



Algorithm Theory

Chapter 1 Divide and Conquer

Part IV: Fast Polynomial Multiplication 1

Representation of Polynomials



Coefficient Representation:

• Polynomial of degree n - 1 defined by coefficients a_0, \dots, a_{n-1} : $p(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_{n-1} x^{n-1}$

Point-value Representation:

- Polynomial p(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

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

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

Operations: Coefficient Representation



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

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:

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

- Naïve solution: Need to compute product $a_i b_j$ for all $0 \le i, j \le n$
- Time: Naïve alg. $O(n^2)$ Karatsuba Alg. $O(n^{1.58496...})$

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Operations: Point-Value Representation



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

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

• Note: We use the same points x_0, \dots, x_{n-1} 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*)
- **Remark:** Need both polynomials at (the same) 2n 1 points.

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

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Operations on Polynomials



Cost depending on representation:

	Coefficient	Point-Value
Evaluation	0 (n)	$O(n^2)$
Addition	0 (n)	0 (n)
Multiplication	O (n ^{1.58})	0 (n)
	default representation	Can we improve this?



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



N > 2n - 1

We will fix X later.

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

- **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
- $p_0(y) = a_0 + a_2 y + a_4 y^2 + \dots + a_{N-2} y^{N/2-1}$ (even coeff.) $p_1(y) = a_1 + a_3 y + a_5 y^2 + \dots + a_{N-1} y^{N/2-1}$ (odd coeff.)

We call the variable y because we will not plug in x into p_0 and p_1

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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. $N/_2 1$
 - $p_0(y) = a_0 + a_2 y + a_4 y^2 + \dots + a_{N-2} y^{N/2-1} \quad \text{(even coeff.)}$ $p_1(y) = a_1 + a_3 y + a_5 y^2 + \dots + a_{N-1} y^{N/2-1} \quad \text{(odd coeff.)}$

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

- We need to compute p(x) for all x ∈ X after recursive calls for polynomials p₀ and p₁:
- Plug $y = x^2$ into $p_0(y)$ and $p_1(y)$:

$$p_0(x^2) = a_0 + a_2 x^2 + a_4 x^4 + \dots + a_{N-2} x^{N-2}$$

$$p_1(x^2) = a_1 + a_3 x^2 + a_5 x^4 + \dots + a_{N-1} x^{N-2}$$

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

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. $N/_2 1$
 - $p_0(y) = a_0 + a_2 y + a_4 y^2 + \dots + a_{N-2} y^{N/2-1} \quad \text{(even coeff.)}$ $p_1(y) = a_1 + a_3 y + a_5 y^2 + \dots + a_{N-1} y^{N/2-1} \quad \text{(odd coeff.)}$

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

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

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

Analysis



Let's get a recurrence recurrence for the given algorithm:

Time for polynomial of degree N with set X: T(N, |X|)

$$T(N, |X|) = 2 \cdot T(\frac{N}{2}, |X^2|) + O(N + |X|)$$

Assume that $|X^2| = |X| = N$:

$$T(N,N) = 2 \cdot T\left(\frac{N}{2}, N\right) + O(N) = \dots = N \cdot \left(T(1,N) + O(N)\right)$$
$$T(1,N) = O(N)$$

Therefore, we get $T(N, |X|) = O(N^2)$.

• We need $|X^2| < |X|$ to get a faster algorithm!



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

• $|X^2| < |X|$ if X contains values x, x' such that $x^2 = x'^2$:

$$X = \{-1, +1\} \implies X^2 = \{+1\}$$

• We also need
$$|(X^2)^2| = |X^4| < |X^2|$$
:

- Can we get a set Y of size 4 such that $Y^2 = \{-1, +1\}$?

- Complex numbers C:
 - Define imaginary constant *i* s.t. $i^2 = -1$
 - Complex numbers: $\mathbb{C} = \{a + i \cdot b \mid a, b \in \mathbb{R}\}$
- $Y = \{-1, +1, -i, +i\} \implies Y^2 = \{-1, +1\}$
- $\forall x \in \mathbb{C} \setminus \{0\}$, there are 2 numbers $y, z \in \mathbb{C}$ s.t. $y^2 = z^2 = x$

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: $\omega_N = e^{2\pi i/N}$ ω_8^2 $\left(i=\sqrt{-1}, \qquad e^{2\pi i}=1\right)$ ω_8^1 ω_{8}^3 ω_8^4 $\omega_8^0 = 1$ Powers of ω_N (roots of unity): $1 = \omega_{N}^{0}, \omega_{N}^{1}, ..., \omega_{N}^{N-1}$ ω_8^7 ω_8^5 ω_8^6 Note: $\omega_N^k = e^{2\pi i k/N} = \cos \frac{2\pi k}{N} + i \cdot \sin \frac{2\pi k}{N}$



