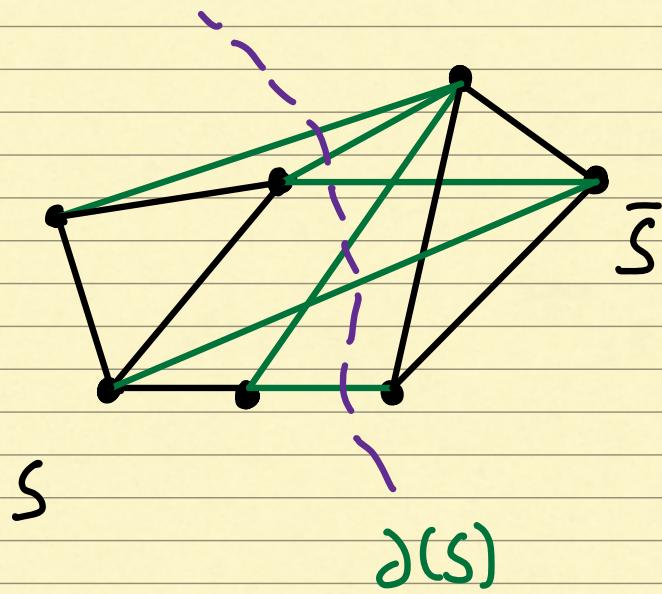


An Invitation to Expander Graphs

(Fernando Granha Jeronimo)
IAS

Unicamp and USP
(Brazil)



Plan

1) What is an expander?

2) Warm up: Expander Mixing Lemma

3) Some Applications

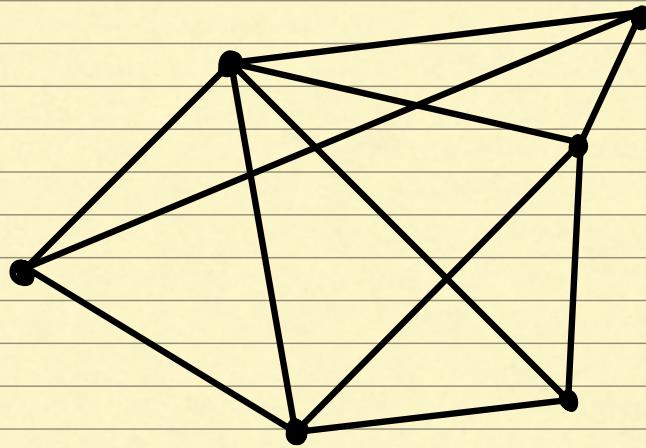
Plan

- 1) What is an expander?
- 2) Warm up: Expander Mixing Lemma
- 3) Some Applications
- 4) Zig-Zag Product
- 5) Near-optimal Expanders

What is an expander?

Informal Def: A well connected yet sparse graph

Graph $G = (V, E)$



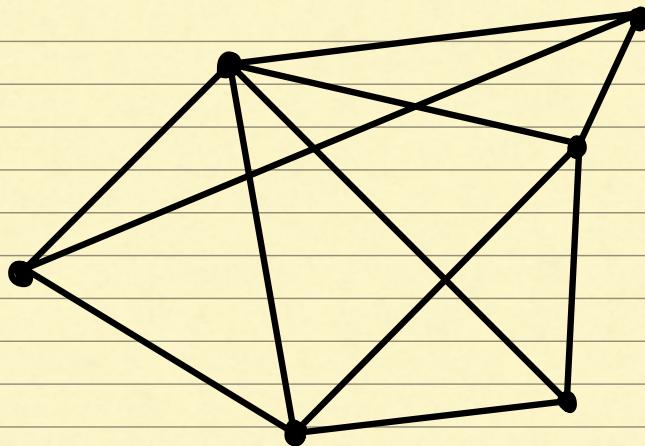
What is an expander?

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"perspectives"

- Combinatorial
- Algebraic
- Random Walk

Graph $G = (V, E)$



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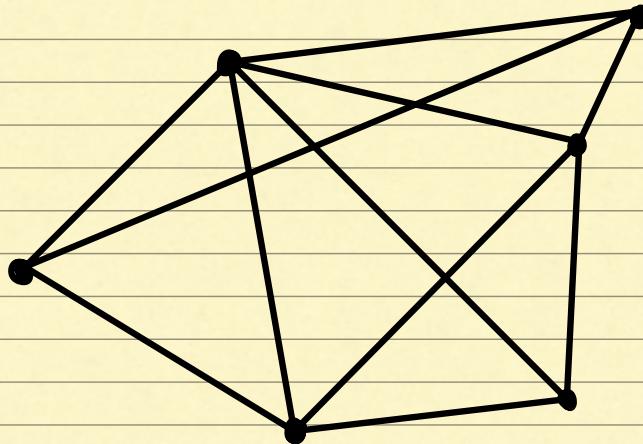
"perspectives"

- Combinatorial
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Graph

$$G = (V, E)$$

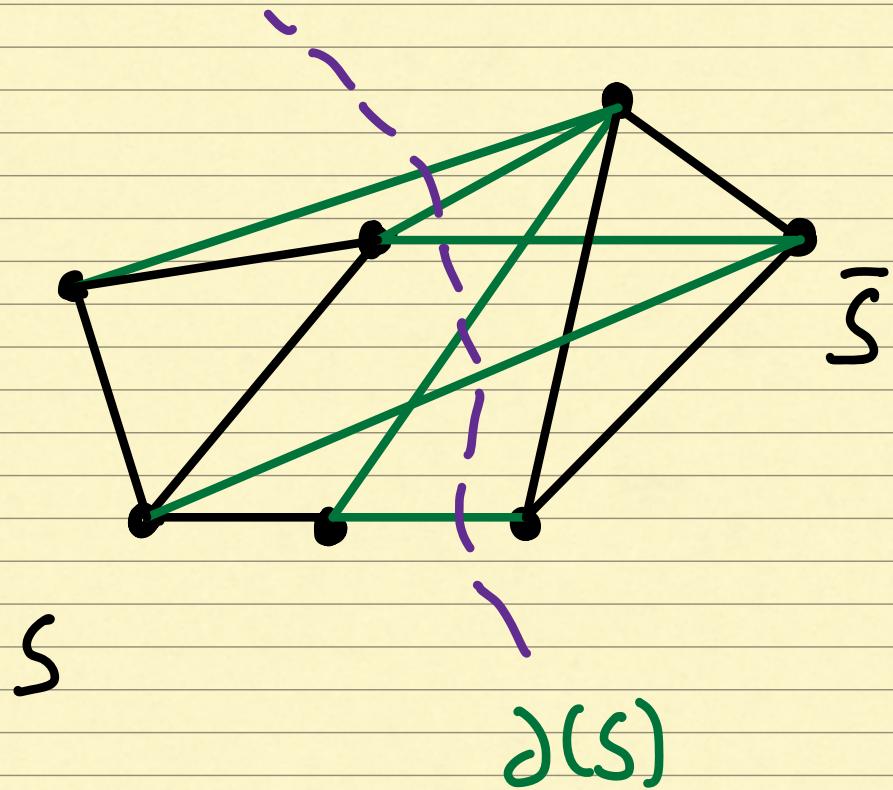
$|E|$ small



Expanders: Combinatorial Def.

Graph $G = (V, E)$ d -regular n -vtx

"No Small Cuts"



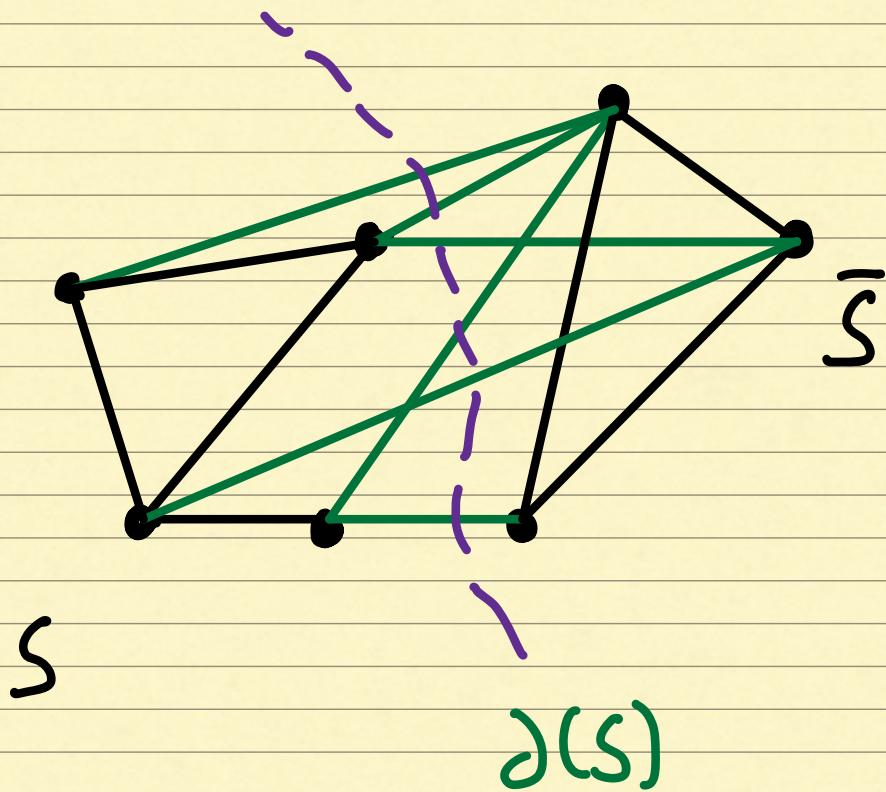
Expanders: Combinatorial Def.

Graph $G = (V, E)$ d -regular n -vtx

"No Small Cuts"

(Edge) Boundary of $S \subseteq V$

$$\partial(S) := \{(s, v) \in E \mid s \in S, v \in \bar{S}\}$$

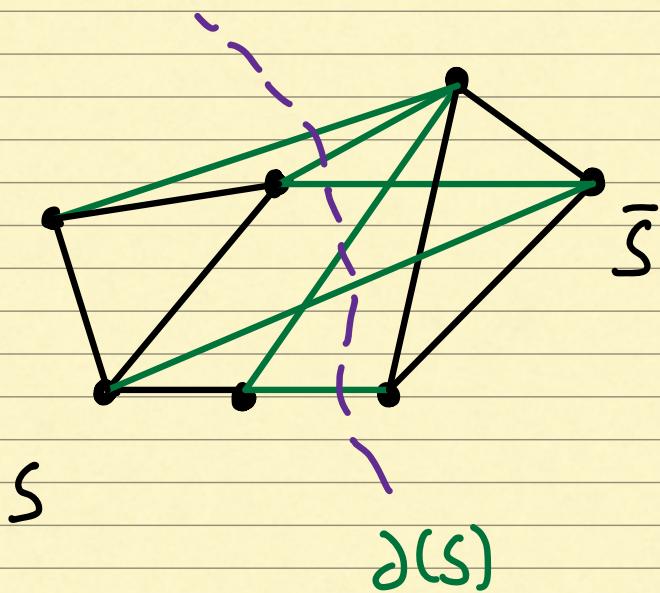


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Conductance of S

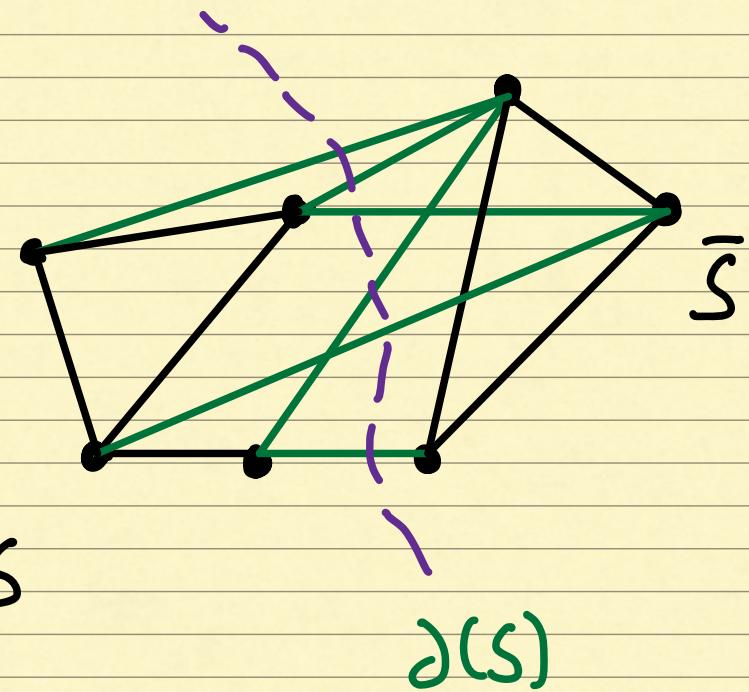
$$\Phi_G(S) := \frac{|\partial(S)|}{d|S|}$$

Expanders: Combinatorial Def.

Graph $G = (V, E)$ d -regular n -vtx

Comb. Expansion

$$\Phi(G) := \min_{\substack{S \subseteq V \\ 1 \leq |S| \leq \frac{n}{2}}} \Phi_G(S)$$

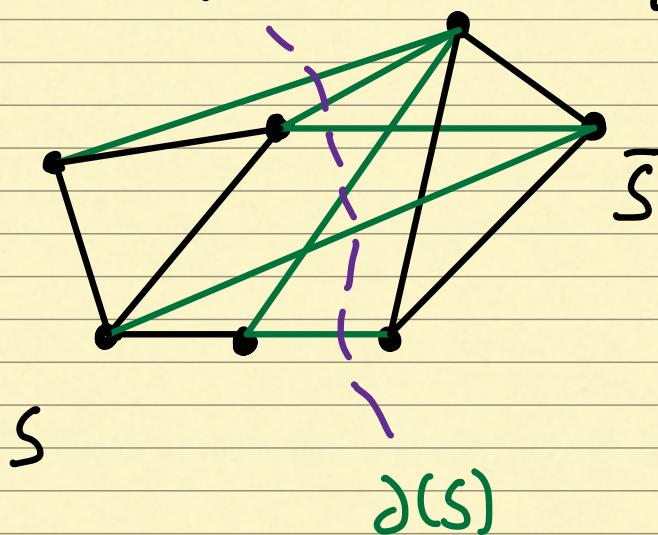


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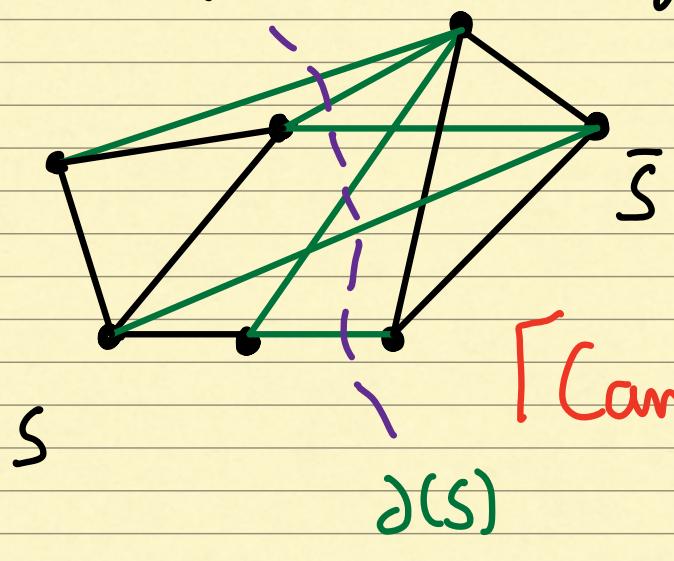
$$\bar{\Phi}(G) := \min_{\substack{S \subseteq V \\ 1 \leq |S| \leq \frac{n}{2}}} \bar{\Phi}_G(S)$$

Want $\bar{\Phi}(G) \geq \epsilon > 0$ for a

constant ϵ

Expanders: Combinatorial Def.

