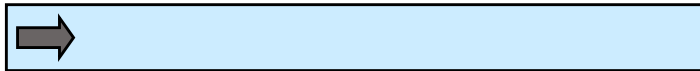


# RANDOMIZED ALGORITHMS

Module V

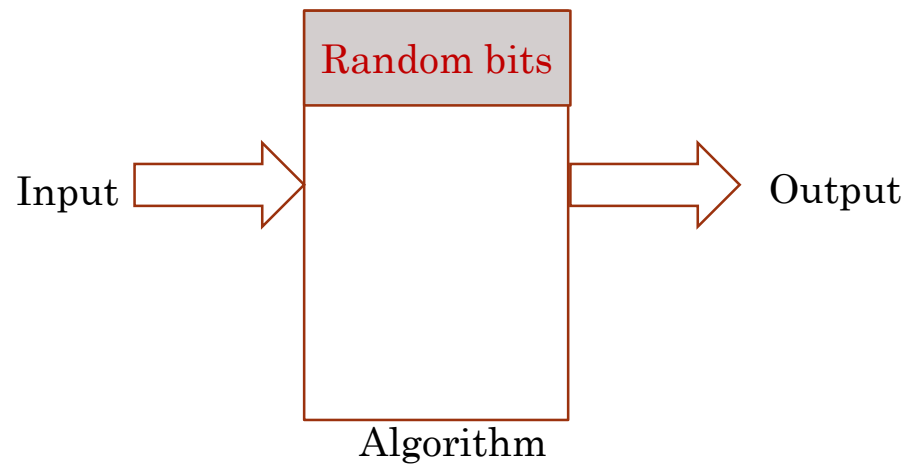
# A short list of categories

- Algorithm types include:
  - Simple recursive algorithms
  - Backtracking algorithms
  - Divide and conquer algorithms
  - Dynamic programming algorithms
  - Greedy algorithms
  - Branch and bound algorithms
  - Brute force algorithms
  - Randomized algorithms



- Also known as Monte Carlo algorithms or stochastic methods

# Randomized Algorithm



- The **output** or the **running time** are functions of the input and random bits chosen.

# Randomized algorithms

- A randomized algorithm is just one that depends on random numbers for its operation
- These are randomized algorithms:
  - Using random numbers to help to find a solution to a problem
  - Using random numbers to improve a solution to a problem
- These are related topics:
  - Getting or generating “random” numbers
  - Generating random data for testing (or other) purposes

# Randomized Algorithms

◆ A **randomized algorithm** performs coin tosses (i.e., uses random bits) to control its execution

◆ It contains statements of the type

$b \leftarrow \text{random}()$

**if**  $b = 0$

do A ...

**else** {  $b = 1$  }

do B ...

◆ Its running time depends on the outcomes of the coin tosses

- We analyze the expected running time of a randomized algorithm under the following assumptions

the coins are unbiased, and  
the coin tosses are  
independent

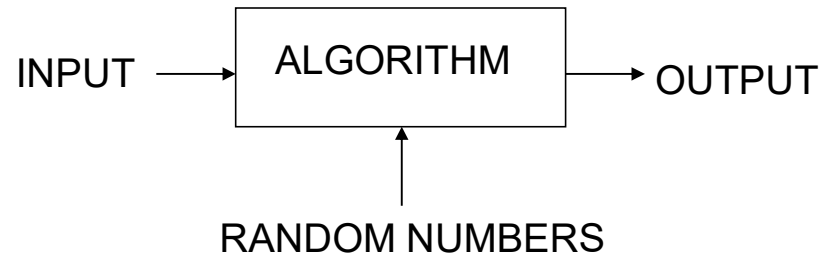
The worst-case running time of a randomized algorithm is often large but has very low probability (e.g., it occurs when all the coin tosses give “heads”)

# Pseudorandom numbers

- The computer is *not capable* of generating truly random numbers
  - The computer can only generate pseudorandom numbers--numbers that are generated by a formula
  - Pseudorandom numbers *look* random, but are perfectly predictable if you know the formula
    - Pseudorandom numbers are good enough for most purposes, but not all--for example, not for serious security applications
  - Devices for generating truly random numbers do exist
    - They are based on radioactive decay, or on lava lamps
- “Anyone who attempts to generate random numbers by deterministic means is, of course, living in a state of sin.”

