

# An Experimental Classification of Maximal Mediated Sets

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# Nonnegative Polynomials

A polynomial  $f \in \mathbb{R}[\mathbf{x}]$  is called **nonnegative** if  $f(\mathbf{x}) \geq 0$  for all  $\mathbf{x} \in \mathbb{R}^n$ .  
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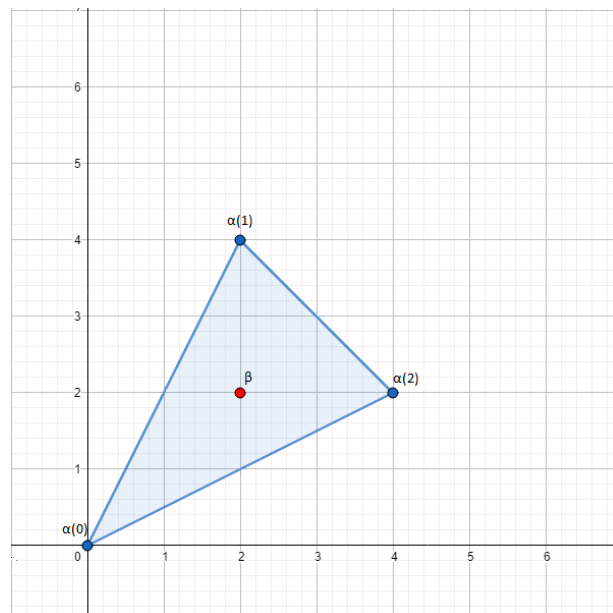
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- The AM-GM inequality can be used to check the nonnegativity the **circuit** polynomials.

# Polynomials Supported on a Circuit

Motzkin Polynomial (1967): Consider the Motzkin polynomial

$$f(x, y) = x^4y^2 + x^2y^4 + 1 - 3x^2y^2$$



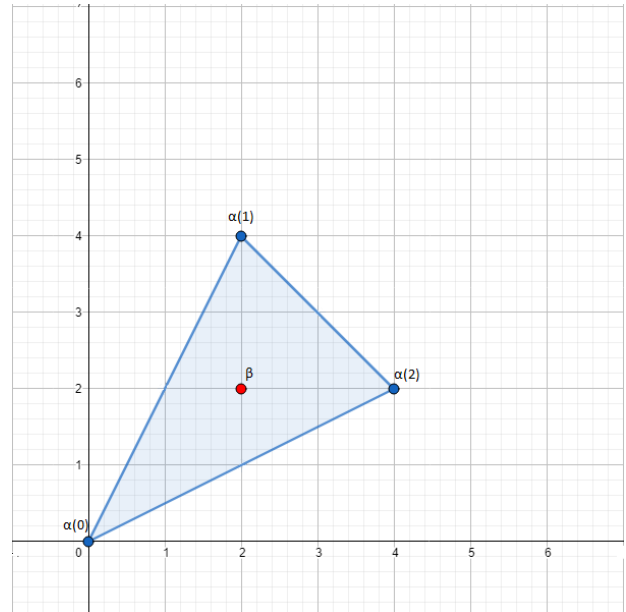


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$f(x, y) \geq 0$  due to the classical AM-GM inequality.



# Maximal Mediated Sets

A set  $L \subseteq \mathbb{Z}^n$  is called  $\Delta$ -mediated if every point in  $L - \Delta$  is midpoint of two distinct points in  $L \cap (2\mathbb{Z})^n$ .

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## Theorem (Reznick (1989), de Wolff, Ilman (2014))

*A nonnegative circuit polynomial  $f$  is SOS if and only if “inner term” is in MMS.*

# MMS Algorithm

Given a set of points  $L \subset \mathbb{Z}^n$ , we define a set of averages:

$$\overline{A}(L) = \left\{ \frac{\mathbf{s} + \mathbf{t}}{2} \mid \mathbf{s}, \mathbf{t} \in L \cap (2\mathbb{Z})^n, \mathbf{s} \neq \mathbf{t} \right\}$$

**Reznick's MMS algorithm(1989):**

**Input:**  $\Delta$ : finite set of points in  $(2\mathbb{Z})^n$

**Output:**  $\Delta^*$ : the  $\Delta$ -mediated subset of  $\mathbb{Z}^n$   
that contains every  $\Delta$ -mediated set

