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1. What is asymptotic notation?

2. Why use asymptotic notation?

3. Types of asymptotic notation

3.1. Big O notation (O -notation)

O -notation provides an asymptotic **upper bound**.

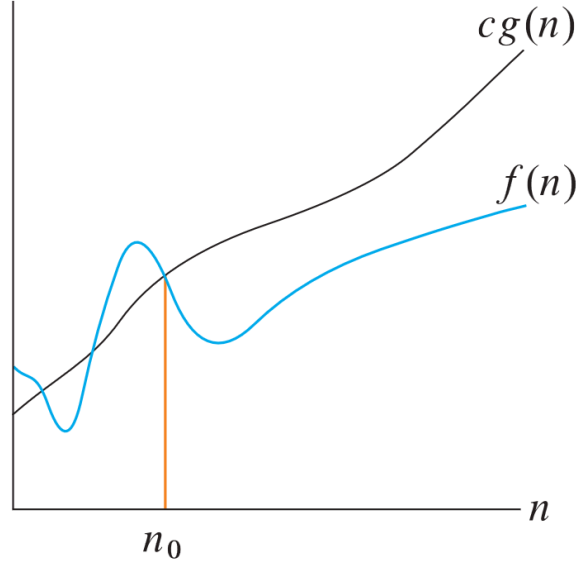


Figure 1: $f(n) = O(g(n))$

Definition 3.1.1:

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

Definition 3.1.2:

$$f(n) := O(g(n)) \Leftrightarrow f(n) \in O(g(n))$$

Example: $\ln(n) = O(n)$

3.2. Big Omega notation (Ω -notation)

Ω -notation provides an asymptotic **lower bound**.

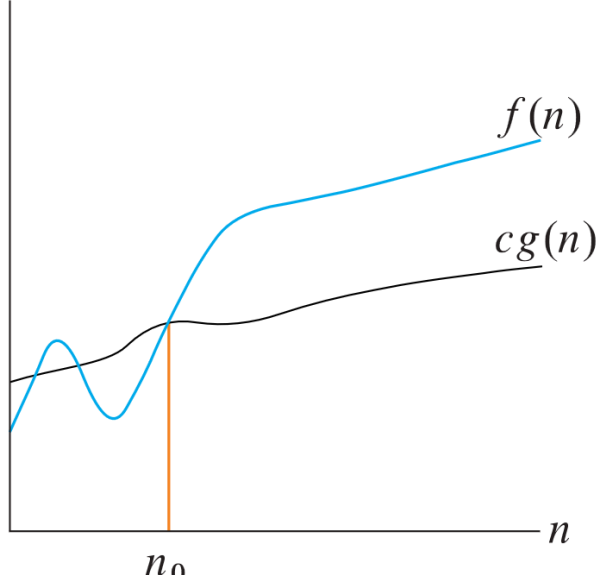


Figure 2: $f(n) = \Omega(g(n))$

Definition 3.2.1:

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

Definition 3.2.2:

$$f(n) := \Omega(g(n)) \Leftrightarrow f(n) \in \Omega(g(n))$$

Example: $n^2 + n = \Omega(n^2)$

3.3. Theta notation (Θ -notation)

Θ -notation provides an asymptotic **tight bound**.

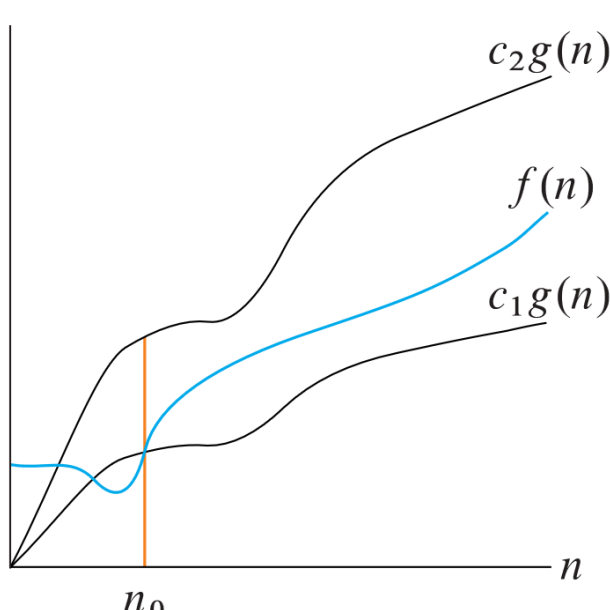


Figure 3: $f(n) = \Theta(g(n))$

Definition 3.3.1:

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

Definition 3.3.2:

$$f(n) := \Theta(g(n)) \Leftrightarrow f(n) \in \Theta(g(n))$$

Example: $\Theta(n^2) = n^2$

3.4. Little o notation (o -notation)

o -notation denotes an **upper bound** that is **not asymptotically tight**

Definition 3.4.1:

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

Proposition 3.4.1:

$$g(n) > 0 \Rightarrow o(g(n)) = \left\{ f(n) : f(n) \geq 0 \text{ and } \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0 \right\}.$$

Definition 3.4.2:

$$f(n) := o(g(n)) \Leftrightarrow f(n) \in o(g(n))$$

Example: $\ln(n) = o(n)$

3.5. Little omega notation (ω -notation)

ω -notation denotes a **lower bound** that is **not asymptotically tight**

Definition 3.5.1:

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

Definition 3.5.2:

$$f(n) := \omega(g(n)) \Leftrightarrow f(n) \in \omega(g(n))$$

Proposition 3.5.1:

$$f(n) := \omega(g(n)) \Rightarrow \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty, \text{ if the limit exists.}$$

Example: $n^2 = \omega(n)$

4. Properties

4.1. Transitivity

$$f(n) = \Theta(g(n)) \text{ and } g(n) = \Theta(h(n)) \Rightarrow f(n) = \Theta(h(n))$$

$$f(n) = O(g(n)) \text{ and } g(n) = O(h(n)) \Rightarrow f(n) = O(h(n))$$

$$f(n) = \Omega(g(n)) \text{ and } g(n) = \Omega(h(n)) \Rightarrow f(n) = \Omega(h(n))$$

$$f(n) = o(g(n)) \text{ and } g(n) = o(h(n)) \Rightarrow f(n) = o(h(n))$$

$$f(n) = \omega(g(n)) \text{ and } g(n) = \omega(h(n)) \Rightarrow f(n) = \omega(h(n))$$

4.2. Reflexivity

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

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

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

4.3. Symmetry

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

4.4. Transpose symmetry

$$f(n) = O(g(n)) \Leftrightarrow g(n) = \Omega(f(n))$$

$$f(n) = o(g(n)) \Leftrightarrow g(n) = \omega(f(n))$$

4.5. Some useful identities

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

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

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

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

5. Common types of asymptotic bound

$$p(n) := \sum_{k=0}^d a_k n^k, \forall k \geq 0 : a_k > 0$$

$$1. p(n) = O(n^k), \forall k \geq d$$

$$2. p(n) = \Omega(n^k), \forall k \leq d$$

$$3. p(n) = \Theta(n^k) \text{ if } k = d$$

$$4. p(n) = o(n^k), \forall k > d$$

$$5. p(n) = \omega(n^k), \forall k < d$$

$$n! = \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \left(1 + \Theta\left(\frac{1}{n}\right)\right)$$

$$\log(n!) = \Theta(n \log(n))$$

6. Methods for proving asymptotic bounds

6.1. Using definitions

Example:

$$\begin{aligned}\ln(n) &\leq n, \forall n \geq 1 \quad (c = 1, n_0 = 1) \\ &\Rightarrow \ln(n) = O(n)\end{aligned}$$

Example:

$$\begin{aligned}0 \leq n^2 &\leq n^2 + n, \forall n \geq 1 \quad (c = 1, n_0 = 1) \\ &\Rightarrow n^2 + n = \Omega(n^2)\end{aligned}$$

Example:

$$\begin{aligned}0 \leq n^2 &\leq n^2 + n \leq 2n^2, \forall n \geq 1 \quad (c_1 = 1, c_2 = 2, n_0 = 1) \\ &\Rightarrow \Theta(n^2) = n^2\end{aligned}$$

Example:

$$\left. \begin{aligned}\ln(n) &\geq 0, \forall n \geq 1 \\ \lim_{n \rightarrow \infty} \frac{\ln(n)}{n} &= \lim_{n \rightarrow \infty} \frac{1}{n} = 0\end{aligned} \right\} \Rightarrow \ln(n) = o(n)$$

Example:

$$\begin{aligned}\forall \varepsilon > 0 : 0 \leq \varepsilon n &< n^2, \forall n \geq \varepsilon + 1 \quad (n_0 = \varepsilon + 1) \\ &\Rightarrow n^2 = \omega(n)\end{aligned}$$

6.2. Substitution method

The substitution method comprises two steps:

- Guess the form of the solution using symbolic constants.
- Use mathematical induction to show that the solution works, and find the constants.

This method is powerful, but it requires experience and creativity to make a good guess.

