Energy Saving Control Approach for Trajectory Tracking of Autonomous Mobile Robots

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Abstract: This research presents an adaptive energy-saving $H_2$ closed-form control approach to solve the nonlinear trajectory tracking problem of autonomous mobile robots (AMRs). The main contributions of this proposed design are as follows: closed-form approach, simple structure of the control law, easy implementation, and energy savings through trajectory tracking design of the controlled AMRs. It is difficult to mathematically obtained this adaptive $H_2$ closed-form solution of AMRs. Therefore, through a series of mathematical analyses of the trajectory tracking error dynamics of the controlled AMRs, the trajectory tracking problem of AMRs can be transformed directly into a solvable problem, and an adaptive nonlinear optimal controller, which has an extremely simple form and energy-saving properties, can be found. Finally, two test trajectories, namely circular and S-shaped reference trajectories, are adopted to verify the control performance of the proposed adaptive $H_2$ closed-form control approach with respect to an investigated $H_2$ closed-form control design.

Keywords: Energy saving; adaptive $H_2$ closed-form control; trajectory tracking

1 Introduction

In recent decades, comprehensive applications of autonomous mobile robots (AMRs) have attracted considerable attention. These AMRs with extended energy endurance, more precise motion ability, and effective control approaches have been applied in the transportation, security, and inspection domains. Thus, a precise motion controller for AMRs and energy saving are becoming increasingly important in the robotics application field, which have been discussed in many studies [1–6]. According to existing studies, it remains difficult to improve the control design when using an extremely simple structure and to realize energy saving in AMRs while accurately and effectively tracking the desired robot trajectory. For trajectory tracking control design, AMRs must be capable of converging the tracking errors of the real trajectory and the desired trajectory as close to zero as possible while considering the influence of modeling uncertainties. A survey of the literature revealed that many studies have focused on the trajectory tracking control of AMRs, for example, trajectory tracking control through backstepping [7–10],

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sliding mode control [11–14], feedback linearization [15–17], neural networks [18–22], fuzzy control [23–28], and the $H_2$ [29,30] approach. In practice, it is challenging to implement microchip operation and torque output with low energy consumption by using the aforementioned control algorithm methodologies or extremely complex theoretical structures.

For these reasons, an innovative nonlinear energy-saving control approach with a simple control structure that can provide high-performance trajectory tracking for AMRs is presented in this paper. To reduce computational costs and output low-energy torque, a novel energy-saving adaptive $H_2$ closed-form control approach for the trajectory tracking of AMRs is developed. Furthermore, this problem is directly solved using a nonlinear time-varying differential equation. Moreover, the proposed adaptive $H_2$ closed-form solution must satisfy an $H_2$ optimal performance index. In such a circumstance, it is extremely difficult to obtain the solution of a nonlinear time-varying differential equation. However, this solution can be expanded and inferred by selecting suitable state variable transformations and performing mathematical analyses of the dynamic equations of the trajectory tracking error. With such a solution, the adaptive $H_2$ closed-form control approach for the trajectory tracking of AMRs will have a direct implementation structure and provide energy saving.

The remainder of this paper is organized as follows. Section 2 describes the mathematical model of trajectory tracking error of AMRs. In Section 3, the adaptive $H_2$ closed-form controller design for AMR trajectory tracking is described. Section 4 illustrates the simulation results obtained for AMRs by using the proposed approach. Finally, our concluding remarks are given in Section 5.

2 Trajectory Tracking Error Mathematical Model

The trajectory tracking error mathematical model of AMRs is presented in this section. Based on the standard trajectory tracking error mathematical equation and the geometry relationship between the AMR and global coordinate systems, a controlled AMR with a nonlinear trajectory tracking error dynamic equation can be inferred as follows.

