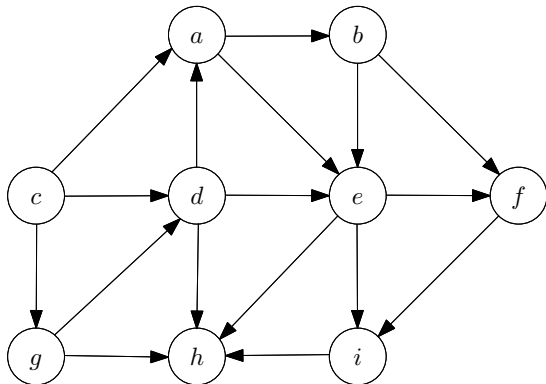


Topological Ordering Problem

Input: a directed acyclic graph (DAG) $G = (V, E)$

Output: 1-to-1 function $\pi : V \rightarrow \{1, 2, 3 \dots, n\}$, so that

- if $(u, v) \in E$ then $\pi(u) < \pi(v)$

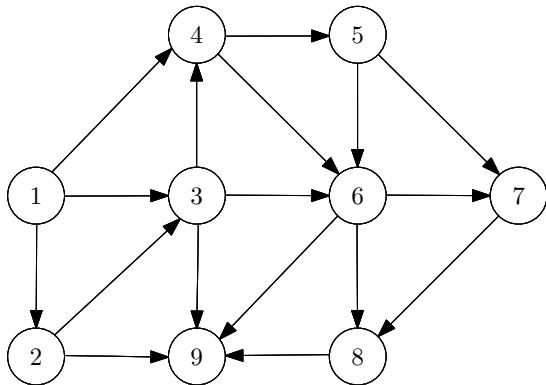


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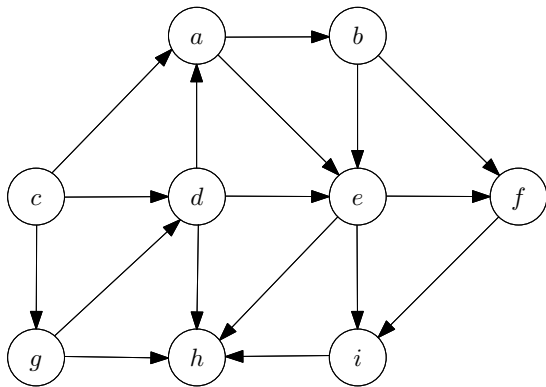
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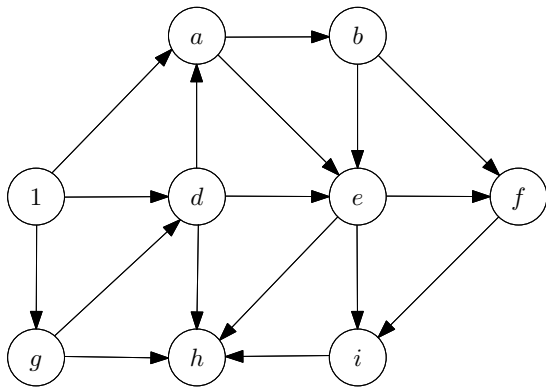
Topological Ordering

- Algorithm: each time take a vertex without incoming edges, then remove the vertex and all its outgoing edges.



