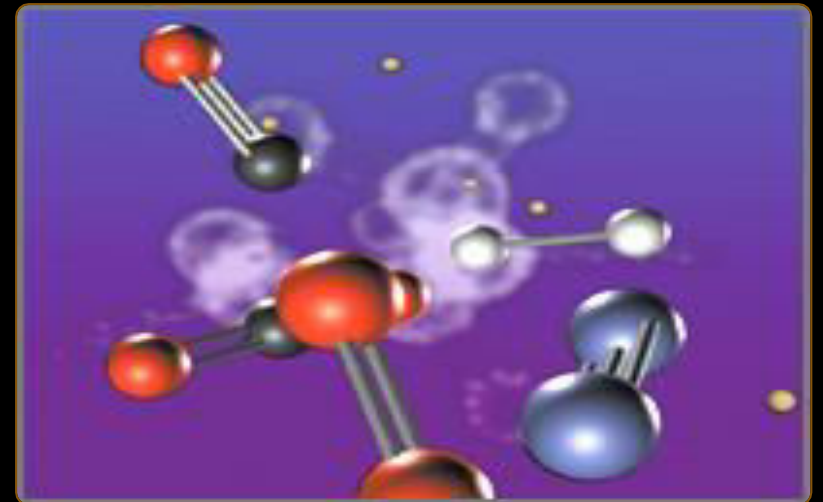


# Advanced Graph Algorithms II

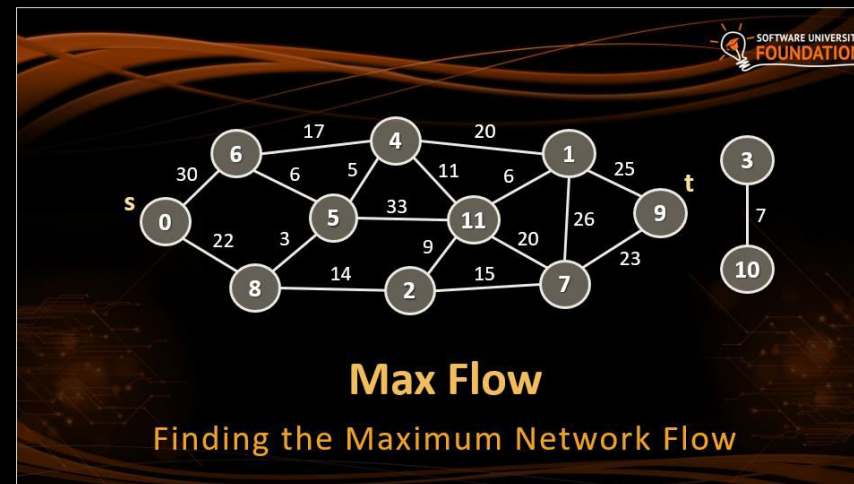
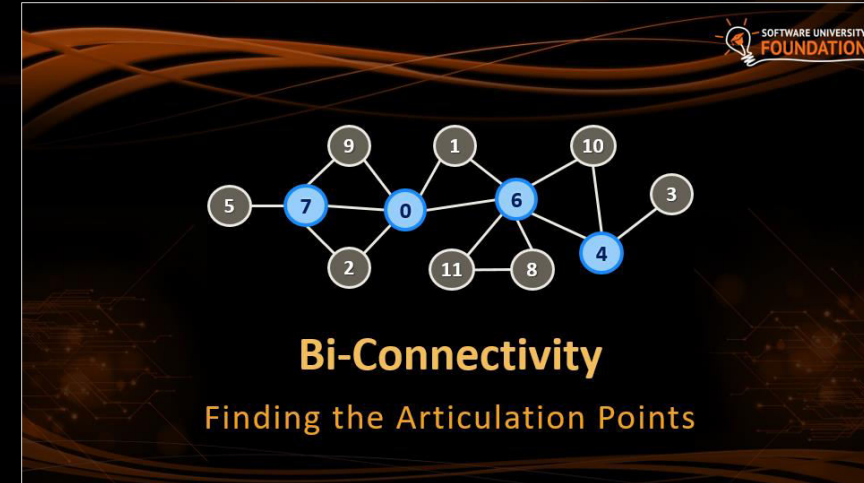
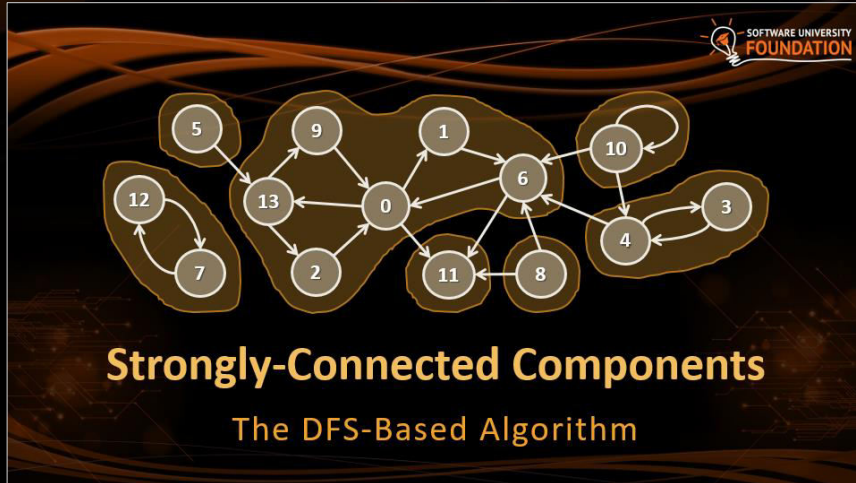
## Strongly Connected Components, Bi-Connectivity, Max Flow



