

Asymptotic Notation, Review of Functions & Summations

Asymptotic Complexity

- ◆ Running time of an algorithm as a function of input size n **for large n** .
- ◆ Expressed using only the **highest-order term** in the expression for the exact running time.
 - ◆ $7n^5 + 2n^4 + 3n^3 + 9n^2 + 4n + 6$
 - ◆ Instead of exact running time, we use *asymptotic notations* such as $O(n^5)$, $\Omega(n)$, $\Theta(n^2)$.
- ◆ Describes behavior of running time functions by setting lower and upper bounds for their values.

Asymptotic Notation

- ◆ $\Theta, O, \Omega, o, \omega$
- ◆ Defined for functions over the natural numbers.
 - ◆ Ex: $f(n) = \Theta(n^2)$.
 - ◆ Describes how $f(n)$ grows in comparison to n^2 .
- ◆ Define a *set* of functions; in practice used to compare two function values.
- ◆ The notations describe different rate-of-growth relations between the defining function and the defined set of functions.

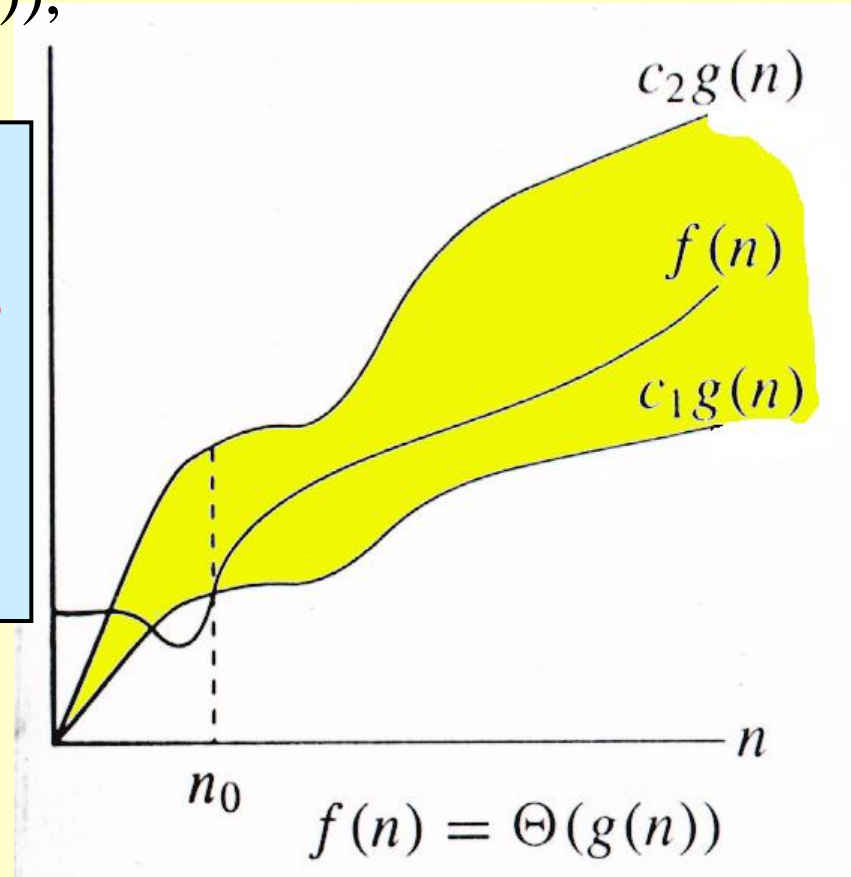
Θ -notation

$g(n) = c$ (a constant), n , n^2 , n^3 , ...

For function $g(n)$, we define $\Theta(g(n))$, big-Theta of n , as a set:

$\Theta(g(n)) = \{f(n) :$
 \exists positive constants c_1, c_2 , and n_0 ,
such that $\forall n \geq n_0$,
we have $0 \leq c_1g(n) \leq f(n) \leq c_2g(n)$
 $\}$

Intuitively: Set of all functions that have the same *rate of growth* as $g(n)$.



$g(n)$ is an *asymptotically tight bound* for any $f(n)$ in the **set**.

