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# The role of "positivity" in moment and polynomial optimization problems

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#### **Truncated K-Moment Problem**

Given an n-dimensional multisequence of degree m,

$$\beta \equiv \beta^{(m)} = \{\beta_i : i \in \mathbb{Z}_+^n, |i| \le m\},\$$

and a closed set  $K \subseteq \mathbb{R}^n$ , find conditions on  $\beta$  so that there exists a positive Borel measure  $\mu$  on  $\mathbb{R}^n$  such that  $supp \ \mu \subseteq K$  and

$$\beta_i = \int_{\mathbb{R}^n} x^i d\mu(x) \ (|i| \le m)$$

 $(x \equiv (x_1, \ldots, x_n), i \equiv (i_1, \ldots, i_n) \in \mathbb{Z}_+^n, x^i := x_1^{i_1} \cdots x_n^{i_n}).$  By analogy, in the *Full K*-Moment Problem, we are given  $\beta^{(\infty)}$ , with moment data of all degrees.

## Multisequence notation and moment matrices

Let  $\beta$  denote an *n*-dimensional real multisequence of degree m,

$$\beta \equiv \beta^{(m)} = \{\beta_i : i \in \mathbb{Z}_+^n, |i| \le m\},\$$

Example. For n=1, m=4,  $\beta^{(4)}:\beta_0,\ldots,\beta_4$ , we associate  $\beta^{(4)}$  to the moment matrix  $M_2$ , with rows and columns indexed by  $1, x, x^2$ , defined by

$$M_2(\beta) \equiv \left( egin{array}{ccc} eta_0 & eta_1 & eta_2 \ eta_1 & eta_2 & eta_3 \ eta_2 & eta_3 & eta_4 \end{array} 
ight).$$

Example. For n = 2, m = 4, consider  $\beta^{(4)}$ :  $\beta_{00}$ ,  $\beta_{10}$ ,  $\beta_{01}$ ,  $\beta_{20}$ ,  $\beta_{11}$ ,  $\beta_{02}$ ,  $\beta_{30}$ ,  $\beta_{21}$ ,  $\beta_{12}$ ,  $\beta_{03}$ ,  $\beta_{40}$ ,  $\beta_{31}$ ,  $\beta_{22}$ ,  $\beta_{13}$ ,  $\beta_{04}$ .

We associate  $\beta^{(4)}$  to the moment matrix  $M_2$ , with rows and columns indexed by 1, x, y,  $x^2$ , xy,  $y^2$ , defined by

$$M_2(\beta) \equiv \begin{pmatrix} \beta_{00} & \beta_{10} & \beta_{01} & \beta_{20} & \beta_{11} & \beta_{02} \\ \beta_{10} & \beta_{20} & \beta_{11} & \beta_{30} & \beta_{21} & \beta_{12} \\ \beta_{01} & \beta_{11} & \beta_{02} & \beta_{21} & \beta_{12} & \beta_{03} \\ \beta_{20} & \beta_{30} & \beta_{21} & \beta_{40} & \beta_{31} & \beta_{22} \\ \beta_{11} & \beta_{21} & \beta_{12} & \beta_{31} & \beta_{22} & \beta_{13} \\ \beta_{02} & \beta_{12} & \beta_{03} & \beta_{22} & \beta_{13} & \beta_{04} \end{pmatrix}.$$

## The Polynomial Optimization Problem

Let  $\mathcal{P}:=\mathbb{R}[x_1,\ldots,x_n]$ . For  $q_0,\ldots,q_k\in\mathcal{P},\ q_0\equiv 1$ , let  $\mathcal{K}_Q$  denote the basic closed semialgebraic set

$$K_Q = \{x \in \mathbb{R}^n : q_i(x) \ge 0 \ (0 \le i \le k)\}.$$

For  $p \in \mathcal{P}$ , we seek to compute (or estimate)

$$p_* := \inf_{x \in K_Q} p(x).$$

Later, we will discuss an algorithm of J.-B. Lasserre [2000] which estimates  $p_*$  based on "moment relaxations", and whose stopping criterion (when  $p_*$  is computed exactly) is based on the theory of the truncated K-moment problem.