Graph $G = (V, E)$ d -regular n -vtx



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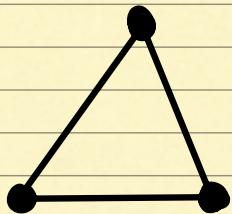
Expanders: Algebraic Def.

Graph $G = (V, E)$ d -regular n -vtx

Adjacency Matrix $A \in \mathbb{R}^{n \times n}$ of G

$$A_{u,v} = \begin{cases} 1 & [u \sim_v v] \\ 0 & \text{otherwise} \end{cases}$$

Ex:



G

$$\begin{pmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = 2 \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

A

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

$$A \vec{I} = d \vec{I}$$



Expanders: Algebraic Def.

Graph $G = (V, E)$ d -regular n -vtx

Adjacency Matrix $A \in \mathbb{R}^{n \times n}$ of G

$$A_{u,v} = \begin{cases} 1 & [u \sim_v v] \\ 0 & \text{otherwise} \end{cases}$$

A is symmetric (over \mathbb{R})



Spectral Theorem



$$A = \sum_{j=1}^n \lambda_j v_j v_j^\top \quad \text{with } \{v_j\}_{j \in [n]} \text{ ONB}$$

$$\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$$

|| ~~λ_1~~

d

$$v_1 = \underbrace{1}_{d}$$

Expanders: Algebraic Def.

Graph $G = (V, E)$ d -regular n -vtx

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$$A_{u,v} = \begin{cases} 1 & [u \sim_v v] \\ 0 & \text{otherwise} \end{cases}$$

$$A = \sum_{j=1}^n \lambda_j v_j v_j^\top$$

$$\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$$

||
d

$\lambda_1 = \frac{1}{\sqrt{n}}$

Algebraic Expansion

$$\lambda(G) := \max \{ |\lambda_2|, |\lambda_n| \}$$

Want $\underline{\lambda(G)}$ small $\ll d$

Expanders: Algebraic Def.

Graph $G = (V, E)$ d -regular n -vtx

Adjacency Matrix $A \in \mathbb{R}^{n \times n}$ of G

$$A = \sum_{j=1}^n \lambda_j v_j v_j^\top \quad \lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$$

Algebraic Expansion

$$\lambda(G) := \max\{|\lambda_2|, |\lambda_n|\}$$

Key Algebraic Perspective

$$A = \underbrace{\frac{d}{n} \vec{I} \vec{I}^\top}_{\text{adj. matrix of complete graph with self loops}} + \lambda E$$

error term $\|E\|_{op} \leq 1$

adj. matrix of complete graph
with self loops

Expanders: Algebraic Def.

Graph $G = (V, E)$ d -regular n -vtx

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Key Algebraic Perspective

$$A = \underbrace{\frac{d}{n} \vec{I} \vec{I}^\top}_{\text{adj. matrix of complete graph with self loops}} + \lambda E$$

error term

$$\|E\|_{op} \leq \frac{1}{n}$$

adj. matrix of complete graph
with self loops

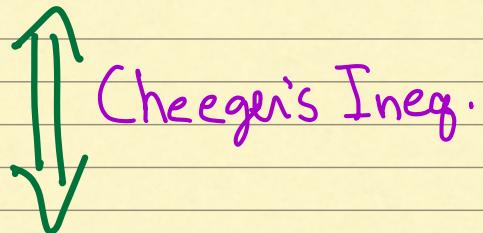
$$A \approx \frac{d}{n} \bar{J}$$

complete graph
with density $\frac{d}{n}$

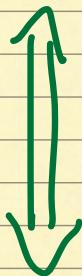
Expander Definitions

Equivalent Views

Combinatorial: $\Phi(G)$ large



Algebraic: $\lambda(G)$ small



Random Walk: Fast Mixing

Warm up: Expander Mixing Lemma

Graph $G = (V, E)$ (n, d, λ) -graph

Suppose we want to compute $|E(S, \bar{S})|$?
(approx)

$$|E(S, \bar{S})| = \mathbf{1}_S^\top A \mathbf{1}_{\bar{S}}$$

Warm up: Expander Mixing Lemma

Graph $G = (V, E)$ (n, d, λ)

Suppose we want to compute $|E(S, \bar{S})|$?
(approx)

$$|E(S, \bar{S})| = \mathbf{1}_S^T A \mathbf{1}_{\bar{S}}$$

Recall that $A = \frac{d}{n} J + \lambda E$

$$(A \approx \frac{d}{n} J)$$

$$|E(S, \bar{S})| \approx \frac{d}{n} \mathbf{1}_S^T J \mathbf{1}_{\bar{S}} = \frac{d}{n} |S| |\bar{S}|$$

Warm up: Expander Mixing Lemma

Graph $G = (V, E)$ (n, d, λ)

$$|E(S, \bar{S})| = \underbrace{\mathbf{1}_S^\top A \mathbf{1}_{\bar{S}}}_{\text{in } \mathbb{R}^n}$$

Recall that $A = \frac{d}{n} J + \lambda E$

$$(A \approx \frac{d}{n} J)$$

$$|E(S, \bar{S})| \approx \frac{d}{n} |S| |\bar{S}|$$

More rigorously,

$$\|E\|_{op} \leq 1$$

$$|E(S, \bar{S})| = \frac{d}{n} |S| |\bar{S}| + \lambda \underbrace{\mathbf{1}_S^\top E \mathbf{1}_{\bar{S}}}_{\text{error term}}$$

$$\pm \lambda \sqrt{|S||\bar{S}|}$$

error term

Some Applications

- Coding Theory
- Complexity Theory
- Hardness of Approximation

:
:
:

Some Applications

- Coding Theory
- Complexity Theory
- Hardness of Approximation
- Pseudorandomness
- Sampling and Counting
- Algorithm Design
- Property Testing
- Metric Spaces
- Group Theory
- Number Theory

⋮

Some Applications

- Coding Theory
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- Group Theory
- Number Theory
- ⋮
- Your new application

Some Applications

(an appetizer)

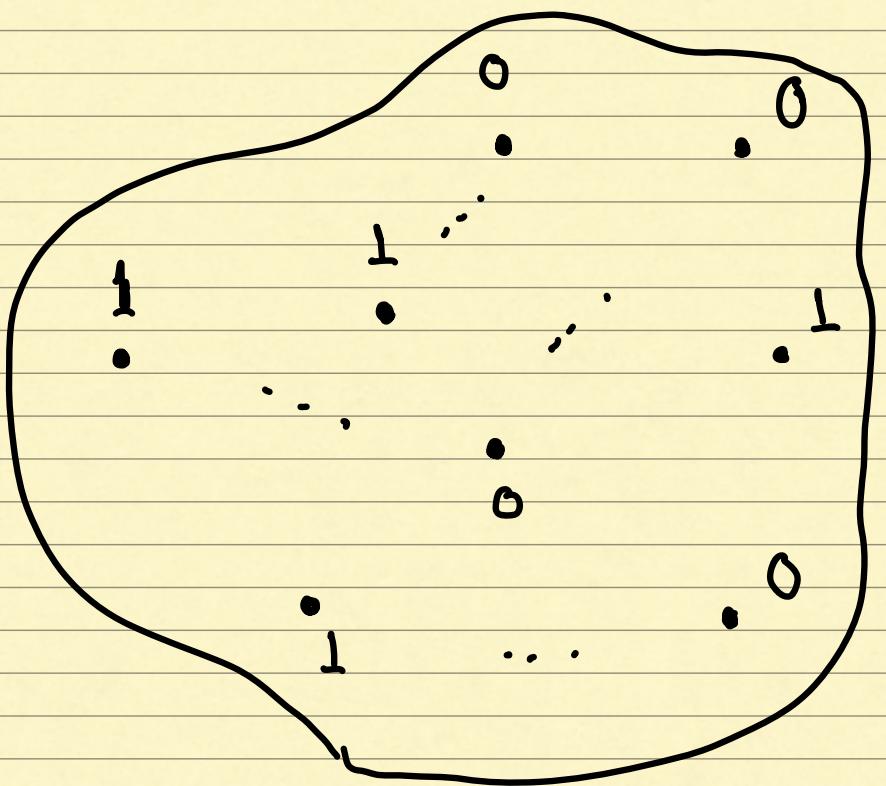
- Derandomization
- Complexity Theory

How does expansion appear?

Chernoff Bound

n O/I values

$$\mu := \frac{\#\text{ of } 1's}{n}$$



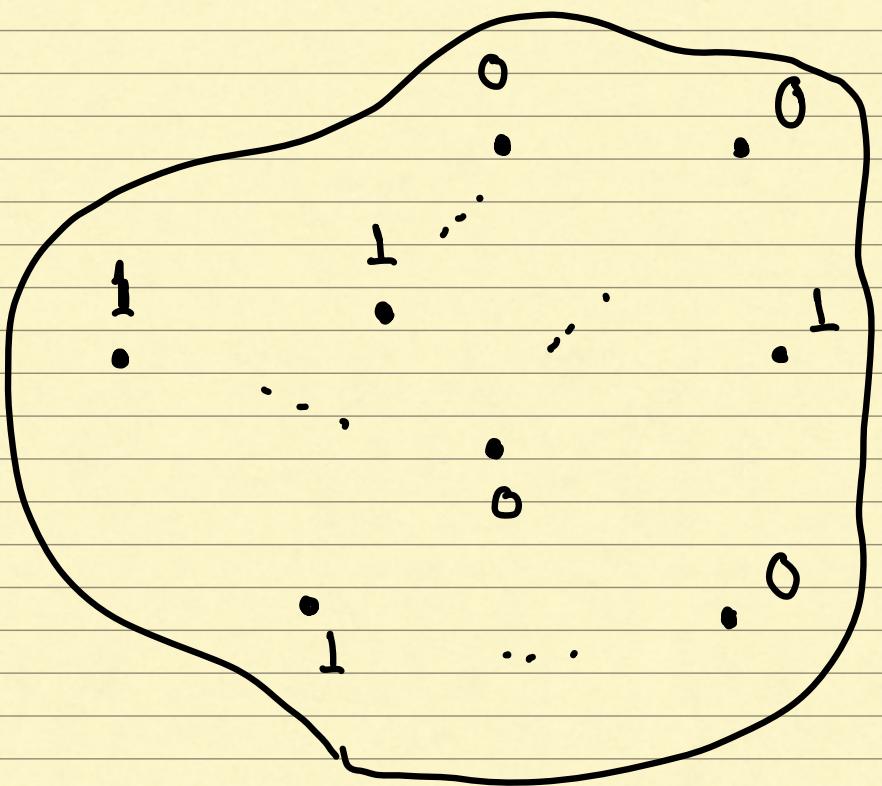
Sample k independent values uniformly

$$X_1, \dots, X_k$$

Chernoff Bound

n 0/1 values

$$\mu := \frac{\# \text{ of } 1's}{n}$$



Sample k independent values uniformly

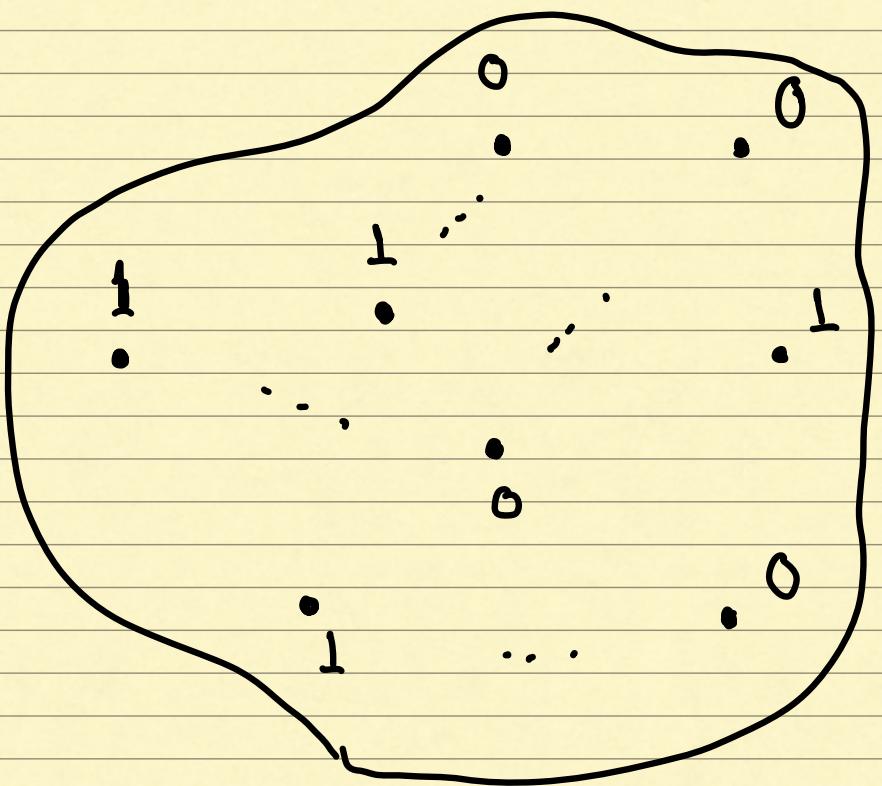
$$X_1, \dots, X_k$$

$$\Pr \left[\left| \frac{1}{k} \sum X_i - \mu \right| > \varepsilon \right] \leq \exp(-\Omega(\varepsilon^2 k))$$

Chernoff Bound

n O/I values

$$\mu := \frac{\# \text{ of } 1's}{n}$$



Sample k independent values uniformly

$$X_1, \dots, X_k$$

$$\Pr \left[\left| \frac{1}{k} \sum X_i - \mu \right| > \varepsilon \right] \leq \exp(-\Omega(\varepsilon^2 k))$$

