

LECTURE 4

We have defined two cones in $\mathbb{R}[x]$:

$P(\mathbb{R}^n)$ = globally positive polynomials

* Σ_n = sums of squares polynomials

and truncated in degree $2d$

$$P(\mathbb{R}^n)_{2d} = P(\mathbb{R}^n) \cap \mathbb{R}[x]_{2d}$$

$$\Sigma_{n,2d} = \Sigma_n \cap \mathbb{R}[x]_{2d}$$

Definition Let $C \subset V$ be a cone. The dimension of C is the dimension of its affine hull (the smallest affine space $H \subset V$ containing C).

Proposition $P(\mathbb{R}^n)$ and Σ_n (resp. $P(\mathbb{R}^n)_{2d}$ and $\Sigma_{n,2d}$) are full-dimensional ^{closed} convex cones of $\mathbb{R}[x]$ (resp. of $\mathbb{R}[x]_{2d}$):

$$\dim P(\mathbb{R}^n) = \dim \Sigma_n = \dim \mathbb{R}[x] = +\infty$$

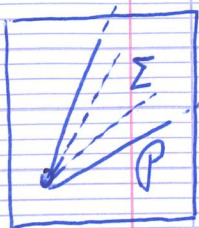
$$\dim P(\mathbb{R}^n)_{2d} = \dim \Sigma_{n,2d} = \dim \mathbb{R}[x]_{2d} = \binom{2d+n}{n}$$

We suppose to work with homogeneous polynomials (for simplicity reasons).

The goal is to ~~also~~ construct the dual cones $(P(\mathbb{R}^n)_{2d})^*$ and $(\Sigma_{n,2d})^*$.

vector space of polynomials

dual set vector space



$$\mathbb{R}[x]_{2d}$$

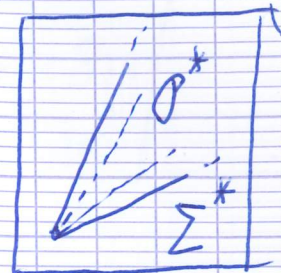
$$\bigcup P(\mathbb{R}^n)_{2d}$$

$$\bigcup \Sigma_{n,2d}$$

$$\mathbb{R}[x]_{2d}^*$$

$$\bigcup (\Sigma_{n,2d})^*$$

$$\bigcup (P(\mathbb{R}^n)_{2d})^*$$



The same for the whole space

$$\mathbb{R}[x]$$

$$\bigcup P(\mathbb{R}^n)$$

$$\bigcup \Sigma_n$$

$$\mathbb{R}[x]^*$$

$$\bigcup (\Sigma_n)^*$$

$$\bigcup (P(\mathbb{R}^n))^*$$

Choose the monomial basis $\langle x^\alpha \rangle_{\alpha \in \mathbb{N}_{2d}^n} = \mathbb{R}[x]_{2d}$

Then $l \in \mathbb{R}[x]_{2d}^*$

$\rightarrow l: \mathbb{R}[x]_{2d} \rightarrow \mathbb{R}$ linear functional

It is determined by its values on x^α :

$$l\left(\sum c_\alpha x^\alpha\right) = \sum c_\alpha l(x^\alpha) \quad \alpha \in \mathbb{N}_{2d}^n$$

So if $l(x^\alpha) = y_\alpha \in \mathbb{R} \quad \forall \alpha$, then we have

the identification of $\mathbb{R}[x]_{2d}^* = \mathbb{R}^{\binom{2d+n}{n}}$

$$\left\{ (y_\alpha)_{\alpha \in \mathbb{N}_{2d}^n} : y_\alpha \in \mathbb{R} \right\}$$

Analogously

$$\mathbb{R}[x]^v = \{ (y_\alpha)_{\alpha \in \mathbb{N}^n} : y_\alpha \in \mathbb{R} \} =: \mathbb{R}^\infty$$

Let $S^2(\mathbb{R}[x]_d)$ be the set of all quadratic forms over $\mathbb{R}[x]_d$:

$$S^2(\mathbb{R}[x]_d) = \{ Q : \mathbb{R}[x]_d \rightarrow \mathbb{R} \quad \left. \begin{array}{l} \text{bilinear} \\ \text{and} \\ \text{symmetric} \end{array} \right\}$$

$$S^2(\mathbb{R}[x]_d) \approx \text{Sym}_{\binom{n+d}{d}}(\mathbb{R})$$

We associate

$$\begin{aligned} \mathbb{R}[x]_{2d}^v &\rightarrow S^2(\mathbb{R}[x]_d) \\ \ell &\mapsto Q_\ell \end{aligned}$$

with

$$Q_\ell(f) = \ell(f^2).$$

Remark What is the matrix of the quadratic form Q_ℓ ? For example, consider the

functional $\ell \in \mathbb{R}[x_1, x_2]_4^v$

$$\begin{aligned} \ell : \mathbb{R}[x_1, x_2]_4 &\rightarrow \mathbb{R} \\ f &\mapsto f(1, 2) \end{aligned}$$

evaluation in $(1, 2)$ is linear!

