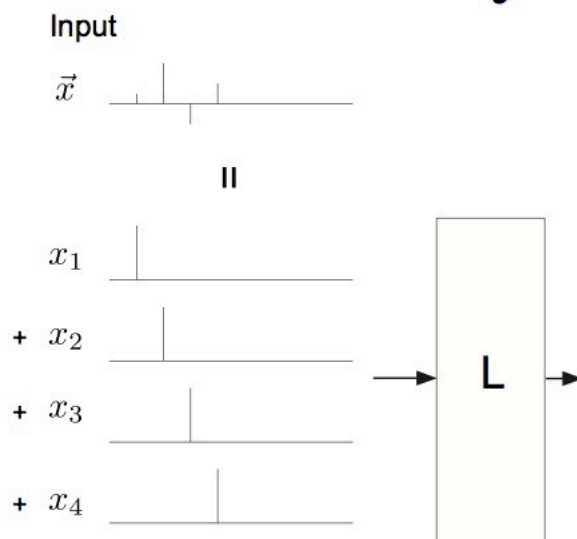


Linear shift-invariant (LSI) systems

- Linearity (covered previously):
“linear combination in, linear combination out”
- Shift-invariance (new):
“shifted vector in, shifted vector out”
- Some examples [on board]

LSI system



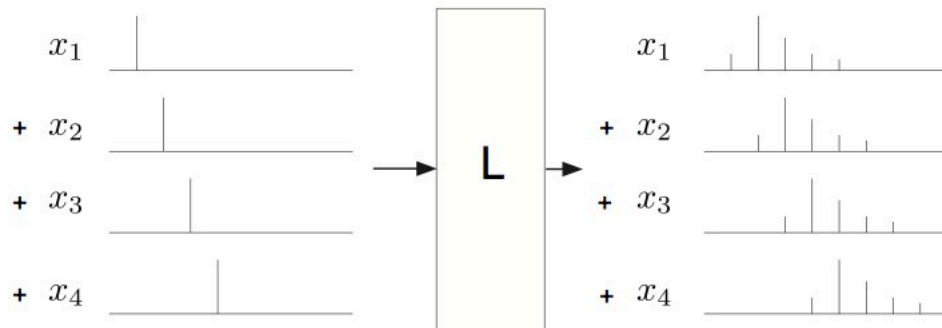
(rewrite as weighted
sum of impulses)

LSI system

Input



||

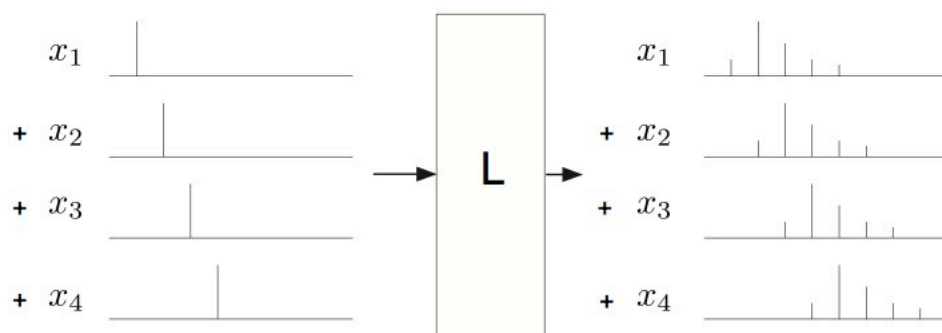


LSI system

Input



||



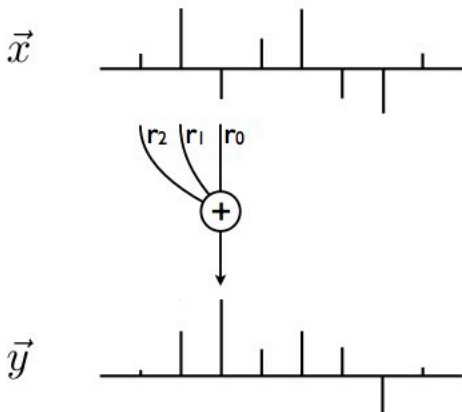
||



Output

LSI systems are characterized by their “impulse response”

