

Signal analysis based on complex wavelet signs

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Local signal analysis by wavelet transform

(Complex) wavelet at scale $a > 0$ and location $b \in \mathbb{R}$

$$\kappa_{a,b}(x) := \frac{1}{\sqrt{a}} \kappa\left(\frac{x-b}{a}\right)$$

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Wavelet coefficient decomposition

$$\langle f, \kappa_{a,b} \rangle = |\langle f, \kappa_{a,b} \rangle| \cdot \frac{\langle f, \kappa_{a,b} \rangle}{|\langle f, \kappa_{a,b} \rangle|} = \underbrace{|\langle f, \kappa_{a,b} \rangle|}_{\text{amplitude}} \cdot \underbrace{\text{sgn} \langle f, \kappa_{a,b} \rangle}_{\text{sign}}$$

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Amplitudes of wavelet coefficients



Classical local regularity
(e.g. Sobolev, Besov)

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Signs of wavelet coefficients



?

Observations and approaches to wavelet signs

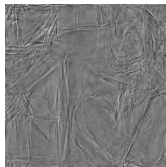
Importance of signs in signals

- Bandpass signals determined by sign information (Logan 1977)
- Signs important for image reconstruction (Oppenheim, Lim 1981)

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Shuffle of sign

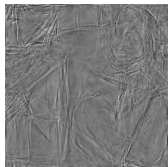


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Phase congruency (Morrone, Owens 1987; Kovesi 1999)

- Heuristic approach to signal analysis based on Fourier-signs
- Successful application to edge detection

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- Lines of constant sign (phase) in scalogram $(a, b) \mapsto \langle f, \kappa_{a,b} \rangle$ converge towards singularities

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Wavelet sign as indicator of local symmetry (Holschneider 1995; Kovesi 1999)

Complex signature wavelets

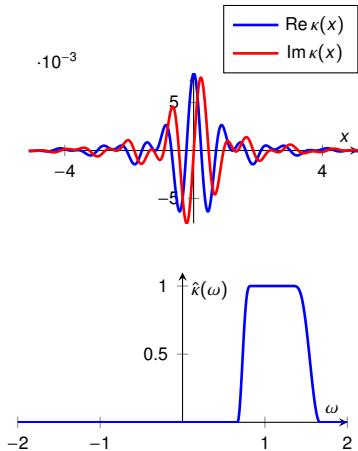
Complex valued function $\kappa \in \mathcal{S}(\mathbb{R}; \mathbb{C})$ is called *signature wavelet* if

- frequency spectrum real and non-negative, i.e.,

$$\hat{\kappa} \geq 0,$$

- support of frequency spectrum compact and one-sided, i.e.,

$$\text{supp } \hat{\kappa} \subset [c, d].$$



$$0 < c < d < \infty$$

The complex signature

Definition (Demaret, Massopust, Størath 2012)

The *signature* of f at location $b \in \mathbb{R}$ is defined by

$$\sigma f(b) = \lim_{a \rightarrow 0} \operatorname{sgn} \langle f, \kappa_{a,b} \rangle,$$

if the limit exists and has the same value for all signature wavelets κ ;
otherwise, we set

$$\sigma f(b) = 0.$$

$f \in \mathcal{S}'(\mathbb{R}; \mathbb{R})$ real-valued tempered distribution; $\operatorname{sgn} z = \frac{z}{|z|}$, $\operatorname{sgn} 0 = 0$

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Idea

- b salient point $\leftrightarrow |\sigma f(b)| = 1$
- b regular point $\leftrightarrow \sigma f(b) = 0$

$f \in \mathcal{S}'(\mathbb{R}; \mathbb{R})$ real-valued tempered distribution; $\operatorname{sgn} z = \frac{z}{|z|}$, $\operatorname{sgn} 0 = 0$

Basic examples

- Bandlimited function f

$$\sigma f(b) = 0, \quad \text{for all } b \in \mathbb{R}.$$

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- Unit step U

$$\sigma U(b) = \begin{cases} i, & \text{for } b = 0, \\ 0, & \text{else.} \end{cases}$$



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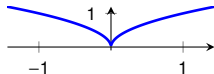
- Unit step U

$$\sigma U(b) = \begin{cases} i, & \text{for } b = 0, \\ 0, & \text{else.} \end{cases}$$



- Cusp $| \bullet |^\gamma$

$$\sigma(| \bullet |^\gamma)(b) = \begin{cases} -1, & \text{for } b = 0, \\ 0, & \text{else,} \end{cases}$$



where $0 < \gamma \leq 1$.

