

# Wavelets

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Analisi tempo-frequenza e multiscala – 01RMQNG

# Uncertainty principle for finite-length sequences

Theorem (Vetterli, Sec. 7.3.4)

Let a nonzero  $x \in \mathbb{C}^N$  have  $N_t$  nonzero entries and DFT  $X$  with  $N_f$  nonzero entries.  
Then,

$$N_t N_f \geq N$$

Proof :

- 1  $x \in \mathbb{C}^N$  have  $N_t$  nonzero entries  $\implies X$  cannot have  $N_t$  consecutive zero entries (interpreted mod  $N$ )
- 2 arrange the points of  $X$  in a circle and choose one nonzero entry to start from, this can be followed by at most  $(N_t - 1)$  zeros, then

$$N_f \geq \left\lceil \frac{N}{N_t} \right\rceil \geq \frac{N}{N_t}$$

# Uncertainty principle for finite-length sequences

Proof of point 1. (by contradiction)

- 1  $x \in \mathbb{C}^N$  have  $N_t$  nonzero entries  $\implies X$  cannot have  $N_t$  consecutive zero entries (interpreted mod  $N$ )

- $x \in \mathbb{C}^N, X = \mathcal{F}(x) \in \mathbb{C}^N, W_N = e^{-2\pi j/N}$
- $x \in \mathbb{C}^N$  have  $N_t$  nonzero entries :  $x_{i_0} = \dots = x_{i_{N_t-1}} \neq 0$
- define  $y \in \mathbb{C}^{N_t} : y_\ell = x_{i_\ell} \neq 0$
- suppose  $N_t$  consecutive zero entries in  $X$  :

$$\exists k \in \{0, \dots, N-1\} : X_{k+m} = 0 \quad \forall m \in \{0, \dots, N_t-1\}$$

- notice

$$0 = X_{k+m} = \sum_{n=0}^{N-1} x_n W_N^{(k+m)n} = \sum_{\ell=0}^{N_t-1} x_{i_\ell} W_N^{(k+m)i_\ell} = \sum_{\ell=0}^{N_t-1} y_\ell z_\ell^{k+m}$$

- in vector form :  $Z_{m,\ell} = z_\ell^{k+m}$ , with  $0 \leq m, n \leq N_t - 1$

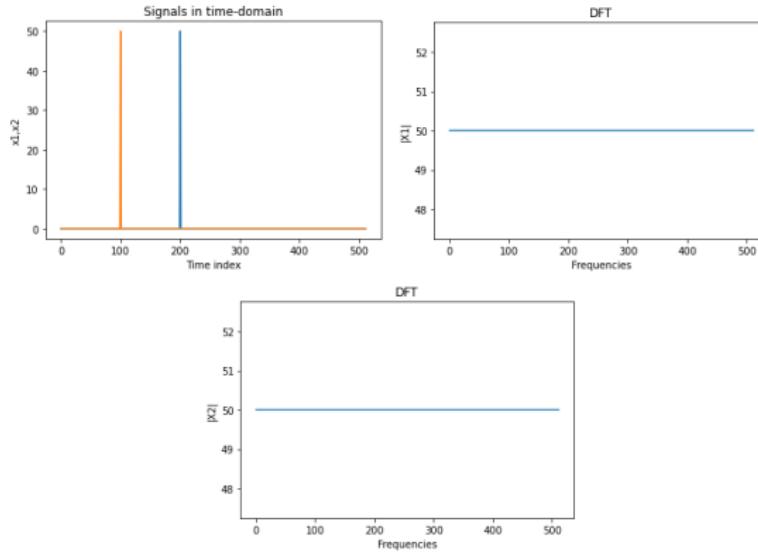
$$0 = Zy$$

- $y$  nonzero  $\implies Z$  not full rank but ...  $Z = \widehat{Z}D$  full rank

- $\widehat{Z}_{m,n} = z_n^m$  Vandermonde matrix (full rank)
- $D = \text{diag}(z_0^k, z_1^k, \dots, z_{N_t-1}^k)$  (full rank)

# Uncertainty principle for finite-length sequences

The equality is achieved...



