Polynomial optimization on finite sets.

Mauricio Velasco Universidad Católica del Uruguay (UCU)

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LECTURE 2: Sparsity in Sums of squares.

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- The Theorem implies that $f_{\min} = f_{(\frac{n}{2}+1)}$ for every quadratic polynomial on the hypercube giving us an upper bound on the exactness degree for quadratic functions.
- This bound is better than the n from the previous lecture. It can be extended [Sakaue, Takeda, Kim, Ito (2017)] to f of any degree (see exercises).
- We will prove later that these upper bounds are in fact sharp.



• Recall that $\mathbb{R}[X] = \mathbb{R}[x_1, \dots, x_n] / (x_j^2 - 1 : j = 1, \dots, n)$ and that the square-free monomials $\chi_I := \prod_{i \in I} x_i$ form a **basis**.

Exercise

Prove that the square-free monomials $\{\chi_I : I \subseteq [n]\}$ for $I \subseteq [n]$ form an orthonormal basis of $\mathbb{R}[X]$ for the inner product

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$$\langle f, g \rangle := \frac{1}{|X|} \sum_{y \in X} f(y)g(y)$$

 A quadratic function f involves only a few of the 2ⁿ elements of this basis

$$f = E_0 + \sum_{i=1}^n E_i x_i + \sum_{i < j} E_{ij} x_i x_j$$

and is therefore sparse.



The proofs of the (Fawzi, Parrilo, Saunderson) and (Sakaue, Takeda, Kim, Ito) Theorem leverage the sparsity of f to prove the existence of sparse sum-of-square certificates for $f - \lambda$ (i.e. certificates $g = s_1^2 + \cdots + s_k^2$ where the s_i involve only a few basis elements).

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The key tool for doing this will be the combinatorics of **chordal graphs**.

Organization:

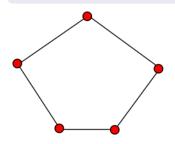
- Preliminaries on Chordal graphs.
- Chordality in SDP and sparsification tools for sum-of-squares.
- **3** Application to Sums-of-squares in the hypercube $\{-1,1\}^4$.

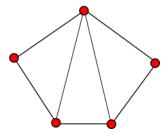
Part 1: A primer on chordal graphs.

Chordal graphs

Definition.

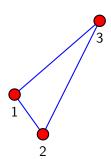
A graph G is chordal if every simple cycle in G of length at least 4 has at least one chord (i.e. an edge that "crosses" the cycle).

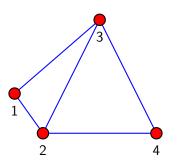


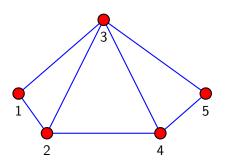


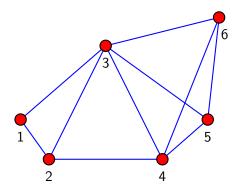


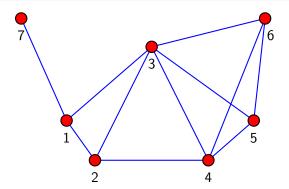












Chordal graphs

Chordal graphs are very useful objects. They appear in numerical linear algebra, semidefinite programming, algebraic geometry, etc. To a first approximation this follows from two facts:

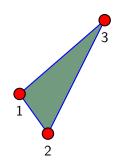
- A perfect elimination ordering allows us to work with chordal graphs inductively very effectively. For instance several NP-hard problems admit efficient (polynomial or even linear time) solutions on chordal graphs.
- ② Chordal graphs are a structured but very rich class. For example every graph is contained in some chordal graph.



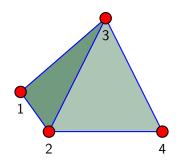
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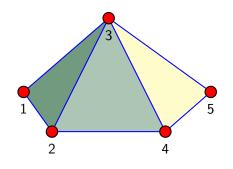
 $\{1, 2\}$



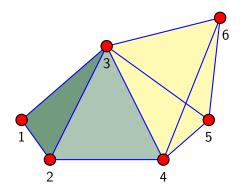
 $\{1, 2, 3\}$



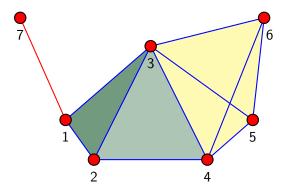
 $\{1,2,3\}\ \{2,3,4\}$



 $\{1,2,3\}\ \{2,3,4\}\quad \{3,4,5\}$



 $\{1,2,3\}\ \{2,3,4\}\ \{3,4,5,6\}$



 $\{1,2,3\}$ $\{2,3,4\}$ $\{3,4,5,6\}$ $\{1,7\}$

Part 2:

Chordality in SDP and sparsification tools for sums-of-squares.

Chordality in SDPs

Let A be a symmetric matrix in $\mathbb{R}^{n \times n}$.

Definition.

An (undirected, loopless) graph H with vertex set [n] is an admissible support graph for A if

$$A_{ij} \neq 0$$
 with $i \neq j \implies \{i,j\} \in E(H)$

Example:

Chordality in SDPs

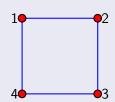
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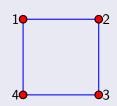
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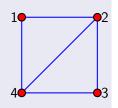
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Example:





Let A be a symmetric matrix and H a support graph for A.

Theorem. (Grone and Johnson, 1984)

Assume H is a chordal graph and let $C_1, \ldots, C_L \subseteq [n]$ be its collection of maximal cliques. The matrix $A \succeq 0$ if and only if there exist PSD matrices B_1, \ldots, B_L with $(B_j)_{s,t} = 0$ outside the edges of C_j such that

$$A=B_1+\cdots+B_L$$

Remark.

