Algorithms of Scientific Computing

Wavelets

Michael Bader

Summer Term 2013

Part I

Haar Wavelets as an Hierarchical Basis

Remember the 1D Hierarchical Basis

- "mother of all hat functions": $\phi(x) := \max\{1 - |x|, 0\}$
- hat functions on level $l \in \mathbb{N}$ with mesh width $h_l = 2^{-l}$ at grid points $x_{l,i} = i \cdot h_l$:

$$\phi_{l,i}(x) := \phi\left(\frac{x - x_{l,i}}{h_l}\right)$$

- hierarchical basis functions on level l:

$$\phi_{l,i}(x) \quad \text{for all } i \in \mathcal{I}_l := \{i : 1 \le i < 2^l,\ i \text{ odd}\}$$

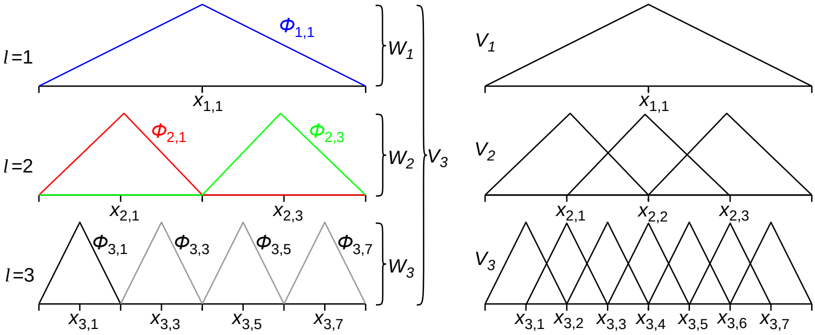
- resulting hierarchical basis

$$\Psi_n := \bigcup_{l=1}^{n} \{\phi_{l,i} : i \in \mathcal{I}_l\}.$$

- with corresponding function spaces:

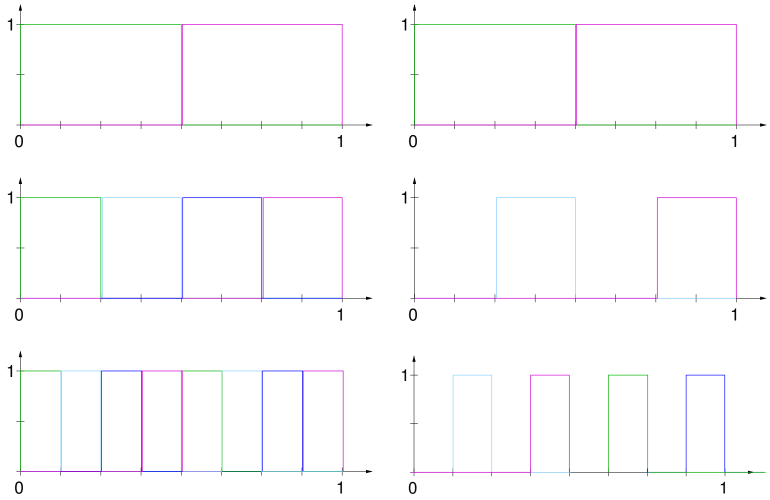
$$W_l := \operatorname{span}\{\phi_{l,i} :\ i \in \mathcal{I}_l\} \quad \text{and} \quad V_n = \bigoplus_{l=1}^{n} W_l$$

Hierarchical vs. Nodal Basis



$\rightarrow$ for piecewise *linear* (basis) functions

Piecewise Constant Basis – Attempt # 1



Piecewise Constant Basis – Attempt # 1 (cont.)

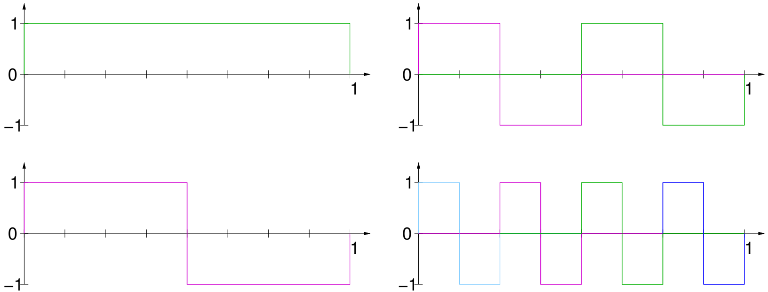
Discussion:

- obviously qualifies as a "hierarchical basis" w.r.t. hierarchical levels and mesh widths
- built from a "mother of all step functions", e.g.:

$$\phi(x) := \begin{cases} 1 & \text{if } 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

- hierarchical basis functions on level l: $\phi_{l,i}(x) := \phi\left(\frac{x - x_{l,i}}{h_l}\right)$
- nodal basis on level l: $V_l = \text{span}\{\phi_{l,i}(x) \colon i = 0, \dots, 2^l - 1\}$
- hierarchical surplus: $W_l = \text{span}\{\phi_{l,i}(x) \colon 1 \leq i < 2^l,\ i \text{ odd}\}$
- would hierarchical surpluses be small in such a setting?
- are functions represented well by coarse-level basis functions?

