

Divide and Conquer

- Polynomial Multiplication
- Integer Multiplication

Polynomial Multiplication



Polynomials: Coefficient Representation

Polynomial. [coefficient representation]

$$A(x) = a_0 + a_1x + a_2x^2 + \cdots + a_{n-1}x^{n-1}$$

$$B(x) = b_0 + b_1x + b_2x^2 + \cdots + b_{n-1}x^{n-1}$$

Add: $O(n)$ arithmetic operations.

$$A(x) + B(x) = (a_0 + b_0) + (a_1 + b_1)x + \cdots + (a_{n-1} + b_{n-1})x^{n-1}$$

Evaluate: $O(n)$ using Horner's method.

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

Multiply (convolve): $O(n^2)$ using brute force.

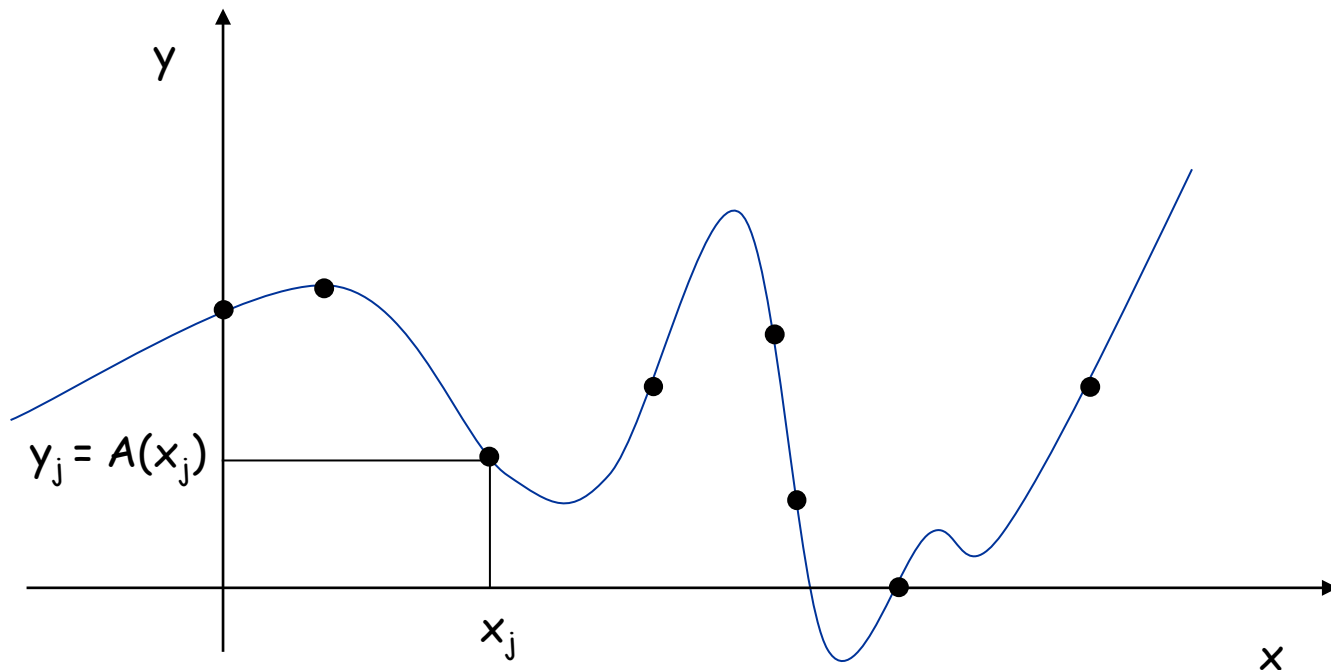
$$A(x) \times B(x) = \sum_{i=0}^{2n-2} c_i x^i, \text{ where } c_i = \sum_{j=0}^i a_j b_{i-j}$$



Polynomials: Point-Value Representation

Fundamental theorem of algebra. [Gauss, PhD thesis] A degree n polynomial with complex coefficients has n complex roots.

