

# Image Processing and Analysis – Math Part: The Discrete Wavelet Transform

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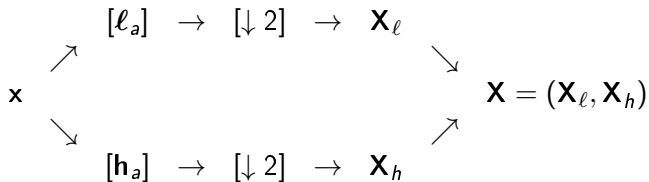
Based upon Chapter 6 of Broughton and Bryan's *Discrete Fourier Analysis and Wavelets* and Maciej Piętko's Lecture Notes from 2010

# Overview

- ▶ Review of the Haar filter bank
- ▶ The general one-stage two-channel filter bank
- ▶ Multistage filter banks
- ▶ The discrete wavelet transform, DWT
- ▶ The 2D DWT

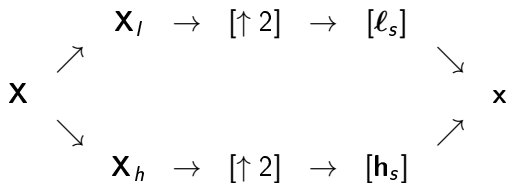
# Review of the Haar Filter Bank

## Analysis Filter Bank



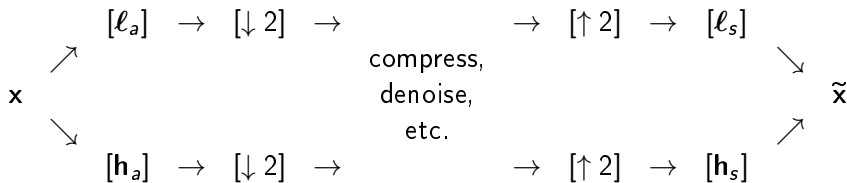
# Review of the Haar Filter Bank

## Synthesis Filter Bank



# Review of the Haar Filter Bank

## Analysis/Synthesis Filter Bank



# The General One-Stage Two-Channel Filter Bank

- ▶ Analysis filters  $\ell_a, \mathbf{h}_a$  need not be 2-point averaging/difference filters
- ▶ Any pair of low-/high-pass finite impulse response filters would do

# The General One-Stage Two-Channel Filter Bank

- ▶ The reconstructed signal  $\mathbf{x}'$  should be a copy of the input signal  $\mathbf{x}$  (possibly delayed)
- ▶ Such a filter bank is called a *perfect reconstruction filter bank*
- ▶ If the above condition is satisfied, synthesis filters are determined from analysis filters

# The General One-Stage Two-Channel Filter Bank

Example: Le Gall 5/3 Filters

- Analysis filters:

$$\ell_a = (\dots, 0, -\frac{1}{8}, \frac{1}{4}, \frac{3}{4}, \frac{1}{4}, -\frac{1}{8}, 0, \dots)$$

$$\mathbf{h}_a = (\dots, 0, -\frac{1}{2}, 1, -\frac{1}{2}, 0, \dots)$$

- Synthesis filters:

$$\ell_s = (\dots, 0, \frac{1}{2}, 1, \frac{1}{2}, 0, \dots)$$

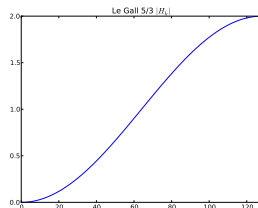
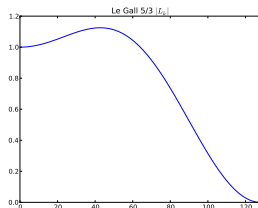
$$\mathbf{h}_s = (\dots, 0, -\frac{1}{8}, -\frac{1}{4}, \frac{3}{4}, -\frac{1}{4}, -\frac{1}{8}, 0, \dots)$$



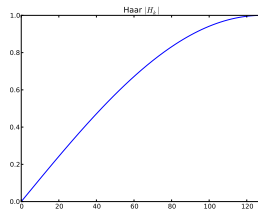
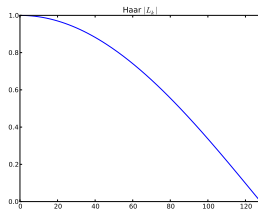
# The General One-Stage Two-Channel Filter Bank

