

5CCS2FC2: Foundations of Computing II

**The Cook-Levin Theorem &
Introduction to Graph Algorithms**

Week 4

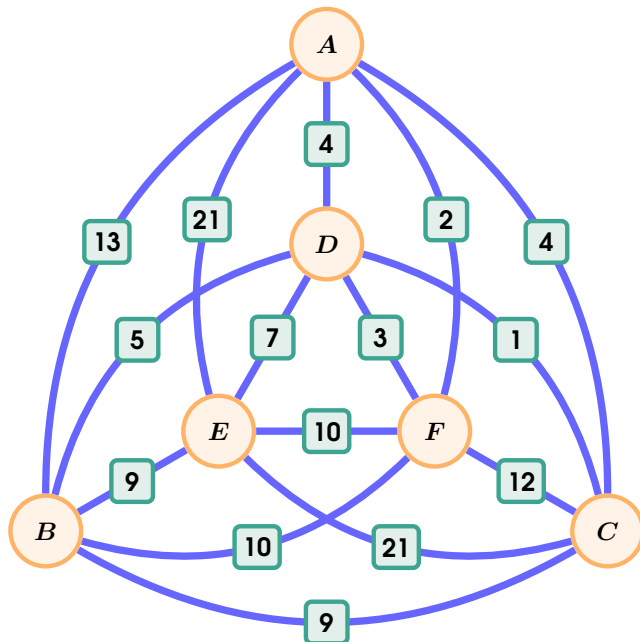
Dr Christopher Hampson

Department of Informatics

King's College London

Warm-up : Find the shortest path

- Find the shortest path through the following graph that visits each vertex **exactly once** and returns to the starting point.



	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>
<i>A</i>	0	13	4	4	7	2
<i>B</i>	13	0	9	13	9	10
<i>C</i>	4	9	0	1	21	12
<i>D</i>	4	13	1	0	7	3
<i>E</i>	7	9	21	7	0	10
<i>F</i>	2	10	12	3	10	0

List of NP-complete Problems

- (Incomplete) List of NP-complete Problems

- The Boolean Satisfiability Problem ✓
- The Graph Colouring Problem ✓
- The Clique problem ✓
- The Hamiltonian Cycle Problem
- The Travelling Salesman Problem (TSP)
- The Knapsack Problem

https://en.wikipedia.org/wiki/List_of_NP-complete_problems

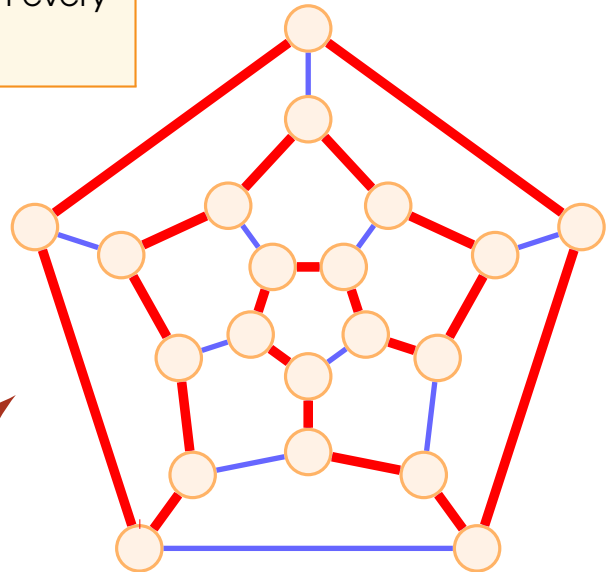
More NP-complete Problems

The Hamiltonian Cycle Problem

Input) An undirected graph $G = (V, E)$

Output) **True** if and only if there is a *closed path* through G that passes through every vertex exactly once

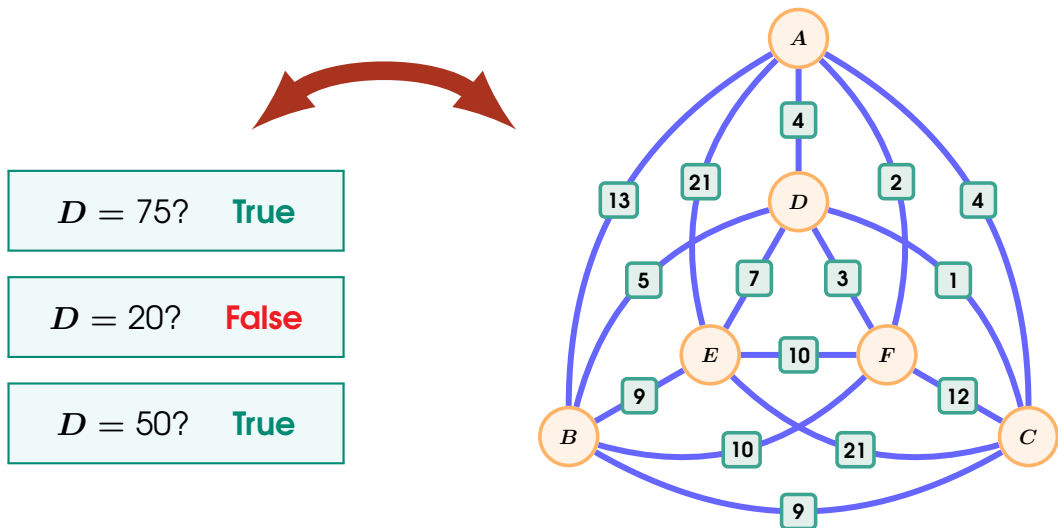
**Has a
Hamiltonian
Cycle**



The Travelling Salesman Problem

Input) A complete weighted graph $G = (V, d)$,
where $d : V \times V \rightarrow \mathbb{R}$, and target distance $D > 0$.

Output) **True** if and only if there is a *closed path* through G of
length $< D$ that passes through every vertex exactly once



The Knapsack Problem

Input) A collection of items $I = \{1, 2, \dots, n\}$ each with weight w_i and value v_i , a target value V and a maximum capacity W .

Output) **True** if and only if there is subset of items $S \subseteq I$ such that $\sum_{i \in S} v_i > V$ and $\sum_{i \in S} w_i < W$.