Corollary. A degree $n-1$ polynomial $A(x)$ is uniquely specified by its evaluation at n distinct values of x .





Polynomials: Point-Value Representation

Polynomial. [point-value representation]

$$A(x): (x_0, y_0), \dots, (x_{n-1}, y_{n-1})$$

$$B(x): (x_0, z_0), \dots, (x_{n-1}, z_{n-1})$$

Add: $O(n)$ arithmetic operations.

$$A(x) + B(x): (x_0, y_0 + z_0), \dots, (x_{n-1}, y_{n-1} + z_{n-1})$$

Multiply: $O(n)$, but need $2n-1$ points.

$$A(x) \times B(x): (x_0, y_0 \times z_0), \dots, (x_{2n-1}, y_{2n-1} \times z_{2n-1})$$

Evaluate: $O(n^2)$ using Lagrange's formula.

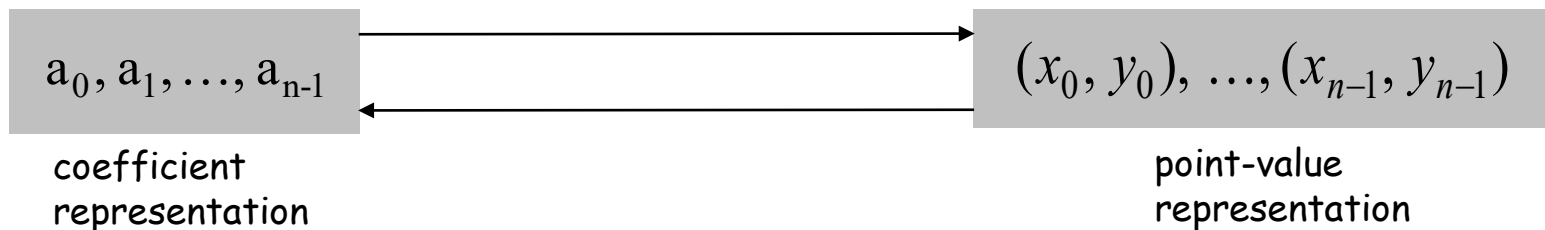
$$A(x) = \sum_{k=0}^{n-1} y_k \frac{\prod_{j \neq k} (x - x_j)}{\prod_{j \neq k} (x_k - x_j)}$$

Converting Between Two Polynomial Representations

Tradeoff. Fast evaluation or fast multiplication. We want both!

Representation	Multiply	Evaluate
Coefficient	$O(n^2)$	$O(n)$
Point-value	$O(n)$	$O(n^2)$

Goal. Make all ops fast by efficiently converting between two representations.



Converting Between Two Polynomial Representations: Brute Force

Coefficient to point-value. Given a polynomial $a_0 + a_1 x + \dots + a_{n-1} x^{n-1}$, evaluate it at n distinct points x_0, \dots, x_{n-1} .

$$\begin{bmatrix} y_0 \\ y_1 \\ y_2 \\ \vdots \\ y_{n-1} \end{bmatrix} = \begin{bmatrix} 1 & x_0 & x_0^2 & \cdots & x_0^{n-1} \\ 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n-1} & x_{n-1}^2 & \cdots & x_{n-1}^{n-1} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_{n-1} \end{bmatrix}$$

$O(n^2)$ for matrix-vector multiply

$O(n^3)$ for Gaussian elimination

↑
Vandermonde matrix is invertible iff x_i distinct

Point-value to coefficient. Given n distinct points x_0, \dots, x_{n-1} and values y_0, \dots, y_{n-1} , find unique polynomial $a_0 + a_1 x + \dots + a_{n-1} x^{n-1}$ that has given values at given points.

Coefficient to Point-Value Representation: Intuition

Coefficient to point-value. Given a polynomial $a_0 + a_1 x + \dots + a_{n-1} x^{n-1}$, evaluate it at n distinct points x_0, \dots, x_{n-1} .

Divide. Break polynomial up into even and odd powers.

