

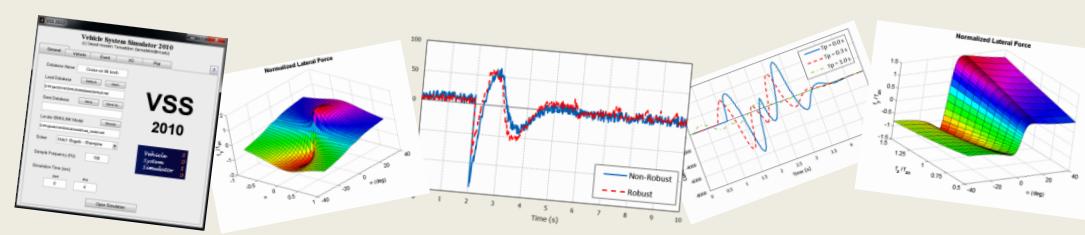


Optimal Vehicle Stability Control with Driver Input and Bounded Uncertainties

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research objectives

- Develop a novel vehicle stability controller that encompasses the driver dynamics
- Develop a vehicle driver-controller interaction model
- Evaluate the effectiveness of the suggested vehicle driver-controller model *
- Evaluate the robustness of the proposed interaction model, and design a robust controller



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* It needs a validated vehicle model that captures the essential dynamics

handling concept

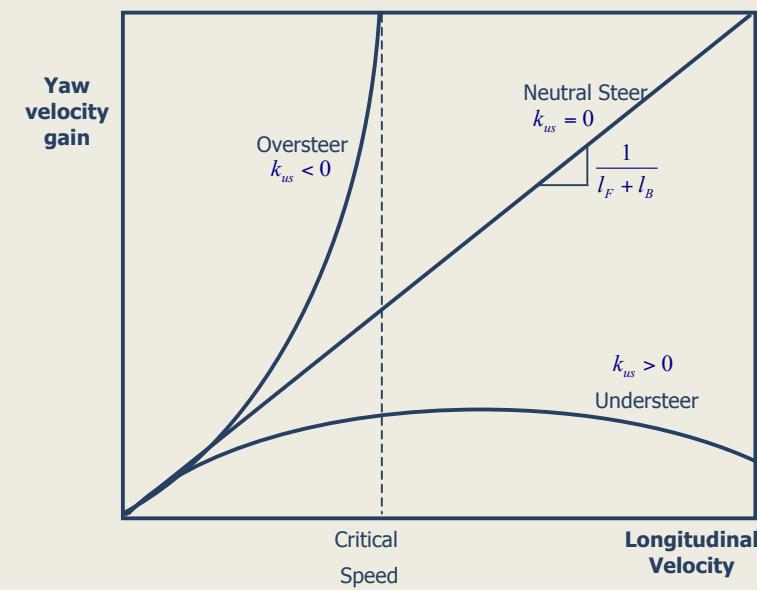
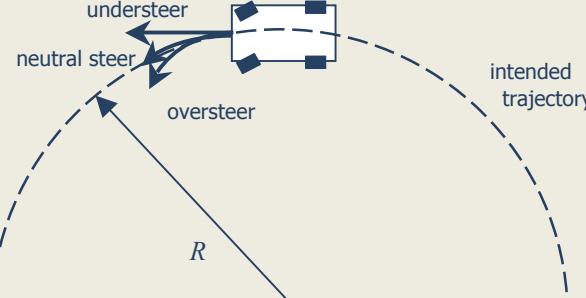
- Cornering: ability of vehicle to travel on curved path

- Yaw velocity gain: $\frac{\dot{\theta}_z}{\delta_F} = \frac{v_x}{(l_F + l_B)(1 + k_{us} v_x^2)}$

- Understeering coefficient: $k_{us} = \frac{m(l_B c_{\alpha B} - l_F c_{\alpha F})}{(l_F + l_B)^2 c_{\alpha F} c_{\alpha B}}$

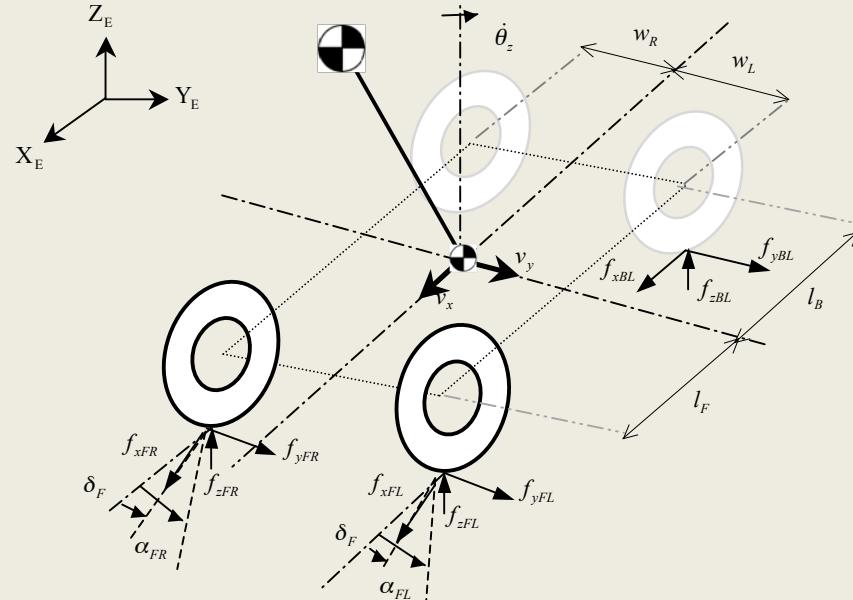
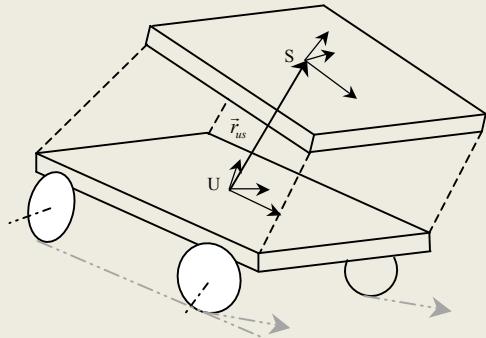


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vehicle dynamics

- **Body dynamics**
- **Wheel/Tire dynamics**



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Longitudinal: $m(\dot{v}_x - v_y \dot{\theta}_z)$ ~~Multibody Equations of Motion~~ = $\sum f_x \sin \delta$

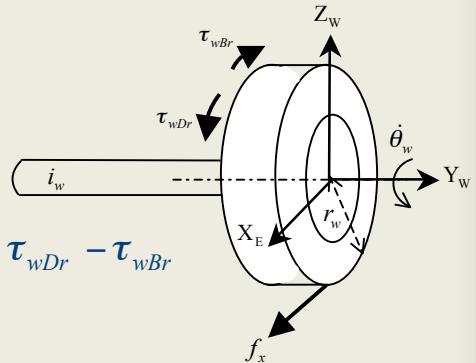
Lateral: $m(\dot{v}_y + v_x \dot{\theta}_z)$ ~~Multibody Equations of Motion~~ = $\sum f_y \cos \delta$

Roll: $m_s h_s (\dot{v}_y + v_x \dot{\theta}_z) + (i_{sx} U + m_s h_s^2) \ddot{\theta}_x + (i_{sz} - i_{sy} - m_s h_s^2) \theta_x \dot{\theta}_z^2 = -k_x \theta_x - c_x \dot{\theta}_x + m_s h_s g \sin \theta_x$

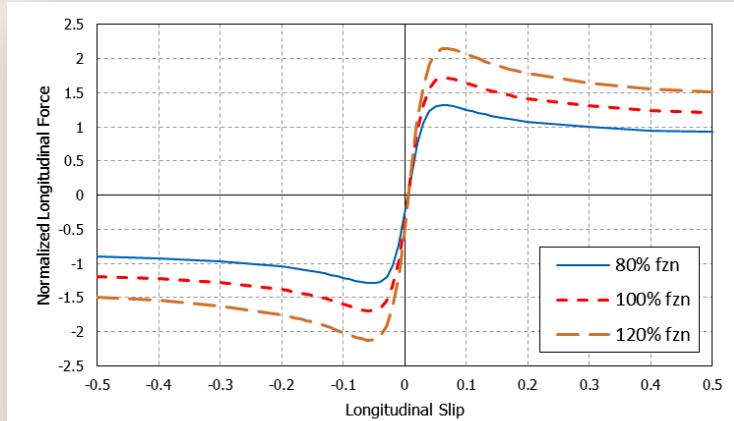
Yaw: $-m_s h_s \theta_x (\dot{v}_x - v_y \dot{\theta}_z) + (i_{uz} + i_{sz}) \dot{\theta}_z = \sum l' f_x \sin \delta - \sum l' f_y \cos \delta + \sum w' f_x \cos \delta - \sum w' f_y \sin \delta$

tire dynamics

- Body dynamics
- **Wheel/Tire dynamics**



$$i_w \ddot{\theta}_w = -f_x r_w + \tau_{wDr} - \tau_{wBr}$$



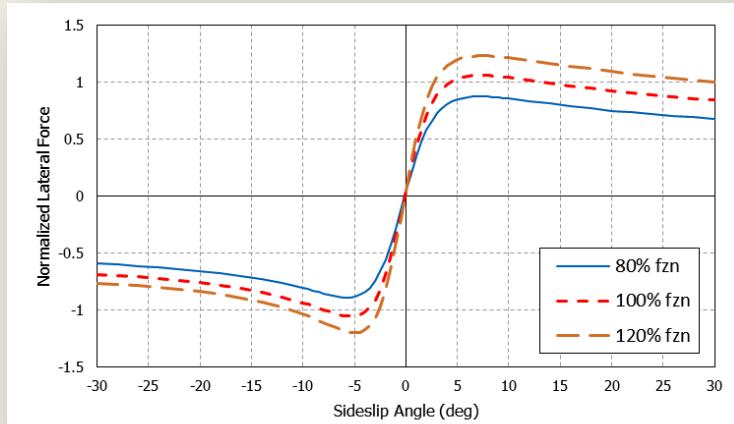
❖ Pacejka “Magic Formula” tire model

$$f = f_0(x, f_z) = D \cdot \sin \left[C \cdot \arctan \left(B \cdot x' - E \cdot (B \cdot x' - \arctan(B \cdot x')) \right) \right] + S_v$$

