

Introduction to Wavelets

Discrete Wavelet Transform

- A wavelet is a function of zero average centered in the neighborhood of $t=0$ and is normalized

$$\int_{-\infty}^{+\infty} \psi(t) dt = 0$$
$$\|\psi\| = 1$$

- The translations and dilations of the wavelet generate a family of functions over which the signal is projected

$$\psi_{u,s}(t) = \frac{1}{\sqrt{s}} \psi\left(\frac{t-u}{s}\right)$$

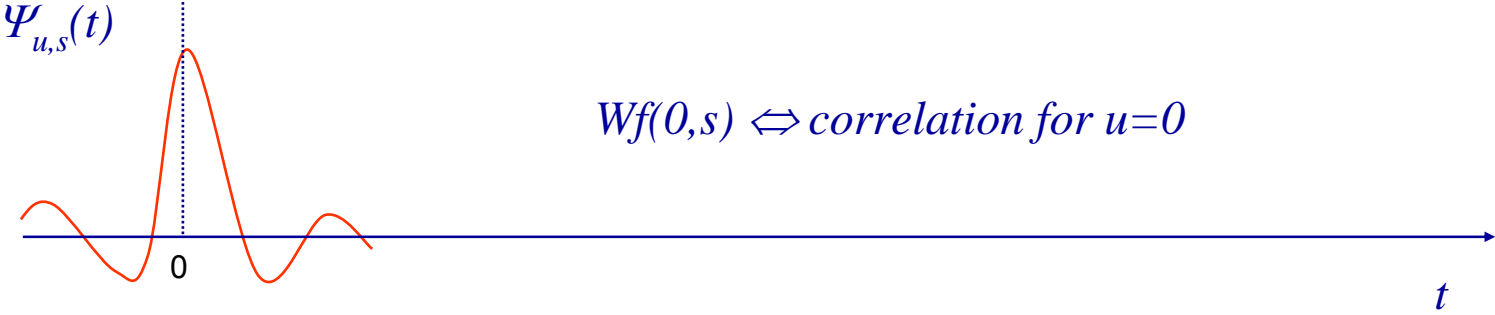
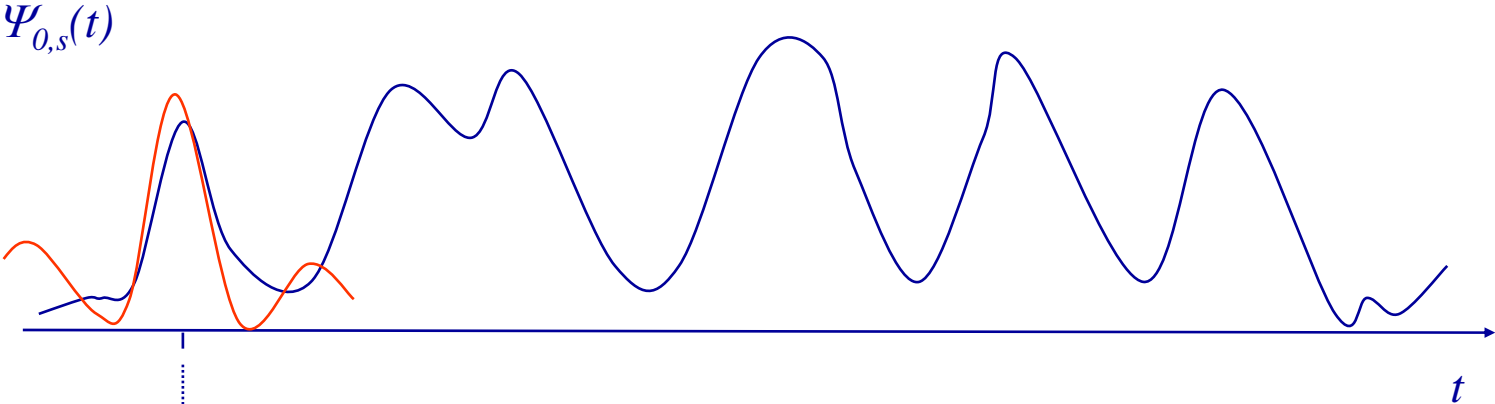
- Wavelet transform of f in $L^2(\mathbb{R})$ at position u and scale s is

$$Wf(u, s) = \langle f, \psi_{u,s} \rangle = \int_{-\infty}^{+\infty} f(t) \frac{1}{\sqrt{s}} \psi^*\left(\frac{t-u}{s}\right) dt$$

