

Tuesdays & Thursdays  
12:00-13:30 ET  
GGBL 2147

*Hello*

# **ME599-004: Data-Driven Methods for Control Systems**

Winter 2024

Instructor: Uduak (*Who-dwak*) Inyang-Udoh



## Uduak Inyang-Udoh

Assistant Professor

3468 GGBL

Postdoc Research Associate



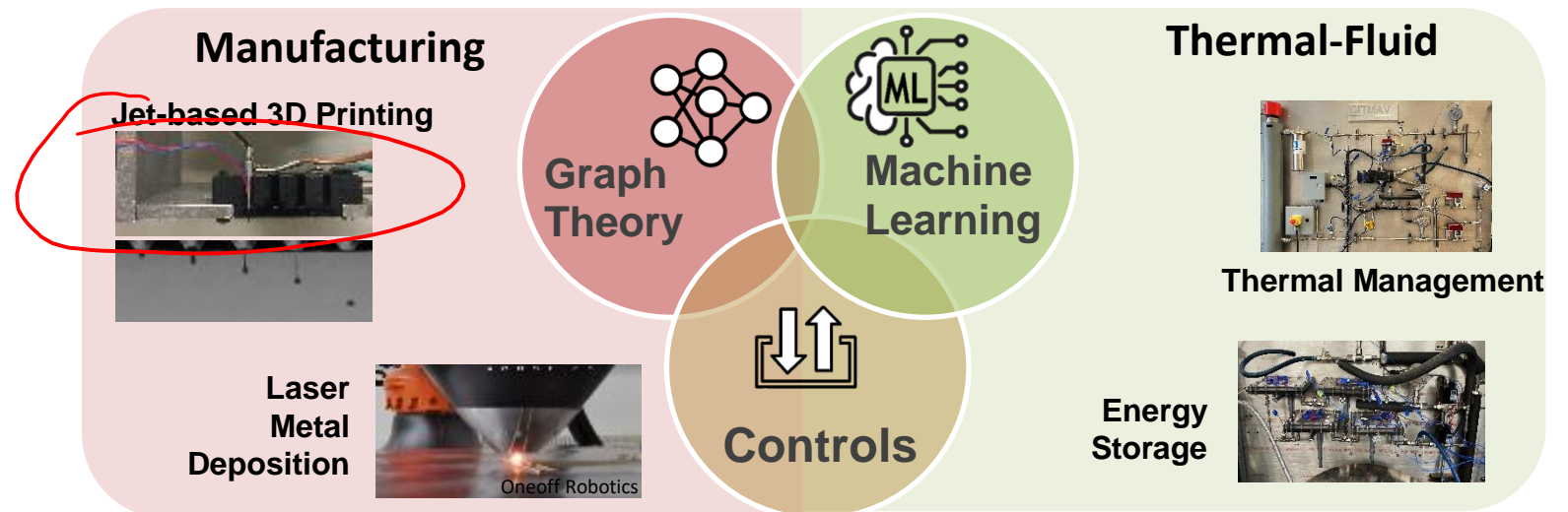
Ph.D. in ME '21



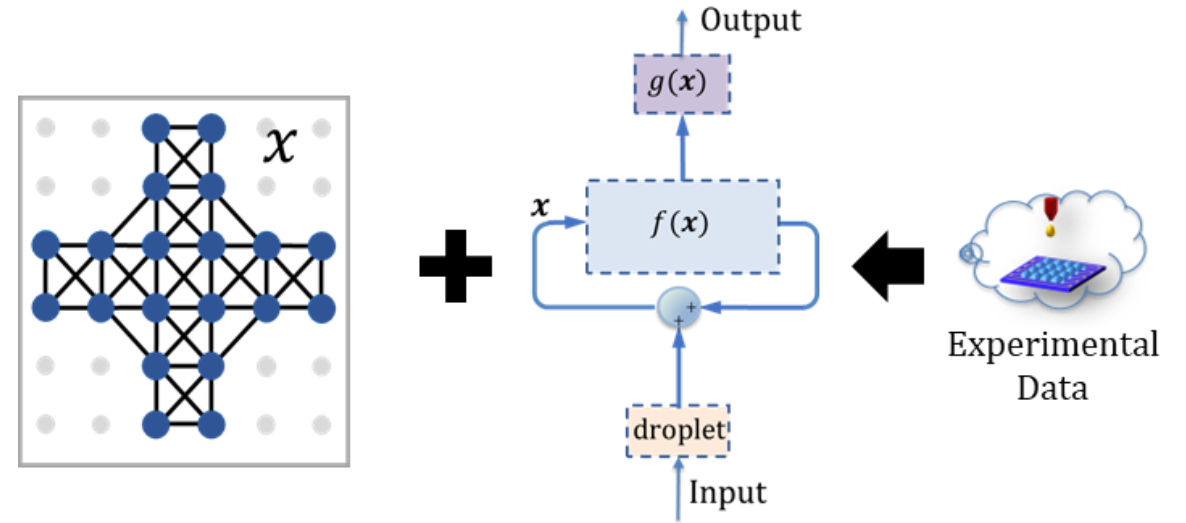
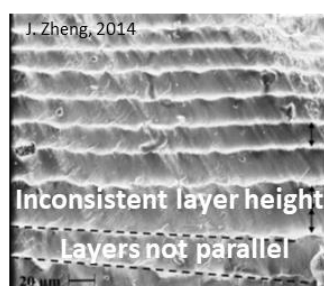
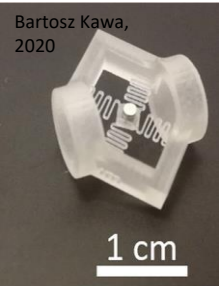
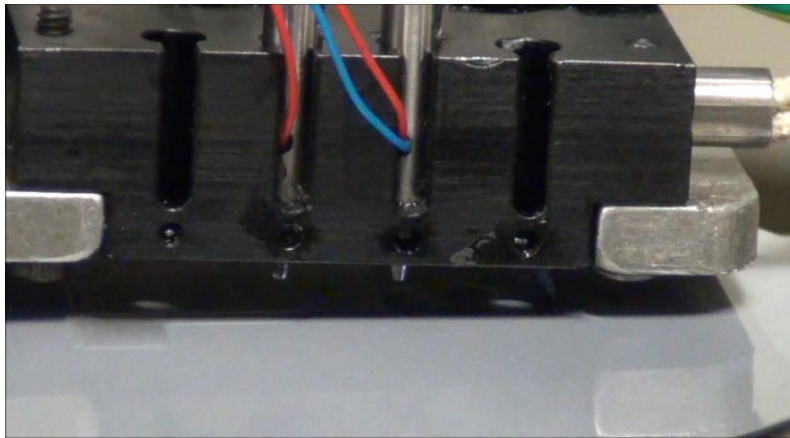
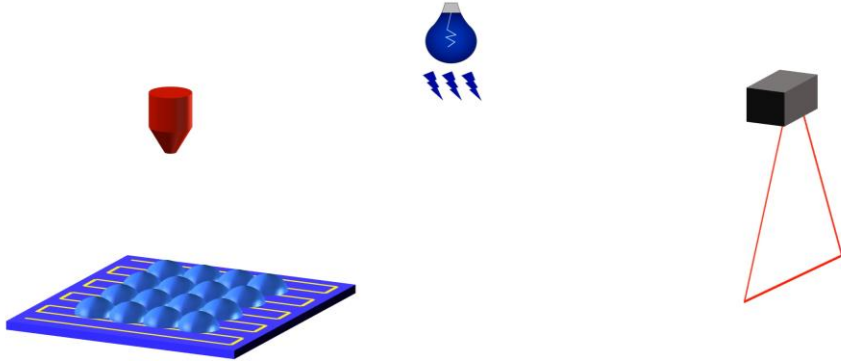
## Autonomous & Intelligent Systems Lab