The Lemma is very useful computationally because it transforms the large semidefinite constraint $A \succeq 0$ into several small semidefinite constraints $B_1 \succeq 0,..., B_L \succeq 0$.

Example:

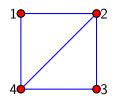
Is the following matrix $A \succeq 0$?

$$A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 2.2 & 1.1 & 0 & 1.1 \\ 2 & 1.1 & 2.2 & 1.1 & 0 \\ 3 & 0 & 1.1 & 2.2 & 1.1 \\ 4 & 1.1 & 0 & 1.1 & 2.2 \end{bmatrix}$$

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Is the following matrix $A \succeq 0$?

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Example:

Is the following matrix $A \succeq 0$?

The chordal cover has cliques $\{1, 2, 4\}$ and $\{2, 3, 4\}$. so

$$A = \left(\begin{array}{cccc} a & b & 0 & f \\ b & d & 0 & e \\ 0 & 0 & 0 & 0 \\ f & e & 0 & g \end{array}\right) + \left(\begin{array}{cccc} 0 & 0 & 0 & 0 \\ 0 & \alpha & \beta & \phi \\ 0 & \beta & \delta & \epsilon \\ 0 & \phi & \epsilon & \iota \end{array}\right)$$

where both matrices are really 3×3 PSD.



Chordal sparsification of sums-of-squares:

Let X be finite. Assume χ_1, \ldots, χ_N is a basis for $\mathbb{R}[X]$ and let $\vec{\chi} := (\chi_1, \ldots, \chi_N)$.

Corollary.

Assume $f = \vec{\chi}^t A \vec{\chi}$ for some $A \succeq 0$ with support H.

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Corollary.

Assume $f = \vec{\chi}^t A \vec{\chi}$ for some $A \succeq 0$ with support H. If C_1, \ldots, C_L are the maximal cliques of any chordal cover of H then f admits a sum-of-squares representation

$$f=t_1+\cdots+t_L$$

with sum-of-square summands $t_i \in \Sigma_{\langle \{\chi_j: j \in C_j\} \rangle}$

Proof.

By Grone and Johnson if $A \succeq 0$ has sparsity graph H then $A = B_1 + \cdots + B_L$ so $\vec{\chi}^t A \vec{\chi} = \vec{\chi}^t B_1 \vec{\chi} + \cdots + \vec{\chi}^t B_L \vec{\chi}$.



Shift sparsification of sums-of-squares:

Assume X is a finite set. We can also **shift the support** as follows,

Lemma.

Suppose $h \in \mathbb{R}[X]$ satisfies $h^2 = 1$ and let $V, W \subseteq \mathbb{R}[X]$ be subspaces. If $hV \subseteq W$ then $\Sigma_V \subseteq \Sigma_W$

Proof.

For any $v \in V$ there is $w \in W$ such that hv = w. Squaring we get

$$w^2 = (hv)^2 = h^2v^2 = v^2$$

summing over the v's we conclude $\Sigma_V \subseteq \Sigma_W$

Remark.

In particular the inclusion $\Sigma_V + \Sigma_W \subseteq \Sigma_W$ holds.

Part 3:

Application: Sparsification in the hypercube $X := \{-1, 1\}^n$.

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• Since $f \ge 0$ we can write an **initial sum-of-squares** representation of g in the Lagrange interpolation basis $f = \sum_{y \in X} f(y) p_y^2$, that is $f = \vec{p}^t D \vec{p}$ for a nonnegative diagonal matrix D.

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Replacing the second expression in the first we obtain

$$f = \vec{\chi}^t A \vec{\chi}$$
 where $A = V^t D V$ is PSD.

It follows that $A_{\chi_I,\chi_I} := \langle f, \chi_I \chi_J \rangle$.



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• We conclude that $A_{\chi_I,\chi_J} \neq 0$ is equivalent to

$$\langle f, \chi_I \chi_J \rangle = \langle f, \chi_{I \triangle J} \rangle \neq 0$$

which, by orthonormality of the χ_I , is equivalent to $I\Delta J$ appearing in f.



Support graph

Our initial support graph for $f \in \mathbb{R}[X]_{\leq 2}$ can be now be constructed

Definition.

Let H be the graph with vertices $\{\chi_I : I \subseteq [n]\}$ and

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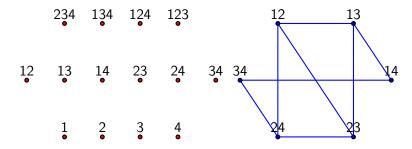
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During the lecture we will focus on n = 4.

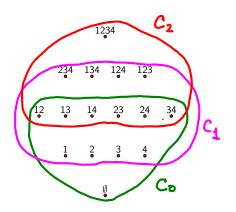


Support graph for n = 4

1234



A chordal cover:



40 × 40 × 42 × 42 × 2 990

The cliques are C_0 , C_1 , C_2 and $\chi_{1234}C_2 \subseteq C_0$.



We have thus proven...

Theorem. (Fawzi, Saunderson, Parrilo for n = 4)

If $f \in \mathbb{R}[X]_{\leq 2}$ is nonnegative then $f = t_0 + t_1$ with $t_0 \in \Sigma_{C_0}$ and $t_1 \in \Sigma_{C_1}$. In particular f is expressible as a sum of squares of functions of degree at most 3 = n/2 + 1.

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Exercise

Prove that if $f \in \mathbb{R}[X]_{\leq d}$ is nonnegative and n is even then f can be written as a sum of squares of elements of $\mathbb{R}[X]_{\leq \frac{n+d}{2}}$.

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More specifically, for a graph H let

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Question.

How to use all these other varieties in optimization?

