



# Control of a Ground Vehicle Using Quadratic Programming Based Control Allocation

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John Plumlee

David M. Bevly

A. Scott Edward Hodel

**Auburn University**



# Outline

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- The Problem and Objectives
- Control Allocation
- Vehicle Model
- Controller Design
- Simulation Results
- Conclusions

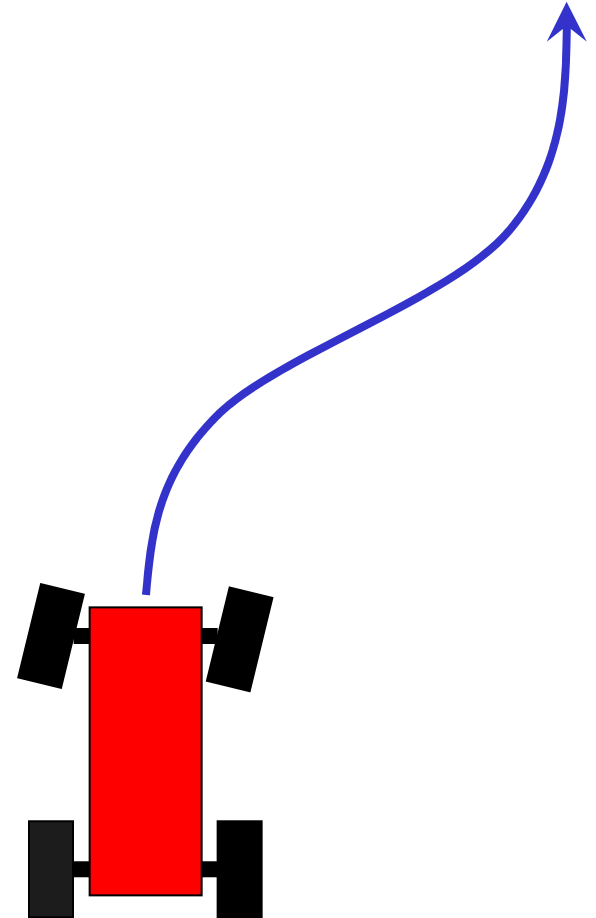
# Problem and Objectives

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We want to track a yaw rate trajectory with a 4 wheeled ground vehicle while minimizing the sideslip angle.

Question: But if there are multiple inputs available, How do you choose which to use and how much?

Answer: Control Allocation!!





# Control Allocation (CA)

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- CA involves generating a set of effector commands that produce a desired control effect while minimizing the control effort.
- A CA approach is generally used for a redundant set of effectors.
- CA allows for reconfiguration in the event of an effector failure.



# Vehicle Model (nonlinear)

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Assumptions made:

- The vehicle runs at constant velocity
- No longitudinal weight transfer (pitch is neglected)
- A Pacejka tire model was used to represent nonlinear tire behavior.



# Vehicle Model (linear)

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The linear vehicle model in state space form:

$$\begin{bmatrix} \dot{\beta} \\ \dot{r} \end{bmatrix} = \begin{bmatrix} \frac{-C_o}{mV} & \frac{-C_1}{mV^2} - 1 \\ \frac{-C_1}{I_z} & \frac{-C_2}{VI_z} \end{bmatrix} \begin{bmatrix} \beta \\ r \end{bmatrix} + \begin{bmatrix} \frac{C_{\alpha f}}{mV} & 0 & 0 \\ a \frac{C_{\alpha f}}{I_z} & \frac{T_f}{2I_z} & \frac{T_r}{2I_z} \end{bmatrix} \begin{bmatrix} \delta \\ \Delta F_{xf} \\ \Delta F_{xr} \end{bmatrix}$$

