CSE214 – Analysis of Algorithms

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https://github.com/FurkanGozukara/Analysis-of-Algorithms-2019

Lecture 2 Asymptotic Notation

Based on Cevdet Aykanat's and Mustafa Ozdal's Lecture Notes - Bilkent

Logaritma Nedir? Logaritma Formülleri Özellikleri

Logaritma Tanımı: a, b ∈ R⁺ ve a≠1 olmak üzere a^x= b denklemini sağlayan x sayısına log_ab denir ve b'nin a tabanında logaritması diye okunur.

1)
$$\log_{a}x = b$$
 ise $x = a^b$ $\log_2 8 = 3 \mid 8 = 2^3$

2)
$$\log_a(A.B) = \log_a A + \log_a B$$
 $\log_2(4 * 8) = \log_2(32) = 5 = \log_2(4) + \log_2(8) = 2 + 3$

3)
$$\log_a(A/B) = \log_a A - \log_a B$$
 $\log_2(16/4) = \log_2(4) = 2 = \log_2(16) - \log_2(4) = 4 - 2 = 2$

4)
$$\log_a A^n = n \cdot \log_a A$$
 $\log_2 8^2 = \log_2 64 = 6 = 2 * \log_2 8 = 2 * 3 = 6$

5)
$$\log_{a} A^n = \frac{n}{m} \log_a A \quad \log_2 B^2 = \log_8 64 = 2 = (2/3) * \log_2 B = 2/3 * 3 = 2$$

6)
$$\log_{(a^n)} x = \frac{1}{n} . \log_{a^n} \log_{a^n} 8 = \log_{8} 8 = 1 = (1/3) * \log_{2} 8 = 1/3 * 3 = 1$$

7)
$$\log_a x = (\log_b x)/(\log_b a)$$
 [taban değiştirme] $\log_4 16 = 2 = \log_2 16 - \log_2 4 = 4 - 2 = 2$

8)
$$a^{\log_2 x} = 2^{\log_2 8} = 2^3 = 8 = 8$$

9)
$$\log_a \sqrt[n]{A} = \frac{1}{n} \log_a A$$
 $\log_2 \sqrt[3]{8} = \log_2 2 = 1 = \left(\frac{1}{3}\right) * \log_2 8 = \frac{1}{3} * 3 = 1$

10)
$$log_{1/a}x = -log_ax$$

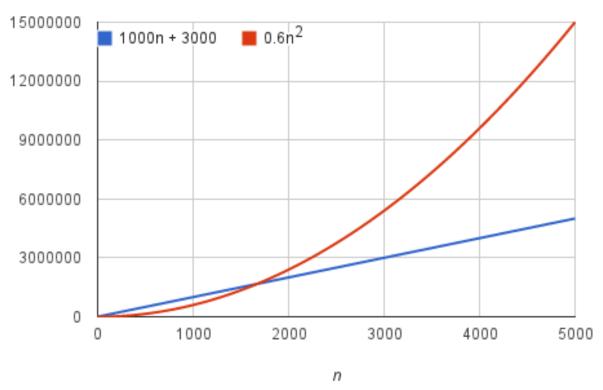
$$log_{1/2}8 = -log_28 = -3 \mid 8 = \left(\frac{1}{2}\right)^{-3}$$

11)
$$\log_a b \cdot \log_b c \cdot \log_c d = \log_a d$$
 $\log_2 4 * \log_4 16 * \log_{16} 256 = 2 * 2 * 2 = 8 = \log_2 256$

12) log_ab=1/log_ba veya log_ab.log_ba=1

What is Asymptotic notation

By dropping the less significant terms and the constant coefficients, we can focus on the important part of an algorithm's running time—its rate of growth—without getting mired in details that complicate our understanding.

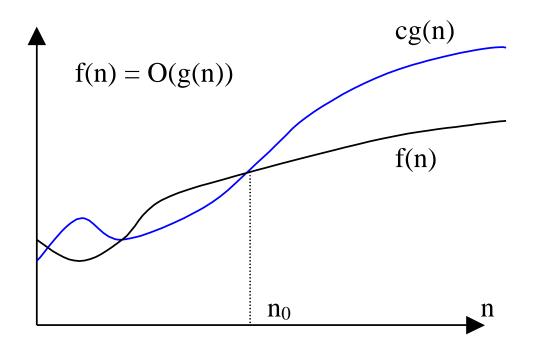


When we drop the constant coefficients and the less significant terms, we use asymptotic notation.

