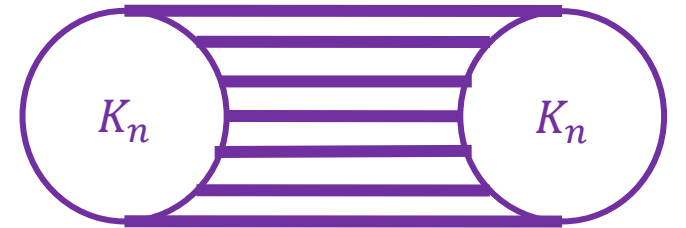


Cheeger Inequalities for Vertex Expansion and Reweighted Eigenvalues

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2nd November @ FOCS 2022



Joint work with:
Tsz Chiu Kwok (Shanghai U of Finance and Economics)
Lap Chi Lau (U Waterloo)

Outline

- Classical Cheeger's inequality
- Vertex expansion
- Reweighted eigenvalues
- Our results
- Proof ideas
- Summary

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Cheeger's inequality

- $G = (V, E)$ undirected

- Conductance of graph: $\phi(G) := \min_{\text{vol}(S) \leq \text{vol}(V)/2} \frac{|\delta(S)|}{\text{vol}(S)}$

$$\mathcal{A} := D^{-\frac{1}{2}} A D^{-\frac{1}{2}}$$

#edges across S

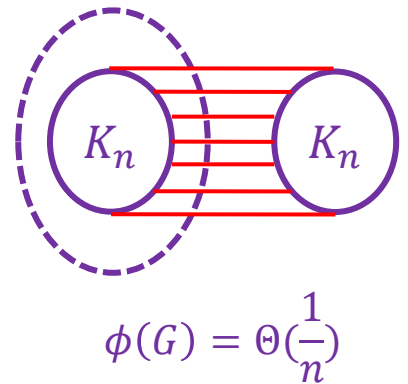
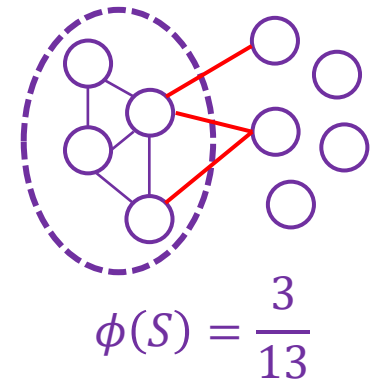
Total degree of S

- Eigenvalues of Laplacian $\mathcal{L} := I - \mathcal{A}$ are $0 = \lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n \leq 2$

Theorem [Cheeger '70, Alon, Milman '85, Alon '86]

$$\frac{\phi^2}{2} \leq \lambda_2 \leq 2\phi$$

- Generalizations abound [Trevisan '09], [LOT '12], [LRTV '12], [KLL0T '13]



λ_2 and mixing time

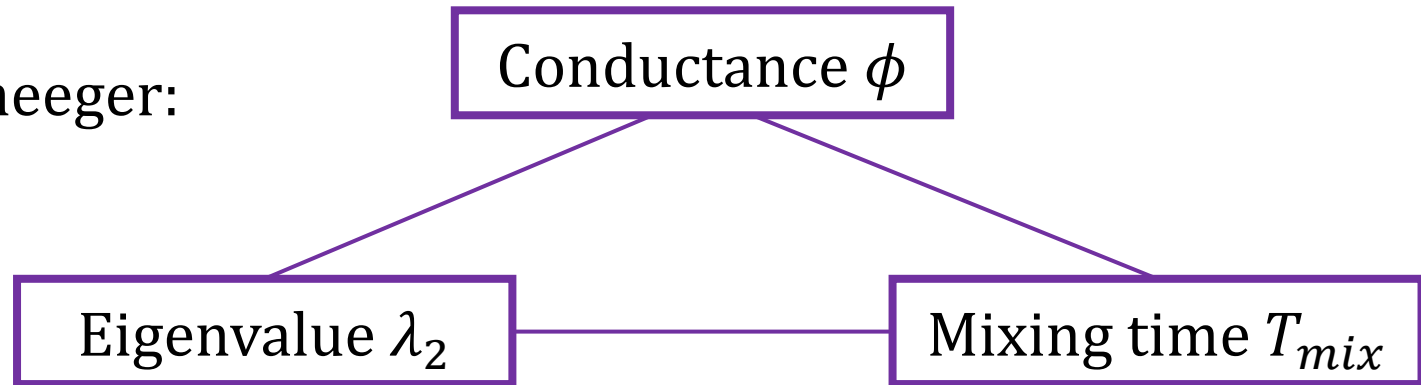
Go to a random neighbor of the current vertex u

- Let P be the canonical random walk on G
- Mixing time is roughly proportional to $1/\lambda_2$:

time needed to get $(1/e)$ -close to s.d. π

$$\frac{1}{\lambda_2} \lesssim T_{mix}(P) \lesssim \frac{1}{\lambda_2} \cdot \log \frac{1}{\pi_{min}}$$

- Summary of classical Cheeger:



Question: is there an analogous theory for vertex expansion?

