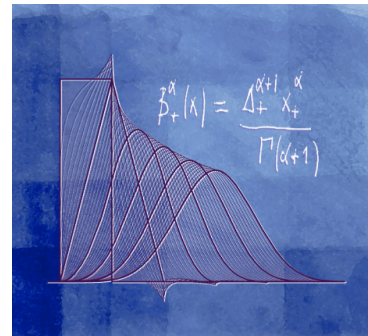


Affine invariance, splines, wavelets and fractional Brownian fields

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Joint work with Pouya Tafti and
Dimitri Van De Ville



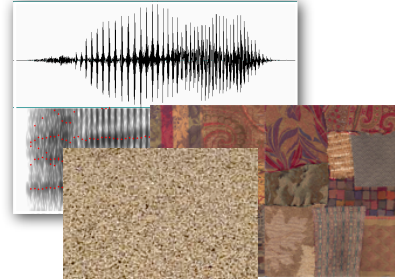
Mathematical Image Processing, Sept. 2007, CIRM, Marseille, France

The quest for invariance

- Invariance to coordinate transformations
 - Primary transformations (X): translation (T), scaling (S), rotation (R), affine (similarity) (A=S+R)
 - A continuous-domain operator L is X-invariant iff. it commutes with X; i.e., $\forall f \in L_2(\mathbb{R}^d), XLf = C_X \cdot LXf$ C_X : normalization constant
 - All classical physical laws are TSR-invariant
- Classical signal/image processing operators are invariant (to various extents)
 - Filters (linear or non-linear): T-invariant
 - Differentiators, Hilbert transform, wavelet transform: TS-invariant
 - Contour/ridge detectors (Gradient, Laplacian, Hessian): TSR-invariant
 - Steerable filters: TR-invariant

Invariant signals

- Natural signals/images often exhibit some degree of invariance (at least locally, if not globally)
 - Stationarity, texture: T-invariance
 - Isotropy (no preferred orientation): R-invariance
 - Self-similarity, fractality: S-invariance



1-3

OUTLINE

- Splines and T-invariant operators
 - Green functions as elementary building blocks
 - Multiresolution revisited
- Imposing scale (resp., affine) invariance
 - Fractional derivatives
 - Fractional (resp., polyharmonic) B-splines
- Fractional wavelets
 - Exact Hilbert-transform pairs of bases
 - Isotropic wavelets
 - Analysis of fractal processes

1-4

General concept of an L-spline

$L\{\cdot\}$: differential operator (translation-invariant)

$\delta(\mathbf{x}) = \prod_{i=1}^d \delta(x_i)$: multidimensional Dirac distribution

Definition

The continuous-domain function $s(\mathbf{x})$ is a **cardinal L-spline** iff.

$$L\{s\}(\mathbf{x}) = \sum_{\mathbf{k} \in \mathbb{Z}^d} a[\mathbf{k}] \delta(\mathbf{x} - \mathbf{k})$$

- Cardinality: the knots (or spline singularities) are on the (multi-)integers
- Generalization: includes polynomial splines as particular case ($L = \frac{d^N}{dx^N}$)

1-5

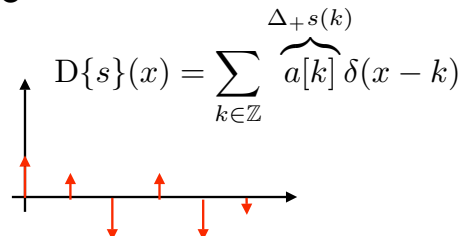
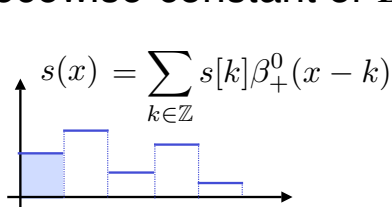
Example: piecewise-constant splines

■ Spline-defining operators

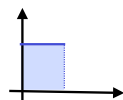
Continuous-domain derivative: $D = \frac{d}{dx} \longleftrightarrow j\omega$

Discrete derivative: $\Delta_+\{\cdot\} \longleftrightarrow 1 - e^{-j\omega}$

■ Piecewise-constant or D-spline



■ B-spline function



$$\beta_+^0(x) = \Delta_+ D^{-1}\{\delta\}(x) \longleftrightarrow \frac{1 - e^{-j\omega}}{j\omega}$$

1-6

Existence of a local, shift-invariant basis?