$$s = 2^j$$

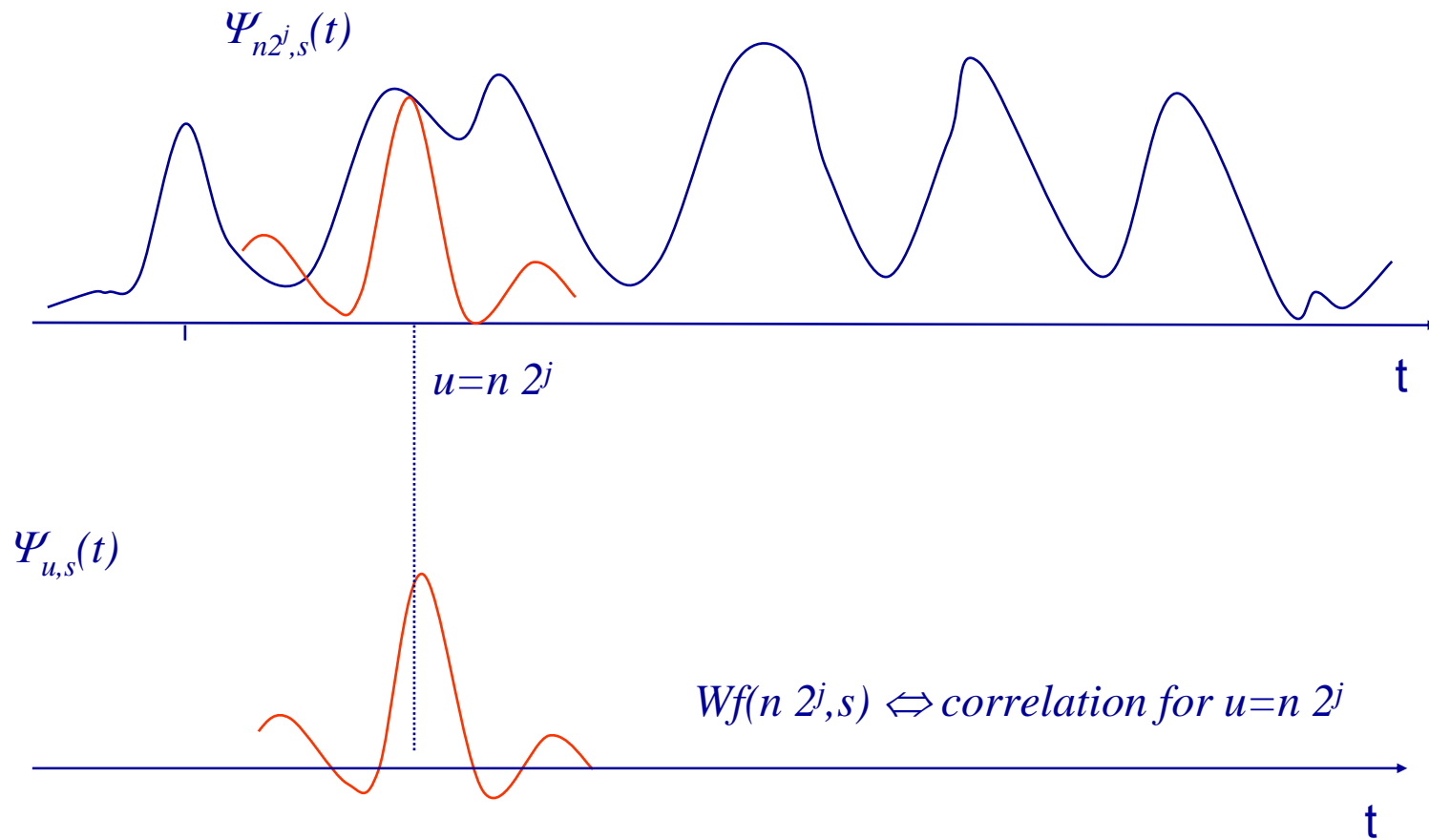
$$u = k \cdot 2^j$$

Wavelet transform

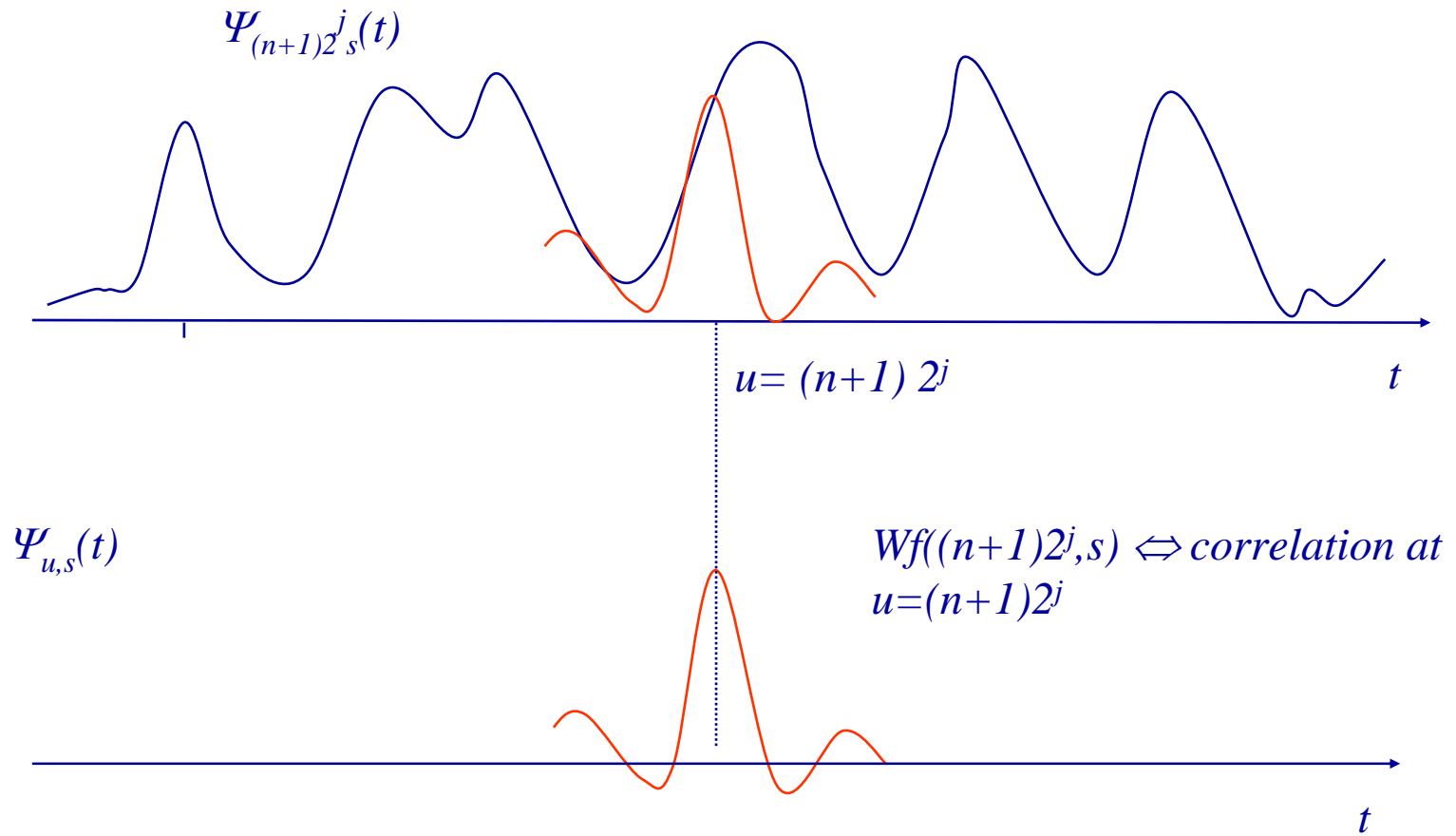


$Wf(0,s) \Leftrightarrow$ correlation for $u=0$

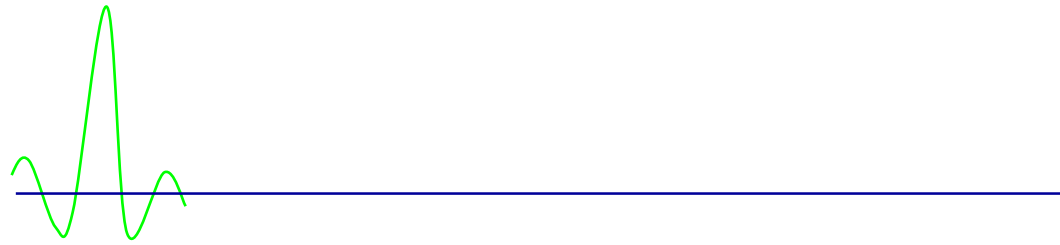
Wavelet transform



Wavelet transform



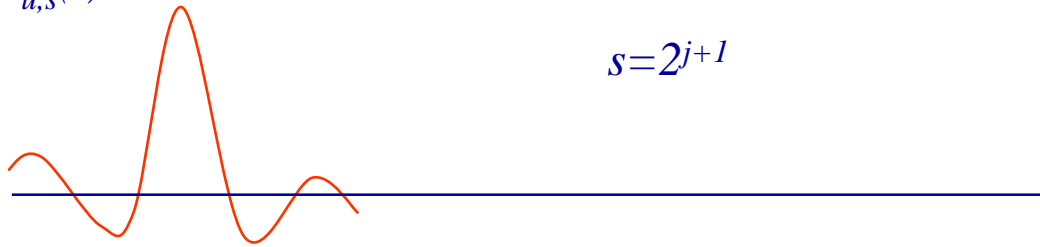
Changing the scale



$\Psi_{u,s}(t)$

$s=2^{j+1}$

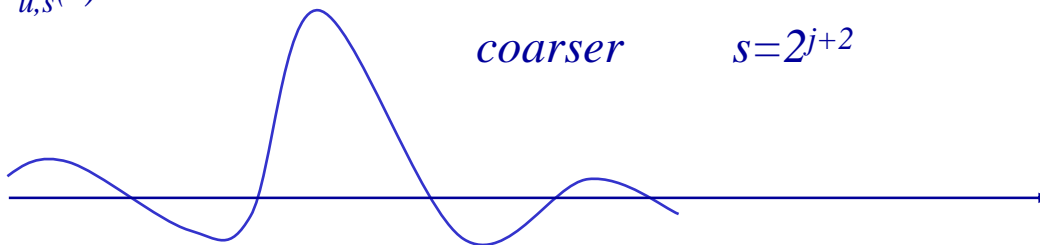
multiresolution



$\Psi_{u,s}(t)$

coarser

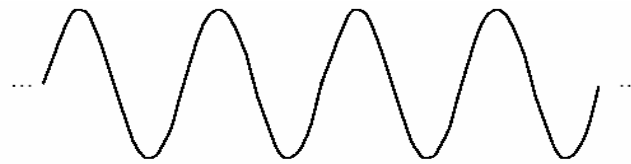
$s=2^{j+2}$



Fourier versus Wavelets

$$F(\omega) = \int_{-\infty}^{\infty} f(t)e^{-j\omega t} dt$$

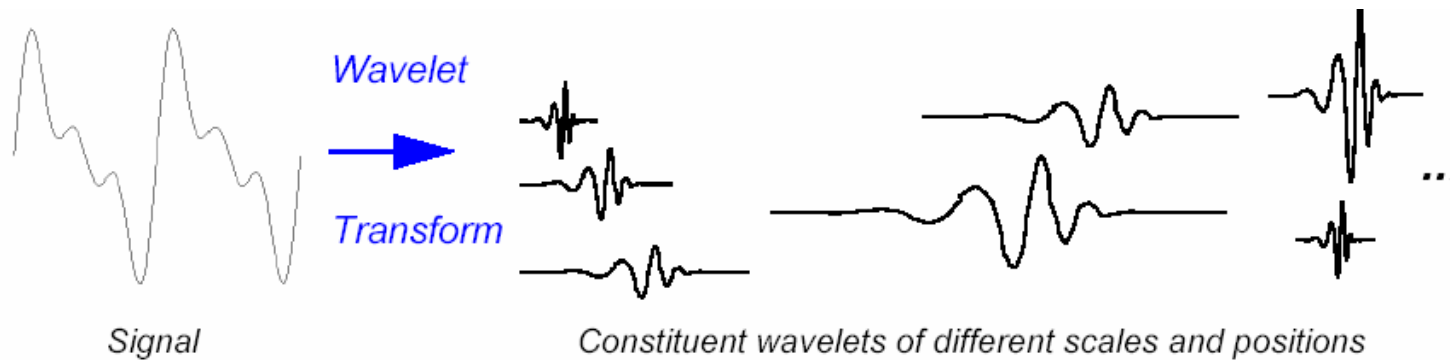
$$C(\text{scale}, \text{position}) = \int_{-\infty}^{\infty} f(t)\psi(\text{scale}, \text{position}, t)dt$$