Monomial basis

$$\langle x_1^2, x_1 x_2, x_2^2 \rangle_{\mathbb{R}} = \mathbb{R}[x_1, x_2]_2$$

$$(f_1 + f_2)(1, 2) = f_1(1, 2) + f_2(1, 2)$$

represented by
the matrix

$$Q_\ell \sim \begin{matrix} & x_1^2 & x_1 x_2 & x_2^2 \\ x_1^2 & \ell(x_1^4) & \ell(x_1^3 x_2) & \ell(x_1^2 x_2^2) \\ x_1 x_2 & \ell(x_1^3 x_2) & \ell(x_1^2 x_2^2) & \ell(x_1 x_2^3) \\ x_2^2 & \ell(x_1^2 x_2^2) & \ell(x_1 x_2^3) & \ell(x_2^4) \end{matrix} = \begin{bmatrix} 1 & 2 & 4 \\ 2 & 4 & 8 \\ 4 & 8 & 16 \end{bmatrix}$$

Hankel
Matrix

Proposition Let $\ell \in \mathbb{R}[x]_{2d}^V$. Then $\ell \in \Sigma_{n,2d}^*$
if and only if Q_ℓ is positive semidefinite.

Proof

\Rightarrow ℓ is positive on sums of squares,
in particular $\forall f \in \mathbb{R}[x]_d$

$$0 \leq \ell(f^2) = Q_\ell(f) \Rightarrow Q_\ell \text{ is psd.}$$

$\Leftarrow \forall f \in \mathbb{R}[x]_d \quad Q_\ell(f) = \ell(f^2) \geq 0.$

$$\rightarrow \ell\left(\sum f_i^2\right) = \sum \ell(f_i^2) \geq 0 \quad \left(\begin{array}{l} \text{sum of positive} \\ \text{numbers is positive} \end{array}\right)$$

We conclude that

$$\Sigma_{n,2d}^* = \left\{ \ell \in \mathbb{R}[x]_{2d}^V : Q_\ell \succeq 0 \right\}$$

It is the feasible set of a SDP program
(a Spectrahedron)
slow Lasserre's book.

What about $P(\mathbb{R}^n)_{2d}^*$?

We want functionals $l: \mathbb{R}[x]_{2d} \rightarrow \mathbb{R}$
that are "positive over positive polynomials".
For instance, point evaluations:

Definition We denote by l_v , for a given
 $v \in \mathbb{R}^n$, the evaluation on v :

$$l_v: \mathbb{R}[x]_{2d} \longrightarrow \mathbb{R}$$
$$f \mapsto f(v)$$

Exercise Prove that $l_v \in \mathbb{R}[x]_{2d}^* \forall v$.

We define

$$K_{n,2d} = \underset{\substack{\text{convex} \\ \text{hull}}}{\text{convex}} \left(\left\{ l_v: \mathbb{R}[x]_{2d} \rightarrow \mathbb{R} \mid v \in \mathbb{R}^n \right\} \right)$$

Theorem $P(\mathbb{R}^n)_{2d}^* = K_{n,2d}$.

Proof

- (1) Prove by exercise that $K_{n,2d}$ is a closed convex cone (exercise). We know the same for $P(\mathbb{R}^n)_{2d}^*$.
- (2) Deduce that $K_{n,2d}^{**} = K_{n,2d}$, $P(\mathbb{R}^n)^{**} = P(\mathbb{R}^n)$.
- (3) Hence it is sufficient to prove that

$W = \mathbb{R}[x]_{2d}$ biduality $W \xrightarrow{\quad} W^V \xrightarrow{\quad} (W^V)^V \approx W$
 $p \mapsto \ell_p$ $q_p(\ell) = \ell(p)$
 $p \mapsto q_p$ injective \Rightarrow isom.

