

Sums of Squares

Lecture ①

Let $\mathbb{R}[x] := \mathbb{R}[x_1, \dots, x_n]$ denote the polynomial ring in the indeterminates x_1, \dots, x_n with real coefficients.

Let $\Sigma^2 \subset \mathbb{R}[x]$ the convex cone of **Sums of Squares** polynomials, that is:

$$\Sigma^2 := \left\{ \sigma \in \mathbb{R}[x] \mid \exists r \in \mathbb{N} \exists h_1, \dots, h_r \in \mathbb{R}[x] \text{ s.t. } \sigma = h_1^2 + \dots + h_r^2 \right\}$$

Example ($n=1$)

$$\text{Let } f(x) = x^4 - 2x^3 + 2x^2 - 2x + 1 = (x^2 - x)^2 + (x - 1)^2$$

• $f(x) \geq 0 \quad \forall x \in \mathbb{R}$, since $f \in \Sigma^2$

$$\bullet \quad f(x) = (1 \ x \ x^2) \begin{pmatrix} 0 \\ -1 \\ 1 \end{pmatrix} (0, -1, 1) \begin{pmatrix} 1 \\ x \\ x^2 \end{pmatrix} +$$

$$+ (1 \ x \ x^2) \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} (-1 \ 1 \ 0) \begin{pmatrix} 1 \\ x \\ x^2 \end{pmatrix} =$$

$$= (1 \ x \ x^2) \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & -1 \\ 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ x \\ x^2 \end{pmatrix} + (1 \ x \ x^2) \begin{pmatrix} 1 & -1 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ x \\ x^2 \end{pmatrix} =$$

$$= (1 \ x \ x^2) \begin{pmatrix} 1 & -1 & 0 \\ -1 & 1 & -1 \\ 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ x \\ x^2 \end{pmatrix} \quad \text{①}$$

Lemmas

0) If $\sigma \in \Sigma^2$, $\sigma(x) \geq 0 \quad \forall x \in \mathbb{R}^n$. ← This is why they are interesting!

1) If $\sigma = h_1^2 + \dots + h_r^2 \in \Sigma^2$ with

$\max_i \deg h_i = d$, then $\deg \sigma = 2d$

2) Σ^2 is a pointed convex cone: $\Sigma^2 \cap -\Sigma^2 = \{0\}$

(in other words, Σ^2 contains no lines)

3) $\Sigma^2 - \Sigma^2 := \{\sigma_1 - \sigma_2 \in \mathbb{R}[x] \mid \sigma_1, \sigma_2 \in \Sigma^2\} = \mathbb{R}[x]$

Proof Exercise.

Proposition Let $f \in \mathbb{R}[x]$, $\deg f = 2d$, $N := \dim \mathbb{R}[x]_{\leq d}$

then $f \in \Sigma^2$ if and only if \exists a positive semidefinite $G \in \mathbb{R}^{N \times N}$ (Gram matrix) such that, if $[x]_d \in \mathbb{R}^N$

denotes the vector of the monomial basis of $\mathbb{R}[x]_{\leq d}$,

$$f(x) = [x]_d^t G [x]_d$$

Proof (\Rightarrow) As in the example.

$$G = U^t U = \begin{pmatrix} 1 & & \\ \mu_1 & \ddots & \\ & \ddots & 1 \end{pmatrix} \begin{pmatrix} -\mu_1 - \\ \vdots \\ -\mu_N - \end{pmatrix}$$

$G \succeq 0$, Cholesky

$$\Rightarrow f(x) = [x]_d^t \begin{pmatrix} 1 & & \\ \mu_1 & \ddots & \\ & \ddots & 1 \end{pmatrix} \begin{pmatrix} -\mu_1 - \\ \vdots \\ -\mu_N - \end{pmatrix} [x]_d$$
$$= ([x]_d^t \begin{pmatrix} 1 \\ \mu_1 \\ \vdots \end{pmatrix})^2 + \dots + ([x]_d^t \begin{pmatrix} 1 \\ \mu_N \\ \vdots \end{pmatrix})^2 \quad \square$$

Proposition the convex cone Σ^2 is closed in $\mathbb{R}[x]$,
 in the finest locally convex topology, * See next page for def.
 i.e. $\Sigma_{2d,n}^2$ is closed in $\mathbb{R}[x]_{\leq 2d} \quad \forall d$.

