

Thursday, September 15, 2022

Reminders:

- test 1 in lab next week
- if you have an accommodations letter, inform your instructor!
- git add, git commit, git push

TODAY: theoretical analysis

- versions 1, 2, 3 of is-sorted

big O definition

- big O proofs

- sorting algorithms

selection sort

merge sort

Version 1 of is-sorted size = n

i loop goes from i=0 to i=size-1

j loop goes from j=i+1 to j=size-1

comparison

How many comparisons are done in total?

1st iteration 2nd iter. 3rd iter
 $(n-1) + (n-2) + (n-3) + \dots + 3 + 2 + 1 = \text{total}$

$$1 + 2 + 3 + \dots + (n-3) + (n-2) + (n-1) = \text{total}$$

$$n + n + n + \dots + n + n + n = 2 \cdot \text{total}$$

$$(n-1)n = 2 \cdot \text{total} \quad \text{total} = \frac{n(n-1)}{2} = \frac{n^2}{2} - \frac{n}{2}$$

Version 2:

i loop goes from i=0 to size-2

comparison

How many comparisons are done in total? $n-1$

Version 3

i loop goes from i=0 to size-2

10k comparisons

How many comparisons are done in total? $10,000(n-1) = 10,000n - 10,000$

version 1: $\frac{n^2}{2} - \frac{n}{2}$

version 2: $n-1$ ← best

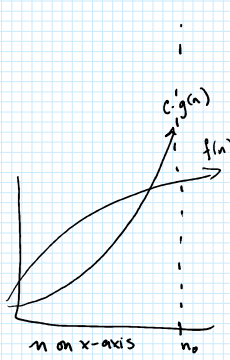
version 3: $10000n - 10000$

version 3 seems better than version 2 for bigger inputs

Definition of big-O:

Let $f(n)$ and $g(n)$ be functions. We say that $f(n)$ is $O(g(n))$

if there exist a constant $c > 0$ and a constant $n_0 \geq 1$ such that $f(n) \leq c \cdot g(n)$ for all $n \geq n_0$.



We say $f(n)$ is asymptotically upper-bounded by $g(n)$.

Version 1: $\frac{n^2}{2} - \frac{n}{2}$ comparisons

want to show this is $O(n^2)$

$$c = \frac{1}{2} \quad c = 100$$

$$n_0 = 1$$

need to check:
 for $n \geq n_0 = 1$, $\frac{n^2}{2} - \frac{n}{2} \leq 1 \cdot n^2$
 $\frac{n^2}{2} - \frac{n}{2} \leq \frac{n^2}{2} \leq n^2 = 1 \cdot n^2 \leq 100n^2$
 (since $n \geq 1$, so $-\frac{n}{2} \leq 0$)

Version 2: $n-1$ comparisons

- Outline
- review classes of algorithms
 - review def of big O
 - proofs of big O
 - sorting
 - o selection sort
 - o mergesort
 - o big O analysis of each

first test - weeks 1 and 2, C++ but not bigO (we only test you on things where you did the lab and got a grade back) (the study guide auto-hides answers so you can check yourself)

- Review classes of algorithms
1. what class grows proportionally to the size of the problem? linear $O(n)$
 2. what's the fastest class of algorithms? constant time $O(1)$
 3. what's an example of a quadratic algorithm? selection sort, bubble sort, insertion sort $O(n^2)$
 4. which class of algorithms is the slowest? factorial $O(n!)$
 5. what class do the fastest sorting algorithms fall into? $O(n \log n)$

Definition of big-O
 We say that $f(n)$ is $O(g(n))$ if there exist constants $c > 0$ and $n_0 \geq 1$ such that $f(n) \leq c \cdot g(n)$ for all $n \geq n_0$.

Proofs

Example 1: Show that $f(n) = n^2 + 3n + 1$ is $O(n^2)$.
 We know $n > 0$ as it's the size of the problem, and n is an integer. We also know $n^2 \leq n^2$ and $3n \leq 3n^2$ and $1 \leq n^2$, so $n^2 + 3n + 1 \leq n^2 + 3n^2 + n^2 = 5n^2$... so as long as we choose a constant $c \geq 5$ and $n_0 \geq 1$, we can say that $f(n)$ is $O(g(n))$ because $n^2 + 3n + 1 \leq 5n^2$.

