



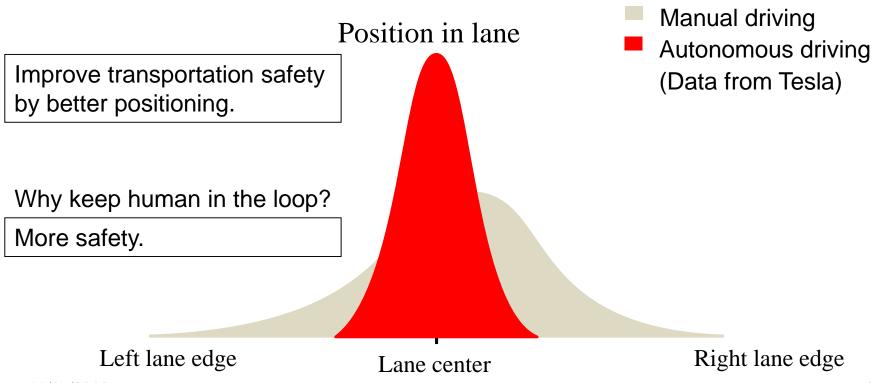
## **Overview**

- Introduction
  - Vehicle Steering Control and Lane Keeping
  - Human in the Loop
- Problem Formulation and Proposed Solution
  - Human-in-the-loop Control Framework
  - A Data-driven Method for Vehicle Steering Control with Human in the Loop
- Simulation





#### **Motivation:**



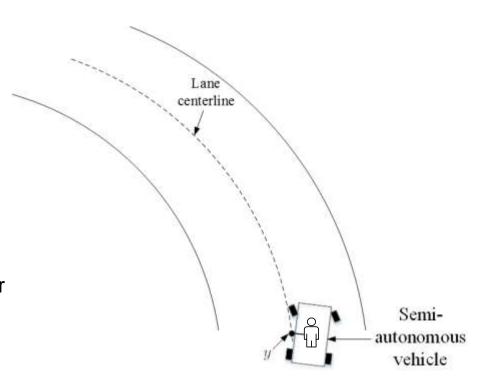


### **Problem description**

Semi-autonomous vehicle:
 To incorporate human driver into the design procedure

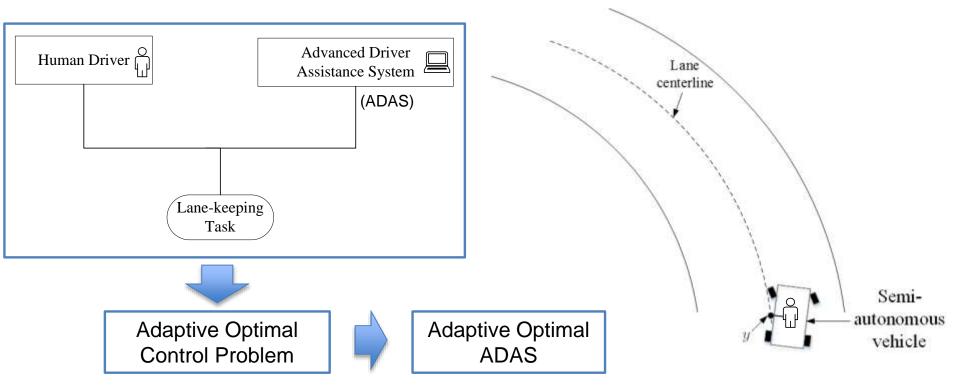
• The goal of our study:

To achieve lane keeping on the curving lane with human in the loop (the driver collaborates with the designed controller



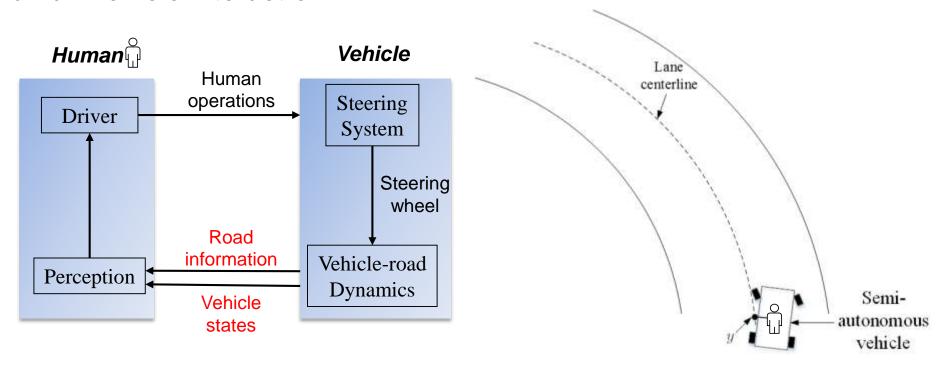


## **Human-in-the-loop Design Procedure**





#### **Human Vehicle Interaction**

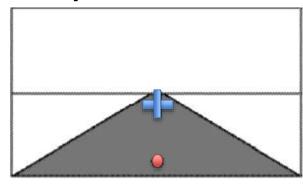




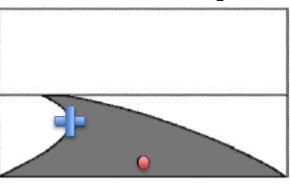
#### **Human driver model**

A human driver model represents the process by which a driver transforms perceived information about the driving situation into an action on the vehicle's actuators.

