

# Fourier Essentials

WTBV

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# key aspects of the Fourier transform

- Fundamental idea: functions/signals have a life both in time/space and in frequency domain — and both aspects are equivalent
- Motivation: Fourier transform can be obtained from Fourier series by a limiting process
- Basic properties of FT
  - Translation and Dilation basic wavelet operations
  - Derivation smoothness properties of wavelets
  - Convolution filtering properties of wavelets
- Advanced properties of FT
  - Time/frequency localization, duality and uncertainty
  - Poisson's formula and sampling
- Fourier transform theory is
  - important immense number of applications
  - not easy making ideas rigorous requires lot of work
  - beautiful leads into a new universe

- Some highlights
  - Trying to make Fourier's ideas precise spawned lots of new mathematics (convergence concepts, set theory, distributions,...)
  - The Cooley-Tukey 1965 paper on FFT is the most frequently cited article in all of mathematics
  - First US patent for a mathematical algorithm for a variant of FFT
  - About 3/4 of all Nobel prizes in physics were awarded for work done with Fourier analysis
  - Other notable Nobel prizes:
    - Crick/Watson/Wilkins (1962): DNA structure by diffraction
    - Cormack/Hounsfield (1979): computed tomography
    - Hauptman/Karle (1985): structure of molecules by X-ray diffraction
    - Lauterbur/Mansfield (2003): MRI
  - Other fields: PDE, Quantum Mechanics, Signals and Systems, Fourier Optics, ...

- A function  $f : \mathbb{R} \rightarrow \mathbb{C}$  is  $p$ -periodic, if  $f(t + p) = f(t)$  for all  $t \in \mathbb{R}$ .
- Equivalently:  
 $f(t)$  is defined in any real interval  $[a, b)$  of length  $b - a = p$  and is extended periodically
- Simple 1-periodic functions are the harmonics  
 $\sin(2\pi kt)$  ( $k \geq 1$ ),  $\cos(2\pi kt)$  ( $k \geq 0$ ),  $\omega_k(t) = e^{2\pi ikt}$  ( $k \in \mathbb{Z}$ )
- Superposition principle: linear combinations of 1-periodic functions are again 1-periodic functions
- Fourier's Idea (1807): "Any" 1-periodic function  $f(t)$  can be represented as a superposition (*Fourier series*) of harmonics, i.e., there are sequences  $(a_k)_{k \geq 0}, (b_k)_{k \geq 1}, (c_k)_{k \in \mathbb{Z}} \in \mathbb{C}$  s.th.

$$f(t) \stackrel{\text{"="}}{=} \frac{a_0}{2} + \sum_{k > 0} a_k \cos(2\pi kt) + b_k \sin(2\pi kt)$$

$$\stackrel{\text{"="}}{=} \sum_{k \in \mathbb{Z}} c_k e^{2\pi ikt}$$

- $\mathcal{L}^2([0, 1))$  : Hilbert space of square-integrable (in the sense of Lebesgue) functions  $f : [0, 1) \rightarrow \mathbb{C}$  with inner product

$$\langle f | g \rangle = \int_0^1 f(t) \overline{g(t)} dt < \infty$$

and norm

$$\|f\|^2 = \langle f | f \rangle = \int_0^1 |f(t)|^2 dt < \infty$$

- The families

$$\left\{ \omega_k(t) = e^{2\pi ikt} \right\}_{k \in \mathbb{Z}}$$

and

$$\{\sin(2\pi kt)\}_{k \geq 1} \cup \{\cos(2\pi kt)\}_{k \geq 0}$$

are orthonormal families (even Hilbert bases) of  $\mathcal{L}^2([0, 1))$ :

$$\langle \omega_k | \omega_\ell \rangle = \int_0^1 e^{2\pi i(k-\ell)t} dt = \delta_{k,\ell}$$

- Similarly for the family of harmonics sin-cos

- Fourier coefficients (*Analysis*)

$$c_k = \widehat{f}[k] = \langle f | \omega_k \rangle = \int_0^1 f(t) e^{-2\pi ikt} dt \quad (k \in \mathbb{Z})$$

- Fourier series (*Synthesis*)

$$f(t) = \sum_{k \in \mathbb{Z}} \langle f | \omega_k \rangle \omega_k(t) = \sum_{k \in \mathbb{Z}} \widehat{f}[k] e^{2\pi ikt}$$

- For  $f \in \mathcal{L}^2([0, 1))$  one has

$$S_N(t) = \sum_{k=-N}^N \widehat{f}[k] e^{2\pi ikt} \xrightarrow{N \rightarrow \infty} f(t)$$

in the sense of  $\mathcal{L}^2$ -convergence (optimal  $\mathcal{L}^2$ -approximation)

- Stronger assertions about convergence are possible, but more difficult to obtain

- Important aspects
  - The Fourier coefficients of  $f(t)$  depend on the behavior of  $f(t)$  over the whole interval  $[0, 1)$
  - The basis functions  $\omega_k(t) = e^{2\pi ikt}$  are
    - perfectly localized w.r.t. frequency
    - not at all localized w.r.t. time/space
  - The family  $\{\omega_k(t) = e^{2\pi ikt}\}_{k \in \mathbb{Z}}$  is a complete basis (Hilbert basis) in  $\mathcal{L}^2([0, 1))$

- Hilbert space of sequences (“discrete signals with finite energy“)

- complex bi-infinite sequences

$$\mathbf{x} = (\dots, x[-1], x[0], x[1], x[2], \dots) = (x[k])_{k \in \mathbb{Z}} \text{ with } x[k] \in \mathbb{C} \text{ (} k \in \mathbb{Z} \text{)}$$

- the relevant vector space is  $\ell^2$

$$\ell^2 = \left\{ \mathbf{x} = (x[k])_{k \in \mathbb{Z}}; \sum_{k \in \mathbb{Z}} |x[k]|^2 < \infty \right\}$$

with inner product  $\langle \mathbf{x} | \mathbf{y} \rangle = \sum_{k \in \mathbb{Z}} x[k] \cdot \overline{y[k]}$   
 and norm  $\|\mathbf{x}\|^2 = \sum_{k \in \mathbb{Z}} |x[k]|^2$

- *Parseval-Plancherel* property: The mapping

$$f(t) \mapsto \left( \widehat{f}[k] \right)_{k \in \mathbb{Z}}$$

- is a linear mapping  $\mathcal{L}^2([0, 1]) \mapsto \ell^2$
- is an *isometry*, which means

$$\langle f | g \rangle_{\mathcal{L}^2} = \int_0^1 f(t) \overline{g(t)} dt = \sum_{k \in \mathbb{Z}} \widehat{f}[k] \overline{\widehat{g}[k]} = \langle \widehat{f} | \widehat{g} \rangle_{\ell^2}$$

- is surjective (the *Riesz-Fischer* theorem)
- Conclusion:  $\mathcal{L}^2([0, 1])$  and  $\ell^2$  are isomorphic as Hilbert spaces

- Everything carries over routinely from  $[0, 1)$  to arbitrary finite intervals  $[a, b)$  and  $p$ -periodic functions with  $p = b - a$
- $\mathcal{L}^2([a, b))$  has a basis of functions

$$\left\{ \omega_k(t/p) = e^{2\pi ikt/p} \right\}_{k \in \mathbb{Z}}$$

and similarly for sin-cos

- the inner product in  $\mathcal{L}^2([a, b))$  is

$$\langle f | g \rangle = \frac{1}{p} \int_a^b f(t) \overline{g(t)} dt$$

- Fourier coefficients (*Analysis*)

$$\widehat{f}[k] = \langle f | \omega_k(t/p) \rangle = \frac{1}{p} \int_a^b f(t) e^{-2\pi ikt/p} dt$$

