

Edge-Disjoint Paths in Networks

(Part 1)

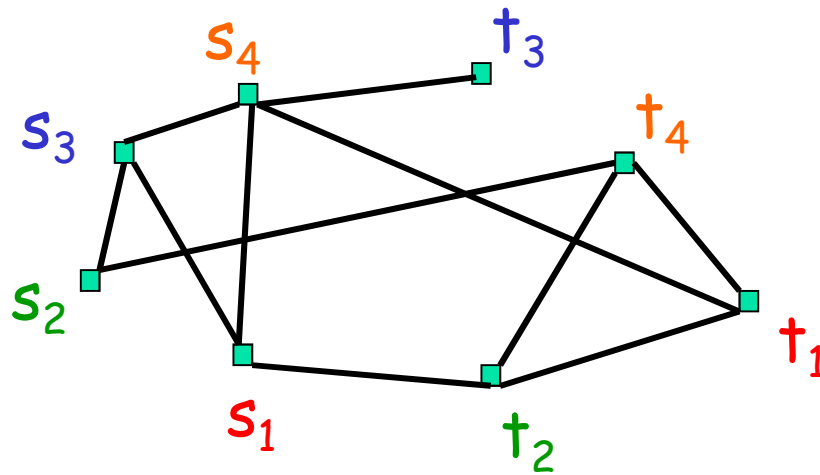
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Edge Disjoint Paths Problem (EDP)

Input: Graph $G(V,E)$, source-sink pairs $s_1t_1, s_2t_2, \dots, s_kt_k$

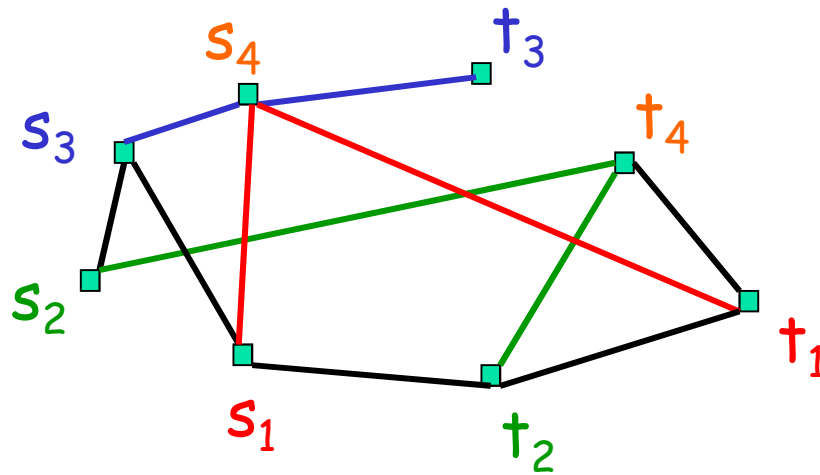
Goal: Route a maximum # of s_i-t_i pairs using edge-disjoint paths



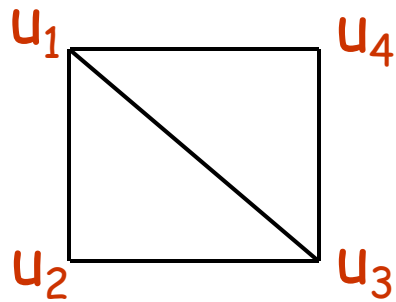
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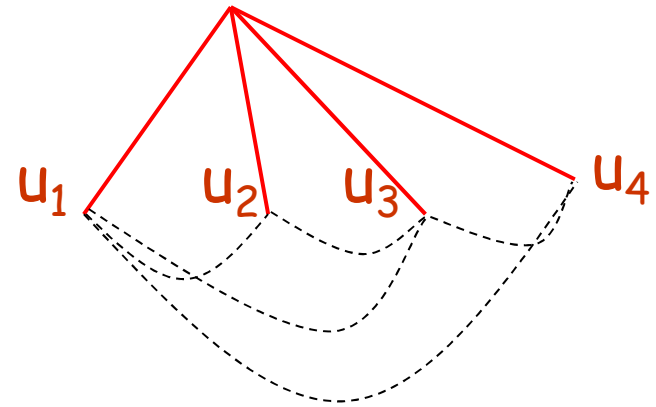
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EDP on Stars



Matching in G



EDP in a Star Graph

(Edges in G become source-sink pairs in the **Star Graph**)

Complexity of EDP

- EDP on star graphs is equivalent to the maximum matching problem in general graphs.
- Even for two pairs, EDP is NP-hard if G is a directed graph [Fortune, Hopcroft, Wylie'80].
- Polynomial-time solvable for constant number of pairs if G is undirected [Roberston, Seymour'88].
- NP-hard even on undirected trees when the edges have capacities.

Coping with Hardness

Settle for **sub-optimal solutions**: route only a fraction of the pairs that can be routed in an optimal solution.

Approximation algorithm A

- Runs in polynomial time
- **Approximation ratio**: how good is A
- Approx ratio α if $A(I) \geq OPT(I) / \alpha$ for all I
- Smaller the α , the closer we are to the optimal.

Overview of the Talk

- Review of classical results for EDP.
- Survey of the current state-of-the-art.
- Key algorithmic ideas underlying recent developments.
- Integrality gap and hardness results.
- Some open problems.

A Greedy Algorithm

- Among the **unrouted** pairs, pick the pair that has the **shortest path** in the current graph.
- **Route** this pair and **remove** all edges on the path from the graph.
- Repeat until no more pairs can be routed.

Clearly gives an **edge-disjoint** routing.

How good is this algorithm?

Analysis of the Greedy Algorithm

n : # of vertices m : # of edges

Fix an optimal solution, say, OPT .

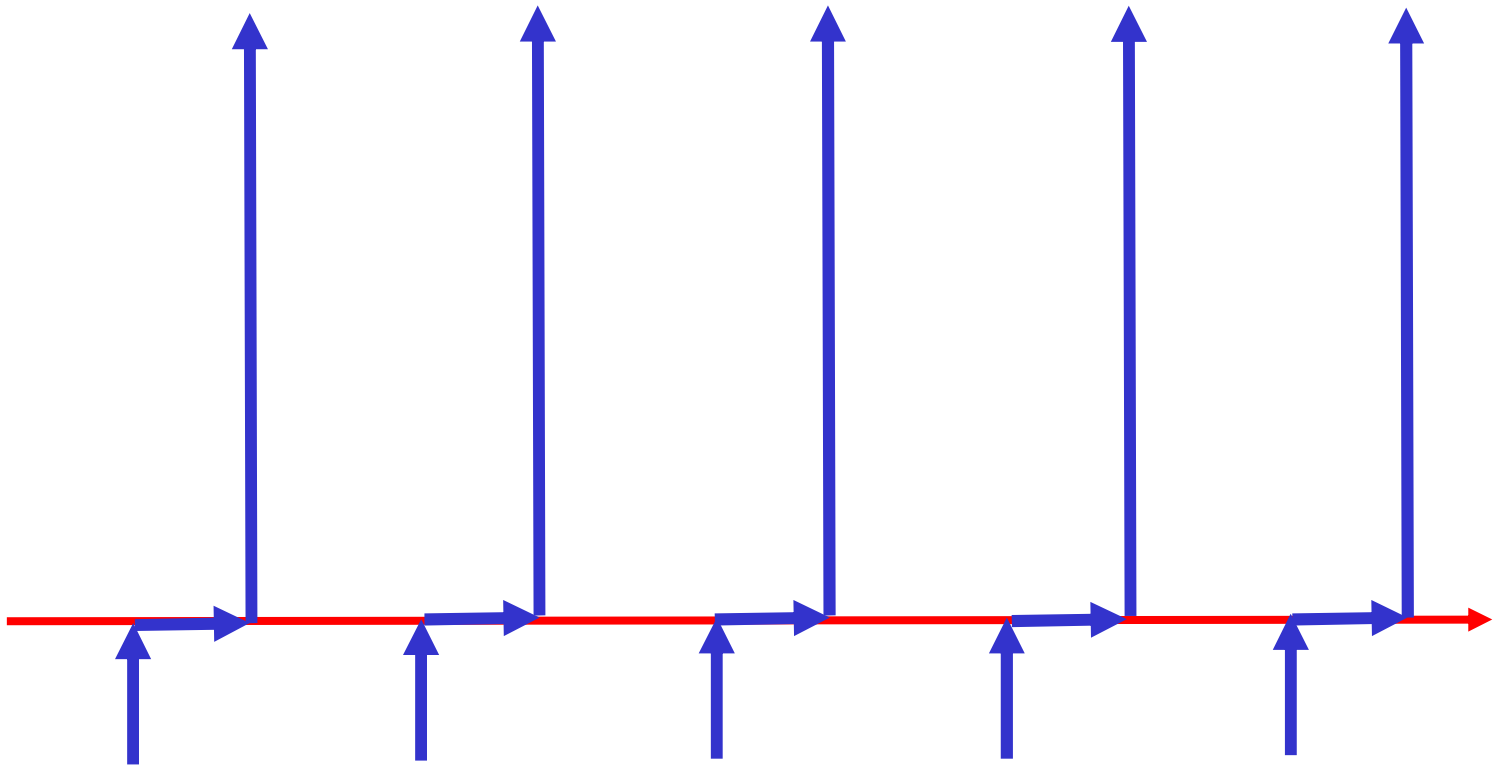
As long as the greedy path has at most $m^{1/2}$ edges, it can destroy at most $m^{1/2}$ paths in OPT .

