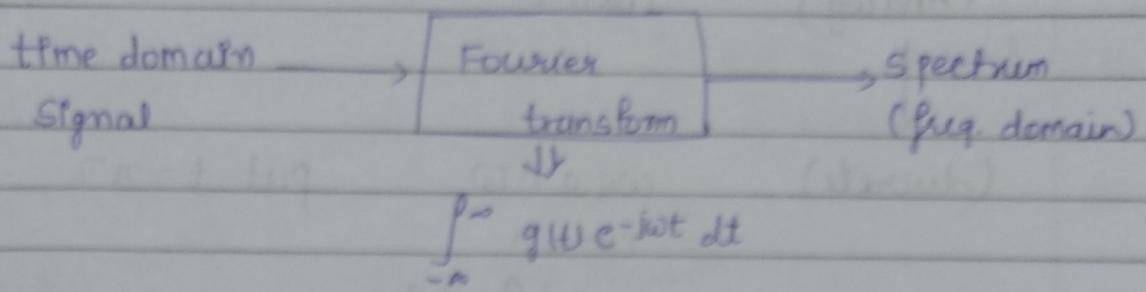
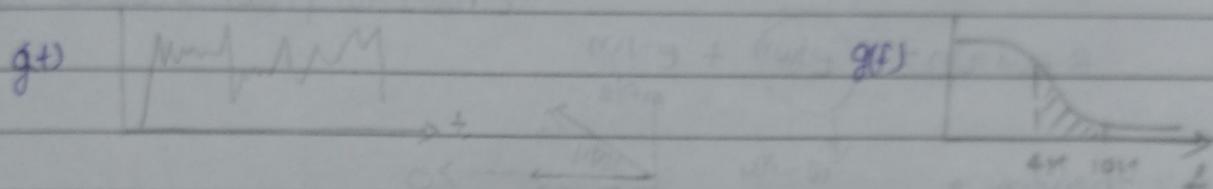


TRANSFORMS



Filtering Concept:



If we have to need only till $4M$ spectrum, then we need Low Pass Filter for that we have designed such a filter that is

$$g(t) \xrightarrow{IFT} \int_0^{2\pi f} g(\omega) e^{j\omega t} d\omega \quad \dots f = 4M$$

Time Domain	Freq. Domain
CT	AP
P	DT
DT	P
AP	CP

DTFT \rightarrow FT \rightarrow FS

$$x(k) = \int_{-\pi/2}^{\pi/2} x(t) e^{-j\omega k t} dt$$

\downarrow

$k = -\infty \text{ to } \infty$

Kernel Basis func

$$x(e^{j\omega}) = \sum_{n=-\infty}^{\infty} x(n) e^{-j\omega n}$$

Analog [max. freq. = ∞ ; $t=0 \rightarrow$ dc]
 discrete [max. freq. = π]

\Rightarrow max. freq. is always π , no matter whatever be the max. F of a signal.

$$F_{\text{max}} = 100 \text{ Hz}$$

Discrete sinusoid concept:

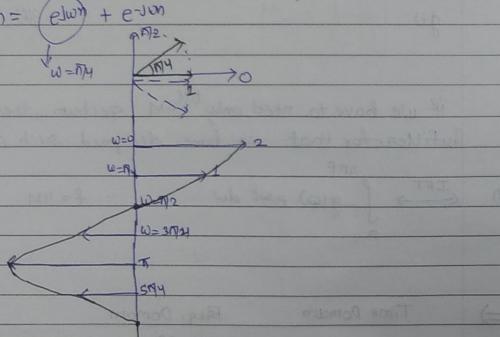
$$(\text{Analog}) \quad \cos \omega t$$



$$(\text{discrete}) \quad \cos(\omega n T_s) \quad \text{put } t = n T_s \\ = \cos \omega n$$

