

EBU6018

Advanced Transform Methods

Multiresolution Analysis (MRA)

Andy Watson

Tackling the Redundancy Problem

- Because the $CWT(a,b)$ is a continuous transform, it encodes more than enough information to reconstruct the signal. A sampled version is sufficient.
- a is scaling factor. b is translation term.
- Let us sample in a dyadic (power of 2) grid:

$$a = 2^{-m} \quad b = n2^{-m}$$

$$CWT(2^{-m}, n2^{-m}) = \frac{1}{\sqrt{|2^{-m}|}} \int_{-\infty}^{\infty} s(t) \psi^* \left(\frac{t - n2^{-m}}{2^{-m}} \right) dt$$

$$d_{m,n} = CWT(2^{-m}, n2^{-m}) = \int_{-\infty}^{\infty} s(t) \psi_{m,n}^* dt$$

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

Discrete (sampled) Transform

If the set of functions $\psi_{m,n}$ forms a frame, then we can recover $s(t)$

$$s(t) = \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} d_{m,n} \hat{\psi}_{m,n}(t)$$

where $\hat{\psi}_{m,n}(t)$ forms the dual frame

For the orthonormal frame, we have $\hat{\psi}_{m,n}(t) = \psi_{m,n}(t)$

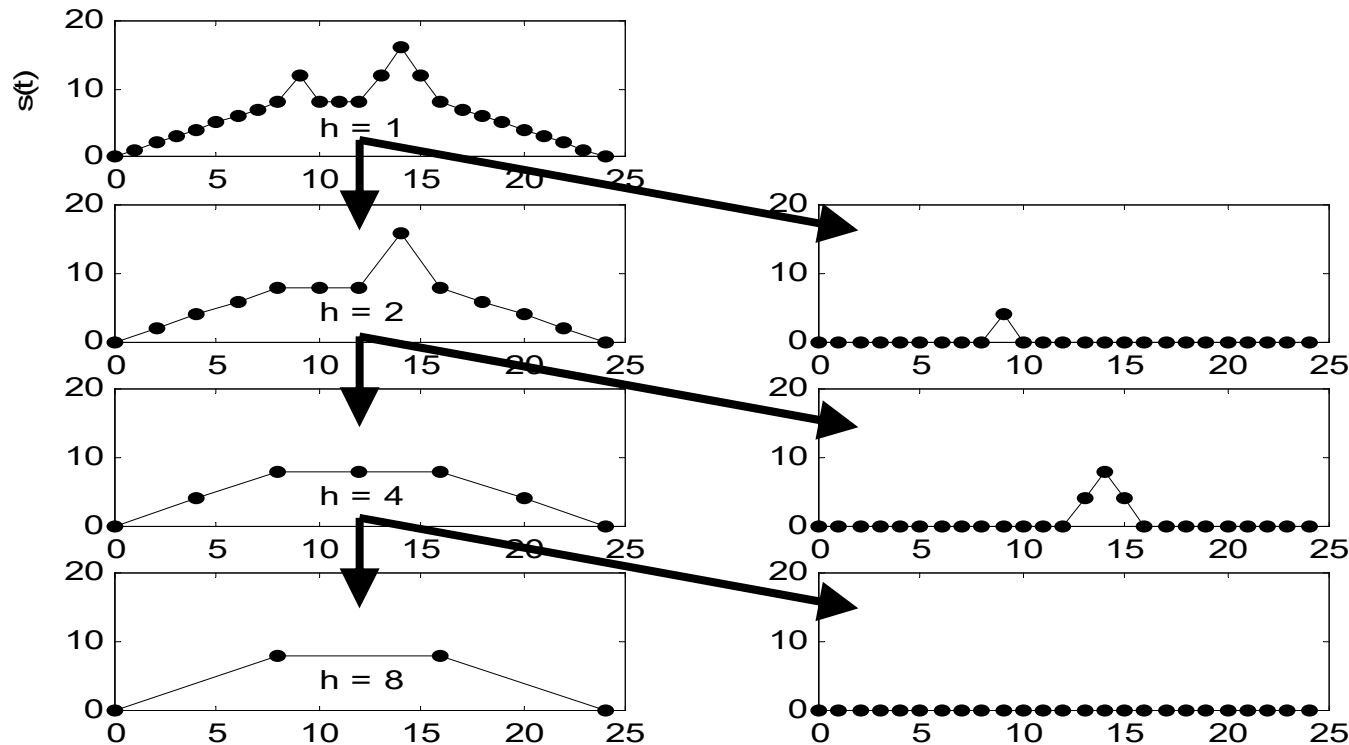
So we get the *wavelet series* (transform and inverse):

