

# Data Structures and Algorithms

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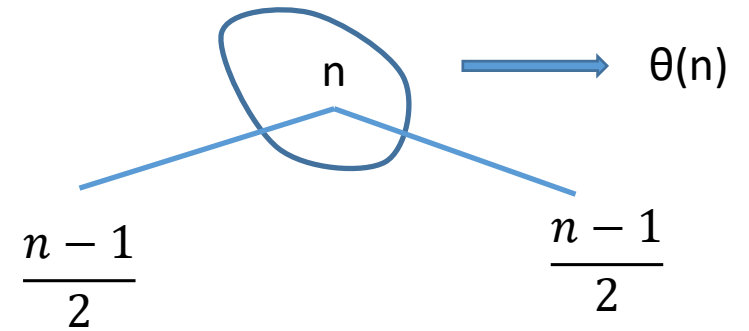
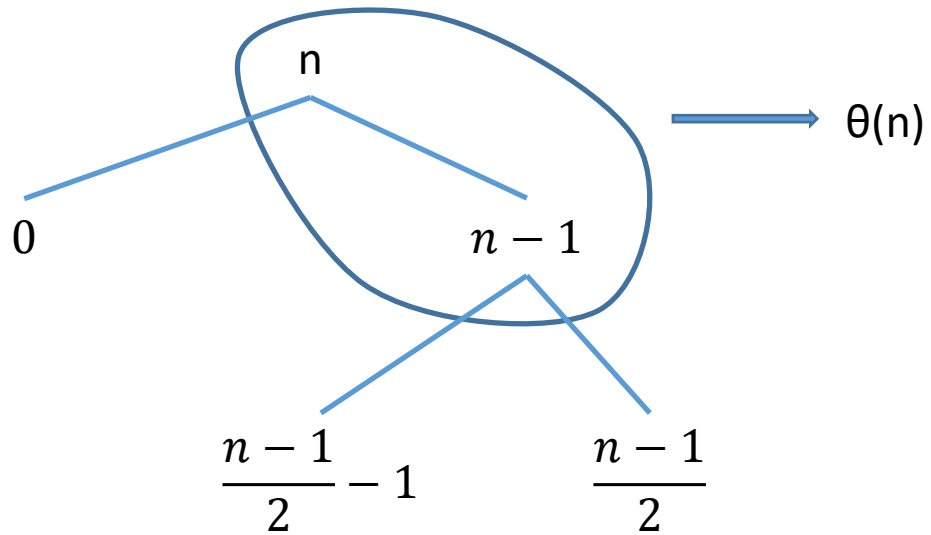
# Analysis of Quick Sort: average case

- The behaviour of quick sort depends upon the relative ordering of the elements in the input array
- All permutations of input numbers are equally likely (assumption)
- On a random input array, some of the splits will be reasonably balanced and some are fairly unbalanced
- In the average case, Partition procedure produces a mix of good (best-case splits) and bad splits ( worst-case splits)
- In a recursion tree for an average-case execution of Partition, the good and bad splits are distributed randomly throughout the tree

# Analysis of Quick Sort: average case

- Suppose that the good and bad splits alternate levels in the tree
- A bad split happens at the root which produces two sub arrays of sizes “0” and “n-1”
- At the next level a good split happens which produces two sub arrays of sizes “(n-1)/2-1” and “(n-1)/2”
- The combination of a bad split followed by a good split produces three sub arrays of sizes 0, (n-1)/2-1 and (n-1)/2
- The partitioning cost of these splits is:  $\theta(n) + \theta(n-1) = \theta(n)$

# Analysis of Quick Sort



# Randomized quick sort

- In a practical situation all permutations are not equally likely
- Can add randomization to an algorithm to obtain a good expected performance over all inputs
- The resulting algorithm is called randomized quick sort
- A randomization technique called “random sampling” is used
- Use a randomly chosen element from given subarray as the pivot instead of the rightmost element
- The input array is expected to get split into reasonably balanced sets on average