#### Representing measures

Suppose we have a measure  $\mu$  as above:

$$eta_i = \int_{\mathbb{R}^n} x^i d\mu(x) \; (|i| \leq m), \; ext{supp} \; \mu \subseteq K$$

 $\mu$  is a *K-representing measure* for  $\beta$ .

For  $K = \mathbb{R}^n$ ,  $\mu$  is a representing measure.

 $\mu$  is a *finitely atomic* K-representing measure if

$$\mu = \sum_{i=1}^k \rho_i \delta_{w_i} \ (\rho_i > 0, w_i \in K).$$

**Question** If there is a K-representing measure, is there a finitely atomic K-representing measure?

A theorem of V. Tchakaloff [1957] provides an affirmative answer for K compact. The complete answer was found 50 years later:

Theorem [C. Bayer and J. Teichmann, 2006]

If  $\beta^{(m)}$  has a K-representing measure, then  $\beta$  has a finitely-atomic K-representing measure  $\mu$ , with card supp  $\mu \leq \dim \mathbb{R}_m[x_1,\ldots,x_n]$ 

#### The Full K-Moment Problem

$$\beta \equiv \beta^{(\infty)} = \{\beta_i : i \in \mathbb{Z}_+^n\}$$

Stieltjes [1894] 
$$K = [0, +\infty)$$

Hamburger [1920] 
$$K = \mathbb{R}$$

Hausdorff [1923] 
$$K = [a, b]$$

Results for FMP suggest results for TMP.

#### Connection between TMP and FMP

## Theorem [Jan Stochel, 2001]

The full multisequence  $\beta \equiv \beta^{(\infty)}$  has a K-representing measure if and only if  $\beta^{(m)}$  has a K-representing measure for every  $m \ge 1$ .

In some cases (one of which is illustrated below), we can use solutions to TMP, together with Stochel's theorem, to solve FMP.

#### **Riesz functional**

$$\mathcal{P} := \mathbb{R}[x_1, \ldots, x_n]$$

$$\beta \equiv \beta^{(\infty)} = \{\beta_i : i \in \mathbb{Z}_+^n\}$$

Riesz functional:  $L_{\beta}: \mathcal{P} \longmapsto \mathbb{R}$ 

$$p \equiv \sum a_i x^i \longmapsto L_{\beta}(\sum a_i x^i) = \sum a_i \beta_i \ (= \int_K p(x) d\mu(x))$$

Note: If  $\beta$  has a K-rep. measure  $\mu$ , then  $L_{\beta}$  is K-positive, i.e.,

$$p|K \geq 0 \Longrightarrow L_{\beta}(p) \geq 0.$$

(For  $K = \mathbb{R}^n$ , we say  $L_{\beta}$  is *positive*.)



#### "Abstract" solution of the Full K-Moment Problem

**Theorem** [M. Riesz, 1923 (n = 1), E.K. Haviland, 1936 ( $n \ge 2$ )]

The full multisequence  $\beta \equiv \beta^{(\infty)}$  has a K-representing measure if and only if  $L_{\beta}$  is K-positive, i.e.,

$$p \in \mathcal{P}, \ p|K \geq 0 \Longrightarrow L_{\beta}(p) \geq 0$$
.

#### A limitation of Riesz-Haviland:

For a general closed set K (even for  $\mathbb{R}^2$ ), there is no concrete structure theorem for K-positive polynomials, so it is difficult to check that  $L_\beta$  is K-positive.

#### **Moment matrices**

Given  $\beta \equiv \beta^{(\infty)}$ , we define the moment matrix  $M \equiv M_{\infty}(\beta)$ :

$$M_{\infty}(\beta) = (\beta_{i+j})_{(i,j)\in\mathbb{Z}_+^n\times\mathbb{Z}_+^n}.$$

 $M_{\infty}(\beta)$  is uniquely determined by

$$\langle M_{\infty}(\beta)\hat{p},\hat{q}\rangle = L_{\beta}(pq) \ \forall p,q \in \mathcal{P},$$

where  $\hat{s}$  denotes the coefficient vector of  $s \in \mathcal{P}$  relative to the basis of monomials in degree-lexicographic order.

