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TGU Lecture Series

Lecture 2:

Elements of the
mathematical theory
of wavelets

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OUTLINE

- ⌘ Classical dyadic wavelets.
 - ⌘ Definition and examples;
 - ⌘ Types of wavelets (MRA, MSF);
 - ⌘ Elements of the MRA method.
- ⌘ Higher dimensions.
 - ⌘ Tensor product constructions;
 - ⌘ Unitary extension principle;
 - ⌘ Composite wavelets;
 - ⌘ “Blubbery blob”.



Important unitary reps

⌘ **Translations $T_k: \mathbf{L}^2(\mathbf{R}) \rightarrow \mathbf{L}^2(\mathbf{R})$.**

$$T_k f(x) = f(x - k), \quad T_k = T^k = T(k)$$

⌘ **Modulations $M_b: \mathbf{L}^2(\mathbf{R}) \rightarrow \mathbf{L}^2(\mathbf{R})$.**

$$M_b f(x) = e^{2\pi i b x} f(x), \quad M_\ell = M^\ell = M(\ell)$$

⌘ **Dilations $D_a: \mathbf{L}^2(\mathbf{R}) \rightarrow \mathbf{L}^2(\mathbf{R})$.**

$$(D_a f)(x) = 1/\sqrt{|a|} f(x/a)$$



Affine systems, wavelets

⌘ **Dyadic affine systems generated by ψ in $L^2(\mathbf{R})$:**

$$\Psi_2(\psi) = \left\{ D_2^j T_k \psi = \psi(2^{-j} \cdot -k), j, k \in \mathbf{Z} \right\}$$

⌘ **ψ in $L^2(\mathbf{R})$ is an orthonormal wavelet if the above Affine system is an orthonormal basis for $L^2(\mathbf{R})$.**

⌘ **Shannon wavelet:** $\hat{\psi}(\xi) = e^{\pi i \xi} \chi_I(\xi)$, $I = [-1, \frac{1}{2}) \cup (\frac{1}{2}, 1]$.

⌘ **Haar wavelet:** $\psi(x) = \chi_{(\frac{1}{2}, 1]}(x) - \chi_{(0, \frac{1}{2}]}(x)$,

⌘ **Exercise:** $\hat{\psi}(\xi) = ?$ Express it using $\text{sinc}(\xi) = \frac{\sin \xi}{\xi}$.



Characterization equations

ψ in $L^2(\mathbf{R})$ is an orthonormal wavelet if and only if the following conditions hold:

⌘ Calderon condition:

$$\sum_{j \in \mathbf{Z}} |\hat{\psi}(2^j \xi)|^2 = 1, \text{ for a.e. } \xi \in \mathbf{R}.$$

⌘ “T-q” condition:

$$\sum_{j \in \mathbf{Z}_+} \hat{\psi}(2^j \xi) \overline{\hat{\psi}(2^j (\xi + q))} = 0, \text{ for all odd } q \text{ and a.e. } \xi \in \mathbf{R}.$$

⌘ **Exercise:** Verify that the above conditions hold for Shannon and Haar wavelet.



Types of dyadic wavelets

⌘ **MSF wavelets (wavelet sets).**

⌘ **MRA wavelets.**

⌘ **Daubechies wavelets.**

⌘ **Lemarié-Meyer wavelets.**

⌘ **Journé wavelet.**



Multiresolution Analysis (MRA)

MRA consists of a (nested) sequence of subspaces V_j in $L^2(\mathbf{R})$ satisfying:

⌘ **$V_j \subset V_{j+1}$ for all j in \mathbf{Z} ;**

⌘ **$f \in V_j$ if and only if $D_2 f \in V_{j-1}$ for all $j \in \mathbf{Z}$;**

⌘ **$\bigcap_{j \in \mathbf{Z}} V_j = \{0\}$;**

⌘ **$\overline{\bigcup_{j \in \mathbf{Z}} V_j} = L^2(\mathbf{R})$;**

⌘ **There exists a function $\varphi \in V_0$ such that $\{T_k \varphi, k \in \mathbf{Z}\}$ is an orthonormal basis for V_0 .**



Scaling equation, filters

⌘ **Scaling equation (time domain):**

$$\varphi\left(\frac{x}{2}\right) = \sum_{k \in \mathbf{Z}} c_k \varphi(x - k).$$

⌘ **Scaling equation (frequency domain):**

$$\hat{\varphi}(2\xi) = m_0(\xi) \hat{\varphi}(\xi)$$

⌘ **Smith-Barnwell equation:**

$$|m_0|^2(\xi) + |m_0|^2\left(\xi + \frac{1}{2}\right) = 1, \text{ for a.e. } \xi.$$

⌘ **Exercise:** Prove the S.-B. equation using $[\hat{\varphi}, \hat{\varphi}](\xi) = 1$, for a.e. ξ .



MRA wavelet construction

⌘ **High-pass filter:**

$$m_1(\xi) = e^{i\pi\xi} \overline{m_0(\xi + \frac{1}{2})}.$$

⌘ **Wavelet from scaling function:**

$$\hat{\psi}(2\xi) = m_1(\xi)\hat{\phi}(\xi)$$

⌘ **Fact:**

$$\psi \in W_0, V_1 = V_0 \oplus W_0.$$

⌘ **Exercise:** Prove the above fact by showing $[\hat{\psi}, \hat{\phi}](\xi) = 0$, for a.e. ξ .



Tensor product construction

⌘ Let φ and ψ be 1-d (dyadic) scaling function and wavelet.

