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Image Pyramids

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Gaussian pyramid



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Downsampling and Upsampling

- **Downsampling** by a factor of 2 (signal with even number of samples):

$$\downarrow_2 (a_0, a_1, a_2, \dots, a_{2N-1}) = (a_0, a_2, a_4, \dots, a_{2N-2})$$

- **Upsampling** by a factor of 2:

$$\uparrow_2 (a_0, a_1, a_2, \dots, a_{N-1}) = (a_0, 0, a_1, 0, a_2, 0, \dots, a_{N-1}, 0)$$

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Image pyramid

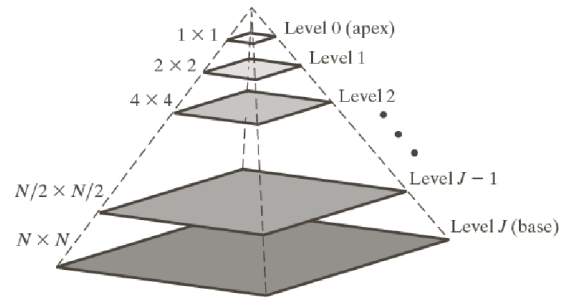


Image of size $N \times N$, with $N = 2^J$.

In a $J + 1$ -level pyramid, the total number of pixels is:

$$N^2 \left(1 + \frac{1}{4^1} + \frac{1}{4^2} + \cdots + \frac{1}{4^J} \right) \leq \frac{4}{3} N^2$$

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Wavelets

The Mathematical Microscope

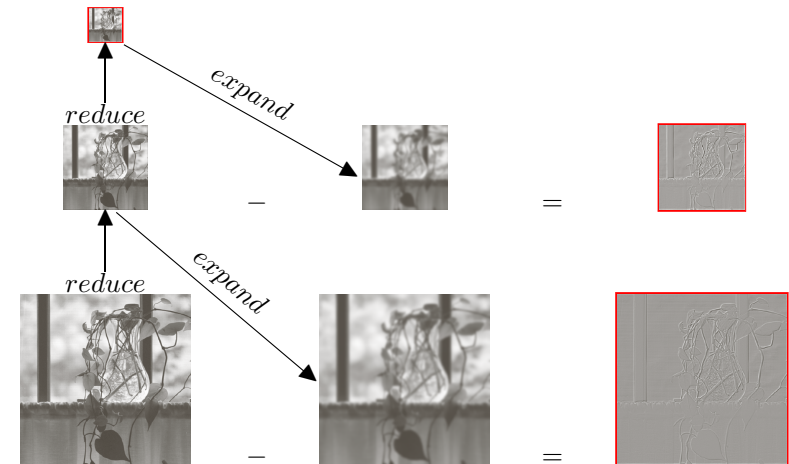
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Laplacian pyramid

Approximations

Predictions

Residuals

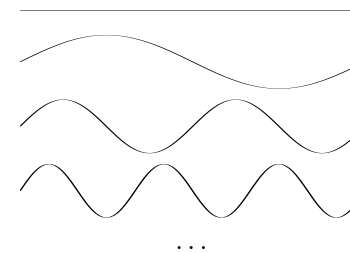


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1-D Fourier analysis



Weighted sum of basic signals



coefficient c_0

coefficient c_1

coefficient c_2

coefficient c_3

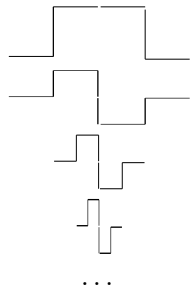
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1-D Wavelets: Haar wavelet



Weighted sum of basic signals



approximation coefficient c_3

detail coefficient d_3

detail coefficient d_2

detail coefficient d_1

...

...

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Simple example: decomposition

Signal decomposition:

$$c_0 = (9, 1, 2, 0)$$

$$c_1 = (\overbrace{9, 1}^{\text{mean}}, \overbrace{2, 0}^{\text{mean}}) = (5, 1) \quad \text{approximation}$$

$$d_1 = (\overbrace{9, 1}^{\text{diff}}, \overbrace{2, 0}^{\text{diff}}) = (4, 1) \quad \text{detail}$$

$$c_2 = (\overbrace{5, 1}^{\text{mean}}) = (3) \quad \text{approximation}$$

$$d_2 = (\overbrace{5, 1}^{\text{diff}}) = (2) \quad \text{detail}$$

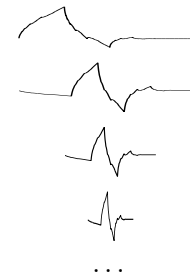
$$\text{mean}(a, b) = (a + b)/2. \quad \text{diff}(a, b) = (a - b)/2.$$

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1-D Wavelets: Daubechies wavelet



Weighted sum of basic signals



approximation coefficient c_3

detail coefficient d_3

detail coefficient d_2

detail coefficient d_1

...

