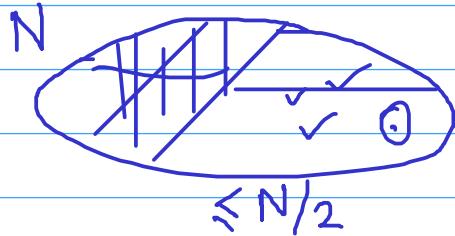


Binary Search (and variants)

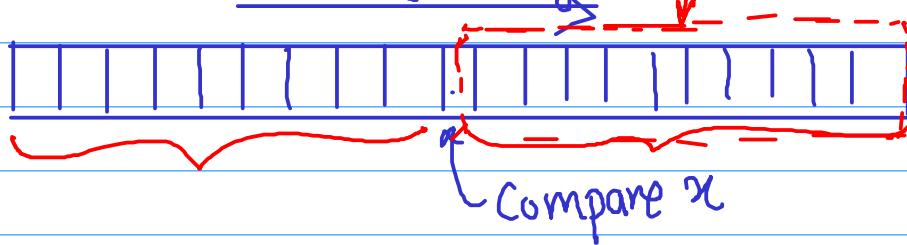
Applicable when with one query, the search space size can be reduced by half.



No. of queries $\log N$

Classic Example:

Given a sorted integer array A and an integer x, find the location of x in A (or say that not present)



Other Examples:

① Looking for a word in a dictionary

② Debugging Code

③ Rice cooking

① Finding [Square root] of an integer a.

Start with a guess $x \in [1, a]$

Check $\underline{x^2 > a}$

No. of rounds $\underline{\log a}$.

What if we want to output the square root as a real number?

Search space? $a \cdot 2^k$

$$\text{No. of queries} \quad \log(a \cdot 2^k) = \log a + k$$

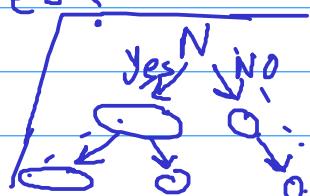
k is the no. of precision bits you are asked for.

_____ x _____

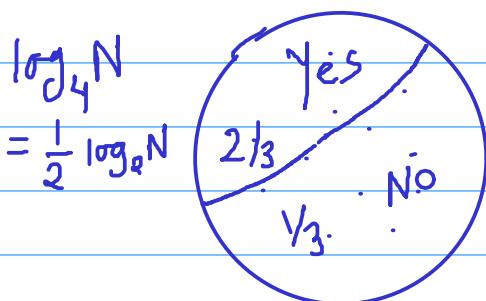
Better than binary search?

Is there any searching scheme that can work in less than $\log_2 N$ queries?

Ans: No.



Argument: If the query is Yes/No-type then it gives only one bit of information



In worst case your new search space size $\geq \underline{\frac{N}{2}}$.

$$\frac{1}{2} \cdot \frac{1}{2} \cdot \dots \cdot \frac{1}{2}^{\log_2 N}$$

Homework ① You have two sorted arrays of integers. Assume all the entries are distinct in/across the two arrays.

Find the median of the union of two arrays by accessing only $O(\log n)$ entries.

n = Size of arrays

HW2

Given an array of integers, and a number S , find a pair of integers in the array whose sum is S .

Trivial : $\binom{n}{2} = O(n^2)$

Another application: suppose for an optimization problem you can test whether the optimal value is greater than a given number W :

How much time will you take to find the optimal value?

$\log(\text{initial Range})$

What if the range is unknown? (Exponential Search)

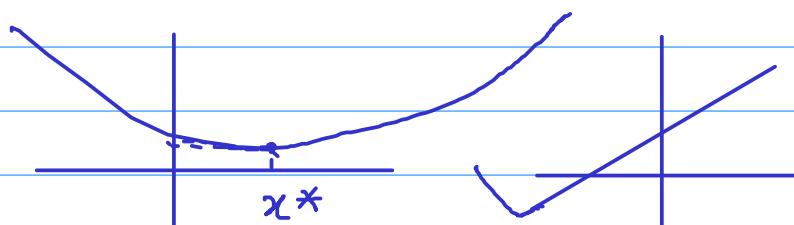
Can we find the optimal value in

$O(\log(\text{optimal value}))$ queries?

Ans: query with $W=0, 2, 2^2, 2^3, \dots$ and stop when optimal value is $\leq 2^k$.

HW3

Let $f: \mathbb{R} \rightarrow \mathbb{R}$ be a convex function



[equivalently
 $f'(x)$ is
non-decreasing]

$f(x)$ is not given explicitly. You can query for $f(x)$ and $f'(x)$ at any point.

Find the point minimizing $f(x)$ (given the promise that x^* exists)

Comment: $f(x) = e^x$ is convex, but has no minimizing point

Analyzing Algorithms

- Comparing different algorithms

Running time:

Why not implement and see?

- Too many inputs
- Too many algorithms
- Processor dependent

Will count the number of basic operations.
(addition / comparison)

Asymptotic Analysis

for Input size n running time $f(n)$

$$3n - 5 \rightarrow O(n)$$

$$2n + 3, 5n^2 - n + 4 \rightarrow O(n)$$

Big-O notation.

$O(n), O(n^2), O(n \log n), O(2^n)$

$\Theta(n)$

$$f(n) = \sqrt{n} \leftarrow O(n)$$

$\cancel{\Theta(n)}$

$$d \cdot n \leq f(n) \leq c \cdot n$$

$$A \leftarrow 100 \cdot n \leftarrow O(n)$$

$$B \leftarrow n^2 + 9 \leftarrow O(n^2)$$

Worst Case Analysis (take the worst bound over all inputs of a fixed size)

Why?

① why not average case analysis?

② It's nice to have worst case guarantees and in many cases we can get it.

Describing Algorithms

Pseudocode / Textual description.
(error prone)
Implementation details.

Combination of the two

Fri, Jan 6

In an array are there two entries whose sum is S.

Sorting + Binary search vs Hashing

Sort the array
for each a_i ,

binary search for $S - a_i$
 $\Theta(n \log n)$

comparison

$\Theta(n \log n \log N)$

Insert all entries into a Hash table
for each a_i

Search $S - a_i$ in this Hash table.

$\Theta(n)$

$h(a_i) = \underline{a_i \bmod n}$

$\Theta(n(\log N \cdot \log n))$

First Design Idea

Reducing to a subproblem



Same problem on a smaller input
(subarray, subgraph)

Assume that you are already given a solution for the subproblem and using that try to build a solution for the original problem.