# Randomized quick sort

Randomized\_Partition(A, p, r)

$i \leftarrow \text{RANDOM}(p, r)$

    exchange  $A[r]$  with  $A[i]$

    return Partition(A, p, r)

Randomized\_QuickSort(A, p, r)

    if( $p < r$ )

$q \leftarrow \text{Randomized\_Partititon}(A, p, r)$

        Randomized\_QuickSort(A, p,  $q-1$ )

        Randomized\_QuickSort(A,  $q+1$ , r)

# Randomized quick sort: Worst-case

$$T(n) = \max_{0 \leq q \leq n-1} (T(q) + T(n - q - 1)) + \theta(n)$$

Use guess-and-test method:  $T(n) \leq cn^2$  for some constant  $c$  (hypothesis)

$$\begin{aligned} T(n) &\leq \max_{0 \leq q \leq n-1} (cq^2 + c(n-q-1)^2) + \theta(n) \\ &= c \max_{0 \leq q \leq n-1} (q^2 + (n-q-1)^2) + \theta(n) \end{aligned}$$

$(q^2 + (n-q-1)^2)$  achieves the maximum at both the end points of the range

$$\max_{0 \leq q \leq n-1} (q^2 + (n-q-1)^2) \leq (n-1)^2 = n^2 - 2n + 1$$

$$\begin{aligned} T(n) &\leq cn^2 - c(2n - 1) + \theta(n) \\ &\leq cn^2 \end{aligned}$$

# Randomized quick sort: Worst-case

$$T(n) = \max_{0 \leq q \leq n-1} (T(q) + T(n - q - 1)) + \theta(n)$$

Use guess-and-test method:  $T(n) \geq cn^2$  for some constant  $c$  (hypothesis)

$$\begin{aligned} T(n) &\geq \max_{0 \leq q \leq n-1} (cq^2 + c(n - q - 1)^2) + \theta(n) \\ &= c \max_{0 \leq q \leq n-1} (q^2 + (n-q-1)^2) + \theta(n) \end{aligned}$$

$(q^2 + (n-q-1)^2)$  achieves the maximum at both the end points of the range

$$\max_{0 \leq q \leq n-1} (q^2 + (n-q-1)^2) = (n-1)^2 = n^2 - 2n + 1$$

$$T(n) \geq cn^2 - c(2n - 1) + \theta(n)$$

$$\geq cn^2, \text{ where } c \text{ is chosen so that } \theta(n) \text{ dominates } c(2n - 1)$$



# A few basics

- Consider a sample space  $S$  and an event  $A$
- Indicator random variable associated with event  $A$  is denoted as  $I\{A\}$  and defined as:

- $I\{A\} = \begin{cases} 1 & \text{if } A \text{ occurs} \\ 0 & \text{if } A \text{ does not occur} \end{cases}$

# A few basics

- The expected value or the expectation of a discrete random variable is:

$$E[X] = \sum_x x \cdot \Pr\{X = x\}$$

- $I\{A\} = \begin{cases} 1 & \text{if } A \text{ occurs} \\ 0 & \text{if } A \text{ does not occur} \end{cases}$

- Linearity of expectation: The expectation of sum of two random variables is the sum of their expectations

$$E[X + Y] = E[X] + E[Y]$$

# A few basics

Experiment: flipping a fair coin

$S = \{H, T\}$ ;  $\Pr\{H\} = \frac{1}{2}$  and  $\Pr\{T\} = \frac{1}{2}$

H: the coin coming up with heads

$X_H$ : indicator random variable associated with H

$$X_H = I\{H\}$$

$$I\{H\} = \begin{cases} 1 & \text{if H occurs} \\ 0 & \text{if T occurs} \end{cases}$$

