

Algo Design 1 : Complexity notions

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Algorithm : Motivation

- Different Approach of programming
 - Definition of the problem to solve
 - Searching for an algorithm
 - Complexity analysis
 - Implementation

Algorithm : definition

- Algorithm – Definition :
- Finite sequence of **well-defined**, computer-implementable instructions, to solve a class of problems or to perform a computation

(Wikipedia)

Algorithm : complexity

- Goal : measure the inherent efficiency of an algorithm
 - As a function of the **input size**
 - Compute the **important elementary operations**
 - **Asymptotic Measure**
 - **Worst** / **Average** -case complexity
 - **Time** / **Space** complexity

Classical example: sorting

- Goal : sort an array of n integers
- Compute the number of comparisons
- Elementary algorithms :
« order of » n^2 comparisons
- Advanced algorithms: $n^2 \Rightarrow n \cdot \log(n)$

Complexity Notations

- $f(n) = \mathcal{O}(g(n))$ iff $0 \leq f(n) \leq c \cdot g(n)$
- $f(n) = \Theta(g(n))$ iff $c \cdot g(n) \leq f(n) \leq c' \cdot g(n)$
- Examples :
 - $n^2 + 3n + 1 = \Theta(n^2) = \Theta(50n^2 + 12345)$
 - $n / \ln(n) = \mathcal{O}(n)$
 - $50n^{10} = \mathcal{O}(n^{10,01})$
 - $2^n = \mathcal{O}(\exp(n))$
 - $\exp(n) = \mathcal{O}(n!)$

Hierarchy of functions

- One can establish a **hierarchy** between functions :

$\log(n) \ll \sqrt{n} \ll n \ll n^2 \ll n^3 \ll 2^n \ll \exp(n) \ll n!$

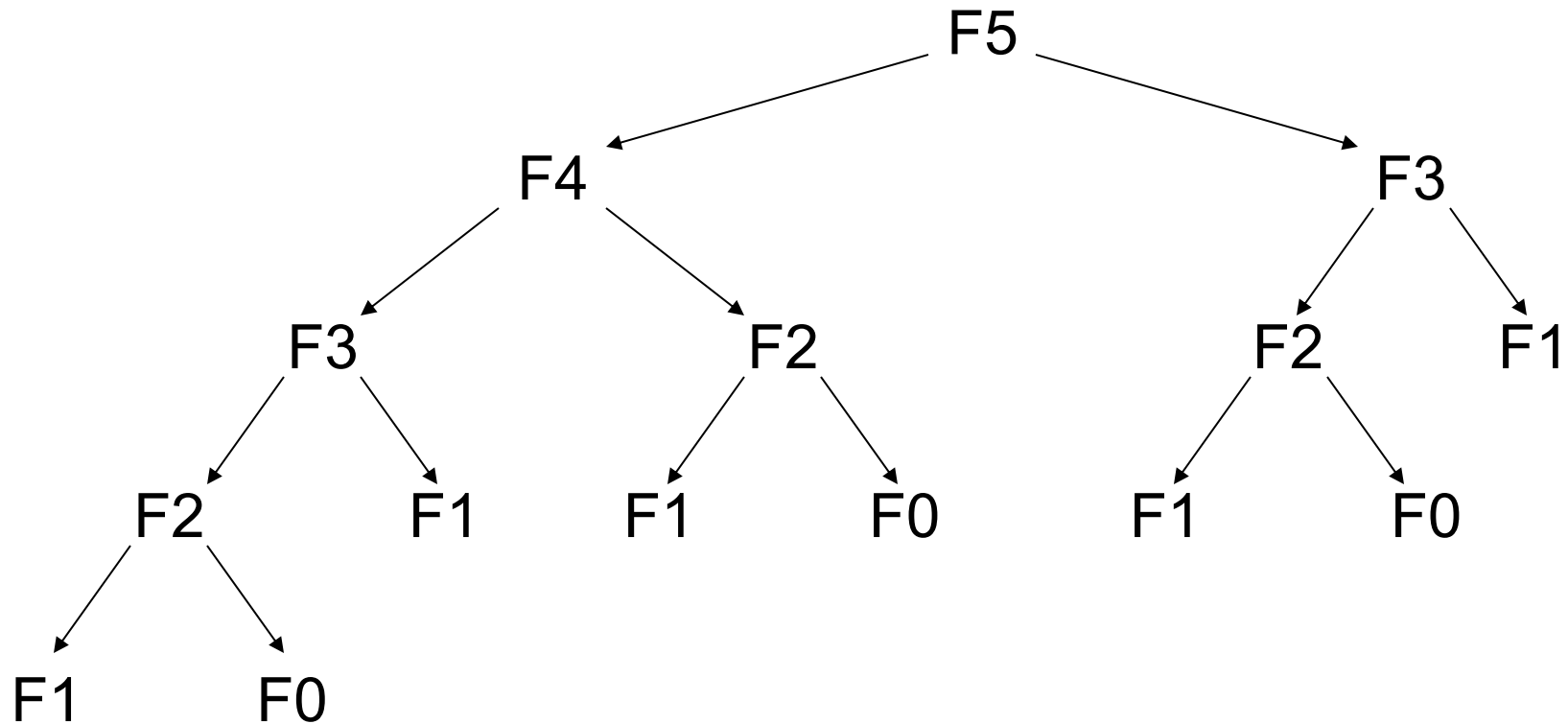
$\log(n)$	3.3	6.6	10
\sqrt{n}	3.1	10	32
n	10	100	1000
$n \log(n)$	33	664	10^4
n^2	100	10^4	10^6
n^3	10^3	10^6	10^9
2^n	10^3	10^{30}	10^{300}
$\exp(n)$	2×10^4	10^{43}	10^{434}
$n!$	3.6×10^6	10^{158}	10^{2568}

