 = -a_1\sqrt{z_1} + a_2\sqrt{z_2}$$
,  $\dot{z}_2 = a_1\sqrt{z_1} - 2a_2\sqrt{z_2} + \frac{u}{c}$ ,  $y = z_1$ , (7)

where  $z_1 = h_2$  is regarded as the system output.

The objective of the control algorithm is to regulate the output  $y(t) = z_1(t) = h_2(t)$  to the desired value  $h_{2d}$ . The dynamics of the coupled-tank system are nonlinear as can be observed. As a result, a transformation will be constructed in order to convert the model into another formula that will facilitate the design of the controller.

For the state  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  we define a transformation  $\mathbf{x} = T(\mathbf{z})$  as,

$$x_1 = z_1, x_2 = -a_1\sqrt{z_1} + a_2\sqrt{z_2}$$
(8)

The inverse transformation  $\mathbf{z} = T^{-1}(\mathbf{x})$  is such

$$z_1 = x_1, z_2 = \left(\frac{a_1\sqrt{x_1} + x_2}{a_2}\right)^2$$
(9)

The model in (7) is therefore expressed as,

$$\dot{\mathbf{x}}_1 = \mathbf{x}_2, \ \dot{\mathbf{x}}_2 = \frac{a_1 a_2}{2} \left( \frac{\sqrt{z_1}}{\sqrt{z_2}} - \frac{\sqrt{z_2}}{\sqrt{z_1}} \right) + \frac{a_1^2}{2} - a_2^2 + \frac{a_2}{2c} \frac{1}{\sqrt{z_2}} \mathbf{u}$$
 (10)

where  $z_1$  and  $z_2$  values are a function of  $x_1$  and  $x_2$  as indicated by (9).

Therefore, system dynamics is represented as,

$$\dot{\mathbf{x}}_1 = \mathbf{x}_2, \ \dot{\mathbf{x}}_2 = \mathbf{f} + \mathbf{\phi}\mathbf{u}, \mathbf{y} = \mathbf{x}_1,$$
 (11)

where,

$$f = \frac{a_1 a_2}{2} \left( \frac{\sqrt{z_1}}{\sqrt{z_2}} - \frac{\sqrt{z_2}}{\sqrt{z_1}} \right) + \frac{a_1^2}{2} - a_2^2, \ \varphi = \frac{a_2}{2c} \frac{1}{\sqrt{z_2}}$$
(12)

#### 3. Controller design

SMC is a type of control algorithm that is robust and nonlinear and based on the Lyapunov method. In this method, an n<sup>th</sup> order uncertain and nonlinear system is transformed into a 1<sup>st</sup> order system. The numerous advantages of SMC include its very straightforward design and durability

in handling dynamic characteristics and environmental disturbances. It is generally acknowledged that SMC offers a reliable solution to the control problem; as a result, it enables adapting to changes in the plant without noticeably diminishing the performance. Although SMC yields discontinuity in control results, it is obvious that the proposed control must direct the trajectory towards the switching surface and then be preserved on this surface. While using the SMC control algorithm, a problem encountered is to make the system respond to follow a particular trajectory.

The following single input single output system will be used to introduce the technique [28]:  $x^{(n)} = f(x) + \phi(x) \cdot u,$ (13)

where f and  $\phi$  are nonlinear functions of the states, u is the input, and x is the vector of states of order n. The control's goal is that the state should remain bound to the vector state trajectory  $x_d(t)$ . Taking into account the surface S(t):

$$S(x,t) = \left(\frac{d}{dt} + \lambda\right)^{n-1} \tilde{x}$$
(14)

where  $\tilde{x} = x - x_d$  denotes the error and  $\lambda$  is a constant > 0. Taking into consideration V =  $0.5S^2$  as a Lyapunov function, the control law must minimize the distance to the surface in (14) as well as all system states trajectory (slide mode). This is to say,

$$0.5\frac{\mathrm{d}}{\mathrm{dt}}\mathrm{S}^2 \le -\eta|\mathrm{S}|\,,\tag{15}$$

where the constant  $\eta$  is > 0. The entire state trajectory was plainly improved, becoming closer to the sliding surface in finite time, and will remain there indefinitely. It is argued that the sliding mode( $\dot{S} = 0$ ) occurred when the system is settled upon the surface.