Outline

- Classical Cheeger's inequality
- **Vertex expansion**
- Reweighted eigenvalues
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Vertex expansion

- $G = (V, E)$ undirected

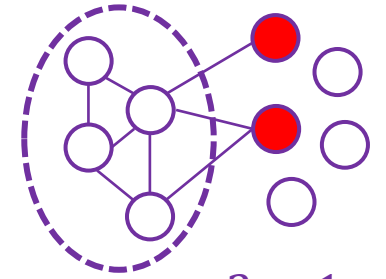
- Vertex expansion of graph: $\psi(G) := \min_{|S| \leq |V|/2} \frac{|N(S)|}{|S|}$

#neighbors of S

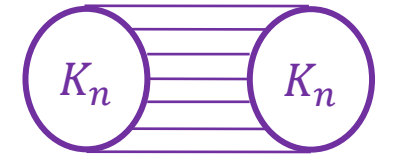
Size of S

- Past work on ψ include:

- Unnormalized Laplacian λ'_2 [Tanner '84], [Alon, Milman '85]
- Spectral quantity λ_∞ [Bobkov, Houdré, Tetali '00]
- SDP relaxation sdp_∞ [Louis, Raghavendra, Vempala '13]
- Extension of ARV [Feige, Hajiaghayi, Lee '08]
- Spectral hypergraph theory [Louis '15], [CLTZ '18]



$$\psi(S) = \frac{2}{4} = \frac{1}{2}$$



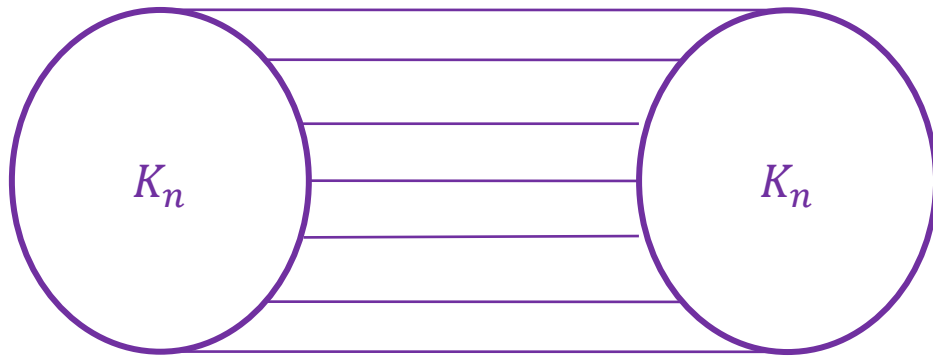
$$\psi(G) = \Theta(1)$$

Lots of past work, but no nice spectral theory for ψ :(

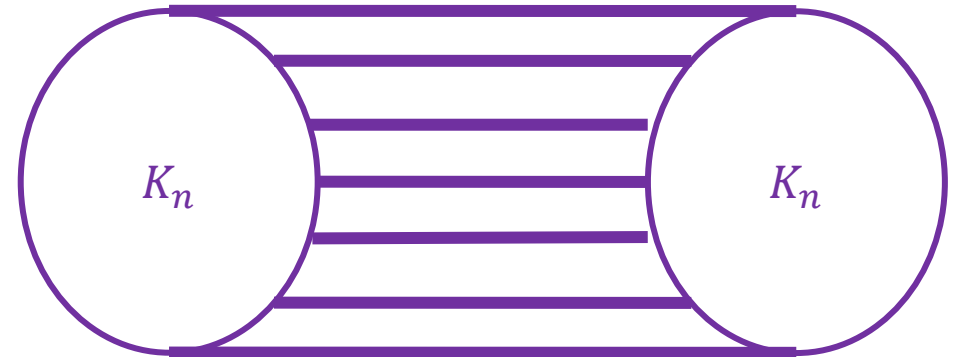
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Reweighting example



