

Corso di Laurea in Fisica - UNITS
ISTITUZIONI DI FISICA
PER IL SISTEMA TERRA

LINEAR SYSTEMS

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Green's function

Green's function (GF) is a basic solution to a linear differential equation, a building block that can be used to construct many useful solutions.

If one considers a linear differential equation written as:

$$L(x)u(x)=f(x)$$

where $L(x)$ is a linear, self-adjoint differential operator, $u(x)$ is the unknown function, and $f(x)$ is a known non-homogeneous term, the GF is a solution of:

$$L(x)u(x,s)=\delta(x-s)$$
$$G(x,s)$$

Why GF is important?

If such a function G can be found for the operator L , then if we multiply the second equation for the Green's function by $f(s)$, and then perform an integration in the s variable, we obtain:

$$\int L(x)G(x,s)f(s)ds = \int \delta(x-s)f(s)ds = f(x) = Lu(x)$$

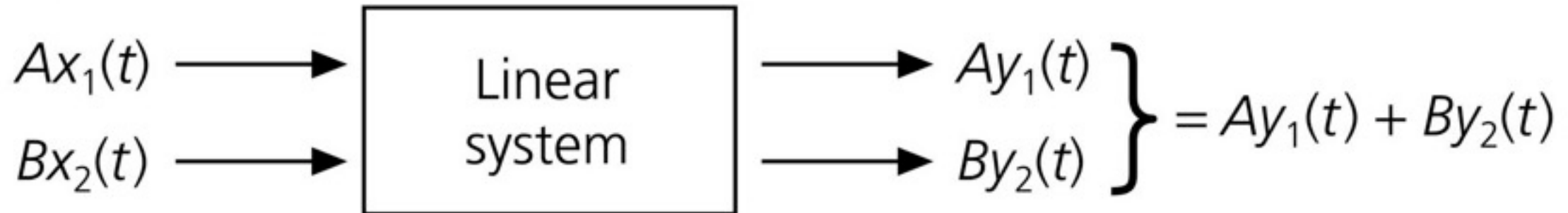
$$L \int G(x,s)f(s)ds = Lu(x)$$

$$u(x) = \int G(x,s)f(s)ds$$

Thus, we can obtain the function $u(x)$ through knowledge of the Green's function, and the source term. This process has resulted from the linearity of the operator L .

Linear Systems

Figure 6.3-1: Definition of a linear system.



$$x(t) = \int x(\tau)\delta(\tau - t)d\tau$$

$$\int x(\tau)h(\tau - t)d\tau$$

$$x(t) * h(t) = y(t)$$

(remember GF definition)

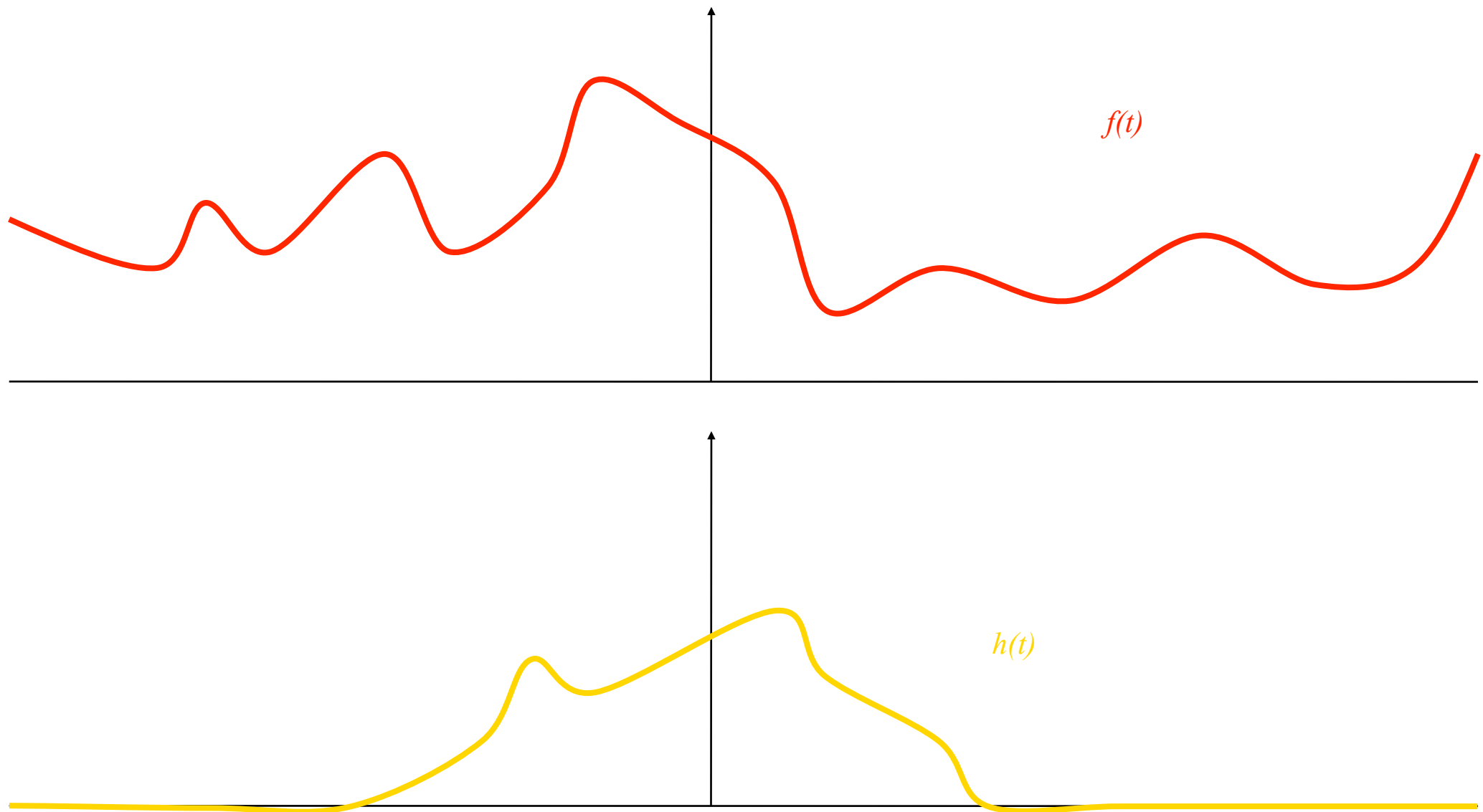
Convolution

 Definition:

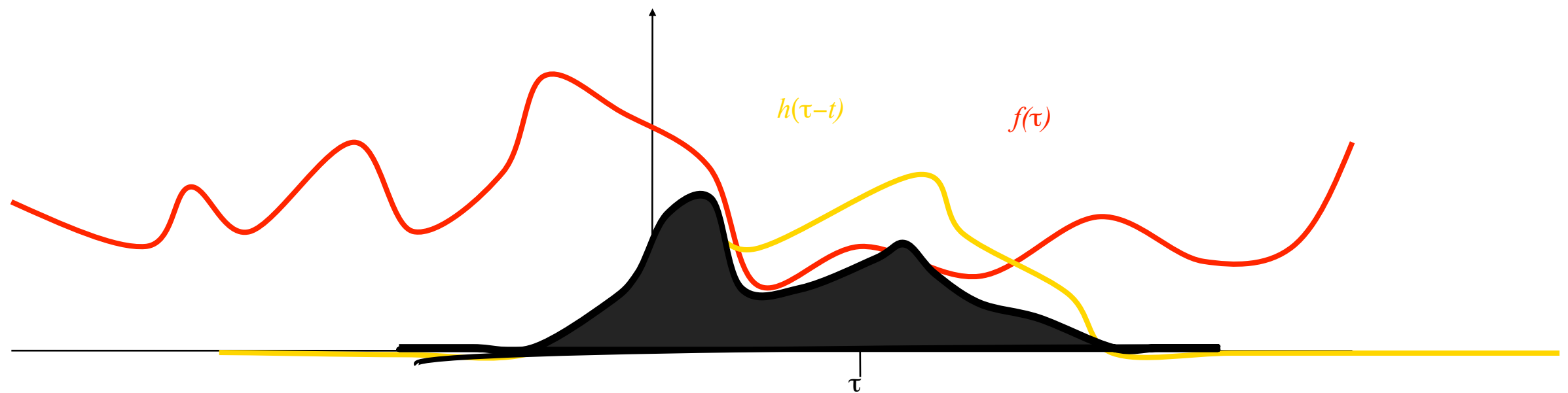
$$f(t) * h(t) = \int_{-\infty}^{\infty} f(\tau)h(t - \tau)d\tau$$

Convolution

Pictorially



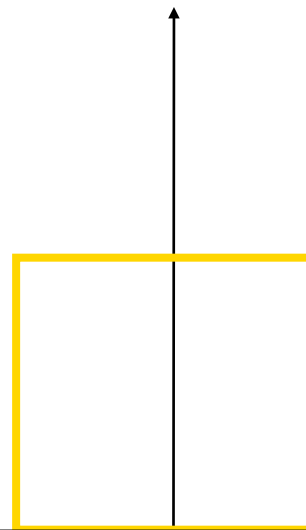
Convolution



Convolution

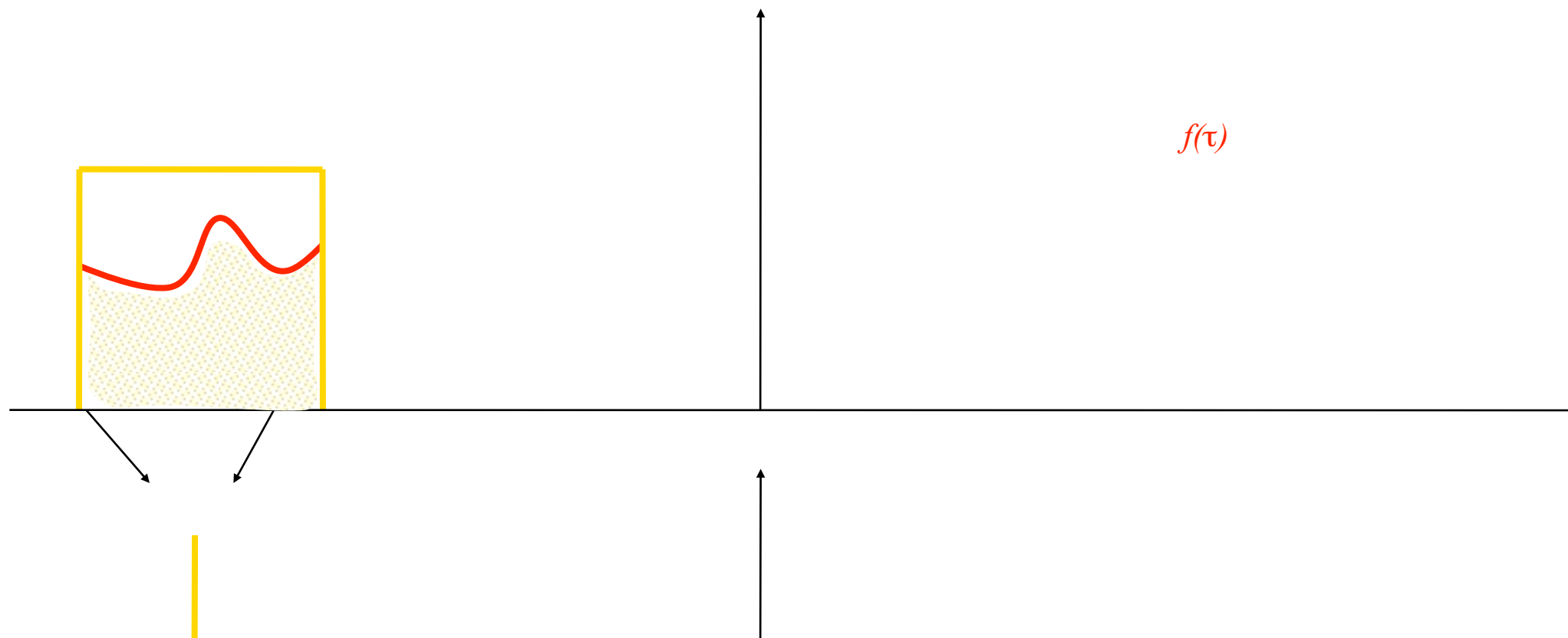
☑ Consider the function (box filter):

$$h(t) = \begin{cases} 0 & t < -\frac{1}{2} \\ 1 & -\frac{1}{2} \leq t \leq \frac{1}{2} \\ 0 & t > \frac{1}{2} \end{cases}$$



