Asymptotic Analysis



Buffalo once roamed the plains in large numbers.

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.
- Describes behavior of function in the limit.
- Written using Asymptotic Notation.

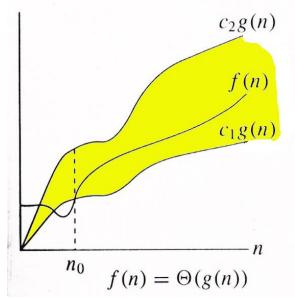
Asymptotic Notation

- \bullet Θ , O, Ω , o, ω
- Defined for functions over the natural numbers.
 - $\mathbf{Ex:} f(n) = \Theta(n^2).$
 - Describes how f(n) grows in comparison to n^2 .
- Defines a set of functions; in practice used to compare growth rate of two functions.
- The notations describe the relationship between the defining function and a defined set of functions.

Θ-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, we have 0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \}
```



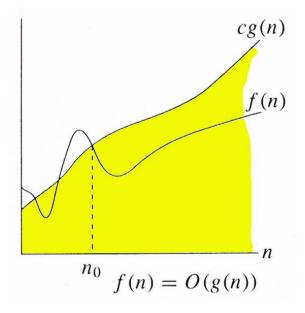
Intuitively: Set of all functions that have the same *rate of growth* as g(n).

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

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).

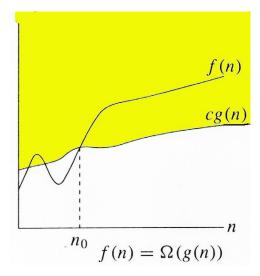
$$f(n) = \Theta(g(n)) \Rightarrow f(n) = O(g(n))$$

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

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)\}
```



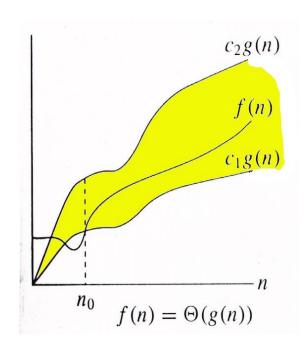
Intuitively: Set of all functions whose *rate of* growth is the same as or higher than that of g(n).

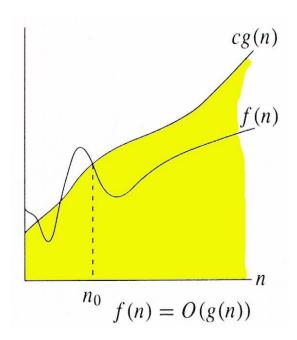
g(n) is an asymptotic lower bound for f(n).

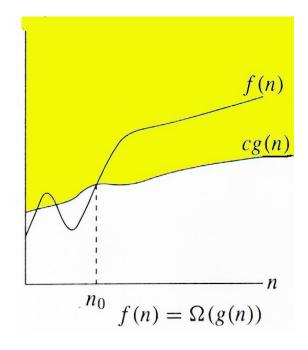
$$f(n) = \Theta(g(n)) \Rightarrow f(n) = \Omega(g(n))$$

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

Relationship Between Θ , O, Ω







Relationship Between Θ , Ω , O

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

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

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 \ge n_0, \text{ we have } 0 \le f(n) < cg(n) \}.
```

Intuitively: Set of all functions whose *rate of* growth is lower than that of g(n).

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

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 \ge n_0, \text{ we have } 0 \le cg(n) < f(n) \}.
```

Intuitively: Set of all functions whose *rate of* growth is higher than that of g(n).

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

Comparison of Functions

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

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

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

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

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

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

Limit Definitions of Asymptotic Notations

- o-notation (Little-o):
 - f(n) becomes insignificant relative to g(n) as n approaches infinity:

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

- \bullet ω -notation (Little-omega):
 - -f(n) becomes arbitrarily large relative to g(n) as n approaches infinity:

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

Limit Definitions of Asymptotic Notations

- ▶ Θ-notation (Big-theta):
 - f(n) relative to g(n) equals a constant, c, which is greater than 0 and less than infinity as n approaches infinity:

$$0 < \lim_{n \to \infty} [f(n) / g(n)] < \infty$$

Limit Definitions of Asymptotic Notations

▶ O-notation (Big-o):

