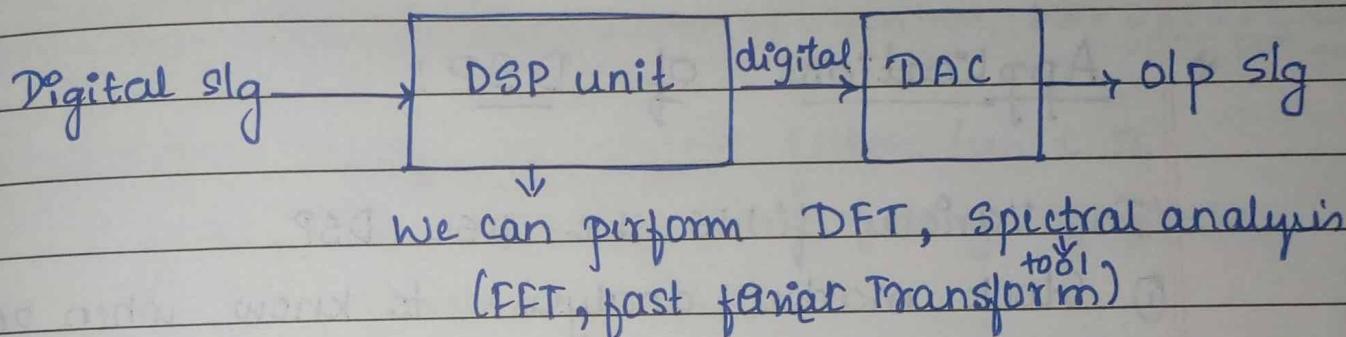
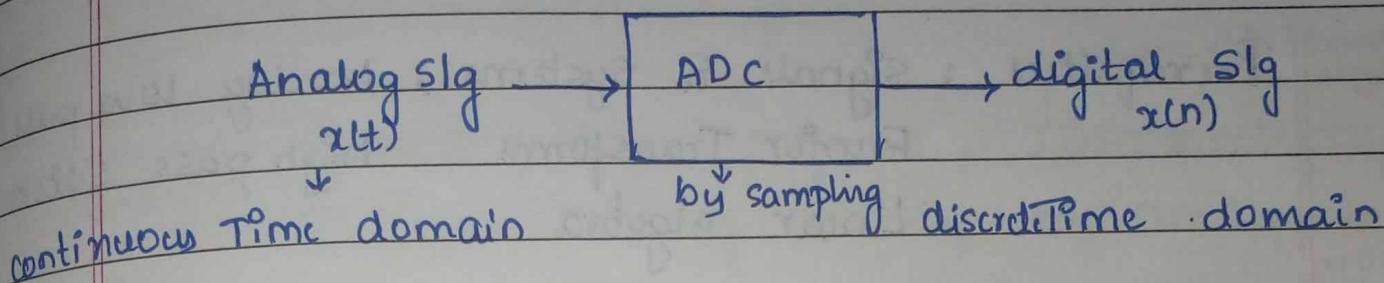


Intro to DSP: "Digital Signal Processing"

- * All slgs in nature are Analog slg
- * But to process we require Digital slgs. to do real tym slg processing



We can use filters to remove noise slg.

Filters → FIR (Finite impulse response filters)
IIR (Infinite " " ")

levels of slg processing :

- I System level ⇒ can use 'c' programming + python, any image processing tool, Audio processing tool or matlab, simulink and so on.

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→ (CMOS)
② Board level \Rightarrow FPGA, Transistor level
(chip level)
Ex: mobile

* Analog Sigs are captured by sensors/hardware

Prerequisites : signals & systems, Analog low pass/
Fourier Transforms high pass filters
Linear algebra

* Applications of DSP:

In various fields we use DSP

- ① Stock market analysis: to know when profits are peak
- ② Biomedical: ECG, EEG analysis
- ③ Radar Sig processing: to detect targets
- ④ Sonar Sig " : in sea to navigate
- ⑤ Communication Sigs: Image/Audio processing
- ⑥ Audio, Video: Speaker Recognitn., Cepstral analysis
- ⑦ Seismic
- ⑧ Satellite/Rocket testing

& types of analysis

→ Time domain analysis:

→ Frequency "

→ Noise removal

Filters → Butterworth

→ IIR → chebyshev

→ FIR → windows filter

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Unit I: ① DFT (Discrete Fourier Transform), IDFT \Leftrightarrow properties
DFT in linear filtering, overlap add, Overlap method
convolution mechanism

② FFT (Fast Fourier Transform) \rightarrow 2 types of algorithm, DIT FFT, DIF FFT

Unit II: Filters

Chapter ③ :- Design of digital filters (unit II)

FIR filters: types, methods of designing filters

④ IIR filters \rightarrow

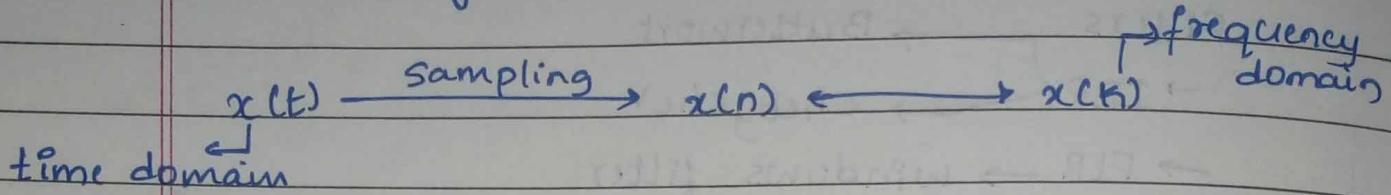
Unit III: Chapter ⑤ Digital Realizatns. for FIR filters

" " " " " IIR filters

Time and frequency domain signals:

$x(t)$ → analog domain

$x(n)$ → digital domain



→ Freq domain Sampling and Reconstruction of discrete time signals.

→ Why DFT (Discrete Fourier Transform) ?

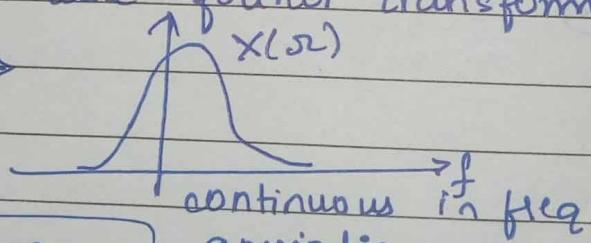
We cannot process a signal directly through digital signal processor

So we will take a part of sig in tym domain $x(t)$
w is continuous in time and Aperiodic in nature.

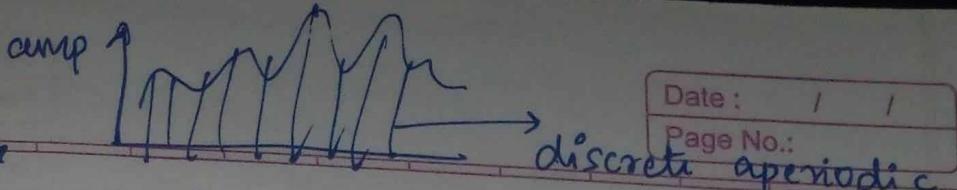
We will now take fourier transform of that

Conti Blg
nuous
aperiodic

CTFT →



$$X(j\omega) = \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt$$

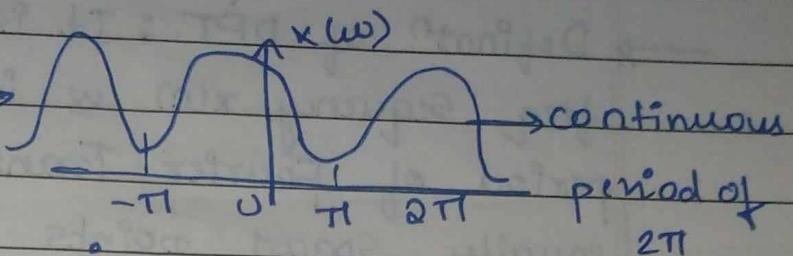


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Now we sample the sig with sampling freq using nyquist criteria

DTaperiodic - DTFT



$$X(w) = \sum_{n=-\infty}^{\infty} x(n) e^{-jwn}$$

$$\omega = \frac{2\pi k}{N} \quad N = \text{total number of samples}$$

$$X\left(\frac{2\pi k}{N}\right) = \sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi kn}{N}}$$

$$X(k) = \sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi kn}{N}} \quad k = 0, 1, 2, \dots, N-1$$

DFT ←

↳ This can be matched to discrete DS processors

IDFT (Inverse discrete fourier Transform)

$$x(n) = \frac{1}{N} \sum_{k=0}^{N-1} X(k) e^{j\frac{2\pi kn}{N}}$$

This is known as N pt IDFT.

* Definitⁿ of DFT and IDFT:

→ Definition of DFT : It is a finite duration discrete freq sequence $x(n)$ \underline{w} is obtained by sampling one period of Fourier Transform Sampling is done at N equally spaced points over the period from $w=0$ to 2π

$$\text{DFT} \Rightarrow x(k) = \sum_{n=0}^{N-1} x(n) e^{-\frac{j2\pi nk}{N}} \quad k=0, 1, 2, \dots, N-1$$

freq domain L, tym domain

$x(n) \xrightarrow{\text{DFT}} x(k)$
 $\xleftarrow{\text{IDFT}}$

$$\text{IDFT} \Rightarrow x(n) = \frac{1}{N} \sum_{k=0}^{N-1} x(k) e^{\frac{j2\pi nk}{N}}, n=0, 1, \dots, N-1$$

Twiddle factor notation

$$w_N = e^{-\frac{j2\pi}{N}}$$

Twiddle factor

$$\therefore \text{DFT} \rightarrow x(k) = \sum_{n=0}^{N-1} x(n) w_N^{kn} \quad k=0, 1, 2, \dots, N-1$$

$$\text{IDFT} \rightarrow x(n) = \frac{1}{N} \sum_{k=0}^{N-1} x(k) w_N^{-kn} \quad k=0, 1, 2, \dots, N-1$$