- 1:  $\Delta^0 \leftarrow \text{Conv}(\Delta) \cap \mathbb{Z}^n$
- 2: **repeat**
- 3:      $\Delta^n \leftarrow \overline{A}(\Delta^{n-1}) \cup \Delta$
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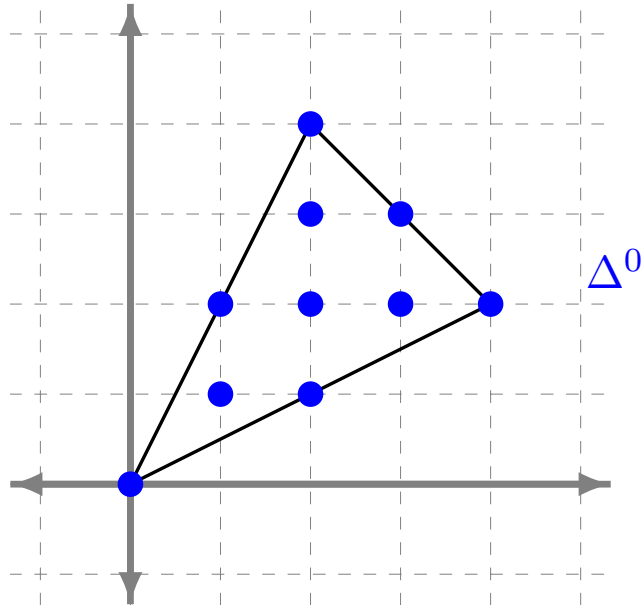
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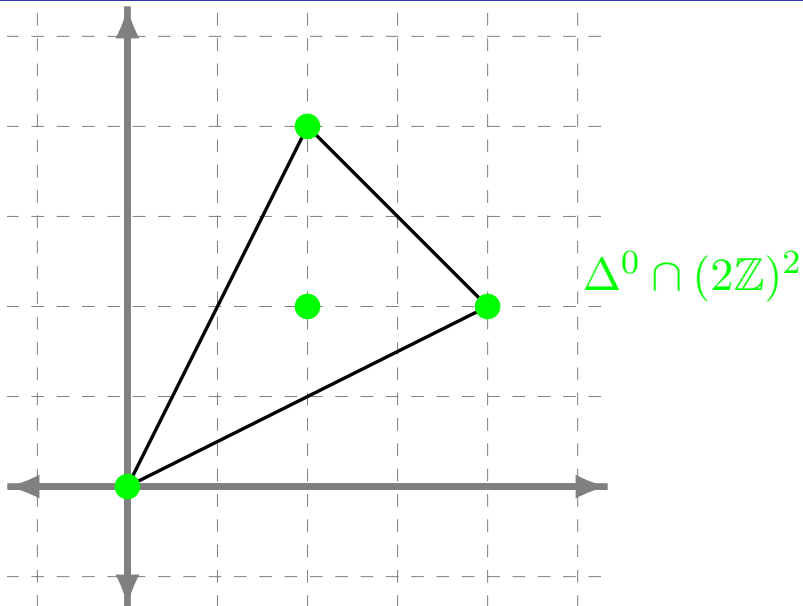
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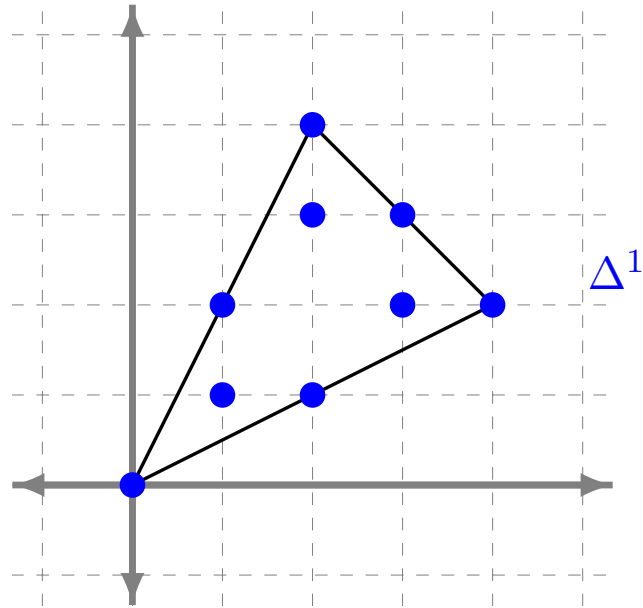
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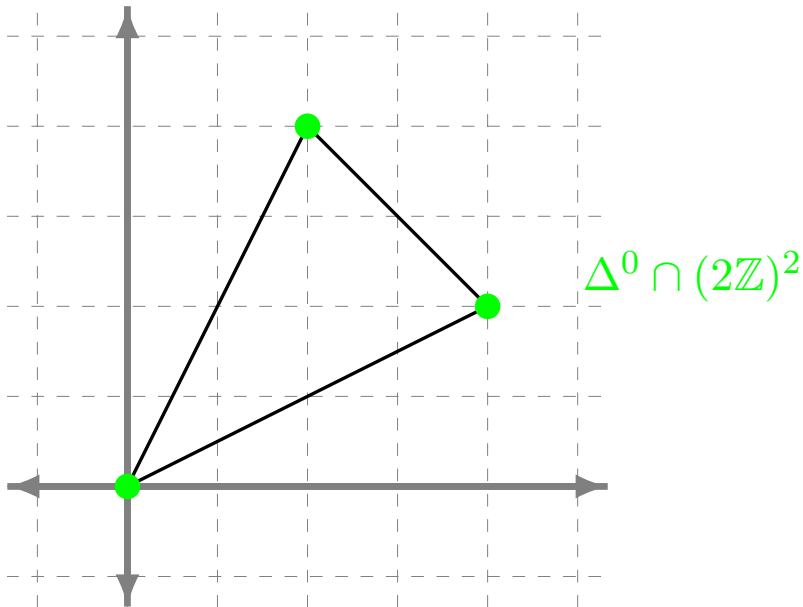
