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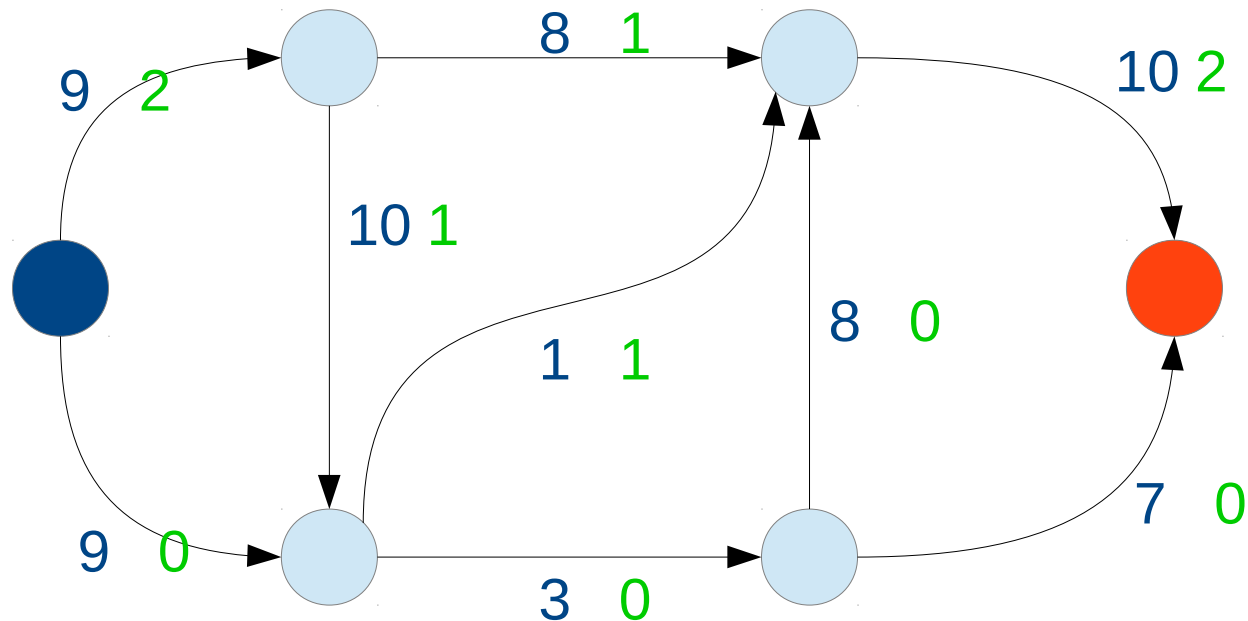
A.A. 2015/2016

# Summary of Lecture 1

- Networks are pervasive
- Networks are too complex in features and size to be managed without advanced tools
- A suitable formal framework is known to model network flows that
  - is « compact »
  - is flexible
  - is tractable

# Summary of Lecture 1

- Given :
  - a graph  $G(V,A)$ , with two special nodes  $s$  and  $t$
  - capacity on each arc
- Find: an optimal flow from  $s$  to  $t$