Mixing time is $\Theta(n)$



Mixing time is $\Theta(1)$

Reweighted eigenvalue

- [BDX '04] “Fastest mixing Markov chain”
- Key idea: eigenvalue λ_2 as proxy for mixing time
- It is the following program:

$$\lambda_2^*(G) := \max_{P \geq 0} 1 - \alpha_2(P)$$

$$\text{subject to } P(u, v) = P(v, u) = 0$$

$$\sum_{v \in V} P(u, v) = 1$$

$$\pi(u)P(u, v) = \pi(v)P(v, u)$$

$$\lambda_2(I - P)$$

$$\forall uv \notin E$$

$$\forall u \in V$$

$$\forall uv \in E.$$

Cheeger's inequality for ψ

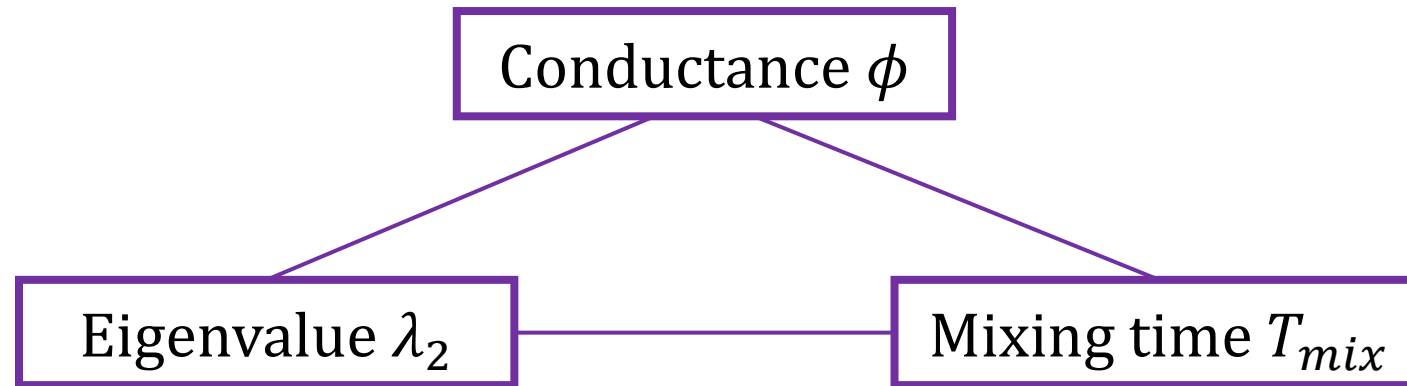
- Relation between λ_2^* and ψ

Theorem [Olesker-Taylor, Zanetti '22] For π uniform,

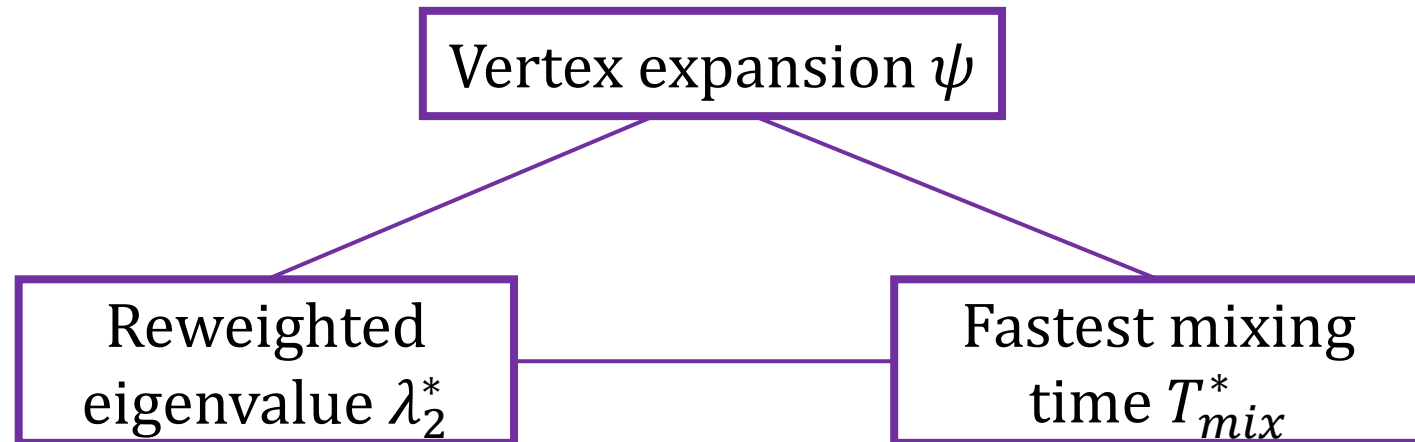
$$\frac{\psi(G)^2}{\log |V|} \lesssim \lambda_2^*(G) \lesssim \psi(G).$$