2.1 AMR Dynamic Equation

In Fig. 1, a schematic of the controlled AMR that has two driving wheels and one omnidirectional wheel is illustrated. The radius of the driving wheels is $r$. The instantaneous position of the controlled AMR in the system reference frame $\{O, X, Y\}$ is denoted by $p$. $(x_c,y_c)$ denotes the center $C$ of the controlled AMR in the system reference frame, and $\theta$ denotes the direction of the AMR frame $\{C, X_c, Y_c\}$. In addition, $d$ denotes the distance between the positions $p$ and $C$. According to the aforementioned descriptions, the universal coordinate frame of the AMR can be expressed as Eq. (1).

$$p = [x_c \quad y_c \quad \theta]^T$$  \hspace{1cm} (1)

Under the nonslipping condition, a standard AMR system usually moves along the orientation of the driving wheels’ axis. Hence, the kinematics of the controlled AMR with constraints can be expressed using the following equation [31]:

$$\dot{p} = \begin{bmatrix} \dot{x}_c \\ \dot{y}_c \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} \cos \theta & -d \sin \theta \\ \sin \theta & d \cos \theta \\ 0 & 1 \end{bmatrix} \begin{bmatrix} v_l \\ \omega \end{bmatrix}$$  \hspace{1cm} (2)

where $v_l$ and $\omega$ are the linear and angular velocities, respectively.
In this study, the dynamics of the controlled AMR are inferred using the Euler–Lagrange method, as expressed in Eq. (3).

\[
I(p)\ddot{p} + C(p, \dot{p})\dot{p} + G(p) = T(p)\tau
\]

where \(I(p)\) denotes a symmetric positive definite inertia matrix, and \(C(p, \dot{p})\) denotes the Coriolis and centripetal matrices. \(I(p)\) and \(C(p, \dot{p})\) satisfy the skew-symmetric property. \(G(p)\) denotes the gravitational vector, \(T(p)\) denotes the \(\tau\) transformation matrix, \(\tau\) denotes the torque vector of the control input, and \(\dot{p}\) and \(\ddot{p}\) denote the velocity and acceleration vectors, respectively. Physically, the gravitational vector \(G(p)\) is zero and can be ignored because the AMR moves in the horizontal plane.

Details of the AMR dynamics are as follows:

\[
I(p) = \begin{bmatrix}
m & 0 & md \sin \theta \\
0 & m & -md \cos \theta \\
md \sin \theta & -md \cos \theta & I
\end{bmatrix}
\]

\[
C(p, \dot{p}) = \begin{bmatrix}
0 & 0 & m \dot{\theta} \cos \theta \\
0 & 0 & m \dot{\theta} \sin \theta \\
0 & 0 & 0
\end{bmatrix}
\]

\[
T(p) = \frac{1}{r} \begin{bmatrix}
\cos \theta & \sin \theta \\
\sin \theta & \cos \theta \\
R & -R
\end{bmatrix}
\]

\[
\tau = \begin{bmatrix}
\tau_r \\
\tau_l
\end{bmatrix}
\]

where \(m = \bar{m} + \Delta m\), is the mass of the controlled AMR which is with a nominal term \(\bar{m}\) and a disturbed term \(\Delta m\). As to \(\tau_r\) and \(\tau_l\), they are the right and left wheel torques, respectively.

### 2.2 Trajectory Tracking Error Dynamics of Controlled AMR

Suppose \(p_t \in C^2\) is the desired tracking trajectory and is twice continuously differentiable, \(\dot{p}_t\) and \(\ddot{p}_t\) are the velocity and acceleration vectors, respectively. Accordingly, the trajectory tracking error of the controlled AMR can be expressed as follows.
\[ e = \begin{bmatrix} \dot{p} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} \dot{p} - \dot{p}_r \\ p - p_r \end{bmatrix} \tag{4} \]

where
\[ p_r = [x \ y \ \theta_d]^T \]

According to Eqs. (3) and (5), the trajectory tracking error dynamics can be described as follows:
\[ \dot{e} = \begin{bmatrix} -I^{-1}(p)C(p, \dot{p}) & 0_{3 \times 3} \\ I_{3 \times 3} \\ 0_{3 \times 3} \end{bmatrix} e + \begin{bmatrix} -\dot{p}_r - I^{-1}(p)C(p, \dot{p})\dot{p}_r \\ 0_{3 \times 3} \\ 0_{3 \times 3} \end{bmatrix} + \begin{bmatrix} I^{-1}(p)T(p)\tau \\ 0_{3 \times 3} \end{bmatrix} \tag{5} \]

