

# Lecture 8

## Graphs IV – Flow Problems and Matching

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Saarland University

June, 2020

# Overview

- ▶ Graphs I: Traversals and Shortest Paths
- ▶ Graphs II: DFS Applications and Friends
- ▶ Graphs III: Trees
- ▶ Graphs IV: Flow Problems and Matching

# Flow Problems

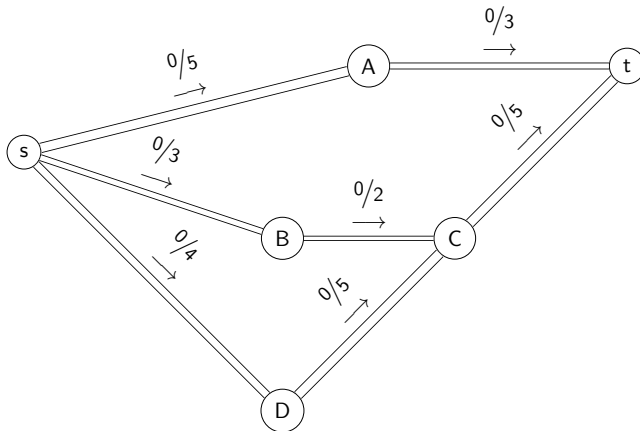
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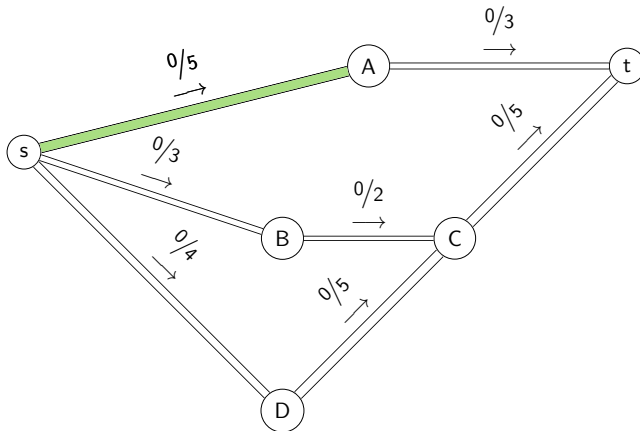
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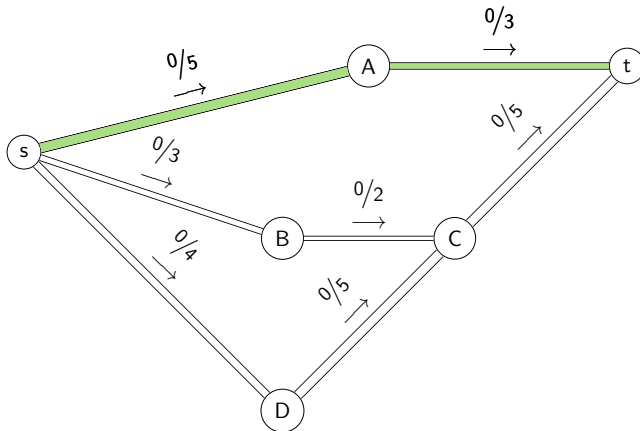
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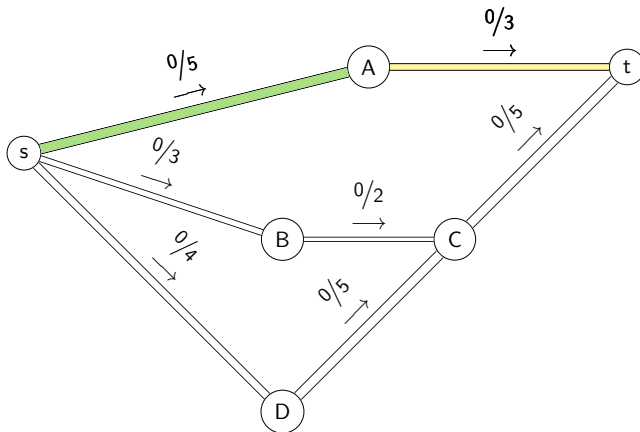
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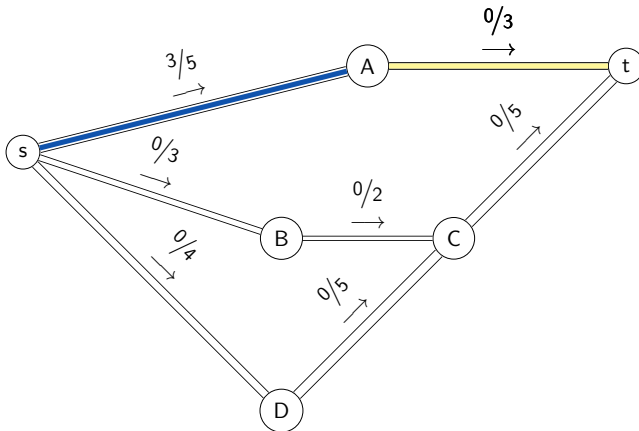
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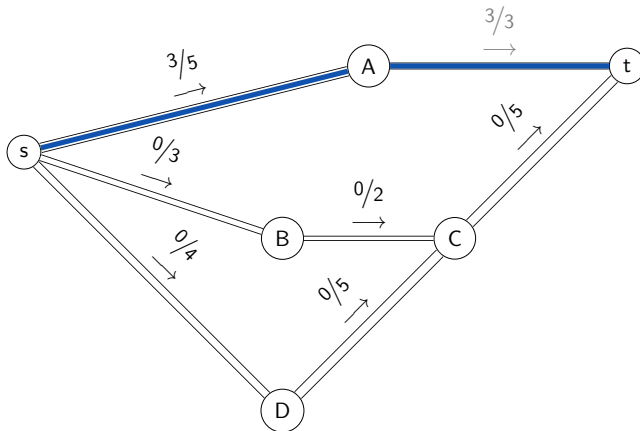




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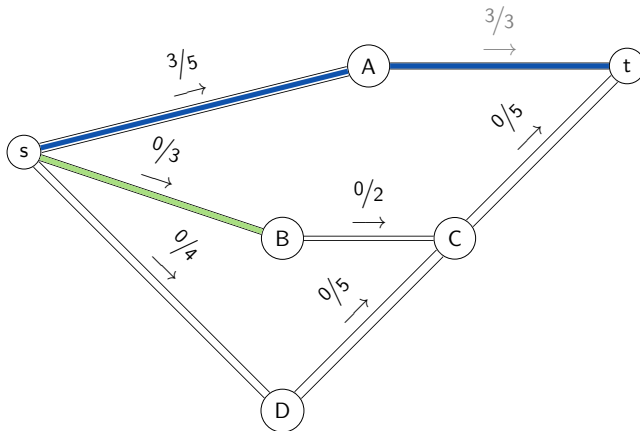
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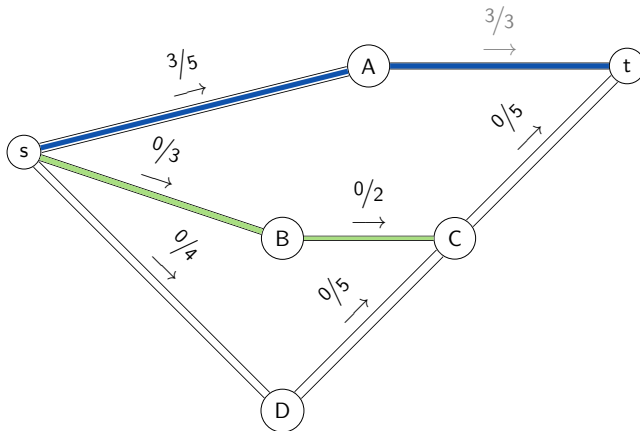
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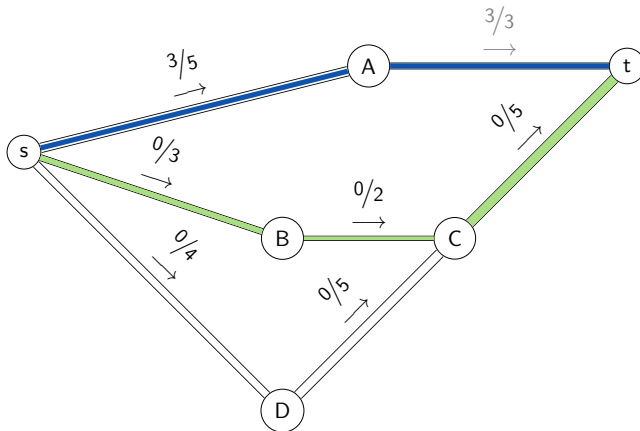
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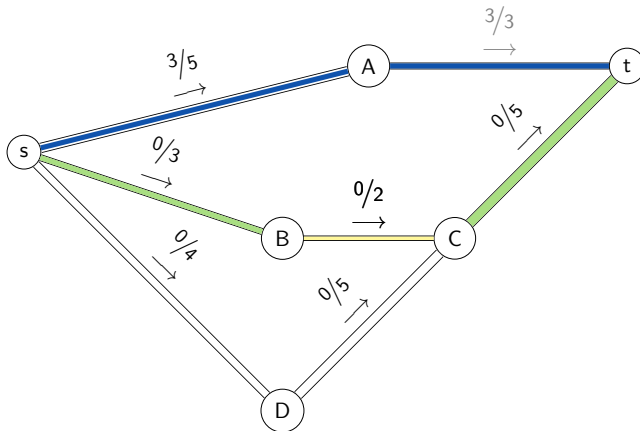
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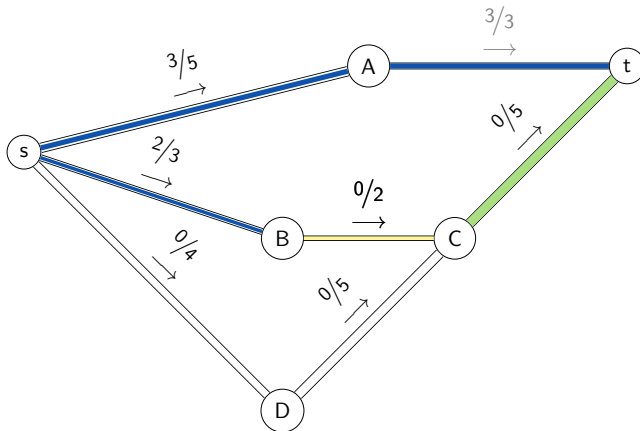
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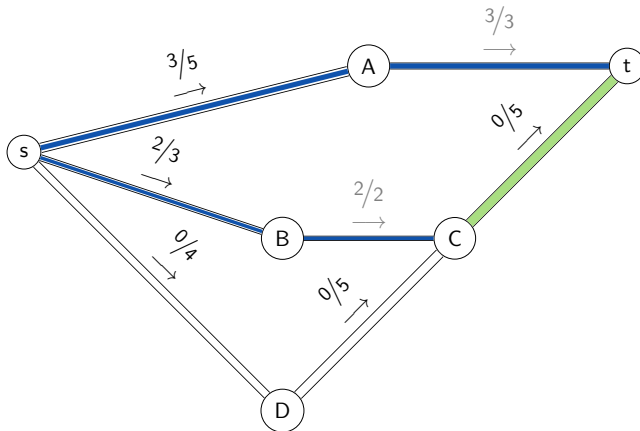
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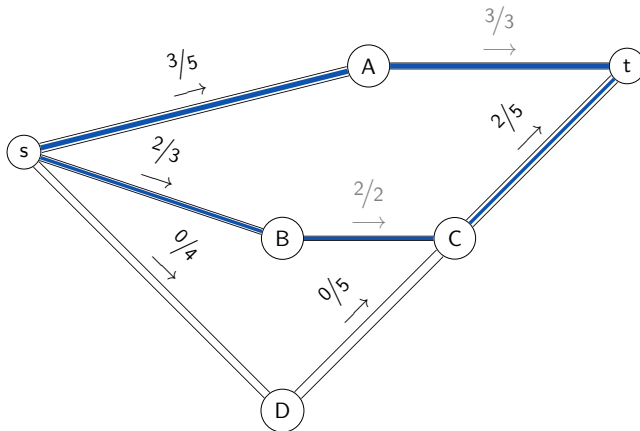
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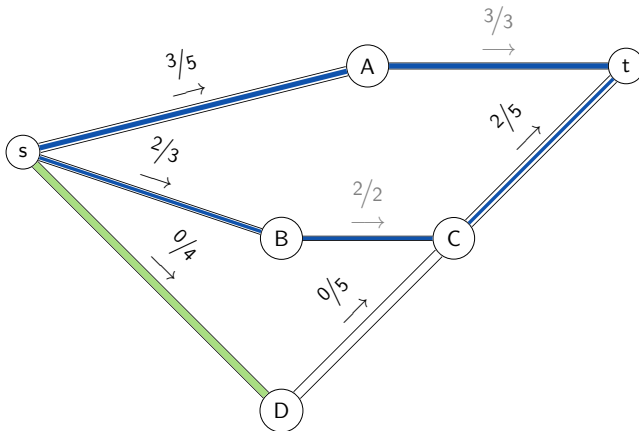




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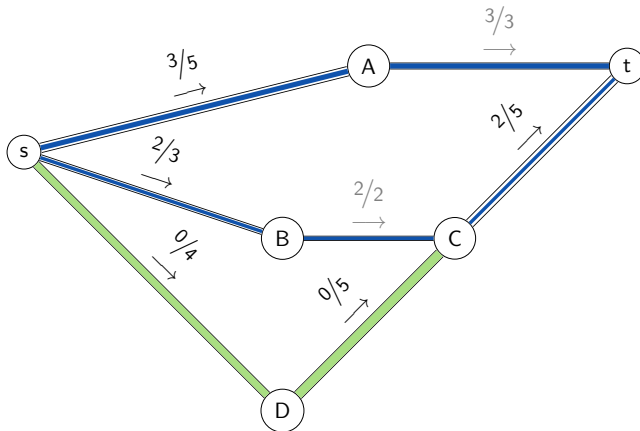
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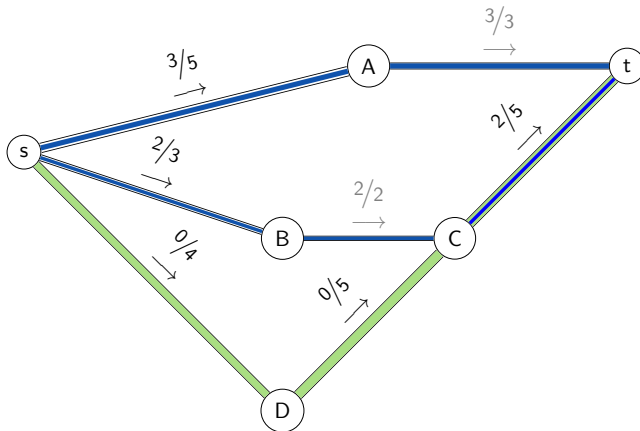
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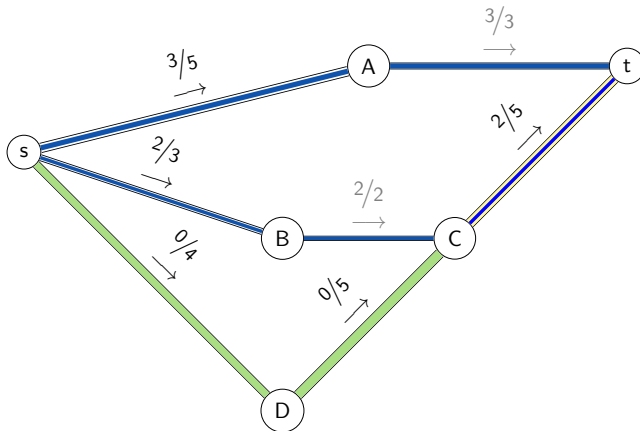
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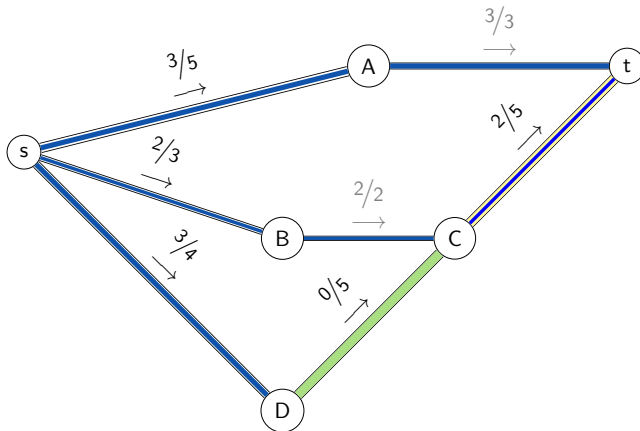
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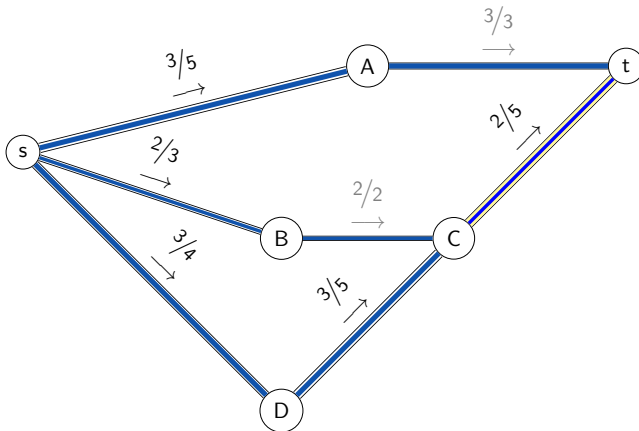
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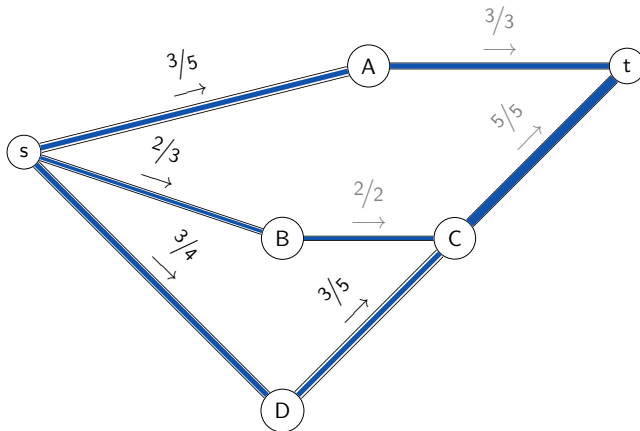
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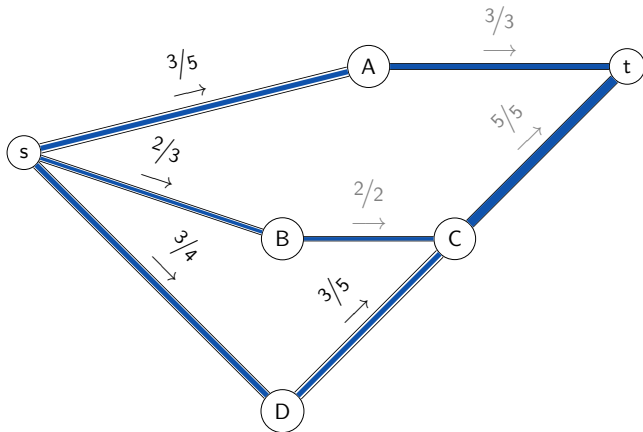
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# Flow Problems

## Solution: Water Flow

At most 8 units of water can flow from  $s$  to  $t$  simultaneously.





# Flow Problems

## Max Flow

### Working Definition: Maximum Flow

Given a directed Graph  $G$  with capacity  $c_e$  for every edge  $e$ , a source vertex  $s$ , and a sink vertex  $t$ , we call the maximum amount of *flow* (e.g. water) that can flow from  $s$  to  $t$  simultaneously the *Maximum Flow* of  $G$ .