Cancellation Lemma:

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

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

Proof: Recall that $\omega_n = e^{2\pi i/n}$, $e^{2\pi i} = 1$

$$\omega_{dn}^{dk} = \left(e^{\frac{2\pi i}{dn}}\right)^{dk} = e^{\frac{2\pi i}{dn} \cdot dk} = e^{\frac{2\pi i}{n} \cdot k} = \omega_n^k$$

$$\omega_n^{k+n} = \left(e^{\frac{2\pi i}{n}}\right)^{k+n} = e^{\frac{2\pi i}{\mathbf{d}n}\cdot(k+n)} = e^{\frac{2\pi i}{n}\cdot k} \cdot e^{2\pi i} = \omega_n^k$$





Claim: If
$$X = \left\{ \omega_{2k}^{j} : j \in \{0, ..., 2k - 1\} \right\}$$
, we have
 $X^{2} = \left\{ \omega_{k}^{j} : j \in \{0, ..., k - 1\} \right\}$, $|X^{2}| = \frac{|X|}{2}$

Proof:

- We just showed: $\omega_{dn}^{dk} = \omega_n^k$, $\omega_n^{k+n} = \omega_n^k$
- Consider some $x = \omega_{2k}^j \in X$:

$$x^{2} = \left(\omega_{2k}^{j}\right)^{2} = \omega_{2k}^{2j} = \omega_{k}^{j}$$

If $j \ge k : \omega_{k}^{j} = \omega_{k}^{j-k}$

• Clearly, $|X^2| = |X|/2$ (|X| = 2k, $|X^2| = k$).

Algorithm Theory

Analysis



New recurrence formula:

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

- W.I.o.g., assume that N is a power of 2
 - We can just add additional coefficients that are equal to 0.
- To compute p(x) for the N different points in X, we need to recursively compute $p_0(x^2)$ and $p_1(x^2)$ for all $x^2 \in X^2$

- p has degree N - 1, p_0 and p_1 have degree $N/_2 - 1$, $|X^2| = |X|/_2$

- Combine step: compute $p(x) = p_0(x^2) + x \cdot p_1(x^2)$ for all $x \in X$
- $|X| = N \implies T(N) \le 2 \cdot T(N/2) + O(N)$

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

Faster Polynomial Multiplication?



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



Discrete Fourier Transform



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

Discrete Fourier Transform (DFT):

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

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

Example



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

•

Roots of unity: $\omega_4^1 = i$ $\omega_4^2 = -1$ $\omega_4^0 = 1$ $\omega_4^3 = -i$

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 :

$$\begin{pmatrix} \omega_4^0, p(\omega_4^0) \end{pmatrix} = (1, p(1)) = (1, 6) \\ \begin{pmatrix} \omega_4^1, p(\omega_4^1) \end{pmatrix} = (i, p(i)) = (i, 15 + 15i) \\ \begin{pmatrix} \omega_4^2, p(\omega_4^2) \end{pmatrix} = (-1, p(-1)) = (-1, -36) \\ \begin{pmatrix} \omega_4^3, p(\omega_4^3) \end{pmatrix} = (-i, p(-i)) = (-i, 15 - 15i)$$

• For
$$a = (0,18, -15,3)$$
:
 $DFT_4(a) = (6, 15 + 15i, -36, 15 - 15i)$

DFT: Recursive Structure



Evaluation for k = 0, ..., N - 1:

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

For the coefficient vector a of p(x):

$$DFT_{N}(a) = \left(p_{0}(\omega_{N/2}^{0}), \dots, p_{0}(\omega_{N/2}^{N/2-1}), p_{0}(\omega_{N/2}^{0}), \dots, p_{0}(\omega_{N/2}^{N/2-1})\right) + \left(\omega_{N/2}^{0}p_{1}(\omega_{N/2}^{0}), \dots, \omega_{N}^{N/2-1}p_{1}(\omega_{N/2}^{N/2-1}), \omega_{N}^{N/2}p_{1}(\omega_{N/2}^{0}), \dots, \omega_{N}^{N-1}p_{1}(\omega_{N/2}^{N/2-1})\right)$$

