6.01 Introduction to EECS via Robotics

Lecture 2: Signals and Systems

Lecturer: Adam Hartz (hz@mit.edu)

As you come in...

- Grab one handout (on the table by the entrance)
- Please sit near the front!

The **intellectual themes** in 6.01 are recurring themes in engineering:

- design of complex systems
- modeling and controlling physical systems
- augmenting physical systems with computation
- building systems that are robust to uncertainty

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In short, 6.01 is about engineering design.

system behavior

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behavior system

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analysis behavior system

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6.01: Technical Content

6.01 is organized into four modules:

- Signals and Systems
- Circuits
- Probabilistic Reasoning
- Artificial Intelligence / Algorithms

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Signals and	Circuits	Probabilistic	AI/
Systems		Reasoning	Algorithms

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- Circuits
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Programming +						
Signals and Systems	Circuits	Probabilistic Reasoning	AI/ Algorithms			

6.01 Structure

- Lecture: Mon, 9:30a-11:00a, in 32-144
- Software Lab: Mon, 11:00a-12:30p, in 34-501
 - practice with material from lecture and readings
 - preparation for design lab
 - useful self-check of understanding
- Design Lab: Wed, 9:30a-12:30p, in 34-501
 - work with a partner (new partner each week)
 - more open-ended interaction with material
 - checkoffs in lab
- Homework: Weekly online readings and exercises
 - Released Monday mornings
 - Due Sunday nights (11pm)
- Midterm Exams: Two 2-hour exams (on calendar)
- Final Exam

6.01: Pedagogy

Most of the learning is in the lab!

- active learning with hands-on exercises
- open-ended problems with multiple correct solutions
- multiple levels of individualized help (partners, LAs, TAs, Instructors)

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Like anything else, you get good by practicing! Labs and homeworks are an opportunity for *learning*; they are not tests!

6.01 Logistics

Course web site: https://mit.edu/6.01

- calendar
- grading policy
- collaboration policy
- reference material
- other items

6.01: Big Ideas

The **intellectual themes** in 6.01 are recurring themes in engineering:

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← Programming →						
Signals and Systems	Circuits	Probabilistic Reasoning	AI/ Algorithms			

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Approach: focus on key concepts to pursue in depth

← Programming → Programmin						
Signals and Systems	Circuits	Probabilistic Reasoning	AI/ Algorithms			

Focus: discrete-time feedback control systems

Modeling and Analyzing Behavior

Consider the "wall finder" problem from Design Lab 1:



In Design Lab 1, we used a "proportional controller" to cause the robot to keep itself a fixed distance from an obstacle.

Which of the following has the right form for the robot's forward velocity in a convergent proportional controller, assuming that k is positive?

- 0. k * (current)
- 1. k * (desired)
- 2. k * (current desired)
- 3. k * (desired current)
- 4. k * (current / desired)
- 5. k * (desired / current)

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Modeling and Analyzing Behavior

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Modeling and Analyzing Behavior

Consider the "wall finder" problem from Design Lab 1:



By tuning the gain k in the proportional controller, we can get drastically different behaviors! Consider the following two behaviors.

Which behavior is better?

- 1. Behavior 1
- 2. Behavior 2
- 3. Both are good
- 4. Both are bad

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Behavior 1 seems sluggish. Behavior 2 overshoots.

Which is better depends largely on context.

Possible metrics: accuracy, speed, overshoot

How can we optimize performance? What are the fundamental limitations on controlling robot?

Modeling and Analyzing System Behavior

Goal: develop a framework for simulating and analyzing behavior

Many pieces:

- mathematical: difference equations, operator notation
- computational: LTI simulator framework
- analysis: responses, system functionals, poles

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- Why? Different perspectives on problem
 - Mathematical model crisp description of behavior; analysis predicts future behavior
 - Computational model experimentally test assumptions; debug thinking underlying mathematical model; match observations with predictions

Modeling and Analyzing System Behavior

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- mathematical: difference equations, operator notation
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- analysis: responses, system functionals, poles
- Why? Different perspectives on problem
 - Mathematical model crisp description of behavior; analysis predicts future behavior
 - Computational model experimentally test assumptions; debug thinking underlying mathematical model; match observations with predictions

Why? – Need ability to test ideas quickly and efficiently, so need simulation and analysis tools to support

Analyzing System Behavior

Consider the "wall finder" again.



Which of the following best represents the robot's measured distance for the example just shown?



5. None of the above

Which of the following best represents the robot's measured distance for the example just shown? **2**.



