# Convolution, Correlation, & Eourier Transforms

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#### Introduction

- A large class of signal processing techniques fall under the category of *Fourier transform* methods
  - These methods fall into two broad categories
    - Efficient method for accomplishing common data manipulations
    - Problems related to the Fourier transform or the power spectrum

## Time & Frequency Domains

- A physical process can be described in two ways
  - In the *time domain*, by the values of some some quantity h as a function of time t, that is h(t),  $-\infty < t < \infty$
  - In the *frequency domain*, by the complex number, H, that gives its amplitude and phase as a function of frequency f, that is H(f), with -∞ < f < ∞
- It is useful to think of h(t) and H(f) as two different representations of the same function
  - One goes back and forth between these two representations by Fourier transforms

#### Fourier Transforms

$$H(f) = \int_{-\infty}^{\infty} h(t)e^{-2\pi i f t} dt$$

$$h(t) = \int_{-\infty}^{\infty} H(f)e^{2\pi i f t} df$$

- If *t* is measured in seconds, then *f* is in cycles per second or Hz
- Other units
  - E.g, if h=h(x) and x is in meters, then H is a function of spatial frequency measured in cycles per meter

#### Fourier Transforms

- The Fourier transform is a linear operator
  - The transform of the sum of two functions is the sum of the transforms

$$h_{12} = h_1 + h_2$$

$$H_{12}(f) = \int_{-\infty}^{\infty} h_{12} e^{-2\pi i f t} dt$$

$$= \int_{-\infty}^{\infty} (h_1 + h_2) e^{-2\pi i f t} dt = \int_{-\infty}^{\infty} h_1 e^{-2\pi i f t} dt + \int_{-\infty}^{\infty} h_2 e^{-2\pi i f t} dt$$

$$= H_1 + H_2$$

#### Fourier Transforms

- h(t) may have some special properties
  - Real, imaginary
  - Even: h(t) = h(-t)
  - Odd: h(t) = -h(-t)
- In the frequency domain these symmetries lead to relations between H(f) and H(-f)

## FT Symmetries

If	Then
h(t) real	$H(-f) = [H(f)]^*$
h(t) imaginary	$H(-f) = -[H(f)]^*$
h(t) even	H(-f) = H(f) (even)
h(t) odd	H(-f) = -H(f)  (odd)
h(t) real & even	H(f) real & even
h(t) real & odd	H(f) imaginary & odd
h(t) imaginary & even	H(f) imaginary & even
h(t) imaginary & odd	H(f) real & odd

## Elementary Properties of FT

$$h(t) \leftrightarrow H(f)$$
 Fourier Pair  $h(at) \leftrightarrow \frac{1}{a} H(f/a)$  Time scaling  $h(t-t_0) \leftrightarrow H(f) e^{-2\pi i f t_0}$  Time shifting

#### Convolution

- With two functions h(t) and g(t), and their corresponding Fourier transforms H(f) and G(f), we can form two special combinations
  - The *convolution*, denoted f = g \* h, defined by

$$f(t) = g * h \equiv \int_{-\infty}^{\infty} g(\tau)h(t - \tau)d\tau$$

#### Convolution

• g\*h is a function of time, and

$$g*h = h*g$$

- The convolution is one member of a transform pair

$$g * h \longleftrightarrow G(f)H(f)$$

- The Fourier transform of the convolution is the product of the two Fourier transforms!
  - This is the Convolution Theorem

#### Correlation

• The *correlation* of *g* and *h* 

$$Corr(g,h) \equiv \int_{-\infty}^{\infty} g(\tau + t)h(t)d\tau$$

- The correlation is a function of *t*, which is known as the lag
  - The correlation lies in the time domain

#### Correlation

• The correlation is one member of the transform pair

$$Corr(g,h) \leftrightarrow G(f)H^*(f)$$

- More generally, the RHS of the pair is G(f)H(-f)
- Usually g & h are real, so  $H(-f) = H^*(f)$
- Multiplying the FT of one function by the complex conjugate of the FT of the other gives the FT of their correlation
  - This is the Correlation Theorem

