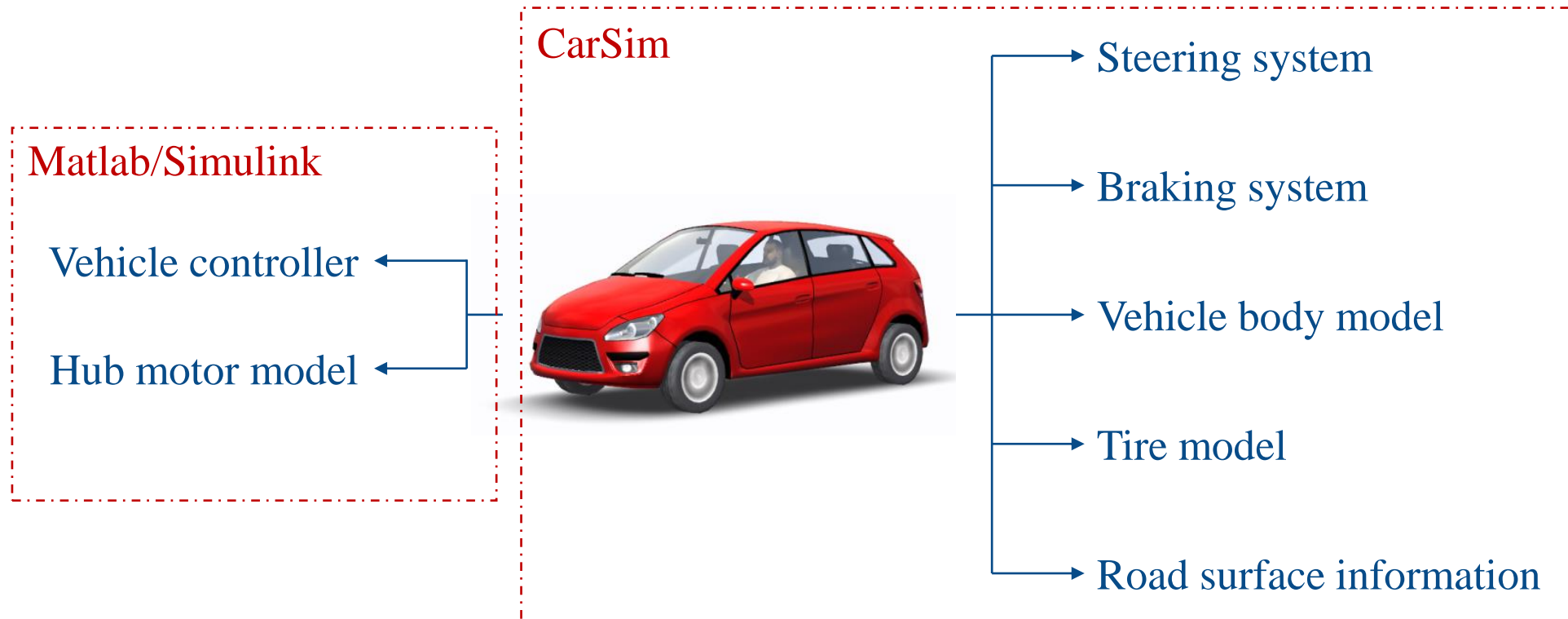




# Collaborative Control of Wire-Controlled Chassis Subsystems Based on Game Theory

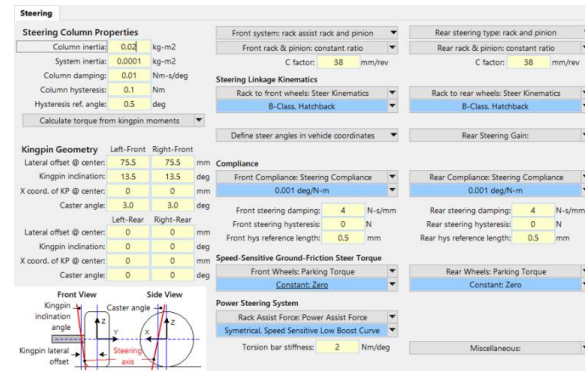


## Overall framework - B-Class, Hatchback

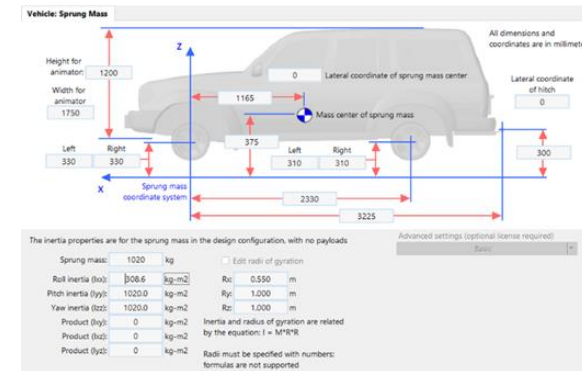




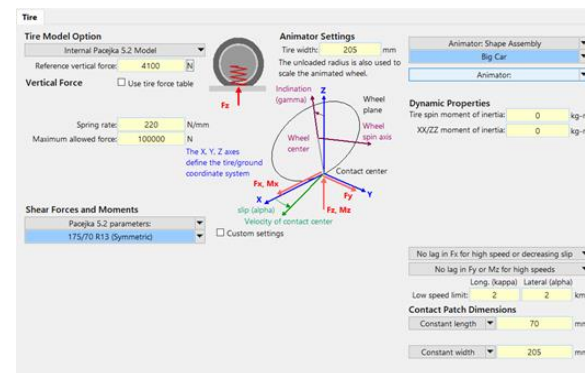
## Using built-in components of Carsim



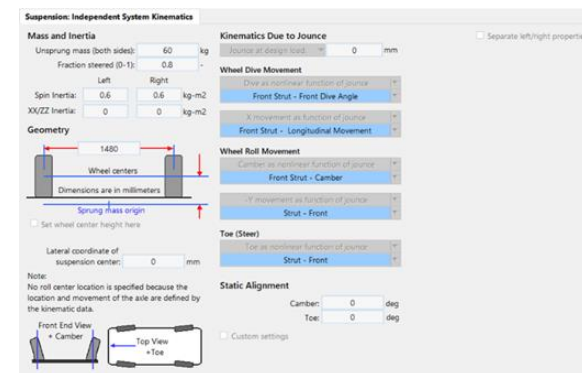
Steering system



Vehicle body model



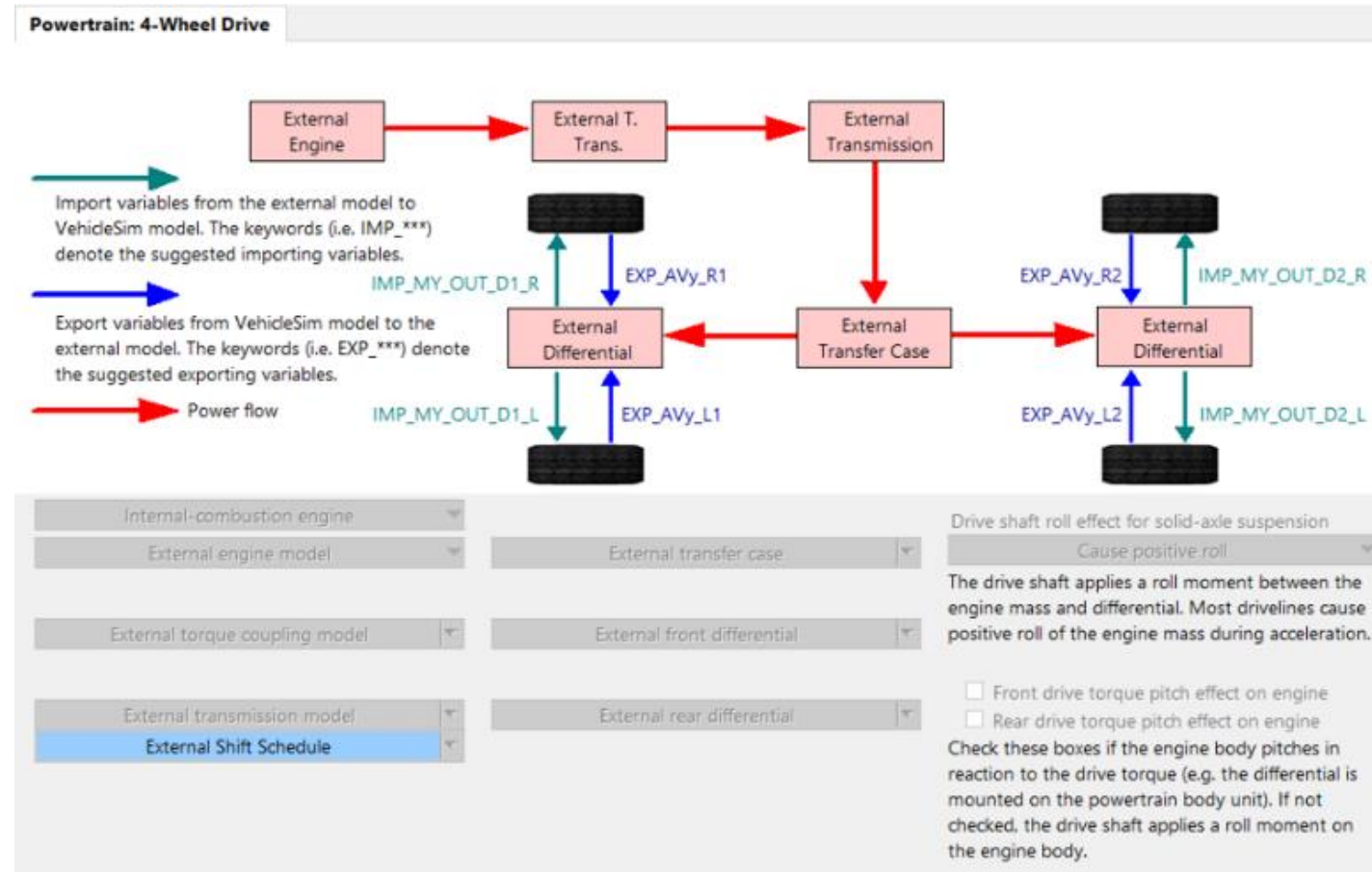
Tire model



Suspension model



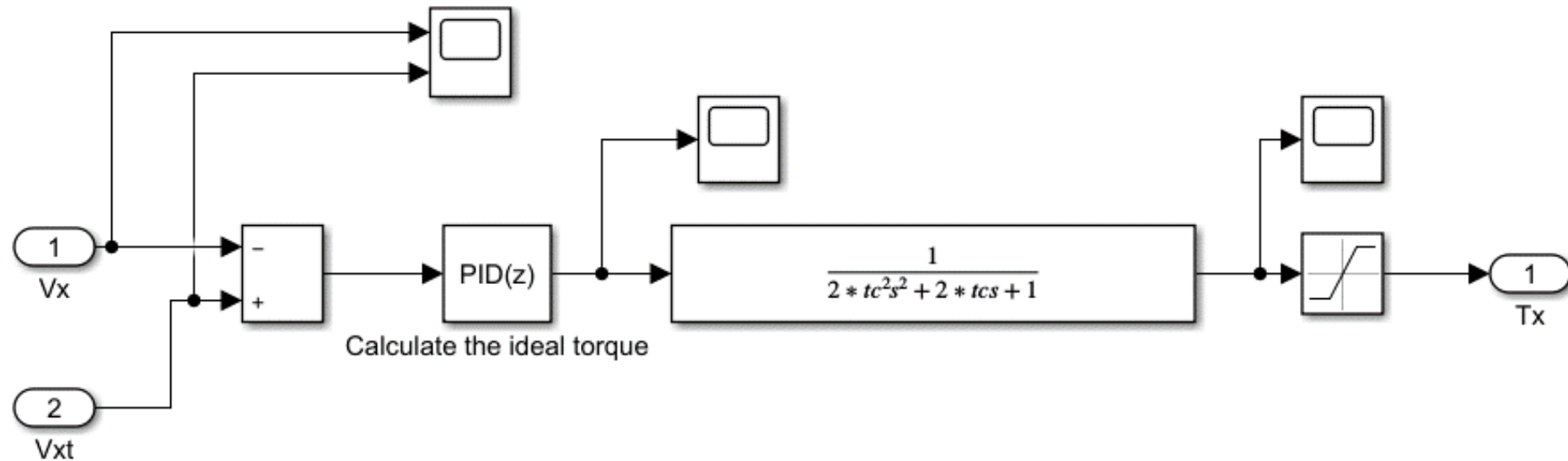
## Main modification part - Powertrain system



Electric vehicle (front wheel steering angle and four-wheel torque)



## Main Modification Section - Powertrain System



$$K(s) = \frac{T_d}{T_{ld}} = \frac{1}{1 + 2\xi s + 2\xi^2 s^2}$$

$$\xi = 0.001$$



## Primary inputs and outputs

Interface	Interface Name	Interface Definitions
Input Interface	IMP_MYUSM_L1	Left front wheel input torque
	IMP_MYUSM_L2	Left rear wheel input torque
	IMP_MYUSM_R1	Right front wheel input torque