We'll see three forms of it: big-Theta Θ notation, big- \mathbf{O} notation, and big-Omega Ω notation.

O-notation: Asymptotic upper bound

f(n) = O(g(n)) if \exists positive constants c, n_0 such that $0 \le f(n) \le cg(n)$, $\forall n \ge n_0$



Asymptotic running times of algorithms are usually defined by functions whose domain are $N=\{0, 1, 2, ...\}$ (natural numbers)

Example

Show that
$$2n^2 = O(n^3)$$

We need to find two positive constants: \mathbf{c} and $\mathbf{n_0}$ such that:

$$0 \le 2n^2 \le cn^3$$
 for all $n \ge n_0$

Choose
$$c = 2$$
 and $n_0 = 1$
 $2n^2 \le 2n^3$ for all $n \ge 1$

Or, choose
$$c = 1$$
 and $n_0 = 2$
 $2n^2 \le n^3$ for all $n \ge 2$

Example

Show that
$$2n^2 + n = O(n^2)$$

We need to find two positive constants: \mathbf{c} and \mathbf{n}_0 such that:

$$0 \le 2n^2 + n \le cn^2$$
 for all $n \ge n_0$

$$2 + (1/n) \le c$$
 for all $n \ge n_0$

Choose
$$c = 3$$
 and $n_0 = 1$

$$\rightarrow$$
 $2n^2 + n \le 3n^2$ for all $n \ge 1$

O-notation

- \square What does f(n) = O(g(n)) really mean?
 - ☐ The notation is a little sloppy
 - One-way equation
 - e.g. $n^2 = O(n^3)$, but we cannot say $O(n^3) = n^2$

 \square O(g(n)) is in fact a set of functions:

$$O(g(n)) = \{f(n): \exists \text{ positive constants } c, n_0 \text{ such that}$$

 $0 \le f(n) \le cg(n), \forall n \ge n_0 \}$

O-notation

$$0 \le f(n) \le cg(n), \forall n \ge n_0$$

 \square In other words: O(g(n)) is in fact:

the set of functions that have asymptotic upper bound g(n)

$$\square$$
 e.g. $2n^2 = O(n^3)$ means $2n^2 \in O(n^3)$

 $2n^2$ is in the set of functions that have asymptotic upper bound n^3

True or False?

$$10^9 n^2 = O(n^2)$$

True

Choose
$$c = 10^9$$
 and $n_0 = 1$
 $0 \le 10^9 n^2 \le 10^9 n^2$ for $n \ge 1$

$$100n^{1.9999} = O(n^2)$$

True

Choose
$$c = 100$$
 and $n_0 = 1$
 $0 \le 100n^{1.9999} \le 100n^2$ for $n \ge 1$

$$10^{-9}n^{2.0001} = O(n^2)$$

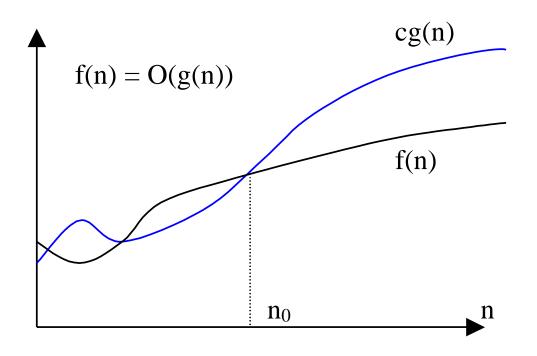
False

$$10^{-9} n^{2.0001} \le cn^2$$
 for $n \ge n_0$

$$10^{-9} \, n^{0.0001} \le c \quad \text{for } n \ge n_0$$
Contradiction

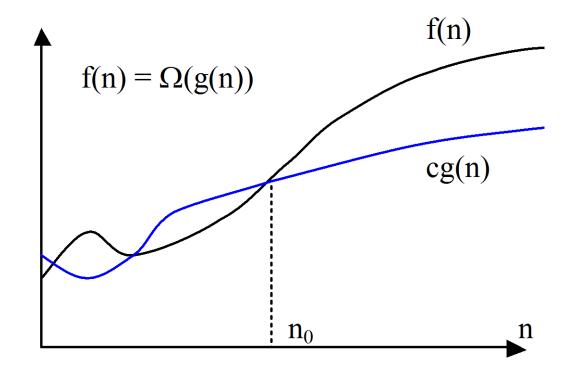
Summary: O-notation: Asymptotic upper bound