Solve the subproblem using the same strategy.

Advantage: useful in analyzing the algorithm.

Implementation: recursive or iterative.

Prob 1:

Find minimum value in a given integer array.



subproblem : min among first $n-1$ values
 m_{n-1}

$$\rightarrow m_n = \min(m_{n-1}, a_n)$$

Recursive

$M(A, i)$:
output min value
among first i values.
if $i=1 \Rightarrow$ output $A[i]$

else
Output $\min(M(A, i-1), A[i])$

Iterative

$m \leftarrow A[1]$
for $i = 2$ to N
 $m \leftarrow \min(m, A[i])$

Maximum Subarray Sum problem.



Subarray - contiguous subset.

Given an integer array (possibly with negative entries), find the subarray with maximum sum.

Naive Algorithm:

Go over all possible subarrays.
and find the one with maximum sum

$O(n^3)$



curr-max;

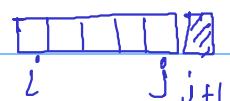
for start = 1 to n.

$s = 0$

for end = start to n

$s = s + A[\underline{\underline{end}}]$;

curr-max := Max (curr-max, s)



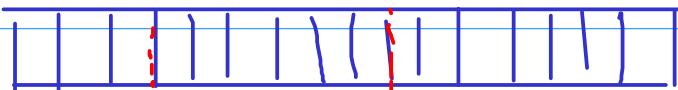
New running time = $O(n^2)$.

Can we improve?

Subproblem



Think about how the subproblem idea can be applied here.



Assume $\text{MaxSubarray}(n-1)$ is given.

Can we compute $\text{MaxSubarray}(n)$ using it?

Subarrays of A are of two kinds.

- which include $A[n]$
- ✓ which are contained in $A[1 \dots n-1]$

$$\text{MaxSubarray}(n) = \max \left\{ \begin{array}{l} \text{MaxSubarray}(n-1) \\ \{ \sum [1 \dots n] \\ \sum [2 \dots n] \\ \vdots \\ \sum [n-1 \dots n] \\ \sum [n] \} \end{array} \right\} O(n)$$

$$T(n) = T(n-1) + O(n) \Rightarrow T(n) = O(n^2)$$

Improvements?

Ask the subproblem to solve more.

$$\max(\sum[n-1..n], \sum[n-2..n], \dots, \underline{\sum[1..n]})$$

$$= \max(A[n] + \sum[n-1], \underline{A[n]} + \underline{\sum[n-2..n-1]}, \dots, \underline{\underline{A[n]}} + \underline{\underline{\sum[1..n-1]}})$$

$$= \underline{A[n]} + \max(\sum[n-1], \sum[n-2..n-1], \dots, \underline{\sum[1..n-1]})$$

Subproblem: $\text{maxsubarray}(n-1)$, $\text{maxsuffixsum}(n-1)$

$\text{max Subarray}(n) = \max$

$$T(n) = T(n-1) + O(1)$$

$$T(n) = O(n)$$

$$\begin{cases} \text{maxsubarray}(n-1) \\ A[n] + \text{maxsuffixsum}(n-1) \\ A[n] \end{cases}$$

$$\text{max suffixsum}(n) = \max(A[n] + \text{maxsuffixsum}(n-1), A[n])$$

$$\text{maxsuffix} = A[1] \quad \text{maxsubarray} = \max(0, A[1])$$

for ($i = 2$ to n) {

$$\text{maxSubarray} = \max \begin{cases} \text{maxSubarray} \\ \text{maxSuffix} + A[i] \\ A[i] \end{cases}$$

$$\text{maxSuffix} = \max \begin{cases} A[i] \\ \text{maxSuffix} + A[i] \end{cases}$$

Principle Used:

When designing recursive / inductive idea,
sometimes it is useful to solve a more general
or harder problem.

Ex

Given share prices for n days p_1, p_2, \dots, p_n .
Want to buy it on one of the days
and sell it on later day. Maximize Profit

7, 9, 3, 4, 6, 4, 2 4 ... $\mathcal{O}(n)$

Ex Celebrity Party one celebrity

Queries: ask i^{th} person, whether j^{th} person
"You know the

$\mathcal{O}(n)$ queries.

X

.

Exponentiation.

Given a, n compute a^n .

Multiplication unit cost.

$$\text{Exp}(a, n) = \text{Exp}(a, n-1) \times a$$

$$a^n = a^{n-1} \times a$$

No. of multiplication = $n-1$

Repeated squaring

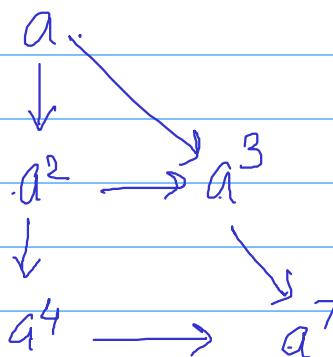
if n is even

$$a^n = (a^{n/2})^2 \quad \left(1 \text{ mult} \right) \quad T(n) = T(n/2) + 2$$

if n is odd

$$a^n = \underbrace{(a^{\frac{n-1}{2}})^2}_{n \log a} \cdot a \quad \left(2 \text{ mult} \right) \quad T(n) = 2 \log n$$

$$a^7 =$$



4 multiplications.

$$\begin{aligned}
 a^{15} &= (a^7)^2 \cdot a && 2 \text{ mult.} \\
 a^7 &= (a^3)^2 \cdot a && 2 \text{ mult.} \\
 a^3 &= a^2 \cdot a && 2 \text{ mult.}
 \end{aligned}
 \quad \left. \begin{array}{l} \\ \\ \end{array} \right\} 6 \text{ multiplications}$$

a^{15} in 5 multiplications?

$$\begin{aligned}
 a^5 &= (a^2)^2 \cdot a && \} 5 \text{ multiplications.} \\
 a^{15} &= a^5 \cdot a^5 \cdot a^5
 \end{aligned}$$

Think

Given n , what is the smallest number of multiplications needed to compute a^n ?

Can you design an efficient algorithm to find the smallest number of multiplications required?

Matrix Exponentiation.

$$F_n = F_{n-1} + F_{n-2} \quad \left. \begin{array}{l} \\ \end{array} \right\} \text{matrix multiplication}$$

Can you compute the n^{th} Fibonacci number in $O(\log n)$ operations?

