

CH08-320201

Algorithms and Data Structures

ADS

Lecture 9

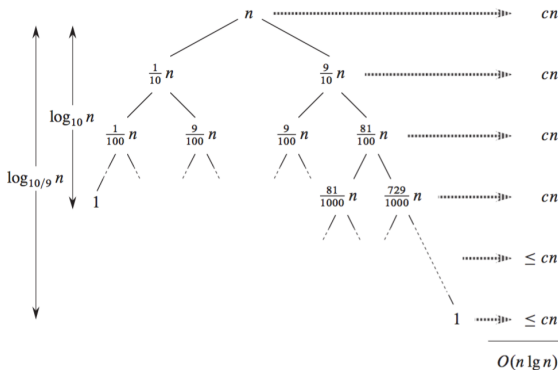
Dr. Kinga Lipskoch

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Runtime Analysis (5)

What if the split is $1/10 : 9/10$?

$$T(n) = T\left(\frac{1}{10}n\right) + T\left(\frac{9}{10}n\right) + \Theta(n)$$



Runtime Analysis

- ▶ What if we alternate between lucky and unlucky choices
 - ▶ $L(n) = 2U(n/2) + \Theta(n)$ lucky
 - ▶ $U(n) = L(n-1) + \Theta(n)$ unlucky
- ▶ Solving:
 - ▶
$$\begin{aligned} L(n) &= 2(L(n/2 - 1) + \Theta(n/2)) + \Theta(n) \\ &= 2L(n/2 - 1) + \Theta(n) \\ &= \Theta(n \lg n) \end{aligned}$$
- ▶ How can we make sure that this is usually happening?

Randomized Quicksort (1)

- ▶ Idea: Partition around a **random** element.
- ▶ Running time is independent of the input order.
- ▶ No assumptions need to be made about the input distribution.
- ▶ No specific input elicits the worst-case behavior.
- ▶ The worst case is determined only by the output of a random-number generator.

Randomized Quicksort (2)

RANDOMIZED-PARTITION(A, p, r)

```
1   $i = \text{RANDOM}(p, r)$   
2  exchange  $A[p]$  with  $A[i]$   
3  return PARTITION( $A, p, r$ )
```

RANDOMIZED-QUICKSORT(A, p, r)

```
1  if  $p < r$   
2       $q = \text{RANDOMIZED-PARTITION}(A, p, r)$   
3      RANDOMIZED-QUICKSORT( $A, p, q - 1$ )  
4      RANDOMIZED-QUICKSORT( $A, q + 1, r$ )
```

Randomized Quicksort (3)

- ▶ Let $T(n)$ be the random variable for the running time of the randomized quicksort on an input of size n (assuming random numbers are independent).

$$X_k = \begin{cases} 1 & \text{if PARTITION generates a } k:n-k-1 \text{ split,} \\ 0 & \text{otherwise.} \end{cases}$$

- ▶ For $k = 0, 1, \dots, n-1$, define indicator random variable
- ▶ $E[X_k] = \Pr\{X_k = 1\} = 1/n$, since all splits are equally likely (assuming elements are distinct).

Randomized Quicksort (4)

Recurrence:

$$T(n) = \begin{cases} T(0) + T(n-1) + \Theta(n) & \text{if } 0 : n-1 \text{ split,} \\ T(1) + T(n-2) + \Theta(n) & \text{if } 1 : n-2 \text{ split,} \\ \vdots & \\ T(n-1) + T(0) + \Theta(n) & \text{if } n-1 : 0 \text{ split,} \end{cases}$$

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

Randomized Quicksort (5)

Calculating expectations:

$$\begin{aligned} E[T(n)] &= E\left[\sum_{k=0}^{n-1} X_k (T(k) + T(n-k-1) + \Theta(n))\right] \\ &= \sum_{k=0}^{n-1} E[X_k (T(k) + T(n-k-1) + \Theta(n))] \\ &= \sum_{k=0}^{n-1} E[X_k] \cdot E[T(k) + T(n-k-1) + \Theta(n)] \\ &= \frac{1}{n} \sum_{k=0}^{n-1} E[T(k)] + \frac{1}{n} \sum_{k=0}^{n-1} E[T(n-k-1)] + \frac{1}{n} \sum_{k=0}^{n-1} \Theta(n) \\ &= \frac{2}{n} \sum_{k=2}^{n-1} E[T(k)] + \Theta(n) \end{aligned}$$

Randomized Quicksort (6)

- ▶ Use substitution method to solve recurrence.
- ▶ Guess: $E[T(n)] = \Theta(n \lg n)$.
- ▶ Prove: $E[T(n)] \leq an \lg n$ for constant $a > 0$.
- ▶ Use:

$$\sum_{k=2}^{n-1} k \lg k \leq \frac{1}{2} n^2 \lg n - \frac{1}{8} n^2$$

(proof by induction)

Randomized Quicksort (7)

Proof:

$$\begin{aligned}E[T(n)] &\leq \frac{2}{n} \sum_{k=2}^{n-1} ak \lg k + \Theta(n) \\E[T(n)] &\leq \frac{2}{n} \sum_{k=2}^{n-1} ak \lg k + \Theta(n) \\&= \frac{2a}{n} \left(\frac{1}{2} n^2 \lg n - \frac{1}{8} n^2 \right) + \Theta(n) \\&= an \lg n - \left(\frac{an}{4} - \Theta(n) \right) \\&\leq an \lg n,\end{aligned}$$

if a is chosen large enough.