Fibonacci sequence

- $F_n = F_{n-1} + F_{n-2}$ if $n > 1$
- $F_0 = F_1 = 1$
- 1, 1, 2, 3, 5, 8, 13, 21, 34, ...
 - ```
int fibo1(int n)
{
 if (n <= 1) return 1;
 else return fibo1(n-1) + fibo1(n-2);
}
```
- Complexity :  $\Theta(\omega^n)$  where  $\omega = (1 + \sqrt{5})/2$



# Fibonacci sequence



- Complexity :  $\Theta(\omega^n)$  where  $\omega = (1 + \sqrt{5})/2$

# Fibonacci sequence (2)

|   |   |   |   |   |   |    |    |    |
|---|---|---|---|---|---|----|----|----|
| 1 | 1 | 2 | 3 | 5 | 8 | 13 | 21 | 34 |
|---|---|---|---|---|---|----|----|----|

F0   F1   F2   F3   F4   F5   F6   F7   **F8**

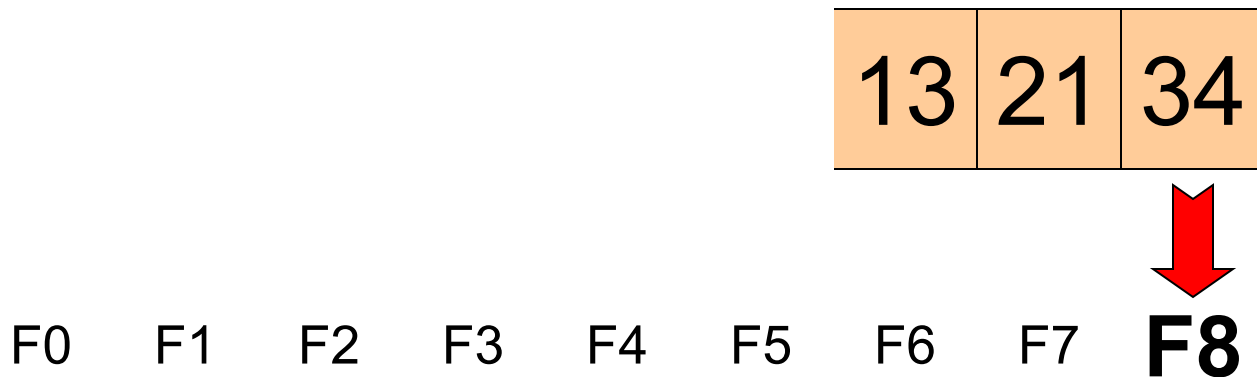


- Complexity :  $\Theta(n)$
- Space complexity :  $\Theta(n)$

# Fibonacci sequence (2)

- ```
int fibo2(int n)
{ int *t;
  int i, res;
  t=(int *)malloc((n+1)*sizeof(int));
  t[0]=1; t[1]=1;
  for(i=2;i<=n;i++) t[i]=t[i-1]+t[i-2];
  res=t[n];
  free(t);
  return res;}
```
- Complexity : $\Theta(n)$
- Space Complexity: $\Theta(n)$

