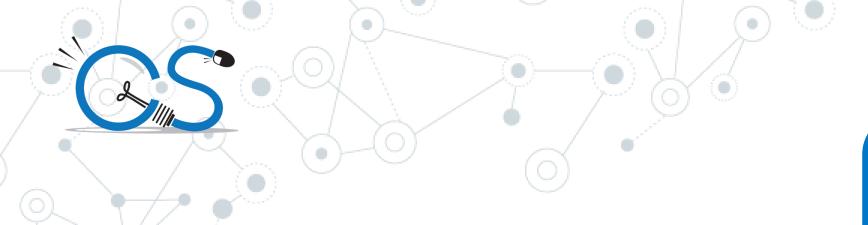


## APPLICATION IN DESIGNING ALGORITHMS: GRAPH ALGORITHMS

## **GROUP 12**

Lê Quang Thiên Phúc Lý Nguyên Thùy Linh





O1 Overview

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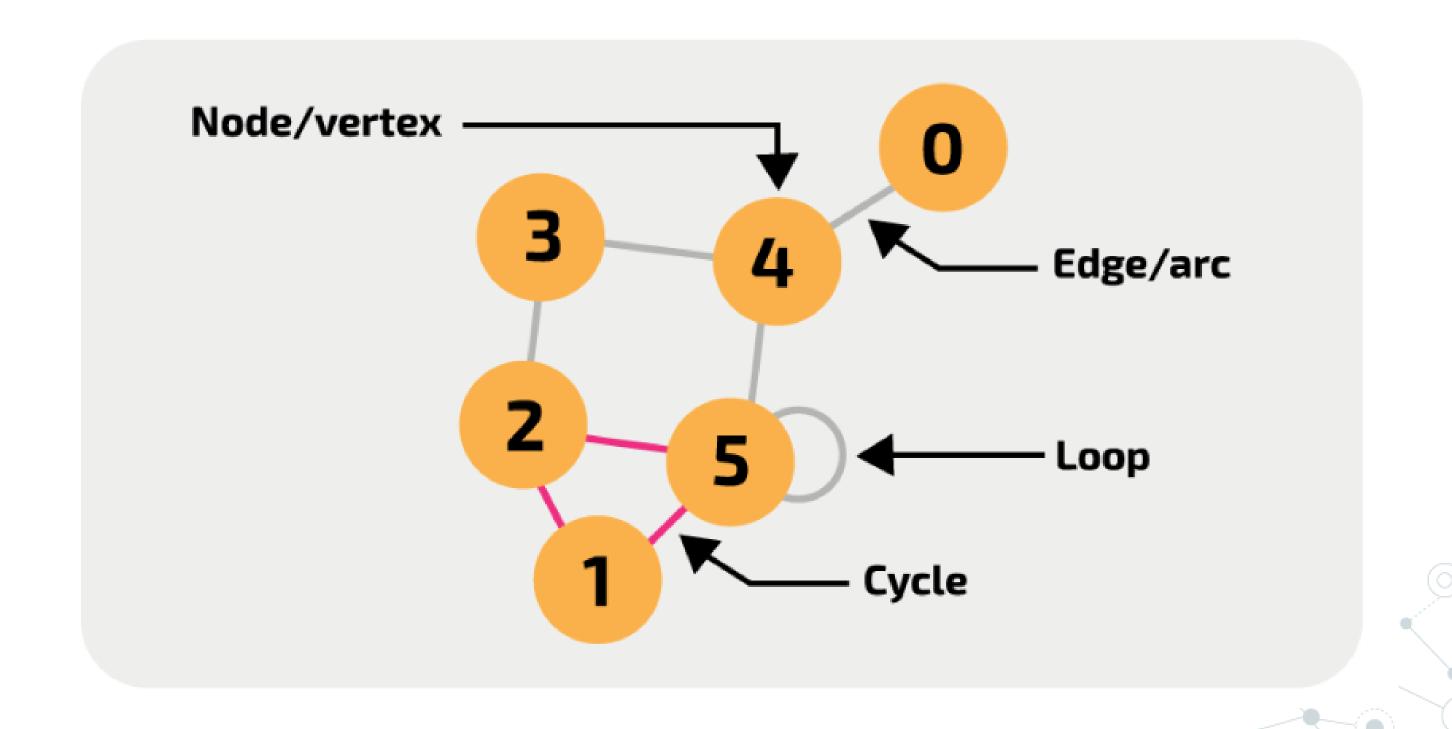