**SoftUni Team**  
Technical Trainers  
Software University  
<http://softuni.bg>



# Table of Contents

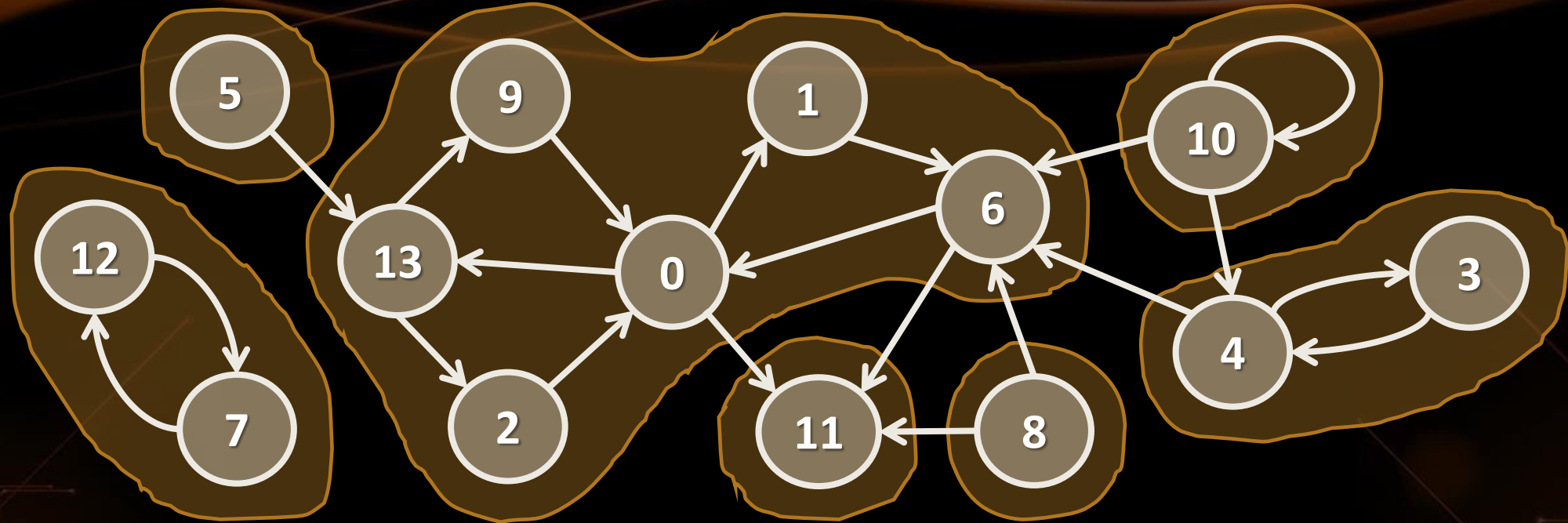


Have a Question?