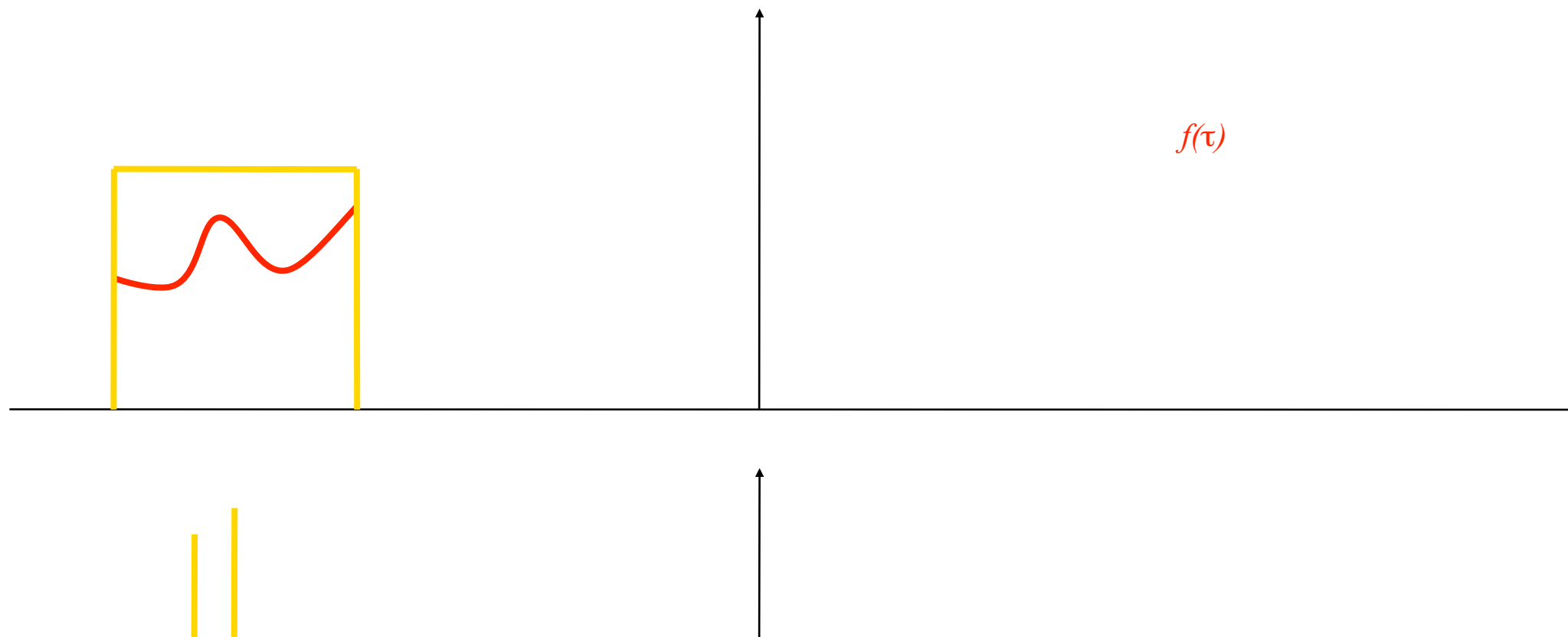
Convolution

This function windows our function $f(t)$



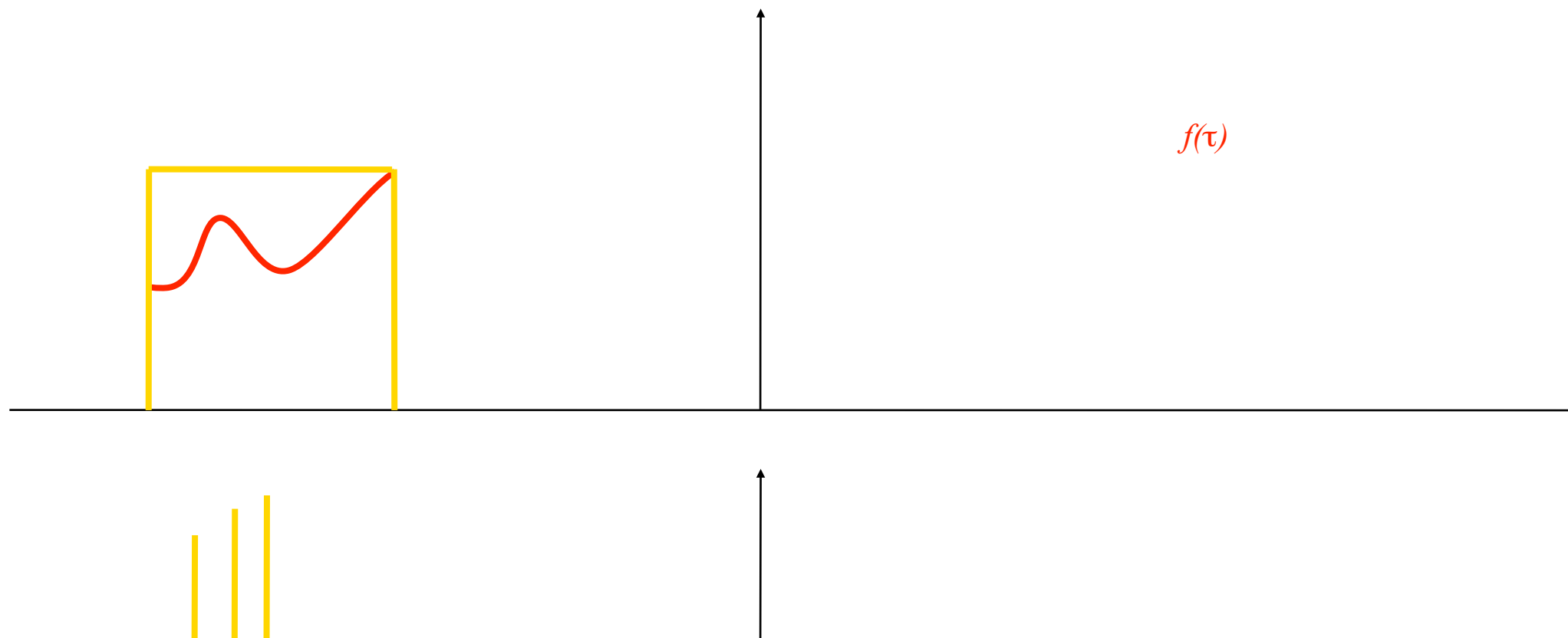
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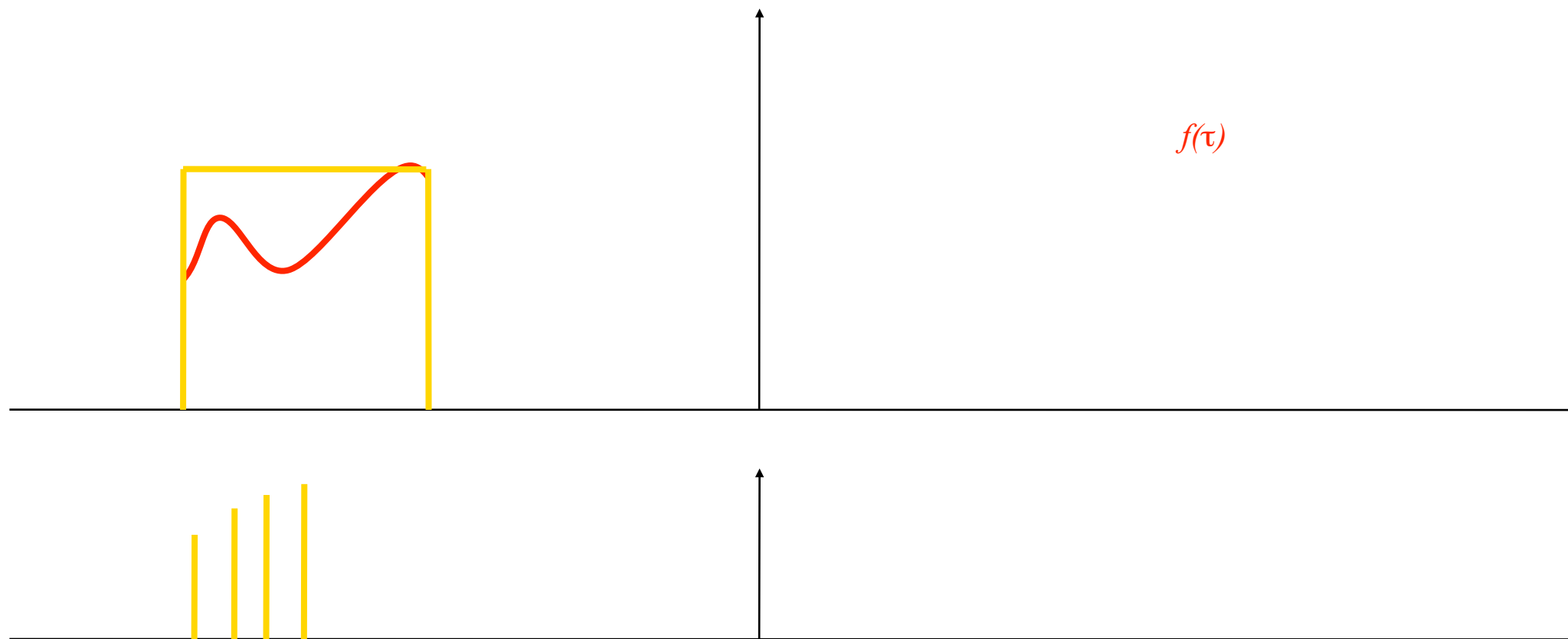
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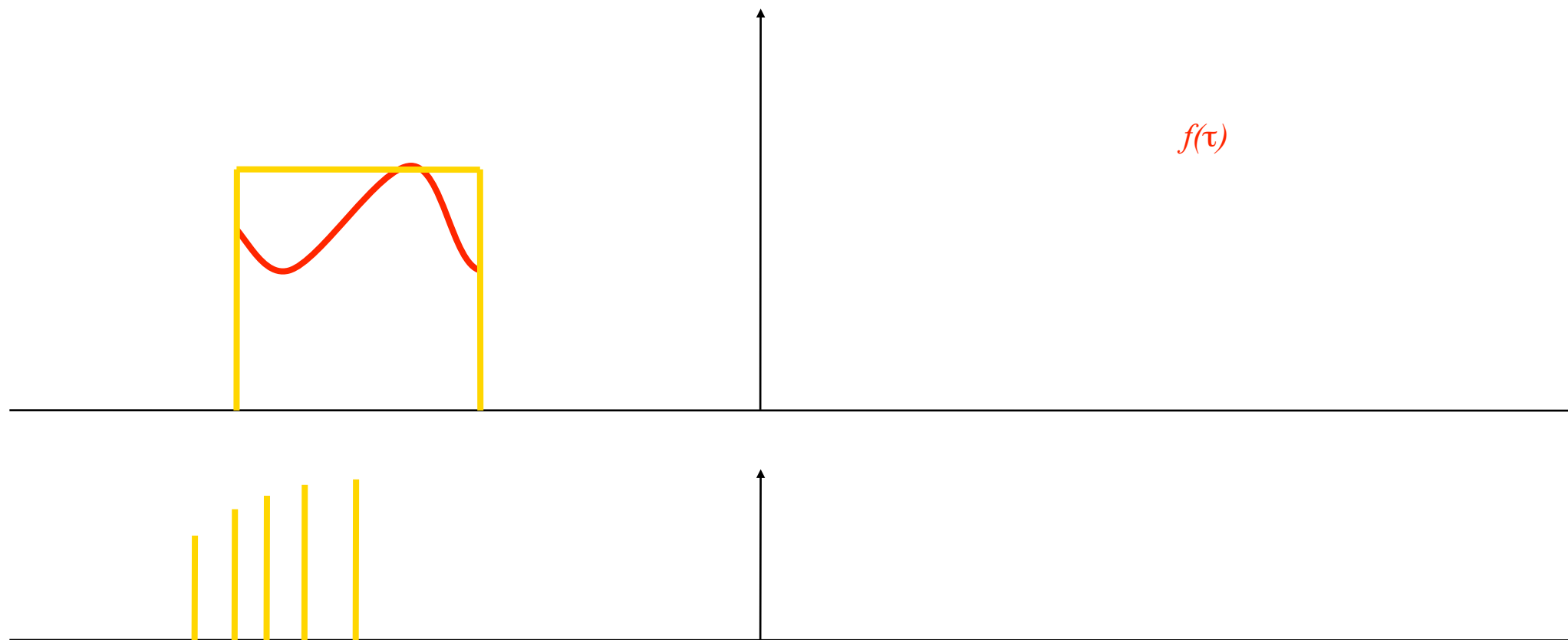
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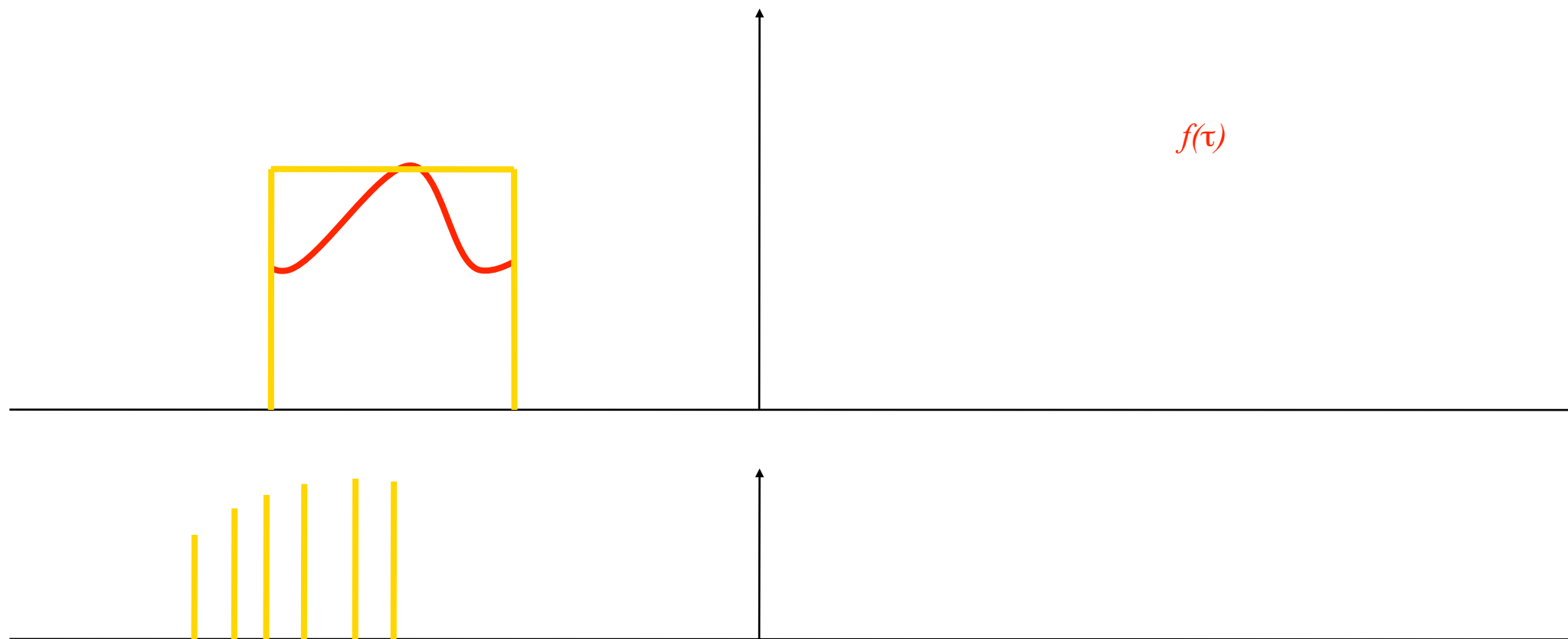
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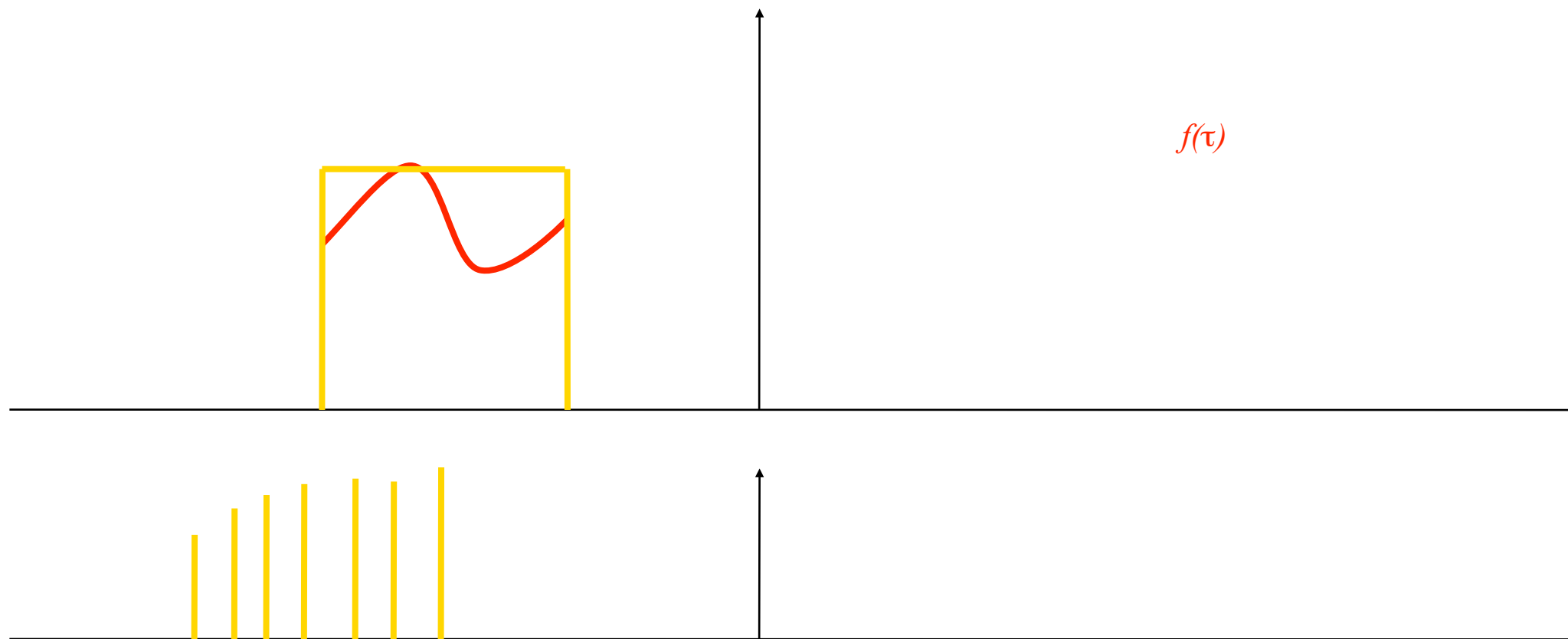
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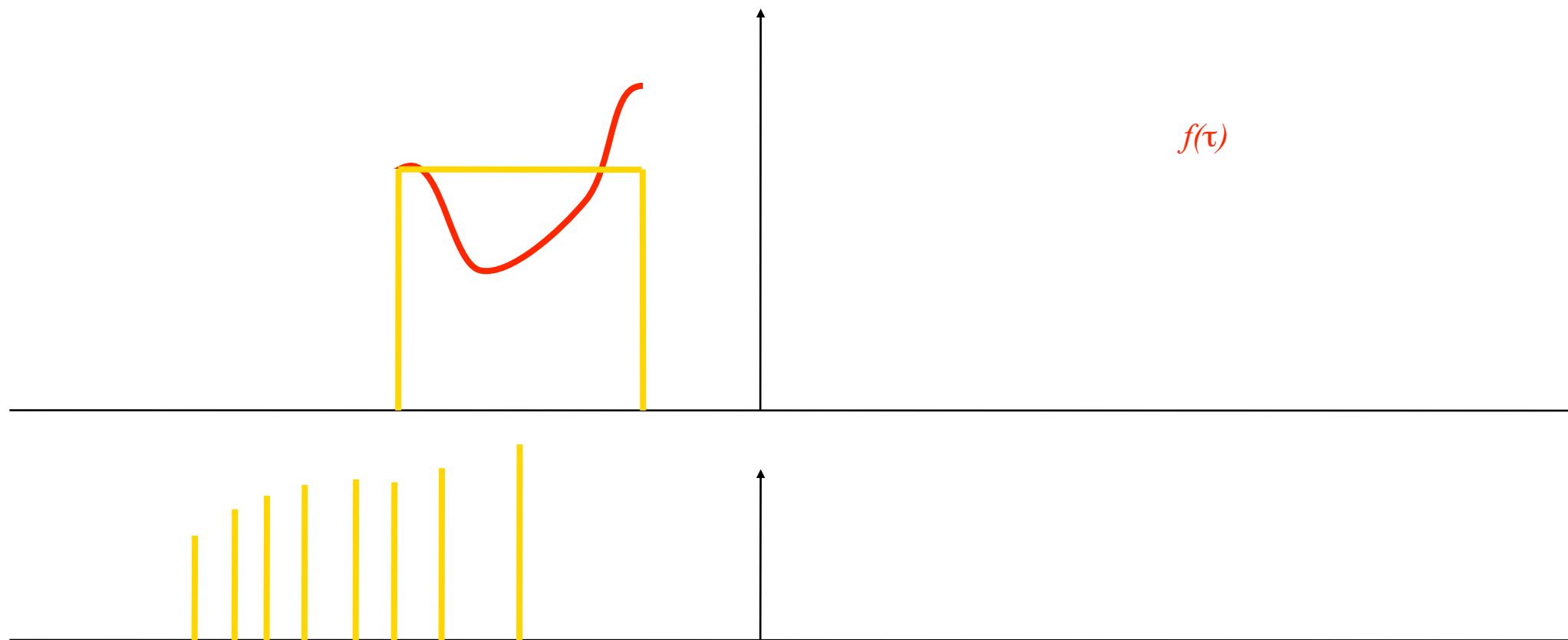