Example: Le Gall 5/3 Filters

Frequency response of the Le Gall 5/3 analysis filters



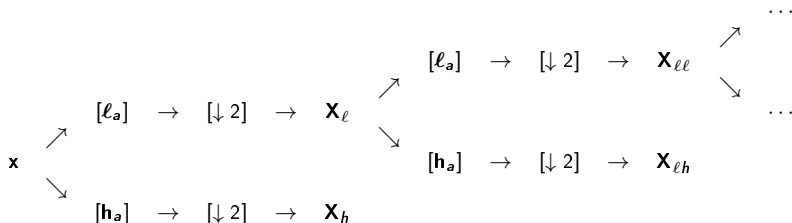
Frequency response of the Haar analysis filters



# Multistage Filter Banks

## General Idea

- ▶ Consider filter bank transform  $\mathbf{x} \rightarrow \mathbf{X} = (\mathbf{X}_\ell \mathbf{X}_h)$
- ▶ Detail coefficients  $\mathbf{X}_h$  are not processed further
- ▶ Approximation coefficients are input to second-stage filter bank
- ▶ Can continue iteratively (“multistage analysis”)



Filter bank tree for multistage analysis

# Multistage Filter Banks

## General Idea

- Multistage analysis filter bank transform

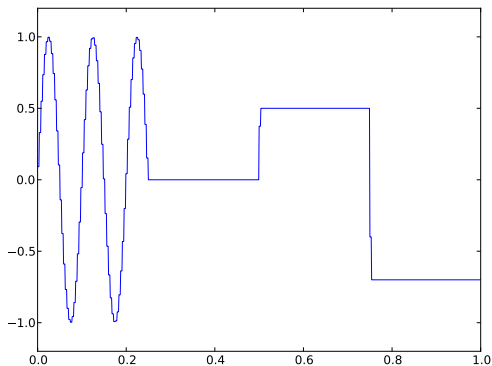
$$\mathbf{x} \rightarrow \begin{pmatrix} \mathbf{x}_\ell \\ \mathbf{x}_h \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{x}_{\ell\ell} \\ \mathbf{x}_{\ell h} \\ \mathbf{x}_h \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{x}_{\ell\ell\ell} \\ \mathbf{x}_{\ell\ell h} \\ \mathbf{x}_{\ell h} \\ \mathbf{x}_h \end{pmatrix} \rightarrow \dots$$

- Multistage synthesis steps are applied in reverse

# Multistage Filter Banks

## 4-Fold Signal Compression with Haar Filter

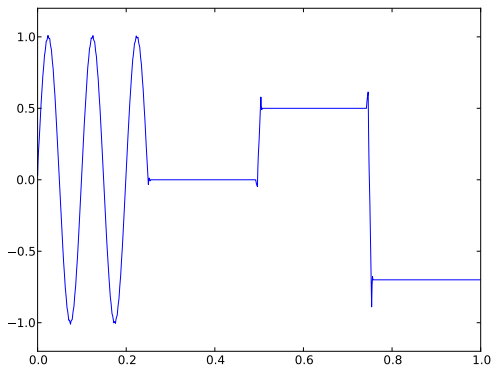
$$\mathbf{x} \rightarrow \begin{pmatrix} \mathbf{X}_{\ell\ell} \\ \mathbf{X}_{\ell h} \\ \mathbf{X}_h \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{X}_{\ell\ell} \\ \mathbf{0} \\ \mathbf{0} \end{pmatrix} \rightarrow \tilde{\mathbf{x}}$$



# Multistage Filter Banks

## 4-Fold Signal Compression with Daubechies 4-Tap Filter

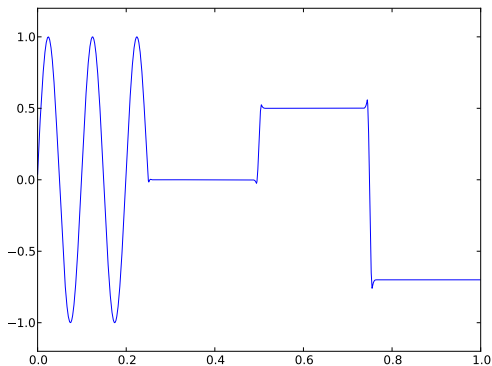
$$\mathbf{x} \rightarrow \begin{pmatrix} \mathbf{x}_{\ell\ell} \\ \mathbf{x}_{\ell h} \\ \mathbf{x}_h \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{x}_{\ell\ell} \\ \mathbf{0} \\ \mathbf{0} \end{pmatrix} \rightarrow \tilde{\mathbf{x}}$$



