

# ALGORITHM ANALYSIS

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## Introduction

- What is an algorithm?
  - a clearly specified **set of simple instructions** to be followed to solve a problem
    - Takes a set of values, as input and
    - produces a value, or set of values, as output
  - May be specified
    - In English
    - As a computer program
    - As a pseudo-code
- Data structures
  - Methods of organizing data
- Program = algorithms + data structures

## Introduction

- Why need algorithm analysis ?
  - Writing a working program is not good enough.
  - The program may be inefficient!
  - If the program is run on a **large data set**, then the running time becomes an issue.

## Example: Selection Problem

- Given a list of  $N$  numbers, determine the  $k^{\text{th}}$  largest, where  $k \leq N$ .
- Algorithm 1:
  - (1) Read  $N$  numbers into an array
  - (2) Sort the array in decreasing order by some simple algorithm
  - (3) Return the element in position  $k$

## Example: Selection Problem (2)

- Algorithm 2:
  - (1) Read the first  $k$  elements into an array and sort them in decreasing order
  - (2) Each remaining element is read one by one
    - If smaller than the  $k^{\text{th}}$  element, then it is ignored
    - Otherwise, it is placed in its correct spot in the array, bumping one element out of the array.
  - (3) The element in the  $k^{\text{th}}$  position is returned as the answer.

## Example: Selection Problem (3)

- Which algorithm is better when
  - $N = 100$  and  $k = 100$ ?
  - $N = 100$  and  $k = 1$ ?
- What happens when  $N = 1,000,000$  and  $k = 500,000$ ?
- There exist better algorithms.

## Algorithm Analysis

- We only analyze **correct** algorithms.
- An algorithm is correct
  - If, for every input instance, it halts with the correct output.
- Incorrect algorithms
  - Might not halt at all on some input instances.
  - Might halt with other than the desired answer.

## Algorithm Analysis (2)

- Analyzing an algorithm
  - **Predicting** the resources that the algorithm requires.
  - Resources include
    - Memory (space)
    - **Computational time** (usually most important)
    - Communication bandwidth (in parallel and distributed computing)

## Algorithm Analysis (3)

- Factors affecting the running time:
  - computer
  - compiler
  - algorithm used
  - input to the algorithm
    - The content of the input affects the running time
    - Typically, the **input size** (number of items in the input) is the main consideration.
      - sorting problem  $\Rightarrow$  the number of items to be sorted
      - multiply two matrices together  $\Rightarrow$  the total number of elements in the two matrices
    - And sometimes the **input order** as well (e.g., sorting algorithms).
- Machine model assumed
  - Instructions are executed one after another, with no concurrent operations  $\Rightarrow$  not parallel computers

## Analysis Model

- It takes exactly one time unit to do any calculation such as
  - + , - , \* , / , % , & , | , && , || , etc.
  - comparison
  - assignment
- There is an infinite amount of memory.
- It does not consider the cost associated with page faulting or swapping.
- It does not include I/O costs (which is usually one or more orders of magnitude higher than computation costs).

## An Example

```
int sum ( int n ) {
    int partialSum;
1      partialSum = 0;
2      for ( int i = 0; i <= n-1; i++ )
3          partialSum += i*i*i;
4      return partialSum;
}
```

- Lines 1 and 4: one unit each
- Line 3:  $4N$
- Line 2:  $1+(N+1)+N=2N+2$
- Total:  $6N+4 \Rightarrow O(N)$

## Running Time Calculations

- Throw away leading constants.
- Throw away low-order terms.
- Compute a Big-Oh running time:
  - An upper bound for running time
  - Never underestimate the running time of a program
  - The program may end earlier, but never later (worst-case running time)

## General Rules for Big-Oh: *for* loops

- *for* loops
  - at most the running time of the statements inside the *for* loop (including tests) times the number of iterations.
- Nested *for* loops

```
for (i=0; i<n; i++)
  for (j=0; j<n; j++)
    k++;
```

- the running time of the statement multiplied by the product of the sizes of all the *for* loops.
- $O(N^2)$

## Consecutive Statements

- Consecutive statements

```
for (i=0; i<n; i++)
  a[i]=0;
for (i=0; i<n; i++)
  for (j=0; j<n; j++)
    a[i] += a[j]+i+j;
```

- These just add.
- $O(N) + O(N^2) = O(N^2)$

## if – then – else

- `if C then S1`  
`else S2`
  - never more than the running time of the test plus the larger of the running times of S1 and S2.

```
if (n > 0)
    for ( int i = 0; i < n; i++ )
        sum += i;
else
    System.out.println( "Invalid input" );
```

## Strategies

- Analyze from the inside out (loops).
- If there are method calls, analyze these first.
- Recursive methods (later):
  - Could be just a hidden “for” loop  $\Rightarrow$  simple.
  - Solve a recurrence  $\Rightarrow$  more complex.



## Worst- / Average- / Best-Case

- Worst-case running time of an algorithm:
  - The **longest** running time for **any input of size  $n$**
  - An upper bound on the running time for any input  
⇒ guarantee that the algorithm will never take longer
  - Example: Sort a set of numbers in increasing order; and the input is in decreasing order
  - The worst case can occur fairly often
    - Example: searching a database for a particular piece of information
- Best-case running time:
  - sort a set of numbers in increasing order; and the input is already in increasing order
- Average-case running time:
  - May be difficult to define what “average” means

## Example

- Given an array of integers, return true if the array contains number 100, and false otherwise.
  - Best case: ?
  - Worst case: ?
  - Average case: ?

