ICCS312 Algorithms and Tractability (Term 1/2021-22)

Lecture 3: Divide and Conquer (I)

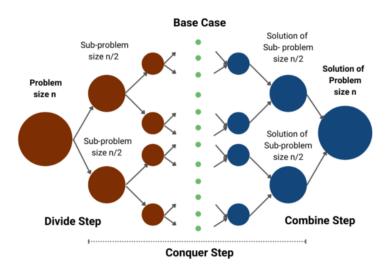
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- Divide-and-Conquer Strategy
- 2 Mergesort
- Sorting lower bound
- 4 Counting inversions

Divide-and-Conquer



Divide-and-Conquer

The divide-and-conquer strategy solves a problem by:

- 1. Divide instance of problem into **WANTE** smaller instances
- Recursively solving these instances
- 3. Obtain solution to original (larger) instance by combining these solutions

Divide-and-Conquer

The real work of a divide conquer algorithm happens at three different places:

- 1. when we partition the problem into smaller sub-problems.
- 2. when the problems are small enough that we can solve them easily.
- 3. when we combine solutions to sub-problems back to a solution for the original problem.

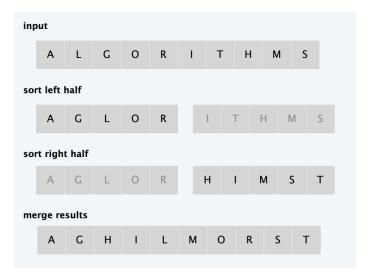
Mergesort

Problem: Given a list *L* of *n* elements from a totally ordered universe, rearrange them in ascending order.

Idea:

- Split the list into left and right halves.
- Recursively sort left half.
- Recursively sort right half.
- Merge two halves to make sorted whole.

Mergesort



Mergesort

T(n) be the mining the of MexeSort(L)