Convolution

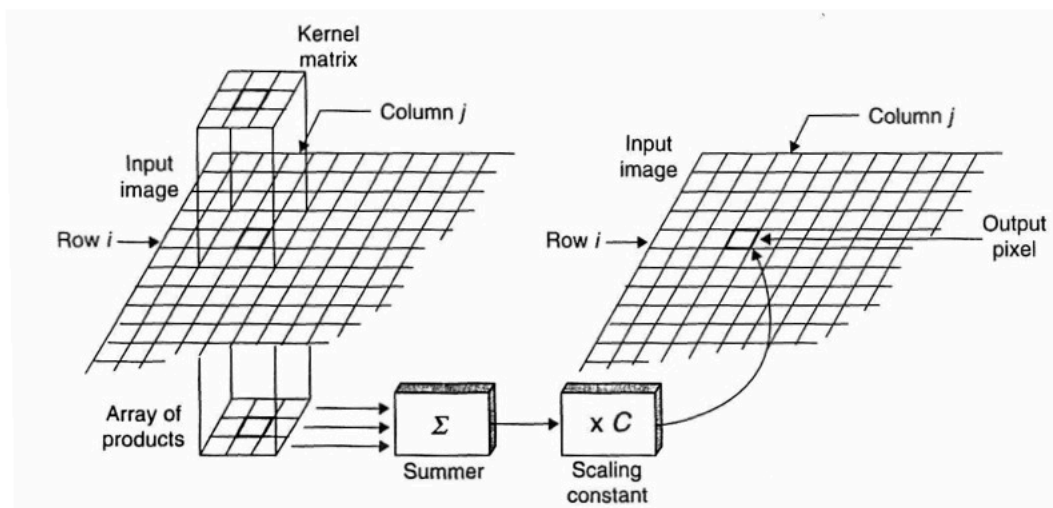


$$\begin{aligned} y(n) &= \sum_m r(n-m)x(m) \\ &= \sum_m r(m)x(n-m) \end{aligned}$$

- Matrix description
- boundaries: zero-padded, reflected, circular
- Examples: impulse, delay, average, difference

