UNIT 1: Fundamentals of the Analysis of Algorithmic Efficiency...

Mathematical Analysis of Non-recursive Algorithms

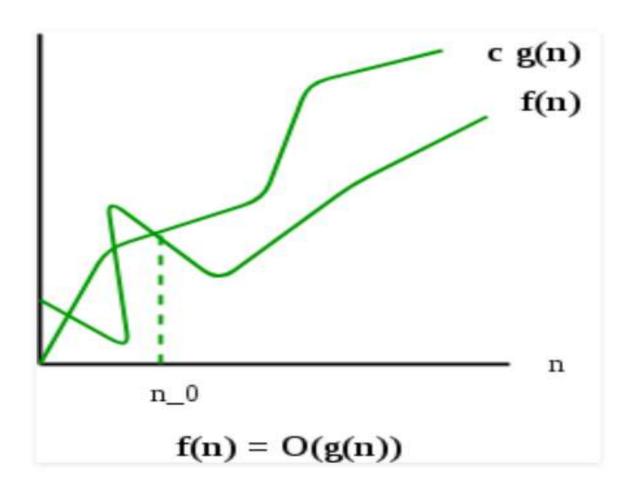
O (Big-O)notation

Definition:

A function f(n) is said to be in O(g(n)), denoted $f(n) \in O(g(n))$,

if f(n) is **bounded above** by some constant multiple of g(n) for all large n, i.e., if there exist some positive constant c and some non-negative integer n_0 such that

 $f(n) \le cg(n)$ for all $n \ge n_0$



Example: O notation

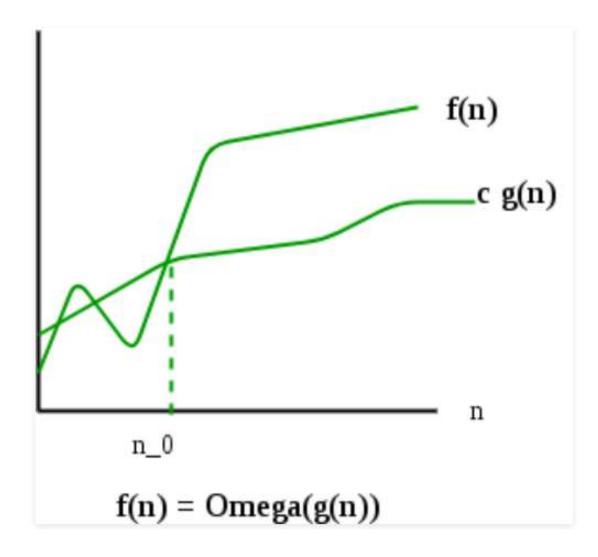
Ω (Omega) notation

Definition:

A function f(n) is said to be in $\Omega(g(n))$, denoted $f(n) \in \Omega(g(n))$,

if f(n) is **bounded below** by some positive constant multiple of g(n) for all large n, i.e., if there exist some positive constant c and some non-negative integer n_0 such that

 $f(n) \ge cg(n)$ for all $n \ge n_0$



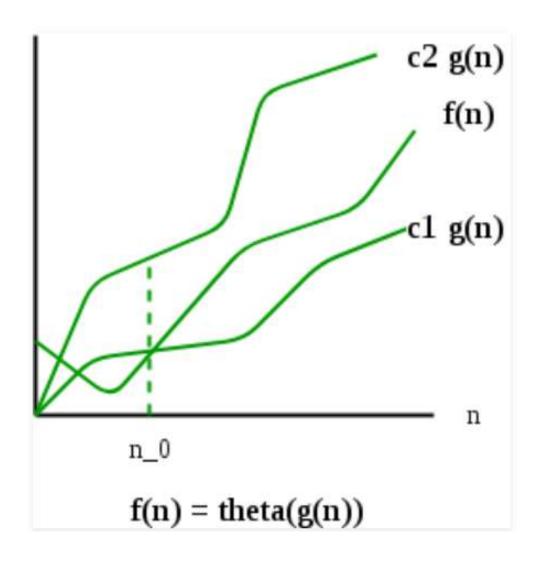
Θ (Theta) notation

Definition:

A function f(n) is said to be in $\Theta(g(n))$, denoted $f(n) \in \Theta(g(n))$,

if f(n) is **bounded both above and below** by some positive constant multiples of g(n) for all large n, i.e., if there exist some positive constant c_1 and c_2 and some non-negative integer n_0 such that

