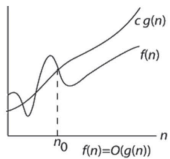


Asymptotic upper bound

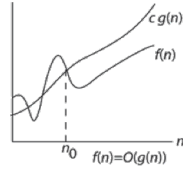
Big-O is like :

$$\begin{aligned}
 x &= O(x) \\
 x &= O(x^2) \\
 x &= O(2^x) \\
 \log x &= O(x) \\
 2^x &= O(2^x) \\
 2^x &= O(3^x)
 \end{aligned}$$



L26: Big-O, Big-Omega and Big-Theta

- Let $f, g: \mathbb{R}^+ \rightarrow \mathbb{R}^+$ *read: f is the big O of g*
- Big-O:** " f is $O(g)$ "
 - Means f grows no faster than g
 - $\exists k, c$ such that for all $n \geq k$ $f(n) \leq cg(n)$

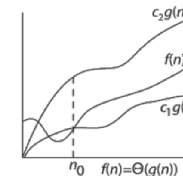


- Big-Omega:** " f is $\Omega(g)$ "
 - Means f grows as least as fast as g
 - $\exists k, c$ such that for all $n \geq k$ $f(n) \geq cg(n)$

*Big-Theta: " f is $\Theta(g)$ "

- Means f grows at the same rate as g
- f is $\Theta(g)$ iff $f = O(g)$ and $f = \Omega(g)$
- $\exists k, c_1, c_2$ such that for all $n > k$:

$$c_1g(n) \leq f(n) \leq c_2g(n)$$



Asymptotic lower bound

Big- Ω ("Big-Omega") is the opposite, like \geq .

$$\begin{aligned}
 f(x) &= O(g(x)) \\
 \text{is the same thing as} \\
 g(x) &= \Omega(f(x))
 \end{aligned}$$

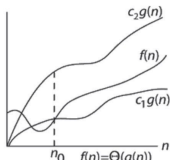
$$\begin{aligned}
 x &= \Omega(x) \\
 x &= \Omega(\log x) \\
 2^x &= \Omega(2^x) \\
 2^x &= \Omega(x)
 \end{aligned}$$

Asymptotic tight bound

Big- Θ is like = :

$$\begin{aligned}
 f(x) &= \Theta(g(x)) \\
 \text{means} \\
 f(x) &= O(g(x)) \\
 \text{AND} \\
 f(x) &= \Omega(g(x))
 \end{aligned}$$

$$\begin{aligned}
 2^x &= \Theta(2^x) \\
 3x^2 + 2 &= \Theta(x^2)
 \end{aligned}$$

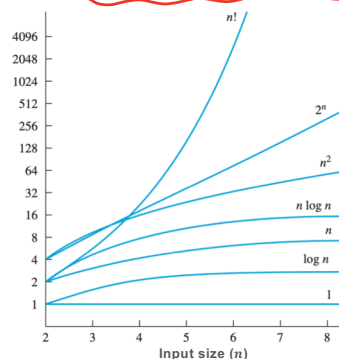


Big-Theta "Cheat Sheet"

Runtime comparison of standard functions

better

worse

 $\Theta(1), \log n, n, n \log n, n^2, n^3, (\text{maybe } n^4, \dots), 2^n, n!$


Scalar multiplication: ignore scalar coefficients

$$f(n) = 1000n^3 \text{ is } \Theta(n^3)$$

Addition: keep largest term

$$f(n) = n^2 + \log n \text{ is } \Theta(n^2)$$

Product: keep all terms

$$f(n) = n^2 \log n \text{ is } \Theta(n^2 \log n)$$

Consider positive-valued functions $f_1(n) = \Theta(g_1(n))$ and $f_2(n) = \Theta(g_2(n))$.• Addition
 $(f_1 + f_2)(n) = \Theta(\max(g_1(n), g_2(n)))$ • Scalar multiplication
 $af(n) = \Theta(f(n))$ • Product
 $(f_1 \cdot f_2)(n) = \Theta(g_1(n) \cdot g_2(n))$