5. None of the above

Lecture 2 (slide 30)

Signals are functions of a single parameter (time, for our purposes)

- signals represented by capital letters: X
- n^{th} sample of X denoted with lowercase: x[n]

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in

signal

out

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- **linear**: output at time *n* is a linear combination of previous inputs and outputs (system consists only of delays, gains, and adders)
- **time-invariant**: output of the system does not depend on the absolute time at which the system was started

Analyzing System Behavior

Represent **systems** (physical, math, computational) by the way they transform an **input signal** into an **output signal**.


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Systems are Modular!

Example: cell phone link



Thinking about complicated systems is *complicated*.

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Thinking about simpler systems is often simpler.

If analyzing a big system, often easier to think about parts and connections between them

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Framework for thinking about complicated systems:

- Primitives
- Means of Combination
- Means of Abstraction
- Recognizing meaningful Patterns

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Example: Python

- Primitives: +, *, ==, !=, ...
- Combination: if, while, f(g(x)), ...
- Abstraction: def

Primitives:

- fundamental elements on which to build systems
- e.g., ints, floats, simple functions (+, *), tests (==, !=)

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Means of Abstraction:

- ways to capture common pattern, treat as if a primitive
- e.g., def both captures a pattern within a procedure body, but also gives it a name
- abstraction has standard interfaces "plug compatible"
- abstracted components are treated as primitives
 - suppress details of abstraction from use of abstraction
 - rely on specifications of abstraction to define interface

Simple Systems

Wire

Difference Equation:

$$y[n] = x[n]$$



Simple Systems

Gain

Difference Equation:

$$y[n] = k \cdot x[n]$$



Simple Systems

Delay

Difference Equation:

$$y[n] = x[n-1]$$



Addition

An adder can be thought of as a system which takes arbitrarily-many inputs and outputs their sum.

Difference Equation:

$$y[n] = x_1[n] + x_2[n] + \ldots + x_k[n]$$



Check Yourself!

Consider the system described by:

$$y[n] = x[n] - x[n-1]$$

What is the output of this system when its input is the "unit sample" sequence $\Delta?$

$$\delta[n] = egin{cases} 1, & ext{if } n=0; \ 0, & ext{otherwise} \end{cases}$$

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This is often called the **unit sample response** of a system – used in acoustic modeling, radar, ultrasound, control electronics, economics, many other domains.

Convenient for step-by-step analysis

Find y[n] given $x[n] = \delta[n]$: y[n] = x[n] - x[n-1]



6.01 Intro to EECS I

Convenient for step-by-step analysis

Find
$$y[n]$$
 given $x[n] = \delta[n]$: $y[n] = x[n] - x[n-1]$
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6.01 Intro to EECS I

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6.01 Intro to EECS I

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. . .

6.01 Intro to EECS I

Representations: Block Diagrams

Graphical representation of the difference equation.

Represent y[n] = x[n] - x[n-1] with a block diagram:



So far...

Primitives:

- Gain
- Delay
- Adder

Representations:

- Difference Equation
- Block Diagram
- Unit Sample Response

From Samples to Signals

"Lumping" all of the (possibly infinite) samples into a single object (the **signal**) simplifies its manipulation.

This is analogous to:

- representing coordinates in 3-space as points
- representing lists of numbers as vectors in linear algebra
- creating an object in Python

From Samples to Signals

Operators manipulate signals rather than individual samples.



Nodes represent whole signals (e.g., X and Y). Boxes **operate** on those signals:

- Delay: shift whole signal to right by 1 time step
- Add: sum two signals
- -1: multiply by -1

Signals are primitives. **Operators** are the means of combination.

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Operator Notation

Systems can now be represented by compact symbolic equations.

Let \mathcal{R} represent the **right-shift operator**:

 $Y = \mathcal{R}\{X\} \equiv \mathcal{R}X$

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where X represents the whole input signal (x[n] for all n) and Y represents the whole output signal (y[n] for all n)

Right-shift means that the entire signal is shifted to the right by one time step.

Check Yourself!

Let $Y = \mathcal{R}X$. Which of the following is/are true?

- 1. y[n] = x[n] for all n
- 2. y[n+1] = x[n] for all n
- 3. y[n] = x[n+1] for all n
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Operator Notation

Systems are concisely represented with operators.

Consider:



Equivalent representation with \mathcal{R} :

$$Y = X - \mathcal{R}X = (1 - \mathcal{R})X$$

Operator Algebra

Cascade of systems: apply second system to output of first



$$Y_1 = (1 - \mathcal{R})X$$
$$Y_2 = (1 - \mathcal{R})Y_1 = (1 - \mathcal{R})(1 - \mathcal{R})X$$
Operator expressions expand and reduce like polynomials!