Attempt # 2: "Hierarchical Haar Basis"



- for each *interval*, we obtain a contribution from each *level*
- course-level representations will consist of *average values*
- each "surplus" level add corrections to averages

Hierarchical Haar Basis

- again a hierarchical basis with "mother Haar function":

$$\psi(x) := \begin{cases} 1 & \text{if } 0 < x < 1 \\ -1 & \text{if } 1 < x < 2 \\ 0 & \text{otherwise} \end{cases}$$

- hierarchical Haar basis functions on level l:

$$\psi_{l,i}(x) := \psi\left(\frac{x - x_{l,i}}{h_l}\right) \quad \text{for all } i \in \mathcal{I}_l := \{i : 0 \le i < 2^l,\ i \text{ even}\}$$

- hierarchical surplus space for each level:

$$W_l := \operatorname{span}\{\psi_{l,i} :\ i \in \mathcal{I}_l\}$$

- space of piecewise constant functions $V_n = \bigoplus_{l=0}^{n} W_l$

  $\rightarrow$ includes a step function on interval $(0,1)$ for $l = 0$

Hierarchical Haar Basis – Coefficients

- consider a piecewise constant function $\in V_1$:

$$s(x) := a\phi_{1,0}(x) + b\phi_{1,1}(x) \begin{cases} a & \text{if } 0 < x < \frac{1}{2} \\ b & \text{if } \frac{1}{2} < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

- condition in interval $0 < x < \frac{1}{2}$:

$$v_{0,0} \underbrace{\psi_{0,0}(x)}_{=\phi_{0,0}(x)} + v_{1,0}\psi_{1,0}(x) = v_{0,0} + v_{1,0} = a$$

- condition in interval $\frac{1}{2} < x < 1$:

$$v_{0,0}\psi_{0,0}(x) + v_{1,0}\psi_{1,0}(x) = v_{0,0} - v_{1,0} = b$$

- solve linear system of equations:

$$v_{0,0} = \frac{1}{2}(a+b) \qquad v_{1,0} = \frac{1}{2}(a-b)$$

Hierarchical Haar Basis – Transformation

- represent a piecewise constant function $s(x) \in V_l$:

$$s(x) = \sum_{i=0}^{2^l-1} c_{l,i}\,\phi_{l,i}(x)$$

- write as coarse function plus hierarchical surplus:

$$s(x) = \underbrace{\sum_i c_{l,i}\,\phi_{l,i}(x)}_{\in V_l} = \underbrace{\sum_i c_{l-1,i}\,\phi_{l-1,i}(x)}_{\in V_{l-1}} + \underbrace{\sum_{i\in\mathcal{I}_l} d_{l,i}\,\psi_{l,i}(x)}_{\in W_l}$$

- examine intervals $x_{l,2i} < x < x_{l,2i+1}$ and $x_{l,2i+1} < x < x_{l,2i+2}$:

$$c_{l-1,i} + d_{l,2i} = c_{l,2i} \qquad \text{and} \qquad c_{l-1,i} - d_{l,2i} = c_{l,2i+1}$$

- leads to formula for $c_{l-1,i}$ and $d_{l,2i}$ (note the even index $2i$):

$$c_{l-1,i} = \tfrac{1}{2}(c_{l,2i} + c_{l,2i+1}) \qquad d_{l,2i} = \tfrac{1}{2}(c_{l,2i} - c_{l,2i+1})$$

Part II

Haar Wavelets as Wavelets

Change of Notation – Scaling Function

- define **scaling function**:

$$\phi(x) := \begin{cases} 1 & \text{if } 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

- nodal basis functions on level l:

$$\phi_{l,k}(x) := 2^{l/2}\phi\left(\frac{x - x_{l,k}}{h_l}\right) = 2^{l/2}\phi\left(\frac{x - k2^{-l}}{2^{-l}}\right) = 2^{l/2}\phi\left(2^l x - k\right)$$

(remember: $x_{l,k} = k \cdot 2^{-l}$ and $h_l = 2^{-l}$)

- scaling with $2^{l/2}$ to be discussed ...
- resulting nodal basis on level l:

$$V_l = \text{span}\{\phi_{l,k}(x) \colon k = 0, \ldots, 2^l - 1\}$$

Change of Notation – Wavelet Functions

- define **mother Haar wavelet**:

$$\psi(x) := \begin{cases} 1 & \text{if } 0 < x < \frac{1}{2} \\ -1 & \text{if } \frac{1}{2} < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

- **Haar wavelet functions** on level l:

$$\psi_{l,k}(x) := 2^{l/2}\psi\left(2^l x - k\right) = 2^{l/2}\psi\left(\frac{x - 2^{-l}k}{2^{-l}}\right) = 2^{l/2}\psi\left(\frac{x - x_{l,k}}{h_l}\right)$$

for $k = 0, \ldots, 2^l - 1$, (but no “stride two”)

- Important changes:
  - shifted numbering of levels: $\psi(x)$ defined on $[0, 1]$
  - thus: supports of $\psi_{l,k}(x)$ and $\psi_{l,k+1}(x)$ no longer overlap
  - index $k = 0, \ldots, 2^l - 1$ used with “stride 1”

Change of Notation – Wavelet Functions (cont.)

- define **mother Haar wavelet**:

$$\psi(x) := \begin{cases} 1 & \text{if } 0 < x < \frac{1}{2} \\ -1 & \text{if } \frac{1}{2} < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

- **Haar wavelet functions** on level l:

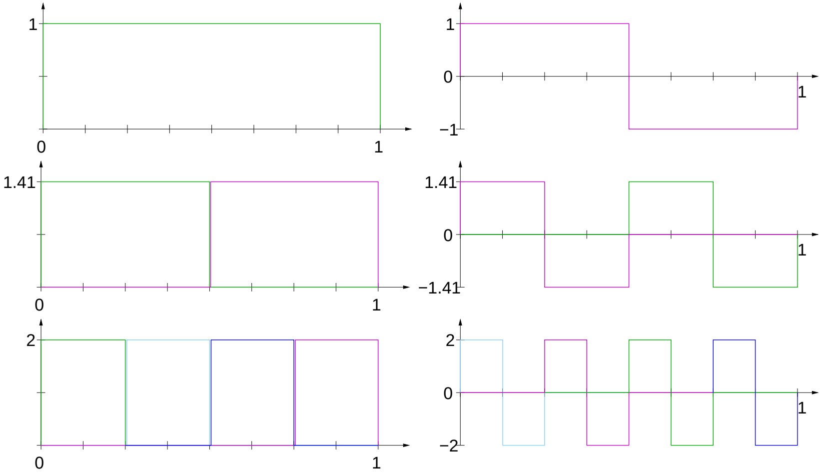
$$\psi_{l,k}(x) := 2^{l/2}\psi\left(2^l x - k\right) \qquad \text{for } k = 0, \dots, 2^l - 1.$$

