

# The moment-LP and moment-SOS approaches in polynomial optimization

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We discuss the optimization problem  $\mathbf{P} : \inf \{f(\mathbf{x}) : \mathbf{x} \in \mathbf{K}\}$  where  $f$  is a polynomial and  $\mathbf{K} \subset \mathbb{R}^n$  is the basic closed semi-algebraic set (assumed to be compact)

$$\mathbf{K} := \{\mathbf{x} \in \mathbb{R}^n : g_j(\mathbf{x}) \geq 0, \quad j = 1, \dots, m\}$$

for some polynomial  $g_j$ ,  $j = 1, \dots, m$ . When one focuses on the *global* minimum  $f^*$  (as opposed to a local optimum), problem  $\mathbf{P}$  can be written as:

$$f^* = \sup \{\lambda : f(\mathbf{x}) - \lambda \geq 0, \quad \forall \mathbf{x} \in \mathbf{K}\}.$$

When  $f$  is a polynomial and  $\mathbf{K}$  is a compact basic semi-algebraic set, powerful positivity certificates of Real Algebraic Geometry allow to express the difficult positivity constraint “ $f(\mathbf{x}) \geq 0$  for all  $\mathbf{x} \in \mathbf{K}$ ” in a way that can be exploited for efficient numerical computation. Indeed one may then define a hierarchy of *convex relaxations*  $(\mathbf{P}_k)$ ,  $k \in \mathbb{N}$ , of  $\mathbf{P}$  which provides a monotone sequence of upper bounds ( $f_k^* \geq f^*$ ) such that  $f_k^* \rightarrow f^*$  as  $k \rightarrow \infty$ . Depending on which type of positivity certificate (e.g. one due to Krivine, Handelman and Vasilescu or one due to Schmüdgen and Putinar) one ends up with solving a hierarchy of either *LP-relaxations* or *SDP-relaxations* (or semidefinite relaxations). In both cases the resulting convex relaxation  $\mathbf{P}_k$  becomes more and more difficult to solve as its size increases with  $k$ .

We then discuss the relative merits and drawbacks of both hierarchies of LP- and SDP-relaxations and their impact not only in optimization (and particularly combinatorial optimization) but also in many areas for solving instances of the so-called *Generalized Problem of Moments* (GMP) with polynomial data (of which Global Polynomial Optimization is in fact the simplest instance).

Finally, we also introduce another characterization of nonnegativity on a closed set  $\mathbf{K} \subset \mathbb{R}^n$  which can be also exploited to now define a monotone non increasing sequence of upper bounds  $(f_k^*)$ ,  $k \in \mathbb{N}$ , that converges to the global minimum  $f^*$  of  $\mathbf{P}$  as  $k \rightarrow \infty$ . Computing each upper bound  $f_k^*$  now boils down to solving a generalized eigenvalue problem associated with some pair of real symmetric matrices whose size increases with  $k$ .

## References

- [1] J.B. Lasserre, *Moments, Positive Polynomials and Their Applications*, Imperial College Press, London, 2010.
- [2] J.B. Lasserre, *An introduction to Polynomial and Semi-Algebraic Optimization*, Cambridge University Press, Cambridge, in Press.
- [3] J.B. Lasserre, *Global optimization with polynomials and the problem of moments*, SIAM J. Optim. **11** (2001), 796–817.