0, 1, 1, 2, 3

$$F_n = \frac{\varphi^n - (-1)^n}{\sqrt{5}}$$

Hint

Design Idea 2

Divide and Conquer

- Divide the problem into multiple subproblems of size $n/2$
- Combine the solutions of the subproblems and build a solution for the original problem.

Example Merge sort-

$$T(n) = a \cdot T(n/2) + f(n)$$

\leftarrow merge
↑ no. of subproblems.

Divide and Conquer might improve the running time for example

$$O(n^2) \rightarrow O(n \log n)$$

Dominating set problem

Non-dominated



(x, y)

dominates
 (α, β)

if $x \geq \alpha$
and $y \geq \beta$

Given a set of points,

find those points

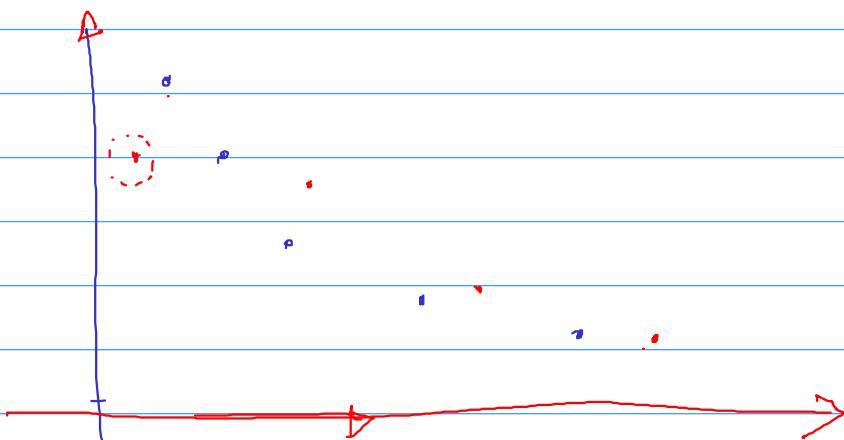
which are not dominated
by any other point.

$$T(n) = T(n-1) + O(n)$$



$O(n \log n)$

Divide and Conquer



Obs

Non-dominated points :

from left to right the y-coordinate decreases

Suppose we have two lists of non-dominated points.

Assume both the lists are sorted in increasing order of x-coordinates. And thus, they will be decreasing order of y-coordinates.

Merge Procedure:

The merge procedure will take two such lists and will output the set of non-dominated points in the union of the two lists.

We have one pointer for each list, which is initially at the start of the list.

While both lists are non-empty }

Let (a_i, b_i) and (c_j, d_j) be the current points in the two lists.

Find which of the two has the smaller x-coordinate, let it be (a_i, b_i)

If $d_j \geq b_i$ then (a_i, b_i) is dominated.

Discard (a_i, b_i) and move the pointer ahead in this list.

Otherwise (a_i, b_i) is non-dominated

Insert (a_i, b_i) in the output list and move the pointer ahead in this list.

}

If one of lists is non-empty

then insert the remaining points in the output list.

Very similar to merge step in mergesort.

$$T(n) = 2T(n/2) + O(n) = \cancel{O(n)}$$

$$\underline{T(n) = O(n \log n)}.$$

Integer Multiplication.

Bit complexity

Adding two n-bit numbers $\rightarrow O(n)$

Multiplying two n-bit numbers

$a \times b$ \rightarrow add a , b times

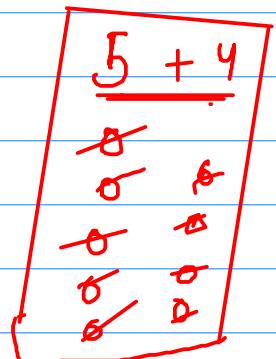
$\uparrow b$
 t_{2^n}

school method

n

$$\begin{array}{r}
 101 \\
 110 \\
 \hline
 \left. \begin{array}{r} 000 \\ 1010 \\ 10100 \end{array} \right\} \begin{array}{l} \leftarrow n \text{ bits} \\ \leftarrow n+1 \\ \leftarrow 2n+1 \text{ bits} \end{array} \\
 \hline
 11110
 \end{array}$$

$O(n^2)$



Can we do better?

Karatsuba [1960] $O(n^{1.58})$

Let's first talk about squaring.

$a \leftarrow n$ bit integer. Find a^2 .

let's try to reduce it to squaring of an $n-1$ bit integer

$$a = 2a' + \epsilon \quad \begin{array}{c} \xleftarrow{\qquad} a' \xrightarrow{\qquad} \end{array} \epsilon = 0 \text{ or } 1$$

\downarrow 2 left shifts

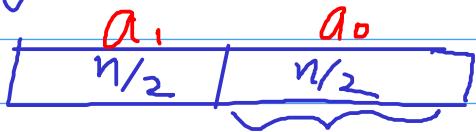
$$a^2 = (2a' + \epsilon)^2 = \underbrace{4 \cdot a'^2}_{\approx 2n+3 \text{ bits}} + \epsilon^2 + \underbrace{4a'\epsilon}_{\approx 2n+2 \text{ bits}}$$

$$T(n) = O(n^2) \Leftarrow T(n) = T(n-1) + O(n)$$

How about divide and conquer?

Reducing to squaring $n/2$ bit integers.

$$a = a_1 \cdot 2^{n/2} + a_0$$



$$a^2 = (2^{n/2} a_1 + a_0)^2$$

$$\begin{aligned} a^2 &= 2^n \cdot a_1^2 + a_0^2 + 2 \cdot 2^{n/2} \cdot a_1 \cdot a_0 ? \\ \text{left shift by } n \text{ bits} &\quad T(n/2) \quad T(n/2) \end{aligned}$$

Can we compute $a_1 \cdot a_0$ via squaring?

$$\rightarrow 2 \cdot a_1 \cdot a_0 = (a_1 + a_0)^2 - a_1^2 - a_0^2$$

$$a^2 = 2^n \cdot a_1^2 + a_0^2 + 2^{n/2} ((a_1 + a_0)^2 - a_1^2 - a_0^2)$$

$$T(n) = 3 T(n/2) + O(n)$$

$$\Rightarrow T(n) = O(n^{\log_2 3}) \approx O(n^{1.58})$$

Solving the recurrence

$$T(n) \leq c \cdot n + 3 T(n/2)$$

$$T(n) \leq c \cdot n + 3 \cdot c n/2 + 3^2 \cdot T(n/4)$$

$$\leq c \cdot n + 3 \cdot c n/2 + 3^2 \cdot c n/2^2 + 3^3 \cdot T(n/8)$$

$$\leq c n + \underbrace{\frac{3}{2} c n + \frac{3^2}{2^2} c n + \dots}_{\frac{3^{\log_2 n}}{2^{\log_2 n - 1}} c n} + 3^{\log_2 n} T(1)$$

$$\begin{aligned} a_1 + a_0 &= 2 \cdot b + e \\ (2b + e)^2 &= 4b^2 + 4be + e^2 \end{aligned}$$

$$T(n) \leq c n \frac{(3/2)^{\log n} - 1}{3/2 - 1} + 3^{\log n}$$

$$\begin{aligned}
 &= 2(cn \cdot n^{\log_{3/2}}) + n^{\log 3} \\
 &= 2c n^{1+\log_{3/2}} + n^{\log 3} \\
 &= O(n^{\log_2 3}) = O(n^{1.585})
 \end{aligned}$$

