CSE 143

Lecture 4: testing and complexity

reading: 13.1-13.2



Tips for testing

- You cannot test every possible input, parameter value, etc.
 - Think of a limited set of tests likely to expose bugs.
- Think about boundary cases
 - Positive; zero; negative numbers
 - Right at the edge of an array or collection's size
- Think about empty cases and error cases
 - 0, -1, null; an empty list or array
- test behavior in combination
 - Maybe add usually works, but fails after you call remove
 - Make multiple calls; maybe size fails the second time only

Interfaces

interface: A list of methods that a class can promise to implement.

- Inheritance gives you an is-a relationship *and* code sharing.
 - A Lawyer can be treated as an Employee and inherits its code.
- Interfaces give you an is-a relationship *without* code sharing.
 - A Rectangle object can be treated as a Shape but inherits no code.
- Always declare variables using the *interface* type.

List<String> list = new ArrayList<String>();

Runtime Efficiency (13.2)

- **efficiency**: measure of computing resources used by code.
 - can be relative to speed (time), memory (space), etc.
 - most commonly refers to run time
- Assume the following:
 - Any single Java statement takes same amount of time to run.
 - A method call's runtime is measured by the total of the statements inside the method's body.
 - A loop's runtime, if the loop repeats N times, is N times the runtime of the statements in its body.





Algorithm growth rates (13.2)

- We measure runtime in proportion to the input data size, N.
 - growth rate: Change in runtime as N changes.
- Say an algorithm runs 0.4N³ + 25N² + 8N + 17 statements.
 - Consider the runtime when N is *extremely large* .
 - We ignore constants like 25 because they are tiny next to N.
 - The highest-order term (N³) dominates the overall runtime.

- We say that this algorithm runs "on the order of" N³.
- or O(N³) for short ("Big-Oh of N cubed")

Complexity classes

• **complexity class**: A category of algorithm efficiency based on the algorithm's relationship to the input size N.

Class	Big-Oh	If you double N,	Example
constant	O(1)	unchanged	10ms
logarithmic	O(log ₂ N)	increases slightly	175ms
linear	O(N)	doubles	3.2 sec
log-linear	O(N log ₂ N)	slightly more than doubles	6 sec
quadratic	O(N ²)	quadruples	1 min 42 sec
cubic	O(N ³)	multiplies by 8	55 min
		•••	
exponential	O(2 ^N)	multiplies drastically	5 * 10 ⁶¹ years

Complexity classes

Big-O Complexity



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Collection efficiency

• Efficiency of our ArrayIntList or Java's ArrayList:

Method	ArrayList
add	O(1)
add(index, value)	O(N)
get	O(1)
remove	O(N)
set	O(1)
size	O(1)

Max subsequence sum

- Write a method maxSum to find the largest sum of any contiguous subsequence in an array of integers.
 - Easy for all positives: include the whole array.
 - What if there are negatives?

index	0	1	2	3	4	5	6	7	8
value	2	1	-4	10	15	-2	22	-8	5

Largest sum: 10 + 15 + -2 + 22 = 45

• (Let's define the max to be 0 if the array is entirely negative.)

• Ideas for algorithms?

Algorithm 1 pseudocode

```
maxSum(a):
 max = 0.
 for each starting index i:
     for each ending index j:
         sum = add the elements from a[i] to a[j].
         if sum > max,
         max = sum.
```

return max.

index	0	1	2	3	4	5	6	7	8
value	2	1	-4	10	15	-2	22	-8	5

Algorithm 1 code

• What complexity class is this algorithm?

• **O(N³).** Takes a few seconds to process 2000 elements.

```
public static int maxSum1(int[] a) {
 int max = 0;
 for (int i = 0; i < a.length; i++) {
     for (int j = i; j < a.length; j++) {
         // sum = add the elements from a[i] to a[j].
         int sum = 0;
         for (int k = i; k \le j; k++) {
             sum += a[k];
         }
         if (sum > max) {
             max = sum;
         }
 return max;
```

Flaws in algorithm 1

- Observation: We are redundantly re-computing sums.
 - For example, we compute the sum between indexes 2 and 5: a[2] + a[3] + a[4] + a[5]
 - Next we compute the sum between indexes 2 and 6:
 a[2] + a[3] + a[4] + a[5] + a[6]
 - We already had computed the sum of 2-5, but we compute it again as part of the 2-6 computation.
 - Let's write an improved version that avoids this flaw.

index	0	1	2	3	4	5	6	7	8
value	2	1	-4	10	15	-2	22	-8	5

Algorithm 2 code

• What complexity class is this algorithm?

• **O(N²).** Can process tens of thousands of elements per second.

```
public static int maxSum2(int[] a) {
 int max = 0;
 for (int i = 0; i < a.length; i++) {
     int sum = 0;
     for (int j = i; j < a.length; j++) {
         sum += a[j];
         if (sum > max) {
             max = sum;
         }
     }
 }
 return max;
```

index	0	1	2	3	4	5	6	7	8
value	2	1	-4	10	15	-2	22	-8	5

A clever solution

• *Claim 1* : A max range cannot start with a negative-sum range.



 Claim 2: If sum(i, j-1) ≥ 0 and sum(i, j) < 0, any max range that ends at j+1 or higher cannot start at any of i through j.

i j-	1 ј	j+1	k
≥ 0	< 0	sum(j+1, k)	
< 0		sum(j+1, k)	
	Ş	sum(?, k) < sum(j+1, k)	

Together, these observations lead to a very clever algorithm...

Algorithm 3 code

- What complexity class is this algorithm?
 - O(N). Handles many millions of elements per second!

```
public static int maxSum3(int[] a) {
 int max = 0;
 int sum = 0;
 int i = 0;
 for (int j = 0; j < a.length; j++) {
     if (sum < 0) { // if sum becomes negative, max range
         i = j; // cannot start with any of i - j-1
         sum = 0; // (Claim 2)
     sum += a[j];
     if (sum > max) {
        max = sum;
 return max;
```

Runtime of first 2 versions

• Version 1:

N	Runtime (ms)	
1000	15	
2000	47	
4000	203	
8000	781	
16000	3110	
32000	12563	
64000	49937	



Version 2:



Input size (N)

Runtime of 3rd version

Version 3:

N	Runtime (ms)
1000	0
2000	0
4000	0
8000	0
16000	0
32000	0
64000	0
128000	0
256000	0
512000	0
le6	0
2e6	16
4e6	31
8e6	47
l.67e7	94
3.3e7	188
6.5e7	453
l.3e8	797
2.6e8	1578



Input size (N)