- $A(x) = a_0 + a_1 x + a_2 x^2 + a_3 x^3 + a_4 x^4 + a_5 x^5 + a_6 x^6 + a_7 x^7.$
- $A_{\text{even}}(x) = a_0 + a_2 x + a_4 x^2 + a_6 x^3.$
- $A_{\text{odd}}(x) = a_1 + a_3 x + a_5 x^2 + a_7 x^3.$
- $A(x) = A_{\text{even}}(x^2) + x A_{\text{odd}}(x^2).$
- $A(-x) = A_{\text{even}}(x^2) - x A_{\text{odd}}(x^2).$

Intuition. Choose two points to be ± 1 .

- $A(1) = A_{\text{even}}(1) + 1 A_{\text{odd}}(1).$
- $A(-1) = A_{\text{even}}(1) - 1 A_{\text{odd}}(1).$

Can evaluate polynomial of degree $\leq n$ at 2 points by evaluating two polynomials of degree $\leq \frac{1}{2}n$ at 1 point.

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- $A(-x) = A_{\text{even}}(x^2) - x A_{\text{odd}}(x^2).$

Intuition. Choose four points to be $\pm 1, \pm i$.

- $A(1) = A_{\text{even}}(1) + 1 A_{\text{odd}}(1).$
- $A(-1) = A_{\text{even}}(1) - 1 A_{\text{odd}}(1).$
- $A(i) = A_{\text{even}}(-1) + i A_{\text{odd}}(-1).$
- $A(-i) = A_{\text{even}}(-1) - i A_{\text{odd}}(-1).$

Can evaluate polynomial of degree $\leq n$ at 4 points by evaluating two polynomials of degree $\leq \frac{1}{2}n$ at 2 points.

Discrete Fourier Transform

Coefficient to point-value. Given a polynomial $a_0 + a_1 x + \dots + a_{n-1} x^{n-1}$, evaluate it at n distinct points x_0, \dots, x_{n-1} .

Key idea: choose $x_k = \omega^k$ where ω is principal n^{th} root of unity.

$$\begin{bmatrix} y_0 \\ y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_{n-1} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 & \dots & 1 \\ 1 & \omega^1 & \omega^2 & \omega^3 & \dots & \omega^{n-1} \\ 1 & \omega^2 & \omega^4 & \omega^6 & \dots & \omega^{2(n-1)} \\ 1 & \omega^3 & \omega^6 & \omega^9 & \dots & \omega^{3(n-1)} \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^{n-1} & \omega^{2(n-1)} & \omega^{3(n-1)} & \dots & \omega^{(n-1)(n-1)} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \\ \vdots \\ a_{n-1} \end{bmatrix}$$

\uparrow \uparrow

Discrete Fourier transform Fourier matrix F_n



Roots of Unity

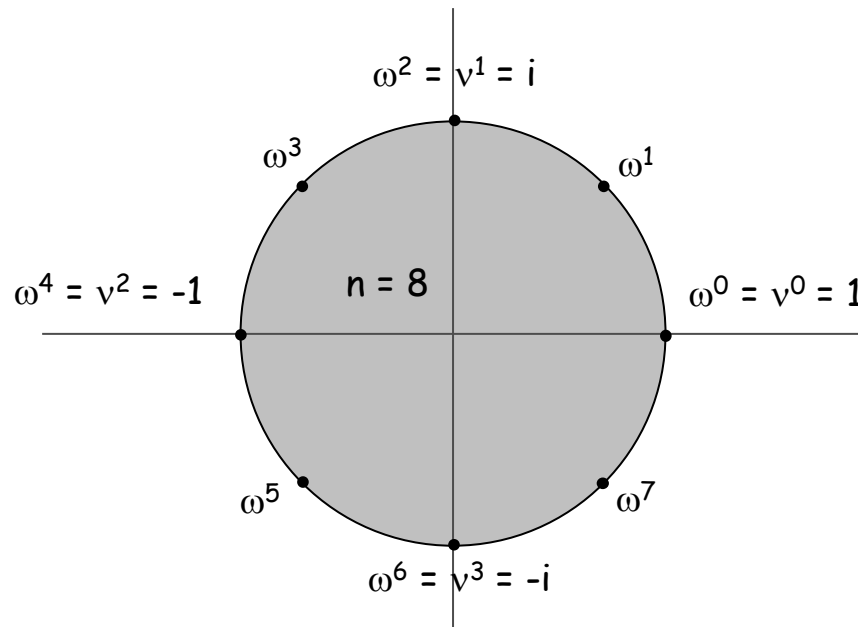
Def. An n^{th} root of unity is a complex number x such that $x^n = 1$.

