

Fast Multidimensional Signal Processing with Shearlab.jl

Héctor Andrade Loarca

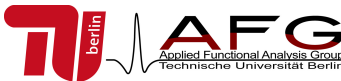
(github: **arsenal9971**)

Notebook and Beamer:

<https://github.com/arsenal997/Shearlab.jl/presentations/PyData>

TU Berlin, BMS

2nd of July, 2017



What is a signal?

Our definition

Function (or something that can be represented as) that contains information about the behavior or attributes of some phenomenon. It can be digital (discrete) or analog (continuous).

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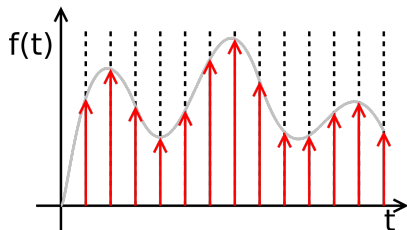


Figure: Digital and continuous one-dimensional signals

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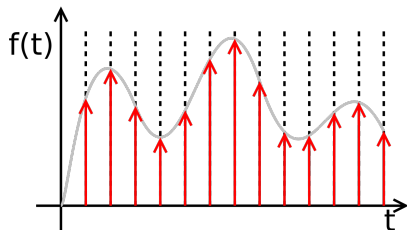


Figure: Digital and continuous one-dimensional signals

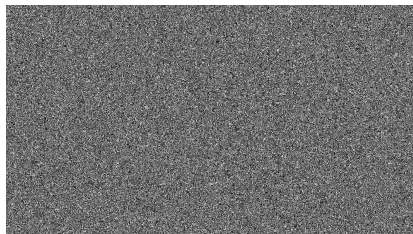


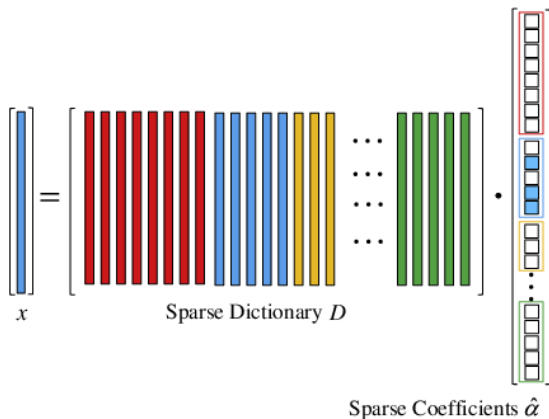
Figure: White noise, not a signal

Sparse representations of signals

- ▶ Relevant information in structured data is sparse, due the high correlation of its elements.

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- ▶ Goal: Find the right dictionary to represent optimally our data.

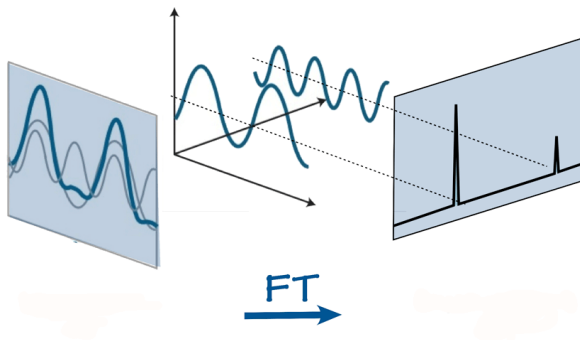


Fourier Transform (Fourier, 1822)

$$\hat{f}(\omega) := \int_{\mathbb{R}^n} f(x) e^{-i\langle x, \omega \rangle} dx$$