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Difference equation:

$$y_2[n] = y_1[n] - y_1[n-1]$$

= $(x[n] - x[n-1]) - (x[n-1] - x[n-2])$
= $x[n] - 2x[n-1] + x[n-2]$

Operator expressions expand and reduce like polynomials!



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Operator notation:

$$Y_2 = (1 - \mathcal{R})(1 - \mathcal{R})X$$
$$= (1 - 2\mathcal{R} + \mathcal{R}^2)X$$

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Expressions involving ${\mathcal R}$ obey all of the familiar laws of algebra. Examples:

• Commutativity:

$$\mathcal{R}(1-\mathcal{R})X = (1-\mathcal{R})\mathcal{R}X$$

• Multiplication distributes over addition:

$$\mathcal{R}(1-\mathcal{R}) = \mathcal{R} - \mathcal{R}^2$$

• Associativity:

$$(2-\mathcal{R})(\mathcal{R}(1-\mathcal{R})) = ((2-\mathcal{R})\mathcal{R})(1-\mathcal{R})$$

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Applies your existing expertise with polynomials to understand block diagrams, and thereby understand systems.

Operator Approach

Facilitates seeing relations among systems. The following are "equivalent" in terms of input/output relationship when starting from rest:



Are the following two systems equivalent?



0. No 1. Yes

Are the following two systems equivalent?



0. No 1. Yes

Reasoning About L and TI

We can use operator notation to show some effects of:

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Time Invariance:

Applying \mathcal{H} to $\mathcal{R}\Delta$ gives the same result as applying \mathcal{R} to $\mathcal{H}\Delta$!

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Linearity:

Applying ${\cal H}$ to 2Δ gives the same result as applying 2 to ${\cal H}\Delta!$

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LTI:

Applying \mathcal{H} to $(2\Delta + 3\mathcal{R}\Delta)$ gives the same result as applying $(2 + 3\mathcal{R})$ to $\mathcal{H}\Delta$!

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Superposition: powerful tool for analyzing systems. Can reason about response to arbitrary systems by thinking only about unit sample response (or *vice versa*).















Bunnies



Bunnies



Bunnies



- y[n] is the total number of pairs of bunnies present at time n (regardless of age)
- x[n] is the number of new pairs of (young) bunnies introduced at time n by an outside source
- Young bunnies mature into adult bunnies in one timestep.
- Each pair of adult bunnies produces a pair of young bunnies on each timestep.

Determine an operator expression for the "bunny" system, relating X and Y.

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Determine an operator expression for the "bunny" system, relating X and Y.

Hint: It may help to introduce two auxilliary signals, one to represent the number of pairs of adult bunnies in the system, and one to represent the number of pairs of child bunnies in the system.

Define new signals C and A, representing children and adults, respectively.

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Want to use algebra to get an expression just involving \boldsymbol{Y} and \boldsymbol{X} and operators

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Y = A + C

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Y = A + C $Y = \mathcal{R}(A + C) + \mathcal{R}A + X$

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Block Diagram

Draw a block diagram for the "bunnies" system:

Simulating

```
>>> children = 1
>>> adults = 0
>>> for i in range(100):
... print(adults + children, end=" ")
... newadults = children + adults
... newchildren = adults
... adults, children = newadults, newchildren
```

. . .
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1 1 2 3 5 8 13 21 34 55 89 144 233 377 610 987 1597 2584 4181 6765 10946 17711 28657 46368 75025 121393 196418 317811 514229 832040 1346269 2178309 3524578 5702887 9227465 14930352 24157817 39088169 63245986 102334155 165580141 267914296 433494437 701408733 1134903170 1836311903 2971215073 4807526976 7778742049 12586269025 20365011074 32951280099 53316291173 86267571272 139583862445 225851433717 365435296162 591286729879 956722026041 1548008755920 2504730781961 4052739537881 ...

Modeling Wall Finder

In order to better understand wall finder, we are going to build mathematical and computational models of the system this week (starting today!). We will break this system into pieces:



We have control over one of these pieces (controller), but the others are part of the system we are trying to control, so we have to try to model what is there. Introduced the signals and systems abstraction.

Introduced several representations of systems: difference equations, block diagrams, state machines, operator expressions, sample responses

Discussed applications of operator notation (you'll get more practice in this week's labs and homeworks!)

Saw one example of using the Signals and Systems approach for modeling, and a few examples of converting between different representations.

Software Lab 2:

Begin Modeling Wall Finder

Design Lab 2: Simulate, Test, and Improve Model