	IMP_MYUSM_R2	Right rear wheel input torque
	IMP_STEER_L1	Left front wheel input steering angle
	IMP_STEER_R1	Right front wheel input steering angle
Output Interface	Beta-Vehicle slip angle	Vehicle center of mass lateral deviation angle
	Vx-Longitudinal speed	Vehicle longitudinal velocity
	Vy-Lateral speed	Vehicle lateral velocity
	AVz-Yaw rate	Vehicle yaw rate
	AVy_L1-Wheel L1 spin	Left front wheel speed
	AVy_R1-Wheel R1 spin	Right front wheel speed
	Fz_L1	Left front wheel vertical tire force
	Fz_L2	Left rear wheel vertical tire force
	Fz_R1	Right front wheel vertical tire force
	Fz_R2	Right rear wheel vertical tire force



## ■ Non-cooperative game theory

- Linear control state equation:

$$\dot{x}(t) = Ax(t) + B_1u_1(t) + B_2u_2(t)$$

- Cost function:

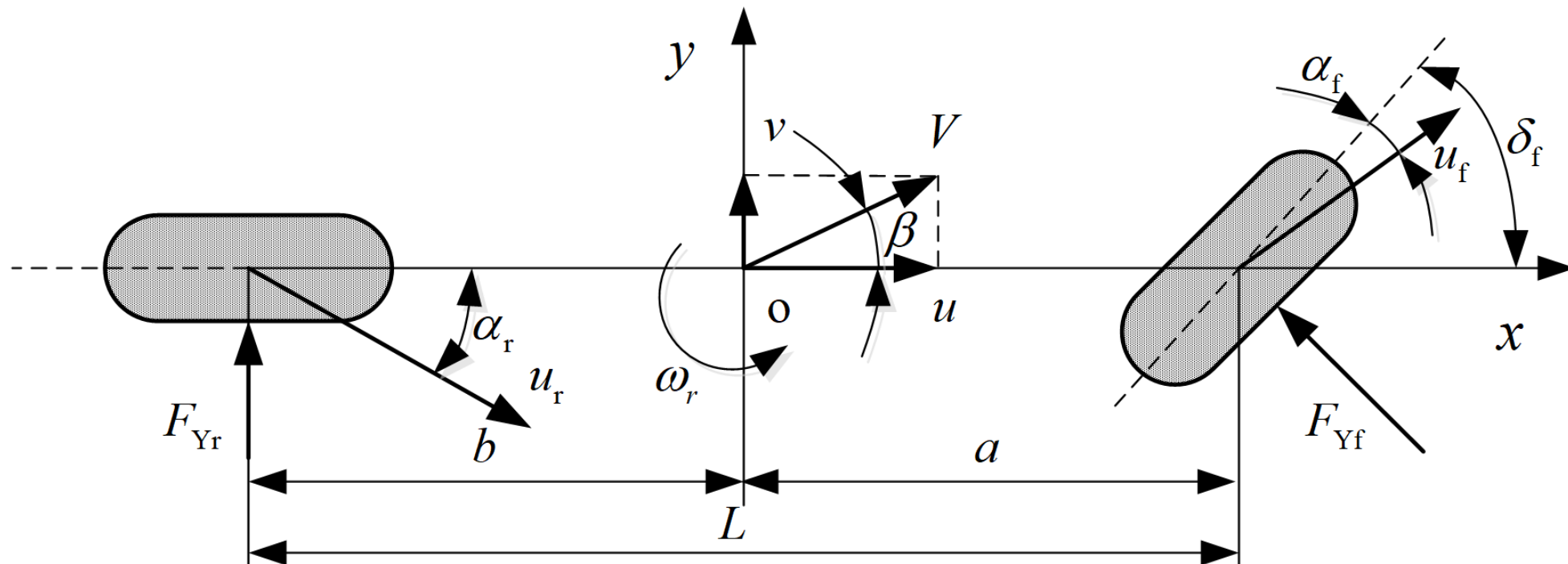
$$\begin{cases} J_1 = \int_0^T \{x^T(t)Q_1x(t) + u_1^T(t)R_{11}u_1(t) + u_2^T(t)R_{12}u_2(t)\}dt \\ J_2 = \int_0^T \{x^T(t)Q_2x(t) + u_1^T(t)R_{21}u_1(t) + u_2^T(t)R_{22}u_2(t)\}dt \end{cases}$$

