Divide and Conquer Algorithms

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- Solve each part recursively.
- Solve base cases by brute force.
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- Solve each part recursively.
- Solve base cases by brute force.
- Efficiently combine solutions for sub-problems into final solution.
- Common use:
 - Partition problem into two equal sub-problems of size n/2.
 - Solve each part recursively.
 - Combine the two solutions in O(n) time.
 - Resulting running time is $O(n \log n)$.

Mergesort

Sort

INSTANCE: Nonempty list $L = x_1, x_2, \ldots, x_n$ of integers.

SOLUTION: A permutation y_1, y_2, \ldots, y_n of x_1, x_2, \ldots, x_n such that $y_i \leq y_{i+1}$, for all $1 \leq i < n$.

• Mergesort is a divide-and-conquer algorithm for sorting.

- 1. Partition L into two lists A and B of size $\lfloor n/2 \rfloor$ and $\lfloor n/2 \rfloor$ respectively.
- 2. Recursively sort A.
- 3. Recursively sort B.
- 4. Merge the sorted lists A and B into a single sorted list.

Merging Two Sorted Lists

• Merge two sorted lists $A = a_1, a_2, \ldots, a_k$ and $B = b_1, b_2, \ldots, b_l$.

Maintain a *current* pointer for each list. Initialise each pointer to the front of the list. While both lists are nonempty:

> Let a_i and b_j be the elements pointed to by the *current* pointers. Append the smaller of the two to the output list. Advance the current pointer in the list that the smaller element belonged to.

EndWhile

Append the rest of the non-empty list to the output.

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• Running time of this algorithm is O(k + l).

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Worst-case running time for *n* elements $(T(n)) \leq$ Worst-case running time for $\lfloor n/2 \rfloor$ elements + Worst-case running time for $\lceil n/2 \rceil$ elements + Time to split the input into two lists + Time to merge two sorted lists.

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- Three basic ways of solving this recurrence relation:
 - 1. "Unroll" the recurrence (somewhat informal method).
 - 2. Guess a solution and substitute into recurrence to check.
 - 3. Guess solution in O() form and substitute into recurrence to determine the

Unrolling the recurrence

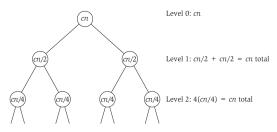


Figure 5.1 Unrolling the recurrence $T(n) \le 2T(n/2) + O(n)$.

Unrolling the recurrence

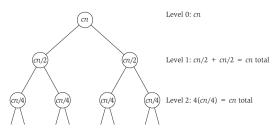


Figure 5.1 Unrolling the recurrence $T(n) \le 2T(n/2) + O(n)$.

- Recursion tree has log n levels.
- Total work done at each level is cn.
- Running time of the algorithm is cn log n.
- Use this method only to get an idea of the solution.

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- Inductive step: Prove $T(n) \leq cn \log n$.

Т

$$\begin{aligned} f(n) &\leq 2T\left(\frac{n}{2}\right) + cn \\ &\leq 2\left(\frac{cn}{2}\log\left(\frac{n}{2}\right)\right) + cn, \text{ by the inductive hypothesis} \\ &= cn\log\left(\frac{n}{2}\right) + cn \\ &= cn\log n - cn + cn \\ &= cn\log n. \end{aligned}$$

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$$= cn\log n.$$

- Why doesn't an attempt to prove $T(n) \le kn$, for some k > 0 work?
- Why is $T(n) \leq kn^2$ a "loose" bound?

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- Guess that the solution is $kn \log n$ (logarithm to the base 2).
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- $k \ge c$ will work.

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- Let m be the smallest power of 2 larger than n.
- $T(n) \leq T(m) = O(m \log m) = O(n \log n)$, because $m \leq 2n$.

Other Recurrence Relations

- ▶ Divide into q sub-problems of size n/2 and merge in O(n) time. Two distinct cases: q = 1 and q > 2.
- ▶ Divide into two sub-problems of size n/2 and merge in $O(n^2)$ time.

T(n) = qT(n/2) + cn, q = 1

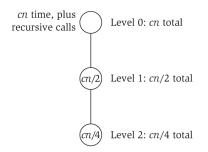


Figure 5.3 Unrolling the recurrence $T(n) \le T(n/2) + O(n)$.

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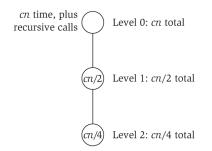


Figure 5.3 Unrolling the recurrence $T(n) \le T(n/2) + O(n)$.

- ► Each invocation reduces the problem size by a factor of 2 ⇒ there are log n levels in the recursion tree.
- At level *i* of the tree, the problem size is $n/2^i$ and the work done is $cn/2^i$.
- Therefore, the total work done is

$$\sum_{i=0}^{i=\log n} \frac{cn}{2^i} = O(n).$$

T(n) = qT(n/2) + cn, q > 2

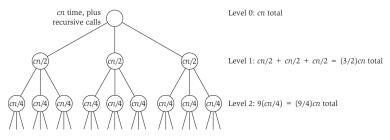


Figure 5.2 Unrolling the recurrence $T(n) \le 3T(n/2) + O(n)$.

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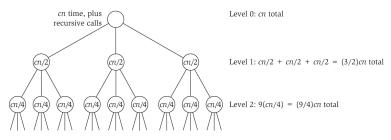


Figure 5.2 Unrolling the recurrence $T(n) \le 3T(n/2) + O(n)$.

- There are log n levels in the recursion tree.
- At level *i* of the tree, there are q^i sub-problems, each of size $n/2^i$.
- The total work done at level *i* is $q^i cn/2^i$.
- Therefore, the total work done is

$$T(n) \leq \sum_{i=0}^{i=\log n} q^i \frac{cn}{2^i} \leq c$$

T(n) = qT(n/2) + cn, q > 2

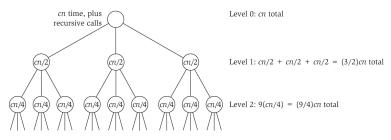


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$$T(n) \leq \sum_{i=0}^{i=\log n} q^i \frac{cn}{2^i} \leq O(n^{\log_2 q}).$$

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