



SIS Scuola Interdipartimentale
delle Scienze, dell'Ingegneria
e della Salute



L. Magistrale in IA (ML&BD)

Scientific Computing
(part 2 – 6 credits)

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The background features a large, faint watermark of the University of Naples Federico II seal. The seal is circular and contains a central figure of a woman (Minerva) holding a cornucopia. The text around the seal includes '1920 - 2020' at the top, 'UNIVERSITA' DEGLI STUDI FEDERICO II' around the inner border, 'ARTHENOPE' at the bottom, and '100° ANNIVERSARIO' at the very bottom.

Contents

- **Windowing and aliasing.**
- **The Nyquist-Shannon Sampling Theorem.**

Windowing effect

When performing spectral analysis of a sample from a physical phenomenon $f(t)$, we are actually using an observation of the phenomenon obtained over a finite interval of time.

From a mathematical point of view, this process is equivalent to the multiplication of the function $f(t)$, which describes the phenomenon, by a **rectangular function $w(t)$** (**rectangular window function of width L**); thus the “observed” function is:

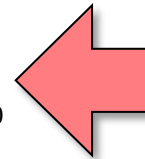
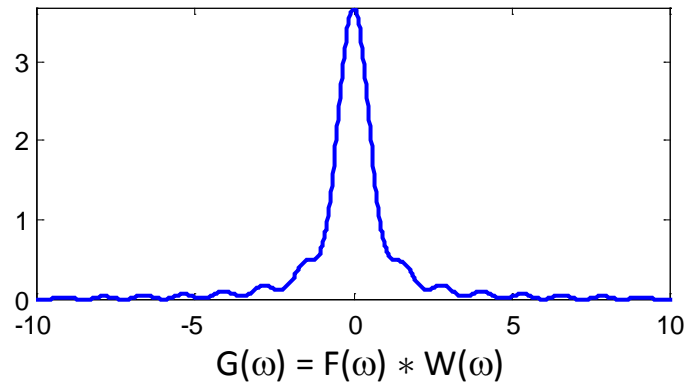
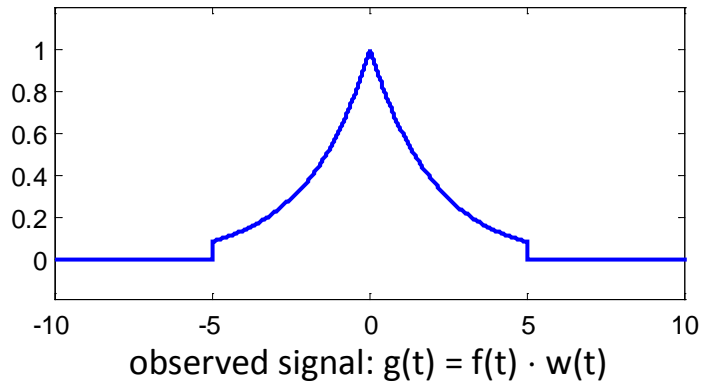
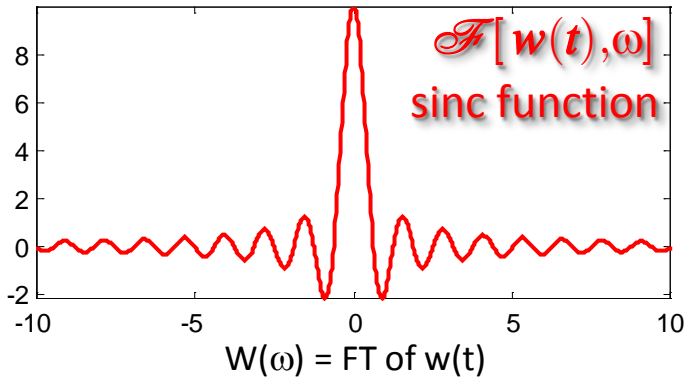
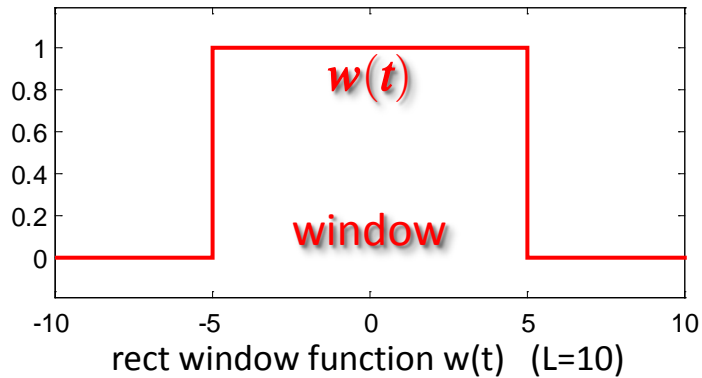
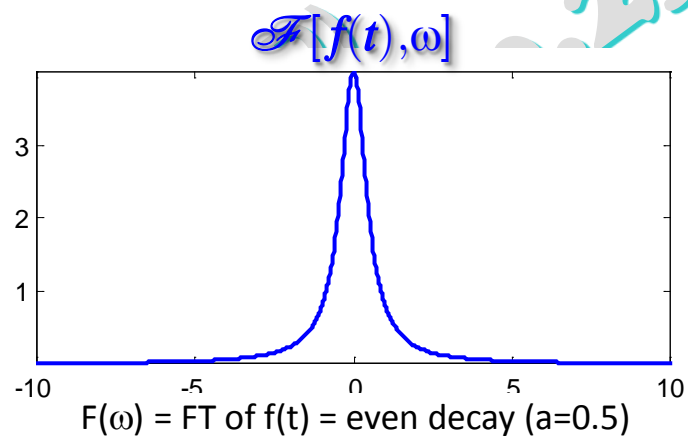
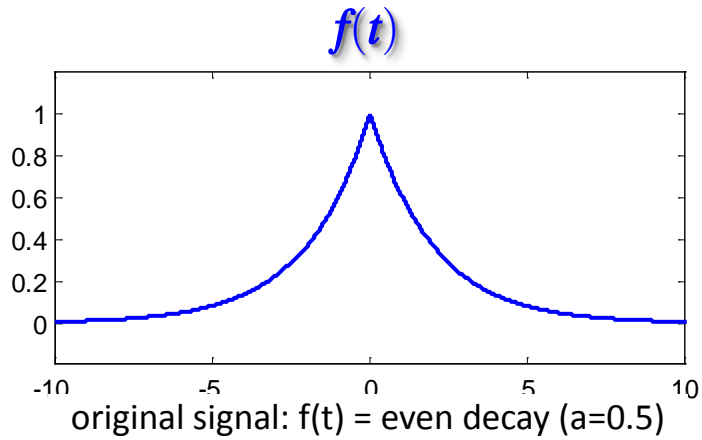
$$g(t) = f(t) \cdot w(t) \qquad \begin{aligned} F(\nu) &= \mathcal{F}[f(t), \nu] \\ W(\nu) &= \mathcal{F}[w(t), \nu] \end{aligned}$$

