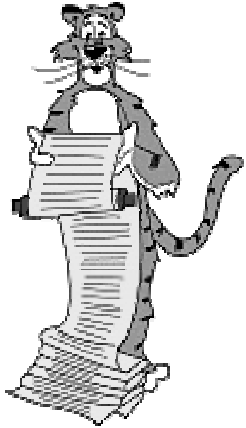


# Fast Fourier Transform



Jean Baptiste Joseph Fourier (1768-1830)

# Fast Fourier Transform

## Applications.

- Perhaps single algorithmic discovery that has had the greatest practical impact in history.
- Optics, acoustics, quantum physics, telecommunications, systems theory, signal processing, speech recognition, data compression.
- Progress in these areas limited by lack of fast algorithms.

## History.

- Cooley-Tukey (1965) revolutionized all of these areas.
- Danielson-Lanczos (1942) efficient algorithm.
- Runge-König (1924) laid theoretical groundwork.
- Gauss (1805, 1866) describes similar algorithm.
- Importance not realized until advent of digital computers.

# Polynomials: Coefficient Representation

## Degree n polynomial.

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

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

## Addition: O(n) ops.

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

## Evaluation: O(n) using Horner's method.

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

## Multiplication (convolution): O(n<sup>2</sup>).

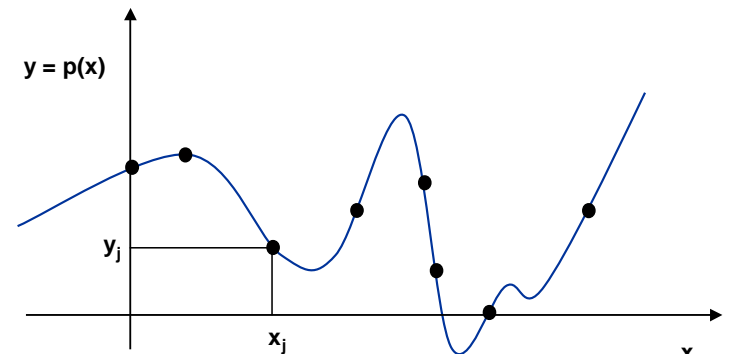
$$p(x) \times q(x) = (a_0b_0) + (a_0b_1 + a_1b_0)x + \dots + \left(\sum_{k=0}^j a_k b_{j-k}\right)x^j + \dots + (a_{n-1}b_{n-1})x^{2n-2}$$

# Polynomials: Point-Value Representation

## Degree n polynomial.

- Uniquely specified by knowing p(x) at n different values of x.

$$\{(x_0, y_0), (x_1, y_1), \dots, (x_{n-1}, y_{n-1})\}, \text{ where } y_k = p(x_k)$$



## Polynomials: Point-Value Representation

Degree  $n$  polynomial.

$$\{(x_0, y_0), (x_1, y_1), \dots, (x_{n-1}, y_{n-1})\}, \text{ where } y_k = p(x_k)$$

$$\{(x_0, z_0), (x_1, z_1), \dots, (x_{n-1}, z_{n-1})\}, \text{ where } z_k = q(x_k)$$

Addition:  $O(n)$ .

$$\{(x_0, y_0 + z_0), (x_1, y_1 + z_1), \dots, (x_{n-1}, y_{n-1} + z_{n-1})\}$$

Multiplication:  $O(n)$ , but need  $2n$  points.

$$\{(x_0, y_0 z_0), (x_1, y_1 z_1), \dots, (x_{2n-1}, y_{2n-1} z_{2n-1})\}$$

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

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

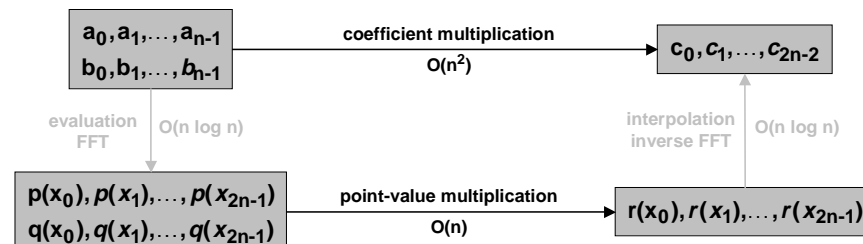
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## Best of Both Worlds

Can we get "fast" multiplication and evaluation?

