# Data Structures

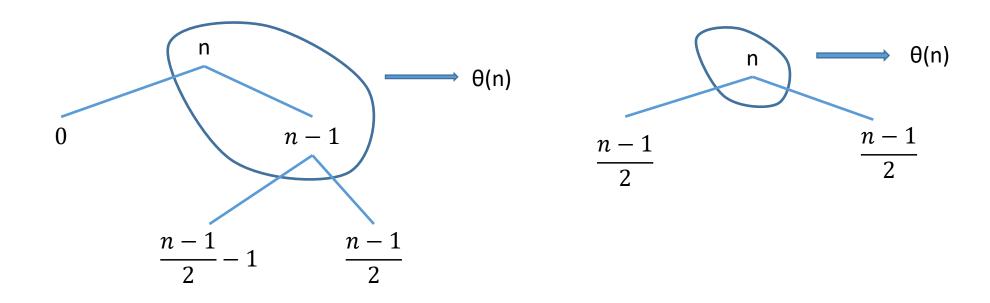
## Analysis of Quick Sort: average case

- 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 RANDOM(p, r)
      exchange A[r] with A[i]
      return Partition(A, p, r)
Randomized QuickSort(A, p, r)
      if(p < r)
             q \leftarrow Randomized Partitition(A, p, r)
             Randomized QuickSort(A, p, q-1)
             Randomized QuickSort(A, q+1, r)
```

## Randomized quick sort: Worst-case

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

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

$$T(n) \le \max_{0 \le q \le n-1} (cq^2 + c(n-q-1)^2) + \theta(n)$$
  
=  $c \max_{0 \le q \le n-1} (q^2 + (n-q-1)^2) + \theta(n)$ 

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

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

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

$$\le cn^2$$

## Randomized quick sort: Worst-case

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

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

$$T(n) \ge \max_{0 \le q \le n-1} (cq^2 + c(n-q-1)^2) + \theta(n)$$

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

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

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

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

 $\geq$  cn<sup>2</sup>, where c is chosen so that  $\theta$ (n) dominates c(2n – 1)

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

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

$$E[X] = \sum_{x} x. \Pr\{X = x\}$$

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

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

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

Experiment: flipping a fair coin

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

H: the coin coming up with heads

X<sub>H</sub>: indicator random variable associated with H

$$X_{H} = I\{H\}$$

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

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

```
E[X_H] = E[I\{H\}]
= 1.Pr{H} + 0. Pr{T}
= 1.1/2 + 0.1/2
=1/2
```

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 ← 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 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

- 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

**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

- Rename the elements in array A as z<sub>1</sub>, z<sub>2</sub>, . . . , z<sub>n</sub>, where z<sub>i</sub> is the i<sup>th</sup> smallest element in A
- $Z_{ij} = \{z_i, z_{i+1}, \ldots, z_j\}$  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,

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

$$\begin{split} \mathsf{E}[\mathsf{X}] &= \mathsf{E} \Big[ \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \mathsf{X}_{\mathsf{i}\mathsf{j}} \Big] \\ &= \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \mathsf{E}[\mathsf{X}_{\mathsf{i}\mathsf{j}}] \\ &= \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \mathsf{Pr} \{ z_{\mathsf{i}} \text{ is compared to } z_{\mathsf{j}} \} \end{split}$$

- We have to compute the quantity,  $Pr\{z_i \text{ is compared to } z_i\}$
- 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?