

Wavelet methods

For the numerical simulation of incompressible fluids

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- **Wavelets for the Navier-Stokes equations**

$$\partial_t \mathbf{u} + (\mathbf{u} \cdot \nabla) \mathbf{u} - \nu \Delta \mathbf{u} + \nabla p = \mathbf{f} \quad , \quad \operatorname{div}(\mathbf{u}) = 0$$

Decomposition of \mathbf{u} in a Riesz basis of $\mathbf{H}_{div,0}$ [Urban96]:

$$\mathbf{u} = \sum_{k \in \mathbb{Z}^d} c_k(t) \phi_k^{df} + \sum_{j \geq 0} \sum_{k \in \mathbb{Z}^d} d_{j,k}(t) \psi_{j,k}^{df}$$

The Navier Stokes equations can be projected on $\mathbf{H}_{div,0}$:

$$\partial_t \mathbf{u} + \mathbb{P}((\mathbf{u} \cdot \nabla) \mathbf{u}) - \nu \Delta \mathbf{u} = \mathbb{P}(\mathbf{f})$$

where we noted \mathbb{P} the Leray projector.

• **Perspectives: Numerical resolution** of incompressible Navier-Stokes equations in dimension 2 and 3 by an adaptive method:

$$\left\{ \begin{array}{l} \frac{\partial \mathbf{u}}{\partial t} + \mathbf{u} \cdot \nabla \mathbf{u} + \nabla p = \nu \Delta \mathbf{u} + f \quad t \geq 0, x \in]0, 1[^n \quad n = 2 \text{ or } 3 \\ \operatorname{div} \mathbf{u} = \nabla \cdot \mathbf{u} = \sum_{i=1}^n \frac{\partial u_i}{\partial x_i} = 0 \\ \mathbf{u}(x, 0) = \mathbf{u}^0(x) \\ + \text{limit conditions (periodic, Dirichlet homogeneous or non homogeneous)} \end{array} \right.$$

With a wavelet discretization:

$$\mathbf{u}(x, t^n) \approx \mathbf{u}_N(x, t^n) = \sum_{\alpha \in A_n} c_\alpha^n \Psi_\alpha(x)$$

with $\operatorname{Card}(A_n) = N$ and $\Psi_\alpha \in \mathbf{H}_{\operatorname{div},0} = \{\mathbf{u} \in L^2 / \operatorname{div}(\mathbf{u}) = 0\}$.

● **Plan :**

I - Divergence-free wavelets

II - Helmholtz decomposition with wavelets

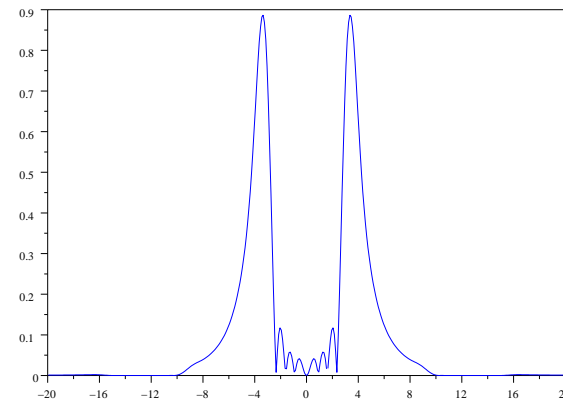
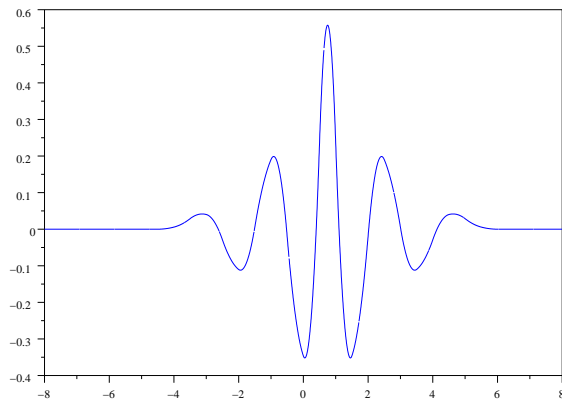
III - Numerical simulations, results

Conclusion - Perspectives

Wavelets

[Meyer90, Daubechies92]

Family of functions $\{\psi_\omega\}_{\omega \in \Omega}$ that forms a basis (sometimes orthogonal) of a function space ($L^2(\mathbb{R})$ for example).



Example of a wavelet with its frequency localisation.

Multiresolution Analysis

Definition [Mallat87]: A Multiresolution Analysis of $L^2(\mathbb{R})$ is defined as a sequence of closed sub-spaces $(V_j)_{j \in \mathbb{Z}}$ verifying :