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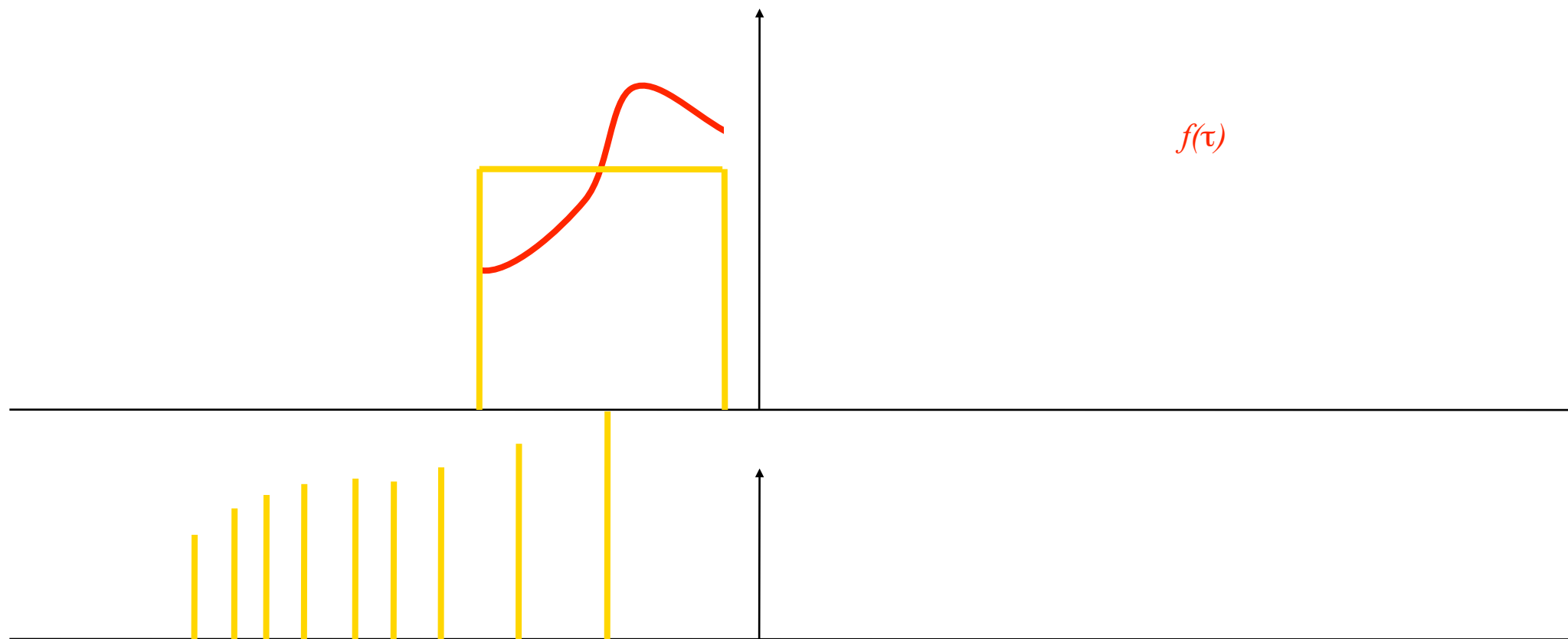
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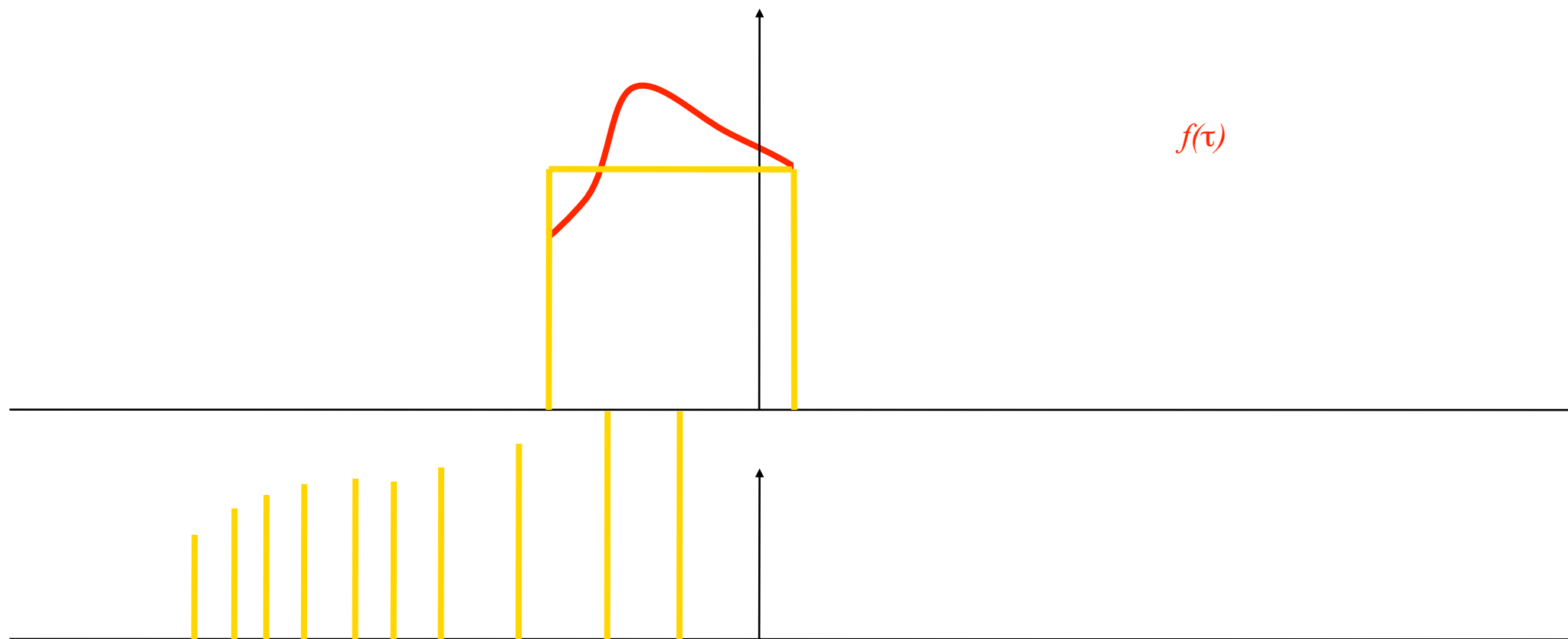
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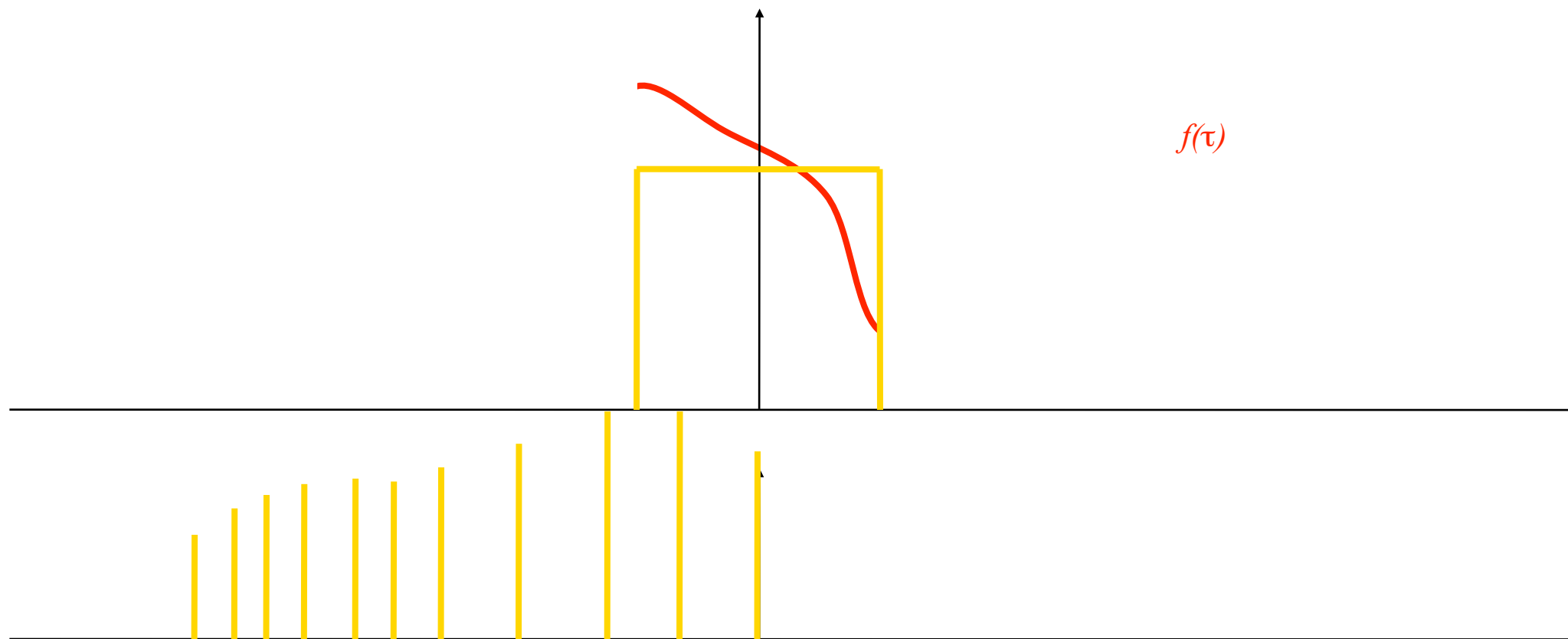
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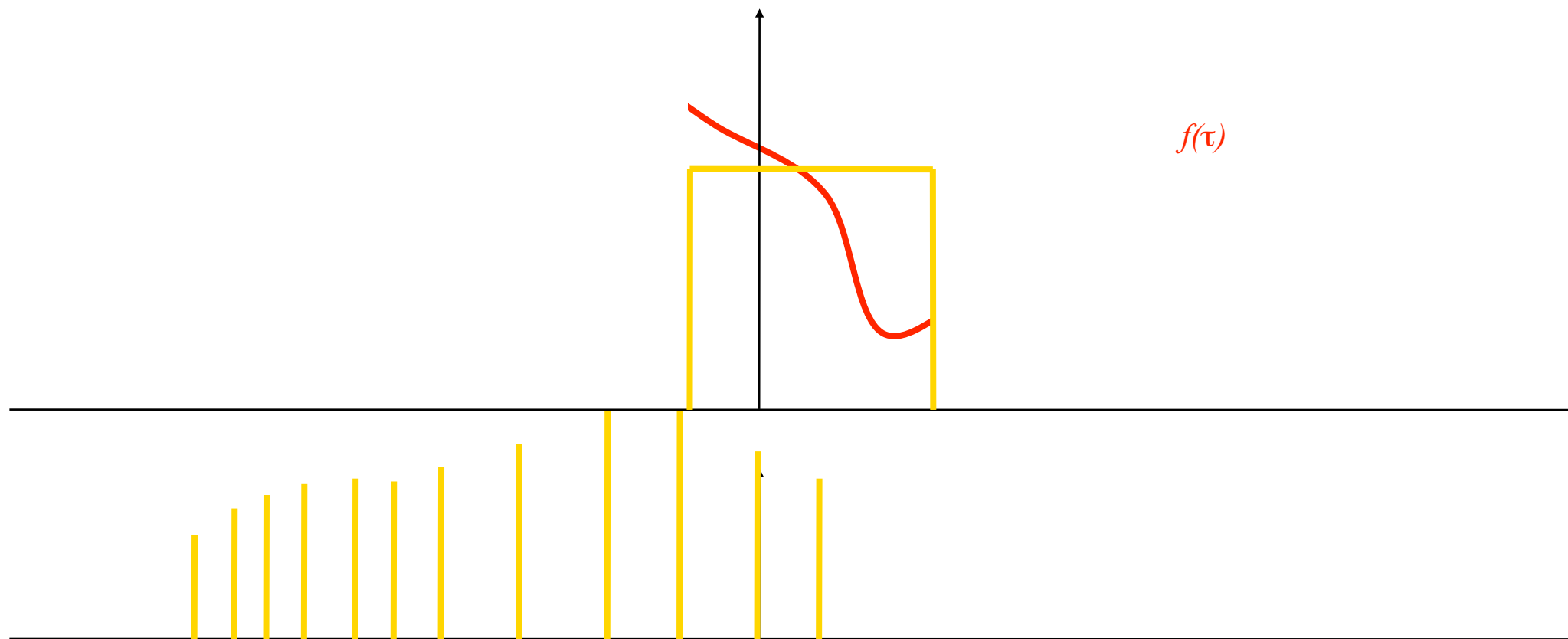
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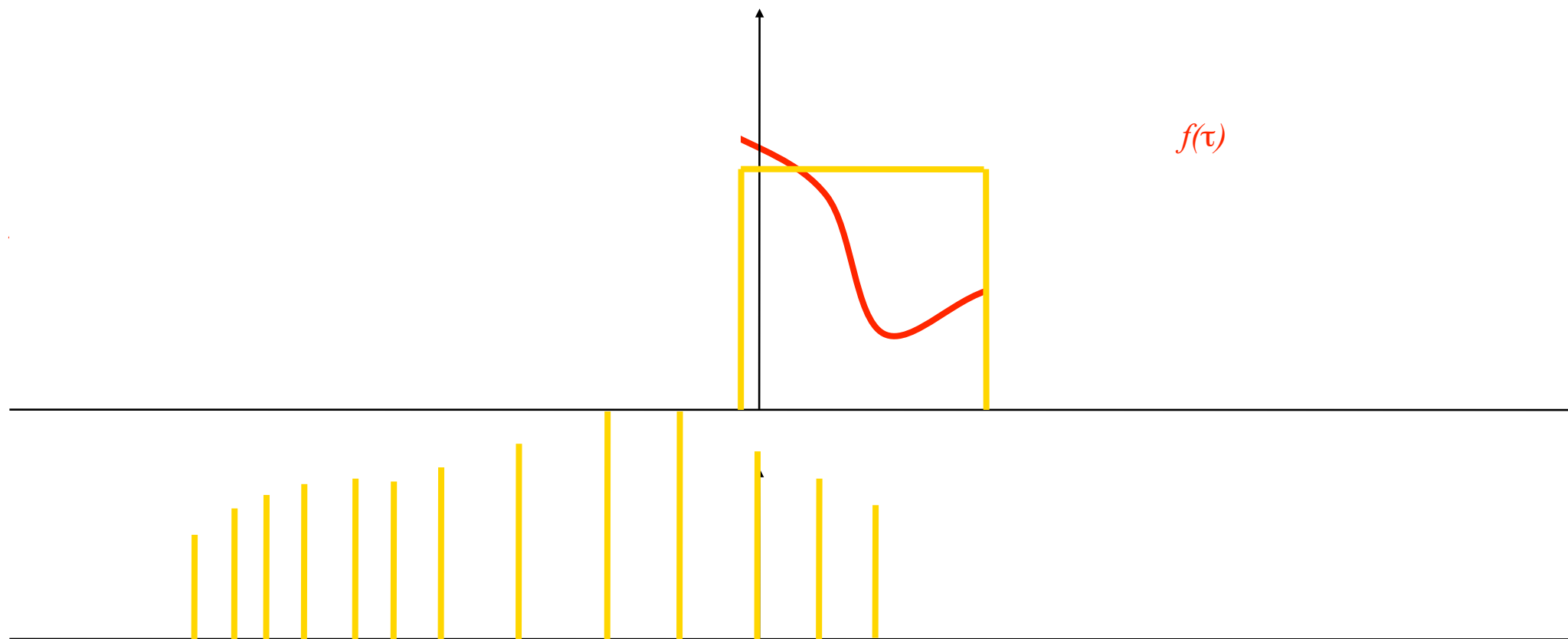
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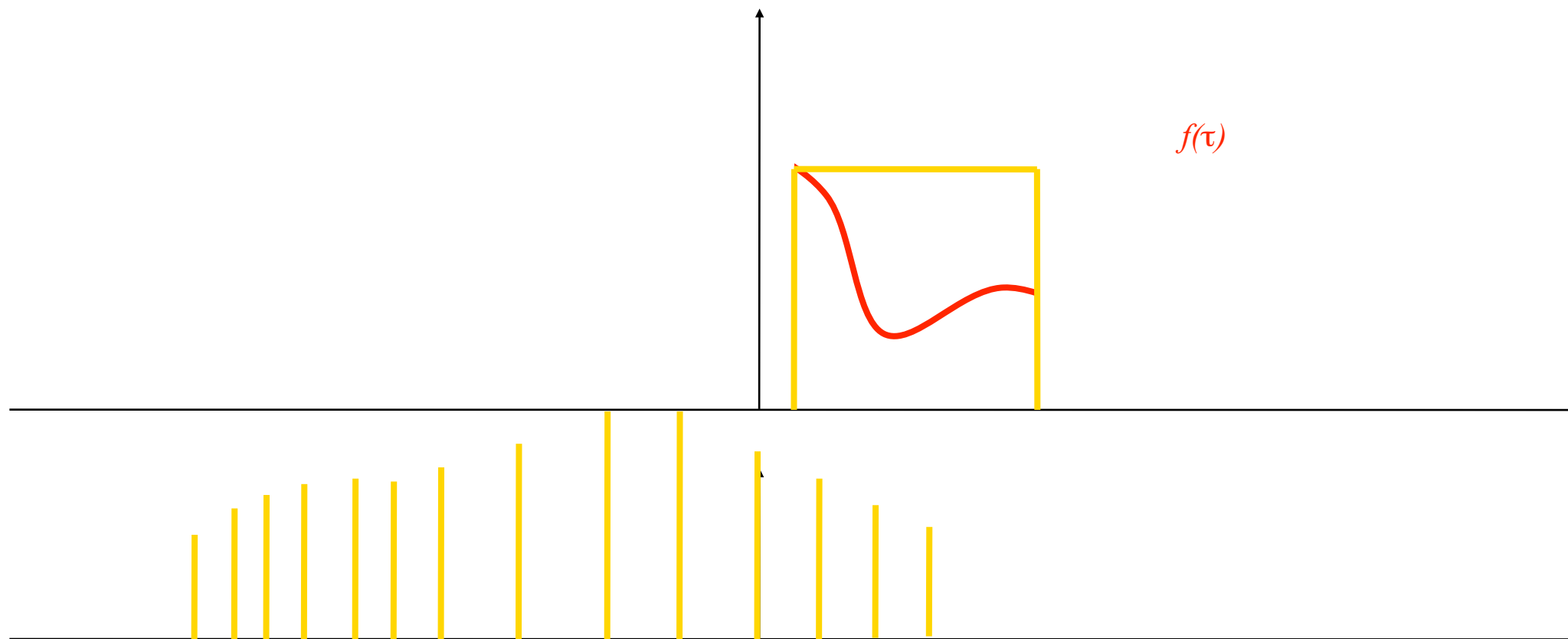
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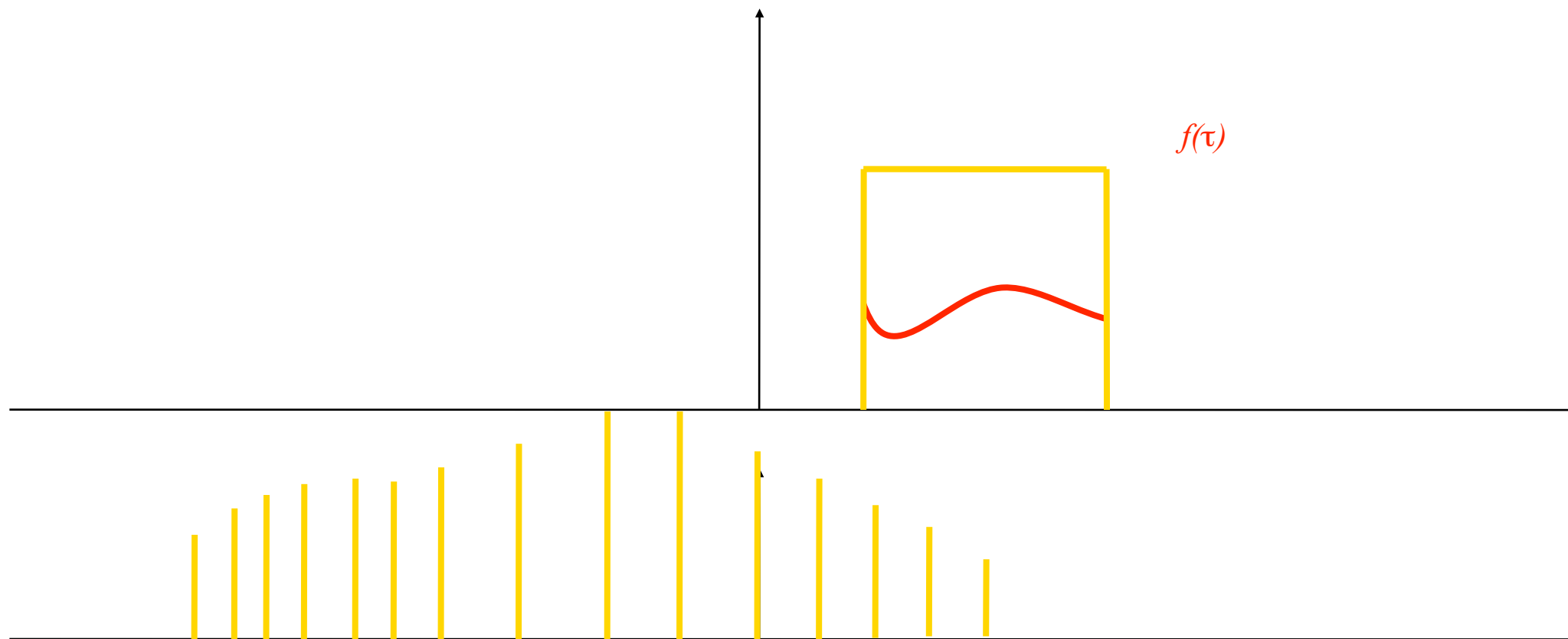
Convolution

