

Master Theorem for Recurrences

Recall recurrences from last day

- recursion tree
- proof by induction

Example - changing variables

$$T(n) = 2T(\lfloor \sqrt{n} \rfloor) + \log n$$

$$\text{Let } m = \log n \quad \text{so } n = 2^m$$

$$T(2^m) = 2T(2^{m/2}) + m$$

$$\text{Let } S(m) = T(2^m)$$

$$\text{so } S(m/2) = T(2^{m/2})$$

Then $S(m) = 2S(m/2) + m$ which we know

$$S(m) \in O(m \log m) \quad \text{so } T(2^m) = O(m \log m)$$

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

We often get recurrences of the form

$$T(n) = a T\left(\frac{n}{b}\right) + c \cdot n^k$$

Note: the recurrence relations you studied in MATH 239 were homogenous i.e. last term was 0.

This arises if we divide problem of size n into

① subproblems of size $\left(\frac{n}{b}\right)$ and do $c \cdot n^k$ extra work.

e.g. $k=1$

- $a=b=2$

$$T(n) = 2T\left(\frac{n}{2}\right) + c \cdot n \quad \text{mergesort}$$

$$O(n \log n)$$

- $a=1 \quad b=2$

$$T(n) = T\left(\frac{n}{2}\right) + c \cdot n$$

$$O(n)$$

- $a=4 \quad b=2$

$$T(n) = 4T\left(\frac{n}{2}\right) + c \cdot n$$

EX. $O(n^2)$

Theorem ("Master Theorem")

$$T(n) = aT\left(\frac{n}{b}\right) + c \cdot n^k$$

↖ floor/ceiling allowed

$$a \geq 1, b > 1, c > 0, k \geq 1$$

Then

$$T(n) \in \begin{cases} \Theta(n^k) & \text{if } a < b^k \quad \text{i.e. } \log_b a < k \\ \Theta(n^k \log n) & \text{if } a = b^k \\ \Theta(n^{\log_b a}) & \text{if } a > b^k \end{cases}$$

Notes

- CLRS has more general version with $f(n)$ in place of cn^k
- You are not responsible for the proof but must know & apply the theorem

A rigorous proof is by induction.

We'll just make sense of it using recursion tree (written out)

$$T(n) = aT\left(\frac{n}{b}\right) + c \cdot n^k$$

$$= a \left[aT\left(\frac{n}{b^2}\right) + c \cdot \left(\frac{n}{b}\right)^k \right] + c \cdot n^k$$

$$= a^2 T\left(\frac{n}{b^2}\right) + a \cdot c \cdot \left(\frac{n}{b}\right)^k + c \cdot n^k$$

$$= \dots = a^3 T\left(\frac{n}{b^3}\right) + a^2 \cdot c \cdot \left(\frac{n}{b^2}\right)^k + a \cdot c \cdot \left(\frac{n}{b}\right)^k + c \cdot n^k$$

$$= \dots = a^{\log_b n} T(1) + \sum_{i=0}^{\log_b n - 1} a^i \cdot c \cdot \left(\frac{n}{b^i}\right)^k$$

$$\boxed{a^{\log_b n} = n^{\log_b a}}$$

change base of log

how far?

$$\frac{n}{b^i} = 1 \text{ so } \boxed{i = \log_b n}$$

$$= n^{\log_b a} T(1) + c \cdot n^k \sum_{i=0}^{\log_b n - 1} \left(\frac{a}{b^k}\right)^i$$

- If $a < b^k$ i.e. $\log_b a < k$
then the second term dominates.
Also Σ is constant. So get $O(n^k)$
 - If $a = b^k$ then Σ is $O(\log n)$
so get $O(n^k \log n)$
 - If $a > b^k$ then the first term dominates
 $O(n^{\log_b a})$.
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More general Master Theorem

$$T(n) = aT\left(\frac{n}{b}\right) + c \cdot n^k \log^l n$$

$$a \geq 1, b > 1, k \geq 0$$

Then

$$T(n) \in \begin{cases} \Theta(n^k \log^l n) & \text{if } a < b^k \text{ } (\log_b a < k) \\ \Theta(n^k \log^{l+1} n) & \text{if } a = b^k \\ \Theta(n^{\log_b a}) & \text{if } a > b^k \end{cases}$$

Divide and Conquer Examples

Counting Inversions

Some web sites try to match your preferences (for music, movies, books) with others.

How do you compare two rankings?

e.g. I like best B D C A worst
You like A D B C

Count: how many pairs do we rank differently?

i.e. how many pairs of lines cross (out of 6 pairs)

BD BA DA CA

$= 4 \cdot 3 / 2$

(the pairs that don't cross are BC DC)

Equivalently my ranking 1 2 3 4

Your ranking 4 2 1 3

and we count # inversions - # pairs out of order in 2nd list.

Brute Force: check all $\binom{n}{2}$ pairs $O(n^2)$

Does sorting help? Doesn't seem to.

Better with divide and conquer

List $a_1 \dots a_n$. count # inversions.

Divide list in two $m = \lceil \frac{n}{2} \rceil$

$A = a_1 \dots a_m$ $B = a_{m+1} \dots a_n$

Recursively count # inversions in each half, r_A, r_B

Combine: $\text{answer} \leftarrow r_A + r_B + r$

$r = \# \text{inversions with one element in } A, \text{ one in } B$

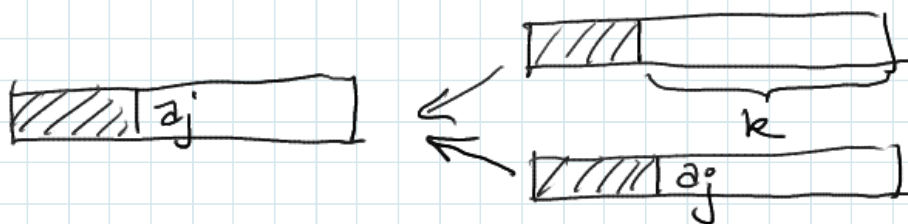
$= \# \text{pairs } a_i, a_j \quad a_i \in A \quad a_j \in B, \quad a_i > a_j$

How do we find r ?

Can we count, for each $a_j \in B$, how many larger elements are there in A ? — r_j

$$r = \sum_{a_j \in B} r_j$$

Think about mergesort: sort A , sort B , merge



When a_j is output to merged list $r_j \leftarrow k$

Whole algorithm

Sort-and-Count (L) — returns sorted L , #inversions

- divide L into A, B
first half, second half
- $(r_A, A) \leftarrow \text{Sort-and-Count}(A)$
- $(r_B, B) \leftarrow \text{Sort-and-Count}(B)$
- $r \leftarrow 0$
- Do merge of A and B

When element is moved from B to output

$r \leftarrow r + \# \text{elements remaining in } A$

- return $(r_A + r_B + r, \text{merged list})$

$$T(n) = 2T\left(\frac{n}{2}\right) + O(n)$$

solution: $T(n) = O(n \log n)$

Question: Is there a better algorithm?

$$O(n \log n / \log \log n) \quad 189$$

$O(n \sqrt{\log n})$ 2010 Timothy Chan et al.
using techniques/model where
sorting is $O(n \log n)$