

{ Csc 80030 | Lecture 10 }

PROBABILISTIC ANALYSIS &
RANDOMIZED ALGORITHMS

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PART 0

- ▶ HOMEWORK: HW5 due May 6
- ▶ PRESENTATIONS:
 - ▶ May 6 – Michael (balls and bins), Ben (sublinear-time algorithm), Peter (hiring problem)
 - ▶ May 13 – Valia (balls and bins), Justin
 - ▶ Homework problem requirement
- ▶ TODAY: expander graphs

PART 1 | expander graphs

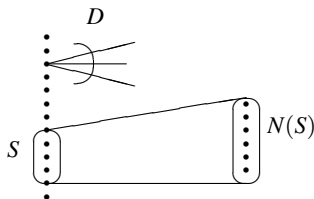
Expander graphs

1. “sparse”

- ▶ constant degree $\implies$ linear # of edges

2. “well-connected”

- ▶ (K, A) vertex expansion: for all sets S of $\leq K$ vertices, the neighborhood $N(S)$ has size $\geq A \cdot |S|$.



- ▶ would like $K = \Omega(N)$ and $A \sim D$.

Expander graphs

1. “sparse”

- ▶ constant degree $\implies$ linear # of edges

2. “well-connected”

- ▶ (K, A) vertex expansion: for all sets S of $\leq K$ vertices, the neighborhood $N(S)$ has size $\geq A \cdot |S|$.
- ▶ edge expansion: # edges leaving S is large (i.e. no small cuts)
- ▶ random walks: converge quickly to uniform, i.e., 2nd eigenvalue is small

QUESTION.

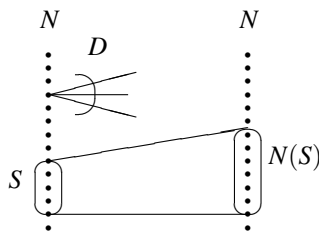
- ▶ Do good expanders exist?
- ▶ Specifically, (K, A) vertex expanders with $K = \Omega(N), A = 1 + \Omega(1)$?

Random graphs are good expanders

- ▶ **THEOREM.** For all $D \geq 3$, there is a constant $\alpha > 0$ s.t. a random D -regular graph on N vertices is an $(\alpha N, D - 1.01)$ vertex expander with prob. $\geq 1/2$.
 - ▶ $D \geq 3$ is necessary – every graph of degree 2 is a poor expander (being a union of cycles and chains).
 - ▶ will only prove a simpler result for bipartite expanders.

Random graphs are good expanders

- **THEOREM.** For all $D \geq 3$, there is a constant $\alpha > 0$ s.t. a random D -regular graph on N vertices is an $(\alpha N, D - 1.01)$ vertex expander with prob. $\geq 1/2$.
- **DEFINITION.** A bipartite graph is a (K, A) vertex expander if for all sets of $\leq K$ left vertices, the neighborhood $N(S)$ has size $\geq A \cdot |S|$.



- **THEOREM.** For all $D \geq 3$, there is a constant $\alpha > 0$ s.t. a random D -left-regular (N, N) -bipartite graph is an $(\alpha N, D - 2)$ vertex expander

Random bipartite expanders

- **THEOREM.** For all $D \geq 3$, there is a constant $\alpha > 0$ s.t. a random D -left-regular (N, N) -bipartite graph is an $(\alpha N, D - 2)$ vertex expander with prob. $\geq 1/2$.

- **STEP 1.** Fix S of size $K \leq \alpha N$. Show that

$$\Pr[|N(S)| \leq (D - 2) \cdot K] \leq \binom{KD}{2K} \left(\frac{KD}{N}\right)^{2K}$$

- **STEP 2.** Take a union bound over S and K .

$$\begin{aligned} \sum_{K=1}^{\alpha N} \binom{N}{K} \binom{KD}{2K} \left(\frac{KD}{N}\right)^{2K} &\leq \sum_{K=1}^{\alpha N} \left(\frac{Ne}{K}\right)^K \left(\frac{KDe}{2K}\right)^{2K} \left(\frac{KD}{N}\right)^{2K} = \sum_{K=1}^{\alpha N} \left(\frac{e^3 D^4 K}{4N}\right)^K \\ &\leq \sum_{K=1}^{\alpha N} \left(\frac{1}{4}\right)^K < \frac{1}{2} \quad \text{setting } \alpha = \frac{1}{e^3 D^4}. \end{aligned}$$

- ▶ Applications
 - ▶ construction of fault-tolerant networks
 - ▶ sorting in $O(\log n)$ time in parallel
 - ▶ derandomization (e.g. of UNDIRECTED S-T CONNECTIVITY)
 - ▶ error-correcting codes
 - ▶ distributed routing
 - ▶ data structures
 - ▶ negative results for integrality gaps for linear programming relaxations and metric embeddings
- ▶ Need *explicit* constructions (i.e. deterministic and efficient)

PART 2 | expansion and eigenvalues

Second eigenvalue

- **DEFINITION.** For an n -vertex regular graph G with random-walk matrix M , we define

$$\lambda(G) = \max_{\pi \in [0,1]^N} \frac{\|\pi M - u\|}{\|\pi - u\|}$$

- **INTERPRETATION.** If $\lambda(G)$ is small, then random walk on G converges quickly to uniform.
 - $\|\pi M - u\| \leq \lambda \|\pi - u\|$
 - $\|\pi M^k - u\| \leq \lambda^k \|\pi - u\|$

Second eigenvalue

- **DEFINITION.** For an n -vertex regular graph G with random-walk matrix M , we define

$$\lambda(G) = \max_{\pi \in [0,1]^N} \frac{\|\pi M - u\|}{\|\pi - u\|}$$

- **CLAIM.** Let $1 = \lambda_1, \lambda_2, \dots, \lambda_N$ be the eigenvalues of M , with $|\lambda_1| \geq |\lambda_2| \geq \dots \geq |\lambda_N|$. Then, $\lambda(G) = |\lambda_2|$ (“2nd eigenvalue”).