$$\# \text{ of random bits} \approx K \cdot \log_2(n)$$

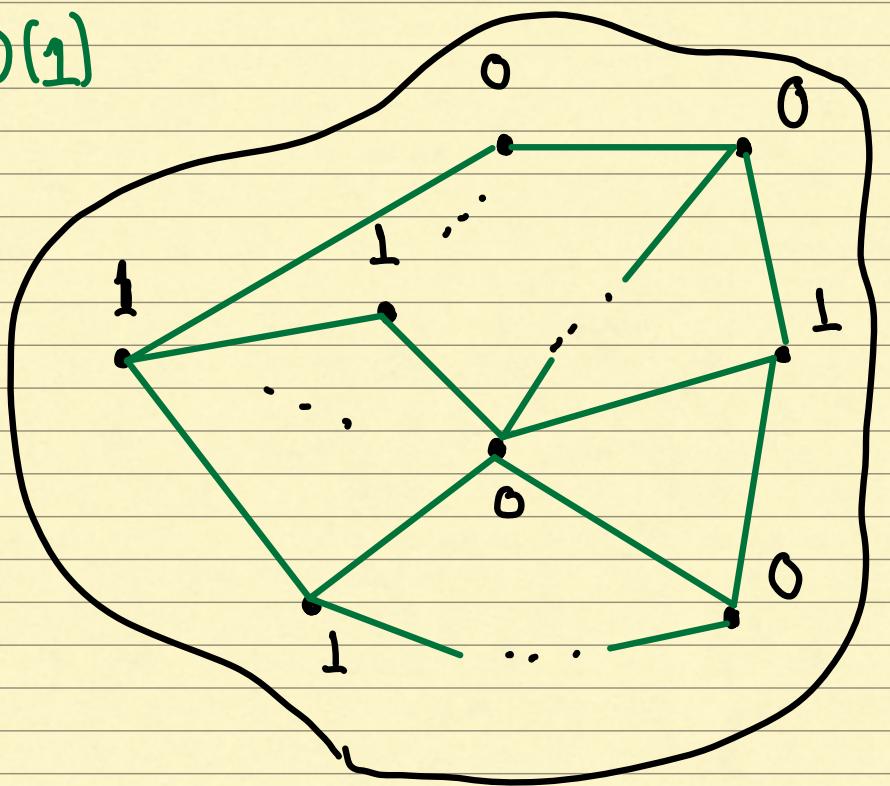
Expander Chernoff Bound

$$G = (V=[n], E)$$

d -regular $d = O(1)$

$$\mu := \frac{\# \text{ of } 1's}{n}$$

n 0/1 values



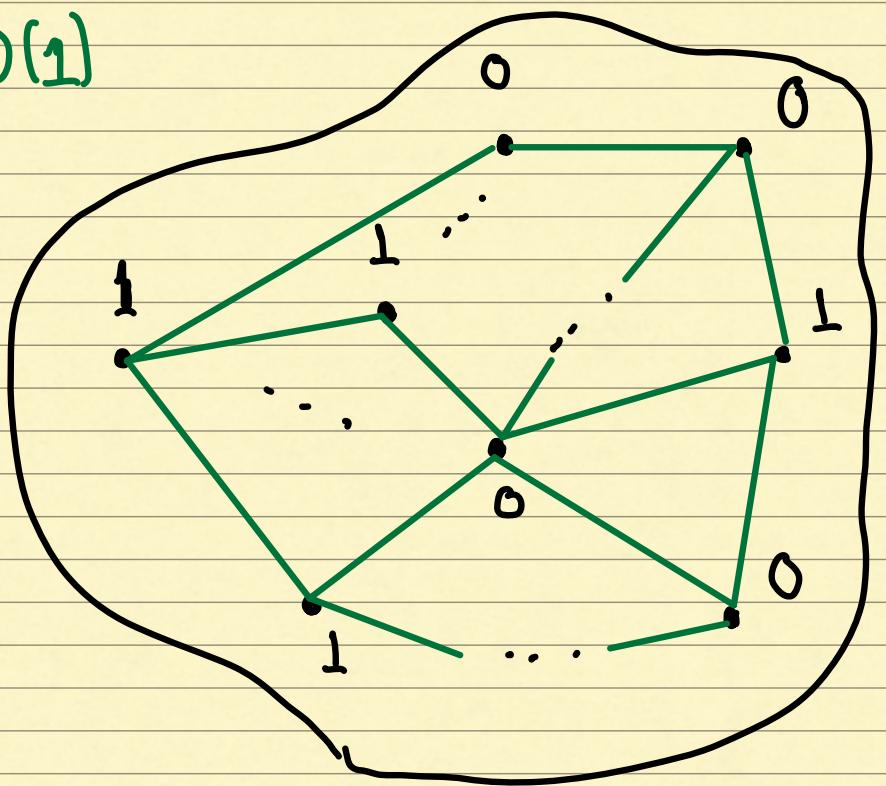
Expander Chernoff Bound

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Sample k values along a random walk

X_1, \dots, X_k on G

Expander Chernoff Bound

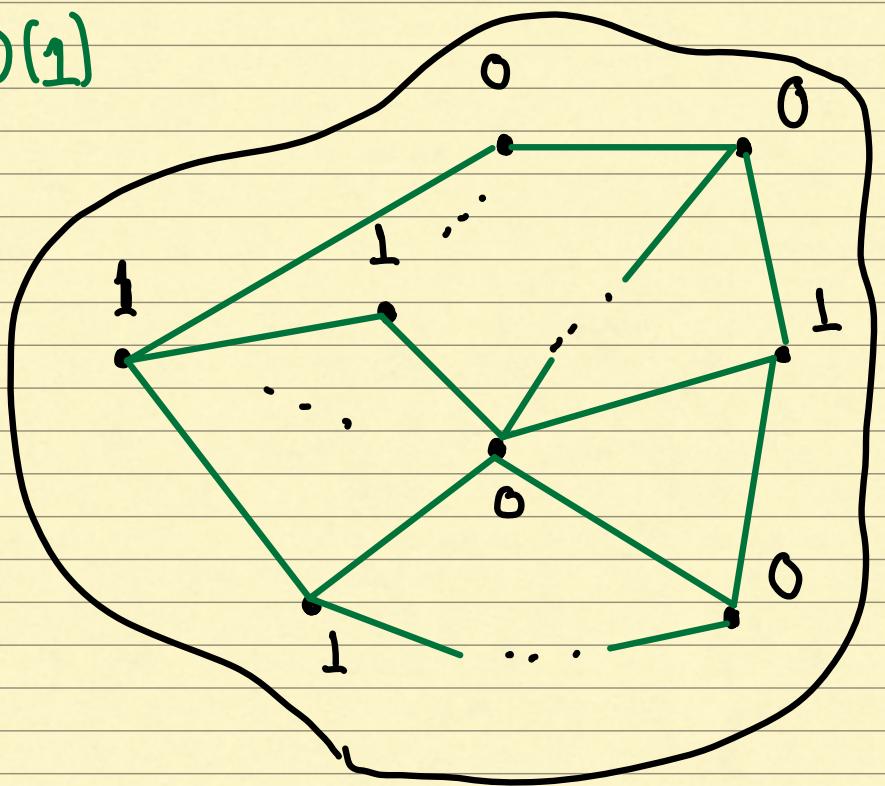
$$G = (V=[n], E)$$

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$$\lambda(G) \leq \lambda$$

$$\mu := \frac{\# \text{ of } 1's}{n}$$

n 0/1 values



Sample k values along a random walk

$$X_1, \dots, X_k$$

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Expander Chernoff Bound

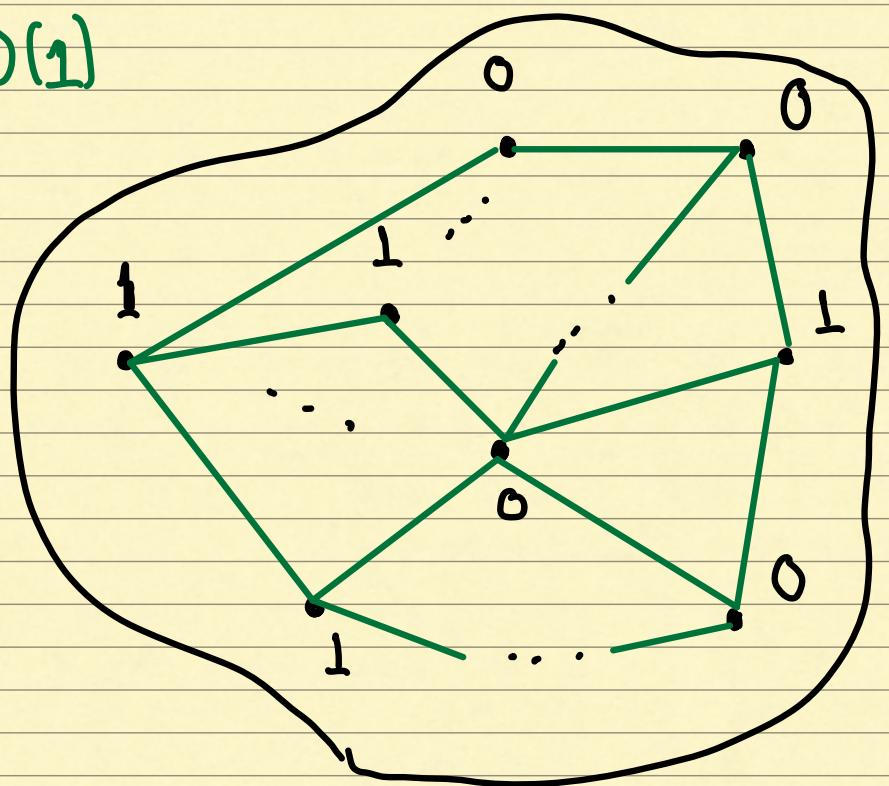
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n 0/1 values



Sample k values along a random walk

X_1, \dots, X_k on G

Theorem [Gillman '93]

$$\Pr \left[\left| \frac{1}{k} \sum X_i - \mu \right| > \varepsilon \right] \leq \exp(-\lambda(\varepsilon^2 k))$$

of random bits $\approx \cancel{k \cdot \log_2(n)} + O(k)$

Complexity: $SL = L$

Input: $G = (V, E)$ an n -vtx graph,
 $s, t \in V$.

Question: \exists $s-t$ path in G ?

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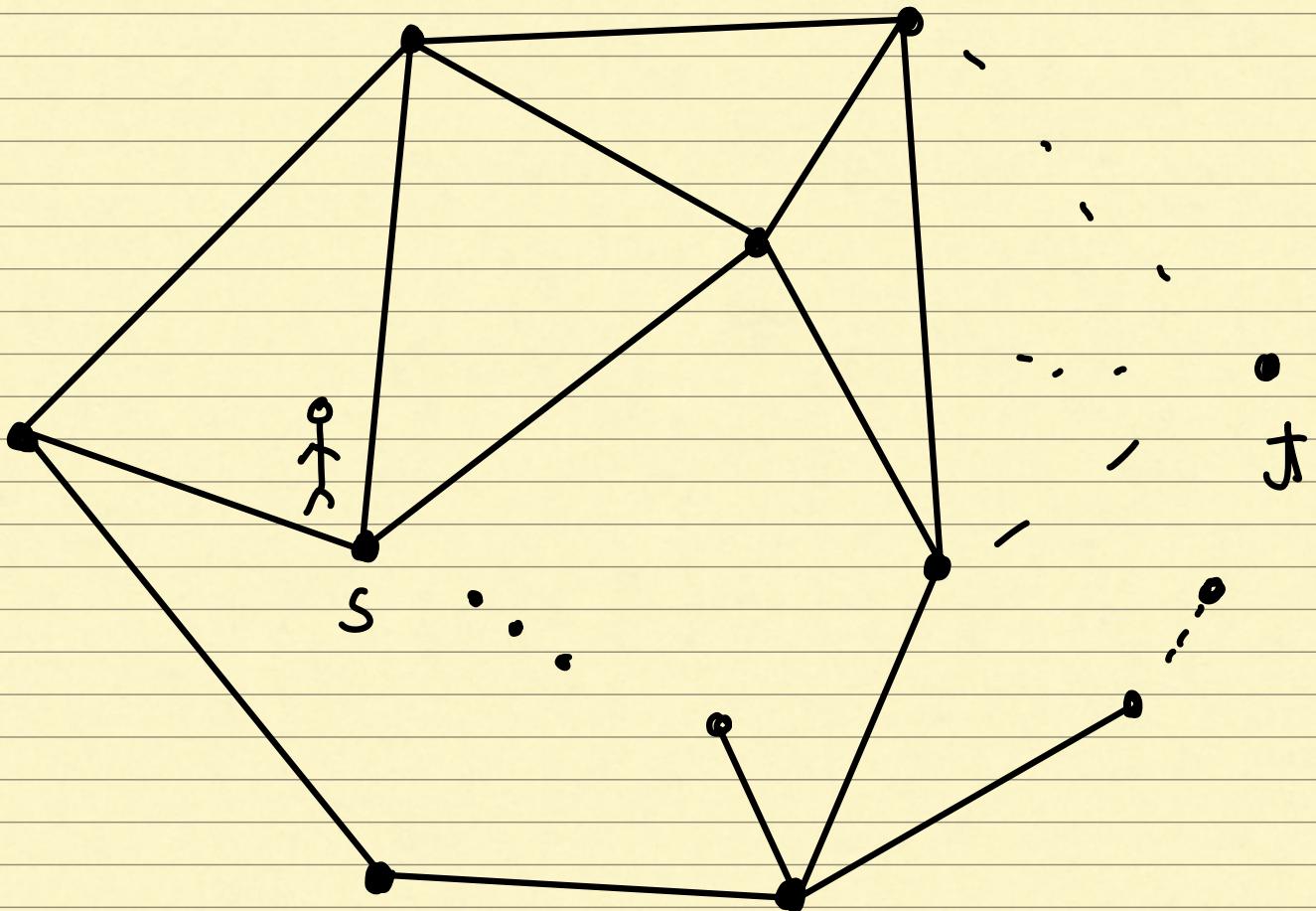
Req: deterministic algorithm with
 $O(\log n)$ memory

Complexity: $SL = L$

Input: $G = (V, E)$ an n -vtx graph,
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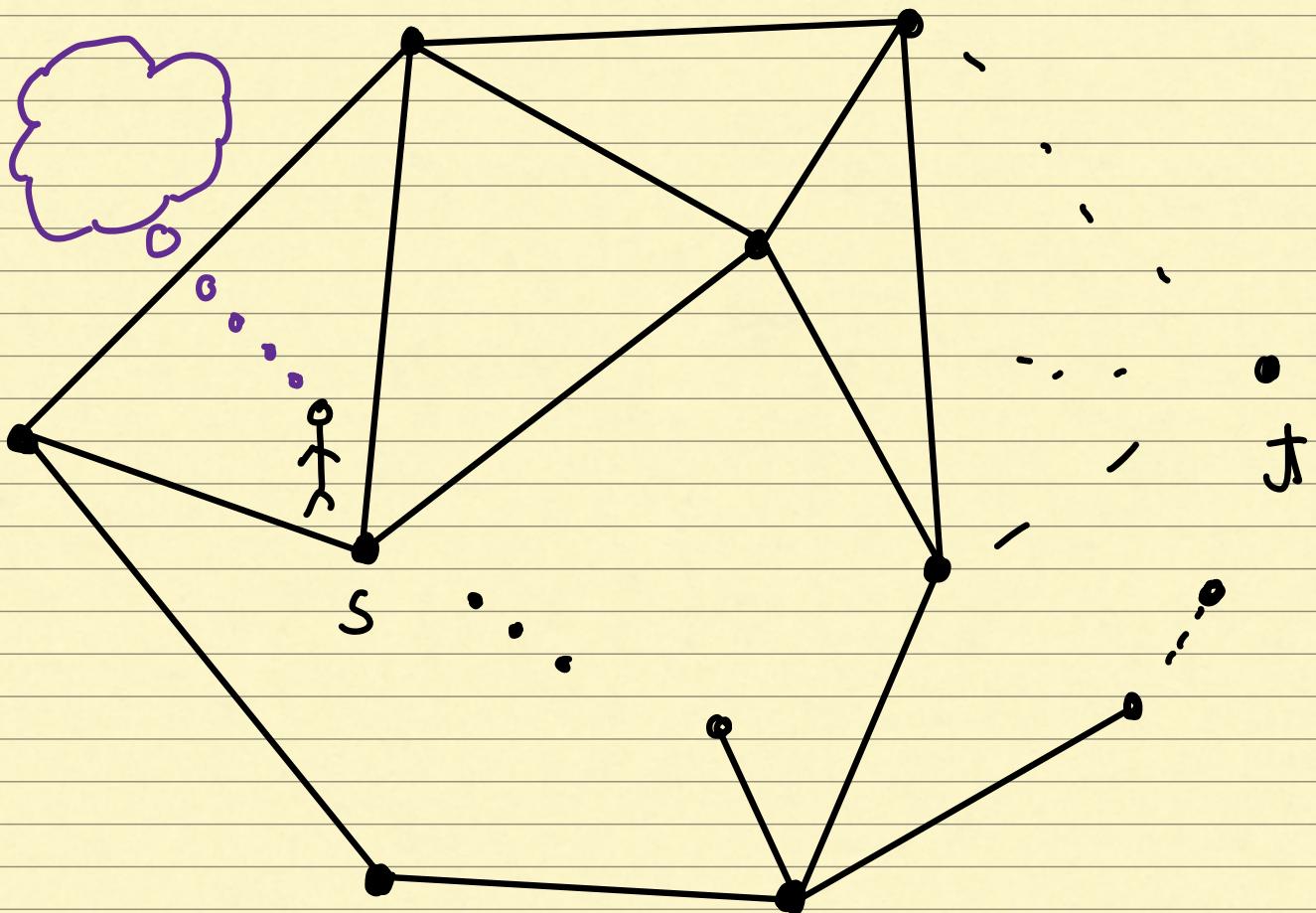
Complexity: $SL = L$

Input: $G = (V, E)$ an n -vert graph,
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Question: \exists $s-t$ path in G ?