Suppose at some point, a path chosen by greedy is longer than $m^{1/2}$. Since there are only m edges, OPT can choose at most $m^{1/2}$ paths from here on.

So greedy gives an $O(m^{1/2})$ -approximation.

Can we improve it ?

Greedy chooses the **red path** and none of the **blue pairs** can be routed as a result.



Multicommodity Flow Relaxation (LP)

- Routing is relaxed to be a **flow** from s_i to t_i .
- A pair can be routed for a **fractional** amount.

x_i : amount of s_i - t_i flow that is routed.

$f(p)$: amount of flow routed on a path p .

$$\text{Max } \sum_i x_i$$

s.t.

$$\forall i \quad x_i = \sum_{s_i-t_i \text{ paths } p} f(p)$$

$$\forall e \quad \sum_{p: e \in p} f(p) \leq 1.$$

$$0 \leq x_i \leq 1.$$

A Simple Rounding Algorithm

- Among the **unrouted** pairs
 - Pick a pair with a **shortest flow path** p s.t. $f(p) > 0$.
 - **Route** this pair along the flow path p .
 - **Discard** all flow paths that share an edge with p (i.e. set $f(p') = 0$ if p' shares an edge with p).
- Repeat until no fractional flow left.

Clearly gives an **edge-disjoint** routing.

But how good is this algorithm?

Analysis of the Rounding Algorithm

n : # of vertices m : # of edges

Let OPT be an optimal fractional solution.

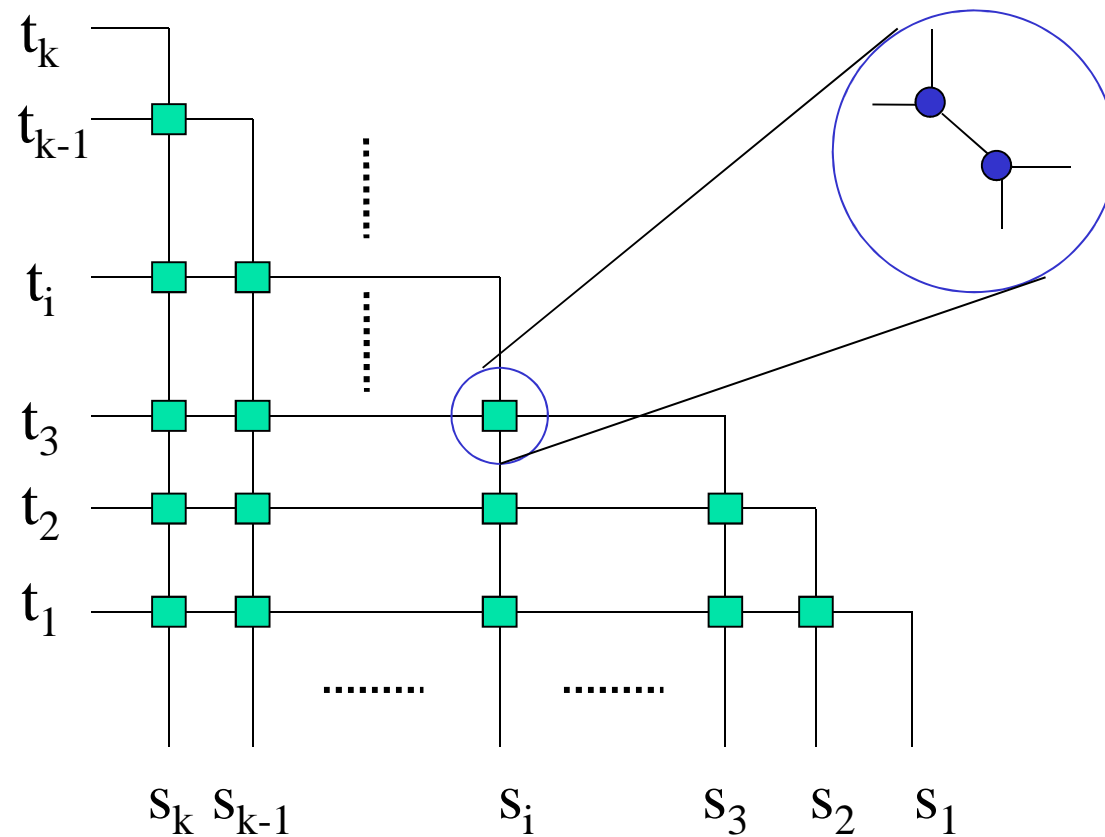
If chosen flow path p has length $\leq m^{1/2}$, routing a pair on p eliminates $\leq m^{1/2}$ units of flow from OPT .
(for every edge on p , we discard at most one unit of flow.)

Once shortest available flow path has length $\geq m^{1/2}$, total remaining fractional flow must be $\leq m^{1/2}$.
(total capacity = m , and each unit of flow consumes $\geq m^{1/2}$ capacity.)

We get an $O(m^{1/2})$ -approximation.