so that the **Fourier Transform $G(\nu)$** of $g(t)$ is then given by:

$$G(\nu) = \mathcal{F}[g(t), \nu] = \mathcal{F}[f(t), \nu] * \mathcal{F}[w(t), \nu] = \mathbf{F(\nu) * W(\nu)}$$

where $\mathbf{F(\nu) * W(\nu)}$ represents the **convolution product** of \mathbf{F} and \mathbf{W} . Remember that \mathbf{W} is a **sinc()**.

Windowing effect: example

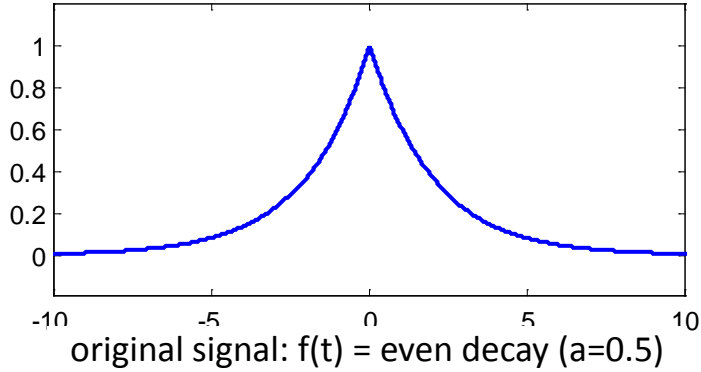


secondary oscillations

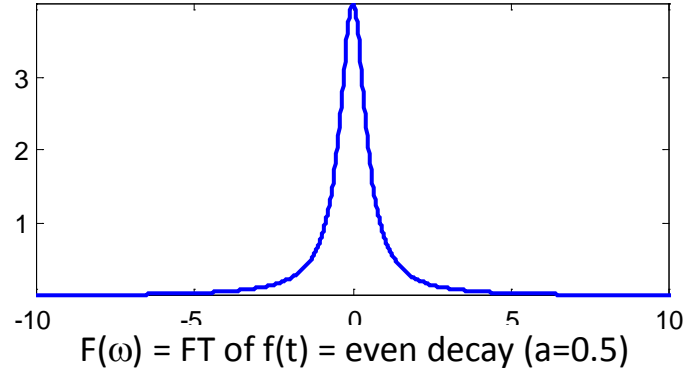
Windowing effect: example (cont.)

