

## Algorithms of Scientific Computing (Algorithmen des Wissenschaftlichen Rechnens) The Wavelet Scaling Function, Haar Wavelets

The wavelet families we look at (e.g. Haar wavelets) are constructed around a *multiresolution analysis*, a nested sequence  $V_n$  of function spaces some of which properties are

$$V_j \subset V_{j+1}, j \in \mathbb{Z} \quad (1)$$

$$\bigcap_{j=-\infty}^{\infty} V_j = \{0\} \quad (2)$$

$$f(t) \in V_l \Leftrightarrow f(2^{-l}t) \in V_0 \quad (3)$$

$$\begin{aligned} V_l &= V_{l-1} \oplus W_{l-1} \\ &= V_{l-2} \oplus W_{l-2} \oplus W_{l-1} \\ &= V_0 \oplus W_0 \oplus W_1 \oplus \dots \oplus W_{l-1}, \end{aligned} \quad (4)$$

with *orthogonal* functions  $f \in V_j$  and  $g \in W_j$ , i.e.  $\langle f, g \rangle = 0$ .

The theory of multiresolution analysis further states the existence of a unique function  $\phi$  which satisfies a so-called *dilation equation* of the form

$$\phi(t) = \sum_{k \in \mathbb{Z}} c_k \cdot \phi(2t - k). \quad (5)$$

and which helps us define *orthonormal nodal bases*  $\{ \phi_{l,k} \}$  for the  $V_l$  with

$$\begin{aligned} \phi_{l,k}(t) &= \phi(2^l t - k) \\ \text{span}\{ \phi_{l,k} \} &= V_l, \quad \langle \phi_{l,k}, \phi_{l,m} \rangle = \delta_{k,m} \quad k, m \in \mathbb{Z}. \end{aligned} \quad (6)$$

The function  $\phi$  is called *father wavelet* or *scaling function*, and together with a *mother wavelet*  $\psi$  it defines the wavelet family. It is not necessary to know a specific formula for  $\phi$ , the dilation equation (5) with its coefficients  $c_k$  together with the theory of multiresolution analysis provide enough information to derive the mother wavelet  $\psi$  as well as *orthonormal wavelet bases*  $\{ \psi_{l,m} \}$  for the  $W_l$  with

$$\begin{aligned} \psi_{l,k}(t) &= \psi(2^l t - k) \\ \text{span}\{ \psi_{l,k} \} &= W_l, \quad \langle \psi_{l,k}, \psi_{l,m} \rangle = \delta_{k,m} \quad k, m \in \mathbb{Z}. \end{aligned} \quad (7)$$

# 1 Cranking The Machine

Typically the scaling function  $\phi$  is not known explicitly, and sometimes a closed-form analytic formula does not even exist. However, for continuous  $\phi$  we can approximate the function to arbitrarily high precision using the “Cascade Algorithm”, a fixed-point method for functions.

In this exercise we want to implement this algorithm by iterating over the expression

$$F(\gamma)(t) = \sum_k c_k \cdot \gamma(2t - k) \quad (8)$$

in order to find the fixed point  $\gamma$  of  $F$ .

Our starting point  $\gamma_0$  will be the hat function

$$\gamma_0(t) = \begin{cases} 1+t & \text{for } -1 \leq t \leq 0 \\ 1-t & \text{for } 0 < t \leq 1 \\ 0 & \text{else} \end{cases} \quad (9)$$

- (i) Over the interval  $[-1; 3]$  plot the approximations of the scaling function  $\phi$  for the Haar wavelet family obtained in the first 7 iterations of the cascade algorithm. Do so by plugging the refinements coefficients  $c_k$ ,  $k = 0, 1$  in (10) into (8) resp. (5).

$$c_0 = c_1 = 1 \quad (10)$$

- (ii) Over the interval  $[-1; 3]$  plot the approximations of the scaling function  $\phi$  for the Daubechies wavelet family obtained in the first 7 iterations of the cascade algorithm. Do so by plugging the refinements coefficients  $c_k$ ,  $k = 0, \dots, 3$  in (11) into (8) resp. (5).

$$c_0 = \frac{1 + \sqrt{3}}{4} \quad c_1 = \frac{3 + \sqrt{3}}{4} \quad c_2 = \frac{3 - \sqrt{3}}{4} \quad c_3 = \frac{1 - \sqrt{3}}{4} \quad (11)$$

## 2 The Haar Wavelet Basis

In this exercise we want to compute the 1- $d$  wavelet transform for the Haar wavelet family and apply it to a signal vector  $\vec{s}$  of length  $m = 2^n$ . The transform can be implemented very efficiently as a “pyramidal algorithm” taking  $\mathcal{O}(m)$  steps. For educational purpose we focus on the  $\mathcal{O}(m^2)$  matrix-based algorithm.

- (i) Write a function that constructs the transformation matrix  $M$  consisting of the basis vectors  $\psi_{l,k}$ ,  $l \leq n$ ,  $0 \leq k \leq 2^n - 1$ .
- (ii) Use Python’s package `numpy.linalg` to invert the matrix.
- (iii) Use the program to compute the transform  $M\vec{s} = \vec{d}$  as well as the reconstructed signal  $M^{-1}\vec{d} = \vec{s}$  of the vector

$$\vec{s} = [1, 2, 3, -1, 1, -4, -2, 4]^T$$

- (iv) Verify the program’s output tracing the steps by hand.