- Nash equilibrium control strategy:

$$\begin{cases} J_1(u_{F1}, u_{F2}) \leq J_2(u_1, u_{F2}) \\ J_2(u_{F1}, u_{F2}) \leq J_2(u_{F1}, u_2) \end{cases}$$



## Upper-level controller design - Two-degree-of-freedom model







## Upper-level controller design - Two-degree-of-freedom model

- Differential equations for the two-degree-of-freedom model:

$$\begin{cases} (k_f + k_r)\beta + \frac{1}{V_x}(ak_f - bk_r)\omega - k_f\delta_f = m(\dot{V}_y + V_x\omega) \\ (ak_f - bk_r)\beta + \frac{1}{V_x}(a^2k_f - b^2k_r)\omega - ak_f\delta_f = I_z\dot{\omega} \end{cases}$$

- State-space equations for the two-degree-of-freedom model:

$$\begin{bmatrix} \dot{\beta} \\ \dot{\omega} \end{bmatrix} = \begin{bmatrix} -\frac{k_f + k_r}{mV_x} & -\frac{ak_f - bk_r}{mV_x^2} - 1 \\ -\frac{ak_f - bk_r}{I_z} & -\frac{a^2k_f + b^2k_r}{I_zV_x} \end{bmatrix} \begin{bmatrix} \beta \\ \omega \end{bmatrix} + \begin{bmatrix} \frac{k_f}{mV_x} \\ \frac{ak_f}{I_z} \end{bmatrix} \delta_f$$



## Upper-level controller design - Two-degree-of-freedom model

- For the sake of convenience in research, redefine the state-space equations:

$$\begin{bmatrix} \Delta \dot{\beta} \\ \Delta \dot{\omega} \end{bmatrix} = \begin{bmatrix} -\frac{k_f + k_r}{mV_x} & -\frac{ak_f - bk_r}{mV_x^2} - 1 \\ -\frac{ak_f - bk_r}{I_z} & -\frac{a^2k_f + b^2k_r}{I_z V_x} \end{bmatrix} \begin{bmatrix} \Delta \beta \\ \Delta \omega \end{bmatrix} + \begin{bmatrix} \frac{k_f}{mV_x} \\ \frac{ak_f}{I_z} \end{bmatrix} \Delta \delta_f + \begin{bmatrix} 0 \\ \frac{1}{I_z} \end{bmatrix} \Delta M$$
  

$$\begin{cases} \Delta \beta = \beta - \beta_{des} \\ \Delta \omega = \omega - \omega_{des} \end{cases} \rightarrow \begin{cases} \beta_{des} = \frac{b - \frac{amV_x^2}{k_r(a+b)}}{(a+b)(1+K_{us}V_x^2)} \delta_f \\ \omega_{des} = \frac{V_x}{(a+b)(1+K_{us}V_x^2)} \delta_f \\ K_{us} = \frac{m(bk_2 - ak_1)}{(a+b)^2 k_f k_r} \end{cases}$$



## ■ Upper-level controller design - Two-degree-of-freedom model

- For ease of computation, simplify the state-space equations:

$$\begin{bmatrix} \Delta \dot{\beta} \\ \Delta \dot{\omega} \end{bmatrix} = \begin{bmatrix} -\frac{k_f + k_r}{mV_x} & -\frac{ak_f - bk_r}{mV_x^2} - 1 \\ -\frac{ak_f - bk_r}{I_z} & -\frac{a^2k_f + b^2k_r}{I_z V_x} \end{bmatrix} \begin{bmatrix} \Delta \beta \\ \Delta \omega \end{bmatrix} + \begin{bmatrix} \frac{k_f}{mV_x} \\ \frac{ak_f}{I_z} \end{bmatrix} \Delta \delta_f + \begin{bmatrix} 0 \\ \frac{1}{I_z} \end{bmatrix} \Delta M$$

$$\dot{x} = Ax_n + B_1u_1 + B_2u_2$$



## ■ Upper-level controller design - Cost function

- Define the game model based on the cost function:

$$\begin{cases} J_1(u_1, u_2) = \int_0^\infty \frac{1}{2} \{x_n^T(t) Q_1 x_n(t) + u_1^T(t) R_{11} u_1(t) + u_2^T(t) R_{12} u_2(t)\} dt \\ J_2(u_1, u_2) = \int_0^\infty \frac{1}{2} \{x_n^T(t) Q_2 x_n(t) + u_1^T(t) R_{21} u_1(t) + u_2^T(t) R_{22} u_2(t)\} dt \end{cases}$$

- The cost function satisfies the following conditions:

$$\begin{cases} J_1(u_1^*, u_2^*) \leq J_1(u_1, u_2^*) \\ J_2(u_1^*, u_2^*) \leq J_2(u_1^*, u_2) \end{cases}$$



## Upper-level controller design - Hamiltonian equation

- Establish the Hamiltonian equation

$$\begin{cases} H_1 = \frac{1}{2}(x_n^T Q_1 x_n + u_1^T R_{11} u_1 + u_2^T R_{12} u_2) + \lambda_1^T (Ax_n + B_1 u_1 + B_2 u_2) \\ H_2 = \frac{1}{2}(x_n^T Q_2 x_n + u_1^T R_{21} u_1 + u_2^T R_{22} u_2) + \lambda_2^T (Ax_n + B_1 u_1 + B_2 u_2) \end{cases}$$

- Construct the extremum condition equation

$$\begin{cases} \frac{\partial H_1}{\partial x_n} = R_{11} u_1 + B_1^T \lambda_1 = 0 \\ \frac{\partial H_2}{\partial x_n} = R_{22} u_2 + B_2^T \lambda_2 = 0 \end{cases} \xrightarrow{\text{solve}} \begin{cases} u_1 = -R_{11}^{-1} B_1^T \lambda_1 \\ u_2 = -R_{22}^{-1} B_2^T \lambda_2 \end{cases}$$



## Upper-level controller design - Hamiltonian equation

- Substitute the simplified state-space equations

$$\begin{cases} u_1 = -R_{11}^{-1} B_1^T \lambda_1 \\ u_2 = -R_{22}^{-1} B_2^T \lambda_2 \end{cases} \longrightarrow \dot{x} = Ax_n + B_1 u_1 + B_2 u_2 \longrightarrow \dot{x}_n = Ax_n - B_1 R_{11}^{-1} B_1^T \boxed{\lambda_1} - B_2 R_{22}^{-1} B_2^T \boxed{\lambda_2}$$

- According to the optimization control conditions

$$\begin{cases} \dot{\lambda}_1 = -\frac{\partial H_1}{\partial x_n} = -A^T \lambda_1 - Q_1 x_n \\ \dot{\lambda}_2 = -\frac{\partial H_2}{\partial x_n} = -A^T \lambda_2 - Q_2 x_n \end{cases}$$

- Write it in terms of the co-state matrix

$$\begin{bmatrix} \dot{x}_n \\ \dot{\lambda}_1 \\ \dot{\lambda}_2 \end{bmatrix} = \begin{bmatrix} A & -B_1 R_{11}^{-1} B_1^T & -B_2 R_{22}^{-1} B_2^T \\ -Q_1 & -A^T & 0 \\ -Q_2 & 0 & -A^T \end{bmatrix} \cdot \begin{bmatrix} x_n \\ \lambda_1 \\ \lambda_2 \end{bmatrix}$$



## Upper-level controller design - Hamiltonian equation

- There is a relationship between state variables

$$\begin{bmatrix} \dot{x}_n \\ \dot{\lambda}_1 \\ \dot{\lambda}_2 \end{bmatrix} = \begin{bmatrix} A & -B_1 R_{11}^{-1} B_1^T & -B_2 R_{22}^{-1} B_2^T \\ -Q_1 & -A^T & 0 \\ -Q_2 & 0 & -A^T \end{bmatrix} \begin{bmatrix} x_n \\ \lambda_1 \\ \lambda_2 \end{bmatrix} \rightarrow \begin{cases} \lambda_1 = G_1 x_n \\ \lambda_2 = G_2 x_n \end{cases}$$