# Multistage Filter Banks

## 4-Fold Signal Compression with Biorthogonal3.3 (8-Tap Filter)

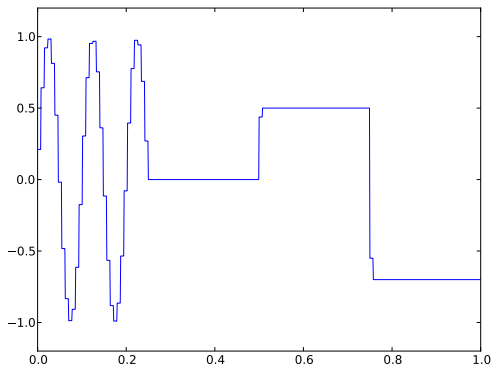
$$\mathbf{x} \rightarrow \begin{pmatrix} \mathbf{X}_{\ell\ell} \\ \mathbf{X}_{\ell h} \\ \mathbf{X}_h \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{X}_{\ell\ell} \\ \mathbf{0} \\ \mathbf{0} \end{pmatrix} \rightarrow \tilde{\mathbf{x}}$$



# Multistage Filter Banks

## 8-Fold Signal Compression with Haar Filter

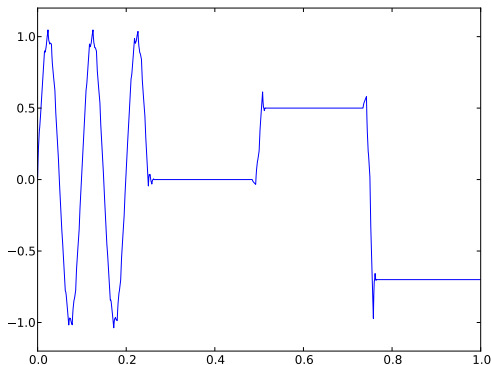
$$\mathbf{x} \rightarrow \begin{pmatrix} \mathbf{X}_{\ell\ell\ell} \\ \mathbf{X}_{\ell\ell h} \\ \mathbf{X}_{\ell h} \\ \mathbf{X}_h \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{X}_{\ell\ell\ell} \\ 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \tilde{\mathbf{x}}$$



# Multistage Filter Banks

## 8-Fold Signal Compression with Daubechies 4-Tap Filter

$$\mathbf{x} \rightarrow \begin{pmatrix} \mathbf{X}_{\ell\ell\ell} \\ \mathbf{X}_{\ell\ell h} \\ \mathbf{X}_{\ell h} \\ \mathbf{X}_h \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{X}_{\ell\ell\ell} \\ 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \tilde{\mathbf{x}}$$

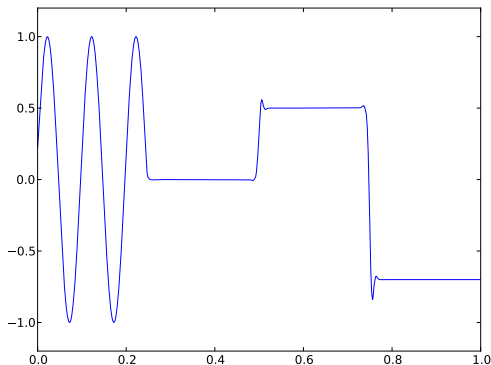




# Multistage Filter Banks

8-Fold Signal Compression with Biorthogonal3.3 (8-Tap Filter)