### Research

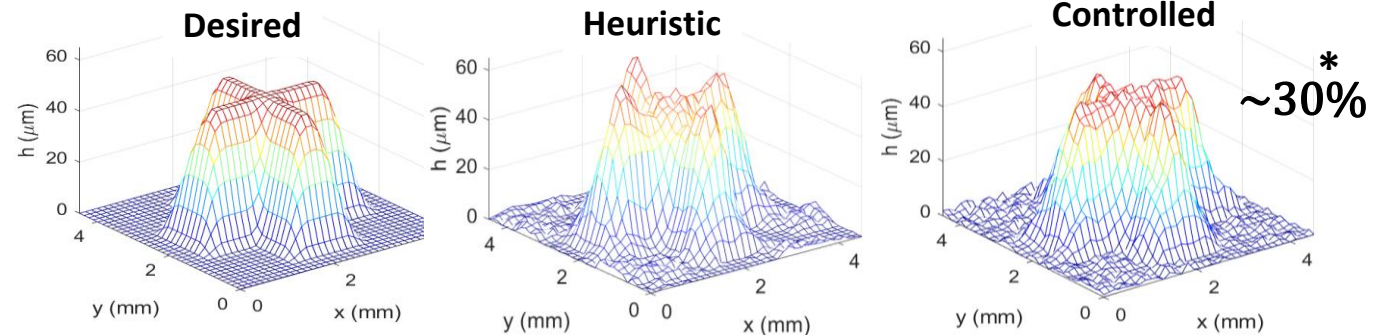
- graph theory + physics-guided machine learning for controls;
- *applications*: advanced manufacturing, thermal-fluid systems.



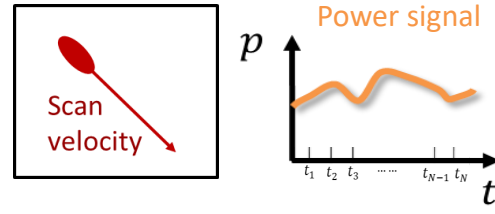
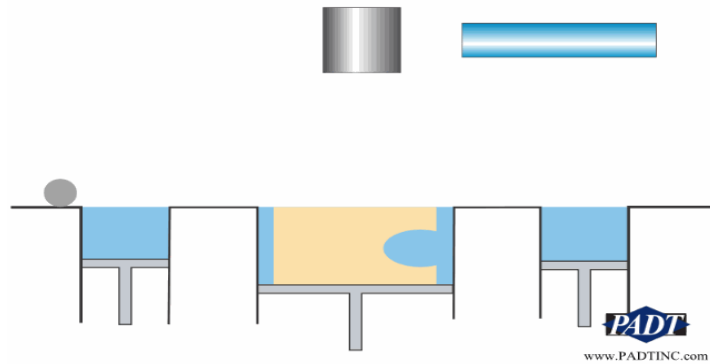
# Input Pattern Control in Inkjet 3D Printing



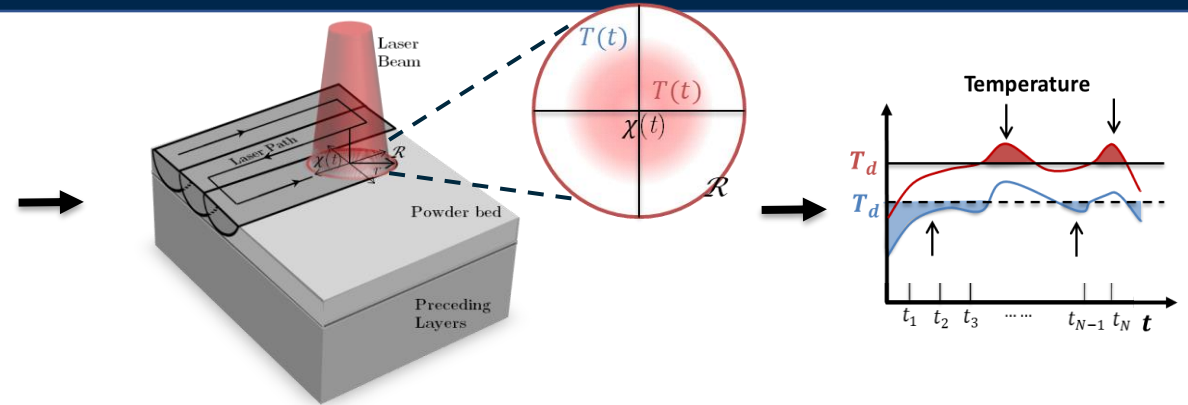
- Learn dynamics
- Lends itself to **in-process learning & control**



# Input Shaping in Laser-Based Additive Manufacturing



Dynamically modulate input?



