

Explicit expander constructions

- We will construct an explicit expander family.
- The main idea is to use three kinds of graph products.

Tradeoff between degree & the spectral gap.

- For the operations we need a new representation of graphs:

Definition: If G is an n -vertex d -degree ^{connected} graph then we label each neighbour of each vertex using $[d]$.

Define a rotation map

$$\hat{G} : [n] \times [d] \rightarrow [n] \times [d]$$

$(v, i) \mapsto (u, j)$, where u is the i -th vertex of v & v the j -th vertex of u .

$\triangleright \hat{G}$ is a permutation on $[n] \times [d]$.
 Pf: It is 1-1 & onto. $\square$

Matrix Product (or Path product)

Definition: For two n -vertex graphs G, G' , with degrees d, d' , the graph GG' is the one with the normalized adjacency matrix AA' . (A, A' are the normalized adjacency matrices of G, G' .)

Proposition: 1) AA' is stochastic.
 2) GG' may have repeated edges, the degree is dd' , & $\# \text{vertices} = n$.

$$\triangleright \lambda(GG') \leq \lambda(G) \cdot \lambda(G').$$

Pf:

$$\lambda(GG') = \max_{u \in \mathbb{T}^\perp} \frac{\|AA'u\|}{\|u\|} = \max_{u \in \mathbb{T}^\perp} \frac{\|AA'u\|}{\|A'u\|} \cdot \frac{\|A'u\|}{\|u\|}$$

$A'u \in \mathbb{T}^\perp \rightarrow \leq \lambda(G) \cdot \lambda(G'). \quad \square$

Theorem: If G, G' are $(n, d, \lambda), (n', d', \lambda')$ -expanders then GG' is $(n, dd', \lambda\lambda')$ -expander.

Pf: Clear. $\square$

$\Rightarrow$ Thus, matrix product improves the spectral gap at the cost of the degree.

Tensor product

Definition: The graph $G \otimes G'$ is the one with the normalized adjacency matrix $A \otimes A'$,
where $A \otimes A' := \begin{pmatrix} A_{1,1} A' & \dots & A_{1,n} A' \\ \vdots & \ddots & \vdots \\ A_{n,1} A' & \dots & A_{n,n} A' \end{pmatrix}_{nn' \times nn'}$.

Proposition: 1) $A \otimes A'$ is symmetric stochastic.
2) $G \otimes G'$ is nn' -vertex, dd' -degree.

$$\triangleright \lambda(G \otimes G') = \max(\lambda(G), \lambda(G')).$$

Pf:

• Let $1 = \lambda_1 \geq |\lambda_2| \geq \dots \geq |\lambda_n|$ &
 $1 = \lambda'_1 \geq |\lambda'_2| \geq \dots \geq |\lambda'_n|$ be the eigenvalues of A & A' resp.

• Then the eigenvalues of $A \otimes A'$ are:
 $\{\lambda_i \lambda'_j \mid i \in [n], j \in [n']\}$.

$$[\because (A \otimes A') \cdot (v \otimes v') = Av \otimes A'v'.]$$

• Thus, the largest ones, apart from 1, are
 $\{\lambda_i \mid i \in [n]\} \cup \{\lambda'_i \mid i \in [n']\}$.

$$\Rightarrow \lambda(G \otimes G') = \max\{\lambda(G), \lambda(G')\}. \quad \square$$

Theorem: If G, G' are $(n, d, \lambda), (n', d', \lambda')$ -expanders,
 then $G \otimes G'$ is $(nn', dd', \max(\lambda, \lambda'))$ - " .

$\Rightarrow$ Tensor product increase the #vertices while preserving the spectral gap.

The Replacement Product

- This product is easier to see as a walk rather than a matrix operation.
- Suppose we have an expander G with a high degree- D . In order to do a walk using fewer random bits, we can use another expander G' with vertex- D & degree- d ($\ll D$).
- This motivates the product $G \circledast G'$:

Definition: Let G, G' be graphs with vertices n, D , degrees D, d & normalized adjacency matrices A, A' .

$H := G \circledast G'$ is a nD -vertex graph st.
i) $\forall u \in V(G)$, H has a copy of G' , say H_u , called a cloud. I.e. $\forall i \in V(G')$, $(u, i) \in V(H)$, & is called the i -th vertex in the u -th cloud.

ii) If $\hat{G}(u,i) = (v,j)$ then $((u,i), (v,j)) \in E(H)$.
 Also, if $(i,j) \in E(G')$ then $\forall u \in V(G), ((u,i), (u,j)) \in E(H)$.