## What is a GRAPH?



#### What is a GRAPH?







# What are differences between geometry and graph?



## Traversing Algorithm

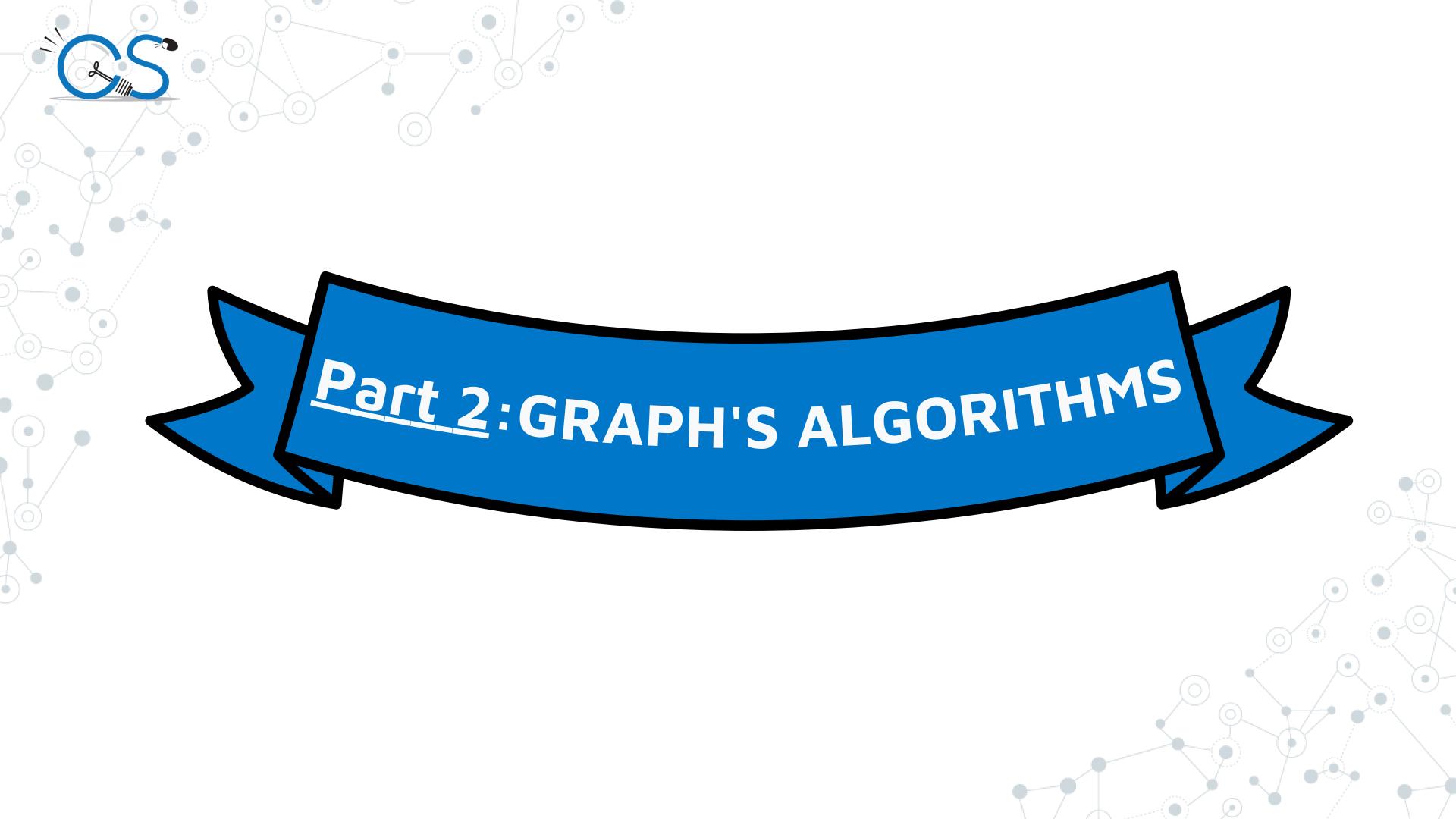
**DFS** 

**BFS** 



	XXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXX	
	DFS	BFS
disc	<ul> <li>explores as deeply as possible along a branch before backtracking</li> <li>processes from farther to nearer</li> </ul>	<ul> <li>explores vertices level by level</li> <li>processes from nearer to farther</li> </ul>
Approach	Backtracking	???
Data structure	stack	queue
Specific use cases	• ???	• finding shortest path (in a map, a network, a puzzle,)
Common use cases	Better if the graph is wide	Better if solutions are shallow
Time complexity	O( V  +  E )  V : numbers of vertices	

|E|: numbers of edges





## Topological Sort

**Problem Statement** 

#### **Algorithm**

- Brute Force
- Backtracking
- Divide & Conquer
- Dynamic Programming
- Comparison



Algorithm

## Part 2: GRAPH'S ALGORITHMS

#### **Topological Sort**

#### **Problem Statement**

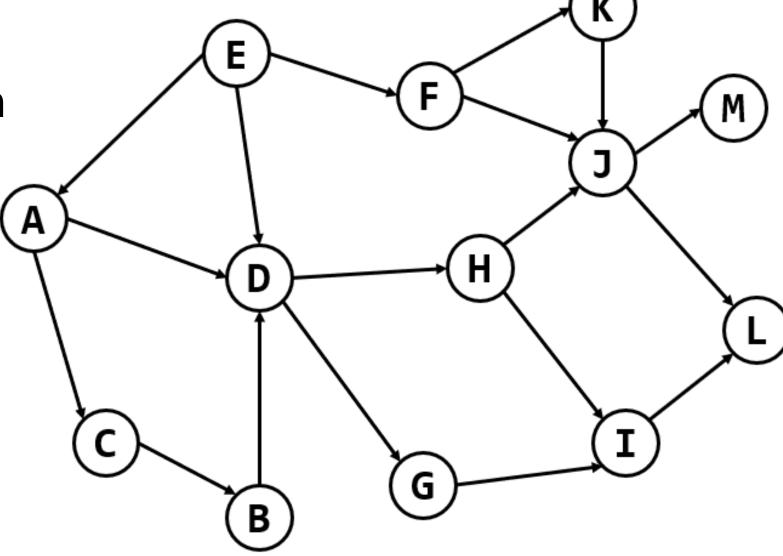
Given a directed acyclic graph, find an disc of vertices so that for every directed edge u-v, vertex u comes before vertex v (finding a topological disc).

Input: a directed acyclic graph

Output: a topological disc

Constraint:

 $|V| \le 3*10^4$ ,  $|E| \le 2*10^5$ 





#### Algorithm

- Backtracking
- Divide & Conquer
- Dynamic Programming
- Comparison

## Part 2: GRAPH'S ALGORITHMS

**Topological Sort** 

#### **Brute Force**

- Let C be any order of the vertices.
- → The number of orders is |V|!.
- To check if C is valid: O(|V|²(|V|+|E|)) (check if all pairs are valid).



How about other approaches?



#### Algorithm

- Brute Force
- Divide & Conquer
- Dynamic Programming
- Comparison

## Part 2: GRAPH'S ALGORITHMS

**Topological Sort** 

## Backtracking

Idea: Iteratively insert vertices into a solution.

- If a newly added vertex can reach any vertices already in the solution, try another vertex.
- This will reduce some solution of brute force but have the same complexity.



#### Algorithm

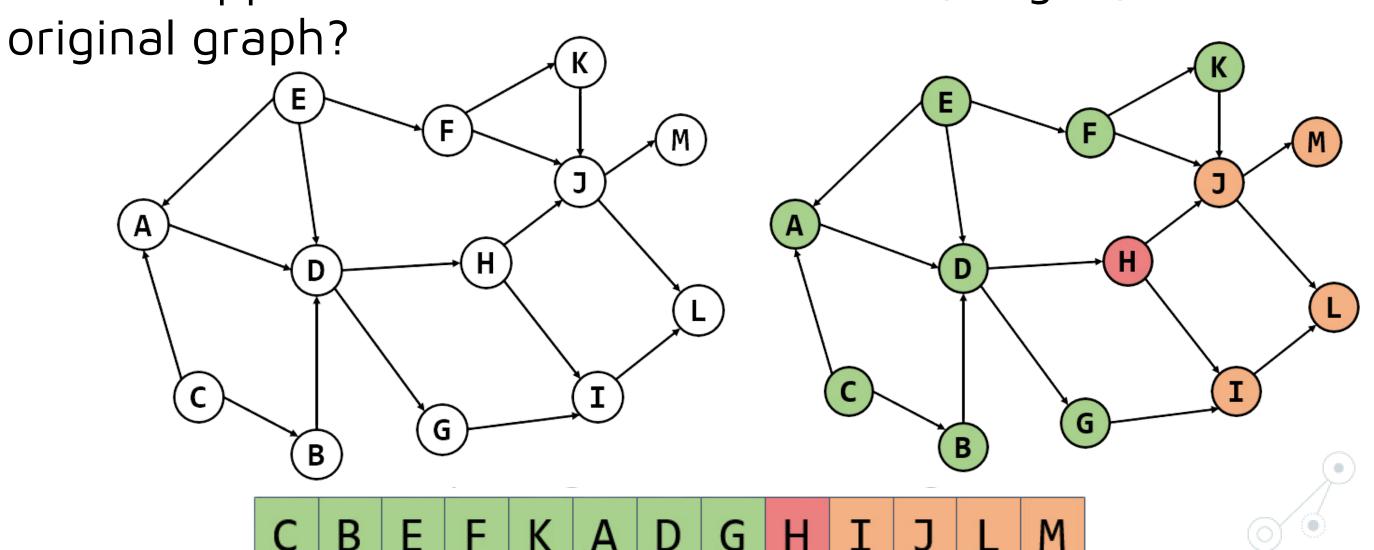
- Brute Force
- Backtracking
- Dynamic Programming
- Comparison

## Part 2: GRAPH'S ALGORITHMS

#### Topological Sort

#### Divide & Conquer

What happens if we remove a vertex, e.g H, from the



This will divide the graph into two independent parts:

the orange part, which can be reached from H, and
the green part, which cannot be reached from H.



