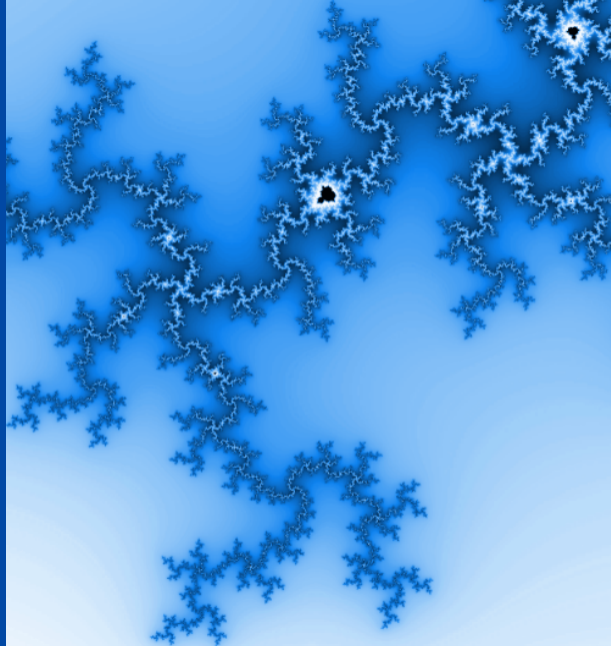


Wavelet representations for stochastic processes defined by multiple Wiener-Itô integrals

Young Scholar Day of the Belgian Mathematical Society

Laurent Loosveldt

20th December 2023



Definition - Step 1

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$$f = \sum_{j_1, \dots, j_d=1}^n a_{j_1, \dots, j_d} \mathbb{1}_{[s_{j_1}, t_{j_1})} \otimes \cdots \otimes \mathbb{1}_{[s_{j_d}, t_{j_d})}, \quad (1)$$

where, \otimes stands for the tensor product, a_{j_1, \dots, j_d} are such that, for all permutation σ , $a_{\sigma(j_1), \dots, \sigma(j_d)} = a_{j_1, \dots, j_d}$ and $a_{j_1, \dots, j_d} = 0$ as soon as two indices j_1, \dots, j_d are equal and, for all $1 \leq \ell \neq \ell' \leq d$, $[s_{j_\ell}, t_{j_\ell}) \cap [s_{j_{\ell'}}, t_{j_{\ell'}}) = \emptyset$.

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$$I_d(f) := \sum_{j_1, \dots, j_d=1}^n a_{j_1, \dots, j_d} (B(t_{j_1}) - B(s_{j_1})) \times \dots (B(t_{j_d}) - B(s_{j_d})). \quad (2)$$

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It is a random variable in $L^2(\Omega)$.

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- ▶ $(I_d(f_j))_j$ is a Cauchy sequence in $L^2(\Omega)$;
- ▶ $I_d(f) := \lim_{j \rightarrow +\infty} I_d(f_j)$ (in $L^2(\Omega)$).

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We use the canonical symmetrization \tilde{f} of f defined, for all $(x_1, \dots, x_d) \in \mathbb{R}^d$, as:

$$\tilde{f}(x_1, \dots, x_d) := \frac{1}{d!} \sum_{\sigma \in \tilde{\mathfrak{S}}_d} f(x_{\sigma(1)}, \dots, x_{\sigma(d)}). \quad (3)$$

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The d th Wiener chaos is defined as the closed linear subspace of $L^2(\Omega)$ generated by the random variables of the form $I_d(f)$

Important facts

Wiener isometry

If $f \in L^2(\mathbb{R}^d)$ and $g \in L^2(\mathbb{R}^{d'})$, we have

$$\mathbb{E}[I_d(f)I_{d'}(g)] = \begin{cases} d!\langle f, g \rangle & \text{if } d = d' \\ 0 & \text{otherwise,} \end{cases} \quad (4)$$

where $\langle \cdot, \cdot \rangle$ stands for the canonical scalar product in $L^2(\mathbb{R}^d)$.

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Connection with Hermite polynomials

if $f \in L^2(\mathbb{R})$ is such that $\|f\|_{L^2(\mathbb{R})} = 1$, then

$$H_d(I_1(f)) = I_d(f^{\otimes d}),$$

where H_d is the d th Hermite polynomial $H_d(x) = (-1)^d e^{x^2/2} D^d e^{-x^2/2}$.

