



Design and Analysis
of Algorithms I

QuickSort

Analysis I: A Decomposition Principle

Necessary Background

Assumption: you know and remember (finite) sample spaces, random variables, expectation, linearity of expectation. For review:

- Probability Review I (video)
- Lehman-Leighton notes (free PDF)
- Wikibook on Discrete Probability

Average Running Time of QuickSort

QuickSort Theorem: for every input array of length n ,
the average running time of QuickSort (with random pivots)
is $O(n \log n)$.

Note: holds for every input. [no assumptions on the data]

- recall our guiding principles!
- "average" is over random choices made by the algorithm (i.e., pivot choices)

Preliminaries

Fix input array A of length n .

Sample space Ω = all possible outcomes of random choices in QuickSort (i.e., pivot sequences).

Key random variable: for $\sigma \in \Omega$,

$C(\sigma)$ = # of comparisons between two input elements made by QuickSort (given random choices σ).

Lemma: running time of QuickSort dominated by comparisons.

Remaining goal: $E[C] = O(n \log n)$.

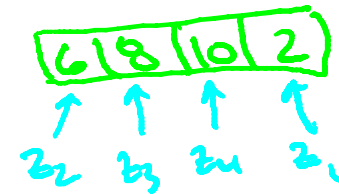
→ \exists constant c s.t. $\forall \sigma \in \Omega$
 $RT(\sigma) \leq c \cdot C(\sigma)$. (see notes)

Building Blocks

Note: Can't apply Master Method [random, unbalanced subproblems]

[A = fixed input array]

Notation: z_i = i^{th} smallest element of A.



For $\sigma \in \Sigma$, indices $i < j$, let

$X_{ij}(\sigma)$ = # of times z_i, z_j get compared
in QuickSort with pivot sequence σ .

Fix two elements of the input array. How many times can these two elements get compared with each other during the execution of QuickSort?

☐ 1

☒ 0 or 1

☐ 0, 1, or 2

☐ Any integer between 0 and $n - 1$

Reason: two elements compared only when one is the pivot, which is excluded from future recursive calls.

Thus: each X_{ij} is an "indicator"
(i.e., 0-1) random variable.

A Decomposition Approach

So: $C(\sigma) = \# \text{ of comparisons between input elements.}$

$X_{ij}(\sigma) = \# \text{ of comparisons between } z_i \text{ and } z_j$

Thus: $\forall \sigma, C(\sigma) = \sum_{i=1}^{n-1} \sum_{j=i+1}^n X_{ij}(\sigma)$ complicated simple!

By linearity of expectation: $E[C] = \sum_{i=1}^{n-1} \sum_{j=i+1}^n E[X_{ij}]$

Since $E[X_{ij}] = 0 \cdot P[X_{ij}=0] + 1 \cdot P[X_{ij}=1] = P[X_{ij}=1]$

Thus: $E[C] = \sum_{i=1}^{n-1} \sum_{j=i+1}^n P[z_i, z_j \text{ get compared}]$ (*) next video

A General Decomposition Principle

1. Identify random variable Y that you really care about.

2. Express Y as sum of indicator random variables:

$$Y = \sum_{e=1}^n X_e$$

3. Apply linearity of expectation:

$$E[Y] = \sum_{e=1}^n \text{Pr}[X_e=1]$$

"just" need to understand those!