Optimize:  $\arg\min_p E$

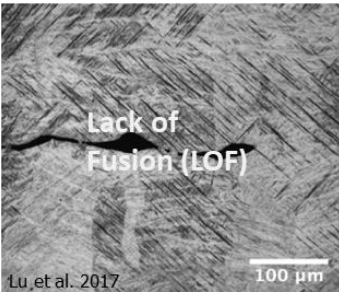
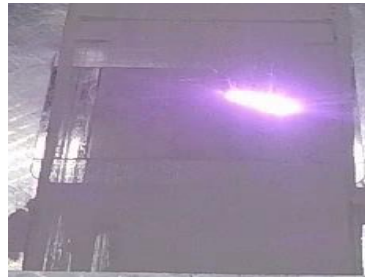
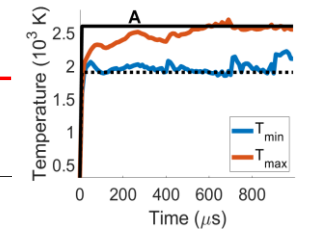
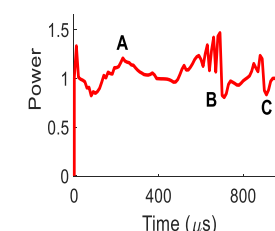
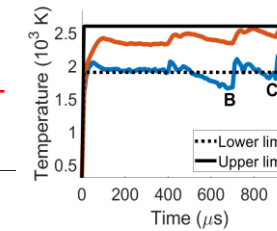
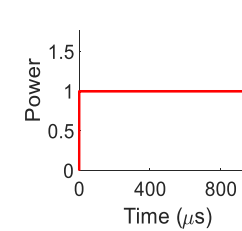
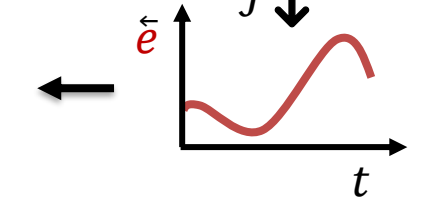
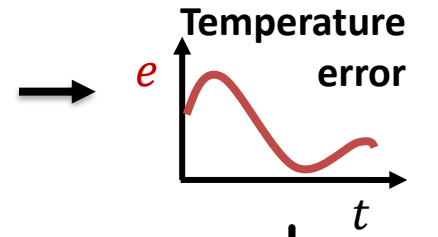
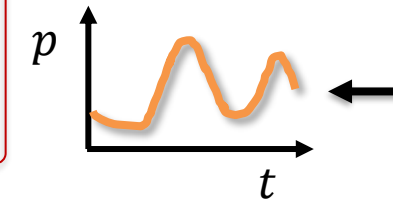
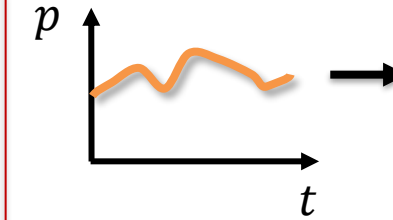
Gradient Descent:

$$p^{new} \leftarrow p^{old} - \alpha G^* e^{old}$$

adjoint

$$p \xrightarrow{G} E$$

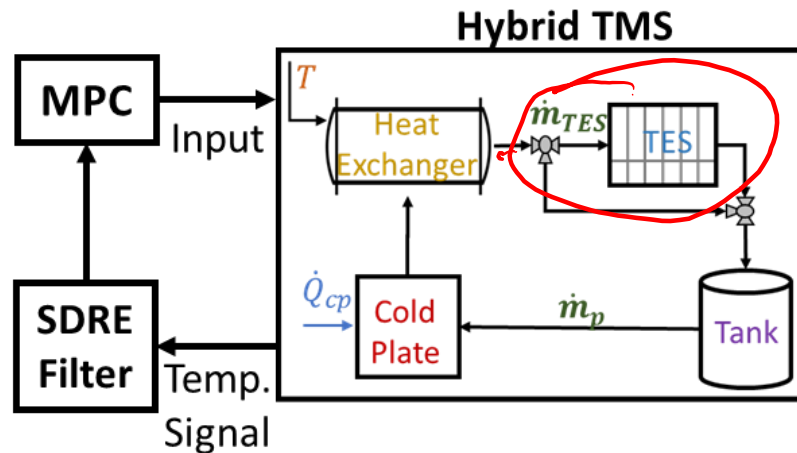
- Unknown, complex
- High dimensional





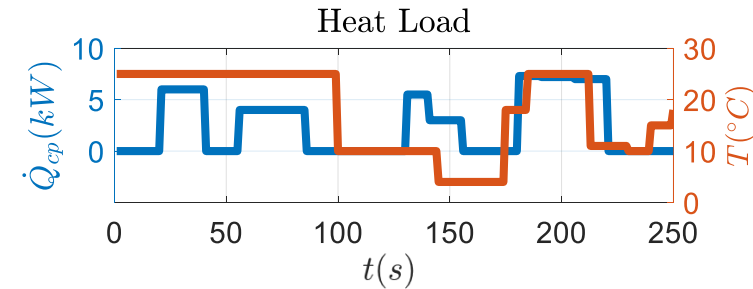
# Control in Hybrid Thermal Management

A TMS with a Thermal energy storage (TES) device → “hybrid TMS”

