



CS 598: Spectral Graph Theory. Lecture 13

Eigenvalues of Random Graphs (or,
Random Graphs are Expanders)

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Today

- Define random graph model.
- Concentration inequalities
- Bounding Raleigh Quotients
- Explain Trace method

Erdos-Renyi model

- Set n of vertices.
- Between every 2 vertices (u, v) add edge with probability p : $G(n, p)$.
- For this lecture $p = \frac{1}{2}$.
- Adjacency matrix has entries $A(i, j) = 0$ with probability $\frac{1}{2}$ and $A(i, j) = 1$ with probability $\frac{1}{2}$.

The Expectation

- Let $M = E(A) = \frac{1}{2}A_{K_n} = \frac{1}{2}(J_n - I_n)$ the expectation of the adjacency matrix
- One eigenvalue is $(n-1)/2$ and the others are $-1/2$.
- We will show that $\|A - M\| \leq 1.34\sqrt{n}$ with exponentially high probability.
- $\|A - M\| = \max_i |\lambda_i(A - M)| = \max_x |x^T R x / x^T x|$, where $R=A-M$.

One Raleigh Quotient

- $R(i,j)$ is mean-zero r.v taking values $\pm 1/2$.
- Fix any unit vector x :
$$x^T R x = \sum_{i < j} 2r_{ij} x(i)x(j).$$
- This is sum of independent r.v, with mean zero.
- **Hoeffding bound** for i.i.d $X_i, |X_i| \in [a_i, b_i], E(X_i) = \mu$:

$$\Pr\left[\sum_{i=1 \text{ to } d} X_i \geq t + \mu \right] \leq e^{-2t^2 / \sum_i (b_i - a_i)^2}$$

One Raleigh Quotient

- **Lemma.** For every unit vector x

$$\Pr_R[|x^T R x| \geq t] \leq 2e^{-t^2}$$

- Kind of useless, there are infinitely many x 's!!!

One Raleigh Quotient

- **Lemma.** Let R be a symmetric matrix and let u be a unit eigenvector of R whose eigenvalue has absolute value $\|R\|$. If x is another unit vector such that $\langle u, x \rangle \geq \sqrt{3}/2$, then $x^T R x \geq \frac{1}{2} \|R\|$.

How many vectors are close?

- **Lemma.** Let v be an arbitrary unit vector and x a random unit vector. Then

$$\Pr \left[\langle x, v \rangle \geq \frac{\sqrt{3}}{2} \right] \geq \frac{1}{\sqrt{\pi n} 2^{n-1}}$$

The Probabilistic Argument

- **Theorem.** Let R be a symmetric matrix as above.

$$\Pr[||R|| \geq t] \leq \sqrt{\pi n} 2^n e^{-t^2/4}$$

An Alternative Argument

- Trace method used to bound eigenvalues of matrices. More powerful sometimes (e.g. eigenvalues of d -regular graphs)
- $Tr(A^k) = \sum_i \lambda_i^k = \sum k - \text{walks}(i, i)$
- $\|R\| \leq \left(Tr(R^k)\right)^{1/k}$