- wavelet space for each level:

$$W_l := \text{span}\left\{\psi_{l,k} : \; k = 0, \dots, 2^l - 1\right\}$$

- definition of function spaces: $V_{l+1} = V_l \oplus W_l$

Haar Wavelet Functions



Haar Wavelets – Transformation

- represent a piecewise constant function $s(x) \in V_l$:

$$s(x) = \sum_{k=0}^{2^l-1} c_{l,k}\, \phi_{l,k}(x)$$

- write as coarse function plus hierarchical surplus:

$$s(x) = \underbrace{\sum_k c_{l,k}\, \phi_{l,k}(x)}_{\in V_l} = \underbrace{\sum_k c_{l-1,k}\, \phi_{l-1,k}(x)}_{\in V_{l-1}} + \underbrace{\sum_k d_{l-,k}\, \psi_{l-1,k}(x)}_{\in W_{l-1}}$$

- transform $c_{l,2k}$ to $c_{l-1,k}$ and $d_{l-1,k}$:

$$c_{l-1,k} = \tfrac{1}{\sqrt{2}}(c_{l,2k} + c_{l,2k+1}) \qquad d_{l-1,k} = \tfrac{1}{\sqrt{2}}(c_{l,2k} - c_{l,2k+1})$$

- backward transform $c_{l-1,k}$ and $d_{l,2k}$ to $c_{l,2k}$ and $c_{l,2k+1}$:

$$c_{l,2k} = \tfrac{1}{\sqrt{2}}(c_{l-1,k} + d_{l-1,k}) \qquad c_{l,2k+1} = \tfrac{1}{\sqrt{2}}(c_{l-1,k} - d_{l-1,k})$$

Haar Wavelets – Transformation (2)

- scheme for wavelet decomposition:

$$
\begin{array}{ccccccccc}
 & & d^{(J-1)} & & d^{(J-2)} & & d^{(J-3)} & & \dots \\
 & \nearrow & & \nearrow & & \nearrow & & \nearrow & \\
c^{(J)} & \rightarrow & c^{(J-1)} & \rightarrow & c^{(J-2)} & \rightarrow & c^{(J-3)} & \rightarrow & \dots
\end{array}
$$

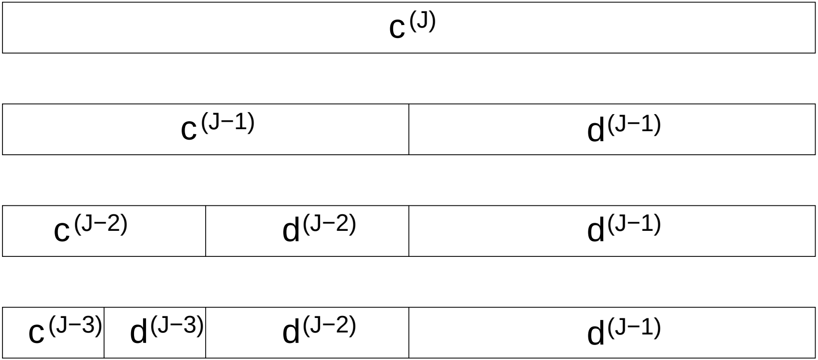
- scheme for assembly:

$$
\begin{array}{ccccccccc}
\dots & & d^{(J-3)} & & d^{(J-2)} & & d^{(J-1)} & & \\
 & \searrow & & \searrow & & \searrow & & \searrow & \\
\dots & \rightarrow & c^{(J-3)} & \rightarrow & c^{(J-2)} & \rightarrow & c^{(J-1)} & \rightarrow & c^{(J)}
\end{array}
$$

- Note: computational effort for transformations is only $\mathcal{O}(N)$

Haar Wavelets – Transformation (3)

Scheme for data structures:



Haar Wavelets – Orthogonality

- Haar wavelets are **orthogonal** functions:

$$\int \psi_{l,i}(x)\,\psi_{m,j}(x)\,dx := \begin{cases} 1 & \text{if } l = m \text{ and } i = j \\ 0 & \text{otherwise} \end{cases}$$

- two different wavelet functions $\psi_{l,i} \neq \psi_{l,j}$ on the same level l

$$\int \psi_{l,i}(x)\,\psi_{l,j}(x)\,dx = 0 \qquad \text{(no overlap of functions!)}$$

- two wavelet functions $\psi_{l,i} \neq \psi_{m,j}$ on different levels $l < m$

$$\int \psi_{l,i}(x)\,\psi_{m,j}(x)\,dx = \psi_{l,i}(x^{+}_{m,j}) \int \psi_{m,j}(x)\,dx = 0$$

- scalar product of a wavelet functions $\psi_{l,i}$ with itself

$$\int \left(\psi_{l,i}(x)\right)^2 dx = \int_{x_{l,i}}^{x_{l,i}+2^{-l}} \left(2^{l/2}\right)^2 dx = 1$$

Haar Wavelets – Summary and Next Steps

Haar wavelets:

- hierarchical basis of **piecewise constant** and . . .
- . . . **orthogonal** basis functions
- $\mathcal{O}(N)$ effort for hierarchical transformation

Haar Wavelets – Summary and Next Steps

Haar wavelets:

- hierarchical basis of **piecewise constant** and ...
- ... **orthogonal** basis functions
- $\mathcal{O}(N)$ effort for hierarchical transformation

Next steps:

- applications in signal and image processing
- extension to 2D (and higher dimensions)
- is there a piecewise linear/polynomial/higher-order orthogonal(!) wavelet basis?