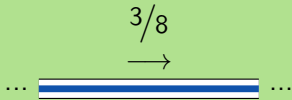
Flow is subject to the conditions that

- ▶ in every vertex, except in  $s$  and  $t$ , the in-flow equals the out-flow
- ▶ no edge can support more flow than its capacity

# Flow Problems

## Ford-Fulkerson

### Some Nomenclature

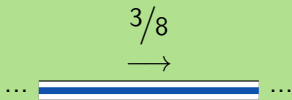


- ▶ edge capacity: 8
- ▶ current flow: 3
- ▶ residual capacity: 5

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## Ford-Fulkerson

### Some Nomenclature



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### Algorithm Idea: Ford-Fulkerson

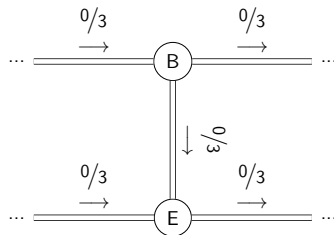
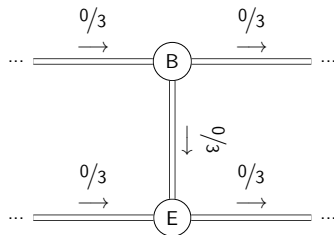
Find a path from  $s$  to  $t$  that can still take some water, a so-called *Augmenting Path*

- ▶ The *bottleneck capacity*  $c$  is the minimum residual capacity along this augmenting path
- ▶ Augment the current flow of each edge on the path by  $c$

The maximum flow equals the total flow added until no Augmenting Path exists

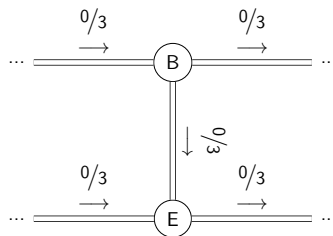
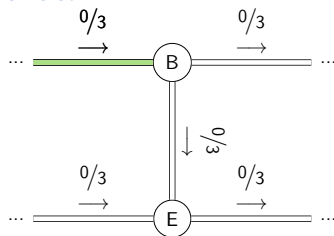
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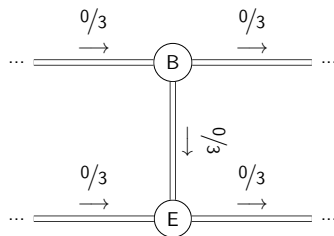
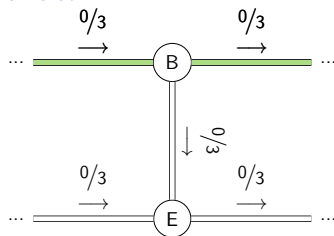
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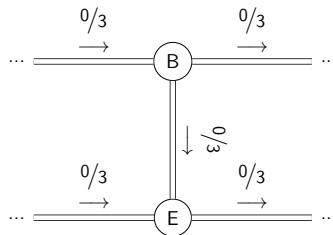
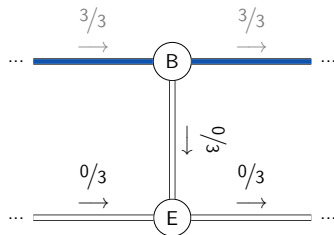
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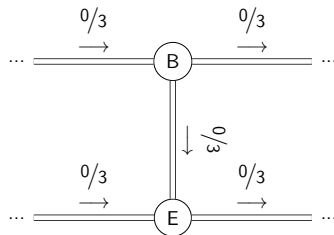
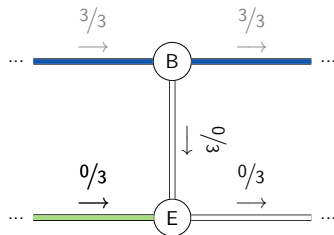
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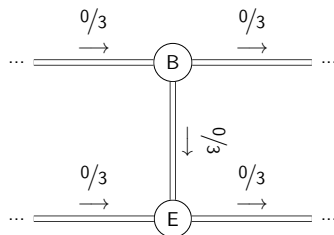
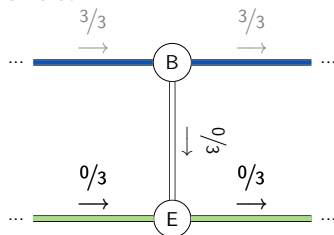
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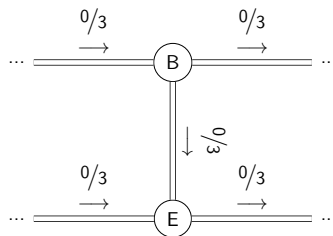
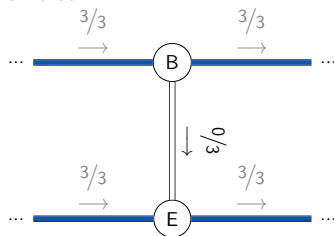
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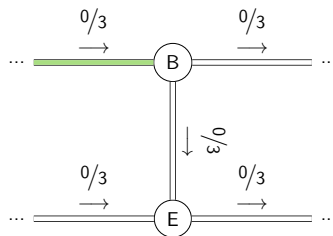
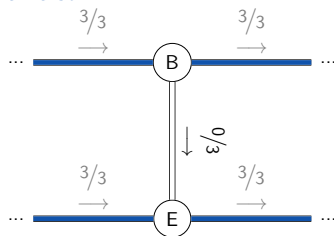
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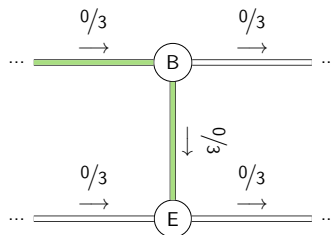
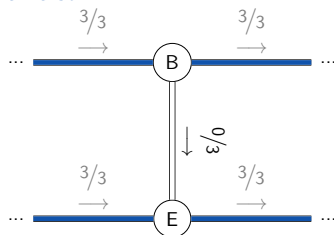
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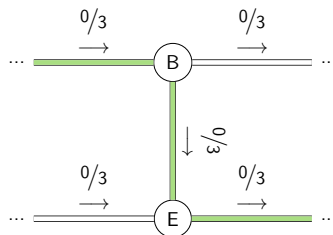
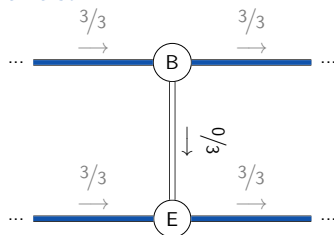
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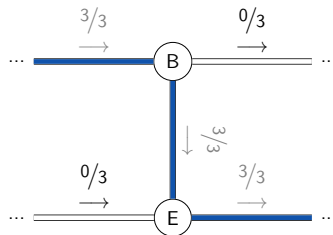
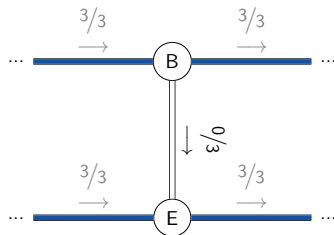
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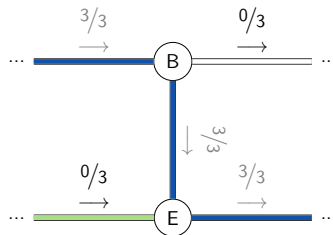
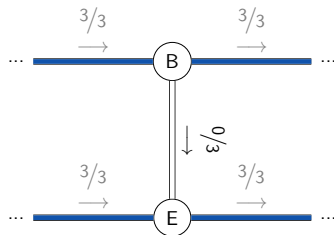
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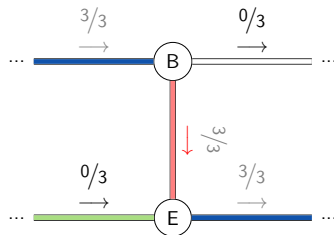
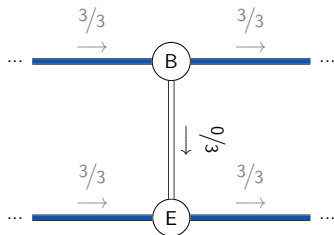
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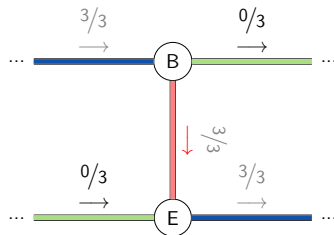
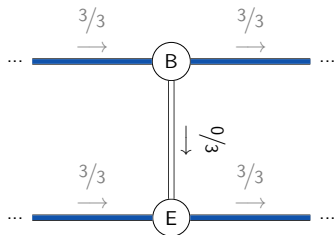
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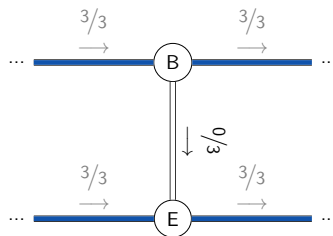
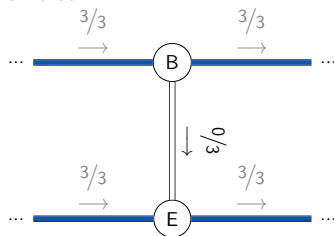
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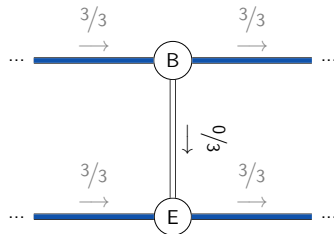
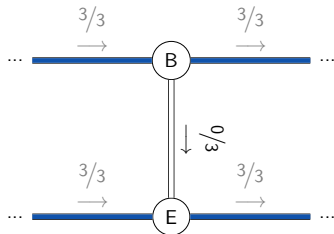
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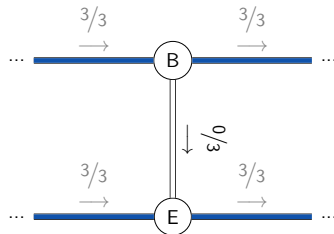
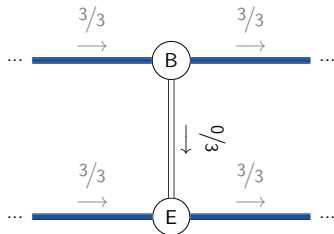


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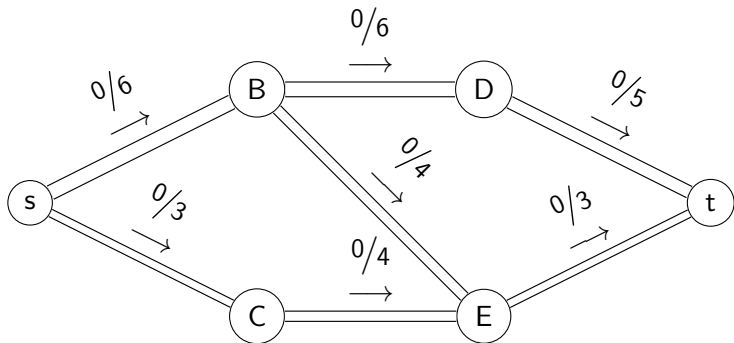
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### Remark: Augmenting Paths

Several Augmenting Paths may be available to choose from. Any such path is sufficient for the correctness of Ford-Fulkerson.

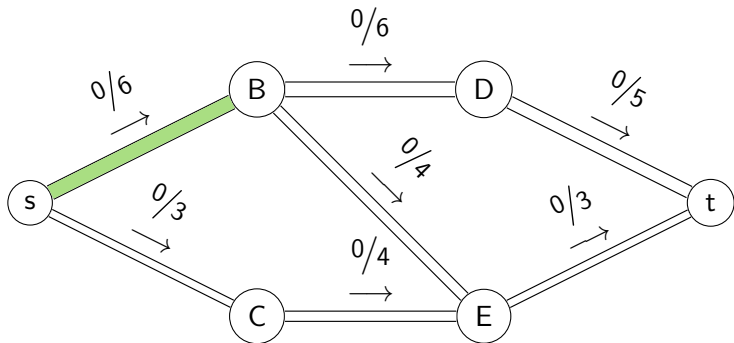
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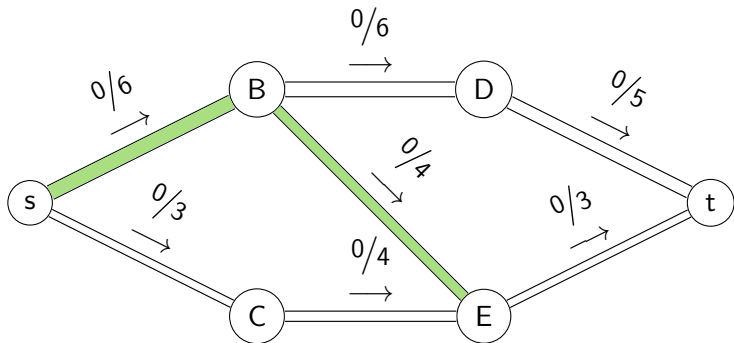
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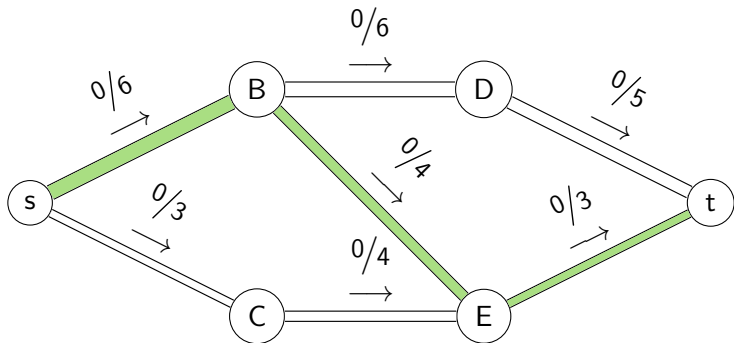
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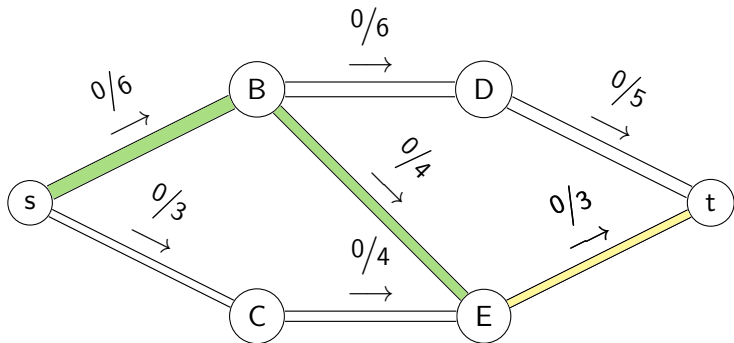
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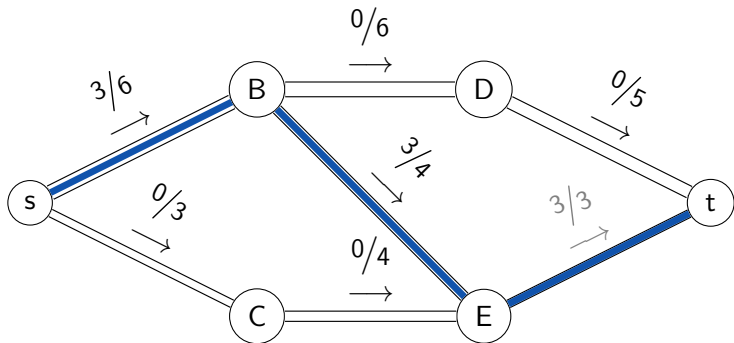
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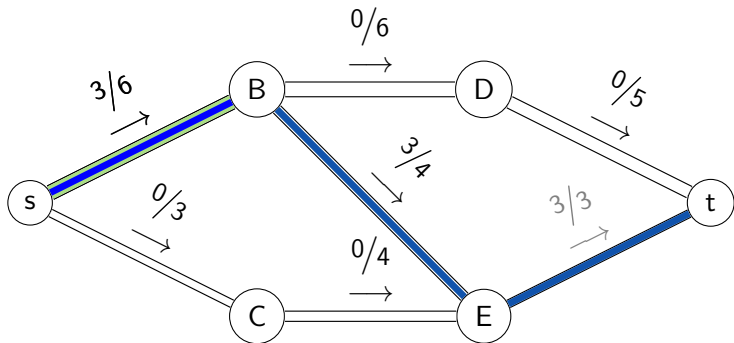
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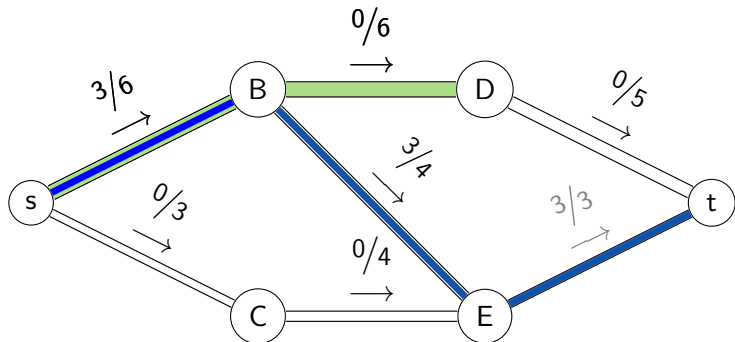
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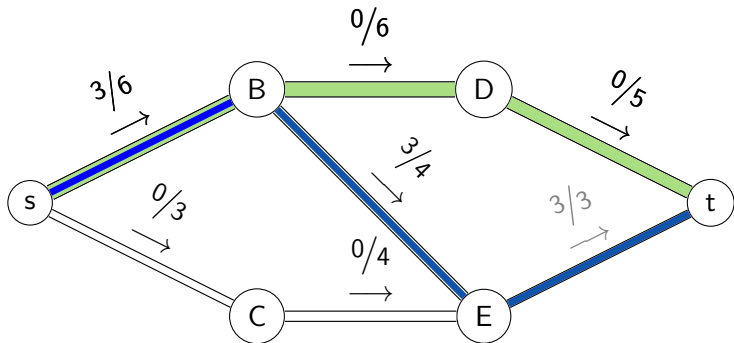
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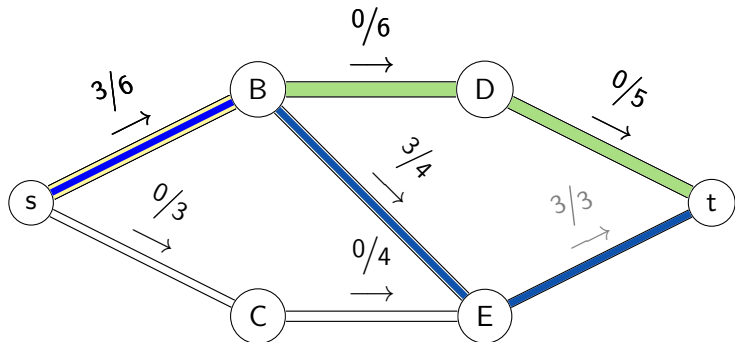
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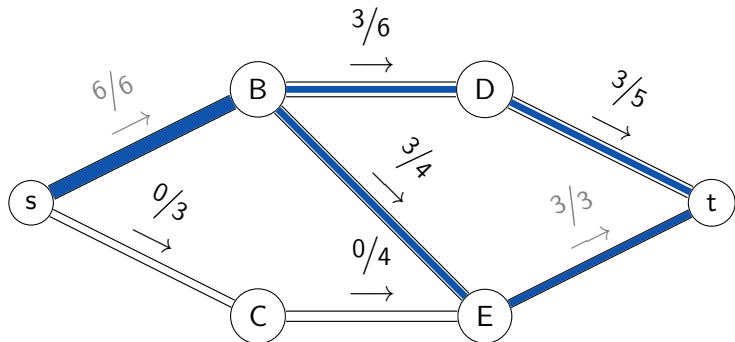
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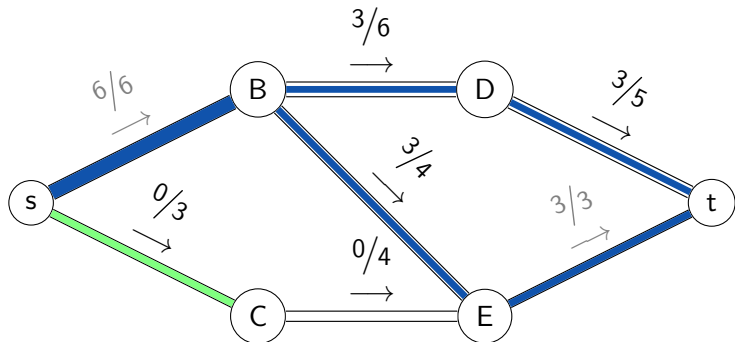
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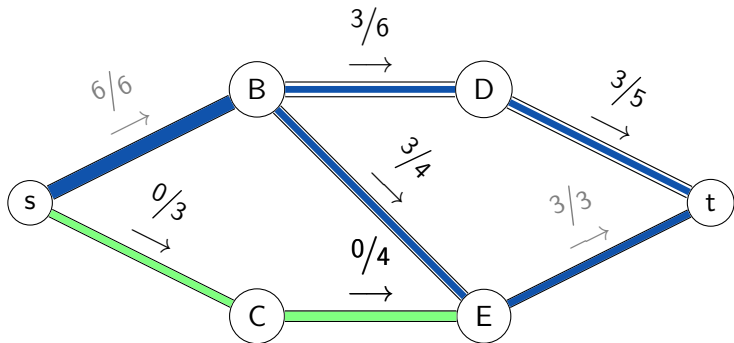
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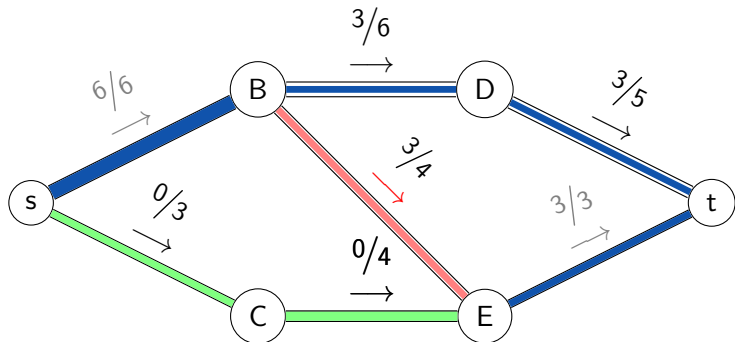
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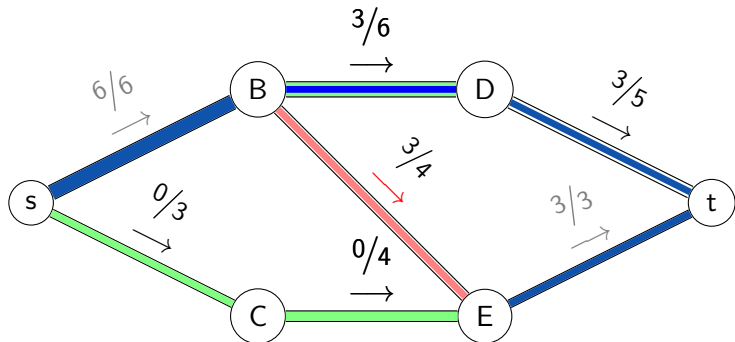
# Flow Problems

