

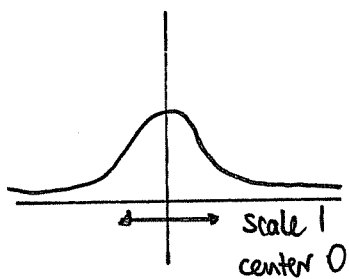
Multi-Resolution Approximation

$$V_0 = \text{span} \{ \varphi(x-j) : j = -\infty \dots \infty \}$$

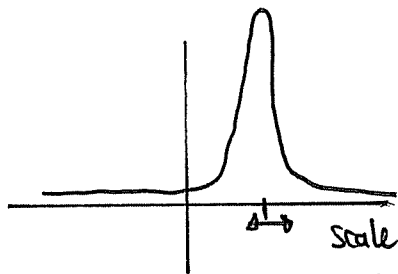
$\varphi(x)$ = scaling function

$$\langle \varphi(x), \varphi(x-j) \rangle = \delta_{0j} = \begin{cases} 1 & j=0 \\ 0 & j \neq 0 \end{cases}$$

$$\varphi_j^k(x) = 2^{k/2} \varphi(2^k x - j)$$



$\varphi(x)$



$\varphi_3^2(x)$

scale $2^{-2} = 1/4$
center $3 \cdot \frac{1}{4} = 3/4$

$$V_k = \text{span} \{ \varphi_j^k(x) : j = -\infty \dots \infty \}$$

$$\dots \subset V_{-1} \subset V_0 \subset V_1 \subset V_2 \subset \dots$$

with

$$\overline{\bigcup_k V_k} = L^2$$

$$\bigcap_k V_k = \{0\}$$

$P_k f(x)$ = \perp projection of $f \in L^2$ onto V_k

$$= \sum_j \langle f, \varphi_j^k \rangle \varphi_j^k(x)$$

interpreted as approximation to f
at resolution 2^{-k} .