- Substitute into the extremum condition equation for a solution

$$\begin{cases} u_1 = -R_{11}^{-1} B_1^T \lambda_1 \\ u_2 = -R_{22}^{-1} B_2^T \lambda_2 \end{cases} \rightarrow \begin{cases} u_1^* = -R_{11}^{-1} B_1^T G_1 x_n \\ u_2^* = -R_{22}^{-1} B_2^T G_2 x_n \end{cases}$$



## ■ Upper-level controller design - Riccati equation

- $G_1$  and  $G_2$  are solutions to the following coupled Riccati equations

$$\begin{cases} (A - T_2 G_2)^T G_1 + G_1 (A - T_2 G_2) - G_1 T_1 G_1 + Q_1 + G_2 T_{12} G_2 = 0 \\ (A - T_1 G_1)^T G_2 + G_2 (A - T_1 G_1) - G_2 T_2 G_2 + Q_2 + G_1 T_{21} G_1 = 0 \end{cases}$$

- Expressions for  $T_{ij}$  and  $T_i$  are as follows

$$\begin{cases} T_i = B_i R_{ii}^{-1} B_i^T & i = 1, 2 \\ T_{ij} = B_i R_{ii}^{-1} R_{ji} R_{ii}^{-1} B_i^T & i, j = 1, 2 \text{ \& } i \neq j \end{cases}$$

**The game process is specifically reflected in the process of solving the coupled Riccati equations.**





## Upper-level controller design - Riccati equation

- Discretize and reconstruct the equations as follows:

$$\begin{cases} (A - T_2 G_2)^T G_1 + G_1 (A - T_2 G_2) - G_1 T_1 G_1 + Q_1 + G_2 T_{12} G_2 = 0 \\ (A - T_1 G_1)^T G_2 + G_2 (A - T_1 G_1) - G_2 T_2 G_2 + Q_2 + G_1 T_{21} G_1 = 0 \end{cases}$$

↓

$$\begin{cases} (A - T_2 G_2^c)^T G_1^{c+1} + G_1^{c+1} (A - T_2 G_2^c) - G_1^{c+1} T_1 G_1^{c+1} + Q_1 + G_2^c T_{12} G_2^c = 0 \\ (A - T_1 G_1^c)^T G_2^{c+1} + G_2^{c+1} (A - T_1 G_1^c) - G_2^{c+1} T_2 G_2^{c+1} + Q_2 + G_1^c T_{21} G_1^c = 0 \end{cases}$$

- Obtain the initial solutions for the decoupled Riccati equations through the following equations:

$$\begin{cases} A^T G_1^0 + G_1^0 A + Q_1 + G_1^0 T_{11} G_1^0 = 0 \\ A^T G_2^0 + G_2^0 A + Q_2 + G_2^0 T_{11} G_2^0 = 0 \end{cases}$$



## Upper-level controller design - Riccati equation

- Treating  $G^0$  as a constant

$$\begin{cases} (A - T_2 G_2^c)^T G_1^{c+1} + G_1^{c+1} (A - T_2 G_2^c) - G_1^{c+1} T_1 G_1^{c+1} + Q_1 + G_2^c T_{12} G_2^c = 0 \\ (A - T_1 G_1^c)^T G_2^{c+1} + G_2^{c+1} (A - T_1 G_1^c) - G_2^{c+1} T_2 G_2^{c+1} + Q_2 + G_1^c T_{21} G_1^c = 0 \end{cases}$$

$\swarrow \quad \nwarrow$ 

$$\begin{cases} A_1^T = (A - T_2 G_2^c)^T \\ A_2^T = (A - T_1 G_1^c)^T \end{cases}$$

$\downarrow$ 

$$\begin{cases} Q_1' = Q_1 + G_2^c T_{12} G_2^c \\ Q_2' = Q_2 + G_1^c T_{21} G_1^c \end{cases}$$

- Transform the equation into the same form as the strong solution

$$\begin{cases} A_1^T G_1^{c+1} + G_1^{c+1} A_1 + Q_1' + G_1^{c+1} T_{11} G_1^{c+1} = 0 \\ A_2^T G_2^{c+1} + G_2^{c+1} A_2 + Q_2' + G_2^{c+1} T_{22} G_2^{c+1} = 0 \end{cases}$$



## ■ Upper-level controller design - Riccati equation

- Iteration termination condition:

$$\begin{cases} \|G_1^{c+1} - G_1^c\| < 10^{-4} \\ \|G_2^{c+1} - G_2^c\| < 10^{-4} \end{cases}$$

- Computed results:

$$\begin{cases} u_1^* = -R_{11}^{-1} B_1^T G_1^{c+1} x_n \\ u_2^* = -R_{22}^{-1} B_2^T G_2^{c+1} x_n \end{cases}$$

- Corresponding results:

$$\begin{cases} \Delta \delta_f = u_1^* \\ \Delta M = u_2^* \end{cases}$$



## Lower-level controller design - Quadratic programming

$$\left\{ \begin{array}{l} \min \varepsilon = \sum_j \frac{T_{xj}^2}{(\mu F_{zj})^2 R^2} \\ s.t. \quad T_x = \sum_j T_{xj} \\ \Delta M = \frac{t_r}{2} \left( \frac{T_{xfr}}{R} - \frac{T_{xfl}}{R} \right) + \frac{t_r}{2} \left( \frac{T_{xrl}}{R} - \frac{T_{xrr}}{R} \right) \\ 0 \leq T_{xj} \leq \min(T_{m\_max}, \mu F_{zj} R) \end{array} \right.$$



## Lower-level controller design - Quadratic programming

$$\left\{ \begin{array}{l} \min \varepsilon = \sum_j \frac{T_{xj}^2}{(\mu F_{zj})^2 R^2} \\ s.t. \quad T_x = \sum_j T_{xj} \\ \Delta M = \frac{t_r}{2} \left( \frac{T_{xfr}}{R} - \frac{T_{xfl}}{R} \right) + \frac{t_r}{2} \left( \frac{T_{xrl}}{R} - \frac{T_{xrr}}{R} \right) \\ 0 \leq T_{xj} \leq \min(T_{m\_max}, \mu F_{zj} R) \end{array} \right.$$

With the objective of minimizing tire utilization



## Lower-level controller design - Quadratic programming

$$\left\{ \begin{array}{l} \min \varepsilon = \sum_j \frac{T_{xj}^2}{(\mu F_{zj})^2 R^2} \\ s.t. \quad T_x = \sum_j T_{xj} \\ \Delta M = \frac{t_r}{2} \left( \frac{T_{xfr}}{R} - \frac{T_{xfl}}{R} \right) + \frac{t_r}{2} \left( \frac{T_{xrl}}{R} - \frac{T_{xrr}}{R} \right) \\ 0 \leq T_{xj} \leq \min(T_{m\_max}, \mu F_{zj} R) \end{array} \right.$$

Vehicle propulsion total torque =  
Sum of torques from all four drive wheels



## Lower-level controller design - Quadratic programming

$$\left\{ \begin{array}{l} \min \varepsilon = \sum_j \frac{T_{xj}^2}{(\mu F_{zj})^2 R^2} \\ s.t. \quad T_x = \sum_j T_{xj} \\ \Delta M = \frac{t_r}{2} \left( \frac{T_{xfr}}{R} - \frac{T_{xfl}}{R} \right) + \frac{t_r}{2} \left( \frac{T_{xrl}}{R} - \frac{T_{xrr}}{R} \right) \\ 0 \leq T_{xj} \leq \min(T_{m\_max}, \mu F_{zj} R) \end{array} \right.$$

Additional total torque should also satisfy allocation conditions



## Lower-level controller design - Quadratic programming

$$\left\{ \begin{array}{l} \min \varepsilon = \sum_j \frac{T_{xj}^2}{(\mu F_{zj})^2 R^2} \\ s.t. \quad T_x = \sum_j T_{xj} \\ \Delta M = \frac{t_r}{2} \left( \frac{T_{xfr}}{R} - \frac{T_{xfl}}{R} \right) + \frac{t_r}{2} \left( \frac{T_{xrl}}{R} - \frac{T_{xrr}}{R} \right) \\ 0 \leq T_{xj} \leq \min(T_{m\_max}, \mu F_{zj} R) \end{array} \right.$$

