

Running time analysis

Lecture 10

Measuring the running “time”

- Goal: Analyze an algorithm written in pseudocode and describe its running time
 - Without having to write code
 - In a way that is independent of the computer used
- To achieve that, we need to
 - Make simplifying assumptions about the running time of each basic (primitive) operations
 - Study how the number of primitive operations depends on the size of the problem solved

• Primitive operation

Simple computer operation that can be performed in time that is always the same, independent of the size of the bigger problem solved (we say: constant time)

- **Assigning a value to a variable:** $x \leftarrow 1$ T_{assign}
 - **Calling a method:** `Expos.addWin()` T_{call}
 - Note: doesn't include the time to execute the method
 - **Returning from a method:** `return x;` T_{return}
 - **Arithmetic operations on primitive types** T_{arith}
 - $x + y$, $r * 3.1416$, x/y , etc.
 - **Comparisons on primitive types:** $x == y$ T_{comp}
 - **Conditionals:** `if (...) then.. else...` T_{cond}
 - **Indexing into an array:** `A[i]` T_{index}
 - **Following object reference:** `Expos.losses` T_{ref}
- Note: Multiplying two bigIntegers is *not* a primitive operation, because the running time depends on the size of the numbers multiplied.

FindMin analysis

Algorithm findMin(A, start, stop)

Input: An array A and indices start and stop

Output: The index of the smallest element of A between start and stop (inclusively)

`minvalue \leftarrow A[start]`

`minindex \leftarrow start`

`index \leftarrow start + 1`

while (index \leq stop) **do** {

if (A[index] < minvalue) **then** {

`minvalue \leftarrow A[index]`

`minindex \leftarrow index`

 }

`index = index + 1`

}

return minindex

Running time

$T_{\text{index}} + T_{\text{assign}}$
 T_{assign}
 $T_{\text{arith}} + T_{\text{assign}}$
 $T_{\text{comp}} + T_{\text{cond}}$
 $T_{\text{index}} + T_{\text{comp}} + T_{\text{cond}}$ repeated
 $T_{\text{index}} + T_{\text{assign}}$ stop-start-1
 T_{assign} times
 $T_{\text{assign}} + T_{\text{arith}}$
 T_{return}

Best case, Worst case

- Running time depends on $n = \text{stop} - \text{start}$
 - But it also depends on the content of the array!
- What kind of array of n elements will give the best running time for findMin?
- The worst running time?

More assumptions

- Counting each type of primitive operations is tedious
- We assume that the running time of each operation is roughly comparable:

$T_{\text{assign}} \approx T_{\text{comp}} \approx T_{\text{arith}} \approx \dots \approx T_{\text{index}} = 1$ primitive operation

- We are only interested in the *number* of primitive operations performed

Worst-case running time for findMin becomes:

Selection Sort

Algorithm SelectionSort(A,n)

Input: an array A of n elements

Output: the array is sorted

i ← 0

while (i < n) **do** {

 minindex ← findMin(A,i,n-1)

 t ← A[minindex]

 A[minindex] ← A[i]

 A[i] ← t

 i ← i + 1

}

**Primitive operations
(worst case) :**

1

2

3 + (6 + 10 (n - 1 - i))

2

3

2

2

} n times

$$\text{Total: } T(n) = 1 + \left(\sum_{i=0}^{n-1} 20 + 10(n-1-i) \right)$$

$$= 1 + 20n + 10(n-1)n - 10 \sum_{i=0}^{n-1} i$$

$$= 1 + 20n + 10(n-1)n - 10(n-1)n/2$$

$$= 1 + 20n + 5(n-1)n = 1 + 15n + 5n^2$$

More simplifications

We have: $T(n) = 1 + 15n + 5n^2$

Simplification #1:

When n is large, $T(n) \approx 5n^2$

Simplification #2:

When n is large, T(n) grows approximately like n^2

We will write T(n) is $O(n^2)$

"T(n) is big-O of n squared"