$\log_{3/2} = \log 3 - \log 2$

What about multiplication?

$$a \cdot b = \frac{(a+b)^2 - a^2 - b^2}{2}$$

$$a \cdot b = \frac{(a+b)^2 - (a-b)^2}{4}$$

Multiplication directly (without going via squaring)

$a \times b$?

$$a = a_1 \cdot 2^{n/2} + a_0$$

$$b = b_1 \cdot 2^{n/2} + b_0$$

$$\begin{aligned}
 ab &= a_1 \cdot b_1 \cdot 2^n + a_0 b_1 \cdot 2^{n/2} + a_1 b_0 \cdot 2^{n/2} + a_0 b_0 \\
 &= a_1 \cdot b_1 \cdot 2^n + (a_0 b_1 + a_1 b_0) \cdot 2^{n/2} + a_0 b_0
 \end{aligned}$$

HW: Can these three terms $a_1 b_1$, $a_0 b_1 + a_1 b_0$, $a_0 b_0$ be computed somehow with three multiplications?

Hint: first compute $(a_1 + a_0)(b_1 + b_0)$

$n/2$ bits

two more multiplications allowed.

Can we instead divide into three parts?

$$\text{Squaring } a = \underline{a_2} \cdot 2^{2n/3} + \underline{a_1} \cdot 2^{n/3} + \underline{a_0}$$

$$a^2 = \underbrace{a_2^2 \cdot 2^{4n/3}} + \underbrace{2a_2 a_1 \cdot 2^n} + \underbrace{(a_1^2 + 2a_0 a_2) 2^{2n/3}} \\ + \underbrace{2a_1 a_0 \cdot 2^{n/3}} + \underbrace{a_0^2 \cdot 2^0}$$

$$T(n) = \alpha T(n/3) + O(n)$$

$$T(n) = O(n^{\log_3 \alpha})$$

current best $n^{1.585}$

$$\log_3 6 = \dots \approx 1.63$$

$$\log_3 5 = \underline{1.46} \Rightarrow O(n^{1.46})$$

Can we compute the desired terms with $5/6$ squarings?
+ $O(n)$ operations.

easy to do with 6 squarings.

$$a_0^2, a_1^2, a_2^2, (a_0 + a_1)^2, (a_0 + a_2)^2, (a_1 + a_2)^2$$

Idea:

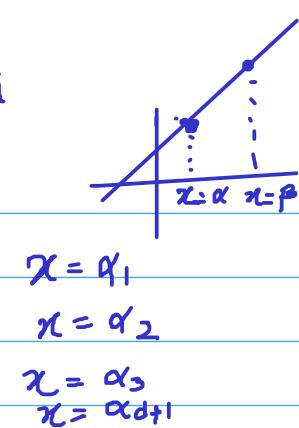
To improve, we need to see it as a squaring of a polynomial.

$$A(x) = a_2 x^2 + a_1 x + a_0$$

$$(A(x))^2 = \underbrace{a_2^2}_{\dots} x^4 + \underbrace{2a_1 a_2}_{\dots} x^3 + \underbrace{(a_1^2 + 2a_0 a_2)}_{\dots} x^2 \\ + \underbrace{2a_0 a_1}_{\dots} x + \underbrace{a_0^2}_{\dots}$$

Polynomial Representations: for degree d

- Coefficients $d+1$
- Roots d
- Evaluations $d+1$



How easy / difficult it is to square a polynomial in evaluation representation?

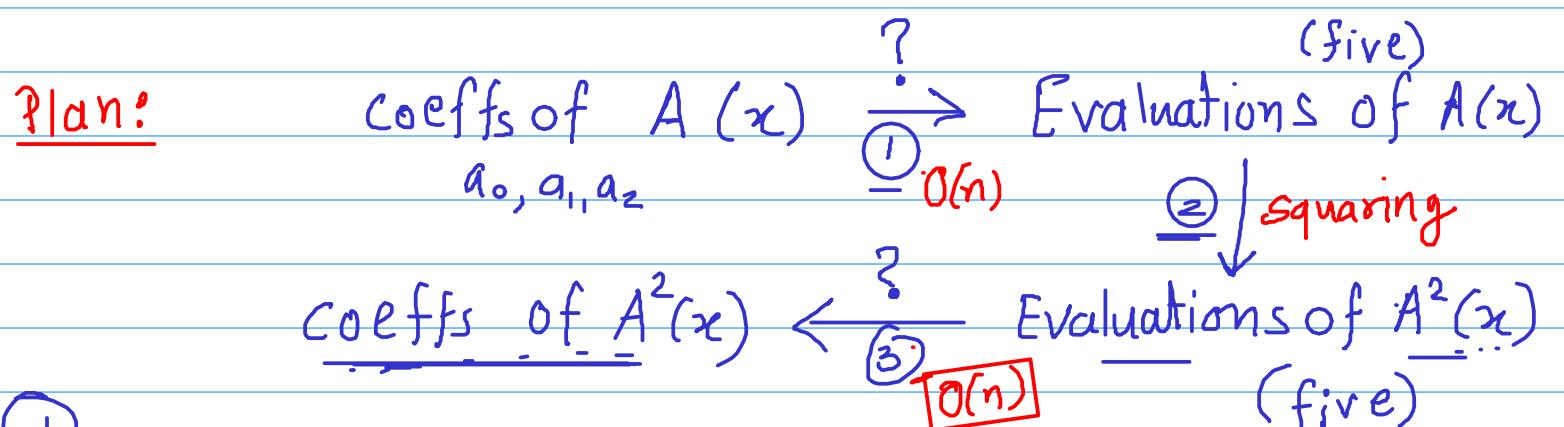
Given evaluations of $A(x)$, computing evaluations of $A^2(x)$?

$$A(\alpha)^2 = (A(\alpha))^2$$

- Each evaluation of $A^2(x)$ needs one squaring
- How many evaluations of $A^2(x)$ needed?