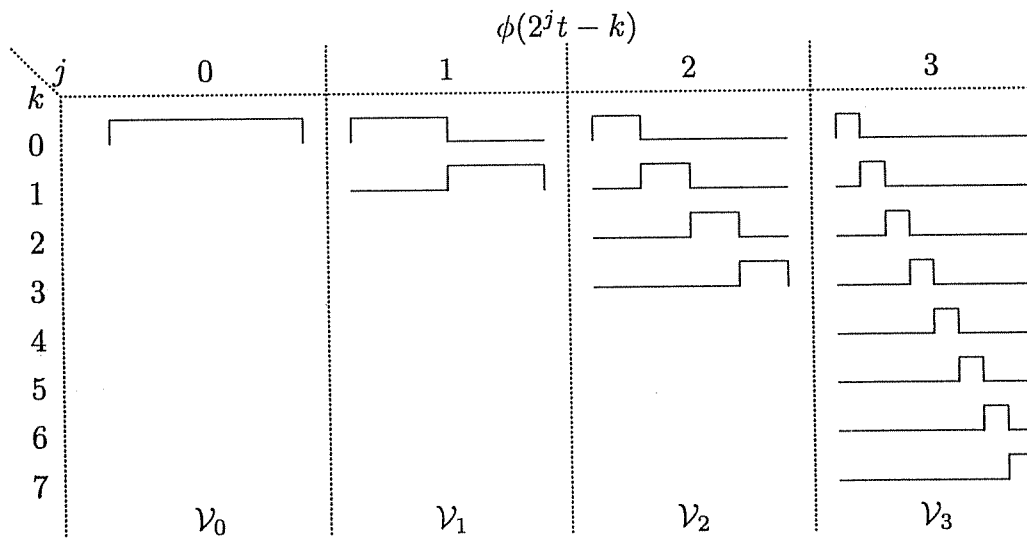


Figure 2.11. Haar Scaling Functions that Span V_j

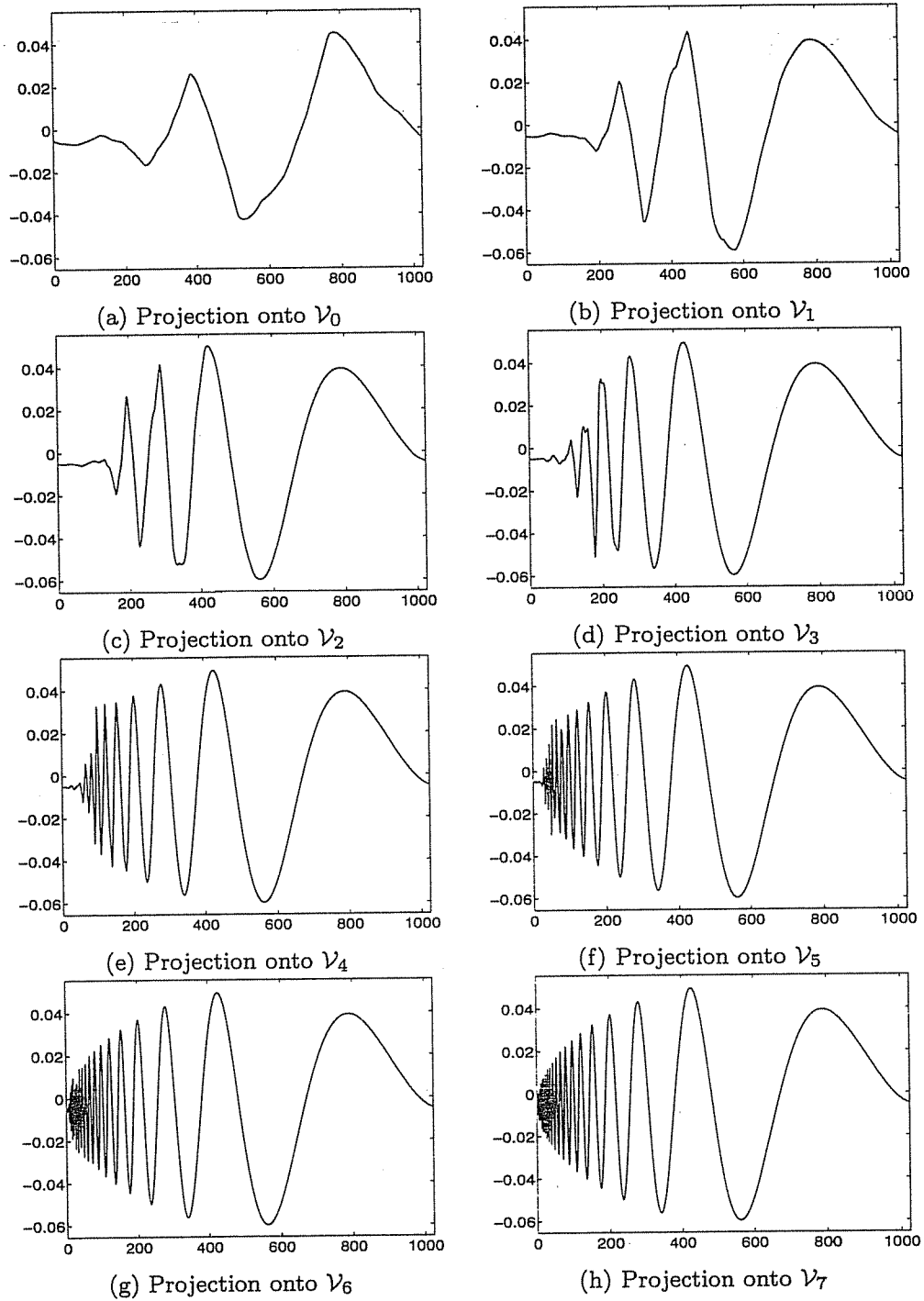


Figure 2.9. Projection of the Doppler Signal onto \mathcal{V} Spaces using ϕ_{D8}

$W_0 = \perp$ complement of V_0 in V_1

$$V_0 \oplus W_0 = V_1$$

Usually

$$W_0 = \text{span} \{ \psi(x-j) : j = -\infty \dots \infty \}$$

$$\langle \psi(x), \psi(x-j) \rangle = \delta_{0j}$$

$$\langle \psi(x), \psi(x-j) \rangle = 0$$

$\psi(x) =$ mother wavelet

Then

$$V_k \oplus W_k = V_{k+1}$$

$$W_k = \text{span} \{ \psi_j^k(x) : j = -\infty \dots \infty \}$$

$Q_k f(x) = \perp$ projection of f onto W_k

$$= \sum_j \langle f, \psi_j^k \rangle \psi_j^k(x)$$

interpreted as fine detail in f at resolution 2^{-k} .