Part III

Wavelets in Signal and Image Processing

Scaling Functions and Wavelet Functions in 2D

Use tensor product, as for hierarchical basis:

- 2D scaling functions on levels l_1, l_2:

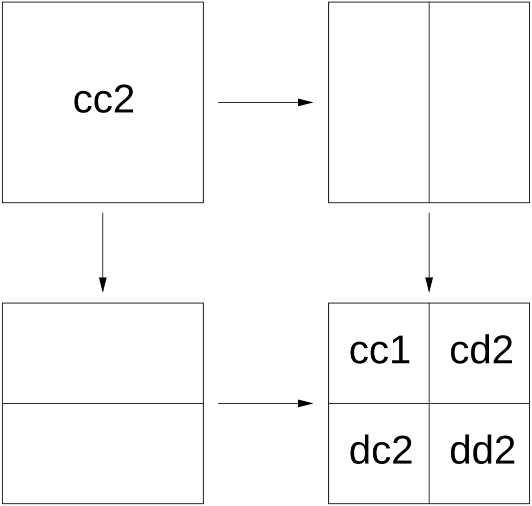
$$\phi_{\vec{l},\vec{k}}(x_1, x_2) := \phi_{l_1,l_2,k_1,k_2}(x_1, x_2) := \phi_{l_1,k_1}(x_1) \cdot \phi_{l_2,k_2}(x_2)$$

- 2D wavelet functions on levels l_1, l_2:

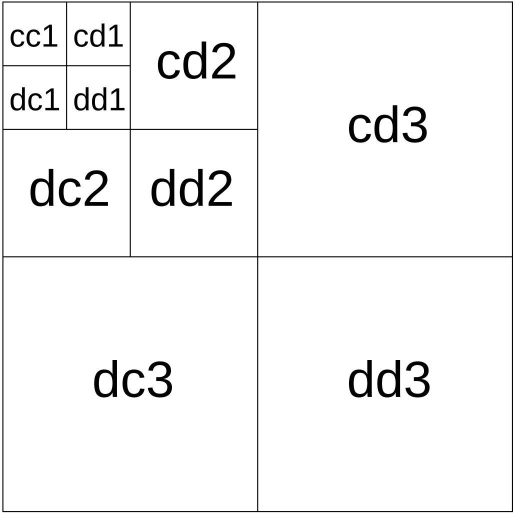
$$\psi_{\vec{l},\vec{k}}(x_1, x_2) := \psi_{l_1,l_2,k_1,k_2}(x_1, x_2) := \psi_{l_1,k_1}(x_1) \cdot \psi_{l_2,k_2}(x_2)$$

- thus straightforward extension to 3D and higher dimensions
- 2D transform equivalent to sequence of 1D transforms (as for Hierarchical Basis and Fourier Transforms)

2D Wavelets – A Single Transformation Step



2D Wavelets – Storage Scheme



Wavelet-Based Compression of Image Data

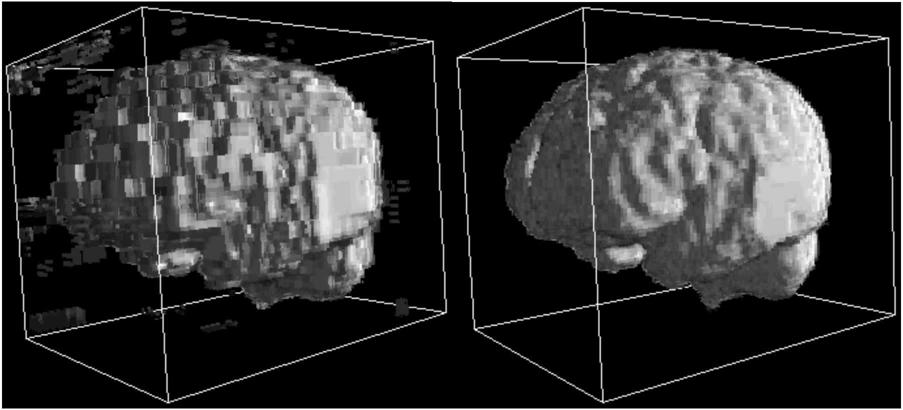
Typical steps for image compression:

1. Conversion of colour model
   (separation of brightness and colour information)
2. **2D discrete Wavelet transform**
3. **Quantisation of the coefficients** ($\rightarrow$ reduce information)
4. efficient encoding
   (loss-less compression of the quantised coefficients)

In practice:

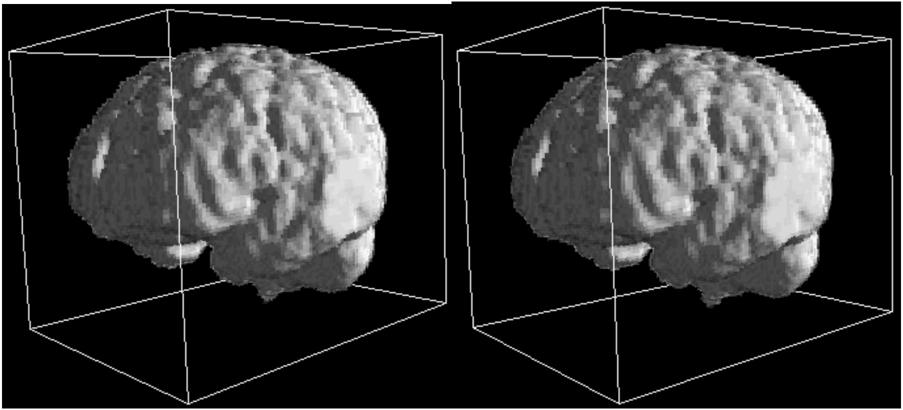
- different algorithms: EZF, SPIHT, ...
- similar to JPEG, but often much better quality
- see, e.g., Walker: “Wavelet-based Image Compression” for full details