the window has been shrunk

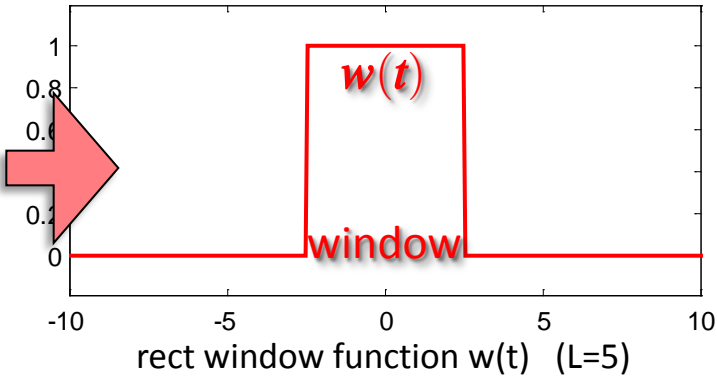
$f(t)$



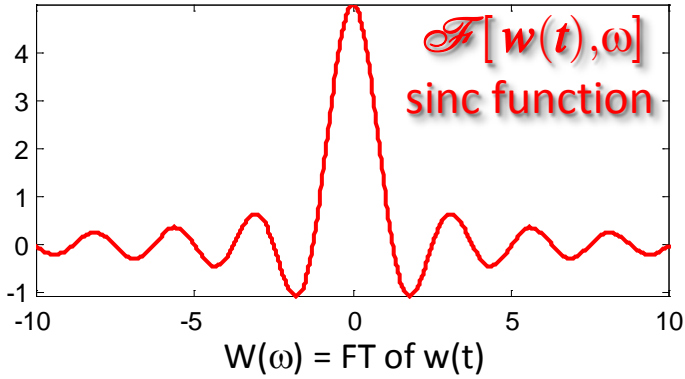
$\mathcal{F}[f(t), \omega]$



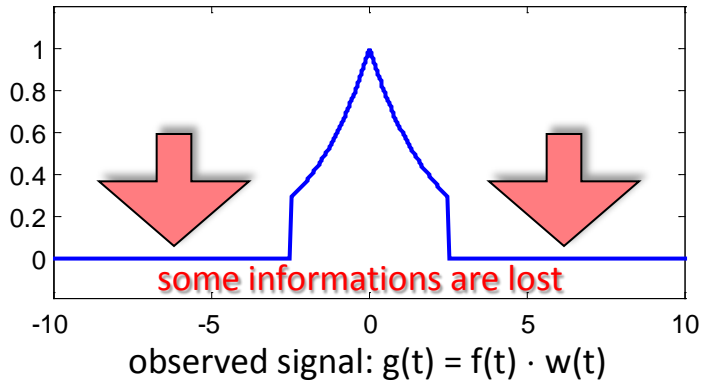
$w(t)$



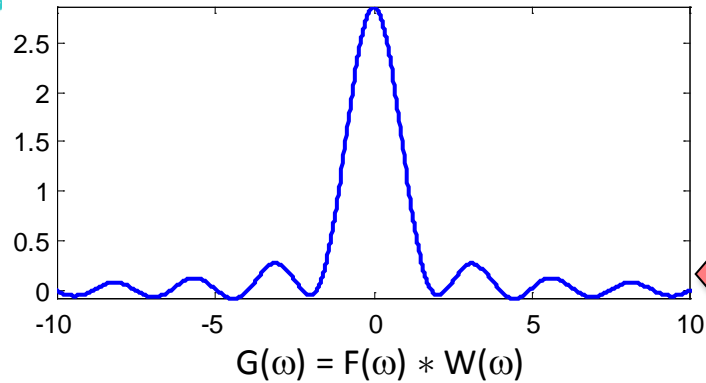
$\mathcal{F}[w(t), \omega]$
sinc function



some informations are lost



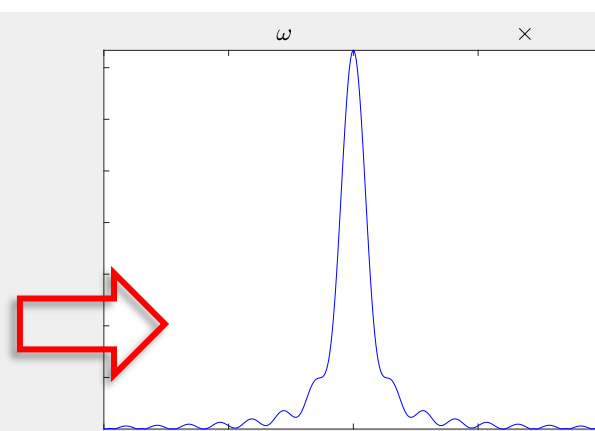
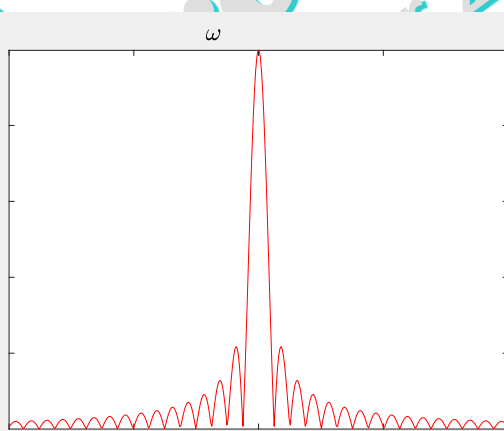
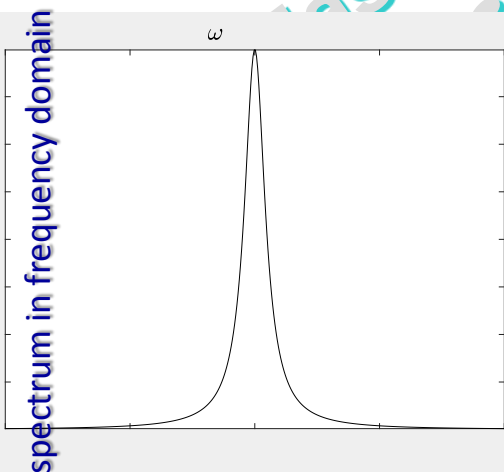
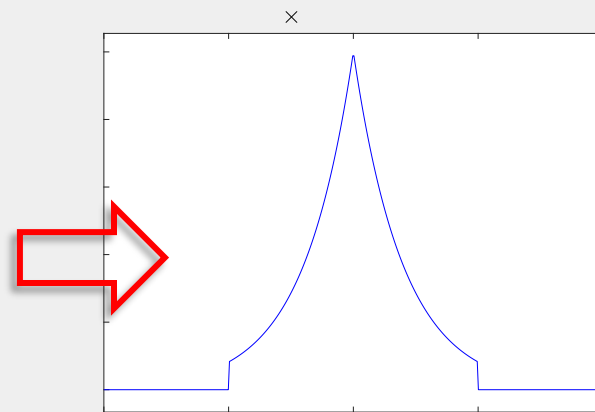
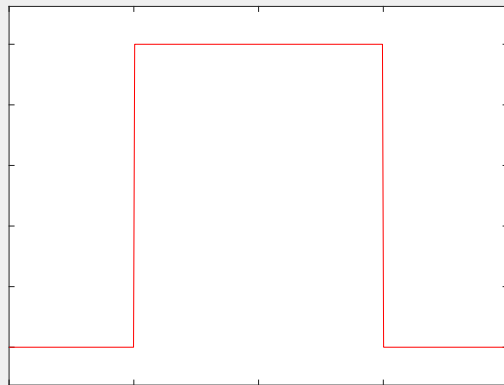
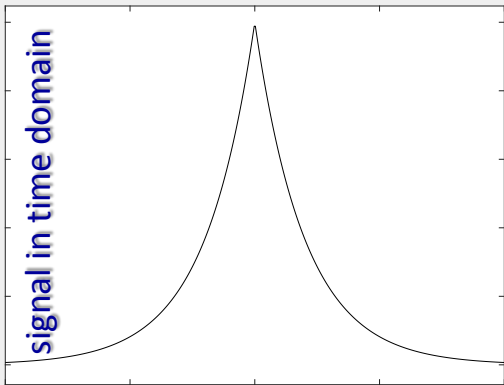
$G(\omega) = F(\omega) * W(\omega)$



