



Algorithm Efficiency

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1



Contents

- A review on algorithm
- Analysis and Big-O notation
- Algorithm efficiency

3

A review on algorithm

What is Algorithm?

- An algorithm is
 - a strictly defined **finite** sequence of **well-defined** steps (statements, often called instructions or commands)
 - that provides the solution to a problem.

Algorithm

- Give some examples of algorithms.

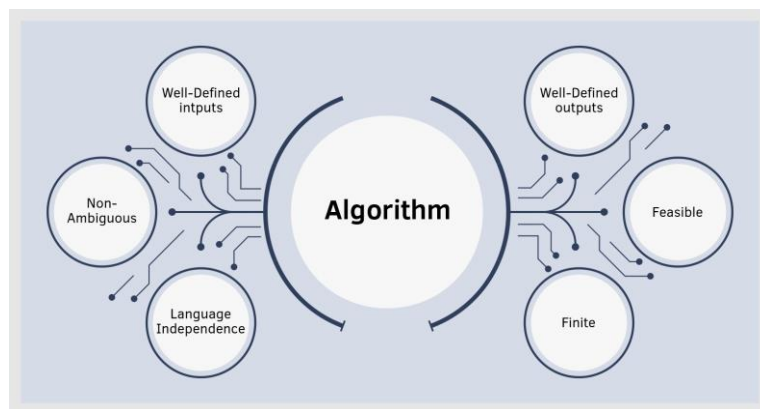
An Example

- Input: No
- Output: what do you think about the output?
- Step 1. Assign $\text{sum} = 0$. Assign $i = 0$.
- Step 2.
 - Assign $i = i + 1$
 - Assign $\text{sum} = \text{sum} + i$
- Step 3. Compare i with 10
 - if $i < 10$, back to step 2.
 - otherwise, if $i \geq 10$, go to step 4.
- Step 4. return sum

Characteristics of Algorithms

- **Finiteness**
For any input, the algorithm must terminate after a finite number of steps.
- **Correctness**
Always correct. Give the same result for different run time.
- **Definiteness**
All steps of the algorithm must be precisely defined.
- **Effectiveness**
It must be possible to perform each step of the algorithm correctly and in a finite amount of time.

Characteristics of Algorithms



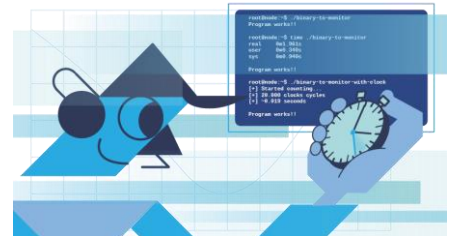
Algorithm Efficiency

- The two factors of Algorithm Efficiency are:
 - **Time Factor:** Time is measured by counting the number of key operations.
 - **Space Factor:** Space is measured by counting the maximum memory space required by the algorithm.

Measuring Efficiency of Algorithms

- Can we compare two algorithms (in time factor) like this?
 - Implement those algorithms (into programs)
 - Calculate the execution time of those programs
 - Compare those two values of time measurement.

Is it fair in this measuring process?



Measuring Efficiency of Algorithms

- Difficulties with comparing programs instead of algorithms
 - How are the algorithms coded?
 - What computer should you use?
 - What data should the programs use?

Measuring Efficiency of Algorithms

- Comparison of algorithms should focus on **significant differences** in efficiency
- Employ mathematical techniques that analyze algorithms **independently** of specific implementations, computers, or data.

Execution Time of Algorithm

- Time complexity is measured by counting the **primitive operations** for the computation that the algorithm needs to perform.
 - Comparisons
 - Assignments
- Derive an algorithm's time requirement as a function of **the problem size**
 - Algorithm A requires $n^2/5$ time unit to solve a problem of size n .
 - Algorithm B requires $5 \times n$ time unit to solve a problem of size n .

