## Quicksort



#### Quicksort

- Sorts in place
- Sorts O(n lg n) in the average case
- Sorts O(n²) in the worst case
  - But in practice, it's quick
  - And the worst case doesn't happen often (but more on this later...)



#### Quicksort

- Another divide-and-conquer algorithm
  - The array A[p..r] is partitioned into two non-empty subarrays A[p..q] and A[q+1..r]
    - Invariant: All elements in A[p..q] are less than all elements in A[q+1..r]
  - The subarrays are recursively sorted by calls to quicksort
  - Unlike merge sort, no combining step: two subarrays form an already-sorted array



#### Divide-and-conquer

1. Divide: Partition the array into two subarrays around a pivot x such that elements in lower subarray  $\leq x \leq$  elements in upper subarray.



- Conquer: Recursively sort the two subarrays.
- *3. Combine*: Trivial (because in place).

Key: Linear-time <u>partitioning</u> procedure.

#### Quicksort Code

```
Quicksort(A, p, r)
    if (p < r)
        q = Partition(A, p, r);
        Quicksort(A, p, q);
        Quicksort(A, q+1, r);
```

#### **Partition**

- Clearly, all the action takes place in the partition() function
  - Rearranges the subarray in place
  - End result:
    - Two subarrays
    - All values in first subarray ≤ all values in second
  - Returns the index of the "pivot" element separating the two subarrays
- How do you suppose we implement this?

#### Partition In Words

- Partition(A, p, r):
  - Select an element to act as the "pivot" (which?)
  - Grow two regions, A[p..i] and A[j..r]
    - All elements in A[p..i] <= pivot</li>
    - All elements in A[j..r] >= pivot
  - Increment i until A[i] >= pivot
  - Decrement j until A[j] <= pivot</p>
  - Swap A[i] and A[j]
  - Repeat until i >= j
  - Return j

Note: slightly different from book's partition()

#### Partition Code

```
Partition(A, p, r)
    x = A[p];
                                      Illustrate on
    i = p - 1;
                             A = \{5, 3, 2, 6, 4, 1, 3, 7\};
    j = r + 1;
    while (TRUE)
        repeat
            j--;
        until A[j] \le x;
        repeat
                                        What is the running time of
            i++;
                                           partition()?
        until A[i] >= x;
        if (i < j)
            Swap(A, i, j);
        else
            return j;
```

#### Partition Code

```
Partition(A, p, r)
    x = A[p];
    i = p - 1;
    j = r + 1;
    while (TRUE)
        repeat
             j--;
        until A[j] \le x;
        repeat
            i++;
                                      partition () runs in O(n) time
        until A[i] >= x;
        if (i < j)
            Swap(A, i, j);
        else
            return j;
```

#### **Analyzing Quicksort**

- What will be the worst case for the algorithm?
  - Partition is always unbalanced
- What will be the best case for the algorithm?
  - Partition is perfectly balanced
- Which is more likely?
  - The latter, by far, except...
- Will any particular input elicit the worst case?
  - Yes: Already-sorted input

#### **Analyzing Quicksort**

In the worst case:

$$T(1) = \Theta(1)$$

$$T(n) = T(n - 1) + \Theta(n)$$

Works out to

$$T(n) = \Theta(n^2)$$



#### **Analyzing Quicksort**

• In the best case:

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

What does this work out to?

$$T(n) = \Theta(n \lg n)$$



#### Improving Quicksort

- The real liability of quicksort is that it runs in O(n²) on already-sorted input
- Book discusses two solutions:
  - Randomize the input array, OR
  - Pick a random pivot element
- How will these solve the problem?
  - By insuring that no particular input can be chosen to make quicksort run in O(n²) time



- Assuming random input, average-case running time is much closer to O(n lg n) than O(n²)
- First, a more intuitive explanation/example:
  - Suppose that partition() always produces a 9-to-1 split. This looks quite unbalanced!
  - The recurrence is thus: T(n) = T(9n/10) + T(n/10) + nUse n instead of O(n) for convenience (how?)
  - How deep will the recursion go? (draw it)

- Intuitively, a real-life run of quicksort will produce a mix of "bad" and "good" splits
  - Randomly distributed among the recursion tree
  - Pretend for intuition that they alternate between best-case (n/2 : n/2) and worst-case (n-1 : 1)
  - What happens if we bad-split root node, then goodsplit the resulting size (n-1) node?
    - We end up with three subarrays, size 1, (n-1)/2, (n-1)/2
    - Combined cost of splits = n + n 1 = 2n 1 = O(n)
    - No worse than if we had good-split the root node!



- Intuitively, the O(n) cost of a bad split (or 2 or 3 bad splits) can be absorbed into the O(n) cost of each good split
- Thus running time of alternating bad and good splits is still O(n lg n), with slightly higher constants
- How can we be more rigorous?



