# Course organization

#### -Introduction (Week 1-2)

- Course introduction
- A brief introduction to molecular biology
- A brief introduction to sequence comparison

#### – Part I: Algorithms for Sequence Analysis (Week 3 - 11)

- Chapter 1-3, Models and theories
  - » Probability theory and Statistics (Week 4)
  - » Algorithm complexity analysis (Week 5)
  - » Classic algorithms (Week 6)
  - » Lab: Linux and Perl
- Chapter 4, Sequence alignment (week 7)
- Chapter 5, Hidden Markov Models (week 8)
- Chapter 6. Multiple sequence alignment (week 10)
- Chapter 7. Motif finding (week 11)
- Chapter 8. Sequence binning (week 11)

– Part II: Algorithms for Network Biology (Week 12 - 16)

# Chapter 2: Algorithm Complexity Analysis

Chaochun Wei Fall 2014

# Contents

- Reading materials
- Why do we need to analyze the complexity of an algorithm?
  - Examples
- Introduction
  - Algorithm complexity
  - "Big O" notation: O()

# Reading

Cormen book:

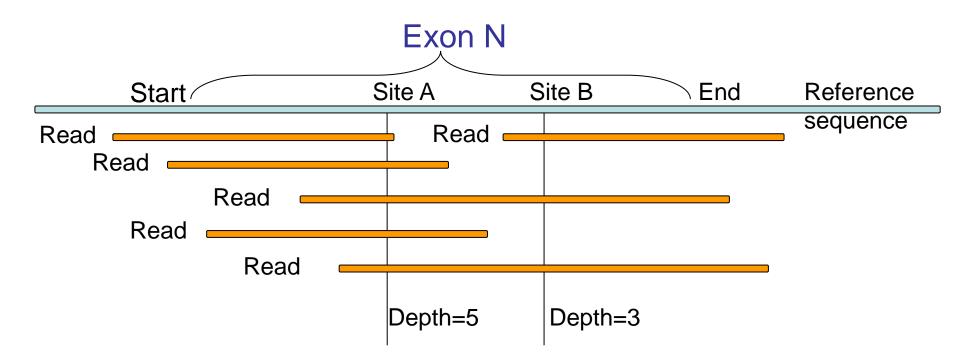
Thomas, H., Cormen, Charles, E., Leiserson, and Ronald, L., Rivest . Introduction to Algorithms, The MIT Press.

(read Chapter 1 and 2, page 1-44).

There are  $\sim$ 60 millions of short reads sequenced from exon regions of a human genome. We need to figure out the how many exons were covered with at least 10 reads.

Steps:

- 1. Reads are aligned to the genome;
- 2. Each alignment is checked to see the exon it covers;
- 3. For each exon, check the number of reads cover the exon;
- 4. For all exons, filter out those with read number < 10.



1 days later

Student: I have created a program to do the analysis. It's running. Teacher: Cool. Let me know when your analysis finishes.

6 days later...

Student: My program has been running for 5 days, and it keeps on running. I have no idea about what is happening and what to do with it.

Teacher: Its core is a sorting algorithm with a complexity of at most O(N\*IgN). It should be done within a few minutes!

Student: What?.....

An **algorithm** is any well-defined computational procedure that takes in some **inputs** and produces some **Outputs**.

Example: Sort an array of numbers 3, 2, 4, 5, 7, 1,  $6 \rightarrow 1$ , 2, 3,4, 5,6,7

An algorithm is any well-defined computational procedure that takes in some inputs and produces some outputs.

# Complexity: a function of input size Time complexity: the running time Space complexity: the memory size required

#### Input size

- •Number of items in the input
  - Sorting problem
  - ●FFT
- Total number of bits needed to represent the input
   Arithmetic operation (+,-,x,/)
- •The value of input
  - •Factorial (N!)

### Multiple input sizes

Need to specify which input size is used
Graph operation (number of Vertices, and edges)

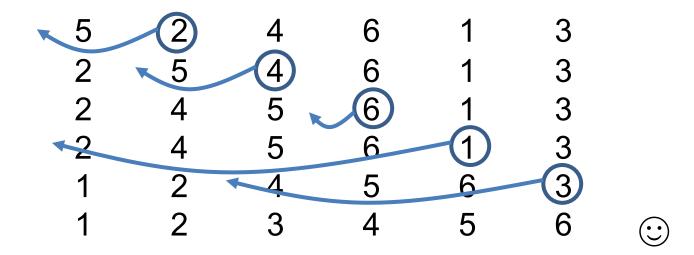
#### Before we start

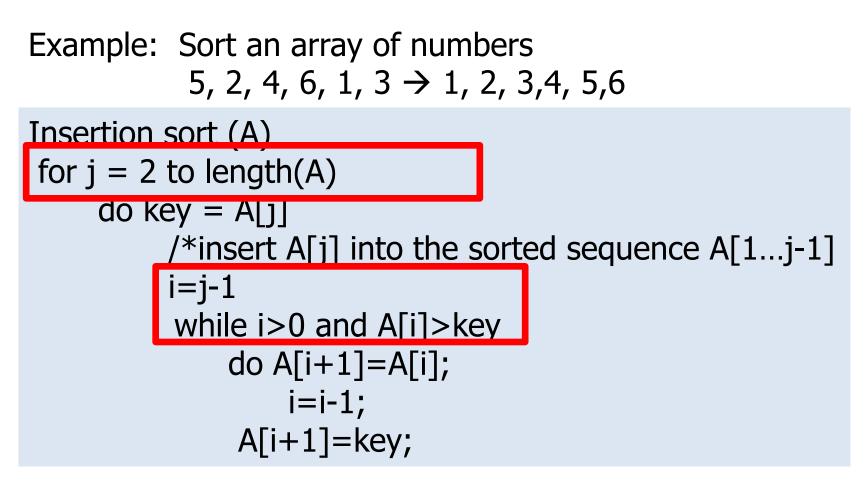
 we use a generic one-processor, randomaccess machine.
 No parallel

