

Design and Analysis of Algorithms (X)
Simple Unit-Capacity Networks

Guoqiang Li
School of Software


## Bipartite Matching

## Definition

Given an undirected graph $G=(V, E)$, subset of edges $M \subseteq E$ is a matching if each node appears in at most one edge in $M$.

## Definition

Given an undirected graph $G=(V, E)$, subset of edges $M \subseteq E$ is a matching if each node appears in at most one edge in $M$.

Maximum matching. Given a graph $G$, find a max-cardinality matching.


## Bipartite Matching

## Definition

A graph $G$ is bipartite if the nodes can be partitioned into two subsets $L$ and $R$ such that every edge connects a node in $L$ with a node in $R$.

## Bipartite Matching

## Definition

A graph $G$ is bipartite if the nodes can be partitioned into two subsets $L$ and $R$ such that every edge connects a node in $L$ with a node in $R$.

Bipartite matching. Given a bipartite graph $G=(L \cup R, E)$, find a max-cardinality matching.


## Max－Flow Formulation

## Formulation．

－Create digraph $G^{\prime}=\left(L \cup R \cup\{s, t\}, E^{\prime}\right)$ ．
－Direct all edges from $L$ to $R$ ，and assign infinite（or unit）capacity．
－Add unit－capacity edges from $s$ to each node in $L$ ．
－Add unit－capacity edges from each node in $R$ to $t$ ．


## Proof of Correctness

## Theorem

1－1 correspondence between matchings of cardinality $k$ in $G$ and integral flows of value $k$ in $G^{\prime}$ ．

## Proof of Correctness

## Theorem

1-1 correspondence between matchings of cardinality $k$ in $G$ and integral flows of value $k$ in $G^{\prime}$.

Proof.

## Proof of Correctness

## Theorem

1-1 correspondence between matchings of cardinality $k$ in $G$ and integral flows of value $k$ in $G^{\prime}$.

Proof.

- Let $M$ be a matching in $G$ of cardinality $k$.
- Consider flow $f$ that sends 1 unit on each of the $k$ corresponding paths.
- $f$ is a flow of value $k$.


## Proof of Correctness

## Theorem

1-1 correspondence between matchings of cardinality $k$ in $G$ and integral flows of value $k$ in $G^{\prime}$.

Proof.

## Proof of Correctness

## Theorem

1-1 correspondence between matchings of cardinality $k$ in $G$ and integral flows of value $k$ in $G^{\prime}$.

Proof.

- Let $f$ be an integral flow in $G^{\prime}$ of value $k$.
- Consider $M=$ set of edges from $L$ to $R$ with $f(e)=1$.


## Proof of Correctness

## Theorem

1-1 correspondence between matchings of cardinality $k$ in $G$ and integral flows of value $k$ in $G^{\prime}$.

Proof. $\Leftarrow$

- Let $f$ be an integral flow in $G^{\prime}$ of value $k$.
- Consider $M=$ set of edges from $L$ to $R$ with $f(e)=1$.
- each node in $L$ and $R$ participates in at most one edge in $M$.
- $|M|=k$ : apply flow-value lemma to cut ( $L \cup\{s\}, R \cup\{t\}$ ).


## Proof of Correctness

## Theorem

1-1 correspondence between matchings of cardinality $k$ in $G$ and integral flows of value $k$ in $G^{\prime}$.

## Proof of Correctness

## Theorem

1-1 correspondence between matchings of cardinality $k$ in $G$ and integral flows of value $k$ in $G^{\prime}$.

## Corollary

Can solve bipartite matching problem via max-flow formulation.

## Proof of Correctness

## Theorem

1-1 correspondence between matchings of cardinality $k$ in $G$ and integral flows of value $k$ in $G^{\prime}$.

## Corollary

Can solve bipartite matching problem via max-flow formulation.

Proof.

## Proof of Correctness

## Theorem

1-1 correspondence between matchings of cardinality $k$ in $G$ and integral flows of value $k$ in $G^{\prime}$.

## Corollary

Can solve bipartite matching problem via max-flow formulation.

## Proof.

- Integrality theorem $\Rightarrow$ there exists a max flow $f^{*}$ in $G^{\prime}$ that is integral.
- 1-1 correspondence $\Rightarrow f^{*}$ corresponds to max-cardinality matching.