Important facts

Product formula

If $f \in L^2(\mathbb{R}^m)$ and $g \in L^2(\mathbb{R}^n)$, we have

$$I_m(f)I_n(g) = \sum_{r=0}^{m \wedge n} r! \binom{m}{r} \binom{n}{r} I_{m+n-2r}(f \otimes_r g), \quad (5)$$

where, for all $0 \leq r \leq m \wedge n$, $f \otimes_r g$ is the $L^2(\mathbb{R}^{m+n-2r})$ function defined, for all $(x_1, \dots, x_{m+n-2r}) \in \mathbb{R}^{m+n-2r}$, through the Lebesgue integral

$$\begin{aligned} & (f \otimes_r g)(x_1, \dots, x_{m+n-2r}) \\ & := \int_{\mathbb{R}^r} f(x_1, \dots, x_{m-r}, s_1, \dots, s_r) g(x_{m-r+1}, \dots, x_{m+n-2r}, s_1, \dots, s_r) ds_1 \dots ds_r, \end{aligned}$$

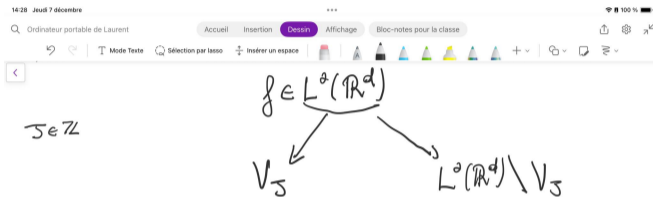
Multiresolution analysis

Definition

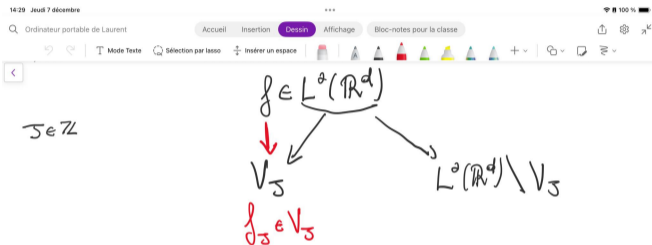
A multiresolution analysis of the Hilbert space $L^2(\mathbb{R}^d)$ is given by a sequence $(V_j)_{j \in \mathbb{Z}}$ of closed linear subspaces of $L^2(\mathbb{R}^d)$ such that

- (a) for all $j \in \mathbb{Z}$, $V_j \subseteq V_{j+1}$;
- (b) $\bigcap_{j \in \mathbb{Z}} V_j = \{0\}$ and $\bigcup_{j \in \mathbb{Z}} V_j$ is dense in $L^2(\mathbb{R}^d)$;
- (c) for all $j \in \mathbb{Z}$, $V_j = \{f(2^j \cdot) : f \in V_0\}$;
- (d) there exists a scaling function $\phi^d \in V_0$ such that the sequence $(\phi^d(\cdot - k))_{k \in \mathbb{Z}^d}$ is an orthogonal basis of V_0 .

Approximation strategy



Approximation strategy



Approximation strategy

14:32 Jeudi 7 décembre

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Mode Texte Sélection par lasso Insérer un espace

100%

$J \in \mathbb{Z}$

$f \in L^2(\mathbb{R}^d)$

V_J

$L^2(\mathbb{R}^d) \setminus V_J$

$f_J \in V_J$

$f - f_J$

$f_J = \sum_{h \in \mathbb{Z}^d} \langle f, 2^{J\frac{d}{2}} \phi^d(2^J \cdot - h) \rangle 2^{J\frac{d}{2}} \phi^d(2^J \cdot - h)$

Approximation strategy

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$\forall j \dots \subseteq V_{j-1} \subseteq V_j \subseteq V_{j+1} \subseteq \dots$

Approximation strategy

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$V_J \dots \subseteq V_{J-1} \subseteq V_J \subseteq V_{J+1} \subseteq \dots$

$V_{J+1} = V_J \oplus W_J$

Approximation strategy

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Mode Texte Sélection par lasso Insérer un espace

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$L^2(\mathbb{R}^d) \setminus V_J = \bigoplus_{j \geq J} W_j$

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A base for the details spaces

In dimension 1 (Meyer)

There exists a function ψ , called mother wavelet, belonging to W_0^1 and such that, for all $j \in \mathbb{Z}$, the sequence $(2^{j/2}\psi(2^j \cdot - k))_{k \in \mathbb{Z}}$ is an orthonormal basis in W_j^1 . Moreover, for all $J \in \mathbb{Z}$, the family

$$\{2^{J/2}\phi^1(2^J x - k) : k \in \mathbb{Z}\} \cup \{2^{j/2}\psi(2^j x - k) : k \in \mathbb{Z}, j \geq J\}$$

is a base in $L^2(\mathbb{R})$, called wavelet base.