Fact. The n^{th} roots of unity are: $\omega^0, \omega^1, \dots, \omega^{n-1}$ where $\omega = e^{2\pi i / n}$.

Pf. $(\omega^k)^n = (e^{2\pi i k / n})^n = (e^{\pi i})^{2k} = (-1)^{2k} = 1$.

Fact. The $\frac{1}{2}n^{\text{th}}$ roots of unity are: $v^0, v^1, \dots, v^{n/2-1}$ where $v = e^{4\pi i / n}$.

Fact. $\omega^2 = v$ and $(\omega^2)^k = v^k$.



Fast Fourier Transform

Goal. Evaluate a degree $n-1$ polynomial $A(x) = a_0 + \dots + a_{n-1} x^{n-1}$ at its n^{th} roots of unity: $\omega^0, \omega^1, \dots, \omega^{n-1}$.

Divide. Break polynomial up into even and odd powers.

- $A_{\text{even}}(x) = a_0 + a_2x + a_4x^2 + \dots + a_{n/2-2} x^{(n-1)/2}.$
- $A_{\text{odd}}(x) = a_1 + a_3x + a_5x^2 + \dots + a_{n/2-1} x^{(n-1)/2}.$
- $A(x) = A_{\text{even}}(x^2) + x A_{\text{odd}}(x^2).$

Conquer. Evaluate degree $A_{\text{even}}(x)$ and $A_{\text{odd}}(x)$ at the $\frac{1}{2}n^{\text{th}}$ roots of unity: $v^0, v^1, \dots, v^{n/2-1}$.

Combine.

- $A(\omega^k) = A_{\text{even}}(v^k) + \omega^k A_{\text{odd}}(v^k), \quad 0 \leq k < n/2$
- $A(\omega^{k+n}) = A_{\text{even}}(v^k) - \omega^k A_{\text{odd}}(v^k), \quad 0 \leq k < n/2$

$$\begin{array}{c} \uparrow \\ \omega^{k+\frac{1}{2}n} = -\omega^k \end{array}$$

$$v^k = (\omega^k)^2 = (\omega^{k+\frac{1}{2}n})^2$$

FFT Algorithm

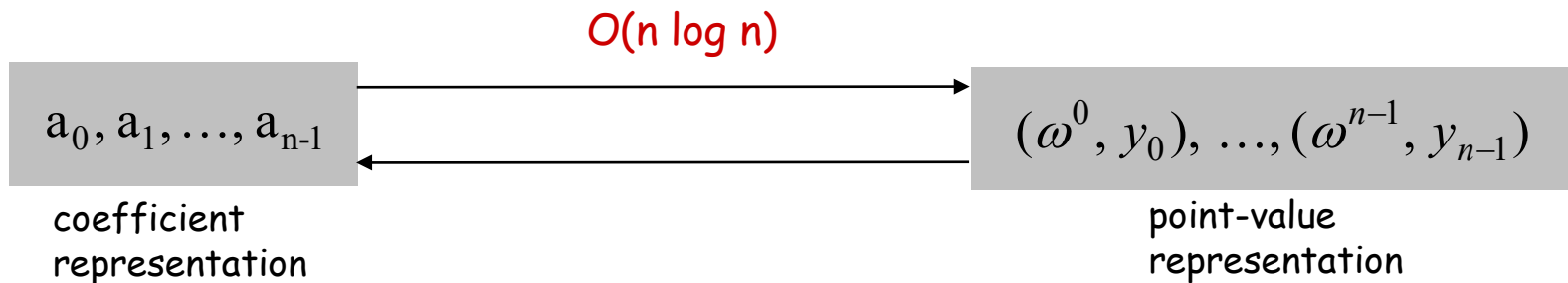
```
fft(n, a0, a1, ..., an-1) {  
    if (n == 1) return a0  
  
    (e0, e1, ..., en/2-1) ← FFT(n/2, a0, a2, a4, ..., an-2)  
    (d0, d1, ..., dn/2-1) ← FFT(n/2, a1, a3, a5, ..., an-1)  
  
    for k = 0 to n/2 - 1 {  
        ωk ← e2πik/n  
        yk ← ek + ωk dk  
        yk+n/2 ← ek - ωk dk  
    }  
  
    return (y0, y1, ..., yn-1)  
}
```

