




IEEE Robust Torque Distribution Control with Energy Optimization for Four-Wheel Electric Vehicles

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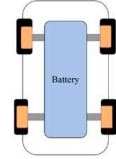




1. Introduction

automotive industry trends

CASE → Control system of 4-IWM EV

- Body motion control
- Energy consumption



Benchmark Problems

Design the controller which has 5 control inputs (steering angle δ , each torque τ_{ij})

Task1: Speed tracking
Task2: Double lane change

Some of the previous researches

- MPC for energy management [1]
- Robustness against disturbances by SMC [2]

We proposed the torque distribution control Strategy which consider both of them

- Robust vehicle motion control
- Energy Optimization

2. Control System Design

The steering angle δ is governed by default controller. This research focuses on **designing a controller that optimizes the torque distribution.**

The control architecture consists of two layers:

- Robust vehicle motion control
- Torque allocation

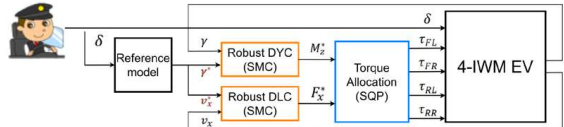


Fig.1 control architecture

2.1 Robust vehicle motion control

It aims **robustness against disturbances and model uncertainties** through DYC and DLC by using SMC.

- DYC** (Direct Yaw Moment Control)
It contributes a yaw moment M_z^* to track the desired yaw rate γ^*
- DLC** (Direct Longitudinal Force Control)
It contributes a longitudinal force F_x^* to track the desired speed v_x^* .

Designing SMC

Hyperplanes: $s_1 = \gamma - \gamma^*, \quad s_2 = v_x - v_x^*$

The control inputs:
$$\begin{cases} M_z = -(f_1 + \widehat{D}_1 + k_{11} \cdot \text{sign}(s_1) + k_{12} \cdot s_1) \\ F_x = -(f_2 + \widehat{D}_2 + k_{21} \cdot \text{sign}(s_2) + k_{22} \cdot s_2) \end{cases}$$

2.2 Torque allocation

It aims to **minimize power consumption**, while satisfying the desired yaw moment M_z^* and longitudinal force F_x^* . SQP is employed to optimize this torque distribution.

$$\min_{\tau_{ij}} \sum_{ij=1}^4 V_b I_{ij}(\tau_{ij})$$

s.t.
$$\begin{cases} I_{ij} = -Q_b \cdot S \cdot C_{ij} \\ -V_b + \sqrt{V_b^2 - 4RP_{ij}} \\ S \cdot C_{ij} = \frac{2RP_{ij}}{2RQ_b} \\ P_{ij} = \tau_{ij} \omega_{ij} \\ F_x^* - F_x(\tau_{ij}) = 0 \\ M_z^* - M_z(\tau_{ij}) = 0 \\ -75 \leq \tau_{ij} \leq 75 \end{cases}$$

3. Results

3.1 Task1: Speed tracking

Fig.2,3 shows the control inputs obtained from the controller. Simulation are presented in Table 1.

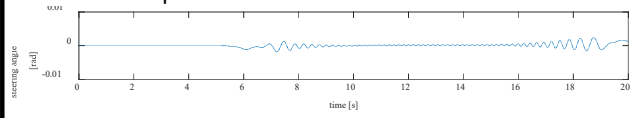


Fig.2 steering angle of Task1

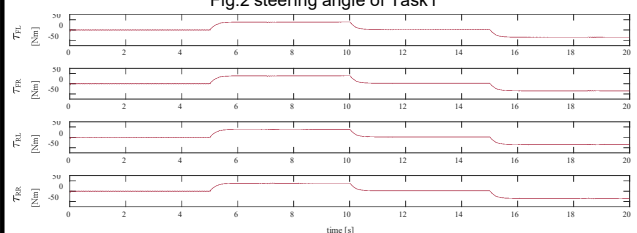


Fig.3 each torque of Task1

- Energy consumption $J_e = \int_0^{t_e} E(t) dt$ $a_z \dots$ Vertical acceleration
- Speed tracking $J_{v1} = \int_0^{t_e} \{v_x(t) - v_x^*(t)\} dt$ $\phi \dots$ Roll angle
- Route following $J_{xy} = \int_0^{t_e} \{[x(t) - x^*(t)]^2 + [y(t) - y^*(t)]^2\} dt$ $\psi \dots$ Pitch angle

J_e	J_{v1}	$\max a_z , (a_{zt})$	$\max \phi , (\phi_t)$	$\max \psi , (\psi_t)$
191.3	13.37	0.7584, (1.4)	0.0078, (0.004)	0.0023, (0.0015)

Table1 simulation results of Task1

3.2 Task2: Double lane change

Fig.4,5 shows the control inputs obtained from the controller. Simulation are presented in Table 2.

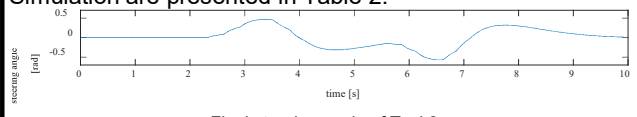


Fig.4 steering angle of Task2

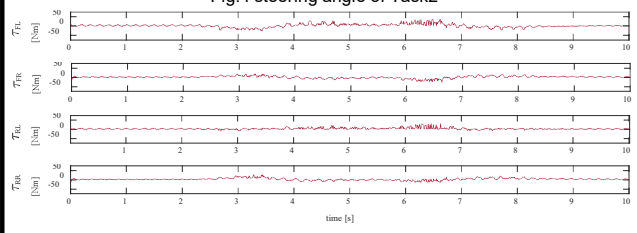


Fig.5 each torque of Task2

J_e	J_{v2}	$\max a_z , (a_{zt})$	$\max \phi , (\phi_t)$	$\max \psi , (\psi_t)$
5.24×10^4	0.752	2.601, (1.4)	0.025, (0.7)	0.0039, (0.14)

Table2 simulation results of Task2

4. Conclusion

- Our study presents the **torque distribution strategy to optimize both robust motion control and energy consumption.**
- The two-layer control strategy demonstrates remarkable performance in terms of **robustness and energy efficiency.**
- However, this proposed method was unable to satisfy the vehicle constraint of **vertical acceleration a_z .**

[1] S. H. Kim and K.-K. K. Kim, 'Model Predictive Control for Energy-Efficient Yaw-Stabilizing Torque Vectoring in Electric Vehicles With Four In-Wheel Motors', IEEE, 2023
[2] S. Ding, L. Liu, and W. X. Zheng, 'Sliding Mode Direct Yaw-Moment Control Design for In-Wheel Electric Vehicles', IEEE, 2017.