A new vertex spectral theory

- Edge:



- Vertex:



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1) Cheeger's inequality for ψ , v.2

Theorem [Olesker-Taylor, Zanetti '22] For π uniform,

$$\frac{\psi(G)^2}{\log |V|} \lesssim \lambda_2^*(G) \lesssim \psi(G).$$

- They left open the following questions:
 - arbitrary distribution π ?
 - $\log |V|$ replaced by $\log d$? ([LRV '13]: SSE-hard to go beyond)

d : max degree

1) Cheeger's inequality for ψ , v.2

- We answered the questions in the affirmative:

Theorem [Kwok, Lau, T. '22] For arbitrary π ,

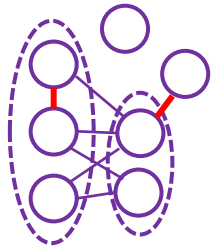
$$\frac{\psi(G)^2}{\log d} \lesssim \lambda_2^*(G) \lesssim \psi(G).$$

- Furthermore, the log-dependence on d is optimal (not just SSE-optimal)

2) Generalizations of Cheeger's inequalities

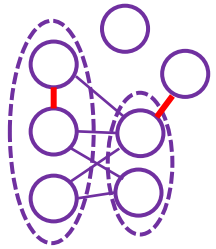
Eigenvalues (edge)	Reweighted eigenvalues (vertex)

2) Generalizations of Cheeger's inequalities

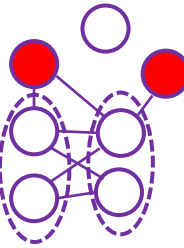


Eigenvalues (edge)	Reweighted eigenvalues (vertex)
<p><u>Bipartite Cheeger</u> [Trevisan '09]</p> $\frac{\phi_B(G)^2}{2} \leq 2 - \lambda_n(G) \leq 2\phi_B(G)$ <p>Relates $2 - \lambda_n$ to "bipartiteness"</p>	

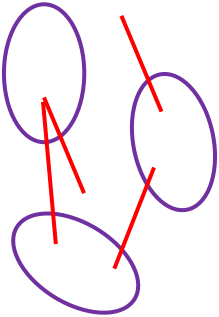
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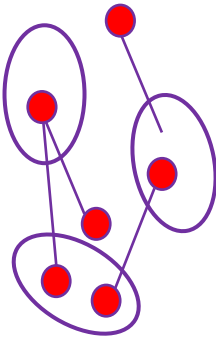
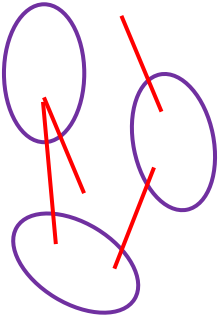


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3) 0/1 polytope with torpid mixing

- A 0/1 polytope is a polytope with vertices in $\{0,1\}^d \subseteq \mathbb{R}^d$

Conjecture (0/1 polytope conjecture)

The graph of any 0/1 polytope has edge expansion ≥ 1 .

- If true implies fast sampling using random walks
- On the sampling side, we obtain the following negative evidence:

Theorem [Kwok, Lau, T. '22]

For fixed k and large enough n , there is a 0/1 polytope Q with $O(n^k)$ vertices and $\psi(Q) \lesssim O_k\left(\frac{1}{n^{k-2}}\right)$.

- As a corollary, for all $\epsilon > 0$ there is a 0/1-polytope with *fastest* mixing time $\Omega(|V|^{1-\epsilon})$ to the uniform distribution

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Dual of λ_2^* program [Roch '05]

Fractional matching

• Primal: $\lambda_2^*(G) := \max_{P \geq 0} 1 - \alpha_2(P)$

subject to $P(u, v) = P(v, u) = 0 \quad \forall uv \notin E$

$\sum_{v \in V} P(u, v) = 1 \quad \forall u \in V$

$\pi(u)P(u, v) = \pi(v)P(v, u) \quad \forall uv \in E.$



Rayleigh quotient + minimax + LP duality

Fractional vertex cover

• Dual: $\gamma(G) := \min_{f: V \rightarrow \mathbb{R}^n, g: V \rightarrow \mathbb{R}_{\geq 0}}$

