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Indian Institute of Technology Hyderabad

Wavelets and Their Applications

EE6310: Image and Video Processing

April 13, 2023

The Why, What, How of Wavelets



- ▶ Why: Time-frequency localization
- ▶ What: Wavelets are a tool that cuts up data or functions into different frequency components and studies each component with a resolution matched to its scale
- ▶ How: Digital filters

Orthogonal Wavelet Filters

- ▶ **Finite-length, orthogonal, discrete wavelets**

- ▶ Examples:

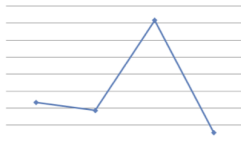
- ▶ Haar:
$$h_0(i) = \begin{cases} \frac{1}{\sqrt{2}}; & i = 0 \\ \frac{1}{\sqrt{2}}; & i = 1 \\ 0; & \text{otherwise} \end{cases}$$

- ▶ Daubechies 4-tap filter:
$$h_0(i) = \frac{1}{4\sqrt{2}} \begin{cases} 1 + \sqrt{3}; & i = 0 \\ 3 + \sqrt{3}; & i = 1 \\ 3 - \sqrt{3}; & i = 2 \\ 1 - \sqrt{3}; & i = 3 \\ 0; & \text{otherwise} \end{cases}$$

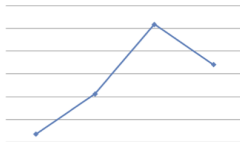
Daubechies 4-tap Wavelet



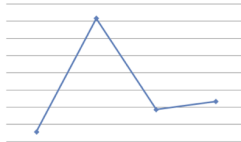
(a) Analysis LP



(b) Analysis HP



(c) Synthesis LP



(d) Synthesis HP

2-D Wavelet Decomposition

- ▶ 2-D analysis filters **decompose** images into high and low frequency bands
- ▶ Information in each band can be analyzed **separately**



Figure: 2-D wavelet decomposition

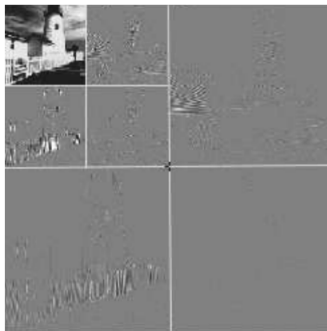
Discrete Wavelet Transform

- ▶ Filter outputs are **wavelet coefficients**
- ▶ Image can be **perfectly** reconstructed from these coefficients
- ▶ Subsampling **heavier** at **lower** frequencies
- ▶ Multiple bands (> 2) created by **iterated filtering** on low-frequency subbands

Discrete Wavelet Transform - Example



(a) Image



(b) 2-level DWT

Image Processing Applications

Myriad of applications. A few popular ones...

- ▶ Image restoration
- ▶ Image compression
- ▶ Image quality assessment
- ▶ Sparse representation
- ▶ Object detection

Wavelet Shrinkage

- ▶ Given $J = I + N$
- ▶ The process of **denoising** is accomplished simply by **thresholding the DWT coefficients** of J
- ▶ A **threshold is applied** to all non-LL subbands

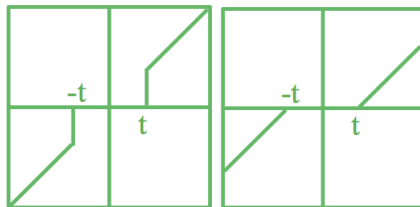
Wavelet Shrinkage

- **Hard Thresholding:** signal zeroed if below threshold

$$J^t(i,j) = \begin{cases} J(i,j); & |J(i,j)| > \tau \\ 0; & |J(i,j)| \leq \tau \end{cases}$$

- **Soft Thresholding:** in addition to threshold, threshold is subtracted if above threshold

$$J^t(i,j) = \text{sgn}[J(i,j)]\max\{0, |J(i,j)| - \tau\}$$



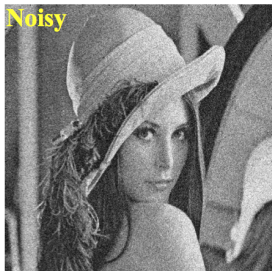
(a) Hard thresholding

(b) Soft thresholding

Wavelet Shrinkage

- ▶ **Soft thresholding** is much more effective. It “shrinks” the wavelet coefficients
- ▶ Works since additive noise **increases coefficient energy**. Soft thresholding **reduces the energy**
- ▶ How to **find threshold**? Can be chosen to **optimize** many criteria such as MSE, Bayesian risk etc.
- ▶ **MMSE Threshold (SureShrink)**: $\tau = \sigma \sqrt{2 \log NM}$ where σ^2 is the estimated noise variance, NM equals the number of pixels in the image
- ▶ **Bayesian threshold (BayesShrink)**: $\tau = \sigma_{\text{signal}} / \sigma_{\text{noise}}$ is estimated for each subband, hence **adaptive**. Large SNR implies little thresholding and vice versa

Wavelet Shrinkage



Comments on Wavelet Shrinkage

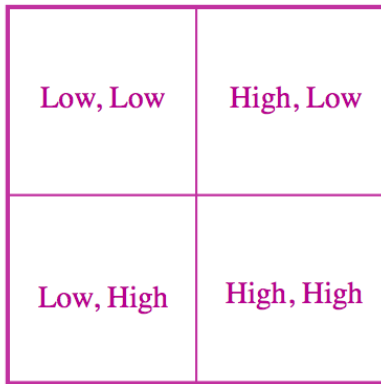
- ▶ Since DWTs can be computed **very fast**, so can **soft-threshold based denoising**
- ▶ The results are **optimal in wavelet space** and the appearance is **reasonably good**

