



CSE408

Maximum Flow

Lecture #30

Maximum Flow

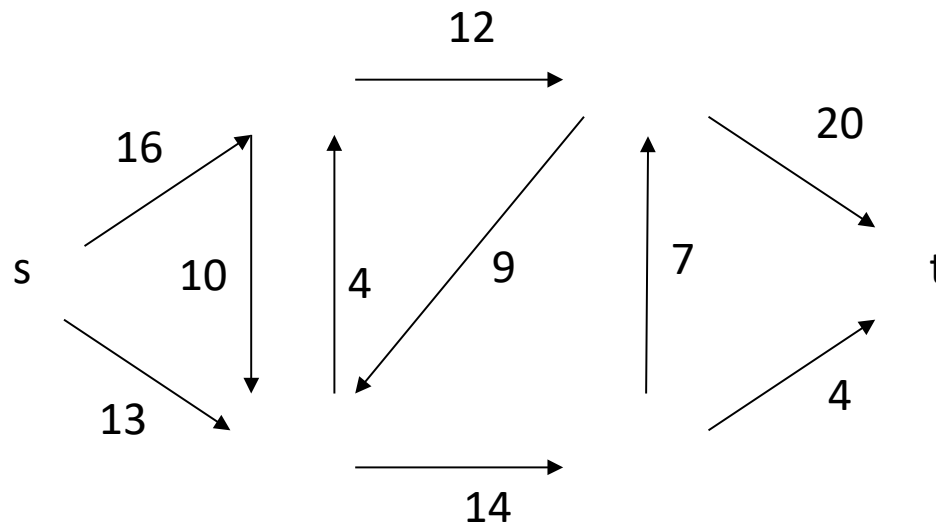


- Maximum Flow Problem
- The Ford-Fulkerson method
- Maximum bipartite matching

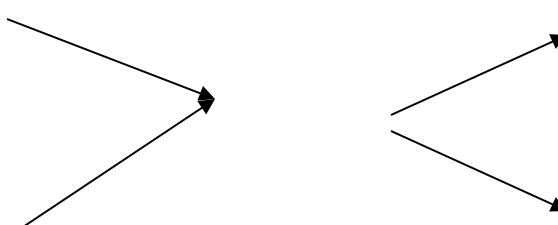
Flow networks:



- A **flow network** $G=(V,E)$: a directed graph, where each edge $(u,v) \in E$ has a nonnegative **capacity** $c(u,v) \geq 0$.
- If $(u,v) \notin E$, we assume that $c(u,v)=0$.
- two distinct vertices : a **source** s and a **sink** t .



- $G=(V,E)$: a flow network with capacity function c .
- s -- the source and t -- the sink.
- A flow in G : a real-valued function $f:V \times V \rightarrow \mathbb{R}$ satisfying the following two properties:
- **Capacity constraint**: For all $u,v \in V$,
we require $f(u,v) \leq c(u,v)$.
- **Flow conservation**: For all $u \in V - \{s,t\}$, we require

$$\sum_{e \text{ in } v} f(e) = \sum_{e \text{ out } v} f(e)$$


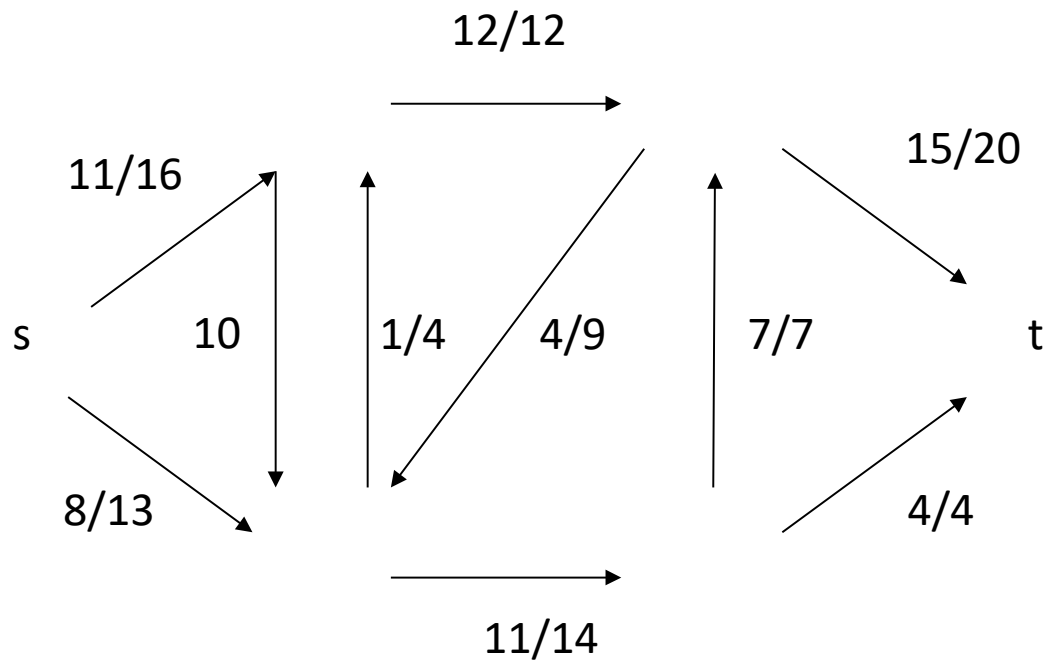
Net flow and value of a flow f:



- The quantity $f(u, v)$ is called the **net flow** from vertex u to vertex v .
- The **value** of a flow is defined as

$$|f| = \sum_{v \in V} f(s, v)$$

- The total flow from source to any other vertices.
- The same as the total flow from any vertices to **the sink**.