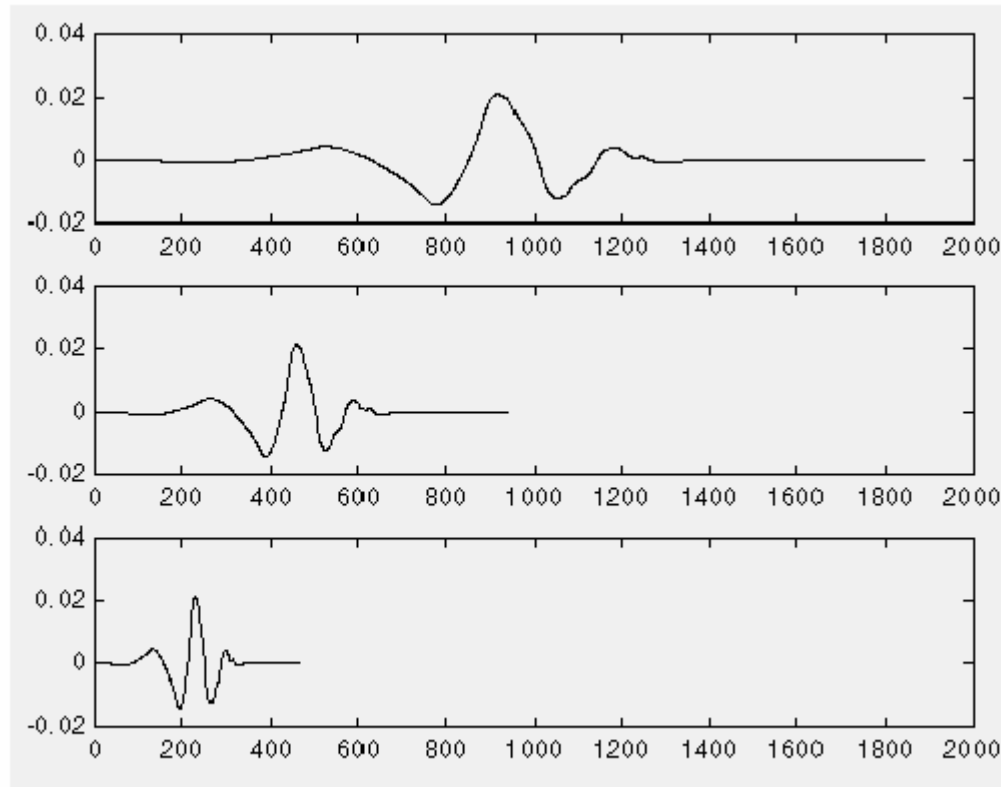
Sine Wave



Wavelet (db10)



Scaling

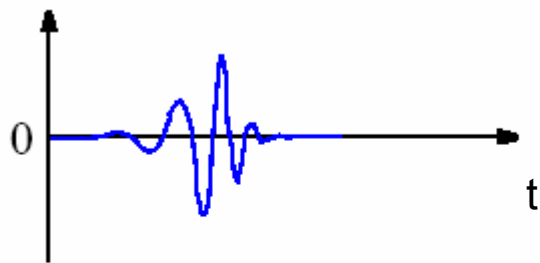


$$f(t) = \psi(t) \quad ; \quad a = 1$$

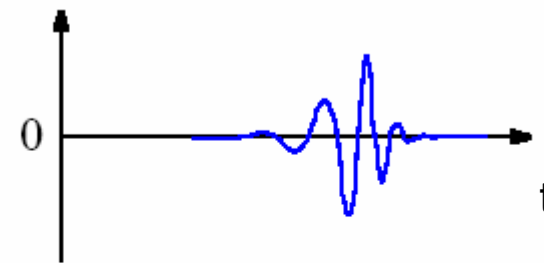
$$f(t) = \psi(2t) \quad ; \quad a = \frac{1}{2}$$

$$f(t) = \psi(4t) \quad ; \quad a = \frac{1}{4}$$

Shifting



Wavelet function
 $\psi(t)$

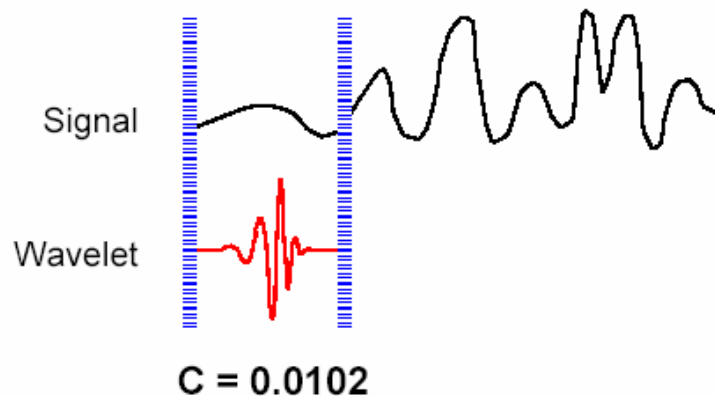


Shifted wavelet function
 $\psi(t-k)$

Recipe

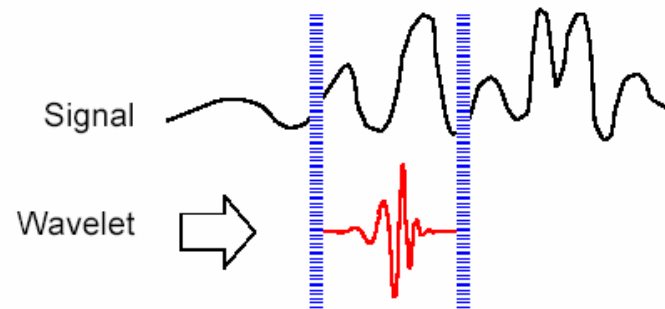
- 1 Take a wavelet and compare it to a section at the start of the original signal.
- 2 Calculate a number, C , that represents how closely correlated the wavelet is with this section of the signal. The higher C is, the more the similarity. More precisely, if the signal energy and the wavelet energy are equal to one, C may be interpreted as a correlation coefficient.

Note that the results will depend on the shape of the wavelet you choose.

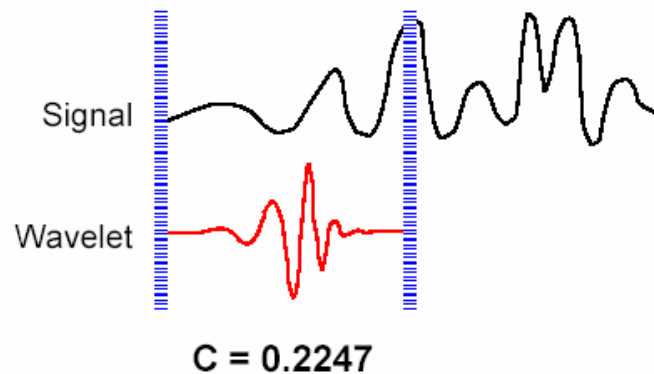


Recipe

- 3 Shift the wavelet to the right and repeat steps 1 and 2 until you've covered the whole signal.



- 4 Scale (stretch) the wavelet and repeat steps 1 through 3.



- 5 Repeat steps 1 through 4 for all scales.

Wavelet Zoom

- WT at position u and scale s measures the local correlation between the signal and the wavelet



Thus, there is a correspondence between wavelet scales and frequency as revealed by wavelet analysis:

- (small) • Low scale $a \Rightarrow$ Compressed wavelet \Rightarrow Rapidly changing details \Rightarrow High frequency ω .
- (large) • High scale $a \Rightarrow$ Stretched wavelet \Rightarrow Slowly changing, coarse features \Rightarrow Low frequency ω .

Frequency domain

- Parseval
$$Wf(u, s) = \int_{-\infty}^{+\infty} f(t) \psi_{u,s}^*(t) dt = \frac{1}{2\pi} \int_{-\infty}^{+\infty} F(\omega) \Psi_{u,s}^*(\omega) d\omega$$

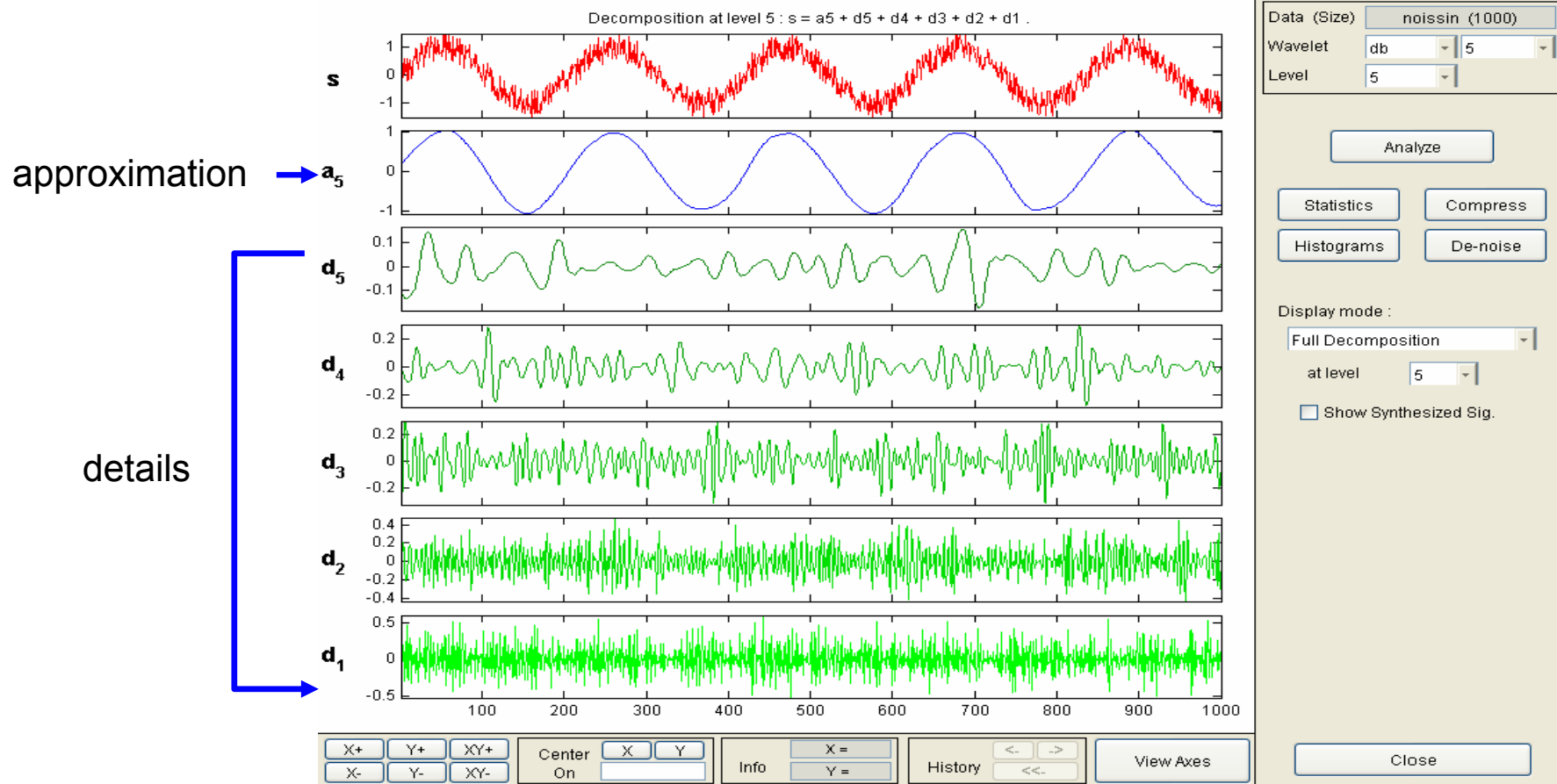
The wavelet coefficients $Wf(u, s)$ depend on the values of $f(t)$ (and $F(\omega)$) in the time-frequency region where the energy of the corresponding wavelet function (respectively, its transform) is concentrated

