



**NUS**  
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| **Computing**

**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).

## Algorithm

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

## Proof

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

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

Since the recurrence reduces by  $1/2$  in 2 iterations, we obtain<sup>1</sup>  $T(n) = T(n/2) + \Theta(n)$ . 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:  $1 \cdot f(n/2) = \frac{n}{2} \leq c f(n)$  for  $c = \frac{1}{2} < 1$ . Thus, by Case 3 of the Master Theorem:  $T(n) = \Theta(n)$ .

<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

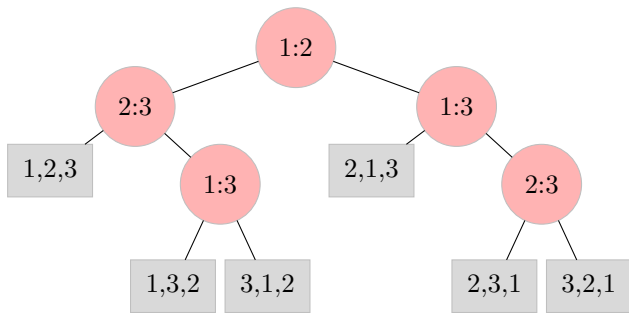


Figure 1: Worst case runtime is the height of the decision tree.

# Polynomial Multiplication (Degree $n$ )

Given two polynomials:

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

$$B(x) = b_n x^n + \dots + b_2 x^2 + b_1 x + b_0$$

Their product:

$$C(x) = A(x) \times B(x) = c_{2n} x^{2n} + \dots + c_2 x^2 + c_1 x + c_0$$

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

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

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

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  $A(x) = a_0 + a_1x + \dots + a_nx^n$ .  
    # B: Coeff [b0, b1, ..., b_n] for  $B(x) = b_0 + b_1x + \dots + b_nx^n$ .  
  
    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

$$A(10) = 352 = 3 \cdot 10^2 + 5 \cdot 10 + 2, \quad \text{i.e. } a_2 = 3, \quad a_1 = 5, \quad a_0 = 2,$$

$$B(10) = 221 = 2 \cdot 10^2 + 2 \cdot 10 + 1, \quad \text{i.e. } b_2 = 2, \quad b_1 = 2, \quad b_0 = 1.$$

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

## Answer

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

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

$$\begin{aligned} c_3 &= a_1 \cdot b_{3-1} + a_2 \cdot b_{3-2} = a_1 \cdot b_2 + a_2 \cdot b_1 \\ &= 5 \cdot 2 + 3 \cdot 2 = 10 + 6 = 16. \end{aligned}$$

$$\begin{aligned} c_2 &= a_0 \cdot b_{2-0} + a_1 \cdot b_{2-1} + a_2 \cdot b_{2-2} \\ &= a_0 \cdot b_2 + a_1 \cdot b_1 + a_2 \cdot b_0 \\ &= 2 \cdot 2 + 5 \cdot 2 + 3 \cdot 1 = 4 + 10 + 3 = 17. \end{aligned}$$

$$\begin{aligned} c_1 &= a_0 \cdot b_{1-0} + a_1 \cdot b_{1-1} = a_0 \cdot b_1 + a_1 \cdot b_0 \\ &= 2 \cdot 2 + 5 \cdot 1 = 4 + 5 = 9. \end{aligned}$$

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

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



**D&C Algorithm**

- 1 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  $\frac{n}{2}$ .

- 2 Compute four smaller polynomial multiplications:

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

- 3 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  $n = 2$ .

## Answer

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

## Computing Partial Products

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

$$\begin{aligned}A_1(10) \times B_2(10) &= (3 \cdot 10 + 5) \times 1 \\ &= 3 \cdot 10 + 5 = 35.\end{aligned}$$

$$\begin{aligned}A_2(10) \times B_1(10) &= 2 \times (2 \cdot 10 + 2) \\ &= 4 \cdot 10 + 4 = 44.\end{aligned}$$