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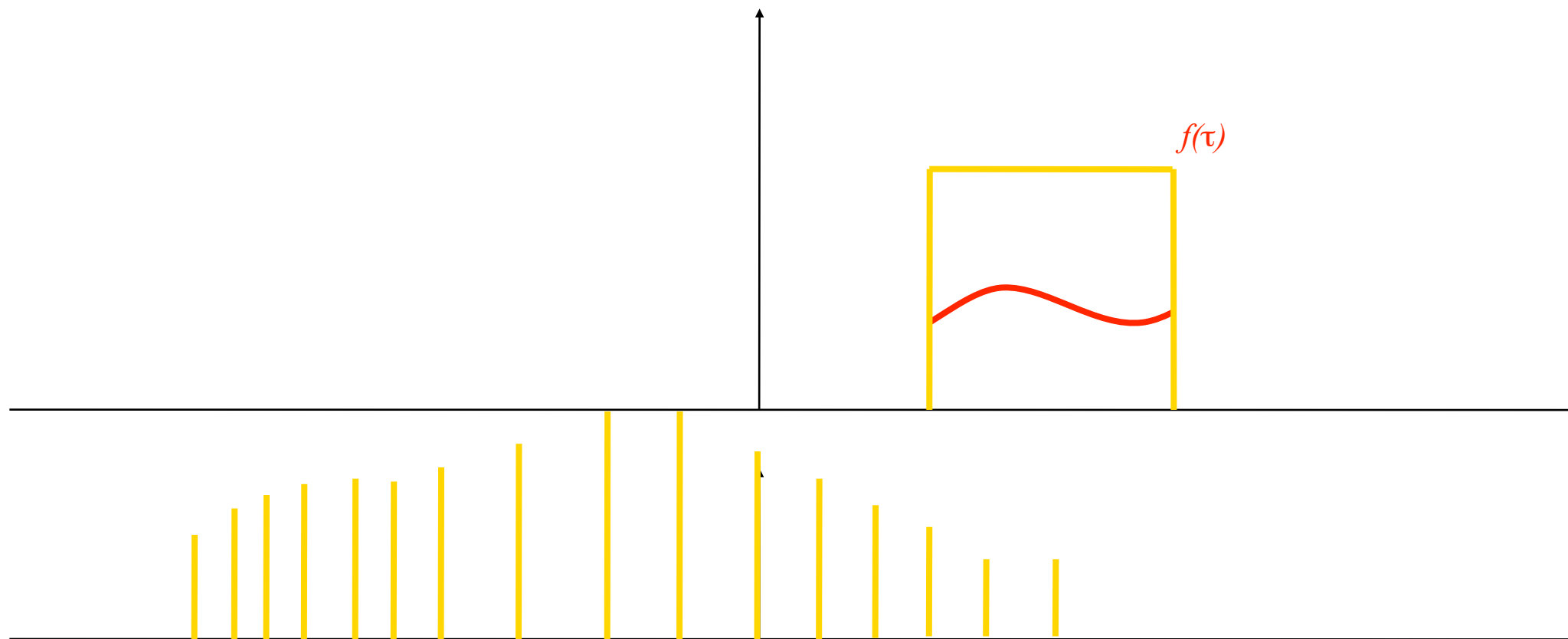
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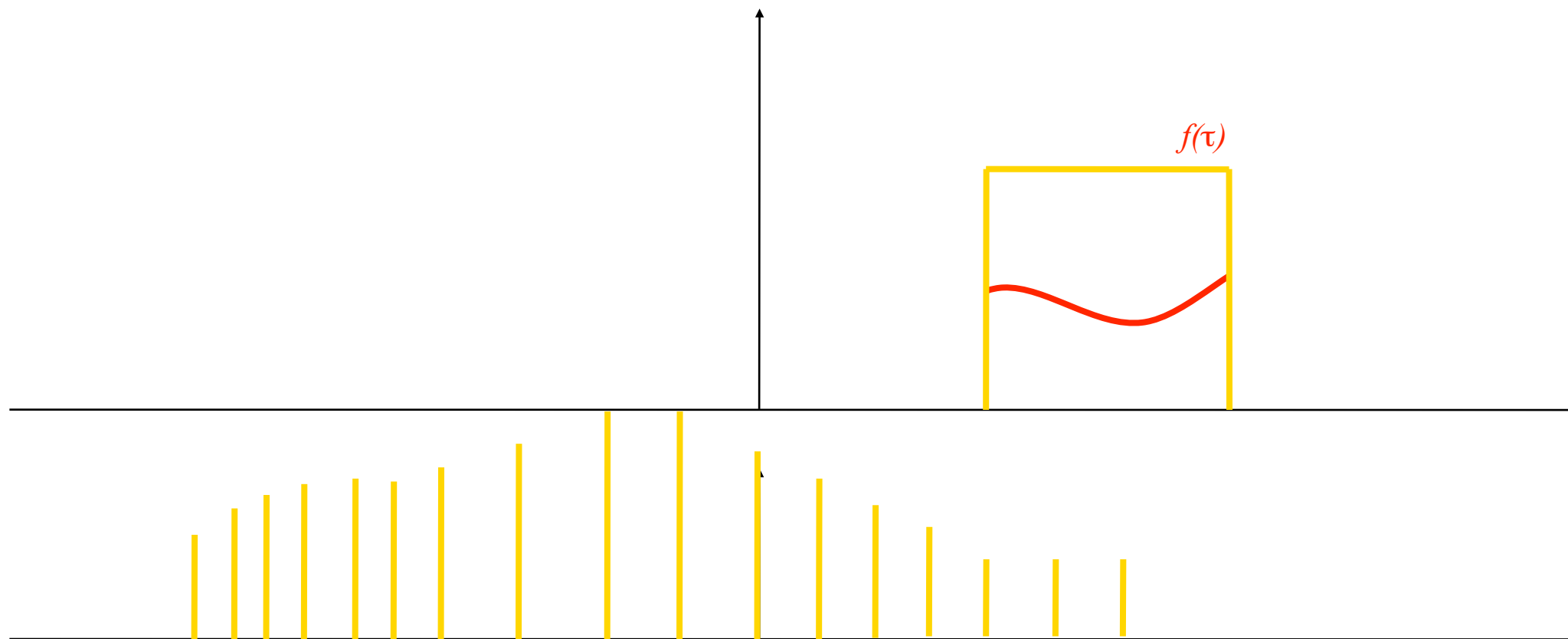
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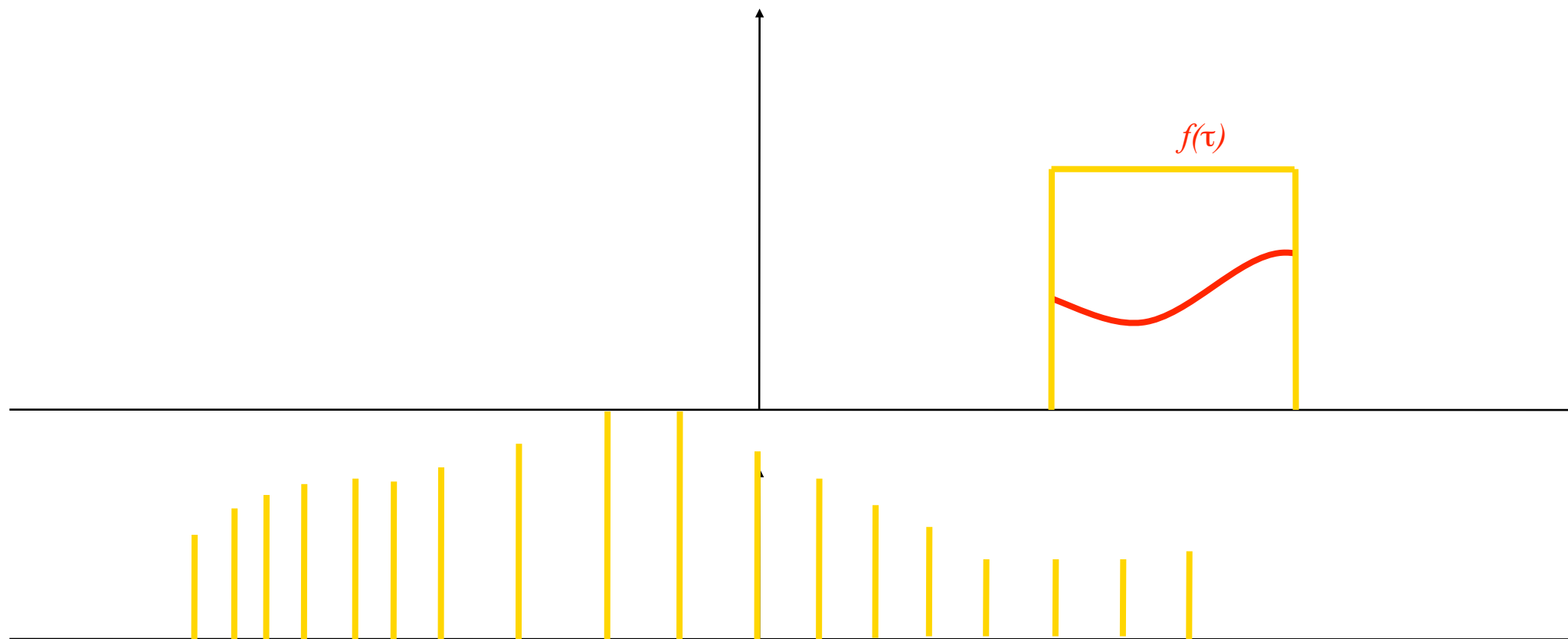
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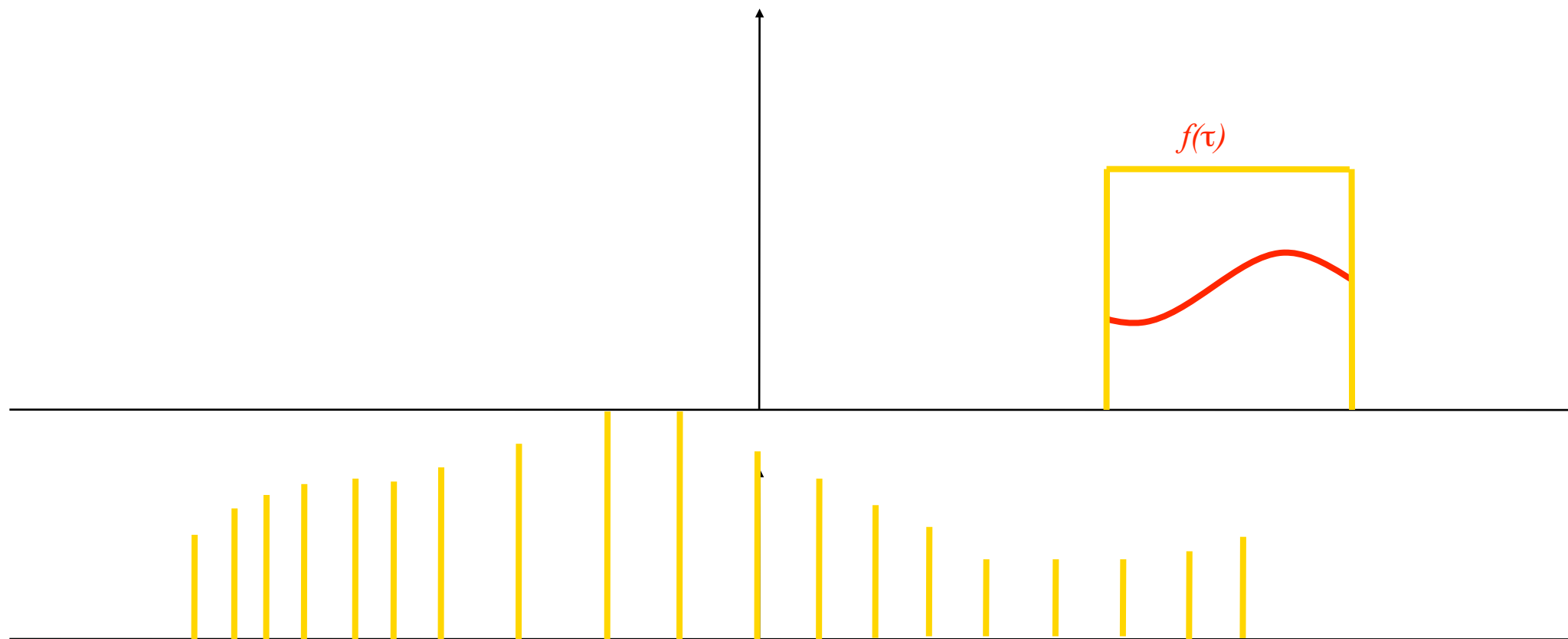
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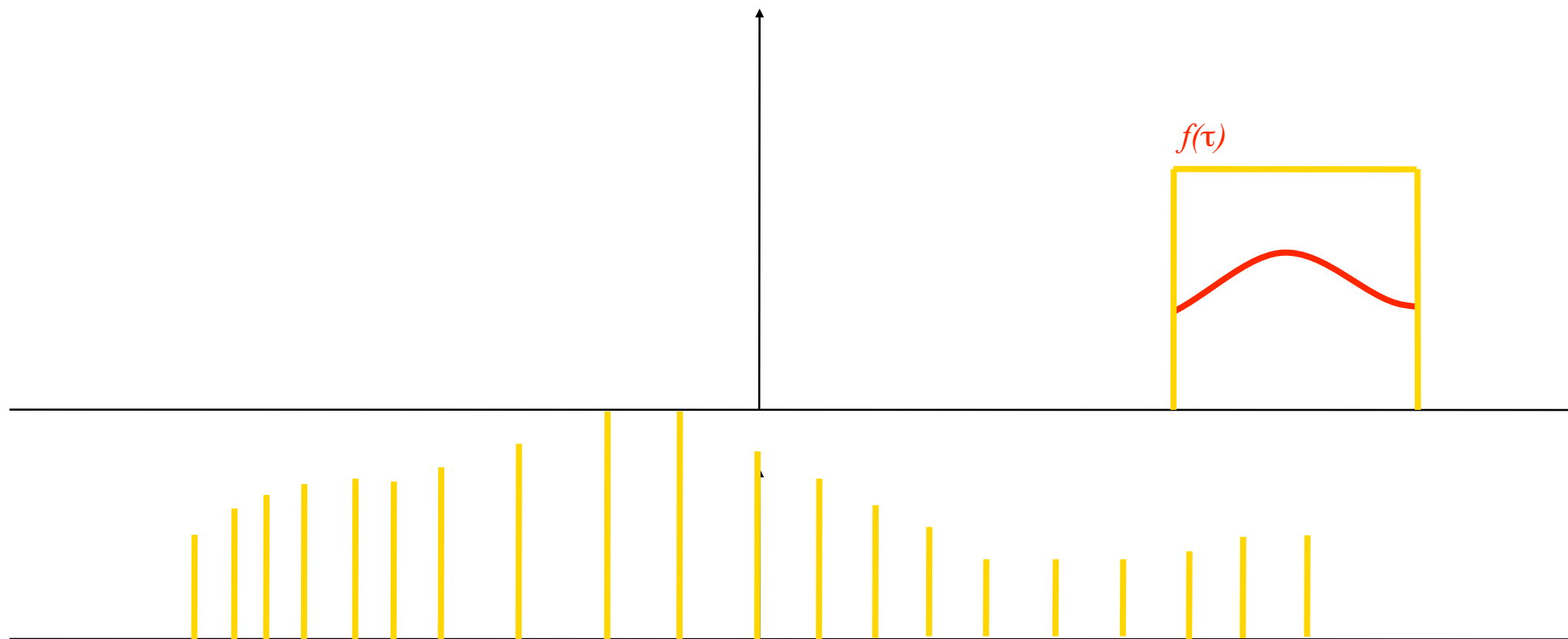
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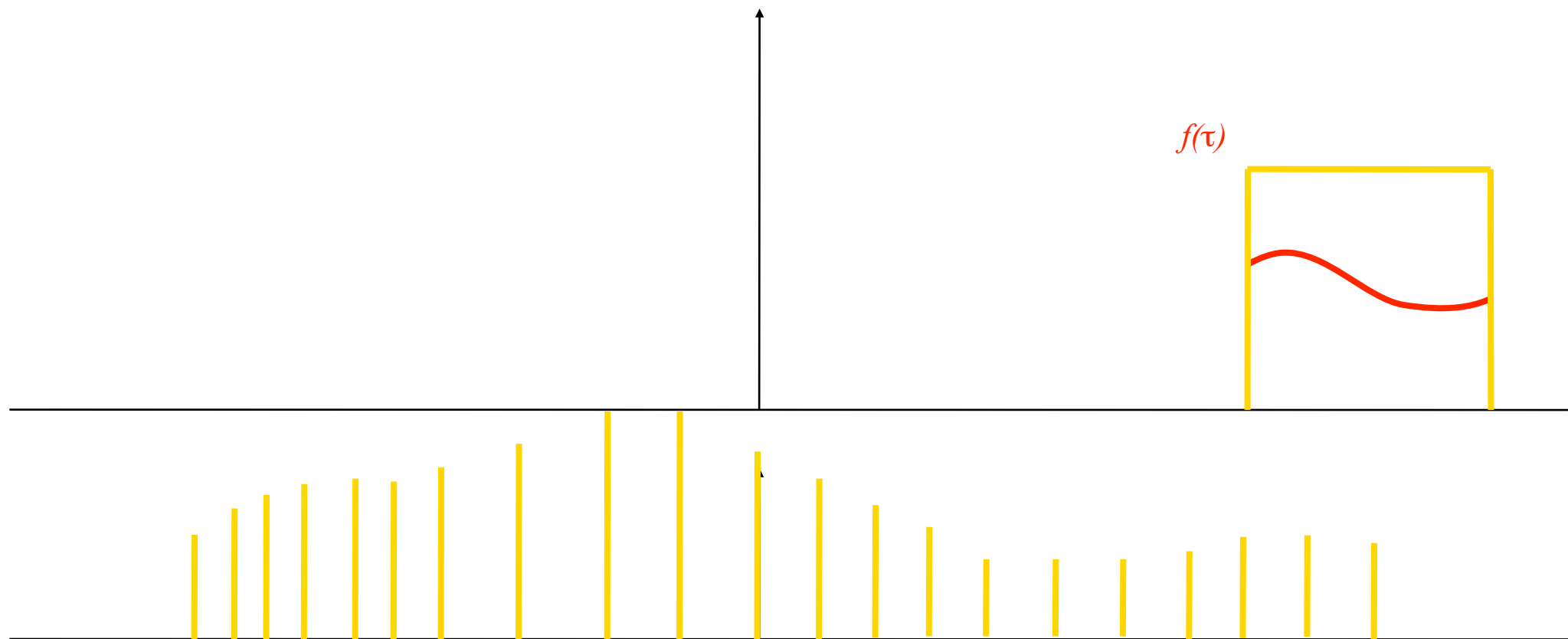
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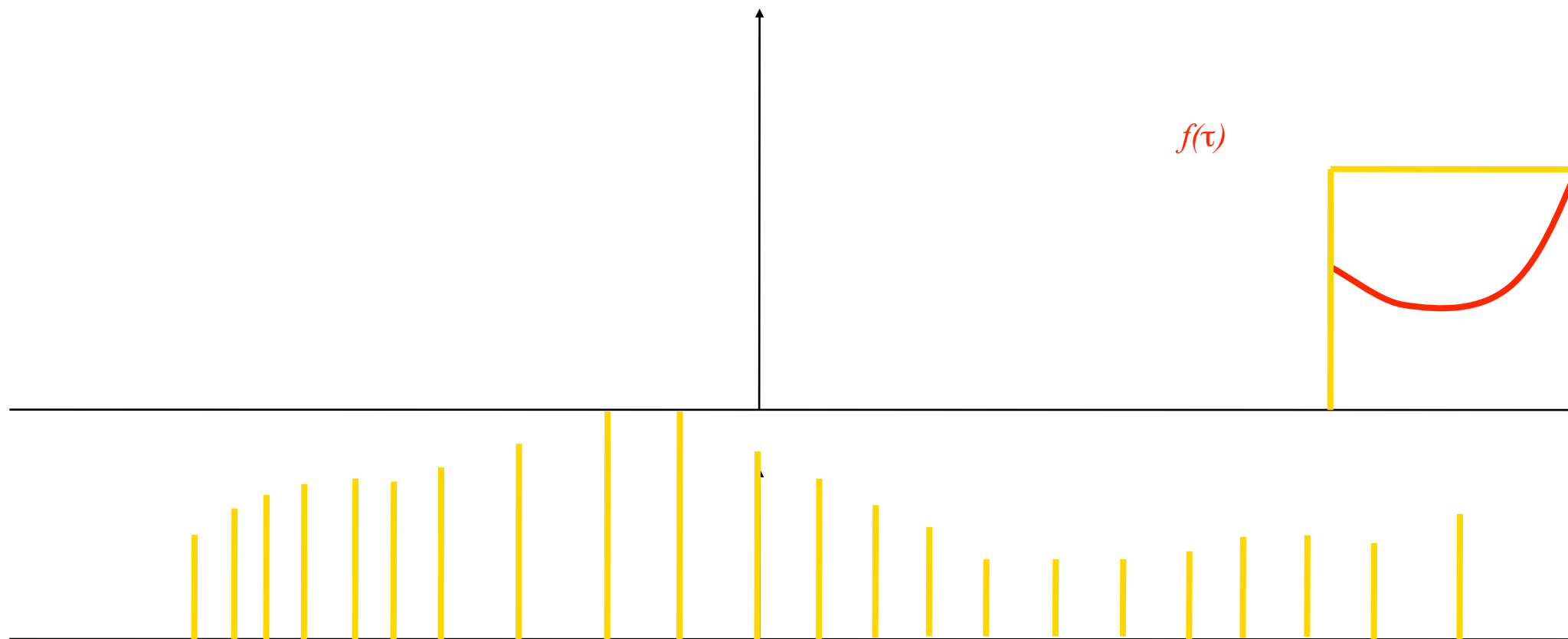
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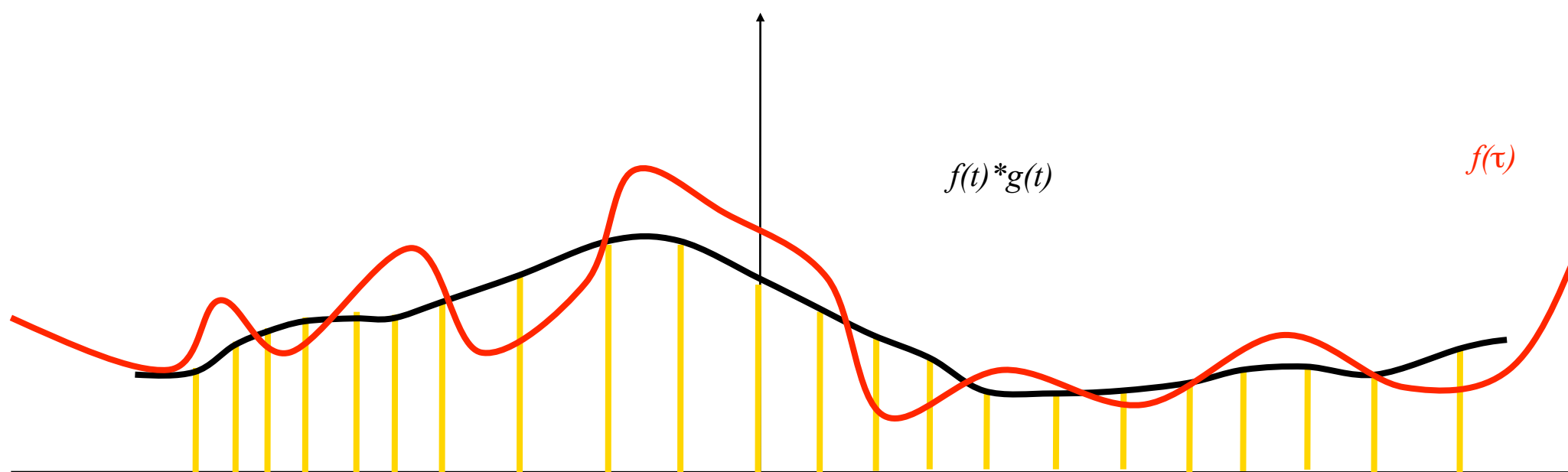
Convolution