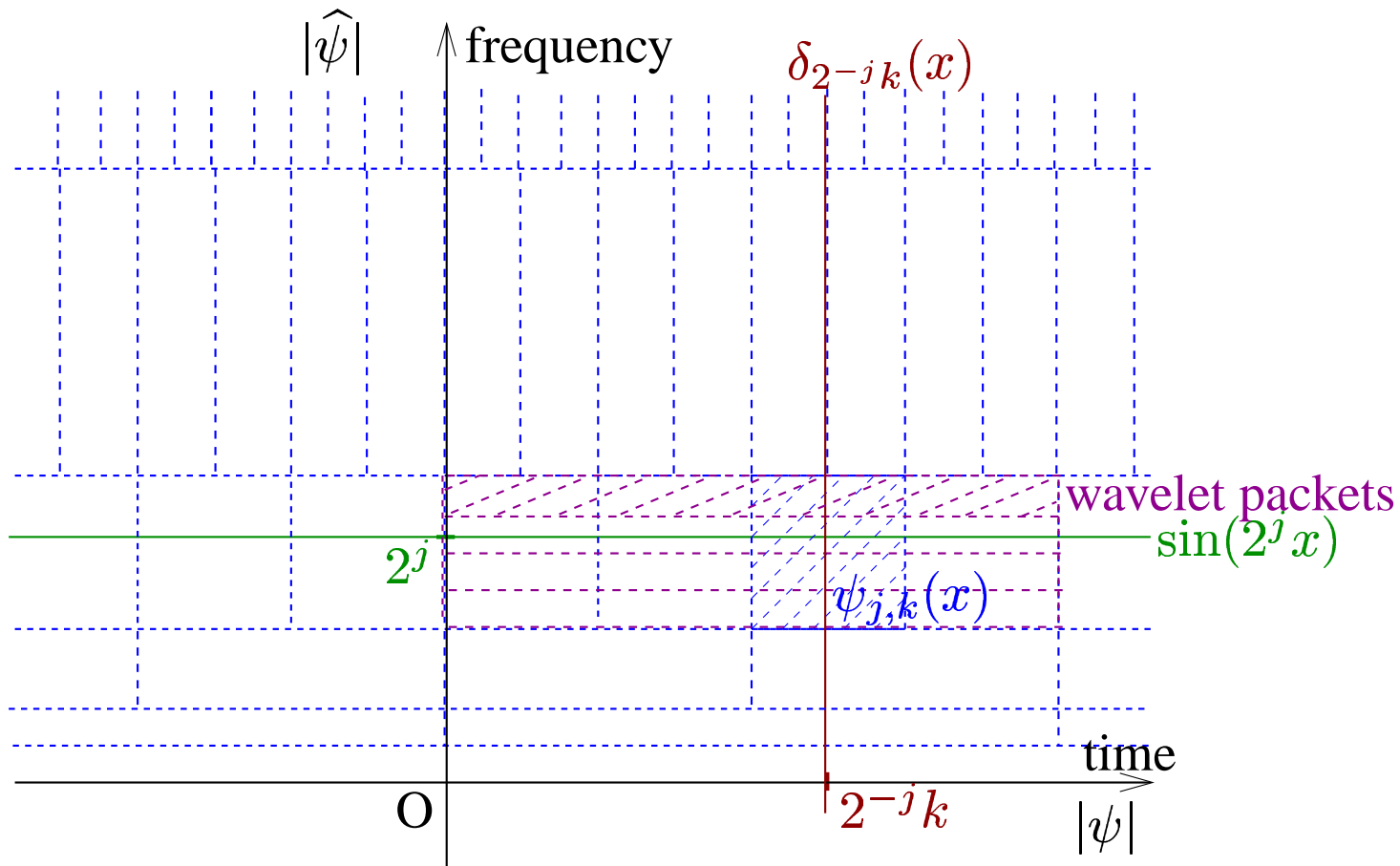
(1) $V_j \subset V_{j+1}, \quad \bigcap_{j \in \mathbb{Z}} V_j = \{0\} \quad \overline{\bigcup_{j \in \mathbb{Z}} V_j} = L^2(\mathbb{R})$

(2) $f \in V_j \iff f(2 \cdot) \in V_{j+1}$ (dilation)

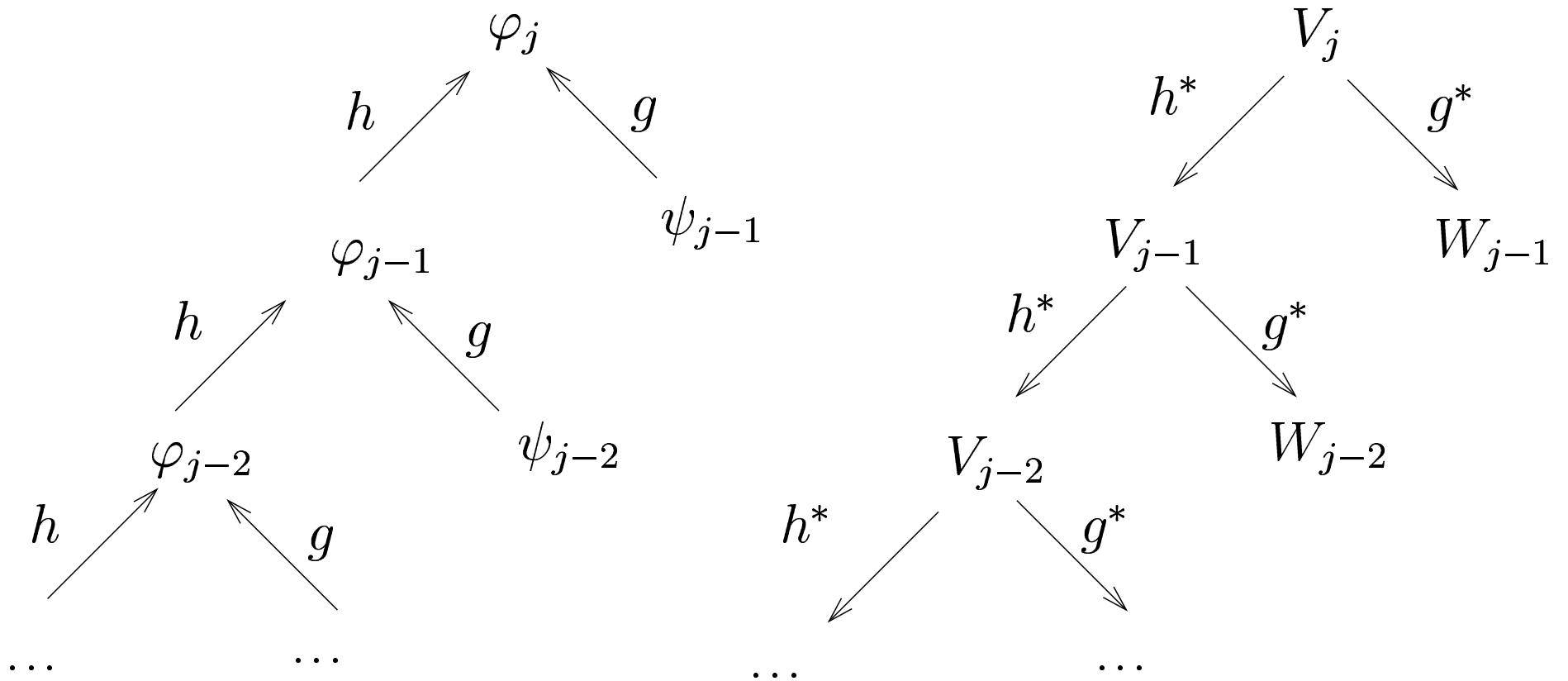
(3) $\exists \varphi \quad / \quad \{\varphi(\cdot - k)\}_{k \in \mathbb{Z}}$ Riesz basis of V_0 .

