### • Get $A_d B_d g_d$

Convert linear model to discrete-time using forward Euler method

$$\dot{x} = A_c x + B_c u + g_c$$

Linearization and discretization the model equation:

$$\begin{cases} \dot{p_x} = v \cos(\phi) \\ \dot{p_y} = v \sin(\phi) \\ \dot{\phi} = \frac{v}{L} \tan(\delta) \\ \dot{v} = a \end{cases} \begin{cases} \dot{p_x} = \bar{v} \cos(\bar{\phi}) + \cos(\bar{\phi}) (v - \bar{v}) - v \sin(\bar{\phi}) (\phi - \bar{\phi}) \\ \dot{p_y} = \bar{v} \sin(\bar{\phi}) + \sin(\bar{\phi}) (v - \bar{v}) + v \cos(\bar{\phi}) (\phi - \bar{\phi}) \\ \dot{\phi} = \frac{\bar{v}}{L} \tan(\bar{\delta}) + \frac{\tan \bar{\delta}}{L} (v - \bar{v}) + \frac{\bar{v}}{L} \frac{1}{\cos^2 \bar{\delta}} (\delta - \bar{\delta}) \\ \dot{v} = a \end{cases}$$

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

$$\frac{\boldsymbol{x}_{k+1} - \boldsymbol{x}_k}{T_s} = \boldsymbol{A}_c \boldsymbol{x}_k + \boldsymbol{B}_c \boldsymbol{u}_k + \boldsymbol{g}_c$$

$$\mathbf{x}_{k+1} = (\mathbf{I} + T_s \mathbf{A}_c) \mathbf{x}_k + T_s \mathbf{B}_c \mathbf{u}_k + T_s \mathbf{g}_c$$
$$\mathbf{A}_d = \mathbf{I} + T_s \mathbf{A}_c \quad \mathbf{B}_d = T_s \mathbf{B}_c \quad \mathbf{g}_d = T_s \mathbf{g}_c$$

#### Set cost function

The position error and polar angle error of the car are required to be as small as possible in the current position and trajectory:

$$J = (x - x_{ref})^2 + (y - y_{ref})^2 + \rho(\phi - \phi_{ref})^2$$

State maxtrix 
$$x = \begin{pmatrix} p_x \\ p_y \\ \phi \\ v \end{pmatrix}$$
, and  $Q = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & \rho & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$ 

So the cost funtion become:

$$J_{i} = (x - x_{ref})^{T} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & \rho & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} (x - x_{ref})$$

Note that when i=horizon:

$$J_N = (x - x_{ref})^T \begin{pmatrix} \rho_N & 0 & 0 & 0 \\ 0 & \rho_N & 0 & 0 \\ 0 & 0 & \rho_N * \rho & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} (x - x_{ref})$$

#### Set the constrain

$$-v_{max} \leqslant v \leqslant v_{max}$$
$$-a_{max} \leqslant a \leqslant a_{max}$$
$$-\delta_{max} \leqslant \delta \leqslant \delta_{max}$$

Note that the constrain of  $\,d\delta\,$  :

$$-d\delta_{max} * dt \le d\delta = \delta_k - \delta_{k-1} \le d\delta_{max} * dt$$

### Solving quadratic programming problems

$$\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_{N-1} \\ x_N \end{bmatrix} = \begin{bmatrix} B & \mathbf{0} & \cdots & \mathbf{0} \\ AB & B & \cdots & \mathbf{0} \\ \vdots & \vdots & \ddots & \vdots \\ A^{N-1}B & A^{N-2}B & \cdots & B \end{bmatrix} \begin{bmatrix} u_0 \\ u_1 \\ \vdots \\ u_{N-1} \end{bmatrix} + \begin{bmatrix} A \\ A^2 \\ \vdots \\ A^N \end{bmatrix} x_0$$

Extended System State:

$$X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_N \end{bmatrix}$$

Extended  $A_0$ :

$$A_0 = \begin{bmatrix} A_0 \\ A_1 A_0 \\ \vdots \\ \prod_{k=0}^{N-1} A_k \end{bmatrix}$$

Extended B, which contains all the control matrices in discrete system equations, stacked:

$$B = \begin{bmatrix} B_0 & 0 & \dots & 0 & 0 \\ A_1 B_0 & B_1 & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ \prod_{k=1}^{N-1} A_k B_0 & \prod_{k=2}^{N-1} A_k B_1 & \dots & A_{N-1} B_{N-2} & B_{N-1} \end{bmatrix}$$

**Extended G**, which contains all the constant terms in discrete system equations, stacked:

$$G = \begin{bmatrix} g_0 \\ A_1 g_0 + g_1 \\ \vdots \\ \sum_{n=0}^{N-2} (\prod_{k=n+1}^{N-1} A_k) g_n + g_{N-1} \end{bmatrix}$$

$$\mathbf{q}_x = -\mathbf{Q}^T \mathbf{x}_{ref}$$

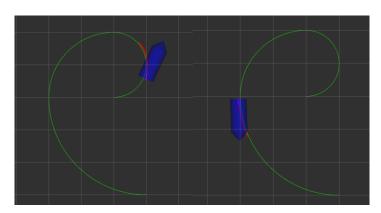
## Get delay MPC new\_x<sub>0</sub>

The initial state of the system can be defined as the amount of state after the current moment  $\tau$ 

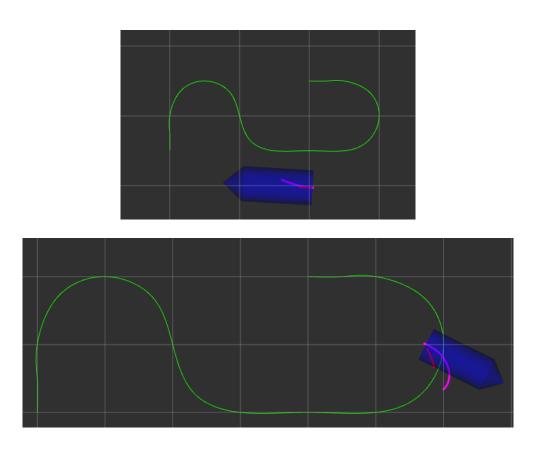
$$\overline{x}_0 = x(t+\tau) \approx \widehat{x}(t+\tau) = A^{\tau}x(t) + \sum_{j=0}^{\tau-1} A^j B u(t-1-j)$$
Past inputs

Let  $N=t+\tau$  we get  $new\_x_0=x_N$  , and according  $X=BB*U+AA*x_0+gg$  , finally we can get  $new\_x_0$  from X

# • Simulation result



MPC performance is affected by the initial position and yaw angle and horizon, so the other two paths perform not very well



# • Bugs:

$$B = \begin{bmatrix} B_0 & 0 & \dots & 0 & 0 \\ A_1 B_0 & B_1 & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ \prod_{k=1}^{N-1} A_k B_0 & \prod_{k=2}^{N-1} A_k B_1 & \dots & A_{N-1} B_{N-2} & B_{N-1} \end{bmatrix}$$

Multiplying A sequentially from right to left causes numerical instability when N\_ is large. Dynamic programming from top to bottom multiplies one A at a time will not cause numerical instability