$\delta$  = steering angle

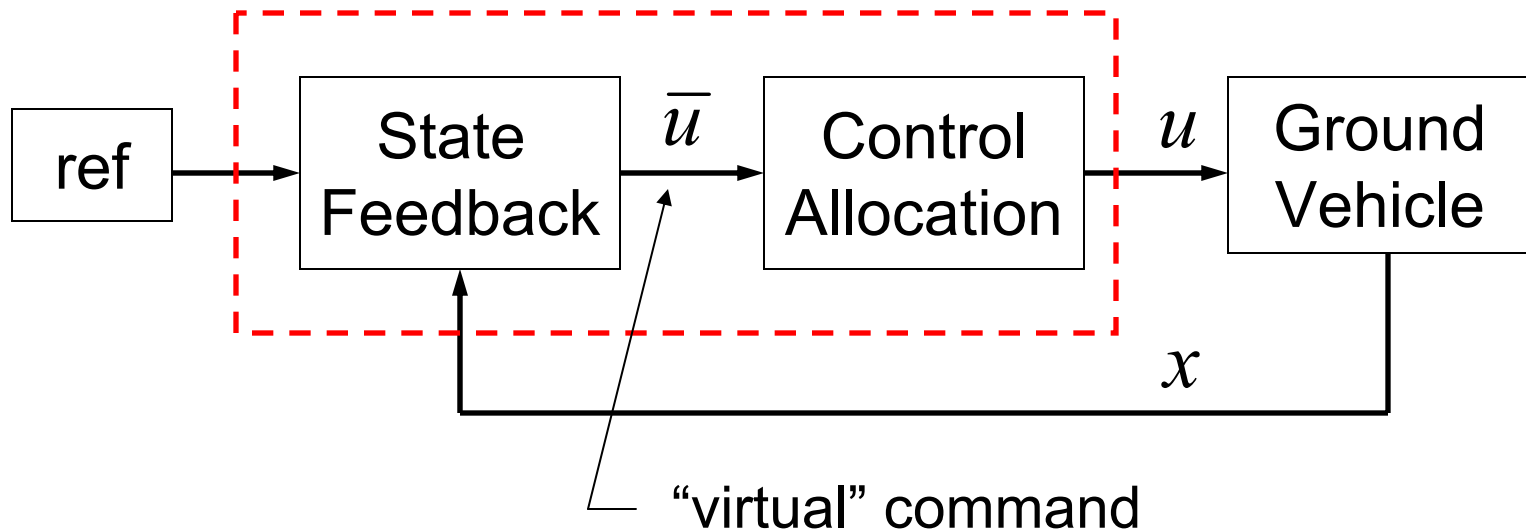
$\Delta F_{xf}$  = differential braking force on front axle

$\Delta F_{xr}$  = differential braking force on rear axle

# Controller Design Approach

Controller has 2 main tasks:

1. Generate Control Effort  $\longrightarrow$  State Feedback
2. Generate Effector Commands  $\longrightarrow$  Control Allocation





# Control Effort

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LQR gains are designed for the following modified system:

$$\begin{bmatrix} \dot{\beta} \\ \dot{r} \\ \dot{\psi} \end{bmatrix} = \begin{bmatrix} \frac{-C_o}{mV} & \frac{-C_1}{mV^2} - 1 & 0 \\ \frac{-C_1}{I_z} & \frac{-C_2}{VI_z} & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \beta \\ r \\ \psi \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \bar{u}_\beta \\ \bar{u}_r \end{bmatrix}$$

This model is used solely for generating the virtual command effort.

The gain matrix  $K$  is applied to the error vector to produce the overall desired effect.

$$\begin{bmatrix} \bar{u}_\beta \\ \bar{u}_r \end{bmatrix} = -K_{2 \times 3} \begin{bmatrix} \beta_{error} \\ r_{error} \\ \psi_{error} \end{bmatrix}$$



# Command Generation

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Physical effector commands are generated by the Control Allocation routine.

## Quadratic Programming (QP)

Given the virtual command  $\bar{u}$ , solve for the effector commands  $u$ .

$$\min u: \frac{1}{2} u^T Q u + c^T u$$

$$\text{subject to: } B u = \bar{u}$$

$$u^- \leq u \leq u^+$$

# Command Generation: slack variables 1

The Sign Preserving Quadratic Programming (SPQP) problem set up as follows:

$$\min u, \sigma : \frac{1}{2} u^T Q_u u + c^T u + \frac{1}{2} Q_\sigma (1 - \sigma_\beta)^2 + \frac{1}{2} Q_\sigma (1 - \sigma_r)^2$$

subject to:  $Bu - \Sigma \bar{u} = 0$

where

$$\begin{array}{c} \text{Ignore virtual} \\ \text{command} \end{array} \rightarrow \begin{bmatrix} u^- \\ 0 \\ 0 \end{bmatrix} \leq \begin{bmatrix} u \\ \sigma_\beta \\ \sigma_r \end{bmatrix} \leq \begin{bmatrix} u^+ \\ 1 \\ 1 \end{bmatrix} \leftarrow \begin{array}{c} \text{achieve} \\ \text{virtual} \\ \text{command} \end{array}$$

$$\Sigma = \begin{bmatrix} \sigma_\beta & 0 \\ 0 & \sigma_r \end{bmatrix}$$

# Command Generation: slack variables 2

Due to the structure of the input matrix  $B$ , it is necessary to add a 4th "pseudo" input  $v$ .

$$B = \begin{bmatrix} \frac{C_{of}}{mV} & 0 & 0 & 1 \\ a\frac{C_{of}}{I_z} & \frac{T_f}{2I_z} & \frac{T_r}{2I_z} & 0 \end{bmatrix} \quad u = \begin{bmatrix} \delta \\ \Delta F_{xf} \\ \Delta F_{xr} \\ v \end{bmatrix} \quad Q = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & Q_v \end{bmatrix}$$

This relaxes the equality constraint placed on the sideslip angle.