FFT Summary

Theorem. FFT algorithm evaluates a degree $n-1$ polynomial at each of the n^{th} roots of unity in $O(n \log n)$ steps.

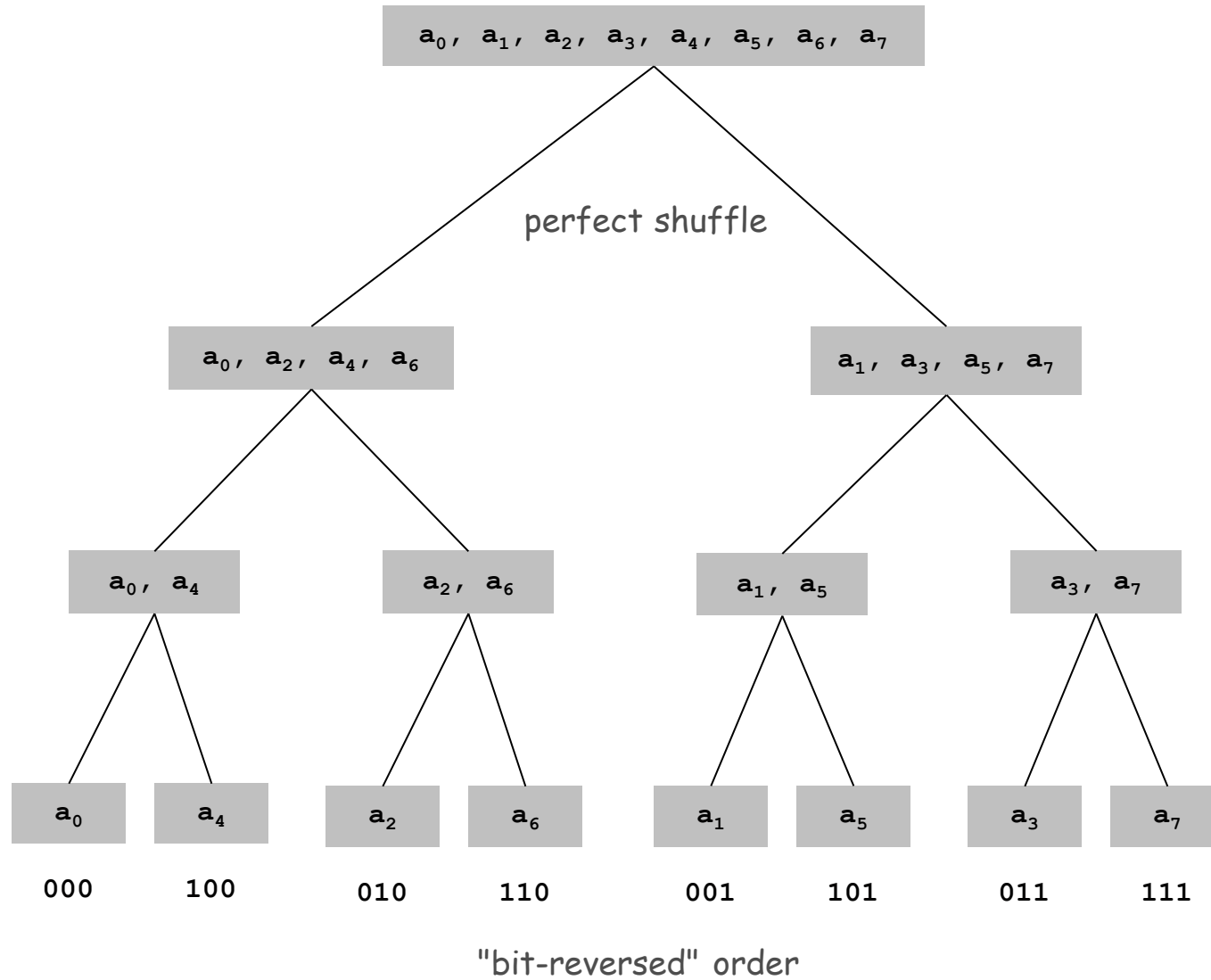
↖
assumes n is a power of 2

Running time. $T(2n) = 2T(n) + O(n) \Rightarrow T(n) = O(n \log n)$.





Recursion Tree



Point-Value to Coefficient Representation: Inverse DFT

Goal. Given the values y_0, \dots, y_{n-1} of a degree $n-1$ polynomial at the n points $\omega^0, \omega^1, \dots, \omega^{n-1}$, find unique polynomial $a_0 + a_1 x + \dots + a_{n-1} x^{n-1}$ that has given values at given points.

$$\begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \\ \vdots \\ a_{n-1} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 & \dots & 1 \\ 1 & \omega^1 & \omega^2 & \omega^3 & \dots & \omega^{n-1} \\ 1 & \omega^2 & \omega^4 & \omega^6 & \dots & \omega^{2(n-1)} \\ 1 & \omega^3 & \omega^6 & \omega^9 & \dots & \omega^{3(n-1)} \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^{n-1} & \omega^{2(n-1)} & \omega^{3(n-1)} & \dots & \omega^{(n-1)(n-1)} \end{bmatrix}^{-1} \begin{bmatrix} y_0 \\ y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_{n-1} \end{bmatrix}$$

