

COMP261

Algorithms and Data Structures

2024 Tri 1

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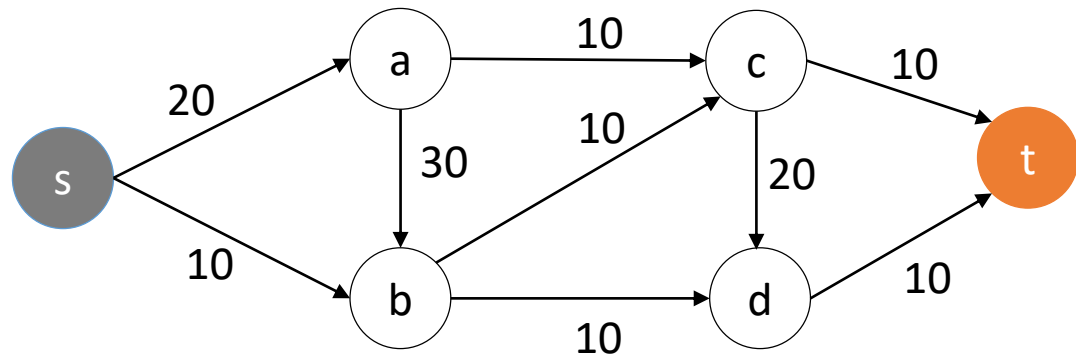
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Office Hours (COMP261): AM414, Thursday 10:00 – 12:00

Recap: Definition

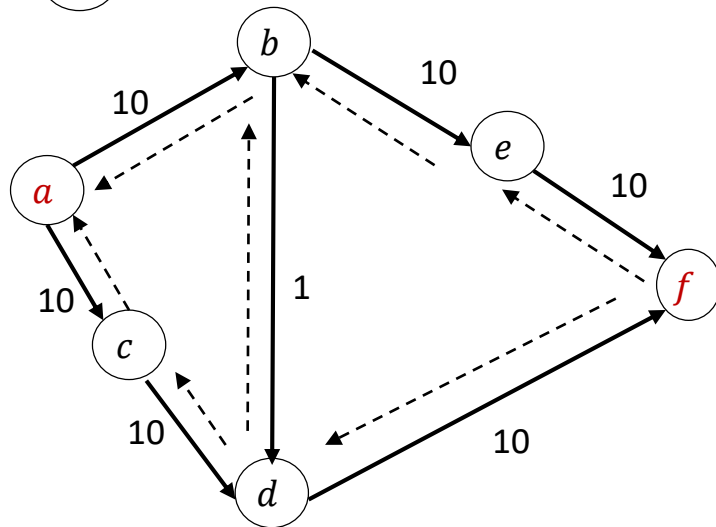
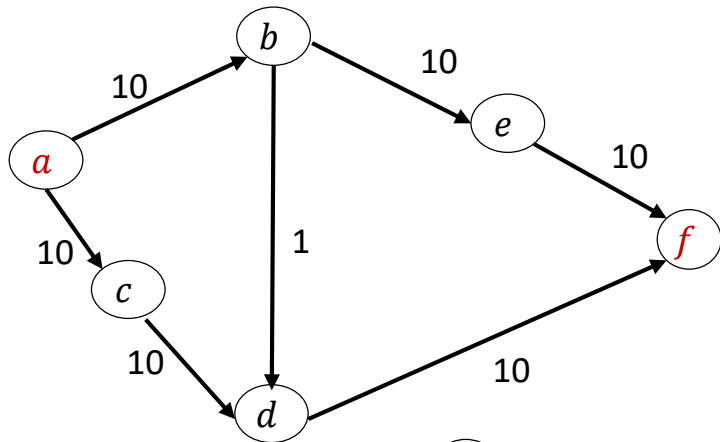
A flow network is a connected, directed graph $G = (V, E)$ where:

- Each edge, e , has a non-negative, integer **capacity** c_e .
- One (or more) vertex is labelled as a **source** $s \in V$.
- One (or more) vertex is labelled as **sink** $t \in V$.
- No edge enters the source and no edge leaves the sink.