Req: deterministic algorithm with
 $O(\log n)$ memory

We can only remember $O(1)$ vertices



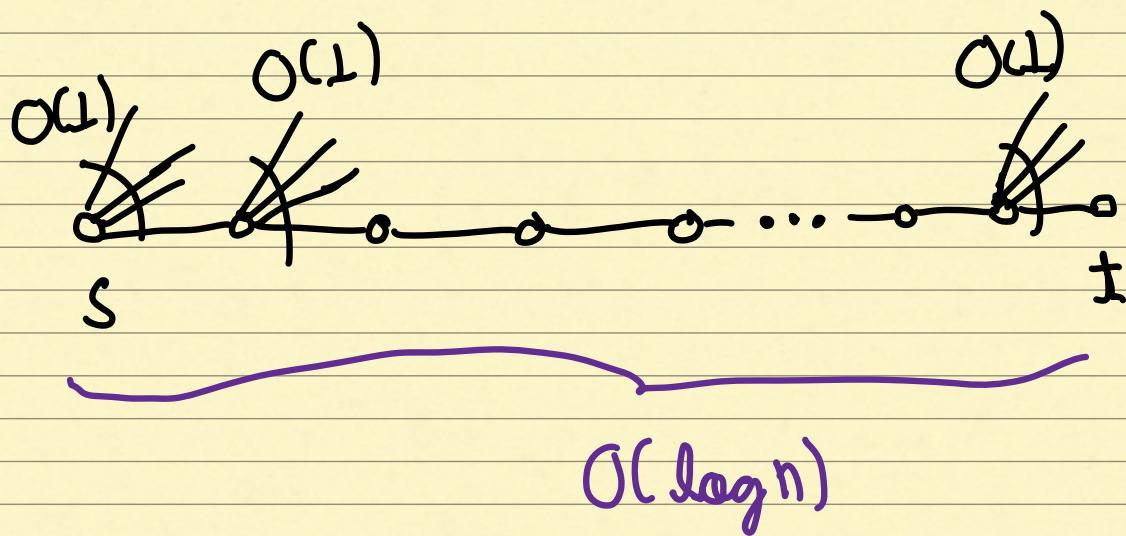
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Input: $G = (V, E)$ an n -vtx graph,
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Observation: This is easy if G is a
constant degree expander
(diameter is $O(\log n)$)



Complexity: $SL = L$

Input: $G = (V, E)$ an n -vertx graph,
 $s, t \in V$.

Question: \exists $s-t$ path in G ?

Req: deterministic algorithm with
 $O(\log n)$ memory

This can be done!

Transform (connected comp.) of G into expander!

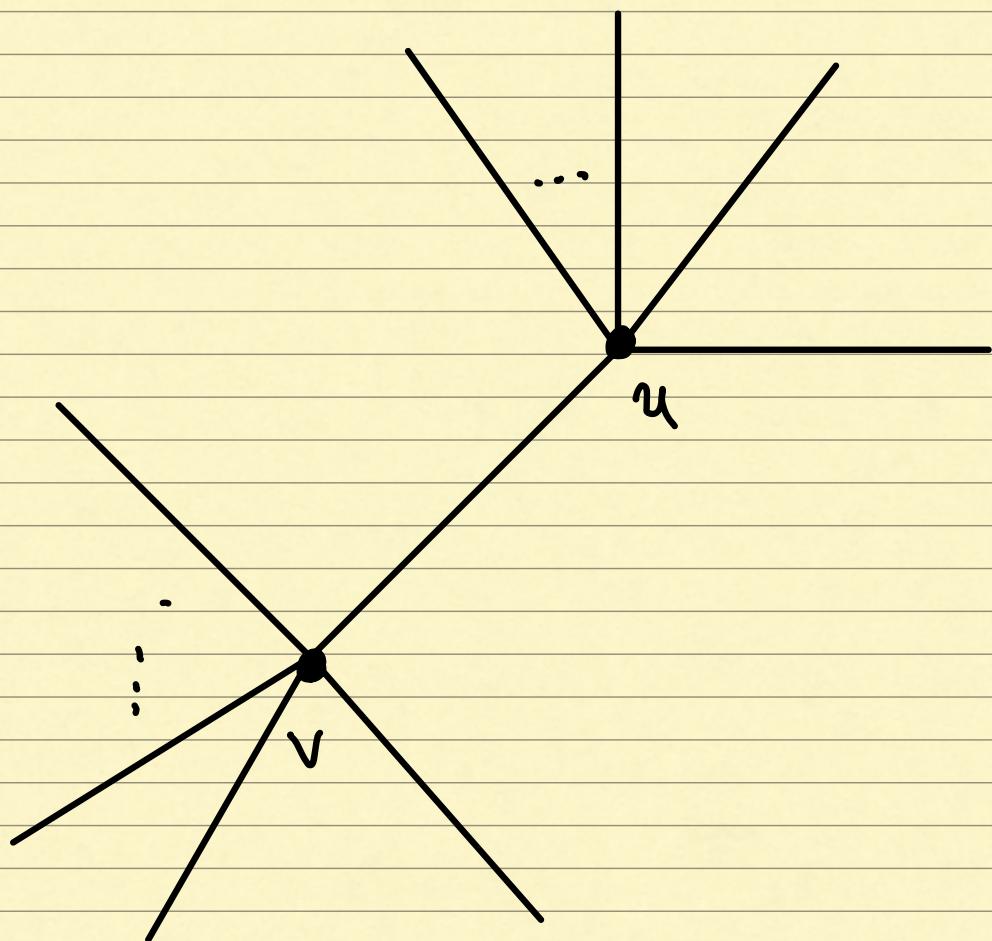
Theorem [Reingold'05] $SL = L$

Key Technique: Zig-Zag product

Towards Zig-Zag

First a simpler goal: Degree Reduction

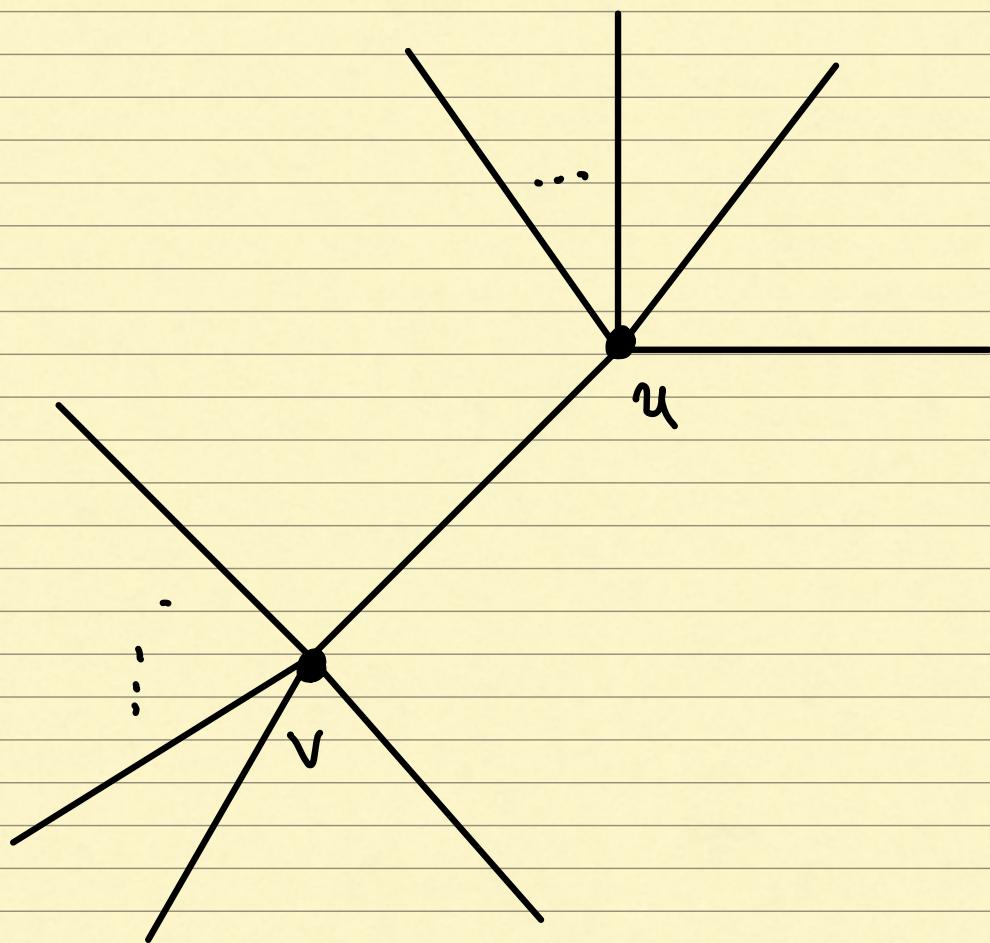
Let G be n -vtx degree D ("large")



Towards Zig-Zag

First a simpler goal: Degree Reduction

Let G be n -vtx degree D ("large")

