# THE MOTION TRAJECTORY-BASED FINITE-TIME CONTROL FOR THE MARINE SURFACE VEHICLE

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# SUMMARY

This brief is devoted to the predesigned motion trajectory-based finite time dynamic positioning (DP) control for a marine surface vehicle (MSV) with unknown external disturbances. Firstly, a preset motion trajectory is presented through establishing the relationship function among position tracking errors and heading tracking error, facilitating the MSV to arrive in the equilibrium point along the pre-designed trajectory. Furthermore, a novel nonsingular and fast terminal sliding mode control (NTSMC) approach is investigated, which ensures faster convergence rate and better stability performance of the close-loop system than the conventional backstepping control approach. What's more, by incorporating the adaptive technique with the NTSMC approach, an adaptive nonsingular and fast terminal sliding mode control (ANTSMC) strategy is addressed. Compared to the NTSMC approach, it strengthens robustness to disturbances and guarantees system states to converge to a closer neighborhood of the equilibrium point. Finally, simulation results illustrate the remarkable effectiveness of proposed control schemes.

## NOMENCLATURE

x	Longitudinal position point (m)
у	Horizontal position point (m)
Ψ	Heading (rad)
u	Longitudinal velocity (m/s)
v	Horizontal velocity (m/s)
r	Yawing angular velocity (rad/s)
MSV	Marine Surface Vehicle
NTSMC	Nonsingular and fast terminal sliding mode control
ANTSMC	Adaptive non-singular and fast terminal sliding mode control

# 1. INTRODUCTION

The dynamic positioning (DP) technology allows a ship to maintain its own position and heading at a fixed location or navigate along a predetermined track exclusively in a way that its own propulsion system counteracts environmental disturbances inducted by waves, currents, and wind (Fossen, & Grovlen, 1998). Compared with traditional anchor mooring positioning, the DP technology has a lot of advantages inclusive of easy operation, high positioning accuracy and avoidance of destroying riverbed, and it has been extensively applied in offshore operations such as wreck investigation, underwater cable laying, and oil drilling (Du *et al*, 2014). Therefore, it is increasingly essential for undertaking of relevant researches of DP technology.

As for DP control issues, nonlinear control methodologies such as backstepping control, sliding mode control, and mode predictive control nowadays have been playing a dominant position instead of previous linear control schemes of PID control and optimal control. Among them, correlational researches

for backstepping control approach were practically active. More excitingly, abundant theoretical and experimental results are accomplished. In (Fossen, & Grovlen, 1998), a vector backstepping technique-based nonlinear control scheme was addressed, achieving a globally uniform asymptotic stability. Based on the control method (Fossen, & Grovlen, 1998), dynamic surface control (DSC) and adaptive technique were incorporated in (Du et al, 2014) to solve the problem of "explosion of complexity" and suppress unknown timevarying disturbances. In (Zhang et al, 2017) "minimal learning parameter" technique was originally introduced into backstepping algorithm, and the on-line learning parameters were reduced substantially, lowering the computational burden. As for the output feedback case, a backstepping method-based output feedback control law was given, and an adaptive observer was deducted to estimate the speed of the vessel and unknown parameters (Do, 2007). In (Du et al, 2015), supposing unknown of the vessel position, heading and speed, an output feedback controller coupling with a high-gain observer was presented, and the consequent simulation results were satisfying.

Nevertheless, overviewing the aforementioned research works, we observe two questions: on one hand, although the backstepping technique-based control approach assists to get good simulation results, the motion trajectory of MSV has never been involved in the past. In fact, we usually hope that the MSV should perform the DP task along a pre-planned trajectory that is satisfying to the practical requirement. On the other hand, the forgoing control methodologies can all achieve the exponential convergence and the tracking errors of closed-loop system were made asymptotically stable or globally uniformly ultimately bounded (GUUB) in infinite time. Whereas, in view of the practical engineering application, it is valuable that tracking errors are stable as early as possible in finite settling time. The finite-time control algorithm (Bhat & Bernstein 1997-Bhat & Bernstein, 2002) was proposed with the advantage of finite-time convergence as well as the interference rejection, and then it was successfully applied in a variety of engineering areas, such as underwater robot, mechanical arm and spacecraft attitude. Unfortunately, there are several literatures on applying the valuable method in DP control of MSV until now. In (Huang, 2018), based on the finite-time Lyapunov theory, an adaptive backstepping DP control was addressed, and model uncertainties and the disturbance upper bound were compensated by utilizing Radial Basis Function (RBF) network. In (Zhang, 2018), a nonsingular backstepping terminal sliding mode control approach was given, incorporating backstepping method and terminal sliding mode method.

In allusion of two issues, we propose a motion trajectorybased finite-time DP control scheme. Inspired by the design method in [30], a motion trajectory is firstly put forward. Next, referring to the design of terminal sliding mode in (Zhu *et al*, 2016), a NTSMC algorithm is proposed. It certifies that the proposed algorithm guarantees faster convergence characteristic and higher tracking accuracy, comparing to conventional asymptotic stability control algorithm. In addition, an adaptive method combined with the NTSMC algorithm is applied in estimating the upper bound of unknown disturbances, which can strengthen robustness to disturbances of the closed-loop system in further, guaranteeing system states to converge to a closer neighborhood of the equilibrium point.

# 2. PRELIMINARIES AND PROBLEM FORMULATION

#### 2.1 PRELIMINARIES

Consider the nonlinear system:

$$\dot{\mathbf{x}} = \boldsymbol{f}(\mathbf{x}, \boldsymbol{u}), \boldsymbol{x} \in \mathbb{R}^n \tag{1}$$

where  $\mathbf{x}$  is the state vector,  $\mathbf{u}$  is control input,  $f(\cdot):\mathbb{R}^n \to \mathbb{R}^n$  is a continuous function, and f(0) = 0. Suppose there exists a time function  $T(\mathbf{x})$ , and the time  $t \ge T(\mathbf{x})$ , satisfying  $\mathbf{x}(t) = 0$ , then the system (1) is finite-time stable[Wang *et al*, 2009].

If for any  $\mathbf{x}(t_0) = \mathbf{x}_0$ , there exists  $T(\varepsilon, \mathbf{x}) < \infty$ , such that  $\|\mathbf{x}(t)\| < \varepsilon$ , for all  $t > t_0 + T$ , and the nonlinear system is the practical finite-time stable (PFS).

