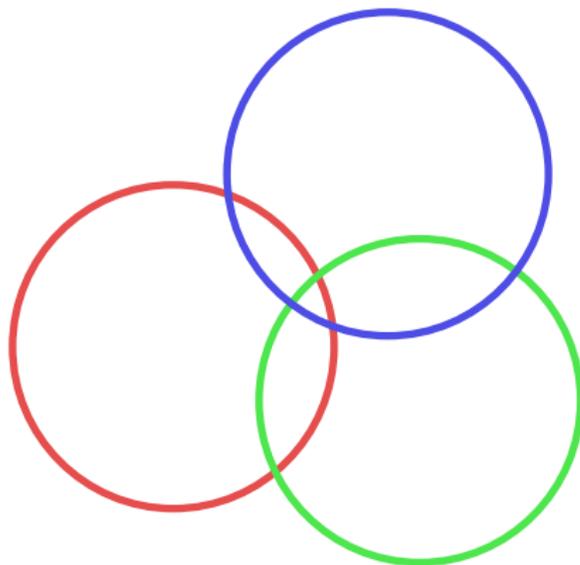


LIMITS OF LINEAR AND SEMIDEFINITE RELAXATIONS FOR COMBINATORIAL PROBLEMS

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- linear and semidefinite programming
- approximation algorithms and computational complexity
- logic and finite model theory



Part I

LINEAR PROGRAMMING RELAXATIONS

Problem:

Given an undirected graph $G = (V, E)$,
find the smallest number of vertices
that **touches** every edge.

Notation:

$$vc(G).$$

Observe:

$A \subseteq V$ is a vertex cover of G
iff
 $V \setminus A$ is an independent set of G

Linear programming relaxation

LP relaxation:

$$\text{minimize } \sum_{u \in V} x_u$$

subject to

$$x_u + x_v \geq 1 \quad \text{for every } (u, v) \in E,$$

$$x_u \geq 0 \quad \text{for every } u \in V.$$

Notation:

$$\text{fvc}(G).$$

Approximation:

$$\text{fvc}(G) \leq \text{vc}(G) \leq 2 \cdot \text{fvc}(G)$$

Integrality gap:

$$\sup_G \frac{\text{vc}(G)}{\text{fvc}(G)}$$

Approximation:

$$\text{fvc}(G) \leq \text{vc}(G) \leq 2 \cdot \text{fvc}(G)$$

Integrality gap:

$$\sup_G \frac{\text{vc}(G)}{\text{fvc}(G)} = 2.$$

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$$\text{fvc}(G) \leq \text{vc}(G) \leq 2 \cdot \text{fvc}(G)$$

Integrality gap:

$$\sup_G \frac{\text{vc}(G)}{\text{fvc}(G)} = 2.$$

Gap examples:

1. $\text{vc}(K_{2n+1}) = 2n$,
2. $\text{fvc}(K_{2n+1}) = \frac{1}{2}(2n + 1)$.

LP tightenings

Add triangle inequalities:

$$\text{minimize } \sum_{u \in V} x_u$$

subject to

$$x_u + x_v \geq 1 \quad \text{for every } (u, v) \in E,$$

$$x_u \geq 0 \quad \text{for every } u \in V,$$

$$x_u + x_v + x_w \geq 2 \quad \text{for every triangle } \{u, v, w\} \text{ in } G.$$

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Integrality gap:

Remains 2.

Gap examples:

Triangle-free graphs with small independence number.

Hierarchy:

Systematic ways of
generating **all** linear inequalities
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$$P = \{x \in \mathbb{R}^n : Ax \geq b\},$$

$$P^{\mathbb{Z}} = \text{convexhull}\{x \in \{0, 1\}^n : Ax \geq b\}.$$

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Produce explicit nested polytopes:

$$P = P^1 \supseteq P^2 \supseteq \dots \supseteq P^{n-1} \supseteq P^n = P^{\mathbb{Z}}$$

P^k : Lasserre/Sums-of-squares (SOS) Hierarchy

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$$Q_0 + \sum_{j=1}^m L_j Q_j + \sum_{i=1}^n (x_i^2 - x_i) Q_i = L \geq 0$$

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$$\deg(Q_0), \deg(L_j Q_j), \deg((x_i^2 - x_i) Q_i) \leq k.$$

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Then:

$$P^k = \{x \in \mathbb{R}^n : L(x) \geq 0 \text{ for each produced } L \geq 0\}$$

