

**Course Name:  
Analysis and  
Design of  
Algorithms**

# Topics to be covered

- 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.
  - Instead of exact running time, say  $\Theta(n^2)$ .
- Describes behavior of function in the limit.
- Written using ***Asymptotic Notation***.

# 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 sizes.
- The notations describe different rate-of-growth relations between the defining function and the defined set of functions.

# $\Theta$ -notation

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

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

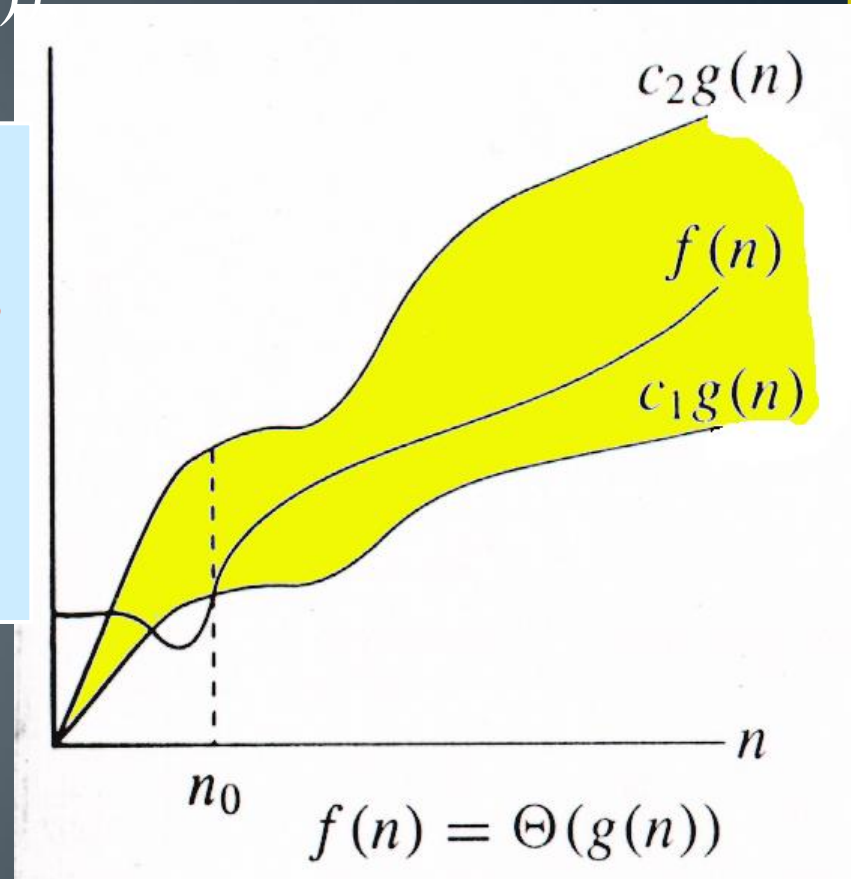
$\exists$  positive constants  $c_1, c_2$ , and  $n_0$ ,  
such that  $\forall n \geq n_0$ ,

$$\text{we have } 0 \leq c_1g(n) \leq f(n) \leq c_2g(n)$$

}

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

$g(n)$  is an *asymptotically tight bound* for  $f(n)$ .