Mathematically, it is difficult to use Eq. (5) to solve the trajectory tracking problem of controlled AMRs because of the structure of the error dynamics between the controlled AMR and the desired trajectory. For simplifying the design complexity of this proposed closed-form control approach, a proportional-derivative (PD)-type transformation \( s(t) \) is defined as follows to reduce the derivative complexity of the proposed control design.
\[ s(t) = \lambda_1 \dot{p} + \lambda_2 \dot{\theta} \tag{6} \]

where \( \lambda_1 \) and \( \lambda_2 \) are positive constants, and from Eq. (6), we have the following:
\[ \dot{s}(t) = -I^{-1}(p)C(p, \dot{p})s(t) + \lambda_2 I^{-1}(p)[-\Theta(e, t)\Delta(e, t) + T(p)\tau] \tag{7} \]

where
\[ \Theta(e, t)\Delta(e, t) = \Theta(p, \dot{p}, \dot{p}_r - \frac{\lambda_1}{\lambda_2} \dot{p}, \dot{p}_r - \frac{\lambda_1}{\lambda_2} \dot{\theta}_d)\Delta(e, t) = I(p)(\dot{p}_r - \frac{\lambda_1}{\lambda_2} \dot{\theta}) + C(p, \dot{p})(\dot{p}_r - \frac{\lambda_1}{\lambda_2} \dot{\theta}) \]

From Eq. (7), the dynamic Eq. (5) of trajectory tracking error can be revised as follows:
\[ \dot{e} = S^{-1} \begin{bmatrix} \dot{s}(t) \\ \dot{\theta}(t) \end{bmatrix} = W(e, t)e(t) + \lambda_2 Q(e, t)[-\Theta(e, t)\Delta(e, t) + T(q)\tau] \tag{8} \]

where
\[ W(e, t) = S^{-1} \begin{bmatrix} -I^{-1}(p)C(p, \dot{p}) \\ \frac{1}{\lambda_2} I_{3 \times 3} \\ -\frac{\lambda_1}{\lambda_2} I_{3 \times 3} \end{bmatrix} S \]

\[ Q(e, t) = S^{-1}A I^{-1}(p) \text{ with } A = \begin{bmatrix} I_{3 \times 3} \\ 0_{3 \times 3} \end{bmatrix} \]

and \( S \) is the state-space transformation matrix as follows:
\[ S = \begin{bmatrix} \lambda_1 I_{3 \times 3} & \lambda_2 I_{3 \times 3} \\ I_{3 \times 3} & 0_{3 \times 3} \end{bmatrix} \tag{9} \]

If \( T(q)\tau \) is selected by Eq. (10)
\[ T(q)\tau = \Theta(e, t)\Delta(e, t) + \frac{1}{\lambda_2} u \tag{10} \]
3 Adaptive $H_2$ Closed-Form Control Approach Design

3.1 Trajectory Tracking Problem of Adaptive $H_2$ Closed-Form

An analytic adaptive $H_2$ control law for the AMR is deduced from the following equations. To inspect Eq. (11), given the weighting matrices $X_{a2}$ and $Z_{a2}$, the adaptive $H_2$ controller design of trajectory tracking with the $H_2$ performance property of AMRs can be solved if there exists a closed-form solution $u_{a2}^*$ and an adaptive rule $\Delta$ that satisfy the following performance index: [32]

$$J(u_{a2}^*) = \min_{u_{a2}} \left[ e^T(t)X_{a2}e(t) + \Delta^T(t)Y_{a2}\Delta(t) + \int_0^t \left[ e^T(t)X_{a2}e(t) + u_{a2}^T(t)Z_{a2}u_{a2}(t) \right] dt \right]$$

The aforementioned performance index can be achieved for all $t_f \in [0, \infty]$, and the positive-definite weighted matrices are $X_{a2} = X_{a2}^T$ and $Y_{a2} = Y_{a2}^T > 0$.

3.2 Adaptive $H_2$ Closed-Form Control Design for AMR

In this section, we solve the AMR trajectory tracking control problem described in Section 2. To this end, we present a novel energy-saving adaptive $H_2$ closed-form control approach for trajectory tracking of the AMR based on the following nonlinear adaptive $H_2$ closed-form control theorem.

