

# Max flow min cut in higher dimensions

Art Duval<sup>1</sup>   Caroline Klivans<sup>2</sup>   Jeremy Martin<sup>3</sup>

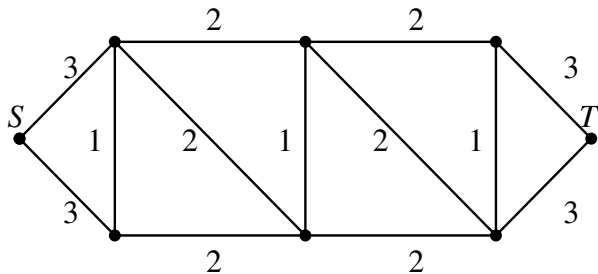
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<sup>3</sup>University of Kansas

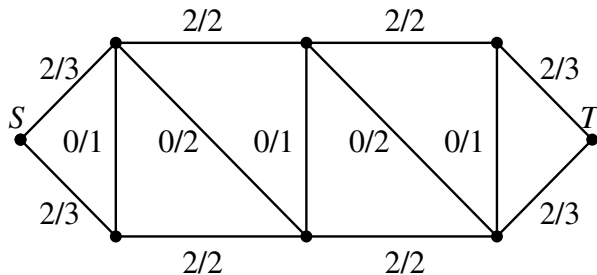
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AMS Special Session on Topological Combinatorics  
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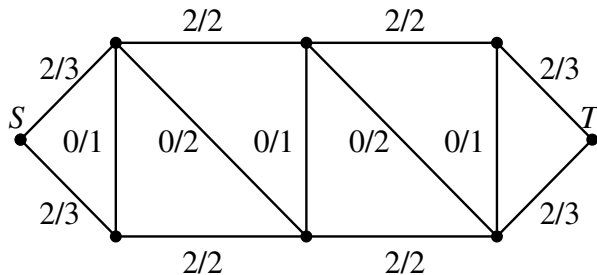
### Definition

**Flow** on  $G$  is an assignment of flow  $x_e$  (non-negative number, and direction) to each edge such that:

- ▶ net flow at each vertex, except  $S$  and  $T$ , is zero; and
- ▶  $|x_e| \leq \kappa_e$ .

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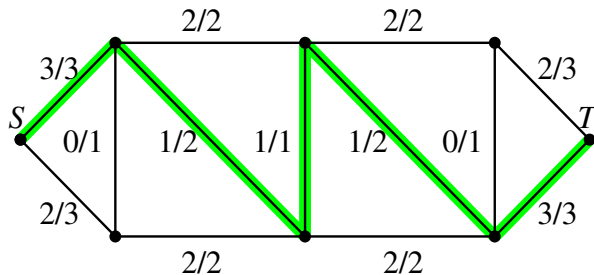
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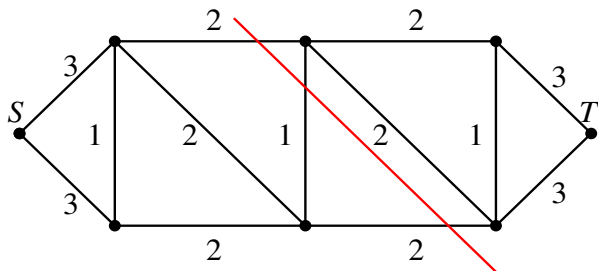
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## Min cut

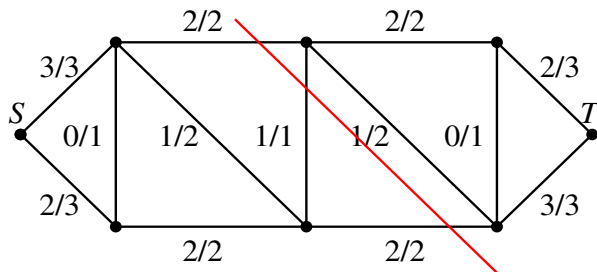


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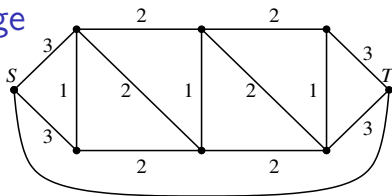
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Clearly,  $\text{value}(\text{flow}) \leq \text{value}(\text{cut})$ , so  $\text{max flow} \leq \text{min cut}$ .

### Theorem (Classic max flow min cut)

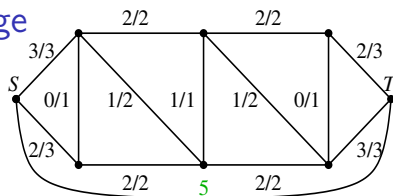
*Max flow = min cut.*

## Add an extra edge





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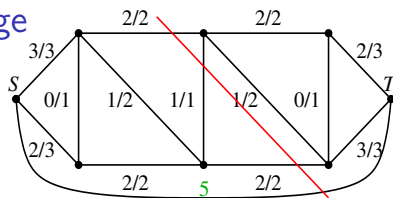
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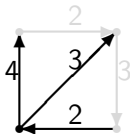
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**Cut** is minimal set of edges, including  $e_0$ , whose removal disconnects  $G$ . **Value** of cut is  $\sum_{e \in \text{cut} \setminus e_0} \kappa_e$ .

