

CS202 - Algorithm Analysis

How to Analyze Algorithms?

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Follow-up

S

156	141	35	94	88	61	111	77
1	2	3	4	5	6	7	8



Algorithm - Find Least Played Song (S)

Input - A set of play counts associated with a variety of songs inside a playlist.

Output - The least played song.

```
1:  $temp \leftarrow S[0]$ 
2:  $res \leftarrow 1$ 
3: for  $i = 1$  to  $|S|$  do
4:   if  $S[i] < temp$  then
5:      $temp \leftarrow S[i]$ 
6:      $res \leftarrow i + 1$ 
7:   end if
8: end for
9: return  $res$ ;
```

Discussion Based On ...

Sedgewick 1.4

The three important techniques ...

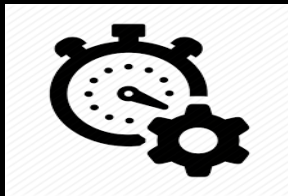
- **Algorithm:** Outline, the essence of a computational procedure, step by step instructions.
- **Program:** An implementation of an algorithm in some programming language such as Java, C, C++, Python, etc ...
- **Data structure:** Organization of data needed to solve the problem.



Algorithm Design Principles ...



Correctness



Efficiency

How is an Algorithm different from a Program?

- Makes use of high level description of the algorithm , instead of testing one of its implementations
- Take into account all possible inputs
- Allows one to evaluate the efficiency of any algorithm in a way that is independent of the hardware and software environment

How do we measure the running time of a Program?

- First write a program to implement the algorithm.
- Execute the program using datasets of varying size and composition.
- Integrate a Python method like **time.time()** to get an accurate measure of the actual running time.

Look at timer.py in repo

Limitation of the Experimental Study approach!

- It is necessary to implement and test the algorithm in order to determine the running time.
- Experiments can be done only on a limited set of inputs, and may not be indicative of the running time on other inputs not included in the experiment.
- In order to compare two algorithms, the same hardware and software environments should be used.

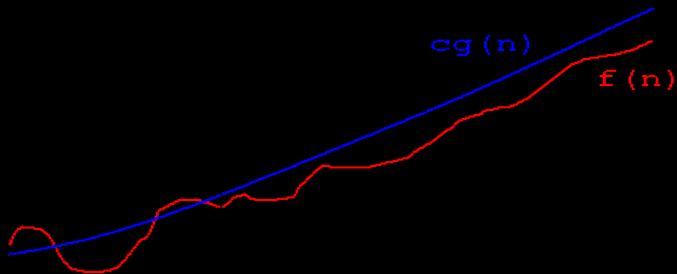
Asymptotic Analysis

- Goal: to simplify analysis of running time by getting rid of "details", which may be affected by specific implementation and hardware.
 - like "rounding": $1,000,001 = 1,000,000$
 - $3n^2 = n^2$
- Capturing the essence: how the running time of an algorithm increases with the size of the input in the limit.
 - Asymptotically more efficient algorithm are best for all but small inputs.

BIG Oh Notation

- Mainly used for worst-case analysis.
- Referred to as Asymptotic upper bound.
- $f(n) = O(g(n))$, if there exists constants c and n_0 , such that $f(n) \leq cg(n)$ for $n \geq n_0$
- $f(n)$ and $g(n)$ are functions over non negative integers

BIG Oh Notation



Asymptotic Growth

- Logarithmic: $O(\log(n))$
- Linear: $O(n)$
- Log-Linear: $O(n \log(n))$
- Quadratic: $O(n^2)$
- Polynomial: $O(n^k)$, $k \geq 1$
- Exponential: $O(a^n)$, $a > 1$

Thumb Rule of Asymptotic Notation

- Drop lower order terms and constant factors
 - $50n\log(n)$ is $O(n\log(n))$
 - $7n - 3$ is $O(n)$
 - $8n^2\log(n) + 5n^2 + n$ is $O(n^2\log(n))$
- Note: Even though $(50n\log(n))$ is $O(n^5)$, it is expected that such an approximation be of as small an order as possible.

Asymptotic Analysis Example-1

Algorithm1 (n)

```
1: Initialize  $i, j, sum = 0$ 
2: for ( $i = 0; i < n; i = i + 1$ ) do
3:   for ( $j = 0; j < n; j = j + 1$ ) do
4:      $sum \leftarrow i + j$ 
5:   end for
6: end for
7: return  $sum$ 
```

$O(n^2)$, Quadratic

Asymptotic Analysis Example-2

Algorithm2 (n)

```
1: Initialize  $i, j, sum = 0$ 
2: for ( $i = 0; i < n; i = i + 2$ ) do
3:   for ( $j = 0; j < n; j = j + 2$ ) do
4:      $sum \leftarrow i + j$ 
5:   end for
6: end for
7: return  $sum$ 
```

$O(n^2)$, Quadratic

Asymptotic Analysis Example-3

Algorithm3 (n)

```
1: Initialize  $x = 0, y = n$   
2: while ( $y > 1$ ) do  
3:    $x \leftarrow x + i$   
4:    $y \leftarrow y/2$   
5: end while  
6: return  $x$ 
```