## Informal Introduction to $O$ , $\Omega$ and $\Theta$

- Given an unsorted array of integers, return true if a number  $k$  is in the array and false otherwise.

```
for( i = 0; i < N; i++ )  
    if ( k == A[i] )  
        return ( true );  
return ( false );
```

- Worst-case running time is  $O(N)$ .  
⇒ The alg has  $O(N)$  running time.
- Best-case running time is  $O(1)$ .  
⇒ The alg has  $\Omega(1)$  running time.

- Given an unsorted array of integers, find and return the maximum value stored in the array.

```
max = A[0];  
for( i = 1; i < N; i++ )  
    if ( max < A[i] )  
        max = A[i];  
return( max );
```

- Worst-case running time is  $O(N)$ .
- Best-case running time is  $O(N)$ .
- ⇒ The alg has  $\Theta(N)$  running time.

20

## Running Time of Algorithms

- Bounds are for **algorithms**, rather than **programs**.
  - Programs are just implementations of an algorithm.
  - Almost always the details of the program do not affect the bounds.
- Bounds are for **algorithms**, rather than **problems**.
  - A problem can be solved with several algorithms, some are more efficient than others.

## Example: Insertion Sort



- 1) Initially  $p = 1$
- 2) Let the first  $p$  elements be sorted
- 3) Insert the  $(p+1)$ th element properly in the list so that now  $p+1$  elements are sorted.
- 4) Increment  $p$  and go to step (3)

## Insertion Sort: Example

Original	34	8	64	51	32	21	Positions Moved
After $p = 1$	8	34	64	51	32	21	1
After $p = 2$	8	34	64	51	32	21	0
After $p = 3$	8	34	51	64	32	21	1
After $p = 4$	8	32	34	51	64	21	3
After $p = 5$	8	21	32	34	51	64	4

## Insertion Sort: Algorithm

```
for (int p=1; p<a.size(); p++)
{
    int tmp=a[p];
    for (j=p; j> 0 && tmp < a[j-1]; j--)
        a[j] = a[j-1];
    a[j] = tmp;
}
```

- Consists of  $N - 1$  passes
- For pass  $p = 1$  through  $N - 1$ , ensures that the elements in positions 0 through  $p$  are in sorted order
  - elements in positions 0 through  $p - 1$  are already sorted
  - move the element in position  $p$  left until its correct place is found among the first  $p + 1$  elements

## Example 2

To sort the following numbers in increasing order:

34 8 64 51 32 21

$p = 1$ ;  $tmp = 8$ ;

$34 > tmp$ , so second element  $a[1]$  is set to 34: {8, 34}...

We have reached the front of the list. Thus, 1st position  $a[0] = tmp=8$

After 1st pass: 8 34 64 51 32 21

(first 2 elements are sorted)

P = 2; tmp = 64;  
 34 < 64, so stop at 3<sup>rd</sup> position and set 3<sup>rd</sup> position = 64  
 After 2nd pass: 8 34 64 51 32 21  
 (first 3 elements are sorted)

P = 3; tmp = 51;  
 51 < 64, so we have 8 34 64 64 32 21,  
 34 < 51, so stop at 2nd position, set 3<sup>rd</sup> position = tmp,  
 After 3rd pass: 8 34 51 64 32 21  
 (first 4 elements are sorted)

P = 4; tmp = 32,  
 32 < 64, so 8 34 51 64 64 21,  
 32 < 51, so 8 34 51 51 64 21,  
 next 32 < 34, so 8 34 34 51 64 21,  
 next 32 > 8, so stop at 1st position and set 2<sup>nd</sup> position = 32,  
 After 4th pass: 8 32 34 51 64 21

P = 5; tmp = 21, ...  
 After 5th pass: 8 21 32 34 51 64

## Analysis: Worst-case Running Time

- What is the worst input?
- Consider a reversed sorted list as input.
- When  $a[p]$  is inserted into the sorted sub-array  $a[0 \dots p-1]$ , we need to compare  $a[p]$  with all elements in  $a[0 \dots p-1]$  and move each element one position to the right  
 $\Rightarrow i$  steps.
- Inner loop is executed  $p$  times, for each  $p = 1, 2, \dots, N-1$   
 $\Rightarrow$  Overall:  $1 + 2 + 3 + \dots + N-1 = \dots = O(N^2)$

## Analysis: Best-case Running Time

- The input is already sorted in the right order.
- When inserting  $a[p]$  into the sorted sub-array  $a[0 \dots p-1]$ , only need to compare  $a[p]$  with  $a[p-1]$  and there is no data movement  
 $\Rightarrow O(1)$
- For each iteration of the outer for-loop, the inner for-loop terminates after checking the loop condition once  
 $\Rightarrow O(N)$  time
- If input is *nearly sorted*, insertion sort runs fast.

## Insertion Sort: Summary

```
for (int p=1; p<a.size(); p++)
{
    int tmp=a[p];
    for (j=p; j> 0 && tmp < a[j-1]; j--)
        a[j] = a[j-1];
    a[j] = tmp;
}
```

- $O(N^2)$
- $\Omega(N)$
- Space requirement is  $O(?)$

## Next time ...

- Growth rates
- $O$ ,  $\Omega$ ,  $\Theta$ ,  $o$
- Reading for this lecture: chapter 4