

**Lecture: Flow Networks** 

**Unit:** 7.4

**Instructor: Carlos C.L** 



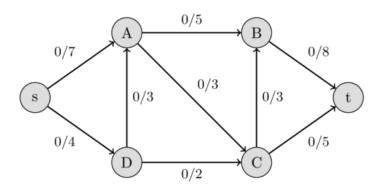
#### Flow network

A directed graph G(V, E) combined with a capacity function c:  $E \rightarrow \{R > 0\}$  and two vertices, a source(s) and a sink(t).

#### **Flow**

A function f:  $E \rightarrow \{R > 0\}$ . It must satisfy the following conditions:

$$f(e) \le c(e)$$
  
  $\Sigma f((u, v)) = \Sigma f((v, u))$  for any node v



We can prove the following holds given this conditions:

$$\Sigma$$
 f((s, u) =  $\Sigma$  f((u, t)) = flow of the network

# Ford-Fulkerson (published 1956)





Our problem is to find the flow function f, that maximizes the flow of the network.

# **Residual Graph**

The residual graph is the same graph but with reverse edges and a different capacity function, the residual capacity.

# **Residual capacity**

How much we can increase the flow for a given edge:

$$rc(e) = c(e) - f(e)$$

# **Reverse edges**

For any edge (u, v) in the original graph, we define a reverse edge (v, u) in the residual graph

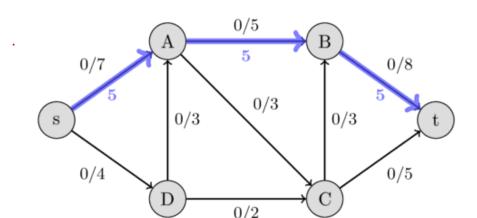
$$f((v, u)) = -f((u, v))$$
  
 $c((v, u)) = 0$   
 $rc((v, u)) = f((u, v))$ 

### **Augmented path**

It's a path on the residual graph with all edges having positive residual capacity.

# **Bottleneck capacity**

The lowest residual capacity on an augmented path.

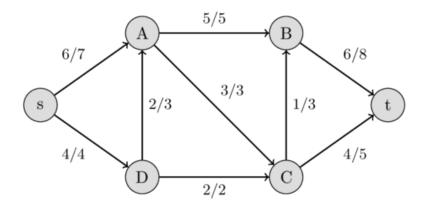


# **Algorithm**

- 1. Find augmented path from s to t (use BFS or DFS)
- 2. Find bottleneck capacity = bc
- 3. Augment each edge on the path

4. Repeat until no augmented paths can be achieved

#### Solution



# **Edmonds Karp**

If we use BFS for step one then we are solving with Edmonds Karp.

### **Complexity:**

Time: O(EF), E = #Aristas, F = Max flow

Time Edmons Karp: O(VE^2)
Space Edmons Karp: O(V+E)

# **Problems:**

**CSES Download Speed** 

**Dinic's Algorithm** 

Solves max flow problem.

<u>Dinic's Algorithm | Network Flow | Graph Theory</u>

# **Layered Network**

A layered network is a new graph network constructed in the following way:

- 1. Build residual network
- 2. **Build level array**level[v] = min distance (unweighted) from s to v using only edges with rc(e) > 0
- 3. Only keep edges where (u, v)

$$level[v] = level[u]+1$$

# Augmenting path on a layered graph

Notice an augmenting path on a layered graph will satisfy: A path e1, ...., e\_k on a residual graph where:

- rc(e\_i)w > 0 for all i
- level[e\_i+1] = level[e\_i]+1 for all i < k</p>

# **Blocking flow**

A flow function, where any path from s to t contains and edge where rc(e) = 0

# **Algorithm**

1. Build level array with BFS

level[v] = min distance (unweighted) from s to v using only edges with rc(e) > 0

- 2. If sink not reached while doing level array return max flow
- 3. Find disjoint (do not share edges) augmenting paths from s to t until a blocking flow is reached, for each augmenting path update flow with bottleneck capacity (bc):

4. Repeat step 1

# Complexity

Time complexity: O(V^2E)

#### **Proof**

We first need to prove the following:

level\_i+1[t] > level\_i[t]

Since level\_i[t] cannot be greater than V, the algorithm must terminate in less than V iterations.

In each iteration we will have to check at most all edges E to find the

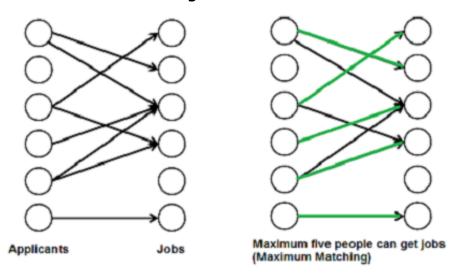
#### For unit networks

A unit network is a network where all edges have unit capacity.

