State Space Modeling

ECE 495/595 Lecture Slides

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Summary and Quick Links

These slides contain the following concepts:

- ▷ Review of state space concepts (Slide 3)
- ▷ Unicycle navigation model (Slide 5)
- ⊳ Holonomic navigation model (Slide 9)
- ▶ Bicycle navigation model (Slide 12)
- ▷ Discrete state space models (Slide 15)

Review of State Space Concepts

- > State space is a method of describing the behavior of a system using a set of first-order differential equations.
- \triangleright State space models consist of a state equation and an observation equation.
- ➤ In a linear state space model, the two equations are typically written as:

Linear State Equation

$$\dot{\mathbf{X}} = \mathbf{AX} + \mathbf{BU}$$

- > State X
- > Control signal U
- > State matrix **A**
- > Input matrix B

Linear Observation Equation

$$z = CX + DU$$

- > Observations z
- > Measurement matrix C
- > Feed-forward matrix **D**

Review of State Space Concepts

▶ When describing a non-linear system, the state space equations are written as:

Non-Linear State Equation

$$\dot{\mathbf{X}} = f(\mathbf{X}, \mathbf{U})$$

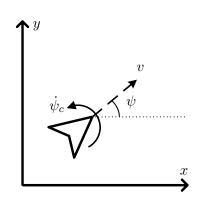
Non-Linear Observation Eqn.

$$\mathbf{z} = h(\mathbf{X}, \mathbf{U})$$

- ▷ Developing navigation algorithms for robots usually begins with representing the system in this form.
- ▶ For now, only the modeling of the state equation is considered. The observation equation will be revisited when discussing sensor fusion techniques.

- Describe how the vehicle's position (x, y) and heading (ψ) are affected by controlling the forward velocity (v) and yaw rate $(\dot{\psi}_c)$
- ▷ Express the system in non-linear state space form:

$$\dot{\mathbf{X}} = f(\mathbf{X}, \mathbf{U})$$



▶ First define the state vector:

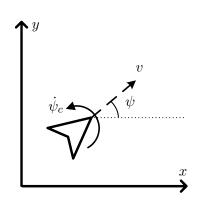
State Vector

$$\mathbf{X} = \begin{bmatrix} x \\ y \\ \psi \end{bmatrix}$$

> ... and the control vector:

Control Vector

$$\mathbf{U} = \begin{bmatrix} v \\ \dot{\psi}_c \end{bmatrix}$$

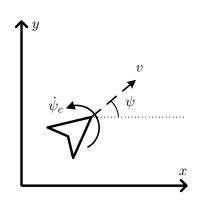


- Need to come up with equations for the derivatives of each state variable.
- \triangleright Project v onto the x axis:

$$\dot{x}=v\cos\psi$$

 \triangleright Project v onto the y axis:

$$\dot{y} = v \sin \psi$$



▶ The actual yaw rate is whatever the command is:

$$\dot{\psi} = \dot{\psi}_c$$

State Vector

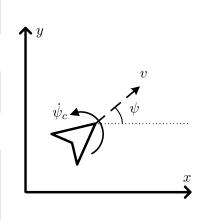
$$\mathbf{X} = \begin{bmatrix} x & y & \psi \end{bmatrix}^T$$

Control Vector

$$\mathbf{U} = \begin{bmatrix} v & \dot{\psi}_c \end{bmatrix}^T$$

State Function

$$f(\mathbf{X}, \mathbf{U}) = \begin{cases} \dot{x} = v \cos \psi \\ \dot{y} = v \sin \psi \\ \dot{\psi} = \dot{\psi}_c \end{cases}$$

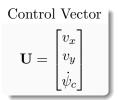


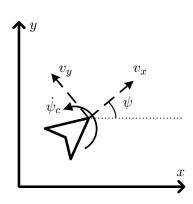
Holonomic Navigation Model

▶ State vector remains the same:

State Vector
$$\mathbf{X} = \begin{bmatrix} x \\ y \\ \psi \end{bmatrix}$$

➤ The control vector has one more signal:





▶ Holonomic vehicles have three degrees of freedom.

