

So, we have a pseudo-poly time algo for Max Flow.  
How do we design a polynomial time algorithm?

One algo that achieves polynomial running time is Edmonds-Karp

MaxFlowEK( $G$ ):

Set  $f[e] = 0 \quad \forall e \in E$

While  $P = \text{FindShortestPath}(s, t, \text{Residual Graph}(G, f))$ :

$f = \text{augment}(f, P)$   
    Update Residual( $G, f$ )

return  $f$

- So, what is the difference between FF and EK?

- FindPath  $\rightarrow$  Find Shortest Path

- the only difference is the selection of the augmenting path

- Why does this lead to a different runtime analysis?

• There is an explicit bound ( $mn$ ) on the total # of bottleneck edges as EK runs. This limits the total number of iterations of the while loop and the overall run time.

Theorem: EK makes at most  $m\Delta$  iterations of the while loop

Proof: Consider laying out the vertices of  $G$  in layers according to a BFS from  $s$ , and let  $t$  be at level  $d$ .

Keeping the layout fixed, consider the sequence of paths found in the resulting residual graph. If a shortest path uses only forward edges, each iteration will cause at least 1 forward edge to be saturated and removed from  $G_f$ , and only backward edges will be added.

This means that  $d$  does not decrease, and as long as  $d$  has not increased (so that only forward edges are being used), at least one forward edge is eliminated per iteration.

Forward edges will be removed at most  $m$  times before either (1)  $d = \infty$  (graph is disconnected and EK terminates) or (2) A path with a non-forward edge is used (so that  $d$  increases by at least 1). We can re-layout  $G$  and apply the same argument again. This shows that the  $s$ - $t$  distance of the selected path never decreases. Further, it increases by at least 1 every (at most)  $m$  iterations. The minimum path length cannot increase beyond  $\Delta$ . This implies we have  $\leq m\Delta$  iterations of the while loop.

Running Time:

How long does each iteration take?

- We can identify the shortest path using BFS ( $O(m+n)$ ) time. However, since we assume every vertex has at least one incident edge,  $n \leq 2m$  and  $O(m+n) = O(m)$
- Given the path  $P$ , augment takes  $O(m)$  time
- Thus, each iteration takes  $O(m)$  time

EK takes  $O(m \cdot mn) = O(m^2n)$  time to find a maximum flow.

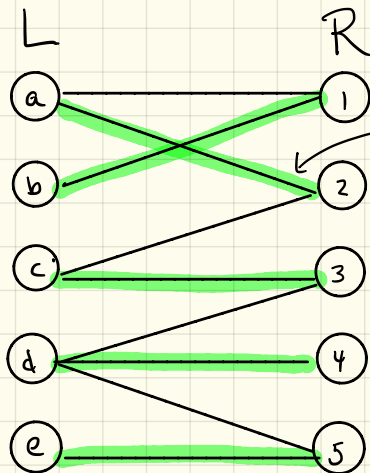
Note: This algorithm's runtime depends only on the specification of the graph and not the edge capacities. It is what we call a "strongly-polynomial" algorithm.

Now, let's look at how to use network flow to solve other interesting problems.

## - Bipartite Matching

E.g. Set of people (L) and jobs (R)

- Each person is only qualified for some subset of jobs
- Each job is done by at most 1 person
- Can model this as a bipartite graph



Here, edges highlighted in green constitute a maximum matching.

- a matching gives an assignment of people to jobs
- Want to accomplish as many tasks as possible
- Each person is assigned 1 job
- So we want a maximum matching (containing as many edges as possible)

## Problem: Maximum Bipartite Matching

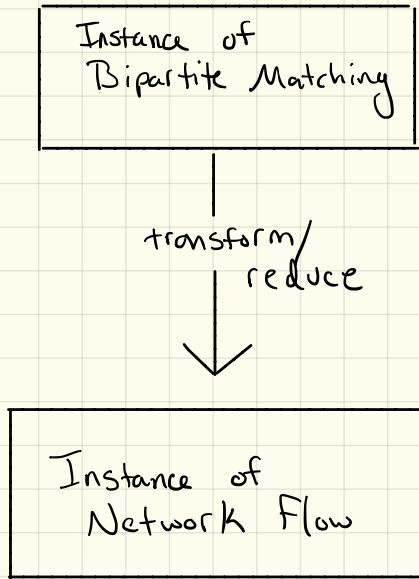
Given a bipartite graph  $G = (A \cup B, E)$ , find a matching  $M \subseteq E$ , such that any node appears in at most one edge in  $M$ , and such that  $M$  is as large as possible.

