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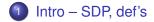
Vector chromatic number – where graph theory meets semidefinite programming

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UNCE seminar — Apr 26, 2012









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Semidefinite programming (SDP)

 $X \succeq 0 \Leftrightarrow X$ is positive semidefinite (PSD) \Leftrightarrow there are vectors v_1, \ldots, v_n so that $X_{i,j} = v_i^T v_j$ (X is Gram matrix)

- SDP: inf $C \bullet X$: $(\forall i) A_i \bullet X = b_i, X \succcurlyeq 0, X \in \mathbb{R}^{n \times n}$
- when all matrices are diagonal, we get classical linear programming
- effectively solvable for every ε > 0 one can in time polynomial to the size of the input and log ¹/_ε approximate solution with precision ε.
- in most cases, duality holds, similarly as for LP

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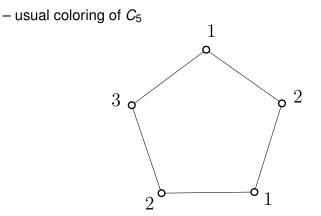
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Vector coloring



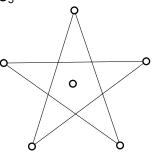
- vector coloring of C_5

We are trying to assign **unit vectors** to vertices of the graph, so that that adjacent vertices are **far apart**.

- (variant: strict vector coloring – all edges have to be of the same length)

Vector coloring

- usual coloring of C₅
- vector coloring of C₅

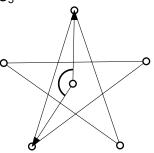


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Vector coloring – definition

Definition (Karger, Motwani and Sudan, 1998)

Given: graph *G* with *n* vertices **Find:** minimal t < 0 s.t. $\exists f : V(G) \to \mathbb{R}^n$ • |f(v)| = 1 $\forall v \in V(G)$ and • $\langle f(u), f(v) \rangle \leq t$ $\forall uv \in E(G)$. t(G) := minimal such *t*. $\chi_v(G) := 1 - \frac{1}{t(G)} \dots$ vector chromatic number

t(G) is defined by a semidefinite program \implies the minimum exists and can be approximated with an absolute error $< \varepsilon$ in time polynomial in *n* and log $\frac{1}{\varepsilon}$.

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 $\chi_{sv}(G) = \vartheta(\bar{G}) =: \bar{\vartheta}(G)$

Original motivation – Θ

Definition (Shannon, 1940's)

Shannon capacity - communication capacity of a channel

$$\Theta(G) = \lim_{n \to \infty} \sqrt[n]{\alpha(G \boxtimes G \boxtimes \cdots \boxtimes G)}$$

Problem: $\Theta(C_5) = ?$.

Theorem (Lovász, 1979)

There is a graph parameter $\vartheta(G)$, such that

• $\vartheta(\mathbf{G}) \geq \alpha(\mathbf{G})$

• $\vartheta(G \boxtimes H) = \vartheta(G)\vartheta(H) \quad \Rightarrow \quad \vartheta(G) \ge \Theta(G)$

• $\vartheta(C_5)$ can be computed easily

... in fact $\vartheta(G)$ can be computed for any G in a polynomial time

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Approximating chromatic number

Theorem (Karger, Motwani, Sudan, 1998)

$$\chi_{\mathbf{v}}(\mathbf{G}) \leq \mathbf{k} \quad \Rightarrow \quad \chi(\mathbf{G}) \leq \Delta^{1-2/k+o(1)}$$

assuming k constant, $\Delta = \Delta(G)$. **Moreover**, such coloring can be found in a polynomial time.

Theorem (Feige, Langberg, Schachtman, 2004)

This is basically best possible.

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Approximating chromatic number

Theorem (Coja-Oghlan, 2005)

Existence of a k-coloring of G(n, p) can be tested in average polynomial time (provided $pn \ge ck^2$).

Idea: The probability that χ_v will not provide the desired bound is exponentially small. Thus in such case we can use some algorithm with exponential running time. **Required ingredience:** understanding of the behaviour of $\chi_v(G(n, p))$ and $\vartheta(G_{n,p})$.