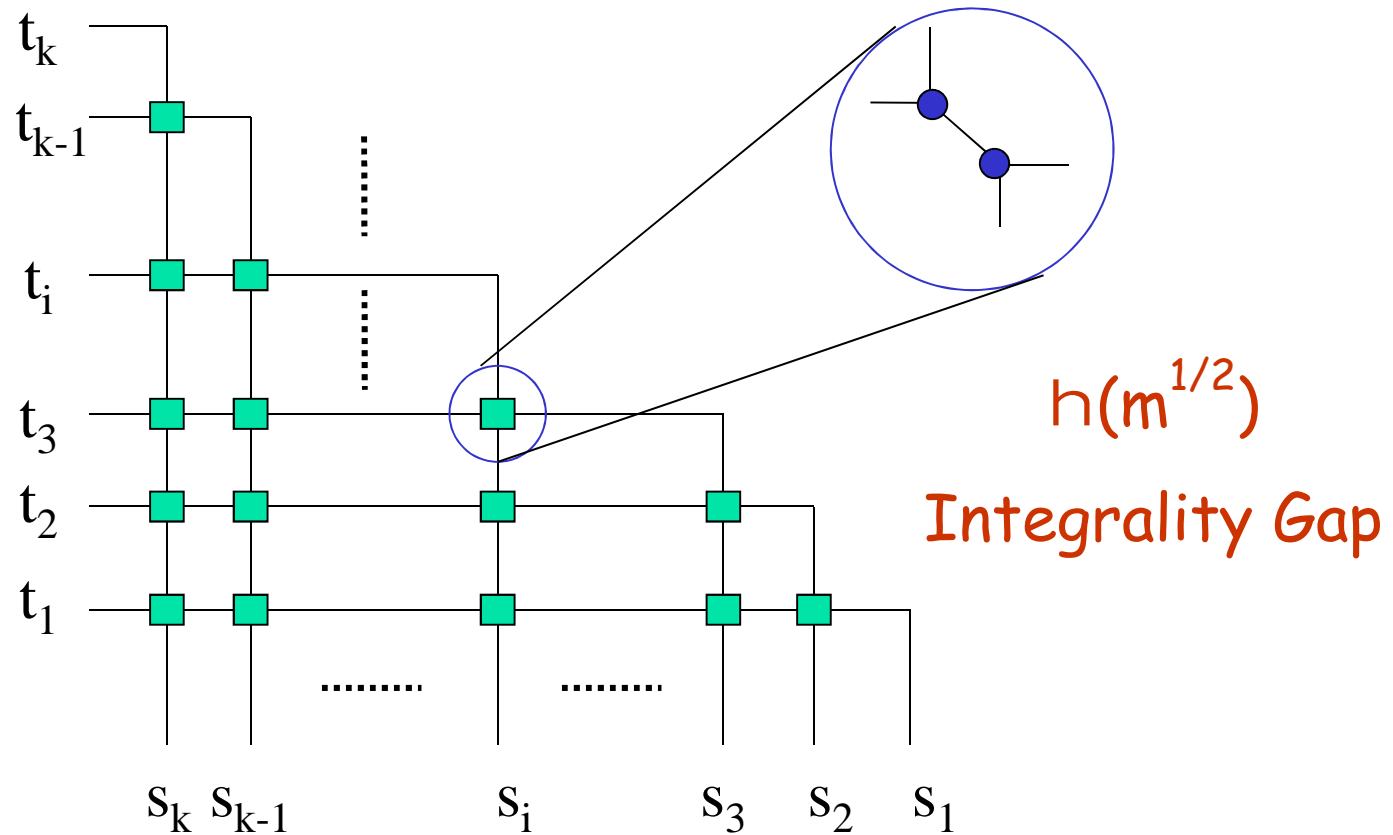
Could we do better?

[Garg, Vazirani, Yannakakis '93]



Could we do better?

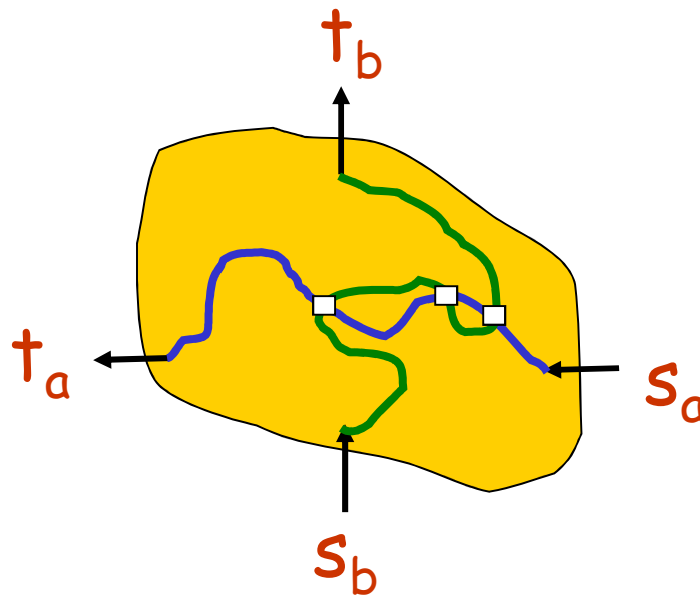
[Garg, Vazirani, Yannakakis '93]



Gap holds for planar undirected graphs

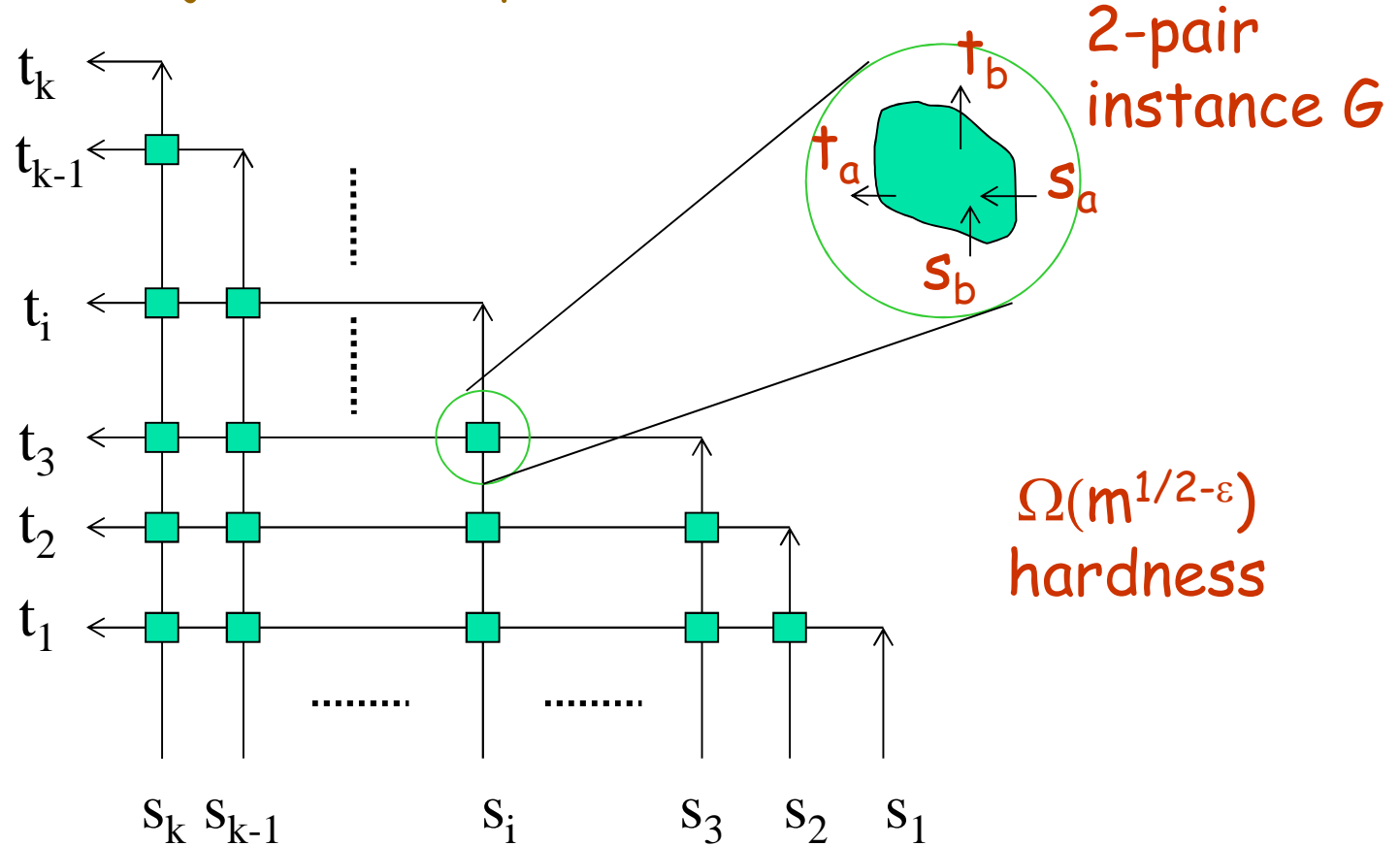
And if the Graph is Directed ...

