

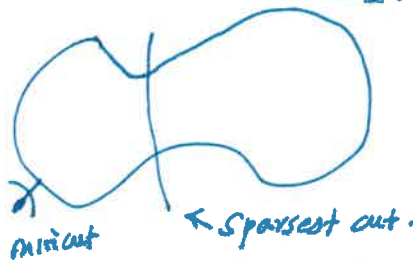
# Lecture 6: Sparsest Cut Problem

Previous lectures: Max Coverage & Set Cover. (Algos, Integrality gaps, Hardness)  
Max Cut (Algos, Int gaps).

Today: third problem, Sparsest cut: see Algos, Int gaps, Hardness  
↑ and next 2-3 lectures. ↑  
LP Spectral SDPs.

Def: Graph  $G=(V,E)$  edges have weights "capacities"  $\{c_e\}_{e \in E}$

Find a bottleneck in the graph "  $\min_{S \subseteq V} \frac{cap(S, \bar{S})}{\min\{|S|, |\bar{S}|\}}$  "  $\sum_{e \in E} c_e$   
edges one end in  $S$   
another in  $\bar{S}$ .



$$\frac{\text{sparsity of cut}}{(S, \bar{S})} = \frac{cap(S, \bar{S})}{|S| \cdot |\bar{S}|}$$

← product of two sides.

Want to find min sparsity cut.  $\Phi(G) = \min_S \Phi(S)$ .

Fact: sparsity of cut  $\in$  edge expansion of cut  $\times \left[ \frac{2}{n}, \frac{1}{n} \right]$

PF: for any cut  $(S, \bar{S})$   $|S| \cdot |\bar{S}| = |S| \cdot (n - |S|) = |S| \times (\text{some factor between } \frac{1}{2} \text{ and } n)$ .  
↑ say  $|S| \leq |\bar{S}|$  □

⇒ sparsity of graph captures edge expansion of graph to within constant factors!

Want to compute the (uniform) sparsest cut

$$\min_S \Phi(S) = \min_S \frac{cap(S, \bar{S})}{|S| \cdot |\bar{S}|}$$

• [Shahrokhi Matulek] NP hard.

So how to approximate it?

[Leighton Rao 88]  $O(\lg n)$  approximation,  $\Omega(\lg n)$  int gap via LPs.

[Arora Rao Vazirani 03]  $O(\sqrt{\lg n})$  apx. via SDPs.

[Naor Young 13]  $\Omega(\lg n)$  int gap.

Several interesting ideas here...

Start from the beginning.

• Leighton Rao LP relaxation.

Set edge lengths  $y_e$  for each edge  $y_e \in [0, 1]$ . as in max cut case.

"Ideal setting"  $y_e \in \{0, 1\}$  captures if edge is uncut or cut.

triangle inequality:  $y_{ij} \leq y_{ik} + y_{kj} \quad \forall k, i, j$ .

if  $y_{ij} = 1$   
 $\Rightarrow y_{ik} \text{ or } y_{kj} = 1$   
in the  $\{0, 1\}$  case

Now: objective function is  $\min \frac{\sum_{i,j \in E} c_{ij} y_{ij}}{\sum_{i,j \in V} y_{ij}}$

subject to  $y$  is a metric (so  $y_{ij} \geq 0, y_{ij} \leq y_{ik} + y_{kj}$ )

Fact: if  $y_{ij} \in \{0, 1\}$  then this is the sparsest cut problem.

But now relaxation so  $LP \text{ obj} \leq OPT$ .

Btw: how to handle ratio objective?

Easy:  $\min \frac{a^T x}{b^T x} \text{ st } x \in K, x \geq 0$   $\Leftrightarrow \min a^T x \text{ st } b^T x = 1, x \in K, x \geq 0$ .

(at least when  $b \geq 0$ ).

Theorem: Given any solution  $y$  to the LP (also called "metric relaxation")

$\exists$  a ~~set~~ cut with sparsity  $\leq O(\log n)$ . LP value.

Thm:  $\exists$  instances where integrality gap of LP =  $\Omega(\log n)$

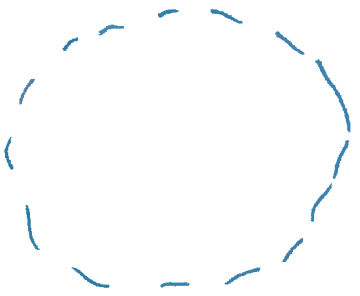
so. all cuts have sparsity  $\geq \Omega(\log n)$ . LP value.

