

The Fast Fourier Transform

March 3, 2017

Polynomial representations

Representation 1: coefficients.

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

- Addition: $O(n)$
- Evaluation: $O(n)$
Horner's method: $a_0 + x(a_1 + x(a_2 + \cdots + x(a_n) \dots))$
- Multiplication (convolution): $O(n^2)$
 $a_0b_0 + (a_0b_1 + a_1b_0)x + (a_0b_2 + a_1b_1 + a_2b_0)x^2 + \dots$

Polynomial representations

Theorem (Fundamental Theorem of Algebra)

Any nonzero degree- n polynomial with complex coefficients has exactly n complex roots.

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Corollary

A degree- n polynomial is uniquely specified by its value at $n + 1$ distinct x -coordinates.

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Proof.

If f and g are two degree- n polynomials which have the same value at each of $n + 1$ distinct x -coordinates, consider the polynomial $f - g$: it has $n + 1$ roots but degree $\leq n$; the only way for this to happen is if $f - g = 0$, that is, $f = g$. □

Polynomial representations

Representation 2: point-value

$$(x_0, f(x_0)), (x_1, f(x_1)), \dots (x_n, f(x_n))$$

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$$(fg)(x_i) = f(x_i)g(x_i)$$

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- Addition: $O(n)$.
- Evaluation: $O(n^2)$ (Lagrange's method)
- Multiplication: $O(n)$

Converting Between Polynomial Representations

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

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

$$a_0, a_1, \dots, a_{n-1} \begin{array}{c} \xrightarrow{\quad ? \quad} \\ \xleftarrow{\quad ? \quad} \end{array} (x_0, y_0), \dots, (x_{n-1}, y_{n-1})$$

Converting Between Polynomial Representations

Brute Force

Coefficient to point-value. Given a polynomial $a_0 + a_1x + \cdots + a_{n-1}x^{n-1}$, evaluate it at n distinct points x_0, \dots, x_{n-1} .

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

(**Vandermonde** matrix—invertible iff x_i distinct)
 $O(n^2)$ for matrix-vector multiplication.

Coefficient to Point-Value: Intuition

Coefficient to point-value. Given $a_0 + a_1x + \cdots + a_{n-1}x^{n-1}$, evaluate it at n distinct points x_0, \dots, x_{n-1} .

Divide. Break polynomial up into even and odd powers.

- $A(x) = a_0 + a_1x + a_2x^2 + a_3x^3 + a_4x^4 + a_5x^5 + a_6x^6 + a_7x^7$.

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Intuition. For two points, choose ± 1 .

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

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

Coefficient to Point-Value: Intuition

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

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

Coefficient to Point-Value: Intuition

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Intuition. For four points, choose $\pm 1, \pm i$.

$$\begin{aligned}A(1) &= A_{\text{even}}(1) + 1A_{\text{odd}}(1) \\ A(-1) &= A_{\text{even}}(1) - 1A_{\text{odd}}(1) \\ A(i) &= A_{\text{even}}(-1) + iA_{\text{odd}}(-1) \\ A(-i) &= A_{\text{even}}(-1) - iA_{\text{odd}}(-1).\end{aligned}$$

Coefficient to Point-Value: Intuition

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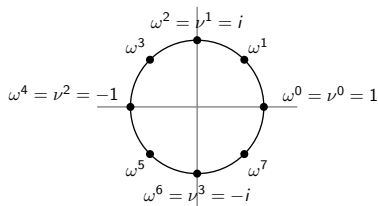
Can evaluate polynomial of degree $\leq n$ at 4 points by evaluating two polynomials of degree $\leq n/2$ at 2 points.

Roots of Unity

Definition

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

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



Discrete Fourier Transform

Coefficient to point-value. Given $a_0 + a_1x + \cdots + a_{n-1}x^{n-1}$, evaluate it at n distinct points x_0, \dots, x_{n-1} .

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

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

(**Fourier matrix F_n**)

Fast Fourier Transform

Goal. Evaluate a degree $n - 1$ polynomial

$A(x) = a_0 + \cdots + a_{n-1}x^{n-1}$ at the n th roots of unity $\omega^0, \dots, \omega^{n-1}$. (Assume n is power of 2.)

Divide. Break up polynomial into even and odd parts.

$$A_{\text{even}}(x) = a_0 + a_2x + a_4x^2 + \cdots + a_{n/2-2}x^{(n-1)/2}$$

$$A_{\text{odd}}(x) = a_1 + a_3x + a_5x^2 + \cdots + a_{n/2-1}x^{(n-1)/2}$$

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

Conquer. Evaluate A_{even} and A_{odd} at $\nu^0, \dots, \nu^{n/2-1}$.

Combine. ($\omega^{k+n/2} = -\omega^k$; $\nu^k = (\omega^k)^2 = (\omega^{k+n/2})^2$)

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

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

FFT: Algorithm

Algorithm 1 FFT

Require: Size n , coefficients a_0, a_1, \dots, a_{n-1}

```
1: if  $n = 1$  then  
2:   return  $a_0$   
3: end if  
4:  $(e_0, e_1, \dots, e_{n/2-1}) \leftarrow \text{FFT}(n/2, a_0, a_2, a_4, \dots, a_{n-2})$   
5:  $(d_0, d_1, \dots, d_{n/2-1}) \leftarrow \text{FFT}(n/2, a_1, a_3, a_5, \dots, a_{n-1})$   
6: for  $k = 0$  to  $n/2 - 1$  do  
7:    $\omega^k \leftarrow e^{2\pi i k / n}$   
8:    $y_k \leftarrow e_k + \omega^k d_k$   
9:    $y_{k+n/2} \leftarrow e_k - \omega^k d_k$   
10: end for  
11: return  $(y_0, y_1, \dots, y_{n-1})$ 
```

FFT Summary

Theorem

The FFT algorithm evaluates a degree $n - 1$ polynomial at each of the n th roots of unity in $O(n \log n)$ time (assuming n is a power of two).