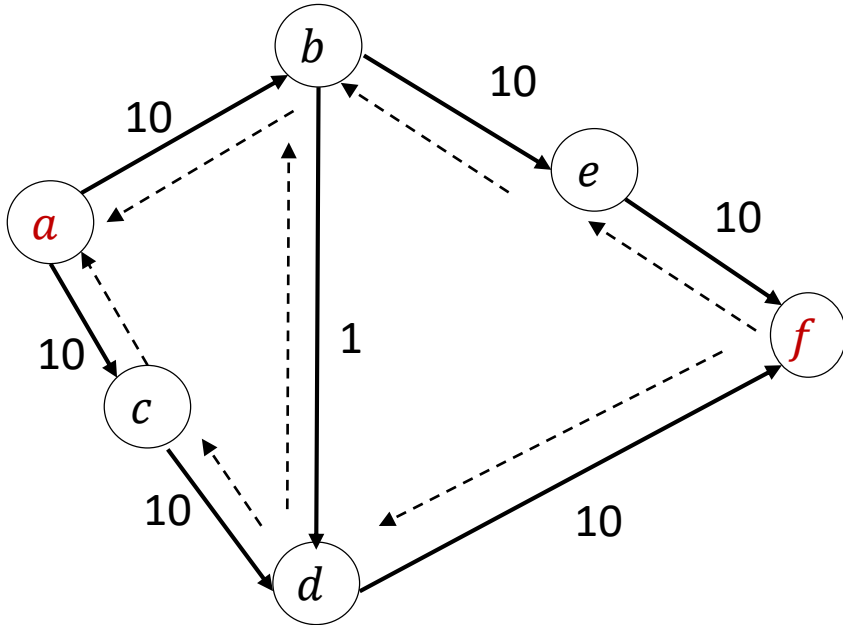
Recap: Solving Maximum flow problem

How can we solve this problem ?



1. Look for paths from source to destination and count them if they have capacities left
2. Repeat until no paths from source to destination is found with capacities left
3. There could be a possibility that the paths we choose are not optimal – have an option of reversing the decisions. **For every edge in the original graph, we add an “imaginary” reverse edge**

Solving Maximum flow problem



$a \rightarrow b \rightarrow d \rightarrow f$ 1

$a \rightarrow c \rightarrow d \rightarrow f$ 9

$a \rightarrow b \rightarrow e \rightarrow f$ 9

$a \rightarrow c \rightarrow d \rightarrow b \rightarrow e \rightarrow f$ 1

Ford-Fulkerson method

It is generally called a method rather than an algorithm, as it encompasses several different implementations with different running times.

- Based on 2 important ideas:
 - Residual Graphs (includes reverse edges)
 - Augmentation paths (path in a residual graph)

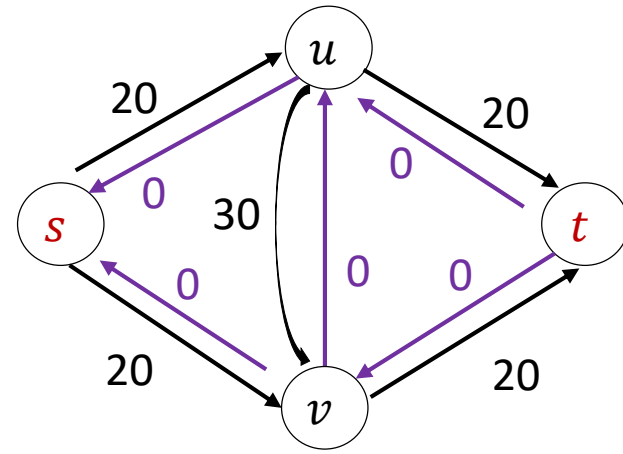
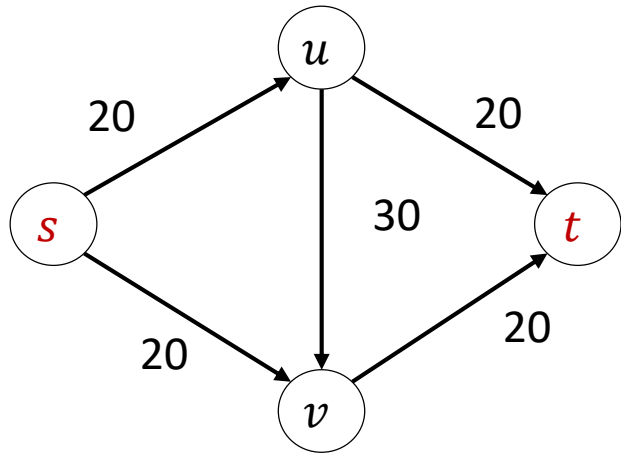
Residual Graph

Given a network G and a flow f , we construct a **residual graph** G_f , representing places where flow can still be added to f , including places where existing flow can be decreased.

