

# CSE 421: Introduction to Algorithms

Fast Fourier Transform  
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## Integer Multiplication

- Given:
  - Two  $n$ -bit integers  $X$  and  $Y$ 
    - $X = a_0 + a_1 2 + a_2 2^2 + \dots + a_{n-2} 2^{n-2} + a_{n-1} 2^{n-1}$
    - $Y = b_0 + b_1 2 + b_2 2^2 + \dots + b_{n-2} 2^{n-2} + b_{n-1} 2^{n-1}$
- Compute:
  - $2n-1$ -bit integer  $XY$ 
    - $XY = a_0 b_0 + (a_0 b_1 + a_1 b_0) 2 + (a_0 b_2 + a_1 b_1 + a_2 b_0) 2^2 + \dots + (a_{n-2} b_{n-1} + a_{n-1} b_{n-2}) 2^{2n-3} + a_{n-1} b_{n-1} 2^{2n-2}$
- Last time: Karatsuba's Algorithm beats naive algorithm, using  $O(n^\alpha)$  where  $\alpha = \log_2 3 = 1.59\dots$

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## Polynomial Multiplication

- Given:
  - Degree  $n-1$  polynomials  $P$  and  $Q$ 
    - $P = a_0 + a_1 x + a_2 x^2 + \dots + a_{n-2} x^{n-2} + a_{n-1} x^{n-1}$
    - $Q = b_0 + b_1 x + b_2 x^2 + \dots + b_{n-2} x^{n-2} + b_{n-1} x^{n-1}$
- Compute:
  - Degree  $2n-2$  Polynomial  $PQ$ 
    - $PQ = a_0 b_0 + (a_0 b_1 + a_1 b_0) x + (a_0 b_2 + a_1 b_1 + a_2 b_0) x^2 + \dots + (a_{n-2} b_{n-1} + a_{n-1} b_{n-2}) x^{2n-3} + a_{n-1} b_{n-1} x^{2n-2}$
- Obvious Algorithm, just like Integer Mult.:
  - Compute all  $a_i b_j$  and collect terms
  - $\Theta(n^2)$  time

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## Divide and Conquer

- Assume  $n=2k$ 
  - $P = P_0 + P_1 x^k$  where  $P_0$  and  $P_1$  are degree  $k-1$  polys
  - Similarly  $Q = Q_0 + Q_1 x^k$
- $PQ = (P_0 + P_1 x^k)(Q_0 + Q_1 x^k) = P_0 Q_0 + (P_1 Q_0 + P_0 Q_1) x^k + P_1 Q_1 x^{2k}$
- Naïve: 4 sub-problems of size  $k=n/2$  plus linear combining  $T(n)=4T(n/2)+cn$  Solution  $T(n) = \Theta(n^2)$
- Karatsuba's: 3 instead of 4:  $A \leftarrow P_0 Q_0$   $B \leftarrow P_1 Q_1$   $C \leftarrow (P_0 + P_1)(Q_0 + Q_1)$  and then  $C - A - B = P_1 Q_0 + P_0 Q_1$  so  $T(n) = 3T(n/2) + cn$  and  $T(n) = O(n^\alpha)$  where  $\alpha = \log_2 3 = 1.59\dots$

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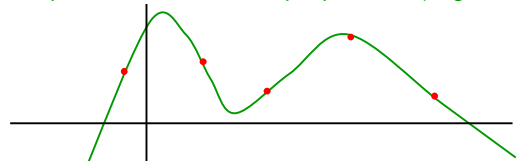
## Integer and Polynomial Multiplication

- Naïve:  $\Theta(n^2)$
- Karatsuba:  $\Theta(n^{1.59\dots})$
- Best known:  $\Theta(n \log n)$ 
  - "Fast Fourier Transform"
  - FFT widely used for signal processing
  - Used in practice in symbolic manipulation systems like Maple
  - MUCH easier for Polynomial Multiplication than for integer multiplication because of ugly details with carries, etc.
    - Schonhage-Strassen (1971) gives  $\Theta(n \log n \log \log n)$
    - Furer (2007) gives  $\Theta(n \log n 2^{O(\log^3 n)})$
    - Harvey, van der Hoeven (2019) finally got  $\Theta(n \log n)$

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## Hints towards FFT: Interpolation

- 2 points determine a unique line (degree 1)
- 3 points determine a unique parabola (degree 2)



Given set of values at  $n$  points  
Can find unique degree  $n-1$  polynomial going through these points

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## Multiplying Polynomials by Evaluation & Interpolation