→ Subject to peak torque limits





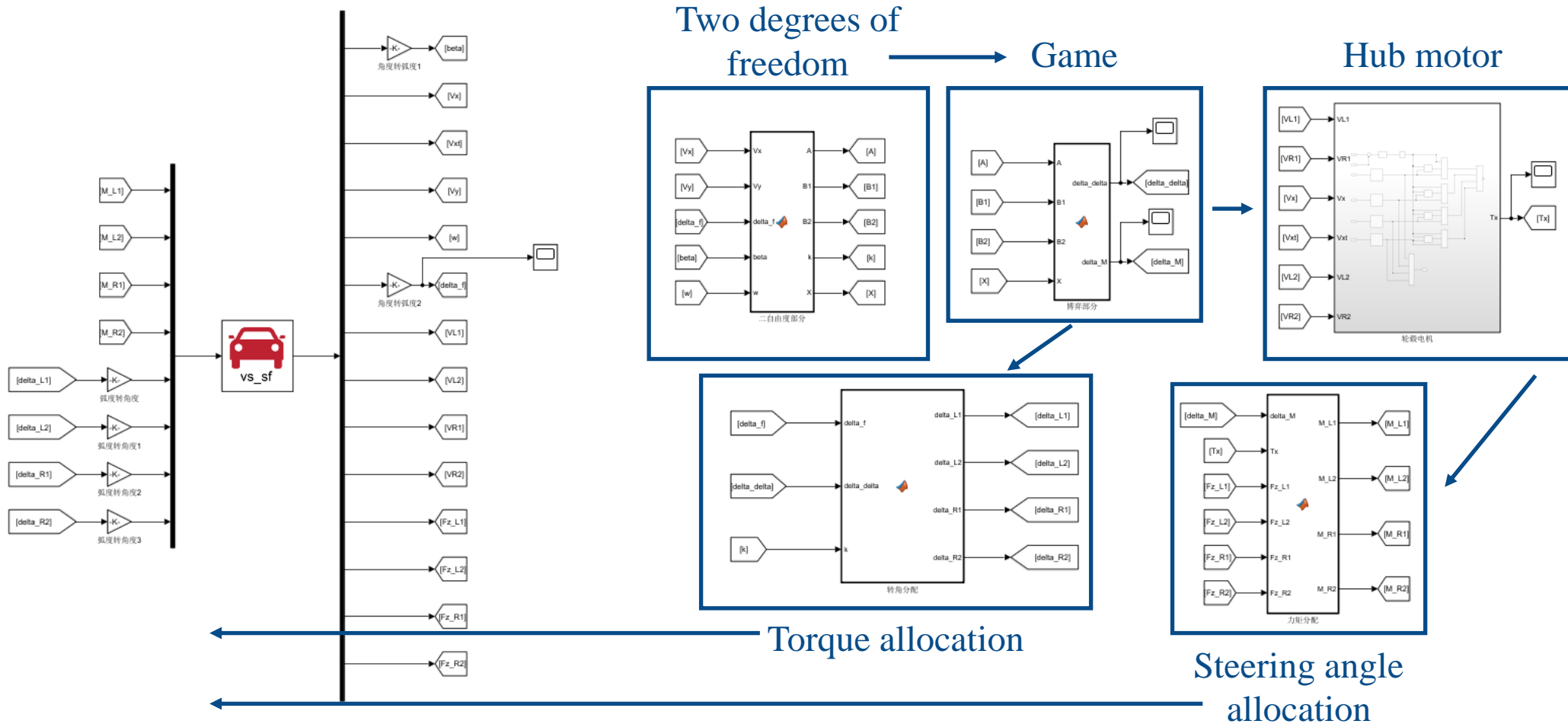
## Lower-level controller design - Quadratic programming

$$\left\{ \begin{array}{l} \min \varepsilon = \sum_j \frac{T_{xj}^2}{(\mu F_{zj})^2 R^2} \\ s.t. \quad T_x = \sum_j T_{xj} \\ \Delta M = \frac{t_r}{2} \left( \frac{T_{xfr}}{R} - \frac{T_{xfl}}{R} \right) + \frac{t_r}{2} \left( \frac{T_{xrl}}{R} - \frac{T_{xrr}}{R} \right) \\ 0 \leq T_{xj} \leq \min(T_{m\_max}, \mu F_{zj} R) \end{array} \right.$$

- A typical quadratic programming problem.
- Solve it using MATLAB toolbox.