Theorem: To obtain the dynamic equation of the trajectory tracking error of the AMR control system described in Eq. (8), the adaptive $H_2$ closed-form control law $f^*(e, t)$ is chosen using the following equation.

$$f^*(e, t) = \Theta(e, t)\Delta(e, t) + \frac{1}{\lambda_2} u_{a2}^*(e, t)$$

where

$$u_{a2}^*(e, t) = -Z_{a2}^{-1}Q^T(e, t)L(e, t)e(t)$$

$$\dot{\Delta}(e, t) = -\lambda_2 Y_{a2}^{-1}\Theta^T(e, t)Q^T(e, t)L(e, t)e(t)$$

and $L(e, t) = L^T(e, t) \geq 0$

If $L(e, t)$ satisfies the following nonlinear $H_2$ time-varying differential Eq. (14), it is equivalent to solving the nonlinear $H_2$ trajectory tracking problem of the AMR in Eq. (12).

$$\dot{L}(e, t) + L(e, t)W(e, t) + W^T(e, t)L(e, t) + X_{a2} - L(e, t)Q(e, t)Z_{a2}^{-1}Q^T(e, t)L(e, t) = 0$$

(14)

It is difficult to determine the adaptive $H_2$ closed-form solution and solve the nonlinear time-varying differential equations in Eqs. (5) and (14). Therefore, this result is a great achievement for the trajectory tracking of AMRs because the adaptive $H_2$ closed-form solution can be directly derived from Eq. (14).
3.3 Adaptive $H_2$ Closed-Form Solution of Nonlinear Time-Varying Differential Equation

In general, it is difficult to solve $L(e, t)$ in time-varying differential Eq. (14). However, Eq. (14) can be further refined to an algebraic Riccati-like equation with the selected matrix $L(e, t)$ by applying the AMR system property.

Because state-space transformation matrix Eq. (9) has been applied to the control design, the solution $L(e, t)$ can be expressed in more explicit forms as follows:

$$L(e, t) = S^T \begin{bmatrix} I(e, t) & 0_{3 \times 3} \\ 0_{3 \times 3} & P \end{bmatrix} S$$  \hspace{1cm} (15)$$

where $P$ is some designed positive distinct symmetric constant matrix implied under some conditions. Furthermore, the solutions of the constant matrices $S$ and $P$ can be obtained from a pair of algebraic Riccati-like equation.

To investigate the second and third terms on the left-hand side of time-varying differential Eq. (14), the following equations can be derived using the dynamic equation of trajectory tracking error in Eq. (5) and the selected relationship in Eq. (15):

$$\dot{L}(e, t) + L(e, t)I(e, t) + I^T(e, t)L(e, t) = \begin{bmatrix} 0_{3 \times 3} & P \\ P & 0_{3 \times 3} \end{bmatrix}$$  \hspace{1cm} (16)$$

$$Q^T(e, t)L(e, t) = A^T S$$  \hspace{1cm} (17)$$

By using the results of Eqs. (16) and (17), time-varying differential Eq. (14) can be rewritten as the following algebraic equation.

$$\begin{bmatrix} 0_{3 \times 3} & P \\ P & 0_{3 \times 3} \end{bmatrix} + X_{a2} - S^T A Z_{a2}^{-1} A^T S = 0$$  \hspace{1cm} (18)$$

In addition, the optimal control law and adaptive law can be expressed as Eqs. (19) and (20), respectively.

$$u_{a2}^*(e, t) = -Z_{a2}^{-1} s(t)$$  \hspace{1cm} (19)$$

$$\dot{\Delta}(e, t) = -\dot{\lambda}_2 Y_{a2}^{-1} \Theta^T(e, t) s(t)$$  \hspace{1cm} (20)$$

where $s(t)$ in Eq. (6). Eqs. (18)–(20) are based on known and applicable matrices or variables. According to the above analysis, the matrix $L(e, t)$ in Eq. (15) is the solution of the time-varying differential Eq. (14) if the matrices $S$ and $P$ satisfy algebraic Eq. (18). Furthermore, the positive definite symmetry of $P$ must be satisfied. For obtaining the solution, the weighting matrices $X_{a2}$, $Y_{a2}$, and $Z_{a2}$ are assumed as in Eqs. (21)–(23), respectively.

