

# Advance Algorithm Analysis

(COMP 502 A)

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# Polynomial operations and representation

A polynomial  $A(x)$  can be written in the following forms:

$$\begin{aligned} A(x) &= a_0 + a_1x + a_2x^2 + \cdots + a_{n-1}x^{n-1} \\ &= \sum_{k=0}^{n-1} a_k x^k \\ &= \langle a_0, a_1, a_2, \dots, a_{n-1} \rangle \quad (\text{coefficient vector}) \end{aligned}$$

The degree of  $A$  is  $n - 1$ .

# Operations on Polynomials

- Evaluation
  - **Horner's Rule:**  $A(x) = a_0 + x(a_1 + x(a_2 + \cdots x(a_{n-1}) \cdots))$ . At each step, a sum is evaluated, then multiplied by  $x$ , before beginning the next step. Thus  $O(n)$  multiplications and  $O(n)$  additions are required.
- **Addition:** Given two polynomials  $A(x)$  and  $B(x)$ , compute  $C(x) = A(x) + B(x)$  ( $\forall x$ ). This takes  $O(n)$  time using basic arithmetic, because  $c_k = a_k + b_k$ .
- **Multiplication:** Given two polynomials  $A(x)$  and  $B(x)$ , compute  $C(x) = A(x) \cdot B(x)$  ( $\forall x$ ).

The degree of the resulting polynomial is twice that of  $A$  or  $B$ . This multiplication is then equivalent to a convolution of the vectors  $A$  and  $\text{reverse}(B)$ .

$$c_k = \sum_{j=0}^k a_j b_{k-j} \text{ for } 0 \leq k \leq 2(n-1).$$

# Multiplication

Naive polynomial multiplication takes  $O(n^2)$ .

$A(x) = 6x^3 + 7x^2 - 10x + 9$  and  $B(x) = -2x^3 + 4x - 5$  as follows:

$$\begin{array}{r} 6x^3 + 7x^2 - 10x + 9 \\ - 2x^3 \phantom{+ 7x^2 - 10x + 9} \\ \hline - 30x^3 - 35x^2 + 50x - 45 \\ 24x^4 + 28x^3 - 40x^2 + 36x \\ - 12x^6 - 14x^5 + 20x^4 - 18x^3 \\ \hline - 12x^6 - 14x^5 + 44x^4 - 20x^3 - 75x^2 + 86x - 45 \end{array}$$

# Multiplication (divide-and-conquer approach)

The *Karatsuba algorithm* is a divide-and-conquer approach for fast multiplication of two polynomials (or integers). It significantly improves the naive  $O(n^2)$  complexity to about  $O(n^{\log_2 3}) \approx O(n^{1.585})$ . Here's how it works:

$$A(x) = 3x^1 + 2x^0$$

$$A_{\text{high}} = 3 \text{ and } A_{\text{low}} = 2$$

$$B(x) = 5x^1 + 4x^0$$

$$B_{\text{high}} = 5 \text{ and } B_{\text{low}} = 4$$

$Z_0$ : Multiply the low parts:

$$Z_0 = A_{\text{low}} \cdot B_{\text{low}} = 2 \cdot 4 = 8$$

$Z_2$ : Multiply the high parts:

$$Z_2 = A_{\text{high}} \cdot B_{\text{high}} = 3 \cdot 5 = 15$$

$Z_1$ : Multiply the sum of the parts:

$$Z_1 = (A_{\text{high}} + A_{\text{low}}) \cdot (B_{\text{high}} + B_{\text{low}})$$

Calculating the sums:

$$(3 + 2) \cdot (5 + 4) = 5 \cdot 9 = 45$$

Now, subtract  $Z_0$  and  $Z_2$  from this result:

$$Z_1 = 45 - 8 - 15 = 22$$

$$P(x) = Z_2 \cdot x^2 + Z_1 \cdot x^1 + Z_0$$

$$P(x) = 15x^2 + 22x + 8$$

Today, we will compute the product in  
 $O(n \cdot \log(n))$   
time via Fast Fourier Transform!

# Representation of polynomials

- Coefficient Representation

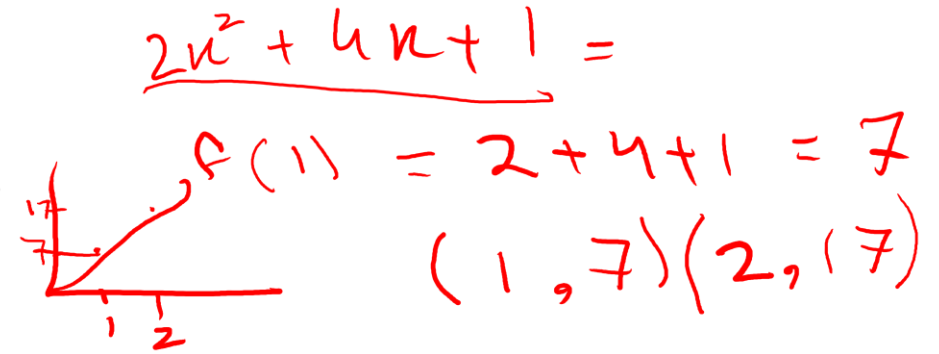
- coefficient vectors  $a = (a_0, a_1, \dots, a_{n-1})$  and  $b = (b_0, b_1, \dots, b_{n-1})$

- Roots and a scale term •  $A(x) = (x - r_0) \cdot (x - r_1) \cdots (x - r_{n-1}) \cdot c$

However, it is impossible to find exact roots with only basic arithmetic operations and kth root operations.

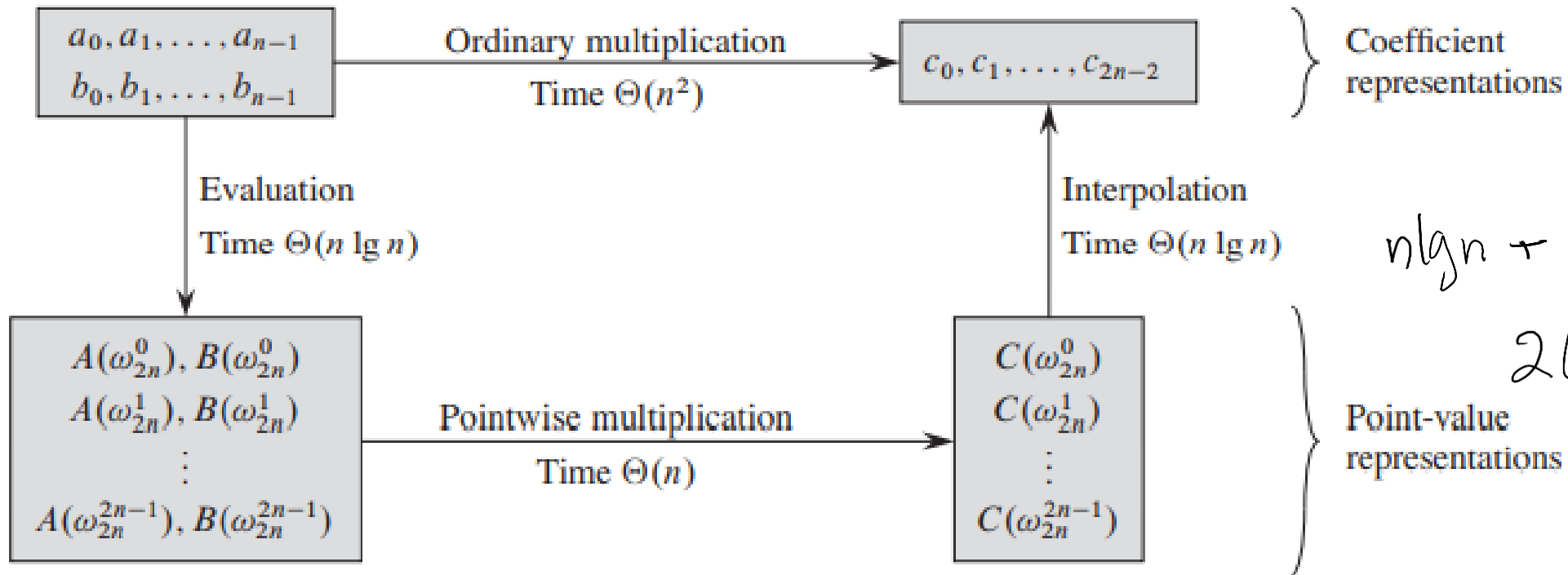
- Point Value Representation (Samples)

$(x_0, y_0), (x_1, y_1), \dots, (x_{n-1}, y_{n-1})$  with  $A(x_i) = y_i \ (\forall i)$



Representation	Evaluation	Addition	Multiplication
Coefficient	$O(n)$	$O(n)$	$O(n^2)$
Point-Value (Samples)	$O(n^2)$	$O(n)$	$O(n)$

# Big Idea



$$\begin{aligned}
 &n \lg n + n + n \lg n \\
 &2(n \lg n) + n \\
 &\Theta(n \lg n)
 \end{aligned}$$

Representation	Evaluation	Addition	Multiplication
Coefficient	$O(n)$	$O(n)$	$O(n^2)$
Point-Value (Samples)	$O(n^2)$	$O(n)$	$O(n)$



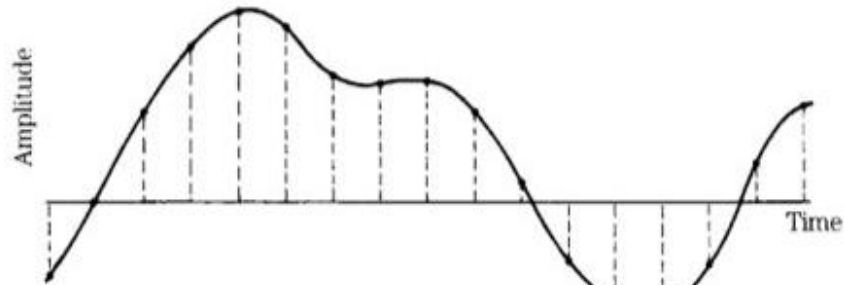
# Fast Polynomial Multiplication

In order to compute the product of two polynomials  $A$  and  $B$ , we can perform the following steps.

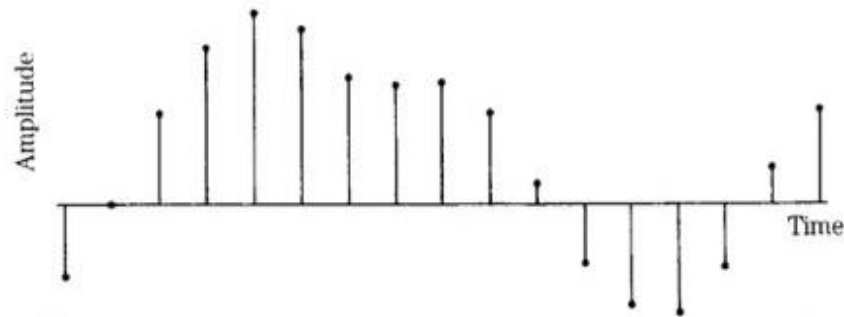
1. Compute  $A^* = FFT(A)$  and  $B^* = FFT(B)$ , which converts both  $A$  and  $B$  from coefficient vectors to a sample representation.
2. Compute  $C^* = A^* \cdot B^*$  in sample representation in linear time by calculating  $C_k^* = A_k^* \cdot B_k^* (\forall k)$ .
3. Compute  $C = IFFT(C^*)$ , which is a vector representation of our final solution.

FFT (Fast Fourier Transform) and IFFT (Inverse Fast Fourier Transform) perform the conversion between Coefficient to Samples and Samples to coefficient representation respectively.

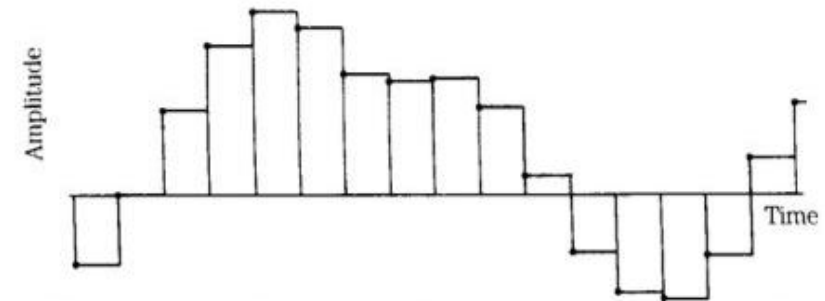
# Continuous and Discrete Signals



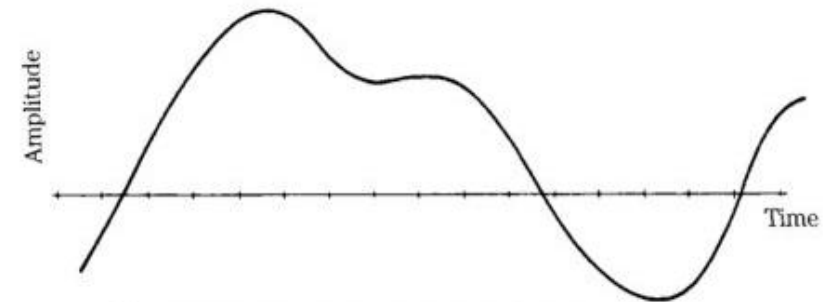
**A** The input analog signal is sampled.



**B** The numerical values of these samples are stored or transmitted (after quantization).



**C** Samples are held to form a staircase representation of the signal.



**D** An output lowpass filter interpolates the staircase to reconstruct the input waveform.

**2-1** With discrete time sampling, a bandlimited signal can be sampled and reconstructed without loss because of sampling.

# Fast Fourier Transform & Inverse Fast Fourier Transform

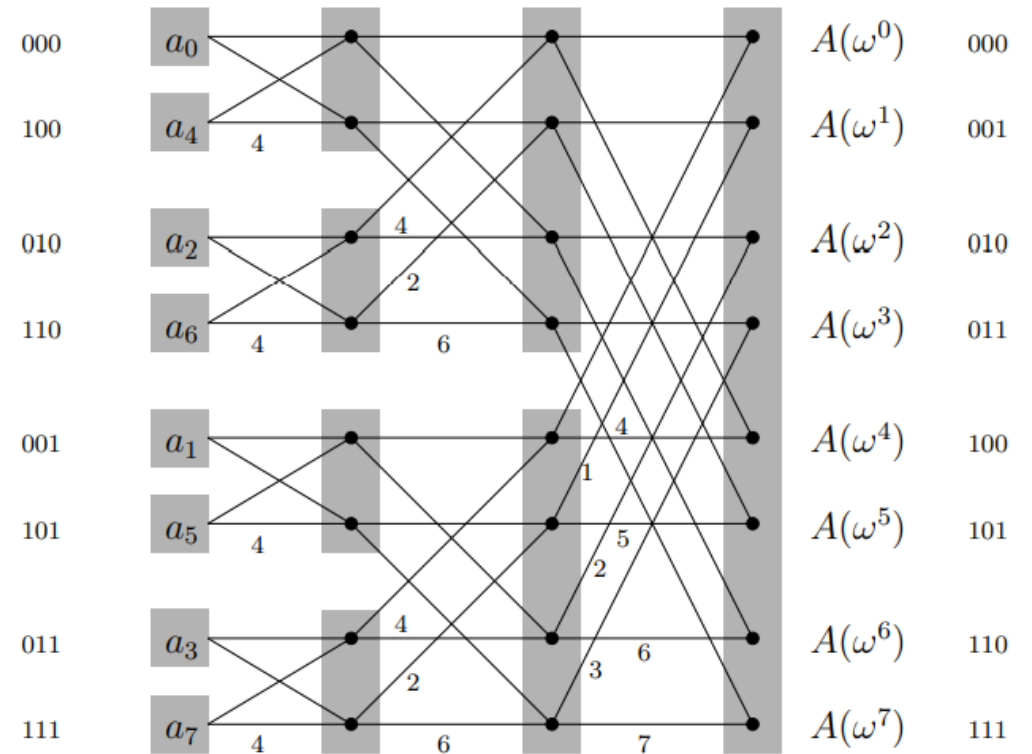
- FFT  $x(n) \rightarrow X(k)$
- IFFT  $X(k) \rightarrow x(n)$
- DIT > Decimation in Time  $n$
- DIF > Decimation in Frequency  $k$

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

$N=4$  (0,1,2,3) **even:** 0,2 **odd:** 1,3

$N=8$  (0,1,2,3,4,5,6,7) **Even:** 0,4,2,6 **Odd:** 1,5,3,7

0, 2, 4, 6  
1, 3, 5, 7



# Twiddle Factor

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

$$W_4^0 = e^{\frac{-j2\pi \cdot 0}{4}} = e^0 = 1$$

$$W_8^0 = e^{\frac{-j2\pi \cdot 0}{8}} = 1$$

$$W_4^1 = e^{\frac{-j2\pi \cdot 1}{4}} = \cos\theta - j\sin\theta$$

$$W_8^0 = 1$$

$$e^{j2\theta} = \cos\theta + j\sin\theta$$

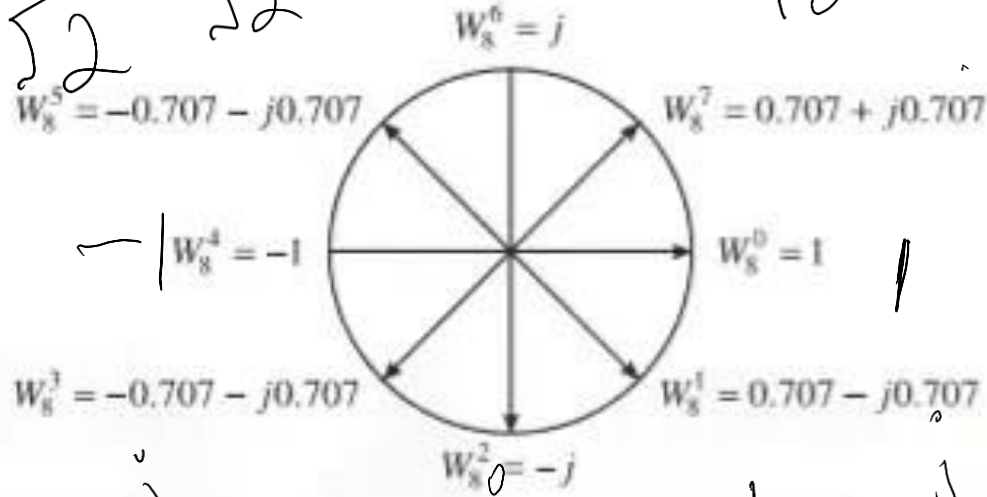
$$W_8^7 =$$

$$W_8^7 = j$$

$$\frac{8}{8 \overline{) 70} \underline{64} 6}$$

$$-\frac{1}{\sqrt{2}} + j\frac{1}{\sqrt{2}}$$

$$\frac{1}{\sqrt{2}} + j\frac{1}{\sqrt{2}}$$



$$-\frac{1}{\sqrt{2}} - j\frac{1}{\sqrt{2}}$$

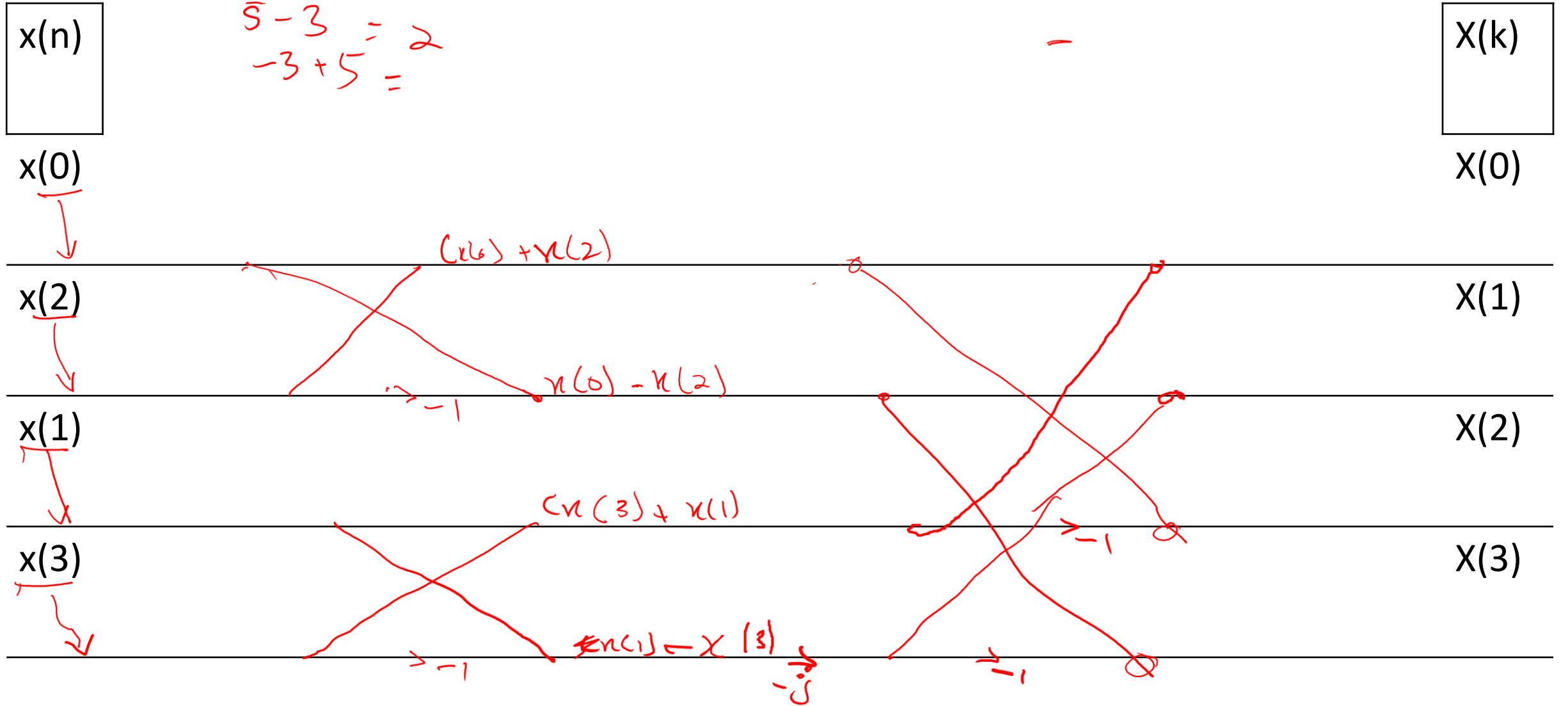
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$$\frac{1}{\sqrt{2}} - j\frac{1}{\sqrt{2}}$$

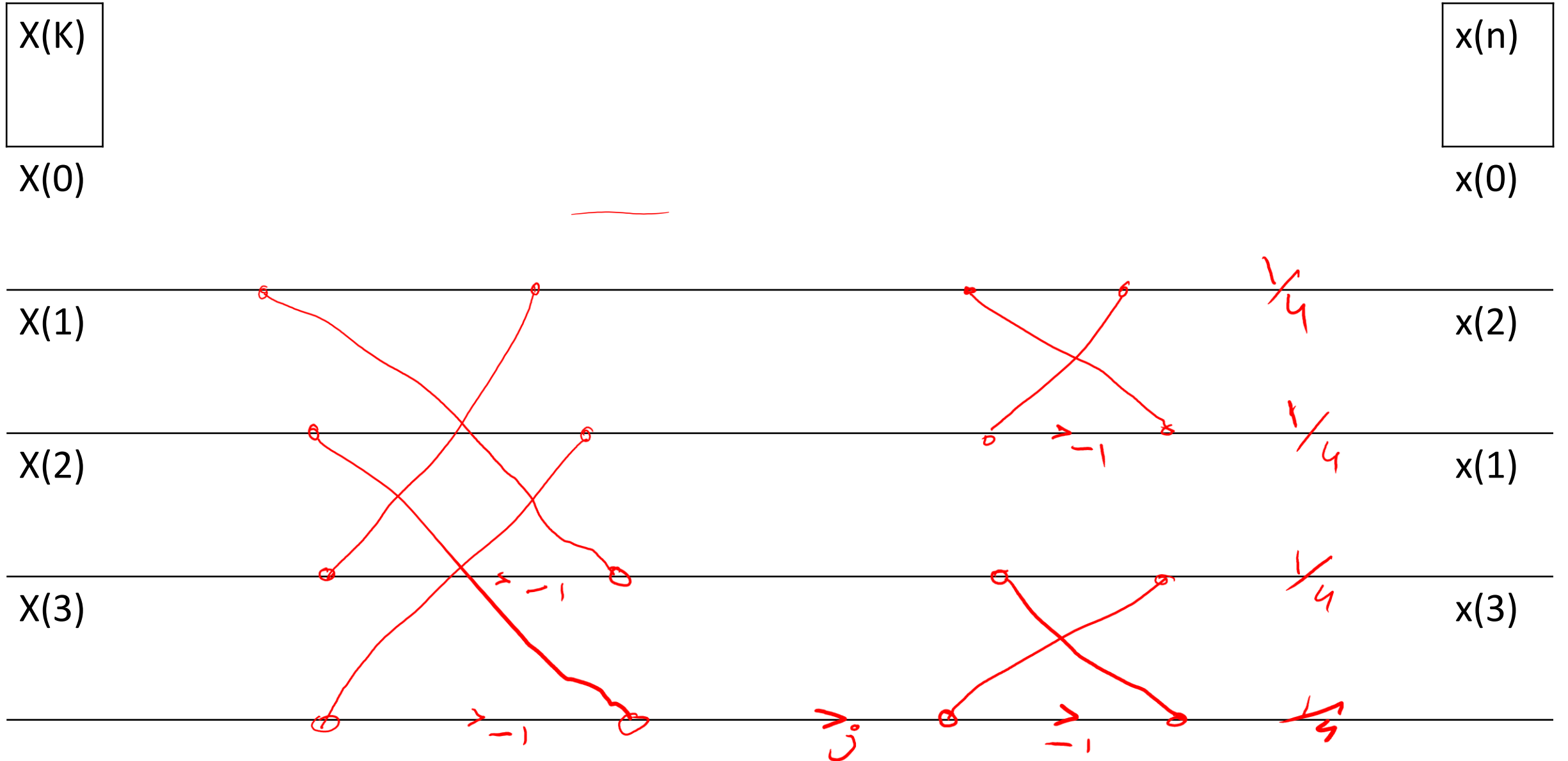
$$e^{j\theta} = \cos\theta + j\sin\theta$$

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

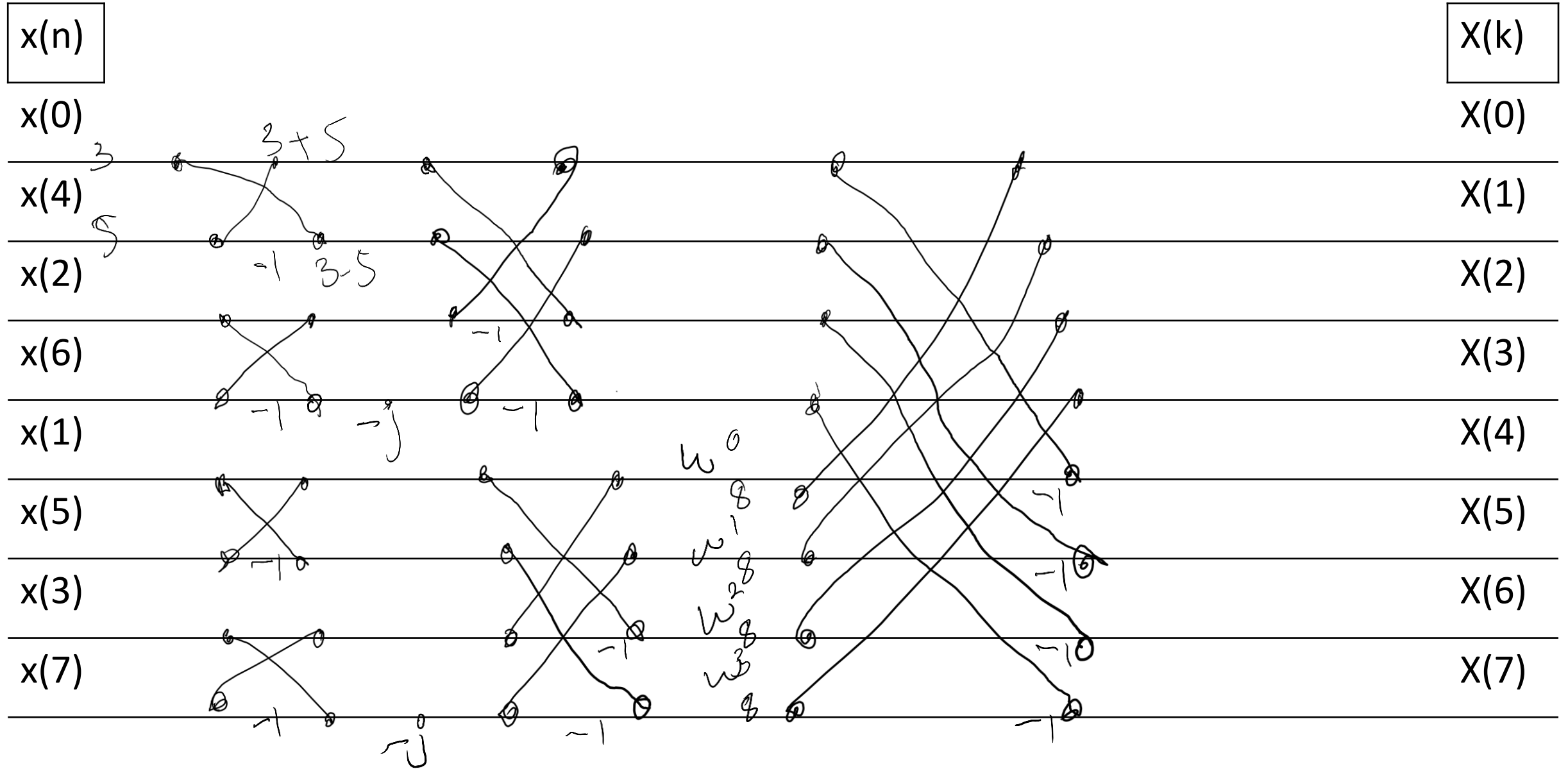
# DIT $x(n) \rightarrow X(k)$ FFT ButterFly Diagram



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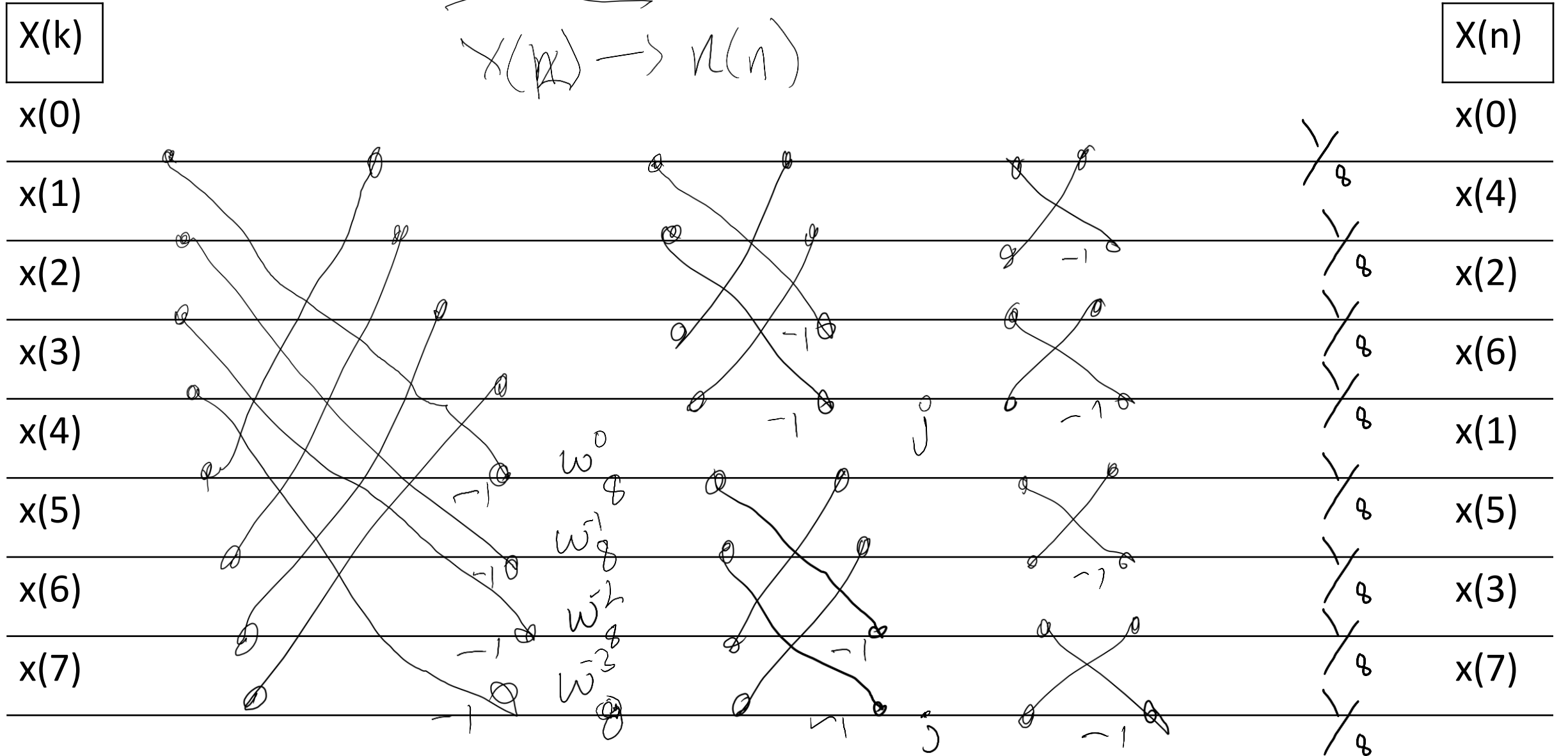


# DIT $x(n) \rightarrow X(k)$ FFT ButterFly Diagram



# DIT $X(k) \rightarrow$ IFFT ButterFly Diagram

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

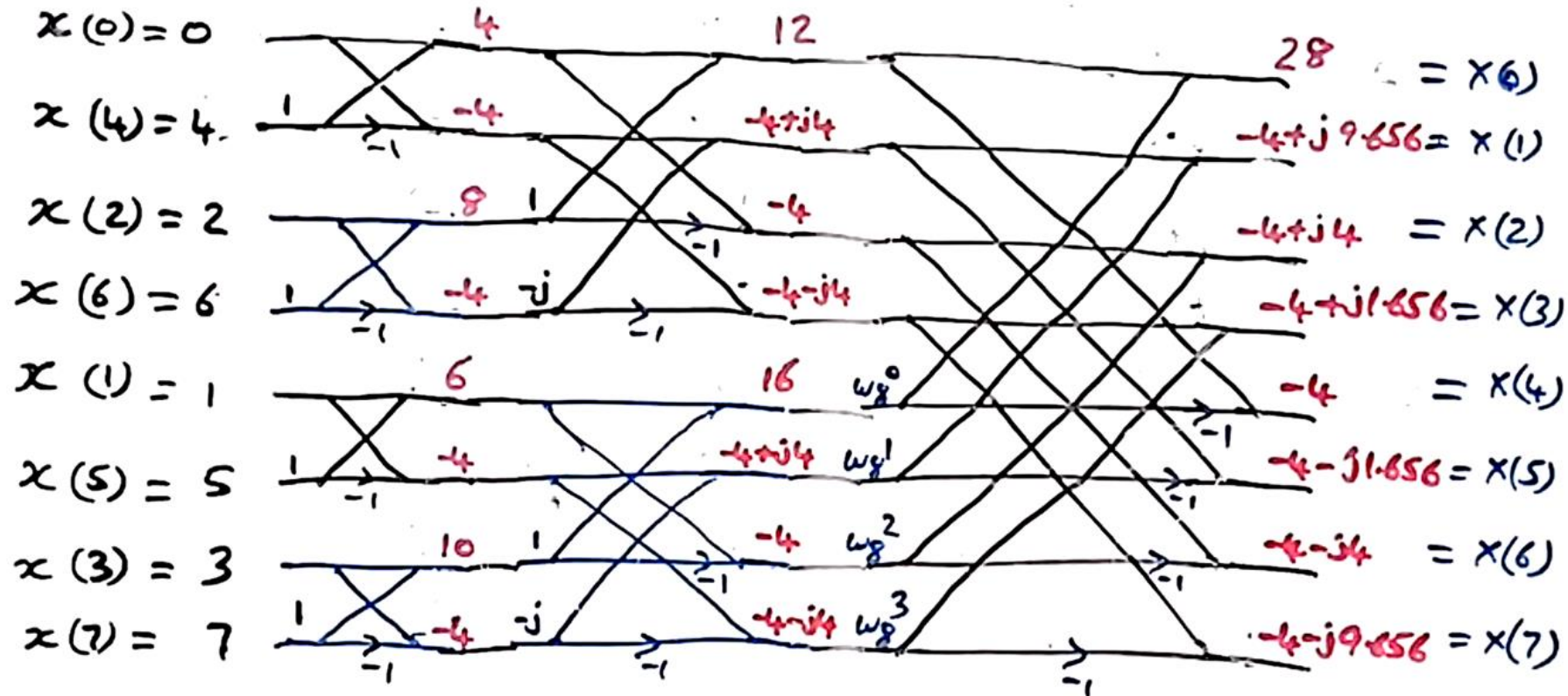




# DIT FFT

$$\omega_8^0 = e^0 = 1$$

$$\omega_8^1 = e^{-j\frac{2\pi}{8}(1)} = \cos \frac{\pi}{4} - j \sin \frac{\pi}{4} =$$



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

$$X(k) = \{28, \underline{-4 + j9.656}, \underline{-4 + j4}, \underline{-4 + j1.656}, -4, -4 - j1.656, -4 - j4, -4 - j9.656\}$$

# DIT $X(k) \rightarrow$ IFFT ButterFly Diagram

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

