

Stability Analysis of Observer-PD Controller

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I. PREFACE

This paper is a supplemented material serving for stability analysis of the combination of PD controller and rain speed observer (RSO), which is a innovatively designed observer introduced in manuscript submitted to IEEE Transactions on Industrial Electronics with paper ID 24-2548. It can also provide reference for the stability analysis of other similar studies.

II. PRELIMINARIES

$\lambda_M(*)$ and $\lambda_m(*)$ represent the maximum and minimum eigenvalues of a matrix, respectively. $\hat{*}$, $\dot{*}$ and $\tilde{*}$ denote the estimation, the first-order time derivative and estimation error of $*$, respectively.

By combining the mathematical expressions of the baseline controller and model dynamics, the relationships between states in the translational loop can be formulated as a state-space equation in the following form, facilitating the evaluation of stability using theoretical analysis tools.

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$$\underbrace{\begin{pmatrix} \dot{e}_p \\ \dot{e}_v \end{pmatrix}}_{\dot{e}} = \underbrace{\begin{pmatrix} \mathbf{0}_{3 \times 3} & \mathbf{I}_{3 \times 3} \\ \frac{-\mathbf{K}_p}{m} & \frac{-\mathbf{K}_v}{m} \end{pmatrix}}_A \underbrace{\begin{pmatrix} e_p \\ e_v \end{pmatrix}}_e + \underbrace{\begin{pmatrix} \mathbf{0}_{3 \times 3} & \mathbf{0}_{3 \times 3} \\ \mathbf{0}_{3 \times 3} & \frac{\mathbf{I}_{3 \times 3}}{m} \end{pmatrix}}_B \underbrace{\begin{pmatrix} \mathbf{0}_{3 \times 1} \\ \tilde{\mathbf{F}}_r^E \end{pmatrix}}_{\tilde{d}}, \quad (1)$$

where $e = [e_p^T, e_v^T]^T$, e_p and e_v represent the tracking errors of position and velocity, respectively, $\tilde{\mathbf{F}}_r^E = \hat{\mathbf{F}}_r^E - \mathbf{F}_r^E = -\vartheta \tilde{v}_r^E$ represents the estimation error of the rain disturbance \mathbf{F}_r^E from RSO and the expression is derived from the designed model of disturbance.

III. STABILITY ANALYSIS

It can be checked that A in (1) has negative definite structure if \mathbf{K}_p and \mathbf{K}_v are positive definite. As a consequence, there must exist a positive definite symmetric matrix M that meets the equation $A^T M + M A = -I$. Subsequently, a Lyapunov function is designed as follows,

$$V = e^T M e + \frac{1}{2} (\tilde{v}_r^E)^T \tilde{v}_r^E. \quad (2)$$

Differentiate V , it can be implied that

$$\begin{aligned} \dot{V} &= \dot{e}^T M e + e^T M \dot{e} + (\tilde{v}_r^E)^T \dot{\tilde{v}}_r^E \\ &= e^T A^T M e + \tilde{d}^T B^T M e + e^T M A e \\ &\quad + e^T M B \tilde{d} + (\tilde{v}_r^E)^T \dot{\tilde{v}}_r^E \\ &= e^T (A^T M + M A) e + 2e^T M B \tilde{d} + (\tilde{v}_r^E)^T \dot{\tilde{v}}_r^E \\ &= -e^T e + 2e^T M B \tilde{d} + (\tilde{v}_r^E)^T \dot{\tilde{v}}_r^E. \end{aligned} \quad (3)$$

Furthermore, by resorting to Young's inequality [1], it can be verified that

$$-e^T e \leq -\frac{e^T M e}{\lambda_M(M)}, \quad (4a)$$

$$-(\tilde{v}_r^E)^T \dot{\tilde{v}}_r^E \leq \frac{1}{4} (\tilde{v}_r^E)^T \tilde{v}_r^E + \delta, \quad (4b)$$

$$e^T M B \tilde{d} \leq \frac{1}{2\varepsilon} \lambda_M^2(MB) e^T e + \frac{\varepsilon}{2} \tilde{d}^T \tilde{d}, \quad (4c)$$

where ε is an arbitrary positive constant, and $\delta \leq (\dot{\tilde{v}}_r^E)^T \tilde{v}_r^E$ is a constant value with an upper bound concerned with $\dot{\tilde{v}}_r^E$.