- time/frequency localization*
- The *position and scale* of high amplitude coefficients allow to characterize the *temporal evolution* of the signal
- Time domain signals (1D) : Temporal evolution
- Spatial domain signals (2D) : Localize and characterize spatial singularities

$$\psi_{u,s}(t) = \frac{1}{\sqrt{s}} \psi\left(\frac{t-u}{s}\right) \Leftrightarrow \Psi_{u,s}(\omega) = \sqrt{s} \Psi(s\omega) e^{-j\omega s}$$

Stretching in time \leftrightarrow Shrinking in frequency (and viceversa)

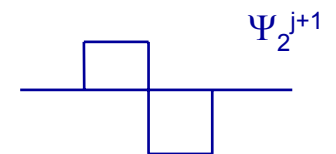
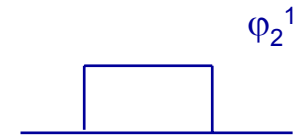
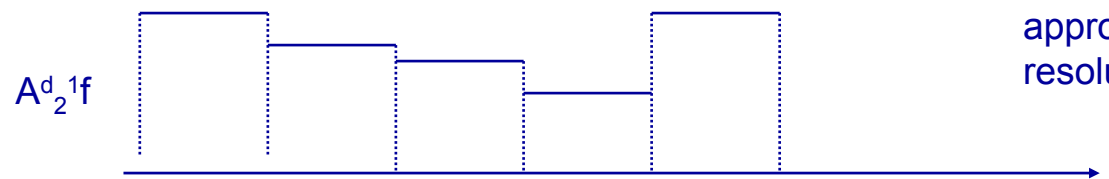
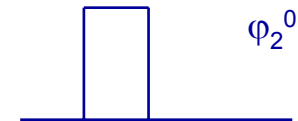
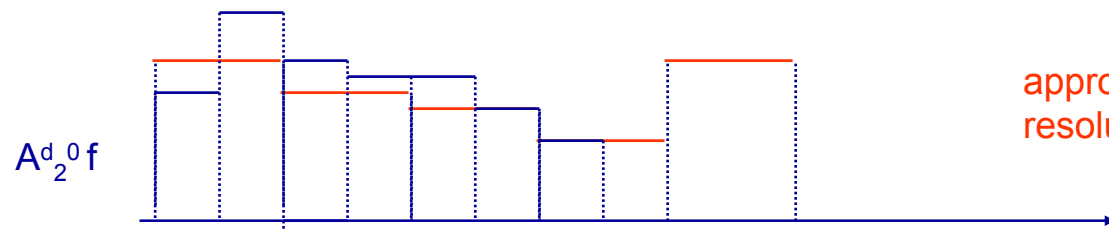
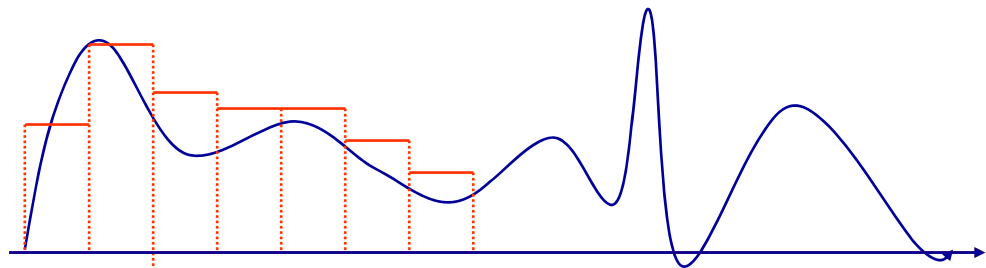
Example



Wavelet representation = approximation + details

approximation ↔ scaling function
details ↔ wavelets

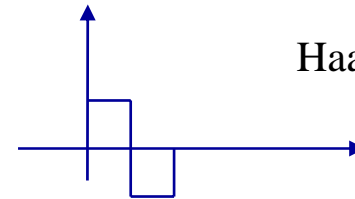
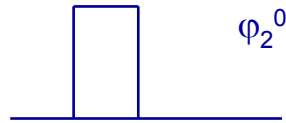
A different perspective



$$A_2^{d,j}f = A_2^{d,j+1}f + d_2^{j+1}f$$

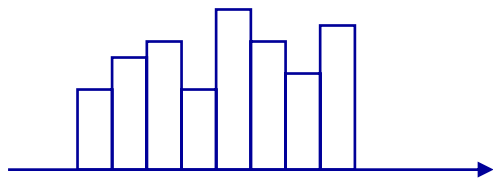
Haar pyramid [Haar 1910]

Haar basis function

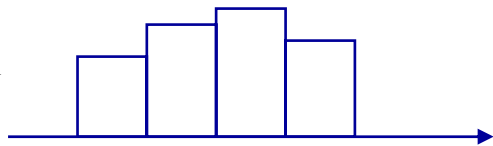


Haar wavelet

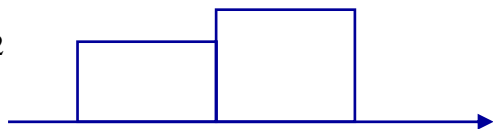
sig₀



sig₁



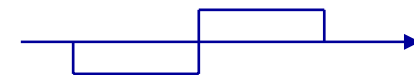
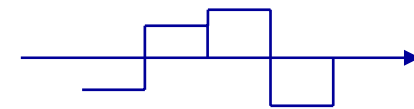
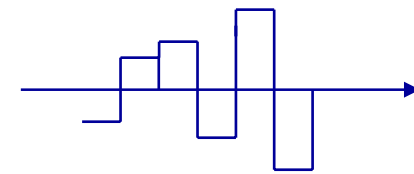
sig₂