❖ LuGre tire model

$$f_x = (\sigma_0 \mu_n + \sigma_1 \dot{\mu}_n + \sigma_2 v_r) f_z$$

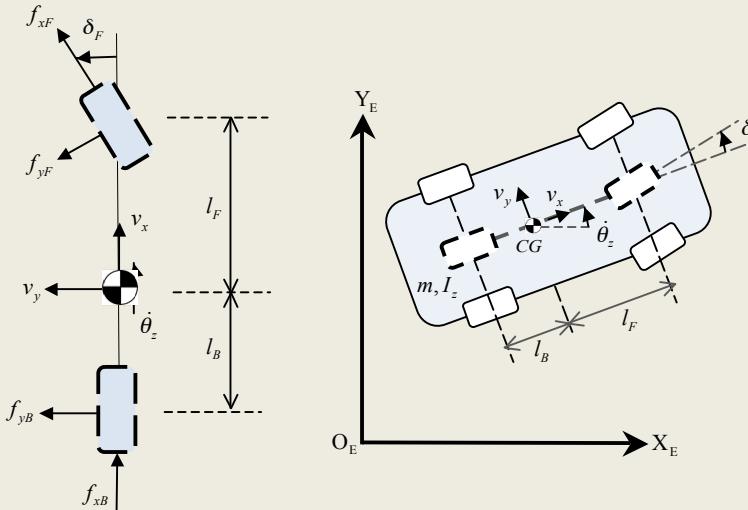
$$\dot{\mu}_n + \left(\frac{\sigma_0 |v_r|}{\mu_c + (\mu_s - \mu_c) e^{-|v_r/v_s|^{\bar{\alpha}}}} + \bar{\kappa} |r_w \dot{\theta}_w| \right) \mu_n = v_r$$



linear 2-DOF vehicle model

- lateral position, lateral velocity, yaw angle, yaw rate

$$x = \begin{pmatrix} y & \dot{y} & \theta_z & \dot{\theta}_z \end{pmatrix}^T$$



$$\dot{x} = \begin{bmatrix} 0 & 1 & v_x & 0 \\ 0 & -\frac{c_{\alpha F} + c_{\alpha B}}{mv_x} & 0 & -v_x - \frac{l_F c_{\alpha F} - l_B c_{\alpha B}}{mv_x} \\ 0 & 0 & 0 & 1 \\ 0 & \frac{l_F c_{\alpha F} - l_B c_{\alpha B}}{i_z v_x} & 0 & -\frac{l_F^2 c_{\alpha F} + l_B^2 c_{\alpha B}}{i_z v_x} \end{bmatrix} x + \begin{bmatrix} 0 \\ \frac{c_{\alpha F}}{r_{st} m} \\ 0 \\ \frac{l_F c_{\alpha F}}{r_{st} i_z} \end{bmatrix} \delta_{sw}$$

traction/braking control systems

- Cruise Control
- Differential Braking Control



$$\tau_{wDr} = \begin{cases} \frac{1}{2} L_{PID} (v_{x,des} - v_x) \cdot [0 \ 0 \ 1 \ 1]^T & , v_{xd} > v_x \\ 0_{1 \times 4} & , v_{xd} \leq v_x \end{cases}$$

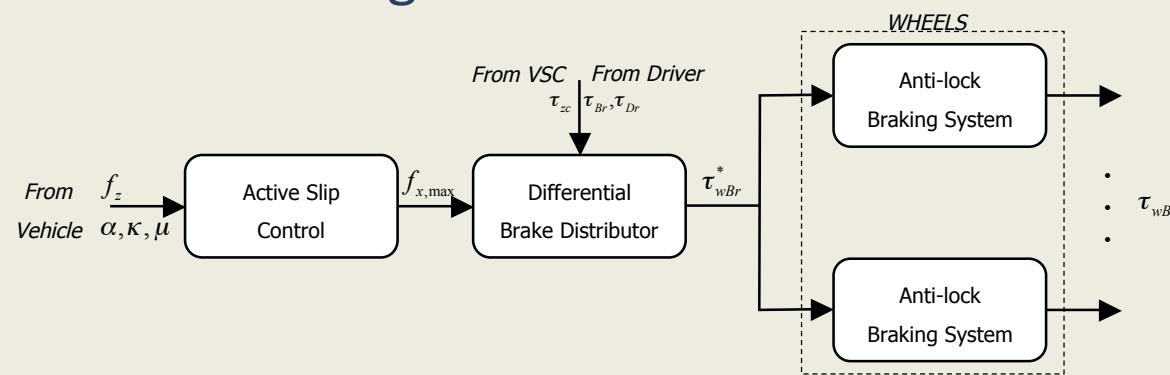
$$L_{PID} = k_{pcc} + k_{icc} \int_0^t dt + k_{dcc} \frac{d}{dt}$$

$$ASC = \begin{cases} 0 & , v_x \leq 0.7v_{avg} \\ \frac{1}{0.15} \left(\frac{v_x}{v_{avg}} - 0.7 \right) & , 0.7v_{avg} < v_x < 0.85v_{avg} \\ 1 & , v_x \geq 0.85v_{avg} \end{cases}$$

$$v_{avg} = \left(\frac{r_w \dot{\theta}_{wBR} + r_w \dot{\theta}_{wBL}}{2} \right)$$

traction/braking control systems

- Cruise Control
- Differential Braking Control



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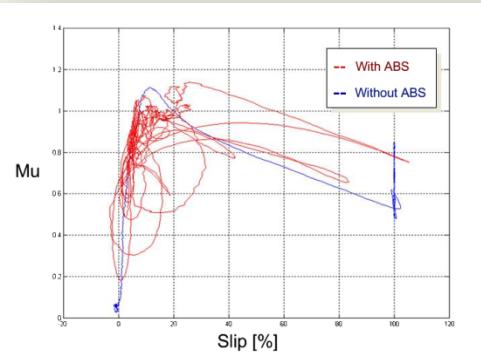
$$f_{x,\max} = \begin{cases} 0 & , \kappa \geq 0.8 \\ \mu f_z \left(\frac{\kappa}{\sqrt{\kappa^2 + \tan^2(\alpha)}} \right) & , \kappa < 0.8 \end{cases}$$

$$\tau_{wBr,i}^* = \max(r_w f_{x,i} + \tau_{wDr,i}, \tau_{wBr,max})$$

$$\begin{cases} f_{xFR} + f_{xFL} + f_{xBR} + f_{xBL} = \frac{\tau_{accel} - \tau_{decel}}{r_w} \\ f_{xFR} - f_{xFL} + f_{xBR} - f_{xBL} = \frac{\tau_{zc}}{w} \\ |f_{xFR}| \leq f_{xFR,max} \\ |f_{xFL}| \leq f_{xFL,max} \\ |f_{xBR}| \leq f_{xBR,max} \\ |f_{xBL}| \leq f_{xBL,max} \end{cases}$$

anti-lock braking system

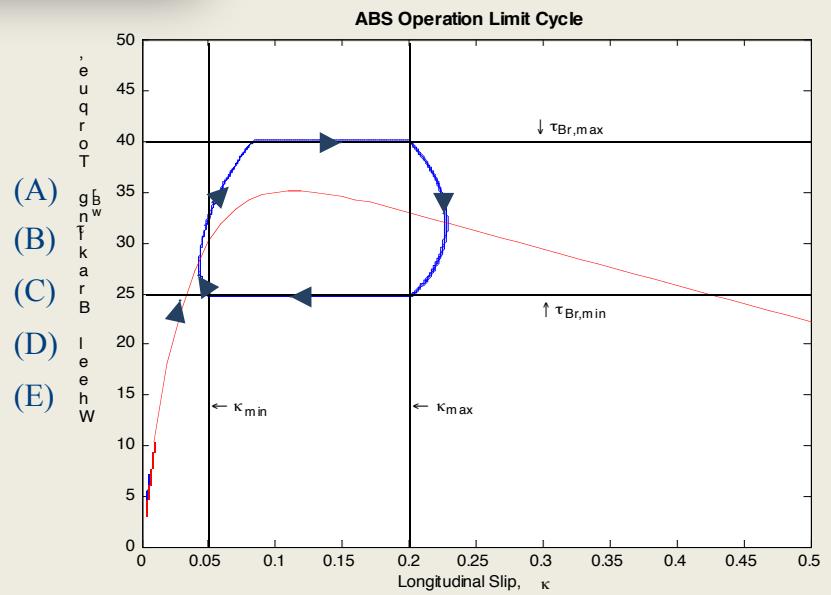
- prevents wheel lock-up during braking
- maintains vehicle stability and steering