Fibonacci sequence (3)



- Complexity : $\Theta(n)$
- Space Complexity : $\Theta(1)$ (constant)

Fibonacci sequence (3)

- ```
int fibo3(int n)
{ int f1,f2,t,i;
 f1=1; f2=1;
 for (i=2;i<=n;i++)
 { t=f2;
 f2=f1+f2;
 f1=t;}
 return f2;}
```
- Complexity :  $\Theta(n)$
- Space Complexity :  $\Theta(1)$  (constant)

# Fibonacci sequence (4)

- $F_n = 1 \times F_{n-1} + 1 \times F_{n-2}$

- $F_{n-1} = 1 \times F_{n-1} + 0 \times F_{n-2}$

$$\begin{pmatrix} F_n \\ F_{n-1} \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \times \begin{pmatrix} F_{n-1} \\ F_{n-2} \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}^{n-1} \times \begin{pmatrix} F_1 \\ F_0 \end{pmatrix}$$

- We get back to compute a « matrix exponentiation»
- Complexity :  $\Theta(\log(n))$
- Space Complexity :  $\Theta(1)$  (constant)

# Matching theory / practice

| n        | 40   | $5 \cdot 10^7$        | $2 \cdot 10^8$     | $2 \cdot 10^9$ |
|----------|------|-----------------------|--------------------|----------------|
| Fibo1(n) | 31 s | Too large computation |                    |                |
| Fibo2(n) | 0 s  | 18 s                  | Segmentation fault |                |
| Fibo3(n) | 0 s  | 4 s                   | 19 s               | 3 min 15       |
| Fibo4(n) | 0 s  | 0 s                   | 0 s                | 0 s            |

# Classical Example : sorting

- Goal : sort an array of  $n$  integers
- Compute the number of comparisons
- Elementary Algorithms:  $O(n^2)$
- Advanced Algorithms:  $O(n \cdot \log(n))$



# Insertion Sorting

- Sort an array  $A$  of  $n$  integers :
- **for**  $i=2$  to  $n$  do  
     $key=A[i]$   
     $j=i-1$   
    **while**  $j>0$  et  $A[j]>key$  do  
         $A[j+1]=A[j]$   
         $j=j-1$   
     $A[j+1]=key$

# Insertion Sorting (2)

- ```
void InsertionSort(int *A, int n)
{
    int i, j, key;
    for (i=1; i<n; i++)
        {
            key=A[i];
            j=i-1;
            while ((j>=0) && (A[j]>key))
                { A[j+1]=A[j];
                  j=j-1;}
            A[j+1]=key;
        }
}
```

Insertion Sorting (3)

- Different style ... to be avoided !!!
- ```
Sort_ugly(int *A, int n) {
 int i=1, j=0, key=* (A+1) ;
 for (; i<n; A[j+1]=key, j=i++, key=A[i])
 while ((j>=0) && (A[j]>key))
 A[j+1]=A[j--]; }
```

# Insertion Sorting (4)

- Worst-case Complexity :  
*table sorted upside down*

$$C_n = \Theta(n^2)$$

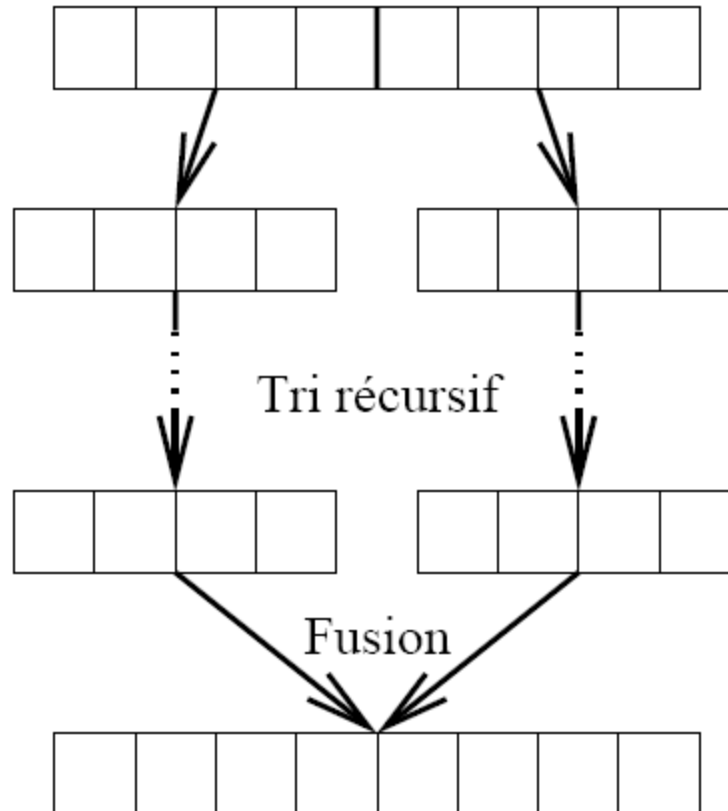
- Average-case complexity :

