Introduction to complexity classes

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Today

- Review
 - Loop invariant and incorrectness examples
- Introduction to complexity classes
 - Big-Oh
 - Big-Omega
 - Big-Theta

Correctness and incorrectness

- To prove an algorithm is correct:
 - Prove that it produces the correct output for every input
 - Trace input to find algorithm's output
 - Prove output is correct
 - Loop invariants for loops
 - Cases for if statements
 - Recursion: generally proof by induction
 - May also use proof by contradiction
- To prove an algorithm is incorrect:
 - Find a counterexample
 - Instance where the algorithm computes an incorrect solution

Loop invariant example

• Algorithm: SelectionSort

• Invariant: After iteration i, data[x] \leq all values that follow it, for all x from 1 up to i

Variant invariant solution

• Invariant: After iteration i, data [x] \leq all values that follow it, for all x from 1 up to i

Proof. (1st iteration) During the first iteration, the min of data is swapped with data[1], so data[1] is less than or equal to all of the other values by the definition of the minimum.

 $((k+1)^{st} \ iteration)$ Suppose that each data[x] is less than or equal to everything that follows it up to data[k] after iteration k. During iteration k+1, the min of data[k+1..n] is swapped to position i, so it will be less than or equal to everything that follows it. Since no previous elements in the array were moved, all of the previous inequalities still hold, so now all data[x] up to k+1 are less than or equal to all elements that follow them.

Hence, the loop invariant holds for every iteration of the loop. \Box *Proof of correctness.* By the loop invariant, after the i=n (last) iteration of the for loop, all n elements of data will be less than or equal to everything that follow them. In particular, they will all (other than data[n]) be less than or equal to the subsequent element, so $data[1] \leq data[2] \leq \ldots \leq data[n]$. Thus, the output of SelectionSort is sorted.

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Loop invariant exercise

• Algorithm: InsertionSort

```
Input:
data: an array of integers to sort
n: the number of values in data
Output: permutation of data such that data[1] ≤ ... ≤ data[n]
Pseudocode:

   for i = 2 to n
        Let ins = data[i]
        Let j = last index of data[1..i-1] <= ins (or 0 if all > ins)
        Shift data[j+1..i-1] one space to the right
        data[j+1] = ins
    end
   return data
```

• **Invariant:** after iteration *i*, data[1..i] is sorted.

Exercise solution

• **Invariant:** after iteration *i*, data[1..i] is sorted.

Proof. (1st iteration) When i = 2, ins becomes data[2]. We have two cases for the value of j. (Case 1) If $data[1] \leq data[2]$, j = 1, nothing is shifted right, and data[2] = ins = data[2]. Since $data[1] \leq data[2]$, data[1] is shifted into data[2] and data[1] = data[2].

(Case 2) If data[1] > data[2], j = 0, data[1] is shifted into data[2], and data[1] = ins, the original value of data[2]. Since this value is smaller than the original value of data[1], data[1] < data[2], and data[1..2] is sorted.

Hence, in both cases, data[1..i] is sorted.

 $((k+1)^{\operatorname{st}} \text{ iteration})$ Suppose that $data[1] \leq data[2] \leq \ldots \leq data[k]$ after iteration k. In iteration k+1, i=k+1, so ins becomes data[k+1]. Then, j becomes the last value of data[1..k] such that $data[j] \leq ins$ (or 0), and data[j+1..i-1] are all shifted right. After this shift, $data[1] \leq data[2] \leq \ldots \leq data[j]$ and $data[j+2] \leq \ldots \leq data[k+1]$ because of the previous iteration. When data[j+1] becomes ins, data[j+2] must be greater than ins, and (if j>0) $data[j] \leq ins$, both by the definition of j. Hence, $data[1] \leq \ldots \leq data[j] \leq data[j+1] < data[j+2] \leq \ldots \leq data[k+1]$, so data[1..k+1] is sorted.

Therefore, by induction, data[1..i] is sorted after every iteration of the for loop.

The importance of pseudocode

Prove that BadSort (below) is not a correct sorting algorithm.

```
Input:
data: an array of integers to sort
       the number of values in data
n:
_{1} for i = n-1 to 1 step -1
    for j = 1 to n-i step i
      if data[j] > data[j+i]
3
        Swap data[j] and data[j+i]
      end
5
   end
 end
```

• What does this even mean?

