



Model Predictive Control

7. Reference Tracking and Disturbance Rejection

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Reference Tracking based on Target Calculation

- Discrete-Time Linear Time-Invariant (LTI) System

$$\mathbf{x}(k+1) = \mathbf{A}\mathbf{x}(k) + \mathbf{B}\mathbf{u}(k) \quad \text{state equation} \quad (7.1)$$

$$\mathbf{y}(k) = \mathbf{C}\mathbf{x}(k) \quad \text{measured output equation} \quad (7.2)$$

$$\mathbf{y}_r(k) = \mathbf{C}_r\mathbf{x}(k) \quad \text{controlled output equation} \quad (7.3)$$

- Symbols

$\mathbf{x}(k) \in \mathbb{R}^n$ state vector

$\mathbf{u}(k) \in \mathbb{U} \subseteq \mathbb{R}^m$ input vector

$\mathbf{y}(k) \in \mathbb{Y} \subseteq \mathbb{R}^p$ measured output vector

$\mathbf{y}_r(k) \in \mathbb{R}^{p_r}$ controlled output vector

$\mathbf{A} \in \mathbb{R}^{n \times n}$ system matrix

$\mathbf{B} \in \mathbb{R}^{n \times m}$ input matrix

$\mathbf{C} \in \mathbb{R}^{p \times n}$ measured output matrix

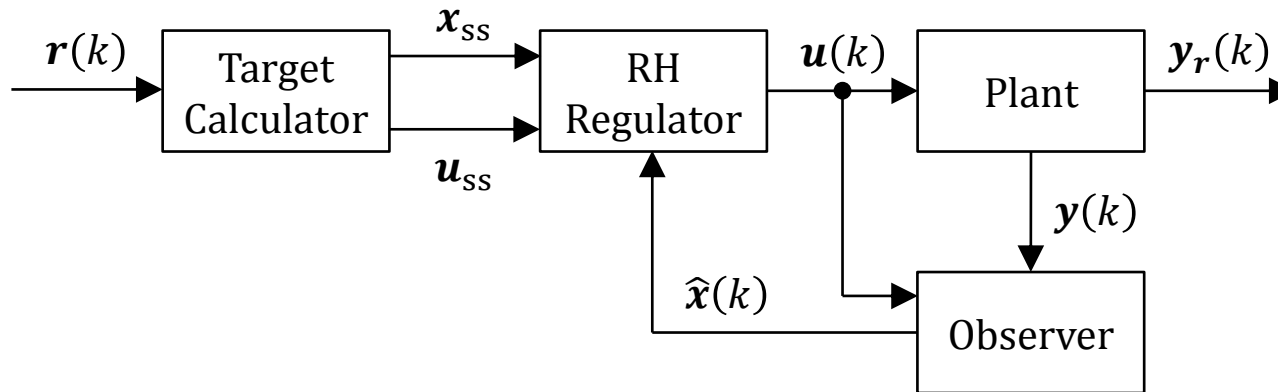
$\mathbf{C}_r \in \mathbb{R}^{p_r \times n}$ controlled output matrix

- Remarks

- The measured output $\mathbf{y}(k)$ is used for the observer
- The controlled output $\mathbf{y}_r(k)$ is considered for reference tracking

Reference Tracking based on Target Calculation

- Structure



- Objective

- Control the discrete-time LTI system such that $y_r(k) \rightarrow r(k)$ if $r(k) \rightarrow \text{const.}$ as $k \rightarrow \infty$

- Approach

- The **target calculator** computes the **target state x_{ss}** and the **target input u_{ss}**
- The **RH regulator** controls the discrete-time LTI system to the **target pair (x_{ss}, u_{ss})**

Reference Tracking based on Target Calculation

- Approach

- Consider that the discrete-time LTI system (7.1)/(7.3) is in the steady **target state** \mathbf{x}_{ss} , i.e.

$$\mathbf{x}_{ss} = \mathbf{A}\mathbf{x}_{ss} + \mathbf{B}\mathbf{u}_{ss} \Leftrightarrow (\mathbf{I}_{n \times n} - \mathbf{A})\mathbf{x}_{ss} - \mathbf{B}\mathbf{u}_{ss} = \mathbf{0}_{n \times 1}$$

$$\mathbf{y}_{r_{ss}} = \mathbf{C}_r \mathbf{x}_{ss} = \mathbf{r} \Leftrightarrow \mathbf{C}_r \mathbf{x}_{ss} = \mathbf{r}$$

- Rewrite the equations in **matrix form**, i.e.