#### Algorithm

- Brute Force
- Backtracking
- Dynamic Programming
- Comparison

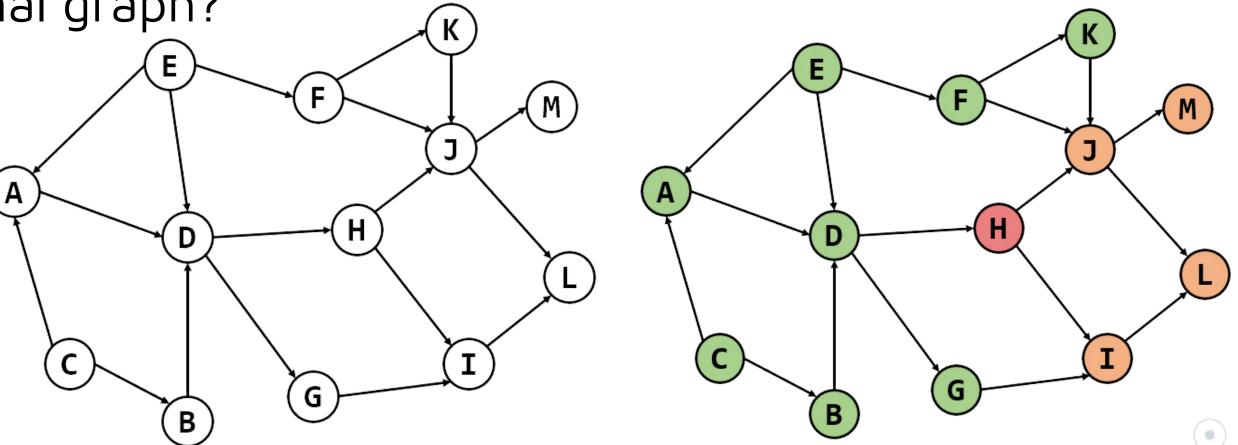
## Part 2: GRAPH'S ALGORITHMS

#### Topological Sort

#### Divide & Conquer

What happens if we remove a vertex, e.g H, from the

original graph?



C B E F K A D G H I J L M

We can build the whole solution in a single container.
 First, we build the orange part, then insert H, then build the green part. That can be done by using DFS.



#### Algorithm

- Brute Force
- Backtracking
- Dynamic Programming
- Comparison

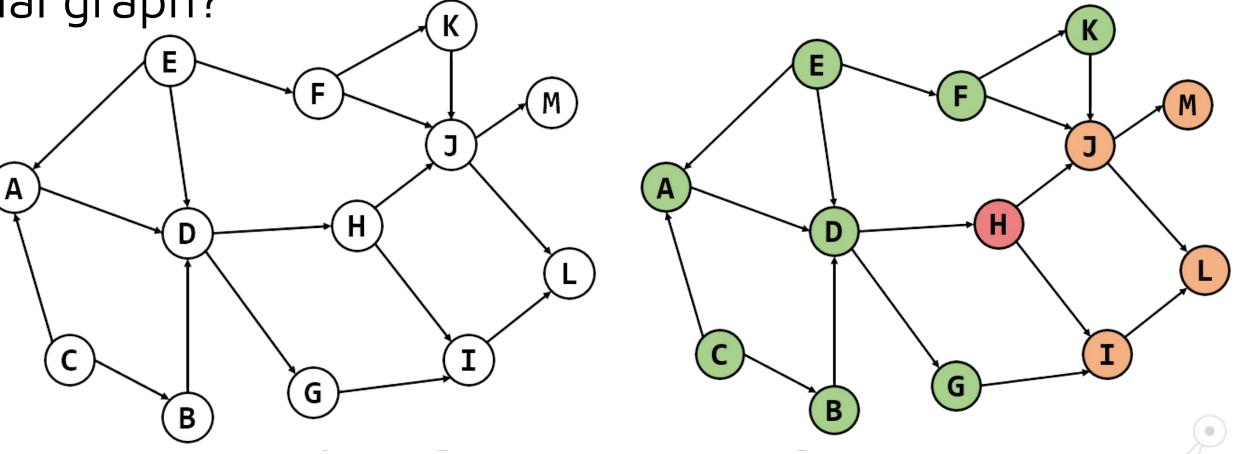
## Part 2: GRAPH'S ALGORITHMS

#### **Topological Sort**

#### Divide & Conquer

What happens if we remove a vertex, e.g H, from the





C B E F K A D G H I J L M

Time Complexity: **O(|V|+|E|)** 



#### Algorithm

- Brute Force
- Backtracking
- Dynamic Programming
- Comparison

## Part 2: GRAPH'S ALGORITHMS

#### Topological Sort

#### Divide & Conquer

#### Pseudocode:

```
ts <- empty list
for each unvisited vertex u:
   topo_sort(u)
function topo_sort(root):
   visited[root] <- True
   for each ajacent vertex v of root:
     if not visited[v]:
        topo_sort(v)
   insert root at the beginning of ts</pre>
```



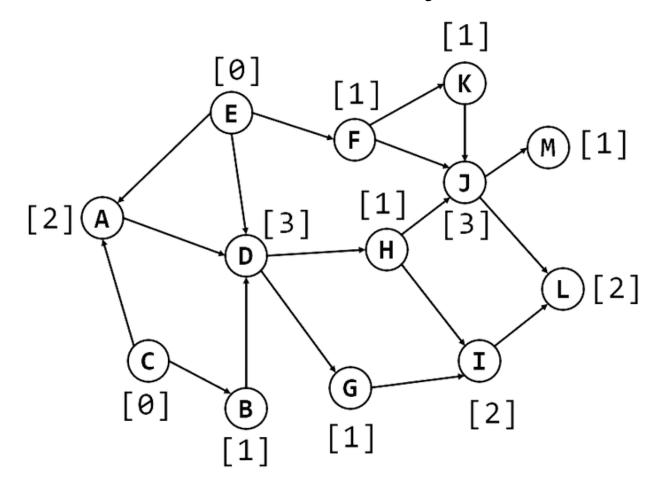
#### Algorithm

- **Brute Force**
- Backtracking
- Divide & Conquer
- Comparison

#### Part 2: GRAPH'S ALGORITHMS

## Topological Sort Dynamic Programming

 Noticing that a vertex can be put at the beginning of a topological order if and only if its in-degree is O



Queue: [E][C]

• In this case, both E and C can be put at the beginning of a topological order.