- Space of cardinal L-splines

$$V_L = \left\{ s(\mathbf{x}) : L\{s\}(\mathbf{x}) = \sum_{\mathbf{k} \in \mathbb{Z}^d} a[\mathbf{k}] \delta(\mathbf{x} - \mathbf{k}) \right\} \cap L_2(\mathbb{R}^d)$$

- Generalized B-spline representation

A “localized” function $\varphi(\mathbf{x}) \in V_L$ is called *generalized B-spline* if it generates a Riesz basis of V_L ; i.e., iff. there exists $(A > 0, B < \infty)$ s.t.

$$A \cdot \|c\|_{\ell_2(\mathbb{Z}^d)} \leq \left\| \sum_{\mathbf{k} \in \mathbb{Z}^d} c[\mathbf{k}] \varphi(\mathbf{x} - \mathbf{k}) \right\|_{L_2(\mathbb{R}^d)} \leq B \cdot \|c\|_{\ell_2(\mathbb{Z}^d)}$$

$$V_L = \left\{ s(\mathbf{x}) = \sum_{\mathbf{k} \in \mathbb{Z}^d} c[\mathbf{k}] \varphi(\mathbf{x} - \mathbf{k}) : \mathbf{x} \in \mathbb{R}^d, c \in \ell_2(\mathbb{Z}^d) \right\}$$

↓
continuous-domain signal
discrete signal
(B-spline coefficients)

1-7

Splines and Green's functions

Definition

$\rho(\mathbf{x})$ is a Green function of the shift-invariant operator L iff $L\{\rho\} = \delta$

$$\rho(\mathbf{x}) \xrightarrow{L\{\cdot\}} \delta(\mathbf{x}) \quad \Rightarrow \quad \delta(\mathbf{x}) \xrightarrow{L^{-1}\{\cdot\}} \rho(\mathbf{x})$$

(+ null-space component?)

- Cardinal L-spline: $L\{s\}(\mathbf{x}) = \sum_{\mathbf{k} \in \mathbb{Z}^d} a[\mathbf{k}] \delta(\mathbf{x} - \mathbf{k})$

Formal integration

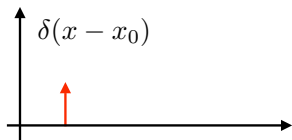
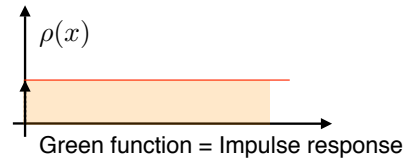
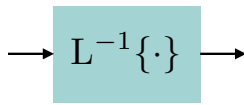
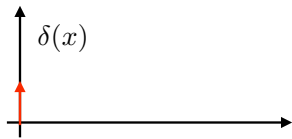
$$\sum_{\mathbf{k} \in \mathbb{Z}^d} a[\mathbf{k}] \delta(\mathbf{x} - \mathbf{k}) \xrightarrow{L^{-1}\{\cdot\}} s(\mathbf{x}) = \sum_{\mathbf{k} \in \mathbb{Z}^d} a[\mathbf{k}] \rho(\mathbf{x} - \mathbf{k})$$

$$\Rightarrow V_L = \text{span} \{ \rho(\mathbf{x} - \mathbf{k}) \}_{\mathbf{k} \in \mathbb{Z}^d}$$

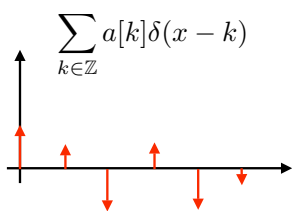
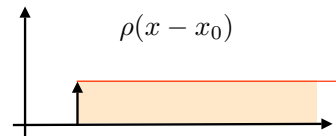
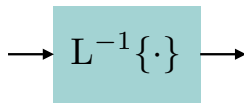
1-8

Example of spline synthesis

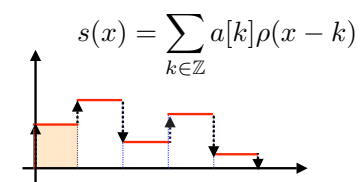
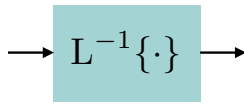
$$L = \frac{d}{dx} \Rightarrow L^{-1}: \text{integrator}$$



Translation invariance



Linearity



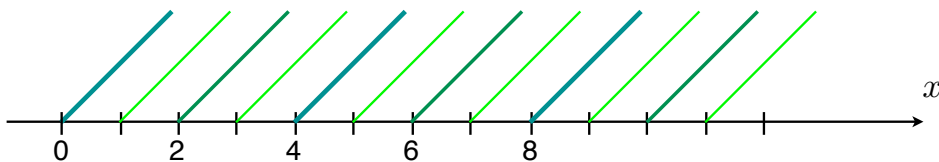
1-9

A fresh (Green's) view of multiresolution

Series of embedded spaces: $V_{(i)} \triangleq \text{span} \{ \rho(\mathbf{x} - 2^i \mathbf{k}) \}_{\mathbf{k} \in \mathbb{Z}^d}$

Example: $L = D^2$ with $\rho(x) = x_+$

$$V_{(2)} \subset V_{(1)} \subset V_{(0)} = V_L$$



Inclusion property: $V_{(j)} \subset V_{(i)}$ for $j \geq i$

■ Technical difficulties

- $\{ \rho(\mathbf{x} - \mathbf{k}) \}_{\mathbf{k} \in \mathbb{Z}^d}$ is not always a Riesz basis (e.g., $\rho \notin L_2(\mathbb{R}^d)$)
- Completeness issue

1-10

IMPOSING INVARIANCE

- Scale-invariant operators
 - Fractional B-splines
- Affine-invariant operators
 - Polyharmonic B-splines

1-11

Scale-invariant operators

Definition: An operator L is scale-invariant iff it commutes with dilation: i.e., $\forall s(\mathbf{x}), L\{s(\cdot)\}(\mathbf{x}/a) = C_a L\{s(\cdot/a)\}(\mathbf{x})$.