$$d_{m,n} = \int_{-\infty}^{\infty} s(t) \psi_{m,n}^* dt$$
$$s(t) = \sum_{m=-\infty}^{\infty} \sum_{n=-\infty}^{\infty} d_{m,n} \psi_{m,n}(t)$$

So: wavelet series is not redundant (finite number of coefficients)

Multiresolution Analysis (MRA)

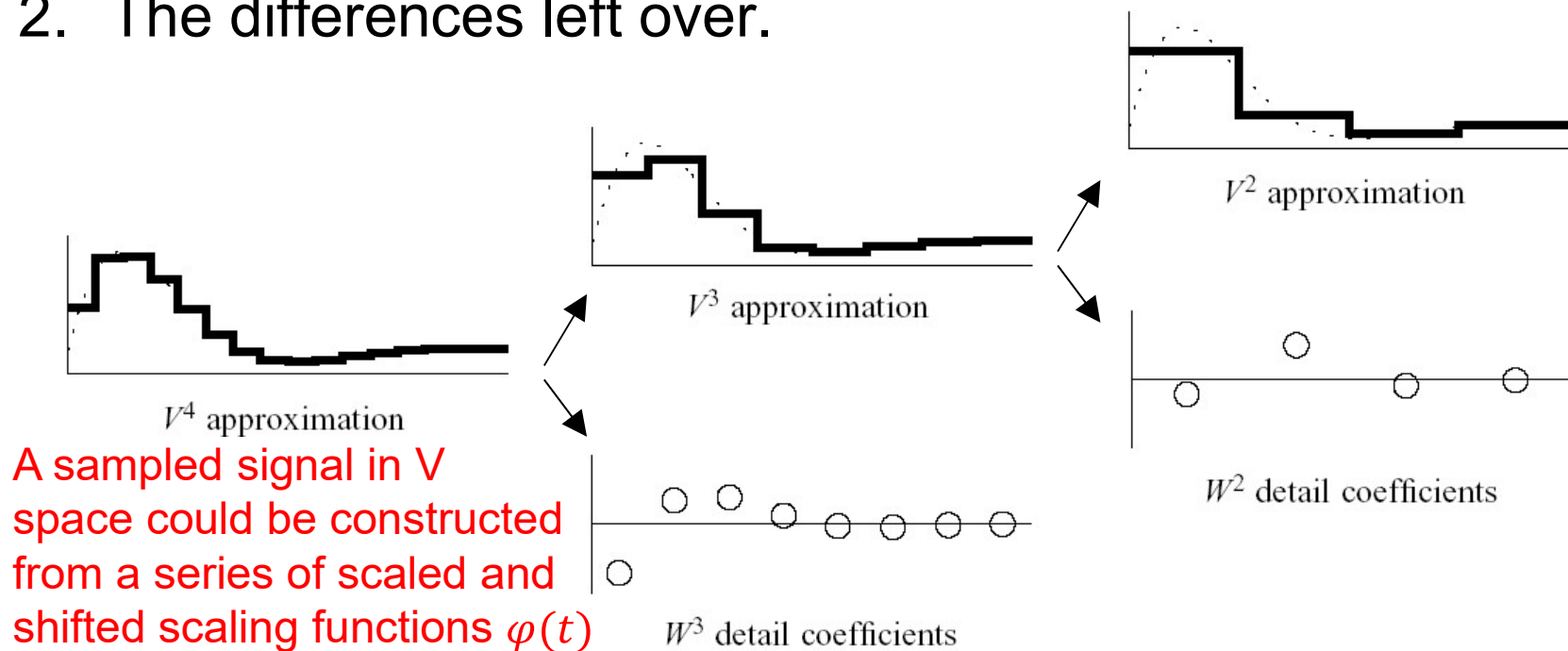
- Objective: To analyze a complicated function by dividing it into several simpler ones and studying them separately.