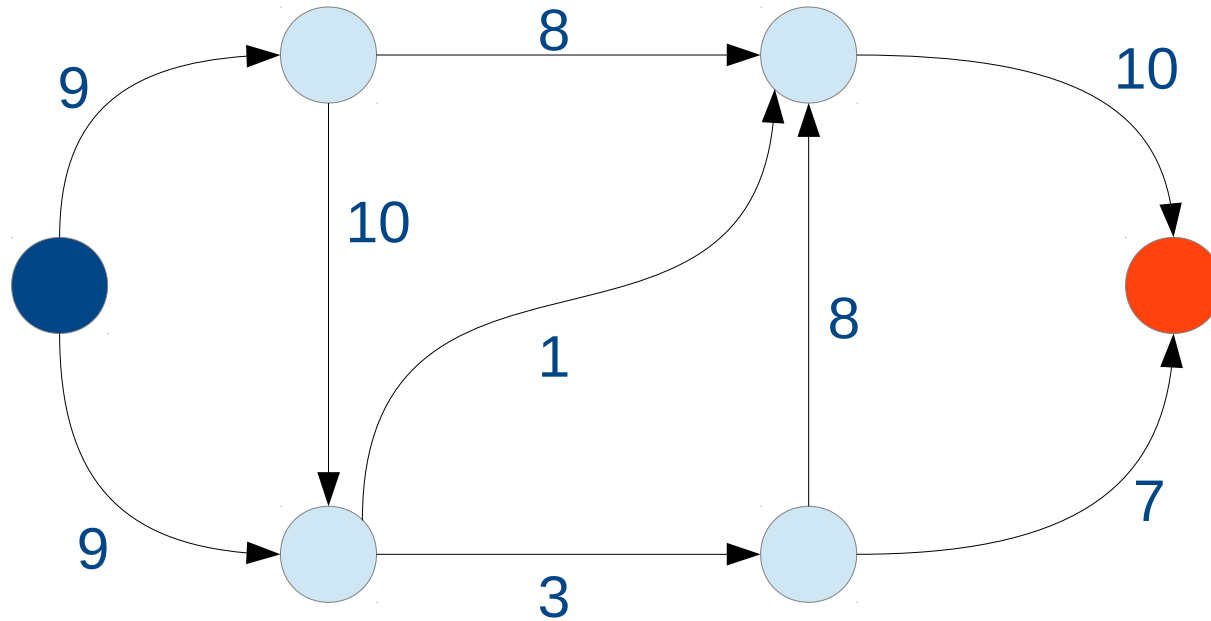


# Summary of Lecture 1

- The framework allows also to
  - Take into account circulations instead of s-t flows
  - Have additional capacities on nodes
  - Have flow lower bounds (on both arcs and nodes)
  - Have multiple sources and multiple sinks
  - Have source and demand points
- by simple modeling gadgets ...
- ... and therefore without any increase in the complexity of the resolution algorithms.