P^k : Sherali-Adams (SA) Hierarchy

Given linear inequalities

$$L_1 \geq 0, \dots, L_m \geq 0$$

produce all linear inequalities of the form

$$Q_0 + \sum_{j=1}^m L_j Q_j + \sum_{i=1}^n (x_i^2 - x_i) Q_i = L \geq 0$$

where

$$Q_i = \sum_{\ell \in J} c_\ell \prod_{i \in A_\ell} x_i \prod_{i \in B_\ell} (1 - x_i) \quad \text{with} \quad c_\ell \geq 0$$

and

$$\deg(Q_0), \deg(L_j Q_j), \deg((x_i^2 - x_i) Q_i) \leq k.$$

Then:

$$P^k = \{x \in \mathbb{R}^n : L(x) \geq 0 \text{ for each produced } L \geq 0\}$$

Example: triangles in P^3

For each triangle $\{u, v, w\}$ in G :

$$\begin{aligned} & Q_0 + \\ & (x_u + x_v - 1)Q_1 + \\ & (x_u + x_w - 1)Q_2 + \\ & (x_v + x_w - 1)Q_3 + \\ & (x_u^2 - x_u)Q_4 + \\ & (x_v^2 - x_v)Q_5 + \\ & (x_w^2 - x_w)Q_6 \\ & = ? \\ & (x_u + x_v + x_w - 2). \end{aligned}$$

$$Q_i = a_i + b_i x_u + c_i x_v + d_i x_w + e_i x_u x_v + f_i x_u x_w + g_i x_v x_w + h_i x_u x_v x_w$$

Solving P^k

Lift-and-project:

- Step 1: **lift** from \mathbb{R}^n up to $\mathbb{R}^{(n+1)^k}$ and linearize the problem
- Step 2: **project** from $\mathbb{R}^{(n+1)^k}$ down to \mathbb{R}^n

Proposition:

Optimization of linear functions over P^k
can be solved in time $\dagger m^{O(1)} n^{O(k)}$.

Proof:

1. for SA- P^k : by linear programming
2. for SOS- P^k : by semidefinite programming

An Important Open Problem

Define

$\text{sa}^k \text{fvc}(G)$: optimum fractional vertex cover of $\text{SA-}P^k$
 $\text{sos}^k \text{fvc}(G)$: optimum fractional vertex cover of $\text{SOS-}P^k$

An Important Open Problem

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Open problem:

$$\sup_G \frac{\text{vc}(G)}{\text{sos}^4 \text{fvc}(G)} \stackrel{?}{<} 2$$

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Gap examples:

Frankl-Rödl Graphs: $FR_\gamma^n = (\mathbb{F}_2^n, \{\{x, y\} : x + y \in A_\gamma^n\})$.

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[Dinur, Safra, Khot, Regev, Kleinberg, Charikar, Hatami, Magen, Georgiou, Lovasz, Arora, Alekhovich, Pitassi; 2000's]

Part II

COUNTING LOGIC

First-order logic of graphs:

$E(x, y)$: x and y are joined by an edge

$x = y$: x and y denote the same vertex

$\neg\phi$: negation of ϕ holds

$\phi \wedge \psi$: both ϕ and ψ hold

$\exists x(\phi)$: there exists a vertex x that satisfies ϕ

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First-order logic with k variables (or width k) :

L^k : collection of formulas for which
all subformulas have at most k free variables.

Example

Paths:

$$P_1(x, y) := E(x, y)$$

$$P_2(x, y) := \exists z_1 (E(x, z_1) \wedge P_1(z_1, y))$$

$$P_3(x, y) := \exists z_2 (E(x, z_2) \wedge P_2(z_2, y))$$

⋮

$$P_{i+1}(x, y) := \exists z_i (E(x, z_i) \wedge P_i(z_i, y))$$

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Bipartiteness of n -vertex graphs:

$$\forall x (\neg P_3(x, x) \wedge \neg P_5(x, x) \wedge \cdots \wedge \neg P_{2\lceil n/2 \rceil - 1}(x, x)).$$

Counting quantifiers

Counting witnesses:

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Counting logic with k variables (or counting width k):

C^k : collection of formulas with counting quantifiers with all subformulas with at most k free variables.

C^k -equivalence:

$G \equiv_k^C H$: G and H satisfy the **same** sentences of C^k .

Combinatorial characterization of C^2 -equivalence

Color-refinement:

1. color each vertex black,
2. color each vertex by number of neighbors in each color-class,
3. repeat 2 until color-classes don't split any more.

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Theorem [Immerman and Lander]

$$G \equiv_2^C H \text{ if and only if } G \equiv^R H$$

LP characterization of color-refinement

Isomorphisms:

1. $G \cong H$,
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LP relaxation of \cong :

$G \equiv^F H$: there exists **doubly stochastic** S such that $GS = SH$.

