

Faculty of Information Technology, Monash University

COMMONWEALTH OF AUSTRALIA

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FIT2004: Algorithms and Data Structures

Week 11: Network Flow

These slides are prepared by [M. A. Cheema](#) and are based on the material developed by [Arun Konagurthu](#) and [Lloyd Allison](#).

Announcements

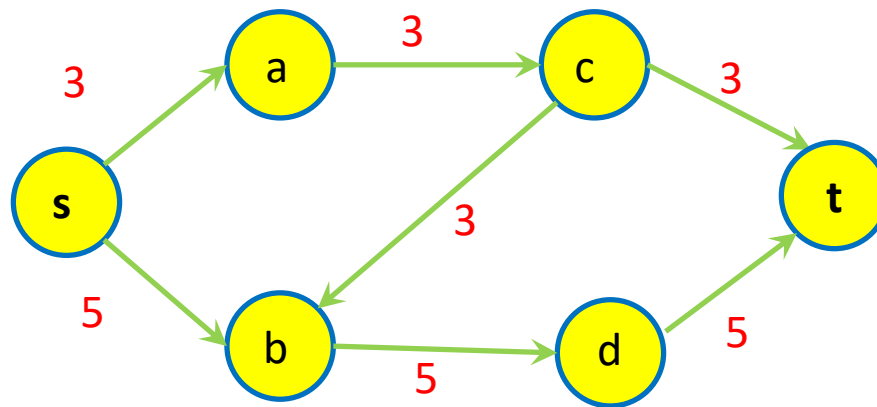
- SETU Feedback and nominations for Teaching Excellence Awards
 - Links on Moodle (on the right)
 - Closes 21st June 2020
- Real-time anonymous feedback:
<https://docs.google.com/forms/d/e/1FAIpQLSdAlUwUNPJtccGebDF-fg41oUCWrt-of5A2mg5I7VtEK6vN3A/viewform>

Outline

1. Maximum Flow Problem
2. Ford-Fulkerson Algorithm
3. Min-cut Max-flow Theorem

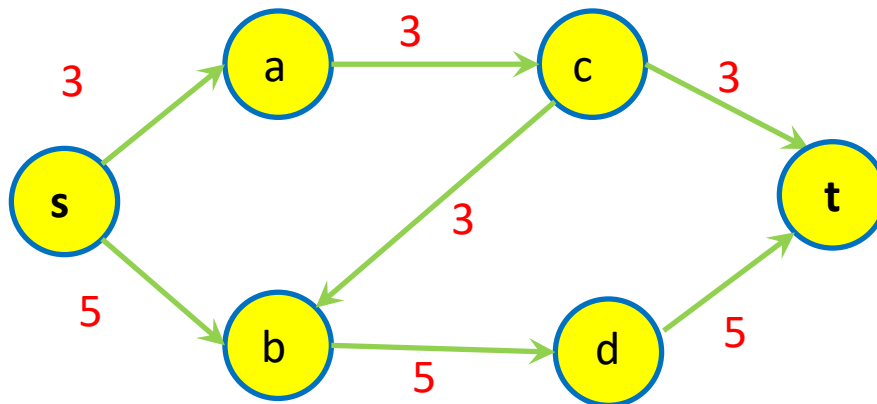
Flow Networks

- A **flow network** is a **connected directed** graph where
 - there is a single source vertex and a single sink/destination vertex;
 - each edge has a given (non-negative) capacity (usually integers)
 - ✦ giving the maximum amount/rate of flow that the edge can carry;
- **Flow networks** model many real-world problems
 - Water flowing through an assembly of pipes.
 - Electric current flowing through electrical circuits.
 - Information flowing through communication networks
 - Can be applied to many scenarios (unrelated to physical flows).



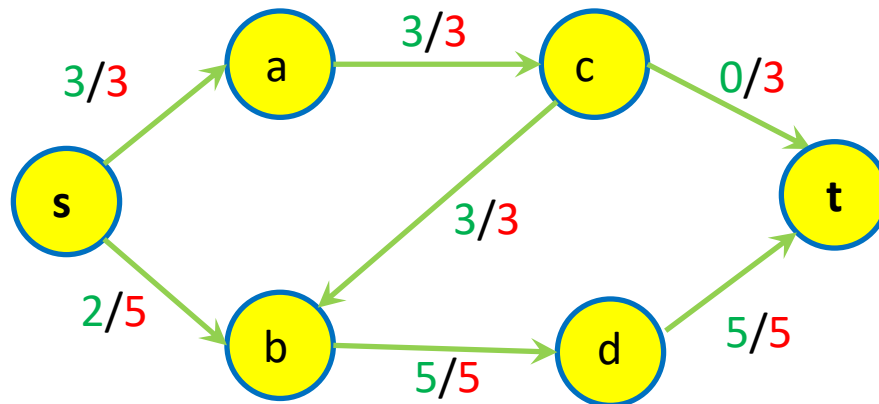
Some basic notations

- Set of all incoming edges to a vertex v : denoted as $E_{in}(v)$
 - $E_{in}(b) = s \rightarrow b, c \rightarrow b$
 - $E_{in}(a) = ?$
- Set of all outgoing edges from a vertex v : denoted as $E_{out}(v)$
 - $E_{out}(b) = b \rightarrow d$
 - $E_{out}(a) = ?$
- Source Vertex: denoted as s (has **no** incoming edges)
- Sink/target vertex: denoted as t (has **no** outgoing edges)



Flow

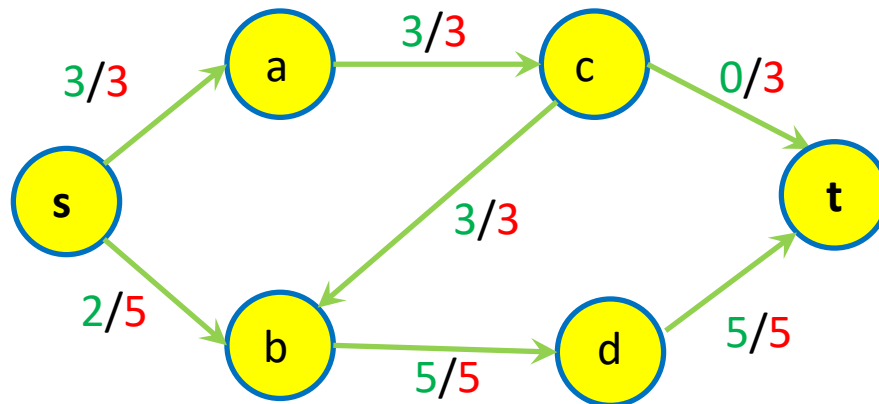
- Flow is an **assignment** of how much material is flowing through each edge in the flow network given its stated edge capacity.



Green numbers indicate flow and red indicate capacity. Flow is not shown if 0

Flow

- All vertices (except source and sink) **conserve** their flow. That is
 - The total amount flowing **into** any vertex (through incoming edges) IS EQUAL TO the total amount flowing **out** of that vertex (through outgoing edges).
 - This key property is called **flow conservation**.



Properties of a Flow Network

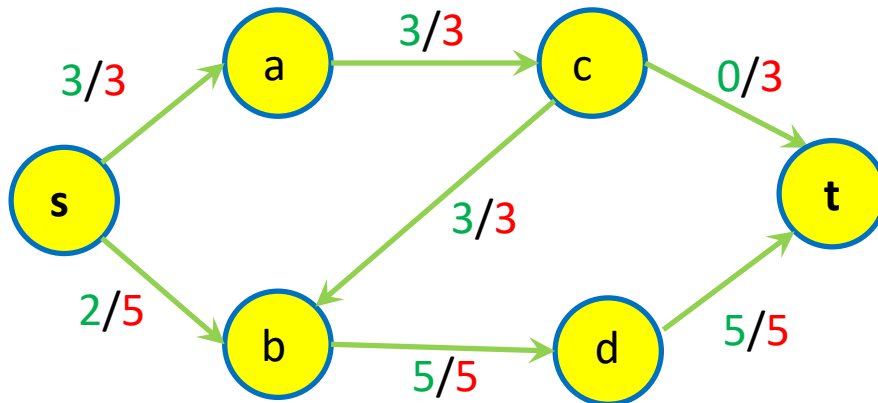
A flow network must satisfy the following two properties.

Property 1: Capacity Constraint

- For each edge e , its flow, denoted as $f(e)$, is bounded by the capacity of its edge, i.e., $0 \leq f(e) \leq c(e)$ where $c(e)$ is the capacity of the edge

Property 2: Flow Conservation

- For any vertex v (except source and sink), the total flow coming into the vertex must be equal to the total flow going out from this vertex – formally

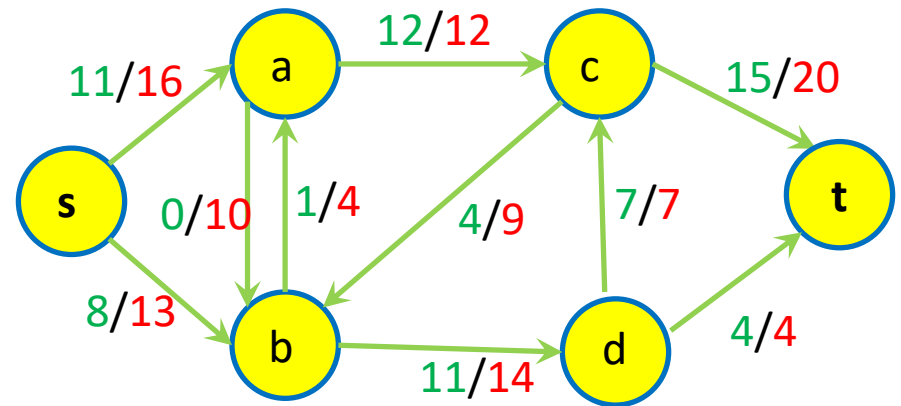
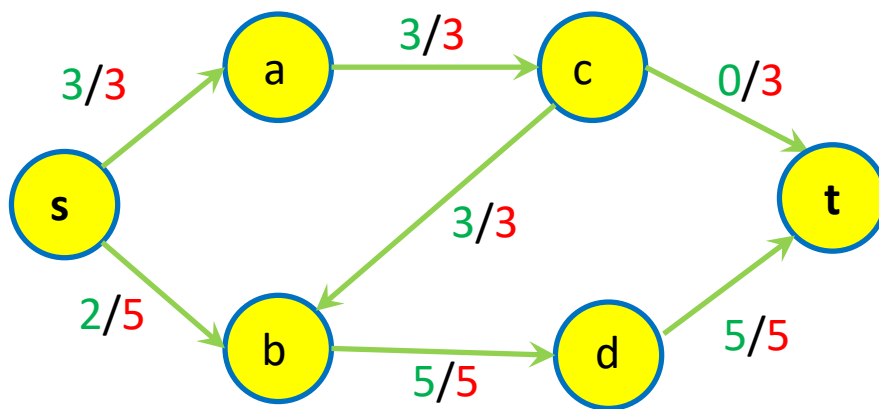


$$\sum_{\forall e_{in} \in E_{in}(v)} f(e_{in}) = \sum_{\forall e_{out} \in E_{out}(v)} f(e_{out}).$$

Maximum-flow Problem

Value of a flow in a network:

- Given that flow network satisfies the capacity constraint and flow conservation properties, flow of a network is the **total flow out of the source vertex**. Equivalently, this is the same as the **total flow into sink vertex**.
- What is the flow value in the flow network at bottom left?
- What is the flow value in the flow network at bottom right?