$n = 0, 1, 2, \dots, N-1$

* Twiddle factors and zero Padding :

$$w_N = e^{-j\frac{2\pi}{N}} \text{ is twiddle factor}$$

$N = 2, 4, 8, 16, \dots$

$$\text{W.K.T } e^{j\theta} = \cos\theta - j\sin\theta$$

For $N=2$

$$w_N = e^{-j\frac{2\pi}{2}} = e^{-j\pi} \\ = \cos\pi - j\sin\pi$$

$$w_2 = -1$$

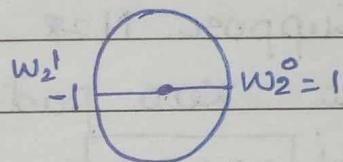
IDE

$$w_N^K = e^{-j\frac{2\pi k}{N}}$$

$$\rightarrow w_2^0 = e^{-j\frac{2\pi \cdot 0}{2}} = 1$$

$$\rightarrow w_2^1 = e^{-j\frac{2\pi \cdot 1}{2}} = e^{-j\pi} = -1$$

Euler's diagram is



When $N=4$ $w_4^K = e^{-j\frac{2\pi k}{4}} = e^{-j\pi k/2}$

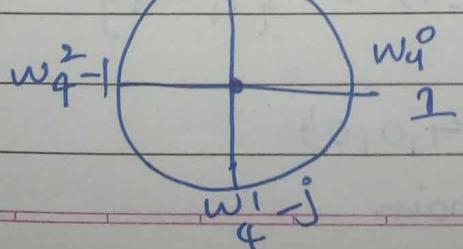
$$w_4^0 = e^{-j\pi \cdot 0} = 1$$

$$w_4^1 = e^{-j\pi/2} = \cos\pi/2 - j\sin\pi/2 = -j$$

$$w_4^2 = e^{-j\pi} = -1$$

$$w_4^3 = e^{-j3\pi/2} = \cos 3\pi/2 - j\sin 3\pi/2 = j$$

$w_4^3 j \sqrt{-1}$ Euler diagram



$$\begin{aligned}
 \text{When } N=8, \quad w_8^k &= e^{-j\frac{2\pi k}{8}} = e^{-j\frac{\pi k}{4}} \\
 w_8^0 &= 1 \\
 w_8^1 &= 0.707 - j0.707 \\
 w_8^2 &= -j \\
 w_8^3 &= -0.707 - j0.707 \\
 w_8^4 &= -1 \\
 w_8^5 &= -0.707 + j0.707 \\
 w_8^6 &= +j \\
 w_8^7 &= 0.707 + j0.707
 \end{aligned}$$

Euler's diagram

→ Zero Padding: Used in signal processing applications
 i.e. Adding slgs of '0' Amplitude in the time domain to end of sequence

$x(n)$

(to time index)

Suppose $N=8$ & we have only $1, +1$,
 then we are going to zero pad remaining 5

i.e., $1, 1, -1, \boxed{\dots \dots}$

↳ going to end sig with '0' Amplitude, not adding any info. only extending the sig with 0 Amplitude

→ If I have to zero pad for $N=4$
 given sig is $x(n) = \{1, -4\}$

Ans is $x(n) = \{1, -4, 0, 0\}$
 (Time domain)

* DFT as a linear Transformation: (Matrix method)

DFT i/p sig be $x_N = \begin{bmatrix} x_0 \\ x_1 \\ \vdots \\ x_{N-1} \end{bmatrix}$ $N \times 1$

$X_N = \begin{bmatrix} X(0) \\ X(1) \\ \vdots \\ X(N-1) \end{bmatrix}$ $N \times 1$ \rightarrow freq domain

By defⁿ

$$X(k) = \sum_{n=0}^{N-1} x(n) e^{-\frac{j2\pi nk}{N}}$$

$$W_N = e^{-\frac{j2\pi}{N}}$$

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

But $N = 4$, $\therefore x(n) = \{x(0), x(1), x(2) \text{ & } x(3)\} \rightarrow$ Time domain

In freq domain $\{X(0), X(1), X(2), X(3)\}$

$$W_4^0 = 1 \quad W_4^1 = j, \quad W_4^2 = -1 \quad W_4^3 = -j \quad (\text{W.KT})$$

$\Rightarrow X(k) = [W_N] x_N$

DFT $\boxed{x(k)} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} x(n)$

↑ complex conjugate

$$\text{IDFT } x(n) = \frac{1}{N} \begin{bmatrix} W_N^* \\ 1 & 1 & 1 & 1 \\ 1 & j & -j & -j \\ 1 & -j & 1 & -1 \\ 1 & -j & -1 & j \end{bmatrix} X(k)$$

here $N=4$

$$\therefore \frac{1}{4} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -j & -j \\ 1 & -j & 1 & -1 \\ 1 & -j & -1 & j \end{bmatrix} X(k)$$

PROPERTIES OF DFT:

① * Linearity & Periodicity Property

1] LINEARITY :-

Suppose we have sig $x_1(n)$ & $x_2(n)$ DFT is $X_1(k)$

$$x_1(n) \xrightarrow{\text{DFT}} X_1(k)$$

$$x_2(n) \xrightarrow{\text{DFT}} X_2(k)$$

then, property states

$$a_1 x_1(n) + a_2 x_2(n) \xrightarrow{\text{DFT}} a_1 X_1(k) + a_2 X_2(k) //$$

$$\text{W.K.T } X_1(k) = \sum_{n=0}^{N-1} x_1(n) e^{-j\frac{2\pi n k}{N}}$$

$$\text{E } X_2(k) = \sum_{n=0}^{N-1} x_2(n) e^{-j\frac{2\pi n k}{N}}$$

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$$\text{DFT } \{a_1 x_1(n) + a_2 x_2(n)\} = \sum_{n=0}^{N-1} [a_1 x_1(n) + a_2 x_2(n)] e^{-j\frac{2\pi n k}{N}}$$

$$= \sum_{n=0}^{N-1} a_1 x_1(n) e^{-j\frac{2\pi n k}{N}} + \sum_{n=0}^{N-1} a_2 x_2(n) e^{-j\frac{2\pi n k}{N}}$$

$$\text{DFT } \{a_1 x_1(n) + a_2 x_2(n)\} = a_1 X_1(k) + a_2 X_2(k) //$$

hence proved.

2] PERIODICITY:

$$x(n) \xrightarrow{\text{DFT}} X(k)$$

$$x(n+N) \xrightarrow{\text{DFT}} X(k)$$

$$X(k+N) = X(k)$$

$$X(k+N) = \sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi n(k+N)}{N}}$$

$$= \sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi n k}{N}} e^{-j\frac{2\pi n N}{N}} \quad ; \quad e^{-j\frac{2\pi n N}{N}} = 1$$

$$X(k+N) = X(k) \rightarrow \text{hence proved.}$$

3] CIRCULAR TIME SHIFT PROPERTY:

If a discrete time signal is circularly shifted in time by m units then its DFT is multiplied by $e^{-j\frac{2\pi m k}{N}}$

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If $x(n) \xleftrightarrow{\text{DFT}} X(k)$ then
 $x(n-m)_N \xleftrightarrow{\text{DFT}} X(k) e^{-j2\pi km/N}$

$\text{DFT} \{x(n-m)\}_N = \sum_{n=0}^{N-1} x(n-m) e^{-j2\pi kn/N}$

but $p = n-m$
 $n = p+m$

$\text{DFT} \{x(n-m)\}_N = \sum_{n=0}^{N-1} x(p)_N e^{-j2\pi k(p+m)/N}$

$= \sum_{n=0}^{N-1} x(p)_N e^{-j2\pi kp/N} \cdot e^{-j2\pi km/N}$

$\text{DFT} \{x(n-m)\}_N = X(k) e^{-j2\pi km/N}$ // hence proved.

4) TIME REVERSAL:

Reversing N point Sequence in time is equivalent to reversing the DFT Sequence

If $\text{DFT} \{x(n)\} = X(k)$

$\text{DFT} \{x(N-n)\} = X(N-k)$ //

$\text{DFT} \{x(N-n)\} = \sum_{n=0}^{N-1} x(N-n) e^{-j2\pi nk/N}$

det

$m = n - N \quad m = N - n$
 $n = N - m$

$= \sum_{n=0}^{N-1} x(m) e^{-j2\pi k(n-M)/N}$

$\text{DFT} \{x(N-n)\} = \sum_{n=0}^{N-1} x(m) e^{-j2\pi km/N} e^{j2\pi km/N}$

$= \sum_{n=0}^{N-1} x(m) e^{-j2\pi km/N} e^{-j2\pi nm} \quad [e^{-j2\pi nm} = 1]$

$\text{DFT} \{x(N-n)\} = \sum_{n=0}^{N-1} x(m) e^{-j2\pi m(N-k)/N}$

$\therefore \text{DFT} \{x(N-n)\} = X(N-k)$ // hence proved.

5) Configuration:

Let $x(n)$ be a complex N point discrete sequence & $x^*(n)$ be its conjugate. Now if
 $\text{DFT} \{x(n)\} = X(k)$ then

$\text{DFT} \{x^*(n)\} = X^*(N-k)$

$\text{DFT} \{x^*(n)\} = \sum_{n=0}^{N-1} x^*(n) e^{-j2\pi nk/N}$

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$$\begin{aligned}
 &= \left[\sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi n k}{N}} \right] * e^{-j\frac{2\pi m}{N}} \\
 &= \left[\sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi n k}{N}} e^{-j\frac{2\pi m n}{N}} \right] * \\
 &= \left[\sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi n k}{N}} e^{-j\frac{2\pi m n}{N}} \right] * \\
 &= \left[\sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi n k(N-m)}{N}} \right] * \\
 &= [X(N-k)]^* \\
 &= X^*(N-k) // \text{hence proved.}
 \end{aligned}$$

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$$\begin{aligned}
 &= \sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi (k-m)n}{N}} \\
 &= X((k-m))_N // \text{hence proved}
 \end{aligned}$$

DTFT TO DFT :-

→ Sampling and reconstruction of discrete time signals:-
 W.K.T Aperiodic finite energy signals have continuous spectra

Let us consider Aperiodic discrete time sig $x(n)$ with Fourier transform. $X(w) = \sum_{n=-\infty}^{\infty} x(n) e^{-jwn}$

We will sample periodically in frequency domain and we will keep spacing bt 2 samples and Δw radians.