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In higher dimension: tensor products

Usually, we use $\{2^{J\frac{d}{2}}\Phi^d(2^J x - k) : k \in \mathbb{Z}^d\} \cup \{2^{j\frac{d}{2}}\Psi(2^j x - k) : k \in \mathbb{Z}^d, j \geq J\}$ with Φ^d , d -tensor products of ϕ with itself and Ψ , d -tensor products of ϕ and ψ where at least one term is ψ .

A base for the details spaces

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Better in our context

$$\{2^{J\frac{d}{2}}\Phi^d(2^J x - k) : k \in \mathbb{Z}^d\} \cup \left\{ \prod_{\ell=1}^d 2^{\frac{j_\ell}{2}} \psi(2^{j_\ell} x_\ell - k_\ell) : k \in \mathbb{Z}^d, \max_{1 \leq \ell \leq d} j_\ell \geq J \right\}$$

Processes in the d th Wiener chaos

General definition

$$\{I_d(K(t, \bullet))\}_{t \geq 0}$$

where, for all $t \geq 0$, the function $K(t, \bullet) \in L^2(\mathbb{R}^d)$.

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Hermite process

$$K(t, x_1, \dots, x_d) = c(h, d) \int_0^t \prod_{\ell=1}^d (s - x_\ell)_+^{\frac{h-1}{d} - \frac{1}{2}} ds$$

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Generalized Hermite processes (Bai, Taqqu)

$$K(t, x_1, \dots, x_d) = \frac{1}{\prod_{\ell=1}^d \Gamma(h_\ell - 1/2)} \int_0^t \prod_{j=1}^d (s - x_j)_+^{h_j - 3/2} ds$$

Expansion strategy

18:03 Jeudi 7 décembre

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Mode Texte Sélection par lasso Insérer un espace

$X(t) = \text{Id}(K(t, \bullet))$

Expansion strategy

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Mode Texte Sélection par lasso Insérer un espace

100%

$\exists \in \mathbb{Z}$

$$X(t) = \text{Id}_d(K(t, \cdot))$$

$$K_{\exists}(t, \cdot) + K_{\exists}^{\perp}(t, \cdot)$$

\uparrow
 V_{\exists}

\uparrow
 $\bigoplus_{j=1,2,3} W_j$

Expansion strategy

18:08 Jeudi 7 décembre

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100%

$\exists \varepsilon \in \mathbb{Z}$

$$X(t) = \text{Id} \left(K(t, \cdot) \right)$$

$$K_{\varepsilon}(t, \cdot) + K_{\varepsilon}^{\perp}(t, \cdot)$$

\uparrow \uparrow

V_{ε} $\oplus_{j=1,2,3} W_j$

$$X_{\varepsilon}(t) = \text{Id} \left(K_{\varepsilon}(t, \cdot) \right)$$

Expansion strategy

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100%

$$X(t) = \text{Id}(K(t, \cdot))$$

$$K_{\mathfrak{J}}(t, \cdot) + K_{\mathfrak{J}^{\perp}}(t, \cdot)$$

$$\uparrow \quad \uparrow$$

$$V_{\mathfrak{J}} \quad \bigoplus_{j \in \mathfrak{J}} W_j$$

$$X_{\mathfrak{J}}(t) = \text{Id}(K_{\mathfrak{J}}(t, \cdot))$$

Approximation process

Expansion strategy

18:09 Jeudi 7 décembre

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$$X(t) = \text{Id} (K(t, \cdot))$$

$$K_J(t, \cdot) + K_J^\perp(t, \cdot)$$

$$\underbrace{K_J(t, \cdot)}_{V_J} \quad \oplus_{j=1,2,3} W_j$$

$$X_J(t) = \text{Id} (K_J(t, \cdot))$$

Approximation process

$$\forall t, X_J(t) \xrightarrow{L^2(\Omega)} X(t) \text{ if } J \rightarrow +\infty$$

