

# Polynomial Representations

(35)

① Coefficient Form: A polynomial  $f(n) = a_0 + a_1 n + \dots + a_{n-1} n^{n-1}$  is represented as  $(a_0, a_1, \dots, a_{n-1})$ .

Examples:  $x+2$  is represented as  $(1, 2)$ ,  $2x+3$  is represented as  $(2, 3)$ , and  $2x^2+7x+6$  is represented as  $(6, 7, 2)$ .

Complexity of adding two polynomials  $(a_0, a_1, \dots, a_{n-1})$  and  $(b_0, b_1, \dots, b_{n-1})$  is  $\Theta(n)$  because

$$(a_0, a_1, \dots, a_{n-1}) + (b_0, b_1, \dots, b_{n-1}) = (a_0 + b_0, a_1 + b_1, \dots, a_{n-1} + b_{n-1})$$

Example:  $(1, 2) + (2, 3) = (3, 5)$

Complexity of multiplying a polynomial  $(a_0, a_1, \dots, a_{n-1})$  with a constant  $c_0$  is  $\Theta(n)$  because

$$c_0(a_0, a_1, \dots, a_{n-1}) = (c_0 a_0, c_0 a_1, \dots, c_0 a_{n-1})$$

Example:  $5(1, 2) = (5, 10)$

Complexity of multiplying two polynomials  $(a_0, a_1, \dots, a_{n-1})$

and  $(b_0, b_1, \dots, b_{n-1})$  is  $\Theta(n^2)$  because

$$(a_0, a_1, \dots, a_{n-1}) \times (b_0, b_1, \dots, b_{n-1})$$

$= (a_0 b_0, a_0 b_1 + a_1 b_0, \dots, a_{n-1} b_{n-1})$  using the convolution algorithm (Iterative Multiply).

Example:  $(1, 2) \times (2, 3) = (1 \times 2, 1 \times 3 + 2 \times 2, 2 \times 3)$

$$= (2, 7, 6)$$

(36)

(2) Point Value Form : A polynomial  $f(n) = a_0 + a_1 n + \dots + a_{m-1} n^{m-1}$   
 is represented as  $\{(n_0, f(n_0)), (n_1, f(n_1)), \dots, (n_{m-1}, f(n_{m-1}))\}$   
 where all  $n_i$ 's are different.

Examples:  $x+2$  is represented as  $\{(0, 2), (1, 3)\}$ ,  
 $2x+3$  is represented as  $\{(2, 7), (3, 9)\}$ , and  
 $2x^2+7x+6$  is represented as  $\{(-1, 1), (0, 6), (1, 15)\}$   
 complexity of adding two polynomials  $\{(n_0, f_1(n_0)), (n_1, f_1(n_1)), \dots, (n_{m-1}, f_1(n_{m-1}))\}$  and  $\{(n_0, f_2(n_0)), (n_1, f_2(n_1)), \dots, (n_{m-1}, f_2(n_{m-1}))\}$   
 is  $\Theta(m)$  because

$$\begin{aligned} & \{ (n_0, f_1(n_0)), (n_1, f_1(n_1)), \dots, (n_{m-1}, f_1(n_{m-1})) \} + \\ & \{ (n_0, f_2(n_0)), (n_1, f_2(n_1)), \dots, (n_{m-1}, f_2(n_{m-1})) \} \\ = & \{ (n_0, f_1(n_0) + f_2(n_0)), (n_1, f_1(n_1) + f_2(n_1)), \dots, \\ & (n_{m-1}, f_1(n_{m-1}) + f_2(n_{m-1})) \} \end{aligned}$$

Complexity of multiplying a constant  $c_0$  with a polynomial  
 $\{(n_0, f(n_0)), (n_1, f(n_1)), \dots, (n_{m-1}, f(n_{m-1}))\}$  is  $\Theta(m)$   
 because  $c_0 \{ (n_0, f(n_0)), (n_1, f(n_1)), \dots, (n_{m-1}, f(n_{m-1})) \}$

$$= \{ (n_0, c_0 f(n_0)), (n_1, c_0 f(n_1)), \dots, (n_{m-1}, c_0 f(n_{m-1})) \}$$

Complexity of multiplying two polynomials

$\{(n_0, f_1(n_0)), (n_1, f_1(n_1)), \dots, (n_{m-1}, f_1(n_{m-1}))\}$  and  
 $\{(n_0, f_2(n_0)), (n_1, f_2(n_1)), \dots, (n_{m-1}, f_2(n_{m-1}))\}$  is  
 $\Theta(m)$  because

$$\left\{ (x_0, f_1(x_0)), (x_1, f_1(x_1)), \dots, (x_{\frac{n}{2}-2}, f_1(x_{\frac{n}{2}-2})) \right\} \times \textcircled{37} \\ \left\{ (x_0, f_2(x_0)), (x_1, f_2(x_1)), \dots, (x_{\frac{n}{2}-2}, f_2(x_{\frac{n}{2}-2})) \right\} = \\ \left\{ (x_0, f_1(x_0) f_2(x_0)), (x_1, f_1(x_1) f_2(x_1)), \dots, \right. \\ \left. (x_{\frac{n}{2}-1}, f_1(x_{\frac{n}{2}-2}) f_2(x_{\frac{n}{2}-2})) \right\}$$

Evaluation using Horner's Rule : We can convert a polynomial  $f(n)$  from coefficient form  $(a_0, a_1, \dots, a_{n-1})$  to point value form  $\{(x_0, f(x_0)), (x_1, f(x_1)), \dots, (x_{n-1}, f(x_{n-1}))\}$  in  $O(n^2)$  time using Horner's Rule for polynomial evaluation.

$$a_0 + a_1 x + \dots + a_{n-2} x^{n-2} + a_{n-1} x^{n-1} =$$

$$a_0 + x(a_1 + x(a_2 + \dots + x(a_{n-2} + x(a_{n-1}) \dots)))$$

Example :  $6 + 7x + 2x^2 = 6 + x(7 + x(2))$

Horner's Rule takes  $O(n)$  time. Applying it  $n$  times will take  $O(n^2)$  time.

Interpolation using Gaussian Elimination : We can convert a polynomial  $f(n)$  from point value form  $\{(x_0, f(x_0)), (x_1, f(x_1)), \dots, (x_{n-1}, f(x_{n-1}))\}$  to coefficient form  $(a_0, a_1, \dots, a_{n-1})$  in  $O(n^3)$  time using the Gaussian Elimination Algorithm.

Example: let's take  $f(x) = \{(-1, 1), (0, 6), (1, 15)\}$

$$= (a_0, a_1, a_2) = a_0 + a_1x + a_2x^2$$

write the ~~three~~ linear equations as in three unknowns

$$a_0 - a_1 + a_2 = 1 \rightarrow (1)$$

$$a_0 = 6 \rightarrow (2)$$

$$a_0 + a_1 + a_2 = 15 \rightarrow (3)$$

First we eliminate  $a_0$ . From equation (2), we put the value of  $a_0 = 6$  in equations (1) and (3) to get two linear equations in two unknowns:

$$6 - a_1 + a_2 = 1 \Leftrightarrow a_1 - a_2 = 5 \rightarrow (4)$$

$$6 + a_1 + a_2 = 15 \Leftrightarrow a_1 + a_2 = 9 \rightarrow (5)$$

Next we eliminate  $a_1$ . From equation (4), we put the value of  $a_1 = a_2 + 5$  in equation (5) to get one linear equation in one unknown:

$$a_2 + 5 + a_2 = 9 \Leftrightarrow 2a_2 = 4 \Rightarrow a_2 = 2 \rightarrow (6)$$

$\Rightarrow a_1 = a_2 + 5 = 7$  and  $a_0 = 6$ . The coefficient form is  $(6, 7, 2)$ .

# Divide and Conquer Graph

(39)

$$\begin{aligned}a_0 - a_1 + a_2 &= 1 \\a_0 &= 6 \\a_0 + a_1 + a_2 &= 15\end{aligned}$$

Divide Step

$$\begin{aligned}a_1 - a_2 &= 5 \\a_1 + a_2 &= 9\end{aligned}$$

$$a_2 = 2$$

Conquer Step

$$\begin{aligned}a_1 &= 7 \\a_0 &= 6\end{aligned}$$

## Complexity of Gaussian Elimination Algorithm.

40

Let  $T(n)$  denote the complexity of Gaussian Elimination Algorithm:

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \rightarrow (1)$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \rightarrow (2)$$

:

$$a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in}x_n = b_i \rightarrow (i)$$

:

:

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n \rightarrow (n)$$

First we eliminate  $x_n$ . From equation (n), we put

$$\text{the value of } x_n = \frac{b_n}{a_{nn}} - \frac{a_{n1}}{a_{nn}} x_1 - \frac{a_{n2}}{a_{nn}} x_2 - \dots - \frac{a_{n(n-1)}}{a_{nn}} x_{n-1}$$

in equations (1), (2), ..., (n-1) to get n-1 linear equations  
in n-1 unknowns:

$$a_{11}'x_1 + a_{12}'x_2 + \dots + a_{1(n-1)}'x_{n-1} = b_1' \rightarrow (1)'$$

$$a_{21}'x_1 + a_{22}'x_2 + \dots + a_{2(n-1)}'x_{n-1} = b_2' \rightarrow (2)'$$

:

:

$$a_{i1}'x_1 + a_{i2}'x_2 + \dots + a_{i(n-1)}'x_{n-1} = b_i' \rightarrow (i)'$$

:

:

$$a_{(n-1)1}'x_1 + a_{(n-1)2}'x_2 + \dots + a_{(n-1)(n-1)}'x_{n-1} = b_{n-1}' \rightarrow (n-1)'$$

Complexity of Divide Step involves putting the value of  $x_n$  in equations (1), (2), ..., (n-1) and simplifying them to get the equations (1)', (2)', ..., (n-1)'. Total complexity of Divide Step =  $(n-1) \times c_m$

$$= c_m(n-1)$$

(41)

Complexity of conquer step involves computing  
the value of  $x_n$ , given the values of  $x_1, x_2, \dots, x_{n-1}$ ,

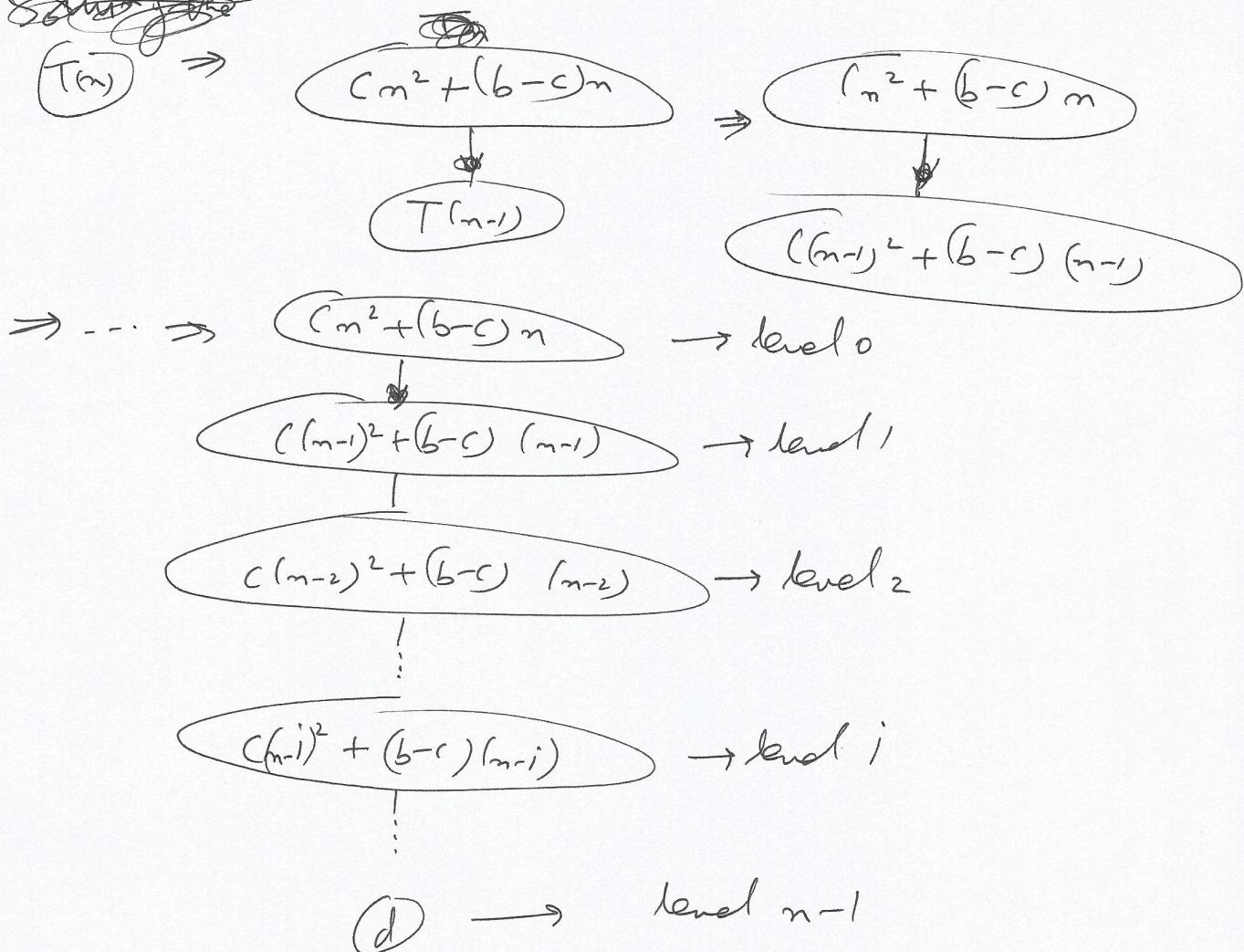
$$= \Theta(b^n)$$

There is only one subproblem of size  $n-1$ .

$$\Rightarrow T(n) = T(n-1) + Cn(n-1) + b^n \text{ for } n > 1$$

$$T(1) = d$$

~~Solve by tree~~



$$\begin{aligned}
 T(n) &= ((n^2 + (b-c)n) + ((n-1)^2 + (b-c)(n-1)) + ((n-2)^2 + (b-c)(n-2)) \\
 &\quad + \dots + ((n-i)^2 + (b-c)(n-i)) + \dots + (C \cdot 2^2 + (b-c) \cdot 2) + d \\
 &= \frac{Cn(n+1)(2n+1)}{6} - C + \frac{(b-c)2n(n+1)}{2} - (b-c) + d \\
 &= \Theta(n^3)
 \end{aligned}$$

## Discrete Fourier Transform (DFT):

(42)

Consider a polynomial  $(a_0, a_1, \dots, a_{n-1}) = f(n)$ . Let  $w_n = e^{i\frac{2\pi}{n}}$  be  $n^{\text{th}}$  root of unity. We consider the point value form of  $f(n)$  at the points  $\{(w_n^0, f(w_n^0)), (w_n^1, f(w_n^1)), \dots, (w_n^{n-1}, f(w_n^{n-1}))\}$  and define the DFT of the vector  $(a_0, a_1, \dots, a_{n-1})$  as the ~~vector~~ (corresponding to the coefficient form of  $f(n)$ ) as the vector  $(f(w_n^0), f(w_n^1), \dots, f(w_n^{n-1}))$  (corresponding to the point value form).

Example:  $\text{DFT}(1, 2) = (1+2, 1-2) = (3, -1)$

$\text{DFT}(2, 3) = (2+3, 2-3) = (5, -1)$ , and

$$\begin{aligned} \text{DFT}(6, 7, 2) &= (6+7+2, 6+7\left(-\frac{1+\sqrt{3}i}{2}\right) + 2\left(-\frac{1-\sqrt{3}i}{2}\right), \\ &\quad 6+7\left(\frac{-1-\sqrt{3}i}{2}\right) + 2\left(\frac{-1+\sqrt{3}i}{2}\right)) \end{aligned}$$

$$= \left(15, \frac{3+5\sqrt{3}i}{2}, \frac{3-5\sqrt{3}i}{2}\right)$$

Complexity of computing DFT using Horner's Rule is  $\Theta(n^2)$ .

# Inverse Discrete Fourier Transform (DFT<sup>-1</sup>): (43)

Consider a polynomial  $(a_0, a_1, \dots, a_{n-1}) = f(x)$ . Let  $w_n = e^{j\frac{2\pi}{n}}$  be  $n^{\text{th}}$  root of unity. We consider the point value form of  $f(x)$  at the points  $\{(w_n^0, f(w_n^0)), (w_n^1, f(w_n^1)), \dots, (w_n^{n-1}, f(w_n^{n-1}))\}$  and consider the vector  $(y_0, y_1, \dots, y_{n-1})$   $= (f(w_n^0), f(w_n^1), \dots, f(w_n^{n-1}))$  (corresponding to the point value form). We define the DFT<sup>-1</sup> of the vector  $(y_0, y_1, \dots, y_{n-1})$  as the vector  $(a_0, a_1, \dots, a_{n-1})$  (corresponding to the coefficient form of  $f(x)$ ).

Examples :  $\text{DFT}^{-1}(3, -1) = (1, 2)$ ,

$\text{DFT}^{-1}(5, -1) = (2, 3)$ , and

$$\text{DFT}^{-1}\left(15, \frac{3+5\sqrt{2}i}{2}, \frac{3-5\sqrt{2}i}{2}\right) = (6, 7, 2)$$

We write the DFT equation as :

$$\begin{bmatrix} y_0 \\ y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_{n-1} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & \cdots & 1 \\ 1 & w_n & w_n^2 & \cdots & w_n^{n-1} \\ 1 & w_n^2 & w_n^4 & \cdots & w_n^{2(n-1)} \\ 1 & w_n^3 & w_n^6 & \cdots & w_n^{3(n-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & w_n^{n-1} & w_n^{2(n-1)} & \cdots & w_n^{(n-1)(n-1)} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \\ \vdots \\ a_{n-1} \end{bmatrix} = A \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix}$$

We multiply both sides by the Matrix

$$\begin{bmatrix} 1 & 1 & 1 & \cdots & 1 \\ 1 & w_n^{-1} & w_n^{-2} & \cdots & w_n^{-n(n-1)} \\ 1 & w_n^{-2} & w_n^{-4} & \cdots & w_n^{-2(n-1)} \\ 1 & w_n^{-3} & w_n^{-6} & \cdots & w_n^{-3(n-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & w_n^{-(n-1)} & w_n^{-2(n-1)} & \cdots & w_n^{-(n-1)(n-1)} \end{bmatrix} = \beta$$

(44)

$$B \begin{bmatrix} y_0 \\ y_1 \\ \vdots \\ y_{n-1} \end{bmatrix} = BA \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix} = C \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix}$$

(i,j)<sup>th</sup> entry of BA is given by

$$\begin{aligned} c_{ij} &= 1 \times 1 + w_n^{-i} \times w_n^j + w_n^{-2i} \times w_n^{2j} + \dots + w_n^{-i(n-1)} \times w_n^{j(n-1)} \\ &= 1 + w_n^{(j-i)} + w_n^{2(j-i)} + \dots + w_n^{(m-1)(j-i)} \end{aligned}$$

for  $i=j$  we get:

$$c_{ii} = 1 + 1 + \dots + 1 = n$$

for  $i \neq j$  we get:

$$c_{ij} = \frac{w_n^{(j-i)n}}{w_n^{(j-i)} - 1} = \frac{(w_n^n)^{j-i} - 1}{w_n^{j-i} - 1} = \frac{1 - 1}{w_n^{j-i} - 1} = 0$$

$$\Rightarrow C = nI = n \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{bmatrix}$$

$$\Rightarrow B \begin{bmatrix} y_0 \\ y_1 \\ \vdots \\ y_{n-1} \end{bmatrix} = nI \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix} = n \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{bmatrix} = \frac{1}{n} B \begin{bmatrix} y_0 \\ y_1 \\ \vdots \\ y_{n-1} \end{bmatrix} = DFT^{-1}(y_0, y_1, \dots, y_{n-1})$$

DFT and

$DFT^{-1}$  can also be computed in time  $O(n^2)$  using Horner's Rule or Matrix Multiplication

Examples: For  $n=2$ , we have  $\omega_2^{-1} = e^{-\frac{2\pi i}{2}} = e^{-i\pi} = -1$ , (45)

$$\text{DFT}^{-1}(3, -1) = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & \omega_2^{-1} \end{bmatrix} \begin{bmatrix} 3 \\ -1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 3 \\ -1 \end{bmatrix} = \begin{pmatrix} \frac{2}{2} \\ \frac{4}{2} \end{pmatrix} = (1, 2)$$

$$\text{DFT}^{-1}(5, -1) = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & \omega_2^{-1} \end{bmatrix} \begin{bmatrix} 5 \\ -1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 5 \\ -1 \end{bmatrix} = \begin{pmatrix} \frac{4}{2} \\ \frac{6}{2} \end{pmatrix} = (2, 3)$$

For  $n=3$ , we have  $\omega_3^{-1} = e^{-\frac{2\pi i}{3}} = \frac{-1-\sqrt{3}i}{2}$ ,  $\omega_3^{-2} = \frac{-1+\sqrt{3}i}{2}$

$$\text{DFT}^{-1}\left(15, \frac{3+5\sqrt{3}i}{2}, \frac{3-5\sqrt{3}i}{2}\right)$$

$$= \frac{1}{3} \begin{bmatrix} 1 & 1 & 1 \\ 1 & \omega_3^{-1} & \omega_3^{-2} \\ 1 & \omega_3^{-2} & \omega_3^{-4} \end{bmatrix} \begin{bmatrix} 15 \\ \frac{3+5\sqrt{3}i}{2} \\ \frac{3-5\sqrt{3}i}{2} \end{bmatrix}$$

$$= \frac{1}{3} \begin{bmatrix} 1 & 1 & 1 \\ 1 & \frac{-1-\sqrt{3}i}{2} & \frac{-1+\sqrt{3}i}{2} \\ 1 & \frac{-1+\sqrt{3}i}{2} & \frac{-1-\sqrt{3}i}{2} \end{bmatrix} \begin{bmatrix} 15 \\ \frac{3+5\sqrt{3}i}{2} \\ \frac{3-5\sqrt{3}i}{2} \end{bmatrix}$$

$$= \frac{1}{3} \begin{bmatrix} 15 + \frac{3+5\sqrt{3}i}{2} + \frac{3-5\sqrt{3}i}{2} \\ 15 + \left(\frac{-1-\sqrt{3}i}{2}\right)\left(\frac{3+5\sqrt{3}i}{2}\right) + \left(\frac{-1+\sqrt{3}i}{2}\right)\left(\frac{3-5\sqrt{3}i}{2}\right) \\ 15 + \left(\frac{-1+\sqrt{3}i}{2}\right)\left(\frac{3+5\sqrt{3}i}{2}\right) + \left(\frac{-1-\sqrt{3}i}{2}\right)\left(\frac{3-5\sqrt{3}i}{2}\right) \end{bmatrix}$$

$$= \frac{1}{3} \begin{bmatrix} 18 \\ 15 + \frac{-3+15}{2} \\ 15 + \frac{-3-15}{2} \end{bmatrix} = \begin{bmatrix} \frac{18}{3} \\ \frac{21}{3} \\ \frac{6}{3} \end{bmatrix} = \begin{pmatrix} 6 \\ 7 \\ 2 \end{pmatrix} = (6, 7, 2)$$

# Fast Fourier Transform (FFT)

(46)

$$\text{Let } f(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_{n-1} x^{n-1}$$

We separate  $f(x)$  into terms having odd and even powers of  $x$  as follows (assuming  $n$  to be a power of 2):

$$\begin{aligned} f(x) &= (a_0 + a_2 x^2 + a_4 x^4 + \dots + a_{n-2} x^{n-2}) + \\ &\quad (a_1 x + a_3 x^3 + a_5 x^5 + \dots + a_{n-1} x^{n-1}) \\ &= f_e(x) + x f_o(x) \end{aligned}$$

$$\text{where } f_e(x) = a_0 + a_2 x^2 + a_4 x^4 + \dots + a_{n-2} x^{\cancel{n-2}} \cancel{+ \dots + a_{\frac{n}{2}-1} x^{\frac{n}{2}-1}}$$

$$\text{and } f_o(x) = a_1 + a_3 x^2 + a_5 x^4 + \dots + a_{n-1} x^{\cancel{n-2}} \cancel{+ \dots + a_{\frac{n}{2}-1} x^{\frac{n}{2}-1}}$$

$$\text{We have } w_m^2 = e^{\frac{2\pi i}{m}} = e^{\frac{2\pi i}{(n/2)}} = w_{n/2}$$

$$\text{and } w_m^{n/2} = e^{\frac{2\pi i}{m} \cdot \frac{m}{2}} = e^{i\pi} = -1$$

$$f(w_m^i) = f_e((w_m^i)^2) + w_m^i f_o((w_m^i)^2)$$

$$= f_e((w_m^2)^i) + w_m^i f_o((w_m^2)^i)$$

$$= f_e(w_{n/2}^i) + w_m^i f_o(w_{n/2}^i)$$

As  $i$  varies from 0 to  $n-1$ , we observe that the values of  $w_{n/2}^i$  repeats at a distance of  $n/2$ .

$$w_{n/2}^{i+n/2} = w_{n/2}^{n/2} \cdot w_{n/2}^i = w_{n/2}^i \rightarrow \textcircled{1}$$

$$\text{and } w_n^{i+\frac{n}{2}} = w_m^{n/2} \cdot w_m^i = -w_m^i \rightarrow \textcircled{2}$$

We have divided the problem of finding DFT of  $f(n)$  of size  $n$  into two subproblems of finding DFT of  $f_e(x)$  and  $f_o(x)$  of size  $\frac{n}{2}$ . (Divide Step)

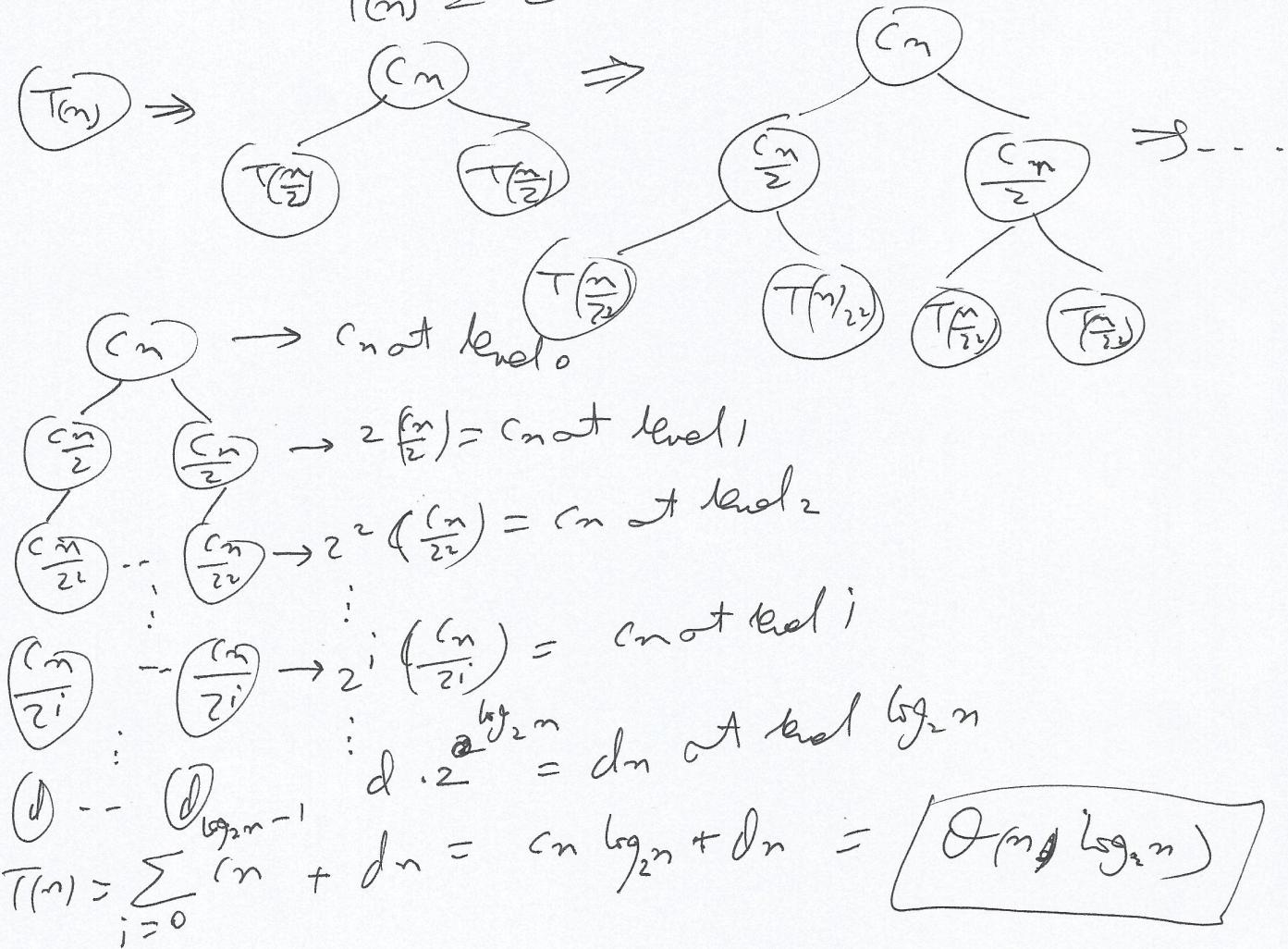
DFT ( $f_{\text{ov}}$ ) is a vector of size  $n$ , while DFT ( $f_{\text{even}}$ ) (47) and DFT ( $f_{\text{odd}}$ ) are vectors of size  $n/2$ . Given the vectors DFT ( $f_{\text{even}}$ ) and DFT ( $f_{\text{odd}}$ ), we recover the vector DFT ( $f_n$ ) as follows (longer step): using equations (1) and (2), we write (for  $0 \leq i \leq \frac{n}{2} - 1$ )

$$f(w_n^i) = f_e(w_{n/2}^i) + w_n^i f_o(w_{n/2}^i)$$

$$f(w_n^{i+\frac{n}{2}}) = f_e(w_{n/2}^i) - w_n^i f_o(w_{n/2}^i)$$

Complexity of FFT: Both Divide Step and Conquer Steps take time linear in input =  $Cn$ . There are two subproblems of size  $\frac{n}{2}$ . We get the following recurrence:  $T(1) = d$

$$T(n) = 2T(\frac{n}{2}) + Cn \quad \text{for } n > 1$$



### Recursive-FFT( $\alpha$ )

(48)

- 1  $n = \alpha.length$  //  $n$  is a power of 2
- 2 if  $n = 1$
- 3     return  $\alpha$
- 4  $w_n \leftarrow e^{2\pi i / n}$
- 5  $w \leftarrow 1$
- 6  $\alpha^{(0)} \leftarrow (\alpha_0, \alpha_2, \dots, \alpha_{n-2})$
- 7  $\alpha^{(1)} \leftarrow (\alpha_1, \alpha_3, \dots, \alpha_{n-1})$
- 8  $y^{(0)} \leftarrow \text{Recursive-FFT}(\alpha^{(0)})$
- 9  $y^{(1)} \leftarrow \text{Recursive-FFT}(\alpha^{(1)})$
- 10 for  $k = 0$  to  $\frac{n}{2} - 1$ 
  - 11  $y_k \leftarrow y_k^{(0)} + w y_k^{(1)}$
  - 12  $y_{k+\frac{n}{2}} \leftarrow y_k^{(0)} - w y_k^{(1)}$
- 13  $w \leftarrow w w_n$
- 14 return  $y$

For computing  $\mathcal{DFT}^{-1}$  using FFT, we replace  $w_n = e^{\frac{2\pi i}{n}}$  with  $e^{-\frac{2\pi i}{n}}$  in line 4 in the above algorithm and return  $\frac{y}{n}$  in line 14 (only in the original problem, we return  $y$ , in subproblems, we return  $y$ ). Recursive-FFT $^{-1}(\alpha)$  is same as Recursive-FFT( $\alpha$ ) except for line 4:  $w_n \leftarrow e^{-\frac{2\pi i}{n}}$ . FFT $^{-1}(\alpha)$  will return  $(\frac{1}{n}) \text{Recursive-FFT}^{-1}(\alpha)$ .

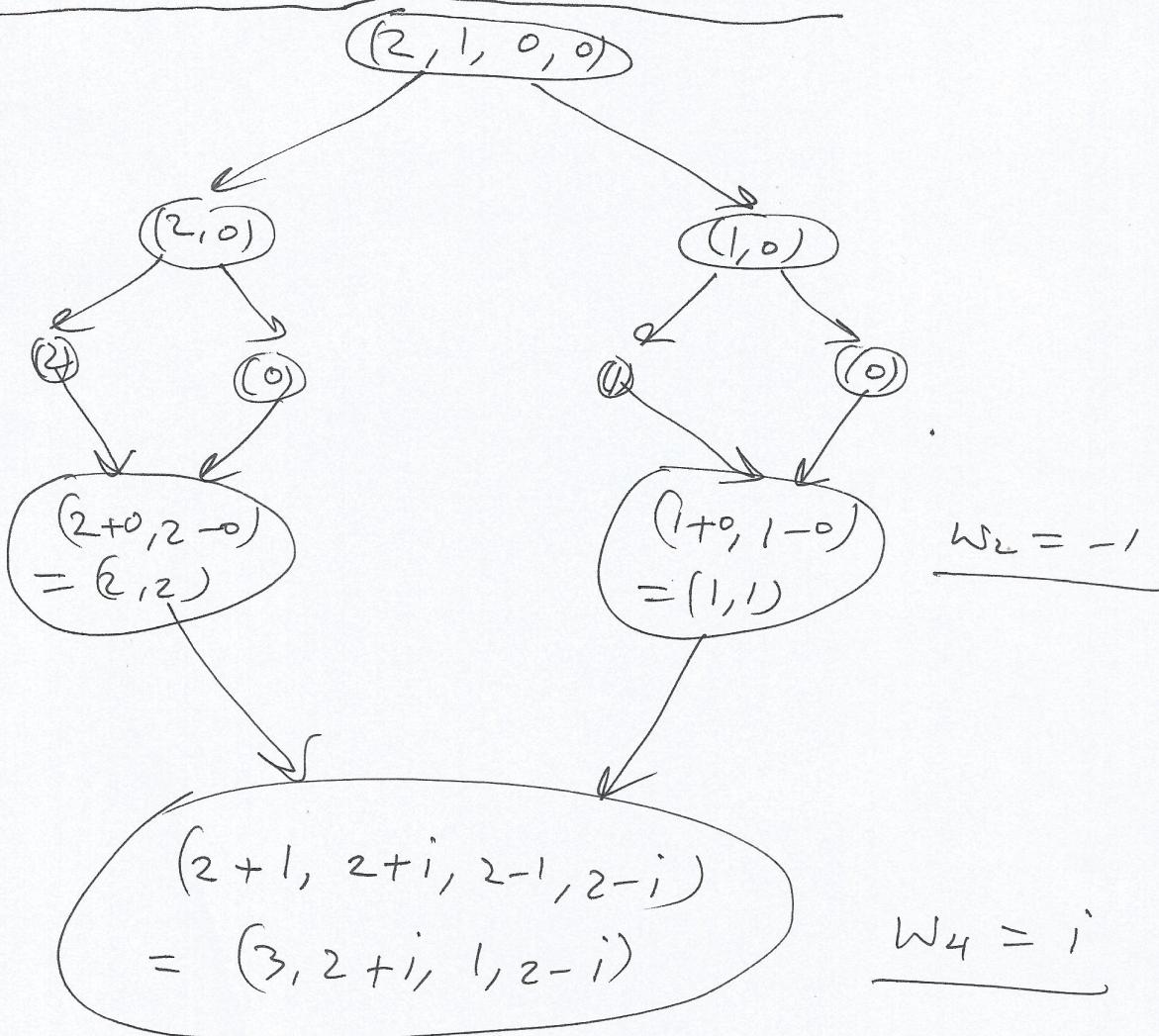
# Polynomial Multiplication using FFT

49

We will multiply  $f_1(n) = n+2$ , and  $f_2(n) = 2n+3$  using the FFT algorithm.  $f_1(n)f_2(n)$  will have 3 terms. We pad all polynomials with 0 to the nearest power of 2 terms. (4 terms in this example). We write  $f_1(n)$  and  $f_2(n)$  in coefficient form as follows:

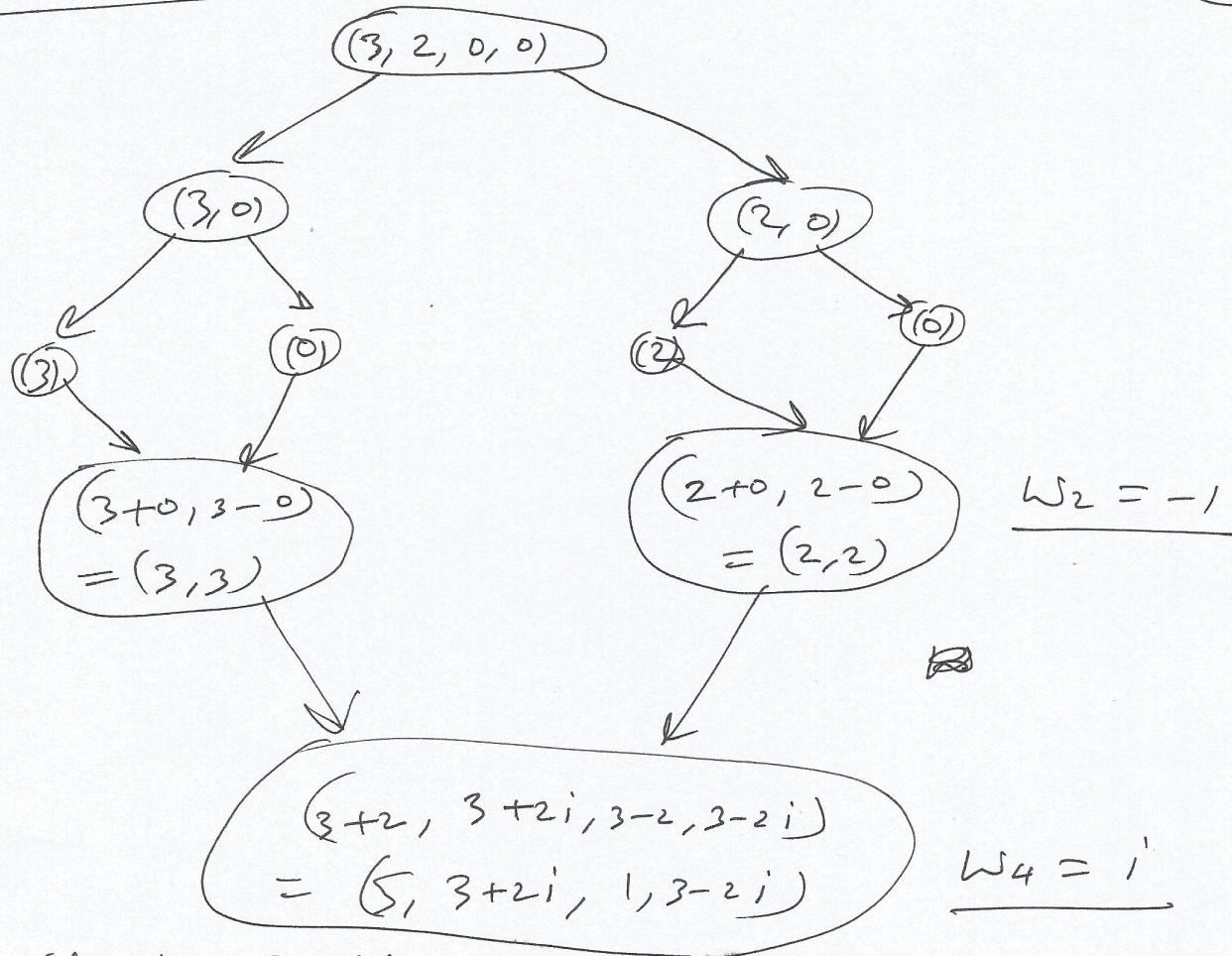
$$f_1(n) = (2, 1, 0, 0) \text{ and } f_2(n) = (3, 2, 0, 0)$$

Divide and conquer graph for  $f_1(n)$



Divide and conquer graph for  $f_2(n)$

50



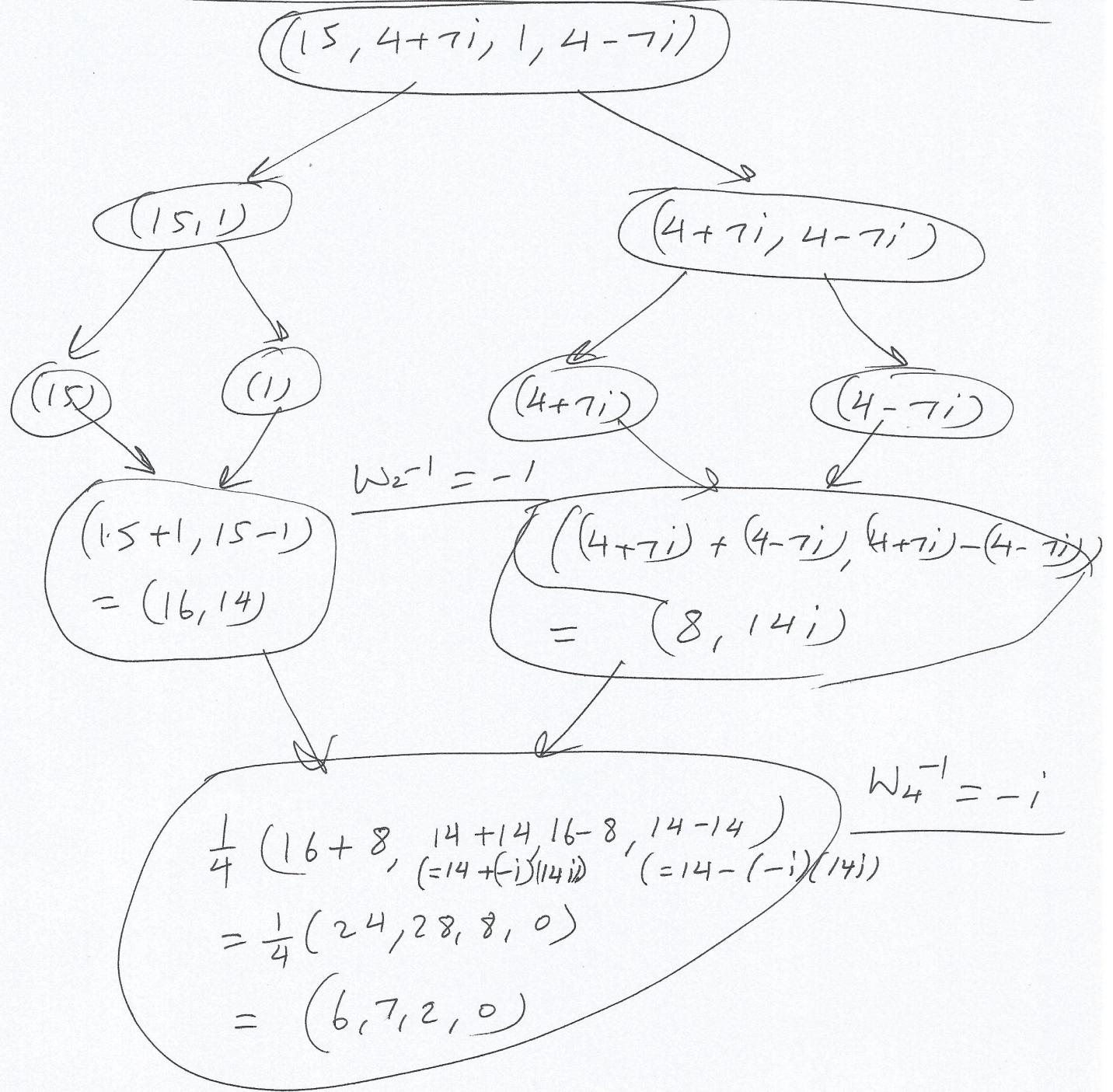
$$DFT(f_1(n)) \cdot DFT(f_2(n)) = (3 \times 5, (2+i) \times (3+2i), 1 \times 1, (2-i) \times (3-2i))$$

$$= (15, 4+7i, 1, 4-7i)$$

Now we compute  $DFT^{-1}(15, 4+7i, 1, 4-7i)$ .

# Divide and Conquer Graph for $DFT^{-1}(f_1(n), f_2(n))$

(51)



$$\begin{aligned}
 & \text{We get } f_1(n) f_2(n) = (6, 7, 2, 0) = 6 + ?n_1 + ?n_2 \\
 & = (n_1 + n_2) (2n_1 + 3)
 \end{aligned}$$