A flow f in G with value $|f| = 19$

Maximum-flow problem:



- Given a flow network G with source s and sink t
- Find a flow of maximum value from s to t .
- How to solve it efficiently?



The Ford-Fulkerson method:



- This section presents the Ford-Fulkerson method for solving the maximum-flow problem. We call it a “method” rather than an “algorithm” because it encompasses several implementations with different running times. The Ford-Fulkerson method depends on three important ideas that transcend the method and are relevant to many flow algorithms and problems: **residual networks, augmenting paths, and cuts**. These ideas are essential to the important max-flow min-cut theorem, which characterizes the value of maximum flow in terms of cuts of the flow network.

Continue:



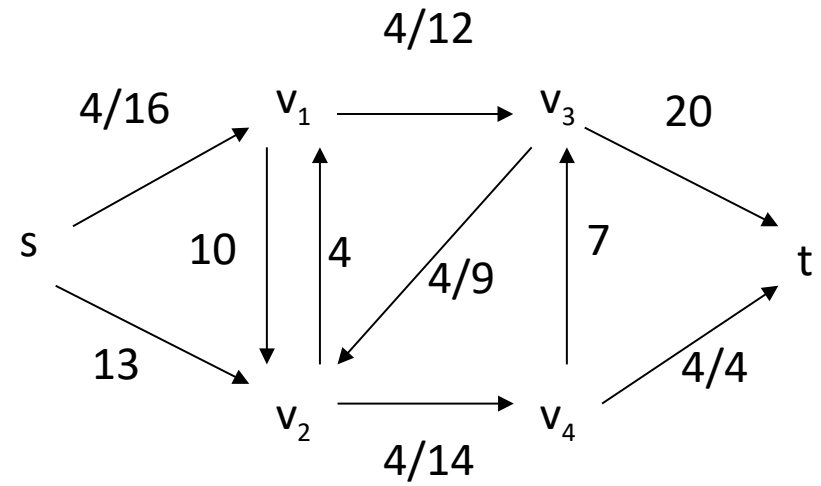
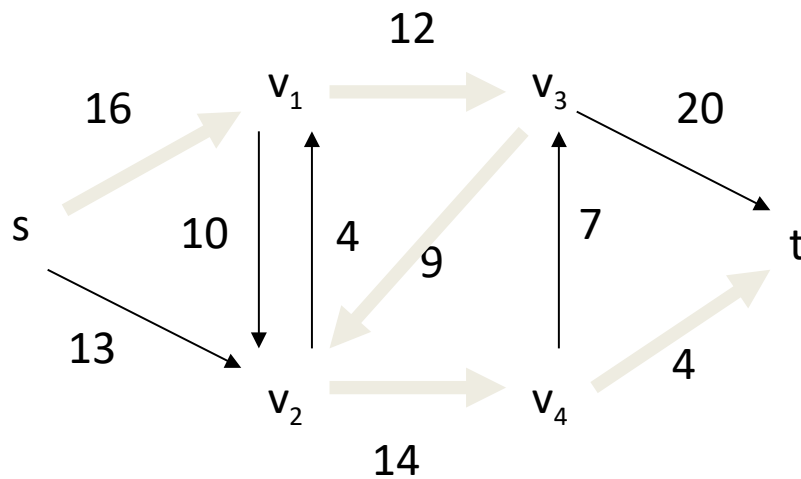
- FORD-FULKERSON-METHOD(G, s, t)
- initialize flow f to 0
- **while** there exists an *augmenting* path p
- **do** *augment* flow f along p
- return f

Residual networks:



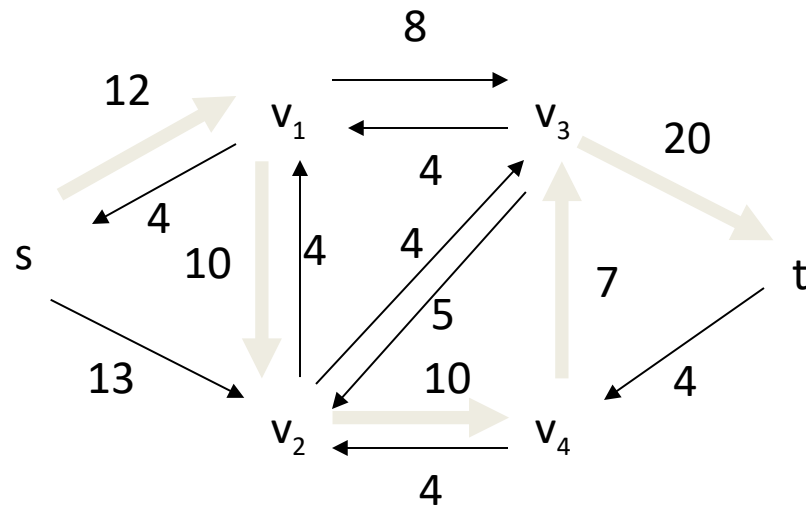
- Given a flow network and a flow, the **residual network** consists of edges that can admit more net flow.
- $G=(V,E)$ --a flow network with source s and sink t
- f : a flow in G .
- The amount of additional net flow from u to v before exceeding the capacity $c(u,v)$ is the **residual capacity** of (u,v) , given by: $c_f(u,v)=c(u,v)-f(u,v)$
in the other direction: $c_f(v, u)=c(v, u)+f(u, v)$.