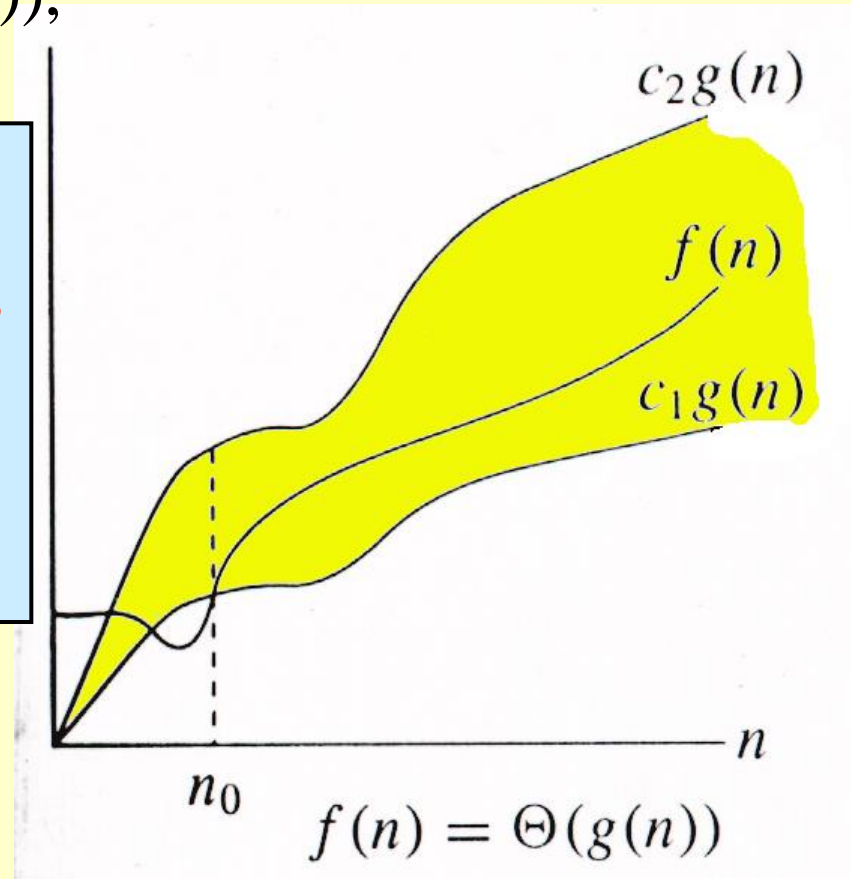
Θ -notation

For function $g(n)$, we define $\Theta(g(n))$, big-Theta of n , as the set:

$\Theta(g(n)) = \{f(n) :$
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 $\}$

Technically, $f(n) \in \Theta(g(n))$.
Older usage, $f(n) = \Theta(g(n))$.
I'll accept either of the forms.

$f(n)$ and $g(n)$ are nonnegative, for large n .



Example

$\Theta(g(n)) = \{f(n) : \exists \text{ positive constants } c_1, c_2, \text{ and } n_0, \text{ such that } \forall n \geq n_0, 0 \leq c_1g(n) \leq f(n) \leq c_2g(n)\}$

- ◆ $10n^2 - 3n = \Theta(n^2)$?
- ◆ What constants for n_0 , c_1 , and c_2 will work?
- ◆ Make c_1 a little smaller than the leading coefficient, and c_2 a little bigger.
- ◆ *To compare orders of growth, look at the leading term (highest-order term).*
- ◆ Exercise: Prove that $n^2/2 - 3n = \Theta(n^2)$

Example

$\Theta(g(n)) = \{f(n) : \exists \text{ positive constants } c_1, c_2, \text{ and } n_0, \text{ such that } \forall n \geq n_0, 0 \leq c_1g(n) \leq f(n) \leq c_2g(n)\}$

- $10n^2 - 3n = \Theta(n^2)$?
- To show that this equation holds, we find $c_1 = 9$, $c_2 = 11$, and $n_0 = 3$ and for $n \geq n_0$, we always have

$$9n^2 \leq 10n^2 - 3n \leq 11n^2.$$

Example

$\Theta(g(n)) = \{f(n) : \exists \text{ positive constants } c_1, c_2, \text{ and } n_0, \text{ such that } \forall n \geq n_0, 0 \leq c_1g(n) \leq f(n) \leq c_2g(n)\}$

- $10n^2 - 3n = \Theta(n^2)$
- $10n^2 - 3n > 9n^2 \quad \Rightarrow \quad n^2 > 3n \quad \Rightarrow \quad n > 3$
- $10n^2 - 3n < 11n^2 \quad \Rightarrow \quad n^2 > -3n \quad \Rightarrow \quad n > -3$

Example

$\Theta(g(n)) = \{f(n) : \exists \text{ positive constants } c_1, c_2, \text{ and } n_0, \text{ such that } \forall n \geq n_0, 0 \leq c_1g(n) \leq f(n) \leq c_2g(n)\}$

- $n^2/2 - 3n = \Theta(n^2)$?
- $c_1 = 1/3 \Rightarrow n^2/2 - 3n > n^2/3$
 $\Rightarrow n^2/6 > 3n \Rightarrow n > 18$
- $c_2 = 1 \Rightarrow n^2/2 - 3n < n^2$
 $\Rightarrow n^2 > -6n \Rightarrow n > -6$
- Then, for $n > n_0 = 18$, we will definitely have
 $n^2/3 < n^2/2 - 3n < n^2$.

