Divide and Conquer Algorithms

T. M. Murali

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Divide and Conquer

- ▶ Break up a problem into several parts.
- Solve each part recursively.
- Solve base cases by brute force.
- ▶ Efficiently combine solutions for sub-problems into final solution.

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- Break up a problem into several parts.
- 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 $L = x_1, x_2, ..., x_n$ 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 $\lceil n/2 \rceil$ 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 $A = a_1, a_2, ..., a_k$ and $B = b_1, b_2, ..., b_l$.
 - 1. Maintain a current pointer for each list.
 - 2. Initialise each pointer to the front of its list.
 - 3. While both lists are nonempty:
 - 3.1 Let a_i and b_i be the elements pointed to by the *current* pointers.
 - 3.2 Append the smaller of the two to the output list.
 - 3.3 Advance the current pointer in the list that the smaller element belonged to.
 - 4. Append the rest of the non-empty list to the output.

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- ▶ Running time of this algorithm is O(k + I).

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Worst-case running time for n elements (T(n)) \le 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 ways of solving this recurrence relation:
 - 1. "Unroll" the recurrence (somewhat informal method).
 - 2. Guess a solution and substitute into recurrence to check.
 - Guess solution in O() form and substitute into recurrence to determine the constants.

Unrolling the recurrence

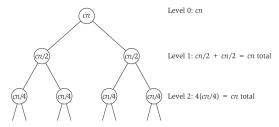


Figure 5.1 Unrolling the recurrence $T(n) \le 2T(n/2) + O(n)$.

Unrolling the recurrence

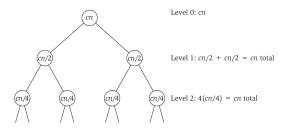


Figure 5.1 Unrolling the recurrence $T(n) \le 2T(n/2) + O(n)$.

- 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.

- ▶ Guess that the solution is $T(n) \le cn \log n$ (logarithm to the base 2).
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- ▶ Inductive step: Prove $T(n) \le cn \log n$.

$$T(n) \leq 2T\left(\frac{n}{2}\right) + cn$$

$$\leq 2\left(\frac{cn}{2}\log\left(\frac{n}{2}\right)\right) + cn, \text{ by the inductive hypothesis}$$

$$= cn\log\left(\frac{n}{2}\right) + cn$$

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

Partial Substitution

- ▶ Guess that the solution is $kn \log n$ (logarithm to the base 2).
- ▶ Substitute guess into the recurrence relation to check what value of *k* will satisfy the recurrence relation.

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- ▶ $k \ge c$ will work.

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- ▶ How do we generalise the proof?

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- ▶ Basic axiom: $T(n) \le T(n+1)$, for all n: worst case running time increases as input size increases.
- ▶ Let m be the smallest power of 2 larger than n.
- $T(n) \leq T(m) = O(m \log m)$

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- ▶ How do we generalise the proof?
- ▶ Basic axiom: $T(n) \le T(n+1)$, for all n: worst case running time increases as input size increases.
- ▶ Let m be the smallest power of 2 larger than n.
- ► $T(n) \le T(m) = O(m \log m) = O(n \log n)$, because $m \le 2n$.

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.



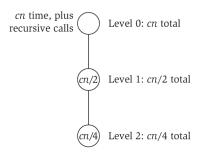


Figure 5.3 Unrolling the recurrence $T(n) \le T(n/2) + O(n)$.

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

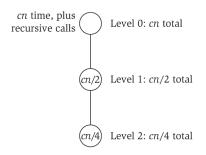


Figure 5.3 Unrolling the recurrence $T(n) \le T(n/2) + O(n)$.

- ▶ Each invocation reduces the problem size by a factor of $2 \Rightarrow$ 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

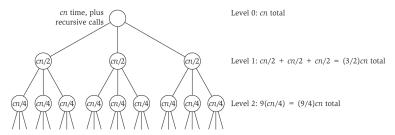


Figure 5.2 Unrolling the recurrence $T(n) \le 3T(n/2) + O(n)$.

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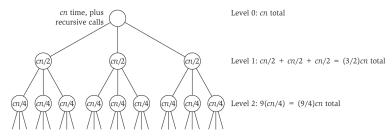


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- ightharpoonup There are $\log n$ levels in the recursion tree.
- ▶ At level *i* of the tree, there are q^i sub-problems, each of size $n/2^i$.
- ▶ The total work done at level i is $q^i cn/2^i$.
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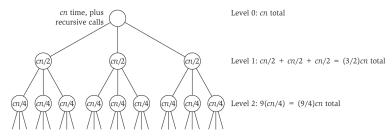


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$$T(n) \leq \sum_{i=0}^{i=\log n} q^i \frac{cn}{2^i} \leq O(n^{\log_2 q}).$$

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