

Linear transform

compute y

$$y = Tx$$

where x is the input vector, y the output vector and T the fixed transform matrix

Example: DFT

1. form (standard in SP): given x_0, \dots, x_{n-1} , compute

$$y_k = \sum_{\ell=0}^{n-1} e^{-2\pi i j k \ell / n} x_\ell, \text{ for } 0 \leq k < n, \quad \boxed{j = \sqrt{-1}}$$

2. form: set $\omega_n = e^{-2\pi i j / n}$ (primitive, n th root of 1)

$$y_k = \sum_{\ell=0}^{n-1} \omega_n^{k\ell} x_\ell, \text{ for } 0 \leq k < n$$

3. form: $x = (x_0, \dots, x_{n-1})^T$, $y = (y_0, \dots, y_{n-1})^T$, compute

$$y = \text{DFT}_n \cdot x, \quad \text{DFT}_n = [\omega_n^{k\ell}]_{0 \leq k, \ell < n}$$

For example:

$$\text{DFT}_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}, \quad \text{DFT}_4 = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & 1 & -j \\ 1 & 1 & -1 & -1 \\ 1 & j & -1 & j \end{bmatrix}$$

Transform algorithms

you want to compute $y = Tx$, T is $n \times n$.

an algorithm is given by a factorization

$$T = T_1 T_2 \dots T_m, \quad \text{all } T_i \text{ are } n \times n$$

namely you can compute y as:

$$t_1 = T_m x$$

$$t_2 = T_{m-1} x$$

$$\vdots$$

$$y = T_1 \cdot t_{m-1}$$

(m steps = m $\Pi V I S$)

this reduces the operations count only if

- the T_i are sparse

- m is not too large

Note: generic NVM (i.e., matrix is not known beforehand) has complexity $\Theta(n^2)$

Example: fast Fourier transform (FFT) for $n=4$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & -1 \\ & 1 & 1 \\ & & 1 & -1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

compare cost:

- by definition, DFT_4 : 12 adds, 4 mults by j (all complex)

- using FFT (4 steps): 8 adds, 1 mult by j

the sparse matrices are structured:

$$= (DFT_2 \otimes I_2) \text{diag}(1, 1, 1, j) (I_2 \otimes DFT_2) L_2^4$$

(explained next)

Structured matrices

- $DFT_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$ (butterfly matrix)

- $I_n = \begin{pmatrix} \ddots & \\ & \ddots & \\ & & 1 \end{pmatrix}$

- $\text{diag}(a_0, \dots, a_{n-1}) = \begin{pmatrix} a_0 & & \\ & \ddots & \\ & & a_{n-1} \end{pmatrix}$

- $A \oplus B = \begin{pmatrix} A & \\ & B \end{pmatrix}$

- $A \otimes B = [a_{k,e} \cdot B]_{0 \leq k, e < n}$ if $A = [a_{k,e}]_{0 \leq k, e < n}$

$A \ n \times \ n, \ B \ m \times \ m \Rightarrow A \otimes B \ nm \times \ nm$

Example:

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \otimes DFT_2 = \begin{bmatrix} 1 & 1 & 2 & 2 \\ 1 & -1 & 2 & -2 \\ \hline 3 & 3 & 4 & 4 \\ 3 & -3 & 4 & -4 \end{bmatrix}$$

↑
coarse
structure

↑
fine
structure

$$I_n \otimes A = \begin{bmatrix} A & & & \\ & A & & \\ & & \dots & \\ & & & A \end{bmatrix}$$

contains n A s

$$A \otimes I_n = \begin{bmatrix} \circ & \circ & \circ & \dots \\ \circ & \circ & \circ & \dots \\ \circ & \circ & \circ & \dots \\ \dots & \dots & \dots & \dots \end{bmatrix}$$

- each block is an $n \times n$ I_n
- all circles together constitute one A
- contains n A s