

## Complexity Analysis: Asymptotic Analysis

### Recall

- What is the measurement of algorithm?
- How to compare two algorithms?
- Definition of Asymptotic Notation

### Today Topic

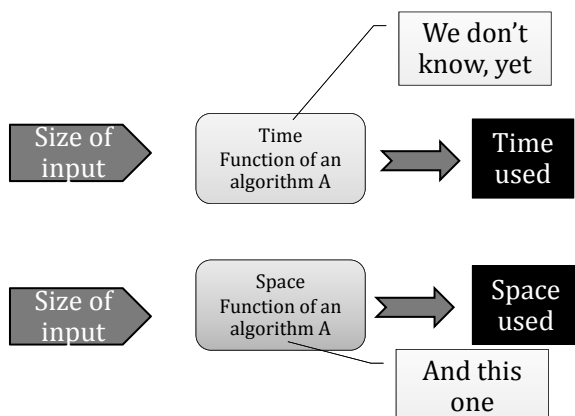
- Finding the asymptotic **upper** bound of the algorithm

### WAIT...

Did we miss something?

The resource function!

### Resource Function



### From Experiment

- We count the number of instructions executed
- Only count some instruction
  - One that's promising

## Why instruction count?

- Does instruction count == time?
  - Probably
  - But not always
- But...
  - Usually, there is a strong relation between instruction count and time if we count the most occurred instruction

## Why count just some?

- Remember the findMaxCount()?
  - Why we count just the max assignment?
  - Why don't we count everything?
- What if, each max assignment takes N instruction
  - That starts to make sense

## Time Function = instruction count

- Time Function = instruction count
- Time Function = instruction count
- Time Function = instruction count

## COMPUTE O()

## Interesting Topics of Upper Bound

- Rule of thumb!
- We neglect
  - Lower order terms from addition
    - E.g.  $n^3 + n^2 = O(n^3)$
  - Constant
    - E.g.  $3n^3 = O(n^3)$

Remember that we use =  
instead of (more correctly)  $\in$

## Why Discard Constant?

- From the definition
  - We can scale by adjusting the constant c
- E.g.  $3n = O(n)$ 
  - Because
    - When we let  $c \geq 3$ , the condition is satisfied

## Why Discard Lower Order Term?

- Consider
  - $f(n) = n^3 + n^2$
  - $g(n) = n^3$
- If  $f(n) = O(g(n))$ 
  - Then, for some  $c$  and  $n_0$ 
    - $c * g(n) - f(n) > 0$
    - Definitely, just use any  $c > 1$

## Why Discard Lower Order Term?

- Try  $c = 1.1$

$$1.1 * g(n) - f(n) = 0.1n^3 - n^2$$

- Does  $0.1n^3 - n^2 > 0$
- It is when
  - $-0.1n > 1$
  - E.g.,  $n > 10$

$$\begin{array}{lcl} 0.1n^3 - n^2 & > & 0 \\ 0.1n^3 & > & n^2 \\ 0.1n^3/n^2 & > & 1 \\ 0.1n & > & 1 \end{array}$$

## Lower Order only?

- In fact,
  - It's only the dominant term that count
- Which one is dominating term?
  - The one that grows faster
- Why?
  - Eventually, it is  $g^*(n)/f^*(n)$ 
    - If  $g(n)$  grows faster,
      - $g(n)/f^*(n) > \text{some constant}$
      - E.g.,  $\lim g(n)/f^*(n) \rightarrow \text{infinity}$

The non-dominant term

The dominant term

## What dominating what?

Left side dominates

$n^a$	$n^b$ (when $a > b$ )
$n \log n$	$n$
$n^2 \log n$	$n \log^2 n$
$c^n$	$n^c$
$\log n$	$1$
$n$	$\log n$

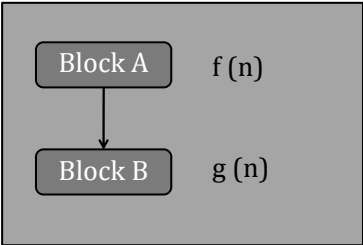
## Putting into Practice

- What is the asymptotic class of
  - $0.5n^3 + n^4 - 5(n-3)(n-5) + n^3 \log^8 n + 25 + n^{1.5}$   **$O(n^4)$**
  - $(n-5)(n^2+3) + \log(n^{20})$   **$O(n^3)$**
  - $20n^5 + 58n^4 + 15n^{3.2} + 3n^2$   **$O(n^{5.2})$**