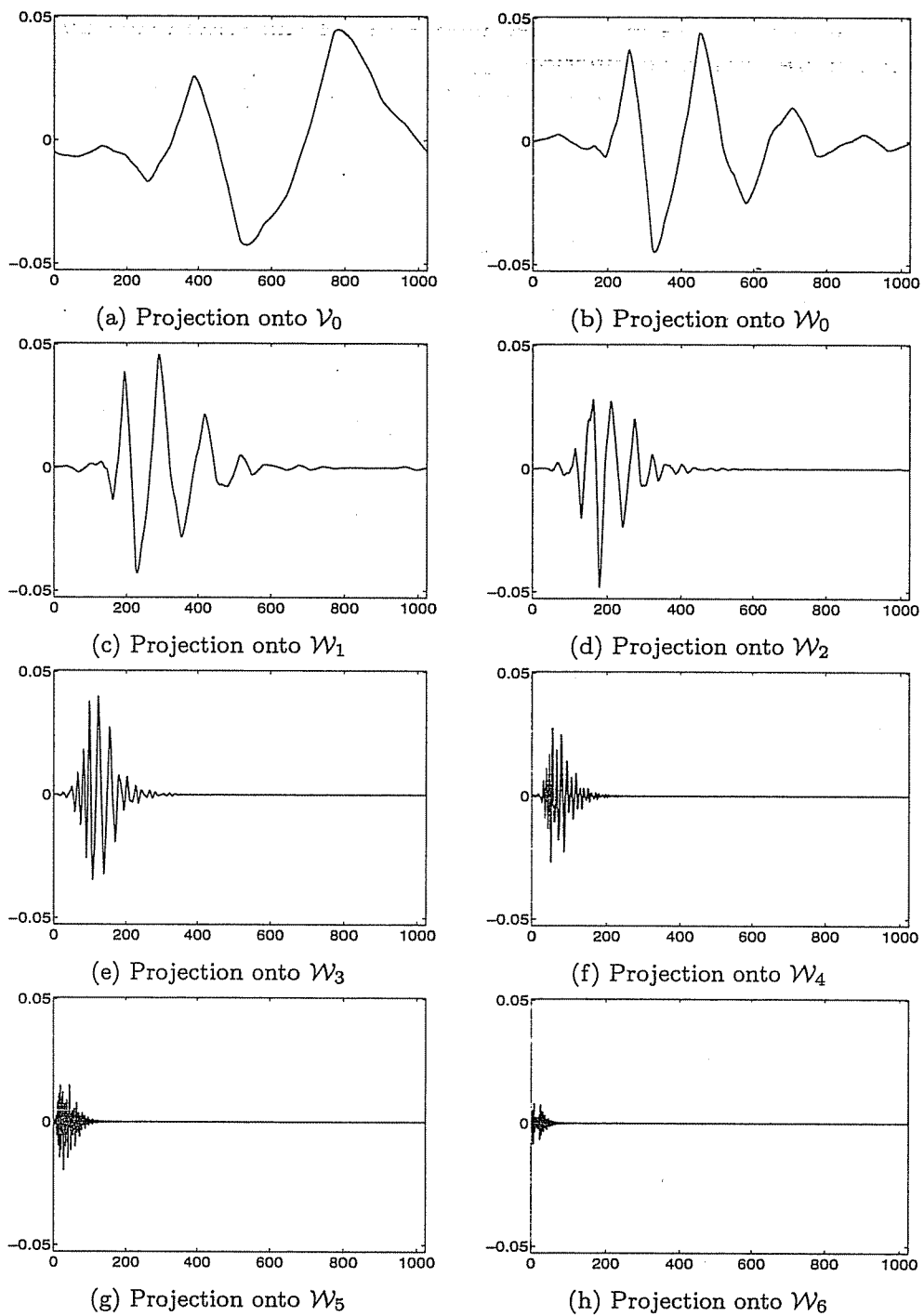
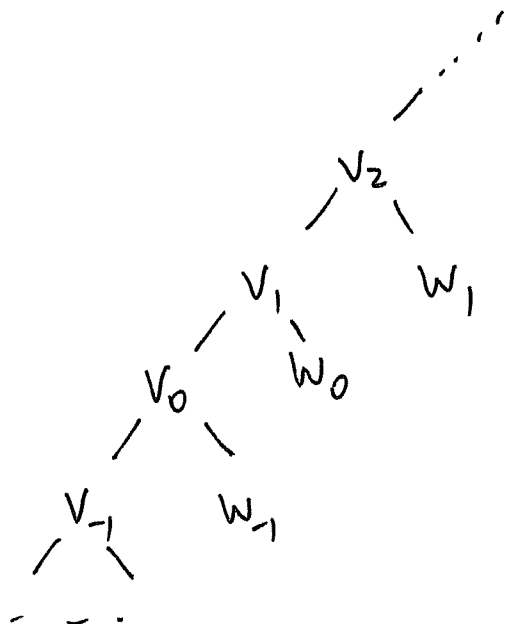


