

1D1020 2018-09-07 #3

Analysing algorithms to understand their performance

Observations, Mathematical models, Running time classes
Complexity theory, memory complexity.

Memory complexity is important.

Strive to solve problems faster with less memory. Environment!

Analysis is needed to understand what is possible

DoS can exploit worst cases

Different views: Functioning, Efficient, Understandable

Estimate running time, compare algorithms

Discrete Fourier transform (DFT, JPEG, MRI)

Brute force: N^2 FFT algorithm: $N \log N$

Big-O Complexity $O(1) \rightarrow O(n!)$

Observe, model, predict, check

Experiments should be reproducible. Falsifiability

Brute force "tests all cases"

Doubling ratio-hypothesis. Run the program and double problem size.
System dependent vs System independent

Mathematical model $\text{running time} = \sum_{\text{operations}} \text{cost}_o \cdot \text{frequency}_o$

Cost is system dependent, frequency is system independent

Cost model. Analyse one primitive operation

Omit lower-order terms for big N

Tilde approximation, Approximate sum with integral

Approximative models are often sufficient.

order-of-growth classes

$O(1)$ constant	Add two numbers
$O(\log N)$ logarithmic	Binary search
$O(N)$ linear	Find max/min
$O(N \log N)$ linearithmic	Mergesort
$O(N^2)$ quadratic	Check all pairs
$O(N^3)$ cubic	Check all triplets
$O(2^N)$ Exponential	Check all subsets

Binary search $\in O(\log N)$

Rule of thumb. Better (lower) time complexity \Rightarrow Faster algorithms

Best case: Lower bound Worst case: Upper bound

Average case: Expected for "random" input.

Amortized complexity

Some operations/cases with high cost

Complexity theory

- How hard?
- Optimal algorithms

Big Theta (Θ) asymptotic complexity

Big Oh (O) Upper bound

Big Omega (Ω) Lower bound

$f(n) = O(g(n))$ if $\exists c > 0$ and $n_0 > 0$ where $f(n) \leq cg(n) \forall n \geq n_0$

$\Omega(g(n)) = \{f(n) \mid \exists c > 0 \text{ and } n_0 > 0 \text{ where } 0 \leq cg(n) \leq f(n) \forall n \geq n_0\}$

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

Class P (polynomial time)

$P = NP?$

Class NP (Non-deterministic Polynomial time)

A problem is NP-complete when it is both NP-hard.

- An NP-hard problem is at least as computationally hard as the hardest problem in NP.

- Design an algorithm
- Prove a lower bound

Is the (theoretical) lower bound lower than the upper bound

- (of the algorithm)?
- Lower the upper bound (design better algorithm)
 - Raise the lower bound (harder)

Golden era of algorithm design: 1970s

In 1D/2D we focus on Tilde notation

Memory complexity

64-bit computer (or more precise OS) \rightarrow 8-byte pointer/address

Object overhead 16 bytes

Reference 8 bytes

Array overhead 24 bytes

Padding

Shallow memory usage

Deep memory usage

Memory profiling

Empiric analysis

Mathematical analysis

Scientific method

boolean:	1 B
byte:	1 B
char:	2 B
int:	4 B
float:	4 B
long:	8 B
double:	8 B