

CS 630 – Fall 2024 – Lab 9
Nov 13, 2024

Problem 1 Consider the *randomized quicksort* algorithm applied to an array $A[1 \dots n]$ containing n distinct elements. In this version of quicksort, the pivot is chosen uniformly at random from the subarray being sorted at each recursive call.

Define the indicator random variable X_{ij} for each pair of indices $1 \leq i < j \leq n$ as follows:

$$X_{ij} = \begin{cases} 1, & \text{if elements } A_i \text{ and } A_j \text{ are compared during the execution of the algorithm,} \\ 0, & \text{otherwise.} \end{cases}$$

Prove that for any pair of indices $i < j$, the probability that A_i and A_j are compared during the execution of the algorithm is:

$$\mathbb{P}(X_{ij} = 1) = \frac{2}{j - i + 1}.$$

Using the result from part (a), compute the expected total number of comparisons made by the randomized quicksort algorithm.

Solution:

(a) **Proof:**

Consider the set of elements between positions i and j inclusive:

$$S = \{A_i, A_{i+1}, \dots, A_j\}.$$

In randomized quicksort, elements A_i and A_j are compared if and only if one of them is chosen as a pivot before any other element in S is chosen as a pivot.

Each element in S has an equal chance of being the first pivot selected from S . There are $|S| = j - i + 1$ elements in S .

The probability that either A_i or A_j is the first pivot chosen from S is:

$$\mathbb{P}(X_{ij} = 1) = \frac{2}{|S|} = \frac{2}{j - i + 1}.$$

Therefore, the probability that A_i and A_j are compared is $\frac{2}{j - i + 1}$.

(b) **Calculation of Expected Total Number of Comparisons:**

The expected total number of comparisons $\mathbb{E}[C]$ is the sum of the probabilities that each pair of elements is compared:

$$\mathbb{E}[C] = \sum_{i=1}^{n-1} \sum_{j=i+1}^n \mathbb{P}(X_{ij} = 1) = \sum_{i=1}^{n-1} \sum_{j=i+1}^n \frac{2}{j-i+1}.$$

Let $k = j - i$, so k ranges from 1 to $n - i$. Rewrite the sum:

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

Swap the order of summation:

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

Simplify the expression:

$$\mathbb{E}[C] = 2 \sum_{k=1}^{n-1} \left(\frac{n}{k+1} - 1 \right).$$

Separate the sums:

$$\mathbb{E}[C] = 2n \sum_{k=1}^{n-1} \frac{1}{k+1} - 2(n-1).$$

Recognize that $\sum_{k=1}^{n-1} \frac{1}{k+1} = H_n - 1$, where H_n is the n -th harmonic number.

Therefore:

$$\mathbb{E}[C] = 2n(H_n - 1) - 2(n-1) = 2nH_n - 2n - 2n + 2 = 2nH_n - 4n + 2.$$

Simplify:

$$\mathbb{E}[C] = 2nH_n - 4n + 2.$$

Alternatively, since H_n grows approximately like $\ln n + \gamma$ (where γ is the Euler-Mascheroni constant), for large n , the expected number of comparisons is approximately $2n \ln n$.

Problem 2 *k*-median

Given n numbers, the k -median is the k th smallest amongst them. (We can assume that numbers are distinct) In this problem you will design an algorithm to find it.

Design an algorithm that takes an unordered length- n array A of numbers, and an integer k as input and returns the k th lowest value in A .

1. design a deterministic (= non-random) algorithm for this that finds the k -median in time $O(n \log n)$.

Solution: The deterministic algorithm is very simple: use a sorting algorithm with runtime $O(n \log n)$ to sort A , e.g. MergeSort and BinarySearchInsertion work well for this. Then return the k th value in the sorted order.

2. design a QuickSort-style randomized algorithm for this same problem.

As it turns out the expected runtime of this is in fact $O(n)$ (linear!) and not $O(n \log n)$ as for QuickSort. For those interested, you can find the analysis of this in the lecture slides from 11/12. (We didn't cover this, it's only for your interest.)

Solution: We run a modified version of QuickSort: In each iteration pick a random pivot, compute the two subarrays of items below and above the pivot. Then recurse on the half that contains the k th lowest element.

Here is how to find which subarray to use: since the QuickSort algorithm in class does the sorting in place, by the end we know that in the sorted order the k -median resides in $A[k]$. Hence, when we make our recursive call around the pivot index q , if $q > k$ then recurse on the left subarray. Otherwise recurse on the right.

The description above is enough, but here is the pseudocode for your convenience:

Algorithm 1: Rnd-kMedian(A , p , r , k)

```
1 if  $p < r$  then
2    $q \leftarrow$  RND-PARTITION( $A, p, r$ )/* identical to partition function */
   from class
3   if  $q > k$  then
4     Rnd-kMedian( $A, p, q-1$ );
5   else
6     Rnd-kMedian( $A, q+1, r$ );
```