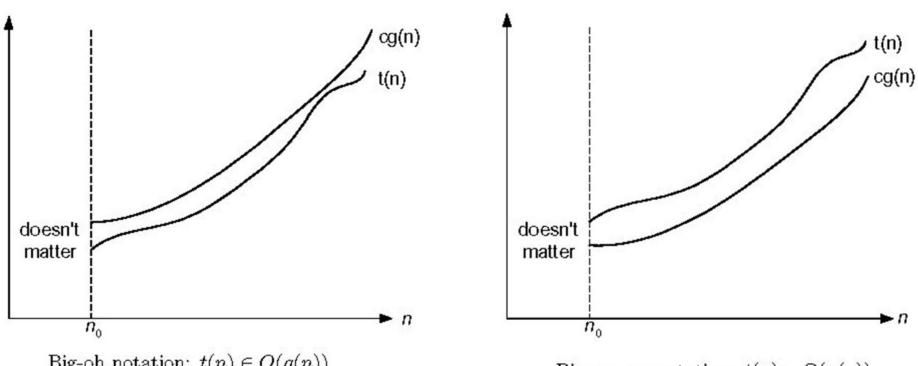
 $c_2g(n) \le f(n) \le c_1g(n)$ for all $n \ge n_0$



Asymptotic order of growth

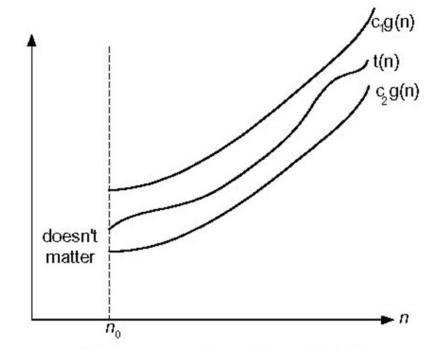
A way of comparing functions that ignores constant factors and small input sizes

- O(g(n)): class of functions f(n) that grow no faster than g(n)
- Θ(g(n)): class of functions f(n) that grow at same rate as g(n)
- $\Omega(g(n))$: class of functions f(n) that grow at least as fast as g(n)



Big-oh notation: $t(n) \in O(g(n))$

Big-omega notation: $t(n) \in \Omega(g(n))$



Big-theta notation: $t(n) \in \Theta(g(n))$

General plan for analyzing efficiency of non-recursive algorithms

- 1. Decide on parameter/s *n* indicating input's size
- 2. Identify algorithm's basic operation
- 3. Check/Determine worst, average, and best cases for input of size n
- 4. Set up a sum for the number of times the basic operation is executed
- Simplify the sum using standard formulas and rules

Example 1: Maximum element

Finding the value of the largest element in a list of n numbers.

Example 1: Maximum element

```
ALGORITHM MaxElement(A[0..n-1])

//Determines the value of the largest element in a given array
//Input: An array A[0..n-1] of real numbers
//Output: The value of the largest element in A

maxval \leftarrow A[0]

for i \leftarrow 1 to n-1 do

if A[i] > maxval

maxval \leftarrow A[i]

return maxval
```

Example 1: Maximum element Algorithm analysis

- 1. input's size: n number of elements in the array
- 2. basic operation: comparison and assignment

if
$$A[i] > maxval$$
 $maxval \leftarrow A[i]$

- 3. No worst, average, and best cases
- 4. Let C(n) = number of times the basic operation is executed.

Algorithm makes one comparison on each iteration of loop, which runs for i=1 to (n-1). Therefore,

$$C(n) = \sum_{i=1}^{n-1} 1$$

Simplifying the sum using standard formulas we get:

$$C(n) = \sum_{i=1}^{n-1} 1$$

$$\sum_{i=1}^{n} 1 = n - 1 + 1$$

= n-1 ∈ Θ(n)

upper bound minus the lower bound plus one

Let's check our understanding...

```
ALGORITHM Mystery(n)

//Input: A nonnegative integer n

S ← 0

for i ←1 to n do

S ← S + i * i

return S
```

- a. What does this algorithm compute?
- b. What is its basic operation?
- c. How many times is the basic operation executed?
- d. What is the efficiency class of this algorithm?