 $f(n) \in O(g(n))$ if \exists positive constants c, n_0 such that $0 \le f(n) \le cg(n)$, $\forall n \ge n_0$



Ω -notation: Asymptotic lower bound

 $f(n) = \Omega(g(n))$ if \exists positive constants c, n_0 such that $0 \le cg(n) \le f(n)$, $\forall n \ge n_0$

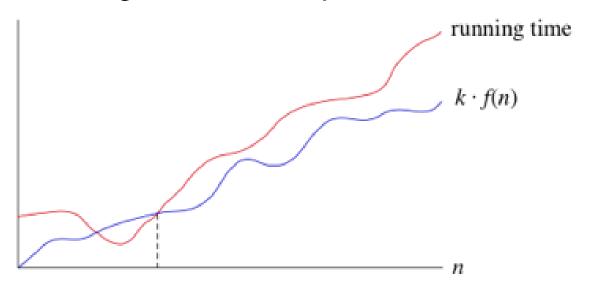


Ω: "big Omega"

What is Asymptotic lower bound

Sometimes, we want to say that an algorithm takes at least a certain amount of time, without providing an upper bound. We use big- Ω notation; that's the Greek letter "omega."

If a running time is $\Omega(f(n))$, then for large enough n, the running time is at least $k \cdot f(n)$, for some constant k. Here's how to think of a running time that is $\Omega(f(n))$:



We say that the running time is "big- Ω of f(n)" We use big- Ω notation for asymptotic lower bounds, since it bounds the growth of the running time from below for large enough input sizes.

Example

Show that
$$2n^3 = \Omega(n^2)$$

We need to find two positive constants: \mathbf{c} and \mathbf{n}_0 such that:

$$0 \le cn^2 \le 2n^3$$
 for all $n \ge n_0$

Choose
$$c = 1$$
 and $n_0 = 1$
 $n^2 \le 2n^3$ for all $n \ge 1$

Example

Show that
$$\sqrt{n} = \Omega(\lg n)$$

We need to find two positive constants: \mathbf{c} and $\mathbf{n_0}$ such that: \mathbf{c} lg $\mathbf{n} \le \sqrt{n}$ for all $\mathbf{n} \ge \mathbf{n_0}$

Choose
$$c = 1$$
 and $n_0 = 16$
 \Rightarrow lg $n \le \sqrt{n}$ for all $n \ge 16$

Ω-notation: Asymptotic Lower Bound

□ $\Omega(g(n)) = \{f(n): \exists \text{ positive constants } c, n_0 \text{ such that } 0 \le cg(n) \le f(n), \forall n \ge n_0 \}$

 \square In other words: Ω (g(n)) is in fact:

the set of functions that have asymptotic lower bound g(n)

True or False?

 $10^{-9} n^{2.0001} = \Omega (n^2)$

True Choose
$$c = 10^9$$
 and $n_0 = 1$ $0 \le 10^9 n^2 \le 10^9 n^2$ for $n \ge 1$
$$cn^2 \le 100 n^{1.9999} = \Omega$$
 False
$$cn^2 \le 100 n^{1.9999} \text{ for } n \ge n_0$$

$$n^{0.0001} \le (100/c) \text{ for } n \ge n_0$$
 Contradiction
$$Choose \ c = 10^9 \text{ and } n_0 = 1$$

True

 $0 \le 10^{-9} n^2 \le 10^{-9} n^{2.0001}$ for $n \ge 1$

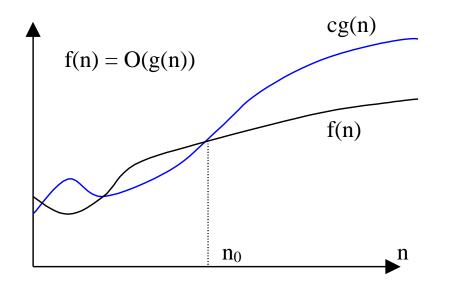
Summary: O-notation and Ω -notation

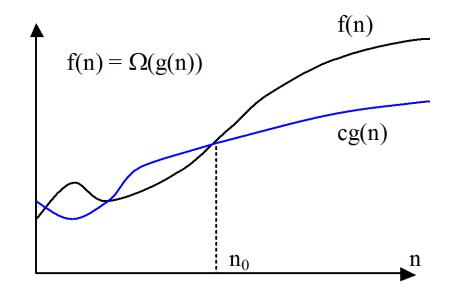
□ O(g(n)): The set of functions with asymptotic upper bound g(n) f(n) = O(g(n)) $f(n) \in O(g(n)) \text{ if } \exists \text{ positive constants } c, n_0 \text{ such that}$