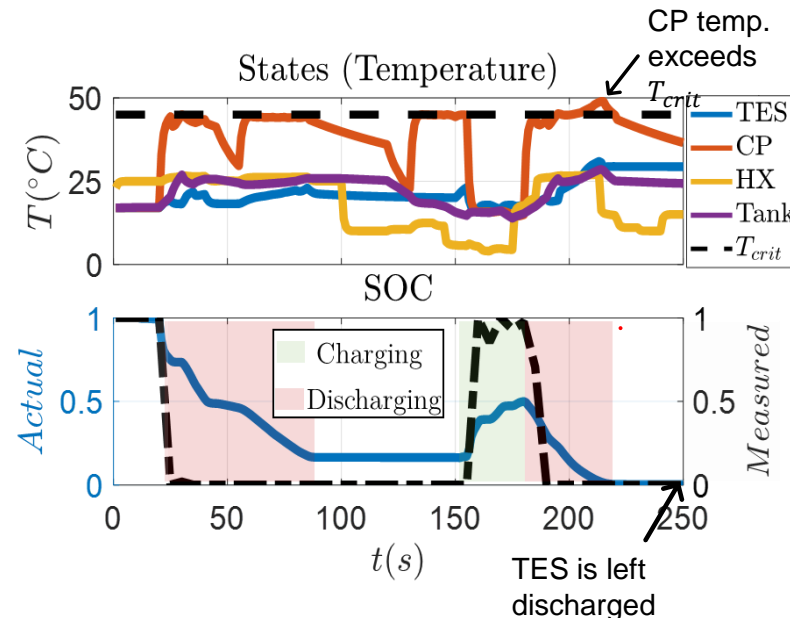


State of charge (SOC): remaining energy storage capacity of TES

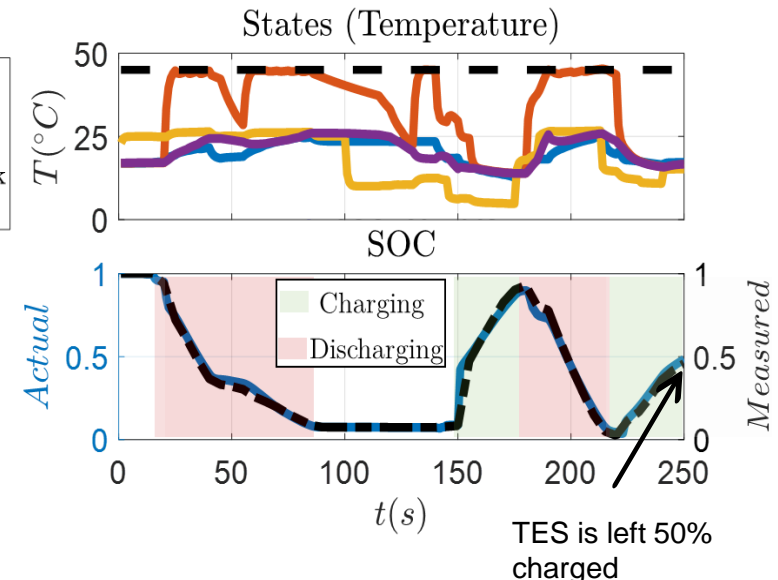
- Designed to keep cold plate (CP) temp. below  $T_{crit}$  while coordinating charging of TES.
- In presence of heat loads  $Q_{cp}$  and chiller temperature  $T_{ch}$



Without SOC Estimate



With SOC Estimate



# Would be nice to ...



- Be familiar with writing codes

have a background

- classical controls (basic knowledge of controls),
- linear algebra (mostly the matrix algebra part)

...even better, to have knowledge of

- state-space modeling

# Assessment



- Homework Problems – 80%
- (Group) Project/Study – 20%
- Quizzes/additional homework problems (optional) – 5% (Bonus points)

# Homework



- Eight (8) problem sets
- At least, one (1) week to complete each
- Mostly entail programming
- Submission report should include:
  - *Method(s)*
  - *Assumption(s)*
  - *Results (results will be graded based on rubric on Canvas)*
- Refer to document on Canvas for more information about homework submission & academic integrity



# (Group) Project/Study



- Exercise the knowledge gained from this course and explore topics that may be beneficial to your graduate research
- Expected to work in teams of 2 or 3, or justify working alone
- Propose topic (I will also have a list of few suggestions) and submit abstract (*tentatively* by March 18)
- Deliverables: Project reports, short presentation
- Grading: Peer review based on rubric provided by instructor

# Bonus Quizzes/Additional Homework Problems (Optional)



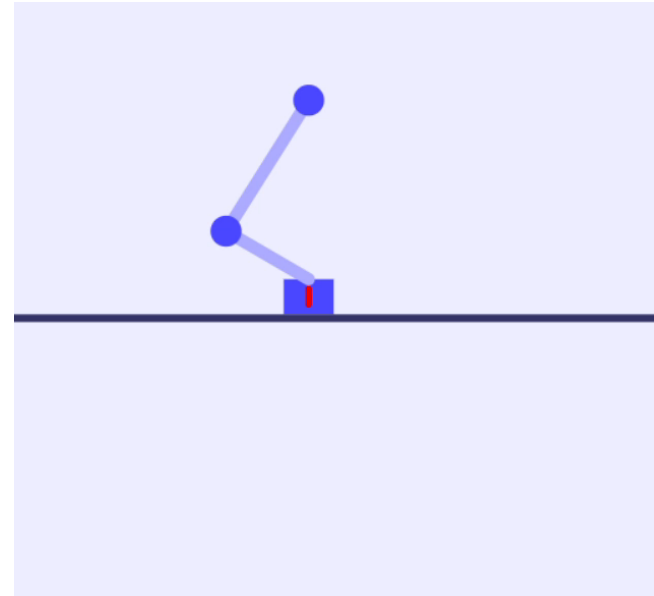
- Few in-class quizzes (optional), time permitting; or
- Optional homework problems

# References



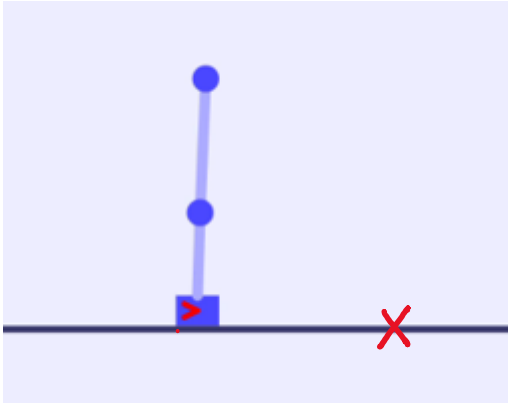
1. Brunton and Kutz, *Data-Driven Science and Engineering: Machine Learning, Dynamical Systems, and Control*, 2nd ed., Cambridge University Press, 2023
2. Ljung, *System Identification: Theory for the User*, 2nd ed., Prentice Hall, 1999
3. Keesman, *System Identification: An Introduction*, Springer, 2011