Quicksort: Conclusion

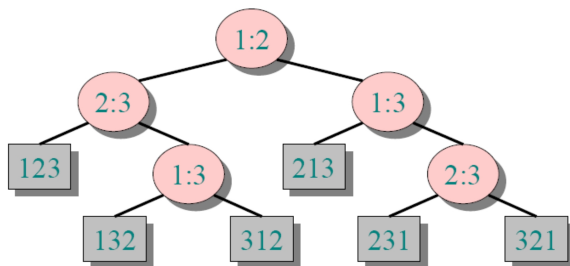
- ▶ Quicksort is a great general-purpose sorting algorithm.
- ▶ Quicksort is often the best practical choice because its expected runtime is $\Theta(n \lg n)$ and the constant is quite small.
- ▶ Quicksort is typically over twice as fast as MergeSort.
- ▶ Quicksort is an in-situ sorting algorithm (debatable).
- ▶ Quicksort has a worst-case runtime of $\Theta(n^2)$ when the array is already sorted.
- ▶ Visualization Randomized Quicksort:
<http://www.sorting-algorithms.com/quick-sort>

Comparison Sorts

- ▶ All sorting algorithms we have seen so far are **comparison sorts**.
- ▶ A comparison sort only uses comparisons to determine the relative order of elements.
- ▶ The best worst-case running time we encountered for comparison sorting was $O(n \lg n)$.
- ▶ Is $O(n \lg n)$ the best we can do?

Decision Tree (1)

- ▶ Sort $\langle a_1, a_2, \dots, a_n \rangle$
- ▶ Each internal node is labeled $i : j$ for $i, j \in \{1, 2, \dots, n\}$.
- ▶ Left subtree shows subsequent comparisons if $a_i \leq a_j$.
- ▶ Right subtree shows subsequent comparisons if $a_i \geq a_j$.

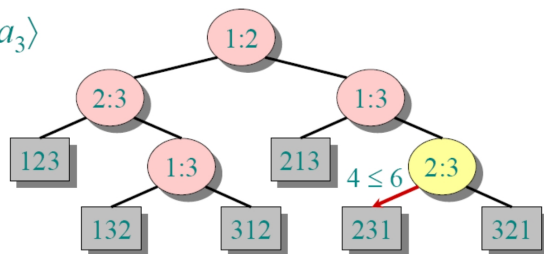


Decision Tree (2)

Example:

Each leaf contains a permutation $\langle \pi(1), \pi(2), \dots, \pi(n) \rangle$ indicating the order $a_{\pi(1)} \leq a_{\pi(2)} \leq \dots \leq a_{\pi(n)}$.

Sort $\langle a_1, a_2, a_3 \rangle$
 $= \langle 9, 4, 6 \rangle$:



Decision Tree Model

A decision tree can model the execution of any comparison sort:

- ▶ One tree for each input size n .
- ▶ View the algorithm as splitting whenever it compares two elements.
- ▶ The tree contains the comparisons along all possible instruction traces.
- ▶ The running time of the algorithm = the length of the path taken.
- ▶ Worst-case running time = height of tree.

Decision Tree Sorting

Theorem:

Any decision tree that can sort n elements must have height $\Omega(n \lg n)$.

Proof:

The tree must contain $\geq n!$ leaves,
since there are $n!$ possible permutations.

A height- h binary tree has $\leq 2^h$ leaves.

Thus, $n! \leq 2^h$.

$$\begin{aligned}\text{Then, } h &\geq \lg(n!) \\ &\geq \lg((n/e)^n) \\ &= n \lg n - n \lg e \\ &= \Omega(n \lg n).\end{aligned}$$

Used Stirling's formula: $n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$ when $n \rightarrow \infty$.

Lower Bound for Comparison Sorting

- ▶ The lower bound for comparison sorting $\Omega(n \lg n)$.
- ▶ Heap Sort and Merge Sort are asymptotically optimal comparison sorting algorithms.

Non-Comparison Sorting?

- ▶ Is it possible to avoid comparisons between elements?
- ▶ Yes, if we can make assumptions on the input data.
- ▶ E.g., trivial case:
 - ▶ **Input:** $A[1...n]$, where $A[j] \in \{1, 2, \dots, n\}$, and $A[i] \neq A[j]$ for all $i \neq j$
 - ▶ **Output:** $B[1...n]$