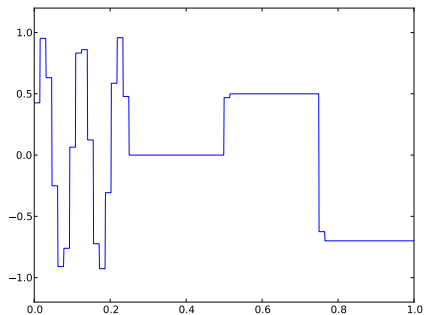
$$\mathbf{x} \rightarrow \begin{pmatrix} \mathbf{X}_{\ell\ell\ell} \\ \mathbf{X}_{\ell\ell h} \\ \mathbf{X}_{\ell h} \\ \mathbf{X}_h \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{X}_{\ell\ell\ell} \\ \mathbf{0} \\ \mathbf{0} \\ \mathbf{0} \end{pmatrix} \rightarrow \tilde{\mathbf{x}}$$



# Multistage Filter Banks

## 16-Fold Signal Compression with Haar Filter

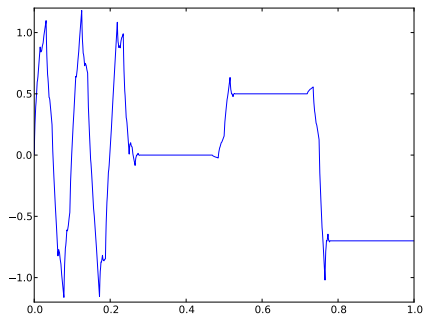
$$\mathbf{x} \rightarrow \begin{pmatrix} \mathbf{X}_{llll} \\ \mathbf{X}_{lllh} \\ \mathbf{X}_{llhh} \\ \mathbf{X}_{lhh} \\ \mathbf{X}_h \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{X}_{llll} \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \tilde{\mathbf{x}}$$



# Multistage Filter Banks

16-Fold Signal Compression with Daubechy 4-Tap Filter

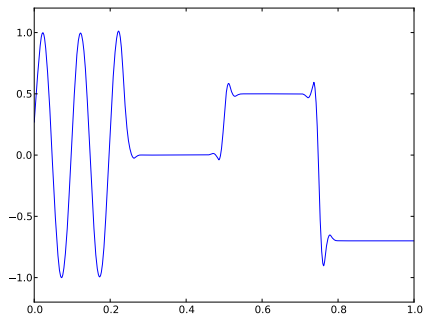
$$\mathbf{x} \rightarrow \begin{pmatrix} \mathbf{X}_{llll} \\ \mathbf{X}_{lllh} \\ \mathbf{X}_{llhh} \\ \mathbf{X}_{lhh} \\ \mathbf{X}_h \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{X}_{llll} \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \tilde{\mathbf{x}}$$



# Multistage Filter Banks

16-Fold Signal Compression with Biorthogonal3.3 (8-Tap Filter)

$$\mathbf{x} \rightarrow \begin{pmatrix} \mathbf{X}_{llll} \\ \mathbf{X}_{lllh} \\ \mathbf{X}_{llh} \\ \mathbf{X}_{lh} \\ \mathbf{X}_h \end{pmatrix} \rightarrow \begin{pmatrix} \mathbf{X}_{llll} \\ \mathbf{0} \\ \mathbf{0} \\ \mathbf{0} \\ \mathbf{0} \end{pmatrix} \rightarrow \tilde{\mathbf{x}}$$



# The Discrete Wavelet Transform

## Basic Idea

- ▶ The above formalism was for signals  $\mathbf{x} \in L^2(\mathbb{Z})$
- ▶ Adopt it for finite length signals  $\mathbf{x} \in \mathbb{C}^N$  by
  - ▶ Extending  $\mathbf{x}$  to  $\tilde{\mathbf{x}} \in L^2(\mathbb{Z})$
  - ▶ Perform the filtering on  $L^2(\mathbb{Z})$  vectors
  - ▶ Truncate the filtered signal to a finite length

# The Discrete Wavelet Transform

## Extension Strategies

- Zero-padding:  $\tilde{x}_k = 0$  for  $k < 0$  and  $k \geq N$

$$\mathbf{x} = (a, b, c) \Rightarrow$$

$$\tilde{\mathbf{x}} = (\dots, 0, 0, 0, 0, 0, 0, a, b, c, 0, 0, 0, 0, 0, 0, \dots)$$