Theorem

The complete family of real scale-invariant 1D convolution operators is given by the fractional derivatives ∂_τ^γ , whose frequency response is

$$\hat{L}(\omega) = (-j\omega)^{\frac{\gamma}{2}-\tau} (j\omega)^{\frac{\gamma}{2}+\tau}$$

$\gamma \in \mathbb{R}^+$: order of the derivative (i.e., $|\hat{L}(\omega)| = |\omega|^\gamma$)

$\tau \in \mathbb{R}$: phase (or asymmetry)

(Unser & Blu, *IEEE-SP*, 2007)

1-12

Fractional B-splines properties

- Stable representation of fractional splines (Riesz basis)

$$V_{\partial_\tau^{\alpha+1}} = \left\{ s(x) = \sum_{k \in \mathbb{Z}} c[k] \beta_\tau^\alpha(x - k) : c[k] \in \ell_2 \right\} \quad \text{Condition: } \alpha > -\frac{1}{2}$$

- Degree versus order of approximation

- Building blocks (Green function of ∂_τ^γ): power functions of degree $\alpha = \gamma - 1$
- Order of approximation: $\gamma = \alpha + 1$

- Reproduction of polynomials

The fractional B-splines $\{\beta_\tau^\alpha(x - k)\}_{k \in \mathbb{Z}}$ reproduce the polynomials of degree $n = \lceil \alpha \rceil$. In particular,

$$\sum_{k \in \mathbb{Z}} \beta_\tau^\alpha(x - k) = 1 \quad (\text{partition of unity})$$

1-15

Scale- and rotation-invariant operators

Definition: An operator L is affine-invariant (or SR-invariant) iff.

$$\forall s(\mathbf{x}), L\{s(\cdot)\}(\mathbf{R}_\theta \mathbf{x}/a) = C_a \cdot L\{s(\mathbf{R}_\theta \cdot /a)\}(\mathbf{x})$$

where \mathbf{R}_θ is an arbitrary $d \times d$ unitary matrix and C_a a constant

- **Invariance theorem**

The complete family of real, scale- and rotation-invariant convolution operators is given by the fractional Laplacians

$$\Delta^{\frac{\gamma}{2}} \quad \xleftrightarrow{\mathcal{F}} \quad \|\boldsymbol{\omega}\|^\gamma$$

- Invariant Green functions (RBF) (Duchon, 1979)

$$\rho(\mathbf{x}) = \begin{cases} \|\mathbf{x}\|^{\gamma-d} \log \|\mathbf{x}\|, & \text{if } \gamma - d \text{ is even} \\ \|\mathbf{x}\|^{\gamma-d}, & \text{otherwise} \end{cases}$$

1-16

Construction of polyharmonic B-splines

Laplacian operator: $\Delta \xleftrightarrow{\mathcal{F}} -\|\boldsymbol{\omega}\|^2$

Discrete Laplacian: $\Delta_d \xleftrightarrow{\mathcal{F}} -\sum_{i=1}^d 4 \sin^2(\omega_i/2) \triangleq -\|2 \sin(\boldsymbol{\omega}/2)\|^2$

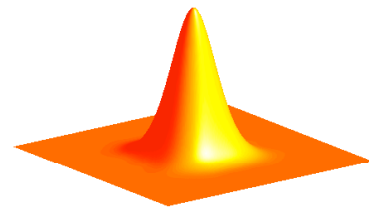
| | | |
|----|----|----|
| 0 | -1 | 0 |
| -1 | 4 | -1 |
| 0 | -1 | 0 |

■ Polyharmonic B-splines (Rabut, 1992)

Discrete operator: localization filter $Q(e^{j\boldsymbol{\omega}})$

$$\frac{\|2 \sin(\boldsymbol{\omega}/2)\|^\gamma}{\|\boldsymbol{\omega}\|^\gamma} \xrightarrow{\mathcal{F}^{-1}} \varphi_\gamma(\mathbf{x})$$

Continuous-domain operator: $\hat{L}(\boldsymbol{\omega})$



1-17

FRACTIONAL WAVELETS

- General scaling relations
- Fractional B-spline wavelets
- Exact Hilbert-transform pairs
- Fractional Mexican-hat wavelets
- Analysis of fractal processes
(multidimensional generalization of pioneering work of Flandrin and Abry)

1-18

General scaling relations (m integer)

Generalized splines of order γ

- Scale-invariant operator of order $\gamma \Leftrightarrow \hat{L}_\gamma(m\omega) = m^\gamma \hat{L}_\gamma(\omega)$

- A cardinal L_γ -spline dilated by m remains an L_γ -spline:

$$L_\gamma\{s(x)\} = \sum_{\mathbf{k} \in \mathbb{Z}^d} a[\mathbf{k}] \delta(x - \mathbf{k}) \Rightarrow L_\gamma\{s(x/m)\} = \frac{1}{m^{\gamma-d}} \sum_{\mathbf{k} \in \mathbb{Z}^d} a[\mathbf{k}] \delta(x - m\mathbf{k})$$