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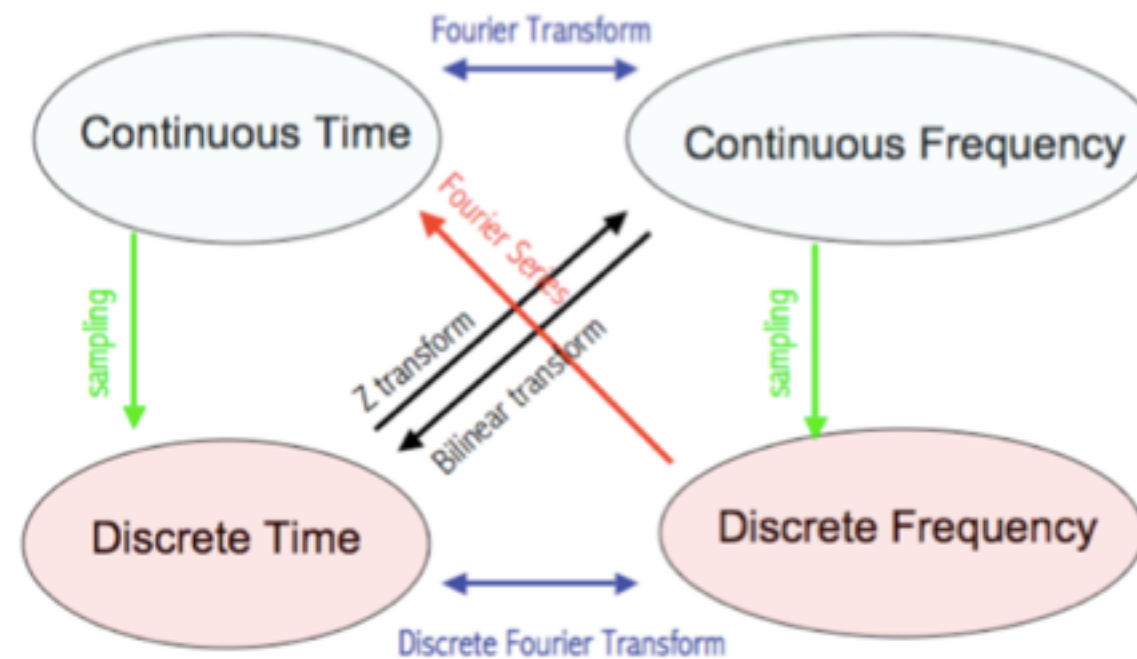