$\omega \rightarrow$ discrete freq. in rad/sec (how much phase is covered rad. in per sample)

$$2 \cos(\omega n) = e^{j\omega n} + e^{-j\omega n}$$



$$\omega = 2 \times 1 / T_s ; T_s = 1 \text{ Hz}$$

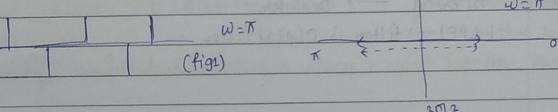
$$\text{discrete} \quad \left\{ \begin{array}{l} \omega = 2\pi f \\ f_s \end{array} \right. \quad \begin{array}{l} \xrightarrow{\text{continuous signal freq.}} \\ \xrightarrow{\text{sampled freq.}} \end{array}$$

$$\text{for max. discrete freq. } \omega_{\text{max}} = \pi \quad : f_s = 2f$$

$$2\pi f = 2\pi F$$

$$f = F / f_s$$

100 Hz is sampled with 200 Hz



$$100 \text{ Hz}$$

$$100T_s$$

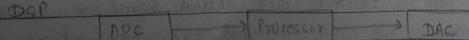
$$(fig 1)$$

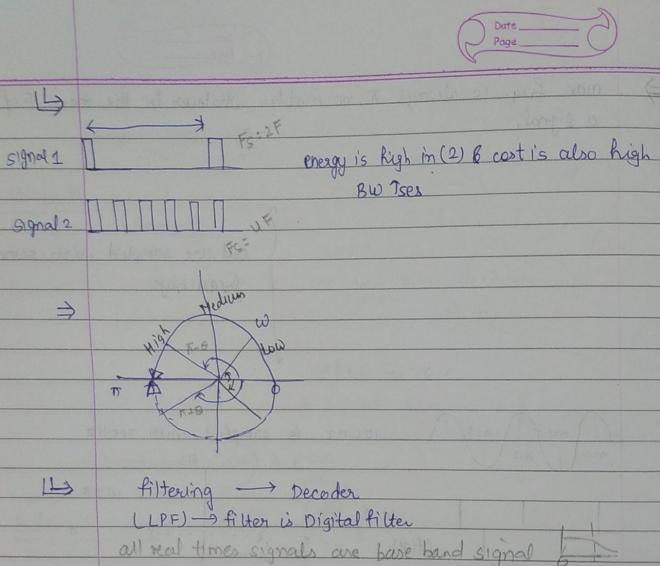
$$100T_s$$

$$(fig 2)$$

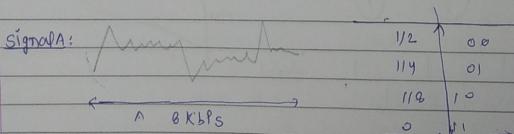
(fig 1) & (fig 2) are same for both signals 100 Hz and $100T_s$ but one thing that differs them is T_s value.

- Basic component of digital sig. is clock
 \Rightarrow clock speed differentiates the two systems.





and anti-aliasing filter is analogous filter.

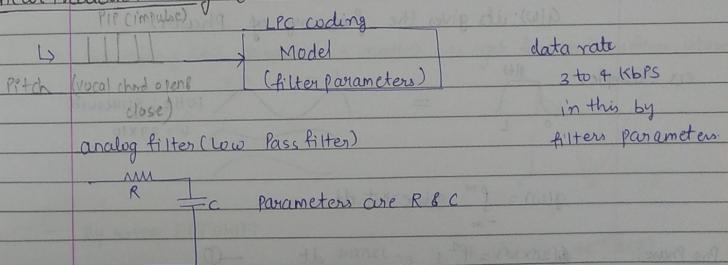


- Signal spectrum is changed by the filter.
minimize the BPS \rightarrow BW uses \rightarrow cost less
 - We don't transmit our actual signal or another form of signal. we only transmit the base band signal with the help of filters. and also we detect our data back with the help of filter (base band filter). we only transmit the parameters of filters.

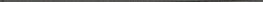
- In cellphones, for transmit the data, we don't transmit the any form of data but we only transmit the parameters of filters.

do signal (anisotrop) \rightarrow SVD \rightarrow impulse, FT showing cell freq. compo
air

ediction coding & required data) ← By filters
(mouth cavity)



- But when we doing sampling, quantization encoder -- then data rate is 64 Kbps which is very high as compare to (model method) Linear prediction coding.

linear  $e = k_1 a + k_2 b + k_3 c + k_4 d$
(linear prediction)

- anti-aliasing filter is analog filter & LPF, HPF, BPF, CIR digital filters.

Transforms

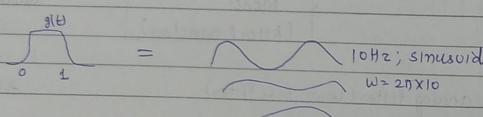
when we talk about transmission or signal processing then spectrum we talk about the spectrum.