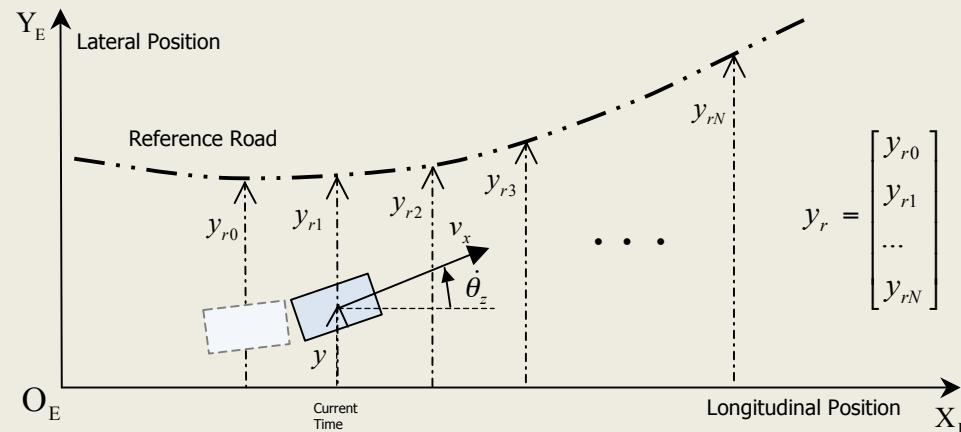
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$$\begin{cases} \tau_{wBr} = \tau_{wBr}^*, & \kappa < \kappa_{\min} \\ \dot{\tau}_{wBr} = k_b, & \kappa_{\min} < \kappa \leq \kappa_{\max} \text{ } \& \tau_{wBr,\min} \leq \tau_{wBr} \leq \tau_{Br,\max} \\ \dot{\tau}_{wBr} = 0, & \kappa_{\min} < \kappa \leq \kappa_{\max} \text{ } \& \tau_{wBr} \geq \tau_{Br,\max} \\ \dot{\tau}_{wBr} = 0, & \kappa_{\min} < \kappa \leq \kappa_{\max} \text{ } \& \tau_{wBr} \leq \tau_{Br,\min} \\ \dot{\tau}_{wBr} = -k_b, & \kappa > \kappa_{\max} \end{cases}$$



driver steering control model



Linear Bicycle Model :

$$x_d^+ = \mathbf{A}_d x_d + \mathbf{B}_d \delta_{sw}$$

Road Preview Model :

$$y_r^+ = \underbrace{\begin{bmatrix} \mathbf{0}_{N \times 1} & \mathbf{I}_{N \times N} \\ \mathbf{0} & \mathbf{0}_{1 \times N} \end{bmatrix}}_{\mathbf{A}_r} y_r + \underbrace{\begin{bmatrix} \mathbf{0}_{N \times 1} \\ 1 \end{bmatrix}}_{\mathbf{B}_r} y_{r(N+1)}$$

$$\underbrace{\begin{bmatrix} x_d^+ \\ y_r^+ \end{bmatrix}}_{z^+} = \underbrace{\begin{bmatrix} \mathbf{A}_d & \mathbf{0} \\ \mathbf{0} & \mathbf{A}_r \end{bmatrix}}_A \underbrace{\begin{bmatrix} x_d \\ y_r \end{bmatrix}}_z + \underbrace{\begin{bmatrix} \mathbf{B}_d \\ \mathbf{0} \end{bmatrix}}_B \delta_{sw} + \underbrace{\begin{bmatrix} \mathbf{0} \\ \mathbf{B}_r \end{bmatrix}}_E y_{r(N+1)}$$



driver–controller interactions

- vehicle stability =
driver decisions + controller performance
- Interaction modeling of driver-controller:
 - independent subsystems
 - collaborative subsystems
- In this research, the interaction modeling of **driver steering control** and **vehicle yaw control** is studied based on “Linear Quadratic Game Theory” approach.



continuous-time interaction model

- Linear bicycle model:

$$\dot{x}(t) = \mathbf{A}_c x(t) + \mathbf{B}_{c1} \underbrace{\delta_{sw}(t)}_{u_1} + \mathbf{B}_{c2} \underbrace{M_{zc}(t)}_{u_2}$$
$$x(t) = [y \quad v_y \quad \theta_z \quad \dot{\theta}_z]^T$$

$$\mathbf{A}_c = \begin{bmatrix} 0 & 1 & V_x & 0 \\ 0 & -\frac{C_{\alpha F} + C_{\alpha B}}{MV_x} & 0 & -V_x - \frac{L_F C_{\alpha F} - L_B C_{\alpha B}}{MV_x} \\ 0 & 0 & 0 & 1 \\ 0 & \frac{L_F C_{\alpha F} - L_B C_{\alpha B}}{I_z V_x} & 0 & -\frac{L_F^2 C_{\alpha F} + L_R^2 C_{\alpha B}}{I_z V_x} \end{bmatrix}, \mathbf{B}_{c1} = \begin{bmatrix} 0 \\ \frac{C_{\alpha F}}{r_{st} M} \\ 0 \\ \frac{L_F C_{\alpha F}}{r_{st} I_z} \end{bmatrix}, \mathbf{B}_{c2} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ \frac{1}{I_z} \end{bmatrix}$$

- Objective function:

$$J_i(u_1, u_2) = \int_0^t (x^T \mathbf{Q}_i x + u_1^T \mathbf{R}_{i1} u_1 + u_2^T \mathbf{R}_{i2} u_2) dt$$

- Nash equilibrium:

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



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LQ Game Theory

- Forming “Hamiltonian” as

$$H_i(x, u_1, u_2, P_i) = x^T Q_i x + u_1^T R_{i1} u_1 + u_2^T R_{i2} u_2 + P_i^T (A_c x + B_{c1} u_1 + B_{c2} u_2)$$

subject to

$$\frac{\partial}{\partial u_i} H_i(x^*, u_1^*, u_2^*, P_i) = 0 \quad , \quad \frac{d}{dt} P_i = -\frac{\partial}{\partial x} H_i(x^*, u_1^*, u_2^*, P_i)$$

where

$$\dot{x}^*(t) = A_c x^*(t) + B_{c1} u_1^* + B_{c2} u_2^*$$

leads to

general form

$$u_i^* = -R_{ii}^{-1} B_{ci}^T P_i(t)$$
$$\frac{d}{dt} P_i = -\frac{\partial}{\partial x} H_i(x^*, u_1^*, u_2^*, P_i) - \frac{\partial}{\partial u_i} H_i(x^*, u_1^*, u_2^*, P_i) \cdot \frac{\partial u_j^*}{\partial x} \quad , \quad (i=1,2)$$

Linear feedback form

$$u_i^*(t) = -R_{ii}^{-1} B_{ci}^T K_i x(t)$$
$$A_c^T K_i + K_i A_c + Q_i - K_i S_i K_i - K_i S_{\hat{i}} K_{\hat{i}} - K_{\hat{i}} S_{\hat{i}} K_i + K_{\hat{i}} S_{\hat{i}} K_{\hat{i}} = 0_n$$
$$where \quad S_i = B_{ci} R_{ii}^{-1} B_{ci}^T \quad , \quad S_{\hat{i}} = B_{c\hat{i}} R_{\hat{i}\hat{i}}^{-1} R_{\hat{i}\hat{i}}^T B_{c\hat{i}}^T$$



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simulation

- Vehicle: same sedan vehicle
- Maneuver: lane-change of $4m$ at 20 m/s
- Control Objective: improved handling performance

$$\dot{\theta}_{z,des} = \frac{V_x}{r_{st}(L_F + L_B)(1 + K_{us}V_x^2)} \delta_{sw}$$

- Driver LQ structure:

Driver: $\mathbf{Q}_1 = \begin{bmatrix} 10 & 0 & 0 & 0 \\ 0 & 0.01 & 0 & 0 \\ 0 & 0 & 0.1 & 0 \\ 0 & 0 & 0 & 0.01 \end{bmatrix}, \mathbf{R}_{11} = 1, \mathbf{R}_{12} = 0$

- Vehicle controller LQ structure:

DYC: $\mathbf{Q}_2 = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0.1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}, \mathbf{R}_{21} = 10, \mathbf{R}_{22} = 10^{-7}$



results

Linear feedback
form

$$u_i^*(t) = -\mathbf{R}_{ii}^{-1} \mathbf{B}_{ci}^T \mathbf{K}_i x(t) = \mathbf{G}_i x(t)$$
$$\mathbf{A}_c^T \mathbf{K}_i + \mathbf{K}_i \mathbf{A}_c + \mathbf{Q}_i - \mathbf{K}_i \mathbf{S}_i \mathbf{K}_i - \mathbf{K}_i \mathbf{S}_{\hat{i}} \mathbf{K}_{\hat{i}} - \mathbf{K}_{\hat{i}} \mathbf{S}_{\hat{i}} \mathbf{K}_i + \mathbf{K}_{\hat{i}} \mathbf{S}_{\hat{i}\hat{i}} \mathbf{K}_{\hat{i}} = \mathbf{0}_n$$

where $\mathbf{S}_i = \mathbf{B}_{ci} \mathbf{R}_{ii}^{-1} \mathbf{B}_{ci}^T$, $\mathbf{S}_{\hat{i}\hat{i}} = \mathbf{B}_{c\hat{i}} \mathbf{R}_{\hat{i}\hat{i}}^{-1} \mathbf{R}_{\hat{i}\hat{i}} \mathbf{R}_{\hat{i}\hat{i}}^{-1} \mathbf{B}_{c\hat{i}}^T$

- Game Theory:



$$\begin{cases} \mathbf{G}_{1(GT)} = [-0.809 \quad -0.146 \quad -8.624 \quad -0.713] \\ \mathbf{G}_{2(GT)} = [\quad 0 \quad 504.08 \quad 0 \quad -10696] \end{cases}$$

- Optimal LQ Control:

$$\begin{cases} \mathbf{G}_{1(LQR)} = [-1.0 \quad -0.215 \quad -13.63 \quad -1.286] \\ \mathbf{G}_{2(LQR)} = [\quad 0 \quad 129.93 \quad 0 \quad -599] \end{cases}$$

- Scenarios:

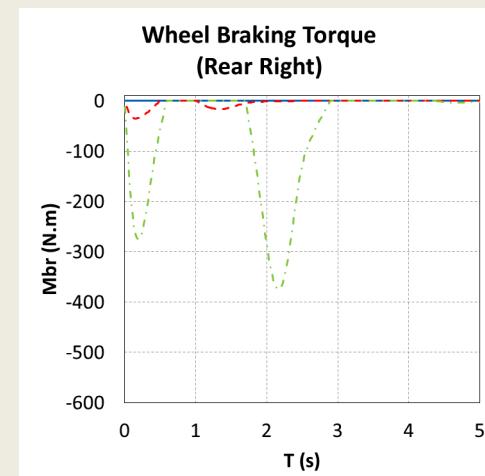
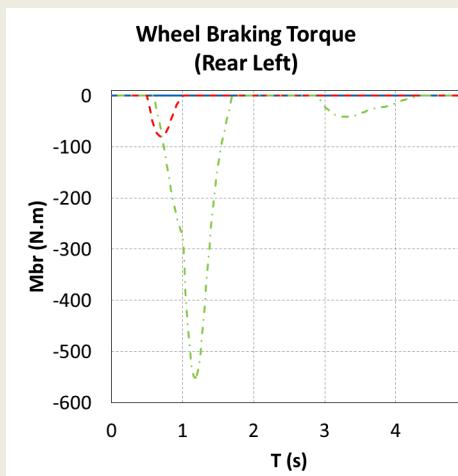
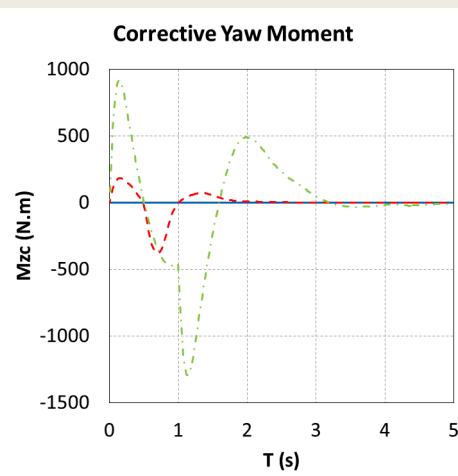
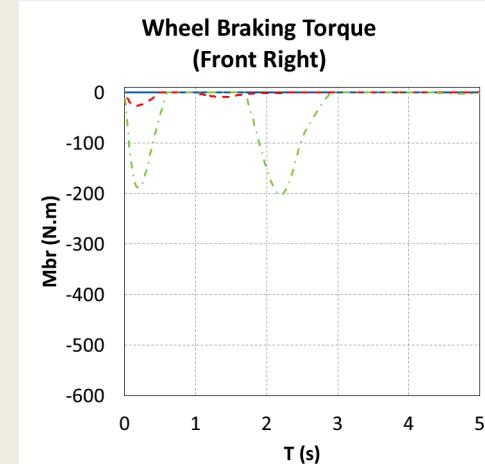
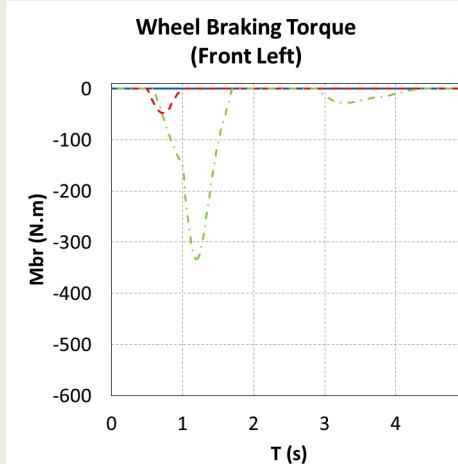
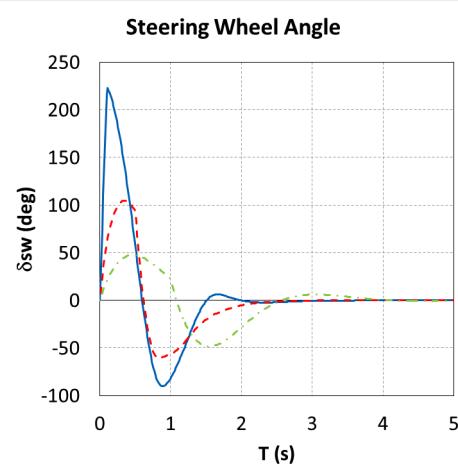
- (A) driver: LQR & vehicle controller: turned off
- (B) driver: LQR & vehicle controller: LQR
- (C) driver: GT & vehicle controller: GT



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results: continuous-time model



— A - - - B - · - C



discrete-time interaction model

- Linear vehicle model with road preview

$$\begin{bmatrix} x_d^+ \\ y_r^+ \end{bmatrix} = \underbrace{\begin{bmatrix} \mathbf{A}_d & 0 \\ 0 & \mathbf{A}_r \end{bmatrix}}_A \begin{bmatrix} x_d \\ y_r \end{bmatrix} + \underbrace{\begin{bmatrix} \mathbf{B}_{d1} \\ 0 \end{bmatrix}}_{\mathbf{B}_1} \delta_{sw} + \underbrace{\begin{bmatrix} \mathbf{B}_{d2} \\ 0 \end{bmatrix}}_{\mathbf{B}_2} M_{zc} + \underbrace{\begin{bmatrix} 0 \\ \mathbf{B}_r \end{bmatrix}}_{\mathbf{E}} y_{r(N+1)}$$

- Objective function

$$J_i = \frac{1}{2} \sum_{k=0}^{\infty} \left(z^T \mathbf{Q}_i z + \sum_{j=1}^2 u_j^T \mathbf{R}_{ij} u_j \right)$$

where

$$\mathbf{Q}_i = \mathbf{N}_i^T \mathbf{Q}_{di} \mathbf{N}_i$$

$$\mathbf{N}_1 = \left[\begin{array}{cccc|ccccc} 1 & 0 & 0 & 0 & 0 & -1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & 0 & 0 & 1/T_s & -1/T_s & 0 & \dots & 0 \\ 0 & 0 & 1 & 0 & 0 & 1/(V_x T_s) & -1/(V_x T_s) & 0 & \dots & 0 \\ 0 & 0 & 0 & 1 & -1/(V_x T_s^2) & 2/(V_x T_s^2) & -1/(V_x T_s^2) & 0 & \dots & 0 \end{array} \right]_{4 \times (4+N)}$$

state gains *road preview gain*

$$\begin{cases} Y_{ref} = Y_{r1} \\ V_{y,ref} \approx \frac{\Delta Y}{T_s} = \frac{Y_{r2} - Y_{r1}}{T_s} \\ \theta_{z,ref} \approx \frac{\Delta Y}{\Delta X} \approx \frac{Y_{r2} - Y_{r1}}{V_x T_s} \\ \dot{\theta}_{z,ref} \approx \frac{\Delta \theta_{z,ref}}{T_s} \approx \frac{\theta_{z,ref}^+ - \theta_{z,ref}^-}{T_s} = \frac{Y_{r2} - 2Y_{r1} + Y_{r0}}{V_x T_s^2} \end{cases}$$

$$\mathbf{N}_2 = \left[\begin{array}{cccc|ccccc} 1 & 0 & 0 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 1 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & 1 & 0 & \dots & 0 \end{array} \right]_{4 \times (4+N)}$$

state gains *road preview gain*



discrete LQ Game Theory

- Suppose \mathbf{P}_i satisfy the coupled Riccati equations for discrete linear quadratic games given by

$$\mathbf{P}_i = \mathbf{Q}_i - \mathbf{G}_1^T \mathbf{R}_{\hat{i}1} \mathbf{G}_1 - \mathbf{G}_2^T \mathbf{R}_{i2} \mathbf{G}_2 + (\mathbf{A} + \mathbf{B}_1 \mathbf{G}_1 + \mathbf{B}_2 \mathbf{G}_2)^T \mathbf{P}_i^+ (\mathbf{A} + \mathbf{B}_1 \mathbf{G}_1 + \mathbf{B}_2 \mathbf{G}_2)$$

where $i = (1, 2)$, and \hat{i} is the counter-coalition, i.e. the player counter-acting to the player with index i , and

$$\mathbf{G}_i = -(\mathbf{R}_{ii} + \mathbf{B}_i^T \mathbf{P}_i^+ \mathbf{B}_i)^{-1} \mathbf{B}_i^T \mathbf{P}_i^+ (\mathbf{A} + \mathbf{B}_{\hat{i}} \mathbf{G}_{\hat{i}})$$

then the following strategy

$$u_i^* = -\mathbf{R}_{ii}^{-1} \mathbf{B}_i^T \mathbf{P}_i^+ z^+ \quad (i = 1, 2)$$

is linear feedback “Nash equilibrium”.

simulation

- Vehicle: same sedan vehicle
 - Maneuver: lane-change of $4m$ at 20 (m/s)
 - Control Objective: improved handling performance
 - Driver/ Vehicle controller LQ structure: same
 - Sampling Frequency: 100 Hz
 - Integration Frequency: 10,000 Hz
-
- Scenarios:
 - 1: no preview time ($T_p = 0s$)
 - 2: short preview time ($T_p = 0.3s$)
 - 3: long preview time ($T_p = 1s$)