Proof Let $N = \dim \mathbb{R}[x]_{\leq d}$. $\exists f \notin \Sigma_{n,2d}^2$, then
 $f = p_1^2 + \dots + p_N^2$ (we can bound the number
 of addends by $N = N(d,n)$, see previous proposition)
 for some $p_i \in \mathbb{R}[x]_{\leq d}$.

Choose good norm $\|\cdot\|$ on $\mathbb{R}[x]$. Say for $p \in \mathbb{R}[x]$
 $\|p\| := \int p d\mu$, where μ has normal (Gaussian)
 distribution. Up to scaling, assume wlog that $\|f\| = 1$.

We have:

$$\|p_i\|^2 \leq \sum_{i=1}^N \|p_i\|^2 = \int \sum_{i=1}^N p_i^2 d\mu = \int f d\mu = \|f\| = 1.$$

thus each p_i lives in the (compact) unit
 ball of $(\mathbb{R}[x]_{\leq d}, \|\cdot\|)$.

If we pick a converging sequence $\{f_k\}_{k \in \mathbb{N}} \subset \Sigma_{2d,n}^2$,
 then writing $\forall k$ a SOS representation for f_k ,
 each SOS addend lives in a compact set,
 and thus we can extract a subsequence s.t.
 every SOS addend converges. Thus

$\lim_{k \rightarrow \infty} f_k \in \Sigma_{2d,n}^2$, and $\Sigma_{2d,n}^2 \subset \mathbb{R}[x]_{\leq 2d}$
 is closed. $\textcircled{3}$

Question: If $f(x) \geq 0 \forall x \in \mathbb{R}^n$, then $f \in \Sigma^2$?

Theorem (Hilbert)

Denote: $\Sigma_{2d,n}^2$ the SOS n -variate polynomials of $\deg \leq 2d$

• $P_{2d,n}$ the n -variate polynomials of $\deg \leq 2d$ that are non-negative on \mathbb{R}^n .

then $\Sigma_{2d,n}^2 = P_{2d,n} \Leftrightarrow (n=1 \text{ or } d=1 \text{ or } (n,d) = (2,2))$

Examples

• Every univariate non-negative polynomial can be written as a sum of 2 squares. (Ex.)

• Let $M(x,y) = 1 - 3x^2y^2 + x^2y^4 + x^4y^2$

• $M(x,y) \notin \Sigma^2$ (Ex.)

• $(x^2+y^2)^2 M(x,y) = x^2y^2(1+x^2+y^2)(-2+x^2+y^2)^2 + (x^2-y^2)^2$
 $\Rightarrow M(x,y) \geq 0$

(Alternative proof: AM-GM inequality, Ex.)

• SOS optimization. Consider $f \in \mathbb{R}[x]$ and

$f^* := \inf \{ f(x) : x \in \mathbb{R}^n \} = \sup \{ \lambda \in \mathbb{R} : f(x) - \lambda \geq 0 \forall x \in \mathbb{R}^n \}$

then $f_{\text{SOS}}^* = \sup \{ \lambda \in \mathbb{R} : f - \lambda \in \Sigma^2 \} \leq f^*$.

