## Lecture 1: Introduction (CLRS 1, 2.1-2.2)

## 1 Introduction

- Class is about *designing* and *analyzing algorithms* 
  - Algorithm: A well-defined procedure that takes an input and computes some output.
    - \* Not a program (but often specified like it): An algorithm can often be implemented in several ways.
  - Design: Methods/ideas for developing (efficient) algorithms.
  - Analysis: Abstract/mathematical comparison of algorithms (without actually implementing them). Think of analysis as a measure of the quality of your algorithm and use it to justify design decisions when you write programs.
- In this class we do all these:
  - come up with solutions for a problem
  - prove that it is correct
  - analyze its running time
- Hopefully the class will show that algorithms matter!

## 2 Algorithm example: Insertion-sort

The problem of sorting is defined as:

- Input: n integers in array A[1..n]
- Output: A sorted in increasing order

Insertion-sort works similarly with sorting a deck of cards. The algorithm is described below in a "pseudo-code" that we will use to describe algorithms.

```
INSERTION-SORT(A)

For j = 2 to n DO

key = A[j]

i = j - 1

WHILE i > 0 and A[i] > key DO

A[i + 1] = A[i]

i = i - 1

OD

A[i + 1] = key

OD
```

How does it work? Example:

5	2	4	6	1	3	j=2	i=1	key=2
5	5	4	6	1	3		i=0	
2	5	4	6	1	3			
2	5	4	6	1	3	j=3	i=2	key=4
2	5	5	6	1	3		i=1	
2	4	5	6	1	3			
2	4	5	6	1	3	j=4	i=3	key=6
2	4	5	6	1	3			
2	4	5	6	1	3	j=5	i=4	key=1
2	4	5	6	6	3		i=3	
2	4	5	5	6	3		i=2	
2	4	4	5	6	3		i=1	
2	2	4	5	6	3		i=0	
1	2	4	5	6	3			
1	2	4	5	6	3	j=6	i=5	key=3
1	2	4	5	6	6		i=4	
1	2	4	5	5	6		i=3	
1	2	4	4	5	6		i=2	
1	2	3	4	5	6			

#### 2.1 Correctness

We prove correctness by finding and proving certain conditions that hold at some point in the algorithm *for any input*. These are called *invariants*.

- Prove the following loop invariant: "A[1..j-1] is sorted" holds at the beginning of each iteration of FOR-loop.
  - When j=n+1 (*Termination*) we have the correct output.
- The loop invariant can be proved by induction (try it!).
- Note: In many cases it is harder to find the right invariant(s) than to prove it (them).

### 2.2 Analysis

- We want to predict the resource use of the algorithm.
- We can be interested in different resources (like main memory, bandwidth), but normally *running time*.
- To analyze running time without actually implementing the algorithm we need a mathematical model of a computer:

#### Random-access machine (RAM) model:

- Instructions executed sequentially one at a time
- All instructions take unit time:
  - \* Load/Store
  - \* Arithmetics (e.g. +, -, \*, /)
  - \* Logic (e.g. >)
- Main memory is infinite

# The running time of an algorithm is the number of instructions it executes in the RAM model of computation.

- RAM model not completely realistic, e.g.
  - main memory not infinite (even though we often imagine it is when we program)
  - not all memory accesses take same time (cache, main memory, disk)
  - not all arithmetic operations take same time (e.g. multiplications expensive)
  - instruction pipelining
  - other processes
- But RAM model often enough to give relatively realistic results (if we don't cheat too much).
- Running time of insertion-sort depends on many things

- How sorted the input is
- How big the input is, etc etc
- Normally we are interested in running time as a function of *input size* 
  - in insertion-sort: n.
- **Best-case running time:** The shortest running time for any input of size *n*. The algorithm will never be faster than this.
- Worst-case running time: The longest running time for *any* input of size *n*. The algorithm will never be slower than this.
- Average-case running time: Be careful: average over what? Must assume an input distribution.
- Let us analyze insertion-sort by assuming that line i in the program use c RAM instructions.
  - How many times are each line of the program executed?
  - Let  $t_j$  be the number of times line 4 (the WHILE statement) is executed in the j'th iteration.

cost times FOR j = 2 to n DO cnkey = A[j]n-1ci = j - 1n-1c $\sum_{j=2}^{n} t_j \\ \sum_{j=2}^{n} (t_j - 1) \\ \sum_{j=2}^{n} (t_j - 1)$ WHILE i > 0 and A[i] > key DO cA[i+1] = A[i]ci = i - 1cOD n-1A[i+1] = keycOD

• Running time: (depends on  $t_j$ )  $T(n) = cn + 2c(n-1) + c\sum_{j=2}^n t_j + 2c\sum_{j=2}^n (t_j-1) + c(n-1)$ 

- Best case:  $t_j = 1$  (already sorted) T(n) = cn + 2c(n-1) + c(n-1) + c(n-1) = 5cn - 4c $= k_1n - k_2$ 

Linear function of n

- Worst case:  $t_j = j$  (sorted in decreasing order)  $T(n) = cn + 2c(n-1) + c \sum_{j=2}^{n} j + 2c \sum_{j=2}^{n} (j-1) + c(n-1)$   $= cn + 2c(n-1) + c(\frac{n(n+1)}{2} - 1) + 2c(\frac{(n-1)n}{2}) + c(n-1)$  = ... $= k_3n^2 + k_4n - k_5$ 

#### Quadratic function of n

Note: We used  $\sum_{j=1}^{n} j = \frac{n(n+1)}{2}$  (Next week!)

- Average case: We assume *n* numbers chosen randomly  $\Rightarrow t_j = j/2$   $T(n) = k_6 n^2 + k_7 n + k_8$ Still Quadratic function of *n* 

• Note:

- We will normally be interested in worst-case running time.
  - \* For some algorithms, worst-case occur fairly often.
  - \* Average case often as bad as worst case (but not always!).
- We will only consider order of growth of running time:
  - \* We already ignored cost of each statement and used the constants c.
  - \* We even ignored c and used  $k_i$ .
  - \* We simply said that best case was linear in n and worst/average case quadratic in n.
  - $\Rightarrow$  O-notation (best case O(n), worst/average case  $O(n^2)$ ) (next lecture!)