## Ford-Fulkerson: Example Run



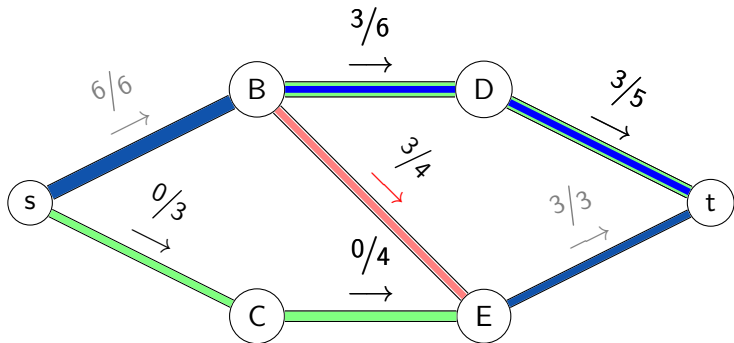
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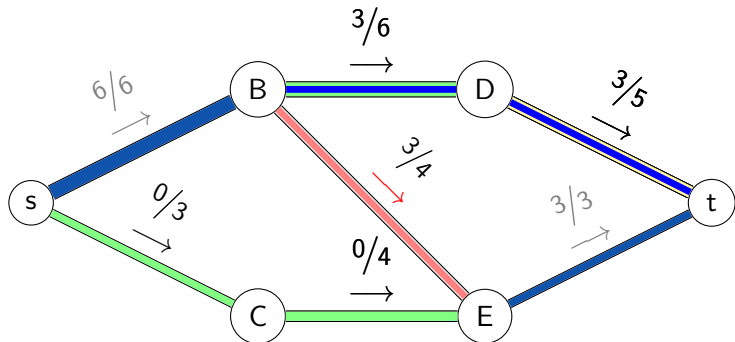
# Flow Problems

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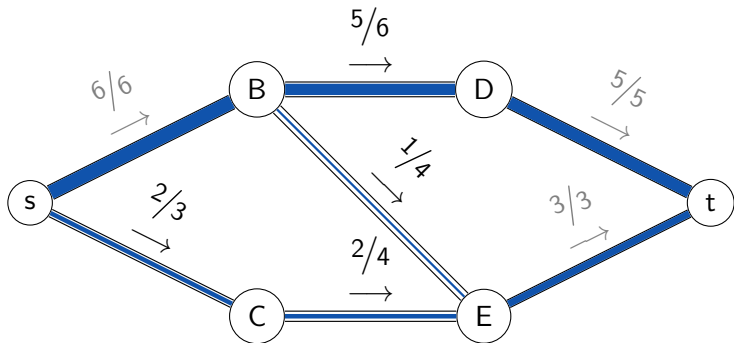
# Flow Problems

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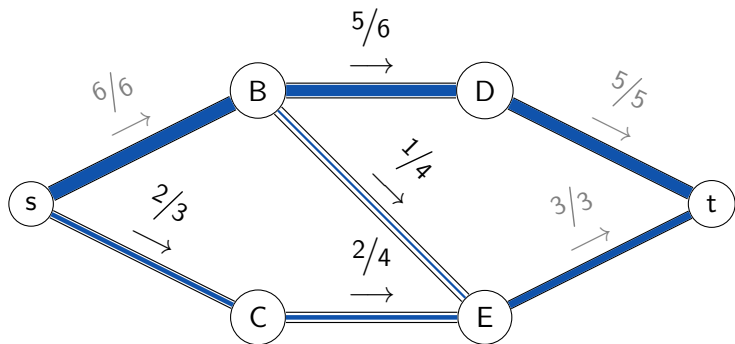
# Flow Problems

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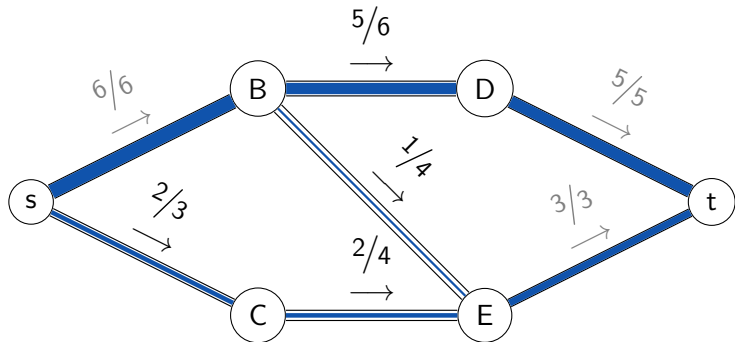


### Result

Since no water gets lost in the system, the maximum flow can be read at  $s$  and  $t$ .

# Flow Problems

## Ford-Fulkerson: Example Run



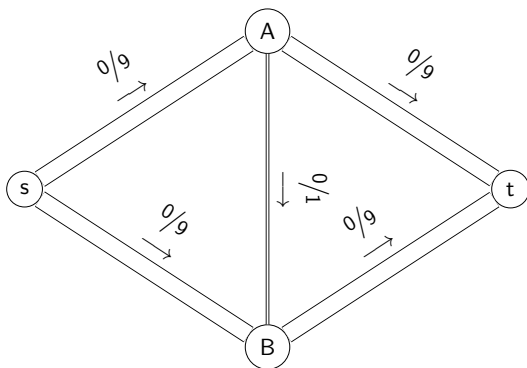
### Result

Since no water gets lost in the system, the maximum flow can be read at  $s$  and  $t$ . The Max Flow value is  $6 + 2 = 5 + 3 = 8$ .



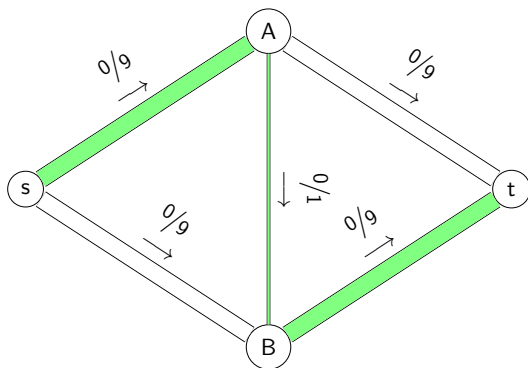
# Flow Problems

## Ford-Fulkerson: Analysis



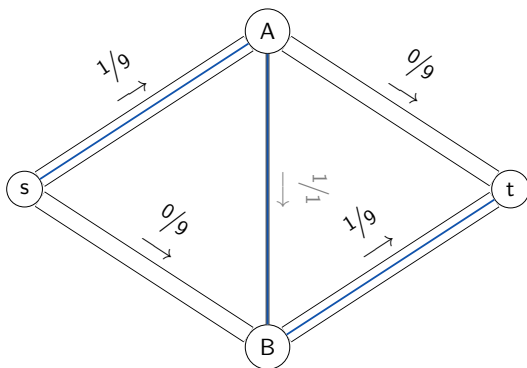
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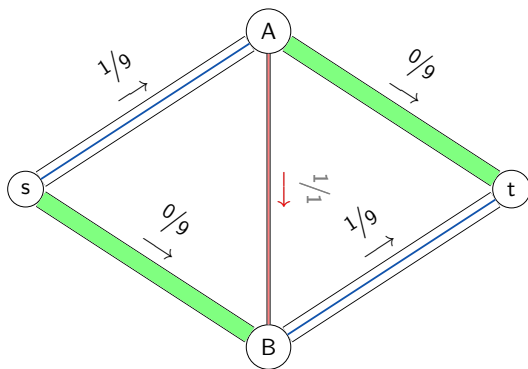
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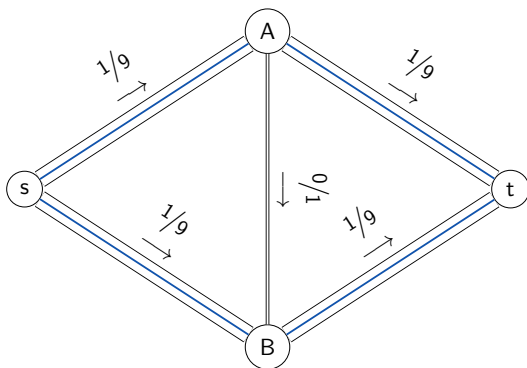
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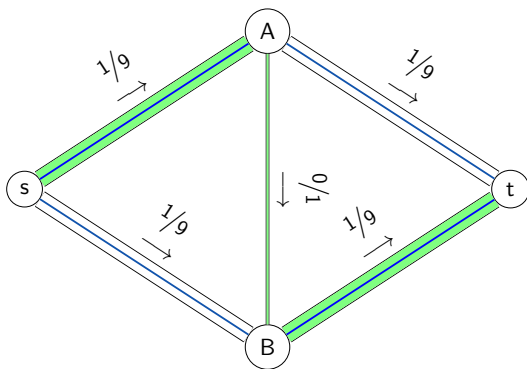
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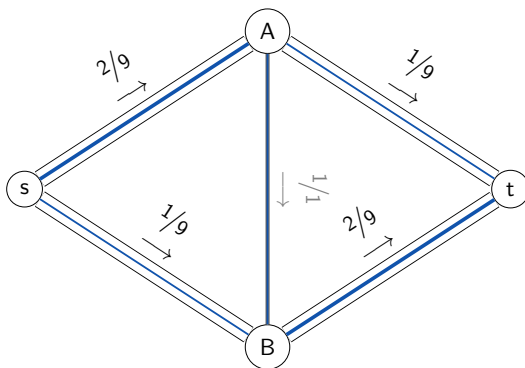
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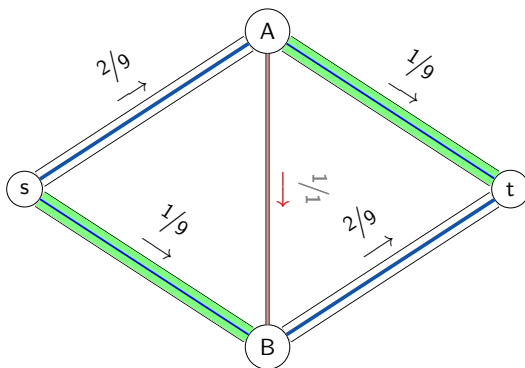
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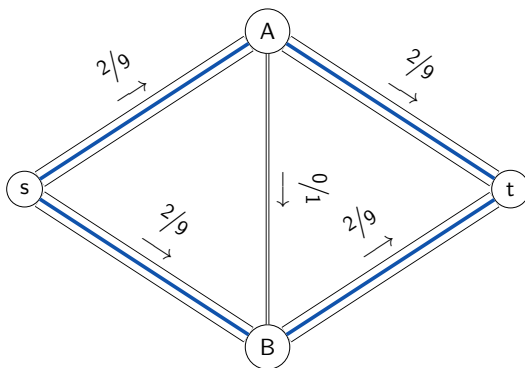
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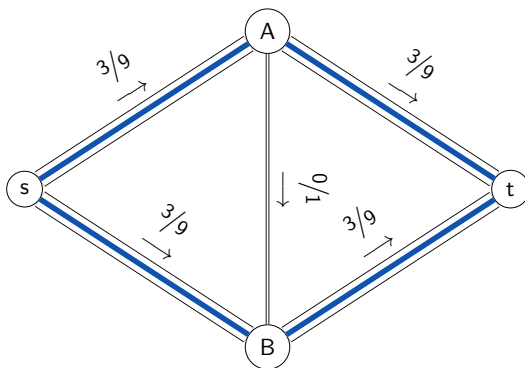
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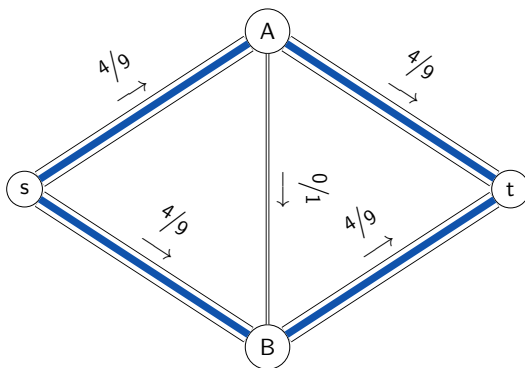
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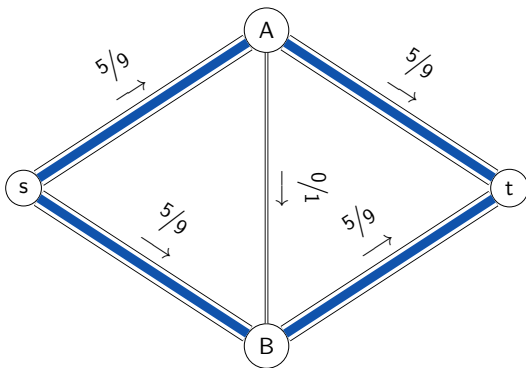
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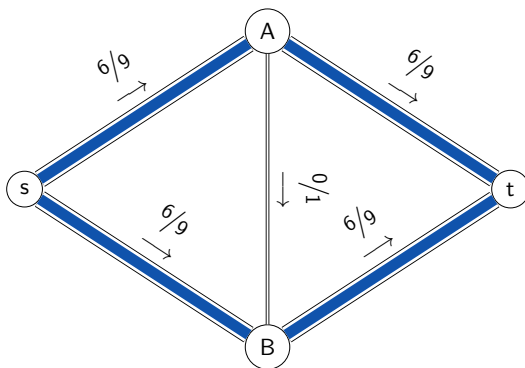
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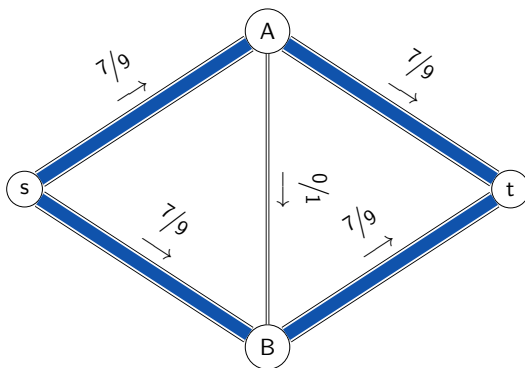
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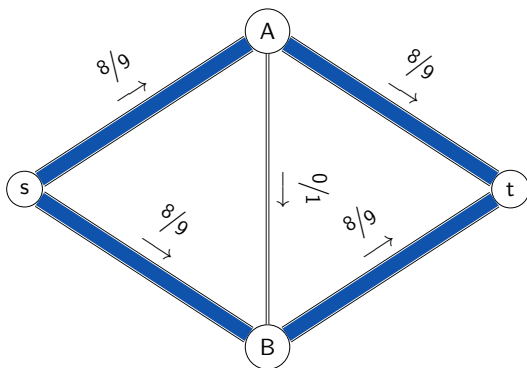
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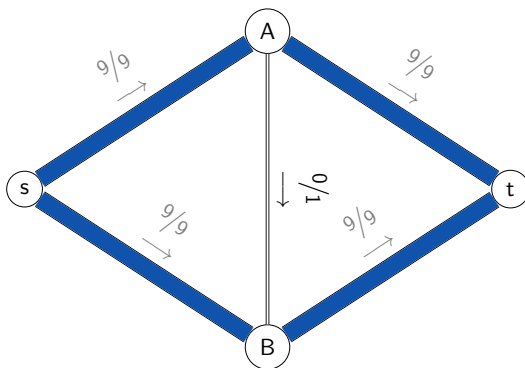
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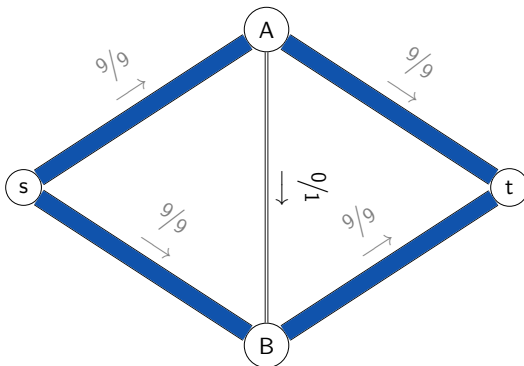
## Ford-Fulkerson: Analysis





# Flow Problems

## Ford-Fulkerson: Analysis



### Analysis of Ford-Fulkerson

FF runs in  $\mathcal{O}(\mathcal{F}^* \cdot E)$  where  $\mathcal{F}^*$  is the maximum flow  
→ Find a better way of finding Augmenting Paths

# Flow Problems

## Edmonds-Karp

### Insight: Ford-Fulkerson

Picking arbitrary Augmenting Paths can lead to  $\mathcal{F}^*$  many iterations.

# Flow Problems

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### Insight (Edmonds & Karp)

Always pick the shortest Augmenting Path – in terms of hops.

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They showed that, after  $\mathcal{O}(VE)$  iterations, no Augmenting Paths exist.

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## Edmonds-Karp

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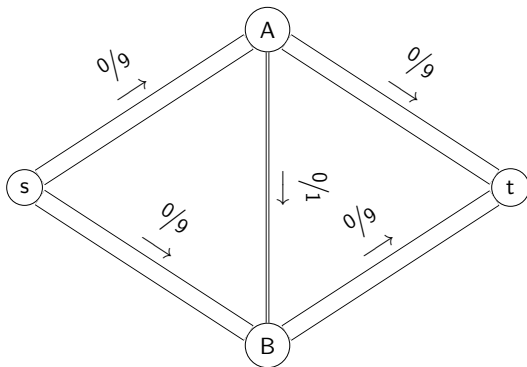
Always pick the shortest Augmenting Path – in terms of hops.  
→ use BFS

They showed that, after  $\mathcal{O}(VE)$  iterations, no Augmenting Paths exist.

Each BFS runs in  $\mathcal{O}(E)$ , totalling to  $\mathcal{O}(VE^2)$ .

# Flow Problems

## Edmonds & Karp's Algorithm

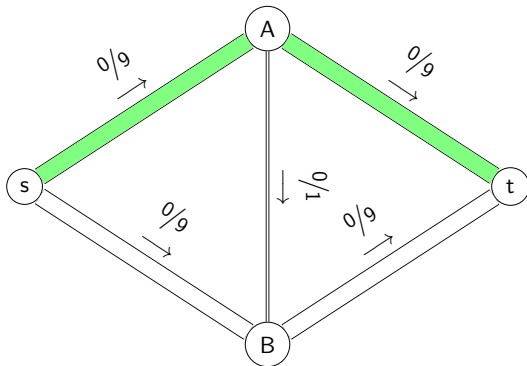


Insight (Edmonds & Karp)

Pick the shortest Augmenting Paths – in terms of hops.