As the surface is entered for the first time, the time is,

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$$t_{\text{reach}} \le \frac{S(0)}{\eta} \tag{16}$$

Using SMC, during its ideal two phases; reaching and sliding, the motion is restricted to the sliding surface [28]. The design process of the sliding mode has two steps. Firstly, designing a switching function S = 0 ensures that the sliding motion fulfills design requirements. Secondly, considering how the control law that will be chosen would describe the sliding mode in order to ensure that the conditions of presence and reaching are met [46]. The three different approaches for designing switching surfaces for a connected tank are described in the following subsections.

#### 3.1 Sliding surface with PD control

Given that H is the reference signal, the system error can be defined as

$$e = (h_2 - H) = (z_1 - H)$$
(17)

Then, PD-SMC will be

$$S = K_{p}e + K_{d}\dot{e} = K_{p}(z_{1} - H) + K_{d}\dot{z}_{1}$$
(18)

Differentiating (18) in terms of time, yields

$$\dot{S} = K_{\rm d} \ddot{z}_{\rm 1} + K_{\rm p} \ \dot{z}_{\rm 1} \tag{19}$$

$$\dot{S} = K_{d} \left( (-a_{1} \dot{z}_{1}/2\sqrt{z_{1}}) + (a_{2} \dot{z}_{2}/2\sqrt{z_{2}}) \right) + K_{p} \dot{z}_{1}$$
(20)

Solving (7) and (20) gives

$$\dot{S} = K_{d} \left[ \frac{a_{1}^{2} - 2a_{2}^{2}}{2} + \left( \frac{a_{1}a_{2}}{2} \right) \left( \frac{\sqrt{z_{1}}}{\sqrt{z_{2}}} - \frac{\sqrt{z_{2}}}{\sqrt{z_{1}}} \right) + \left( \frac{a_{2}}{2C\sqrt{z_{2}}} \right) u \right] + K_{p} \left( -a_{1}\sqrt{z_{1}} + a_{2}\sqrt{z_{2}} \right)$$
(21)

In order to verify the Lyapunov criterion, it is assumed that

$$\dot{S} = -K \operatorname{sgn}(S) \tag{22}$$

where 
$$sgn(S) = \begin{cases} +1, & \text{if } S > 0, \\ 0, & \text{if } S = 0, \\ -1, & \text{if } S < 0, \end{cases}$$

substituting in (21)

$$-K \operatorname{sgn}(S) = K_{d} \left[ \frac{a_{1}^{2} - 2a_{2}^{2}}{2} + \left( \frac{a_{1}a_{2}}{2} \right) \left( \frac{\sqrt{z_{1}}}{\sqrt{z_{2}}} - \frac{\sqrt{z_{2}}}{\sqrt{z_{1}}} \right) + \left( \frac{a_{2}}{2C\sqrt{z_{2}}} \right) u \right] + K_{p} \left( -a_{1}\sqrt{z_{1}} + a_{2}\sqrt{z_{2}} \right)$$
(23)

gives,

$$u = \left(\frac{2C\sqrt{z_2}}{a_2}\right) \left[ -\frac{a_1^2}{2} + a_2^2 - \left(\frac{a_1a_2}{2}\right) \left(\frac{\sqrt{z_1}}{\sqrt{z_2}} - \frac{\sqrt{z_2}}{\sqrt{z_1}}\right) \right] - (K_p K_d) (-a_1 \sqrt{z_1} + a_2 \sqrt{z_2}) - (K/K_d) \operatorname{sgn}(S)$$
(24)

The system states are now approaching the hyperplane using the control law in the equation above. The error vectors are compelled to tend to zero. Concurrently, the height  $h_2(t)$  will converge to the required value H. The SMC ensures that the output will asymptotically converge to the desired value.

The chattering problem affects the switching function. The chattering is caused by the sign function in the control signal. This implies that the control may change the value at any time without delay. Using the saturation function [28] could minimize chattering.

$$K \operatorname{sat}(\frac{S}{\Delta}) = \begin{cases} +1 & \text{for } (S/\Delta) \ge 1\\ S/\Delta & \text{for } -1 < S < 1\\ -1 & \text{for } (S/\Delta) \le 1 \end{cases}$$
(25)

where K > 0 is the switching gain and  $\Delta$  is the width of the boundary layer.