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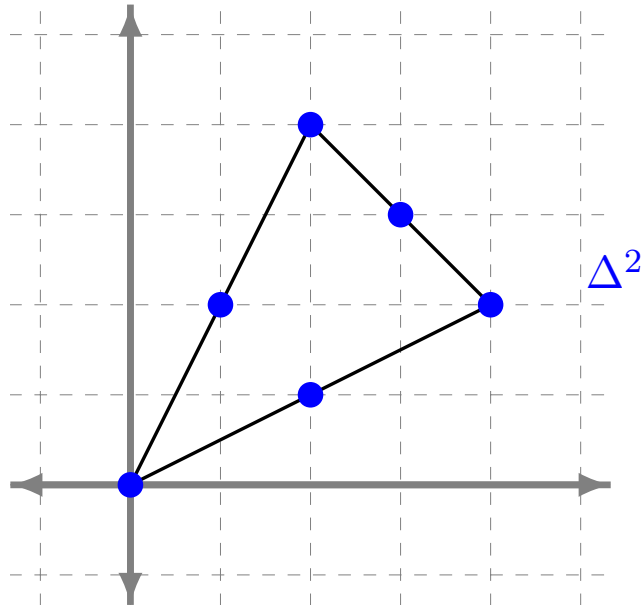
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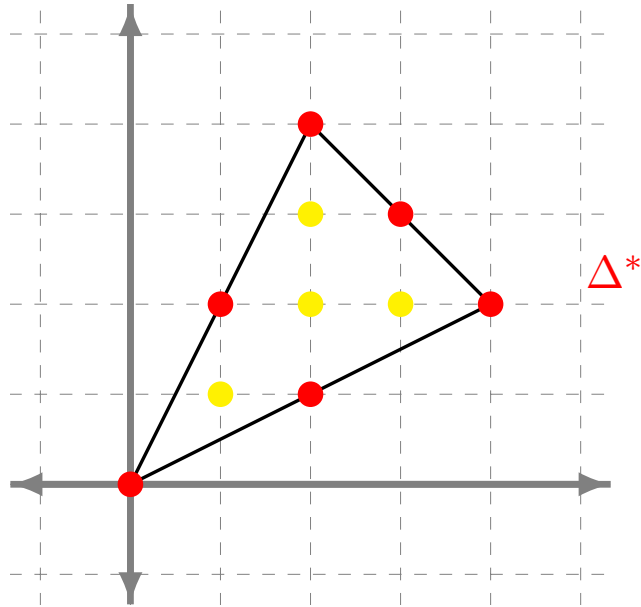
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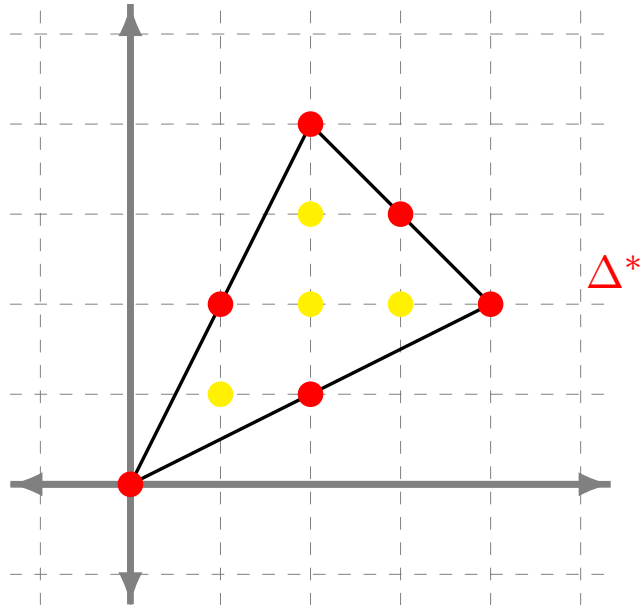
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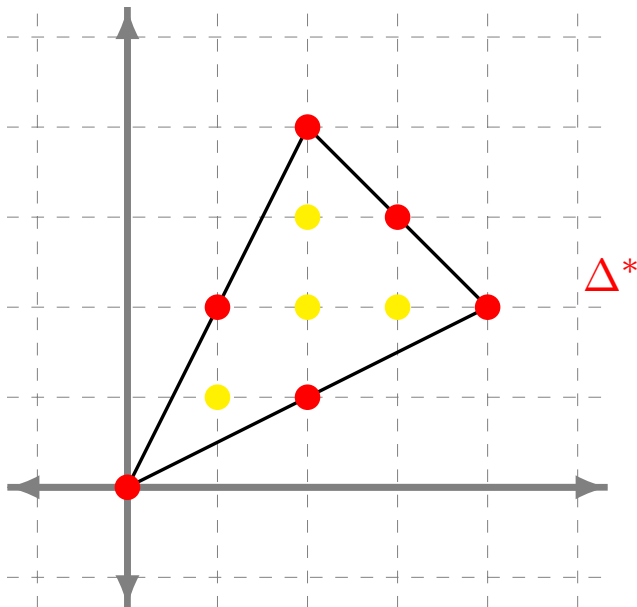
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**Fact:**

