

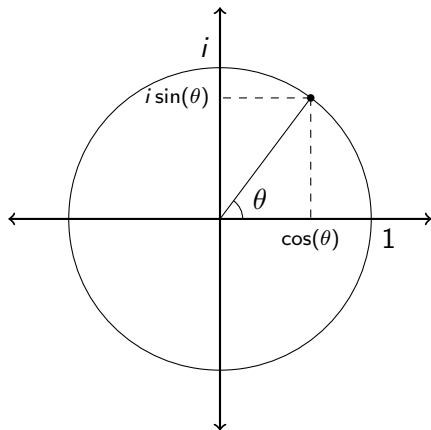
Creative Computing II

Christophe Rhodes
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Autumn 2009, Tuesdays, 10:00–15:00
Winter 2010, Tuesdays, 13:30–17:30

Signals

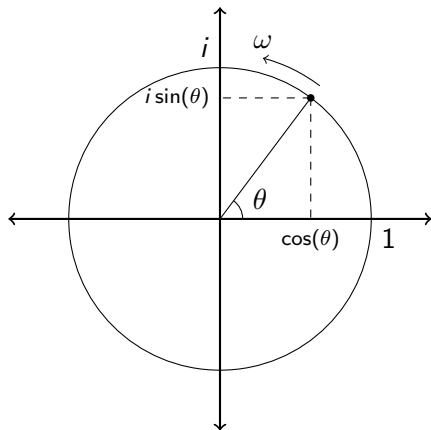
The Complex Exponential



$$e^{i\theta} = \cos(\theta) + i \sin(\theta); e^{-i\theta} = \cos(\theta) - i \sin(\theta).$$

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$$e^{i\omega t} = \cos(\omega t) + i \sin(\omega t); e^{-i\omega t} = \cos(\omega t) - i \sin(\omega t).$$

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Sinusoidal functions

Functional relations:

$$\blacktriangleright \cos(\omega t) = \frac{e^{i\omega t} + e^{-i\omega t}}{2}$$

$$\blacktriangleright \sin(\omega t) = \frac{e^{i\omega t} - e^{-i\omega t}}{2i}$$

Identities:

$$\blacktriangleright e^{i\pi} = -1 \text{ (Euler's Identity)}$$

$$\blacktriangleright e^{i\frac{\pi}{2}} = i$$

$$\blacktriangleright e^{2\pi i} = 1$$

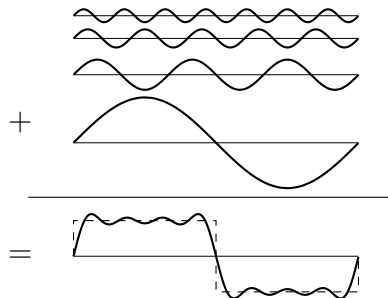
$$\blacktriangleright (\cos(\theta) + i \sin(\theta))^n = \cos(n\theta) + i \sin(n\theta) \text{ (de Moivre's formula)}$$

Signals

Fourier Series

Square wave:

$$s_f(t) = \sum_{k=1}^{\infty} \frac{1}{2k-1} \sin(2\pi(2k-1)ft)$$



Signals

Fourier Series

Fourier Series:

- ▶ Any signal can be written as a weighted sum of sin and cos terms: a **Fourier Series**.
- ▶ For a signal of length L , all sinusoids have angular frequencies that are integer multiples of $\frac{2\pi}{L}$.
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Fourier Analysis of Signals:

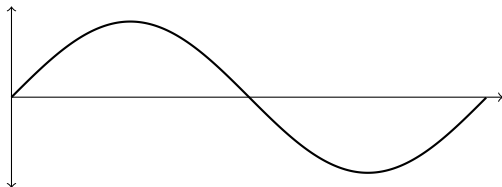
- ▶ Extraction of frequency components for a given signal;
- ▶ Dot-product multiply by complex exponential signal;
- ▶ Magnitude and phase of result give magnitude and phase of corresponding sinusoid.

Signals

Fourier Series

How does this work?

- ▶ dot-product of sinusoid with *exactly itself* gives a non-zero result;
- ▶ all other dot-products between sinusoids give zero.
- ▶ sinusoids are **orthogonal** basis functions.