**sli.do**

**#DsAlgo**



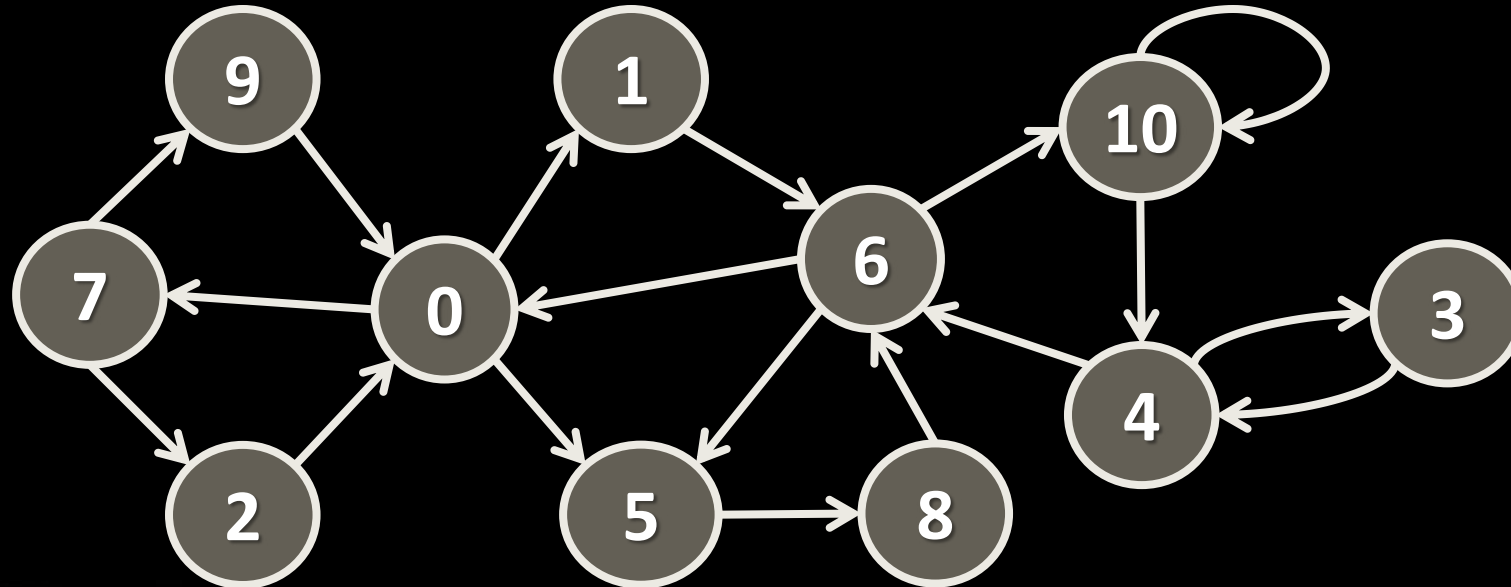


# Strongly-Connected Components

The DFS-Based Algorithm

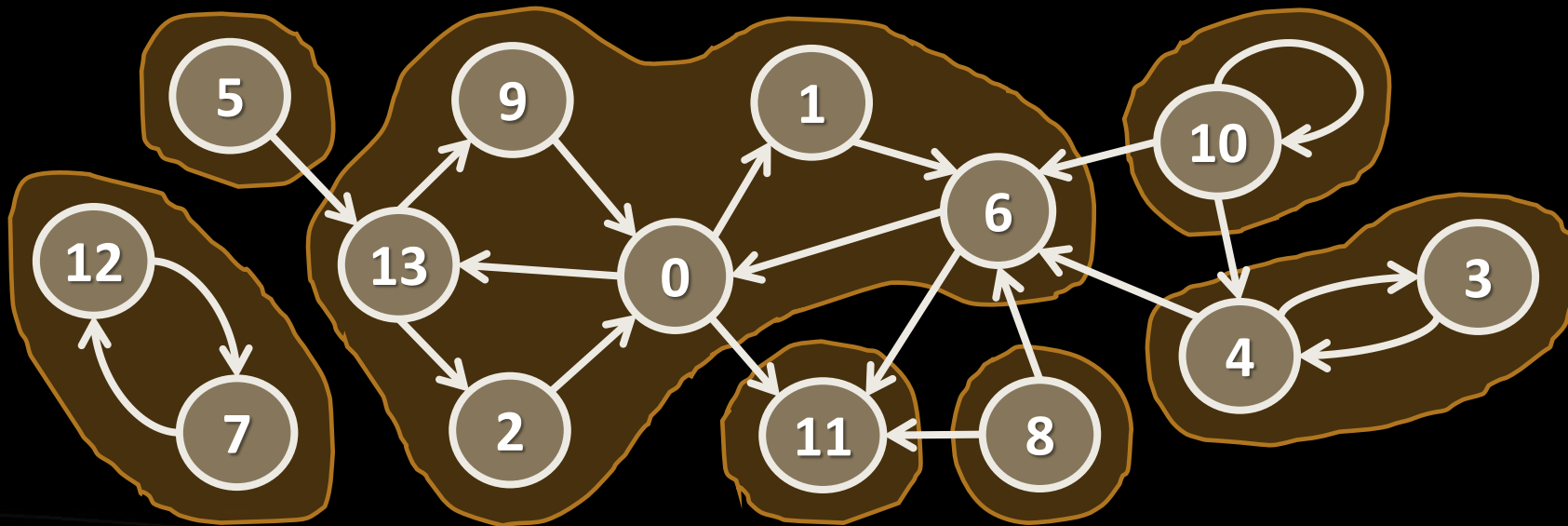
# Strongly-Connected Components

- A directed graph is **strongly-connected**
  - When every two vertices are connected by path
- Example of strongly-connected graph:



# Strongly-Connected Components

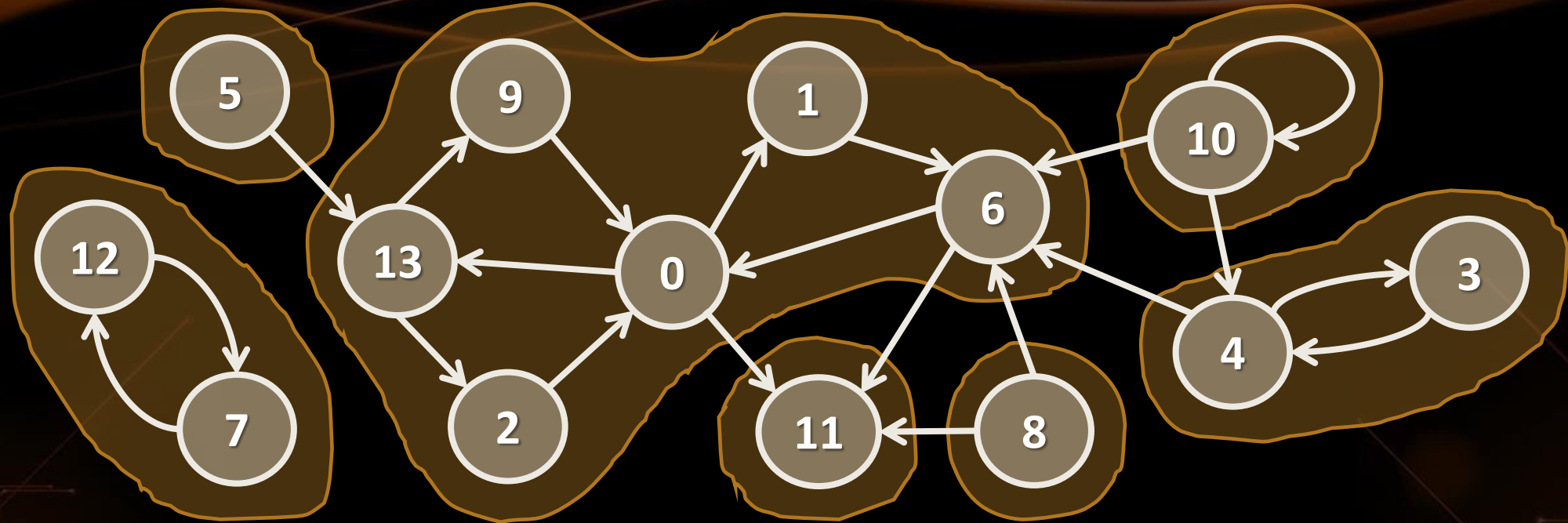
- **Strongly-connected component** is a maximal strongly-connected subgraph (component with paths between any two nodes)
- A directed graph can be decomposed into strongly-connected components, e.g.



# Kosaraju–Sharir Algorithm

- Let **G** be a directed graph and **S** be an empty stack
- For each vertex **v** in **G**
  - Call **DFS(v)** to traverse the graph (visit each node once)
  - Each time **DFS(v)** finishes (before recursive return), push **v** onto **S**
- Build the **reverse graph G'** (reverse all edges from **G**)
- While **S** is non-empty:
  - Pop the top vertex **v** from **S**
  - if **v** is not visited, call **ReverseDFS(v)** to find the next strongly-connected component

