

# Basics of Algorithm Analysis

- We measure running time as a function of  $n$ , the size of the input (in bytes assuming a reasonable encoding).
- We work in the RAM model of computation. All “reasonable” operations take “1” unit of time. (e.g.  $+$ ,  $*$ ,  $-$ ,  $/$ , array access, pointer following, writing a value, one byte of I/O...)

## What is the running time of an algorithm

- Best case (seldom used)
- Average case (used if we understand the average)
- Worst case (used most often)

## We measure as a function of $n$ , and ignore low order terms.

- $5n^3 + n - 6$  becomes  $n^3$
- $8n \log n - 60n$  becomes  $n \log n$
- $2^n + 3n^4$  becomes  $2^n$

# Asymptotic notation

## big-O

$O(g(n)) = \{f(n) : \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq f(n) \leq cg(n) \text{ for all } n \geq n_0\} .$

Alternatively, we say

$f(n) = O(g(n))$  if there exist positive constants  $c$  and  $n_0$  such that  $0 \leq f(n) \leq cg(n)$  for all  $n \geq n_0$

Informally,  $f(n) = O(g(n))$  means that  $f(n)$  is asymptotically less than or equal to  $g(n)$ .

## big-Ω

$\Omega(g(n)) = \{f(n) : \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq cg(n) \leq f(n) \text{ for all } n \geq n_0\} .$

Alternatively, we say

$f(n) = \Omega(g(n))$  if there exist positive constants  $c$  and  $n_0$  such that  $0 \leq cg(n) \leq f(n)$  for all  $n \geq n_0$  .

Informally,  $f(n) = \Omega(g(n))$  means that  $f(n)$  is asymptotically greater than or equal to  $g(n)$ .

## big- $\Theta$

$f(n) = \Theta(g(n))$  if and only if  $f(n) = O(g(n))$  and  $f(n) = \Omega(g(n))$ .

Informally,  $f(n) = \Theta(g(n))$  means that  $f(n)$  is asymptotically equal to  $g(n)$ .

### INFORMAL summary

- $f(n) = O(g(n))$  roughly means  $f(n) \leq g(n)$
- $f(n) = \Omega(g(n))$  roughly means  $f(n) \geq g(n)$
- $f(n) = \Theta(g(n))$  roughly means  $f(n) = g(n)$
- $f(n) = o(g(n))$  roughly means  $f(n) < g(n)$
- $f(n) = w(g(n))$  roughly means  $f(n) > g(n)$

We use these to **classify** algorithms into classes, e.g.  $n$ ,  $n^2$ ,  $n \log n$ ,  $2^n$ .

See chart for justification

# Polynomial Time

The size of a problem instance typically is described by parameters such as:

- number of nodes  $n$  or  $V$
- number of edges  $m$  or  $E$
- largest capacity  $U$
- largest cost (in absolute value)  $C$

**Input size:** The **size** of the input, which consists of a list of nodes and edges and their capacities and costs is typically

$$\Theta(n + m + m \log U + m \log C)$$

.

- A **polynomial algorithm** is one whose running time is polynomial in the input, i.e. is polynomial in  $n$ ,  $m$ ,  $\log U$ ,  $\log C$ .
- A **strongly polynomial algorithm** is one whose running time is polynomial in the size of the graph and independent of the size of the numbers, i.e. is polynomial in  $n$ ,  $m$ .
- A **pseudo-polynomial algorithm** is one whose running time is polynomial in the size of the graph and the magnitude of the numbers, i.e. is polynomial in  $n$ ,  $m$ ,  $U$ ,  $C$ .

## Commentary (with trivial interpretations excluded)

- Strongly polynomial and polynomial algorithms are polynomial algorithms. Pseudo-polynomial algorithms are not polynomial algorithms.
- Strongly polynomial algorithms are mainly a theoretical concept and do not tend to get used in practice.

We will typically shoot for polynomial algorithms.

## Some Graph terminology

- node, vertex
- edge, arc
- directed undirected
- head tail
- path
- cycle
- acyclic
- bipartite graph
- tree
- forest
- cut
- $s$ - $t$  cut
- connectivity
- strong connectivity
- bipartite graph

# Easily Solved Graph Problems

- Connectivity
- Strong Connectivity
- Spanning trees
- Bipartiteness
- Topological Search
- Depth-first Search
- Breadth-first Search

# Other Basics

## Basic Data Structures

- Arrays
- Linked Lists
- Stack - LIFO
- Queue - FIFO
- Binary tree
- Hash table



# Dictionary Operations on ordered set

- Insert
- Delete
- Find
- Min, Max
- Successor, Predecessor
- IncreaseKey, DecreaseKey

## Comments

- Some form of a **balanced binary tree** supports all dictionary operations in  $O(\log n)$  time
- A hash table supports Insert, Delete and Find in  $O(1)$  expected time

# Graph Storage

- An **adjacency matrix** is an  $n$  by  $n$  matrix in which  $A[i, j]$  stores values related to edge  $(i, j)$ .
- An **adjacency list** is a length  $n$  array  $L$  of linked lists, where entry  $L[i]$  is a list of all edges adjacent to vertex  $i$ .