



Chapter 1

Divide and Conquer

Algorithm Theory
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Divide-And-Conquer Principle

- Important algorithm design method
- Examples from basic alg. & data structures class (Informatik 2):
 - Sorting: Mergesort, Quicksort
 - Binary search
- Further examples
 - Median
 - **Comparing orders**
 - Convex hull / Delaunay triangulation / Voronoi diagram
 - **Closest pairs**
 - Line intersections
 - **Polynomial multiplication / FFT**
 - ...

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.

Running Time

Recurrence relation:

$$T(n) = 2 \cdot T(n/2) + c \cdot n, \quad T(1) = a$$

Solution:

- Same as for computing number of number of inversions, merge sort (and many others...)

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

Recurrence relation

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

Cases

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

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

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

$$T(n) = \Theta(f(n))$$

- $f(n) = \Theta(n^c \cdot \log^k n)$, $c = \log_b a$

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

Polynomials

Real polynomial p in one variable x :

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

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:

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

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

Divide-&-Conquer Polynomial Multiplication

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

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

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

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

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

- **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:**

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

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

$$T(n) = 4T(n/2) + O(n)$$

- Leads to $T(n) = \Theta(n^2)$ like the naive algorithm...
 - follows immediately by using the master theorem

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: $T(n) = O(n^{1.58496\dots})$ (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

Point-value representation:

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

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

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: $O(n^2)$**

Operations: Point-Value Representation



$$\begin{aligned} p &= \{(x_0, p(x_0)), \dots, (x_{n-1}, p(x_{n-1}))\} \\ q &= \{(x_0, q(x_0)), \dots, (x_{n-1}, q(x_{n-1}))\} \end{aligned}$$

- Note: we use the **same points** x_0, \dots, x_n for both polynomials

Addition:

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

- **Time:** $O(n)$

Multiplication:

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

- **Time:** $O(n)$

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

Operations on Polynomials

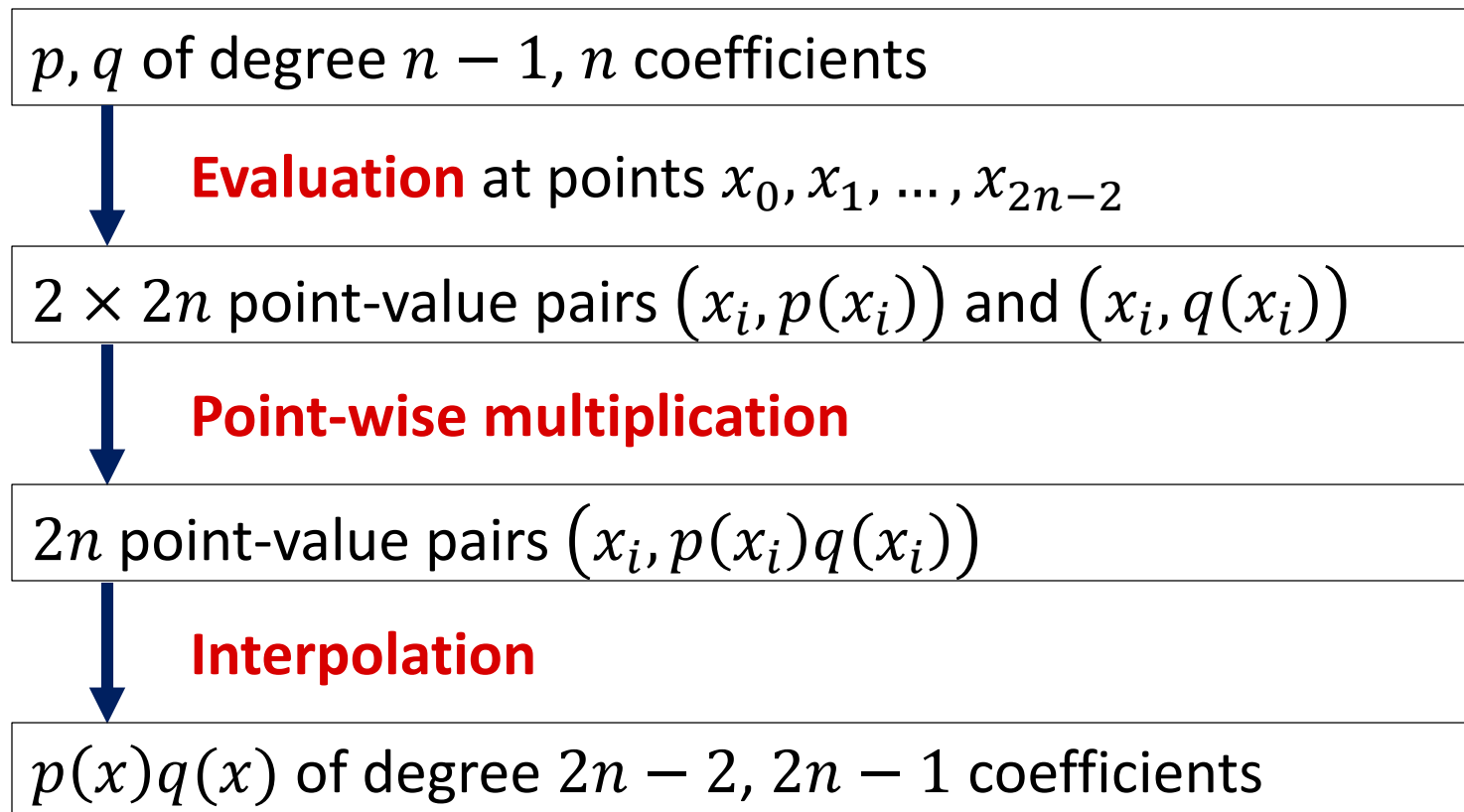
Cost depending on representation:

	Coefficient	Roots	Point-Value
Evaluation	$O(n)$	$O(n)$	$O(n^2)$
Addition	$O(n)$	∞	$O(n)$
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 $< n$):



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

Goal: Compute $p(x)$ for all x in a given set X

- Where X is of size $|X| = N$
- Assume that 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
- $p_0(y) = a_0 + a_2y + a_4y^2 + \dots + a_{N-2}y^{N/2-1}$ (even coeff.)
 $p_1(y) = a_1 + a_3y + a_5y^2 + \dots + a_{N-1}y^{N/2-1}$ (odd coeff.)

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 $p_0(y)$ and $p_1(y)$ for all $y \in X^2$
 - Where $X^2 := \{x^2 : x \in X\}$
- Generally, we have $|X^2| = |X|$

Analysis

Recurrence formula for the given algorithm:

Faster Algorithm?

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

Choice of X

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

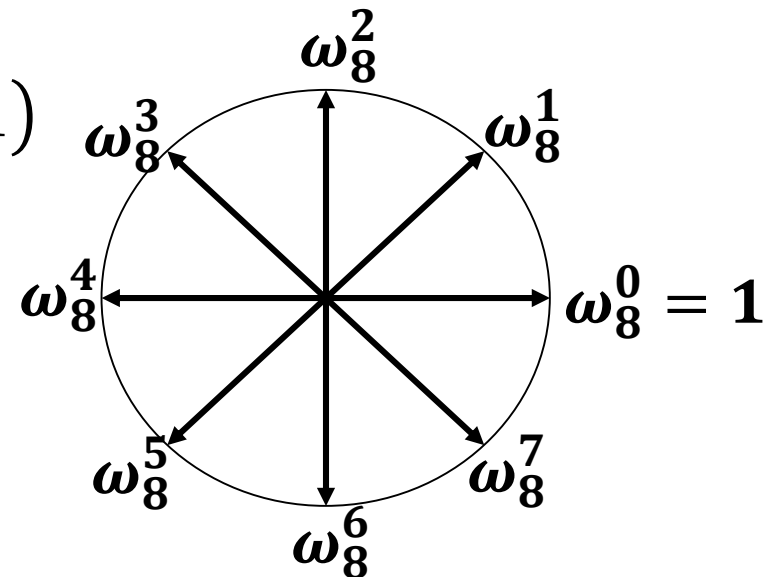
Consider the N complex roots of unity:

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

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

Powers of ω_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:

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

- **Proof:**

Properties of the Roots of Unity

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

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

New recurrence formula:

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

Faster Polynomial Multiplication?