## Quiz 1

What is running time of Ford－Fulkerson algorithms to find a max－cardinality matching in a bipartite graph？

A．$O(|E|+|V|)$
B．$O(|E||V|)$
C．$O\left(|E \| V|^{2}\right)$
D．$O\left(|E|^{2}|V|\right)$

## Quiz 2

Which max－flow algorithm to use for bipartite matching？

A．Ford－Fulkerson：$O(|E| \cdot|V| \cdot C)$ ．
B．Capacity scaling：$O\left(|E|^{2} \cdot \log C\right)$ ．
C．Shortest augmenting path：$O\left(|E|^{2}|V|\right)$ ．
D．Dinitz＇algorithm：$O\left(|E \| V|^{2}\right)$ ．

## Perfect Matchings in Bigraphs

## Definition

Given a graph $G=(V, E)$, a subset of edges $M \subseteq E$ is a perfect matching if each node appears in exactly one edge in $M$.

## Perfect Matchings in Bigraphs

## Definition

Given a graph $G=(V, E)$, a subset of edges $M \subseteq E$ is a perfect matching if each node appears in exactly one edge in $M$.
Q. When does a bipartite graph have a perfect matching?

## Perfect Matchings in Bigraphs

## Definition

Given a graph $G=(V, E)$, a subset of edges $M \subseteq E$ is a perfect matching if each node appears in exactly one edge in $M$.
Q. When does a bipartite graph have a perfect matching?

Structure of bipartite graphs with perfect matchings.

- Clearly, we must have $|L|=|R|$.
- Which other conditions are necessary?
- Which other conditions are sufficient?


## Perfect Matchings in Bigraphs

## Notation.

Let $S$ be a subset of nodes, and let $N(S)$ be the set of nodes adjacent to nodes in $S$.

## Perfect Matchings in Bigraphs

## Notation.

Let $S$ be a subset of nodes, and let $N(S)$ be the set of nodes adjacent to nodes in $S$.
Observation. If a bipartite graph $G=(L \cup R, E)$ has a perfect matching, then $|N(S)| \geq|S|$ for all subsets $S \subseteq L$.

## Perfect Matchings in Bigraphs

## Notation.

Let $S$ be a subset of nodes, and let $N(S)$ be the set of nodes adjacent to nodes in $S$.
Observation. If a bipartite graph $G=(L \cup R, E)$ has a perfect matching, then $|N(S)| \geq|S|$ for all subsets $S \subseteq L$.

Proof. Each node in $S$ has to be matched to a different node in $N(S)$.


## Hall's Marriage Theorem

## Theorem (Frobenius 1917, Hall 1935)

Let $G=(L \cup R, E)$ be a bipartite graph with $|L|=|R|$. Then, graph $G$ has a perfect matching iff $|N(S)| \geq|S|$ for all subsets $S \subseteq L$.

## Hall's Marriage Theorem

## Theorem (Frobenius 1917, Hall 1935)

Let $G=(L \cup R, E)$ be a bipartite graph with $|L|=|R|$. Then, graph $G$ has a perfect matching iff $|N(S)| \geq|S|$ for all subsets $S \subseteq L$.

Proof.
$\Rightarrow$

## Hall＇s Marriage Theorem

## Theorem（Frobenius 1917，Hall 1935）

Let $G=(L \cup R, E)$ be a bipartite graph with $|L|=|R|$ ．Then，graph $G$ has a perfect matching iff $|N(S)| \geq|S|$ for all subsets $S \subseteq L$ ．

Proof．
This was the previous observation．


## Hall's Marriage Theorem

Proof. $\Leftarrow$

## Hall's Marriage Theorem

## Proof.

Suppose $G$ does not have a perfect matching.

## Hall's Marriage Theorem

## Proof.

Suppose $G$ does not have a perfect matching.
Formulate as a max-flow problem and let $(A, B)$ be a min cut in $G^{\prime}$.

## Hall's Marriage Theorem

## Proof.

Suppose $G$ does not have a perfect matching.
Formulate as a max-flow problem and let $(A, B)$ be a min cut in $G^{\prime}$.
By max-flow min-cut theorem, $\operatorname{cap}(A, B)<|L|$.