$K_{n,2d}^* = \mathcal{P}(\mathbb{R}^n)_{2d}$. Indeed, this is true:

$K_{n,2d}^* = \{ p \in \mathbb{R}[x]_{2d} : q_p(\ell) \geq 0 \ \forall \ell \in K_{n,2d} \} =$

$(\mathbb{R}[x]_{2d})_{2d}^* \approx \mathbb{R}[x]_{2d} = \{ p \in \mathbb{R}[x]_{2d} : q_p(\ell) \geq 0 \ \forall v \in \mathbb{R}^n \} =$

canonical isomorphism

$= \{ p \in \mathbb{R}[x]_{2d} : \ell_v(p) \geq 0 \ \forall v \} =$

because $K_{n,2d}$ is the conical hull of $\{\ell_v\}$

$= \{ p \in \mathbb{R}[x]_{2d} : p(v) \geq 0 \ \forall v \in \mathbb{R}^n \} = \mathcal{P}(\mathbb{R}^n)_{2d}$

We deduce that $\mathcal{P}(\mathbb{R}^n)_{2d}^*$ is the conical hull of the point evaluations. □

Moment Problem

Moments of measures will be the key to understand what is $\mathcal{P}(\mathbb{R}^n)_{2d}^*$ and to solve our original problem, that is:

(SA0) $f^* = \inf_{x \in S(g)} f^*(x)$
 s.t. $x \in S(g) = \{ y \in \mathbb{R}^n : g_i(y) \geq 0 \ \forall i=1, \dots, m \}$

Definition

A Borel measure on \mathbb{R}^n is a function

$$\mu: \mathbb{R}^n \rightarrow \mathbb{R} \quad \text{such that}$$

(0) $\mu(x) \geq 0 \quad \forall x \in \mathbb{R}^n$ (nonnegative)

(0) If $\{E_k\}_{k \in \mathbb{N}}$ is a family of ^{pairwise} disjoint sets on which μ is defined, then

$$\mu\left(\bigcup_{k=1}^{\infty} E_k\right) = \sum_{k=1}^{\infty} \mu(E_k) \quad \left(\begin{array}{l} \text{countable} \\ \text{additivity on} \\ \text{disjoint sets} \end{array}\right).$$

A σ -algebra on a set X is a family

\mathcal{F} of subsets of X , such that

(0) $X \in \mathcal{F}$

(0) $Y \in \mathcal{F} \rightarrow X \setminus Y \in \mathcal{F}$

(0) $\{Y_k\}_{k \in \mathbb{N}} \subset \mathcal{F} \rightarrow \bigcup_{k=1}^{\infty} Y_k \in \mathcal{F}$.

The ~~sets~~ elements of \mathcal{F} are called measurable sets. If X is a topological space, the

smallest σ -algebra containing the open sets of X is called the Borel algebra \mathcal{B} , and every $Y \in \mathcal{B}$ is called a Borel set.

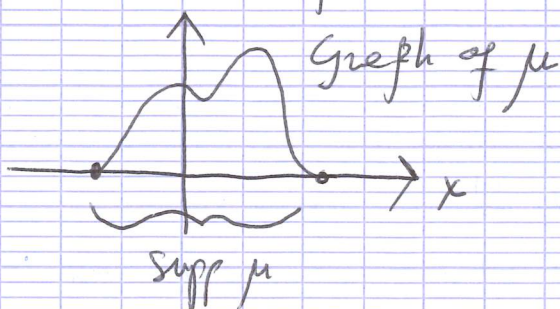
10h 30

The support of a measure μ is the complement of the largest open set where μ vanishes:

$$\text{Supp}(\mu) = \mathbb{R}^n \setminus \bigcup_{\substack{A \subset \mathbb{R}^n \\ \text{open set}}} \{A : \mu(A) = 0\}.$$

↓ closed set!

Example:



Example (Gaussian measure)

Let λ be the Lebesgue measure on \mathbb{R}^n . Then the standard Gaussian measure is

$$E \subset \mathbb{R}^n \text{ set: } \mu(E) = \frac{1}{\sqrt{2\pi}^n} \int_E e^{-\frac{1}{2} \sum x_i^2} d\lambda$$

is a Borel measure on \mathbb{R}^n , with $\mu(\mathbb{R}^n) = 1$ (probability measure).

$\mu(E)$ probability of the event E .

$$\text{Supp } \mu = \mathbb{R}^n$$

Example (Dirac's delta)

For $\bar{x} \in \mathbb{R}^n$, we denote the Dirac measure with support \bar{x} the measure $\delta_{\bar{x}}$ such that

$$\forall E \subset \mathbb{R}^n$$

$$\delta_{\bar{x}}(E) = \begin{cases} 0 & \text{if } \bar{x} \notin E \\ 1 & \text{if } \bar{x} \in E \end{cases}$$

For instance $n=1$, $x=2$ then

$$\delta_2((0,1)) = 0, \quad \delta_2((0,3)) = 1, \quad \delta_2(\{5\}) = 0$$

$$\delta_2(\{2,3\}) = 1.$$

Let μ be a measure with $\text{supp}(\mu) \subset S(g)$ for $S(g)$ a given semi-algebraic set.

