

Analysis of Algorithms

- Analysis of algorithms
 - Running time
 - Space
- Asymptotic notation
 - O , Θ , and Ω -notation.
- Experimental analysis of algorithms

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Analysis of Algorithms

- **Goal.** **Determine** and **predict** computational resources and correctness of algorithms.
- **Examples.**
 - Does my route finding algorithm work?
 - How quickly can I answer a query for a route?
 - Can it scale to 10k queries per second?
 - Will it run out of memory with large maps?
 - How many cache-misses does the algorithm generate per query? And how does this affect performance?
- **Primary focus.**
 - Correctness, running time, space usage.
 - Theoretical and experimental analysis.

Running time

- **Running time.** Number of **steps** an algorithm performs on an input of size n .
- **Steps.**
 - Read/write to memory ($x := y$, $A[i]$, $i = i + 1$, ...)
 - Arithmetic/boolean operations ($+$, $-$, $*$, $/$, $\%$, $\&\&$, $\|\|$, $\&$, $|$, \wedge , \sim)
 - Comparisons ($<$, $>$, $=<$, $=>$, $=$, \neq)
 - Program flow (if-then-else, while, for, goto, function call, ..)
- **Terminology.** Running time, time, time complexity.

Running time

- **Worst-case running time.** Maximal running time over all inputs of size n .
- **Best-case running time.** Minimal running time over all inputs of size n .
- **Average-case running time.** Average running time over all inputs of size n .

- **Terminology.** Time = worst-case running time (unless otherwise stated).
- **Other variants.** Amortized, randomized, deterministic, non-deterministic, etc.

Space

- **Space.** Number of **memory cells** used by the algorithm
- **Memory cells.**
 - Variables and pointers/references = 1 memory cells.
 - Array of length k = k memory cells.
- **Terminology.** Space, memory, storage, space complexity.

Analysis of Algorithms

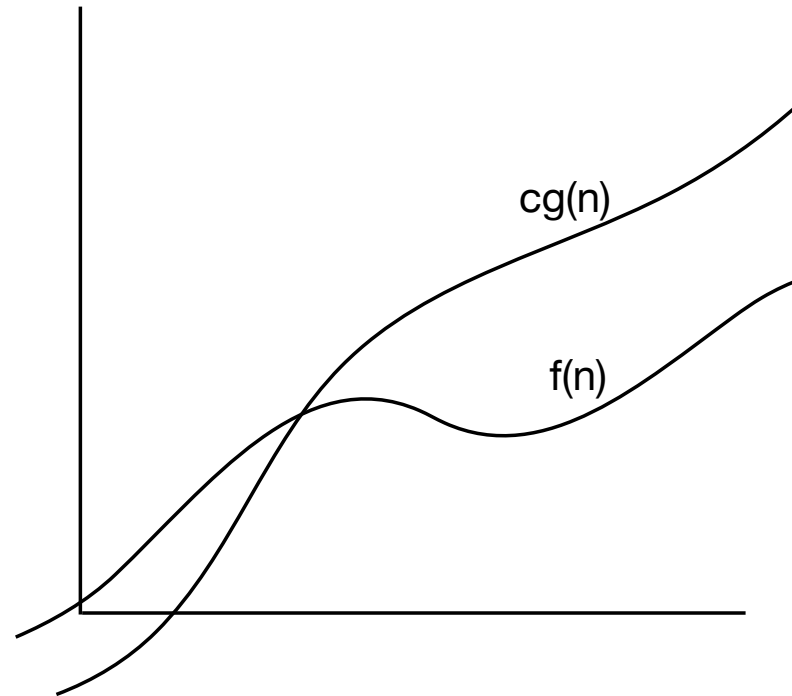
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Asymptotic Notation

- [Asymptotic notation](#).
 - O , Θ , and Ω -notation.
 - Notation to **bound** the **asymptotic** growth of functions.
 - Fundamental tool for talking about time and space of algorithms.

O-notation

- **Definition.** $f(n) = O(g(n))$ if $f(n) \leq cg(n)$ for large n .