# Why this Course?

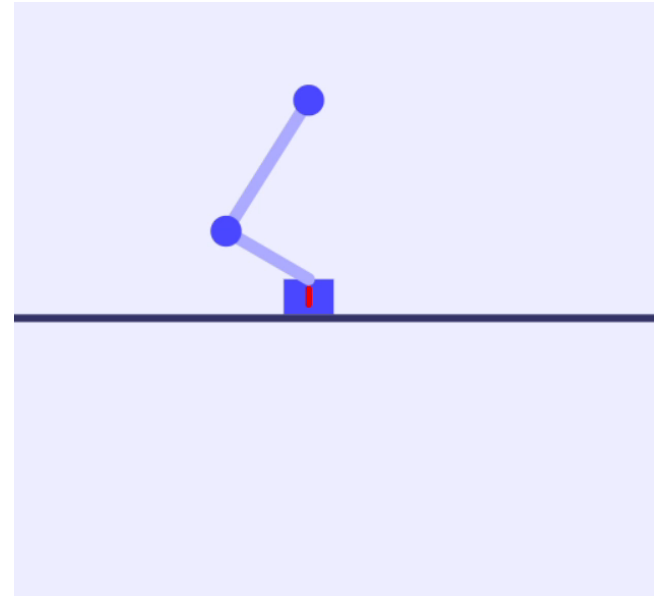


Double inverted pendulum on a guide rail

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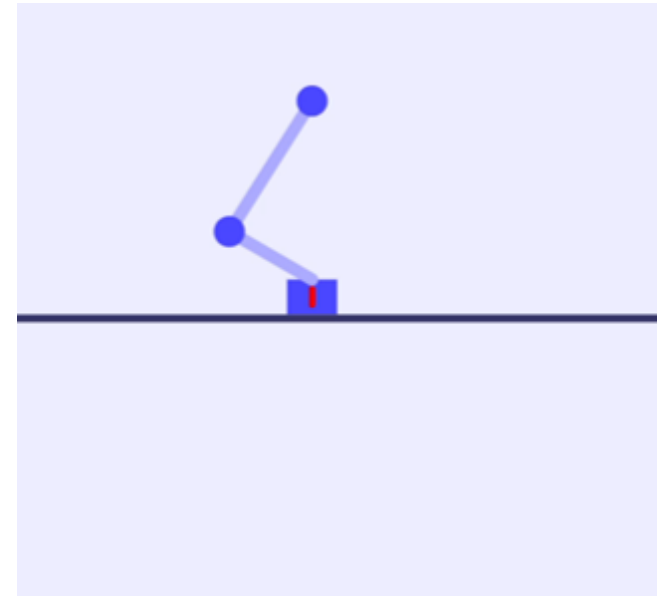
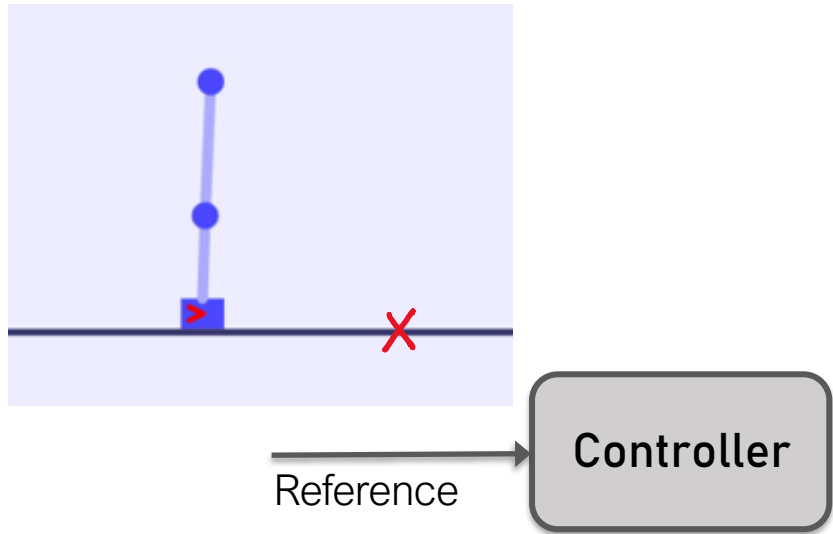


Reference



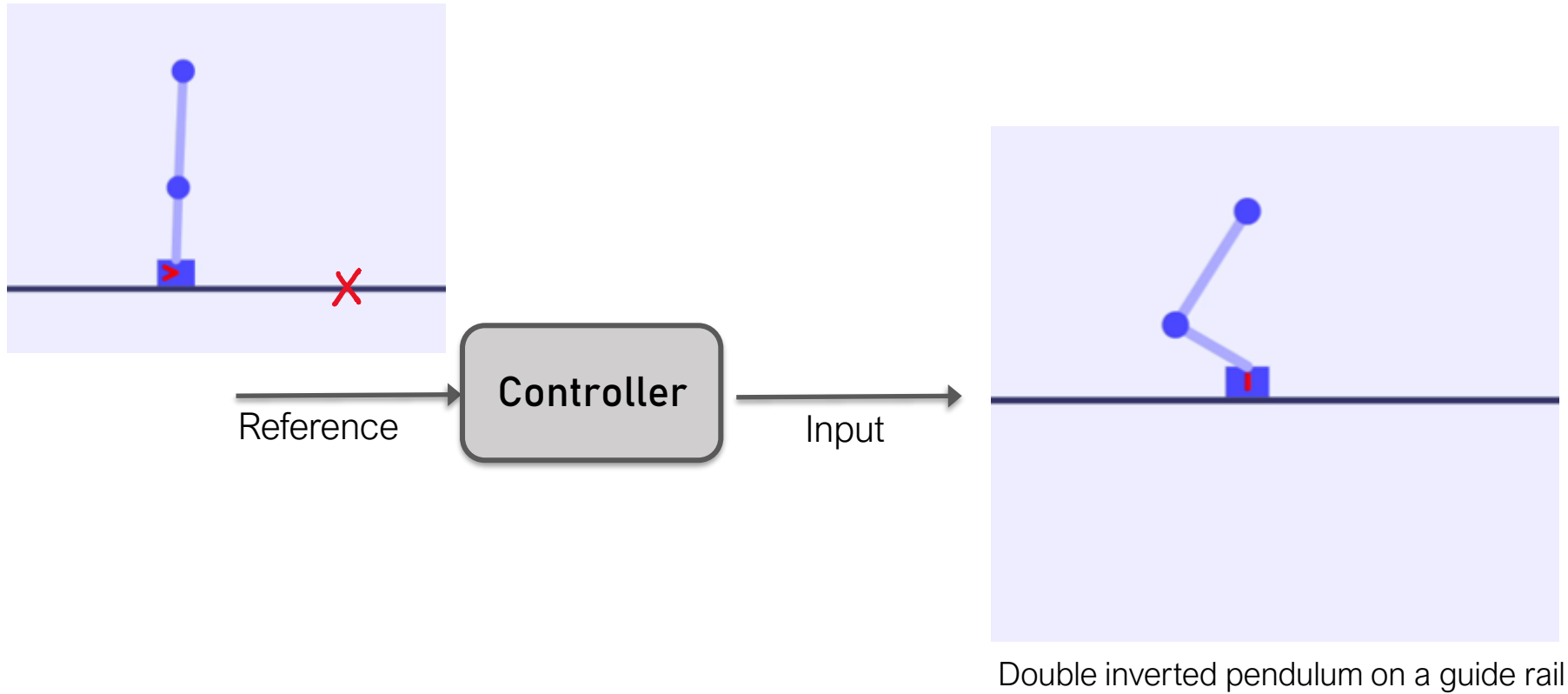
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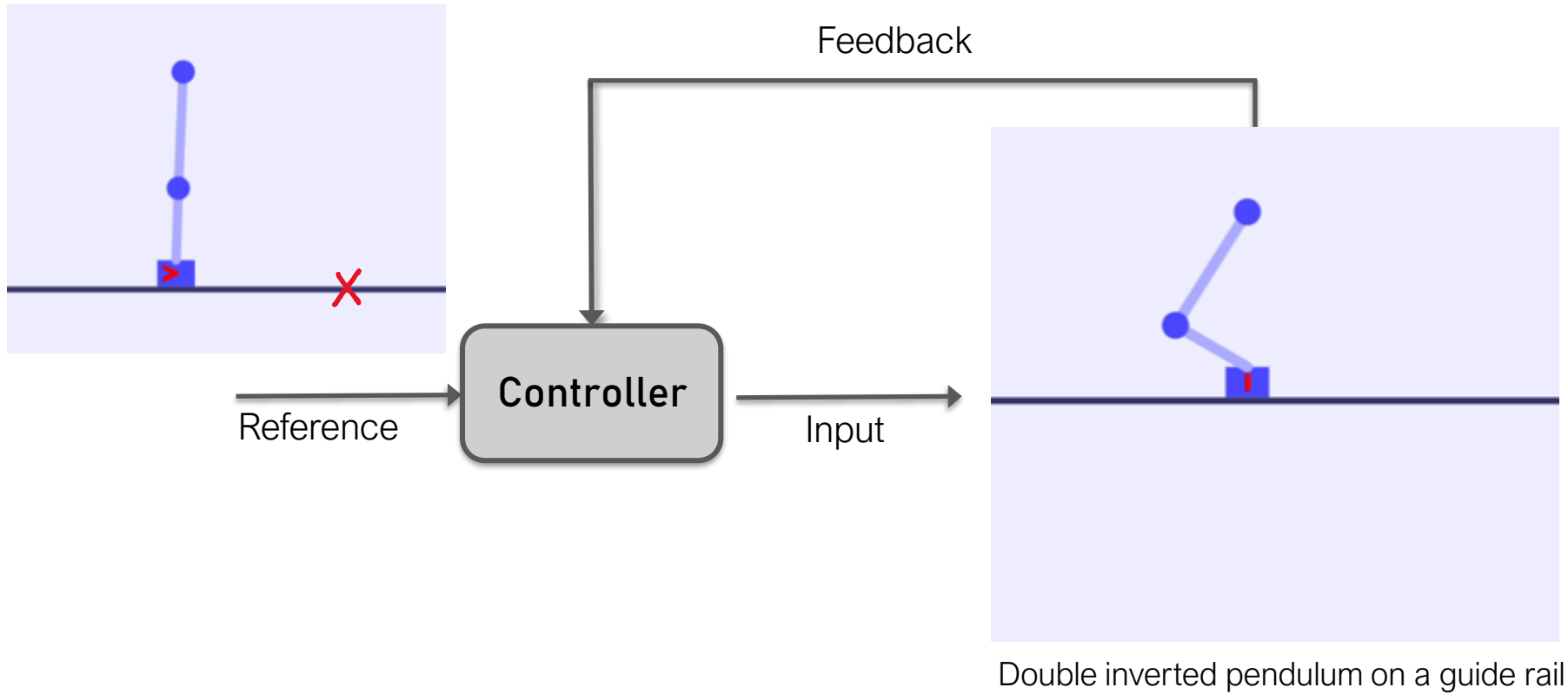
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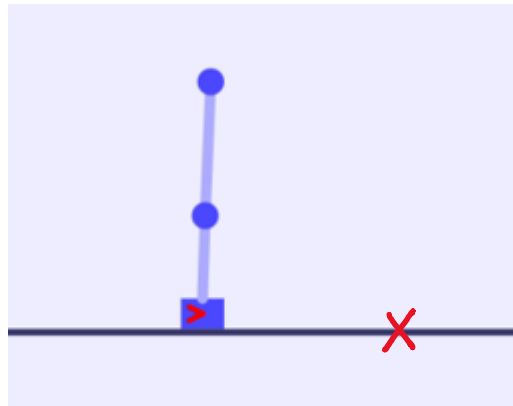