$$0 \le f(n) \le cg(n), \forall n \ge n_0$$

 $\square \ \Omega(g(n)): \text{ The set of functions with asymptotic lower bound } g(n)$ $f(n) = \Omega(g(n))$ $f(n) \in \Omega(g(n)) \ \exists \ positive \ constants \ c, \ n_0 \ such \ that$ $0 \le cg(n) \le f(n), \ \forall n \ge n_0$

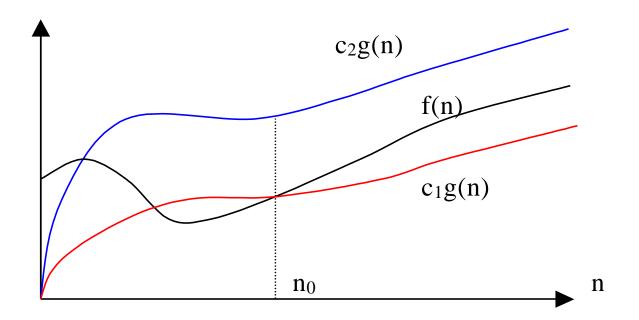
Summary: O-notation and Ω -notation





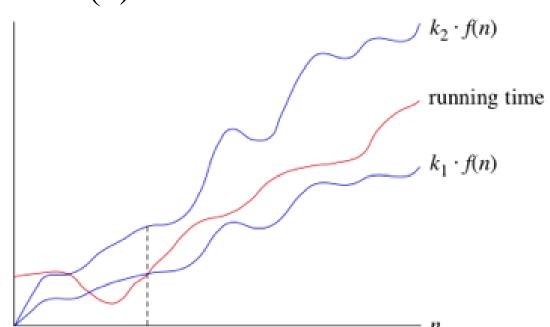
Θ-notation: Asymptotically tight bound

□ $f(n) = \Theta(g(n))$ if \exists positive constants c_1 , c_2 , n_0 such that $0 \le c_1 g(n) \le f(n) \le c_2 g(n)$, $\forall n \ge n_0$



What is Asymptotically tight bound

When we say that a particular running time is $\Theta(n)$, we're saying that once n gets large enough, the running time is at least $k1 \cdot f(n)$ and at most $k2 \cdot f(n)$ for some constants k1 and k2. Here's how to think of $\Theta(n)$:



Once n gets large enough, the running time is between $k1 \cdot f(n)$ and $k2 \cdot f(n)$

Example

Show that
$$2n^2 + n = \Theta(n^2)$$

We need to find 3 positive constants: c_1 , c_2 and n_0 such that:

$$0 \le c_1 n^2 \le 2n^2 + n \le c_2 n^2$$
 for all $n \ge n_0$

$$c_1 \le 2 + (1/n) \le c_2$$
 for all $n \ge n_0$

Choose
$$c_1 = 2$$
, $c_2 = 3$, and $n_0 = 1$

→
$$2n^2 \le 2n^2 + n \le 3n^2$$
 for all $n \ge 1$

Example

Show that
$$\frac{1}{2}n^2 - 2n = \Theta(n^2)$$

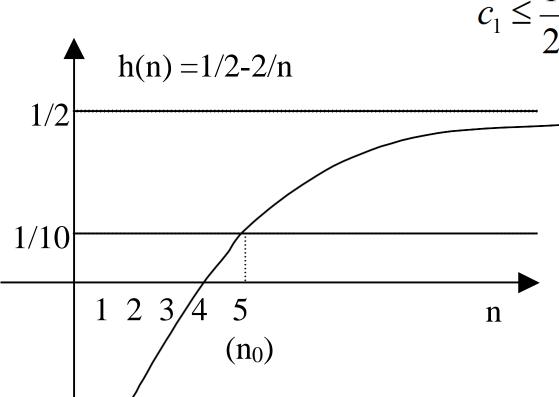
We need to find 3 positive constants: $\mathbf{c_1}$, $\mathbf{c_2}$ and $\mathbf{n_0}$ such that:

$$0 \le c_1 n^2 \le \frac{1}{2} n^2 - 2n \le c_2 n^2 \quad \text{for all } n \ge n_0$$

$$c_1 \le \frac{1}{2} - \frac{2}{n} \le c_2 \quad \text{for all } n \ge n_0$$

Example (cont'd)

