

- Today:
- Wavelet transforms for 2-D signals
  - Preview of wavelets from filter banks
- 

$s[k] \quad k=0, \dots, N-1$

$\mathbb{R}^N \text{ or } \mathbb{C}^N$

$\vec{s} = \begin{bmatrix} s[0] \\ s[1] \\ \vdots \\ s[N-1] \end{bmatrix}$

2-dim'l signal

$s[j,k] \quad j, k=0, \dots, N-1$

$s = \begin{bmatrix} s[0,0] & s[0,1] & \dots & s[0,N-1] \\ \vdots & & & \\ \vdots & & & \\ s[N-1,0] & - & - & - & s[N-1,N-1] \end{bmatrix}$

$\mathbb{R}^{N^2} \text{ or } \mathbb{C}^{N^2}$

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know: when we write vectors as columns  $\begin{bmatrix} \vdots \\ \vdots \end{bmatrix} \in \mathbb{C}^N$   
 linear transformations can write as  
 matrices  $\in \mathbb{C}^{N^2}$

when we write vectors as matrices  $\begin{bmatrix} \phantom{x} \\ \phantom{x} \\ \phantom{x} \end{bmatrix} \in \mathbb{C}^{N^2}$   
 space of lin transformations  
 is now  $\mathbb{C}^{N^4}$

awkward to write down:

But certain nice kinds of lin transformations  
 are easier to describe.

$$\mathbb{C}^{N^2} \longleftrightarrow M_N(\mathbb{C})$$

given  $A, B \in M_N(\mathbb{C})$  get a lin transformation

$$M_N(\mathbb{C}) \xrightarrow{f} M_N(\mathbb{C}) \text{ via}$$

$$T \longmapsto ATB = f(T)$$

$$T+T' \longmapsto A(T+T')B = ATB + AT'B$$

$$\lambda T \longmapsto A(\lambda T)B \quad \text{"} f(T+T') = f(T) + f(T') \text{"}$$

$$f(\lambda T) = \lambda ATB = \lambda f(T)$$

2-d wavelet transforms are given by these kinds  
 of maps:

given a matrix  $S = \left[ \begin{array}{c} \\ \\ \end{array} \right]_{N \times D}$

think of this as a signal,  
can consider each column as a 1-d signal

$$\left[ s[* , 0] \mid s[* , 1] \mid \dots \right]$$

can apply  $T_a$  to each column.

$$\left[ \begin{array}{c} s[* , 0] \mid s[* , 1] \mid \dots \\ S \end{array} \right] \rightsquigarrow \left[ \begin{array}{c} T_a s[* , 0] \mid T_a s[* , 1] \mid \dots \\ T_a S \end{array} \right]$$

$$T \left[ v_0 \mid v_1 \mid v_2 \mid \dots \mid v_{N-1} \right] = \left[ T v_0 \mid T v_1 \mid \dots \mid T v_{N-1} \right]$$

$$\left[ \begin{array}{c} w_0 \\ \hline w_1 \\ \hline \vdots \\ \hline w_{N-1} \end{array} \right] T^t = \left[ \begin{array}{c} T w_0 \\ \hline T w_1 \\ \hline \vdots \\ \hline T w_{N-1} \end{array} \right]$$

$$\left[ \begin{array}{c} s[0, \neq] \\ \hline s[1, \neq] \\ \hline \vdots \\ \hline s[N-1, \neq] \end{array} \right] T_a^t = \left[ \begin{array}{c} T_a s[0, \neq] \\ \hline T_a s[1, \neq] \\ \hline \vdots \\ \hline T_a s[N-1, \neq] \end{array} \right]$$

Def: The 2-d wavelet transform of  $s$  is

$$T_a s T_a^t = \left[ \begin{array}{c|c} t_0^v & t_1^v \\ \hline d_0^v & d_1^v \end{array} \right] T_a^t$$

$$= \left[ \begin{array}{c|c} \text{hor \& wnt trend} & \text{horiz detail of wnt trend} \\ \hline \text{vertical detail of horz trend} & \text{wnt detail of vertical detail} \end{array} \right]$$

$$\approx \begin{bmatrix} \text{blurred or} \\ \text{denoised} \\ \text{image} & \text{vertical} \\ & \text{features} \\ \hline \text{horizontal} \\ \text{features} & \text{diagonal} \\ & \text{features} \end{bmatrix}$$

Step 6 : shift focus from periodic & finite fixed length signals to signals of arbitrary length.

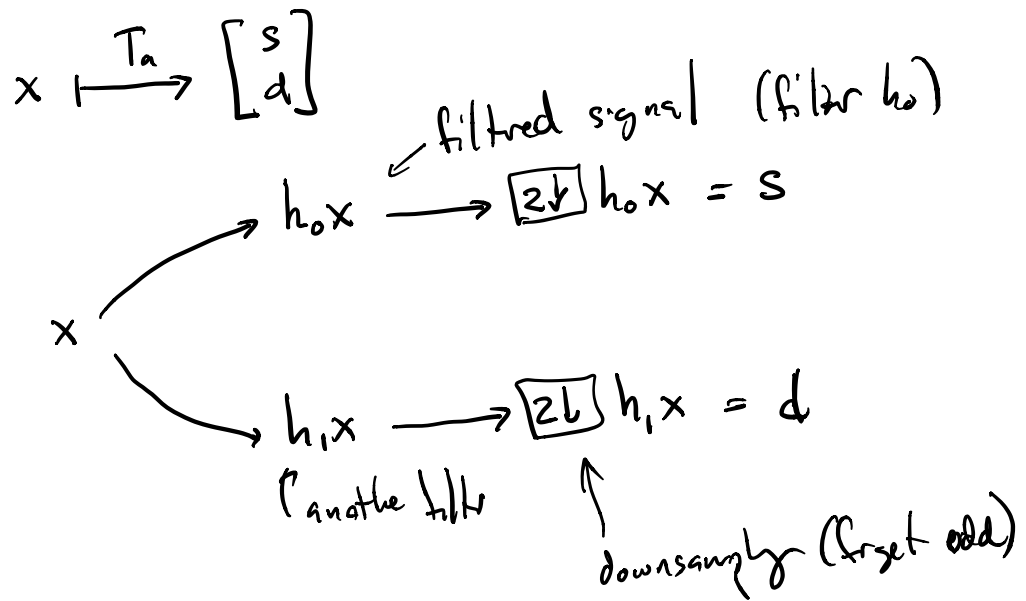
Punchline: all tools we've developed transfer painlessly to this context.

Nice language (trashed from DFT) to talk about frequencies among signals of arbitrary length

"Frequency spectrum"?

Filters in this context. (still described by (linear) convolution)

Step 2: Wavelet transforms for these.



The search for wavelets with good (desirable) properties  
 $\Leftrightarrow$  the search for good pairs of filters  $(h_0, h_1)$