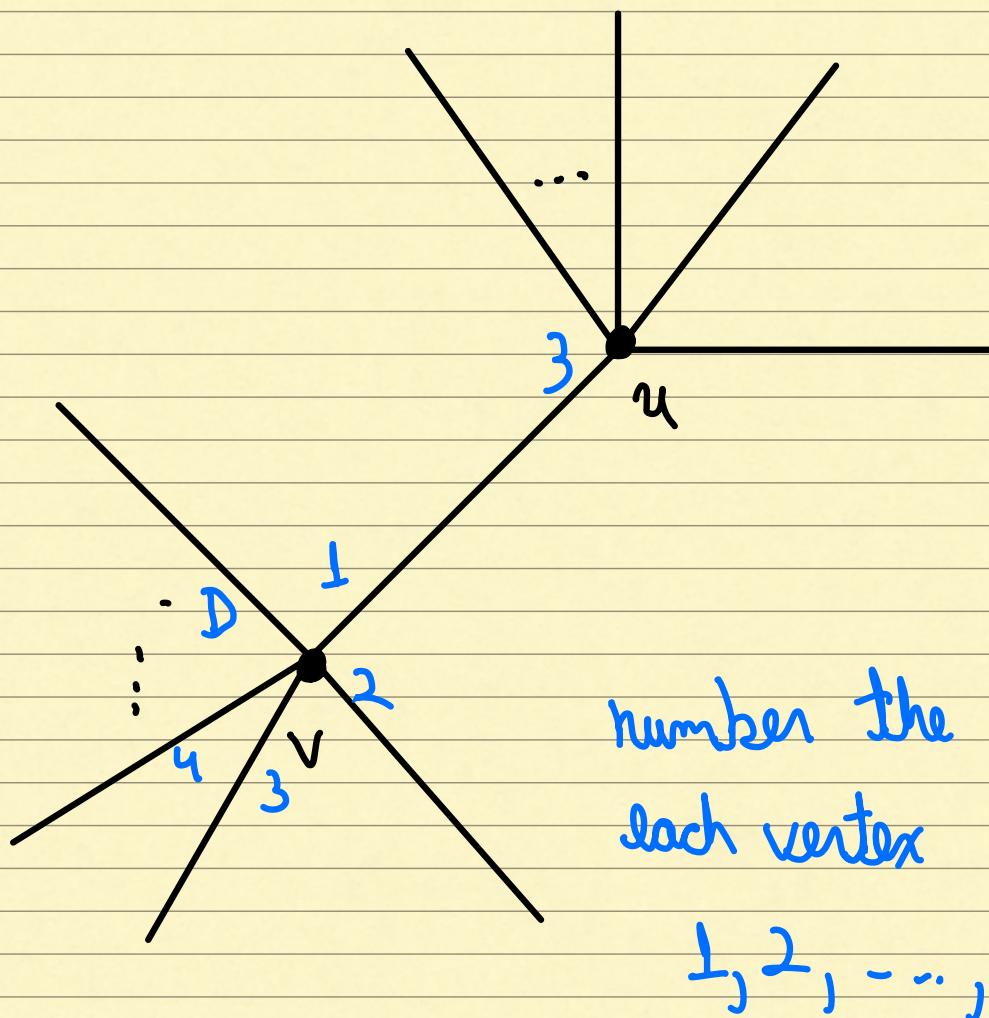


Mint: Use a D -vtx graph H of degree $d \ll D$

Towards Zig-Zag

First a simpler goal: Degree Reduction

Let G be n -vtx degree D



number the edges of
each vertex using
 $1, 2, \dots, D$

Hint: Use a D -vtx graph H of degree $d \ll D$

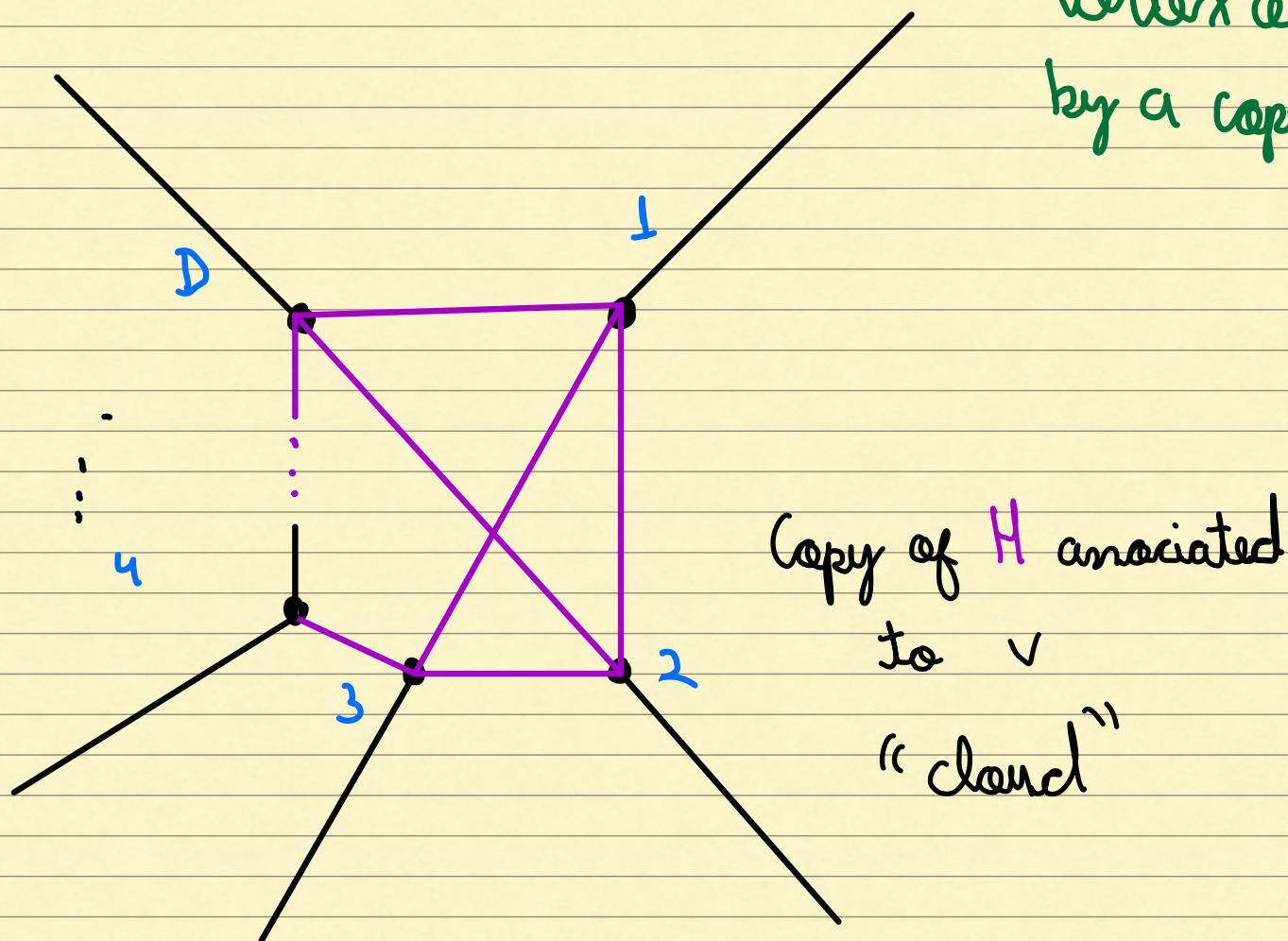
$$H = (\underline{V = [D]}, E)$$

Towards Zig-Zag

First a simpler goal: Degree Reduction

Let G be n -vtx degree D

Replace each vertex of G by a copy of H



$$H = (V=[D], E)$$

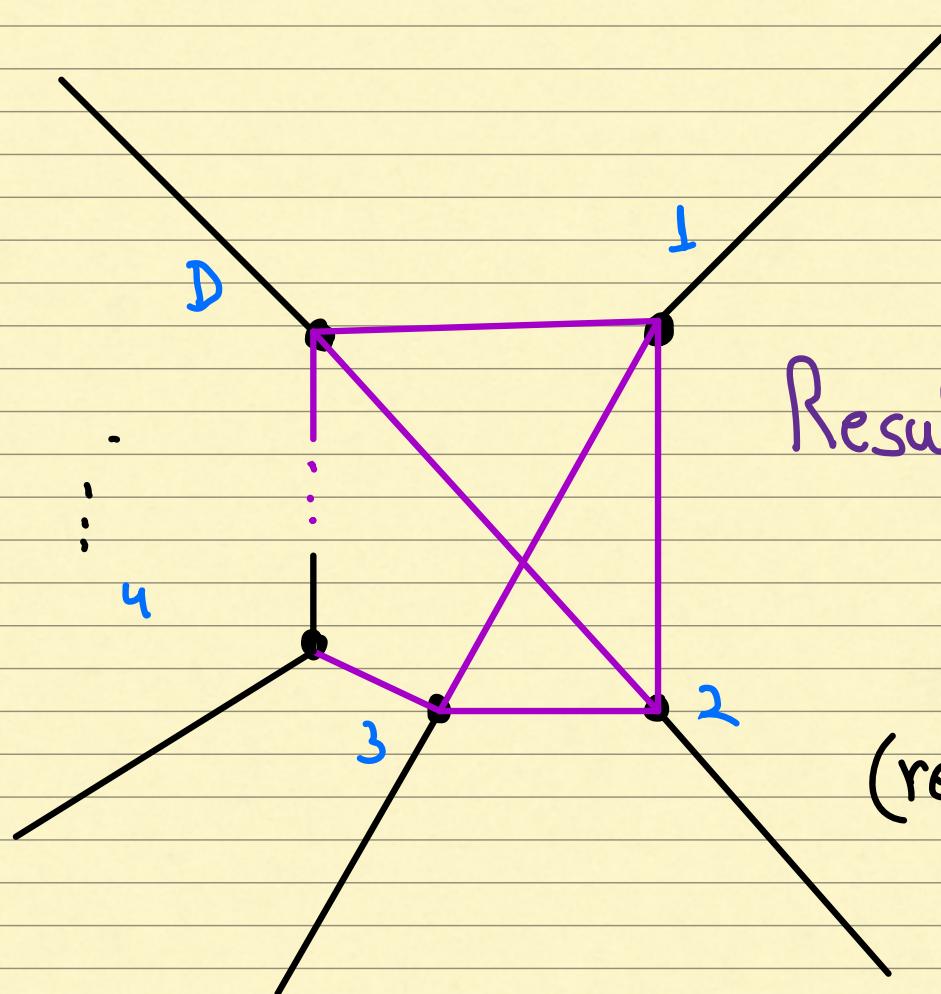
degree $d \ll D$

Towards Zig-Zag

First a simpler goal: Degree Reduction

Let G be n -vtx degree D

Replace each vertex of G by a copy of H



Resulting Graph:

$G \circledast H$

(replacement product)

deg: $d+1$



$$H = (V=[D], E)$$

degree $d \ll D$

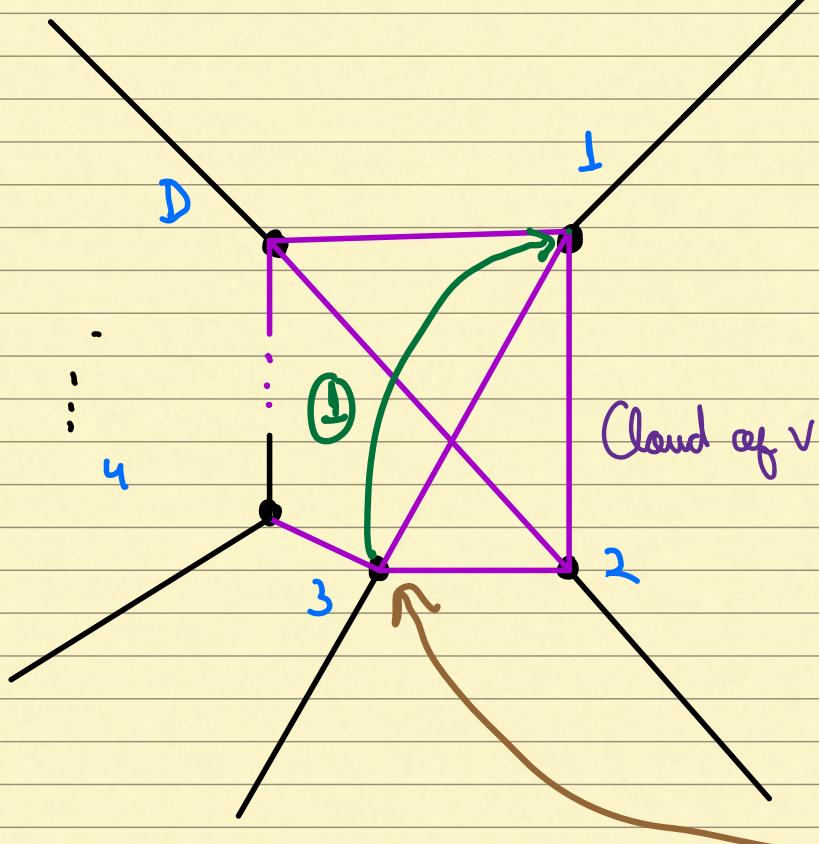
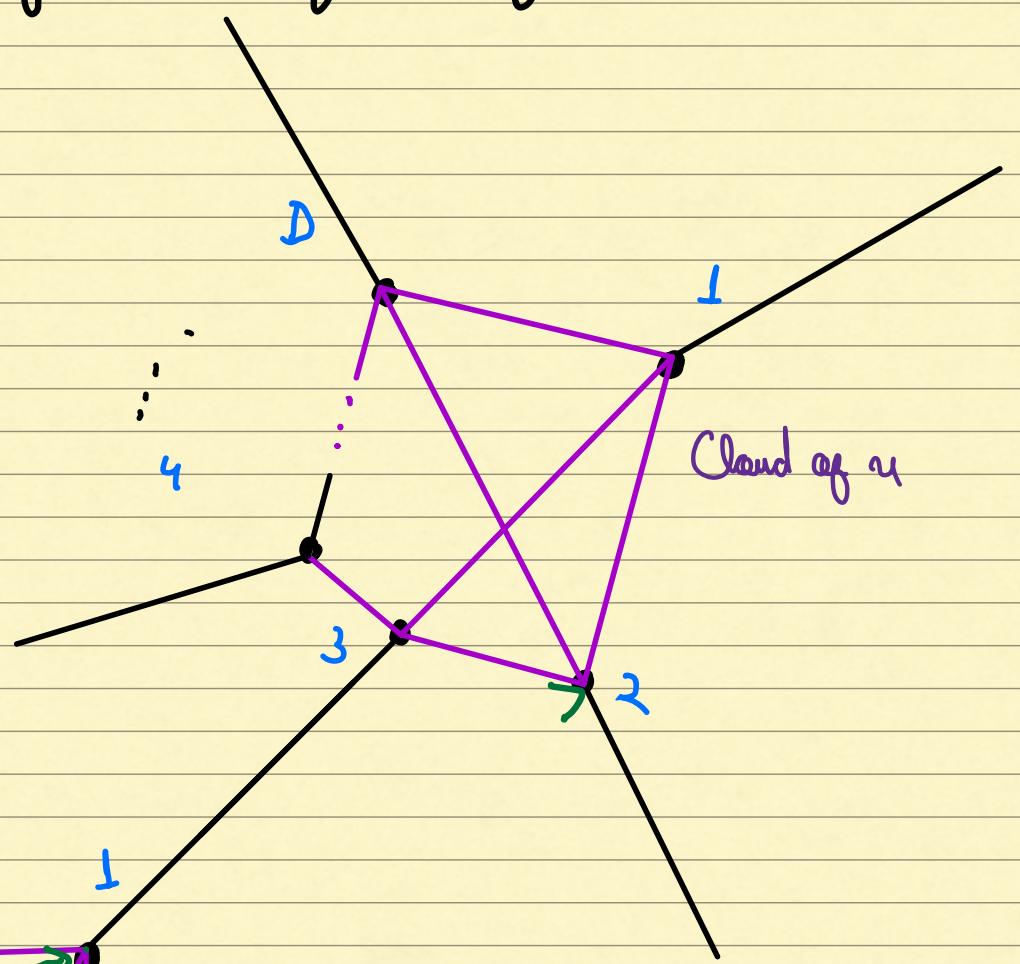
Zig-Zag Product

$G \otimes H$ is obtained from $G \oplus H$
by a "zig-zag walk"

Zig-Zag Product

$G \otimes H$ is obtained from $G \circledast H$
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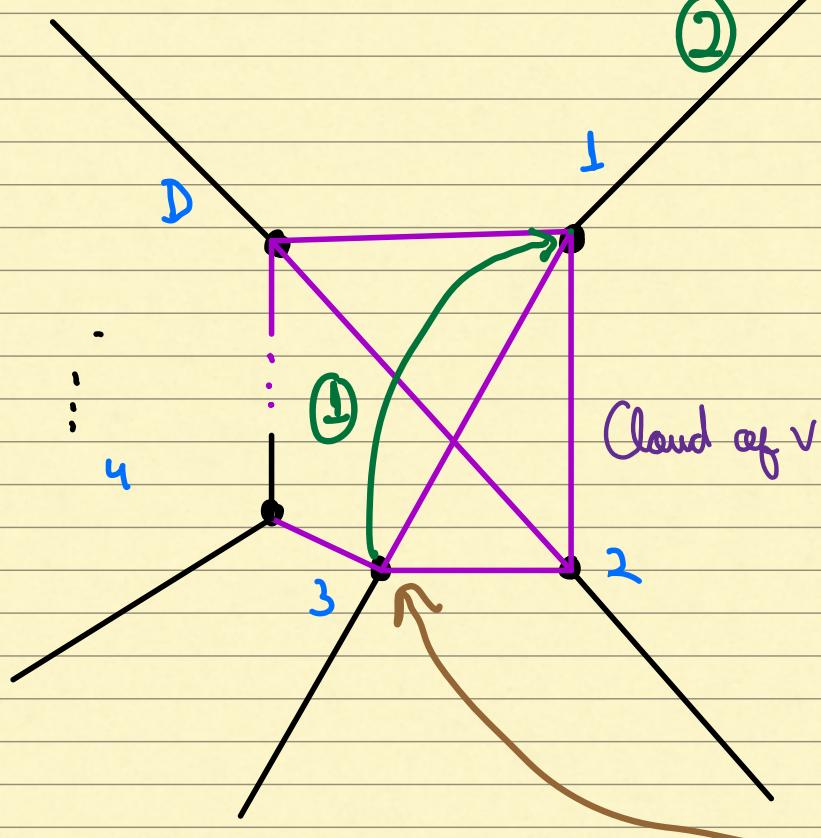
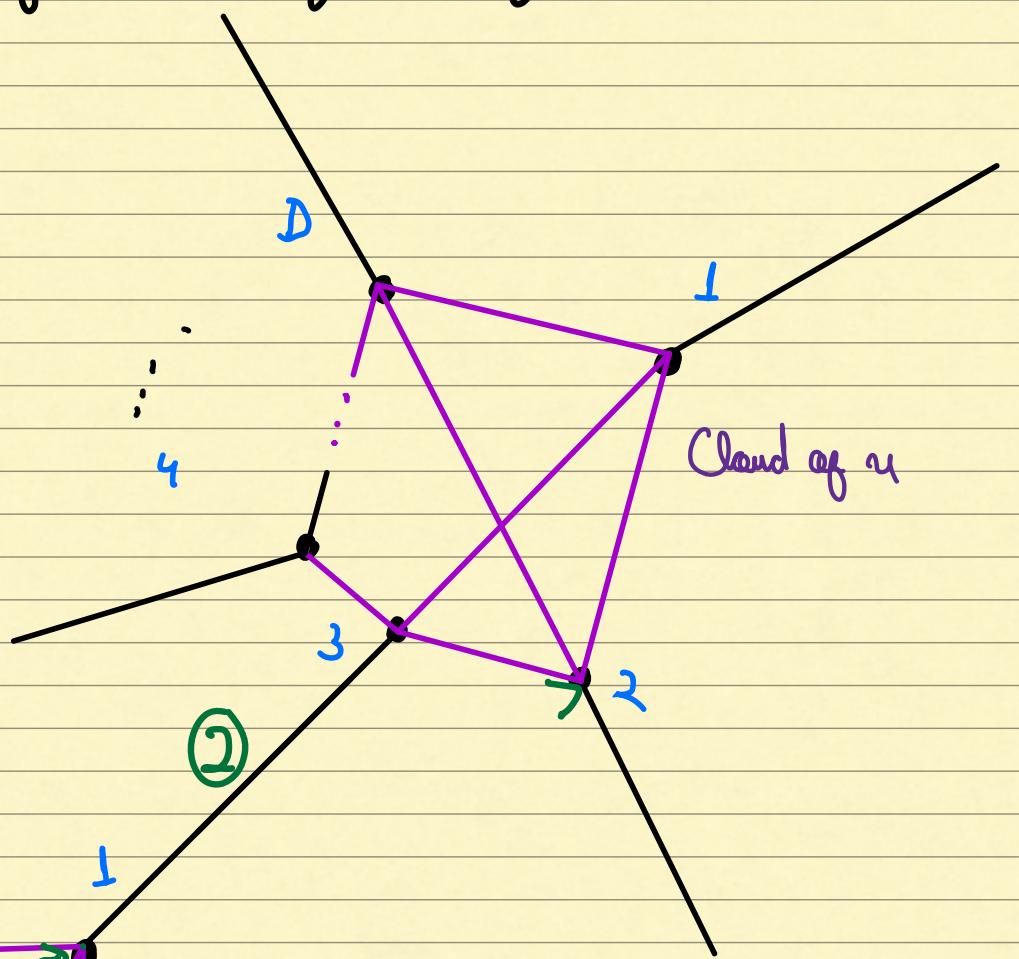
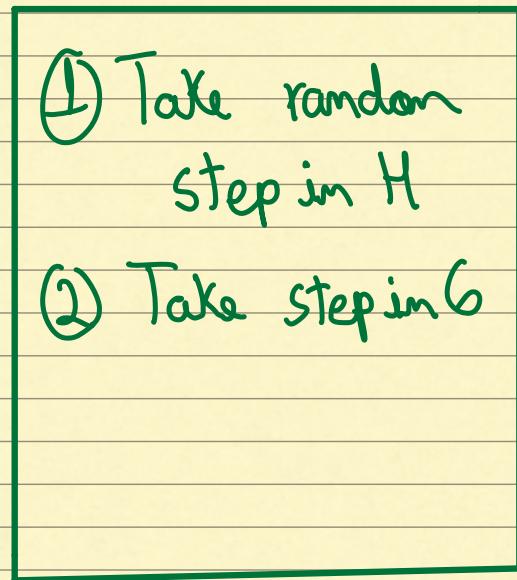
① Take random step in H