### Two-point visual model [D. Salvucci, 2004]

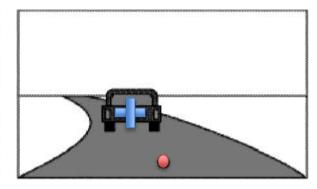


(a) Straight lane



(b) Curving lane

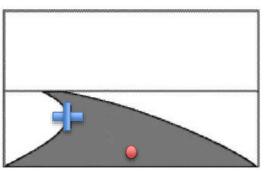
Near and far points



(c) With a leading vehicle



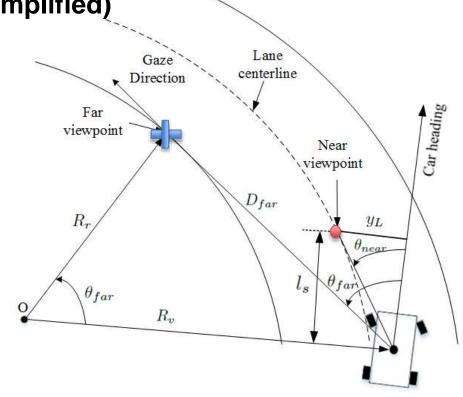
# Mathematical modeling for a driver (simplified)



The information from the two viewpoints can be approximated by **two visual angles**  $\theta_{near}$  and  $\theta_{far}$  [Saleh, 2013].

$$T_d = K_{near}\theta_{near} + K_{far}\theta_{far}$$

Driver's control torque







#### Driver's control torque

$$T_d = K_{near}\theta_{near} + K_{far}\theta_{far}$$

Vary from person to person



Adaptive ADAS provide more personalized service

Minimal intrusiveness and good performance

 $\begin{array}{c|c} \text{Visual Information} \\ \theta_{near}, \ \theta_{far} \end{array} \begin{array}{c} \text{Hum an } \\ \text{Driver} \end{array} \begin{array}{c} \text{Steering Torque} \\ \text{Road} \end{array}$ 

#### **Block Diagram**



Adaptive optimal control problem



Data-driven method

[1] W. Gao, Z. P. Jiang and K. Ozbay. "Data-driven adaptive optimal control of connected vehicles." *IEEE Transactions on ITS* (2016).

[2] Z. P. Jiang and Y. Jiang. "Robust adaptive dynamic programming for linear and nonlinear systems: An overview." *European Journal of Control* 19.5 (2013): 417-425.

Lane



# Data-driven Formulation for Vehicle Steering Control with Human in the Loop

Find an adaptive optimal ADAS (controller) to solve following optimization problem

$$\min_{u} J = \sum_{k=0}^{\infty} [qy^{2}(k) + ru^{2}(k)]$$

Lateral deviation from lane centerline

Assistance torque from ADAS with minimal intrusiveness

with minimal intrusiveness  $+ {\it T_d}) + Dw(k)$  Road information (curvature)

s.t.  $x(k+1) = Ax(k) + B(\mathbf{u}(\mathbf{k}) + \mathbf{T}_{\mathbf{d}}) + Dw(k)$  (Vehicle-road dynamics with human in the loop)

Note: Vehicle and human parameters are unknown.

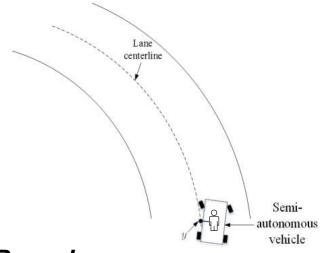
Semi-

vehicle



# Data-driven Method for Vehicle Steering Control with Human in the Loop

- Collect data
  - Vehicle internal states (sideslip angle, yaw rate, steering angle, change rate of steering angle)
  - Vehicle-road position states (lateral deviation, vehicle heading angle, road curvature)
  - Driver's control torque
     Sensors: camera, line sensors and GPS.
- Use data-driven adaptive dynamic programming (ADP) to find an adaptive optimal ADAS to improve the performance of lane keeping



#### Remark:

ADP integrates ideas from reinforcement learning and dynamic programming to design model-free adaptive optimal controllers.



#### Simulations on a circle:

Red:

Human driver

Blue:

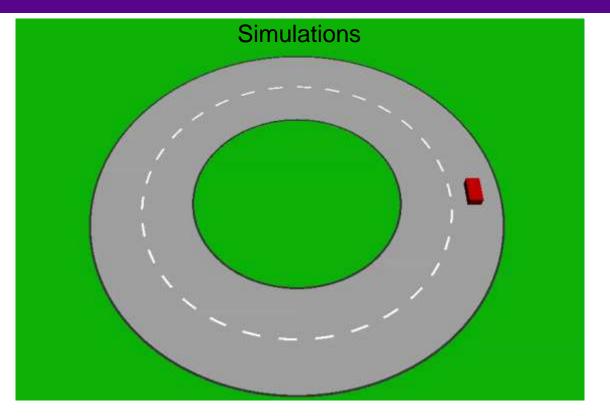
Human driver and ADAS

Performance

Red: About 20-30 cm

deviation

Blue: minimized lateral deviation



Animation made by Manuel Serrano



# Thank you!