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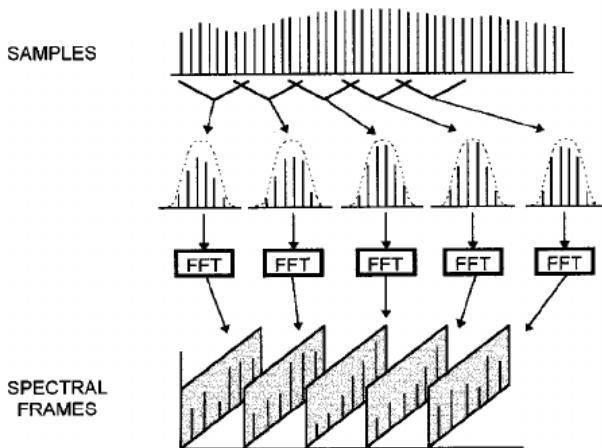


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$$\begin{aligned}\mathcal{W}_\psi f(a, b) &= \int_{\mathbb{R}} f(t) a^{-\frac{1}{2}} \overline{\psi\left(\frac{t-b}{a}\right)} dt \\ &= (f * D_a \bar{\psi}^*)(b), \quad (a, b) \in \mathbb{R}^+ \times \mathbb{R}\end{aligned}$$

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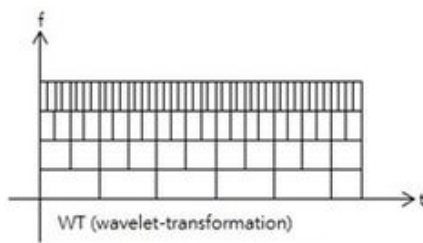
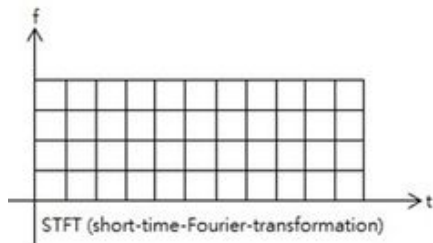
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Cartoon-like functions

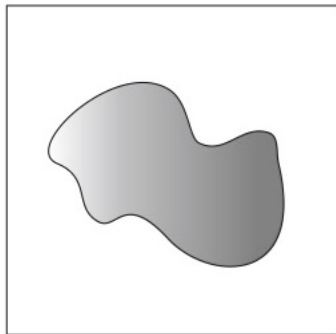
Definition

Let $f : \mathbb{R}^2 \rightarrow \mathbb{C}$, $f \in \mathcal{E}^2(\mathbb{R}^2)$ if $f = f_0 + \chi_B f_1$, with $B \subset [0, 1]^2$, $\partial B \in C^2$ and with bounded curvature. Moreover, $f_i \in C^2(\mathbb{R}^2)$ with $\|f_i\|_{C^2} \leq 1$ and $\text{supp} f_i \subset [0, 1]^2$ for $i = 0, 1$.

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Optimal error for 2D signals

Best N-term approx. error (Donoho, 2001)

Let $\{\psi_\lambda\}_{\lambda \in \Lambda} \subset L^2(\mathbb{R}^2)$ a frame. The optimal best N-Term approximation error for any $f \in \mathcal{R}^2(\mathbb{R}^2)$ is

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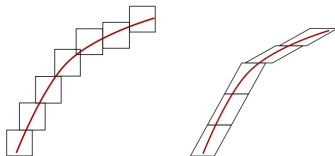
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How to solve this? Scaling and Shearing