Piecewise Approximation

Basic Concept: Decompose a fine-resolution signal into

1. A coarse-resolution version of the signal, and
2. The differences left over.



So: what about wavelets? Theory coming next...

Multiresolution Analysis

- If we remove the fine detail from the signal in space V^4
- and put that in space W^3 then the approximate signal left in V^3 and the detail in W^3 have no elements in common.
- So the signal in V space and the signal in W space at a given level of decomposition are orthogonal.
- Similarly for V^2 and W^2 , etc.
- Starting at level 4 is arbitrary.

Multiresolution Analysis

With the Haar Function we saw that the scaling function can be constructed from a series of scaled and translated scaling functions. That is:

A scaling function $\phi(t)$ generates a nested subsequence of subspaces $\{V_j\}$,

$$\{0\} \leftarrow \cdots \subset V_{-1} \subset V_0 \subset V_1 \subset \cdots$$

and satisfies a *dilation (refinement) equation*

$$\phi(t) = \sum_k p_k \phi(at - k) \quad \text{for integer } k$$

for some $a > 0$ and coefficients $\{p_k\}$

We consider $a = 2$ that corresponds to octave-scales, that is, dyadic.

Spaces of functions

The space V_0 is generated by $\{\phi(t - k)\}$ for integers k

In general, V_{n-1} is generated by $\{\phi(2^{j-1}t - k)\}$

Then: $V_n = \bigoplus_{j=-\infty}^n W_j$ (orthogonal sum of subspaces).

$$V_{n+1} = \bigoplus_{j=-\infty}^{n+1} W_j = \left(\bigoplus_{j=-\infty}^n W_j \right) \oplus W_{n+1} = V_n \oplus W_{n+1}$$

Subspace W_{n+1} is the *orthogonal complementary subspace* of V_n in V_{n+1} .

That is, we can reconstruct the original signal by adding together V and W at lower levels. (We could also provide compression by omitting the coefficients at low levels.)

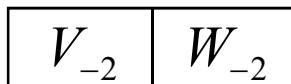
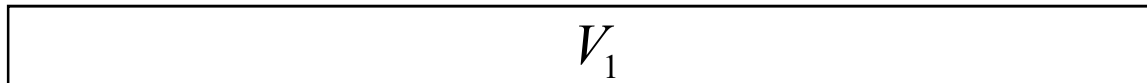
Multiresolution Analysis (cont)

Subspaces $\{V_j\}$ are nested
while subspaces $\{W_j\}$ are mutually orthogonal.

Consequently,

$$\begin{cases} V_j \cap V_l = V_j & l > j; \\ W_j \cap W_l = \{0\} & j \neq l; \\ V_j \cap W_l = \{0\} & j \leq l. \end{cases}$$

\vdots



\vdots

i.e. V_1 can be decomposed
into combination of V_0 and W_0 , etc.

Multiresolution Analysis (cont.)

If the Scaling function and the Wavelet function are orthogonal, as they are for the Haar Function, then we can say that because V and W are orthogonal:

Subspaces $\{W_j\}$ are generated by *wavelets* $\psi(t)$

And $\{V_j\}$ is generated by scaling functions $\phi(t)$

In other words, any $f_j(t) \in V_j$

can be written as
$$f_j(t) = \sum_k c_k^j \phi(2^j t - k),$$

And.....

Multiresolution Analysis (cont.)

..... any function $g_j(t) \in W_j$
can be written as $g_j(t) = \sum_k d_k^j \psi(2^j t - k),$
for some coefficients

$$\{c_k^j\}_{k \in \mathbb{Z}}, \{d_k^j\}_{k \in \mathbb{Z}}$$

Furthermore,

$$f(t) \in V_j \Leftrightarrow f(2t) \in V_{j+1}$$

$$f(t) \in V_j \Leftrightarrow f\left(t + \frac{1}{2^j}\right) \in V_j$$

Multiresolution Analysis (cont.)