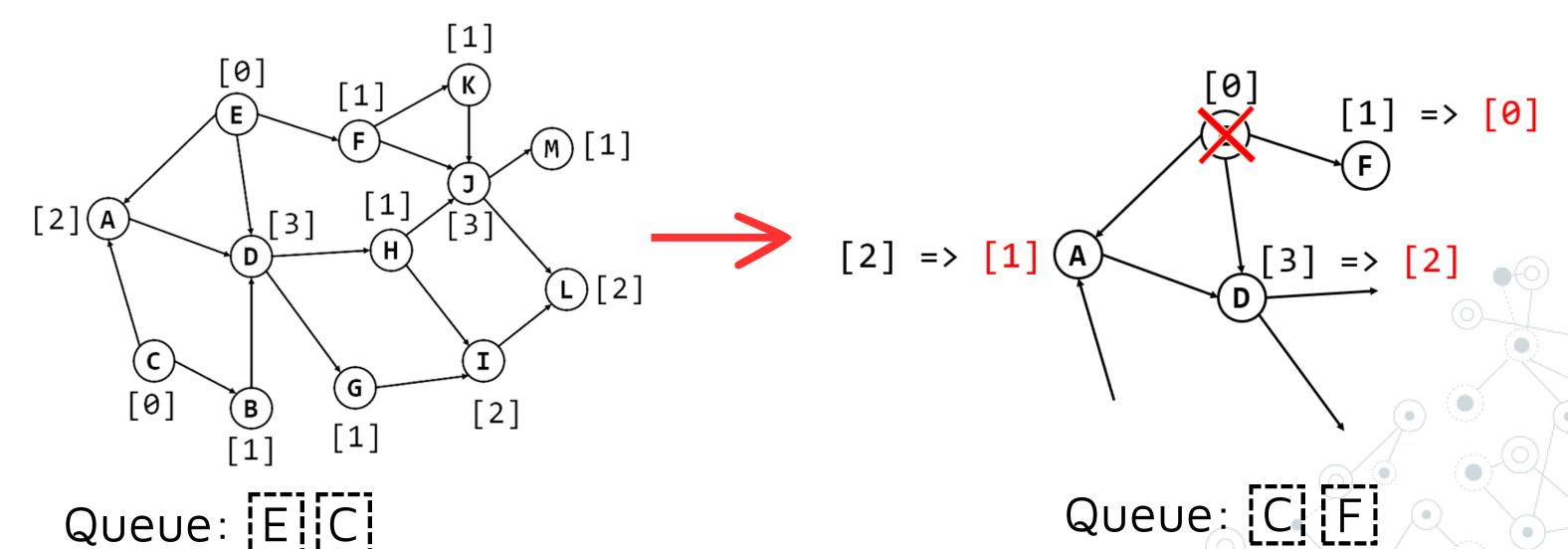
#### Algorithm

- Brute Force
- Backtracking
- Divide & Conquer
- Comparison

## Part 2: GRAPH'S ALGORITHMS

## Topological Sort Dynamic Programming

Now if we remove E from the graph (insert it to the topological order), we can efficiently calculate the in-degrees of its neighbors.



F now have in degree of O. Both C and F can be put right after E



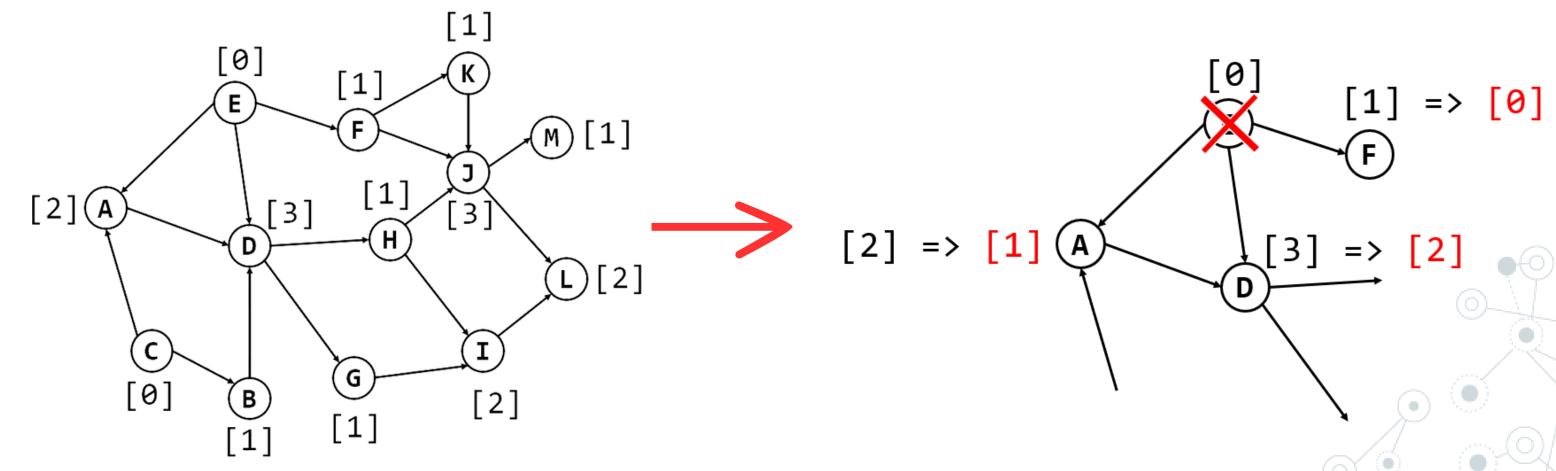
#### Algorithm

- Brute Force
- Backtracking
- Divide & Conquer
- Comparison

## Part 2: GRAPH'S ALGORITHMS

## Topological Sort Dynamic Programming

Now if we remove E from the graph (insert it to the topological order), we can efficiently calculate the in-degrees of its neighbors.