# Flow Problems

## Edmonds & Karp's Algorithm



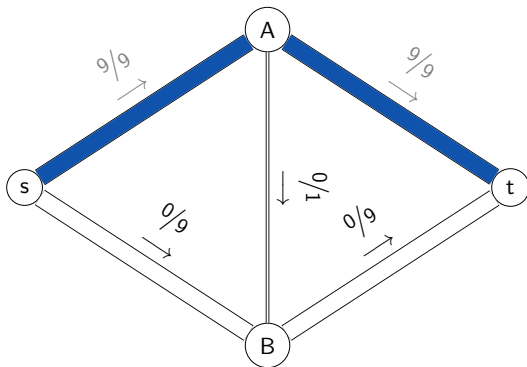
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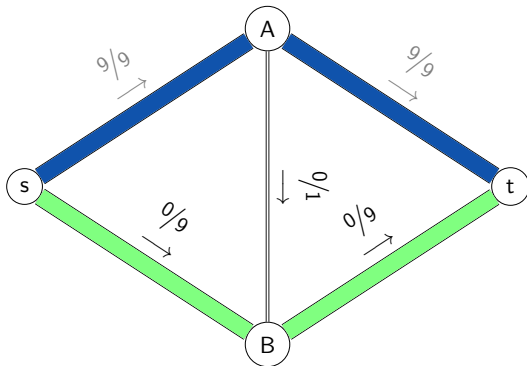


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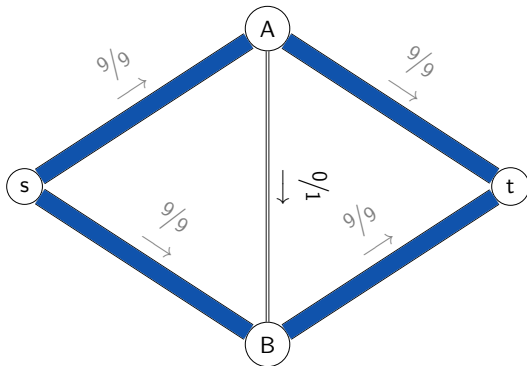


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# Flow Problems

## Edmonds & Karp's Algorithm



Insight (Edmonds & Karp)

Pick the shortest Augmenting Paths – in terms of hops.

# Flow Problems

## Edmonds & Karp's Algorithm

```
1  vector<vector<int>> capacity; // adjacency matrix
2  vector<vector<int>> adj;      // adjacency list
3  vector<int> parent;          // used in BFS
4
5  int main() {
6      int s, t, V, E;
7      cin >> V >> E >> s >> t;
8      adj.resize(V);
9      capacity.assign(V, vector<int>(V, 0));
10     int u, v, c;
11     for (int i = 0; i < E; i++) {
12         cin >> u >> v >> c;
13         adj[u].push_back(v);
14         adj[v].push_back(u);
15         capacity[u][v] += c; // forward capacity
16     }
17     int out = maxflow(s, t);
18     cout << out << endl;
19 }
```

- Capacity could be an adjacency list  
→ Code will be in CMS

# Flow Problems

## Edmonds & Karp's Algorithm

```
1  int maxflow(int s, int t) {
2      int totalflow = 0, u;
3      while (true) {
4          bfs(s); // build bfs tree
5          if (parent[t] == -1) break; // unreachable
6          int bottleneck = INF;
7          u = t; // find bottleneck capacity
8          while (u != s) {
9              int v = parent[u];
10             bottleneck = min(bottleneck, capacity[v][u]);
11             u = v;
12         }
13         u = t; // update capacities along path
14         while (u != s) {
15             int v = parent[u];
16             capacity[v][u] -= bottleneck;
17             capacity[u][v] += bottleneck;
18             u = v;
19         }
20         totalflow += bottleneck;
21     }
22     return totalflow;
23 }
```

# Flow Problems

## Edmonds & Karp's Algorithm

```
1 void bfs(int s) {
2     parent.assign(adj.size(), -1);
3     parent[s] = -2; // s is visited
4
5     queue<int> Q;
6     Q.push(s);
7     while (!Q.empty()) {
8         int u = Q.front(); Q.pop();
9         for (int v : adj[u])
10             if (parent[v] == -1 and capacity[u][v] > 0) {
11                 Q.push(v);
12                 parent[v] = u;
13             }
14     }
15 }
```

- ▶ Almost plain BFS. Tracks parent pointers and checks if capacity is nonzero

# Flow Problems

## Recap

### Maximum Flow Algorithms

Algorithm	Concept	Running Time
Ford-Fulkerson	Augmenting Paths	$\mathcal{O}(\mathcal{F}^* E)$
Edmonds-Karp	Augmenting Paths	$\mathcal{O}(VE^2)$
Dinic	Blocking Flow	$\mathcal{O}(V^2 E)$

Dinic's Algorithm is worth a glance and useful to have at hand, but Edmonds-Karp's is usually sufficient for competitive programming.



# Modeling Flow Problems

## The Golden Hammer of Network Flow

Once you know Edmonds-Karp's algorithm, every problem starts to look like a Max Flow instance.

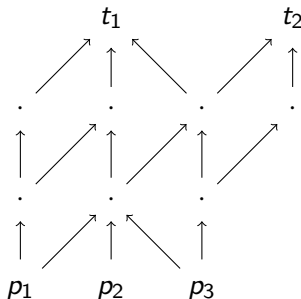


# Modeling Flow Problems

## Hiking

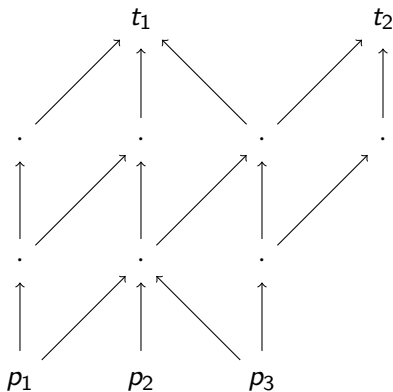
### Problem: Hiking

You are a big fan of hiking. Your favorite hike starts any of the parking lots  $p_1$ ,  $p_2$  or  $p_3$  and ends at either one of the tops  $t_1$  or  $t_2$ . Sadly, you get bored very easily. How many different hikes starting at any parking lot can you find such that you visit no path segment twice?



# Modeling Flow Problems

## Hiking: Modeling

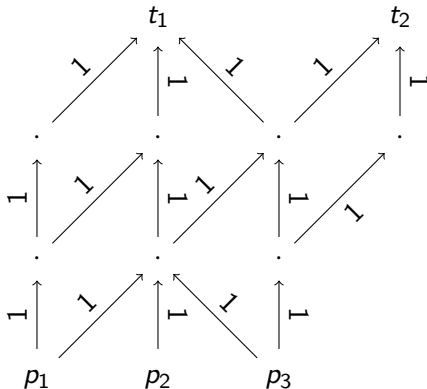


Hiking: Idea

See Hiking as a Max Flow instance

# Modeling Flow Problems

## Hiking: Modeling



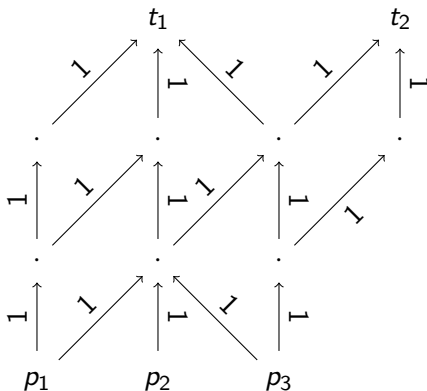
### Hiking: Idea

See Hiking as a Max Flow instance

- ▶ Edges have unit capacities

# Modeling Flow Problems

## Hiking: Modeling



### Hiking: Idea

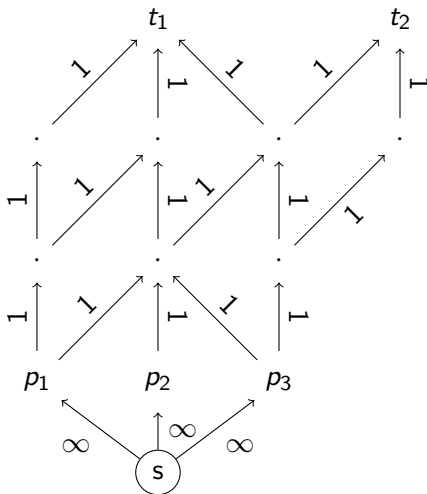
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S

# Modeling Flow Problems

## Hiking: Modeling



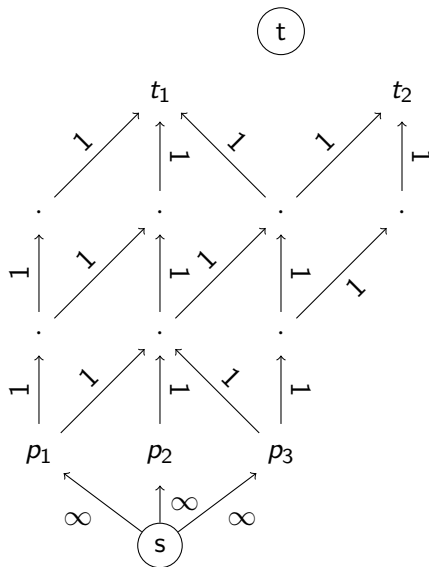
### Hiking: Idea

See Hiking as a Max Flow instance

- ▶ Edges have unit capacities
- ▶ Add super source, connect to  $p_i$

# Modeling Flow Problems

## Hiking: Modeling



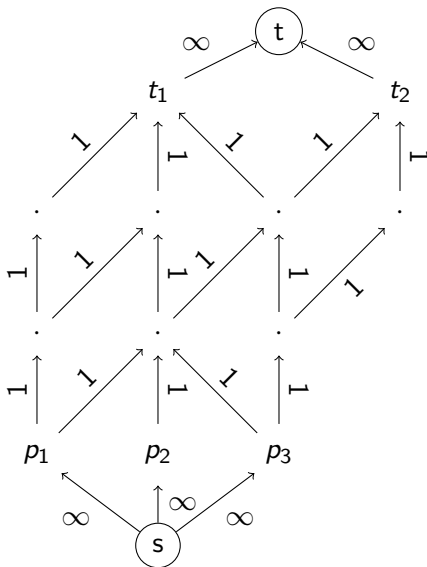
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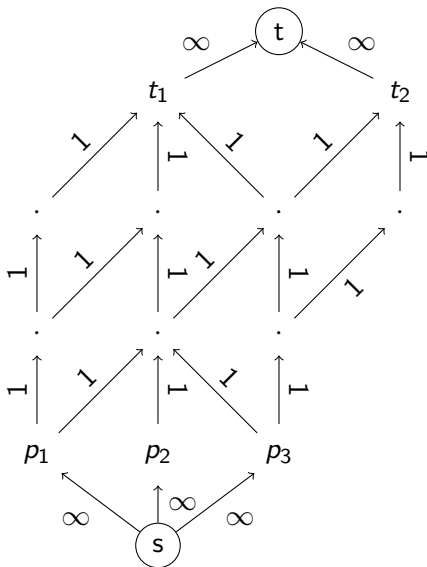
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# Modeling Flow Problems

## Hiking: Modeling



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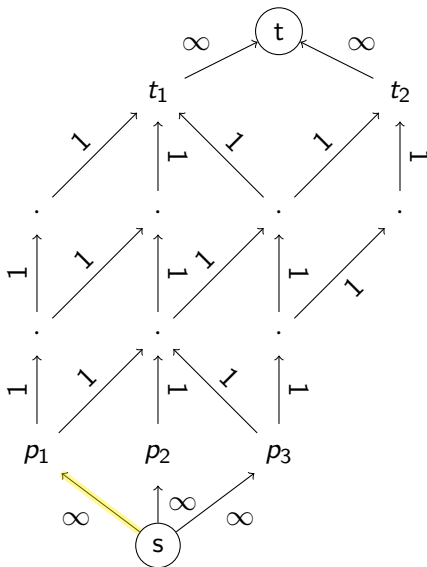
- ▶ Edges have unit capacities
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Finally,  $\text{MaxFlow}(s, t)$  computes the number of edge-disjoint paths between  $p_i$  to  $s_j$



# Modeling Flow Problems

## Hiking: Modeling



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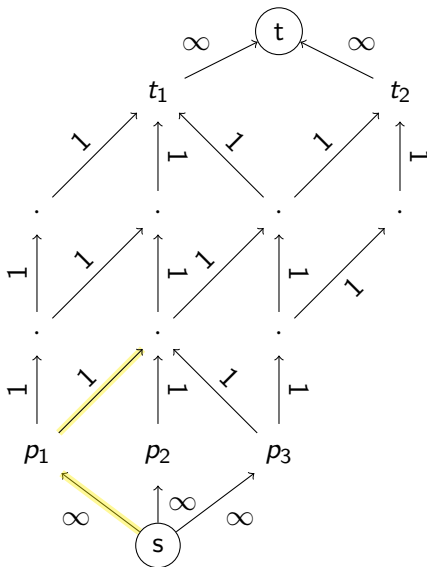
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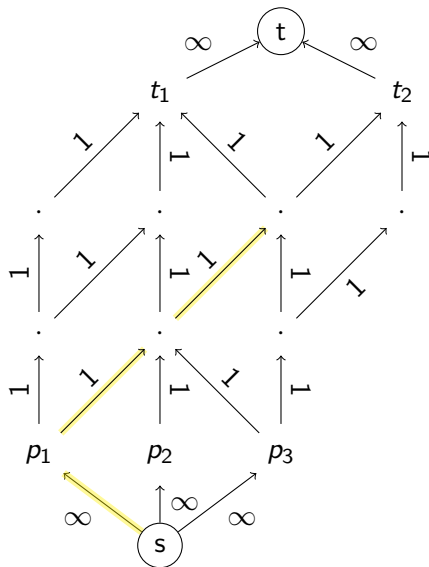
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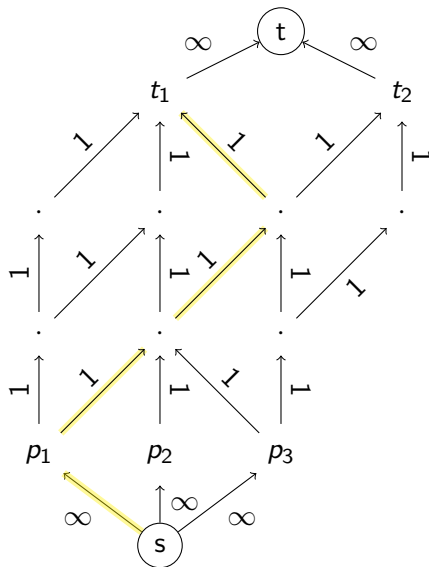
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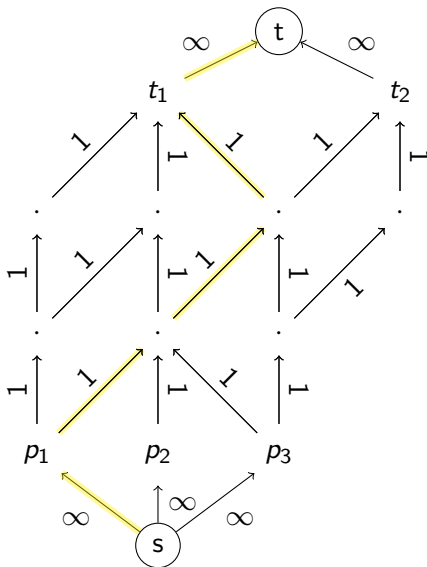
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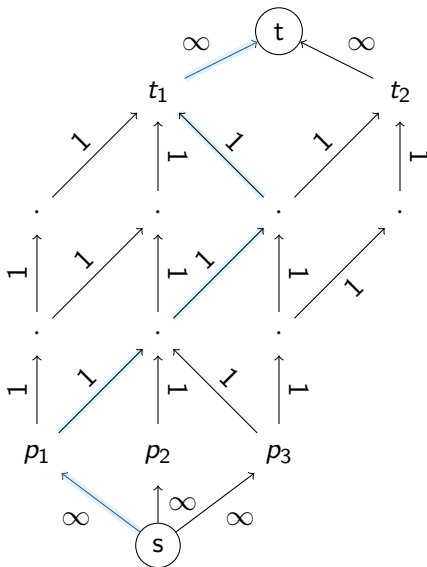
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### Hiking: Solution

There are four edge-disjoint paths from the parking lots to the summits.

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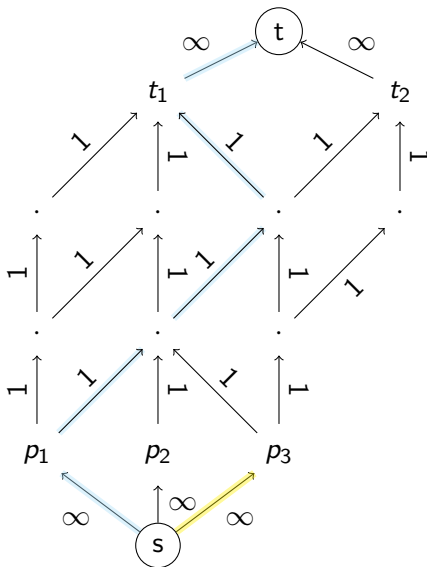
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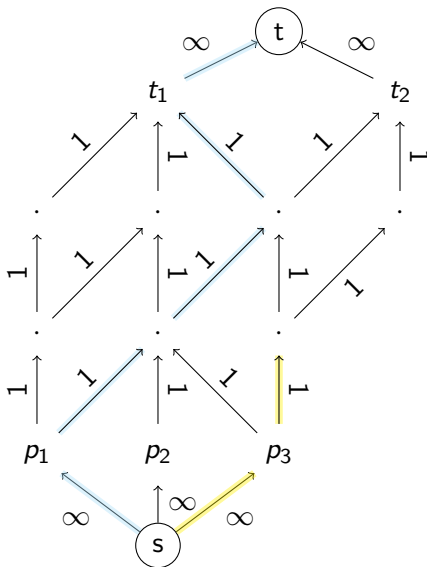
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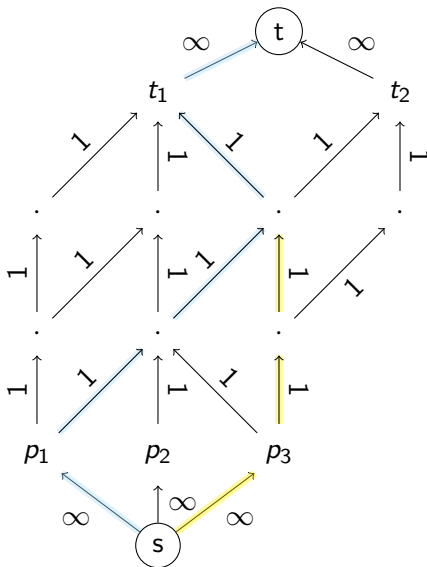
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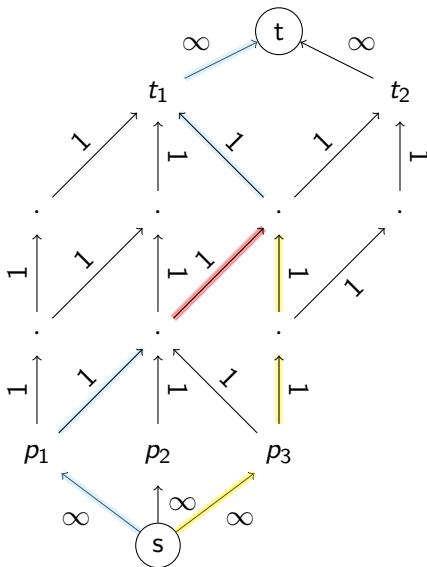
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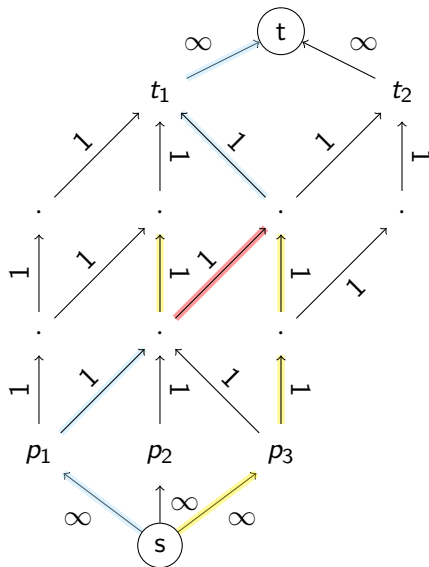
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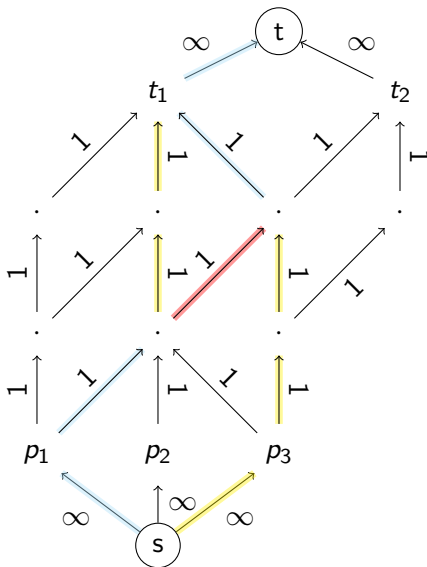
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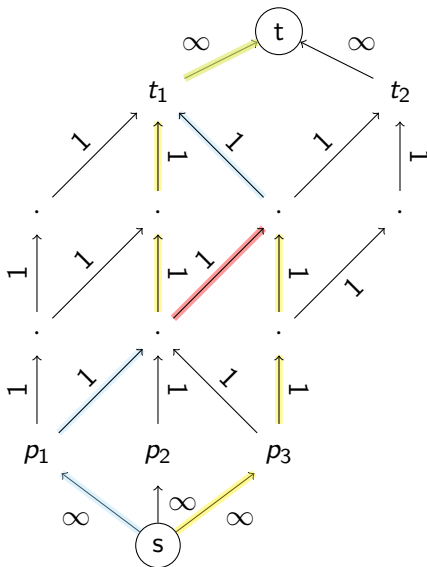
Finally,  $\text{MaxFlow}(s, t)$  computes the number of edge-disjoint paths between  $p_i$  to  $s_i$

## Hiking: Solution

There are four edge-disjoint paths from the parking lots to the summits.