## Fourier analysis

- DFT : decomposition of a signal into sinusoidal basis functions of different frequencies
- DFT is invertible (no information lost)      -> possiamo sempre recuperare interamente il segnale

## Wavelet analysis

-> permette di rappresentare bene segnali transitori

- DWT : decomposition of a signal into a set of mutually orthogonal wavelet basis functions
- Wavelet functions are spatially localized      -> le ondine sono spazialmente localizzate, quindi sono non nulle solo su una parte del segnale: hanno supporto non nullo su un sottointervallo del segnale
- functions are dilated, translated and scaled versions of a common mother wavelet
- DWT refers not just to a single transform, but rather a set of transforms (Haar, Daubechies, ...)      -> quindi a una famiglia di trasformate
- DWT is invertible (no information lost)      -> è possibile la ricostruzione completa del segnale

# One-dimensional DWT

Esempio trasformata DWT monodimensionale

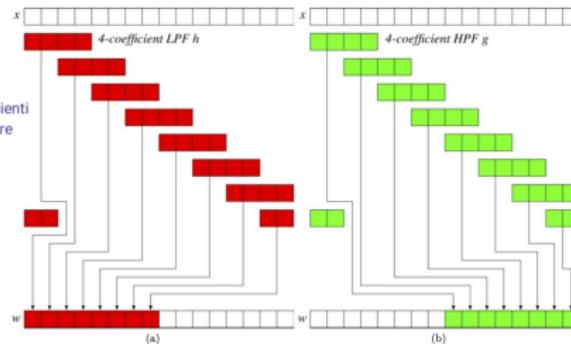
## The (one-dimensional) DWT

- operates on a real-valued vector  $x$  of length  $2^n$ ,  $n \in 2, 3, \dots$ ,
- results in a transformed vector  $w$  of equal length.

→ La DWT trasforma il segnale in un altro segnale della stessa lunghezza

1. Applicazione filtro passa basso
2. Applicazione filtro passa alto

Il segnale viene filtrato a incrementi di due e i coefficienti risultantici vengono messe nelle prime posizioni libere

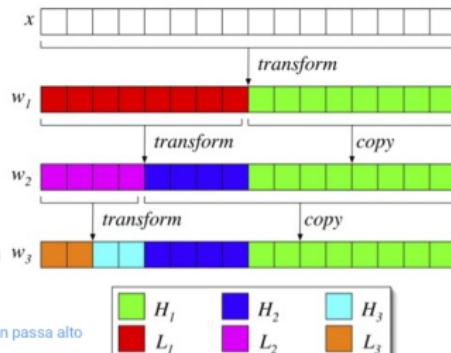


- The original signal is low-pass filtered in increments of two, and the resulting coefficients are grouped as the first eight elements of the vector.
- The original signal is high-pass filtered in increments of two, and the resulting coefficients are grouped as the last eight elements of the vector.

# Three-level wavelet transform

Repeat identical procedure :

Le trasformate si possono fare a diversi livelli (come abbiamo visto in teoria)



- values  $L_3$  are the result of three consecutive low-pass filters,
- values  $H_3$  are the result of two consecutive low-pass filtering operations followed by a high-pass filter
- values  $H_2$  are the result of a low-pass filter followed by a high-pass filter
- values  $H_1$  are the result of one high-pass filter.
- the highest frequencies will be isolated and localized in values  $H_1$  of  $w_3$ , intermediate frequencies will be isolated and localized in values  $H_2$  of  $w_3$ , etc
- lower frequencies, the resolution is decimated by half for each level of the wavelet transform

# Filter coefficients

If  $h$  low-pass filter of length  $n$  and  $g$  high-pass filter, then they lead to basis functions with properties of compactness and orthogonality...

$$g_k = (-1)^k h_{n-k-1}, k \in \{0, \dots, n-1\}$$

Examples :

■  $n = 2$  :

$$h = [c_0 \quad c_1], \quad g = [c_1 \quad -c_0]$$

(e.g. Haar :  $h = [1/\sqrt{2}, \quad 1/\sqrt{2}]$

■  $n = 4$  :

$$h = [c_0 \quad c_1 \quad c_2 \quad c_3], \quad g = [c_3 \quad -c_2 \quad -c_1 \quad -c_0]$$

(e.g. Daubechies-4 :  $h = [\frac{1+\sqrt{3}}{4\sqrt{2}}, \quad \frac{3+\sqrt{3}}{4\sqrt{2}}, \quad \frac{3-\sqrt{3}}{4\sqrt{2}}, \quad \frac{1-\sqrt{3}}{4\sqrt{2}}]$

Press, William H., Brian P. Flannery, Saul A. Teukolsky, and William T. Vetterling. Numerical Recipes in C : The Art of Scientific Computing. 2nd ed., Cambridge University Press, 1992.