Subspaces $\{V_j\}$ are called “approximation subspaces” and $\{W_j\}$ are the “wavelet subspaces”.

At any level j , V_j contains the smooth part and W_j contains the “details” of the original function.

$j \uparrow \Rightarrow f_j(t)$, a finer approximation of $f(t)$,

$j \downarrow \Rightarrow f_j(t)$, a coarser approximation of $f(t)$.

Two-Scale Relations

Since $\phi(t) \in V_0 \subset V_1$, and $\psi(t) \in W_0 \subset V_1$;
we should be able to write $\phi(t)$ and $\psi(t)$ in terms of the
bases that generate V_1 .

In other words, there exists two sequences $\{p_k\}, \{q_k\}$
such that

$$\phi(t) = \sum_k p_k \phi(2t - k),$$

$$\psi(t) = \sum_k q_k \phi(2t - k).$$

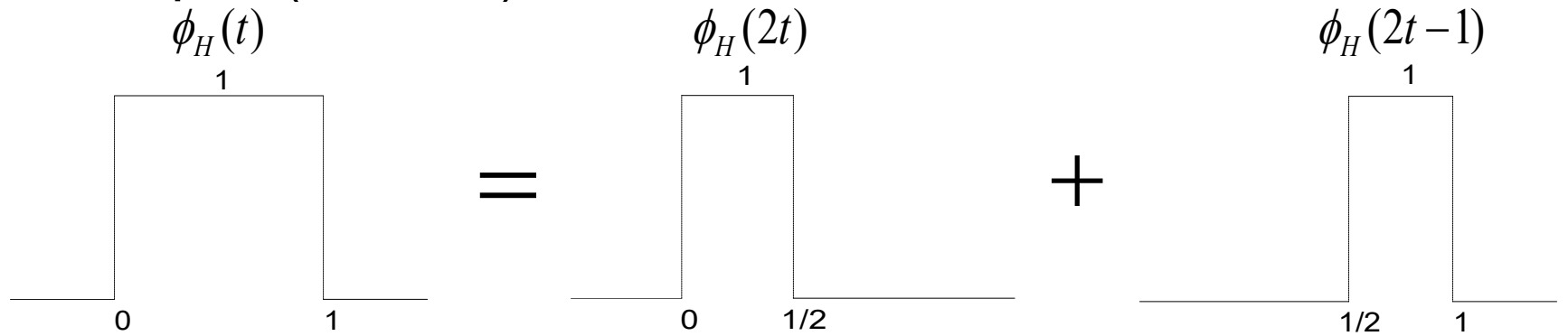
In general, for any j , the relationship between V_j, W_j with
 V_{j+1} is governed by