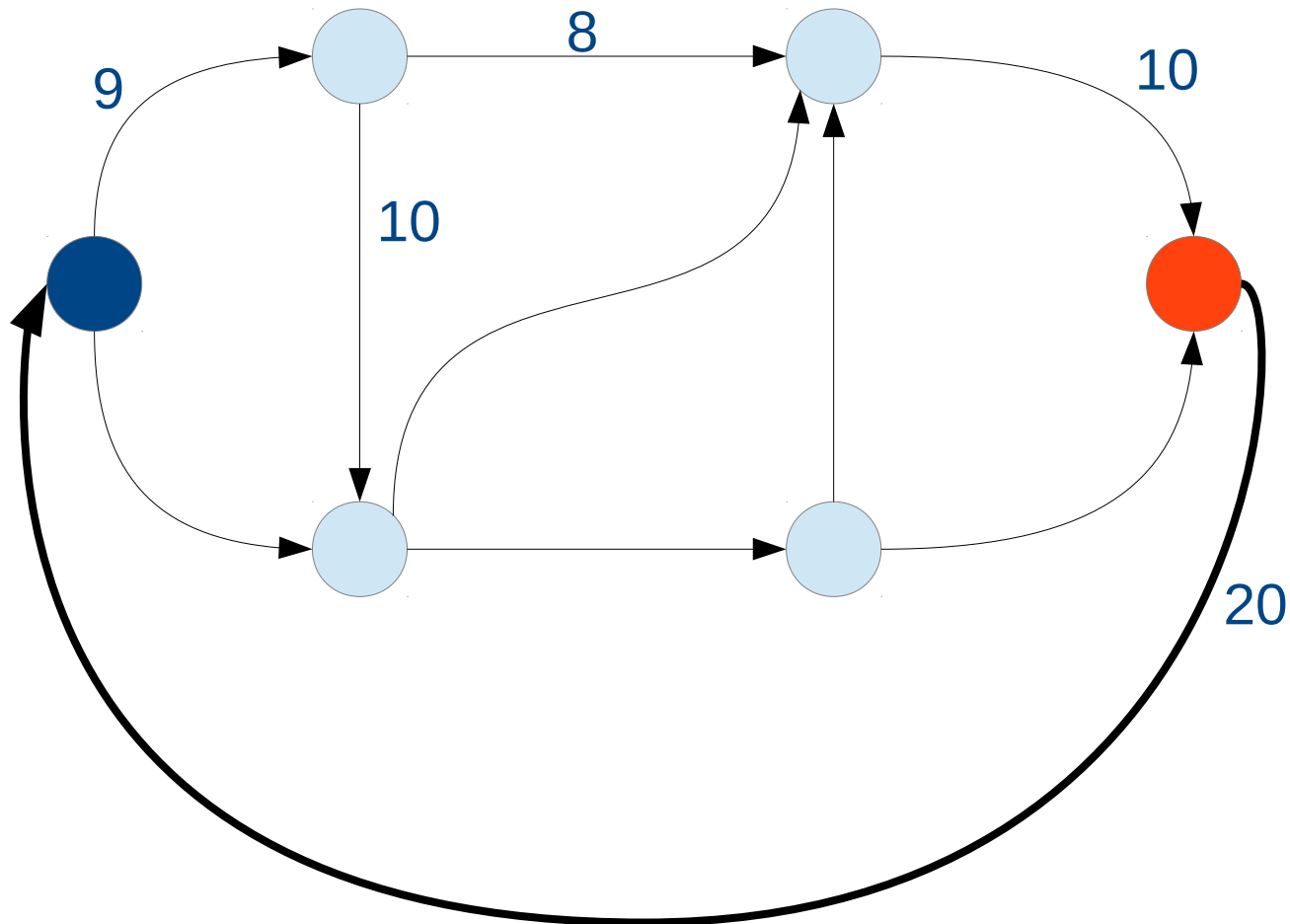
# Summary

- Basic model



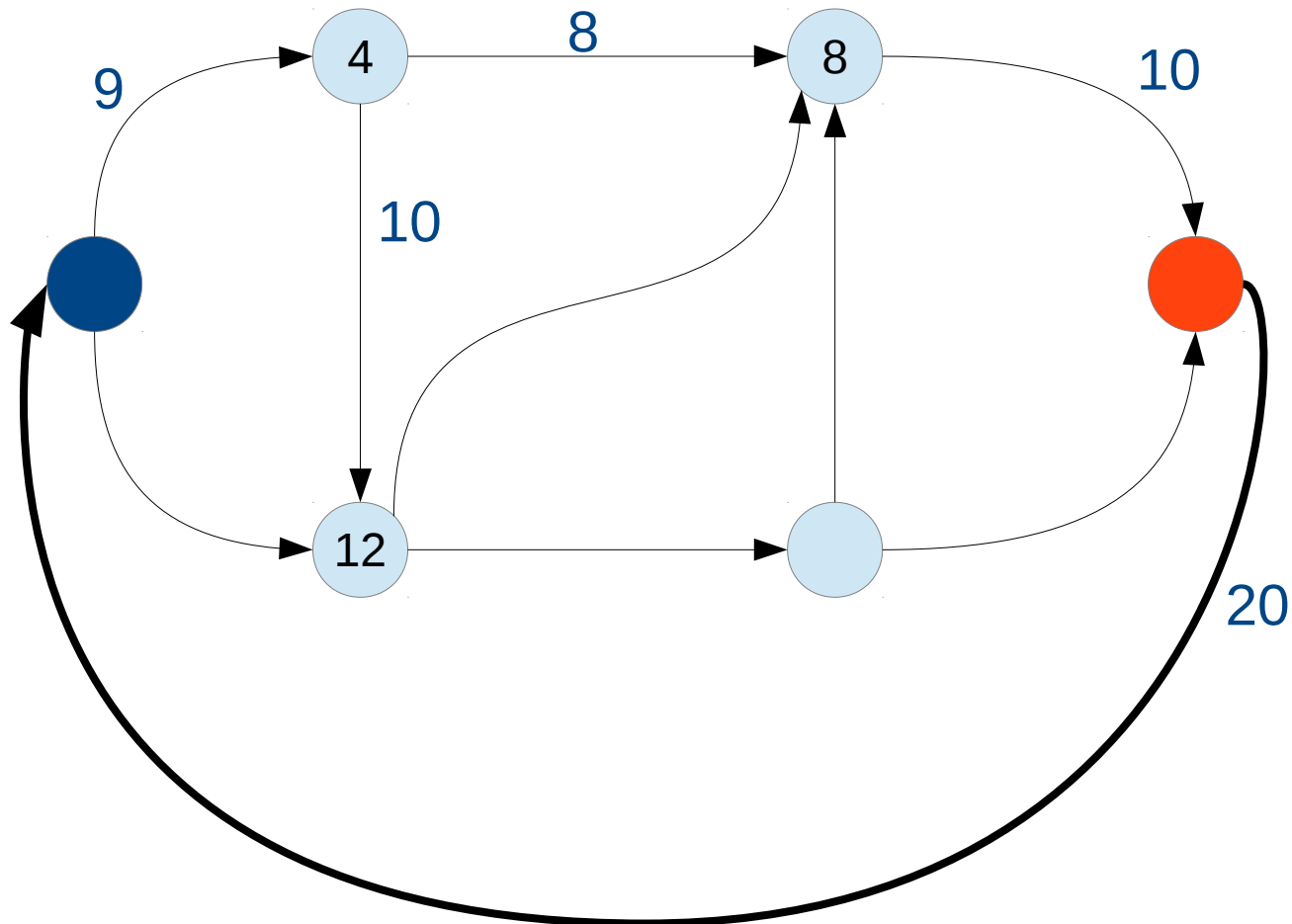
# Summary

- Flow circulation



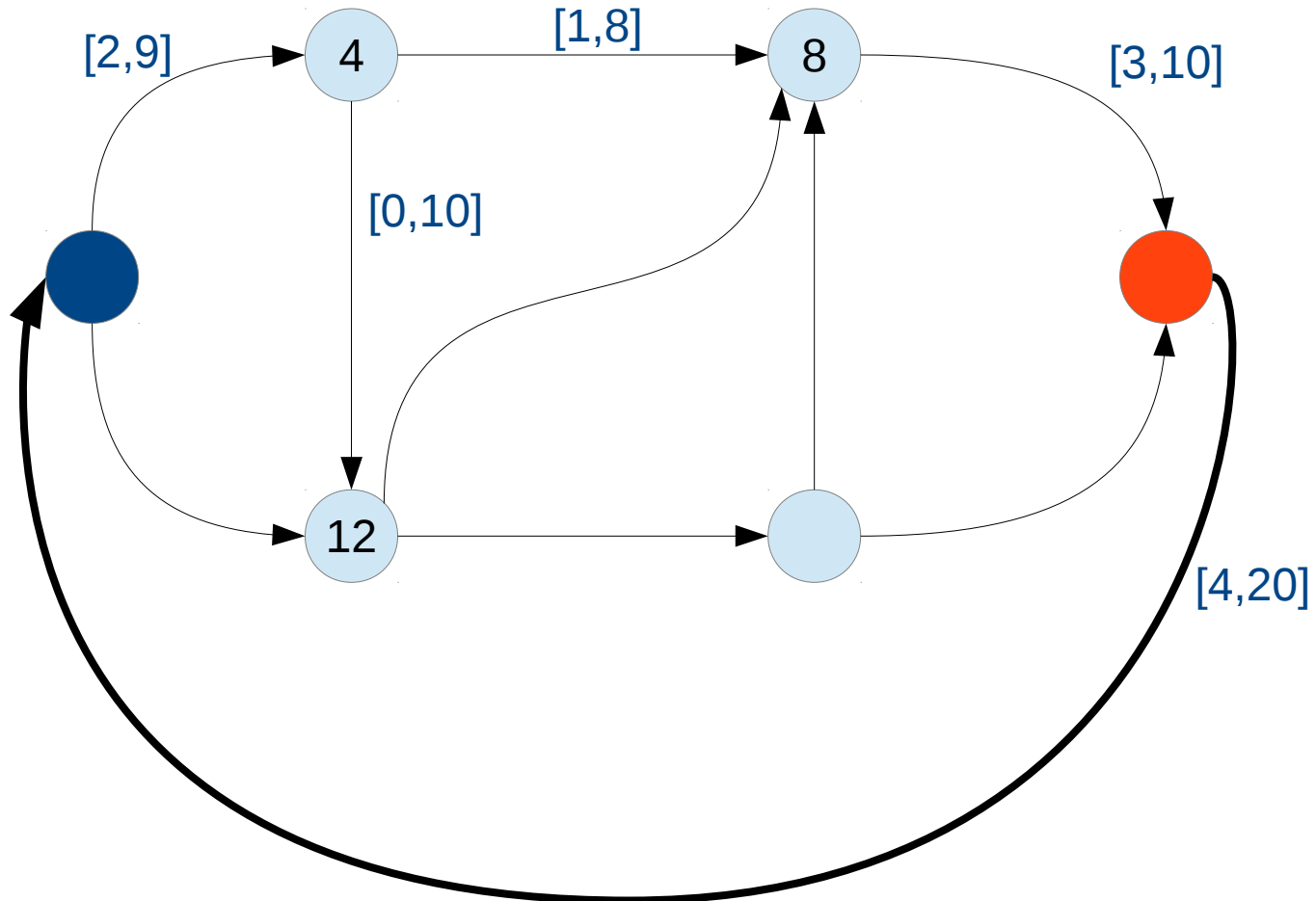
# Summary

- Capacities on nodes



