

# New algorithms for Disjoint Paths and Routing Problems

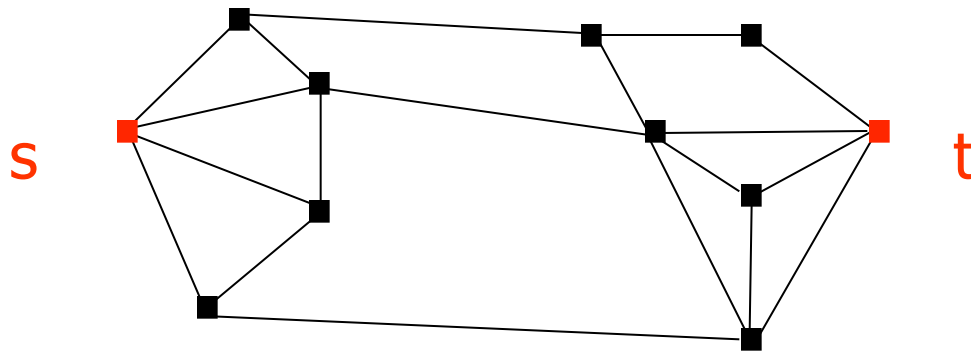
**Chandra Chekuri**

Dept. of Computer Science  
Univ. of Illinois (UIUC)

# Menger's Theorem

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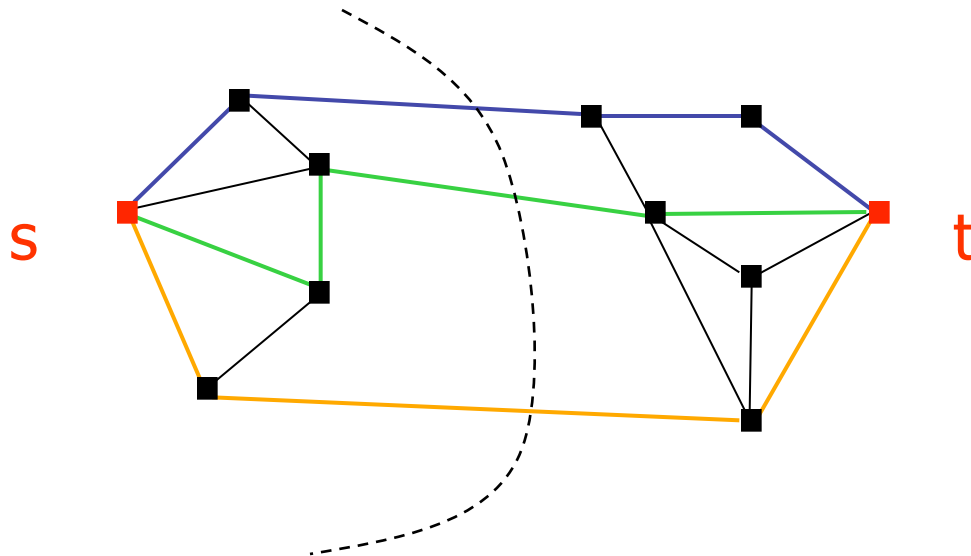
**Theorem:** The maximum number of  $s$ - $t$  edge-disjoint paths in a graph  $G=(V,E)$  is equal to minimum number of edges whose removal disconnects  $s$  from  $t$ .



# Menger's Theorem

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# Max-Flow Min-Cut Theorem

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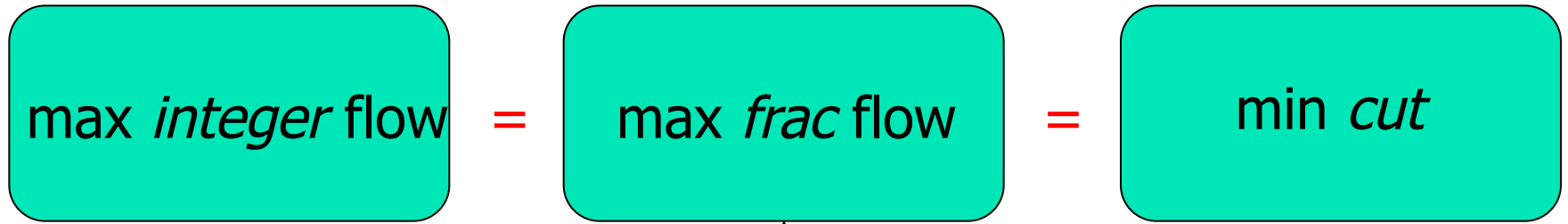
[Ford-Fulkerson]

**Theorem:** The maximum  $s$ - $t$  flow in an edge-capacitated graph  $G=(V,E)$  is equal to minimum  $s$ - $t$  cut. If capacities are integer valued then max fractional flow is equal to max integer flow.

# Computational view

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difficult to compute directly



max *integer* flow

=

max *frac* flow

=

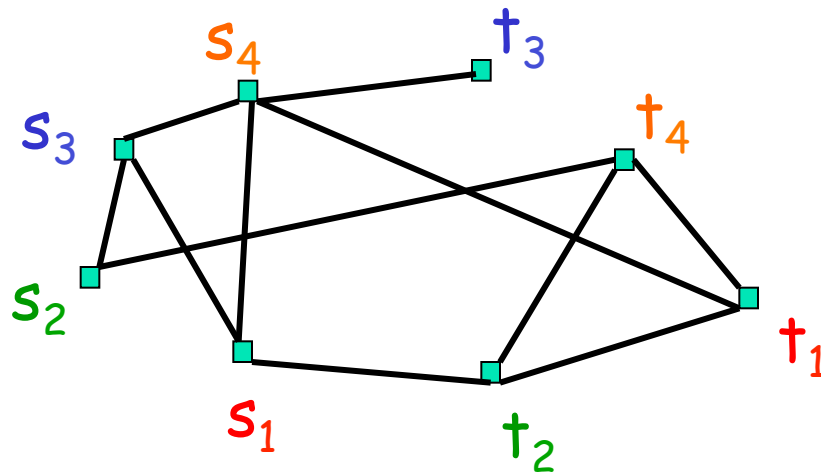
min *cut*

“easy” to compute

# Multi-commodity Setting

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Several pairs:  $s_1t_1, s_2t_2, \dots, s_kt_k$

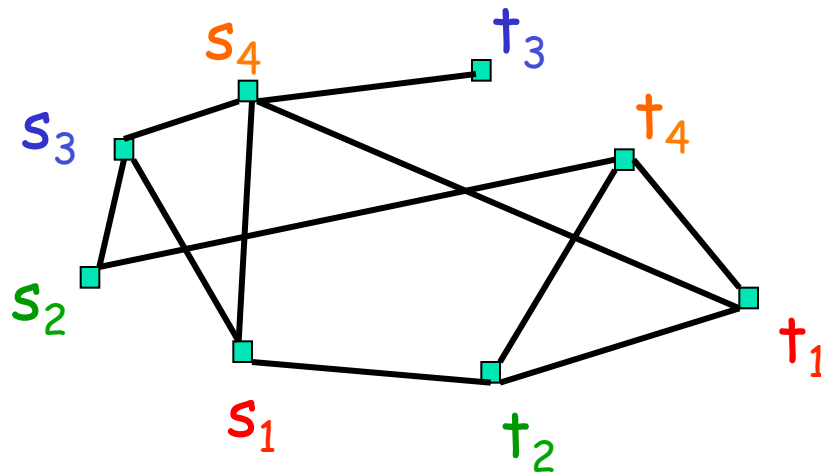


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Can *all* pairs be connected via edge-disjoint paths?