\uparrow Inverse DFT \uparrow Fourier matrix inverse $(F_n)^{-1}$

Inverse FFT

Claim. Inverse of Fourier matrix is given by following formula.

$$G_n = \frac{1}{n} \begin{bmatrix} 1 & 1 & 1 & 1 & \dots & 1 \\ 1 & \omega^{-1} & \omega^{-2} & \omega^{-3} & \dots & \omega^{-(n-1)} \\ 1 & \omega^{-2} & \omega^{-4} & \omega^{-6} & \dots & \omega^{-2(n-1)} \\ 1 & \omega^{-3} & \omega^{-6} & \omega^{-9} & \dots & \omega^{-3(n-1)} \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & \omega^{-(n-1)} & \omega^{-2(n-1)} & \omega^{-3(n-1)} & \dots & \omega^{-(n-1)(n-1)} \end{bmatrix}$$

Consequence. To compute inverse FFT, apply same algorithm but use $\omega^{-1} = e^{-2\pi i / n}$ as principal n^{th} root of unity (and divide by n).

Inverse FFT: Proof of Correctness

Claim. F_n and G_n are inverses.

Pf.

$$(F_n G_n)_{kk'} = \frac{1}{n} \sum_{j=0}^{n-1} \omega^{kj} \omega^{-jk'} = \frac{1}{n} \sum_{j=0}^{n-1} \omega^{(k-k')j} = \begin{cases} 1 & \text{if } k = k' \\ 0 & \text{otherwise} \end{cases}$$

summation lemma

Summation lemma. Let ω be a principal n^{th} root of unity. Then

$$\sum_{j=0}^{n-1} \omega^{kj} = \begin{cases} n & \text{if } k \equiv 0 \pmod{n} \\ 0 & \text{otherwise} \end{cases}$$

Pf.