Example of residual network



(a)

Example of Residual network (continued)



(b)

Fact 1:



- Let $G=(V,E)$ be a flow network with source s and sink t , and let f be a flow in G .
- Let G_f be the residual network of G induced by f , and let f' be a flow in G_f . Then, the flow sum $f+f'$ is a flow in G with value
- $f+f'$: the flow in the same direction will be added.
the flow in different directions will be cancelled.

Augmenting paths:

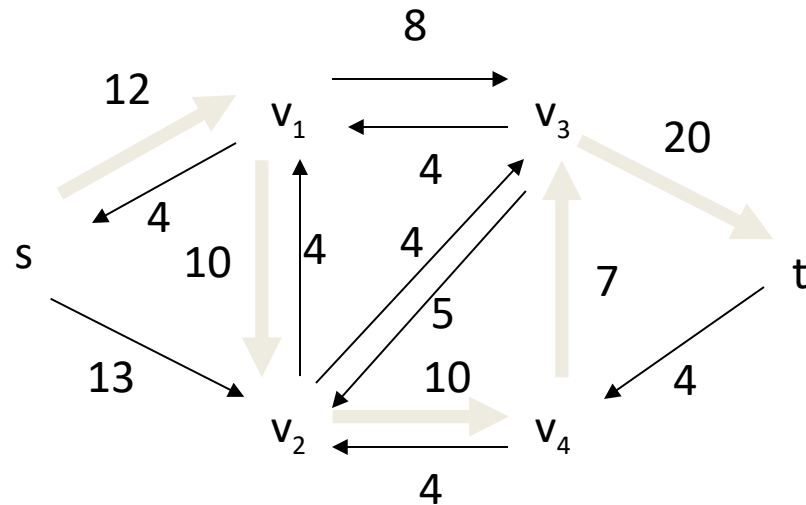


- Given a flow network $G=(V,E)$ and a flow f , an **augmenting path** is a simple path from s to t in the residual network G_f .
- **Residual capacity** of p : the maximum amount of net flow that we can ship along the edges of an augmenting path p , i.e., $c_f(p)=\min\{c_f(u,v):(u,v) \text{ is on } p\}$.



The residual capacity is 1.

Example of an augment path (bold edges)



(b)

The basic Ford-Fulkerson algorithm:

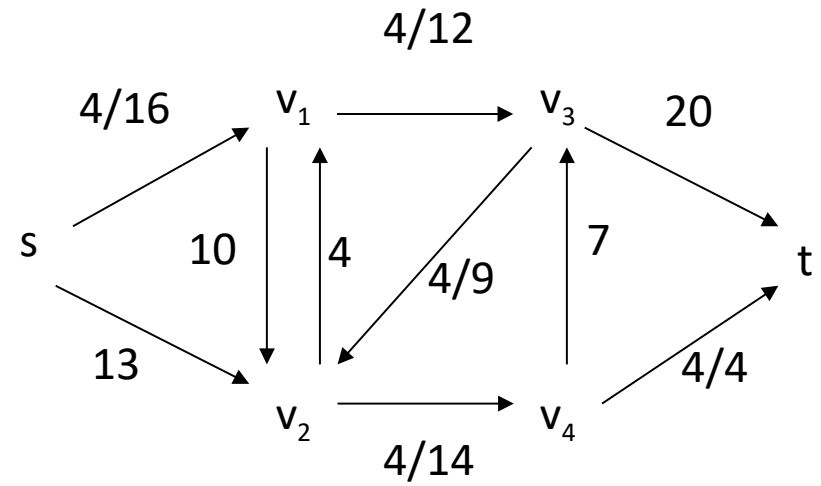
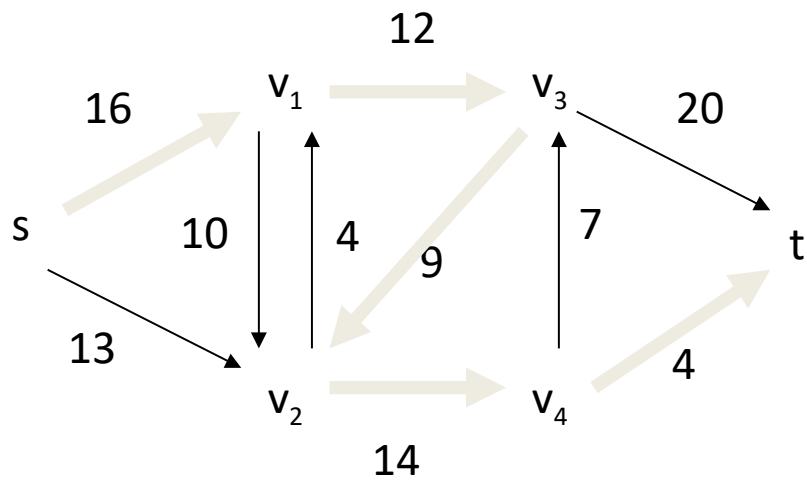


- FORD-FULKERSON(G, s, t)
- **for** each edge $(u, v) \in E[G]$
- **do** $f[u, v] \leftarrow 0$
- $f[v, u] \leftarrow 0$
- **while** there exists a path p from s to t in the residual network G_f
- **do** $c_f(p) \leftarrow \min\{c_f(u, v) : (u, v) \text{ is in } p\}$
- **for** each edge (u, v) in p
- **do** $f[u, v] \leftarrow f[u, v] + c_f(p)$
-

