## **Asymptotic Notations**

## **Definitions and Properties**

**Definition 1** (Big-O). Let f = f(n) and g = g(n) be two positive functions over integers n. We way f = O(g), if there exists positive number c > 0 and integer  $N \ge 0$  such that  $f(n) \le c \cdot g(n)$  for all  $n \ge N$ .

Similarly, we can define Big-O for multiple-variable functions.

**Definition 2** (Big-O). Let f = f(m,n) and g = g(m,n) be two positive functions over integers m and n. We way f = O(g), if there exists positive number c > 0 and integers  $M \ge 0$  and  $N \ge 0$  such that  $f(m,n) \le c \cdot g(m,n)$  for all  $m \ge M$  and  $n \ge N$ .

Intuitively, Big-O is analogous to "\le \".

Example. Let f(m,n) = 4m + 4n + 5 and g(m,n) = m + n. We now show that f = O(g), using above definition. To show it, we need to find c, M, and N. What are good choices for them? There are lots of choices; one set of it is: c = 7, M = 1, and N = 1. Let's verify:  $f(m,n) - c \cdot g(m,n) = 4m + 4n + 5 - 7m - 7n = 5 - 3m - 3n \le 5 - 3 - 3 = -1 \le 0$ , where we use that  $m \ge M = 1$  and  $n \ge N = 1$ . This proves that f = O(g).

**Definition 3** (Big-Omega). Let f = f(n) and g = g(n) be two positive functions over integers n. We way  $f = \Omega(g)$ , if there exists positive number c > 0 and integer  $N \ge 0$  such that  $f(n) \ge c \cdot g(n)$  for all  $n \ge N$ .

Similarly, we can define Big-Omega for multiple-variable functions.

**Definition 4** (Big-O). Let f = f(m,n) and g = g(m,n) be two positive functions over integers m and n. We way  $f = \Omega(g)$ , if there exists positive number c > 0 and integers  $M \ge 0$  and  $N \ge 0$  such that  $f(m,n) \ge c \cdot g(m,n)$  for all  $m \ge M$  and  $n \ge N$ .

Intuitively, Big-Omega is analogous to "\ge "."

Example. Let f(m,n) = 4m + 4n + 5 and g(m,n) = m + n. We now show that  $f = \Omega(g)$ , using above definition. To show it, we need to find c, M, and N. We can choose: c = 1, M = 0, and N = 0. Let's verify:  $f(m,n) - c \cdot g(m,n) = 4m + 4n + 5 - m - n = 5 + 3m + 3n \ge 5 \ge 0$ , where we use that  $m \ge M = 0$  and  $n \ge N = 0$ . This proves that  $f = \Omega(g)$ .

**Claim 1.** f = O(g) if and only if  $g = \Omega(f)$ .

*Proof.* We have

$$\begin{split} f &= O(g) \\ \Leftrightarrow &\; \exists \; c > 0, N \geq 0, \; \text{s.t.} \; f(n) \leq c \cdot g(n), \forall n \geq N \\ \Leftrightarrow &\; \exists \; c > 0, N \geq 0, \; \text{s.t.} \; 1/c \cdot f(n) \leq g(n), \forall n \geq N \\ \Leftrightarrow &\; \exists \; c' = 1/c > 0, N \geq 0, \; \text{s.t.} \; g(n) \geq c' \cdot f(n), \forall n \geq N \\ \Leftrightarrow &\; g = \Omega(f) \end{split}$$

**Definition 5** (Big-Theta). We say  $f = \Theta(g)$  if and only if f = O(g) and  $f = \Omega(g)$ .

Intuitively, Big-Theta is analogous to "=".

Example. Let f(m,n) = 4m + 4n + 5 and g(m,n) = m + n. We have  $f = \Theta(g)$  as we proved that f = O(g)

Below we give an equivalent description of Big-Theta.

**Fact 1.** Let f and g be two positive functions. Then  $f = \Theta(g)$  if and only if  $\lim_{n \to \infty} f(n)/g(n) = c > 0$ .

Example. Let f(m,n) = 4m + 4n + 5 and g(m,n) = m + n. We now show  $f = \Theta(g)$  using above fact.  $\lim_{m \to \infty, n \to \infty} f/g = \lim_{m \to \infty, n \to \infty} (4m + 4n + 5)/(m + n) = 4 > 0$ . Hence,  $f = \Theta(g)$ .

**Definition 6** (small-o). Let f = f(n) and g = g(n) be two positive functions. We say f = o(g) if and only if  $\lim_{n\to\infty} f(n)/g(n) = 0$ .

Intuitively, small-o is analogous to "<".

*Example.* Let f(n) = n and  $g(n) = n^2$ . We now show f = o(g) using above definition.  $\lim_{n \to \infty} f/g = \lim_{n \to \infty} 1/n = 0$ . Hence, f = o(g).

## **Commonly-Used Functions in Algorithm Analysis**

In theoretical computer science, we often see following categories of functions.

- 1. logarithmic functions:  $\log \log n$ ,  $\log n$ ,  $(\log_n)^2$ ;
- 2. polynomial functions:  $\sqrt{n} = n^{0.5}$ , n,  $n \log n$ ,  $n^{1.001}$ ;
- 3. exponential functions:  $2^n$ ,  $n2^n$ ,  $3^n$ ;
- 4. factorial functions: *n*!;

In above lists, any logarithmic function is small-o of any polynomial function: for example,  $(\log n)^2 = o(n^{0.01})$ ; any polynomial function is small-o of any exponential function: for example,  $n^2 = o(2^n)$ ; any exponential function is small-o of any factorial function: for example,  $n^2 = o(n!)$ . Within each category, a function to the left is small-o of a function to the right, for example  $n \log n = o(n^{1.001})$ .

## **Merge-Sort**

We now start introducing the first algorithm-design technique: divide-and-conquer. A typical divide-and-conquer algorithm follows the framework below.

- 1. partition the original problem into smaller problems;
- 2. recursively solve all subproblems;
- 3. combine the solutions of the subproblems to obtain the solution of the original problem.

We use sorting as the first problem to demonstrate designing divide-and-conquer algorithms. Recall that the *sorting* problem is to find the sorted array (say, in increasing order) S' of a given array S. We now design a divide-and-conquer algorithm for it. For any recursive algorithm, we always need to clearly define the recursion. In this case, we define function merge-sort (S) returns the sorted array (in ascending order) of S.

The idea is to sort the first half and second half of *S*, by recursively call the merge-sort function. How to obtain the sorted array of *S* then with the two sorted half-sized arrays? We have introduced such an algorithm to merge two sorted arrays into a single sorted array. That's exactly the algorithm we need here in the combining step.

```
Algorithm merge-sort (S[1 \cdots n])

if n \le 1: return S;

S'_1 = \text{merge-sort} (S[1 \cdots n/2]);

S'_2 = \text{merge-sort} (S[n/2 + 1 \cdots n]);

return merge-two-sorted-arrays (S'_1, S'_2);

end algorithm;
```