sig₃

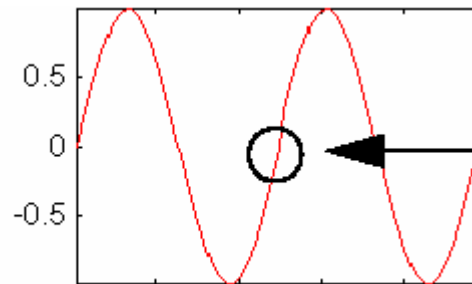


details

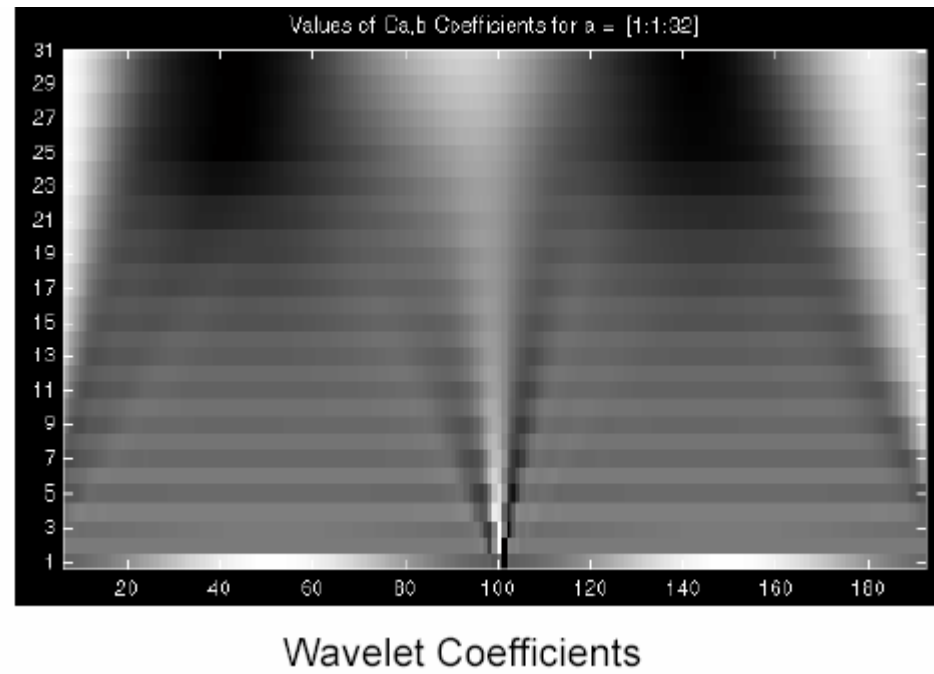
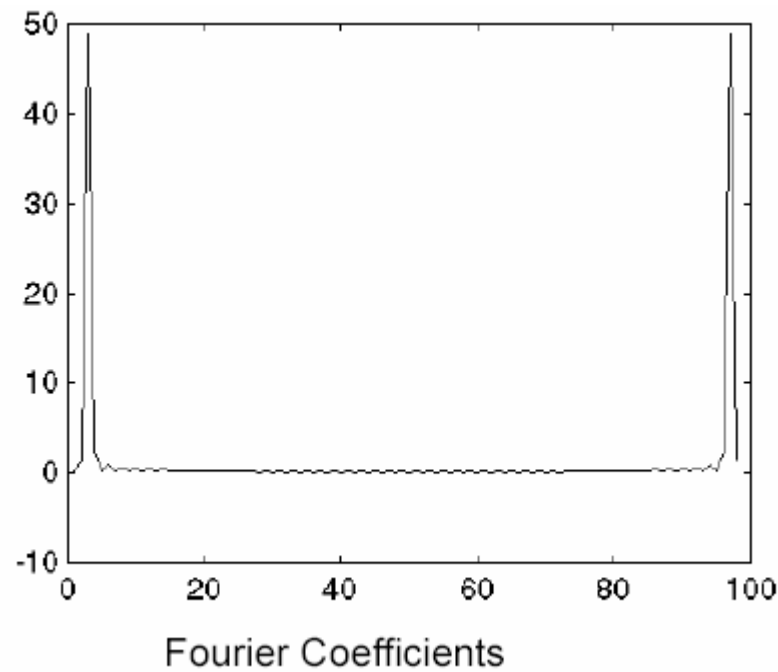


signal=approximation at scale n + details at scales 1 to n

What wavelets can do?



Sinusoid with a small discontinuity



Wavelets and linear filtering

- The WT can be rewritten as a convolution product and thus the transform can be interpreted as a linear filtering operation

$$Wf(u, s) = \langle f, \psi_{u,s} \rangle = \int_{-\infty}^{+\infty} f(t) \frac{1}{\sqrt{s}} \psi^* \left(\frac{t-u}{s} \right) dt = f * \bar{\psi}_s(u)$$

$$\bar{\psi}_s(t) = \frac{1}{\sqrt{s}} \psi^* \left(\frac{-t}{s} \right)$$

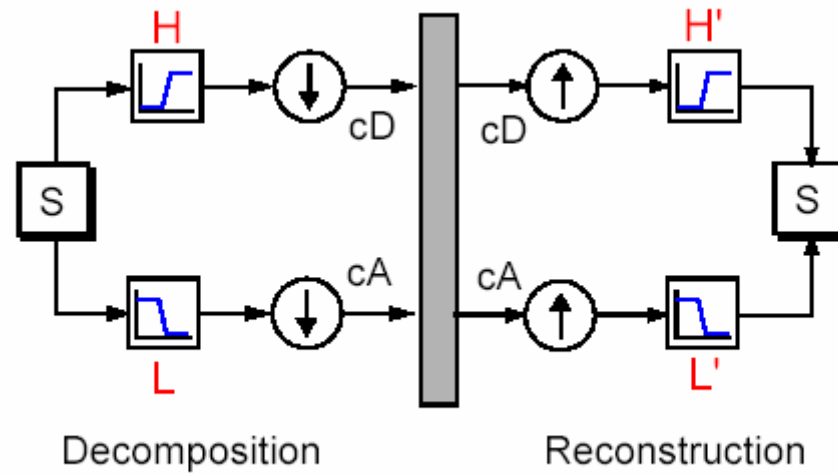
$$\hat{\bar{\psi}}_s(\omega) = \sqrt{s} \hat{\psi}^*(s\omega)$$

$$\hat{\psi}(0) = 0$$

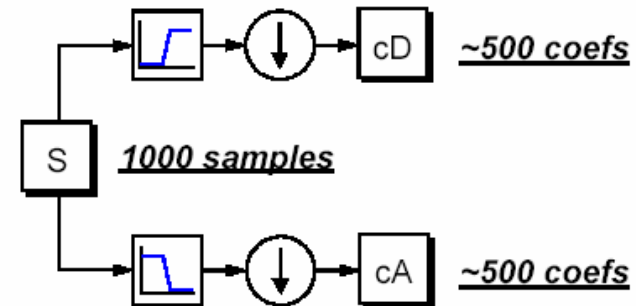
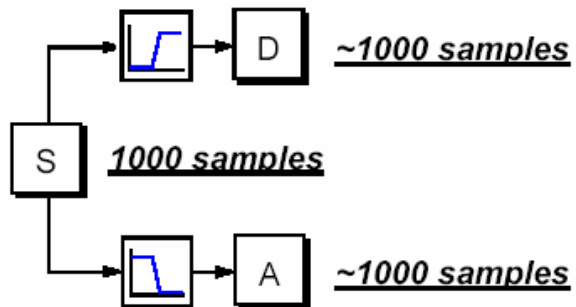
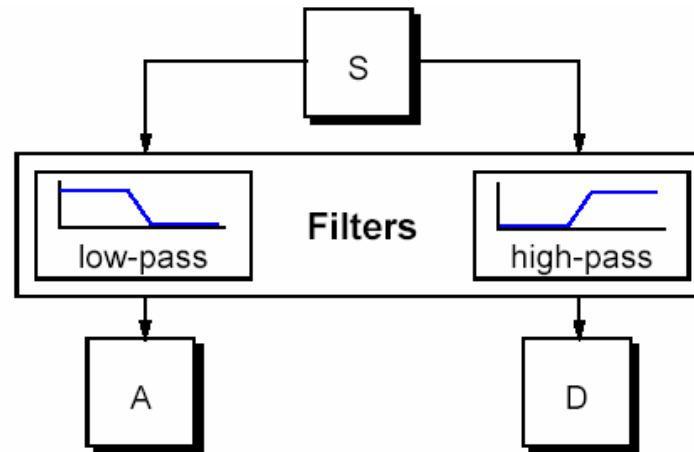
→ band-pass filter

Wavelets & filterbanks

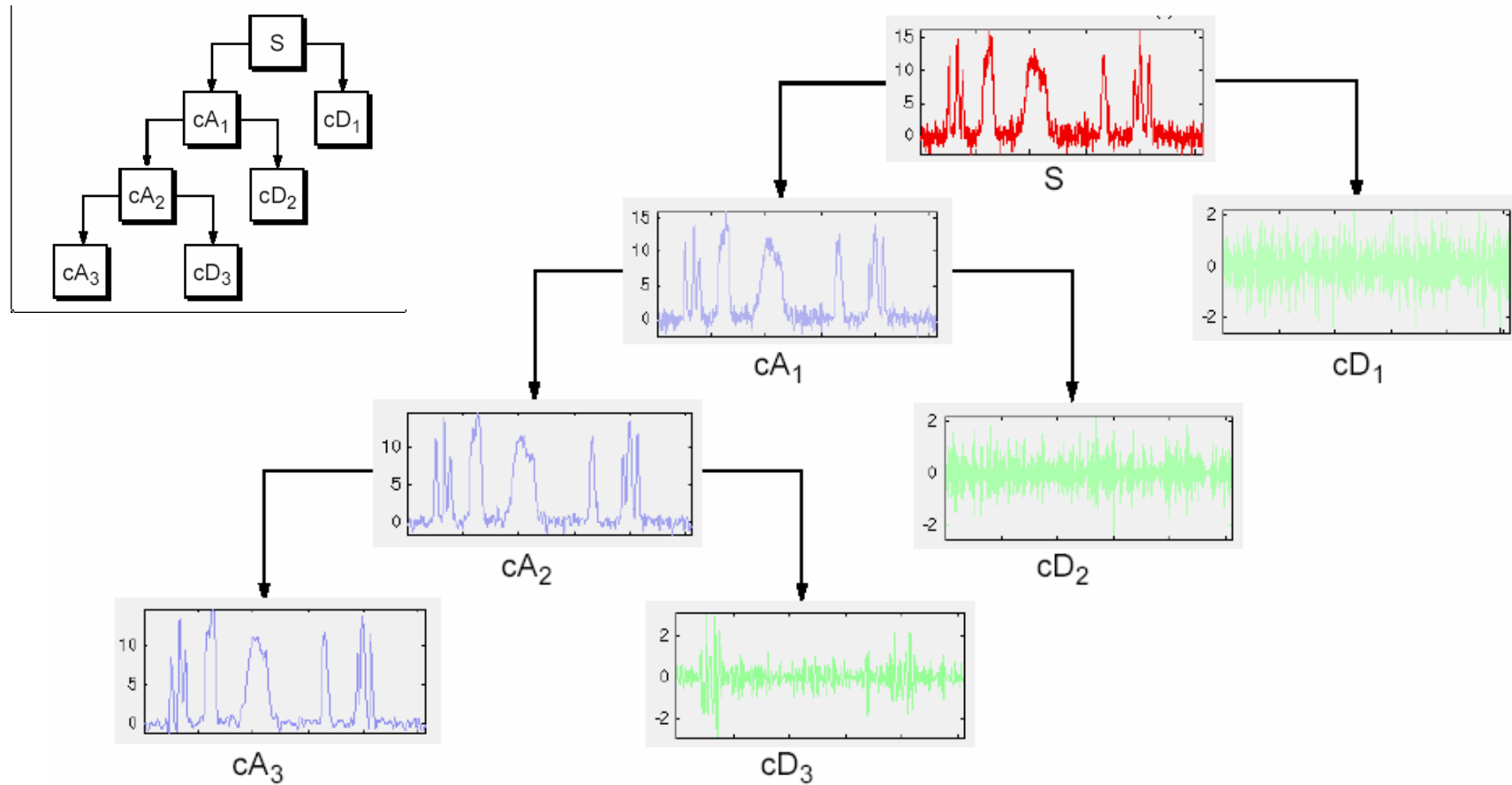
Quadrature Mirror Filter (QMF)