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Theorem [Tinhofer]

$G \equiv^R H$ if and only if $G \equiv^F H$.

Higher levels of SA Hierarchy

SA-levels of fractional isomorphism:

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Moreover:

1. This interleaving is **strict** for $k > 2$ [Grohe-Otto 2015]
2. A combined LP characterizes \equiv_k^{C} **exactly** [Grohe-Otto 2015]
3. Alternative (and independent) formulation by [Malkin 2014]

SA and SOS-levels of fractional isomorphism:

1. $G \equiv_k^{\text{SA}} H$: the degree- k SA level of $\text{iso}(G, H)$ is **feasible**.
2. $G \equiv_k^{\text{SOS}} H$: the degree- k SOS level of $\text{iso}(G, H)$ is **feasible**.

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Theorem [AA and Ochremiak 2018]: There exists $c > 1$ such that:

$$G \equiv_{ck}^{\text{SA}} H \implies G \equiv_k^{\text{SOS}} H \implies G \equiv_k^{\text{SA}} H.$$

Part III

APPLICATIONS

Local LPs (and SDPs)

Basic k -local LPs:

1. one **variable** $x_{\mathbf{u}}$ for each k -tuple $\mathbf{u} \in V^k$,
2. one **inequality** $\sum_{\mathbf{u} \in V^k} a_{\mathbf{u}, \mathbf{v}} \cdot x_{\mathbf{u}} \geq b_{\mathbf{v}}$ for every k -tuple $\mathbf{v} \in V^k$,
3. coefficients $a_{\mathbf{u}, \mathbf{v}}$ **depend only** on the type $\text{atp}_G(\mathbf{u}, \mathbf{v})$,
4. coefficients $b_{\mathbf{v}}$ **depend only** on the type $\text{atp}_G(\mathbf{v})$.

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k -local LP:

Union of basic k -local LPs
with coefficients $a_{t(\mathbf{x}, \mathbf{y})}$ and $b_{t(\mathbf{y})}$ indexed
by isomorphism types $t(\mathbf{x}, \mathbf{y})$ and $t(\mathbf{y})$.

Example 1: fractional vertex cover

Fractional vertex cover: Given a graph $G = (V, E)$

$$\sum_{u \in V} x_u \leq W$$

$$x_u + x_v \geq 1 \quad \text{for every } (u, v) \in E,$$

$$x_u \geq 0 \quad \text{for every } u \in V.$$

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1. Objective function: basic 1-local LP
2. Edge constraint: basic 2-local LP
3. Positive constraint: basic 1-local LP

Example 2: fractional matching polytope

Fractional matching polytope: Given a graph $G = (V, E)$

$$\sum_{uv \in E} x_{uv} \geq W$$

$$x_{uv} = x_{vu}$$

for every $u, v \in V$

$$\sum_{v \in V} x_{uv} \leq 1$$

for every $u \in V$

$$0 \leq x_{uv} \leq 1$$

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$$\sum_{v \in V} x_{uv} \leq 1 \quad \text{for every } u \in V$$

$$0 \leq x_{uv} \leq 1 \quad \text{for every } u, v \in V$$

1. Objective function: basic 2-local LP
2. Symmetry constraint: two basic 2-local LPs
3. Degree-at-most-one constraint: basic 2-local LP

Example 3: metric polytope

Metric polytope: Given a graph $G = (V, E)$

$$\frac{1}{2} \sum_{uv \in E} x_{uv} \geq W$$

$$x_{uv} = x_{vu} \quad \text{for every } u, v \in V$$

$$x_{uw} \leq x_{uv} + x_{vw} \quad \text{for every } u, v, w \in V$$

$$x_{uv} + x_{vw} + x_{uw} \leq 2 \quad \text{for every } u, v, w \in V$$

$$0 \leq x_{uv} \leq 1 \quad \text{for every } u, v \in V$$

1. Objective function: basic 2-local LP
2. Symmetry constraint: two basic 2-local LPs
3. Triangle inequality: basic 3-local LP
4. Perimetric inequality: basic 3-local LP
5. Unit cube constraint: two basic 2-local LPs

Preservation of local LPs and SDPs

Theorem Let P be a k -local LP or SDP.

1. LP: If $G \equiv_k^C H$, then $P(G)$ is feasible iff $P(H)$ is feasible.
2. SDP: If $G \equiv_{ck}^C H$, then $P(G)$ is feasible iff $P(H)$ is feasible.