$\triangleright H = G \circledast G'$ is nD -vertex, $(d+1)$ -degree.

Pf:

- Any vertex in H corresponds to a vertex (u,i) in H_u , $u \in V(G)$.
 - $\because H_u \cong G'$, (u,i) has d neighbours in the cloud.
 - Further property (ii) of the definition adds one edge outside the cloud.
- $\Rightarrow H$ has degree $(d+1)$. $\square$

Zig-Zag Product (has more memory)
(like self-loops)

- We consider length-3 paths that zig-zag the clouds.

Zig-Zag product
↓

Definition: On the vertex set $V(G \circledast G')$ define $H := \underline{G \circledast G'}$ s.t.
 $((u,i), (v,j))$ is an edge if $\exists \ell, k$ with $((u,i), (u,\ell)), ((u,\ell), (v,k)), ((v,k), (v,j)) \in E(G \circledast G')$.

▷ $G \otimes G'$ is nd -vertex, d^2 -degree.

▷ Its normalized adjacency matrix is

$$A \otimes A' := B \hat{A} B, \text{ where}$$

$$\hat{A}[(u,i), (v,j)] = \begin{cases} 1 & , \text{ if } \hat{G}(u,i) = (v,j), \\ 0 & , \text{ else.} \end{cases}$$

$$\& B[(u,i), (v,j)] = \begin{cases} A'[i,j] & , \text{ if } u=v, \\ 0 & , \text{ else.} \end{cases}$$

Pf:

• $A \otimes A'$ encodes the definition of $G \otimes G'$.

$$\cdot \text{ Also, } (A \otimes A') \cdot \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} = B \hat{A} B \cdot \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} = B \hat{A} \cdot \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}$$

$$= B \cdot \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} \quad [\because \hat{A} \text{ merely permutes }]$$

$$= \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}$$

$\Rightarrow A \otimes A'$ is stochastic, symmetric. $\square$

Theorem (Reingold, Vadhan, Wigderson, 2002): If $\lambda(G) = a$ & $\lambda(G') = b$ then $\lambda(G \otimes G') \leq a + 2b + b^2$.

Proof: • Let $M := A \otimes A'$.

• Recall that $M = B \hat{A} B$, where $\hat{A}$ is a permutation & $B = I_n \otimes A'$.

• Write $B = I_n \otimes \frac{1}{D} J + I_n \otimes \underbrace{(A' - \frac{1}{D} J)}_E$, where J is the all-one matrix.

• Define $\bar{J} := I_n \otimes \frac{1}{D} J$ & $\bar{E} := I_n \otimes E$.

$$\Rightarrow M = B \hat{A} B = (\bar{J} + \bar{E}) \cdot \hat{A} \cdot (\bar{J} + \bar{E})$$

$$= \bar{J} \hat{A} \bar{J} + \bar{J} \hat{A} \bar{E} + \bar{E} \hat{A} \bar{J} + \bar{E} \hat{A} \bar{E}.$$

• For each of these matrices we will upper bound the matrix norm $\|A\| := \max_{\substack{x \in \mathbb{R}^n \\ \|x\|=1}} \|Ax\|$.

$$\lambda(A) = \max_{\substack{x \in \mathbb{R}^n \\ \|x\|=1}} \|Ax\|$$

• Let $\underline{\lambda}(A) := \max_{\|x\|=1} \|Ax\|$. equals largest eigenvalue

• Clearly, $\|E\| \leq \underline{\lambda}(E) = \underline{\lambda}(A' - \frac{J}{D})$

• Let $\lambda_1, \dots, \lambda_D$ be the eigenvalues of A' s.t. $1 = |\lambda_1| \geq |\lambda_2| \geq \dots \geq |\lambda_D|$. We can write $A' = \sum_{i=1}^D \lambda_i v_i v_i^T$ using the orthonormal eigenvectors of A' .

$$\Rightarrow \|E\| \leq \underline{\lambda}(A' - \frac{J}{D}) = \underline{\lambda}(\sum_{i=2}^D \lambda_i v_i v_i^T) \leq |\lambda_2| = \|A'\| = b.$$

• Going back, we have by the triangle inequality:

$$\|M\| \leq \|\bar{J}\hat{A}\bar{J}\| + \|\bar{J}\hat{A}\bar{E}\| + \|\bar{E}\hat{A}\bar{J}\| + \|\bar{E}\hat{A}\bar{E}\|$$

$$\leq \|\bar{J}\hat{A}\bar{J}\| + \|\bar{J}\| \cdot \|\hat{A}\| \cdot \|\bar{E}\| + \dots$$