Figure 2.10. Projection of the Doppler Signal onto \mathcal{W} Spaces using ψ_{D8}

Infinite Setting



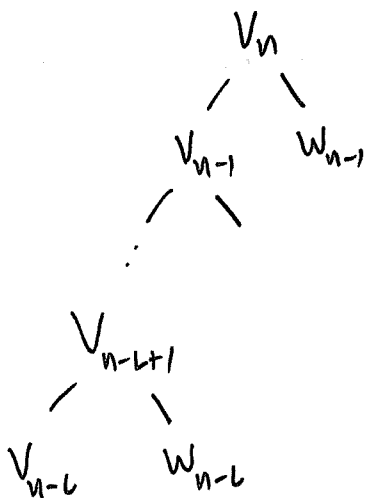
$$V_j = V_{j+1}$$

$$W_j \perp W_k \quad j \neq k$$

$$\overline{\bigoplus W_k} = L^2$$

$$f(x) = \sum_k Q_k f(x) = \sum_k \sum_j \langle f, \psi_j^k \rangle \psi_j^k(x)$$

Finite Setting



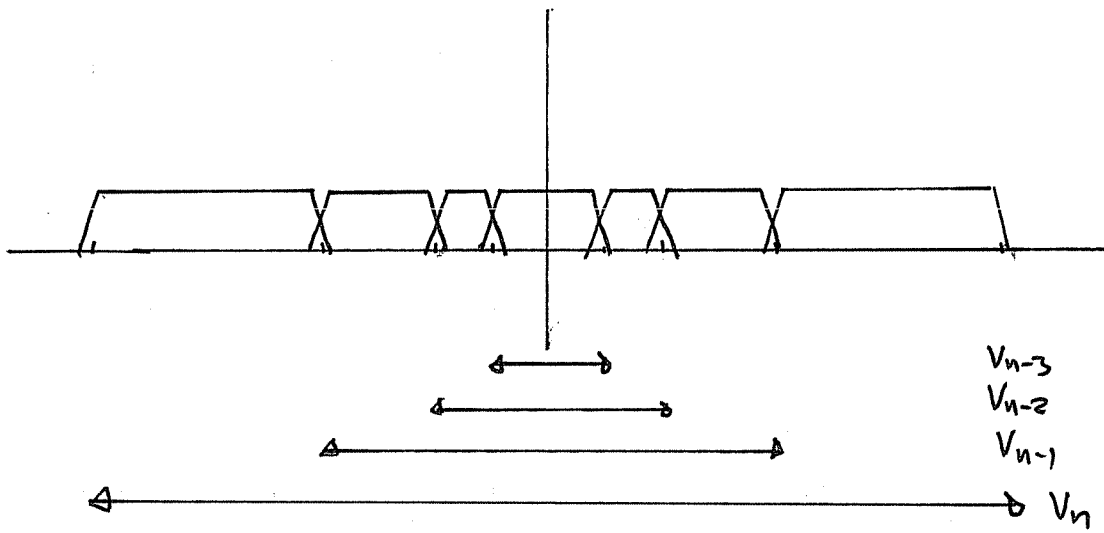
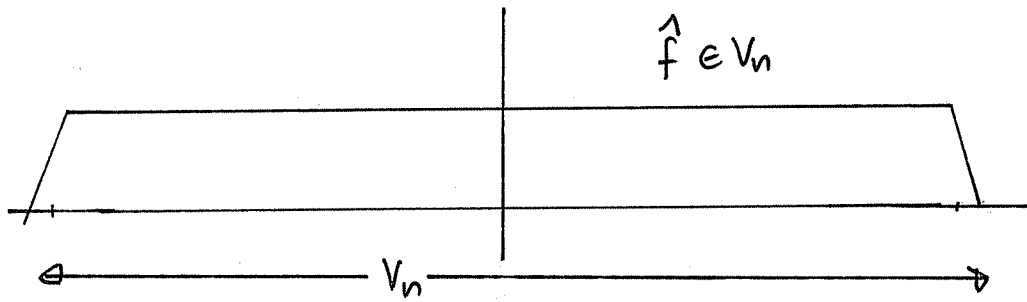
assume $f \in V_n$ (or start with $P_n f$)

Then

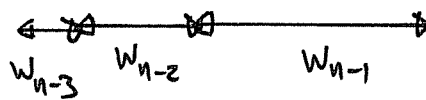
$$\begin{aligned} f &= P_{n-1} f \oplus Q_{n-1} f \\ &= P_{n-2} f \oplus Q_{n-2} f \oplus Q_{n-1} f \\ &= \dots \\ &= P_{n-l} f \oplus Q_{n-l} f \oplus \dots \oplus Q_{n-1} f \end{aligned}$$