From the stability analysis of the RSO, i.e., $\dot{\tilde{\mathbf{v}}}_r^E = -\mathbf{L}\boldsymbol{\vartheta}\tilde{\mathbf{v}}_r^E - \dot{\mathbf{v}}_r^E$, where $\mathbf{L}\boldsymbol{\vartheta}$ is a diagonal matrix, it can be obtained that

$$\begin{aligned} (\tilde{\mathbf{v}}_r^E)^\top \dot{\tilde{\mathbf{v}}}_r^E &= -\mathbf{L}\boldsymbol{\vartheta}(\tilde{\mathbf{v}}_r^E)^\top \tilde{\mathbf{v}}_r^E - (\tilde{\mathbf{v}}_r^E)^\top \dot{\mathbf{v}}_r^E \\ &\leq -\lambda_m(\mathbf{L}\boldsymbol{\vartheta})(\tilde{\mathbf{v}}_r^E)^\top \tilde{\mathbf{v}}_r^E + \frac{1}{4}(\tilde{\mathbf{v}}_r^E)^\top \tilde{\mathbf{v}}_r^E + \delta, \end{aligned} \quad (5a)$$

$$\tilde{\mathbf{d}}^\top \tilde{\mathbf{d}} = (\tilde{\mathbf{v}}_r^E)^\top \boldsymbol{\vartheta}^\top \boldsymbol{\vartheta} \tilde{\mathbf{v}}_r^E \leq \lambda_M(\boldsymbol{\vartheta}^\top \boldsymbol{\vartheta})(\tilde{\mathbf{v}}_r^E)^\top \tilde{\mathbf{v}}_r^E. \quad (5b)$$

Combining (3) - (5), (3) can be adjusted that

$$\begin{aligned} \dot{V} &\leq -\mathbf{e}^\top \mathbf{e} + \frac{1}{\varepsilon} \lambda_M^2(\mathbf{M}\mathbf{B}) \mathbf{e}^\top \mathbf{e} + \varepsilon \tilde{\mathbf{d}}^\top \tilde{\mathbf{d}} + (\tilde{\mathbf{v}}_r^E)^\top \dot{\tilde{\mathbf{v}}}_r^E \\ &\leq -\frac{\varepsilon - \lambda_M^2(\mathbf{M}\mathbf{B})}{\varepsilon \lambda_M(\mathbf{M})} \mathbf{e}^\top \mathbf{M} \mathbf{e} + \varepsilon \lambda_M(\boldsymbol{\vartheta}^\top \boldsymbol{\vartheta})(\tilde{\mathbf{v}}_r^E)^\top \tilde{\mathbf{v}}_r^E \\ &\quad - \lambda_m(\mathbf{L}\boldsymbol{\vartheta})(\tilde{\mathbf{v}}_r^E)^\top \tilde{\mathbf{v}}_r^E + \frac{1}{4}(\tilde{\mathbf{v}}_r^E)^\top \tilde{\mathbf{v}}_r^E + \delta \\ &\leq -\sigma_1 \mathbf{e}^\top \mathbf{M} \mathbf{e} - \sigma_2 (\tilde{\mathbf{v}}_r^E)^\top \tilde{\mathbf{v}}_r^E + \delta, \end{aligned} \quad (6)$$

where $\sigma_1 = \frac{\varepsilon - \lambda_M^2(\mathbf{M}\mathbf{B})}{\varepsilon \lambda_M(\mathbf{M})}$ and $\sigma_2 = \lambda_m(\mathbf{L}\boldsymbol{\vartheta}) - \varepsilon \lambda_M(\boldsymbol{\vartheta}^\top \boldsymbol{\vartheta}) - \frac{1}{4}$. σ_1 can be ensured to be positive with $\varepsilon - \lambda_M^2(\mathbf{M}\mathbf{B}) > 0$. σ_2 can be guaranteed to be positive by adjusting \mathbf{L} properly. We can obtain that $\dot{V} \leq -\sigma_1 V + \delta$ where $\gamma = \min\{\sigma_1, 2\sigma_2\}$, which implies that

$$0 \leq V(t) \leq e^{-\gamma t} \left[V(0) - \frac{\delta}{\gamma} \right] + \frac{\delta}{\gamma}. \quad (7)$$

IV. CONCLUSION

It can be concluded from (7) that all signals of the translational loop are globally uniformly bounded. According to (5b) and the boundedness of $V(t)$, we can find that the estimation error of RSO can converge to a small residual set related to δ .

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