

# Chapter 6 Trajectory Following Homework

## 1、MPC Controller Design

### 1.1 System Model

Due to low vehicle speed, we utilize the kinematic model as the system model:

$$\begin{cases} \dot{p}_x = v \cos(\phi) \\ \dot{p}_y = v \sin(\phi) \\ \dot{\phi} = \frac{v}{L} \tan(\delta) \\ \dot{v} = a \end{cases}$$

State:  $X = [x \ y \ \phi \ v]^T$ , Input:  $U = [a \ \delta]$

- *Linearization*

For improving the solving efficiency, we linearize the nonlinear system model. There are two different method.

- Linearization along the state trajectory:

We can linearize the system model based on the result of last run.

- **Linearization along the reference trajectory**

we can linearize the system model based on the reference state.

Define the linearization point as follows:

$$\bar{X} = [\bar{p}_x \ \bar{p}_y \ \bar{\phi} \ \bar{v}]^T, \bar{U} = [\bar{a} \ \bar{\delta}]^T$$

The linear model:

$$\begin{bmatrix} \dot{p}_x \\ \dot{p}_y \\ \dot{\phi} \\ \dot{v} \end{bmatrix} = \begin{pmatrix} 0 & 0 & -\bar{v} \sin \bar{\phi} & \cos \bar{\phi} \\ 0 & 0 & \bar{v} \cos \bar{\phi} & \sin \bar{\phi} \\ 0 & 0 & 0 & \frac{\tan \bar{\delta}}{L} \\ 0 & 0 & 0 & 0 \end{pmatrix} \begin{bmatrix} p_x \\ p_y \\ v \\ \phi \end{bmatrix} + \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ 0 & \frac{\bar{v}}{L} \frac{1}{\cos^2 \bar{\delta}} \\ 1 & 0 \end{pmatrix} \begin{bmatrix} a \\ \delta \end{bmatrix} + \begin{pmatrix} \bar{v} \sin \bar{\phi} \\ -\bar{v} \cos \bar{\phi} \\ -\bar{v} \frac{\bar{\delta}}{L \cos^2 \bar{\delta}} \\ 0 \end{pmatrix}$$

- Discretization

For computer calculation, we discretize the continuous system with forward Euler difference:

$$\begin{aligned} A_{\text{discrete}} &= I + \Delta T \times A_{\text{linearized}} \\ B_{\text{discrete}} &= \Delta T \times B_{\text{linearized}} \\ g_{\text{discrete}} &= \Delta T \times g_{\text{linearized}} \end{aligned}$$

### 1.2 Define Cost Function

The objective is to follow the trajectory accurately, the cost function can be defined as follows:

$$\begin{aligned}
& \min \sum_{k=1}^N (x_k - \bar{x}_k)^T Q (x_k - \bar{x}_k) + u_{k-1}^T R u_{k-1} \\
& \text{s.t. } x_0 = x, \\
& x_{k+1} = f(x_k, u_k), \quad k = 0, \dots, N-1 \\
& \underline{x} \leq x_k \leq \bar{x}, \quad k = 1, \dots, N \\
& \underline{u} \leq u_k \leq \bar{u}, \quad k = 0, \dots, N-1
\end{aligned}$$

### 1.3 Prediction

$$\eta = AAx_0 + BBu + G$$

$$\eta = [x_1 \ x_2 \ \dots \ x_N]^T, U = [u_0 \ u_1 \ u_2 \ \dots \ u_{N-1}]^T$$

$$\begin{aligned}
AA &= \begin{bmatrix} A_0 \\ A_1 A_0 \\ \vdots \\ \prod_{k=0}^{N-1} A_k \end{bmatrix} \\
BB &= \begin{bmatrix} B_0 & 0 & \dots & 0 & 0 \\ A_1 B_0 & B_1 & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ \prod_{k=1}^{N-1} A_k B_0 & \prod_{k=2}^{N-1} A_k B_1 & \dots & A_{N-1} B_{N-2} & B_{N-1} \end{bmatrix} \\
G &= \begin{bmatrix} g_0 \\ A_1 g_0 + g_1 \\ \vdots \\ \sum_{n=0}^{N-2} \left( \prod_{k=n+1}^{N-1} A_k \right) g_n + g_{N-1} \end{bmatrix}
\end{aligned}$$

Thus we can remove the equality constrain of the optimize problem.

### 1.4 OSQP Solver

we adopt the QP solver OSQP to solve the QP problem.

OSQP solves convex quadratic programs (QPs) of the form

$$\begin{aligned}
& \text{minimize} && \frac{1}{2} x^T P x + q^T x \\
& \text{subject to} && l \leq A x \leq u
\end{aligned}$$

where  $x \in \mathbf{R}^n$  is the optimization variable. The objective function is defined by a positive semidefinite matrix  $P \in \mathbf{S}_+^n$  and vector  $q \in \mathbf{R}^n$ . The linear constraints are defined by matrix  $A \in \mathbf{R}^{m \times n}$  and vectors  $l$  and  $u$  so that  $l_i \in \mathbf{R} \cup \{-\infty\}$  and  $u_i \in \mathbf{R} \cup \{+\infty\}$  for all  $i \in \{1, \dots, m\}$ .

Substituting the prediction model into the cost function, we have

$$U^T (BB^T Q BB) U + 2(x_0 A A^T Q BB + G G^T Q BB - q_x Q BB) U$$

$$q_x = [\bar{x}_1, \dots, \bar{x}_N]$$

## 2、Result