# The Discrete Wavelet Transform

## Extension Strategies

- Periodic extension  $\tilde{x}_k = x_{x \bmod N}$  (like DFT)

$$\mathbf{x} = (a, b, c) \Rightarrow$$

$$\tilde{\mathbf{x}} = (\dots, a, b, c, a, b, c, a, b, c, a, b, c, a, b, c, \dots)$$

# The Discrete Wavelet Transform

## Extension Strategies

- ▶ Half-point symmetric extension: reflect the signal about one of its endpoints, then extend periodically (like DCT), e.g.,

$$\mathbf{x} = (a, b, c) \Rightarrow$$

$$\tilde{\mathbf{x}} = (\dots, a, b, c, c, b, a, a, b, c, c, b, a, a, b, c, \dots)$$



# The Discrete Wavelet Transform

## Definition of the DWT

- ▶ Suppose  $\mathbf{x} \in \mathbb{C}^N$  is extended periodically to  $\tilde{\mathbf{x}}$
- ▶ We apply analysis filters as  $\tilde{\mathbf{x}} * \mathbf{f}$ , where  $\mathbf{f}$  is either  $\ell_a$  or  $\mathbf{h}_a$
- ▶ Key observation: if  $\mathbf{f}$  is a FIR filter with at most  $N$  taps, then  $\tilde{\mathbf{x}} * \mathbf{f}$  is exactly the circular convolution  $\mathbf{x} * \mathbf{f}$  extended periodically
- ▶ Downsampled vectors  $\tilde{\mathbf{X}}_\ell = D(\tilde{\mathbf{x}} * \ell_a)$  and  $\tilde{\mathbf{X}}_h = D(\tilde{\mathbf{x}} * \mathbf{h}_a)$  are periodic with period  $N/2$
- ▶ Truncate  $\tilde{\mathbf{X}}_\ell$  and  $\tilde{\mathbf{X}}_h$  to  $\mathbf{X}_\ell$  and  $\mathbf{X}_h$  by taking components in the range  $0 \leq k \leq N/2 - 1$
- ▶ Concatenate them to produce  $\mathbf{X} = (\mathbf{X}_\ell, \mathbf{X}_h) \in \mathbb{C}^N$
- ▶ The mapping  $\mathbf{x} \rightarrow \mathbf{X}$  is called a *discrete wavelet transform* (DWT)

# The Discrete Wavelet Transform

## DWT Properties

- ▶ The DWT is linear
- ▶ The DWT can be represented as an  $N \times N$  matrix
- ▶ If the filters are FIR filters for a perfect reconstruction filter bank, then the DWT is invertible

# The Discrete Wavelet Transform

Example: the Haar DWT

- ▶  $\mathbf{x} = (a, b, c, d) \in \mathbb{C}^4$
- ▶  $\ell_a = (\frac{1}{2}, \frac{1}{2}, 0, 0)$ ,  $\mathbf{h}_a = (\frac{1}{2}, -\frac{1}{2}, 0, 0)$
- ▶  $\mathbf{x} * \ell_a = \frac{1}{2}(a + d, b + a, c + b, d + c)$  ext. periodically
- ▶  $\mathbf{x} * \mathbf{h}_a = \frac{1}{2}(a - d, b - a, c - b, d - c)$  ext. periodically
- ▶ Downsampling and truncating:

$$\mathbf{X}_\ell = \frac{1}{2}(a + d, c + b)$$

$$\mathbf{X}_h = \frac{1}{2}(a - d, c - b)$$

- ▶ The DWT of  $\mathbf{x}$  is

$$\mathbf{X} = \frac{1}{2}(a + d, c + b, a - d, c - b)$$

# The Discrete Wavelet Transform

Example: the Haar DWT

Matrix view of the DWT:

$$\mathbf{X} = \mathbf{W}_4^a \mathbf{x}$$

where

$$\mathbf{W}_4^a = \frac{1}{2} \begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & -1 \\ 0 & -1 & 1 & 0 \end{pmatrix}$$