[$\because \hat{A}, \bar{J}, \bar{E}$ map $\bar{1}^\perp$ to $\bar{1}^\perp$.]

$$\leq \|\bar{J}\hat{A}\bar{J}\| + 1 \cdot 1 \cdot b + b \cdot 1 \cdot 1 + b \cdot 1 \cdot b$$

• Now we estimate $\|\bar{J}\hat{A}\bar{J}\|$.

$$\triangleright \bar{J}\hat{A}\bar{J} = A \otimes \bar{D}^{-1} \cdot J.$$

$$\text{Pf: } (\bar{J}\hat{A}\bar{J})_{(u,i),(v,j)} = \bar{J}_{(u,i),-} \cdot \hat{A} \cdot \bar{J}_{-, (v,j)}$$

$$= \frac{1}{D} \cdot A_{u,v} = (A \otimes \bar{D}^{-1} \cdot J)_{(u,i),(v,j)}. \quad \square$$

• Now note that the eigenvalues of J are $\{1, 0, \dots, 0\}$.

This is because $J \cdot \bar{1} = \bar{1}$ &
 $J \cdot \bar{x} = 0$, $\forall \bar{x} \in \bar{1}^\perp$.

$$\Rightarrow \|A \otimes \bar{D}^{-1} \cdot J\| = \|A\| = a.$$

$$\Rightarrow \|\bar{J}\hat{A}\bar{J}\| = a.$$

$$\Rightarrow \|M\| \leq a + 2b + b^2. \quad \square$$

Theorem: If G, G' are $(n, D, \lambda_1), (D, d, \lambda_2)$ -expanders then $G \otimes G'$ is $(nD, d^2, \lambda_1 + 2\lambda_2 + \lambda_2^2)$ - " .

- Thus, it reduces the degree dramatically when $D \gg d$, without increasing the λ too much.

The Construction

- Using the three products we now give a strongly explicit construction of expanders,
i.e. given (u, i) the i -th neighbour of u can be computed in $\text{poly}(\lg |V|)$ time.

Theorem: $\exists$ strongly explicit (d^2, λ) -expander family for poly-many constants $d \in \mathbb{N}$ & $\lambda \in (0, 1)$.

Pf: • We will recursively construct $\{G_k\}_{k \geq 1}$
s.t. G_k has $2^{O(k)}$ vertices.

• Let H be a $(d^8, d, 0.04)$ -expander.

It can be found by the known constructions, or randomly, for a constant d .

• Also, let $G_1 := H^2$ be $(d^8, d^2, 0.04^2)$ -expander.

• Initialize $G_2 = G_1$.

• For odd $k > 2$, $G_{k+1} = G_k := (G_{\frac{k-1}{2}}^{\otimes 2}) \circledast H$.

Claim: $\forall$ odd k , G_k is $(d^{8k}, d^2, 0.1)$ -expander.

Pf:

• True for $k=1$.

• By induction, #vertices $(G_k) = d^{8 \cdot \frac{k-1}{2} \times 2} \cdot d^8 = d^{8k}$.

• Similarly, $\deg(G_{\frac{k-1}{2}}^{\otimes 2}) = d^{2 \cdot 2 \cdot 2} = d^8 = |V(H)|$
 $\Rightarrow$ the zig-zag product is defined & reduces the degree to d^2 .

• Finally, $\lambda(G_k) \leq 0.1^2 + 2 \cdot (0.04) + (0.04)^2$
 $< \underline{0.1}$. $\square$

- The algorithm for finding the neighbours of $v \in V(G_k)$ is recursive:

Say, listing a row in $G_{\frac{k-1}{2}}$ takes time $T(\frac{k-1}{2})$.

$\Rightarrow$ We can list a row in $G_{\frac{k-1}{2}}^{\otimes 2}$ in time $2 \cdot T(\frac{k-1}{2})$. (It's d^4 -sparse.)

$\Rightarrow$ Listing all the needed d^4 -rows in the other $G_{\frac{k-1}{2}}^{\otimes 2}$ takes $d^4 \cdot 2 \cdot T(\frac{k-1}{2})$ time.

$$\Rightarrow T(k) = O(d^4 \cdot T(\frac{k-1}{2}))$$

$$\Rightarrow T(k) = d^{O(\lg k)} = k^{O(\lg d)}$$

$$= \text{poly}(k) = \text{poly-log}(\#V(G_k)).$$

constt. d $\square$

— Now, before we derandomize " $\text{Upath} \in \text{RL}$ " we will need another estimate for $\lambda(G \otimes G')$.

