

A Generalization of SAGE Certificates for Constrained Optimization

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Joint work with Venkat Chandrasekaran and Adam Wierman (Caltech).

Nonnegativity and optimization



Given a function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ and a set $X \subset \mathbb{R}^n$, we have

$$\begin{aligned} f_X^* &= \inf\{f(\mathbf{x}) : \mathbf{x} \text{ in } X\} \\ &= \sup\{\gamma : f(\mathbf{x}) \geq \gamma \text{ for all } \mathbf{x} \text{ in } X\} \\ &= \sup\{\gamma : f - \gamma \text{ is nonnegative over } X\}. \end{aligned}$$

How do we test if f (or $f - \gamma$) is nonnegative over X ?

- This is usually NP-Hard, even for $X = \mathbb{R}^n$.
- Desire *sufficient conditions* that f is nonnegative over X .
- Different sufficient conditions for different function classes.

Signomials and polynomials



Parameters α_i in \mathbb{R}^n , c_i in \mathbb{R} .

$$\mathbf{x} \mapsto \sum_{i=1}^m c_i \exp(\alpha_i \cdot \mathbf{x})$$

Uncountable basis.

Complexity measured by number of terms m .

Nonnegativity certificates

- SAGE

Parameters α_i in \mathbb{N}^n , c_i in \mathbb{R} .

$$\mathbf{x} \mapsto \sum_{i=1}^m c_i \prod_{j=1}^n x_j^{\alpha_{ij}}$$

Countable basis.

Complexity measured by degree $\max_{ij} \alpha_{ij}$.

Select nonnegativity certificates

- SOS
- SONC
- SAGE

The signomial nonnegativity cones



Write $f = \text{Sig}(\boldsymbol{\alpha}, \mathbf{c})$ to mean $f(\mathbf{x}) = \sum_{i=1}^m c_i \exp(\boldsymbol{\alpha}_i \cdot \mathbf{x})$.

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Define the nonnegativity cone for signomials over exponents $\boldsymbol{\alpha}$:

$$C_{\text{NNS}}(\boldsymbol{\alpha}) \doteq \{ \mathbf{c} : \text{Sig}(\boldsymbol{\alpha}, \mathbf{c})(\mathbf{x}) \geq 0 \text{ for all } \mathbf{x} \text{ in } \mathbb{R}^n \}.$$

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These nonnegativity cones exhibit affine-invariance:

$$C_{\text{NNS}}(\boldsymbol{\alpha}) = C_{\text{NNS}}(\boldsymbol{\alpha} - \mathbf{1}\mathbf{u}) = C_{\text{NNS}}(\boldsymbol{\alpha}\mathbf{V})$$

for all row vectors \mathbf{u} in \mathbb{R}^n and all invertible \mathbf{V} in $\mathbb{R}^{n \times n}$.

SAGE certificates for signomials



Definition. A nonnegative signomial with at most one negative coefficient is an “**AM/GM Exponential**,” or an “AGE function.”

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For each k , have cone of coefficients for AM/GM Exponentials

$$C_{\text{AGE}}(\alpha, k) \doteq \{c : c_{\setminus k} \geq \mathbf{0} \text{ and } c \text{ in } C_{\text{NNS}}(\alpha)\}.$$

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We take sums of AGE cones to obtain the **SAGE cone**

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Crucial question: How to represent the AGE cones?

The convex duality behind AGE cones

Fix α in $\mathbb{R}^{m \times n}$, and c in \mathbb{R}^m satisfying $c_{\setminus k} \geq \mathbf{0}$.

Does c belong to $C_{\text{NNS}}(\alpha)$?