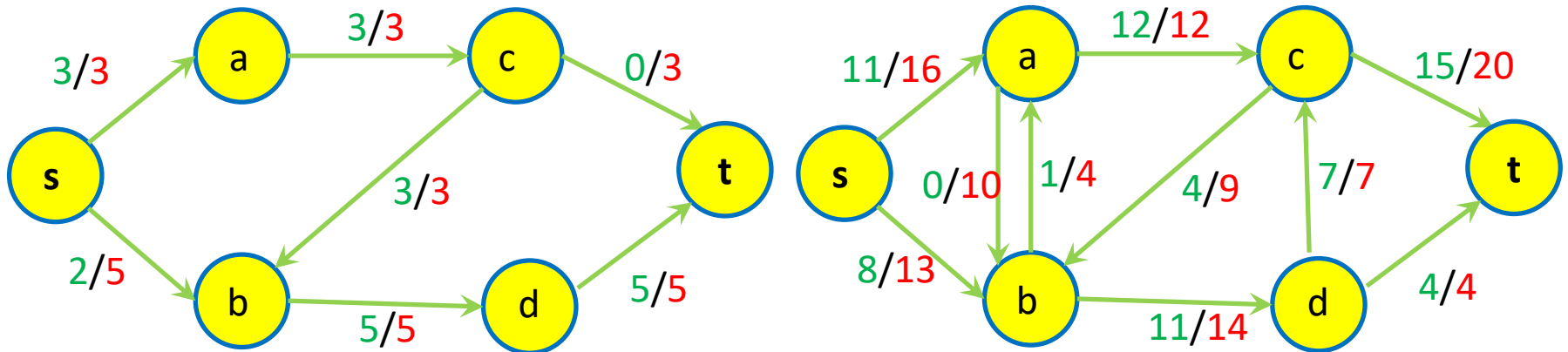
Maximum-flow Problem

Value of a flow in a network:

- Given that flow network satisfies the capacity constraint and flow conservation properties, flow of a network is the **total flow out of the source vertex**.

Maximum-flow problem

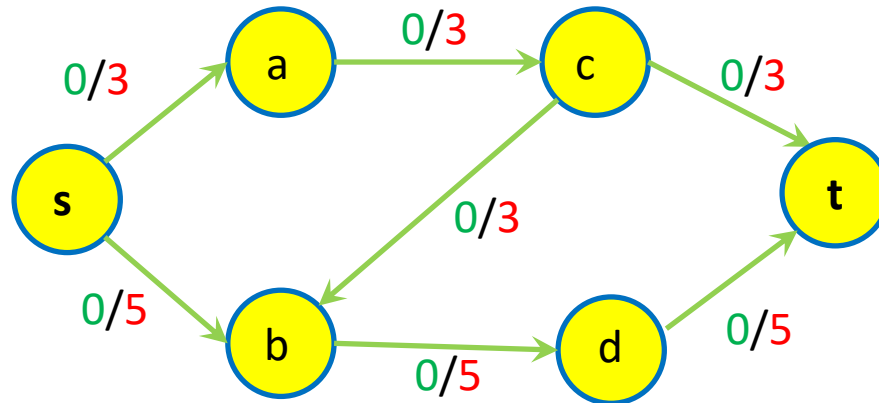
- Given a flow network, determine the maximum value of the flow that can be sent from source s to sink t without violating the flow network properties.



Outline

1. Maximum Flow Problem
2. Ford-Fulkerson Algorithm
3. Min-cut Max-flow Theorem

Ford Fulkerson Intuition

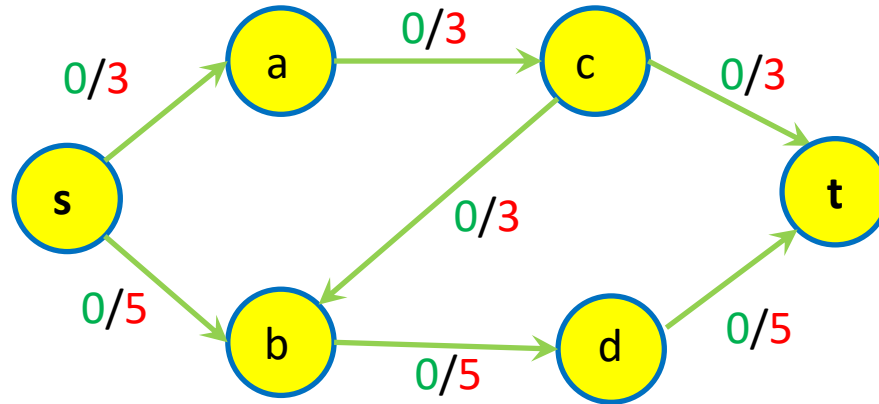


How can we increase the flow in the graph above?

Quiz time!

<https://flux.qa> - RFIBMB

Ford Fulkerson Intuition

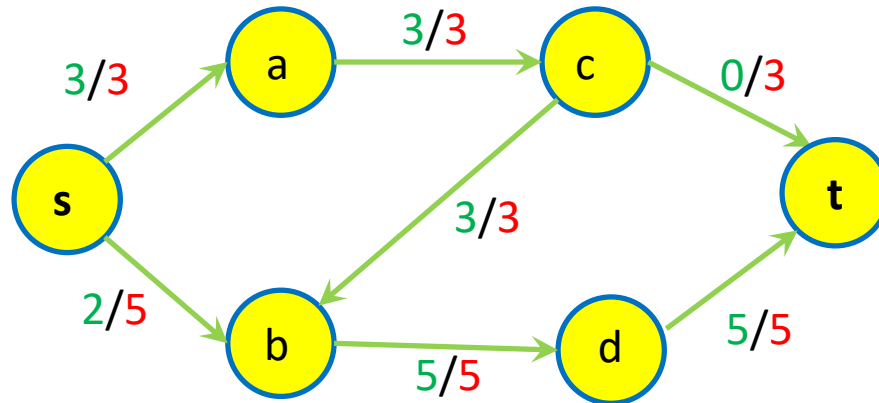


How can we increase the flow in the graph above?

1. Choose a path from source to sink
2. Increase flow along it

Seems easy enough!

Ford Fulkerson Intuition

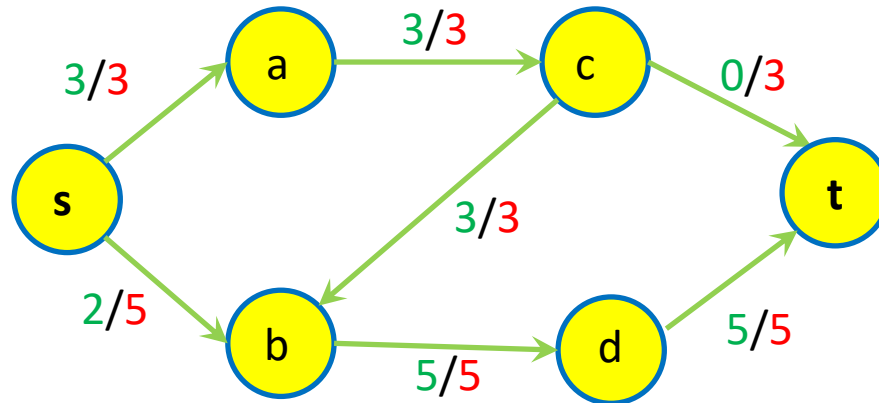


Can we increase the flow in the above network?

Quiz time!

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Ford Fulkerson Intuition

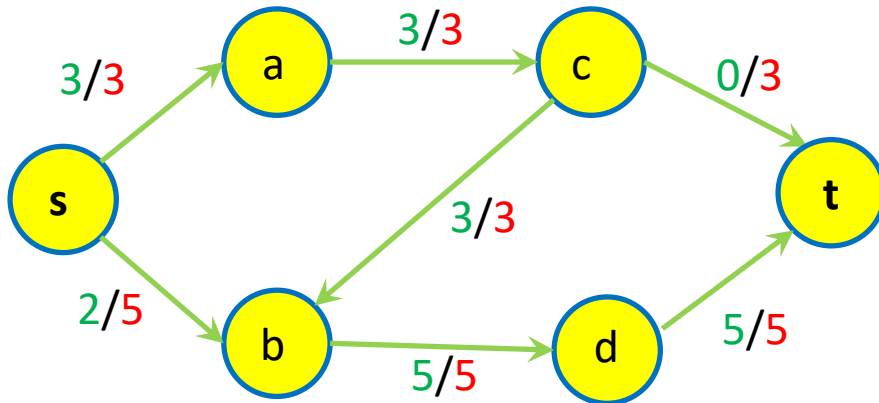


Can we increase the flow in the above network?

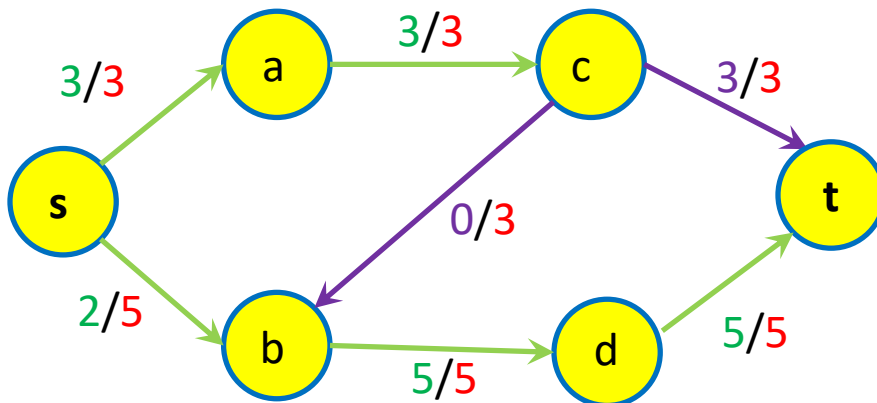
We can!

But there is no path from source to sink with spare capacity...

