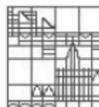


GNS construction

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- $\mathbb{R}[x] := \mathbb{R}[x_1, \dots, x_n]$, $x = (x_1, \dots, x_n)$
- $\mathbb{R}[x]_t := \{p \in \mathbb{R}[x] : \deg(p) \leq t\}$
- $\mathbb{R}[x]_t^* := \{L : \mathbb{R}[x]_t \rightarrow \mathbb{R} \text{ linear form}\}$
- $\alpha = (\alpha_1, \dots, \alpha_n) \in \mathbb{N}^n$
- $x^\alpha := x_1^{\alpha_1} \cdots x_n^{\alpha_n} \in \mathbb{R}[x]_d \leftrightarrow \alpha \in \mathbb{N}_d^n$, (i.e.
 $|\alpha| := \alpha_1 + \dots + \alpha_n \leq d$)

Definition

We say that a linear form $L \in \mathbb{R}[x]_{2d}^*$ is integration with respect to a measure μ , if $L(p) = \int p d\mu \quad \forall p \in \mathbb{R}[x]_{2d}$.

Definition

Given a linear form $L \in \mathbb{R}[x]_{2d}^*$, we can define its respective moment matrix, indexed by $(x^\alpha)_{\alpha, |\alpha| \leq d}$, in this way:

$$M_L = ((L(x^{\alpha+\beta}))_{\alpha, \beta, |\alpha| \leq d, |\beta| \leq d}$$

Example

For $L = \text{ev}(0, 1) \in \mathbb{R}[x_1, x_2]_2^*$, its moment matrix is:

$$M_L = \begin{pmatrix} L(1) & L(x) & L(y) \\ L(x) & L(x^2) & L(xy) \\ L(y) & L(xy) & L(y^2) \end{pmatrix} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \end{pmatrix}$$

Bayer and Teichmann Theorem

A linear form $L \in \mathbb{R}[x]_{2d}^$, is a integration with respect to a measure if and only if there exists an integer N , nodes $x_1, \dots, x_N \in \mathbb{R}^n$ and weights $\lambda_1 > 0, \dots, \lambda_N > 0$, such that, $L(p) = \sum_{i=1}^N \lambda_i p(x_i), \forall p \in \mathbb{R}[x]_{2d}$.*

Remark

If $L \in \mathbb{R}[x]_{2d}^$, and L is integration with respect to a measure, then $L(\sum \mathbb{R}[x]_{2d}^2) \subseteq \mathbb{R}_{\geq 0}$, that is $M_L \geq 0$*

Given $L \in \mathbb{R}[x]_{2d}^*$, and $L(\sum \mathbb{R}[x]_d^2) \subseteq \mathbb{R}_{\geq 0}$

- Is L , or at least $L|_{\mathbb{R}[x]_k}$ for some $k \leq 2d$, integration with respect to a measure?
- If L is integration with respect to a measure, How can I recover the nodes and weights of such a measure?

For this we will use a variation of the **Gelfand-Naimark-Segal** construction.

Given $L \in \mathbb{R}[x]_{2d}^*$, and $L(\sum \mathbb{R}[x]_d^2) \subseteq \mathbb{R}_{\geq 0}$. We would like to find :

- x_1, \dots, x_N points in \mathbb{R}^n , $\lambda_1, \dots, \lambda_N > 0$ weights such that

$$L = \sum_{i=1}^N \lambda_i \text{ev}(x_i)$$

That is similar to find:

- A finite dimensional space V and commuting self-adjoint endomorphisms M_1, \dots, M_n of V and $a \in V$ such that

$$L(p) = \langle p(M_1, \dots, M_n)a, a \rangle$$

Idea: if we had $L(p^2) > 0$ for all $p \neq 0$:

- $V := \mathbb{R}[x]$
- $\langle p, q \rangle := L(pq)$
- $M_i : \mathbb{R}[x] \rightarrow \mathbb{R}[x] : p \mapsto x_i p, i \in \{1, \dots, n\}$
- $a := 1 \in \mathbb{R}[x]$

Let $L \in \mathbb{R}[x]_{2d}^*$ with $L(\sum \mathbb{R}[x]^2) \subseteq \mathbb{R}_{\geq 0}$