Example

$\Theta(g(n)) = \{f(n) : \exists \text{ positive constants } c_1, c_2, \text{ and } n_0, \text{ such that } \forall n \geq n_0, 0 \leq c_1g(n) \leq f(n) \leq c_2g(n)\}$

- ◆ Is $3n^3 \in \Theta(n^4)$?
- ◆ How about $2^{2n} \in \Theta(2^n)$?

Example

$\Theta(g(n)) = \{f(n) : \exists \text{ positive constants } c_1, c_2, \text{ and } n_0, \text{ such that } \forall n \geq n_0, 0 \leq c_1g(n) \leq f(n) \leq c_2g(n)\}$

- Is $3n^3 \in \Theta(n^4)$?
- If it is true, we can find $c_1, c_2,$ and n_0 such that for $n > n_0$, we have

$$c_1n^4 \leq 3n^3 \leq c_2n^4.$$

$$c_1n^4 \leq 3n^3 \Rightarrow n \leq 3/c_1.$$

- It is a contradiction. So, $3n^3 \notin \Theta(n^4)$?

Example

$\Theta(g(n)) = \{f(n) : \exists \text{ positive constants } c_1, c_2, \text{ and } n_0, \text{ such that } \forall n \geq n_0, 0 \leq c_1g(n) \leq f(n) \leq c_2g(n)\}$

- How about $2^{2n} \in \Theta(2^n)$?
- If it is true, we can find c_1, c_2 , and n_0 such that for $n > n_0$, we have

$$c_1 2^n \leq 2^{2n} \leq c_2 2^n.$$

$$2^{2n} \leq c_2 2^n \Rightarrow 2^n \leq c_2 \Rightarrow n \leq \log_2 c_2.$$

- It is a contradiction. So, $2^{2n} \notin \Theta(2^n)$?

O-notation

For function $g(n)$, we define $O(g(n))$, big-O of n , as the set:

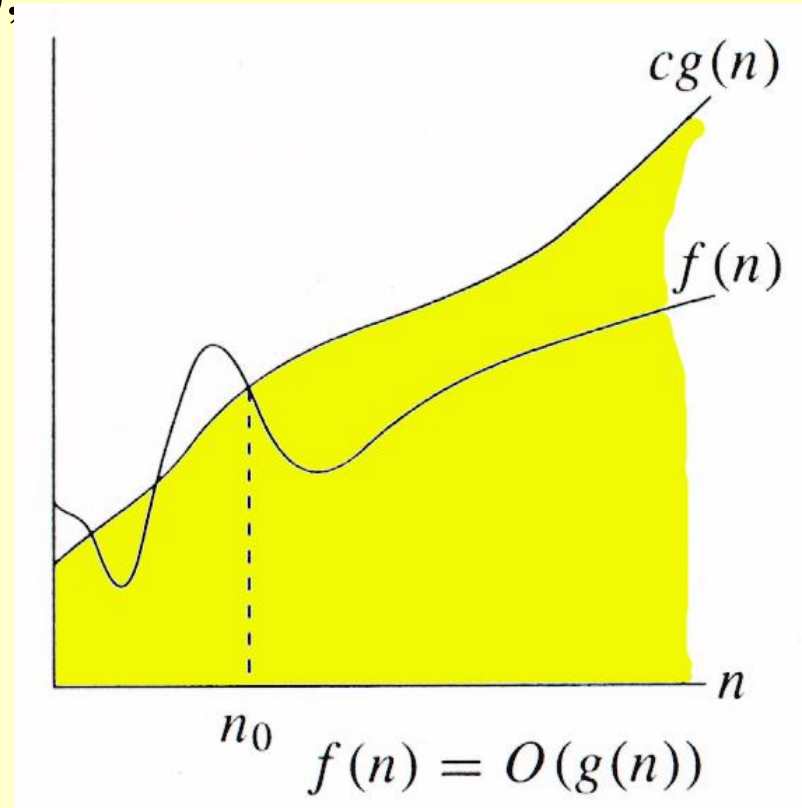
$O(g(n)) = \{f(n) :$
 \exists positive constants c and n_0 ,
such that $\forall n \geq n_0$,
we have $0 \leq f(n) \leq cg(n) \}$

Intuitively: Set of all functions whose *rate of growth* is the same as or lower than that of $g(n)$.