[Fortune, Hopcroft, Wylie '80] Given a directed graph $G(V,E)$ and two pairs s_a-t_a and s_b-t_b , it is NP-hard to decide if we can route s_a to t_a and s_b to t_b on edge-disjoint paths.



Hardness of Approximation

[Guruswami, K, Rajaraman, Shepherd, Yannakakis '99]



EDP in Directed Acyclic Graphs

[Chekuri, K, Shepherd '06]

$O(n^{1/2})$ -approximation for EDP in DAGs via rounding of the multicommodity flow relaxation.

[Chalermsook, Laekhanukit, Nanongkai '14]

$\Omega(n^{1/2-\epsilon})$ -hardness for EDP in DAGs for any $\epsilon > 0$.

What if we allow congestion ...

EDP with congestion c : up to c paths can share an edge.

What happens to the integrality gap for $c \geq 2$?

Key question: does EDP become well-approximable with constant congestion?

What if we allow congestion ...

Randomized Rounding [Raghavan-Thompson '87]

- Route pair (s_i, t_i) with probability x_i .
- If (s_i, t_i) is chosen, pick an s_i - t_i flow path p for routing: choose with probability proportional to $f(p)$.

$O(1)$ -approximation with congestion $\Theta(\lg n / \lg \lg n)$.

[Raghavan and Thompson '87]

$O(n^{1/c})$ -approximation with constant congestion c .

[Srinivasan' 97], [Baveja-Srinivasan' 00],[Azar-Regev'01].

Could we do better?

EDP with Congestion in Directed Graphs

[Andrews, Zhang '06] [Chuzhoy, Guruswami, K, Talwar '07]

- Integrality gap of the flow relaxation is roughly $n^{1/(3c)}$ for c up to $\Theta(\log n / \log \log n)$.
- Also, $n^{\Omega(1/c)}$ -hardness for c up to $\Theta(\log n / \log \log n)$.
- Randomized rounding is *essentially optimal* for the *directed* edge-disjoint paths problems.

EDP with Congestion in Undirected Graphs

[Andrews, Zhang '05]

[Andrews, Chuzhoy, Guruswami, K, Talwar, Zhang '05]

- Integrality gap of the flow relaxation is at least $(\log n)^{1/(c+1)}$ with congestion c .
- $\Omega(\log^{1/(c+1)} n)$ -hardness with congestion c .

Undirected Graphs: State of the Art

$\Omega(\log^{1/(c+1)} n)$ -hardness with constant congestion c .

$O(n^{1/c})$ -approximation with constant congestion c .

- $O(\log n)$ -approximation with congestion 2 for planar graphs [Chekuri, K, Shepherd '05]
- $\text{Polylog}(n)$ approximation with no congestion for graphs with large minimum cut [Rao-Zhou '06]
- $\text{Polylog}(n)$ approximation with $\text{poly}(\lg \lg n)$ congestion in arbitrary graphs [Andrews '10]
- $\text{Polylog}(n)$ approximation with constant congestion in arbitrary graphs [Chuzhoy '12] [Chuzhoy-Li '12]

Well-linked Decomposition Framework for EDP

[Chekuri, K, Shepherd '04, '05]

- Start with a multicommodity flow solution but use it only to partition the graph into **well-linked** instances. We ignore the flow paths !
- Show that any well-linked instance contains a routing structure called a **crossbar** on which EDP is easy to solve.
- Route the given source-sink pairs using the crossbar.

Instance of EDP

G : the underlying graph.

$X : \{s_1, t_1, s_2, t_2, \dots, s_k, t_k\}$ is the terminal set.

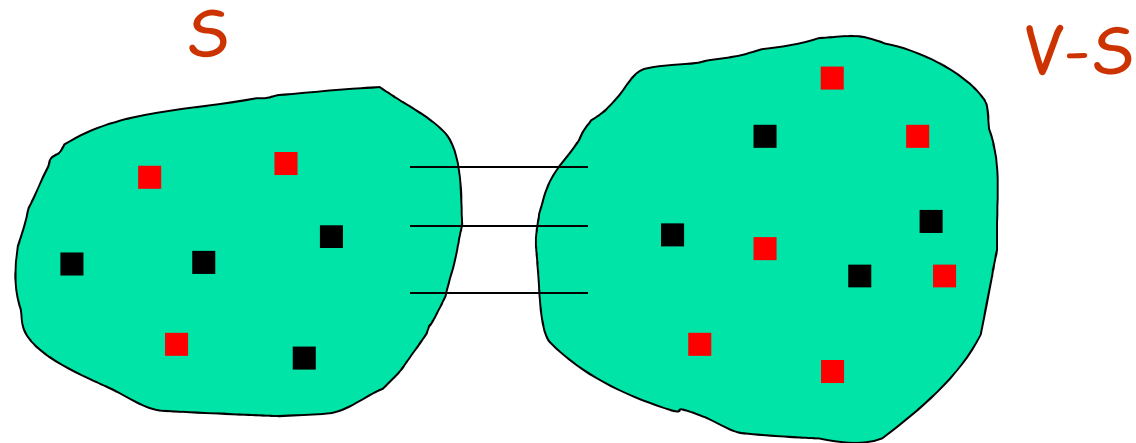
Assume w.l.o.g. that s_i, t_i are distinct and have degree 1 in the graph.

The goal is to route a given matching on X .

Also assume w.l.o.g. that degree of any vertex in G is bounded by 4.

Well-linked Set

Subset X is *well-linked* in G if for any partition $(S, V-S)$:
of edges cut \geq # of X vertices in the smaller side.



$$\forall S \subset V \text{ s.t. } |S \cap X| \leq |X|/2, |E(S, V-S)| \geq |S \cap X|.$$

Instance of EDP

G : the underlying graph.

$X : \{s_1, t_1, s_2, t_2, \dots, s_k, t_k\}$ is the terminal set.

Assume w.l.o.g. that s_i, t_i are distinct.

The goal is to route a given matching on X .

Well-linked Instance of EDP

G : the underlying graph.

$X : \{s_1, t_1, s_2, t_2, \dots, s_k, t_k\}$ is the terminal set.

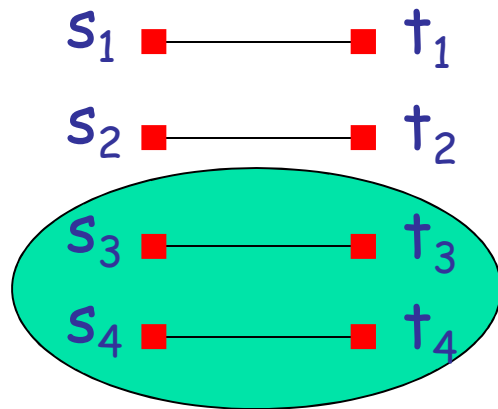
Assume w.l.o.g. that s_i, t_i are distinct.

The goal is to route a given matching on X .

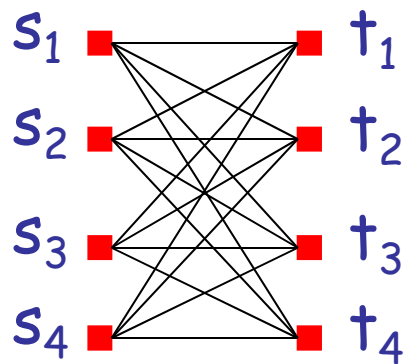
X is well-linked in G .

Theorem [Chekuri, K, Shepherd '05] Any instance of EDP can be reduced to a collection of well-linked instances with only a $\text{polylog}(n)$ factor loss in the solution value.

