

CMSC 451: Divide and Conquer

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Based on Sections 5.1–5.3 of *Algorithm Design* by Kleinberg & Tardos.

Greedy Recap

Greedy algorithms are usually very natural.

Many problems have nice greedy solutions:

- 1 Topological Sorting (ch. 3)
- 2 Interval Scheduling (4.1)
- 3 Interval Partitioning (4.1)
- 4 Minimizing Lateness (4.2)
- 5 Optimal Scheduling (4.3)
- 6 Shortest Paths (Dijkstra's) (4.4)
- 7 Minimum Spanning Tree (4.5)
- 8 Maximum Separation Clustering (4.7)
- 9 Matroids: Max-Weight

Greedy Recap, 2

We've seen some general patterns for algorithms:

- 1 Sort and then scan (matroid greedy algorithm)
- 2 TreeGrowing

And for proof techniques:

- 1 **Greedy “stays ahead”**: any choice the OPT can make, the greedy can make and will
- 2 **Exchange**: we can transform an OPT solution into greedy with series of small changes.
- 3 **Matroid**: Hereditary Subset System with Augmentation Property

Divide and Conquer

Divide and Conquer is a different framework.

Related to *induction*:

- Suppose you have a “box” that can solve problems of size $\leq k < n$
- You use this box on some subset of the input items to get partial answers
- You combine these partial answers to get the full answer.

But: you construct the “box” by recursively applying the same idea until the problem is small enough to be solved by brute force.

Merge Sort

```
MergeSort(L):  
    if |L| = 2:  
        return [min(L), max(L)]  
    else:  
        L1 = MergeSort(L[0, |L|/2])  
        L2 = MergeSort(L[|L|/2+1, |L|-1])  
        return Combine(L1, L2)
```

- In practice, you sort in-place rather than making new lists.
- `Combine(L1,L2)` walks down the sorted lists putting the smaller number onto a new list. Takes $O(n)$ time
- Total time: $T(n) \leq 2T(n/2) + cn$.

To Solve a Recurrence

Given a recurrence such as $T(n) \leq 2T(n/2) + cn$, we want a simple upper bound on the total running time.

Two common ways to “solve” such a recurrence:

- 1 Unroll the recurrence and see what the pattern is.
Typically, you'll draw the recursion tree.
- 2 Guess an answer and prove that it's right.

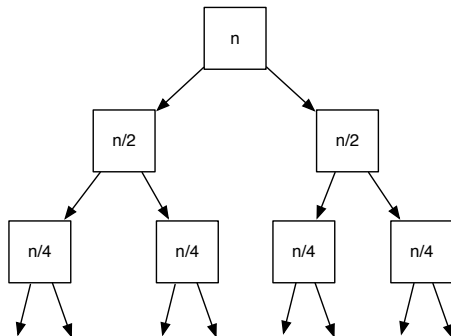
Solving Recurrences

Draw the first few levels of the tree.

Write the amount of **work done at each level** in terms of the level.

Figure out the **height** of the tree.

Sum over all levels of the tree.



$$T(n) \leq 2T(n/2) + cn$$

Each level is cn . There are $\log n$ levels, so $T(n)$ is $O(n \log n)$.

Substitution Method

Substitution method is based on induction. We:

- 1 Show $T(k) \leq f(k)$ for some small k .
- 2 Assume $T(k) \leq f(k)$ for all $k < n$.
- 3 Show $T(n) \leq f(n)$.

$$T(n) \leq 2T(n/2) + cn$$

Base Case: $2c \log 2 = 2c \geq T(2)$

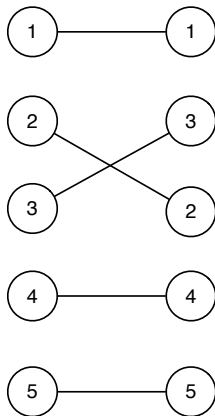
Induction Step:

$$\begin{aligned}T(n) &\leq 2T(n/2) + cn \\&\leq 2c(n/2) \log(n/2) + cn \\&= cn[(\log n) - 1] + cn \\&= cn \log n\end{aligned}$$

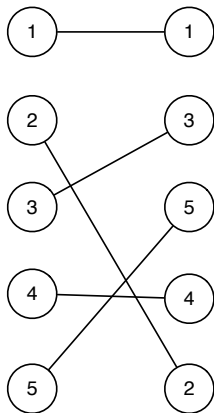
Counting Inversions

Comparing Rankings

Suppose two customers **rank** a list of movies.



similar



more different

A measure of distance

What's a good measure of how dissimilar two rankings are?

A measure of distance

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We can count the number of inversions:

- Assume one of the rankings is $1, 2, 3, \dots, n$.
- Denote the other ranking by a_1, a_2, \dots, a_n .
- An inversion is a pair (i, j) such that $i < j$ but $a_j < a_i$.

Two identical rankings have no inversions.

How many inversions do opposite rankings have?

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$$\binom{n}{2}$$

How can we count inversions quickly?