## Hiking: Modeling



## See Hiking as a Max Flow instance

- ▶ Edges have unit capacities
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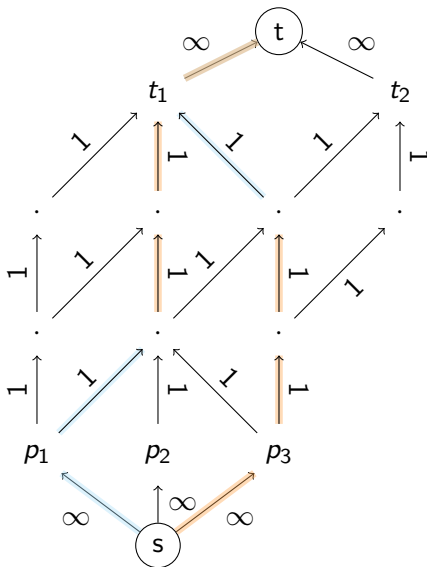
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# Modeling Flow Problems

## Hiking: Modeling



### Hiking: Idea

See Hiking as a Max Flow instance

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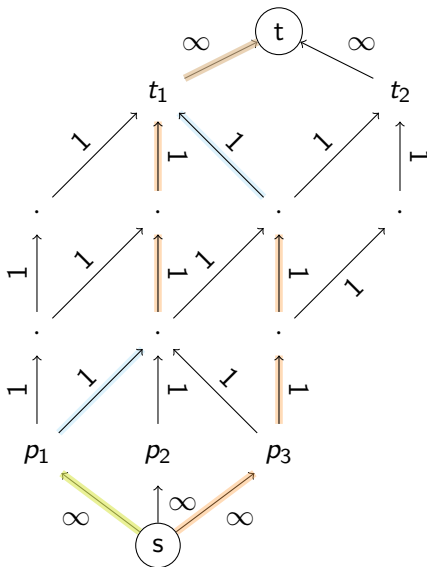
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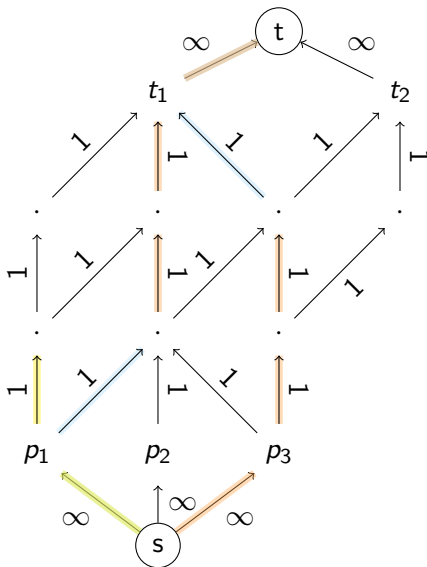
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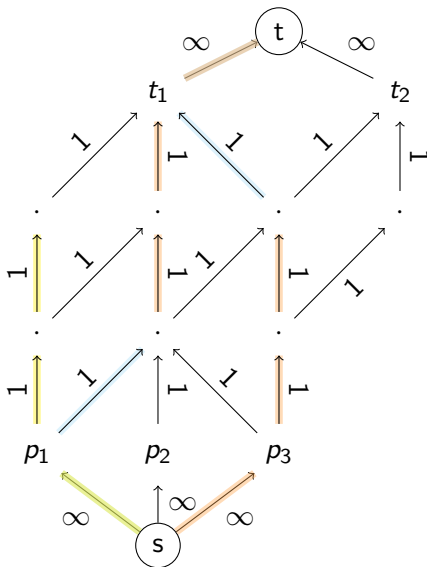
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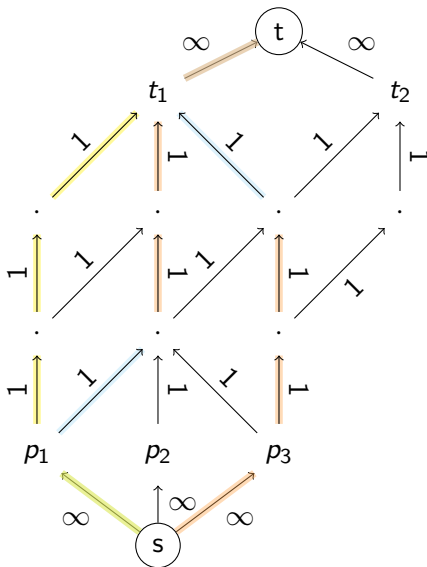
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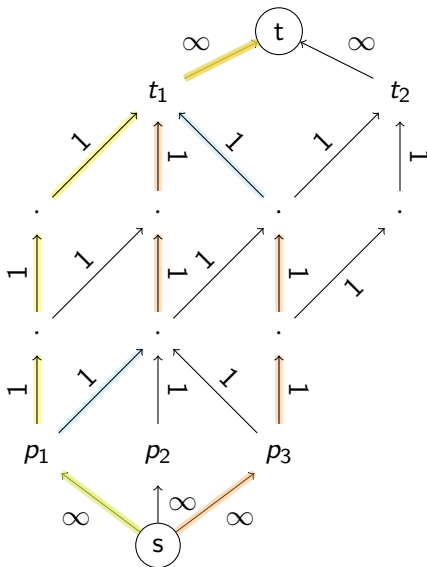
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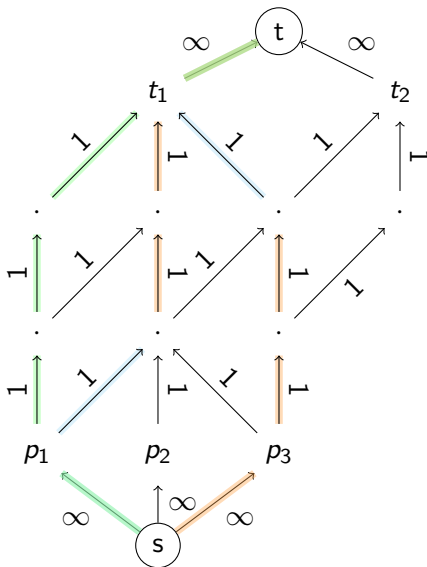
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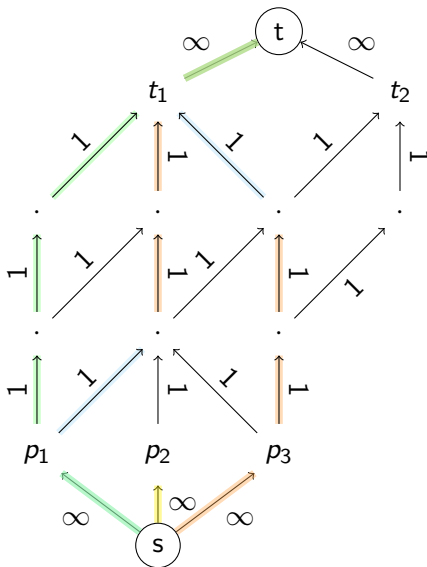
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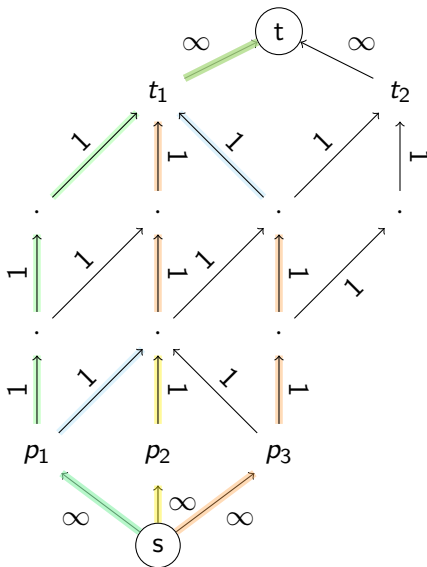
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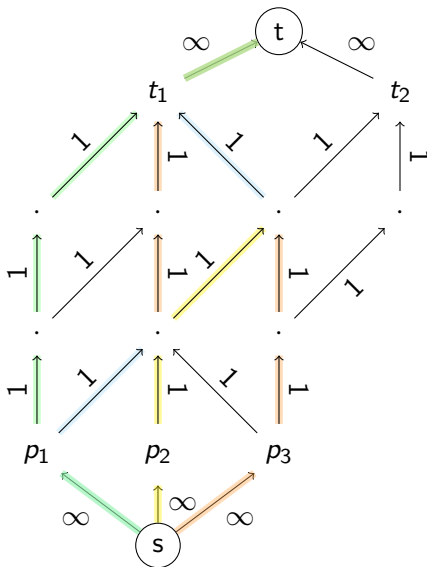
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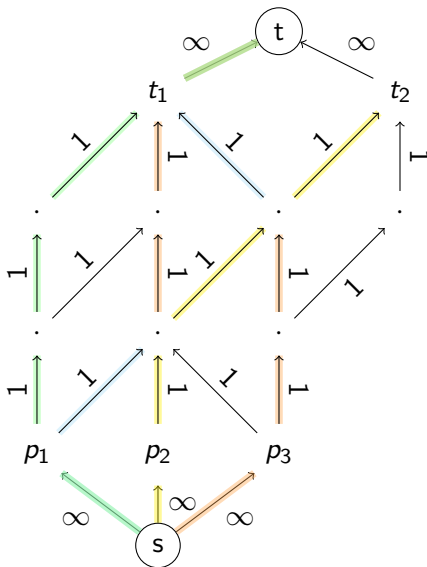
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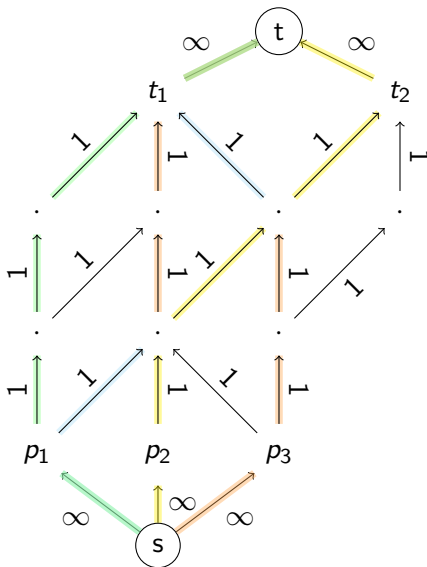
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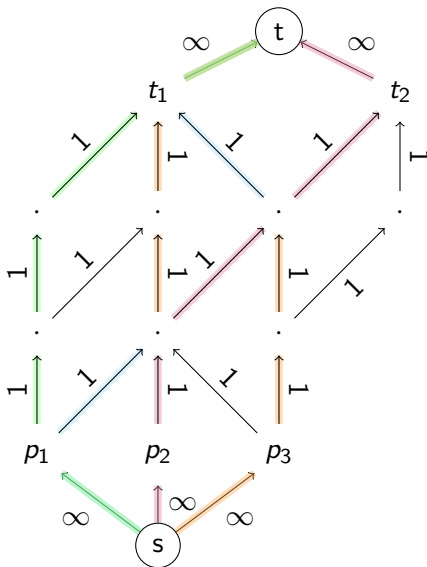
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# Modeling Flow Problems

## Lily Pads

### Problem: Lily Pads

A bunch of frogs are sitting on lily pads. They want to jump from leaf to leaf in order to reach a big lotus flower in the center  $(0,0)$ . Each frog can jump a distance of up to three units. However, some leaves are brittle and can only be jumped off once. Given the coordinates of the frogs and lily pads, how many frogs can meet at the flower?

## Lily Pads: Modeling



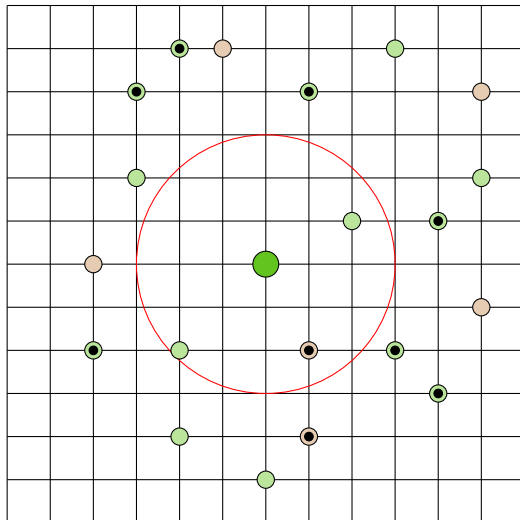
## See Lily Pads as a Max Flow instance



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# Modeling Flow Problems

## Lily Pads: Modeling



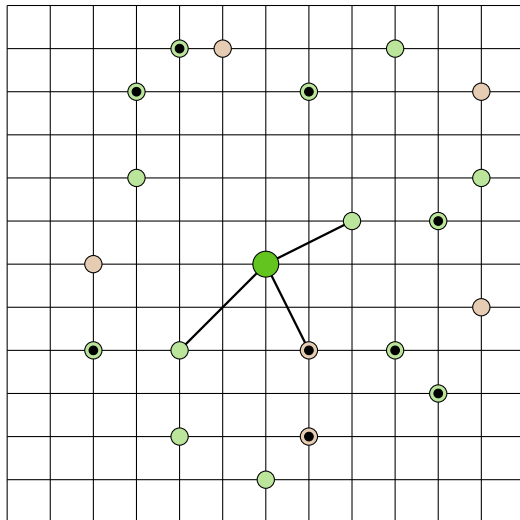
### Lily Pads: Idea

See Lily Pads as a Max Flow instance

► Model Graph

# Modeling Flow Problems

## Lily Pads: Modeling



● Lily ● Brittle ● Frog ● Lotus Flower

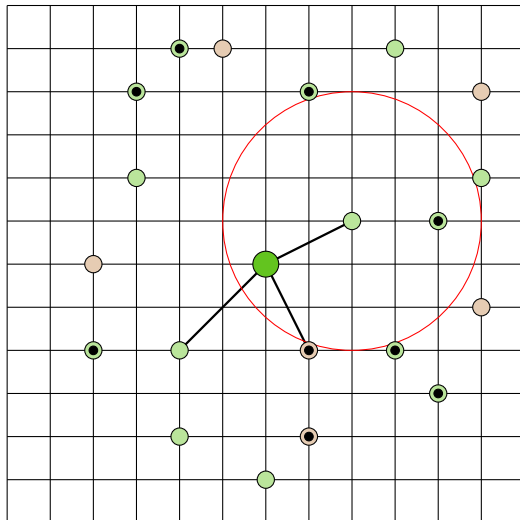
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# Modeling Flow Problems

## Lily Pads: Modeling



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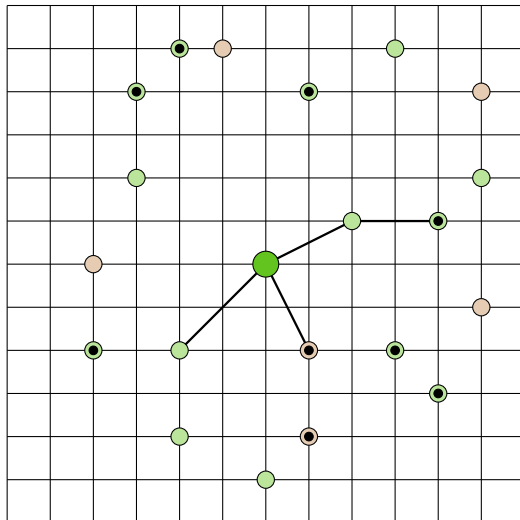
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See Lily Pads as a Max Flow instance

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# Modeling Flow Problems

## Lily Pads: Modeling



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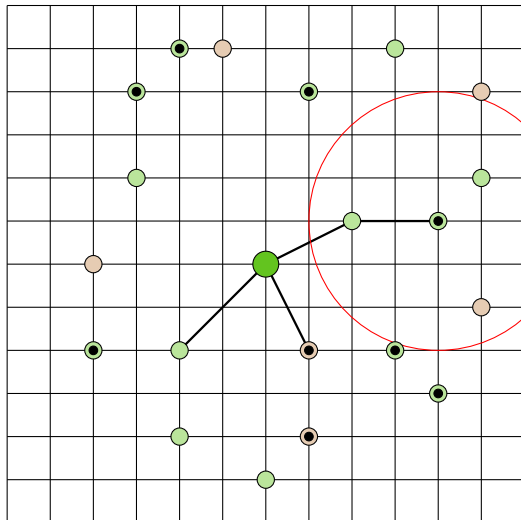
See Lily Pads as a Max Flow instance

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# Modeling Flow Problems

## Lily Pads: Modeling



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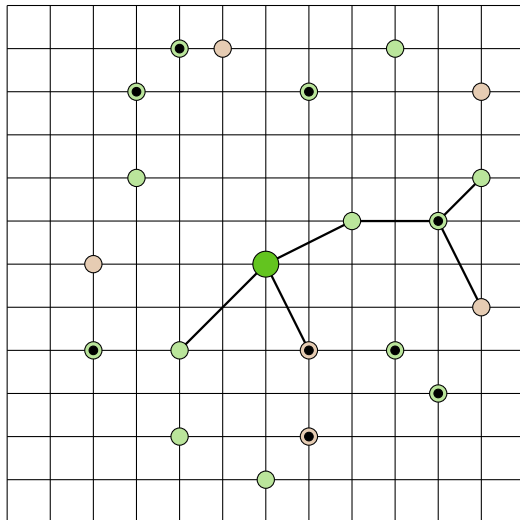
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# Modeling Flow Problems

## Lily Pads: Modeling



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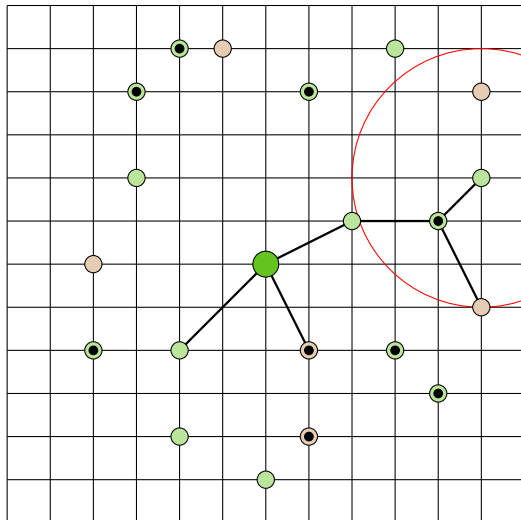
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See Lily Pads as a Max Flow instance

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# Modeling Flow Problems

## Lily Pads: Modeling



### Lily Pads: Idea

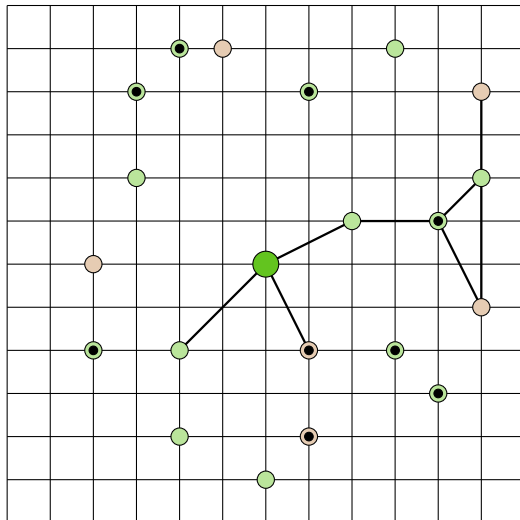
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# Modeling Flow Problems

