## Lecture 16

Friday, May 12, 2017 9:42 AM

## The Sparsest Cut Problem

Input: Undirected G=(V,E)
· Costs Ce on edges

Output: SEV

Objective: Minimize (25)

RATIO

## LP-Rehaation

As last time, we have "distance" variables where in the "integral" soln, d(u,v) = 1 if u,v are in sep. sides of the cut & 0 %

: OPT >

S d(u, σ) u, ν ενχν note this is d satisfies Δ-ineq precisely ISI. ISI if d(u, ν) = 1 15... -3 ncl= Y pairs [ EUNT ? NS = 1

Ratio ~> Constraint

min  $\sum c(e) d(e)$ OPT >

. Zdur = 1

a satisfies metric constraints.

ROUNDING - Try 1 - Solve LP - Pick a node & Carbitanily for now; we will choose appropriately later) - Let R := man d(s,r) - Sample P € (O,R) u.a.r. - S:= {u | d(s,u) < 9} As last time, IE [c(85)] = f. LP But this time we "also have a denominator". Ideally, we would're liked to analyze  $\mathbb{E}\left[\frac{c(\delta s)}{|s|\sqrt{|s|}}\right]$ This is a rother unwieldy object. Instead let's look HE [ 181. 131] ... \*\* Note these two NOT the same! Why is (\* interesting?

: if we show  $\frac{\mathbb{E}\left[c(8S)\right]}{\mathbb{E}\left[c(8S)\right]} \leq \alpha \cdot \mathbb{E}$ 

F Sich 13 II
E [181.13]
then, by moving Jerms & LINEARITY OF EXPECTATION,
EXPECTATION,
E (88) - 4.LP. ISI.131 ≤ 0
The of the possible cuts that our algorithm returns satisfies
returns satisfies
c(25*) ≤ d.LP. 15*1
⇒ sparsity (s*) ≤ <. LP
Why would we be able to get our hands on 5th?
Why would we be able to get our hands on 5th?  Only a O(m) different e's are "interesting"  the can deterministically enumerate.
It suffices to upper bound # [c(85)]
#[ISI.131]
, G
The numerator we already known to be LP/R  The denominator is again a bit problematic  pince
The denominator is again a bit problematic
151 & 151 are not independent (in fit they so

on |S|, ie, |S| 2 1

· · · For now we use a trivial LOWER BOUND

: IE[ISI.ISI] > IE[IS] -....#/

$$\mathbb{E}\left[\overline{|S|}\right] = \frac{1}{R} \sum_{s} (\# \text{ of vertices } u \text{ st.d}(s,u) > g)$$

this really should be an integration.

$$= \frac{1}{R} \sum_{s=1}^{R} \frac{1}{2} \sum_{s=1}^{R} \frac{$$

$$\sum_{u,v} J(u,v) = 1$$

$$\frac{1}{1-\left[\frac{1}{2}\left(\frac{1}{2}\right)\right]} \leq \frac{1}{1-\left[\frac{1}{2}\right]} \leq \frac{1}{1-\left[$$

$$\frac{L}{|E[|s|\cdot|s|]} \le n \cdot Lt$$

An n-factor algorithm ....

Try 2

- The algorithm started cutting around s, a Singleton, & i 151>1 was the only thing we could use.

- Modifation:

- let T ⊆ V s.t. [T] ≈ @(n); say 17/2 n/3

-  $R := \max_{r} d(T, r)$ ;  $S \in_{R} (o, R)$ 

- S:= {u | d(T, u) < 9)

Once again: -- IE [cost (OS)] = 1.LP

- ISI > ITI > n/3 ... T C S

- What about IE [ISI]?

As before,

 $\mathbb{E}\left[\overline{S}\right] = \frac{1}{R} \sum_{g} (\# of vs st d(T, v) > p)$   $= \frac{1}{R} \sum_{g} \sum_{u:d(T,1)>0} 1$ 

$$= \frac{1}{R} \underbrace{\frac{1}{3}}_{3} \underbrace{\frac{1}{4nR}}_{1:d(T,n) \ge g}$$

$$= \frac{1}{R} \underbrace{\frac{1}{3}}_{3:g \le d(T,n)}$$

$$= \frac{1}{R} \underbrace{\frac{1}{3}}_{3:g \le d(T,n)}_{1:g \le g \le d(T,n)}$$

$$= \underbrace{\frac{1}{N}}_{3:g \le d(T,n)}_{1:g \le g \le g \le d(T,n)}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g}_{1:g \le g \le g \le g \le g \le g}_{1:g \le g \le g}_{1:g \le g \le g \le g}_{1:g \le g \le g \le g}_{1:g \le g \le g}_{1:g \le g \le g}_{1:g \le g \le g \le g}_{$$

⇒ 
$$E[S|-|S|]$$
 ≥  $\Omega(\frac{1}{R})$ 

⇒  $E[S|-|S|]$  =  $O(LP)$ 
 $E[S|-|S|]$