Using the saturation function to rewrite equation (24), gives,

$$u = \left(\frac{2C\sqrt{z_2}}{a_2}\right) \left[ -\frac{a_1^2}{2} + a_2^2 - \left(\frac{a_1a_2}{2}\right) \left(\frac{\sqrt{z_1}}{\sqrt{z_2}} - \frac{\sqrt{z_2}}{\sqrt{z_1}}\right) - \frac{(K_p)}{K_d} \right) (-a_1\sqrt{z_1} + a_2\sqrt{z_2}) - \left(\frac{K}{K_d}\right) \operatorname{sat}\left(\frac{S}{\Delta}\right)$$
(26)

### 3.2 Sliding surface with PID control

The PID model assists in quickly bringing the output of the system to the designed value ensuring minimal error and overshoot [47]. Because of its low cost and simplicity, it is the widely used controller in the industry.

The PID-SMC is constructed as:

$$S = K_{p}e + K_{I} \int edt + K_{d}\dot{e} = K_{p}(z_{1} - H) + K_{I} \int (z_{1} - H)dt + K_{d}\dot{z}_{1}$$
(27)

Differentiating (27) in terms of time leads to

$$\dot{S} = K_p \dot{z}_1 + K_I (z_1 - H) + K_d \ddot{z}_1$$
 (28)

Solving (7) and (28), gives

$$\dot{S} = K_{p}(a_{2}\sqrt{z_{2}} - a_{1}\sqrt{z_{1}}) + K_{I}(z_{1} - H) + K_{d}\left[\frac{a_{1}^{2} - 2a_{2}^{2}}{2} + \left(\frac{a_{1}a_{2}}{2}\right)\left(\frac{\sqrt{z_{1}}}{\sqrt{z_{2}}} - \frac{\sqrt{z_{2}}}{\sqrt{z_{1}}}\right) + \left(\frac{a_{2}}{2C\sqrt{z_{2}}}\right)u\right]$$
(29)

Providing a procedure same as introduced in (24), for S to be zero,

$$u = \left(\frac{2C\sqrt{z_2}}{a_2}\right) \left[ -\left(\frac{K_P}{K_d}\right) \left(a_2\sqrt{z_2} - a_1\sqrt{z_1}\right) - \left(\frac{K_I}{K_d}\right) \left(z_1 - H\right) - \frac{a_1^2}{2} + a_2^2 - \left(\frac{a_1a_2}{2}\right) \left(\frac{\sqrt{z_1}}{\sqrt{z_2}} - \frac{\sqrt{z_2}}{\sqrt{z_1}}\right) \right] - \left(\frac{K_I}{K_d}\right) \operatorname{sat}\left(\frac{S}{\Delta}\right)$$
(30)

#### **3.3 Dynamic sliding mode controller**

A dynamic SMC is introduced [36] to reduce chattering caused by static-SMC (PD-SMC). Assuming the scalers  $\alpha_1$ ,  $\alpha_2$  are positive, then sliding surface S is defined as follows,

$$S = \ddot{x}_1 + \alpha_1 \dot{x}_1 + \alpha_2 (z_1 - H)$$
(31)

where  $x_1$  is defined (11) and (12).

Sub. (11) and (12) in (31) gives,

$$S = \left(\frac{a_1 a_2}{2}\right) \left(\frac{\sqrt{z_1}}{\sqrt{z_2}} - \frac{\sqrt{z_2}}{\sqrt{z_1}}\right) + \frac{a_1^2}{2} - a_2^2 + \frac{a_2}{2C} \frac{1}{\sqrt{z_2}} u + \alpha_1 \left(-a_1 \sqrt{z_1} + a_2 \sqrt{z_2}\right) + \alpha_2 (z_1 - H)$$
(32)

Differentiating equation (31) in terms of time, and substituting from (8) and (10),

$$\dot{S} = \ddot{x}_{1} + \alpha_{1} \ddot{x}_{1} + \alpha_{2} \dot{x}_{1}$$

$$= f_{1} + \alpha_{1} \left( \frac{a_{1}a_{2}}{2} \left( \frac{\sqrt{z_{1}}}{\sqrt{z_{2}}} - \frac{\sqrt{z_{2}}}{\sqrt{z_{1}}} \right) + \frac{a_{1}^{2}}{2} - a_{2}^{2} + \frac{a_{2}}{2C} \frac{1}{\sqrt{z_{2}}} u \right) + \alpha_{2} (a_{2}\sqrt{z_{2}} - a_{1}\sqrt{z_{1}}) + \frac{a_{2}}{2C} \frac{1}{\sqrt{z_{2}}} \dot{u} - \frac{a_{2}}{4C} \frac{1}{\sqrt{z_{2}^{3}}} \left( a_{1}\sqrt{z_{1}} - 2a_{2}\sqrt{z_{2}} + \frac{1}{C}u \right) u$$
(33)