If  $L_{\beta}$  is positive (in particular, if  $\beta$  has a representing measure), then  $M_{\infty}(\beta) \succeq 0$  (positive semidefinite):

$$\langle M_{\infty}(\beta)\hat{p},\hat{p}
angle = L_{\beta}(p^2) = \int p^2 d\mu \geq 0.$$

Summary so far: rep. meas.  $\iff L_{\beta}$  pos.  $\implies M \succeq 0$ .

#### **Sums of squares**

There is one situation where the "concrete" condition  $M \succeq 0$  readily implies that  $L_{\beta}$  is positive. Consider the following property:  $(H_{n,d})$  Every  $p \in \mathcal{P}_d$  with  $p | \mathbb{R}^n \geq 0$  can be expressed as  $p = \sum p_i^2$ . If  $(H_{n,d})$  holds and  $M \succeq 0$ , then  $L_{\beta}$  is positive:

$$L_{\beta}(p) = L_{\beta}(\sum p_i^2) = \sum \langle M \hat{p}_i, \hat{p}_i \rangle \geq 0.$$

Hilbert's theorem on sums of squares [D. Hilbert, 1888]

$$(H_{n,d})$$
 holds  $\iff$   $n = 1$ , or  $(n,d) = (2,4)$ , or  $d = 2$ .

The moment problem can be solved concretely in the positive cases of Hilbert's theorem; we will discuss the first two cases in the sequel.

We consider FMP in the first case of Hilbert's theorem, n = 1.

## Theorem [Hamburger, 1920]

Let n=1,  $K=\mathbb{R}$ . The full multisequence  $\beta\equiv\beta^{(\infty)}$  has a representing measure if and only if  $M_{\infty}(\beta)\succeq 0$ .

#### Proof.

For 
$$p \in \mathbb{R}[x]$$
,  $p|\mathbb{R} \geq 0 \Longrightarrow p = r^2 + s^2$  for some  $r, s \in \mathbb{R}[x]$ .  
Then  $L_{\beta}(p) = L_{\beta}(r^2) + L_{\beta}(s^2) = \langle M\hat{r}, \hat{r} \rangle + \langle M\hat{s}, \hat{s} \rangle \geq 0$ .  
Apply Riesz' Theorem.

## Conditions for solving TMP (with R. Curto) K-positivity in TMP

$$\beta \equiv \beta^{(m)}$$

$$\mathcal{P}_k := \{ p \in \mathcal{P} : deg \ p \le k \}$$

Riesz functional:  $L_{\beta}: \mathcal{P}_m \longrightarrow \mathbb{R}$ 

$$p \equiv \sum a_i x^i \longmapsto L_{\beta}(\sum a_i x^i) = \sum a_i \beta_i \ (= \int_{\mathbb{R}^n} p(x) d\mu(x))$$

If  $\beta$  has a K-representing measure, then  $L_{\beta}$  is K-positive, i.e.,

$$p \in \mathcal{P}_m, \ p|K \geq 0 \Longrightarrow L_{\beta}(p) \geq 0$$
.



If  $\beta \equiv \beta^{(m)}$  has a K-representing measure, then  $L_{\beta}$  is K-positive.

Tchakaloff's Thm. implies that the converse is true for K compact. For the noncompact case, we introduce an example.

For n=1,  $K=\mathbb{R}$ ,  $\beta\equiv\beta^{(4)}$ , consider the moment problem with  $\beta_0=\beta_1=\beta_2=\beta_3=1,\ \beta_4=2.$ 

We will show below that  $L_{\beta}$  is positive.

Is there a representing measure?

Question What is the analogue of R-H for TMP?