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Algorithm: MergeSort(L)
If len(L) == 1: return L
                              - bare core
Divide L into two halves A and B.
A = MergeSort(A) \leftarrow T(TN/2T)
  = MergeSort(B) \rightarrow T([n/2])
L = Merge(A,B) \qquad \bigcirc (n)
return I.
```

• What is running time?



Analyzing Mergesort

Let T(n) denote the maximum number of compares to mergesort a list of length n

We will have the following recurrence relation.

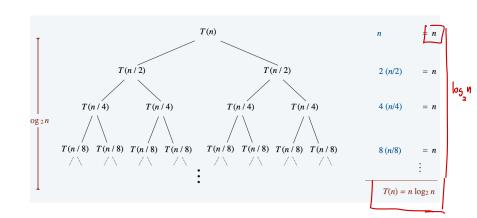
$$T(n) \le \begin{cases} 0 & \text{if } n = 1\\ T(\lfloor n/2 \rfloor) + T(\lfloor n/2 \rfloor) + n & \text{if } n > 1 \end{cases}$$

Proposition

If T(n) satisfies the following recurrence, then $T(n) = n \log_2 n$ (assuming n is a power of 2).

$$T(n) = \begin{cases} 0 & \text{if } n = 1\\ 2T(n/2) + n & \text{if } n > 1 \end{cases}$$







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Proof

- Base case: when n = 1, $T(1) = 0 = n \log_2 n$
- Inductive hypothesis: assume $T(n) = n \log_2 n$
- Goal: show that $T(2n) = 2n \log_2(2n)$



Proof. (Cont.)

$$T(2n) = 2T(n) + 2n$$
 by our def.
 $= 2n \log_2 n + 2n$ by our IH
 $= 2n(\log_2(2n) - 1) + 2n$
 $= 2n \log_2(2n)$

Back to Mergesort

Proposition

If T(n) *satisfies the following recurrence, then* $T(n) \le n \lceil \log_2 n \rceil$.

$$T(n) \leq \begin{cases} 0 & \text{if } n = 1 \\ T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + n & \text{if } n > 1 \end{cases}$$

Proof. [by strong induction]

- Base case: when n = 1, $T(1) = 0 = n \log_2 n$
- Let $n_1 = \lfloor n/2 \rfloor$ and $n_2 = \lceil n/2 \rceil$
- Note that $n_1 + n_2 = n$
- Induction step: assume true for $n \downarrow \downarrow \downarrow 1, 2, ..., n-1$



Back to Mergesort

Proof. (Cont.)

$$T(n) \leq T(n_1) + T(n_2) + n$$

$$\leq n_1 \lceil \log_2 n_1 \rceil + n_2 \lceil \log_2 n_2 \rceil + n \qquad \text{IH}$$

$$\leq n_1 \lceil \log_2 n_2 \rceil + n_2 \lceil \log_2 n_2 \rceil + n$$

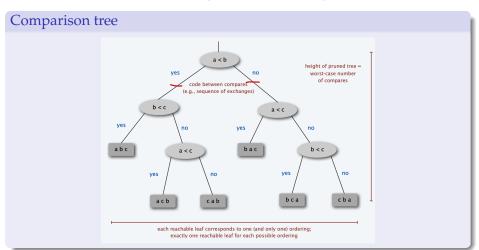
$$= n \lceil \log_2 n_2 \rceil + n$$

$$\leq n (\lceil \log_2 n \rceil - 1) + n$$

$$= n \lceil \log_2 n \rceil$$

Detour: Sorting lower bound

Question: Are there a faster algorithm than Mergesort?



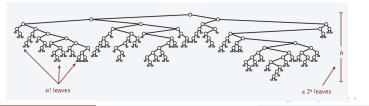
Sorting lower bound

Theorem

Any deterministic compare-based sorting algorithm must make $\Omega(n \log n)$ compares in the worst-case.

Proof.

Assume the input array contain distinct elements a_1, a_2, \ldots, a_n . Worst-case number of compares equals to the height *h* of the pruned comparison tree. Binary tree of height h has $\leq 2^h$ leaves. Potentially, n!different ordering $\Rightarrow n!$ reachable leaves.



Sorting lower bound

Theorem

Any deterministic compare-based sorting algorithm must make $\Omega(n \log n)$ compares in the worst-case.

Proof. (Cont.)

$$2^h \geqslant \text{number of leaves} \geqslant n!$$