Queue: [E][C]

Queue: [C] [F]

The number of such calculations is the number of edges. Time Complexity: **O(|V|+|E|)** 



#### Algorithm

- Brute Force
- Backtracking
- Divide & Conquer
- Comparison

## Part 2: GRAPH'S ALGORITHMS

## **Topological Sort**

#### Dynamic Programming

Pseudocode:

```
in_deg <- a list of size |V| with all Ø
ts <- empty list
zero_deg <- empty list</pre>
for each vertex u:
  for each child v of u:
    in_deg[v] <- in_deg[v]+1</pre>
  for each vertex u:
    if in_deg[u]=Ø:
      insert u to zero_deg
  while zero_deg is not empty:
    u <- any element of zero_deg</pre>
    insert u to ts
    for each child v of u:
      in_deg[v] ← in_deg[v] - 1
      if in_deg[v]=Ø:
         insert v to zero_deg
```



#### Algorithm

- Brute Force
- Backtracking
- Divide & Conquer
- Dynamic Programming

#### Part 2: GRAPH'S ALGORITHMS

## Topological Sort Comparison

#### DAC

- Run faster for large graphs
- May fail to detect cycles without further improvement

#### DP

- Run slow for large graphs
- Can always to detect cycles without further improvement





**Problem Statement** 

#### **Algorithm**

- Problem Analysis
- Finding Bridges
- Finding Articulation points



Algorithm

#### Part 2: GRAPH'S ALGORITHMS

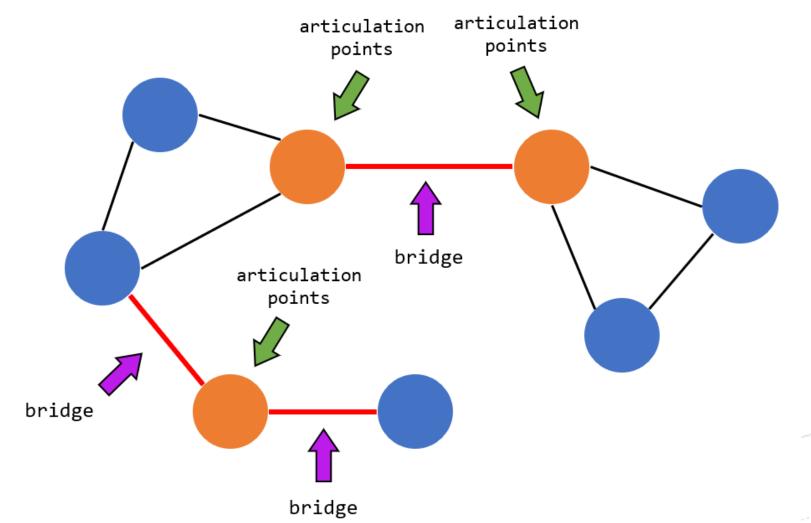
## Tarjan's Algorithm Problem Statement

Given an undirected graph, find all bridges and articulation points in the graph.

Input: an undirected graph

Output: a list of bridges, a list of articulation points

Constraint: |V|≤3\*10^4, |E|≤2\*10^5





#### **Algorithm**

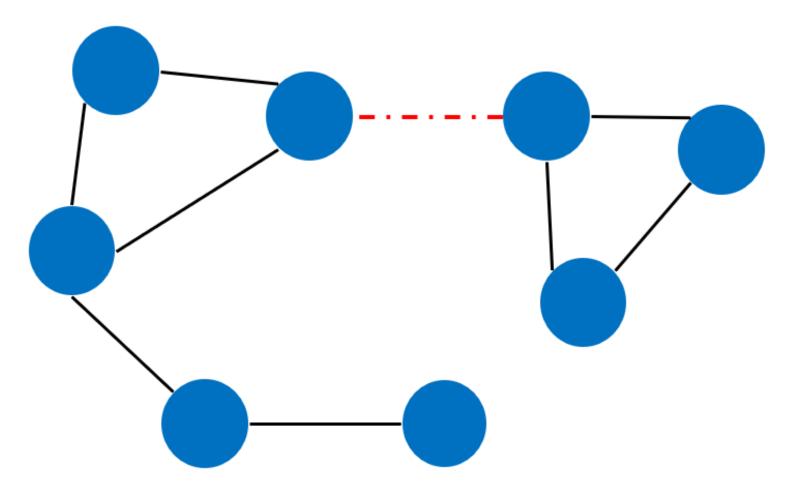
- Finding Bridges
- Finding Articulation points

## Part 2: GRAPH'S ALGORITHMS

#### Tarjan's Algorithm

#### **Problem Analysis**

**Subproblem:** given an edge in the graph, determine if it is a bridge or not



Counting number of connected components before and after removing an edge  $\longrightarrow$  O(|V|+|E|)



**Algorithm** 

- Finding Bridges
- Finding Articulation points

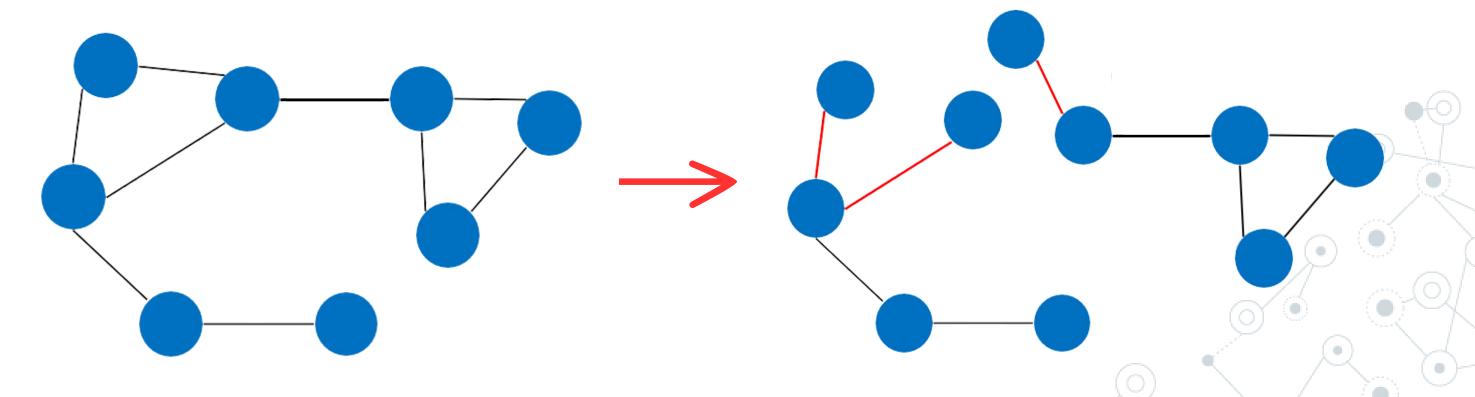
## Part 2: GRAPH'S ALGORITHMS

#### Tarjan's Algorithm

#### Problem Analysis

#### Better approach?

- Optimization problem: Greedy Approach X
- Independent subproblems: Backtracking
- ⇒ Non bridges may becomes bridges.



- → Backtracking X
  - Optimal Structures?