secondary oscillations
increase in amplitude

Example by means of Symbolic Math Toolbox

```
syms t real; f=exp(-abs(t)/2); figure; ezplot(f,[-10 10])  
L=10; w=rectangularPulse(-L/2,+L/2,t); figure; ezplot(w,[-10 10])  
figure; ezplot( f*w, [-10 10])  
F=fourier(f); figure; h=ezplot(abs(F),[-10 10]); set(h,'Color','k')  
W=fourier(w); figure; h=ezplot(abs(W),[-10 10]); set(h,'Color','r')  
G=fourier(f*w); figure; h=ezplot(abs(G),[-10 10]); set(h,'Color','b')
```



Exercise: What happens for $L=5$?

Aliasing effect

When performing spectral analysis of a sample from a physical phenomenon $f(t)$, we are actually using a discrete set of observations of the phenomenon obtained by equispaced time series.

From a mathematical point of view, this process is equivalent to the multiplication of the function $f(t)$, which describes the phenomenon, by a **Comb function** $\delta_T(t)$ (of step **T**); thus the “observed” function is:

$$h(t) = f(t) \cdot \delta_T(t) \quad \begin{aligned} F(\nu) &= \mathcal{F}[f(t), \nu] \\ \delta_{1/T}(\nu) &= \mathcal{F}[\delta_T(t), \nu] \end{aligned}$$

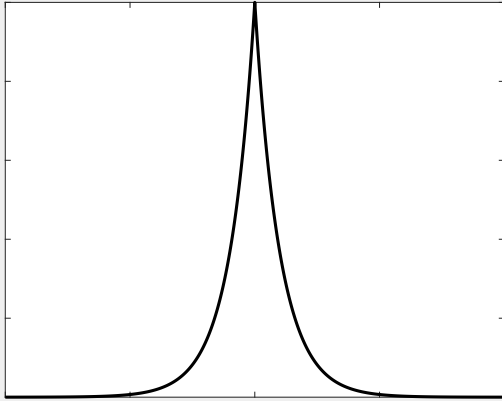
so that the **Fourier Transform** $H(\nu)$ of $h(t)$ is then given by:

$$H(\nu) = \mathcal{F}[h(t), \nu] = \mathcal{F}[f(t), \nu] * \mathcal{F}[\delta_T(t), \nu] = F(\nu) * \delta_{1/T}(\nu) / T$$

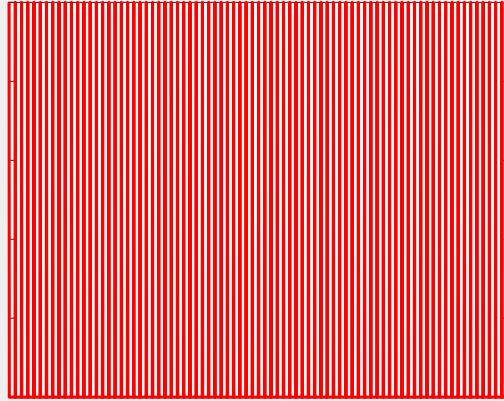
where $F(\nu) * \delta_{1/T}(\nu) / T$ represents the **convolution product** of F and $\delta_{1/T} / T$. Remember that the Fourier Transform of a comb function is still a comb function, but with a **period** equal to the **reciprocal** of the period of the original comb function.