Item 1

Weight : 10

Value : £ 60

Item 2

Weight : 20

Value : £ 100

Item 3

Weight : 30

Value : £ 120

$V = \text{£ } 270$

$W = 70$

True

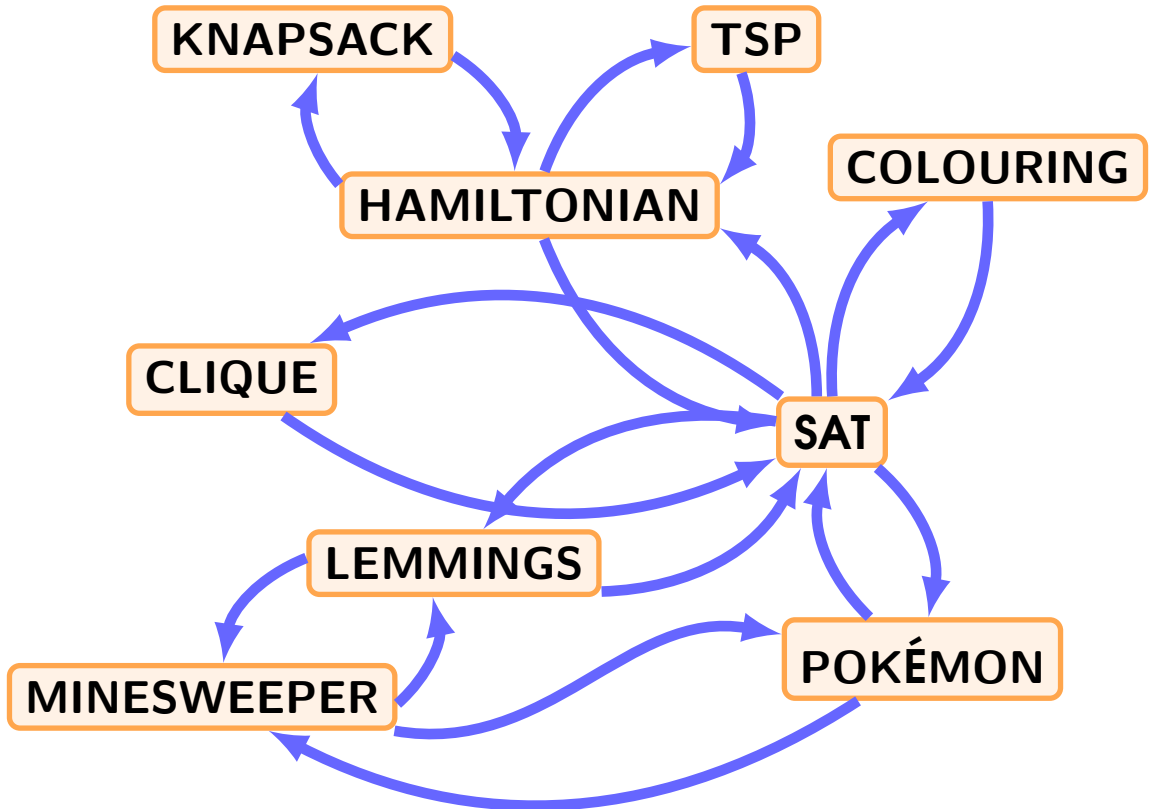
$V = \text{£ } 200$

$W = 40$

False



Relationships between NP-complete Problems



The Cook-Levin Theorem

(The Existence of NP-Complete Problems)

The Cook-Levin Theorem

Theorem (The Cook-Levin Theorem) The Boolean Satisfiability problem **SAT** is **NP**-complete.

(i.e. SAT is among the hardest problems in NP)

Proof (sketch): (you do NOT need to memorise this proof)

We want to show that **every** problem $X \in \mathbf{NP}$ can be reduced to **SAT**.

Step 1) Let $X \in \mathbf{NP}$ be **any problem** belonging to the class **NP**.

By **definition** there is some NDTM \mathcal{M} such that

\mathcal{M} has a polynomial-length computation that accepts $w \iff w \in X$

(for all input words $w \in \Sigma^*$)

The Cook-Levin Theorem

Step 2) For each **state** $q \in Q$, **symbol** $a \in \Sigma$ and **integers** $i, t \leq T(n)$

cell $_{i,t}(a)$ = Cell i of the \mathcal{M} 's tape contains symbol a at time t

head $_{i,t}$ = The tape head is in position i at time t

state $_{q,t}$ = The machine is in state q at time t

(where $T(n) \in O(n^k)$ is the worst-case termination-time of \mathcal{M})

Step 3) With these **propositional variables**, we make the following claim:

Claim: We can write down a propositional formula that describes the behaviour of a NDTM on a given input

$F_{\mathcal{M},w} = \mathcal{M}$ has a computation that accepts w

The Cook-Levin Theorem

Step 4) We then have that

$$w \in X \iff F_{\mathcal{M},w} \text{ is satisfiable}$$

which is to say that $X \leq_p \mathbf{SAT}$.

Step 5) Since we have assumed **nothing special** about X other than it can be solved by a NDTM in polynomial time (and space), we must have that

$$X \leq_p \mathbf{SAT} \quad \text{for all } X \in \mathbf{NP}$$

which is to say that **SAT** is **NP-complete**

Q.E.D.

Graph Algorithms

Data Structures and Representations

- Data Structures for Graphs

- The type of **data structure** used to store a graph can affect the efficiency of your algorithms, and memory requirements

List of Edges

Edges	Edges
(1, 5)	(2, 1)
(3, 4)	(5, 2)
(1, 2)	(5, 1)
(2, 4)	(3, 2)
(2, 3)	(4, 2)
(2, 5)	(4, 3)

Adjacency List

Vertex	Successors
1	2, 5
2	1, 3, 4, 5
3	2, 4
4	2, 3, 5
5	1, 2, 4

Adjacency Matrix

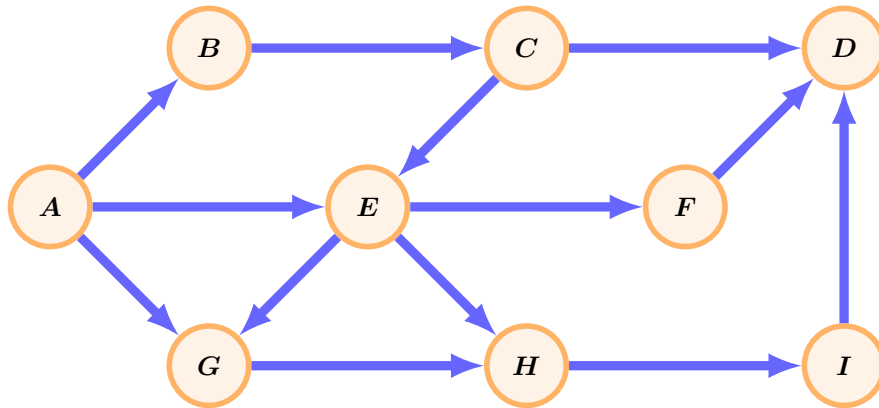
	1	2	3	4	5
1	0	1	0	0	1
2	1	0	1	1	1
3	0	1	0	1	0
4	0	1	1	0	1
5	1	1	0	1	0

