

Fourier transformation

Explains the mathematical concept of space, norm, and basis. Introduces Fourier transform and convolution.

WS 16/17

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Mathematical methods

Concepts of space and Fourier transform are connected and very important.

- Investigation of vibrating string, planet orbits
- Fourier transformation (1759-1805, Lagrange, Gauss, Fourier, Euler)
- Hilbert space (1900, Hilbert, Lebesgue, Quantum mechanics)

Euclidean space \mathbb{R}^N

\mathbf{x} and \mathbf{y} are vectors (x_1, x_2, \dots, x_N) and (y_1, y_2, \dots, y_N) .

Scalar product:

$$\langle \mathbf{x}, \mathbf{y} \rangle := \sum_{n=1}^N x_n y_n$$

Norm:

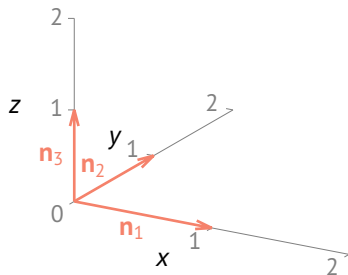
$$\|\mathbf{x}\| := \langle \mathbf{x}, \mathbf{x} \rangle^{\frac{1}{2}} = \left(\sum_{n=1}^N x_n^2 \right)^{\frac{1}{2}}$$

Basis:

$$\mathbf{n}_1 = (1, 0, \dots, 0), \mathbf{n}_2 = (0, 1, \dots, 0), \dots, \mathbf{n}_n = (0, 0, \dots, 1)$$

Euclidean space \mathbb{R}^N

Example: \mathbb{R}^3



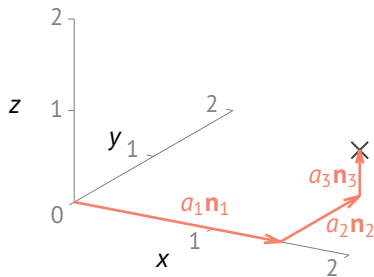
A space is spanned by its basis, here $\mathbf{n}_1, \mathbf{n}_2, \mathbf{n}_3$.

$$\langle \mathbf{n}_i, \mathbf{n}_j \rangle = \delta_{ij}$$

$$\|\mathbf{n}_i\| = \|\mathbf{n}_j\| = 1$$

Euclidean space \mathbb{R}^N

Example: \mathbb{R}^3



Any point can be constructed by a linear combination of the basis.

$$\mathbf{x} = a_1 \mathbf{n}_1 + a_2 \mathbf{n}_2 + a_3 \mathbf{n}_3 = \sum_{n=1}^3 a_n \mathbf{n}_n$$

$$a_n = \langle \mathbf{x}, \mathbf{n}_n \rangle$$

Complex space \mathbb{C}

$z_1 = x_1 + iy_1$ and $z_2 = x_2 + iy_2$ are complex numbers.

Scalar product:

$$\langle z_1, z_2 \rangle := z_1 z_2^* = (x_1 + iy_1)(x_2 - iy_2)$$

Norm:

$$\|z_1\| := \langle z_1, z_1 \rangle^{\frac{1}{2}} = (z_1 z_1^*)^{\frac{1}{2}}$$

Basis:

$$1, i$$

Hilbert space L^2

$f(x)$ and $g(x)$ are square-integrable functions.

Scalar product:

$$\langle f, g \rangle := \int_a^b f(x) g(x)^* dx$$

Norm:

$$\|f\|^2 := \langle f, f \rangle = \int_a^b |f(x)|^2 dx$$

Basis:

$$e^{i\omega x} = \cos(\omega x) + i \sin(\omega x) \quad \text{trigonometric functions}$$

Fourier series

Every periodic signal can be approximated by a finite summation of basis functions.

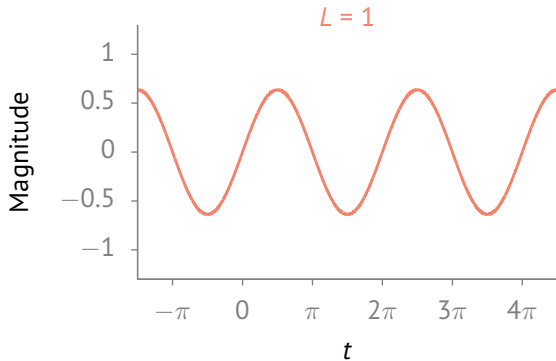
$$f(x) = \sum_{n=1}^N a_n e^{in\omega x}$$

$$a_n(\omega) = \langle f(x), e^{-in\omega x} \rangle = \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} f(x) e^{-in\omega x} dx$$