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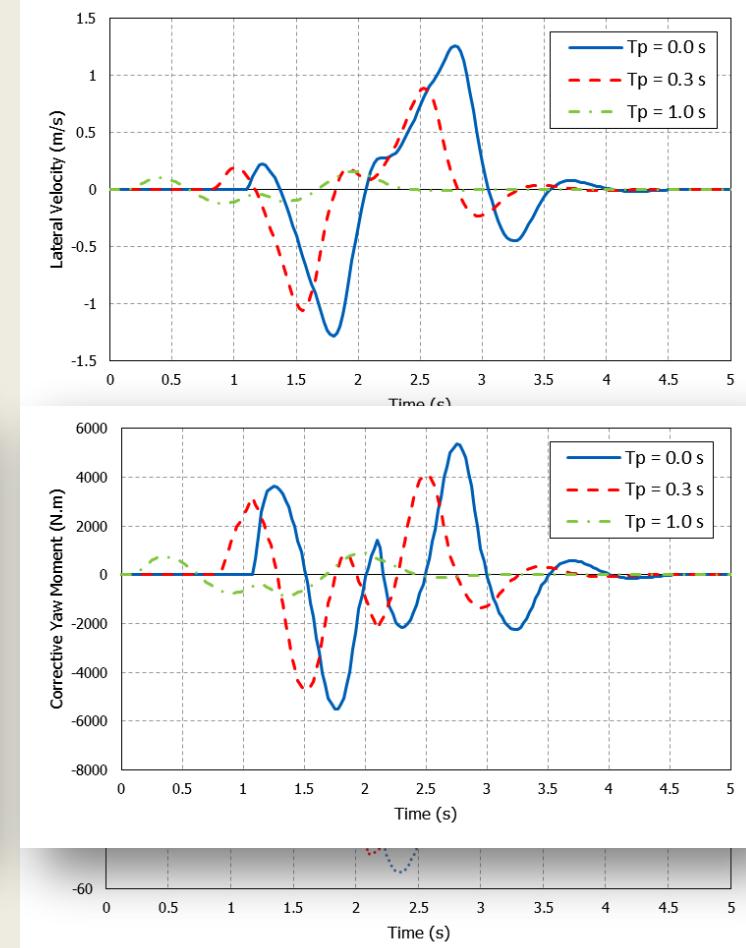
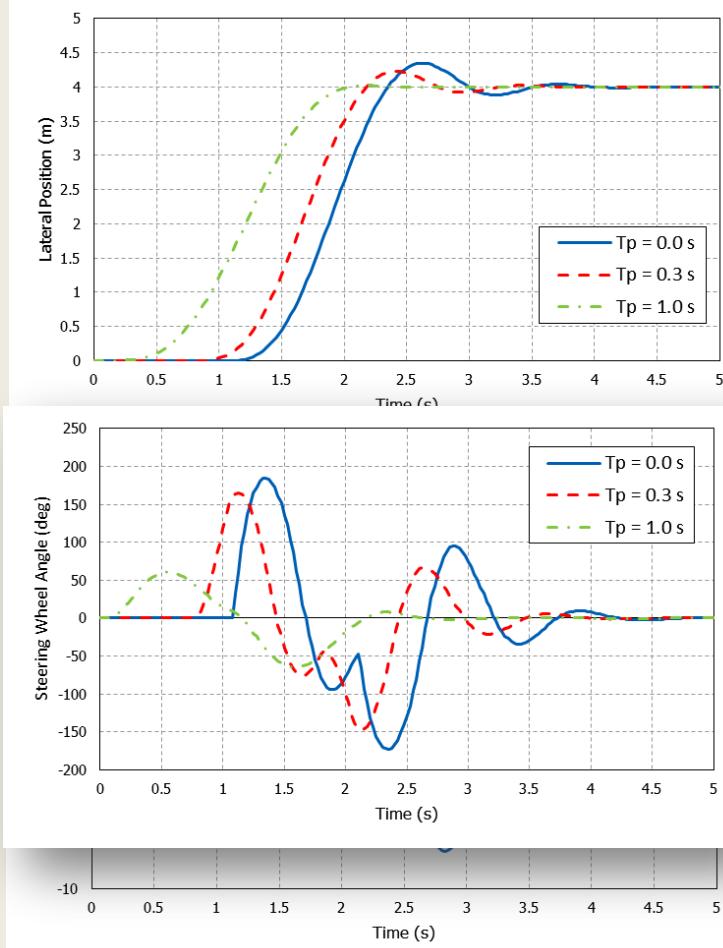
results: discrete-time model



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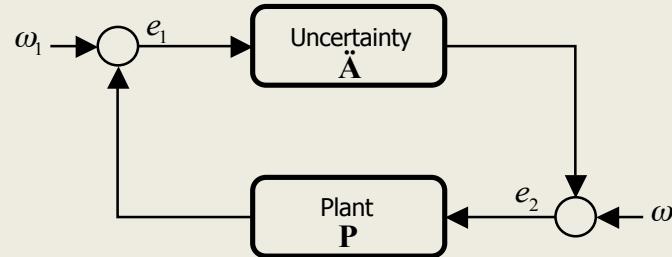
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μ -sensitivity analysis

Small Gain Theorem [Zames, 1966]

Consider the feedback interconnection:



Suppose P is the plant and let $\gamma > 0$. Then this feedback interconnection is internally stable for all unstructured uncertainty Δ with $\|\ddot{\Delta}\|_\infty \leq 1/\gamma$ if and only if $\|P\|_\infty < \gamma$.

Extension [Packard, 1993]

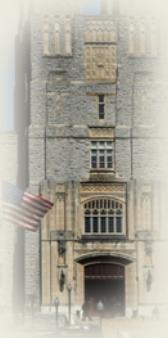
For the same feedback interconnection, assuming that $\gamma > 0$, this feedback interconnection is internally stable for all structured Δ with $\|\ddot{\Delta}\|_\infty \leq 1/\gamma$ if and only if $\sup_{\delta \in \ddot{\Delta}} \mu_{\ddot{\Delta}}(P) < \gamma$.

* The structured singular value of P with respect to uncertainty structure Δ is defined as

$$\mu_{\ddot{\Delta}}(P) = \frac{1}{\min(\bar{\sigma}(\ddot{\Delta}) : |\mathbf{I} - P\ddot{\Delta}| = 0)}, \text{ the maximum singular value of } P\ddot{\Delta}$$

sensitivity analysis of driver gains

- Effect of uncertainty in lateral position gain



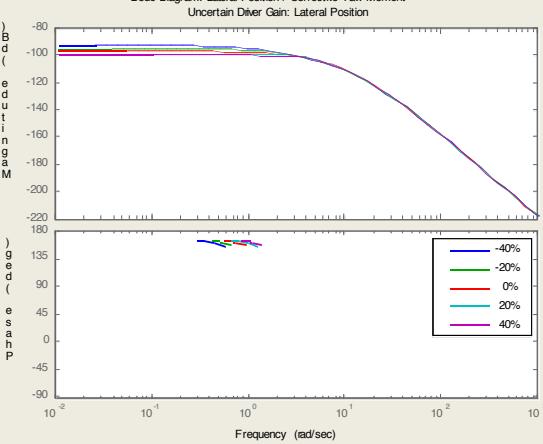
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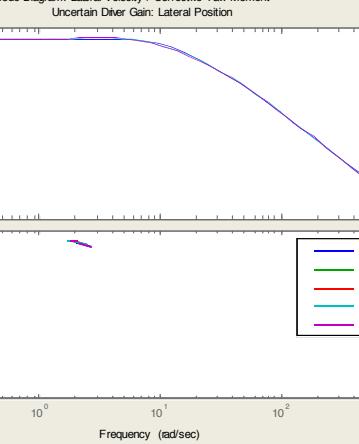
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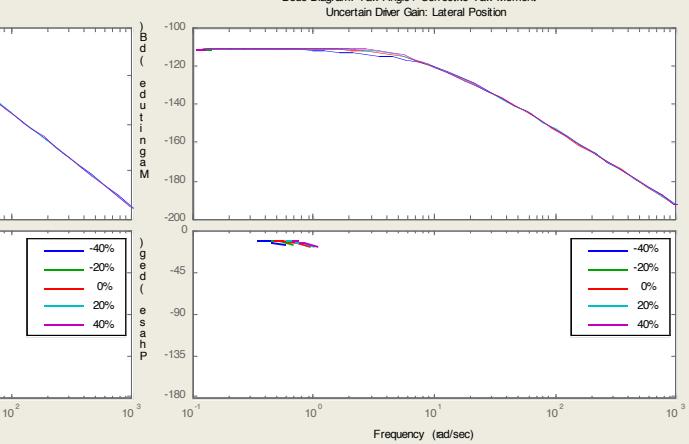
Bode Diagram: Lateral Position / Corrective Yaw Moment
Uncertain Driver Gain: Lateral Position



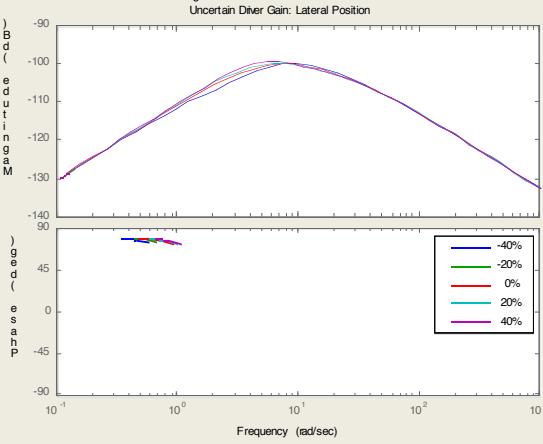
Bode Diagram: Lateral Velocity / Corrective Yaw Moment
Uncertain Driver Gain: Lateral Position