- Fourier series (*Synthesis*)

$$f(t) = \sum_{k \in \mathbb{Z}} \widehat{f}[k] e^{2\pi ikt/p}$$

- The *Gibbs-Wilbraham* phenomenon

- describes the convergence of the approximations  $s_N(t)$  at a jump discontinuity of the function  $f(t)$
- typical example:  $f(t)$  as extension of  $\chi_{[-1/2,1/2)}(t)$  to a 2-periodic function
- Fourier coefficients

$$\widehat{f}[k] = \frac{1}{2} \int_{-1}^1 \chi_{[-1/2,1/2)}(t) \frac{e^{-\pi ikt}}{\sqrt{2}} dt = \begin{cases} 1/2 & k = 0 \\ 0 & k \neq 0 \text{ and even} \\ \frac{(-1)^{(k-1)/2}}{\pi k} & k \text{ odd} \end{cases}$$

- Fourier series

$$\frac{1}{2} + \frac{2}{\pi} \cos(\pi t) - \frac{2}{3\pi} \cos(3\pi t) + \frac{2}{5\pi} \cos(5\pi t) \mp \dots$$

- Approximation

$$S_N(t) = \frac{1}{2} + \frac{2}{\pi} \sum_{n=1}^N (-1)^{n-1} \frac{\cos((2n-1)\pi t)}{2n-1}$$

- graphical display

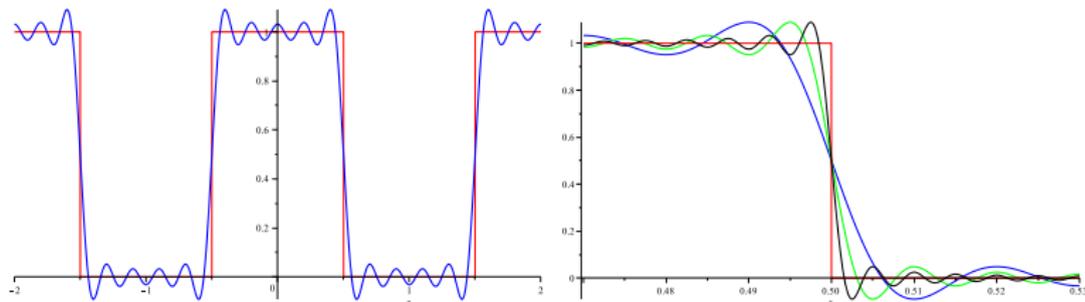


Figure:  $S_5$  (left),  $S_{50}$ ,  $S_{100}$ ,  $S_{200}$  (right)

- Notabene: the “overshooting” of the approximation does NOT disappear as  $N \rightarrow \infty$ !

- Consider a function  $f(t)$  which vanishes outside a finite interval  $[-L_0, L_0)$ , and for  $L \geq L_0$  consider

$$f_L(t) = \begin{cases} f(t) & \text{for } |t| \leq L_0 \\ 0 & \text{for } L_0 \leq |t| \leq L \end{cases}$$

as a  $2L$ -periodic function

- Fourier coefficients (analysis)

$$\hat{f}_L[k] = \frac{1}{2L} \int_{-L}^L f_L(t) e^{-2\pi ikt/2L} dt$$

- Synthesis formula

$$f_L(t) = \sum_{k \in \mathbb{Z}} \hat{f}_L[k] e^{2\pi ikt/2L}$$

- Now define for all  $s \in \mathbb{R}$  and  $L \geq L_0$

$$\widehat{f}(s) = \int_{-L}^L f_L(t) e^{-2\pi i s t} dt$$

This definition is independent of  $L$  !

- Then for all  $s \in \mathbb{R}$  of the form  $s = \frac{k}{2L}$  with  $k \in \mathbb{Z}$  it is true that

$$\widehat{f}(s) = 2L \cdot \widehat{f}_L[k]$$

- For  $L \geq L_0$  one has

$$g_L : \frac{k}{2L} \mapsto 2L \cdot \widehat{f}_L[k] \quad (= \widehat{f}(\frac{k}{2L})) \quad (k \in \mathbb{Z})$$

as a discrete function

- Conclusion: For  $L \rightarrow \infty$  the graphs of the discrete functions  $g_L$  “converge” to the graph of a function  $s \mapsto \widehat{f}(s)$  defined on  $\mathbb{R}$

- Furthermore

$$f_L(t) = \sum_{k \in \mathbb{Z}} \hat{f}_L[k] e^{2\pi i k t / 2L} = \sum_{k \in \mathbb{Z}} \frac{1}{2L} \hat{f}\left(\frac{k}{2L}\right) e^{2\pi i (k/2L) t}$$

The right-hand side is the Riemann sum for the integral

$$\int_{\mathbb{R}} \hat{f}(s) e^{2\pi i s t} ds$$

- Thus for  $L \rightarrow \infty$  one expects a synthesis formula

$$f(t) := f_{\infty}(t) = \int_{\mathbb{R}} \hat{f}(s) e^{2\pi i s t} ds$$

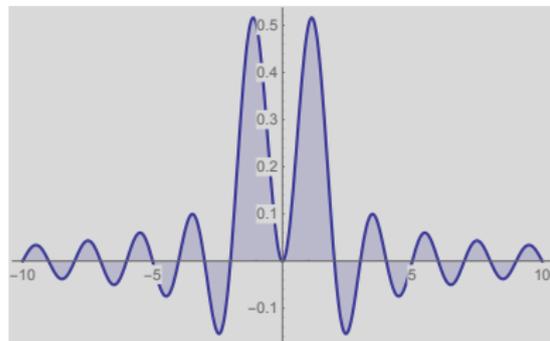
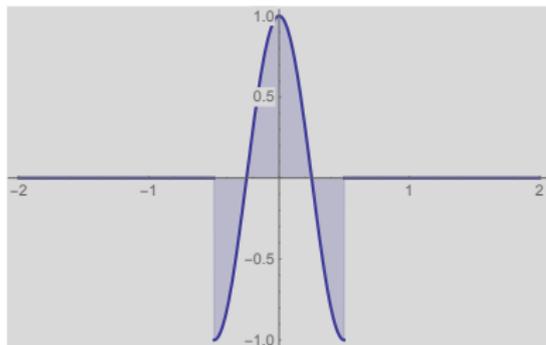
- together with an analysis formula

$$\hat{f}(s) = \int_{\mathbb{R}} f(t) e^{-2\pi i s t} dt$$

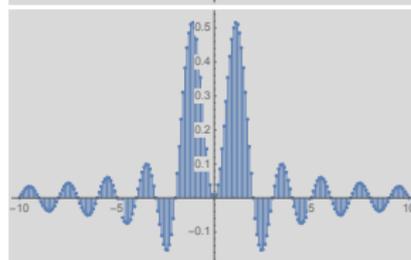
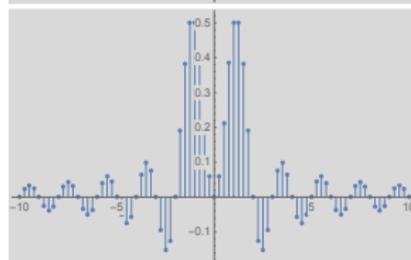
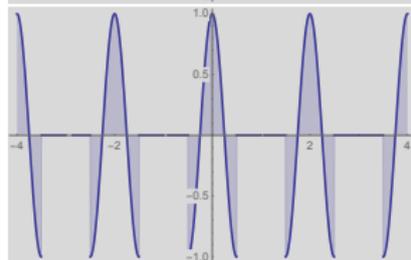
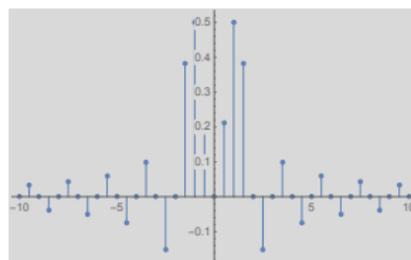
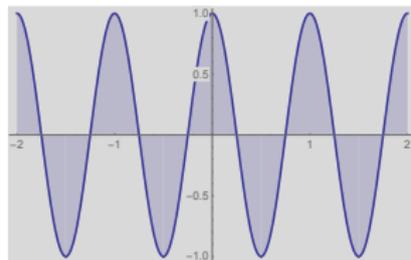
- Example: A 1-periodic function and its Fourier transform

$$f(t) = \cos(2\pi t) \quad |t| \leq 1/2$$

$$\hat{f}(s) = \frac{s \sin \pi s}{\pi - \pi s^2}$$



- Schematic display of the transition Fourier series  $\rightarrow$  Fourier transform for  $f(t) = \cos(2\pi t)$  with  $|t| \leq 1/2$  and  $L = 1, 2, 4$