Breadth-First-Search

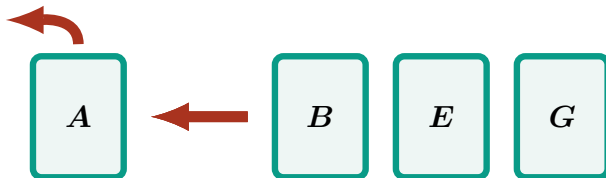
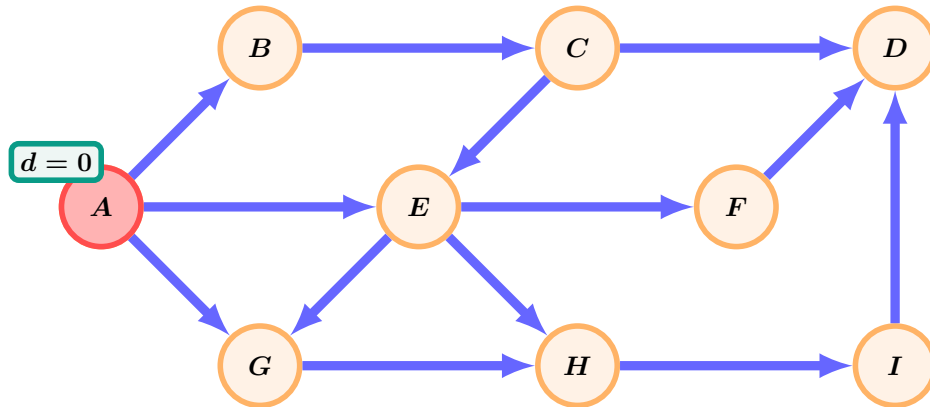
Breadth-First-Search (BFS)

- Step 1)** Select a root node $r \in V$ and set $d(r) = 0$.
- Step 2)** Add v to a queue.
- Step 3)** De-queue the first element u from the queue.
- Step 4)** For each vertex v adjacent to u , if $d(v)$ is not yet defined, then set $d(v) = d(u) + 1$ and add v to the queue.
- Step 5)** Repeat from Step 3 until the queue is empty.

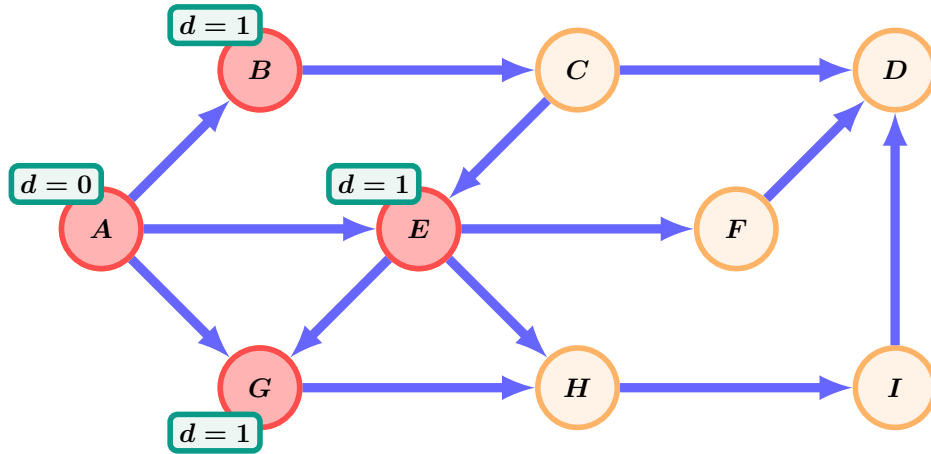
Breadth-First-Search



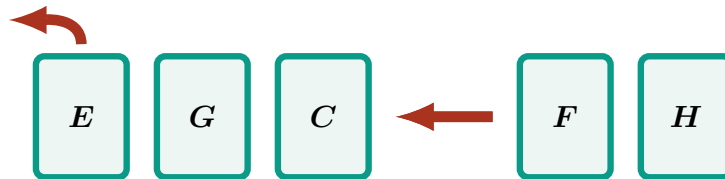
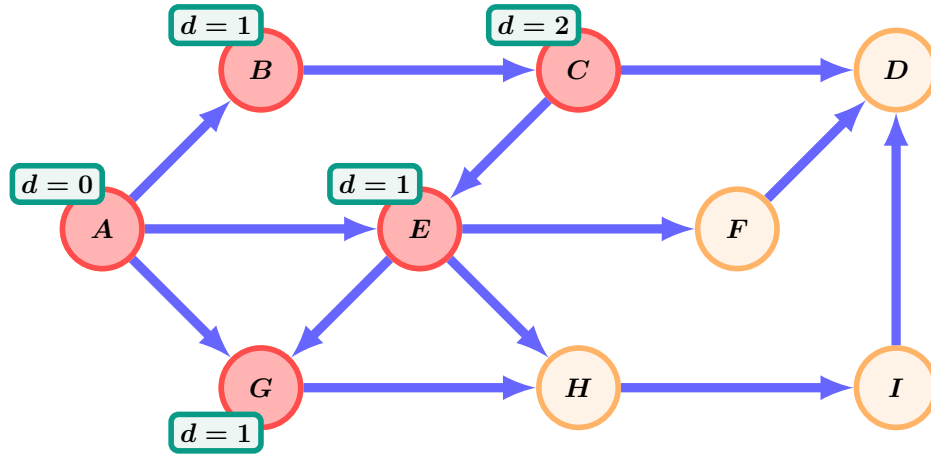
Breadth-First-Search



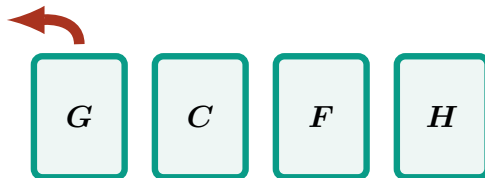
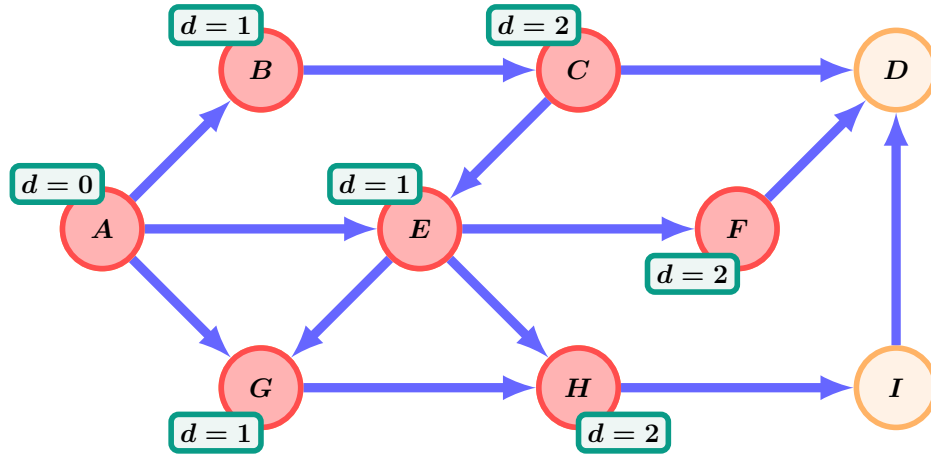
Breadth-First-Search