Example: 3D Image Compression



(wavelet-based compression of raster data, A. Dehmel)

Example: 3D Image Compression (2)



(wavelet-based compression of raster data, A. Dehmel)

From Fourier Transform to Wavelets

(Discrete) Fourier Transform:

$$f(x) \sim \sum c_k e^{ikx} \qquad \text{or} \qquad f_n = \sum F_k e^{i\pi kn/N}$$

- f contains only spatial information
- c_k, F_k contain only frequency information
- no relation between frequency and location

From Fourier Transform to Wavelets

(Discrete) Fourier Transform:

$$f(x) \sim \sum c_k e^{ikx} \qquad \text{or} \qquad f_n = \sum F_k e^{i\pi kn/N}$$

- f contains only spatial information
- c_k, F_k contain only frequency information
- no relation between frequency and location

Windowed Fourier Transform:

$$f(x) = \frac{1}{2\pi} \int\int F(u,k) g(x-u) e^{ikx}\, dk\, du, \quad F(u,k) = \int f(x) g(x-u) e^{-ikx}\, dx$$

- $F(u,k)$: frequency k at location u
- $g(\xi)$ a window function
  $\rightarrow$ narrow windows do not allow to locate coarse frequencies
  $\rightarrow$ but wide windows decrease accuracy in location

From Fourier Transform to Wavelets (2)

Continuous Wavelet Transform:

$$W(a,b) = \int f(t)\psi_a^b(t)\,dt \quad \text{and} \quad f(x) = \frac{1}{C_\psi}\int\int W(a,b)\frac{\psi_a^b(x)}{a^2}\,da\,db$$

- continuous in a and b
- $\psi_a^b(t) = |a|^{-1/2}\psi\left(\frac{t-b}{a}\right)$ with "mother wavelet" ψ
- infinitely many (redundant) coefficients $\rightarrow$ not practicable

From Fourier Transform to Wavelets (2)

Continuous Wavelet Transform:

$$W(a,b) = \int f(t)\psi_a^b(t)\,dt \quad \text{and} \quad f(x) = \frac{1}{C_\psi}\int\int W(a,b)\frac{\psi_a^b(x)}{a^2}\,da\,db$$

- continuous in a and b
- $\psi_a^b(t) = |a|^{-1/2}\psi\left(\frac{t-b}{a}\right)$ with "mother wavelet" ψ
- infinitely many (redundant) coefficients $\rightarrow$ not practicable

Multiresolution Analysis/Discrete Wavelet Transform:

- restrict (a,b) to discrete values $(a,b) := \left(\frac{1}{2^j}, \frac{k}{2^j}\right)$
- thus discrete wavelet functions:

$$\psi_{j,k} = \psi_k^{2^{-j}} = 2^{j/2}\psi(2^j t - k)$$

- combines frequency and location: higher spatial resolution for higher frequencies

Part IV

More Complicated Wavelets

Mother and Father Wavelets – General Situation

- **mother wavelet** $\psi(x)$
- **father wavelet** $\phi(x)$, also called **scaling function**
- basis built from scaling functions on each level l:

$$\phi_{l,k}(x) := 2^{l/2}\phi\left(2^l x - k\right) \qquad V_l := \operatorname{span}\{\phi_{l,k}(x)\}$$

- surplus basis built from wavelet functions on each level l:

$$\psi_{l,k}(x) := 2^{l/2}\psi\left(2^l x - k\right) \qquad W_l := \operatorname{span}\{\psi_{l,k}(x)\}$$

- definition of function spaces: $V_{l+1} = V_l \oplus W_l$
- wavelet functions are **orthonormal**:

$$\langle \psi_{l,k}(x), \psi_{m,j}(x) \rangle = \int \psi_{l,k}(x)\psi_{m,j}(x)\,\mathrm{d}x = \begin{cases} 1 & \text{if } l = m \text{ and } k = j \\ 0 & \text{otherwise} \end{cases}$$

- also: scaling functions orthonormal on each level

Scaling and Wavelet Functions

- note: $\phi_{l-1,k} \in V_l \supset V_{l-1}$, and also $\psi_{l-1,k} \in V_l = V_{l-1} \oplus W_{l-1}$
- hence, all $\phi_{l-1,k}$ and $\psi_{l-1,k}$ can be uniquely represented via the basis functions of V_l, i.e., the $\phi_{l,k}$:

$$\phi_{l-1,0}(x) = \sum_i p_i\, \phi_{l,i}(x) = 2^{l/2} \sum_i p_i\, \phi\left(2^l x - i\right)$$

$$\psi_{l-1,0}(x) = \sum_i q_i\, \phi_{l,i}(x) = 2^{l/2} \sum_i q_i\, \phi\left(2^l x - i\right)$$

- p_i and q_i are non-zero for only a few i
- for Haar wavelets:

$$p_0 = \frac{1}{\sqrt{2}},\; p_1 = \frac{1}{\sqrt{2}}, \quad \text{all other } p_i = 0$$

$$q_0 = \frac{1}{\sqrt{2}},\; q_1 = -\frac{1}{\sqrt{2}}, \quad \text{all other } q_i = 0$$

Scaling and Wavelet Functions (2)