- G_f is defined as follows:
 - G_f contains same nodes as G .
 - **Forward edges**: for each edge $e = (u, v)$ of G for which $f_e < c_e$, include an edge $e' = (u, v)$ in G_f with capacity $c_e - f_e$
 - **Backward edges (represent decreasing flow)**: for each edge $e = (u, v)$ of G with $f_e > 0$, include an edge $e' = (v, u)$ in G_f with capacity f_e

Residual network

- Initially flow is 0 for all edges.



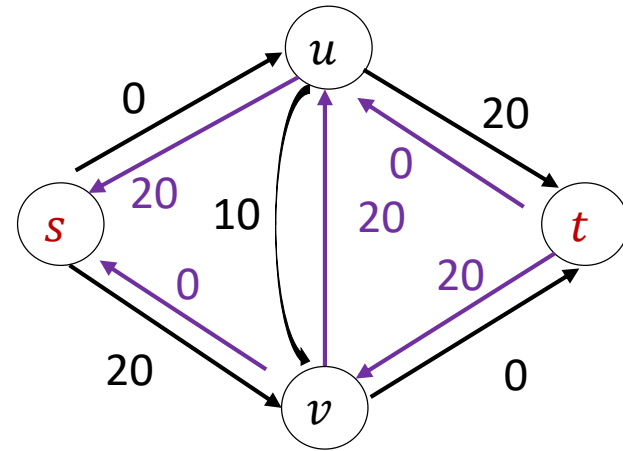
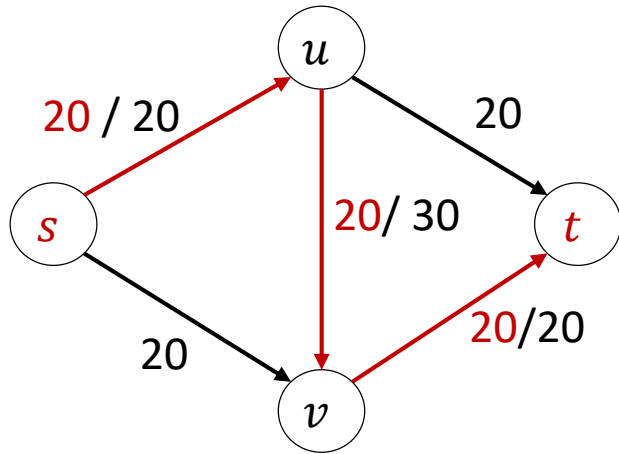
Residual network

G_f contains same nodes as G .

Forward edges: for each edge $e = (u, v)$ of G for which $f_e < c_e$, include an edge $e' = (u, v)$ in G_f with capacity $c_e - f_e$

Backward edges: for each edge $e = (u, v)$ of G with $f_e > 0$, include an edge $e' = (v, u)$ in G_f with capacity f_e