- The relevant Hilbert space for the Fourier transform is  $\mathcal{L}^2(\mathbb{R})$ , the vector space of (Lebesgue-)square-integrable functions on  $\mathbb{R}$
- Inner product and norm in  $\mathcal{L}^2(\mathbb{R})$

$$\langle f | g \rangle = \int_{\mathbb{R}} f(t) \overline{g(t)} dt \quad \|f\|^2 = \langle f | f \rangle = \int_{\mathbb{R}} |f(t)|^2 dt$$

- Serious defect: simple functions like polynomials, trigonometric functions and complex exponentials do NOT belong to  $\mathcal{L}^2(\mathbb{R})$ ; in particular, the family

$$\{\omega_s(t) = e^{2\pi ist}\}_{s \in \mathbb{R}}$$

cannot be a basis of the Hilbert space  $\mathcal{L}^2(\mathbb{R})$  !

- For integrable functions  $f(t)$  the inner products

$$\widehat{f}(s) := \langle f | \omega_s(t) \rangle = \int_{\mathbb{R}} f(t) e^{-2\pi ist} dt$$

are nevertheless well defined!

- Definition: For “suitable” functions  $f : \mathbb{R} \rightarrow \mathbb{C}$  their *Fourier transform*  $\hat{f} : \mathbb{R} \rightarrow \mathbb{C}$  is defined by

$$\text{(Analysis)} \quad \hat{f}(s) = \int_{-\infty}^{\infty} f(t) e^{-2\pi i s t} dt \quad (s \in \mathbb{R})$$

Often denoted as  $\mathcal{F}_t[f(t)]$  or  $\mathcal{F}[f]$  instead of  $\hat{f}$

- *Inversion formula*: If the function  $f(t)$  is sufficiently well-behaved, one expects that it can be reconstructed from its Fourier transform  $\hat{f}$  by:

$$\text{(Synthesis)} \quad f(t) = \int_{-\infty}^{\infty} \hat{f}(s) e^{2\pi i s t} ds \quad (t \in \mathbb{R})$$

Often denoted as  $f = \mathcal{F}_s^{-1}[\hat{f}(s)]$  or  $f = \mathcal{F}^{-1}[\hat{f}]$

- If this holds, then  $f(t)$  is “continuous linear combination” (superposition) of harmonics (complex exponentials)
- $\hat{f}(s)$  is the amplitude or intensity of  $\omega_s(t) = e^{2\pi i s t}$  in  $f(t)$

- Attention! In the literature there are many slightly different conventions used of the definition of the *Fourier transform*. The type of expressions used is

$$\hat{f}(s) = \sqrt{\frac{|b|}{(2\pi)^{1-a}}} \int_{-\infty}^{\infty} f(t) e^{i b s t} dt$$

- with the following conventions
  - $(a, b) = (0, 1)$  (modern physics, Mathematica)
  - $(a, b) = (1, -1)$  (mathematics, systems theory, Maple)
  - $(a, b) = (-1, 1)$  (classical physics)
  - $(a, b) = (0, -2\pi)$  (signal processing, *this lecture*)
- The formula for the inverse transform has to be adapted accordingly

- Comments:

- Fourier transform is a linear transformation, it is even *unitary* (=complex-orthogonal) transform ( $\rightarrow$  Parseval-Plancherel)
- The definition of the Fourier transform makes sense if  $f \in \mathcal{L}^1(\mathbb{R})$ , i.e., if  $f$  is integrable (in the sense of Lebesgue):  $\|f\|_1 = \int_{\mathbb{R}} |f(t)| dt < \infty$
- For the inversion formula to make sense, one should have  $\hat{f} \in \mathcal{L}^1(\mathbb{R})$ , which unfortunately is not guaranteed, it holds, however, e.g., if  $f \in \mathcal{L}^1(\mathbb{R})$  is continuous
- The complex exponentials  $t \mapsto e^{2\pi ist}$  belong neither to  $\mathcal{L}^1(\mathbb{R})$ , nor to  $\mathcal{L}^2(\mathbb{R})$ , i.e., they cannot be taken as basis functions
- In order to get a satisfactory theory of the Fourier transform one has to extend the space of admissible functions ( $\rightarrow$  *distributions*)

- Examples (1)

$f(t)$	$\longleftrightarrow$	$\hat{f}(s)$
$\chi_{[-a,a]}(t)$	$\longleftrightarrow$	$\frac{\sin(2\pi as)}{\pi s}$
$(1 - [\frac{t}{a}]) \cdot \chi_{[-a,a]}(t)$	$\longleftrightarrow$	$\frac{\sin^2(\pi as)}{a(\pi s)^2}$
$e^{-a t }$	$\longleftrightarrow$	$\frac{2a}{a^2 + (2\pi s)^2}$
$e^{-at^2}$	$\longleftrightarrow$	$\sqrt{\frac{\pi}{a}} e^{-(\pi s)^2/a}$

- Fourier transform can/must be extended to cover familiar functions
- Examples (2)

$f(t)$	$\longleftrightarrow$	$\widehat{f}(s)$
1	$\longleftrightarrow$	$\delta(s)$
$e^{2\pi i a t}$	$\longleftrightarrow$	$\delta(a - s)$
$1 + a t + b t^2$	$\longleftrightarrow$	$\delta(s) + \frac{ia \delta'(s)}{2\pi} - \frac{b \delta''(s)}{4\pi^2}$
$\frac{1}{1 + a t^2}$	$\longleftrightarrow$	$\frac{\pi}{\sqrt{a}} \left( e^{2\pi s/\sqrt{a}} \theta(-2\pi s) + e^{-2\pi s/\sqrt{a}} \theta(2\pi s) \right)$

where

- $\theta(t) = \chi_{t>0}(t)$  denotes Heaviside's jump function
- $\delta(t) = \frac{d}{dt} \theta(t)$  denotes Diracs Delta- "function"

- A possible definition of the *distribution*  $\delta(t)$  is furnished by

$$\delta : f(t) \mapsto f(0)$$

for “sufficiently well-behaved” functions  $f(t)$  (“test functions”), often written as

$$\int_{-\infty}^{\infty} \delta(t) f(t) dt = f(0)$$

- No function in the traditional sense can have this property, so  $\delta(t)$  is not a function, but a linear functional

- Translation vs. Modulation

$$\widehat{f(t - a)}(s) = e^{-2\pi ias} \cdot \widehat{f(t)}(s)$$

- Dilation (Scaling)

$$\sqrt{a} \widehat{f(at)}(s) = \frac{1}{\sqrt{a}} \widehat{f(t)}\left(\frac{s}{a}\right)$$