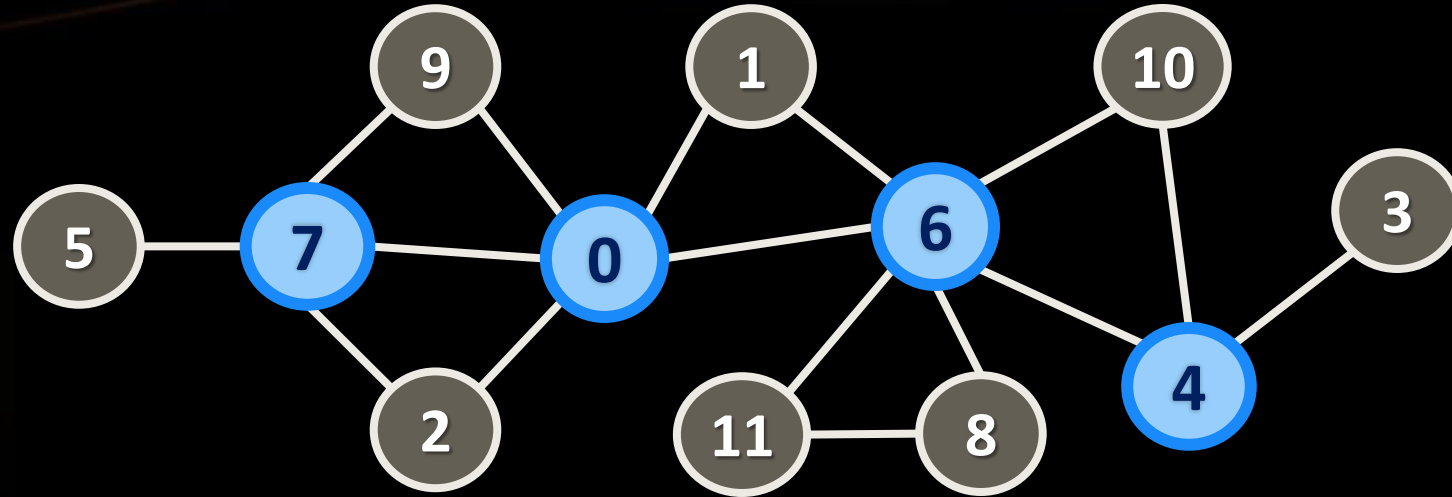




# Kosaraju–Sharir Algorithm

Lab



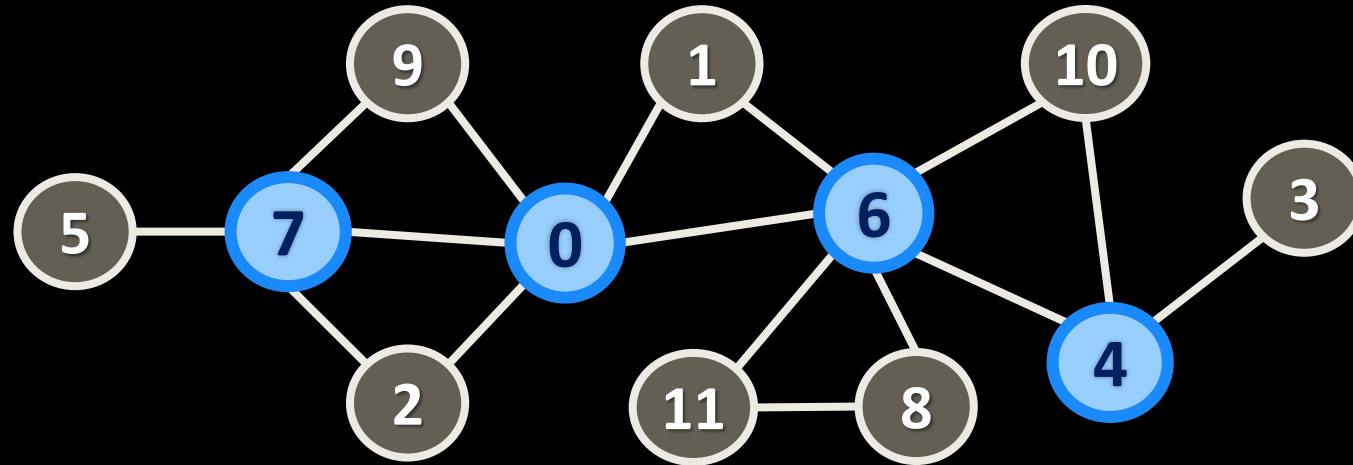


# Bi-Connectivity

## Finding the Articulation Points

# Articulation Points

- In a connected undirected graph an **articulation point** is a node that when removed, splits the graph into several components
- The blue nodes below are the **articulation points**: 7, 0, 6, 4



- There are 6 **bi-connected components**:
  - {5, 7}, {0, 2, 7, 9}, {1, 0, 6}, {6, 8, 11}, {4, 6, 10}, {3, 4}

# Articulation Points – The Slow Algorithm

- The straightforward algorithm –  $O(n * (m + n))$ 
  - Remove each node and check whether the graph stays connected

```
foreach  $v \in$  graph nodes
  temporary remove  $v$ 
  check for connectivity with DFS( $u$ ), where  $u \neq v$ 
  if the graph is not connected, print  $v$ 
  restore  $v$  back in the graph
```

- Check for graph connectivity by DFS traversal + counting the number of visited nodes

# Articulation Points – The Fast Algorithm

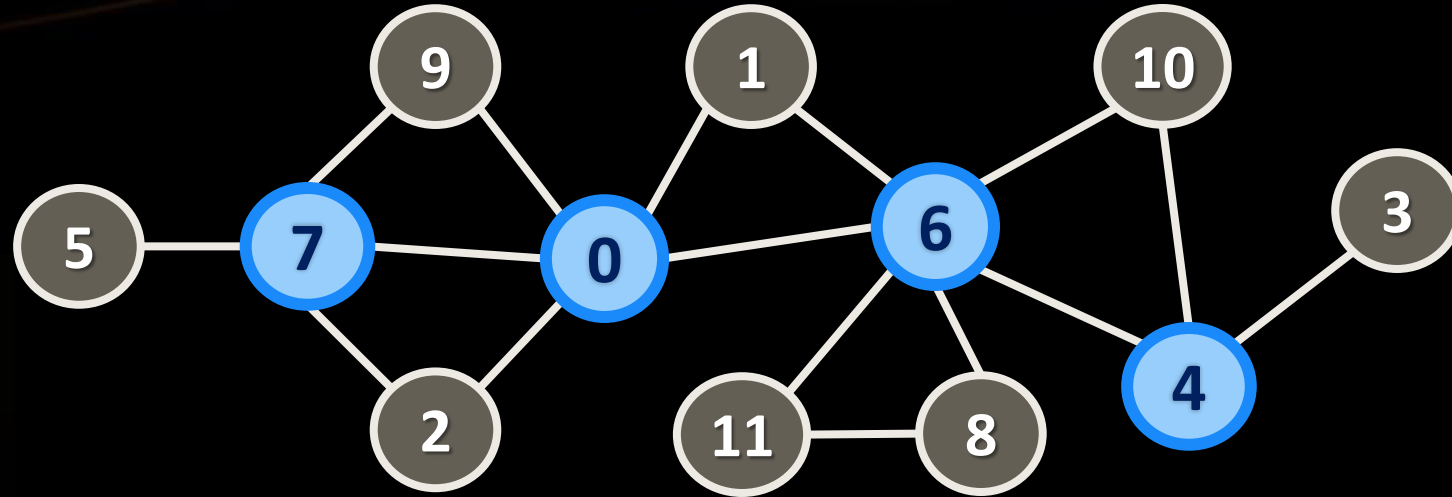
- The fast (linear) algorithm for finding articulation points is based on DFS traversal with some computations (Hopcroft, Tarjan, 1973)
- Run a DFS while maintaining the following information:
  - The  $\text{depth}(v)$  of each vertex in the DFS tree (once it gets visited)
  - For each vertex  $v$ , the lowest depth of neighbors of all descendants of  $v$  in the DFS tree, called the  $\text{lowpoint}(v)$
- A non-root vertex  $v$  is an articulation point if there is a child  $y$  of  $v$  such that  $\text{lowpoint}(y) \geq \text{depth}(v)$
- The root vertex is an articulation point if it has at least two children