# Summary

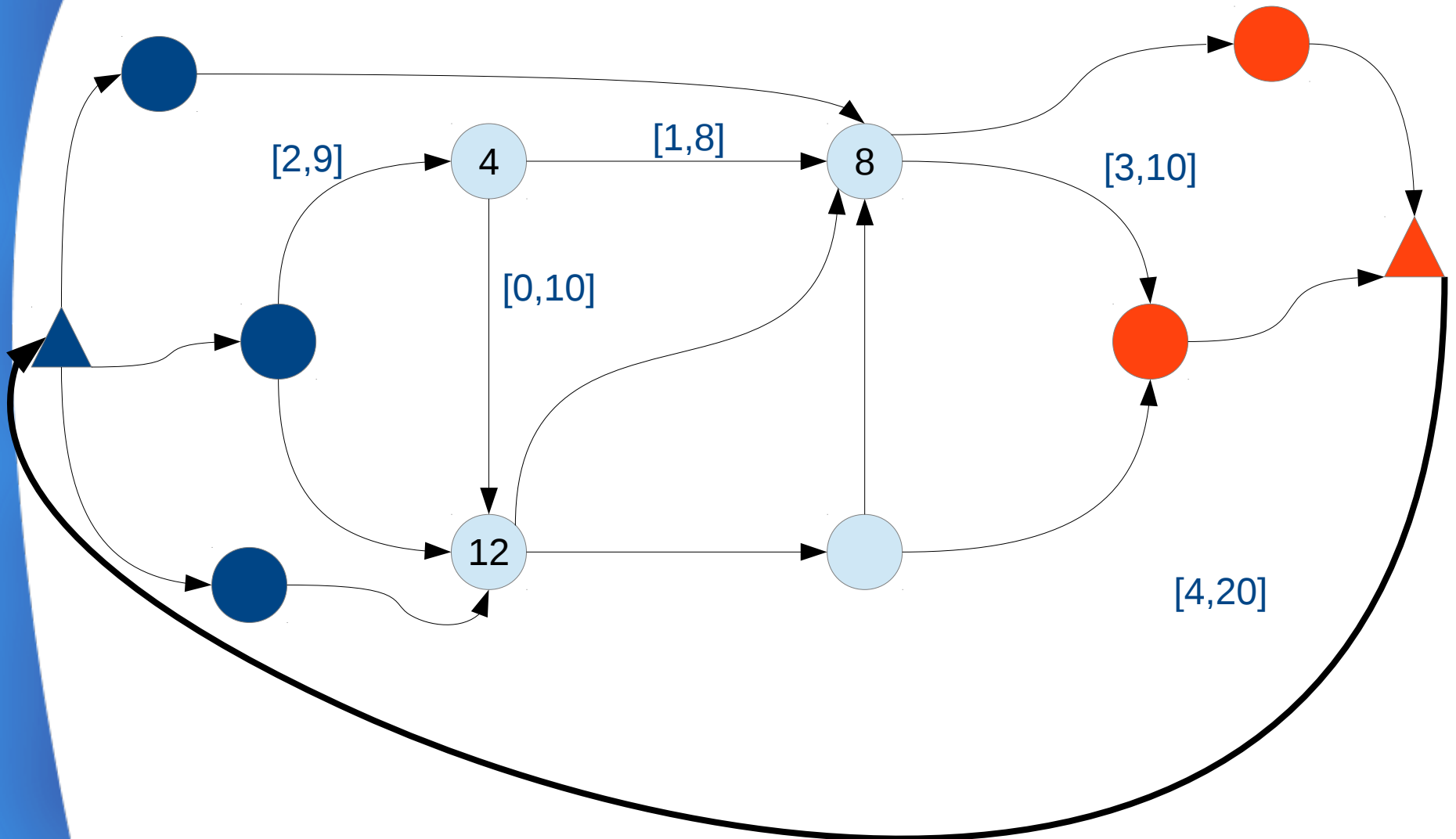
- Flow lower bounds





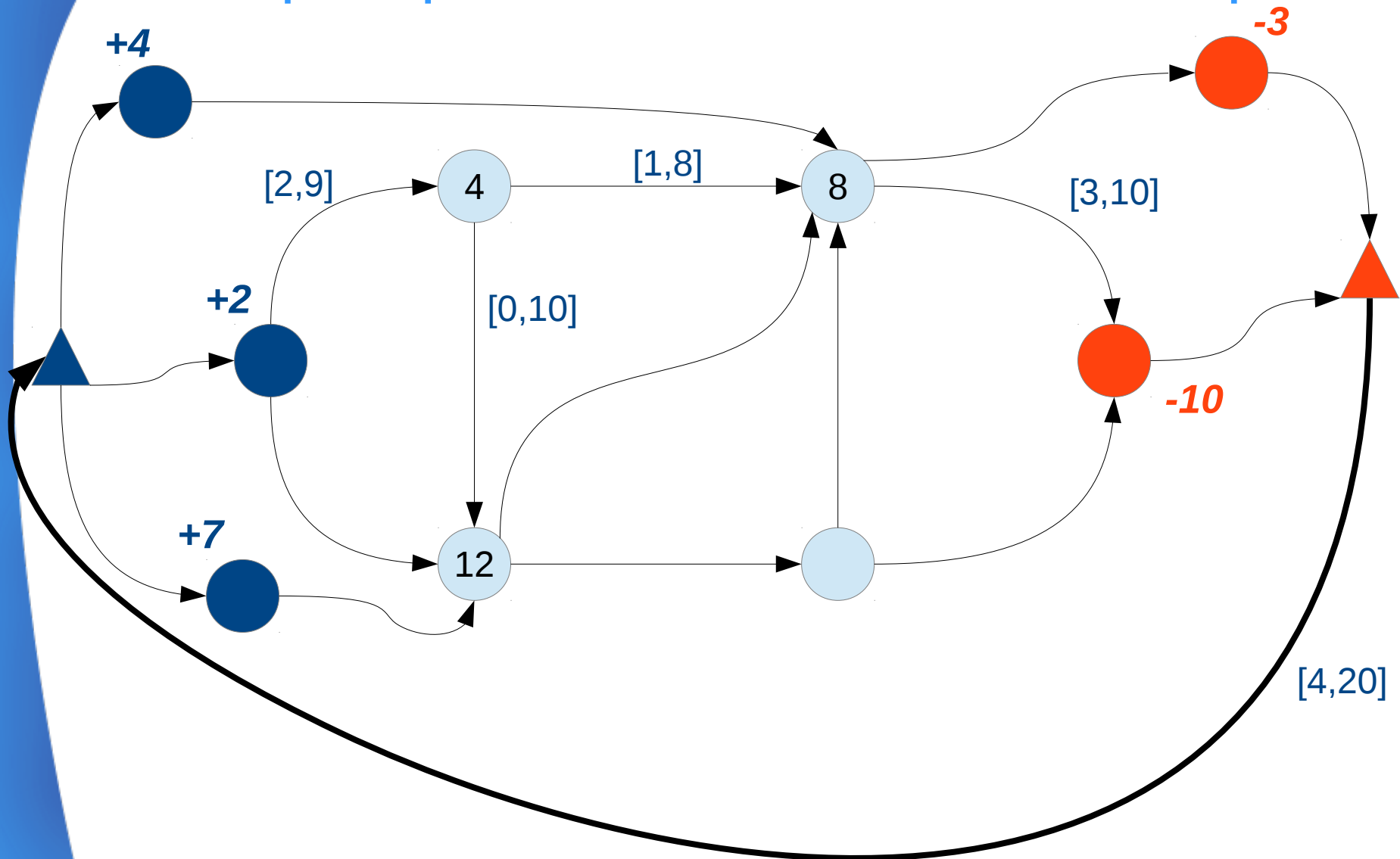
# Summary

- Multiple sources and multiple sinks



# Summary

Multiple «production» and «demand» points



# Examples

- Analysis of (telecom) network connectivity
  - Single path
  - Link disjoint paths
  - «robustness» against link failures
- Tango dancing problem (max matching)
- Matrix rounding problem