$g(n)$ is an *asymptotic upper bound* for any $f(n)$ in the **set**.

$f(n) = \Theta(g(n)) \Rightarrow f(n) = O(g(n))$.

$\Theta(g(n)) \subset O(g(n))$.



Examples

$O(g(n)) = \{f(n) : \exists \text{ positive constants } c \text{ and } n_0, \text{ such that } \forall n \geq n_0, \text{ we have } 0 \leq f(n) \leq cg(n) \}$

- ◆ Any linear *function* $an + b$ is in $O(n^2)$. **How?**
- ◆ Show that $3n^3 = O(n^4)$ for appropriate c and n_0 .
- ◆ Show that $3n^3 = O(n^3)$ for appropriate c and n_0 .

Examples

$O(g(n)) = \{f(n) : \exists \text{ positive constants } c \text{ and } n_0, \text{ such that } \forall n \geq n_0, \text{ we have } 0 \leq f(n) \leq cg(n) \}$

- Any linear *function* $an + b$ is in $O(n^2)$. **How?**
- To answer this question, we set $c = 1$, to see whether we have $an + b < n^2$ for $n >$ a constant n_0 .
- To determine the value of n_0 , we will solve an equation: $n^2 - an - b = 0$.
- We get $n_0 = \frac{a + \sqrt{a^2 + 4b}}{2}$

Examples

$O(g(n)) = \{f(n) : \exists \text{ positive constants } c \text{ and } n_0, \text{ such that } \forall n \geq n_0, \text{ we have } 0 \leq f(n) \leq cg(n) \}$

- Show that $3n^3 = O(n^4)$ for appropriate c and n_0 .
- The answer is obviously *yes*, since for any $n > n_0 = 4$, we must have $n^4 > 3n^3$.

- Show that $3n^3 = O(n^3)$ for appropriate c and n_0 .
- The answer is also *yes*, since we can take $c = 4$, and for any $n > n_0 = 1$, we must have $cn^3 > 3n^3$.

Ω -notation

For function $g(n)$, we define $\Omega(g(n))$, big-Omega of n , as the set:

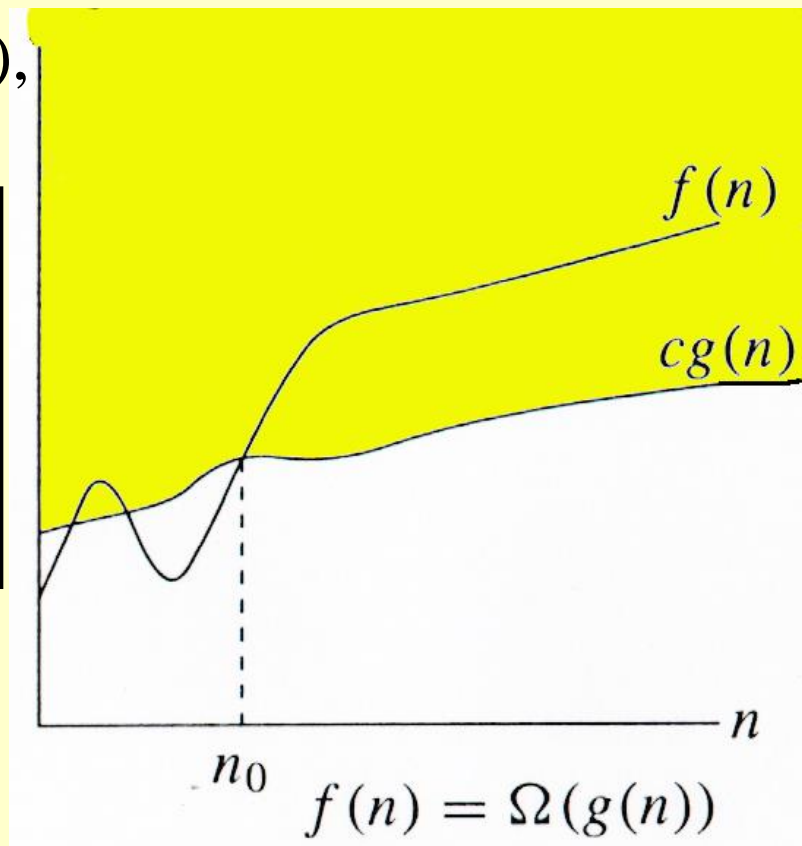
$\Omega(g(n)) = \{f(n) :$
 \exists positive constants c and n_0 ,
such that $\forall n \geq n_0$,
we have $0 \leq cg(n) \leq f(n)\}$

Intuitively: Set of all functions whose *rate of growth* is the same as or higher than that of $g(n)$.