Because $\lambda(G)$ is, in general, large & then the previous bound is useless.

Theorem (Reingold, Trevisan, Vadhan 2005): If $\lambda(G) = 1 - \varepsilon$ & $\lambda(G') = 1 - \delta$ then $\lambda(G \otimes G') \leq 1 - \varepsilon \delta^2$.

Proof:

- As before, $M := A \otimes A' = B \hat{A} B$, where $B = I_n \otimes A'$ & $\hat{A}$ is the rotation map of G .
- We can express A' as $D^{-1} \delta J + (1 - \delta) C$, where J is the all-1 matrix & C is a matrix s.t. $\|C\| \leq 1$.

Again use
 $A' = \sum_{i=1}^n \lambda_i v_i v_i^T$

(Show that $\|\frac{1}{1-\delta} \cdot (A' - \frac{\delta}{D} J)\| \leq \|x\|$.)

$$\Rightarrow B = I_n \otimes (\delta D^{-1} J + (1 - \delta) C) =: \delta \bar{J} + (1 - \delta) \bar{C}$$

$$\Rightarrow M = (\delta \bar{J} + (1 - \delta) \bar{C}) \cdot \hat{A} \cdot (\delta \bar{J} + (1 - \delta) \bar{C})$$

$$= \delta^2 \bar{J} \hat{A} \bar{J} + (1 - \delta^2) \cdot F$$

$$\text{where } F := \frac{\delta}{1 + \delta} \cdot (\bar{J} \hat{A} \bar{C} + \bar{C} \hat{A} \bar{J}) + \frac{1 - \delta}{1 + \delta} \cdot \bar{C} \hat{A} \bar{C}$$

with $\|F\| \leq 1$. ($\because \|C\| \leq 1$, $\hat{A}$ is a permutation)

$$\Rightarrow M = \delta^2 \cdot A \otimes D^{-1} J + (1 - \delta^2) \cdot F$$

$$\Rightarrow \lambda(M) = \|M\| \leq \delta^2 \cdot \|A \otimes D^{-1} J\| + (1 - \delta^2) \cdot \|F\|$$

$$\leq \delta^2 \cdot (1 - \varepsilon) + (1 - \delta^2) \cdot 1$$

$$= (1 - \varepsilon \delta^2)$$

□

Theorem (Reingold 2005): $\text{Upath} \in \mathbb{L}$.

Proof:

- Let G be the given undirected n -vertex graph (s be the start vertex).

- Idea: To apply the graph products on G to get $\tilde{G}$ s.t. the connected component of s in $\tilde{G}$ is an expander (constant λ, d).

The shortest paths in $\tilde{G}$ are then only $O(\lg n)$ in length.

- Let d be a large enough constant s.t. a $(d^{16}, d, 1/2)$ -expander H exists.

- Wlog assume G to be $D := d^{16}$ -regular.

- Now we transform G as follows:

$$G_0 := G, \quad G_{i+1} := (G_i \otimes H)^8 \text{ for } i \geq 0.$$

▷ G_k is an (nd^{16k}, d^{16}) -graph.

Pf: • by induction, $\#V(G_k) = nd^{16(k-1)} \cdot d^{16} = nd^{16k}$,
& $\deg(G_k) = (d^{16})^8 = d^{16}$. □

- We have seen before that for any graph,
$$\lambda(G) \leq 1 - \frac{1}{8 \cdot D \cdot n^3} = 1 - \frac{1}{8 \cdot d^{16} \cdot n^3}.$$

- If $\lambda(G_{k-1}) =: 1 - \varepsilon$ then we have
$$\lambda(G_k) \leq \left(1 - \varepsilon \cdot \frac{1}{4}\right)^8$$

$$\Rightarrow 1 - \lambda(G_k) \geq 1 - \left(1 - \frac{\varepsilon}{4}\right)^8 = \binom{8}{1} \cdot \frac{\varepsilon}{4} - \binom{8}{2} \cdot \left(\frac{\varepsilon}{4}\right)^2 + \dots$$

- Note that for $\varepsilon < 1/2$, the above expression means that the spectral gap increases by a multiple of $\varepsilon/8$.

$$\Rightarrow \text{for } \ell = O(\lg n), \quad 1 - \lambda(G_\ell) \geq 1/2.$$

- Thus, G_ℓ has $O(\lg n)$ -length shortest paths & constant degree d^{16} .

- All that remains is to show that the walk in G_ℓ is possible in $O(\lg n)$ -space.

(Use Reingold's recursive data structure.)

□