Scaling function

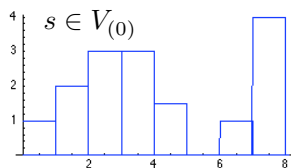
- Generalized B-spline: $\varphi_\gamma(x) = \sum_{\mathbf{k} \in \mathbb{Z}^d} q[\mathbf{k}] \rho(x - \mathbf{k}) \xleftrightarrow{\mathcal{F}} \frac{Q(e^{j\omega})}{\hat{L}_\gamma(\omega)}$

- General m -scale relation: $\varphi_\gamma(x/m) = \sum_{\mathbf{k} \in \mathbb{Z}^d} h_{\gamma,m}[\mathbf{k}] \varphi_\gamma(x - \mathbf{k})$

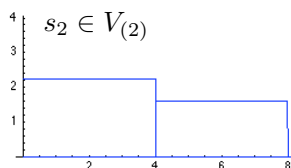
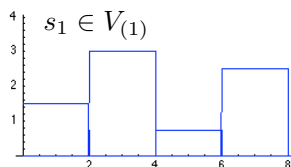
- Refinement filter: $H_{\gamma,m}(e^{j\omega}) = \frac{|m|^d \hat{\varphi}_\gamma(m\omega)}{\hat{\varphi}_\gamma(\omega)} = \frac{1}{m^{\gamma-d}} \frac{Q(e^{jm\omega})}{Q(e^{j\omega})}$

1-19

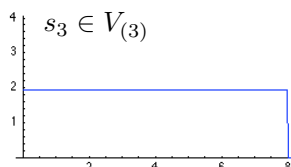
Multiresolution analysis of L_2



- Multiresolution basis functions: $\varphi_{i,\mathbf{k}}(x) = 2^{-id/2} \varphi\left(\frac{x-2^i \mathbf{k}}{2^i}\right)$
- Subspace at resolution i : $V_{(i)} = \text{span}\{\varphi_{i,\mathbf{k}}\}_{\mathbf{k} \in \mathbb{Z}^d}$



Two-scale relation $\Rightarrow V_{(i)} \subset V_{(j)}$, for $i \geq j$

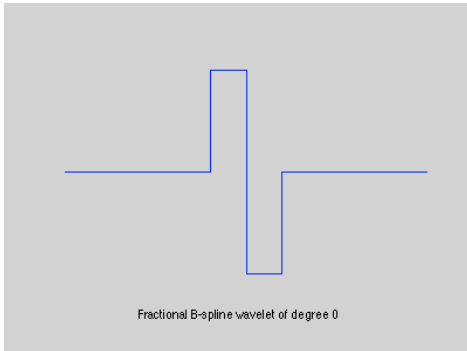


Partition of unity $\Leftrightarrow \overline{\bigcup_{i \in \mathbb{Z}} V_{(i)}} = L_2(\mathbb{R}^d)$

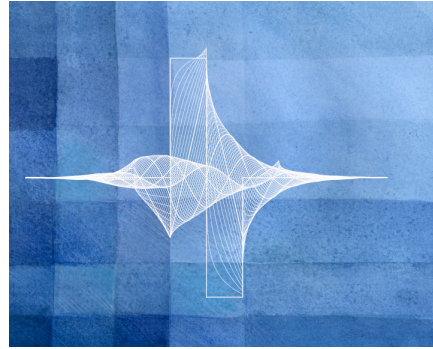
1-20

Fractional B-spline wavelets

$$\psi_\tau^\alpha(x/2) = \sum_{k \in \mathbb{Z}} \underbrace{\sum_{n \in \mathbb{Z}} (-1)^n h_{\tau,2}^\alpha[n] \beta_0^{2\alpha+1}(n+k-1) \beta_\tau^\alpha(x-k)}_{g_{\text{wave}}[k]}$$



Causal B-spline: $\tau = \frac{\alpha+1}{2}$



(Unser and Blu, *SIAM Review*, 2000)

1-21

Fractional B-spline wavelet properties

- Semi-orthogonal basis of $L_2(\mathbb{R})$: $\{2^{-i/2} \psi_\tau^\alpha(x/2^i - k)\}_{(i,k) \in \mathbb{Z}^2}$

- Multiscale fractional derivative behavior:

$$\langle \psi_\tau^\alpha \left(\frac{\cdot - x}{a} \right), f \rangle = \partial_\tau^{\alpha+1} \left\{ \phi \left(\frac{\cdot}{a} \right) * f \right\} (x)$$

$\phi(x)$: smoothing spline kernel of order $2\alpha + 2$

- Spline family closed under fractional differentiation; in particular, $\psi_{\tau-\frac{1}{2}}^\alpha(x)$ is the Hilbert transform of $\psi_\tau^\alpha(x)$
- Optimal time-frequency localization: convergence to Gabor functions as α increases
- Fast filterbank algorithm (using FFT)