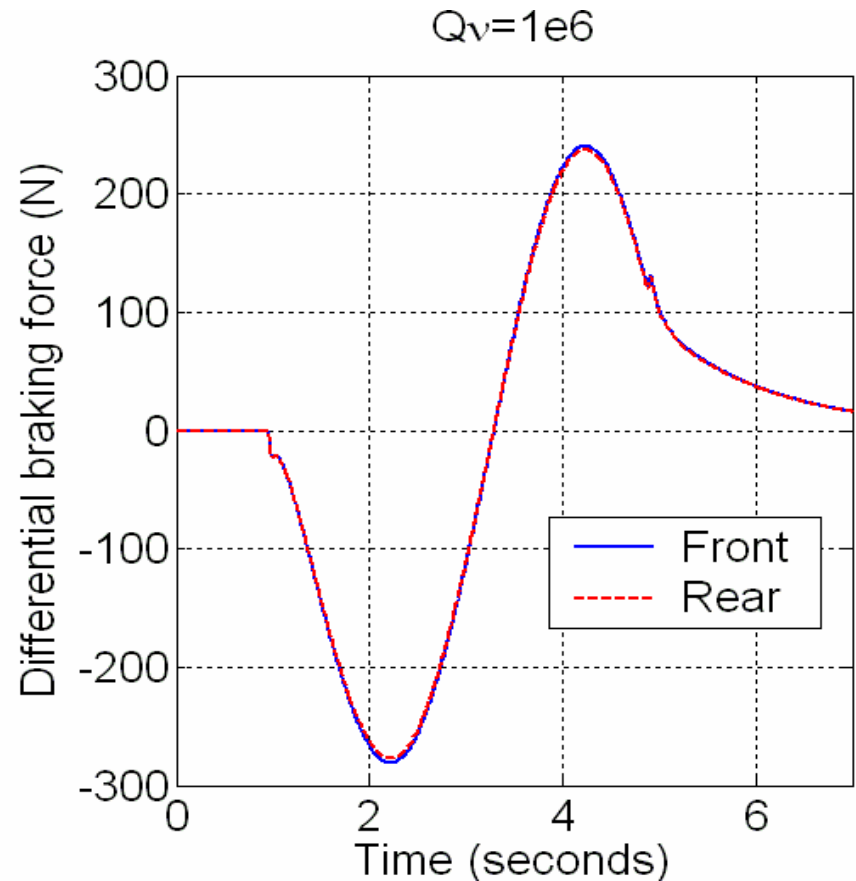
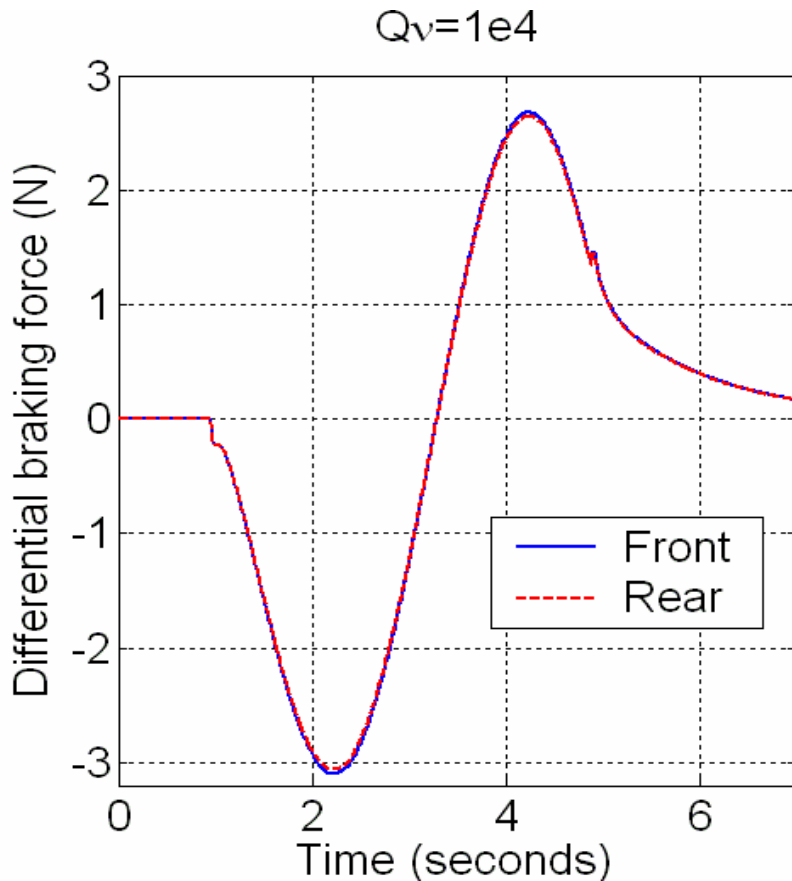
# Simulation