Examples

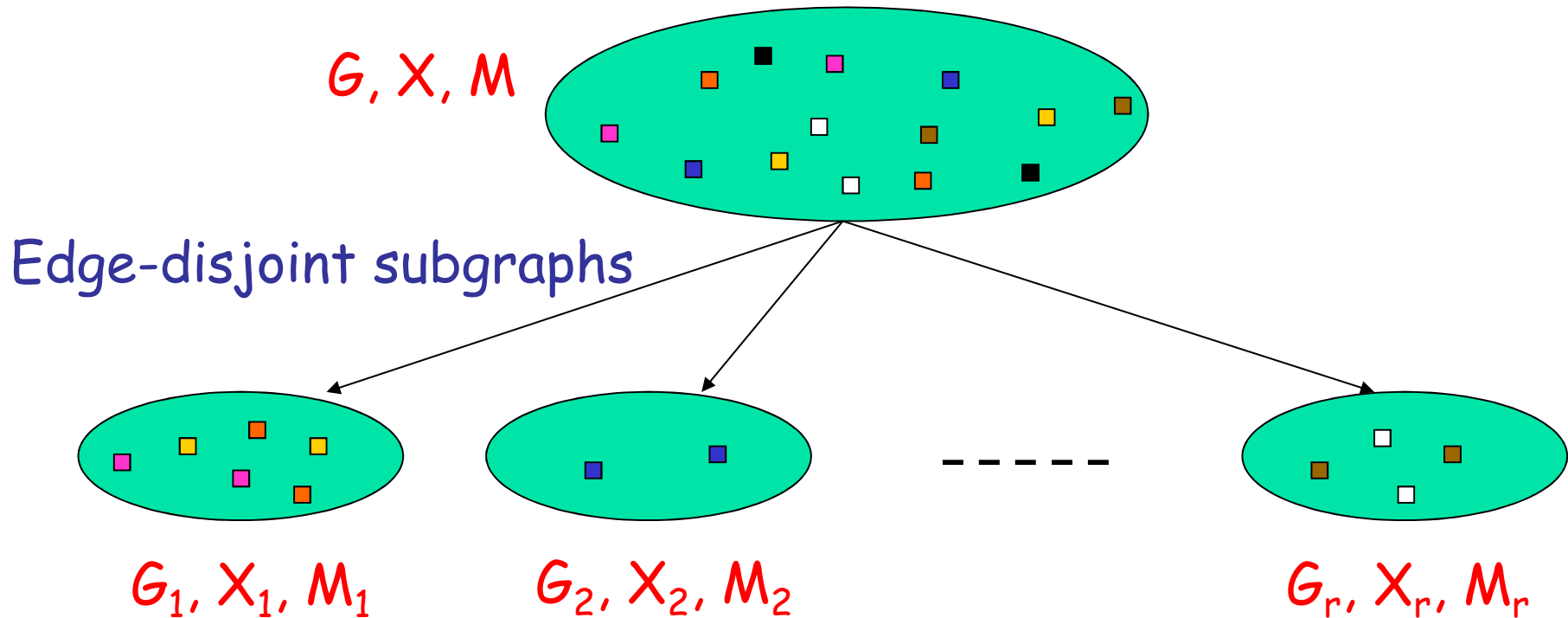


Not a well-linked instance



A well-linked instance

Well-linked Decomposition



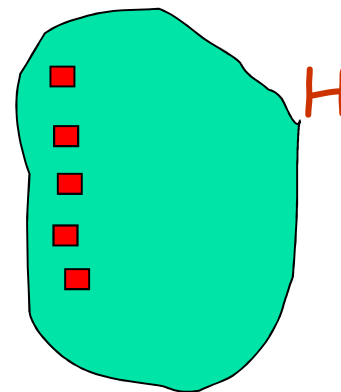
$$M_i \subset M$$

X_i is well-linked in G_i

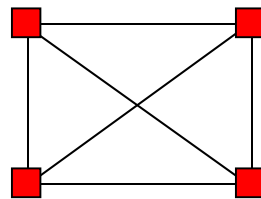
$$\sum_i |X_i| \geq \text{OPT}/\text{polylog}(k)$$

Crossbars

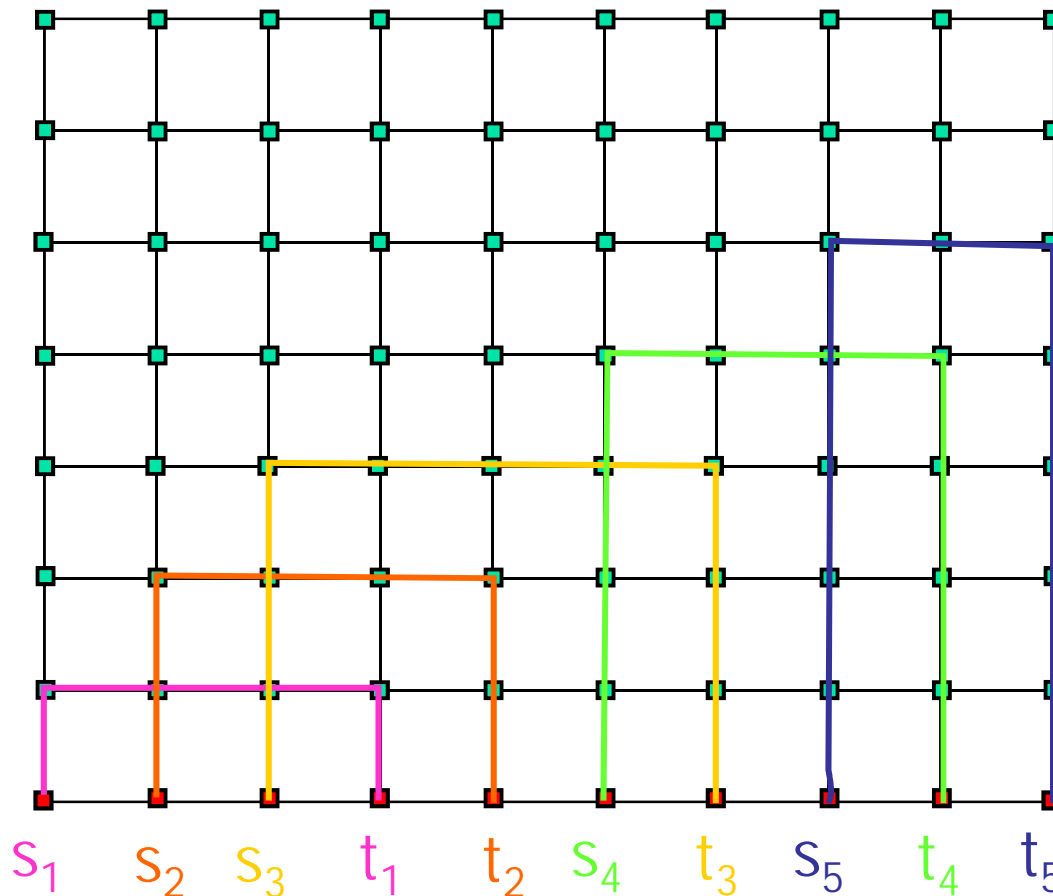
$H(V,E)$ is a cross-bar with respect to an interface $I \subseteq V$ if any matching on I can be routed using edge-disjoint paths.



Ex: a complete graph is a cross-bar with $I=V$



Grids as Crossbars



First row is
the interface

Application: EDP in Planar Graphs

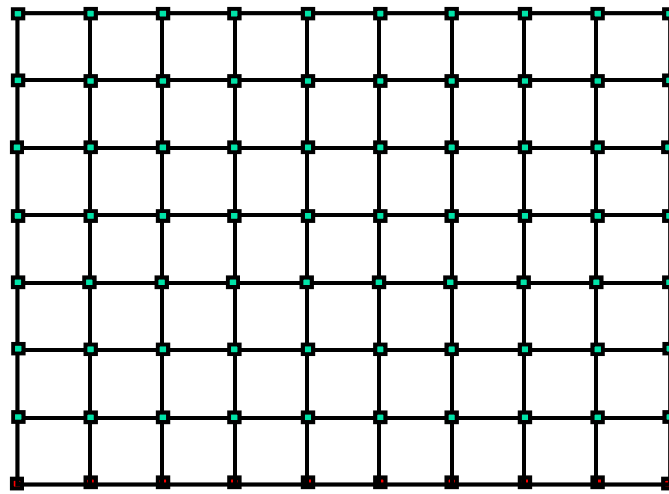
- Solve the multicommodity flow relaxation.
- Use the solution to partition the given instance into planar well-linked instances.
- Find a crossbar in each well-linked instance, and route using the crossbar.