Execution Time of Algorithm

- Traversal of linked nodes – example:

```

Node<ItemType>* curPtr = headPtr;      ← 1 assignment
while (curPtr != nullptr)             ← n + 1 comparisons
{
    cout << curPtr->getItem() < endl;    ← n writes
    curPtr = curPtr->getNext();          ← n assignments
} // end while
    
```

- Assignment: a time units.
- Comparison: c time units.
- Write: w time units.
- Displaying data in linked chain of n nodes requires time proportional to n

Execution Time of Algorithm

- Nested loops

```
for (i = 1 through n)
  for (j = 1 through i)
    for (k = 1 through 5)
      Task T
```

- Task T requires t time units.

Previous Example

- **Step 1.** Assign $\text{sum} = 0$. Assign $i = 0$.
- **Step 2.**
 - Assign $i = i + 1$
 - Assign $\text{sum} = \text{sum} + i$
- **Step 3.** Compare i with **10**
 - if $i < 10$, back to step 2.
 - otherwise, if $i \geq 10$, go to step 4.
- **Step 4.** Return sum

How many

- Assignments?
- Comparisons?

Another Example

- **Step 1.** Assign $\text{sum} = 0$. Assign $i = 0$.
- **Step 2.**
 - Assign $i = i + 1$
 - Assign $\text{sum} = \text{sum} + i$
- **Step 3.** Compare i with n
 - if $i < n$, back to step 2.
 - otherwise, if $i \geq n$, go to step 4.
- **Step 4.** Return sum

How many

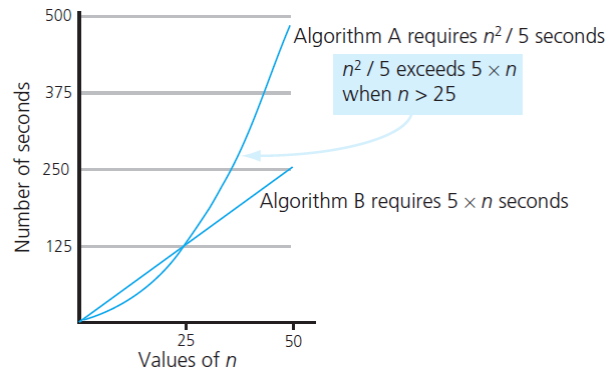
- Assignments?
- Comparisons?

Algorithm Growth Rates

- Measure algorithm's time requirement **as a function** of problem size
- Compare algorithm efficiencies for **large problems**
- Look only at **significant differences**.

Algorithm Growth Rates

- Time requirements as a function of the problem size n



Analysis and Big O Notation

Big O Notation

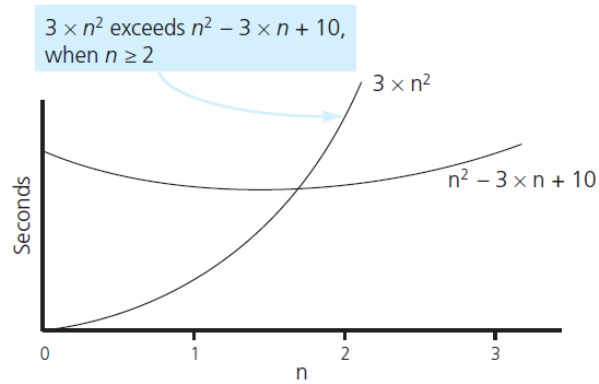
○ Definition:

- Algorithm A is order $f(n)$
 - Denoted $O(f(n))$
- If constants k and n_0 exist
- Such that A requires **no more** than $k \times f(n)$ time units to solve a problem of size $n \geq n_0$.

Example

- An algorithm requires $n^2 - 3 \times n + 10$ (time units). What is the order of algorithm?
 - Hint: Find the values k va n_0 .

Example



The graphs of $3 \times n^2$ and $n^2 - 3 \times n + 10$

Another Example

- How about the order of an algorithm requiring
 $(n + 1) \times (a + c) + n \times w$
time units?

Another Example

- Another algorithm requires $n^2 + 3 \times n + 2$ time units. What is the order of this algorithm?