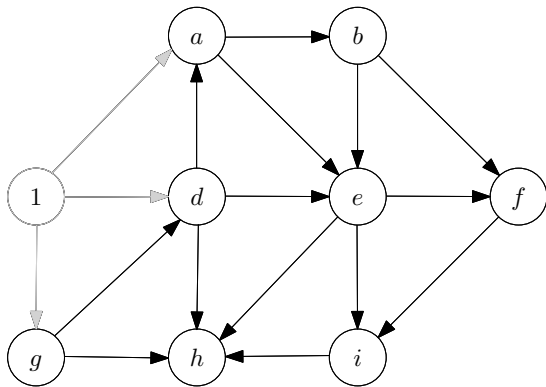
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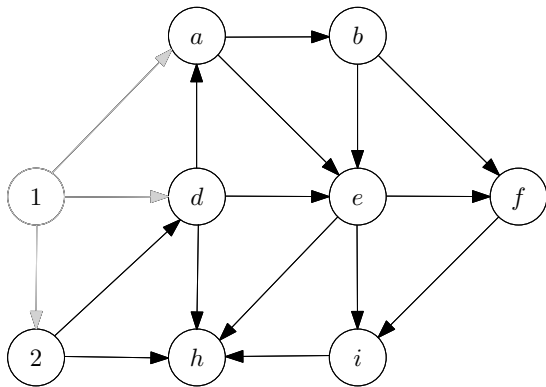
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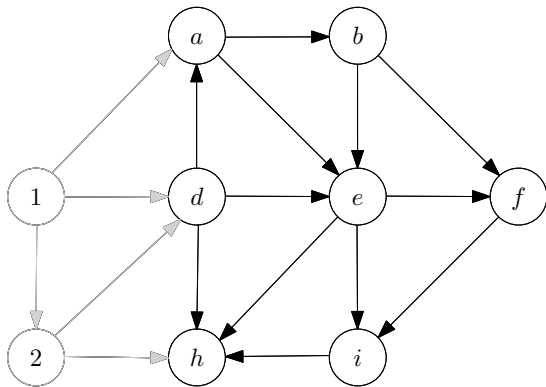
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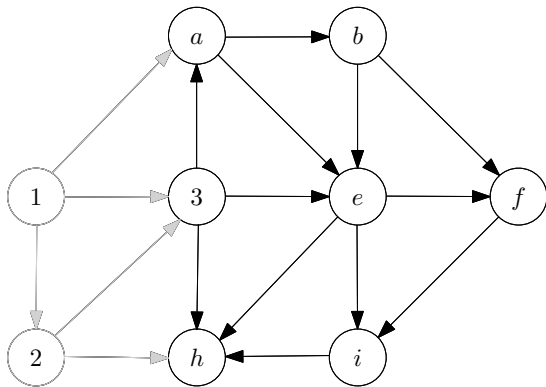
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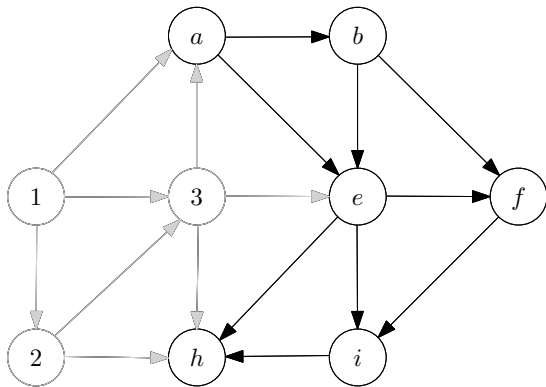
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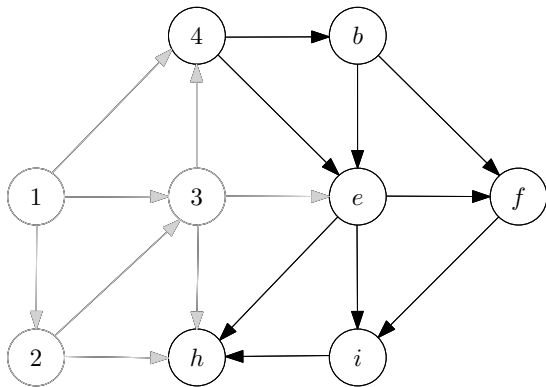