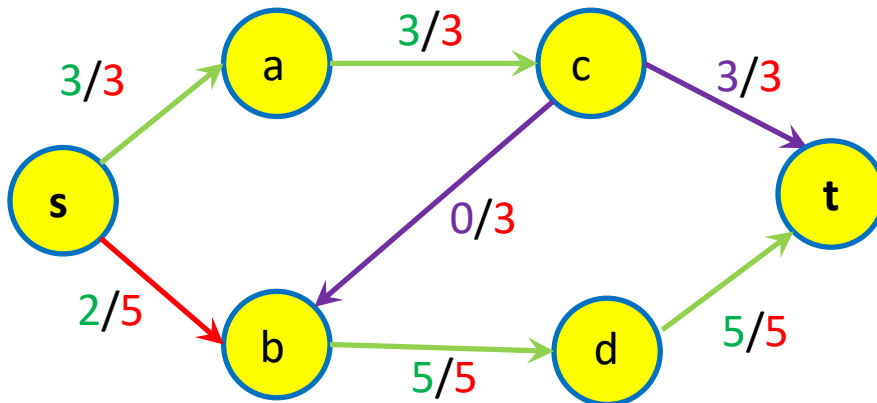
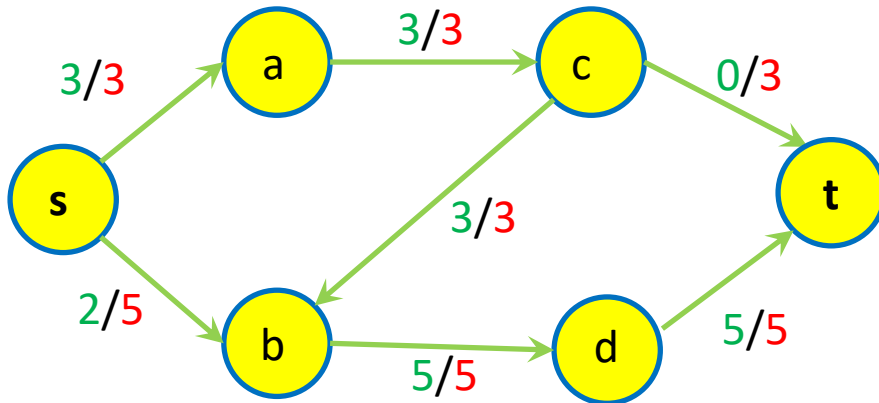
Ford Fulkerson Intuition



- There is not path from s to t with spare capacity
- Instead, we “redirect” the 3 units on the edge c->b
- They are now going to t

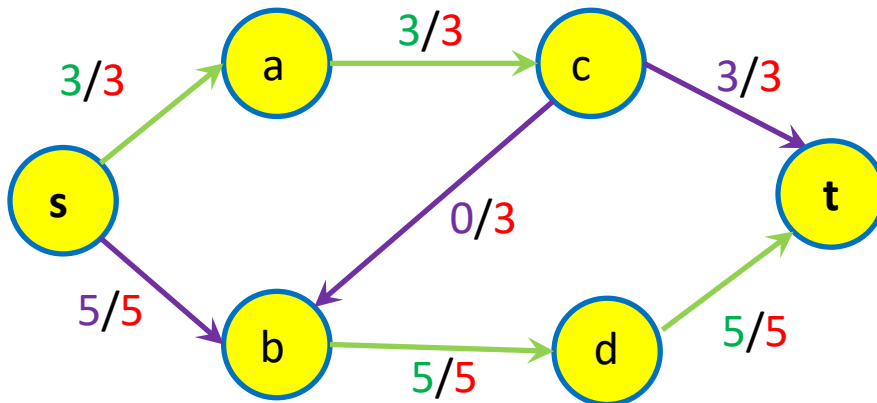
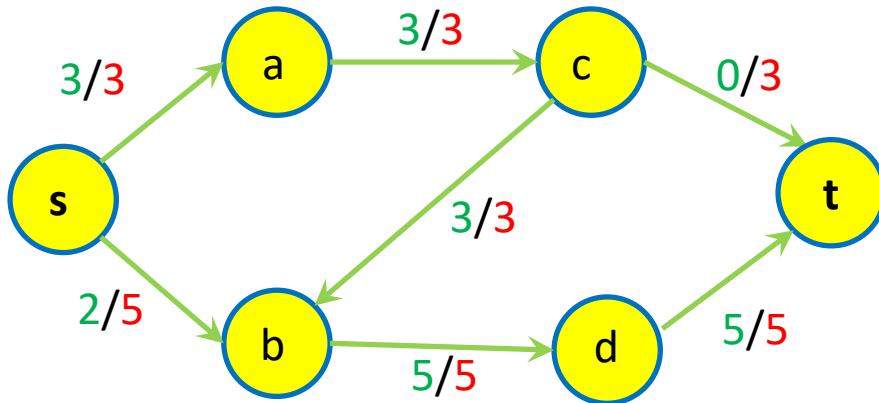


Ford Fulkerson Intuition



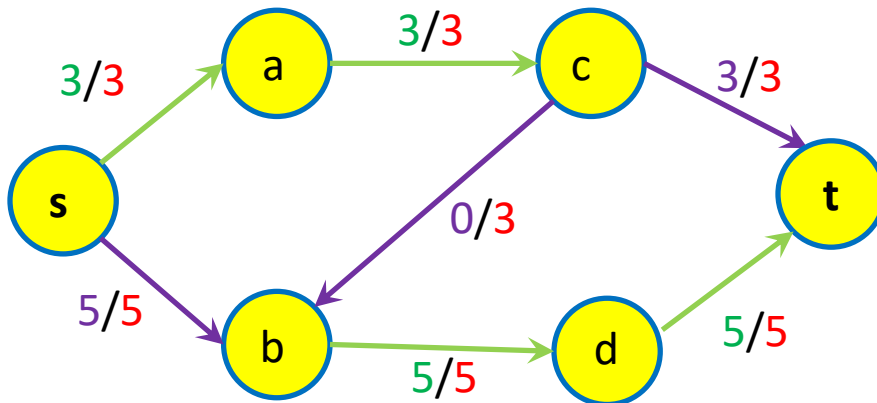
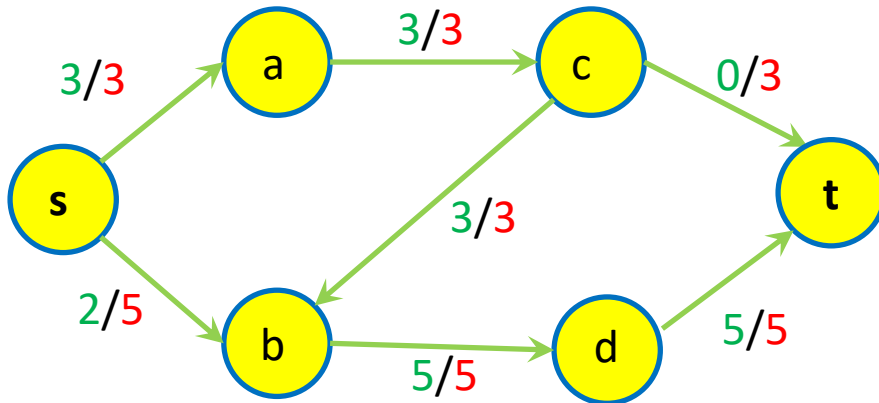
- There is not path from s to t with spare capacity
- Instead, we “redirect” the 3 units on the edge c->b
- They are now going to t
- In the second diagram, the flow through b is not conserved
- We can send 3 more units along s->b to both increase total flow and conserve flow through b

Ford Fulkerson Intuition



- There is not path from s to t with spare capacity
- Instead, we “redirect” the 3 units on the edge $c \rightarrow b$
- They are now going to t
- This means that we need to send 3 more units into b (because of flow conservation)
- In this case, the extra flow comes from s

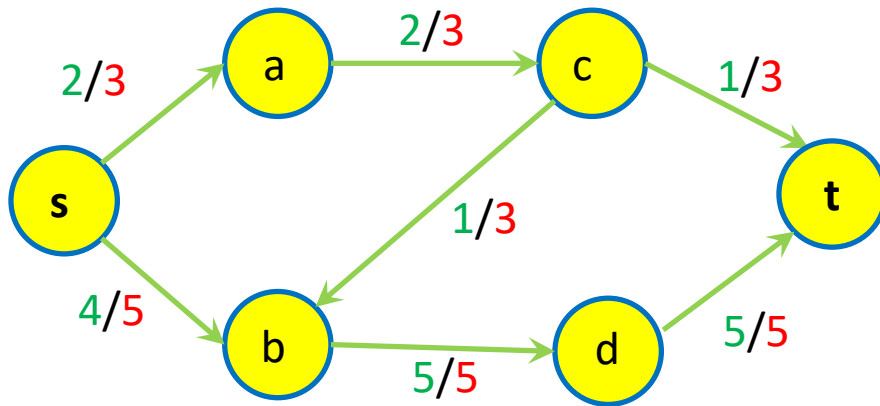
Ford Fulkerson Intuition



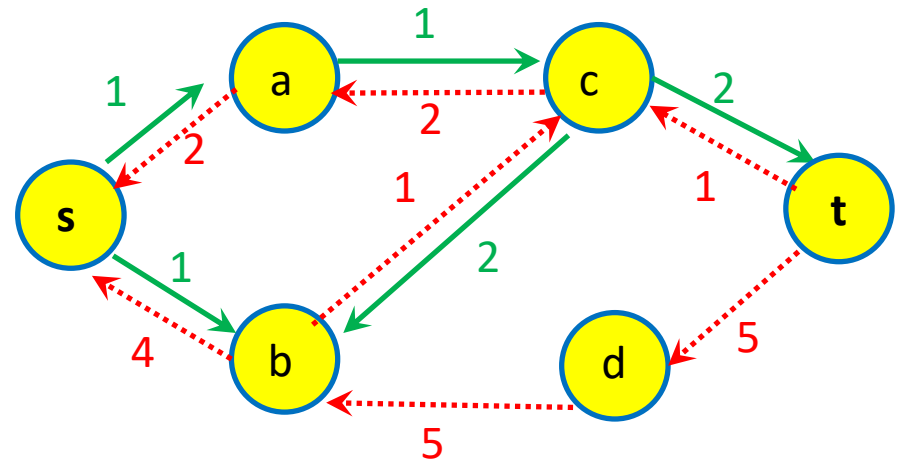
- What actually happened here?
- We increased the total flow by 3:
 - We sent 3 units along $s \rightarrow b, c \rightarrow t$
 - We “removed” 3 units of flow from $c \rightarrow b$
- Our path was $s \rightarrow b \rightarrow c \rightarrow t$, but we had a backwards edge...