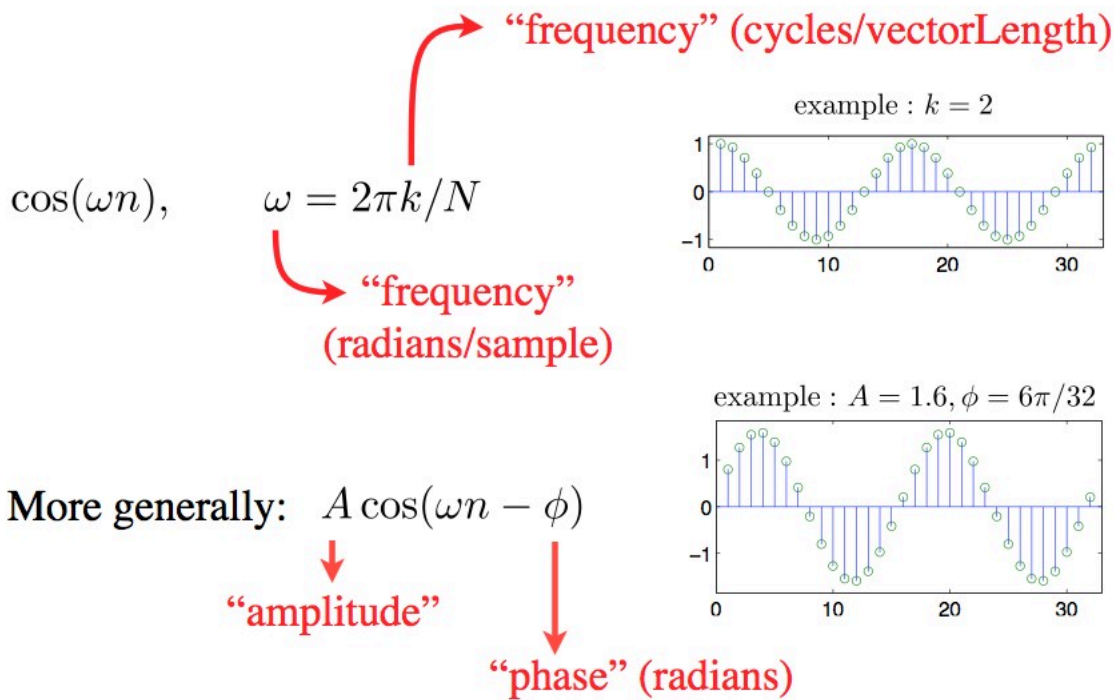
2D convolution

- sliding window



[figure c/o Castleman]

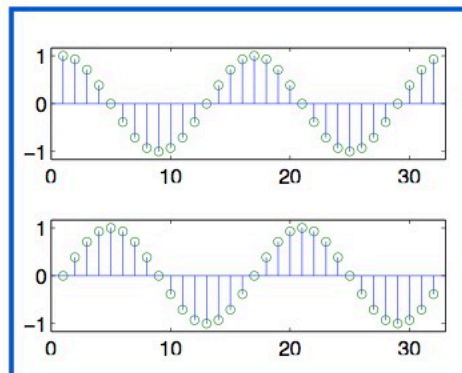
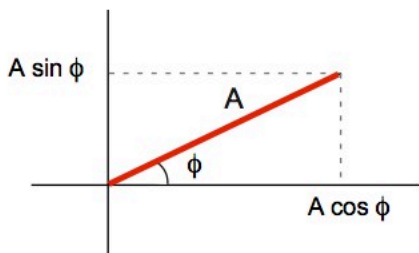
Discrete Sinusoids



Shifting Sinusoids

$$A \cos(\omega n - \phi) = \underline{A \cos(\phi)} \underline{\cos(\omega n)} + \underline{A \sin(\phi)} \underline{\sin(\omega n)}$$

fixed cos/sin vectors:

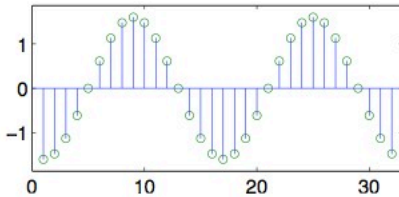


A *shifted* sinusoidal vector can be written as a weighted sum of two *fixed* sinusoidal vectors!

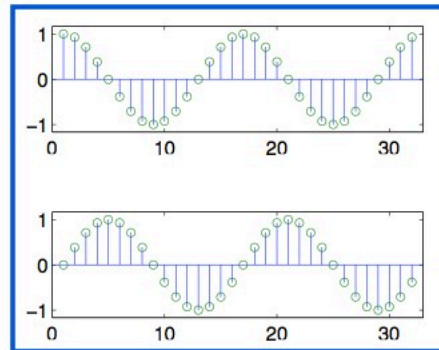
Shifting Sinusoids

$$A \cos(\omega n - \phi) = \underbrace{A \cos(\phi)}_{\text{fixed cos vector}} \underbrace{\cos(\omega n)}_{\text{fixed sin vector}} + \underbrace{A \sin(\phi)}_{\text{fixed cos vector}} \underbrace{\sin(\omega n)}_{\text{fixed sin vector}}$$

$$A = 1.6, \phi = 2\pi/12$$



fixed cos/sin vectors:



A *shifted* sinusoidal vector can be written as a weighted sum of two *fixed* sinusoidal vectors!