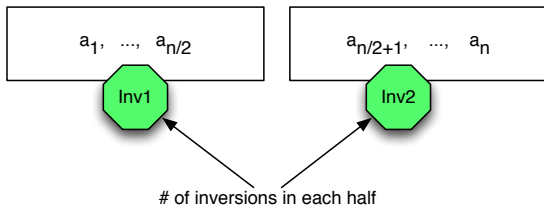
How can we count inversions quickly?

- **Brute Force:** check every pair: $O(n^2)$.
- Some sequences might have $O(n^2)$ inversions, so you might think that it might take as much as $O(n^2)$ time to count them.
- In fact, with divide and conquer, you can count them in $O(n \log n)$ time.

Basic Divide and Conquer

Count the number of inversions in the sequence a_1, \dots, a_n .

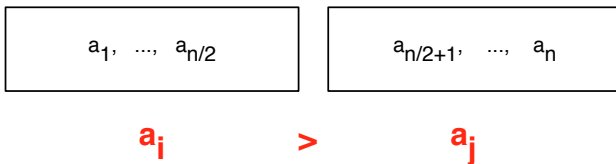
Suppose I told you the number of inversions in the first half of the list and in the second half of the list:



What kinds of inversions are not accounted for in $Inv1 + Inv2$?

Half-Crossing Inversions

The inversions we have to count during the merge step:

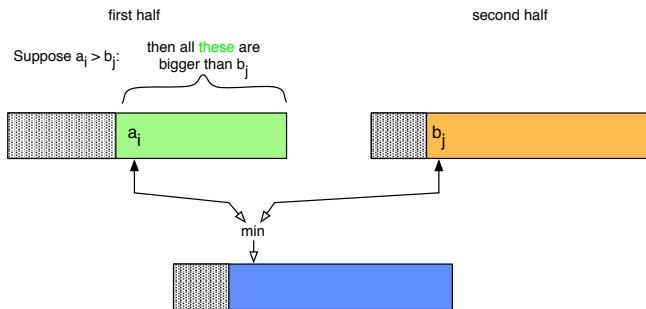


The crux is that we have to count these kinds of inversion in $O(n)$ time.

What if each of the half lists were sorted?

Suppose each of the half lists were sorted.

If we find a pair $a_i > a_j$, then we can infer many other inversions:



Each of the green items is an inversion with b_j .

Merge-and-Count

```
MergeAndCount(SortedList A, SortedList B):  
    a = b = CrossInvCount = 0  
    OutList = empty list  
    While a < |A| and b < |B|:      // not at end of a list  
        next = min(A[a], B[b])  
        OutList.append(next)  
  
        If B[b] == next:  
            b = b + 1  
            CrossInvCount += |A| - a //inc by # left in A  
        Else  
            a = a + 1  
    EndWhile  
    Append the non-empty list to OutList  
    Return CrossInvCount and OutList
```

Sorted!

Note that MergeAndCount will produce a sorted list as well as the number of cross inversions.

SortAndCount

```
SortAndCount(List L):
```

```
  If  $|L| == 1$ : Return 0
```

```
  A, B = first & second halves of L
```

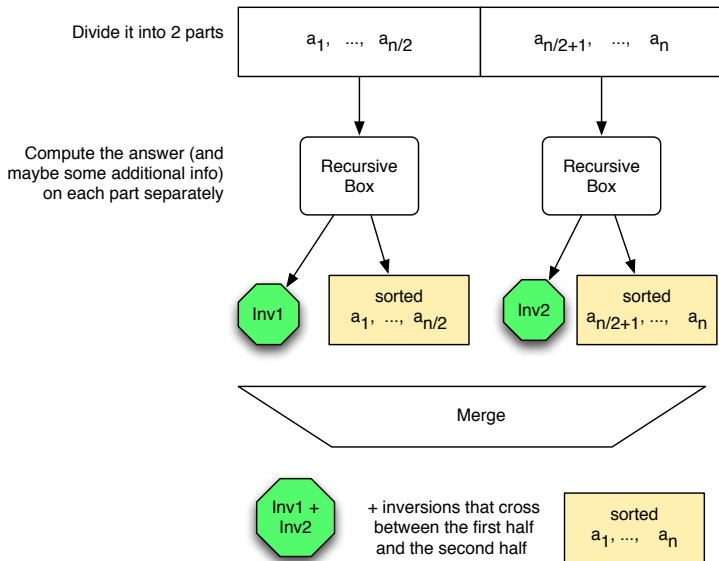
```
  invA, SortedA = SortAndCount(A)
```

```
  invB, SortedB = SortAndCount(B)
```

```
  crossInv, SortedL = MergeAndSort(SortedA, SortedB)
```

```
  Return invA + invB + crossInv and SortedL
```

Algorithm Schematic



Running time?

What's the running time of `SortAndCount`?

Running time?

What's the running time of SortAndCount?

Break the problem into two halves.

Merge takes $O(n)$ time.

$$T(n) \leq 2T(n/2) + cn$$

\implies Total running time is $O(n \log n)$.