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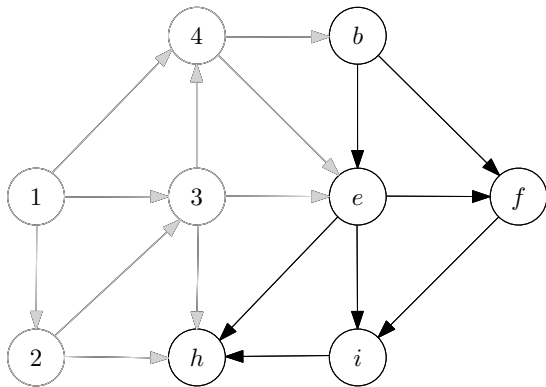
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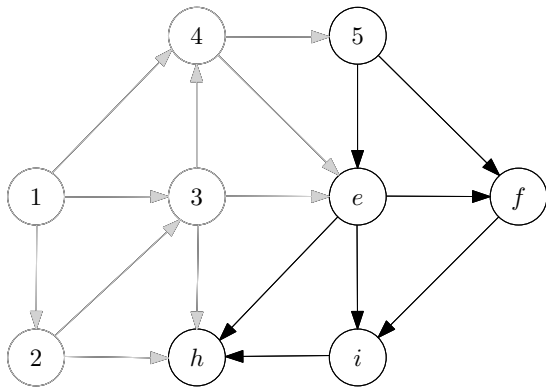
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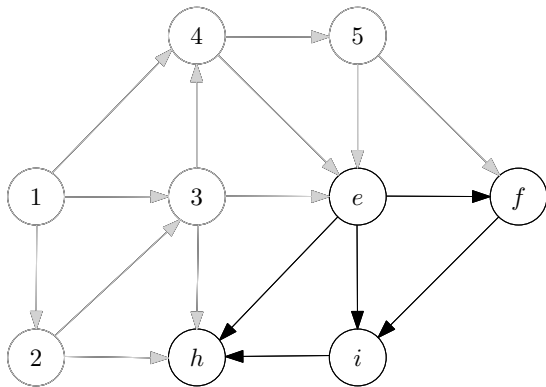
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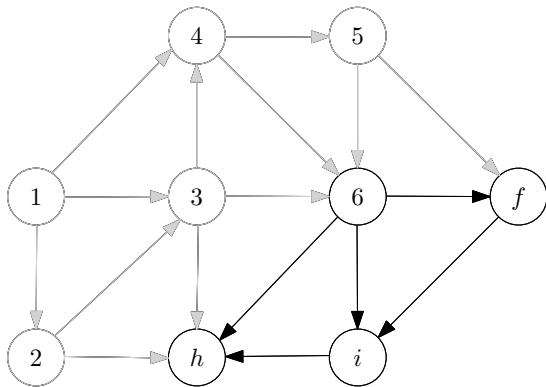
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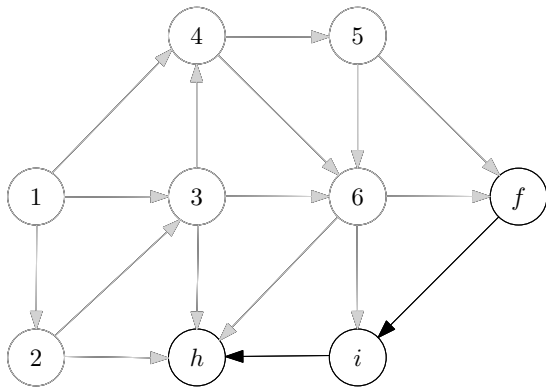
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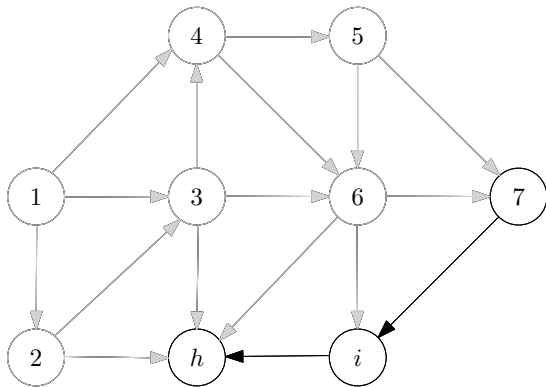