We only consider on the spectrum

Fourier transform: $G(j\omega) = \int_{-\infty}^{\infty} g(t) e^{-j\omega t} dt$

$G(j\omega)$; it gives the frequency of phasor.

as example:



$$g(2\pi \times 10) = \int_0^1 1 \cdot e^{-j2\pi \times 10 t} dt \quad \text{--- (1)}$$

its gives the strength of phasor at $f=10$ Hz

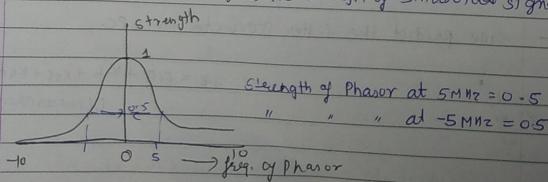
$$\text{because } \cos(2\pi \times 10 t) = e^{j(2\pi \times 10 t)} + e^{-j(2\pi \times 10 t)}$$

neg. Phase:

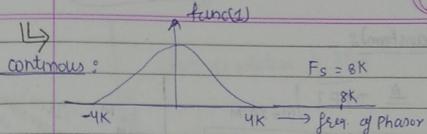
$$g(-2\pi \times 10) = \int_0^1 1 \cdot e^{j2\pi \times 10 t} dt \quad \text{--- (2)}$$

its gives the strength of phasor at $f=10$ Hz

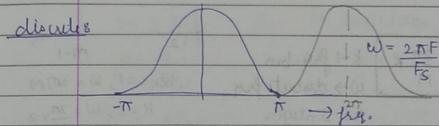
$G(j\omega) + G(j\omega)$ → gives the strength of sinusoidal signal.



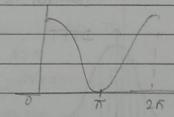
$$\text{Strength of sinusoidal signal} = 0.5 + 0.5 = 1.00$$



spectrum will remain same by doing sampling on the func(f). If it's not then it may distorted.

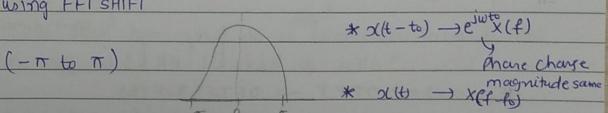


- By using FFT command in matlab, we get fourier transform from $(0 \text{ to } 2\pi)$



- By using FFT SHIFT

$(-\pi \text{ to } \pi)$



$$* x(t - t_0) \rightarrow e^{j\omega_0 t} X(f)$$

phase change

$$* x(t) \rightarrow X_{FF}(f)$$

magnitude same

- When we shifted the signal in time domain, the spectrum of phase also changes / shift in freq. domain.

DFT: discrete time fourier transform

$$X(e^{j\omega}) = \sum_{n=0}^{N-1} x(n) e^{-j\omega n}$$

discrete freq.: $-\pi \text{ to } \pi$

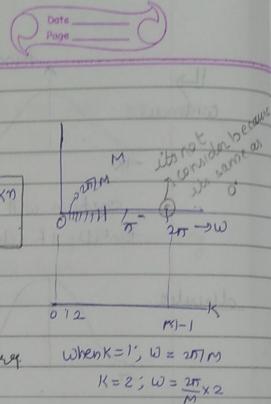
here discrete freq. are infinite. So by doing DFT, we can get finite frequencies

DFT (Discrete Fourier Transform):

$$DFT \triangleq DTFT \Big|_{W=2\pi/M}$$

M-Point DFT:

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



$$\begin{cases} W = 2\pi \\ M \end{cases} \begin{cases} K \rightarrow \text{freq. bin} \\ w \rightarrow \text{discrete freq.} \end{cases}$$

K = Integer

$$\begin{cases} w = 2\pi \\ M \end{cases} \begin{cases} K = 2; w = \frac{2\pi}{M} \times 2 \\ M-1 \end{cases}$$

DFT is used to find the unknown frequency.