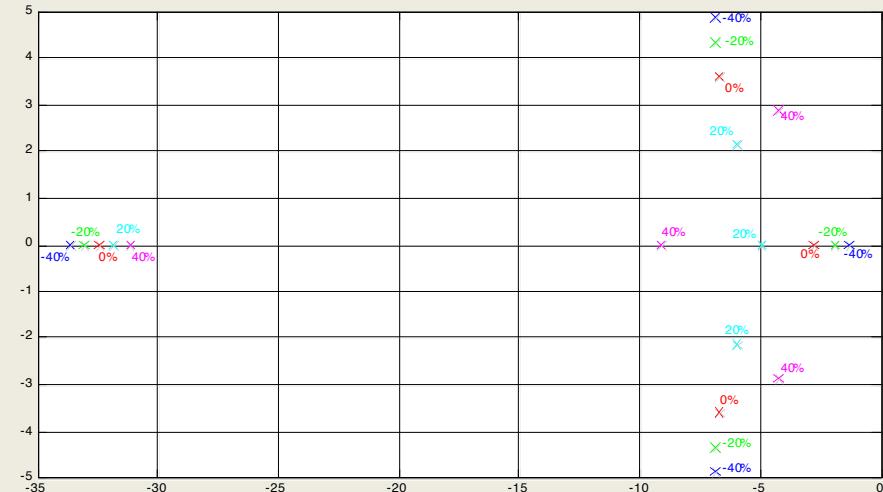
Bode Diagram: Yaw Angle / Corrective Yaw Moment
Uncertain Driver Gain: Lateral Position



Bode Diagram: Yaw Rate / Corrective Yaw Moment
Uncertain Driver Gain: Lateral Position



Vehicle-Driver System Poles
Uncertain Driver Gain: Lateral Position



sensitivity analysis of driver gains

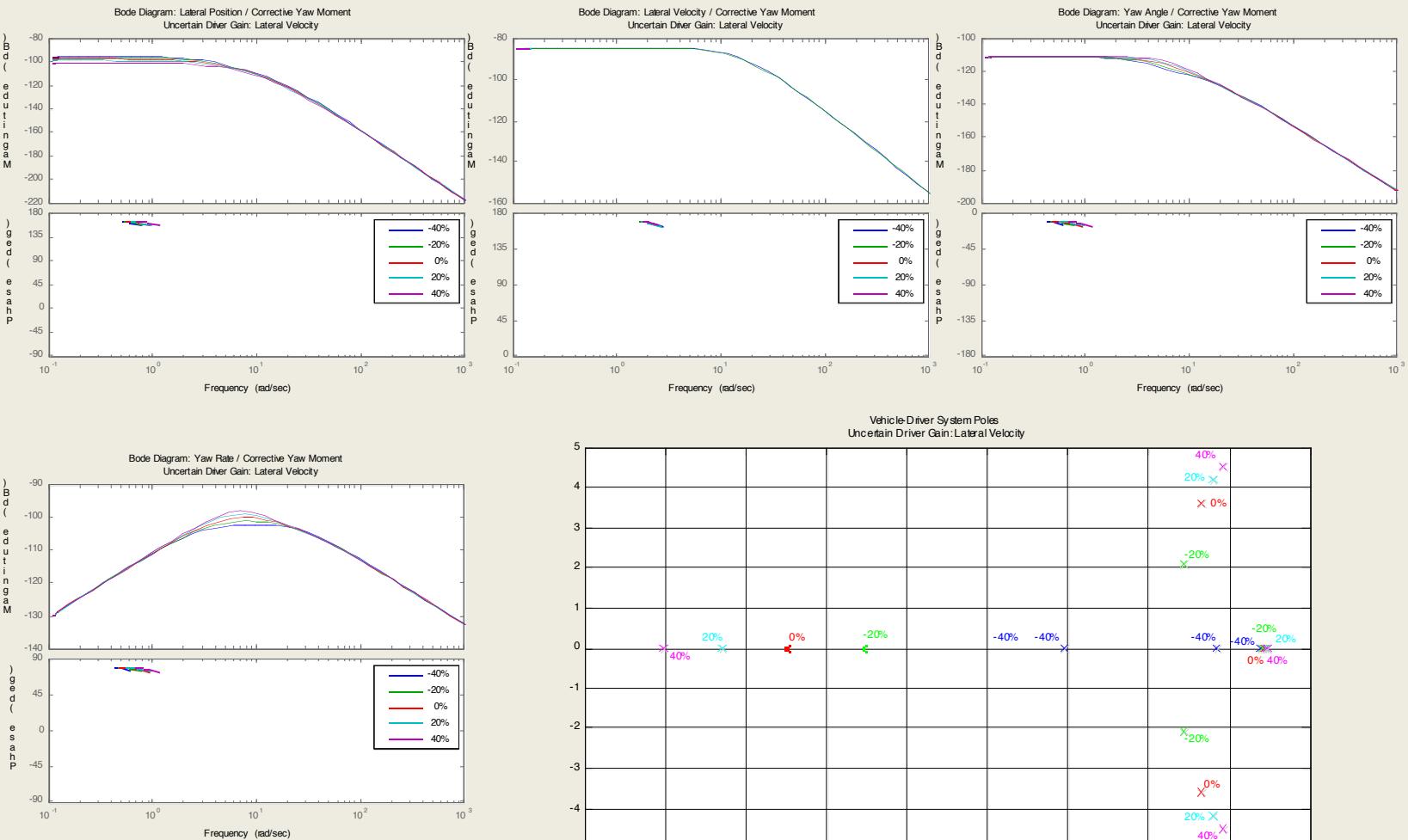
- Effect of uncertainty in lateral velocity gain



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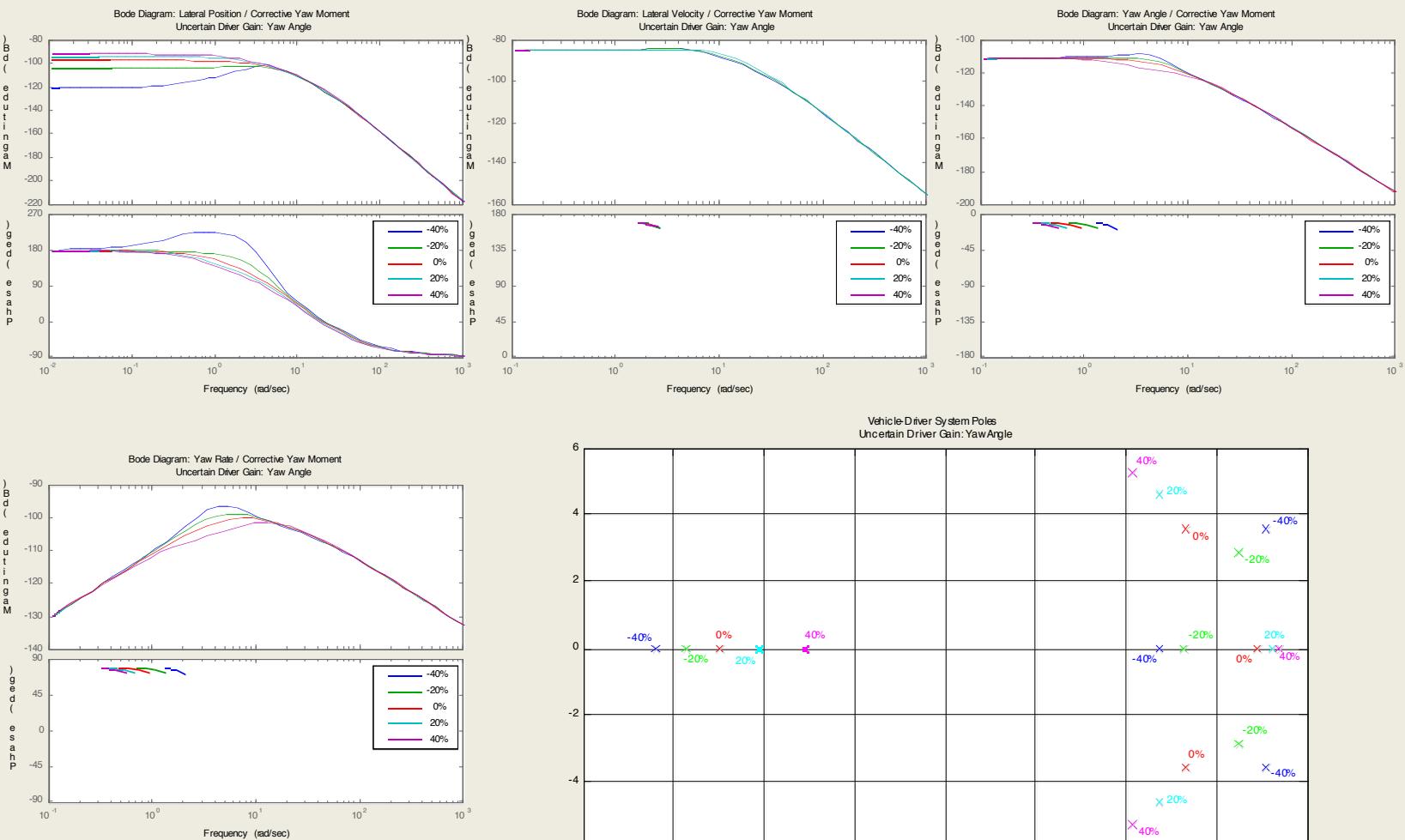
- Effect of uncertainty in yaw angle gain



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sensitivity analysis of driver gains