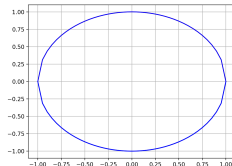
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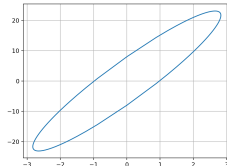
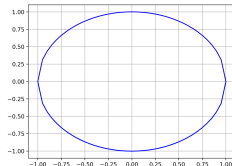
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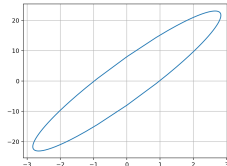
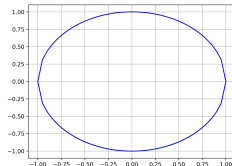
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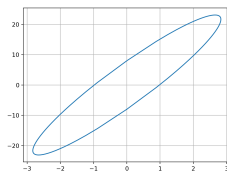
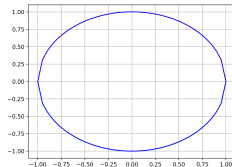
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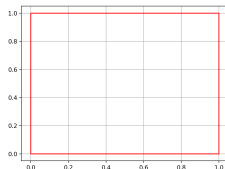
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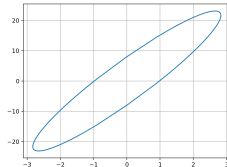
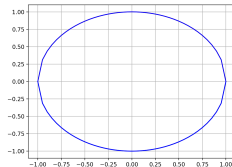
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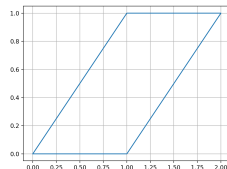
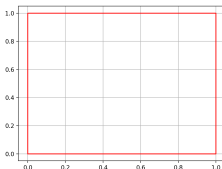
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$$\mathcal{SH}(\psi) = \{\psi_{j,k,m}(x) = 2^{3j/4} \psi(S_k A_j x - m) : (j, k) \in \mathbb{Z}^2, m \in \mathbb{Z}^2\}$$

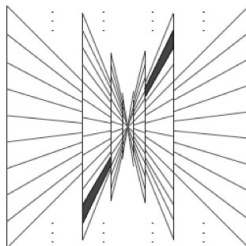
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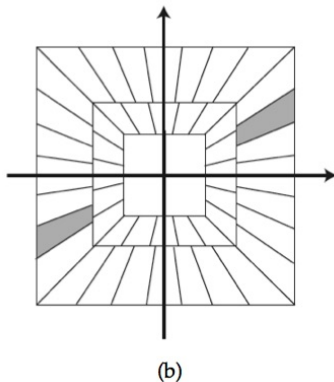
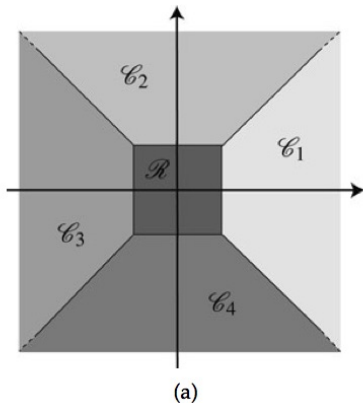


Cone-based shearlet transform

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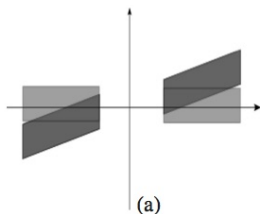
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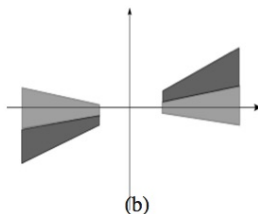
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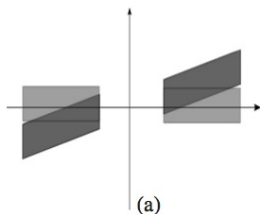
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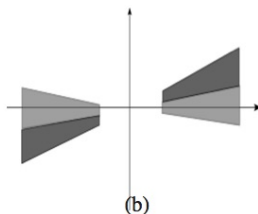
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► Best N -term approximation error