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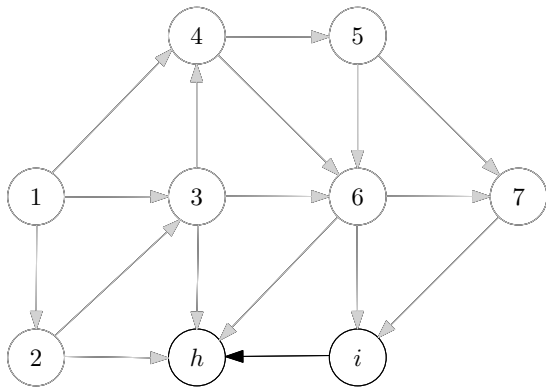
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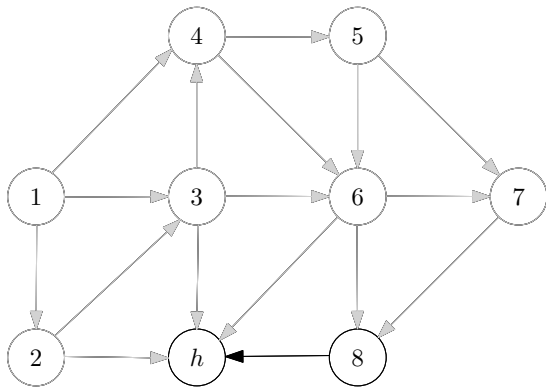
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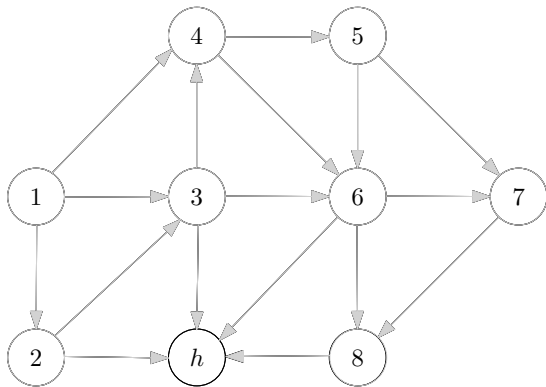
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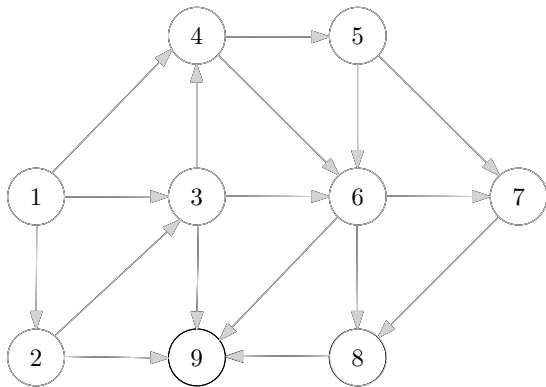
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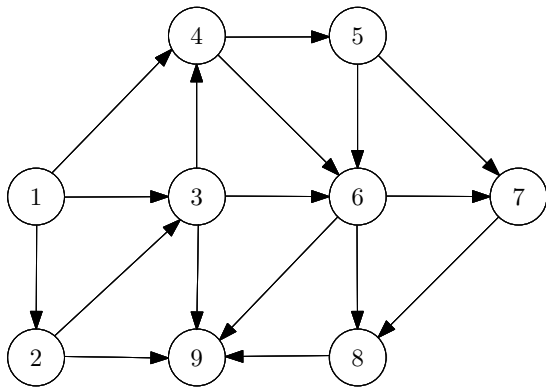
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Q: How to make the algorithm as efficient as possible?