Lemma 1: considering the system (1), suppose there exists a continuous definite function  $V(\mathbf{x})$ , such that  $\dot{V}(\mathbf{x}) \leq -\mathbf{c}V^{a}(\mathbf{x})$  for some c > 0, 0 < a < 1, then the system (1) is finite-time stable <sup>[18]</sup> and the finite-time *T* satisfies

$$T_{reach} \le \frac{V^{1-a}(x_0)}{c(1-a)} \tag{2}$$

where  $V(\mathbf{x}_0)$  is the initial value of  $V(\mathbf{x})$ .

Lemma 2: considering the system (1), suppose there exists a continuous definite function  $V(\mathbf{x})$ , such that  $\dot{V}(\mathbf{x}) \leq -\mathbf{c}V^a(\mathbf{x}) + \zeta$  for some  $\lambda > 0, 0 < a < 1$  and  $0 < \zeta < \infty$ . Then the system (1) is practical finite-time stable (PFS)<sup>[19]</sup> and the finite-time satisfies

$$T_{reach} \le \frac{V^{1-a}(x_0)}{c\theta_0(1-a)} \tag{3}$$

where  $0 < \theta \le 1$ , and  $V(\mathbf{x}_0)$  is the initial value of  $V(\mathbf{x})$ .

Lemma 3: if any  $x_i, i = 1, 2, \dots, n$ , and  $r \in (0, 1)$ , the following inequalities hold<sup>[20]</sup>:

$$\sum_{i=1}^{n} |x_i|^{1+r} \ge \left(\sum_{i=1}^{n} |x_i|^2\right)^{\frac{1+r}{2}}$$
(4)

Lemma 4: for any  $a \in \mathbb{R}, b \in \mathbb{R}, p \ge 1$  is a constant, the following inequalities hold<sup>[Qian & Li, 2006]</sup>:

$$|a+b|^{p} \le 2^{p-1} |a^{p}+b^{p}|, (|a|+|b|)^{\frac{1}{p}} \le |a|^{\frac{1}{p}} + |b|^{\frac{1}{p}}$$
(5)

Lemma 5: for c,d and any real-valued function  $\gamma(x, y) > 0$ , the inequalities hold<sup>[Qian & Lin, 2015]</sup>:

$$|x|^{c} |y|^{d} \leq \frac{c}{c+d} \gamma(x,y) |x|^{c+d} + \frac{d}{c+d} \gamma(x,y)^{-\frac{c}{d}} |y|^{c+d}$$
(6)

Assumption 1:  $\boldsymbol{\tau}_{w} = [\boldsymbol{\tau}_{wu}, \boldsymbol{\tau}_{wv}, \boldsymbol{\tau}_{wr}]^{\mathrm{T}}$  is the unknown external disturbance (wind, wave), satisfying  $|\boldsymbol{\tau}_{w}| \leq \boldsymbol{\tau}_{w}^{*} < \infty, \boldsymbol{\tau}_{w}^{*} = [\boldsymbol{\tau}_{wu}^{*}, \boldsymbol{\tau}_{wv}^{*}, \boldsymbol{\tau}_{wr}^{*}]^{\mathrm{T}}$  is the upper bound, which is unknown vector, only for analysis.

#### 2.2 DYNAMIC MODEL OF MSV

When analysing the motion of MSV in three DOF (degree of freedom), we need to establish two coordinate frames as shown in Figure. 1.

The earth-fixed reference frame  $X_0OY_0$  is considered to be inertial, with  $OX_0$  -axis pointing to north and  $OY_0$  -axis pointing to east.  $\boldsymbol{\eta} = [x, y, \psi]^T$  is expressed as the 3 DOF position (x, y) and heading angle  $(\psi)$  of the vessel in this inertial frame. Besides, the body-fixed frame *XAY* is attached to the vessel with its origin *A* coincident with the centre of gravity. The *AX*-axis points to head of the vessel, and the *AY*-axis is perpendicular to the *AX*-axis and points to starboard side. Let  $v = [u, v, r]^T$  be the corresponding velocity in surge, sway and yaw, respectively in the body-fixed frame.



Figure. 1. Earth-fixed and body-fixed reference frames

The vectorial model of MSV is expressed as [Fossen 2002]

$$\dot{\boldsymbol{\eta}} = \boldsymbol{R}(\boldsymbol{\psi})\boldsymbol{v} \tag{7}$$

$$\boldsymbol{M}\boldsymbol{\dot{\boldsymbol{v}}} + \boldsymbol{D}(\boldsymbol{v})\boldsymbol{v} = \boldsymbol{\tau}_{v} + \boldsymbol{\tau}_{w} + \boldsymbol{\omega}_{b}$$
(8)

$$\boldsymbol{\eta}_{output} = \boldsymbol{\eta} + \boldsymbol{\omega}_{\eta} \tag{9}$$

$$\boldsymbol{R}(\boldsymbol{\psi}) = \begin{bmatrix} \cos \boldsymbol{\psi} & -\sin \boldsymbol{\psi} & 0\\ \sin \boldsymbol{\psi} & \cos \boldsymbol{\psi} & 0\\ 0 & 0 & 1 \end{bmatrix}$$
(10)

$$\boldsymbol{M} = \begin{bmatrix} m - X_{ii} & 0 & 0 \\ 0 & m - Y_{ii} & m X_G - Y_{ii} \\ 0 & m X_G - Y_{ii} & I_z - N_{ii} \end{bmatrix}$$
(11)

$$\boldsymbol{D}(\boldsymbol{v})\boldsymbol{v} = \begin{bmatrix} -X_{u}u - X_{|\mu|u} |u|u + Y_{v}vr + Y_{r}rr \\ -X_{u}ur - Y_{v}v - Y_{r}r - Y_{|\nu|v} |v|v - Y_{|\nu|r} |v|r \\ (X_{u} - Y_{v})uv - Y_{v}uv - N_{v}v - N_{r}r - N_{|\nu|v} |v|v - N_{|\nu|r} |v|r \end{bmatrix}$$
(12)