# The Discrete Wavelet Transform

Example: the Haar IDWT

- ▶ The IDWT is governed by the synthesis filters
- ▶  $\ell_s = (1, 0, 0, 1)$ ,  $\mathbf{h}_s = (1, 0, 0, -1)$
- ▶ Let  $\mathbf{X} = (A, B, C, D)$
- ▶ Upsampling gives  $U(\mathbf{X}_\ell) = (A, 0, B, 0)$ ,  $U(\mathbf{X}_h) = (C, 0, D, 0)$
- ▶ Convolution with synthesis filters:

$$U(\mathbf{X}_\ell) * \ell_s = (A, B, B, A)$$

$$U(\mathbf{X}_h) * \mathbf{h}_s = (C, -D, D, -C)$$

- ▶ The IDWT of  $\mathbf{X}$  is

$$\mathbf{x} = (A + C, B - D, B + D, A - C)$$

# The Discrete Wavelet Transform

Example: the Haar IDWT

Matrix view of the IDWT:

$$\mathbf{x} = \mathbf{W}_4^s \mathbf{X}$$

where

$$\mathbf{W}_4^s = \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & -1 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & -1 & 0 \end{pmatrix}$$

Check that  $\mathbf{W}_4^a \mathbf{W}_4^s = \mathbf{I}$

# The Discrete Wavelet Transform

Example: the Haar IDWT

Changing the normalisation gives

$$\mathbf{W}_4^a = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 \\ 1 & 0 & 0 & -1 \\ 0 & -1 & 1 & 0 \end{pmatrix}$$

$$\mathbf{W}_4^s = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & -1 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & -1 & 0 \end{pmatrix}$$

i.e., an orthogonal matrix  $\mathbf{W}_4^s = (\mathbf{W}_4^a)^T$  (this is *not* the case for all wavelets)

# The Discrete Wavelet Transform

## Matrix Representation of Multistage DWT

- ▶ Let  $\mathbf{W}_N^a, \mathbf{W}_N^s \in M_{N,N}(\mathbb{C})$  be the analysis and synthesis matrices for the  $N$ -point one-stage DWT
- ▶ Assume  $N = 2^p$
- ▶ Two-stage transform has the matrix

$$\mathcal{W}_2^a = \begin{pmatrix} \mathbf{W}_{N/2}^a & 0 \\ 0 & \mathbf{I}_{N/2} \end{pmatrix} \mathbf{W}_N^a$$

where  $\mathbf{I}_{N/2}$  is the  $\frac{N}{2} \times \frac{N}{2}$  identity matrix

- ▶ The inverse transform

$$\mathcal{W}_2^s = \mathbf{W}_N^s \begin{pmatrix} \mathbf{W}_{N/2}^s & 0 \\ 0 & \mathbf{I}_{N/2} \end{pmatrix}$$



# The Discrete Wavelet Transform

## Matrix Representation of Multistage DWT

- An  $r$ -stage DWT matrix obtained by iteration

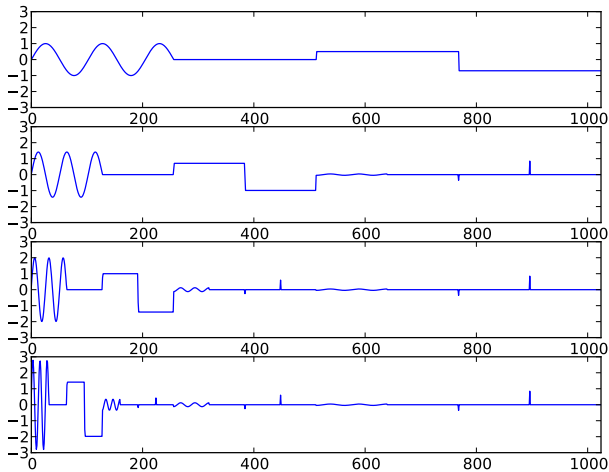
$$\mathcal{W}_r^a = \begin{pmatrix} \mathbf{W}_{N/2^{r-1}}^a & 0 \\ 0 & \mathbf{I}_{N(1-1/2^{r-1})} \end{pmatrix} \cdots \begin{pmatrix} \mathbf{W}_{N/2}^a & 0 \\ 0 & \mathbf{I}_{N/2} \end{pmatrix} \mathbf{W}_N^a$$