- do for all scaling functions $\phi_{l-1,k}$:

$$\begin{aligned}\phi_{l-1,k}(x) &= 2^{l/2}\sum_i p_i\,\phi\left(2^l x - 2k - i\right)\\ &= 2^{l/2}\sum_i p_{i-2k}\,\phi\left(2^l x - i\right) \qquad \text{(with } i \rightsquigarrow 2k+i\text{)}\\ &= \sum_i p_{i-2k}\,\phi_{l,i}(x)\end{aligned}$$

- and similar for wavelet functions: $\psi_{l-1,k}(x) = \sum_i q_{i-2k}\,\phi_{l,i}(x)$
- for Haar wavelets:
  p_{i-2k} and q_{i-2k} are non-zero only for $i = 2k$ and $i = 2k+1$:

$$\begin{aligned}\phi_{l-1,k}(x) &= \tfrac{1}{\sqrt{2}}\,\phi_{l,2k}(x) + \tfrac{1}{\sqrt{2}}\,\phi_{l,2k+1}(x)\\ \psi_{l-1,k}(x) &= \tfrac{1}{\sqrt{2}}\,\phi_{l,2k}(x) - \tfrac{1}{\sqrt{2}}\,\phi_{l,2k+1}(x)\end{aligned}$$

Wavelet Transformations and Filtering

- consider a signal function represented on (fine) level $l+1$:

$$f_{l+1}(x) = \sum_i c_i^{(l+1)} \phi_{l+1,k}(x)$$

- and decomposition $f_{l+1} = f_l + g_l$, where $f_l \in V_l$ and $g_l \in W_l$:

$$\begin{aligned} f_{l+1}(x) &= \sum_i c_i^{(l+1)} \phi_{l+1,i}(x) = \sum_j c_j^{(l)} \phi_{l,j}(x) + \sum_j d_j^{(l)} \psi_{l,j}(x) \\ &= \sum_j \left(c_j^{(l)} \sum_i p_{i-2j} \phi_{l+1,i}(x) \right) + \sum_j \left(d_j^{(l)} \sum_i q_{i-2j} \phi_{l+1,i}(x) \right) \\ &= \sum_i \phi_{l+1,i}(x) \sum_j \left(p_{i-2j} c_j^{(l)} + q_{i-2j} d_j^{(l)} \right) \end{aligned}$$

- two different representations of $f_{l+1}(x)$, but $\{\phi_{l+1,k}(x)\}$ a basis:

$$\Rightarrow \quad c_i^{(l+1)} = \sum_j \left(p_{i-2j} c_j^{(l)} + q_{i-2j} d_j^{(l)} \right)$$

Wavelet Transformations and Filtering (2)

- p_i and q_i determine transformation of coefficients:

$$c_i^{(l+1)} = \sum_j \left(p_{i-2j} c_j^{(l)} + q_{i-2j} d_j^{(l)} \right)$$

- solves assembly:
  for given f_l and g_l (i.e., given coefficients $c_j^{(l)}$ and $d_j^{(l)}$),
  find coefficients $c_i^{(l+1)}$ for f_{l+1}

- for Haar wavelets:

$$\text{even } i: \quad c_i^{(l+1)} = \frac{1}{\sqrt{2}} c_{i/2}^{(l)} + \frac{1}{\sqrt{2}} d_{i/2}^{(l)}$$

$$\text{odd } i: \quad c_i^{(l+1)} = \frac{1}{\sqrt{2}} c_{(i-1)/2}^{(l)} - \frac{1}{\sqrt{2}} d_{(i-1)/2}^{(l)}$$

Wavelet Transformations and Filtering (3)

- now: fine-level representation given as

$$f_{l+1}(x) = \sum_i c_i^{(l+1)} \phi_{l+1,i}(x)$$

- wanted: decomposition $f_{l+1} = f_l + g_l$ with

$$f_l(x) + g_l(x) = \sum_j c_j^{(l)} \phi_{l,j}(x) + \sum_j d_j^{(l)} \psi_{l,j}(x)$$

- use that $\{\phi_{l,k}(x)\}$ and $\{\psi_{l,k}(x)\}$ are **orthonormal** basis for V_l and W_l, and $V_l \perp W_l$:

$$\begin{aligned} \Rightarrow \quad c_j^{(l)} &= \langle f_{l+1}(x), \phi_{l,j}(x) \rangle = \left\langle \sum_i c_i^{(l+1)} \phi_{l+1,i}(x), \phi_{l,j}(x) \right\rangle \\ &= \sum_i c_i^{(l+1)} \langle \phi_{l+1,i}(x), \phi_{l,j}(x) \rangle \\ &= \ldots \end{aligned}$$

Wavelet Transformations and Filtering (4)

- continued:

$$
\begin{aligned}
c_j^{(l)} &= \left\langle f_{l+1}(x), \phi_{l,j}(x) \right\rangle = \ldots \\
&= \sum_i c_i^{(l+1)} \left\langle \phi_{l+1,i}(x), \sum_k p_{k-2j}\, \phi_{l+1,k}(x) \right\rangle \\
&= \sum_i c_i^{(l+1)} \sum_k p_{k-2j} \left\langle \phi_{l+1,i}(x), \phi_{l+1,k}(x) \right\rangle = \sum_i c_i^{(l+1)} p_{i-2j}
\end{aligned}
$$

- similar computation for $d_j^{(l)}$, and therefore:

$$
c_j^{(l)} = \sum_i p_{i-2j}\, c_i^{(l+1)} \qquad d_j^{(l)} = \sum_i q_{i-2j}\, c_i^{(l+1)}
$$

- again, for Haar wavelets:

$$
c_j^{(l)} = \frac{1}{\sqrt{2}}\, c_{2j}^{(l+1)} + \frac{1}{\sqrt{2}}\, c_{2j+1}^{(l+1)} \qquad d_j^{(l)} = \frac{1}{\sqrt{2}}\, c_{2j}^{(l+1)} - \frac{1}{\sqrt{2}}\, c_{2j+1}^{(l+1)}
$$

