

C&O 355
Mathematical Programming
Fall 2010
Lecture 21

[N. Harvey](#)

Topics

- Max Weight Spanning Tree Problem
- Spanning Tree Polytope
- Separation Oracle using Min s-t Cuts

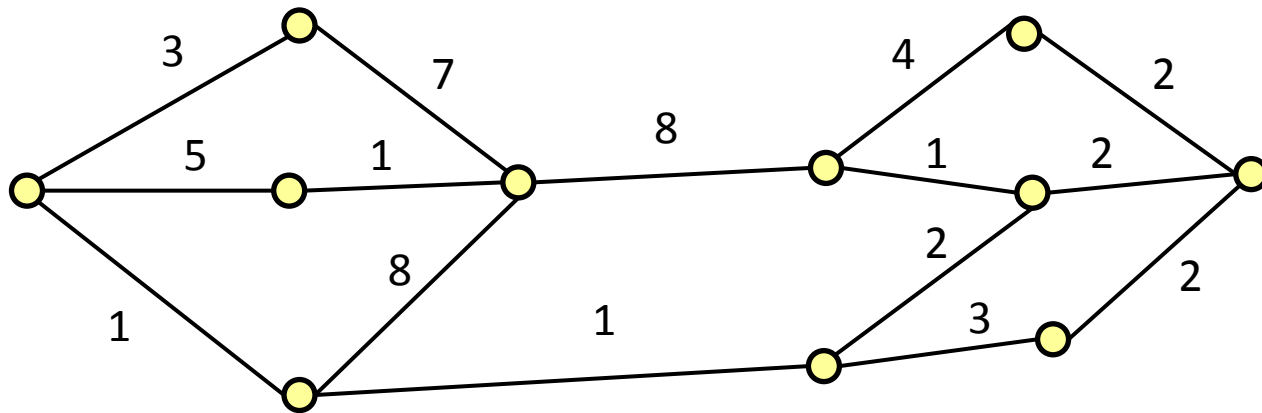


Warning!

The point of this lecture is to do things in an unnecessarily complicated way.

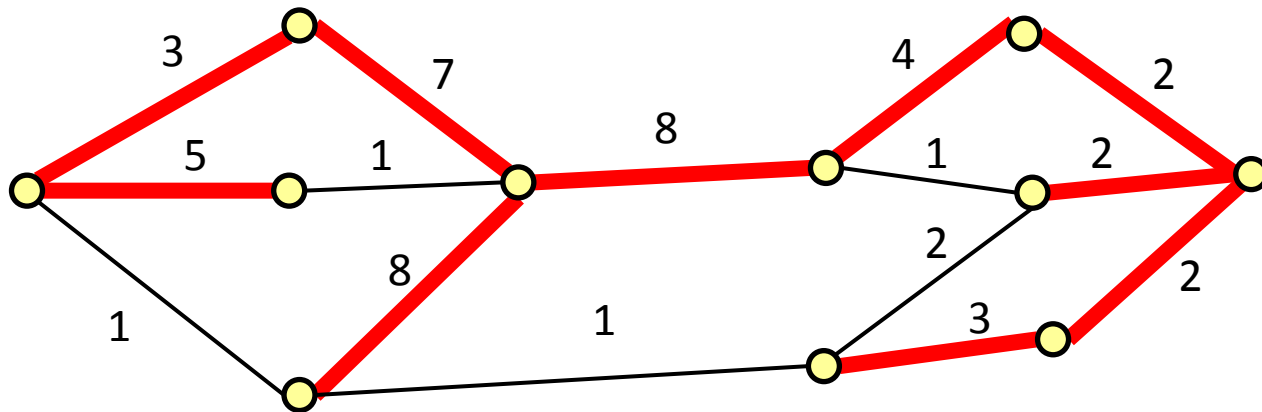
Spanning Tree

- Let $G = (V, E)$ be a connected graph, $n = |V|$, $m = |E|$
- Edges are weighted: $w_e \in \mathbb{R}$ for every $e \in E$
- **Def:** A set $T \subseteq E$ is a **spanning tree** if (these are equivalent)
 - $|T| = n - 1$ and T is acyclic
 - T is a maximal acyclic subgraph
 - T is a minimal connected, spanning subgraph



