

# DFT and FFT

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# The Outline

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# The Outline

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# State of the art

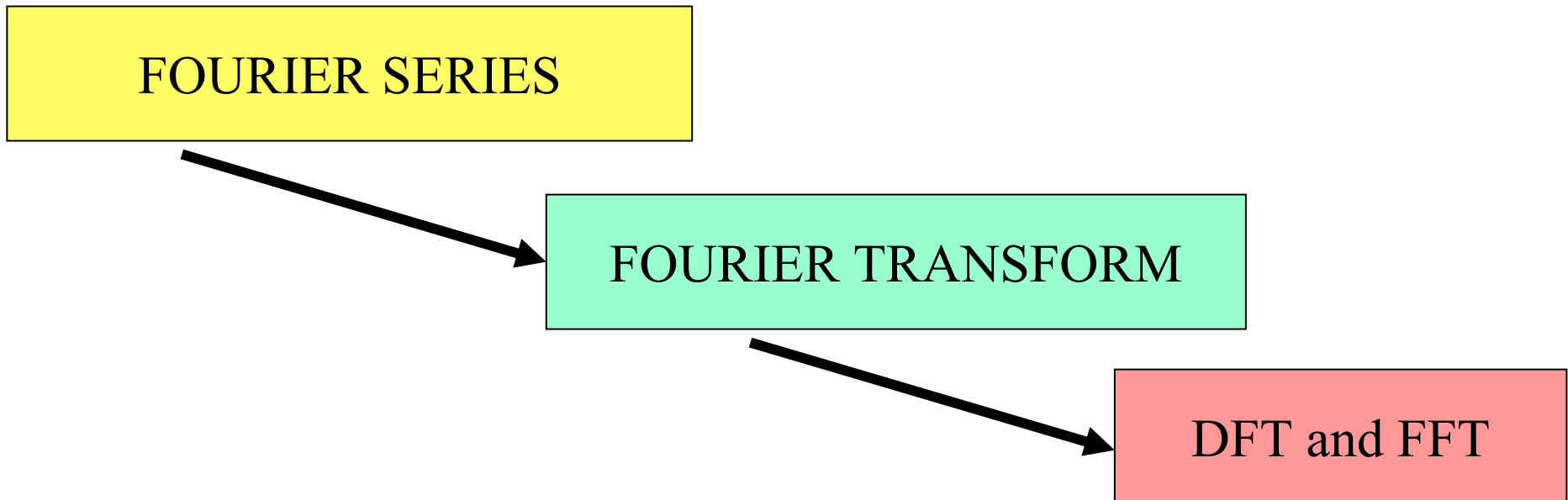
- The transformation of discrete data between the time and frequency is useful in analysis of the signal
- Fourier Transform is used for its transformation
- Voltage v.s. time become magnitude v.s. frequency and phase v.s frequency
- The domains provide complementary information about the same data

# The purpose of Discrete Transform

- Discrete transforms, particularly the discrete cosine transform are used in the data compression of speech and video signals to allow transmission with reduced bandwidth
- It is also used in image processing to obtain a reduced set of features for pattern recognition purposes
- For these computation the transformation from the frequency to the time domain is important

# What kind of transformation ?

- The Discrete Fourier Transform (DFT) and Fast Fourier Transform (FFT) are the best known and the most important



# Fourier Series

- Any periodic waveform ,  $f(t)$  can be represented as the sum of an infinite number of sinusoidal and cosinusoidal terms

$$f(t) = a_0 + \sum_{n=1}^{\infty} a_n \cos(n\omega t) + \sum_{n=1}^{\infty} b_n \sin(n\omega t)$$

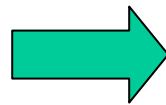
- Remember that  $f(t)$  is often a varying **voltage** v.s. **time** waveform



## Fourier series to the unit of volts

- Fourier series may be written more compactly by using exponential
- The series become relation of  $d_n$  as the unit of volts, it is very useful in pulse graphic