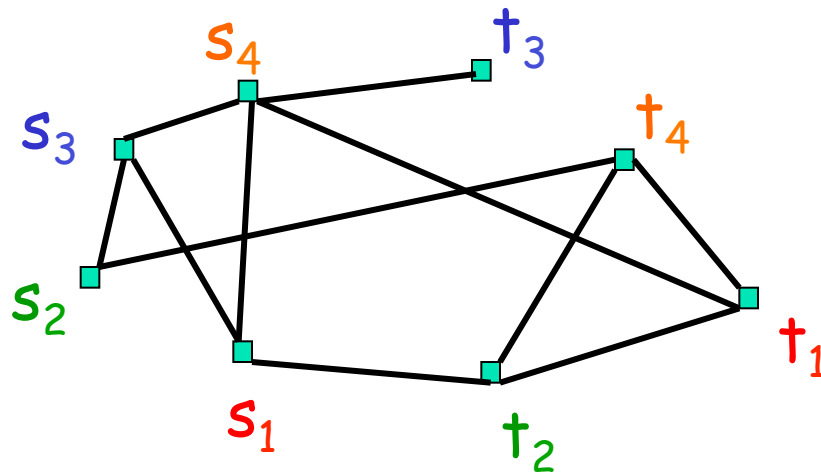


# Multi-commodity Setting

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Several pairs:  $s_1t_1, s_2t_2, \dots, s_kt_k$

Can *all* pairs be connected via edge-disjoint paths?  
*Maximize* number of pairs that can be connected



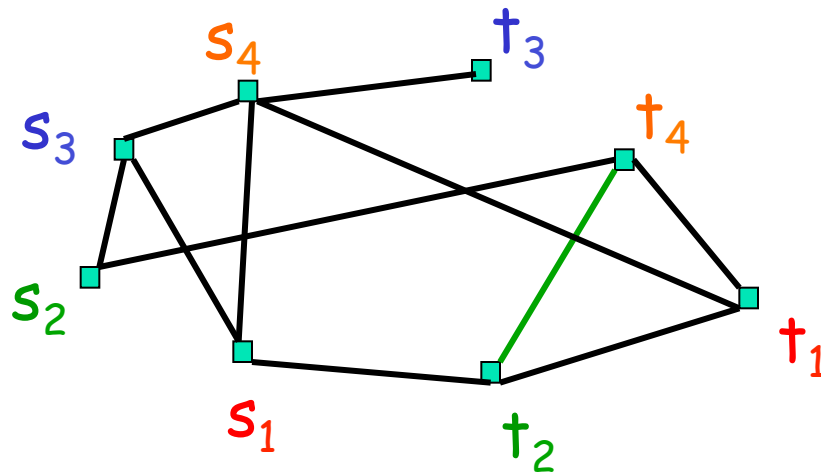


# Maximum Edge Disjoint Paths Prob

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Input: Graph  $G(V,E)$ , node 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

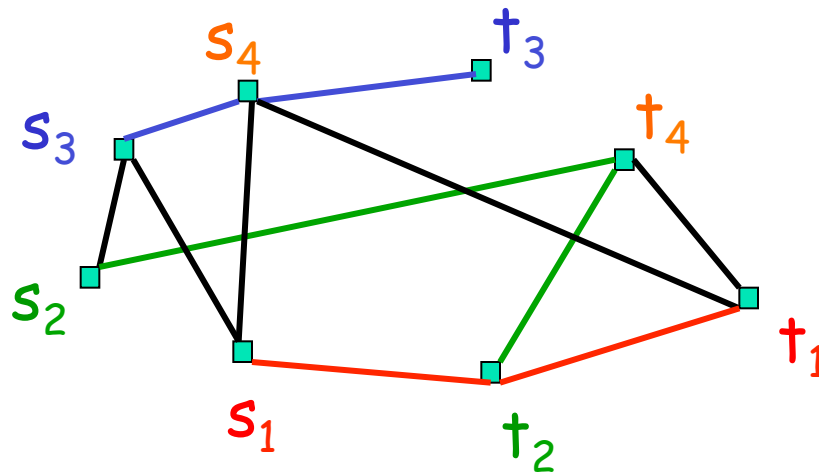


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# Motivation

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Basic problem in combinatorial optimization

Applications to VLSI, network design and routing, resource allocation & related areas

Related to significant theoretical advances

- Graph minor work of Robertson & Seymour
- Randomized rounding of Raghavan-Thompson
- Routing/admission control algorithms

# Computational complexity of MEDP

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Directed graphs: 2-pair problem is NP-Complete  
[Fortune-Hopcroft-Wylie' 80]

Undirected graphs: for any fixed constant  $k$ , there  
is a polynomial time algorithm  
[Robertson-Seymour' 88]

NP-hard if  $k$  is part of input

# Approximation

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Is there a good *approximation algorithm*?

- polynomial time algorithm
- for *every* instance  $I$  returns a solution of value at least  $OPT(I)/\alpha$  where  $\alpha$  is approx ratio

How useful is the flow *relaxation*?

- What is its integrality gap?

