

1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100

# Algorithms

[illegible][illegible]

# Algorithms

## READING



Network Flow: Sections 7.1-7.3

Network Flow Applications: Sections 7.5, 7.7

Randomized Algorithm: Section 12.2

**Suggested.** As many exercises as you can do from  
Chapter 7 of the textbook.

# Recitations

Thu, Apr 27 2023- Fida (OH by appt)

2:30 PM - 3:45 PM

Location Instructions

West Admin (A3-002)

Tue, May 9 2023- Shan

2:30 PM - 3:45 PM

Location Instructions

West Admin (A3-002)

Thu, May 11 2023- Fida (send questions before May 8)

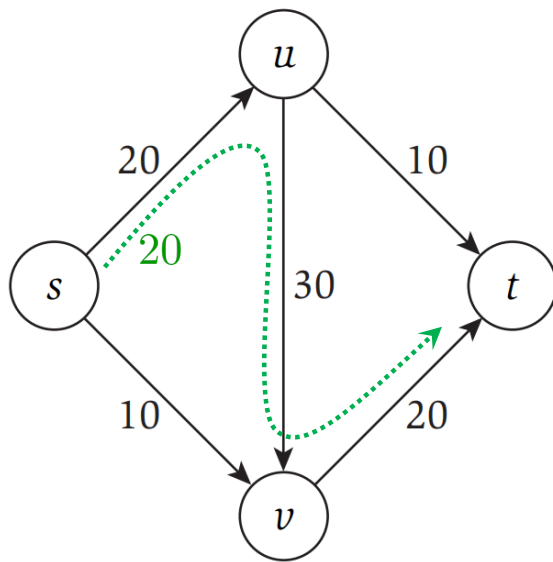
Class time

Class location

# NETWORK FLOWS

## Ford Fulkerson Algorithm.

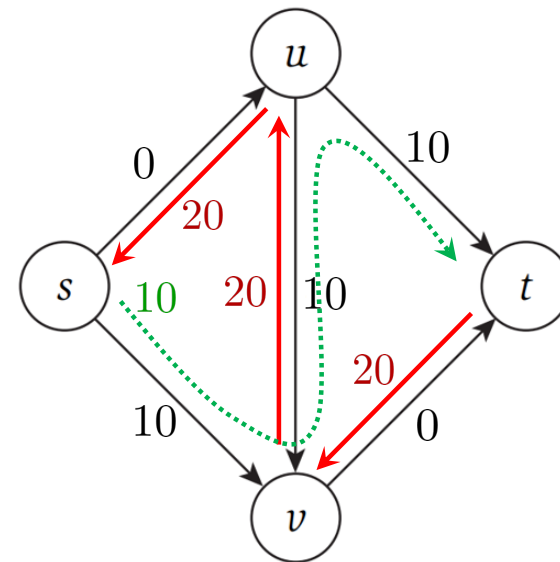
Repeatedly find a path in  $G_f$  from  $s$  to  $t$  in which all edges have some leftover capacity and send flow equal to the capacity of the *bottleneck edge*.