Expansion strategy

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100%

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$$X(t) = \text{Id}(K(t, \cdot))$$

$$K_{\varepsilon}(t, \cdot) + K_{\varepsilon}^{\perp}(t, \cdot)$$

\uparrow V_{ε} \uparrow $\bigoplus_{j=1,2,3} W_j$

$$X_{\varepsilon}(t) = \text{Id}(K_{\varepsilon}(t, \cdot)) \quad X_{\varepsilon}^{\perp}(t) = \text{Id}(K_{\varepsilon}^{\perp}(t, \cdot))$$

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$$K_J(t, \cdot) + K_J^\perp(t, \cdot)$$

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$$X_J^\perp(t) = \text{Id}(K_J^\perp(t, \cdot))$$

Approximative process
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Details process

Expansion strategy

16:10 Jeudi 7 décembre

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$$K_{\varepsilon}(t, \cdot) + K_{\varepsilon}^{\perp}(t, \cdot)$$

\uparrow
 V_{ε}

\uparrow
 $\bigoplus_{j=1,2,3} W_j$

$$X_{\varepsilon}(t) = \text{Id} \left(K_{\varepsilon}(t, \cdot) \right)$$

Approximation process

$$\forall t, X_{\varepsilon}(t) \xrightarrow{L^2(\Omega)} X(t) \text{ if } \varepsilon \rightarrow +\infty$$

$$X^{\perp}_{\varepsilon}(t) = \text{Id} \left(K_{\varepsilon}^{\perp}(t, \cdot) \right)$$

Details process

$$\forall t, X(t) - X_{\varepsilon}(t) = X^{\perp}_{\varepsilon}(t)$$

Application to generalized Hermite processes

We want more!

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- ▶ More explicit expression

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$$\text{for all } t \geq 0 \quad K_J^\perp(t, \bullet) = \sum_{\substack{(\mathbf{j}, \mathbf{k}) \in (\mathbb{Z}^d)^2 \\ \max_{\ell \in \llbracket 1, d \rrbracket} j_\ell \geq J}} \langle K(t, \bullet), \bigotimes_{\ell=1}^d \psi_{j_\ell, k_\ell} \rangle \bigotimes_{\ell=1}^d \psi_{j_\ell, k_\ell}$$

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- ▶ Uniform convergence on compact set for the approximation process, with rate of convergence.

$$\text{for all } t \geq 0 \quad K_J^\perp(t, \bullet) = \sum_{\substack{(\mathbf{j}, \mathbf{k}) \in (\mathbb{Z}^d)^2 \\ \max_{\ell \in \llbracket 1, d \rrbracket} j_\ell \geq J}} \langle K(t, \bullet), \bigotimes_{\ell=1}^d \psi_{j_\ell, k_\ell} \rangle \bigotimes_{\ell=1}^d \psi_{j_\ell, k_\ell}$$

$$\text{for all } t \geq 0 \quad X_J^\perp(t) = \sum_{\substack{(\mathbf{j}, \mathbf{k}) \in (\mathbb{Z}^d)^2 \\ \max_{\ell \in \llbracket 1, d \rrbracket} j_\ell \geq J}} \langle K(t, \bullet), \bigotimes_{\ell=1}^d \psi_{j_\ell, k_\ell} \rangle I_d \left(\bigotimes_{\ell=1}^d \psi_{j_\ell, k_\ell} \right) \text{ in } L^2(\Omega).$$

Explicit expression of the details process

$$\langle K(t, \bullet), \bigotimes_{\ell=1}^d \psi_{j_\ell, k_\ell} \rangle = 2^{j_1(1-h_1)+\dots+j_d(1-h_d)} \int_0^t \prod_{\ell=1}^d \psi_{h_\ell}(2^{j_\ell} s - k_\ell) ds$$

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$$I_d \left(\bigotimes_{\ell=1}^d \psi_{j_\ell, k_\ell} \right) = \prod_{\ell=1}^p H_{n_\ell} \left(I_1 \left(2^{j_\ell/2} \psi(2^{j_\ell} \cdot - k_\ell) \right) \right)$$

where n_ℓ is the multiplicity of (j_ℓ, k_ℓ) in (\mathbf{j}, \mathbf{k}) .