$\sum_{v \in V} \pi(v)g(v)$

subject to $\sum_{v \in V} \pi(v) \|f(v)\|^2 = 1$ Normalization constraints

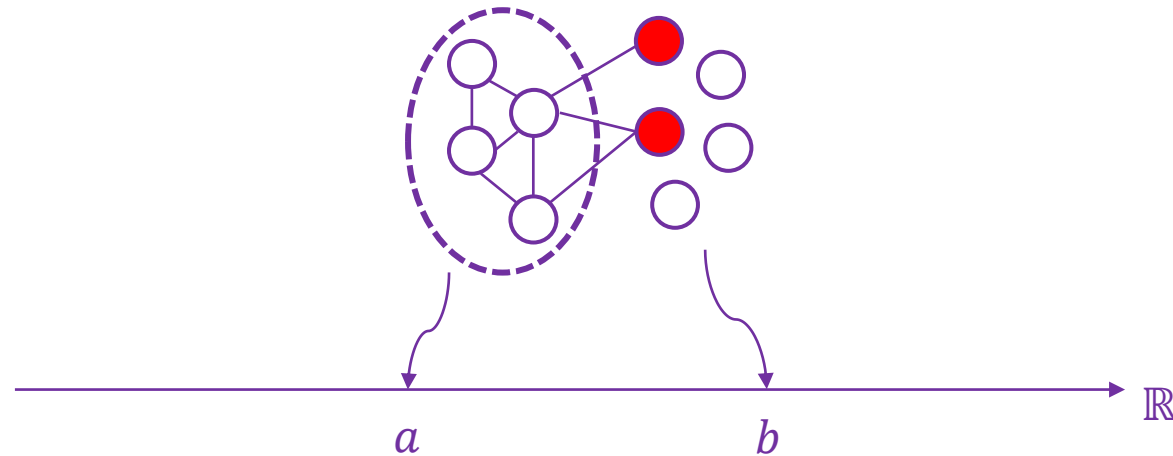
$\sum_{v \in V} \pi(v)f(v) = \vec{0}$

$g(u) + g(v) \geq \|f(u) - f(v)\|^2 \quad \forall uv \in E.$

$\gamma^{(k)}(G)$ if $f: V \rightarrow \mathbb{R}^k$

Easy direction

- Every vertex cut S can be realized as a two-point embedding



- Cover all crossing edges using $g(u) = [u \in N(S)] \cdot (a - b)^2$
- Check that $\sum_u \pi(u)g(u) \leq 2 \psi(S)$

Hard direction: overview

Theorem [Olesker-Taylor, Zanetti '22] For π uniform,

$$\frac{\psi(G)^2}{\log |V|} \lesssim \lambda_2^*(G) \lesssim \psi(G).$$

1. Take dual: $\gamma^{(n)}(G) = \lambda_2^*(G)$
2. J-L lemma: $\gamma^{(1)}(G) \lesssim \log |V| \cdot \gamma^{(n)}(G)$
3. Round 1D solution to matching conductance

Theorem [Kwok, Lau, T. '22] For arbitrary π ,

$$\frac{\psi(G)^2}{\log d} \lesssim \lambda_2^*(G) \lesssim \psi(G).$$

1. Take dual: $\gamma^{(n)}(G) = \lambda_2^*(G)$
2. Gaussian projection: $\gamma^{(1)}(G) \lesssim \log d \cdot \gamma^{(n)}(G)$
[Jain, Pham, Vuong '22] showed the same for π uniform
3. Round 1D solution to “**directed vertex expansion**”

Key intermediary is “directed program”

Directed program

- Original “vertex cover” constraint:

$$g(u) + g(v) \geq \|f(u) - f(v)\|^2 \quad \forall uv \in E$$

- Direct all edges “appropriately”
- Constraint changed to $g(v) \geq \|f(u) - f(v)\|^2 \quad \forall u \rightarrow v$

$$\text{Fact: } \gamma^{(k)}(G) \leq \vec{\gamma}^{(k)}(G) \leq 2\gamma^{(k)}(G)$$

$$g(v) \geq g(u)$$



$$g(u) + g(v) \geq \|f(u) - f(v)\|^2$$



$$g'(v) \geq \|f(u) - f(v)\|^2$$

$$g'(u) := 2g(u) \quad \forall u \in V$$

Better (analysis of) projection

$$\Pi: \mathbb{R}^n \rightarrow \mathbb{R}^{O(\log n)}$$

- Using J-L:
 - We ensure all $\|\Pi f(u) - \Pi f(v)\|^2 \approx \|f(u) - f(v)\|^2$
 - Go to $O(\log n)$ dimensions, then “take the best coordinate”