Convolution

- ☑ This particular convolution smooths out some of the high frequencies in $f(t)$.



Various spaces and transforms



Signal type	Continuous time	Discrete time	Transform Domain
Finite duration	Laplace	z	Continuous complex frequency (s-plane)
Finite duration	Fourier	Discrete-time Fourier (DTFT)	Continuous real frequency
Periodic	Fourier Series	Discrete Fourier Series (DFS)	Discrete real frequency

LS & FT

Figure 6.3-1: Definition of a linear system.

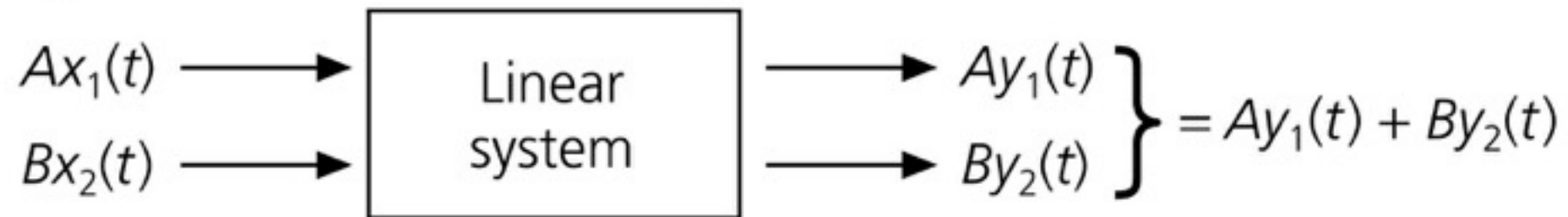
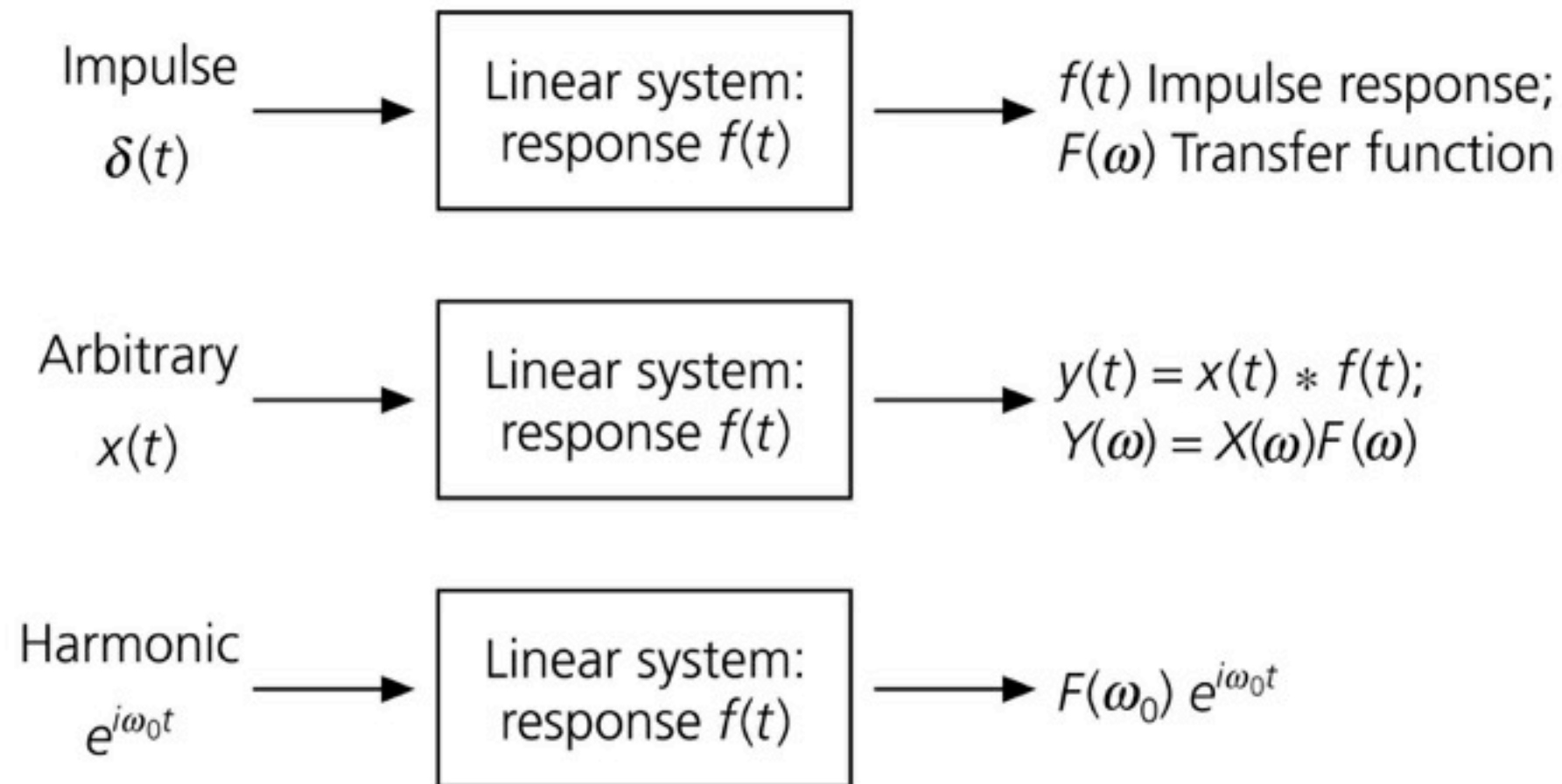


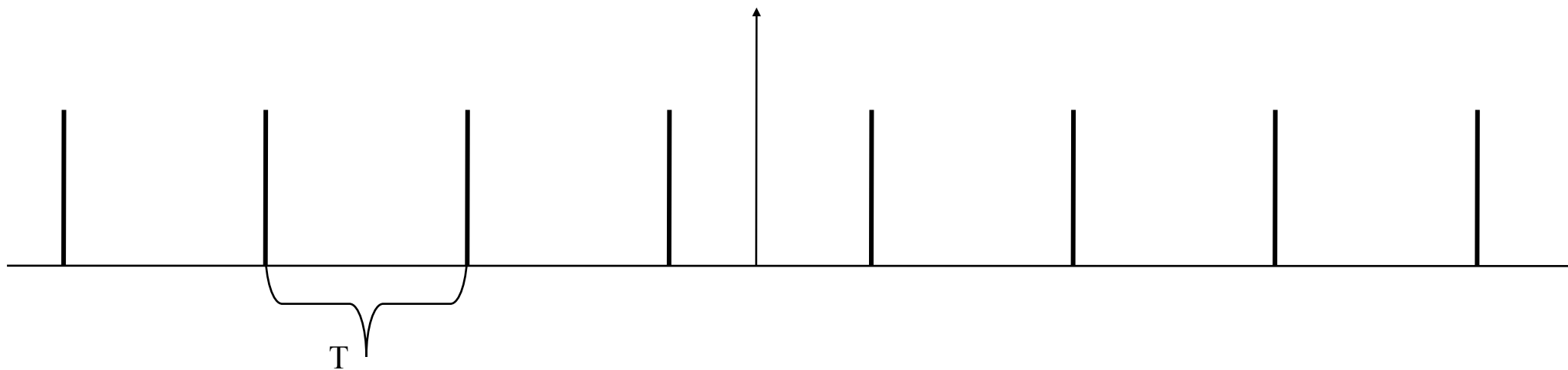
Figure 3.3-29: Seismic section before and after deconvolution.



Sampling Function

- ✓ A Sampling Function or Impulse Train is defined by:

$$S_T(t) = \sum_{k=-\infty}^{\infty} \delta(t - kT)$$



Sampling Function

- ✓ The Fourier Transform of the Sampling Function is itself a sampling function.
- ✓ The sample spacing is the inverse.

$$S_T(t) \Leftrightarrow S_{\frac{1}{T}}(\omega)$$

Convolution Theorem

- ✓ The convolution theorem states that convolution in the spatial domain is equivalent to multiplication in the frequency domain, and viceversa.