where  $\mathbf{R}(\psi)$  denotes the velocity transformation matrix.  $\boldsymbol{\tau}_{\nu} = [\boldsymbol{\tau}_{u}, \boldsymbol{\tau}_{v}, \boldsymbol{\tau}_{r}]^{\mathrm{T}}$  is the effective control input,  $\boldsymbol{\tau}_{w} = [\boldsymbol{\tau}_{wu}, \boldsymbol{\tau}_{wv}, \boldsymbol{\tau}_{wr}]^{\mathrm{T}}$  is external disturbance induced by waves, wind.  $\boldsymbol{M}$  is inertia matrix which includes hydrodynamic additional mass. It is symmetric positive definite, i.e.,  $\boldsymbol{M} = \boldsymbol{M}^{T} > 0$ .  $\boldsymbol{m}$  and  $I_{z}$  are mass of the MSV, moment of inertia, while  $Y_{\psi}, Y_{\psi}$  and  $N_{\psi}$  denote added mass and added moment of inertia. The expression  $\boldsymbol{D}(\boldsymbol{v})\boldsymbol{v}$  that consists of the Coriolis and centripetal force/moment and the nonlinear damping ones is the nonlinear hydrodynamic function.  $X_u, X_{|u|u}, Y_v, N_v$ , etc., denote the corresponding hydrodynamic derivatives.  $\boldsymbol{\omega}_b \in \mathbb{R}^3$  and  $\boldsymbol{\omega}_\eta \in \mathbb{R}^3$  are process noise vectors and measurement noise vector, respectively.

Actually, the abovementioned vectorial model, i.e., equations (7), (8) and (9), is the simulation model. In fact, according to the purpose of the controller design in this brief, the following design vectorial model is chosen, in which we replace nonlinear function D(v)v with the linear hydrodynamic function Dv and ignore the process noise vector and measurement noise vector.

$$\dot{\boldsymbol{\eta}} = \boldsymbol{R}(\boldsymbol{\psi})\boldsymbol{\upsilon} \tag{13}$$

$$\boldsymbol{M}\boldsymbol{\dot{\boldsymbol{v}}} + \boldsymbol{D}\boldsymbol{\boldsymbol{v}} = \boldsymbol{\tau}_{\boldsymbol{v}} + \boldsymbol{\tau}_{\boldsymbol{w}} \tag{14}$$

$$\boldsymbol{D}\boldsymbol{v} = \begin{bmatrix} -X_u u \\ -Y_v v - X_u r \\ -N_v v - N_r r \end{bmatrix}$$
(15)

#### 2.3 MOTION TRAJECTORY DESIGN

Here, a motion trajectory generating method is given, by establishing a relationship function among position tracking errors  $\eta_{e1}$ ,  $\eta_{e2}$  and heading error  $\eta_{e3}$ .

$$\eta_e = \eta - \eta_r$$

where  $\boldsymbol{\eta}_e = [\eta_{e1}, \eta_{e2}, \eta_{e3}]^{\mathrm{T}}$  is tracking error vector,  $\boldsymbol{\eta}_r = [x_r, y_r, \psi_r]^{\mathrm{T}}$  is the reference tracking attitude.

$$\eta_{e1} = x - x_r, \eta_{e2} = y - y_r, \eta_{e3} = \psi - \psi_r.$$

Design the following relationship function

$$h = \frac{180}{\pi} \eta_{e3} + \eta_{e2} - \eta_{e1}$$

The projection of *h* within *xy* plane is  $h_1 = \eta_{e2} - \eta_{e1}$ , while, the projection of *h* within *x\u03c6* plane is  $h_2 = \eta_{e3} - \frac{\pi}{180} \eta_{e1}$ .

When 
$$h_1 = h_2 = 0$$
,  $y = x - x_r + y_r$ ,  $\psi = \frac{\pi}{180}(x - x_r) + \psi_r$ .

#### 2.4 CONTROLLER DESIGN

In the kinematic design, a virtual control law is presented based on the backstepping technique; in the dynamic design, the virtual control law is regarded as the tracking target and a novel backstepping NTSMC approach is investigated. Step 1: the deduction of virtual control law  $a_{\nu}$ 

Define the transforming tracking error and velocity error

$$\boldsymbol{\eta}_{eh} = [\boldsymbol{\eta}_{e1}, \boldsymbol{h}_1, \boldsymbol{h}_2]^{\mathrm{T}}$$
(16)

$$\boldsymbol{v}_e = \boldsymbol{v} - \boldsymbol{a}_v \tag{17}$$

where  $\eta_{eh}$  is transforming tracking error,  $v_e$  is the velocity error, let  $\boldsymbol{a}_{\nu} = -\boldsymbol{R}_{h}(\boldsymbol{\psi})^{-1}\boldsymbol{K}_{1}Sig(\boldsymbol{\eta}_{eh})^{a}$  virtual control law.

Where

$$Sig(\boldsymbol{\eta}_{eh})^{a} = \begin{bmatrix} |\eta_{eh}(1)|^{a} \operatorname{sgn}(\eta_{eh}(1)) \\ |\eta_{eh}(2)|^{a} \operatorname{sgn}(\eta_{eh}(2)) \\ |\eta_{eh}(3)|^{a} \operatorname{sgn}(\eta_{eh}(3)) \end{bmatrix},$$
  
$$\boldsymbol{K}_{1} = diag(k_{11}, k_{12}, k_{13}), 0 < a < 1$$

The derivative of (16)

$$\dot{\boldsymbol{\eta}}_{eh} = \left[\dot{\boldsymbol{\eta}}_{e1}, \dot{\boldsymbol{h}}_{1}, \dot{\boldsymbol{h}}_{2}\right]^{\mathrm{T}}$$
(18)

$$\dot{h}_{1} = \dot{y} - \dot{y}_{r} - \dot{\eta}_{e1} = (\sin\psi - \cos\psi)u + (\cos\psi + \sin\psi)v - \dot{y}_{r} + \dot{x}_{r}$$
(19)

$$\dot{h}_2 = \dot{\eta}_{e3} - \frac{\pi}{180} \dot{\eta}_{e1} = r - \frac{\pi}{180} (\cos \psi u - \sin \psi v) - \dot{\psi}_r + \frac{\pi}{180} \dot{x}_r$$
(20)

