Fall 2022 ME424 Modern Control and Estimation

Lecture Note 6 Control Design and Testing in Drake with Python

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Outline

- Short introduction to Drake
- Example 1: Observer and Controller Design
- Example 2: Cart-Pole Balancing
- From regulation to tracking control
- Example 3: DC Motor speed tracking control

What is Simulation?

- Real-world physics are often described by functions, ODE or PDE
- All simulators essentially solve the ODEs and/or PDEs corresponding to a physical process of interest
- Three pillars of a simulator:
 - 1. Constructing the differential equations/models

2. Solving differential equations



3. Visualization of the simulation results

Popular simulators in robotics



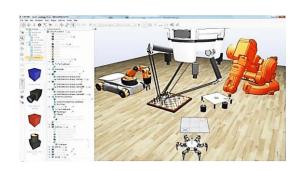
Mujoco (Roboti LLC)



PyBullet (open source)



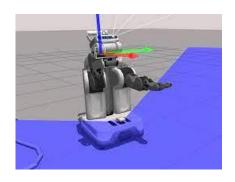
ISAAC (NVIDIA)



V-REP (CoppeliaSim)

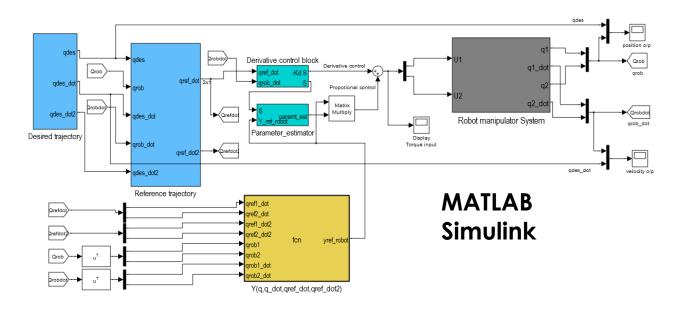


Webot (Cyberbotics)

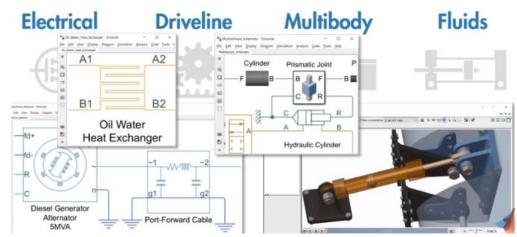


Gazebo

Popular simulators for control systems

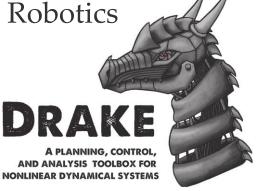






Drake: Model-Based Design and Verification for Robotics

 Happy marriage between MATLAB Simulink with and robotic simulators



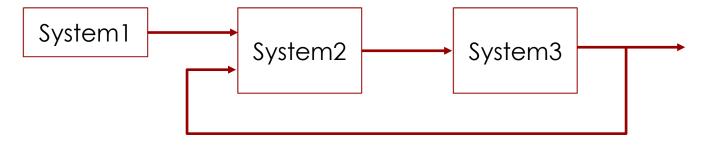
 Great support for dynamic system modeling, optimization, robotic kinematics and dynamics

Very accurate simulation (contact handling)

Visualization is not great but good enough

Open-source and support Python

Drake: Block Diagram Overview



• How to define a system block (static or dynamic)?

How to connect blocks to make the overall diagram?

How to simulate?

Drake: How to Define a Static System?

```
class StaticSysExample(LeafSystem):
    def __init__(self, myParameter):
        LeafSystem.__init__(self)
        self.DeclareVectorInputPort("u1", BasicVector(num_input1))
        self.DeclareVectorInputport("u2", BasicVector(num_input2))
        self.DeclareVectorOutputPort("y", BasicVector(num_output), self.CalcOutputY)

def CalcOutputY(self, context, output):
    u1 = self.get_input_port(0).Eval(context)
    u2 = self.get_input_port(1).Eval(context)
    y = "your output function"
    output.SetFromVector(y)
```

BasicVector:

- a = BasicVector(3) # 3-d vector
- a.SetFromVector([arrray]) # from numpy array to Basic vector
- a.CopyToVector() # from BasicVector to numpy array