$k \rightarrow$ freq. bin

$$N = 32$$

$$f_s = 64 \text{ sample/sec}$$

$$\begin{aligned} &\text{freq. resolution} \triangleq \frac{2\pi}{M} \\ &M = 32 \quad f_s = 64 \text{ sample/sec} \quad \text{For 32 samples/sec} \\ &W_f = \frac{2\pi \times 5}{32} \quad w_2 = \frac{2\pi \times 27}{32} \\ &F_1 = \frac{2\pi \times 5 \times 64}{32 \times 32} \quad F_2 = \frac{2\pi \times 27 \times 64}{32 \times 32} \\ &= 10 \text{ Hz} \quad = 54 \text{ Hz} \end{aligned}$$

max. discrete freq. = π ; and range between 0 to π
only one w_k is selected $w = \frac{10\pi}{32}$

analogue freq. $F = 10 \text{ Hz}$

Example: which freq. bin, what do we have max. power if the PIP was at 11 Hz instead of 10 Hz?

Sol:-

$$\begin{aligned} F &= 11 \text{ Hz} \\ w &= 2\pi F = \frac{2\pi \times 11}{64} = \frac{11\pi}{32} \end{aligned}$$

$$w = \frac{2\pi \cdot k}{M}$$

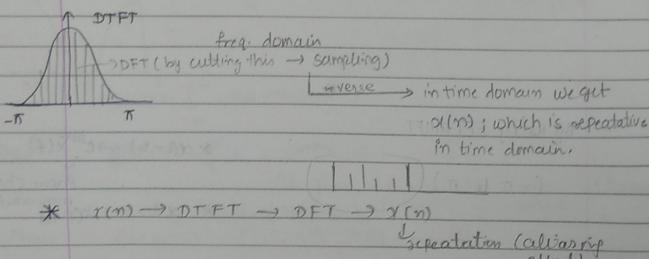
$$\frac{11\pi}{32} = \frac{2\pi \cdot k}{32}$$

$$k = \frac{11}{2} = 5.5 \notin \mathbb{Z}$$

$$*\quad \text{freq. resolution} \triangleq \frac{2\pi}{M} = \frac{2\pi}{32}$$

$$*\quad \text{DFT} \rightarrow 0 \text{ to } 2\pi$$

- sampling in one domain is repetition in another domain occurs.



$$x(n) \rightarrow DTFT \rightarrow DFT \rightarrow x(n)$$

aliasing effect

- to remove or overcome from the anti-aliasing filter;

$$M \geq N$$

too many points between them
In original signal, how much samples are there

- time limited Signal \rightarrow Band Unlimited Signal



Example: Rohit samples the signal ^{Unknown} at 64 sample/second. He selects 32 samples and computes 32 points DFT.

$x(n) \rightarrow$ discrete time signal
with $0 \leq n \leq N-1$

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

$$K \rightarrow 0 \leq k \leq M-1$$

$N \rightarrow$ in time domain (how many samples)

$M \rightarrow$ in freq. domain (how many parts in DFT, 0 to ω_0)

* K can never be fractional no. It should be integer always.

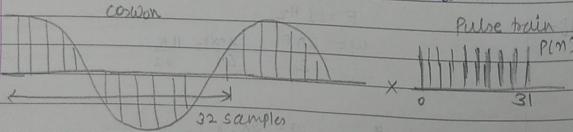
- DFT missed the spectrum, because DFT always takes the value of K^{integer} .

the graph spectrum achieved is 0 at all points.

Modulation property

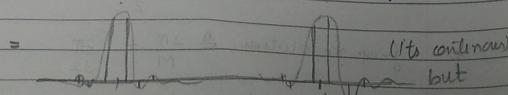
To get the spectrum, modulation property of Fourier transform is used.

Modulation: $x(t) \cos \omega_c t \longleftrightarrow X(\omega + \omega_c) + X(\omega - \omega_c)$



32 samples = $[\cos \omega_0 n] \cdot P(n)$ = spectrum of pulse is sinc func
spectrum of pulse is sinc func

ω_0 freq.
 ω_0 freq.