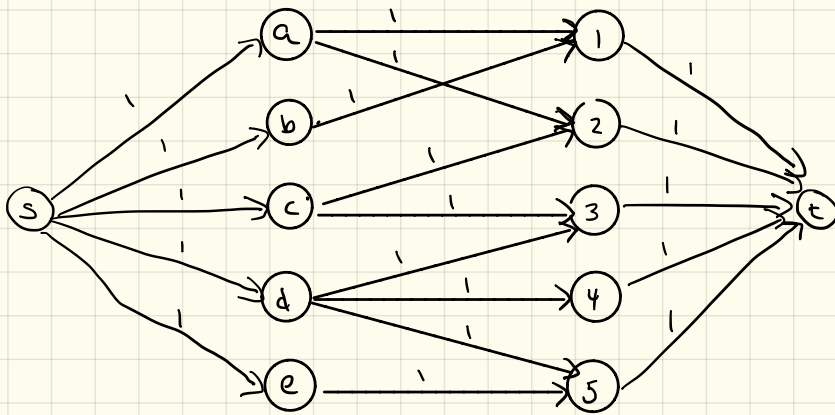
### Note:

- We are given the bipartition; no need to find it
- $S$  is a perfect matching if every vertex is matched
- Maximum is not maximal  $\Rightarrow$  a greedy algo will give a maximal matching.

## Key Concept: Reduction

- Given an instance of Maximum Bipartite Matching
- Create an instance of network flow such that the solution of the flow problem can be easily used to find the solution to the bipartite matching problem.





Transformation:

- 1) Given bipartite graph  $G = (A \cup B, E)$ , direct edges from  $A$  to  $B$
- 2) Add new vertices  $s$  and  $t$
- 3) Add an edge from  $s$  to every vertex in  $A$
- 4) Add an edge from every vertex in  $B$  to  $t$
- 5) Make all edge capacities 1
- 6) Solve Max Flow on this graph  $G'$

Claim: The edges used in the Max Flow will correspond to the largest possible matching.

## Important Notes:

- Since capacities are integers, flows will be integral
- Since capacities are all 1, every edge is used completely or not at all
- If  $M$  is the set of edges from  $A$  to  $B$  we use then
  - 1)  $M$  is a matching
  - 2)  $M$  is the largest possible matching

Theorem: The  $A \rightarrow B$  edges of our flow constitute a maximum bipartite matching in  $G$

Lemma:  $M$  is a matching

Proof: We can choose at most 1 edge leaving any node in  $A$  and at most 1 edge entering any node in  $B$ . Otherwise, we would not satisfy the balance constraints, and our solution wouldn't be a valid flow!

Lemma:  $M$  is of maximum size

Proof: If there is a matching of  $k$  edges, there is a flow of value  $k$ :  
-  $f$  has 1 unit along each of the  $k$  edges,  $\leq 1$  unit leaves and enters every node but  $s$  and  $t$ .



If there is a flow of value  $k$ , there is a matching with  $k$  edges

- We find a maximum flow with (say)  $k$  edges.

- This corresponds to a matching of  $k$  edges

- If there were a matching with  $k' > k$  edges, we would have found a flow with a value  $> k$ , contradicting that  $f$  was maximum.

- Hence,  $M$  is maximum.

## Running Time?

- Consider the FF algo. The runtime is bounded by  $O(m'C)$  where  $m'$  is the # of edges and

$$C = \sum_{e \text{ out of } S} c_e$$

-  $C = |A| = n$

- the # of edges in  $G'$  is equal to the number of edges in  $G$  ( $m$ ) plus  $2n$

- So, the running time is  $O((m+2n)n) = O(mn+n^2) = O(mn)$

This leads immediately to:

Theorem: We can find a maximum bipartite matching on a graph with  $n$  vertices and  $m$  edges in  $O(mn)$  time.