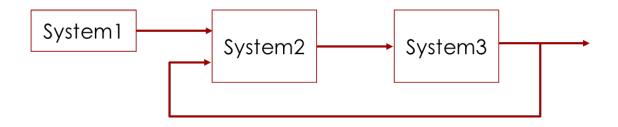
Drake: How to Define a Continuous-Time Dynamic System?

```
class CTSysExample(LeafSystem):
   def init (self, myParameter):
       LeafSystem.__init__(self)
       self.DeclareContinuousState(num state)
       self.DeclareVectorInputPort("u", BasicVector(num input))
        self.DeclareVectorOutputPort("y", BasicVector(num output), self.CalcOutputY)
   def DoCalcTimeDerivatives(self, context, derivatives):
       x = context.get_continuous_state_vector().CopyToVector()
       u = self.get_input_port(0).Eval(context)
       xdot = "your vector field"
       derivatives.get mutable vector().SetFromVector(xdot)
   def CalcOutputY(self, context, output):
       x = context.get_continuous_state_vector().CopyToVector()
       y = "your output function"
       output.SetFromVector(y)
```

Drake: How to Define a Discrete-Time Dynamic System?

```
class DTLinearSys(LeafSystem):
    def init (self):
        LeafSystem.__init__(self)
        self.DeclareDiscreteState(num state)
        self.DeclareVectorInputPort("u", BasicVector(num_input))
        self.DeclareVectorOutputPort("y", BasicVector(num output), self.CalcOutputY)
        self.DeclarePeriodicDiscreteUpdate(0.5) # dt
    def DoCalcDiscreteVariableUpdates(self, context, events, discrete state):
        x = context.get discrete state vector().CopyToVector()
        u = self.get_input_port(0).Eval(context)
        xnext = 0.8*x + np.sin(u)
        discrete state.get mutable vector().SetFromVector(xnext)
    def CalcOutputY(self, context, output):
        x = context.get discrete state vector().CopyToVector()
        u = self.get input port(0).Eval(context)
        v = x + 2*u
        output.SetFromVector(v)
```

Drake: Block Diagram Construction

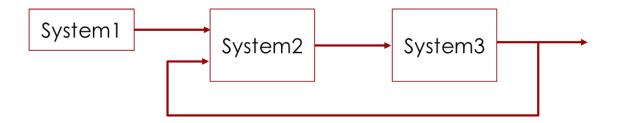


```
builder = DiagramBuilder()
Sys1 = builder.AddSystem(Sys1)
Sys2 = builder.AddSystem(Sys2)
Sys3 = builder.AddSystem(Sys3)

builder.Connect(Sys1.get_output_port(0), Sys2.get_input_port(0))
builder.Connect(Sys2.get_output_port(0), Sys3.get_input_port(0))
builder.Connect(Sys3.get_output_port(0), Sys2.get_input_port(1))

logger_output = LogOutput(Sys3.get_output_port(0), builder)
diagram = builder.Build()
```

Drake: Simulate a Block Diagram



```
simulator = Simulator(diagram)
simulator.set_target_realtime_rate(1)
context = simulator.get_mutable_context()
context.SetContinuousState(CT_x0)
context.SetDiscreteState(DT_x0)
simulator.AdvanceTo(sim_time)
```

Regarding **context** class:

- get_continuous_state_vector()
- get_discrete_state_vector()
- get_mutable_continuous_state_vector()
- get_mutable_discrete_state_vector()

- SetContinuousState(..)
- SetDiscreteState(..)

Drake: Simple Simulation Examples

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Example 1: design output feedback controller for plant

$$A_c = \begin{bmatrix} 33 & -60 \\ 20 & -33.2 \end{bmatrix}, \qquad B_c = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \qquad C_c = \begin{bmatrix} 2 & 1 \end{bmatrix}$$

• **Step 1:** Discretization: e.g. with sampling time T = 0.01

$$A_d = (I + A_c T) = \begin{bmatrix} 1.33 & -0.6 \\ 0.2 & 0.668 \end{bmatrix}, \qquad B_d = B_c T = \begin{bmatrix} 0.01 \\ 0.01 \end{bmatrix}, \qquad C_d = C_c = \begin{bmatrix} 2 & 1 \end{bmatrix}$$

• Step 2: Select desired closed-loop eigs (suppose we want the continuous time poles: $s_{1,2} = \{-2+j, -2-j\}$

Step 3: Design feedback gain K

• Step 4: Observer eigs: suppose we want: observer_ $s_{1,2} = \{-8+j, -8+j\}$

• Step 5: Observer gain *L*

Observer dynamical system:

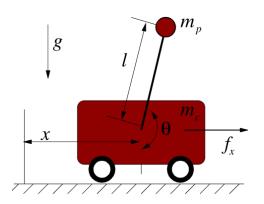
$$\hat{x}_{k+1} = A\hat{x}_k + Bu_k + L(y_k - C\hat{x}_k) = (A - LC)\hat{x}_k + [B \ L]\begin{bmatrix} u_k \\ y_k \end{bmatrix}$$

Simulation Testing

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Cart-Pole Model



$$egin{aligned} \ddot{x} = & rac{1}{m_c + m_p \sin^2 heta} \Big[f_x + m_p \sin heta (l \dot{ heta}^2 + g \cos heta) \Big] \ \ddot{ heta} = & rac{1}{l(m_c + m_p \sin^2 heta)} \Big[-f_x \cos heta - m_p l \dot{ heta}^2 \cos heta \sin heta - (m_c + m_p) g \sin heta \Big] \end{aligned}$$

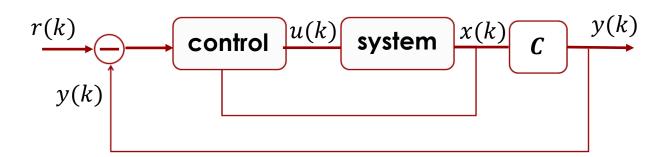
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Robust tracking problem:

$$x(k+1) = Ax(k) + Bu(k) + B_d d(k)$$
$$y(k) = Cx(k)$$

- d(k): disturbance entering the systems
- Goal: design u to make output y(k) track a reference r(k)
- To simplify discussion, we assume:
 - r(k) and y(k): scalar
 - u(k): full state feedback (add observer if full state is not available)



• Illustrating example: consider a simple scalar system

$$x(k + 1) = a x(k) + u(k),$$
 $y(k) = x(k)$
Suppose tracking reference: $r(k) = r \neq 0$

• Linear feedback doesn't work: $u(k) = -Kx(k) \Rightarrow x(k+1) = (a-K)x(k)$

• Can we compute the correct input? E.g. assume |a| < 1, then $x(k) = a^k x_0 + \sum_{j=0}^{k-1} a^{k-j-1} u(j)$

Challenges for more general tracking problems:

$$x(k+1) = Ax(k) + Bu(k) + B_d d(k)$$
$$y(k) = Cx(k)$$

- $x(k) \in \mathbb{R}^n$ can be multi-dimensional $(a \to A)$, so can input u(k)
- System may be unstable
- Uncertainties:
 - reference r(k) may not be know a priori
 - we have nontrivial **unknown** disturbance d(k)
- How to improve transient performance (track promptly)

■ Introduce an `integral state'': z(k + 1) = z(k) + r(k) - y(k)

- Extended state space: $\tilde{x}(k) = \begin{bmatrix} x(k) \\ z(k) \end{bmatrix}$
- Feedback control: $u(k) = \begin{bmatrix} K_x & K_z \end{bmatrix} \begin{bmatrix} x(k) \\ z(k) \end{bmatrix}$
- CL dynamics: $\tilde{x}(k+1) = \begin{bmatrix} A & 0 \\ -C & 1 \end{bmatrix} \tilde{x}(k) + \begin{bmatrix} B \\ 0 \end{bmatrix} u(k) + \begin{bmatrix} B_d d(k) \\ r(k) \end{bmatrix}$

• Under mild conditions: (\tilde{A}, \tilde{B}) is controllable so we can design \tilde{K} such that $\tilde{A} + \tilde{B}\tilde{K}$ has desired eigenvalues

• Closed-loop dynamics:
$$\tilde{x}(k+1) = (\tilde{A} + \tilde{B}\tilde{K})\tilde{x}(k) + \begin{vmatrix} B_d d(k) \\ r(k) \end{vmatrix}$$

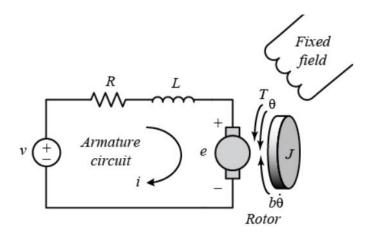
- For constant or slowly changing $d(k) \approx d$, $r(k) \approx r$,
 - the extended state $\tilde{x}(k)$ converges to a finite value.
 - Thus, $z(k + 1) = z(k) \Rightarrow$ error becomes zero

Proof and Discussions

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DC Motor Speed Tracking Control Example



Summary