

CS3230

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Computer Science

T05 - Week 6

# D&C, Sorting, and Average-Case Analysis

CS3230 – Design and Analysis of Algorithms

#### **Loop Invariant**

- ➤ GeeksforGeeks Loop Invariants
- > StackExchange Tips for Constructing Basic Loop Invariants

#### Induction

- https://leetcode.com/problem-list/recursion/
- ▶ Brilliant Writing a proof by Induction
- Khan Academy Verifying an algorithm (also invariant)

#### D&C

In general, this requires training your thinking processes (which is v hard):

- https://leetcode.com/problem-list/divide-and-conquer/
- > T04 Q5: Split by the largest direction (row or column).

# **T04 Q5:** How to achieve $\Theta(n)$

#### Algorithm

- Split the matrix always in the larger (width or height)
- 2 Same algorithm as before.

### Proof

Assuming m > n, and also vice versa for n > m:

$$T(m,n) = T\left(\frac{m}{2}, n\right) + \Theta(n)$$
$$= \left[T\left(\frac{m}{2}, \frac{n}{2}\right) + \Theta\left(\frac{m}{2}\right)\right] + \Theta(n)$$

Since the recurrence reduces by 1/2 in 2 iterations, we obtain  $T(n) = T(n/2) + \Theta(n)$ . Since  $a=1,\ b=2,\ d=\log_2 1=0,$  and  $f(n)=n\in\Omega(n^{d+\epsilon})$  for any  $\epsilon>0$ . Furthermore, the regularity condition is satisfied, as:  $1\cdot f(n/2)=\frac{n}{2}\leq cf(n)$  for  $c=\frac{1}{2}<1$ . Thus, by Case 3 of the Master Theorem:  $T(n)=\Theta(n)$ .

<sup>&</sup>lt;sup>1</sup>You may work it out exactly, but... Lazy.

#### A decision tree consists of:

- > Vertices (Internal): A comparison
- **> Branches:** Outcome of the comparison
- > Leaves: Output/decision for the input

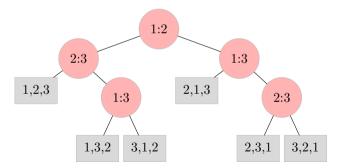


Figure 1: Worst case runtime is the height of the decision tree.

# Polynomial Multiplication (Degree n)

Given two polynomials:

$$A(x) = a_n x^n + \dots + a_2 x^2 + a_1 x + a_0$$

$$B(x)=b_nx^n+\cdots+b_2x^2+b_1x+b_0$$

Their product:

$$C(x) = A(x) \times B(x) = c_{2n}x^{2n} + \dots + c_2x^2 + c_1x + c_0$$

where all coefficients  $a_i, b_i, c_i$  are integers.

Brute Force Approach:  $O(n^2)$  Complexity

Brute Force Approach. 
$$O(n)$$
 Complexity

 $\forall_i \in [2n..0], \quad c_i = \sum_{i=0}^n a_j \cdot b_{i-j}, \quad \text{where } 0 \leq i-j \leq n.$ 

$$V_i \in [2n..0], \quad C_i = \sum_{j=0}^n a_j \cdot b_{i-j}, \quad \text{where } 0 \leq i-j \leq n$$

Assuming integer addition and multiplication take O(1) time, this approach runs in  $O(n^2)$ .

## Brute Force Approach (code)

```
def poly_mult_bruteforce(A, B):
    # A: Coeff [a0, a1, ..., a_n] for A(x) = a0 + a1*x + ... + a n*x^n.
    # B: Coeff [b0, b1, ..., b_n] for B(x) = b0 + b1*x + ... + b n*x^n.
   n = len(A) - 1 # Degree of the polynomial A or B
    result = [0] * (2 * n + 1) # Result in (2n + 1) coefficients
    # Compute each coefficient c i for the product polynomial C(x)
    for i in range(2 * n + 1):
        for j in range(max(0, i - n), min(i, n) + 1):
            result[i] += A[j] * B[i - j]
    return result
```

Let x = 10 to visualize this as base-10 multiplication with n = 2.