# Current knowledge

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- If  $P \neq NP$ , problem is hard to approximate to within polynomial factors in *directed graphs*
- In undirected graphs, problem is quite open
  - upper bound -  $O(n^{1/2})$  [C-Khanna-Shepherd' 06]
  - lower bound -  $\Omega(\log^{1/2-\epsilon} n)$  [Andrews etal' 06]
- Main approach is via flow relaxation

# Current knowledge

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- Main approach is via flow relaxation

Rest of talk: focus on *undirected* graphs

# Flow relaxation

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For each pair  $s_i, t_i$  allow fractional flow  $x_i \in [0, 1]$

Flow for each pair can use *multiple* paths

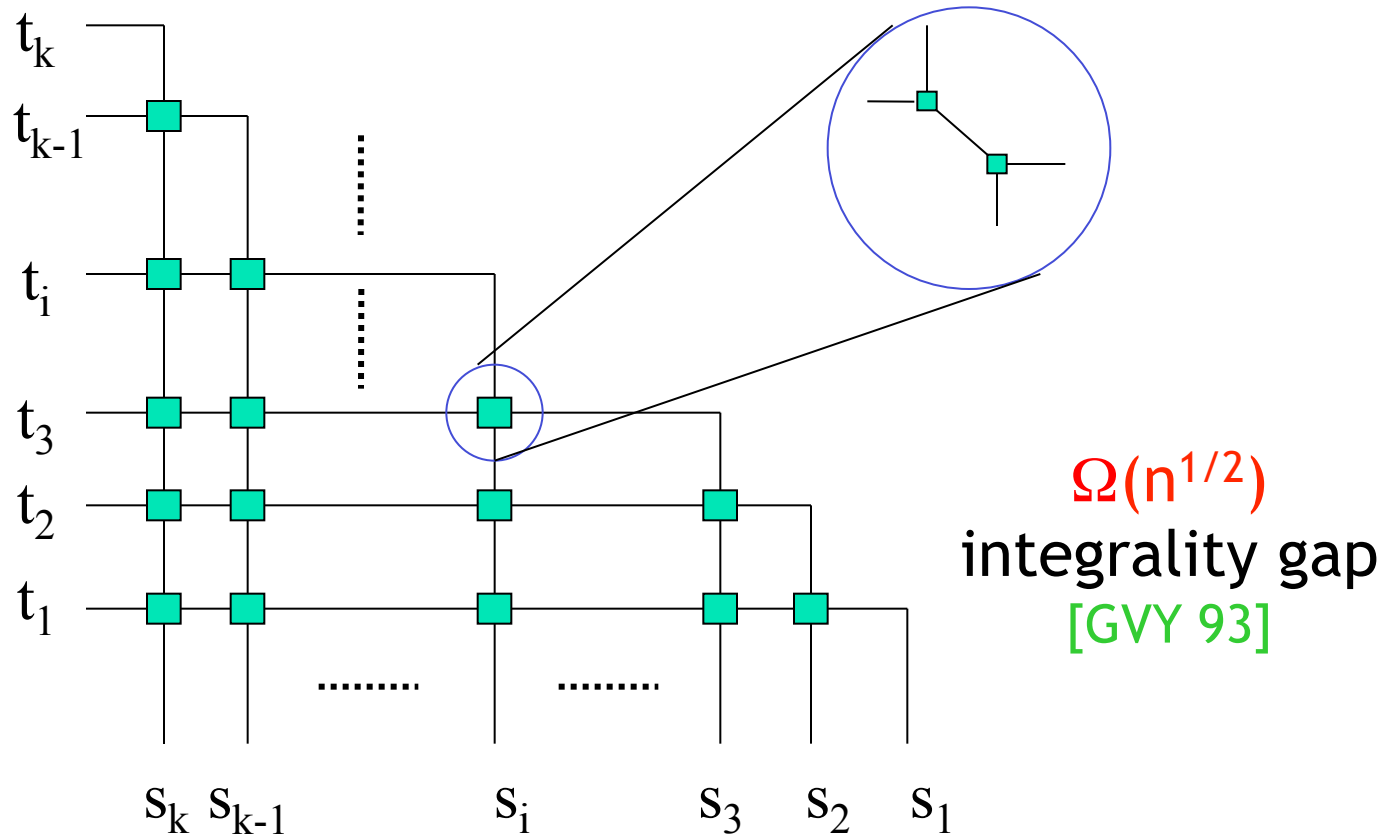
Total flow for all pairs on each edge  $e$  is  $\leq 1$

Total fractional flow =  $\sum_i x_i$

Relaxation can be solved in polynomial time using linear programming (faster approximate methods also known)



# Example



max integer flow = **1**, max fractional flow =  **$k/2$**

# Overcoming integrality gap

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Two approaches:

- Allow some small congestion  $c$ 
  - up to  $c$  paths can use an edge

# Overcoming integrality gap

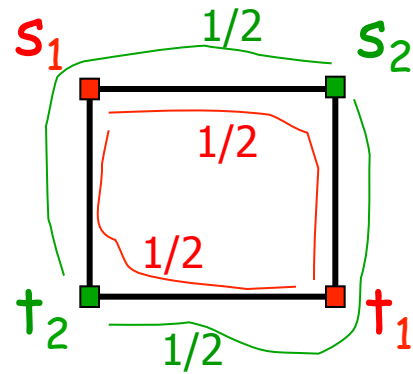
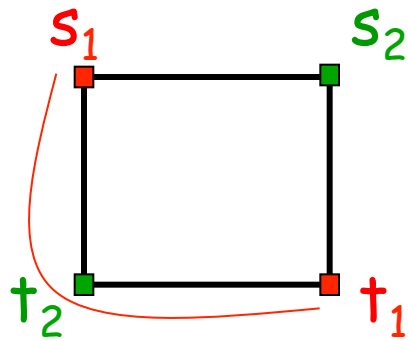
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Two approaches:

- Allow some small congestion  $c$ 
  - up to  $c$  paths can use an edge
  - $c=2$  is known as half-integer flow path problem
- All-or-nothing flow problem
  - $s_i t_i$  is routed if *one unit of flow* is sent for it (can use multiple paths) [C.-Mydlarz-Shepherd' 03]

# Example

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# Prior work on approximation

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Greedy algorithms or randomized rounding of flow

- polynomial approximation ratios in general graphs.  $O(n^{1/c})$  with congestion  $c$
- better bounds in various special graphs: trees, rings, grids, graphs with high expansion

No techniques to take advantage of relaxations:  
congestion or all-or-nothing flow

# New framework

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[C-Khanna-Shepherd]

New framework to understand flow relaxation

Framework allows near-optimal approximation algorithms for planar graphs and several other results

Flow based relaxation is much better than it appears

New connections, insights, and questions

# Some results

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**OPT**: optimum value of the flow relaxation

**Theorem:** In *planar graphs*

- can route  $\Omega(\text{OPT}/\log n)$  pairs with  $c=2$  for both edge and node disjoint problems
- can route  $\Omega(\text{OPT})$  pairs with  $c=4$

**Theorem:** In any graph  $\Omega(\text{OPT}/\log^2 n)$  pairs can be routed in all-or-nothing flow problem.

# Flows, Cuts, and Integer Flows

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Multicommodity: several pairs