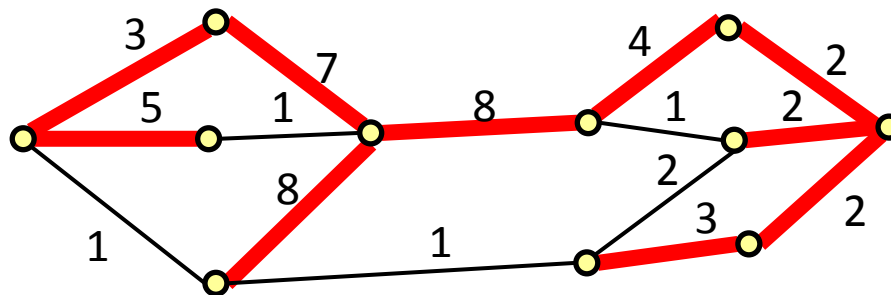
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- **Def:** $T \subseteq E$ is a **max weight spanning tree** if it maximizes $\sum_{e \in T} w_e$ over all spanning trees.



A Simple Properties of Trees

- For any $C \subseteq E$, let $\kappa(C) = \#$ connected components in (V, C)
- **Examples:** $\kappa(E)=1$ and $\kappa(\emptyset)=n$
- **Claim:** Suppose T is a spanning tree.
For every $C \subseteq E$, $|T \cap C| \leq n - \kappa(C)$.
- **Proof:** Let the connected components of (V, C) be $(V_1, C_1), (V_2, C_2), \dots$
- So $V = \bigcup_i V_i$ and $C = \bigcup_i C_i$.
- Since $T \cap C_i$ is acyclic, $|T \cap C_i| \leq |V_i| - 1$.
- So $|T \cap C| = \sum_{i=1}^{\kappa(C)} |T \cap C_i| \leq \sum_{i=1}^{\kappa(C)} (|V_i| - 1) = n - \kappa(C)$. ■

Characteristic Vectors

- **Notation:** We consider vectors x assigning real numbers to the edges in E . We write this as $x \in \mathbb{R}^E$.
- **Notation:** For $C \subseteq E$, let $x(C) = \sum_{e \in C} x_e$.
- **Examples:**
 - the edge weights are $w \in \mathbb{R}^E$
 - For $T \subseteq E$, the **characteristic vector** of T is $x \in \mathbb{R}^E$ where

$$x_e = \begin{cases} 1 & (\text{if } e \in T) \\ 0 & (\text{otherwise}) \end{cases}$$

For any $C \subseteq E$, $x(C) = |T \cap C| \leq n - \kappa(C)$.

This is a **linear inequality** in x : $\sum_{e \in C} x_e \leq n - \kappa(C)$

Spanning Tree Polytope

- Since we know all these linear inequalities, why not assemble them into a polyhedron?

- Let $P_{ST} = \left\{ \begin{array}{l} x(E) = n - 1 \\ x(C) \leq n - \kappa(C) \quad \forall C \subseteq E \\ x \geq 0 \end{array} \right\}$

- **Note:**

- P_{ST} is a polyhedron, because $x(E)$ and $x(C)$ are linear functions of x
- P_{ST} is a polytope, because $x_e = x(\{e\}) \leq 1$, so P_{ST} is bounded
- If x is the characteristic vector of a spanning tree, then $x \in P_{ST}$

The Main Theorems

- **Theorem 1:** The LP $\max \{ w^T x : x \in P_{ST} \}$ can be solved in polynomial time.
(In fact, we'll do this by the ellipsoid method!)
- **Theorem 2:** [Edmonds '71]
The extreme points of P_{ST} are precisely the characteristic vectors of spanning trees of G .
- **Corollary:** A max weight spanning tree can be found in polynomial time.
- **Proof:** Solve the LP and find an extreme point x .
 x is the characteristic vector of a tree T of weight $w^T x$.
Since $w^T x \geq w^T y$ for any other extreme point y ,
it follows that T is a max weight spanning tree. ■

“Polynomial Time”