 \square Choose 3 positive constants: c_1 , c_2 , n_0 that satisfy:



$$c_1 \le \frac{1}{2} - \frac{2}{n} \le c_2 \qquad \text{for all } n \ge n_0$$

$$\frac{1}{10} \le \frac{1}{2} - \frac{2}{n} \quad \text{for } n \ge 5$$

$$\frac{1}{2} - \frac{2}{n} \le \frac{1}{2} \quad \text{for } n \ge 0$$

Example (cont'd)

 \square Choose 3 constants: c_1 , c_2 , n_0 that satisfy:

$$c_1 \le \frac{1}{2} - \frac{2}{n} \le c_2$$
 for all $n \ge n_0$

$$\frac{1}{10} \le \frac{1}{2} - \frac{2}{n} \quad \text{for } n \ge 5$$

$$\frac{1}{2} - \frac{2}{n} \le \frac{1}{2} \quad \text{for } n \ge 0$$

$$\frac{1}{2} - \frac{2}{n} \le \frac{1}{2} \qquad \text{for } n \ge 0$$

Therefore, we can choose::
$$c_1 = \frac{1}{10}$$
 $c_2 = \frac{1}{2}$ $n_0 = 5$

Θ-notation: Asymptotically tight bound

- □ Theorem: leading constants & low-order terms don't matter
- Justification: can choose the leading constant large enough to make high-order term dominate other terms

True or False?

$$10^9 \mathrm{n}^2 = \Theta (\mathrm{n}^2)$$

True

$$100n^{1.9999} = \Theta(n^2)$$

False

$$10^{-9} n^{2.0001} = \Theta(n^2)$$

False

Θ-notation: Asymptotically tight bound

□ In other words: $\Theta(g(n))$ is in fact:

the set of functions that have asymptotically tight bound g(n)

Θ-notation: Asymptotically tight bound

□ <u>Theorem</u>:

```
f(n) = \Theta(g(n)) if and only if f(n) = O(g(n)) \text{ and } f(n) = \Omega(g(n))
```

- □ In other words:
 - Θ is stronger than both O and Ω

Example

 \square Prove that 10^{-8} $n^2 \neq \Theta(n)$

Before proof, note that $10^{-8}n^2 = \Omega$ (n) but $10^{-8}n^2 \neq O(n)$

Proof by contradiction:

Suppose positive constants c_2 and n_0 exist such that:

$$10^{-8}n^2 \le c_2 n \qquad \text{for all } n \ge n_0$$

$$10^{-8} \text{n} \le \text{c}_2 \qquad \text{for all } \text{n} \ge \text{n}_0$$

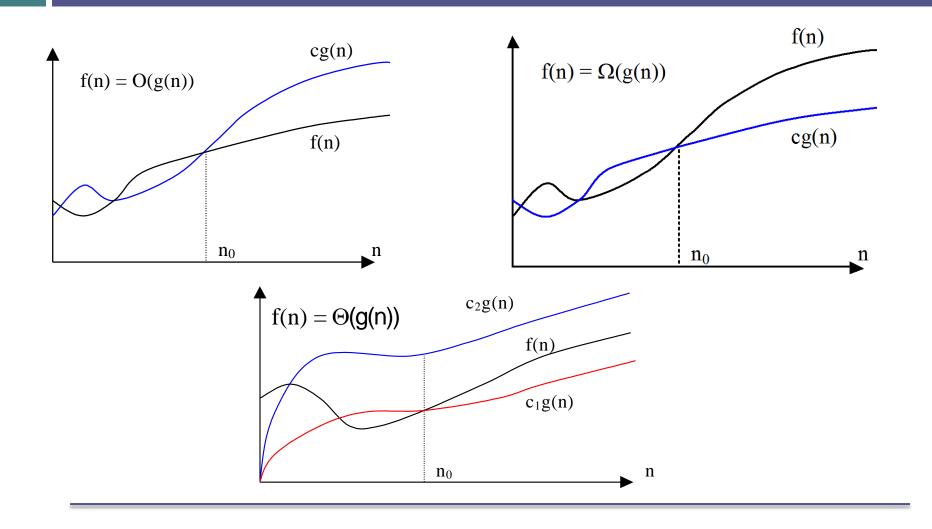
Contradiction: c_2 is a constant

Summary: O, Ω , and Θ notations

 \Box O(g(n)): The set of functions with asymptotic upper bound g(n)