1-22

Multidimensional, nonseparable wavelets

Search for a **single wavelet** that generates a basis of $L_2(\mathbb{R}^d)$ and that is a multi-scale version of the operator L ; i.e., $\psi = L^* \phi$ where ϕ is a suitable smoothing kernel

■ General operator-based construction

- Basic space V_0 generated by the integer shifts of the Green function ρ of L :

$$V_0 = \text{span}\{\rho(\mathbf{x} - \mathbf{k})\}_{\mathbf{k} \in \mathbb{Z}^d} \text{ with } L\rho = \delta$$

- Orthogonality between V_0 and $W_0 = \text{span}\{\psi(\mathbf{x} - \frac{1}{2}\mathbf{k})\}_{\mathbf{k} \in \mathbb{Z}^d \setminus 2\mathbb{Z}^d}$

$$\begin{aligned} \langle \psi(\cdot - \mathbf{x}_0), \rho(\cdot - \mathbf{k}) \rangle &= \langle \phi, L\rho(\cdot - \mathbf{k} + \mathbf{x}_0) \rangle \\ &= \langle \phi, \delta(\cdot - \mathbf{k} + \mathbf{x}_0) \rangle = \phi(\mathbf{k} - \mathbf{x}_0) = 0 \end{aligned}$$

(can be enforced via a judicious choice of ϕ (interpolator) and \mathbf{x}_0)

- Works in arbitrary dimensions and for any dilation matrix \mathbf{D}

1-23

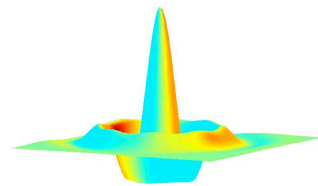
New isotropic basis: the Mexican-Hat

$$\psi_\gamma(\mathbf{x}) = (-\Delta)^{\frac{\gamma}{2}} \phi_{2\gamma}(\mathbf{x})$$

Gaussian-like polyharmonic spline kernel:

$$\phi_{2\gamma}(\mathbf{x}) \approx \exp\left(-\frac{\|\mathbf{x}\|^2}{2\gamma/9}\right)$$

$$\psi_{(i,\mathbf{k})} \triangleq |\det(\mathbf{D})|^{-i/2} \psi_\gamma(\mathbf{D}^{-i}\mathbf{x} - \mathbf{D}^{-1}\mathbf{k})$$



- The set $\{\psi_{(i,\mathbf{k})}\}_{(\mathbf{k} \in \mathbb{Z}^d \setminus \mathbf{D}\mathbb{Z}^d, i \in \mathbb{Z})}$ is a semi-orthogonal basis of $L_2(\mathbb{R}^d)$ for $\gamma > 1$:

$$\forall f \in L_2(\mathbb{R}^d), \quad f = \sum_{i \in \mathbb{Z}} \sum_{\mathbf{k} \in \mathbb{Z}^d \setminus \mathbf{D}\mathbb{Z}^d} \langle f, \psi_{(i,\mathbf{k})} \rangle \tilde{\psi}_{(i,\mathbf{k})} = \sum_{i \in \mathbb{Z}} \sum_{\mathbf{k} \in \mathbb{Z}^d \setminus \mathbf{D}\mathbb{Z}^d} \langle f, \tilde{\psi}_{(i,\mathbf{k})} \rangle \psi_{(i,\mathbf{k})}$$

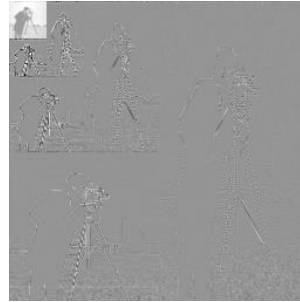
$\{\tilde{\psi}_{(i,\mathbf{k})}\}$: dual wavelet basis of $\{\psi_{(i,\mathbf{k})}\}$

- The Mexican-Hat wavelet analysis implements a multiscale version of the Laplace operator and is perfectly reversible (one-to-one transform)
- The wavelet transform has a fast filterbank algorithm (Van De Ville, IEEE-IP, 2005)

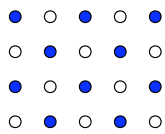
1-24

Mexican-Hat wavelet basis

■ Nonredundant transform



Quincunx sampling pattern



$$\mathbf{D} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

First decomposition level (one-to-one):

$$\varphi(\mathbf{D}^{-1}\mathbf{x} - \mathbf{k})$$

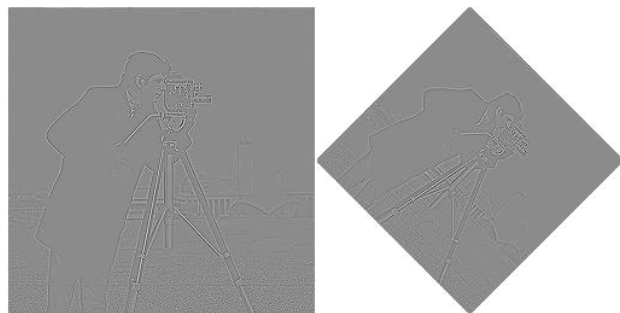
Scaling functions
(dilated by $\sqrt{2}$)

$$\psi(\mathbf{D}^{-1}\mathbf{x} - \mathbf{k})$$

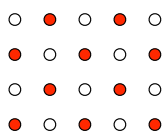
Wavelets
(dilated by $\sqrt{2}$)