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

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

Evaluation at points $\omega_{2n}^0, \omega_{2n}^1, \dots, \omega_{2n}^{2n-1}$

$2 \times 2n$ 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

Discrete Fourier Transform

- The values $p(\omega_N^i)$ for $i = 0, \dots, N - 1$ uniquely define a polynomial p of degree $< N$.

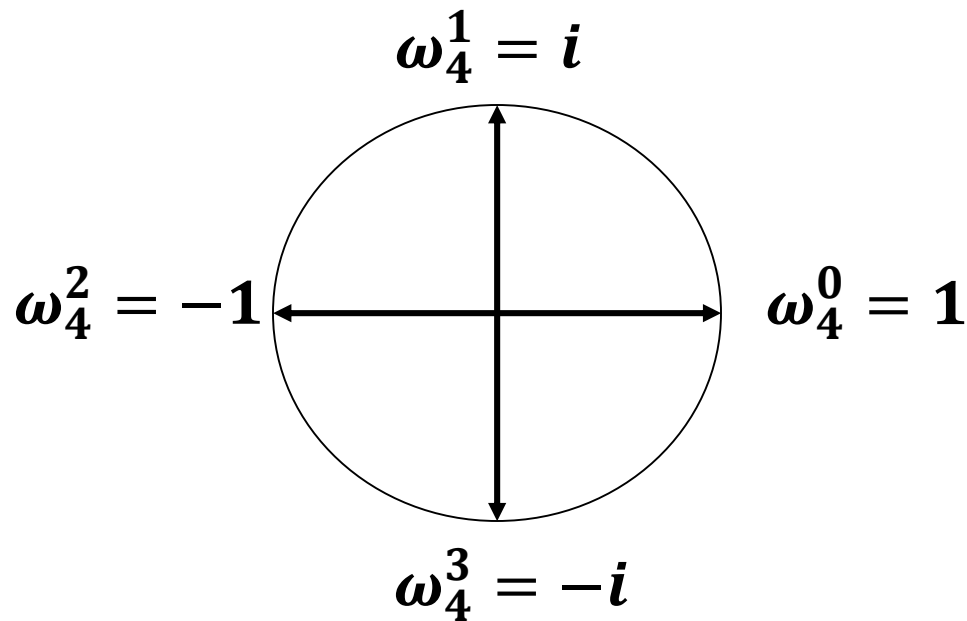
Discrete Fourier Transform (DFT):

- Assume $a = (a_0, \dots, a_{N-1})$ is the coefficient vector of poly. p
$$(p(x) = a_{N-1}x^{N-1} + \dots + a_1x + a_0)$$

$$\text{DFT}_N(a) := \left(p(\omega_N^0), p(\omega_N^1), \dots, p(\omega_N^{N-1}) \right)$$

Example

- Consider polynomial $p(x) = 3x^3 - 15x^2 + 18x$
- Choose $N = 4$
- Roots of unity:



Example

- Consider polynomial $p(x) = 3x^3 - 15x^2 + 18x$
- $N = 4$, roots of unity: $\omega_4^0 = 1, \omega_4^1 = i, \omega_4^2 = -1, \omega_4^3 = -i$
- Evaluate $p(x)$ at ω_4^k :

$$\left(\omega_4^0, p(\omega_4^0)\right) = (1, p(1)) = (1, 6)$$

$$\left(\omega_4^1, p(\omega_4^1)\right) = (i, p(i)) = (i, 15 + 15i)$$

$$\left(\omega_4^2, p(\omega_4^2)\right) = (-1, p(-1)) = (-1, -36)$$

$$\left(\omega_4^3, p(\omega_4^3)\right) = (-i, p(-i)) = (-i, 15 - 15i)$$

- For $a = (0, 18, -15, 3)$:

$$\mathbf{DFT}_4(a) = (6, 15 + 15i, -36, 15 - 15i)$$

DFT: Recursive Structure

Evaluation for $k = 0, \dots, N - 1$:

$$\begin{aligned}
 p(\omega_N^k) &= p_0((\omega_N^k)^2) + \omega_N^k \cdot p_1((\omega_N^k)^2) \\
 &= \begin{cases} p_0(\omega_{N/2}^k) + \omega_N^k \cdot p_1(\omega_{N/2}^k) & \text{if } k < N/2 \\ p_0(\omega_{N/2}^{k-N/2}) + \omega_N^k \cdot p_1(\omega_{N/2}^{k-N/2}) & \text{if } k \geq N/2 \end{cases}
 \end{aligned}$$

For the coefficient vector a of $p(x)$:

$$\begin{aligned}
 \text{DFT}_N(a) &= \left(p_0(\omega_{N/2}^0), \dots, p_0(\omega_{N/2}^{N/2-1}), p_0(\omega_{N/2}^0), \dots, p_0(\omega_{N/2}^{N/2-1}) \right) \\
 &\quad + \left(\omega_N^0 p_0(\omega_{N/2}^0), \dots, \omega_N^{N/2-1} p_0(\omega_{N/2}^{N/2-1}), \omega_N^{N/2} p_0(\omega_{N/2}^0), \dots, \omega_N^{N-1} p_0(\omega_{N/2}^{N/2-1}) \right)
 \end{aligned}$$

Example

For the coefficient vector a of $p(x)$:

$$\begin{aligned} \text{DFT}_N(a) = & \left(p_0(\omega_{N/2}^0), \dots, p_0(\omega_{N/2}^{N/2-1}), p_0(\omega_{N/2}^0), \dots, p_0(\omega_{N/2}^{N/2-1}) \right) \\ & + \left(\omega_N^0 p_0(\omega_{N/2}^0), \dots, \omega_N^{N/2-1} p_0(\omega_{N/2}^{N/2-1}), \omega_N^{N/2} p_0(\omega_{N/2}^0), \dots, \omega_N^{N-1} p_0(\omega_{N/2}^{N/2-1}) \right) \end{aligned}$$

$N = 4$:

$$\begin{aligned} p(\omega_4^0) &= p_0(\omega_2^0) + \omega_4^0 p_1(\omega_2^0) \\ p(\omega_4^1) &= p_0(\omega_2^1) + \omega_4^1 p_1(\omega_2^1) \\ p(\omega_4^2) &= p_0(\omega_2^0) + \omega_4^2 p_1(\omega_2^0) \\ p(\omega_4^3) &= p_0(\omega_2^1) + \omega_4^3 p_1(\omega_2^1) \end{aligned}$$

Need: $\left(p_0(\omega_2^0), p_0(\omega_2^1) \right)$ and $\left(p_1(\omega_2^0), p_1(\omega_2^1) \right)$

(DFTs of coefficient vectors of p_0 and p_1)

Summary: Computation of DFT_N

- Divide-and-conquer algorithm for $\text{DFT}_N(p)$:

1. Divide

$$N \leq 1: \text{DFT}_1(p) = a_0$$

$N > 1$: Divide p into p_0 (even coeff.) and p_1 (odd coeff.).

2. Conquer

Solve $\text{DFT}_{N/2}(p_0)$ and $\text{DFT}_{N/2}(p_1)$ recursively

3. Combine

Compute $\text{DFT}_N(p)$ based on $\text{DFT}_{N/2}(p_0)$ and $\text{DFT}_{N/2}(p_1)$

