

Expander Graphs and Their Applications (VIII)

Yijia Chen
Shanghai Jiaotong University

Last Lecture

The Zig-Zag Theorem

Theorem

If G is an (n, m, α) -graph and H an (m, d, β) -graph. Then $G \otimes H$ is an $(nm, d^2, \varphi(\alpha, \beta))$ -graph, where

$$\varphi(\alpha, \beta) = \frac{1}{2}(1 - \beta^2)\alpha + \frac{1}{2}\sqrt{(1 - \beta^2)^2\alpha^2 + 4\beta^2}.$$

Corollary

If G is an (n, m, α) -graph and H an (m, d, β) -graph. Then

$$1 - \hat{\lambda}(G \otimes H) \geq \frac{1}{2}(1 - \beta^2)(1 - \alpha)$$

Lower Bound of $d - \lambda$ in Arbitrary Graphs

Theorem

Let G be a connected, d -regular, and non-bipartite graph with n vertices. Then

$$d - \lambda \geq \frac{1}{n^2}.$$

Reingold's Algorithm for Graph Reachability Problem

Theorem (**Omer Reingold**, 2005)

There is an algorithm that solves the reachability problem on *any* graph $G = (V, E)$ using space $O(\log |V|)$.

Let H be a $(d^{16}, d, 1/2)$ -graph for some constant $d \in \mathbb{N}$.

Input: A d^{16} -regular graph G with $n := |V(G)|$ and two vertices $s, t \in V(G)$.

Output: Is there a path between s and t in G ?

1. Choose some $\ell = O(\log n)$ such that

$$\left(1 - \frac{1}{d^{16}n^2}\right)^{2^\ell} \leq \frac{1}{2}.$$

2. $G_0 \leftarrow G$.
3. **for** $i = 1$ **to** ℓ **do** $G_i \leftarrow (G_{i-1} \otimes H)^8$.
4. Check if s and t are connected in G_ℓ .

The Analysis of Reingold's Algorithm

For all $i \in [\ell]$ and every *connected component* C of G , we define

$$C_0 := C \quad \text{and} \quad C_i := (C_{i-1} \otimes H)^8$$

It is easy to see

$$G_i := \bigcup_{C \text{ a connected component of } G} C_i$$

Proposition

$$\hat{\lambda}(C_i) \leq \max \left\{ \hat{\lambda}^2(C_{i-1}), \frac{1}{2} \right\}.$$

The Analysis of Reingold's Algorithm (cont'd)

Proposition *For every connected component C of G and all $i \in [\ell]$*

$$\hat{\lambda}(C_i) \leq \max \left\{ \hat{\lambda}^2(C_{i-1}), \frac{1}{2} \right\}.$$

Corollary

For every connected component C of G

$$\hat{\lambda}(C_\ell) \leq \frac{1}{2}.$$

The Base Graphs

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In the following, we

- ▶ provide some *explicit construction*,
- ▶ and prove their existence by the *probabilistic method*.

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If G is an (n, d, α) -graph, then G^i is an (n, d^i, λ^i) -graph.

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$$\underline{\vec{u} \otimes \vec{v}} \in \mathbb{R}^{n_1 \cdot n_2}$$

whose (i, j) th entry is $u_i \cdot v_j$ for every $i \in [n_1]$ and $j \in [n_2]$.

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Let A be an $n_1 \times n_1$ matrix and B an $n_2 \times n_2$ matrix, then their tensor product $A \otimes B$ is an $n_1 \cdot n_2 \times n_1 \cdot n_2$ matrix such that for every $(i_1, j_1), (i_2, j_2) \in [n_1] \times [n_2]$

$$(A \otimes B)_{(i_1, j_1), (i_2, j_2)} = A_{i_1, i_2} \cdot B_{j_1, j_2}.$$

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Lemma

$$(A \otimes B)(\vec{u} \otimes \vec{v}) = (A\vec{u}) \otimes (B\vec{v}).$$

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If α is an eigenvalue for A with corresponding eigenvector \vec{u} , and β is an eigenvalue for B with corresponding eigenvector \vec{v} , then $\alpha \cdot \beta$ is an eigenvalue for $A \otimes B$ with corresponding eigenvector $\vec{u} \otimes \vec{v}$.