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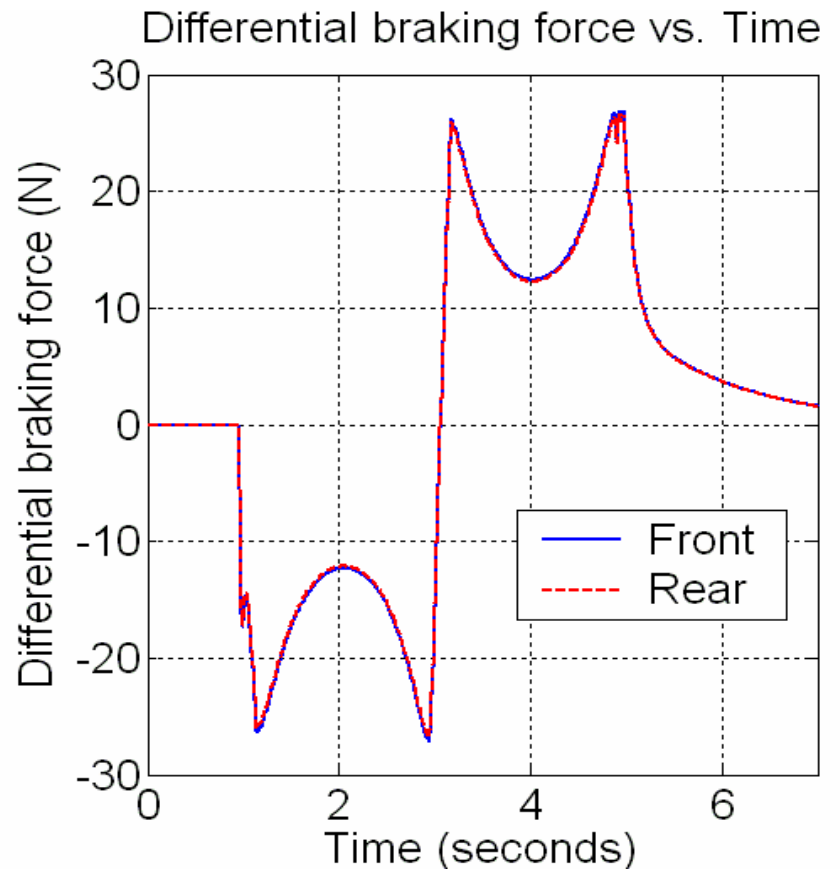
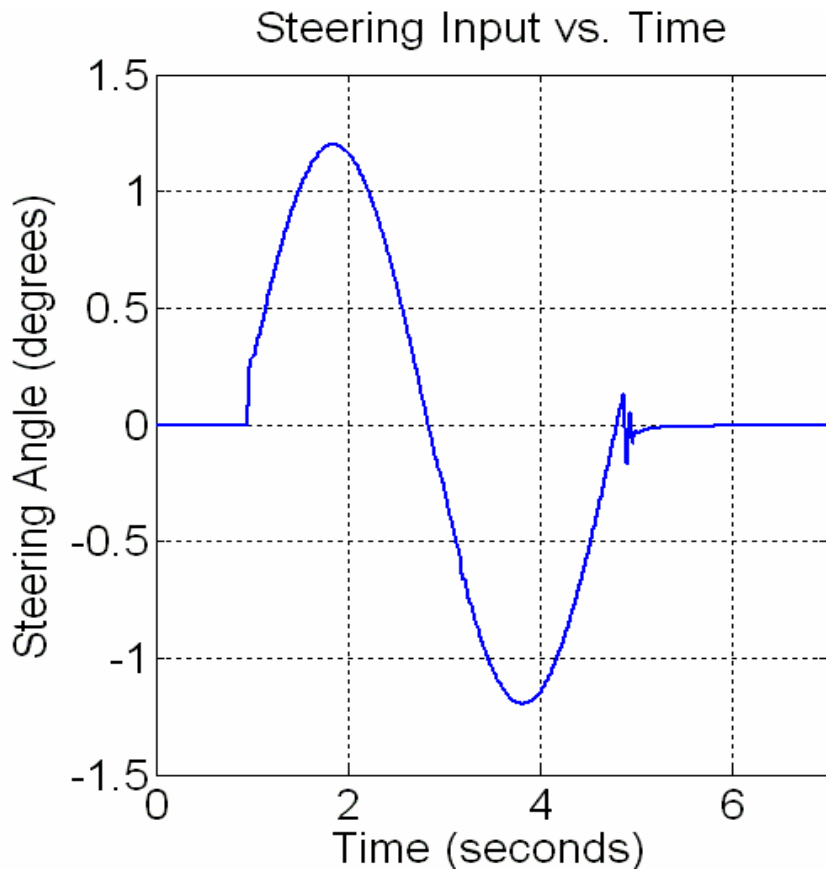
- The vehicle is simulated traveling at constant velocities of 45, 55, and 65mph.
- Scenarios examined: nominal, brake failure, and steering failure.
- Penalty on pseudo effector:  $Q_v=10^4, 10^6$ .
- The desired yaw rate trajectory simulates a double lane change maneuver.
- Effector limitations: steering angle=  $\pm 0.5rad$   
braking=  $0.75 * F_{zi}$