Planar Well-linked Instances

Theorem [Robertson, Seymour, Thomas '94]

If G is a planar graph with k well-linked terminals, then with congestion 2 , an $\Omega(k) \times \Omega(k)$ grid H can be embedded in G .

Routing pairs in X using H

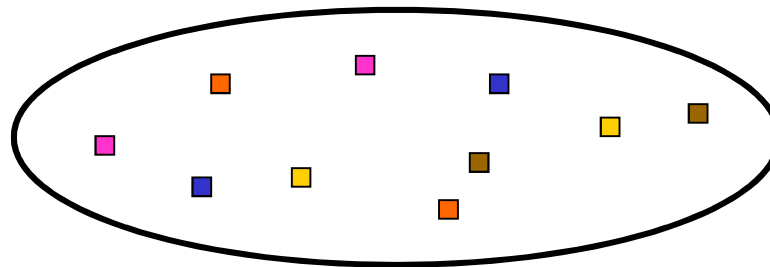


Grid

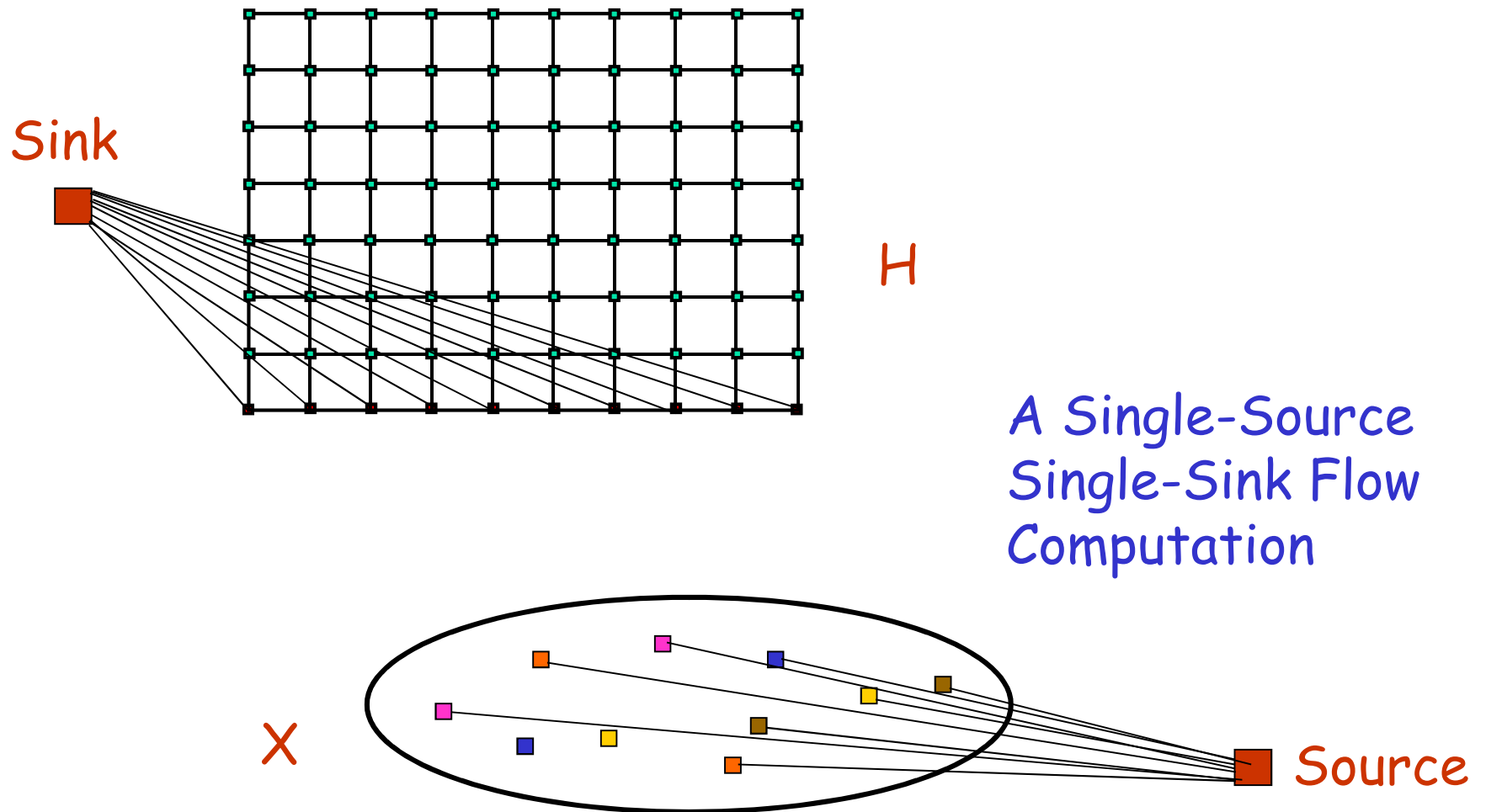
H

Interface I

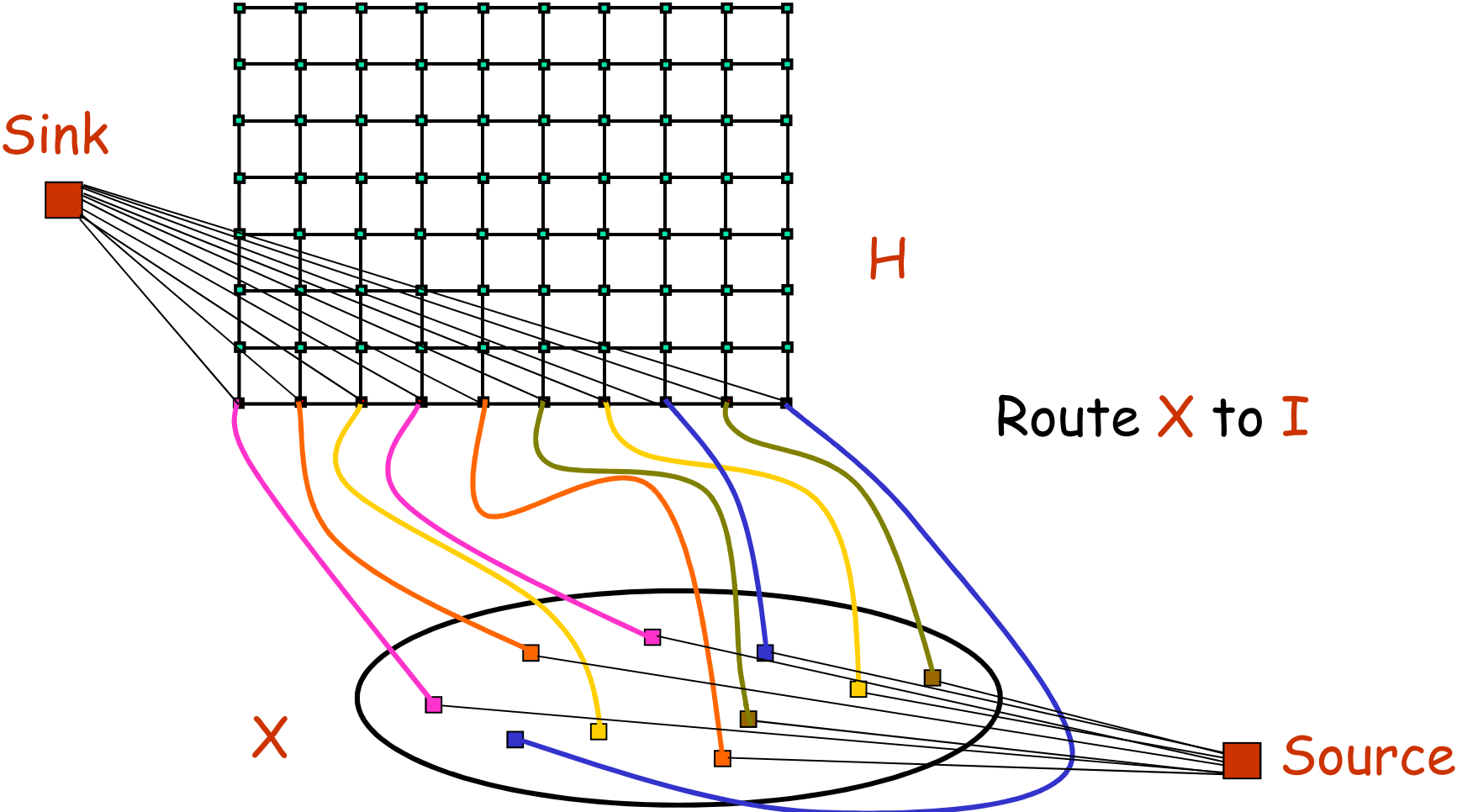
Terminals X



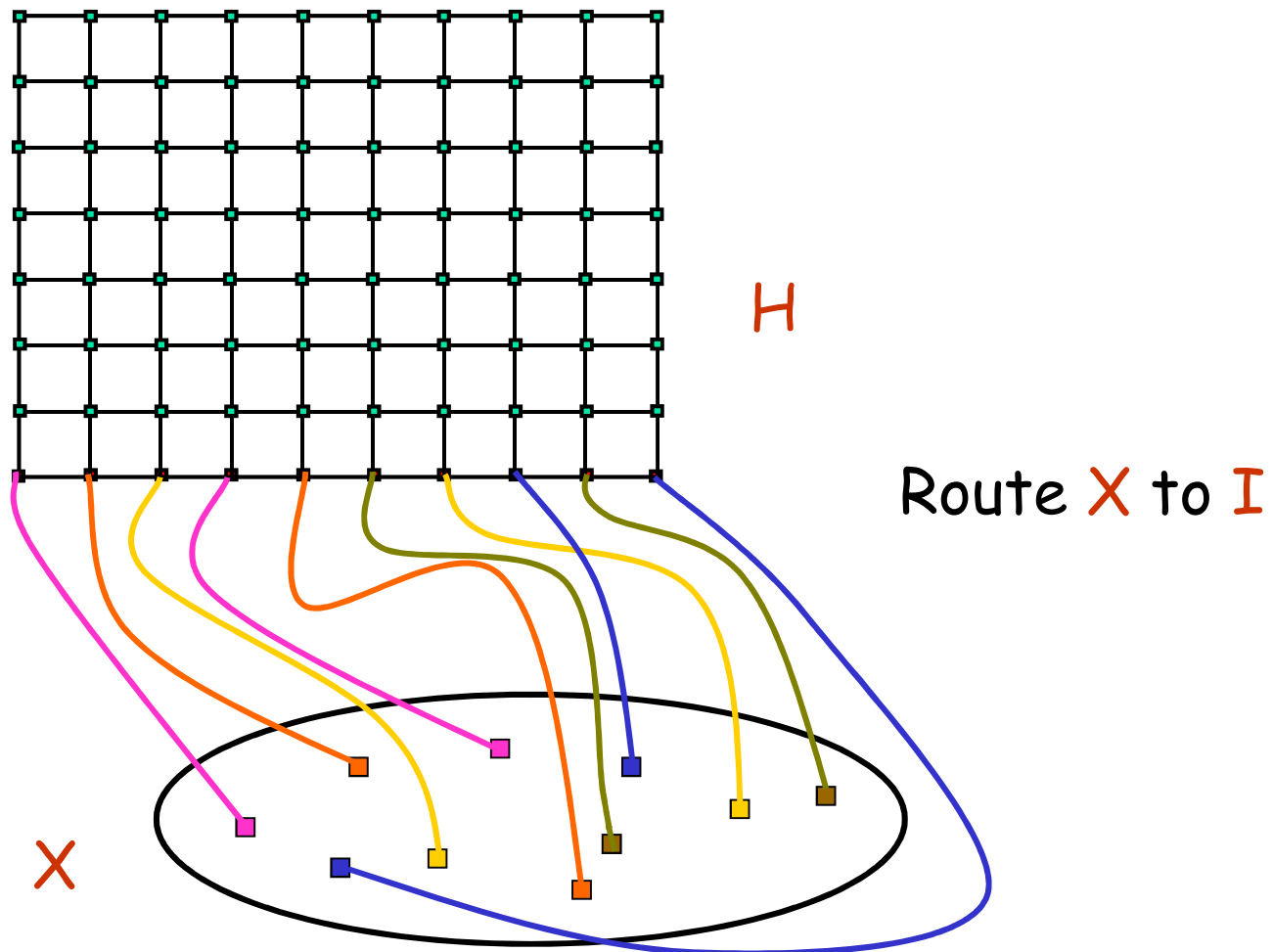
Routing pairs in X using H



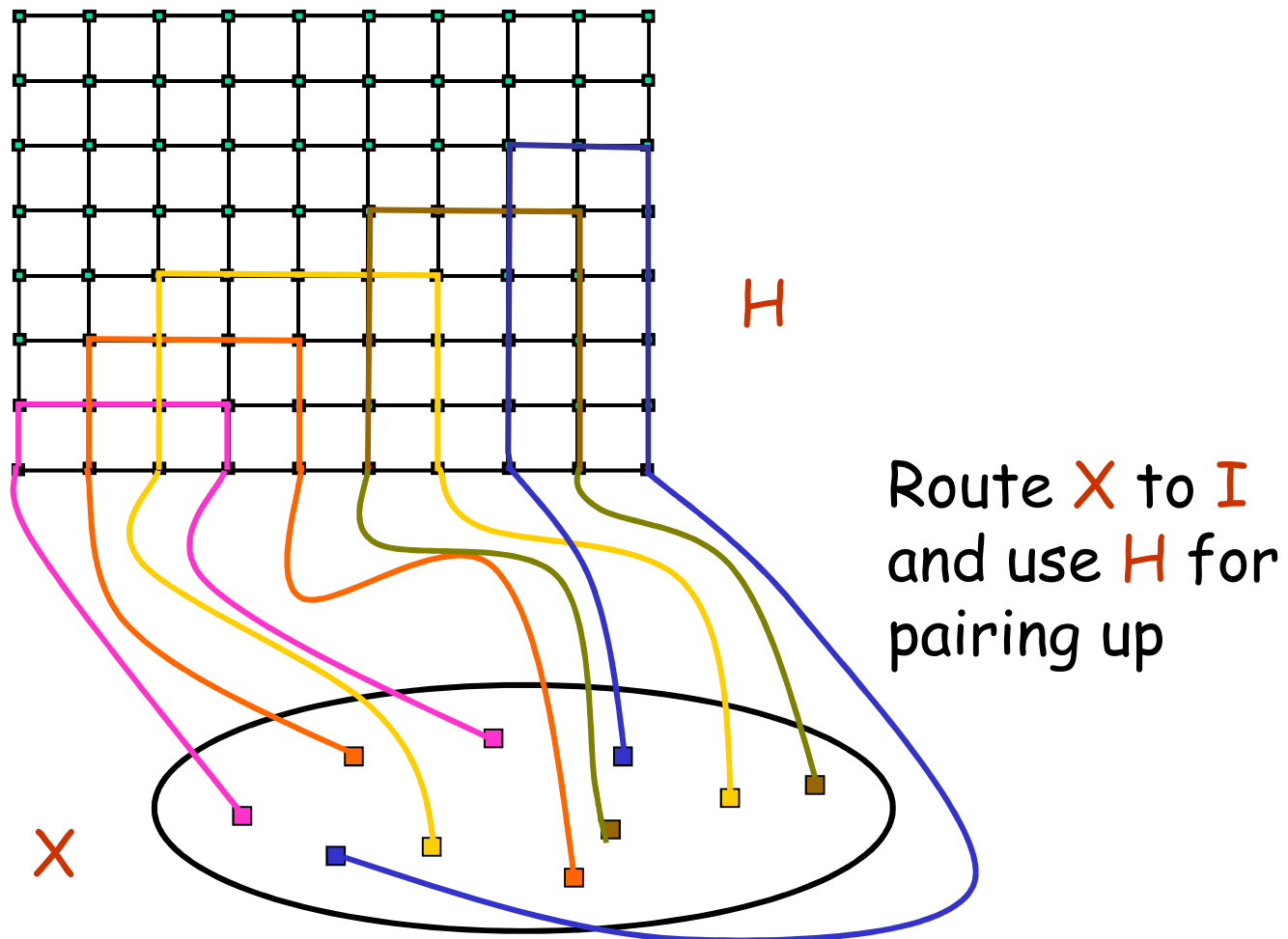
Routing pairs in X using H



Routing pairs in X using H



Routing pairs in X using H



EDP in Planar Graphs

Theorem [Chekuri, K, Shepherd '05]

EDP in planar graphs can be approximated to within a factor of $O(\log n)$ with congestion 2.

- Prior to this, only $n^{\Omega(1)}$ -approximation was known for any constant congestion.
- Recall that the integrality gap is $\Omega(n^{1/2})$ when no congestion is allowed.
- Later, $O(1)$ -approximation with congestion 4, and then $O(1)$ -approximation with congestion 2 [Chekuri, K, Shepherd '06; Seguin-Charbonneau, Shepherd '11].

Well-linked Decomposition

Flow Well-Linked Sets

A subset X is **flow**-well-linked in G if the following multicommodity flow is feasible in G :

$$\text{for } u, v \text{ in } X, d(uv) = 1/|X|$$

An instance of product multicommodity flow on X .

Cut vs Flow Well-Linked Sets

X flow-linked $\Rightarrow X$ is cut-linked