Representation	Multiplication	Evaluation
coefficient	$O(n^2)$	$O(n)$
point-value	$O(n)$	$O(n^2)$
FFT	$O(n \log n)$	$O(n \log n)$

 **Yes! Convert back and forth between two representations.**



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## Converting Between Representations: Naïve Solution

Evaluation (coefficient to point-value).

- Given a polynomial  $p(x) = a_0 + a_2 x^1 + \dots + a_{n-1} x^{n-1}$ , choose  $n$  distinct points  $\{x_0, x_1, \dots, x_{n-1}\}$  and compute  $y_k = p(x_k)$ , for each  $k$  using Horner's method.
- $O(n^2)$ .

Interpolation (point-value to coefficient).

- Given  $n$  distinct points  $\{x_0, x_1, \dots, x_{n-1}\}$  and  $y_k = p(x_k)$ , compute the coefficients  $\{a_0, a_1, \dots, a_{n-1}\}$  by solving the following linear system of equations.
- Note Vandermonde matrix is invertible iff  $x_k$  are distinct.
- $O(n^3)$ .

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

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## Fast Interpolation: Key Idea

Key idea: choose  $\{x_0, x_1, \dots, x_{n-1}\}$  to make computation easier!

- Set  $x_k = x_j^i$ ?

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

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## Fast Interpolation: Key Idea

Key idea: choose  $\{x_0, x_1, \dots, x_{n-1}\}$  to make computation easier!

- Set  $x_k = x_j$ ?
- Use negative numbers: set  $x_k = -x_j$  so that  $(x_k)^2 = (x_j)^2$ .
  - set  $x_k = -x_{n/2+k}$

$$\begin{pmatrix} 1 & 17 & (17)^2 & (17)^3 & \dots & (17)^{n-1} \\ 1 & 5 & (5)^2 & (5)^3 & \dots & (5)^{n-1} \\ \vdots & \vdots & \vdots & \vdots & \dots & \vdots \\ \hline 1 & -17 & (-17)^2 & (-17)^3 & \dots & (-17)^{n-1} \\ 1 & -5 & (-5)^2 & (-5)^3 & \dots & (-5)^{n-1} \\ \vdots & \vdots & \vdots & \vdots & \dots & \vdots \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_{n-1} \end{pmatrix} = \begin{pmatrix} y_0 \\ y_1 \\ y_2 \\ \vdots \\ y_{n-1} \end{pmatrix}$$

- $E = p_{\text{even}}(x^2) = a_0 + a_2 17^2 + a_4 17^4 + a_6 17^6 + \dots + a_{n-2} 17^{n-2}$
- $O = x p_{\text{odd}}(x^2) = a_1 17 + a_3 17^3 + a_5 17^5 + a_7 17^7 + \dots + a_{n-1} 17^{n-1}$
- $y_0 = E + O, \quad y_{n/2} = E - O$

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## Fast Interpolation: Key Idea

Key idea: choose  $\{x_0, x_1, \dots, x_{n-1}\}$  to make computation easier!

- Set  $x_k = x_j$ ?
- Use negative numbers: set  $x_k = -x_j$  so that  $(x_k)^2 = (x_j)^2$ .
  - set  $x_k = -x_{n/2+k}$
- Use complex numbers: set  $x_k = \omega^k$  where  $\omega$  is  $n^{\text{th}}$  root of unity.
  - $(x_k)^2 = (-x_{n/2+k})^2$
  - $(x_k)^4 = (-x_{n/4+k})^4$
  - $(x_k)^8 = (-x_{n/8+k})^8$