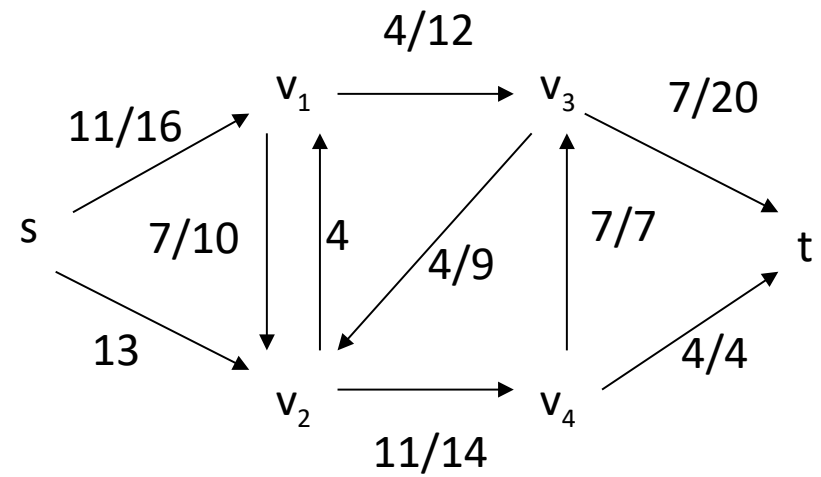
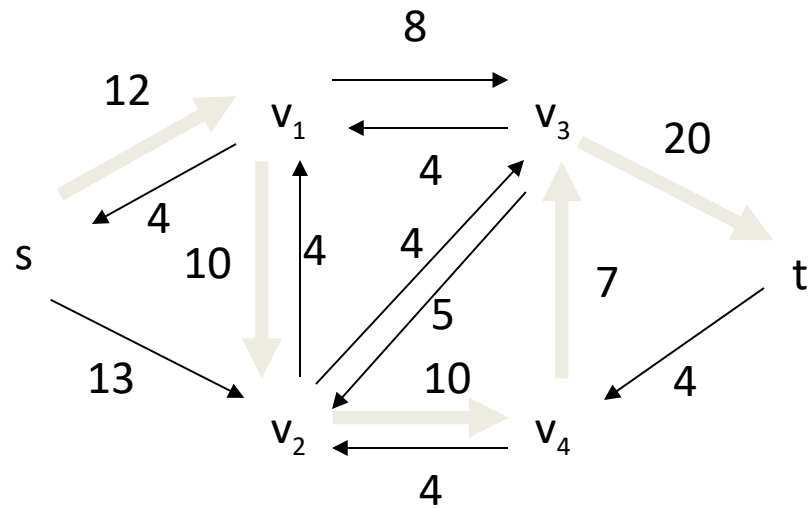
Example:



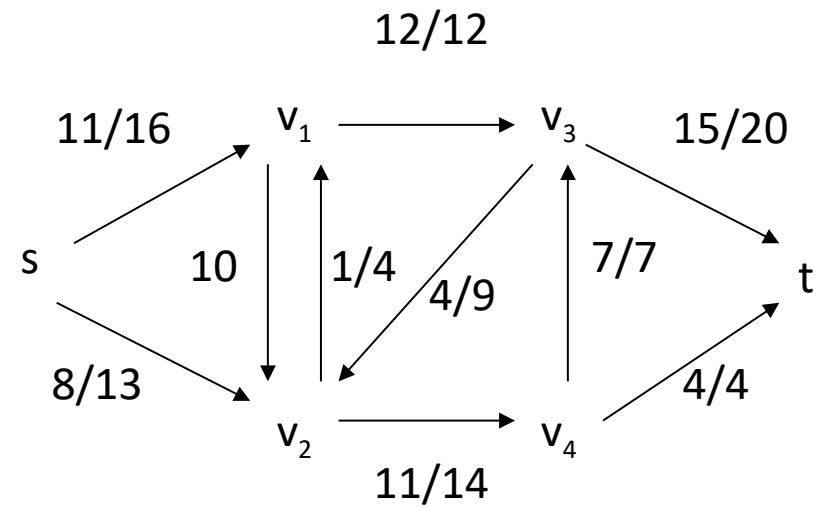
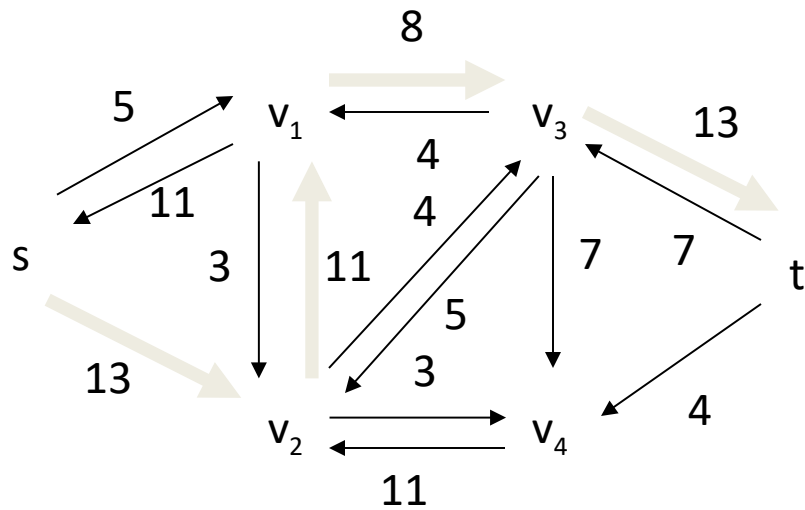
- The execution of the basic Ford-Fulkerson algorithm.
- (a)-(d) Successive iterations of the **while** loop: The left side of each part shows the residual network G_f from line 4 with a shaded augmenting path p . The right side of each part shows the new flow f that results from adding f_p to f . The residual network in (a) is the input network G . (e) The residual network at the last **while** loop test. It has no augmenting paths, and the flow f shown in (d) is therefore a maximum flow.