Let's check our understanding...

```
Algorithm Mystery(n)

//Input: A nonnegative integer n

S \leftarrow 0

for i \leftarrow 1 to n do

S \leftarrow S + i * i

return S
```

computes the sum of 'squares of numbers' from 1 to n, i.e., $S = 1*1 + 2*2 + 3*3 \dots + n*n$

What does this algorithm compute?

A.
$$n^2$$

B.
$$\sum_{i=1}^{n} i$$

C.
$$\sum_{i=1}^{n} i^2$$

D.
$$\sum_{i=1}^{n} 2i$$

Example 2: Unique elements

Check whether all the element in a given list are distinct.

Example 2: Unique elements

```
ALGORITHM UniqueElements (A[0..n-1])

//Determines whether all the elements in a given array are distinct

//Input: An array A[0..n-1]

//Output: Returns "true" if all the elements in A are distinct

// and "false" otherwise

for i \leftarrow 0 to n-2 do

for j \leftarrow i+1 to n-1 do

if A[i] = A[j] return false

return true
```

Example 2: Unique elements Algorithm analysis

- input's size: n array size
- 2. basic operation: comparison

if
$$A[i] = A[j]$$

- 3. Worst, average, and best cases may exists (depends on implementation)
- 4. Let C(n) = number of times the basic operation is executed.

Example 2: Unique elements

```
ALGORITHM UniqueElements (A[0..n-1])

//Determines whether all the elements in a given array are distinct

//Input: An array A[0..n-1]

//Output: Returns "true" if all the elements in A are distinct

// and "false" otherwise

for i \leftarrow 0 to n-2 do

for j \leftarrow i+1 to n-1 do

if A[i] = A[j] return false

return true

Flag = 1;
```

Example 2: Unique elements Algorithm analysis

- input's size: n array size
- 2. basic operation: comparison

if
$$A[i] = A[j]$$

- 3. No worst, average, and best cases
- 4. Let C(n) = number of times the basic operation is executed.

Algorithm makes one comparison on each iteration of inner for loop, which runs for j=i+1 to (n-1); which in turn runs for each iteration of outer for loop with i=0 to n-2. Therefore,

$$C(n) = \sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1$$

Simplifying the sum using standard formulas we get:

$$C(n) = \sum_{i=0}^{n-2} [(n-1) - (i+1) + 1]$$

$$= n(n-1)/2 \in \Theta(n^2)$$

$$C(n) = \sum_{i=0}^{n-2} [(n-1) - (i+1) + 1]$$

$$=\sum_{i=0}^{n-2} n - 1 - i$$

$$=\sum_{i=0}^{n-2}(n-1)-\sum_{i=0}^{n-2}i$$

$$= (n-1)\sum_{i=0}^{n-2} 1 - \sum_{i=0}^{n-2} i$$

$$= (n-1)\sum_{i=0}^{n-2} 1 - \sum_{i=0}^{n-2} i$$

$$= (n-1)(n-2-0+1) - \sum_{i=0}^{n-2} i$$

$$= (n-1)^2 - \sum_{i=0}^{n-2} i$$

$$= (n-1)^2 - \frac{(n-2)(n-1)}{2}$$

$$\sum_{i=1}^{n} i = \frac{n(n+1)}{2}$$

$$= (n-1)^2 - \frac{(n-2)(n-1)}{2}$$

$$= (n-1) ((n-1) - \frac{(n-2)}{2})$$

$$=(n-1)(\frac{2n-2-n+2}{2})$$

$$=(n-1)(\frac{n}{2}) = \frac{n(n-1)}{2} = \frac{1}{2}n^2$$

$$=\Theta(n^2)$$

Example 3: Matrix multiplication

Example 2: Matrix multiplication

```
ALGORITHM MatrixMultiplication(A[0..n-1, 0..n-1], B[0..n-1, 0..n-1])

//Multiplies two n-by-n matrices by the definition-based algorithm

//Input: Two n-by-n matrices A and B

//Output: Matrix C = AB

for i \leftarrow 0 to n-1 do

C[i, j] \leftarrow 0.0

for k \leftarrow 0 to n-1 do

C[i, j] \leftarrow C[i, j] + A[i, k] * B[k, j]

return C
```

Example 3: Matrix multiplication Algorithm analysis

- 1. input's size: n matrix order
- basic operation: multiplication, addition, assignment

$$C[i, j] \leftarrow C[i, j] + A[i, k] * B[k, j]$$

- 3. No worst, average, and best cases
- 4. Let C(n) = number of times the basic operation is executed.