$$C_n = \Theta(n^2)$$

# Merge Sort

- Sort an array  $A$  between the indices  $p$  &  $r$  :
- **if**  $p < r$  **then**
  - $q = (p+r) / 2$
  - `mergesort (A, p, q)`
  - `mergesort (A, q+1, r)`
  - `merge (A, p, q, r)`
- Complexity...

# Merge Sort



# Quicksort

- Recursive sort based on partitioning
- `if p < r then`
  - `q = partition(A, p, r)`
  - `quicksort(A, p, q - 1)`
  - `quicksort(A, q + 1, r)`
- Worst-case complexity :  $C_n = \Theta(n^2)$
- Average-case Complexity :  $C_n = \Theta(n \cdot \log(n))$

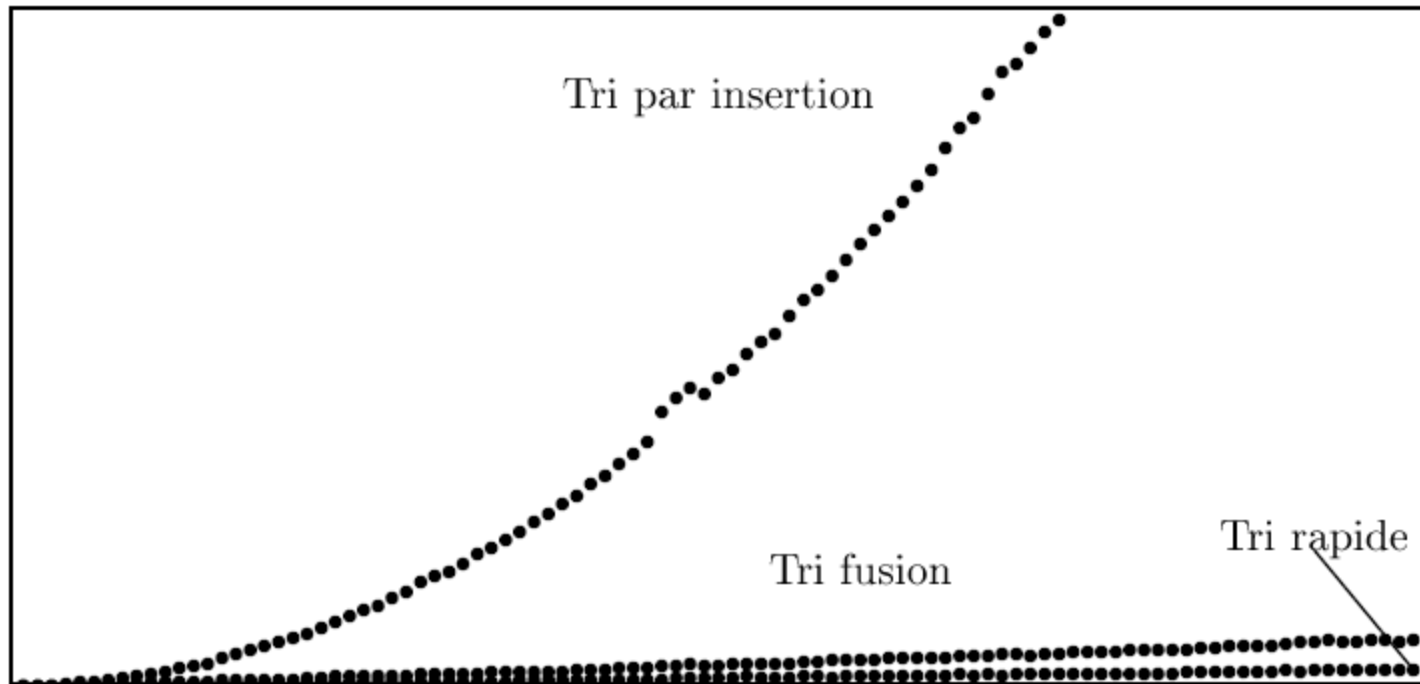
# Sort Problem

| Algorithm                | Worst-case                | Average                   |
|--------------------------|---------------------------|---------------------------|
| Insertion<br>Bubble Sort | $\Theta(n^2)$             | $\Theta(n^2)$             |
| Quicksort                | $\Theta(n^2)$             | $\Theta(n \cdot \log(n))$ |
| Merge Sort               | $\Theta(n \cdot \log(n))$ | $\Theta(n \cdot \log(n))$ |



