

WAVELETS:
A DIFFERENT WAY
TO LOOK AT
SUBBAND CODING.

Ingrid DAUBECHIES
AT&T Bell Laboratories

(until July 1st 1990:
visiting Mathematics Department
University of Michigan)

APRIL 30, 1990

WAVELETS and all that

"Wavelets"



Technique to cut up { data functions }
operators
into different frequency components, and
to study each component with a resolution
matched to their scale.

This technique was "invented" independently
in several different fields

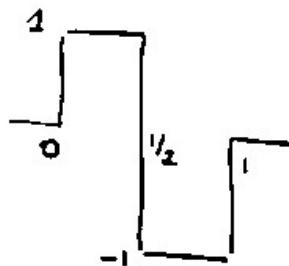
- pure mathematics: harmonic analysis
(Calderon)
- quantum mechanics: coherent states
(Aslateen - Klunder)
- engineering: signal analysis
(QMF filters - Esteban & Galland
Smith & Barnwell
Jean Morlet)

Recently (last three years): synthesis between
different approaches → very fertile for all branches.

9.

ORTHONORMAL BASES OF WAVELETS.

Old example: Haar basis.



$$\rightarrow \psi(x)$$

$$\psi_{jk}(x) = 2^{-j/2} \psi(2^{-j}x - k)$$

$$j, k \in \mathbb{Z}$$

orthonormal basis for $L^2(\mathbb{R})$

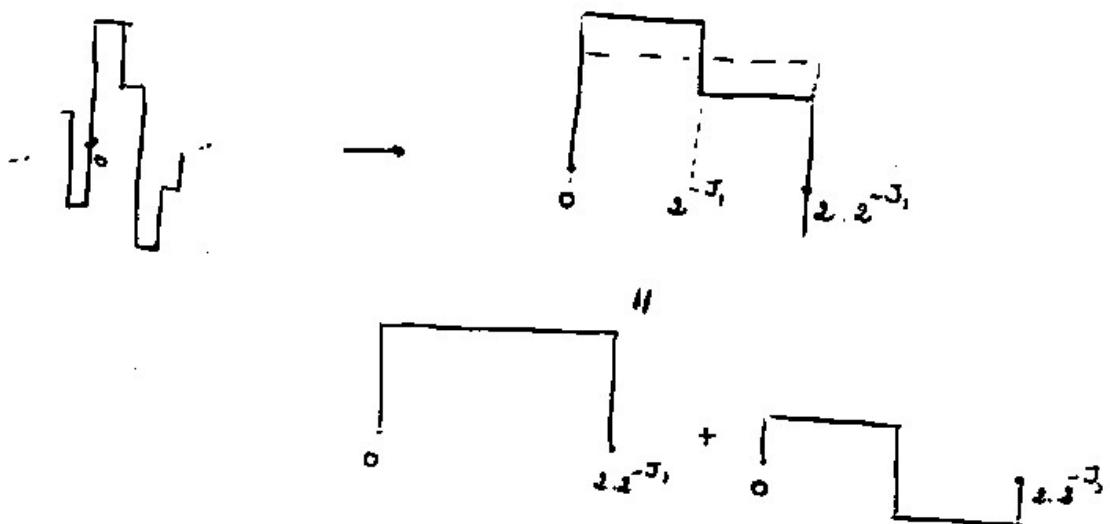
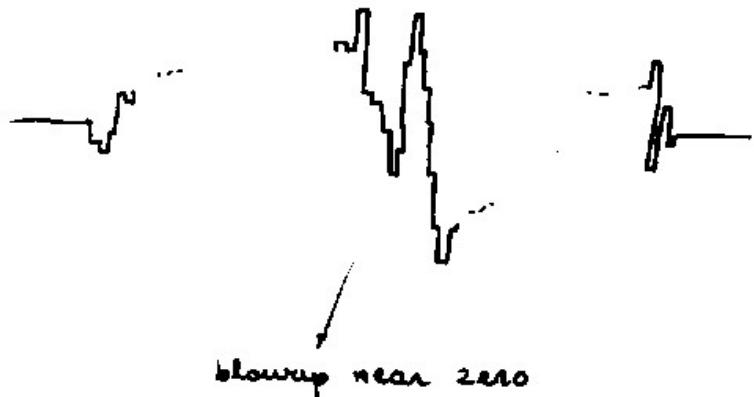
$$|\text{support } \psi| = 1$$

$$|\text{support } \psi_{jk}| = 2^j$$

$$\text{center of support } \psi_{jk} = (k + \frac{1}{2})2^j$$

Prog that ψ_{jk} constitute orthonormal basis?

Sufficient to show that functions with support $\subset [-2^{J_0}, 2^{J_0}]$, piecewise constant on intervals $[k2^{-J_1}, (k+1)2^{-J_1}]$, can be written as combination of ψ_{jk} .

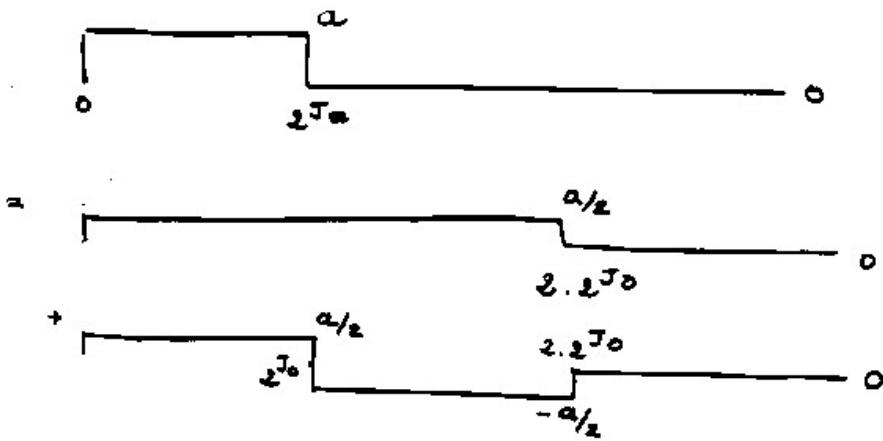


$$\Rightarrow f = \sum c_{j_1, l} \Psi_{j_1, l} + \text{fct. piecewise constant in } [k 2^{-(j_1-1)}, (k+1) 2^{-(j_1-1)}]$$

= ...

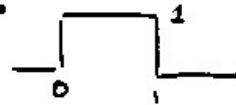
$$= \sum_{j=-J_1-J_0}^{J_1-1} c_{j, l} e^{\Psi_{j, l}}$$

↓
keep going!



$$= \sum_{k=1}^K 2^{-k} a \psi(2^{-J_0-k} x) + 2^{-K} a \phi(2^{-J_0-K} x)$$

where ϕ is



$$\text{But } \| 2^{-K} a \phi(2^{-J_0-K} \cdot) \|_{L^2}^2 \approx 2^{-2K} (a)^2 2^{J_0+K} \rightarrow 0 \text{ for } K \rightarrow \infty.$$

⇒ done!

In fact: proof uses multiresolution analysis.

- introduces "averaging" function ϕ
- space V_j spanned by $\phi(2^{-j} x - k)$
- $V_j \subset V_{j-1}$
- $\text{Proj}_{V_{j-1}} f = \text{Proj}_{V_j} f + \text{expansion in the } \psi_k$.

. Orthonormal bases of wavelets.

For very special ψ :

$$2^{-m/2} \psi(2^{-m}t - n) = \psi_{mn}(t)$$

are orthonormal basis.

N.B. $b_0 = 1$ \Rightarrow not really a restriction

$a_0 = 2$: computationally easy.

other a_0 also possible. (in fact
all rational values are allowed)

These are associated to a beautiful
mathematical construction:

Multiresolution analysis. (S. Mallat,
Y. Meyer)

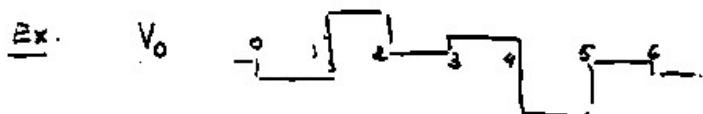
. ladder of spaces

$$\dots \subset V_1 \subset V_0 \subset V_{-1} \subset \dots$$

$$\bigcap_{m \in \mathbb{Z}} V_m = \{0\}$$

$$\bigcup_{m \in \mathbb{Z}} V_m = L^2(\mathbb{R}).$$

V_d : describes functions in which all scales
finer than 2^d are left out.



- general framework for construction of orthonormal wavelet bases.

$$\cdots \subset V_2 \subset V_1 \subset V_0 \subset V_{-1} \subset V_{-2} \subset \cdots$$

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

$$f \in V_j \iff f(2^j \cdot) \in V_0$$

$\exists \phi \in V_0$ so that ϕ_{0k} are o.n. basis for V_0

$$\phi_{0k}(x) = \phi(x - k).$$

MULTIRESOLUTION ANALYSIS.