Wavelet space W_j defined by : $V_{j+1} = V_j \oplus^\perp W_j$.

Time-frequency partition with wavelets



Filtering Schema: decomposition – recomposition



Divergence-free wavelets

Proposition (Malgouyres): [Lemarié92] Let (φ_1, ψ_1) be an MRA. If $\varphi_1 \in C^{1+\epsilon}$ for a certain $\epsilon > 0$, then there is an MRA (φ_0, ψ_0) such that:

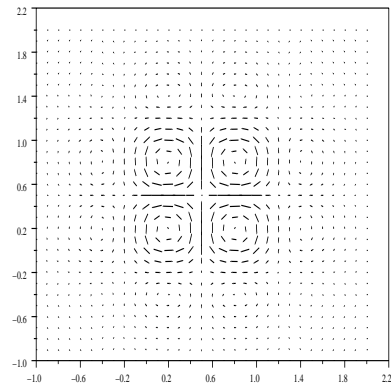
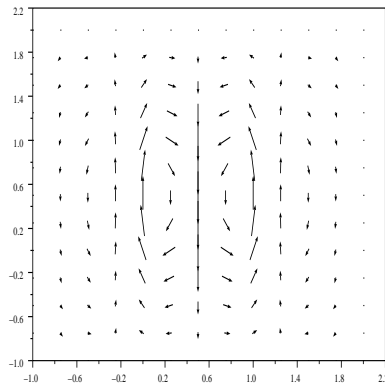
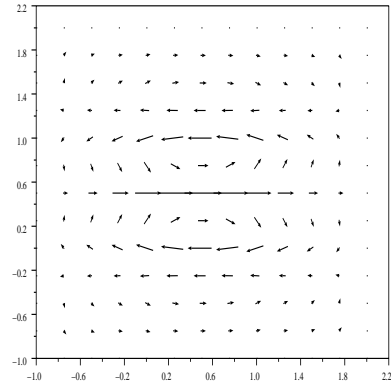
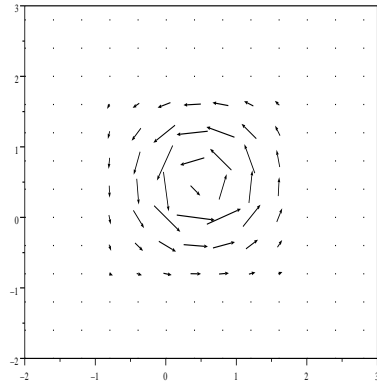
$$\varphi_1'(x) = \varphi_0(x) - \varphi_0(x - 1)$$

wavelets : $\psi_0(x) = \frac{1}{4}\psi_1'(x)$

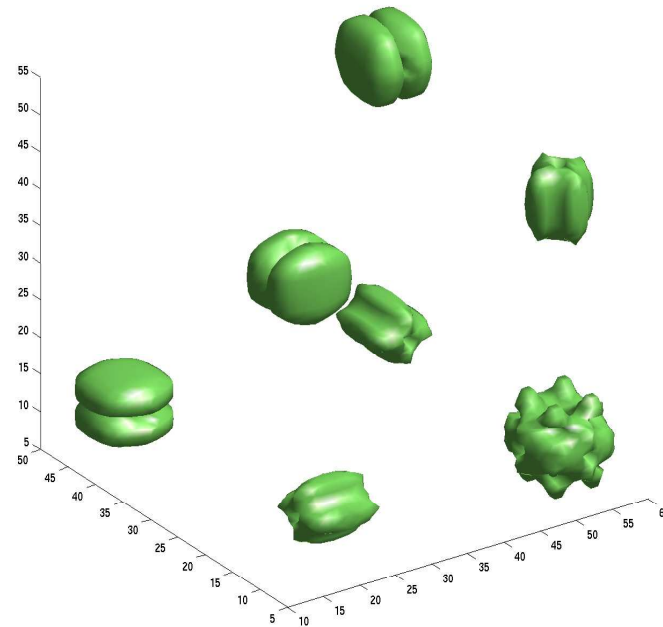
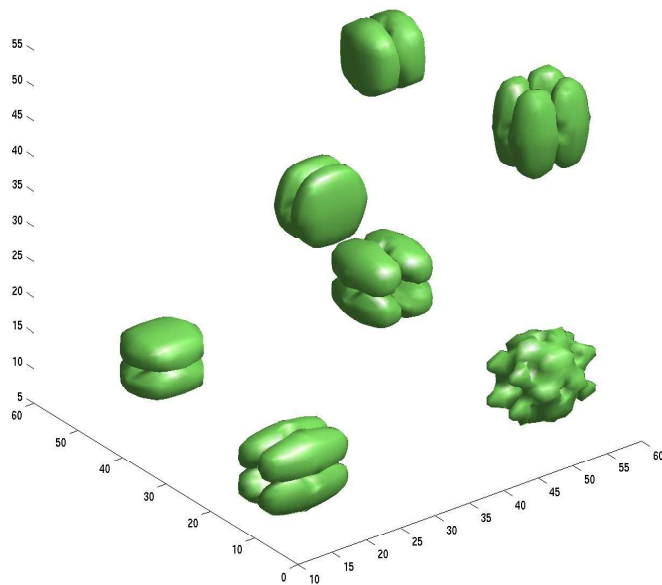
Theorem: There exist $(d - 1)(2^d - 1)$ vectorial functions $\overrightarrow{\Gamma_{\omega, i}} = (\Gamma_{\omega, i, 1}; \dots; \Gamma_{\omega, i, d})$, $\omega \in E^*$ and $i \neq i_0$ that, when translated and dilated, form an inconditionnelle basis of $\mathbf{H}_{div, 0}$.