$$f(t) = \sum_{n=-\infty}^{\infty} d_n e^{jn\omega t}$$



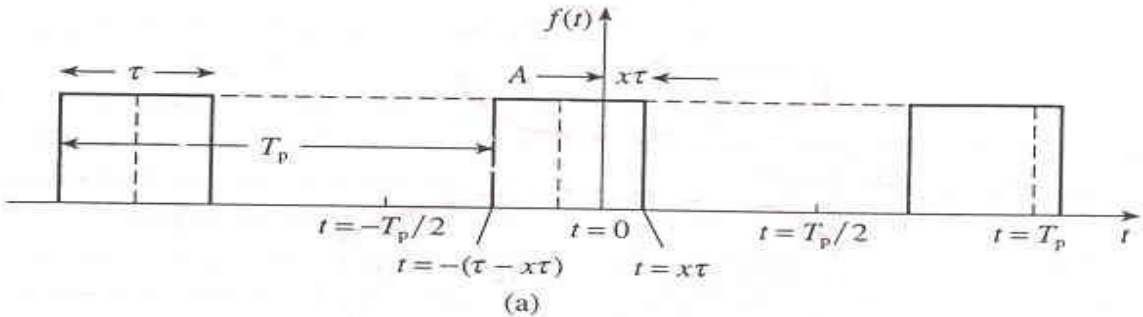
$$d_n = \frac{1}{T_p} \int_{-T_p/2}^{T_p/2} f(t) e^{-jn\omega t} dt$$

# The graphic example

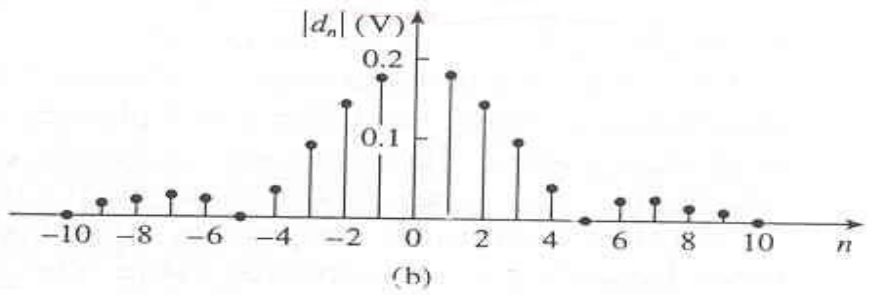
- Here are the example, look at the periodic unipolar pulse waveform shown in figure 4.1 (a) . Deliberate choice of time origin to be offset from the centre and edge of a pulse is intended to allow illustration of the phase feature of the Fourier series .
- By substituting appropriate values into the formula of  $d_n$ , we can get the graphic below :

