Eva Hostalkova

WAVELET TRANSFORM

Eva Hostalkova

Dept of Computing and Control Engineering

Institute of Chemical Technology, Prague

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- Short-Time Fourier Transform (STFT)

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- Continuous Wavelet Transform (CWT)
- Wavelet Functions Properties

4 Discrete Wavelet Transform

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5 Wavelet Transform Applications

- Discontinuity Detection in the ECG Signal
- Image Compression
- Image Segmentation
- Noise Reduction by Wavelet Shrinkage

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Wavelet Transform (WT) History

- 19th cent. Jean B. Fourier: frequency analysis
- 1909 Alfred Haar: Haar function (not yet WT)
- since 1990s WT related research and applications

Vavelet Functions

- Wavelet = a small wave, i.e. an oscillatory function zero outside a bounded interval (having compact support)
- Some real, some complex
- Designed as much smooth and symmetric as possible
- Seldom analytic expression, mostly parametric (e.g. Morlet, Haar or Daubechies fcns)

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WT Analysis

• For non-stationary signals (with time-varying frequency content)

NT Applications in Signal Processing

- Noise reduction
- Detection of trends and discontinuities in higher derivatives
- Compression (JPEG2000, FBI fingerprints database)
- Image edge detection
- Watermarking
- Features extraction for image segmentation

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- Decomposes signals (real or complex) into the orthogonal basis of complex exponentials e^{-j2πft} of frequencies f
- FT for the angular frequency $\omega = 2\pi f$ and time *t*:

$$X(\omega) = \int_{-\infty}^{+\infty} x(t) e^{-j\omega t} dt \qquad (1)$$

where $X(\omega)$ is the Fourier transform of the function x(t)

• Converges for a piece-wise smooth x(t) or finite energy:

$$\int_{-\infty}^{\infty} |x(t)|^2 dt < \infty$$
 (2)

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Inverse F

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} X(\omega) e^{j\omega t} d\omega \qquad ($$

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Inverse FT

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FT Analysis of Non-Stationary Signals

 Discrete FT for the discrete freq. index k=0,1,...,N-1 and discrete time n (for the unit sampling frequency)

$$X(k) = \sum_{n=0}^{N-1} x(n) e^{-j2\pi kn/N}$$
(4)

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Discrete Short-Time Fourier Transform (STFT)

- Non-stationary signal analysis (localizes frequency components within time intervals)
- STFT of a discrete signal x(n) for k = 0, 1, ..., N-1:

$$X(u,k) = \sum_{n=0}^{N-1} x(n) w(n-u) e^{-j2\pi k n/N}$$
 (5)

where u denotes the window position

Jncertainty (Heisenberg) Principle

$$\Delta T \,\Delta f = 1 \tag{6}$$

- $\Delta f = F_s/N$... frequency resolution (distance between adjacent spectral samples)
- $\Delta T = N T_s \dots$ time resolution (window length)
- N ... number of samples per window
- $F_s = 1/T_s \dots$ sampling frequency

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STFT - Uncertainty Tradeoff

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Multi-Resolution Analysis (MRA)

- The wavelet transform deals with the uncertainty principle by MRA
- Analyzing signals at different frequency bands with different resolution:
 - Higher frequencies: good ΔT , poor Δf
 - Lower frequencies: good Δf , poor ΔT
- Convenient for most real-world signals composed of:
 - Long-lasting lower frequencies (approximations)
 - Short-lasting higher frequencies (details main information)

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Continuous Wavelet Transform (CWT)

- FT: correlation of a signal of finite energy with complex exponentials $e^{i2\pi ft}$ of different frequencies f
- WT: correlation of a signal of finite energy with delated and shifted versions ψ_{u,s} of the mother wavelet ψ:

$$\psi_{u,s}(t) = \frac{1}{\sqrt{s}} \psi\left(\frac{t-u}{s}\right) \tag{7}$$

.

- where u is the shift of ψ along the signal \rightarrow time
- and s is scale (dilation of $\psi \rightarrow$ the inverse of frequency)
- $1/\sqrt{s}$ ensures energy normalization
- CWT of the signal x(t): $W_x^{\psi}(s, u) = \langle x, \psi_{u,s} \rangle = \frac{1}{\sqrt{s}} \int_{-\infty}^{+\infty} x(t) \psi^*\left(\frac{t-u}{s}\right) dt$ (8)
 - where * denotes the complex conjugate pair
 and () stands for the input product
 - and $\langle \rangle$ stands for the inner produc

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<u>CWT Computation Procedure</u> (s=5)

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CWT Properties

• Linearity: implied by the properties of the inner product

• Shift invariance:

- A shift in the signal along the time axis
 → the equivalent shift of the wavelet coefficients without any changes
- Does not apply to the discrete WT (great drawback)



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Figure: Comparison of DWT and nearly invariant complex WT (CWT is even better - completely shift invariant)

Wavelet Transform CWT

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Orthogonality and orthonormality