- If k is a multiple of n then $\omega^k = 1 \Rightarrow$ sums to n .
- Each n^{th} root of unity ω^k is a root of $x^n - 1 = (x - 1)(1 + x + x^2 + \dots + x^{n-1})$.
- if $\omega^k \neq 1$ we have: $1 + \omega^k + \omega^{k(2)} + \dots + \omega^{k(n-1)} = 0 \Rightarrow$ sums to 0. ▀

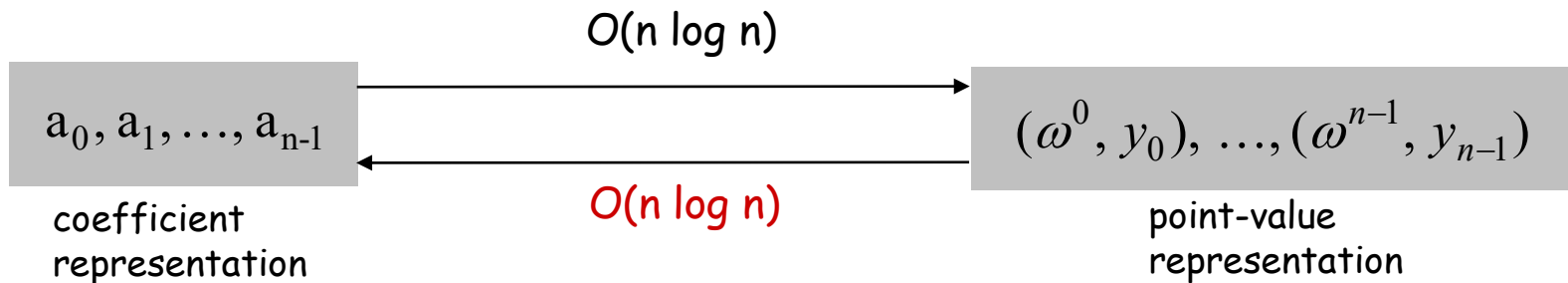
Inverse FFT: Algorithm

```
ifft(n, a0, a1, ..., an-1) {  
    if (n == 1) return a0  
  
    (e0, e1, ..., en/2-1) ← FFT(n/2, a0, a2, a4, ..., an-2)  
    (d0, d1, ..., dn/2-1) ← FFT(n/2, a1, a3, a5, ..., an-1)  
  
    for k = 0 to n/2 - 1 {  
        ωk ← e-2πik/n  
        yk ← (ek + ωk dk) / n  
        yk+n/2 ← (ek - ωk dk) / n  
    }  
  
    return (y0, y1, ..., yn-1)  
}
```

