

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	c	n
$key = A[j]$	c	$n - 1$
$i = j - 1$	c	$n - 1$
WHILE $i > 0$ and $A[i] > key$ DO	c	$\sum_{j=2}^n t_j$
$A[i + 1] = A[i]$	c	$\sum_{j=2}^n (t_j - 1)$
$i = i - 1$	c	$\sum_{j=2}^n (t_j - 1)$
OD		
$A[i + 1] = key$	c	$n - 1$
OD		

- 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)
$$\begin{aligned} T(n) &= cn + 2c(n - 1) + c(n - 1) + c(n - 1) \\ &= 5cn - 4c \\ &= k_1n - k_2 \end{aligned}$$

Linear function of n
 - **Worst case:** $t_j = j$ (sorted in decreasing order)
$$\begin{aligned} T(n) &= cn + 2c(n - 1) + c \sum_{j=2}^n j + 2c \sum_{j=2}^n (j - 1) + c(n - 1) \\ &= cn + 2c(n - 1) + c\left(\frac{n(n+1)}{2} - 1\right) + 2c\left(\frac{(n-1)n}{2}\right) + c(n - 1) \\ &= \dots \\ &= k_3n^2 + k_4n - k_5 \end{aligned}$$

Quadratic function of n

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