# Inverse DWT

La DWT è invertibile e esiste un algoritmo che dai coefficienti della DWT possiamo ritornare al segnale originale

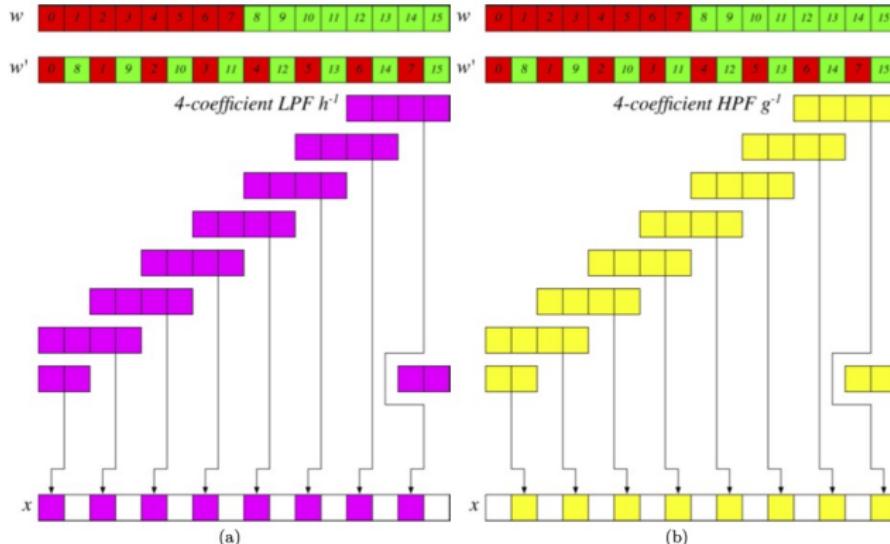
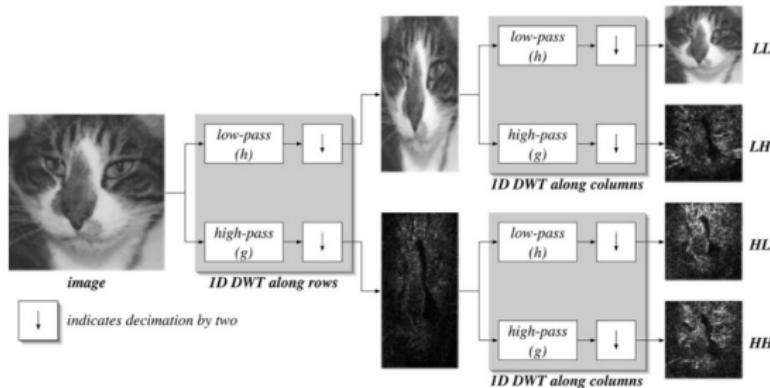


Figure 5: Illustration of the inverse DWT for a one-level DWT  $w$  of length 16. First, the low-pass and high-pass elements of  $w$  are interleaved. Then, (a) the inverse low-pass filter  $h^{-1}$  is applied in increments of two, and (b) the inverse high-pass filter  $g^{-1}$  is applied in increments of two.

# DWT in two dimensions

Two-dimensional DWT through repeated application of the one-dimensional DWT :



- apply a one-level, one- dimensional DWT along the rows of the image.
- apply a one-level, one-dimensional DWT along the columns of the transformed image
- four distinct bands :
  - LL band  $\leadsto$  down-sampled (by a factor of two) version of the original image
  - LH band  $\leadsto$  preserve localized horizontal features
  - HL band  $\leadsto$  to preserve localized vertical features in the original image
  - HH band  $\leadsto$  isolate localized high-frequency point features in the image.