# The graphic example (cont'd)

Waveform



Amplitude Spectrum



Phase Spectrum

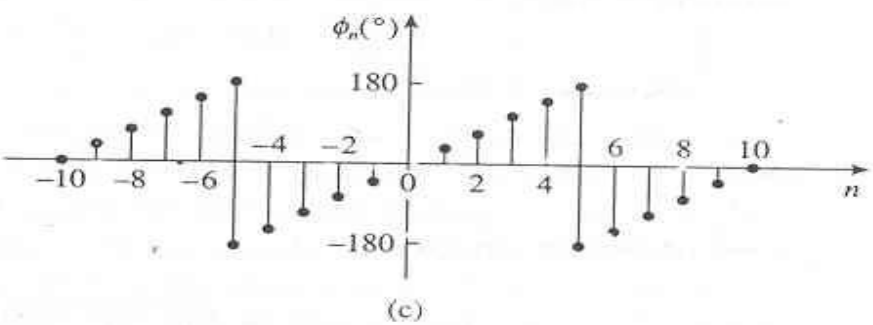


Figure 4.1

# The Fourier Transform

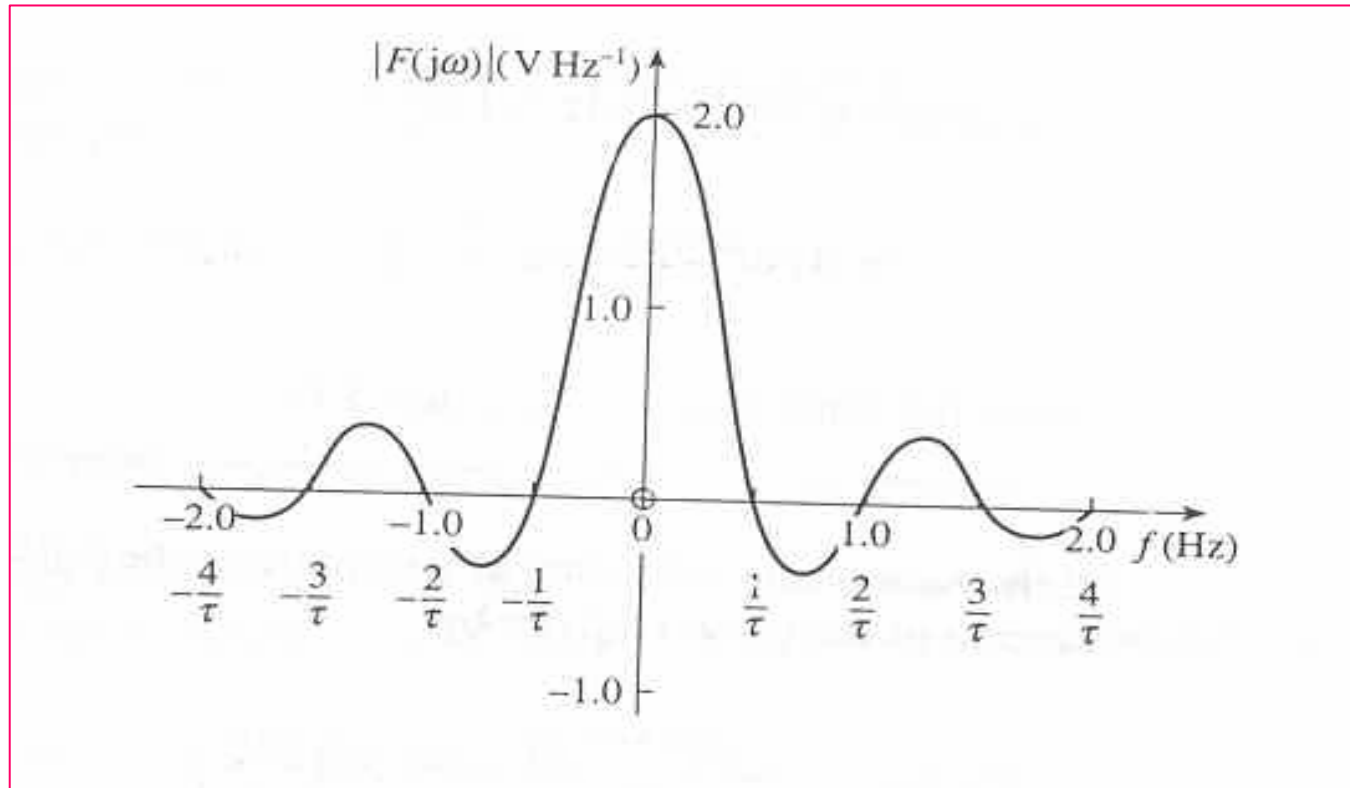
- The Fourier series approach has to be modified when the waveform is not periodic
- By using the following formula, we can change from the discrete frequency variable  $n\omega$  to the continuous variable  $\omega$
- So that, the **amplitude** and **phase** spectra become continuous

$$F(j\omega) = \frac{d(\omega)}{d\omega / 2\pi} = \int_{-\infty}^{\infty} f(t)e^{-j\omega t} dt$$

# What is its transform ?

- $F(j \omega)$  is complex and is known as the Fourier integral or more commonly = Fourier Transform
- Consider the graphic example on Figure 4.1 , by using Fourier transform we can transform *amplitude discrete pulse* to be *amplitude spectrum*
- Look at figure 4.2 on the next slide
- This is the key term of Fourier Transform that can be improved to be DFT and FFT