Aliasing effect: example

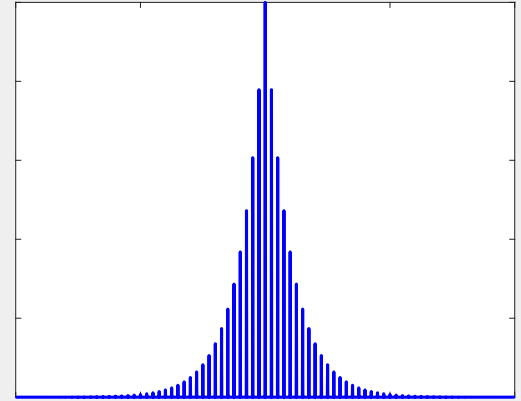
$f(t)$



$\delta_T(t)$



$f(t) \cdot \delta_T(t)$

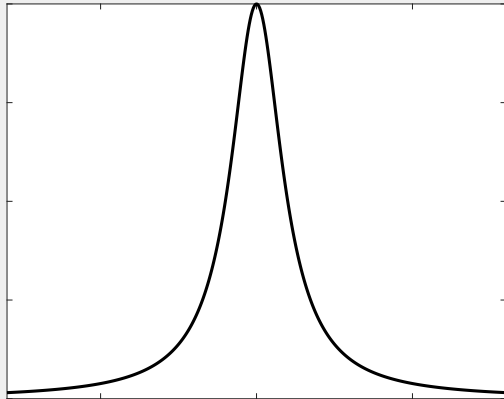


original signal: $f(t) = \text{even decay } e^{-|t|}$

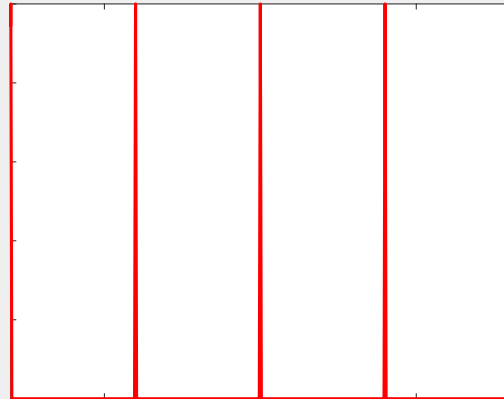
pulse train $\delta_T(t)$, $T = 0.25$

observed signal: $h(t) = f(t) \cdot \delta_T(t)$

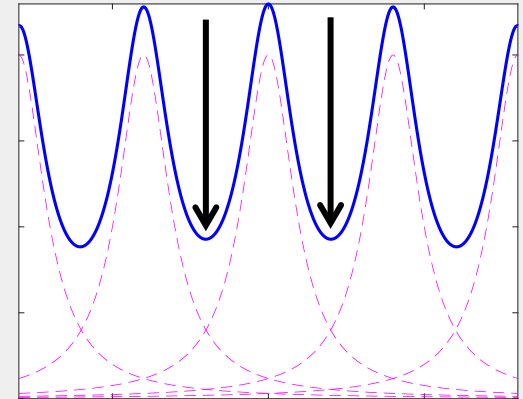
$\mathcal{F}[f(t), \omega]$



$\mathcal{F}[\delta_T(t), \omega]$



$\mathcal{F}[f(t), \omega] * \mathcal{F}[\delta_T(t), \omega]$
Nyquist frequency

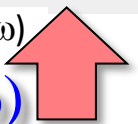


$F(\omega) = 2/(1+\omega^2) = \text{FT of } f(t) = e^{-|t|}$

$\delta_p(\omega) = \text{FT of } \delta_T(t)$, $P = 1/T = 4$

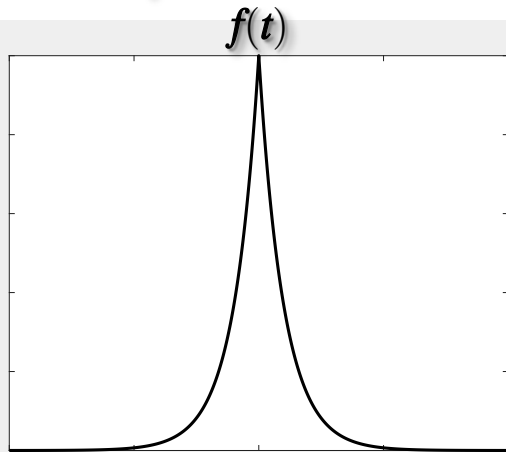
$H(\omega) = F(\omega) * \delta_{1/T}(\omega)$

Superposition of $F(\omega)$

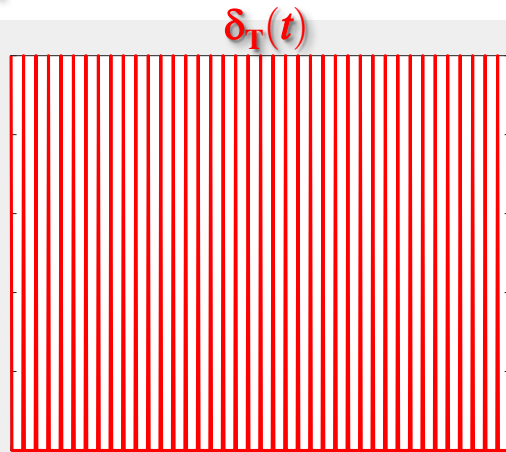


Aliasing effect: example (cont.)

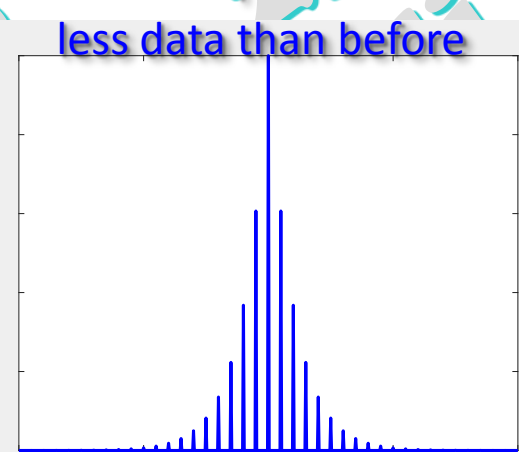
Now, let's double the period T of the Comb Function δ_T