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Theorem

Let A be an $n_1 \times n_1$ matrix with eigenvalues $\alpha_1, \dots, \alpha_{n_1}$ and corresponding **orthogonal** eigenvectors $\vec{u}_1, \dots, \vec{u}_{n_1}$. And let B be an $n_2 \times n_2$ matrix with eigenvalues $\beta_1, \dots, \beta_{n_2}$ and corresponding **orthogonal** eigenvectors $\vec{v}_1, \dots, \vec{v}_{n_2}$. Then $A \otimes B$ has eigenvalue

$$\alpha_1 \cdot \beta_1, \alpha_2 \cdot \beta_1, \dots, \alpha_{n_1} \cdot \beta_1, \alpha_1 \cdot \beta_2, \dots, \alpha_{n_1} \cdot \beta_{n_2}$$

with corresponding **orthogonal** eigenvectors

$$\vec{u}_1 \otimes \vec{v}_1, \vec{u}_2 \otimes \vec{v}_1, \dots, \vec{u}_{n_1} \otimes \vec{v}_1, \vec{u}_1 \otimes \vec{v}_2, \dots, \vec{u}_{n_1} \otimes \vec{v}_{n_2}.$$

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$$(n_1 \cdot n_2, d_1 \cdot d_2, \max\{\alpha_1, \alpha_2\})\text{-graph.}$$

Proof.

$$\begin{aligned}\hat{\lambda}(G_1 \otimes G_2) &= \max \left\{ |\hat{\lambda}_i(G_1) \cdot \hat{\lambda}_j(G_2)| \mid i \in [n_1], j \in [n_2], \text{ and } (i \neq 1 \text{ or } j \neq 1) \right\} \\ &= \max \{ \hat{\lambda}(G_1), \hat{\lambda}(G_2) \}.\end{aligned}$$

□

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Then \underline{AP}_q is a graph with vertex set \mathbb{F}_q^2 and edge set

$$\left\{ \begin{array}{l} \text{an edge between } (a, b) \text{ and } (c, d) \\ | a, b, c, d \in \mathbb{F}_q \text{ and } ac = b + d \end{array} \right\}.$$

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Equivalently, we connect the vertex (a, b) to all points on the line

$$\mathcal{L}_{a,b} := \{(x, y) \mid y = ax - b\}.$$

The Affine Plane (cont'd)

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Proof. Let \underline{M} be the $q^2 \times q^2$ normalized adjacency matrix of AP_q .

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Proof. Let M be the $q^2 \times q^2$ normalized adjacency matrix of AP_q . The entry of M^2 in row (a, b) and column (a', b') is exactly the number of common neighbors of (a, b) and (a', b') *divided by q^2* , i.e.,

$$|L_{a,b} \cap L_{a',b'}|/q^2.$$

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- ▶ If $a = a'$ and $b = b'$, then $|L_{a,b} \cap L_{a',b'}| = q$

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Thus $\lambda(M) = 1/\sqrt{q}$. □

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Choosing some sufficiently large q , we can get a $(d^4, d, 1/4)$ -graph or a $(d^{16}, d, 1/2)$ graph.

The Probabilistic Method

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$\mathbb{P}(k)$ is a random perfect matching on $[2k]$.

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4. Output (V, E) .

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An Important Warning: $\mathbb{R}_d(k)$ is not uniformly distributed over all d -regular graphs on vertices $[2k]$.

Main Theorem (Restated)

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Theorem

There exists a constant $c > 0$ such that for all sufficiently large $k \in \mathbb{N}$

$$\Pr [h(\mathbb{R}_3(k)) \geq c] > 0$$

Proof.

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Consider the event

$E :=$ there exists a subset $S \subseteq V$ with $|S| \leq k = |V|/2$ and $|\partial S| \leq c \cdot |S|$.

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$$\begin{aligned} \Pr[E] &\leq \sum_{i \in [k]} \binom{2k}{i} \binom{2k-i}{c \cdot i} \Pr[\Gamma([i]) \subseteq [i + c \cdot i]] \\ &\leq \sum_{i \in [k]} \binom{2k}{i} \binom{2k}{c \cdot i} \Pr[\Gamma([i]) \subseteq [i + c \cdot i]]. \end{aligned}$$