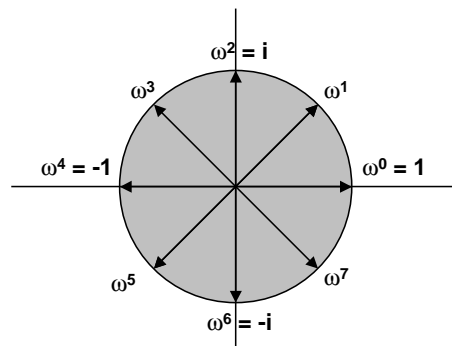
$$\begin{pmatrix} y_0 \\ y_1 \\ y_2 \\ y_3 \\ \vdots \\ y_{n-1} \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 & 1 & \dots & 1 \\ 1 & \omega^1 & \omega^2 & \omega^3 & \dots & \omega^{n-1} \\ 1 & \omega^2 & \omega^4 & \omega^6 & \dots & \omega^{2(n-1)} \\ 1 & \omega^3 & \omega^6 & \omega^9 & \dots & \omega^{3(n-1)}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

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## Roots of Unity

An  $n^{\text{th}}$  root of unity is a complex number  $z$  such that  $z^n = 1$ .

- $\omega = e^{2\pi i/n}$  = principal  $n^{\text{th}}$  root of unity.
- $e^{it} = \cos t + i \sin t$ .
- $i^2 = -1$ .
- There are exactly  $n$  roots of unity:  $\omega^k, k = 0, 1, \dots, n-1$ .



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## Roots of Unity: Properties

L1: Let  $\omega$  be the principal  $n^{\text{th}}$  root of unity. If  $n > 0$ , then  $\omega^{n/2} = -1$ .

- Proof:  $\omega = e^{2\pi i/n} \Rightarrow \omega^{n/2} = e^{\pi i} = -1$ . (Euler's formula)

L2: Let  $n > 0$  be even, and let  $\omega$  and  $\nu$  be the principal  $n^{\text{th}}$  and  $(n/2)^{\text{th}}$  roots of unity. Then  $(\omega^k)^2 = \nu^k$ .

- Proof:  $(\omega^k)^2 = e^{(2k)2\pi i/n} = e^{(k)2\pi i/(n/2)} = \nu^k$ .

L3: Let  $n > 0$  be even. Then, the squares of the  $n$  complex  $n^{\text{th}}$  roots of unity are the  $n/2$  complex  $(n/2)^{\text{th}}$  roots of unity.

- Proof: If we square all of the  $n^{\text{th}}$  roots of unity, then each  $(n/2)^{\text{th}}$  root is obtained exactly twice since:
  - L1  $\Rightarrow \omega^{k+n/2} = -\omega^k$
  - thus,  $(\omega^{k+n/2})^2 = (\omega^k)^2$
  - L2  $\Rightarrow$  both of these =  $\nu^k$
  - $\omega^{k+n/2}$  and  $\omega^k$  have the same square

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## Divide-and-Conquer

Given degree  $n$  polynomial  $p(x) = a_0 + a_1x^1 + a_2x^2 + \dots + a_{n-1}x^{n-1}$ .

- Assume  $n$  is a power of 2, and let  $\omega$  be the principal  $n^{\text{th}}$  root of unity.
- Define even and odd polynomials:
  - $p_{\text{even}}(x) := a_0 + a_2x^1 + a_4x^2 + a_6x^3 + \dots + a_{n-2}x^{n/2-1}$
  - $p_{\text{odd}}(x) := a_1 + a_3x^1 + a_5x^2 + a_7x^3 + \dots + a_{n-1}x^{n/2-1}$
  - $p(x) = p_{\text{even}}(x^2) + x p_{\text{odd}}(x^2)$
- Reduces problem of
  - evaluating degree  $n$  polynomial  $p(x)$  at  $\omega^0, \omega^1, \dots, \omega^{n-1}$  to
  - evaluating two degree  $n/2$  polynomials at:  $(\omega^0)^2, (\omega^1)^2, \dots, (\omega^{n-1})^2$ .
- L3  $\Rightarrow p_{\text{even}}(x)$  and  $p_{\text{odd}}(x)$  only evaluated at  $n/2$  complex  $(n/2)^{\text{th}}$  roots of unity.

