# Transform Methods & Signal Processing lecture 11

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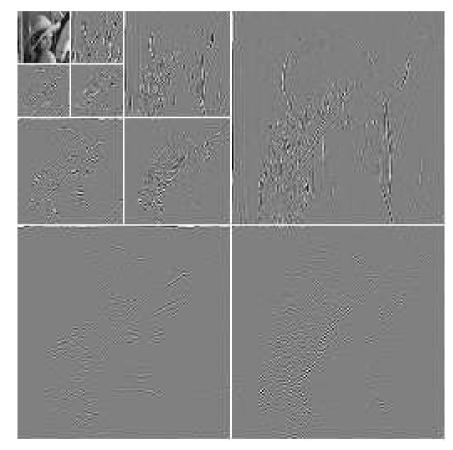
September 8, 2010

## 2D Wavelets

Generalizing Wavelets to 2D is not quite as simple as generalizing the Fourier transform to higher dimensions.

#### Simple extension (for separable wavelet bases)





#### Separable wavelet bases

Take any orthonormal wavelet basis  $\{\psi_{n,j}\}_{n,j\in\mathbb{Z}}$  of  $L^2(\mathbb{R})$ , then a separable wavelet basis for  $L^2(\mathbb{R}^2)$  is

$$\{\psi_{n_1,j_1}\psi_{n_2,j_2}\}_{n_1,n_2,j_1,j_2\in\mathbb{Z}}$$

- but basis above mixes resolutions at different scales  $j_1$  and  $j_2$
- separable MRAs lead to constructions that are products of functions dilated to the same scale
- can construct non-separable bases, but used less often
- build approximation spaces  $V_j^2 = V_j \otimes V_j$  such that these are separable, i.e., basis looks like  $\phi_{n_1,n_2,j}^2(x_1,x_2) = \phi_{n_1,j}(x_1)\phi_{n_2,j}(x_2)$

## 2D scaling functions

#### Scaling functions

$$\phi_{n_1,n_2,j}^2(x_1,x_2) = \phi_{n_1,j}(x_1)\phi_{n_2,j}(x_2) = \frac{1}{2^j}\phi\left(\frac{x_1}{2^j} - n_1\right)\phi\left(\frac{x_2}{2^j} - n_2\right)$$

Approximation  $\hat{f}_j = \sum_{n_1,n_2,j} \left\langle f, \phi^2_{n_1,n_2,j} \right\rangle \phi^2_{n_1,n_2,j}$  where

$$\langle f, \phi_{n_1, n_2, j}^2 \rangle = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x_1, x_2) \phi_{n_1, n_2, j}^2(x_1, x_2) dx_1 dx_2$$

which separates into two integrals if f separates.