$$\sigma_N(f, \{\psi_{j,k,m}\}_{j,k,m}) \sim N^{-1}(\log(N))^{3/2}$$

► Matlab

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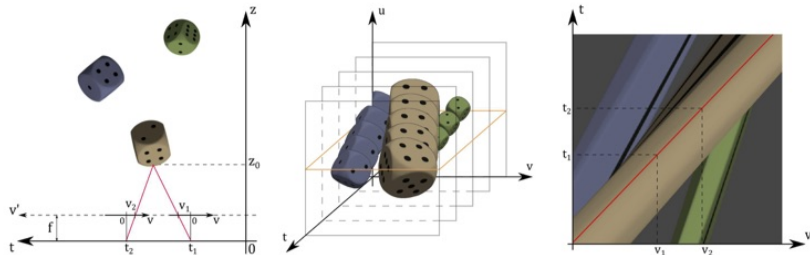
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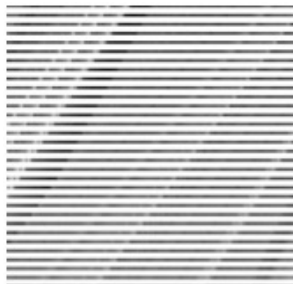
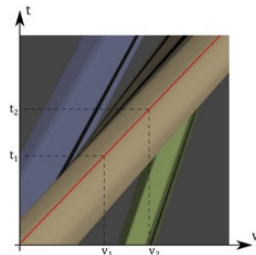
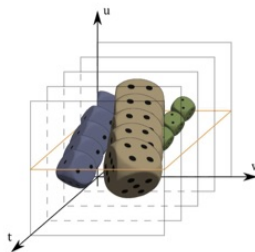
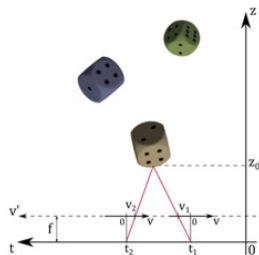
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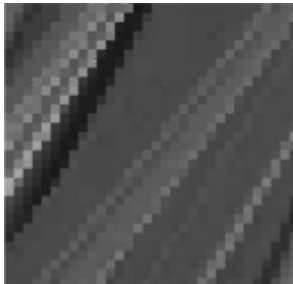
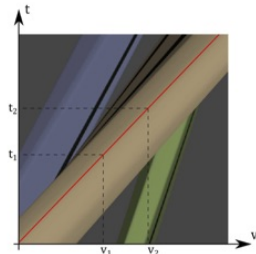
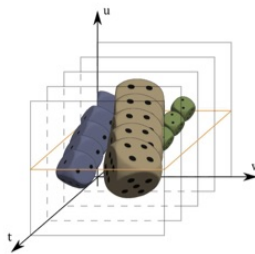
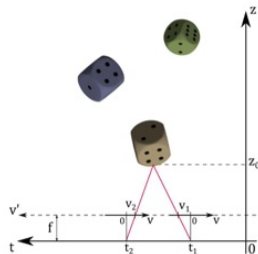
Current application: Light Field Recovery



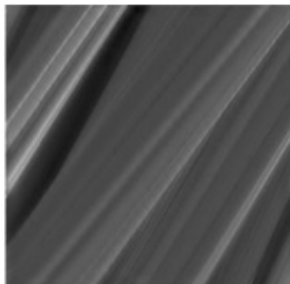
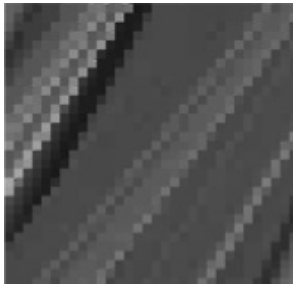
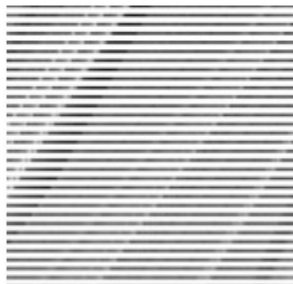
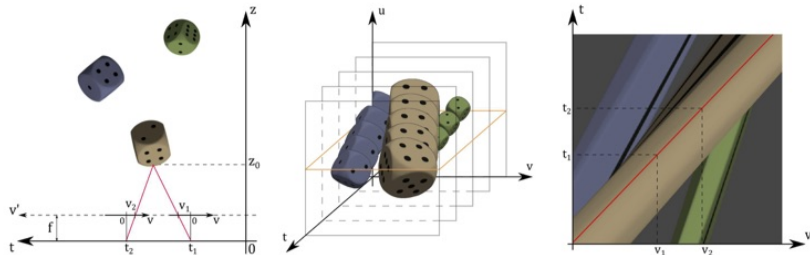
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Thanks!

Questions?