Prove that BadSort (below) is not a correct sorting algorithm.

```
Input:
data: an array of integers to sort
n: the number of values in data
1 for i = n-1 to 1 step -1
2   Consider every ith value of data, starting at data[1]
3   if two of these values are out of order,
4   Swap them
5 end
```

• Prove that BadSort (below) is not a correct sorting algorithm.

```
Input:
data: an array of integers to sort
n: the number of values in data
1 for i = n-1 to 1 step -1
2 Consider every ith value of data, starting at data[1]
3 if two of these values are out of order,
4 Swap them
5 end
```

i = 4

• Input: 3 2 5 4 1

• Prove that BadSort (below) is not a correct sorting algorithm.

```
Input:
data: an array of integers to sort
n: the number of values in data
1 for i = n-1 to 1 step -1
2 Consider every ith value of data, starting at data[1]
3 if two of these values are out of order,
4 Swap them
5 end
```

$$i = 3$$

• Input: 1 2 5 4 3

• Prove that BadSort (below) is not a correct sorting algorithm.

```
Input:
data: an array of integers to sort
n: the number of values in data
1 for i = n-1 to 1 step -1
2 Consider every ith value of data, starting at data[1]
3 if two of these values are out of order,
4 Swap them
5 end
```

i = 2

• Input: 1 2 5 4 3

• Prove that BadSort (below) is not a correct sorting algorithm.

```
Input:
data: an array of integers to sort
n: the number of values in data
1 for i = n-1 to 1 step -1
2 Consider every ith value of data, starting at data[1]
3 if two of these values are out of order,
4 Swap them
5 end
```

i = 1

• Input: 1 2 3 4 5

Prove that BadSort (below) is not a correct sorting algorithm.

```
Input:
data: an array of integers to sort
n: the number of values in data
1 for i = n-1 to 1 step -1
2   Consider every ith value of data, starting at data[1]
3   if two of these values are out of order,
4   Swap them
5 end
```

Counterexample: [1, 3, 5, 2, 4], which outputs [1, 3, 2, 4, 5]

Complexity analysis

- What are the factors that contribute to the running time of an algorithm?
 - Processor speed
 - Number of instructions executed
 - Cache coherency
 - Resource conflicts (network, hard disk, etc.)
- Which of these are important when comparing algorithms?
 - Processor speed affects fast and slow algorithms equally
 - Not an important factor
- What can we most reliably control when designing an algorithm?
 - Number of instructions executed

RAM model of computation

Set of assumptions that make analysis more reasonable

Assumptions

- 1. All "basic" operations (assignment, arithmetic, branching, etc.) take 1 operation
 - Loops and functions do not qualify
- 2. Memory access is instantaneous
 - All variables are in registers
- 3. We have "infinite" memory

Cons

- Different operations take different number of clock cycles
- Cache locality has significant impact on performance
- Virtual memory can slow performance

Pros

Can actually analyze algorithms

RAM model example

```
data: an array of integers to find the min

n: the number of values in data
```

Min algorithm:

```
min = 1
for i = 2 to n

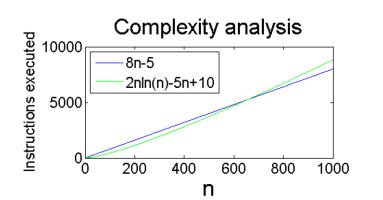
if data[i] < data[min]
min = i
min = i
end
end
return min</pre>
```

Question: is this better or worse than an algorithm that takes at most $2n \ln n - 5n + 10$ ops?

Better unless n < 658

Ops per line Times executed 1 1 2 n-1 (or n) 4 n-1 1 ??? ($\leq n-1$) 0 n/a 1 n-1

Total ops: $\leq 8(n-1) + 3 = 8n - 5$



Big-Oh notation

- Technique for *abstracting away details* of complexity
 - Can be used for time complexity, space complexity, etc.
- **Main idea:** most important aspect of complexity is *how fast it grows* relative to input size
 - Focus on asymptotic (eventual) growth rate
 - "Fast" functions will eventually pass "slow" functions for large n
 - Coefficients only matter if growth rate is similar
 - Predicting behavior for small n is difficult and often pointless
- Big-Oh notation
 - Organizes growth rates into classes
 - Three main symbols: $O(f(n)), \Omega(f(n)), \Theta(f(n))$
 - Analogous to "at least", "at most", and "similar to" f(n)

Big-Oh

• Upper bound ("at most")

f(n) = O(g(n)) if and only if there exist positive constants c and n_0 such that $f(n) \leq cg(n)$ for all $n \geq n_0$.

- We say "g(n) dominates f(n)" when f(n) = O(g(n))
- Notation weirdness:
 - O, Ω , and Θ are classes (sets) of functions
 - BUT: we use = to assign class, not ∈

Example

- Prove that
$$7n^2 + 19n - 4444 = O(n^2)$$
.

Proof. If
$$n \geq 19$$
,

$$7n^{2} + 19n - 4444 \le 7n^{2} + 19n$$

 $\le 7n^{2} + n^{2}$
 $= 8n^{2}$

Therefore, there exist positive constants c=8 and $n_0=19$ such that $7n^2+19n-4444 \le cn^2$ for all $n \ge n_0$.

Coming up

- Big-Omega and Big-Theta
- Big-Oh notation
- Big-Oh properties
- Logarithms review
- Homework 3 is due Friday
- **Homework 2** is due tomorrow
- Recommended readings: Sections 1.2, 2.2, and 3.1
- **Practice problems:** 2.2-, 2.2-3, 3.1-1