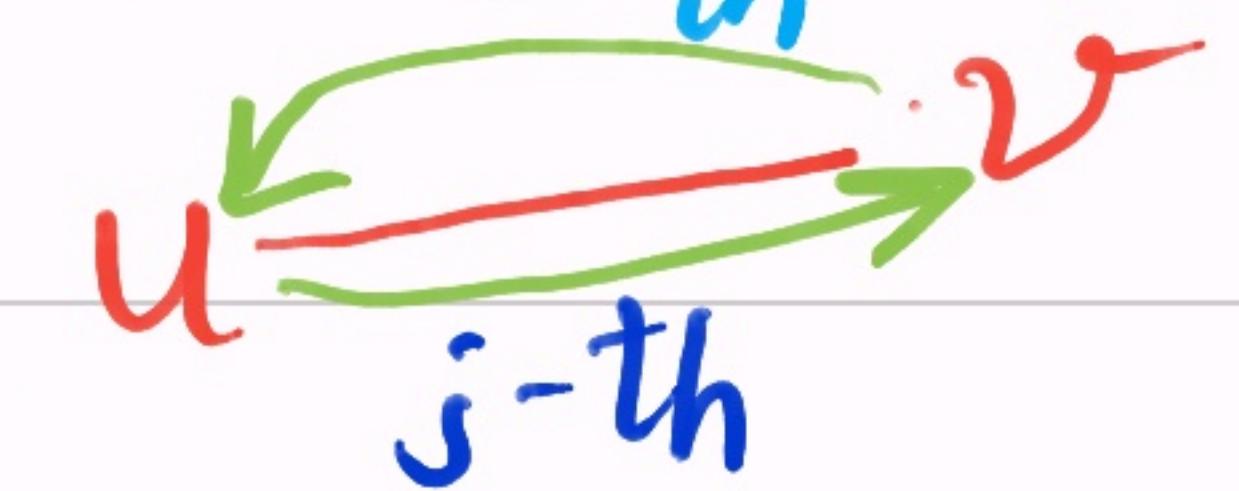


Explicit Expander Constructions

- We will construct an explicit expander family.
- Idea is to use 4 kinds of graph products.

Tradeoff: between degree & spectral-gap.
(we want constants)

- We'll need a new graph representation: $i\text{-th}$
- Defn: If G is an n -vertex d -degree connected graph then we label each neighbor of each vertex using $[d]$. Define a rotation map
- $$\widehat{G}: [n] \times [d] \rightarrow [n] \times [d]; \quad (v, i) \mapsto (u, j)$$
- 

where u is the i -th vertex of v & v j -th vertex of u .

► \hat{G} is a permutation of $[n] \times [d]$.

Pf: $\because \hat{G}$ is 1-1 & onto. \square

- Given G & G' (in the above representation) we'll define 4 products:

1. Path product
2. Tensor product
3. Replacement product
4. Zig-Zag product

► Combining them we'll improve expansion!

Matrix Product (or Path product)

Defn: For two n -vertex graphs G, G' with degrees d, d' ,
the graph $\underline{GG'}$ is the one with the normalized
adjacency matrix $A \cdot A'$.

Proposition: 1) AA' is stochastic.

2) GG' may have repeated edges, $\deg = dd'$,
 $(1 = \text{all-one}) \# \text{vertices} = n$.

Pf: 1) $AA' \cdot \underline{1} = A \cdot \underline{1} = 1 \Rightarrow AA'$ row-sum is 1.

\Rightarrow (by symmetry) AA' is stochastic.

2) $v \in V(G)$, v has d neighbors u . $u \in V(G')$ has d' neighbors.

\Rightarrow Neighbors of $u \in V(GG')$ are $d \times d'$. \square

Claim: $\lambda(GG') \leq \lambda(G) \cdot \lambda(G')$. as spectral gap increases

Proof: $\lambda(GG') = \max_{\substack{u \in \mathbb{T}^L \\ \text{spectral norm}}} \frac{\|AA'u\|}{\|u\|} = \max \frac{\|AA'u\|}{\|A'u\|} \cdot \frac{\|A'u\|}{\|u\|}$

$[\langle A'u, \bar{1} \rangle = \langle u, \bar{A}'\bar{1} \rangle = \langle u, \bar{1} \rangle = 0] \leq \lambda(A) \cdot \lambda(A')$. \square

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

- Thus, matrix product improves the spectral gap at the cost of the degree.