send 20 units of flow  
along  $s \rightarrow u \rightarrow v \rightarrow t$



## RESIDUAL GRAPH

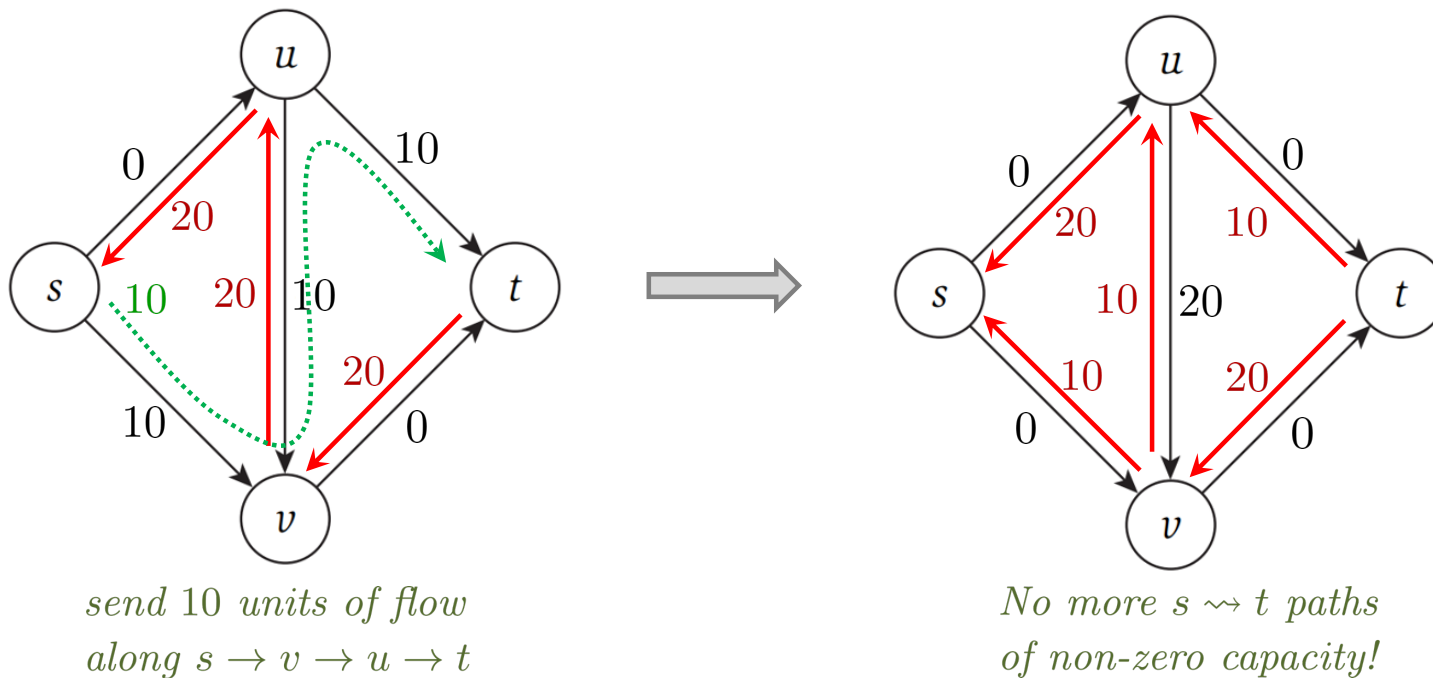


send 10 units of flow  
along  $s \rightarrow v \rightarrow u \rightarrow t$

# NETWORK FLOWS

## Ford Fulkerson Algorithm.

Repeatedly find a path in  $G_f$  from  $s$  to  $t$  in which all edges have some leftover capacity and send flow equal to the capacity of the *bottleneck edge*.