# Example 4: survey design

Taken from J. Kleinberg, E. Tardos, « Algorithm Design »

- We would like to design a survey of products used by consumers (i.e., « Consumer  $i$ : what did you think of product  $j$ ? »).
- The  $i$ th consumer agreed in advance to answer a certain number of questions in the range  $[c_i, d_i]$
- Similarly, for each product  $j$  we would like to have at least  $p_j$  opinions about it, but not more than  $q_j$
- Each consumer can be asked about a subset of the products which they consumed.
- The question is how to assign questions to consumers, so that we get all the informations we need, and each consumer is being asked a valid number of questions.

# Part II – Properties of Flows

- Another good feature of Network Flows :
  - structural properties can be proved by the design of efficient algorithms!
- ... i.e. theory and computation fit nicely!

# Lecture Plan

- Shortest path problems and algorithms
- Flows and Cuts
- Max flow algorithms
- Combinatorial properties of max flows
- Application examples

# Shortest Path Problems

- Given a graph  $G(V,A)$
- Given a source node  $s$  and a destination node  $t$
- Given a length function  $l: A \rightarrow \mathbb{R}_+$
- Find a path from  $s$  to  $t$  of minimum overall length
- Path: sequence of adjacent nodes
- Length of a path: sum of lengths of arcs connecting pairs of subsequent nodes

# Shortest Path Algorithms

- Most famous one: Dijkstra
  - Keep a set of « labels »  $d(i)$ , one for each  $i$  in  $V$
  - ... encoding « tentative » distances from  $s$  to  $i$
  - Iteratively pick one node, and try to improve the labels of its neighbors
  - Until no more improvements are possible

```
Procedure update(i) {  
  foreach j in  $\partial^+(i)$  do  
    if  $d(j) > d(i) + l(i,j)$  then  
       $d(j) = d(i) + l(i,j)$   
}
```

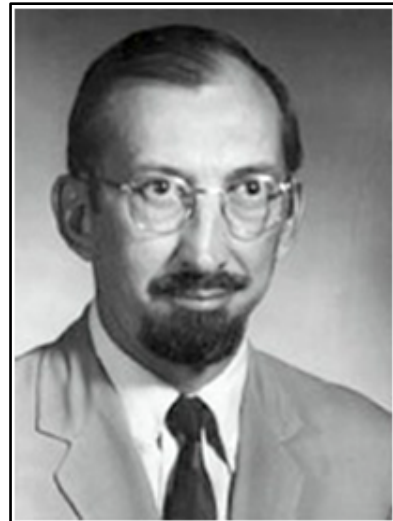




- Slides 8 – 12 from  
    « Network Flows », by J. Orlin
- Dijkstra Algorithm example  
    by J. Orlin
- Proof of correctness (blackboard)
- Summary: we have an efficient way for finding (special) paths between nodes of networks