# Results (pseudo effector)

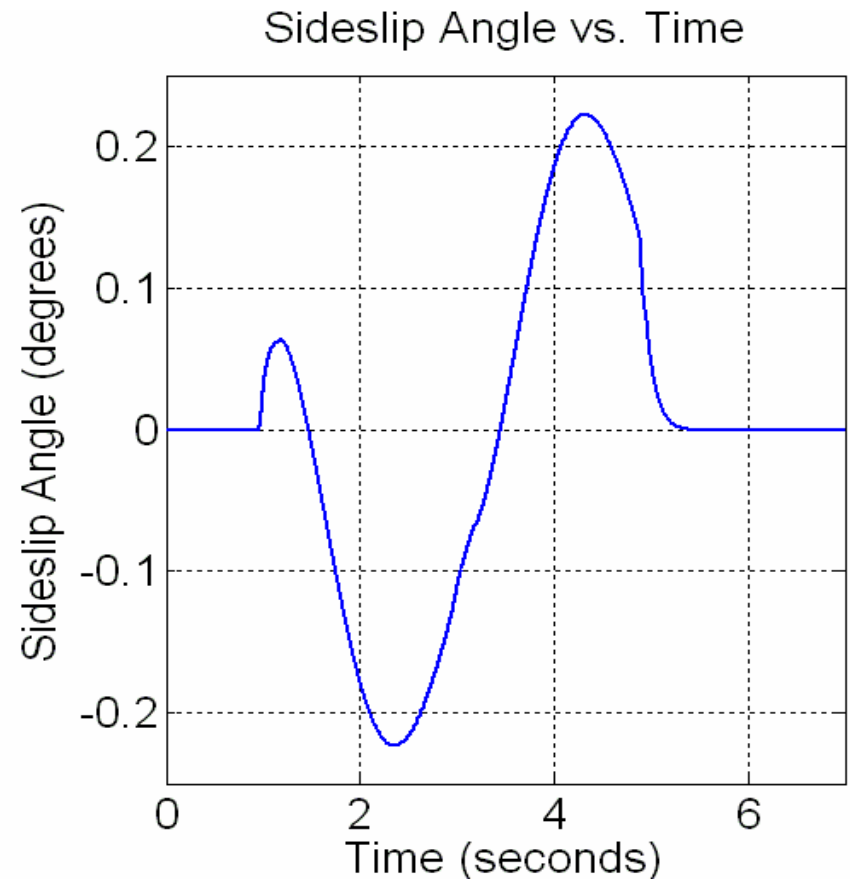
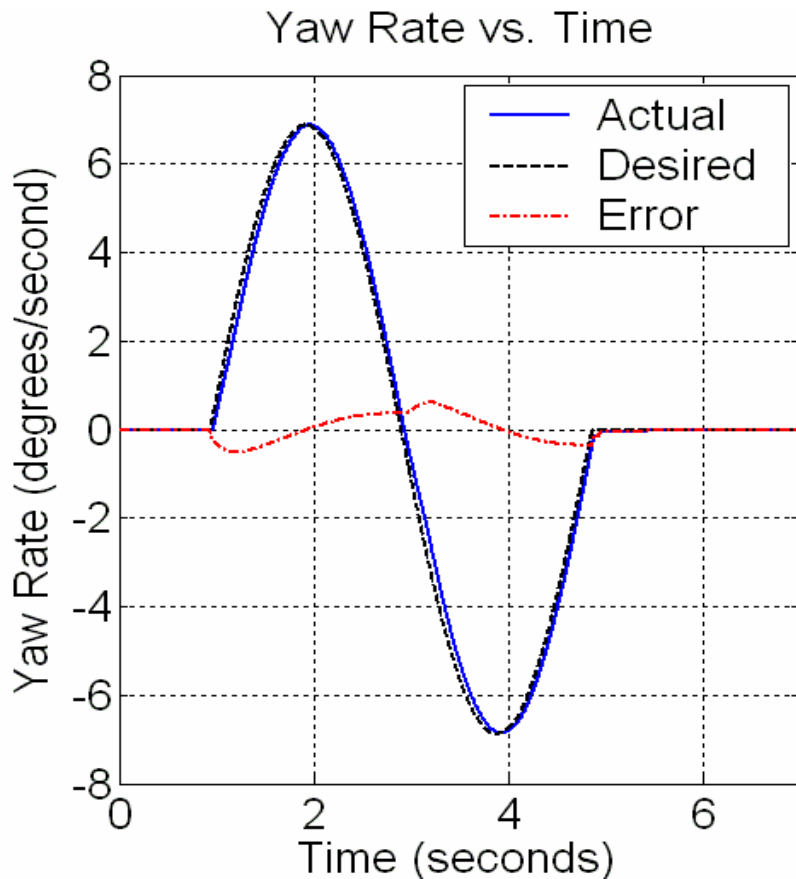
## Nominal , 55mph, QP