Algorithm makes one multiplication on each iteration of innermost for loop, which runs for k=0 to n-1; which in turn runs for each iteration of j=0 to n-1; which in turn runs for each iteration of outer for loop with i=0 to n-1. Therefore,

$$C(n) = \sum_{i=0}^{n-1} \sum_{j=0}^{n-1} \sum_{k=0}^{n-1} 1$$

Simplifying the sum using standard formulas we get:

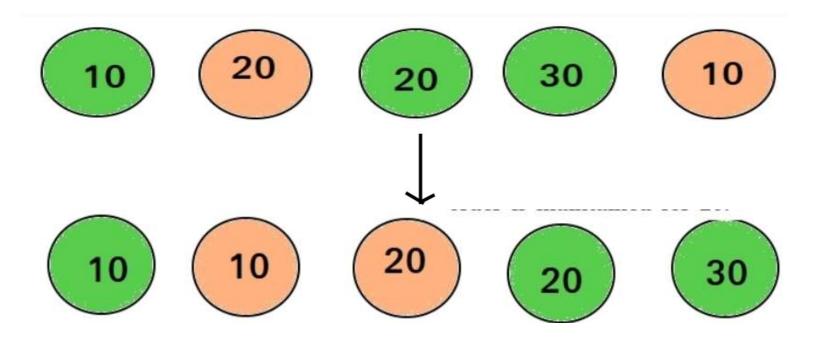
$$= n^3 \in \Theta(n^3)$$

Brute force

A straightforward approach to solving a problem, usually directly based on the problem's statement and definitions of the concepts involved.

Stable sorting algorithm

A sorting algorithm is said to be **stable** if two objects with equal keys appear in the same order in sorted output as they appear in the input unsorted array.



Example: Insertion Sort, Merge Sort, Bubble Sort

In-place sorting algorithm

Algorithm that do not use extra space for manipulating the input but may require a small though non constant extra space for its operation

Example:

Bubble sort, Selection sort, Insertion sort,

Heapsort, Shell sort

Selection sort

- in-place comparison-based algorithm
- performs well on a small list
- works by repeatedly going through the list of items, each time selecting an item according to its ordering and placing it in the correct position in the sequence.

$$A_0 \le A_1 \le \dots \le A_{i-1} \mid A_i, \dots, A_{min}, \dots, A_{n-1}$$

in their final positions the last n - i elements

Selection sort

```
ALGORITHM SelectionSort(A[0..n-1])
    //Sorts a given array by selection sort
    //Input: An array A[0..n-1] of orderable elements
    //Output: Array A[0..n-1] sorted in ascending order
    for i \leftarrow 0 to n-2 do
         min \leftarrow i
         for j \leftarrow i + 1 to n - 1 do
             if A[j] < A[min]
                  min \leftarrow j
         swap A[i] and A[min]
```

Visualization tools

https://visualgo.net/bn/sorting



https://www.cs.usfca.edu/~galles/visualization/ComparisonSort.html



Selection sort: Algorithm analysis

- 1. input's size: n Number of elements to be sorted
- 2. basic operation: key comparison, swapping

if
$$A[j] < A[min] \quad min \leftarrow j$$

swap $A[i]$ and $A[min]$

- 3. No worst, average, and best cases
- 4. Let C(n) = number of times the basic operation is executed.

Algorithm makes one key comparison on each iteration of innermost for loop, which runs for j=i+1 to n-1; which in turn runs for each iteration of outer for loop with i=0 to n-2. Therefore,

$$C(n) = \sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1$$

Simplifying the sum using standard formulas we get:

$$=\frac{(n-1)n}{2}\in\Theta(\mathsf{n}^2)$$

Selection sort

Disadvantage:

- poor efficiency when dealing with a huge list of items
- its performance is easily influenced by the initial ordering of the items before the sorting process. Thus, it is only suitable for a list of few elements that are in random order.

Let's check our understanding...

```
ALGORITHM BubbleSort(A[0..n-1])

//Sorts a given array by bubble sort

//Input: An array A[0..n-1] of orderable elements

//Output: Array A[0..n-1] sorted in ascending order

for i \leftarrow 0 to n-2 do

for j \leftarrow 0 to n-2-i do

if A[j+1] < A[j]

swap A[j] and A[j+1]
```

Next session...

Fundamentals of the Analysis of Algorithmic Efficiency...

Mathematical Analysis of Recursive algorithms