$(E = \{0, 1\}^d, \Gamma$ constructed by tensor products of $\varphi_0, \varphi_1, \psi_0$ and $\psi_1)$

Example of divergence-free wavelets in 2D



Example of divergence-free wavelets in 3D



Vorticity isosurfaces of the 3D isotropic divergence-free wavelets

Divergence-free wavelet transform

$$\Psi_{df}^{(1,1)}(x_1, x_2) = \begin{vmatrix} \psi_1(x_1)\psi_0(x_2) \\ -\psi_0(x_1)\psi_1(x_2) \end{vmatrix} \quad \Psi_{cf}^{(1,1)}(x_1, x_2) = \begin{vmatrix} \psi_1(x_1)\psi_0(x_2) \\ \psi_0(x_1)\psi_1(x_2) \end{vmatrix}$$

$$\begin{array}{ccc} \text{Standard wavelet transform} & & \\ u_1 & \longrightarrow & d_{1j,\mathbf{k}}^\varepsilon \\ u_2 & \longrightarrow & d_{2j,\mathbf{k}}^\varepsilon \end{array} \left. \vphantom{\begin{array}{ccc} \text{Standard wavelet transform} \\ u_1 \\ u_2 \end{array}} \right\} \begin{array}{ccc} \text{linear combinations} & & \\ & \longrightarrow & d_{dfj,\mathbf{k}}^\varepsilon \\ & & d_{cfj,\mathbf{k}}^\varepsilon \end{array}$$

Anisotropic divergence-free wavelets

$$\Psi_{\mathbf{j},\mathbf{k}}^{\text{div}}(x_1, x_2) = \begin{cases} 2^{j_2} \psi_1(2^{j_1} x_1 - k_1) \psi_0(2^{j_2} x_2 - k_2) \\ -2^{j_1} \psi_0(2^{j_1} x_1 - k_1) \psi_1(2^{j_2} x_2 - k_2) \end{cases}$$

with $\mathbf{j} = (j_1, j_2) \in \mathbb{Z}^2$ the scale and $\mathbf{k} = (k_1, k_2) \in \mathbb{Z}^2$ the position.

The operation on the coefficients:

$$\begin{bmatrix} d_{\mathbf{j},\mathbf{k}}^{\text{div}} \\ d_{\mathbf{j},\mathbf{k}}^{\text{n}} \end{bmatrix} = \frac{1}{2^{2j_1} + 2^{2j_2}} \begin{bmatrix} 2^{j_2} & -2^{j_1} \\ 2^{j_1} & 2^{j_2} \end{bmatrix} \begin{bmatrix} d_{1\mathbf{j},\mathbf{k}} \\ d_{2\mathbf{j},\mathbf{k}} \end{bmatrix}$$

is an orthogonal basis change (orthogonal matrix).

Anisotropic divergence-free wavelets in n -D:

$$\Psi_{i, \mathbf{j}, \mathbf{k}}^{\text{div}}(x_1, \dots, x_n) = \begin{cases} -2^{j_i + j_1} \prod_{\ell} \psi_{\delta_{1, \ell}}(2^{j_\ell} x_\ell - k_\ell) \\ \vdots \\ \left(\sum_{\ell \neq i} 2^{2j_\ell} \right) \prod_{\ell} \psi_{\delta_{i, \ell}}(2^{j_\ell} x_\ell - k_\ell) \\ \vdots \\ -2^{j_i + j_n} \prod_{\ell} \psi_{\delta_{n, \ell}}(2^{j_\ell} x_\ell - k_\ell) \end{cases}$$

$$\Psi_{i, \mathbf{j}, \mathbf{k}}^n(x_1, \dots, x_n) = \begin{cases} 2^{j_1} \prod_{\ell} \psi_{\delta_{1, \ell}}(2^{j_\ell} x_\ell - k_\ell) \\ \vdots \\ 2^{j_n} \prod_{\ell} \psi_{\delta_{n, \ell}}(2^{j_\ell} x_\ell - k_\ell) \end{cases}$$

