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Chapter 1

Elementary analysis

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What is an algorithm?

An algorithm is a finite set of precise instructions for performing a computation or for solving a problem.





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An algorithm is a finite set of precise instructions for performing a computation or for solving a problem.

Properties of algorithms

- **Input** from a specified set,
- **Output** from a specified set (solution),
- **Definiteness** of every step in the computation,
- **Correctness** of output for every possible input,
- **Finiteness** of the number of calculation steps,
- **Effectiveness** of each calculation step and
- **Generality** for a class of problems.

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- Generally, not much interested in time and space complexity for small inputs.
- Given two algorithms A and B for solving problem P .

Input size	Algorithm A	Algorithm B
n	$5000n$	1.2^n
10	50,000	6
100	500,000	2,817,975
1,000	5,000,000	1.5×10^{79}
100,000	5×10^8	1.3×10^{7918}



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100	500,000	2,817,975
1,000	5,000,000	1.5×10^{79}
100,000	5×10^8	1.3×10^{7918}

- B cannot be used for large inputs, while A is still feasible.
- So what is important is the **growth** of the complexity functions.
- Growth of time and space complexity with increasing input size n is a suitable measure for the comparison of algorithms.

Types of formulas for basic operation count

- Exact formulas, e.g., $C(n) = n(n - 1)/2$.



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Types of formulas for basic operation count



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- Exact formulas, e.g., $C(n) = n(n - 1)/2$.
- Formula indicating order of growth with specific multiplicative constant e.g., $C(n) \approx 0.5n^2$.

Types of formulas for basic operation count



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Types of formulas for basic operation count

- Exact formulas, e.g., $C(n) = n(n - 1)/2$.
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- Most important: Order of growth within a constant multiple as $n \rightarrow \infty$



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- Most important: Order of growth within a constant multiple as $n \rightarrow \infty$

Asymptotic growth rate

A way of comparing functions that ignores constant factors and small input sizes

- $O(g(n))$: class of functions $f(n)$ that grow **no faster than** $g(n)$
- $\Theta(g(n))$: class of functions $f(n)$ that grow **at the same rate as** $g(n)$
- $\Omega(g(n))$: class of functions $f(n)$ that grow **at least as fast as** $g(n)$



Complexity classes - a small vocabulary

- Constant: $O(1)$ (independent on the input size)
- Sub-linear or logarithmic: $O(\log n)$
- Linear: $O(n)$
- Quasi-linear: $O(n \log n)$
- Quadratic: $O(n^2)$
- Cubic: $O(n^3)$
- Polynomial: $O(n^p)$ ($O(n^2)$, $O(n^3)$, etc)
- Quasi-polynomial: $O(n^{\log(n)})$
- Exponential: $O(2^n)$
- Factorial: $O(n!)$



Asymptotic upper bound - worst case

Asymptotic upper bound “big O”

$T(n) = O(f(n))$ **iff** $\exists c \in \mathbb{R}^+, c > 0$ and $\exists n_0 \in \mathbb{N}, n_0 > 0$
such that $\forall n > n_0: T(n) \leq c \times f(n)$



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Example

Let $T(n) = 2n + 3n^3 + 5$. $T(n)$ is in $O(n^3)$ with:

- $(c = 8 \text{ and } n_0 = 1)$ or $(c = 5 \text{ and } n_0 = 2)$

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Principle: the lower-order terms are negligible.

Asymptotic lower bound - best case

“big Omega”

$T(n) = \Omega(f(n))$ iff $\exists c \in \mathbb{R}^+, c > 0$ and $\exists n_0 \in \mathbb{N}, n_0 > 0$
such that $\forall n > n_0: T(n) \geq c \times f(n)$



Asymtotic lower bound - best case



“big Omega”

$T(n) = \Omega(f(n))$ **iif** $\exists c \in \mathbb{R}^+, c > 0$ and $\exists n_0 \in \mathbb{N}, n_0 > 0$
such that $\forall n > n_0: T(n) \geq c \times f(n)$

Example

Let $T(n) = 2n + 3n^3 + 5$. $T(n)$ is in $\Omega(n^3)$ with:

- $(c = 1 \text{ and } n_0 = 1)$

Asymptotic approximating bound - average case



“big Theta”

$T(n) = \Theta(f(n))$ **iff** $\exists c_1, c_2 \in R^+, c_1 > 0, c_2 > 0$ and $\exists n_0 \in N,$
 $n_0 > 0$

such that $\forall n > n_0: c_1 \times f(n) \leq T(n) \leq c_2 \times f(n)$

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such that $\forall n > n_0: c_1 \times f(n) \leq T(n) \leq c_2 \times f(n)$

Property

$T(n) = O(f(n))$ and $T(n) = \Omega(f(n)) \implies T(n) = \Theta(f(n))$

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“big Theta”

$T(n) = \Theta(f(n))$ **iff** $\exists c_1, c_2 \in \mathbb{R}^+, c_1 > 0, c_2 > 0$ and $\exists n_0 \in \mathbb{N}, n_0 > 0$

such that $\forall n > n_0: c_1 \times f(n) \leq T(n) \leq c_2 \times f(n)$

Property

$T(n) = O(f(n))$ and $T(n) = \Omega(f(n)) \implies T(n) = \Theta(f(n))$

Example

Let $T(n) = 2n + 3n^3 + 5$.

