

## Lecture 4: Cheeger's Inequalities cont'd

*In which we finish the proof of Cheeger's inequalities.*

It remains to prove the following statement.

**Lemma 1** *Let  $\mathbf{y} \in \mathbb{R}_{\geq 0}^V$  be a vector with non-negative entries. Then there is a  $0 < t \leq \max_v \{y_v\}$  such that*

$$\phi(\{v : y_v \geq t\}) \leq \sqrt{2R_L(\mathbf{y})}$$

We will provide a probabilistic proof. Without loss of generality (multiplication by a scalar does not affect the Rayleigh quotient of a vector) we may assume that  $\max_v y_v = 1$ . We consider the probabilistic process in which we pick  $t > 0$  in such a way that  $t^2$  is uniformly distributed in  $[0, 1]$  and then define the non-empty set  $S_t := \{v : y_v \geq t\}$ .

We claim that

$$\frac{\mathbb{E} E(S_t, V - S_t)}{\mathbb{E} d|S_t|} \leq \sqrt{2R_L(\mathbf{y})} \quad (1)$$

Notice that Lemma 1 follows from such a claim, because of the following useful fact.

**Fact 2** *Let  $X$  and  $Y$  be random variables such that  $\mathbb{P}[Y > 0] = 1$ . Then*

$$\mathbb{P} \left[ \frac{X}{Y} \leq \frac{\mathbb{E} X}{\mathbb{E} Y} \right] > 0$$

**PROOF:** Call  $r := \frac{\mathbb{E} X}{\mathbb{E} Y}$ . Then, using linearity of expectation, we have  $\mathbb{E} X - rY = 0$ , from which it follows  $\mathbb{P}[X - rY \leq 0] > 0$ , but, whenever  $Y > 0$ , which we assumed to happen with probability 1, the conditions  $X - rY \leq 0$  and  $\frac{X}{Y} \leq r$  are equivalent.  $\square$

It remains to prove (1).

To bound the denominator, we see that

$$\mathbb{E} d|S_t| = d \cdot \sum_{v \in V} \mathbb{P}[v \in S_t] = d \sum_v y_v^2$$

because

$$\mathbb{P}[v \in S_t] = \mathbb{P}[y_v \geq t] = \mathbb{P}[y_v^2 \geq t^2] = y_v^2$$

To bound the numerator, we say that an edge is *cut* by  $S_t$  if one endpoint is in  $S_t$  and another is not. We have

$$\begin{aligned} \mathbb{E} E(S_t, V - S_t) &= \sum_{\{u,v\} \in E} \mathbb{P}[\{u,v\} \text{ is cut}] \\ &= \sum_{\{u,v\} \in E} |y_v^2 - y_u^2| = \sum_{\{u,v\} \in E} |y_v - y_u| \cdot (y_u + y_v) \end{aligned}$$

Applying Cauchy-Schwarz, we have

$$\mathbb{E} E(S_t, V - S_t) \leq \sqrt{\sum_{\{u,v\} \in E} (y_v - y_u)^2} \cdot \sqrt{\sum_{\{u,v\} \in E} (y_v + y_u)^2}$$

and applying Cauchy-Schwarz again (in the form  $(a + b)^2 \leq 2a^2 + 2b^2$ ) we get

$$\sum_{\{u,v\} \in E} (y_v + y_u)^2 \leq \sum_{\{u,v\} \in E} 2y_v + 2y_u^2 = 2d \sum_v y_v^2$$

Putting everything together gives

$$\frac{\mathbb{E} E(S_t, V - S_t)}{\mathbb{E} d|S_t|} \leq \sqrt{2 \frac{\sum_{\{u,v\} \in E} (y_v - y_u)^2}{d \sum_v y_v^2}}$$

which is (1).