Small Improvement

Polynomial p of degree $N - 1$:

$$\begin{aligned}
 p(\omega_N^k) &= \begin{cases} p_0(\omega_{N/2}^k) + \omega_N^k \cdot p_1(\omega_{N/2}^k) & \text{if } k < N/2 \\ p_0(\omega_{N/2}^{k-N/2}) + \omega_N^k \cdot p_1(\omega_{N/2}^{k-N/2}) & \text{if } k \geq N/2 \end{cases} \\
 &= \begin{cases} p_0(\omega_{N/2}^k) + \omega_N^k \cdot p_1(\omega_{N/2}^k) & \text{if } k < N/2 \\ p_0(\omega_{N/2}^{k-N/2}) - \omega_N^{k-N/2} \cdot p_1(\omega_{N/2}^{k-N/2}) & \text{if } k \geq N/2 \end{cases}
 \end{aligned}$$

Need to compute $p_0(\omega_{N/2}^k)$ and $\omega_N^k \cdot p_1(\omega_{N/2}^k)$ for $0 \leq k < N/2$.

Example $N = 8$

$$p(\omega_8^0) = p_0(\omega_4^0) + \omega_8^0 \cdot p_1(\omega_4^0)$$

$$p(\omega_8^1) = p_0(\omega_4^1) + \omega_8^1 \cdot p_1(\omega_4^1)$$

$$p(\omega_8^2) = p_0(\omega_4^2) + \omega_8^2 \cdot p_1(\omega_4^2)$$

$$p(\omega_8^3) = p_0(\omega_4^3) + \omega_8^3 \cdot p_1(\omega_4^3)$$

$$p(\omega_8^3) = p_0(\omega_4^0) - \omega_8^0 \cdot p_1(\omega_4^0)$$

$$p(\omega_8^3) = p_0(\omega_4^1) - \omega_8^1 \cdot p_1(\omega_4^1)$$

$$p(\omega_8^3) = p_0(\omega_4^2) - \omega_8^2 \cdot p_1(\omega_4^2)$$

$$p(\omega_8^3) = p_0(\omega_4^3) - \omega_8^3 \cdot p_1(\omega_4^3)$$

Fast Fourier Transform (FFT) Algorithm

Algorithm FFT(a)

- Input: Array a of length N , where N is a power of 2
- Output: $\text{DFT}_N(a)$

if $n = 1$ **then return** a_0 ; // $a = [a_0]$

$d^{[0]} := \text{FFT}([a_0, a_2, \dots, a_{N-2}]);$

$d^{[1]} := \text{FFT}([a_1, a_3, \dots, a_{N-1}]);$

$\omega_N := e^{2\pi i/N}$; $\omega := 1$;

for $k = 0$ **to** $N/2 - 1$ **do** // $\omega = \omega_N^k$

$x := \omega \cdot d_k^{[1]}$;

$d_k := d_k^{[0]} + x$; $d_{k+N/2} := d_k^{[0]} - x$;

$\omega := \omega \cdot \omega_N$

end;

return $d = [d_0, d_1, \dots, d_{N-1}]$;

Example

$$p(x) = 3x^3 - 15x^2 + 18x + 0, \quad a = [0, 18, -15, 3]$$

Faster Polynomial Multiplication?

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

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

Evaluation at $\omega_{2n}^0, \omega_{2n}^1, \dots, \omega_{2n}^{2n-1}$ using **FFT**

$2 \times 2n$ 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

Interpolation

Convert point-value representation into coefficient representation

Input: $(x_0, y_0), \dots, (x_{n-1}, y_{n-1})$ with $x_i \neq x_j$ for $i \neq j$

Output:

Degree- $(n - 1)$ polynomial with coefficients a_0, \dots, a_{n-1} such that

$$\begin{aligned} p(x_0) &= a_0 + a_1x_0 + a_2x_0^2 + \dots + a_{n-1}x_0^{n-1} = y_0 \\ p(x_1) &= a_0 + a_1x_1 + a_2x_1^2 + \dots + a_{n-1}x_1^{n-1} = y_1 \\ &\vdots \\ p(x_{n-1}) &= a_0 + a_1x_{n-1} + a_2x_{n-1}^2 + \dots + a_{n-1}x_{n-1}^{n-1} = y_{n-1} \end{aligned}$$

→ linear system of equations for a_0, \dots, a_{n-1}