## Lily Pads: Modeling



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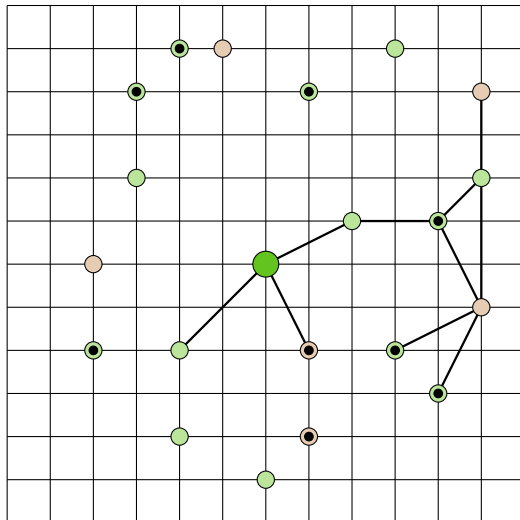


## See Lily Pads as a Max Flow instance

- ▶ Model Graph

# Modeling Flow Problems

## Lily Pads: Modeling



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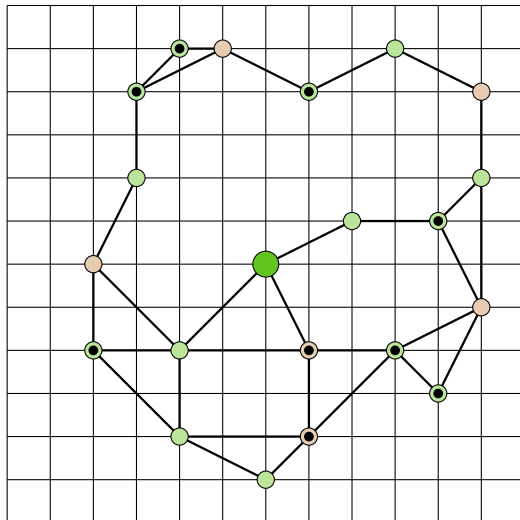
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# Modeling Flow Problems

## Lily Pads: Modeling



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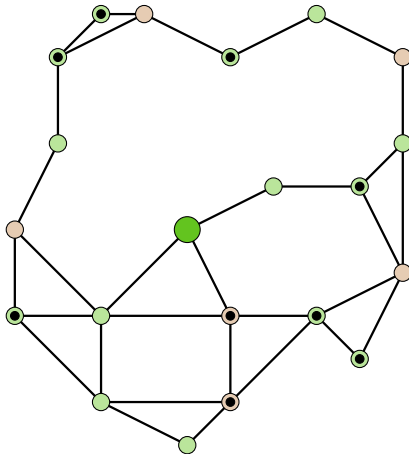
### Lily Pads: Idea

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# Modeling Flow Problems

## Lily Pads: Modeling



### Lily Pads: Idea

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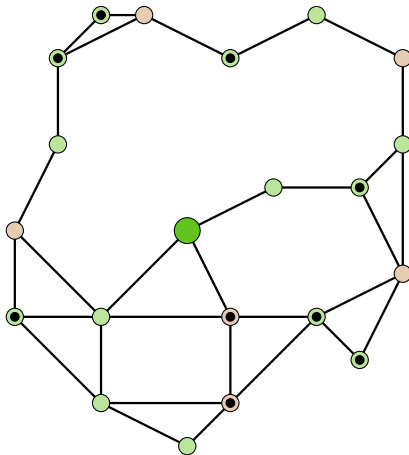
► Model Graph

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# Modeling Flow Problems

## Lily Pads: Modeling



### Lily Pads: Idea

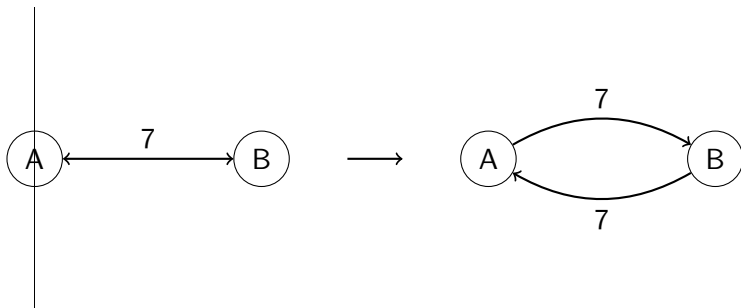
See Lily Pads as a Max Flow instance

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- ▶ Undirected Edges of capacity  $\infty$

● Lily ● Brittle ● Frog ● Lotus Flower

# Modeling Flow Problems

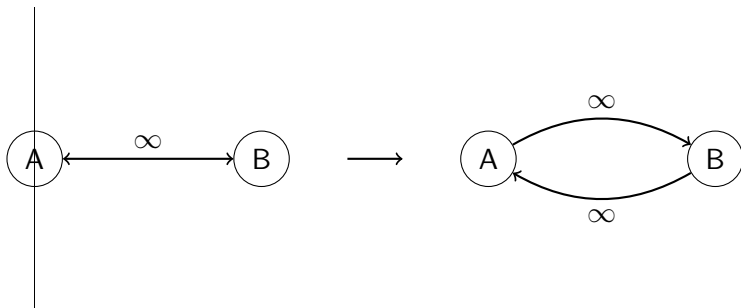
## Modeling Undirected Edges



```
1  int main() {  
2      // setting up capacities  
3      capacity[u][v] = 7; // forward capacity  
4      capacity[v][u] = 7; // backward capacity  
5      // ...  
6  }
```

# Modeling Flow Problems

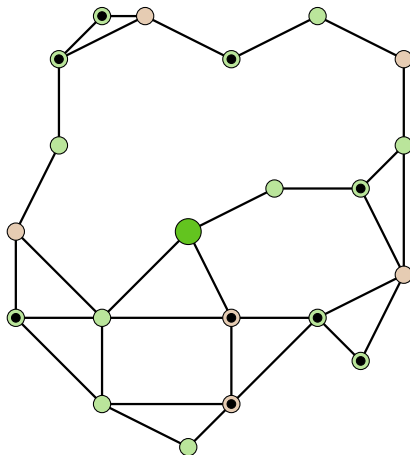
## Modeling Undirected Edges



```
1 int main() {  
2     // setting up capacities  
3     capacity[u][v] = INF / 2; // prevent overflow when  
4     capacity[v][u] = INF / 2; // updating flow  
5     // ...  
6 }
```

# Modeling Flow Problems

## Lily Pads: Modeling



### Lily Pads: Idea

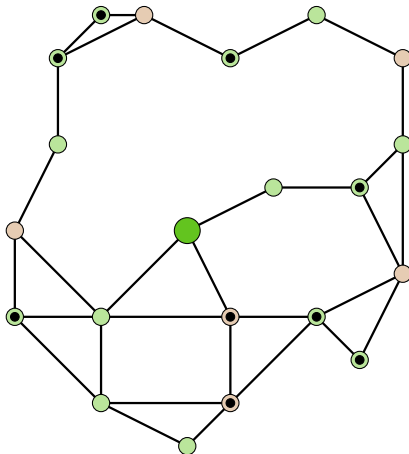
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● Lily ● Brittle ● Frog ● Lotus Flower

# Modeling Flow Problems

## Lily Pads: Modeling



### Lily Pads: Idea

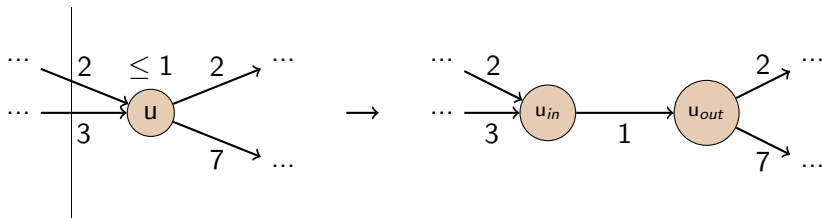
See Lily Pads as a Max Flow instance

- ▶ Model Graph
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- ▶ Brittle Leaves  
→ Vertex Capacities

● Lily ● Brittle ● Frog ● Lotus Flower

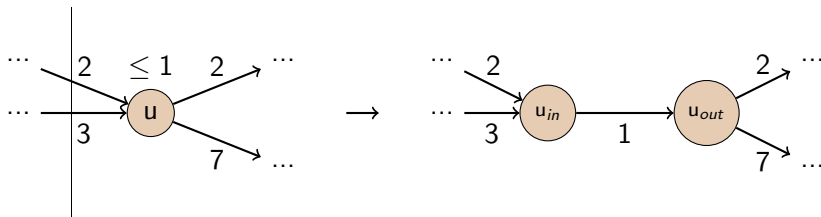
# Modeling Flow Problems

## Modeling Vertex Capacities



# Modeling Flow Problems

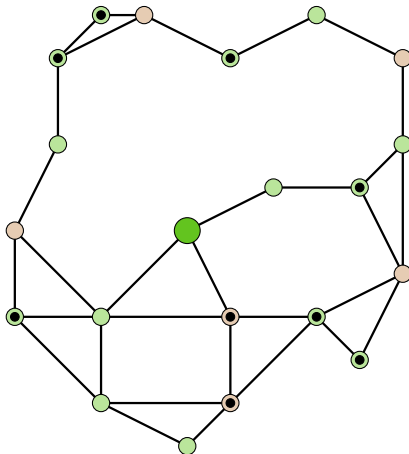
## Modeling Vertex Capacities



```
1 vector<vector<int>> adj(2*V);
2 int u, v;
3 // access to duplicate nodes
4 int u_in = u; int u_out = u + V;
5 int v_in = v; int v_out = v + V;
6 // edge u → v
7 adj[u_out].push_back(v_in);
8 adj[v_in].push_back(u_out);
9
10 // vertex capacity
11 capacity[u_in][u_out] = 1;
12 capacity[u_out][u_in] = 0;
```

# Modeling Flow Problems

## Lily Pads: Modeling



● Lily ● Brittle ● Frog ● Lotus Flower

### Lily Pads: Idea

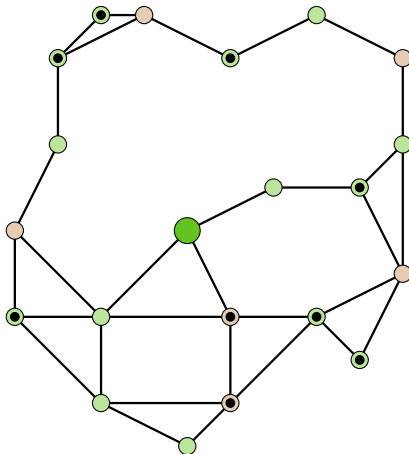
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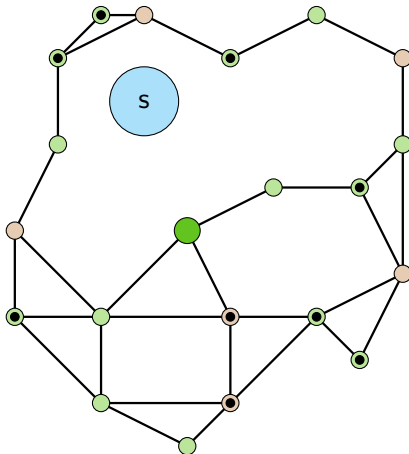
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See Lily Pads as a Max Flow instance

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→ Vertex Capacities
- ▶ Add super source  
→ Capacity 1 per frog

# Modeling Flow Problems

## Lily Pads: Modeling



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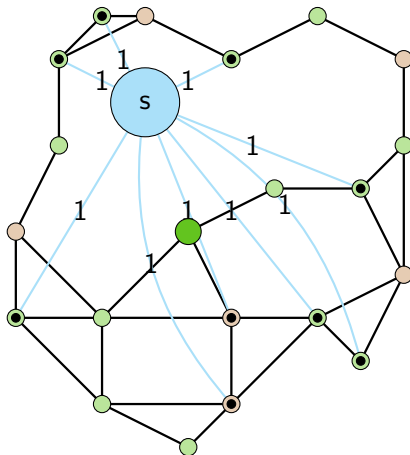
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# Modeling Flow Problems

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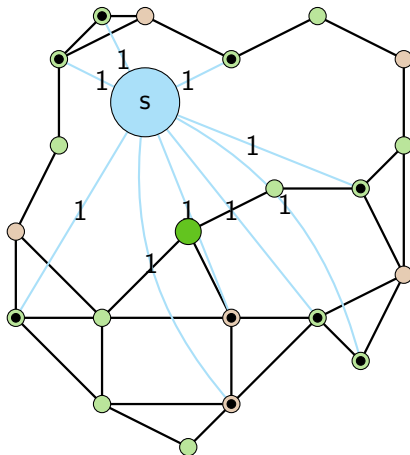
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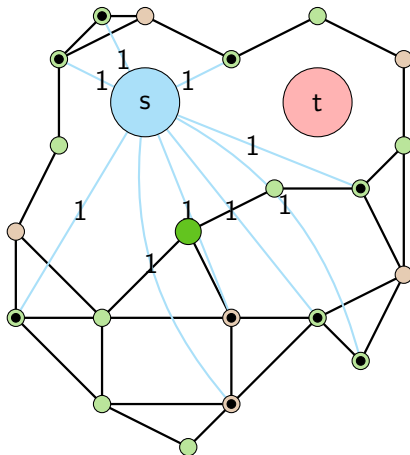
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# Modeling Flow Problems

## Lily Pads: Modeling



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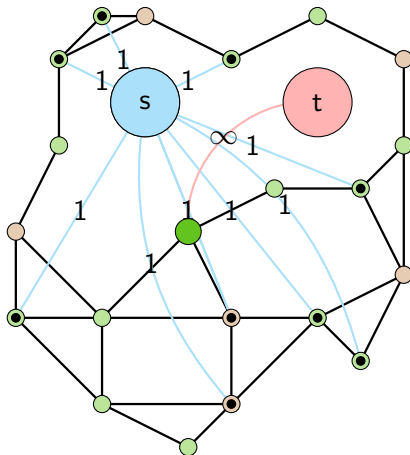
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# Modeling Flow Problems

## Lily Pads: Modeling



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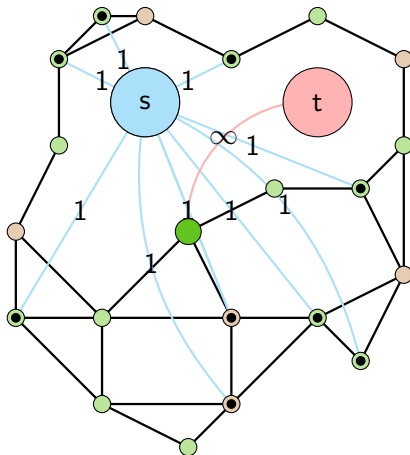
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# Modeling Flow Problems

## Lily Pads: Modeling



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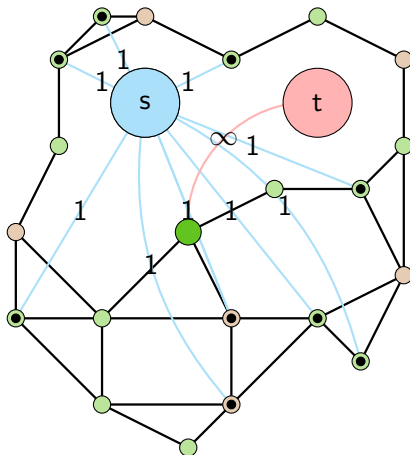
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Run `MaxFlow(s, t)`

● Lily ● Brittle ● Frog ● Lotus Flower

# Modeling Flow Problems

## Lily Pads: Modeling



● Lily ● Brittle ● Frog ● Lotus Flower

## Lily Pads: Idea

See Lily Pads as a Max Flow instance

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- ▶ Add super source  $\rightarrow$  Capacity 1 per frog
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Run  $\text{MaxFlow}(s, t)$

## Lily Pads: Solution

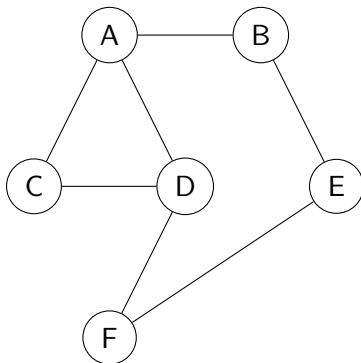
At most seven frogs can make it to the lotus flower.



# Matching

## Matching

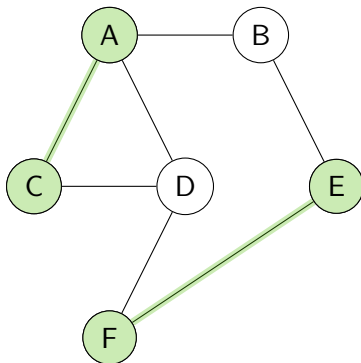
A matching of a graph  $G$  is a subset  $M \subseteq E$  such that no two edges in  $M$  share a common vertex.



# Matching

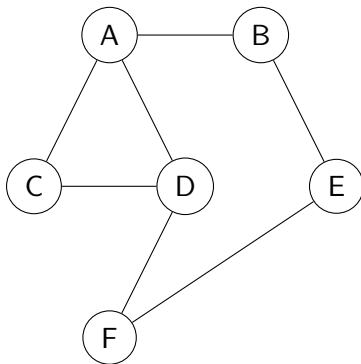
## Matching

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## Maximum Cardinality Matching

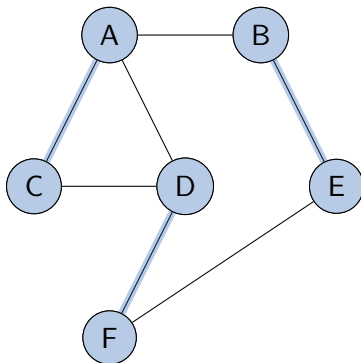
A matching  $M$  is called a Maximum Cardinality Matching if among all matchings  $M'$  of  $G$ ,  $|M| \geq |M'|$ .



# Matching

## Maximum Cardinality Matching

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# Matching

## Bipartite Matching

### Baby-Sitting

You are baby-sitting a bunch of kids. If you can, you try to keep them busy by themselves. How many kids can you have enjoy their favorite activities at the same time?