Suppose we
are here

Zig-Zag Product

$G \otimes H$ is obtained from $G \circledast H$
by a "zig-zag walk"

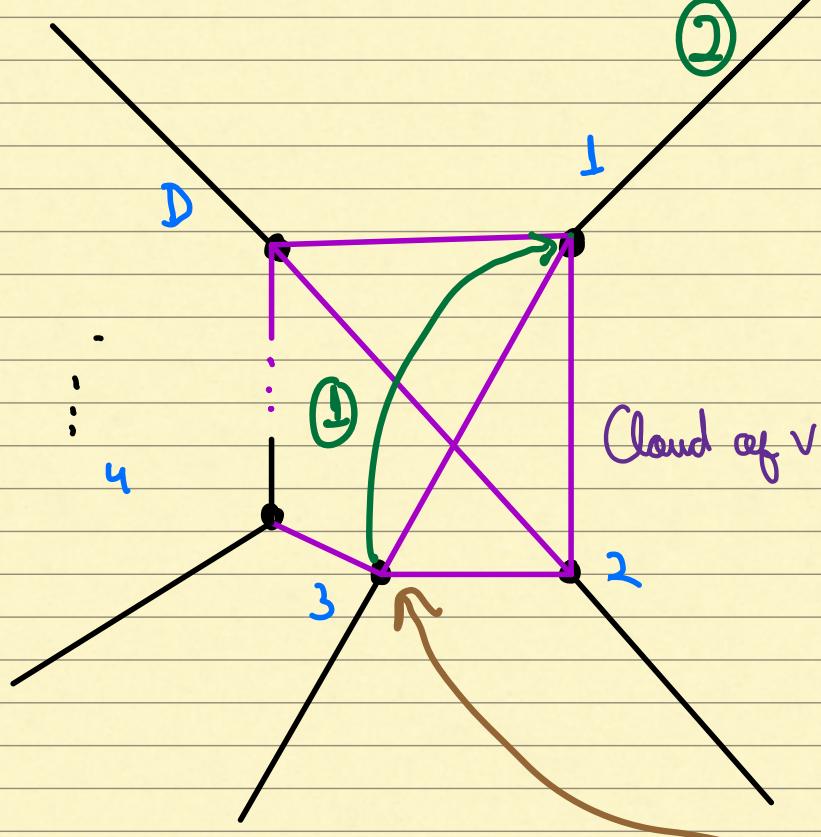
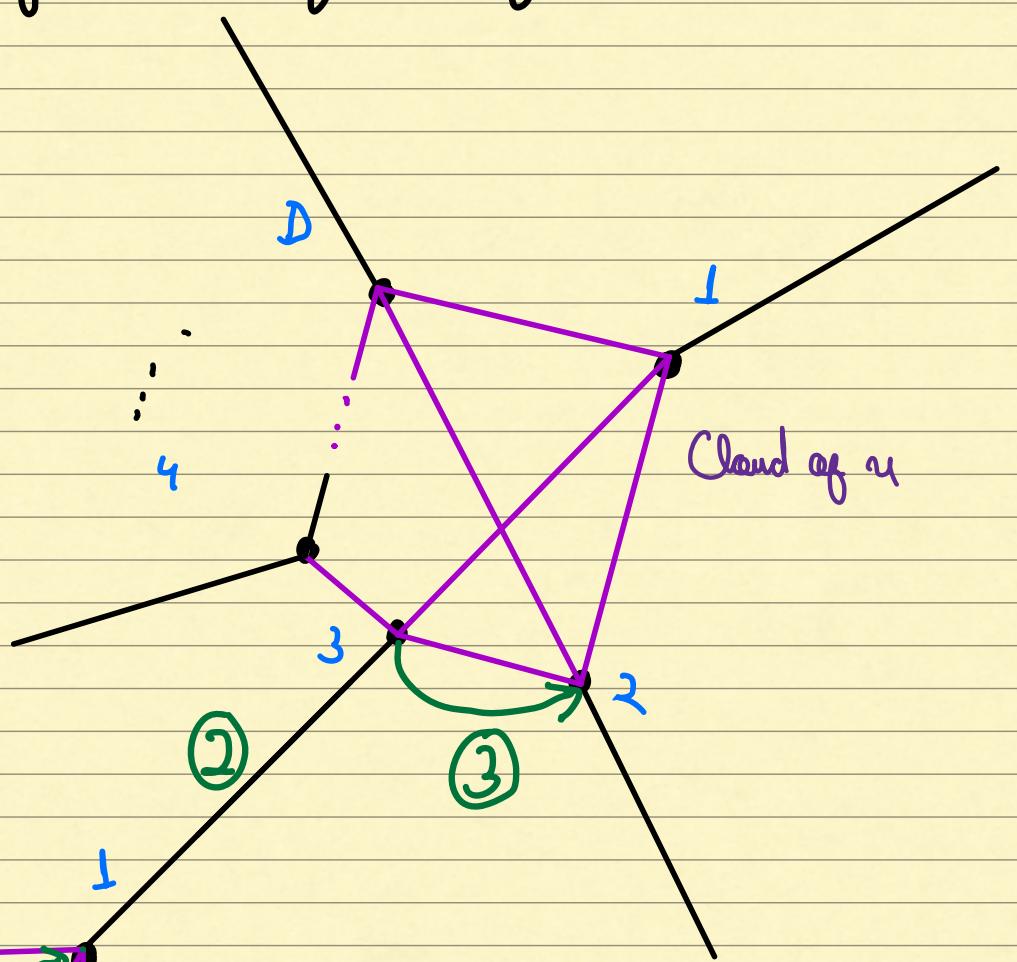


Suppose we
are here

Zig-Zag Product

$G \otimes H$ is obtained from $G \circledast H$
by a "zig-zag walk"

- ① Take random step in H
- ② Take step in G
- ③ Take random step in H



Suppose we
are here

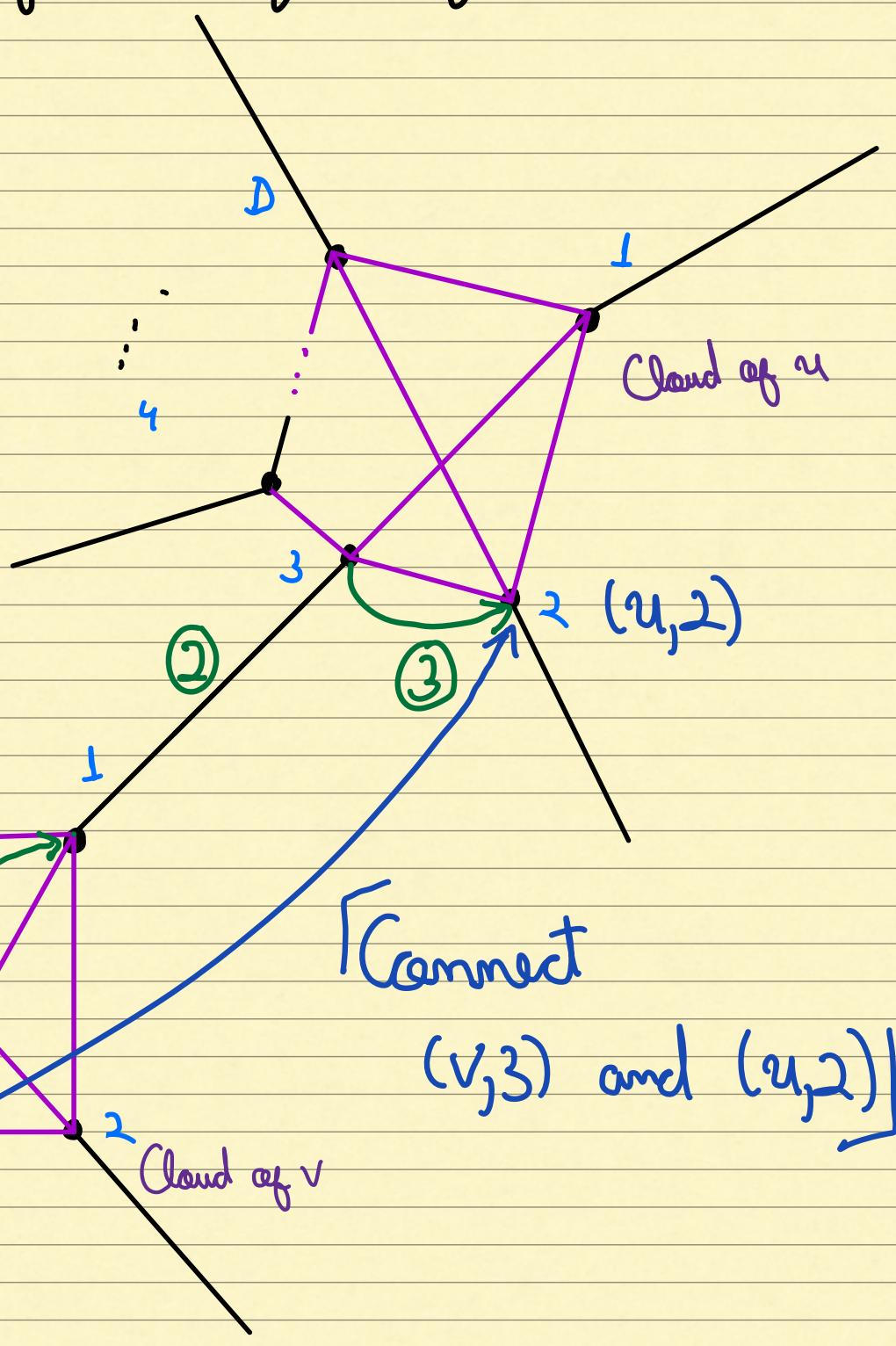
Zig-Zag Product

$G \otimes H$ is obtained from $G \circledast H$
by a "zig-zag walk"

(1) Take random step in H

(2) Take step in G

(3) Take random step in H



Zig-Zag Product

Theorem [Reingold - Vadhan - Wigderson '00]

If G is (n, D, λ_1) -graph and

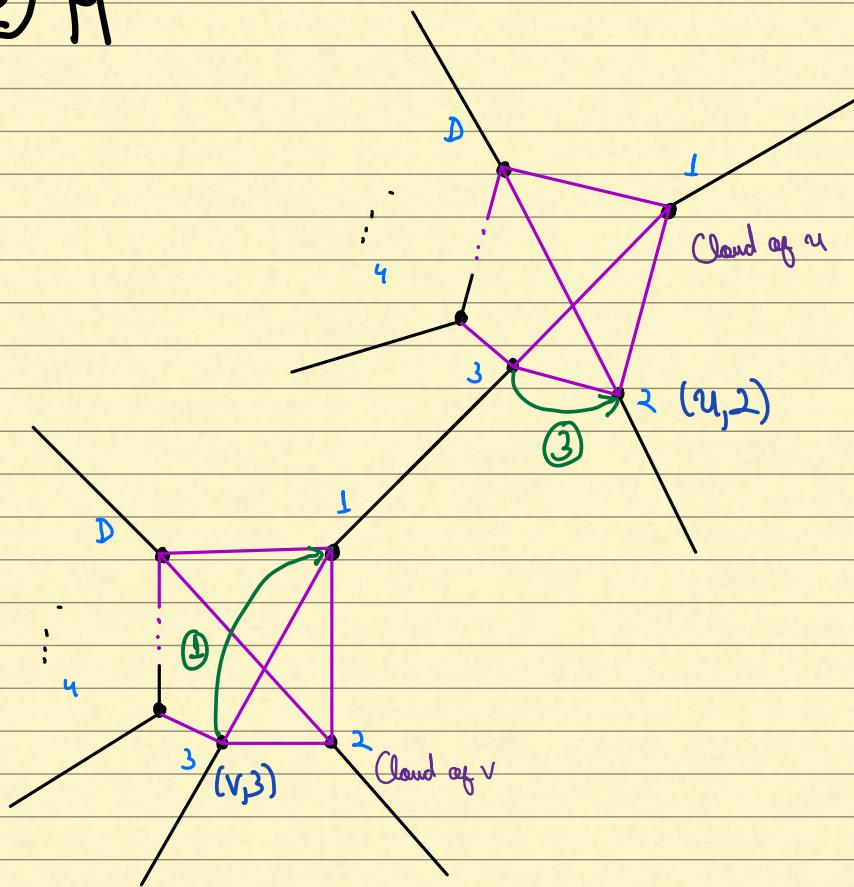
H is (D, d, λ_2) -graph, then

$G \otimes H$ is $(nD, d^2, f(\lambda_1, \lambda_2))$ -graph

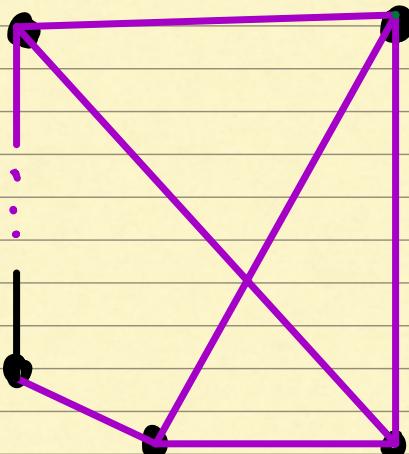
(We will bound $f(\lambda_1, \lambda_2)$ soon)