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## FFT Algorithm

```

FFT ( $n, a_0, a_1, a_2, \dots, a_{n-1}$ )
if ( $n == 1$ ) //  $n$  is a power of 2
    return  $a_0$ 

 $\omega \leftarrow e^{2\pi i / n}$ 
 $(e_0, e_1, e_2, \dots, e_{n/2-1}) \leftarrow \text{FFT}(n/2, a_0, a_2, a_4, \dots, a_{n-2})$ 
 $(d_0, d_1, d_2, \dots, d_{n/2-1}) \leftarrow \text{FFT}(n/2, a_1, a_3, a_5, \dots, a_{n-1})$ 

for  $k = 0$  to  $n/2 - 1$ 
     $y_k \leftarrow e_k + \omega^k d_k$ 
     $y_{k+n/2} \leftarrow e_k - \omega^k d_k$ 

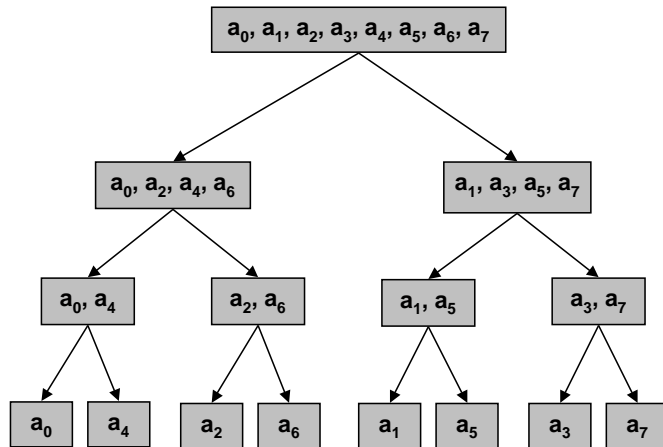
return  $(y_0, y_1, y_2, \dots, y_{n-1})$ 
    
```

$O(n)$  complex multiplies  
if we pre-compute  $\omega^k$ .

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

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## Recursion Tree



"bit-reversed" order

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## Proof of Correctness

Proof of correctness. Need to show  $y_k = p(\omega^k)$  for each  $k = 0, \dots, n-1$ , where  $\omega$  is the principal  $n^{\text{th}}$  root of unity.

$$p(\omega^k) = \sum_{j=0}^{n-1} a_j \omega^{kj}$$

- Base case.  $n = 1 \Rightarrow \omega = 1$ . Algorithm returns  $y_0 = a_0 = a_0 \omega^0$ .
- Induction step. Assume algorithm correct for  $n/2$ .
  - let  $v$  be the principal  $(n/2)^{\text{th}}$  root of unity
  - $e_k = p_{\text{even}}(v^k) = p_{\text{even}}(\omega^{2k})$  by Lemma 2
  - $d_k = p_{\text{odd}}(v^k) = p_{\text{odd}}(\omega^{2k})$  by Lemma 2
  - recall  $p(x) = p_{\text{even}}(x^2) + x p_{\text{odd}}(x^2)$

$$\begin{aligned}
 y_k &= e_k + \omega^k d_k \\
 &= p_{\text{even}}(\omega^{2k}) + \omega^k p_{\text{odd}}(\omega^{2k}) \\
 &= p(\omega^k)
 \end{aligned}$$

$$\begin{aligned}
 y_{k+n/2} &= e_k - \omega^k d_k \\
 &= p_{\text{even}}(\omega^{2k}) - \omega^k p_{\text{odd}}(\omega^{2k}) \\
 &= p_{\text{even}}(\omega^{2k}) + \omega^{k+n/2} p_{\text{odd}}(\omega^{2k}) \\
 &= p_{\text{even}}(\omega^{2k+n}) + \omega^{k+n/2} p_{\text{odd}}(\omega^{2k+n}) \\
 &= p(\omega^{k+n/2})
 \end{aligned}$$

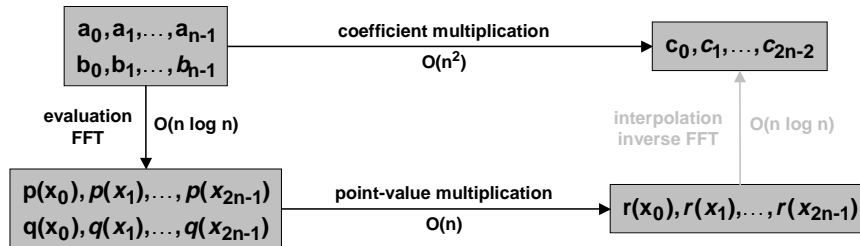
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## Best of Both Worlds

Can we get "fast" multiplication and evaluation?