- There are many algorithms for finding maximum weight spanning trees

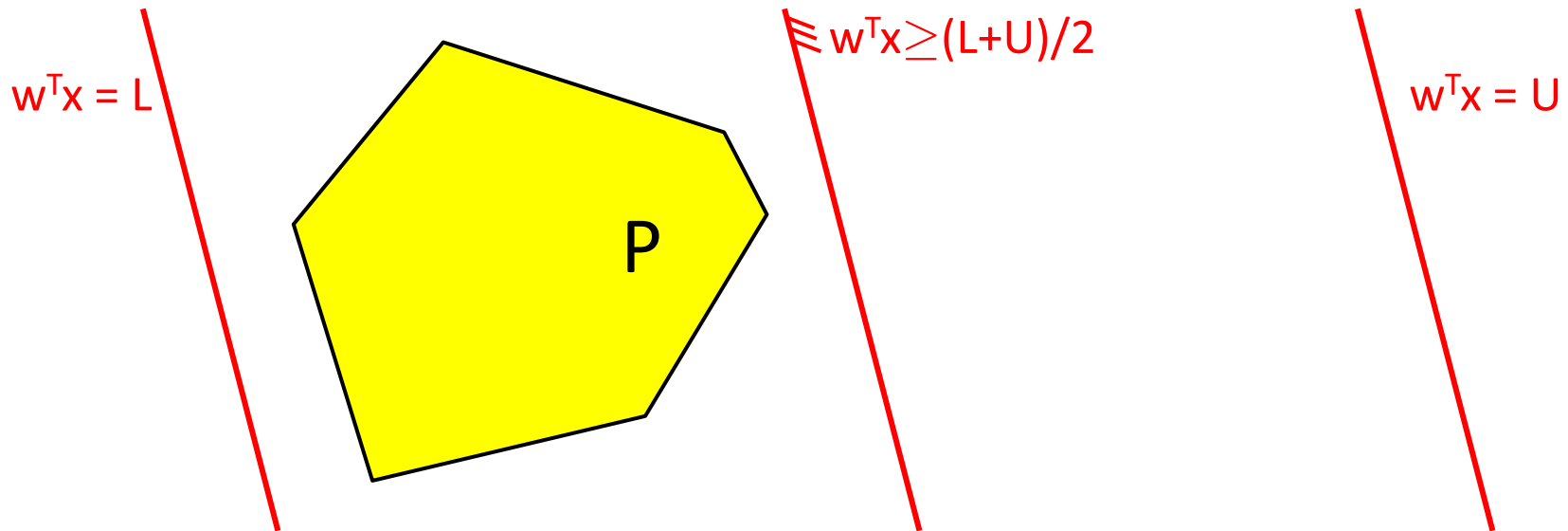
Name	Running Time
Prim's Algorithm	$O(m \log n)$
Kruskal's Algorithm	$O(m \log n)$
...with fancy data structures	$O(m + n \log n)$
...even fancier data structures	$O(m \alpha(m, n))$
Karger-Klein-Tarjan Algorithm	$O(m)$ (randomized)
Pettie-Ramachandaran Algorithm	Optimal deterministic alg, but runtime is unknown

- Our algorithm has running time something like $O(m^{12})$
- Hopelessly impractical! But illustrates important ideas.

Ellipsoid Method for Solving LPs

(from Lecture 8)

- Ellipsoid method can **find a feasible point** in P
i.e., it can **solve a system of inequalities**
- But we want to **optimize**, i.e., solve $\max \{ w^T x : x \in P \}$
- **One approach:** Binary search for optimal value
 - Suppose we know optimal value is in interval $[L, U]$
 - Add a new constraint $w^T x \geq (L+U)/2$
 - If LP still feasible, replace L with $(L+U)/2$ and repeat
 - If LP not feasible, replace U with $(L+U)/2$ and repeat



Applying the Ellipsoid Method

- By binary search, we need to decide feasibility of

$$P_{ST} = \left\{ \begin{array}{l} w^T x \geq W \\ x(E) = n - 1 \\ x(C) \leq n - \kappa(C) \quad \forall C \subseteq E \\ x \geq 0 \end{array} \right\}$$

- **Main obstacle:** Huge number of constraints! (2^m)
- **Recall:** Ellipsoid method works for **any convex set** P , as long as you can give a separation oracle.

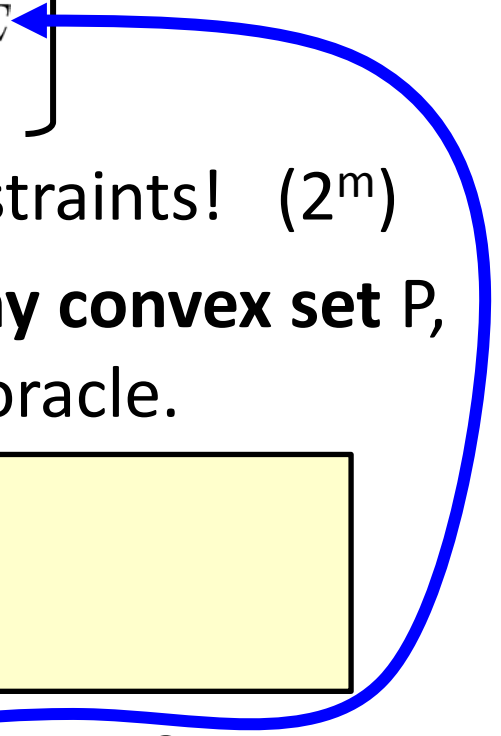
Separation Oracle

Is $z \in P$?