- Not necessary for wavelet analysis (biorthogonality for more freedom in wavelet filters construction)
- A set of shifted and dilated versions of the mother wavelet $\psi_{u,s}$, is orthonormal on the interval [a, b] when:

$$\langle \psi_{u,s}, \psi_{v,r} \rangle = \int_{a}^{b} \psi_{u,s}(t) \psi_{v,r}^{*}(t) dt = \begin{cases} 1 & \text{for } u = v, s = r \\ 0 & \text{otherwise} \end{cases}$$
(9)

• A set of $\psi_{u,s}$, is orthogonal on the interval [a, b] when:

$$\langle \psi_{u,s}, \psi_{v,r} \rangle = \int_{a}^{b} \psi_{u,s}(t) \, \psi_{v,r}^{*}(t) \, dt = \begin{cases} c & \text{for } u = v, s = r \\ 0 & \text{otherwise} \end{cases}$$
(10)
where c is a constant, * denotes a complex conjugate

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Wavelet Functions Properties I

• Admissibility condition

$$\int_{-\infty}^{\infty} \frac{|\Psi(\omega)|^2}{|\omega|} \, d\omega < +\infty \tag{11}$$

- where $\Psi(\omega)$ is the FT of $\psi(t)$
- Required for no loss of information during the analysis nor the synthesis
- The basis do not have to be orthogonal

Consequences of Admissibility Condition

• Oscillatory function (zero mean):

$$\int_{-\infty}^{\infty} \psi(t) \, dt = 0 \tag{12}$$

- Band-pass spectrum ($\Psi(\omega)$ vanishes at $\omega = 0$):
 - $|\Psi(\omega)|^2|_{\,\omega=0} = 0 \tag{13}$

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Wavelet Functions Properties II

• Time dilation of ψ causes shift and compression of the magnitude frequency spectrum $|\Psi|$





NEWLAND WAVELET:

$$\psi(t) = \frac{1}{j \frac{\pi}{2} t} \left(e^{j \pi t} - e^{j \frac{\pi}{2} t} \right)$$

WAVELET SCALING & DILATION:

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

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Desired Wavelet Functions Properties

- Compact support (non-zero only on a restricted interval) not necessary
- Regularity:
 - Smoothness of ψ and vanishing of $|\Psi(\omega)|$ for large ω (small scales)
 - Vanishing moments concept (see literature)
- Symmetry (liner phase)

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Discrete Wavelet Transform (DWT)

- CWT provides us with redundant signal representation
- DWT derive by critically sampling the CWT (great reduction in the number of dilations and shifts)

$$\psi_{u,s}(t) = \frac{1}{\sqrt{s}} \psi\left(\frac{t-u}{s}\right) \Rightarrow \psi_{j,k}(t) = \frac{1}{\sqrt{s_0^j}} \psi\left(\frac{t-k \tau_0 s_0^j}{s_0^j}\right)$$
(14)

Sampling on the Dyadic Grid

- $s_0 = 2 \Rightarrow$ the scale $s = 2^j$
- $\tau_0 = 1 \Rightarrow$ the time translation $u = k 2^j$
- *t* denotes time, *j*, *k* are integers

$$\psi_{j,k}(t) = \frac{1}{\sqrt{2^j}} \psi\left(\frac{t - k2^j}{2^j}\right)$$

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(15)

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The spectrum of the wavelet function corresponds to a high-pass filter (or a band-pass filter for higher levels)
The spectrum of the scaling function corresponds to a low pass filter

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CWT versus DWT

- Correlation
- Dilating the wavelets

Convolution

• Keeping filters the same and down-sampling the previous level output by 2

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Convolution and Downsampling by 2

• The taps of the high-pass filter h_d and the low-pass l_d filter are derived from the wavelet and the scaling function, resp., of a chosen family (e.g. Daubechies, symlets, etc.)

Convolution and Downsampling by 2

• Approximation coefficients of the first level

$$A_{1}[n] = \sum_{k=-\infty}^{\infty} l_{d}[k] \times [2n-k]$$
(16)

• Detail coefficients of the first level $D_1[n] = \sum_{k=-\infty}^{\infty} h_d[k] \times [2n-k]$ (17)

• The j-th level
$$A_j[n] = \sum_{k=-\infty}^{\infty} l_d[k] A_{j-1}[2n-k]$$
 (18)
 $D_j[n] = \sum_{k=-\infty}^{\infty} h_d[k] A_{j-1}[2n-k]$ (19)

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Subband Coding & the Frequency Spectrum



• Dilated by $2 \Rightarrow$ the spectrum is compressed and shifted

 Finite number of dilations ⇒ advantageous to use a low-pass filter - derived from the scaling function φ(t) (the counter part of the wavelet filter at each level creating a filter bank)

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Wavelet Transform	DWT Matrix										
	• The Haar filters										
ntroduction	$l_d = 1/\sqrt{2} \cdot [1, 1]$										
Fourier Transform	α / [/]	()									
Continuous Fourier Transform STFT	$h_d = 1/\sqrt{2} \cdot [1, -1]$	(21)									
Vavelet Transform MRA	• DWT matrix for the Haar filters										
CWT Wavelets Properties	(110000)										
DWT	-1 1 0 0 0 0										
DWT and CWT Subband Coding	0 0 1 1 0 0										
Matrix Interpretation	$1 0 0 -1 1 \dots 0 0$	(22)									
NT Applications	$\mathbf{v}\mathbf{v} = \frac{1}{\sqrt{2}} \cdot \mathbf{v}$	(22)									
Discontinuity Detection											
Image Compression											
Image Segmentation											
Noise Reduction	$\begin{pmatrix} 0 & 0 & 0 & 0 & \dots & -1 & 1 \end{pmatrix}$										