Representation	Multiplication	Evaluation
coefficient	$O(n^2)$	$O(n)$
point-value	$O(n)$	$O(n^2)$
FFT	$O(n \log n)$	$O(n \log n)$

 **Yes! Convert back and forth between two representations.**



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## Inverse FFT

Forward FFT: given  $\{a_0, a_1, \dots, a_{n-1}\}$ , compute  $\{y_0, y_1, \dots, y_{n-1}\}$ .

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

Inverse FFT: given  $\{y_0, y_1, \dots, y_{n-1}\}$  compute  $\{a_0, a_1, \dots, a_{n-1}\}$ .

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

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## Inverse FFT

Great news: same algorithm as FFT, except use  $\omega^{-1}$  as "principal"  $n^{\text{th}}$  root of unity (and divide by  $n$ ).

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

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

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## Inverse FFT: Proof of Correctness

Summation lemma. Let  $\omega$  be a primitive  $n^{\text{th}}$  root of unity. Then

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

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

Claim:  $F_n$  and  $F_n^{-1}$  are inverses.

$$\begin{aligned} (F_n F_n^{-1})_{i i'} &= \sum_{j=0}^{n-1} \omega^{ij} \frac{\omega^{-j i'}}{n} \\ &= \frac{1}{n} \sum_{j=0}^{n-1} \omega^{(i-i')j} \\ &= \begin{cases} 1 & \text{if } i = i' \\ 0 & \text{otherwise} \end{cases} \end{aligned}$$

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## Inverse FFT: Algorithm

```

IFFT (n, a0, a1, a2, . . . , an-1)
if (n == 1) // n is a power of 2
    return a0

ω ← e-2πi/n
(e0, e1, e2, . . . , en/2-1) ← FFT(n/2, a0, a2, a4, . . . , an-2)
(d0, d1, d2, . . . , dn/2-1) ← FFT(n/2, a1, a3, a5, . . . , an-1)

for k = 0 to n/2 - 1
    yk ← (ek + ωk dk) / n
    yk+n/2 ← (ek - ωk dk) / n

return (y0, y1, y2, . . . , yn-1)
    
```

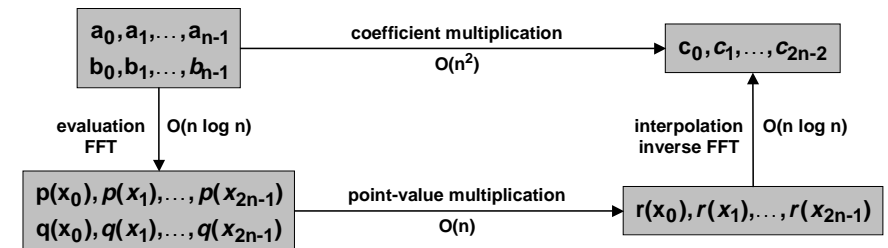
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Can we get "fast" multiplication and evaluation?

Representation	Multiplication	Evaluation
coefficient	$O(n^2)$	$O(n)$
point-value	$O(n)$	$O(n^2)$
FFT	$O(n \log n)$	$O(n \log n)$

 **Yes! Convert back and forth between two representations.**



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## Integer Arithmetic

Multiply two  $n$ -digit integers:  $a = a_{n-1} \dots a_1 a_0$  and  $b = b_{n-1} \dots b_1 b_0$ .

- Form two degree  $n$  polynomials.

- Note:  $a = p(10)$ ,  $b = q(10)$ .

$$p(x) = \sum_{j=0}^{n-1} a_j x^j$$

$$q(x) = \sum_{j=0}^{n-1} b_j x^j$$

- Compute product using FFT in  $O(n \log n)$  steps.

- Evaluate  $r(10) = a \times b$ .

$$r(x) = p(x) \times q(x)$$

- Problem:**  $O(n \log n)$  complex arithmetic steps.

**Solution.**

- Strassen (1968): carry out arithmetic to suitable precision.

$$- T(n) = O(n T(\log n)) \Rightarrow T(n) = O(n \log n (\log \log n)^{1+\epsilon})$$

- Schönhage-Strassen (1971): use modular arithmetic.

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

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