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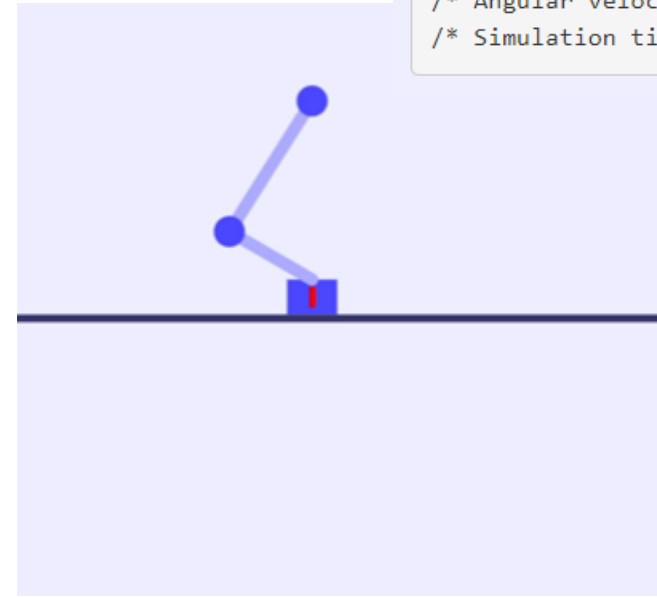
# Why this Course?