- Using directed program:

- Objective written as $\sum_u \pi(u) \max_{v:u \rightarrow v} \|f(u) - f(v)\|^2$

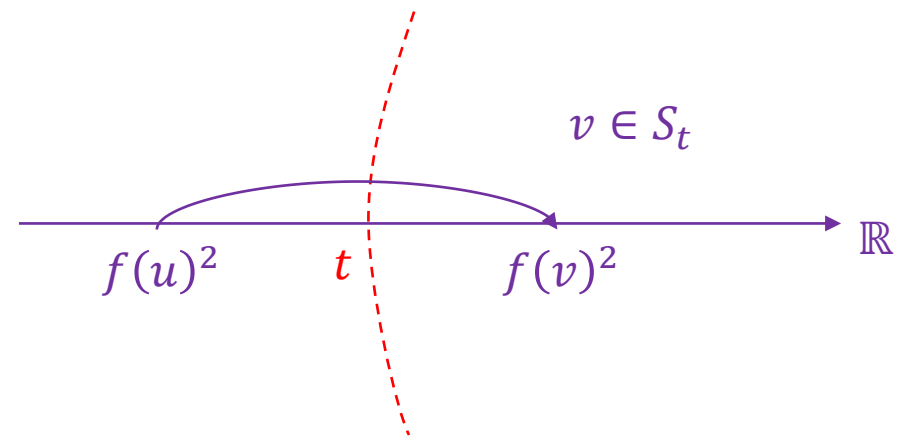
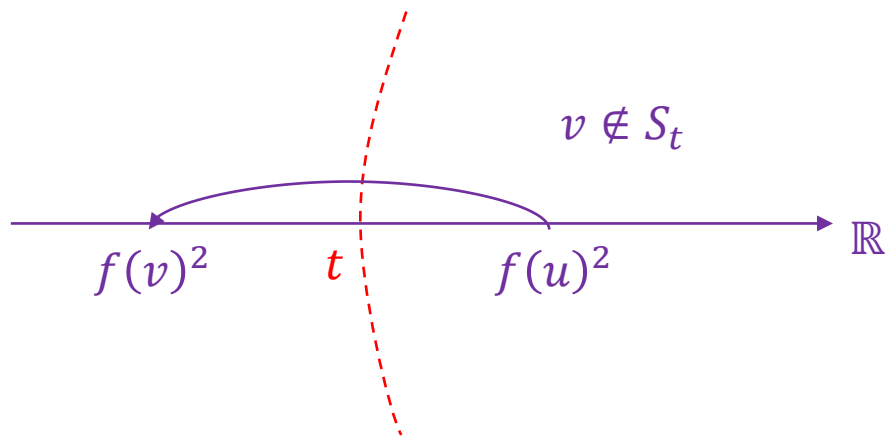
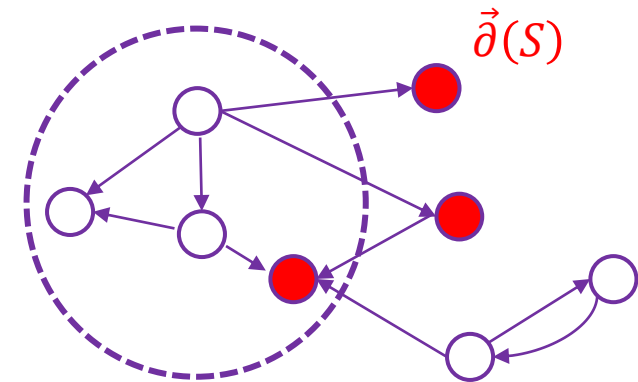
- Projected objective: $\sum_u \pi(u) \max_{v:u \rightarrow v} (\Pi f(u) - \Pi f(v))^2$

$$\Pi: \mathbb{R}^n \rightarrow \mathbb{R}$$

- Expected max. of d squared standard Gaussians is $O(\log d)$
 - Linearity of expectations to conclude $\gamma^{(1)}(G) \leq O(\log d) \cdot \gamma^{(n)}(G)$