Residual Network

- Residual network has the same vertices as the original network.
- For every directed edge $u \rightarrow v$ in flow network, we add two edges in the residual network:
 - **Forward edge/Residual edge:** An edge in the same direction as $u \rightarrow v$ with the residual/remaining capacity
 - **Backward edge/Reversible flow edge:** An edge in the direction opposite to $u \rightarrow v$ (i.e., $v \rightarrow u$) with weight equal to the current flow of $u \rightarrow v$ in the flow network



Flow Network: Green numbers indicate flow and red indicate capacity. Flow is not shown if 0

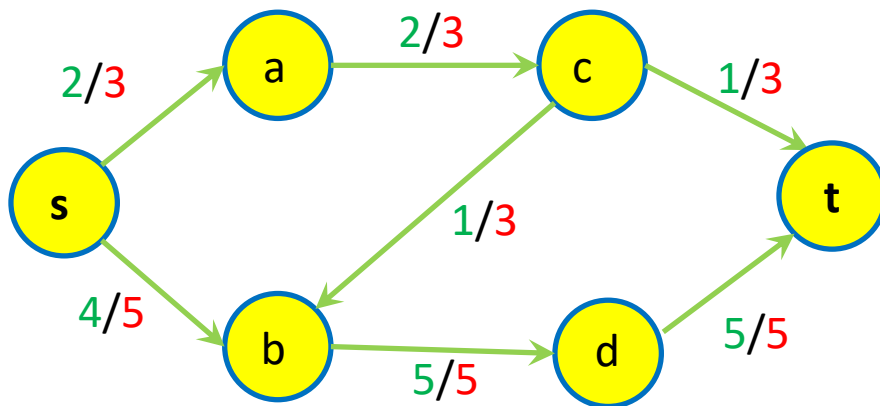


Residual Network: Where possible red edges indicate reversible flow and green indicate residual capacity

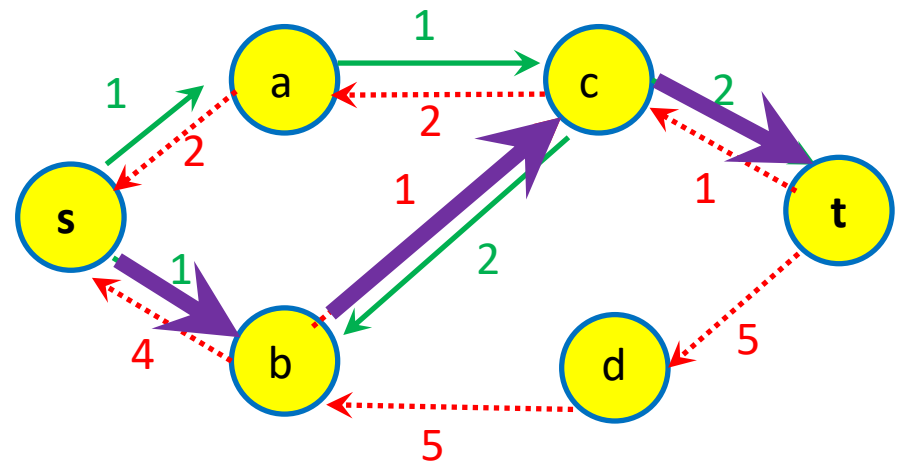
Augmenting Path in Residual Network

Augmenting path is any simple path (a path without repeating vertices) from source s to target t .

- E.g., $s \rightarrow b \rightarrow c \rightarrow t$ (shown in purple edges)
- **Residual capacity** of a path is the minimum edge weight on this path (e.g., 4 in the example)
- For each edge along this path, we can push additional flow equal to the “residual capacity of the path” in the flow network, e.g., 1 along each edge on $s \rightarrow b \rightarrow c \rightarrow t$




Flow Network: Green numbers indicate flow and red indicate capacity. Flow is not shown if 0



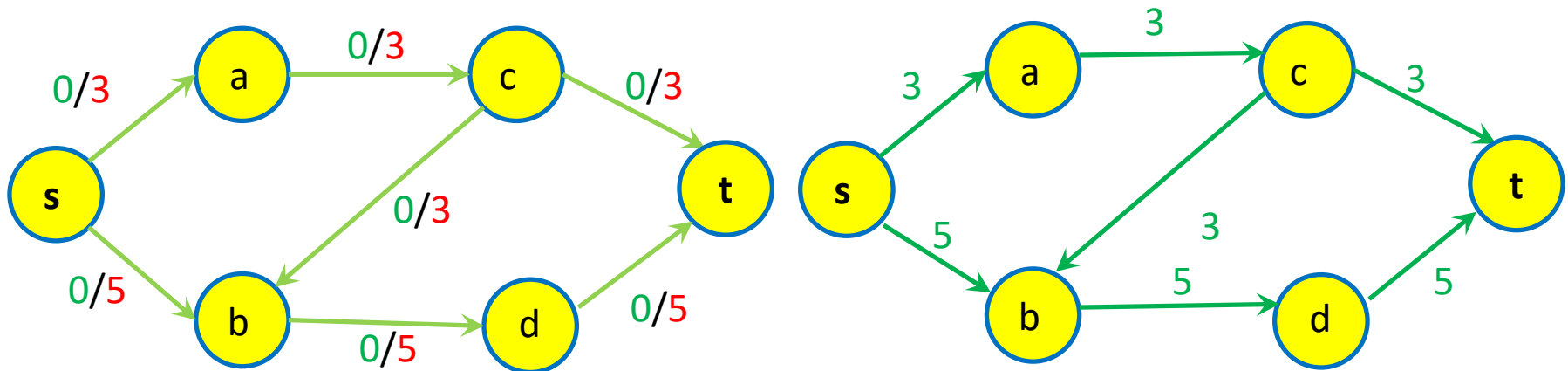
Residual Network: Where possible red edges indicate reversible flow and green indicate residual capacity

Ford-Fulkerson Method – Worked example (by hand in lecture)

Algorithm 75 The Ford-Fulkerson method

- 1: **function** MAX_FLOW($G = (V, E), s, t$)
 - 2:  Set initial flow f to 0 on all edges
 - 3: **while** there exists an augmenting path p in the residual network G_f **do**
 - 4: Augment the flow f along the augmenting path p
 - 5: **return** f
-

Total flow: 0

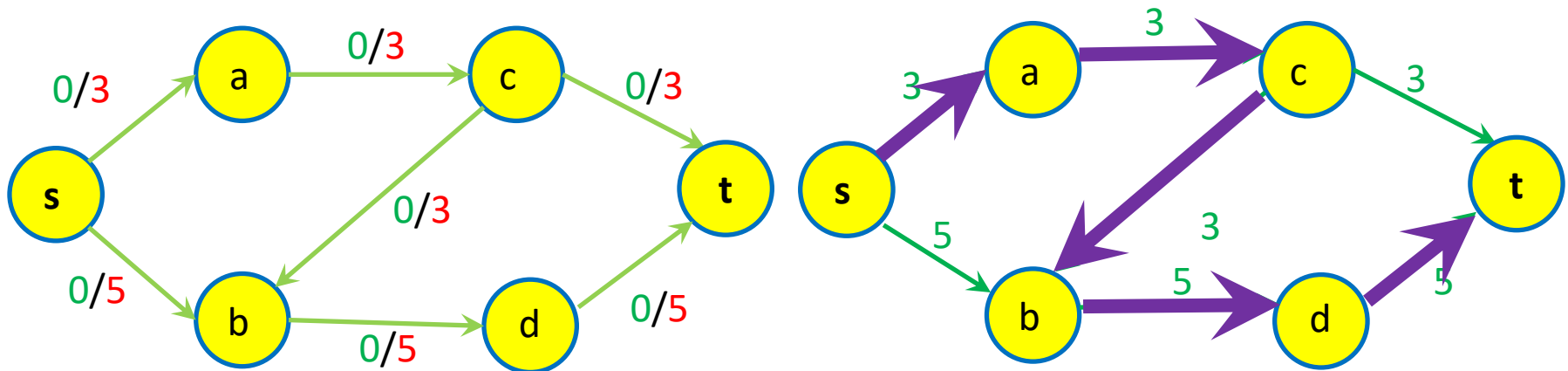


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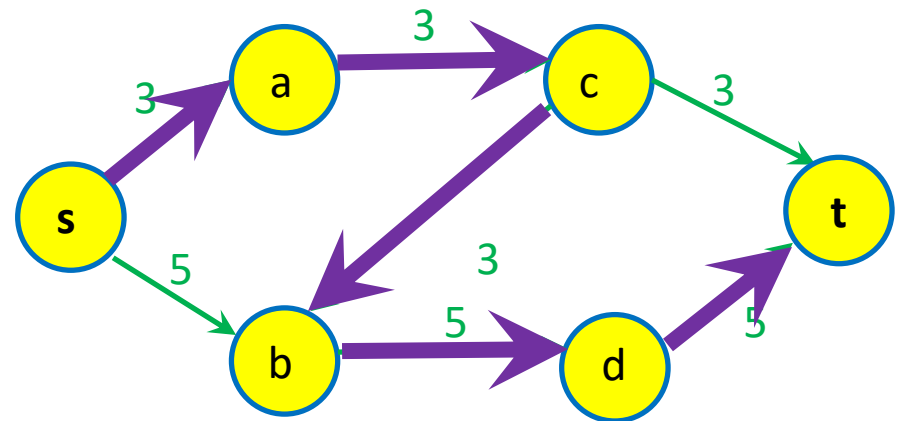
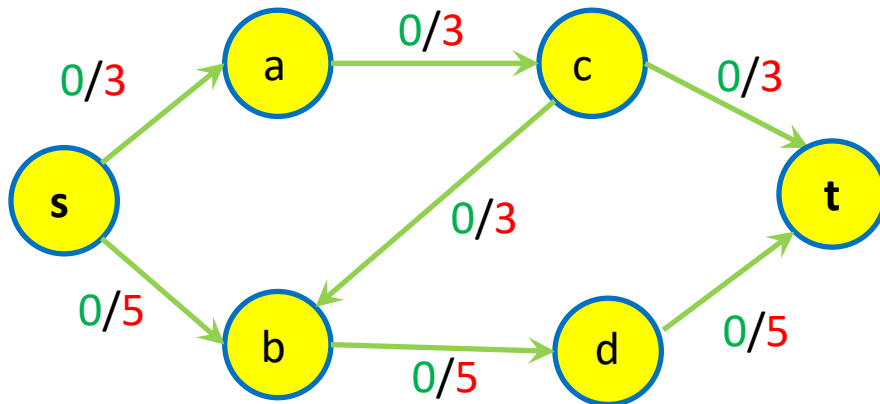