#### **Algorithm**

**Problem Analysis** 

Finding Articulation points

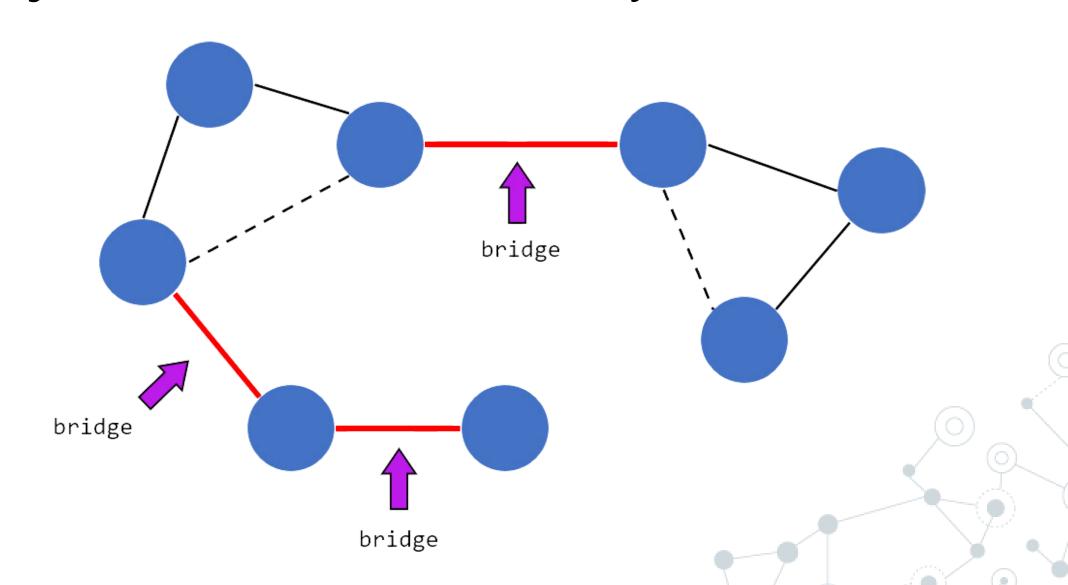
## Part 2: GRAPH'S ALGORITHMS

#### Tarjan's Algorithm

## Finding Bridges

On the DFS tree, vertex u is the parent of vertex v:  $\Rightarrow$  edge (u, v) is a **bridge** if and only if v **doesn't** 

have any other links to u or any ancestors of u.





#### **Algorithm**

**Problem Analysis** 

Finding Articulation points

## Part 2: GRAPH'S ALGORITHMS

#### Tarjan's Algorithm

## Finding Bridges

disc: 4

lowlink: 0

disc: 2

lowlink: 2

disc: 5

lowlink: 5

disc: 6

disc: 7

lowlink: 5

lowlink: 5

Giving each vertex an id showing the time we have discovered it

⇒ we can find the earliest disc: 0 vertex that a certain vertex can link to using the following formulas:

lowlink[u] = disc[u] for each neighbor v of u:

lowlink[u] = min(disc[v], lowlink[u]) if v is visited and not parent lowlink[u] = min(lowlink[v], lowlink[v]) if v is not visited

disc: 1

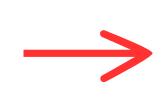
lowlink: 1

disc: 3

start

lowlink: 0

lowlink: 3



Dynamic programming

Time complexity: O(|V|+|E|)



#### Algorithm

Problem Analysis

Finding Articulation points

#### Part 2: GRAPH'S ALGORITHMS

#### Tarjan's Algorithm

#### Finding Bridges

#### Pseudocode:

```
function bridges():
  visited <- a list of size |V| with all Ø
  disc <- a list of size |V| with all -1
  low_link <- a list of size |V| with all -1</pre>
  parent <- a list of size |V| with all -1
  current <- Ø
  bridge_count <- Ø</pre>
  for each vertex v in the graph:
    if not visted[u]:
      bridge_helper(u)
  return bridge_count
```



#### Algorithm

Problem Analysis

Finding Articulation points

#### Part 2: GRAPH'S ALGORITHMS

#### Tarjan's Algorithm

## Finding Bridges

```
function bridge_helper(u):
  visited[u] <- 1</pre>
  disc[u] <- current</pre>
  low_link[u] <- current</pre>
  current <- current + 1
  for each adjacent vertex v of u:
    if not visited[v]:
      parent[v] <- u</pre>
      bridge_helper(v)
      low_link[u] <- min(low_link[u], low_link[v])</pre>
      if low_link[v] > disc[u]:
         bridge_count <- bridge_count + 1</pre>
    else if parent[u] != v
      low_link[u] <- min(low_link[u], disc[v])</pre>
```



#### Algorithm

- Problem Analysis
- Finding Bridges

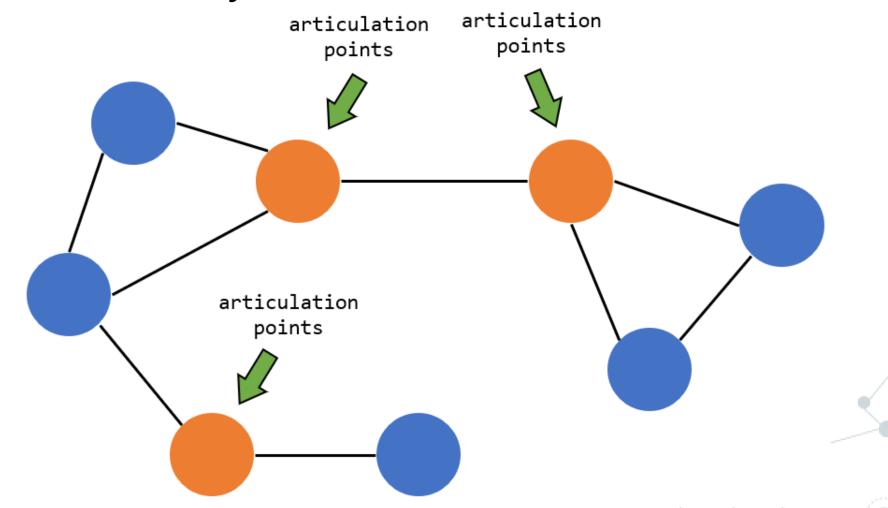
#### Part 2: GRAPH'S ALGORITHMS

## Tarjan's Algorithm Finding Articulation points

Same idea as finding articulation points.