wavelet decomposition of f through l levels

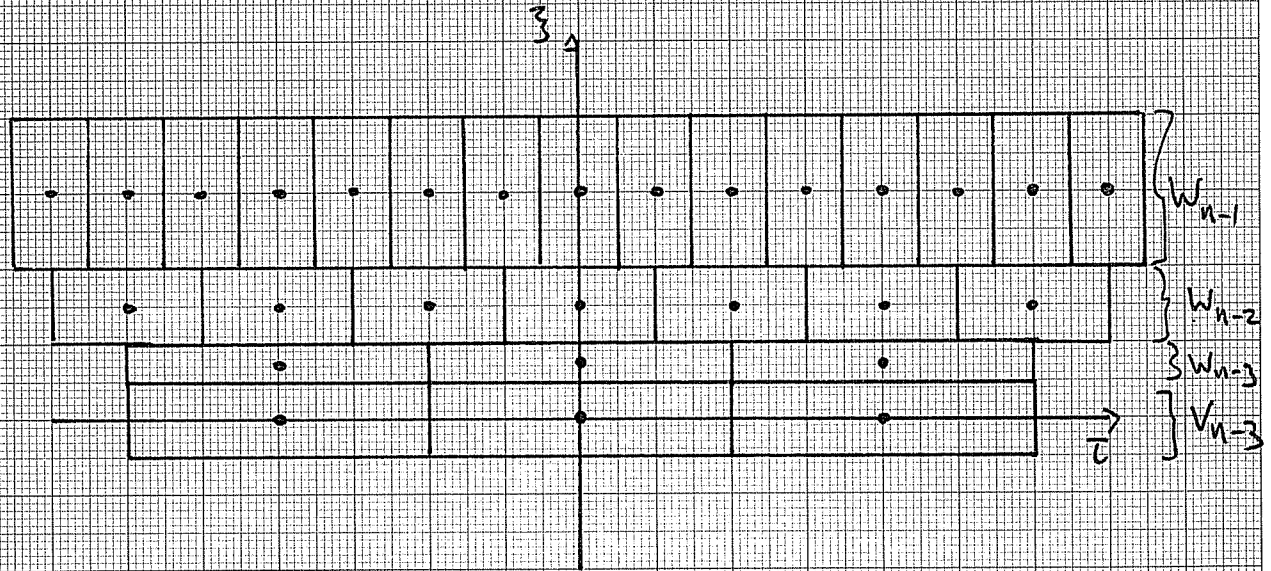
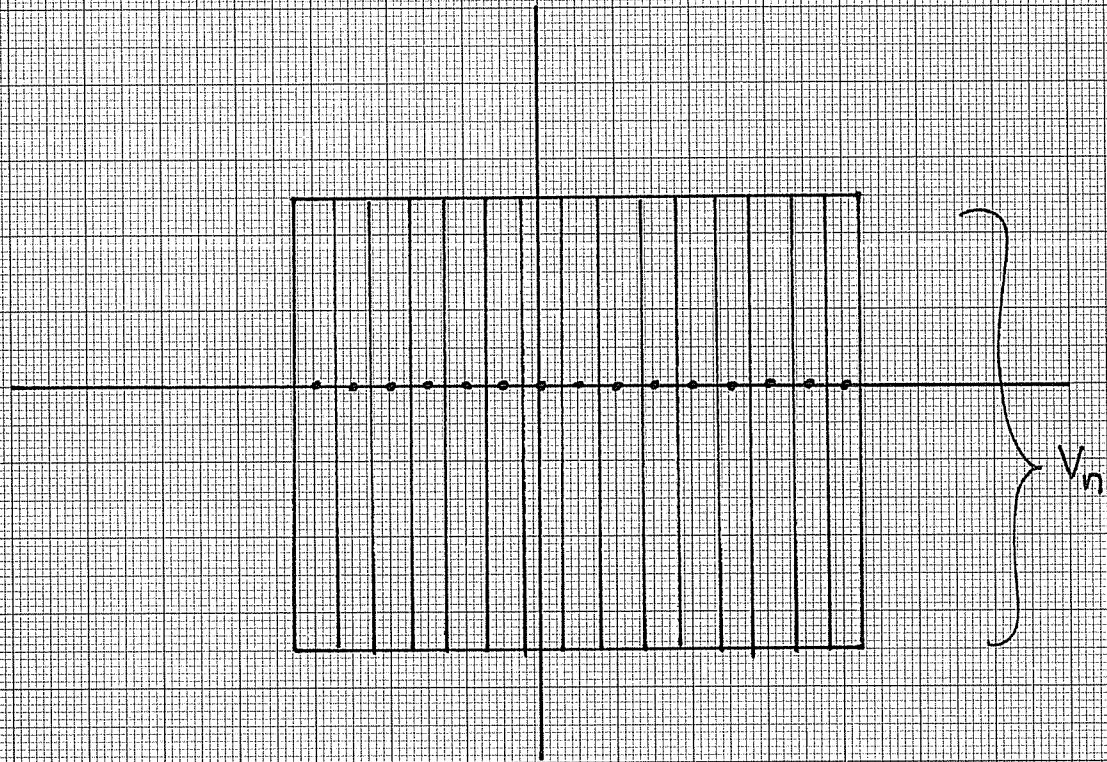
Frequency Picture



same here



Time-Frequency Picture



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Shorthand Notation

$$s_j^n = \langle f, \varphi_j^n \rangle$$

$$\vec{s}^n = (s_j^n)_{j \in \mathbb{Z}}$$

$$d_j^n = \langle f, \psi_j^n \rangle$$

$$\vec{d}^n = (d_j^n)_{j \in \mathbb{Z}}$$

$$V_n = V_{n-1} \oplus W_{n-1} \oplus W_{n-2} \oplus \dots \oplus W_{n-1}$$

$$f \in V_n = P_{n-1}f \oplus Q_{n-1}f \oplus Q_{n-2}f \oplus \dots \oplus Q_{n-1}f$$

represented by

$$\vec{s}^n \quad \vec{s}^{n-1} \quad \vec{d}^{n-1} \quad \vec{d}^{n-2} \quad \dots \quad \vec{d}^{n-1}$$