### **Given Polynomials**

$$A(10) = 352 = 3 \cdot 10^2 + 5 \cdot 10 + 2, \qquad \text{i.e. } a_2 = 3, \quad a_1 = 5, \quad a_0 = 2, \\ B(10) = 221 = 2 \cdot 10^2 + 2 \cdot 10 + 1, \qquad \text{i.e. } b_2 = 2, \quad b_1 = 2, \quad b_0 = 1.$$

Compute the coefficients of  $C(10)=A(10)\times B(10)=77,792$  using the  $O(n^2)$  algorithm.

# U

# **Answer** Using the $O(n^2)$ algorithm, we compute:

$$c_4 = a_2 \cdot b_{4-2} = a_2 \cdot b_2 = 3 \cdot 2 = 6.$$

$$c_3 = a_1 \cdot b_{3-1} + a_2 \cdot b_{3-2} = a_1 \cdot b_2 + a_2 \cdot b_1$$

$$= 5 \cdot 2 + 3 \cdot 2 = 10 + 6 = 16$$

$$c_2 = a_0 \cdot b_{2-0} + a_1 \cdot b_{2-1} + a_2 \cdot b_{2-2}$$

$$= a_0 \cdot b_2 + a_1 \cdot b_1 + a_2 \cdot b_0$$

$$= 2 \cdot 2 + 5 \cdot 2 + 3 \cdot 1 = 4 + 10 + 3 = 17.$$

$$c_1 = a_0 \cdot b_{1-0} + a_1 \cdot b_{1-1} = a_0 \cdot b_1 + a_1 \cdot b_0$$

$$= 2 \cdot 2 + 5 \cdot 1 = 4 + 5 = 9$$

$$c_0 = a_0 \cdot b_{0-0} = a_0 \cdot b_0 = 2 \cdot 1 = 2.$$

Hence,  $C(10) = 6 \cdot 10^4 + 16 \cdot 10^3 + 17 \cdot 10^2 + 9 \cdot 10 + 2 = 77792$ .

Rewrite the polynomials:

$$A(x) = x^{\frac{n}{2}} \cdot A_1(x) + A_2(x), \quad B(x) = x^{\frac{n}{2}} \cdot B_1(x) + B_2(x)$$

where  $A_1(x), A_2(x), B_1(x), B_2(x)$  are polynomials of degree at most  $\frac{n}{2}$ .

Compute four smaller polynomial multiplications:

$$A_1(x) \times B_1(x), \quad A_1(x) \times B_2(x), \quad A_2(x) \times B_1(x), \quad A_2(x) \times B_2(x)$$

Compute the final polynomial:

$$C(x) = x^n \cdot [A_1(x) \times B_1(x)] + x^{\frac{n}{2}} \cdot [A_1(x) \times B_2(x) + A_2(x) \times B_1(x)] + A_2(x) \times B_2(x)$$

Use this algorithm to multiply two polynomials of degree n=2.

Given:  $A(10) = 352 = 10 \cdot (3 \cdot 10 + 5) + 2$ ,  $B(10) = 221 = 10 \cdot (2 \cdot 10 + 2) + 1$ 

## **Computing Partial Products**

$$\begin{split} A_1(10) \times B_1(10) &= (3 \cdot 10 + 5) \times (2 \cdot 10 + 2) \\ &= 6 \cdot 10^2 + 16 \cdot 10 + 10 \\ &= 600 + 160 + 10 = 770. \end{split}$$