$$X_{a2} = \begin{bmatrix} \sigma_{11}^2 & 0_{12} \\ 0_{21}^T & \sigma_{22}^2 I_{3 \times 3} \end{bmatrix}$$  \hspace{1cm} (21)$$

$$Y_{a2} = \beta^2 I_{3 \times 3}$$  \hspace{1cm} (22)$$

$$Z_{a2} = \alpha^2 I_{3 \times 3}$$  \hspace{1cm} (23)$$

where $\alpha$ and $\beta$ denote some positive scale.
Using the definitions of A and S in Eqs. (8) and (9) and Eqs. (21)–(23), algebraic Eq. (18) can be solved using the following equalities:

\[ \lambda_1 = o_{11} \quad (24) \]
\[ \lambda_2 = o_{22} \quad (25) \]
\[ P = o_{11}o_{22}I_{3\times3} - O_{12} \quad (26) \]

with \( O_{12} < o_{11}o_{22}I_{3\times3} \)

Then, the following adaptive \( H_2 \) control law can be used to solve the trajectory tracking problem of adaptive \( H_2 \) closed-form control.

\[ f^*(e, t) = \Theta(e, t)\dot{\Delta}(e, t) + \frac{1}{\lambda_2} u^*_{o2}(e, t) \quad (27) \]

where

\[ u^*_{o2}(e, t) = -\frac{1}{\lambda_2} (o_{11}\dot{p} + o_{22}\dot{p}) \]
\[ \dot{\Delta}(e, t) = -\frac{o_{22}}{\beta} \Theta^T(e, t)(o_{11}\dot{p} + o_{22}\dot{p}) \]

4 Simulation Results

In this section, a verification scenario with the \( H_2 \) closed-form and adaptive \( H_2 \) closed-form control approach for trajectory tracking of a circle and an S shape is presented using the MATLAB software application. According to the aforementioned simulation results, this adaptive \( H_2 \) closed-form control approach will be certified the performances of trajectory tracking and energy saving of the AMR are more excellent than \( H_2 \) closed-form control approach.

4.1 Configuration of Simulation Environment

To construct the simulation environment, the following parameters of the practical AMR are employed: \( R = 17.8 \text{ (cm)} \), \( r = 6.5 \text{ (cm)} \), and \( d = 14 \text{ (cm)} \), and the mass \( m \) is with a nominal value \( m = 10 \text{ (kg)} \) and a disturbed value \( \Delta m \) of 20% variation of \( m \). This is a real case involving the application of practical hardware in this simulation scenario, and the desired circular and S-shaped trajectories for the verification scenario are generated using Eqs. (28) and (29), respectively. In addition, the original states of the desired trajectory are \( x_0 = 0 \text{ (meter)} \), \( y_0 = 0 \text{ (meter)} \), and \( \omega_d = 3^\circ/\text{s} \) in the circular and S-shaped simulation scenarios.

\[ \begin{cases} x = x_0 + r_d \cos(\theta_d) \\ y = y_0 + r_d \sin(\theta_d) \end{cases} \quad (28) \]
\[ \begin{cases} x = x_0 + r_d \cos(2\theta_d) \\ y = y_0 + 2r_d \sin(\theta_d) \end{cases} \quad (29) \]

where \( r_d \) denotes the radius of the desired trajectory, and \( \theta_d = \int_0^t \omega_d dt \) denotes the desired rotation angle with the desired constant angular velocity \( \omega_d \), respectively.
4.2 Simulation Results

Figs. 2–8 show the simulation results of the AMR driven by the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches for tracking a desired circular trajectory with a radius of 3.8 m. The verification results of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches for tracking this circular trajectory are displayed in Figs. 2 and 3. These results indicate that the trajectory tracking performance of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches for the desired circular trajectory is adequate. The circular trajectory tracking errors along the x-y axis and angle-to-convergence rates obtained using the $H_2$ closed-form and adaptive $H_2$ closed-form control approach are displayed in Figs. 4–6. The torque performance results of trajectory tracking have outstanding convergence rates that approach zero rather quickly, as indicated in Figs. 7 and 8. Especially, the adaptive $H_2$ closed-form control approach can track the desired circular trajectory of the AMR more rapidly and yield superior trajectory tracking performance and energy savings than the $H_2$ closed-form control approach.