NP-hard

Polytime via LP

NP-hard

max *integer* flow

$\leq$

max *frac* flow

$\leq$

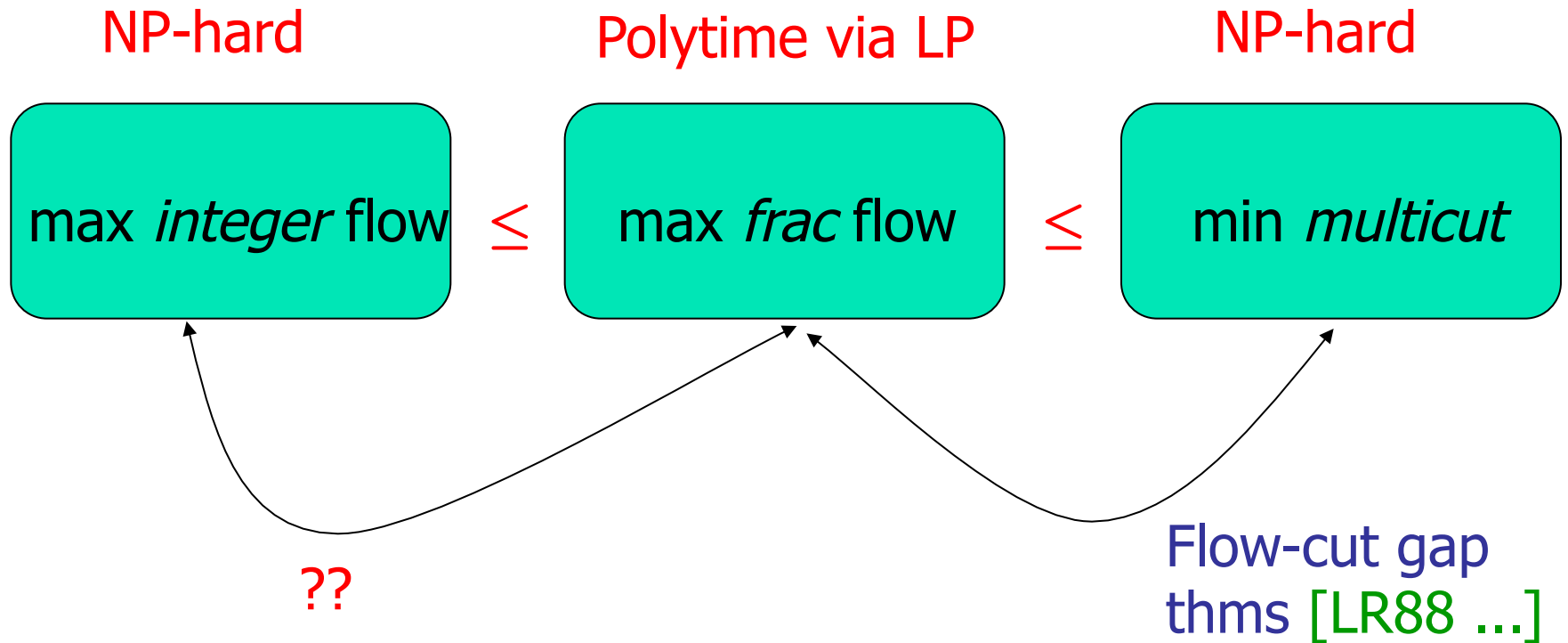
min *multicut*



# Flows, Cuts, and Integer Flows

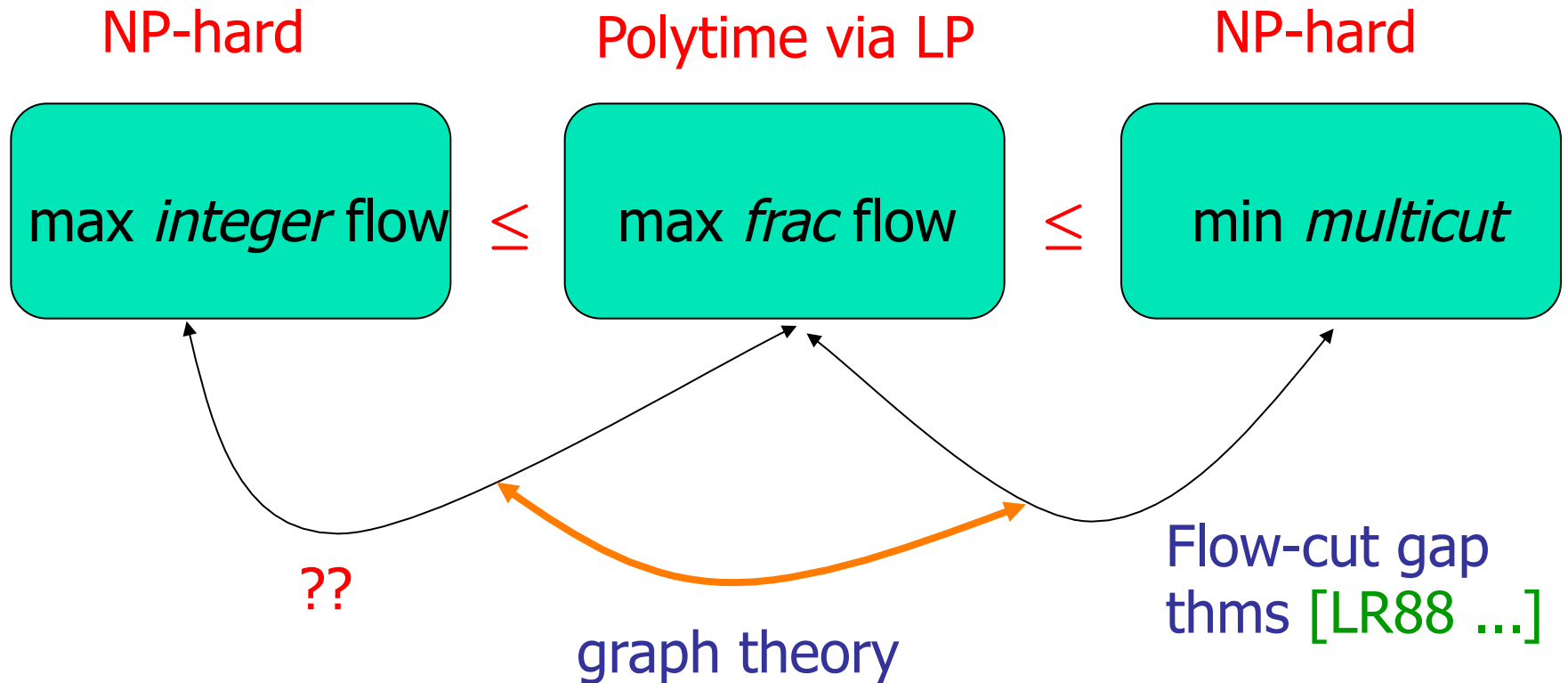
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Multicommodity: several pairs