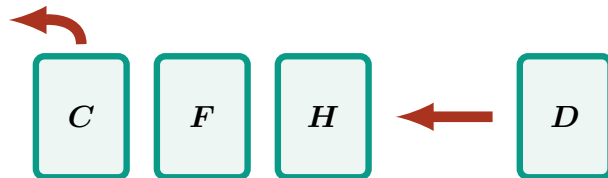
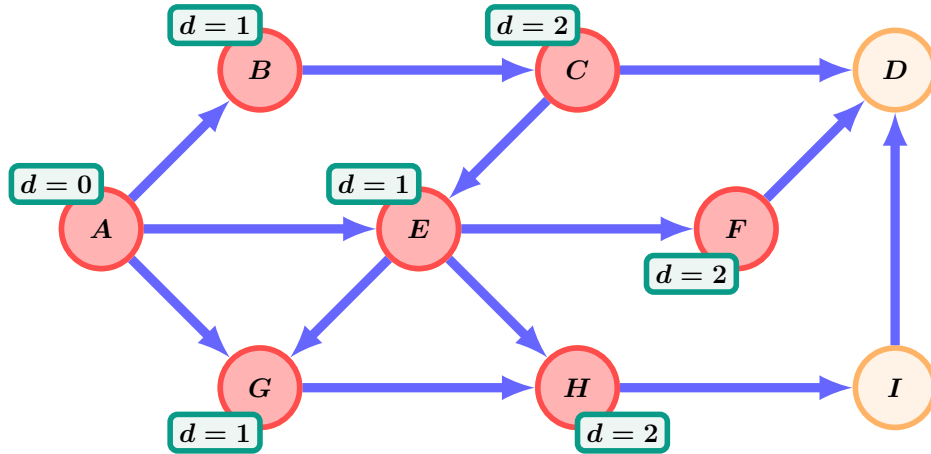
Breadth-First-Search



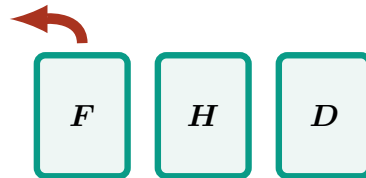
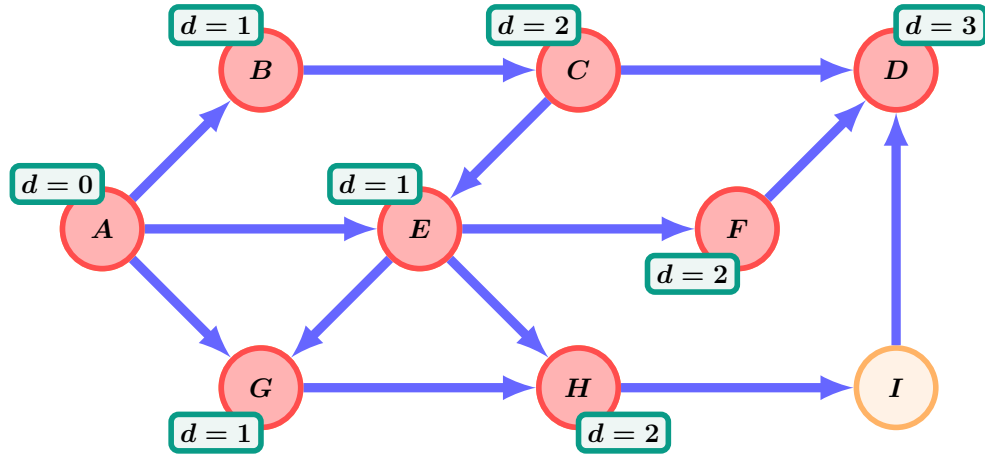
Breadth-First-Search



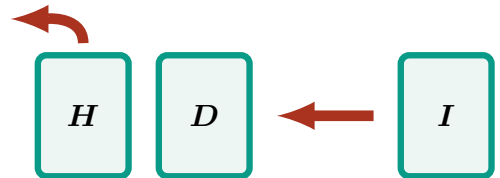
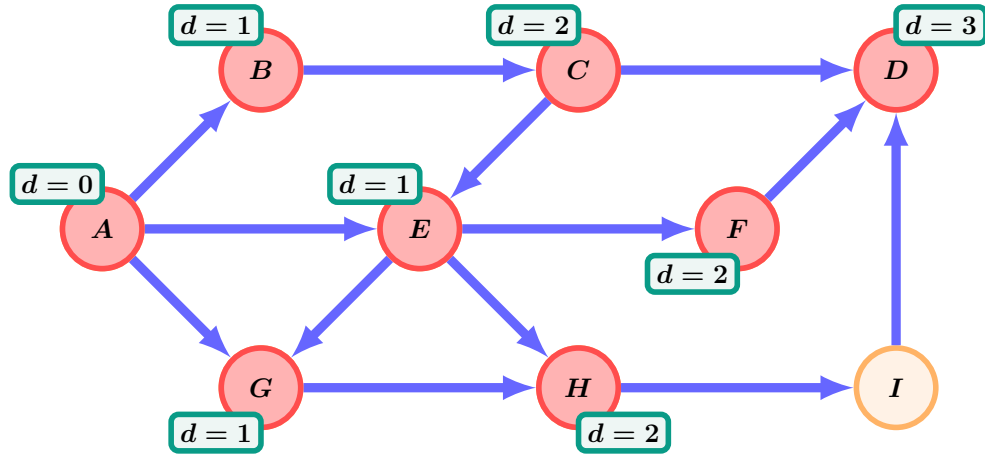
Breadth-First-Search



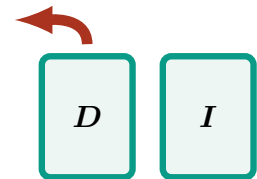
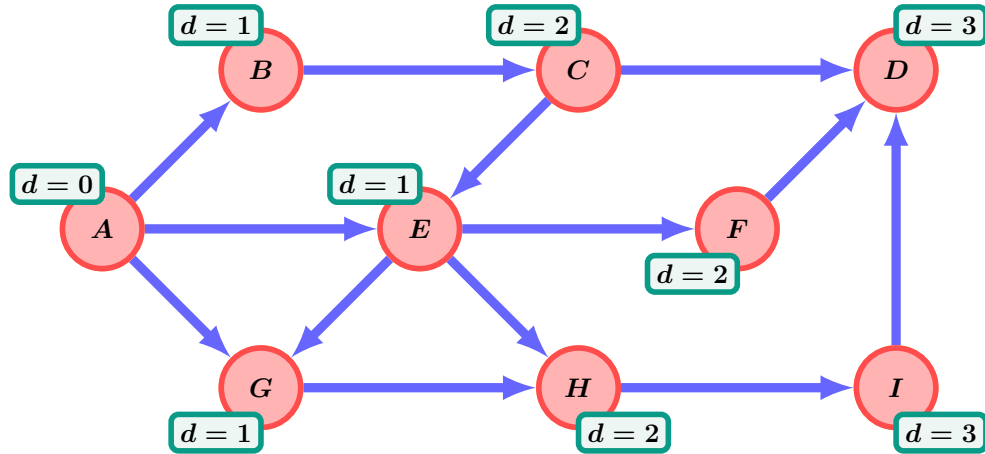
Breadth-First-Search



