# CS 255, Spring 2014, SJSU

## Course introduction: Analysis of Insertion Sort

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## Agenda

- ► Course introduction.
- ▶ Warmup problem to get us started in analysis of algorithms.
- ▶ Review a problem that you should already know.
- ▶ Describe it in pseudocode (as in the textbook).
- ▶ Start using asymptotic notation for algorithm running time.

### Course introduction

- ▶ This is a course on design and analysis of algorithms.
- ▶ Not a programming class, but assumes you know how to program.
- ▶ Assumes you know elementary algorithms and data structures.
- ▶ Grading: 7 HWs, 2 midterms, 1 final exam.
- ▶ See greensheet and tentative schedule in canvas for further details.

# The sorting problem

- ▶ Input: A sequence of *n* numbers  $\langle a_1, a_2, \ldots, a_n \rangle$
- ▶ Output: A permutation  $\langle a'_1, a'_2, \dots, a'_n \rangle$  of the input sequence such that:

$$a_1' \leq a_2' \leq \ldots \leq a_n'$$

- ► Assume sequence is stored in an array.
- Example:
  - ► Input: ⟨8, 2, 4, 9, 3, 6⟩
  - ▶ Output: ⟨2, 3, 4, 6, 8, 9⟩

### **Insertion Sort**

- ▶ Good for sorting small arrays.
- ► Works more or less in the same way we would sort a hand of cards when picking them one by one.
  - ▶ The cards in our hand are always sorted.
  - ▶ When we pick a new card we insert it in the correct position.

### Pseudocode

(convention: arrays start at position 1)

```
INSERTION-SORT(A)

1 for j = 2 to length[A]

2 key = A[j]

3 // Insert A[j] into the sorted sequence A[1..j-1]

4 i = j-1

5 while i > 0 and A[i] > key

6 A[i+1] = A[i]

7 i = i-1

8 A[i+1] = key
```

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### Example

```
8 2 4 9 3 6  // input
2 8 4 9 3 6  // end of 1st iteration
2 4 8 9 3 6  // end of 2nd iteration
2 4 8 9 3 6  // end of 3rd iteration
2 3 4 8 9 6  // end of 4th iteration
2 3 4 6 8 9  // end of 5th iteration
```

### Correctness

- Loop Invariant: At the beginning of each iteration of the **for** loop, the subarray A[1..j-1] is sorted.
- ▶ We use this invariant to prove the algorithm is correct.

#### Correctness

Need to prove 3 things about the loop invariant:

- 1. Initialization: It is true before the 1st iteration.
- 2. <u>Maintenance</u>: If it is true before a given iteration, it remains true at the beginning of the next iteration.
- 3. Termination: When the loop ends, the array A[1..n] is sorted.

Loop invariants

The utilization of loop invariants is analogous to mathematical induction:

- ▶ Initialization → base case.
- ► <u>Maintenance</u> → indutive step.
- ► <u>Termination</u>: → no correspondence in mathematical induction. Here the "induction" stops when the loop terminates.

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# How to analyze the algorithm's running time?

It depends of the input:

- ► Sorting 100000 numbers takes more time than sorting 30 numbers.
- ► It can take different time for arrays of the same size (ex: INSERTION-SORT is faster if the input is already sorted.)

# How to analyze the algorithm's running time?

The input size depends on the problem:

- ▶ Usually it is the number of input elements (*n* in the sorting problem).
- ▶ But it can be something else:
  - ▶ In graph algorithms, the input size is usually expressed in terms of 2 quantities: number of nodes, and number of edges.

## Running time

The <u>running time</u> of an algorithm on a given input is the number of primitive operations executed.

- ▶ Primitive operations: assignments, comparisions, etc.
- ► Each line of the pseudocode requires a constant amount of time (independent of the input size).
- ▶ Different lines might take different amounts of time.
- ► Function calls also take constant time, but its execution might not.

Analysis of Insertion-Sort

- For j = 2, 3, ..., n, let  $t_j$  be the number of times that the **while** loop test is executed for that value of j.
- ► (NOTE: **for** and **while** loop tests are executed one more time than the loop bodies. Why?)
- ▶ Running time depends on the  $t_{j's}$  (which depend on the input).