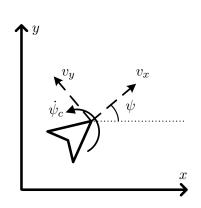
Holonomic Navigation Model

- ▷ Derive expressions for the derivatives of state variables.
- \triangleright Project v_x and v_y onto the x axis:

$$\dot{x} = v_x \cos \psi - v_y \sin \psi$$

 \triangleright Project v_x and v_y onto the y axis:

$$\dot{y} = v_x \sin \psi + v_y \cos \psi$$



▶ The actual yaw rate is whatever the command is:

$$\dot{\psi} = \dot{\psi}_c$$

Holonomic Navigation Model

State Vector

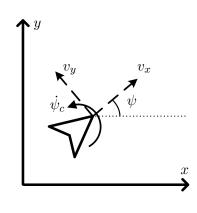
$$\mathbf{X} = \begin{bmatrix} x & y & \psi \end{bmatrix}^T$$

Control Vector

$$\mathbf{U} = \begin{bmatrix} v_x & v_y & \dot{\psi}_c \end{bmatrix}^T$$

State Function

$$f(\mathbf{X}, \mathbf{U}) = \begin{cases} \dot{x} = v_x \cos \psi - v_y \sin \psi \\ \dot{y} = v_x \sin \psi + v_y \cos \psi \\ \dot{\psi} = \dot{\psi}_c \end{cases}$$



- ▶ The kinematic constraints of a car can be modeled simplistically by a bicycle navigation model.
- \triangleright The state vector remains the same as the other vehicles, but the control vector becomes the combination of speed vand steering wheel angle α_s :

State Vector
$$\mathbf{X} = \begin{bmatrix} x \\ y \\ \psi \end{bmatrix}$$

Control Vector
$$\mathbf{U} = \begin{bmatrix} v \\ \alpha_s \end{bmatrix}$$

▷ Recall the relationship between steering wheel angle, yaw rate and speed:

$$\alpha_s = \gamma \tan^{-1} \left(\frac{L\dot{\psi}}{v} \right)$$

 \triangleright Rearranging this, the yaw rate $\dot{\psi}$ can be represented in terms of the control variables v and α_s , with the steering ratio γ and wheelbase L of the vehicle as constant parameters:

$$\dot{\psi} = \frac{v}{L} \tan \left(\frac{\alpha_s}{\gamma} \right)$$

State Vector

$$\mathbf{X} = \begin{bmatrix} x \\ y \\ \psi \end{bmatrix}$$

Control Vector

$$\mathbf{U} = \begin{bmatrix} v \\ \alpha_s \end{bmatrix}$$

State Function

$$f(\mathbf{X}, \mathbf{U}) = \begin{cases} \dot{x} = v \cos \psi \\ \dot{y} = v \sin \psi \\ \dot{\psi} = \frac{v}{L} \tan \left(\frac{\alpha_s}{\gamma}\right) \end{cases}$$

State Space Discretization

- ▶ In order to implement a state space controller or observer in a computer, it must first be *discretized*.
- ▶ Discretization involves approximating the time derivative of the continuous state space equation.
- ➤ There are many techniques to discretize a system, each with different implications when analyzed in frequency domain.

State Space Discretization

➤ The simplest derivative approximation is the <u>Forward Euler</u> method:

Forward Euler Derivative Approximation

$$\frac{dg(t)}{dt} \approx \frac{g(t+\Delta t) - g(t)}{\Delta t}$$

▶ Applying the conventional discrete equation notation:

$$g(t) \triangleq g_k, \quad g(t + n\Delta t) \triangleq g_{k+n}, \quad \Delta t \triangleq T_s$$

$$\left. \frac{dg}{dt} \right|_k = \frac{g_{k+1} - g_k}{T_s} \Rightarrow g_{k+1} = g_k + T_s \left. \frac{dg}{dt} \right|_k$$

State Space Discretization

Equivalent state equations

$$\underbrace{\dot{\mathbf{X}} = f(\mathbf{X}, \mathbf{U})}_{\text{Continuous}} \quad \Rightarrow \quad \underbrace{\mathbf{X}_{k+1} = f(\mathbf{X}_k, \mathbf{U}_k)}_{\text{Discrete}}$$

▶ Apply forward Euler approximation:

$$\left. \frac{d\mathbf{X}}{dt} \right|_k \approx \frac{\mathbf{X}_{k+1} - \mathbf{X}_k}{T_s} \Rightarrow \mathbf{X}_{k+1} = \mathbf{X}_k + T_s \left. \frac{d\mathbf{X}}{dt} \right|_k$$

 \triangleright where $d\mathbf{X}/dt$ can be replaced by the continuous state equation:

$$\mathbf{X}_{k+1} = \mathbf{X}_k + T_s f(\mathbf{X}_k, \ \mathbf{U}_k)$$

▶ Apply the forward Euler approximation: