



# **Chapter 1**

# **Divide and Conquer**

**Algorithm Theory**  
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# Formulation of the D&C principle

Divide-and-conquer method for solving a problem instance of size  $n$ :

## 1. Divide

$n \leq c$ : Solve the problem directly.

$n > c$ : Divide the problem into  $k$  subproblems of sizes  $n_1, \dots, n_k < n$  ( $k \geq 2$ ).

## 2. Conquer

Solve the  $k$  subproblems in the same way (recursively).

## 3. Combine

Combine the partial solutions to generate a solution for the original instance.

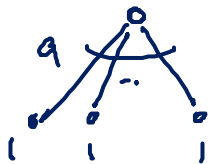
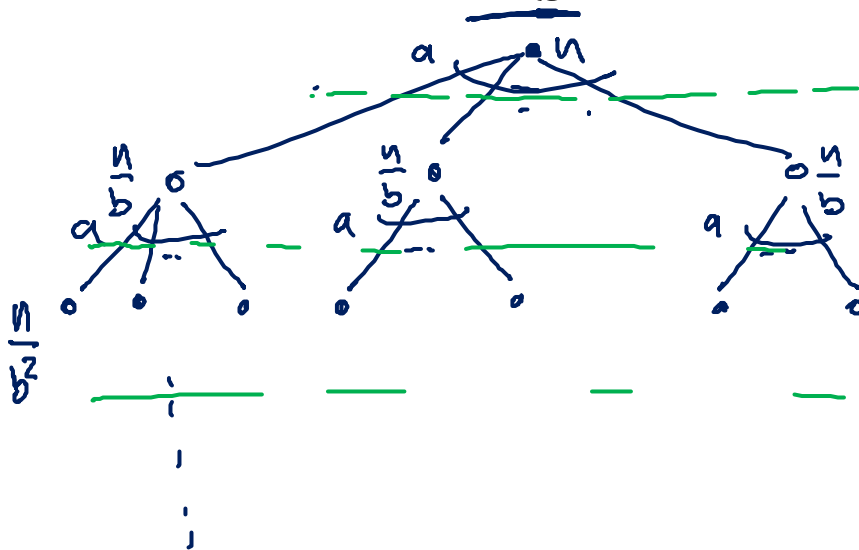
# Recurrence Relations $\log(a^{\log_b n}) = \frac{\log n}{\log b} \cdot \log a$

Recurrence relation

cost for divide & combine

$$\underline{T(n)} = \underline{a} \cdot T\left(\frac{n}{b}\right) + \underline{O(n^c)},$$

$$T(n) = O(1) \text{ for } n \leq n_0$$



$$n^c$$

$$a \cdot \left(\frac{n}{b}\right)^c = \frac{a}{b^c} \cdot n^c$$

$$a^2 \left(\frac{n}{b^2}\right)^c = \left(\frac{a}{b^c}\right)^2 \cdot n^c$$

$$\vdots$$

$$\vdots$$

$$\vdots$$

$$a^{\log_b n} = n^{\log_b a}$$

if  $a < b^c$   
( $\log_b a < c$ ):  
 $T(n) = O(n^c)$

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if  $\log_b a > c$ :  
 $T(n) = \Theta(n^{\log_b a})$

---

if  $\log_b a = c$ :  
 $T(n) = n^c \cdot \log n$

# Recurrence Relations: Master Theorem

## Recurrence relation

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + \underline{f(n)}, \quad T(n) = O(1) \text{ for } n \leq n_0$$

## Cases

- $f(n) = O(n^c)$ ,  $c < \log_b a$

$$\underline{T(n) = \Theta(n^{\log_b a})}$$

- $f(n) = \Omega(n^c)$ ,  $c > \log_b a$

$$\underline{T(n) = \Theta(f(n))}$$

- $f(n) = \Theta(n^c \cdot \underline{\log^k n})$ ,  $k \geq 0, c = \log_b a$

$$\underline{T(n) = \Theta(n^c \cdot \log^{k+1} n)}$$

$$T(n) = 2 \cdot T(n/2) + O(n)$$

$$T(n) = O(n \cdot \log n)$$

# Polynomials

Real polynomial  $p$  in one variable  $x$ :  $a_i \in \mathbb{R}$

$$p(x) = a_{n-1}x^{n-1} + \dots + a_1x^1 + a_0$$

Coefficients of  $p$ :  $a_0, a_1, \dots, a_n \in \mathbb{R}$

**Degree** of  $p$ : largest power of  $x$  in  $p$  ( $n - 1$  in the above case)

**Example:**

$$p(x) = 3x^3 - 15x^2 + 18x$$

$$a_0 = 0, a_1 = 18, a_2 = -15, a_3 = 3$$

Set of all real-valued polynomials in  $x$ :  $\mathbb{R}[x]$  (polynomial ring)

# Operations: Evaluation

- Given: Polynomial  $p \in \mathbb{R}[x]$  of degree  $n - 1$

$$p(x) = \underbrace{a_{n-1}x^{n-1}} + \underbrace{a_{n-2}x^{n-2}} + \dots + a_1x + a_0$$

given some value  $x_0 \in \mathbb{R}$ ,

compute  $p(x_0)$

# Operations: Evaluation

- Given: Polynomial  $p \in \mathbb{R}[x]$  of degree  $n - 1$

$$p(x) = a_{n-1}x^{n-1} + a_{n-2}x^{n-2} + \dots + a_1x + a_0$$

- **Horner's method** for evaluation at specific value  $x_0$ :

$$p(x_0) = (\dots ((\underline{a_{n-1}}x_0 + \underline{a_{n-2}})x_0 + a_{n-3})x_0 + \dots + a_1)x_0 + a_0$$

- Pseudo-code:

$p := a_{n-1}; i := n - 1;$

**while** ( $i > 0$ ) **do**

$i := i - 1;$

$p := p \cdot x_0 + a_i$

- Running time:  $O(n)$

# Operations: Addition

- Given: Polynomials  $p, q \in \mathbb{R}[x]$  of degree  $n - 1$

$$p(x) = a_{n-1}x^{n-1} + a_{n-2}x^{n-2} + \dots + a_1x + a_0$$

$$q(x) = b_{n-1}x^{n-1} + b_{n-2}x^{n-2} + \dots + b_1x + b_0$$

- Compute sum  $p(x) + q(x)$ :

$$p(x) + q(x)$$

$$= (a_{n-1}x^{n-1} + \dots + a_0) + (b_{n-1}x^{n-1} + \dots + b_0)$$

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

$\mathcal{O}(n)$  time



# Operations: Multiplication

- Given: Polynomials  $p, q \in \mathbb{R}[x]$  of degree  $n - 1$

$$p(x) = a_{n-1}x^{n-1} + \dots + a_1x + a_0$$

$$q(x) = b_{n-1}x^{n-1} + \dots + b_1x + b_0$$

$$c_0 = a_0 \cdot b_0, \quad c_1 = a_1 \cdot b_0 + a_0 \cdot b_1, \quad c_2 = a_0 \cdot b_2 + a_1 \cdot b_1 + a_2 \cdot b_0$$

- Product  $p(x) \cdot q(x)$ :

$$\begin{aligned} p(x) \cdot q(x) &= (a_{n-1}x^{n-1} + \dots + a_0) \cdot (b_{n-1}x^{n-1} + \dots + b_0) \\ &= \underline{c_{2n-2}}x^{2n-2} + c_{2n-3}x^{2n-3} + \dots + c_1x + \underline{c_0} \end{aligned}$$

- Obtaining  $c_i$ : what products of monomials have degree  $i$ ?

$$\text{For } 0 \leq i \leq 2n - 2: \underline{c_i} = \sum_{j=0}^i \underline{a_j b_{i-j}}$$

where  $a_i = b_i = 0$  for  $i \geq n$ .

- Running time naïve algorithm:  $\mathcal{O}(n^2)$

# Faster Multiplication?

$$(a_{n-1}, a_{n-2}, \dots, a_{n/2} \mid a_{n/2-1}, \dots, a_0)$$



- Multiplication is slow ( $\Theta(n^2)$ )
- Try **divide-and-conquer** to get a faster algorithm
- Assume: degree is  $n - 1$ ,  $n$  is even ( $n$  is a power of 2)
- Divide polynomial  $p(x) = a_{n-1}x^{n-1} + \dots + a_0$  into 2 polynomials of degree  $n/2 - 1$ :

$$\underline{p_0}(x) = a_{n/2-1}x^{n/2-1} + \dots + a_0$$

$$\underline{p_1}(x) = a_{n-1}x^{n/2-1} + \dots + a_{n/2}$$

$$\underline{p}(x) = \underline{p_1}(x) \cdot \underline{x^{n/2}} + p_0(x)$$

- Similarly:  $q(x) = q_1(x) \cdot x^{n/2} + q_0(x)$

# Use Divide-And-Conquer

- **Divide:**

$$p(x) = p_1(x) \cdot x^{n/2} + p_0(x), \quad q(x) = q_1(x) \cdot x^{n/2} + q_0(x)$$

- **Multiplication:**

$$p(x)q(x) = \underbrace{p_1(x)q_1(x)} \cdot x^n + \underbrace{(p_0(x)q_1(x) + p_1(x)q_0(x))} \cdot x^{n/2} + \underbrace{p_0(x)q_0(x)}$$

- 4 multiplications of degree  $n/2 - 1$  polynomials:

$$\underline{T(n)} = 4\underline{T(n/2)} + \underline{O(n)}$$

Master theorem!  
 $a=4, b=2, \log_b a = 2 > 1$   
 $\hookrightarrow T(n) = n^{\log_b a} = O(n^2)$

- Leads to  $T(n) = \Theta(n^2)$  like the naive algorithm...

- follows immediately by using the master theorem

# More Clever Recursive Solution

- Recall that

$$p(x)q(x) = \overbrace{p_1(x)q_1(x)}^A \cdot x^n + \underbrace{(p_0(x)q_1(x) + p_1(x)q_0(x))}_{B} \cdot x^{n/2} + \underbrace{p_0(x)q_0(x)}_C$$

- Compute  $\underline{r(x)} = \underbrace{(p_0(x) + p_1(x))}_{\text{poly. of size } \frac{n}{2}} \cdot \underbrace{(q_0(x) + q_1(x))}_{\text{size } \frac{n}{2}}$ :

$$r(x) = \underbrace{p_0(x) \cdot q_0(x)}_C + \underbrace{p_0(x)q_1(x) + p_1(x)q_0(x)}_B + \underbrace{p_1(x) \cdot q_1(x)}_A = A + B + C$$

compute :  $r(x), A, C$   
 $B = r(x) - A - C$

$$T(n) = 3 \cdot T(n/2) + O(n)$$

# Karatsuba Algorithm

- Recursive multiplication:

$$r(x) = (p_0(x) + p_1(x)) \cdot (q_0(x) + q_1(x))$$

$$p(x)q(x) = p_1(x)q_1(x) \cdot x^n$$

$$+ (r(x) - p_0(x)q_0(x) + p_1(x)q_1(x)) \cdot x^{n/2}$$

$$+ p_0(x)q_0(x)$$

- Recursively do **3 multiplications of degr.  $(n/2 - 1)$ -polynomials**

$$T(n) = 3T(n/2) + O(n)$$

- Gives:  $T(n) = O(n^{\log_2 3})$  (see Master theorem)

# Representation of Polynomials

## Coefficient representation:

- Polynomial  $p(x) \in \mathbb{R}[x]$  of degree  $n - 1$  is given by its  $n$  coefficients  $a_0, \dots, a_{n-1}$ :

$$p(x) = a_{n-1}x^{n-1} + \dots + a_1x + a_0$$

- Coefficient vector  $\mathbf{a} = (a_0, a_1, \dots, a_{n-1})$
- Example:

$$p(x) = 3x^3 - 15x^2 + 18x$$

- The most typical (and probably most natural) representation of polynomials

# Representation of Polynomials

## Product of linear factors:

- Polynomial  $p(x) \in \mathbb{C}[x]$  of degr.  $n - 1$  is given by its  $n - 1$  roots

$$p(x) = \underline{a_{n-1}} \cdot (x - x_1) \cdot (x - x_2) \cdot \dots \cdot (x - x_{n-1})$$

- Example:

$$p(x) = 3x(x - 2)(x - 3)$$

- Every polynomial has exactly  $n - 1$  roots  $x_i \in \mathbb{C}$  (s.t.  $p(x_i) = 0$ )
  - Polynomial is uniquely defined by the  $n - 1$  roots and  $a_{n-1}$
- We will not use this representation...

# Representation of Polynomials

## Point-value representation:

- Polynomial  $p(x) \in \mathbb{R}[x]$  of degree  $n - 1$  is given by  $n$  point-value pairs:

$$p = \{(\underline{x_0}, \underline{p(x_0)}), (\underline{x_1}, \underline{p(x_1)}), \dots, (\underline{x_{n-1}}, \underline{p(x_{n-1})})\}$$

where  $x_i \neq x_j$  for  $i \neq j$ .

- Example: The polynomial

$$p(x) = 3x(x - 2)(x - 3)$$

is uniquely defined by the four point-value pairs  $(0,0), (1,6), (2,0), (3,0)$ .



# Operations: Coefficient Representation



$$p(x) = a_{n-1}x^{n-1} + \dots + a_0, \quad q(x) = b_{n-1}x^{n-1} + \dots + b_0$$

**Evaluation:** Horner's method: Time  $O(n)$

**Addition:**

$$p(x) + q(x) = (a_{n-1} + b_{n-1})x^{n-1} + \dots + (a_0 + b_0)$$

- Time:  $O(n)$

**Multiplication:**

$$p(x) \cdot q(x) = c_{2n-2}x^{2n-2} + \dots + c_0, \quad \text{where } c_i = \sum_{j=0}^i a_j b_{i-j}$$

- Naive solution: Need to compute product  $a_i b_j$  for all  $0 \leq i, j \leq n$
- Time:  $O(n^2)$   $\rightarrow$  can be improved to  $O(n^{1.58\dots})$

# Operations: Linear Factors (Roots)

$$p(x) = a_{n-1} \cdot (x - x_1) \cdot \dots \cdot (x - x_{n-1})$$
$$q(x) = b_{n-1} \cdot (x - x_1) \cdot \dots \cdot (x - x_{n-1})$$

## Evaluation:

- Just plug in the value where the poly. is evaluated: **Time  $O(n)$**

## Multiplication:

- Concatenate the two representations: **Time  $O(n)$**

## Addition:

- Need to find the roots of  $p(x) + q(x)$
- For polynomials of degree  $> 4$ , this is not possible by using basic arithmetic operations ( $+$ ,  $-$ ,  $\cdot$ ,  $/$ ,  $\sqrt[a]{b}$ )
- In the usual computational model impossible
  - Numerically, the roots can be computed to arbitrary precision

# Operations: Point-Value Representation

$$\rightarrow p = \{(x_0, p(x_0)), \dots, (x_{n-1}, p(x_{n-1}))\}$$

$$\rightarrow q = \{(x_0, q(x_0)), \dots, (x_{n-1}, q(x_{n-1}))\}$$

- Note: we use the **same points**  $x_0, \dots, x_{n-1}$  for both polynomials

## Addition:

$$p + q = \{(\underline{x_0}, \underline{p(x_0) + q(x_0)}), \dots, (x_{n-1}, p(x_{n-1}) + q(x_{n-1}))\}$$

- Time:  $O(n)$

## Multiplication:

$$\underline{p} \cdot q = \{(x_0, p(x_0) \cdot q(x_0)), \dots, (\underline{x_{2n-2}}, p(x_{2n-2}) \cdot q(x_{2n-2}))\}$$

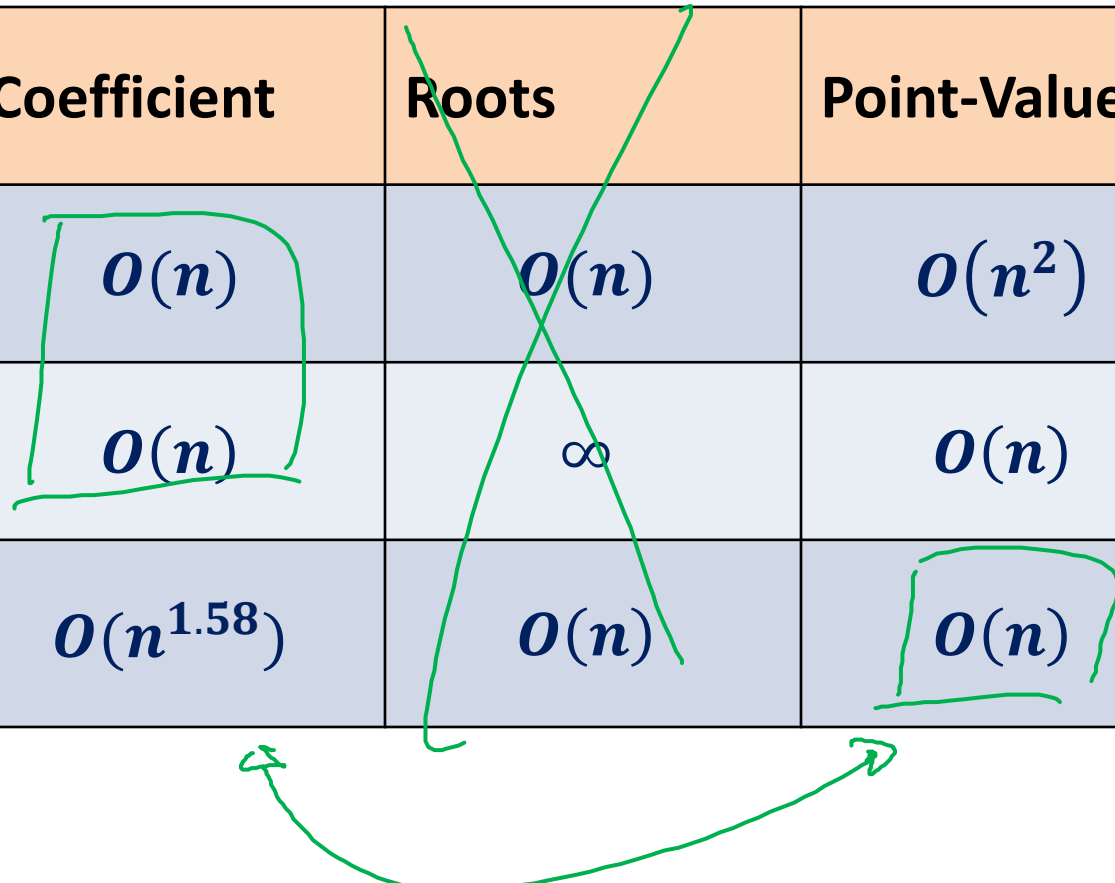
- Time:  $O(n)$

**Evaluation:** Polynomial interpolation can be done in  $O(n^2)$

# Operations on Polynomials

Cost depending on representation:

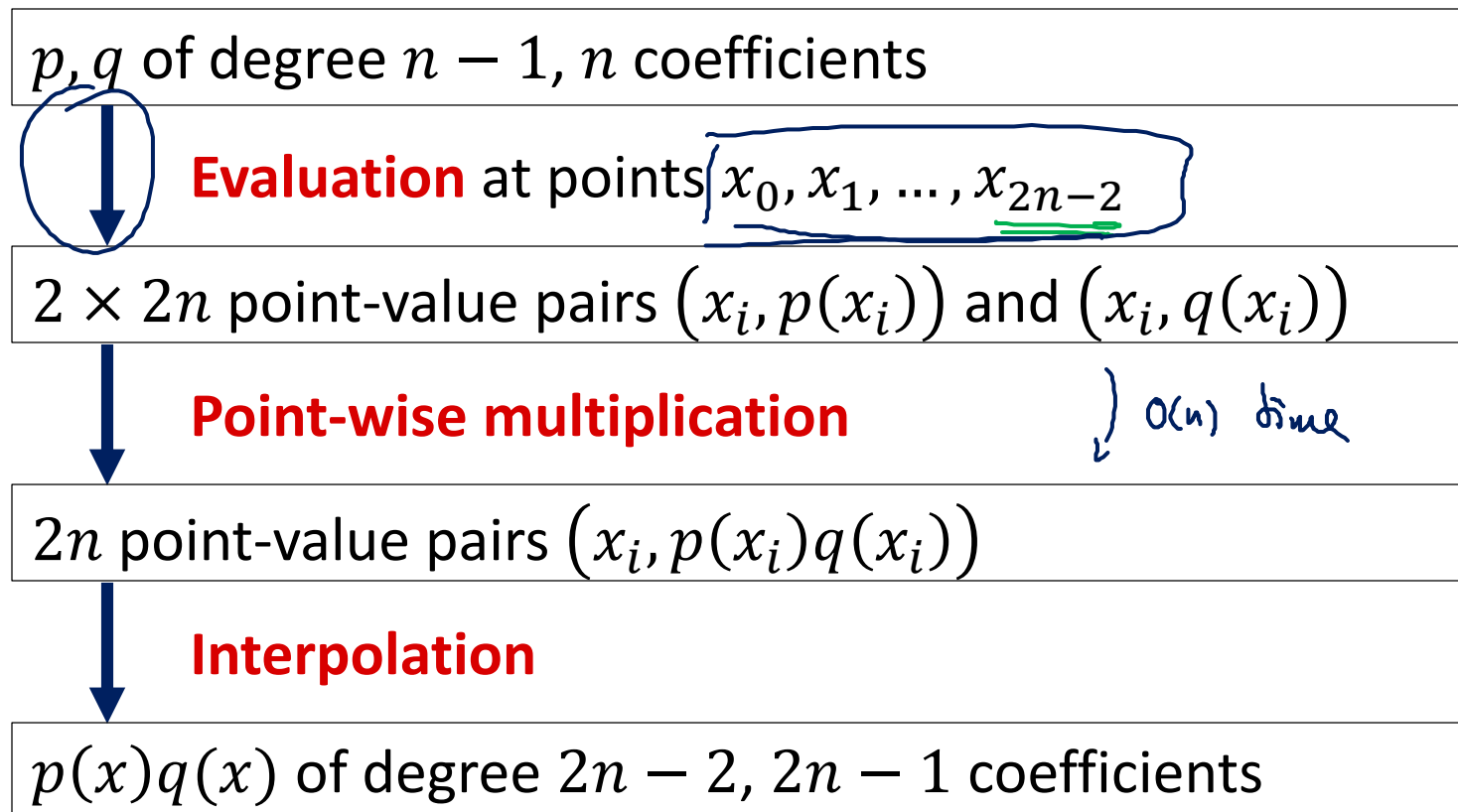
	Coefficient	Roots	Point-Value
Evaluation	$O(n)$	$O(n)$	$O(n^2)$
Addition	$O(n)$	$\infty$	$O(n)$
Multiplication	$O(n^{1.58})$	$O(n)$	$O(n)$



# Faster Polynomial Multiplication?

Multiplication is fast when using the **point-value representation**

**Idea** to compute  $p(x) \cdot q(x)$  (for polynomials of degree  $< n$ ):



# Coefficients to Point-Value Representation



**Given:** Polynomial  $\underline{p(x)}$  by the coefficient vector  $\underline{(a_0, a_1, \dots, a_{N-1})}$

**Goal:** Compute  $\underline{p(x)}$  for all  $x$  in a given set  $\underline{X}$

- Where  $X$  is of size  $\underline{|X| = N}$
- Assume that  $\underline{N}$  is a power of 2

## Divide and Conquer Approach

- Divide  $p(x)$  of degree  $N - 1$  ( $N$  is even) into 2 polynomials of degree  $N/2 - 1$  differently than in Karatsuba's algorithm

$$\begin{aligned} \rightarrow p_0(y) &= a_0 + a_2y + a_4y^2 + \dots + a_{N-2}y^{N/2-1} && \text{(even coeff.)} \\ \rightarrow p_1(y) &= a_1 + a_3y + a_5y^2 + \dots + a_{N-1}y^{N/2-1} && \text{(odd coeff.)} \end{aligned}$$

# Coefficients to Point-Value Representation

**Goal:** Compute  $p(x)$  for all  $x$  in a given set  $X$  of size  $|X| = N$

- Divide  $p(x)$  of degr.  $N - 1$  into 2 polynomials of degr.  $N/2 - 1$

$$\rightarrow p_0(y) = a_0 + \underline{a_2}y + \underline{a_4}y^2 + \dots + a_{N-2}y^{N/2-1} \quad (\text{even coeff.})$$

$$\rightarrow p_1(y) = a_1 + a_3y + a_5y^2 + \dots + a_{N-1}y^{N/2-1} \quad (\text{odd coeff.})$$

**Let's first look at the "combine" step:**

- We need to compute  $p(x)$  for all  $x \in X$  after recursive calls for polynomials  $p_0$  and  $p_1$ :

$$\forall x \in X : p(x) = p_0(x^2) + x \cdot p_1(x^2)$$

recursively compute  $p_0(y), p_1(y)$  for all  $y \in X^2$

$$X^2 := \{x^2 : x \in X\}$$

cost of divide  
& combine!

$$\underline{O(N + |X|)}$$

# Coefficients to Point-Value Representation



**Goal:** Compute  $p(x)$  for all  $x$  in a given set  $X$  of size  $|X| = N$

- Divide  $p(x)$  of degr.  $N - 1$  into 2 polynomials of degr.  $N/2 - 1$

$$p_0(y) = a_0 + a_2y + a_4y^2 + \cdots + a_{N-2}y^{N/2-1} \quad (\text{even coeff.})$$

$$p_1(y) = a_1 + a_3y + a_5y^2 + \cdots + a_{N-1}y^{N/2-1} \quad (\text{odd coeff.})$$

**Let's first look at the "combine" step:**

$$\forall x \in X : p(x) = p_0(x^2) + x \cdot p_1(x^2)$$

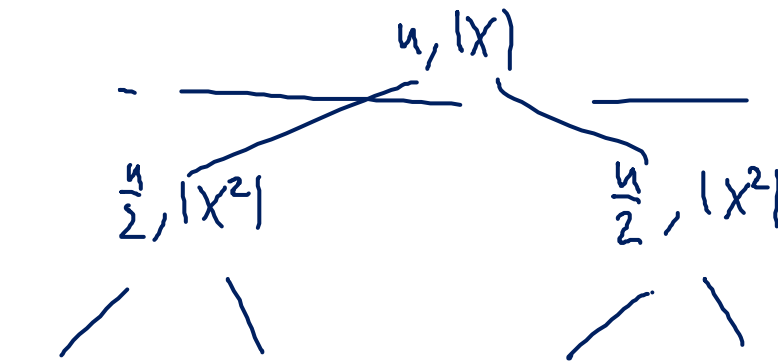
- Recursively compute  $\underline{p_0(y)}$  and  $\underline{p_1(y)}$  for all  $\underline{y \in X^2}$ 
  - Where  $X^2 := \{x^2 : x \in X\}$

- Generally, we have  $\underline{|X^2|} = \underline{|X|}$



Recurrence formula for the given algorithm:

$$T(n, |X|) = 2T\left(\frac{n}{2}, |X^2|\right) + \mathcal{O}(n + |X|) \leq 2T\left(\frac{n}{2}, |X|\right) + \mathcal{O}(n + |X|)$$

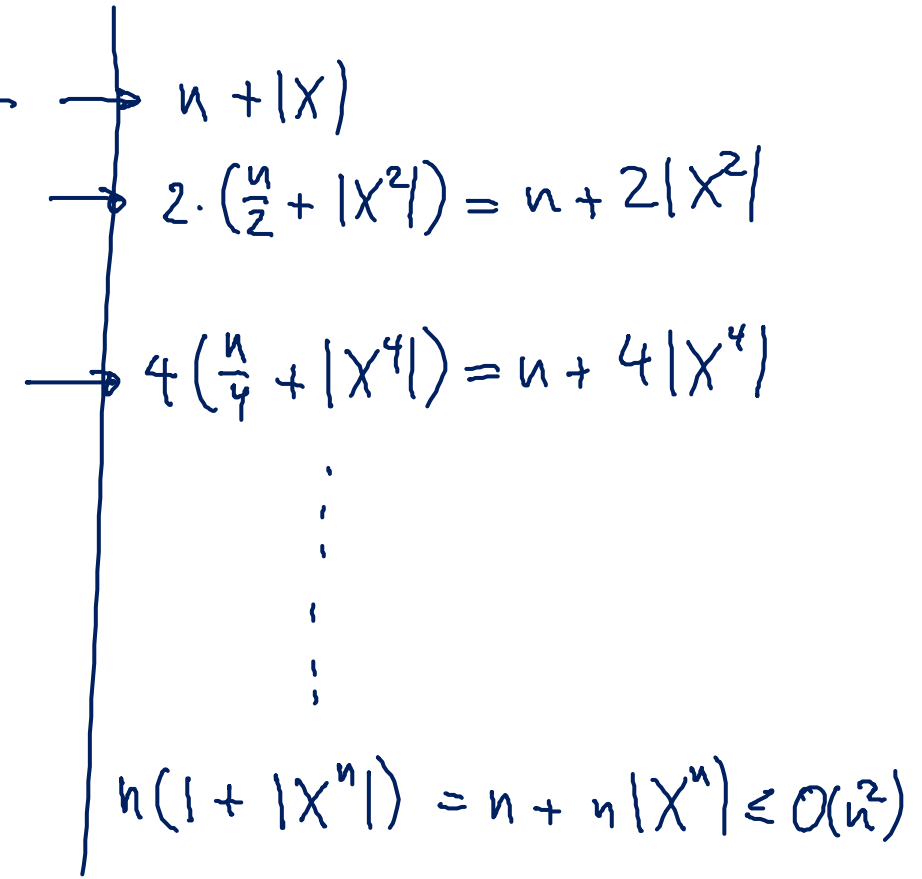


$$|X^2| \stackrel{?}{=} |X|$$

$\{-1, 1\}$

$$\text{if } |X^{2^k}| = \frac{|X^k|}{2},$$

we would get an  $\mathcal{O}(n \log n)$  alg.



# Faster Algorithm?

- In order to have a faster algorithm, we need  $|X^2| < |X|$

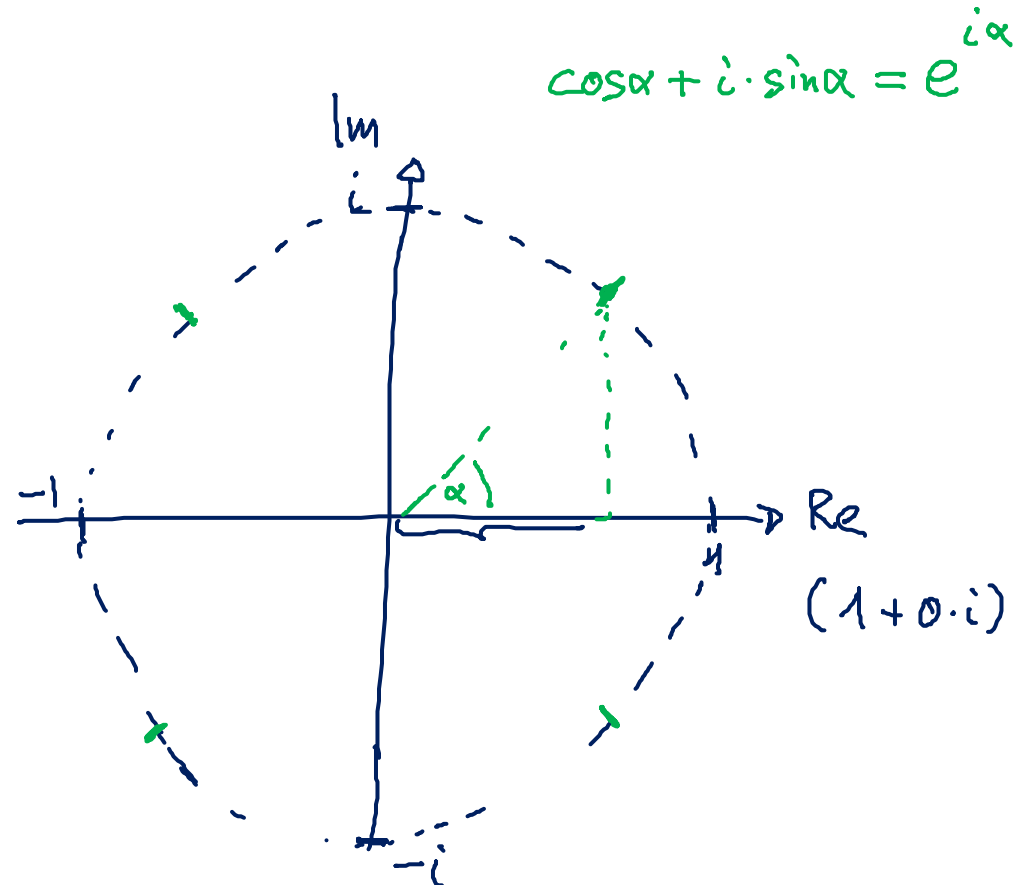
$$n \rightarrow \frac{n}{2} \rightarrow \frac{n}{4}$$

$$\{1\}$$

$$\{-1, 1\}$$

$$\{i, -i, -1, 1\}$$

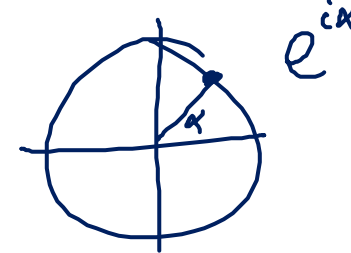
$$\left\{1, -1, i, -i, \frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}}i, \frac{1}{\sqrt{2}} - \frac{1}{\sqrt{2}}i, \frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}}i, -\frac{1}{\sqrt{2}} - \frac{1}{\sqrt{2}}i\right\}$$



# Choice of $X$

- Select points  $x_0, x_1, \dots, x_{N-1}$  to evaluate  $p$  and  $q$  in a clever way

Consider the  $N$  complex roots of unity:

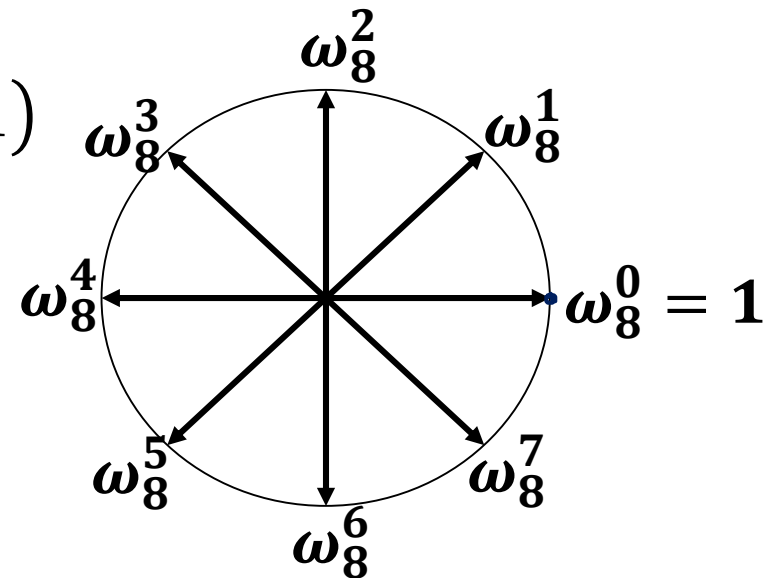


**Principle root of unity:**  $\omega_N = e^{2\pi i/N}$

$$(i = \sqrt{-1}, \quad e^{2\pi i} = 1)$$

**Powers of  $\omega_N$  (roots of unity):**

$$1 = \omega_N^0, \omega_N^1, \dots, \omega_N^{N-1}$$



Note:  $\omega_N^k = e^{2\pi i k/N} = \cos \frac{2\pi k}{N} + i \cdot \sin \frac{2\pi k}{N}$

# Properties of the Roots of Unity

- **Cancellation Lemma:**

For all integers  $n > 0$ ,  $k \geq 0$ , and  $d > 0$ , we have:

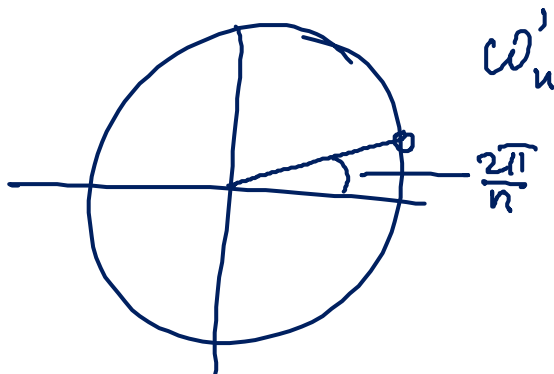
$$\omega_{dn}^{dk} = \omega_n^k,$$

$$\omega_n^{k+n} = \omega_n^k$$

- **Proof:**  $\omega_n^k = e^{i \cdot \frac{2\pi}{n} \cdot k}$

$$\omega_{dn}^{dk} = e^{i \frac{2\pi}{dn} \cdot dk} = \omega_n^k$$

$$\omega_n^{k+n} = e^{i \frac{2\pi}{n} k} \cdot \underbrace{e^{i \frac{2\pi}{n} n}}_{e^{2\pi i} = 1} = \omega_n^k$$

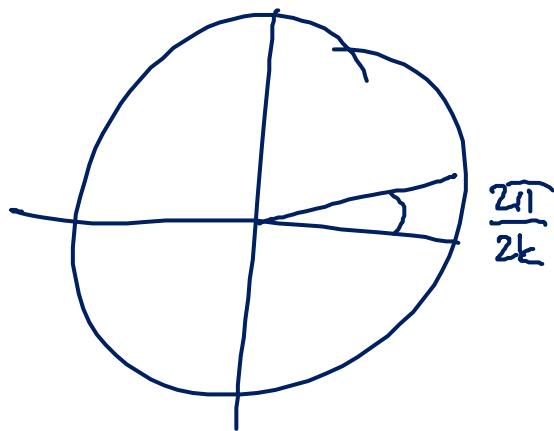


# Properties of the Roots of Unity

**Claim:** If  $X = \{\omega_{2k}^i : i \in \{0, \dots, 2k - 1\}\}$ , we have

$$X^2 = \{\omega_k^i : i \in \{0, \dots, k - 1\}\}, \quad |X^2| = \frac{|X|}{2}$$

$$\begin{aligned} X^2 &= \{\omega_{2k}^{2i} = \omega_k^i, i \in \{0, \dots, 2k - 1\}\} \\ &= \{\omega_k^i, i \in \{0, \dots, k - 1\}\} \end{aligned}$$



New recurrence formula:

$$T(N, |X|) \leq 2 \cdot T\left(\frac{N}{2}, \frac{|X|}{2}\right) + O(N + |X|)$$

for  $|X| = N$

$$\leadsto O(N \log N)$$

# Faster Polynomial Multiplication?

Idea to compute  $\underline{p(x)} \cdot \underline{q(x)}$  (for polynomials of degree  $< n$ ):

$p, q$  of degree  $n - 1$ ,  $n$  coefficients

FFT

**Evaluation** at points  $\omega_{2n}^0, \omega_{2n}^1, \dots, \omega_{2n}^{2n-1}$   $O(n \log n)$

$2 \times 2n$  point-value pairs  $(\omega_{2n}^k, p(\omega_{2n}^k))$  and  $(\omega_{2n}^k, q(\omega_{2n}^k))$

**Point-wise multiplication**

$2n$  point-value pairs  $(\omega_{2n}^k, p(\omega_{2n}^k)q(\omega_{2n}^k))$

**Interpolation**

$p(x)q(x)$  of degree  $2n - 2$ ,  $2n - 1$  coefficients