

# SC2001/ CX2101: Algorithm Design and Analysis

**Week 8: Review Lecture** 

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#### Revision of Complexity Analysis

- Complexity analysis expressed in big-O, big- $\theta$ , big- $\Omega$  gives the growth rate of a function compared with another function the order of magnitude of increase in the function when n increases. E.g f(n) = O(n<sup>2</sup>), g(n) = O(n)
- Two functions with the same complexity class may have very different running times. E.g f(n) = 2n<sup>2</sup>, and g(n) = 10n<sup>2</sup>

#### Revision of Complexity Analysis

Arrange the following functions in increasing order of their big-O time complexity

```
f(n)= n<sup>2</sup>, logn, nlgn, n!, 2<sup>n</sup>, 3<sup>n</sup>, lgn, n, n<sup>1/2</sup>, n<sup>5</sup>, log(n<sup>2</sup>), 2<sup>2n</sup>, 1000, n<sup>n</sup>
```

```
log_{10}(x) = ln(x) / ln(10)

log_{10}(x) = lg(x) / lg(10)

ln(x) = log_{10}(x) / log_{10}(e)

That is,

log_a(x) = c * log_b(x)
```

#### **Revision of Complexity Analysis**

f(n)= n<sup>2</sup>, logn, nlgn, n!, 2<sup>n</sup>, 3<sup>n</sup>, lgn, n, n<sup>1/2</sup>, n<sup>5</sup>, log(n<sup>2</sup>), 2<sup>2n</sup>, n<sup>2</sup>, 1000, n<sup>n</sup>

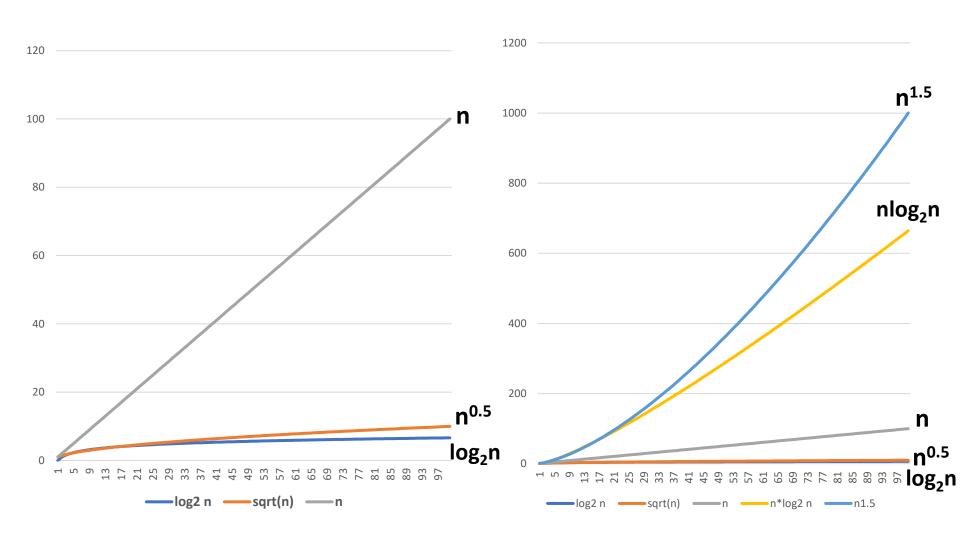
1) 
$$1000 = \theta(1) = O(1)$$

7) 
$$n^{5}$$

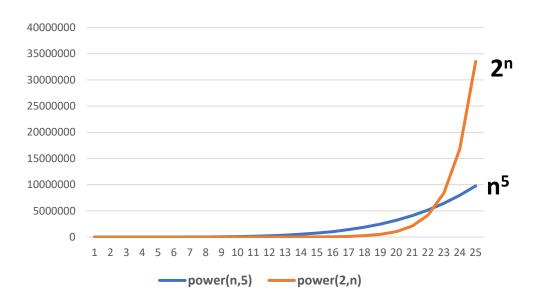
2) 
$$\log n$$
,  $\log (n^2) = \theta(\log n)$ 