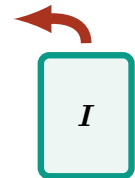
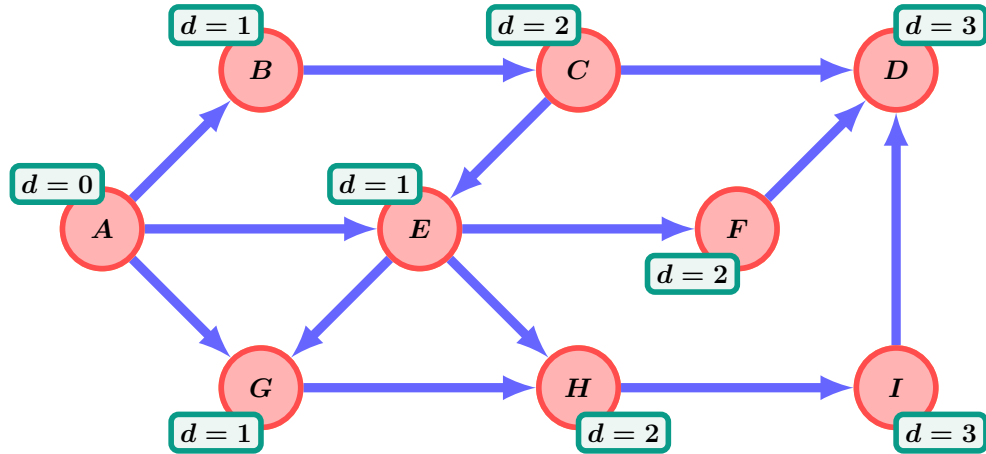
Breadth-First-Search



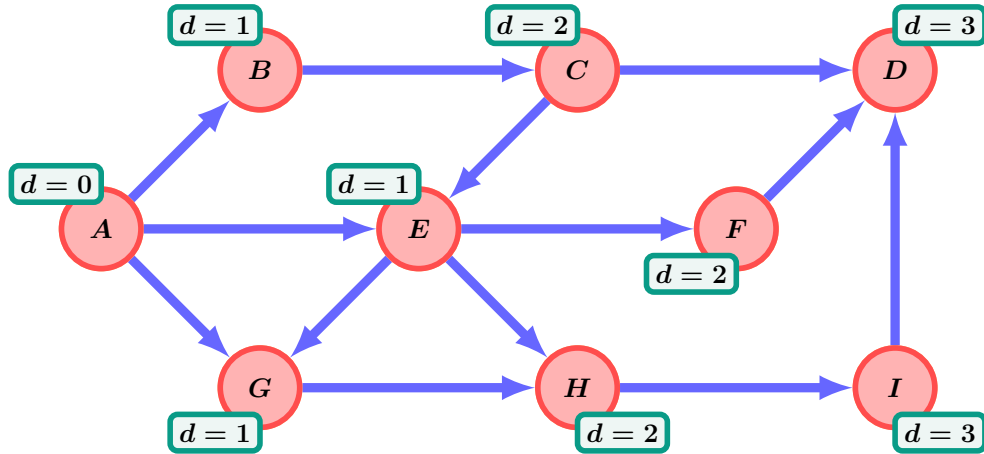
Breadth-First-Search



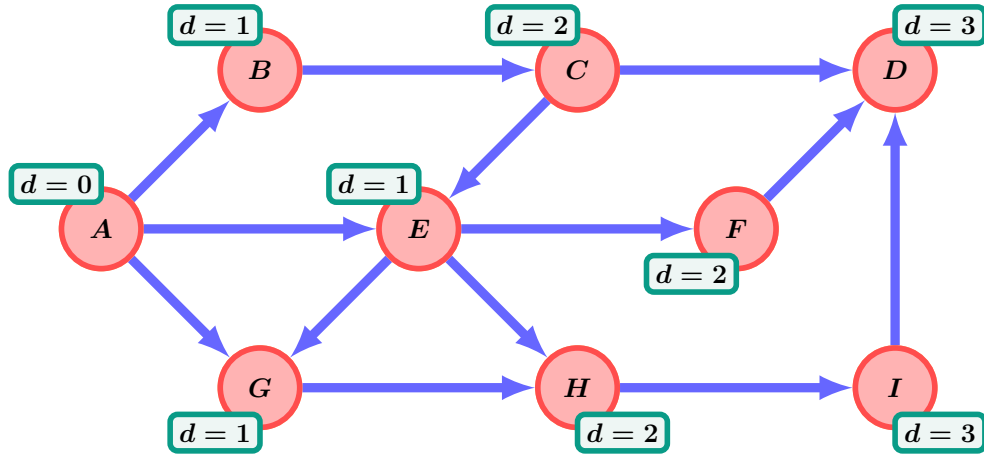
Breadth-First-Search



Breadth-First-Search



Breadth-First-Search



Queue Empty so Terminate!

Breadth-First-Search

Theorem The worst-case termination time for Breadth-First-Search is $O(|V| + |E|)$.

Proof:

- Each vertex is enqueued at most once and so the number of times that **Step 3** and **Step 5** are repeated is at most $O(|V|)$ -times.
- Furthermore, for each vertex u we repeat **Step 4** at most $|\text{Adj}(u)|$ -times (where $\text{Adj}(u) \subseteq E$ is the set of adjacent vertices) In total, we repeat **Step 4** at most $O(|E|)$ -times, since

$$|\text{Adj}(u_1)| + |\text{Adj}(u_2)| + \dots + |\text{Adj}(u_n)| = |E|$$

Q.E.D.