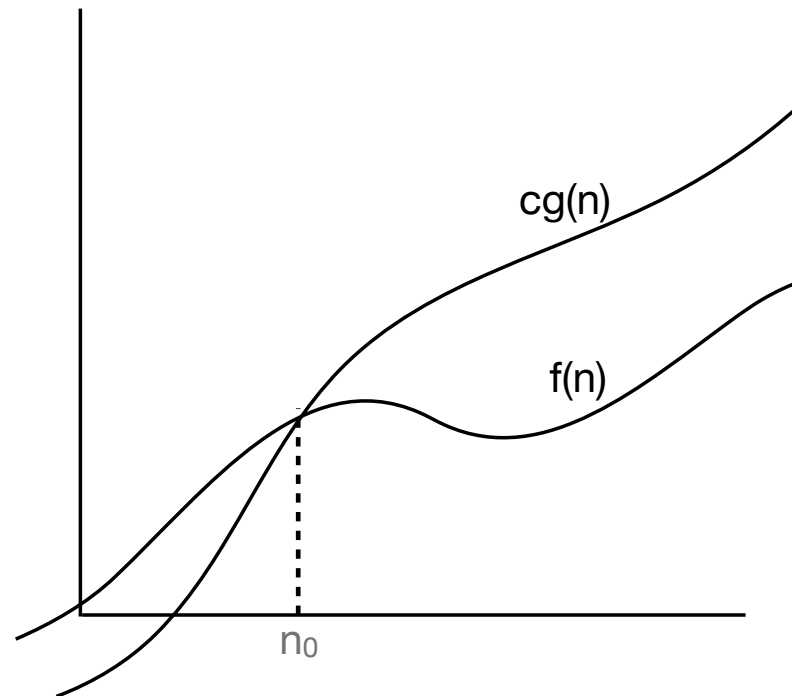


O-notation

- **Example.** $f(n) = O(n^2)$ if $f(n) \leq cn^2$ for large n .
- $5n^2 = O(n^2)$?
 - $5n^2 \leq 5n^2$ for large n .
- $5n^2 + 3 = O(n^2)$?
 - $5n^2 + 3 \leq 6n^2$ for large n .
- $5n^2 + 3n = O(n^2)$?
 - $5n^2 + 3n \leq 6n^2$ for large n .
- $5n^2 + 3n^2 = O(n^2)$?
 - $5n^2 + 3n^2 = 8n^2 \leq 8n^2$ for large n .
- $5n^3 = O(n^2)$?
 - $5n^3 \geq cn^2$ for all constants c for large n .

O-notation

- **Definition.** $f(n) = O(g(n))$ if $f(n) \leq cg(n)$ for large n .
- **Definition.** $f(n) = O(g(n))$ if exists constants $c, n_0 > 0$, such that for all $n \geq n_0$, $f(n) \leq cg(n)$.



O-notation

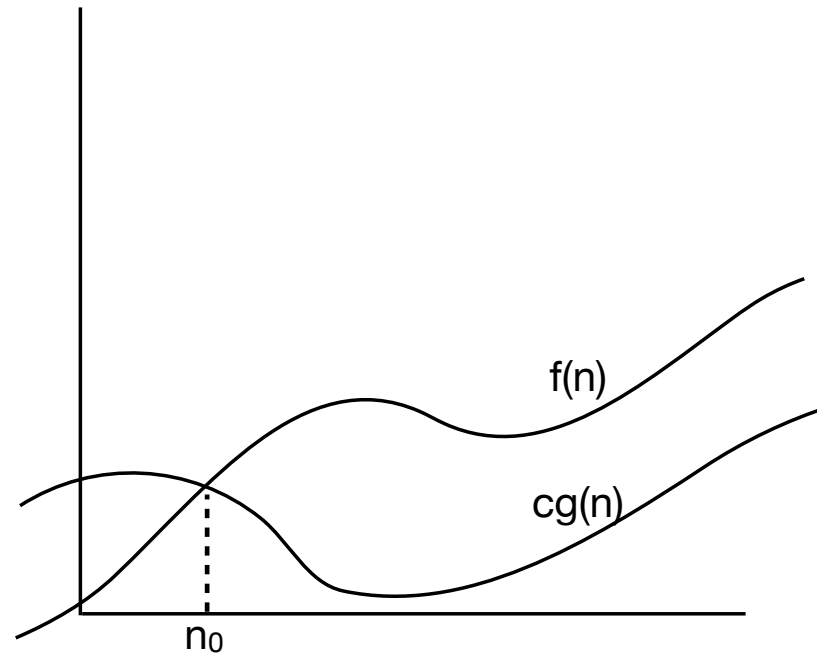
- Notation.
 - $O(g(n))$ is a **set** of functions.
 - Think of $=$ as \in or \subseteq .
 - $f(n) = O(n^2)$ is ok. $O(n^2) = f(n)$ is not!

O-notation

- $f(n) = O(g(n))$ if $f(n) \leq cg(n)$ for large n .
- **Exercise.**
 - Which are true? ($\log^k n$ is $(\log n)^k$)
 - $3n + 2n^3 - n^2 = O(n^2)$
 - $3n^2 + \log n = O(n^3)$
 - $5n^7 + 2^n + = O(n^7)$
 - $n \log^3 n = O(n^2 \log n)$
 - $4n^2 + \log n = O(n^3)$
 - $n(n+3)/1000 + 10000 \log^4 n = O(n^2)$