W.K.T $X(w)$ is periodic, period is 2π

If we take only samples of the fundamental freq range That is enough

Let us take N equidistant samples in the interval $0 \leq w \leq 2\pi$ with spacing $\Delta w = \frac{2\pi}{N}$

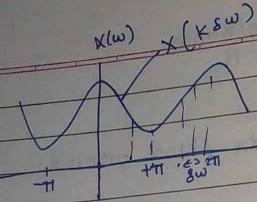
6) CIRCULAR FREQUENCY SHIFT PROPERTY:

The circular frequency shift property of DFT says that if a discrete time signal is multiplied by $e^{j\frac{2\pi m n}{N}}$ then $X((k-m))_N$

Proof:

→ If DFT $\{x(n)\} = X(k)$ then
 $DFT \{x(n) e^{j\frac{2\pi m n}{N}}\} = X((k-m))_N$

$$\begin{aligned}
 DFT \{x(n) e^{j\frac{2\pi m n}{N}}\} &= \sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi m n}{N}} e^{-j\frac{2\pi k n}{N}}
 \end{aligned}$$



Freq. domain sampling of fourier transform

Consider N number of samples in freq. domain

$$x(w) = \sum_{n=-\infty}^{\infty} x(n) e^{-jwn}$$

$$\text{Evaluate this @ } w = \frac{2\pi k}{N} - j \frac{2\pi kn}{N}$$

$$x(k) = x\left(\frac{2\pi k}{N}\right) = \sum_{n=-\infty}^{\infty} x(n) e^{-j \frac{2\pi kn}{N}}$$

↓ subdivids RHS summation

each should have only N terms

$$x\left(\frac{2\pi k}{N}\right) = \sum_{n=-N}^{-1} x(n) e^{-j \frac{2\pi kn}{N}} + \sum_{n=0}^{N-1} x(n) e^{-j \frac{2\pi kn}{N}} +$$

$$\sum_{n=N}^{2N-1} x(n) e^{-j \frac{2\pi kn}{N}} + \dots$$

$$x\left(\frac{2\pi k}{N}\right) = \sum_{l=-\infty}^{\infty} \sum_{n=LN}^{(L+1)N-1} x(n) e^{-j \frac{2\pi kn}{N}}$$

change $n \Rightarrow n-lN$ & interchange the summation

$$x\left(\frac{2\pi k}{N}\right) = \sum_{n=0}^{N-1} \sum_{l=-\infty}^{\infty} x(n-lN) e^{-j \frac{2\pi k(n-lN)}{N}}$$

$$\text{def } x_p(n) = \sum_{l=-\infty}^{\infty} x(n-lN)$$

fourier series expansion of $x_p(n)$ is

$$x_p(n) = \sum_{k=0}^{N-1} C_k e^{j \frac{2\pi kn}{N}}$$

C_k = fourier coefficient

$$C_k = \frac{1}{N} \sum_{n=0}^{N-1} x_p(n) e^{-j \frac{2\pi kn}{N}}$$

$$\therefore x_p(n) = \frac{1}{N} \sum_{k=0}^{N-1} x\left(\frac{2\pi k}{N}\right) e^{j \frac{2\pi nk}{N}}$$

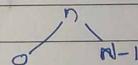
→ It provides reconstruction of periodic sig. $x_p(n)$ from $x(w)$

let Relationship bt $x_p(n)$ and $x(n)$.

for $x(n)$ $N \geq L \rightarrow$ no aliasing

$N < L \rightarrow$ aliasing

Conclusion: spectrum of an aperiodic discrete time signal with finite duration (L) can be exactly recovered from samples $w_p = \frac{2\pi k}{N}$ if $\underline{N \geq L}$



$$x_1(n) \circledast x_2(n) = \sum_{m=0}^{N-1} x_1(m) x_2((n-m))_N$$

CIRCULAR CONVOLUTION

property
If $\text{DFT}\{x_1(n)\} = X_1(k)$ & $\text{DFT}\{x_2(n)\} = X_2(k)$ then
 $\text{DFT}\{x_1(n) \circledast x_2(n)\} = X_1(k)X_2(k)$

proof: $X_1(k) = \sum_{n=0}^{N-1} x_1(n) e^{-j\frac{2\pi nk}{N}}$

$$X_1(k) = \sum_{m=0}^{N-1} x_1(m) e^{-j\frac{2\pi km}{N}}$$

$$X_2(k) = \sum_{n=0}^{N-1} x_2(n) e^{-j\frac{2\pi nk}{N}}$$

change $n \rightarrow p$

$$X_2(k) = \sum_{n=0}^{N-1} x_2(p) e^{-j\frac{2\pi pk}{N}}$$

$$\begin{aligned} \text{IDFT} : \text{DFT}^{-1}\{X_1(k)X_2(k)\} &= \frac{1}{N} \sum_{K=0}^{N-1} X_1(K)X_2(K)e^{j\frac{2\pi KnK}{N}} \\ &= \frac{1}{N} \sum_{K=0}^{N-1} \left[\sum_{m=0}^{N-1} x_1(m) e^{-j\frac{2\pi km}{N}} \right] \left[\sum_{n=0}^{N-1} x_2(n) e^{-j\frac{2\pi nk}{N}} \right] \\ &= \frac{1}{N} \sum_{m=0}^{N-1} x_1(m) \sum_{p=0}^{N-1} x_2(p) \sum_{K=0}^{N-1} e^{j\frac{2\pi k(n-m-p)}{N}} \end{aligned}$$

$\text{dft } n-m-p = qN \quad q \text{ is integer}$

$$\therefore \sum_{K=0}^{N-1} e^{\frac{j2\pi k(n-m-p)}{N}} = \sum_{K=0}^{N-1} e^{\frac{j2\pi kqN}{N}} = \sum_{K=0}^{N-1} (e^{j2\pi q})^k = \sum_{K=0}^{N-1} 1^k$$

$$\text{and } \sum_{p=0}^{N-1} x_2(p) = \sum_{m=0}^{N-1} x_2(n-m-qN) = \sum_{m=0}^{N-1} x_2(n-m), \text{ mod } N$$

$$= \sum_{m=0}^{N-1} x_2((n-m))_N$$

$$\begin{aligned} \therefore \text{DFT}\{X_1(k)X_2(k)\} &= \frac{1}{N} \sum_{m=0}^{N-1} x_1(m) \sum_{n=0}^{N-1} x_2((n-m))_N \\ &= \sum_{m=0}^{N-1} x_1(m) x_2((n-m))_N \\ &= x_1(n) * x_2(n) \end{aligned}$$

$$\therefore X_1(k)X_2(k) = \text{DFT}\{x_1(n) \circledast x_2(n)\} \parallel.$$

Multiplication Property of DFT:

If $DFT\{x_1(n)\} = X_1(k)$ & then
 $DFT\{x_1(n)x_2(n)\} = \frac{1}{N} [X_1(k) \otimes X_2(k)]$

$$\begin{aligned} x_1(n) &= \frac{1}{N} \sum_{k=0}^{N-1} X_1(k) e^{j2\pi kn/N} = \frac{1}{N} \sum_{m=0}^{N-1} X_1(m) e^{-j2\pi km/N} \quad \because k=m \\ DFT\{x_1(n)x_2(n)\} &= \sum_{n=0}^{N-1} x_1(n)x_2(n) e^{-j2\pi kn/N} \\ &= \sum_{n=0}^{N-1} \left[\sum_{m=0}^{N-1} X_1(m) e^{-j2\pi mn/N} \right] \left[x_2(n) e^{-j2\pi kn/N} \right] \\ &= \frac{1}{N} \sum_{m=0}^{N-1} X_1(m) \left[\sum_{n=0}^{N-1} x_2(n) e^{-j2\pi n(K-m)/N} \right] \\ &= \frac{1}{N} \sum_{m=0}^{N-1} X_1(m) X_2((K-m))_N = \frac{1}{N} [X_1(k) \otimes X_2(k)] \end{aligned}$$

$$\therefore DFT\{x_1(n)x_2(n)\} = \frac{1}{N} [X_1(k) \otimes X_2(k)] //$$

hence proved

CIRCULAR CORRELATION:

If we have 2 Nlp sigs $x(n)$ and $y(n)$ then my circular correlatn property

$$\bar{x}y(m) = \sum_{n=0}^{N-1} x(n)y^*((n-m))_N$$

If $DFT\{x(n)\} = X(k)$ & $DFT\{y(n)\} = Y(k)$
By correlation property

$$DFT\{\bar{x}y(m)\} = DFT\left\{\sum_{n=0}^{N-1} x(n)y^*((n-m))_N\right\}$$

PROOF: $X(k) = \sum_{n=0}^{N-1} x(n) e^{-j2\pi nk/N}$ and $Y(k) = \sum_{n=0}^{N-1} y(n) e^{-j2\pi nk/N}$
 $y(k) = \sum_{p=0}^{N-1} y(p) e^{-j2\pi pk/N}$

$$\begin{aligned} DFT\{x(k)y^*(k)\} &= \frac{1}{N} \sum_{K=0}^{N-1} x(k)y^*(k)e^{-j2\pi Kk/N} \\ &= \frac{1}{N} \sum_{K=0}^{N-1} \left[\sum_{n=0}^{N-1} x(n) e^{-j2\pi nk/N} \right] \left[\sum_{p=0}^{N-1} y(p) e^{-j2\pi pk/N} \right]^* e^{-j2\pi Kk/N} \end{aligned}$$

$$= \frac{1}{N} \sum_{n=0}^{N-1} \sum_{p=0}^{N-1} \sum_{K=0}^{N-1} x(n) y^*(p) e^{-j2\pi K(m-n+p)/N}$$

Let us consider this part.

consider $\sum_{k=0}^{N-1} e^{j2\pi k(m-n+p)}$, but $m-n+p = \frac{q}{N}$ integer

$$= \sum_{k=0}^{N-1} e^{j2\pi k q/N} - \sum_{k=0}^{N-1} e^{j2\pi q k}$$

$$= \sum_{k=0}^{N-1} 1^k = N$$

c. but us take $\sum_{p=0}^{N-1} y^*(p)$ from that eqn

$$\text{but } m-n+p = qN \\ p = n-m+qN$$

$$\therefore \sum_{p=0}^{N-1} y^*(p) = \sum_{n=0}^{N-1} y^*(n-m+qN)$$

$$= \sum_{n=0}^{N-1} y^*(n-m), \text{ mod } N$$

$$\sum_{p=0}^{N-1} y^*(p) = \sum_{n=0}^{N-1} y^*((n-m))_N$$

$$\therefore \text{DFT} \{x(k)y^*(k)\} = \frac{1}{N} \sum_{n=0}^{N-1} x(n) \sum_{n=0}^{N-1} y^*((n-m))_N \cdot N$$

$$= \sum_{n=0}^{N-1} x(n) y^*((n-m))_N$$

$$= \bar{r}_{xy}(m)$$

$$\therefore X(k)y^*(k) = \text{DFT} \{ \bar{r}_{xy}(m) \}.$$

PARSEVAL'S THEOREM (energy)

- If $\text{DFT} \{x_1(n)\} = X_1(k)$ & $\text{DFT} \{x_2(n)\} = X_2(k)$ then

$$\sum_{n=0}^{N-1} x_1(n)x_2^*(n) = \frac{1}{N} \sum_{k=0}^{N-1} X_1(k)X_2^*(k)$$

proof: $X_1(k) = \sum_{n=0}^{N-1} x_1(n)e^{-j2\pi nk/N}$

$$x_2(n) = \sum_{k=0}^{N-1} X_2(k)e^{j2\pi nk/N}$$

$$\sum_{k=0}^{N-1} X_1(k)X_2^*(k) = \sum_{k=0}^{N-1} \left[\sum_{n=0}^{N-1} x_1(n)e^{-j2\pi nk/N} \right] X_2^*(k)$$

Rearrange = $\sum_{n=0}^{N-1} x_1(n) \left[\sum_{k=0}^{N-1} X_2^*(k)e^{-j2\pi nk/N} \right]$

$$= \sum_{n=0}^{N-1} x_1(n) \underbrace{\left[\sum_{k=0}^{N-1} X_2(k)e^{-j2\pi nk/N} \right]}_{\text{conjugate}}^*$$

$$\sum_{k=0}^{N-1} X_1(k)X_2^*(k) = \sum_{n=0}^{N-1} x_1(n)x_2^*(n) // \text{proved}$$

Problems on Circular Convolution

1) Time domain (matrix method)

If given $x(n)$ and $h(n)$.