Easiness Proof:

given metric  $\{y_{ij}\}_{ij}$  st.  $\sum_{ij} y_{ij} = 1$

have LP =  $\sum_{ij} C_{ij} y_{ij}$ .

( $\Rightarrow$  avg distance =  $\frac{1}{\binom{n}{2}}$ )



Want to find a good sparsity cut.

Algorithm:

(1) Find a "Low Diameter Decomposition" with  $R = \frac{1}{4n^2}$  radius

that cuts "few" edges.  $\leq O(\log n) \cdot \frac{LP}{R}$

$\uparrow$  smaller than avg distance.

(2) if each piece has  $\leq \frac{n}{3}$  nodes then collect pieces into

two parts  $L, \bar{L}$  each having  $\geq \frac{n}{3}$  nodes greedily.

output  $L$ .

$\Rightarrow$  sparsity  $\leq \frac{O(\log n) \cdot \frac{LP_{value}}{R}}{\binom{n}{3} \binom{2n}{3}} = O(\log n) \cdot LP$  😊

(3) if  $\exists$  piece that has  $> \frac{n}{3}$  nodes ("dense" piece), all these nodes at dist  $\leq 2R$  from each other

then contract this set of nodes into "root"

~~to~~ Lay out vertices in order of distance from this root  $\uparrow$  according to  $y$ .

Find best "level cut".

Claim: cut  $\leq O(LP)$  !

So two things to show:

- ① that given any metric  $y$ , can do this Low Diam Decomp.
- ② and that dense case is easy.

Low Diam Decomposition

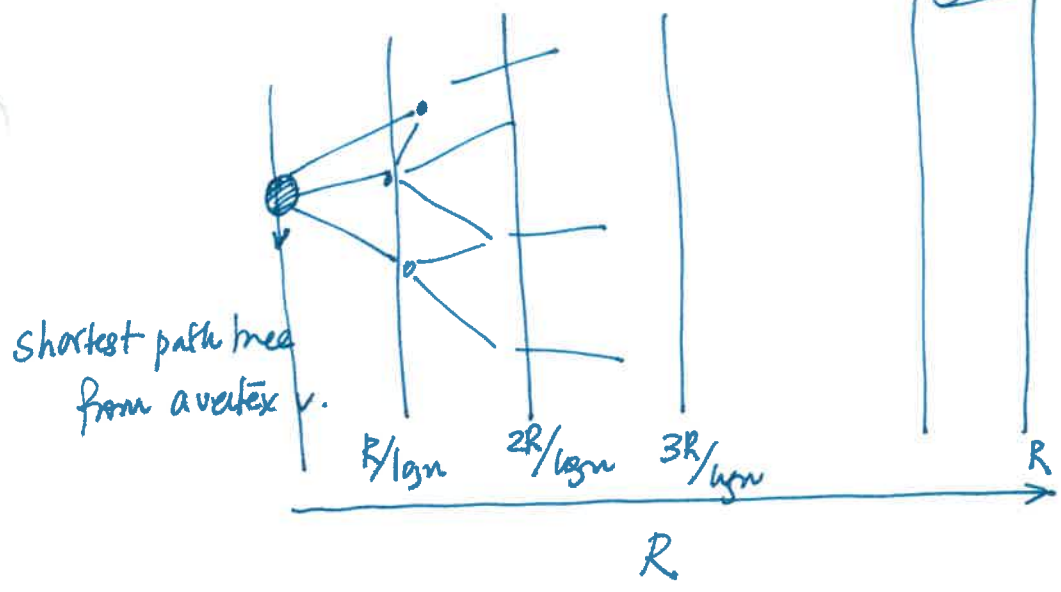
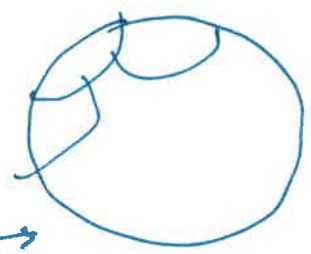
Many ways of doing this

Let's see a "region growing" argument.

~~Prove~~

Imagine each edge has cross section = capacity  $C_{ij}$   
length =  $y_{ij}$

$\Rightarrow$  volume =  $C_{ij} y_{ij}$



Let  $V_i$  = volume in layers ~~between~~ between  $v$  and  $\frac{iR}{\ln n}$

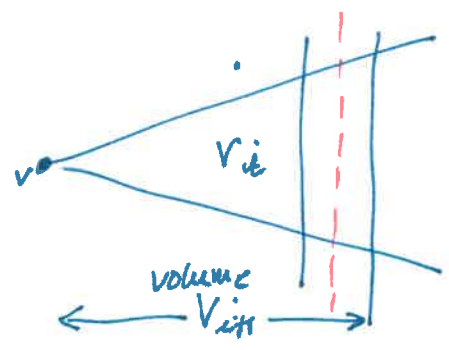
Assume  $V_0 = \frac{LP}{n}$ . Original volume.