Wavelet Transformations and Filtering – Summary

Wanted: decomposition $f_{l+1} = f_l + g_l$ with

- coarser representation $f_l(x) = \sum c_j^{(l)} \phi_{l,j}(x)$ with

$$c_j^{(l)} = \sum_i p_{i-2j}\, c_i^{(l+1)}$$

  corresponds to a **low-pass filter** (averaging)

- oscillatory surplus $g_l(x) = \sum d_j^{(l)} \psi_{l,j}(x)$ with

$$d_j^{(l)} = \sum_i q_{i-2j}\, c_i^{(l+1)}$$

  corresponds to a **high-pass filter** (difference computation)

- and reconstruction: $c_i^{(l+1)} = \sum_j \left(p_{i-2j} c_j^{(l)} + q_{i-2j} d_j^{(l)} \right)$

How to Determine the Filtering Coefficients?

- we need coefficients for low-pass and high-pass filter:

$$c_j^{(l)} = \sum_i p_{i-2j}\, c_i^{(l+1)} \qquad d_j^{(l)} = \sum_i q_{i-2j}\, c_i^{(l+1)}$$

- reconstruction then: $c_i^{(l+1)} = \sum_j \left(p_{i-2j} c_j^{(l)} + q_{i-2j} d_j^{(l)} \right)$
- requires **scaling equation** for scaling and wavelet functions:

$$\phi_{l-1,k}(x) = \sum_i p_{i-2k}\, \phi_{l,i}(x) \qquad \psi_{l-1,k}(x) = \sum_i q_{i-2k}\, \phi_{l,i}(x)$$

- requires **orthogonal** scaling and wavelet functions:
  - $\phi_{l,k} \perp \phi_{l,j}$ and $\psi_{l,k} \perp \psi_{l,j}$ for $k \neq j$
  - $\psi_{l,k} \perp \phi_{m,j}$ if $m \leq l$ and arbitrary k, j (i.e., $W_l \perp V_m$)

How to Determine the Wavelet Functions? (2)

- **scaling equation** for mother and father wavelet:

$$\phi(x) = \sqrt{2}\sum_k p_k\,\phi(2x-k) \qquad \psi(x) = \sqrt{2}\sum_k q\,\phi(2x-k)$$

  also called **dilation equation**

- for Haar wavelet:

$$\phi(x) = \phi(2x) + \phi(2x-1) \qquad \psi(x) = \phi(2x) - \phi(2x-1)$$

- for more complicated wavelets:
  - more than 2 non-zeros p_k (and q_k)
  - p_k and q_k determined to satisfy orthogonality
  - often no analytical expression for $\phi(x)$ and $\psi(x)$ available
  - obtain $\phi(x)$ and $\psi(x)$ as solutions of the scaling equation
    $\rightarrow$ see worksheet “cranking the machine”

Towards More Complicated Wavelets

"Wish List:"

- orthonormal basis of scaling functions on each level:

$$\langle \phi_{l,k}(x), \phi_{l,j}(x) \rangle = \begin{cases} 1 & \text{if and } k = j \\ 0 & \text{otherwise} \end{cases}$$

- scaling/wavelet functions obey top scaling equation:

$$\phi_{l-1,k}(x) = \sum_i p_{i-2k}\, \phi_{l,i}(x) \qquad \psi_{l-1,k}(x) = \sum_i q_{i-2k}\, \phi_{l,i}(x)$$

- scaling/wavelet functions have **compact support**
  $\rightsquigarrow$ $p_i \neq 0$ only for few i (same for q_i)

- "vanishing moments" of wavelet functions:

$$\int \psi(t)\, dt = 0 \qquad \int t\psi(t)\, dt = 0 \qquad \text{etc.}$$

Towards More Complicated Wavelets (2)

orthonormal basis of scaling functions:

- on each level:

$$\langle \phi_{l,k}(x), \phi_{l,j}(x) \rangle = \begin{cases} 1 & \text{if and } k = j \\ 0 & \text{otherwise} \end{cases}$$

- combine with scaling equation and compact support:

$$\phi_{l-1,k}(x) = \sum_i p_{i-2k}\, \phi_{l,i}(x) \qquad \text{where } p_i \neq 0 \text{ only for few } i$$

and obtain:

$$\langle \phi_{l-1,k}(x), \phi_{l-1,m}(x) \rangle = \left\langle \sum_i p_{i-2k}\, \phi_{l,i}(x), \sum_j p_{j-2m}\, \phi_{l,j}(x) \right\rangle$$

$$= \sum_i p_{i-2k} \sum_j p_{j-2m} \langle \phi_{l,i}(x), \phi_{l,j}(x) \rangle = \sum_i p_{i-2k} p_{i-2m}$$

- in particular (for $k = m$): $\sum_i (p_{i-2k})^2 = \sum p_i^2 = 1$

Towards More Complicated Wavelets (3)

- in addition – for $k = 0$ and arbitrary $m \neq 0$:

$$\langle \phi_{l-1,0}(x), \phi_{l-1,m}(x) \rangle = \sum_i p_i p_{i-2m} = 0$$

- similar argument: scaling and wavelet functions are orthogonal!

$$\langle \phi_{l-1,0}(x), \psi_{l-1,0}(x) \rangle = \left\langle \sum_i p_i \, \phi_{l,i}(x), \sum_j q_j \, \phi_{l,j}(x) \right\rangle$$

$$= \sum_i p_i \sum_j q_j \left\langle \phi_{l,i}(x), \phi_{l,j}(x) \right\rangle = \sum_i p_i q_i \stackrel{!}{=} 0$$