Decomposition

compute $\vec{s}^{n-1}, \vec{d}^{n-1}$ from \vec{s}^n

Reconstruction

compute \vec{s}^n from $\vec{s}^{n-1}, \vec{d}^{n-1}$

} apply
recursively

Calculating \tilde{S}^{n-1} from \tilde{S}^n

$$f(x) = \sum_j s_j^n \varphi_j^n(x)$$

$$P_{n-1} f(x) = \sum_j s_k^{n-1} \varphi_k^{n-1}(x)$$

$$\begin{aligned} \text{where } s_k^{n-1} &= \langle f, \varphi_k^{n-1} \rangle = \langle \sum_j s_j^n \varphi_j^n, \varphi_k^{n-1} \rangle \\ &= \sum_j \underbrace{\langle \varphi_j^n, \varphi_k^{n-1} \rangle}_{\text{need to compute this}} s_j^n \end{aligned}$$

Fundamental Trick: $V_{n-1} \subset V_n$

\Rightarrow every function in V_{n-1} can be written as a linear combination of basis functions of V_n

In particular

$$\varphi_0^n(x) = \sum_i h_i \varphi_i^{n-1}(x) = \sum_i \langle \varphi_0^n, \varphi_i^{n-1} \rangle \varphi_i^{n-1}(x)$$

$$\Leftrightarrow \boxed{\varphi(x) = \sqrt{2} \sum_i h_i \varphi(2x-i)}$$

(Recursion Relation)

then

$$\begin{aligned}\varphi_k^{n-1}(x) &= 2^{\frac{n-1}{2}} \varphi(2^{n-1}x - k) \\ &= 2^{\frac{n-1}{2}} \left\{ \sqrt{2} \sum_i h_i \varphi(2^n x - 2k - i) \right\} \\ &= 2^{\frac{n}{2}} \sum_i h_i \varphi(2^n x - (2k+i)) \\ &= \sum_i h_i \varphi_{2k+i}^n(x) = \sum_i h_{i-2k} \varphi_i^n(x)\end{aligned}$$

so

$$\begin{aligned}\langle \varphi_j^n, \varphi_k^{n-1} \rangle &= \langle \varphi_j^n, \sum_i h_{i-2k} \varphi_i^n \rangle \\ &= \sum_i \underbrace{\langle \varphi_j^n, \varphi_i^n \rangle}_{\delta_{ij}} \bar{h}_{i-2k} = \bar{h}_{j-2k}\end{aligned}$$

we get

$$s_k^{n-1} = \sum_j \bar{h}_{j-2k} s_j^n$$

convolution (filtering)
followed by downsampling

likewise

$$d_k^{n-1} = \sum_j \bar{g}_{j-2k} s_j^n$$

where

$$\psi(x) = \sqrt{2} \sum_i g_i \varphi(2x - i)$$

Reconstruction:

$$s_j^n = \sum_k h_{j-2k} s_k^{n-1} + \sum_k g_{j-2k} d_k^{n-1}$$

upsampling
followed by convolution

finite length vectors

$$(s_0^u \dots s_{N-1}^u) \quad \text{length } N$$

method 1 extend vector by zeros

disadvantage: if h, g have length L
then $\vec{s}^{u-1}, \vec{d}^{u-1}$ have $\frac{N}{2} + (\frac{L}{2} - 1)$
nonzero coefficients each

method 2 periodize

disadvantage: edge effects

operation count

vector length N , filter length L

level 1: $2LN$ multiplications

level 2: $2L(\frac{N}{2})$

level k : $2L(\frac{N}{2^{k-1}})$

total: $2L(N + \frac{N}{2} + \frac{N}{4} + \dots) \approx 4LN = O(N)$

compare to FFT: $O(N \log N)$