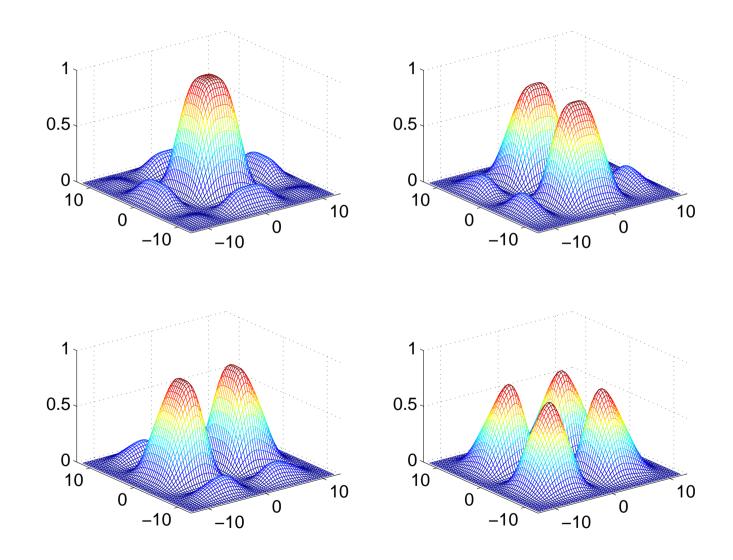
#### 2D Wavelets

#### We get 3 mother wavelets

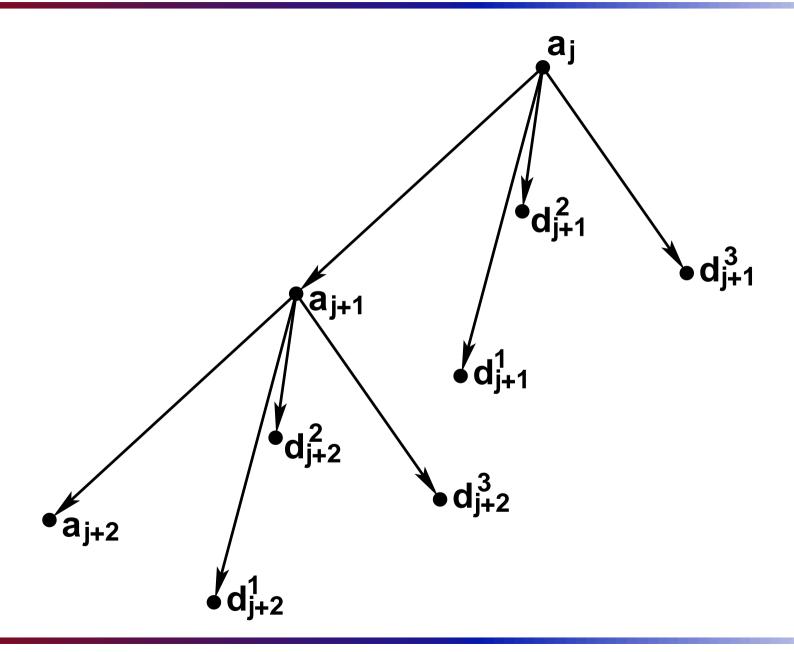
$$\psi^{1}(x_{1}, x_{2}) = \phi(x_{1})\psi(x_{2}) 
\psi^{2}(x_{1}, x_{2}) = \psi(x_{1})\phi(x_{2}) 
\psi^{3}(x_{1}, x_{2}) = \psi(x_{1})\psi(x_{2})$$

from which we derive the wavelets by dilation and 2D translations.

## 2D Wavelet filter spectrum



#### 2D MRA tree



#### 2D Wavelet Filters

$$a_j(n_1,n_2) = \langle f, \phi_{n_1,n_2,j} \rangle$$
 and  $d_j^k(n_1,n_2) = \langle f, \psi_{n_1,n_2,j}^k \rangle$ 

one step of the decomposition takes the form

$$a_{j+1}(n_1, n_2) = [a_j * \bar{h}\bar{h}] (2n_1, 2n_2)$$

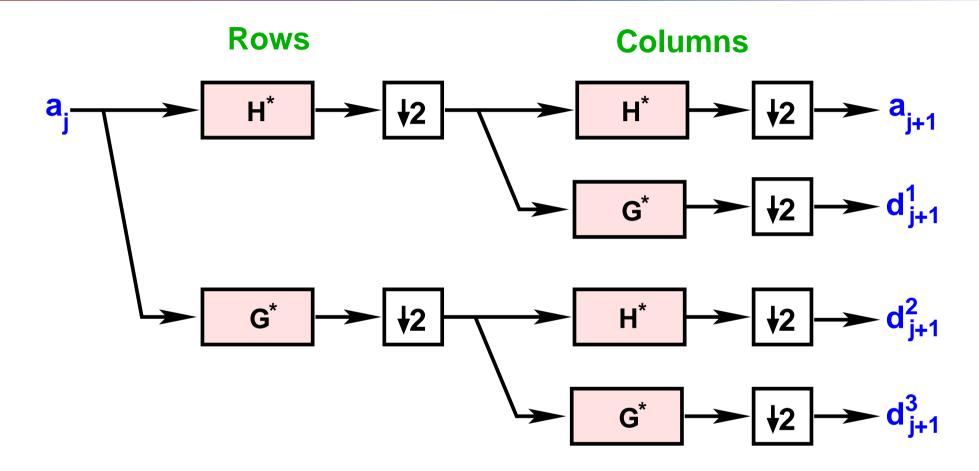
$$d_{j+1}^1(n_1, n_2) = [a_j * \bar{h}\bar{g}] (2n_1, 2n_2)$$

$$d_{j+1}^2(n_1, n_2) = [a_j * \bar{g}\bar{h}] (2n_1, 2n_2)$$

$$d_{j+1}^3(n_1, n_2) = [a_j * \bar{g}\bar{g}] (2n_1, 2n_2)$$

- Notation  $\bar{h}(n) = h(-n)$
- Product hg means  $[hg](n_1,n_2) = h(n_1)g(n_2)$
- 2D convolution can be performed as two 1D convolutions (downsample between doing rows and columns)

