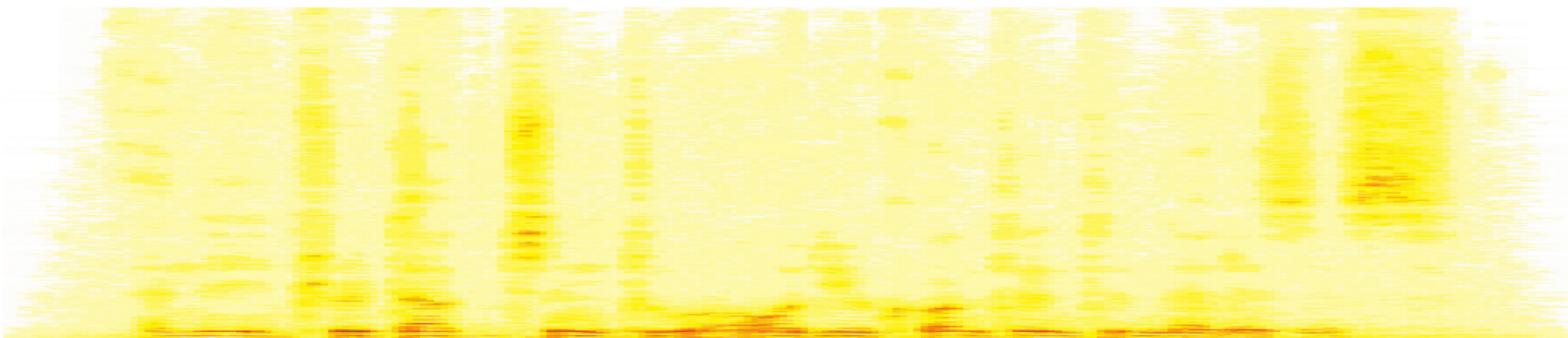


# Introduction to Audio Content Analysis

## Module 2.3: Fundamentals — Convolution

alexander lerch



# introduction

## overview

corresponding textbook section

[Chapter 2 — Fundamentals](#): pp. 14–18

[Appendix A — Convolution Properties](#): pp. 181–183

### ● lecture content

- LTI systems
- convolution
- filter examples

### ● learning objectives

- basic understanding of linearity and time-invariance
- basic understanding of the convolution operation
- ability to implement simple filters



# introduction

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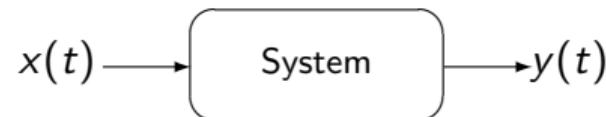


# systems

## introduction

a system:

- any process producing an output signal in response to an input signal



**name examples for systems in signal processing**

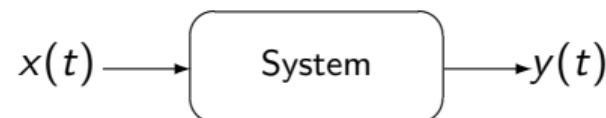


# systems

## introduction

a system:

- any process producing an output signal in response to an input signal



**name examples for systems in signal processing**



- filters, effects
- vocal tract
- room
- (audio) cable
- ...

# systems

## LTI systems

### LTI: Linear Time-Invariant Systems

are a great model for many real-world systems

- **linearity**

- ① *homogeneity*:  $f(ax) = af(x)$
- ② *superposition* (additivity):  $f(x + y) = f(x) + f(y)$

- **time invariance**:  $f(x(t - \tau)) = f(x)(t - \tau)$

# convolution

## introduction

### convolution

convolution operation describes the **output of an LTI system**:

$$y(t) = (x * h)(t) := \int_{-\infty}^{\infty} x(\tau)h(t - \tau)d\tau$$

$$y(i) = (x * h)(i) := \sum_{j=-\infty}^{\infty} x(j)h(i - j)$$

overview



systems



convolution



filter examples



summary



# convolution

## animation



matlab source: [matlab/animateConvolution.m](#)



# convolution

## properties

- **identity:**

$$x(i) = \delta(i) * x(i)$$

- **commutativity:**

$$h(i) * x(i) = x(i) * h(i)$$

- **associativity:**

$$(g(i) * h(i)) * x(i) = g(i) * (h(i) * x(i))$$

- **distributivity:**

$$g(i) * (h(i) + x(i)) = (g(i) * h(i)) + (g(i) * x(i))$$

- **circularity:**

$$h(i) \text{ periodic} \Rightarrow y(i) = h(i) * x(i) \text{ periodic}$$

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# filter

## example 1: Moving Average

$$y(i) = \sum_{j=0}^{\mathcal{J}-1} b(j) \cdot x(i-j)$$

- replaces current sample with average of  $\mathcal{J}$  samples
- smooths a signal (low pass)
- IR: rectangular
- linear phase, but inefficient for many coefficients
- Finite Impulse Response (FIR)

# filter

## example 2: Single-Pole

$$y(i) = (1 - \alpha) \cdot x(i) + \alpha \cdot y(i - 1)$$

- **recursive system:** output depends on previous *output*
- the larger alpha, the less the current input is taken into account (low pass)
- alpha from 0...1
- efficient, but non-linear phase
- Infinite Impulse Response (IIR)

# summary

## lecture content

### ● LTI system

- good model for many real-world system
- linear (homogeneity, superposition) and time-invariant
- impulse response reflects all system properties

### ● convolution

- operation that computes the output of an LTI system from the input

### ● low pass filters

- often used to smooth a signal
- typical examples are moving average (FIR) and single pole (IIR)