Let us use matrix method of convolution.

If $h(n) = \{h(0), h(1), h(2), h(3)\}$ and

$x(n) = \{x(0), x(1), x(2), x(3)\}$

$$\begin{array}{c} \downarrow \\ \begin{bmatrix} h(0) & h(3) & h(2) & h(1) & | & x(0) \\ h(1) & h(0) & h(3) & h(2) & | & x(1) \\ h(2) & h(1) & h(0) & h(3) & | & x(2) \\ h(3) & h(2) & h(1) & h(0) & | & x(3) \end{bmatrix} = \begin{bmatrix} x(0)h(0) + x(1)h(3) + \dots \\ x(1)h(1) + x(2)h(2) + \dots \\ x(2)h(2) + x(3)h(1) + \dots \\ x(3)h(3) + x(0)h(2) + \dots \end{bmatrix} \\ = \begin{bmatrix} y(0) \\ y(1) \\ y(2) \\ y(3) \end{bmatrix} \end{array}$$

$$y(n) = \{y(0), y(1), y(2), y(3)\}$$

2) Frequency domain $x(n) \& h(n)$ (matrix method)

Step 1: Find $X(k)$ and $H(k)$

" " $\Rightarrow X(k)H(k)$ (element by element mul.)

& say $Y(k) = X(k)H(k)$

Step 3: Take IDFT of $Y(k)$ to obtain $y(n)$.

Ex:

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$$X(k) = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} \begin{bmatrix} x(0) \\ x(1) \\ x(2) \\ x(3) \end{bmatrix} = \begin{bmatrix} X(0) \\ X(1) \\ X(2) \\ X(3) \end{bmatrix}$$

$$H(k) = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} \begin{bmatrix} h(0) \\ h(1) \\ h(2) \\ h(3) \end{bmatrix} = \begin{bmatrix} H(0) \\ H(1) \\ H(2) \\ H(3) \end{bmatrix}$$

$$Y(k) = \begin{bmatrix} X(0) \\ X(1) \\ X(2) \\ X(3) \end{bmatrix} \begin{bmatrix} H(0) \\ H(1) \\ H(2) \\ H(3) \end{bmatrix} = \begin{bmatrix} Y(0) \\ Y(1) \\ Y(2) \\ Y(3) \end{bmatrix}$$

$$Y(n) = IDFT(Y(k)) = \frac{1}{N} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & 1 & j \end{bmatrix} \begin{bmatrix} Y(0) \\ Y(1) \\ Y(2) \\ Y(3) \end{bmatrix} = \begin{bmatrix} y(0) \\ y(1) \\ y(2) \\ y(3) \end{bmatrix}$$

here $N=4$

$$y(n) = \{y(0), y(1), y(2), y(3)\}$$

Both by time domain and by frequency domain $y(n)$ should be same. (Verification).

Problems on Circular Convolution

1) Time domain (matrix method)

If given $x(n)$ and $h(n)$.

Let us use matrix method of convolution.

If $h(n) = \{h(0), h(1), h(2), h(3)\}$ and

$x(n) = \{x(0), x(1), x(2), x(3)\}$

↓

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

$$= \begin{bmatrix} y(0) \\ y(1) \\ y(2) \\ y(3) \end{bmatrix}$$

$$y(n) = \{y(0), y(1), y(2), y(3)\}$$

2) Frequency domain $x(n) * h(n)$ (matrix method)

Step 1: Find $X(k)$ and $H(k)$

" a: $X(k)H(k)$ (element by element mul)

& say $Y(k) = X(k)H(k)$

Step 3: Take IDFT of $Y(k)$ to obtain $y(n)$.

Ex:

$$X(k) = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} \begin{bmatrix} x(0) \\ x(1) \\ x(2) \\ x(3) \end{bmatrix} = \begin{bmatrix} X(0) \\ X(1) \\ X(2) \\ X(3) \end{bmatrix}$$

$$H(k) = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -j & -1 & j \\ 1 & -1 & 1 & -1 \\ 1 & j & -1 & -j \end{bmatrix} \begin{bmatrix} h(0) \\ h(1) \\ h(2) \\ h(3) \end{bmatrix} = \begin{bmatrix} H(0) \\ H(1) \\ H(2) \\ H(3) \end{bmatrix}$$

$$Y(k) = \begin{bmatrix} X(0) \\ X(1) \\ X(2) \\ X(3) \end{bmatrix} \begin{bmatrix} H(0) \\ H(1) \\ H(2) \\ H(3) \end{bmatrix} = \begin{bmatrix} Y(0) \\ Y(1) \\ Y(2) \\ Y(3) \end{bmatrix}$$

$$y(n) = IDFT(Y(k)) = \frac{1}{N} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & j & -1 & -j \\ 1 & -1 & 1 & -1 \\ 1 & -j & 1 & j \end{bmatrix} \begin{bmatrix} Y(0) \\ Y(1) \\ Y(2) \\ Y(3) \end{bmatrix} = \begin{bmatrix} y(0) \\ y(1) \\ y(2) \\ y(3) \end{bmatrix}$$

Here $N=4$

$$y(n) = \{y(0), y(1), y(2), y(3)\}$$

Both by time domain and by frequency domain
 $y(n)$ should be same. (Verification).

OVERLAP ADD METHOD:

↳ for finding linear convolution, segmenting given sequence

$$\text{Suppose } x(n) = \{x(0), x(1), x(2), x(3), x(4), x(5), x(6), x(7), x(8)\}$$

$x(n)\}$

$$h(n) = \{h(0), h(1), h(2)\} \rightarrow M=3 \quad N=8$$

choose segment length $L_s = 5$ (we can vary)

↳

$$1^{\text{st}} \text{ Segment } [x(0), x(1), x(2), 0, 0]$$

$$2^{\text{nd}} \text{ Segment } [x(3), x(4), x(5), 0, 0]$$

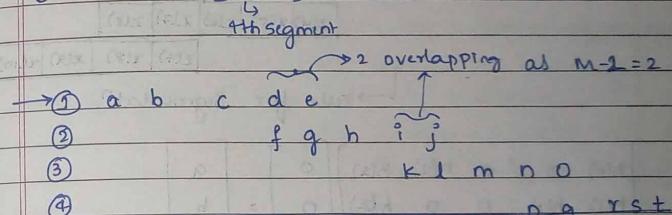
$$3^{\text{rd}} \text{ " } [x(6), x(7), x(8), 0, 0]$$

$$4^{\text{th}} \text{ " } [x(9), 0, 0, 0, 0]$$

$$\begin{array}{c|ccccc} & \multicolumn{5}{c}{1^{\text{st}} \text{ segment}} \\ \hline h(0) & 0 & 0 & h(2) & h(1) & x(0) & a \\ h(1) & h(0) & 0 & 0 & h(2) & x(1) & b \\ h(2) & h(1) & h(0) & 0 & 0 & x(2) & c \\ 0 & h(2) & h(1) & h(0) & 0 & 0 & d \\ 0 & 0 & h(2) & h(1) & h(0) & 0 & e \end{array}$$

$$\begin{array}{c|ccccc} & \multicolumn{5}{c}{2^{\text{nd}} \text{ segment}} \\ \hline h(0) & 0 & 0 & h(2) & h(1) & x(3) & f \\ h(1) & h(0) & 0 & 0 & h(2) & x(4) & g \\ h(2) & h(1) & h(0) & 0 & 0 & x(5) & h \\ 0 & h(2) & h(1) & h(0) & 0 & 0 & i \\ 0 & 0 & h(2) & h(1) & h(0) & 0 & j \end{array}$$

3rd segment		Date: / /	Page No.:
$x(6)$	k		
$x(7)$	t	(3)	
$x(8)$	m		
0	n		
0	0		
$x(9)$	p		
0	q	(4)	
0	r		
0	s		
0	t		



$$y(n) = \{a, b, c, (d+e), (e+f), h, (i+k), (j+l), m, (n+p), (o+q), r, s, t\}$$

OVERLAP SAVE METHOD: (Segmented method)

↳ To find linear convolution.

$$x(n) = \{x(0), x(1), x(2), x(3), x(4), x(5), x(6)\}$$

$$h(n) = \{h(0), h(1), h(2)\} \rightarrow M=3,$$

$$M-1 = 2.$$

$$y(n) = ?$$

YPOX

M-1

0	0	x(0)	x(1)	x(2)	
zeros		x(0)	x(1)	x(2)	
		x(1)	x(2)	x(3)	x(4)
		x(2)	x(3)	x(4)	x(5)
		x(3)	x(4)	x(5)	x(6)
		x(4)	x(5)	x(6)	x(7)
		x(5)	x(6)	x(7)	x(8)
		x(6)	x(7)	x(8)	x(9)
		x(7)	x(8)	x(9)	x(10)
					0

↳ rule for Segmentation.

$$\begin{bmatrix} h(0) & 0 & 0 & h(2) & h(4) \\ h(1) & h(0) & 0 & 0 & h(2) \\ h(2) & h(1) & h(0) & 0 & 0 \\ 0 & h(2) & h(1) & h(0) & 0 \\ 0 & 0 & h(2) & h(1) & h(0) \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ x(0) \\ x(1) \\ x(2) \end{bmatrix} = \begin{bmatrix} a \\ b \\ c \\ d \\ e \end{bmatrix}$$

h(0)
same as
previous
segment

$$\begin{bmatrix} x(1) \\ x(2) \\ x(3) \\ x(4) \\ x(5) \end{bmatrix} = \begin{bmatrix} f \\ g \\ h \\ i \\ j \end{bmatrix}$$