Mexican-Hat wavelet analysis

■ Pyramid decomposition: redundancy 2



Quincunx sampling pattern



First decomposition level:

$$\varphi(\mathbf{D}^{-1}\mathbf{x} - \mathbf{k})$$

Scaling functions

$$\psi(\mathbf{D}^{-1}(\mathbf{x} - \mathbf{k}))$$

Wavelets
(redundant by 2)

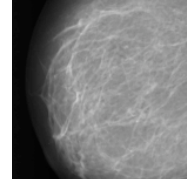
Self-similar image models

■ Motivation: most images exhibit some level of self-similarity

- 2D projection of a 3D scene: objects appear with various degrees of magnification (depth-dependent scaling)
- Step edges and singularities give rise to $1/\omega$ spectral decay
- Natural/biological growth processes often generate fractal structures



(Penland, 1984; Mumford 2001)



■ Stochastic fractals: fractional Brownian fields

- d -dimensional version of Mandelbrot's fractional Brownian motion: self-similar analogs of stationary processes—loosely referred to as “ $1/\omega$ -processes”

- Distributional solution of the **stochastic partial differential equation**:

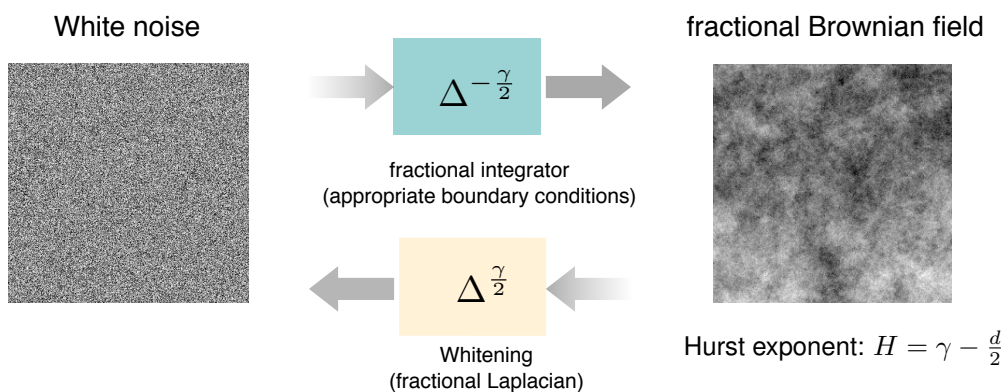
$$\Delta^{\frac{H}{2} + \frac{d}{4}} B_H = W, \quad \text{where } W \text{ is white Gaussian noise}$$

- **Key finding**: Polyharmonic splines are the optimal function spaces for the MMSE estimation of such processes from their noisy samples (Blu, 2007; Tirosh, 2006)

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fBm: innovation model

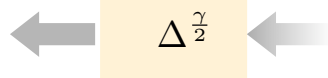
Formalism: Gelfand's theory of generalized stochastic processes



■ Deterministic counterpart

Train of Dirac impulses:

$$\sum_{\mathbf{k} \in \mathbb{Z}^d} a[\mathbf{k}] \delta(\mathbf{x} - \mathbf{k})$$



Polyharmonic spline

$$s(\mathbf{x}) = \sum_{\mathbf{k} \in \mathbb{Z}^d} a[\mathbf{k}] \rho_\gamma(\mathbf{x} - \mathbf{k})$$

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Wavelet analysis of fBm: whitening revisited

■ Operator-like behavior of wavelet

- Analysis wavelet: $\psi_\gamma = \Delta^{\frac{\gamma}{2}} \phi(\mathbf{x}) = \Delta^{\frac{H}{2} + \frac{d}{4}} \psi'_{\gamma'}(\mathbf{x})$
- Reduced-order wavelet: $\psi'_{\gamma'}(\mathbf{x}) = \Delta^{\frac{\gamma'}{2}} \phi(\mathbf{x})$ with $\gamma' = \gamma - (H + \frac{d}{2}) > 0$

■ Stationarizing effect of wavelet analysis

- Analysis of fractional Brownian field with exponent H :

$$\langle B_H, \psi_\gamma \left(\frac{\cdot - \mathbf{x}_0}{a} \right) \rangle \propto \langle \Delta^{\frac{H}{2} + \frac{d}{4}} B_H, \psi'_{\gamma'} \left(\frac{\cdot - \mathbf{x}_0}{a} \right) \rangle = \langle W, \psi'_{\gamma'} \left(\frac{\cdot - \mathbf{x}_0}{a} \right) \rangle$$

- Equivalent spectral noise shaping: $S_{\text{wave}}(e^{j\omega}) = \sum_{\mathbf{n} \in \mathbb{Z}^d} |\hat{\psi}'_{\gamma'}(\omega + 2\pi\mathbf{n})|^2$
 \Rightarrow Extent of wavelet-domain whitening depends on flatness of $S_{\text{wave}}(e^{j\omega})$
- “Whitening” effect is the same at all scales up to a proportionality factor
 \Rightarrow fractal exponent can be deduced from the log-log plot of the variance