X cut-linked $\Rightarrow X$ is flow-linked with congestion $\beta(G)$

$\beta(G)$ - flow-cut gap for product multicommodity instances in G

Fractional Version

π : a non-negative weight function on X

$\pi(v)$: weight of v in X

X is π -cut-linked: for all $S \subseteq V$ with $\pi(S \cap X) \leq \pi(X)/2$,
 $|E(S, V-S)| \geq \pi(S \cap X)$

X is π -flow-linked: multicommodity flow instance with
 $d(uv) = \pi(u) \pi(v) / \pi(X)$ is feasible in G

Well-linked Instance

G : the underlying graph.

$X : \{s_1, t_1, s_2, t_2, \dots, s_k, t_k\}$ is the terminal set.

Assume w.l.o.g. that s_i, t_i are distinct.

The goal is to route a given matching on X .

X is well-linked in G .

Fractional Well-linked Instance

G : the underlying graph.

$X : \{s_1, t_1, s_2, t_2, \dots, s_k, t_k\}$ is the terminal set.

Assume w.l.o.g. that s_i, t_i are distinct.

The goal is to route a given matching on X .

X is π -well-linked in G , and for each pair s_j, t_j we have $\pi(s_j) = \pi(t_j)$.

Assume that for $0 \leq \pi(v) \leq 1$ all $v \in X$.

Decomposition using Sparse Cuts

We now describe the process for creating fractional flow well-linked instances.

Start with the LP solution for the given instance.

f_j : flow for pair $s_j t_j$.

$f = \sum_i f_j$ is the total flow in LP.

Define π to be $\pi(s_j) = \pi(t_j) = f_j$.

Decomposition Algorithm

$\beta(G)$ - flow-cut gap in G

If X is $\pi/(10 \beta(G) \log k)$ -flow-linked then stop;