# Matching

## Bipartite Matching

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Tom

Bob

Amy

Anna

TV

Games

Blocks

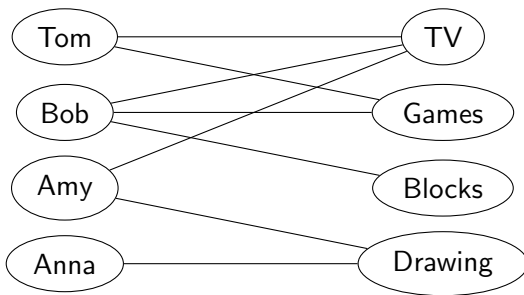
Drawing

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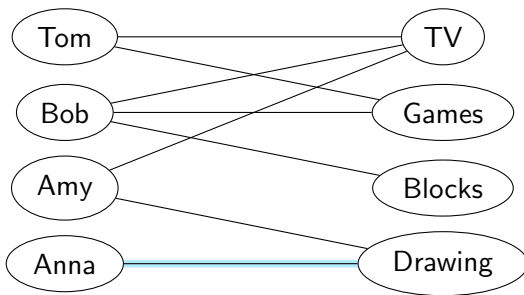


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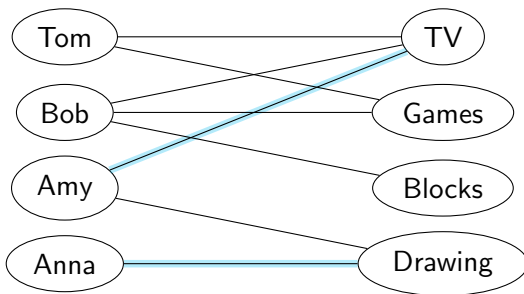


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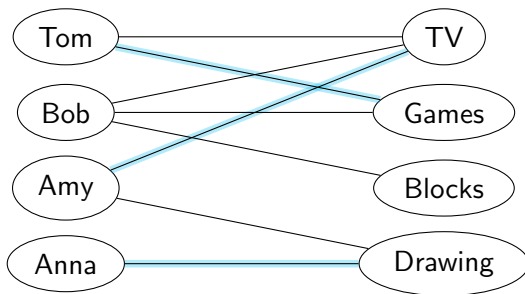


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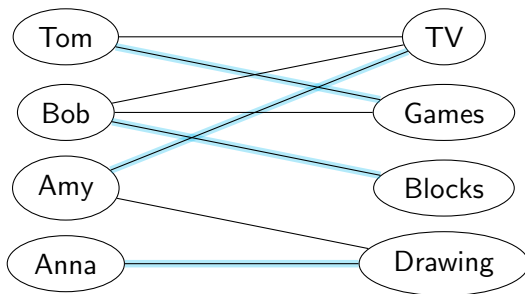


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# Matching

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### Maximum Cardinality Bipartite Matching

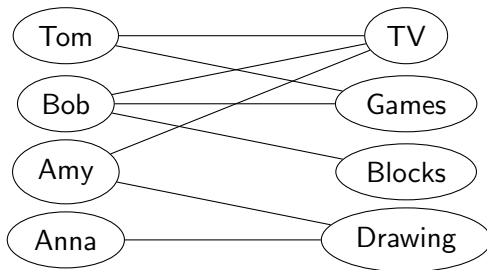
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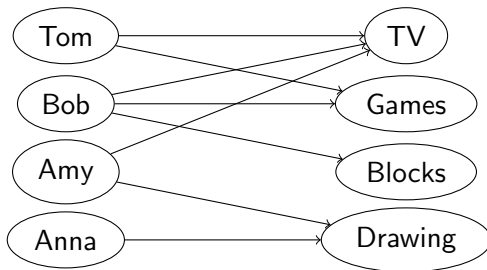


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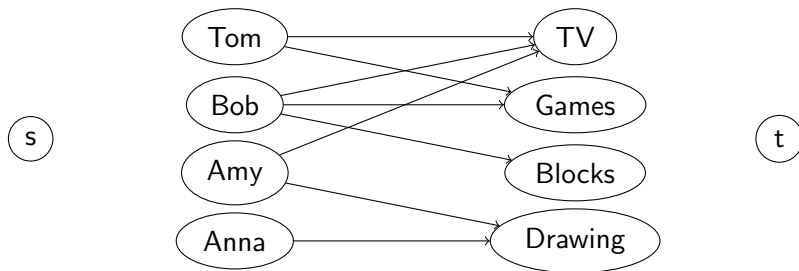


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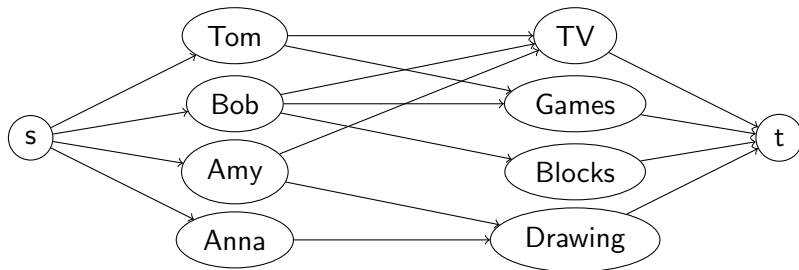


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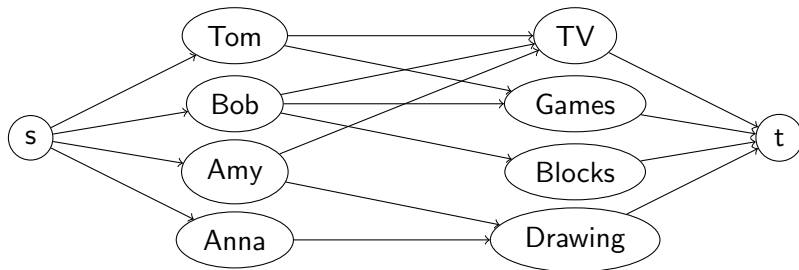
# Matching

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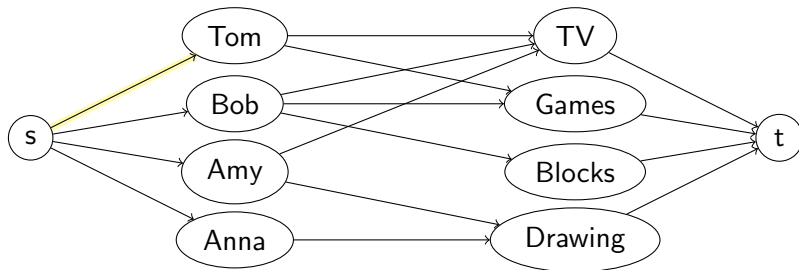
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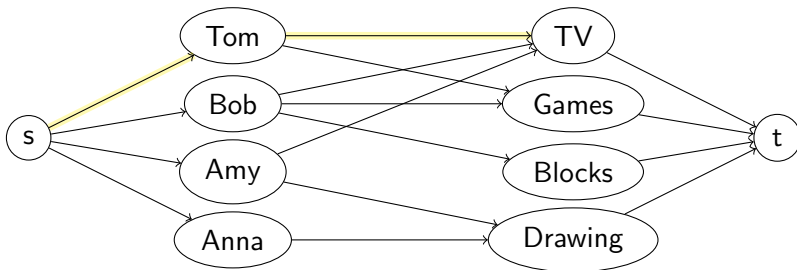


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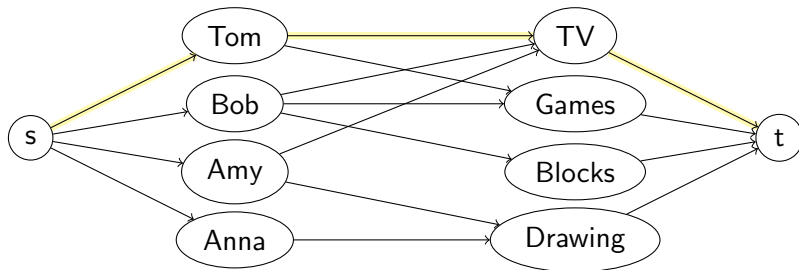
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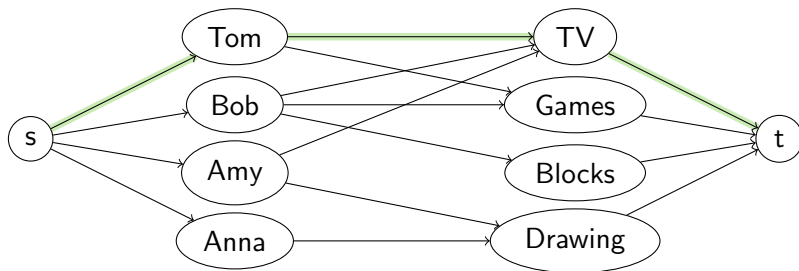


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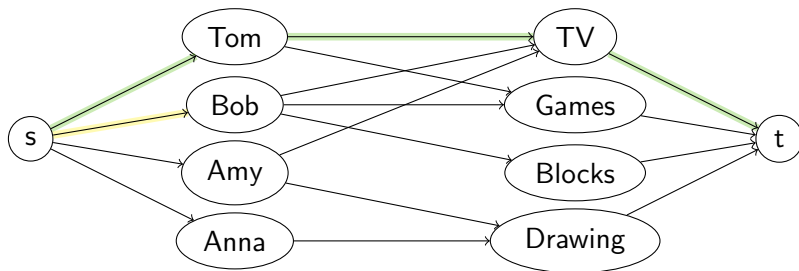


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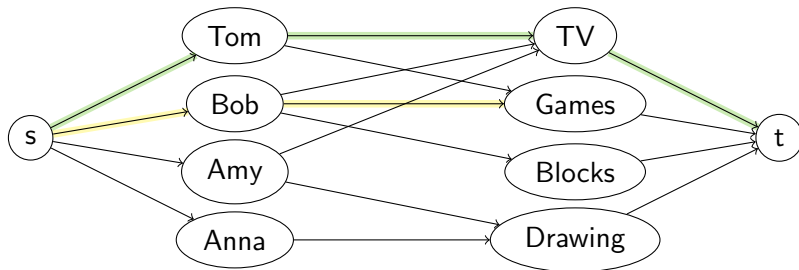
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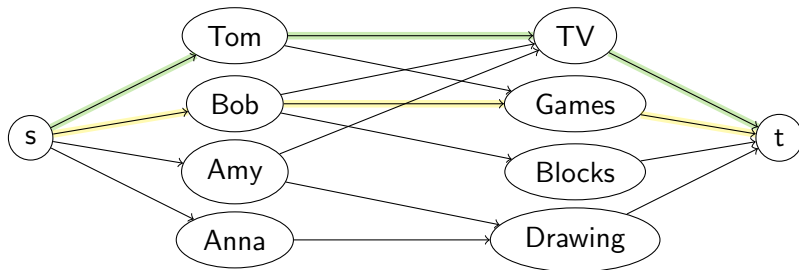


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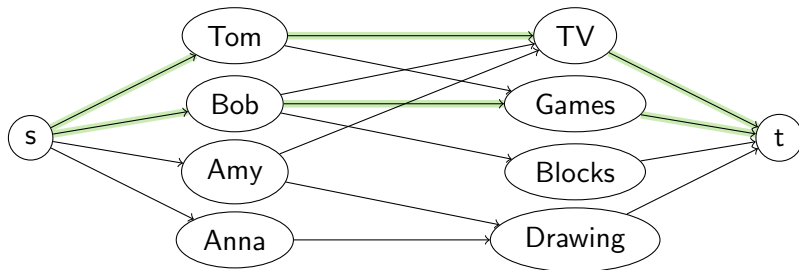


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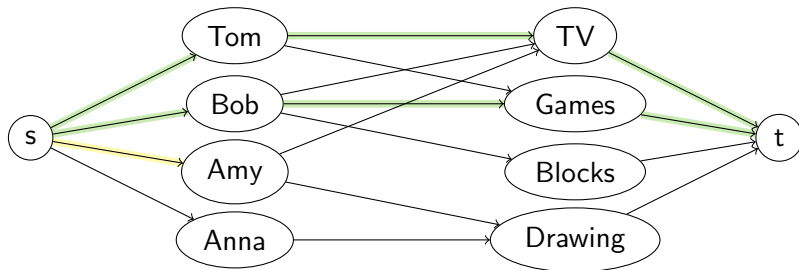


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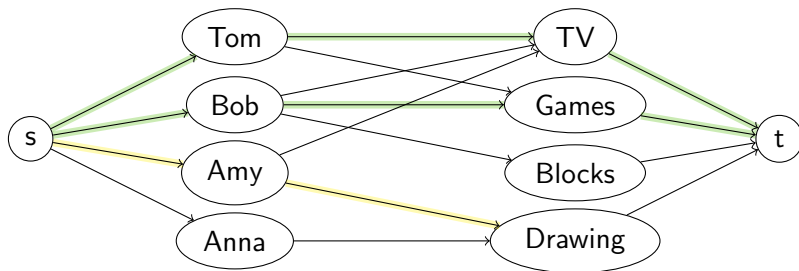


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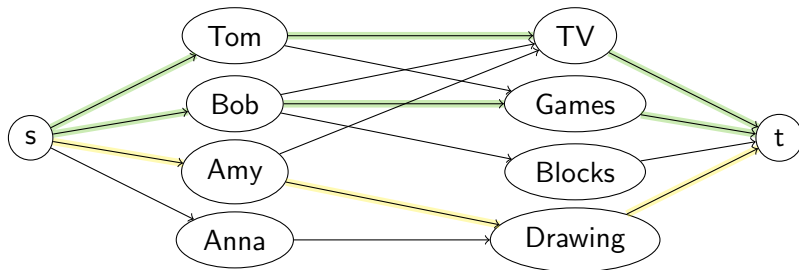
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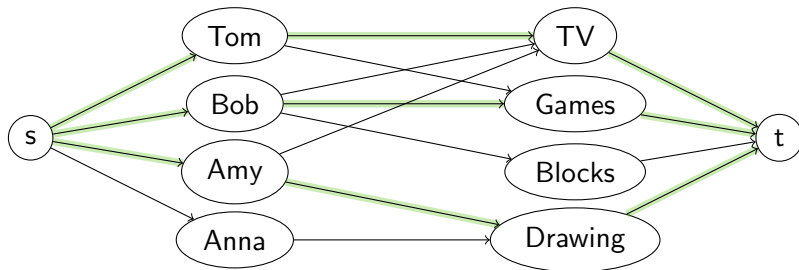


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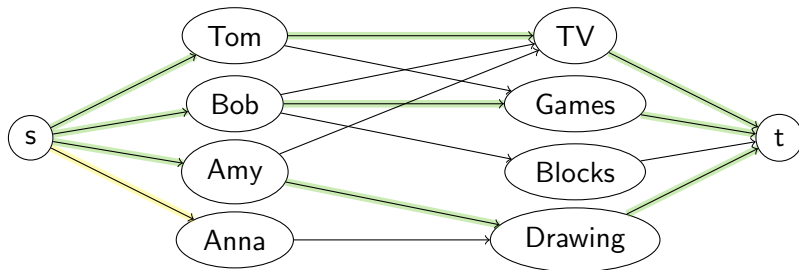
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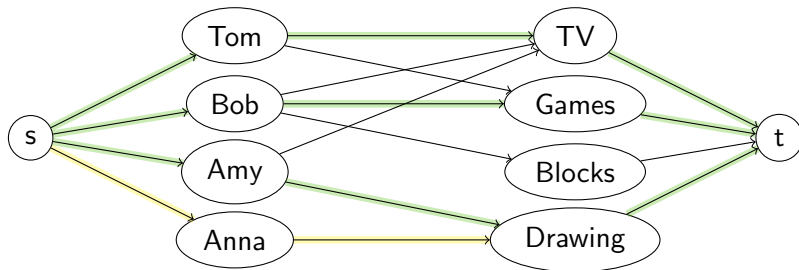


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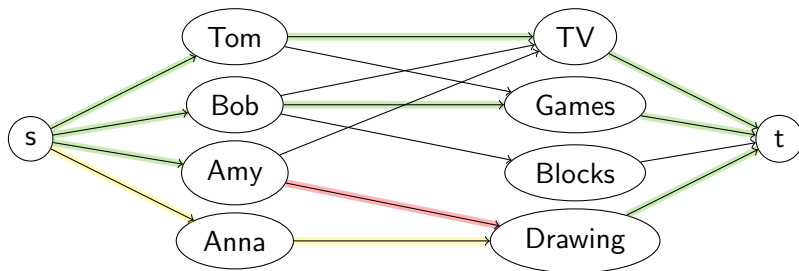
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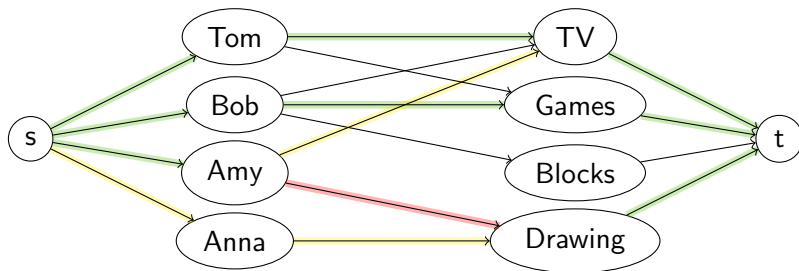


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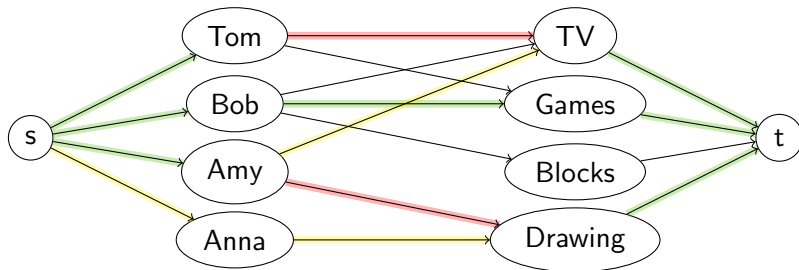


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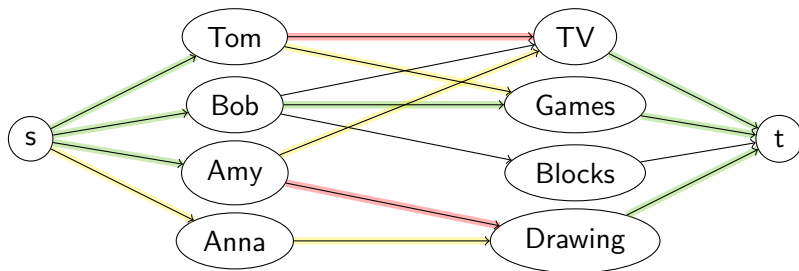
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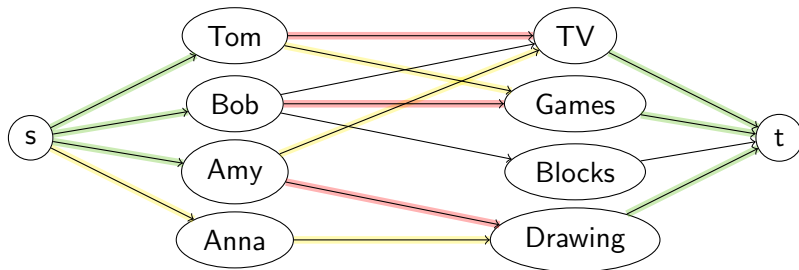


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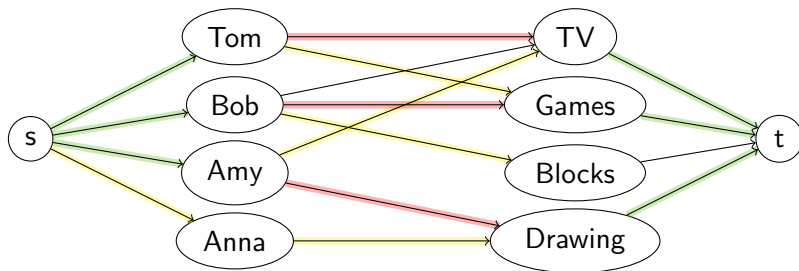


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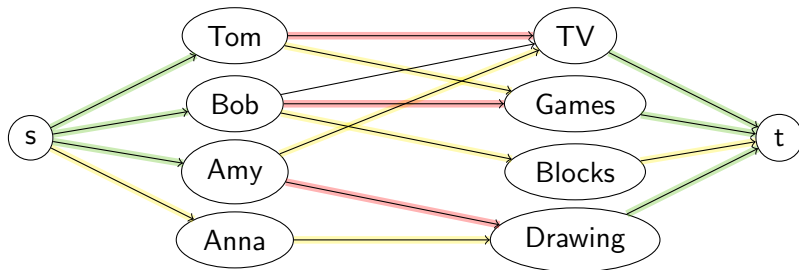
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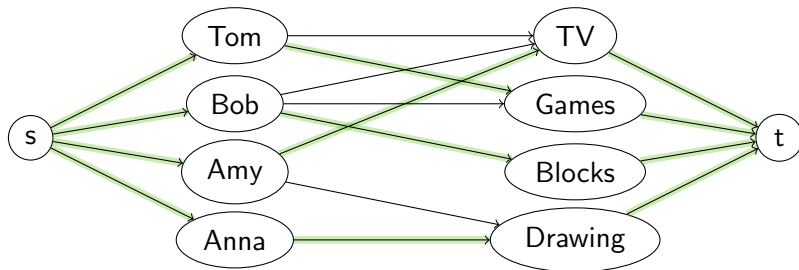
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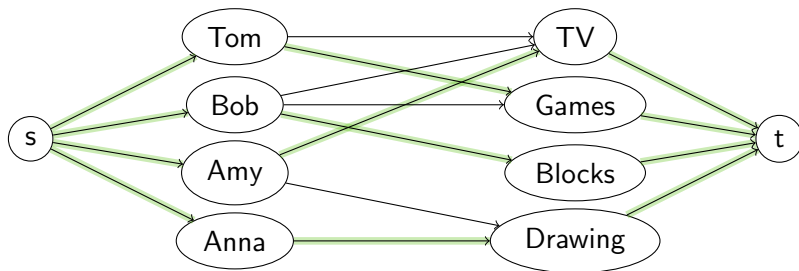
# Matching

## Bipartite Matching

### Problem: Baby-Sitting

How many kids can keep themselves busy?

→ The MCBM's size is equal to the Maximum Flow from  $s$  to  $t$





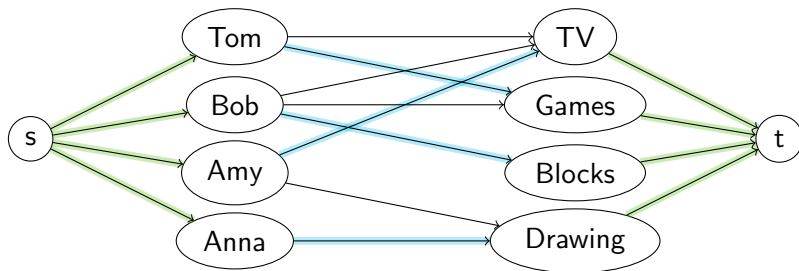
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# Matching

## Bipartite Matching

### Remark: Matchings

Assignment problems – in particular bipartite matching – are abundant in competitive programming.