## 2D Wavelet Block Diagram



#### 2D Wavelet Reconstruction

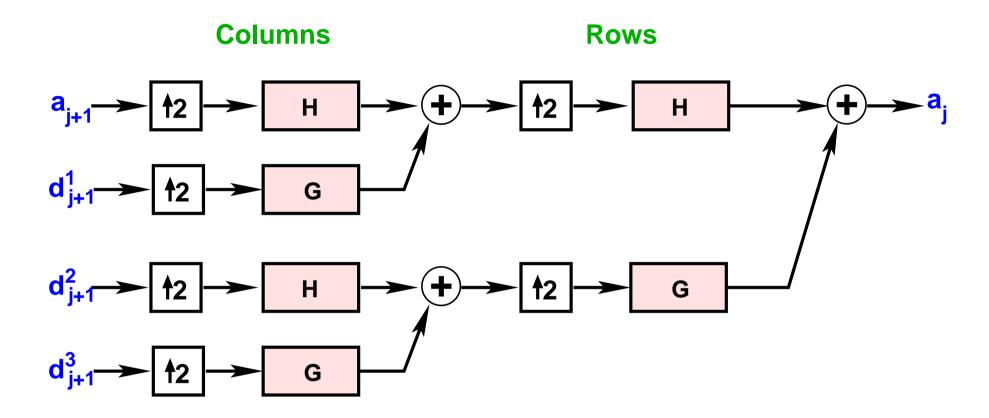
Representation  $\{a_{J}, \{d_{j}^{1}, d_{j}^{2}, d_{j}^{3}\}_{L < j \leq j}\}$ 

■ define upsampled 2D image  $\breve{y}(n_1, n_2)$  made by inserting rows and columns of zeros in between existing rows and columns

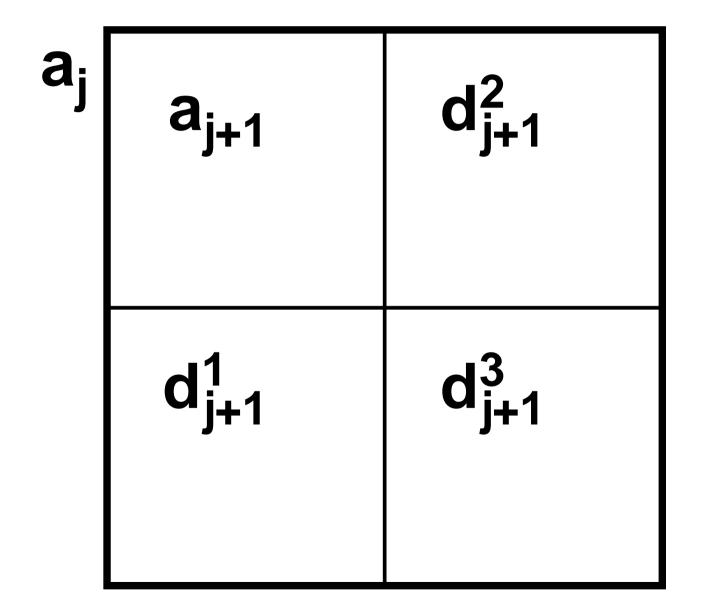
#### Reconstruction

$$a_j = \breve{a}_{j+1} * hh + \breve{d}_{j+1} * hg + \breve{d}_{j+1} * gh + \breve{d}_{j+1} * gg$$