$$= 3 \cdot 10 + 5 = 35.$$

$$A_2(10) \times B_1(10) = 2 \times (2 \cdot 10 + 2)$$

 $A_1(10) \times B_2(10) = (3 \cdot 10 + 5) \times 1$ 

$$= 4 \cdot 10 + 4 = 44.$$

$$A_2(10) \times B_2(10) = 2 \times 1 = 2.$$

## Compute the final polynomial

$$\begin{split} C(10) &= 10^2 \cdot (A_1(10) \times B_1(10)) \\ &+ 10 \cdot (A_1(10) \times B_2(10) + A_2(10) \times B_1(10)) \\ &+ A_2(10) \times B_2(10) \\ \\ &= 10^2 \cdot (6 \cdot 10^2 + 16 \cdot 10 + 10) \\ &+ 10 \cdot (3 \cdot 10 + 5 + 4 \cdot 10 + 4) + 2 \\ \\ &= 6 \cdot 10^4 + 16 \cdot 10^3 + 10 \cdot 10^2 + 7 \cdot 10^2 + 9 \cdot 10 + 2 \\ \\ &= 6 \cdot 10^4 + 16 \cdot 10^3 + 17 \cdot 10^2 + 9 \cdot 10 + 2 \\ \\ &= 60 \ 000 + 16 \ 000 + 1700 + 90 + 2 \\ \\ &= 77 \ 992 \end{split}$$

What is the time complexity of that recursive D&C algorithm?

#### Answer

$$T(n) = 4 \cdot T(n/2) + O(n).$$

- **>** There are **4 multiplications** of polynomials of degree  $\frac{n}{2}$ .
- **Combining results** requires O(n) work.

Since a=4, b=2, and  $d=\log_2 4=2$ , and  $f(n)\in O(n^{d-\epsilon})$  for some  $\epsilon>0$ , by Case 1 of the Master Theorem, we get:

$$T(n) \in \Theta(n^d) = \Theta(n^2).$$

Thus, this is no better than naive polynomial multiplication.

#### Karatsuba Algorithm

Compute two smaller polynomial multiplications:

$$A_1(x) \times B_1(x), \quad A_2(x) \times B_2(x).$$

Compute one multiplication with two additions:

$$[A_1(x) + A_2(x)] \times [B_1(x) + B_2(x)].$$

Apply the identity, two subtractions

$$\begin{split} A_1(x) \times B_2(x) + A_2(x) \times B_1(x) &= [A_1(x) + A_2(x)] \times [B_1(x) + B_2(x)] \\ &- A_1(x) \times B_1(x) - A_2(x) \times B_2(x). \end{split}$$

4 Compute C(x)

What is the time complexity of Karatsuba's algorithm?

$$T(n) = 3 \cdot T(n/2) + O(n).$$

- Now, there are **only 3 multiplications** of polynomials of degree  $\frac{n}{2}$ .
- ightharpoonup Additional work still takes O(n).

Since a=3, b=2, and  $d=\log_2 3=1.58\ldots$ , and  $f(n)=O(n)=O(n^{d-\epsilon})$  for some  $\epsilon>0$ , by Case 1 of the Master Theorem, we get:

$$T(n) \in \Theta(n^d) = \Theta(n^{\log_2 3}) = \Theta(n^{1.58...}).$$

$$T(n) = 3 \cdot T(n/2) + O(n).$$

- Now, there are **only 3 multiplications** of polynomials of degree  $\frac{n}{2}$ .
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$$T(n) \in \Theta(n^d) = \Theta(n^{\log_2 3}) = \Theta(n^{1.58\dots}).$$

#### Remarks

- **Practical Application**: This method is in CPython for multiplying large integers.
- **> Beyond Karatsuba**: Can be improved further to  $O(n \log n)$  using more advanced techniques.

You are given **243** balls, where one is heavier while the rest have the same weight. You (your friend) have a balance scale and must determine the heavier ball while minimizing the worst-case number of weighings.

- ightharpoonup The balance scale provides only **comparison results** (<, =, or >).
- Each weighing has a cost.

With these information,

- January What is the minimum number of weighings needed?
- **b.** What is the lower bound for **any** algorithm solving this problem?

### Minimum Number of Weighings

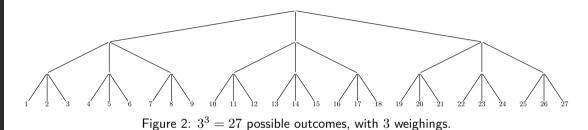
- $\blacksquare$  Divide the balls into three equal groups: A, B, and C.
- $\blacksquare$  Weigh group A against group B.
  - $\longrightarrow$  If A=B, the heavier ball is in group C.
  - $\gg$  If A>B, the heavier ball is in group A.
    - ightharpoonup If A < B, the heavier ball is in group B.