Substituting (19) and (20) into (18), obtain

$$\dot{\boldsymbol{\eta}}_{eh} = \begin{bmatrix} \cos\psi & -\sin\psi & 0\\ \sin\psi - \cos\psi & \cos\psi + \sin\psi & 0\\ -\frac{\pi}{180}\cos\psi & \frac{\pi}{180}\cos\psi & 1 \end{bmatrix} \begin{bmatrix} u\\ v\\ r \end{bmatrix} - \begin{bmatrix} \dot{x}_r\\ \dot{y}_r - \dot{x}_r\\ \dot{\psi}_r - \frac{\pi}{180}\dot{x}_r \end{bmatrix}$$
(21)  
Let  $\boldsymbol{R}_h(\psi) = \begin{bmatrix} \cos\psi & -\sin\psi & 0\\ \sin\psi - \cos\psi & \cos\psi + \sin\psi & 0\\ -\frac{\pi}{180}\cos\psi & \frac{\pi}{180}\sin\psi & 1 \end{bmatrix}$   
 $\dot{\boldsymbol{\eta}}_{rh} = \begin{bmatrix} \dot{x}_r\\ \dot{y}_r - \dot{x}_r\\ \dot{\psi}_r - \frac{\pi}{180}\dot{x}_r \end{bmatrix}$   
 $\dot{\boldsymbol{\eta}}_{eh} = \boldsymbol{R}_h(\psi)\boldsymbol{v} - \dot{\boldsymbol{\eta}}_{rh}$ (22)

Substituting (17) into (22) yields  

$$\dot{\boldsymbol{\eta}}_{eh} = \boldsymbol{R}_h(\psi)\boldsymbol{a}_v + \boldsymbol{R}_h(\psi)\boldsymbol{v}_e - \dot{\boldsymbol{\eta}}_{rh}$$
(23)

Design the Lyapunov candidate

$$V_1 = \frac{1}{2} \boldsymbol{\eta}_{eh}^{\mathrm{T}} \boldsymbol{\eta}_{eh}$$
(24)

Time differentiation of  $V_1$  gives

$$\dot{V}_{1} = \boldsymbol{\eta}_{eh}^{\mathrm{T}} \left[ \boldsymbol{R}_{h}(\boldsymbol{\psi}) \boldsymbol{a}_{\upsilon} + \boldsymbol{R}_{h}(\boldsymbol{\psi}) \boldsymbol{v}_{e} - \dot{\boldsymbol{\eta}}_{rh} \right]$$
(25)

Take the virtual control law  $a_{\nu}$ , get

$$\dot{V}_1 = -\boldsymbol{\eta}_{eh}^{\mathrm{T}} \boldsymbol{K}_1 Sig(\boldsymbol{\eta}_{eh})^a + \boldsymbol{\eta}_{eh}^{\mathrm{T}} \boldsymbol{R}_h(\boldsymbol{\psi}) \boldsymbol{v}_e \qquad (26)$$

If  $v_e = 0$ ,  $\dot{V}_1 = -\sum_{i=1}^3 k_{1i} |\eta_{ehi}|^{1+a}$ . According to Lemma 1, we can obtain

$$\dot{V}_{1} = -\sum_{i=1}^{3} K_{1i} \left| \eta_{ehi} \right|^{1+a} \le -2^{\frac{1+a}{2}} \lambda_{\min}(K_{1}) V_{1}^{\frac{1+a}{2}}$$

Step 2: obtain the control  $\tau_{\text{NTSMC}}$ , when ignoring the external disturbances.

Refer to (Zhu et al, 2016), a non-singular and fast terminal sliding mode surface is designed.

$$\boldsymbol{s} = \boldsymbol{v}_e + \boldsymbol{d}_1 \int_0^t \boldsymbol{v}_e d\tau + \boldsymbol{d}_2 \int_0^t Sig(\boldsymbol{v}_e)^q d\tau$$
(27)

Design the NTSMC approach

$$\boldsymbol{\tau}_{\text{NTSMC}} = \boldsymbol{K}_2 Sig(\boldsymbol{s})^a + \boldsymbol{D}\boldsymbol{v} + \boldsymbol{M}\dot{\boldsymbol{a}}_{\nu} - \boldsymbol{M}\boldsymbol{d}_1\boldsymbol{v}_e - \boldsymbol{M}\boldsymbol{d}_2 Sig(\boldsymbol{v}_e)^q - \boldsymbol{s}^{-1}\boldsymbol{\eta}_{eh}^{\text{T}}\boldsymbol{R}_h(\boldsymbol{\psi})\boldsymbol{v}_e$$
(28)

where

(22)

$$Sig(\boldsymbol{v}_{e})^{q} = \begin{bmatrix} |\upsilon_{e}(1)|^{q} \operatorname{sgn}(\upsilon_{e}(1)) \\ |\upsilon_{e}(2)|^{q} \operatorname{sgn}(\upsilon_{e}(2)) \\ |\upsilon_{e}(3)|^{q} \operatorname{sgn}(\upsilon_{e}(3)) \end{bmatrix}, 0.5 < q < 1,$$
  
$$\boldsymbol{d}_{1} = diag(\boldsymbol{d}_{11}, \boldsymbol{d}_{12}, \boldsymbol{d}_{13}), \boldsymbol{d}_{2} = diag(\boldsymbol{d}_{21}, \boldsymbol{d}_{22}, \boldsymbol{d}_{23}),$$
  
$$\boldsymbol{K}_{2} = diag(\boldsymbol{k}_{21}, \boldsymbol{k}_{22}, \boldsymbol{k}_{23})$$

Design the Lyapunov candidate

$$V_2 = V_1 + \frac{1}{2} \boldsymbol{s}^{\mathrm{T}} \boldsymbol{M} \boldsymbol{s}$$

The derivative of  $V_2$  along with (26) and (27)

$$\dot{V}_{2} = \dot{V}_{1} + \boldsymbol{s}^{\mathrm{T}} \boldsymbol{M} \dot{\boldsymbol{s}} = -\boldsymbol{\eta}_{eh}^{\mathrm{T}} \boldsymbol{K}_{1} Sig(\boldsymbol{\eta}_{eh})^{a} + \boldsymbol{\eta}_{eh}^{\mathrm{T}} \boldsymbol{R}_{h}(\boldsymbol{\psi}) \boldsymbol{v}_{e} - \boldsymbol{s}^{\mathrm{T}} \boldsymbol{M} (\dot{\boldsymbol{v}}_{e} + \boldsymbol{d}_{1} \boldsymbol{v}_{e} + \boldsymbol{d}_{2} \operatorname{Sig}(\boldsymbol{v}_{e})^{q})$$
(29)