# Flows, Cuts, and Integer Flows

Multicommodity: several pairs



# Part II: Details

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# New algorithms for routing

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1. Compute maximum **fractional flow**
2. Use fractional flow solution to **decompose** input instance into a collection of **well-linked** instances.
3. **Well-linked** instances have **nice properties** – exploit them to route

# Some simplifications

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Input: undir graph  $G=(V,E)$  and pairs  $s_1t_1, \dots, s_kt_k$

$X = \{s_1, t_1, s_2, t_2, \dots, s_k, t_k\}$  -- *terminals*

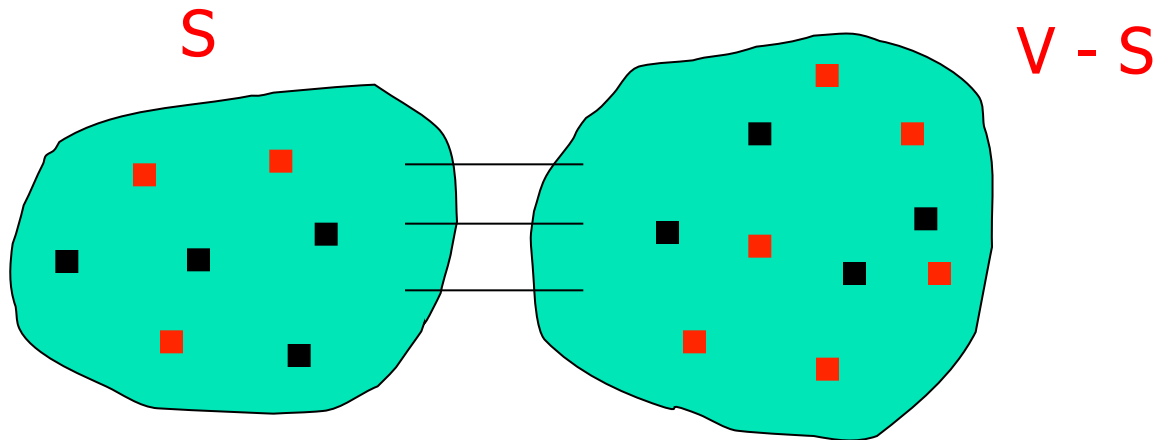
Assumption: wlog each terminal in only one pair

Instance:  $(G, X, M)$  where  $M$  is matching on  $X$

# Well-linked Set

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Subset  $X$  is *well-linked* in  $G$  if for every partition  $(S, V-S)$ , # of edges cut is at least # of  $X$  vertices in smaller side



for all  $S \subset V$  with  $|S \cap X| \leq |X|/2$ ,  $|\delta(S)| \geq |S \cap X|$

# Well-linked instance of EDP

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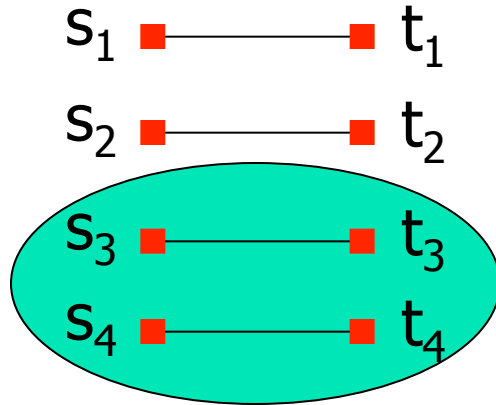
Input instance:  $(G, X, M)$

$X = \{s_1, t_1, s_2, t_2, \dots, s_k, t_k\}$  – terminal set

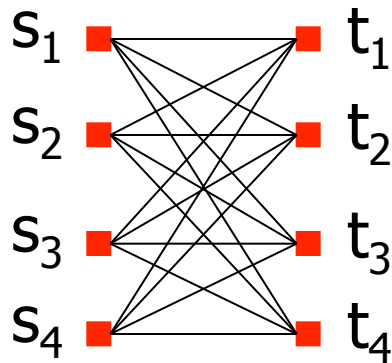
Instance is well-linked if  $X$  is well-linked in  $G$

# Examples

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Not a well-linked instance



A well-linked instance



# New algorithms for routing

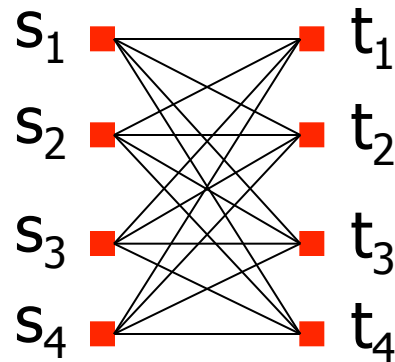
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1. Compute maximum **fractional flow**
2. Use fractional flow solution to **decompose** input instance into a collection of **well-linked** instances.
3. **Well-linked** instances have **nice properties** – exploit them to route

# Advantage of well-linkedness

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LP value does not depend on input matching **M**

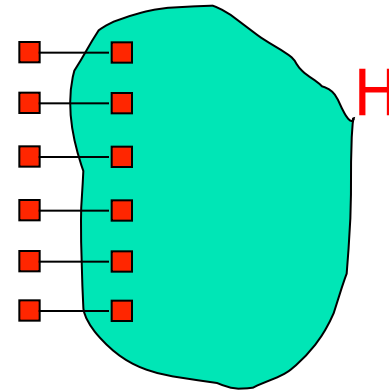


**Theorem:** If  $X$  is well-linked, then for *any* matching on  $X$ , LP value is  $\Omega(|X|/\log |X|)$ . For planar  $G$ , LP value is  $\Omega(|X|)$