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$$\begin{aligned} \text{Sig}(\alpha, c)(x) \geq 0 &\Leftrightarrow \text{Sig}(\alpha - \mathbf{1}\alpha_k, c)(x) \geq 0 \\ &\text{Sig}(\alpha_{\setminus k} - \mathbf{1}\alpha_k, c_{\setminus k})(x) \geq -c_k. \end{aligned}$$

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$$\inf_{\mathbf{x} \in \mathbb{R}^n} \text{Sig}(\alpha_{\setminus k} - \mathbf{1}\alpha_k, c_{\setminus k})(\mathbf{x}) \geq -c_k$$

holds **if and only if**

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SAGE certificates for signomials



The preceding slides established that

$$C_{\text{AGE}}(\boldsymbol{\alpha}, k) = \{ \mathbf{c} : \mathbf{c}_{\setminus k} \geq \mathbf{0}, \text{ and there exists } \boldsymbol{\nu} \text{ in } \mathbb{R}^{m-1} \text{ satisfying} \\ D(\boldsymbol{\nu}, \mathbf{c}_{\setminus k}) - \boldsymbol{\nu}^\top \mathbf{1} \leq c_k \text{ and } [\boldsymbol{\alpha}_{\setminus k} - \mathbf{1}\alpha_k]^\top \boldsymbol{\nu} = \mathbf{0} \}.$$

This allows us to optimize over

$$C_{\text{SAGE}}(\boldsymbol{\alpha}) = \sum_{k=1}^m C_{\text{AGE}}(\boldsymbol{\alpha}, k)$$

with a convex program of size $O(m^2)$.

Handling conditional nonnegativity

Want sufficient conditions that f is nonnegative over X .

Typical approach: search for a function \mathcal{L} for which we can prove

$$0 \leq \mathcal{L}(\mathbf{x}) \text{ for all } \mathbf{x} \text{ in } \mathbb{R}^n \quad \text{and} \quad \mathcal{L}(\mathbf{x}) \leq f(\mathbf{x}) \text{ for all } \mathbf{x} \text{ in } X$$

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Concretely, we fix a representation $X = \{\mathbf{x} : g(\mathbf{x}) \geq \mathbf{0}\}$, and use

$$\mathcal{L} = f - \sum_i \lambda_i g_i$$

with dual variables $\lambda_i : \mathbb{R}^n \rightarrow \mathbb{R}_+$. Apply your favorite (tractable!) proof system to ensure that λ_i and \mathcal{L} are nonnegative functions.

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We will be taking a different route.

X-SAGE Signomials

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X-SAGE certificates for signomials



Definition. A signomial which is nonnegative over X and which has at most one negative coefficient is called an “X-AGE function.”

For each k , have cone of coefficients for X-AGE functions

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This definition assumes nothing of the set X !

Basic properties



Conditional SAGE cones ...

- are order-reversing in the set-argument.

If $X_2 \subset X_1$, then $C_{\text{SAGE}}(\boldsymbol{\alpha}, X_1) \subset C_{\text{SAGE}}(\boldsymbol{\alpha}, X_2)$.

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- *are tractable whenever X is a tractable convex set.*

Representing X -AGE cones

Fix α in $\mathbb{R}^{m \times n}$, and c in \mathbb{R}^m satisfying $c_{\setminus k} \geq \mathbf{0}$. Convex $X \subset \mathbb{R}^n$.

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$$\begin{aligned} \sigma_X(\lambda) + D(\nu, c_{\setminus k}) - \nu^\top \mathbf{1} &\leq c_k, \text{ and} \\ [\alpha_{\setminus k} - \mathbf{1}\alpha_k]^\top \nu + \lambda &= \mathbf{0}. \end{aligned}$$

Representing X -AGE cones

The preceding slides established that when X is convex,

$$C_{\text{AGE}}(\boldsymbol{\alpha}, k, X) = \{ \mathbf{c} : \mathbf{c}_{\setminus k} \geq \mathbf{0}, \text{ and there exist } \boldsymbol{\nu} \text{ in } \mathbb{R}^{m-1}, \boldsymbol{\lambda} \text{ in } \mathbb{R}^n \\ \text{satisfying } [\boldsymbol{\alpha}_{\setminus k} - \mathbf{1}\boldsymbol{\alpha}_k]^\top \boldsymbol{\nu} + \boldsymbol{\lambda} = \mathbf{0}, \\ \text{and } \sigma_X(\boldsymbol{\lambda}) + D(\boldsymbol{\nu}, \mathbf{c}_{\setminus k}) - \boldsymbol{\nu}^\top \mathbf{1} \leq c_k \}.$$

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For implementations, suppose $X = \{ \mathbf{x} : \mathbf{A}\mathbf{x} + \mathbf{b} \in K \}$.