- An  $r$ -stage IDWT matrix

$$\mathcal{W}_r^s = \mathbf{W}_N^s \begin{pmatrix} \mathbf{W}_{N/2}^s & 0 \\ 0 & \mathbf{I}_{N/2} \end{pmatrix} \cdots \begin{pmatrix} \mathbf{W}_{N/2^{r-1}}^s & 0 \\ 0 & \mathbf{I}_{N(1-1/2^{r-1})} \end{pmatrix}$$

# The Discrete Wavelet Transform

Example: Multistage Transform



# The 2D Discrete Wavelet Transform

## Matrix Formulation, One-Stage Transform

- In analogy with the 2D DFT and the 2D DCT, the 2D DWT can be computed as the matrix product

$$\hat{\mathbf{A}} = \mathbf{W}_m^a \mathbf{A} (\mathbf{W}_n^a)^T$$

where  $\mathbf{W}_m^a$  and  $\mathbf{W}_n^a$  are the one-dimensional DCT matrices

- The inverse transform is

$$\mathbf{A} = \mathbf{W}_m^s \hat{\mathbf{A}} (\mathbf{W}_n^s)^T$$

# The 2D Discrete Wavelet Transform

## Matrix Formulation, One-Stage Transform

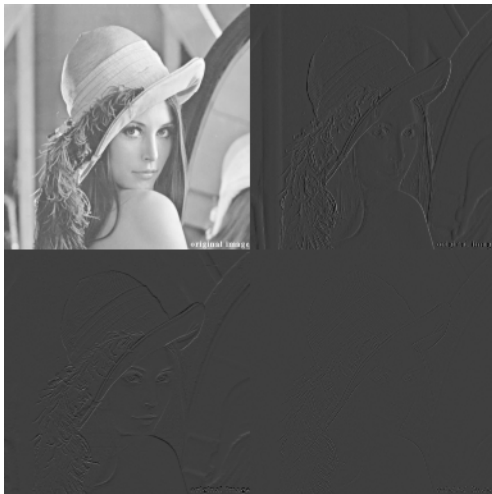
Interpretation of the result

$$\begin{aligned}\hat{\mathbf{A}} &= \mathbf{W}_m^a \mathbf{A} (\mathbf{W}_n^a)^T \\ &= \begin{pmatrix} \text{LL (approximation)} & \text{HL (vertical details)} \\ \text{LH (horizontal details)} & \text{HH (diagonal details)} \end{pmatrix}\end{aligned}$$

(where L represents low-pass, and H high-pass filtering)

# The 2D Discrete Wavelet Transform

## Example, One-Stage Transform



# The 2D Discrete Wavelet Transform

## Matrix Formulation, Multistage Transform

- The multistage 2D DWT can be computed as the matrix product

$$\hat{\mathbf{A}} = \mathcal{W}_m^a \mathbf{A} (\mathcal{W}_n^a)^T$$

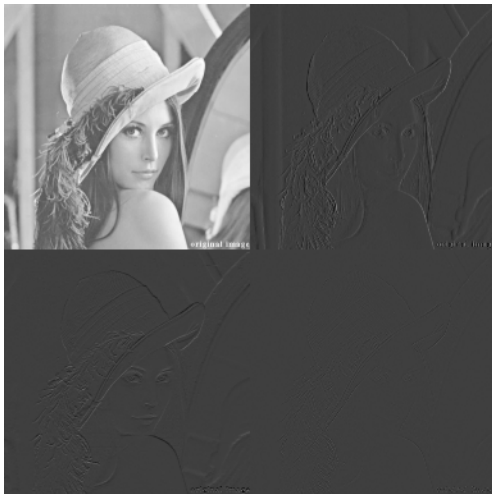
where  $\mathcal{W}_m^a$  and  $\mathcal{W}_n^a$  are the one-dimensional multistage DCT matrices