Fourier series

Example 1: sawtooth wave

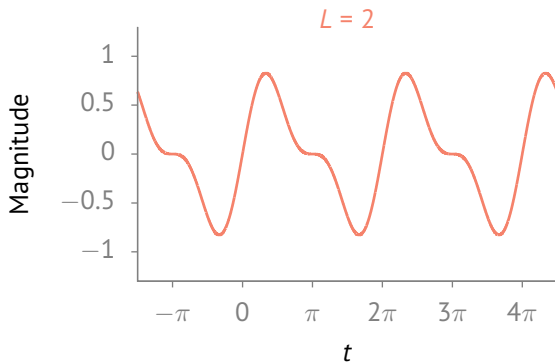
$$\text{sawtooth}(t) = \sum_{n=1}^N \frac{2 \sin(nt)}{n\pi}$$



Fourier series

Example 1: sawtooth wave

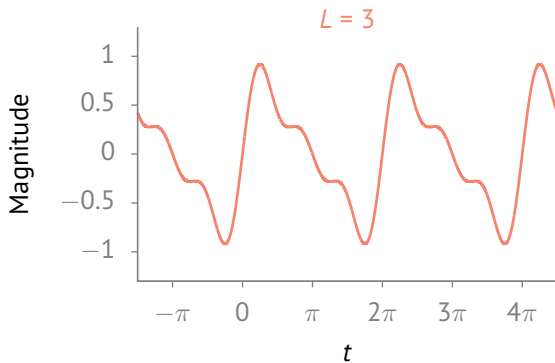
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Fourier series

Example 1: sawtooth wave

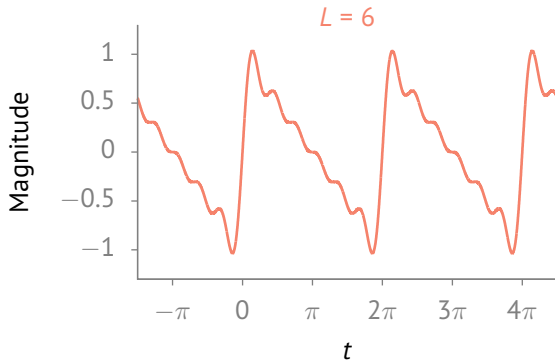
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Fourier series

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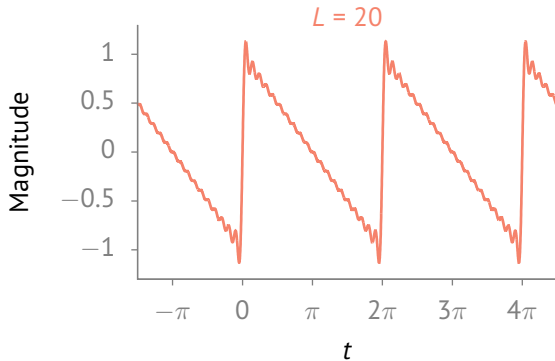
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Fourier series

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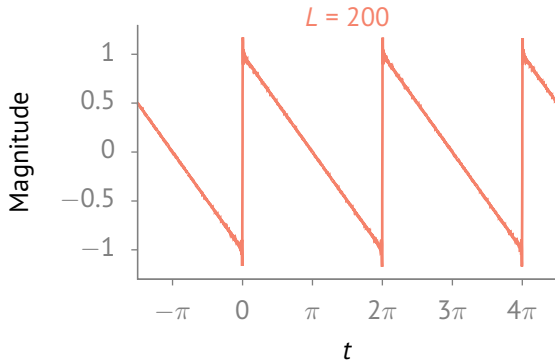
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Fourier series

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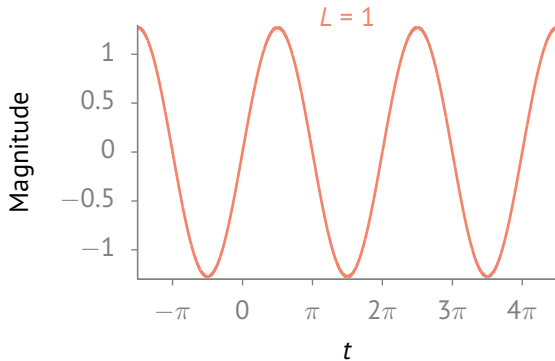
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Fourier series

