# Plant Modeling for an Autonomous Vehicle

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## Objective and Contributions

### Objective

- Provide accurate plant models of each autonomous vehicle subsystem to be used for designing controllers **Contribution**
- Determine if System Identification or Neural Network modeling produces better models
- Non-linearity modeling

## Applications

- Use in testing to help develop more accurate vehicle controllers
- Create a guide for modeling future vehicle subsystems

## Problem Setup

- Conducted a literature review to look for existing solutions
- System Architecture

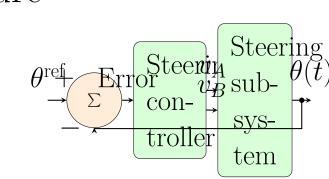


Figure 1:Steering subsystem block diagram.

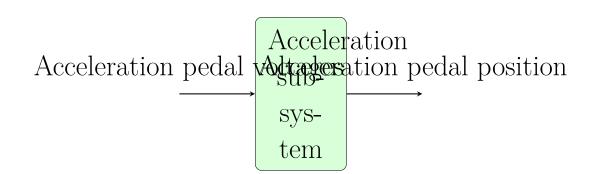


Figure 2:Acceleration subsystem block diagram

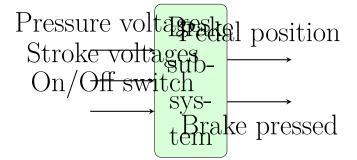


Figure 3:Brake subsystem block diagram

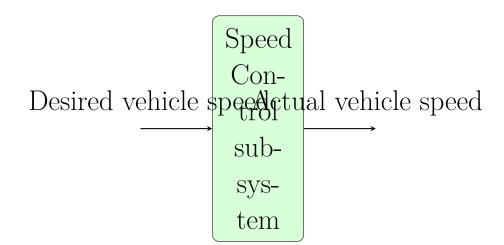


Figure 4:Speed Control subsystem block diagram

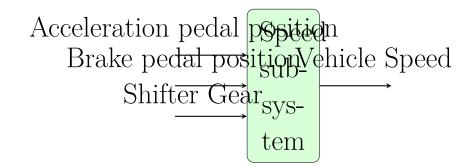


Figure 5:Speed subsystem block diagram

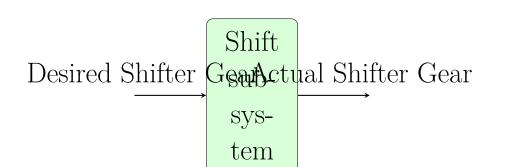


Figure 6:Shift subsystem block diagram

- Data Collection Setup
- Lexus RX450H vehicle platform
- Laptop
- PACMod ECU
- CANCase
- CAN Bus

Transfer Function Modeling
Figure 7:A desired orientation is given by a user. The difference between this input and the actual position is calculated. The controller the calculates the proper amount of voltage to apply to the DC motors.

• Employ state-space representation of 2-DOF helicopter:

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u}$$

Use state feedback law to minimize the quadratic cost function:  $\mathbf{u} = -\mathbf{K}\mathbf{x}$ 

$$J(\mathbf{u}) = \int_0^\infty (\mathbf{x}^T \mathbf{Q} \mathbf{x} + \mathbf{u}^T \mathbf{R} \mathbf{u} + 2\mathbf{x}^T \mathbf{N} \mathbf{u}) dt$$

3 Find the solution S to the Riccati equation

$$\mathbf{A}^T \mathbf{S} + \mathbf{S} \mathbf{A} - (\mathbf{S} \mathbf{B} + \mathbf{N}) \mathbf{R}^{-1} (\mathbf{B}^T \mathbf{S} + \mathbf{N}^T) + \mathbf{Q} = 0$$

• Calculate gain, **K** 

$$\mathbf{K} = \mathbf{R}^{-1}(\mathbf{B}^T\mathbf{S} + \mathbf{N}^T)$$

# Neural Network Modeling Utilizes gain calculated in LQR

- Added Kalman filter to reduce external disturbances to the system

Figure 8: Noise resistant 2-DOF helicopter model.

 Neural Network Algorithm
 Uses neural network based on difference between desired and actual orientation to determine optimal gain

Figure 9:ADP Neural Network

## Experimental Results Figure 10:Experimental Setup

### (a)(b)

Figure 11:ADP experimental results for (a) the main rotor and (b) the tail rotor given a step input

### (a)(b)

Figure 12:Comparison between P and PI control for a step input is shown for (a) the main rotor and (b) the tail rotor

Figure 13:(a) Time = 0 and (b) Time = 10

## Conclusion and Future Work

- Using Neural Networks produced more accurate models than System Identification
- Test models using Hardware-in-the-Loop
- Create new vehicle controllers