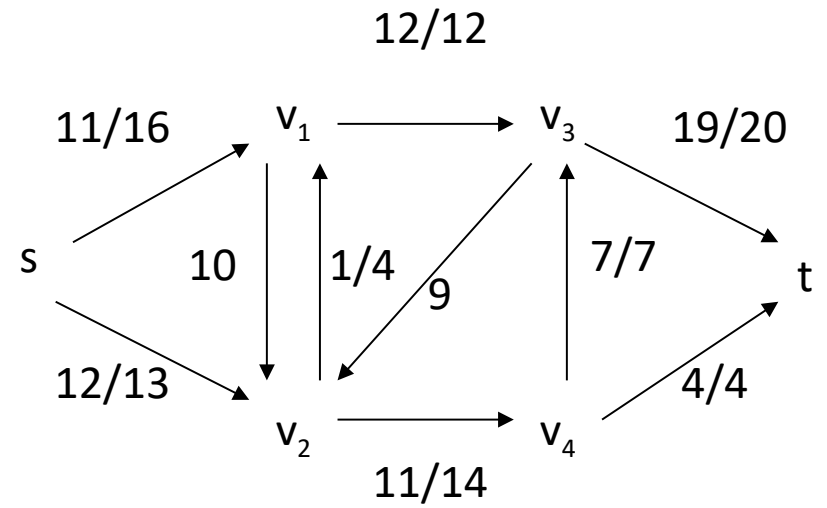
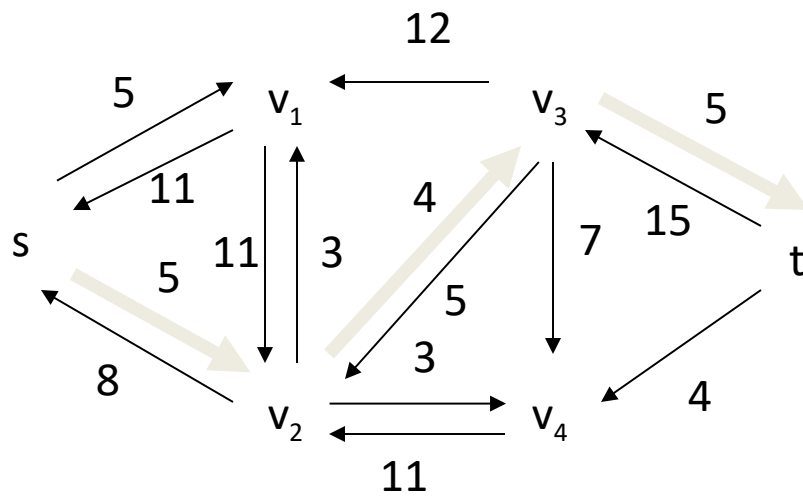
(a)



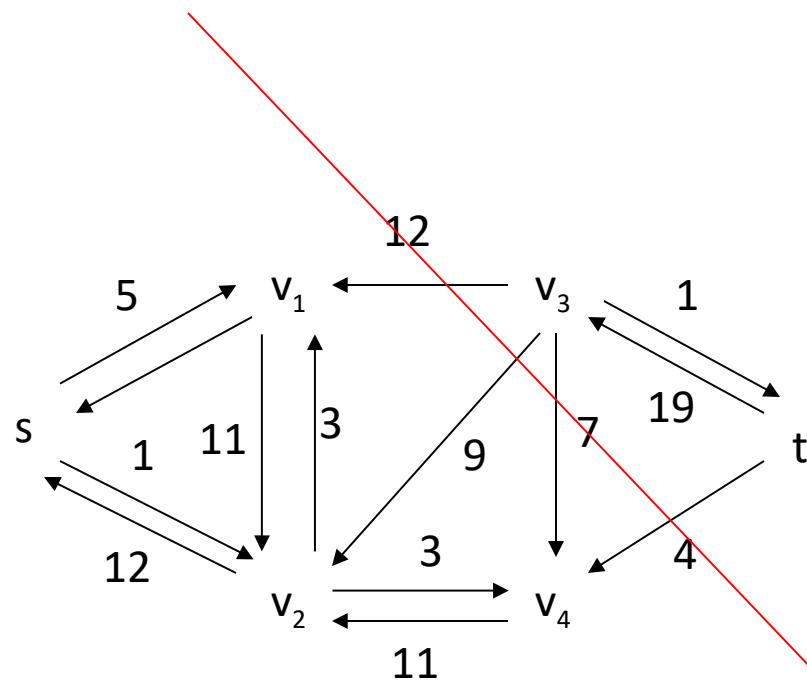
(b)



(c)



(d)



(e)

Time complexity:



- If each $c(e)$ is an *integer*, then time complexity is $O(|E|f^*)$, where f^* is the maximum flow.
- Reason: each time the flow is increased by at least one.
- This might not be a polynomial time algorithm since f^* can be represented by $\log(f^*)$ bits. So, the input size might be $\log(f^*)$.

