

Convolution, Correlation,
&
Fourier 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 h as a function of time t , that is $h(t)$, $-\infty < t < \infty$
 - In the *frequency domain*, by H that gives its amplitude and phase as a function of frequency f , that is $H(f)$, with $-\infty < f < \infty$
 - In general h and H are complex numbers
- 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

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

$$\begin{aligned} H_{12}(f) &= \int_{-\infty}^{\infty} h_{12} e^{-2\pi i f t} dt \\ &= \int_{-\infty}^{\infty} (h_1 + h_2) e^{-2\pi i f t} dt = \int_{-\infty}^{\infty} h_1 e^{-2\pi i f t} dt + \int_{-\infty}^{\infty} h_2 e^{-2\pi i f t} dt \\ &= H_1 + H_2 \end{aligned}$$

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

$$h(t) \leftrightarrow H(f) \quad \text{Fourier Pair}$$

$$h(at) \leftrightarrow \frac{1}{a} H(f/a) \quad \text{Time scaling}$$

$$h(t - t_0) \leftrightarrow H(f)e^{-2\pi i f t_0} \quad \text{Time shifting}$$

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

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

- 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

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

- 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

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

- 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

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

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

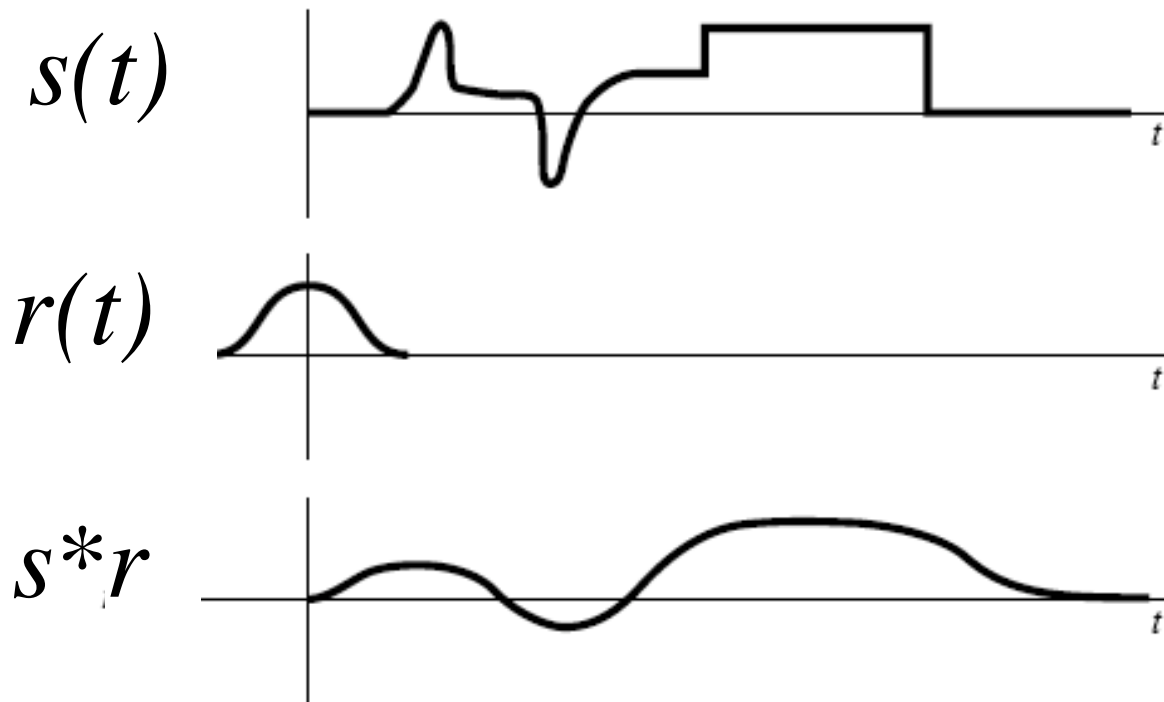
Convolution

- 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
 - Smearred 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_j
- 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 $j-1$
 - Repeat for all values of k