## Hall's Marriage Theorem

## Proof.

Suppose $G$ does not have a perfect matching.
Formulate as a max-flow problem and let $(A, B)$ be a min cut in $G^{\prime}$.
By max-flow min-cut theorem, $\operatorname{cap}(A, B)<|L|$.
Define $L_{A}=L \cap A, L_{B}=L \cap B, R_{A}=R \cap A$.

## Hall's Marriage Theorem

## Proof. $\Leftarrow$

Suppose $G$ does not have a perfect matching.
Formulate as a max-flow problem and let $(A, B)$ be a min cut in $G^{\prime}$.
By max-flow min-cut theorem, $\operatorname{cap}(A, B)<|L|$.
Define $L_{A}=L \cap A, L_{B}=L \cap B, R_{A}=R \cap A$.
$\operatorname{cap}(A, B)=\left|L_{B}\right|+\left|R_{A}\right| \Rightarrow\left|R_{A}\right|<\left|L_{A}\right|$
Min cut can't use $\infty$ edges $\Rightarrow N\left(L_{A}\right) \subseteq R_{A}$.
$\left|N\left(L_{A}\right)\right| \leq\left|R_{A}\right|<\left|L_{A}\right|$.
Choose $S=L_{A}$.

## Bipartite Matching

Problem. Given a bipartite graph, find a max-cardinality matching.

| year | worst case | technique | discovered by |
| :---: | :---: | :---: | :---: |
| 1955 | $O(\|E\|\|V\|)$ | augmenting path | Ford-Fulkerson |
| 1973 | $O\left(\|E\|\|V\|^{1 / 2}\right)$ | blocking flow | Hopcroft-Karp, Karzanov |
| 2004 | $O\left(\|V\|^{2.378}\right)$ | fast matrix multiplication | Mucha-Sankowsi |
| 2013 | $\tilde{O}\left(\|E\|^{10 / 7}\right)$ | electrical flow | Madry |
| $20 x x$ | $? ? ?$ |  |  |

## Quiz 3

Which of the following are properties of the graph $G=(V, E)$ ?
A. $G$ has a perfect matching.
B. Hall's condition is satisfied: $|N(S)| \geq|S|$ for all subsets $S \subseteq V$.
C. Both A and B.
D. Neither A nor B.


## Nonbipartite Matching

Problem. Given an undirected graph, find a max-cardinality matching.

## Nonbipartite Matching

Problem. Given an undirected graph, find a max-cardinality matching.

- Structure of nonbipartite graphs is more complicated.
- But well understood. [Tutte-Berge formula, Edmonds-Gallai]
- Blossom algorithm: $O\left(n^{4}\right)$. [Edmonds 1965]
- Best known: $O\left(m n^{1 / 2}\right)$. [Micali-Vazirani 1980, Vazirani 1994]


## Hackathon Problem

Hackathon problem.

- Hackathon attended by $n$ Harvard students and $n$ Princeton students.
- Each Harvard student is friends with exactly $k>0$ Princeton students; each Princeton student is friends with exactly $k$ Harvard students.
- Is it possible to arrange the hackathon so that each Princeton student pair programs with a different friend from Harvard?


## Hackathon Problem

Mathematical reformulation. Does every $k$-regular bipartite graph have a perfect matching?
Example. Boolean hypercube.


Hackathon Problem

## Theorem

Every $k$-regular bipartite graph $G$ has a perfect matching.

## Hackathon Problem

## Theorem

Every $k$-regular bipartite graph $G$ has a perfect matching.

Proved by Hall's Marriage Theorem, DIY!

## Hackathon Problem: Another Proof

## Proof.

- Size of max matching = value of max flow in $G^{\prime}$.
- It is easy to construct the following flow

$$
f(u, v)= \begin{cases}1 & \text { if } u=s \text { or } v=t \\ 1 / k & \text { otherwise }\end{cases}
$$



- The value of flow $f$ is $n \Rightarrow G^{\prime}$ has a perfect matching.

Hall's Theorem by Max-Flow

## Simple Unit-Capacity Networks

## Simple Unit-Capacity Networks

## Definition

A flow network is a simple unit-capacity network if:

- Every edge has capacity 1.
- Every node (other than $s$ or $t$ ) has exactly one entering edge, or exactly one leaving edge, or both.


## Simple Unit-Capacity Networks

Property. Let $G$ be a simple unit-capacity network and let $f$ be a $0-1$ flow. Then, residual network $G_{f}$ is also a simple unit-capacity network.