So, $T(n)$ is in $O(n^3)$ and in $\Omega(n^3)$.

Consequently, $T(n)$ is in $\Theta(n^3)$.

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Other properties

Not transitive

- $f(n) = n^2; g(n) = n$
- $\Rightarrow f(n) = O(n^2) = g(n)$ but $f(n) \neq g(n)$



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Transitivity

- $f(n) = O(g(n)) \ \& \ g(n) = O(h(n)) \Rightarrow f(n) = O(h(n))$
- $f(n) = \Omega(g(n)) \ \& \ g(n) = \Omega(h(n)) \Rightarrow f(n) = \Omega(h(n))$
- $f(n) = \Theta(g(n)) \ \& \ g(n) = \Theta(h(n)) \Rightarrow f(n) = \Theta(h(n))$



Other properties

Not transitive

- $f(n) = n^2; g(n) = n$
- $\Rightarrow f(n) = O(n^2) = g(n)$ but $f(n) \neq g(n)$

Transitivity

- $f(n) = O(g(n)) \ \& \ g(n) = O(h(n)) \Rightarrow f(n) = O(h(n))$
- $f(n) = \Omega(g(n)) \ \& \ g(n) = \Omega(h(n)) \Rightarrow f(n) = \Omega(h(n))$
- $f(n) = \Theta(g(n)) \ \& \ g(n) = \Theta(h(n)) \Rightarrow f(n) = \Theta(h(n))$

Additivity

- $f(n) = O(h(n)) \ \& \ g(n) = O(h(n)) \Rightarrow f(n) + g(n) = O(h(n))$
- $f(n) = \Omega(h(n)) \ \& \ g(n) = \Omega(h(n)) \Rightarrow f(n) + g(n) = \Omega(h(n))$
- $f(n) = \Theta(h(n)) \ \& \ g(n) = \Theta(h(n)) \Rightarrow f(n) + g(n) = \Theta(h(n))$



Exercise



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- $T(n) = 3 + 5n^2 \Rightarrow T(n) = \Theta(n^2)$?



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- $T(n) = 3 + 5n^2 \Rightarrow T(n) = \Theta(n^2)$?
- if $T(n) = \begin{cases} 2n + 5 & \text{if } n \text{ is even} \\ n^2 - n + 1 & \text{if } n \text{ is odd} \end{cases}$, then $T(n) = O(?)$ and $T(n) = \Omega(?)$.

Exercise

Compare the asymptotic behaviours of

① 2^n and 10^n

② $\log_2 n$ and $\log_{10} n$



Compare the asymptotic behaviours of

- ① 2^n and 10^n
- ② $\log_2 n$ and $\log_{10} n$

- ① Prove that for any positive functions f and g , $f(n) + g(n)$ and $\max(f(n); g(n))$ are asymptotically equivalent.
- ② Give a (necessary and sufficient) condition on positive functions f and g to ensure that $f(n) + g(n)$ and $f(n)$ are asymptotically equivalent.



Common asymptotic behaviours



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Size	Approximate computational time					
n	$\Theta(\log n)$	$\Theta(n)$	$\Theta(n \log n)$	$\Theta(n^2)$	$\Theta(2^n)$	$\Theta(n!)$
10	$3 \cdot 10^{-9}s$	$10^{-8}s$	$3 \cdot 10^{-8}s$	$10^{-7}s$	$10^{-6}s$	$3 \cdot 10^{-3}s$
10^2	$7 \cdot 10^{-9}s$	$10^{-7}s$	$7 \cdot 10^{-7}s$	$10^{-5}s$	$4 \cdot 10^{13}y$	*
10^3	$10^{-8}s$	$10^{-6}s$	$10^{-5}s$	$10^{-3}s$	*	*
10^4	$1, 3 \cdot 10^{-8}s$	$10^{-5}s$	$10^{-4}s$	$10^{-1}s$	*	*
10^5	$1, 7 \cdot 10^{-8}s$	$10^{-4}s$	$2 \cdot 10^{-3}s$	10s	*	*
10^6	$2 \cdot 10^{-8}s$	$10^{-3}s$	$2 \cdot 10^{-2}s$	17m	*	*

Example

- 1 **Var** int: $d = 0$
- 2 **For** i from 1 to n **do**
 - 1 $d = d + 1$
 - 2 $a[i] = a[i] \times a[i] + d \times d$
- 3 **Endfor**



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Example

```
1 (1) Var int:  $d = 0$ 
2 For  $i$  from 1 to  $n$  do
  1  $d = d + 1$ 
  2  $a[i] = a[i] \times a[i] + d \times d$ 
3 Endfor
```