The Edmonds-Karp algorithm



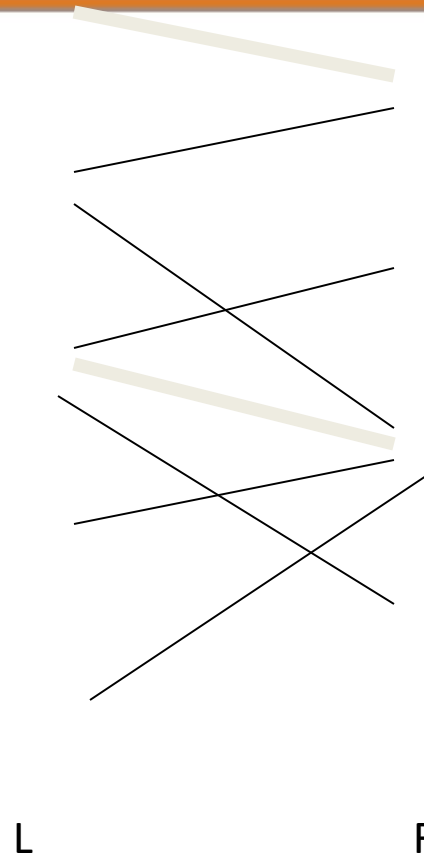
- Find the augmenting path using breadth-first search.
- Breadth-first search gives the shortest path for graphs (Assuming the length of each edge is 1.)
- Time complexity of Edmonds-Karp algorithm is $O(VE^2)$.
- The proof is very hard and is not required here.

Maximum bipartite matching:

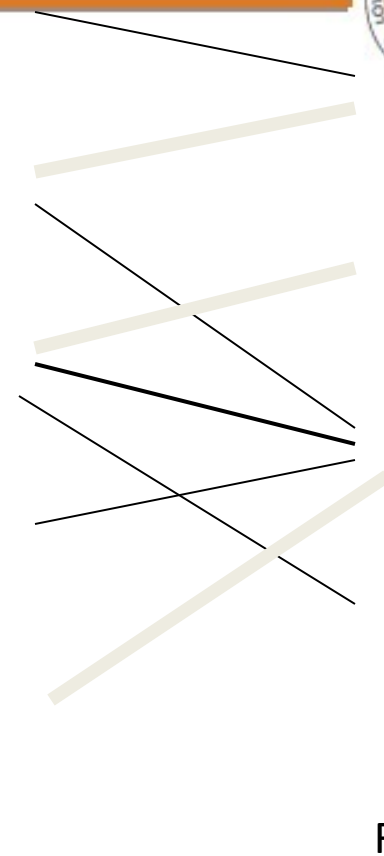


- Bipartite graph: a graph (V, E) , where $V=L \cup R$, $L \cap R = \text{empty}$, and for every $(u, v) \in E$, $u \in L$ and $v \in R$.
- Given an undirected graph $G=(V,E)$, a **matching** is a subset of edges $M \subseteq E$ such that for all vertices $v \in V$, at most one edge of M is incident on v . We say that a vertex $v \in V$ is **matched** by matching M if some edge in M is incident on v ; otherwise, v is **unmatched**. A **maximum matching** is a matching of maximum cardinality, that is, a matching M such that for any matching M' , we have

$$|M| \geq |M'|$$



(a)



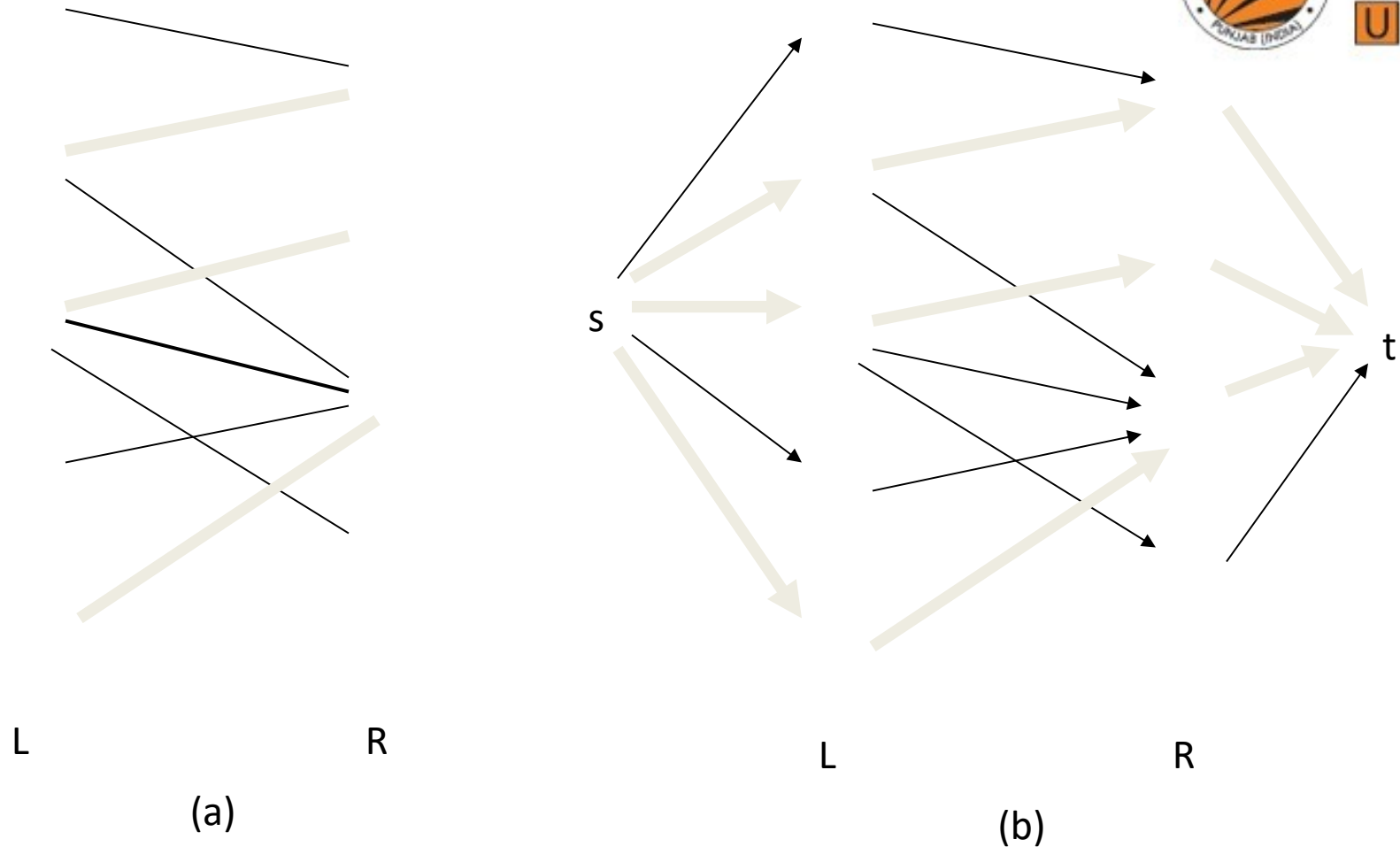
(b)