Exercises

1. $f(n) = 5n^3$. Which are true? **DE**
(A) $f = O(\log n)$ (B) $f = O(n)$ (C) $f = O(n^2)$ (D) $f = O((n+1)^3)$ (E) $f = O(n^3)$ 2. $f(n) = n^2 + (n+1)^2 + (n+2)^2 - 100n$. **CDE**
(A) $f = O(\log n)$ (B) $f = O(n)$ (C) $f = O(n^2)$ (D) $f = O((n+1)^3)$ (E) $f = O(n^3)$ 3. $f(n) = \log(5n^3)$. So f is $\Theta(n^3)$. *highest bound*4. $f(n) = (n+1)^3 - n^3 + 5n + 5,000 = 3n^2 + 8n + 5001$
So f is $\Theta(n^2)$ 5. Which is **not** true for the f in Question 4? **E**
(A) $f = \Omega(1)$ (B) $f = \Omega(\log n)$ (C) $f = \Omega(n)$ (D) $f = \Omega(n^2)$ (E) $f = \Omega(n^3)$

1, log n, n, n log n, n^2, n^3, ... n^4, n!

better ← Θ → worse

← Ω →

Even More Examples

Consider functions f and g with $f, g \geq 0 \forall n$ where:

- f is $O(n^3)$ and $\Omega(n)$
- g is $\Theta(\log n)$

Fill in the blanks by finding the largest lower bound and smallest upper bound on h_1 and h_2 .a) If $f + g$ is $\Theta(h_1(n))$, then $n \leq h_1(n) \leq n^3$ b) If $f \cdot g$ is $\Theta(h_2(n))$, then $n \log n \leq h_2(n) \leq n^3 \log n$
 $f_1(n) = \Theta(g_1(n))$ and $f_2(n) = \Theta(g_2(n))$ means there exists constants c_1, c_2, k_1 and c_3, c_4, k_2 such that

$$c_1g_1(n) \leq f_1(n) \leq c_2g_1(n) \text{ for all } n \geq k_1$$

$$c_3g_2(n) \leq f_2(n) \leq c_4g_2(n) \text{ for all } n \geq k_2$$

Adding these inequalities gives

$$c_1g_1(n) + c_3g_2(n) \leq (f_1 + f_2)(n) \leq c_2g_1(n) + c_4g_2(n) \text{ for all } n \geq \max(k_1, k_2)$$

$$\text{Now } c_2g_1(n) + c_4g_2(n) \leq (c_2 + c_4) \max(g_1(n), g_2(n))$$

$$\text{and } c_1g_1(n) + c_3g_2(n) \geq c \max(g_1(n), g_2(n)) \text{ for } c = \min(c_1, c_3)$$

This gives $(f_1 + f_2)(n) = \Theta(\max(g_1(n), g_2(n)))$

Time Complexity of Algorithms

Find the Big- Θ runtime of each algorithm, given its pseudocode:

(a) **procedure** hello_goodbye(n : integer)

```
sum := 0
for i := 1 to 2n
  print "hello world"
  sum := sum + 2
```

n times $\rightarrow \Theta(n)$

```
for i := 1 to n
```

```
  for j := 1 to n
```

```
    print "goodbye"
    sum := sum + 1
```

n^2 times $\rightarrow \Theta(n^2)$

\downarrow
 $\Theta(n^2)$

(b) **procedure** foo(n : integer)

```
a := 0
i := 1
while i < n
  a := a + i
  i := i * 2
return a
```

procedure square_matrix_mult(A : $n \times n$ matrix, B : $n \times n$ matrix)

```
for i := 1 to n
```

```
  for j := 1 to n
```

```
     $c_{ij} := 0$ 
```

```
    for k := 1 to n
```

```
       $c_{ij} := c_{ij} + a_{ik} * b_{kj}$ 
```

```
return C
```