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### Pseudocode

```
INSERTION-SORT(A)

1 for j = 2 to length[A]
```

$$2 key = A[i]$$

3 // Insert 
$$A[j]$$
 into the sorted sequence  $A[1..j-1]$ 

4 
$$i = j - 1$$

5 **while** 
$$i > 0$$
 and  $A[i] > key$ 

$$A[i+1] = A[i]$$

$$7 i = i - 1$$

$$8 A[i+1] = key$$

### Number of times each line is executed

▶ line 1: *n* 

▶ line 2: n-1

▶ line 3: *n* − 1

▶ linel 4: *n* − 1

▶ line 5:  $t_2 + t_3 + \ldots + t_n$ 

▶ line 6:  $(t_2-1)+(t_3-1)+\ldots+(t_n-1)$ 

▶ line 7:  $(t_2 - 1) + (t_3 - 1) + \ldots + (t_n - 1)$ 

▶ line 8: *n* − 1

## Running time

- $ightharpoonup \sum_{i}$  (cost of line i) imes (num times line i is executed)
- ▶ Let *c<sub>i</sub>* be the cost of line *i*.
- ▶ Let T(n) be the running time of INSERTION-SORT.

$$T(n) = c_1 n + c_2(n-1) + c_3(n-1) + c_4(n-1)$$

$$+ c_5 \sum_{j=2}^{n} t_j + c_6 \sum_{j=2}^{n} (t_j - 1) + c_7 \sum_{j=2}^{n} (t_j - 1) + c_8(n-1)$$

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#### Best case

- ▶ The input array is already sorted. That is,
- ▶  $A[i] \le key$  at the beginning of the **while** loop  $\implies$  all  $t_{j's} = 1$

$$T(n) = c_1 n + c_2 (n-1) + c_4 (n-1) + c_5 (n-1) + c_8 (n-1)$$
  
=  $(c_1 + c_2 + c_4 + c_5 + c_8) n - (c_2 + c_4 + c_5 + c_8)$ 

- ightharpoonup T(n) = an + b, with a and b constants.
- ightharpoonup T(n) is a linear function of n.

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### Worst case

- ▶ The input array is reverse sorted.
- Need to compare *key* with all elements to the left of position  $j \implies j-1$  comparisons.
- ▶ **while** loop ends when i = 0 (j 1 comparisons, j tests)  $\Longrightarrow$   $t_j = j$

$$\sum_{j=2}^{n} t_j = \sum_{j=2}^{n} j$$

$$\sum_{j=2}^{n} (t_j - 1) = \sum_{j=2}^{n} (j - 1)$$

(arithmetic progressions)

### Worst case

$$T(n) = c_1 n + c_2 (n-1) + c_4 (n-1) + c_5 \frac{(2+n)(n-1)}{2}$$

$$+ c_6 \frac{n(n-1)}{2} + c_7 \frac{n(n-1)}{2} + c_8 (n-1)$$

$$= \left(\frac{c_5}{2} + \frac{c_6}{2} + \frac{c_7}{2}\right) n^2$$

$$+ \left(c_1 + c_2 + c_4 + \frac{c_5}{2} - \frac{c_6}{2} - \frac{c_7}{2} + c_8\right) n$$

$$- \left(c_2 + c_4 + c_5 + c_8\right)$$

- ▶  $T(n) = an^2 + bn + c$ , with a, b, c constants.
- ightharpoonup T(n) is a quadratic function of n.

## Average case

- ▶ In general it's difficult or even impossible to calculate.
- ▶ Need several assumptions. For example, all input sequences are equally likely (which can be far from true).

Average case

- ► Suppose the input of INSERTION-SORT is an array of *n* numbers randomly generated from a uniform distribution.
- ▶ On average, *key* in A[j] is less than half of the elements in A[1..j-1] and is larger than the other half.
- $\implies$  on average the **while** loop has to look at half of the subarray  $A[1..j-1] \implies t_j=j/2$
- ► The running time is approx half of the worst case running time. Still a quadratic function of *n*.

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## Best, worst, and average cases

- ▶ Best case analysis is usually irrelevant.
- ► Why?
- ▶ ⇒ We can have a very bad sorting algorithm which has linear running time in the best case, and exponential running time in the worst (and average) case.
- Any hint of such an algorithm?

# Best, worst, and average cases

- ▶ We are usually interested in the <u>worst case</u> analysis: The longest running time on any given input of size *n*.
- ► Why?
  - Gives us an upper bound for the total execution time, regardless of the input.
  - ► Average case is usually as bad as the worst case, and much harder to analyze.
  - ▶ In many cases, the worst case happens very often (ex: searching for an element that is not in a collection.)

## Order of growth

- ▶ Make several simplifications and focus on the essentials.
  - eliminate lower order terms.
  - ▶ ignore the coefficient of the higher order term.
- ► Example: The worst case running time of INSERTION-SORT is given by  $an^2 + bn + c$ 
  - eliminating lower order terms  $\implies an^2$
  - eliminating the coefficient  $\implies n^2$
- ▶ IMPORTANT: Cannot say that  $T(n) = n^2$
- ightharpoonup but we can say that T(n) has a growth rate proportional to  $n^2$ .

## Order of growth

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- ▶ We say that  $T(n) = \Theta(n^2)$  to capture the notion that the order of growth is  $n^2$ .
- ► This is called asymptotic analysis.
- ➤ We usually say that an algorithm is more efficient than another if its worst case running time is of a lower order (i.e. has a lower asymptotic complexity).