Sinusoids & LSI

$$x(n) = \cos(\omega n)$$

$$y(n) = \sum_m r(m) \cos(\omega(n - m)) \quad (\text{convolution})$$

Sinusoids & LSI

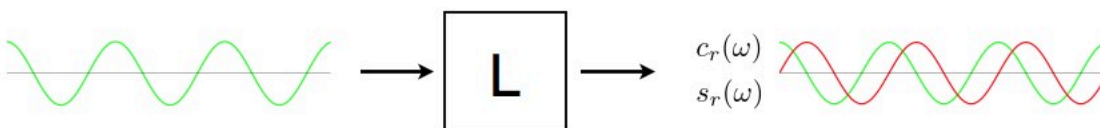
$$x(n) = \cos(\omega n)$$

$$\begin{aligned}
 y(n) &= \sum_m r(m) \cos(\omega(n-m)) && \text{(convolution)} \\
 &= \sum_m r(m) \cos(\omega m) \cos(\omega n) + \sum_m r(m) \sin(\omega m) \sin(\omega n)
 \end{aligned}$$

Sinusoids & LSI

$$x(n) = \cos(\omega n)$$

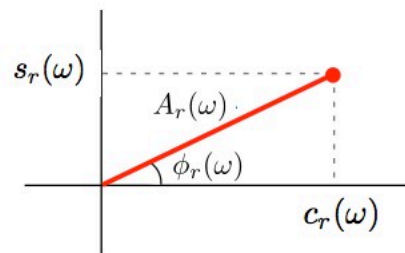
$$\begin{aligned}
 y(n) &= \sum_m r(m) \cos(\omega(n-m)) && \text{(convolution)} \\
 &= \sum_m r(m) \cos(\omega m) \cos(\omega n) + \sum_m r(m) \sin(\omega m) \sin(\omega n) \\
 &= \underset{\substack{\downarrow \\ c_r(\omega)}}{\sum_m r(m) \cos(\omega m)} \cos(\omega n) + \underset{\substack{\downarrow \\ s_r(\omega)}}{\sum_m r(m) \sin(\omega m)} \sin(\omega n)
 \end{aligned}$$



Sinusoids & LSI

$$x(n) = \cos(\omega n)$$

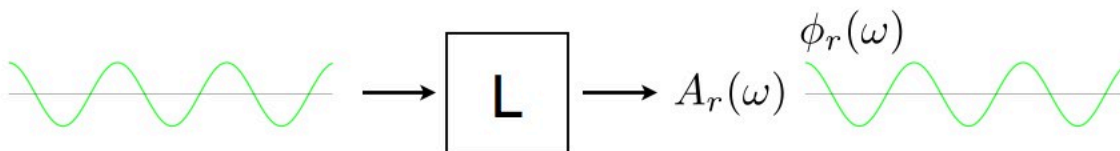
$$\begin{aligned}
 y(n) &= \sum_m r(m) \cos(\omega(n-m)) && \text{(convolution)} \\
 &= \sum_m r(m) \cos(\omega m) \cos(\omega n) + \sum_m r(m) \sin(\omega m) \sin(\omega n) \\
 &= c_r(\omega) \cos(\omega n) + s_r(\omega) \sin(\omega n) \\
 &= A_r(\omega) \cos(\phi_r(\omega)) \cos(\omega n) + A_r(\omega) \sin(\phi_r(\omega)) \sin(\omega n)
 \end{aligned}$$



Sinusoids & LSI

$$x(n) = \cos(\omega n)$$

$$\begin{aligned}
 y(n) &= \sum_m r(m) \cos(\omega(n-m)) \\
 &= \sum_m r(m) \cos(\omega m) \cos(\omega n) + \sum_m r(m) \sin(\omega m) \sin(\omega n) \\
 &= c_r(\omega) \cos(\omega n) + s_r(\omega) \sin(\omega n) \\
 &= A_r(\omega) \cos(\phi_r(\omega)) \cos(\omega n) + A_r(\omega) \sin(\phi_r(\omega)) \sin(\omega n) \\
 &= A_r(\omega) \cos(\omega n - \phi_r(\omega))
 \end{aligned}$$