# Graphic : amplitude spectrum of a 2 V pulse



Amplitude  
Spectrum

Figure 4.2 a

# Graphic : energy spectrum of a 2 V pulse

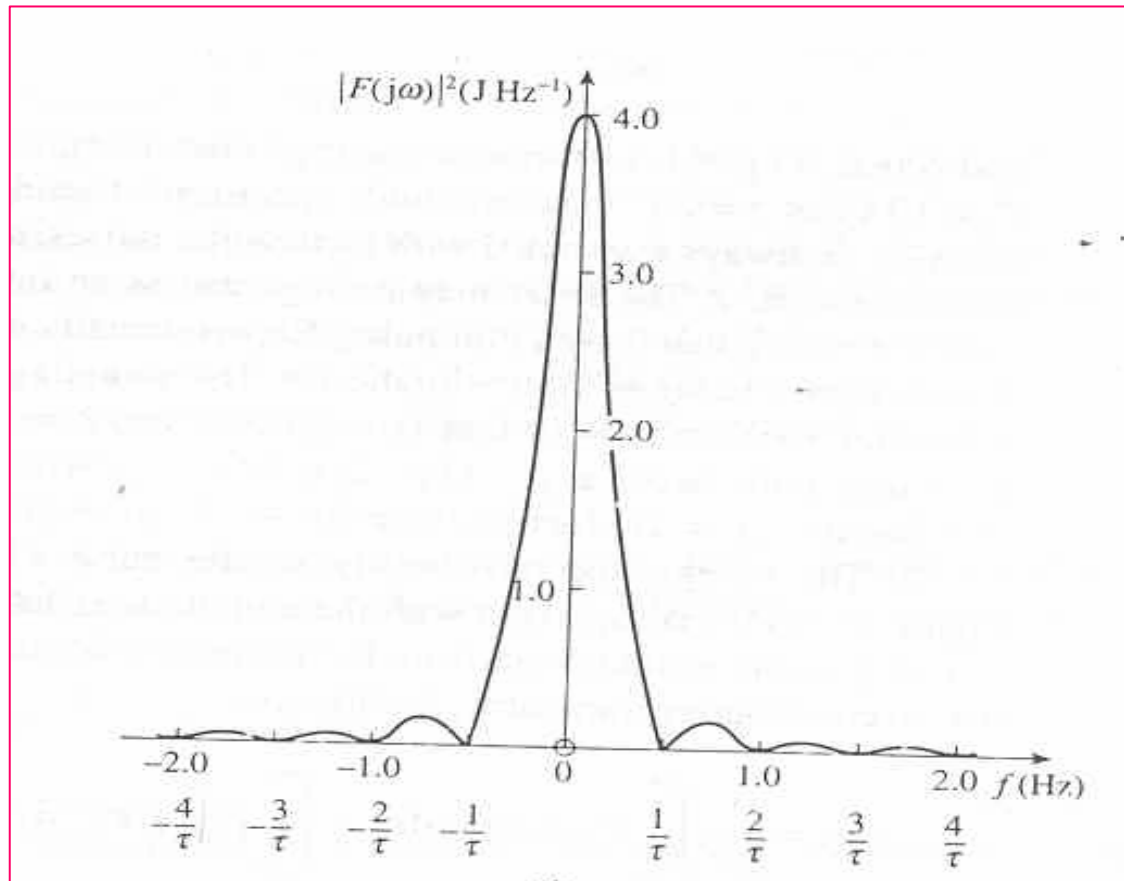


Figure 4.2 b

Energy  
Spectrum

# DFT and its inverse

- In practice the fourier components of data are obtained by digital computation rather than by analog processing
- This is achieved using a sample-and-hold circuit followed by an AD converter
- The problem is Fourier Transform can be used only for continuous data, while the data is commonly discrete and probably non-periodic



## DFT and its inverse (cont'd)

- However there is an analog transform for use with discrete data, known as Discrete Fourier Transform (DFT)
- The first assumption is : consider a waveform has been sampled at regular time intervals  $T$  to produce the sample sequence :

$$\{x(nT)\} = x(0), x(T), \dots, x[(N-1)T]$$

- Where  $n$  is the sample number from  $n=0$  to  $n=N-1$

## DFT and its inverse (cont'd)

- The data values  $x(nT)$  will be real only when representing the values of a time series such a voltage waveform
- The DFT of  $x(nT)$  is then defined as the sequence of complex values

$$\{X(k\Omega)\} = X(0), X(\Omega), X(2\Omega), \dots, X[(N-1)\Omega]$$

- Note that  $\Omega$  is the first harmonic frequency given by  $\Omega = 2\pi/NT$

## DFT and its inverse (cont'd)

- Related to our formula on previous slides, we can compare those equations, so that we have the DFT values  $X(k)$ , is given by :

$$X(k) = F_D [x(nT)] = \sum_{n=0}^{N-1} x(nT) e^{-jk\Omega nT}$$

- Where  $k = 0, 1, \dots, N-1$
- Watch out the **exponent variable**, don't miss it

$$-jk\Omega nT$$

# DFT and its inverse (cont'd)

- Because of  $\Omega=2\pi/NT$  , then the exponent variable of that formula can be convert to :

$$X(k) = \sum_{n=0}^{N-1} x(nT) e^{-j k \Omega n T}$$

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

## DFT and its inverse (cont'd)

- To make computation more easy, these equation are very useful :

$$e^{j\theta} = \cos \theta + j \sin \theta$$
$$e^{-j\theta} = \cos \theta - j \sin \theta$$

- So that, DFT is the key for having transformation from **time** to **frequency** domain

# Example of DFT

Evaluate the DFT of sequence  $\{1,0,0,1\}$

Answer :

Assume that this data represent four consecutive Voltages  $x(0)=1, x(T)=0, x(2T)=0, x(3T)=0$   
The data recorded at time intervals,  $T$  and  $N=4$ .  
Since  $N-1=3$ , then it is required to find the complex values of  $X(k)$  for  $k=0, k=1, k=2, k=3$ ,

Define each variable

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

## Example of DFT (cont'd)

For  $k=0$

$$\begin{aligned} X(0) &= \sum_{n=0}^3 x(nT) e^{-j \frac{2\pi \cdot n \cdot 0}{N}} \\ &= x(0) + x(T) + x(2T) + x(3T) \\ &= 1 + 0 + 0 + 1 \\ &= 2 \end{aligned}$$

Watch that  $X(0)=2$  is entirely real, of magnitude 2  
And phase angle  $\Phi(0)=0$

## Example of DFT (cont'd)

For  $k=1$

$$\begin{aligned}
 X(1) &= \sum_{n=0}^3 x(nT) e^{-j \frac{2\pi \cdot n \cdot 1}{N}} \\
 &= 1 \cdot e^{-j \frac{2\pi \cdot 0 \cdot 1}{4}} + 0 \cdot e^{-j \frac{2\pi \cdot 1 \cdot 1}{4}} + 0 \cdot e^{-j \frac{2\pi \cdot 2 \cdot 1}{4}} + 1 \cdot e^{-j \frac{2\pi \cdot 3 \cdot 1}{4}} \\
 &= 1 + 0 + 0 + 1 \cdot e^{-j \frac{3\pi}{2}} \\
 &= 1 + \cos\left(\frac{3\pi}{2}\right) - j \sin\left(\frac{3\pi}{2}\right)
 \end{aligned}$$

$$= 1 + j$$

$$\text{Magnitude} = \sqrt{2} \quad \text{Phase} = 45^\circ$$



## Example of DFT (cont'd)

For  $k=2$ 

$$\begin{aligned} X(1) &= \sum_{n=0}^3 x(nT) e^{-j \frac{2\pi \cdot n \cdot 2}{N}} \\ &= 1 \cdot e^{-j \frac{2\pi \cdot 0 \cdot 2}{4}} + 0 \cdot e^{-j \frac{2\pi \cdot 1 \cdot 2}{4}} + 0 \cdot e^{-j \frac{2\pi \cdot 2 \cdot 2}{4}} + 1 \cdot e^{-j \frac{2\pi \cdot 3 \cdot 2}{4}} \\ &= 1 + 0 + 0 + 1 \cdot e^{-j3\pi} \\ &= 1 - 1 \\ &= 0 \end{aligned}$$

## Example of DFT (cont'd)

For  $k=3$

$$X(1) = \sum_{n=0}^3 x(nT) e^{-j \frac{2\pi \cdot n \cdot 3}{N}}$$

$$= 1 \cdot e^{-j \frac{2\pi \cdot 0 \cdot 3}{4}} + 0 \cdot e^{-j \frac{2\pi \cdot 1 \cdot 3}{4}} + 0 \cdot e^{-j \frac{2\pi \cdot 2 \cdot 3}{4}} + 1 \cdot e^{-j \frac{2\pi \cdot 3 \cdot 3}{4}}$$

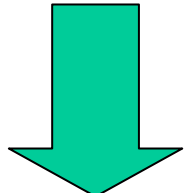
$$= 1 + 0 + 0 + 1 \cdot e^{-j \frac{9\pi}{2}}$$

$$= 1 - j$$

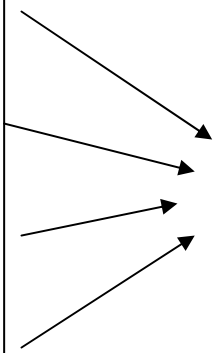
$$\text{Magnitude} = \sqrt{2} \quad \text{Phase} = -45^\circ$$

## Example of DFT (cont'd)

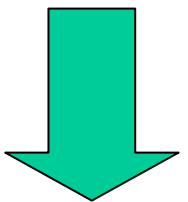
From those computation we have the DFT for time series  $\{1,0,0,1\}$ , given by the complex sequence



$$\begin{aligned} X(0) &= 2 \\ X(1) &= 1 + j \\ X(2) &= 0 \\ X(3) &= 1 - j \end{aligned}$$



$$\text{DFT} = \{ 2 , 1 + j , 0 , 1 - j \}$$



**PLOT THE GRAPHIC**

# Cyclical property of DFT

- The fact that  $X(k + N) = X(k)$
- It means DFT is periodic with period  $N$
- The DFT components are repetitive
- This is the cyclical property of DFT
- The amplitude spectrum of an  $N$ -point DFT is symmetrical about harmonic  $N/2$  when both the zero and  $(N+1)$ th harmonics are included in the plot

So that.. what is N point DFT ?

## IDFT as its inverse

- Inverse DFT is very useful in transformation from the **frequency** to the **time** domain
- The value of IDFT is  $x(nT)$  that commonly defined as  $F_D^{-1}[X(k)]$

$$F_D^{-1}[X(k)] = \frac{1}{N} \sum_{k=0}^{N-1} X(k) \cdot e^{j \frac{2\pi \cdot n}{N} k}$$

Please try to invert the previous example

# Properties of DFT

## 1. Symmetry

$\text{Re} [X(N-k)] = \text{Re} X(k)$  where Re is real part

## 2. Even functions

If  $x_e(n)$  is an even function, then :  $x_e(n) = x_e(-n)$

$$F_D[x_e(n)] = \sum_{n=0}^{N-1} x_e(n) \cos(k\Omega nT)$$

# Properties of DFT

## 3. Odd functions

If  $x_o(n)$  is an even function, then :  $x_o(n) = -x_o(-n)$

$$F_D[x_o(n)] = -j \cdot \sum_{n=0}^{N-1} x_o(n) \sin(k\Omega nT)$$

## 4. Parseval's Theorem

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

← The normalized energy  
In the signal

# Properties of DFT

## 5. Delta Function

$$F_D[\delta(nT)] = 1$$

## 6. The linear cross-correlation computed using DFT

$$r_{c x_1 x_2}(j) = F_D^{-1}[X_1^*(k) X_2(k)] \longrightarrow \text{Circular correlation}$$

$$r_{x_1 x_2}(j) = F_D^{-1}[X_{1a}^*(k) X_{2a}(k)] \longrightarrow \text{Linear-cross corr.}$$



# Computational Complexity of DFT

- A large number of multiplications and additions are required for the calculations of DFT
- For  $N$ -point DFT there will be  $N^2$  multiplications and  $N.( N - 1 )$  additions
- Just imagine for thousands of data there will be millions of multiplications and additions, that needs a lot of **memory**, **time** and **cost**
- So what should we do for reducing this number ?

# Fast Fourier Transform (FFT)

- FFT is an algorithm that is useful for speeding up the computation
- When applied in the time domain, the algorithm is referred to as a decimation-in-time (DIT) FFT
- Decimation refers to the significant reduction in the number of calculations performed on time domain data
- The computational savings will be seen to increase as  $N^2 - (N/2) \log_2 N$

# FFT Algorithm

- The notation can be re-written as :

$$X_1(k) = \sum_{n=0}^{N-1} x_n e^{-j \frac{2\pi n k}{N}} \quad k = 0, \dots, N-1$$

- The factor  $e^{-j 2\pi / N}$  is written as  $W_N$ , so that :

$$X_1(k) = \sum_{n=0}^{N-1} x_n W_N^{k n} \quad , k = 0, \dots, N-1$$

# FFT Algorithm (cont'd)

- Remember these useful property :

$$W_N = e^{-j2\pi/N}$$
$$W_N^2 = W_{N/2}$$
$$W_N^{(k+N/2)} = -W_N^k$$

Find out where  
do these come from

- The data sequences  $X_1$  is divided into two equal sequences (even and odd )  $X_{11}$  and  $X_{12}$

## FFT Algorithm (cont'd)

- Because of those property, we can say that the DFT  $X_1(k)$  can be expressed in terms of two DFTs :  $X_{11}(k)$  and  $X_{12}(k)$  with the factor  $W_{N/2}^k$
- In the equation  $X_1(k) = X_{11}(k) + W_{N/2}^k \cdot X_{12}(k)$
- The number of  $k$  is only from 0 to  $N-1$
- The factor  $W_N^k$  needs calculation once only

# The Butterfly Method

- The FFT algorithm is applied in Butterfly method
- This method gives a very simple calculation to determine the value of DFT
- You can compare the time that you need for having that value between '*DFT classic method*' and '*FFT Butterflies method*'
- The point is you have to know exactly the FFT butterflies, at least for 8-point DFT on the figure below :