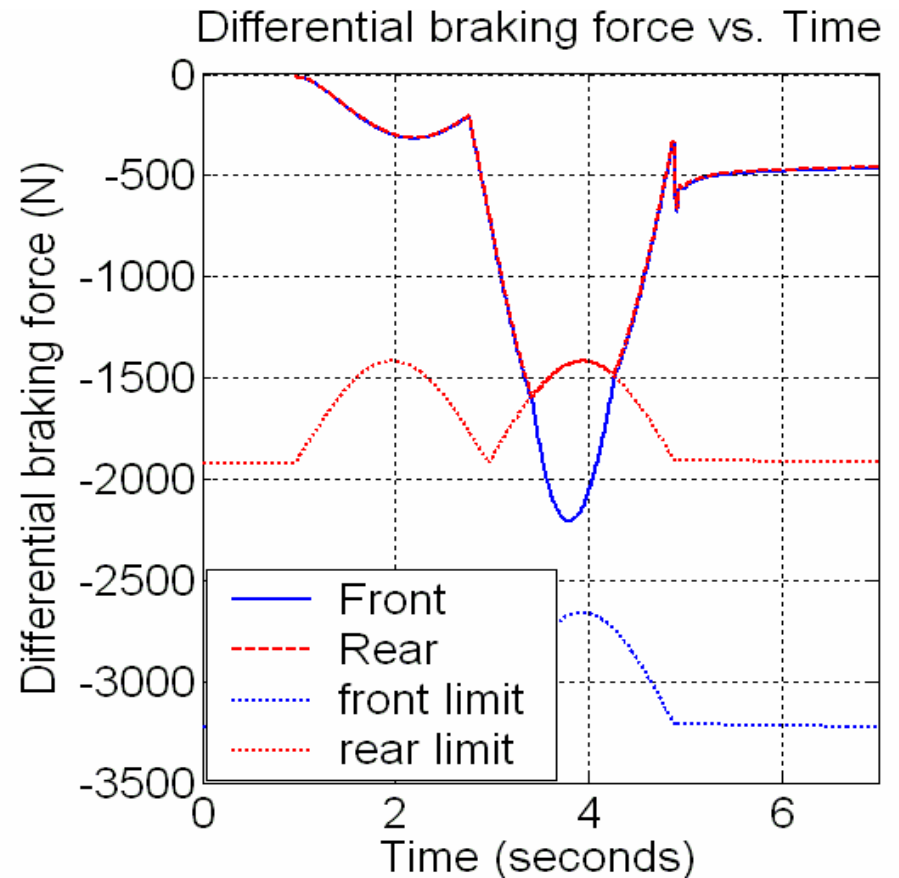
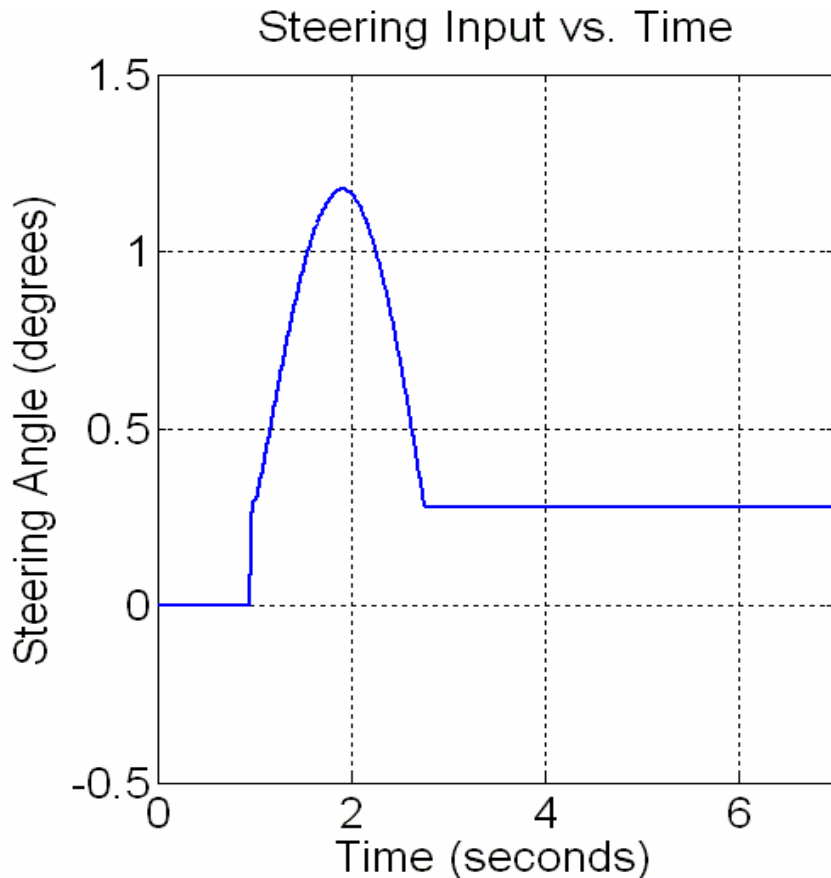
# Results – Nominal, 45mph SPQP