Then: \exists associated orthonormal wavelet basis.

W_0 : orthogonal complement in V_1 of V_0
 $V_0 \oplus W_0 = V_1$, $W_0 \perp V_0$.

$\exists \psi$ in W_0 so that ψ_{0k} are orthonormal basis for W_0

$$\Rightarrow \text{Proj}_{V_1} = \text{Proj}_{V_0} + \text{expansion in } \psi_{0k}.$$

$$V_1 \subset V_0 \subset V_{-1} \subset V_{-2} \subset \cdots$$

$$W_0 \subset W_1 \subset W_{-1} \subset W_{-2} \subset \cdots$$

$$f \in W_j \iff f(2^j x) \in W_0.$$

$\Rightarrow W_i$: all orthogonal, and $\bigoplus W_i = L^2(\mathbb{R})$

- W_j dilated version of W_0
 \downarrow
 o.n. basis $\psi(x-k)$
- $\Rightarrow \psi_{jk}(x) = e^{-\delta/2} \psi(2^{-j}x-k) \quad k \in \mathbb{Z}$
 o.n. basis in W_j
- $\Rightarrow \{\psi_{jk}; j, k \in \mathbb{Z}\}$ o.n. basis for $L^2(\mathbb{R})$
- recipe for ψ :
 - $\phi \in V_0 \subset V_{-1}$ ← o.n. basis $\phi_{-1,k}$
 - $\phi(x) = \sum_n h_n \phi_{-1,n}(x)$
 $= \sqrt{2} \sum_n h_n \phi(2x-n)$
 - $h_n = \langle \phi, \phi_{-1,n} \rangle$
 - $\psi(x) = \sqrt{2} \sum_n (-1)^n h_{-n+1} \phi(2x-n)$

To prove existence + recipe for ψ , analyze in detail what W_0 really represents.

A crucial role in this analysis is played by the trigonometric polynomial

$$m_0(\xi) = \frac{1}{\sqrt{2}} \sum_n h_n e^{-2\pi i n \xi}.$$

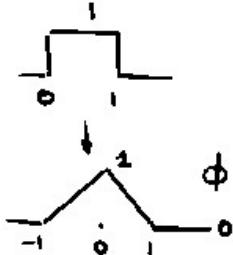
$$|m_0(\xi)|^2 + |m_0(\xi + \frac{1}{2})|^2 = 1.$$

to generalize Haar basis:

generalize the associated multiresolution analysis.

Two paths.

generalize
piecewise constant"
 ↓
 linear
 quadratic
 cubic
 (splines)



But ϕ_{ok} not orthonormal!

↳ orthonormalization
trick.

$$\hat{\phi}(z) = \frac{\phi(z)}{\left(\sum_{k} |\phi(z+k)|^2 \right)^{1/2}}$$

$\hat{\phi}_{ok}$ orthonormal
span same space
as ϕ_{ok}

→ can be used to
construct Ψ .

orthonormalization trick loses

compact support wanted
arbitrarily high regularity.

only finite # of h_n allowed.

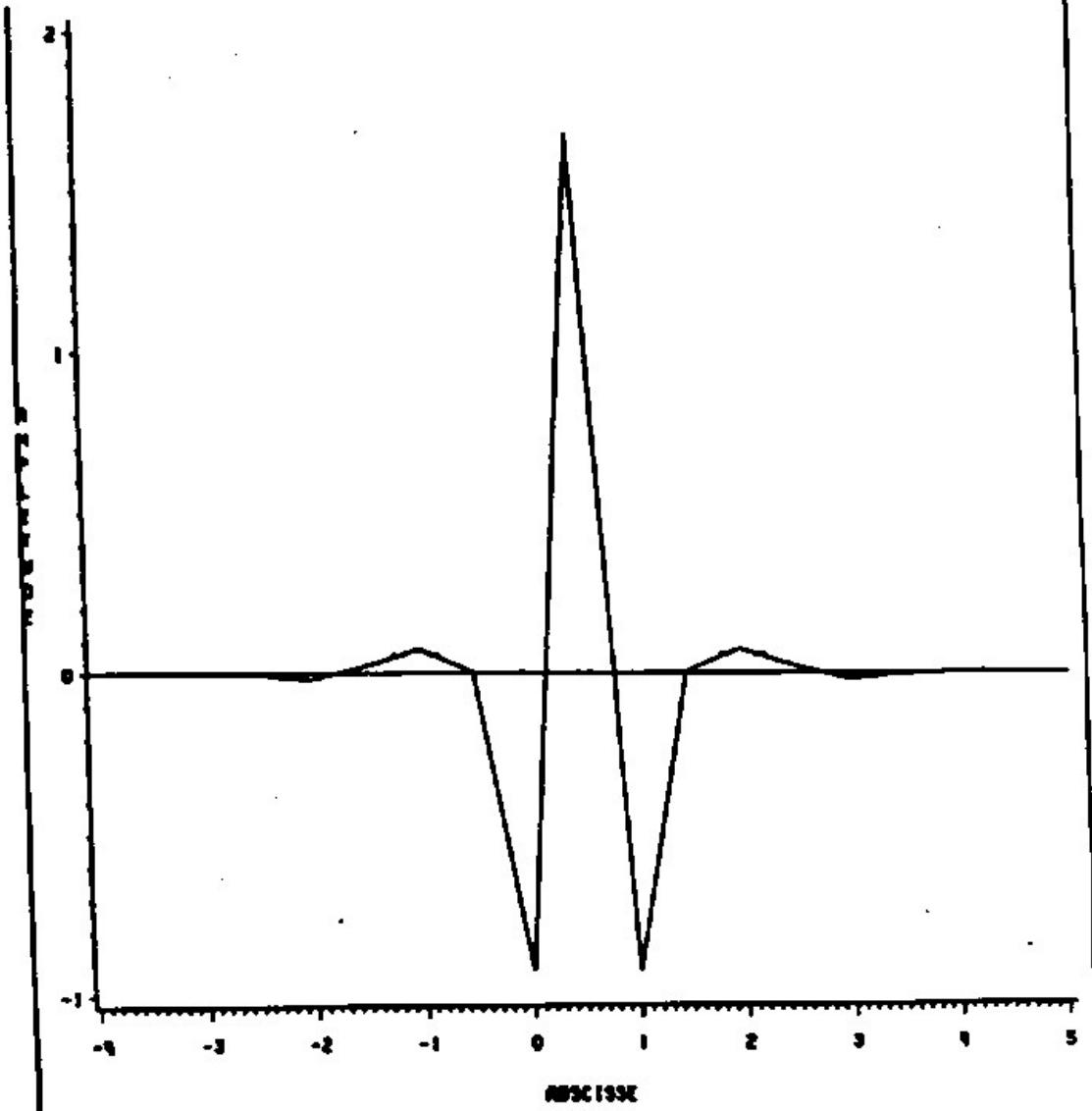
$$m_0(z) = \frac{1}{\sqrt{2}} \sum_n h_n e^{-2\pi i n z}$$

$$|m_0(z)|^2 + |m_0(z+1/2)|^2 = 1$$

$$\begin{aligned} \hat{\Phi}(z) &= m_0(z/2) \hat{\phi}(z/2) \\ &= \prod_{j=1}^{\infty} m_0(2^{-j} z) \end{aligned}$$

- strategy to construct m_0 so that infinite product has decay
- check that strategy works!

ONDELETTE:



• Wanted: $m_0(\xi) = \sum_{n=0}^{-N} \frac{1}{\sqrt{n}} h_n e^{-2\pi i n \xi}$

$$|m_0(\xi)|^2 + |m_0(\xi + \frac{1}{2})|^2 = 1$$

$\prod_{j=1}^{\infty} m_0(2^j \xi)$ decays for $|\xi| \rightarrow \infty$.

• $m_0(\xi) = \left(\frac{1 + e^{-2\pi i \xi}}{2} \right)^L \mathcal{E}(\xi)$

$$\frac{1 + e^{-2\pi i \xi}}{2} = e^{-\pi i \xi} \cos \pi \xi$$

$$\prod_{j=1}^{\infty} \cos(2^j \pi \xi) = \frac{\sin \pi \xi}{\pi \xi}$$

⇒ if sufficient control over $\prod_{j=1}^{\infty} \mathcal{E}(2^j \xi)$,
then ϕ will have good decay.

• $\Rightarrow (\cos \pi \xi)^{2L} \underbrace{|\mathcal{E}(\xi)|^2}_{\text{polynomial in } \cos 2\pi \xi} + (\sin \pi \xi)^{2L} |\mathcal{E}(\xi + \frac{1}{2})|^2 = 1$

↓
→ polynomial in $\sin^2 \pi \xi$

$$(1-y)^L P(y) + y^L P(1-y) = 1$$

$$P(y) = \frac{1}{(1-y)^L} + O(y^L)$$

$$= \sum_{l=0}^{L-1} \binom{L-1+l}{l} y^l + O(y^L)$$

$$\Rightarrow |\mathcal{F}(\xi)|^2 = \sum_{l=0}^{L-1} \binom{L-1+l}{l} (\cos \pi \xi)^{el}$$

- use a lemma by Riesz to "extract square root"

$$\rightarrow \mathcal{F}(\xi) = \sum_{l=0}^{L-1} f_l e^{-2\pi i \xi l}$$

f_l real.