## Flows and boundary

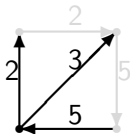


$$\begin{pmatrix} -1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & -1 & 0 \\ 0 & -1 & -1 & 0 & 1 \\ 0 & 0 & 0 & 1 & -1 \end{pmatrix} \begin{pmatrix} 2 \\ 4 \\ 3 \\ 3 \\ 2 \end{pmatrix} = -4 - 3 + 2 = -5$$

Assign orientation to each edge (flow going “backwards” gets negative value)

$$\text{netflow}(v) = \sum_{v=e^+} x_e - \sum_{v=e^-} x_e = \sum_{v \in e} (-1)^{\varepsilon(e,v)} x_e = (\partial x)_v$$

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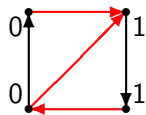
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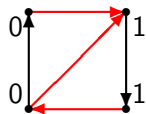
So net flow condition is  $\partial x = 0$ .

## Cuts and coboundary



Assign 1 to every vertex on one side of the cut, 0 to every vertex on the other side.

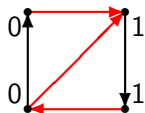
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Edges in cut are those that have both 0 and 1 endpoints.

# Linear programming

Flow is now a linear program

- ▶ Find vector  $x$  (in edge space)
- ▶  $\partial x = 0$  ( $x$  is in flow space)
- ▶  $-\kappa_e \leq x_e \leq \kappa_e$  (can omit  $e_0$ )
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The dual program is (can easily be reworked to say):

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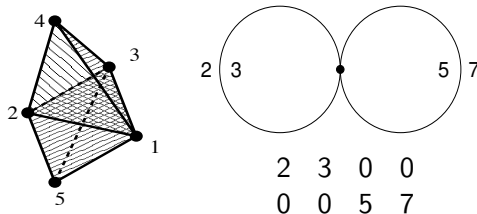
Linear programming says the solutions are equal; with some effort we can show the solution to the dual LP is the min cut problem.

# Cell complexes

## Definition

A **cell complex**  $X$  is a finite CW-complex (i.e., collection of cells of different dimensions), with say  $n$  facets and  $p$  ridges, and a  $p \times n$  **cellular boundary matrix**  $\partial \in \mathbb{Z}^{p \times n}$ .

## Example

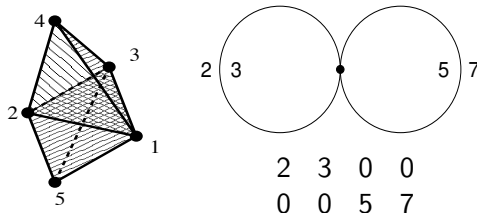


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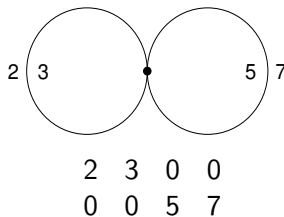
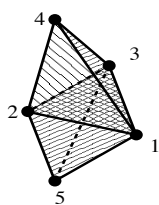
## Remark

Any  $\mathbb{Z}$  matrix can be the boundary matrix of a cell complex.

## Cellular matroids

- ▶ Matroid whose elements are columns of boundary matrix
- ▶ Dependent sets are the supports of the kernel of the boundary matrix

### Example



# Flow space and circuits

## Definition

$i$ -dimensional **flow space** of cell complex  $X$  is

$$\text{Flow}_i(X) = \ker(\partial_i : C_{i-1}(X, \mathbb{R}) \rightarrow C_{i-1}(X, \mathbb{R})).$$

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Circuits are the circuits (minimal dependent sets) of cellular matroid.

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## Example

Bipyramid



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Cut space is the rowspace of the boundary matrix; cut space and flow space are orthogonal complements.

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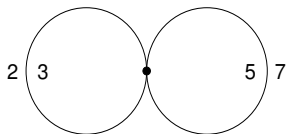
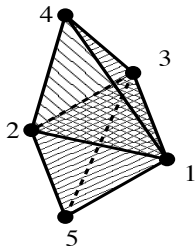
Cocircuits are the cocircuits of cellular matroid

## Topological interpretation of cocircuits

### Remark

Cocircuits are minimal for increasing  $(i - 1)$ -dimensional homology instead of decreasing  $i$ -dimensional homology

### Examples



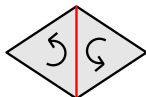
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Find a codimension-1 cycle on the complex, and attach a facet  $f_0$  filling that cycle. We are trying to maximize circulation on that designated facet (around that cycle), while making all circulation balance on each codimension-1 face (**ridge**).



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### Theorem (DKM)

*The max flow equals the value of **some** solution to the dual LP whose support is a cocircuit.*



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Fine print:

- ▶ normalize cut vector by specifying its value is 1 on  $f_0$ , the added filling-in facet
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Questions:

- ▶ Is there an analogue to Ford-Fulkerson? That is, a combinatorial algorithm that would construct the “min cut”, without relying on linear programming?
- ▶ What happens when we restrict to integers?