Data Structures and Algorithms

Spring 2009-2010

Outline

- Sorting Algorithms (contd.)
 - $o(n^2)$ Algorithms
 - Solving Divide-and-Conquer and the Master Theorem

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Heapsort

- Have seen that we can build a heap in linear time using n/2 calls of trickleDown()
- If we now call deleteMin() n times, we will get the elements in smallest-to-largest (non-decreasing) order
- Where to store them as we peel them off?
- Do we need an extra array?
- No. Each time a deleteMin() is called, place the result in the slot just opened up by the shrunk array
- Problem with this is that after n deleteMin()s the elements now appear largest-to-smallest (non-increasing)
- Two solutions:
 - Use a (max)heap with the deleteMax() operation
 - 2 After the n deleteMin() s reverse the contents of the array to get it smallest-to-largest; reversing an array can be done in O(n)-time using just O(1) additional storage

Mergesort

- Mergesort strategy: To sort an array of numbers
 - Sort the first half of the array
 - Sort the second half of the array
 - Merge the two halves of the array
- Running time analysis:
 - Sorting an array comprising one element takes unit time
 - To sort n numbers we perform two sorts on n/2 numbers and then merge the resulting arrays
 - Merging two arrays of length n/2 each takes time n since we can make one

Mergesort (contd.)

Recurrence relation:

$$T(1) = 1$$

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

- Assume *n* is a power of 2 $(n = 2^k, k = \log n)$
- Can argue then, that if n is not a power of 2, it falls between two powers of two: $2^{\lfloor k \rfloor} < n < 2^{\lceil k \rceil}$
- Two techniques to solving the recurrence

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

Mergesort Solution 1

Divide through by *n*,

$$\frac{T(n)}{n} = \frac{T(n/2)}{\frac{n}{2}} + 1$$
and letting $T'(n) = T(n)/n$

$$T'(n) = T'(n/2) + 1$$

$$= T'(n/4) + 1 + 1$$

$$\vdots$$

$$= T'(1) + \underbrace{1 + \dots + 1}_{k}$$

Thus,

$$T'(n) = T(n)/n = 1 + \log n$$

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

Mergesort Solution 2

$$T(n) = 2T(n/2) + n$$
, and so,
 $T(n/2) = 2T(n/4) + n/2$

Therefore

$$T(n) = 2(2T(n/4) + n/2) + n$$

$$= 4T(n/4) + 2n$$

$$\vdots$$

$$T(n) = 2^k T(n/2^k) + k \times n$$
When $k = \log n$, $n/2^k = 1$ (since $n = 2^k$ for some k)
$$T(n) = nT(1) + n\log n$$

$$= n + n\log n$$

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

Did you know?

$$a^{\log_c n} = (c^{\log_c a})^{\log_c n} = c^{\log_c a \log_c n}$$
$$= c^{\log_c n \log_c a} = (c^{\log_c n})^{\log_c a} = n^{\log_c a}$$

In summary

$$a^{\log_c n} = n^{\log_c a}$$

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General Solutions to Divide and Conquer

How do we solve recurrences of the form

$$T(n) = aT(\frac{n}{c}) + bn$$

 $T(1) = b$

- Since we are dividing by c, we will assume that n is a power of c; that is, $n = c^k$ or, $k = log_c n$
- Of course, this is not the case in general but it is **always** the case that n is sandwiched between two powers of c: $c^k < n < c^{k+1}$

$$T(n) = aT(\frac{n}{c}) + bn$$

$$= a[aT(\frac{n}{c^2}) + b\frac{n}{c}] + bn$$

$$= a^2T(\frac{n}{c^2}) + b\frac{a}{c}n + bn$$

$$= a^2[aT(\frac{n}{c^3}) + \frac{bn}{c^2}] + b\frac{a}{c}n + bn$$

$$= a^3T(\frac{n}{c^3}) + b\frac{a^2}{c^2}n + b\frac{a}{c}n + bn$$

$$\vdots$$

$$= a^{\log_c n}T(1) + bn[\frac{a^{\log_c n - 1}}{c^{\log_c n - 1}} + \dots + \frac{a}{c} + 1]$$

$$T(n) = a^{\log_c n} T(1) + bn \sum_{i=0}^{\log_c n-1} \left(\frac{a}{c}\right)^i$$

$$= a^{\log_c n} b \frac{n}{n} + bn \sum_{i=0}^{\log_c n-1} \left(\frac{a}{c}\right)^i, \quad \text{since } T(1) = b$$

$$= bn \frac{a^{\log_c n}}{c^{\log_c n}} + bn \sum_{i=0}^{\log_c n-1} \left(\frac{a}{c}\right)^i, \quad \text{since } c^{\log_c n} = n$$

$$= bn \sum_{i=0}^{\log_c n} \left(\frac{a}{c}\right)^i$$

$$= bn \left[\frac{\left(\frac{a}{c}\right)^{\log_c n+1} - 1}{\frac{a}{c} - 1}\right], \quad \frac{a}{c} \neq 1$$

- There are three cases to consider with this expression for T(n)
 - $\mathbf{0} \ a = c$
 - \mathbf{e} a < c
 - Θ a > c
- In each case we will reduce the expression as much as we can and then argue that since n is always bounded from above and below by a power of c, then T(n) will be " Θ " of the function of n that we derive.
- See also the Master Theorem page on Wikipedia

Case
$$\bullet$$
 $\frac{a}{c} = 1$ $(a = c)$ then

$$T(n) = bn [\log_c n + 1] = \Theta(n \log_c n)$$

Case 2 $\frac{a}{c}$ < 1 (a < c) then

$$\lim_{n\to\infty} \left(\frac{a}{c}\right)^{\log_c n+1} \to 0$$

$$bn\left[\frac{\left(\frac{a}{c}\right)^{\log_{c}n+1}-1}{\frac{a}{c}-1}\right] = bn\left[\frac{1-\left(\frac{a}{c}\right)^{\log_{c}n+1}}{1-\frac{a}{c}}\right]$$

$$\approx bn\left[\frac{1}{\frac{c-a}{c}}\right]$$

$$= bn\frac{c}{c-a}$$

$$= \Theta(n)$$

Case 3 $\frac{a}{c} > 1$ (a > c) then

$$\lim_{n\to\infty} \left(\frac{a}{c}\right)^{\log_c n+1} \to \infty$$

so the remaining terms are insignificant, and the running time should grow very rapidly Using the "Did you know?" fact, $a^{\log_c n} = n^{\log_c a}$

$$T(n) = bn \left(\frac{a}{c}\right)^{\log_c n + 1}$$

$$= bn \frac{a}{c} \left(\frac{a}{c}\right)^{\log_c n}$$

$$= \frac{ab}{c} n \frac{n^{\log_c a}}{n} = \Theta(n^{\log_c a})$$