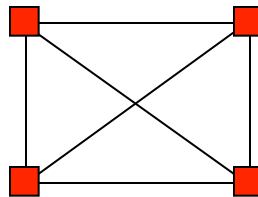
# Crossbars

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

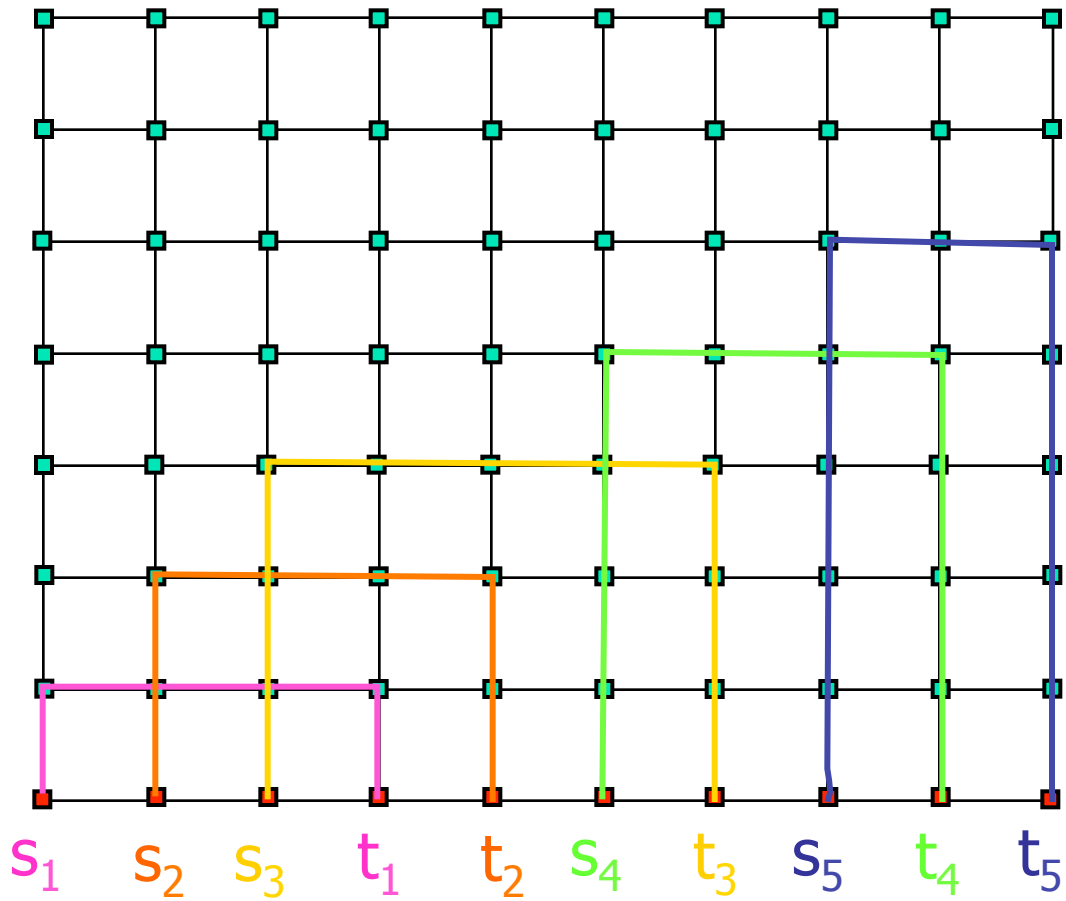


Ex: a complete graph is a cross-bar with  $I=V$



# Grids as crossbars

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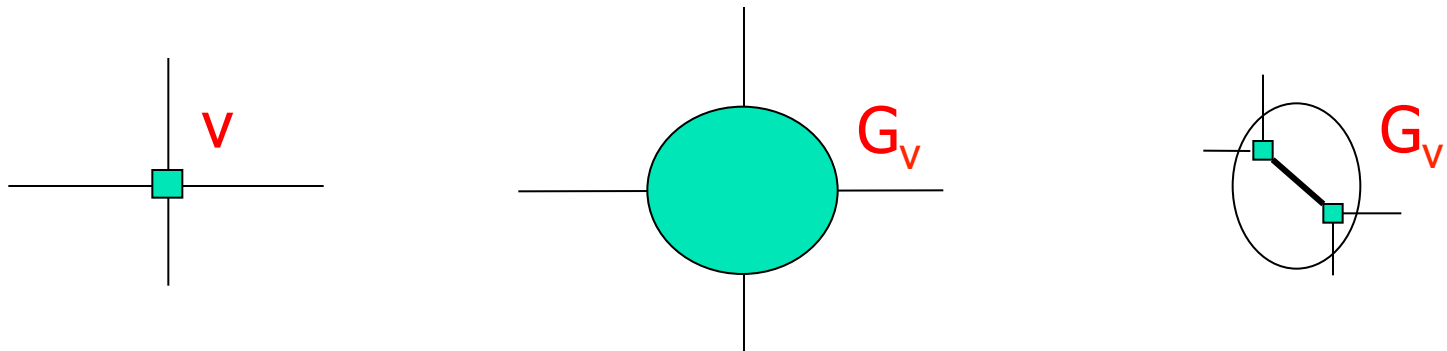


First row is  
*interface*

# Grids in Planar Graphs

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**Theorem[RST94]:** If  $G$  is *planar graph* with *treewidth*  $h$ , then  $G$  has a grid minor of size  $\Omega(h)$  as a subgraph.