Tensor Product

Defn: The graph $\underline{G \otimes G'}$ is the one with the normalized adjacency matrix $\underline{A \otimes A'}$.

where, $A \otimes A' := \begin{pmatrix} A_{11} A' & \cdots & A_{1n} A' \\ \vdots & & \vdots \\ A_{n1} A' & \cdots & A_{nn} A' \end{pmatrix}_{nn' \times nn'}$.

Proposition: 1) $A \otimes A'$ is symmetric stochastic.

2) $G \otimes G'$ is nn' -vertex, dd' -degree.

Pf: 1) $(A \otimes A') \cdot 1_{nn'} = (A \otimes A') \cdot (1_n \otimes 1_{n'}) = A \cdot 1_n \otimes A' \cdot 1_{n'} = 1_n \otimes 1_{n'} = 1_{nn'}$.

2) n -th "row" is $(A_{1j} A', \dots, A_{nj} A')$. Each row has wt. $d \times d'$. \square

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

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 eigenvalues of
A & A' resp.

- $[(A \otimes A') \cdot (v \otimes v')] = (Av) \otimes (A'v')$ \Rightarrow eigenvalues
of $A \otimes A'$ are: $\{ \lambda_i \lambda'_j \mid (i, j) \in [n] \times [n'] \}$.
- Thus, the largest ones, apart from 1, are in:
 $\{ \lambda_i \mid i \in [2..n] \} \cup \{ \lambda'_j \mid j \in [2..n'] \}$.
 $\Rightarrow \lambda(G \otimes G') = \max(\lambda(G), \lambda(G'))$. □

Theorem: G, G' are $(n, d, \lambda), (n', d', \lambda')$ -expanders resp.
 $\Rightarrow G \otimes G'$ is $(nn', dd', \max(\lambda, \lambda'))$ -expander.

- Tensor product increases #vertices (& degree), while preserving the spectral-gap.

The Replacement Product

- We want to reduce the degree.
- This product is easier seen as a walk rather than a matrix operation.

Idea: Use an expander G' with D -vertex & d -degree, to pick the neighbor in a D -degree graph G . ($d \ll D$)

- This is similar to the idea to reduce random-bits.
 - This motivates the product $G \circledast G'$:

Defn: Let G, G' be graphs with vertices n, D & degrees D, d & normalized adj. matrices A, A' resp.

H := $G \circledast G'$ is (nD) -vertex graph s.t.

- i) $\forall u \in V(G)$, H has a copy of G' , say H_u , called a cloud. I.e. $\forall i \in V(G')$, $(u, i) \in V(H)$; and is called the i th vertex in u -th cloud.
- ii) For $(i, j) \in E(G')$, put $\forall u \in V(G), ((u, i), (u, j)) \in E(H)$.
- iii) If $\hat{G}(u, i) = (v, j)$ then put $((u, i), (v, j)) \in E(H)$.

Claim: H is nD -vertex, (dH) -degree.

Pf: $\circ V(H) = V(G) \times V(G')$.

- Neighbors of $(u, i) \in V(H)$ are d -many in the cloud H_u . Plus, one more given by $\hat{G}(u, i)$.
 $\Rightarrow \deg(H) = d+1.$ □

- Replacement product reduces degree D to $d+1 < D$.

Zig-Zag Product

- Idea: Consider length-3 paths that zig-zag the clouds. ($\text{cloud} \rightarrow \text{out} \rightarrow \text{another cloud}$)

Defn: On the vertex-set $V(G \circledR G')$ define zig-zag
 $H := \underline{G \circledR G'}$ s.t. $((u, i), (v, j)) \in E(H)$ if $\exists l, k$ with
 $((u, i), (u, l)), ((u, e), (v, k)), ((v, k), (v, j)) \in E(G \circledR G')$.