Residual network

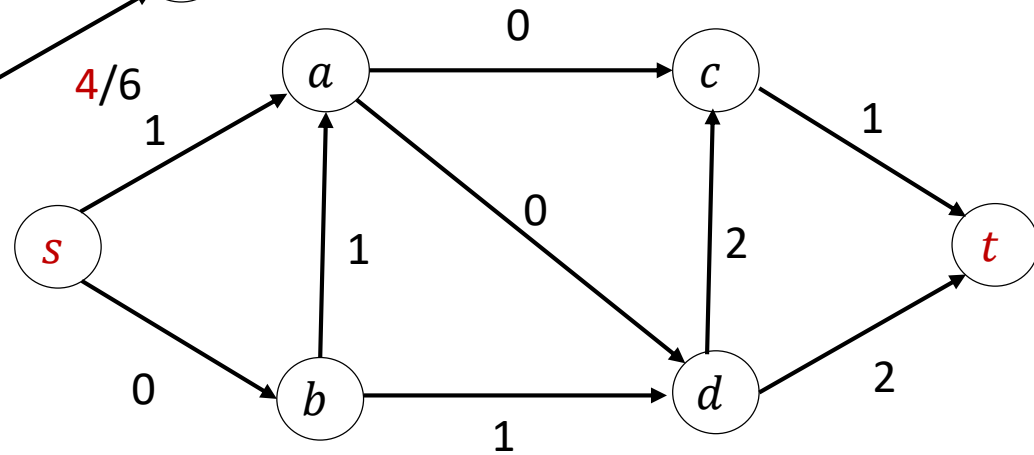
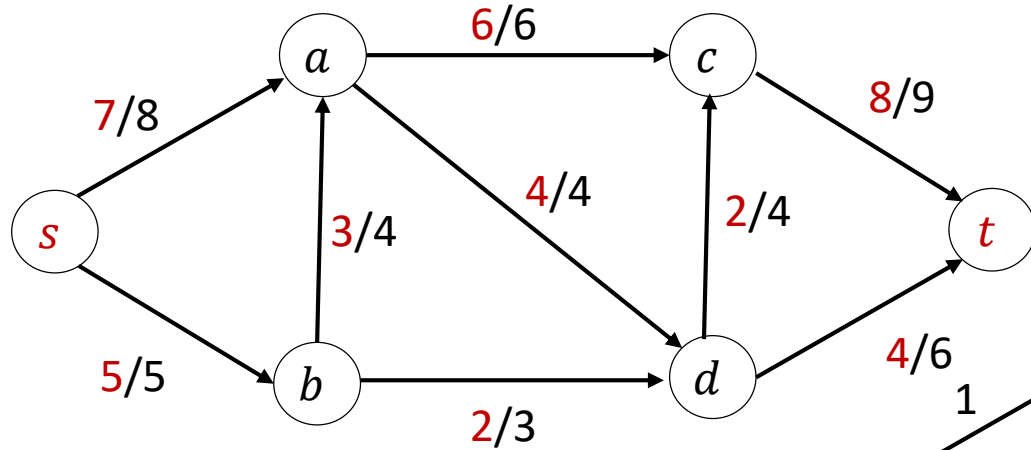


G_f contains same nodes as G .

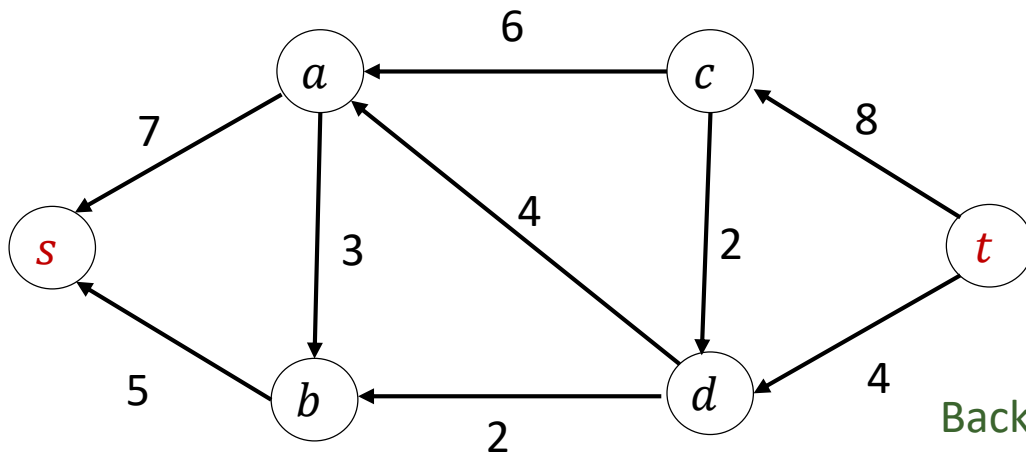
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Backward edges: for each edge $e = (u, v)$ of G with $f_e > 0$, include an edge $e' = (v, u)$ in G_f with capacity f_e