Reference

Controller

Input

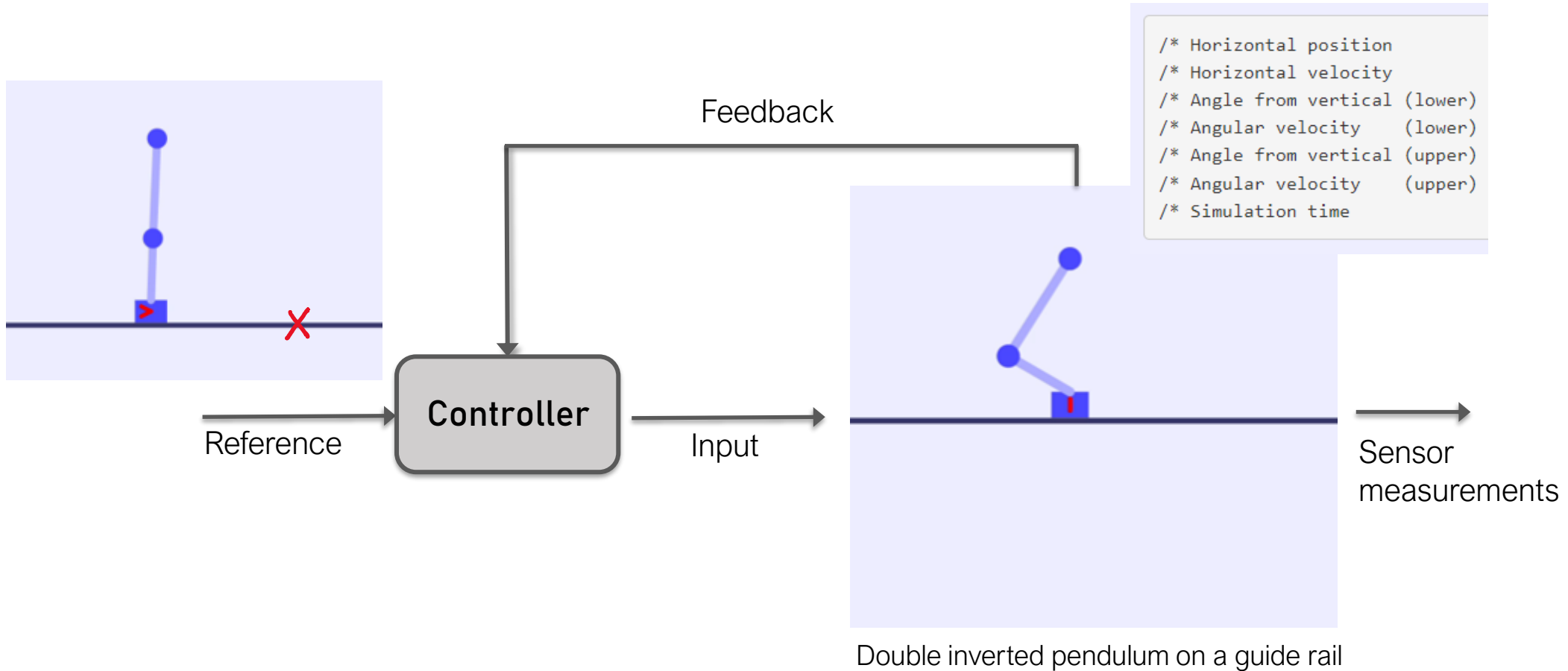


Double inverted pendulum on a guide rail

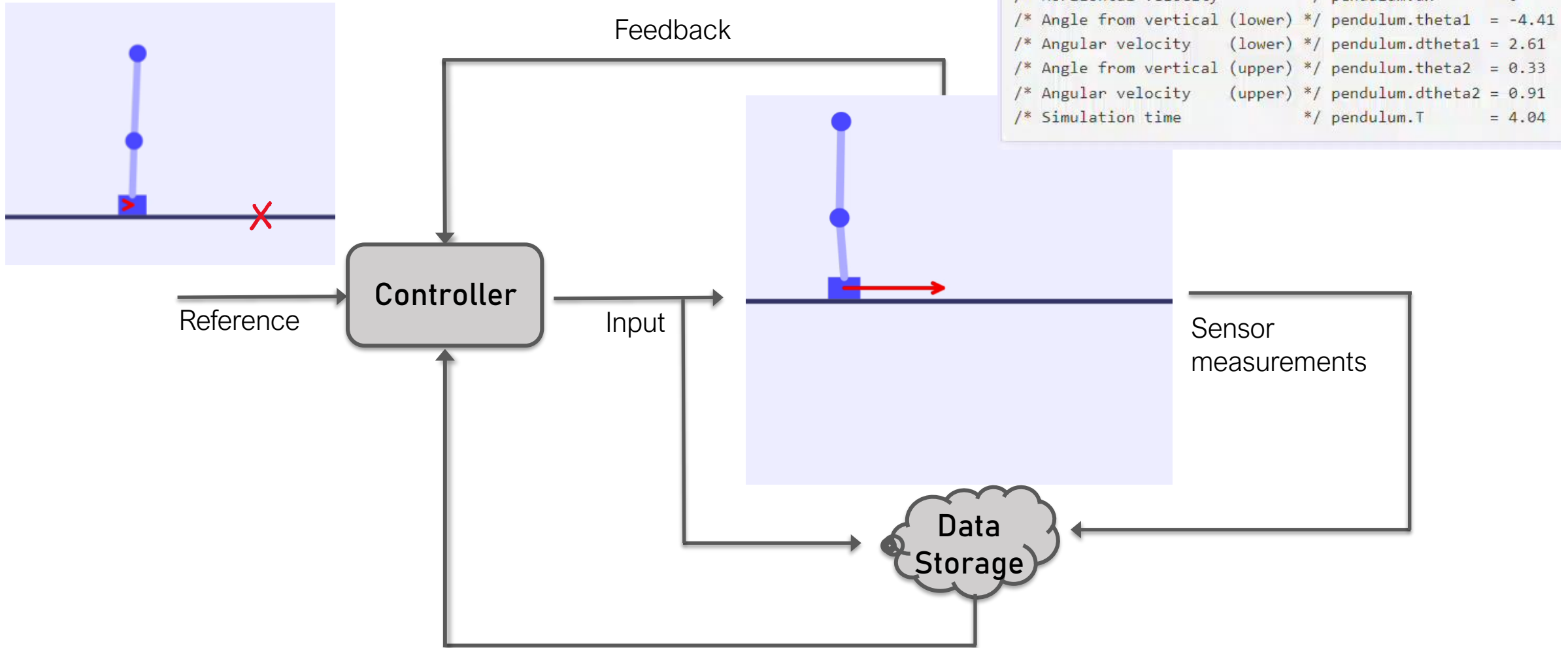
Sensor  
measurements

```
/* Horizontal position  
/* Horizontal velocity  
/* Angle from vertical (lower)  
/* Angular velocity (lower)  
/* Angle from vertical (upper)  
/* Angular velocity (upper)  
/* Simulation time
```

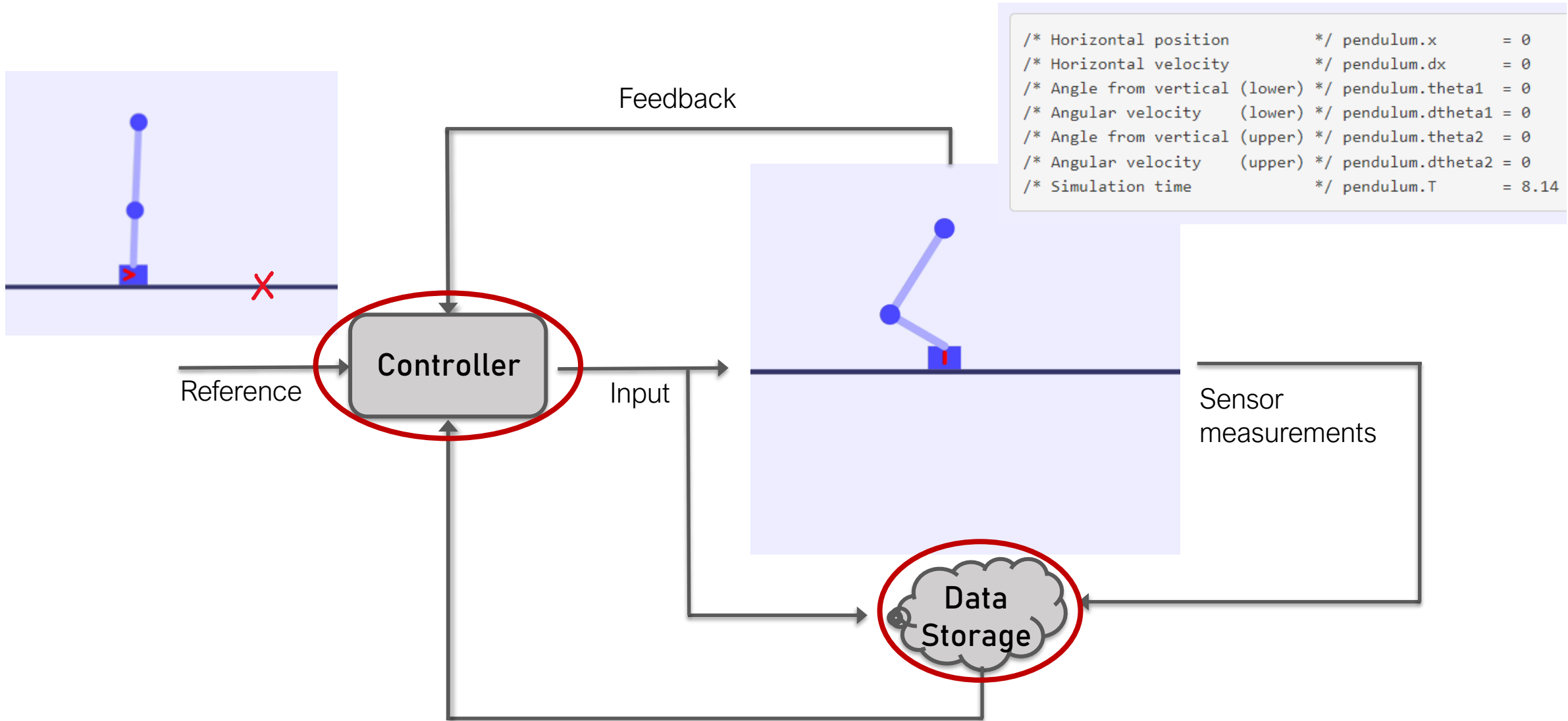
# Why this Course?