# A few basics

The expected number of heads in one flip of coin is the expected value of our indicator  $X_H$ :

$$\begin{aligned} E[X_H] &= E[I\{H\}] \\ &= 1 \cdot \Pr\{H\} + 0 \cdot \Pr\{T\} \\ &= 1 \cdot 1/2 + 0 \cdot 1/2 \\ &= 1/2 \end{aligned}$$

Lemma: Given a sample space  $S$  and an event  $A$  in  $S$ , let  $X_A = I\{A\}$ . Then

$$E[X_A] = \Pr\{A\}$$

# Quick sort: Analysis

Algorithm Partition(A, p, r)

$x \leftarrow A[r]$

$i \leftarrow p-1$

for  $j \leftarrow p$  to  $r-1$

    if( $A[j] \leq x$ )

$i \leftarrow i + 1$

        exchange  $A[i]$  and  $A[j]$

exchange  $A[i+1]$  and  $A[r]$

return  $i+1$

# Randomized quick sort: Expected running time

- Randomized quick sort works similar to quick sort except for pivot selection
- Analyse Randomized Quick\_Sort by discussing Quick\_Sort and Partition algorithms (randomly selected pivot)
- Consider an array of  $n$  distinct elements
- The running time of Quick\_Sort is dominated by the time spent in Partition procedure
- How many times an element is selected as pivot?
- Observation 1: An element selected as pivot never included in the future recursive calls to Quick\_Sort and Partition
- There can be at most  $n$  calls to Partition

# Randomized quick sort: Expected running time

- One call to Partition:
  - constant amount of time and
  - Amount of time proportional to number of iterations of the **for** loop
- In each iteration of **for** loop, the pivot is compared with an element in the array (pivot-array element comparison)
- What is the total time spent in the **for** loop over all calls to Partition procedure?
- By counting the number of times the pivot is compared to an array element, we can bound the total time spent in the **for** loop

# Randomized quick sort: Expected running time

**Lemma:** Let  $X$  be the number of pivot-array element comparisons performed over the entire execution of Quick\_Sort on an  $n$ -element array. Then the running time of Quick\_Sort is  $O(n + X)$

**Proof:** The algorithm makes at most  $n$  calls to Partition procedure

Each call does a constant amount of work and executes the **for** loop some number of times

Each iteration of **for** loop performs one pivot-array element comparison

We have to compute “ $X$ ”, the total number of comparisons performed over all calls to Partition



# Randomized quick sort: Expected running time

- Rename the elements in array  $A$  as  $z_1, z_2, \dots, z_n$ , where  $z_i$  is the  $i^{\text{th}}$  smallest element in  $A$
- $Z_{ij} = \{z_i, z_{i+1}, \dots, z_j\}$  be the set of elements between  $z_i$  and  $z_j$  inclusive
- Observation 2: each pair of elements are compared at most once, why?
- Define indicator random variable as:  
 $X_{ij} = I\{z_i \text{ is compared to } z_j\}$   
(during entire execution of the algorithm)
- Since each pair of elements is compared at most once,

$$X = \sum_{i=1}^{n-1} \sum_{j=i+1}^n X_{ij}$$

# Randomized quick sort: Expected running time

$$\begin{aligned} E[X] &= E\left[\sum_{i=1}^{n-1} \sum_{j=i+1}^n X_{ij}\right] \\ &= \sum_{i=1}^{n-1} \sum_{j=i+1}^n E[X_{ij}] \\ &= \sum_{i=1}^{n-1} \sum_{j=i+1}^n \Pr\{z_i \text{ is compared to } z_j\} \end{aligned}$$

- We have to compute the quantity,  $\Pr\{z_i \text{ is compared to } z_j\}$
- Consider an input array  $A = \{41, 11, 21, 51, 81, 61, 31, 71, 101, 91\}$
- Assume that the first call to Partition separates this array into two sets:  $\{41, 11, 21, 51, 31\}$  and  $\{81, 71, 101, 91\}$
- An element from either of these sets will ever be compared with the elements in the other set?