- The inverse transform is

$$\mathbf{A} = \mathcal{W}_m^s \hat{\mathbf{A}} (\mathcal{W}_n^s)^T$$

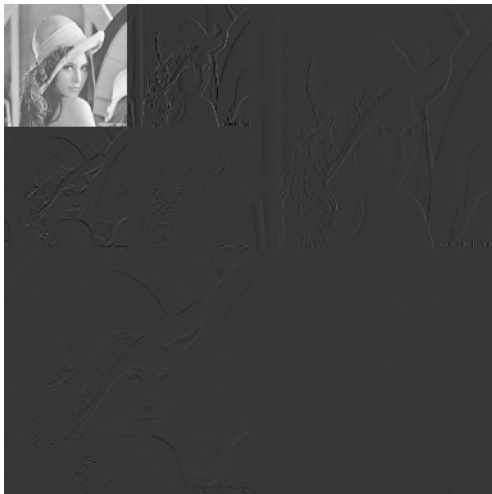
# The 2D Discrete Wavelet Transform

Example, Multi-Stage Transform



# The 2D Discrete Wavelet Transform

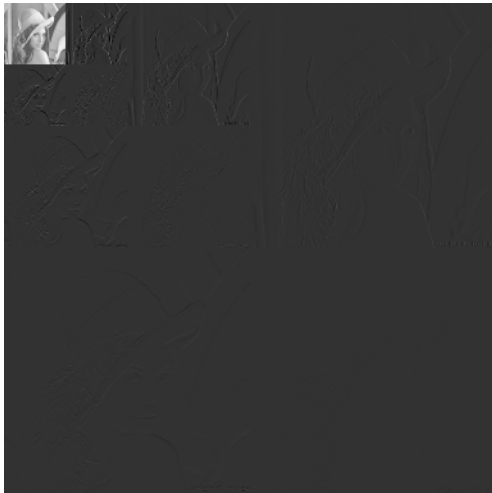
Example, Multi-Stage Transform





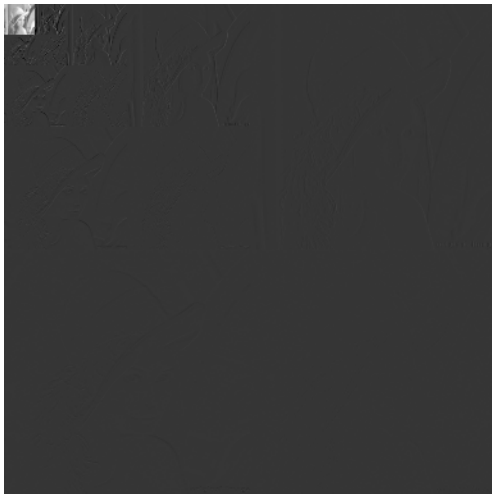
# The 2D Discrete Wavelet Transform

Example, Multi-Stage Transform



# The 2D Discrete Wavelet Transform

Example, Multi-Stage Transform



# The 2D Discrete Wavelet Transform

Example, Multi-Stage Compression, 'haar'

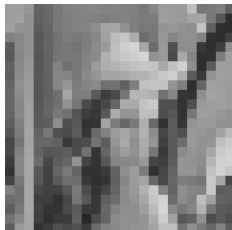
4-fold



16-fold



64-fold



256-fold

# The 2D Discrete Wavelet Transform

Example, Multi-Stage Compression, 'db2'

4-fold



16-fold



64-fold



256-fold

# The 2D Discrete Wavelet Transform

Example, Multi-Stage Compression, 'bior3.3'

4-fold



16-fold



64-fold



256-fold

# The 2D Discrete Wavelet Transform

## Comparison of Transformations

Transform	1D	2D
DFT	$\mathbf{X} = \mathbf{F}_N \mathbf{x}$	$\hat{\mathbf{A}} = \mathbf{F}_N \mathbf{A} \mathbf{F}_M^T$
DCT	$\mathbf{X} = \mathbf{C}_N \mathbf{x}$	$\hat{\mathbf{A}} = \mathbf{C}_N \mathbf{A} \mathbf{C}_M^T$
DWT	$\mathbf{X} = \mathcal{W}_N^a \mathbf{x}$	$\hat{\mathbf{A}} = \mathcal{W}_N^a \mathbf{A} \mathcal{W}_M^{a^T}$

# Overview

- ▶ Review of the Haar filter bank
- ▶ The general one-stage two-channel filter bank
- ▶ Multistage filter banks
- ▶ The discrete wavelet transform, DWT
- ▶ The 2D DWT

# Exercises

## The Discrete Wavelet Transform

Do exercise 6.3 in the text book