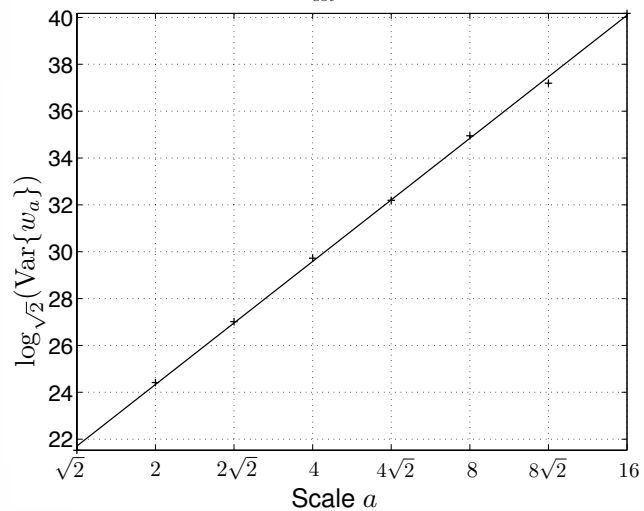
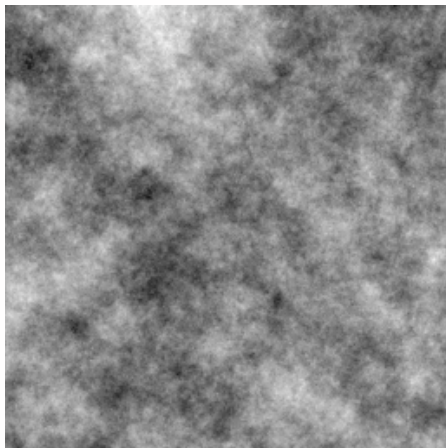
1-29

Wavelet analysis of fractional Brownian fields

Theoretical scaling law : $\text{Var}\{w_a[k]\} = \sigma_0^2 \cdot a^{(2H+d)}$

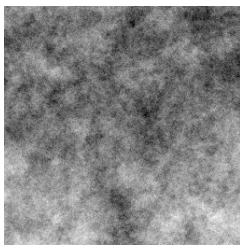
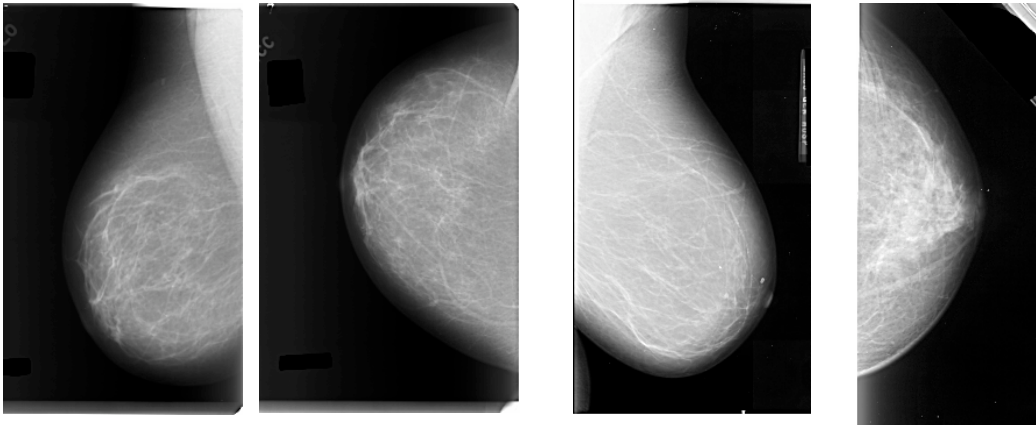
log-log plot of variance

$H_{\text{est}} = 0.31$



1-30

Fractals in bioimaging: fibrous tissue



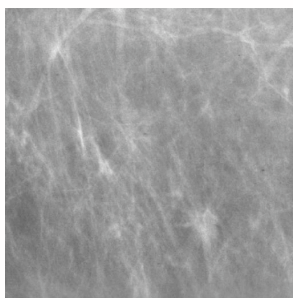
DDSM: University of Florida

(Digital Database for Screening Mammography)

(Laine, 1993; Li *et al.*, 1997)