Proof.

$$T(n) = 2T(n/2) + O(n) \implies T(n) = O(n \log n).$$



$$a_0, a_1, \dots, a_{n-1} \begin{array}{c} \xrightarrow{O(n \log n)} \\ \xleftarrow{?} \end{array} (x_0, y_0), \dots, (x_{n-1}, y_{n-1})$$

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 the unique polynomial $a_0 + a_1x + \dots + a_{n-1}x^{n-1}$ that fits the given points.

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

Inverse FFT!

Claim

The inverse of the Fourier matrix F_n is given by

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

Corollary

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

Inverse FFT: Proof of Correctness

Theorem

F_n and G_n are inverse.

Lemma

Let ω be a principal n th root of unity. Then

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

Proof.

- If k is a multiple of n then $\omega^{kj} = 1$.
- Every n th root of unity ω^k is a root of $x^n - 1 = (x - 1)(1 + x + x^2 + \dots + x^{n-1})$. Hence if $\omega^k \neq 1$, we have $1 + \omega^k + \omega^{k(2)} + \dots + \omega^{k(n-1)} = 0$.



Inverse FFT: Proof of Correctness

Theorem

F_n and G_n are inverse.

Note $(F_n)_{ij} = \omega^{ij}$, and $(G_n)_{ij} = \frac{1}{n}\omega^{-ij}$.

Proof.

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



Inverse FFT: Algorithm

Algorithm 2 IFFT

Require: Size n , coefficients a_0, a_1, \dots, a_{n-1}

- 1: **if** $n = 1$ **then**
 - 2: **return** a_0
 - 3: **end if**
 - 4: $(e_0, e_1, \dots, e_{n/2-1}) \leftarrow \text{IFFT}(n/2, a_0, a_2, a_4, \dots, a_{n-2})$
 - 5: $(d_0, d_1, \dots, d_{n/2-1}) \leftarrow \text{IFFT}(n/2, a_1, a_3, a_5, \dots, a_{n-1})$
 - 6: **for** $k = 0$ **to** $n/2 - 1$ **do**
 - 7: $\omega^k \leftarrow e^{-2\pi i k / n}$
 - 8: $y_k \leftarrow (e_k + \omega^k d_k) / n$
 - 9: $y_{k+n/2} \leftarrow (e_k - \omega^k d_k) / n$
 - 10: **end for**
 - 11: **return** $(y_0, y_1, \dots, y_{n-1})$
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Inverse FFT Summary

Theorem

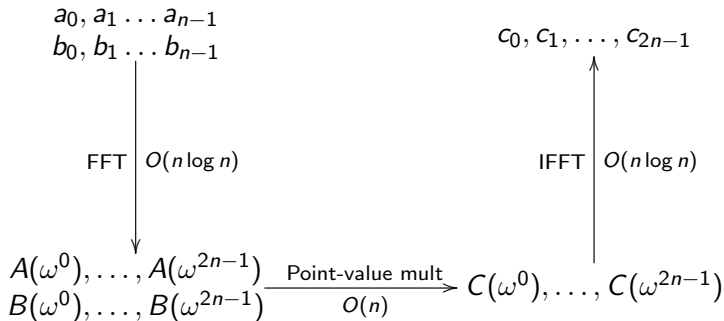
The inverse FFT algorithm interpolates a degree $n - 1$ polynomial, given values at each of the n th roots of unity, in $O(n \log n)$ time.

$$a_0, a_1, \dots, a_{n-1} \begin{array}{c} \xrightarrow{O(n \log n)} \\ \xleftarrow{O(n \log n)} \end{array} (x_0, y_0), \dots, (x_{n-1}, y_{n-1})$$

Polynomial Multiplication

Theorem

We can multiply two degree $n - 1$ polynomials in $O(n \log n)$ time.



FFT in Practice

Fastest Fourier transform in the West.

[Frigo and Johnson, <http://www.fftw.org>]

- Optimized C library.
- Features: DFT, DCT, real, complex, any size, any dimension.
- Won 1999 Wilkinson Prize for Numerical Software.
- Portable, competitive with vendor-tuned code.

Implementation details.

- Instead of executing predetermined algorithm, it evaluates your hardware and uses a special-purpose compiler to generate an optimized algorithm catered to “shape” of the problem.
- $O(n \log n)$, even for prime sizes.

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 $c = a \times b$.

FFT algorithm.

- Form polynomials, **e.g.** $A(x) = a_0 + a_1x + \dots + a_{n-1}x^{n-1}$.
- Note: $a = A(2)$, $b = B(2)$.
- Compute $C(x) = A(x)B(x)$ via FFT.
- Evaluate $C(2) = ab$.
- Running time: $O(n \log n)$ complex **arithmetic operations**.

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

Practice. GNU Multiple Precision Arithmetic Library (GMP) claims to be “the fastest bignum library on the planet.” For multiplication it uses brute force ($O(n^2)$), Karatsuba ($O(n^{1.59})$), Toom-Cook (generalization of Karatsuba), and FFT, depending on n .