Algorithm Theory

Example



For the coefficient vector a of p(x):

$$DFT_{N}(a) = \left(p_{0}(\omega_{N/2}^{0}), \dots, p_{0}(\omega_{N/2}^{N/2-1}), p_{0}(\omega_{N/2}^{0}), \dots, p_{0}(\omega_{N/2}^{N/2-1})\right) + \left(\omega_{N/2}^{0}p_{1}(\omega_{N/2}^{0}), \dots, \omega_{N}^{N/2-1}p_{1}(\omega_{N/2}^{N/2-1}), \omega_{N}^{N/2}p_{1}(\omega_{N/2}^{0}), \dots, \omega_{N}^{N-1}p_{1}(\omega_{N/2}^{N/2-1})\right)$$

N = 4:

$$p(\omega_4^0) = p_0(\omega_2^0) + \omega_4^0 p_1(\omega_2^0)$$

$$p(\omega_4^1) = p_0(\omega_2^1) + \omega_4^1 p_1(\omega_2^1)$$

$$p(\omega_4^2) = p_0(\omega_2^0) + \omega_4^2 p_1(\omega_2^0)$$

$$p(\omega_4^3) = p_0(\omega_2^1) + \omega_4^3 p_1(\omega_2^1)$$

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

(DFTs of coefficient vectors of p_0 and p_1)

Algorithm Theory

Summary: Computation of DFT_N



• Divide-and-conquer algorithm for $DFT_N(p)$:

1. Divide

- $N \leq 1$: DFT₁(p) = a_0
- N > 1: Divide p into p_0 (even coeff.) and p_1 (odd coeff).

2. Conquer

Solve $DFT_{N/2}(p_0)$ and $DFT_{N/2}(p_1)$ recursively

3. Combine

Compute $DFT_N(p)$ based on $DFT_{N/2}(p_0)$ and $DFT_{N/2}(p_1)$

Small Constant Improvement



Polynomial p of degree N - 1:

$$p(\omega_{N}^{k}) = \begin{cases} p_{0}(\omega_{N/2}^{k}) + \omega_{N}^{k} \cdot p_{1}(\omega_{N/2}^{k}) & \text{if } k < N/2 \\ p_{0}(\omega_{N/2}^{k-N/2}) + \omega_{N}^{k} \cdot p_{1}(\omega_{N/2}^{k-N/2}) & \text{if } k \ge N/2 \\ \end{cases}$$
$$= \begin{cases} p_{0}(\omega_{N/2}^{k}) + \omega_{N}^{k} \cdot p_{1}(\omega_{N/2}^{k}) & \text{if } k < N/2 \\ p_{0}(\omega_{N/2}^{k-N/2}) - \omega_{N}^{k-N/2} \cdot p_{1}(\omega_{N/2}^{k-N/2}) & \text{if } k \ge N/2 \end{cases}$$

•
$$\omega_N^{k-N/2} = e^{\frac{2\pi i}{N} \cdot (k-N/2)} = e^{\frac{2\pi i}{N} \cdot k} \cdot e^{-\frac{2\pi i}{N} \cdot \frac{N}{2}} = \omega_N^k \cdot e^{-\pi i} = -\omega_N^k$$

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

Example N = 8



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

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

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

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

$$p(\omega_8^4) = p_0(\omega_4^0) - \omega_8^0 \cdot p_1(\omega_4^0)$$

$$p(\omega_8^5) = p_0(\omega_4^1) - \omega_8^1 \cdot p_1(\omega_4^1)$$

$$p(\omega_8^6) = p_0(\omega_4^2) - \omega_8^2 \cdot p_1(\omega_4^2)$$

$$p(\omega_8^7) = p_0(\omega_4^3) - \omega_8^3 \cdot p_1(\omega_4^3)$$

Fast Fourier Transform (FFT) Algorithm



Algorithm FFT(a)

- Input: Array *a* of length *N*, where *N* is a power of 2
- Output: $DFT_N(a)$

if n = 1 then return a_0 ; $//a = [a_0]$ $d^{[0]} \coloneqq \text{FFT}([a_0, a_2, \dots, a_{N-2}]);$ $d^{[1]} \coloneqq \text{FFT}([a_1, a_3, \dots, a_{N-1}]);$ $\omega_N \coloneqq e^{2\pi i/N}; \omega \coloneqq 1;$ for k = 0 to $N/_2 - 1$ do $//\omega = \omega_N^k$ $x \coloneqq \omega \cdot d_{\nu}^{[1]};$ $d_k \coloneqq d_k^{[0]} + x; d_{k+N/2} \coloneqq d_k^{[0]} - x;$ $\omega \coloneqq \omega \cdot \omega_N$

end;

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

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