Example 3: Show that $f(n) = 4n^4 - 5n^3 + 6n - 7n + 8$ is $O(n^4)$ assuming $n > 0$.
 We know $0 > -5n^3$ and $0 > -7n$. So $4n^4 - 5n^3 + 6n^2 - 7n + 8 \leq 4n^4 - 0 + 6n^2 - 0 + 8n^4 = 10n^4$. So as long as $c \geq 10$ and $n_0 \geq 1$, we know that $f(n)$ is $O(n^4)$.

Example 4: Is $f(n) = 4n^2 \dots$
 $O(n^2)$? yes
 $O(n^3)$? yes
 $O(n^{100})$? yes

Remember that big O provides an upper bound, but the definition does not require it to be a tight upper bound. Obviously we prefer tighter upper bounds because they give us a better sense of the algorithm's efficiency. Example 5: suppose $f(n) = n^2$ and $g(n) = 7n^2 + 3n$, which of the following is the best answer: [discussion about why we do this, is this good, generally we just say $O(n^2)$ and not anything weirder like $O(7n^2 + 4)$]. Goal: take an unsorted array of elements and rearrange them such that they are in ascending order.

Sanity check:

if $n = 1000$ & then $n = 3000$, how much longer will this take?
 If our analysis is right, a problem 3 times larger should take 9 times longer

```
onscreen demo
./sortTest 1000 selectSort
output shows the last 10 elements as a sanity check for "did it sort correctly?"

time ./sortTest 1000 selectSort
takes .009s
triple 4
takes .02s
10,000 took 0.1s
30,000 took 0.9s
so you have to sometimes pick a big enough input size to have the difference in runtime actually show.
```

```
Mergesort
• example of a divide-and-conquer style algorithm (think: like binary search)
• typically have recursive solutions
  o base case, where no recursion is necessary
  o recursive cases, where we use the same approach on smaller versions of the problem
• conquer by putting separate results from recursion back together

reminder: a recursive function is a function that calls itself

mergeSort(array, size):
if size <= 2:
    return // you're all done! that array is definitely sorted
copy the first half of the array into a new array B
copy the second half of the array into a new array C
mergeSort(B, size/2)
mergeSort(C, size/2)
merge(B, C, array) // helper function merge takes two arrays and puts them back in sorted order in the original array.

merge function:
```

$$c = \frac{1}{n_0} \quad c = 100$$

$$\frac{n^2}{2} = \frac{n}{2} \leq \frac{n^2}{2} \leq n^2 = 1 \cdot n^2 \leq 100n^2$$

↑ \downarrow $c \cdot n \geq 1, \text{ so } -\frac{n}{2} \leq 0$

Version 2: $n-1$ comparisons

want to show this is $O(n)$

need to check:
for $n \geq n_0 = 1, n-1 \leq 1 \cdot n$

$$c = \frac{1}{n_0} \quad c = 5$$

$$n-1 \leq n \quad \text{for } n \geq 1 \quad \checkmark$$

$$n_0 = 1 \quad n = 18$$

$$\text{for } n \geq 18, n-1 \leq 5n \quad \checkmark$$

Version 3: $10,000n - 10,000$ comparisons

want to show this is $O(n)$

need to check:
for $n \geq n_0 = 1, 10,000n - 10,000 \leq 8 \cdot n$

$$c = \frac{10,001}{n_0} \quad c = 8$$

$$10,000n - 10,000 \leq 10,000n \leq 10,001 \cdot n \quad \checkmark$$

$$(10,000 - 8)n \leq 10,000$$

Version 1: $O(n^2)$

Version 2: $O(n)$

Version 3: $O(n)$

Classes of algorithms by runtime (fastest to slowest)