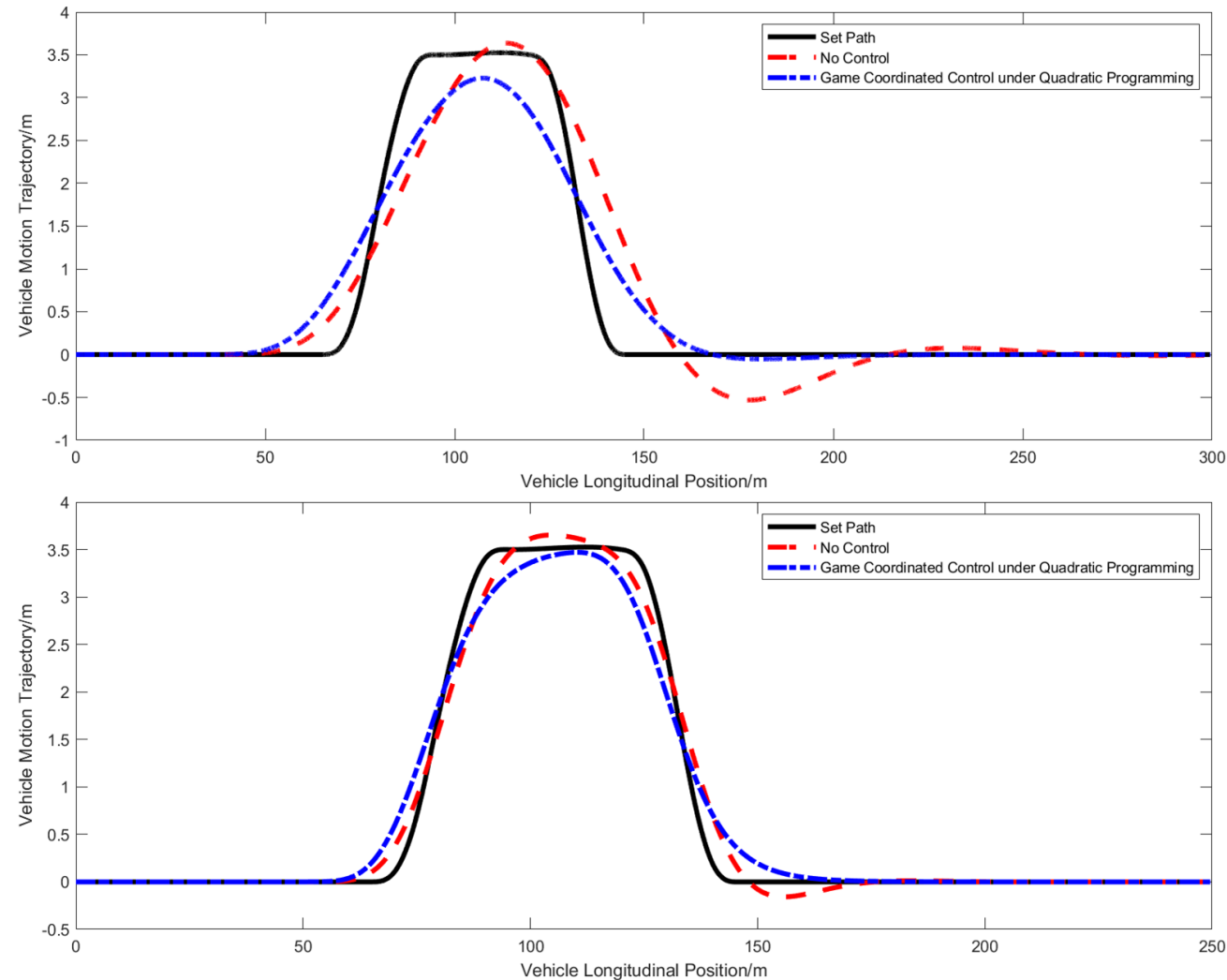
## Building a Simulink model





## Vehicle trajectory tracking performance

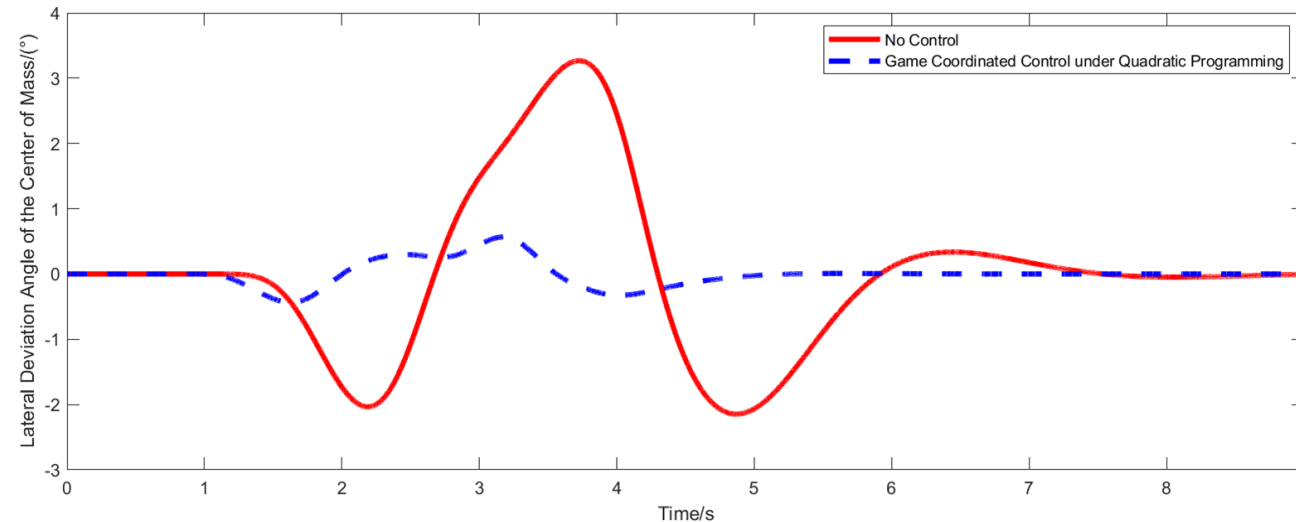
- Vehicle speed: 120 km/h
  - Road surface adhesion coefficient: 0.85
- 
- Vehicle speed: 60 km/h
  - Road surface adhesion coefficient: 0.3



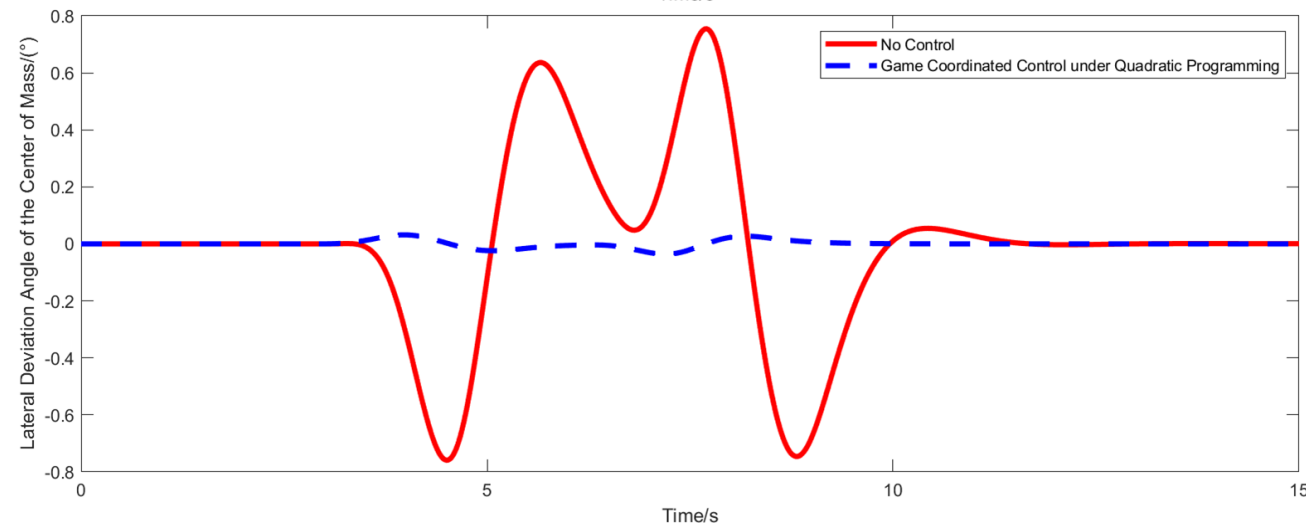


## Simulation results for lateral deviation angle of the center of mass

- Vehicle speed: 120 km/h
- Road surface adhesion coefficient: 0.85



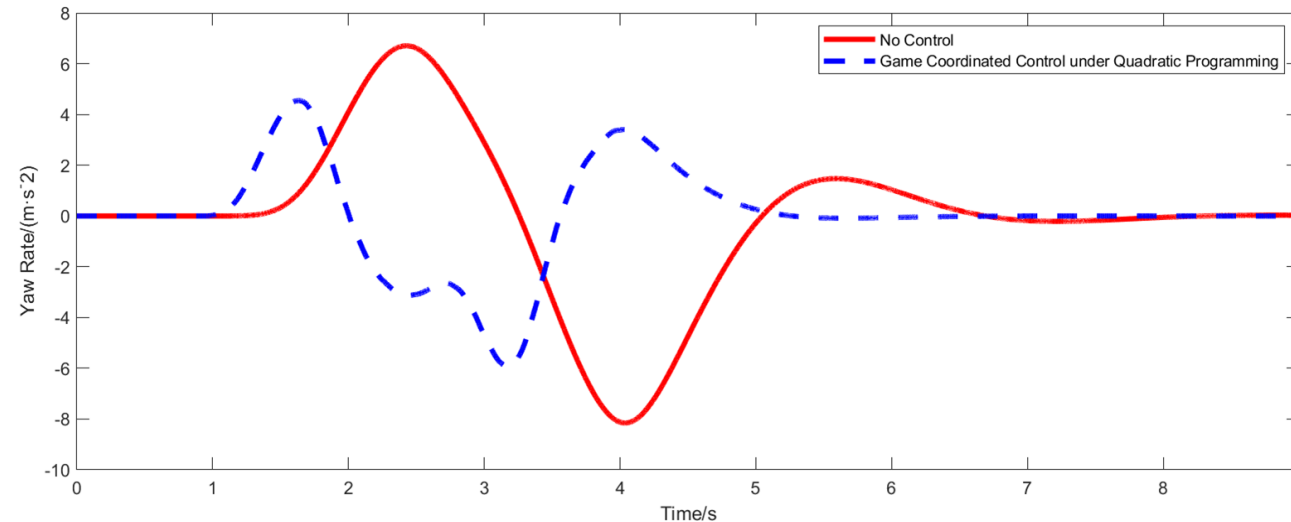
- Vehicle speed: 60 km/h
- Road surface adhesion coefficient: 0.3