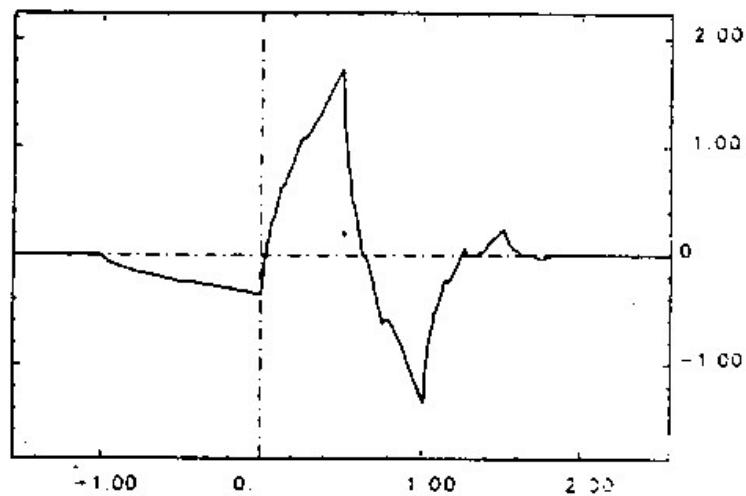
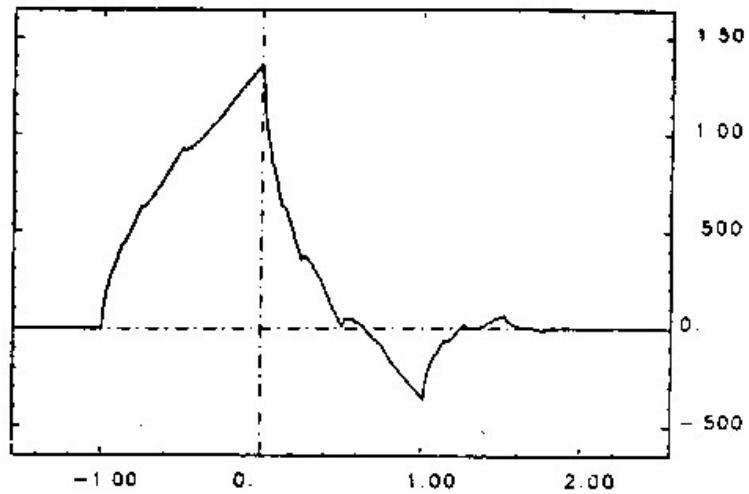
- $\prod_{j=1}^{\infty} |\mathcal{F}(2^{-j}\xi)| \leq C(1+|\xi|)^{-\mu L}$

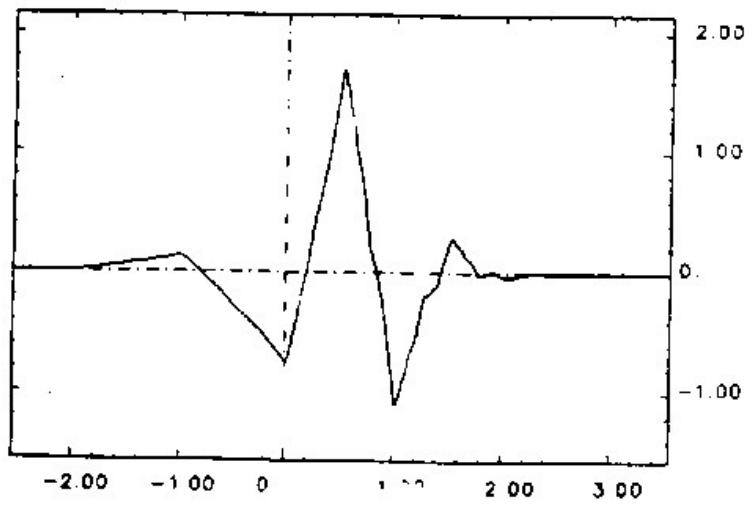
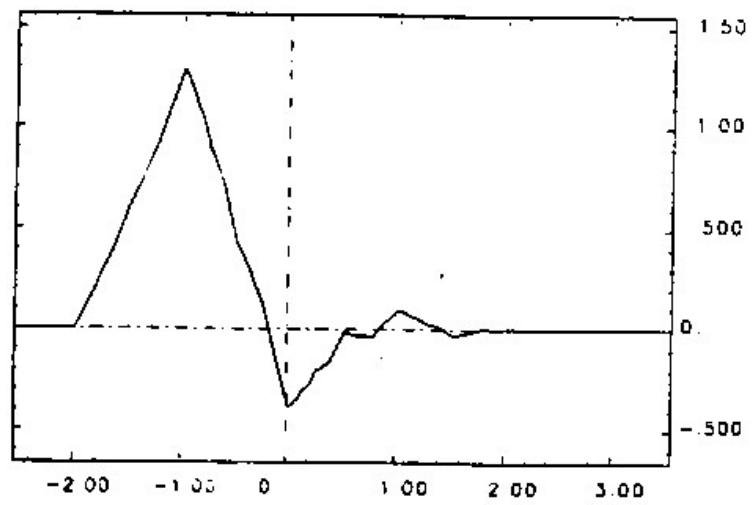
$$\mu \approx 0.81$$

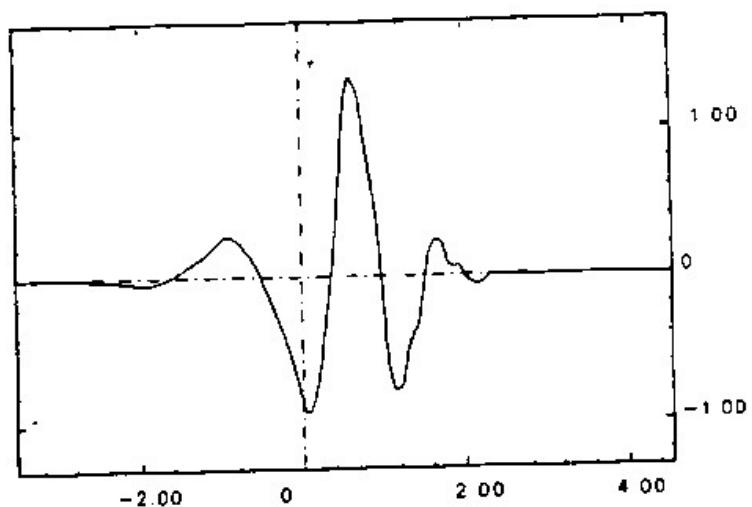
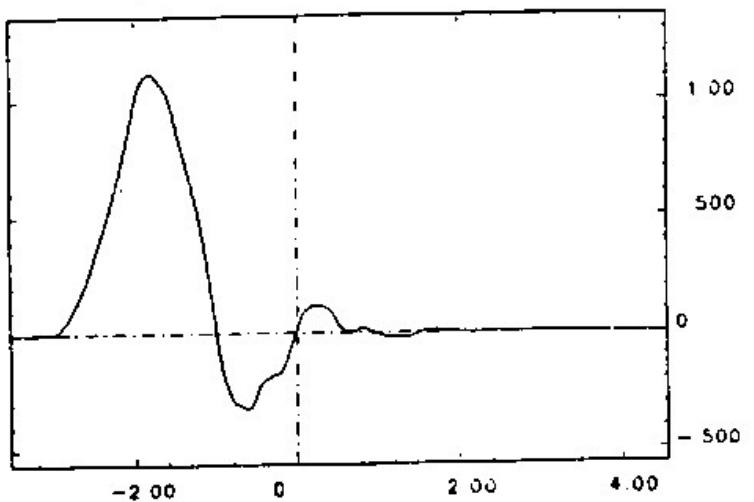
$$\Rightarrow |\hat{\phi}(\xi)| \leq C(1+|\xi|)^{-\nu L}$$