- $-f(n) \in O(g(n)) \Rightarrow f(n) \in \Theta(g(n)) \text{ and } f(n) \in O(g(n))$
- f(n) relative to g(n) equals some value less than infinity as n approaches infinity:

$$\lim_{n\to\infty} \left[f(n) / g(n) \right] < \infty$$

$ightharpoonup \Omega$ -notation (Big-omega):

- $-f(n) \in \Omega(g(n)) \Rightarrow f(n) \in \Theta(g(n)) \text{ and } f(n) \in \omega(g(n))$
- f(n) relative to g(n) equals some value greater than 0 as n approaches infinity:

$$0 < \lim_{n \to \infty} [f(n) / g(n)]$$

Examples

Use limit definitions to prove:

$$-10n - 3n \in O(n^2)$$

$$-3n^4 \in \Omega(n^3)$$

$$- n^2/2 - 3n \in \Theta(n^2)$$

$$-2^{2n}\in\Theta(2^n)$$

Examples

Use limit definitions to prove:

-
$$10n - 3n \in O(n^2)$$

$$\lim_{n \to \infty} [10n - 3n / n^2] = 0$$
- $3n^4 \in \Omega(n^3)$

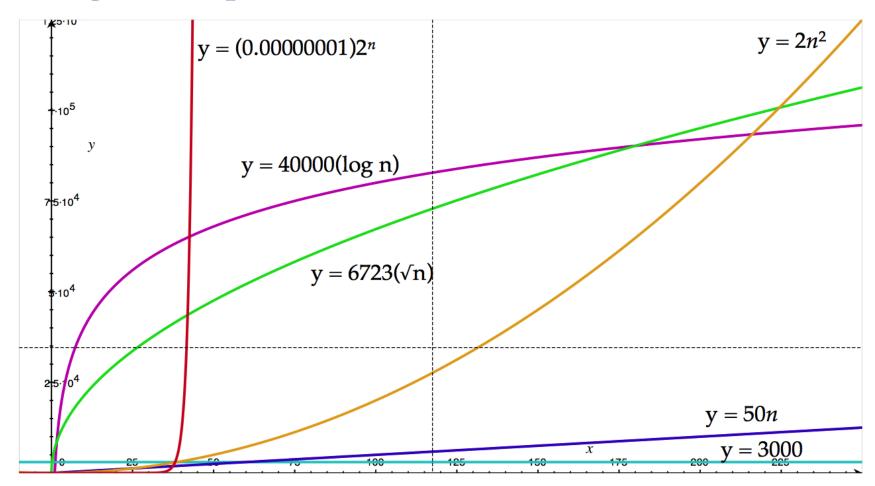
$$\lim_{n \to \infty} [3n^4 / n^3] = \infty$$
- $n^2/2 - 3n \in \Theta(n^2)$

$$\lim_{n \to \infty} [n^2/2 - 3n / n^2] = 1/2$$
- $2^{2n} \in \Theta(2^n)$

$$\lim_{n \to \infty} [2^{2n} / 2^n] = 4$$

Why Do We Care?

Big-O Examples



Complexity and Tractability

				T(n)			
n	n	$n \log n$	n^2	n^3	n^4	n^{10}	2^n
10	.01ms	.03ms	.1ms	1ms	10ms	10s	1ms
20	.02ms	.09ms	.4ms	8ms	160ms	2.84h	1ms
30	.03ms	.15ms	.9ms	27ms	810ms	6.83d	1s
40	.04ms	.21ms	1.6ms	64ms	2.56ms	121d	18m
50	.05ms	.28ms	2.5ms	125ms	6.25ms	3.1y	13d
100	.1ms	.66ms	10ms	1ms	100ms	3171y	$4^{10^{13}}$ y
10^{3}	1ms	9.96ms	1ms	1s	16.67m	$3.17^{\circ}10^{13}$ y	$32^{10^{283}}$ y
10^{4}	10ms	130ms	100ms	16.67m	115.7d	$3.17^{\circ}10^{23}$ y	
10^{5}	100ms	1.66ms	10s	11.57d	3171y	$3.17^{\circ}10^{33}$ y	
10^{6}	1ms	19.92ms	16.67m	31.71y	$3.17^{\circ}10^{7}$ y	$3.17^{\circ}10^{43}$ y	

Assuming the microprocessor performs 1 billion ops/s.