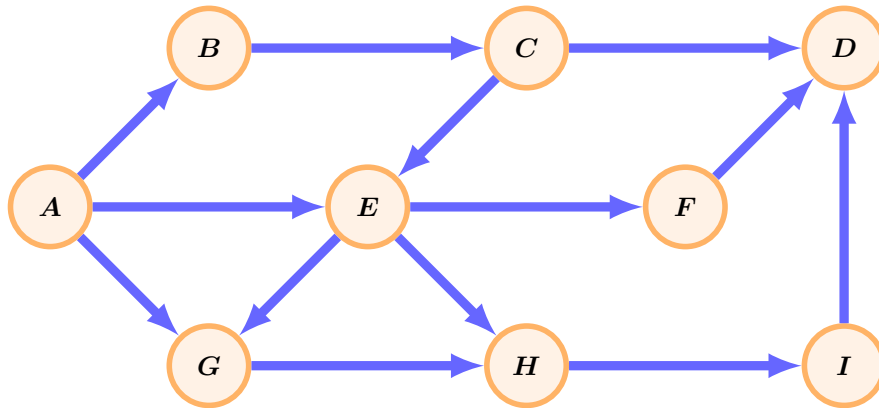
Depth-First-Search

Depth-First-Search (DFS)

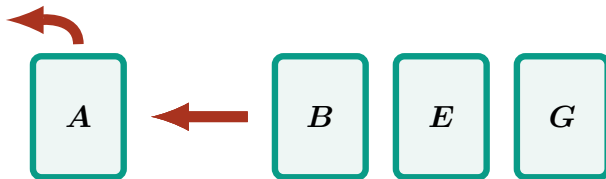
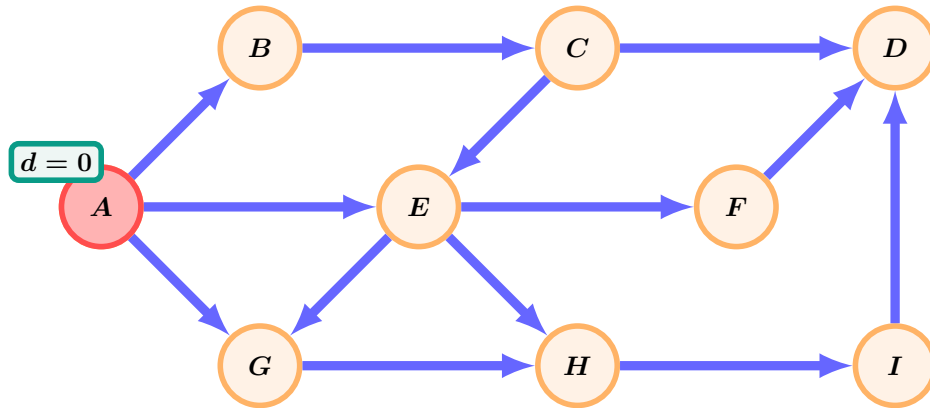
- Step 1)** Select a root node $r \in V$ and set $d(r) = 0$.
- Step 2)** Add v to a stack.
- Step 3)** Pop the first element u from the stack.
- Step 4)** For each vertex v adjacent to u , if $d(v)$ is not yet defined, then set $d(v) = d(u) + 1$ and push v to the stack.
- Step 5)** Repeat from Step 3 until the stack is empty.

(the implementation appears differently in Cormen et.al, but the principle is the same)

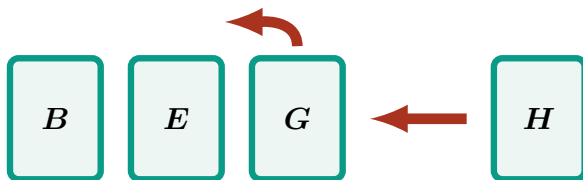
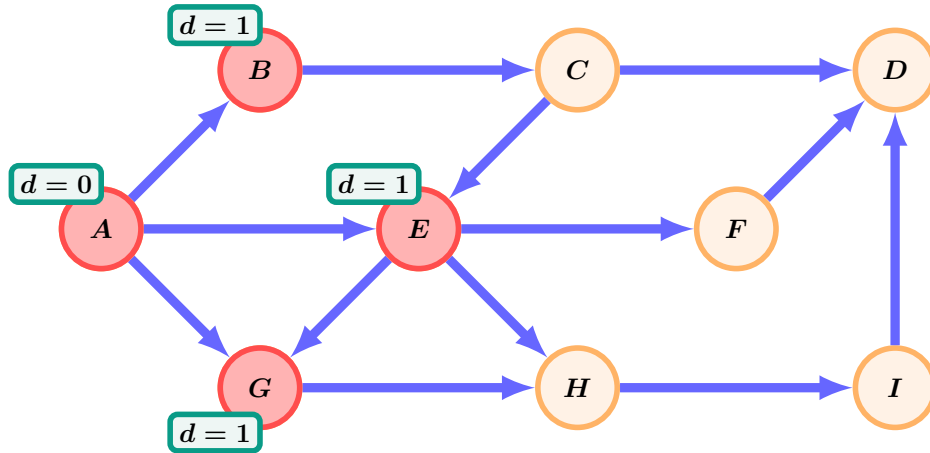
Depth-First-Search



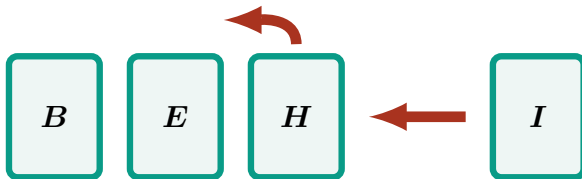
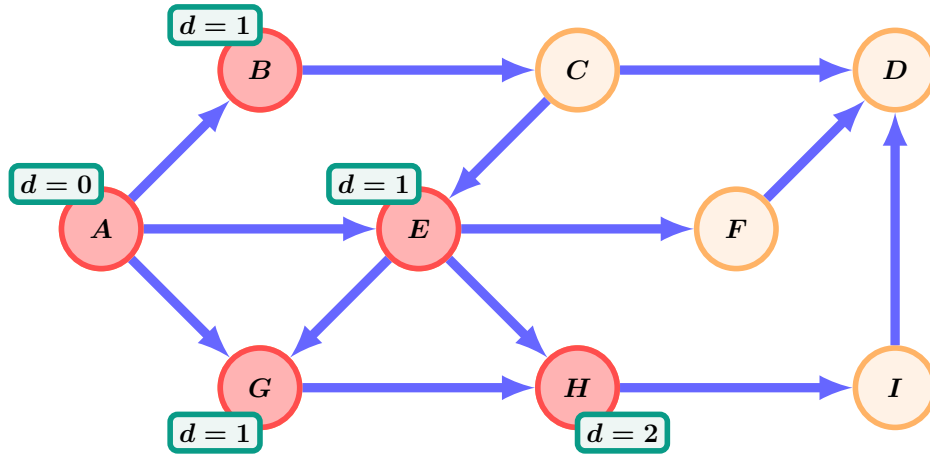
Depth-First-Search