#### Moment matrices for TMP

For  $\beta \equiv \beta^{(2d)}$ , we define the *d*-th order moment matrix  $M_d(\beta)$ :

$$M_d(\beta) = (\beta_{i+j})_{(i,j) \in \mathbb{Z}_+^n \times \mathbb{Z}_+^n : |i|, |j| \le d}.$$

 $M_d(\beta)$  is uniquely determined by

$$\langle M_d(y)\hat{p},\hat{q}\rangle = L_{\beta}(pq) \ \forall p,q \in \mathcal{P}_d.$$

## Positivity condition for TMP

#### **Necessary condition 1: positivity**

If  $\beta$  has a representing measure, then  $M_d(\beta) \succeq 0$ :

$$\langle M_d(\beta)\hat{p},\hat{p}\rangle = L_\beta(p^2) = \int p^2 d\mu \geq 0$$

In the preceding example, we have

$$M_2(\beta) \equiv \left( \begin{array}{ccc} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 2 \end{array} \right).$$

 $M(2) \succeq 0$  and therefore (by sos)  $L_{\beta}$  is positive. Does  $\beta$  have a representing measure?

#### Recursiveness

Columns of 
$$M_d(\beta)$$
: 1,  $X_1, \ldots, X_n, \ldots, X_1^d, \ldots, X_n^d$ 

A dependence relation in *Col*  $M_d(\beta)$  can be denoted by  $p(X) \equiv p(X_1, \dots, X_n) = 0$  for some  $p \in \mathcal{P}_d$ .

#### **Necessary condition 2: recursiveness**

 $M_d(\beta)$  is recursively generated:

$$p(X) = 0 \Longrightarrow (pq)(X) = 0$$
 whenever  $pq \in \mathcal{P}_d$ .

In the example, X = 1, but  $X^2 \neq X$ , so  $L_{\beta}$  is positive, but there is no measure. The direct analogue of R-H fails for TMP.



#### An analogue of the Riesz-Haviland theorem for TMP

## Theorem [C-F, 2008]

Let  $\beta \equiv \beta^{(2d)}$  or  $\beta \equiv \beta^{(2d+1)}$ .  $\beta$  has a K-representing measure  $\iff L_{\beta}$  admits a K-positive extension  $L_{\tilde{\beta}}: \mathcal{P}_{2d+2} \longrightarrow \mathbb{R}$ .

Issue: In general it is very difficult to establish K-positivity.

## Example of a concrete truncated moment theorem

We return to the first case of Hilbert's theorem, n = 1.

## **Theorem** [C-F, 1991]

Let n=1.  $\beta \equiv \beta^{(2d)}$  has a representing measure  $\iff M_d(\beta) \succeq 0$  and  $M_d(\beta)$  is recursively generated.

## Using TMP to solve FMP

## Another proof of Hamburger's Theorem

#### **Theorem**

Let n=1. The full multisequence  $\beta \equiv \beta^{(\infty)}$  has a representing measure if and only if  $M_{\infty}(\beta) \succeq 0$ .

#### Proof.

Suppose  $M_{\infty} \succeq 0$ . Then, for each d,  $M_d(\beta)$  is positive semidefinite and recursively generated, so  $\beta^{(2d)}$  has a representing measure. Stochel's theorem now implies that  $\beta^{(\infty)}$  has a representing measure.

#### The variety of a moment matrix

The *variety* of  $\beta \equiv \beta^{(2d)}$  (or of  $M_d(\beta)$ ):

$$\mathcal{V}(\beta) = \bigcap_{p \in \mathcal{P}_d, p(\mathbf{X}) = \mathbf{0}} \mathcal{Z}(p),$$

where 
$$\mathcal{Z}(p) = \{x \in \mathbb{R}^n : p(x) = 0\}.$$

#### The variety condition

#### **Proposition**

If  $\mu$  is a representing measure, then supp  $\mu \subseteq \mathcal{V}(\beta)$  and

rank 
$$M_d(\beta) \leq card$$
 supp  $\mu \leq card$   $V(\beta)$ .

**Necessary condition 3: variety condition** 

$$r \equiv rank \ M_d(\beta) \le v \equiv card \ \mathcal{V}(\beta).$$

In the previous example, r = 2, X = 1,  $V(\beta) = \{1\}$ , v = 1.

#### The flat extension theorem

Recall: If  $\mu$  is a representing measure, then

$$r \equiv rank \ M_d(\beta) \leq card \ supp \ \mu.$$

## **Theorem** [C-F, 1996, 2005]

 $\beta \equiv \beta^{(2n)}$  has an r-atomic representing measure  $\iff$   $M_d(\beta) \succeq 0$  and  $M_d(\beta)$  has a *flat*, i.e., rank-preserving, moment matrix extension  $M_{d+1}$ . In this case, an r-atomic representing measure  $\mu$  can be explicitly constructed with  $supp \ \mu = \mathcal{V}(M_{d+1})$ .