Look for layer  $i$  s.t.  $V_{i+1} \leq 2V_i$

Fact:  $\exists$  such a layer, else each time volume doubles (and more!)

Starts at  $\frac{LP}{n}$ , so after  $\ln n$  levels, must be  $> LP$ , impossible!

So  $\exists$  layer  $i$  st.  $V_{i+1} \leq V_i \times 2$ .



~~Find a random cut in~~

Find the cheapest cut in  $[\frac{iR}{\log n}, \frac{(i+1)R}{\log n}]$

no more expensive than random ~~cut~~ level cut

But each edge in that range cut w.p.

$$\frac{\text{length of edge within that range}}{(R/\log n)}$$

$$\Rightarrow E[\text{cut value}] = \sum_{\substack{ij \\ \text{in that} \\ \text{range}}} c_{ij} \cdot y_{ij} \frac{(\text{in that range})}{(R/\log n)}$$

$$\leq \frac{V_{i+1} - V_i}{R/\log n}$$

$$\leq \frac{V_{i+1}}{R} \log n$$

$$\leq \frac{2V_i}{R} \log n$$

$V_i = \text{~~fake~~ volume that is definitely cut away.$

$\Rightarrow$  we are paying at most  $\frac{2V_i}{R} \log n$

but cutting away  $V_i$  volume. (so we can "charge" to this removed volume).

Eventually when process stops, total edges cut

$$\leq \frac{2 \cdot (\text{total volume cut away})}{R} \cdot \log n \leq \frac{2(LP + n \frac{LP}{n})}{R} \cdot \log n = 4LP \cdot \log n / R$$

fake volume added to roots.

This proves the Low diam decomp property

cut graph into pieces of radius  $\leq R$

and total edges cut  $\leq \frac{LP}{R} \cdot O(\log n)$ .

Notes: Can give randomized versions too

see [Bartal, Calinescu Karloff Rabani, Miller Peng Xu]

this is a "local" version, "act and charge to the left, to the volume removed".

Called "region growing", used widely in algorithm design for approx algo's, distributed computation, etc.

Originally due to [Awerbach Peleg], [Linial Saks].

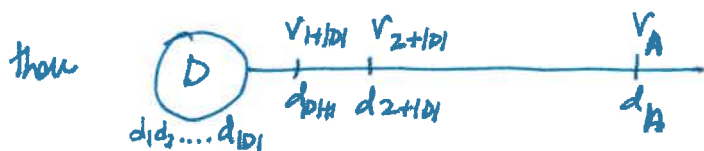
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Finally: Handle dense cluster case.

Sps.  $\exists$  set  $D \subseteq V$  of size  $\geq n/3$ , such that all distances in  $D$  are  $\leq 2R$ .

$\uparrow$  "dense"

$$\leq \frac{2}{4n^2} \text{ (avg)}$$



• Arrange vertices according to distance from  $D$ .  $d_1 \leq d_2 \leq \dots \leq d_n$ .

$$d_v = \min_{u \in D} y_{uv} = \text{distance of } v \text{ to closest vertex in } D.$$

• Consider sets  $S_i = \{v_1, v_2, \dots, v_i\}$ .

Output set  $S_i$  with least ~~dens~~ sparsity.  $\phi(S_i) = \frac{e(S_i, \bar{S}_i)}{|S_i| \cdot |\bar{S}_i|}$

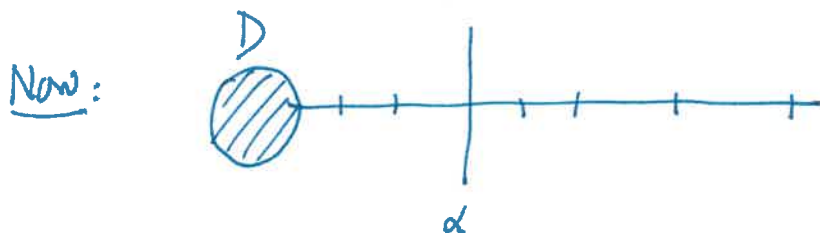
• Claim: if  $|D| \geq n/3$ , there is a set  $S_i$  with sparsity  $\leq O(LP)$ .

