



Polynomial Non-negativity

let $f(x)$ be a polynomial function of x .

problem: $\exists x$ st $f(x) < 0$?

- 1° No: f is called PSD. need proof or certificate
- 2° Yes: give an x st $f(x) < 0$

Sum-of-Square

let f be a polynomial,

f is non-negative $\Leftrightarrow \exists$ polynomial g_1, \dots, g_m st. $f = g_1^2 + \dots + g_m^2 \rightarrow$ Sum-of-Square decomposition

In general non-negativity does not imply SOS

We can write any polynomial as a quadratic function of monomials

Eg. $f(x,y) = 4x^4 + 4xy^2 - 7x^2y^2 - 2xy^4 + 10y^4$

$$= \begin{bmatrix} x^2 \\ xy \\ y^2 \end{bmatrix} \begin{bmatrix} 4 & 2 & -1 \\ 2 & -7+1 & 1 \\ -1 & 1 & 10 \end{bmatrix} \begin{bmatrix} x^2 \\ xy \\ y^2 \end{bmatrix}$$

$$= b(x,y)^T Q(x) b(x,y)$$

With $\lambda = 6$

$$Q(x) = \begin{bmatrix} 4 & 2 & -6 \\ 2 & 5 & 1 \\ -6 & 1 & 10 \end{bmatrix} = \begin{bmatrix} 0 & 2 \\ 2 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 0 & 2 & 1 \\ 2 & 1 & -3 \end{bmatrix} \text{ is PSD}$$

$$f(x,y) = \left\| \begin{bmatrix} 0 & 2 & 1 \\ 2 & 1 & -3 \end{bmatrix} \begin{bmatrix} x^2 \\ xy \\ y^2 \end{bmatrix} \right\|^2$$

$$= \underbrace{(2xy + y^2)^2}_{g_1^2} + \underbrace{(2x^2 + xy - 3y^2)^2}_{g_2^2}$$

Suppose $f = \text{polynomial}(x_1, \dots, x_n)$ of degree $2d$

let $b(x_1, \dots, x_n)$ be a vector of all monomials with degree $\leq d$

f is SOS $\Leftrightarrow \exists Q \succeq 0$ st $f(x_1, \dots, x_n) = b(x_1, \dots, x_n)^T Q b(x_1, \dots, x_n)$

affine constraint on Q by matching coeffs

if $Q \succeq 0$ and $f(x_1, \dots, x_n) = b(x_1, \dots, x_n)^T Q b(x_1, x_n)$

can reconstruct the sum-of-squares by

$$b(x,y)^T Q b(x,y) = b(x,y)^T U^T U b(x,y) = \|U b(x,y)\|^2$$

$Q = U^T U$ is the Cholesky factorization

Polynomial Function Global Optimization

let $F(x,y)$ be a polynomial

$$\min_{x,y} F(x,y)$$

$$\downarrow$$

$$\max_f$$

$$\text{s.t. } F(x,y) \geq f \quad \forall x,y$$

$$\downarrow$$

$$\max_f$$

$$\text{s.t. } F(x,y) - f \in \text{sos}$$

$$\downarrow$$

$$\left. \begin{array}{l} \max_f \\ \text{s.t. } \begin{bmatrix} 1 & b(x,y) \end{bmatrix} \underbrace{\begin{bmatrix} -f & \dots \\ \vdots & \vdots \\ 1 & b(x,y) \end{bmatrix}}_Q \begin{bmatrix} 1 \\ b(x,y) \end{bmatrix} = f(x,y) \\ \underbrace{\begin{bmatrix} -f & \dots \\ \vdots & \vdots \\ 1 & b(x,y) \end{bmatrix}}_Q \succeq 0 \end{array} \right\} \text{SDP}$$

Remark:

$$Q \succeq 0 \Rightarrow f \leq 0$$

$$\text{and } \min_x f(x) \leq 0 \text{ since } f(x) = 0$$

(if f contains constant term, just drop it)

S-Procedure

consider $\begin{cases} \text{a scalar polynomial } p(x) \\ \text{a vector of polynomial } g(x) = \begin{bmatrix} g_1(x) \\ \vdots \\ g_m(x) \end{bmatrix} \end{cases}$

1° if \exists polynomial multiplier $\lambda(x) = \begin{bmatrix} \lambda_1(x) \\ \vdots \\ \lambda_m(x) \end{bmatrix}$ s.t.

$$\begin{cases} p(x) + \lambda(x)^T g(x) \in \text{sos} \\ \lambda_i(x) \in \text{sos} \quad \forall i=1, \dots, m \end{cases}$$

then $g(x) \leq 0 \Rightarrow p(x) \geq 0$

proof:

consider the optimization problem

$$\min_x p(x)$$

$$\text{s.t. } g(x) \leq 0 \quad \text{dual polynomial } \lambda(x)$$

$$L(x, \lambda) = p(x) + \lambda(x)^T g(x) \in \text{sos}$$

$$\text{thus } L(x, \lambda) \geq 0 \quad \forall x$$

$$\text{thus } 0 \leq \inf_x L(x, \lambda)$$

$$\leq \inf_{g(x) \leq 0} p(x) + \underbrace{\sum_i \lambda_i(x) \cdot g_i(x)}_{\leq 0}$$

$$\leq \inf_{g(x) \leq 0} p(x)$$

same proof as for weak duality

2° if \exists polynomial multiplier $\lambda(x) = \begin{bmatrix} \lambda_{11}(x) \\ \vdots \\ \lambda_{nn}(x) \end{bmatrix}$ st

$p(x) + \lambda(x)^T g(x) \in \text{SOS}$
then $g(x) = 0 \Rightarrow p(x) \geq 0$