## Asymptotic Notation from Program Flow

- Sequence
- Conditions
- Loops
- Recursive Call

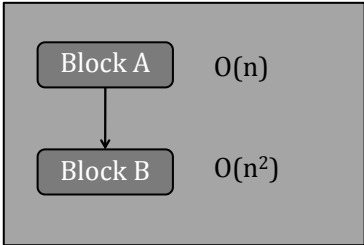
Sequence



$f(n) + g(n) =$   
 $O(\max(f(n),g(n)))$

Example

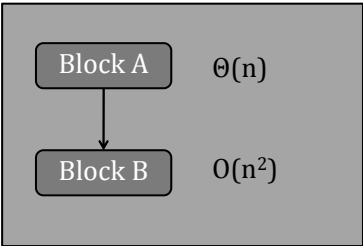
$f(n) + g(n) = O(\max(f(n),g(n)))$



$O(n^2)$

Example

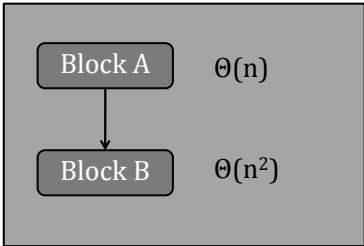
$f(n) + g(n) = O(\max(f(n),g(n)))$



$O(n^2)$

Example

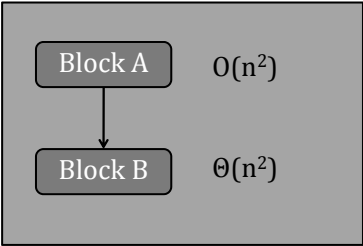
$f(n) + g(n) = O(\max(f(n),g(n)))$



$\Theta(n^2)$

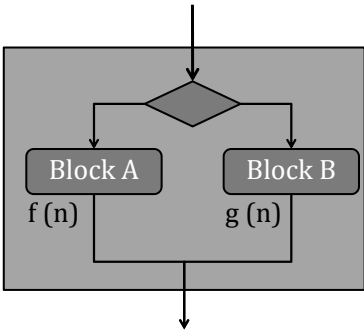
Example

$f(n) + g(n) = O(\max(f(n),g(n)))$



$\Theta(n^2)$

Condition



$O(\max(f(n),g(n)))$

## Loops

```
for (i = 1; i <= n; i++) {
    P(i)
}
```

$$\sum_{i=1}^n t_i$$

Let P(i)  
takes time  $t_i$

## Example

```
for (i = 1; i <= n; i++) {
    sum += i;
}
```

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

sum += i  $\rightarrow \Theta(1)$

## Why don't we use max( $t_i$ )?

- Because the number of terms is not constant

```
for (i = 1; i <= n; i++) {
    sum += i;
}
```

$\Theta(n)$

```
for (i = 1; i <= 1000000; i++) {
    sum += i;
}
```

$\Theta(1)$

With large constant

## Example

```
for (j = 1; j <= n; j++) {
    for (i = 1; i <= n; i++) {
        sum += i;
    }
}
```

sum += i  $\rightarrow \Theta(1)$

$$\begin{aligned} \sum_{j=1}^n \sum_{i=1}^n \Theta(1) &= \sum_{j=1}^n \Theta(n) \\ &= \Theta(n) + \Theta(n) + \dots + \Theta(n) \\ &= \Theta(n^2) \end{aligned}$$

## Example

```
for (j = 1; j <= n; j++) {
    for (i = 1; i <= j; i++) {
        sum += i;
    }
}
```

sum += i  $\rightarrow \Theta(1)$

$$\begin{aligned} \sum_{j=1}^n \sum_{i=1}^j \Theta(1) &= \sum_{j=1}^n \Theta(j) \\ &= \sum_{j=1}^n cj \\ &= c \sum_{j=1}^n j \\ &= c \left( \frac{n(n+1)}{2} \right) \\ &= \Theta(n^2) \end{aligned}$$

## Example : Another way

```
for (j = 1; j <= n; j++) {
    for (i = 1; i <= j; i++) {
        sum += i;
    }
}
```

sum += i  $\rightarrow \Theta(1)$

$$\begin{aligned} \sum_{j=1}^n \sum_{i=1}^j \Theta(1) &= \sum_{j=1}^n \Theta(j) \\ &= \sum_{j=1}^n O(n) \\ &= O(n^2) \end{aligned}$$