If not, find a vector a s.t. $a^T x < a^T z \quad \forall x \in P$

Applying the Ellipsoid Method

- By binary search, we need to decide feasibility of

$$P_{ST} = \left\{ \begin{array}{l} w^T x \geq W \\ x(E) = n - 1 \\ x(C) \leq n - \kappa(C) \quad \forall C \subseteq E \\ x \geq 0 \end{array} \right\}$$


- **Main obstacle:** Huge number of constraints! (2^m)
- **Recall:** Ellipsoid method works for **any convex set** P , as long as you can give a separation oracle.

Separation Oracle

Is $z \in P$?

If not, find a **violated constraint**.

- How quickly can we test **these** constraints?
- **We'll show:** This can be done in time polynomial in n .

Ellipsoid method inside Ellipsoid method

Everything runs in polynomial time!

Minimum Spanning Tree Problem

Solve by Ellipsoid Method
Separation oracle uses...

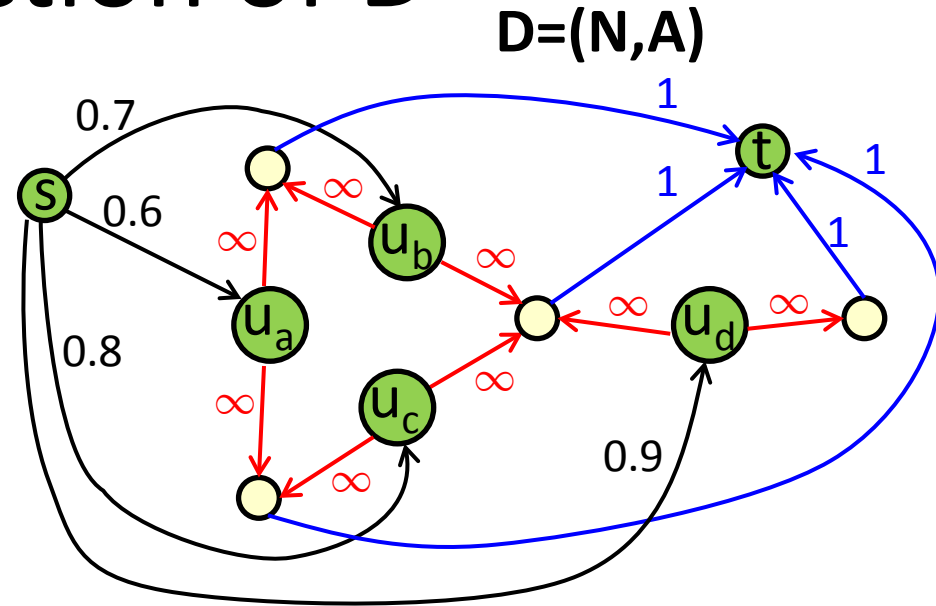
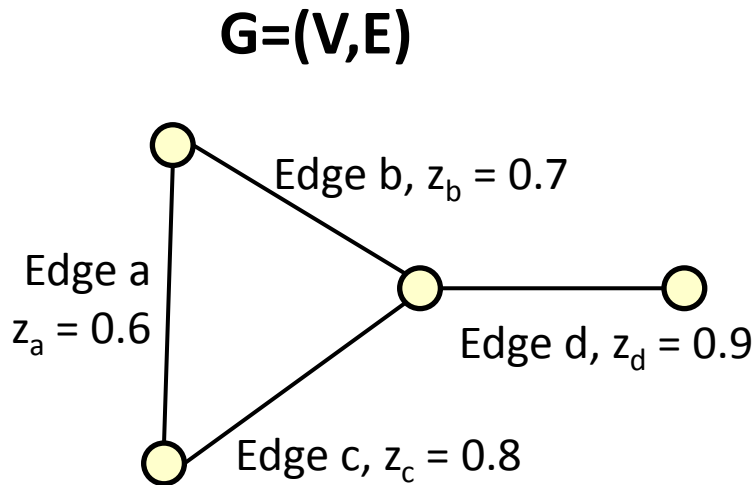
Minimum S-T Cut Problem

Solve by Ellipsoid Method!

