

Low-rank Sum of Squares has  
no spurious local minima

$$\text{Sum of Squares : } p(x) = \sum_{i=1}^r u_i(x)^2$$

- certifies that  $p(x) \geq 0$
- enables pdy. optimization  $\max_{\text{s.t.}} X$   
 $p(x) - r \geq 0$
- Applications in control, comb. opt., signal processing

How to solve?

Semidefinite programming

$$\langle A_i, X \rangle = b_i, \quad X \succeq 0$$

$$\begin{matrix} \downarrow \text{sos} \\ X \in S^n \end{matrix}$$

$$b(x) = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \end{bmatrix} \in \mathbb{R}^n \quad p(x) = b(x)^T Q b(x)$$

$$Q \succeq 0$$

B-M Method

$$f(u) = \sum_i (\langle A_i, uu^T \rangle - b_i)^2 \quad u \in \mathbb{R}^{h \times r}$$

$$\begin{matrix} \downarrow \text{sos} \\ f_p(\vec{u}) = \left\| \sum_{i=1}^r u_i(x)^2 - p(x) \right\|^2 \end{matrix} \xrightarrow{\text{inner product on polynomials}}$$

Linear constraints on PSD cone.

Solved with nonlinear optimization methods

Typically solved with interior-point methods

TF(u) fast to compute when r small.

Affine  $O(h^3)$  iteration complexity ( $h \approx 1000$ )

Nonconvex problem so might be stuck in local minima

Smaller r, more non-convex

When does second-order critical point  $\Rightarrow$  global optimum?

$r \geq h$  : always

$r \gtrsim \sqrt{hn}$  : almost always  
 $\downarrow$   
no. of constraints

$\exists$  instance that has spurious SDP when  $r=h-1$ .

low rank: matrix completion/sensing, planted statistical problems...

Fact: If univariate  $p(x) \geq 0$ , then  $p(x) = u_1(x)^2 + u_2(x)^2$ .

Thm: If univariate  $p(x) \geq 0$ ,  $f_p(\vec{u}) = \|u_1^2 + u_2^2 - p\|^2$  has no spurious SOSCPs. ( $r=2!$ )

+  $\nabla f_p(\vec{u})$  can be computed in  $O(n \log n)$ -time if suitable  $\|\cdot\|^2$  chosen.

+ first-order method for univariate SOS of million-degree polynomials ( $\sim 30$  mins)

Proof:

$$A_{\vec{u}}(\vec{v}) = \sum_{i=1}^r u_i(x) v_i(x), \quad \sigma(\vec{u}) = A_{\vec{u}}(\vec{u}) \text{ so } f_p(\vec{u}) = \|\sigma(\vec{u}) - p\|^2$$

$$\begin{aligned} \nabla f_p(\vec{u})(\vec{v}) &= \langle A_{\vec{u}}(\vec{v}), \sigma(\vec{u}) - p \rangle \\ &= 0 \quad \forall \vec{v} \end{aligned} \quad \left. \begin{array}{l} \\ \end{array} \right\} \Rightarrow \vec{u} \text{ is SOCp}$$

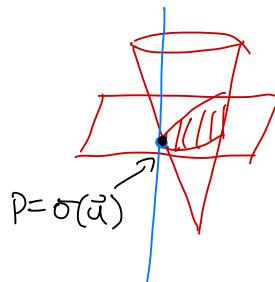
$$\nabla^2 f_p(\vec{u})(\vec{v}, \vec{w}) = \langle \sigma(\vec{v}), \sigma(\vec{u}) - p \rangle + 2 \|A_{\vec{u}}(\vec{v})\|^2 \geq 0 \quad \forall \vec{v}, \vec{w}$$

Want to show:  $\forall p, \forall \vec{u}, \nabla f_p(\vec{u}) = 0, \nabla^2 f_p(\vec{u}) \succeq 0 \Rightarrow f_p(\vec{u}) = 0 \quad (*)$

Insight: for fixed  $\vec{u}$ , finding  $p$  counterexample to  $(*)$  is convex problem

$$\{p \mid \nabla f_p(\vec{u}) = 0\}$$

$$\{p \mid \nabla^2 f_p(\vec{u}) \succeq 0\}$$



Certificate:  $\forall \vec{u}$ , find  $\vec{\lambda}, \vec{v}_i$  s.t.

$$\nabla f_p(\vec{u})(\vec{\lambda}) + \sum_i \nabla^2 f_p(\vec{u})(\vec{v}_i, \vec{v}_i) = -f_p(\vec{u})$$

$$\Rightarrow -f_p(\vec{u}) \geq 0 \Rightarrow f_p(\vec{u}) = 0.$$

$$\nabla f_p(\vec{u})(\vec{\lambda}) = \langle u_1 \lambda_1 + u_2 \lambda_2, u_1^2 + u_2^2 - p \rangle$$

If  $u_1(x)$  and  $u_2(x)$  are coprime, then  $\exists \lambda_1, \lambda_2$  s.t.

$$u_1 \lambda_1 + u_2 \lambda_2 = (\text{any polynomial}) = p - u_1^2 - u_2^2$$

$$\nabla^2 f_p(\vec{u})(\vec{\lambda}) = - \|u_1^2 + u_2^2 - p\|^2 \quad \checkmark$$

On the other hand, suppose  $u_1 = u_2$ .

$$\text{Let } p = \frac{1}{2} 2t_i^2, \vec{\lambda} = (-u_1, -u_2), \vec{v}_2 = (t_i, -t_i).$$

$$\nabla f_p(\vec{u})(\vec{\lambda}) = - \langle \sigma(\vec{u}), \sigma(\vec{u}) - p \rangle$$

$$\nabla^2 f_p(\vec{u})(\vec{v}, \vec{v}) = \langle 2t_i^2, \sigma(\vec{u}) - p \rangle + 2 \|u_1 t - u_2 t\|^2$$

$$(*) = \left\langle \underbrace{\frac{p}{2} 2t_i^2}_{\vec{w}} - \sigma(\vec{u}), \sigma(\vec{u}) - p \right\rangle = -f_p(\vec{u}) \quad \checkmark$$

$u_1$  and  $u_2$  share common factor:  $u_1 = u'_1 g, u_2 = u'_2 g$ .

Always can always find  $\vec{\lambda}$  s.t.

$$\nabla f_p(\vec{u})(\vec{\lambda}) = \langle w g, \sigma(\vec{u}) - p \rangle$$

To make second term of Hessian zero, choose  $\vec{v} = \begin{pmatrix} -u'_1 t \\ u'_2 t \end{pmatrix}$

$$\nabla^2 f_p(\vec{u})(\vec{v}, \vec{v}) = \langle t^2 (u'_1^2 + u'_2^2), \sigma(\vec{u}) - p \rangle$$

Need to find  $w, t_i$ , s.t.  $p = w g + \left( \frac{1}{2} t_i^2 \right) \underbrace{(u'_1^2 + u'_2^2)}$

$$p \equiv sq \pmod{g}$$

Suppose  $\gcd(q, g) = 1$ , then  $\exists a, b$  s.t.  $aq + bg = 1$ .

Lemma: if  $a(x) > 0 \quad \forall x \in \{x \in \mathbb{R} \mid g(x)=0\}$ , then

$\exists \lambda, s \text{ s.t. } a = s^2 + \lambda g$ .

$$s^2 q + \lambda' g = 1 \quad \underbrace{s^2 p q}_{\sum_i t_i^2} + \lambda g = p$$

Otherwise,  $\gcd(q, g) = h$