

DFT

- DTFT

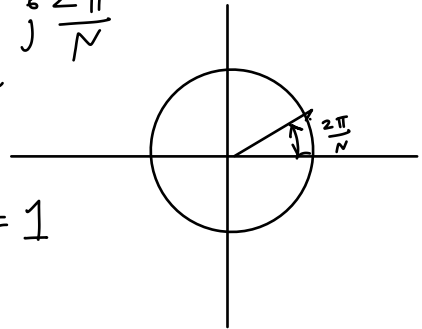
- N -point to N -point transform
i.e. $N \times N$ matrix.

$$X^d(k) = \sum_{n=0}^{N-1} x(n) W_N^{-nk}$$

$$k = 0, \dots, N-1$$

$$W_N = e^{j \frac{2\pi}{N}}$$

$$W_N^N = 1$$



$$\begin{array}{l}
 k=0 \\
 k=1 \\
 \vdots \\
 k=N-1
 \end{array}
 \begin{bmatrix}
 X^d(0) \\
 X^d(1) \\
 \vdots \\
 X^d(N-1)
 \end{bmatrix}
 =
 \begin{bmatrix}
 1 & 1 & 1 & 0 & 0 & 0 & 1 \\
 1 & W^{-1} & W^{-2} & W^{-3} & \dots & W^{-(N-1)} \\
 1 & W^{-2} & W^{-4} & W^{-6} & \dots & W^{-2(N-1)} \\
 \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\
 1 & W^{-(N-1)} & \dots & \dots & \dots & W^{-(N-1)(N-1)}
 \end{bmatrix}
 \begin{bmatrix}
 X(0) \\
 X(1) \\
 \vdots \\
 X(N-1)
 \end{bmatrix}
 \begin{array}{l}
 n=0 \\
 n=1 \\
 \vdots \\
 n=N-1
 \end{array}$$

DFT Matrix of size $N \times N$

$$\underline{X^d} = F_N \underline{X}$$

$N \times N$
matrix

$$F = F^t$$

where F^t is the
transpose
(not conj transpose)

Inverse DFT

$$x(n) = \frac{1}{N} \sum_{k=0}^{N-1} X^d(k) W_N^{nk}$$

So

$$F^{-1} = \frac{1}{N} F^* = \frac{1}{N} F^H$$

F^H is "Hermitian transpose"

or conjugate transpose.

$$F^{-1} F = F F^{-1} = I$$

$$F^{-1} = \frac{1}{N} F^H$$

$$\rightarrow F^H F = N \cdot I$$
$$F \cdot F^H = N \cdot I$$

Parseval's property:

$$E = \sum_{n=0}^{N-1} |x(n)|^2 \quad \text{"Energy" of the signal.}$$

$$= \sum_{n=0}^{N-1} \text{conj}(x(n)) \cdot x(n)$$

$$= X^H X \quad \text{where } X = \begin{bmatrix} x(0) \\ x(1) \\ \vdots \\ x(N-1) \end{bmatrix} \quad \text{Signal as a column vector}$$

$$\sum_{k=0}^{N-1} |X^d(k)|^2 = (X^d)^H X^d$$

$$= (F X)^H (F X) = X^H \overset{N \cdot I}{F^H F} X$$

$$= X^H (N \cdot I) X = N X^H X = N \sum_{n=0}^{N-1} |x(n)|^2$$

$$E = \sum_{n=0}^{N-1} |x(n)|^2 = \frac{1}{N} \sum_{k=0}^{N-1} |X^d(k)|^2$$

key property for Parseval's Energy Identity

$$F^{-1} = \alpha F^H$$

If a matrix F satisfies $F^{-1} = F^H$
then F is a "unitary" matrix.

If F is also real then $F^{-1} = F^t$
and F is called a "orthonormal" matrix.

And F represents a transform with Parseval's property.

convolution as a matrix-vector multiplication

$$y = h * x$$

Linear convolution.

$$\begin{bmatrix} y(0) \\ y(1) \\ y(2) \\ y(3) \\ y(4) \end{bmatrix} = \begin{bmatrix} h(0) & 0 & 0 \\ h(1) & h(0) & 0 \\ h(2) & h(1) & h(0) \\ 0 & h(2) & h(1) \\ 0 & 0 & h(2) \end{bmatrix} \begin{bmatrix} x(0) \\ x(1) \\ x(2) \end{bmatrix}$$

5x1

5x3

3x1

$$\boxed{y = H x}$$

← TOEPLITZ

MATRIX

[const along
each diag]

Circular convolution as $m \times n$ -vec mult...

$$y = h \circledast x$$

$$\begin{bmatrix} y(0) \\ y(1) \\ y(2) \end{bmatrix} = \begin{bmatrix} h(0) & h(2) & h(1) \\ h(1) & h(0) & h(2) \\ h(2) & h(1) & h(0) \end{bmatrix} \begin{bmatrix} x(0) \\ x(1) \\ x(2) \end{bmatrix}$$

3×1

3×3

3×1

circulant matrix.

Convolution property of the DFT

$$X^d = Fx \quad H^d = Fh \quad Y^d = Fy$$

$$\Rightarrow Y^d(k) = H^d(k) X^d(k) \Leftrightarrow Y^d = \text{diag}(H^d) X^d$$

$$\begin{bmatrix} Y^d(0) \\ Y^d(1) \\ \vdots \\ Y^d(N-1) \end{bmatrix} = \begin{bmatrix} H^d(0) X^d(0) \\ H^d(1) X^d(1) \\ \vdots \\ H^d(N-1) X^d(N-1) \end{bmatrix} = \begin{bmatrix} H^d(0) & & & \\ & H^d(1) & & \\ & & \ddots & \\ & & & H^d(N-1) \end{bmatrix} \begin{bmatrix} X^d(0) \\ X^d(1) \\ \vdots \\ X^d(N-1) \end{bmatrix}$$

$$= Fy = \text{diag}(H^d) X^d$$

$$F^{-1} [Fy = \text{diag}(Fh) Fx] \Rightarrow y = F^{-1} \text{diag}(Fh) Fx$$

$$y = h \circledast x \Leftrightarrow y = Hx$$

↑
circulant matrix

So the DFT matrix diagonalizes
any circulant matrix.

$$H = F^{-1} \text{diag}(Fh) F$$

↑
circulant matrix (h)

matrix factorization result.

$$H^f(\omega) = A(\omega) e^{j\theta(\omega)}$$

$$\begin{aligned} |H^f(\omega)| &= |A(\omega) e^{j\theta(\omega)}| \\ &= |A(\omega)| |e^{j\theta(\omega)}| \\ &= |A(\omega)| \end{aligned}$$

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