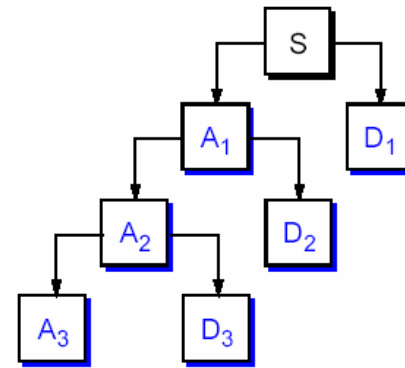
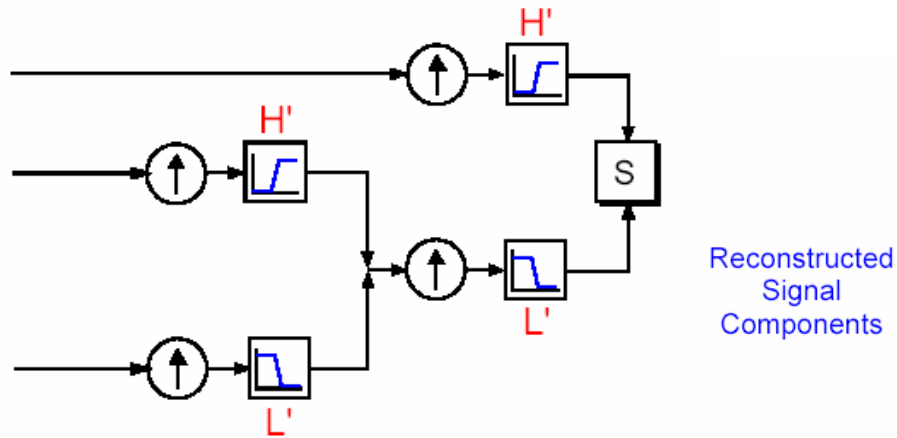
Analysis or decomposition



Analysis or decomposition

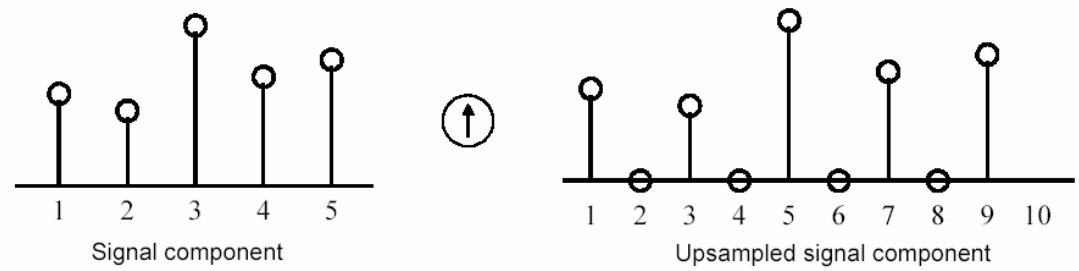


Synthesis or reconstruction

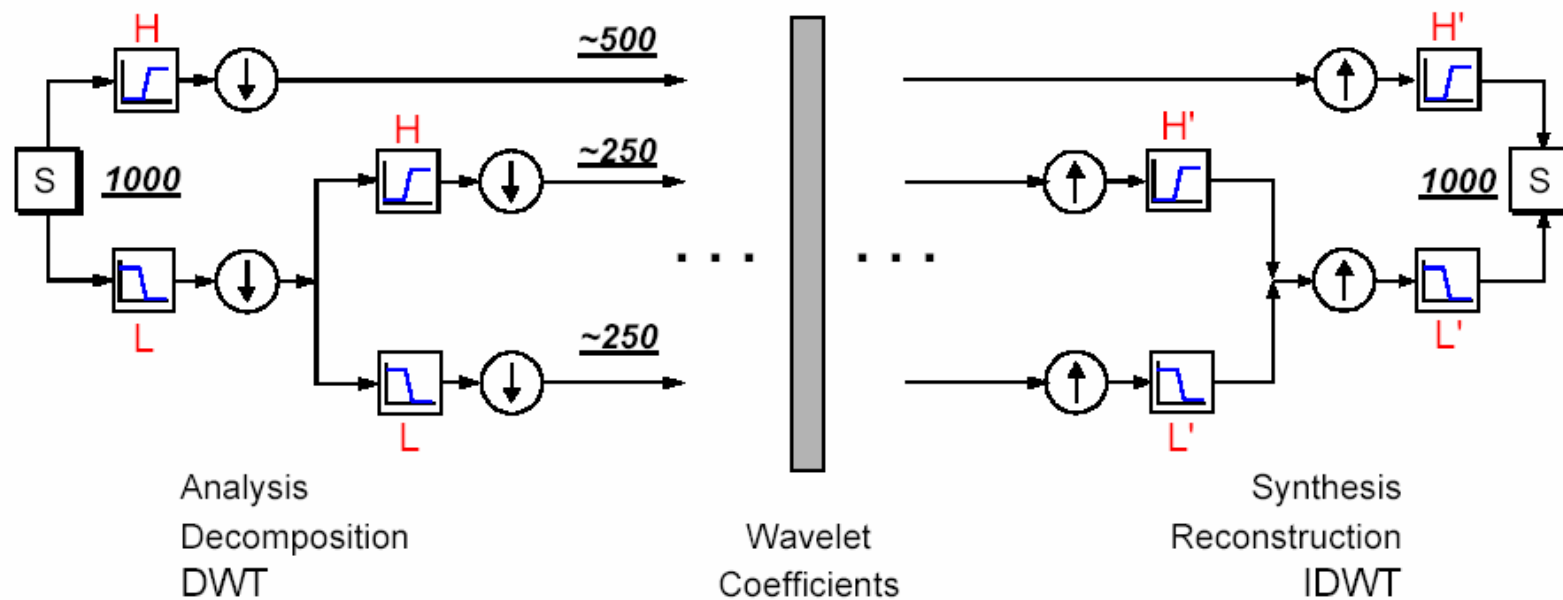


$$\begin{aligned}
 S &= A_1 + D_1 \\
 &= A_2 + D_2 + D_1 \\
 &= A_3 + D_3 + D_2 + D_1
 \end{aligned}$$