# Why this Course?



# Why this Course?



*introduce* approaches for controlling systems using data

## 1. Data Analysis & Machine Learning Preliminaries

- Fourier Transforms & Applications, Dimensionality Reduction
- Regression Analysis: Least Squares, Nonlinear regression
- Review of unconstrained optimization
- Classification: Discriminants, Support Vector Machine, Clustering, Neural Networks

## 2. Dynamical Systems & Control (P)review

- ODE's, Transfer functions, State Space representations
- Classical and linear controls theories; nonlinear control

## 3. Learning Control

- Iterative Learning Control
- Reinforcement Learning (RL)

## 4. Learning Models for Controls

- Classical System Identification (ID) and Controls
- Koopman Operator, Dynamic Mode Decomposition (DMD), Sparse Identification of Nonlinear Dynamics (SINDy)

## 5. Deep Learning for Controls & Recent Topics

- Deep neural networks
  - Multilayer NN,
  - Recurrent NNs,
  - Convolutional NN,
  - Autoencoders,
  - Generative Adversarial Networks
- Physics-Informed Deep Learning

Prof Daniel Broude





# Quick Linear Algebra Review

# Vector (Linear) Spaces

A Vector Space  $(V, F)$  is a set of vectors  $V$  and a field of scalars  $F$ , along with two operations: vector addition  $(+)$  and scalar multiplication  $(\cdot)$ ; such that

*Addition*  $(+)$  :

- (i) associative  $(x + y) + \cancel{z} = x + (y + z) \quad \forall x, y \in V.$
- (ii) commutative  $x + y = y + x.$
- (iii)  $\exists$  additive identity  $0 \in V$  such that  $x + 0 = 0 + x = x.$
- (iv)  $\exists$  additive inverse, i.e.,  $\forall x \in V, \exists(-x)$  such that  $x + (-x) = 0$

$\mathbb{R}, \mathbb{C}$

# Vector (Linear) Spaces

*Scalar Multiplication*  $(\cdot)$  :

(v)  $(\alpha\beta) \cdot x = \alpha \cdot (\beta \cdot x) \quad \forall x \in V, \quad \forall \alpha, \beta \in F.$

(vi)  $1 \cdot x = x$ , where 1 is the multiplicative identity for the field  $F$ .

(vii)  $0 \cdot x = 0$ , where 0 is the additive identity for the field  $F$ .

(viii) distributive (1)  $\forall x \in V, \forall \alpha, \beta \in F (\alpha + \beta) \cdot x = \alpha \cdot x + \beta \cdot x.$

(ix) distributive (2)  $\forall x, y \in V, \forall \alpha \in F \alpha \cdot (x + y) = \alpha \cdot x + \alpha \cdot y.$

Examples  $(\mathbb{R}, \mathbb{R}), (\mathbb{R}^n, \mathbb{R}), (\mathbb{R}^n, \mathbb{C})$   
 $\downarrow \quad \downarrow \quad \downarrow \quad \downarrow$   
 $\mathbb{V} \quad \mathbb{F} \quad \mathbb{V} \quad \mathbb{F}$

# Subspaces



Let  $(V, F)$  be a linear space (vector space) and  $W \subset V$ . Then,  $(W, F)$  is called a *subspace* of  $(V, F)$  if  $(W, F)$  itself is a vector space (with the same inherited operations).

Ex. •  $W = \mathbb{R}^+$  and  $F = \mathbb{R}$  Is  $(W, F)$  a subspace of  $(V, F)$  where  $V = \mathbb{R}$ ?

$$\alpha = 3 \quad \beta = 2$$

$$\alpha = 1 \quad \beta = -8$$

$$\alpha v_1 + \beta v_2 = 6 - 8 = -2 \notin \mathbb{R}^+$$

# Linear Independence



Suppose  $(V, F)$  is a vector space. The set of vectors  $\{v_1, v_2, \dots, v_p\}$  is said to be *linearly independent* iff  $\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_p v_p = 0 \Leftrightarrow \alpha_1 = \alpha_2 = \dots = \alpha_p = 0$ .

The set of vectors is said to be linearly dependent iff  $\exists$  scalars  $\alpha_1, \alpha_2, \dots, \alpha_p$  not all zero, such that,  $\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_p v_p = 0$ .

# Basis

Suppose  $(V, F)$  is a linear space. Then a set of vectors  $B = \{b_1, b_2, \dots, b_n\}$  is called a *basis* if

- (i)  $\{b_1, b_2, \dots, b_n\}$  spans  $V$ ; and
- (ii)  $\{b_1, b_2, \dots, b_n\}$  is a linearly independent set.

Coordinate Representation of Vectors

$$x \in V, \quad x = \sum_{i=1}^n \xi_i b_i \quad \xi_i \text{ is a scalar}$$

$$x \in \mathbb{R}^n$$

$$\begin{bmatrix} x \\ y \end{bmatrix} = \alpha \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \beta \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