A vertex u is an articulation point if and only if:

- u is the root and have more than 1 child
- u is not the root and a child of u does not have any other links to any ancestors of u





#### Algorithm

- Problem Analysis
- Finding Bridges

#### Part 2: GRAPH'S ALGORITHMS

## Tarjan's Algorithm Finding Articulation points

#### Pseudocode:

```
function ap():
  visited <- a list of size |V| with all Ø
  disc <- a list of size |V| with all -1
  low_link <- a list of size |V| with all -1
  parent <- a list of size |V| with all -1
  current <- Ø
  ap_count <- Ø
  branches <- Ø
  root <- -1
  for each vertex u in the graph:
    if not visited[u]:
      branches <- Ø
      root <- u
      ap_helper(u)
  return ap_count
```



#### **Algorithm**

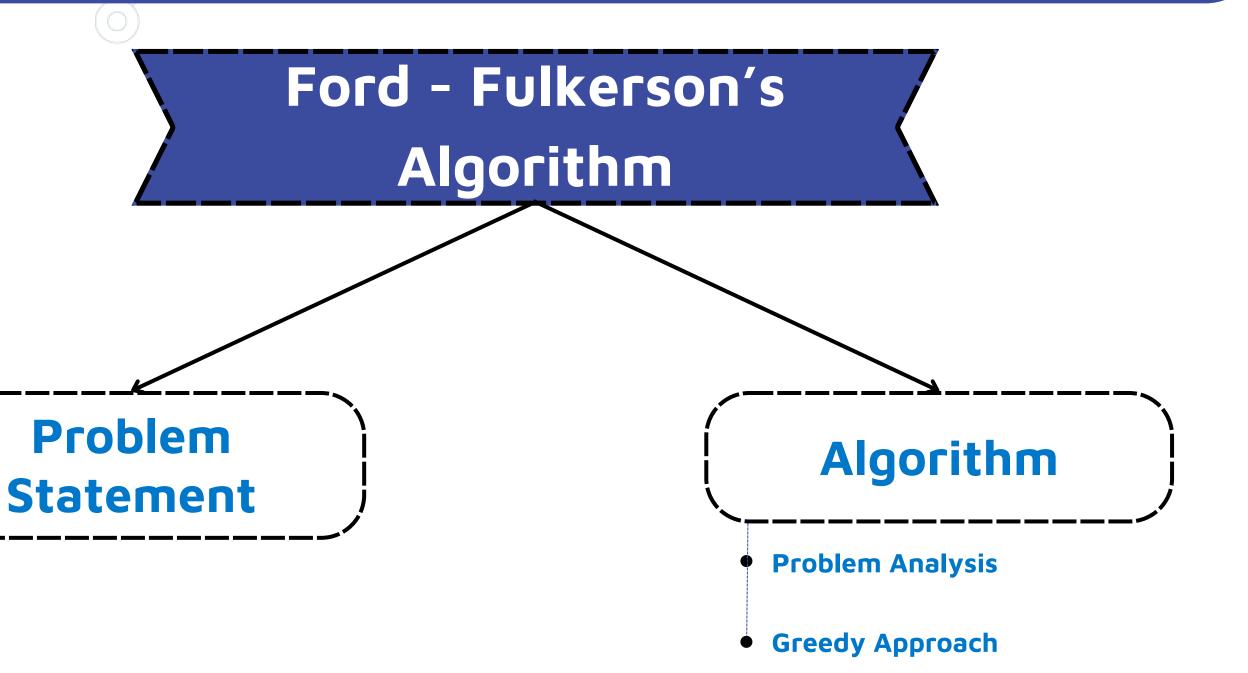
- Problem Analysis
- Finding Bridges

#### Part 2: GRAPH'S ALGORITHMS

## Tarjan's Algorithm Finding Articulation points

```
function ap_helper(u):
  visited[u] <- 1</pre>
  disc[u] <- current</pre>
  low_link[u] <- current</pre>
  current <- current + 1
  is_ap <- Ø
  for each adjacent vertex v of u:
    if not visited[v]:
      parent[v] <- u</pre>
      ap_helper(v)
      if u = root:
        branches <- branches + 1
         if branches > 1:
           is_ap = 1
      else:
         low_link[u] <- min(low_link[u], low_link[v])</pre>
         if low_link[v] >= disc[u]:
           is_ap = 1
    else if parent[u] != v
      low_link[u] <- min(low_link[u], disc[v])</pre>
  ap_count <- ap_count + is_ap</pre>
```







**Algorithm** 

Greedy Approach

## Part 2: GRAPH'S ALGORITHMS

#### Ford - Fulkerson's Algorithm

#### **Problem Analysis**

capacity

0/9

0/10

0/10

sink

0/6

Given a pipe network with a source, a sink and capacities of each pipe. Find the maximum amount of water can be sent from source to sink so that:

- total water flowing from source = total water flowing to sink
- total water flowing to v = total water flowing from v (v is not source or sink)
- water flowing through a pipe ≤ capacity of that pipe

Input: a positive-weighted directed graph
Output: max flow value (mfv)
Constraint: mfv ≤ 500, |V|≤500

source S 0/2



Algorithm

Greedy Approach

## Part 2: GRAPH'S ALGORITHMS

## Ford - Fulkerson's Algorithm

#### **Problem Analysis**

#### Brute force approach

Let C be the combination of all possible flow value for each edges.

• The number of combinations is the product of all capacities.

To check if C is valid  $\Rightarrow$  O(|V|+|E|).

The total complexity: O((|V|+|E|)\*product of capacities)

#### Backtracking approach

We may not be able to check for validity after each insertion



## Ford - Fulkerson's Algorithm

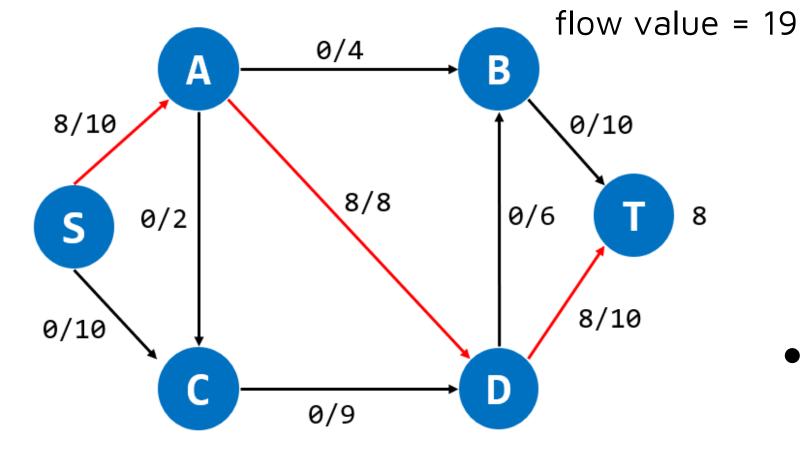
#### Greedy Approach

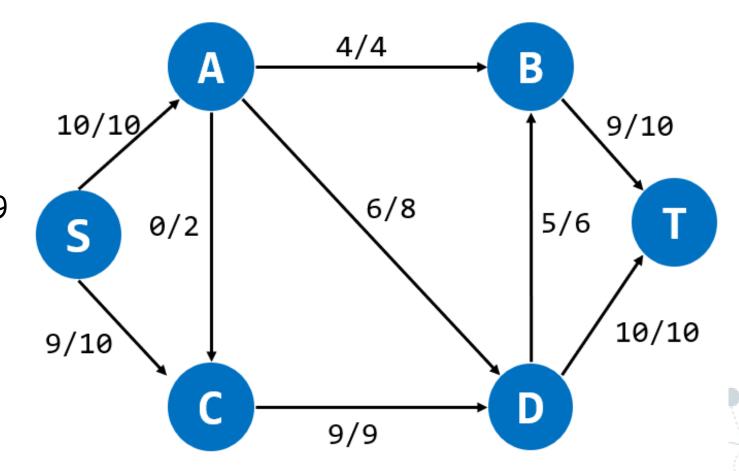
Problem Statement

**Algorithm** 

Problem Analysis

 We can continuingly add water into the network until we cannot do so anymore.





