

Algorithm Analysis

Complexity Analysis

- Establishing the relationship between the input size and the algorithm/program time (and/or space) requirement.
- Estimate the time and space requirement for a given input size
- Compare algorithms
- Time Complexity: counting operations
- Count the number of operations that an algorithm will perform
- Asymptomatic analysis
- The Big O notation
- How fast time/space requirements increase as the input size approaches infinity

```
`` `cpp // number of outputs // t(n) = n for(i = 0; i < n, i++){ cout << A[i] << endl; }
```

```
// number of comparisons
```

```
// t(n) = n-1
```

```
template
```

```
bool IsSorted(T *A, int n){
```

```
bool sorted = true;
```

```
for(int i=0; i<n-1; i++){
```

```
if(A[i] > A[i+1]){
```

```
sorted = false;
```

```
}
```

```
}
```

```
return sorted;
```

```
}
```

Algorithm analysis covers the **worst case** (most of the time).

The average case is more useful, however, it is more

difficult to calculate.

The complexity of a function is its **input size**. For instance...

$t(n) = 1000n$ vs. $t(n) = 2n^2$

if n doubled...

$t(n) = 1000n \dots \frac{1000 \cdot 2n}{1000n} = 2$

– Time Will Double

$t(n) = 2n^2 \dots 2(2n^2) = 4n^2$

– Time increases by 4x

Scaling Analysis

– The constant factor does not change the growth rate and can be ignored

– **We can ignore the slower-growing terms.**

– Ex. $n^2 + n + 1 \dots n^2$

– capturing $O(n^2)$

Example Problem

Algorithm 1: $t_1(n) = 100n + n^2$

– insert – n , delete – $\log(n)$, lookup – 1

Algorithm 2: $t_2(n) = 10n^2$

– insert – $\log(n)$, delete – n , lookup – $\log(n)$

Which is faster if an application has as many inserts but few deletes and lookups?

Asymptotic Complexity Analysis

– Compares the **growth rate** of two functions

- Variables & Values - Nonnegative integers
- Dependent on eventual (asymptotic) behavior
- Independent of constant multipliers, and lower-order effects

Big "O" Notation

$$f(n) = O(g(n))$$

$$\text{iff } \exists c, n_0 > 0 \mid 0 < f(n) < cg(n) \forall n \geq n_0$$

- if there exists two positive constants $c > 0$ & $n_0 > 0$ such that $f(n) \leq cg(n)$ for all $n \geq n_0$

Big "Omega" Notation

$$f(n) = \Omega(g(n))$$

$$\text{iff } \exists c, n_0 > 0 \mid 0 < cg(n) < f(n) \forall n \geq n_0$$

- $f(n)$ is asymptotically lower bounded by $g(n)$

Big "Theta" Notation

$$f(n) = \Theta(g(n))$$

$$\text{iff } \exists c_1, c_2, n_0 > 0 \mid 0 < c_1g(n) < f(n) < c_2g(n) \forall n \geq n_0$$

- $f(n)$ has the same long-term growth as $g(n)$

Examples

$$f(n) = 3n^2 + 17$$

$$\Omega(1), \Omega(n), \Omega(n^2) \rightarrow \text{lower bounds}$$

- Chose $\Omega(n^2)$ because it is the **closest** to the lower bound

- Set \exists to 3. So... $3(n^2)$

$$O(n^2), O(n^3) \rightarrow \text{upper bounds}$$

- Chose $O(n^3)$ because it is the **closest** to the upper-bound

$\theta(n^2)$ \rightarrow **exact bound**

Why $f(n) \neq O(n)$?

Transitivity

$f(n) = O(g(n)) \rightarrow (a \leq b)$

$f(n) = \Omega(g(n)) \rightarrow (a \geq b)$

$f(n) = \theta(g(n)) \rightarrow (a = b)$

Additive property

– If $e(n) = O(g(n))$ and $f(n) = O(h(n))$

– Then...

– $e(n) + f(n) = O(g(n)) + O(h(n))$

Function | Name

– | –

c | Constant

$\log(N)$ | Logarithmic

$\log^2(N)$ | Log-squared

N | Linear

$N \log N$ |

N^2 | Quadratic

N^3 | Cubic

2^n | Exponential

Running time Calculations – Loops

```
```cpp
```

```
for(j=0; j < n; ++j){
```

```
// 3 Atomics
```

```
}
```

- Each iteration has 3 atomics so  $3n$
- Cost of the iteration itself ( $c * n$ ,  $c$  is a constant)
- Complexity  $\theta(3n + c * n) = \theta(n)$

### Running time Calculations - Loops with a break

```
for(j=0; j < n; ++j){
// 3 Atomics
 if(condition) break;
}
```

- Upper bound -  $O(4n) = (n)$
- Lower bound -  $\Omega(4) = \Omega(1)$
- Complexity -  $O(n)$

## Complexity Analysis

- Find  $n$  = input size - Find atomic activities count - Find  $f(n)$  = the number of atomic activities done by an input

### Sequential Search

```
```cpp for(size_t i= 0; i < a.size() ;i++){ if(a[i] == x){return} } //  $\theta(n)$  time
complexity ```
```

If then for loop

```
```cpp if(condition) i=0; else for(j =0; j < n; j++) a[j] = j; //  $\theta(n)$  time
complexity ```
```

### Nested Loop Search

```
```cpp for(j =0; j < n; j++){ // 2 atomics for(k =0; k
```