Let $\xi_\ell = 2^{j_\ell}$ and $|\xi|^2 = \sum_{\ell=1}^n 2^{2j_\ell}$,

$$\begin{bmatrix} d_{1\mathbf{j},\mathbf{k}}^{\text{div}} \\ d_{2\mathbf{j},\mathbf{k}}^{\text{div}} \\ \vdots \\ \vdots \\ d_{n\mathbf{j},\mathbf{k}}^{\text{div}} \\ d_{\mathbf{j},\mathbf{k}}^{\text{n}} \end{bmatrix} = \begin{bmatrix} 1 - \frac{\xi_1^2}{|\xi|^2} & -\frac{\xi_2\xi_1}{|\xi|^2} & \cdots & \cdots & -\frac{\xi_n\xi_1}{|\xi|^2} & \frac{\xi_1}{|\xi|} \\ -\frac{\xi_1\xi_2}{|\xi|^2} & 1 - \frac{\xi_2^2}{|\xi|^2} & \ddots & \ddots & -\frac{\xi_n\xi_2}{|\xi|^2} & \frac{\xi_2}{|\xi|} \\ \vdots & \vdots & \ddots & \ddots & \vdots & \vdots \\ \vdots & \vdots & \ddots & \ddots & \vdots & \vdots \\ -\frac{\xi_1\xi_n}{|\xi|^2} & -\frac{\xi_2\xi_n}{|\xi|^2} & \cdots & -\frac{\xi_{n-1}\xi_n}{|\xi|^2} & 1 - \frac{\xi_n^2}{|\xi|^2} & \frac{\xi_n}{|\xi|} \\ \frac{\xi_1}{|\xi|} & \frac{\xi_2}{|\xi|} & \cdots & \frac{\xi_{n-1}}{|\xi|} & \frac{\xi_n}{|\xi|} & 0 \end{bmatrix} \begin{bmatrix} d_{1\mathbf{j},\mathbf{k}} \\ d_{2\mathbf{j},\mathbf{k}} \\ \vdots \\ \vdots \\ d_{n\mathbf{j},\mathbf{k}} \\ 0 \end{bmatrix}$$

Matrix of size $(n + 1) \times (n + 1)$, orthogonal.

II - Helmholtz decomposition

Principal

Vector field $\mathbf{u} \in (L^2(\mathbb{R}^n))^n$, decomposition with

$$\mathbf{u} = \mathbf{u}_{\text{div}} + \mathbf{u}_{\text{curl}} \quad \text{where} \quad \mathbf{u}_{\text{div}} = \mathbf{curl} \psi \quad \mathbf{u}_{\text{curl}} = \nabla p$$

the functions $\mathbf{curl} \psi$ and ∇p are orthogonal in $(L^2(\mathbb{R}^n))^n$ and we have uniqueness.

$$(L^2(\mathbb{R}^n))^n = \mathbf{H}_{\text{div } 0}(\mathbb{R}^n) \oplus^\perp \mathbf{H}_{\text{curl}, 0}(\mathbb{R}^n)$$

In N-S, importance of this decomposition to project the term $\mathbf{u} \cdot \nabla \mathbf{u}$ onto $\mathbf{H}_{\text{div } 0}(\mathbb{R}^n)$.

Leray projector (in Fourier)

$$\mathbf{u} = \mathbf{u}_{\text{div}} + \nabla p$$

$$\Delta p = \text{div} \mathbf{u} = \partial_1 u_1 + \dots + \partial_n u_n$$

In Fourier,

$$\hat{p} = -\frac{i}{|\xi|^2} \sum_{l=1}^n \xi_l \hat{u}_l$$

et

$$\hat{\mathbf{u}}_{\text{div}} = \hat{\mathbf{u}} - \begin{bmatrix} \xi_1 \sum_{l=1}^n \xi_l \hat{u}_l \\ \xi_2 \sum_{l=1}^n \xi_l \hat{u}_l \\ \vdots \\ \xi_n \sum_{l=1}^n \xi_l \hat{u}_l \end{bmatrix}$$

Wavelet Helmholtz decomposition

We want to write:

$$\mathbf{v} = \mathbf{v}_{\text{div}} + \mathbf{v}_{\text{curl}}$$

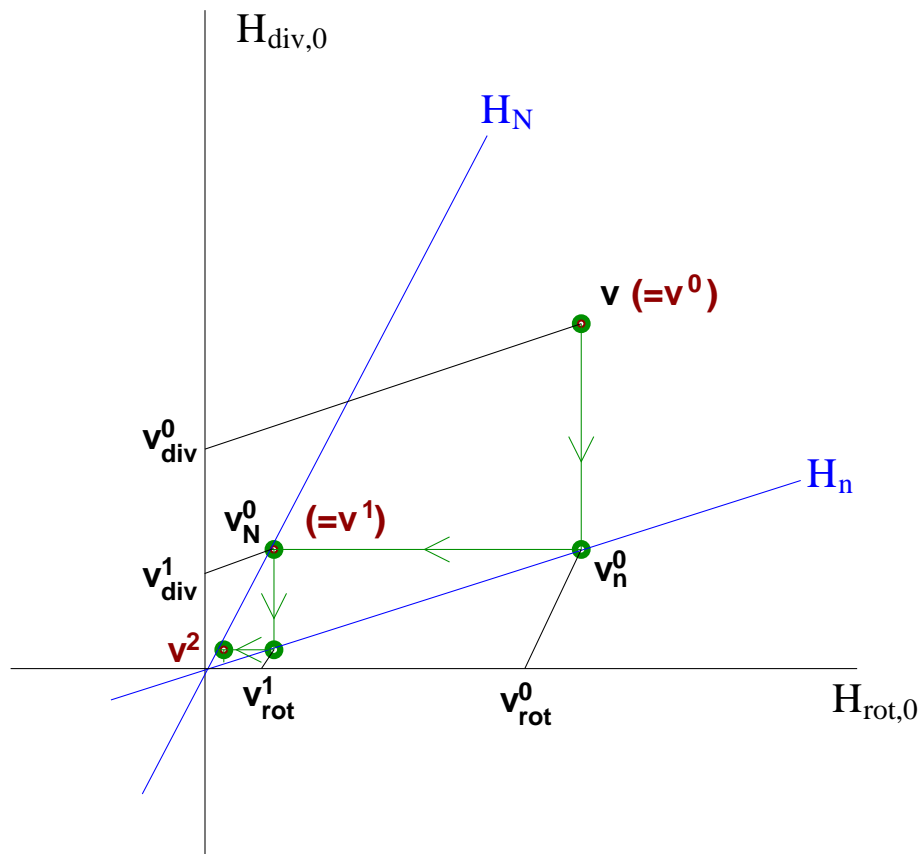
with

$$\mathbf{v}_{\text{div}} = \sum_{\mathbf{j}, \mathbf{k}} d_{\text{div } \mathbf{j}, \mathbf{k}} \Psi_{\text{div } \mathbf{j}, \mathbf{k}} \quad \text{and} \quad \mathbf{v}_{\text{curl}} = \sum_{\mathbf{j}, \mathbf{k}} d_{\text{curl } \mathbf{j}, \mathbf{k}} \Psi_{\text{curl } \mathbf{j}, \mathbf{k}}$$

Problem : the projectors on the divergence-free wavelet basis and on the gradient wavelet basis are *biorthogonal* projectors.

→ Iterative method to find \mathbf{v}_{div} et \mathbf{v}_{curl} .

Construction of the sequences $\mathbf{v}_{\text{div}}^p$ and $\mathbf{v}_{\text{curl}}^p$



Convergence processus for the sequences with $H_N = \text{vect}\{\Psi_{\mathbf{j},\mathbf{k}}^N\}$ and $H_n = \text{vect}\{\Psi_{\mathbf{j},\mathbf{k}}^n\}$.

Theorem: Convergence in dimension 2 for Shannon wavelets.

Proof (Kai Bittner): Looking at the proximity of H_n to $H_{\text{rot},0}$, we find a convergence criteria.

If there are $q_n, q_N \in \mathbb{R}$ such that:

$$\forall \mathbf{f}_n \in H_n, \quad \|\mathbb{P} \mathbf{f}_n\|_{L^2} \leq q_n \|\mathbb{Q} \mathbf{f}_n\|_{L^2} \quad (1)$$