$O(\log(n))$, Logarithmic

Asymptotic Analysis Example-4

Algorithm4 (n)

```
1: Initialize  $i, j, sum = 0$ 
2: for ( $i = 0; i < n; i = i + 2$ ) do
3:   for ( $j = 0; j < n; j = j * 2$ ) do
4:      $sum \leftarrow i + j$ 
5:   end for
6: end for
7: return  $sum$ 
```

$O(n \log(n))$, Log-Linear

Asymptotic Analysis Example-5

Algorithm5 (n)

```
1: Initialize  $i, j, sum = 0$ 
2: for ( $i = 1; i \leq n; i = i + 1$ ) do
3:   for ( $(j = 1; j \leq i; j = j + 1)$  do
4:     while ( $j > 1$ ) do
5:        $j = j/2$ 
6:        $sum = i + j$ 
7:     end while
8:   end for
9: end for
10: return  $sum$ 
```

Guess?

Defective Coin Problem



Given a set of coins of size n , where $n = 2^k$

The weight of all coins are equal except one.

The one with the different weight is the **Defective**

Assume a weigh scale is given to measure the weight of coin(s) in constant time.

Write an Algorithm to detect the defective one.

Defective Coin Problem

- First, let us solve this problem. (**Brute Force**)
- Next, find a way to solve it fast.

Brute Force Algorithm

Algorithm - FDC(C)

Input - Number of coins. An internal scale maintains the coin weights.

Output - The position of the defective coin.

```
1: for  $i = 0$  to  $C - 1$  do
2:   if  $scale(i) \neq scale(i + 1)$  then
3:     if  $scale(i) < scale(i + 1)$  then
4:       return  $i + 1$ 
5:     else
6:       return  $i + 2$ 
7:   end if
8: end if
9: end for
```

$O(n)$, linear time

Divide and Conquer Algorithm

Algorithm - FDC(*low*, *high*)

Input - Range of coins with the low and high. An internal scale maintains the coin weights.

Output - The position of the defective coin.

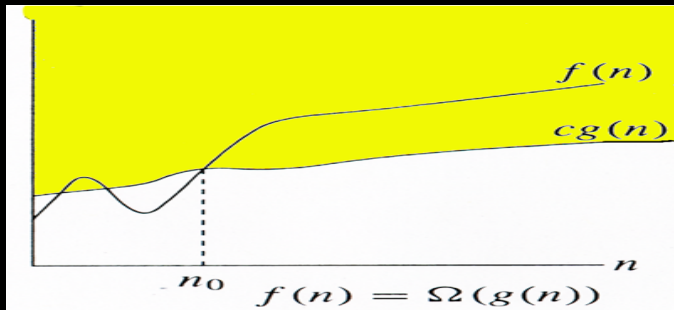
```
1: first  $\leftarrow$  SCALE(low to (low + high)/2);
2: second  $\leftarrow$  SCALE((low + high)/2 to high);
3: if (high - low) is equal to 1 then
4:   if first < second then
5:     return low;
6:   else
7:     return high;
8:   end if
9: else
10:  if first < second then
11:    return FDC(low, (low + high)/2);
12:  else
13:    return FDC((low + high)/2, high);
14:  end if
15: end if
```

$O(\log(n))$, logarithmic time

BIG Omega Notation

- The "big-Omega" or Ω - notation.
- It is generally used to describe best case running time or lower bound of algorithmic problems.
- $f(n) = \Omega(g(n))$ if there exists constant c and n_0 such that $cg(n) \leq f(n)$ for $n \geq n_0$.
- E.g., lower-bound of searching in an unsorted array is $\Omega(n)$.

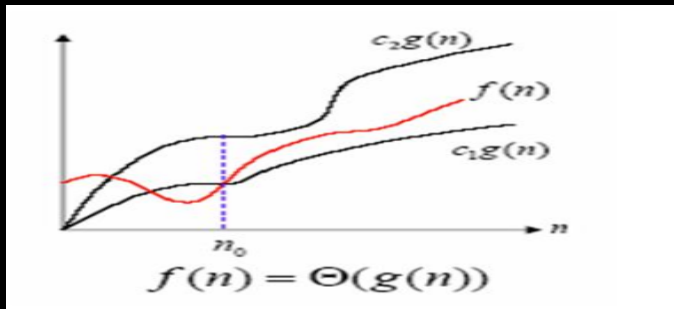
BIG Omega Notation



BIG Theta Notation

- The "big-Theta" or Θ - notation.
- Asymptotic tight bound.
- It is generally used to describe running time in between best and worst case. For example: **average** running time of an algorithmic problem.
- $f(n) = \Theta(g(n))$ if there exists constant c_1, c_2 , and n_0 such that
$$c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for } n \geq n_0$$
- $f(n) = \Theta(g(n))$ if and only if $f(n) = O(g(n))$ and $f(n) = \Omega(g(n))$.

BIG Theta Notation



Reading Assignment

Sedgewick 1.4

Questions?

Please ask your Questions!