Wavelet-based Image Coding

- ▶ **Idea:** Compress images in the wavelet domain
- ▶ Image blocks not used: **no blocking artifacts!**
- ▶ Forms the basis of **JPEG2000 standard**
- ▶ Requires **perfect reconstruction** filter banks

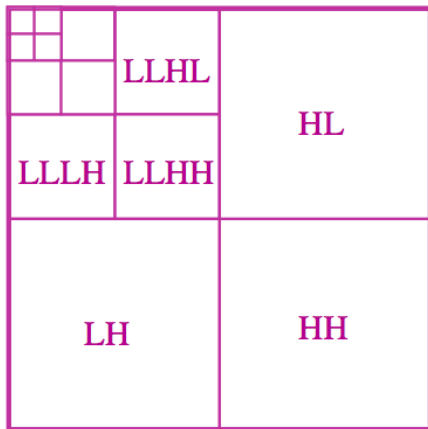
Wavelet Decomposition

- Recall that wavelet filters **separate or decompose** the image into high and low frequency **subbands**

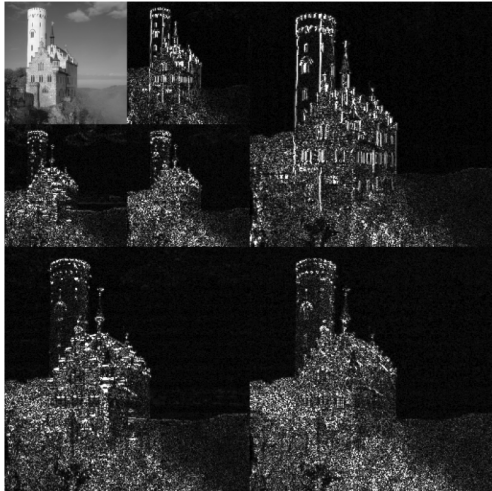


Wavelet Decomposition

- ▶ Typically frequency division is done iteratively on the **lowest-frequency bands**

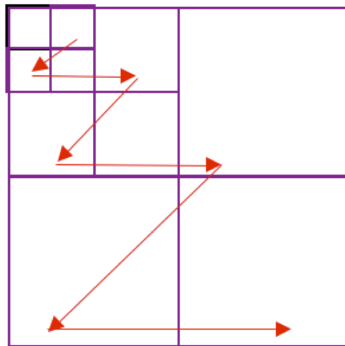


DWT of an Image

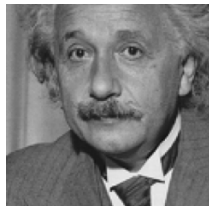


Zero-tree Encoding Concept

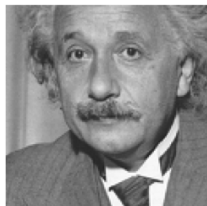
- ▶ Many **complex** wavelet-based compression techniques
- ▶ Popular technique is based on **zero-trees**: similar to JPEG, **scan from lower to higher frequencies** and **run-length code** the zero coefficients



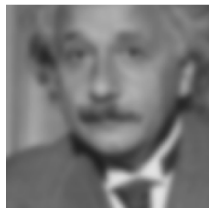
Wavelet-based Image Quality Assessment



(a) Reference



(b) Mean



(c) Blurred



(d) JPEG

Figure: All images have same MSE (309)!

Wavelet-based Image Quality Assessment

Outline:

- ▶ Statistical modeling of wavelet coefficients - the Gaussian Scale Mixture model
- ▶ $X = xU$ where $z \geq 0$ is a scalar random variable, $U \sim \mathcal{N}(0, Q)$ is Gaussian random vector
- ▶ Model the IQA process as a communication channel:
$$Y = g * X + N$$
- ▶ Image quality is measured as the mutual information between X and Y , $I(X; Y)$

Compressive sensing

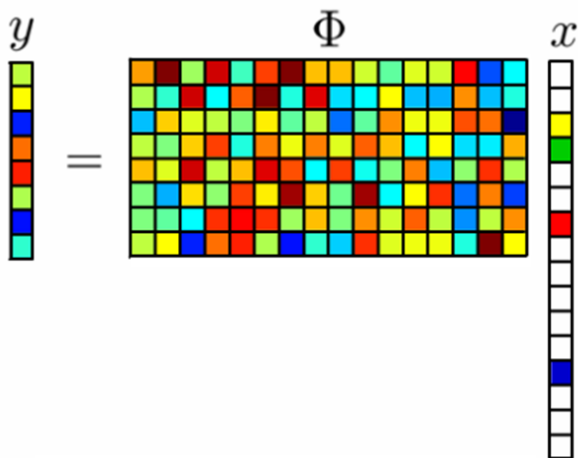


Image courtesy: <http://nuit-blanche.blogspot.fr/>

Compressive sensing

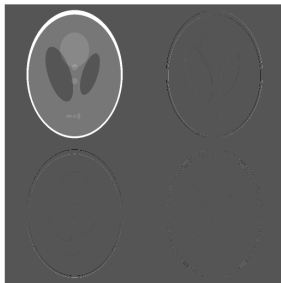
The diagram illustrates the compressive sensing equation $y = \Phi \Psi \alpha$. On the left, a vertical vector y is shown with 8 colored squares (light green, yellow, orange, red, light green, blue, cyan, and light blue). This is followed by an equals sign. To the right of the equals sign is a large grid representing the product of two matrices, Φ and Ψ . Matrix Φ is a 10x10 grid of colored squares. Matrix Ψ is a 10x10 grid of colored squares. To the right of matrix Ψ is a vertical vector α with 10 squares, mostly white, with a few colored squares (yellow, green, red, blue) indicating non-zero coefficients.

Image courtesy: <http://nuit-blanche.blogspot.fr/>

Sparse representation using wavelets



(a) Image



(b) Subband decomposition