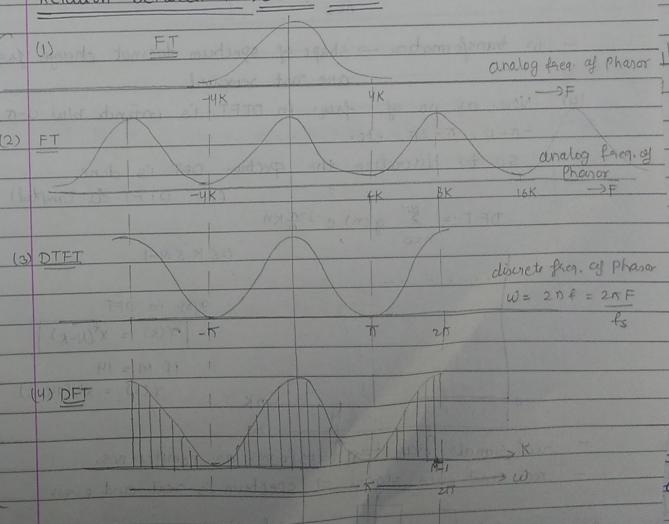


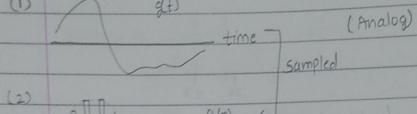
to get only 32 samples of a signal then multiply it with pulse train (32 samples 0 to 31)

\Rightarrow 32 samples of signal is achieved.

- The DFT of impulse train is discrete sinc. as we get max. peak at $k=5 \& 6$

Relation Between FT, DTFT & DFT





$g(n) | g(t)$ is quantized $\Rightarrow g(n)$ is achieved (discrete signal)
 $FT = \int_{-\infty}^{\infty} g(t) e^{-j\omega t} dt$

(3) $G(e^{j\omega}) = DTFT = \sum_{n=0}^{N-1} g(n) e^{-j\omega n}$ (discrete spectrum)

- In transformation \rightarrow shape of spectrum does not change freq. are not removed

(4) Now, as no. of freq. in DTFT is infinite b/w $0-\pi$, $-\pi-0$, $\pi-2\pi$ etc.

so to discretize the spectrum DFT is done

(i.e. DTFT is sampled)

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

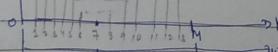
$0 \leq k \leq M-1$

Sym in DFT

$$[x(k) = x^*(N-k)]$$

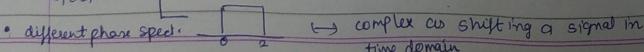
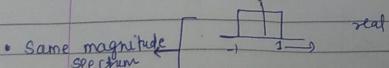
If $M = 14$

$$x(8) = x^*(14-8)$$

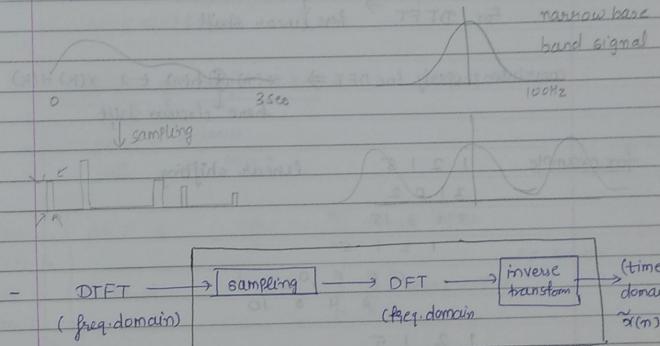


- real signals may have transforms as complex nos.

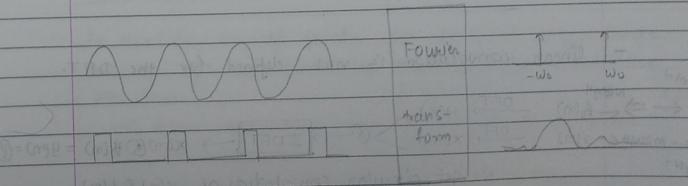
- real and even signal \Rightarrow spectrum is real and even



- Sampling in one domain results into periodicity in other domain.



② Periodicity in one-domain results in sampling ②



It's always not true.

- DFT is always periodic in both domains.