Example

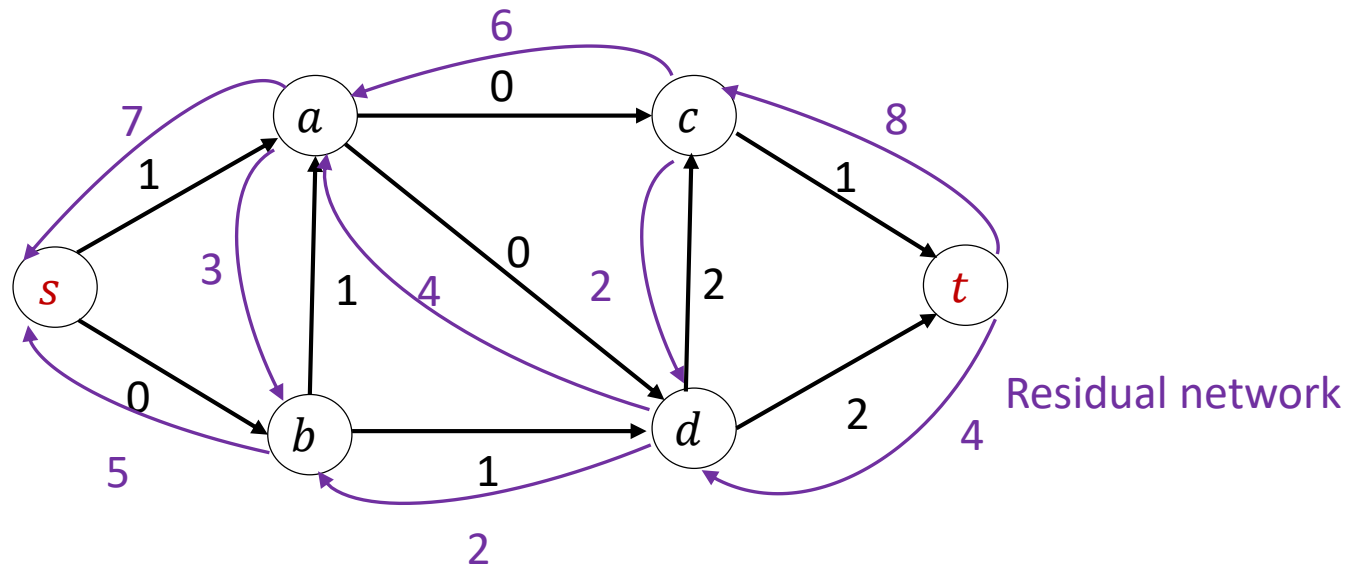
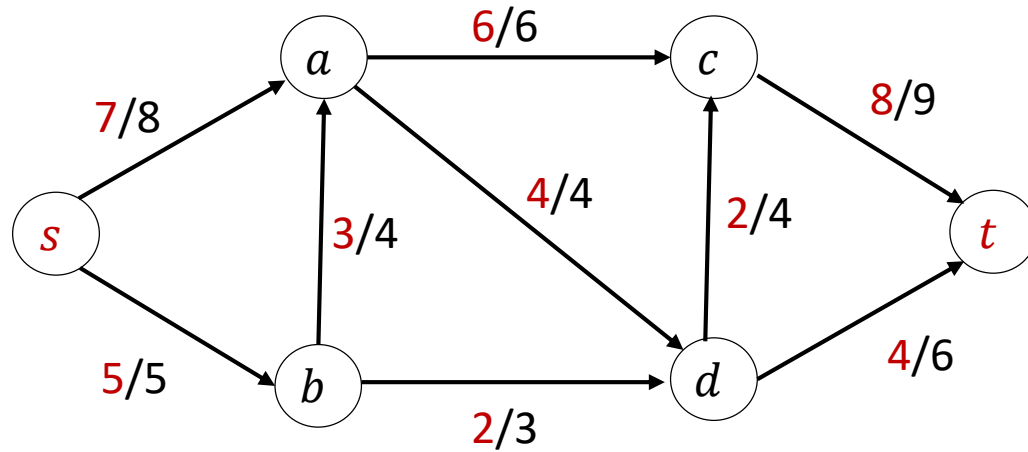


