

Polynomial Optimization: Moment Relaxations.

$$\underline{g} = (g_1, \dots, g_m), \quad g_i \in \mathbb{R}[X].$$

$$\text{Today: } S = S(\underline{g}) = \{x \in \mathbb{R}^n \mid g_1(x) \geq 0 \dots g_m(x) \geq 0\}$$

is always assumed to be compact.

Final goal: understand the dual of the

$$\text{SDP program } f_{S, \pi}^* = \sup \{ \lambda \in \mathbb{R} \mid f - \lambda \in M(\underline{g})_{2\pi} \}$$

Dual SDP: denote $\text{vec}(f)$ the vector of coeffs of f in the monomial basis $(\text{vec}(f) = \begin{pmatrix} f_0 \\ \vdots \end{pmatrix})$ if $f = \sum f_\alpha x^\alpha$

then the dual can be written as

$$f_{\text{mom}, \pi}^* = \inf \left\{ (-y_\alpha) \begin{pmatrix} 1 \\ f_0 \\ \vdots \end{pmatrix} \in \mathbb{R} \mid \begin{pmatrix} 1 \\ y_\alpha \end{pmatrix} \in \mathbb{R}^N, \quad M_\pi(\underline{y}) \succeq 0 \right. \\ \left. M_{\kappa_i}(g_i \cdot \underline{y}) \succeq 0 \quad \forall i \right\}$$

$y_0 = 1.$

where:

- $N := \dim \mathbb{R}[X]_{\leq \pi}, \quad \kappa_i := \pi - \lfloor \frac{\deg g_i}{2} \rfloor,$

- $M_\pi(\underline{y}) := (y_{\alpha+\beta})_{|\alpha|, |\beta| \leq \pi} \in \mathbb{R}^N$

- $M_{\kappa_i}(g_i \cdot \underline{y}) := \left(\sum_{\gamma} g_i^\delta y_{\alpha+\beta+\gamma} \right)_{|\alpha|, |\beta| \leq \kappa_i} \quad (g_i = \sum_{\gamma} g_i^\delta x^\gamma)$

→ Make the example $f = x, g = 1 - x^2, \pi = 2.$

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Moreover (with many simplifications):

if $\text{rk}(M_{n-1}(y)) = \text{rk} M_n(y) = k$ for $y = (y_\alpha) : (-y_\alpha^{-1}) \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} = f_{\text{min},n}^x$ (i.e. y is the output of the interior point method) then we have $f_{\text{min},n}^x = f^*$ (finite convergence), and there are k minimizers that can be extracted from $M_n(y)$.

Question: what's happening? \rightarrow Duality

Definition 1 Borel (a Radon) measure supported on S is a σ -finite, positive measure on the σ -algebra generated by the compact subsets of S .

We denote the set of Borel measures by $\mathcal{M}(S)$.

Concrete examples:

- Dirac deltas δ_x (for $x \in S$). We have, for ACS:

$$\delta_x(A) = \begin{cases} 1 & \text{if } x \in A \\ 0 & \text{if } x \notin A \end{cases}, \quad \text{and}$$
$$\int f d\delta_x = f(x) \quad \forall f.$$

- Restriction of Lebesgue measure. $\int_S f(x) dx = \int f d\mu$

$$\mu(A) = \text{Lebesgue measure of } A \cap S.$$

$$\int f d\mu = \int_S f(x) dx.$$

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→ there are many other things inside $M(S)$:

Gaussian measures restricted to S ,

integration along curves $C \subset S, \dots$. We have:

$$f^* = \inf \left\{ \int f d\mu \mid \mu \in M(S), \int 1 d\mu = 1 \right\} \quad (\text{Ex.})$$

Basic property of integration ($\mu \in M(S)$):

$$\int a f + b g d\mu = a \int f d\mu + b \int g d\mu \quad (a, b \in \mathbb{R})$$

→ Integration is linear!

$\mathbb{R}[x]$ is an \mathbb{R} -vector space, $\mathbb{R}[x]^v = \{ L: \mathbb{R}[x] \rightarrow \mathbb{R} \mid$

We denote $L_\mu: \mathbb{R}[x] \rightarrow \mathbb{R} \in \mathbb{R}[x]^v$ "row vectors"
 $f \mapsto \int f d\mu$ $\left. \begin{array}{l} L \text{ is linear} \\ \end{array} \right\}$

the moments of μ are the sequence of integration of monomials:
$$\left\{ \int x^\alpha d\mu \right\}_{\alpha \in \mathbb{N}^n} = \left\{ \int x_1^{\alpha_1} \dots x_n^{\alpha_n} d\mu \right\}_{\alpha \in \mathbb{N}^n} = \left\{ L_\mu(x^\alpha) \right\}_{\alpha \in \mathbb{N}^n}$$

More generally, for $L \in \mathbb{R}[x]^v$ the sequence:

$\left\{ L(x^\alpha) \right\}_{\alpha \in \mathbb{N}^n} = \left\{ y_\alpha \right\}_{\alpha \in \mathbb{N}^n}$ is called pseudo-moment sequence
 $L(x^\alpha) \in \mathbb{R}$

→ Pseudo-moments determine uniquely L :

if $f = \sum_{\alpha} f_{\alpha} x^{\alpha}$, then

We substitute the monomial x^{α} by the value y_{α} .