- and wavelet functions of one level are orthogonal:

$$\langle \psi_{l,k}(x), \psi_{l,m}(x) \rangle = 0 \quad \rightsquigarrow \sum_i q_i q_{i-2(k-m)} = \begin{cases} 0 & \text{if } k \neq m \\ 1 & \text{if } k = m \end{cases}$$

- to satisfy these requirements: $q_k = (-1)^k p_{1-k}$

Daubechies Wavelets (D4)

- setting: $\phi(x) = 0$ outside of interval $[0, 3]$
  $\rightarrow$ non-zero coefficients are p_0, p_1, p_2, and p_3
- orthogonality requires $\sum p_i^2 = 1$ and $\sum p_i p_{i-2m} = 0$:

$$p_0^2 + p_1^2 + p_2^2 + p_3^2 = 1 \quad \text{and} \quad p_0 p_2 + p_1 p_3 = 0$$

- plus vanishing moments $\int \psi(t)\,dt = 0$ and $\int t\psi(t)\,dt = 0$ together with $q_k = (-1)^k p_{1-k}$ leads to

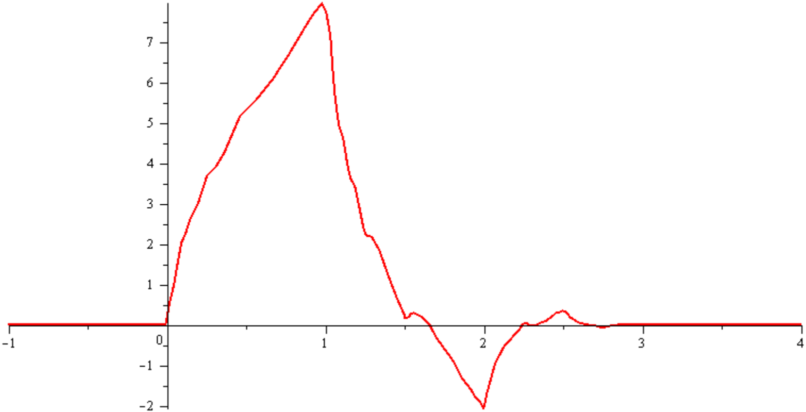
$$-p_0 + p_1 - p_2 + p_3 = 0 \quad \text{and} \quad -p_1 + 2p_2 - p_3 = 0$$

- one solution to this system:

$$p_0 = \frac{1+\sqrt{3}}{4\sqrt{2}}, \quad p_1 = \frac{3+\sqrt{3}}{4\sqrt{2}}, \quad p_2 = \frac{3-\sqrt{3}}{4\sqrt{2}}, \quad p_3 = \frac{1-\sqrt{3}}{4\sqrt{2}}$$

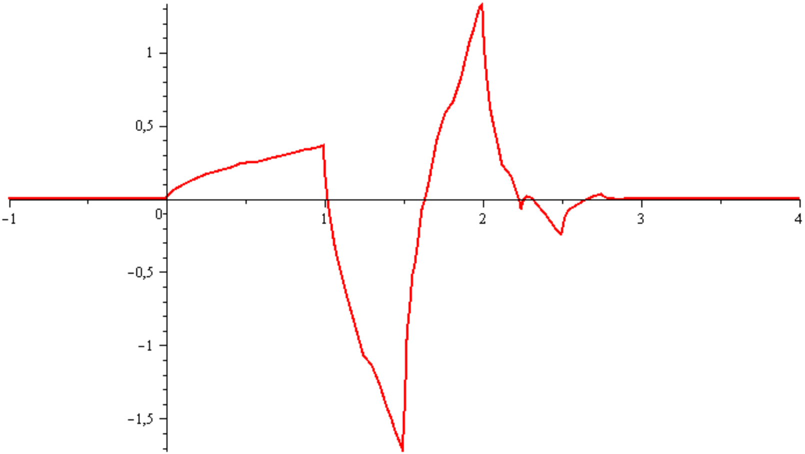
Daubechies Wavelets (D4) – Scaling Function

no analytical expression available $\rightarrow$ iterative approximation



see tutorials: $\rightarrow$ **“cranking the machine”**

Daubechies Wavelets (D4) – Wavelet Function



Finally: Multiresolution Analysis

Definition: **Multiresolution Analysis**

- nested sequence of function spaces:

$$\cdots \subset V_0 \subset V_1 \subset V_2 \subset V_3 \subset \dots$$

- with a scaling function ϕ such that $\phi(2^l x - k)$ is an orthonormal Basis of V_l (and $V_l = \text{span}\{\phi_{l,k} \colon k \in \mathbb{Z}\}$)
- $\bigcup V_l$ is **dense** in $L^2(\mathbb{R})$
- V_l satisfy **separation property**: $\bigcap V_l = \{0\}$
- $f(t) \in V_l$ if and only if $f(2^{-l}t) \in V_0$

Finally: Multiresolution Analysis

Definition: **Multiresolution Analysis**

- nested sequence of function spaces:

$$\cdots \subset V_0 \subset V_1 \subset V_2 \subset V_3 \subset \ldots$$

- with a scaling function ϕ such that $\phi(2^l x - k)$ is an orthonormal Basis of V_l (and $V_l = \text{span}\{\phi_{l,k} \colon k \in \mathbb{Z}\}$)
- $\bigcup V_l$ is **dense** in $L^2(\mathbb{R})$
- V_l satisfy **separation property**: $\bigcap V_l = \{0\}$
- $f(t) \in V_l$ if and only if $f(2^{-l}t) \in V_0$

Last but not least: find coefficients c_k such that $s(x) \approx \sum c_k \phi_{l,k}(x)$?

$\rightarrow$ use orthogonality: $c_k = \langle s(x), \phi_{l,k}(x) \rangle$ (orthogonal projection to space V_l)