$\triangleright G \circledast G'$ is nD -vertex, d^2 -degree. $[d^2 \ll D]$

Pf: • #neighbors of (u, i) in $G \circledast G' = d \times 1 \times d = d$. \square

\triangleright Its normalized adj. matrix is $\underline{A \circledast A'} := B \hat{A} B$,
where $\underline{\hat{A}}[(u, i), (v, j)] := \begin{cases} 1, & \text{if } \hat{G}(u, i) = (v, j) \\ 0, & \text{else} \end{cases}$

& $\underline{B}[(u, i), (v, j)] := \begin{cases} A'[i, j], & u = v \\ 0, & u \neq v \end{cases}$

Pf: • $A \circledast A'$ encodes the defn. of $G \circledast G'$. \square

• Note: $(A \circledast A') \cdot 1 = B \hat{A} B \cdot 1 = B \hat{A} \cdot 1 = B \cdot 1 = 1$.

$\triangleright A \circledast A'$ is stochastic symmetric.

Spectral analysis of Zig-Zag

Theorem (Reingold, Vadhan, Wigderson '02): $\lambda(G) = a$ & $\lambda(G') = b \implies \lambda(G \otimes G') \leq a + 2b + b^2$.

Proof: • Let $M := A \otimes A'$.

• Recall: $M = B \hat{A} B$; \hat{A} is permutation & $B = I_n \otimes A'$.

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

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

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

• Each of these four we'll upper-bound the matrix norm: $\lambda(A) := \|A\| := \max_{x \in \mathbb{T}^L} \|Ax\| / \|x\|$. or 2nd largest eigenval.

• Let $L(A)$:= $\max_{\mathbf{x}} \|\mathbf{Ax}\| / \|\mathbf{x}\|$ or largest eigenval

$$\triangleright \|\bar{\mathbf{E}}\| \leq L(\bar{\mathbf{E}}) = L(A' - J/D).$$

• Let $\lambda_1, \dots, \lambda_D$ be the eigenvalues of A' s.t.

$$1 = \lambda_1 \geq |\lambda_2| \geq \dots \geq |\lambda_D|.$$

\triangleright Write $A' = \sum_{i \in [D]} \lambda_i \cdot v_i v_i^T$, for orthonormal eigenvectors $\{v_i\}_i$ of A' .

$$\Rightarrow \|\bar{\mathbf{E}}\| \leq L(A' - J/D) = L(A' - \lambda_1 \cdot v_1 v_1^T) = L\left(\sum_{i>1}^D \lambda_i \cdot v_i v_i^T\right)$$

$$\left[\left\| \sum_{i>1} \lambda_i \cdot v_i v_i^T \cdot \bar{\mathbf{x}} \right\|^2 = \left\| \sum_{i>1} \lambda_i \cdot v_i \cdot \langle v_i, \bar{\mathbf{x}} \rangle \right\|^2 = \sum \left\| \lambda_i \cdot v_i \cdot \langle v_i, \bar{\mathbf{x}} \rangle \right\|^2 \right]$$

$$\text{Say, } \bar{\mathbf{x}} = \sum \alpha_i v_i \Rightarrow \text{RHS} = \sum_{i>1} \lambda_i^2 \cdot \alpha_i^2 \leq \lambda_2^2.$$

$$\leq |\lambda_2| = \|A'\| = b.$$

- Going back, we're by triangle-inequality:

$$\begin{aligned}\|M\| &\leq \|\bar{J} \hat{A} \bar{J}\| + \|\bar{J} \hat{A} \bar{E}\| + \|\bar{E} \hat{A} \bar{J}\| + \|\bar{E} \hat{A} \bar{E}\| \\ &\leq " + \|\bar{J}\| \cdot \|\hat{A}\| \cdot \|\bar{E}\| + \|\bar{E}\| \cdot \|\hat{A}\| \cdot \|\bar{J}\| + \|\bar{E}\| \cdot \|\hat{A}\| \cdot \|\bar{E}\|.\end{aligned}$$

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

$$\begin{aligned}&\leq \|\bar{J} \hat{A} \bar{J}\| + 1 \cdot 1 \cdot b + b \cdot 1 \cdot 1 + b \cdot 1 \cdot b \\ &= " + (2b + b^2).\end{aligned}$$

- Now consider $\bar{J} \hat{A} \bar{J}$:

$\triangleright \bar{J} \hat{A} \bar{J} = A \otimes J/D.$

(Why?)