# $\Theta$ -notation

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

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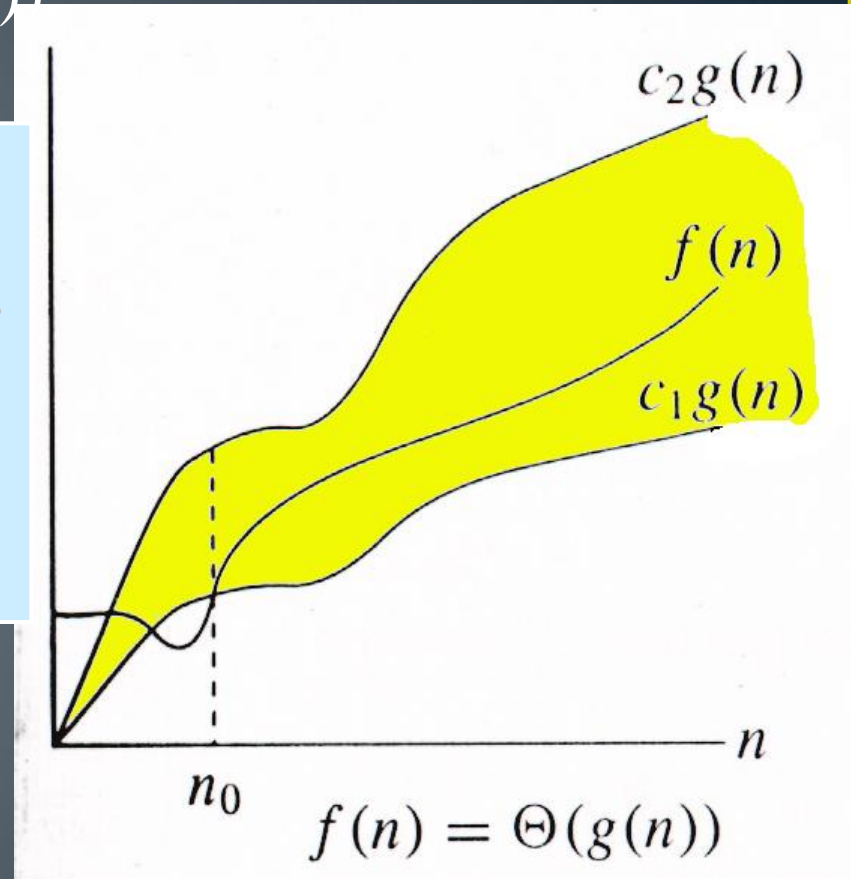
}

Technically,  $f(n) \in \Theta(g(n))$ .

Older usage,  $f(n) = \Theta(g(n))$ .

I'll accept either...

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



# Example

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

- $10n^2 - 3n = \Theta(n^2)$
- 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.***
- Exercise: Prove that  $n^2/2 - 3n = \Theta(n^2)$

# Example

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

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



# O-notation

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

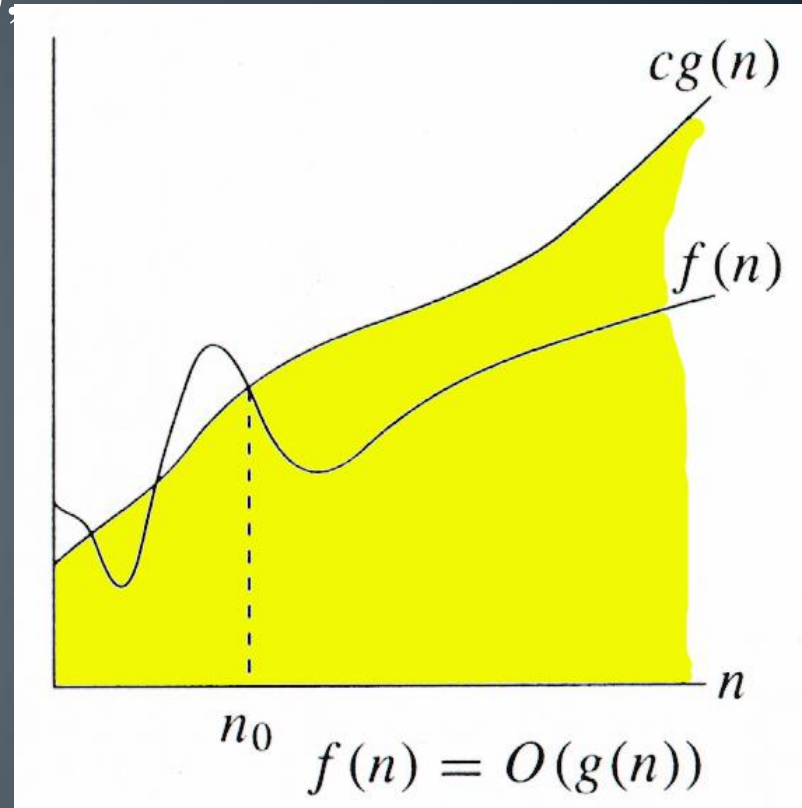
$O(g(n)) = \{f(n) :$   
 **$\exists$  positive constants  $c$  and  $n_0$ ,**  
**such that  $\forall n \geq n_0$ ,**  
**we have  $0 \leq f(n) \leq cg(n)$  }**

*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  $f(n)$ .