" " mul are 4×4] " Date: 11/11/2022 Page No.:

same as 1st	x(4)	k
" one segment	x(5)	l
	x(6)	m
	x(7)	n
	x(8)	o
	x(9)	p
same as	x(10)	q
1st segment	x(11)	r
	x(12)	s
	x(13)	t

as $M-1=2$, discard 1st 2 element of all matrix
i.e. $\{a, b, f, g, k, l, p, q\}$
 $y(n) = \{c, d, e, h, i, j, m, n, o, r, s, t\}$
 $y(n) = \{y(0), y(1), y(2), y(3), y(4), y(5), y(6), y(7), y(8), y(9), y(10), y(11)\}$

RELATIONSHIP OF DFT TO OTHER TRANSFORMS

1] DTFT to DFT

By defn
 \Rightarrow DTFT is $X(e^{j\omega}) = \sum_{n=0}^{N-1} x(n)e^{-j\omega n}$ $n = 0, 1, \dots, N-1$

DFT is $X(k) = \sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi}{N}nk}$

$X(k) = X(\omega) @ \omega = \frac{2\pi k}{N}, k = 0, 1, 2, 3, \dots, N-1$

\therefore If we want to find DFT, we have to evaluate
 DTFT @ $\omega = \frac{2\pi k}{N}$

2] Z Transform to DFT:

$\Rightarrow X(z) = \sum_{n=0}^{N-1} x(n) z^{-n} \rightarrow ①$

DFT $X(k) = \sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi}{N}nk} \rightarrow ②$

from ① & ②
 w.k.t $|e^{-j\frac{2\pi}{N}k}| = 1$ $|e^{-j\frac{2\pi}{N}n}| = 2\pi k$

\therefore On comparing 1 & 2

$\therefore z = e^{j\frac{2\pi}{N}k}$

unit circle

$$X(k) = X(z) @ z = e^{j\frac{2\pi}{N}k}$$

\therefore On unit circle evaluation is performed, N values are equally spaced

3] FS to DFT:

w.k.t
 $x_p(n) = \sum_{k=0}^{N-1} C_k e^{-j\frac{2\pi}{N}nk} \rightarrow$ periodic sig

where C_k are fourier coeff

$$C_k = \frac{1}{N} \sum_{n=0}^{N-1} x_p(n) e^{-j\frac{2\pi}{N}nk} \rightarrow ③$$

$$x(n) = x_p(n) \rightarrow 0 \leq n \leq N-1$$

$$= 0 \rightarrow \text{otherwise}$$

because of this relationship

$$C_k = \frac{1}{N} X(k) \rightarrow ④$$

$$X(k) = \sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi}{N}nk}$$

compare eq 3 & 4
 $\rightarrow X(k) = N C_k //$

CHAPTER - 2

FAST FOURIER TRANSFORM (FFT) ALGORITHMS

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→ In this we have 2 algorithms:

DTT : Decimation in time

DIF : Decimation in frequency

→ FFT is tool to find discrete fourier transform (DFT) efficiently

→ In this the symmetry property of the twiddle factors is used

* Symmetry property is:

$$W_N^{K+N/2} = -W_N^K$$

* Periodicity property is:

$$W_N^{K+N} = W_N^K \quad W_N = e^{-j\frac{2\pi}{N}}$$

→ In FFT, the number of complex multiplications are

$$\frac{N}{2} \log_2 N$$

→ In FFT, the number of complex additions are

$$N \log_2 N$$

$$2 \log_2 N$$

2 :

DIT (Decimation In Time); FFT algorithm

→ We use following properties to do DIT

i) Symmetry:

$$W_N^{K+N/2} = -W_N^K$$

ii) Periodicity

$$W_N^{K+N} = W_N^K$$

Radix 2 DITFFT Algorithm

Ex: 4, 8, 16, ... 2048

In case of DITFFT

$x(n)$
even odd

$$x_{e(n)} = x(2n)$$

$$x_{o(n)} = x(2n+1)$$

→ Derivation of DITFFT

$$\text{W.K.T } X(K) = \sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi n K}{N}} = \sum_{n=0}^{N-1} x(n) W_N^{nk},$$

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

$$= \sum_{n=0}^{N-1} x(2n) W_N^{2nk} + \sum_{n=0}^{N-1} x(2n+1) W_N^{2nk}$$

$$X(K) = \sum_{n=0}^{\frac{N}{2}-1} x(2n) W_N^{2nk} + W_N^{kN} \sum_{n=0}^{\frac{N}{2}-1} x(2n+1) W_N^{2nk}$$

Suppose $N=4$, $w_4^0, w_4^1, w_4^2, w_4^3$

$$w_N^2 = \left(e^{-j\frac{\pi}{N}}\right)^2 = w_{N/2}^2$$

$$x(k) = \sum_{n=0}^{N-1} x_e(n) w_{N/2}^{nk} + w_{N/2}^{nk} \sum_{n=0}^{N-1} x_o(n) w_{N/2}^{nk}$$

$$x(k) = x_e(k) + w_N^k x_o(k)$$

Let us symbolise $x(k)$ as

$$x(k) = g(k) + w_N^k h(k)$$

$$x(k) = g(k) + w_N^k h(k)$$

$$x(k) = g\left(\frac{k+N}{2}\right) + w_N^{k+N/2} h\left(\frac{k+N}{2}\right)$$

Break $g(k)$ & $h(k)$ into $\frac{N}{2}$ pt (Decimation)

$$g(k) = \sum_{r=0}^{\frac{N}{2}-1} g(r) w_{N/2}^{rk}$$

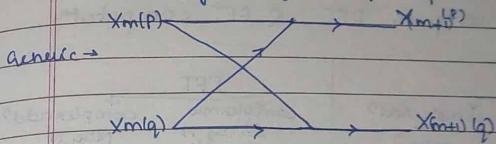
$$g(k) = \sum_{l=0}^{\frac{N}{2}-1} g(2l) w_{N/4}^{rk} + w_{N/2}^{rk} \sum_{l=0}^{\frac{N}{2}-1} g(2l+1) w_{N/4}^{rk}$$

similarly

$$h(k) = \sum_{r=0}^{\frac{N}{2}-1} h(r) w_{N/2}^{rk}$$

$$h(k) = \sum_{r=0}^{\frac{N}{2}-1} h(2r) w_{N/4}^{rk} + w_{N/2}^{rk} \sum_{l=0}^{\frac{N}{2}-1} h(2r+1) w_{N/4}^{rk}$$

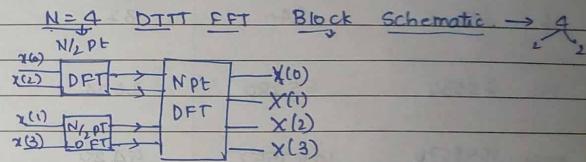
Butterfly diagram for previous eqn



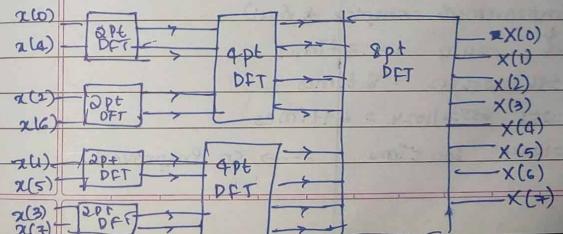
$x_{m+1}(p) \quad x_{m+1}(q)$

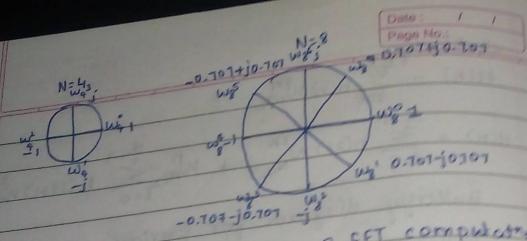
$$x_{m+1}(p) = x_m(p) + x_m(q)$$

$$x_{m+1}(q) = [x_m(p) - x_m(q)] w_N^q$$



for $N=8$ DIT FFT Block schematic





Now we compare DFT & FFT computations.

	DFT	FFT
N	complex mult N ^(*)	complex add? N(N-1)
8	64	52
16	256	240
256	65536	65280
1024	1048576	1047552
		5420
		10240

$$\text{Complex mult } \frac{N}{2} \log_2 N$$

$$\text{Complex add? } N \log_2 N$$

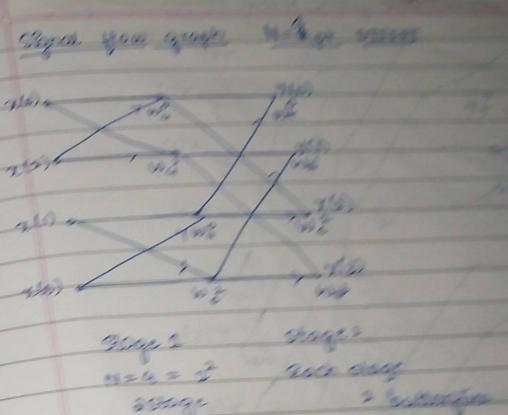
Speed improvement complex \rightarrow final

$$N=8, 64/10 = 5.3 \text{ times}$$

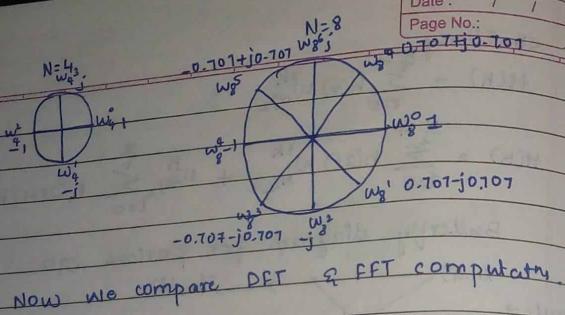
$$N=16, 256/32 = 8 \text{ times}$$

$$N=256, 65536/1024 = 64 \text{ times}$$

$$N=1024, 1048576 \rightarrow \text{so improved}$$



Stage 1: $w_8^{4j}, w_8^4, w_8^{-4}, w_8^{-4j}$
 Stage 2: $w_8^4 = 2^2$
 Stage 3: 2nd stage
 Stage 4: 2 butterflies



Now we compare DFT & FFT computation.