## Simple Unit-Capacity Networks

Property. Let $G$ be a simple unit-capacity network and let $f$ be a $0-1$ flow. Then, residual network $G_{f}$ is also a simple unit-capacity network.

Example. Bipartite matching.

## Simple Unit-Capacity Networks

Property. Let $G$ be a simple unit-capacity network and let $f$ be a $0-1$ flow. Then, residual network $G_{f}$ is also a simple unit-capacity network.

Example. Bipartite matching.


## Simple Unit-Capacity Networks

## Shortest-augmenting-path algorithm.

- Normal augmentation: length of shortest path does not change.
- Special augmentation: length of shortest path strictly increases.


## Theorem (Even-Tarjan 1975)

In simple unit-capacity networks, Dinitz'algorithm computes a maximum flow in $O\left(|E \| V|^{1 / 2}\right)$ time.

## Simple Unit-Capacity Networks

## Shortest-augmenting-path algorithm.

- Normal augmentation: length of shortest path does not change.
- Special augmentation: length of shortest path strictly increases.


## Theorem (Even-Tarjan 1975)

In simple unit-capacity networks, Dinitz'algorithm computes a maximum flow in $O\left(|E \| V|^{1 / 2}\right)$ time.

Proof.

## Simple Unit-Capacity Networks

## Shortest-augmenting-path algorithm.

- Normal augmentation: length of shortest path does not change.
- Special augmentation: length of shortest path strictly increases.


## Theorem (Even-Tarjan 1975)

In simple unit-capacity networks, Dinitz'algorithm computes a maximum flow in $O\left(|E \| V|^{1 / 2}\right)$ time.

Proof.

- Lemma 1. Each phase of normal augmentations takes $O(|E|)$ time.
- Lemma 2. After $|V|^{1 / 2}$ phases, $\operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$.
- Lemma 3. After $\leq|V|^{1 / 2}$ additional augmentations, flow is optimal.


## Simple Unit－Capacity Networks

## Lemma 3

After $\leq|V|^{1 / 2}$ additional augmentations，flow is optimal．

## Simple Unit－Capacity Networks

## Lemma 3

After $\leq|V|^{1 / 2}$ additional augmentations，flow is optimal．

Proof．Each augmentation increases flow value by at least 1.

## Simple Unit－Capacity Networks

Phase of normal augmentations．
－Construct level graph $L_{G}$ ．
－Start at $s$ ，advance along an edge in $L_{G}$ until reach $t$ or get stuck．
－If reach $t$ ，augment flow；update $L_{G}$ ；and restart from $s$ ．
－If get stuck，delete node from $L_{G}$ and go to previous node．
construct level graph


## Simple Unit-Capacity Networks

Phase of normal augmentations.

- Construct level graph $L_{G}$.
- Start at $s$, advance along an edge in $L_{G}$ until reach $t$ or get stuck.
- If reach $t$, augment flow; update $L_{G}$; and restart from $s$.
- If get stuck, delete node from $L_{G}$ and go to previous node.
advance



## Simple Unit-Capacity Networks

Phase of normal augmentations.

- Construct level graph $L_{G}$.
- Start at $s$, advance along an edge in $L_{G}$ until reach $t$ or get stuck.
- If reach $t$, augment flow; update $L_{G}$; and restart from $s$.
- If get stuck, delete node from $L_{G}$ and go to previous node.



## Simple Unit-Capacity Networks

Phase of normal augmentations.

- Construct level graph $L_{G}$.
- Start at $s$, advance along an edge in $L_{G}$ until reach $t$ or get stuck.
- If reach $t$, augment flow; update $L_{G}$; and restart from $s$.
- If get stuck, delete node from $L_{G}$ and go to previous node.
advance



## Simple Unit－Capacity Networks

Phase of normal augmentations．
－Construct level graph $L_{G}$ ．
－Start at $s$ ，advance along an edge in $L_{G}$ until reach $t$ or get stuck．
－If reach $t$ ，augment flow；update $L_{G}$ ；and restart from $s$ ．
－If get stuck，delete node from $L_{G}$ and go to previous node．

## retreat



## Simple Unit-Capacity Networks

Phase of normal augmentations.

