COMP20007 Design of Algorithms

Master Theorem

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Lecture 10

Semester 1, 2020

Divide and Conquer

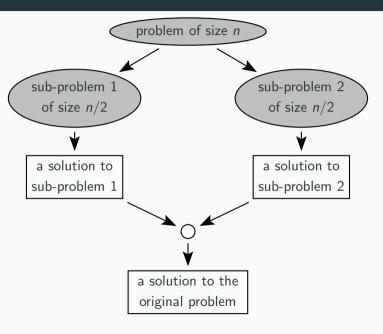
We earlier studied recursion as a powerful problem solving technique.

The divide-and-conquer strategy tries to make the most of this:

- 1. Divide the given problem instance into smaller instances.
- 2. Solve the smaller instances recursively.
- 3. Combine the smaller solutions to solve the original instance.

This works best when the smaller instances can be made to be of equal size.

Split-Solve-and-Join Approach



Divide-and-Conquer Algorithms

You have seen:

- Tree traversal
- Closest pair

You will learn later:

- Mergesort
- Quicksort

Divide-and-Conquer Recurrences

What is the time required to solve a problem of size n by divide-and-conquer?

For the general case, assume we split the problem into b instances (each of size n/b), of which a need to be solved:

$$T(n) = aT(n/b) + f(n)$$

where f(n) expresses the time spent on dividing a problem into b sub-problems and combining the a results.

(A very common case is
$$T(n) = 2T(n/2) + n$$
.)

How do we find closed forms for these recurrences?

The Master Theorem

(A proof is in Levitin's Appendix B.)

For integer constants $a \ge 1$ and b > 1, and function f with $f(n) \in \Theta(n^d), d \ge 0$, the recurrence

$$T(n) = aT(n/b) + f(n)$$

(with T(1) = c) has solutions, and

$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Note that we also allow a to be greater than b.

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

$$a = 2, b = 2, d = 1$$

$$1 \times n$$

$$2 \times n/2$$

$$4 \times n/4$$

$$\vdots$$

$$(\log_2 n \text{ times})$$

T(n) = 4T(n/4) + n a = 4, b = 4, d = 1



$$T(n) = 2T(n/2) + n^2$$

$$a = 2, b = 2, d = 2$$

Here $a < b^d$ and we simply get n^d .