- circular shift

- in DFT not linear shift, it rotates shift, [which comes out that comes in]

$$- x((n-n_0)_N) \xrightarrow{\text{circular shift}} g(t-t_0) \xrightarrow{\text{linear shift}} g(t) e^{-j\omega_0 t}$$

rotating/circular shift.

$$x(n) - h(n) \rightarrow y(n) = x(n) * h(n)$$

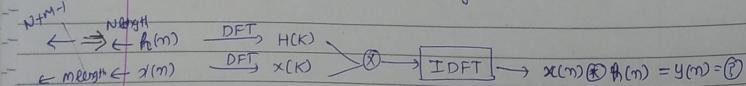
convolution property $\rightarrow x(n) * h(n) \leftrightarrow X(e^{j\omega n}) H(e^{j\omega n})$
for DTFT \rightarrow here linear shift

convolution property for DFT $\rightarrow x(n) * h(n) \leftrightarrow X(k) H(k)$
here circular shift

for example $\begin{matrix} 1 & 2 & 1 & 5 \\ 3 & 1 & 0 & 2 \\ 3 & 6 & 3 & 15 \\ 1 & 2 & 1 & 5 \\ 0 & 0 & 0 & 0 \\ 8 & 4 & 2 & 10 \end{matrix}$ linear shifting

$\begin{matrix} 1 & 2 & 1 & 5 \\ 3 & 1 & 0 & 2 \\ 3 & 6 & 3 & 15 \\ 5 & 1 & 2 & 1 \\ 0 & 0 & 0 & 0 \\ 10 & 2 & 4 & 2 \end{matrix}$ circular shifting

- linear convolution is not defined for the DFT.



we get circular convolution of $x(n) * h(n)$

but as we know that we always have linear convolution. So we have to change this circular to linear



- same length of $h(m)$ & $x(n)$ are required for linear convolution. so we added the $n+m-1$
- O/P of any filter ($h(m)$) is convolution.
 $\rightarrow h(m) \rightarrow y(m) = x(n) * h(m) = \sum x(k) h(n-k)$

By doing $\sum x(k) h(n-k)$, we also get the linear convolution but its requirement of multiplication and division is more as compare to DFT. So processor work will more in $\sum x(k) h(n-k)$.

Notation of Sanjit Mitra :

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

$$w_N^k = e^{-j2\pi/N \cdot k}$$

can be written in matrix form

$$\left. \begin{array}{l} \text{DFT} \Rightarrow x(k) = \sum_{n=0}^{N-1} x(n) e^{-j\frac{2\pi}{N} kn} \\ 0 \leq k \leq M-1 \\ \text{IDFT} \Rightarrow x(n) = \frac{1}{N} \sum_{k=0}^{M-1} x(k) e^{j\frac{2\pi}{N} kn} \end{array} \right\} \quad N, k \geq 0$$

$$X = x \cdot D \Rightarrow D = \begin{bmatrix} 1 & 1 & \dots & 1 \\ w_N & w_N^2 & \dots & w_N^{M-1} \\ \vdots & \vdots & \ddots & \vdots \\ w_N^{N-1} & w_N^{2(N-1)} & \dots & w_N^{(M-1)(N-1)} \end{bmatrix}$$

Inverse of processing speed of processor

In reality no. of points for DFT is very large
So no. of operations of multiplication & addition etc. Tries

\rightarrow time taken and power consumed by processor was more.

\Rightarrow we need to \downarrow the time taken.

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

for $N=2$; $M=2$ \rightarrow min. M for $N=2$ i.e. $M \geq N$

$$x(k) = x(0) e^{-j\frac{2\pi}{M} k(0)} + x(1) e^{-j\frac{2\pi}{M} k(1)}$$

$$x(0) = x(0) + x(1)$$

$$x(1) = x(0) + x(1) e^{-j\pi}$$

So the basic idea to decrease time was to convert 1000 samples to 8 samples and then computing.

1000 samples $\rightarrow x(k) =$

```

graph TD
    Root["+-+--+-+--+-+--"]
    Root -- odd --> Node1Odd[odd]
    Root -- even --> Node1Even[even]
    Node1Odd -- odd --> Node2Odd1[odd]
    Node1Odd -- even --> Node2Even1[even]
    Node1Even -- odd --> Node2Odd2[odd]
    Node1Even -- even --> Node2Even2[even]
    Node2Odd1 -- 0 --> Leaf1[0]
    Node2Odd1 -- E --> Leaf2[E]
    Node2Even1 -- 0 --> Leaf3[0]
    Node2Even1 -- E --> Leaf4[E]
    Node2Odd2 -- 0 --> Leaf5[0]
    Node2Odd2 -- E --> Leaf6[E]
    Node2Even2 -- 0 --> Leaf7[0]
    Node2Even2 -- E --> Leaf8[E]
  