Common Growth-Rate Functions

- $f(n) =$
 - 1: Constant
 - $\log_2 n$: Logarithmic
 - n : Linear
 - $n \times \log_2 n$: Linearithmic
 - n^2 : Quadratic
 - n^3 : Cubic
 - 2^n : Exponential

Common Growth-Rate Functions

- Order of growth of some common functions

$$O(1) < O(\log_2 n) < O(n) < O(n \times \log_2 n) < O(n^2) < O(n^3) < O(2^n)$$

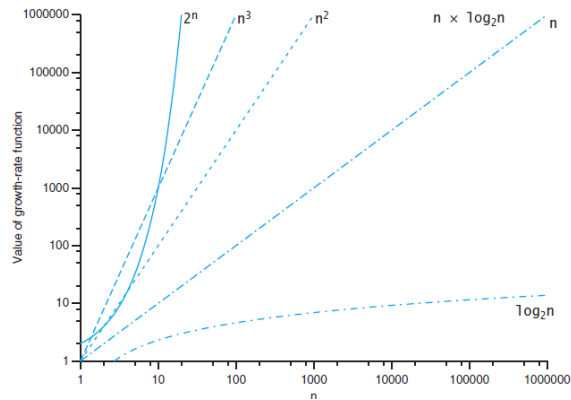
Common Growth-Rate Functions

- A comparison of growth-rate functions in **tabular form**

Function	n					
	10	100	1,000	10,000	100,000	1,000,000
1	1	1	1	1	1	1
$\log_2 n$	3	6	9	13	16	19
n	10	10^2	10^3	10^4	10^5	10^6
$n \times \log_2 n$	30	664	9,965	10^5	10^6	10^7
n^2	10^2	10^4	10^6	10^8	10^{10}	10^{12}
n^3	10^3	10^6	10^9	10^{12}	10^{15}	10^{18}
2^n	10^3	10^{30}	10^{301}	$10^{3,010}$	$10^{30,103}$	$10^{301,030}$

Common Growth-Rate Functions

- A comparison of growth-rate functions in **graphical form**



Properties of Growth-Rate Functions

- Ignore low-order terms
- Ignore a multiplicative constant in the high-order term
- $O(f(n)) + O(g(n)) = O(f(n) + g(n))$

Some Useful Results

○ Constant Multiplication:

- If $f(n)$ is $O(g(n))$ then $c.f(n)$ is $O(g(n))$, where c is a constant.

○ Polynomial Function:

- $f(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$ is $O(x^n)$.

Some Useful Results

○ Summation Function:

- If $f_1(n)$ is $O(g_1(n))$ and $f_2(n)$ is $O(g_2(n))$
- Then $f_1(n) + f_2(n)$ is $O(\max(g_1(n), g_2(n)))$

○ Multiplication Function:

- If $f_1(n)$ is $O(g_1(n))$ and $f_2(n)$ is $O(g_2(n))$
- Then $f_1(n) \times f_2(n)$ is $O(g_1(n) \times g_2(n))$

Quiz

Are these functions of order $O(x)$?

- a) $f(x) = 10$
- b) $f(x) = 3x + 7$
- c) $f(x) = 2x^2 + 2$

Quiz

What are the order of the following functions?

- $f(n) = (2 + n) * (3 + \log_2 n)$
- $f(n) = 11 * \log_2 n + n/2 - 3542$
- $f(n) = n * (3 + n) - 7 * n$
- $f(n) = \log_2(n^2) + n$

Notes

- Use like this:
 - $f(x)$ is $O(g(x))$, or
 - $f(x)$ is of order $g(x)$, or
 - $f(x)$ has order $g(x)$

Algorithm Efficiency

Algorithm Efficiency

- Best case scenario
- Worst case scenario
- Average case scenario

An Algorithm to Analyze

- Input:
- Output:
- **Step 1.** Set the first integer the temporary maximum value (temp_max).
- **Step 2.** Compare the current value with the temp_max.
 - If it is greater than, assign the current value to temp_max.
- **Step 3.** If there is other integer in the list, move to next value. Back to step 2.
- **Step 4.** If there is no more integer in the list, stop.
- **Step 5.** return temp_max (the maximum value of the list).