To meet the Lyapunov stability criterion, equation (33) is modified, where  $\dot{S} = -Ksat(S/\Delta)$ , as follows,

$$\dot{u} = \frac{-2C\sqrt{z_2}}{a_2} \left[ f_1 + \alpha_1 \left( \frac{a_1 a_2}{2} \left( \frac{\sqrt{z_1}}{\sqrt{z_2}} - \frac{\sqrt{z_2}}{\sqrt{z_1}} \right) + \frac{a_1^2}{2} - a_2^2 + \frac{a_2}{2C} \frac{1}{\sqrt{z_2}} u \right) + \alpha_2 (-a_1 \sqrt{z_1} + a_2 \sqrt{z_2}) + \text{Ksat} \left( \frac{S}{\Delta} \right) \right] + \frac{1}{2z_2} \left( a_1 \sqrt{z_1} - 2a_2 \sqrt{z_2} + \frac{1}{C} u \right) u$$
(34)

where,

$$f_{1} = \frac{-a_{1}a_{2}(z_{1}+z_{2})}{4\sqrt{(z_{1}z_{2})^{3}}} * \left(a_{1}z_{1}^{\frac{3}{2}} - 2a_{2}z_{1}\sqrt{z_{2}} + \frac{1}{c}z_{1}u + a_{1}z_{2}\sqrt{z_{1}} - a_{2}z_{2}^{\frac{3}{2}}\right)$$
(35)

The state trajectories co-occurring with the discontinuous function in equation (22) reveal a finite-time accessibility to zero starting from random initial conditions providing that K > 0. Since S is forced to zero, the output  $y = z_1 = h_2$  is regulated by the second-order dynamics  $\ddot{y} + \alpha_1 \dot{y} + \alpha_2 (y - H) = 0$  after such a finite time as a result, because  $\alpha_1$  and  $\alpha_2$  are positive scalars, the output y(t) will converge asymptotically to the required value H.

### 4. Results of Simulation

There are four to five tuning parameters for each control algorithm. In this case, it is impractical to select controller parameters through trial and error, as has been done in many prior studies, to obtain specific response requirements. The optimization toolbox of MATLAB is utilized in this study to determine the controller parameters' optimum values. The toolbox offers many tools for minimizing or maximizing objectives and functions while meeting specific restrictions. In this research, the method of gradient descent optimization is applied for its simplicity and effectiveness. Figure 2 depicts the optimization constraints applied in the three algorithms. The rise time (based on 90% of the final value) is set to 35 seconds, settling time (based on  $\pm 5\%$  of the final value) is set to 100 seconds, overshoot is set to 8%, and undershoot is set to 5%.

Two objective functions are chosen. The first one is the integral of the square of the difference in height between the level of liquid,  $h_{2}$ , and the required height,  $h_{2d}$ . This is referred to as the index of error and is given by,

$$Q_1 = \int (h_2 - h_{2d})^2 \, dt \tag{36}$$

The integral of the square of the control signal time derivative, u, is the second objective function. This function will be referred to as the index of chattering and is defined as

$$Q_2 = \int \left(\frac{du}{dt}\right)^2 . dt \tag{37}$$

As the rate of chattering of the signal reduces, so does this index given by (37). By minimizing (36) and (37) and maintaining the signal within the limitations specified above, the controller's optimum performance is provided. The optimization algorithm seeks the optimal combinations of controller parameters to achieve this goal. Table 1 illustrates the controller parameters search range.