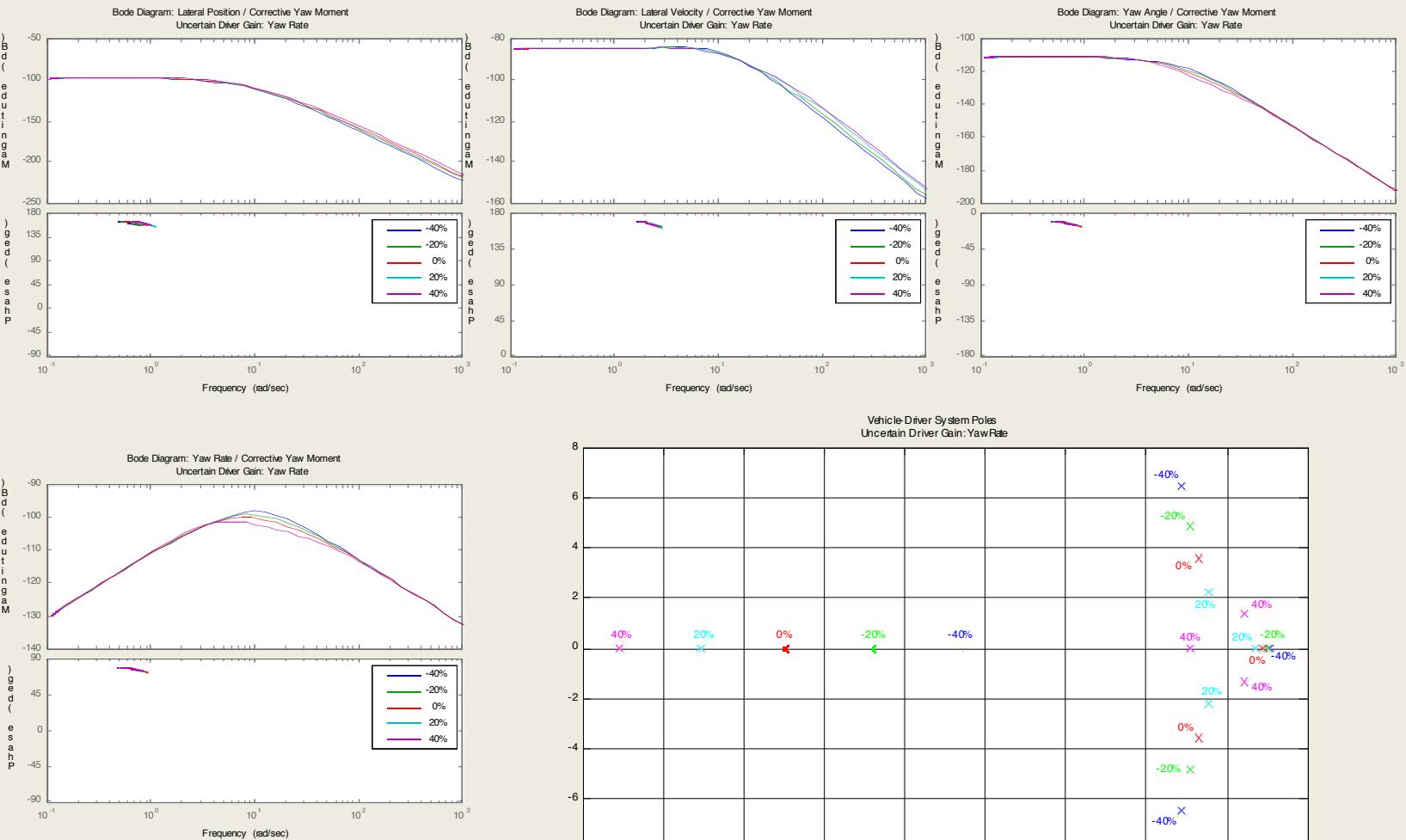
- Effect of uncertainty in yaw rate gain



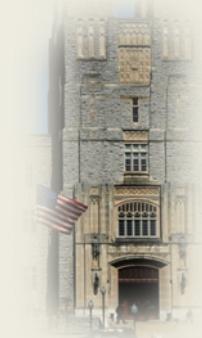
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μ -sensitivity plots



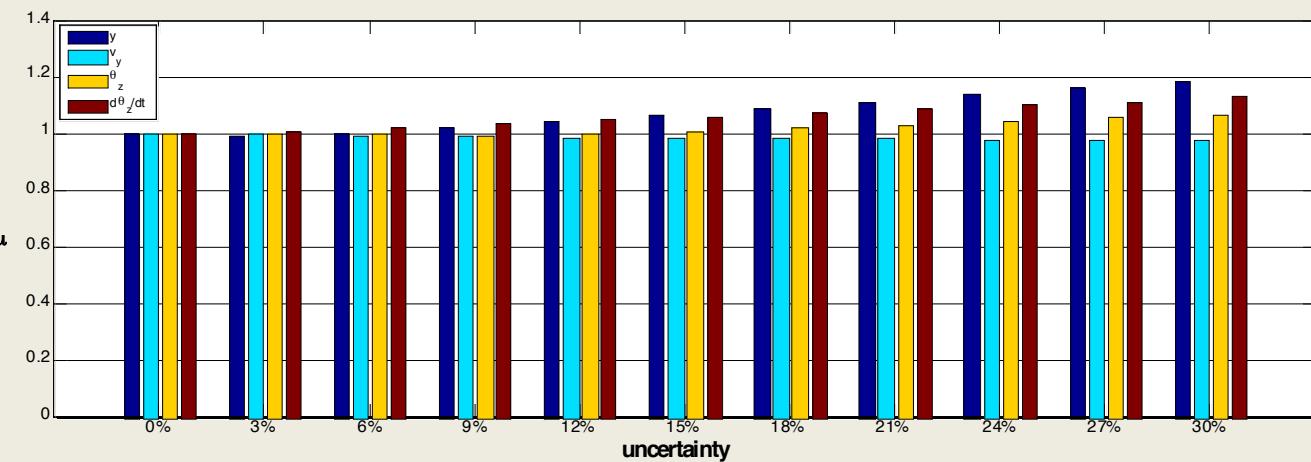
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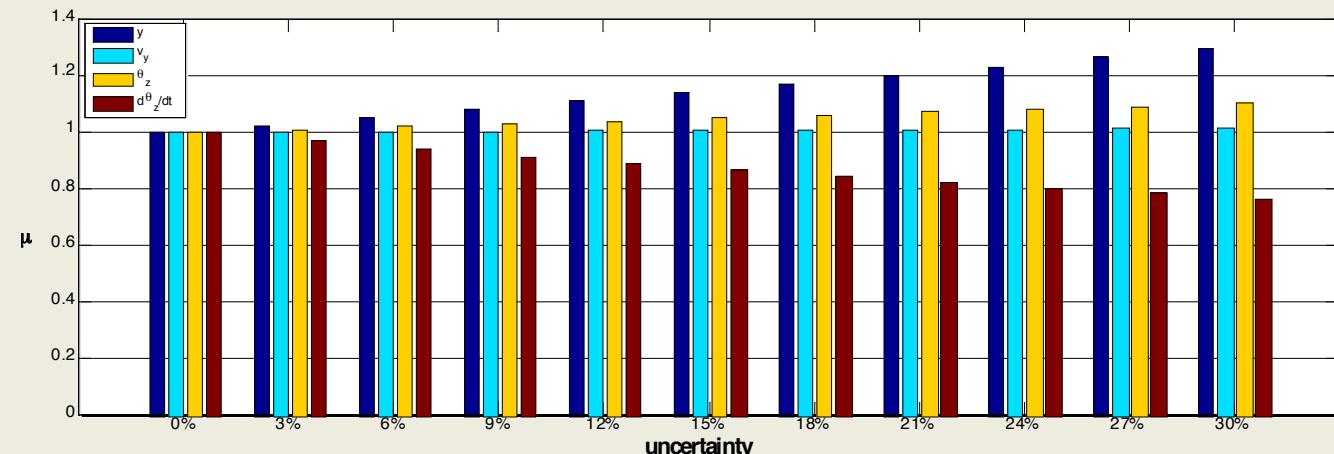
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Game Theory



LQR



robust interaction model

- Assume that uncertainties exist in driver model (output)

$$\dot{x}(t) = \mathbf{A}_c x(t) + \mathbf{B}_{c1} \left(\underbrace{\delta_{sw}(t)}_{u_1} + \bar{\delta}(t) \right) + \mathbf{B}_{c2} \underbrace{M_{zc}(t)}_{u_2}$$

- The uncertainty is assumed to be smooth disturbances and bounded by

$$\forall t \geq 0, \quad \bar{\delta}(t) \leq \mathbf{F}x(t) + \bar{\delta}_0$$

with $\mathbf{F} \in \mathbb{R}^{1 \times 4}$ and $\bar{\delta}_0 > 0$

- To obtain robust control design, the “Integral Sliding Model” approach is applied:

$$\begin{cases} u_1 = u_{o1} \\ u_2 = u_{o2} + \delta_{opt} \end{cases}$$



optimal control
disturbance rejecting control



Integral Sliding Mode control design

Uncertain Model :

$$\dot{x}(t) = \mathbf{A}_c x(t) + \mathbf{B}_{c1} \left(\underbrace{u_{sw} + (\bar{\delta})(t)}_{u_1} \right) + \mathbf{B}_{c2} \underbrace{M_{zc}(t)}_{u_2}$$

↓

$$\dot{x}(t) = \mathbf{A}_c x(t) + \mathbf{B}_{c1} (u_{o1}(t) + \bar{\delta}(t)) + \mathbf{B}_{c2} (u_{o2}(t) + u_\delta(t))$$

Sliding Surface :

$$s(x, t) = s_o(x, t) + s_\delta(x, t)$$

$$\begin{aligned}\dot{s}(x, t) &= \dot{s}_o(x, t) + \dot{s}_\delta(x, t) = \frac{\partial s_o(x, t)}{\partial x} \dot{x} + \frac{\partial s_o(x, t)}{\partial t} + \dot{s}_\delta(x, t) \\ &= \frac{\partial s_o(x, t)}{\partial x} \left(\mathbf{A}_c x(t) + \mathbf{B}_{c1} (u_{o1}(t) + \bar{\delta}(t)) + \mathbf{B}_{c2} (u_{o2}(t) + u_\delta(t)) \right) + \frac{\partial s_o(x, t)}{\partial t} + \dot{s}_\delta(x, t)\end{aligned}$$

