

Jul 10

140 Dis

Asymptotic Order of Growth

Upper bounds. $T(n)$ is $O(f(n))$ if there exist constants $c > 0$ and $n_0 \geq 0$ such that for all $n \geq n_0$ we have $T(n) \leq c \cdot f(n)$.

Lower bounds. $T(n)$ is $\Omega(f(n))$ if there exist constants $c > 0$ and $n_0 \geq 0$ such that for all $n \geq n_0$ we have $T(n) \geq c \cdot f(n)$.

Tight bounds. $T(n)$ is $\Theta(f(n))$ if $T(n)$ is both $O(f(n))$ and $\Omega(f(n))$.

Ex: $T(n) = 32n^2 + 17n + 32$.

- $T(n)$ is $O(n^2)$, $O(n^3)$, $\Omega(n^2)$, $\Omega(n)$, and $\Theta(n^2)$.
- $T(n)$ is not $O(n)$, $\Omega(n^3)$, $\Theta(n)$, or $\Theta(n^3)$.

Properties

Transitivity.

- If $f = O(g)$ and $g = O(h)$ then $f = O(h)$.
- If $f = \Omega(g)$ and $g = \Omega(h)$ then $f = \Omega(h)$.
- If $f = \Theta(g)$ and $g = \Theta(h)$ then $f = \Theta(h)$.

Additivity.

- If $f = O(h)$ and $g = O(h)$ then $f + g = O(h)$.
- If $f = \Omega(h)$ and $g = \Omega(h)$ then $f + g = \Omega(h)$.
- If $f = \Theta(h)$ and $g = O(h)$ then $f + g = \Theta(h)$.

Special functions

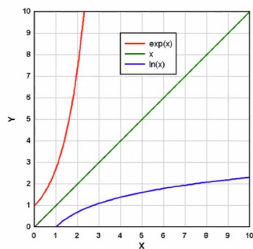
Polynomial $O(n^a)$, a independent of n

- $O(n)$ linear
- $O(n^2)$ quadratic
- $O(n^3)$ cubic
- $O(\log n)$ logarithms
 - $O(\log n) = O(n^\epsilon)$

logarithms grow more slowly than polynomial

Nonpolynomial

- $O(n!)$, $O(3^n)$



Practice

$2n$	$O(n)$
$2n + 3$	$O(n)$
$2n + 10,000,000,000$	$O(n)$
$2n - 10,000,000,000$	$O(n)$
$3n^2 + 2n + 3$	$O(n^2)$
$3n^2 + 10000000000n + 3$	$O(n^2)$
$2\log(n)$	$O(\log(n))$
$2\log_{100}(n)$	$O(\log(n))$
$2n + \log(n)$	$O(n)$
$3^n + n^{100}$	$O(3^n)$

$$\log_a b = \frac{\log_c a}{\log_c b}$$

PSEUDO Code Practice: Simple

Assumption about one operation:

- compare two numbers
- math operations(+, /, log)
- assign value to an array element

```
(1)
for (int i=1; i<=n; i++)
{
    sum = sum + i;
}
```

$O(n)$

```
■ (2)
for (int i=1; i<=n; i++)
{
    sum = sum + i;
    System.out.println("Hello");
}
```

$O(n)$

PSEUDO Code Practice: Simple

```
(3)
for (int i=1; i<=n; i=i+2)
{
    sum = sum + i;
}
```

```
■ (4)
for (int i=1; i<=n; i=i*2)
{
    sum = sum + i;
}
```

n=5, i=1,3,5
n=10, i=1,3,5,7,9
loop times: n/2
 $O(n)$

n=5, i=1,2,4
n=20, i=1,2,4,8,16
 $O(\log(n))$

PSEUDO Code Practice: Simple

```

(5)
for (int i=1; i<=n; i=i*3)
{
    sum = sum + i;
}

n=5, i=1,3
n=20, i=1,3,9
loop times:  $\log_3(n) - 1$ 
O(log(n))
    
```

```

(6)
for (int i=1; i<=n; i++)
{
    for(int j=1; j<=n; j++)
    {
        A[i,j]=i*j;
    }
}
    
```

$O(n^2)$

Practice

Practice

2. Suppose you have algorithms with the six running times listed below. (Assume these are the exact number of operations performed as a function of the input size n .) Suppose you have a computer that can perform 10^{10} operations per second, and you need to compute a result in at most an hour of computation. For each of the algorithms, what is the largest input size n for which you would be able to get the result within an hour?

- (a) n^2
 (b) n^3
 (c) $100n^2$
 (d) $n \log n$ ← fastest
 (e) 2^n
 (f) 2^{2^n} ← slowest

$3600 \cdot 10^{10}$

One hour: 3.6×10^{13} operations

n^2	6,000,000
n^3	33019
$100n^2$	600,000
$n \log n$	1.29×10^{12} (different results for different base)
2^n	45
2^{2^n}	5