Time Complexity Analysis

Asymptotic notation

- In mathematical analysis, asymptotic analysis, also known as asymptotics, is a method of describing limiting behavior.
- As an illustration, suppose that we are interested in the limiting behavior or growth properties of a function f(n) as n becomes very large.
- If $T(n)=f(n)=n^2+3n$, then as n becomes very large, the term 3n becomes insignificant compared to n^2 .

n	n^2	3 <i>n</i>
2	4	6
50	2500	150
150	22,500	450
500	2,50,000	1500
1000	10,00,000	3000

- The function f(n) is said to be "asymptotically equivalent to n^2 , as $n \to \infty$ ".
- This is often written symbolically as $f(n) \sim n^2$, which is read as "f(n) is asymptotic to n^2 ".
- The idea is we compare the growth rate of the given function f(n) with the known behaviors of standard functions represented by g(n).

- $T(n) = f(n) = n^3 + 3n^2 + 2n + 1$
- What is f(n) asymptotically equivalent to as $n \rightarrow \infty$?
- What is g(n)?

- The family of g(n) includes 1, n, n^2 , n^k , $\log n$, $n\log n$, 2^n , n!
- The same idea with a specfic notation is adopted in the Algorithm Analysis which is called as asymptotic notation/analysis.
- In the case of Algorithm Analysis, the function T(n)- the step count takes the place of f(n), and we estimate the growth rate of T(n) in terms of g(n).

$$g(n) = 1$$
 Constant function

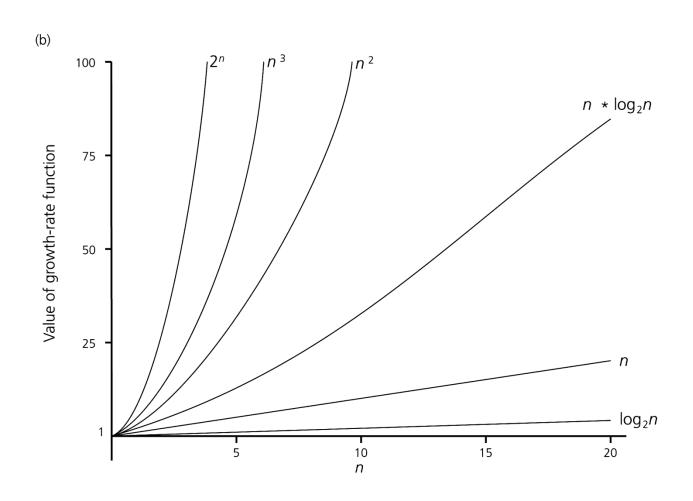
$$g(n) = \log n$$
 Logarithmic function

$$g(n) = n$$
 Linear function

$$g(n) = n^2$$
 Quadratic function

$$g(n) = n^3$$
 Cubic function

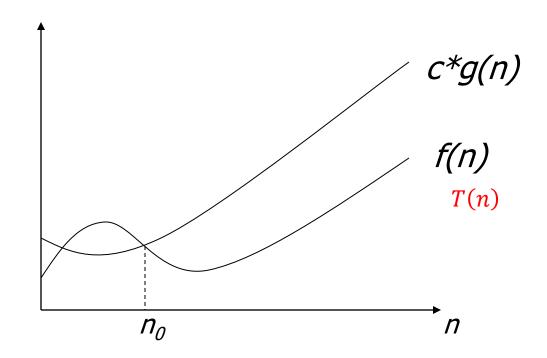
$$g(n) = 2^n$$
 Exponential function



- The analysis is more meaningful because we wound be interested in the behavior of the algorithm as the input size n increases.
- The asymptotic analysis defines identifies three kinds of g(n)-

- g(n) whose growth rate is greater or equal to the growth rate of T(n).
- g(n) whose growth rate is same as T(n)
- g(n) whose growth rate is smaller or equal to the growth rate of T(n).

• T(n) is said to be O(g(n)) iff there exists constants c and n_0 such that $T(n) \le c$. g(n) for $n \ge n_0$



- T(n) will have growth rate which is smaller than g(n). In other words g(n) is the upper bound on T(n). g(n) should be as small as possible.
- T(n) = 3n + 2 = O(n)
- $3n + 2 \le 4n$ for $n \ge 2$, c = 4 and $n_0 = 2$

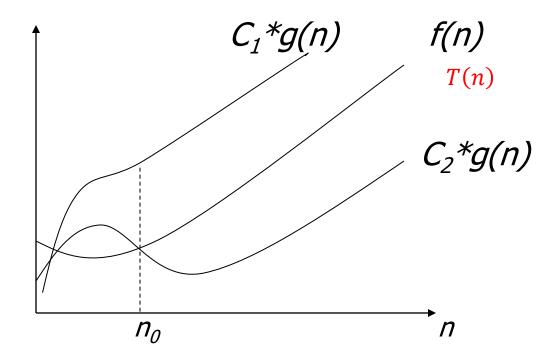
- $T(n) = 10n^2 + 4n + 2$
- Find c, n_0 , g(n)
- $10n^2 + 4n + 2 \le 11 n^2$ for $n \ge 5$
- $T(n) = O(n^2)$
- $T(n) = n^2 + 6 * 2^n = O(2^n)$
- Find c, n_0 , g(n)
- $n^2 + 6 * 2^n \le 7 * 2^n$ for $n \ge 4$
- $T(n) = O(2^n)$

• T(n) is said to be $\Omega(g(n))$ iff there exists constants c and n_0 such that $T(n) \ge c$. g(n) for $n \ge n_0$. g(n) should be as large as possible.

f(n) T(n) c*g(n) n

- T(n) will have growth rate which is larger than g(n). In other words g(n) is the lower bound on T(n).
- $T(n) = 3n + 2 = \Omega(n)$
- $3n + 2 \ge 3n$ for $n \ge 1$, c = 3 and $n_0 = 1$
- $T(n) = 10n^2 + 4n + 2 = \Omega(n^2)$
- $10n^2 + 4n + 2 \ge n^2$ for $n \ge 1$

• T(n) is said to be $\theta(g(n))$ iff there exists constants c_1 , c_2 and n_0 such that $T(n) \le c_1 g(n)$ and $T(n) \ge c_2 g(n)$ for $n \ge n_0$



- T(n) will have growth rate which is same as g(n).
- If T(n) is a polynomial of degree n, then
- $T(n) = 3n^2 + 2n + 3 = \theta(n^2)$?
- $T(n) = 3n^2 + 2n + 3 \le 4n^2$ for $n \ge 3$
- $T(n) = 3n^2 + 2n + 3 \ge n^2$ for $n \ge 1$

- T(n) is said to be o(g(n)) iff for any positive constant c, T(n) < c. g(n) for $n \ge n_0$
- Suppose T(n) = 2n + 3
- $T(n) = 2n + 3 < 6n \text{ for } n \ge 1$
- but
- $T(n) = 2n + 3 < cn \text{ for } n \ge 1$ for c = 1,2,3,4,5

- $T(n) = 2n + 3 < cn^2 \text{ for } n \ge 1$
- Hence
- $T(n) = o(n^2)$