Else

Find an approximate sparse cut $(S, V-S)$ w.r.t. π in G

Remove flow on edges of the cut $(S, V-S)$

$G_1 = G[S], G_2 = G[V-S]$

Recurse on G_1, G_2 with the remaining flow

Analysis

Suppose the remaining graphs at end of recursion are:

(G_1, X_1, π_1) , (G_2, X_2, π_2) , ..., (G_r, X_r, π_r)

π_i is the remaining flow for X_i

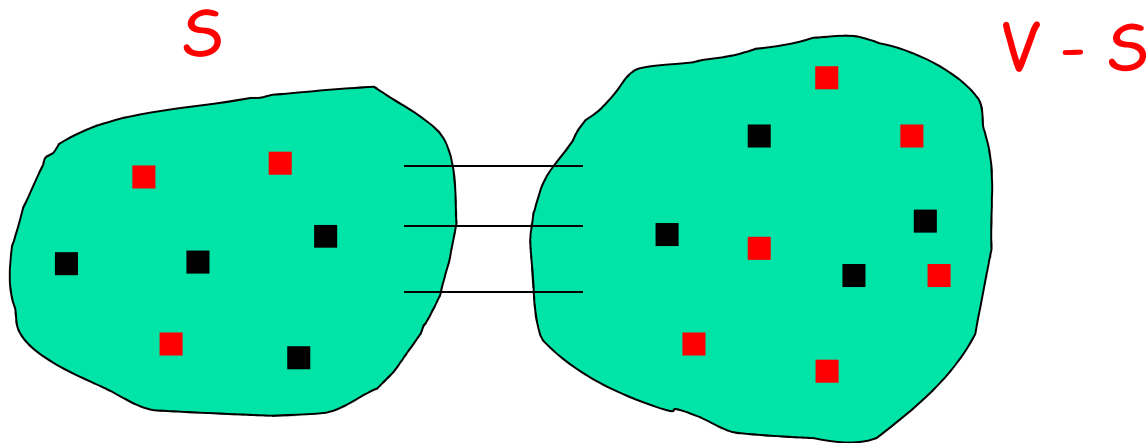
X_i is $\pi_i / (10 \beta(G) \log k)$ flow-linked in G_i

$\sum_i \pi_i(X_i) \geq (\text{Original flow}) - (\# \text{ of edges cut})$

Bounding the # of Edges Cut

X is not $\pi/(10 \beta(G) \log k)$ flow-linked

\Rightarrow # of edges in the cut $(S, V-S) \leq \pi(S)/(10 \log k)$



Analysis Continued ...

Claim: total number of edge cut is at most $f/2$.

$T(x)$: max # of edges cut if started with flow x

$$T(f) \leq T(f_1) + T(f_2) + f_1 / (10 \log k)$$

$$\Rightarrow T(f) \leq f/2.$$

Thus $\sum_i \pi_i(X_i) \geq f/2$.

Each X_i is $\pi_i / (10 \beta(G) \log k)$ flow-well-linked.

Fractional to Integer Well-linked

Theorem[Chekuri, K, Shepherd '05]

Given an input instance G, X, M where X is π -flow well-linked, we can recover G, X', M' such that

- X' is $\frac{1}{2}$ -flow well-linked,
- $|X'| = \Omega(\pi(X))$, and
- $M' \subseteq M$, is a matching defined over X' .

Proof Idea: Use a spanning tree to cluster fractional mass into integral units.

A similar result can be shown for cut well-linked.

Thank You!