Inverse FFT Summary

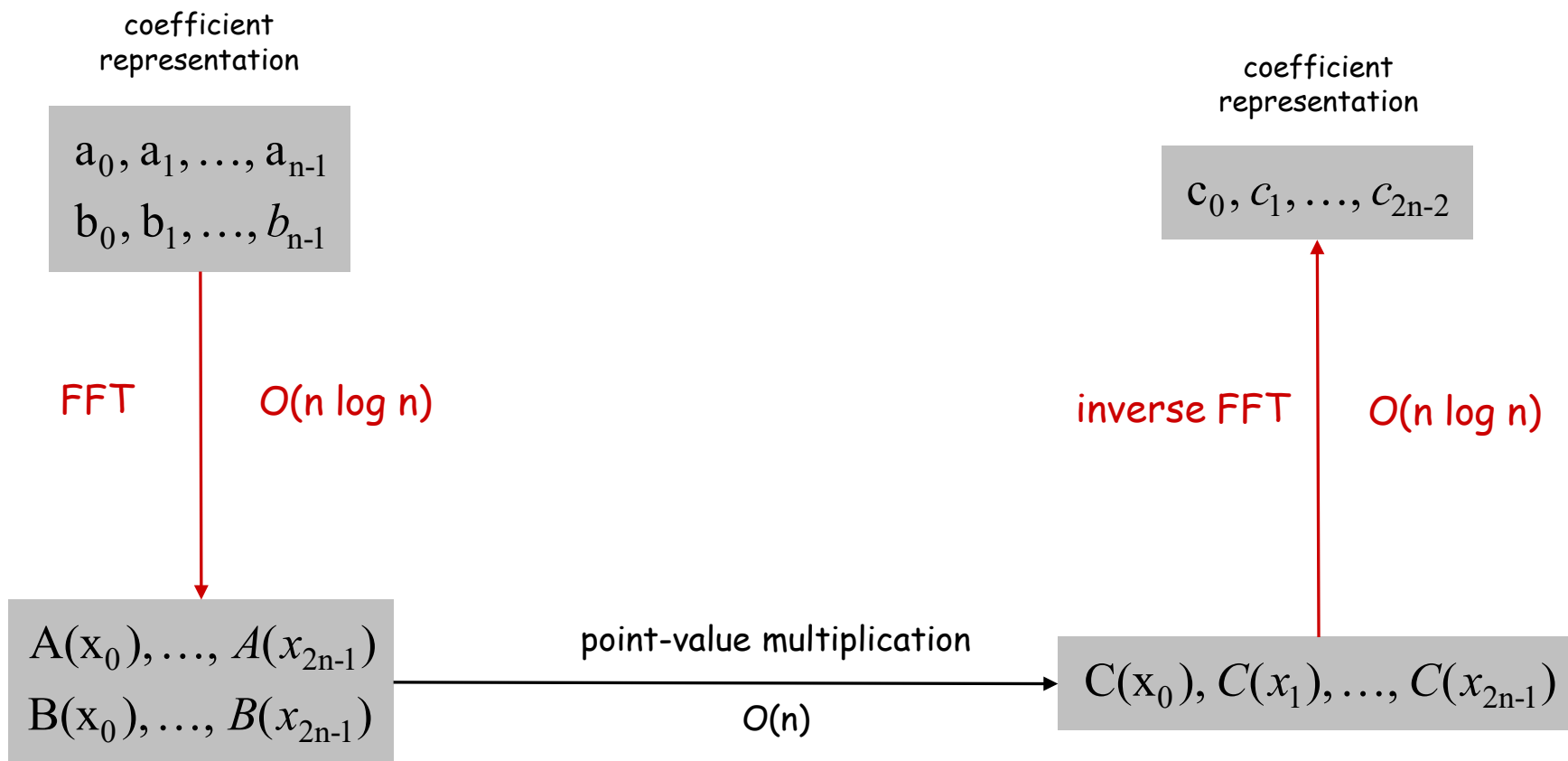
Theorem. Inverse FFT algorithm interpolates a degree $n-1$ polynomial given values at each of the n^{th} roots of unity in $O(n \log n)$ steps.

↑
assumes n is a power of 2



Polynomial Multiplication

Theorem. Can multiply two degree $n-1$ polynomials in $O(n \log n)$ steps.



Integer Multiplication

Integer Multiplication

Integer multiplication. Given two n bit integers $a = a_{n-1} \dots a_1 a_0$ and $b = b_{n-1} \dots b_1 b_0$, compute their product $c = a \times b$.

Convolution algorithm.

- Form two polynomials.
- Note: $a = A(2)$, $b = B(2)$.
- Compute $C(x) = A(x) \times B(x)$.
- Evaluate $C(2) = a \times b$.
- Running time: $O(n \log n)$ complex arithmetic operations.

$$A(x) = a_0 + a_1x + a_2x^2 + \dots + a_{n-1}x^{n-1}$$

$$B(x) = b_0 + b_1x + b_2x^2 + \dots + b_{n-1}x^{n-1}$$

Theory. [Schönhage-Strassen 1971] $O(n \log n \log \log n)$ bit operations.

Theory. [Fürer 2007] $O(n \log n 2^{O(\log^* n)})$ bit operations.



References

References

- Section 5.6 of the text book "algorithm design" by Jon Kleinberg and Eva Tardos
- The original slides were prepared by Kevin Wayne. The slides are distributed by Pearson Addison-Wesley.