- Construct level graph $L_{G}$.
- Start at $s$, advance along an edge in $L_{G}$ until reach $t$ or get stuck.
- If reach $t$, augment flow; update $L_{G}$; and restart from $s$.
- If get stuck, delete node from $L_{G}$ and go to previous node.
advance



## Simple Unit-Capacity Networks

Phase of normal augmentations.

- Construct level graph $L_{G}$.
- Start at $s$, advance along an edge in $L_{G}$ until reach $t$ or get stuck.
- If reach $t$, augment flow; update $L_{G}$; and restart from $s$.
- If get stuck, delete node from $L_{G}$ and go to previous node.
augment



## Simple Unit－Capacity Networks

## Phase of normal augmentations．

－Construct level graph $L_{G}$ ．
－Start at $s$ ，advance along an edge in $L_{G}$ until reach $t$ or get stuck．
－If reach $t$ ，augment flow；update $L_{G}$ ；and restart from $s$ ．
－If get stuck，delete node from $L_{G}$ and go to previous node．


## Simple Unit-Capacity Networks

Phase of normal augmentations.

- Construct level graph $L_{G}$.
- Start at $s$, advance along an edge in $L_{G}$ until reach $t$ or get stuck.
- If reach $t$, augment flow; update $L_{G}$; and restart from $s$.
- If get stuck, delete node from $L_{G}$ and go to previous node.


## Simple Unit-Capacity Networks

Phase of normal augmentations.

- Construct level graph $L_{G}$.
- Start at $s$, advance along an edge in $L_{G}$ until reach $t$ or get stuck.
- If reach $t$, augment flow; update $L_{G}$; and restart from $s$.
- If get stuck, delete node from $L_{G}$ and go to previous node.


## Lemma 1

A phase of normal augmentations takes $O(|E|)$ time.

## Simple Unit-Capacity Networks

Phase of normal augmentations.

- Construct level graph $L_{G}$.
- Start at $s$, advance along an edge in $L_{G}$ until reach $t$ or get stuck.
- If reach $t$, augment flow; update $L_{G}$; and restart from $s$.
- If get stuck, delete node from $L_{G}$ and go to previous node.


## Lemma 1

A phase of normal augmentations takes $O(|E|)$ time.

Proof.

- $O(|E|)$ to create level graph $L_{G}$.
- $O(1)$ per edge (each edge involved in at most one advance, retreat, and augmentation).
- $O(1)$ per node (each node deleted at most once)


## Computational Geometry

## Lemma 2

After $|V|^{1 / 2}$ phases, $\operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$.

## Computational Geometry

## Lemma 2

After $|V|^{1 / 2}$ phases, $\operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$.

Proof.

## Computational Geometry

## Lemma 2

After $|V|^{1 / 2}$ phases, $\operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$.

Proof.
level graph Lc for flow $\mathbf{f}$


## Computational Geometry

## Lemma 2

After $|V|^{1 / 2}$ phases， $\operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$ ．

Proof．
After $|V|^{1 / 2}$ phases，length of shortest augmenting path is $>|V|^{1 / 2}$ ．

## Computational Geometry

## Lemma 2

After $|V|^{1 / 2}$ phases, $\operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$.

Proof.
After $|V|^{1 / 2}$ phases, length of shortest augmenting path is $>|V|^{1 / 2}$.
Thus, level graph has $\geq|V|^{1 / 2}$ levels (not including levels for $s$ or $t$ )

## Computational Geometry

## Lemma 2

After $|V|^{1 / 2}$ phases, $\operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$.

## Proof.

After $|V|^{1 / 2}$ phases, length of shortest augmenting path is $>|V|^{1 / 2}$.
Thus, level graph has $\geq|V|^{1 / 2}$ levels (not including levels for $s$ or $t$ )
Let $1 \leq h \leq|V|^{1 / 2}$ be a level with min number of nodes $\Rightarrow\left|V_{h}\right| \leq|V|^{1 / 2}$.

## Computational Geometry

## Lemma 2

After $|V|^{1 / 2}$ phases, $\operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$.

## Proof.

After $|V|^{1 / 2}$ phases, length of shortest augmenting path is $>|V|^{1 / 2}$.
Thus, level graph has $\geq|V|^{1 / 2}$ levels (not including levels for $s$ or $t$ )
Let $1 \leq h \leq|V|^{1 / 2}$ be a level with min number of nodes $\Rightarrow\left|V_{h}\right| \leq|V|^{1 / 2}$.
Let $A=\{v: \ell(v)<h\} \cup\{v: \ell(v)=h$ and $v$ has $\leq 1$ outgoing residual edge $\}$.