Figure 2. Signal constraints

<b>Table 1.</b> Controller parameters search range				
Coefficient	Range	Coefficient	Range	
K	zero→∞	$\alpha_1$	$zero \rightarrow \infty$	
Δ	$zero \rightarrow \infty$	$\alpha_2$	$zero \rightarrow \infty$	
Kp	$zero \rightarrow \infty$			
$K_I$	$zero \rightarrow \infty$			
$K_d$	$zero \rightarrow \infty$			

 Table 1. Controller parameters search range

A 0.1-meter step input is applied to the connected tank system. Table 2 displays the parameters identified by the optimization technique for each controller to make the response meet the constraints indicated in Fig. 4 and fulfill both objective functions (36) and (37). Figures 3-5 depict the three control algorithms' responses to this disruption. Those figures show that PD-SMC, PID-SMC, and Dynamic-SMC have all met the desired level (steady-state error = 0) The PID-

SMC has a maximum overshoot of 4.9 percent, while the PD-SMC has the least overshoot of 1.59 percent. As observed in figures 3-b to 5-b, the three algorithms almost have no chattering. Table 3 compares the rising time, overshoot percentage, steady-state error, index of error, and settling time (five percent criterion) for the three controllers. According to the table, PD-SMC has the lowest index of error.

Table 2. Controllers tuning parameters				
Parameter	PD-SMC	PID-SMC	Dynamic-SMC	
K	2	10.2	1.036	
Δ	0.832	47.213	0.406	
K <sub>p</sub>	8.55	1.828	-	
K <sub>I</sub>	-	1.0*10 <sup>-5</sup>	-	
K <sub>d</sub>	50.865	1.099	-	
$\alpha_1$	-	-	0.233	
$\alpha_2$	_	_	0.012	

0.11 2.5 0.1 0.09 0.08 0.102 u (m<sup>3</sup>/sec) 0.07 1.5 0.101  $h_2(m)$ 0.06 0.05 0.04 0.03 0.5 0.02 0 i 0 0.01 50 100 150 0 Time(s) 100 150 Time(s) (B) (A)

Figure 3. Performance of PD-SMC (A) 0.1 m step response, (B) Control signal



Figure 4. Performance of PID-SMC (A) 0.1 m step response, (B) Control signal



Figure 5. Performance of dynamic-SMC (A) 0.1 m step response, (B) Control signal

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Algorithm	Rise time (s)	Overshoot (%)	Steady- state error (m)	Error index (m <sup>2</sup> .s)	Settling time (s)
PD-SMC	29	1.59	0	0.116	34
PID-SMC	28	4.85	0	0.117	33.6
Dynamic-SMC	31		0	0.158	40

**Table 3.** Step response comparison for the three SMC algorithms

## 5. Experimental Results

#### 5.1 System description

This section discusses the description of the experimental setup used to validate the mathematical model and test the performance of the developed control algorithm. The system shown in Fig. 6 consists of four water tanks, the dimensions of each tank are  $15 \times 15 \times 35$  cm<sup>3</sup>, with a maximum capacity of 7.88 L. These tanks can be connected in different configurations to facilitate investigating different coupled tank problems. However, the system will be used here to investigate the horizontal coupled tank problem. The main water tank at the bottom feeds two separately operated variable speed centrifugal pumps to control the output flow rates. All connections are made by  $\frac{1}{2}$ " copper pipes with 13 manually controlled ball valves.

Fig. 7 shows a schematic drawing of the main components used in controlling and monitoring the tank levels. The system consists of:

• Two calibrated level sensors (eTape – resistive type) with voltage divider circuit.

- A Variable Frequency Drive (VFD); frequency inverter (AC-220 V) connected with a PWM to an analogue converter circuit.
- Centrifugal pump (model QB 60 with max. flow = 35 L/min and max. head = 35 m)
- Microcontroller (Arduino Mega 2560).
- Host computer for data monitoring and recording



Figure 6. Coupled tanks liquid level system



Figure 7. Main components of the horizontal coupled tank system



Figure 8. Open-loop test

As shown in Fig. 7, tanks 1 and 2 are connected horizontally via a manually controlled ball valve, which is mounted on the flow channels. The system is configured as follows; the pump discharges water into tank 1, which discharges the water to tank 2 through an intermediate valve to provide dynamic coupling between the two tanks. To minimize the disturbance in the water level and the water splashes inside the tanks from the flowing water, the pipe inside tank 1 from

which the water enters the tank, is filled by holes. The height of the liquid level in the two tanks  $(h_1, h_2)$  is measured by two level sensors providing the feedback signal to the controller. The outputs from the two-level sensors are connected to the analog inputs of the microcontroller through a voltage divider circuit to convert the sensor resistance change into equivalent voltage.