$g(n)$ is an *asymptotic lower bound* for any $f(n)$ in the **set**.

$$f(n) = \Theta(g(n)) \Rightarrow f(n) = \Omega(g(n)).$$

$$\Theta(g(n)) \subset \Omega(g(n)).$$



Example

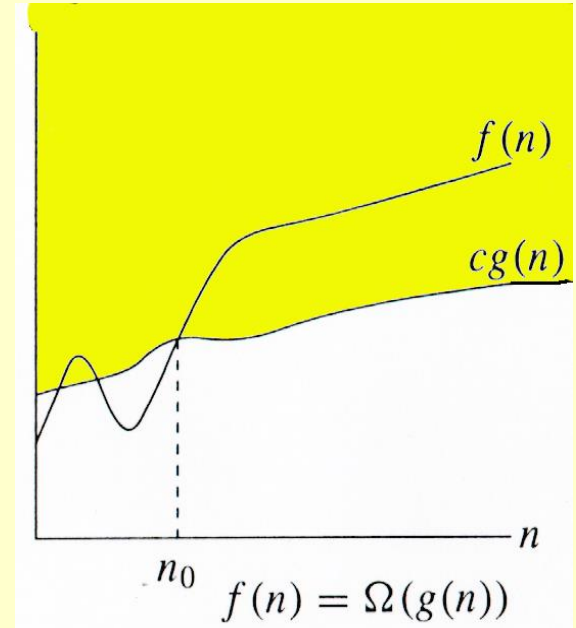
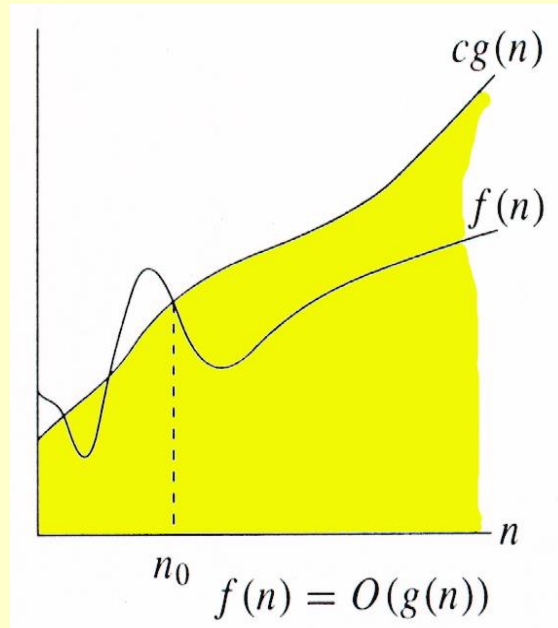
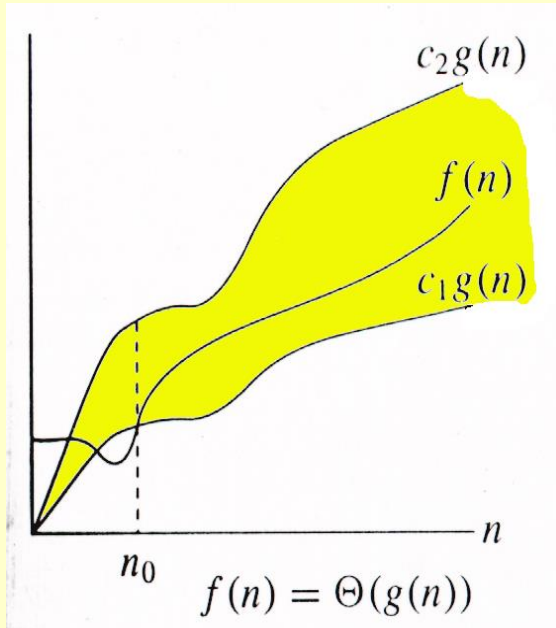
$\Omega(g(n)) = \{f(n) : \exists \text{ positive constants } c \text{ and } n_0, \text{ such that } \forall n \geq n_0, \text{ we have } 0 \leq cg(n) \leq f(n)\}$

- $\sqrt{n} = \Omega(\log_2 n)$. Choose c and n_0 .
- For this purpose, we need to determine constants c and n_0 , such that for any $n \geq n_0$, we have

$$c \log_2 n \leq \sqrt{n}$$

- We can $c = 1$ and $n_0 = 25$ since $\log_2 25 < \log_2 32 = 5 = \sqrt{25}$
- We can also prove that $\sqrt{n} - \log_2 n$ is an increasing function.