Each weighing reduces the balls by 1/3, which goes:

$$243 \xrightarrow{\rm 1st} 81 \xrightarrow{\rm 2nd} 27 \xrightarrow{\rm 3rd} 9 \xrightarrow{\rm 4th} 3 \xrightarrow{\rm 5th} 1.$$

After 5 weighings, the last ball must be the heavier one.

### **Optimal Weighings**



▶ Each weighing divides the balls into at most<sup>2</sup> 3 groups.

- $\triangleright$  A full **ternary tree** of height h has at most:  $3^h$  leaves.
- Since there are 243 possible outcomes, a tree of height 4 is insufficient.
- > Thus, at least 5 weighings are necessary.

<sup>&</sup>lt;sup>2</sup>weighings may not divide the balls into three

# Question 6 [G]

You are given an array A[1..n] that is sorted in **non-increasing order**. Your task is to find the largest index i such that  $A[i] \ge i$ . Design an efficient algorithm to solve this problem.

To guide your approach, consider the following properties of the sorted array:

- ▶ If  $A[j] \ge j$ , then it must hold that (left)  $A[j-1] \ge j-1$ , unless j=1.
- ightharpoonup If A[j] < j, then it must follow that (right) A[j+1] < j+1, unless j=n.

For ease of notation, assume that the array is extended such that A[0] > 0 and A[n+1] < n+1. Thus, there is a unique i such that  $A[i] \ge i$  but A[i+1] < i+1.

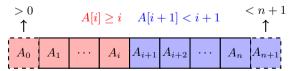


Figure 3: Key observation: Red implies left is red, Blue implies right is blue.

#### Method 1: Linear Search

- **>** Perform a linear search to find the largest i such that  $A[i] \geq i$ .
- ightharpoonup This takes O(n) time.

#### Method 2: Binary Search

- > Use binary search, leveraging the given properties of the array.
- **This reduces the time complexity to**  $O(\log n)$ .

#### Method 3: Exponential Search + Binary Search

- I Find the smallest k where  $A[2^k] < 2^k$  by testing k = 0, 1, 2, ...
  - $\gg$  If k=0, we are already done.
- ightharpoonup Otherwise, this ensures that  $A[2^k] < 2^k$  while  $A[2^{k-1}] > 2^{k-1}$ .
- **2** Apply **binary search** in the range  $[2^{k-1}, 2^k]$  to find the largest i such that  $A[i] \geq i$ .
- **This** approach runs in  $O(\log i)$  time, where i is the final answer.

# Question 7 [G] Bogosort repeatedly shuffles the array until it happens to be sorted. Analyze its **best-case**.

worst-case, and average-case time complexity for an array of length n.

### **Algorithm 1:** Bogosort(A[0..n-1])

- 1 while not IsSorted(A) do
  - RandomlyShuffle(A)
- 3 return A
- 4 Function IsSorted(A): for  $i \leftarrow 1$  to n-1 do
- if A[i] < A[i-1] then

return false

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return true

**Note:** RandomlyShuffle runs in O(n) using the Fisher-Yates shuffle.

#### Best-case

- ▶ If the array is already sorted, only one IsSorted check is needed.
- **>** Time complexity: O(n).

#### Worst-case

> Unbounded; the algorithm may never terminate as shuffles are random.

#### Average-case

- ▶ n! possible permutations, (assume) each equally likely.
- **>** Probability of a correct permutation in one shuffle: 1/n!
- Expected number of iteration: n!.
- $\rightarrow$  O(n) for RandomlyShuffle and
  - •• O(11) for Landonny Straine and
  - $\gg O(n)$  for IsSorted.
- **>** Total expected runtime:  $O(n \cdot n!)$ .

# Practical [Optional]

Practical repo: To help you further your understanding, not compulsory; Work for Snack!

- Bruteforce implementation is given, poly\_mult\_bruteforce.
- $\blacksquare$  Implement the D&C algorithm in code, poly\_mult\_dc .
- Check that you get this output:

Brute force result: [2, 9, 17, 16, 6]
Divide and Conquer result: [2, 9, 17, 16, 6]