Proof:

First consider  $1 = \sum_{ij} y_{ij} \leq \sum_{ij} (d_i + d_j + 2R) \leftarrow \Delta\text{-ineq and } D \text{ has small diam.}$

$$= n \cdot \sum_i d_i + n \cdot \sum_i d_j + \underbrace{n^2 \cdot 2R}_{= \frac{1}{2} \text{ by our choice}}$$

$$\Rightarrow \sum_i d_i \geq \frac{1}{4n}$$



Pick a radius  $\alpha$ , look at ball  $B_\alpha = \{\text{all vertices at distance } \leq \alpha \text{ from } D\}$   
 $\alpha \in [0, 1]$   
 $\bar{B}_\alpha = \text{complement.}$

$$LP = \int_{\alpha=0}^1 \text{cap}(B_\alpha, \bar{B}_\alpha) d\alpha \geq \int \text{Alg.} \cdot |B_\alpha| \cdot |\bar{B}_\alpha| d\alpha \leftarrow \text{We chose best sparsity set}$$

$$\geq \text{Alg.} \cdot \frac{n}{3} \cdot \int |B_\alpha| d\alpha$$

$$= \sum_v d_v$$

$$\geq \text{Alg.} \cdot \frac{n}{3} \cdot \frac{1}{4n} = \frac{\text{Alg.}}{12}$$

$$\Rightarrow \text{Alg.} \leq 12 LP.$$



# How good is the LP?

Sadly, integrality gap =  $\Omega(\log n)$ .

Examples come from the other important class of counter examples  
constant degree expander graphs.

Expander Graphs:  $G=(V,E)$  expander if  $\forall$  sets  $S$  that are "small enough",  
the neighborhood of  $S$  is "large".

Every set "expands"  
small.

Formally: an  $d$ -edge expander  $G$  is graph s.t.  
 $\forall |S| \leq |V|/2, |E(S, \bar{S})| \geq \alpha \cdot d \cdot |S|$ .

let's make it  $d$ -regular.

technically we will give graph families

Fact/Thm:  $d$ -expander graphs exist ~~with~~  $\alpha = \text{constant}$ .

when we say graph is expander without specifying  $\alpha$ , usually mean  $\alpha = \Theta(1)$  universal constant.

Examples: ~~random~~ take constant # of random matchings.

But even deterministic constructions are known.

• Say vertex set is  $\mathbb{F}_p$ ,  $p$  prime

each  $i$  connected to  $i+1, i-1, 1/i$

• Cayley graphs of certain groups [Margulis, Gabber-Galil]

• And zig-zag product (maybe discuss later).



Thm: Integrality gap of metric relaxation =  $\Omega(\log n)$ .

Pf: take constant-degree expander with  $d = \Theta(1)$ .  
 $d = 3$  (say). expansion.

Set  $c_e = 1 \forall e \in E$

Fact:  $\forall$  vertex in expander, at most  $3^k$  vertices at hop distance  $\leq k$ .

$\Rightarrow$  at least  $n/2$  vertices at distance  $\Omega(\log n)$  from any vertex.

Let  $y_{ij} = \frac{c_e}{n^2 \log n}$ . And define  $y_{ij}$  = shortest path distance b/w  $i \& j$  according to those edge lengths.  
↑  
"length of edge"

$\Rightarrow \sum_i y_{ij} \approx 1$  for suitable constant, since  $\Omega(n^2)$  vertex pairs at hop-distance  $\Omega(\log n)$  from each other.

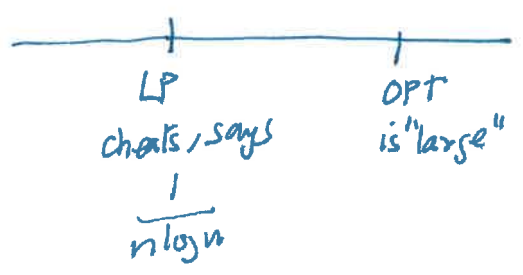
Now: LP value =  $\sum_{e \in E} c_e y_e = O\left(\frac{1}{n^2 \log n}\right) \cdot O(n) = O\left(\frac{1}{n \log n}\right)$ .

But expansion of all cuts =  $\Omega(1)$

$\Rightarrow$  sparsity of all cuts =  $\Omega\left(\frac{1}{n}\right)$ .

sparsity  $\approx$  expansion  $\times \Theta\left(\frac{1}{n}\right)$ .  
from page 1 of these notes

$\Rightarrow$  gap of  $\Omega(\log n)$  between LP and OPT.



Next Lecture(s): Make connection to

- metric embeddings and
- eigenvalues/vectors