Time complexity: O(V^1/2E)

# **Unweighted Bipartite Matching**

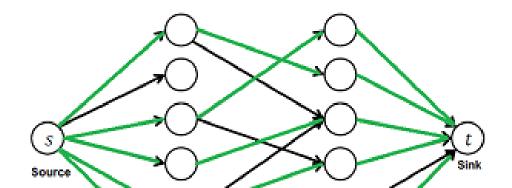
We have a bipartite graph representing applicants and jobs. There is an edge (applicant, job) if the applicant has the requirements to do the job.



We want to maximize the number of applicants that get a job.

#### Idea

- 1. Transform problem to max flow problem
- 2. Make an imaginary source and target, connect source to applicant and jobs to applicants
- 3. Find max flow in this graph





The maximum flow from source to sink is five units. Therefore, maximum five people can get jobs.

#### **Problems:**

**CSES School Dance** 

# **Matching Problems**

Common matching variations  Easier Harder								
Easi	ler	Bipartite	Non-Bipartite					
	Unweighted Edges	<ul> <li>Max flow algorithms</li> <li>Repeated augmenting paths with dfs</li> <li>Hopcroft-Karp</li> </ul>	• Edmond's blossom algorithm					
	Weighted Edges	<ul> <li>Min cost max flow algorithms</li> <li>Hungarian algorithm (perfect matching)</li> <li>LP network simplex</li> </ul>	• DP solution for small graphs					
Hard	ler							

#### **Resources:**

<u>Unweighted Bipartite Matching | Network Flow | Graph Theory</u>

# Assignment problem

# First Example:

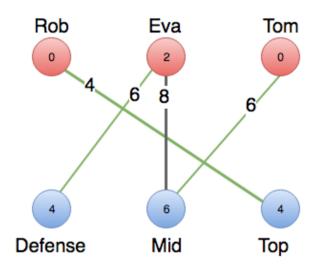
There are N jobs and N workers. Some jobs can only done by some workers, and we know the cost it takes for the worker to do the job. Make a 1:1 assignment (each worker gets assigned only one job and each job gets assigned only one worker) so that the total cost is minimized.



V4	0	0	1	1	1	0	V40 W4
<b>V</b> 5	0	0	0	1	1	1	V50 W5
<b>V</b> 6	0	0	0	1	1	0	V6 0 0 W6

# **Second Example:**

A football coach wants to assign a position to each of its players, he knows the skill of each player for the positions. Make a 1:1 assign (each player has only one position and each position has only one player) so that the total skill is maximized.



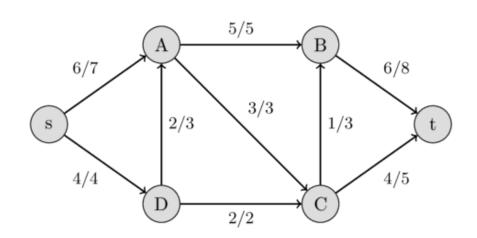
#### Min Cost Max Flow Problem

Suppose we have a network flow, G(V, E, capacity) and a cost function, cost:  $E \rightarrow \{R > 0\}$ .

cost(e) = cost per unit flow for edge e

Total  $cost(G) = \Sigma cost(e)*flow(e)$ 

We want to find the max flow in this network and the lowest possible total cost to achieve this flow.



# **Algorithm:**

- 1. Find augmented path from s to t with the lowest possible cost (use Bellman Ford or SPFA)
- 2. Find bottleneck capacity = bc
- 3. Augment each edge on the path

4. Repeat until no augmented paths can be achieved

# Solve Initial problem using Min cost Max Flow:

 What nodes should we create and how should we connect them to the graph?

# **Complexity:**

Time: O(V^5)

# **Hungarian Algorithm**

It can solve the assignment problem on O(V^3)

# **Problems:**

**CSES Police Chase** 

#### **Resources:**

<u>Brilliant – Matching Problem</u>

# **Topological Sort**

### **Motivation**

Suppose we are given a set of classes and an ordering in which the classes must be taken, this is, a number of pairs (x,y) where:

(x, y) implies x must be taken before y Return any valid ordering (permutation of courses) that is valid, this is, we can take the courses following that

# ordering.

Unsorted graph

Topologically sorted graph

0 6 1 5 4 2 3

# **Algorithm**

1. Build degree array:

degree[i] = #Nodes pointing to i

- 2. Create queue, which will contains nodes with degree[v] = 0
- 3. Pop node u from queue and update degree for its child's: degree[v] -= 1 for all v child's of u
- 4. Add all nodes with degree 0 to the queue.
- 5. Repeat from step 3 until there are no nodes left

#### **Problems**

<u>Leetcode Course Schedule</u> <u>CSES Game Routes</u>