Another Algorithm to Analyze

- Input:
- Output:

- Step 1. Assign $i = 0$
- Step 2. While $i < n$ and $x \neq a_i$, increase i by 1.
 while ($i < n$ and $x \neq a_i$)
 $i = i + 1$
- Step 3.
 - If $i < n$, return i .
 - Otherwise ($i \geq n$), return -1 to tell that x does not exist in list a .

Another Algorithm to Analyze

- Use comparisons for counting.

- Worst case:
 - When it occurs?
 - How many operations?

- Best case:
 - When it occurs?
 - How many operations?

Another Algorithm to Analyze

- Use comparisons for counting.
- Average case:
 - If x is found at position i^{th} , the number of comparisons is $2i + 1$.
 - The average number of comparisons is:

$$\frac{3+5+7+\dots+(2n+1)}{n} = \frac{2(1+2+3+\dots+n)+n}{n} = \frac{2 \frac{n(n+1)}{2} + n}{n} = n + 2$$

Analysis of Algorithms

- Decide **n** – the input size
- Identify the algorithm's **basic operation** (as a rule, it is in the innermost loop)
- Check whether the number of times the basic operation is executed depends only on n
 - If it depends on some additional property, specify the **worst-case** for Big-Oh
- Set up a **sum** expressing the number of times the algorithm's basic operation is executed.
- Find a closed-form formula for the count and **establish its order of growth**.

Analysis of Algorithms

- **Example:** Check whether all the elements in a given array of n elements are distinct.

```

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 ← 0 to n - 2 do
    for j ← i + 1 to n - 1 do
        if A[i] = A[j]
            return false
    return true
    
```

Basic operation

Analysis of Algorithms

- **Worst-case:**

$$\begin{aligned}
 C_{worst}(n) &= \sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1 = \sum_{i=0}^{n-2} [(n-1) - (i+1) + 1] = \sum_{i=0}^{n-2} (n-1-i) \\
 &= \sum_{i=0}^{n-2} (n-1) - \sum_{i=0}^{n-2} i = (n-1) \sum_{i=0}^{n-2} 1 - \frac{(n-2)(n-1)}{2} \\
 &= (n-1)^2 - \frac{(n-2)(n-1)}{2} = \frac{(n-1)n}{2} \approx \frac{1}{2}n^2
 \end{aligned}$$

Keeping Your Perspective

- If problem size always small, ignore an algorithm's efficiency
- Weigh trade-offs between algorithm's time and memory requirements
- Compare algorithms for both style and efficiency

Exercises

Exercise

- Propose an algorithm to calculate the value of S defined below. What order does the algorithm have?

$$S = 1 + \frac{1}{2} + \frac{1}{6} + \dots + \frac{1}{n!}$$

- How many comparisons, assignments are there in the following code fragment with the size n ?

```
sum = 0;
for (i = 0; i < n; i++)
{
    std::cin >> x;
    sum = sum + x;
}
```

Exercise

How many assignments are there in the following code fragment with the size n ?

```
for (i = 0; i < n ; i++)
    for (j = 0; j < n; j++)
    {
        C[i][j] = 0;
        for (k = 0; k < n; k++)
            C[i][j] = C[i][j] + A[i][k]*B[k][j];
    }
```


Exercise

- Give the order of growth (as a function of N) of the running time of the following code fragment:

```
int sum = 0;
for (int n = N; n > 0; n /= 2)
    for (int i = 0; i < n; i++)
        sum++;
```

Exercise

- Give the order of growth (as a function of N) of the running time of the following code fragment:

```
int sum = 0;
for (int i = 1; i < N; i *= 2)
    for (int j = 0; j < i; j++)
        sum++;
```

Exercise

- Give the order of growth (as a function of N) of the running time of the following code fragment:

```
int sum = 0;
for (int i = 1; i < N; i *= 2)
    for (int j = 0; j < N; j++)
        sum++;
```

Questions and Answers