Signature at regular regions

Theorem (Demaret, Massopust, Storath 2012)

If f is smooth in a neighborhood of $b \in \mathbb{R}$ and if

$$f^{(k)}(b) = 0, \quad \text{for all } k \in \mathbb{N}_0,$$

then

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Sketch of proof

- Show that all moments of $G(u) = \langle f, \kappa_{\frac{1}{u}, b} \rangle$, $u \in \mathbb{R}$, vanish.
- Conclude that either $\operatorname{Re} G$ and $\operatorname{Im} G$ have infinitely many sign changes or are equal to 0 for u large enough; thus

$$\lim_{u \rightarrow \infty} \operatorname{sgn} G(u) = \lim_{a \rightarrow 0} \operatorname{sgn} \langle f, \kappa_{a, b} \rangle$$

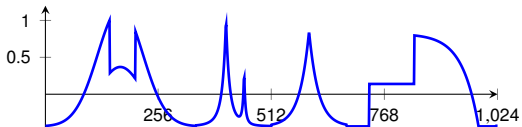
does not exist or is equal to 0. In either case, $\sigma f(b) = 0$.

Locally polynomial signals

Corollary

If f coincides on an open set $U \subset \mathbb{R}$ with a polynomial then

$$\sigma f(b) = 0, \quad \text{for every } b \in U.$$



A piecewise polynomial signal from Wavelab (Donoho et. al.)

Jump discontinuities

Theorem (Demaret, Massopust, Storath 2012)

If f has a jump discontinuity at b , then

$$\sigma f(b) = \begin{cases} +i, & \text{if } f(b-) < f(b+), \\ -i, & \text{if } f(b-) > f(b+). \end{cases}$$

Jump discontinuities

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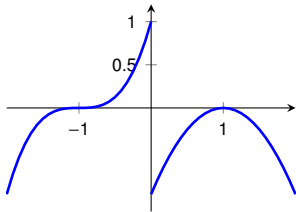
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Example

$$f(x) = \begin{cases} (x+1)^3, & \text{for } x < 0, \\ -(x-1)^2, & \text{for } x \geq 0. \end{cases}$$

$$\sigma f(b) = \begin{cases} -i, & \text{if } b = 0, \\ 0, & \text{else.} \end{cases}$$

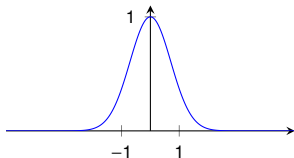


Differences to classical singular support

Gaussian function

$$f(x) = e^{-x^2}$$

$$\{0\} \subset \text{supp } \sigma f \not\subset \text{sing supp } f = \emptyset$$



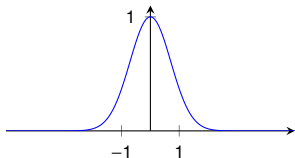
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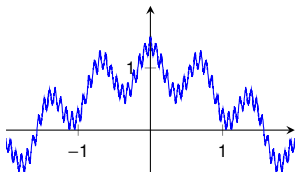
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Weierstrass function

$$f(x) = \sum_{n=0}^{\infty} r^n \cos(t^n x)$$

$$\mathbb{R} = \text{sing supp } f \not\subset \text{supp } \sigma f = \emptyset.$$



$$\text{sing supp } f = \{b \in \mathbb{R} : f \text{ not } C^\infty \text{ at } b\}$$

Signature complementary to classical regularity

Fractional power of Laplacian

$$(-\Delta)^r f = \mathcal{F}^{-1}(|\bullet|^{2r} \cdot \hat{f})$$

Fractional Hilbert transform

$$\mathcal{H}^\alpha f = \mathcal{F}^{-1}(e^{-i\alpha \frac{\pi}{2} \cdot \text{sgn}(\bullet)} \cdot \hat{f}).$$

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Local Sobolev regularity

Addition

$$s_{(-\Delta)^r f} = s_f - 2r$$

Invariant

$$s_{\mathcal{H}^\alpha f} = s_f$$

$r, \alpha \in \mathbb{R}; s_f(b) := \sup \{s \in \mathbb{R} : \exists \varphi \in \mathcal{D}(\mathbb{R}), \varphi(b) \neq 0, \text{ so that } \varphi f \in H^s(\mathbb{R})\}$

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$$s_{(-\Delta)^r f} = s_f - 2r$$

Invariant

$$s_{\mathcal{H}^\alpha f} = s_f$$

Signature

Invariant

$$\sigma((-\Delta)^r f) = \sigma f$$

Rotation

$$\sigma(\mathcal{H}^\alpha f) = e^{i\alpha \frac{\pi}{2}} \cdot \sigma f$$

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Geometric interpretation

Step right



Signature $+i$

Cusp upwards



Signature $+1$

Step left



Signature $-i$

Cusp downwards



Signature -1

Geometric interpretation

Step right



Signature $+i$

Cusp upwards



Signature $+1$

Step left



Signature $-i$

Cusp downwards



Signature -1

Imaginary signature



locally antisymmetric

Real signature



locally symmetric

Discretization (I)