- Any degree  $n-1$  polynomial  $R(y)$  is determined by  $R(y_0), \dots, R(y_{n-1})$  for any  $n$  distinct  $y_0, \dots, y_{n-1}$
- To compute  $PQ$  (assume degree at most  $n/2-1$ )
  - Evaluate  $P(y_0), \dots, P(y_{n-1})$
  - Evaluate  $Q(y_0), \dots, Q(y_{n-1})$
  - Multiply values  $P(y_i)Q(y_i)$  for  $i=0, \dots, n-1$
  - Interpolate to recover  $PQ$

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## Interpolation

- Given values of degree  $n-1$  polynomial  $R$  at  $n$  distinct points  $y_0, \dots, y_{n-1}$ 
  - $R(y_0), \dots, R(y_{n-1})$
- Compute coefficients  $c_0, \dots, c_{n-1}$  such that
  - $R(x) = c_0 + c_1x + c_2x^2 + \dots + c_{n-1}x^{n-1}$
- System of linear equations in  $c_0, \dots, c_{n-1}$ 

$$c_0 + c_1y_0 + c_2y_0^2 + \dots + c_{n-1}y_0^{n-1} = R(y_0) \quad \text{known}$$

$$c_0 + c_1y_1 + c_2y_1^2 + \dots + c_{n-1}y_1^{n-1} = R(y_1) \quad \text{known}$$

...

$$c_0 + c_1y_{n-1} + c_2y_{n-1}^2 + \dots + c_{n-1}y_{n-1}^{n-1} = R(y_{n-1}) \quad \text{unknown}$$

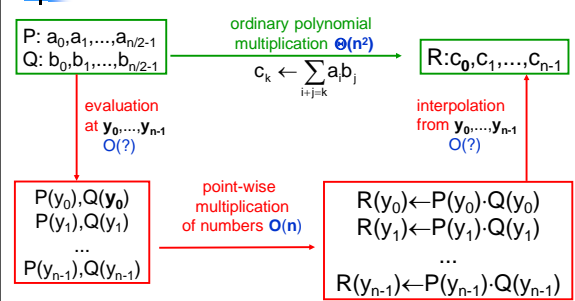
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## Interpolation: $n$ equations in $n$ unknowns

- Matrix form of the linear system
 
$$\begin{pmatrix} 1 & y_0 & y_0^2 & \dots & y_0^{n-1} \\ 1 & y_1 & y_1^2 & \dots & y_1^{n-1} \\ \dots & \dots & \dots & \dots & \dots \\ 1 & y_{n-1} & y_{n-1}^2 & \dots & y_{n-1}^{n-1} \end{pmatrix} \begin{pmatrix} c_0 \\ c_1 \\ c_2 \\ \dots \\ c_{n-1} \end{pmatrix} = \begin{pmatrix} R(y_0) \\ R(y_1) \\ \dots \\ R(y_{n-1}) \end{pmatrix}$$
- Fact: Determinant of the matrix is  $\prod_{i < j} (y_i - y_j)$  which is not 0 since points are distinct
  - System has a unique solution  $c_0, \dots, c_{n-1}$

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## Hints towards FFT: Evaluation & Interpolation



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## Karatsuba's algorithm and evaluation and interpolation

- Karatsuba's algorithm can be thought of as a way of multiplying two degree 1 polynomials (which have 2 coefficients) using only 3 multiplications
  - $PQ = (P_0 + P_1z)(Q_0 + Q_1z)$ 

$$= P_0Q_0 + (P_1Q_0 + P_0Q_1)z + P_1Q_1z^2$$
  - Evaluate at 0, 1 plus compute  $P_1Q_1$ 
    - $A = P(0)Q(0) = P_0Q_0$
    - $B = P_1Q_1$
    - $C = P(1)Q(1) = (P_0 + P_1)(Q_0 + Q_1)$
  - Alternative: replace  $B$  by the following: Evaluate at -1
    - $D = P(-1)Q(-1) = (P_0 - P_1)(Q_0 - Q_1)$
  - Interpolating, product is  $A + (C-D)/2 z + [(C+D)/2 - A] z^2$

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## Evaluation at Special Points

- Evaluation of polynomial at 1 point takes  $O(n)$  time
  - So  $2n$  points (naively) takes  $O(n^2)$ —no savings
  - But the algorithm works no matter what the points are...
- So... choose points that are related to each other so that evaluation problems can share subproblems

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### The key idea: Evaluate at related points