We fix the monomial basis $\langle x^\alpha \rangle_{\alpha \in \mathbb{N}^n}$ of $\mathbb{R}[x]$. We define the moments of μ as the (infinite sequence)

$$y_\alpha = \int_{S(g)} x^\alpha d\mu$$

Measure with
support in $S(g)$
 μ

 \longrightarrow
 $\longleftarrow ?$

Moments of
measures
 $(y_\alpha)_{\alpha \in \mathbb{N}^n}$

Moment Problem Given a sequence $(y_\alpha)_{\alpha \in \mathbb{N}^n}$ of real numbers, find a measure μ such that (y_α) is the sequence of its moments

on a given set $S(g)$.

Definition We say that $(y_\alpha)_{\alpha \in \mathbb{N}^n}$ has a representing measure if there exists μ such that $y_\alpha = \int x^\alpha d\mu \quad \forall \alpha \in \mathbb{N}^n$.

Truncated Moment Problem We substitute $\mathbb{R}[x]_{\leq d}$ to $\mathbb{R}[x]$ and \mathbb{N}_d^n to \mathbb{N}^n , and we have finite sequences.

Special cases of moment problems (MP)

* Hamburger ~~moment~~ MP : $S(g) = \mathbb{R}$

* Stieltjes MP : $S(g) = \mathbb{R}_{\geq 0}$, $y_d \geq 0$.

Q. Why MP is related to SAO?

Definition For $y = (y_\alpha)_{\alpha \in \mathbb{N}^n} \in \mathbb{R}$, the Riesz functional $\mathcal{L}_y : \mathbb{R}[x] \rightarrow \mathbb{R}$ is defined by

$$\text{For } f = \sum_{\alpha \in \mathbb{N}^n} c_\alpha x^\alpha \mapsto \mathcal{L}_y(f) = \sum_{\alpha \in \mathbb{N}^n} c_\alpha y_\alpha.$$

\mathcal{L}_y "linearizes" polynomials.

Theorem (Riesz-Haviland)

Let $y = (y_\alpha)_{\alpha \in \mathbb{N}^n}$, $S \subset \mathbb{R}^n$ be closed. There exists a Borel measure μ on S such that

$$y_\alpha = \int_S x^\alpha dx \quad \forall \alpha \in \mathbb{N}^n$$

if and only if $L_y(f) \geq 0$ for all $f \in \mathcal{P}(S)$.

Hence we have another characterization of the dual cone $(\mathcal{P}(\mathbb{R}^n))^*$ of $\mathcal{P}(\mathbb{R}^n)$:

$$\mathcal{P}(\mathbb{R}^n)^* = \left\{ y \in (\mathbb{R}^{\mathbb{N}^n}) : \exists \mu \text{ such that } y_\alpha = \int x^\alpha d\mu \forall \alpha \right\}$$

the same for $\mathcal{P}(S)^*$ ($\text{supp}(\mu) \subset S$).

A second link with SAO comes by looking at the problem:

$$(GMP) \quad m^* = \inf_{\substack{\mu \\ \text{supp } \mu \subset S(g) \\ \int_{S(g)} d\mu = 1}} \int_{S(g)} f d\mu$$

We recall the SAO

$$(SAO) \quad f^* = \inf_{x \in S(y)} f(x).$$

Theorem With the notation above, $f^* = m^*$.

Proof for any M such that $f(x) \leq M$, $\bar{x} \in S(y)$ is an admissible measure and

$$\int_{S(y)} f d\bar{\mu}_x = f(\bar{x}) \leq M$$

$$\rightarrow m^* = -\infty.$$

(2) $f^* > -\infty$, $f^* \in \mathbb{R}$. Since $f(x) \geq f^*$ $\forall x \in S(y)$, then $\int_{S(y)} f d\mu \geq \int_{S(y)} f^* d\mu = f^* \int_{S(y)} d\mu = f^*$ $\forall \mu$ such that $\int_{S(y)} d\mu = 1$.

Hence $m^* \geq f^*$.

Viceversa, for $\bar{x} \in S(y)$, we consider $\bar{\mu}_x$,

which is a feasible measure:

$$m^* \leq \int_{S(y)} f d\bar{\mu}_x = f(\bar{x})$$

passing to Inf we have $m^* \leq f^*$ \square