Topological Ordering

- Algorithm: each time take a vertex without incoming edges, then remove the vertex and all its outgoing edges.

Q: How to make the algorithm as efficient as possible?

A:

- Use linked-lists of outgoing edges
- Maintain the in-degree d_v of vertices
- Maintain a queue (or stack) of vertices v with $d_v = 0$

topological-sort(G)

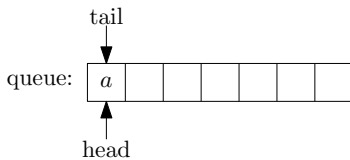
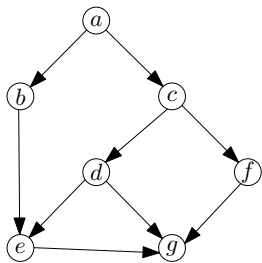
```
1: let  $d_v \leftarrow 0$  for every  $v \in V$ 
2: for every  $v \in V$  do
3:   for every  $u$  such that  $(v, u) \in E$  do
4:      $d_u \leftarrow d_u + 1$ 
5:  $S \leftarrow \{v : d_v = 0\}, i \leftarrow 0$ 
6: while  $S \neq \emptyset$  do
7:    $v \leftarrow$  arbitrary vertex in  $S, S \leftarrow S \setminus \{v\}$ 
8:    $i \leftarrow i + 1, \pi(v) \leftarrow i$ 
9:   for every  $u$  such that  $(v, u) \in E$  do
10:     $d_u \leftarrow d_u - 1$ 
11:    if  $d_u = 0$  then add  $u$  to  $S$ 
12: if  $i < n$  then output "not a DAG"
```

- S can be represented using a queue or a stack
- Running time = $O(n + m)$

S as a Queue or a Stack

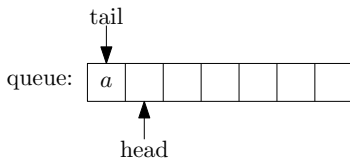
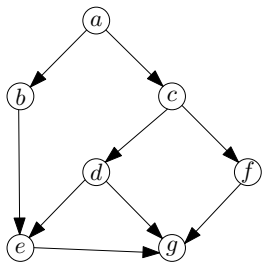
DS	Queue	Stack
Initialization	$head \leftarrow 1, tail \leftarrow 0$	$top \leftarrow 0$
Non-Empty?	$head \leq tail$	$top > 0$
Add(v)	$tail \leftarrow tail + 1$ $S[tail] \leftarrow v$	$top \leftarrow top + 1$ $S[top] \leftarrow v$
Retrieve v	$v \leftarrow S[head]$ $head \leftarrow head + 1$	$v \leftarrow S[top]$ $top \leftarrow top - 1$

Example



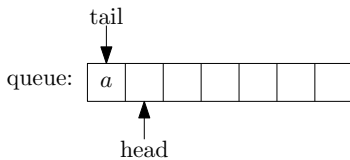
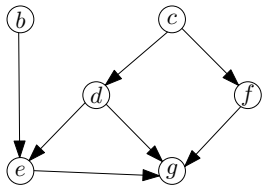
	a	b	c	d	e	f	g
degree	0	1	1	1	2	1	3

Example



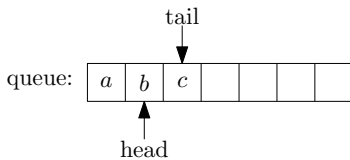
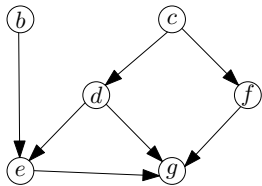
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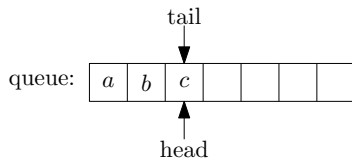
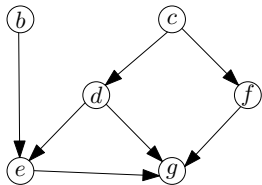
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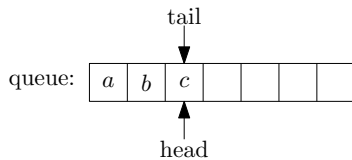
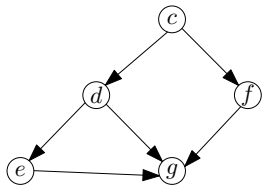
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Example



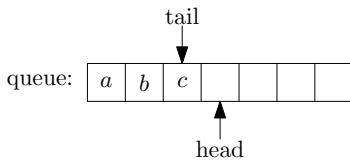
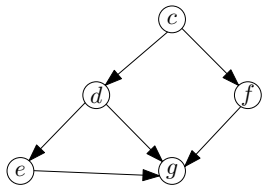
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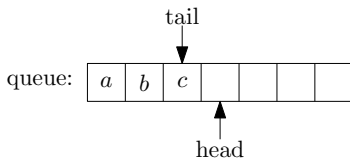
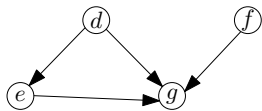
	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
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Example