N	DFT		FFT	
	complex mul \downarrow N^2	complex add? \downarrow $N(N-1)$	complex mul \downarrow $\frac{N}{2} \log_2 N$	complex add? \downarrow $N \log_2 N$
8	64	52	12	24
16	256	240	32	64
256	65536	65280	1024	2048
1024	1048576	1047552	5120	10240

Speed improvement complex * (mul)

$$N=8, \quad 64/12 = 5.3 \text{ times}$$

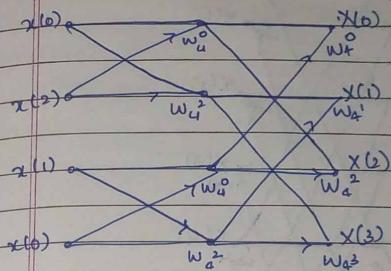
$$N=16, \quad 256/32 = 8 \text{ times}$$

$$N=256, \quad 65536/1024 = 64 \text{ times}$$

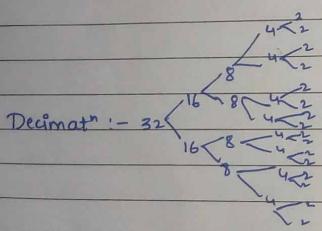
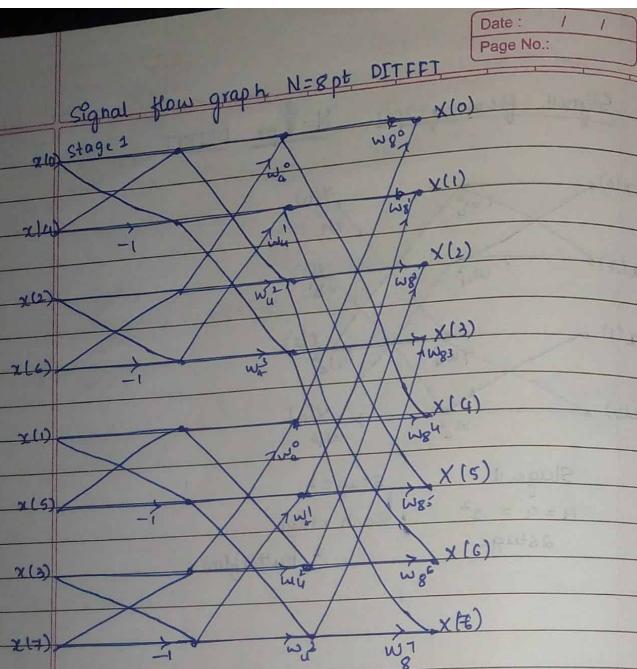
$$N=1024, \quad 1048576/1024 = 1024 \text{ times} \rightarrow 50 \text{ improved.}$$

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Signal flow graph $N = 8$ pt DITFFT



Stage 1
 $N = 4 = 2^2$
Each stage
2 stage
2 butterflies



Decimation in Frequency (DIF) FFT Algorithm

Input values are in order

Output values are bit reversed.

→ Derivatⁿ

$$\text{By defn } X(k) = \sum_{n=0}^{N-1} x(n) W_N^{kn}$$

$$X(k) = \sum_{n=0}^{\frac{N}{2}-1} x(n) W_N^{kn} + \sum_{n=\frac{N}{2}}^{N-1} x(n) W_N^{kn}$$

$$\text{let } r = n - \frac{N}{2}$$

$$X(k) = \sum_{n=0}^{\frac{N}{2}-1} x(n) W_N^{kn} + \sum_{r=0}^{\frac{N}{2}-1} x(r + \frac{N}{2}) W_N^{k(r+N/2)}$$

r is dummy variable $\therefore r$ is replaced by n

$$\therefore X(k) = \sum_{n=0}^{\frac{N}{2}-1} x(n) W_N^{kn} + \sum_{n=0}^{\frac{N}{2}-1} x(n + \frac{N}{2}) W_N^{k(n+N/2)}$$

$$W_N^{kn} W_N^{kN/2} = (-1)^k$$

$$X(k) = \sum_{n=0}^{\frac{N}{2}-1} [x(n) + (-1)^k x(n + \frac{N}{2})] W_N^{2kn}$$

even $k = 2r$, $r = 0, 1, 2, \dots, \frac{N}{2}-1$

$$X(2r) = \sum_{n=0}^{\frac{N}{2}-1} [x(n) + (-1)^{2r} x(n + \frac{N}{2})] W_N^{2rn}$$

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$$= \sum_{n=0}^{\frac{N}{2}-1} [x(n) + x\left(n + \frac{N}{2}\right)] W_N^{rn}$$

odd $k = 2r+1, r = 0, 1, \dots, \frac{N}{2}-1$

$$X(2r+1) = \sum_{n=0}^{\frac{N}{2}-1} [x(n) + (-1)^{2r+1} x\left(n + \frac{N}{2}\right)] W_N^r W_N^{2rn}$$

even $p(n) = x(n) + x\left(n + \frac{N}{2}\right)$

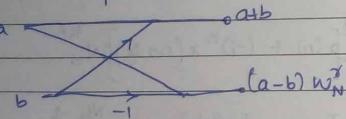
$$\text{odd } o(n) = [x(n) - x\left(n + \frac{N}{2}\right)] W_N^{rn}$$

$$\therefore X(2r) = \sum_{n=0}^{\frac{N}{2}-1} o(n) W_N^{rn}$$

$$X(2r+1) = \sum_{n=0}^{\frac{N}{2}-1} o(n) W_N^{rn} \quad r = 0, 1, \dots, N/2-1$$

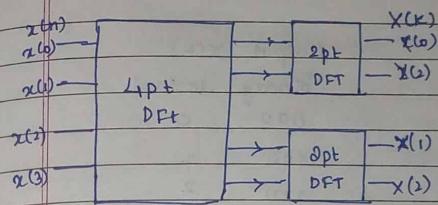
DIFFFT Butterfly pattern

Generic pattern is

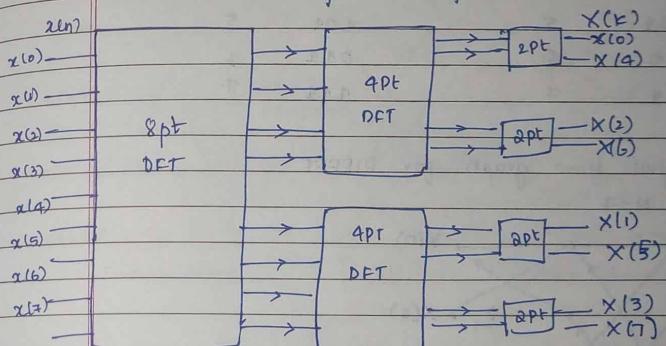


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Block Schematic for $N = 4$ pt DIFFFT



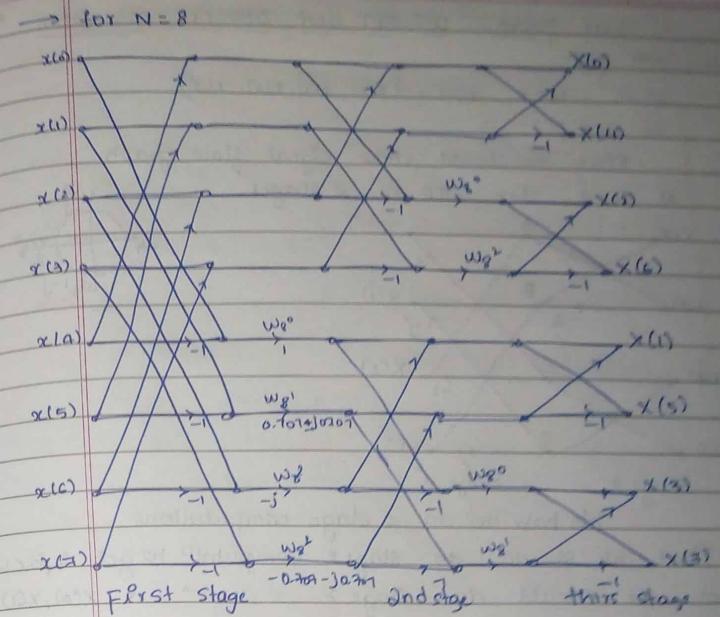
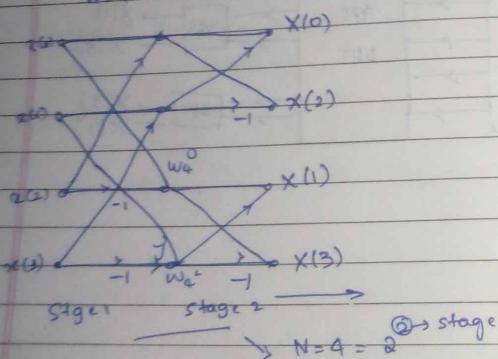
Block Schematic for $N = 8$ pt DIFFFT



In place computation
Bit Reversed pattern

Input $x(n)$	Output $X(k)$
Binary decimal	Binary decimal
000 0	000 0
001 1	100 4
010 2	010 2
011 3	110 6
100 4	001 2
101 5	101 5
110 6	011 3
111 7	111 7

Signal flow graph for DIFFER
 $N=4$



$N = 8 = 2^3 \rightarrow$ stage

In each stage $N/2$ butterflies will be per stage

Computational complexity

$\frac{N \log N}{2} \rightarrow$ Complex multiplications

$N \log N \rightarrow$ Complex additions

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→ Four Point DIT FFT and DIF FFT Algorithm

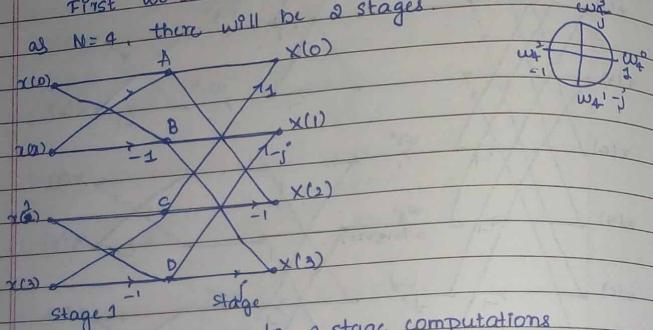
→ DIT FFT

Given, $N = 4$, $x(n) = \{x(0), x(1), x(2), x(3)\}$

To solve,

First we should draw signal flow graph.

as $N = 4$, there will be 2 stages.



We have to do 2 stage computations.