A bipartite graph $G=(V,E)$ with vertex partition $V=L \cup R$. (a) A matching with cardinality 2. (b) A maximum matching with cardinality 3.

Finding a maximum bipartite matching:



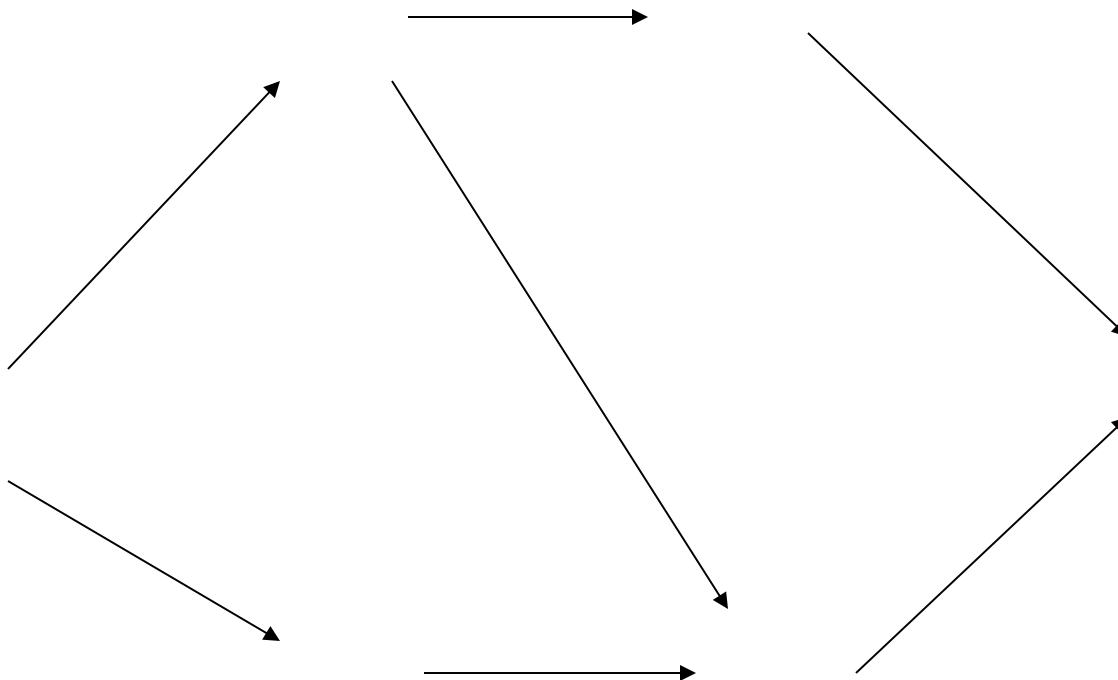
- We define the **corresponding flow network** $G'=(V',E')$ for the bipartite graph G as follows. Let the source s and sink t be new vertices not in V , and let $V'=V\cup\{s,t\}$. If the vertex partition of G is $V=L\cup R$, the directed edges of G' are given by $E'=\{(s,u):u\in L\}\cup\{(u,v):u\in L,v\in R,\text{ and } (u,v)\in E\}\cup\{(v,t):v\in R\}$. Finally, we assign unit capacity to each edge in E' .
- We will show that a matching in G corresponds directly to a flow in G' 's corresponding flow network G' . We say that a flow f on a flow network $G=(V,E)$ is **integer-valued** if $f(u,v)$ is an integer for all $(u,v)\in V^*V$.

