

Discrete Fourier Transform

Suppose $x \in \mathbb{C}^N$, then the N -point DFT of x is given by

$$F: \mathbb{C}^N \rightarrow \mathbb{C}^N$$
$$x \mapsto \hat{x} \text{ with}$$

$$\hat{x}(k) = \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} x_n e^{-2\pi i k n / N}$$

so $\hat{x}(k) = \langle x, \varphi^{(k)} \rangle = \varphi^{(k)*} x$

$$\varphi^{(k)} = \begin{pmatrix} 1 \\ e^{2\pi i k / N} \\ e^{2\pi i \cdot 2k / N} \\ \vdots \\ e^{2\pi i (N-1)k / N} \end{pmatrix} \cdot \frac{1}{\sqrt{N}}$$

(= inner product with complex exponential of frequency k/N)

Matrix representation:

Let $w := w_N = e^{-2\pi i / N} = N^{\text{th}}$ root of unity

$$\hat{x} = Fx$$

$$\hat{x} = \frac{1}{\sqrt{N}} \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & \omega & \omega^2 & \dots & \omega^{N-1} \\ 1 & \omega^2 & \omega^4 & \dots & \omega^{2N-2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^{N-1} & \omega^{2N-2} & \dots & \omega^{(N-1)^2} \end{bmatrix} x$$

DFT Matrix (note the adjoint)

Theorem: F is a unitary matrix

Proof: Check that $\langle \varphi^{(k)}, \varphi^{(l)} \rangle = \delta_{k,l}$
 $= \begin{cases} 1, & k=l \\ 0, & \text{oth} \end{cases}$

Corollary: $\hat{x} = Fx$

$$\Rightarrow x = F^* \hat{x}$$



Inverse DFT.

Some comments:

Continuous time Fourier transform:

- $f \in L^1$ (or L^2 with appropriate "work")

$$\hat{f}(\omega) = \mathcal{C} \int_{-\infty}^{\infty} f(t) e^{-i\omega t} dt \approx \langle f, \varphi_{\omega} \rangle$$

$\varphi_{\omega}(t) = e^{-i\omega t}$

gives the "coefficients" of f at frequency ω .

- The DFT gives the frequency coefficients of x at frequency $\frac{k}{N}$.

Convenient Properties of the DFT

- $$x(n) = \frac{1}{\sqrt{N}} \sum_{k=0}^{N-1} \hat{x}(k) e^{2\pi i kn/N}$$

$$= \langle \hat{x}, \bar{\varphi}^{(k)} \rangle$$

(Inversion)

$$\left. \begin{array}{l} x \mapsto \hat{x} \\ y \mapsto \hat{y} \end{array} \right\} \Rightarrow \underbrace{\langle x, y \rangle = \langle \hat{x}, \hat{y} \rangle}_{\text{Parseval}}$$

$$\left. \begin{array}{l} \|x\|^2 = \|\hat{x}\|^2 \\ \hookrightarrow \sum_{n=0}^{N-1} |x_n|^2 \end{array} \right\} \text{Plancherel}$$

Phase shift:

Suppose $x \mapsto \hat{x}$

Let $y : y(n) = x(n) e^{2\pi i m n / N}$

then $\hat{y}(k) = \hat{x}(k - m)$

Proof \exists exercise (change of variable)

• Convolution: (circular)

Define the circular convolution of $a, b \in \mathbb{C}^N$ by

$$(a \circledast b)_n := \sum_{l=0}^{N-1} a_l b_{(n-l) \bmod N}$$

Then

(I) if $x \mapsto \hat{x}$, $y \mapsto \hat{y}$

$$\frac{1}{\sqrt{N}}(x \circledast y)_n = \left[F^{-1}(\hat{x} \circ \hat{y}) \right]_n$$

pointwise multiplication

(II)

$$F(x \circ y)_k = \frac{1}{\sqrt{N}} \sum_{n=0}^{N-1} x_n y_n e^{-\frac{2\pi i k n}{N}}$$

$$= \frac{1}{\sqrt{N}} (\hat{x} \circledast \hat{y})_k$$

Proof \circ HW

Corollary: The DFT matrix
diagonalizes any circulant
matrix.