- Derivation vs. Multiplication

$$\frac{d}{dt} \widehat{f(t)}(s) = 2\pi is \cdot \widehat{f(t)}(s)$$

- Convolution

$$(f \star g)(t) := \int_{-\infty}^{\infty} f(x) g(t - x) dx$$

Convolution theorem

$$\widehat{(f \star g)}(t)(s) = \widehat{f(t)}(s) \cdot \widehat{g(t)}(s)$$

- Dilation

- The  $a$ -Dilation  $(D_a f)(t)$  of a function  $f(t)$  is defined as

$$(D_a f)(t) = \sqrt{a} f(at)$$

- Dilation means *stretching* (for  $0 < a < 1$ ) resp. *squeezing* (for  $a > 1$ ) of the graph of  $f$  so that the norm is conserved

$$\|D_a f\| = \|f\|$$

- The behavior of the Fourier transform w.r.t. dilation can be succinctly described by

$$\widehat{D_a f} = D_{1/a} \widehat{f}$$

This antagonistic property is one of the characteristics of the Fourier transform ( $\rightarrow$  uncertainty relation)

- Derivation vs. multiplication

Under suitable conditions on  $f(t)$  by partial integration or by interchanging integration and derivation:

- $\widehat{f'(t)}(s) = (2\pi is) \cdot \widehat{f}(s)$

$$\begin{aligned}\widehat{f'(t)}(s) &= \int_{\mathbb{R}} f'(t) e^{-2\pi ist} dt \\ &= e^{-2\pi ist} f(t) \Big|_{t \rightarrow -\infty}^{t \rightarrow +\infty} + (2\pi is) \cdot \int_{\mathbb{R}} f(t) e^{-2\pi ist} dt \\ &= (2\pi is) \cdot \int_{\mathbb{R}} f(t) e^{-2\pi ist} dt\end{aligned}$$

- $\widehat{t \cdot f(t)}(s) = \frac{-1}{2\pi i} \cdot \frac{d}{ds} \widehat{f}(s)$

$$\widehat{t \cdot f(t)}(s) = \int_{\mathbb{R}} t \cdot f(t) e^{-2\pi ist} dt = \frac{-1}{2\pi i} \cdot \frac{d}{ds} \int_{\mathbb{R}} f(t) e^{-2\pi ist} dt$$

- Derivation: Smoothness and vanishing at infinity

- Riemann-Lebesgue Lemma*

$$f \in \mathcal{L}^1(\mathbb{R}) \Rightarrow \begin{cases} \widehat{f} \text{ is uniformly continuous on } \mathbb{R} \\ \text{and } \lim_{|s| \rightarrow \infty} \widehat{f}(s) = 0 \end{cases}$$

- “ $t^N \cdot f(t) \in \mathcal{L}^1(\mathbb{R})$ ” means:  $f(t)$  vanishes fast as  $t \rightarrow \pm\infty$ :

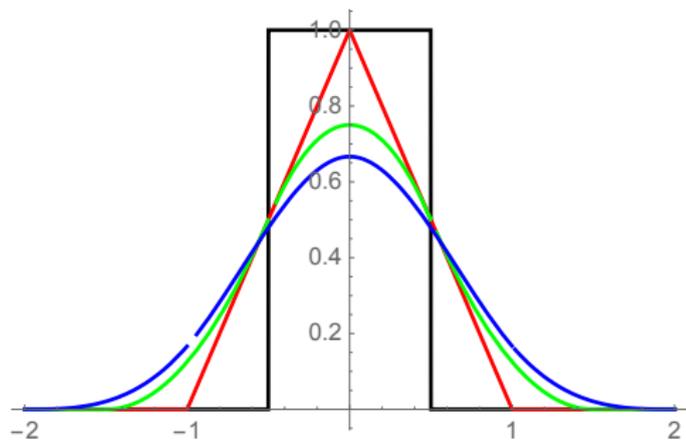
$$\int_{\mathbb{R}} |t^N f(t)| dt < \infty, \text{ so typically } f(t) \in \mathcal{O}(t^{-N-1-\varepsilon})$$

- The faster a function  $f(t)$  vanishes as  $t \rightarrow \pm\infty$ , the smoother (higher order differentiable) is  $\widehat{f}(s)$  – and conversely

$$\left. \begin{array}{l} f(t) \in \mathcal{L}^1(\mathbb{R}) \\ t^N \cdot f(t) \in \mathcal{L}^1(\mathbb{R}) \end{array} \right\} \Leftrightarrow \begin{cases} \widehat{f} \in \mathcal{C}^N(\mathbb{R}) \text{ and} \\ \frac{d^k}{ds^k} \widehat{f}(s) = \frac{-1}{(2\pi i)^k} \widehat{t^k f(t)}(s) \quad (0 \leq k \leq N) \end{cases}$$

- Derivation and multiplication with the variable* are “complementary”

- B-Spline functions and their Fourier transforms



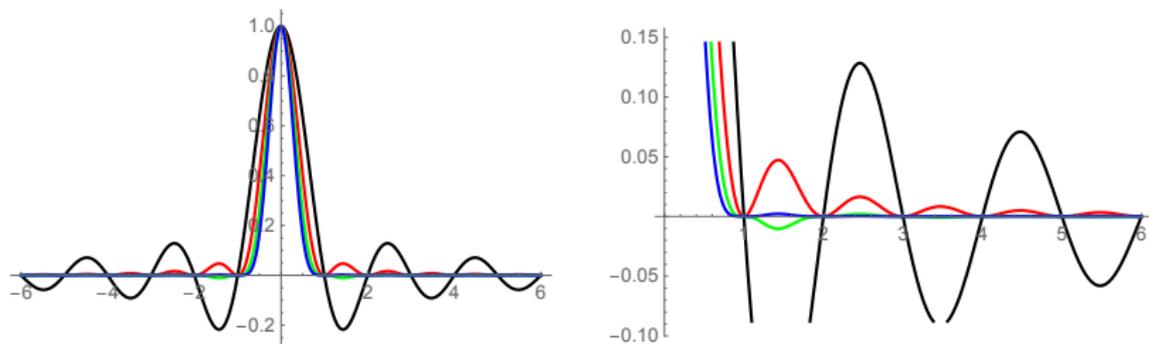
Iterated convolutions of the box function  $b(t)$

$$b^{*n}(t) = (b \star b \star \cdots \star b)(t) \quad (n \text{ factors})$$

$b^{*n}$  is  $(n - 2)$ -fold differentiable

$n$ : 1=black, 2=red, 3=green, 4 = blue

- B-Spline functions and their Fourier transforms



The Fourier transforms are the functions

$$\widehat{b^{*n}}(s) = \text{sinc}(\pi s)^n = \frac{\sin(\pi s)^n}{(\pi s)^n} \in \mathcal{O}(s^{-n})$$

$n$ : 1=black, 2=red, 3=green, 4 = blue

- Definition of convolution

$$(f \star g)(t) := \int_{-\infty}^{\infty} f(x) g(t-x) dx$$

- If  $g(t) = \omega_s(t) = e^{2\pi i s t}$ :

$$\begin{aligned} (f \star \omega_s)(t) &= \int_{-\infty}^{\infty} f(x) e^{2\pi i s(t-x)} dx \\ &= e^{2\pi i s t} \int_{-\infty}^{\infty} f(x) e^{-2\pi i s x} dx = \widehat{f}(s) \cdot \omega_s(t) \end{aligned}$$