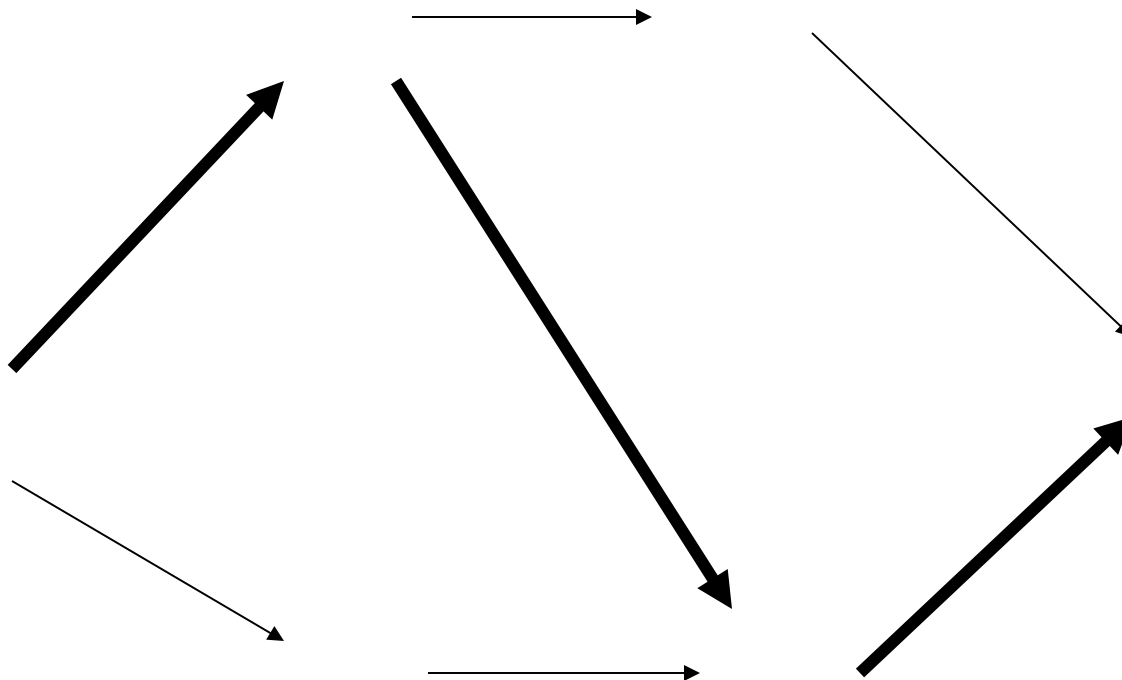


(a) The bipartite graph $G=(V,E)$ with vertex partition $V=L\cup R$. A maximum matching is shown by shaded edges. (b) The corresponding flow network. Each edge has unit capacity. Shaded edges have a flow of 1, and all other edges carry no flow.

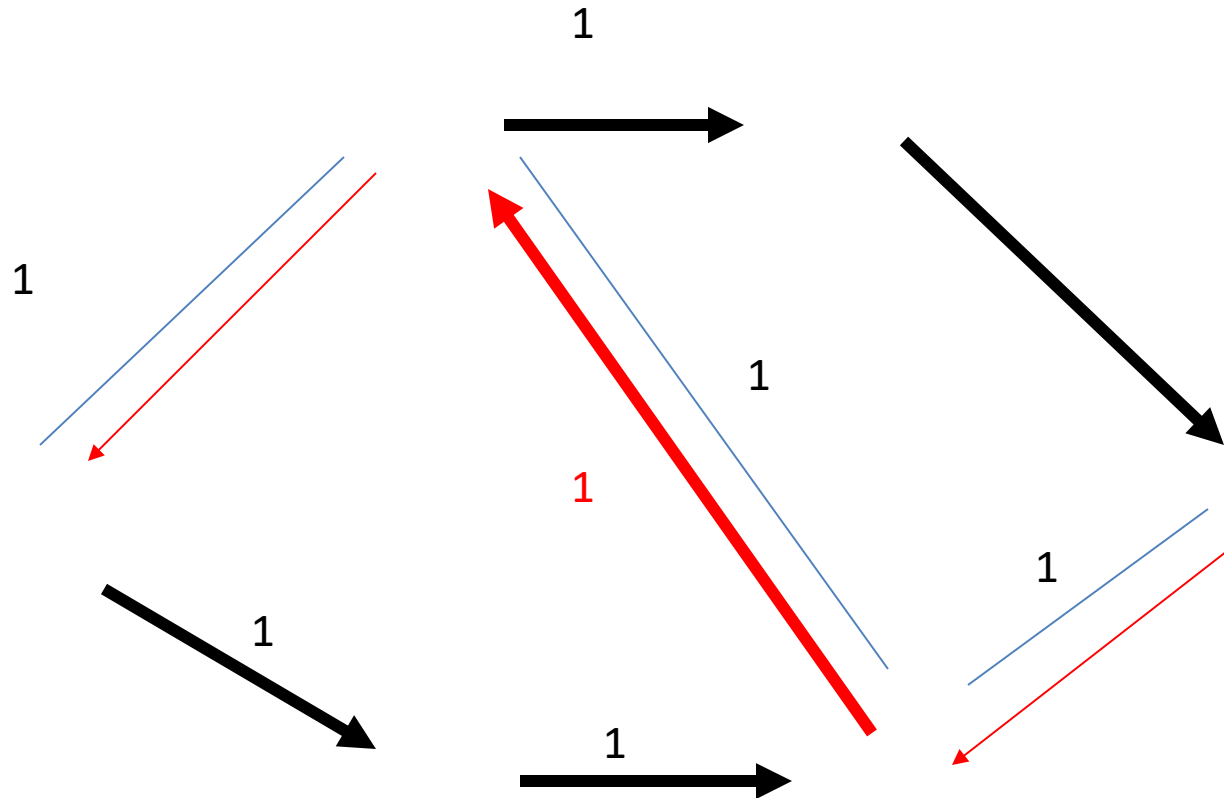
- Lemma .
- Let $G=(V,E)$ be a bipartite graph with vertex partition $V=L\cup R$, and let $G'=(V',E')$ be its corresponding flow network. If M is a matching in G , then there is an integer-valued flow f in G' with value $|M|$. Conversely, if f is an integer-valued flow in G' , then there is a matching M in G with cardinality $|M|=|f|$.
- Reason: The edges incident to s and t ensures this.
 - Each node in the first column has in-degree 1
 - Each node in the second column has out-degree 1.
 - So each node in the bipartite graph can be involved once in the flow.

Example:





Aug. path:



Residual network. Red edges are new edges in the residual network. The new aug. path is bold. Green edges are old aug. path. old flow=1.



Thank You !!!