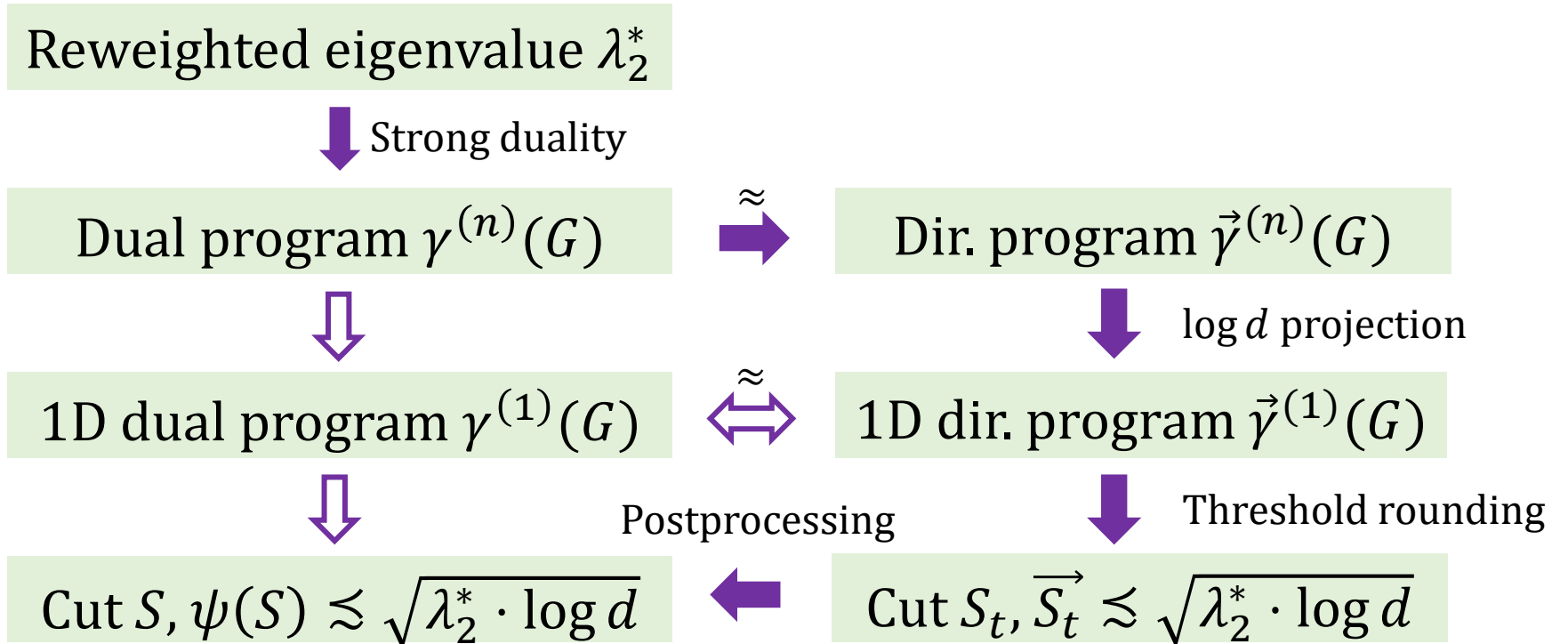
Threshold rounding

- Take 1D solution $f: V \rightarrow \mathbb{R}$
- Consider directed vertex boundary $\vec{\partial}(S)$
- Define $S_t := \{v \in V: f(v)^2 > t\}$



$$\Pr[v \in \vec{\partial}(S_t)] \propto \max_{u: u \rightarrow v} |f(u)^2 - f(v)^2|$$

Recap



Proxy for λ_k^*

- We don't know how to write λ_k^* as a convex program!
- Idea: consider $\sigma_k^* := (\lambda_1 + \lambda_2 + \dots + \lambda_k)^*$

- Dual looks like

$$\kappa(G) := \min_{f:V \rightarrow \mathbb{R}^n, g:V \rightarrow \mathbb{R}_{\geq 0}}$$

subject to

$$\sum_{v \in V} \pi(v)g(v)$$

$$g(u) + g(v) \geq \|f(u) - f(v)\|^2 \quad \forall uv \in E$$

$$\sum_{v \in V} \pi(v)f(v)f(v)^T \preceq I_n$$

$$\sum_{v \in V} \pi(v) \|f(v)\|^2 = k.$$

Fractional vertex cover

Normalization
constraints
(but different)