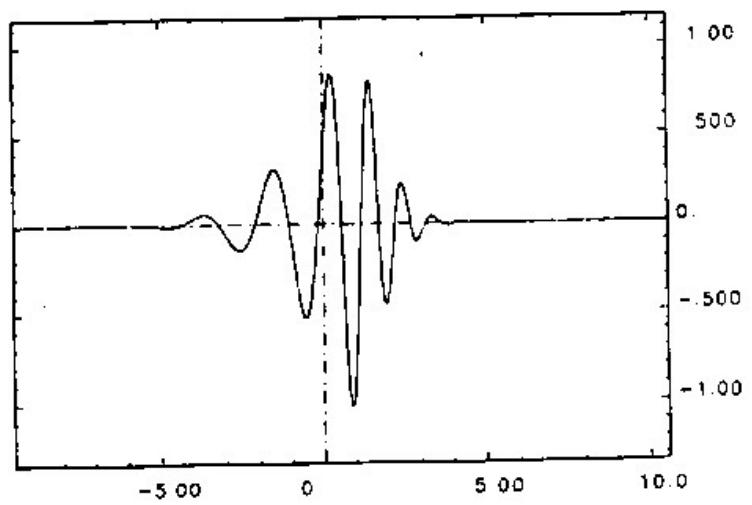
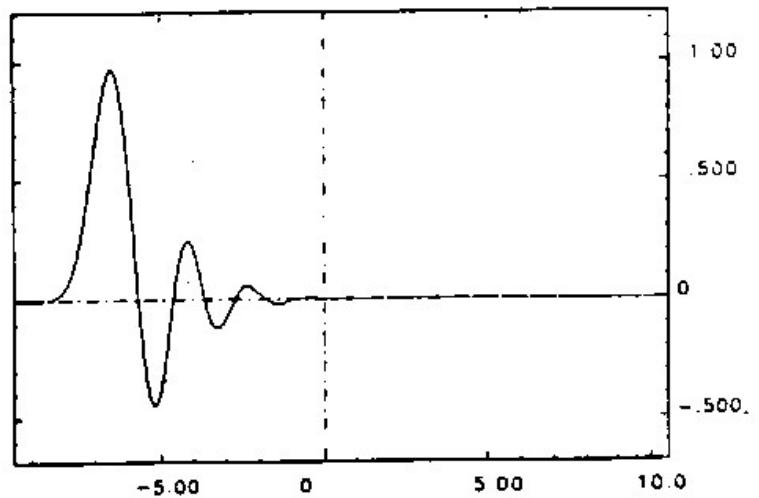
$$\nu \approx 0.19$$

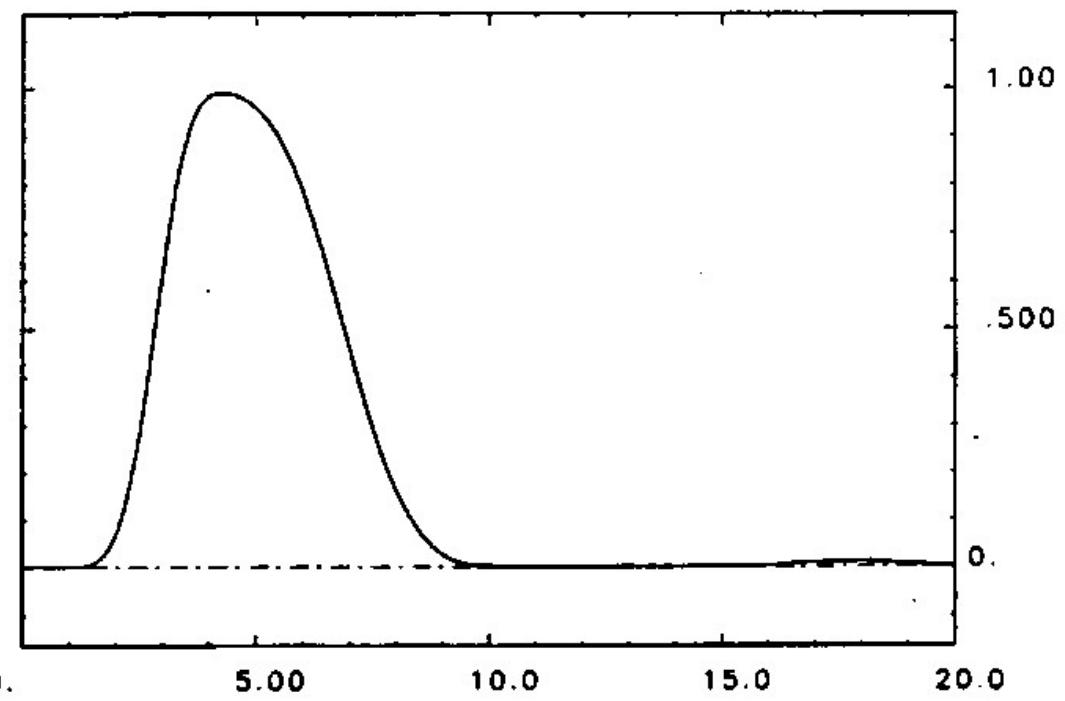
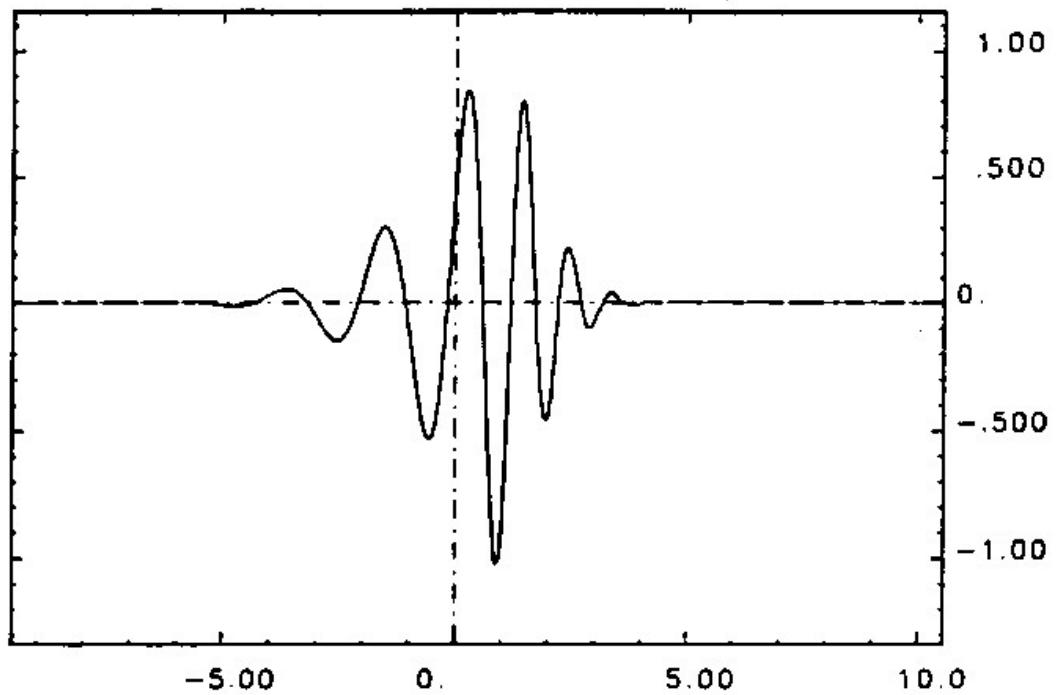
\Rightarrow arbitrarily high regularity!











What does all this have to do with
subband coding?

$$f \in V_0 \quad f = \sum_n f_n \phi_{0n}$$

$$V_1 \oplus W_1$$

$$f = s + d \quad s = \sum_k s_k \phi_{1k}$$

$$d = \sum_k d_k \psi_{1k}$$

$\phi_{1k} = \sum_n h_{n-2k} \phi_{0n}$
$\psi_{1k} = \sum_n g_{n-2k} \phi_{0n}$

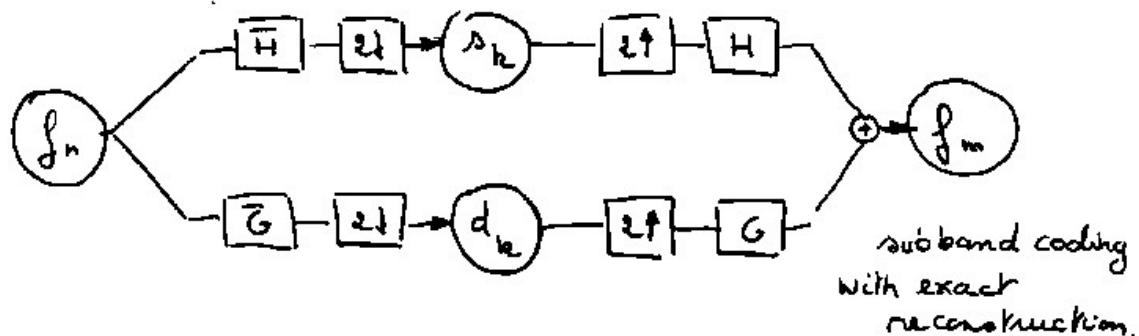
$$g_n = (-1)^n h_{-n+1}$$

$$\Rightarrow s_k = \sum_n h_{n-2k} f_n$$

$$d_k = \sum_n g_{n-2k} f_n$$

Inverse transform: uses transposed matrix

$$f_m = \sum_k [h_{m-2k} s_k + g_{m-2k} d_k]$$

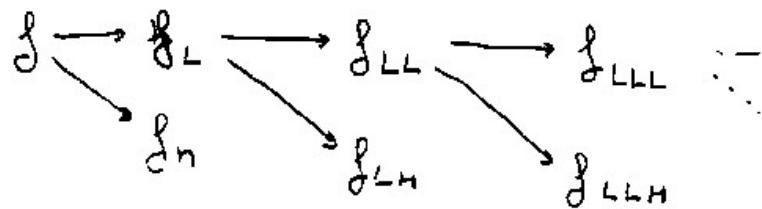


→ the "filter coefficients" h_n, g_n coming from an orthonormal basis of wavelets correspond exactly to the filters in an exact reconstruction subband coding scheme.