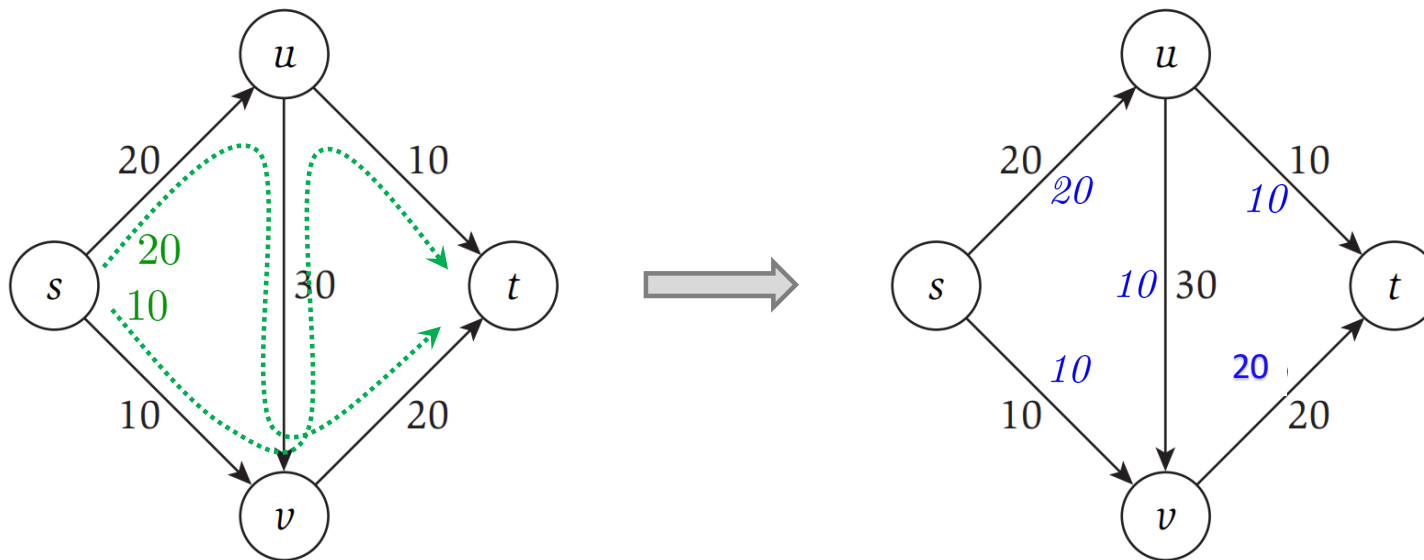


Total flow: 30 units.

# NETWORK FLOWS

## Ford Fulkerson Algorithm.

Repeatedly find a path in  $G_f$  from  $s$  to  $t$  in which all edges have some leftover capacity and send flow equal to the capacity of the *bottleneck edge*.



The paths along which we sent flows are called *augmenting paths*.

# NETWORK FLOWS

Max-Flow

```
Initially  $f(e) = 0$  for all  $e$  in  $G$ 
While there is an s-t path in the residual graph  $G_f$ 
    Let  $P$  be a simple s-t path in  $G_f$ 
     $f' = \text{augment}(f, P)$ 
    Update  $f$  to  $f'$ 
    Update the residual graph  $G_f$  to be  $G_{f'}$ 
EndWhile
Return  $f$ 
```

FLOW VALUE IS 0

IF A PATH IS FOUND,  
AUGMENT IS CALLED

RESIDUAL GRAPH IS  
UPDATED

$\text{augment}(f, P):$

```
Let  $b = \text{bottleneck}(P, f)$ 
For each edge  $(u, v) \in P$ 
```

\*\*Here we are calculating the flow being passed on each edge. So for backward edges  $u \rightarrow v$ , we need to decrease the flow of  $v \rightarrow u$  by  $b$ .

```
    If  $e = (u, v)$  is a forward edge then
```

```
        increase  $f(e)$  in  $G$  by  $b$ 
```

```
    Else  $((u, v)$  is a backward edge, and let  $e = \begin{pmatrix} (v, u) \\ \underline{u, v} \end{pmatrix}$ )
```

```
        decrease  $f(e)$  in  $G$  by  $b$ 
```

```
    Endif
```

```
Endfor
```

```
Return( $f$ )
```

SAY A FLOW WITH VALUE  $f$  IS PRODUCED

WHAT IS AN UPPER BOUND FOR THE NUMBER PATHS FOUND?

CALCULATES BOTTLENECK

SENDS FLOW EQUAL TO  
BOTTLENECK ON ALL PATH  
EDGES

FORWARD EDGES INCREASE  
FLOW, BACKWARD EDGES  
DECREASE THE FLOW.

## NETWORK FLOWS

**Claim.** If all edge capacities are (non-negative) integers, then the Ford Fulkerson algorithm yields an optimal solution and runs in  $O((m + n) \cdot f)$  time, where  $f$  is the value of the flow returned.