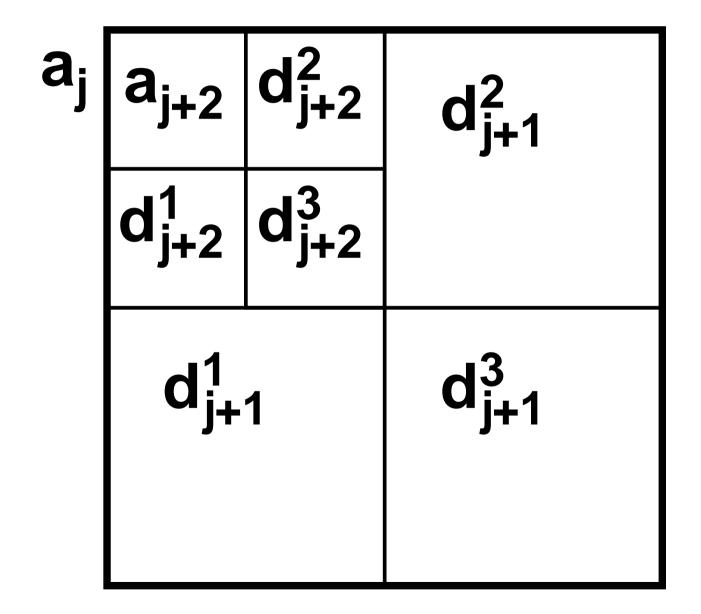
### 2D Wavelet Block Diagram



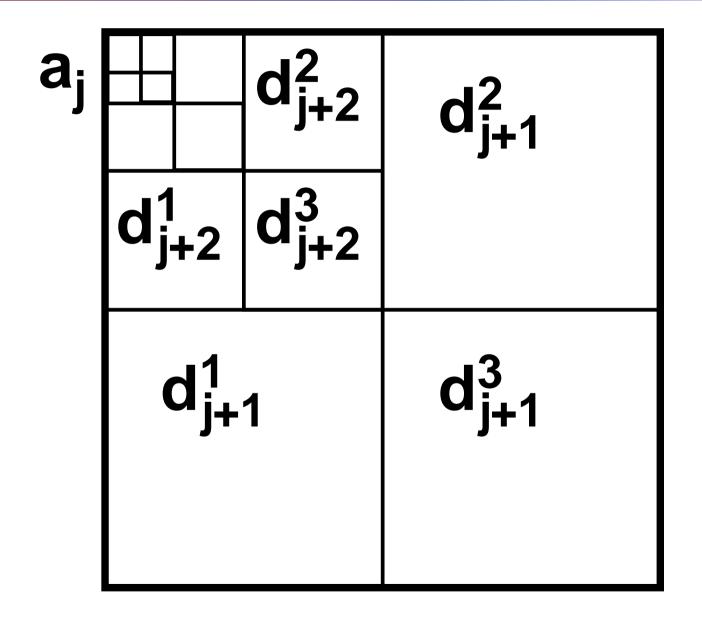
## 2D Layout



#### 2D Layout

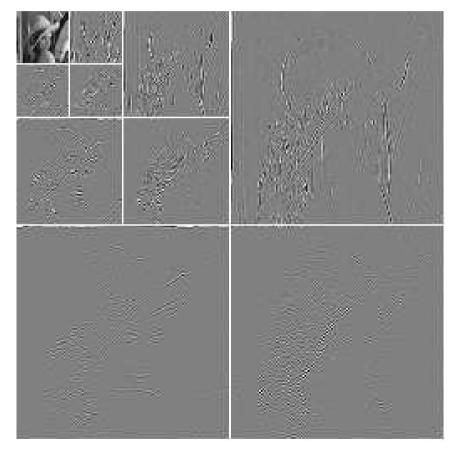


### 2D Layout

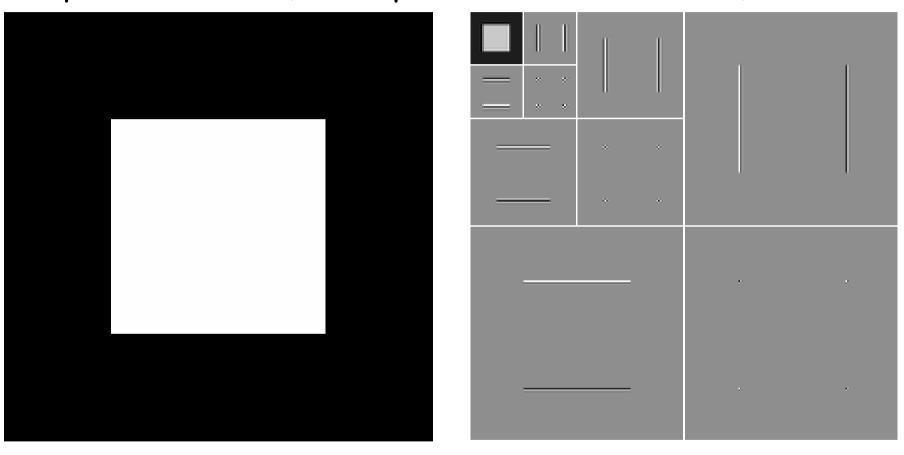


#### Simple extension (for separable wavelet bases)





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## Applications

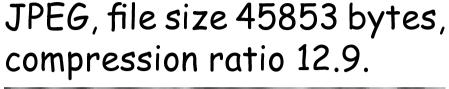
- $\blacksquare$  FBI have  $\sim$  30 million fingerprints
  - actually more like 200 million (repeats etc)
  - 30-50,000 more per day
- 1993 started converting from ink on cards (transmitted by fax) to digital storage
- 500 pixels per inch resolution. 256 grey levels (8 bits)
  - one fingerprint, 700,000 pixels, and 6 MB storage
  - 200 TB for whole database
  - have to transmit cards (3 hours on slow modem)
- compression is needed

Basic idea, quantize in the transform space.