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$$\sigma_X(\boldsymbol{\lambda}) \doteq \sup_{\mathbf{x} \in X} \boldsymbol{\lambda}^\top \mathbf{x} \leq \inf \{ \mathbf{b}^\top \boldsymbol{\eta} : \mathbf{A}^\top \boldsymbol{\eta} + \boldsymbol{\lambda} = \mathbf{0}, \boldsymbol{\eta} \in K^\dagger \},$$

and equality holds generically.

Dual SAGE relaxations



Let $f = \text{Sig}(\boldsymbol{\alpha}, \mathbf{c})$ have $\alpha_1 = 0$.

The primal and dual SAGE relaxations for f_X^* are

$$\begin{aligned} f_X^{\text{SAGE}} &= \sup\{\gamma : \mathbf{c} - \gamma(1, 0, \dots, 0) \text{ in } C_{\text{SAGE}}(\boldsymbol{\alpha}, X)\} \\ &= \inf\{\mathbf{c}^\top \mathbf{v} : v_1 = 1 \text{ and } \mathbf{v} \text{ in } C_{\text{SAGE}}(\boldsymbol{\alpha}, X)^\dagger\}. \end{aligned}$$

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When X is convex, the dual X-SAGE cone can be expressed as

$$\begin{aligned} C_{\text{SAGE}}(\boldsymbol{\alpha}, X)^\dagger &= \text{cl}\{\mathbf{v} : \text{some } \mathbf{z}_1, \dots, \mathbf{z}_m \text{ in } \mathbb{R}^n \text{ satisfy} \\ &\quad v_k \log(\mathbf{v}/v_k) \geq [\boldsymbol{\alpha} - \mathbf{1}\alpha_k]\mathbf{z}_k \\ &\quad \text{and } \mathbf{z}_k/v_k \in X \text{ for all } k \text{ in } [m]\}. \end{aligned}$$

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Solution recovery? Consider vectors $\mathbf{x}_k = \mathbf{z}_k/v_k$ for k in $[m]$.

A Signomial Example



$$\inf_{\mathbf{x} \in \mathbb{R}^3} f(\mathbf{x}) \doteq 0.5 \exp(x_1 - x_2) - \exp x_1 - 5 \exp(-x_2)$$

$$\text{s.t. } 100 - \exp(x_2 - x_3) - \exp x_2 - 0.05 \exp(x_1 + x_3) \geq 0$$

$$\exp \mathbf{x} - (70, 1, 0.5) \geq \mathbf{0}$$

$$(150, 30, 21) - \exp \mathbf{x} \geq \mathbf{0}$$

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Compute $f_X^{\text{SAGE}} = -147.85713 \leq f_X^*$, recover feasible

$$\mathbf{x}^* = (5.01063529, 3.40119660, -0.48450710)$$

satisfying $f(\mathbf{x}^*) = -147.66666$. *This is actually optimal!*

X -SAGE polynomials

Geometric-form signomials



If $\mathbf{x} > 0$, then

$$\mathbf{x} \mapsto \sum_{i=1}^m c_i \mathbf{x}^{\alpha_i}$$

is defined for any real α_i .

For $X \subset \mathbb{R}_{++}^n$ and α in $\mathbb{R}^{m \times n}$, define

$$C_{\text{NNS}}^{\text{GEOM}}(\alpha, X) = \{c : \sum_{i=1}^m c_i \mathbf{x}^{\alpha_i} \geq 0 \text{ for all } \mathbf{x} \text{ in } X\}.$$

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For $X \subset \mathbb{R}_{++}^n$ and α in $\mathbb{R}^{m \times n}$, define

$$C_{\text{NNS}}^{\text{GEOM}}(\alpha, X) = \{c : \sum_{i=1}^m c_i \mathbf{x}^{\alpha_i} \geq 0 \text{ for all } \mathbf{x} \text{ in } X\}.$$