# Articulation Points – Fast Algorithm

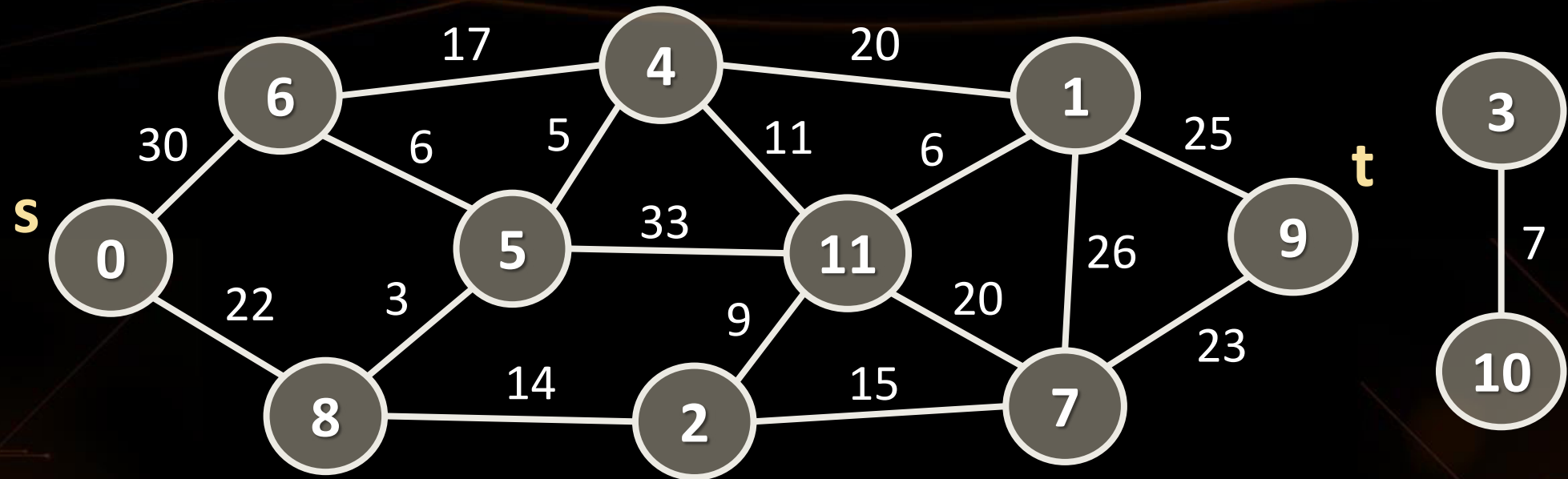
```
FindArticulationPoints(node, d)
    visited[node] = true
    depth[node] = d
    lowpoint[node] = d
    childCount = 0
    isArticulation = false
    for each childNode in childNodes[node]
        if not visited[childNode]
            parent[childNode] = node
            FindArticulationPoints(childNode, d + 1)
            childCount = childCount + 1
            if lowpoint[childNode] >= depth[node]
                isArticulation = true
            lowpoint[node] = Min(lowpoint[node], lowpoint[childNode])
        else if childNode <> parent[node]
            lowpoint[node] = Min(lowpoint[node], depth[childNode])
    if (parent[node] <> null and isArticulation) or (parent[node] == null and childCount > 1)
        print node as articulation point
```

Time complexity:  $O(N + M)$



# Finding the Articulation Points

Lab



# Max Flow

Finding the Maximum Network Flow

# Max Flow Problem

- Weighted directed / undirected graph
  - With capacities assigned to the edges  $c(u \rightarrow v)$
- Goal: compute the maximum flow from node  $s$  to node  $t$
- Each edge  $\{u \rightarrow v\}$  holds certain flow  $\text{flow}(u \rightarrow v)$ 
  - $\text{flow}(u \rightarrow v) \leq c(u \rightarrow v)$
- For each node input flow == output flow
- Max flow ==  $\text{sum}(\text{flow}(s \rightarrow v)) = \text{sum}(\text{flow}(u \rightarrow t))$ 
  - Total output flow from  $s$  == total input from to  $t$



# Ford-Fulkerson Max-Flow Algorithm

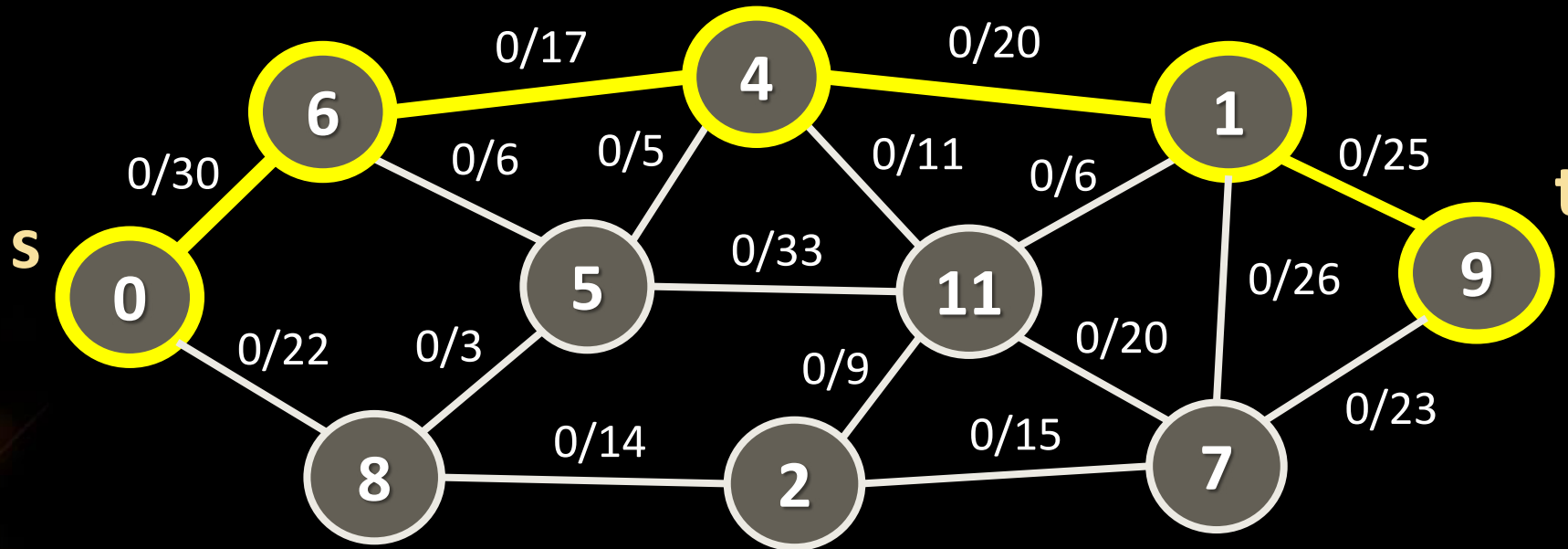
- Ford-Fulkerson can be summarized in the following steps:
  1. Start from zero flow:  $\text{flow}(u \rightarrow v) = 0$  for each edge
  2. While possible:
    - Find **augmenting path**  $p$  such that
      - $p$  is a valid path from  $s$  to  $t$  and for each edge  $\{u \rightarrow v\} \in p \rightarrow c(u \rightarrow v) > 0$
      - The  $\text{flow}(p)$  for the augmenting path  $p$  in  $s \rightarrow t$  is the minimum capacity  $c$  of each edge  $\{u \rightarrow v\}$  in the path  $s \rightarrow t$
      - Modify the capacities of the edges in the path  $p$ :
        - For each edge  $\{u \rightarrow v\} \in p \rightarrow c(u \rightarrow v) = c(u \rightarrow v) - \text{flow}(p)$
        - For each edge  $\{u \rightarrow v\} \in p \rightarrow c(v \rightarrow u) = c(v \rightarrow u) + \text{flow}(p)$
    - Add  $\text{flow}(p)$  to the maximum flow