- constant $O(1)$
ex: return the last element in an array
- logarithmic $O(\log_2 n)$
ex: binary search
- linear $O(n)$
ex: is-sorted version 2, 3; linear
- $O(n \log n)$
ex: mergesort
- quadratic $O(n^2)$
ex: is-sorted version 1, selection sort, bubblesort, insertion sort
- $O(n^3)$
- exponential $O(2^n)$
- factorial $O(n!)$

big O practice:

Definition of big-O:

Let $f(n)$ and $g(n)$ be functions.
We say that $f(n)$ is $O(g(n))$

if there exist a constant $c > 0$
and a constant $n_0 \geq 1$ such that
 $f(n) \leq c \cdot g(n)$ for all $n \geq n_0$.

To prove that
 $f(n)$ is $O(g(n))$
we need to find
values for c and n_0
that satisfy the definition.

Ⓐ Show that $f(n) = n^2 + 6n + 2$ is $O(n^2)$.

Assume: $c = \frac{9}{3} \quad n_0 = 3$

Want to show: $n^2 + 6n + 2 \leq 9 \cdot n^2$ for $n \geq n_0 = 3$.

$$n^2 + 6n + 2 \leq n^2 + 6n^2 + 2 \quad \text{if } 6n \leq 6n^2 \text{ for } n \geq 3$$

$$\leq n^2 + 6n^2 + 2n^2 \quad \text{if } 2 \leq 2n^2 \text{ for } n \geq 3$$

$$= 9n^2 \quad \checkmark$$

Ⓑ Show that $f(n) = 4n^5 - 3n^4 + 8n^3 - 7n^2 + 12$
is $O(n^5)$. (Safe to assume n always > 0 .)

Ⓒ The function $f(n) = 20n^3$ is ...

- $O(n^2)$? no $c = 20$
- $O(n^3)$? yes $n_0 = 2$
- $O(n^4)$? yes $c = 20$
- $O(n^5)$? yes $n_0 = 1$

$O(n^{100})$? yes

In general, we want our O bounds to be tight (close to the function f)
in order to be useful.

Ⓓ Let $f(n) = n^2$ and $g(n) = 8n^2 + n$.
What best describes these functions?

copy one n into an array into a new array c
copy the second half of the array into a new array c
mergesort($B, \text{size}/2$)
mergesort($C, \text{size}/2$)

this is the DIVIDING

merge(B, C, array) // helper function merge takes two arrays and puts them back in sorted order in the original array.

merge function:

live blackboard crossbreeding,
with giant playing cards!

```
merge(A, sizeA, B, sizeB, C):
  i=0, j=0, k=0 // indices for A, B, and C, respectively
  while i<sizeA and j<sizeB:
    if A[i] <= B[j]:
      C[k++] = A[i++] // this uses the values k and i, then increments each of them
    else:
      C[k++] = B[j++]
  while i<sizeA: // to handle the leftover items if they are in A
    C[k++] = A[i++]
  while j<sizeB: // ditto
    C[k++] = B[j++]
```

live class demo of mergesort w/ students
(as stack frames)

how much work to get to base case? $O(\log_2 n)$

how much work to merge 2 lists of size $n/2$? $O(n)$

$O(n \log_2 n)$ work overall

1. $f(n)$ is $O(g(n))$ ← true
2. $g(n)$ is $O(f(n))$ ← true $8n^2 + n \leq C \cdot n^2$ for $n \geq n_0 = 1$
3. both 1 and 2 ← best answer!
4. neither 1 nor 2

SORTING

problem: take an unsorted array of elements and rearrange them to be in ascending (increasing) order

Selection Sort: pseudocode

```

SelectSort(array, size)
  for i = size - 1 down to 1
    indexOfMax = findMax(array, i)
    swap(array, i, indexOfMax)

```

```

findMax(array, end)
  indexOfMax = 0
  for i = 1 to end
    if array[i] > array[indexOfMax]
      indexOfMax = i
  return indexOfMax

```

example run of selection sort:

① [7, 6, 9, 1, 3, 5, 4]