Example 2: square wave

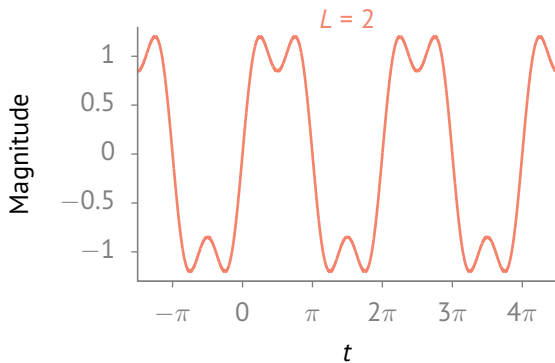
$$\text{square}(t) = \sum_{n=1}^N \frac{4 \sin((2n-1)t)}{(2n-1)\pi}$$



Fourier series

Example 2: square wave

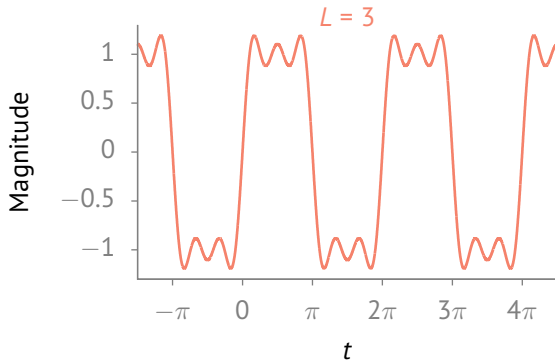
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Fourier series

Example 2: square wave

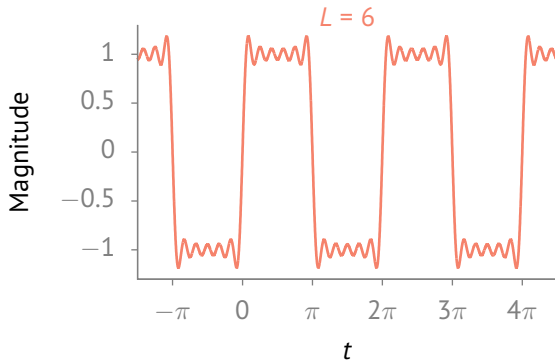
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Fourier series

Example 2: square wave

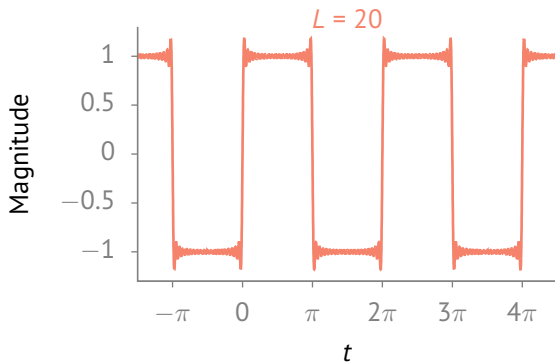
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Fourier series

Example 2: square wave

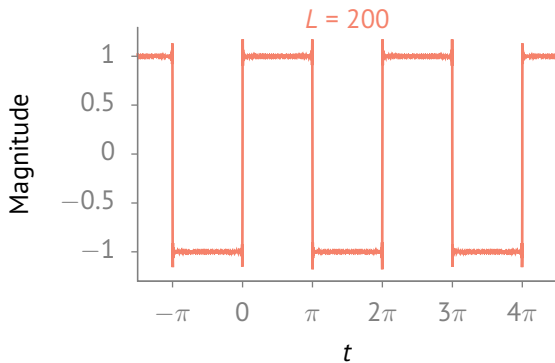
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Fourier series

Example 2: square wave

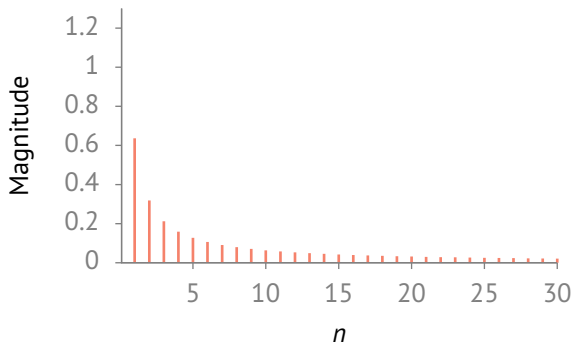
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Fourier series

Example 1: sawtooth wave

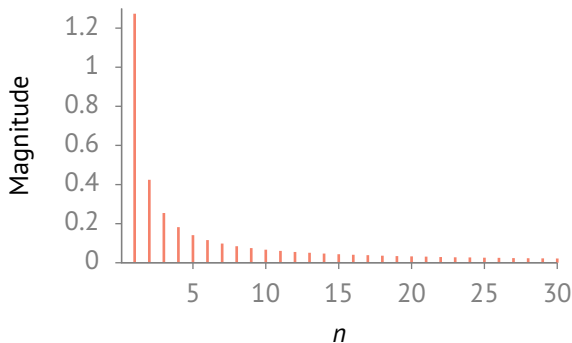
$$a_n = \frac{1}{T_0} \int_{-\frac{T_0}{2}}^{\frac{T_0}{2}} \text{sawtooth}(t) e^{-int} dt$$