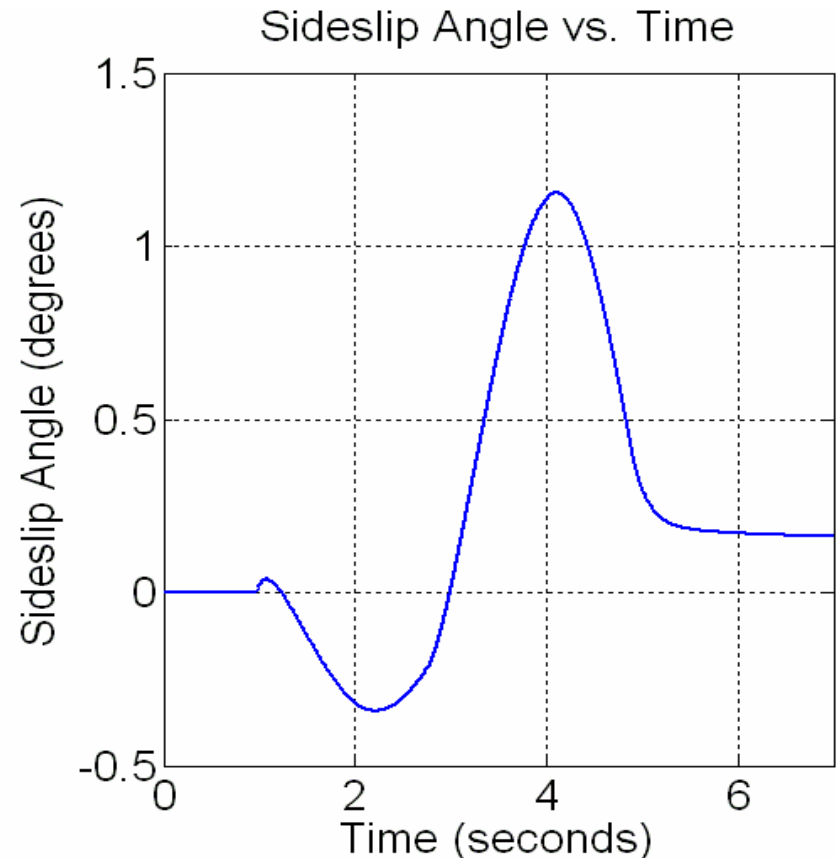
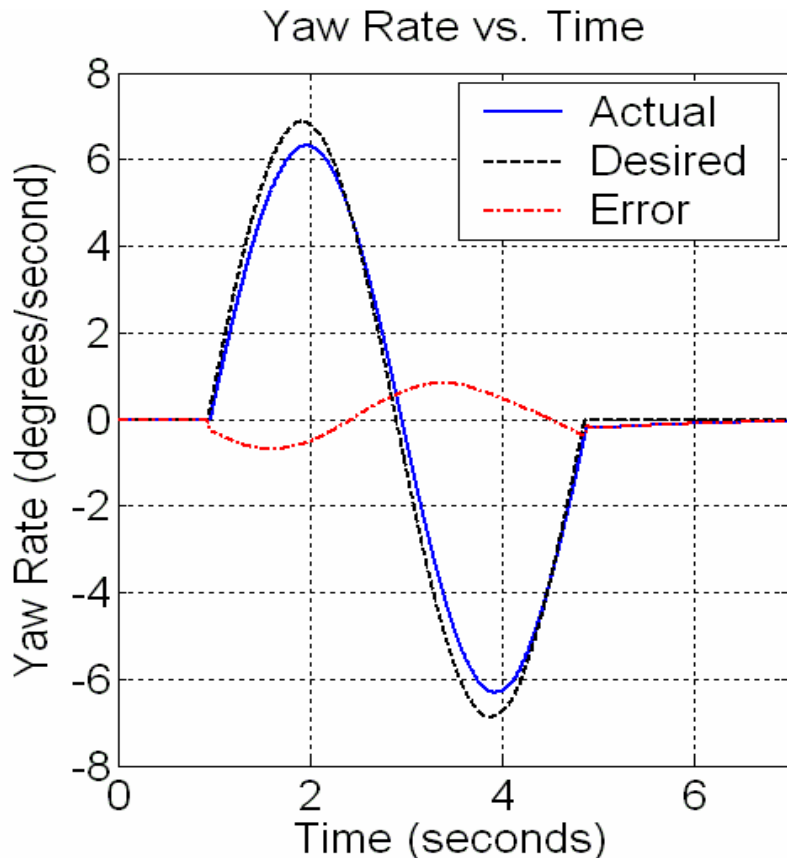
# Results – Nominal, 45mph SPQP



# Results – Steering failure, 55mph QP



# Results – Steering failure, 55mph QP





# Conclusions

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- QP provides an intelligent reconfiguration of control effort in the event of a failure and also obeys effector limits.
- SPQP gives the best yaw rate tracking but standard QP provides the best sideslip minimization.
- The quadratic penalty placed on the pseudo effector significantly affects the results.



# Current Work

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- Gain insight for controller tuning through further investigation of the effects of the pseudo effector weight  $Q_v$ .
- Address differential braking saturation limits. Better estimate of limits using the friction circle.
- Study the effects of additional inputs such as rear steering and individual torque control of each wheel.