Separation Oracle: Game Plan

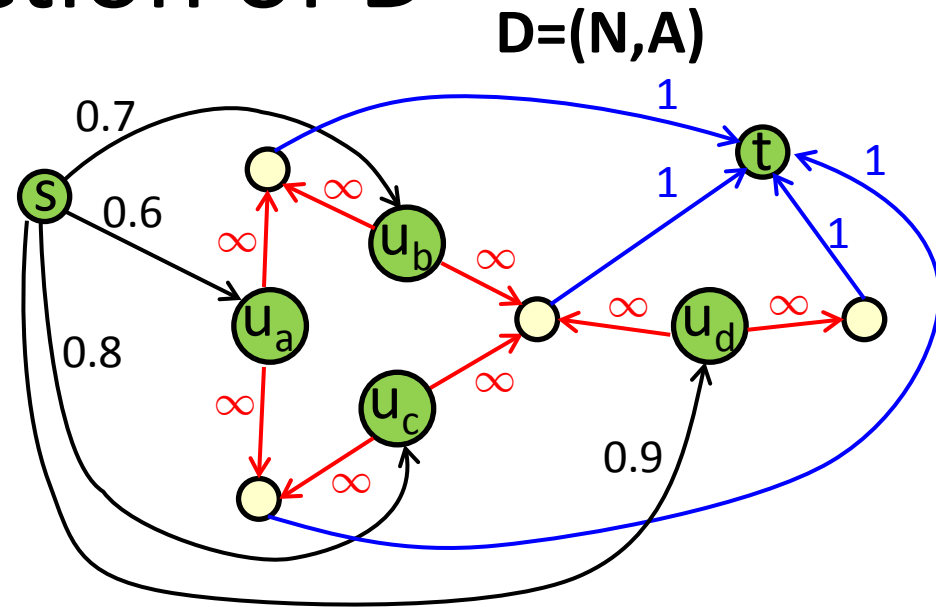
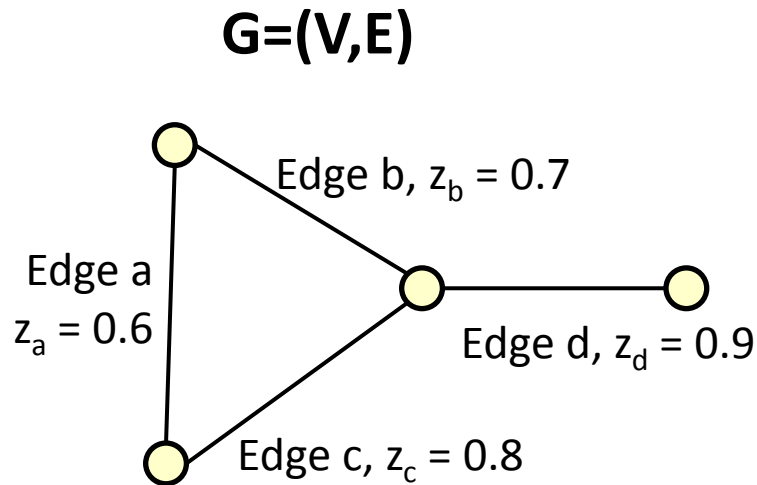
- We have graph $G=(V,E)$ and a point $z \in \mathbb{R}^E$
- We construct digraph $D=(N,A)$ with capacities $c \in \mathbb{R}^A$
- If s-t min cut in D is:
 - **Small:** this shows that z violates a constraint of P
 - **Large:** this shows that z is feasible for P