Figure 2: Verification result of $H_2$ closed-form control approach with a circular trajectory from $x_c = 0$ m, $y_c = 3.8$ m
Figure 3: Verification result of the adaptive $H_2$ closed-form control approach with a circular trajectory from $x_c = 0$ m, $y_c = 3.8$ m

Figure 4: Tracking error results of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches for the x-axis with a circular trajectory from $x_c = 0$ m, $y_c = 3.8$ m
**Figure 5:** Tracking error results of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches for the y-axis with a circular trajectory from $x_c = 0$ m, $y_c = 3.8$ m

**Figure 6:** Tracking error results of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches for angle with a circular trajectory from $x_c = 0$ m, $y_c = 3.8$ m

**Figure 7:** Verification results of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches for left torque with a circular trajectory from $x_c = 0$ m, $y_c = 3.8$ m
In the second simulation scenario, the verification results of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches for an S-shaped trajectory with a radius of 3.8 m are displayed in Figs. 9 and 10. The tracking errors of the adaptive $H_2$ closed-form control approach for the desired S-shaped trajectory illustrate an outstanding performance, as displayed in Figs. 11–13. Moreover, the torque performance is notable, as illustrated in Figs. 14 and 15. Finally, the proposed adaptive $H_2$ closed-form control approach can track the desired S-shaped trajectory faster than the $H_2$ closed-form control approach, and the energy-saving effect of the adaptive $H_2$ closed-form control approach is superior to that of the $H_2$ closed-form control design in this simulation scenario.

**Figure 8:** Verification results of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches for right torque with a circular trajectory from $x_c = 0$ m, $y_c = 3.8$ m

**Figure 9:** Verification result of $H_2$ closed-form control approach for S-shaped trajectory from $x_c = 0$ m, $y_c = 3.8$ m
**Figure 10:** Verification result of adaptive $H_2$ closed-form control approach for S-shaped trajectory from $x_c = 0$ m, $y_c = 3.8$ m

**Figure 11:** Tracking error of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches along $x$-axis for S-shaped trajectory from $x_c = 0$ m, $y_c = 3.8$ m
**Figure 12:** Tracking error of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches along y-axis for S-shaped trajectory from $x_c = 0$ m, $y_c = 3.8$ m

**Figure 13:** Tracking error of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches in terms of angle for S-shaped trajectory from $x_c = 0$ m, $y_c = 3.8$ m

**Figure 14:** Verification result of the $H_2$ closed-form and adaptive $H_2$ closed-form control approaches in terms of left torque for S-shaped trajectory from $x_c = 0$ m, $y_c = 3.8$ m
5 Conclusions

Suboptimal trajectory tracking designs have been studied for autonomous mobile wheel robots in the past decades, and most of them have achieved acceptable control performance. However, they have disadvantages such as their extremely complex control structures, such as the sliding mode and backstepping control methods. For simultaneously achieving satisfactory tracking performance and a simple control structure, an analytical adaptive nonlinear control scheme was developed to track the trajectory of autonomous mobile wheel robots in this study. The proposed adaptive control design consists of an adaptive cancellation term that is used to cancel the nonlinear component of tracking errors and an optimal control term to minimize the power consumption when tracking the desired trajectories. Thus, the proposed control method has an impressive property; that is, without knowing the system parameters of autonomous mobile wheel robots, the desired trajectory tracking performance can be maintained by exploiting the adaptive learning ability of the proposed method. The simulation results indicate that the proposed adaptive nonlinear control method delivers promising trajectory tracking performance for WMRs because the tracking errors quickly converge to zero when a large amount of modeling uncertainties appear. Therefore, the proposed method has the advantages of being able to execute tasks such as the uploading and downloading of goods and regular patrolling.

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