## Example

Your turn

```
for (j = 2; j <= n-1; j++) {  
  for (i = 3; i <= j; i++) {  
    sum += i;  
  }  
}
```

sum += i  $\rightarrow \Theta(1)$

## Example : While loops

```
While (n > 0) {  
  n = n - 1;  
}
```

$\Theta(n)$

## Example : While loops

```
While (n > 0) {  
  n = n - 10;  
}
```

$\Theta(n/10) = \Theta(n)$

## Example : While loops

```
While (n > 0) {  
  n = n / 2;  
}
```

$\Theta(\log n)$

## Example : Euclid's GCD

```
function gcd(a, b) {  
  while (b > 0) {  
    tmp = b  
    b = a mod b  
    a = tmp  
  }  
  return a  
}
```

## Example : Euclid's GCD

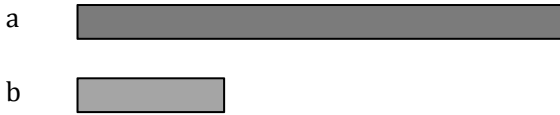
```
function gcd(a, b) {  
  while (b > 0) {  
    tmp = b  
    b = a mod b  
    a = tmp  
  }  
  return a  
}
```

Until the  
modding one is  
zero

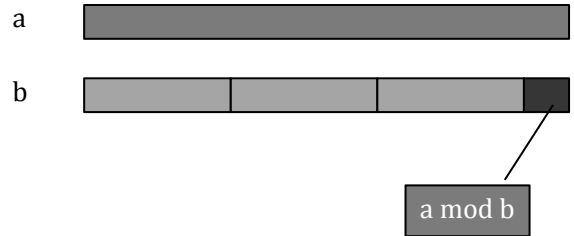
How many  
iteration?

Compute mod  
and swap

## Example : Euclid's GCD



## Example : Euclid's GCD

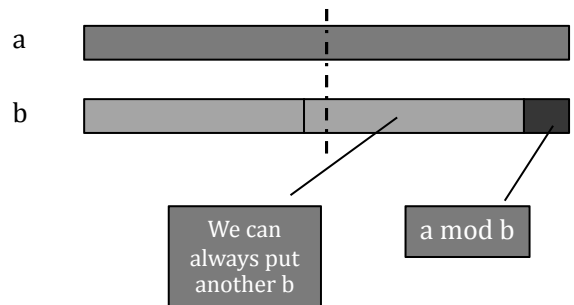


If  $a > b$   
 $a \bmod b < a / 2$

## Case 1: $b > a / 2$



## Case 1: $b \leq a / 2$



We can  
 always put  
 another  $b$

## Example : Euclid's GCD

```
function gcd(a, b) {
  while (b > 0) {
    tmp = b
    b = a mod b
    a = tmp
  }
  return a
}
```

$O(\log n)$

$B$  always reduces at  
 least half

## RECURSIVE PROGRAM

## Code that calls itself

```
void recur(int i) {  
    // checking termination  
  
    // do something  
  
    // call itself  
}
```

## Find summation from 1 to i

```
int sum(int i) {  
    //terminating condition  
    if (i == 1) //return 1;  
  
    // recursive call  
    int result;  
    result = i + sum(i-1);  
    return result;  
}  
  
void main() { printf("%d\n",sum()); }
```

## Order of Call: Printing Example

```
void seq(int i) {  
    if (i == 0)  
        return;  
  
    printf("%d ",i);  
    seq(i - 1);  
}
```

```
void seq(int i) {  
    if (i == 0)  
        return;  
  
    seq(i - 1);  
    printf("%d ",i);  
}
```

## Draw Triangle

```
void drawtri(int start,int i,int n) {  
    if (i <= n) {  
        for (int j = start;j <= i+start-1;j++) {  
            printf("%d ",j);  
        }  
        printf("\n");  
        drawtri(start,i + 1,n);  
    }  
}
```

## Programming Recursive Function

- First, define what will the function do
  - Usually, it is related to “smaller” instant of problem
- Write the function to do that in the trivial case
- Call itself to do the defined things

## Analyzing Recursive Programming

- Use the same method, count the most occurred instruction
- Needs to consider every calls to the function



## Example

```
void sum(int i) {
    if (i == 0) return 0;

    int result;
    result = i + sum(i - 1);
    return result;
}
```