Example:

$$T(n) := \begin{cases} \Theta(1), & \forall n : 4 > n \geq 2 \\ T(\lfloor \frac{n}{2} \rfloor) + d \quad (d > 0), & \forall n \geq 4 \end{cases}$$

To guess the solution easily, we will assume that: $T(n) = T(\frac{n}{2}) + d$

$$\begin{aligned}T(n) &= T\left(\frac{n}{2}\right) + d \\ &= T\left(\frac{n}{4}\right) + 2d \\ &= T\left(\frac{n}{2^k}\right) + (k-1)d \\ &= T(c) + \left(\log\left(\frac{n}{c}\right) - 1\right)d \\ &= d \log(n) + (T(c) - \log(c) - d)\end{aligned}$$

So we will make a guess: $T(n) = O(\log(n))$

Define $c := \max\{T(2), T(3), d\}$

Assume $T(n) \leq c \log(n)$, $\forall n : k > n$

$$\begin{aligned}T(k) &= T\left(\left\lfloor \frac{k}{2} \right\rfloor\right) + d \\ &\leq c \log\left(\left\lfloor \frac{k}{2} \right\rfloor\right) + d \\ &\leq c \log\left(\frac{k}{2}\right) + d \\ &\leq c \log(k) - c + d \\ &\leq c \log(k) \quad (1)\end{aligned}$$

$$T(n) \leq c \log(n) \quad \forall n : 4 > n \geq 2 \quad (2)$$

From (1), (2) $\Rightarrow T(n) = O(\log(n))$

6.3. Master theorem

Theorem 6.3.1 (Master theorem):

$$T(n) := aT\left(\frac{n}{b}\right) + f(n)$$

where:

- $a > 0$
- $b > 1$
- $\exists n_0 > 0 : f(n) > 0, \forall n \geq n_0$

$$\Rightarrow T(n) = \begin{cases} \Theta(n^{\log_b a}), & \text{if } \exists \varepsilon > 0 : f(n) = O(n^{\log_b a - \varepsilon}) \\ \Theta(n^{\log_b a} \log(n)^{k+1}), & \text{if } \exists k \geq 0 : f(n) = \Theta(n^{\log_b a} \log(n)^k) \\ \Theta(f(n)), & \text{if } \begin{cases} \exists \varepsilon > 0 : f(n) = \Omega(n^{\log_b a + \varepsilon}) \\ \exists n_0 > 0, c < 1 : af(\frac{n}{b}) \leq cf(n), \forall n \geq n_0 \end{cases} \end{cases}$$

Example: Solve the recurrence for merge sort: $T(n) = 2T(\frac{n}{2}) + \Theta(n)$

We have $f(n) = \Theta(n) = \Theta(n^{\log_2 2} \log(n)^0)$, hence

$T(n) = \Theta(n^{\log_2 2} \log(n)^1) = \Theta(n \log(n))$ (according to 2nd case of Theorem 6.3.1)

6.4. Akra-Bazzi method

Theorem 6.4.1 (Akra-Bazzi method):

$$T(x) := g(x) + \sum_{i=1}^k a_i T(b_i x + h_i(x))$$

where:

- $a_i > 0, \forall i \geq 1$
- $0 < b_i < 1, \forall i \geq 1$
- $\exists c \in \mathbb{N} : |g'(x)| = O(x^c)$
- $|h_i(x)| = O\left(\frac{x}{\log(x)^2}\right)$

$$\Rightarrow T(x) = \Theta\left(x^p \left(1 + \int_1^x \frac{g(u)}{u^{p+1}} du\right)\right)$$

where: $\sum_{i=1}^k a_i b_i^p = 1$

Example: Solve the recurrence: $T(x) = T(\frac{x}{2}) + T(\frac{x}{3}) + T(\frac{x}{6}) + x \log(x)$

$$\begin{aligned}|(x \log x)'| &= |\log x + 1| \leq x, \forall x \geq 1 \\ &\Rightarrow |(g(x))'| = O(x) \quad (1)\end{aligned}$$

$$|h_{i(x)}| = 0 = O\left(\frac{x}{\log(x)^2}\right) \quad (2)$$

$$\left(\frac{1}{2}\right)^1 + \left(\frac{1}{3}\right)^1 + \left(\frac{1}{6}\right)^1 = 1 \quad (3)$$

From (1), (2), and (3), we can apply Theorem 6.4.1 to get:

$$\begin{aligned}T(x) &= \Theta\left(x \left(1 + \int_1^x \frac{u \log(u)}{u^2} du\right)\right) \\ &= \Theta\left(x \left(1 + \int_1^x \frac{\log(u)}{u} du\right)\right) \\ &= \Theta\left(x \left(1 + \frac{1}{2} \log(u)^2 \Big|_1^x\right)\right) \\ &= \Theta\left(x \left(1 + \frac{1}{2} \log(x)^2\right)\right) \\ &= \Theta\left(x + \frac{1}{2} x \log(x)^2\right) \\ &= \Theta(x \log(x)^2)\end{aligned}$$