- Convolution by a fixed function  $f(t)$

$$\mathcal{C}_f : g(t) \mapsto (f \star g)(t)$$

is a linear transformation which has

- the complex exponentials  $\omega_s(t) = e^{2\pi i s t}$  as *eigenfunctions*
- with Fourier transform value  $\widehat{f}(s)$  as the corresponding *eigenvalues*
- Convolution with  $\delta(t)$  replicates  $f(t)$

$$(\delta \star f)(t) = \int_{-\infty}^{\infty} \delta(x) f(t-x) dx = f(t)$$

- The convolution theorem

$$\begin{array}{ccc}
 f, g & \xrightarrow{\mathcal{F}} & \widehat{f}, \widehat{g} \\
 \downarrow \star & & \downarrow \cdot \\
 f \star g & \xrightarrow{\mathcal{F}} & \widehat{f \star g} = \widehat{f} \cdot \widehat{g}
 \end{array}$$

- Main application of convolution

“Filtering in the frequency domain”

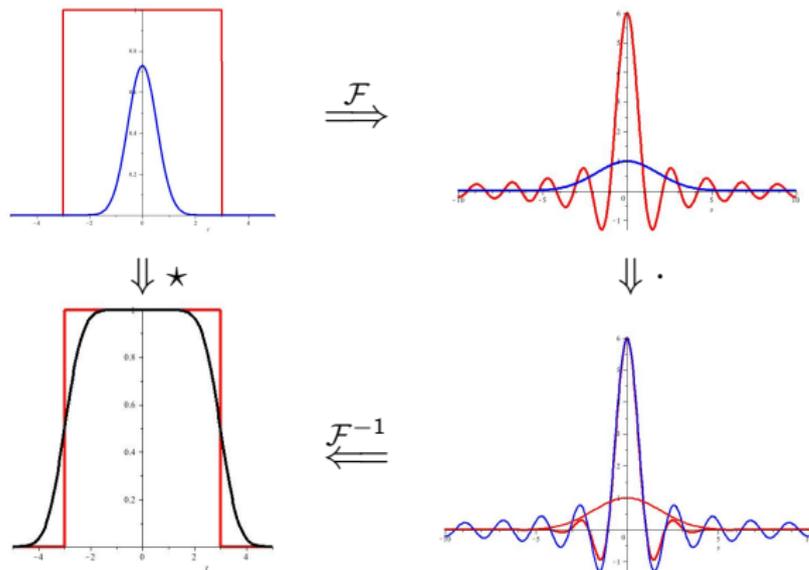
$$\begin{array}{ccc}
 f, g & \xrightarrow{\mathcal{F}} & \widehat{f}, \widehat{g} \\
 \downarrow \star & & \downarrow \cdot \\
 f \star g = \mathcal{F}^{-1}(\widehat{f} \cdot \widehat{g}) & \xleftarrow{\mathcal{F}^{-1}} & \widehat{f} \cdot \widehat{g}
 \end{array}$$

- Proof of the convolution theorem (sketch)

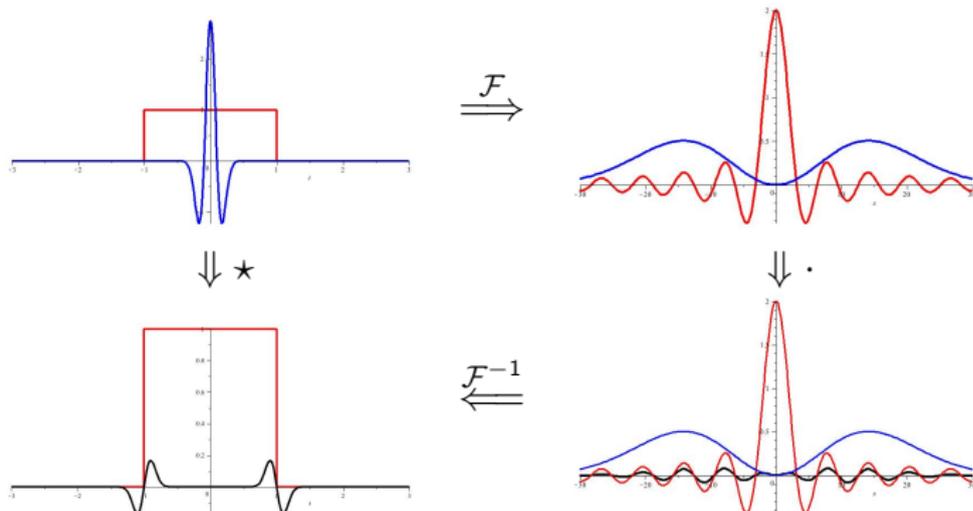
$$\begin{aligned}
 \widehat{f \star g}(s) &= \int_{\mathbb{R}} (f \star g)(t) e^{-2\pi i s t} dt && \text{def. of FT} \\
 &= \int_{\mathbb{R}} \int_{\mathbb{R}} f(x) g(t-x) dx e^{-2\pi i s t} dt && \text{def. of } \star \\
 &= \int_{\mathbb{R}} \int_{\mathbb{R}} f(x) e^{-2\pi i s x} g(t-x) e^{-2\pi i s (t-x)} dx dt \\
 &= \int_{\mathbb{R}} f(x) e^{-2\pi i s x} \int_{\mathbb{R}} g(t-x) e^{-2\pi i s (t-x)} dt dx && \int_t \int_x \equiv \int_x \int_t \\
 &= \int_{\mathbb{R}} f(x) e^{-2\pi i s x} \int_{\mathbb{R}} g(t) e^{-2\pi i s t} dt dx && t \mapsto t+x \\
 &= \int_{\mathbb{R}} f(x) e^{-2\pi i x s} \widehat{g}(s) dx && \text{def. of FT} \\
 &= \widehat{f}(s) \cdot \widehat{g}(s) && \text{def. of FT}
 \end{aligned}$$

The crucial point is the change of the order of integration!

- Low-pass filtering with a Gauss filter



- High-pass filtering with a Mexhat filter



- A fundamental consequence of the convolution theorem:

The *Parseval-Plancherel identity*: Fourier transform is an isometry!

- For  $f, g \in \mathcal{L}^2$  s.th. also  $\widehat{f}, \widehat{g} \in \mathcal{L}^2$ , one has

$$\langle f | g \rangle = \langle \widehat{f} | \widehat{g} \rangle \quad \text{and in particular} \quad \begin{cases} \|f\| = \|\widehat{f}\| \\ f \perp g \Leftrightarrow \widehat{f} \perp \widehat{g} \end{cases}$$

Sketch of proof

Define  $\widetilde{g}(t) = \overline{g(-t)}$  and check that  $\widehat{\widetilde{g}}(s) = \overline{\widehat{g}(s)}$ , then

$$\begin{aligned} \langle \widehat{f} | \widehat{g} \rangle &= \int \widehat{f}(s) \cdot \overline{\widehat{g}(s)} ds = \int \widehat{f}(s) \cdot \widehat{\widetilde{g}}(s) ds = \\ &= \int (\widehat{f \star \widetilde{g}})(s) ds = (f \star \widetilde{g})(0) = \int f(t) \cdot \widetilde{g}(-t) dt \\ &= \int f(t) \cdot \overline{g(t)} dt = \langle f | g \rangle \end{aligned}$$

- Uncertainty relation

- For  $f(t) \in \mathcal{L}^2(\mathbb{R})$  with

$$\|f\|^2 = \int_{\mathbb{R}} |f(t)|^2 dt = 1$$

then  $t \mapsto |f(t)|^2$  can be seen as a probability density function on  $\mathbb{R}$

- Expectation and variance of this probability density are given by

$$\mu(f) = \int_{\mathbb{R}} t |f(t)|^2 dt \quad \sigma^2(f) = \int_{\mathbb{R}} (t - \mu(f))^2 |f(t)|^2 dt$$