Zig-Zag Product

$G \circledast H$

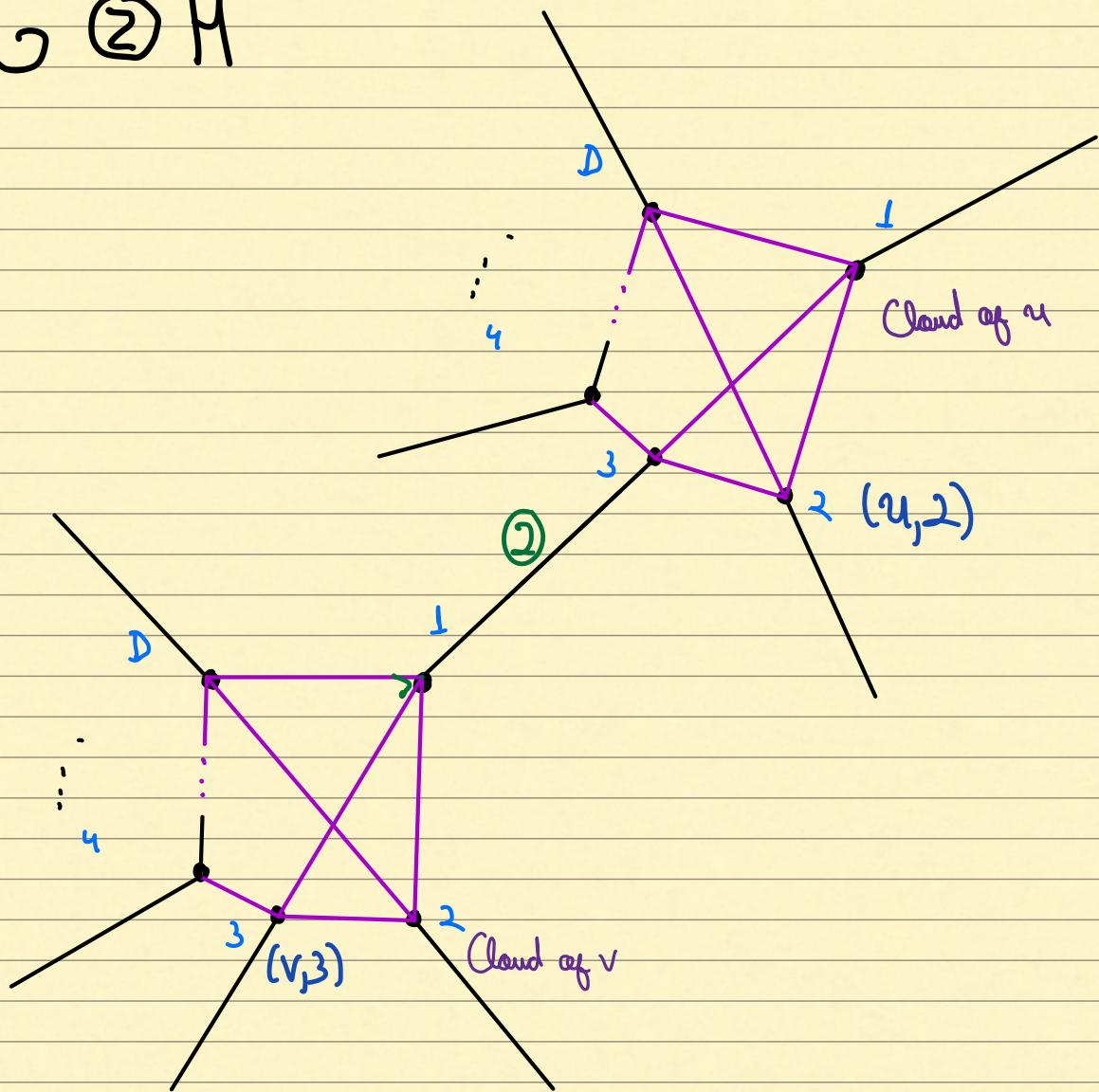


Let A_H be the
normalized adj. matrix
of H



Zig-Zag Product

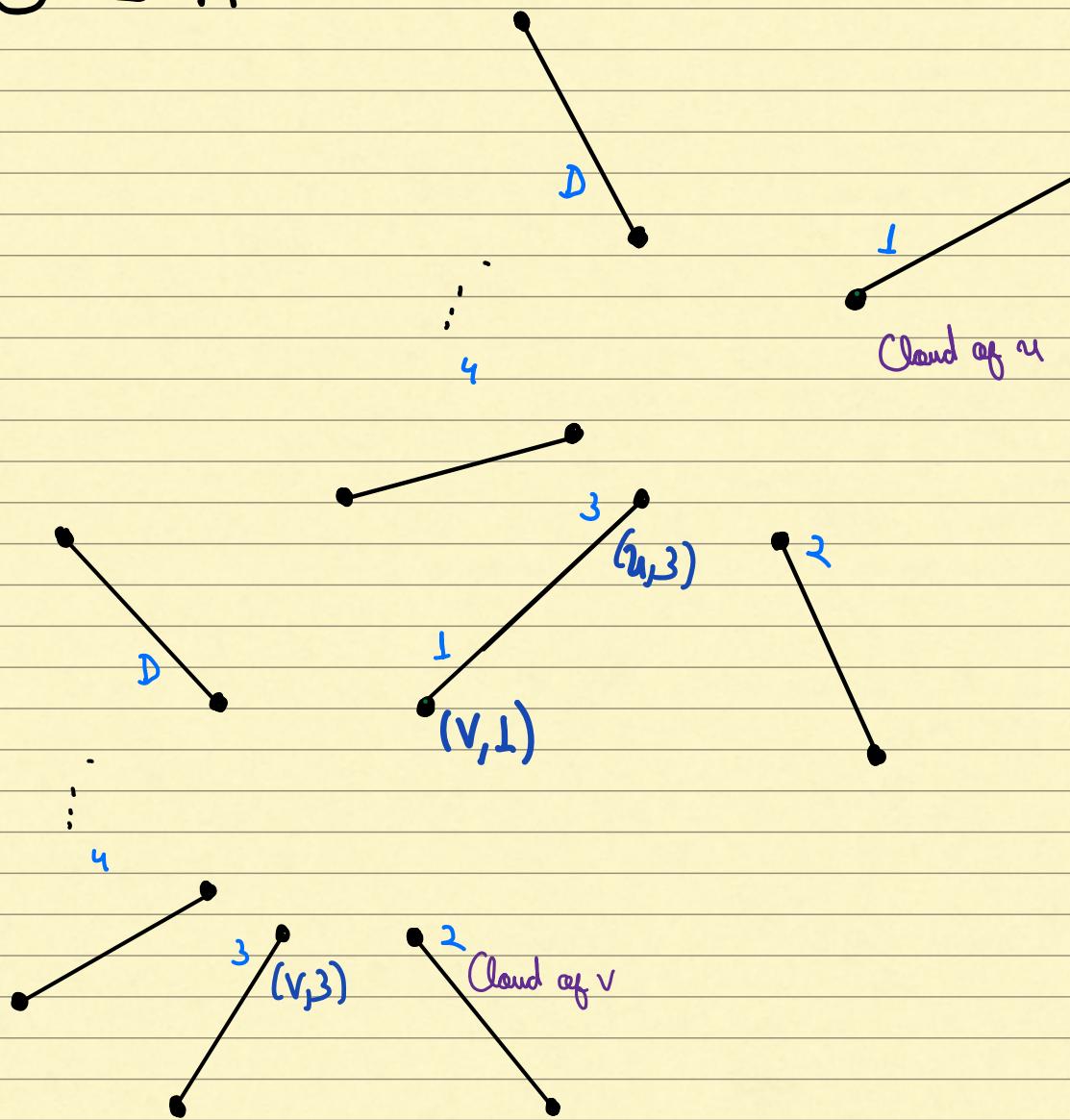
$G \circledast H$



Let \tilde{G} be the
matrix of the action
 $V(G) \times V(H)$
of G on H

Zig-Zag Product

$G \circledast H$



Let \tilde{G} be the
matrix of the action
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Zig-Zag Product

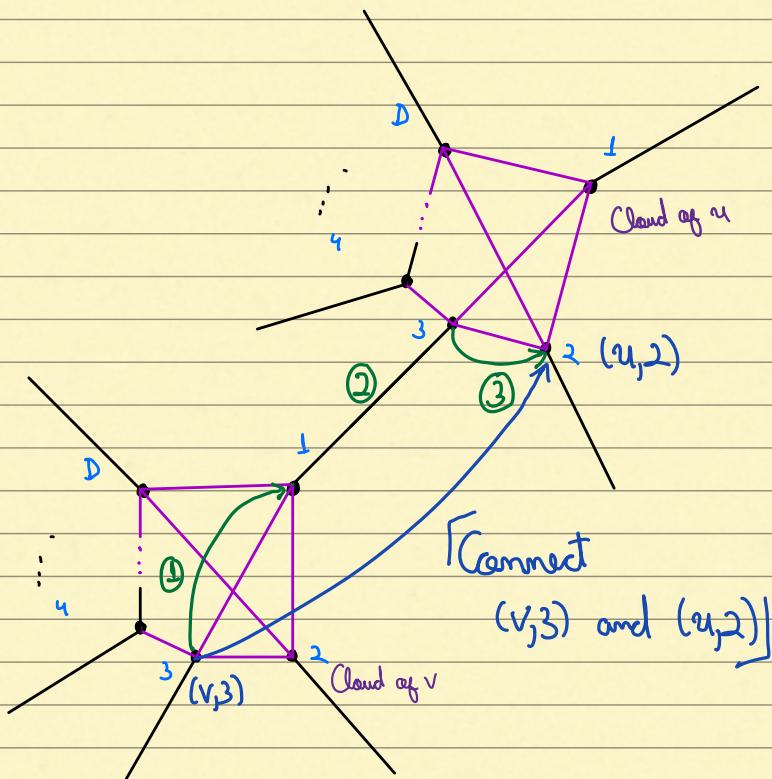
$G \otimes H$

Operators

- ① Take random step in H
- ② Take step in G
- ③ Take random step in H

$$I \otimes A_H \sim \tilde{G}$$

$$I \otimes A_H$$



Let A_H be the normalized adj. matrix of H

Let \tilde{G} be the matrix of the action $V(G) \times V(H)$ of G on \mathbb{R}

Zig-Zag Product

$G \otimes H$

- ① Take random step in H
- ② Take step in G
- ③ Take random step in H

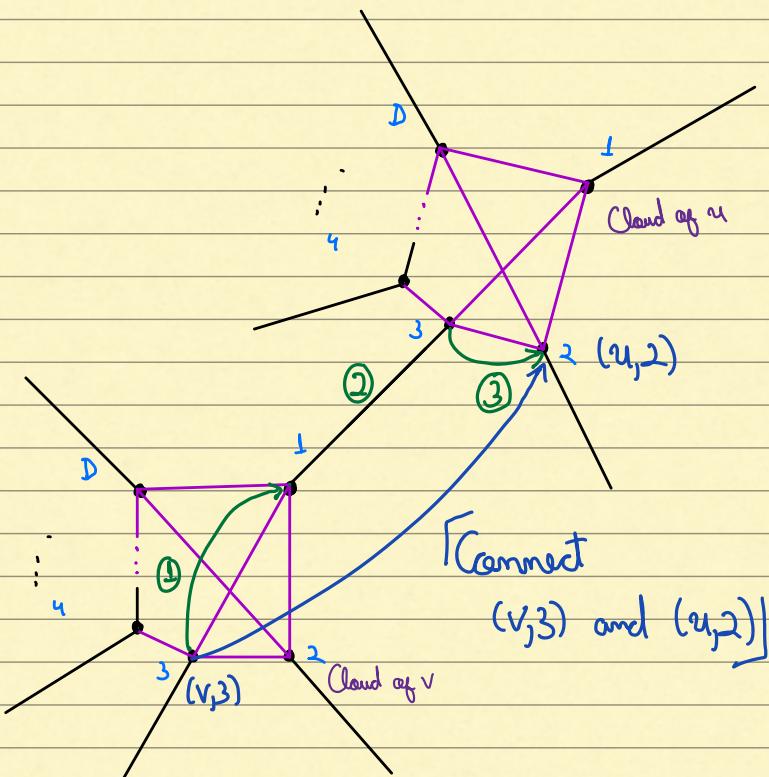
$$I \otimes A_H$$

$$\tilde{G}$$

$$I \otimes A_H$$

Adj. mat. $G \otimes H$

$$(I \otimes A_H) \tilde{G} (I \otimes A_H)$$



Let A_H be the normalized adj. matrix of H

Let \tilde{G} be the matrix of the action $V(G) \times V(H)$ of G on \mathbb{R}

Zig-Zag Product

Special Case : $G \circledcirc H$ with H complete graph with self loops

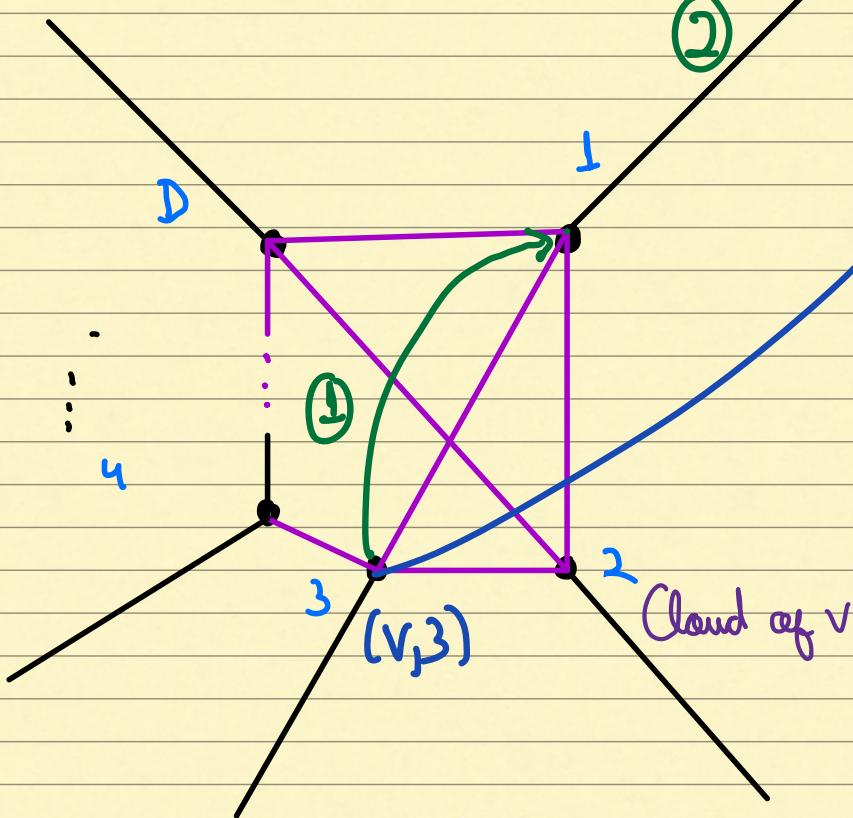
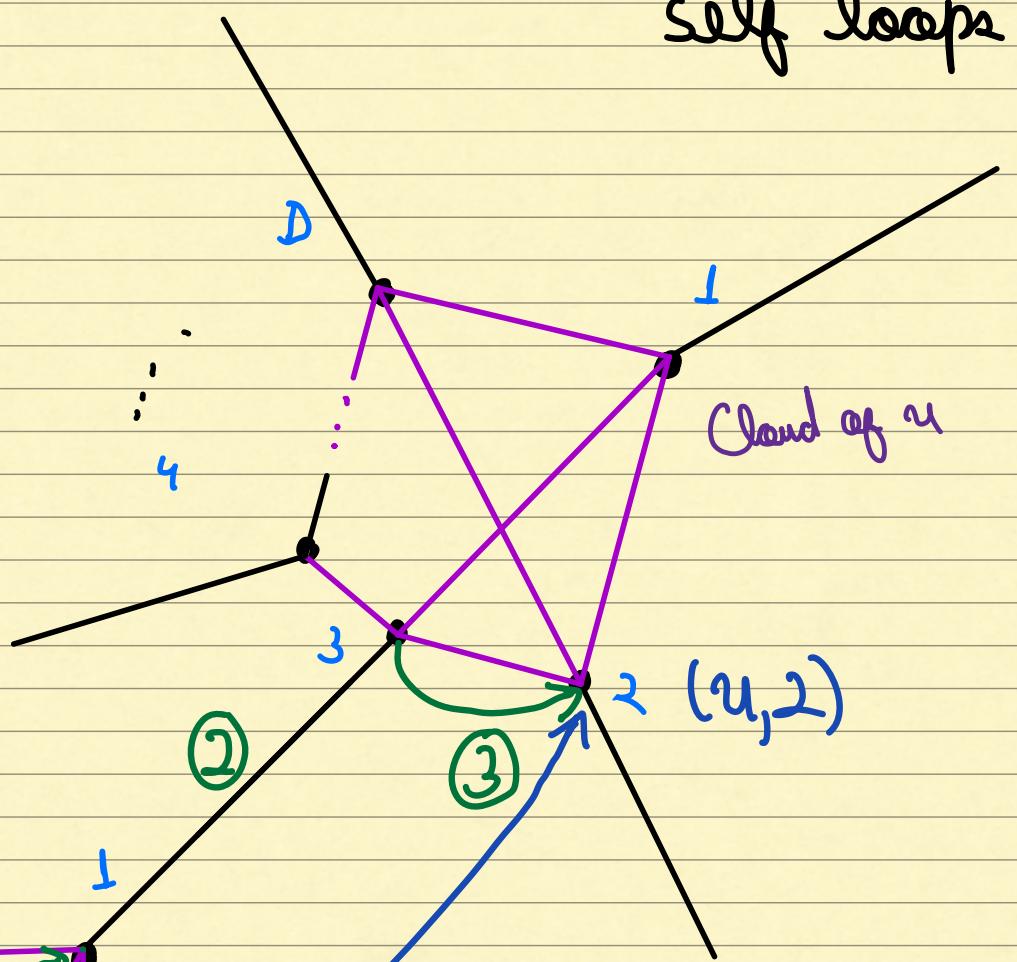
Zig-Zag Product