Grid minor is crossbar with congestion  $2$

# Back to Well-linked sets

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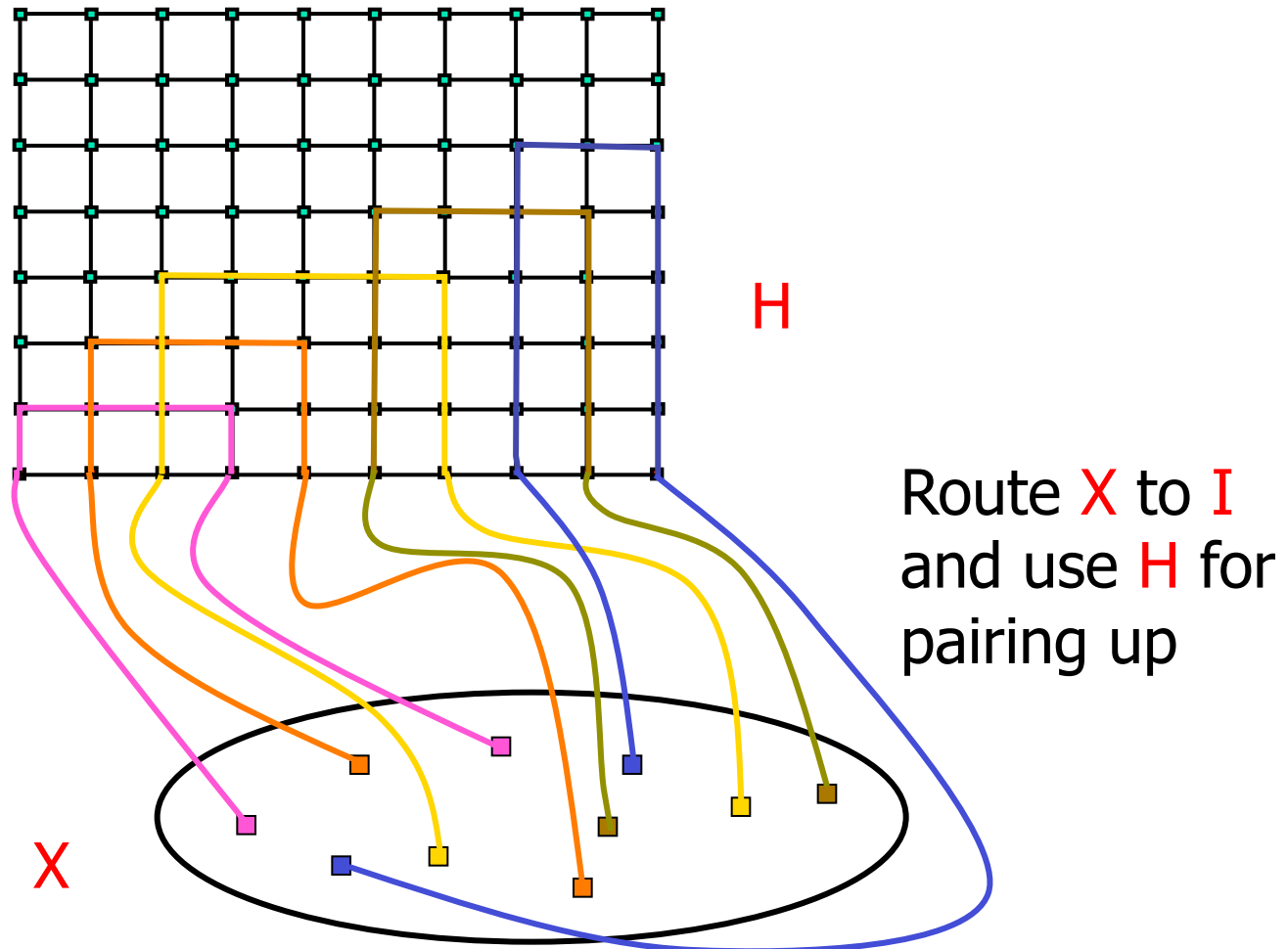
Claim:  $X$  is *well-linked* implies  $\text{treewidth} = \Omega(|X|)$

$X$  *well-linked*  $\Rightarrow G$  has grid minor  $H$  of size  $\Omega(|X|)$

Q: how do we route  $M = (s_1 t_1, \dots, s_k t_k)$  using  $H$  ?

# Routing pairs in X using H

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# Several technical issues

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- What if  $X$  cannot reach  $H$ ?
- $H$  is smaller than  $X$ , so can *pairs* reach  $H$ ?
- Can  $X$  reach  $H$  without using edges of  $H$ ?
- Can  $H$  be found in polynomial time?



# General Graphs?

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Grid-theorem extends to graphs that exclude a fixed minor [RS, DHK' 05]

For general graphs, need to prove following:

**Conjecture:** If  $G$  has treewidth  $h$  then it has an *approximate crossbar* of size  $\Omega(h/\text{polylog}(n))$

Crossbar  $\Leftrightarrow$  LP relaxation is good

# Reduction to Well-linked case

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Given  $G$  and  $k$  pairs  $s_1t_1, s_2t_2, \dots, s_kt_k$

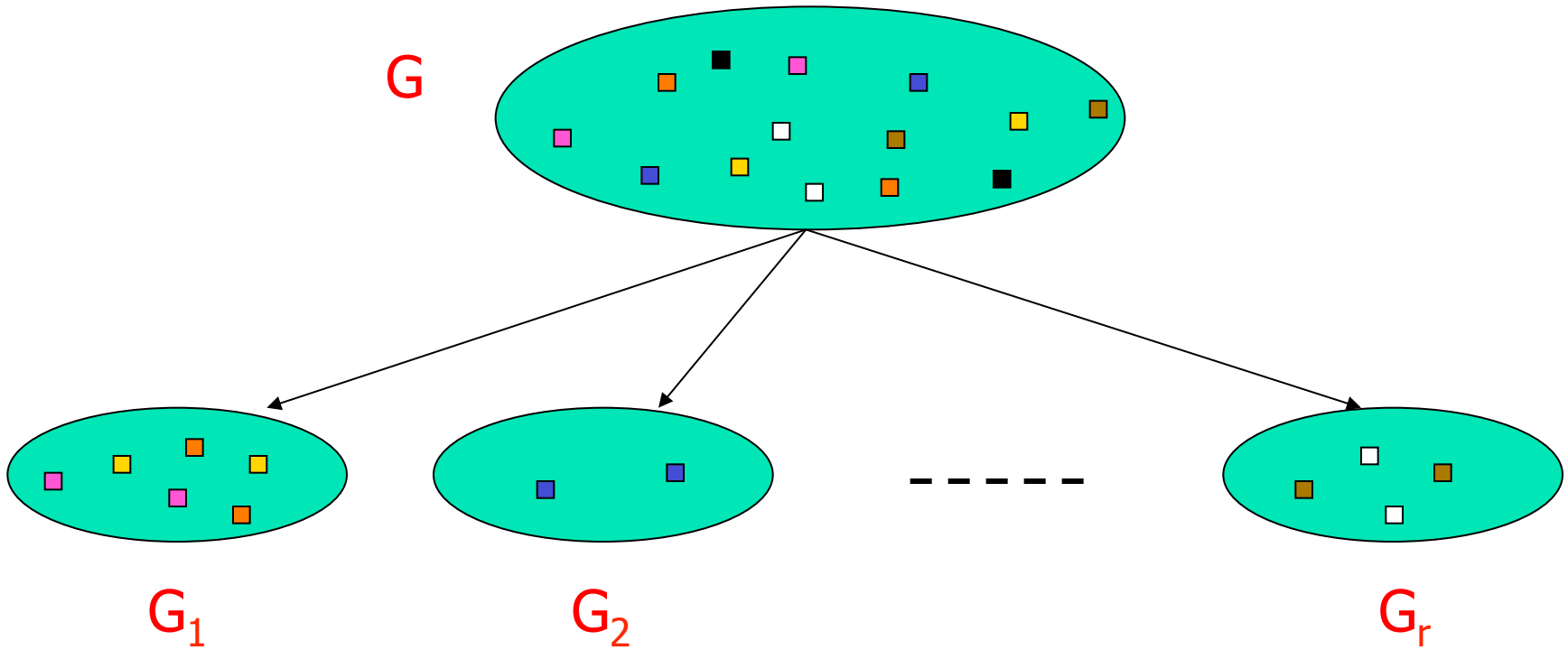
$$X = \{s_1, t_1, s_2, t_2, \dots, s_k, t_k\}$$