- $$P(x) = a_0 + a_1x + a_2x^2 + a_3x^3 + a_4x^4 + \dots + a_{n-1}x^{n-1}$$

$$= a_0 + a_2x^2 + a_4x^4 + \dots + a_{n-2}x^{n-2}$$

$$+ a_1x + a_3x^3 + a_5x^5 + \dots + a_{n-1}x^{n-1}$$

$$= P_{\text{even}}(x^2) + x P_{\text{odd}}(x^2)$$
- $$P(-x) = a_0 - a_1x + a_2x^2 - a_3x^3 + a_4x^4 - \dots - a_{n-1}x^{n-1}$$

$$= a_0 + a_2x^2 + a_4x^4 + \dots + a_{n-2}x^{n-2}$$

$$- (a_1x + a_3x^3 + a_5x^5 + \dots + a_{n-1}x^{n-1})$$

$$= P_{\text{even}}(x^2) - x P_{\text{odd}}(x^2)$$

where  $P_{\text{even}}(z) = a_0 + a_2z + a_4z^2 + \dots + a_{n-2}z^{n/2-1}$   
and  $P_{\text{odd}}(z) = a_1 + a_3z + a_5z^2 + \dots + a_{n-1}z^{n/2-1}$

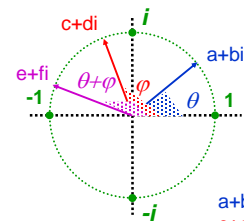
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### The key idea: Evaluate at related points

- So... if we have half the points as negatives of the other half
  - i.e.,  $y_{n/2} = -y_0, y_{n/2+1} = -y_1, \dots, y_{n-1} = -y_{n/2-1}$
 then we can reduce the size  $n$  problem of evaluating degree  $n-1$  polynomial  $P$  at  $n$  points to evaluating 2 degree  $n/2 - 1$  polynomials  $P_{\text{even}}$  and  $P_{\text{odd}}$  at  $n/2$  points  $y_0^2, \dots, y_{n/2-1}^2$  and recombine answers with  $O(1)$  extra work per point
- But to use this idea recursively we need half of  $y_0^2, \dots, y_{n/2-1}^2$  to be negatives of the other half
  - If  $y_{n/4}^2 = -y_0^2$ , say, then  $(y_{n/4}/y_0)^2 = -1$
  - Motivates use of complex numbers as evaluation points

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### Complex Numbers $i^2 = -1$



To multiply complex numbers:  
1. add angles  
2. multiply lengths  
(all length 1 here)

$$e+fi = (a+bi)(c+di)$$

$$a+bi = \cos \theta + i \sin \theta = e^{i\theta}$$

$$c+di = \cos \phi + i \sin \phi = e^{i\phi}$$

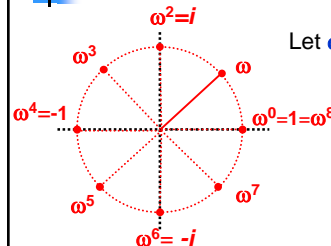
$$e+fi = \cos(\theta+\phi) + i \sin(\theta+\phi) = e^{i(\theta+\phi)}$$

$$e^{2\pi i} = 1$$

$$e^{\pi i} = -1$$

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### Primitive $n^{\text{th}}$ root of 1 $\omega = \omega_n = e^{i2\pi/n}$



Let  $\omega = \omega_n = e^{i2\pi/n}$   
 $= \cos(2\pi/n) + i \sin(2\pi/n)$

$$i^2 = -1$$

$$e^{2\pi i} = 1$$

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### Facts about $\omega = e^{2\pi i/n}$ for even $n$

- $\omega = e^{2\pi i/n}$  for  $i = \sqrt{-1}$
- $\omega^n = 1$
- $\omega^{n/2} = -1$
- $\omega^{n/2+j} = -\omega^j$  for all values of  $j$
- $\omega^2 = e^{2\pi i/k}$  where  $k=n/2$
- $\omega^j = \cos(2j\pi/n) + i \sin(2j\pi/n)$  so can compute with powers of  $\omega$
- $\omega^j$  is a root of  $x^n - 1 = (x-1)(x^{n-1} + x^{n-2} + \dots + 1) = 0$   
but for  $j \neq 0, \omega^j \neq 1$  so  $\omega^{j(n-1)} + \omega^{j(n-2)} + \dots + 1 = 0$

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### The key idea for $n$ even

- $$P(\omega) = a_0 + a_1\omega + a_2\omega^2 + a_3\omega^3 + a_4\omega^4 + \dots + a_{n-1}\omega^{n-1}$$

$$= a_0 + a_2\omega^2 + a_4\omega^4 + \dots + a_{n-2}\omega^{n-2}$$

$$+ a_1\omega + a_3\omega^3 + a_5\omega^5 + \dots + a_{n-1}\omega^{n-1}$$

$$= P_{\text{even}}(\omega^2) + \omega P_{\text{odd}}(\omega^2)$$
- $$P(-\omega) = a_0 - a_1\omega + a_2\omega^2 - a_3\omega^3 + a_4\omega^4 - \dots - a_{n-1}\omega^{n-1}$$

$$= a_0 + a_2\omega^2 + a_4\omega^4 + \dots + a_{n-2}\omega^{n-2}$$

$$- (a_1\omega + a_3\omega^3 + a_5\omega^5 + \dots + a_{n-1}\omega^{n-1})$$

$$= P_{\text{even}}(\omega^2) - \omega P_{\text{odd}}(\omega^2)$$

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

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## The recursive idea for n a power of 2