“Sinusoid in, sinusoid out”

Discrete Fourier transform (DFT)

- Construct an orthogonal matrix of sin/cos pairs, at frequency multiples of $2\pi/N$ radians/sample, i.e., $2\pi k/N$, for $k = 0, 1, 2, \dots, N/2$
- For $k = 0$ and $k = N/2$, only need the cosine part of the pair
- When we transform a vector using this matrix, think of output as paired coordinates
- Common to plot these pairs as amplitude/phase

[all details on board]

The Fourier family

		signal domain	
frequency domain		continuous	discrete
	continuous	Fourier transform	discrete-time Fourier transform
	discrete	Fourier series	discrete Fourier transform

you are here

The “fast Fourier transform” (FFT) is a computationally efficient implementation of the DFT (runs in $N \log N$ time, instead of N^2).

Sinusoids & LSI

$$x(n) = \cos(\omega n)$$

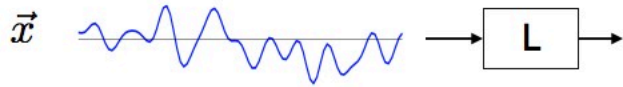
$$\begin{aligned} y(n) &= \sum_m r(m) \cos(\omega(n-m)) \\ &= \sum_m r(m) \cos(\omega m) \cos(\omega n) + \sum_m r(m) \sin(\omega m) \sin(\omega n) \\ &= c_r(\omega) \cos(\omega n) + s_r(\omega) \sin(\omega n) \\ &= A_r(\omega) \cos(\phi_r(\omega)) \cos(\omega n) + A_r(\omega) \sin(\phi_r(\omega)) \sin(\omega n) \\ &= A_r(\omega) \cos(\omega n - \phi_r(\omega)) \end{aligned}$$

NOTE: Change in amplitude and phase come from Fourier transform of the impulse response!

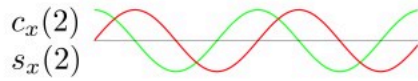
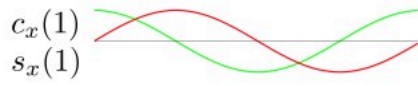
Fourier & LSI



Fourier & LSI



||

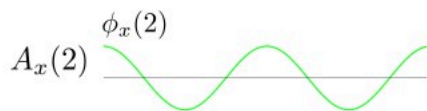


note: only 3 (of many) components shown

Fourier & LSI

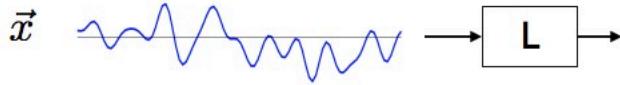


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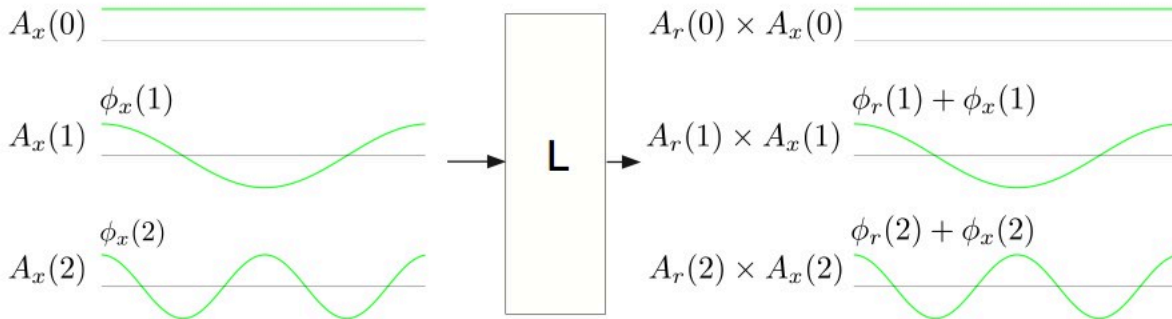


note: only 3 (of many) components shown

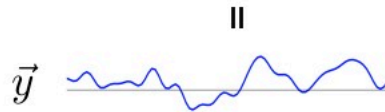
Fourier & LSI



||



note: only 3 (of many) components shown



LSI systems are characterized by their *frequency response*, specified by the Fourier Transform of their impulse response