We know how to solve problem if  $X$  is well-linked

**Q:** can we reduce general case to well-linked case?

# Decomposition

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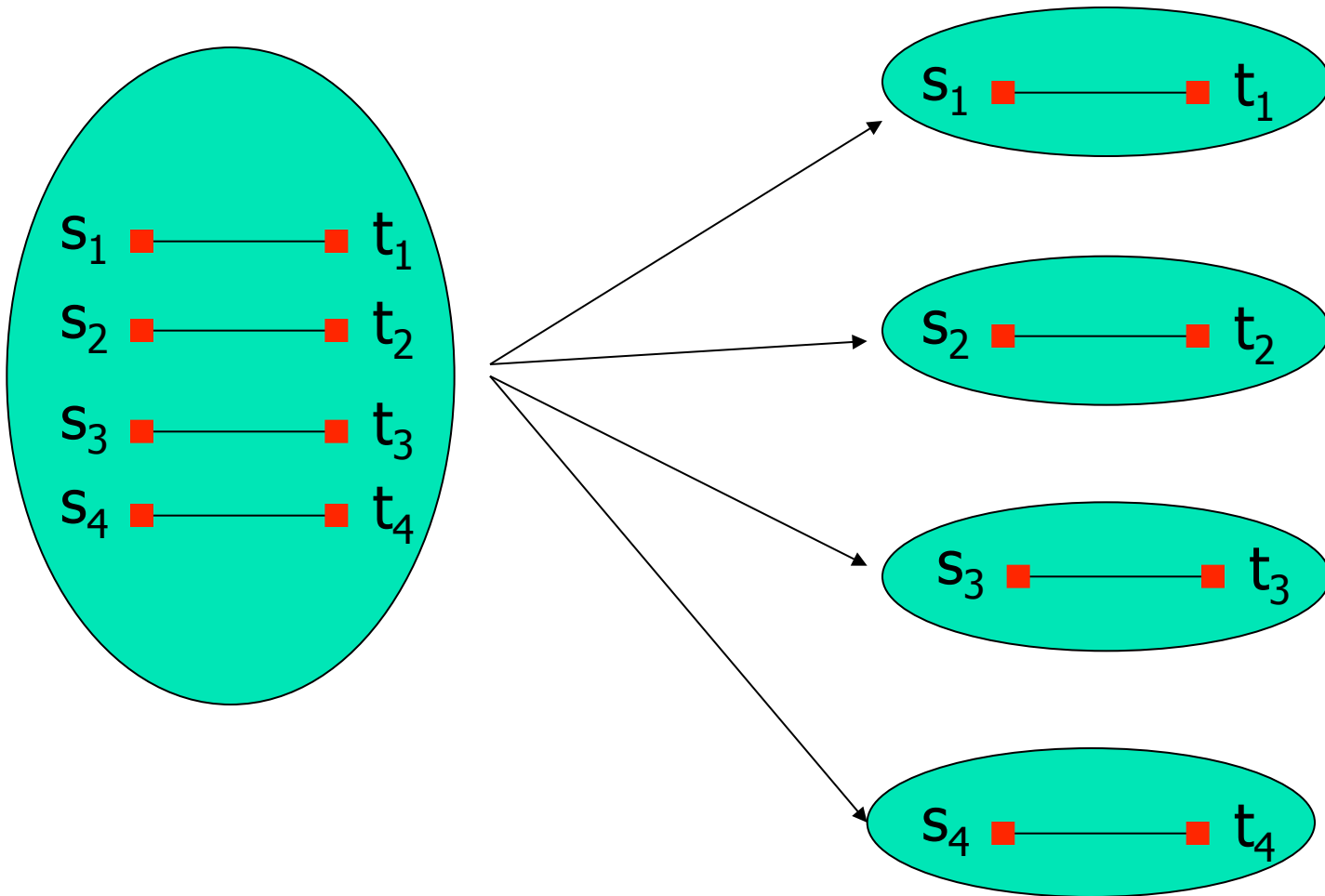


$X_i$  is well-linked in  $G_i$

$$\sum_i |X_i| \geq \text{OPT}/\beta$$

# Example

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# Decomposition

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- $\beta = O(\log k \gamma)$  where  $\gamma =$  worst gap between flow and cut
- $\gamma = O(\log k)$  using [Leighton-Rao' 88]
  - $\gamma = O(1)$  for planar graphs [Klein-Plotkin-Rao' 93]

Decomposition based on LP solution

Recursive algorithm using *separator* algorithms

Need to work with *approximate* and *weighted* notions of well-linked sets

# Decomposition Algorithm

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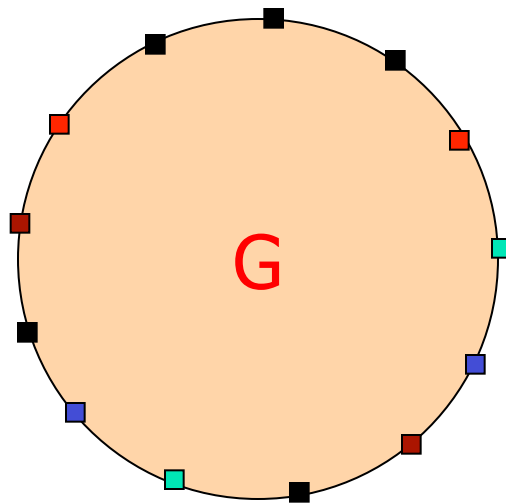
## Weighted version of well-linkedness

- each  $v \in X$  has a weight
  - weight determined by LP solution
  - weight of  $s_i$  and  $t_i$  equal to  $x_i$  the flow in LP soln
- 
- $X$  is *well-linked* implies *no sparse cut*
  - If sparse cut exists, break the graph into two
  - Recurse on each piece
  - Final pieces determine the decomposition

# OS instance

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Planar graph  $G$ , all terminals on single (outer) face



**Okamura-Seymour Theorem:** If all terminals lie on a single face of a planar graph then the cut-condition implies a *half-integral* flow.

# Decomposition into OS instances

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Given instance  $(G, X, M)$  on planar graph  $G$ ,  
algorithm to decompose into OS-type instances  
with only a constant factor loss in value

Contrast to well-linked decomposition that loses a  
 $\log n$  factor

Using OS-decomposition and several other ideas,  
can obtain  $O(1)$  approx using  $c=4$



# Conclusions

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- New approach to disjoint paths and routing problems in undirected graphs
- Interesting connections including new proofs of flow-cut gap results via the “primal” method
- Several open problems
  - Crossbar conjecture: a new question in graph theory
  - Node-disjoint paths in planar graphs -  $O(1)$  approx with  $c = O(1)$ ?
  - Congestion minimization in planar graphs.  $O(1)$  approximation?

# Thanks!

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