

## Problem Set 5

*Due Date: 27 June 2017*

1. In this exercise, we'll look at a different way of bounding the largest eigenvalue and obtaining an approximation algorithm for the maximum cut problem; we'll do part of the problem today, and the other part in tomorrow's problem set. Let  $\lambda_n$  be the maximum eigenvalue of the normalized Laplacian  $\mathcal{L}$ , and let  $y$  be the corresponding eigenvector, so that

$$\lambda_n = \frac{y^T L_G y}{y^T D y} = \max_x \frac{x^T L_G x}{x^T D x} = \max_z \frac{z^T \mathcal{L} z}{z^T z}.$$

Assume that  $\max_i |y(i)| \leq 1$ . Let  $OPT$  denote the number of edges in a maximum cut, and let  $S^* \subset V$  denote the set of vertices associated with that set, so that  $|\delta(S^*)| = OPT$ .

- (a) Prove that if  $OPT \geq (1 - \epsilon)|E|$ , then  $\lambda_n \geq 2(1 - \epsilon)$ .
- (b) Suppose we construct a solution  $x \in \{-1, 0, +1\}^n$  as in Trevisan's algorithm (that is, pick  $t \in (0, 1]$  uniform, and let  $x(i) = -1$  if  $y(i) \leq -\sqrt{t}$ ,  $x(i) = 1$  if  $y(i) \geq \sqrt{t}$ , and  $x(i) = 0$  otherwise). Let sets  $L = \{i \in V : x(i) = -1\}$ ,  $R = \{i \in V : x(i) = 1\}$ ,  $S = L \cup R$ , and  $V - S = \{i \in V : x(i) = 0\}$ .

It can be shown that for all  $0 \leq \beta \leq 1$ ,

$$E[|\delta(L, R)| + \beta|\delta(S)|] \geq \beta(1 - \beta) \sum_{(i,j) \in E} (y(i) - y(j))^2.$$

(If you want some extra work, you can prove this yourself; it might help to know Bergström's inequality, which states that for  $a, b \geq 0$  and  $0 \leq \beta \leq 1$ ,  $\beta(1 - \beta)(a + b)^2 \leq (1 - \beta)a^2 + \beta b^2$ ).

Prove that if  $\lambda_n \geq 2(1 - \epsilon)$ , then

$$E[|\delta(L, R)| + \beta|\delta(S)|] \geq 2(1 - \epsilon)\beta(1 - \beta)E[2|E(S)| + |\delta(S)|].$$