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

**Compute the final polynomial**

$$\begin{aligned}C(10) &= 10^2 \cdot (A_1(10) \times B_1(10)) \\ &\quad + 10 \cdot (A_1(10) \times B_2(10) + A_2(10) \times B_1(10)) \\ &\quad + A_2(10) \times B_2(10) \\ &= 10^2 \cdot (6 \cdot 10^2 + 16 \cdot 10 + 10) \\ &\quad + 10 \cdot (3 \cdot 10 + 5 + 4 \cdot 10 + 4) + 2 \\ &= 6 \cdot 10^4 + 16 \cdot 10^3 + 10 \cdot 10^2 + 7 \cdot 10^2 + 9 \cdot 10 + 2 \\ &= 6 \cdot 10^4 + 16 \cdot 10^3 + 17 \cdot 10^2 + 9 \cdot 10 + 2 \\ &= 60\,000 + 16\,000 + 1\,700 + 90 + 2 \\ &= 77\,992\end{aligned}$$

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

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**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

- 1 Compute two smaller polynomial multiplications:

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

- 2 Compute one multiplication with two additions:

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

- 3 Apply the identity, two subtractions

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

- 4 Compute  $C(x)$

What is the time complexity of Karatsuba's algorithm?

## Answer

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

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

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

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

## Answer

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## 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.

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

With these information,

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

## Answer

### Minimum Number of Weighings

- 1 Divide the balls into three equal groups:  $A$ ,  $B$ , and  $C$ .
- 2 Weigh group  $A$  against group  $B$ .
  - » If  $A = B$ , the heavier ball is in group  $C$ .
  - » If  $A > B$ , the heavier ball is in group  $A$ .
  - » If  $A < B$ , the heavier ball is in group  $B$ .

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

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

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

## Optimal Weighings

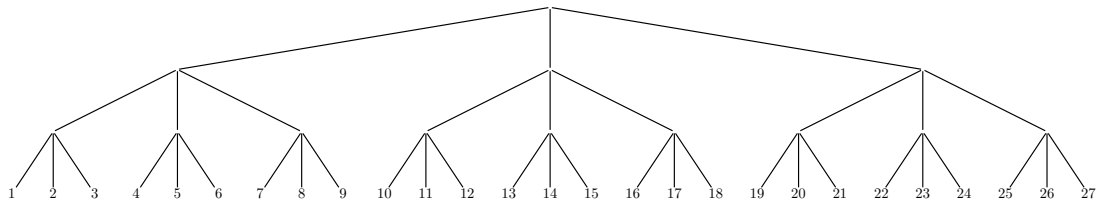


Figure 2:  $3^3 = 27$  possible outcomes, with 3 weighings.

- Each weighing divides the balls into at most<sup>2</sup> 3 groups.
- 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>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  $A[i] \geq i$ . Design an efficient algorithm to solve this problem.

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

- If  $A[j] \geq j$ , then it must hold that (left)  $A[j - 1] \geq j - 1$ , unless  $j = 1$ .
- 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  $A[i] \geq i$  but  $A[i + 1] < i + 1$ .

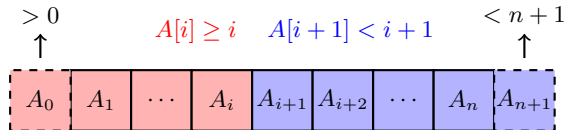


Figure 3: Key observation: Red implies left is red, Blue implies right is blue.

## Answer

### Method 1: Linear Search

- › Perform a linear search to find the largest  $i$  such that  $A[i] \geq i$ .
- › 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

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

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  $n$ .

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

---

```
1 while not IsSorted( $A$ ) do
2   RandomlyShuffle( $A$ )
3 return  $A$ 
4 Function IsSorted( $A$ ):
5   for  $i \leftarrow 1$  to  $n - 1$  do
6     if  $A[i] < A[i - 1]$  then
7       return false
8   return true
```

---

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

## Answer

### 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:  $n!$ ,
  - ››  $O(n)$  for `RandomlyShuffle` and
  - ››  $O(n)$  for `IsSorted`.
- › **Total expected runtime:**  $O(n \cdot n!)$ .

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

- 1 Bruteforce implementation is given, `poly_mult_bruteforce` .
- 2 Implement the D&C algorithm in code, `poly_mult_dc` .
- 3 Check that you get this output:

```
Brute force result: [2, 9, 17, 16, 6]
```

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