# Edmonds-Karp Max Flow Algorithm

- If we use **Breadth-first Search** to Find the augmenting path we get the **Edmonds-Karp** algorithm
  1. Start from zero flow:  $\text{flow}(u \rightarrow v) = 0$  for each edge
  2. While possible:
    - Find an **augmenting path** **p** from **s** to **t** using BFS such that:
      - For each  $\{u \rightarrow v\} \in \mathbf{p} \rightarrow c(u \rightarrow v) > 0$
    - Keep track of the parent for each visited vertex
    - Reconstruct the path **p** using the parents
      - Set **flow(p)** as the smallest capacity **c** in the path **p**
      - Modify the capacities of the edges in the path **p** as in Ford-Fulkerson:
    - Add **flow(p)** to the maximum flow

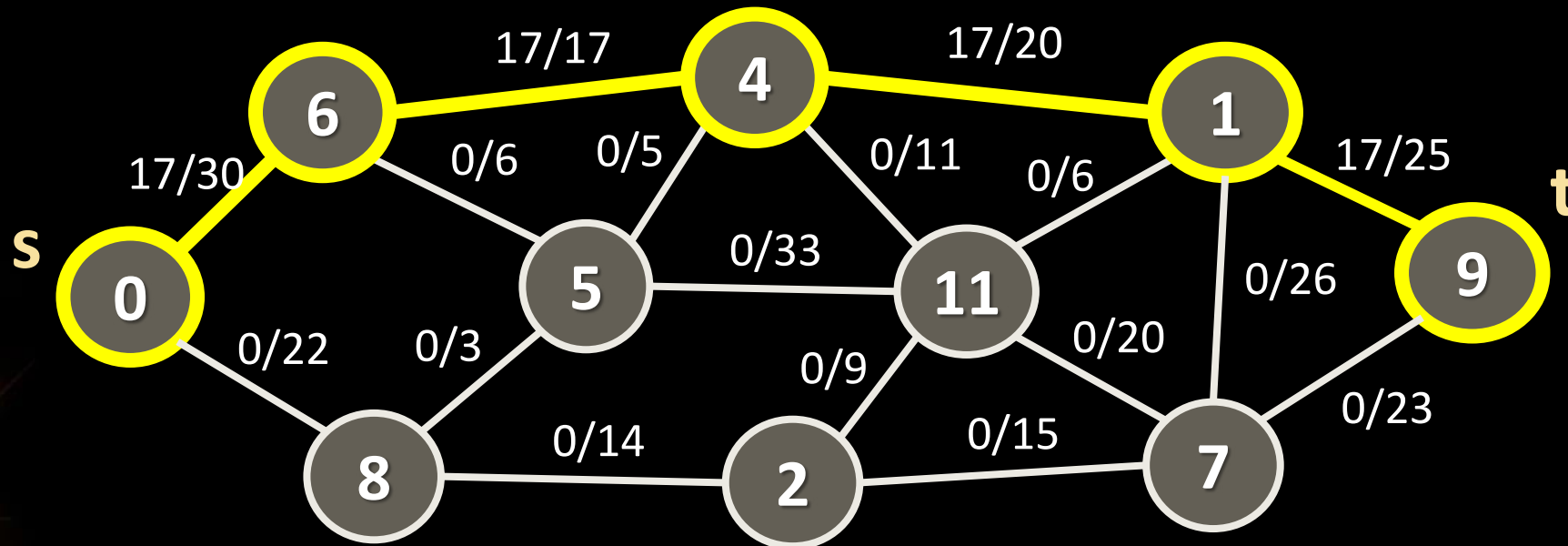
# Edmonds-Karp – Step #1

- Start from empty flows through all edges
  - Find an augmenting path:  $0 \rightarrow 6 \rightarrow 4 \rightarrow 1 \rightarrow 9$  (increment = 17)
- Time complexity:  $O(VE^2)$