# Matching theory / practice

## Computation Time

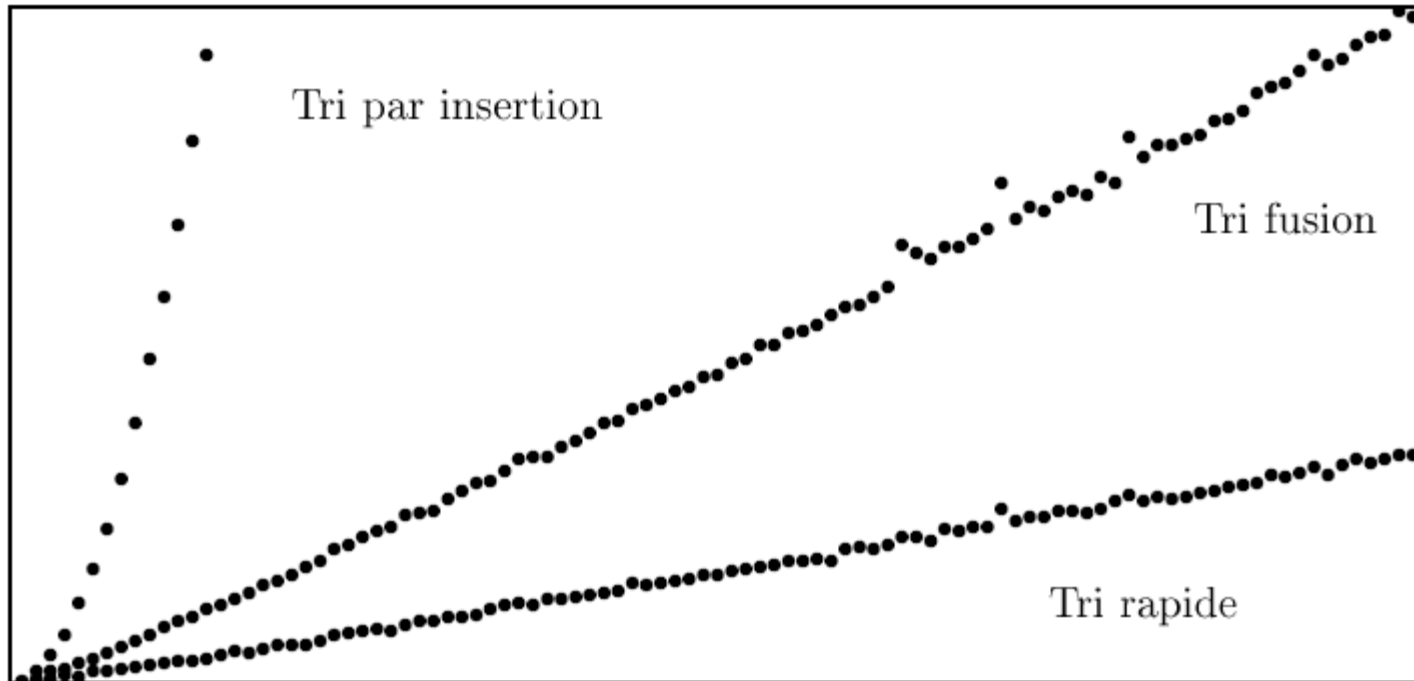


n=5000

FIG. 1.1 – *Complexité moyenne expérimentale*

# Matching theory / practice

## Computation Time



n=5000

FIG. 1.2 – *Complexité moyenne expérimentale (agrandissement)*

# Matching theory / practice

## Computation Time

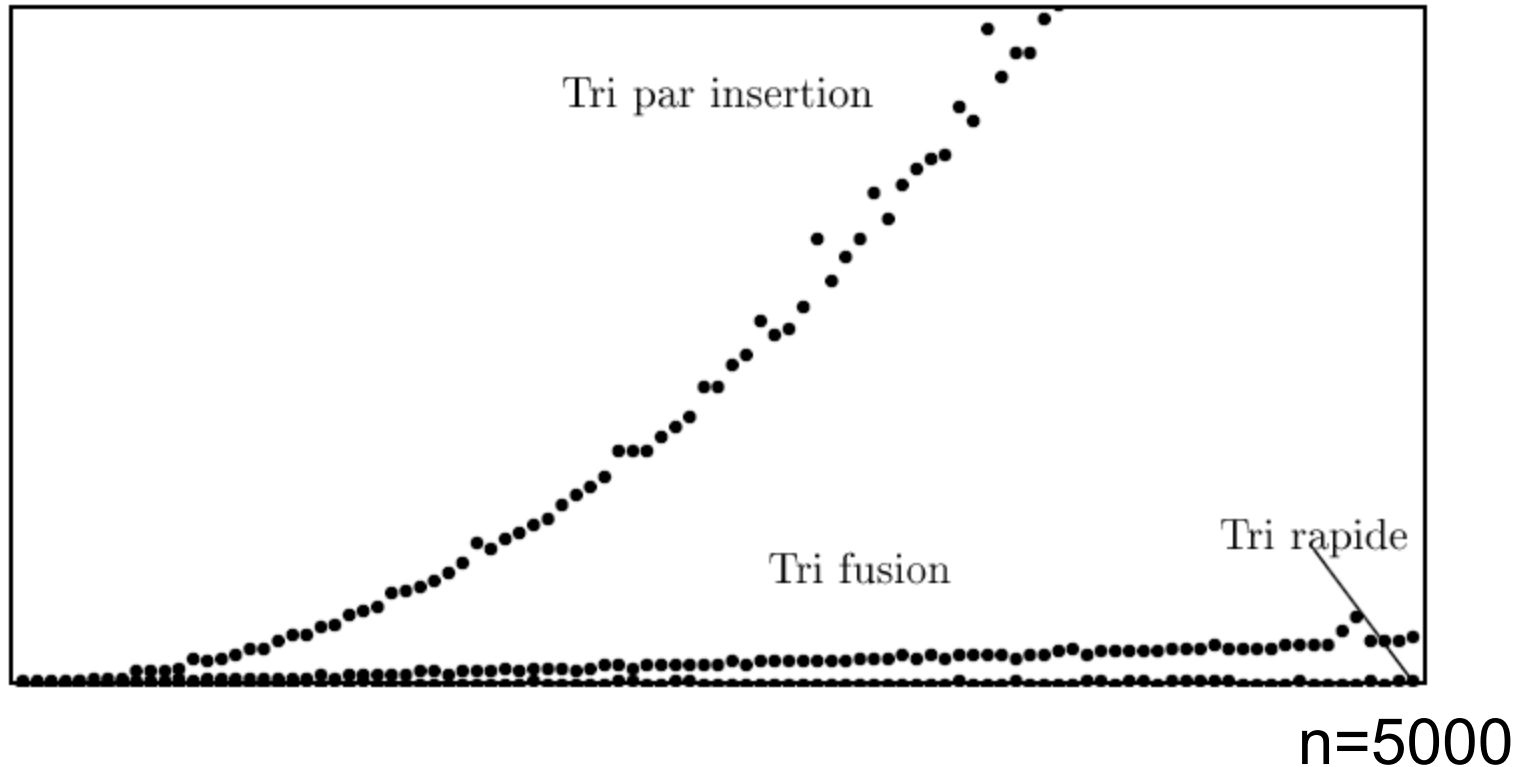