## Computational Geometry

## Lemma 2

After $|V|^{1 / 2}$ phases, $\operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$.

## Proof.

After $|V|^{1 / 2}$ phases, length of shortest augmenting path is $>|V|^{1 / 2}$.

Thus, level graph has $\geq|V|^{1 / 2}$ levels (not including levels for $s$ or $t$ )

Let $1 \leq h \leq|V|^{1 / 2}$ be a level with min number of nodes $\Rightarrow\left|V_{h}\right| \leq|V|^{1 / 2}$.

Let $A=\{v: \ell(v)<h\} \cup\{v: \ell(v)=h$ and $v$ has $\leq 1$ outgoing residual edge $\}$.
$\operatorname{cap}_{f}(A, B) \leq\left|V_{h}\right| \leq|V|^{1 / 2} \Rightarrow \operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$

## Computational Geometry



## Theorem (Even-Tarjan 1975)

In simple unit-capacity networks, Dinitz' algorithm computes a maximum flow in $O\left(|E||V|^{1 / 2}\right)$ time.

Proof.

- Lemma 1. Each phase take $O(|E|)$ time.
- Lemma 2. After $|V|^{1 / 2}$ phase, $\operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$
- Lemma 3. After $\leq|V|^{1 / 2}$ additional augmentations.


## Theorem (Even-Tarjan 1975)

In simple unit-capacity networks, Dinitz' algorithm computes a maximum flow in $O\left(|E||V|^{1 / 2}\right)$ time.

Proof.

- Lemma 1. Each phase take $O(|E|)$ time.
- Lemma 2. After $|V|^{1 / 2}$ phase, $\operatorname{val}(f) \geq \operatorname{val}\left(f^{*}\right)-|V|^{1 / 2}$
- Lemma 3. After $\leq|V|^{1 / 2}$ additional augmentations.


## Corollary

Dinitz' algorithm computes maximum-cardinality bipartite matching in $O\left(|E||V|^{1 / 2}\right)$ time.

## Disjoint Paths

## Edge-Disjoint Paths

## Definition

Two paths are edge-disjoint if they have no edge in common.

## Edge-Disjoint Paths

## Definition

Two paths are edge-disjoint if they have no edge in common.

Edge-disjoint paths problem. Given a digraph $G=(V, E)$ and two nodes $s$ and $t$, find the max number of edge-disjoint $s \rightsquigarrow t$ paths.

## Edge-Disjoint Paths

## Definition

Two paths are edge-disjoint if they have no edge in common.

Edge-disjoint paths problem. Given a digraph $G=(V, E)$ and two nodes $s$ and $t$, find the max number of edge-disjoint $s \rightsquigarrow t$ paths.

Max-flow formulation. Assign unit capacity to every edge.

## Edge－Disjoint Paths

## Theorem

1－1 correspondence between $k$ edge－disjoint $s \rightsquigarrow t$ paths in $G$ and integral flows of value $k$ in $G^{\prime}$ ．

## Edge-Disjoint Paths

## Theorem

1-1 correspondence between $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$ and integral flows of value $k$ in $G^{\prime}$.

Proof.

## Edge-Disjoint Paths

## Theorem

1-1 correspondence between $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$ and integral flows of value $k$ in $G^{\prime}$.

Proof.

- Let $P_{1}, \ldots, P_{k}$ be $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$.


## Edge-Disjoint Paths

## Theorem

1-1 correspondence between $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$ and integral flows of value $k$ in $G^{\prime}$.

## Proof.

- Let $P_{1}, \ldots, P_{k}$ be $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$.
- Set $f(e)= \begin{cases}1 & \text { edge } e \text { participates in some path } P_{j} \\ 0 & \text { otherwise }\end{cases}$


## Edge-Disjoint Paths

## Theorem

1-1 correspondence between $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$ and integral flows of value $k$ in $G^{\prime}$.

## Proof.

- Let $P_{1}, \ldots, P_{k}$ be $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$.
- Set $f(e)= \begin{cases}1 & \text { edge } e \text { participates in some path } P_{j} \\ 0 & \text { otherwise }\end{cases}$
- Since paths are edge-disjoint, $f$ is a flow of value $k$.


