

AI Lab-Class

Algorithm Analysis

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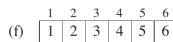
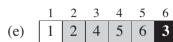
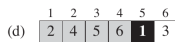
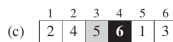
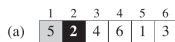
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Asymptotic Algorithm Analysis

“Duuude, yesterday I implemented an algorithm that sorts an array in only 10 seconds!!!” Nice, **but:**

- Performance and runtime of an algorithm depend on the **input** and available **hardware**.
- Not only the input size matters, but also the properties of the input, e.g. if it is pre-sorted or somehow ordered in advance.

Motivation: Insertion Sort's running time



INSERTION-SORT(A)

```
1  for  $j = 2$  to  $A.length$ 
2       $key = A[j]$ 
3      // Insert  $A[j]$  into the sorted sequence  $A[1..j-1]$ .
4       $i = j - 1$ 
5      while  $i > 0$  and  $A[i] > key$ 
6           $A[i + 1] = A[i]$ 
7           $i = i - 1$ 
8       $A[i + 1] = key$ 
```

Motivation: Insertion Sort's running time

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

- t_j : number of times the while-condition is tested in iteration j
- Best case (array pre-sorted): $t_j = 1$
- Worst case (array sorted in descending order): $t_j = j - 1$

Motivation: Insertion Sort's running time

General case:

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

Best case ($t_j = 1$):

$$\begin{aligned} T(n) = & c_1 n + c_2(n-1) + c_4(n-1) + c_5(n-1) + c_8(n-1) \\ = & (c_1 + c_2 + c_4 + c_5 + c_8)n - (c_2 + c_4 + c_5 + c_8) . \end{aligned}$$

Worst case ($t_j = j - 1$):

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

Concentrate on the worst case!

- Worst-case running time gives an upper bound for any input. A guarantee, that the algorithm will never take longer!
- The worst case appears fairly often (database lookups).
- The average case is often roughly as bad as the worst case.

Motivation: Asymptotic algorithm analysis

- Precise analysis of an algorithm in a computational model is tedious. So far, we haven't even touched recursive algorithms!
- Needed: A metric that is simple to write and manipulate, shows the most important resource requirements while suppressing the tedious details.
- To the rescue: Asymptotic notation!

O-Notation

- Let $g : \mathbb{N} \rightarrow \mathbb{R}_+$ be an arbitrary function.
- The set of functions $f : \mathbb{N} \rightarrow \mathbb{R}_+$, which do not grow faster than the function g after a specific point n_0 , is denoted as $O(g(n))$.
- More specifically:
$$O(g(n)) := \{f(n) \mid \exists c \in \mathbb{R} \text{ and } \exists n_0 \in \mathbb{N} : 0 \leq f(n) \leq c \cdot g(n) \forall n \geq n_0\}$$

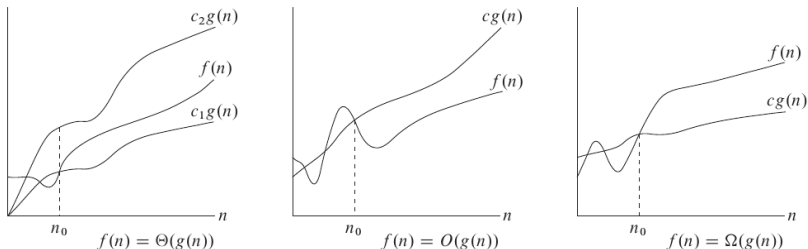


Figure 1: O , Θ and Ω notation