$$\begin{pmatrix} \mathbf{I}_{n \times n} - \mathbf{A} & -\mathbf{B} \\ \mathbf{C}_r & \mathbf{0}_{p_r \times m} \end{pmatrix} \begin{pmatrix} \mathbf{x}_{ss} \\ \mathbf{u}_{ss} \end{pmatrix} = \begin{pmatrix} \mathbf{0}_{n \times 1} \\ \mathbf{r} \end{pmatrix} \quad (7.4)$$

- The steady **target pair** $(\mathbf{x}_{ss}, \mathbf{u}_{ss})$ can be then calculated from (7.4) provided that a solution exists
- Introduce the **state deviation** and **input deviation**

$$\tilde{\mathbf{x}}(k) = \hat{\mathbf{x}}(k) - \mathbf{x}_{ss}, \quad \tilde{\mathbf{u}}(k) = \mathbf{u}(k) - \mathbf{u}_{ss}$$

- This leads to the **discrete-time LTI state equation**

$$\tilde{\mathbf{x}}(k+1) = \hat{\mathbf{x}}(k+1) - \mathbf{x}_{ss} = \mathbf{A}\hat{\mathbf{x}}(k) + \mathbf{B}\mathbf{u}(k) - (\mathbf{A}\mathbf{x}_{ss} + \mathbf{B}\mathbf{u}_{ss}) = \mathbf{A}\tilde{\mathbf{x}}(k) + \mathbf{B}\tilde{\mathbf{u}}(k) \quad (7.5)$$

Reference Tracking based on Target Calculation

- Approach

the **discrete-time quadratic cost function**

$$\tilde{V}_N(\tilde{\mathbf{x}}(k), \tilde{\mathbf{U}}(k)) = \tilde{\mathbf{x}}^T(k+N) \mathbf{P} \tilde{\mathbf{x}}(k+N) + \sum_{i=0}^{N-1} \tilde{\mathbf{x}}^T(k+i) \mathbf{Q} \tilde{\mathbf{x}}(k+i) + \tilde{\mathbf{u}}^T(k+i) \mathbf{R} \tilde{\mathbf{u}}(k+i) \quad (7.6)$$

the **state** and **input constraints**

$$\begin{aligned} \mathbf{C}(\tilde{\mathbf{x}}(k+i) + \mathbf{x}_{ss}) &\in \mathbb{Y}, i = 1, 2, \dots, N \\ \tilde{\mathbf{u}}(k+i) + \mathbf{u}_{ss} &\in \mathbb{U}, \quad i = 0, 1, \dots, N-1 \end{aligned} \quad (7.7)$$

- Rewrite **Problem 4.1/5.1** w.r.t. the state equation (7.5), cost function (7.6) and constraints (7.7), i.e.

$$\begin{aligned} &\min_{\tilde{\mathbf{U}}(k)} \tilde{V}_N(\tilde{\mathbf{x}}(k), \tilde{\mathbf{U}}(k)) \\ &\text{subject to } \begin{cases} \tilde{\mathbf{x}}(k+i+1) = \mathbf{A} \tilde{\mathbf{x}}(k+i) + \mathbf{B} \tilde{\mathbf{u}}(k+i), i = 0, 1, \dots, N-1 \\ \mathbf{C}(\tilde{\mathbf{x}}(k+i) + \mathbf{x}_{ss}) \in \mathbb{Y}, & i = 1, 2, \dots, N \\ \tilde{\mathbf{u}}(k+i) + \mathbf{u}_{ss} \in \mathbb{U}, & i = 0, 1, \dots, N-1 \end{cases} \end{aligned} \quad (7.8)$$

Reference Tracking based on Target Calculation

- **Remark on the Optimization Problem**

- Problem (7.8) can be formulated as a **quadratic program** using the methods from Chapter 4 and 5
- Problem (7.8) relates to a **regulation problem**, i.e. $\tilde{\mathbf{x}}(k) \rightarrow \mathbf{0}, \tilde{\mathbf{u}}(k) \rightarrow \mathbf{0}$ as $k \rightarrow \infty$
- The **reference tracking problem** is, however, simultaneously addressed since

$$\tilde{\mathbf{x}}(k) \rightarrow \mathbf{0}, \tilde{\mathbf{u}}(k) \rightarrow \mathbf{0} \Rightarrow \hat{\mathbf{x}}(k) \rightarrow \mathbf{x}_{ss}, \mathbf{u}(k) \rightarrow \mathbf{u}_{ss} \Rightarrow \mathbf{y}_r(k) \rightarrow \mathbf{r}$$