...

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Simple example: reconstruction

Signal reconstruction.

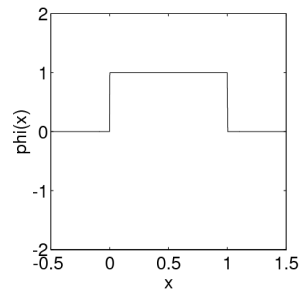
From the coefficients in $c_2 = (3)$, $d_2 = (2)$, $d_1 = (4, 1)$ we get:

$$c_0 = \begin{pmatrix} 9 \\ 1 \\ 2 \\ 0 \end{pmatrix} = 3 \begin{pmatrix} 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} + 2 \begin{pmatrix} 1 \\ 1 \\ -1 \\ -1 \end{pmatrix} + 4 \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + 1 \begin{pmatrix} 0 \\ 0 \\ 1 \\ -1 \end{pmatrix}$$

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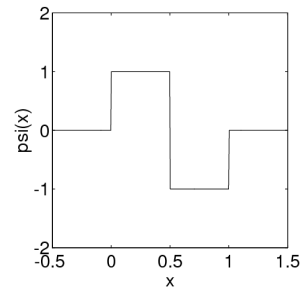
Haar scaling function

$$\phi(x) = \begin{cases} 1 & -\frac{1}{2} \leq x \leq \frac{1}{2} \\ 0 & \text{otherwise} \end{cases}$$



Haar wavelet

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



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Orthonormal basis functions:

$$\phi_{j,k}(x) = 2^{-j/2} \phi(2^{-j}x - k) \quad \text{scaling functions}$$

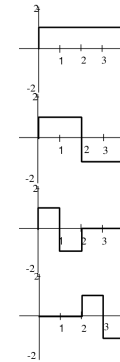
$$\psi_{j,k}(x) = 2^{-j/2} \psi(2^{-j}x - k) \quad \text{wavelet functions}$$

$$\phi_{2,0}(x) \leftrightarrow \frac{1}{2}(1, 1, 1, 1)$$

$$\psi_{2,0}(x) \leftrightarrow \frac{1}{2}(1, 1, -1, -1)$$

$$\psi_{1,0}(x) \leftrightarrow \frac{1}{\sqrt{2}}(1, -1, 0, 0)$$

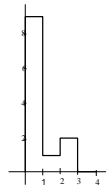
$$\psi_{1,1}(x) \leftrightarrow \frac{1}{\sqrt{2}}(0, 0, 1, -1)$$



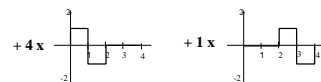
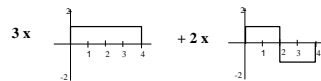
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Wavelet representation

$$f = c_2(0)\phi_{2,0} + d_2(0)\psi_{2,0} + d_1(0)\psi_{1,0} + d_1(1)\psi_{1,1}$$



=



$$c_2 = (3), d_2 = (2), d_1 = (4, 1)$$

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Decomposition: general form

We can write the operations in general form:

- For c_j : First filter c_{j-1} with a **low pass filter**
 $h_\phi = \frac{1}{2}(1, 1)$, then downsample by a factor of 2.
- For d_j : First filter c_{j-1} with a **band pass filter**
 $h_\psi = \frac{1}{2}(1, -1)$, then downsample by a factor of 2.

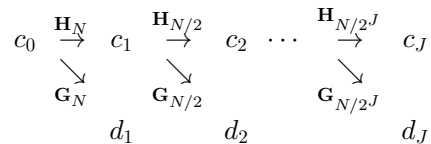
$$c_j = \mathbf{H}c_{j-1} = \downarrow_2 (h_\phi * c_{j-1})$$

$$d_j = \mathbf{G}c_{j-1} = \downarrow_2 (h_\psi * c_{j-1})$$

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Input: signal c_0 of length $N = 2^K$ ($J \leq K$).

Output: signals $c_J, d_J, d_{J-1}, \dots, d_1$.



Mallat's pyramid algorithm for decomposition.

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We can write the operations in general form:

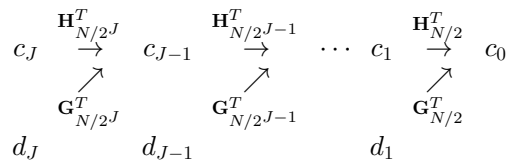
- Upsample c_j by a factor of 2, then filter the result with a **low pass filter**
 $\tilde{h}_\phi = (1, 1)$.
- Upsample d_j by a factor of 2, then filter the result with a **band pass filter**
 $\tilde{h}_\psi = (1, -1)$.
- Add the results.

$$c_{j-1} = \tilde{h}_\phi * (\uparrow_2 c_j) + \tilde{h}_\psi * (\uparrow_2 d_j)$$

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Input: signals $c_J, d_J, d_{J-1}, \dots, d_1$.