#### Solution to TMP based on moment matrix extensions

## **Theorem** [C-F, 2005]

 $\beta \equiv \beta^{(2d)}$  has a representing measure  $\iff M_d(\beta)$  admits a positive extension  $M_{d+k}$  (for some  $k \geq 0$ ), and  $M_{d+k}$  has a flat extension  $M_{d+k+1}$ .

Note: When the strategy of this theorem can be implemented, this method circumvents the difficulty of positivity for  $L_{\beta^{(2d+2)}}$  in the truncated R-H theorem. In this case, is there some way to recognize directly that for  $M_{d+1}$  (as above),  $L_{\beta^{(2d+2)}}$  is positive?

## TMP for K a planar curve of degree 2

## Theorem [C-F, 2005]

Let n=2 and suppose p(X)=0 in  $M_d(\beta)$  for some  $p\in\mathcal{P}_2$ . Then  $\beta^{(2d)}$  has a representing measure (necessarily supported in  $\mathcal{Z}_p)\Longleftrightarrow M_d(\beta)$  is positive and recursively generated, and rank  $M_d(\beta)\leq card$   $\mathcal{V}_\beta$ . In this case, either  $M_d(\beta)$  has a flat extension  $M_{d+1}$ , or  $M_d(\beta)$  has a positive extension  $M_{d+1}$ , which in turn has a flat extension  $M_{d+2}$ .

#### Note

- (i) This result solves the bivariate quartic moment problem  $(n=2, \beta \equiv \beta^{(4)})$  in the case when  $M_2(\beta)$  is singular. For the case when  $M_2(\beta) \succ 0$  we will use alternate methods based on approximation and convexity.
- (ii) The above result does not extend to  $y = x^3$  [F, 2008]

## Approximation methods (with Jiawang Nie)

Let 
$$\eta = \dim \mathcal{P}_{2d}$$
, so  $\beta^{(2d)} \in \mathbb{R}^{\eta}$ .

$$\mathcal{R}_{n,d} := \{ \beta \in \mathbb{R}^{\eta} : \beta \text{ has a } K - representing measure} \}$$
, convex cone

$$\mathcal{S}_{n,d} := \{ \beta \in \mathbb{R}^{\eta} : L_{\beta} \text{ is } K - positive \}, \text{ convex cone }$$

Theorem [F-Nie, 2009]

$$S_{n,d} = \overline{\mathcal{R}_{n,d}}$$
.

#### Strict K-positivity and representing measures

 $L_{\beta}$  is *strictly K-positive* if

$$p \in \mathcal{P}_{2d}, \ p|K \geq 0, \ p|K \not\equiv 0 \Longrightarrow L_{\beta}(p) > 0.$$

*K* is *determining* if  $p \in \mathcal{P}_{2d}$ ,  $p|K \equiv 0 \Longrightarrow p \equiv 0$ .

Theorem [F-Nie, 2009]

If K is determining and  $L_{\beta}$  is strictly K-positive, then  $\beta$  has a K-representing measure.

#### Proof.

The hypotheses imply that

$$\beta \in interior(\mathcal{S}_{n,d}) = interior(closure(\mathcal{R}_{n,d}))$$

= interior(
$$\mathcal{R}_{n,d}$$
)  $\subseteq \mathcal{R}_{n,d}$ .



## Bivariate quartic moment problem

Consider the second case of Hilbert's theorem, when n=2, d=4, and consider the corresponding moment problem for  $\beta^{(4)}$ . For  $M_2(\beta)$  singular, the problem was solved by [Curto-F, 2005] (above).

## Theorem [F-Jiawang Nie, 2009]

Let n=2. If  $M_2(\beta) \succ 0$ , then  $\beta$  has a representing measure.

#### Proof.

Let  $K=\mathbb{R}^2$ , determining. Since  $M_2(\beta)\succ 0$ , Hilbert's theorem implies that  $L_\beta$  is strictly K-positive: If  $p|\mathbb{R}^2\geq 0$ ,  $p\not\equiv 0$ , then  $p=\sum p_i^2$  (with some  $p_i\not\equiv 0$ ), so  $L_\beta(p)=\sum \langle M_2(\beta)\hat{p}_i,\hat{p}_i\rangle>0$ . Apply the previous theorem.

## Lasserre's method for polynomial optimization

For simplicity, we consider the polynomial optimization problem for  $K_Q=\mathbb{R}^n$ , i.e.,  $Q=\{q_0\equiv 1\}$ . Let  $p\in\mathbb{R}[x_1,\ldots,x_n]$ . For  $2t\geq deg\ p$ , the t-th Lasserre "moment relaxation" for  $p_*\equiv \inf_{x\in\mathbb{R}^n}p(x)$  is defined by

$$p_t := \inf\{L_{\beta}(p) : \beta \equiv \beta^{(2t)}, \ \beta_0 = 1, \ M_t(\beta) \succeq 0\}.$$

Then  $p_t \leq p_*$ , and for  $t' \geq t$ ,  $p_{t'} \geq p_t$ ; thus, $\{p_t\}$  is convergent, and  $p^{mom} \equiv \lim_{t \to \infty} p_t \leq p_*$ . In general, for fixed t,  $p_t$  is not necessarily attained at any  $\beta$ . Assuming that the infimum is attained, at some optimal sequence  $\beta \equiv \beta^{\{t\}}$ , we seek criteria so that  $L_{\beta}(p) = p_*$ , so that we have finite convergence of  $\{p_s\}$  at stage t.

## Lasserre's stopping criterion

Assume at stage t that  $\beta \equiv \beta^{\{t\}}$  has a representing measure  $\mu$ . Then

$$p_*=p_*eta_0=p_*\int 1d\mu\leq\int pd\mu=L_eta(p)=p_t\leq p_*,$$

so we have convergence at stage t. Although the existence of a representing measure for  $\beta^{\{t\}}$  is difficult to ascertain in general, Lasserre focuses on the easy-to-check case when  $M_t(\beta)$  is flat, i.e.,  $rank\ M_t(\beta) = rank\ M_{t-1}(\beta)$ . In this case,  $\beta$  has a  $rank\ M_t$ -atomic representing measure, and the atoms are the global minimizers for p.

Can we find a more general, but still concrete, stopping criterion?

## A more general stopping criterion

## Theorem [LF-Jiawang Nie, 2010]

Let  $\beta \equiv \beta^{\{t\}}$ . If  $L_{\beta}$  is positive, then  $p_t = p_*$ .

Of course, in general, positivity for  $L_{\beta}$  is very difficult to check. In current work we are studying a class for which positivity is clear. Let  $\mathcal{F}_d := \{\beta \equiv \beta^{(2d)} : M_d(y) \succeq 0 \text{ is flat} \}$  (a subset of  $\mathbb{R}^{\rho}$ , where  $\rho \equiv \rho_{2d} = \dim \mathcal{P}_{2d}$ ). Consider  $\overline{\mathcal{F}_d}$ , the closure. If  $\beta = \lim_{k \to \infty} \beta^{[k]}$ ,

with each  $M_d(\beta^{[k]})$  positive and flat, then each  $L_{\beta^{[k]}}$  is positive, so  $L_\beta$  is positive. Thus,  $M_d(\beta)\succeq 0$ , and

$$\operatorname{rank} M_d(\beta) \leq \liminf_{k \to \infty} \operatorname{rank} M_d(\beta^{[k]} = \liminf_{k \to \infty} \operatorname{rank} M_{d-1}(\beta^{[k]} \leq \rho_{d-1}.$$

If  $M_d(\beta) \succeq 0$  and rank  $M_d(\beta) \leq \rho_{d-1}$ , does  $\beta$  belong to  $\overline{\mathcal{F}_d}$ ?



#### On limits of positive flat moment matrices

Theorem [F-Nie, 2010]

Let n=1, or d=1, or n=d=2. If  $M_d(\beta) \succeq 0$  and rank  $M_d(\beta) \leq \rho_{d-1}$ , then  $\beta \in \overline{\mathcal{F}_d}$ .