# The FFT Butterflies for an 8-point DFT

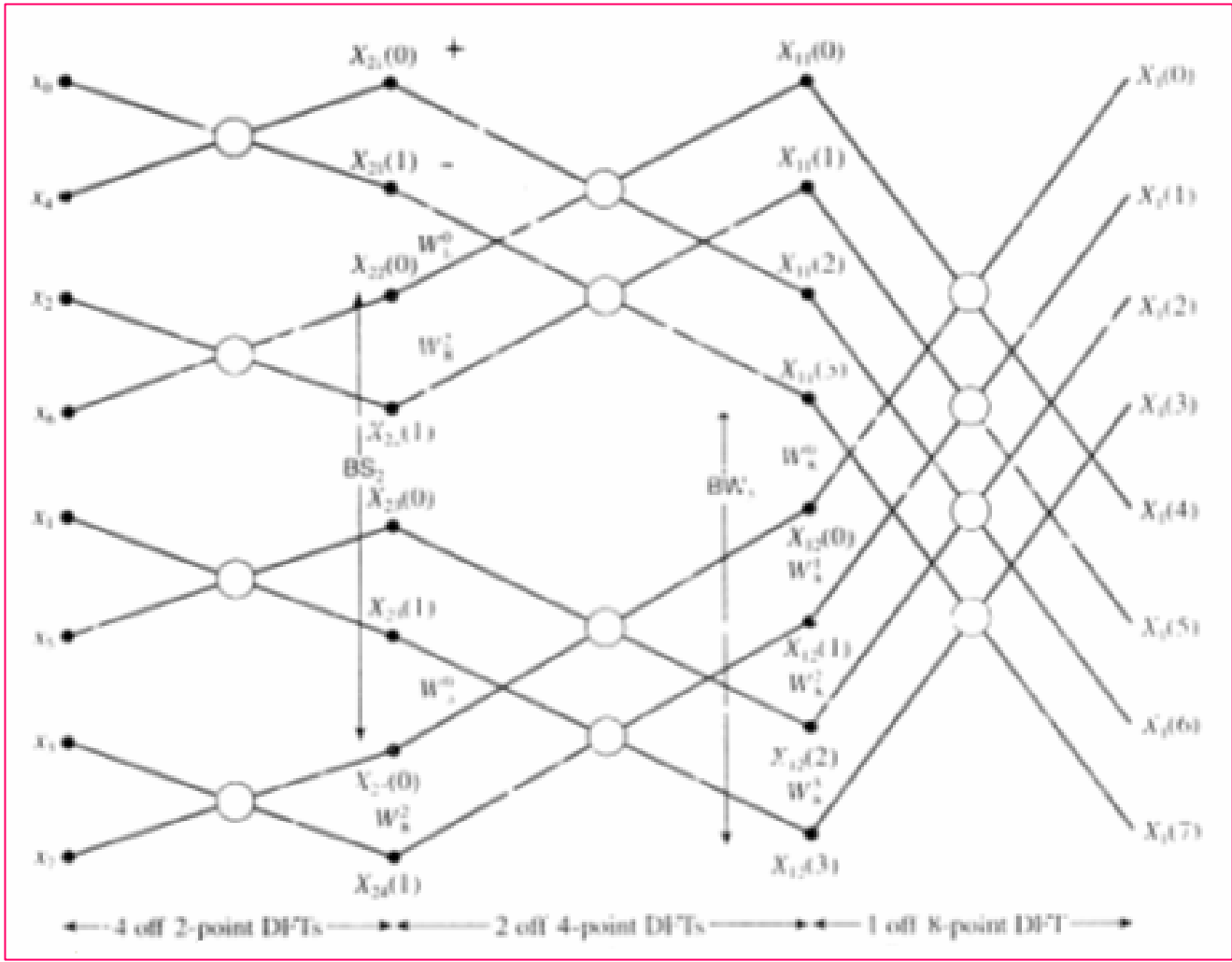


Figure 4.3

## Example for 4-point DFT

Find out the DFT of a sequence  $\{1, 0, 0, 1\}$

Answer

Note that its only 4-point DFT, so that from the previous figure points  $x_0, x_4, x_2, x_6$  are replaced by  $x_0, x_2, x_1, x_3$  and the require DFT values are :