$$\bullet \Delta \cup \bar{A}(\Delta) \subseteq \Delta^* \subseteq \text{conv}(\Delta) \cap \mathbb{Z}^n$$

**Task:** Decide how dense  $\Delta^*$  is in  $\text{conv}(\Delta) \cap \mathbb{Z}^n$ , so we define the **h-ratio**:

$$\mathcal{H}(\Delta) = \frac{|\Delta^* - (\Delta \cup \bar{A}(\Delta))|}{|(\text{conv}(\Delta) \cap \mathbb{Z}^n) - (\Delta \cup \bar{A}(\Delta))|}$$



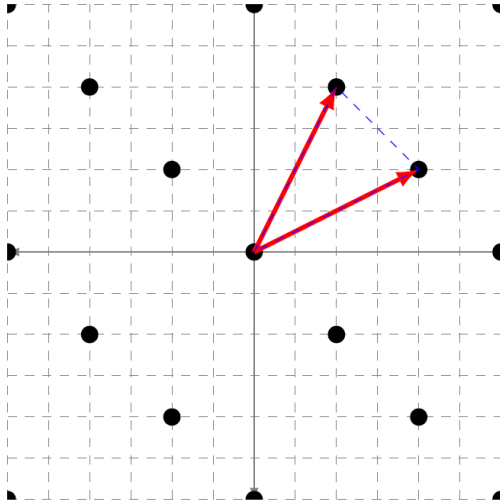
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**Observations:**

- Even though the algorithm looks easy, there are too many simplices to consider. In dimension 4, maximal degree 8 there are more than 300000 simplices to check. This makes it hard to write to database.
- Thus, we need to get rid of the redundant data. In fact, instead of simplicies one can consider the underlying lattice.

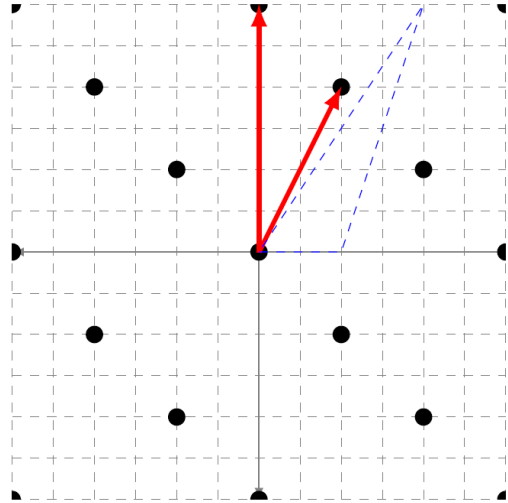




$$\Delta_1 = \left\{ \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 4 \end{bmatrix}, \begin{bmatrix} 4 \\ 2 \end{bmatrix} \right\}$$

$$M_{\Delta_1} = \begin{bmatrix} 2 & 4 \\ 4 & 2 \end{bmatrix}$$

$$L_{\Delta_1} = \langle (2, 4), (4, 2) \rangle$$



$$\Delta_2 = \left\{ \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 \\ 6 \end{bmatrix} \right\}$$

$$M_{\Delta_2} = \begin{bmatrix} 2 & 4 \\ 0 & 6 \end{bmatrix}$$

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$$T(\mathbf{x}) = \begin{bmatrix} 1 & 0 \\ 2 & -1 \end{bmatrix} \mathbf{x}$$

Thank you for your attention!