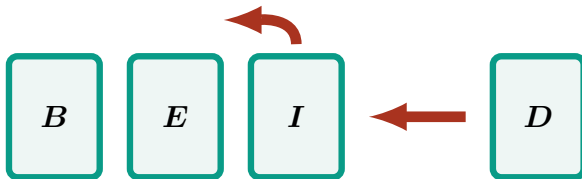
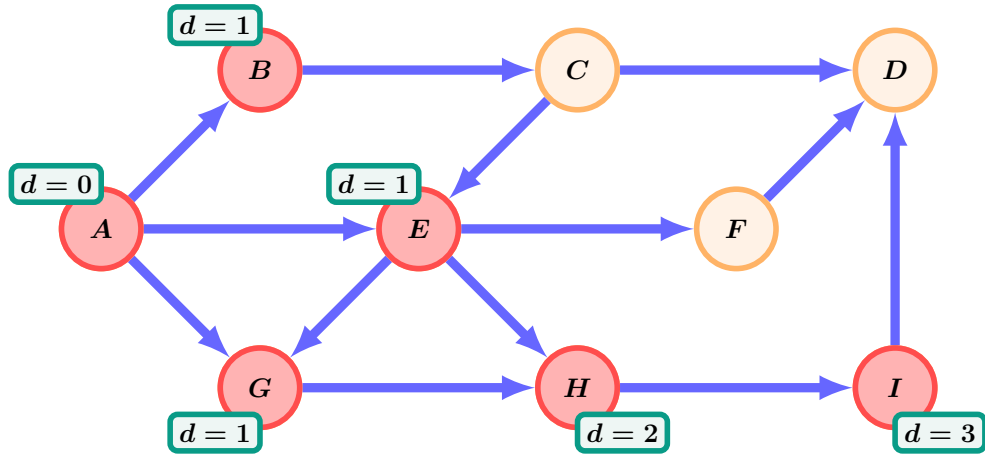
Depth-First-Search



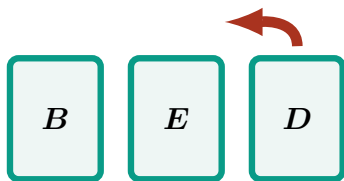
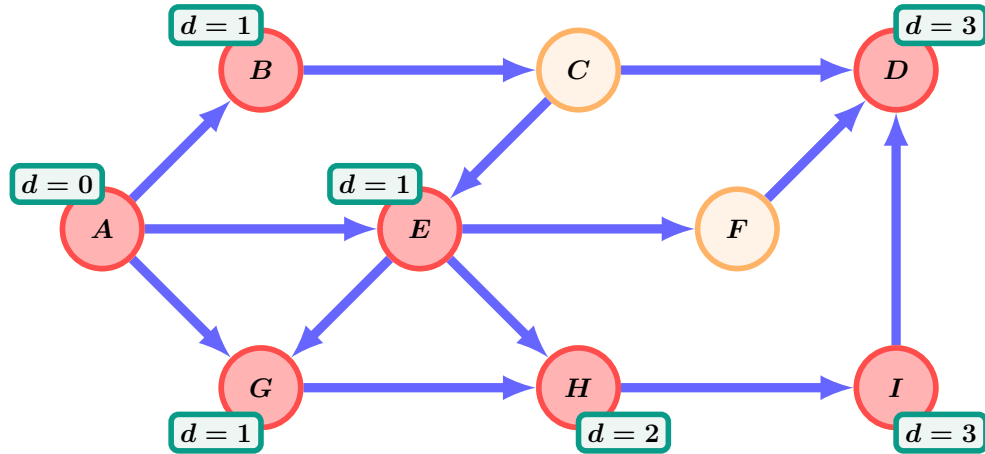
Depth-First-Search



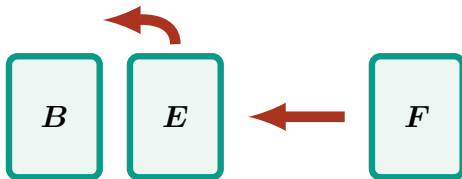
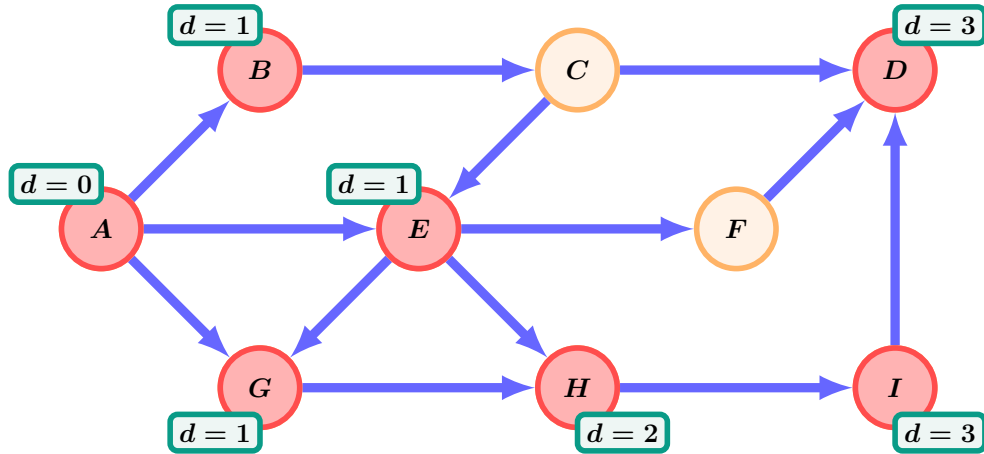
Depth-First-Search



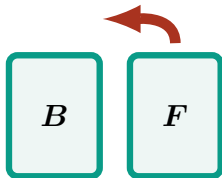
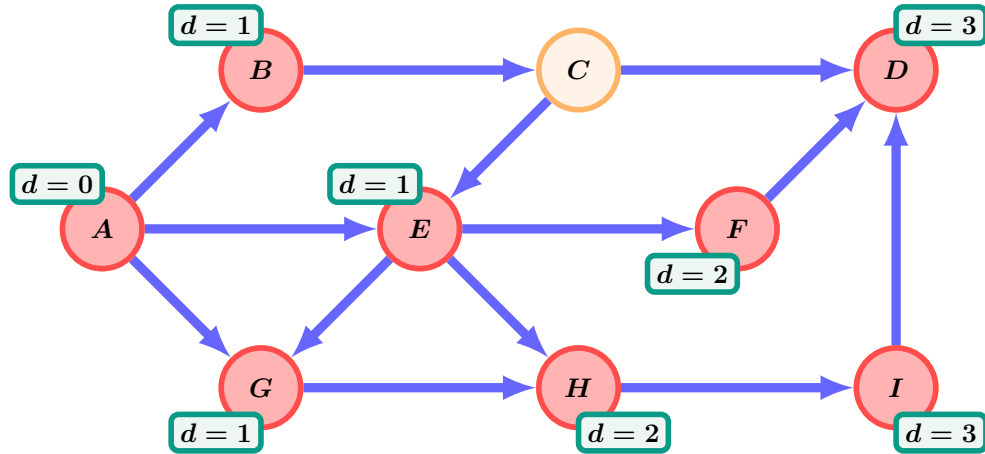
Depth-First-Search