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Total flow: 0



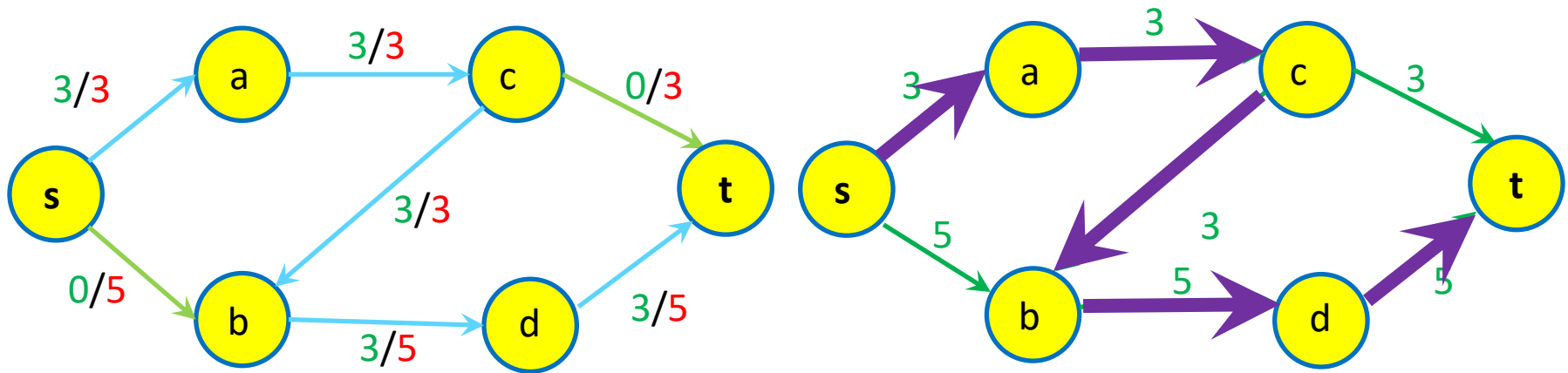
Path found: capacity = 3

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Total flow: 3



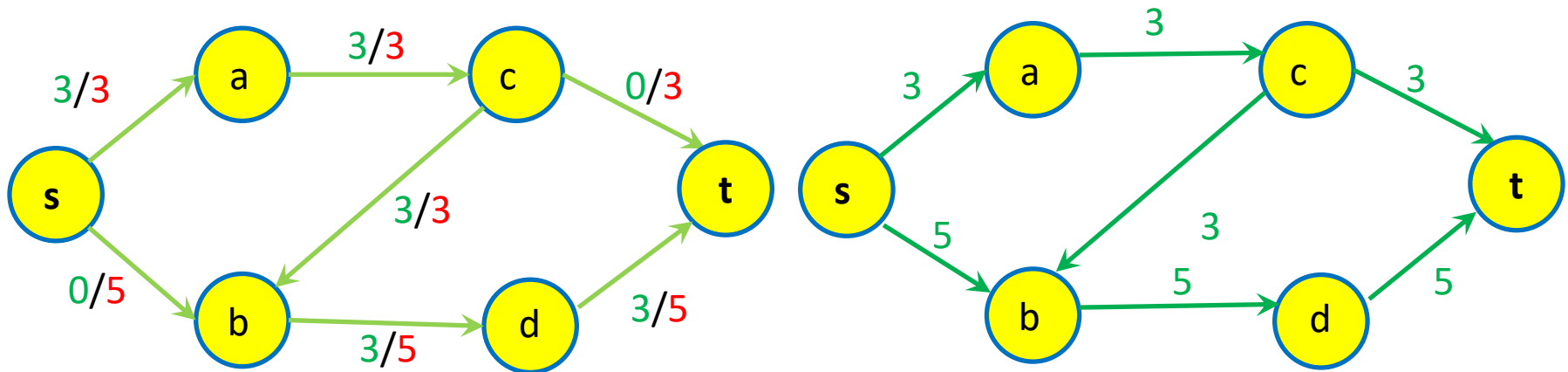
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Total flow: 3



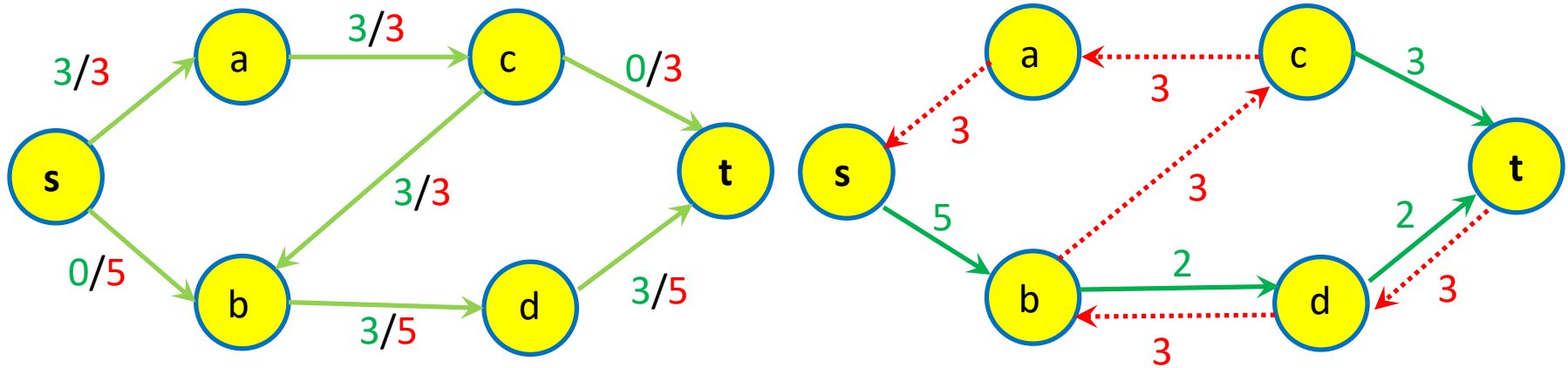
Update residual to match new flows

Ford-Fulkerson Method – Worked example (by hand in lecture)

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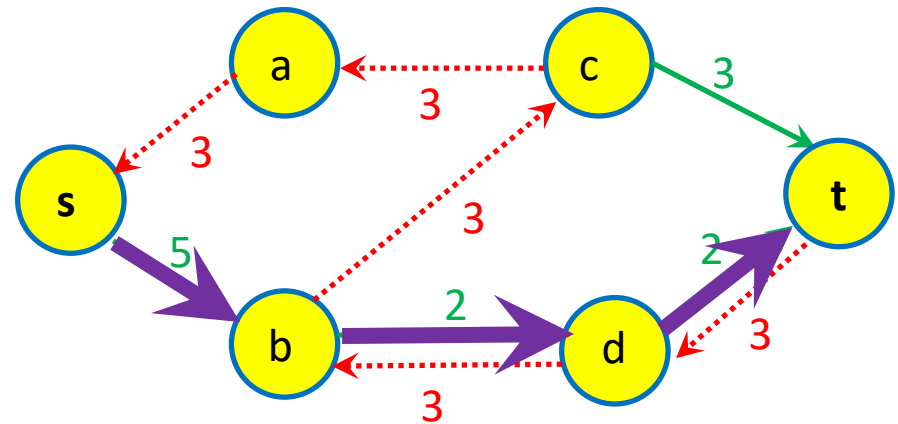
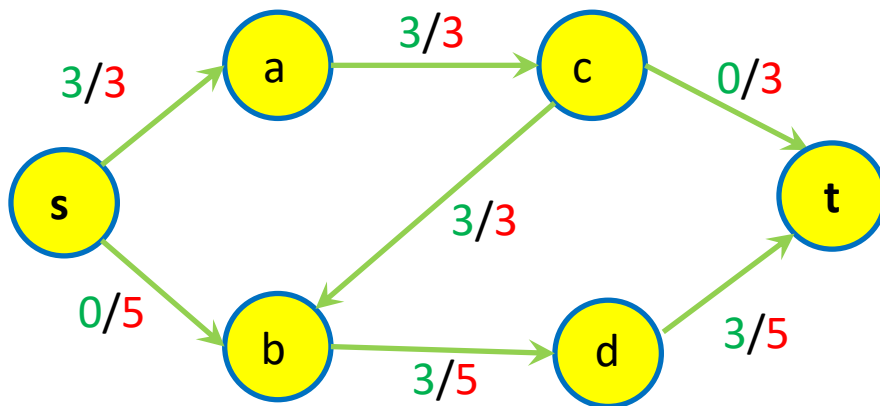
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Total flow: 3



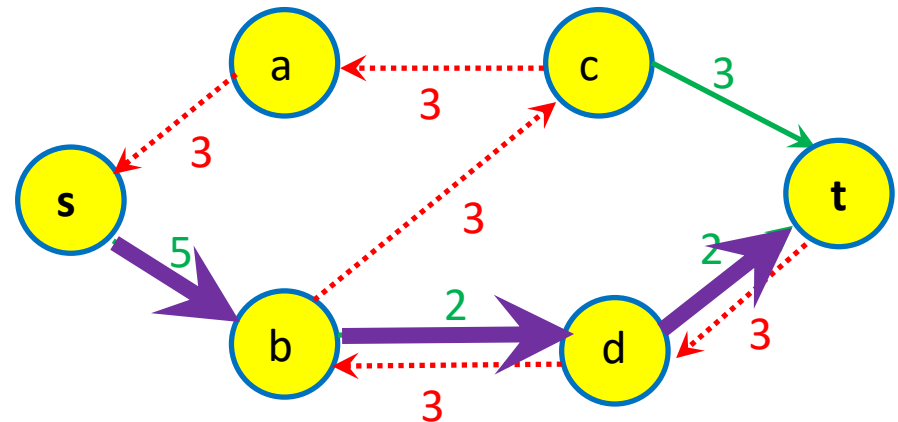
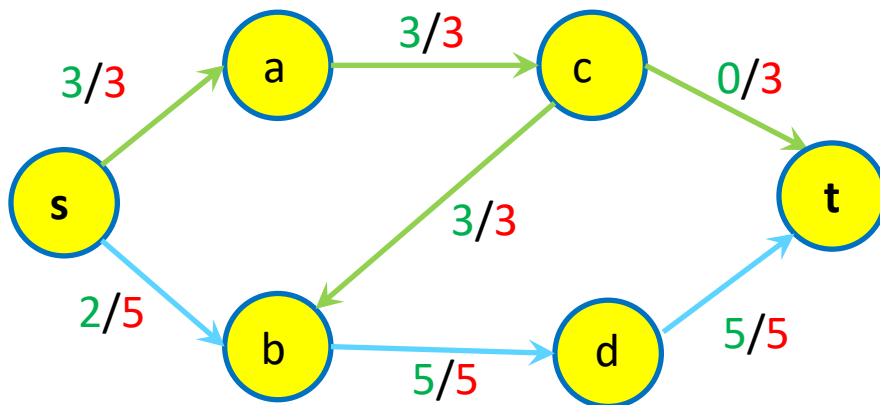
Path found: capacity = 2

Ford-Fulkerson Method – Worked example (by hand in lecture)

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 - 5: **return** f
-

Total flow: 5



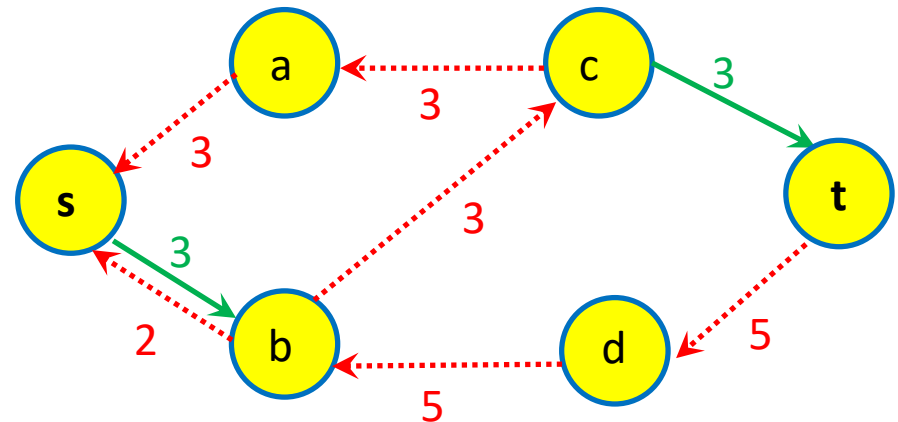
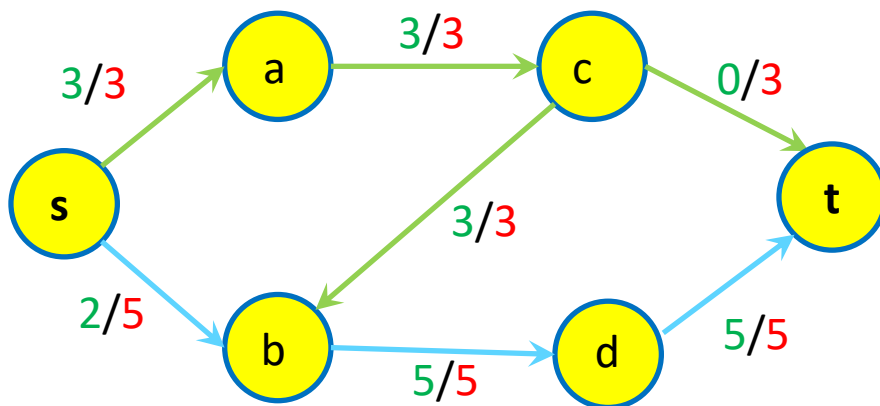
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Total flow: 5