First we should do stage 1 computation to get A, B, C, D

then we should do stage 2 " " $\{x(0), x(1), x(2), x(3)\}$

$$A = x(0) + x(3)$$

$$B = x(0) - x(3)$$

$$C = x(1) + x(2)$$

$$D = x(1) - x(2)$$

$$x(0) = A + C$$

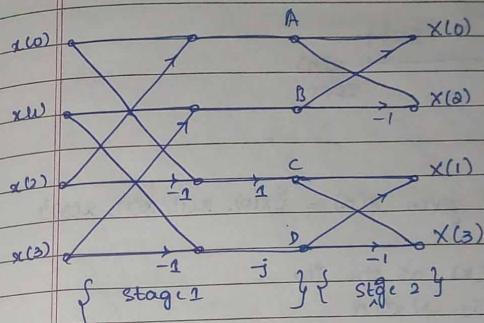
$$x(1) = B + D(-j)$$

$$x(2) = A - C$$

$$x(3) = B + JD$$

$$x(k) = \{x(0), x(1), x(2), x(3)\}$$

→ DIF FFT + pbl



$$A = x(2) + x(0)$$

$$B = x(3) + x(1)$$

$$C = x(0) - x(2)$$

$$D = x(1) - x(3)(-j)$$

$$x(0) = A + B$$

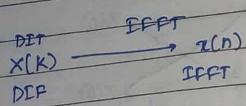
$$x(2) = A - B$$

$$x(1) = C + D$$

$$x(3) = C - D$$

$$x(k) = \{x(0), x(1), x(2), x(3)\}_{II}$$

Four Point DIT FFT and Four Point DIF IFFT Algorithm :



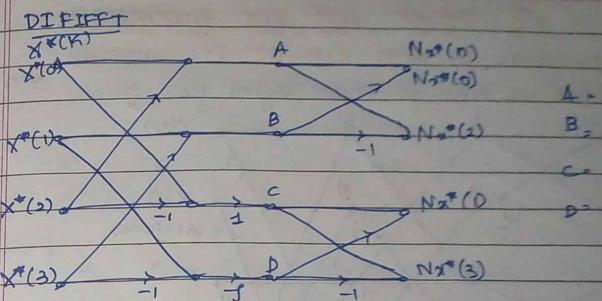
DIT IFFT :
It will be given $x(k) = \{x(0), x(1), x(2), x(3)\}$

Procedure:

- Take $X^*(k)$ as the input
 - Output is $Nx^*(n)$
-
- $X^*(0)$ A $Nx^*(0)$
 $X^*(2)$ B $Nx^*(1)$
 $X^*(1)$ C $Nx^*(2)$
 $X^*(3)$ D $Nx^*(3)$
- $$A = X^*(0) + X^*(2)$$
- $$B = X^*(0) - X^*(2)$$
- $$C = X^*(1) + X^*(3)$$
- $$D = X^*(1) - X^*(3)$$

- Take output $x(n) \rightarrow$ complex conjugate & divide by N to get $x(n)$

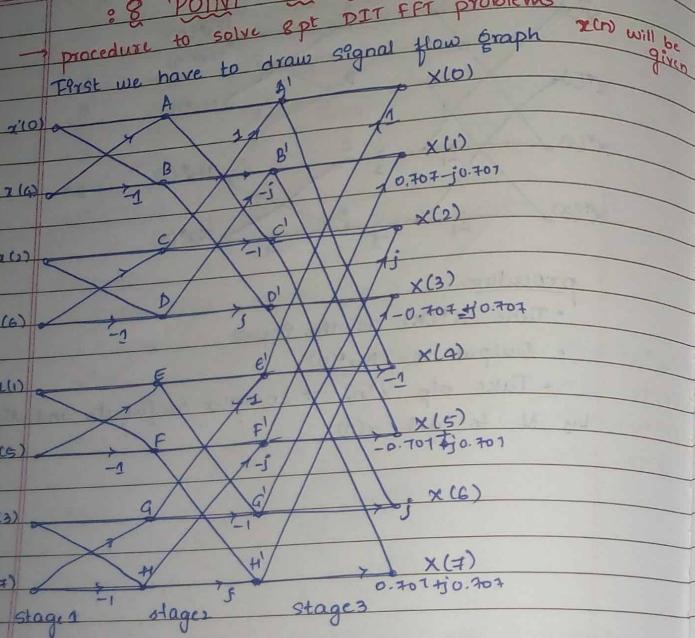
IFFT : inverse fast fourier transform



Procedure:

- Take $X^*(k)$ as the input
- Output is $Nx^*(n)$
- Take output $x(n) \rightarrow$ complex conjugate and divide by N to get $x(n)$.

8 POINT DIT FFT ALGORITHM



→ stage 2 calc's:

$$\begin{aligned} A' &= A + C(1) & F' &= F + H(-j) \\ B' &= B + D(-j) & G' &= G'E + G(-1) \\ C' &= C A + C(-1) & H' &= F + H(j) \\ D' &= B + D(j) \\ E' &= F + G \end{aligned}$$

→ stage 3 calc's

$$\begin{aligned} x(0) &= A' + E' & x(4) &= E' - A' \\ x(1) &= B' + F'(-0.707 - j0.707) & x(5) &= F'(-0.707 + j0.707) + B' \\ x(2) &= C' + G'(j) & x(6) &= G(j) + C' \\ x(3) &= D' + H'(-0.707 - j0.707) & x(7) &= D' + H'(-0.707 + j0.707) \end{aligned}$$

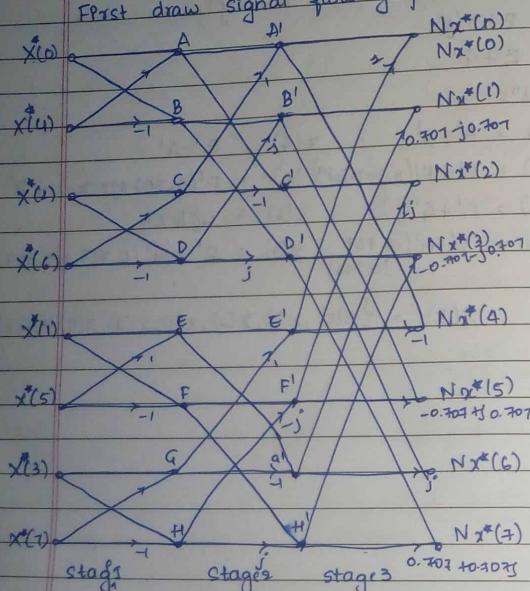
$$x(k) = \{x(0), x(1), x(2), x(3), x(4), x(5), x(6), x(7)\}$$

→ Stage 1 calc's:
 $A = x(0) + x(4)$ $F = x(1) - x(5)$
 $B = x(0) - x(4)$ $G = x(3) + x(7)$
 $C = x(2) + x(6)$ $H = x(5) - x(1)$
 $D = x(2) - x(6)$
 $E = x(1) + x(5)$

8 POINT DIT FFT ALGORITHM:

→ procedure to solve 8 point DIT FFT problems
 $X(k)$ will be given, we will should find $x(n)$.

First draw signal flow graph.

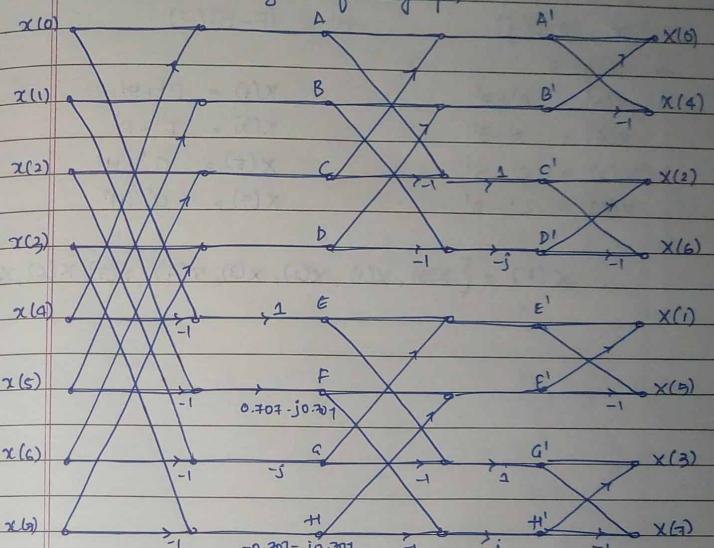


stage 1
procedure

- 1) Take $X^*(k)$ as input
- 2) Find A, B, C, D, E, F, G, H
- 3) Find $A', B', C', D', E', F', G', H'$
- 4) Find $\alpha_{lpt} N x^*(n) \rightarrow$ Find $x(n)$ by taking complex conjugate & divide by $N(8)$.

8 POINT DIF FFT ALGORITHM:

→ Procedure to solve 8pt DIF FFT problems
 First draw signal flow graph



→ Find A, B, C, D, E, F, G, H

→ " $A', B', C', D', E', F', G', H'$

$$\begin{aligned} A &= x(0) + x(4) & E &= [x(0) - x(4)] \\ B &= x(1) + x(5) & F &= [x(1) - x(5)] [0.707 - j 0.707] \\ C &= x(2) + x(6) & G &= [x(2) - x(6)] - j \\ D &= x(3) + x(7) & H &= [x(3) - x(7)] [-0.707 - 0j 0.707] \end{aligned}$$

Stage 2

$$A' = A + C$$

$$B' = B + D$$

$$C' = A - C$$

$$D' = (B - D)(-j)$$

$$E' = E + G$$

$$F' = F + H$$

$$G' = E - G$$

$$H' = (F - H)(-j)$$

→ Stage 3

$$X(0) = A' + B'$$

$$X(4) = A' - B'$$

$$X(2) = C' + D'$$

$$X(6) = C' - D'$$

$$X(1) = E' + F'$$

$$X(3) = E' - F'$$

$$X(5) = G' + H'$$

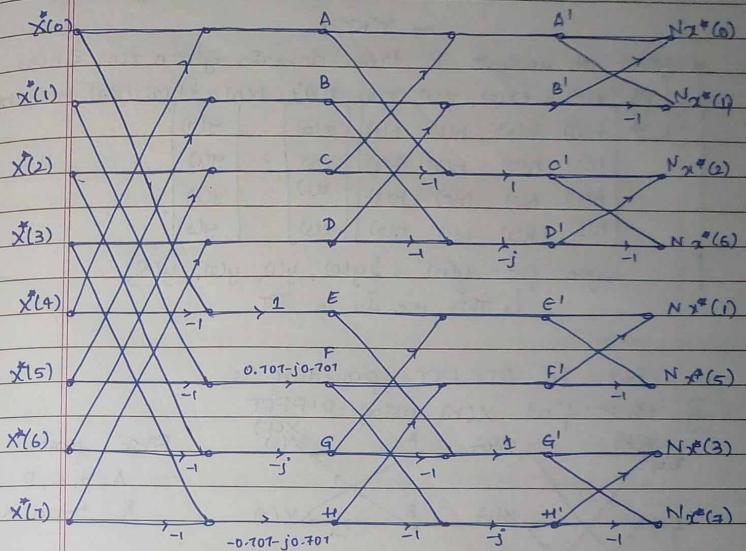
$$X(7) = G' - H'$$

$$X(k) = \{X(0), X(1), X(2), X(3), X(4), X(5), X(6), X(7)\}$$

8 POINT DIF FFT ALGORITHM:

→ Procedure to solve 8 point DIF FFT problems:

$$X(k) \rightarrow x(n)$$



→ Take $X^*(k)$ as PIP

→ Find $A, B, C, D, E, F, G, H \& A', B', C', D', E', F', G', H'$

→ Find Olp $Nx^*(n) \rightarrow$ Find Olp $x(n)$ by find complex conjugating and dividing by $N (= 8)$