Depending on the error signal value which is the difference between the measured level and the desired level, the controller sends a control signal (u) in the pulse width modulator (PWM) to the digital output pin of the microcontroller. Consequently, the PWM signal is sent to the PWM/Analog converter circuit to the frequency inverter. The frequency inverter, which is powered by an AC 220V power supply, controls the pump speed, hence controlling the amount of flow flowing into tank 1. The output of the two-level sensors and the control signal (u) are monitored and recorded using an external mode of the microcontroller which is connected to the host computer.

#### **5.2 Open-loop test**

In order to verify the derived mathematical model experimentally, an open loop test is implemented such that both the model and the physical system are subjected to a step input flow rate. The experiment is carried out considering all pipelines are filled with water. The level signal is recorded with time for both tests. As can be seen from Fig. 8, there is a little discrepancy between both signals due to model accuracy. However, as will be seen from the controllers' performance curves in the next section, the proposed SMC control algorithms can handle this discrepancy.

#### 5.3 Closed-loop test

In this section, experimental validation of the developed controllers' performances is carried out. The system is given a reference value of 10 cm starting from 4 cm height. The test rig is connected to the host computer through the Arduino Mega 2560 board by using Simulink-MATLAB. First, the PD-SMC model is loaded to the board which operates in external mode for monitoring sensors' readings ( $h_1$ ,  $h_2$ ) and the control signal (u). The adopted sample time is 1 second which is found suitable for the system dynamics (closed-loop system time constant  $\approx$  30 seconds). As can be seen from Fig. 9, the control signal suffered from a high level of fluctuation which affects the performance of the actuator (the frequency inverter and the pump). The actuators' sensitivity to that level of noise hindered the completion of the experimental validation under the effect of sensor noise. Such a high level of chattering could harm the used actuator system.

It is worth noting that all algorithms' controller parameters have been tuned to minimize chattering and error when there is no sensor noise. In order to treat that, a Gaussian-white-noise with SNR (signal-to-noise ratio) with a band of 19-23 dB is superimposed to the sensor in the simulation experiments. Then, in the presence of sensor noise, the optimization method is rerun to obtain the optimum controller parameters., as illustrated in Table 4.



Figure 9. Response of PD-SMC when sensor noise is present: (A) Step response, (B) control

Algorithm Parameter	PD-SMC	PID-SMC	Dynamic-SMC
K	$7.281*10^{6}$	$2.148*10^{12}$	0.146
Δ	$4.366*10^7$	9.600*10 <sup>14</sup>	2.179
K <sub>p</sub>	$3.259*10^{6}$	$6.494*10^7$	-
K	-	$6.182*10^{6}$	-
$K_d$	$2.042*10^{7}$	3.890*10 <sup>8</sup>	-
α1	-	_	0.3398
α2	_	_	0.0493

**Table 4.** New tuning parameters after adding white noise

In order to mitigate sensor noise, a low-pass filter with a cut-off frequency of approximately 1 Hz is added to the controller algorithm. The cut-off frequency is chosen to filter out the sensor noise. Fig. 10 shows the performance of the PD-SMC after adding the designed low-pass filter.



Figure 10. Response of PD-SMC with low-pass filter: (A) 0.1 m step response, (B) Control signal

The same experiment is carried out by using PID-SMC and dynamic-SMC models, see Fig. 11 and Fig. 12. As illustrated in the graphs, PD-SMC showed the minimum value of steady-state error and fluctuation level in the control signal when compared with the other control algorithms, PID-SMC, and dynamic-SMC.







Figure 12. Performance of dynamic-SMC with low-pass filter: (A) 0.1 m step response, (B) control signal

# 6. Conclusion

In the current paper, the three SMC algorithms' performance for controlling the level of liquid in a horizontally connected tank system is explored. Simulation of the problem facilitates selecting the controller parameter based on optimizing a performance index. The control algorithms are highly sensitive to sensor noise which necessitates filter-out the noises in the band of frequencies of concern. PD-SMC has a superior performance over the other two algorithms, PID-SMC and dynamic-SMC both in simulation and experimental results.

In future work of this research, mathematical analysis of the controller sensitivity to sensor noise level is to be done. Also, a second-dynamic SMC is to be included in the investigation.

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## • Conflict of Interest

The authors have no conflicts of interest to declare.

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