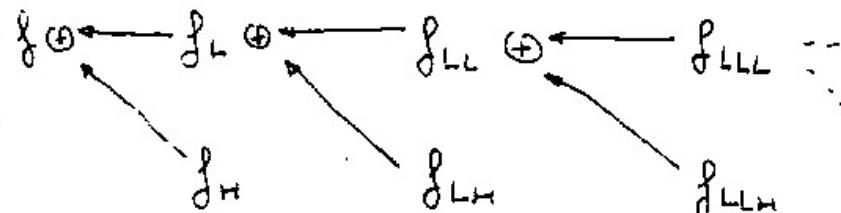
(→ compact support for ϕ, ψ important!
leads to FIR filters).

What role does regularity play?

Suppose you have the decomposition



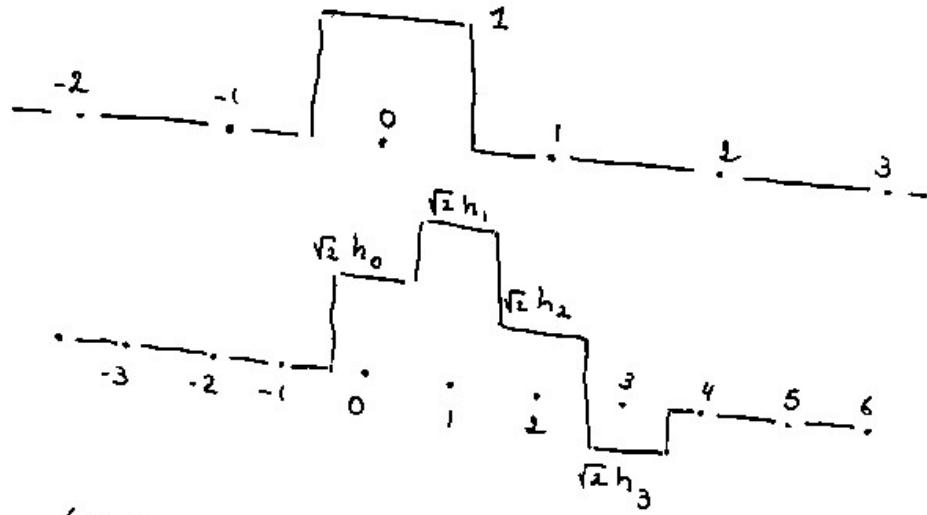
with reconstruction



What does a sequence

$$f_{LLL} \dots 0 0 0 1 0 0 0 \dots$$

correspond to?



$$(f_0)_n = \delta_{n,0}$$

$$(f_j)_k = \sqrt{2} \sum_n h_{n-2k} (f_{j-1})_n$$

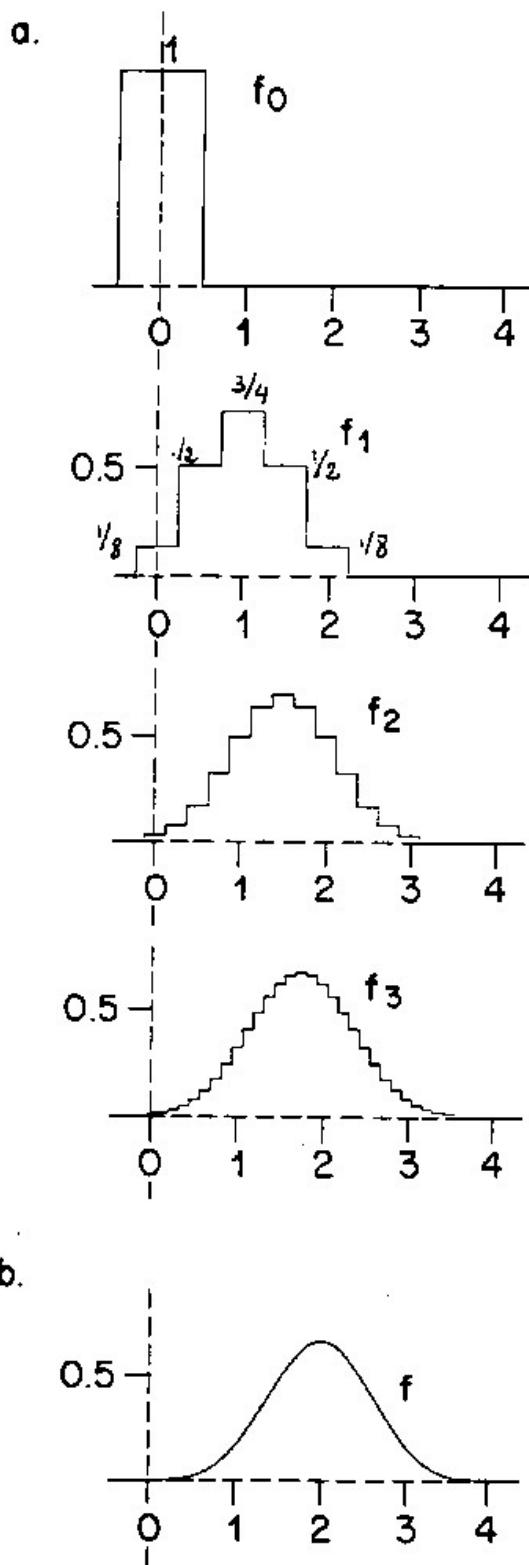
$F_j(x)$: piecewise constant on $[x \cdot 2^{-j}, (x+1) \cdot 2^{-j}]$

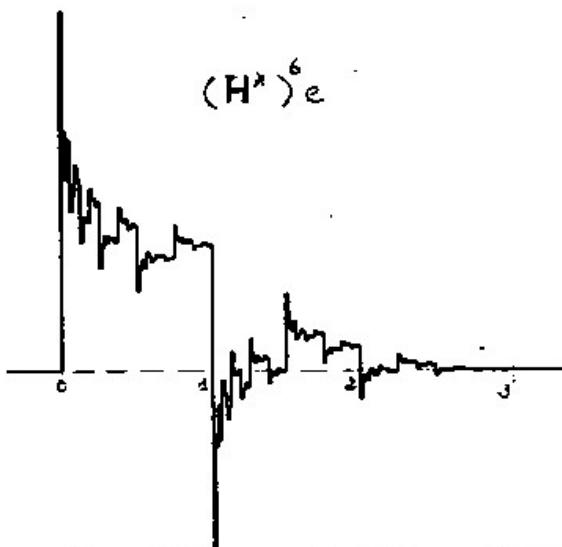
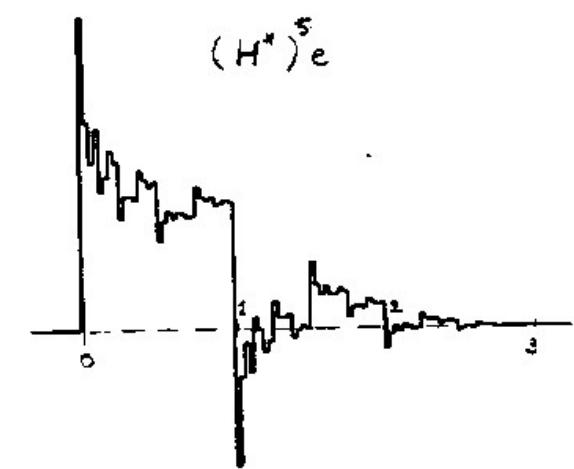
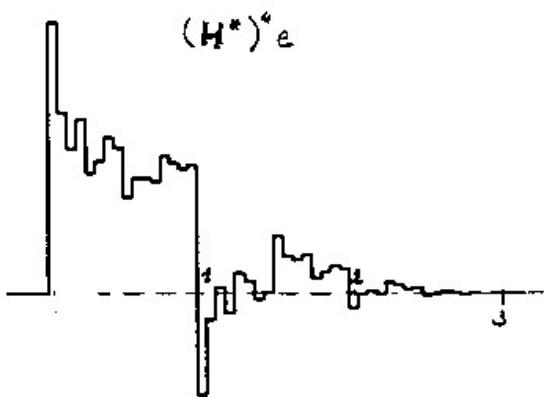
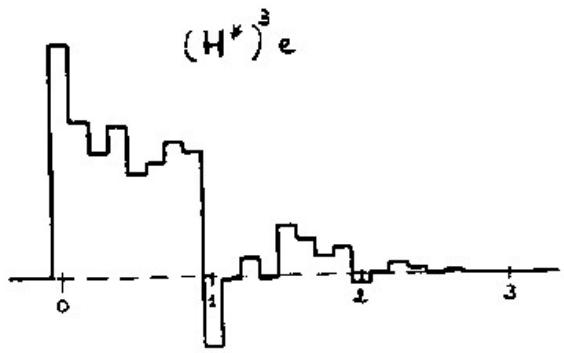
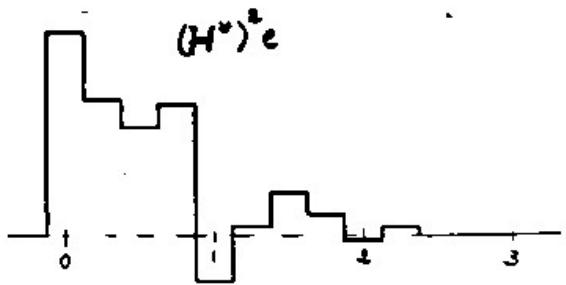
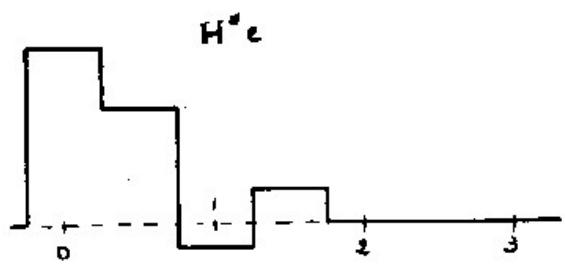
$$F_j(x) = \sqrt{2} \sum_n h_n F_{j-1}(2x-n)$$