$$\phi(2^j t) = \sum_k p_k \phi(2^{j+1} t - k),$$

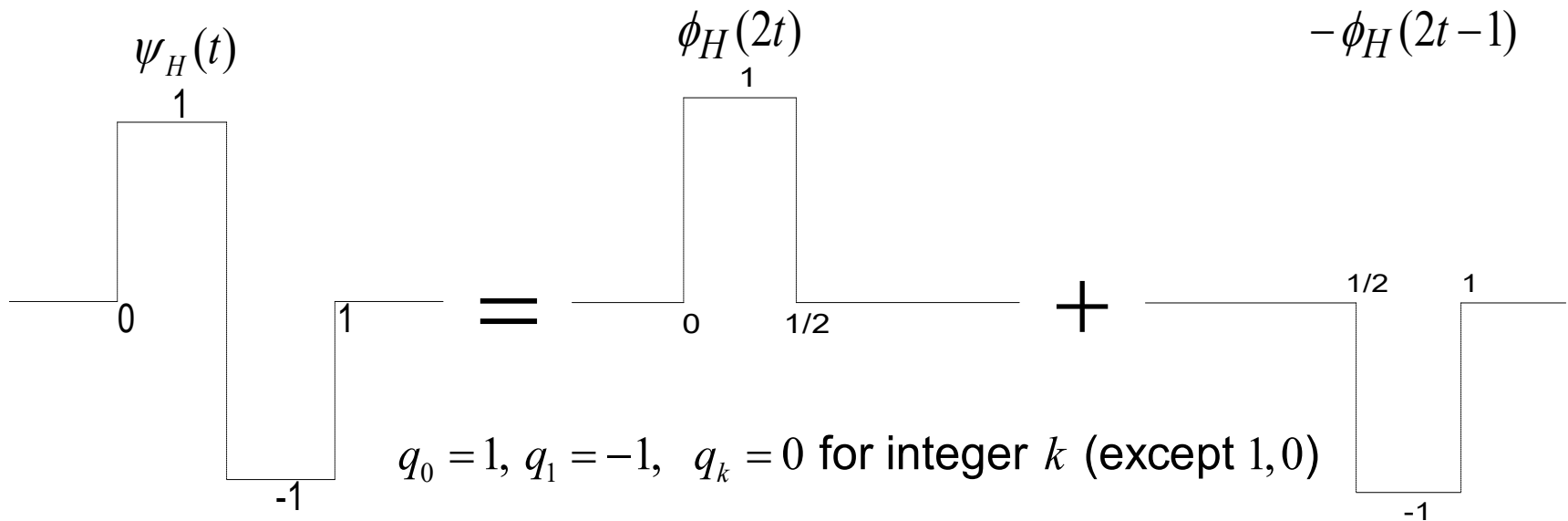
$$\psi(2^j t) = \sum_k q_k \phi(2^{j+1} t - k).$$

Two-Scale Relations (cont.)

Example: (H: Haar)



$p_0 = p_1 = 1, \quad p_k = 0$ for integer k (except 1,0)



$q_0 = 1, \quad q_1 = -1, \quad q_k = 0$ for integer k (except 1,0)

Decomposition Relation

Since $V_1 = V_0 + W_0$ and $\phi(2t), \phi(2t-1) \in V_1$, there exist two pairs of sequences $(\{a_{2k}\}, \{b_{2k}\})$