—John von Neumann

# Randomized Algorithms



- In addition to input, algorithm takes a source of random numbers and makes random choices during execution;
- Behavior can vary even on a fixed input;

# Types of Randomized Algorithms

## Randomized **Las Vegas** Algorithms:

- Output is always correct
- Running time is a **random variable**

**Example:** Randomized Quick Sort

## Randomized **Monte Carlo** Algorithms:

- Output may be incorrect with some probability
- Running time is deterministic.

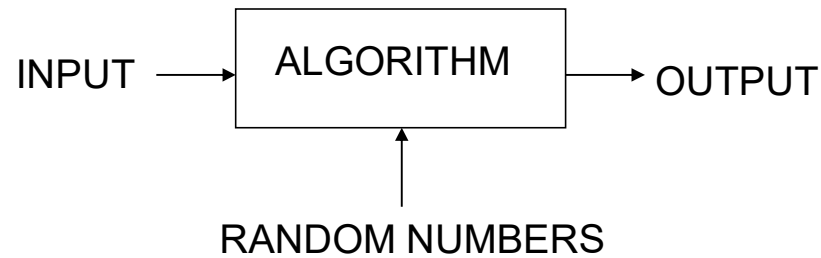
**Example:** Randomized algorithm for **approximate median**



# Motivation for Randomized Algorithms

- Simplicity;
- Performance;
- Reflects reality better (Online Algorithms);
- For many hard problems helps obtain better complexity bounds when compared to deterministic approaches;

# Las Vegas Randomized Algorithms



**Goal:** Prove that for all input instances the algorithm solves the problem correctly and the expected number of steps is bounded by a polynomial in the input size.

**Note:** The expectation is over the random choices made by the algorithm.

# Probabilistic Analysis of Algorithms



Input is assumed to be from a probability distribution.

**Goal:** Show that for all inputs the algorithm works correctly and for most inputs the number of steps is bounded by a polynomial in the size of the input.

**Example** : randomized Quick Sort

# QuickSort( $S$ )

QuickSort( $S$ )

{     If ( $|S| > 1$ )

        Pick and remove an element  $x$  from  $S$ ;

$(S_{<x}, S_{>x}) \leftarrow \text{Partition}(S, x)$ ;

        return( Concatenate(QuickSort( $S_{<x}$ ),  $x$ ,  
QuickSort( $S_{>x}$ ))

}

# QuickSort( $S$ )

When the input  $S$  is stored in an array  $A$

QuickSort( $A, l, r$ )

```
{   If ( $l < r$ )  
     $x \leftarrow A[l]$ ;  
     $i \leftarrow \text{Partition}(A, l, r, x)$ ;  
    QuickSort( $A, l, i - 1$ );  
    QuickSort( $A, i + 1, r$ )  
}
```

- **Average** case running time:  $O(n \log n)$
- **Worst** case running time:  $O(n^2)$
- **Distribution sensitive**: Time taken depends upon the initial permutation of  $A$ .

# Randomized QuickSort(*S*)

When the input *S* is stored in an array *A*

QuickSort(*A*, *l*, *r*)

{     If (*l* < *r*)

*x* ← *A*[*l*];

*i* ← Partition(*A*, *l*, *r*, *x*);

        QuickSort(*A*, *l*, *i* - 1);

        QuickSort(*A*, *i* + 1, *r*)

}

an element selected **randomly** uniformly  
from *A*[*l*..*r*];

- **Distribution** insensitive: Time taken does not depend on initial permutation of *A*.
- Time taken **depends** upon the **random** choices of pivot elements.
  1. For a given input, Expected(**average**) running time:  $O(n \log n)$
  2. **Worst** case running time:  $O(n^2)$

# Randomized Quick Sort

## Randomized-Partition( $A, p, r$ )

1.  $i \leftarrow \text{Random}(p, r)$
2. exchange  $A[r] \leftrightarrow A[i]$
3. return **Partition**( $A, p, r$ )

## Randomized-Quicksort( $A, p, r$ )

1. if  $p < r$
2.   then  $q \leftarrow \text{Randomized-Partition}(A, p, r)$
3.       **Randomized-Quicksort**( $A, p, q-1$ )
4.       **Randomized-Quicksort**( $A, q+1, r$ )



# Randomized Quick Sort

- Exchange  $A[r]$  with an element chosen at random from  $A[p\dots r]$  in **Partition**.
- The pivot element is equally likely to be any of input elements.
- *For any given input, the behavior of Randomized Quick Sort is determined not only by the input but also by the random choices of the pivot.*
- We add randomization to Quick Sort to obtain for any input the expected performance of the algorithm to be good.