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Then

$$\begin{aligned} \prod_{j=1}^{\lceil i/2 \rceil} \frac{i + c \cdot i - 2j + 1}{2k - 2j + 1} &\leq \prod_{j=0}^{\lceil i/2 \rceil - 1} \frac{i + c \cdot i - 2j}{2k - 2j} \\ &= \frac{2^{\lceil i/2 \rceil} \cdot \prod_{j=0}^{\lceil i/2 \rceil - 1} (\lceil (i + c \cdot i)/2 \rceil - j)}{2^{\lceil i/2 \rceil} \cdot \prod_{j=0}^{\lceil i/2 \rceil - 1} (k - j)} \\ &= \binom{\lceil (i + c \cdot i)/2 \rceil}{\lceil i/2 \rceil} \bigg/ \binom{k}{\lceil i/2 \rceil} \end{aligned}$$

Binary Entropy

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Definition

The binary entropy function $H_2 : (0, 1) \rightarrow \mathbb{R}$ is defined by

$$H_2(p) = -p \log_2 p - (1 - p) \log_2(1 - p).$$

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Lemma

For $p \in (0, 1)$ and $n \in \mathbb{N}$

$$\binom{n}{\lceil p \cdot n \rceil} \approx 2^{H(p) \cdot n}.$$

Proof. (cont'd)

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We have seen

$$\begin{aligned} & \Pr [\text{there exists an } S \subseteq V \text{ with } |S| \leq k \text{ and } |\partial S| \leq c \cdot |S|] \\ & \leq \sum_{i \in [k]} \binom{2k}{i} \binom{2k}{c \cdot i} \Pr [\Gamma([i]) \subseteq [i + c \cdot i]] \\ & \leq \sum_{i \in [k]} \binom{2k}{i} \binom{2k}{c \cdot i} \binom{\lceil (i + c \cdot i)/2 \rceil}{\lceil i/2 \rceil}^3 / \binom{k}{\lceil i/2 \rceil}^3 \end{aligned}$$

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Then

$$\begin{aligned} & \log_2 \left(\binom{2k}{i} \binom{2k}{c \cdot i} \binom{\lceil (i + c \cdot i)/2 \rceil}{\lceil i/2 \rceil}^3 / \binom{k}{\lceil i/2 \rceil}^3 \right) \\ & \approx H_2(i/2k) \cdot 2k + H_2(c \cdot i/2k) \cdot 2k \\ & \quad + 3 \cdot H_2(i/(i + c \cdot i)) \cdot (i + c \cdot i)/2 - 3H_2(i/2k) \cdot k \\ & = - (H_2(i/2k) - 2H_2(c \cdot i/2k)) \cdot k + 3 \cdot H_2(1/(1 + c)) \cdot (i + c \cdot i)/2. \end{aligned}$$

Binary Entropy (cont'd)

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Lemma

Let $p \in (0, 1/2)$ and $\varepsilon \in (0, 1/2)$. Then

$$H_2(\varepsilon \cdot p) \leq \delta \cdot H_2(p)$$

for $\delta := -4 \cdot \varepsilon \cdot \log \varepsilon$.

Proof.

$$\begin{aligned} H_2(\varepsilon \cdot p) &= -\varepsilon \cdot p \cdot \log(\varepsilon \cdot p) - (1 - \varepsilon \cdot p) \log(1 - \varepsilon \cdot p) \\ &\leq -2 \cdot \varepsilon \cdot p \cdot \log(\varepsilon \cdot p) \\ &\leq 4 \cdot \varepsilon \cdot p \cdot \log \varepsilon \cdot \log p \\ &\leq -4 \cdot \varepsilon \cdot \log \varepsilon \cdot H_2(p). \end{aligned}$$

□

Proof. (cont'd)

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For $c \in (0, 1/2)$, recall

$$\begin{aligned} & \log_2 \left(\binom{2k}{i} \binom{2k}{c \cdot i} \binom{\lceil (i + c \cdot i)/2 \rceil}{\lceil i/2 \rceil}^3 \right) / \left(\binom{k}{\lceil i/2 \rceil}^3 \right) \\ & \approx - (H_2(i/2k) - 2H_2(c \cdot i/2k)) \cdot k + 3 \cdot H_2(1/(1+c)) \cdot (i + c \cdot i)/2 \\ & \leq - (1 - c') \cdot H_2(i/2k) \cdot k + c' \cdot i, \end{aligned}$$

for $c' := \max \{ -8 \cdot c \cdot \log c, 3 \cdot H_2(1/(1+c)) \cdot (1+c)/2 \}$.

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For $c \in (0, 1/2)$, recall

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Now

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for some appropriate c (independent of k).

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