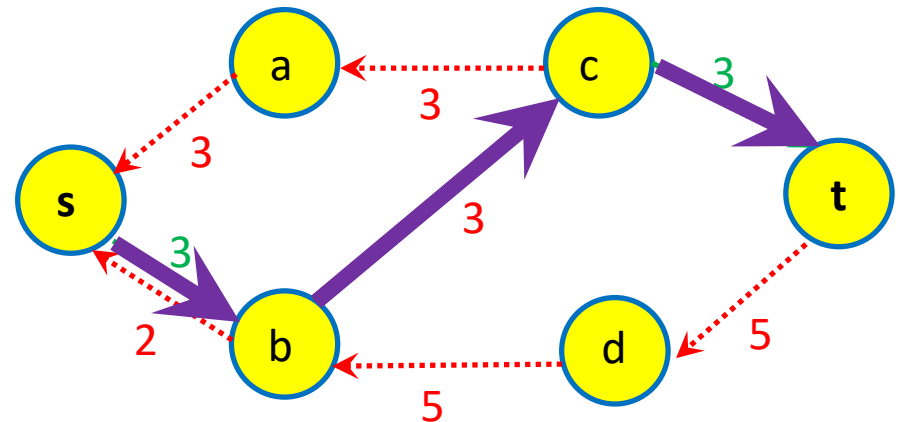
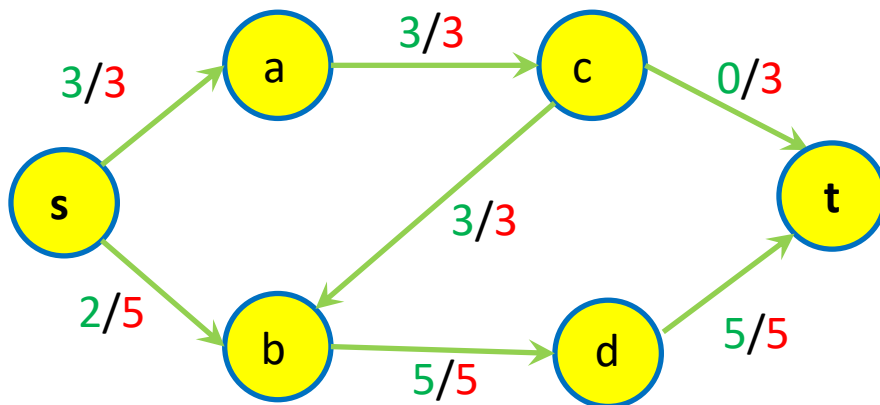
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Total flow: 5



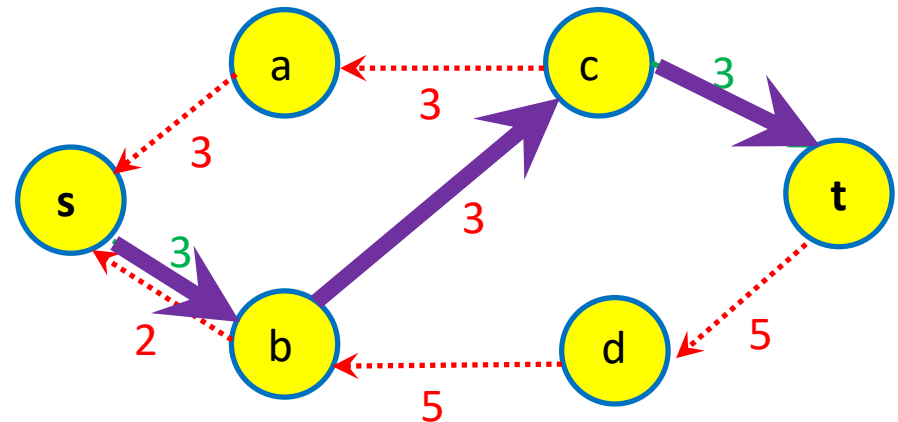
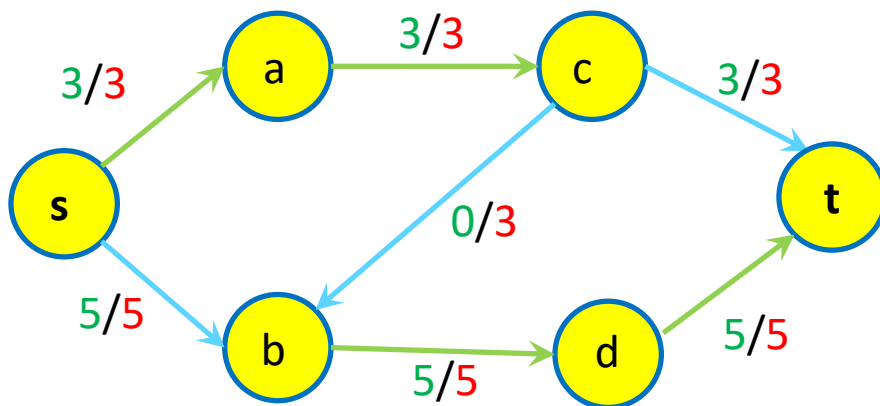
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Ford-Fulkerson Method – Worked example (by hand in lecture)

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Total flow: 8



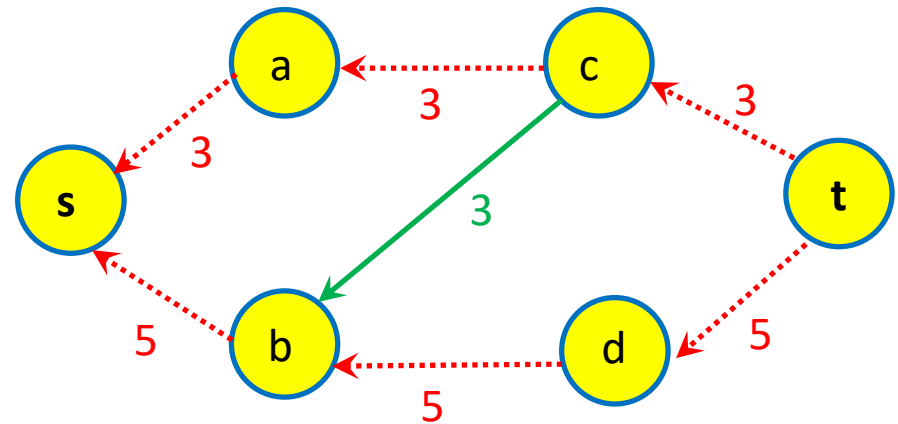
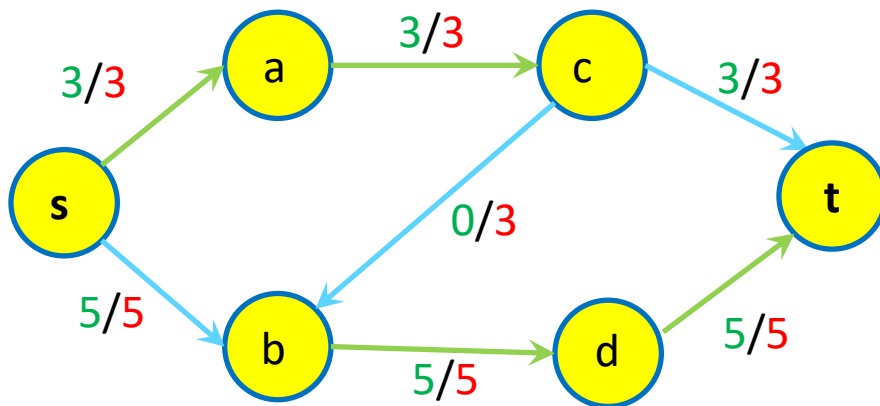
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Ford-Fulkerson Method – Worked example (by hand in lecture)

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Total flow: 8



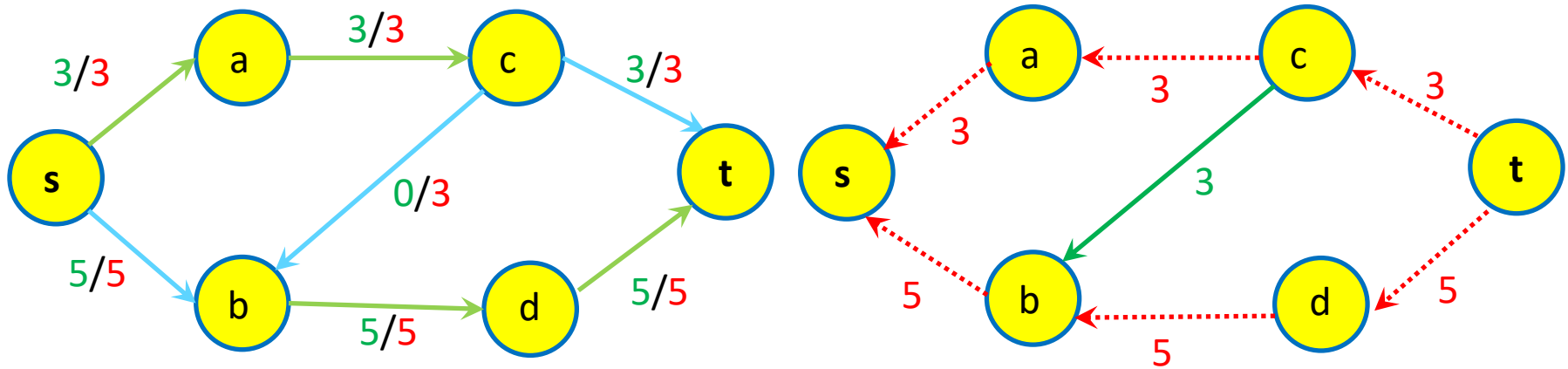
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-

Total flow: 8



No more augmenting paths!

Complexity Analysis

Algorithm 75 The Ford-Fulkerson method

```
1: function MAX_FLOW( $G = (V, E), s, t$ )
2:   Set initial flow  $f$  to 0 on all edges
3:   while there exists an augmenting path  $p$  in the residual network  $G_f$  do
4:     Augment the flow  $f$  along the augmenting path  $p$ 
5:   return  $f$ 
```

Cost of finding an augmenting path:

- Using BFS, cost is $O(V+E)$ (or just $O(E)$ since the graph is connected)

Augmenting flow along a path:

- length of path $\leq V$, so cost is $O(V)$

Updating the residual:

- Same amount of work as updating the flows along the augmenting path, $O(V)$

Total work in one iteration of the loop:

- **$O(V+E) = O(E)$**

Complexity Analysis

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```

Total work in one iteration of the loop:

- $O(V+E) = O(E)$

How many iterations?