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

$$= (A \otimes J/D)_{(u,i), (v,j)}.$$

□

• Note that the eigenvalues of J are $\{1, 0, 0, -1, 0\}$. (Why?)

$$[\because J \cdot \bar{T} = \bar{T} \text{ & } \forall \bar{x} \in \bar{T}^\perp, J \cdot \bar{x} = 0.]$$

$$\Rightarrow \|A \otimes JD\| = \lambda(A) = \|A\| = a.$$

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

$$\Rightarrow \|M\| \leq (a + 2b + b^2).$$

D

Theorem: G, G' are (n, D, λ_1) , (D, d, λ_2) -expanders

$\Rightarrow G \circledast G'$ is $(nD, d^2, \lambda_1 + 2\lambda_2 + \lambda_2^2)$ -expander.

- Thus, ' \circledast ' reduces the degree dramatically (if $d \ll D$),
without increasing the λ too much!

The Construction

- Using the products we give a strongly explicit family of expanders,
i.e. given (u, i) , the i -th neighbor of u , is $\text{poly}(t_s, |V|)$ -time computable.

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

Pf: • We'll recursively construct $\{G_k\}_{k \geq 1}$ st. G_k has $2^{O(k)}$ vertices.

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

[H could be found randomly, or by known constructions.]

- Let $G_1 := H^2$ be $(d^8, d^2, 0.04^2)$ -expander.
- Let $G_2 := G_1$.
- For odd $k \geq 3$, $G_{k+1} := G_k := \left(G_{\frac{k-1}{2}} \otimes G_{\frac{k-1}{2}}\right)^2 \circledast H$.

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

Pf: • True for $k=1$.

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

• $\deg(G_{\frac{k-1}{2}}^{\otimes 2})^2 = (d^{2 \times 2})^2 = d^8$.

$\Rightarrow \deg(\text{---} \circledast H) = d^2$.

• Finally, $\lambda(G_k) \leq (0.1)^2$

$$+ 2 \times (0.04) + (0.04)^2 < 0.1. \quad \square$$

The algorithm to find neighbors, in G_K , of $v \in V(G_K)$ is also recursive:

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

\Rightarrow Listing a row in $G_{\frac{K-1}{2}}^{\otimes 2}$ takes $2 \cdot T\left(\frac{k-1}{2}\right)$.
is d^4 -sparse

\Rightarrow Listing all the needed d^4 -rows in $G_{\frac{K-1}{2}}^{\otimes 2}$ (in zig-zag product) takes $d^4 \cdot 2 \cdot T\left(\frac{k-1}{2}\right)$ time

\Rightarrow Listing a row in G_K takes time $T(k) = O(d^4 \cdot T\left(\frac{k-1}{2}\right))$.

$\Rightarrow T(k) = O(d^{4+4k}) = O(k^{4+4d}) = \text{poly}(k)$
 $= \text{poly.log}(\#V(G_K))$ [$\because d$ is constant]. \square

- To derandomize "Upath $\in RL$ ", we'll need operations to make input G an expander, with changing the connectivity!
- To use zig-zag product we'll need another estimate on $\lambda(G \otimes G')$.
- For the input graph, $\lambda(G)$ may be large in general ($\approx 1 - 1/\text{poly}(n)$).
To make it small, we prove a different bound.

Theorem (Reingold, Trevisan, Vadhan'05): $\lambda(G) = 1 - \varepsilon$ &
 $\lambda(G') = 1 - \delta$ $\Rightarrow \lambda(G \otimes G') \leq 1 - \varepsilon \delta^2$ \leftarrow multiplicative

Proof: • As before, $M := A \otimes A' = B \hat{A} B$, where $B = I_n \otimes A'$
& \hat{A} is rotation map of G .