Complex exponentials: “bundling” sine and cosine

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

$$e^{i\omega n} \longrightarrow \boxed{\text{L}} \longrightarrow A_r(\omega) e^{i(\omega n - \phi_r(\omega))} = A_r(\omega) e^{-i\phi_r(\omega)} e^{i\omega n} = \tilde{r}(\omega) e^{i\omega n}$$

Complex exponentials: “bundling” sine and cosine

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

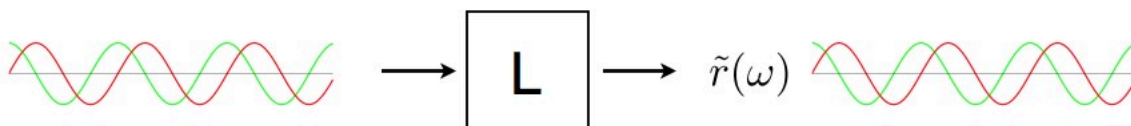
$$e^{i\omega n} \longrightarrow \boxed{\text{L}} \longrightarrow A_r(\omega) e^{i(\omega n - \phi_r(\omega))} = A_r(\omega) e^{-i\phi_r(\omega)} e^{i\omega n} = \tilde{r}(\omega) e^{i\omega n}$$

F.T. of impulse response!

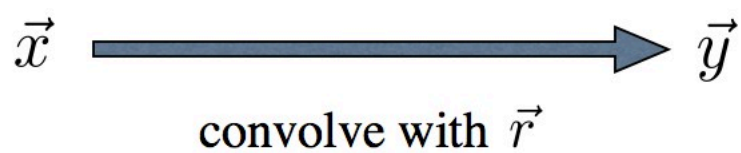
Complex exponentials: “bundling” sine and cosine

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

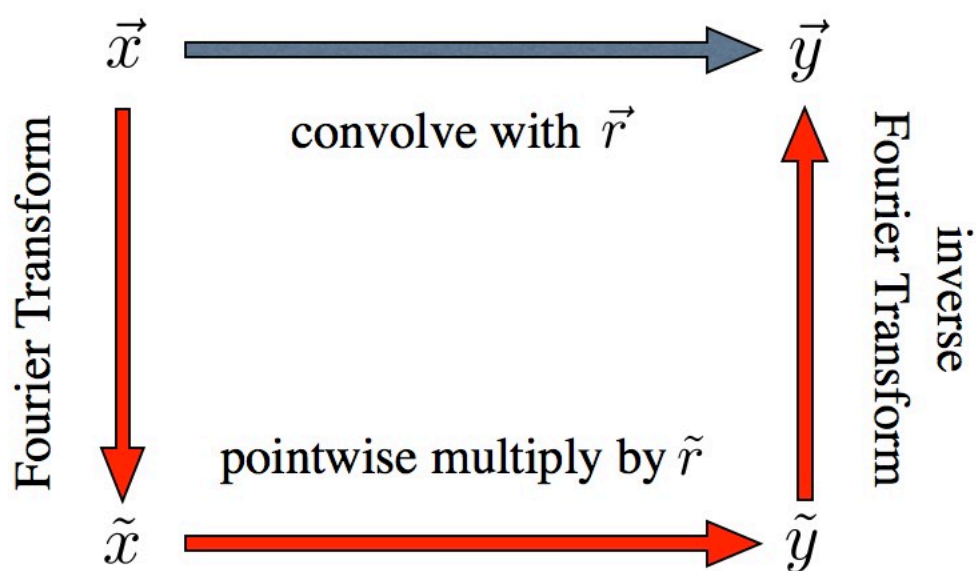
$$e^{i\omega n} \longrightarrow \boxed{\text{L}} \longrightarrow A_r(\omega) e^{i(\omega n - \phi_r(\omega))} = A_r(\omega) e^{-i\phi_r(\omega)} e^{i\omega n} = \tilde{r}(\omega) e^{i\omega n}$$