- Because of the Parseval-Plancherel identity one also has  $\|\widehat{f}\| = 1$ ;  $\mu(\widehat{f})$  and  $\sigma^2(\widehat{f})$  are defined analogously
  - Then the *Heisenberg inequality* holds:

$$\sigma^2(f) \cdot \sigma^2(\widehat{f}) \geq \frac{1}{(4\pi)^2}$$

(For a proof see the Lecture Notes)

- Examples

$f(t, a)$	$\sigma^2(f)$	$\hat{f}(s, a)$	$\sigma^2(\hat{f})$
$\sqrt{a} \chi_{[-1/2, 1/2]}(at)$	$\frac{1}{12a^2}$	$\sqrt{a} \frac{\sin(\pi s/a)}{\pi s}$	$\infty$
$\sqrt{\frac{3}{2a}} (1 -  at ) \cdot \chi_{-1/a, 1/a}(t)$	$\frac{1}{10a^2}$	$\sqrt{\frac{3}{2a}} \frac{(\sin(\pi s/a))^2}{\pi^2 s^2}$	$\frac{3a^2}{4\pi^2}$
$\sqrt{a} e^{-a t }$	$\frac{1}{2a^2}$	$2 \frac{a^{3/2}}{a^2 + 4\pi^2 s^2}$	$\frac{a^2}{4\pi^2}$
$\sqrt[4]{\frac{2a}{\pi}} e^{-at^2}$	$\frac{1}{4a}$	$\sqrt[4]{\frac{2a}{\pi}} e^{-\frac{\pi^2 s^2}{a}}$	$\frac{a}{4\pi^2}$

- Graphical illustration of uncertainty: *Heisenberg boxes*

- For any function  $f(t)$  and  $a > 0, b \in \mathbb{R}$  let

$$f_{a,b}(t) = \sqrt{a} \cdot f(at - b), \quad \mu_{a,b} = \mu(f_{a,b}), \quad \sigma_{a,b}^2 = \sigma^2(f_{a,b}),$$

and similarly for  $\hat{f}(s)$

- Then

$$\mu_{a,b} = \frac{\mu + b}{a}, \quad \sigma_{a,b}^2 = \frac{\sigma^2}{a^2}, \quad \hat{\mu}_{a,b} = a\hat{\mu}, \quad \hat{\sigma}_{a,b}^2 = a^2\hat{\sigma}^2$$

- The Heisenberg box for the function  $f(t)$  is the rectangle in the  $(s, t)$ -plane centered at  $(\mu_{a,b}, \hat{\mu}_{a,b})$  and with side lengths  $(\sigma_{a,b}, \hat{\sigma}_{a,b})$ . This box characterizes the simultaneous uncertainty of  $f(t)$  in the time/space domain and in the frequency domain  
The box area  $\sigma_{a,b} \cdot \hat{\sigma}_{a,b} \geq \frac{1}{4\pi}$  is independent of scaling  $a$  and translation  $b$  !

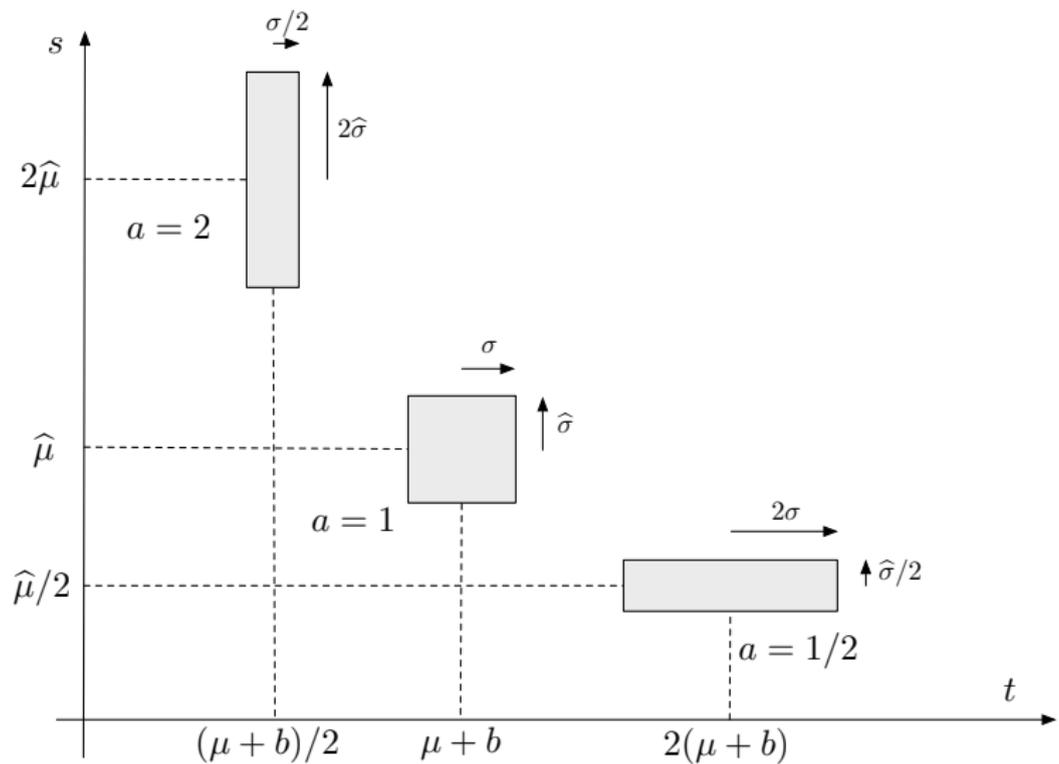


Figure: Heisenberg boxes for  $f_{a,b}(t)$  with  $a = 1/2$ ,  $a = 1$  and  $a = 2$

- *Poisson's formula*

For any sufficiently well-behaved function  $f : \mathbb{R} \rightarrow \mathbb{C}$  there is a relation

- between the values  $f(k)$  ( $k \in \mathbb{Z}$ ) at integer arguments
- and the values  $\hat{f}(s - n)$  ( $n \in \mathbb{Z}$ ) of its Fourier transform

$$\sum_{n=-\infty}^{\infty} \hat{f}(s - n) = \sum_{k=-\infty}^{\infty} f(k) e^{-2\pi i s k} \quad (s \in \mathbb{R})$$

- Note: the sum on the l.h.s. defines a 1-periodic function, the sum on the r.h.s. is a Fourier series
- In particular (take  $s = 0$ )

$$\sum_{n=-\infty}^{\infty} \hat{f}(n) = \sum_{k=-\infty}^{\infty} f(k)$$

- Equivalent version of Poisson's formula (for  $a > 0$ )

$$\sum_{n=-\infty}^{\infty} f(t - n/a) = a \cdot \sum_{k=-\infty}^{\infty} \hat{f}(k \cdot a) e^{2\pi i t k a}$$

- Sketch of proof (case  $a = 1$  suffices):

$\phi(t) = \sum_n f(t - n)$  is 1-periodic,

so if it has a Fourier series  $\phi(t) = \sum_{k \in \mathbb{Z}} \varphi[k] e^{2\pi i k t}$ , then

$$\begin{aligned} \varphi[k] &= \int_0^1 \phi(t) e^{-2\pi i k t} dt = \sum_{n \in \mathbb{Z}} \int_0^1 f(t - n) e^{-2\pi i k t} dt \\ &= \sum_{n \in \mathbb{Z}} \int_n^{n+1} f(t) e^{-2\pi i k t} dt = \int_{\mathbb{R}} f(t) e^{-2\pi i k t} dt = \hat{f}(k) \end{aligned}$$