Five evaluations → five squarings.

But we are really interested in coeff of $A^2(x)$.



$$A(x) = a_2 x^2 + a_1 x + a_0$$

$$A(x=\alpha) = a_2 \alpha^2 + a_1 \alpha + a_0$$

$$x = 0, 1, -1, 2, -2$$

$a_0, a_1, a_2 \rightarrow \frac{n}{3}$ bits

$O(n)$

$$A(0) = a_0 \quad A(1) = a_0 + a_1 + a_2 \quad A(-1) = a_0 - a_1 + a_2$$

$$\underbrace{n/3 + 4 \text{ bits}}_{\leftarrow} \quad A(2) = a_0 + 2a_1 + 4a_2$$

② $A(0), A(1), A(-1), A(2), A(-2)$

↓ square

$A^2(0), A^2(1), A^2(-1), A(2), A(-2)$

$5T(n/3) + O(n)$

③ Define $S(x) := A^2(x)$

How are coeffs and Evaluations of $S(x)$ are related?

$$S(0) = a_0^2$$

$$S(1) = a_0^2 + 2a_0a_1 + a_1^2 + 2a_0a_2 + 2a_1a_2 + a_2^2$$

$$S(2) = a_0^2 + 2 \cdot 2a_0a_1 + 4(a_1^2 + 2a_0a_2) + 8 \cdot 2a_1a_2 + 16a_2^2$$

$$\begin{bmatrix} S(0) \\ S(1) \\ S(2) \\ S(-1) \\ S(-2) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 & 16 \\ 1 & -1 & 1 & -1 & 1 \\ 1 & -2 & 4 & -8 & 16 \end{bmatrix} \begin{bmatrix} a_0^2 \\ 2a_0a_1 \\ a_1^2 + 2a_0a_2 \\ 2a_1a_2 \\ a_2^2 \end{bmatrix}$$

$$\text{eval-vector} = \begin{bmatrix} M \end{bmatrix}_{5 \times 5} \quad \text{coeff-vector}$$

$$M^{-1}, \text{eval-vector} = \underbrace{\text{coeff-vector}}$$

Can we do this in $O(n)$?

We can pre-compute M^{-1} and store it.

Once computed, M^{-1} can be used to square any integer.

M^{-1} . eval-vector
↑
 5×5

$$(n/3 + 4) \times 2 = O(n)$$

All entries of M^{-1} are constants.

Hw Multiplication/division of an n bit integer with a constant can be done in $O(n)$ bit operations.

need 25 multiplications and 20 additions.
Overall $O(n)$ time.

$$T(n) = 5T(n/3) + O(n)$$

$$T(n) \approx O(n^{1.46}) \quad \text{Toom-Cook}$$

Integer Multiplication. History

1960 Karatsuba $O(n^{1.585})$

Toom Cook $O(n^{1.46})$

can be further generalized by dividing the integers into more parts

and get better and better time complexity.

But, the time complexity will remain something like $O(n^{1+\varepsilon})$ for $\varepsilon > 0$.

{ 1971 Schönhage Strassen $O(n \log n \log \log n)$

{ 2005 Fürer $O(n \log n 2^{\log^* n})$

{ 2019 Harvey, van der Hoeven $O(n \log n)$

→ Ideas: polynomial evaluation (also known as discrete fourier transform)
divide and conquer and other ideas.

Last class

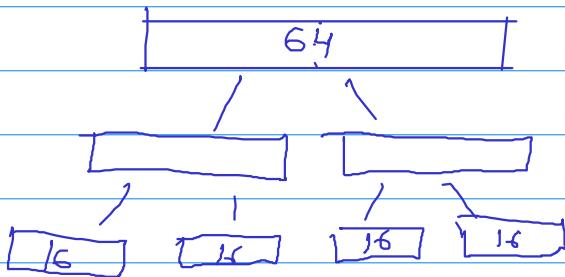
divide and conquer for "integer multiplication"

Karatsuba $O(n^{\log 3}) \approx O(n^{1.58})$

In practice:

may be slower than the school method say for 64 bit int

combination of Karatsuba and school method may be better.



Similar ideas can be applied to

- Matrix multiplication
- P_0

• Matrix multiplication

A and B are $n \times n$ matrices
find $A \cdot B$

$$\begin{bmatrix} & & \\ & & \\ \end{bmatrix} \begin{bmatrix} & & \\ & & \\ \end{bmatrix}$$

Naive algorithm $\mathcal{O}(n^2 \cdot n) = \mathcal{O}(n^3)$

Divide and conquer

Verify

$$\left[\begin{array}{c|c} A_1 & A_2 \\ \hline A_3 & A_4 \end{array} \right] \left[\begin{array}{c|c} B_1 & B_2 \\ \hline B_3 & B_4 \end{array} \right] = \left[\begin{array}{cc} A_1B_1 + A_2B_3 & A_1B_2 + A_2B_4 \\ A_3B_1 + A_4B_3 & A_3B_2 + A_4B_4 \end{array} \right]$$

Assume a subroutine for $n/2 \times n/2$ matrices

no. of multiplications = 8

no. of additions = 4

Recurrence $T(n) = 7T(n/2) + O(n^2)$ Strassen

$$T(n) = \boxed{O(n^3)} \quad O(n^{\log 7}) \approx O(n^{2.8})$$

current $O(n^{2.37...})$ want $O(n^2 \log^c n)$

Puzzle (Secret sharing)

A resource shared ownership - n people

Should be accessible only if - at least k of them together.

Polynomial Multiplication

$$a(x) = a_0 + a_1x + a_2x^2 + \dots + a_dx^d$$

$$b(x) = b_0 + b_1x + b_2x^2 + \dots + b_dx^d$$

$$\begin{aligned} a \cdot b &= a_0b_0 + (a_1b_0 + a_0b_1)x + (a_2b_0 + a_1b_1 + a_0b_2)x^2 \\ &\quad + \dots + a_db_d x^{2d} \\ &= \underbrace{\sum_{j=0}^{2d} x^j}_{\text{Naive algorithm}} \left(\sum_i a_i b_{j-i} \right) \end{aligned}$$

Naive algorithm $\mathcal{O}(d^2)$

(unit cost arithmetic operations)

Karatsuba? Verify.