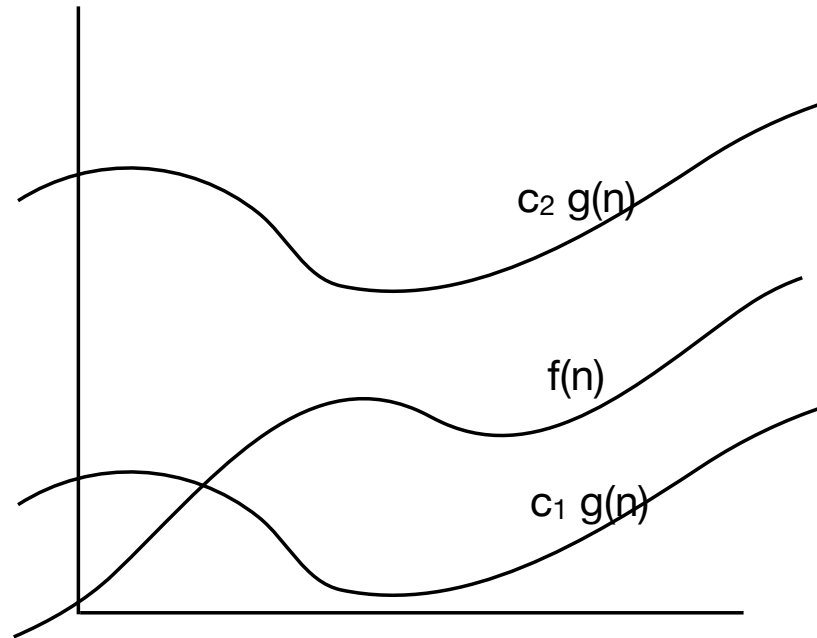
Ω -notation

- **Definition.** $f(n) = \Omega(g(n))$ if $f(n) \geq cg(n)$ for large n .
- **Definition.** $f(n) = \Omega(g(n))$ if exists constants $c, n_0 > 0$, such that for all $n \geq n_0$, $f(n) \geq cg(n)$



Θ -notation

- **Definition.** $f(n) = \Theta(g(n))$ if $f(n) = O(g(n))$ and $f(n) = \Omega(g(n))$



Asymptotic Notation

- $f(n) = O(g(n))$ if $f(n) \leq cg(n)$ for large n .
- $f(n) = \Omega(g(n))$ if $f(n) \geq cg(n)$ for large n .
- $f(n) = \Theta(g(n))$ if $f(n) = O(g(n))$ *and* $f(n) = \Omega(g(n))$.

- **Exercise.** Which are true? ($\log^k n$ is $(\log n)^k$)
 - $n \log^3 n = O(n^2)$
 - $2^n + 5n^7 = \Omega(n^3)$
 - $n^2(n - 5)/5 = \Theta(n^2)$
 - $4 n^{1/100} = \Omega(n)$
 - $n^3/300 + 15 \log n = \Theta(n^3)$
 - $2^{\log n} = O(n)$
 - $\log^2 n + n + 7 = \Omega(\log n)$

Asymptotic Notation

- Basic properties.

- Any polynomial grows proportional to its leading term.

$$a_0 + a_1n + a_2n^2 + \cdots + a_dn^d = \Theta(n^d)$$

- All logarithms are asymptotically the same.

$$\log_a(n) = \frac{\log_b n}{\log_b a} = \Theta(\log_c(n)) \quad \text{for all constants } a, b > 0$$

- Logarithms grow slower than any polynomials.

$$\log(n) = O(n^d) \quad \text{for all } d > 0$$

- Polynomials grow slower than any exponentials.

$$n^d = O(r^n) \quad \text{for all } d > 0 \text{ and } r > 1$$

Typical Running Times

```
for i = 1 to n  
  <  $\theta(1)$  time operation >
```

```
for i = 1 to n  
  for j = 1 to n  
    <  $\theta(1)$  time operation >
```

```
for i = 1 to n  
  for j = i to n  
    <  $\theta(1)$  time operation >
```

Typical Running Times

$$T(n) = \begin{cases} T(n/2) + \Theta(1) & \text{if } n > 1 \\ \Theta(1) & \text{if } n = 1 \end{cases}$$

$$T(n) = \begin{cases} 2T(n/2) + \Theta(n) & \text{if } n > 1 \\ \Theta(1) & \text{if } n = 1 \end{cases}$$

$$T(n) = \begin{cases} 2T(n/2) + \Theta(1) & \text{if } n > 1 \\ \Theta(1) & \text{if } n = 1 \end{cases}$$

$$T(n) = \begin{cases} T(n/2) + \Theta(n) & \text{if } n > 1 \\ \Theta(1) & \text{if } n = 1 \end{cases}$$

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Experimental Analysis

- **Challenge.** Can we experimentally estimate the theoretical running time?
- **Doubling technique.**
 - Run program and measure time for different input sizes.
 - Examine the time increase when we **double** the size of the input.
- **Example.**
- Input size x 2 and time x 4.
- \Rightarrow Algorithm probably runs in quadratic time.
 - $T(n) = cn^2$
 - $T(2n) = c(2n)^2 = c2^2n^2 = c4n^2$
 - $T(2n)/T(n) = 4$

n	time	ratio
5000	0	-
10000	0,2	-
20000	0,6	3
40000	2,3	3,8
80000	9,4	4
160000	37,8	4

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