## Edmonds-Karp – Step #2

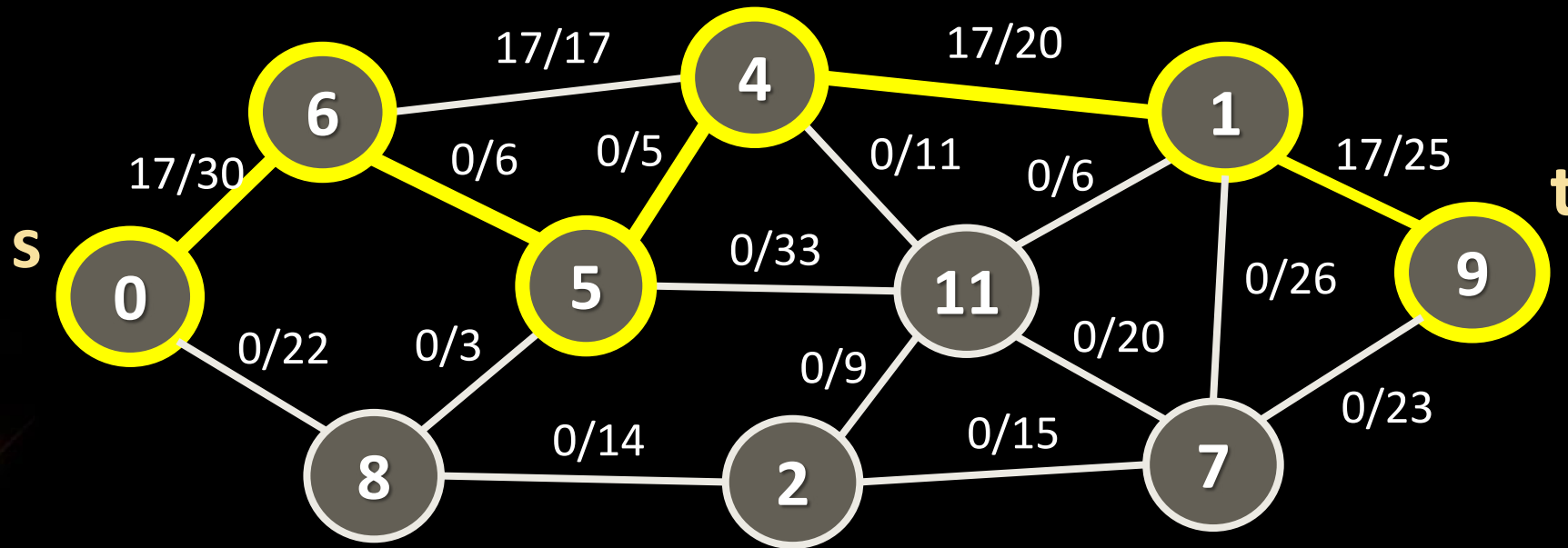
- Augment the flow through the path:  $0 \rightarrow 6 \rightarrow 4 \rightarrow 1 \rightarrow 9$  (increment = 17)
- Current max flow = 17





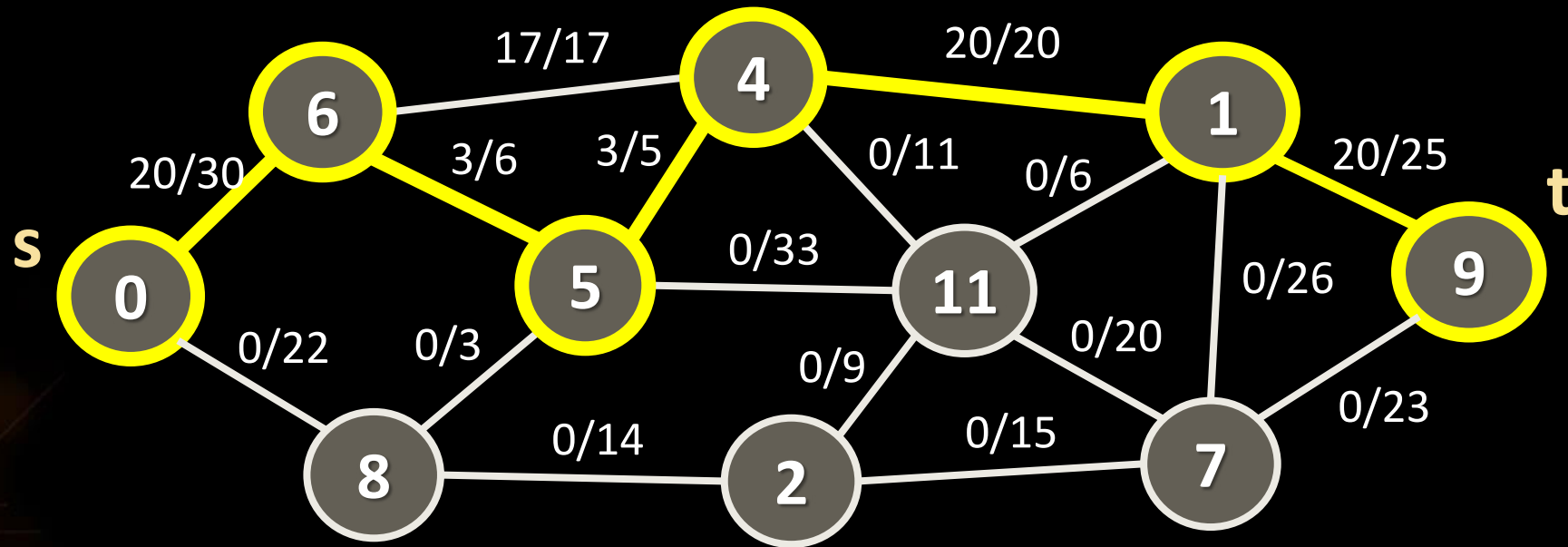
# Edmonds-Karp – Step #3

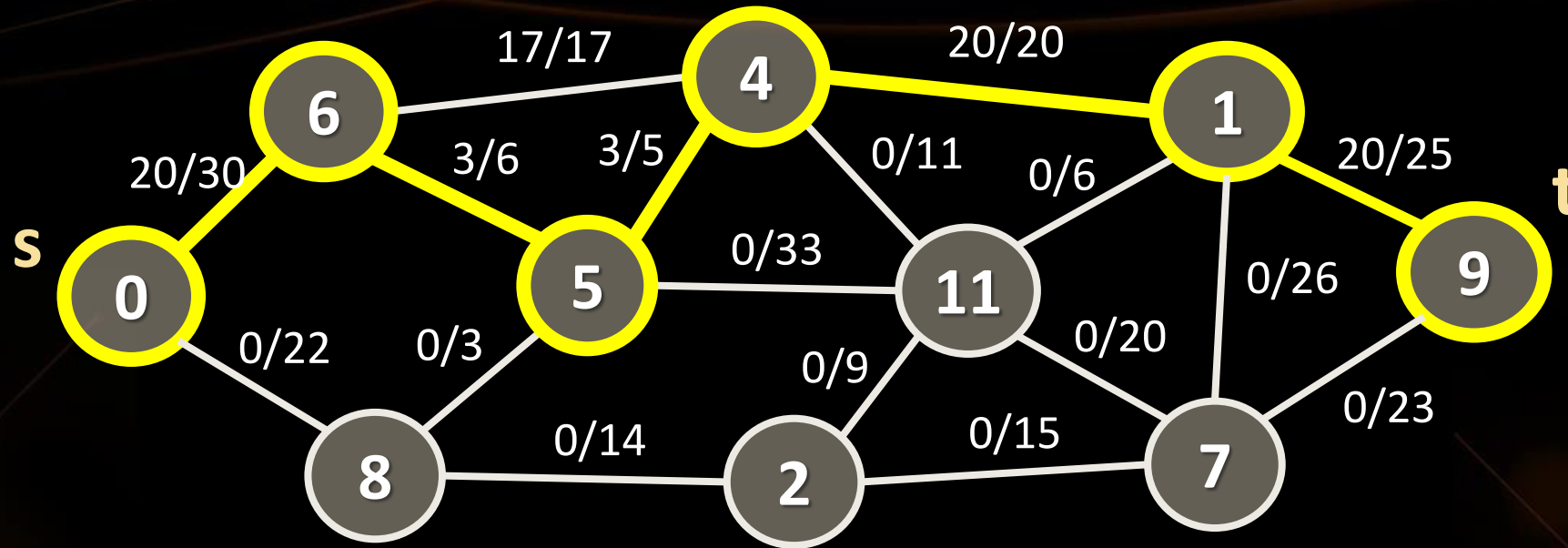
- Find an augmenting path:  $0 \rightarrow 6 \rightarrow 5 \rightarrow 4 \rightarrow 1 \rightarrow 9$   
(increment = 3)