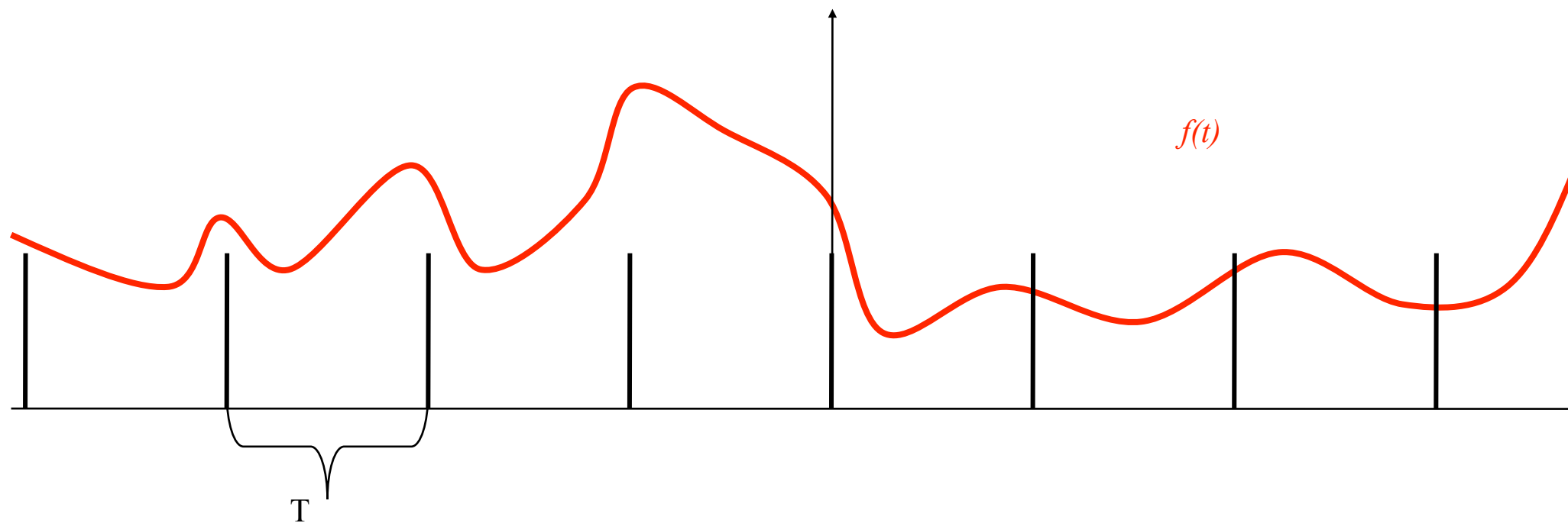
$$f(t) * g(t) \Leftrightarrow F(\omega)G(\omega)$$

$$f(t)g(t) \Leftrightarrow F(\omega) * G(\omega)$$

Convolution Theorem

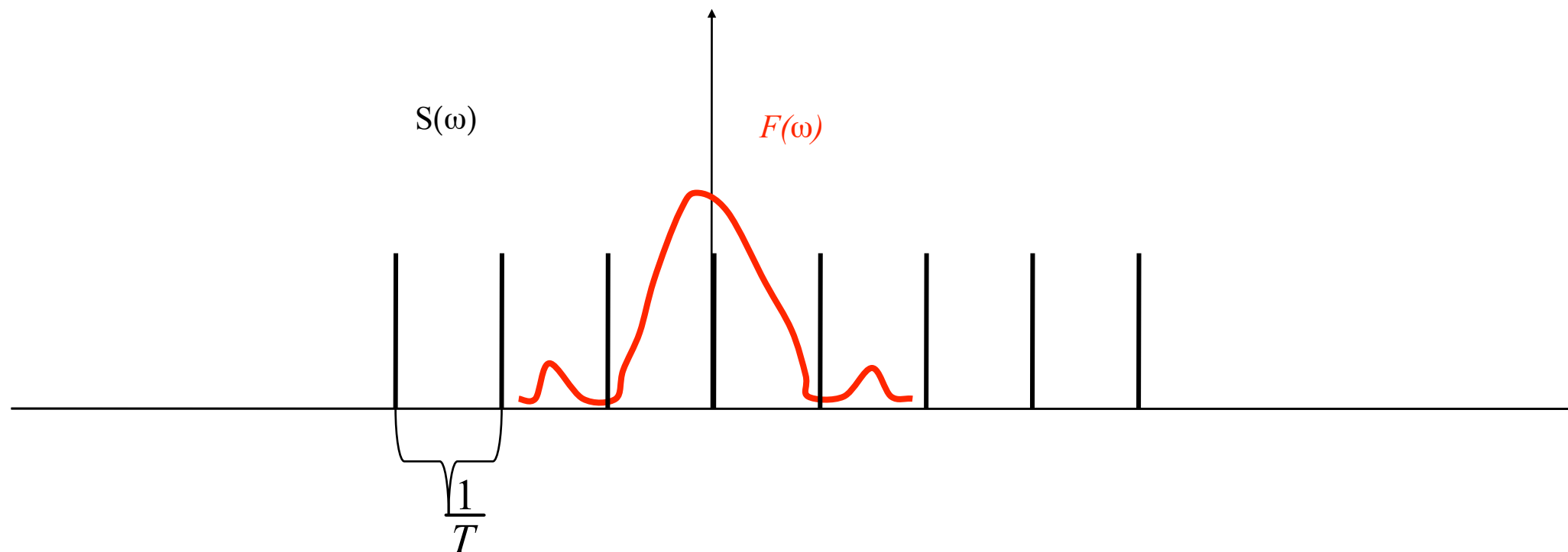
✓ This powerful theorem can illustrate the problems with our point sampling and provide guidance on avoiding aliasing.

✓ Consider: $f(t) S_T(t)$



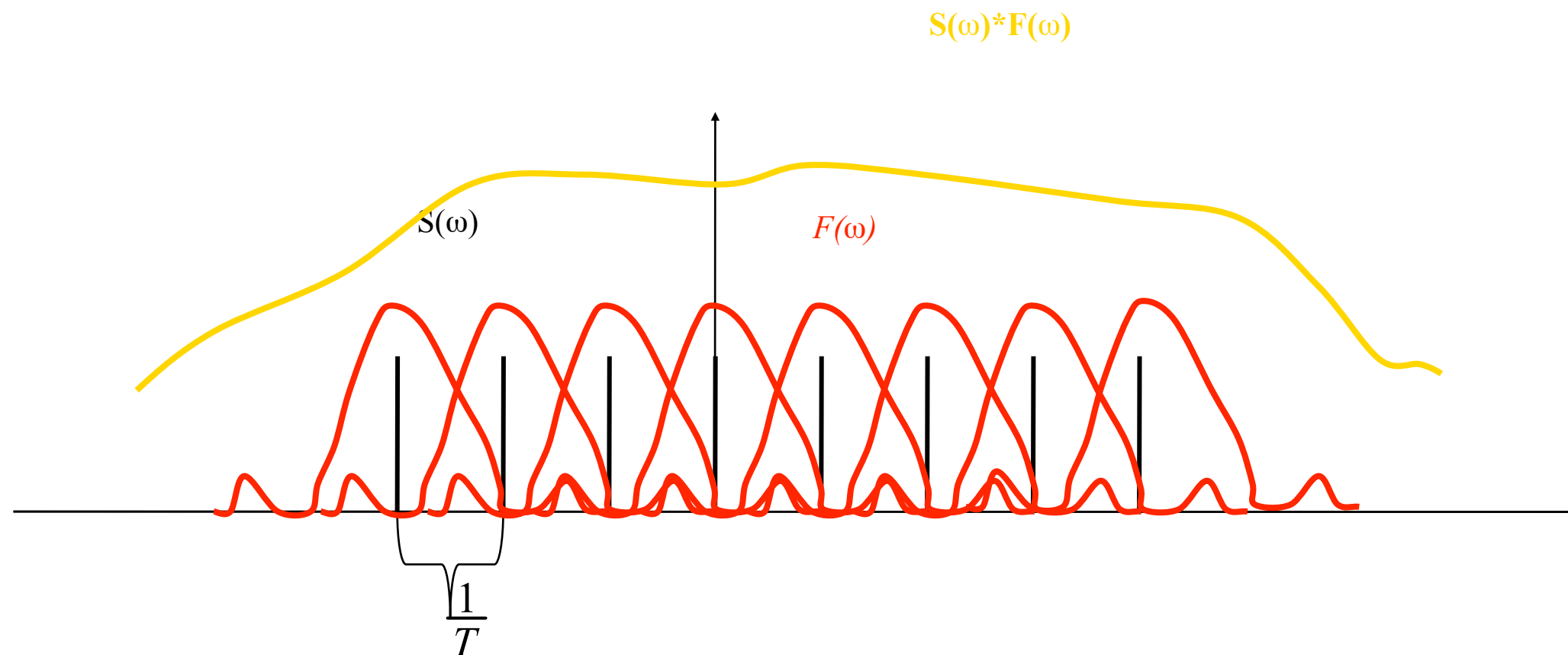
Convolution Theorem

- ✓ What does this look like in the Fourier domain?



Convolution Theorem

☑ In Fourier domain we would convolve



Aliasing

- ☑ What this says, is that any frequencies greater than a certain amount will appear intermixed with other frequencies.
- ☑ In particular, the higher frequencies for the copy at $1/T$ intermix with the low frequencies centered at the origin.

Aliasing and Sampling

- ☑ Note, that the sampling process introduces frequencies out to infinity.
- ☑ We have also lost the function $f(t)$, and now have only the discrete samples.
- ☑ This brings us to our next powerful theory.

Sampling Theorem

- ✓ The Shannon Sampling Theorem
- ✓ A band-limited signal $f(t)$, with a cutoff frequency of λ , that is sampled with a sampling spacing of T may be perfectly reconstructed from the discrete values $f[nT]$ by convolution with the $\text{sinc}(t)$ function, provided:

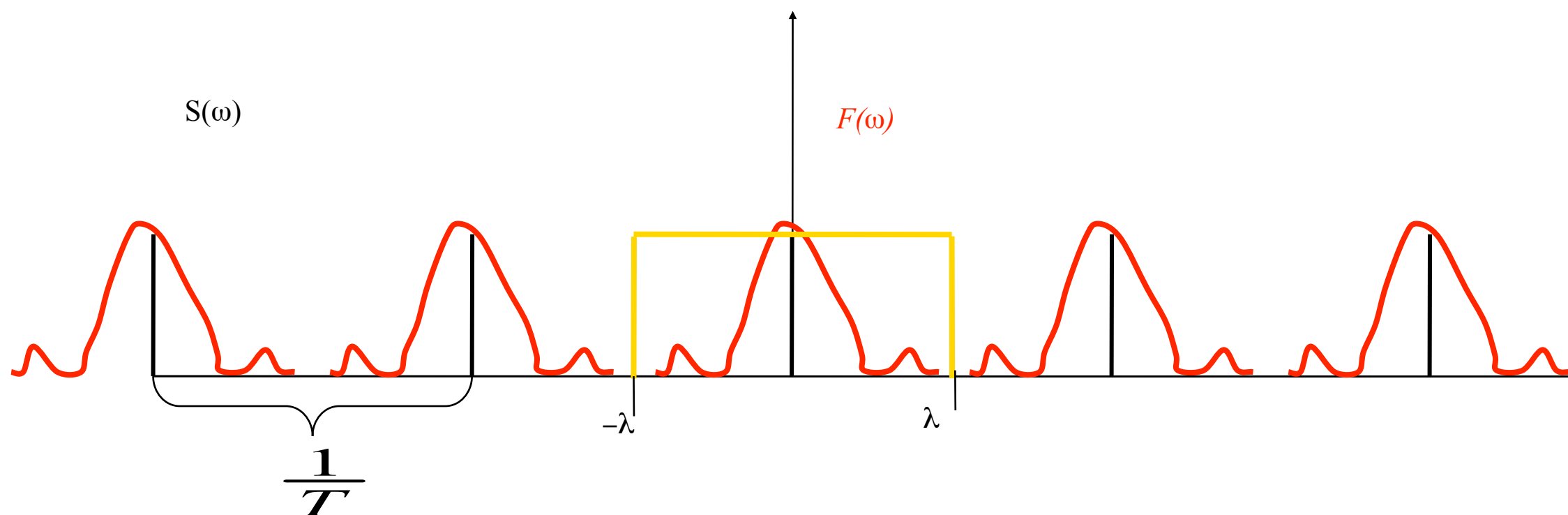
$$\lambda < \frac{1}{2T}$$

Sampling Theory

- ☑ Why is this?
- ☑ The Nyquist limit will ensure that the copies of $F(\omega)$ do not overlap in the frequency domain.
- ☑ I can completely reconstruct or determine $f(t)$ from $F(\omega)$ using the Inverse Fourier Transform.

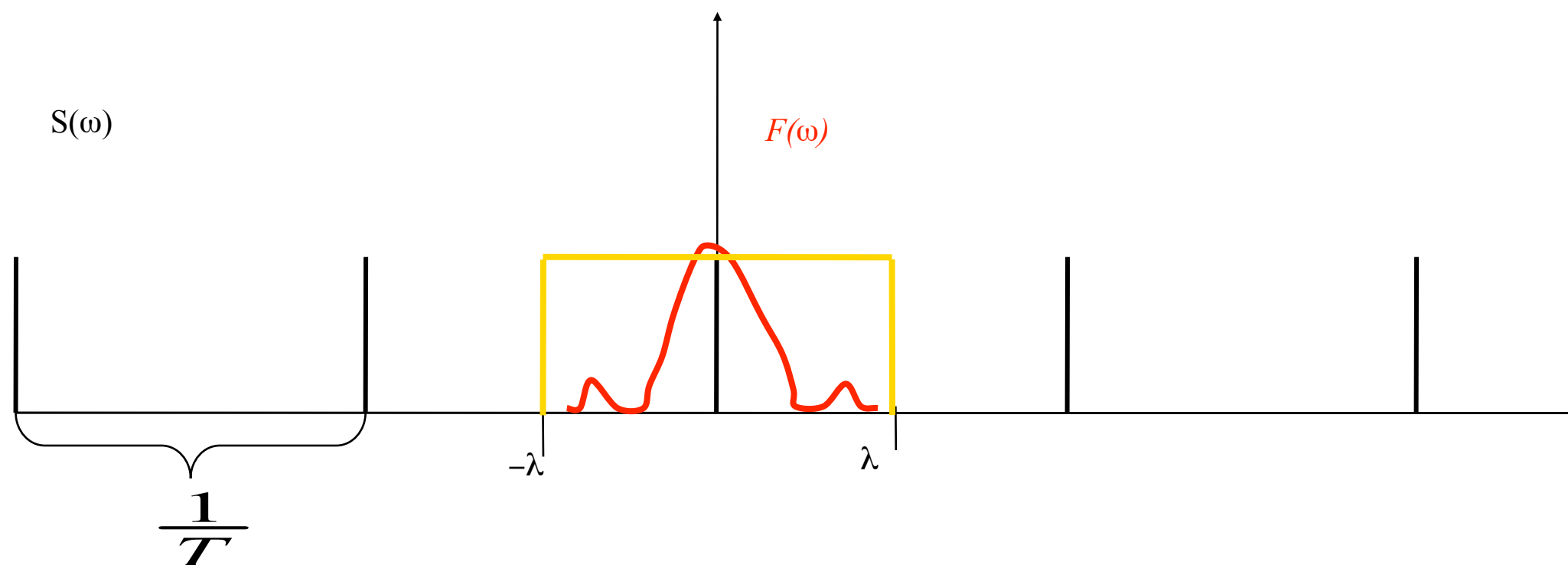
Sampling Theory

- ✓ In order to do this, I need to remove all of the shifted copies of $F(\omega)$ first.
- ✓ This is done by simply multiplying $F(\omega)$ by a box function of width 2λ .