$\underset{j \rightarrow \infty}{\longrightarrow}$ fixed point of T

$(TF)(x) = \sqrt{2} \sum_n h_n F(2x-n)$.
But this fixed point is ϕ !

$$\phi(x) = \sqrt{2} \sum_n h_n \phi(2x-n)$$





\Rightarrow Regularity is a good idea.

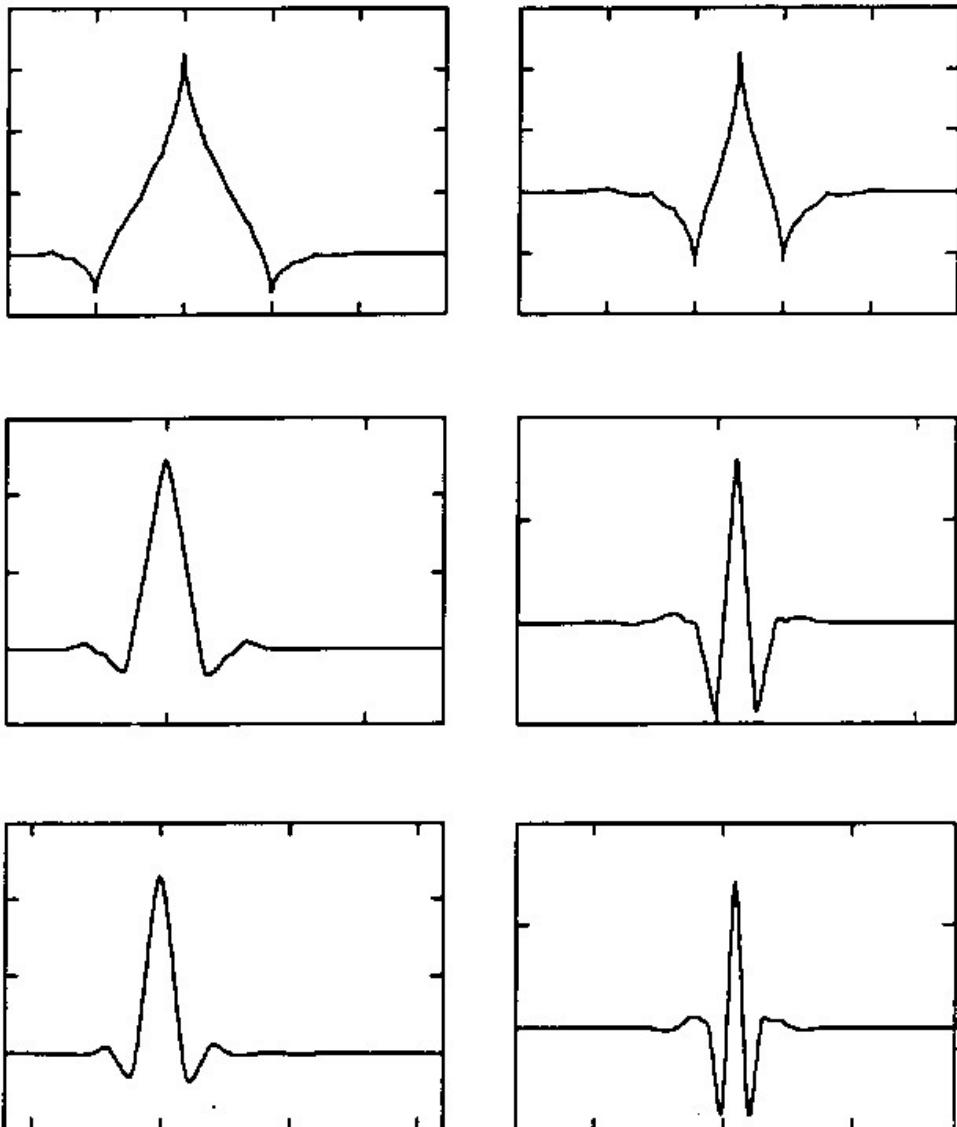
Regularity forces constraints on filters.

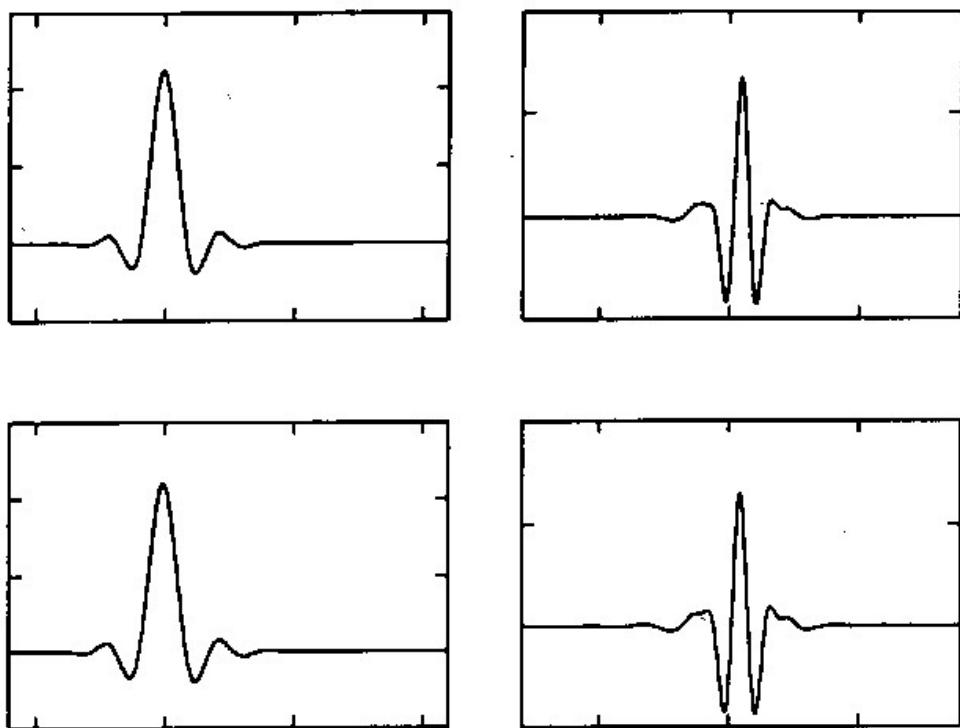
$$\phi, \psi \in C^k$$

$$\Rightarrow m_0(\zeta) = \left(\frac{1 + e^{-2\pi i \zeta}}{2} \right)^k \vartheta(\zeta).$$

lowpass filter has zero of order k
at $\zeta = 1/2$.

Same is true for generalizations of orthonormal wavelet bases to higher dimensions, or to other dilation factors than 2.





Bioorthogonal wavelet bases.