- $U_L := \{p \in \mathbb{R}[x]_d : \forall q \in \mathbb{R}[x]_d : L(pq) = 0\}$ **GNS kernel**
- $V_L := \frac{\mathbb{R}[x]_d}{U_L}$ **GNS representation space of L**
- $\langle \bar{p}, \bar{q} \rangle_L := L(pq)$ ($p, q \in \mathbb{R}[x]_d$) **GNS scalar product**
- Then we have built $(V_L, \langle, \rangle_L)$ a Euclidean Vector Space

We had a euclidean vector space $(V_L, \langle, \rangle_L)$:

- $\Pi_L: V_L \rightarrow \{\bar{p} : p \in \mathbb{R}[x]_{d-1}\}$ **orthogonal projection**
- $M_{L,i}: \Pi_L V_L \rightarrow \Pi_L V_L : \bar{p} \rightarrow \Pi_L(\overline{X_i p})$ ($p \in \mathbb{R}[x]_{d-1}$) and $i \in \{1, \dots, n\}$.

$M_{L,i}$, called **i-th truncated GNS multiplication operator**, is self-adjoint endomorphism of $\Pi_L V_L$

Simultaneous Diagonalization

Let V be a finite euclidean vector space, and M_1, \dots, M_n , commuting selfadjoint endomorphisms of V . Then there exists an orthonormal basis of V which contains all the common eigenvectors of M_i for $i = 1, \dots, n$.

Proposition (Schweighofer)

Let $L \in \mathbb{R}[x]_{2d}^$ with $L(\sum \mathbb{R}[x]_d^2) \subseteq \mathbb{R}_{\geq 0}$. Suppose that the truncated GNS multiplication operators of L commute. And we set:*

$$W_L := \left\{ \sum_{i=1}^m p_i q_i : m \in \mathbb{N}_0, p_i, q_i \in \mathbb{R}[x]_{d-1} + U_L \right\} \supseteq \mathbb{R}[x]_{2(d-1)}$$

Then $L|_{W_L}$ is integration with respect to a finitely supported measure.

Definition

Let $L \in \mathbb{R}[x]_{2d}^*$ with $L(\mathbb{R}[x]_d^2) \subseteq \mathbb{R}_{\geq 0}$. We say L is flat if $\mathbb{R}[x]_{d-1} + U_L = \mathbb{R}[x]_d$.

Proposition (Schweighofer)

Let $L \in \mathbb{R}[x]_{2d}^*$ with $L(\mathbb{R}[x]_d^2) \subseteq \mathbb{R}_{\geq 0}$ and $L' := L|_{\mathbb{R}[x]_{2(d-1)}}$. Then the following are equivalent:

- L is flat.
- $\Pi_L(V_L) = V_L$
- For all $\alpha \in \mathbb{N}^d$: ($|\alpha| = d \Rightarrow \exists p \in \mathbb{R}[x]_{d-1}$ such that $x^\alpha - p \in U_L$)
- $\dim(V_{L'}) = \dim(V_L)$
- $(L(X^{\alpha+\beta}))_{|\alpha|, |\beta| \leq d-1}$ and $(L(X^{\alpha+\beta}))_{|\alpha|, |\beta| \leq d}$ have the same rank.

Proposition (Schweighofer)

Let $L \in \mathbb{R}[x]_{2d}^$ with $L(\mathbb{R}[x]_d^2) \subseteq \mathbb{R}_{\geq 0}$. If L is flat then the GNS multiplication operators commute.*

Corollary

Let $L \in \mathbb{R}[x]_{2d}^$ with $L(\mathbb{R}[x]_d^2) \subseteq \mathbb{R}_{\geq 0}$ in one variable. Then $L|_{\mathbb{R}[x]_{2(d-1)}}$ comes from a measure.*

Corollary (Curto and Fialkow, 1996)

Let $L \in \mathbb{R}[x]_{2d}^$ with $L(\mathbb{R}[x]_d^2) \subseteq \mathbb{R}_{\geq 0}$. If L is flat then $L|_{\mathbb{R}[x]_{2(d-1)}}$ comes from a measure.*

$M_{L,i}$ commute does not necessarily imply that L is flat:

Example

Let $L = \frac{1}{3}(ev(1, 2) + ev(2, 3) + ev(4, 5)) \in \mathbb{R}[x_1, x_2]_4^*$, (with points in the line $y = x + 1$) here we have:

$$M_{L,1} = \begin{pmatrix} 3 & 1,6330 \\ 1,6330 & 3 \end{pmatrix} \quad M_{L,2} = \begin{pmatrix} 4 & 1,6330 \\ 1,6330 & 4 \end{pmatrix}$$

$M_{L,1}M_{L,2} = M_{L,2}M_{L,1}$ and L is not flat. And

$L' := \frac{1}{2}(ev(1.37, 2.37) + ev(4.63, 5.63))$ with $L' = L|_{\mathbb{R}[x]_2}$.

Example

Let $L = \frac{1}{6}(ev(1, 1) + ev(-1, -1) + ev(-1, 1) + ev(1, -1) + ev(2, -2) + ev(-2, 2)) \in \mathbb{R}[x_1, x_2]_6^*$, (points in $x^2 = y^2$) here L is not flat and the operators commute.

Interested

$L \in \mathbb{R}[x]_{2d}^*$ with $L(\sum \mathbb{R}[x]_d^2) \subseteq \mathbb{R}_{\geq 0}$ such that L is not necessarily flat and $M_{L,i}$ commute

Remark

Let $L = \sum_{i=1}^r \lambda_i \text{ev}_{x_i} \in \mathbb{R}[x, y]_{2d}^*$ (two variables) such that $x_i \in \{(x, y) \mid y = mx + n\} \subset \mathbb{R}^2$. Then the GNS multiplication operators commute.

Remark

Let $L = \sum_{i=1}^r \lambda_i \text{ev}_{x_i} \in \mathbb{R}[x]_{2d}^*$ with $r \leq \dim(\mathbb{R}[x]_{d-1})$ then "almost always" L is flat. For example for $n = 2$ if we take points in the circle sometimes L is not flat.

Polynomial optimization problem: minimize $p[x]$ over $x \in \mathbb{R}^n$, such that $p \in \mathbb{R}[x]$. We denote $\deg(p) := d$.

First SDP (Nie-Demmel-Sturmfels)

(NDS_k)

min $L(p)$ with:

- $L \in \mathbb{R}[x]_{2k}^*$
- $L(\sum \mathbb{R}[x]_k^2) \subseteq \mathbb{R}_{\geq 0}$
- $L(\frac{\partial p}{\partial x_j}(x^\alpha)) = 0$

for all α with $|\alpha| + d - 1 \leq 2k$

Second SDP (Heuristic-Program)

(H_k, λ)

min $(1 - \lambda)L(p) + \lambda E$

- $L \in \mathbb{R}[x]_{2k}^*$
- $L(\sum \mathbb{R}[x]_k^2) \subseteq \mathbb{R}_{\geq 0}$
- $L \approx \text{flat}$

where $\lambda \in [0, 1]$ is fixed.

Some examples:

Example 1

With NDS_4 for $p = x^4y^2 + x^2y^4 + 1 - 3x^2y^2$, the *Moztkin polynomial*. I get the minimizers :
 $(-16,8583, 0)$, $(0, -16,8583)$, $(\pm 1, \pm 1)$, $(16,583, 0)$, $(0, 16,583)$

Example 2

With NDS_4 for $p = x^6 + y^6 + 1 - x^4y^2 - x^2y^4 - x^4 - y^4 - x^2 - y^2 + 3x^2y^2$, the *Robinson polynomial*. I get the minimizers: $(\pm 1, \pm 1)$, $(1, 0)$, $(0, 1)$, $(-1, 0)$, $(0, -1)$

Example 3

With NDS_4 for the polynomial $x^2y^2(x^2 + y^2 - 1)$. I get the minimizers
 $(-14,89, 0)$, $(0, -14,89)$, $(0, 14,89)$, $(14,89, 0)$, $(\pm 0,5774, \pm 0,5774)$

Example 4

With $H_{3,1/60}$ for the *Moztkin polynomial* I get (approximately) the minimizers $(\pm 1, \pm 1)$, $(0, 0)$

Dankeschön!