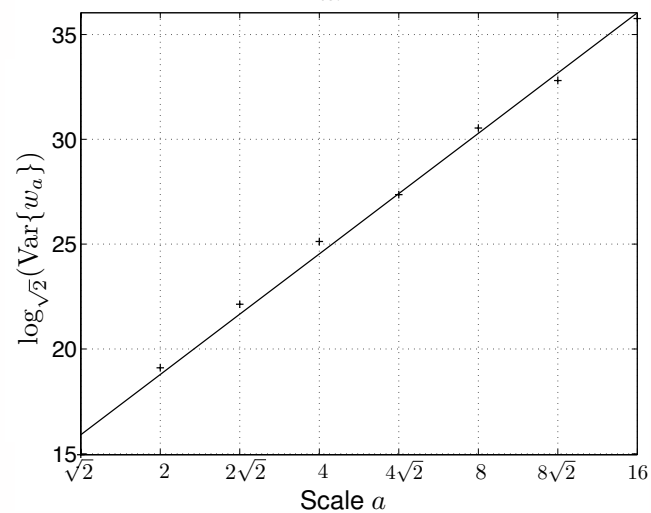
1-31

Wavelet analysis of mammograms



log-log plot of variance

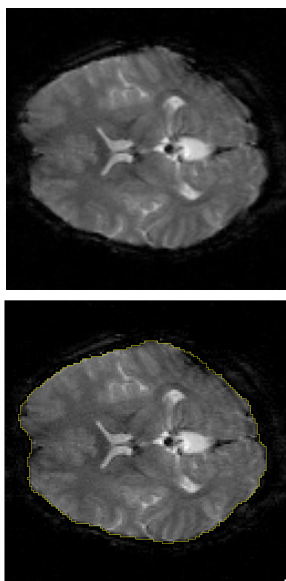
$H_{est} = 0.44$



Fractal dimension: $D = 1 + d - H = 2.56$ with $d = 2$ (topological dimension)

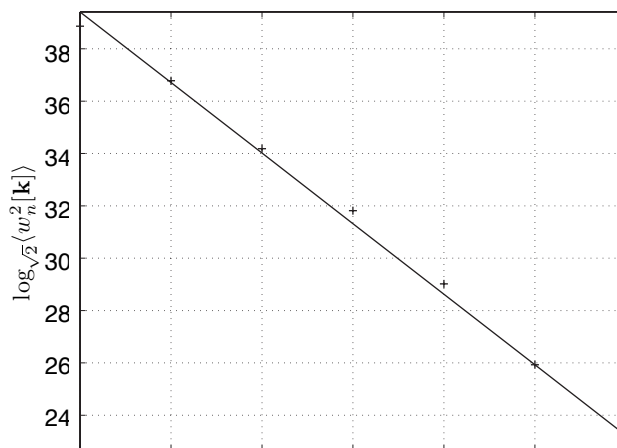
1-32

Wavelet analysis of fMRI data



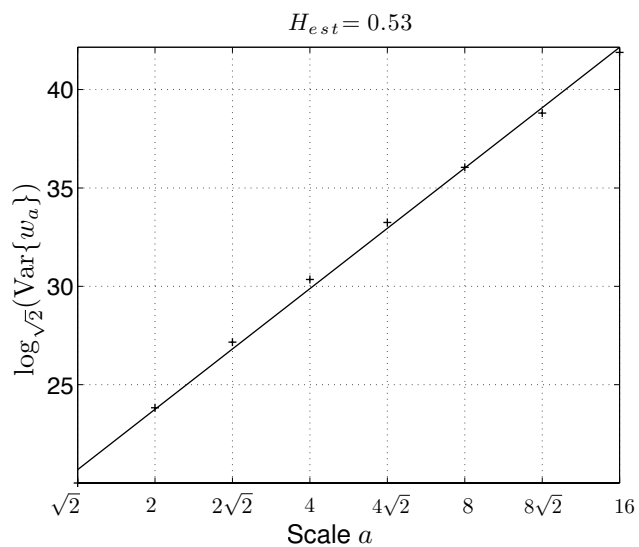
Brain: courtesy of Jan Kybic

Fractal dimension: $D = 1 + d - H = 2.65$ with $d = 2$ (topological dimension)



1-33

...and some non-biomedical images...



1-34

CONCLUSION

- Affine invariance is universal
 - Natural and biomedical images
 - Operators: fractional derivatives
 - Optimal functions spaces: splines (universal interpolators)
 - Fractal processes
- Invariant operators yield “matched” wavelet bases
 - Existence of “localized” B-spline bases
 - Enforcement of multiresolution property \Rightarrow wavelets
 - Wavelet = multiscale version of operator
 - “Invariant” wavelets: more operator-like (e.g., better isotropy)
 - Extended family of **fractional** wavelets: tunable, closed under fractional differentiation
 - Promising tool for (multi-D) signal processing and fractal analyses

1-35

The end: Thank you!

■ Key collaborators

T. Blu, D. Van De Ville, P. Tafti, I. Khalidov, K. Chaudhury, K. Balac

■ Selected papers

- M. Unser, T. Blu, "Fractional Splines and Wavelets," *SIAM Review*, 42(1), pp. 43-67, March 2000.
- M. Unser, T. Blu, "Self-similarity—Part I: Splines and Operators," *IEEE Trans. Signal Processing*, 55(4), pp. 1352-1363, 2007.
- T. Blu, M. Unser, "Self-similarity—Part II: Optimal Estimation of Fractal Processes," *IEEE Trans. Signal Processing*, 55(4), pp. 1364-1378, 2007.
- D. Van De Ville, T. Blu, M. Unser, "Isotropic Polyharmonic B-Splines: Scaling Functions and Wavelets," *IEEE Trans. Image Processing*, 14(11), pp. 1798-1813, 2005.

- Preprints and demos: <http://bigwww.epfl.ch/>

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