- \square $\Omega(g(n))$: The set of functions with asymptotic lower bound g(n)
- \square $\Theta(g(n))$: The set of functions with asymptotically tight bound g(n)
- \Box f(n) = Θ (g(n)) if and only if f(n) = O(g(n)) and f(n) = Ω (g(n))

Summary: O, Ω , and Θ notations



o ("small o") Notation Asymptotic upper bound that is <u>not tight</u>

Reminder: Upper bound provided by O ("big O") notation can be tight or not tight:

e.g.
$$2n^2 = O(n^2)$$
 is asymptotically tight $2n = O(n^2)$ is not asymptotically tight $n = O(n^2)$ both true

o-Notation: An upper bound that is not asymptotically tight

o ("small o") Notation Asymptotic upper bound that is <u>not tight</u>

□
$$o(g(n)) = \{f(n): \text{ for any constant } c > 0,$$

∃ a constant $n_0 > 0$, such that
$$0 \le f(n) < cg(n), \forall n \ge n_0\}$$

e.g.,
$$2n = o(n^2)$$
, any positive c satisfies
but $2n^2 \neq o(n^2)$, $c = 2$ does not satisfy

ω ("small omega") Notation Asymptotic lower bound that is <u>not tight</u>

□
$$\omega(g(n)) = \{f(n): \text{ for } \underbrace{\text{any}}_{\text{constant } c > 0},$$

∃ a constant $n_0 > 0$, such that
$$0 \le cg(n) < f(n), \forall n \ge n_0\}$$

lim $f(n)$ — ∞

e.g.,
$$n^2/2 = \omega(n)$$
, any positive c satisfies
but $n^2/2 \neq \omega(n^2)$, $c = 1/2$ does not satisfy

Analogy to the comparison of two real numbers

$$\Box f(n) = O(g(n)) \longleftrightarrow a \le b$$

$$\Box f(n) = \Omega(g(n)) \longleftrightarrow a \ge b$$

$$\Box f(n) = \Theta(g(n)) \longleftrightarrow a = b$$

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

True or False?

$5n^2 = O(n^2)$	True	$n^2 \lg n = O(n^2)$	False
$5n^2 = \Omega(n^2)$	True	$n^2 \lg n = \Omega(n^2)$	True
$5n^2 = \Theta(n^2)$	True	$n^2 \lg n = \Theta(n^2)$	False
$5n^2 = o(n^2)$	False	$n^2 \lg n = o(n^2)$	False
$5n^2 = \omega(n^2)$	False	$n^2 \lg n = \omega(n^2)$	True
$2^n = O(3^n)$	True		
$2^n = \Omega(3^n)$	False	$2^n = o(3^n)$	True
$2^n = \Theta(3^n)$	False	$2^n = \omega(3^n)$	False

Asymptotic Growth Rates Comparison

n	constant O(1)	logarithmic O(log n)	linear O(n)	N-log-N O(n log n)	quadratic $O(n^2)$	cubic $O(n^3)$	exponential $O(2^n)$
2	1	1	2	2	4	8	4
4	1	2	4	8	16	64	16
8	1	3	8	24	64	512	256
16	1	4	16	64	256	4,096	65536
32	1	5	32	160	1,024	32,768	4,294,967,296
64	1	6	64	384	4,069	262,144	1.84 x 10 ¹⁹

Using O-Notation to Describe Running Times

- □ Used to bound worst-case running times
 - □ Implies an upper bound runtime for arbitrary inputs as well

□ Example:

"Insertion sort has worst-case runtime of O(n²)"

Note: This $O(n^2)$ upper bound also applies to its running time on every input.