⌘ 2-d “tensor product” scaling function and wavelets:

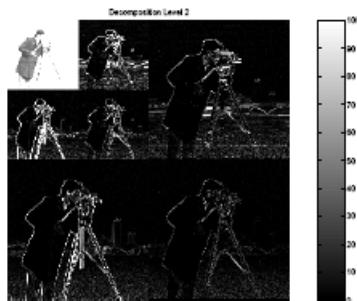
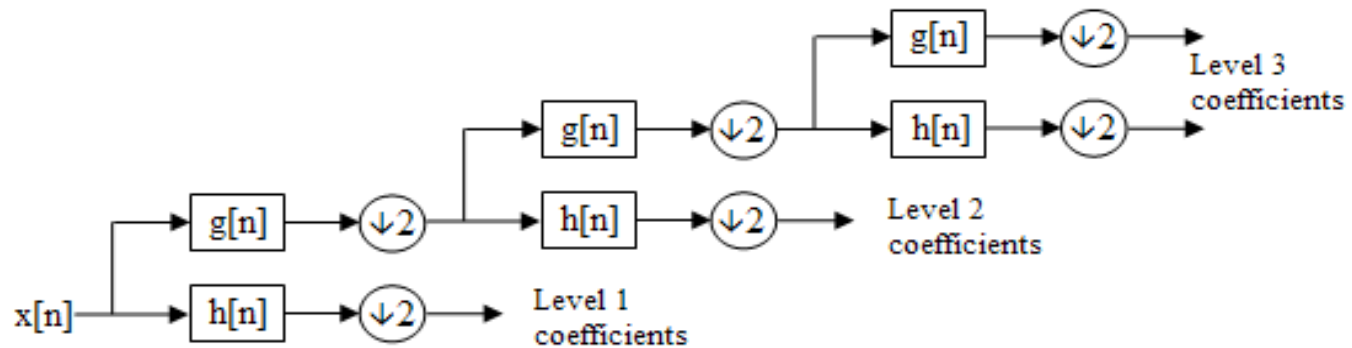
$$\Phi(\eta, \xi) = \varphi(\eta)\varphi(\xi), \quad \Psi_1(\eta, \xi) = \psi(\eta)\varphi(\xi),$$

$$\Psi_2(\eta, \xi) = \varphi(\eta)\psi(\xi), \quad \Psi_3(\eta, \xi) = \psi(\eta)\psi(\xi).$$

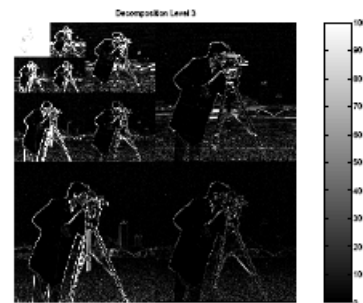
⌘ **Exercise:** Prove that the affine system generated by the three 2-d wavelet is indeed an o.n.b.

Image processing

⌘ Wavelet decomposition:



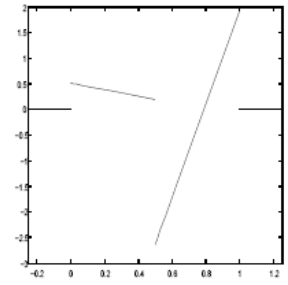
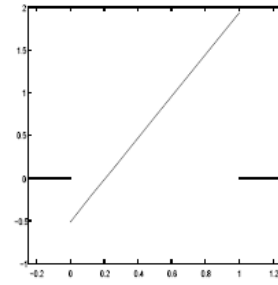
2 level Haar wavelet decomposition



3 level Haar wavelet decomposition

2-d wavelets

Composite wavelets

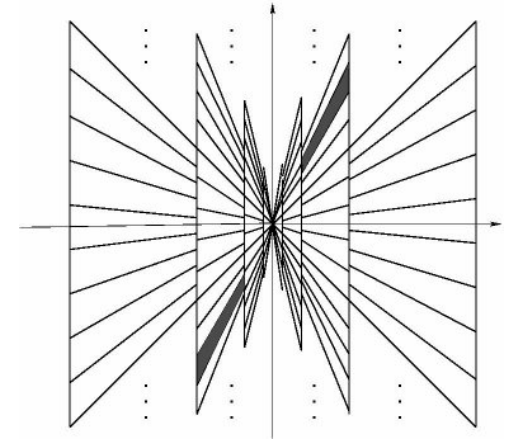


⌘ **Composite affine systems generated by ψ in $L^2(\mathbf{R}^d)$:**

$$\Psi_{a,b}(\psi) = \left\{ D_a^j D_b T_k \psi, j, k \in \mathbf{Z} \right\}$$

⌘ **Shearlets (<http://www.shearlet.org/>):**

$$B = \left\{ b = \begin{pmatrix} 1 & k \\ 0 & 1 \end{pmatrix}; k \in \mathbf{Z} \right\}$$



⌘ **Piece-wise linear 1-d example:**

$$\varphi(x) = \begin{cases} \frac{1}{\sqrt{2}} [2\sqrt{3}x + (1 - \sqrt{3})] & \text{for } x \in [0, 1], \\ 0 & \text{for } x \notin [0, 1] \end{cases}$$

“Blubbery Blob”

- ⌘ Low pass filter matrix in 1-d:

$$M_0(\xi) = \frac{1}{2} \begin{pmatrix} \alpha + \beta e^{-2\pi i \xi} & \gamma e^{-2\pi i \xi} + \delta e^{-4\pi i \xi} \\ \gamma e^{2\pi i \xi} + \delta e^{4\pi i \xi} & \alpha + \beta e^{2\pi i \xi} \end{pmatrix}$$

- ⌘ **Exercise:** Compute the “tensor-product” low-pass filter matrix for dilation $[0,2; 1,0]$.