Fourier series

Example 2: square wave

$$a_n = \frac{1}{T_0} \int_{-\frac{T_0}{2}}^{\frac{T_0}{2}} \text{square}(t) e^{-int} dt$$



Fourier transformation

Temporal Fourier transformation:

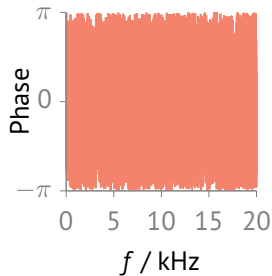
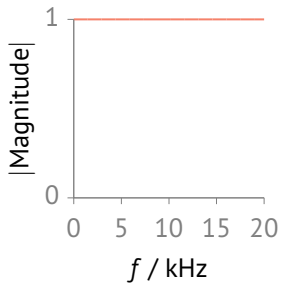
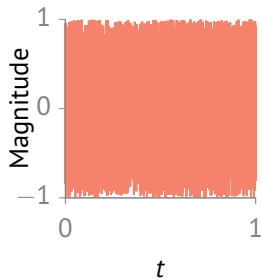
$$S(\omega) = \mathcal{F} \{s(t)\} := \int_{-\infty}^{\infty} s(t) e^{-i\omega t} dt$$

Inverse temporal Fourier transformation:

$$s(t) = \mathcal{F}^{-1} \{S(\omega)\} := \frac{1}{2\pi} \int_{-\infty}^{\infty} S(\omega) e^{i\omega t} d\omega$$

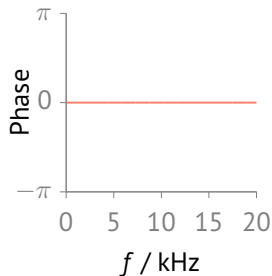
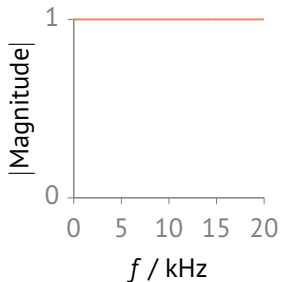
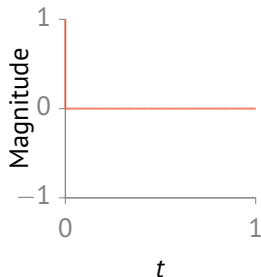
Fourier transformation

White noise



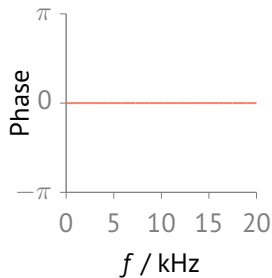
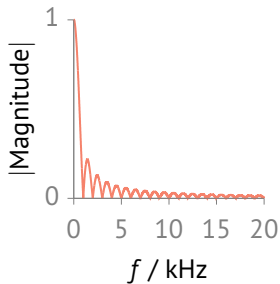
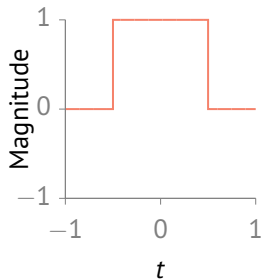
Fourier transformation

Dirac function



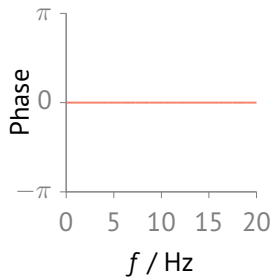
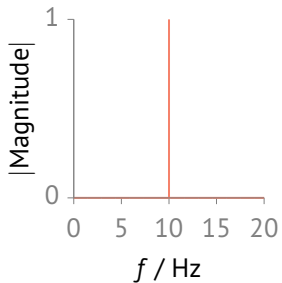
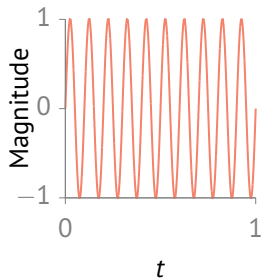
Fourier transformation

Rectangular function



Fourier transformation

Sinusoid



Convolution theorem

Convolution:

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

Convolution theorem:

$$\mathcal{F}\{f * g\} = \mathcal{F}\{f\} \cdot \mathcal{F}\{g\}$$

Take home

- Every acoustical signal is a combination of sinusoid tones with different frequency, amplitude and phase
- Fourier transformation is a way to characterize time signals