Reminder: If $\sigma f(b) \neq 0$, then

$$\sigma f(b) = \lim_{j \rightarrow \infty} \operatorname{sgn} \langle f, \kappa_{a_j, b} \rangle, \quad \text{for } a_j \rightarrow 0.$$

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Observations

- Cesàro-sequence converges to signature,

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{j=1}^N \operatorname{sgn} \langle f, \kappa_{a_j, b} \rangle = \sigma f(b).$$

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- Modulus of Cesàro-sequence tends to 1,

$$\lim_{N \rightarrow \infty} \left| \frac{1}{N} \sum_{j=1}^N \operatorname{sgn} \langle f, \kappa_{a_j, b} \rangle \right| = |\sigma f(b)| = 1.$$

Discretization (II)

In practice: finite number of scale samples $\{a_j\}_{j=1}^N$

Idea

Consider the finite sum

$$\bar{w}_b := \frac{1}{N} \sum_{j=1}^N \operatorname{sgn} \langle f, \kappa_{a_j, b} \rangle,$$

as N -th element of Cesàro-sequence

- $|\bar{w}_b| \ll 1 \rightarrow$ estimate signature as 0
- $|\bar{w}_b| \approx 1 \rightarrow$ estimate signature by orientation in complex plane of \bar{w}_b

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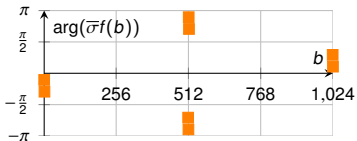
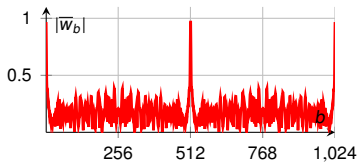
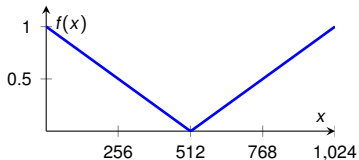
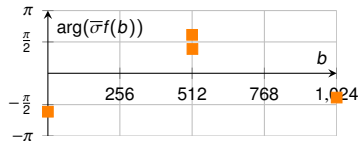
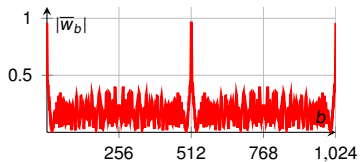
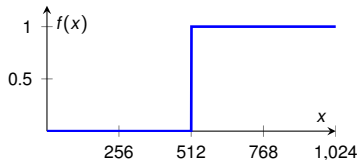
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Discrete signature

$$\bar{\sigma}f(b) := \begin{cases} 0, & \text{if } |\bar{w}_b| < \tau, \\ \operatorname{sgn} \bar{w}_b, & \text{if } |\bar{w}_b| \geq \tau, \end{cases}$$

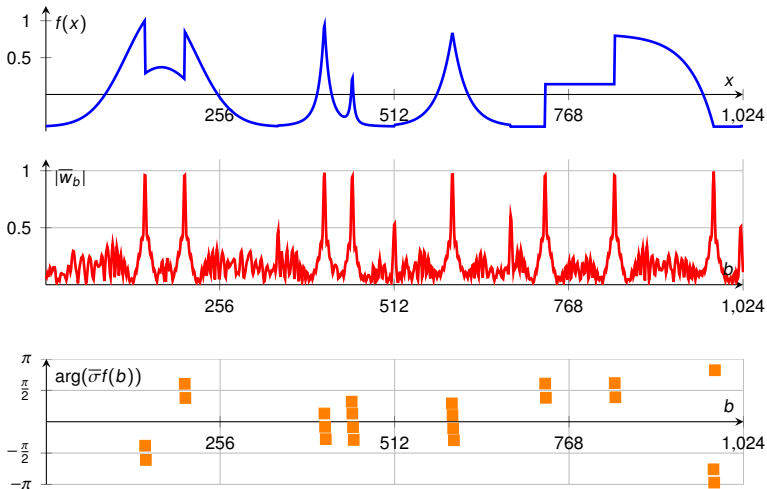
where $\tau \in [0, 1]$ is an empirical threshold parameter

Experiment - Jump and cusp



Threshold $\tau = 0.7$

Experiment - Piecewise polynomial



Threshold $\tau = 0.7$

Summary

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- Signature as new rigorous approach to sign-based signal analysis
- Detection of salient points in signals and determination of local symmetry
- Discrete signature for numerical estimation

Main reference

- L. Demaret, P. Massopust, and M. Storath. [Signal analysis based on complex wavelet signs](#).
Submitted, 2012.
[Preprint arXiv:1208.4578v1](#)

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Difference to phase congruency

Phase congruency (Kovesi 1999)

$$\widetilde{\text{PC}}(b) = \frac{\sum_{j=1}^N \langle f, \kappa_{a_j, b} \rangle}{\sum_{j=1}^N |\langle f, \kappa_{a_j, b} \rangle|}.$$

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