- *Shannon-Nyquist* sampling theorem

If a signal  $f : \mathbb{R} \rightarrow \mathbb{C}$  is *band-limited* in the sense that

$$|s| > \frac{1}{2a} \implies \widehat{f}(s) = 0,$$

then  $f(t)$  can be perfectly reconstructed from its discrete *sampling values*  $f(k \cdot a)$  ( $k \in \mathbb{Z}$ ) by

$$\begin{aligned} f(t) &= \sum_{k \in \mathbb{Z}} f(k \cdot a) \cdot \frac{\sin\left(\frac{\pi}{a}(t - k \cdot a)\right)}{\frac{\pi}{a}(t - k \cdot a)} \\ &= \sum_{k \in \mathbb{Z}} f(k \cdot a) \cdot \operatorname{sinc}\left(\frac{1}{a}(t - k \cdot a)\right) \end{aligned}$$

This is Shannon's formula  $\left(\operatorname{sinc}(x) = \frac{\sin(\pi x)}{\pi x}\right)$

- From the band-limiting condition, only the ( $n = 0$ )-term from

$$\sum_{n \in \mathbb{Z}} \hat{f}(s - n/a)$$

in Poisson's formula survives, so that

$$\hat{f}(s) = a \sum_{k \in \mathbb{Z}} f(k \cdot a) \cdot e^{-2\pi i s k \cdot a}$$

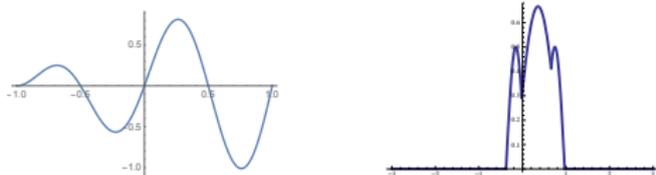
and thus

$$\begin{aligned} f(t) &= \int_{\mathbb{R}} \hat{f}(s) e^{2\pi i s t} ds \\ &= \int_{-1/(2a)}^{1/(2a)} \hat{f}(s) e^{2\pi i s t} ds \\ &= a \sum_{k \in \mathbb{Z}} f(k \cdot a) \int_{-1/(2a)}^{1/(2a)} e^{2\pi i s(t - k \cdot a)} ds \\ &= \sum_{k \in \mathbb{Z}} f(k \cdot a) \cdot \frac{\sin \frac{\pi}{a}(t - k \cdot a)}{\frac{\pi}{a}(t - k \cdot a)} \end{aligned}$$

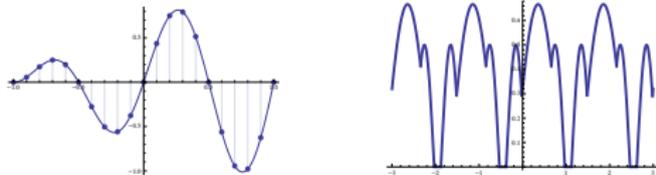
- What sampling really means?

- Sampling a continuous signal with frequency  $a$  means: repeating its spectrum periodically with distance  $a$

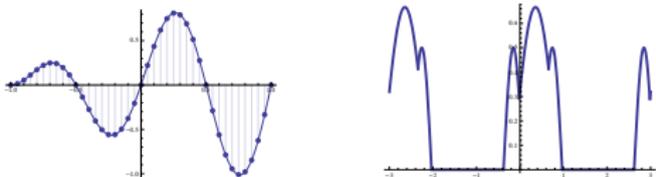
a function and its spectrum



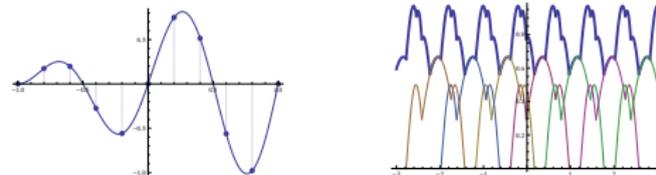
sampling with limit frequency



oversampling



undersampling



- In a purely formal way:

$$\delta(s) = \widehat{1}(s) = \int_{\mathbb{R}} e^{-2\pi i s t} dt$$

The integral doesn't make sense, but ...

- ... if  $\delta$  appears under an integral, it may work

$$\begin{aligned} \int_{\mathbb{R}} f(t) \delta(t) dt &= \int_{\mathbb{R}} f(t) \int_{\mathbb{R}} e^{-2\pi i s t} ds dt = \\ &= \int_{\mathbb{R}} \int_{\mathbb{R}} f(t) e^{-2\pi i s t} dt ds = \int_{\mathbb{R}} \widehat{f}(s) ds = f(0) \end{aligned}$$

which motivates the common definition (given earlier)

- Another characteristic property:  $\delta \star f = f$

$$(\delta \star f)(t) = \int_{\mathbb{R}} \delta(s) \cdot f(t - s) ds = f(t)$$

i.e.,  $\delta(t)$  acts as a neutral element w.r.t. convolution

No “proper” function can have this property. Therefore

$$\widehat{f}(s) = \widehat{(\delta \star f)}(s) = \widehat{\delta}(s) \cdot \widehat{f}(s) \implies \widehat{\delta} \equiv 1$$

- Translation of  $\delta$

- definition

$$\delta_a(t) = \delta(t - a) \quad \text{or} \quad \int f(t) \delta_a(t) dt = f(a) \quad \text{or} \quad \delta_a : f(t) \mapsto f(a)$$

- multiplication with a function

$$f(t) \cdot \delta_a(t) \equiv f(a) \cdot \delta_a(t)$$

- convolution with  $\delta_a$  is translation

$$(f \star \delta_a)(t) = \int f(t - x) \delta_a(x) dx = f(t - a)$$

- the Fourier transform of  $\delta_a$  is

$$\widehat{\delta_a}(s) = e^{-2\pi ias}$$

- DIRAC's comb

- definition

$$\text{III}(t) = \sum_{k \in \mathbb{Z}} \delta_k(t)$$

- multiplication (the sampling property)

$$f(t) \cdot \text{III}(t) = \sum_{k \in \mathbb{Z}} f(t) \delta_k(t) = \sum_{k \in \mathbb{Z}} f(k) \delta_k(t)$$

- convolution (the periodizing property)

$$(f \star \text{III})(t) = \sum_{k \in \mathbb{Z}} (f \star \delta_k)(t) = \sum_{k \in \mathbb{Z}} f(t - k)$$

- the Fourier transform of  $\text{III}(t)$  is

$$\widehat{\text{III}}(s) = \text{III}(s)$$

- DIRAC's comb and POISSON's formula

$$\begin{aligned}
 \sum_{k \in \mathbb{Z}} \widehat{f}(t - k) &= (\widehat{f} \star \text{III})(t) && \text{III-convolution} \\
 &= (\widehat{f} \star \widehat{\text{III}})(t) && \text{FT of III} \\
 &= \widehat{(f \cdot \text{III})}(t) && \text{convolution theorem} \\
 &= \widehat{\left( \sum_{k \in \mathbb{Z}} f(k) \delta_k \right)}(t) && \text{definition of III} \\
 &= \sum_{k \in \mathbb{Z}} f(k) \widehat{\delta}_k(t) && \text{linearity of FT} \\
 &= \sum_{k \in \mathbb{Z}} f(k) e^{-2\pi i k t} && \text{FT of } \delta_k
 \end{aligned}$$