- Assuming integer flows and capacities...
- Each iteration, flow grows by at least 1
- If maximum flow in graph is F
- Maximum number of iterations is also F

Total work:

- $O(EF)$

Complexity Analysis

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```

Total work:

- $O(EF)$

Not examinable:

- This looks polynomial
- It isn't because F is a number, so its **value** is exponential in the space required to store it
- It can be proven that the complexity is $O(VE^2)$ when using BFS to find augmenting paths, which is polynomial

Proof of Correctness

Algorithm 75 The Ford-Fulkerson method

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5:   return  $f$ 
```

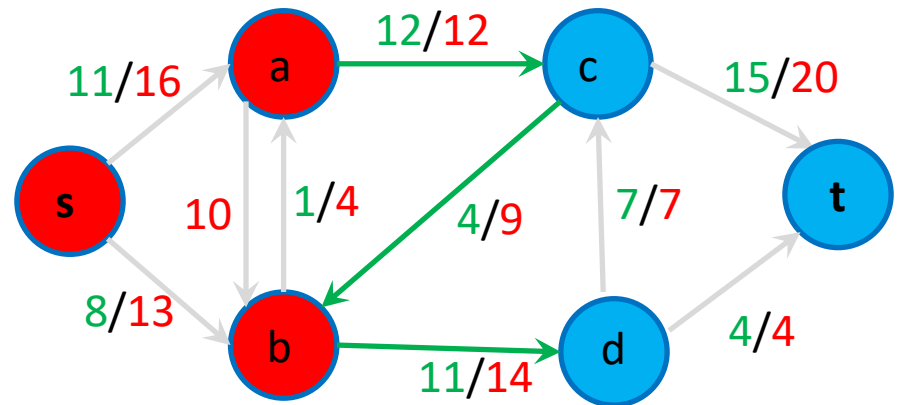
- **Does the algorithm terminate?**
 - Yes (assuming all capacities are integers), because
 - the flow always increases by at least 1 and the algorithm terminates when flow is equal to the maximum flow
- When the algorithm terminates (i.e., there is no augmenting path in residual network), the flow of the network is the maximum flow.
- We will need to understand “min-cut and max-flow” theorem to prove this fact

Outline

1. Maximum Flow Problem
2. Ford-Fulkerson Algorithm
3. **Min-cut Max-flow Theorem**

Flow and capacity of a cut

- A cut (S, T) of a flow network partitions the **vertices** of the network into two **disjoint** partitions S and T such that source s is in S and target t is in T .
 - E.g., $S = \{s, a, b\}$ and $T = \{t, c, d\}$
- Cut-set of a cut (S, T) is the set of edges that “cross” the cut, i.e., each edge connects one vertex in S with another in T .
 - E.g., the cut-set for the example is $a \rightarrow c$, $b \rightarrow d$, $c \rightarrow b$ (green edges)
 - The edges that have direction from a vertex in S to a vertex in T are called outgoing edges of the cut.
 - ✦ E.g., $a \rightarrow c$ and $b \rightarrow d$ are the outgoing edges of the cut
 - The edges that have direction from a vertex in T to a vertex in S are called incoming edges of the cut.
 - ✦ E.g., $c \rightarrow b$ is an incoming edge of the cut.



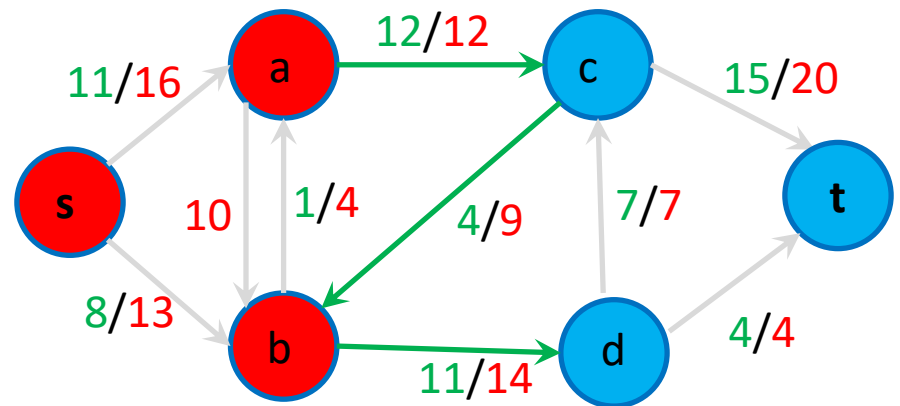
Flow and capacity of a cut

- Capacity of a cut (S, T) is **just** the total capacity of its **outgoing** edges
 - E.g., capacity of the cut in the example is $12 + 14 = 26$
- Flow of a cut (S, T) is Total flow of outgoing edges – total flow of incoming edges
 - E.g., flow in the example is $12 + 11 - 4 = 19$

Is it true that flow of a cut is always less than or equal to the capacity of the cut?

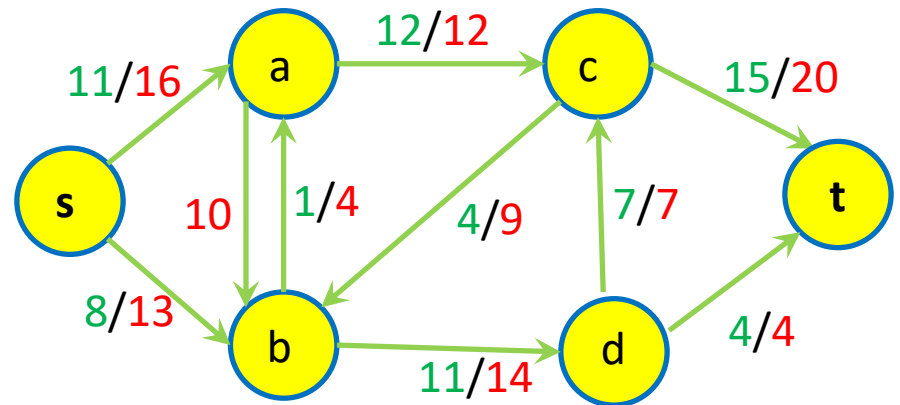
Yes, because

- Flow of an edge \leq capacity of an edge
- Capacity of a cut does not subtract capacities for incoming edges



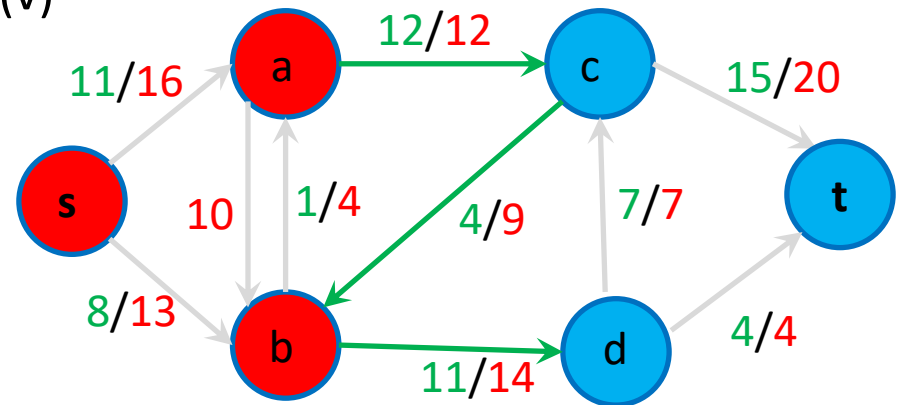
Flow and capacity of a cut

- Capacity of a cut (S,T) : the total capacity of its outgoing edges
- Flow of a cut (S,T) : Total flow of its outgoing edges – total flow of its incoming edges
- Assume $S = \{s, a, b, c, d\}$ and $T = \{t\}$.
 - What is the capacity of this cut?
 - What is the flow of this cut?
- Assume $S = \{s, a, b, d\}$ and $T = \{c, t\}$.
 - What is the capacity of this cut?
 - What is the flow of this cut?
- Assume $S = \{s, a\}$ and $T = \{b, c, d, t\}$.
 - What is the capacity of this cut?
 - What is the flow of this cut?
- What is the flow value of this network?
- Note: flow of all of the above cuts is 19
 - which is the same as flow of the network.
- I.e., flow of **every** cut = flow of the network
- Let's prove this formally



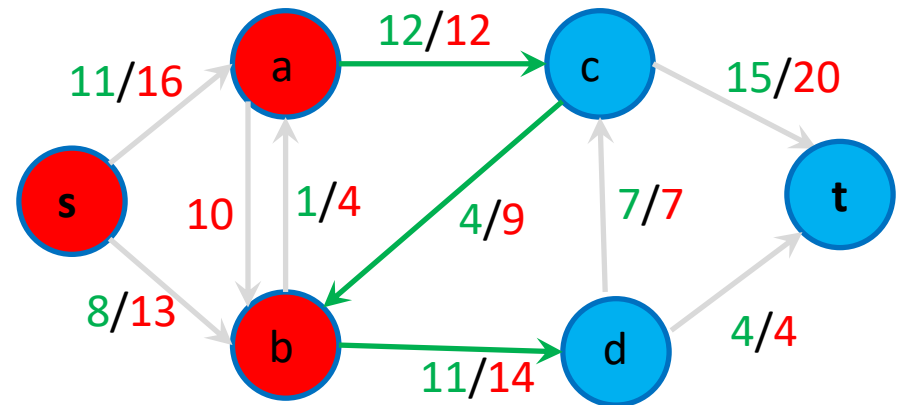
Flow of a cut = Flow of the network

- Let $F^{\text{out}}(v)$ be the total flow going out of a vertex and $F^{\text{in}}(v)$ be the total flow coming in the vertex
- Recall that flow of a network is the total flow going out from the source s .
 - Flow of the network = $F^{\text{out}}(s)$
- Flow conservation property: $F^{\text{out}}(v) - F^{\text{in}}(v) = 0$ for every vertex except s and t
- Flow of the network = $F^{\text{out}}(s)$
- Flow of the network = $F^{\text{out}}(s) + \sum_{v \in S \setminus s} F^{\text{out}}(v) - F^{\text{in}}(v)$
 - recall S is the cut containing s and excluding t
- Since $F^{\text{in}}(s) = 0$, we can rewrite the flow as.
- Flow of the network = $\sum_{v \in S} F^{\text{out}}(v) - F^{\text{in}}(v)$



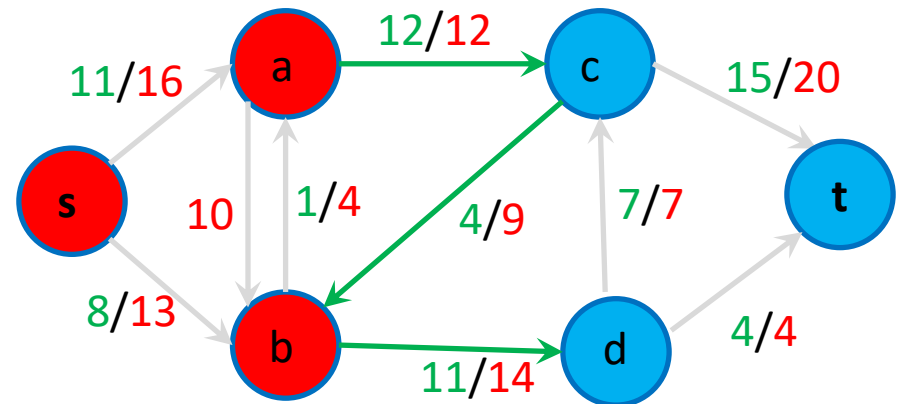
Flow of a cut = Flow of the network

- Flow of the network = $\sum_{v \in S} F^{\text{out}}(v) - F^{\text{in}}(v)$