- $f = \sum_{j,k} \langle f, \psi_{jk} \rangle \tilde{\psi}_{jk}$
correspond to analysis filters
+ synthesis filters
- symmetric $\psi, \tilde{\psi}$ possible (\rightarrow linear phase filters!)
- regularity constraints:

$$\psi \in C^k \Rightarrow \int dx x^l \tilde{\psi}(x) = 0 \quad l=0,..,k-1$$

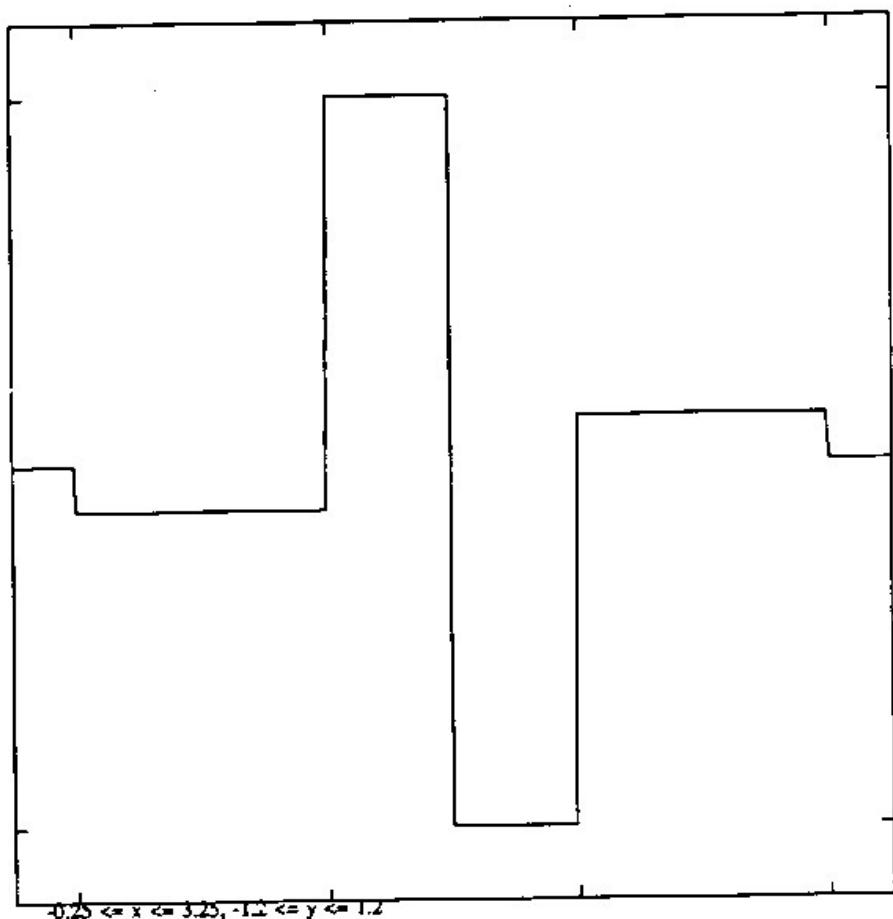
$$\Rightarrow m_0(\xi) = \left(\frac{1 + e^{-2\pi i \xi}}{\xi} \right)^k \tilde{f}(\xi)$$

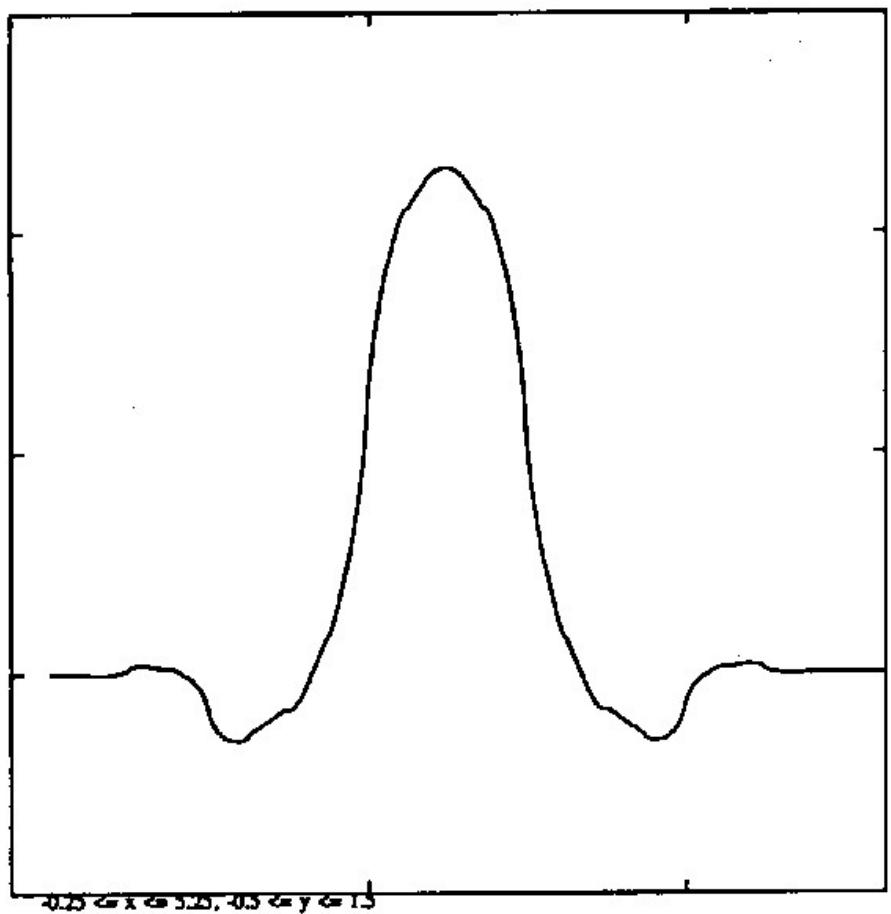
regularity on both $\psi, \tilde{\psi}$

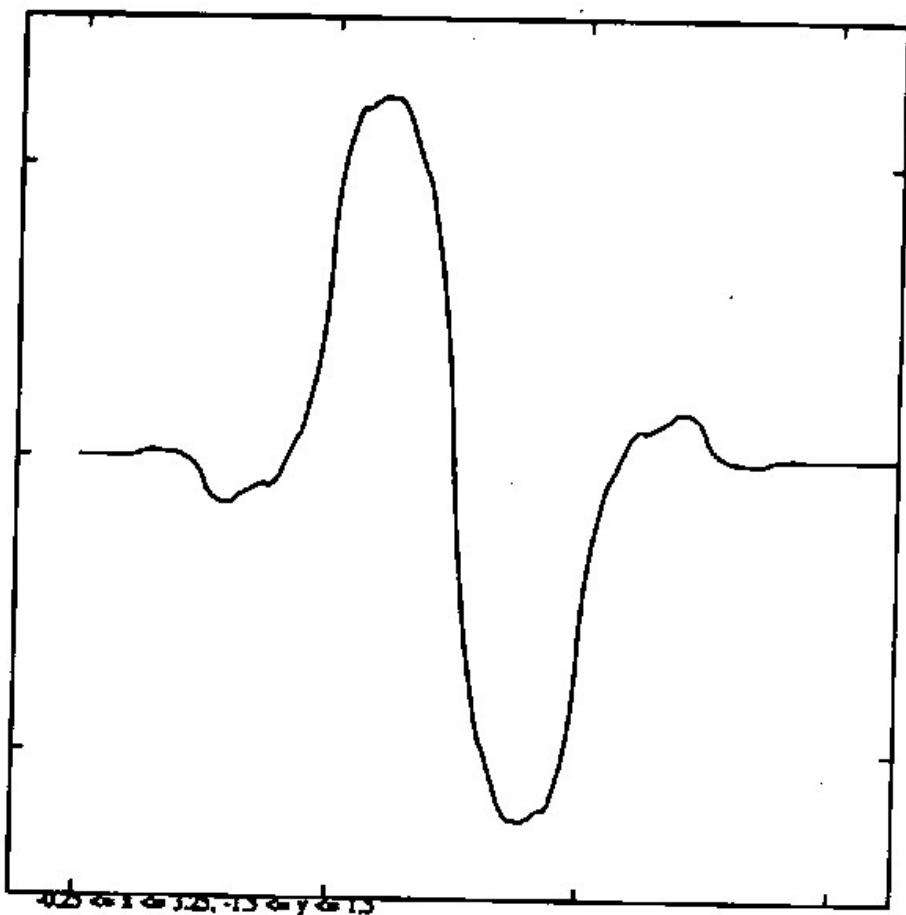
\rightarrow both m_0, \tilde{m}_0 need factorization of this kind.