Relations Between Θ , O , Ω



Relations Between Θ , Ω , O

Theorem : For any two functions $g(n)$ and $f(n)$,
 $f(n) = \Theta(g(n))$ iff
 $f(n) = O(g(n))$ and $f(n) = \Omega(g(n))$.

- ◆ That is, $\Theta(g(n)) = O(g(n)) \cap \Omega(g(n))$
- ◆ In practice, asymptotically tight bounds are obtained from asymptotic upper and lower bounds.

Running Times

- ◆ “Running time is $O(f(n))$ ” \Rightarrow Worst case is $O(f(n))$
- ◆ $O(f(n))$ bound on the worst-case running time $\Rightarrow O(f(n))$ bound on the running time of every input.
- ◆ $\Theta(f(n))$ bound on the worst-case running time $\not\Rightarrow \Theta(f(n))$ bound on the running time of every input.
- ◆ “Running time is $\Omega(f(n))$ ” \Rightarrow Best case is $\Omega(f(n))$
- ◆ Can still say “Worst-case running time is $\Omega(f(n))$ ”
 - ◆ Means worst-case running time is given by some unspecified function $g(n) \in \Omega(f(n))$.

Example

- ◆ *Insertion sort* takes $\Theta(n^2)$ in the worst case, so sorting (as a *problem*) is $O(n^2)$. Why?
- ◆ Any sort algorithm must look at each item, so sorting is $\Omega(n)$.
- ◆ In fact, using (e.g.) merge sort, sorting is $\Theta(n \lg n)$ in the worst case.
 - ◆ Later, we will prove that we cannot hope that any comparison sort to do better in the worst case.

Asymptotic Notation in Equations

- ◆ Can use asymptotic notation in equations to replace expressions containing lower-order terms.

- ◆ For example,

$$\begin{aligned}4n^3 + 3n^2 + 2n + 1 &= 4n^3 + 3n^2 + \Theta(n) \\ &= 4n^3 + \Theta(n^2) = \Theta(n^3). \quad \text{\textbf{\underline{How to interpret?}}}\end{aligned}$$

- ◆ In equations, $\Theta(f(n))$ always stands for an *anonymous function* $g(n) \in \Theta(f(n))$

- ◆ In the example above, $\Theta(n^2)$ stands for $3n^2 + 2n + 1$.

o-notation

For a given function $g(n)$, the set little- o :

$$o(g(n)) = \{f(n): \forall c > 0, \exists n_0 > 0 \text{ such that} \\ \forall n \geq n_0, \text{ we have } 0 \leq f(n) < cg(n)\}.$$

$f(n)$ becomes insignificant relative to $g(n)$ as n approaches infinity:

$$\lim_{n \rightarrow \infty} [f(n)/g(n)] = 0$$

$g(n)$ is an ***upper bound*** for $f(n)$ that is not asymptotically tight.

Observe the difference in this definition from previous ones. **Why?**

little-o:

$$o(g(n)) = \{f(n):$$

$\forall c > 0, \exists n_0 > 0$ such that $\forall n \geq n_0,$

we have $0 \leq f(n) < cg(n)$ }.

big-O:

$$O(g(n)) = \{f(n) :$$

\exists positive constants c and n_0 , such that $\forall n \geq n_0,$

we have $0 \leq f(n) \leq cg(n)$ }

ω -notation

For a given function $g(n)$, the set little-omega:

$$\omega(g(n)) = \{f(n): \forall c > 0, \exists n_0 > 0 \text{ such that} \\ \forall n \geq n_0, \text{ we have } 0 \leq cg(n) < f(n)\}.$$

$f(n)$ becomes arbitrarily large relative to $g(n)$ as n approaches infinity:

$$\lim_{n \rightarrow \infty} [f(n)/g(n)] = \infty.$$

$g(n)$ is a **lower bound** for $f(n)$ that is not asymptotically tight.

little- ω :

$\omega(g(n)) = \{f(n): \forall c > 0, \exists n_0 > 0 \text{ such that } \forall n \geq n_0,$
we have $0 \leq cg(n) < f(n)\}$.

big- Ω :

$\Omega(g(n)) = \{f(n) :$
 $\exists \text{ positive constants } c \text{ and } n_0, \text{ such that } \forall n \geq n_0,$
we have $0 \leq cg(n) \leq f(n)\}$