- For simplicity, assume:
  - All inputs distinct (no repeats)
  - Slightly different partition () procedure
    - partition around a random element, which is not included in subarrays
    - all splits (0:n-1, 1:n-2, 2:n-3, ..., n-1:0) equally likely
- What is the probability of a particular split happening?
- Answer: 1/n

- So partition generates splits

   (0:n-1, 1:n-2, 2:n-3, ..., n-2:1, n-1:0)
   each with probability 1/n
- If T(n) is the expected running time,

$$T(n) = \frac{1}{n} \sum_{k=0}^{n-1} [T(k) + T(n-1-k)] + \Theta(n)$$

- What is each term under the summation for?
- What is the  $\Theta(n)$  term for?



• So...  $T(n) = \frac{1}{n} \sum_{k=0}^{n-1} [T(k) + T(n-1-k)] + \Theta(n)$ 

$$= \frac{2}{n} \sum_{k=0}^{n-1} T(k) + \Theta(n)$$
 Write it on the board



- We can solve this recurrence using the dreaded substitution method
  - Guess the answer
  - Assume that the inductive hypothesis holds
  - Substitute it in for some value < n</li>
  - Prove that it follows for n



- We can solve this recurrence using the dreaded substitution method
  - Guess the answer
    - What's the answer?
  - Assume that the inductive hypothesis holds
  - Substitute it in for some value < n</li>
  - Prove that it follows for n



- We can solve this recurrence using the dreaded substitution method
  - Guess the answer
    - $T(n) = O(n \lg n)$
  - Assume that the inductive hypothesis holds
  - Substitute it in for some value < n</li>
  - Prove that it follows for n



- We can solve this recurrence using the dreaded substitution method
  - Guess the answer
    - $T(n) = O(n \lg n)$
  - Assume that the inductive hypothesis holds
    - What's the inductive hypothesis?
  - Substitute it in for some value < n</li>
  - Prove that it follows for n



- We can solve this recurrence using the dreaded substitution method
  - Guess the answer
    - $T(n) = O(n \lg n)$
  - Assume that the inductive hypothesis holds
    - $T(n) \le an \lg n + b$  for some constants a and b
  - Substitute it in for some value < n</li>
  - Prove that it follows for n



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  - Guess the answer
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  - Substitute it in for some value < n</li>
    - What value?
  - Prove that it follows for n

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  - Guess the answer
    - $T(n) = O(n \lg n)$
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    - $T(n) \le an \lg n + b$  for some constants a and b
  - Substitute it in for some value < n</li>
    - The value k in the recurrence
  - Prove that it follows for n

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  - Guess the answer
    - $T(n) = O(n \lg n)$
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    - $T(n) \le an \lg n + b$  for some constants a and b
  - Substitute it in for some value < n</li>
    - The value k in the recurrence
  - Prove that it follows for n
    - Grind through it...

$$T(n) = \frac{2}{n} \sum_{k=0}^{n-1} T(k) + \Theta(n)$$

The recurrence to be solved

$$\leq \frac{2}{n} \sum_{k=0}^{n-1} (ak \lg k + b) + \Theta(n)$$

Plug in inductive hypothesis

$$\leq \frac{2}{n} \left[ b + \sum_{k=1}^{n-1} \left( ak \lg k + b \right) \right] + \Theta(n)$$
 Expand out the k=0 case

$$= \frac{2}{n} \sum_{k=1}^{n-1} (ak \lg k + b) + \frac{2b}{n} + \Theta(n)$$
 2b/n is just a constant, so fold it into  $\Theta(n)$ 

$$= \frac{2}{n} \sum_{k=1}^{n-1} (ak \lg k + b) + \Theta(n)$$

Note: leaving the same recurrence as the book

$$T(n) = \frac{2}{n} \sum_{k=1}^{n-1} (ak \lg k + b) + \Theta(n)$$

$$= \frac{2}{n} \sum_{k=1}^{n-1} ak \lg k + \frac{2}{n} \sum_{k=1}^{n-1} b + \Theta(n)$$
Distribute the summation
$$= \frac{2a}{n} \sum_{k=1}^{n-1} k \lg k + \frac{2b}{n} (n-1) + \Theta(n)$$
Evaluate the summation:
$$b + b + \dots + b = b (n-1)$$

$$\leq \frac{2a}{n} \sum_{k=1}^{n-1} k \lg k + 2b + \Theta(n)$$
Since  $n-1 < n$ ,  $2b(n-1)/n < 2b$ 