- **Remark on the Target Calculator**

- Generally it is not possible to control the state $\mathbf{x}(k)$ to an arbitrary target state \mathbf{x}_{ss}
- E.g. it is not possible to maintain a constant position and a constant velocity of a car simultaneously
- A **sufficient condition** for the **existence of a solution of (7.4)** for any reference input \mathbf{r} is that

$$\begin{pmatrix} \mathbf{I}_{n \times n} - \mathbf{A} & -\mathbf{B} \\ \mathbf{C}_r & \mathbf{0}_{p_r \times m} \end{pmatrix} \in \mathbb{R}^{(n+p_r) \times (n+m)} \text{ has full rank } n + p_r$$

- This implies that \mathbf{C}_r must have full rank and number of controlled outputs $p_r \leq$ number of inputs m
- The solution may not be unique

Reference Tracking based on Target Calculation

- Remark on the Constrained Case

- For the constrained case target pair $(\mathbf{x}_{ss}, \mathbf{u}_{ss})$ must fulfill output constraint \mathbb{Y} and input constraint \mathbb{U}
- For this purpose the target calculator based on (7.4) must be modified to

$$\begin{aligned} \min_{\mathbf{x}_{ss}, \mathbf{u}_{ss}} \quad & \frac{1}{2} \left((\mathbf{C}_r \mathbf{x}_{ss} - \mathbf{r})^T \mathbf{Q}_{ss} (\mathbf{C}_r \mathbf{x}_{ss} - \mathbf{r}) + (\mathbf{u}_{ss} - \mathbf{u}_{ss}^{\text{unc}})^T \mathbf{R}_{ss} (\mathbf{u}_{ss} - \mathbf{u}_{ss}^{\text{unc}}) \right) \\ \text{subject to} \quad & \begin{cases} \begin{pmatrix} \mathbf{I}_{n \times n} - \mathbf{A} & -\mathbf{B} \\ \mathbf{C}_r & \mathbf{0}_{p_r \times m} \end{pmatrix} \begin{pmatrix} \mathbf{x}_{ss} \\ \mathbf{u}_{ss} \end{pmatrix} = \begin{pmatrix} \mathbf{0}_{n \times 1} \\ \mathbf{r} \end{pmatrix} \\ \mathbf{C} \mathbf{x}_{ss} \in \mathbb{Y} \\ \mathbf{u}_{ss} \in \mathbb{U} \end{cases} \end{aligned} \quad (7.9)$$

where $\mathbf{u}_{ss}^{\text{unc}}$ is the target input resulting from (7.4) for the unconstrained case and $\mathbf{Q}_{ss} = \mathbf{Q}_{ss}^T \succcurlyeq \mathbf{0}$ and $\mathbf{R}_{ss} = \mathbf{R}_{ss}^T \succ \mathbf{0}$ are weighting matrices

- Problem (7.9) can be formulated as a quadratic program
- Feasibility of Problem (7.9) is discussed in [RM09, Section 1.5.1]

Reference Tracking based on Target Calculation

- Remark on Receding Horizon Control

1. Estimate the current state $\hat{\mathbf{x}}(k)$
2. Solve Problem (7.9) for the given reference input $\mathbf{r}(k)$ to determine the target pair $(\mathbf{x}_{ss}, \mathbf{u}_{ss})$
3. Solve Problem (7.8) for $\tilde{\mathbf{x}}(k) = \hat{\mathbf{x}}(k) - \mathbf{x}_{ss}$ to determine the optimal input sequence $\tilde{\mathbf{U}}^*(k)$
4. Compute first element of the optimal input sequence $\tilde{\mathbf{u}}^*(k) = (\mathbf{I}_{m \times m} \quad \mathbf{0}_{m \times m} \quad \cdots \quad \mathbf{0}_{m \times m}) \tilde{\mathbf{U}}^*(k)$
5. Implement the optimal input $\mathbf{u}^*(k) = \tilde{\mathbf{u}}^*(k) + \mathbf{u}_{ss}$
6. Increment the time instant $k := k + 1$ and go to 1.

- Further Remarks

- The extension for state constraints and controlled output constraints is straightforward
- The structure on Slide 7-3 is essentially equivalent to the state-command structure on Slide 2-49ff
- Disturbances and uncertainties in \mathbf{A} , \mathbf{B} and \mathbf{C}_r lead to a steady-state error or offset
- More details and references are given in [RM09, Section 1.5.1], [BBM17, Section 12.7], and [MR93]

Miscellaneous

- [MR93] Kenneth R. Muske and James B. Rawlings. Model predictive control with linear models. *AIChE Journal*, 39(2):262–287, 1993.
- [PR03] Gabriele Pannocchia and James B. Rawlings. Disturbance models for offset-free model predictive control. *AIChE Journal*, 49(2):426–437, 2003.