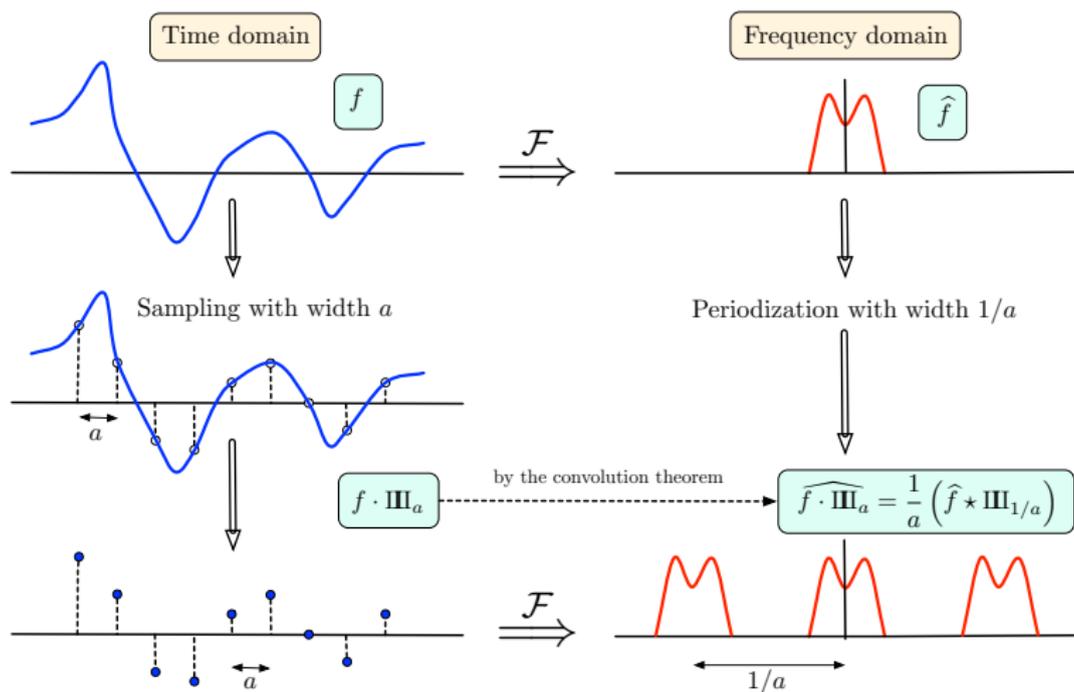


Figure: The sampling scheme

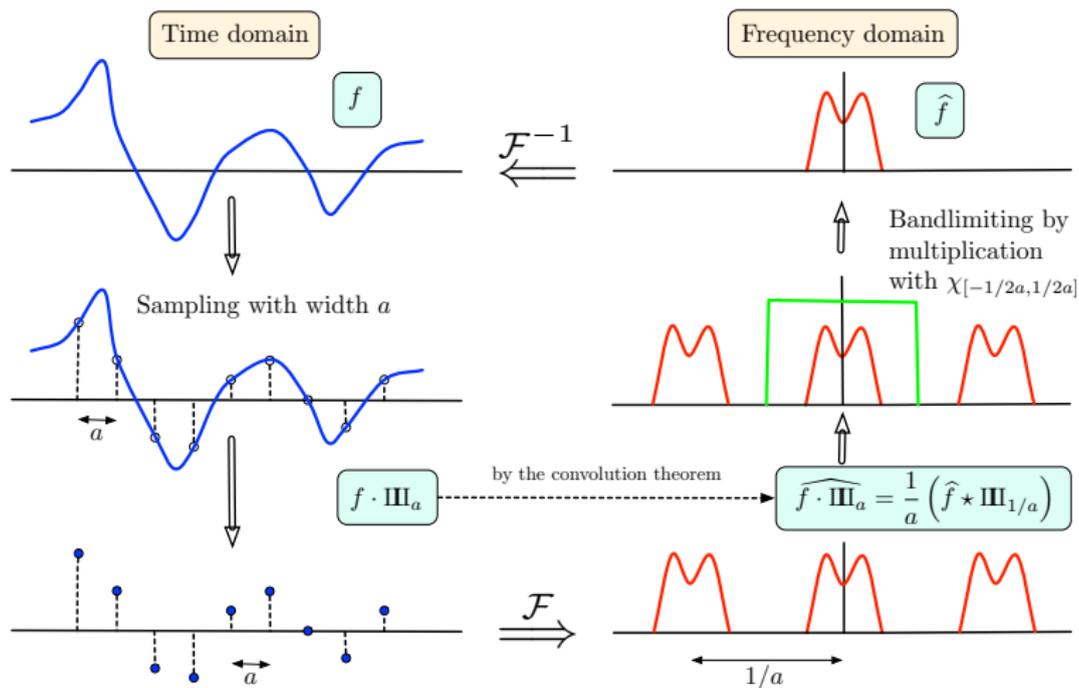


Figure: Reconstructing a sampled bandlimited signal

- $\delta_a(t) : f(t) \mapsto f(a)$
- $\text{III}_a(t) = \sum_{k \in \mathbb{Z}} \delta_{k \cdot a}(t) = \frac{1}{a} \sum_{k \in \mathbb{Z}} e^{2\pi i k t / a}$

	$\delta_a(t)$	$\text{III}_a(t)$
action on $f(t)$	$f(a)$	$\sum_{k \in \mathbb{Z}} f(k \cdot a)$
product with $f(t)$	$f(a) \cdot \delta_a(t)$	$\sum_{k \in \mathbb{Z}} f(k \cdot a) \cdot \delta_{k \cdot a}(t)$
scaling with $p > 0$	$\frac{1}{p} \delta_{a/p}(t)$	$\frac{1}{p} \text{III}_{a/p}(t)$
convolution with $f(t)$	$f(t - a)$	$\sum_{k \in \mathbb{Z}} f(t - k \cdot a)$
Fourier transform	$e^{-2\pi i a s}$	$\frac{1}{a} \text{III}_{1/a}(s)$

- periodizing a function  
 $f(t) \mapsto \sum_{k \in \mathbb{Z}} f(t - k \cdot a) = (f \star \text{III}_a)(t)$
- sampling a function  $f(t) \mapsto \sum_{k \in \mathbb{Z}} f(k \cdot a) \cdot \delta_{k \cdot a}(t) = (f \cdot \text{III}_a)(t)$

$f(t)$  a  $b$ -bandlimited function

$\Rightarrow$  the copies of  $\widehat{f}(s)$  contained in  $f \cdot \widehat{\mathbb{I}}_{1/b}$  do not overlap

$\Rightarrow \widehat{f}$  can be recovered by

$$\widehat{f} = \Pi_b \cdot f \cdot \widehat{\mathbb{I}}_{1/b} = b \cdot \Pi_b \cdot (\widehat{f} \star \mathbb{I}_b)$$

where  $\Pi_b(t) = \chi_{[-b/2, b/2]}(t)$ . Now compute:

$$\begin{aligned} f &= \mathcal{F}^{-1}(b \cdot \Pi_b \cdot (\mathcal{F}(f) \star \mathbb{I}_b)) && \text{inverse Fourier transform} \\ &= b \cdot \mathcal{F}^{-1}(\Pi_b) \star \mathcal{F}^{-1}(\mathcal{F}(f) \star \mathbb{I}_b) && \text{convolution theorem} \\ &= b \cdot \mathcal{F}^{-1}(\Pi_b) \star (f \cdot \mathcal{F}^{-1}(\mathbb{I}_b)) && \text{convolution theorem} \\ &= \mathcal{F}^{-1}(\Pi_b) \star (f \cdot \mathbb{I}_{1/b}) && \text{iFT of } \mathbb{I}_b \end{aligned}$$

which gives the celebrated Shannon-formula

$$f(t) = \text{sinc}(bt) \star \sum_{k \in \mathbb{Z}} f(k/b) \delta(t - k/b) = \sum_{k \in \mathbb{Z}} f(k/b) \text{sinc}(b(t - k/b))$$

where

$$\text{sinc}(t) = \frac{\sin(\pi t)}{\pi t} = \mathcal{F}^{-1}(\Pi_1)(t)$$