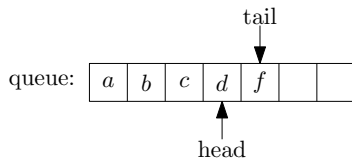
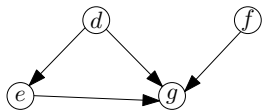
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Example



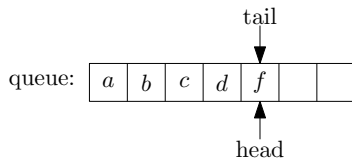
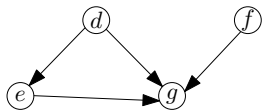
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Example



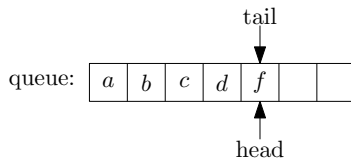
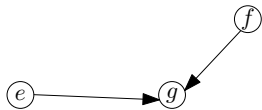
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Example



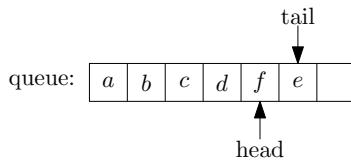
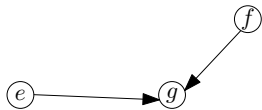
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Example



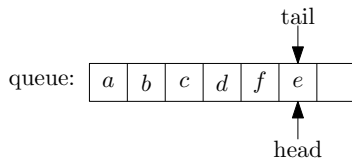
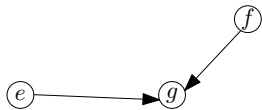
	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
degree	0	0	0	0	0	0	2

Example



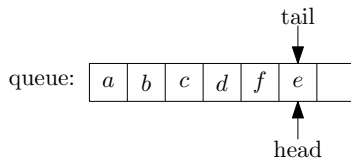
	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
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Example



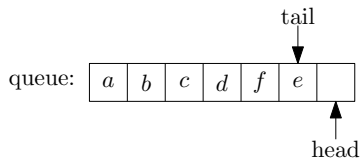
	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
degree	0	0	0	0	0	0	2

Example



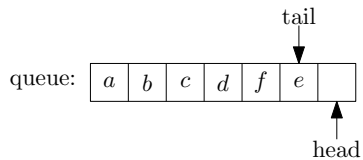
	a	b	c	d	e	f	g
degree	0	0	0	0	0	0	1

Example



	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
degree	0	0	0	0	0	0	1

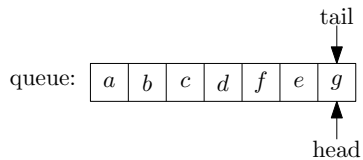
Example



⑨

	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
degree	0	0	0	0	0	0	0

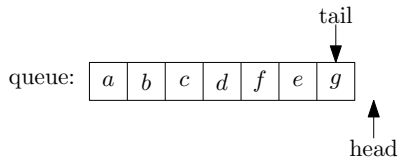
Example



⑨

	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
degree	0	0	0	0	0	0	0

Example



⑨

	<i>a</i>	<i>b</i>	<i>c</i>	<i>d</i>	<i>e</i>	<i>f</i>	<i>g</i>
degree	0	0	0	0	0	0	0

Outline

- 1 Graphs
- 2 Connectivity and Graph Traversal
 - Types of Graphs
- 3 Bipartite Graphs
 - Testing Bipartiteness
- 4 Topological Ordering
 - Applications: Word Ladder

Def. Word: A string formed by letters.

Def. Adjacency words: Word A and B are adjacent if they differ in exactly one letter.

e.g. *word* and *work*; *tell* and *tall*; *askbe* and *askee*.

Def. Word Ladder: Players start with one word and, in a series of steps, change or transform that word into another word.

Def. Word Ladder: Players start with one word and, in a series of steps, change or transform that word into another word.

- The objective is to make the change in the smallest number of steps, with each step involving changing a **single letter** of the word to create a new valid word.