Using O-Notation to Describe Running Times

 \square Abuse to say "running time of insertion sort is $O(n^2)$ "

- □ For a given n, the actual running time <u>depends on</u> the particular input of size n
 - □ i.e., running time is not only a function of n

□ However, worst-case running time is only a function of n

Using O-Notation to Describe Running Times

□ When we say:

"Running time of insertion sort is $O(n^2)$ ",

what we really mean is:

"Worst-case running time of insertion sort is $O(n^2)$ "

or equivalently:

"No matter what particular input of size n is chosen, the running time on that set of inputs is $O(n^2)$ "

Using Ω -Notation to Describe Running Times

- □ Used to bound best-case running times
 - ☐ Implies a lower bound runtime for arbitrary inputs as well

□ Example:

"Insertion sort has best-case runtime of $\Omega(n)$ "

Note: This $\Omega(n)$ lower bound also applies to its running time on every input.

Using Ω -Notation to Describe Running Times

□ When we say:

"Running time of algorithm A is $\Omega(g(n))$ ",

what we mean is:

"For any input of size n, the runtime of A is at least a constant times g(n) for sufficiently large n"

Using Ω -Notation to Describe Running Times

□ *Note*: It's not contradictory to say:

"worst-case running time of insertion sort is $\Omega(n^2)$ "

because there exists an input that causes the algorithm to take $\Omega(n^2)$.

Using Θ-Notation to Describe Running Times

□ Consider 2 cases about the runtime of an algorithm:

- □ Case 1: Worst-case and best-case not asymptotically equal
 - \rightarrow Use Θ -notation to bound worst-case and best-case runtimes separately
- □ <u>Case 2</u>: Worst-case and best-case <u>asymptotically equal</u>
 - \rightarrow Use Θ -notation to bound the runtime for any input

Using Θ-Notation to Describe Running Times Case 1

- □ Case 1: Worst-case and best-case not asymptotically equal
 - → Use Θ -notation to bound the worst-case and best-case runtimes <u>separately</u>
 - □ We can say:
 - "The worst-case runtime of insertion sort is $\Theta(n^2)$ "
 - "The best-case runtime of insertion sort is $\Theta(n)$ "
 - □ But, we can't say:
 - "The runtime of insertion sort is $\Theta(n^2)$ for every input"

Using Θ-Notation to Describe Running Times Case 2

- □ Case 2: Worst-case and best-case asymptotically equal
 - \rightarrow Use Θ -notation to bound the runtime for any input
 - □ e.g. For merge-sort, we have:

$$T(n) = O(nlgn)$$

$$T(n) = \Omega(nlgn)$$

$$T(n) = \Theta(nlgn)$$

Using Asymptotic Notation to Describe Runtimes Summary

- \square "The <u>worst case</u> runtime of Insertion Sort is $O(n^2)$ "
 - \triangleright Also implies: "The runtime of Insertion Sort is $O(n^2)$ "
- \square "The <u>best-case</u> runtime of Insertion Sort is $\Omega(n)$ "
 - \triangleright Also implies: "The runtime of Insertion Sort is $\Omega(n)$ "
- \square "The worst case runtime of Insertion Sort is $\Theta(n^2)$ "
 - \triangleright But: "The runtime of Insertion Sort is not $\Theta(n^2)$ "
- \Box "The <u>best case</u> runtime of Insertion Sort is $\Theta(n)$ "
 - \triangleright But: "The runtime of Insertion Sort is not $\Theta(n)$ "

Using Asymptotic Notation to Describe Runtimes Summary

- \Box "The <u>worst case</u> runtime of Merge Sort is $\Theta(nlgn)$ "
- \Box "The <u>best case</u> runtime of Merge Sort is $\Theta(nlgn)$ "
- \Box "The runtime of Merge Sort is $\Theta(nlgn)$ "
 - > This is true, because the best and worst case runtimes have asymptotically the same tight bound $\Theta(nlgn)$

Asymptotic Notation in Equations

- Asymptotic notation appears <u>alone on the RHS (right hand</u> <u>side)</u> of an equation:
 - implies set membership

e.g.,
$$n = O(n^2)$$
 means $n \in O(n^2)$

- Asymptotic notation appears on the RHS of an equation
 - □ stands for <u>some</u> anonymous function in the

set e.g.,
$$2n^2 + 3n + 1 = 2n^2 + \Theta(n)$$
 means:
$$2n^2 + 3n + 1 = 2n^2 + h(n), \text{ for } \underline{some} \ h(n) \in \Theta(n)$$

i.e.,
$$h(n) = 3n + 1$$

Asymptotic Notation in Equations

- Asymptotic notation appears <u>on the LHS</u> of an equation:
 - > stands for <u>any</u> anonymous function in the set

```
e.g., 2n^2 + \Theta(n) = \Theta(n^2) means:

for <u>any</u> function g(n) \in \Theta(n)

\exists <u>some</u> function h(n) \in \Theta(n^2)

such that 2n^2 + g(n) = h(n)
```