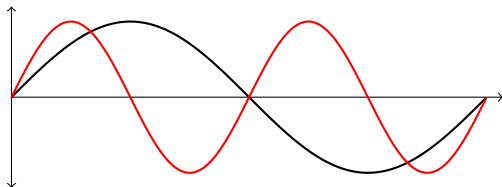


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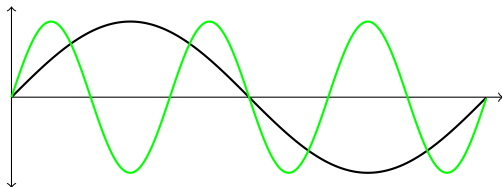


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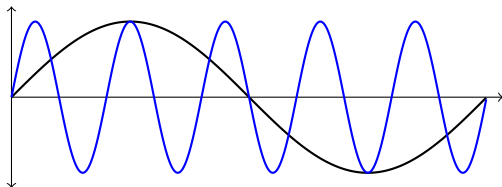


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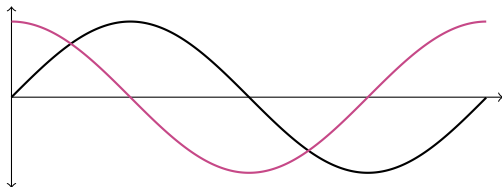


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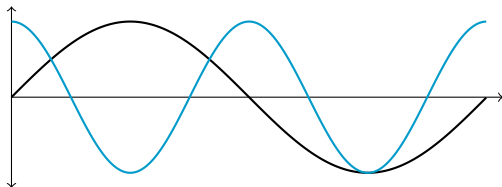


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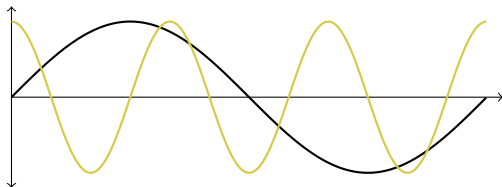


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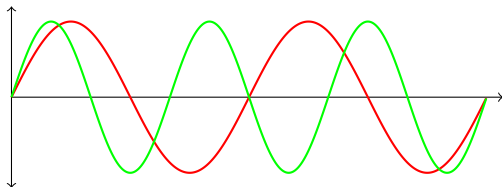


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Note: $\mathcal{F}(x)$ sometimes notated as \tilde{x} .

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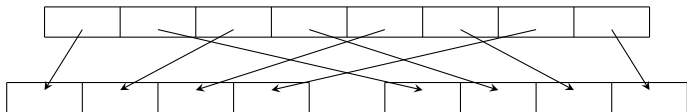
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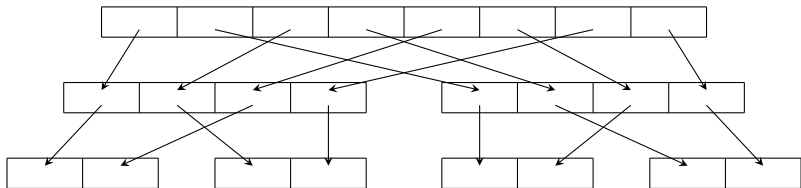
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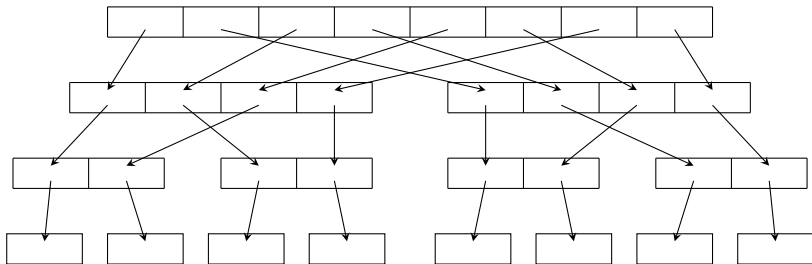
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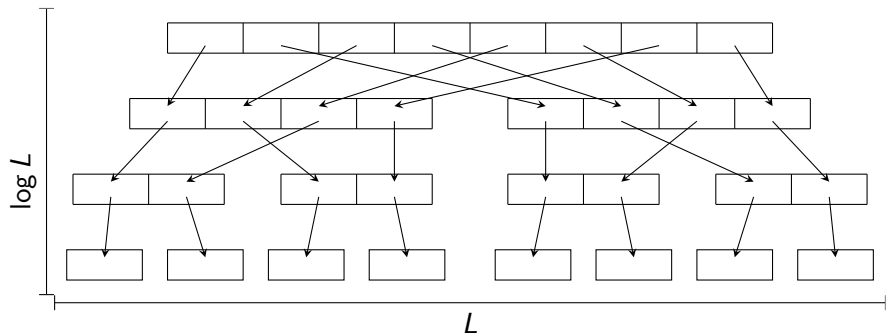
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Overall complexity $O(L \log L)$

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Systems

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Octave:

- ▶ `ifft(fft(h,length([h x])-1).*fft(x,length([h x])-1))`

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$$y = \mathcal{F}^{-1}(\mathcal{F}(h) \times \mathcal{F}(x))$$

so

$$\mathcal{F}(y) = \mathcal{F}(h) \times \mathcal{F}(x)$$

- ▶ $\mathcal{F}(h)$ is the **frequency response** of the system.
- ▶ the frequency spectrum of the output signal is the product of the spectrum of the input and the frequency response of the system.