



Chapter 7 Randomization

Algorithm Theory WS 2017/18

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Randomized Quicksort



Quicksort:

S $S_{\ell} < v$ **function** Quick (*S*: sequence): sequence; {returns the sorted sequence *S*} begin if $\#S \leq 1$ then return S **else** { choose pivot element v in S; partition S into S_{ℓ} with elements < v, and S_r with elements > v**return** Quick(S_{ℓ}) v Quick(S_r) end;



Randomized Quicksort: pick uniform random element as pivot

Running Time of sorting n elements:

- Let's just count the number of comparisons
- In the partitioning step, all n-1 non-pivot elements have to be compared to the pivot
- Number of comparisons:

n-1 + #comparisons in recursive calls

• If rank of pivot is r:
recursive calls with r-1 and n-r elements

Law of Total Expectation



- Given a random variable X and
- a set of events A_1, \dots, A_k that partition Ω
 - E.g., for a second random variable Y, we could have $A_i \coloneqq \{\omega \in \Omega : Y(\omega) = i\}$

Law of Total Expectation

$$\mathbb{E}[X] = \sum_{i=1}^{\kappa} \mathbb{P}(A_i) \cdot \mathbb{E}[X \mid A_i] = \sum_{y} \mathbb{P}(Y = y) \cdot \mathbb{E}[X \mid Y = y]$$

Example:

- X: outcome of rolling a die
- $A_0 = \{X \text{ is even}\}, A_1 = \{X \text{ is odd}\}$



Random variables:

- C: total number of comparisons (for a given array of length n)
- R: rank of first pivot
- C_{ℓ} , C_r : number of comparisons for the 2 recursive calls

$$\mathbb{E}[C] = n - 1 + \mathbb{E}[C_{\ell}] + \mathbb{E}[C_r]$$

Law of Total Expectation:

$$\mathbb{E}[C] = \sum_{\substack{r=1\\n}}^{n} \mathbb{P}(R=r) \cdot \mathbb{E}[C|R=r]$$

$$= \sum_{r=1}^{n} \mathbb{P}(R=r) \cdot (n-1+\mathbb{E}[C_{\ell}|R=r] + \mathbb{E}[C_{r}|R=r])$$



We have seen that:

$$\mathbb{E}[C] = \sum_{r=1}^{n} \mathbb{P}(R=r) \cdot (n-1+\mathbb{E}[C_{\ell}|R=r] + \mathbb{E}[C_r|R=r])$$

Define:

• T(n): expected number of comparisons when sorting n elements

$$\mathbb{E}[C] = T(n)$$

$$\mathbb{E}[C_{\ell}|R = r] = T(r - 1)$$

$$\mathbb{E}[C_r|R = r] = T(n - r)$$

Recursion:

$$T(n) = \sum_{r=1}^{n} \frac{1}{n} \cdot (n-1+T(r-1)+T(n-r))$$

$$T(0) = T(1) = 0$$



Theorem: The expected number of comparisons when sorting n elements using randomized quicksort is $T(n) \le 2n \ln n$.

$$T(n) = \sum_{r=1}^{n} \frac{1}{n} \cdot (n-1+T(r-1)+T(n-r)), \qquad T(0) = 0$$



Theorem: The expected number of comparisons when sorting n elements using randomized quicksort is $T(n) \le 2n \ln n$.

$$T(n) \le n - 1 + \frac{4}{n} \cdot \int_{1}^{n} x \ln x \, dx$$

$$\int x \ln x \, dx = \frac{x^2 \ln x}{2} - \frac{x^2}{4}$$

Alternative Analysis



Array to sort: [7,3,1,10,14,8,12,9,4,6,5,15,2,13,11]

Viewing quicksort run as a tree:

Comparisons



- Comparisons are only between pivot and non-pivot elements
- Every element can only be the pivot once:
 - → every 2 elements can only be compared once!
- W.l.o.g., assume that the elements to sort are 1, 2, ..., n
- Elements i and j are compared if and only if either i or j is a pivot before any element h: i < h < j is chosen as pivot
 - i.e., iff i is an ancestor of j or j is an ancestor of i

$$\mathbb{P}(\text{comparison betw. } i \text{ and } j) = \frac{2}{j-i+1}$$

Counting Comparisons



Random variable for every pair of elements (i, j):

$$\mathbf{X}_{ij} = \begin{cases} 1, & \text{if there is a comparison between } i \text{ and } j \\ 0, & \text{otherwise} \end{cases}$$

Number of comparisons: X

$$X = \sum_{i < j} X_{ij}$$

• What is $\mathbb{E}[X]$?



Theorem: The expected number of comparisons when sorting n elements using randomized quicksort is $T(n) \le 2n \ln n$.