$\Theta(n)$

## Example 2

```
void sum(int i) {
    if (i == 0) return 0;

    int result;
    result = i + sum(i - 1);
    int count = 0;
    for (int j = 0; j < i; j++) {
        count++;
    }
    return result;
}
```

$\Theta(n^2)$

## Theorem

- $T(n) = \sum T(a_i n) + O(N)$
- If  $\sum a_i < 1$  then
  - $T(n) = O(n)$

$$T(n) = T(0.7n) + T(0.2n) + T(0.01n) + 3n \\ = O(n)$$

## Recursion

```
void try( n ){
    if ( n <= 0 ) return 0;

    for ( j = 1; j <= n ; j++)
        sum += j;

    try ( n * 0.7 )
    try ( n * 0.2 )
}
```

## Recursion

```
void try( n ){
    if ( n <= 0 ) return 0;
    for ( j = 1; j <= n ; j++)
        sum += j;
    try ( n * 0.7 )
    try ( n * 0.2 )
}
```

terminating

$\Theta(1)$

process

$\Theta(n)$

recursion

$T(0.7n) + T(0.2n)$

$T(n)$

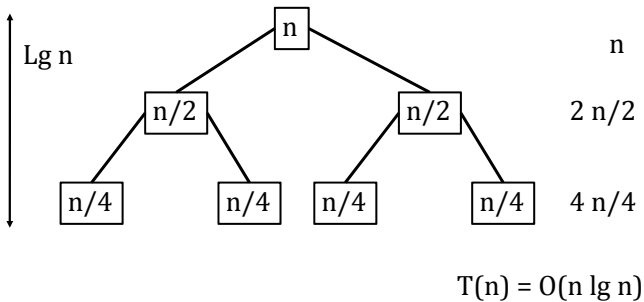
$$T(n) = T(0.7n) + T(0.2n) + O(n)$$

## Guessing and proof by induction

- $T(n) = T(0.7n) + T(0.2n) + O(n)$
- Guess:  $T(n) = O(n)$ ,  $T(n) \leq cn$
- Proof:
- Basis: obvious
- Induction:
  - Assume  $T(i < n) = O(i)$
  - $T(n) \leq 0.7cn + 0.2cn + O(n)$
  - $= 0.9cn + O(n)$
  - $= O(n)$  <<< dominating rule

## Using Recursion Tree

- $T(n) = 2T(n/2) + n$



## Master Method

- $T(n) = aT(n/b) + f(n)$   $a \geq 1, b > 1$

- Let  $c = \log_b(a)$

- $f(n) = O(n^{c-\epsilon}) \rightarrow T(n) = \Theta(n^c)$
- $f(n) = \Theta(n^c) \rightarrow T(n) = \Theta(n^c \log n)$   
 $- f(n) = \Theta(n^c \log^k n) \rightarrow T(n) = \Theta(n^c \log^{k+1} n)$
- $f(n) = \Omega(n^{c+\epsilon}) \rightarrow T(n) = \Theta(f(n))$

## Master Method : Example

- $T(n) = 9T(n/3) + n$
- $a = 9, b = 3, c = \log_3 9 = 2, n^c = n^2$
- $f(n) = n = O(n^{2-0.1})$
- $T(n) = \Theta(n^c) = \Theta(n^2)$

The case of  
 $f(n) = O(n^{c-\epsilon})$

## Master Method : Example

- $T(n) = T(n/3) + 1$
- $a = 1, b = 3, c = \log_3 1 = 0, n^c = 1$
- $f(n) = 1 = \Theta(n^c) = \Theta(1)$
- $T(n) = \Theta(n^c \log n) = \Theta(\log n)$

The case of  
 $f(n) = \Theta(n^c)$

## Master Method : Example

- $T(n) = 3T(n/4) + n \log n$
- $a = 3, b = 4, c = \log_4 3 < 0.793, n^c < n^{0.793}$
- $f(n) = n \log n = \Omega(n^{0.793})$
- $a f(n/b) = 3((n/4) \log(n/4)) \leq (3/4) n \log n = d f(n)$
- $T(n) = \Theta(f(n)) = \Theta(n \log n)$

The case of  
 $f(n) = \Omega(n^{c+\epsilon})$

## Conclusion

- Asymptotic Bound is, in fact, very simple
  - Use the rule of thumbs
    - Discard non dominant term
    - Discard constant
- For recursive
  - Make recurrent relation
    - Use master method
    - Guessing and proof
    - Recursion Tree