$\Theta(n^3)$

Time Complexity of Linear Search

procedure linear_search(x : integer, a_1, a_2, \dots, a_n : integers)

```
i := 1
```

```
while(i ≤ n and x ≠ ai)
  i := i + 1
```

at most n times
 $\Rightarrow \Theta(n)$

```
if i ≤ n then location := i
else location := 0
```

```
return location
```

Time Complexity of Binary Search

procedure binary_search (x : integer, a_1, a_2, \dots, a_n : increasing integers)

```
i := 1 {i is left endpoint of search interval}
```

```
j := n {j is right endpoint of search interval}
```

```
while i < j
```

```
  m := [(i + j)/2]
```

```
  if x > am then i := m + 1
```

```
  else j := m
```

```
if x = ai then location := i
```

```
else location := 0
```

```
return location {location is the subscript i of the term ai equal to x, or 0 if x is not found}
```

Updates cut remaining list in half each time:

$j-i \approx n$, then $n/2$, then $n/4$, ...

So loop iterates $\log n$ times.

find the midpoint

update i to midpoint, or

update j to midpoint

procedure bar(n : integer)

```
a := (n * n - 7) / 2
```

```
for i := 1 to n
```

```
  j := n
```

```
  while j > 1
```

```
    print "hi"
```

```
    j := j / 2
```

```
print "bye"
```

```
for i := 1 to 500n
```

```
  print "203 is fun!"
```

$\log n$ $n \Rightarrow \Theta(n \log n)$

$500n \Rightarrow \Theta(n)$

Putting it all together, the algorithm is $\Theta(n \log n)$

Let $T(n)$ be the number of steps needed.

$$\Rightarrow T(n) = T\left(\frac{n}{2}\right) + 1 = T\left(\frac{n}{4}\right) + 1 + 1 = T\left(\frac{n}{8}\right) + 1 + 1 + 1 = \dots = T\left(\frac{n}{2^k}\right) + k$$

$$T(1) = 0 \Rightarrow k = \log_2 n$$

$$\Rightarrow T(n) = T(n) + \log_2 n = \log_2 n$$

Examples: Runtimes Galore!

Determine the Θ estimate for each of the following functions.

a) $f(n) = (n^3 + n^2)(n^2 + 50,000)$

$= \Theta(n^5)$ $\underbrace{n^3}_{n^3} \underbrace{n^2}_{n^2}$

e) $f(n) = (3^n + n!)(2^n + n^2)$

$= \Theta(n! 2^n)$ $\underbrace{3^n}_{n!} \underbrace{n!}_{2^n}$

b) $f(n) = \left(5n + \frac{n}{2} + 7\right)(3n^4 + 8n!)$

$= \Theta(n! n)$ $\underbrace{5n}_{n!} \underbrace{n}_{n!} \underbrace{7}_{n!}$

f) $f(n) = (3^n + n!)(2^n + n^2)$

$= \Theta(n! 2^n)$ $\underbrace{3^n}_{n!} \underbrace{n!}_{2^n}$

c) $f(n) = (n^5 + 2^n + \log n)(3n + 4n \log n)$

$= \Theta(2^n n \log n)$ $\underbrace{n^5}_{2^n} \underbrace{2^n}_{n \log n} \underbrace{\log n}_{n \log n}$

g) $f(n) = n^2 \cdot 3^n + \frac{n}{3} \cdot n^3 \log n$

$= \Theta(n^2 3^n)$ $\underbrace{n^2}_{n^2} \underbrace{3^n}_{3^n} \underbrace{\frac{n}{3} \cdot n^3 \log n}_{n^2 \log n}$

d) $f(n) = 10n(\log n^2 + \log n^3)(n^2 + 1)$

$= \Theta(n^2 \log n)$ $\underbrace{10n}_{n^2} \underbrace{(\log n^2 + \log n^3)}_{\log n} \underbrace{(n^2 + 1)}_{n^2}$

$\log n^k = k \log n$
 $\Rightarrow \Theta(\log n)$