$f(t)$



$\delta_T(t)$



less data than before

original signal: $f(t) = \text{even decay } e^{-|t|}$

$$\mathcal{F}[f(t), \omega]$$

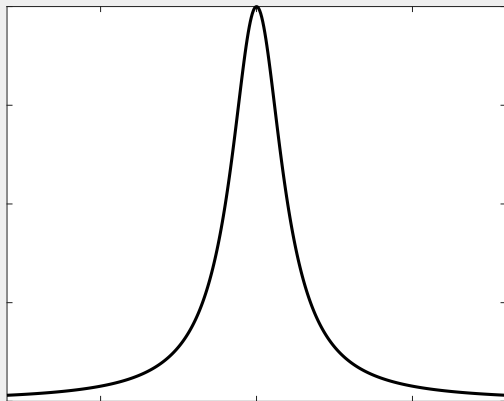
pulse train $\delta_T(t)$, $T = 0.5$

$$\mathcal{F}[\delta_T(t), \omega]$$

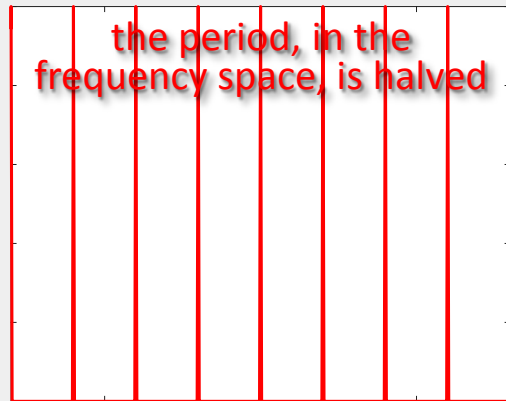
observed signal: $h(t) = f(t) \cdot \delta_T(t)$

$$\mathcal{F}[f(t), \omega] * \mathcal{F}[\delta_T(t), \omega]$$

Nyquist frequency

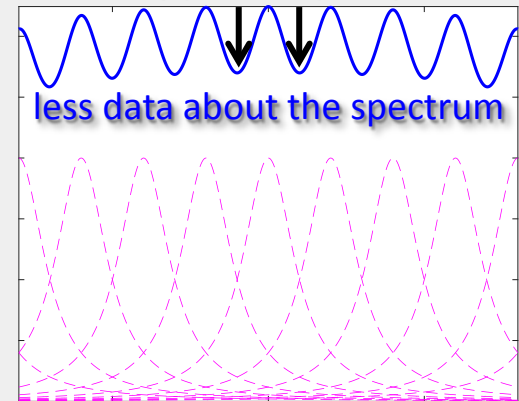


$$F(\omega) = 2/(1+\omega^2) = \text{FT of } f(t) = e^{-|t|}$$



$$\delta_p(\omega) = \text{FT of } \delta_T(t), P = 1/T = 2$$

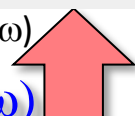
the period, in the frequency space, is halved



less data about the spectrum

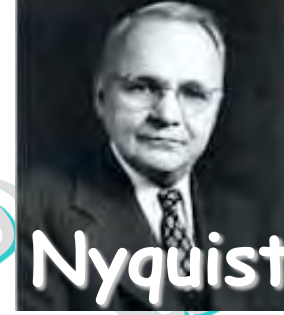
$$H(\omega) = F(\omega) * \delta_{1/T}(\omega)$$

Superposition of $F(\omega)$



The Nyquist-Shannon Sampling Theorem

stated by Harry Nyquist and proved by Claude Shannon in 1948



Nyquist



Shannon

w.r.t. the circular frequency ν

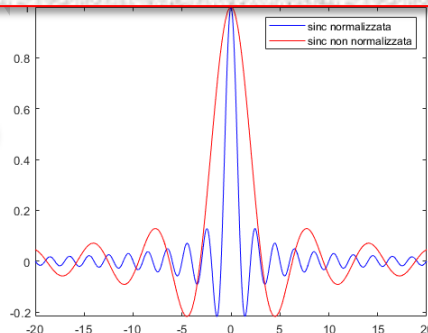
If the *FT* of $f(t)$, $F(\nu)$ [$\omega=2\pi\nu$], is such that

$$F(\nu)=0, |\nu| > H/2 \quad (f \text{ is a band limited function of bandwidth } H)$$

then the values of f at the sampling points $t_k=k\Delta t$, taken with frequency $f_s=1/\Delta t$, if $f_s > H$ (greater than twice the maximum frequency) allow f to be fully restored, i.e.:

$$f(t) = \sum_{k=-\infty}^{+\infty} f(k\Delta t) \operatorname{sinc}\left(\frac{t}{\Delta t} - k\right)$$

$$\operatorname{sinc}(x) = \frac{\sin(\pi x)}{\pi x}$$



sinc:
“sine cardinal” function

new expansion basis

normalized sinc: $\sin(\pi x)/(\pi x)$
unnormalized sinc: $\sin(x)/(x)$

The **Aliasing Error** is governed by the **Sampling Theorem**.

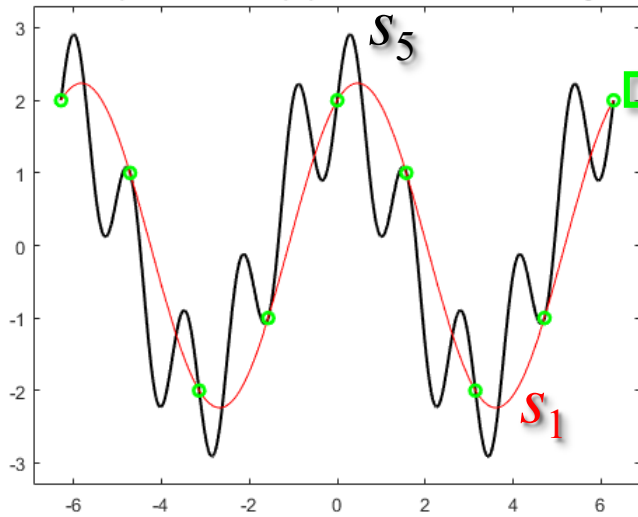
The **Sampling Theorem** can be thought of as the conversion of an analog signal into a discrete form by taking the sampling frequency at least as twice the maximum frequency of the input analog signal.

This **Theorem** is the basis of digital recordings able to reach a degree of fidelity much higher than analog ones.

Example: application of Sampling Theorem

Let us consider two signals: $s_5 = 2\cos(t) + \sin(5t)$ and $s_1 = 2\cos(t) + \sin(t)$.

We want to reconstruct s_5 starting from a sequence of its equispaced samples. We sample s_5 with a step $h = \pi/2$ in the interval $[-2\pi, +2\pi]$ getting 9 samples.



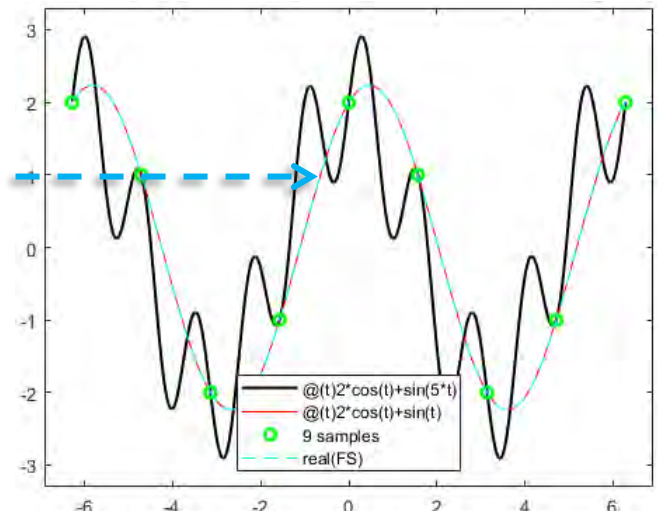
But s_5 and s_1 share these samples and we are not able to distinguish the two signals starting from these 9 samples, since the sampling rate **does not satisfy** the Sampling Theorem.

max freq in s_5 : $\omega_{\max} = 5 \Rightarrow v = \omega / (2\pi)$

min useful freq for s_5 : $f_s > 2v_{\max} = 5/\pi$

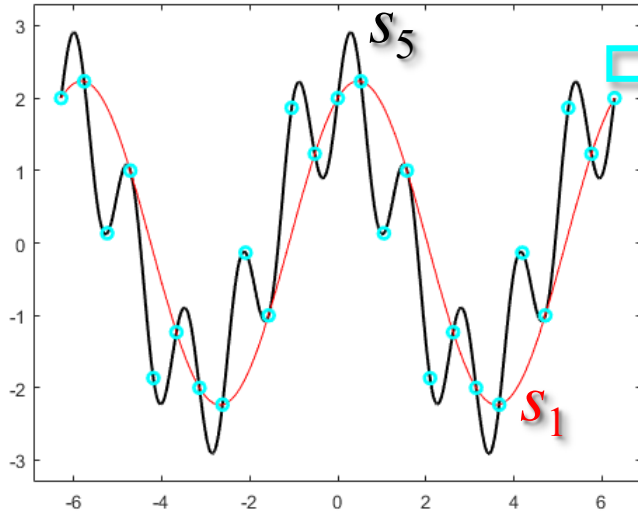
According to the **Sampling Theor.** we should use a step: $\Delta t < \pi/5$ but used step = $\pi/2 > \pi/5$

In facts, if we reconstruct a signal by means of the **Fourier Series** starting from the 9 samples (whose frequency is $f_s = 2/\pi < 5/\pi = 2v_{\max}$), we get s_1 and not s_5 (aliasing).



Example: application of Sampling Theorem (cont.)

Now we sample s_5 with a step $h=\pi/6 < \pi/5$, that satisfies the Sampling Theor.: we get 25 samples.



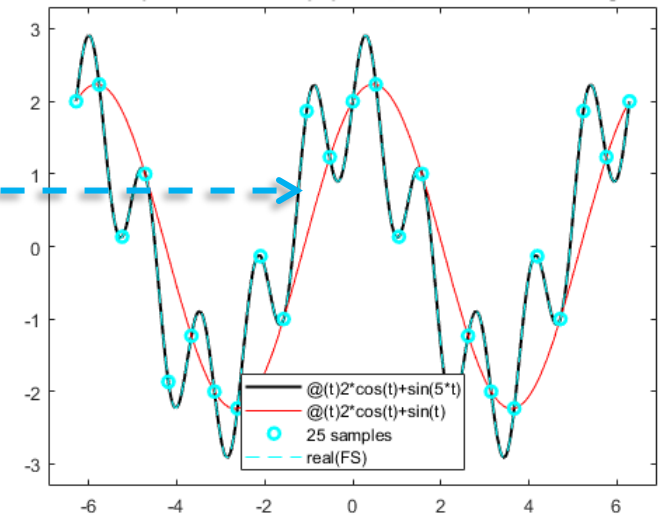
The figure shows that some of the 25 samples of s_5 no longer fall on s_1 , so that we are able to distinguish the two signals starting from these 25 samples.