Construction of D



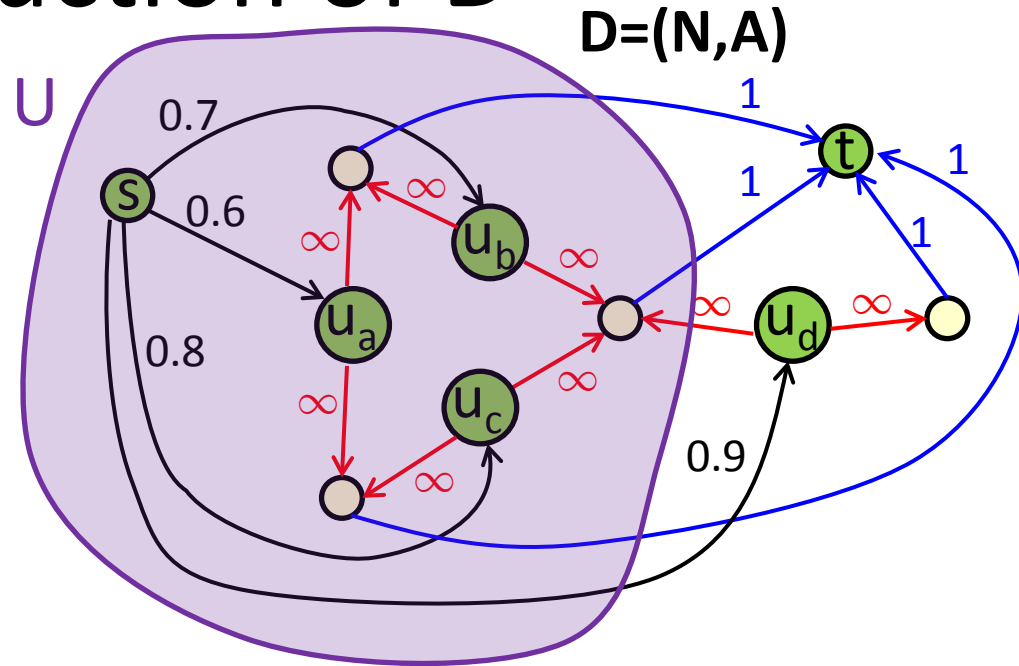
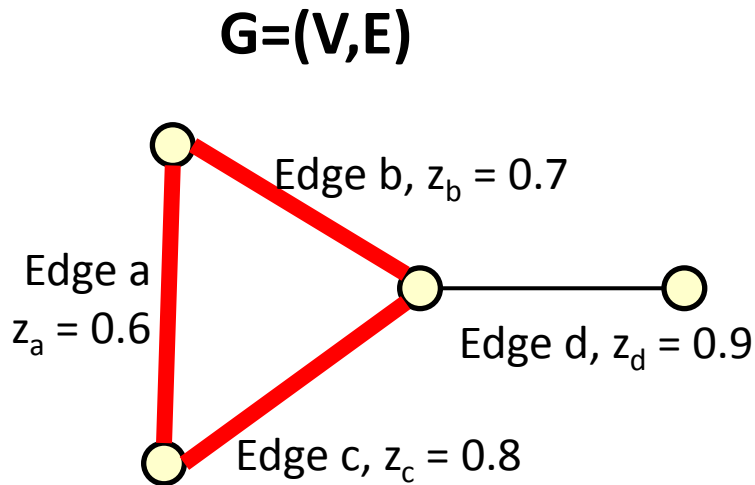
- **Nodes of D:** $N = V \cup \{s,t\} \cup \{u_e : e \in E\}$
- **Arcs of D:**
 - Arc (s,u_e) of capacity z_e for every edge $e \in E$
 - Arc (v,t) of capacity 1 for every node $v \in V$
 - Infinite capacity arcs $(u_{\{v,w\}}, v)$ and $(u_{\{v,w\}}, w)$ for all $\{v,w\} \in E$

Construction of D



- **Lemma:** z is feasible \Leftrightarrow every s-t cut has capacity $\geq n$ (except for the cut with only black edges).

Construction of D



- **Lemma:** z is feasible \Leftrightarrow every s - t cut has capacity $\geq n$ (except for the cut with only black edges).
- **Example:**
 - $z(\text{C}) = 2.1 > n - \kappa(\text{C}) = 2$, so z is infeasible.
 - The s - t cut $\delta^+(U)$ in D has capacity $3.9 < n = 4$.

Separation Oracle Summary

- **Input:** $G=(V,E)$ and $z \in \mathbb{R}^E$
- Construct the graph $D=(N,A)$ and arc capacities
- For each $v \in V$
 - Temporarily add an infinity capacity arc (s,v)
 - Compute the s - t min cut value q
(by the Ellipsoid Method)
 - If $q < n$
 - We obtain a set $S \subseteq V$ s.t. $z(E[S]) > |S| - 1$
 - **Halt:** return this violated constraint
 - Remove the temporary arc
- End for
- **Halt:** z is feasible

Summary

- Some combinatorial objects are described by LPs of exponential size
- Even if an LP has exponential size, the ellipsoid method might be able to solve it “efficiently”, if a separation oracle can be designed
- The separation oracle might use ellipsoid method too
- Ellipsoid method gives impractical algorithms, but these can be a “proof of concept” for realistic algorithms