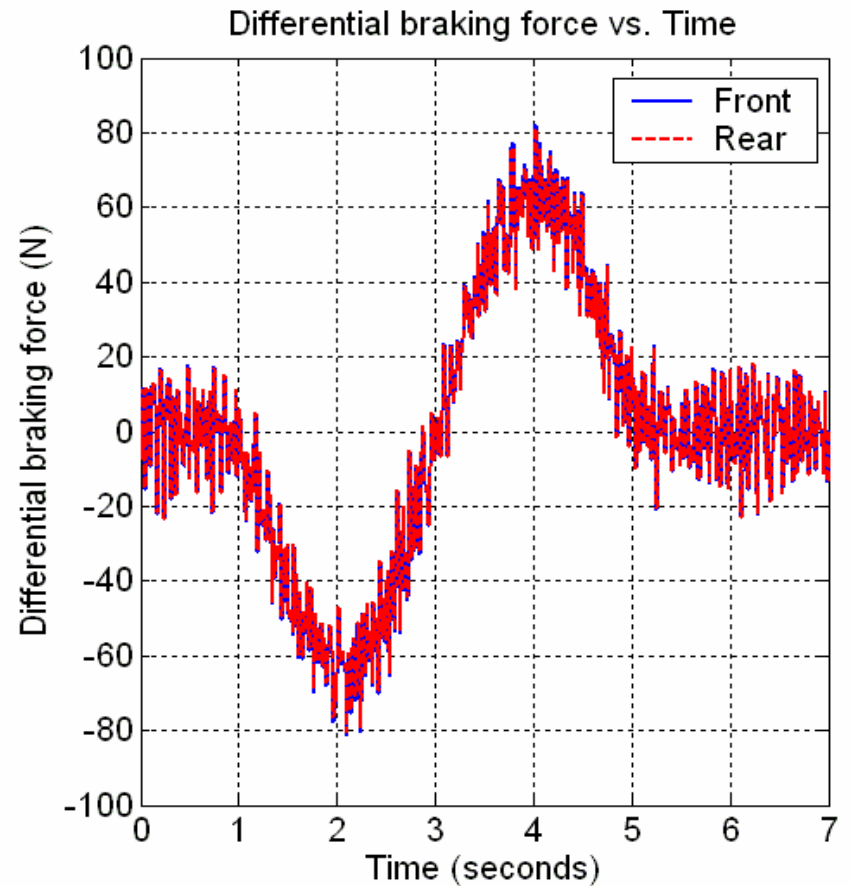
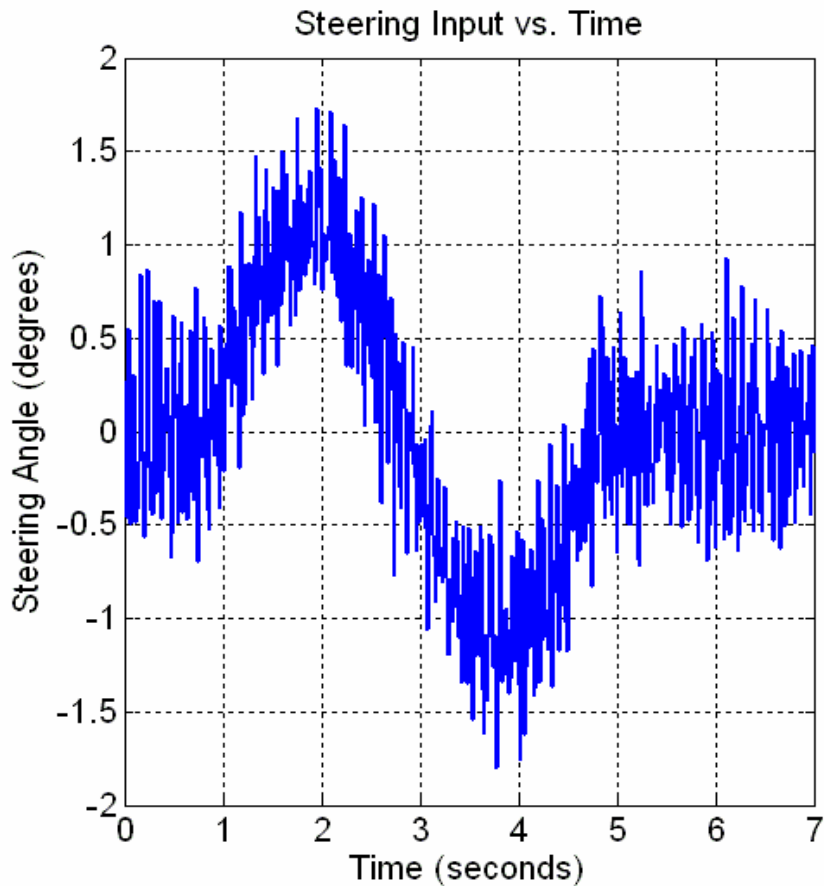


# Questions

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# Answers – Commands w/ noisy state estimates



# Answers – tracking w/ noise