7. Finding asymptotic bound of an algorithm

7.1. Exact step-counting analysis

The asymptotic bound of an algorithm can be calculated by following the steps below:

- Break the program into smaller segments
- Find the number of operations performed in each segment
- Add up all the number of operations, call it $T(n)$
- Find the asymptotic bound of $T(n)$

Example: Analyze insertion sort

We have the following analysis:

INSERTION-SORT(A, n)	<i>cost</i>	<i>times</i>
1 for $i = 2$ to n	c_1	n
2 $key = A[i]$	c_2	$n - 1$
3 <i>// Insert $A[i]$ into the sorted subarray $A[1 : i - 1]$.</i>	0	$n - 1$
4 $j = i - 1$	c_4	$n - 1$
5 while $j > 0$ and $A[j] > key$	c_5	$\sum_{i=2}^n t_i$
6 $A[j + 1] = A[j]$	c_6	$\sum_{i=2}^n (t_i - 1)$
7 $j = j - 1$	c_7	$\sum_{i=2}^n (t_i - 1)$
8 $A[j + 1] = key$	c_8	$n - 1$

Figure 4: Pseudo code for insertion sort with analysis

where:

- c_k denotes the cost of k^{th} line
- t_i denotes the number of times the while loop test in line 5 is executed for given i

From the analysis, we can see that:

$$\begin{aligned}
 T(n) &= c_1 n + c_2(n - 1) + c_4(n - 1) + c_5 \sum_{i=2}^n t_i \\
 &\quad + c_6 \sum_{i=2}^n (t_i - 1) + c_7 \sum_{i=2}^n (t_i - 1) + c_8(n - 1)
 \end{aligned}$$

In the best case (when the array is already sorted), we have $t_i = 1$ for all i .

$$\begin{aligned}
 \Rightarrow T(n) &= c_1 n + c_2(n - 1) + c_4(n - 1) + c_5 \sum_{i=2}^n 1 \\
 &\quad + c_6 \sum_{i=2}^n (1 - 1) + c_7 \sum_{i=2}^n (1 - 1) + c_8(n - 1) \\
 &= c_1 n + c_2(n - 1) + c_4(n - 1) + c_5 n + c_8(n - 1) \\
 &= (c_1 + c_2 + c_4 + c_5 + c_8)n - c_2 - c_4 - c_8 \\
 \Rightarrow T(n) &= \Omega(n)
 \end{aligned}$$

In the worst case, we have $t_i = i$ for all i .

$$\begin{aligned}
 \Rightarrow T(n) &= c_1 n + c_2(n - 1) + c_4(n - 1) + c_5 \sum_{i=2}^n i \\
 &\quad + c_6 \sum_{i=2}^n (i - 1) + c_7 \sum_{i=2}^n (i - 1) + c_8(n - 1) \\
 &= c_1 n + c_2(n - 1) + c_4(n - 1) + c_5 \left(\frac{n(n + 1)}{2} - 1 \right) \\
 &\quad + c_6 \frac{n(n - 1)}{2} + c_7 \frac{n(n - 1)}{2} + c_8(n - 1) \\
 &= \left(\frac{c_5}{2} + \frac{c_6}{2} + \frac{c_7}{2} \right) n^2 \\
 &\quad + \left(c_1 + c_2 + c_4 + \frac{c_5}{2} - \frac{c_6}{2} - \frac{c_7}{2} + c_8 \right) n - c_2 - c_4 - c_5 - c_8 \\
 \Rightarrow T(n) &= O(n^2)
 \end{aligned}$$

In conclusion, we have $T(n) = \Omega(n)$ and $T(n) = O(n^2)$

7.2. Recurrence relation

Example: Calculate asymptotic bound of merge sort

Define $T(n)$ as the running time of the algorithm.

From the implementation of merge sort, we have: $T(n) = 2T\left(\frac{n}{2}\right) + \Theta(n)$

Applying Theorem 6.3.1, we can conclude that $T(n) = \Theta(n \log(n))$ (2nd case with $b = 2$, $a = 2$, $k = 0$)

8. References

- <https://mitpress.mit.edu/9780262046305/introduction-to-algorithms/>
- [https://en.wikipedia.org/wiki/Master_theorem_\(analysis_of_algorithms\)](https://en.wikipedia.org/wiki/Master_theorem_(analysis_of_algorithms))
- https://en.wikipedia.org/wiki/Akra%E2%80%93Bazzi_method#Formulation
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