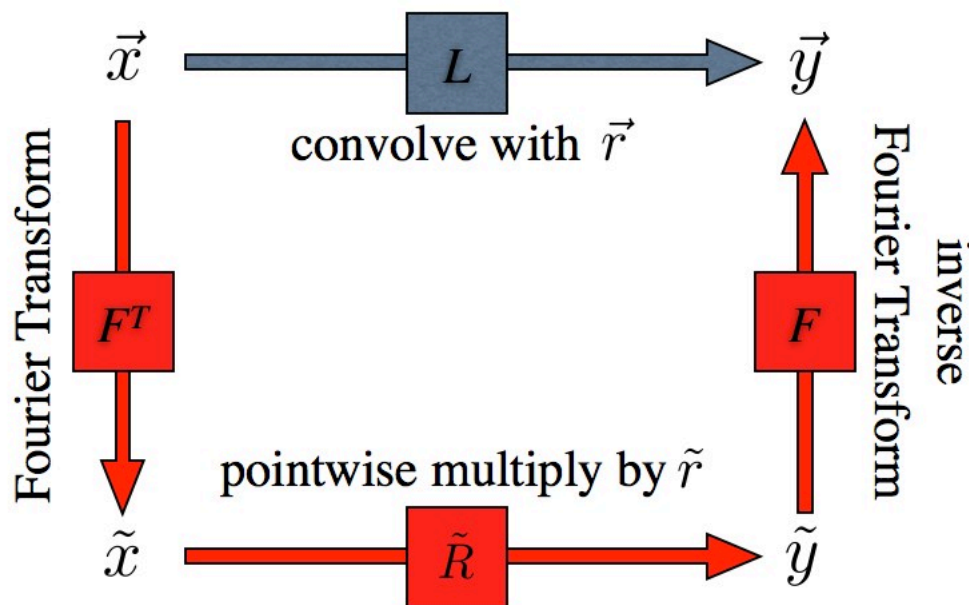
The “convolution theorem”



The “convolution theorem”



The “convolution theorem”



In matrix form: $L\vec{x} = F\tilde{R}F^T\vec{x}$

Recap

- Linear system
 - => defined by superposition
 - => characterized by a matrix
- Linear Shift-invariant (LSI) system
 - => defined by superposition and shift-invariance
 - => characterized by impulse response
 - => alternatively, characterized by frequency response (the Fourier Transform of the impulse response!)

Discrete Fourier transform (with complex numbers)