- Write $\pi = u + c_2 v_2 + \dots + c_N v_N$.
- Hence, $\pi M = u + c_2 \lambda_2 v_2 + \dots + c_N \lambda_N v_N$.

$$\begin{aligned} \text{Now, } \|\pi M - u\|^2 &= c_2^2 \lambda_2^2 \|v_2\|^2 + \dots + c_N^2 \lambda_N^2 \|v_N\|^2 \\ &\leq \lambda_2^2 (c_2^2 \|v_2\|^2 + \dots + c_N^2 \|v_N\|^2) \\ &= \lambda_2^2 \|\pi - u\|^2 \end{aligned}$$

Expansion and eigenvalues

- ▶ **THEOREM.** If G is a regular graph with 2nd eigenvalue λ , then G is an $(N/2, 2 - \lambda)$ vertex expander.
 - ▶ smaller λ , better expansion.
 - ▶ $2 - \lambda$ is “best possible.”
 - ▶ In fact, $\forall \alpha \in [0, 1]$, G is an $(\alpha N, 1/((1 - \alpha)\lambda^2 + \alpha))$ vertex expander.

Expansion and eigenvalues

- ▶ **THEOREM.** If G is a regular graph with 2nd eigenvalue λ , then $\forall$ subsets S of size $\leq \alpha N$, $|N(S)| \geq |S|/(1 - \alpha)\lambda^2 + \alpha$.
- ▶ **DEFINITIONS.** For a prob. distribution π ,
 - ▶ The *collision probability* of π is the prob. that two independent samples are equal, i.e., $\text{CP}(\pi) = \sum_x \pi_x^2$.
 - ▶ The *support* of π is $\text{Supp}(\pi) = \{x : \pi_x > 0\}$.
- ▶ **CLAIMS.** For any prob. distribution $\pi \in [0, 1]^N$,
 - ▶ $\text{CP}(\pi) = \|\pi\|^2 = \|\pi - u\|^2 + 1/N$.
 - ▶ $\text{CP}(\pi) \geq 1/\text{Supp}(\pi)$. (since $(\sum_x \pi_x^2) \cdot (\sum_x 1) \geq (\sum_x \pi_x)^2$)
- ▶ **KEY IDEA.** set π to be uniform over S .

Expansion and eigenvalues

► **THEOREM.** If G is a regular graph with 2nd eigenvalue λ , then $\forall$ subsets S of size $\leq \alpha N$, $|N(S)| \geq |S|/(1 - \alpha)\lambda^2 + \alpha$.

► STEP 1. for all π , $\|\pi M - u\|^2 \leq \lambda^2 \|\pi - u\|^2$.

1st claim $\implies \text{CP}(\pi M) - \frac{1}{N} \leq \lambda^2 (\text{CP}(\pi) - \frac{1}{N})$.

► STEP 2. set π to be uniform over S . Then,

2nd claim $\implies \text{CP}(\pi M) \geq \text{Supp}(\pi M) = 1/|N(S)|$.

Also, $\text{CP}(\pi) = 1/|S|$.

► COMBINING, $\frac{1}{|N(S)|} - \frac{1}{N} \leq \lambda^2 (\frac{1}{|S|} - \frac{1}{N})$

An equivalence

- ▶ **THEOREM.** Let $\mathcal{G}$ be an infinite family of D -regular multigraphs for a constant D . Then, the following two conditions are equivalent:
 - ▶ “good” vertex expansion:
 $\exists$ constant $\delta_1 < 1$ s.t. every $G \in \mathcal{G}$ is an $(N/2, 2 - \delta_1)$ vertex expander.
 - ▶ “small” 2nd eigenvalue:
 $\exists$ constant $\delta_2 < 1$ s.t. every $G \in \mathcal{G}$ has 2nd eigenvalue at most δ_2 .
- ▶ **NOTE.** (2) $\implies$ (1) as before; the other direction is “hard” (omitted).
- ▶ The term “expander” typically refers to a family of regular graphs of constant degree satisfying one of the two equivalent conditions.

PART 3 | towards explicit constructions

Iterative constructions of expanders

► **APPROACH.** iterative constructions

1. Start with a “constant size” expander H .
2. Repeatedly apply graph operations to get better expanders.
increase # nodes, keep degree and 2nd eigenvalue bounded.

► **DEFINITION.** a (N, D, λ) -graph is a D -regular graph on N vertices with 2nd eigenvalue $\leq \lambda$.

Squaring of graphs

- ▶ **GRAPH SQUARING.** If $G = (V, E)$ is a D -regular graph, then $G^2 = (V, E')$ is a D^2 -regular graph on the same vertex set, E' comprises pairs of vertices connected by a path of length 2 in G .
- ▶ **LEMMA.** If G is a (N, D, λ) -graph, then G^2 is a (N, D^2, λ^2) -graph.
 - ▶ If M is random-walk matrix for G , then M^2 is random-walk matrix for G^2 .

THE END | next, ...