Substituting  $\tau_{\text{NTSMC}}$  into (29) yields

$$\dot{V}_{2} = -\eta_{eh}^{\mathrm{T}} \mathbf{K}_{1} Sig(\eta_{eh})^{a} - \mathbf{s}^{\mathrm{T}} \mathbf{K}_{2} Sig(\mathbf{s})^{a}$$

$$\leq -2^{\frac{1+a}{2}} \lambda_{\min}(\mathbf{K}_{1}) V_{1}^{\frac{1+a}{2}} - 2^{\frac{1+a}{2}} \frac{\lambda_{\min}(\mathbf{K}_{2})}{\lambda_{\max}(\mathbf{M})^{\frac{1+a}{2}}} (\frac{1}{2} \mathbf{s}^{\mathrm{T}} \mathbf{M} \mathbf{s})^{\frac{1+a}{2}}$$

$$\leq -m V_{2}^{\frac{1+a}{2}}$$

where  $m = 2^{\frac{1+a}{2}} \min \left\{ \lambda_{\min}(\boldsymbol{K}_1), \frac{\lambda_{\min}(\boldsymbol{K}_2)}{\lambda_{\max}(\boldsymbol{M})^{\frac{1+a}{2}}} \right\}$ 

 $\lambda_{\min}(\bullet)$  is the minimum eigenvalue of a matrix and  $\lambda_{\max}(\bullet)$  is the maximum eigenvalue of a matrix.

According to Lemma 1, and the closed-loop system is finite time stable under the condition of ignoring external disturbances. And the tracking errors will converge to the equilibrium point in finite time

$$T_{reach} \le \frac{V_2^{\frac{1+a}{2}}(\mathbf{x}(0))}{m(1 - \frac{1+a}{2})}$$

Step 3: obtain the practical ANTSMC strategy  $\tau_{\text{ANTSMC}}$ 

$$\boldsymbol{\tau}_{\text{ANTSMC}} = \boldsymbol{\tau}_{\text{NTSMC}} - \hat{\boldsymbol{\tau}}_{w}^{*}$$
(30)

where  $\boldsymbol{\tau}_{\text{NTSMC}}$  is the control scheme in (28),  $\hat{\boldsymbol{\tau}}_{w}^{*} = \left[\hat{\boldsymbol{\tau}}_{wu}^{*}, \hat{\boldsymbol{\tau}}_{wv}^{*}, \hat{\boldsymbol{\tau}}_{wr}^{*}\right]^{\text{T}}$  is the upper bound estimated vector of disturbances, and the adaptive law

$$\dot{\hat{\boldsymbol{\tau}}}_{w}^{*} = \boldsymbol{K}_{3} \left[ \boldsymbol{\vartheta} \boldsymbol{s} - \boldsymbol{K}_{4} (\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0}) \right]$$
(31)

$$\begin{split} &\mathcal{G} = diag(\tanh(\frac{s(1)}{\varepsilon_1}), \tanh(\frac{s(2)}{\varepsilon_2}), \tanh(\frac{s(3)}{\varepsilon_3})), \\ &\mathbf{K}_3 = diag(k_{31}, k_{32}, k_{33}), \\ &\mathbf{K}_4 = diag(k_{41}, k_{42}, k_{43}), \\ &\varepsilon_1, \\ &\varepsilon_2, \\ &\varepsilon_3 \text{ are design parameters.} \end{split}$$

 $\tilde{\boldsymbol{\tau}}_{w}^{*} = \hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{*}$  is estimating error vector,  $\boldsymbol{\tau}_{w}^{*}$  is upper bound vector of external disturbances,  $\boldsymbol{\tau}_{w}^{0} = \left[\boldsymbol{\tau}_{wu}^{0}, \boldsymbol{\tau}_{wv}^{0}, \boldsymbol{\tau}_{wr}^{0}\right]^{\mathrm{T}}$  is the design constant vector.

Design the Lyapunov candidate

$$V_3 = V_1 + \frac{1}{2} \boldsymbol{s}^{\mathrm{T}} \boldsymbol{M} \boldsymbol{s} + \frac{1}{2} \tilde{\boldsymbol{\tau}}_{w}^{* \mathrm{T}} \boldsymbol{K}_3^{-1} \tilde{\boldsymbol{\tau}}_{w}^{*}$$
(32)

Derivate  $V_3$  along with (26) and (27)

$$\dot{V}_{3} = \dot{V}_{1} + \boldsymbol{s}^{\mathrm{T}} \boldsymbol{M} \dot{\boldsymbol{s}} + \tilde{\boldsymbol{\tau}}_{w}^{*\mathrm{T}} \boldsymbol{K}_{3}^{-1} \dot{\boldsymbol{\tau}}_{w}^{*} = -\boldsymbol{\eta}_{eh}^{\mathrm{T}} \boldsymbol{K}_{1} Sig(\boldsymbol{\eta}_{eh})^{a} - \boldsymbol{s}^{\mathrm{T}} \boldsymbol{K}_{2} Sig(\boldsymbol{s})^{a} + \sum_{i=1}^{3} |\boldsymbol{s}(i)| \boldsymbol{\tau}_{w}^{*}(i) - \boldsymbol{s}^{\mathrm{T}} \boldsymbol{\vartheta} \boldsymbol{\tau}_{w}^{*} - \tilde{\boldsymbol{\tau}}_{w}^{*\mathrm{T}} \boldsymbol{K}_{4} (\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0})$$

$$(33)$$

According to Assumption 1, and substituting (14), (17), (28) and (31) into (33) yields

$$\dot{V}_{3} = \dot{V}_{1} + \boldsymbol{s}^{\mathrm{T}} \boldsymbol{M} \dot{\boldsymbol{s}} + \tilde{\boldsymbol{\tau}}_{w}^{*\mathrm{T}} \boldsymbol{K}_{3}^{-1} \dot{\boldsymbol{\tau}}_{w}^{*} = -\boldsymbol{\eta}_{eh}^{\mathrm{T}} \boldsymbol{K}_{1} Sig(\boldsymbol{\eta}_{eh})^{a} - \boldsymbol{s}^{\mathrm{T}} \boldsymbol{K}_{2} Sig(\boldsymbol{s})^{a} + \sum_{i=1}^{3} |\boldsymbol{s}(i)| \boldsymbol{\tau}_{w}^{*}(i) - \boldsymbol{s}^{\mathrm{T}} \boldsymbol{\vartheta} \boldsymbol{\tau}_{w}^{*} - \tilde{\boldsymbol{\tau}}_{w}^{*\mathrm{T}} \boldsymbol{K}_{4} (\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0})$$

$$(34)$$

Applying hyperbolic tangent function  $tanh(\bullet)$ , and for any  $\zeta > 0, a \in \mathbb{R}$ , get

$$0 \le \left|a\right| - a \tanh(\frac{a}{\zeta}) \le \beta \zeta$$

where  $\beta$  is a constant, satisfying  $\beta = e^{-(\beta+1)}, \beta = 0.2785$ .