such that
$$\phi(2t) = \sum_k \{a_{2k}\phi(t-k) + b_{2k}\psi(t-k)\};$$

$$\phi(2t-1) = \sum_k \{a_{2k-1}\phi(t-k) + b_{2k-1}\psi(t-k)\}.$$

Combining these two relations, we have

$$\phi(2t-l) = \sum_k \{a_{2k-l}\phi(t-k) + b_{2k-l}\psi(t-k)\}.$$

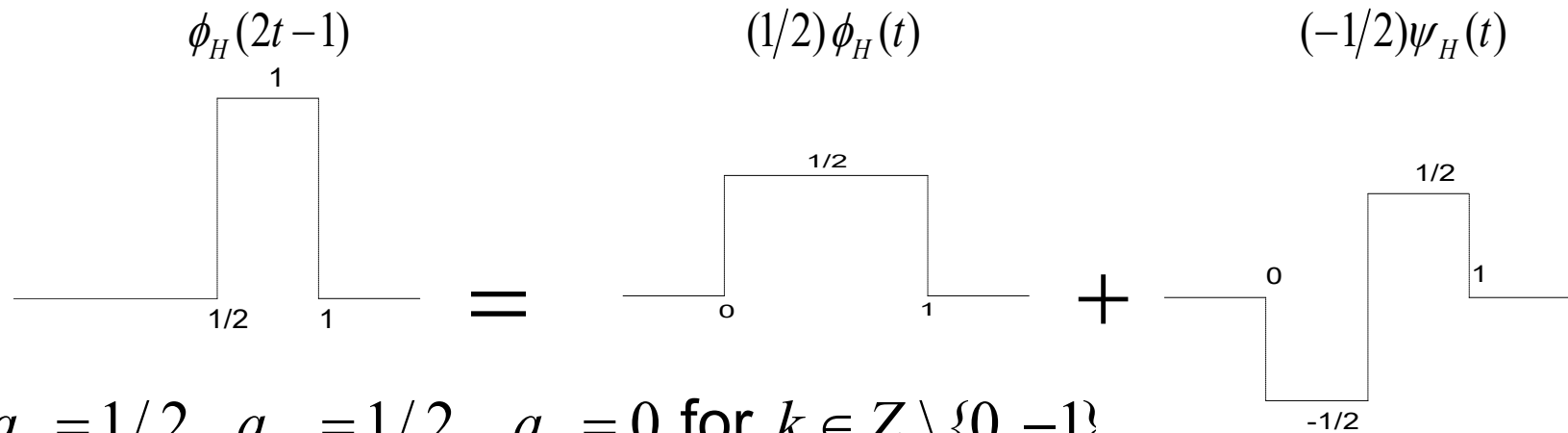
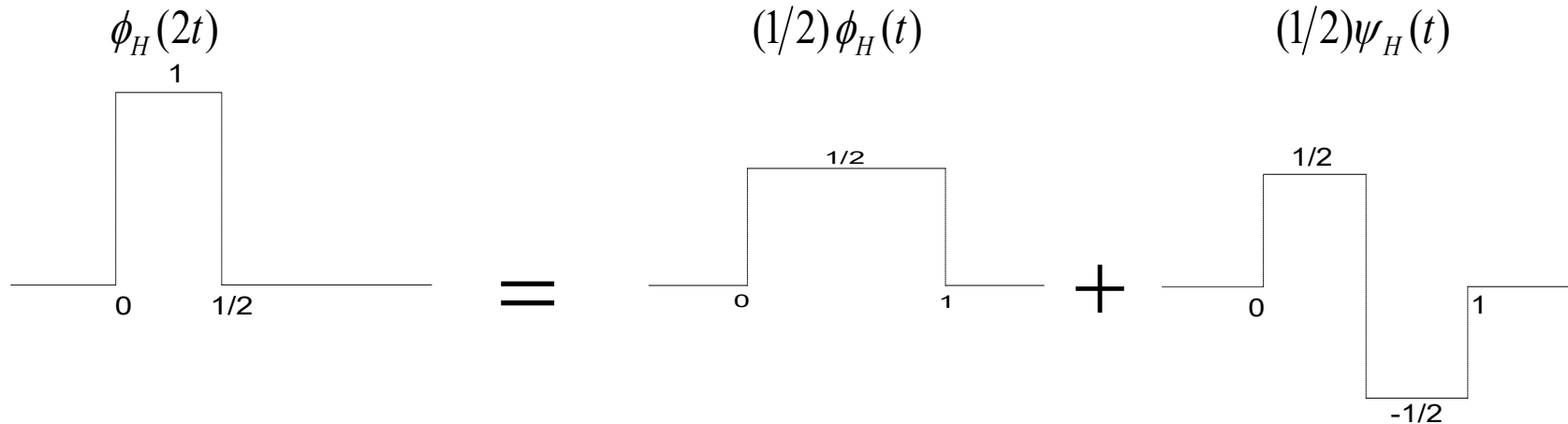
for all integer l

In general, we have

$$\phi(2^{j+1}t-l) = \sum_k \{a_{2k-l}\phi(2^j t-k) + b_{2k-l}\psi(2^j t-k)\}.$$

Decomposition Relation (cont.)

Example: (H: Haar)



$$a_0 = 1/2, \quad a_{-1} = 1/2, \quad a_k = 0 \text{ for } k \in \mathbb{Z} \setminus \{0, -1\}$$

$$b_0 = 1/2, \quad b_{-1} = -1/2, \quad b_k = 0 \text{ for } k \in \mathbb{Z} \setminus \{0, -1\}$$

Summary

We have seen that a series of sampled values of a continuous signal can be divided into two sequences:

- One which is approximations
- The other fine detail
- These two sequences are orthogonal
- The approximations can be derived as a series of scaled and shifted scaling functions, each with a coefficient (same as a linear piecewise approximation)
- The fine detail can be derived as a series of scaled and shifted wavelet functions, each with a coefficient
- These two sequences can be recombined to form the original sequence