From a change of variables $\mathbf{x} \mapsto \exp \mathbf{y}$, we have

$$C_{\text{NNS}}^{\text{GEOM}}(\alpha, X) = C_{\text{NNS}}(\alpha, \log X).$$

Geometric-form signomials

If $\mathbf{x} > 0$, then

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Thus we naturally define

$$C_{\text{SAGE}}^{\text{GEOM}}(\alpha, X) \doteq C_{\text{SAGE}}(\alpha, \log X).$$

The polynomial X -nonnegativity cones

Fix α in $\mathbb{N}^{m \times n}$. Write $f = \text{Poly}(\alpha, c)$ to mean

$$f(\mathbf{x}) = \sum_{i=1}^m c_i \mathbf{x}^{\alpha_i}, \quad \text{where} \quad \mathbf{x}^{\alpha_i} \doteq \prod_{j=1}^n x_j^{\alpha_{ij}}.$$

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The matrix α and the set X induce a nonnegativity cone

$$C_{\text{NNP}}(\alpha, X) = \{\mathbf{c} : \text{Poly}(\alpha, \mathbf{c})(\mathbf{x}) \geq 0 \text{ for all } \mathbf{x} \text{ in } X\}.$$

X -SAGE certificates for polynomials



We call $f = \text{Poly}(\alpha, c)$ an X -**AGE** polynomial if

- 1 c belongs to $C_{\text{NNP}}(\alpha, X)$, and
- 2 at most one “ i ” has $c_i x^{\alpha_i} < 0$ for some $x \in X$.

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In conic form, we can express $C_{\text{AGE}}^{\text{POLY}}(\boldsymbol{\alpha}, i, X) =$

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Define the X-SAGE polynomial cone in the natural way:

$$C_{\text{SAGE}}(\alpha, X) = \sum_{i=1}^m C_{\text{AGE}}^{\text{POLY}}(\alpha, i, X).$$

Representing $C_{\text{SAGE}}^{\text{POLY}}(\alpha, X)$

Consider the case when X is contained within a single orthant.

W.l.o.g, take $X \subset \mathbb{R}_+^n$.

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we have

$$C_{\text{SAGE}}^{\text{POLY}}(\boldsymbol{\alpha}, X) = C_{\text{SAGE}}(\boldsymbol{\alpha}, Y).$$

Representing $C_{\text{SAGE}}^{\text{POLY}}(\alpha, X)$

Define the set of signomial-representative coefficient vectors

$$\text{SR}(\alpha, c) = \{ \hat{c} : \hat{c}_i = c_i \text{ whenever } \alpha_i \text{ is in } 2\mathbb{N}^n, \text{ and} \\ \hat{c}_i \leq -|c_i| \text{ whenever } \alpha_i \text{ is not in } 2\mathbb{N}^n \}.$$

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$$C_{\text{SAGE}}^{\text{POLY}}(\boldsymbol{\alpha}, X) = \{ \mathbf{c} : \text{SR}(\boldsymbol{\alpha}, \mathbf{c}) \cap C_{\text{SAGE}}(\boldsymbol{\alpha}, Y) \text{ is nonempty} \}.$$

Working with $C_{\text{SAGE}}^{\text{POLY}}(\boldsymbol{\alpha}, X)$

How can we formulate a problem to appeal to previous theorems?

Examples include

$$-a \leq x_j \leq a, \quad \|\mathbf{x}\|_p \leq a, \quad |\mathbf{x}^{\alpha_i}| \geq a, \quad \text{and} \quad x_j^2 = a$$

where $a > 0$ is a fixed constant.

A polynomial example

Minimize

$$f(\mathbf{x}) = -64 \sum_{i=1}^7 \prod_{j \in [7] \setminus \{i\}} x_j$$

over the box $X = [-1/2, 1/2]^7$.

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For numeric computation (with MOSEK, on a large workstation)

- the X -SAGE relaxation takes 0.01 seconds to solve
- the earliest tight SOS relaxation takes 90 seconds to solve.

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