*Proof.* In every iteration, the algorithm sends at least one additional unit of flow.

The algorithm therefore has  $\leq f$  iterations.

In any iteration, we can find an augmenting path in  $O(m + n)$  time using BFS/DFS on the residual graph.

This implies that the running time is  $O((m + n) \cdot f)$ .

WHY OPTIMAL?

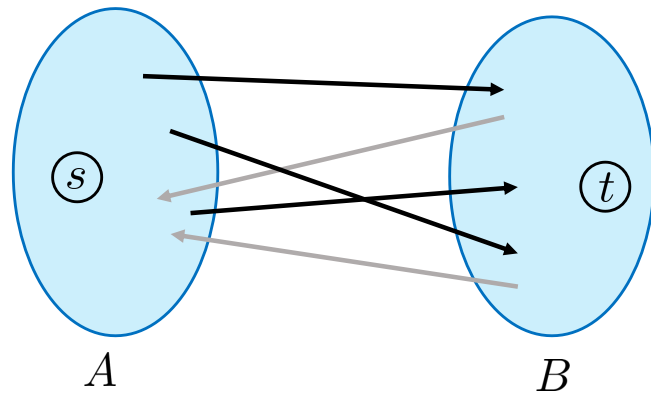


# NETWORK FLOWS

## TWO OBSERVATIONS

# NETWORK FLOWS

An **s-t cut** in the graph is a partition  $(A, B)$  of the vertices into two sets  $A$  and  $B$  s.t.  $s \in A$  and  $t \in B$ .



The capacity of the  $s$ - $t$  cut, denoted  $c(A, B)$  is the sum of the capacities of the edges going *from  $A$  to  $B$* .

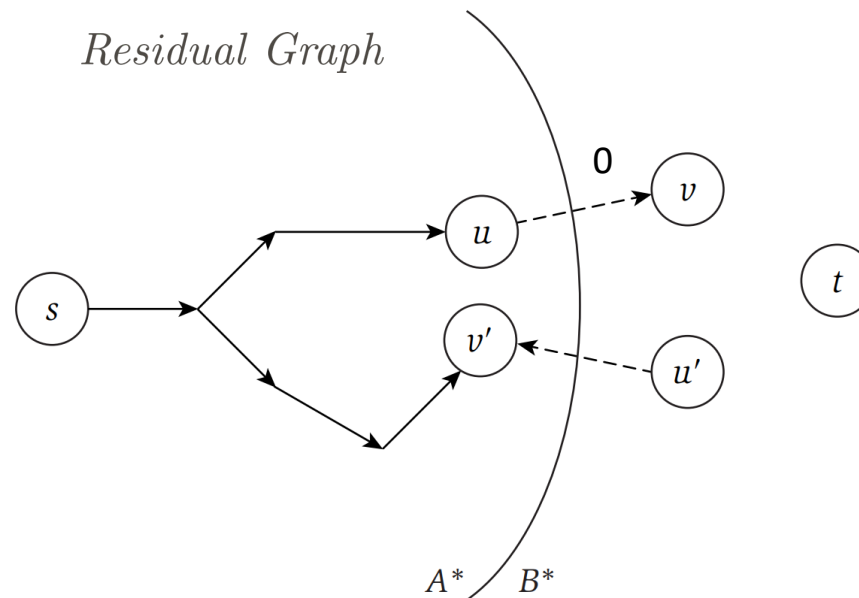
*Observation.* No flow can have value more than the capacity of an  $s$ - $t$  cut.

# NETWORK FLOWS

Consider the residual graph corresponding to a maximum flow  $f$ .

Let  $A^*$  be the set of all nodes  $a$  in  $G$  s.t. there is an  $s \rightsquigarrow a$  path in  $G_f$   
i.e. all edges have non-zero (residual) capacity.