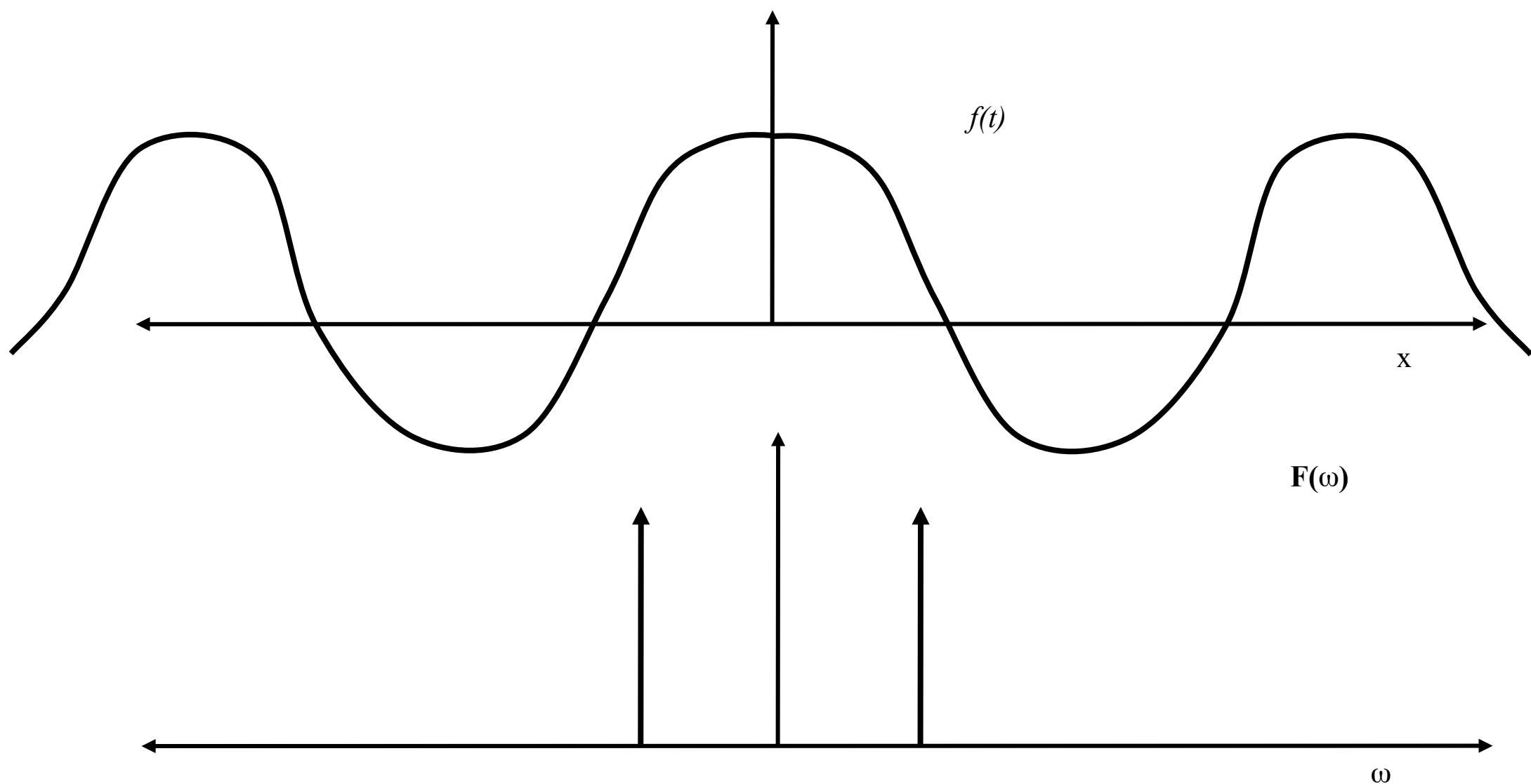
Sampling Theory

- ✓ In order to do this, I need to remove all of the shifted copies of $F(\omega)$ first.
- ✓ This is done by simply multiplying $F(\omega)$ by a box function of width 2λ .



A Cosine Example

Consider the function $f(t) = \cos(2\pi t)$.



Sampling Theory

✓ So, given $f[nT]$ and an assumption that $f(t)$ does not have frequencies greater than $1/2T$, we can write the formula:

$$f[nT] = f(t) S_T(t) \Leftrightarrow F(\omega)^* S_T(\omega)$$

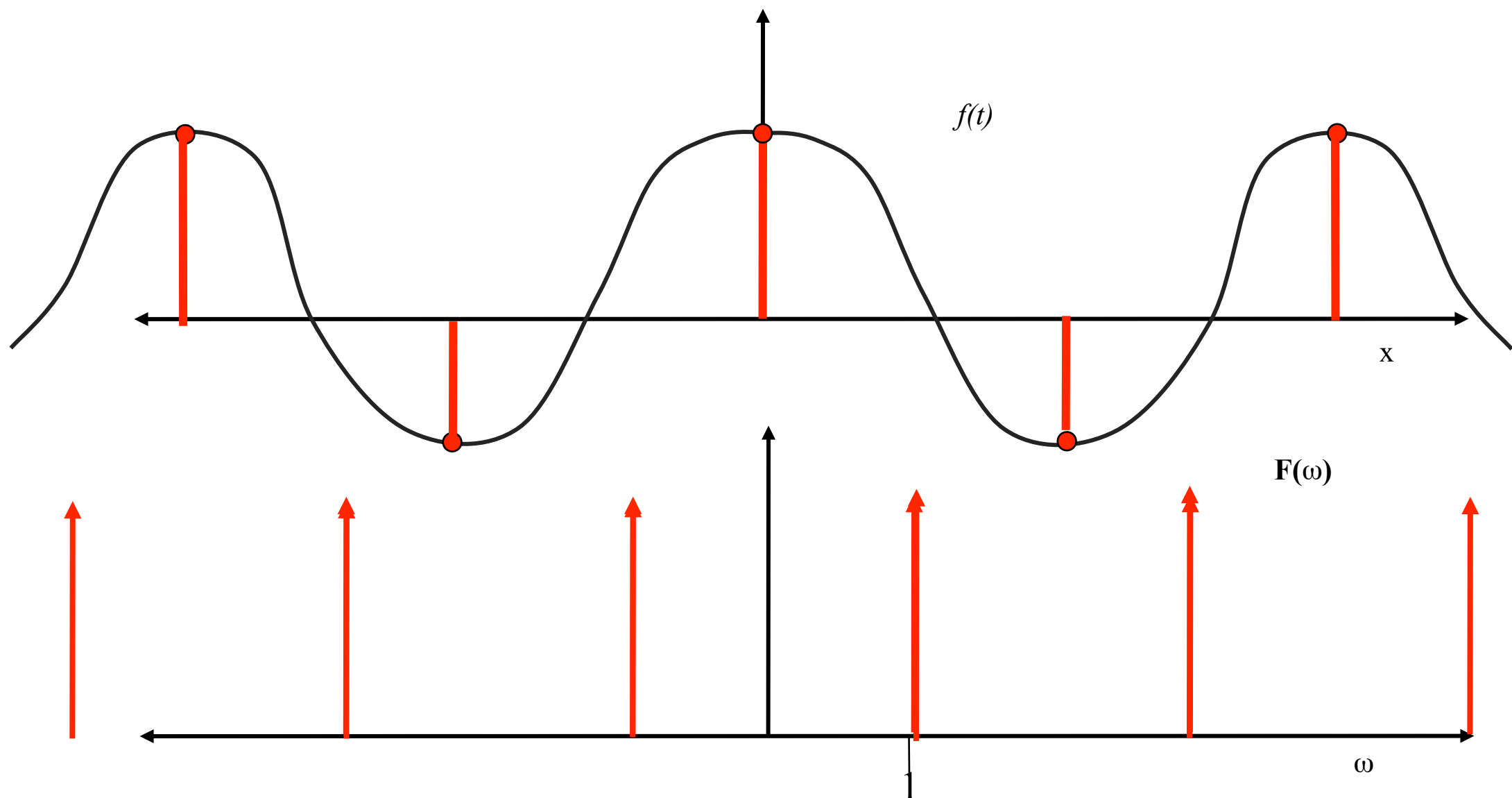
$$F(\omega) = (F(\omega)^* S_T(\omega)) \text{BOX}_{1/2T}(\omega)$$

therefore,

$$f(t) = f[nT] * \text{sinc}(t)$$

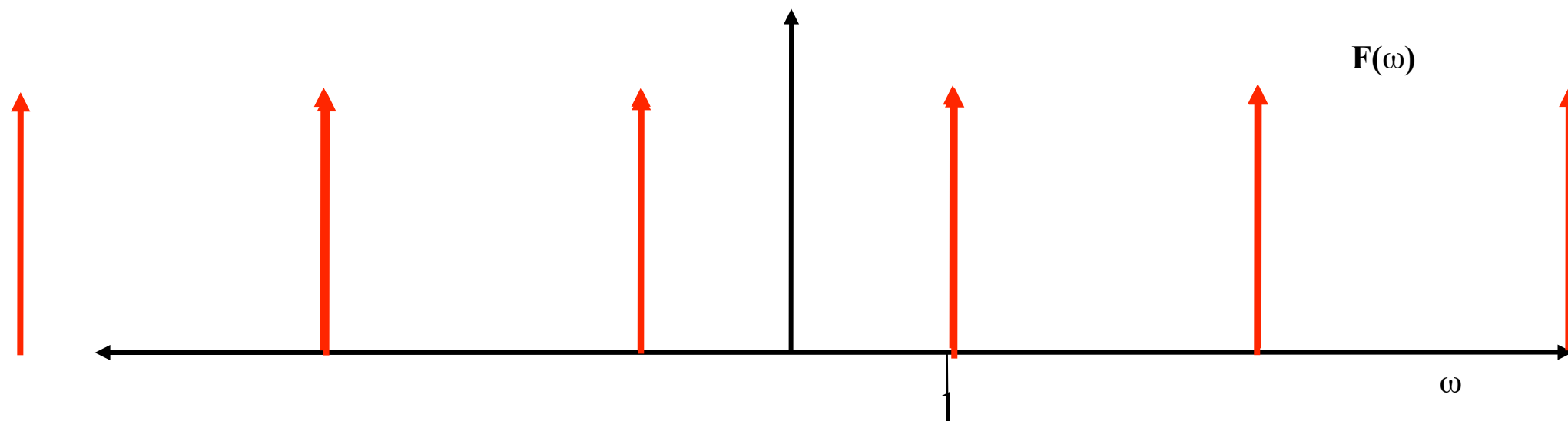
A Cosine Example

Now sample it at $T=1/2$



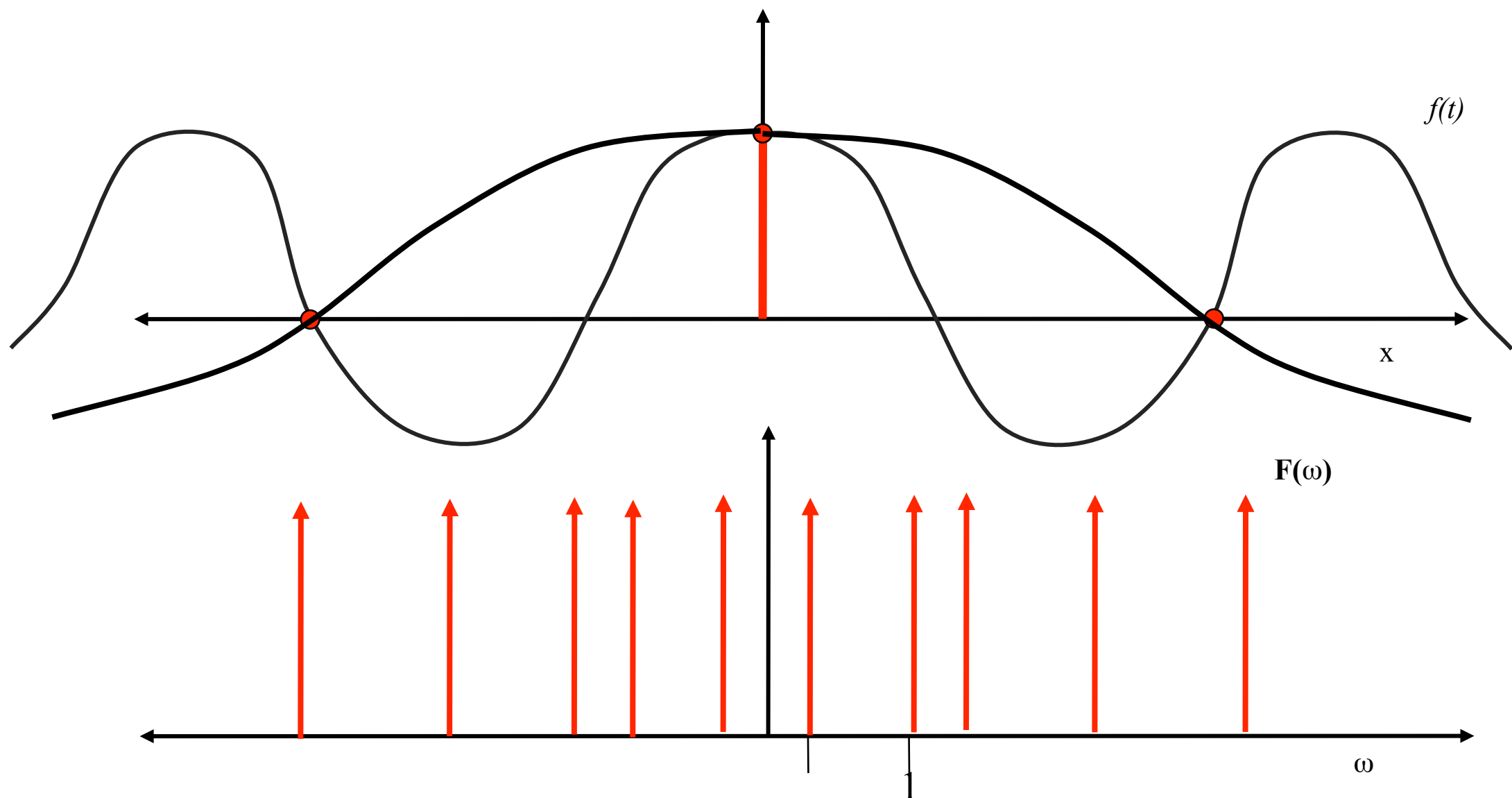
A Cosine Example

- ✓ Problem:
- ✓ The amplitude is now wrong or undefined.
- ✓ Note however, that there is one and only one cosine with a frequency less than or equal to 1 that goes through the sample pts.



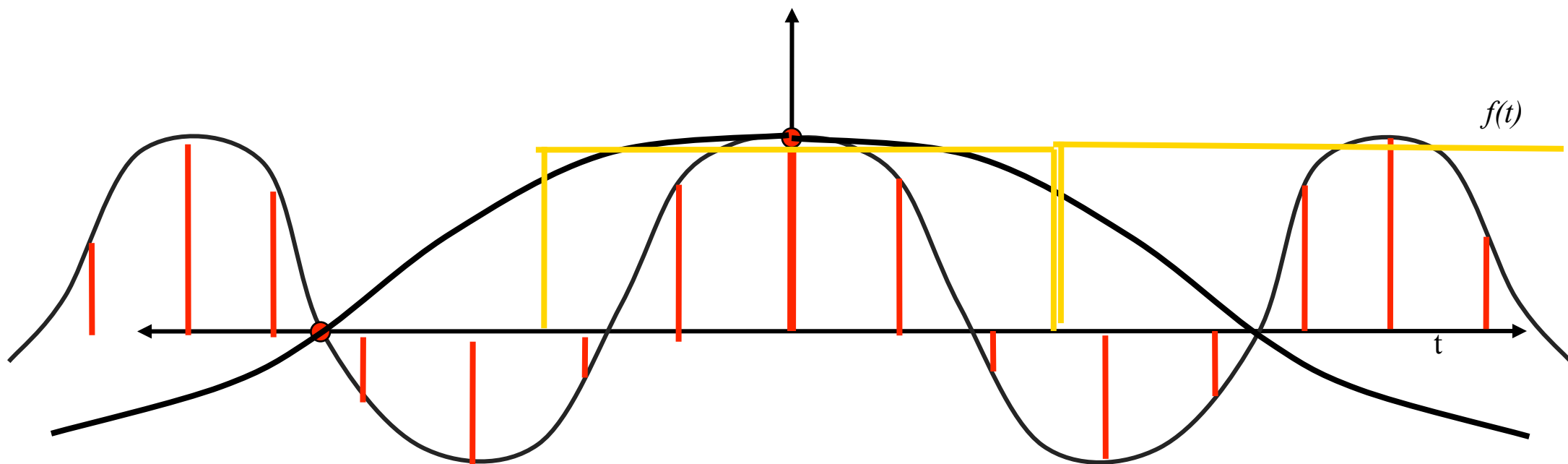
A Cosine Example

What if we sample at $T=2/3$?



Supersampling

- ✓ Supersampling increases the sampling rate, and then integrates or convolves with a box filter, which is finally followed by the output sampling function.



Sampling and Anti-aliasing

- ✓ The problem:
- ✓ The signal is not band-limited.
- ✓ Uniform sampling can pick-up higher frequency patterns and represent them as low-frequency patterns.