$$X_{11}(0), X_{11}(1), X_{11}(2), X_{11}(3),$$

Therefore our calculation only up to second step



## Example for 4-point DFT (cont'd)

$$X_{21}(0) = x_0 + x_2 = 1$$

$$X_{21}(1) = x_0 - x_2 = 1$$

$$X_{22}(0) = x_1 + x_3 = 1$$

$$X_{22}(1) = x_1 - x_3 = -1$$

$$X_{11}(0) = X_{21}(0) + W_8^0 X_{22}(0) = 1 + 1 = 2$$

$$X_{11}(1) = X_{21}(1) + W_8^2 X_{22}(1) = 1 + e^{-j\pi/2}(-1) = 1 + j$$

$$X_{11}(2) = X_{21}(0) - W_8^0 X_{22}(0) = 1 - 1 = 0$$

$$X_{11}(3) = X_{21}(1) - W_8^2 X_{22}(1) = 1 - j$$

## Example for 4-point DFT (cont'd)

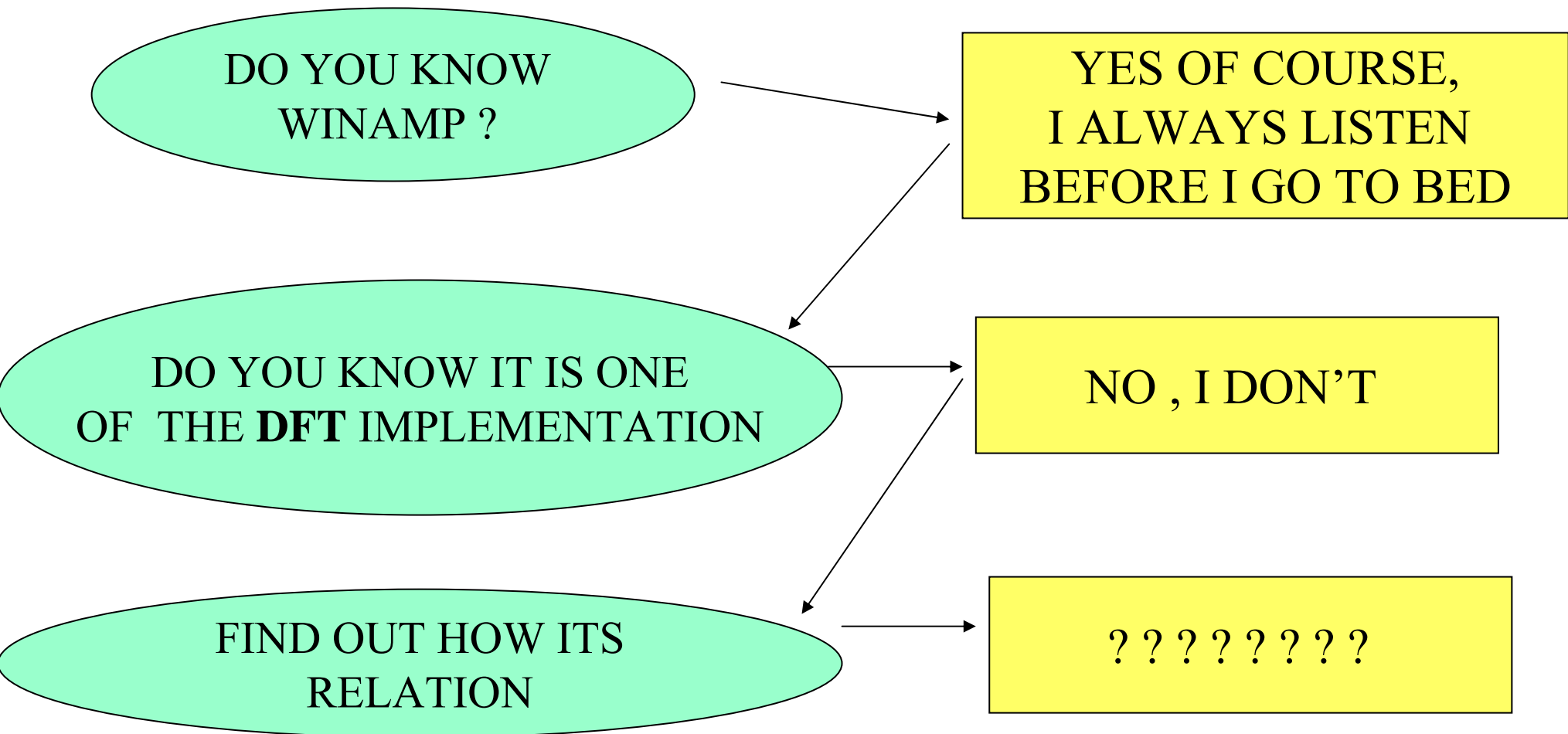
From those value we know that those are the same as the value that we determine from the 'DFT classical method'

So that, we may say : By using FFT, the computational savings, increase as the number of data increases



What about if you try by using computer program such as C++ or MATLAB ?


# Applications of the DFT and FFT



# End of this session



PREPARE  
FOR  
THE REVIEW



Before that...  
ARE YOU INTEREST WITH THIS TOPIC ?

# Review

1. Do the exercise on text book [ *Ifeachor* ] page 158-160. The number that you've to do is related to your absence number, but now its reversed
2. Repeat example 3.3 on text book [*Ifeachor*] page 114, by using MATLAB. Print out the result.

You have to be able for using other program,  
not only MATLAB