

# Analysis, Big O and Growth of Functions

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## 1 Book Keeping

- Reading posted
- Lab 1 available

## 2 Analysis of Algorithms

**Problem:** a general description of input parameters and the properties that an optimal solution should have

**Instance:** a specific example of a problem with all parameters specified

- Example: Given a weighted graph, find the cheapest Hamiltonian Cycle (TSP)
- A "problem" can have many instances

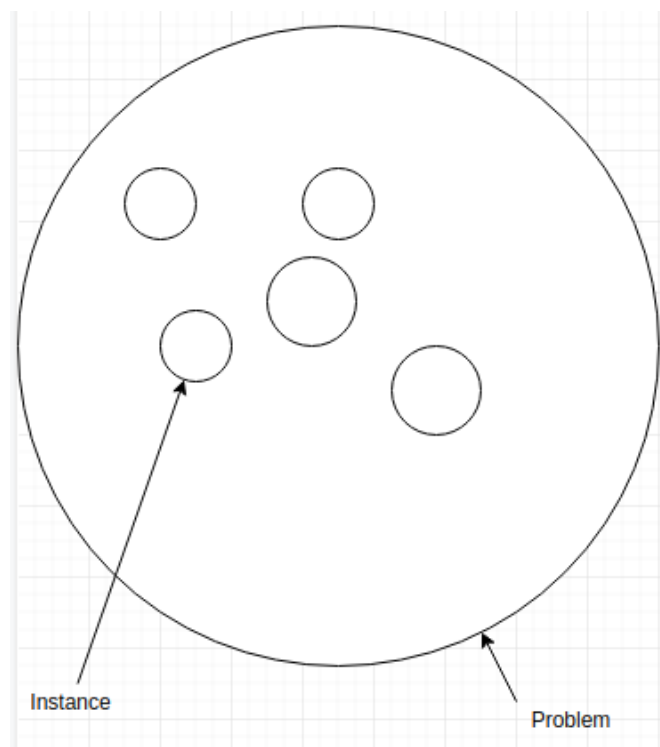


Figure 1: instance\_problem

- An algorithm solves all instances of problem

- Many algorithms, what is most efficient?
- What is efficient?
  - Memory
  - Time
  - CPU cycles
  - Disk Space
  - I/O bandwidth
  - Power
- Efficiency usually defined as using smallest time
- Index runtimes by instance size
- "Instance Size" not always well defined - can have multiple params (edges, nodes)

### 3 Example: Insertion Sort

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

Figure 2: Cor p.24

- Best case: already sorted.  $T(j) = 1, T(n) = an + b \rightarrow$  linear
- Worst case: reverse sorted:  $T(j) = j, T(n) = \frac{n(n+1)}{2} \approx an^2 + bn + c \rightarrow$  quadratic

**Time Complexity Function:** The largest amount of time for an algorithm needed to solve the problem for a given instance size.

- Even Time-Complexity function considered too complicated for daily use
- Asymptotic notation used instead

## 4 Asymptotic Notation

For a given function  $g(n)$ ,  $O(g(n)) = f(n)$  there exist positive constants  $k$  and  $n_0$  such that  $f(n) \leq K g(n)$  for all  $n \geq n_0$

Less formally:  $O(g(n))$  is the set of functions that are asymptotically less than  $g(n)$  for large  $n$ .

### Example

I claim that  $f(n) = an^2 + bn + c = O(n^2)$ . If so, then there should exist positive constants  $k$  and  $n_0$  such that

$$an^2 + bn + c \leq kn^2$$

$$a + b/n + \frac{c}{n^2} \leq k$$

$$k = a + 1$$

$n_0$  is intersection

### Summary

- For insertion sort, worst case runtime (time complexity function) is  $an^2 + bn + c$  so the complexity is  $O(n^2)$
- Also  $O(n^3), O(n^4)$  etc.
- Worst case runtime is  $O(n^2)$
- Worst case runtime **itself** is upper bound on run time
- $O(n^2)$  is then an upper bound on the general runtime as well!

Polynomial-time Algorithm: an algorithm whose time complexity function is  $O(p(n))$  for some polynomial  $p(n)$

Exponential-time Algorithm: an algorithm that is not polynomial time

### EXPONENTIAL VERY BAD