- ▶ Numerous MCBM algorithms exist
- ▶ Solving MCBM with Max Flow is usually sufficient

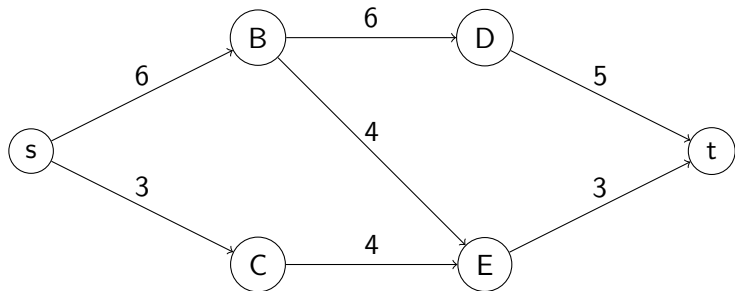
### Solving Flow Problems

- ▶ Determine that the problem is indeed a flow problem
- ▶ Model the flow network
- ▶ Run Edmonds-Karp's Algorithm

# Minimum Cuts

## Problem: Sabotage

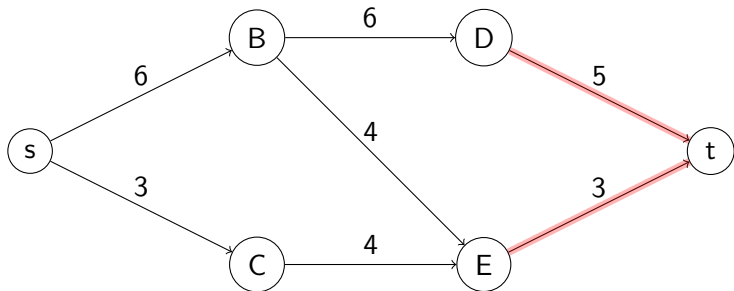
You want to prevent sabotages to your road system. In particular, you need to guarantee traffic from  $s$  to  $t$ . Some roads are easier to sabotage than others, thus every road has a sabotage cost  $c$ . Which roads would an attacker sabotage to stop traffic flow?



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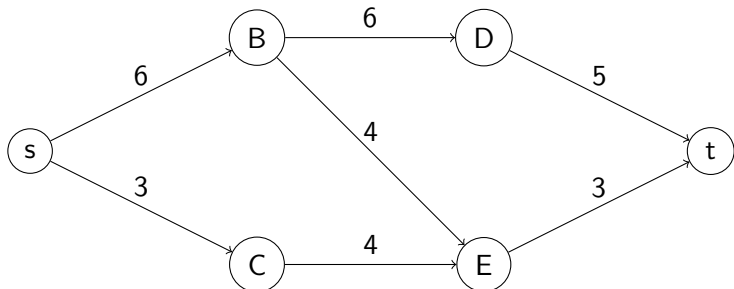
# Minimum Cuts

## Definition: $s$ - $t$ -Cut

Let  $s$  and  $t$  be two vertices of a graph. An  $s$ - $t$ -cut is a set of edges in a graph, whose removal disconnects  $s$  and  $t$ .

## Definition: Minimum $s$ - $t$ -Cut

In a weighted graph, an  $s$ - $t$ -Cut  $C$  is called a minimum cut if the sum of the edge weights of  $C$  is the minimum among all  $s$ - $t$ -cuts.



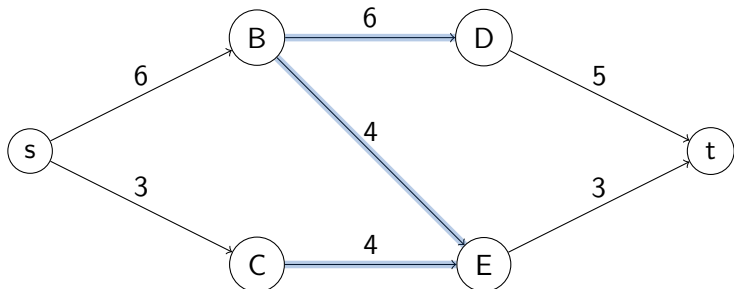
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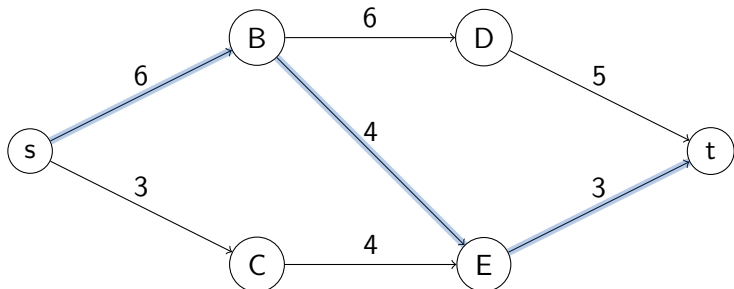
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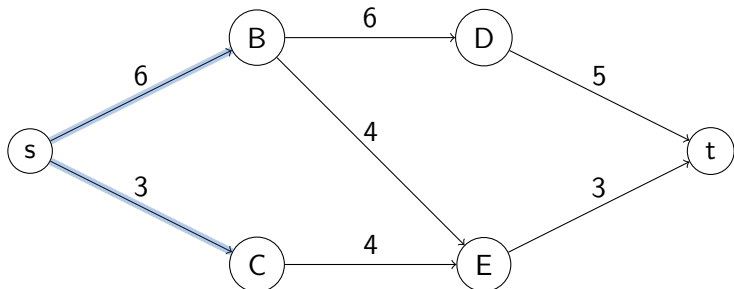
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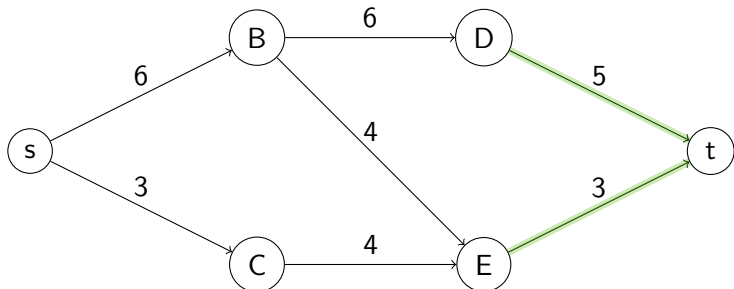
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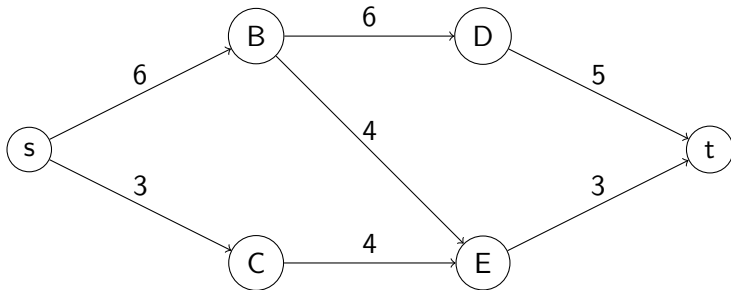
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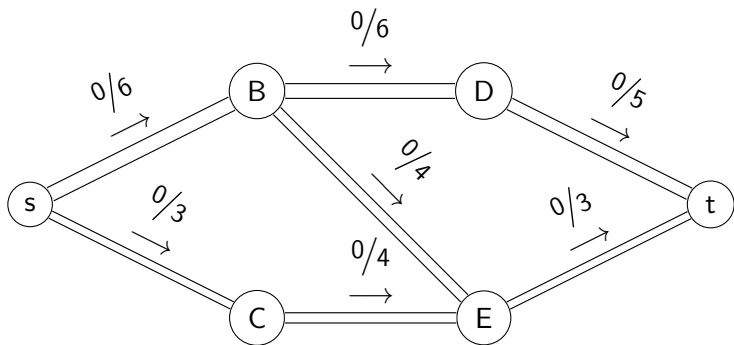
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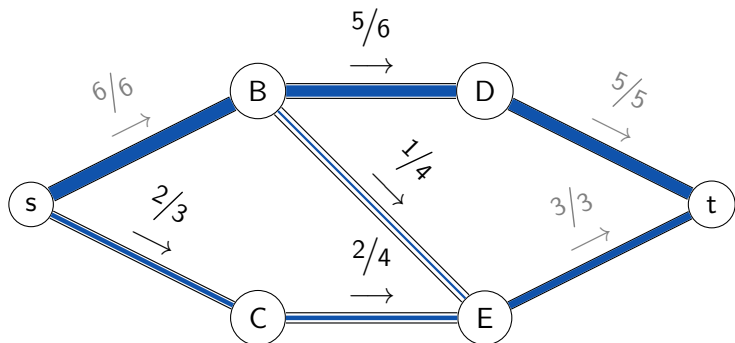
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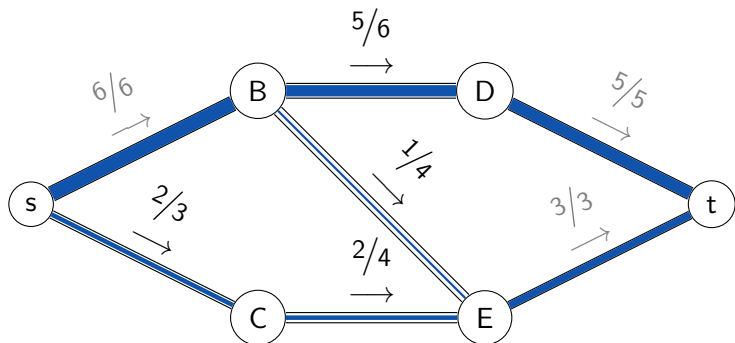
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# Minimum Cuts

## Finding Minimum Cuts

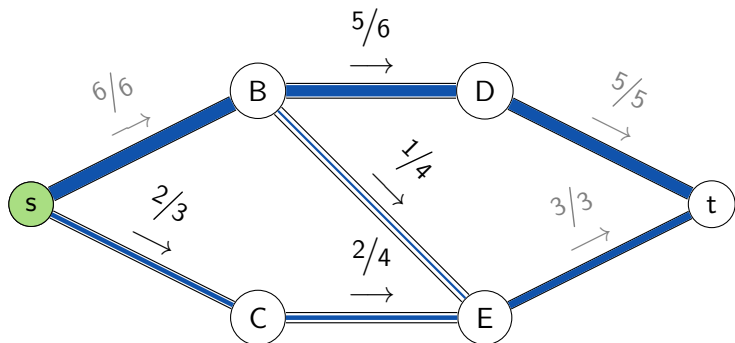
- ▶ Explore all vertices reachable from  $s$  using only edges with positive residual capacity (in the direction of traversal)
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# Minimum Cuts

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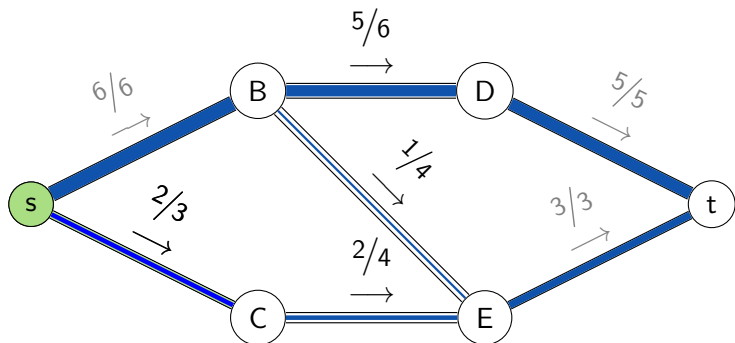
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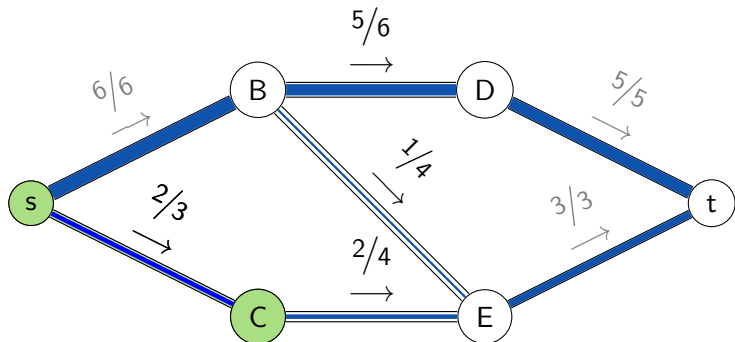
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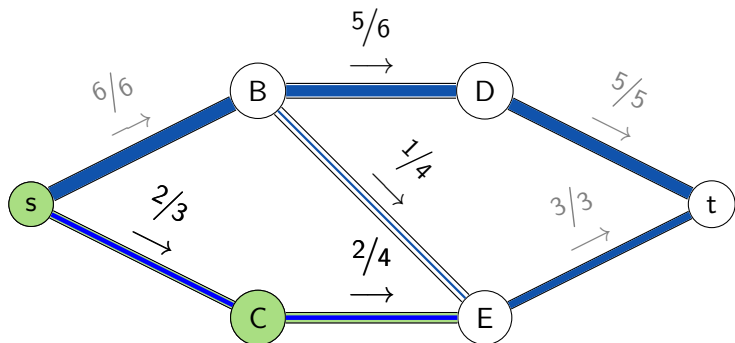




# Minimum Cuts

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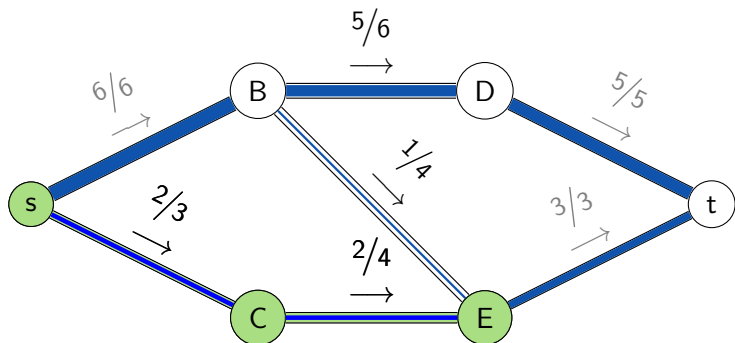
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# Minimum Cuts

## Finding Minimum Cuts

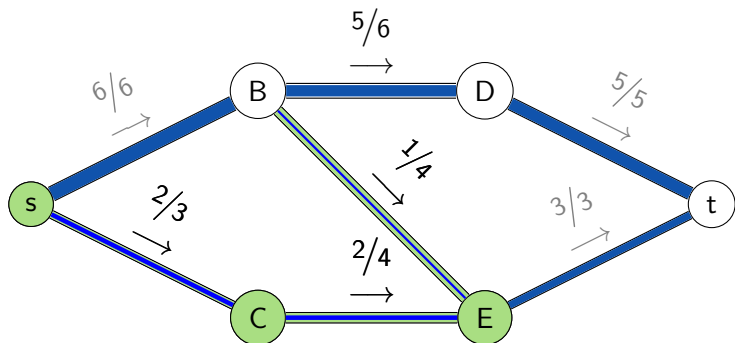
- ▶ Explore all vertices reachable from  $s$  using only edges with positive residual capacity (in the direction of traversal)
- ▶ All edges with zero residual capacity leaving this set are cut edges



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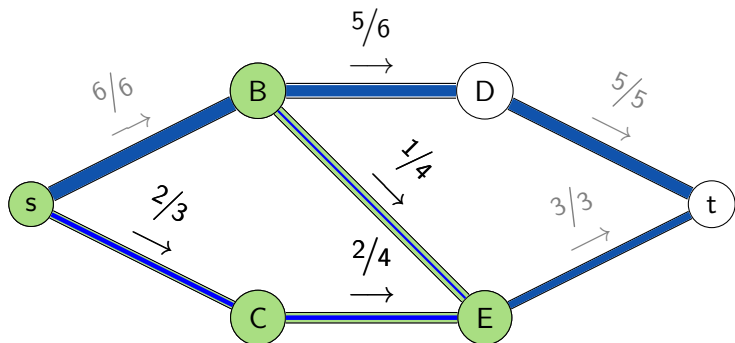
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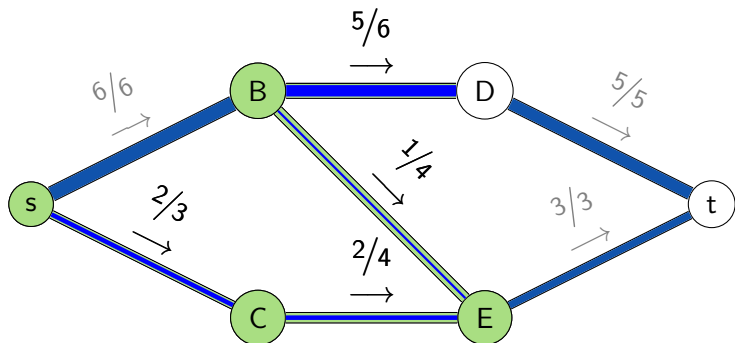
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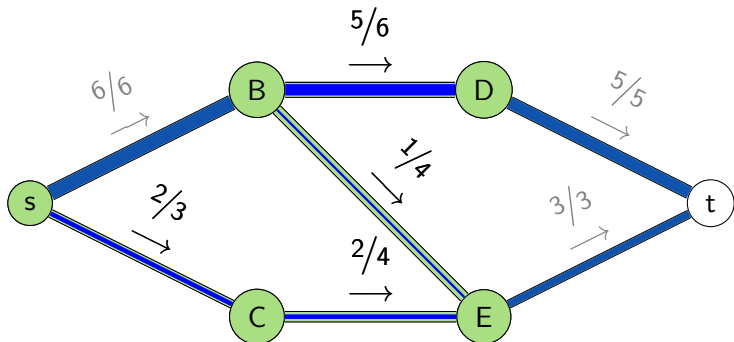
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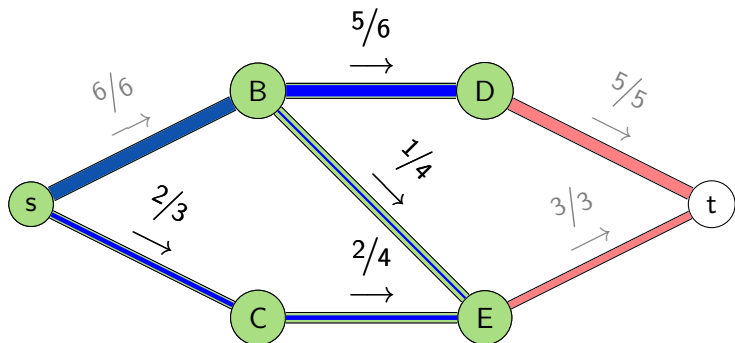
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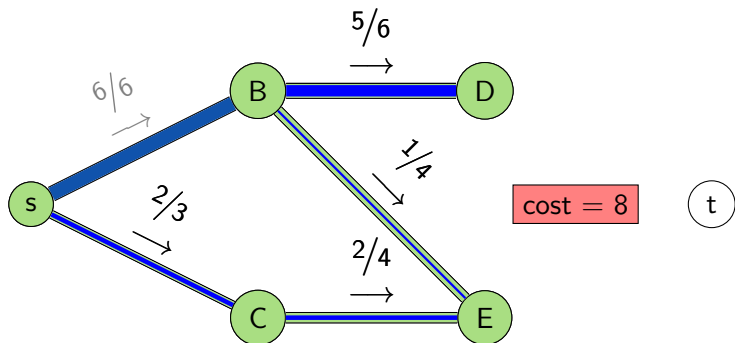
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# Flow Problems & Matching

## Today's Recap

We discussed

- ▶ Maximum Flow
- ▶ Algorithms of Ford-Fulkerson & Edmonds-Karp
- ▶ How to model Flow Networks
- ▶ Bipartite Matchings
- ▶ Minimum Cuts

# Conclusion

End of the graph lecture series. You are now familiar with

- ▶ Graph Traversals and Applications
  - ▶ DFS and BFS
  - ▶ Computing Toposort
  - ▶ Finding Bridges & Articulation Points
  - ▶ Finding (Strongly) Connected Components
  - ▶ Determining LCAs
  - ▶ Solving MaxFlow and Matching
- ▶ Shortest Path Algorithms
  - ▶ BFS for unweighted graphs
  - ▶ Dijkstra, Bellman Ford, SPFA
  - ▶ Floyd Warshall
- ▶ Special Types of Graphs
  - ▶ DAGs (recall Toposorts)
  - ▶ Bipartite Graphs
  - ▶ Eulerian Graphs
  - ▶ Planar Graphs
  - ▶ Trees and Spanning Trees
  - ▶ Flow Graphs (and how to model)