• Express A' as $\delta \cdot J/D + (1-\delta) \cdot C$, where $J = \text{all-one}$
& C is a matrix st. $\|C\| \leq 1$. $\bar{x} \in T^\perp$

[Pf: • $A' = \sum_{i \in [D]} \lambda_i \cdot v_i v_i^T \Rightarrow \| (A' - \delta \cdot J/D) \cdot \bar{x} \|$ ↗
 $= \| ((1-\delta) \cdot J/D + (1-\delta) \cdot v_2 v_2^T + \dots) \cdot \bar{x} \|$
 $\leq \| (1-\delta) \cdot v_2 \cdot \langle v_2, \bar{x} \rangle + \lambda_3 \cdot v_3 \cdot \langle v_3, \bar{x} \rangle + \dots \|$
 $\leq (1-\delta) \cdot \|\bar{x}\|$ $\left[\because v_i \text{'s form orthonormal basis} \right]$
 $\Rightarrow \left\| \frac{1}{(1-\delta)} \cdot (A' - \delta \cdot J/D) \right\| \leq 1.$ $\square \quad \boxed{\quad}$

$$\Rightarrow B = I_n \otimes (sJ_D + (1-s)C) =: s\bar{J} + (1-s)\bar{C}$$

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

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

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

$$\bar{J} := I_n \otimes J_D \quad \& \quad \bar{C} := I_n \otimes C.$$

$\triangleright \|F\| \leq 1$. [Pf: $\|\bar{C}\| \leq 1$, $\|\bar{J}\| \leq 1$ & \hat{A} is permutation.]

$$\Rightarrow M = s^2 \cdot (A \otimes J_D) + (1-s^2) \cdot F$$

$$\Rightarrow \lambda(M) = \|M\| \leq s^2 \cdot \|A \otimes J_D\| + (1-s^2) \cdot \|F\|$$

$$\leq \underbrace{s^2 \cdot (1-\varepsilon)}_{-\text{re part}} + \underbrace{(1-s^2) \cdot 1}_{\text{close to } 1!} = 1 - \varepsilon s^2.$$

□

-re part

close to 1!

Theorem (Reingold'05): $\text{Upath} \in \text{L}$.

Proof: • Let \underline{G} be the given graph; undirected, \underline{n} -vertex ($s = \text{start-vertex}$).

Idea: Apply the graph-products on \underline{G} to get \tilde{G} st. the connected component of s , in \tilde{G} , is an expander (with $\lambda, d = \text{constant}$).

\Rightarrow Shortest-paths in this connected component of \tilde{G} have length $\leq O(\lg n)$.

\Rightarrow Guessing in logspace possible!

• Let \underline{d} be a large constant st. $(d^{16}, d, 0.5)$ -expander H exists.

- Wlog assume G to be regular of $\deg = D \stackrel{:=}{=} d^{16}$. (connected)
- Now, transform G as follows:

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

* compute only locally, in logspace.

$\triangleright G_k$ is (nd^{16k}, d^{16}) -graph.

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

• $\deg(G_k) = (d^2)^8 = d^{16}$. \square

- Recall that for any connected graph G ,

$$\lambda(G) \leq 1 - \frac{1}{8Dn^3} = 1 - \left(\frac{1}{8d^{16} \cdot n^3}\right).$$

$$\triangleright \lambda(G_{k_1}) = 1 - \underline{\varepsilon} \Rightarrow \lambda(G_k) \leq \left(1 - \frac{\varepsilon}{4}\right)^8$$

$$\begin{aligned} \Rightarrow 1 - \lambda(G_k) &\geq 1 - \left(1 - \frac{\varepsilon}{4}\right)^8 = 8 \cdot \left(\frac{\varepsilon}{4}\right) - \frac{8 \cdot 7}{2} \cdot \left(\frac{\varepsilon}{4}\right)^2 + \dots \\ &= 2\varepsilon - \frac{7}{4}\varepsilon^2 + \dots = \varepsilon \cdot \left(2 - \frac{7\varepsilon}{8}\right) + \dots \end{aligned}$$

[Note: $\varepsilon < \frac{1}{2} \Rightarrow \dots > \varepsilon \cdot \left(2 - \frac{7}{16}\right) = \varepsilon \cdot \frac{25}{16} > \varepsilon$.]

$$\triangleright 1 - \lambda(G_k) > (1 - \lambda(G_{k_1})) \cdot (\text{constant} > 1).$$

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

$\Rightarrow \triangleright G_\ell$ has $O(\lg n)$ -length shortest paths & $\deg = d^{16}$.

Exercise: Do the walk in G_ℓ in logspace.

[Use Reingold's recursive data structure.]