# Example: Sort an array of numbers 5, 2, 4, 6, 1, $3 \rightarrow 1$ , 2, 3,4, 5,6

```
Insertion sort (A)
for j = 2 to length(A)
do key = A[j]
/*insert A[j] into the sorted sequence A[1...j-1]
i=j-1
while i>0 and A[i]>key
do A[i+1]=A[i];
i=i-1;
A[i+1]=key;
```

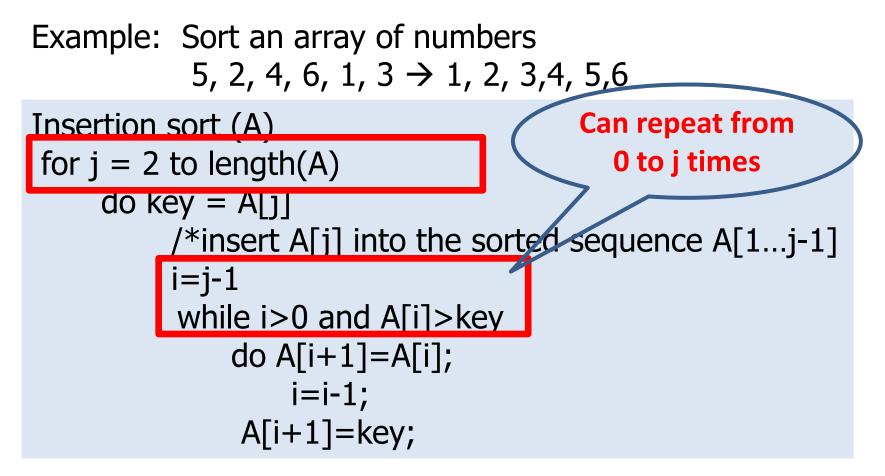
Example: Sort an array of numbers 5, 2, 4, 6, 1,  $3 \rightarrow 1$ , 2, 3,4, 5,6





Algorithm time complexity:  $O(N^2)$ 

#### Worst-case and average-case analysis



Algorithm time complexity:  $O(N^2)$ 

Algorithm complexity

#### Order of growth

Example: Sort an array of numbers 5, 2, 4, 6, 1,  $3 \rightarrow 1$ , 2, 3,4, 5,6

Insertion sort: Algorithm run time complexity: O(N<sup>2</sup>) Order of growth: 2

Algorithm complexity

#### O-notation (big-O notation): Asymptotic upper bound

 $O(g(n)) = \{f(n): \text{ there exist positive constants c}$ and  $n_0$  such that  $0 \le f(n) \le c g(n)$  for all  $n \ge n_0\}$ 

Note about O-notation operations:  $O(k_1*N^2+k_2*N^3)=O(N^3)$  for constants  $k_1$ ,  $k_2$ 

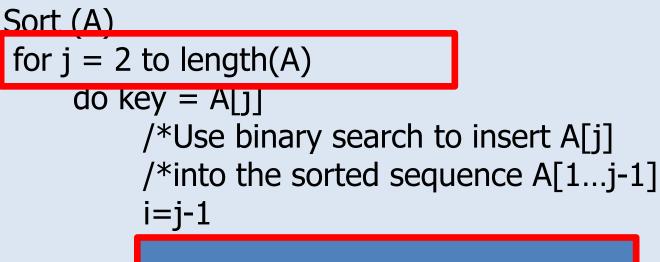
#### O-notation (big-O notation): Asymptotic upper bound

Example: Sort an array of numbers 5, 2, 4, 6, 1,  $3 \rightarrow 1$ , 2, 3,4, 5,6

Insertion sort: algorithm time complexity: O(N<sup>2</sup>)

#### Sorting with time complexity of O(N\*logN)

Example: Sort an array of numbers 5, 2, 4, 6, 1,  $3 \rightarrow 1$ , 2, 3,4, 5,6



Binary\_search(A[j], A[1...j-1],)

Algorithm complexity

#### Sorting

# Example: Sort an array of numbers 5, 2, 4, 6, 1, $3 \rightarrow 1$ , 2, 3,4, 5,6

There are a lot of sorting algorithms: Heap sort (O(N\*logN)) Merge sort (O(N\*logN)) \*Quick sort (worst-case O(N<sup>2</sup>), average O(N\*logN))

Algorithm complexity

#### Merge sort

Time Complexity: 
$$T(N) = \begin{cases} O(1); if N = 1\\ 2T(N/2) + O(N); if N > 1 \end{cases}$$
 22  
Solve it: T(N) = O(N\*logN)

#### Space complexity

Example: Sort an array of numbers 5, 2, 4, 6, 1,  $3 \rightarrow 1$ , 2, 3,4, 5,6

Need an array of size N: A[1...N], and 3 temporary variables O(N)

Example: Sequence alignment

Need a two-dimension array of size N\*M, and a constant number of temporary variables O(N\*M) or O(max(N, M))

## Other issues

Output size

•Blast: output can be a problem

Input/Output method/place/mode

Speed

screen << hard disk << memory</li>

• Programming language

Speed

●Perl < java < C++ <C