Convolution (discrete)

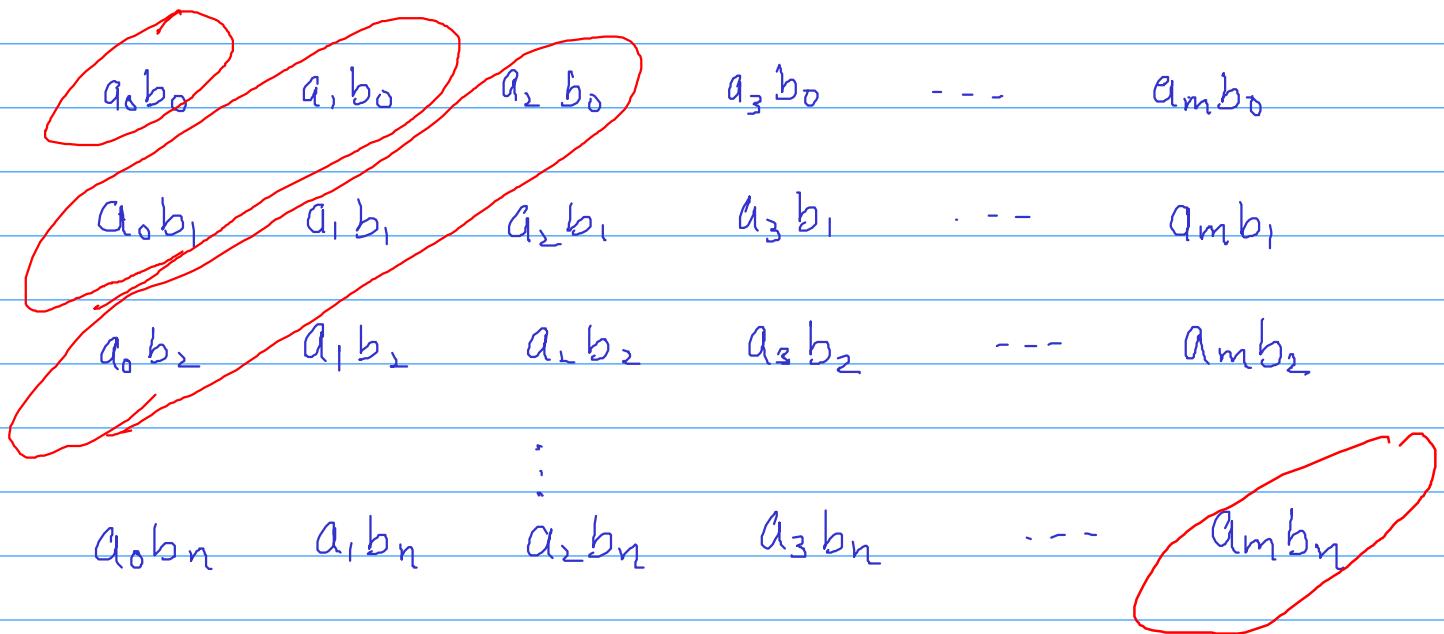
$$a = (a_0, a_1, a_2, \dots, a_m) \in \mathbb{R}^{m+1}$$

$$b = (b_0, b_1, b_2, \dots, b_n) \in \mathbb{R}^{n+1}$$

$$a * b = \underbrace{\quad \quad \quad}_{n+m+1}$$

$\mathcal{O}(mn)$

$$(a_0 b_0, a_1 b_0 + a_0 b_1, a_2 b_0 + a_1 b_1 + a_0 b_2, \dots, (\sum_i a_i b_{j-i}), \dots, a_m b_n)$$



Sliding window

$a_0 \quad a_1 \quad a_2 \quad a_3 \quad a_4 \quad a_5 \quad a_6$

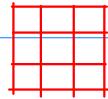
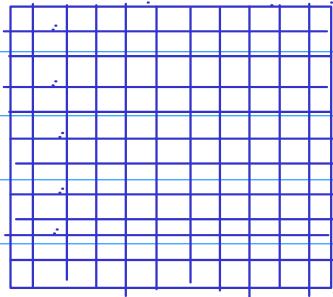
$b_2 \quad b_1 \quad b_0$

Applications

- Signal processing
 - Smoothening of noisy data
e.g. 7-day averages of covid cases

- Image processing

2d convolution.



polynomial in 2 variables

- Probability

A dice	outcome	1	2	3	4	5	6
prob	0.2	0.1	0.05	0.3	0.15	0.2	

Roll two dice and take sum of the two

compute probabilities of all outcomes.

2, 3, 4, ---, 12

0.04 0.06

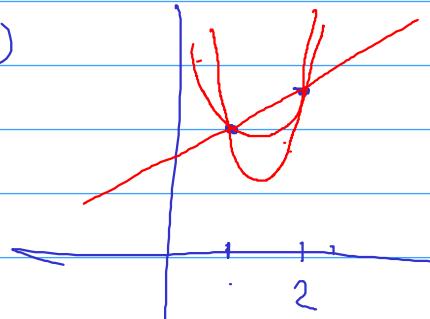
Convolution?

Polynomial multiplication / convolution

- Faster algorithms.

Representation of polynomials

$$a(x)$$



- Coefficients

- Roots

- Evaluations. x_1, x_2, \dots, x_{d+1}

$$a(x_1), a(x_2), \dots, a(x_{d+1})$$

Claim Given $d+1$ evaluations, there is a
HW unique degree d polynomial satisfying those.

Computation in evaluation representation

	Coeff	Roots	Evaluations.
Multiplication	$O(d^2)$	$O(d)$	$O(d)$
addition	$O(d)$?	$O(d)$

How efficiently can we compute evaluations from coefficients?