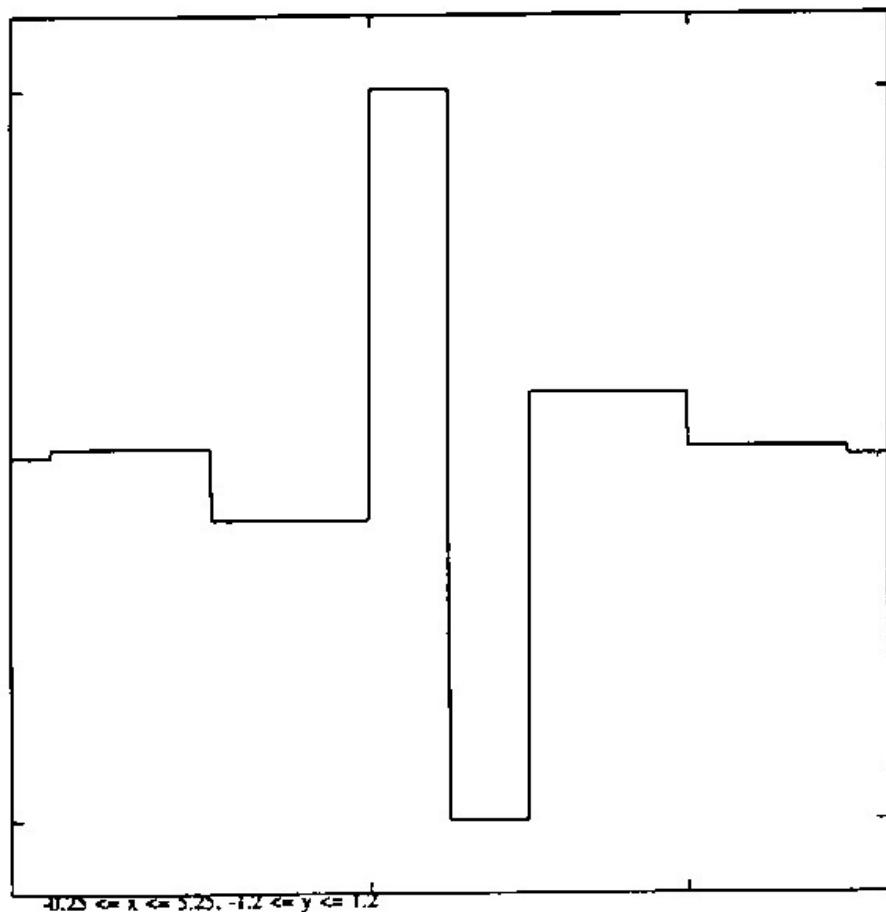
Examples:

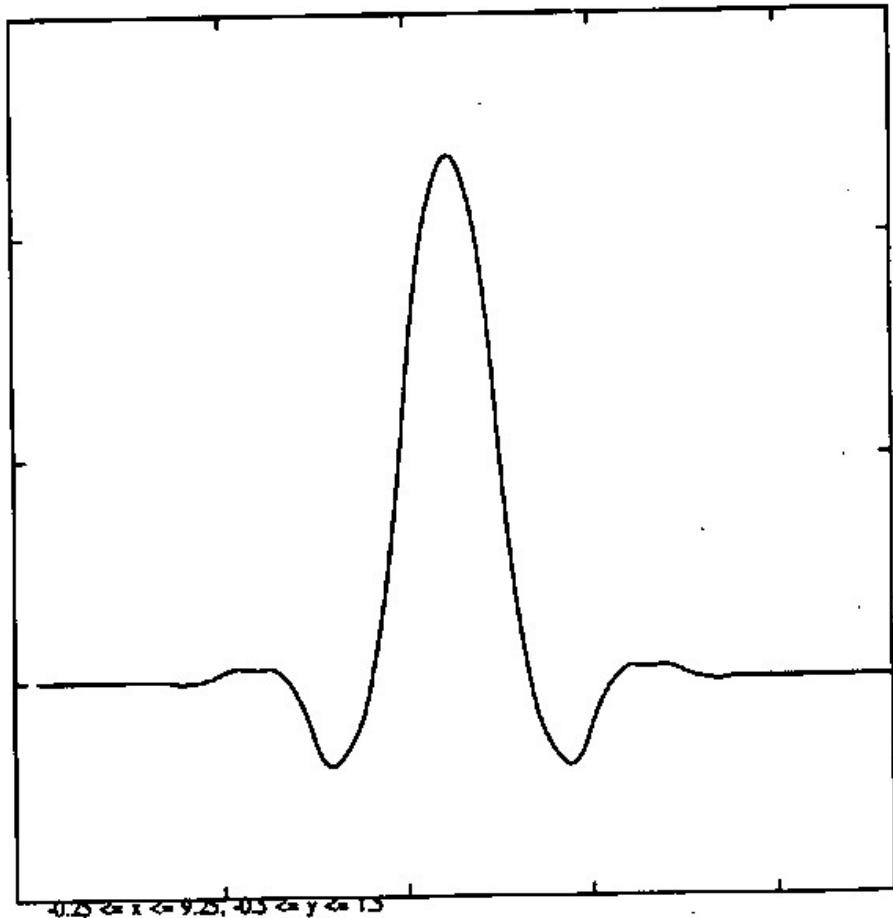
- m_0 binomial (\Rightarrow B-spline)
- family of more and more regular \tilde{m}_0
- rearrange previous examples (cf Vetterli)
- m_0, \tilde{m}_0 both very close to orthonormal case
 $\approx m_0(\xi) = -.05e^{-4\pi i \xi} + .25e^{-2\pi i \xi} + .6 + .25e^{2\pi i \xi} - .05e^{4\pi i \xi}$

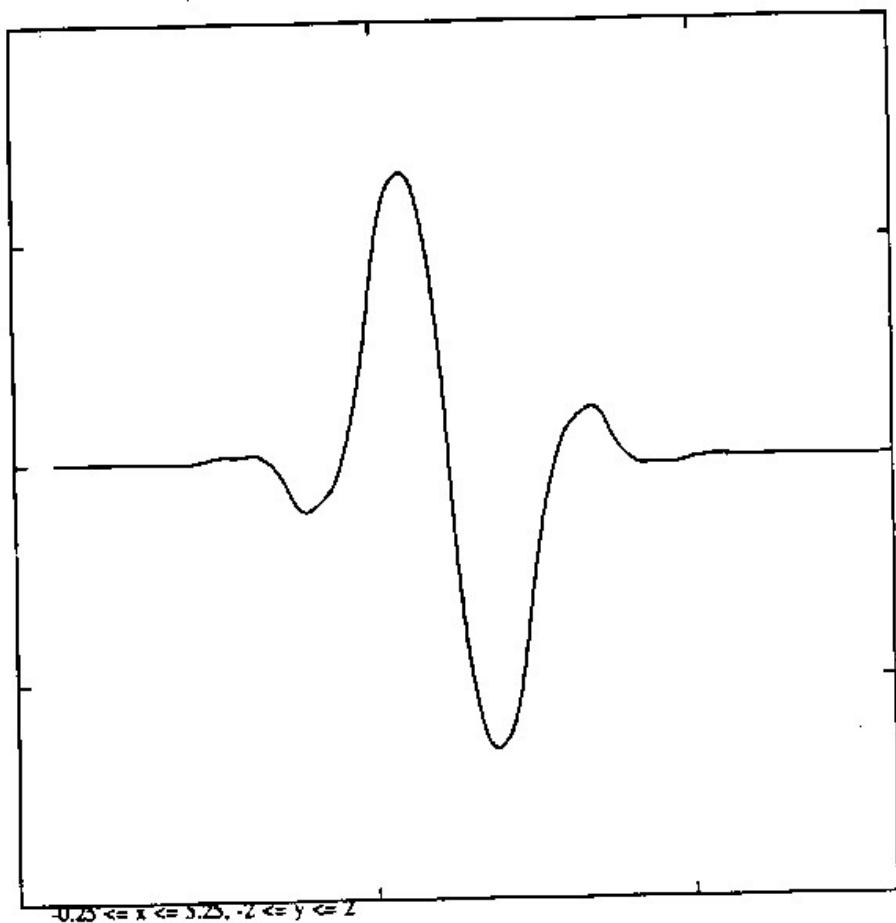












"Wavelets": more than just orthonormal bases!

Continuous case

$$\psi_{a,b}(x) = \frac{1}{\sqrt{a}} \psi\left(\frac{x-b}{a}\right)$$

$$\int dx \psi(x) = 0 \quad \psi \text{ symmetric}$$

$$f = C_4^{-1} \int_0^\infty \frac{da}{a^2} \int_{-\infty}^\infty db \quad \langle f, \psi_{ab} \rangle \psi_{ab}$$

$$\langle f, \psi_{ab} \rangle = \int dr f(r) \psi_{ab}(r)$$

Frames

$$a = a_0^m \quad m \in \mathbb{Z} \quad (a_0 > 1 \text{ fixed})$$

$$b = n b_0 \quad a_0^m \quad n \in \mathbb{Z} \quad (b_0 > 0 \text{ fixed})$$

$$\psi_{mn}(x) = a_0^{-m/2} \psi(a_0^{-m}x - nb_0)$$

2 dual points of view:

- characterize f by $\langle f, \psi_{mn} \rangle$

- find α_{mn} so that $f = \sum_{m,n} \alpha_{mn} \psi_{mn}$.

CONCLUSION.

- Subband coding with exact reconstruction
 - ↔ orthonormal wavelet bases
 - biorthogonal wavelet bases
- Regularity !
- More to wavelets than subband coding .