RHS provides coarser level of detail than LHS

Asymptotic Growth Rates Comparison

n	constant O(1)	logarithmic O(log n)	linear O(n)	N-log-N O(n log n)	quadratic $O(n^2)$	cubic $O(n^3)$	exponential $O(2^n)$
2	1	1	2	2	4	8	4
4	1	2	4	8	16	64	16
8	1	3	8	24	64	512	256
16	1	4	16	64	256	4,096	65536
32	1	5	32	160	1,024	32,768	4,294,967,296
64	1	6	64	384	4,069	262,144	1.84 x 10 ¹⁹

Quiz: Asymptotic notation

For the functions, n^k and c^n , what is the asymptotic relationship between these functions?

Assume that $k \ge 1$ and $c \ge 1$ are constants.

- \bigcap n^k is $O(c^n)$
- (B) n^k is $\Omega(c^n)$
- \bigcap n^k is $\Theta(c^n)$

For the functions, n^k and c^n , what is the asymptotic relationship between these functions?

Assume that $k \ge 1$ and $c \ge 1$ are constants.

- CORRECT (SELECTED) n^k is $O(c^n)$
- $igcap_n^k$ is $\Omega(c^n)$
- igorplus INCORRECT n^k is $\Theta(c^n)$

For the functions, $\lg n$ and $\log_8 n$, what is the asymptotic relationship between these functions?

- B $\lg n \text{ is } \Omega(\log_8 n)$
- c $\lg n \text{ is } \Theta(\log_8 n)$

For the functions, $\lg n$ and $\log_8 n$, what is the asymptotic relationship between these functions?

- CORRECT (SELECTED) $\lg n$ is $O(\log_8 n)$
- CORRECT (SELECTED) $\lg n$ is $\Omega(\log_8 n)$
- CORRECT (SELECTED) $\lg n$ is $\Theta(\log_8 n)$

What is the asymptotic relationship between the functions $n^3 \lg n$ and $3n \log_8 n$?

- $n^3 \lg n \text{ is } \Theta(3n \log_8 n)$

What is the asymptotic relationship between the functions $n^3 \lg n$ and $3n \log_8 n$?

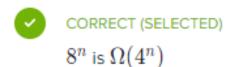
- $\begin{array}{c} \frown & \text{INCORRECT} \\ n^3 \lg n \text{ is } O(3n \log_8 n) \end{array}$
- CORRECT (SELECTED) $n^3 \lg n$ is $\Omega(3n \log_8 n)$
- $\begin{array}{c} \frown & \text{INCORRECT} \\ & n^3 \lg n \text{ is } \Theta(3n \log_8 n) \end{array}$

For the functions, 8^n and 4^n , what is the asymptotic relationship between these functions?

- \bigcirc 8ⁿ is $O(4^n)$
- lacksquare 8 n is $\Omega(4^n)$
- \bigcirc 8ⁿ is $\Theta(4^n)$

For the functions, 8^n and 4^n , what is the asymptotic relationship between these functions?







$$\lg a^b = b \lg a$$

For the functions, $\lg n^{\lg 17}$ vs. $\lg 17^{\lg n}$, what is the asymptotic relationship between these functions?

- B $\lg n^{\lg 17}$ is $\Omega(\lg 17^{\lg n})$
- \bigcirc $\lg n^{\lg 17}$ is $\Theta(\lg 17^{\lg n})$

For the functions, $\lg n^{\lg 17}$ vs. $\lg 17^{\lg n}$, what is the asymptotic relationship between these functions?

- CORRECT (SELECTED) $\lg n^{\lg 17}$ is $O(\lg 17^{\lg n})$
- CORRECT (SELECTED) $\lg n^{\lg 17} \text{ is } \Omega(\lg 17^{\lg n})$
- CORRECT (SELECTED) $\lg n^{\lg 17}$ is $\Theta(\lg 17^{\lg n})$