Comparison of Functions

$$f \leftrightarrow g \approx a \leftrightarrow b$$

$$f(n) = O(g(n)) \approx a \leq b$$

$$f(n) = \Omega(g(n)) \approx a \geq b$$

$$f(n) = \Theta(g(n)) \approx a = b$$

$$f(n) = o(g(n)) \approx a < b$$

$$f(n) = \omega(g(n)) \approx a > b$$

Limits

- ◆ $\lim_{n \rightarrow \infty} [f(n) / g(n)] = 0 \Rightarrow f(n) \in o(g(n))$
- ◆ $\lim_{n \rightarrow \infty} [f(n) / g(n)] < \infty \Rightarrow f(n) \in O(g(n))$
- ◆ $0 < \lim_{n \rightarrow \infty} [f(n) / g(n)] < \infty \Rightarrow f(n) \in \Theta(g(n))$
- ◆ $0 < \lim_{n \rightarrow \infty} [f(n) / g(n)] \Rightarrow f(n) \in \Omega(g(n))$
- ◆ $\lim_{n \rightarrow \infty} [f(n) / g(n)] = \infty \Rightarrow f(n) \in \omega(g(n))$
- ◆ $\lim_{n \rightarrow \infty} [f(n) / g(n)]$ undefined \Rightarrow can't say

Properties

◆ **Transitivity**

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

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

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

$$f(n) = o(g(n)) \ \& \ g(n) = o(h(n)) \Rightarrow f(n) = o(h(n))$$

$$f(n) = \omega(g(n)) \ \& \ g(n) = \omega(h(n)) \Rightarrow f(n) = \omega(h(n))$$

◆ **Reflexivity**

$$f(n) = \Theta(f(n))$$

$$f(n) = O(f(n))$$

$$f(n) = \Omega(f(n))$$

Properties

◆ **Symmetry**

$$f(n) = \Theta(g(n)) \text{ iff } g(n) = \Theta(f(n))$$

◆ **Complementarity**

$$f(n) = O(g(n)) \text{ iff } g(n) = \Omega(f(n))$$

$$f(n) = o(g(n)) \text{ iff } g(n) = \omega(f(n))$$

Common Functions

Monotonicity

◆ $f(n)$ is

- ◆ **monotonically increasing** if $m \leq n \Rightarrow f(m) \leq f(n)$.
- ◆ **monotonically decreasing** if $m \geq n \Rightarrow f(m) \geq f(n)$.
- ◆ **strictly increasing** if $m < n \Rightarrow f(m) < f(n)$.
- ◆ **strictly decreasing** if $m > n \Rightarrow f(m) > f(n)$.

Exponentials

◆ Useful Identities:

$$a^{-1} = \frac{1}{a}$$

$$(a^m)^n = a^{mn}$$

$$a^m a^n = a^{m+n}$$

◆ Exponentials and polynomials

$$\lim_{n \rightarrow \infty} \frac{n^b}{a^n} = 0$$

$$\Rightarrow n^b = o(a^n)$$

Logarithms

$x = \log_b a$ is the
exponent for $a = b^x$.

Natural log: $\ln a = \log_e a$

Binary log: $\lg a = \log_2 a$

$$\lg^2 a = (\lg a)^2$$

$$\lg \lg a = \lg (\lg a)$$

$$a = b^{\log_b a}$$

$$\log_c (ab) = \log_c a + \log_c b$$

$$\log_b a^n = n \log_b a$$

$$\log_b a = \frac{\log_c a}{\log_c b}$$

$$\log_b (1/a) = -\log_b a$$

$$\log_b a = \frac{1}{\log_a b}$$

$$a^{\log_b c} = c^{\log_b a}$$

Logarithms and exponentials – Bases

- ◆ If the base of a logarithm is changed from one constant to another, the value is altered by a constant factor.
 - ◆ **Ex:** $\log_{10} n * \log_2 10 = \log_2 n$.
 - ◆ Base of logarithm is not an issue in asymptotic notation.
- ◆ Exponentials with different bases differ by an exponential factor (not a constant factor).
 - ◆ **Ex:** $2^n = (2/3)^n * 3^n$.

Polylogarithms

- ◆ **For $a \geq 0, b > 0$,** $\lim_{n \rightarrow \infty} (\lg^a n / n^b) = 0$,
so $\lg^a n = o(n^b)$, and $n^b = \omega(\lg^a n)$
 - ◆ Prove using L'Hopital's rule repeatedly
- ◆ $\lg(n!) = \Theta(n \lg n)$
 - ◆ Prove using Stirling's approximation (in the text) for $\lg(n!)$.

Exercise

Express functions in A in asymptotic notation using functions in B.