$d+1$ coefficients \longrightarrow $d+1$ evaluations $O(d^2)$

$$a_0 + a_1 x_1 + a_2 x_1^2 + a_3 x_1^3 + \dots + a_d x_1^d \quad O(d)$$

$$a_0 + a_1 x_2 + a_2 x_2^2 + a_3 x_2^3 + \dots + a_d x_2^d \quad \underline{x_0 = 0}$$

May be we can choose evaluation points cleverly

and find correlations among evaluations?

$$a(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_{d-1} x^{d-1}$$

$$\boxed{a(1)} = a_0 + a_1 + a_2 + \dots + a_{d-1} \quad \left\{ d-1 \text{ additions } (d \text{ is even}) \right.$$

$$\boxed{a(-1)} = a_0 - a_1 + a_2 - a_3 + a_{d-1} \quad \left\{ d-1 \text{ additions} \right.$$

$$a_0 + a_2 + a_4 + \dots + a_{d-2} \quad \left\{ \frac{d}{2}-1 \text{ additions} \right.$$

$$a_1 + a_3 + \dots + a_{d-1} \quad \left\{ \frac{d}{2}-1 \text{ additions} \right.$$

$2d-2$ additions $\rightarrow d$ additions.

work reduced by half.

$$a(x) = a_0 + a_1 x + a_2 x^2 + a_3 x^3 + \dots$$

$$a(-x) = a_0 - a_1 x + a_2 x^2 - a_3 x^3 + \dots$$

$$a_{\text{even}}(x) = a_0 + a_2 x + a_4 x^2 + \dots \quad \text{degree } \frac{d-1}{2}$$

$$a_{\text{odd}}(x) = a_1 + a_3 x + a_5 x^2 + \dots \quad \text{degree } \frac{d-1}{2}$$

$$a(x) = a_{\text{even}}(x^2) + x \cdot a_{\text{odd}}(x^2)$$

$$a_0 + a_2 x^2 + a_4 x^4 + x(a_1 + a_3 x^2 + a_5 x^4)$$

$$a(-x) = a_{\text{even}}(x^2) - x \cdot a_{\text{odd}}(x^2)$$

One Degree $d-1$
2 evaluations

$\xrightarrow{\hspace{1cm}}$ 2 degree $\frac{d-1}{2}$ polynomials
1 evaluation.

Two points - a polynomial of deg d

$$a_{\text{even}} = a_0 + a_2 x + a_4 x^2 + \dots + a_{d-1} x^{\frac{d-1}{2}}$$

$$a_{\text{odd}} = a_1 + a_3 x + a_5 x^2 + \dots + a_d x^{\frac{d-1}{2}}$$

$$a(x) = a_{\text{even}}(x^2) + x a_{\text{odd}}(x^2)$$

$$a(x), a(-x)$$

$$a(x) = a_{\text{even}}(x^2) + x a_{\text{odd}}(x^2)$$

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

deg d, two evaluations, one poly



$$2 \leq d$$

deg $\frac{d-1}{2}$, one evaluation, two polys

$$2 \cdot \leq \frac{d-1}{2}$$

$$\alpha_1, \alpha_2, \dots, \alpha_{2d+1}$$

$$\alpha_1, -\alpha_1, \alpha_2, -\alpha_2, \dots$$

deg d poly at K points

two polynomials, degree $\frac{d-1}{2}$, $\frac{K}{2}$ points

Old set $\alpha_1, -\alpha_1, \alpha_2, -\alpha_2, \alpha_3, -\alpha_3, \dots$

New set of points $\alpha_1^2, \alpha_2^2, \alpha_3^2, \alpha_4^2, \dots$

Need $\alpha_1^2 = -\alpha_2^2$ for applying same trick again

4 points $i, -i, -i, i$ $i = \sqrt{-1}$

4 points $i, -i, -1, 1$

2 points $-1, 1$

1 point 1

$i \leftarrow 4^{\text{th}} \text{ root of unity}$

$8^{\text{th}} \text{ root of unity}$

Apply this K times

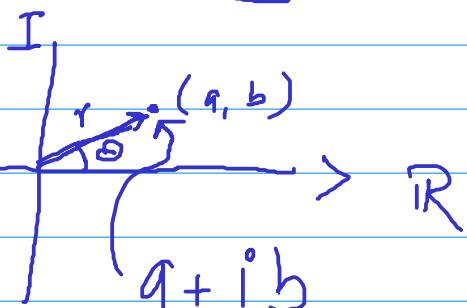
2^k th root of unity

$a+ib$

$$= r e^{i\theta}$$

$$r = \sqrt{a^2 + b^2}$$

$$\tan \theta = \frac{b}{a}$$



ω is k^{th} root of unity

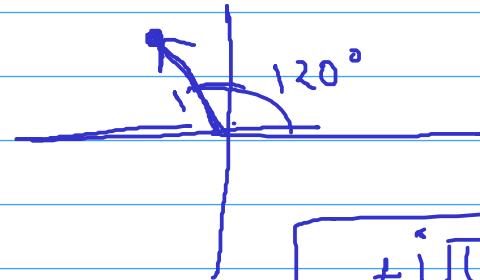
$$\omega^k = 1$$

↓

$$k=3$$

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

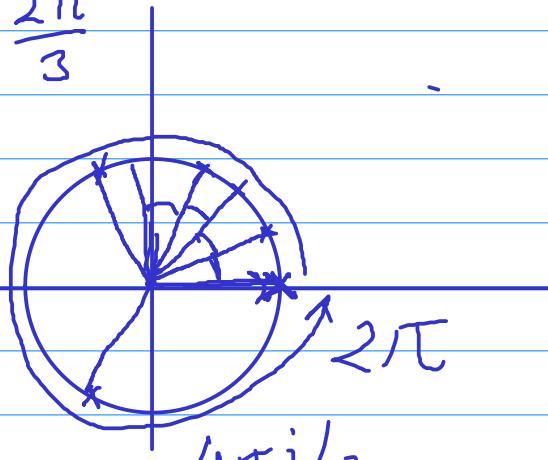
$$(e^{2\pi i/3})^3 = 1$$



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

$$r e^{i\theta} \quad r = 1 \\ \theta = \frac{2\pi}{3}$$

$$(e^{4\pi i/3})^3 = 1$$



$$\cdot e^0, e^{2\pi i/3}, e^{4\pi i/3}$$

$$k^{\text{th}} \text{ roots} \quad e^0, e^{2\pi i/k}, e^{4\pi i/k}, \dots, e^{\frac{k-1}{k} \cdot 2\pi i}$$