Proof:

Linearity of expectation:

For all random variables $X_1, ..., X_n$ and all $a_1, ..., a_n \in \mathbb{R}$,

$$\mathbb{E}\left[\sum_{i}^{n} a_{i} X_{i}\right] = \sum_{i}^{n} a_{i} \mathbb{E}[X_{i}].$$



Theorem: The expected number of comparisons when sorting n elements using randomized quicksort is $T(n) \le 2n \ln n$.

$$\mathbb{E}[X] = 2\sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \frac{1}{j-i+1} = 2\sum_{i=1}^{n-1} \sum_{k=2}^{n-i+1} \frac{1}{k}$$

Quicksort: High Probability Bound



- We have seen that the number of comparisons of randomized quicksort is $O(n \log n)$ in expectation.
- Can we also show that the number of comparisons is $O(n \log n)$ with high probability?

Recall:

On each recursion level, each pivot is compared once with each other element that is still in the same "part"

Counting Number of Comparisons



- We looked at 2 ways to count the number of comparisons
 - recursive characterization of the expected number
 - number of different pairs of values that are compared

Let's consider yet another way:

- Each comparison is between a pivot and a non-pivot
- How many times is a specific array element x compared as a non-pivot?

Value x is compared as a non-pivot to a pivot once in every recursion level until one of the following two conditions apply:

- 1. x is chosen as a pivot
- 2. x is alone

Successful Recursion Level



- Consider a specific recursion level ℓ
- Assume that at the beginning of recursion level ℓ , element x is in a sub-array of length K_{ℓ} that still needs to be sorted.
- If x has been chosen as a pivot before level ℓ , we set $K_{\ell}\coloneqq 1$

Definition: We say that recursion level ℓ is successful for element x iff the following is true:

$$K_{\ell+1} = 1$$
 or $K_{\ell+1} \le \frac{2}{3} \cdot K_{\ell}$

Successful Recursion Level



Lemma: For every recursion level ℓ and every array element x, it holds that level ℓ is successful for x with probability at least $^1/_3$, independently of what happens in other recursion levels.

Number of Successful Recursion Levels



Lemma: If among the first ℓ recursion levels, at least $\log_{\frac{3}{2}}(n)$ are successful for element x, we have $K_{\ell} = 1$.

Chernoff Bounds



- Let $X_1, ..., X_n$ be independent 0-1 random variables and define $p_i \coloneqq \mathbb{P}(X_i = 1)$.
- Consider the random variable $X = \sum_{i=1}^{n} X_i$
- We have $\mu \coloneqq \mathbb{E}[X] = \sum_{i=1}^n \mathbb{E}[X_i] = \sum_{i=1}^n p_i$

Chernoff Bound (Lower Tail):

$$\forall \delta > 0$$
: $\mathbb{P}(X < (1 - \delta)\mu) < e^{-\delta^2 \mu/2}$

Chernoff Bound (Upper Tail):

$$\forall \delta > 0 \colon \mathbb{P}(X > (1+\delta)\mu) < \left(\frac{e^{\delta}}{(1+\delta)^{1+\delta}}\right)^{\mu} < e^{-\delta^{2}\mu/3}$$
holds for $\delta \leq 1$

Chernoff Bounds, Example



Assume that a fair coin is flipped n times. What is the probability to have

1. less than n/3 heads?

2. more than 0.51n tails?

3. less than $n/2 - \sqrt{c \cdot n \ln n}$ tails?



- Independent Bernoulli random variables $X_1, X_2, ..., X_n$
- $\mathbb{P}(X_i = 1) \geq p_i, X \coloneqq \sum_{i=1}^n X_i, \mu \coloneqq \sum_{i=1}^n p_i \geq \mathbb{E}[X]$

Chernoff Lower Tail: $\mathbb{P}(X < (1 - \delta)\mu) < e^{-\delta^2\mu/2}$

Recall

• Markov Inequality: Given non-negative rand. var. $Z \ge 0$

$$\forall t > 0 \colon \mathbb{P}(Z > t) < \frac{\mathbb{E}[Z]}{t}$$

Independent random variables Y, Z:

$$\mathbb{E}[Y \cdot Z] = \mathbb{E}[Y] \cdot \mathbb{E}[Z]$$



• $\mathbb{P}(X_i = 1) \ge p_i, X \coloneqq \sum_{i=1}^n X_i, \mu \coloneqq \sum_{i=1}^n p_i \ge \mathbb{E}[X]$

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Number of Comparisons for *x*



Lemma: For every array element x, with high probability, as a non-pivot, x is compared to a pivot at most $O(\log n)$ times.

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Number of Comparisons



Theorem: With high probability, the total number of comparisons is at most $O(n \log n)$.

Types of Randomized Algorithms



Las Vegas Algorithm:

- always a correct solution
- running time is a random variable
- Example: randomized quicksort, contention resolution

Monte Carlo Algorithm:

- probabilistic correctness guarantee (mostly correct)
- fixed (deterministic) running time
- **Example:** primality test