Let  $B^*$  be the set of the remaining vertices.  $B^* = V - A^*$



*Observation.* Value of  $f$  = Capacity of the  $s$ - $t$  cut  $(A^*, B^*)$ .

# NETWORK FLOWS

We have made the following observations.

*Observation.* No flow can have value more than the capacity of an  $s$ - $t$  cut.

*Observation.* Corresponding to a maximum flow  $f$ , there is an  $s$ - $t$  cut  $(A^*, B^*)$   
s.t.  $\nu(f) = c(A^*, B^*)$ . *value of  $f$  = capacity of  $(A^*, B^*)$*

These two together imply:

**Max flow = Min Cut**

The maximum value of a flow is equal to the minimum capacity of an  $s$ - $t$  cut.

## NETWORK FLOWS

**Claim.** If all edge capacities are (non-negative) integers, then the Ford Fulkerson algorithm yields an optimal solution and runs in  $O((m + n) \cdot f)$  time, where  $f$  is the value of the flow returned.

*Proof.* In every iteration, the algorithm sends at least one additional unit of flow.

The algorithm therefore has  $\leq f$  iterations.

In any iteration, we can find an augmenting path in  $O(m + n)$  time using BFS/DFS on the residual graph.

This implies that the running time is  $O((m + n) \cdot f)$ .

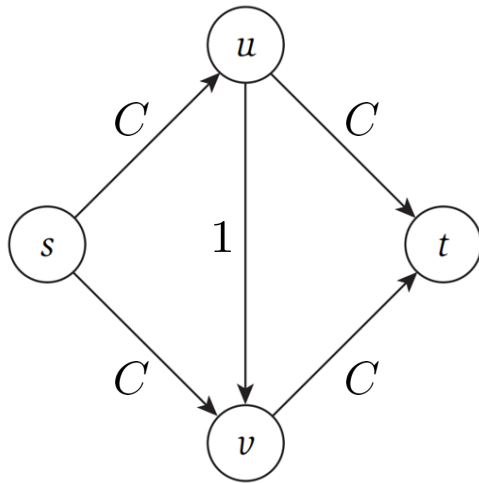
### WHY OPTIMAL?

The algorithm stops only when there is no augmenting path, implying that there is an  $s$ - $t$  cut whose capacity is equal to the value of the flow returned by the algorithm.

Thus, there cannot be a flow with larger value. ■

## NETWORK FLOWS

$O((m + n) \cdot f)$  is not a polynomial running time since  $f$  can be large.

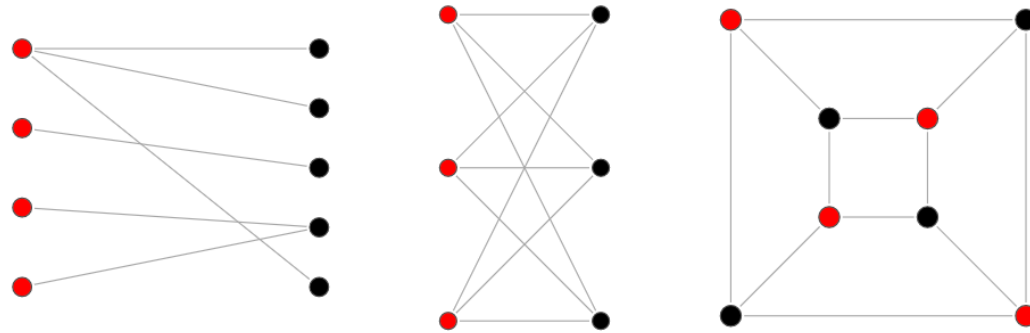


EXERCISE. THE ALGORITHM MAY TAKE  $2C$  STEPS (VALUE OF THE MAXIMUM FLOW)  
DESCRIBE HOW

# BIPARTITE GRAPHS

A bipartite graph, also called a bigraph, is a

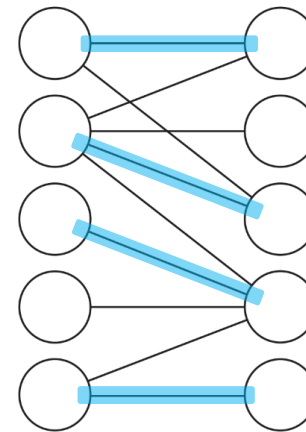
- set of graph vertices decomposed into two disjoint sets
- no two graph vertices within the same set are adjacent.