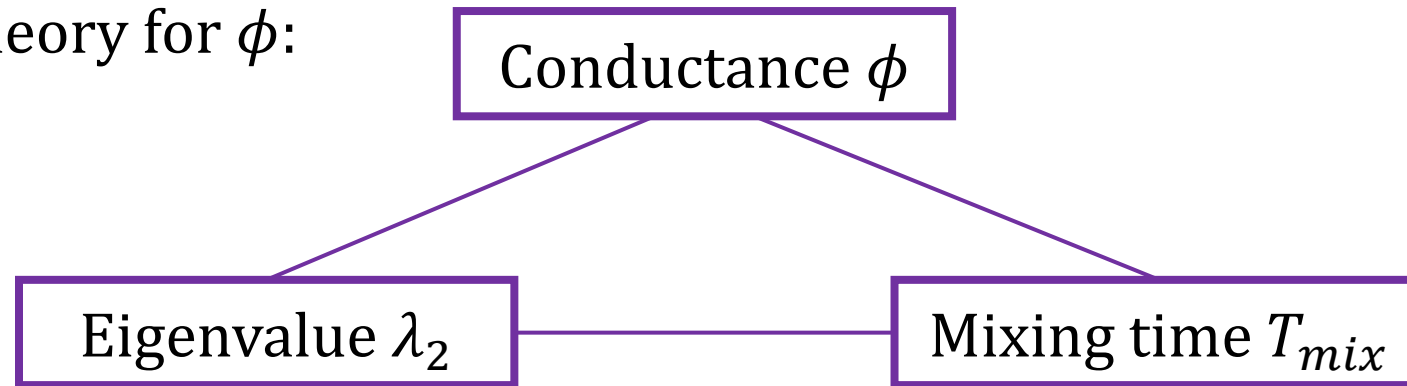
- Convex program! (Factor k loss)

Outline

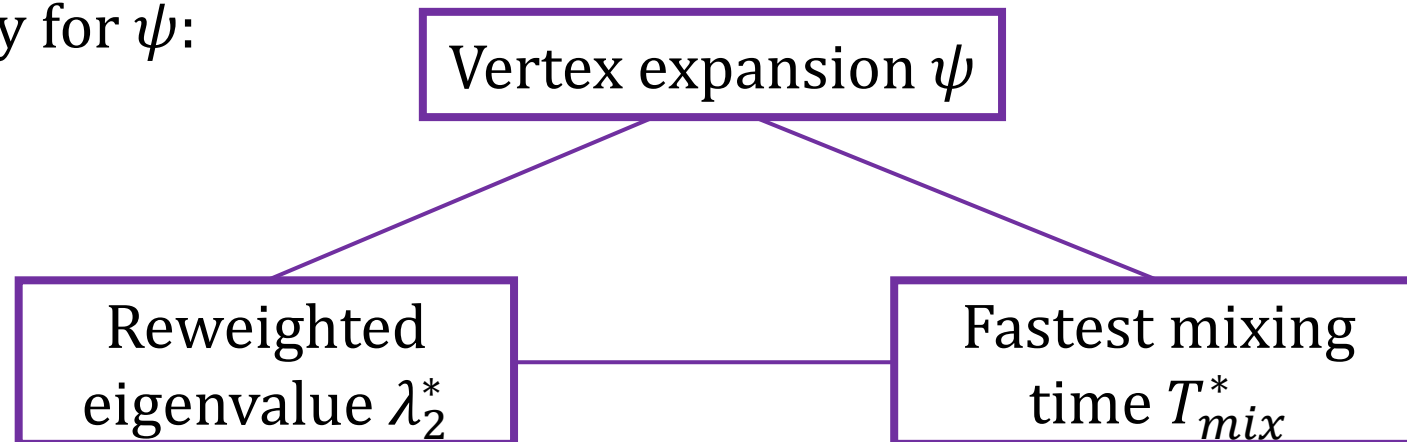
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- **Summary**

I) Features of new spectral theory

- Classical spectral theory for ϕ :



- New spectral theory for ψ :



II) Cheeger's inequalities

Eigenvalues	Reweighted eigenvalues
<u>Cheeger</u> [Cheeger '70, Alon, Milman '85, Alon '86] Relates λ_2 to ϕ	[Kwok, Lau, T. '22] Relates λ_2^* to ψ
<u>Bipartite Cheeger</u> [Trevisan '09] Relates $2 - \lambda_n$ to "bipartiteness"	[Kwok, Lau, T. '22] Relates " $2 - \lambda_n^*$ " to "bipartite v. expn"
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What about reweighted versions of other spectral results?

Open questions

Extension of spectral theory

- Small-set vertex expansion
- Hypergraphs, directed graphs, etc.

Algorithms

- Fast(er) algorithms for approximating ψ or computing λ^*
- Local algorithms for ψ

Reweighting & approximation

- Arora-Ge conjecture for graph coloring
- Steurer's conjecture (true \Rightarrow SUBEXP sparsest cut)

Thank you!