A

B

$$5n^2 + 100n$$

$$3n^2 + 2$$

$$A \in \Theta(B)$$

$$A \in \Theta(n^2), n^2 \in \Theta(B) \Rightarrow A \in \Theta(B)$$

$$\log_3(n^2)$$

$$\log_2(n^3)$$

$$A \in \Theta(B)$$

$$\log_b a = \log_c a / \log_c b; A = 2 \lg n / \lg 3, B = 3 \lg n, A/B = 2/(3 \lg 3)$$

$$n^{\lg 4}$$

$$3^{\lg n}$$

$$A \in \omega(B)$$

$$a^{\log b} = b^{\log a}; B = 3^{\lg n} = n^{\lg 3}; A/B = n^{\lg(4/3)} \rightarrow \infty \text{ as } n \rightarrow \infty$$

$$\lg^2 n$$

$$n^{1/2}$$

$$A \in o(B)$$

$$\lim_{n \rightarrow \infty} (\lg^a n / n^b) = 0 \text{ (here } a = 2 \text{ and } b = 1/2) \Rightarrow A \in o(B)$$

Summations – Review

Review on Summations

- ◆ Why do we need summation formulas?

For computing the running times of iterative constructs (loops). (CLRS – Appendix A)

Example: Maximum Subvector

Given an array $A[1 \dots n]$ of numeric values (can be positive, zero, and negative) determine the subvector $A[i \dots j]$ ($1 \leq i \leq j \leq n$) whose sum of elements is maximum over all subvectors.

1	-2	2	2
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Review on Summations

```
MaxSubvector(A, n)
  maxsum ← 0;
  for i ← 1 to n
    do for j = i to n
      sum ← 0
      for k ← i to j
        do sum += A[k]
      maxsum ← max(sum, maxsum)
  return maxsum
```

$$\blacklozenge T(n) = \sum_{i=1}^n \sum_{j=i}^n \sum_{k=i}^j 1$$

◆NOTE: This is not a simplified solution. What *is* the final answer?

Review on Summations

- ◆ **Constant Series:** For integers a and b , $a \leq b$,

$$\sum_{i=a}^b 1 = b - a + 1$$

- ◆ **Linear Series (Arithmetic Series):** For $n \geq 0$,

$$\sum_{i=1}^n i = 1 + 2 + \cdots + n = \frac{n(n+1)}{2}$$

- ◆ **Quadratic Series:** For $n \geq 0$,

$$\sum_{i=1}^n i^2 = 1^2 + 2^2 + \cdots + n^2 = \frac{n(n+1)(2n+1)}{6}$$

Review on Summations

- ◆ **Cubic Series:** For $n \geq 0$,

$$\sum_{i=1}^n i^3 = 1^3 + 2^3 + \cdots + n^3 = \frac{n^2(n+1)^2}{4}$$

- ◆ **Geometric Series:** For real $x \neq 1$,

$$\sum_{k=0}^n x^k = 1 + x + x^2 + \cdots + x^n = \frac{x^{n+1} - 1}{x - 1}$$

For $|x| < 1$,
$$\sum_{k=0}^{\infty} x^k = \frac{1}{1-x}$$

Review on Summations

- ◆ **Linear-Geometric Series:** For $n \geq 0$, real $c \neq 1$,

$$\sum_{i=1}^n ic^i = c + 2c^2 + \cdots + nc^n = \frac{-(n+1)c^{n+1} + nc^{n+2} + c}{(c-1)^2}$$

- ◆ **Harmonic Series:** n th harmonic number, $n \in \mathbb{I}^+$,

$$\begin{aligned} H_n &= 1 + \frac{1}{2} + \frac{1}{3} + \cdots + \frac{1}{n} \\ &= \sum_{k=1}^n \frac{1}{k} = \ln(n) + O(1) \end{aligned}$$

Review on Summations

◆ Telescoping Series:

$$\sum_{k=1}^n a_k - a_{k-1} = a_n - a_0$$

◆ Differentiating Series: For $|x| < 1$,

$$\sum_{k=0}^{\infty} kx^k = \frac{x}{(1-x)^2}$$

Review on Summations

◆ **Approximation by integrals:**

- ◆ For monotonically increasing $f(n)$

$$\int_{m-1}^n f(x)dx \leq \sum_{k=m}^n f(k) \leq \int_m^{n+1} f(x)dx$$

- ◆ For monotonically decreasing $f(n)$

$$\int_m^{n+1} f(x)dx \leq \sum_{k=m}^n f(k) \leq \int_{m-1}^n f(x)dx$$

◆ How?

Review on Summations

◆ *n*th harmonic number

$$\sum_{k=1}^n \frac{1}{k} \geq \int_1^{n+1} \frac{dx}{x} = \ln(n+1)$$

$$\sum_{k=2}^n \frac{1}{k} \leq \int_1^n \frac{dx}{x} = \ln n$$

$$\Rightarrow \sum_{k=1}^n \frac{1}{k} \leq \ln n + 1$$

Reading Assignment

- ◆ Chapter 4 of CLRS.