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

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



# 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  $n_0$ .

# $\Omega$ -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  $\forall n \geq n_0$ ,

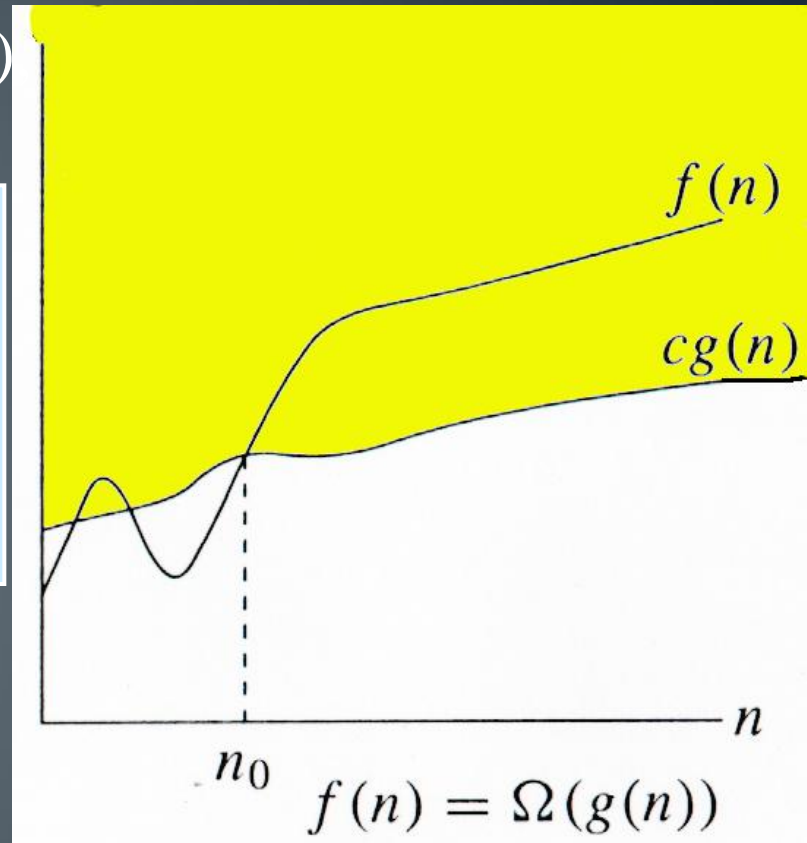
we have  $0 \leq cg(n) \leq f(n)$  }

*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  $f(n)$ .