FIG. 1.3 – Complexité expérimentale dans le cas d'un tableau déjà trié

# Matching theory / practice

## Computation Time

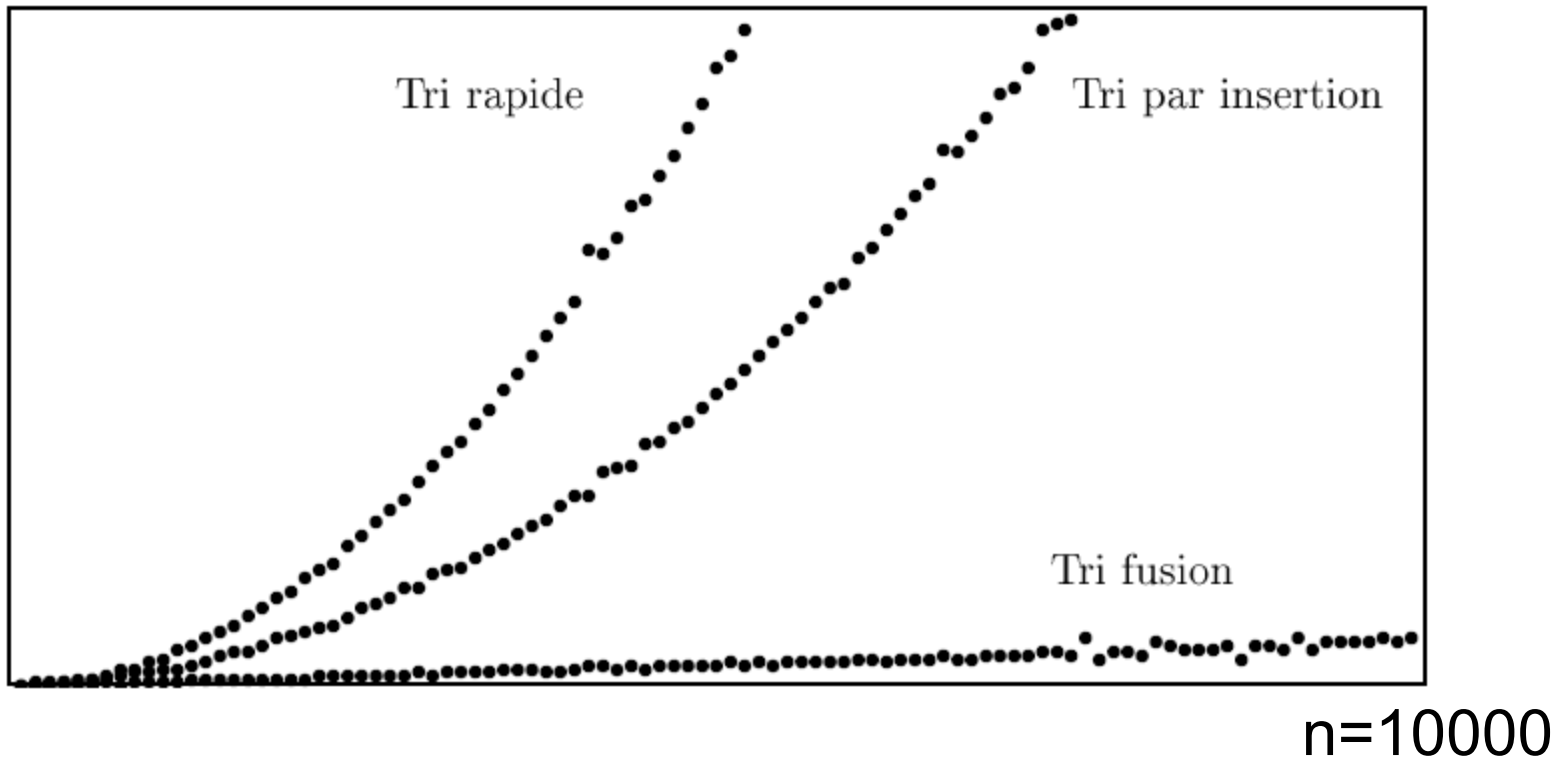


FIG. 1.4 – Complexité expérimentale dans le cas d'un tableau inversement trié

# Example: Matrix product

- Goal : multiply 2  $n \times n$  matrices
- Compute number of additions and multiplications between elements
- Elementary Algorithms :  $O(n^3)$
- Advanced Algorithms :  $O(n^{2,376})$  !!!

# Efficient Algorithms ?

- Average-case Complexity
- Worst-case Complexity
- Easy implementation
- Efficiency in practice
- Hybrid Algorithms

**The End**