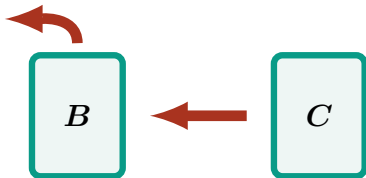
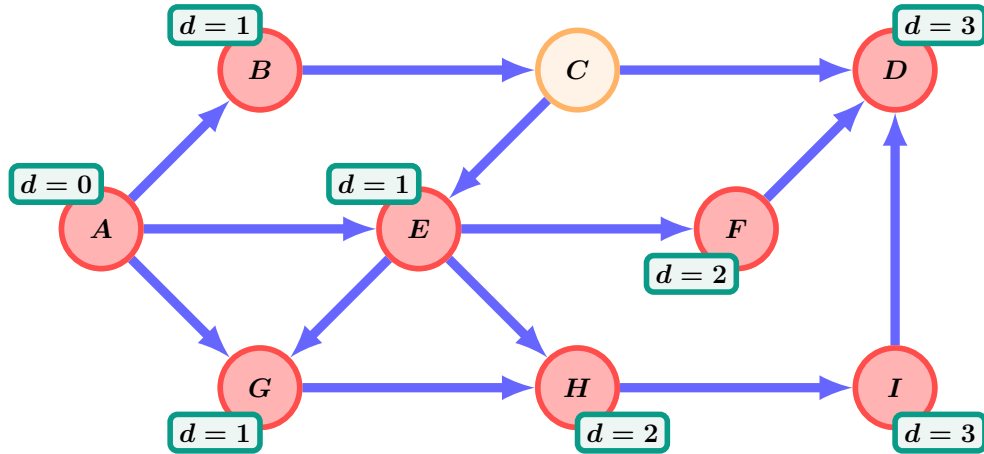
Depth-First-Search



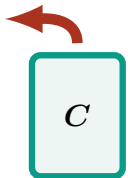
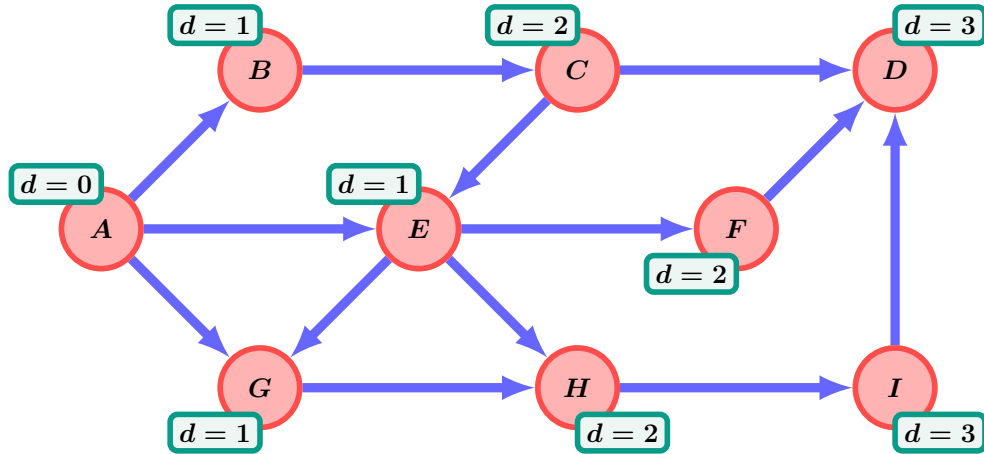
Depth-First-Search



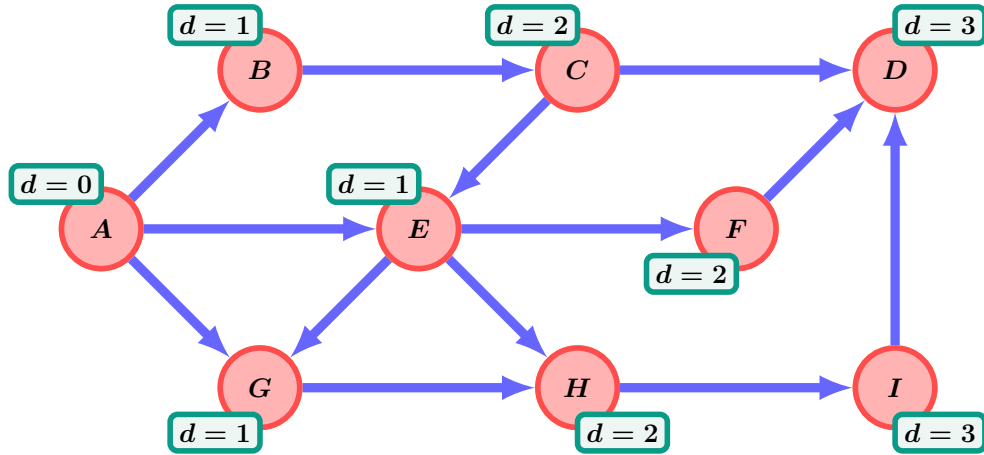
Depth-First-Search



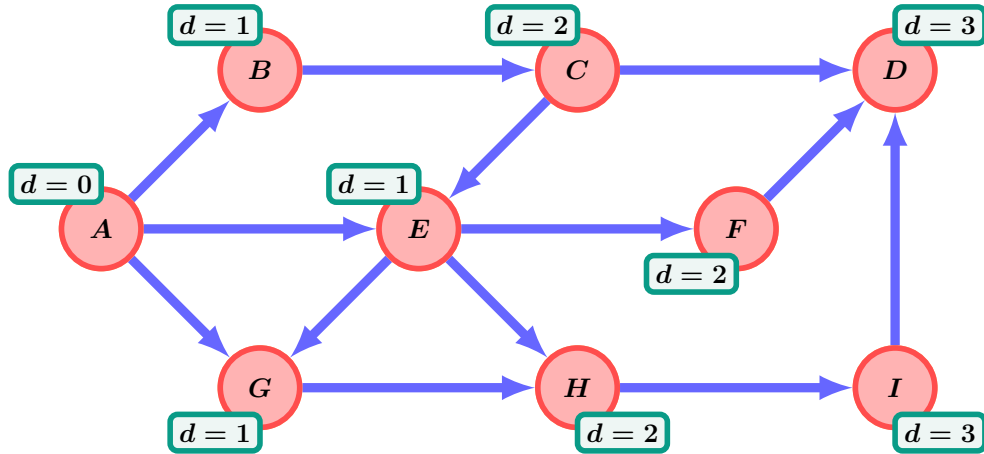
Depth-First-Search



Depth-First-Search



Depth-First-Search



Stack Empty so Terminate!

Depth-First-Search

Theorem The worst-case termination time for Depth-First-Search is $O(|V| + |E|)$.

Proof:

- The proof is almost identical to that of the **Breadth-First-Search**, replacing the **queue** for a **stack**.

Q.E.D.

End of Slides!



Feedback

- Let me know how you found today's lecture!



<https://goo.gl/forms/Q7NWfrgQiK0fKSJd2>