$$\tilde{r}_k = \sum_{n=0}^{N-1} r_n e^{-i\omega_k n}$$

$$r_n = \sum_{k=0}^{N-1} \tilde{r}_k e^{i\omega_k n} \quad (\text{inverse})$$

$$\text{where } \omega_k = \frac{2\pi k}{N}$$

Visualizing the (discrete) Fourier transform

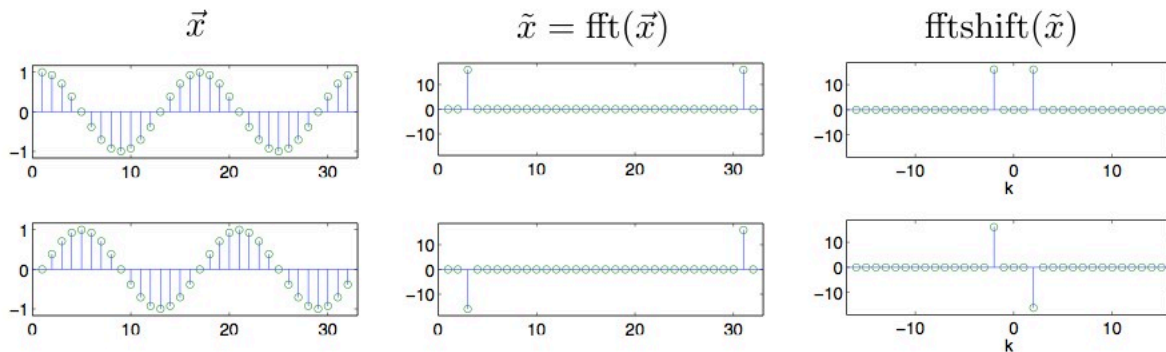
- Typically, plot Amplitude (and possibly Phase, on a separate graph), instead of real/imaginary parts
- Two conventional choices for frequency axis:
 - Plot frequencies from $k=0$ to $k=N/2$
 - Plot frequencies from $k=-N/2$ to $N/2-1$

Dealing with DFT in Matlab...

$$e^{i\omega n} = \cos(\omega n) + i \sin(\omega n)$$

$$\cos(\omega n) = \frac{1}{2}(e^{i\omega n} + e^{-i\omega n}) \quad \sin(\omega n) = \frac{-i}{2}(e^{i\omega n} - e^{-i\omega n})$$

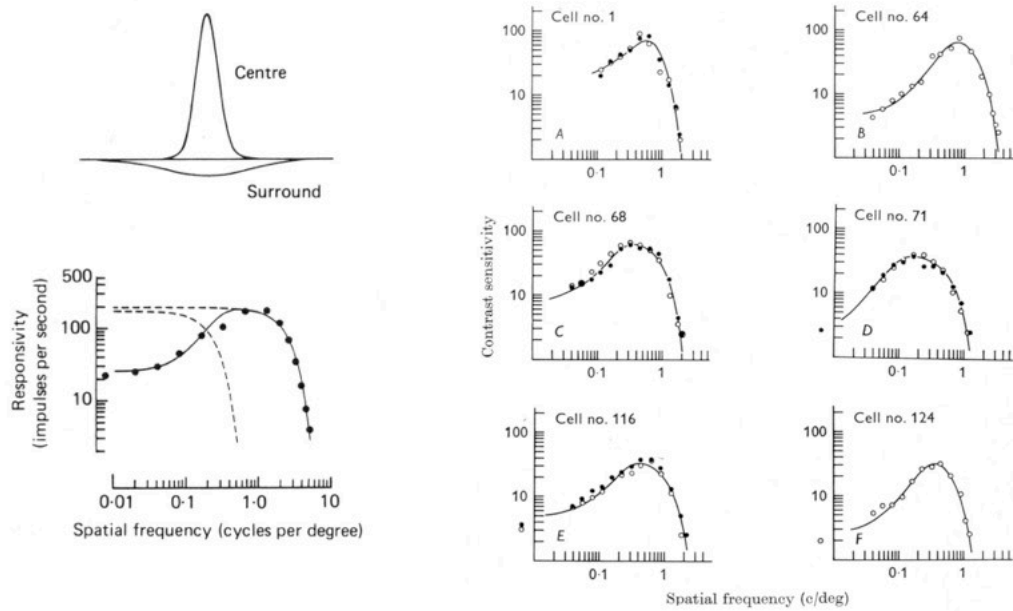
Matlab examples, N=32 (note indexing and amplitudes):



Some examples

- constant
- sinusoid
- impulse
- Gaussian - “lowpass”
- DoG (difference of 2 Gaussians) - “bandpass”
- Gabor (Gaussian windowed sinusoid) - “bandpass”

Retinal ganglion cells (1D)



Enroth-Cugell and Robson (1984)