```

Here $x(k)$ is divided into many 2 samples pair (even & odd).

for N-Points

$$x(k) \rightarrow 2N^2(x) \rightarrow N/2 \log_2 N] - \text{FFT}$$

$$2N^2 (+) \rightarrow N \log_2 N]$$

and if no. of samples are odd then a zero valued sample is added \rightarrow called as zero padding.

$$\text{DTFT : } X(e^{j\omega}) = \sum_{n=0}^{N-1} x(n) e^{-j\omega n}$$

\Rightarrow multiplication of spectrum by $e^{-j\omega t}$ in freq domain, this cause real spectrum will change into conjugate spectrum

*Exo DTFT $x(e^{j\omega}) = \sum_{n=0}^{N-1} x(n) e^{-jn\omega}$; what is the condition for $x(e^{j\omega})$ DTFT to exist?
Sol: i.e. magnitude is finite for existence of DTFT

$$x(e^{j\omega}) < \infty$$

$$x(e^{j\omega}) = x(0)e^{-j\omega 0} + x(1)e^{-j\omega(1)} + x(2)e^{-j\omega(2)} + x(3)e^{-j\omega(3)}$$

mag (Amplitude): $|x(e^{j\omega})| = |x(0)| + |x(1)| + |x(2)| + |x(3)| + \dots$

$$\text{So: } \left| \sum_{m=0}^{N-1} |x(m)| \right| < \infty \quad \text{for existence of DTFT}$$

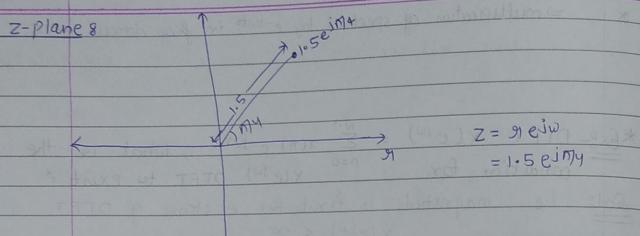
$$x(e^{j\omega}) = \sum_{n=0}^{\infty} x(n) e^{-jn\omega} \quad \begin{matrix} \text{how to make this convergence} \\ \text{absolute} \\ \text{or summable for existence} \\ \text{of DTFT} \end{matrix}$$

- if we don't know about the SIS then how can we find out its response. By giving know I/P & SIS the O/P we can evaluate the SIS response.
 - I/P Signals :- Unit Step, ramp
but Unit step signal is $\sum_{n=1}^{\infty} x(n) \cdot 1^n < \infty$ not satisfy this

By doing z -transform, we can satisfy this eqn ①

$$\begin{aligned} \text{Z-transform of } X(e^{j\omega}) &= \sum_{n=0}^{\infty} x(n) e^{-jn\omega} \xrightarrow[n \rightarrow \infty]{\text{added one series when } n \rightarrow \infty} \\ &\quad \sum_{m=0}^{N-1} r^{-m} e^{-jm\omega} x(m) \\ &= \sum_{m=0}^{N-1} x(m) (re^{j\omega})^{-m} \end{aligned}$$

$$X(z) = \sum_{n=0}^{\infty} x(n) z^{-n} \quad \text{"Z-transform"} \quad z = r e^{j\omega}$$



To analyze implementation the digital signal, **z-transform** is used.

* In Laplace $s = - + j\omega$

?

* By putting $\omega_1 = 1$ $\rightarrow H(z) = (z-1)$
 $\rightarrow z\text{-transform} = DTFT$

ROC (Region of Convergence)

for what value of z , the **z-transform** exist?

Example

$$H(z) = \frac{(z-1)(z-0.5)}{z(z+1)}$$

→ zeros
→ poles

$H(z)$ is the SIS response of **z-plane**.

Date _____
Page _____

concept — Sensors + 50Hz ($\omega = 2\pi f/f_s$) → for suppressed the 50Hz freq. → for this at this freq. $H(z) = 0$; zeros → $z_1 = 9e^{j\omega}$ → converted into **z-plane** → $H(z) = (z-2)$

- Digital SIS design: if we want to suppress any freq. we find the 'z' of the freq., we keep at zero. The nearby surrounding freq. which is unwantedly suppressed is lifted by assigning poles at those frequencies.