Selecting $\mathbf{G}^T (\mathbf{B}_{c1} \bar{\delta}(t) + \mathbf{B}_{c2} u_\delta(t))$

$$s_o(x, t) = \mathbf{G}^T x(t)$$

$$\dot{s}_\delta(x, t) = -\frac{\partial s_o(x, t)}{\partial t} - \frac{\partial s_o(x, t)}{\partial x} \left(\mathbf{A}_c x(t) + \sum_{i=1}^2 \mathbf{B}_{ci} u_{oi}(t) \right) = -\mathbf{G}^T \left(\mathbf{A}_c x(t) + \sum_{i=1}^2 \mathbf{B}_{ci} u_{oi}(t) \right)$$

$$s_\delta(x_0, t_0) = -s_o(x_0, t_0)$$



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Integral Sliding Mode control design

Lyapunov Candidate :

$$V(x, t) = \frac{1}{2} s^2(x, t) \geq 0 , \quad \forall t$$

$$\dot{V}(x, t) = s(x, t) \dot{s}(x, t) = s(x, t) \left(\mathbf{G}^T \left(\mathbf{B}_{c1} \bar{\delta}(t) + \mathbf{B}_{c2} u_\delta(t) \right) \right)$$

Selecting the disturbance rejecting controller as

$$u_\delta = -\mathbf{B}_{c2}^T S \left(\mathbf{B}_{c2} \mathbf{B}_{c2}^T \mathbf{B}_{c1}^{-1} \mathbf{R}_{c1}^{-1} \left(\mathbf{F} x(t) + \bar{\delta}_0 \right) \bar{\delta}_0 \operatorname{sgn}(s(x, t)) \right) \leq 0 , \quad \forall t$$



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The robust uncertain system :

$$\dot{x}(t) = \underbrace{\left(\mathbf{A}_c - \mathbf{B}_{c1} \mathbf{F} \operatorname{sgn}(s(x, t)) \right)}_{\mathbf{A}_{eq}} x(t) + \mathbf{B}_{c1} u_{o1}(t) + \mathbf{B}_{c2} u_{o2}(t) - \underbrace{\mathbf{B}_{c1} \bar{\delta}_0 \operatorname{sgn}(s(x, t))}_{\omega(t)} + \mathbf{B}_{c1} \bar{\delta}(t)$$

with sliding surface :

$$s(x, t) = \mathbf{G}^T x(t) - \mathbf{G}^T \int_0^t \left(\mathbf{A}_c x(\varepsilon) + \sum_{i=1}^2 \mathbf{B}_{ci} u_{oi}(\varepsilon) \right) d\varepsilon$$

Integral Sliding Mode control design



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Suppose $\hat{\mathbf{E}}_i$ satisfy the coupled Riccati equations given by

$$\mathbf{A}_{eq}^T \mathbf{K}_i + \mathbf{K}_i \mathbf{A}_{eq} + \mathbf{Q}_i - \mathbf{K}_i \mathbf{S}_i \mathbf{K}_i - \mathbf{K}_i \mathbf{S}_{\hat{i}} \mathbf{K}_{\hat{i}} - \mathbf{K}_{\hat{i}} \mathbf{S}_{\hat{i}} \mathbf{K}_i + \mathbf{K}_{\hat{i}} \mathbf{S}_{\hat{i}\hat{i}} \mathbf{K}_{\hat{i}} = \mathbf{0}$$

and the shifting vectors k_i satisfy

$$\mathbf{A}_{eq}^T k_i + \mathbf{K}_i \mathbf{S}_i k_i - \mathbf{K}_{\hat{i}} \mathbf{S}_{\hat{i}} k_{\hat{i}} + \mathbf{K}_{\hat{i}} \mathbf{S}_{\hat{i}\hat{i}} k_{\hat{i}} = \mathbf{0}$$

The robust uncertain system

$$\dot{x}(t) = (\mathbf{A}_c - \mathbf{B}_{c1} \mathbf{F} \operatorname{sgn}(s(x, t)))x(t) + \mathbf{B}_{c1} u_{o1}(t) + \mathbf{B}_{c2} u_{o2}(t) - \underbrace{\mathbf{B}_{c1} \bar{\delta}_0 \operatorname{sgn}(s(x, t))}_{\omega(t)} + \mathbf{B}_{c1} \bar{\delta}(t)$$
$$\mathbf{S}_i = \mathbf{B}_{ci} \mathbf{R}_{ii}^{-1} \mathbf{B}_{ci}^T, \quad \mathbf{A}_{eq\hat{i}\hat{i}} = \mathbf{B}_{\hat{c}\hat{i}} \mathbf{R}_{\hat{i}\hat{i}}^{-1} \mathbf{R}_{\hat{i}\hat{i}}^T \mathbf{B}_{\hat{c}\hat{i}}^T$$

The sliding surface :

$$u_{oi}^*(t) = -\mathbf{R}_{pi}^{-1} \mathbf{B}_{ci}^T (\mathbf{K}_i x(t) + k_i)$$
$$s(x, t) = \mathbf{G}^T x(t) - \mathbf{G}^T \int_0^2 \mathbf{A}_c x(\varepsilon) + \sum_{i=1}^2 \mathbf{B}_{ci} u_{oi}(\varepsilon) d\varepsilon$$

is a linear feedback Nash equilibrium for the above uncertain game system.

simulation

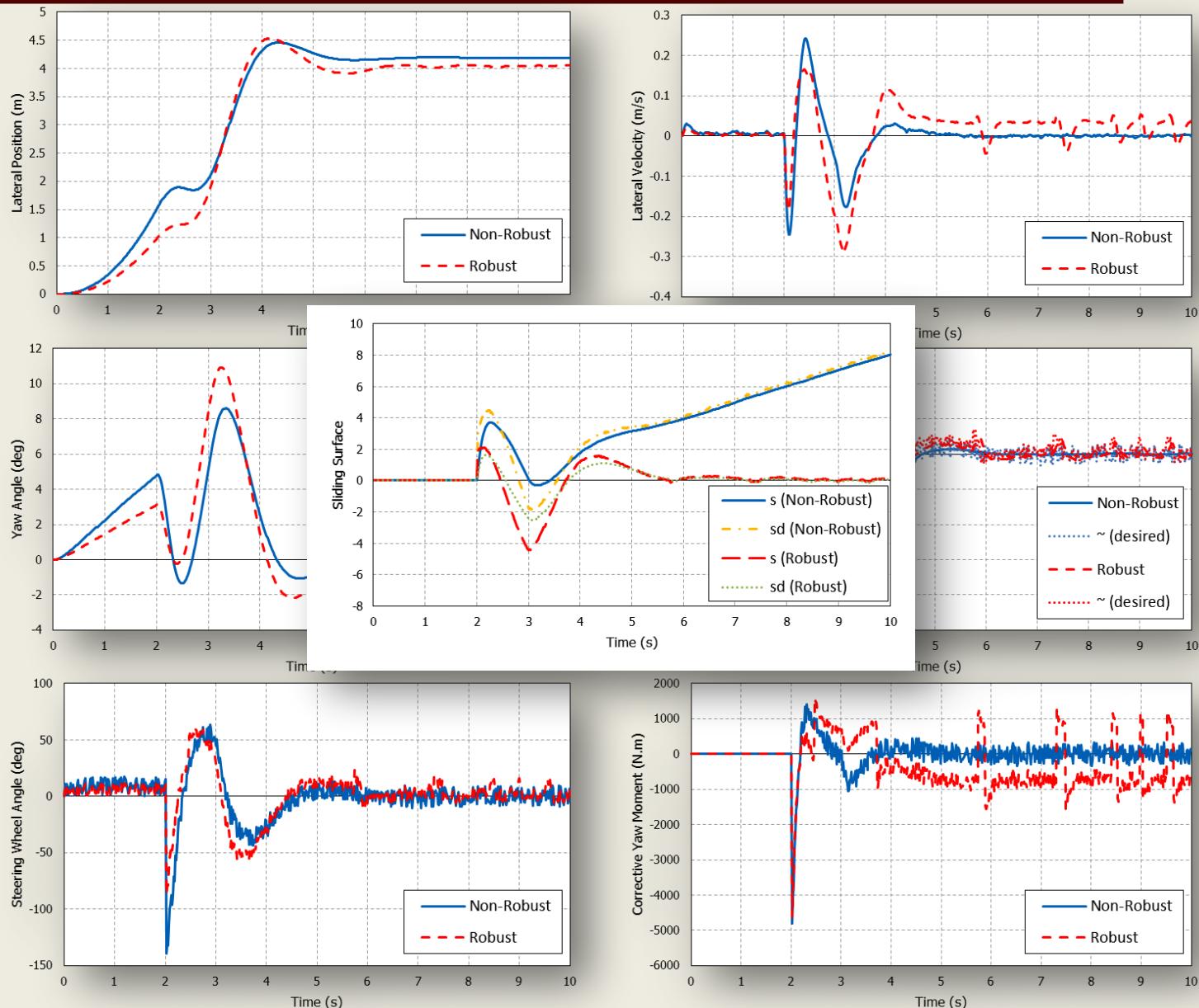
- Vehicle: same sedan vehicle
- Maneuver: lane-change of $5m$ at 20 m/s
- Control Objective: improved handling performance
- Driver/ Vehicle controller LQ structure: same
- Driver steering angle uncertainty: $F = (20\%)B_1$, $\bar{\delta}_0 = 10^\circ$

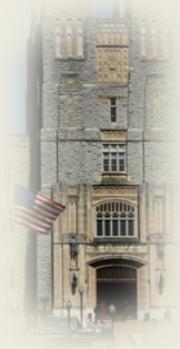


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results





conclusions

- Extension of discrete preview driver steering control model to include lateral velocity and yaw rate was developed
- Introduction of modified Cruise Control and Differential Braking systems for application of the proposed control algorithm
- A new structure for optimal linear vehicle steering and yaw control has been devised based on linear quadratic Game Theory
 - * Continuous-time interaction model
 - * Discrete-time interaction model with preview-time
 - * Robust interaction model with sensitivity analysis



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Thank You!