Word Ladder Problem

Input: Two words S and T , a list of words $A = \{W_1, W_2, \dots, W_k\}$.

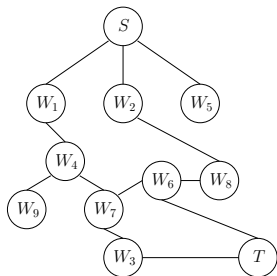
Output: “The smallest word ladder” if we can change S to T by moving between adjacency words in $A \cup \{S, T\}$;
Otherwise, “No word ladder”.

Example:

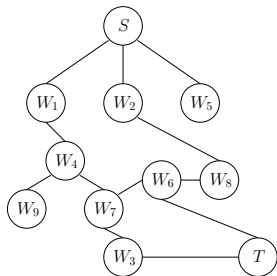
- $S = \text{“a e f g h”}$, $T = \text{“d l m i h”}$
- $W_1 = \text{“a e f i h”}$, $W_2 = \text{“a e m g h”}$, $W_3 = \text{“d l f i h”}$
 $W_4 = \text{“s e f i h”}$, $W_5 = \text{“a d f g h”}$, $W_6 = \text{“d e m i h”}$
 $W_7 = \text{“d e f i h”}$, $W_8 = \text{“d e m g h”}$, $W_9 = \text{“s e m i h”}$

Example:

- $S = \text{"a e f g h"} , T = \text{"d l m i h"}$
- $W_1 = \text{"a e f i h"} , W_2 = \text{"a e m g h"} , W_3 = \text{"d l f i h"}$
 $W_4 = \text{"s e f i h"} , W_5 = \text{"a d f g h"} , W_6 = \text{"d e m i h"}$
 $W_7 = \text{"d e f i h"} , W_8 = \text{"d e m g h"} , W_9 = \text{"s e m i h"}$



- Each vertex corresponds to a word.
- Two vertices are adjacent if the corresponding words are adjacent.



- Each vertex corresponds to a word.
- Two vertices are adjacent if the corresponding words are adjacent.
- Hints: Given vertex v , check its nearest neighbor.

CSE 431/531B: Algorithm Analysis and Design (Fall 2023)

Greedy Algorithms

Lecturer: Kelin Luo

*Department of Computer Science and Engineering
University at Buffalo*

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Trivial Algorithm for an Optimization Problem

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- However, trivial algorithm often runs in **exponential** time, as the number of potential solutions is often exponentially large.

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Trivial Algorithm for an Optimization Problem

Enumerate all valid solutions, compare them and output the best one.

- However, trivial algorithm often runs in **exponential** time, as the number of potential solutions is often exponentially large.
- $f(n)$ is a **polynomial** if $f(n) = O(n^k)$ for some **constant** $k > 0$.

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Goals of algorithm design

- 1 Design efficient algorithms to solve problems
- 2 Design more efficient algorithms to solve problems

Common Paradigms for Algorithm Design

- Greedy Algorithms: shortest path problem
- Divide and Conquer: merge-sort, binary search
- Dynamic Programming: Fibonacci number

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- Hard to see correctness. Mostly, it is not correct. E.g. $\min f(x)$

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- Build up the solutions in steps
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Def. A strategy is safe: there is always an optimum solution that agrees with the decision made according to the strategy.

Outline

- 1 Toy Example: Box Packing
- 2 Interval Scheduling
- 3 Offline Caching
 - Heap: Concrete Data Structure for Priority Queue
- 4 Data Compression and Huffman Code
- 5 Summary

Box Packing

Input: n boxes of capacities c_1, c_2, \dots, c_n

m items of sizes s_1, s_2, \dots, s_m

Can put **at most 1** item in a box

Item j can be put into box i if $s_j \leq c_i$

Output: A way to put as many items as possible in the boxes.

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Example:

- Box capacities: 60, 40, 25, 15, 12
- Item sizes: 45, 42, 20, 19, 16
- Can put 3 items in boxes: 45 \rightarrow 60, 20 \rightarrow 40, 19 \rightarrow 25

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Designing a Reasonable Strategy for Box Packing

- Q: Take box 1. Which item should we put in box 1?