#### Autocorrelation

- The correlation of a function with itself is called its *autocorrelation*.
  - In this case the correlation theorem becomes the transform pair

$$Corr(g,g) \leftrightarrow G(f)G^*(f) = |G(f)|^2$$

- This is the **Wiener-Khinchin Theorem** 

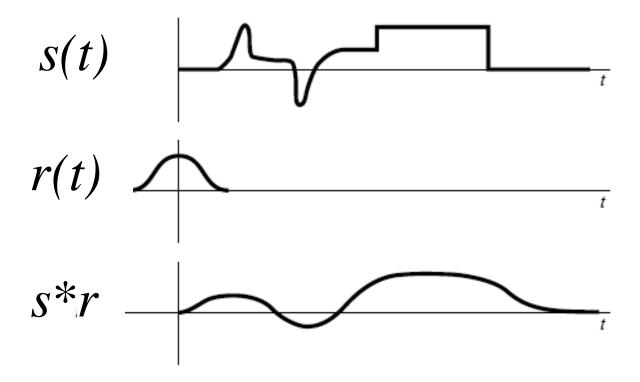
#### Convolution

- Mathematically the convolution of r(t) and s(t), denoted r\*s=s\*r
- In most applications *r* and *s* have quite different meanings
  - s(t) is typically a signal or data stream, which goes on indefinitely in time
  - -r(t) is a response function, typically a peaked and that falls to zero in both directions from its maximum

## The Response Function

- The effect of convolution is to smear the signal s(t) in time according to the recipe provided by the response function r(t)
- A spike or delta-function of unit area in s which occurs at some time  $t_0$  is
  - Smeared into the shape of the response function
  - Translated from time 0 to time  $t_0$  as  $r(t t_0)$

#### Convolution



- The signal s(t) is convolved with a response function r(t)
  - Since the response function is broader than some features in the original signal, these are smoothed out in the convolution

#### Fourier Transforms & FFT

- Fourier methods have revolutionized many fields of science & engineering
  - Radio astronomy, medical imaging, & seismology
- The wide application of Fourier methods is due to the existence of the **fast Fourier transform** (FFT)
- The FFT permits rapid computation of the discrete Fourier transform
- Among the most direct applications of the FFT are to the convolution, correlation & autocorrelation of data

#### The FFT & Convolution

- The convolution of two functions is defined for the continuous case
  - The convolution theorem says that the Fourier transform of the convolution of two functions is equal to the product of their individual Fourier transforms

$$g * h \leftrightarrow G(f)H(f)$$

- We want to deal with the discrete case
  - How does this work in the context of convolution?

#### Discrete Convolution

- In the discrete case s(t) is represented by its sampled values at equal time intervals  $s_i$
- The response function is also a discrete set  $r_k$ 
  - $r_0$  tells what multiple of the input signal in channel j is copied into the output channel j
  - $-r_1$  tells what multiple of input signal j is copied into the output channel j+1
  - $-r_{-1}$  tells the multiple of input signal j is copied into the output channel j-l
  - Repeat for all values of k

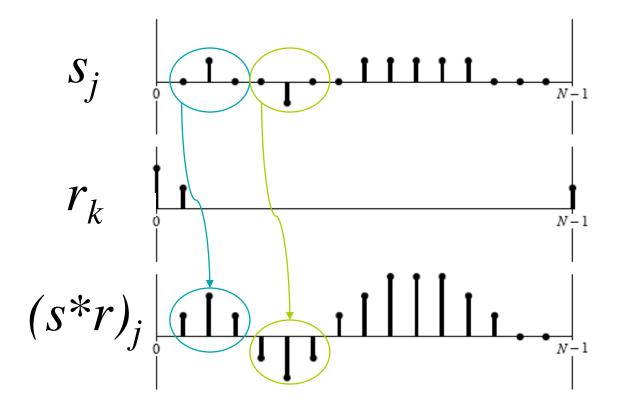
#### Discrete Convolution

• Symbolically the discrete convolution is with a response function of finite duration,

N, is
$$(s*r)_{j} = \sum_{k=-N/2+1}^{N/2} s_{k}r_{j-k}$$

$$(s*r)_{j} \longleftrightarrow S_{l}R_{l}$$

#### Discrete Convolution



- Convolution of discretely sampled functions
  - Note the response function for negative times wraps around and is stored at the end of the array  $r_k$

### Examples

- Java applet demonstrations
  - Continuous convolution
    - http://www.jhu.edu/~signals/convolve/
  - Discrete convolution
    - http://www.jhu.edu/~signals/discreteconv/