Special Case : $G \circledast H$ with H complete graph with self loops

① Take random step in H

② Take step in G

③ Take random step in H



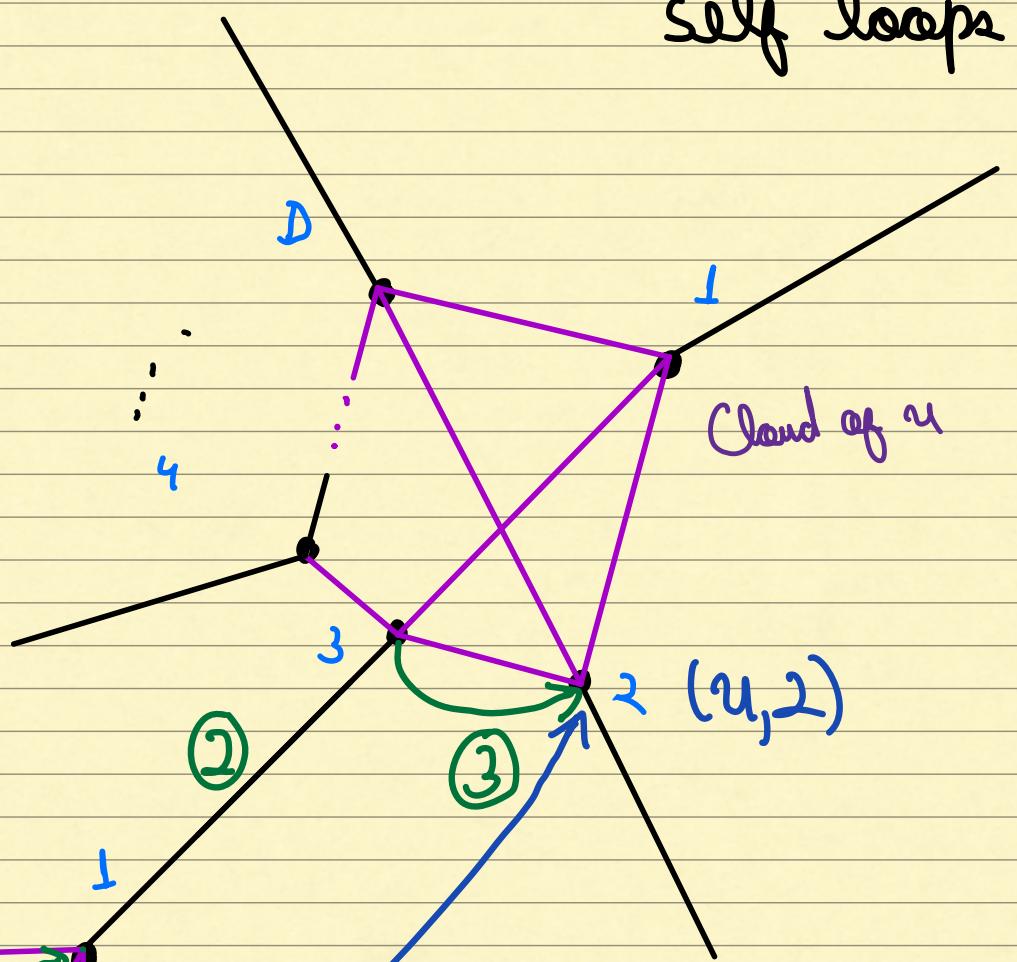
Zig-Zag Product

Special Case : $G \otimes H$ with H complete graph with self loops

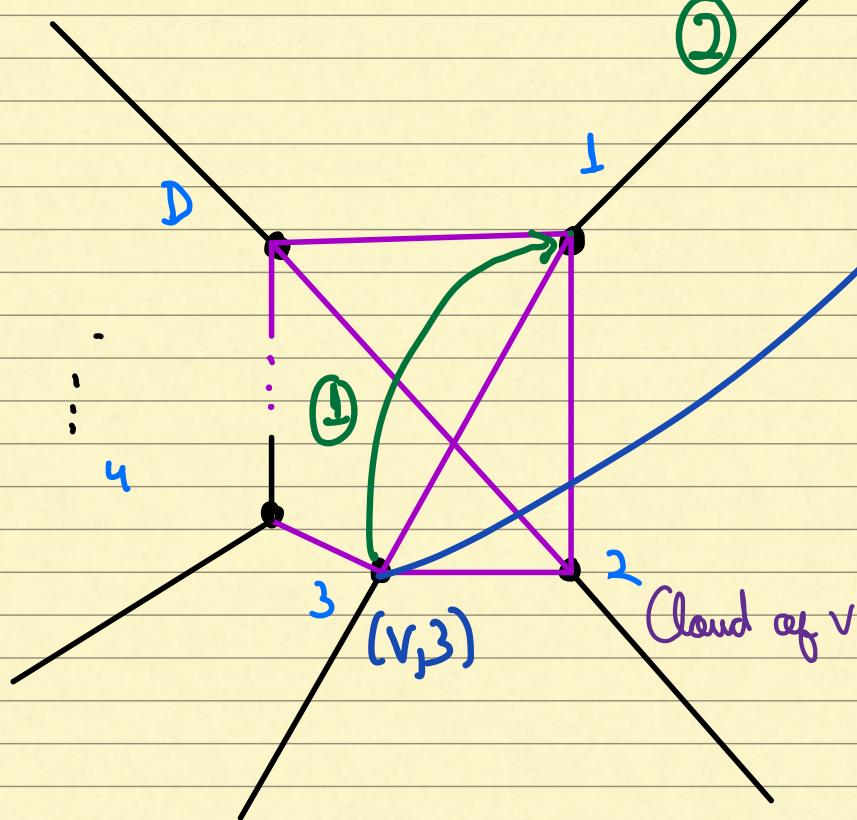
① Take random step in H

② Take step in G

③ Take random step in H



① + ② amounts
to a random
step on G

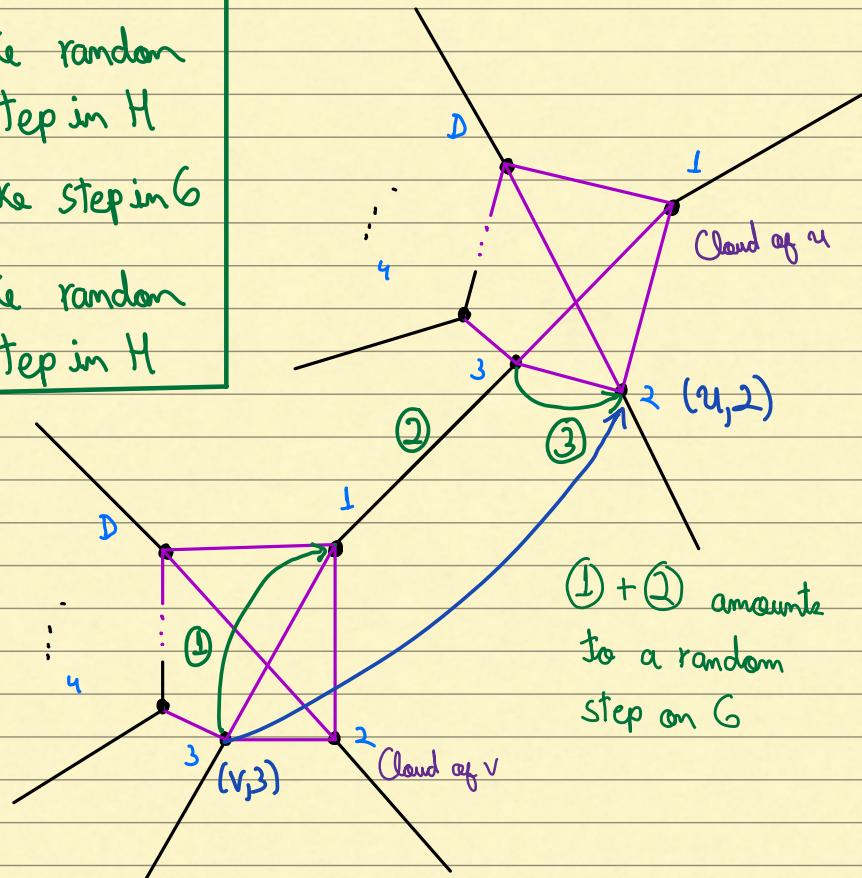


Zig-Zag Product

Special Case: $G \circledcirc H$ with H complete graph with self loops

$$(I \otimes J_p) \tilde{\otimes} (I \otimes J_D) = A_G \otimes J_D$$

- ① Take random step in H
- ② Take step in G
- ③ Take random step in H



Zig-Zag Product

Special Case : $G \circledcirc H$ with H complete graph with self loops

Fact: $(I \otimes J_D) \tilde{\otimes} (I \otimes J_D) = A_G \otimes J_D$

$$\lambda(A_G \otimes J_D) = \lambda(G)$$

Zig-Zag Product

Special Case : $G \circledcirc H$ with H complete graph with self loops

Fact: $(I \otimes J_D) \tilde{\otimes} (I \otimes J_D) = A_G \otimes J_D$

$$\lambda(A_G \otimes J_D) = \lambda(G)$$

$$|V(G \circledcirc H)| = |V(G)| \cdot |V(H)| \quad \checkmark$$

$$\lambda(G \circledcirc H) = \lambda(G)$$

Degree $G \circledcirc H$ is D^2

X

Zig-Zag Product

$G \circledast H$ with H an expander

Recall that $A_H \approx J_D$

Fact: $(I \otimes J_D) \tilde{\otimes} (I \otimes J_D) = A_G \otimes J_D$

$$\lambda(A_G \otimes J_D) = \lambda(G)$$

Zig-Zag Product

$G \circledcirc H$ with H an expander

Recall that $A_H \approx J_D$

$$A_H = \tilde{J}_D + \lambda E \quad \text{with } \|E\|_{op} \leq 1$$

$$(I \otimes A_H) \tilde{\otimes} (I \otimes A_H) =$$

$$(I \otimes \tilde{J}_D) \tilde{\otimes} (I \otimes \tilde{J}_D)$$

$$+ \lambda (I \otimes \tilde{J}_D) \tilde{\otimes} (I \otimes E)$$

$$+ \lambda (I \otimes E) \tilde{\otimes} (I \otimes \tilde{J}_D)$$

$$+ \lambda^2 (I \otimes E) \tilde{\otimes} (I \otimes E)$$

Zig-Zag Product

$G \circledcirc H$ with H an expander

Recall that $A_H \approx J_D$

$$A_H = \bar{J}_D + \lambda E \quad \text{with } \|E\|_{op} \leq 1$$

$$(I \otimes A_H) \tilde{\otimes} (I \otimes A_H) =$$

$$(I \otimes \bar{J}_D) \tilde{\otimes} (I \otimes \bar{J}_D) \rightarrow \lambda_1$$

$$\begin{aligned} &+ \lambda (I \otimes \bar{J}_D) \tilde{\otimes} (I \otimes E) \\ &+ \lambda (I \otimes E) \tilde{\otimes} (I \otimes \bar{J}_D) \} \rightarrow 2\lambda_2 \\ &+ \lambda^2 (I \otimes E) \tilde{\otimes} (I \otimes E) \rightarrow \lambda_2^2 \end{aligned}$$

$\text{Fact: } (I \otimes \bar{J}_D) \tilde{\otimes} (I \otimes \bar{J}_D) = A_G \otimes \bar{J}_D$ $\lambda(A_G \otimes \bar{J}_D) = \lambda(G)$
--

Zig-Zag Product

$G \circledcirc H$ is obtained from $G \circledast H$
by a "zig-zag walk"

Theorem [Reingold - Vadhan - Wigderson '00]

If G is (n, D, λ_1) -graph and
 H is (D, d, λ_2) -graph, then

$G \circledcirc H$ is $(nD, d^2, \lambda_1 + 2\lambda_2 + \lambda_2^2)$ -graph

Zig-Zag Product

$G \circledcirc H$ is obtained from $G \circledast H$
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Theorem [Reingold - Vadhan - Wigderson '00]

If G is (n, D, λ_1) -graph and
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$G \circledcirc H$ is $(nD, d^2, \lambda_1 + 2\lambda_2 + \lambda_2^2)$ -graph

[Construct larger and larger expanders

Starting from a constant size one]

Near-optimal Spectral Expanders

Theorem [Alon - Boppana]

$$\lambda(G) \geq 2\sqrt{d-1} - o_n(1)$$

Ramanujan Graphs

$$\lambda(G) \leq 2\sqrt{d-1}$$

Near-optimal Spectral Expanders

Ramanujan Graphs

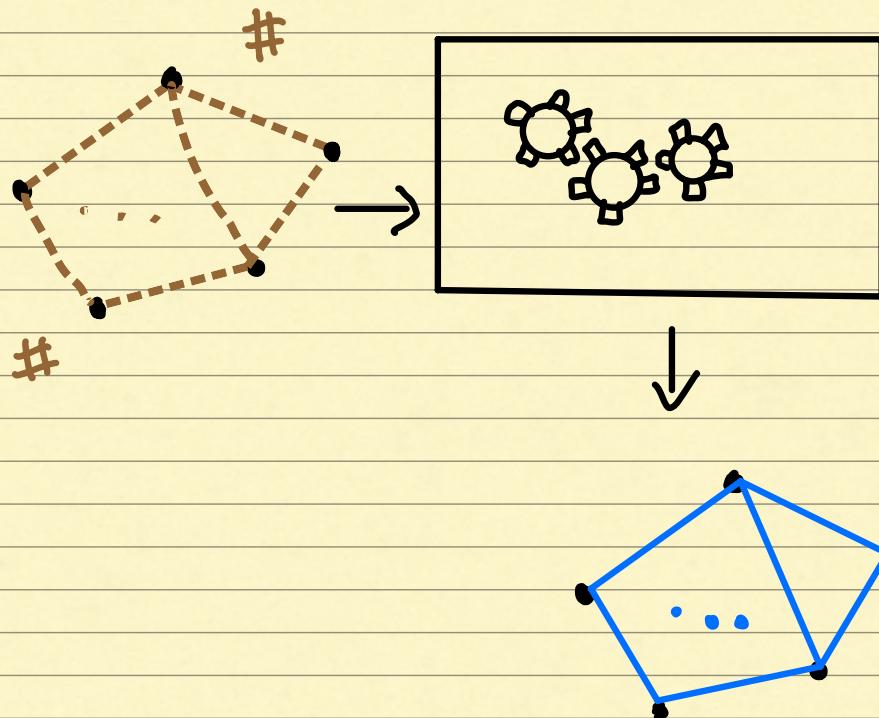
$$\lambda(G) \leq 2\sqrt{d-1}$$

(What is required to be a near
optimal expander?)

Near-optimal Spectral Expanders

Theorem [J-Mittal-Ray-Wigderson'22]

Any expander can be transformed
into an almost Ramanujan one



Near-optimal Spectral Expanders

Theorem [J-Mittal-Roy-Wigderson'22]

Any expander can be transformed
into an almost Ramanujan one

Corollary [JMRW'22]

All "expanding groups" admit
almost optimal expanders

Applications to

- "Generic" Expanders
- Cayley Graphs
- Quantum Expanders
- Monotone Expanders
- Dimension Expanders
- Codes

Near-optimal Spectral Expanders

Theorem [J-Mittal - Roy - Wigderson '22]

Any expander can be transformed
into an almost Ramanujan one

Key Technique : higher-order zig-zag
for operators

(building on the work of
Ben-Aroya and Ta-Shma)

Thank You!