$$h \ge \log_2 n! \ge \log_2 e \left(\frac{n}{e}\right)^n = \log_2 e + \log_2 \left(\frac{n}{e}\right)^n = \ln \log_2 n - n / \ln 2$$
By Stirling's formula = ...

Stirling's formula: $n! \ge e(\frac{n}{e})^n$

- Given a list of integers *L* of size *n*.
- We say that two elements L[i] and L[j] form an inversion if L[i] > 1L[i] and i < i.
- So if array is already sorted then inversion count is 0 and, on the other hand, if array is sorted in reverse order then the inversion count is at its maximum.
- Our goal here is to design an algorithm to count the number of inversions given an array.

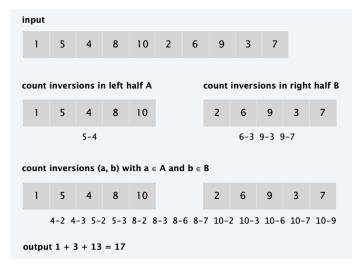
- Naive solution: Consider all possible pairs and just count.
- What is the running time for this?

 \triangle (n^2)

• Can we do better?

- Naive solution: Consider all possible pairs and just count.
- What is the running time for this?
- Can we do better?
- Yes, let's do it divide-and-conquer style!

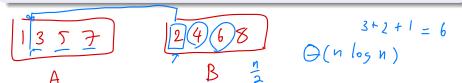
- Divide: separate list into two halves A and B.
- Conquer: recursively count inversion in each list.
- Combine: count inversions (a, b) with $a \in A$ and $b \in B$
- Return sum of three counts

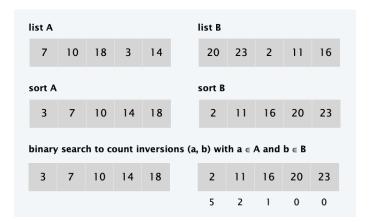


- But, how to count inversions (a, b) with $a \in A$ and $b \in B$?
- Easy if both *A* and *B* are sorted.

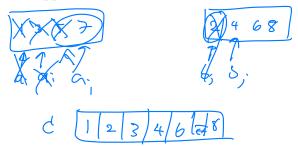
Idea: Binary Search

- Assume A and B are already sorted
- For each element $b \in B$
 - binary search in *A* to find how elements in *A* are greater than *b*.





- We can do better by performing Merge-and-Count
- Assume A and B are sorted.
- Scan *A* and *B* from left to right.
- Compare a_i and b_i .
- If $a_i < b_i$, then a_i is not inverted with any element in B.
- If $a_i > b_j$, then b_i is inverted with any element left in A.
- Append smaller element to sorted list C.



Inv. Cout : 0+3+0

If
$$len(L) == 1$$
: return $(0,L)$ $(0,L)$

Divide I, into A and B.

$$(rA, A) = Sort-And-Count(A)$$
 $T(N/2)$
 $(rA, B) = Sort-And-Count(B)$ $T(N/2)$

return (rA+rB+rAB, L)

• What is running time?

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

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

Discussion: Nuts & Bolts

Problem: A disorganized carpenter has a mixed pile of *n* nuts and *n* bolts.

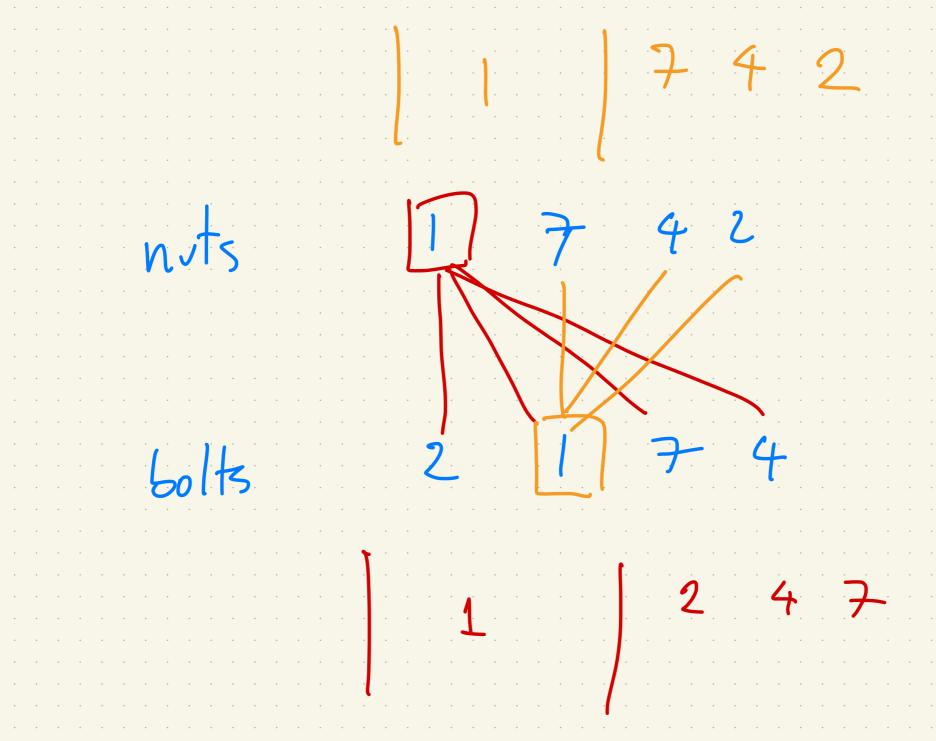
- The goal is to find the corresponding pairs of nuts and bolts.
- Each nut fits exactly one bolt and each bolt fits exactly one nut.
- By fitting a nut and a bolt together, the carpenter can see which one is bigger (but cannot directly compare either two nuts or two bolts).

```
Example:
```

```
nuts = [1,7,4,2]
bolts = [2,1,7,4]
Allowed Operations: Compare(nuts[i], bolts[j])
```

• Challenge: Design an algorithm that makes $O(n \log n)$ compares.

bolt (12)



nut & Solt (nts, Solts): [p = nuts[0]

Solts, bdfs, b= partition (bolts, p)

nuls, nuts, - = powtition (nuts, bg) TCM-m) whe bolts (nuts, bolts,)
TCM-m) what he lets (nuts, bolts,) $7(n) \leq T(m) + T(n-m) + O(m)$