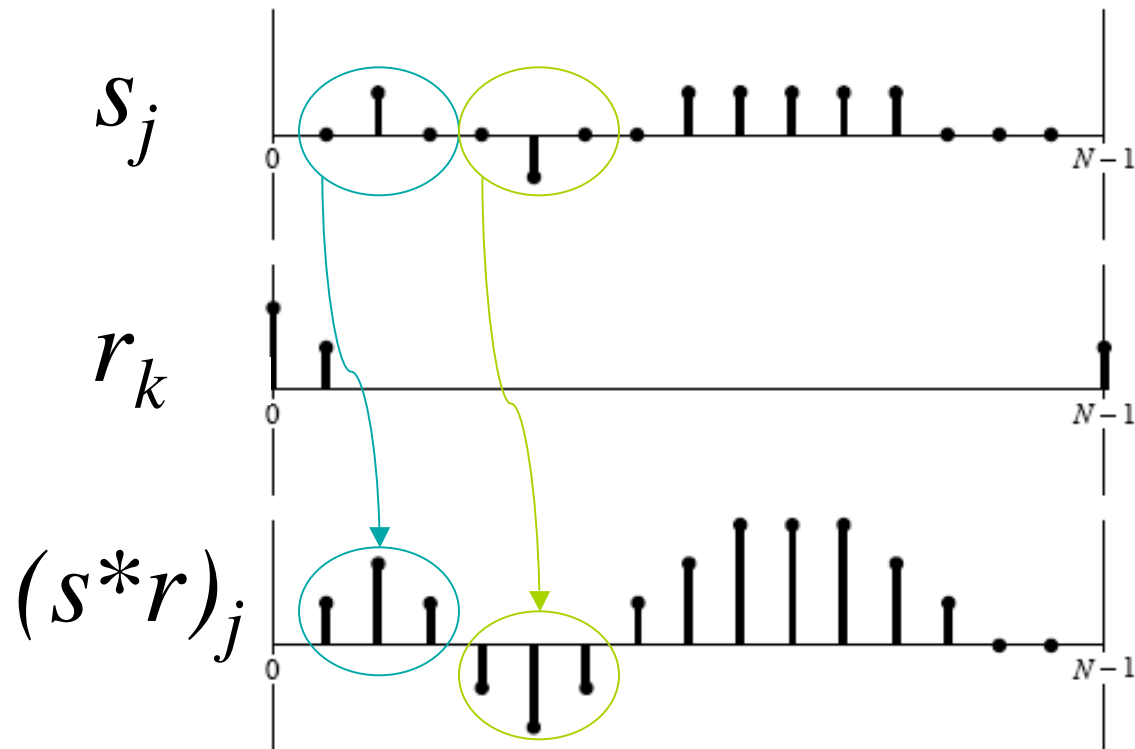
Discrete Convolution

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

$$(s * r)_j = \sum_{k=-N/2+1}^{N/2} s_k r_{j-k}$$

$$(s * r)_j \leftrightarrow S_l R_l$$

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/>

Suppose that f and g are functions of time

$$f(t) = \int_{-\infty}^{\infty} F(\nu) e^{2\pi i \nu t} d\nu \quad \text{and} \quad g(t) = \int_{-\infty}^{\infty} G(\nu) e^{2\pi i \nu t} d\nu$$

The convolution $f * g$ says

$$\begin{aligned} f * g &= \int_{-\infty}^{\infty} g(t') f(t - t') dt' \\ &= \int_{-\infty}^{\infty} g(t') \left[\int_{-\infty}^{\infty} F(\nu) e^{2\pi i \nu (t - t')} d\nu \right] dt' \end{aligned}$$

Swap the order of integration

$$\begin{aligned} &= \int_{-\infty}^{\infty} F(\nu) \left[\int_{-\infty}^{\infty} g(t') e^{-2\pi i \nu t'} dt' \right] e^{2\pi i \nu t} d\nu \\ &= FT [F(\nu)G(\nu)] \end{aligned}$$

Voila!