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Example

```
1 (1) Var int:  $d = 0$ 
2 (n) For  $i$  from 1 to  $n$  do
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Example

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    1 (1)  $d = d + 1$ 
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```



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Example

- ① (1) **Var** int: $d = 0$
- ② (n) **For** i from 1 to n **do**
 - ① (1) $d = d + 1$
 - ② (1) $a[i] = a[i] \times a[i] + d \times d$
- ③ **Endfor**



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Example

```
1. (1) Var int:  $d = 0$   
2. ( $n$ ) For  $i$  from 1 to  $n$  do  
    1. (1)  $d = d + 1$   
    2. (1)  $a[i] = a[i] \times a[i] + d \times d$   
3. Endfor
```

Number of elementary operations: $1 + n \times (1 + 1) = 2n + 1$.



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Linear loop example

1. **Var** int: $i = 1$
2. **While** $i \leq n$ **do**
 1. Write “Bonjour”
 2. $i = i + 1$
3. **EndWhile**



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Linear loop example

1. **Var** int: $i = 1$
2. **While** $i \leq n$ **do**
 1. Write "Bonjour"
 2. $i = i + 1$
3. **EndWhile**

1. **Var** int: $i = n$
2. **While** $i \geq 1$ **do**
 1. Write "Bonjour"
 2. $i = i - 1$
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Linear loop example

1. **Var** int: $i = 1$
2. **While** $i \leq n$ **do**
 1. Write "Bonjour"
 2. $i = i + 1$
3. **EndWhile**

1. **Var** int: $i = n$
2. **While** $i \geq 1$ **do**
 1. Write "Bonjour"
 2. $i = i - 1$
3. **EndWhile**

Number of elementary operations: $2n + 1$.



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Logarithmic loop example

1. **Var** int: $i = 1$
2. **While** $i \leq n$ **do**
 1. Write "Bonjour"
 2. $i = i \times 2$
3. **EndWhile**



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Logarithmic loop example

1. **Var** int: $i = 1$
2. **While** $i \leq n$ **do**
 1. Write "Bonjour"
 2. $i = i \times 2$
3. **EndWhile**

1. **Var** int: $i = n$
2. **While** $i \geq 1$ **do**
 1. Write "Bonjour"
 2. $i = i/2$
3. **EndWhile**



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Logarithmic loop example

1. **Var** int: $i = 1$
2. **While** $i \leq n$ **do**
 1. Write "Bonjour"
 2. $i = i \times 2$
3. **EndWhile**

1. **Var** int: $i = n$
2. **While** $i \geq 1$ **do**
 1. Write "Bonjour"
 2. $i = i/2$
3. **EndWhile**

Number of elementary operations: $1 + \log_2(n)$.



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Nested loop example

Nb of iterations = nb of iterations of external loop \times nb of iterations of internal loop



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Nested loop example

Nb of iterations = nb of iterations of external loop \times nb of iterations of internal loop

1. **Var** int: $i = 1$
2. **While** $i \leq n$ **do**
 1. **Var** int: $j = 1$
 2. **While** $j \leq n$ **do**
 1. Write “Bonjour”
 2. $j = j \times 3$
 3. **EndWhile**
 4. $i = i + 1$
3. **EndWhile**



Nested loop example

Nb of iterations = nb of iterations of external loop \times nb of iterations of internal loop

1. **Var** int: $i = 1$
2. **While** $i \leq n$ **do**
 1. **Var** int: $j = 1$
 2. **While** $j \leq n$ **do**
 1. Write “Bonjour”
 2. $j = j \times 3$
 3. **EndWhile**
 4. $i = i + 1$
3. **EndWhile**

Number of elementary operations: $1 + n + n \times \log_3(n)$.





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Function XYZ(array: $a[]$)

1. **Var** int: i
2. **For** i from 1 to n **do**
 1. **Var** int: $t = a[i]$
 2. **Var** int: j
 3. **For** j from $i - 1$ to 0 **do**
 1. $a[j + 1] = a[j]$
 4. **EndFor**
 5. $a[j + 1] = t$
3. **EndFor**



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 2. (1) **Var** int: j
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2. (n) **For** i from 1 to n **do**
 1. (1) **Var** int: $t = a[i]$
 2. (1) **Var** int: j
 3. (?) **For** j from $i - 1$ to 0 **do**
 1. $a[j + 1] = a[j]$
 4. **EndFor**
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 4. **EndFor**
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Counting number of elementary operations

Give algorithms having number of elementary operations as below.

- $T_1(n) = 3 + 5n$
- $T_2(n) = n \log_2 n$
- $T_3(n) = n^3$
- $T_4(n) = (3n)!$
- $T_5(n) = \log_2(3n)$
- $T_6(n) = 2 \log_3(2n)$
- $T_7(n) = n^2 \log_4 n$
- $T_8(n) = \sqrt{n}$
- $T_9(n) = \sqrt[3]{n^2}$
- $T_{10}(n) = 2^n$
- $T_{11}(n) = n!$