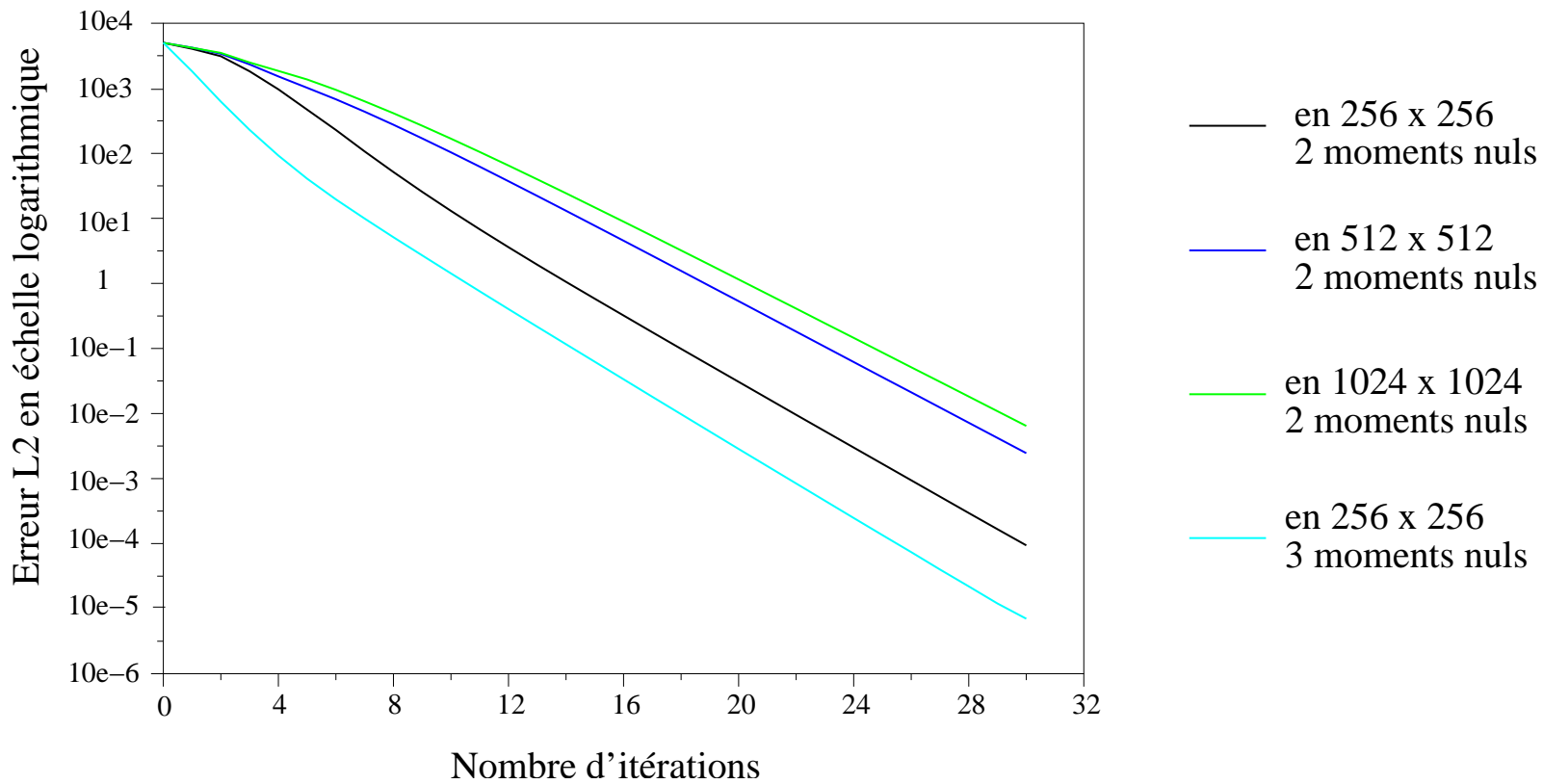
and

$$\forall \mathbf{f}_N \in H_N, \quad \|\mathbb{Q} \mathbf{f}_N\|_{L^2} \leq q_N \|\mathbb{P} \mathbf{f}_N\|_{L^2} \quad (2)$$

then

$$\|\mathbf{v}^{p+1}\|_{L^2} \leq q_n q_N \|\mathbf{v}^p\|_{L^2} \quad (3)$$

Numerically the convergence have been tested successfully on variate 2D and 3D fields.



Problem of the frequency localisation of the wavelets

Convergence rate linked (\sim proportional) to :

$$\rho = \iint_{\xi \in \mathbb{R}^2} \frac{(\xi_1^2 - \xi_2^2)^2}{(\xi_1^2 + \xi_2^2)^2} \left| \widehat{\psi}_1(\xi_1) \widehat{\psi}_1(\xi_2) \right|^2 d\xi$$

► Problem on the frequency localisation of the wavelets.

- in Fourier

- ponderation function:

$$\omega(\xi) = \frac{(\xi_1^2 - \xi_2^2)^2}{(\xi_1^2 + \xi_2^2)^2}$$

- Size of the compact support:

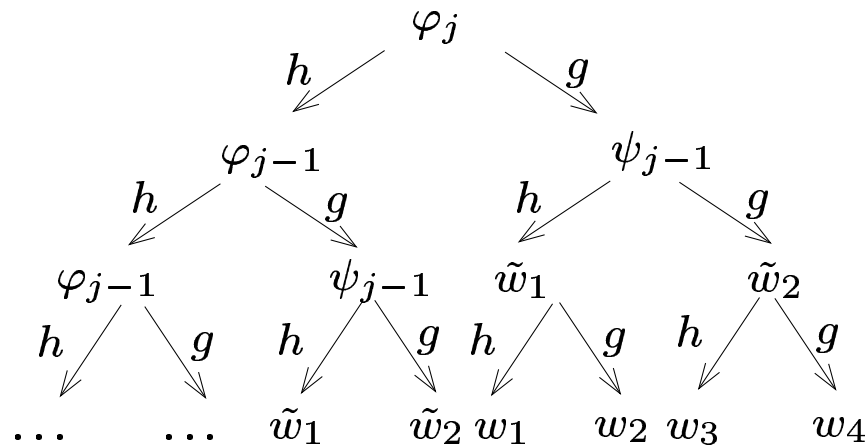
$$\left| \widehat{\psi}_1(\xi_1) \widehat{\psi}_1(\xi_2) \right|$$

Conclusion : we must get a **better localisation** in frequency for ψ_1 .

The wavelet packets

Definition : $w_n(x)$, $n \in \mathbb{N}$, packets associated to the scale function φ :

$$\hat{w}_n(\xi) = \prod_{j=1}^N m_{\epsilon_j} \left(\frac{\xi}{2^j} \right) \hat{\varphi} \left(\frac{\xi}{2^N} \right), \quad n = \sum_{j=1}^N \epsilon_j 2^{j-1}, \quad \epsilon_j \in \{0, 1\}.$$



In general, fail to control the frequency localisation.

Frequency target

- With the Shannon wavelets

- with the Walsh packets

By iteration, $w_\omega \sim \cos(2\pi\omega \cdot +\theta)$ for

$$\omega = \sum_{j \geq 1} \epsilon_j 2^{-j}$$

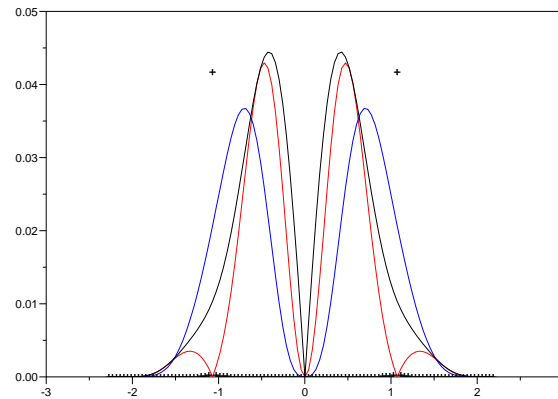
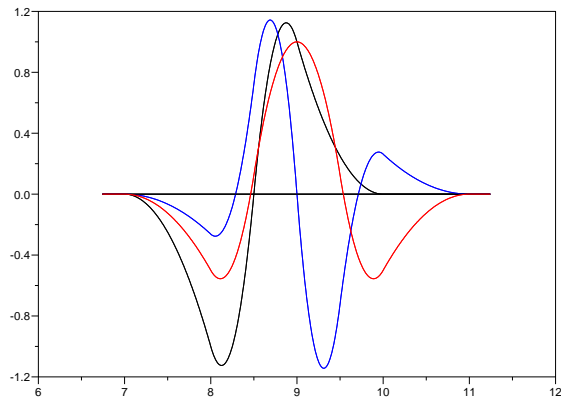
Packets modulation