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- W includes both convolution and down-sampling by 2
- W is orthonormal \Rightarrow W \cdot W^T = I

/avelet Transform	DWT Matrix									
Eva Hostalkova	• The Haar filters									
ıction	$I_d =$	$1/\sqrt{2} \cdot [1, 1]$	(20)							
Transform	_	, , , ,	()							
ious Fourier Transform	$h_d = 1$	$1/\sqrt{2} \cdot [1, -1]$	(21)							
et Transform	• DWT matrix for the Haar filters									
ts Properties		. 0 0 0	0)							
	-1 1	. 00 0	0							
nd CWT	0.0) 1 1 0	0							
d Coding	1 0 0		õ							
Provention -	$\mathbf{W} = \frac{1}{\sqrt{2}} \cdot \begin{bmatrix} 0 & 0 \end{bmatrix}$	$0 -1 1 \dots 0$	0 (22)							
oplications	$\sqrt{2}$									
ompression			:							
Segmentation	0 0) 0 0 1	1							
leduction) 0 0 1	1							
nces		0 01	1 /							
	• Mingludge hath convolu	tion and down come	alian hu O							

DWT Matrix Interpretation

- W includes both convolution and down-sampling by 2 (the filters are shifted by 2 samples)
- W is orthonormal ⇒ W · W^T = I where I stands for the identity matrix

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Wavele DWT DWT Subban Matrix WT A Discon Image Image Noise I Refere

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$\frac{1}{\sqrt{2}}$	$\begin{pmatrix}1\\-1\\0\\0\end{pmatrix}$	1 1 0 0	0 0 1 -1	0 0 1 1	· · · · · · · ·	0 0 0 0	0 0 0 0	$ \begin{pmatrix} x(0) \\ x(1) \\ x(2) \\ x(3) \end{pmatrix} $	_	$\left(\begin{array}{c} A_{1}(0) \\ D_{1}(0) \\ A_{1}(1) \\ D_{1}(1) \end{array}\right)$	
	: 0 0	: 0 0	: 0 0	: 0 0	•••• ••••	: 1 -1	: 1 1)	$\begin{pmatrix} \vdots \\ x(N-1) \\ x(N) \end{pmatrix}$		$ \begin{vmatrix} \vdots \\ A_1(\frac{N}{2}-1) \\ D_1(\frac{N}{2}-1) \end{vmatrix} $	

WT Decomposition and Reconstruction

Signal decomposition

$$\mathbf{N} \cdot \mathbf{x} = \mathbf{w} \tag{23}$$

- Signal reconstruction
- $\mathbf{x} = \mathbf{W}^{-1} \cdot \mathbf{w} \tag{24}$
- $\bullet\,$ Signal reconstruction for the orthonormal DWT

$$\mathbf{x} = \mathbf{W}^{\mathsf{T}} \cdot \mathbf{w} \tag{25}$$

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$\frac{1}{\sqrt{2}}$	$\begin{pmatrix}1\\-1\\0\\0\end{pmatrix}$	1 1 0 0	0 0 1 -1	0 0 1 1	· · · · · · · · · · ·	0 0 0 0	0 0 0 0	$ \begin{pmatrix} x(0) \\ x(1) \\ x(2) \\ x(3) \end{pmatrix} $	_	$\left(egin{array}{c} A_1(0) \ D_1(0) \ A_1(1) \ D_1(1) \end{array} ight)$	
	: 0 0	: 0 0	: 0 0	: 0 0	•••• ••••	: 1 -1	: 1 1 /	$\left(\begin{array}{c} \vdots \\ x(N-1) \\ x(N) \end{array}\right)$		$ert rac{ert}{N} A_1(rac{N}{2}\!-\!1) \ D_1(rac{N}{2}\!-\!1) ight)$	

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- Discrete and Continuous Wavelet Transform
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- Discontinuity Detection in the ECG Signal
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- Image Segmentation
- Noise Reduction by Wavelet Shrinkage

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 The Parseval theorem: the energy ε_x conveyed in the signal x equals the energy of the coefficients w obtained through an orthonormal transform

$$\varepsilon_x = \sum_{n=0}^{N-1} |x_n|^2 = \sum_{k=0}^{N-1} |w_k|^2$$
 (26)



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(a) ORIGINAL IMAGE



(c) DWT SEGMENTATION



- Image preprocessing
- Watershed transform 2
- DWT features extraction 3
- Features classification using a neural network

(b) WATERSHED TRANSFORM



(d) RECOGNIZED REGIONS SUPERIMPOSED ON ORIGINAL IMAGE



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Thresholding procedure type? Threshold level?



Wavelet function? Number of levels?

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