 It is proved that at that point, a maximum flow is found.

At each iteration, we just need to find a way from source to sink and add the amount of water equal to the bottleneck. However, there is a problem...



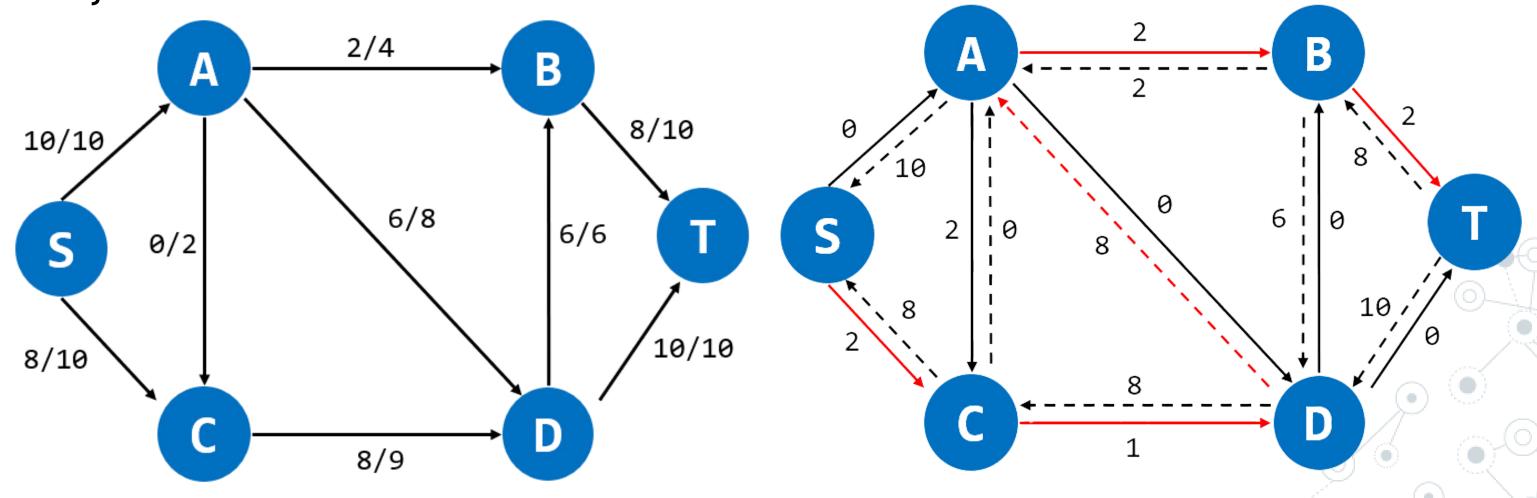
## Ford - Fulkerson's Algorithm Greedy Approach

**Problem Statement** 

**Algorithm** 

**Problem Analysis** 

The current flow has "blocked" any water to be sent. How can we adjust it?



introducing reverse edges (which tell us the number of subtances currently in the original edges), we can let some water to flow into another pipe.



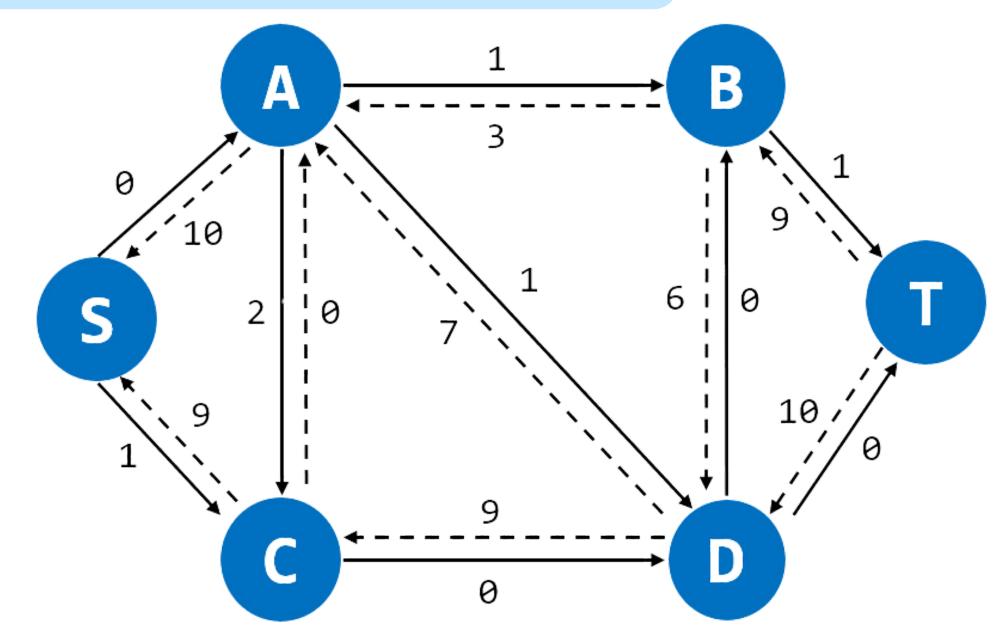
## Ford - Fulkerson's Algorithm

#### **Greedy Approach**

Problem Statement

**Algorithm** 

Problem Analysis



In the worst case: we have to traverse the whole network and can only add 1 more unit of water to the flow.

Time complexity: **O(mfv\*(|V|+|E|)** 



## Ford - Fulkerson's Algorithm Greedy Approach

**Problem Statement** 

Algorithm

Problem Analysis

```
for each edge (u, v) of the graph:
  add edge (v, u) of capacity Ø to the graph
function ford_fulkerson():
  flow_value <- Ø</pre>
 visited <- a list of size |V| with all Ø
  parent <- a list of size |V| with all -1
 bottleneck <- infinity</pre>
  while find_aug_path(source): # find a way from source to sink
    s <- sink
    while s != source:
      graph[parent[s]][s] <- graph[parent[s]][s] - bottleneck</pre>
      graph[s][parent[s]] <- graph[s][parent[s]] + bottleneck</pre>
    s <- parent[s]
    flow_value += bottleneck
    visited <- a list of size |V| with all Ø
    parent <- a list of size |V| with all -1
 return max_flow
```

Pseudocode:

