

# Sum of Squares: Part 1

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# Introduction

Reason two basic problems about polynomial inequalities

1. Feasibility of polynomial system.

$$\begin{aligned} p_1(\mathbf{x}) &\geq 0 \\ &\vdots \\ p_m(\mathbf{x}) &\geq 0 \end{aligned} \tag{1}$$

Is there an  $\mathbf{x}$  such that (1) is satisfied?

2. Checking non-negativity: Is  $q(\mathbf{x}) \geq 0, \forall \mathbf{x}$  satisfying (1) ?

## Justification: Feasibility

Feasibility checking is highly expressive.

- Example: MaxCut.
- . Input:  $G = (V, E)$ ,  $|V| = n$ .
- . Goal: Find  $S \subseteq V$  such that  $|E(S, \bar{S})|$  is maximized.
- . Polynomial Feasibility: For some  $\beta \in \mathbb{Z}^+$ ,

$$\frac{1}{4} \sum_{(i,j) \in E} (\mathbf{x}_i - \mathbf{x}_j)^2 = \beta$$

$$\mathbf{x}_i^2 = 1, \quad \forall 1 \leq i \leq n.$$

- .  $n + 1$  degree-2 polynomials. Enumerate over polynomial number of values of  $\beta$  and solve MaxCut.

Other examples: MaxClique, Max 3-SAT, Knapsack. Therefore, polynomial feasibility checking problem is *NP*-Hard.

## Goal

- ▶ Analyze a relaxation for the feasibility problem, and try to find interesting situations where one can get a poly-time algorithms.

## Justification: Non-Negativity

- ▶ Checking non-negativity: Given  $f : \{-1, 1\}^n \rightarrow \mathbb{R}$  with rational coefficients, decide if:
  - $f \geq 0, \forall \mathbf{x} \in \{-1, 1\}^n$ , or,
  - find an  $\mathbf{x} \in \{-1, 1\}^n$  such that  $f(\mathbf{x}) < 0$ .
- ▶ Example MaxCut: Decide if  $\text{MaxCut} \leq c$ .
  - Let  $f_G(\mathbf{x}) \stackrel{\text{def}}{=} \frac{1}{4} \sum_{(i,j) \in E} (\mathbf{x}_i - \mathbf{x}_j)^2$ .
  - Decide if  $c - f_G(\mathbf{x}) \geq 0, \forall \mathbf{x} \in \{-1, 1\}^n$ .

## Certifying Non-Negativity

Given  $f : \{-1, 1\}^n \rightarrow \mathbb{R}$ , find an “efficiently verifiable” certificate of non-negativity.

# SoS Certificates

Definition (SoS cert of non-neg (or) SoS proof of non-neg)

A degree- $d$  SoS certificate of non-negativity of  $f : \{-1, 1\}^n \rightarrow \mathbb{R}$  is a list of polynomials  $g_1, \dots, g_r : \{-1, 1\}^n \rightarrow \mathbb{R}$ , such such that

- $\deg(g_i) \leq d/2$ , and
- $f(\mathbf{x}) = \sum_{i \leq r} g_i^2(\mathbf{x}), \forall \mathbf{x} \in \{-1, 1\}^n$ .

## Efficiently Verifiable (?)

Polynomials  $f, g_1, \dots, g_r$  are represented as a vector of coefficients.

1. How large is  $r$ ? ( $\leq n^d$ , see later)
2. How large are coefficients of  $g_i$ ?



# Efficiently Verifiable

## Proposition (Efficiently Verifiable)

Suppose  $r \leq n^d$ , all coefficients of  $g_i$  are bounded in magnitude by  $2^{\text{poly}(n^d)}$ . Then the identity  $f = \sum_{i \leq r} g_i^2$  over all  $\mathbf{x} \in \{-1, 1\}^n$  can be checked in  $\text{poly}(n^d)$  time.

## Proof.

- Given  $g_i$ , can compute  $g_i^2$ , and  $\sum_{i \leq r} g_i^2$  in polynomial time.
- Check if  $(f - \sum_{i \leq r} g_i^2)(\mathbf{x}) = 0, \forall \mathbf{x} \in \{-1, 1\}^n$ .
- Using the fact that coefficient vector representation is unique, just check if  $f - \sum_{i \leq r} g_i^2 = \mathbf{0}$



## Fact (Unique Representation)

$\forall f : \{-1, 1\}^n \rightarrow \mathbb{R}$ , there exists a unique representation of  $f$ : The multi-linear representation of  $f$

$$f = \sum_{S \subseteq [n]} \hat{f}(S) \prod_{i \in S} x_i.$$

(this representation is its Fourier transform)

$\implies$  coefficient vector representation is unique.

(multilinear representation exists because  $x_i^2 = 1$ )

# Are non-negative functions always certifiable?

## Proposition (Certifiability of non-negative functions)

*Let  $f : \{-1, 1\}^n \rightarrow \mathbb{R}$  be non-negative over  $\{-1, 1\}^n$ . Then, there exists a  $\deg(2n)$ -SoS certificate of non-negativity.*

### Proof.

- Consider  $g : \{-1, 1\}^n \rightarrow \mathbb{R}$ , and  $g(\mathbf{x}) = \sqrt{f(\mathbf{x})}$ .
- Every function on  $\{-1, 1\}^n$  is a polynomial of  $\deg \leq n$ .
- $f = g^2 \implies \deg(2n)$ -SoS Certificate.



# Tensor Notation

- Suppose vector  $\mathbf{v} \in \mathbb{R}^n$ .
- $\mathbf{v}^{\otimes 2} \in \mathbb{R}^{n^2}$ , where  $\mathbf{v}(i, j) = \mathbf{v}_i \mathbf{v}_j$ .
- $\mathbf{v}^{\otimes k} \in \mathbb{R}^{n^k}$ .

# Proving Efficient Verifiability

## Theorem (PSD Matrices and SoS Certificates)

$f : \{-1, 1\}^n \rightarrow \mathbb{R}$  has a  $\deg(d)$ -SoS certificate of non-negativity iff there exists a matrix  $A$  such that  $A \succcurlyeq 0$ , and

$$f(\mathbf{x}) = \left\langle (1, \mathbf{x})^{\otimes \frac{d}{2}}, A \cdot (1, \mathbf{x})^{\otimes \frac{d}{2}} \right\rangle .$$

► Parsing Notation:

- $(1, \mathbf{x}) \in \mathbb{R}^{(n+1)}$ .
- $(1, \mathbf{x})^{\otimes \frac{d}{2}}$ : populate in a vector all possible monomials in the variable  $\mathbf{x}$  of degree at most  $d/2$ .
- $A \in \mathbb{R}^{(n+1)^{d/2} \times (n+1)^{d/2}}$ .

## Proof

- If Part:

$$\begin{aligned} f(\mathbf{x}) &= \left\langle (\mathbf{1}, \mathbf{x})^{\otimes \frac{d}{2}}, A \cdot (\mathbf{1}, \mathbf{x})^{\otimes \frac{d}{2}} \right\rangle \\ &= \left\langle (\mathbf{1}, \mathbf{x})^{\otimes \frac{d}{2}}, (B^T B) \cdot (\mathbf{1}, \mathbf{x})^{\otimes \frac{d}{2}} \right\rangle \\ &= \left\langle B \cdot (\mathbf{1}, \mathbf{x})^{\otimes \frac{d}{2}}, B \cdot (\mathbf{1}, \mathbf{x})^{\otimes \frac{d}{2}} \right\rangle \\ &= \left\| B \cdot (\mathbf{1}, \mathbf{x})^{\otimes \frac{d}{2}} \right\|^2. \end{aligned} \tag{2}$$

- Let  $g_i(\mathbf{x}) = \left\langle e_i, B \cdot (\mathbf{1}, \mathbf{x})^{\otimes \frac{d}{2}} \right\rangle$ , i.e.,  $i$ -th entry of the vector.
- $B$  is a matrix of constants, applied to monomials of degree at most  $d/2$ , therefore,  $\deg(g_i) \leq d/2$ .
- Therefore,

$$f(\mathbf{x}) = \sum_{i=1}^{(n+1)^{\otimes d/2}} g_i^2(\mathbf{x}).$$

## Proof Cont...

- Only if Part: Suppose  $f$  has a degree- $d$  SoS certificate.

$$f = \sum_{i \leq r} g_i^2, \text{ and } \underbrace{g_i(\mathbf{x})}_{\deg \leq d/2} = \left\langle \mathbf{v}_i, (1, \mathbf{x})^{\otimes \frac{d}{2}} \right\rangle.$$

$$\begin{aligned} f(\mathbf{x}) &= \sum_{i \leq r} \left\langle \mathbf{v}_i, (1, \mathbf{x})^{\otimes \frac{d}{2}} \right\rangle^2 \\ &= \left\langle (1, \mathbf{x})^{\otimes \frac{d}{2}}, \underbrace{\left( \sum_{i \leq r} \mathbf{v}_i \mathbf{v}_i^T \right)}_{\succeq 0} \cdot (1, \mathbf{x})^{\otimes \frac{d}{2}} \right\rangle. \end{aligned}$$



# Efficient Verifiability

## Corollary (Bound on $r$ )

*If  $f : \{-1, 1\}^n \rightarrow \mathbb{R}$  has a degree- $d$  SoS certificate, then it has a certificate with  $r \leq (n + 1)^{d/2}$ .*

Proof.

Follows from (2):

$$f(\mathbf{x}) = \left\| B \cdot (1, \mathbf{x})^{\otimes \frac{d}{2}} \right\|^2 .$$





# Efficient Verifiability

## Lemma (Bit-complexity of SoS proofs)

*Suppose  $f$  has a degree- $d$  SoS certificate over  $\{-1, 1\}^n$ . Then, we can find a degree- $d$  SoS certificate for  $f + \varepsilon'$  in time  $\text{poly}(n^d, \log 1/\varepsilon', \text{size}(f))$ .*

- ▶  $\{-1, 1\}^n$  is important, and doesn't necessarily hold for other domains.

## Proof

Since we are given  $f$ , we know that it can be efficiently represented. Therefore, we try to bound the entries of  $A$  in terms of  $f$ .

- $f(\mathbf{x}) = \left\langle (1, \mathbf{x})^{\otimes \frac{d}{2}}, A \cdot (1, \mathbf{x})^{\otimes \frac{d}{2}} \right\rangle$ , for some  $A$ .
- $f = \sum_u \hat{f}_u \mathbf{x}_u$ , where  $\mathbf{x}_u = \prod_{i \in u} \mathbf{x}_i$ .
- Expanding the inner product, we see  $\hat{f}_u = \sum_{S, T} A_{S, T}$ , such that  $\text{odd}(S + T) = u$ , and  $|S|, |T| \leq d/2$ .

$$\hat{f}_\emptyset = \sum_S A_{S, S} = \text{tr}(A) = \sum_i \underbrace{\lambda_i(A)}_{\geq 0}.$$

$$\|A\|_F^2 = \sum_{S, T} A_{S, T}^2 = \sum_i \lambda_i^2(A) \leq \hat{f}_\emptyset^2.$$

We do not know if entries of  $A$  are rational, therefore, above proof doesn't suffice. We now try to find an  $A$ .

## Connection between SDP and SoS: Find A

Proof Cont...

Recall

$$f(\mathbf{x}) = \left\langle \underbrace{(1, \mathbf{x})^{\otimes \frac{d}{2}}, A \cdot (1, \mathbf{x})^{\otimes \frac{d}{2}}}_{\stackrel{\text{def}}{=} g_A(\mathbf{x})} \right\rangle,$$

Then, we form the following constraints:

1.  $A \succcurlyeq 0$ .
2.  $\text{mult}(g_A(\mathbf{x})) = \text{mult}(f(\mathbf{x}))$ .

Therefore, we get the following SDP feasibility problem:

$$\forall u \subseteq [n] : \hat{f}_u = \sum_{\text{odd}(S+T)=u} A_{S,T}; \quad \left( (n+1)^d \text{constraints} \right)$$
$$A \succcurlyeq 0.$$

*It is unknown if we can decide the feasibility of this system.*

Therefore, we try to solve it approximately.

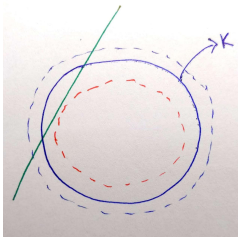
# Tools for Approximately Solving SDP

## Definition (Weak Separation Oracle)

Let  $K \subseteq \mathbb{R}^N$  be a convex set. Weak Separation Oracle:

- Input: Rational vector  $\mathbf{x} \in \mathbb{R}^N$ , and  $\varepsilon > 0$ .
- Output: Either
  - Correctly asserts that  $\mathbf{x} \in K + \mathcal{B}(0, \varepsilon)$ , or,
  - Returns an “almost separating hyperplane”, i.e., returns  $\mathbf{y} \neq \mathbf{0} \in \mathbb{R}^N$ , such that

$$\langle \mathbf{y}, \mathbf{x} \rangle > \langle \mathbf{y}, \mathbf{z} \rangle - \varepsilon \|\mathbf{y}\|_2, \forall \mathbf{z} \in K.$$



# Tools for Approximately Solving SDP

## Theorem (Grötschel, Lovász, Schrijver '81)

Let  $K$  be a closed, convex, and bounded set. Suppose there exists  $R > r > 0$ , such that  $\mathcal{B}(\mathbf{p}, r) \subseteq K \subseteq \mathcal{B}(\mathbf{0}, R)$ . Assume that we have a poly-time weak-separation oracle for  $K$ . Then given any rational vector  $\mathbf{v} \in \mathbb{R}^N$ , we can compute a rational vector  $\mathbf{x} \in \mathbb{R}^N$  such that

1.  $\mathbf{x} \in K$ .
2.  $\langle \mathbf{v}, \mathbf{x} \rangle \geq \langle \mathbf{v}, \mathbf{z} \rangle - \varepsilon, \forall \mathbf{z} \in K$ .

Running time:  $\text{poly}(\log R/r + \log 1/\varepsilon + N)$ .

- ▶ Interpreting theorem: If I have a convex set  $K$  with non-empty interior, with a weak separation oracle, then I can approximately maximize  $\mathbf{v}^\top \mathbf{x}$  over  $K$ .

## Proof Cont...

Applying this to our problem. We define the following:

$$S = \left\{ A \mid A \succcurlyeq 0, \langle C_i, A \rangle = b_i, \forall i, \|A\|_F^2 \leq \hat{f}_\emptyset^2 \right\} .$$

We note that  $S$  is convex, bounded, closed. Now,

$$\mathcal{B}(\mathbf{p}, r) \not\subseteq S .$$

Therefore, relax the equality constraints. And find a point in  $S'$ ,

$$S' = \left\{ A \mid A \succcurlyeq 0, \langle C_i, A \rangle = [b_i - \varepsilon, b_i + \varepsilon], \forall i, \|A\|_F^2 \leq \hat{f}_\emptyset^2 \right\} .$$

Now,  $\mathcal{B}(\mathbf{p}, r) \subseteq S'$  because for any point  $A \in S$ , then  $A + \delta I \in S'$  for  $\delta$  small enough.

# Applying the Theorem

## Proof Cont...

- We can find some  $f'$  such that  $f'$  has a degree- $d$  SoS certificate and  $|\hat{f}_u - \hat{f}'_u| \leq \varepsilon$ .
- Note:  $f(\mathbf{x}) = f'(\mathbf{x}) + (f - f')(\mathbf{x})$ .
- Small coefficient:  $\sum_{|u| \leq d} |\hat{f}_u - \hat{f}'_u| \leq \varepsilon(n+1)^d$ .
- Let  $L = \sum_{|u| \leq d} |\hat{f}_u - \hat{f}'_u|$ .
- Then,  $L + f - f'$  has a degree  $d$ -SoS certificate.
- Then, it implies,  $L + f$  has a degree- $d$  SoS certificate, i.e.,  $\varepsilon(n+1)^d + f$  has a degree- $d$  SoS certificate.
- $\varepsilon = \varepsilon' \mathcal{O}(n^{-d})$  finishes the proof.



$L + f - f'$  has a degree  $d$ -SoS certificate

Proof.

Claim

Let  $f = \sum_{|S| \leq d} \hat{f}_S \mathbf{x}_S$ . Then  $(1 - \mathbf{x}_S)$  and  $(1 + \mathbf{x}_S)$  has a degree- $d$  SoS certificate.

- $f = \sum_S |\hat{f}_S| \left( \text{sign}(\hat{f}_S) \mathbf{x}_S \right)$ .
- By **claim**:  $\sum_S |\hat{f}_S| \left( 1 + \text{sign}(\hat{f}_S) \mathbf{x}_S \right)$  has a degree- $d$  SoS certificate.

$$\begin{aligned} \sum_S |\hat{f}_S| \left( 1 + \text{sign}(\hat{f}_S) \mathbf{x}_S \right) &= \sum_S |\hat{f}_S| + \sum_S |\hat{f}_S| \text{sign}(\hat{f}_S) \mathbf{x}_S \\ &= \sum_S |\hat{f}_S| + f. \end{aligned}$$

- Note: to see connection with  $L$ , function  $f$  used in the claim here is  $f - f'$  in the previous proof.





## Proof of Claim

Proof.

- Let  $|S| \leq d$ .
- $S = T_1 \cup T_2$ ,  $|T_1| \leq |T_2| \leq d/2$ .
- Then  $\mathbf{x}_S = \mathbf{x}_{T_1} \cdot \mathbf{x}_{T_2}$ .

$$\begin{aligned}(\mathbf{x}_{T_1} - \mathbf{x}_{T_2})^2 &= \mathbf{x}_{T_1}^2 + \mathbf{x}_{T_2}^2 - 2\mathbf{x}_{T_1}\mathbf{x}_{T_2} \\ &= 2 - 2\mathbf{x}_{T_1}\mathbf{x}_{T_2} \\ \therefore (1 - \mathbf{x}_S) &= \frac{1}{2}(\mathbf{x}_{T_1} - \mathbf{x}_{T_2})^2.\end{aligned}$$

- Similarly for  $(1 + \mathbf{x}_S)$ .



Halfway Through

# Sum of Squares: Part 1 (continued)

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# What if Degree- $d$ SoS Certificate Doesn't Exist for $f$ ?

In that case

1.  $\exists \mathbf{x}$ , such that  $f(\mathbf{x}) < 0$ , or,
  2. If  $d \leq 2n$ , then  $f$  may be non-negative and yet a degree- $d$  SoS certificate doesn't exist.
- ▶ **Ideally**, if  $f$  does not have a degree- $d$  SoS certificate, we would like the “algorithm” to output an  $\mathbf{x}$  such that  $f(\mathbf{x}) < 0$ .
  - ▶ However, that may not always be possible.
  - ▶ To achieve that **ideal** aim, we construct an object called *Pseudo-distribution*.

# Towards Constructing Pseudo-distribution

## Fact

The set  $SoS_d \subseteq \mathbb{R}^{2^n}$ , where

$$SoS_d \stackrel{\text{def}}{=} \{f \mid f \text{ has a degree-}d \text{ SoS certificate}\},$$

is a closed, convex cone.

## Theorem (Hyperplane Separation Theorem)

Suppose  $K \subseteq \mathbb{R}^N$  is a convex set. Let  $\mathbf{v} \notin K$ . Then there exists a hyperplane  $\mathcal{H} = \{\mathbf{x} \mid \langle \mathbf{u}, \mathbf{x} \rangle \geq 0\}$ , such that  $K \subseteq \mathcal{H}$ , and  $\mathbf{v} \notin \mathcal{H}$ .

## Towards Constructing Pseudo-distribution

Suppose  $p \notin \text{SoS}_d$ . Then, there exists  $\mu$  such that  $p$  is on one side of the hyperplane (defined by  $\mu$ ) and  $\text{SoS}_d$  on the other side, i.e.,

$$\sum_{\mathbf{x} \in \{-1,1\}^n} \mu(\mathbf{x}) \cdot p(\mathbf{x}) < 0,$$

$$\sum_{\mathbf{x} \in \{-1,1\}^n} \mu(\mathbf{x}) \cdot f(\mathbf{x}) \geq 0, \quad \forall f \in \text{SoS}_d$$

$$\sum_{\mathbf{x} \in \{-1,1\}^n} \mu(\mathbf{x}) = 1, \quad (\text{by scaling}).$$

- Hypothetical: Suppose  $\mu \geq 0$ , then it describes a probability distribution over  $\{-1, 1\}^n$ .
- . Therefore, there exists a distribution such that for  $p \notin \text{SoS}_d$

$$\mathbb{E}_{\mu} p < 0,$$

$$\mathbb{E}_{\mu} f \geq 0,$$

$$\forall f \in \text{SoS}_d.$$

# Pseudo-distribution

- ▶ Notation: Pseudo-expectation (when  $\mu$  is not non-negative)

$$\tilde{\mathbb{E}}_{\mu} f = \sum_{\mathbf{x} \in \{-1, 1\}^n} \mu(\mathbf{x}) \cdot f(\mathbf{x}).$$

## Definition (Pseudo-distribution)

A degree- $d$  pseudo-distribution over  $\{-1, 1\}^n$  is a function  $\mu : \{-1, 1\}^n \rightarrow \mathbb{R}$  such that  $\tilde{\mathbb{E}}_{\mu}$  satisfies:

1.  $\tilde{\mathbb{E}}_{\mu} 1 = 1$ .
2.  $\forall f : \deg(f) \leq \frac{d}{2}, \tilde{\mathbb{E}}_{\mu} f^2 \geq 0$ .

## Fact

*Every degree  $\geq 2n$  pseudo-distribution  $\mu$  is an actual probability distribution.*

## Proof Sketch.

- Define indicator polynomial  $f_{\mathbf{y}} : \{-1, 1\}^n \rightarrow \mathbb{R}$ , such that  $f_{\mathbf{y}}(\mathbf{y}) = 1$ , and  $f_{\mathbf{y}}(\mathbf{x}) = 0$ ,  $\forall \mathbf{x} \neq \mathbf{y}$ . Moreover,  $\deg(f) \leq n$ .
- By definition  $\tilde{\mathbb{E}}_{\mu} f_{\mathbf{y}}^2 \geq 0$ .
- Construct the distribution  $\text{prob}(\mathbf{y}) \stackrel{\text{def}}{=} \tilde{\mathbb{E}}_{\mu} f_{\mathbf{y}}^2$ . And this distribution is the pseudo-distribution  $\mu$ .





## Specifying Pseudo-distributions

Pseudo-distributions can be specified as a vector  $\mu' \in \mathbb{R}^{n^d}$ .

### Claim

*For all degree- $d$  pseudo-distributions, there exists a degree- $d$  multi-linear polynomial  $\mu' : \{-1, 1\}^n \rightarrow \mathbb{R}$ , such that  $\tilde{\mathbb{E}}_{\mu} p = \tilde{\mathbb{E}}_{\mu'} p$  for all  $p$  such that  $\deg(p) \leq d$ .*

### Proof Sketch.

Write  $\mu$  and  $p$  in multilinear form.

$$\mu(\mathbf{x}) = \sum_{S \subseteq [n]} \hat{\mu}_S \mathbf{x}_S$$

$$p(\mathbf{x}) = \sum_{S \subseteq [n], |S| \leq d} \hat{p}_S \mathbf{x}_S.$$

$\tilde{\mathbb{E}}_{\mu} p = \langle \mu, p \rangle = \langle \mu', p \rangle$ , where  $\mu'$  is a degree  $\leq d$  part of  $\mu$ . □

► Notation- Pseudo-moments:  $\tilde{\mathbb{E}}\mu(1, \mathbf{x})^{\otimes d}$ .

(Expectation of a vector) expectations of degree  $\leq d$  monomials.

# Pseudo-distribution and PSD Matrices

## Proposition

$\mu$  is a degree- $d$  pseudo-distribution iff

1.  $\tilde{\mathbb{E}}_{\mu} \mathbf{1} = 1,$

2.  $\underbrace{\tilde{\mathbb{E}}_{\mu}(\mathbf{1}, \mathbf{x})^{\otimes \frac{d}{2}} \left( (\mathbf{1}, \mathbf{x})^{\otimes \frac{d}{2}} \right)^{\top}}_{\text{degree-}d \text{ pseudo-moment matrix}} \succeq 0.$

- Parsing notation: The  $S, T$ -th entry of the pseudo-moment matrix is  $\tilde{\mathbb{E}}_{\mu} \mathbf{x}_S \mathbf{x}_T = \tilde{\mathbb{E}}_{\mu} \mathbf{x}_{\text{odd}(S+T)}.$

## Proof.

- Let  $f$  be a degree- $d/2$  polynomial.
- $\tilde{\mathbb{E}}_{\mu} f^2 = c (\hat{f})^{\top} M_{d/2} \hat{f}$ , because

$$\begin{aligned}\tilde{\mathbb{E}}_{\mu} f^2 &= \tilde{\mathbb{E}}_{\mu} \left( \sum_S \hat{f}_S \mathbf{x}_S \right)^2 = \tilde{\mathbb{E}}_{\mu} \sum_{S,T} \hat{f}_S \hat{f}_T \mathbf{x}_S \mathbf{x}_T \\ &= \sum_{S,T} \hat{f}_S \hat{f}_T \tilde{\mathbb{E}}_{\mu} \mathbf{x}_S \mathbf{x}_T.\end{aligned}$$

- But since  $f^2$  is a degree- $d$  SoS, the quadratic form is positive, and therefore,  $M_{d/2} \succcurlyeq 0$ .



# Formal Pseudo-expectation Proof

## Theorem

For every  $f$ , every even  $d \in \mathbb{Z}_{\geq 0}$ , there exists a degree- $d$  SoS certificate of  $f$  iff

$$\forall \text{ degree-}d \text{ pseudo-distribution over } \{-1, 1\}^n, \tilde{\mathbb{E}}_{\mu} f \geq 0.$$

## Proof

- If Part: If  $f$  has a degree- $d$  SoS certificate,  
$$\tilde{\mathbb{E}}_{\mu} f = \sum_{i \leq r} \tilde{\mathbb{E}}_{\mu} g_i^2 \geq 0.$$
- Only If Part: Suppose  $f \notin \text{SoS}_d$ , then there exists a hyperplane with  $\mu$  as the normal vector such that

$$\begin{aligned} \tilde{\mathbb{E}}_{\mu} f &< 0, \\ \tilde{\mathbb{E}}_{\mu} g^2 &\geq 0, \quad \forall g \text{ such that } \deg(g) \leq d/2. \end{aligned}$$

Need:  $\tilde{\mathbb{E}}_{\mu} 1 > 0$ .

$$\tilde{\mathbb{E}}_{\mu} \mathbf{1} > 0$$

Proof Cont...

- We know,  $\exists L > 0$  such that  $L + f$  has a degree- $d$  SoS certificate. We have  $\tilde{\mathbb{E}}_{\mu} f < 0$ , and

$$\tilde{\mathbb{E}}_{\mu}(L + f) \geq 0$$

$$\tilde{\mathbb{E}}_{\mu} L \geq -\tilde{\mathbb{E}}_{\mu} f$$

$$L \tilde{\mathbb{E}}_{\mu} \mathbf{1} \geq -\tilde{\mathbb{E}}_{\mu} f$$

$$\tilde{\mathbb{E}}_{\mu} \mathbf{1} \geq \frac{-\tilde{\mathbb{E}}_{\mu} f}{L} > 0.$$



# Pseudo-distribution and SDP Connection

## Theorem

Suppose  $f$  does not have a degree- $d$  SoS certificate. Then, there exists a poly  $(n^d, \log 1/\varepsilon, \text{size}(f))$ -time algorithm to compute a pseudo-distribution  $\mu$  such that  $\tilde{\mathbb{E}}_\mu f < \varepsilon$ .

## Proof Sketch.

Form a SDP system- Make variables for all possible monomials:  
 $\tilde{\mathbb{E}}_\mu \mathbf{x}_S, |S| \leq d$ .

- Constraint 1:  $\tilde{\mathbb{E}}_\mu 1 = 1$ .
- Constraint 2:  $\tilde{\mathbb{E}}_\mu (1, \mathbf{x})^{\otimes \frac{d}{2}} \left( (1, \mathbf{x})^{\otimes \frac{d}{2}} \right)^T \succeq 0$ .
- Constraint 3:  $\tilde{\mathbb{E}}_\mu f < 0$ .

Use SDP solving up to slack  $\varepsilon$  to obtain the theorem statement.



# SoS Algorithms

Algorithms for computing  
SoS certificates, and Pseudo-distributions  
are called *SoS algorithms*.