Application to generalized Hermite processes

Tools for the approximation process

- ▶ fractional scaling function :

$$\widehat{\Phi}_{\Delta}^{(\delta)}(\xi) = \left(\frac{1 - e^{-i\xi}}{i\xi} \right)^{\delta} \widehat{\phi}(\xi) \quad \forall \xi \neq 0 \quad \text{and} \quad \widehat{\Phi}_{\Delta}^{(\delta)}(0) = 1$$

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$$\mu_{J,\mathbf{k}} = \sum_{m=0}^{\lfloor d/2 \rfloor} (-1)^m \sum_{P \in \mathcal{P}_m^{(d)}} \prod_{r=1}^m \mathbb{E}[g_{J,k_{l_r}} g_{J,k_{l'_r}}] \prod_{s=m+1}^{d-m} g_{J,k_{l'_s}} \quad \text{with} \quad g_{J,k} = I_1(2^{J/2} \phi(2^J \dots - k))$$

Tools for the approximation process

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The approximation process can be expressed, for all $t \in \mathbb{R}_+$, as:

$$X_J(t) = 2^{-J(h_1 + \dots + h_d - d)} \sum_{\mathbf{k} \in \mathbb{Z}^d} \left(\int_0^t \prod_{\ell=1}^d \Phi_{\Delta}^{(h_{\ell} - 1/2)}(2^J s - k_{\ell}) ds \right) \sigma_{J, \mathbf{k}}^{(\mathbf{h})}, \quad (6)$$

where the series is convergent in $L^2(\Omega)$. Moreover this series is also almost surely uniformly convergent in t on each compact interval of \mathbb{R}_+ .

Rate of convergence

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For any compact interval $I \subset \mathbb{R}_+$, there exists an almost surely finite random variable C (depending on I) for which one has, almost surely, for each $J \in \mathbb{N}$,

$$\|X - X_J\|_{I,\infty} = \|X_{\mathbf{h},J}^{(d,\perp)}\|_{I,\infty} \leq C J^{\frac{d}{2}} 2^{-J(h_1 + \dots + h_d - d + 1/2)}. \quad (7)$$

Towards numerical simulation

In the proof of the last Theorem, we notice that the rate of convergence is mainly governed by the terms

$$2^{j_1(1-h_1)+\dots+j_d(1-h_d)} I_d \left(\bigotimes_{\ell=1}^d \psi_{j_\ell, k_\ell} \right) \int_0^t \prod_{\ell=1}^d \psi_{h_\ell}(2^{j_\ell} s - k_\ell) ds$$

for which there is $\ell \in \llbracket 1, d \rrbracket$ and

$$k_\ell \in D_j^1(t) := \{k \in \mathbb{Z} : [k2^{-j} - 2^{-ja}, k2^{-j} + 2^{-ja}] \subseteq [0, t]\},$$

with $\frac{1}{2} < a < 1$.

Towards numerical simulation

Definition

$$\mathcal{I}_J^1(t) := \{\mathbf{k} \in (D_J^1(t))^d : \max_{\ell, \ell' \in \llbracket 1, d \rrbracket} |k_\ell - k_{\ell'}| \leq 2^{\varepsilon J}\},$$

Towards numerical simulation

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$$\mathcal{I}_J^1(t) := \{\mathbf{k} \in (D_J^1(t))^d : \max_{\ell, \ell' \in \llbracket 1, d \rrbracket} |k_\ell - k_{\ell'}| \leq 2^{\varepsilon J}\},$$

The simulation process at scale J

$$S_J(t) = 2^{-J(h_1 + \dots + h_d - d + 1)} \sum_{\mathbf{k} \in \mathcal{I}_J^1(t)} \sigma_{J, \mathbf{k}}^{(\mathbf{h})} \int_{\mathbb{R}} \prod_{\ell=1}^d \Phi_{\Delta}^{(h_\ell - 1/2)}(s - k_\ell) ds.$$

Towards numerical simulation

A.Ayache, J.Hamonier, L.L.

For any compact interval $I \subset \mathbb{R}_+$, there exists an almost surely finite random variable C (depending on I) for which one has, almost surely, for each $J \in \mathbb{N}$,

$$\|X - S_J\|_{I, \infty} \leq C J^{\frac{d}{2}} 2^{-J(h_1 + \dots + h_d - d + 1/2)}. \quad (8)$$