

### Convolution Integral

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

$\int_{-\infty}^{\infty}$  a window of weights moved over the data  
 $x(t-\tau)$  data

look at it like an operation that involves a "running mean" where  $h(\tau)$  represents some weighting function and that varies at a lag time  $\tau$  and some data  $x(t-\tau)$  collected at a past time  $(t-\tau)$ . The "running mean" is over the length of the filter, infinite as the limits of the integral indicates

Convolution Theorem says that if

$$\begin{aligned}
 Y(f) &= \mathcal{F}(y(t)) \\
 H(f) &= \mathcal{F}(h(t)) \\
 X(f) &= \mathcal{F}(x(t))
 \end{aligned}
 \quad \mathcal{F}(\cdot) \equiv \int_{-\infty}^{\infty} \cdot e^{-j2\pi ft} dt$$

and

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

then

$$Y(f) = H(f) \cdot X(f)$$

A <sup>convolution</sup> multiplication in the time domain of two time series corresponds to a multiplication in the frequency domain (the reverse holds also, i.e., a multiplication in the time domain corresponds to a convolution in the frequency domain).

Proof.: 
$$Y(f) = \int_{-\infty}^{\infty} \underbrace{\int_{-\infty}^{\infty} h(\tau) x(t-\tau) d\tau}_{g(t)} e^{-j2\pi ft} dt$$

$\mathcal{F}\{g(t)\}$

*does not depend on  $t$*       *does not depend on  $\tau$*

$$= \int_{-\infty}^{\infty} h(\tau) \int_{-\infty}^{\infty} x(t-\tau) e^{-j2\pi ft} dt d\tau$$

*switch order of integrals*

$\sigma = t - \tau$   
 $\frac{d\sigma}{dt} = 1$

$$= \int_{-\infty}^{\infty} h(\tau) \int_{-\infty}^{\infty} x(\sigma) e^{-j2\pi f(\sigma + \tau)} d\sigma d\tau$$

$\sigma = t - \tau$

$$= \int_{-\infty}^{\infty} h(\tau) e^{-j2\pi f\tau} \underbrace{\int_{-\infty}^{\infty} x(\sigma) e^{-j2\pi f\sigma} d\sigma}_{X(f)} d\tau$$

~~$\mathcal{F}\{x(t-\tau)\} = X(f)$~~  *does not depend on  $\sigma$  or  $t$  or  $\tau$*

$$= X(f) \underbrace{\int_{-\infty}^{\infty} h(\tau) e^{-j2\pi f\tau} d\tau}_{\mathcal{F}\{h(\tau)\} = H(f)}$$

$$= H(f) \cdot X(f)$$

### Properties of convolution

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

(ii)  $h * (a * x) = (h * a) * x$  associative

(iii)  $h * (x_1 + x_2) = h * x_1 + h * x_2$  distributive with respect to addition

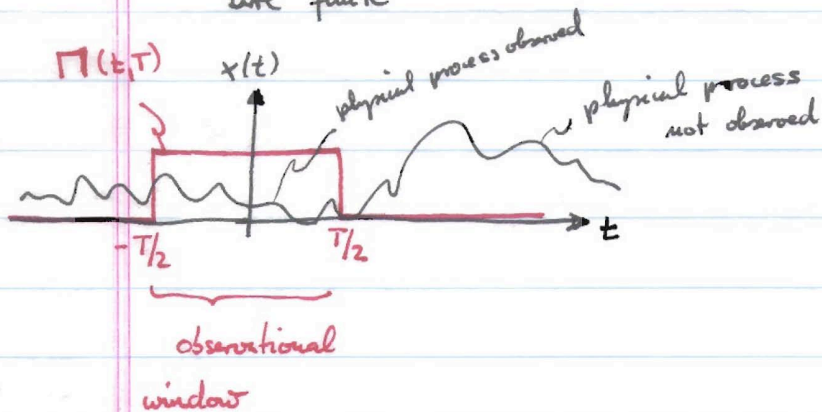
(iv)  $\mathcal{F}(h * x) = H \cdot X$  convolution theorem

class #3



### Record length

Now, apply the convolution theorem to data that are continuous but are finite



$$x(t) = \cos 2\pi f \cdot t$$

$$\hat{x}(t, T) = \cos(2\pi f \cdot t) \cdot \Pi(t, T)$$

$$\mathcal{F}(\Pi(t, T)) = T \operatorname{sinc}(fT) = T \frac{\sin(fT)}{\pi f T}$$

$$\mathcal{F}(\hat{x}(t, T)) =$$

$$\begin{aligned}
\tilde{F}(\hat{x}(t, T)) &= \tilde{F}(x(t)) * \tilde{F}(\Pi(t, T)) \\
&= X(f) * T \cdot \text{sinc}(fT) \\
&= \int_{-\infty}^{\infty} T \text{sinc}(f'T) X(f-f') df' \\
&= \int_{-\infty}^{\infty} T \text{sinc}(f'T) \frac{1}{2} [\delta(f+f_0-f') + \delta(f-f_0-f')] df' \\
&= \frac{T}{2} \left\{ \text{sinc}[(f+f_0)T] + \text{sinc}[(f-f_0)T] \right\}
\end{aligned}$$

Compare this result with the earlier one, i.e., on p. 14

$$\tilde{F}(x(t)) = \frac{1}{2} \left\{ \delta(f+f_0) + \delta(f-f_0) \right\}$$

*this is referred to as leakage*

$$\Rightarrow \tilde{F}(x(t) \Pi(t, T)) \neq \tilde{F}(x(t))$$

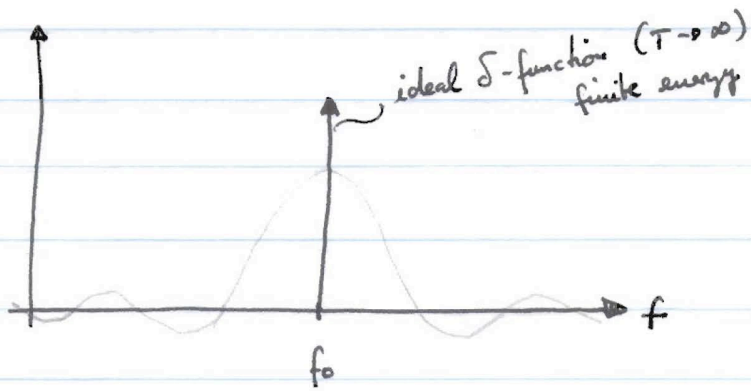
But

$$\lim_{T \rightarrow \infty} T \text{sinc}[(f+f_0)T] = \delta(f+f_0)$$

~~Useful~~

The function  $\text{sinc } fT = \frac{\sin(\pi fT)}{\pi fT}$

needs to be made wide in the time domain ( $T \rightarrow \infty$ )  
in order to reflect the narrow  $\delta$ -function like character in  
the frequency domain (little leakage, little sidelobes of the sinc fct.)



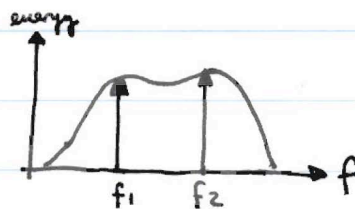
$\frac{2}{T}$

width of the 1<sup>st</sup> sidelobe 93% of energy

1. resolution problem arises as the sidelobes of two sinc fctn run into each other

2. false "mean" or "zero frequency" component

3. two frequencies may overlap



$(f_2 - f_1) < \frac{2}{T}$

but for  $A_1 \gg A_2$ :