Properties of k^{th} roots of unity

$$k \rightarrow \text{even} \quad \textcircled{1} \quad \omega = e^{2\pi i/k}$$

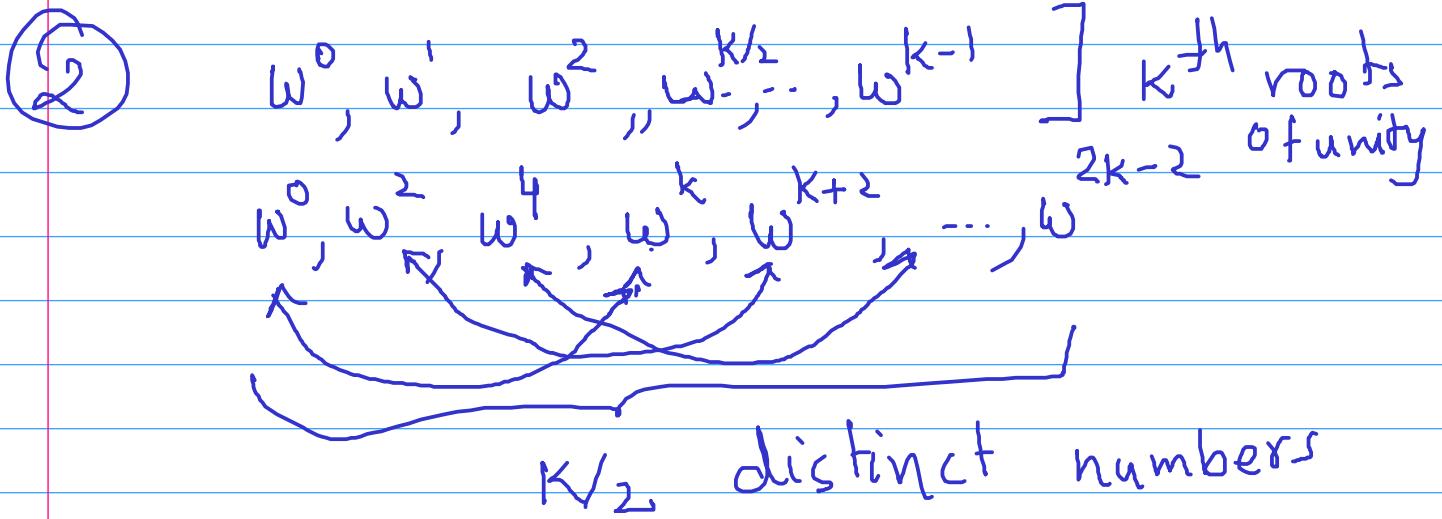
$$\Rightarrow \boxed{\omega^{k/2} = e^{i\pi} = -1}$$

$$\omega^0, \omega^1, \omega^2, \omega^3, \dots, \omega^{k-1}$$

$$-\omega^j = e^{i\pi} \omega^j = \omega^{k/2} \omega^j = \omega^{j+\frac{k}{2}}$$

$$\omega^{k/2} = -1$$

$$-\omega^j = \omega^{j+k/2}$$



$\frac{k}{2}$ th roots .

$$e^0, e^{2\pi i/k}, e^{4\pi i/k}, e^{6\pi i/k}, \dots$$

—————*

(3) $w^0 + w + w^2 + \dots + w^{k-1} = 0$

$$w^j = -w^{j+k/2}$$

|



Polynomial Evaluation

(discrete Fourier Transform)

$a(x) \deg d-1$

d is a power of 2

Evaluate $a(x)$ over d points

$O(d^2)$ arithmetic operations

the d^{th} roots of unity.

$O(d \log d)$ arithmetic operations.

$$w \leftarrow e^{2\pi i / d}$$

$$w^0, w^1, w^2, w^3, \dots, w^{d-1}$$

$$a(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_{d-1} x^{d-1}$$

$$a_{\text{odd}} = a_1 + a_3 x + a_5 x^2 + \dots + a_{d-1} x^{\frac{d}{2}-1}$$

$$a_{\text{even}} = a_0 + a_2 x + a_4 x^2 + \dots + a_{d-2} x^{\frac{d}{2}-1}$$

$$\rightarrow a(x) = \underline{a_{\text{even}}(x^2)} + x \cdot \underline{a_{\text{odd}}(x^2)}$$

Evaluate $a(x)$ over $w^0, w^1, w^2, w^3, \dots, w^{d-1}$



Evaluate $a_{\text{even}}(x)$ over $w^0, w^2, w^4, \dots, w^{d-2}$

$$a_{\text{odd}}(x) = \gamma^0, \gamma^1, \gamma^2, \dots, \gamma^{d/2-1}$$

One polynomial of degree $d-1$
evaluate at d^{th} roots of unity
(d points)



two polynomials of degree $\frac{d}{2}-1$
evaluate at $\frac{d}{2}$ th roots of unity
($d/2$ points)

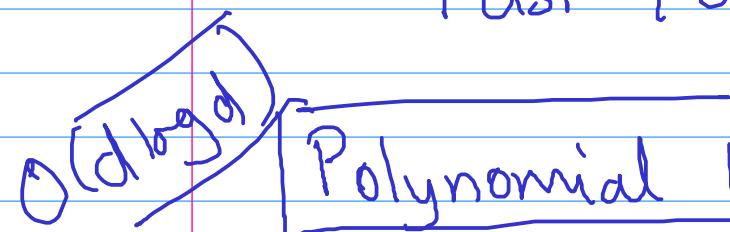
$T(d) = \text{No. of arithmetic operations}$
in evaluating a degree $d-1$
polynomial at d th roots of 1.

$$T(d) = 2 T\left(\frac{d}{2}\right) + O(d)$$

[d additions
 d multiplication]

$$T(d) = O(d \log d)$$

Fast Fourier Transform (FFT)



Polynomial Multiplication / Convolution

- $O(d \log d)$ ① Evaluate
- $O(d)$ ② Multiply these Evaluations
- $O(d \log d)$ ③ compute coeffs from Evaluations

$$a(x) = a_0 + a_1 x + \dots + a_{d-1} x^{d-1}$$

Evaluate over d^{th} roots of unity

$$\begin{bmatrix} a(w^0) \\ a(w^1) \\ a(w^2) \\ \vdots \\ a(w^{d-1}) \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 & \dots & 1 \\ 1 & w & w^2 & w^3 & \dots & w^{d-1} \\ 1 & w^2 & w^4 & \dots & w^{2d-2} \\ 1 & w^{d-1} & \dots & w^{(d-1)^2} & \dots & w^{(d-1)^2} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_{d-1} \end{bmatrix}$$

$\overbrace{\quad\quad\quad\quad\quad\quad\quad\quad\quad}^M$

$M^{-1} \cdot \text{Eval} = \text{coeff}$ $O(d \log d)$

Claim $M' = \begin{bmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & w^{-1} & w^{-2} & & w^{-(d-1)} \\ 1 & w^{-2} & w^{-4} & \dots & \\ 1 & w^{-(d-1)} & & \dots & w^{-(d-1)^2} \end{bmatrix}$

HW $MM' = dI$

$M^{-1} \cdot \text{eval.}] O(d \log d)$ operations