max freq in s_5 : $\omega_{\max} = 5 \Rightarrow v = \omega / (2\pi)$

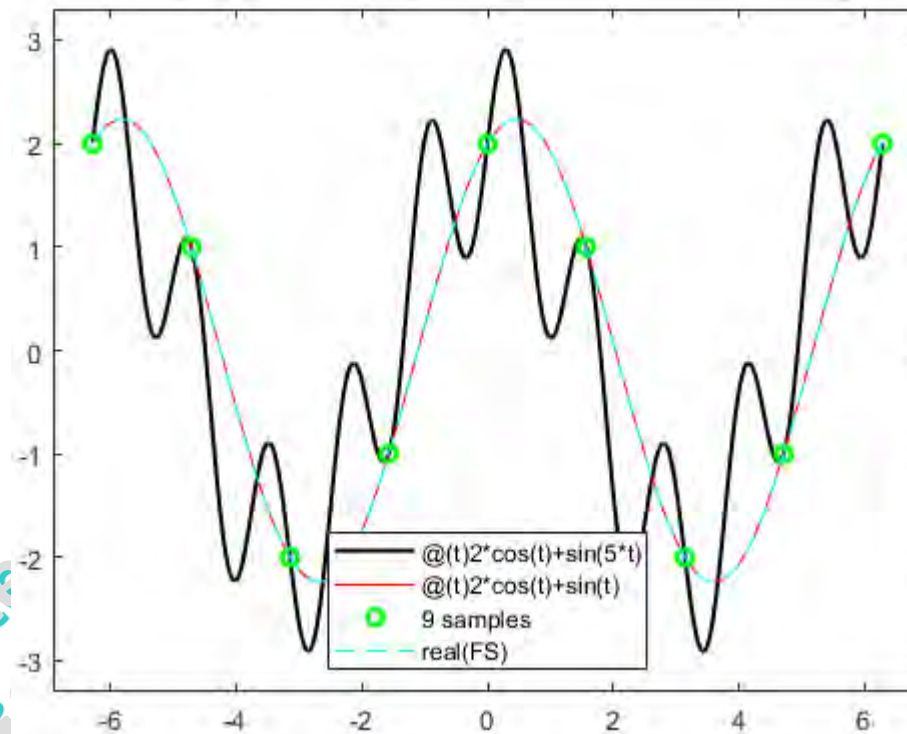
min useful freq for s_5 : $f_s > 2v_{\max} = 5/\pi \Rightarrow$

According to the Sampling Theor. we should use a step: $\Delta t < \pi/5$ and used step = $\pi/6 < \pi/5$

In facts, if we reconstruct a signal by means of the **Fourier Series** starting from the 25 samples (whose frequency is $f_s = 6/\pi > 5/\pi = 2v_{\max}$), now we get s_5 (no aliasing).



Quiz



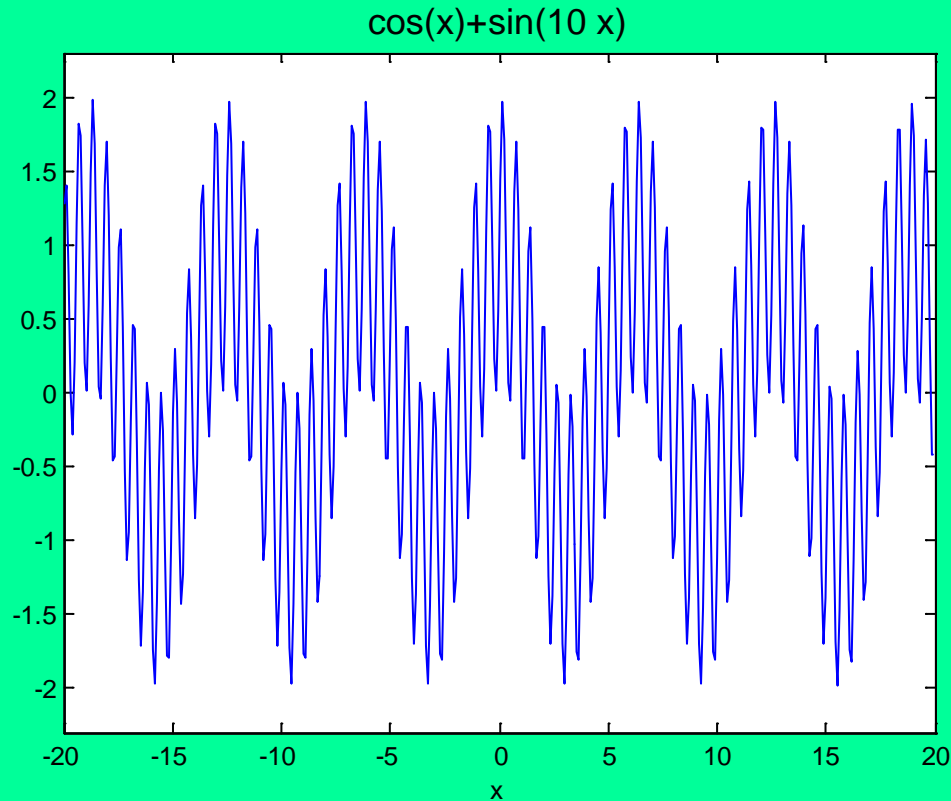
Why, starting from the samples of $s_5 = 2\cos(t) + \sin(5t)$ taken with frequency $f_s = 2/\pi < 5/\pi = 2\nu_{\max}$, are we able to reconstruct the signal $s_1 = 2\cos(t) + \sin(t)$ with angular frequency = 1, and not:

$$s_k = 2\cos(t) + \sin(kt), \quad k=2,3,4$$

with maximum frequency $\omega = k < 5$?

Exercise

What is the Nyquist frequency of the following signal?



Check it by means the **Sampling Theor.** and display what happens with a lower frequency.