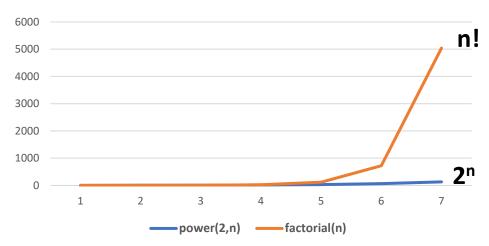
3) 
$$n^{1/2}$$

#### How f(n) increases when n increases



#### How f(n) increases when n increases





## Solving Recurrences (1)

- 1) The substitution method guess and check
- The iteration method expand (iterate) the recurrence
- 3) The master method use the manual

#### Comments:

- The master method is the easiest for certain types of recurrences and finds the tight bound.
- The substitution method is powerful for almost all types of recurrences and finds the upper bound.

## Example of the substitution method

$$T(1) = 0$$
 // worst case of quicksort  
 $T(n) = T(n-1) + n - 1$   
Guess  $T(n) = O(n^2)$  so we try to prove  $T(n) <= c*n^2$ 

#### Proof:

(1) 
$$T(1) = 0 \le c*1^2$$
 (so c=1)

(2) Assume 
$$T(k) \le k^2$$
, prove that  $T(k+1) \le (k+1)^2$ 

We have 
$$T(k+1) = T(k) + k$$

$$<= k^2 + k <= k^2 + 2*k + 1 = (k+1)^2$$

So 
$$T(n) \le n^2$$
 for all  $n \ge 1$ 

## Example of the iteration method

Suppose that, instead of using E[middle] as pivot, QuickSort also can use the median of E[first], E[(first + last)/2] and E[last]. How many key comparisons will QuickSort do in the worst case to sort *n* elements? (Remember to count the comparisons done in choosing the pivot.) Before partition: (P is the median)

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After partition: (3 comparisons to get the median, n-3 comparisons for partition)

$$T(1)=0$$

$$T(2) = 1$$

<P

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

## Example of the iteration method

$$T(n) = T(n-2) + n$$
  
=  $T(n-4) + n - 2 + n$   
=  $T(n-6) + n - 4 + n - 2 + n$   
=  $T(n-8) + n - 6 + n - 4 + n - 2 + n$ 

$$T(1) = 0$$
  
 $T(2) = 1$   
 $T(n) = T(n-2) + n$ 

If n = 2k (i.e. n is even)
$$T(n) = T(2) + (4 + 6 + 8 + ... + n) // k-1 \text{ terms}$$

$$= 1 + \frac{k-1}{2}(4 + n)$$

$$= O(n^2)$$

$$a_i = a_1 - a_2$$

$$s_k = \frac{k}{2}(6)$$

$$a_i = a_1 + (i-1)d$$
  
 $s_k = \frac{k}{2}(a_1 + a_k)$ 

## Example of the iteration method

```
If n = 2k+1 (n is odd)
T(n) = T(1) + (3 + 5 + 7 + ... + n) 	 // k \text{ terms}
= \frac{k}{2}(3 + n)
= O(n^2)
a_i = a_1 + (i-1)d
s_k = \frac{k}{2}(a_1 + a_k)
```

## Example of the master method

Multiplying two nxn matrices

$$W(n) = 7W(n/2) + 15n^2/4$$

n 
$$\log b^a = n \log 2^7 = n^{\frac{\ln 7}{\ln 2}} = n^{2.8075}$$
  
 $f(n) = 15n^2/4$   
 $= O(n^{2.8075-0.5})$   
 $W(n) = \theta(n^{2.8075})$ 

$$\log_b x = \frac{\log_d x}{\log_d b}$$

## Additional example of the substitution method

Recurrence for the best case of mergesort:

$$T(1)=0$$

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

Guess T(n) = O(nlgn)



Proof: consider n is a power of 2.

$$(1) T(1) = 0 \le 1*lg1$$

(2) Assume that  $T(2^k) \le k^* 2^k$ , prove that  $T(2^{k+1}) \le (k+1)^* 2^{k+1}$ .

$$T(2^{k+1}) = 2T(2^k) + 2^k$$
  
 $<= 2^* k^* 2^k + 2^k$   
 $<= k^* 2^{k+1} + 2^k + 2^k$   
 $= (k+1)^* 2^{k+1}$   
Thus  $T(n) = O(n \lg n)$