Forward edges



Backward edges

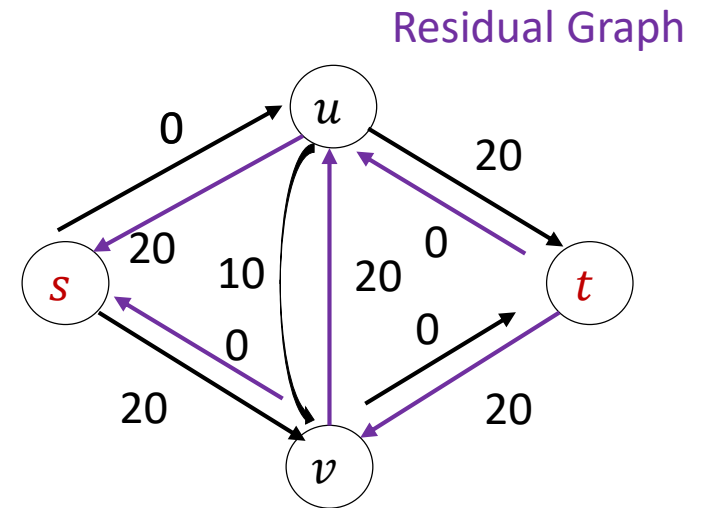
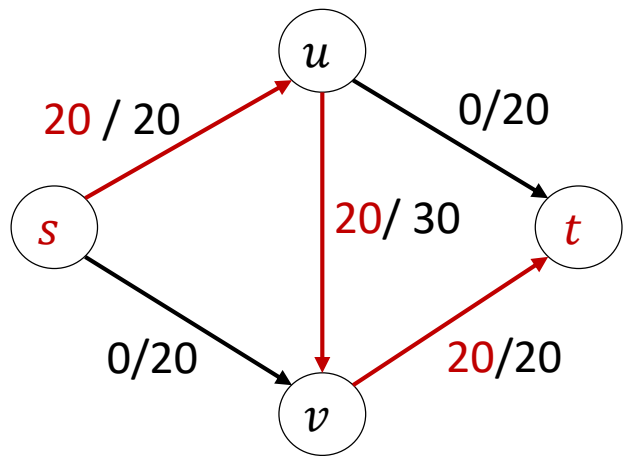
Example



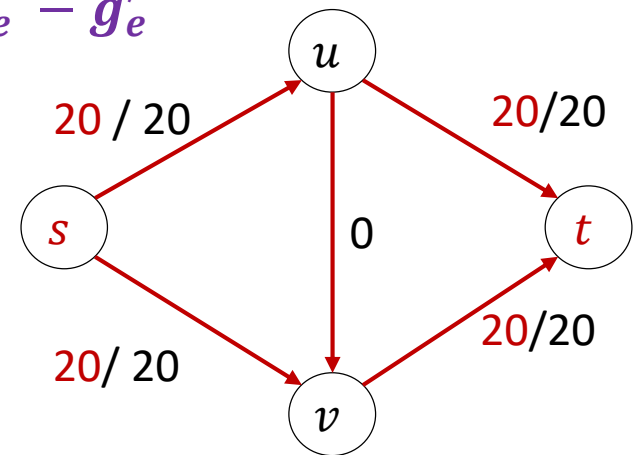
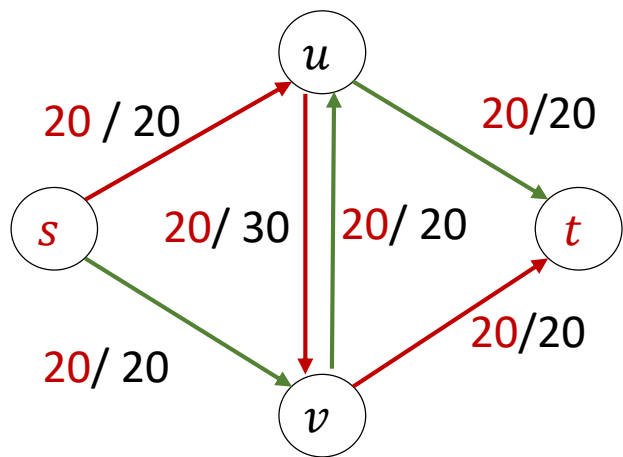
Augmentation Path

- **Augmentation path**: Given a network G and a flow f . Augmentation path is any flow g on residual graph G_f from source s to destination t .
- Augmentation path g can be added to f to get a new flow on G .
 - g_e (forward edge) adds to f_e
 - g'_e (backward edge) subtracts from f_e (equivalent of reversing our previous decision)