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

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

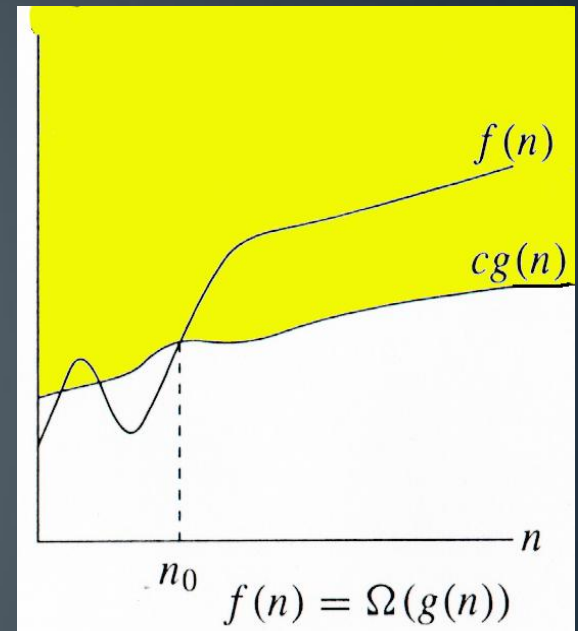
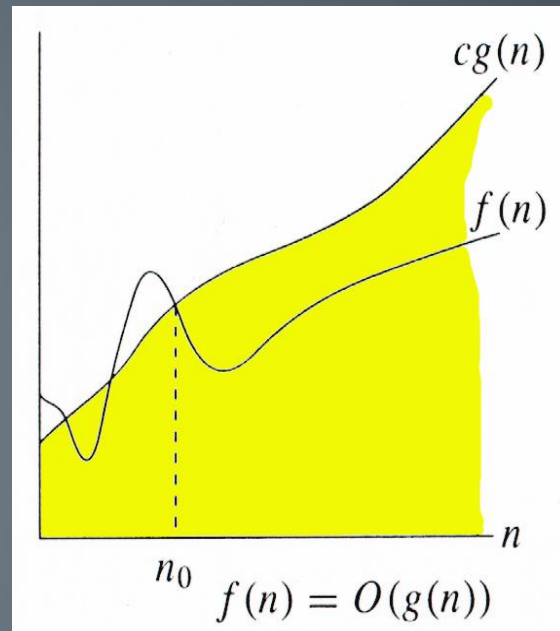
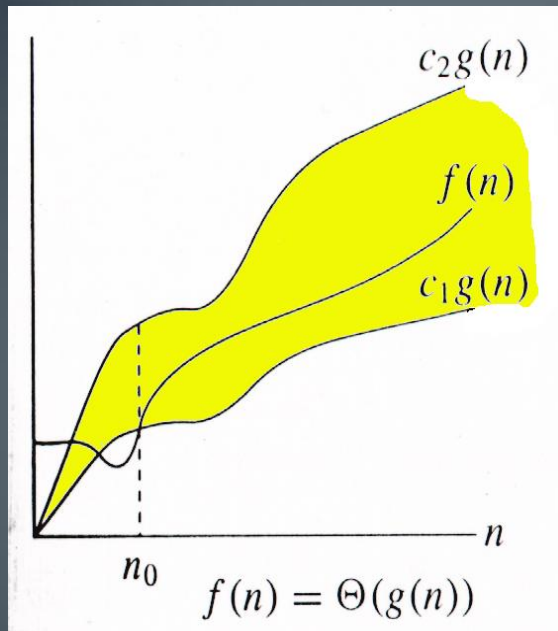


# Example

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

- $\sqrt{n} = \Omega(\lg n)$ . Choose  $c$  and  $n_0$ .

# Relations Between $\Theta$ , $O$ , $\Omega$



## Relations Between $\Theta$ , $O$ , $\Omega$

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

- I.e.,  $\Theta(g(n)) = O(g(n)) \cap \Omega(g(n))$
- 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  $\Rightarrow \Theta(f(n))$  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  $\Omega(f(n))$ ”

# 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,

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

- 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  $3n^2 + 2n + 1$ .

**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?**

**Definition**  
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.

# 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

$$\begin{aligned} f(n) = \Theta(g(n)) \ \& \ g(n) = \Theta(h(n)) \ \Rightarrow \ f(n) = \Theta(h(n)) \\ f(n) = O(g(n)) \ \& \ g(n) = O(h(n)) \ \Rightarrow \ f(n) = O(h(n)) \\ f(n) = \Omega(g(n)) \ \& \ g(n) = \Omega(h(n)) \ \Rightarrow \ f(n) = \Omega(h(n)) \\ f(n) = o(g(n)) \ \& \ g(n) = o(h(n)) \ \Rightarrow \ f(n) = o(h(n)) \\ f(n) = \omega(g(n)) \ \& \ g(n) = \omega(h(n)) \ \Rightarrow \ f(n) = \omega(h(n)) \end{aligned}$$

- **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 a 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  $f(n)$

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

# Review on Summations

- ***n*th harmonic number**

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

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

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