"A theoretical study show that we have to target"

Ideal Packet :

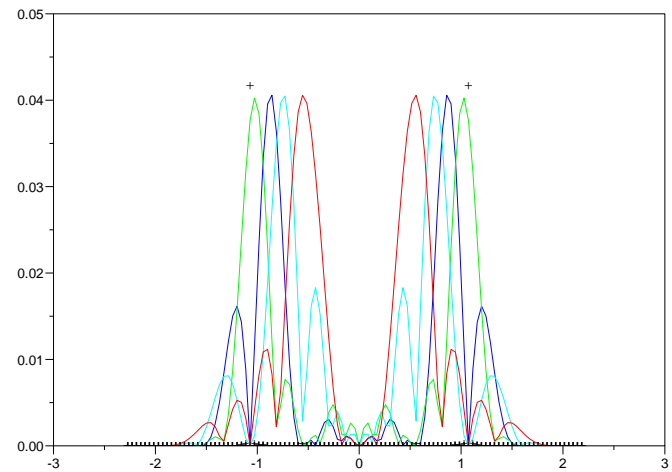
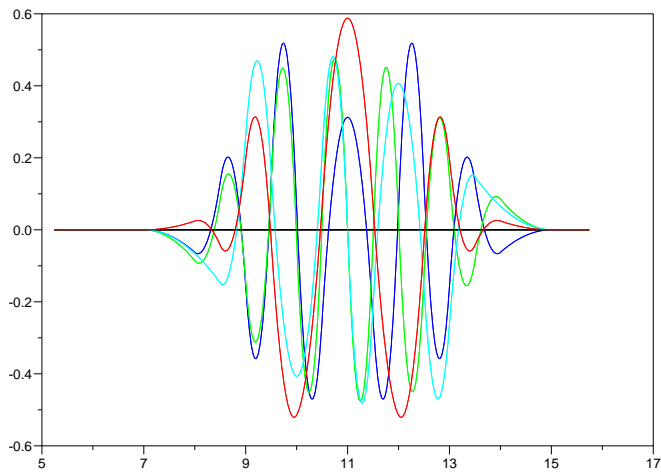
$$\psi_{\omega}(x) = \sum_{k \in \mathbb{Z}} \cos(2\pi\omega k + \theta) \varphi(2x - k)$$

Examples :



Quadratic spline wavelet packets with 2 wavelets

Packets with 4 quadratic spline wavelets



Numerical schema for Navier-Stokes

\mathbb{P} Leray projector with wavelets (give the pressure directly) :

$$\mathbb{P}[(\mathbf{u} \cdot \nabla)\mathbf{u}] = (\mathbf{u} \cdot \nabla)\mathbf{u} - \nabla p$$

Δ operator is linear.

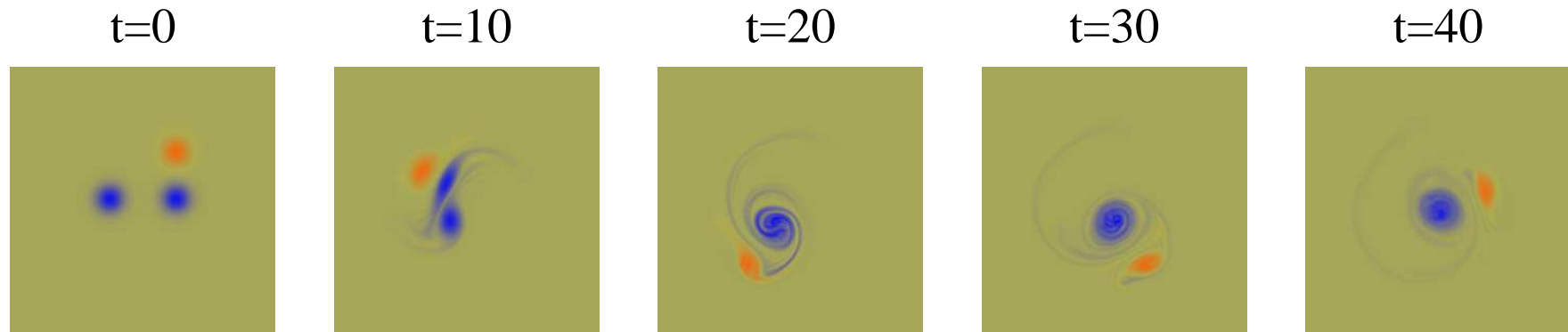
Euler explicite in time :

$$\mathbf{u}^{n+1} = \mathbf{u}^n - \delta t \mathbb{P}[(\mathbf{u} \cdot \nabla)\mathbf{u}]^n + \delta t \nu \Delta \mathbf{u}^n$$

on the wavelet coefficients :

$$d_{i\mathbf{j},\mathbf{k}}^{\text{div},n+1} = d_{i\mathbf{j},\mathbf{k}}^{\text{div},n} - \delta t d_{i\mathbf{j},\mathbf{k}}^{\text{div}}(\mathbb{P}[(\mathbf{u} \cdot \nabla)\mathbf{u}]) + \delta t \nu d_{i\mathbf{j},\mathbf{k}}^{\text{div}}(\Delta \mathbf{u}^n)$$

Test with the simulation “fusion of 3 vortices”



- **wavelet code** splines of degree 1 and 2 the simplest (~ 30 iterations for Helmholtz)

- **Runge-Kutta** schema of order 2 for the time evolution

- 256^2 grid

Results are visually identical to a **spectral code** in 256^2 .

Conclusion

Assets

- Calculation in $O(n)$
- Scale separation
- Non linear approximation

Perspectives

- Get adaptativity
- Limit conditions
- Complex geometries