Output: signal c_0 of length $N = 2^K$.

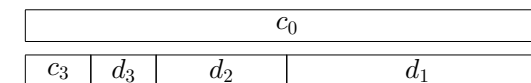


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Signal of size N , with $N = 2^K$. In a J -level pyramid ($J \leq K$), the total number of points is:

$$N \left(\underbrace{\frac{1}{2^1}}_{d_1} + \underbrace{\frac{1}{2^2}}_{d_2} + \dots + \underbrace{\frac{1}{2^J}}_{d_J} + \underbrace{\frac{1}{2^J}}_{c_J} \right) = N$$

The wavelet decomposition is **non-redundant**: a signal of size N is mapped to a signal of size N .



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It is convenient to **normalize** the filters:

Decomposition low pass filter : $h_\phi = \frac{1}{\sqrt{2}}(1, 1)$

Decomposition band pass filter : $h_\psi = \frac{1}{\sqrt{2}}(1, -1)$

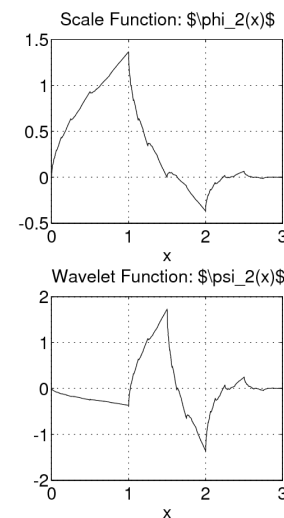
Reconstruction low pass filter : $\tilde{h}_\phi = \frac{1}{\sqrt{2}}(1, 1)$

Reconstruction band pass filter : $\tilde{h}_\psi = \frac{1}{\sqrt{2}}(1, -1)$

Now the decomposition and reconstruction filters are the same:
 $h_\phi = \tilde{h}_\phi$, $h_\psi = \tilde{h}_\psi$.

This is an example of an **orthonormal wavelet basis**.

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4-tap filter:

$$h_\phi(0) = \frac{1 + \sqrt{3}}{4\sqrt{2}}$$

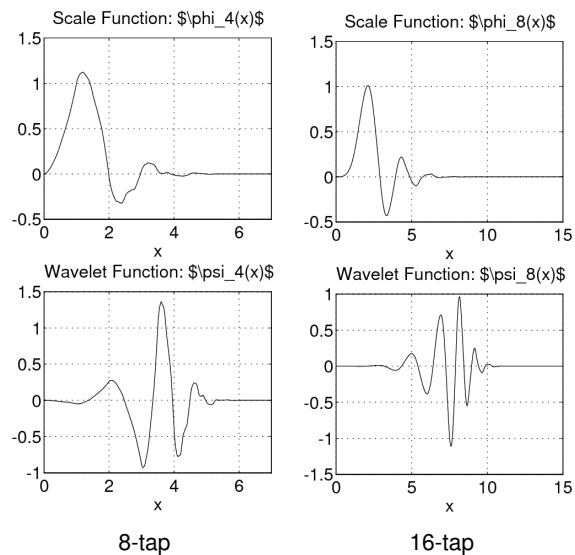
$$h_\phi(1) = \frac{3 + \sqrt{3}}{4\sqrt{2}}$$

$$h_\phi(2) = \frac{3 - \sqrt{3}}{4\sqrt{2}}$$

$$h_\phi(3) = \frac{1 - \sqrt{3}}{4\sqrt{2}}$$

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More Daubechies wavelets

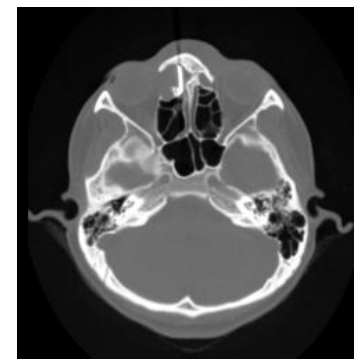


8-tap

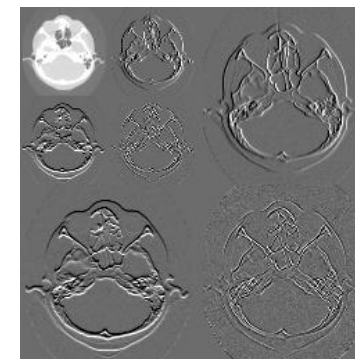
16-tap

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2-D Wavelet analysis



CT image



decomposition (Haar wavelet)

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Local histograms