# Edmonds-Karp – Step #4

- Augment the flow through the path:  $0 \rightarrow 6 \rightarrow 5 \rightarrow 4 \rightarrow 1 \rightarrow 9$  (increment = 3)
- Current max flow = 20





# Edmonds-Karp Algorithm

## Lab

# Terminology

- **Level Graph  $G_L$  :**

- A tree built by using BFS from the **source  $s$**
- Each node in the tree is assigned a value representing its distance from the start
- The distance is the number of edges from the **source  $s$**

- **Blocking flow:**

- Combination of all **augmenting paths** that can be built on  $G_L$
- Since  $G_L$  is created using BFS, any path  **$p$**  from  **$s \rightarrow t$**  is going to have the least amount of edges possible



# Dinic/Dinitz Max-Flow Algorithm

- Dinitz can be summarized in the following steps:
  1. Start from zero max flow:  $\mathbf{m} = 0$
  2. Construct the level Graph  $\mathbf{G}_L$  using BFS
    - If  $\mathbf{t}$  is **unreachable** in  $\mathbf{G}_L \rightarrow$  return the max flow  $\mathbf{m}$
  3. Find **blocking flow**  $\mathbf{f}$  in  $\mathbf{G}_L$  using DFS
  4. Add flow  $\mathbf{f}$  to the max flow  $\mathbf{m}$

# Dinitz Algorithm Pseudocode – 1/3

```
private static int[] childCounter;    // Track blocked edges
private static int[] bfsDist;        // Distances in the Level Graph
private static List<int>[] edges;    // adjacency list
private static int[][] capacities;
private static int endNode;

static int Dinic(int source, int destination)
{
    int result = 0;
    while (Bfs(source, destination))    // While we can find a path from source to sink
    {
        for (int i = 0; i < childCounter.Length; i++)
        {
            childCounter[i] = 0;        // Reset blocked edges on each Level Graph
        }

        int delta;
        do
        {
            delta = Dfs(source, int.MaxValue);    // Each delta is the flow from an augmenting path
            result += delta;
        }
        while (delta != 0);
    }
    return result;
}
```

Time complexity:  $O(V^2E)$

# Dinitz Algorithm Pseudocode – 2/3

```
static bool Bfs(int src, int dest)
{
    for (int i = 0; i < bfsDist.Length; i++)
    {
        bfsDist[i] = -1;          // Reset distances in Level Graph
    }
    bfsDist[src] = 0;
    Queue<int> queue = new Queue<int>();
    queue.Enqueue(src);
    while (queue.Count > 0)
    {
        int currentNode = queue.Dequeue();
        for (int i = 0; i < edges[currentNode].Count; i++)
        {
            int child = edges[currentNode][i];
            if (bfsDist[child] < 0 && capacities[currentNode][child] > 0) // If node has not been visited
            {
                bfsDist[child] = bfsDist[currentNode] + 1;
                queue.Enqueue(child);
            }
        }
    }
    return bfsDist[dest] >= 0;    // If there is a path to the sink return true
}
```

Time complexity:  $O(V^2E)$

# Dinitz Algorithm Pseudocode – 3/3

```
static int Dfs(int source, int flow)
{
    if (source == endNode)    // If we reach the sink return the flow
    {
        return flow;
    }
    for (int i = childCounter[source]; i < edges[source].Count; i++, childCounter[source]++)
    {
        int child = edges[source][i];
        if (capacities[source][child] <= 0) continue;    // If the edge has no more room skip
        if (bfsDist[child] == bfsDist[source] + 1)    // Only check vertexes following the Level Graph
        {
            int augmentationPathFlow = Dfs(child, Math.Min(flow, capacities[source][child]));
            if (augmentationPathFlow > 0)
            {
                capacities[source][child] -= augmentationPathFlow;    // Fix capacities
                capacities[child][source] += augmentationPathFlow;
                return augmentationPathFlow;
            }
        }
    }
    return 0;    // If no path is found return 0 - path is blocked
}
```

Time complexity:  $O(V^2E)$



# Summary

1. Strongly-connected components
  - Use DFS + reverse DFS algorithm
2. Articulation points → use modified DFS
3. Maximum flow:
  - Ford-Fulkerson
  - Edmonds-Karp
  - Dinitz



# Advanced Graph Algorithms II



## Questions?



# License

- This course (slides, examples, labs, videos, homework, etc.) is licensed under the "Creative Commons Attribution-NonCommercial-ShareAlike 4.0 International" license



- Attribution: this work may contain portions from
  - "Fundamentals of Computer Programming with C#" book by Svetlin Nakov & Co. under CC-BY-SA license
  - "Data Structures and Algorithms" course by Telerik Academy under CC-BY-NC-SA license

# Trainings @ Software University (SoftUni)

- Software University – High-Quality Education, Profession and Job for Software Developers
  - [softuni.bg](http://softuni.bg)
- Software University Foundation
  - <http://softuni.foundation/>
- Software University @ Facebook
  - [facebook.com/SoftwareUniversity](https://facebook.com/SoftwareUniversity)
- Software University Forums
  - [forum.softuni.bg](http://forum.softuni.bg)



**Software  
University**