# Max Flow algorithms

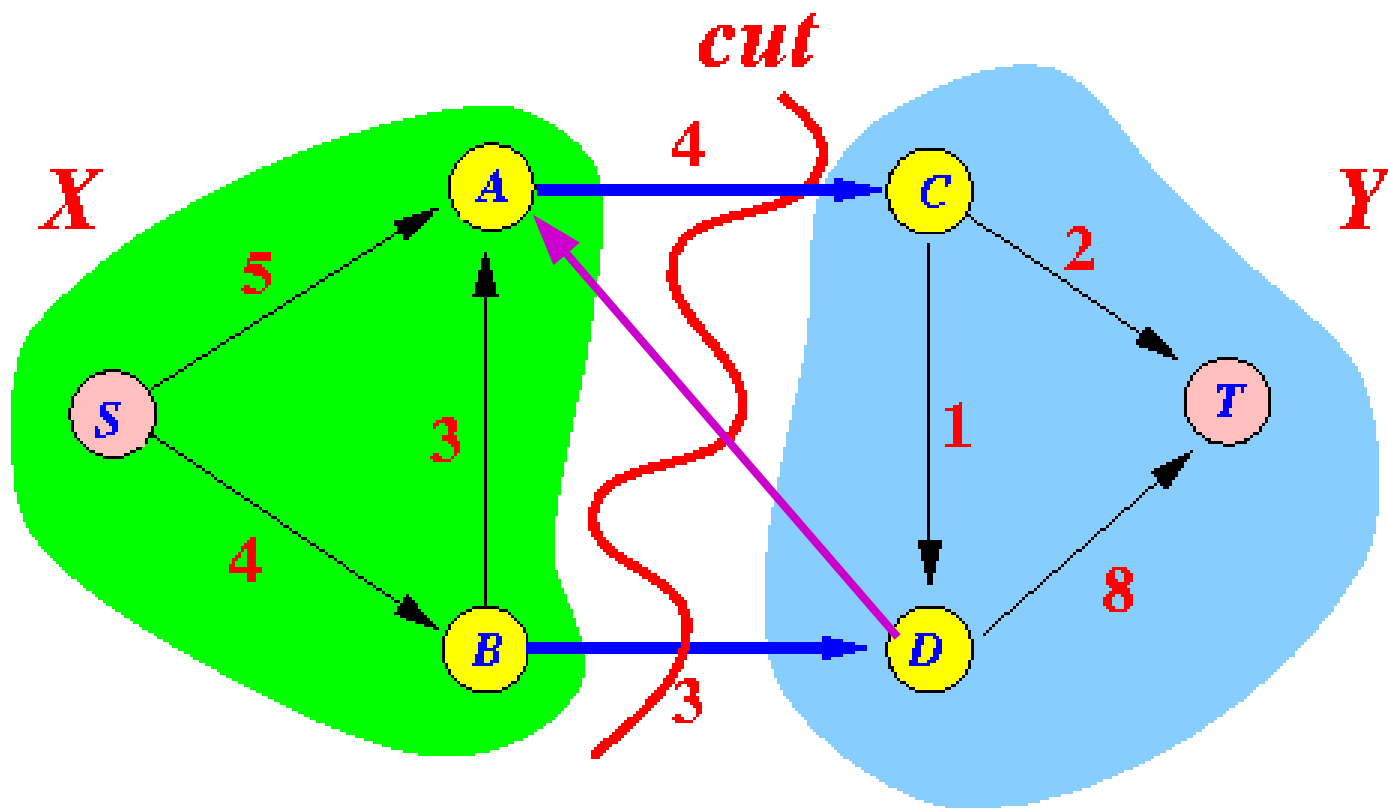
- Working with residual networks, slides  
« Network Flows », by J. Orlin
- Ford Fulkerson Algorithm example  
by J. Orlin



# Flows and Cuts

- Intuitively (s-t) **Cut**: a set of arcs whose removal makes sink unreachable from source (no more path exists going from s to t)
- Formally, a cut is a **partition**  $[S, V \setminus S]$  of the nodes of the graph (s-t cut if s is in S and t in  $V \setminus S$ )
- Arcs of the cut are those having one endpoint in S and the other in  $V \setminus S$ ;
  - forward arc:  $(i,j)$  with i in S and j in  $V \setminus S$
  - backward arc:  $(i,j)$  with i in  $V \setminus S$  and j in S
- The **capacity** of a cut is the sum of capacities of its **forward arcs**
- A cut is **minimum** if its capacity is minimum

# Flows and cuts



$$\text{Cut} = \{ (A, C), (B, D) \}$$

# Properties of max flows

- Obs. 1: if caps are **integral**, I always increment flows by **integers**, and therefore obtain only **integral flows**!
- Obs. 2: Intuitively, how can I detect if a flow is maximum? No more augm. path!
- Formally, I need to compare flows and cuts !
  - Weak duality theorem for flows and cuts
  - Optimality conditions for max flows

# Weak duality for max flow

- Def. **Flow across a cut**  $[S,T]$  is the sum of flow on forward arcs minus the sum of flow on backward arcs
- Claim 1:
  - If  $f()$  is a feasible flow and  $[S,T]$  is an  $(s-t)$  cut
  - The the flow across the cut = flow from  $s$  = flow into  $t$
  - Proof: blackboard
- Claim 2:
  - If  $f()$  is a feasible flow and  $[S,T]$  is an  $(s-t)$  cut
  - Then the **flow** across the cut is at most equal to the **capacity** of the cut  $[S,T]$
  - Proof: blackboard
- → If  $f()$  is a feasible flow and  $[S,T]$  is an  $(s-t)$  cut
  - Then the **flow** from  $s$  to  $t$  is **at most** equal to the **capacity** of the cut  $[S,T]$

# Optimality conditions for max flows

- The following are equivalent :
  - 1 - A flow  $f()$  is maximum
  - 2 - There is no augmenting path in the residual network
  - 3 - There is an (s-t) cut whose capacity equals the s-t flow of  $f()$
- Proof: blackboard

# Cuts and Flows

- Recall: find **how many** paths are there from source to destination, having **no common arc**?
- **Menger's Theorem (1927)**. The *max* number of *edge*-disjoint s-t paths is equal to the *min* number of *edges* whose removal disconnects destination from source.

K. Menger "Zur allgemeinen Kurventheorie". Fund. Math. 10: 96–115 (1927)





# Cuts and Flows

Let  $F$  be a subset of  $A$

• Proof  $\leftarrow$

(suppose the removal of  $F$  disconnects  $t$  from  $s$ , and  $|F| = k$ )

- All  $s$ - $t$  paths use at least one edge of  $F$ .
- Hence, the number of edge-disjoint paths is **at most  $k$** .