We write SOS optimization as an SDP more explicitly. Write $f = [x]_d^t G [x]_d$, $G \succeq 0$. Consider f as a 1×1 matrix with entries in $\mathbb{R}[x]$. Then:

$$f = \text{tr}(f) = \text{tr}([x]_d^t G [x]_d) = \text{tr}(G [x]_d [x]_d^t)$$

$$\text{Write } A = [x]_d [x]_d^t = \sum_{|\alpha| \leq 2d} A_\alpha x^\alpha$$

for some $A_\alpha \in \mathbb{R}^{N \times N}$

$$\left(\begin{aligned} \text{e.g.: } \begin{pmatrix} 1 \\ x \end{pmatrix} \begin{pmatrix} 1, x \end{pmatrix} &= \begin{pmatrix} 1 & x \\ x & x^2 \end{pmatrix} \\ &= \underbrace{\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}}_{A_0} \cdot 1 + \underbrace{\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}}_{A_1} x + \underbrace{\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}}_{A_2} x^2 \end{aligned} \right)$$

Therefore:

$$f = \sum_{|\alpha| \leq 2d} \text{tr}(G A_\alpha) x^\alpha \quad (f_0 = \text{tr}(G A_0))$$

and:

$$\begin{aligned} f_{\text{SOS}}^* - f_0 &= \sup_{\lambda} \lambda - f_0 \quad \text{s.t.} \quad f - \lambda \in \Sigma_{n, 2d}^2 = \\ &= \sup \text{tr}(G A_0) \quad \text{s.t.} \\ &G \succeq 0, \text{tr}(G A_\alpha) = f_\alpha \quad \forall 0 \neq |\alpha| \leq 2d. \end{aligned}$$

$$= \sup \langle G, A_0 \rangle \quad \text{s.t.}$$

$$G \succeq 0, \langle G, A_\alpha \rangle = f_\alpha \quad \forall 0 \neq |\alpha| \leq 2d.$$

(A_α, f_α for $|\alpha| \leq 2d$: data of the SDP. $G \succeq 0$: variable) (5)

Representation theorems

Since $\Sigma_{2d,n}^2 \neq P_{2d,n}$, we cannot certify always non-negativity using an SOS representation.

Question (Hilbert's 17th problem):

Can we always represent non-negative polynomials using ratios of SOS? Answer: yes!

Theorem (Artin) $\forall f \in \mathbb{R}[x]$ is ≥ 0 on \mathbb{R}^n , $\exists p_1, \dots, p_m, q_1, \dots, q_m \in \mathbb{R}[x]$ s.t.

$$f = \left(\frac{p_1}{q_1}\right)^2 + \dots + \left(\frac{p_m}{q_m}\right)^2$$

Remarks

- Unknown denominator(s) \rightarrow difficult to write as an SDP.
- Unknown degrees: the degrees of p_i, q_i can be higher than $\deg f$.

Theorem (Reznick) $\forall f \in \mathbb{R}[x]$ is homogeneous

and $f(x) > 0$ for $x \neq 0$, $d = \deg f$, and denoting

$$\varepsilon(f) = \frac{\inf\{f(u) : u \in S^{n-1}\}}{\sup\{f(u) : u \in S^{n-1}\}} > 0,$$

then $(x_1^2 + \dots + x_n^2)^r f \in \Sigma^2$ if $r \geq \frac{nd(d-1)}{(4 \log 2) \varepsilon(f)} - \frac{n+d}{2}$

Other kind of effective representation theorems:

Theorem (Polyá, Powers, Reznick) If:

- $f \in \mathbb{R}[X]$ is homogeneous, $d = \deg f$;
- $f(x) > 0$ for $x \in \Delta_n = \left\{ x \in \mathbb{R}^n : x_i \geq 0, \sum x_i = 1 \right\}$;
- $f = \sum_{|\alpha|=d} c(\alpha) \frac{d!}{\alpha_1! \dots \alpha_n!} X^\alpha$, $\|f\| := \max_{\alpha} |c(\alpha)|$;
- $\varepsilon(f) := \frac{\inf \{ f(u) : u \in \Delta_n \}}{\|f\|}$;

Then $(x_1 + \dots + x_n)^r f$ has positive coefficients

if $r > \frac{d(d-1)}{2 \varepsilon(f)} - d$

For representations of polynomials over \mathbb{C} , the parameter $\varepsilon(f)$ does not appear in the complexity estimates! \rightarrow More about $\varepsilon(f)$ in the upcoming lectures.