# MATCHINGS IN BIPARTITE GRAPHS

A matching in a graph is a subset of edges s.t. each vertex appears in at most one edge.

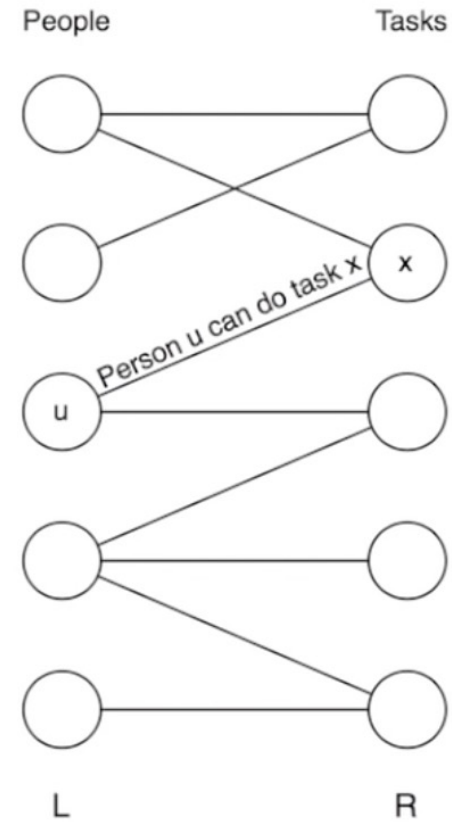
FIND A MATCHING





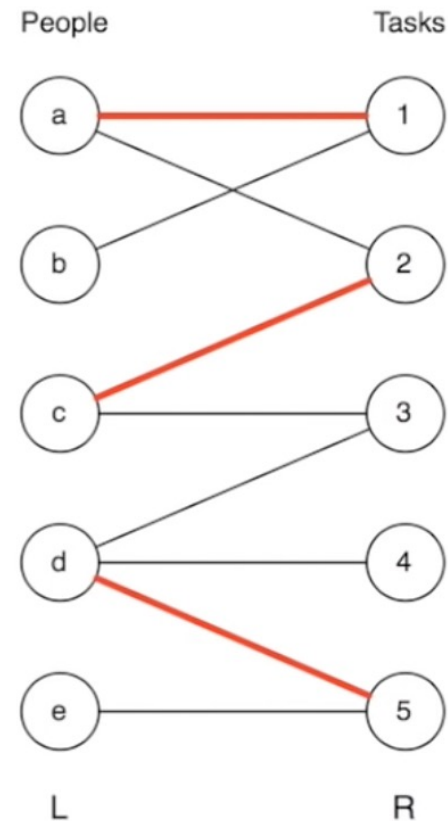
# MATCHINGS IN BIPARTITE GRAPHS

- Suppose we have a set of people  $L$  and set of jobs  $R$ .
- Each person can do only some of the jobs.
- Can model this as a bipartite graph  $\rightarrow$



# MATCHINGS IN BIPARTITE GRAPHS

- A **matching** gives an assignment of people to tasks.
- Want to get as many tasks done as possible.
- So, want a **maximum matching**: one that contains as many edges as possible.
- (This one is not maximum.)



## MATCHINGS IN BIPARTITE GRAPHS

A matching in a graph is a subset of edges s.t. each vertex appears in at most one edge.

**Task.** Find a matching with the maximum number of edges in a bipartite graph.

