CS202 - Algorithm Analysis How to Analyze Algorithms?

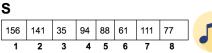
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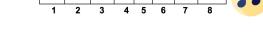
Allegheny College

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A Follow-up on Practical1





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: $temp \leftarrow 200$
- **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;



A Follow-up on Practical1



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;

Alternative



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 ...



(a) (Correctness)



(b) (Efficiency)

How is an Algorithm different from a Program?

- Makes uses 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 Java method like System.currentTimeMillis() to get an accurate measure of the actual running time.

Limitation of the Experimental Study approach!

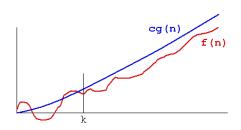
- It is necessary to implement and test the algorithm in order to determine the running time.
- Experiments can be done only on a limited set of inputs, and may not be indicative of the running time on other inputs not included in the experiment.
- In order to compare two algorithms, the same hardware 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 k , such that $f(n) \le cg(n)$ for $n \ge k$
- f(n) and g(n) are functions over non negative integers



Asymptotic Growth

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

Thumb Rule of Asymptotic Notation

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

Algorithm 1(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(n²), Quadratic



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(n²), Quadratic



Algorithm3(n)

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

O(log(n)), Logarithmic



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(nlog(n)), Log-Linear



```
Algorithm5 (n)
 1: Initialize i, j, sum = 0
 2: for (i = 1; i \le n; i = i + 1) do
     for ((i = 1; i <= i; i = i + 1)) do
        while (i > 0) do
 4:
          i = i/2
 5:
           sum = i + j
 6:
       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(W)
Input - A set of coin weights associated with a collection of coins.
Output - The position of the defective coin.
 1: for i = 0 to |W| - 1 do
       if W[i] \ll W[i+1] then
 3:
         if W[i] < W[i+1] then
 4:
            return i
 5:
         else
 6:
            return i+1
 7:
         end if
 8:
       end if
 9: end for
```

O(n), linear time

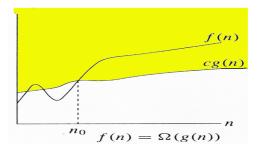
Divide and Conquer Algorithm

```
Algorithm - FDC(W, low, high)
Input - A set of coin weights associated with a collection of coins.
Output - The position of the defective coin.
 1: first \leftarrow SCALE(low to (low + high)/2);
 2: second \leftarrow SCALE((low + high)/2 \text{ to } high);
 3: if (high - low) is equal to 1 then
 4:
      if first < second then
 5:
         return low:
      else
 7:
         return high;
 8:
      end if
 9 else
10.
       if first < second then
11:
         return FDC(W, low, (low + high)/2);
12.
       else
13:
         return FDC(W, (low + high)/2, high);
       end if
14:
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 $cg(n) \le f(n)$ for $n \ge n_0$.
- E.g., lower-bound of searching in an unsorted array is $\Omega(n)$.

BIG Omega Notation

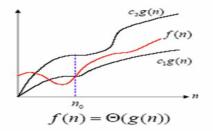


BIG Theta Notation

- 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 $c_1 g(n) \le f(n) \le c_2 g(n)$ for $n \ge n_0$
- $f(n) = \Theta(g(n))$ 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 if there are any Questions!