upsampling

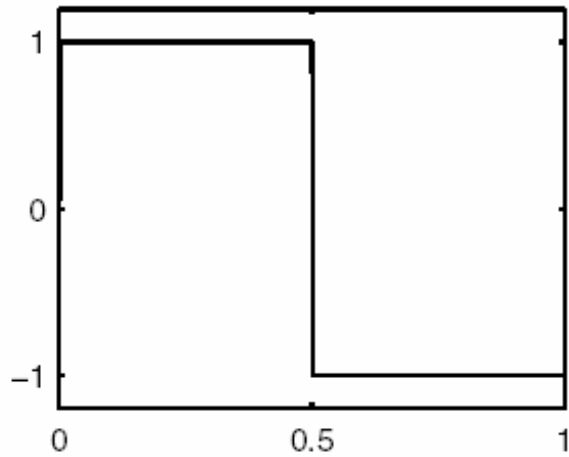


Multi-scale analysis



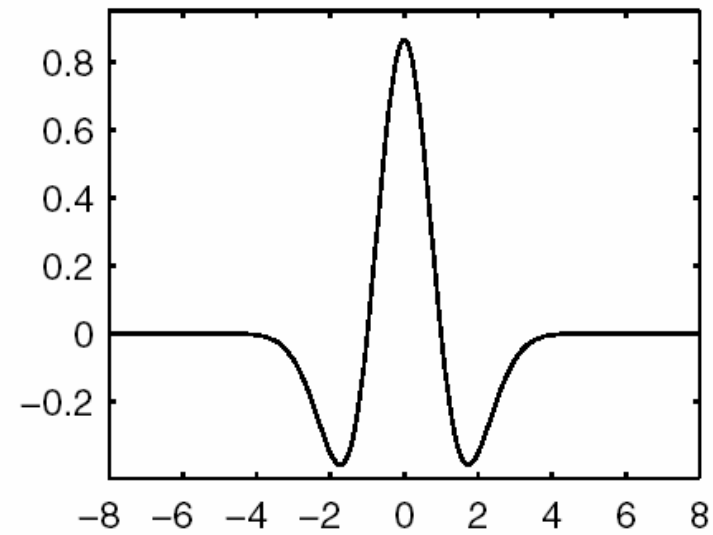
Famous wavelets

Haar



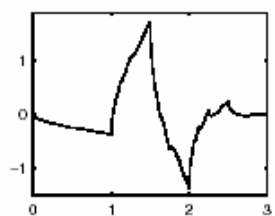
Wavelet function psi

Mexican hat

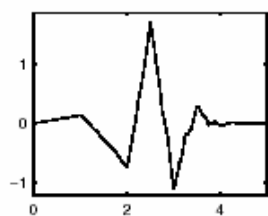


Wavelet function psi

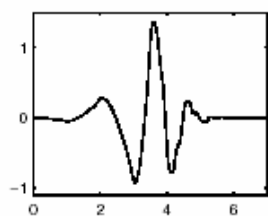
Daubechie's



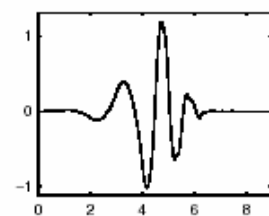
db2



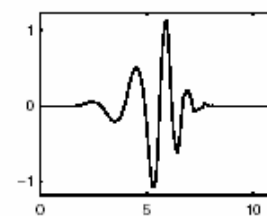
db3



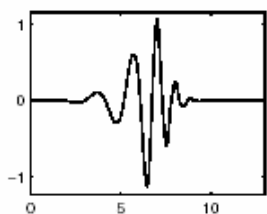
db4



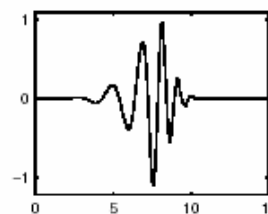
db5



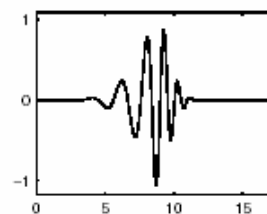
db6



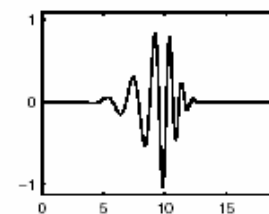
db7



db8



db9



db10

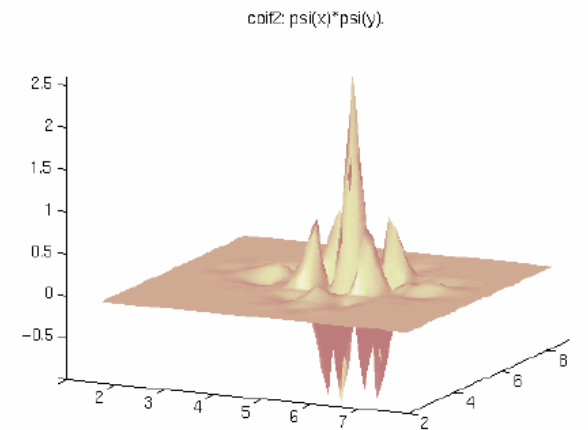
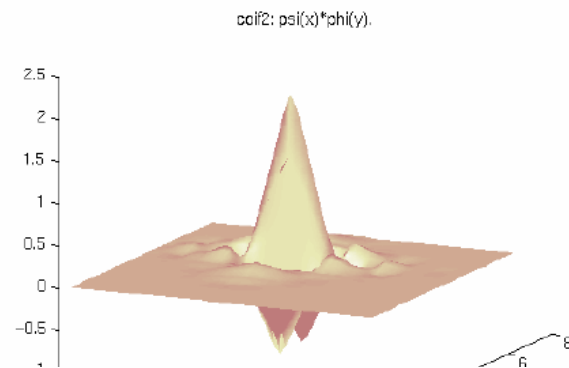
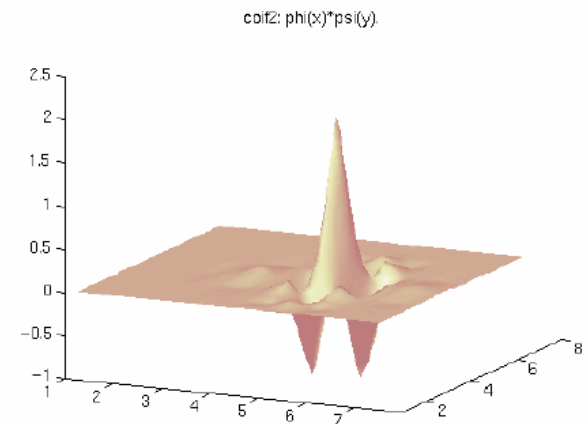
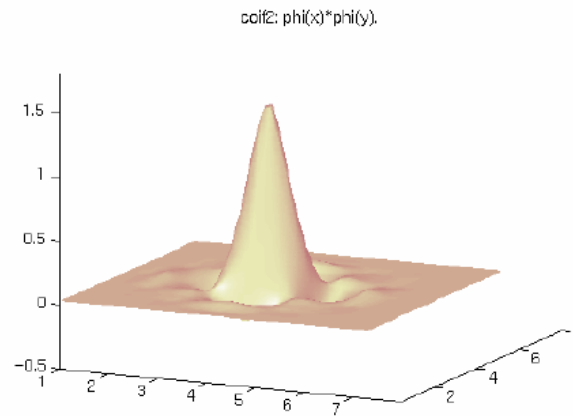
Bi-dimensional wavelets

$$\varphi(x, y) = \varphi(x)\varphi(y)$$

$$\psi^1(x, y) = \varphi(x)\psi(y)$$

$$\psi^2(x, y) = \psi(x)\varphi(y)$$

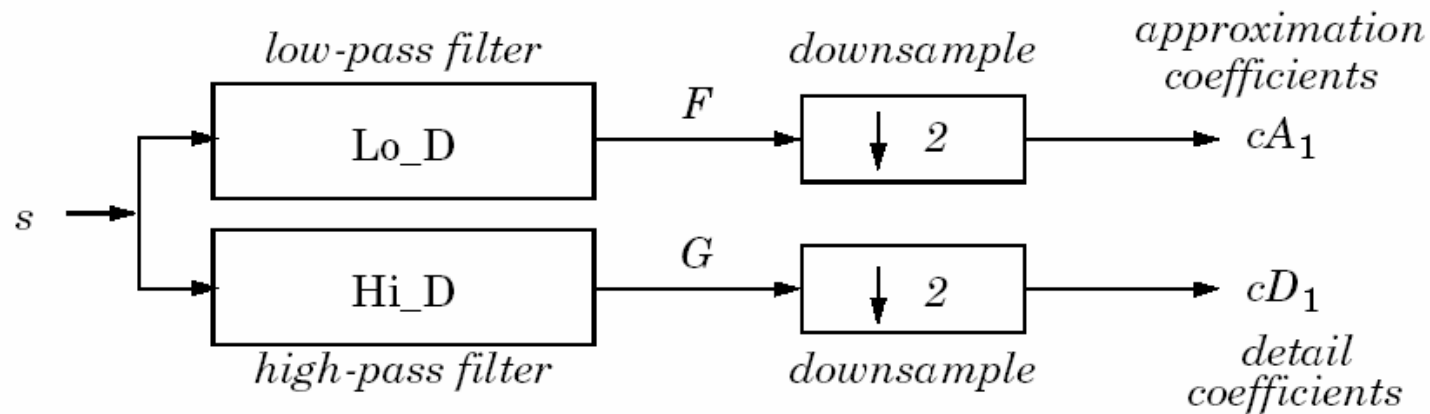
$$\psi^3(x, y) = \psi(x)\psi(y)$$



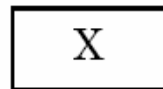
$$\frac{1}{\sqrt{a_1 a_2}} \psi\left(\frac{x_1 - b_1}{a_1}, \frac{x_2 - b_2}{a_2}\right) \text{ where } (x = (x_1, x_2) \in \mathbb{R}^2)$$

Fast wavelet transform algorithm (DWT)

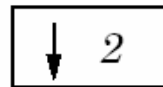
Decomposition step



where



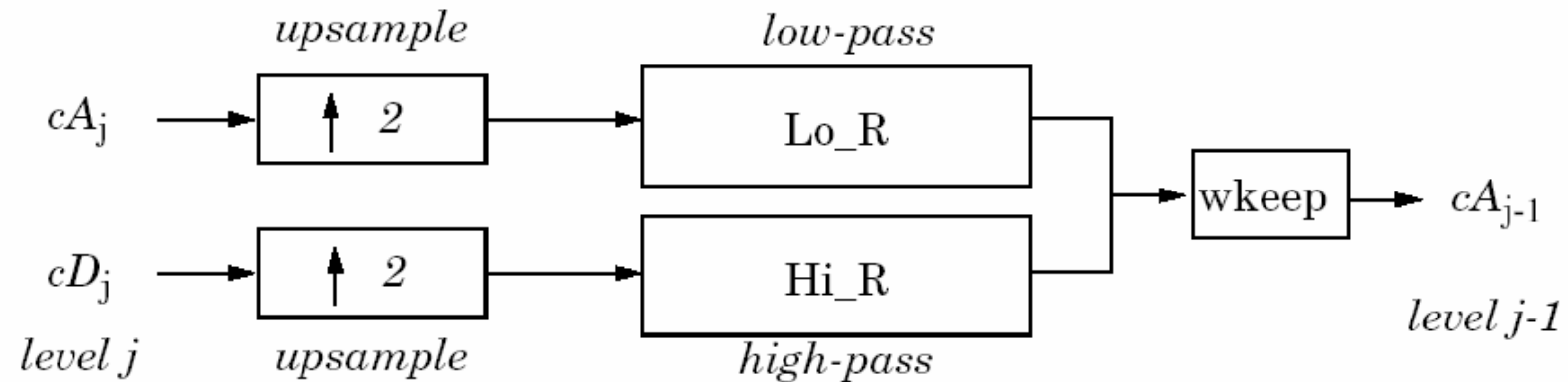
Convolve with filter X.



