

# Global Optimization via the dual SONC cone & linear Programming

arXiv:2002.09368

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02. December 2020



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# Goal

Minimise exponential sums with finite support  $A \subset \mathbb{R}^n$ , coefficient vector  $c \in \mathbb{R}^A$  and  $x \in \mathbb{R}^n$ .

$$f(x) = \sum_{\alpha \in A} c_{\alpha} e^{\langle x, \alpha \rangle}$$

Special case  $A \subset \mathbb{N}^n$ : Polynomials on  $\mathbb{R}_{>0}^n$ .

## Problems

- $\min_{x \in \mathbb{R}^n} f(x)$  is not convex.
- Solving  $\min_{x \in \mathbb{R}^n} f(x)$  is NP-hard in general.

# Solution

Solving  $\min_{x \in \mathbb{R}^n} f(x)$  is equivalent to solving

$$\max \{ \gamma \in \mathbb{R} : f(x) - \gamma \geq 0 \text{ for all } x \in \mathbb{R}^n \}.$$

Tool for finding lower bounds:

## Certificates of Nonnegativity

- imply nonnegativity,
- hold for a large class of exponential sums/polynomials,
- are easier to test than nonnegativity.

# Certificate Examples

## Sum of Squares (SOS)

If  $f$  can be written as **SOS**,  $f$  is nonnegative.

Requires semidefinite programming.

## Sum of nonnegative Circuits (SONC)

Reznick (1989), Ilman & de Wolff (2016):  
If  $f$  can be written as **SONC**,  $f$  is nonnegative.

Requires geometric/relative entropy programming.  
One interpretation: Use **AM-GM inequality** to certify nonnegativity.

$$\frac{1}{n} \sum_{i=1}^n t_i \geq \sqrt[n]{\prod_{i=1}^n t_i} \implies \frac{x^4 y^2 + x^2 y^4 + 1}{3} \geq \sqrt[3]{x^6 y^6}$$
$$\implies x^4 y^2 + x^2 y^4 - 3x^2 y^2 + 1 \geq 0$$

# Outline

## Remember

$$f(x) = \sum_{\alpha \in A} c_{\alpha} e^{\langle x, \alpha \rangle}$$

Consider SONC cone for fixed support set and **fixed sign distribution**  $A = A^+ \cup A^- \subset \mathbb{R}^n$ , where  $A^{\pm} = \{\alpha \in A : \pm c_{\alpha} > 0\}$

$$\begin{aligned} \mathcal{S}_{A^{\pm}} &= \left\{ f \in \mathbb{R}^{A^+ \cup A^-} : f \text{ is SONC} \right\} \\ &\subset \left\{ f \in \mathbb{R}^{A^+ \cup A^-} : f \geq 0 \right\}. \end{aligned}$$

## Our Contribution

Consider **dual SONC cone**  $\check{\mathcal{S}}_{A^{\pm}}$  and show:

- Containment in  $\check{\mathcal{S}}_{A^{\pm}}$  is certificate of nonnegativity.
- Optimizing over  $\check{\mathcal{S}}_{A^{\pm}}$  is **linear program**.

# The Dual SONC Cone

For  $f \in \mathbb{R}^A$  with coefficient vector  $c \in \mathbb{R}^A$  consider the natural duality pairing

$$v(f) = \sum_{\alpha \in A^+} v_\alpha c_\alpha + \sum_{\beta \in A^-} v_\beta c_\beta.$$

**Dual (signed) SONC cone**

$$\check{\mathcal{S}}_{A^\pm} = \{v \in \check{\mathbb{R}}^A : v(f) \geq 0 \forall f \in \mathcal{S}_{A^\pm}\}$$

For every  $v \in \check{\mathbb{R}}^A$  associate

$$f = \#A^- \cdot \sum_{\alpha \in A^+} v_\alpha e^{\langle x, \alpha \rangle} + \sum_{\beta \in A^-} v_\beta e^{\langle x, \beta \rangle}.$$

# Dual Cone contained in Primal Cone

Using a particular representation of  $\check{\mathcal{S}}_{A^\pm}$ , we can show:

## Proposition

$$\check{\mathcal{S}}_{A^\pm} \subset \mathcal{S}_{A^\pm}$$

→ containment in  $\check{\mathcal{S}}_{A^\pm}$  is certificate of nonnegativity.

# Optimizing over Dual Cone is LP

Using a different representation of  $\check{\mathcal{S}}_{A^\pm}$  yields:

## Theorem

Let  $f = \#A^- \cdot \sum_{\alpha \in A^+} v_\alpha e^{\langle x, \alpha \rangle} + \sum_{\beta \in A^-} v_\beta e^{\langle x, \beta \rangle}$ .

The following linear feasibility program in  $\#A^-$  variables  $(\tau^{(\beta)})_{\beta \in A^-}$  verifies containment in  $\check{\mathcal{S}}_{A^\pm}$ :

$$\ln \left( \frac{|v_\beta|}{v_\alpha} \right) \leq (\alpha - \beta)^T \tau^{(\beta)} \quad \forall \beta \in A^-, \alpha \in A^+$$

## Result

Finding  $\max \{ \gamma \in \mathbb{R} : f(x) - \gamma \in \check{\mathcal{S}}_{A^\pm} \}$   
can be done by solving an LP.

→ Lower bound for

$\max \{ \gamma \in \mathbb{R} : f(x) - \gamma \geq 0 \}$ .



# Conclusion

- Effective algorithm for optimizing over the dual SONC cone.
- Computationally more stable than existing methods as it only requires linear programming.
- Yields, in general, worse bounds than preexisting approaches, since we are only optimizing over a subcone of the primal SONC cone.
- Promising runtimes.

# References

Thank you for your  
attention!