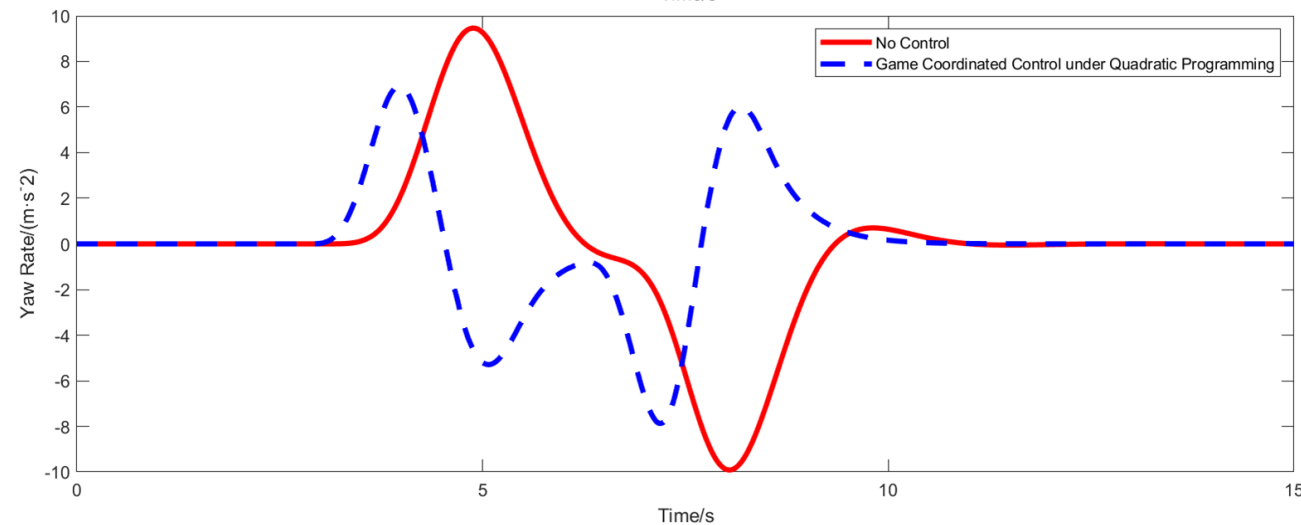


## Simulation results for yaw rate

- Vehicle speed: 120 km/h
- Road surface adhesion coefficient: 0.85



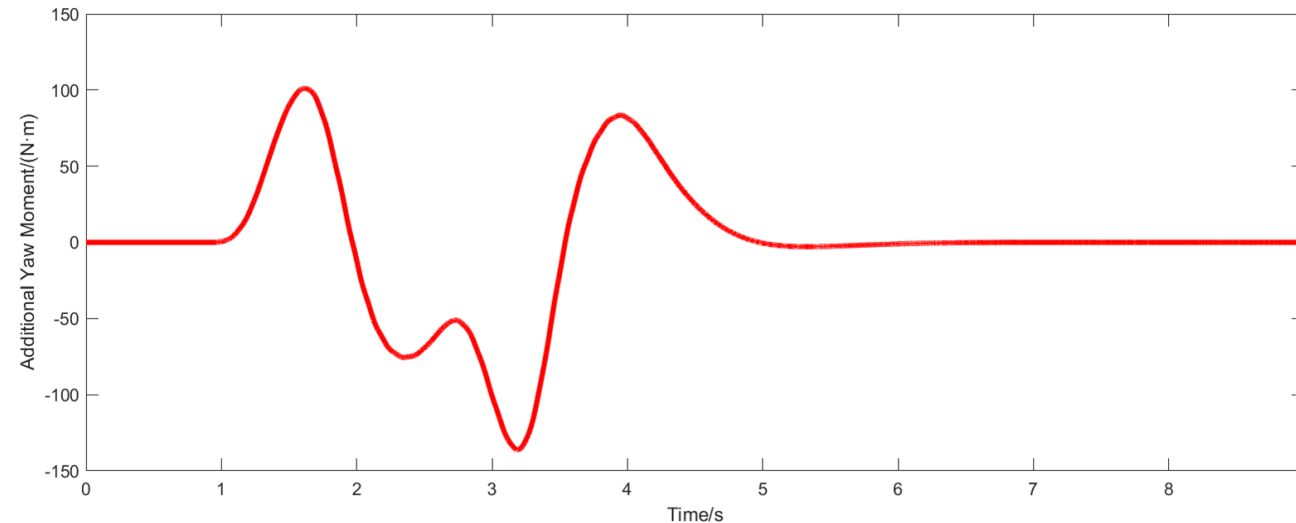
- Vehicle speed: 60 km/h
- Road surface adhesion coefficient: 0.3



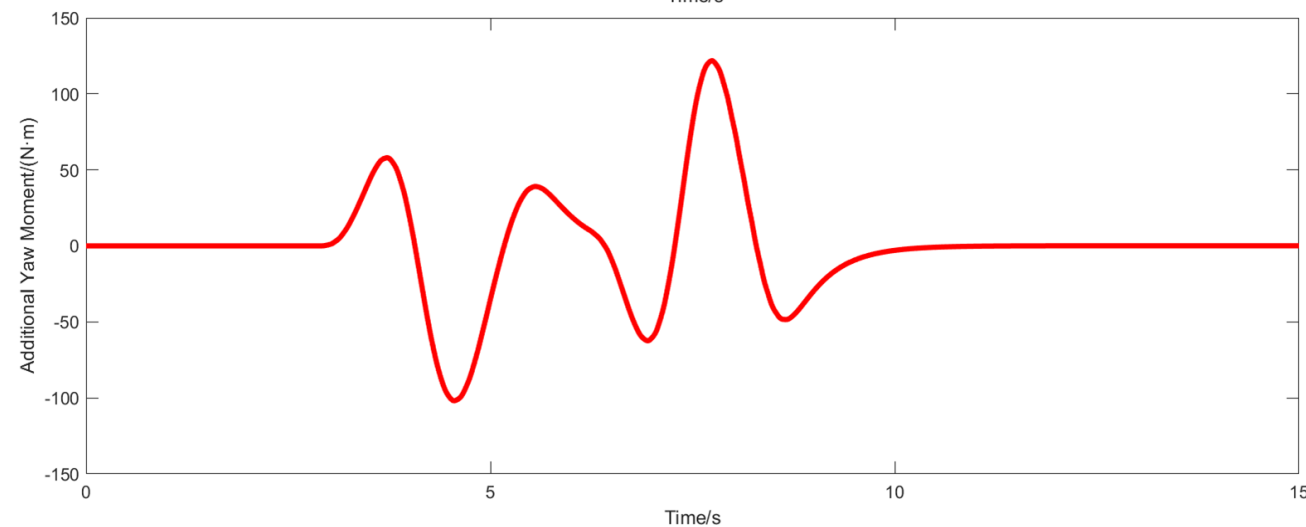


## Simulation results for additional yaw moment

- Vehicle speed: 120 km/h
- Road surface adhesion coefficient: 0.85



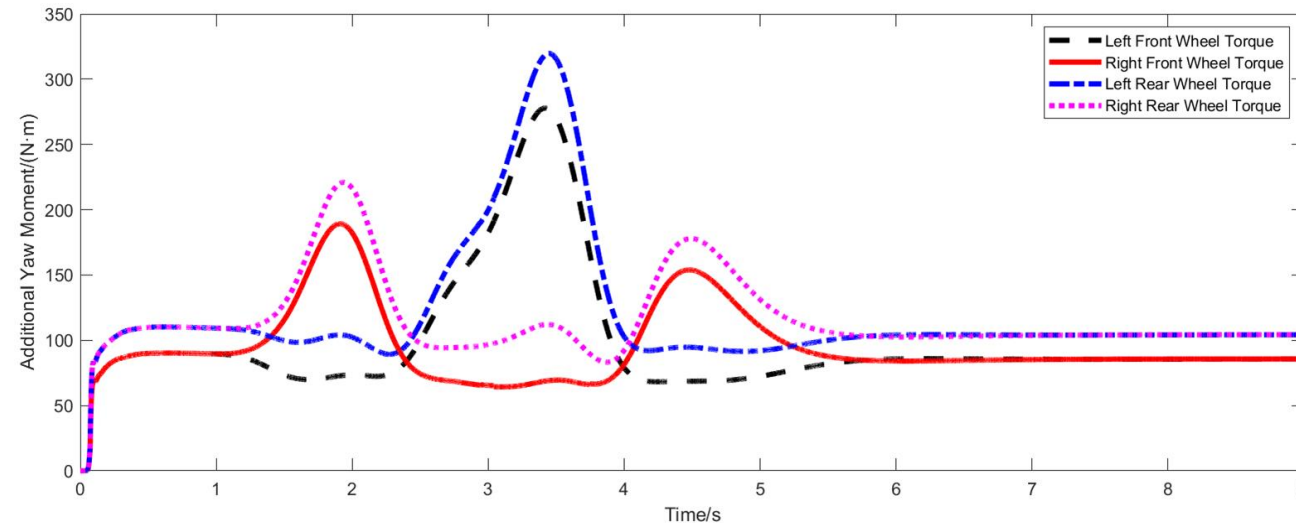
- Vehicle speed: 60 km/h
- Road surface adhesion coefficient: 0.3