Example

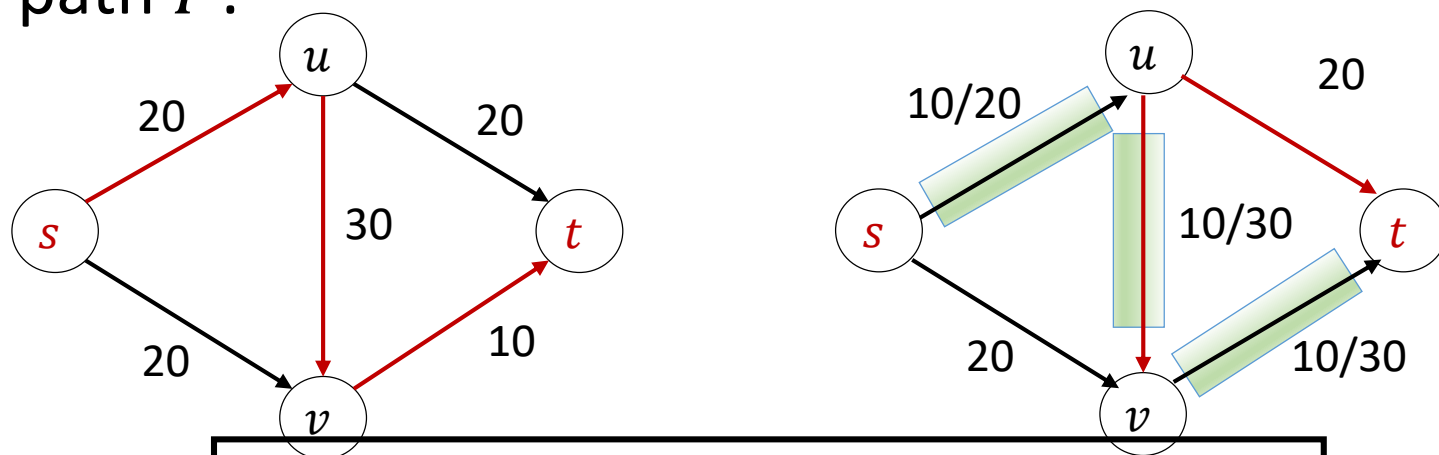


Overall flow is given by:
 $f_e + g_e - g'_e$



Augmenting Paths

- How much flow can be added in each step?
- **Bottleneck** (P, G_f) : the smallest capacity in G_f on any edge of P . If $\text{bottleneck}(P, G_f) > 0$ then we can increase the flow by sending $\text{bottleneck}(P, G_f)$ along the path P .



$$\text{Bottleneck}(P, G_f) = \min\{(c_e) : e \text{ is in } P\}$$

Ford-Fulkerson method

- Follows a greedy approach. Iteratively increase the value of flow.

FordFulkerson(G, s, t)

Let $f_e = 0$ for all edges (no flow anywhere)

Initialize **Residual Graph** RG // for every forward edge in the original graph G add a reverse edge with a capacity 0

$maxFlow = 0$

Repeat

Find some P path from s to t in RG such that $c_e > 0$ for all edges in P

if P exists

$pathFlow = Bottleneck(P, RG)$

$maxFlow = maxFlow + pathFlow$

Output ($P, pathflow$) as an augmentation path

Update RG

Endif

Until the RG has no more augmentation paths.

Output $maxFlow$

Note: Ford-Fulkerson does not state how to find augmentation paths.

Updating Residual Graph

For each edge $e = (u, v) \in P$:

$$f_{(u,v)} = f_{(u,v)} + pathFlow \text{ //add flow}$$

$$c_{(u,v)} = c_{(u,v)} - pathFlow \text{ //reduce capacity}$$

$$c_{(v,u)} = c_{(v,u)} + pathFlow \text{ //increase capacity in reverse edge}$$

EndFor

Edmonds-Karp algorithm is an implementation of the Ford-Fulkerson method that uses **BFS** for finding augmenting paths.

BFS(RG, s, t)

Define *augPath* as ArrayList of edges

q := *queue*()

q.push(*s*)

backpointer(*v*) = *null* for all *v* // backpointer data-structure to hold
//edges that lead to the vertex

while !*q.isEmpty*()

cur := *q.pull*()

 for each outedge *e* of *cur* in *RG* do

 if *e.toCity* ≠ *s* and *backpointer*(*e.toCity*) == *null* and *e.cap* ≠ 0

backpointer(*e.toCity*) := *e*

 if(*backpointer*(*t*) ≠ *null*) // found a path from *s* to *t*. Build it now from reverse

pathEdge = *backpointer*(*t*)

 while(*pathEdge* ≠ *null*)

augPath.add(*pathEdge*)

pathEdge = *backpointer*(*pathEdge.fromCity*)

 endWhile

Collections.reverse(*augPath*)

 return *augPath*

 endIf

q.push(*e.toCity*)

 endIf

 endFor

endWhile

return *null*

Next Lecture

- Example – Edmonds-Karp algorithm