- Goal:
  - Evaluate  $P$  at  $1, \omega, \omega^2, \omega^3, \dots, \omega^{n-1}$
- Now
  - $P_{\text{even}}$  and  $P_{\text{odd}}$  have degree  $n/2-1$  where
  - $P(\omega^k) = P_{\text{even}}(\omega^{2k}) + \omega^k P_{\text{odd}}(\omega^{2k})$
  - $P(-\omega^k) = P_{\text{even}}(\omega^{2k}) - \omega^k P_{\text{odd}}(\omega^{2k})$
- Recursive Algorithm
  - Evaluate  $P_{\text{even}}$  at  $1, \omega^2, \omega^4, \dots, \omega^{n-2}$
  - Evaluate  $P_{\text{odd}}$  at  $1, \omega^2, \omega^4, \dots, \omega^{n-2}$
  - Combine to compute  $P$  at  $1, \omega, \omega^2, \dots, \omega^{n/2-1}$
  - Combine to compute  $P$  at  $-1, -\omega, -\omega^2, \dots, -\omega^{n/2-1}$  (i.e. at  $\omega^{n/2}, \omega^{n/2+1}, \omega^{n/2+2}, \dots, \omega^{n-1}$ )

$\omega^2$  is  $e^{2\pi i/n}$  where  $k=n/2$  so problems are of same type but smaller size

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## Analysis and more

- Run-time
  - $T(n) = 2 \cdot T(n/2) + cn$  so  $T(n) = O(n \log n)$
- So much for evaluation ... what about interpolation?
  - Given
    - $r_0 = R(1), r_1 = R(\omega), r_2 = R(\omega^2), \dots, r_{n-1} = R(\omega^{n-1})$
  - Compute
    - $c_0, c_1, \dots, c_{n-1}$  s.t.  $R(x) = c_0 + c_1 x + \dots + c_{n-1} x^{n-1}$

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## Interpolation $\approx$ Evaluation: strange but true

- Non-obvious fact:
  - If we define a new polynomial  $S(x) = r_0 + r_1 x + r_2 x^2 + \dots + r_{n-1} x^{n-1}$  where  $r_0, r_1, \dots, r_{n-1}$  are the evaluations of  $R$  at  $1, \omega, \dots, \omega^{n-1}$
  - Then  $c_k = S(\omega^k)/n$  for  $k=0, \dots, n-1$
  - Relies on the fact the interpolation (inverse) matrix has  $ij$  entry  $\omega^{(ij)/n}$  instead of  $\omega^j$
- So...
  - evaluate  $S$  at  $1, \omega^{-1}, \omega^{-2}, \dots, \omega^{-(n-1)}$  then divide each answer by  $n$  to get the  $c_0, \dots, c_{n-1}$
  - $\omega^{-1}$  behaves just like  $\omega$  did so the same  $O(n \log n)$  evaluation algorithm applies!

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## Why this is called the discrete Fourier transform

- Real Fourier series
  - Given a real valued function  $f$  defined on  $[0, 2\pi]$  the Fourier series for  $f$  is given by  $f(x) = a_0 + a_1 \cos(x) + a_2 \cos(2x) + \dots + a_m \cos(mx) + \dots$  where
 
$$a_m = \frac{1}{2\pi} \int_0^{2\pi} f(x) \cos(mx) dx$$
  - is the component of  $f$  of frequency  $m$
  - In signal processing and data compression one ignores all but the components with large  $a_m$  and there aren't many since

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## Why this is called the discrete Fourier transform

- Complex Fourier series
  - Given a function  $f$  defined on  $[0, 2\pi]$  the complex Fourier series for  $f$  is given by  $f(z) = b_0 + b_1 e^{iz} + b_2 e^{2iz} + \dots + b_m e^{miz} + \dots$  where
 
$$b_m = \frac{1}{2\pi} \int_0^{2\pi} f(z) e^{-miz} dz$$
  - is the component of  $f$  of frequency  $m$
  - If we **discretize** this integral using values at  $n$   $2\pi/n$  apart equally spaced points between  $0$  and  $2\pi$  we get

$$\bar{b}_m = \frac{1}{n} \sum_{k=0}^{n-1} f_k e^{-2kmiz/n} = \frac{1}{n} \sum_{k=0}^{n-1} f_k \omega^{-km} \text{ where } f_k = f(2k\pi/n)$$

just like interpolation!

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