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'Just do it' proof for LP:

1. Let $\{x_u\}$ be a feasible solution for $P(G)$.
2. Let $\{X_{u,v}\}$ be a feasible solution for $\text{sa}^k \text{iso}(G, H)$.
3. Define:

$$y_v := \sum_{u \in G^k} X_{u,v} \cdot x_u.$$

4. Check that $\{y_v\}$ is a feasible solution for $P(H)$.

More examples of local LPs

More examples:

1. maximum flows (2-local)
2. if P is r -local LP, then sa^k - P is rk -local LP.
3. if P is r -local LP, then sos^k - P is rk -local SDP.

Back to integrality gaps for vertex cover

Goal:

For large k and every $\epsilon > 0$ find graphs G and H such that

1. $G \equiv_{C_{c \cdot 2k}}^C H$
2. $\text{vc}(G) \geq (2 - \epsilon)\text{vc}(H)$

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It would follow that:

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Proof:

$$\begin{aligned} \text{vc}(G) &\geq (2 - \epsilon)\text{vc}(H) && \text{by 2.} \\ &\geq (2 - \epsilon)\text{sos}^k \text{fvc}(H) && \text{obvious} \\ &\geq (2 - \epsilon)\text{sos}^k \text{fvc}(G) && \text{by 1. and 2-locality} \end{aligned}$$

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1. $G \equiv_{C_{c \cdot 2k}}^C H$
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A weak (easy) case: $k = 1$ with gap = 2

Choose:

$G =$ any d -regular expander graph (i.e., $\lambda_2(G) \ll \lambda_1(G)$),

$H =$ any d -regular bipartite graph.

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$H =$ any d -regular bipartite graph.

Then:

$\text{vc}(G) = (1 - \epsilon)n$ by expansion

$\text{vc}(H) = n/2$ by bipartition

$G \equiv^R H$ by regularity

$G \equiv_2^C H$ by Tinhofer's Theorem

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$H =$ any d -regular bipartite graph.

Then:

$\text{vc}(G) = (1 - \epsilon)n$ by expansion

$\text{vc}(H) = n/2$ by bipartition

$G \equiv^R H$ by regularity

$G \equiv_2^C H$ by Tinhofer's Theorem

Tight in two ways:

$G \not\equiv_3^C H$ bipartiteness is C^3 -definable,
 $G \equiv_2^C H \implies \text{vc}(G) \leq 2\text{vc}(H)$ [AA-Dawar 2018]

A different weak (harder) case: $k = \Omega(n)$ but gap = 1.08

Theorem [AA-Dawar 2018]

There exist graphs G_n and H_n such that

1. $G_n \equiv_{\Omega(n)}^C H_n$
2. $\text{vc}(G_n) \geq 1.08 \cdot \text{vc}(H_n)$

Part IV

PROOF INGREDIENTS

1/3: Locally consistent systems of linear equations

Ingredient 1: A linear system $Ax = b$ over \mathbb{F}_2 where:

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Probabilistic construction:

1. set $m = cn$ for a large constant $c = c(\epsilon)$
2. choose three ones uniformly at random in each row of A
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Half-deterministic construction:

1. set $m = cn$ for a large constant $c = c(\epsilon)$
2. let A be incidence matrix of bipartite expander
3. choose b uniformly at random in \mathbb{F}_2^n .

2/3: Indistinguishable systems of linear equations

Ingredient 2: A pair of linear systems S_0 and S_1 over \mathbb{F}_2 where:

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Construction of S_0 :

1. start with $Ax = b$ from previous section
2. duplicate each variable $x \mapsto (x^{(0)}, x^{(1)})$
3. replace each equation $x_i + x_j + x_k = b$ by 8 equations

$$x_i^{(u)} + x_j^{(v)} + x_k^{(w)} = b + u + v + w$$

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Construction of S_1 :

1. same but start with $Ax = 0$ (the homogeneous system)

3/3: Reduction to vertex cover

Ingredient 3: A pair of graphs G_0 and G_1 where:

1. $G_0 \equiv_{\Omega(n)}^C G_1$
2. $\text{vc}(G_0) \geq 26m$
3. $\text{vc}(G_1) \leq 24m$

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Construction:

a standard reduction from \mathbb{F}_2 -SAT to vertex cover

Open Problem 1

$$\sup_G \frac{vc(G)}{\text{sos}^4 fvc(G)} > 1.36?$$

Open Problem 2

find strongly regular graphs G and H with same parameters
so that $\text{vc}(G) \geq (2 - \epsilon)\text{vc}(H)$.

Acknowledgments

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