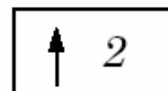
Keep the even indexed elements
(see dyaddown).

Fast wavelet transform algorithm (DWT)

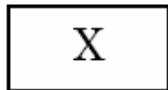
Reconstruction Step



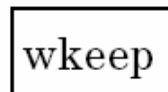
where



Insert zeros at odd-indexed elements.

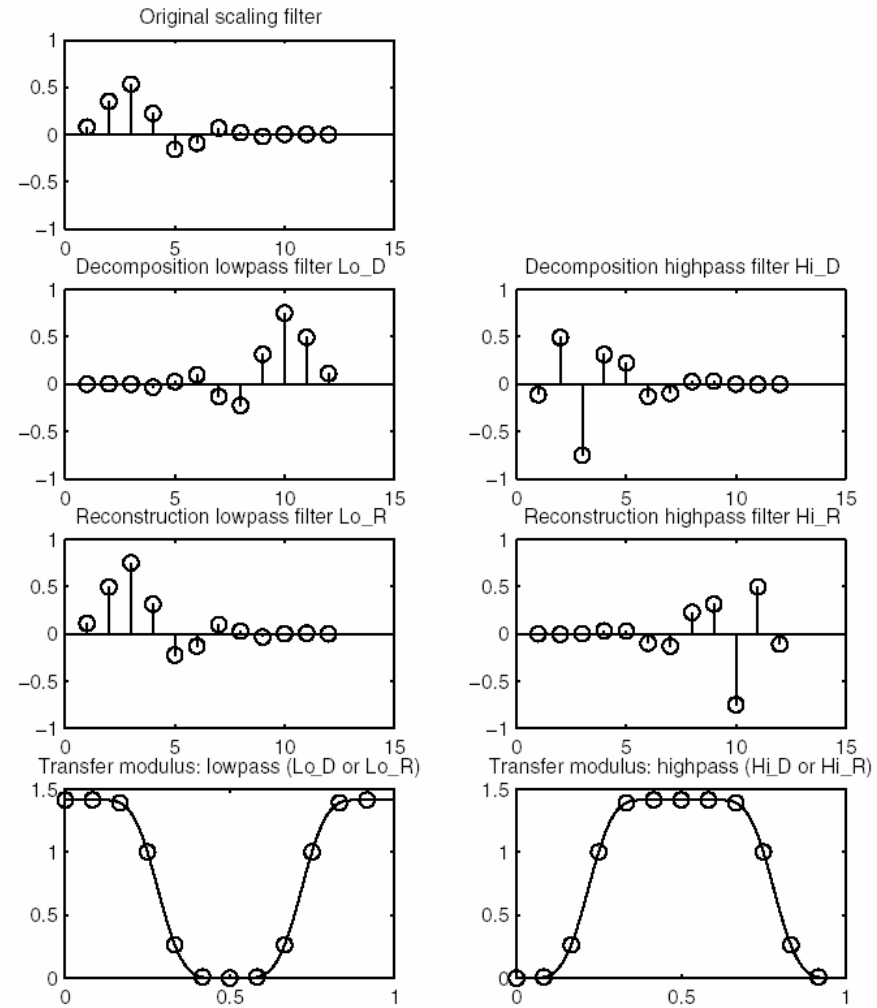


Convolve with filter X.



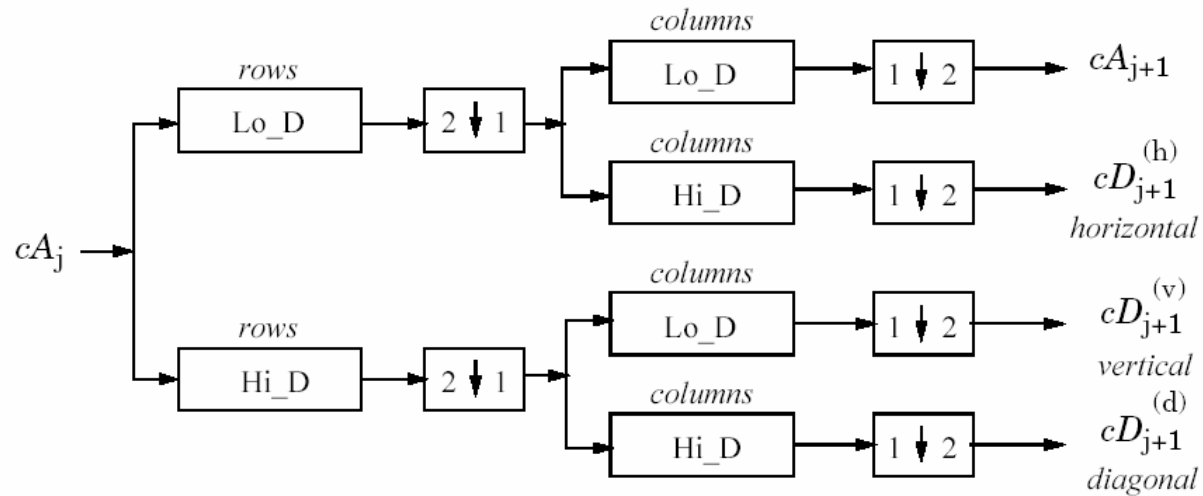
Take the central part of U with the convenient length.

Filters



Fast DWT for images

Decomposition Step



where $\begin{matrix} \boxed{2 \downarrow 1} \end{matrix}$ Downsample columns: keep the even indexed columns.

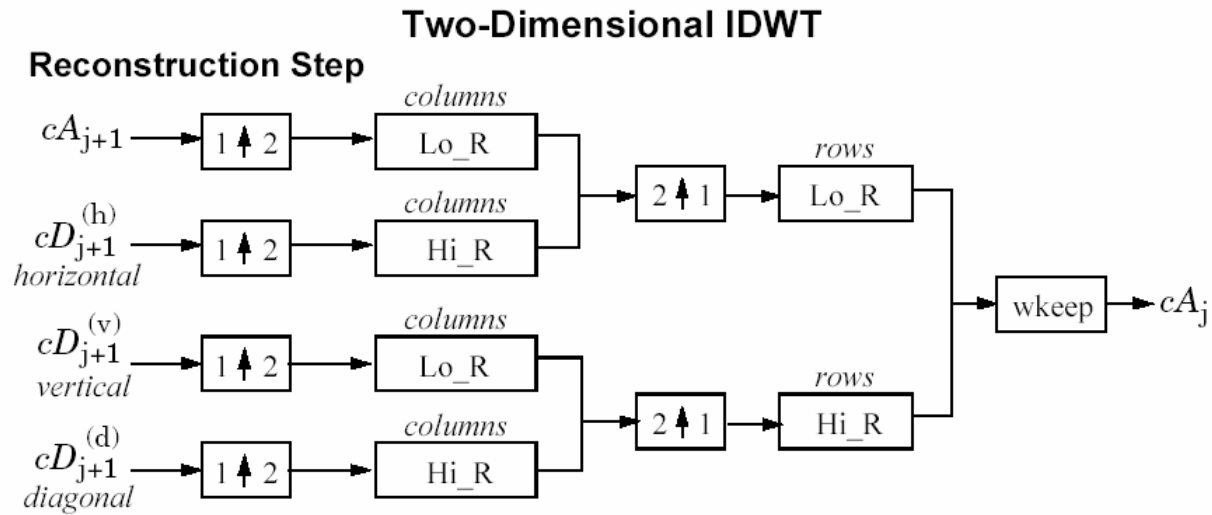
$\begin{matrix} \boxed{1 \downarrow 2} \end{matrix}$ Downsample rows: keep the even indexed rows.

$\begin{matrix} \text{rows} \\ \boxed{X} \end{matrix}$ Convolve with filter X the rows of the entry.

$\begin{matrix} \text{columns} \\ \boxed{X} \end{matrix}$ Convolve with filter X the columns of the entry.

Initialization $CA_0 = s$ for the decomposition initialization.

Fast DWT for images



where

- $2 \uparrow 1$ Upsample columns: insert zeros at odd-indexed columns.
- $1 \uparrow 2$ Upsample rows: insert zeros at odd-indexed rows.
- $\begin{matrix} \text{rows} \\ \boxed{X} \end{matrix}$ Convolve with filter X the rows of the entry.
- $\begin{matrix} \text{columns} \\ \boxed{X} \end{matrix}$ Convolve with filter X the columns of the entry.

Subband structure for images