$$\dot{V}_{3} = \dot{V}_{1} + \boldsymbol{s}^{\mathrm{T}} \boldsymbol{M} \dot{\boldsymbol{s}} + \tilde{\boldsymbol{\tau}}_{w}^{*\mathrm{T}} \boldsymbol{K}_{3}^{-1} \dot{\boldsymbol{\tau}}_{w}^{*} = -\boldsymbol{\eta}_{eh}^{\mathrm{T}} \boldsymbol{K}_{1} Sig(\boldsymbol{\eta}_{eh})^{a} - \boldsymbol{s}^{\mathrm{T}} \boldsymbol{K}_{2} Sig(\boldsymbol{s})^{a} \\ - \tilde{\boldsymbol{\tau}}_{w}^{*\mathrm{T}} \boldsymbol{K}_{4}(\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0}) + 0.2785 \boldsymbol{E}^{\mathrm{T}} \boldsymbol{\tau}_{w}^{*}.$$

where  $\boldsymbol{E} = [\varepsilon_1, \varepsilon_2, \varepsilon_3]^{\mathrm{T}}$ .

$$\dot{V}_{3} \leq -2^{\frac{1+a}{2}} \lambda_{\min}(\mathbf{K}_{1}) V_{1}^{\frac{1+a}{2}} - 2^{\frac{1+a}{2}} \frac{\lambda_{\min}(\mathbf{K}_{2})}{\lambda_{\max}(\mathbf{M})^{\frac{1+a}{2}}} (\frac{1}{2} \mathbf{s}^{\mathrm{T}} \mathbf{M} \mathbf{s})^{\frac{1+a}{2}} -\tilde{\boldsymbol{\tau}}_{w}^{*\mathrm{T}} \mathbf{K}_{4} (\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0}) + 0.2785 \mathbf{E}^{\mathrm{T}} \boldsymbol{\tau}_{w}^{*} \leq -m V_{2}^{\frac{1+a}{2}} - \tilde{\boldsymbol{\tau}}_{w}^{*\mathrm{T}} \mathbf{K}_{4} (\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0}) + 0.2785 \mathbf{E}^{\mathrm{T}} \boldsymbol{\tau}_{w}^{*}$$
(35)

Adopting  $n = \frac{1+a}{2}, V_3^n = (V_1 + \frac{1}{2}s^T Ms + \frac{1}{2}\tilde{\tau}_w^{*T}K_3^{-1}\tilde{\tau}_w^*)^n$ 

According to lemma 4, yields

$$V_3^n \le V_1^n + \left(\frac{1}{2}\boldsymbol{s}^{\mathrm{T}}\boldsymbol{M}\boldsymbol{s}\right)^n + \left(\frac{1}{2}\tilde{\boldsymbol{\tau}}_w^{*\mathrm{T}}\boldsymbol{K}_3^{-1}\tilde{\boldsymbol{\tau}}_w^{*}\right)^n \qquad (36)$$

According to (35) and (36), get

$$\dot{V}_{3} \leq -mV_{3}^{\frac{1+\alpha}{2}} + m(\frac{1}{2}\tilde{\boldsymbol{\tau}}_{w}^{*T}\boldsymbol{K}_{3}^{-1}\tilde{\boldsymbol{\tau}}_{w}^{*})^{n} - \tilde{\boldsymbol{\tau}}_{w}^{*T}\boldsymbol{K}_{4}(\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0}) + 0.2785\boldsymbol{E}^{T}\boldsymbol{\tau}_{w}^{*}$$
(37)

According to Lemma 5, obtain

$$m(\frac{1}{2}\tilde{\boldsymbol{\tau}}_{w}^{*^{\mathrm{T}}}\boldsymbol{K}_{3}^{-1}\tilde{\boldsymbol{\tau}}_{w}^{*})^{n} \leq \frac{1}{2}\tilde{\boldsymbol{\tau}}_{w}^{*^{\mathrm{T}}}\boldsymbol{K}_{3}^{-1}\tilde{\boldsymbol{\tau}}_{w}^{*} + (1-n)n^{n/1-n}m^{n/1-n}$$

$$\dot{V}_{3} \leq -mV_{3}^{\frac{1+a}{2}} + m(\frac{1}{2}\tilde{\boldsymbol{\tau}}_{w}^{*T}\boldsymbol{K}_{3}^{-1}\tilde{\boldsymbol{\tau}}_{w}^{*})^{n} \leq \frac{1}{2}\tilde{\boldsymbol{\tau}}_{w}^{*T}\boldsymbol{K}_{3}^{-1}\tilde{\boldsymbol{\tau}}_{w}^{*} + (1-n)n^{n/1-n}m^{\frac{1}{1-n}} -\frac{1}{2}\tilde{\boldsymbol{\tau}}_{w}^{*T}\boldsymbol{K}_{4}\tilde{\boldsymbol{\tau}}_{w}^{*} + \frac{1}{2}(\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0})^{T}\boldsymbol{K}_{4}(\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0}) + 0.2785\boldsymbol{E}^{T}\boldsymbol{\tau}_{w}^{*}$$

$$(38)$$

$$\dot{V}_{3} \leq -mV_{3}^{\frac{n-2}{2}} - \frac{1}{2}\tilde{\boldsymbol{\tau}}_{w}^{*T}(\boldsymbol{K}_{4} - \boldsymbol{K}_{3}^{-1})\tilde{\boldsymbol{\tau}}_{w}^{*} + (1-n)n^{\frac{n}{1-n}}m^{\frac{1}{1-n}} + \frac{1}{2}(\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0})^{T}\boldsymbol{K}_{4}(\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0}) + 0.2785\boldsymbol{E}^{T}\boldsymbol{\tau}_{w}^{*}$$

$$(39)$$

Choose  $K_4$  and  $K_3^{-1}$  appropriately to make sure  $K_4 - K_3^{-1} \ge 0$ , so

$$\dot{V}_3 \le -mV_3^n + \delta \tag{40}$$

$$\delta = (1-n)n^{n_{1-n}}m^{\frac{1}{n-n}} + \frac{1}{2}(\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0})^{\mathrm{T}}\boldsymbol{K}_{4}(\hat{\boldsymbol{\tau}}_{w}^{*} - \boldsymbol{\tau}_{w}^{0}) + 0.2785\boldsymbol{E}^{\mathrm{T}}\boldsymbol{\tau}_{w}^{*}$$

According to Lemma 2, the error signals of close-loop system converge to the following stability domain  $V_3 \leq \left(\frac{\delta}{m(1-\theta_0)}\right)^{\frac{2}{1+a}}$  in finite time  $T \leq \frac{V_3^{1-a}(V_3(\mathbf{x}_0))}{m\theta_0(1-a)}$ .