This summation gets its own set of slides later

$$T(n) \le \frac{2a}{n} \sum_{k=1}^{n-1} k \lg k + 2b + \Theta(n)$$

The recurrence to be solved

$$\leq \frac{2a}{n} \left( \frac{1}{2} n^2 \lg n - \frac{1}{8} n^2 \right) + 2b + \Theta(n)$$
 We'll prove this later

$$= an \lg n - \frac{a}{4}n + 2b + \Theta(n)$$

Distribute the (2a/n) term

$$= an \lg n + b + \left(\Theta(n) + b - \frac{a}{4}n\right) \quad \text{Remember, our goal is to get}$$

$$T(n) \leq an \lg n + b$$

 $\leq an \lg n + b$ 

Pick a large enough that an/4 dominates  $\Theta(n)+b$  30

- So  $T(n) \le an \lg n + b$  for certain a and b
  - Thus the induction holds
  - Thus  $T(n) = O(n \lg n)$
  - Thus quicksort runs in O(n lg n) time on average (phew!)
- Oh yeah, the summation...



$$\sum_{k=1}^{n-1} k \lg k = \sum_{k=1}^{\lceil n/2 \rceil - 1} k \lg k + \sum_{k=\lceil n/2 \rceil}^{n-1} k \lg k$$

$$\leq \sum_{k=1}^{\lceil n/2 \rceil - 1} k \lg k + \sum_{k=\lceil n/2 \rceil}^{n-1} k \lg n$$

The lg k in the second term is bounded by lg n

$$= \sum_{k=1}^{\lceil n/2 \rceil - 1} k \lg k + \lg n \sum_{k=\lceil n/2 \rceil}^{n-1} k$$

Move the lg n outside the summation



$$\sum_{k=1}^{n-1} k \lg k \le \sum_{k=1}^{\lceil n/2 \rceil - 1} k \lg k + \lg n \sum_{k=\lceil n/2 \rceil}^{n-1} k$$

The summation bound so far

$$\leq \sum_{k=1}^{\lceil n/2 \rceil - 1} k \lg(n/2) + \lg n \sum_{k=\lceil n/2 \rceil}^{n-1} k$$
 The  $\lg k$  in the first bounded by  $\lg n/2$ 

The lg k in the first term is

$$= \sum_{k=1}^{\lceil n/2 \rceil - 1} k (\lg n - 1) + \lg n \sum_{k=\lceil n/2 \rceil}^{n-1} k$$

$$\lg n/2 = \lg n - 1$$

$$= \left(\lg n - 1\right)^{\left\lceil n/2\right\rceil - 1} \sum_{k=1}^{n-1} k + \lg n \sum_{k=\left\lceil n/2\right\rceil}^{n-1} k \quad \begin{array}{l} \textit{Move (lg n - 1) outside the} \\ \textit{summation} \end{array}$$



$$\sum_{k=1}^{n-1} k \lg k \le \left(\lg n - 1\right)^{\left\lceil n/2\right\rceil - 1} \sum_{k=1}^{n-1} k + \lg n \sum_{k=\left\lceil n/2\right\rceil}^{n-1}$$
 The summation bound so far

$$=\lg n\sum_{k=1}^{\lceil n/2\rceil-1}k-\sum_{k=1}^{\lceil n/2\rceil-1}k+\lg n\sum_{k=\lceil n/2\rceil}^{n-1}k$$
 Distribute the (lg n - 1)

$$= \lg n \sum_{k=1}^{n-1} k - \sum_{k=1}^{\lceil n/2 \rceil - 1} k$$

$$= \lg n \left( \frac{(n-1)(n)}{2} \right) - \sum_{k=1}^{\lceil n/2 \rceil - 1} k$$

The summations overlap in range; combine them

The Guassian series



$$\sum_{k=1}^{n-1} k \lg k \le \left(\frac{(n-1)(n)}{2}\right) \lg n - \sum_{k=1}^{\lceil n/2 \rceil - 1} k$$

The summation bound so far

$$\leq \frac{1}{2} [n(n-1)] \lg n - \sum_{k=1}^{n/2-1} k$$

Rearrange first term, place upper bound on second

$$\leq \frac{1}{2} \left[ n(n-1) \right] \lg n - \frac{1}{2} \left( \frac{n}{2} \right) \left( \frac{n}{2} - 1 \right)$$
 X Guassian series

$$\leq \frac{1}{2} \left( n^2 \lg n - n \lg n \right) - \frac{1}{8} n^2 + \frac{n}{4} \qquad \text{Multiply it all out}$$



$$\sum_{k=1}^{n-1} k \lg k \le \frac{1}{2} \left( n^2 \lg n - n \lg n \right) - \frac{1}{8} n^2 + \frac{n}{4}$$

$$\le \frac{1}{2} n^2 \lg n - \frac{1}{8} n^2 \text{ when } n \ge 2$$

Done!!!



#### Quicksort in practice

- Quicksort is great general-purpose sorting algorithm.
- Quicksort is typically over twice as fast as merge sort.
- Quicksort can benefit substantially from code tuning.
- Quicksort behaves well even with caching and virtual memory.