• Proof  $\rightarrow$

(suppose max number of edge-disjoint paths is  $k$ )

- Then max flow value is  $k$ .
- A cut  $(S, T)$  of capacity  $k$  exists (max flow min cut duality)
- Let  $F$  be set of edges going from  $S$  to  $T$ .
- $|F| = k$ , and  $F$  disconnects  $t$  from  $s$ .

# Example 6: image segmentation

Taken from J. Kleinberg, E. Tardos, « Algorithm Design»

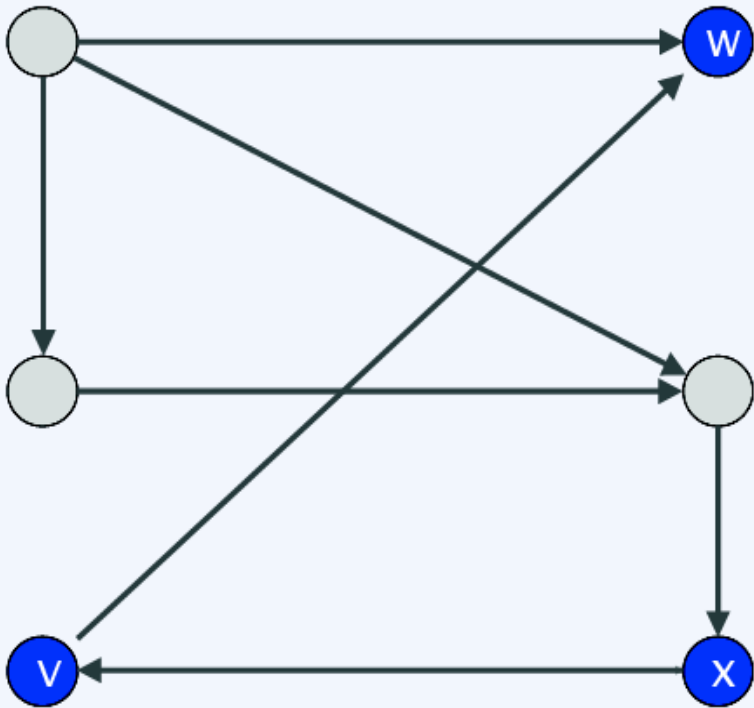
- You are given an image, say a set  $V$  of pixels
- For each pixel  $i$  in  $V$  :
  - $a_i$  is the likelihood that  $i$  is foreground
  - $b_i$  is the likelihood that  $i$  is background
  - $p_{ij}$  is the similarity between  $i$  and  $j$
- Goal: find a partition of pixels in foreground and background, optimizing accuracy and smoothness



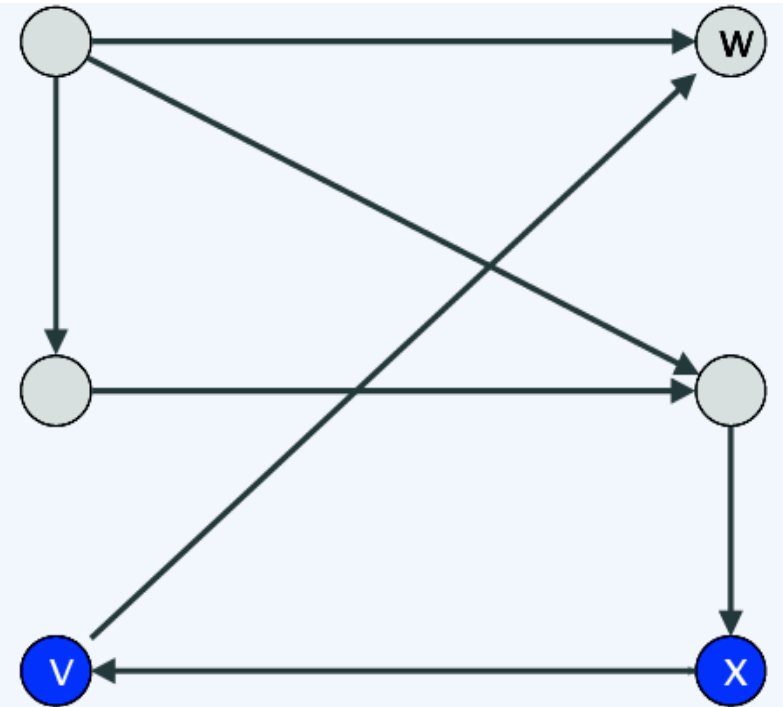
# Example 7: project management

Taken from J. Kleinberg, E. Tardos, « Algorithm Design»

- Set of possible projects  $P$
- Each project  $i$  has an associated coefficient  $p_i$ 
  - Profit if  $p_i > 0$
  - Cost if  $p_i < 0$
- Set of prerequisites  $E$ : if  $(i,j)$  in  $E$ ,  $i$  cannot be done unless also  $j$  is done
- A subset  $A$  of projects is feasible if each prerequisite of a project in  $A$  is also in  $A$
- Goal: find a feasible subset of projects of maximum revenue (maximum weight closure problem)



**{ v, w, x } is feasible**



**{ v, x } is infeasible**