# 3. SIMULATION

In this section, one MSV model (Zhang *et al*, 2017) is selected as the plant and comparative experiments with the results in (Du *et al*, 2014) are illustrated in order to certify the performance of the proposed control algorithm.

The MSV model parameters are listed bellow

$$\begin{split} X_{\dot{u}} &= -0.7212 \times 10^{6}, Y_{\dot{v}} = -3.6921 \times 10^{6}, \\ Y_{\dot{r}} &= -1.0234 \times 10^{6}, I_{z} - N_{\dot{r}} = 3.7454 \times 10^{9}, \\ X_{u} &= 5.0242 \times 10^{4}, Y_{v} = 2.7229 \times 10^{5}, \\ Y_{r} &= -4.3933 \times 10^{6}, Y_{|v|v} = 1.7860 \times 10^{4}, \\ X_{|u|u} &= 1.0179 \times 10^{3}, Y_{|v|r} = -3.0068 \times 10^{5}, \\ N_{v} &= -4.3821 \times 10^{6}, N_{r} = 4.1894 \times 10^{8}, \\ N_{|v|v} &= -2.4684 \times 10^{5}, N_{|v|r} = 6.5759 \times 10^{6}, m = 4.591 \times 10^{6} \end{split}$$

The initial position and heading of the MSV model is  $\eta(0) = [20m, 20m, 20^{\circ}]^{T}$ , the desired reference attitude is  $\eta_r = [0m, 0m, 0^{\circ}]^{T}$ . The other initial state is  $\upsilon(0) = [0m/s, 0m/s, 0^{\circ}/s]^{T}$ .

As to the external disturbances, the sea wind and irregular wind-generated wave are involved in simulations.

The wind speed is  $V_{wind} = 4m / s$ , and wind direction is  $\psi_{wind} = 60 \deg deg$ .

The disturbance forces and moment, i.e., the wind forces and moment and 2<sup>nd</sup>-order wave forces and moment, are described as follows

$$\begin{split} \overline{X}_{wind} &= \frac{1}{2} \rho_{\alpha} A_{f} U_{R}^{2} C_{wx}(\alpha_{R}) \\ \overline{Y}_{wind} &= \frac{1}{2} \rho_{\alpha} A_{s} U_{R}^{2} C_{wy}(\alpha_{R}) \\ \overline{N}_{wind} &= \frac{1}{2} \rho_{\alpha} A_{s} L_{oa} U_{R}^{2} C_{wn}(\alpha_{R}) \\ X_{wave}^{D} &= \rho L_{oa} \cos \chi \sum_{i=1}^{m} C_{Xw}^{D} (2\pi\omega_{i}^{2} / g) S_{\zeta\zeta}(\omega_{i}) \Delta \omega \\ Y_{wave}^{D} &= \frac{1}{2} \rho L_{oa} \sin \chi \sum_{i=1}^{m} C_{Yw}^{D} (2\pi\omega_{i}^{2} / g) S_{\zeta\zeta}(\omega_{i}) \Delta \omega \\ N_{wave}^{D} &= \frac{1}{2} \rho L_{oa}^{2} \sin \chi \sum_{i=1}^{m} C_{Nw}^{D} (2\pi\omega_{i}^{2} / g) S_{\zeta\zeta}(\omega_{i}) \Delta \omega \\ \tau_{w} &= \left[ \tau_{wu}, \tau_{wv}, \tau_{wr} \right]^{T}, \tau_{wu} = \overline{X}_{wind} + X_{wave}^{D} \\ \tau_{wv} &= \overline{Y}_{wind} + Y_{wave}^{D}, \tau_{wr} = N_{wind} + N_{wave}^{D} \end{split}$$

Please refer to the literature (Jia & Yang, 1999) to obtain specified meanings of disturbance model parameters in detail.

The controller parameters are all given  $k_{11} = 0.105, k_{12} = 0.18, k_{13} = 0.038,$ 

$$\begin{aligned} & k_{11} = 0.025, k_{12} = 0.025, k_{13} = 0.005, \\ & k_{21} = 7.3 \times 10^5, k_{22} = 9.5 \times 10^5, k_{23} = 9.5 \times 10^7, \\ & k_{31} = 9.3 \times 10^5, k_{32} = 1.5 \times 10^6, k_{33} = 2.0 \times 10^9, \\ & k_{41} = 7.3 \times 10^{-8}, k_{42} = 8.3 \times 10^{-8}, k_{43} = 1.6 \times 10^{-10}, \\ & d_{11} = 0.09, d_{12} = 0.1, d_{13} = 0.01, d_{21} = 0.1, \\ & d_{22} = 0.08, d_{23} = 0.013, \varepsilon_1 = 0.2, \varepsilon_2 = 0.01, \varepsilon_3 = 0.01. \end{aligned}$$

Simulation results are shown in Figures. 2-6. In addition, Tables 1-2 give the results of performance comparisons.



Figure. 2. Trajectory of MSV in xy plane



Figure. 3. Variation curves of control forces  $\tau_u$ ,  $\tau_v$  and moment  $\tau_r$ 



Figure. 4. Variation curves of actual position x, y and heading  $\psi$ 

Table 1 Tracking error comparisons

Performance		Backstepp ing control in[2]	Proposed $\boldsymbol{\tau}_{\text{NTSMC}}$	Proposed $\tau_{\text{ANTSMC}}$
	$x_e(m)$	1.5671	1.3125	1.2567
MAE	$y_e(\mathbf{m})$	1.6086	1.3174	1.2664
	$\psi_e(\text{deg})$	2.2455	1.5424	1.4086

As showed in Figure. 2(1), the proposed adaptive control scheme  $\tau_{ANTSMC}$  is capable of overcoming influence of external disturbances, arrives smoothly along the straight-line motion trajectory ( $y = x - x_r + y_r$ ) in xy plane and maintains the preset terminal position successfully at last. It is obviously the shortest distance between original position and terminal position with the advantage of saving time and energy. Of course, we can also devise other curve motion trajectories. For example, suppose that there is a wreck shown in Figure. 2(2) on the connection line between initial position and terminal position, we consider to design a curve motion trajectory in xy plane in Figure. 2(2). MSV is obviously able to reach the destination along the curve trajectory, avoiding colliding with the wreck.

Next, Figures. 3-4 demonstrate control inputs and the convergence trajectory comparisons of system states, i.e., x, y and  $\psi$ , in detail under three control methodologies ( $\tau_{\text{ANTSMC}}, \tau_{\text{NTSMC}}$  and the backstepping control method) in detail. In Figure. 4, we can discern easily that the finite time control scheme  $\tau_{\text{NTSMC}}$  can achieve superior tracking

performance in terms of rapid response, in comparison with backstepping control scheme in (Du *et al*, 2014). It means that the control scheme  $\tau_{\text{NTSMC}}$  ensures faster convergence rate of the closed-loop system than backstepping control scheme in (Du *et al*, 2014).

Moreover, the qualification analysis is summarized in Table 1, where the mean absolute error (MAE) is used to evaluate stability performance of three control algorithms. We can observe obviously that steady-state errors of  $x_e$ ,  $y_e$  and  $\psi_e$  under the control scheme  $\tau_{\text{NTSMC}}$  reduce to 1.3125, 1.3174 and 1.5424, lowering by 16.2%, 18.1% and 31.3%, respectively in comparison with that under the backstepping control scheme in (Du *et al*, 2014). Results illustrate that the finite time control scheme  $\tau_{\text{NTSMC}}$  poses better stability performance.

What's more, as observed in Table 1, due to adaptive estimation of external disturbances, the control scheme  $\tau_{\text{ANTSMC}}$  further shrinks the steady-state errors, dropping to 1.2567, 1.2664 and 1.4086, respectively, compared to the controller  $\tau_{\text{NTSMC}}$ . That proves that the control scheme  $\tau_{\text{ANTSMC}}$  guarantees steady-state errors to converge to the closest neighbourhood of the equilibrium point among three control algorithms.

Based on the analysis above, it is clear that the proposed finite time control scheme  $\tau_{\text{ANTSMC}}$  facilities the fastest convergence rate and best stability performance among three control algorithms.

In order to verify robustness superiority of the finite time control algorithm, we directly rise the wind speed from 4m/s to 12m/s, and continue to carry out the other comparison simulations as shown in Figures. 5-6.



Figure. 5. Variation curves  $\tau_u$ ,  $\tau_v$  and  $\tau_r$  with  $V_{wind} = 4m / s$ 



Figure. 6. Variation curves of x, y and  $\psi$  with  $V_{wind} = 12m / s$ 

Table 2 Tracking error	comparisons	with V <sub>wind</sub>	= 12m/s
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Performance		Backstepping control in[2]	Proposed	Proposed
			$\pmb{ au}_{ ext{NTSMC}}$	$ au_{ ext{ANTSMC}}$
MAE	$x_e(\mathbf{m})$	2.1124	1.3230	1.2577
	$y_e(\mathbf{m})$	1.7613	1.3250	1.2674
	$\psi_e(\text{deg})$	3.8383	2.6909	1.7475

According to disturbance model in this section, the rise of wind speed from  $4m/s \tan 2m/s$  results in the strength of disturbance forces and moment. Therefore, we can see that the jitters of control inputs  $\tau_u$ ,  $\tau_v$  and  $\tau_r$  in Figure. 5 are stronger than that in Figure. 3 in order to sustain the stability of the close system.

As viewed in Figure. 6 and Table 2, we can see obviously that there is a relevant rise of steady-state errors, i.e.,  $x_e, y_e$  and  $\psi_e$ , in comparison with that in Figure. 4 and Table 1, because of the growth of the disturbance forces and moment (wind and wave). Especially, as we can see, the steady-state errors under the backstepping control scheme climb to 2.1124, 1.7613 and 3.8383, respectively, increasing by 34.8%, 9.5% and 70.9%, respectively. Besides, there is a minor increase of steady-state errors under the control scheme  $au_{_{
m NTSMC}}$  . The error values go up to 1.3230, 1.3250 and 2.6909, respectively. However, they are still less than that under backstepping control scheme in (Du et al, 2014). In consequence, it proves that robustness to disturbances of the finite time control algorithm is better than the backstepping control algorithm. Furthermore, duo to the addition of disturbance compensation in the control scheme  $\pmb{ au}_{\mathrm{ANTSMC}}$  , the steady-state errors under the control scheme  $\tau_{\text{ANTSMC}}$  stay the lowest, just increasing from 1.2567, 1.2664, and 1.4086 to 1.2577, 1.2674 and 1.7475, respectively.

Overall, all abovementioned simulation results and data comparison analysis testify that the finite time control algorithm finishes the advantages of faster convergence rate, better stability performance and more excellent robustness to disturbances, in comparison with the backstepping control algorithm.

Remark: the curve motion trajectory is implemented by the following method.

Design the following relationship function

$$h = \eta_{e3} + \eta_{e2} - 20\sin(\frac{\pi}{40}\eta_{e1})$$

The projection of *h* within *xy* plane is  $h_1 = \eta_{e2} - 20\sin(\frac{\pi}{40}\eta_{e1})$ . For the detail meanings of parameters, please refer to section 2.3.

### 4. CONCLUSION

In this brief, one NTSMC approach for DP control is originally derived based on a motion trajectory. The motion trajectory between initial point and the terminal point is established according to requirements of engineering application by designing a relationship function among position tracking errors  $\eta_{e1}, \eta_{e2}$  and heading error  $\eta_{e3}$ . Next, it demonstrates that the NTSMC approach not only raises the convergence rate of the closed-loop system, but also lower the state errors, compared with the conventional backstepping control approach. Furthermore, by employing adaptive technique and the NTSMC approach, we present an ANTSMC approach, which further strengthens robustness to disturbances, and decline steady-state errors to the lowest among three control algorithms. At last, finial simulation results certify remarkably outstanding performances of the ANTSMC approach in terms transient and steadystate responses.

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