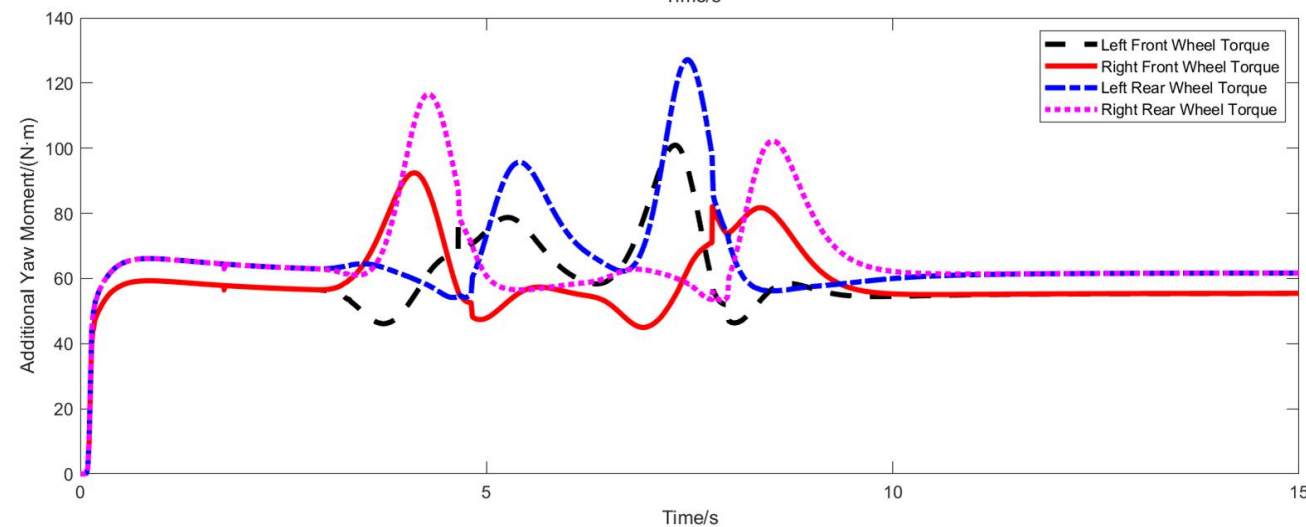


## Simulation results for four-wheel torque

- Vehicle speed: 120 km/h
- Road surface adhesion coefficient: 0.85



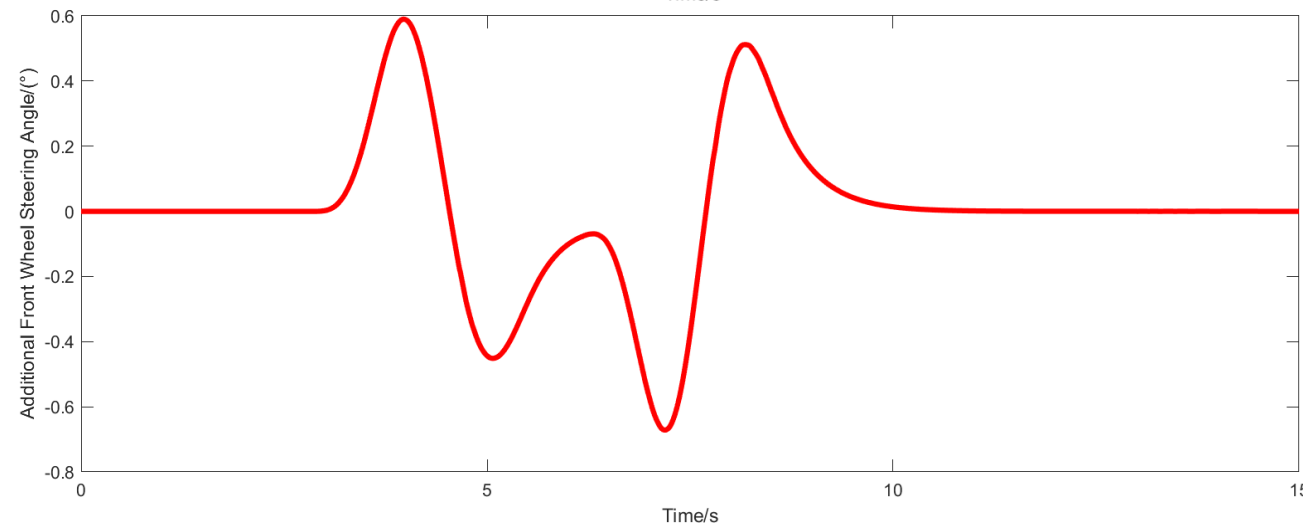
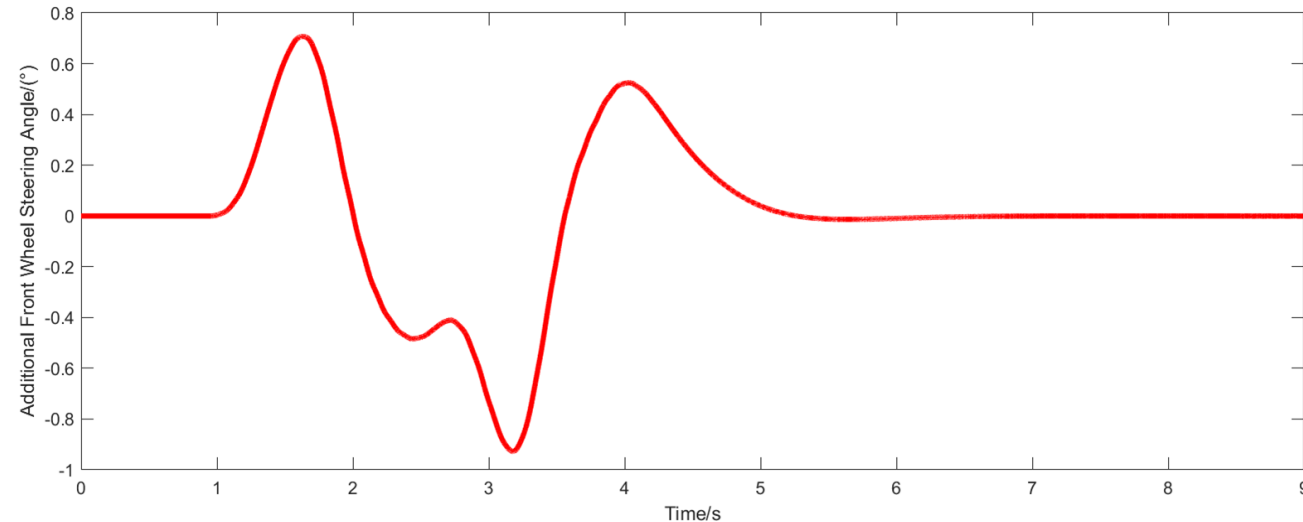
- Vehicle speed: 60 km/h
- Road surface adhesion coefficient: 0.3





## Simulation results for additional front wheel steering angle

- Vehicle speed: 120 km/h
  - Road surface adhesion coefficient: 0.85
- 
- Vehicle speed: 60 km/h
  - Road surface adhesion coefficient: 0.3

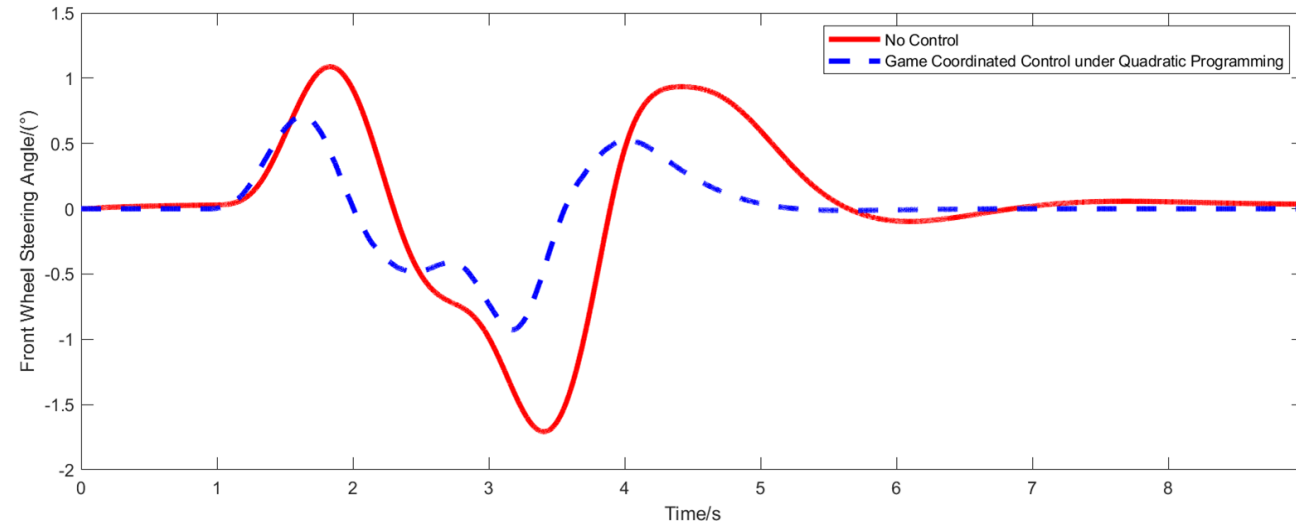






## Simulation results for front wheel steering angle

- Vehicle speed: 120 km/h
- Road surface adhesion coefficient: 0.85



- Vehicle speed: 60 km/h
- Road surface adhesion coefficient: 0.3