- Each vertex v in S (red vertices) can have two types of edges
 - Grey edges (the edges that connect the vertex to another vertex in S)
 - Green edges (the edges that connect the vertex to a vertex in T)
 - Let $F^{\text{out-grey}}(v)$ be the total flow out from v via grey edges. Similarly, $F^{\text{in-grey}}(v)$ be the total flow coming to v via grey edges.
 - Let $F^{\text{out-green}}(v)$ be the total flow out from v via green edges. Similarly, $F^{\text{in-green}}(v)$ be the total flow coming to v via green edges.
 - We have $F^{\text{out-green}}(v) + F^{\text{out-grey}}(v) = F^{\text{out}}(v)$ and $F^{\text{in-green}}(v) + F^{\text{in-grey}}(v) = F^{\text{in}}(v)$



Flow of a cut = Flow of the network

- We have $F^{\text{out-green}}(v) + F^{\text{out-grey}}(v) = F^{\text{out}}(v)$ and $F^{\text{in-green}}(v) + F^{\text{in-grey}}(v) = F^{\text{in}}(v)$
- From before, Flow of the network = $\sum_{v \in S} F^{\text{out}}(v) - F^{\text{in}}(v)$
- Flow of the network = $\sum_{v \in S} F^{\text{out-green}}(v) + F^{\text{out-grey}}(v) - (F^{\text{in-green}}(v) + F^{\text{in-grey}}(v))$
- Flow of the network = $\sum_{v \in S} F^{\text{out-green}}(v) - F^{\text{in-green}}(v) + F^{\text{out-grey}}(v) - F^{\text{in-grey}}(v)$
- Note that $\sum_{v \in S} F^{\text{out-grey}}(v) - F^{\text{in-grey}}(v) = 0$ because each grey edge appears once as an incoming edge for one vertex and once as an outgoing edge for another vertex.
- Flow of the network = $\sum_{v \in S} F^{\text{out-green}}(v) - F^{\text{in-green}}(v)$
- Flow of the network = *Flow of the cut*



Min-cut Max-Flow Theorem

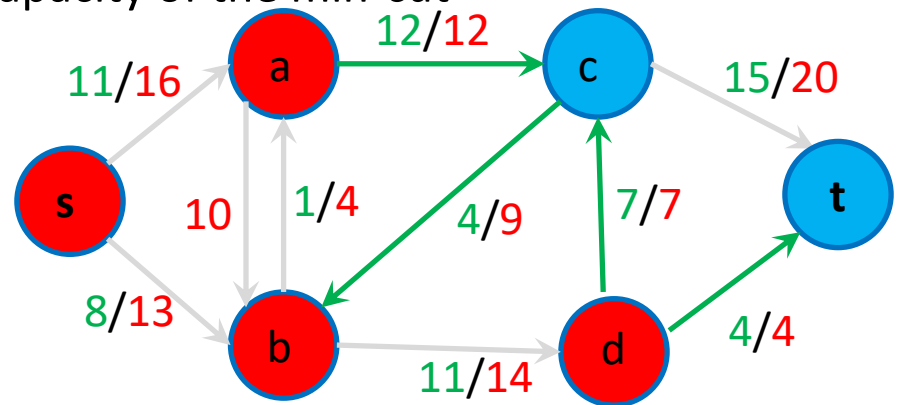
- Min-cut of a flow network is the cut with the minimum capacity

We know that

1. Flow of a cut \leq capacity of the cut
 2. Flow of **every** cut = Flow of the network
- Therefore, Maximum possible flow of the network \leq capacity of every cut
 - Or, Maximum possible flow of the network \leq capacity of min-cut
 - What if we can find a cut such that the flow of the network = capacity of the cut
 - This would mean flow of the network is the maximum possible (we have found maximum possible flow)
 - The cut is the min-cut of the flow network

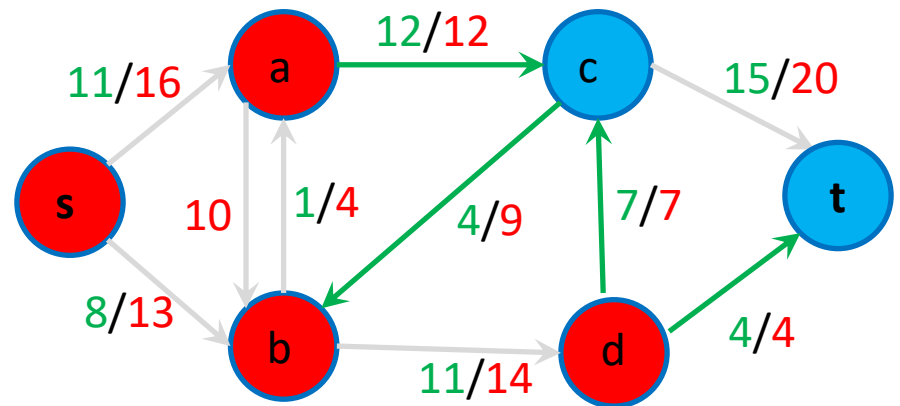
Min-cut Max-Flow Theorem

- Maximum possible flow of a network = capacity of the min-cut



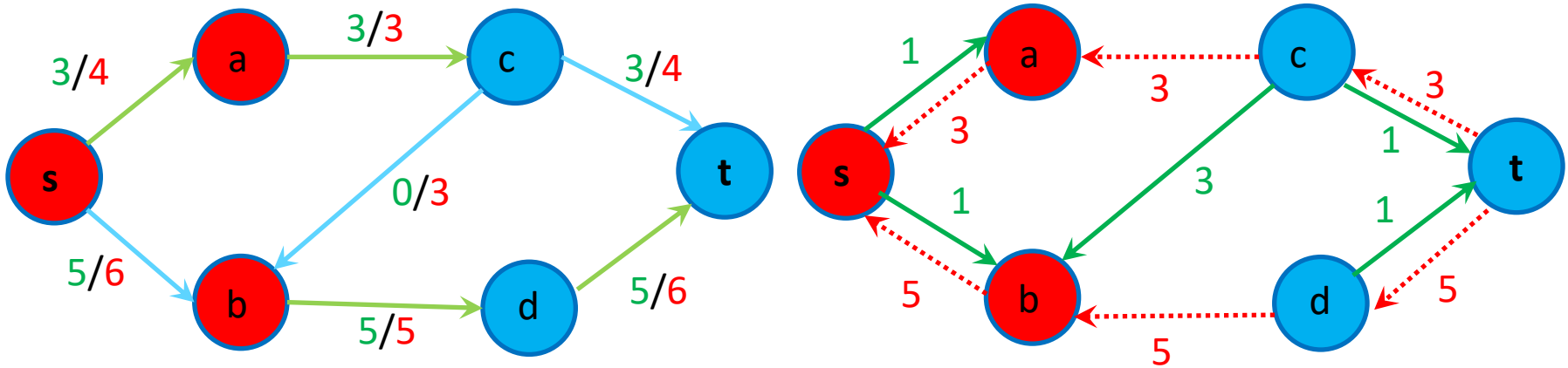
Min-cut Max-Flow Theorem

- Capacity of a cut (S,T) is the total capacity of its **outgoing** edges
- Flow of a cut (S,T) = Total flow of its **outgoing** edges – total flow of its **incoming** edges
- Flow of a cut = capacity of the cut when:
 1. Flow on **each outgoing** edge of the cut is equal to the capacity of the edge; AND
 2. Flow on **each incoming** edge of the cut is zero
- We show that when the algorithm terminates, there exists a cut for which both of the above two conditions hold which imply that the flow of the cut is equal to its capacity.
 - This guarantees that the flow is maximum



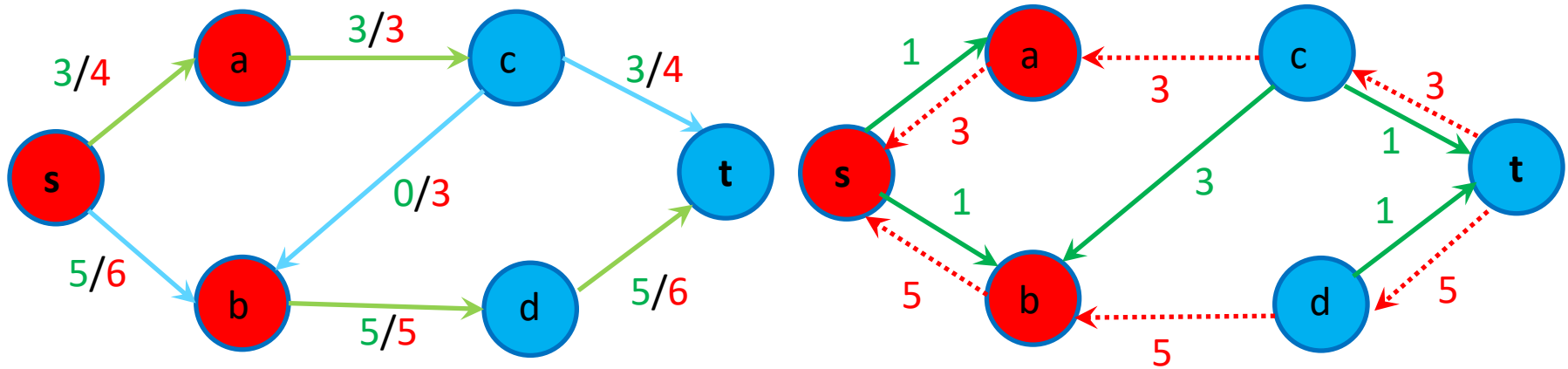
Proof of correctness

- Suppose the algorithm has terminated (there does not exist any augmenting path in the residual network).
- We define a cut (S, T) such that
 - S contains every vertex v that is reachable from s in **the residual network**.
 - T contains every other vertex. Note t cannot be in S because it is not reachable from S (no augmenting path)



Proof of correctness

- Flow of this cut = Capacity of this cut because
 - For each outgoing edge $a \rightarrow c$, its flow is equal to the capacity of the edge
 - ✦ Otherwise, we would have an edge $a \rightarrow c$ in the residual network which would mean c is reachable from s but we know this is not the case as c is not in S .
 - For each incoming edge $c \rightarrow b$, its flow is zero.
 - ✦ Otherwise, there would be an edge $b \rightarrow c$ in the residual network implying c is reachable from s but we know this is not the case as c is not in S .
- Therefore, the flow is maximum when the algorithm terminates



Summary

Take home message

- Maximum flow of a network is equal to its min-cut and can be found using Ford-Fulkerson

Things to do (this list is not exhaustive)

- Make sure you understand
 - the two algorithms
 - understand why Ford-Fulkerson is correct
- Start preparing for the final exam

Coming Up Next

- Topological sorting and final exam
- Revision – bring any questions you would like to discuss to next week's lecture!