$$L(f) = L\left(\sum_{\alpha} f_{\alpha} x^{\alpha}\right) = \sum_{\alpha} f_{\alpha} L(x^{\alpha}) = \sum_{\alpha} f_{\alpha} y_{\alpha} =$$

$$= \left(- y_{\alpha} - \right) \begin{pmatrix} 1 \\ f_{\alpha} \\ \vdots \end{pmatrix}$$

In coordinates, linear functionals are row vectors of pseudo-moments (2)

the Moment Problem: Given a sequence of real numbers $\{y_\alpha\}_{\alpha \in \mathbb{N}^n}$, $\exists \mu \in \mathcal{M}(S)$:

$$y_\alpha = \int x^\alpha d\mu = L_\mu(x^\alpha) \quad \forall \alpha \in \mathbb{N}^n?$$

Equivalently: given $L \in \mathbb{R}[x]^{\vee}$, $\exists \mu \in \mathcal{M}(S)$
s.t. $L = L_\mu$?

theorem (Haviland) Given $L: \mathbb{R}[x] \rightarrow \mathbb{R}$ linear,

$\exists \mu \in \mathcal{M}(S) : L = L_\mu$ if and only if,

$\forall f \in \mathbb{R}[x]$, $f \geq 0$ on S , $L(f) \geq 0$. " $\mathcal{M}(S) = \mathcal{P}(S)^*$ "

"only if" is clear: if $f \geq 0$ on S ,

$$L_\mu(f) = \int f d\mu \geq 0 \quad \text{by definition of measure and integral}$$

theorem (Putinar)

Given $L: \mathbb{R}[x] \rightarrow \mathbb{R}$ linear, $\exists \mu \in \mathcal{M}(S_{\text{reg}}) : L = L_\mu$

if and only if $\forall q \in \mathcal{H}(g)$, $L(q) \geq 0$

if and only if $\forall f \in \mathbb{R}[x]$, $L(f^2) \geq 0$, $L(f^2 g_1) \geq 0, \dots,$

$$L(f^2 g_n) \geq 0.$$

Applying L to a polynomial means to substitute the monomials x^α by the values

$$y_\alpha = L(x^\alpha). \quad \text{therefore:}$$

$$\begin{aligned}
L(f^2) &= L\left(\langle - \mid f \mid - \rangle \begin{pmatrix} 1 \\ x^k \\ \vdots \end{pmatrix} \langle - \mid x^\alpha \mid - \rangle \begin{pmatrix} 1 \\ f_\alpha \\ \vdots \end{pmatrix}\right) = \\
&= L\left(\langle - \mid f \mid - \rangle \begin{pmatrix} x^{\alpha+\beta} \\ \vdots \end{pmatrix}_{\alpha, \beta \in \mathbb{N}^n} \begin{pmatrix} 1 \\ f_\alpha \\ \vdots \end{pmatrix}\right) \\
&= \langle - \mid f \mid - \rangle \begin{pmatrix} L(x^{\alpha+\beta}) \\ \vdots \end{pmatrix}_{\alpha, \beta} \begin{pmatrix} 1 \\ f_\alpha \\ \vdots \end{pmatrix} = \\
&= \langle - \mid f \mid - \rangle \begin{pmatrix} y_{\alpha+\beta} \\ \vdots \end{pmatrix}_{\alpha, \beta} \begin{pmatrix} 1 \\ f_\alpha \\ \vdots \end{pmatrix}
\end{aligned}$$

therefore, if $M(y) = (y_{\alpha+\beta})_{\alpha, \beta}$ is the moment matrix,

$$L(f^2) \geq 0 \quad \forall f \Leftrightarrow M(y) \succeq 0.$$

In a similar way: $L(g_i f^2) \geq 0 \quad \forall f \Leftrightarrow$
 $\Leftrightarrow M(g_i \cdot y) \succeq 0.$

Problem: infinite matrices!

Solution: Restriction to $\mathbb{R}[x]_{\leq r}$, or equivalently,
 $1 \leq \alpha \leq r.$

$$\rightarrow M_r(y), \dots$$

truncated moment and
localizing matrices.

theorem (Curtis-Fisker)

If $\text{rk } M_{\pi}(\underline{y}) = \text{rk } M_{\pi+1}(\underline{y}) = k$, then

$y_{\alpha} = \int x^{\alpha} d\mu$ is a moment linear functional,

where μ is a (convex) sum of k Dirac deltas.

Moreover, we can extract the points associated with the Dirac deltas from $M_{\pi+1}(\underline{y})$.

Example ($n=2$). Assume $y_0=1, y_1=0, y_2=0$.

$$M_0(\underline{y}) = (y_0) = 1$$

$$M_1(\underline{y}) = \begin{pmatrix} y_0 & y_1 \\ y_1 & y_2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$

} same rank = 1

• $y_1 = \int x^1 d\delta_0$, δ_0 is the Dirac delta at the origin.

The point corresponds to the vector $\begin{pmatrix} 1 \\ 0 \end{pmatrix} \in \ker M_1(\underline{y})$.

Other formulation:

$$f_{\text{mom}, \pi}^* = \inf \left\{ L(f) \in \mathbb{R} \mid L \in \mathbb{R}[x]_{\leq 2\pi}^{\vee}, L(M(\underline{y})_{2\pi}) \geq 0, L(1)=1 \right\}$$

Properties

$$\bullet f_{\text{mom}, \pi}^* \leq f_{\text{mom}, \pi+1}^* \leq f^* \quad \forall \pi.$$

$$\bullet f_{\text{pos}, \pi}^* \leq f_{\text{mom}, \pi}^* \quad \forall \pi.$$

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