- use Wavelet transform (in 2D)
- steps
  - wavelet transform
  - quantize coefficients
  - entropy encoding
- called WSQ (Wavelet Scalar Quantization)



Original, file size 589,824 bytes.







http://www.c3.lanl.gov/~brislawn/FBI/FBI.html

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## Compression

#### Compression

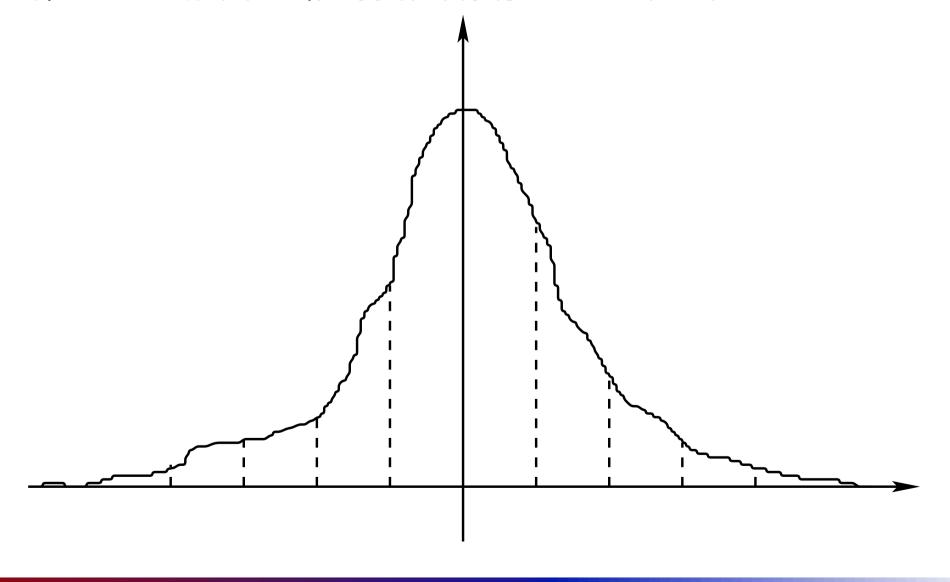
- Simplest version just perform algorithm above
- more general: quantize wavelet coefficients by a fixed step

$$\hat{d}(j,k) = Q \operatorname{sign}(d(j,k)) \left[ \frac{|d(j,k)|}{Q} \right]$$

- more general: use a quantization table
- lacksquare Inverse Wavelet Transform of  $\{\hat{d}(j,k)\}_{j,k}$

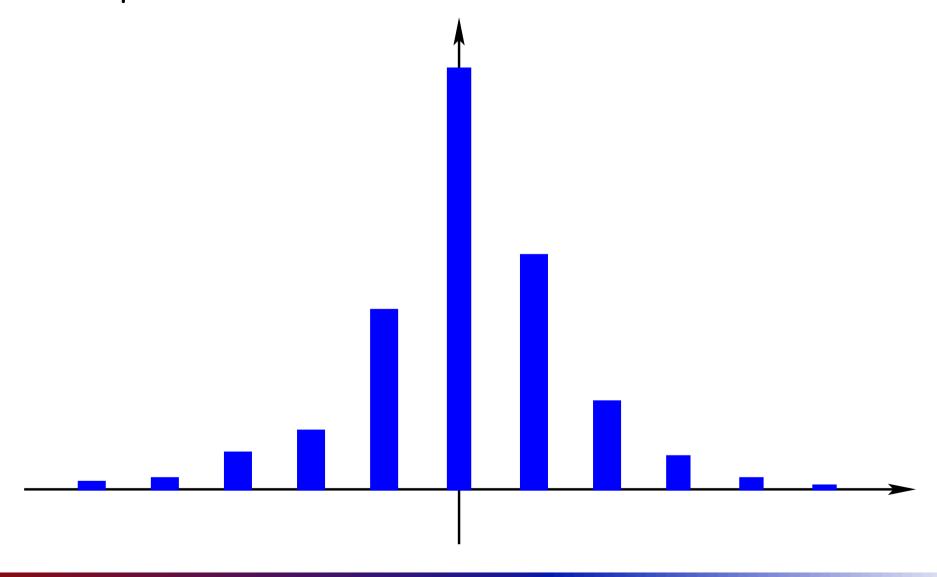
#### Quantization

We start with some continuous distribution



#### Quantization

Then quantize the distribution into a number of levels



#### JPEG 2000

JPEG 2000 uses wavelets rather than the DCT