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Approximating cubical chromatic number

Theorem (Š., 2010+)

For every graph G we have

$$\chi_{\boldsymbol{v}}(\boldsymbol{G}) \leq \chi_{\boldsymbol{q}}(\boldsymbol{G}) \leq \frac{\pi}{2} \chi_{\boldsymbol{v}}(\boldsymbol{G})$$

Here χ_q is the cubical chromatic number.

Vector chromatic number as a worthy graph parameter

Observation

$$G \to H \quad \Rightarrow \chi_{\nu}(G) \leq \chi_{\nu}(H)$$

(And the same is true for $\chi_{sv} = \bar{\vartheta}$.)

Thus χ_{ν} is a "coloring-type" parameter and we may understand it as a mean to understand the homomorphism structure of graphs.

Compared to other such parameters (χ , χ_c , χ_f , ...) this one is computationally tractable (there is a polynomial-time algorithm to approximate it).

Example 1: values for random graph

Coja-Oghlan – concentrated on an interval of width one Computational evidence: converges to a normal distribution, if p = const. $\chi_{sv}(G(100,.5))$ 1.5 0.5

two-point distribution, rather surprising phenomenon.

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Example 1: values for random graph

Coja-Oghlan – concentrated on an interval of width one Computational evidence: converges to a normal distribution, if p = const.

However, if p = o(1), it seems that $\chi_V(G(n, p))$ converges to a two-point distribution, rather surprising phenomenon. $\chi_{sv}(G(200,.1))$

Example 2: product (Hedetniemi) conjecture

Observation

G imes H o GG imes H o H

Thus: $\chi(\mathbf{G} \times \mathbf{H}) \leq \min{\{\chi(\mathbf{G}), \chi(\mathbf{H})\}}.$

Conjecture (Hedetniemi, 1966)

 $\chi(\mathbf{G}\times\mathbf{H})=\min\{\chi(\mathbf{G}),\chi(\mathbf{H})\}.$

Note: a simple exercise is to show that $\chi(G \Box H) = \max{\chi(G), \chi(H)}$: we have $G, H \rightarrow G \Box H$ and it is not hard to actually provide the coloring.

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Example 2: product conjecture for $\bar{\vartheta}$

Observation

G imes H o GG imes H o H

Thus: $\bar{\vartheta}(G \times H) \leq \min\{\bar{\vartheta}(G), \bar{\vartheta}(H)\}.$

Conjecture (Š., 2011)

 $\bar{\vartheta}(G \times H) = \min\{\bar{\vartheta}(G), \bar{\vartheta}(H)\}.$

Example 2: product conjecture for $\bar{\vartheta}$ – approach

Compared to the normal chromatic number, we can use SDP duality – so that we have a certificate for $\chi_{\nu}(G) \ge c$ (whereas for normal coloring one only has certificate for $\chi(G) \le c$ – namely a coloring using *c* colors).

Using this idea and some variants of $\bar{\vartheta}$, one can find that

$$\bar{\vartheta}(G \Box H) \bar{\vartheta}(G \times H) \geq \bar{\vartheta}(G) \bar{\vartheta}(H)$$
.

As $\bar{\vartheta}(G \Box H) \ge \max{\{\bar{\vartheta}(G), \bar{\vartheta}(H)\}}$, it's enough to show also the reverse inequality for $\bar{\vartheta}(G \Box H)$.

Example 2: product conjecture for $\bar{\vartheta}$ – matrix completion

... it's enough to show also the reverse inequality for $\bar{\vartheta}(G \Box H)$ (this was trivial for χ).

It leads to the following conjecture about matrices:

Conjecture (Š., 2012)

Let A, B be PSD matrices with 1's on diagonals. Then

 $I \otimes B + A \otimes I - I \otimes I$

can be completed to a PSD matrix.

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Example 3 – graphs on surfaces?

- Can you show that $\chi_{\nu}(G) < 5$ for every planar graph *G*?
- Can you show version of the Hadwiger conjecture?