Proof: HW

—x—

Applications: (just a couple of examples)

- Signal analysis: understanding
the frequency content of a signal

- extension: STFT

$$x \in \mathbb{C}^N \mapsto \tilde{x} : \tilde{x}(k, \ell) =$$

$$\sum_{n=0}^{N-1} x(n) \underbrace{w(n-\ell)}_{\substack{\leftarrow \\ \text{compactly supported} \\ \text{or approximately so}}} e^{-2\pi i k n / N}$$

Exercise: • Visualize the STFT of a few seconds of music

• Use W supported on approximately ~~60~~ - 120 msec.

• Compression (e.g. Via the DCT)

JPEG works this way.

• Major advantage: the
Fast Fourier Transform
algorithm.

Normally to compute a matrix-vector product of the form

$$\hat{x} = Fx$$

$\downarrow \qquad \downarrow$
 $N \times N \qquad N \times 1$

we need $O(N^2)$ operations.

In the case of the DFT,

the FFT algorithm allows us to do it in $O(N \log N)$!

Idea (Cooley-Tukey 1965, Gauss?)

Let $N = 2^n$ and ignore the $\frac{1}{\sqrt{N}}$ normalization for now.

$$\hat{x}_k = \sum_{n=0}^{N-1} x_n e^{-2\pi i k n / N}$$

$$= \sum_{n \text{ even}} (\quad) + \sum_{n \text{ odd}} (\quad)$$

$$= \underbrace{\sum_{m=0}^{N/2-1} x_{2m} e^{-2\pi i (2m) k/N}}_{e_k}$$

$$+ \underbrace{\sum_{m=0}^{N/2-1} x_{2m+1} e^{-2\pi i (2m+1) k/N}}_{e^{-2\pi i k/N} \sum_{m=0}^{N/2-1} x_{2m+1} e^{-2\pi i k m / \frac{N}{2}}} = e^{-2\pi i k/N} O_k$$

So $\hat{x}_k = e_k + e^{-2\pi i k/N} O_k$

\downarrow
 $\frac{N}{2}$ point DFT

\swarrow $\frac{N}{2}$ pt DFT
of $(x_1, x_3, \dots, x_{2n-1})$

of $(x_0, x_2, \dots, x_{2n-2})$

for $k \in \{0, \dots, N/2-1\}$

Similarly

$$\hat{x}_{k+N/2} = e_k - e^{-2\pi i k/N} O_k$$

for $k \in \{N/2, \dots, N-1\}$

Now, we can compute each of the 2 $N/2$ pt DFTs by repeating this even/odd splitting

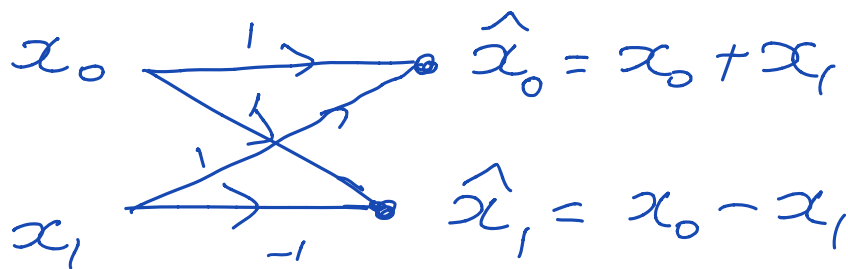
$\Rightarrow N/4$ pt DFT $\rightarrow N/8$ pt DFT
 $\rightarrow \dots \rightarrow 2$ pt DFT

(Butterfly) \leftarrow

$$F_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

Examples:

2 pt DFT:



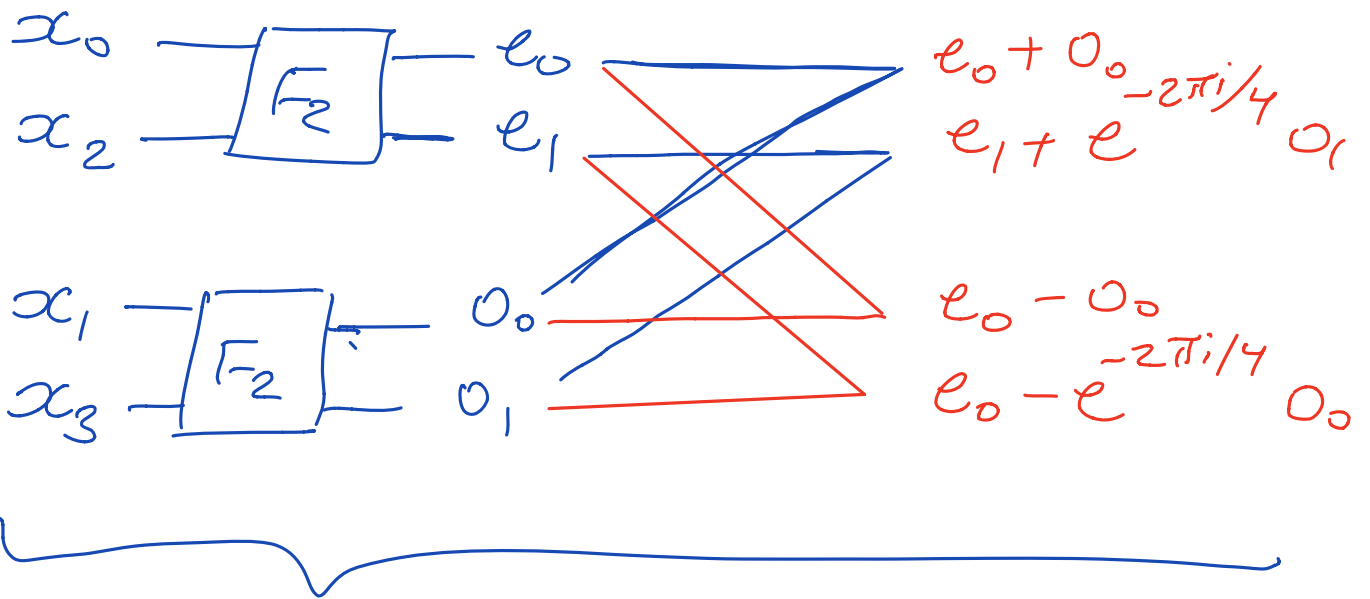
write this as



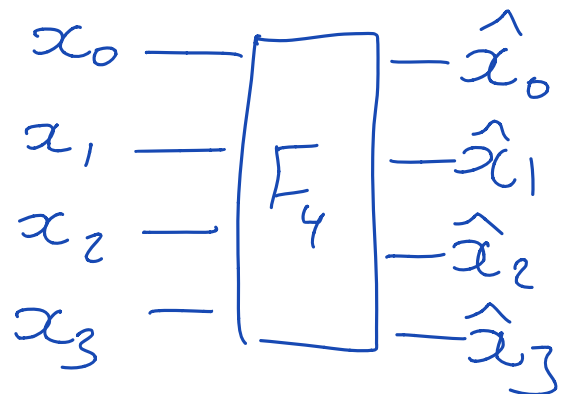
4 pt DFT

Recall $\hat{x}_k = e_k + e^{-2\pi i k/N} O_k$

$\hat{x}_{k+N/2} = e_k - e^{-2\pi i k/N} O_k$



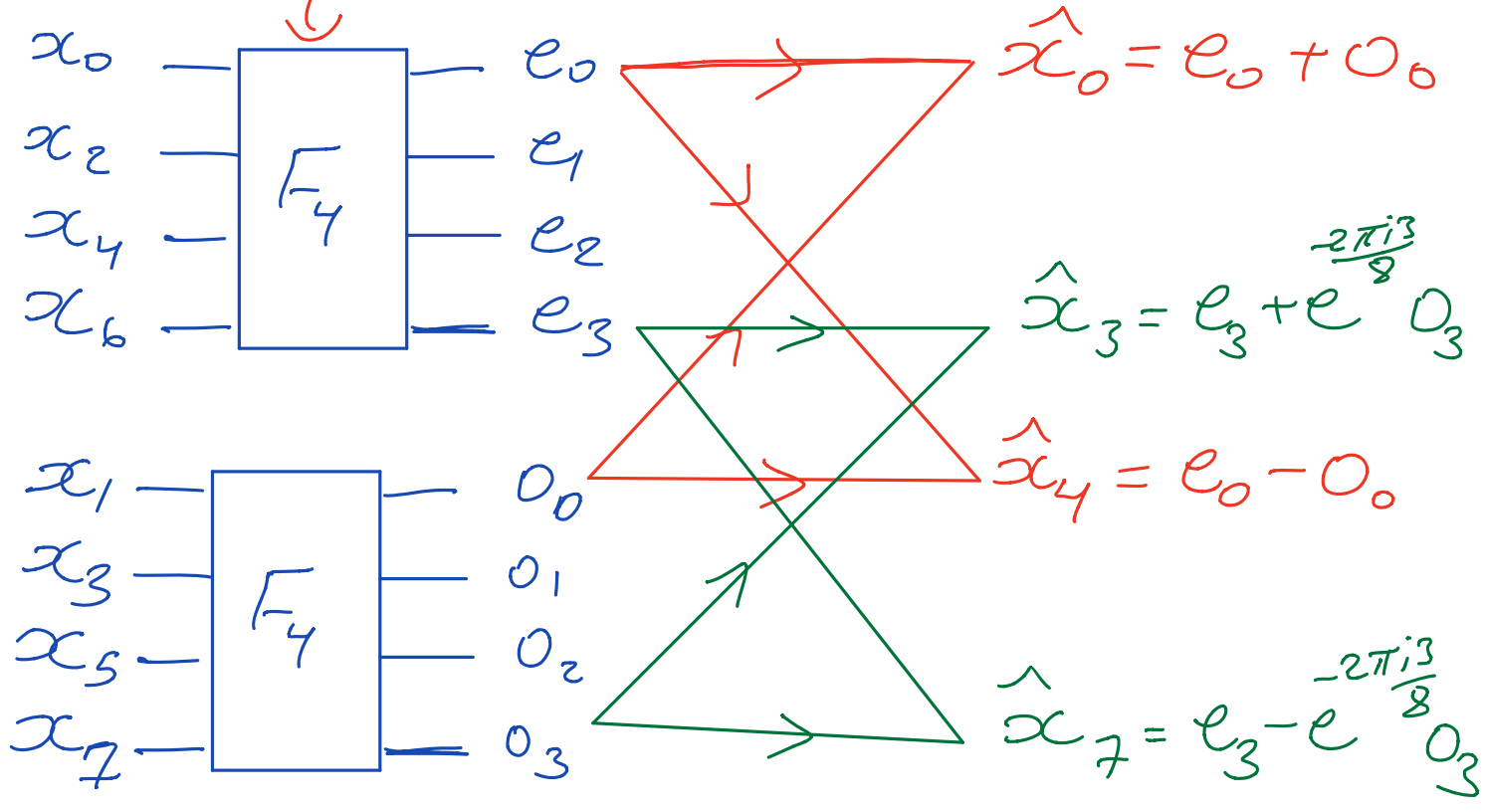
write this as



8 pt DFT

Same idea

as above



Computational Complexity

- We have $\log_2(N)$ levels
 - Each level has $N/2$ butterfly operations
 - Each butterfly is
 - 2 complex adds
 - & 1 complex mult.
- $\Rightarrow O(N \log N)$

Remarks: Need to reorder the input (due to even/odd splitting at every stage)

binary rep
of index

bit reversal

$x_0 \rightarrow$	000	\rightarrow	000
$x_4 \rightarrow$	100		001
$x_2 \rightarrow$	010		010
$x_6 \rightarrow$	110		011
$x_1 \rightarrow$	001		100
$x_5 \rightarrow$	101		101
$x_3 \rightarrow$	011		110
$x_7 \rightarrow$	111		111

What about F^{-1} ?

Can show that with unitary DFT

$$x = F(\hat{x}) \Rightarrow O(N \log N)$$

also