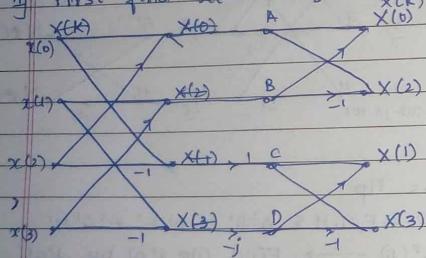
$$x(n) = \{x(0), x(1), x(2), x(3), x(4), x(5), x(6), x(7)\}$$

CIRCULAR CONVOLUTION USING FFT & IFFT

→ Circular convolution using DIF FFT
→ $Y(K) = ?$

* First do analysis in time domain given $x(n) \otimes h(n)$
 If $x(n) = \{x(0), x(1), x(2), x(3)\}$, $h(n) = \{h(0), h(1), h(2), h(3)\}$
 $\rightarrow h(0) \quad h(3) \quad h(2) \quad h(1)$ $x(0) \quad y(0)$
 $h(1) \quad h(0) \quad h(3) \quad h(2)$ $x(1) \quad y(1)$
 $h(2) \quad h(1) \quad h(0) \quad h(3)$ $x(2) \quad y(2)$
 $h(3) \quad h(2) \quad h(1) \quad h(0)$ $x(3) \quad y(3)$
 you get $y(n) = \{y(0), y(1), y(2), y(3)\}$
 This we do in FT

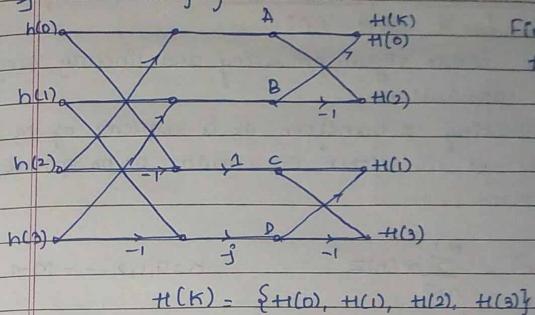
But in DIF FFT procedure is
First find $X(k)$ using DIFFFT



$$X(k) = \{x(0), x(1), x(2), x(3)\}$$

First find
A, B, C, D
& then $X(k)$

→ Same way find $H(k)$

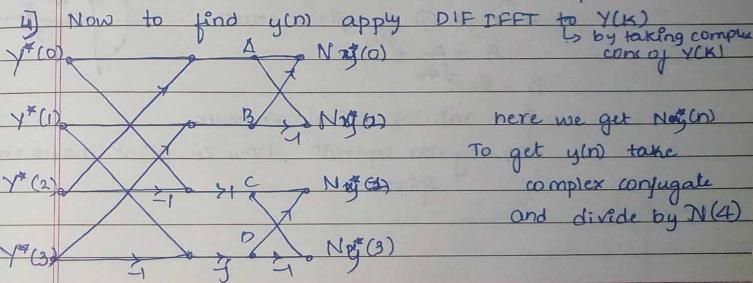


$$H(k) = \{H(0), H(1), H(2), H(3)\}$$

3] Find $Y(k) = X(k)H(k)$

$$Y(k) = \begin{bmatrix} X(0) \\ X(1) \\ X(2) \\ X(3) \end{bmatrix} \begin{bmatrix} H(0) \\ H(1) \\ H(2) \\ H(3) \end{bmatrix}$$

→ not matrix mul,
do element by
element multi
-plcatn



here we get $Ny^*(n)$
To get $y(n)$ take
complex conjugate
and divide by $N(4)$

CHIRP-Z TRANSFORM (CZT Algorithm)

- Used in radar signal processing and linearly varying frequency applications.
- For evaluating z transform of a sequence of M pts in z-plane w/ lie either on circular or on spiral contour in z-plane.

$$X(z_k) = \sum_{n=0}^{N-1} x(n) z_k^{-n} \quad k = 0, 1, 2, \dots, M-1$$

If this reduces to DFT

$$z_k = e^{j2\pi k/N} = WN^k, \quad k = 0, 1, \dots, N-1$$

Let z_k be a point on the spiral contour centred about the origin

$$\rightarrow z_k = AB^{-k}$$

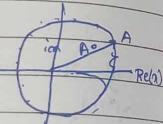
$$A \rightarrow \text{pt @ w/ spiral starts}$$

$$A = A_0 e^{j\theta_0}$$

$$B = B_0 e^{j\phi_0}$$

$B \rightarrow$ rate of spiral contour

\hookrightarrow angular speed, how z_k point moves on spiral



Case 1: $|B| < 1$

$\{z_k\} \rightarrow$ It spirals towards the origin

Case 2: $|B| > 1$

$\{z_k\} \rightarrow$ spirals away from the origin

Case 3: $|B| = 1$

$\{z_k\} \rightarrow$ will be on circle A_0

* Special case

$$\begin{cases} \text{If } A_0 = 1 \text{ and } B = e^{-j2\pi/N} \\ \rightarrow M = N \end{cases}$$

It corresponds to DFT if contour is a unit circle

$$\rightarrow X(z_k) = \sum_{n=0}^{N-1} x(n) A^{-n} B^{kn}, \quad 0 \leq k \leq N-1$$

$$\text{W.K.T. } nk = \frac{1}{2} [n^2 + k^2 - (k-n)^2]$$

$$\therefore X(z_k) = \sum_{n=0}^{N-1} x(n) A^{-n} B^{\frac{n^2}{2}} B^{\frac{k^2}{2}} B^{\frac{(k-n)^2}{2}}$$

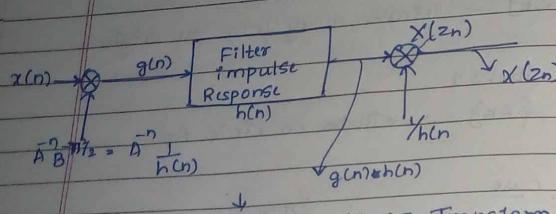
det us consider 2 functns $g(n) = x(n) A^{-n} B^{\frac{n^2}{2}}$
and $h(n) = B^{\frac{k^2}{2}} B^{\frac{(k-n)^2}{2}}$

$$\therefore X(z_k) = \sum_{n=0}^{N-1} g(n) \frac{1}{h(n)} (h(k-n)) \quad 0 \leq n \leq N-1$$

Now consider

$$X(z) = \frac{1}{h(n)} [g(n) * h(n)]$$

↳ This is computation of Chirp X transform algorithm



Generating Chirp z Transform

$$h(n) = \left[e^{-j\phi_0} \right] \frac{n^2}{2} = e^{\frac{jn^2\phi_0}{2}} - e^{\frac{jwn}{2}}$$

☞ complex exponential sequence linearly increasing frequency is called "Chirp".

GOERTZEL ALGORITHM :- To find DFT

- This algorithm is linear filtering technique
- Efficient tool to compute DFT when small number of frequency points are there.
- This algorithm makes use of periodicity property of w_N^{-kn}

→ Let us take, given signal $x(m)$

$$x(k) = \sum_{m=0}^{N-1} x(m) w_N^{km} w_N^{-kn}$$

$$x(k) = \sum_{m=0}^{N-1} x(m) w_N^{-k(N-m)}$$

$$x(k) = \sum_{m=-\infty}^{\infty} x(m) w_N^{-k(N-m)}$$

$$\text{let } y_k(n) = \sum_{m=-\infty}^{\infty} x(m) w_N^{-(n-m)k}$$

$$x(k) = y_k(n) \Big|_{n=N}$$

$$y_k(n) = x(n) * w_N^{-kn}$$

Transfer func of Goertzel filter is

$$H_k(z) = \sum_{n=0}^{\infty} h_k(n) z^{-n} = \sum_{n=0}^{\infty} (w_N^{-k} z^1)^n$$

$$\text{WKT } \sum_{n=0}^{\infty} a^n = \frac{1}{1-a}, |a| < 1$$

$$H_k(z) = \frac{1}{1 - w_N^{-k} z^{-1}}$$

$$\frac{Y_k(z)}{X(z)} = \frac{1}{1 - w_N^{-k} z^{-1}}$$

Cross multiply and take inverse Z transform

$$Y_k(n) = w_N^{-k} Y_k(n-1) = x(n)$$

Take $H_k(z)$ & multiply and divide by $1 - w_N^{-k} z^{-1}$

$$H_k(z) = \frac{1 - w_N^{-k-1} z}{(1 - w_N^{-k-1} z)(1 - w_N^{-k} z^{-1})}$$

$$H_k(z) = \frac{1 - w_N^{-k} z}{1 - w_N^{-k} z - w_N^{-k-1} z^{-1} + z^{-2}}$$

$$H_k(z) = \frac{1 - w_N^{-k-1} z}{1 - e^{\frac{j2\pi k}{N}} z^{-1} - e^{\frac{-j2\pi k}{N}} z^{-1} + z^{-2}}$$

$$H_k(z) = \frac{1 - w_N^{-k} z}{1 - 2\cos\left(\frac{2\pi k}{N}\right) z^{-1} + z^{-2}}$$

$$= H_1(z) + h_0(z)$$

$$\text{let } V_k(z) = X(z) + h_1(z)$$

$$= 1 \cdot x(z)$$

$$1 - 2\cos\left(\frac{2\pi k}{N}\right) z^{-1} + z^{-2}$$

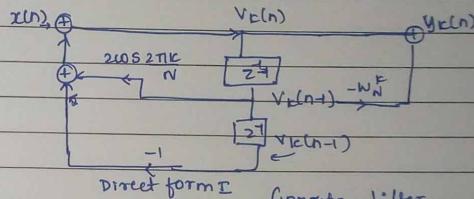
(cross mult & then take inverse Z transform)
 $V_k(n) = 2\cos\left(\frac{2\pi k}{N}\right) V_k(n-1) - V_k(n-2) + x(n)$

$$Y_k(z) = H_2(z) V_k(z)$$

$$Y_k(z) = (1 - w_N^{-k-1} z) V_k(z)$$

$$Y_k(n) = V_k(n) - w_N^{-k} V_k(n-1)$$

$$V_k(-1), V_k(-2) \Rightarrow \text{zero}$$



→ It's an ^{very} efficient algorithm as
 Each iteration will have only one real multiplication & two additions.

$x(n) \Rightarrow \text{real}$

$N+1$ real multiplications only