## Edge-Disjoint Paths

## Theorem

1-1 correspondence between $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$ and integral flows of value $k$ in $G^{\prime}$.

Proof.

## Edge-Disjoint Paths

## Theorem

1-1 correspondence between $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$ and integral flows of value $k$ in $G^{\prime}$.

## Proof.

- Let $f$ be an integral flow in $G^{\prime}$ of value $k$.
- Consider edge $(s, u)$ with $f(s, u)=1$.
- by flow conservation, there exists an edge $(u, v)$ with $f(u, v)=1$.
- continue until reach $t$, always choosing a new edge
- Produces $k$ (not necessarily simple) edge-disjoint paths.


## Edge－Disjoint Paths

## Theorem

1－1 correspondence between $k$ edge－disjoint $s \rightsquigarrow t$ paths in $G$ and integral flows of value $k$ in $G^{\prime}$ ．

## Edge-Disjoint Paths

## Theorem

1-1 correspondence between $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$ and integral flows of value $k$ in $G^{\prime}$.

## Corollary

Can solve edge-disjoint paths problem via max-flow formulation.

## Edge-Disjoint Paths

## Theorem

1-1 correspondence between $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$ and integral flows of value $k$ in $G^{\prime}$.

## Corollary

Can solve edge-disjoint paths problem via max-flow formulation.

Proof.

## Edge-Disjoint Paths

## Theorem

1-1 correspondence between $k$ edge-disjoint $s \rightsquigarrow t$ paths in $G$ and integral flows of value $k$ in $G^{\prime}$.

## Corollary

Can solve edge-disjoint paths problem via max-flow formulation.

## Proof.

- Integrality theorem $\Rightarrow$ there exists a max flow $f^{*}$ in $G^{\prime}$ that is integral.
- 1-1 correspondence $\Rightarrow f^{*}$ corresponds to max number of edge-disjoint $s \rightsquigarrow t$ paths in $G$.


## Network Connectivity

## Definition

A set of edges $F \subseteq E$ disconnects $t$ from $s$ if every $s \rightsquigarrow t$ path uses at least one edge in $F$ ．

## Network Connectivity

## Definition

A set of edges $F \subseteq E$ disconnects $t$ from $s$ if every $s \rightsquigarrow t$ path uses at least one edge in $F$ ．

Network connectivity．Given a digraph $G=(V, E)$ and two nodes $s$ and $t$ ，find minimal number of edges whose removal disconnects $t$ from $s$ ．

## Menger's Theorem

## Theorem (Menger 1927)

The max number of edge-disjoint $s \rightsquigarrow t$ paths equals the min number of edges whose removal disconnects $t$ from $s$.

## Menger's Theorem

## Theorem (Menger 1927)

The max number of edge-disjoint $s \rightsquigarrow t$ paths equals the min number of edges whose removal disconnects $t$ from $s$.

Proof.

## Theorem (Menger 1927)

The max number of edge-disjoint $s \rightsquigarrow t$ paths equals the min number of edges whose removal disconnects $t$ from $s$.

## Proof.

- Suppose the removal of $F \subseteq E$ disconnects $t$ from $s$, and $|F|=k$.
- Every $s \rightsquigarrow t$ path uses at least one edge in $F$.
- Hence, the number of edge-disjoint paths is $\leq k$.


## Menger＇s Theorem

## Theorem（Menger 1927）

The max number of edge－disjoint $s \rightsquigarrow t$ paths equals the min number of edges whose removal disconnects $t$ from $s$ ．

Proof．

## Menger's Theorem

## Theorem (Menger 1927)

The max number of edge-disjoint $s \rightsquigarrow t$ paths equals the min number of edges whose removal disconnects $t$ from $s$.

## Proof.

- Suppose max number of edge-disjoint $s \rightsquigarrow t$ paths is $k$.
- Then value of max flow $=k$.
- Max-flow min-cut theorem $\Rightarrow$ there exists a cut $(A, B)$ of capacity $k$.
- Let $F$ be set of edges going from $A$ to $B$.
- $|F|=k$ and disconnects $t$ from $s$.

Referred Materials

## Referred Materials

- Content of this lecture comes from Section 7.5 in [KT05].

