

## Robust Univariate Optimization: stronger results in 1D?

My interest is intrigued by a recent result on the famous S-Lemma. It is shown in [1] that in one dimension the S-Lemma can be strengthened to have a stronger form. Namely, the lemma is stated as follows, which says under some regularity condition, the S-Lemma holds for any univariate polynomials (with arbitrary degrees). This is in contradiction to multivariate polynomials where we know a similar result only holds for quadratic functions.

**Lemma 1.** *Assume  $\gcd(g, g') = 1$  (regularity condition). Then the following two statements are equivalent:*

- (i)  $\forall x \in \mathbb{R}, f(x) \geq 0$  implies  $\forall x \in \mathbb{R}, g(x) \geq 0$ ;
- (ii) *There exists nonnegative polynomials  $h_1(x) \geq 0, h_2(x) \geq 0$  and  $h_3(x) \geq 0$  such that  $h_1(x)f(x) = h_2(x)g(x) + h_3(x)$ .*

Moreover, if we have the decomposition  $f(x) = f_0(x)\prod_{i=1}^I f_i(x)$  and  $g(x) = g_0(x)\prod_{j=1}^J g_j(x)$ , where  $f_i(x)$  and  $g_j(x)$  have degree  $\leq 2$  ( $\forall i, j$ ) and denote  $\deg f_0(x) = 2s$  and  $\deg g_0(x) = 2r$  then  $\deg h_1(x) = 2s + 2\lceil \frac{\deg f(x) - 2r}{2} \rceil \left( \frac{\deg g(x) - 2s}{2} - 1 \right)$ ,  $\deg h_2(x) = 2r + 2\lceil \frac{\deg g(x) - 2s}{2} \rceil \left( \frac{\deg f(x) - 2r}{2} - 1 \right)$ .

It is well known S-lemma is useful in deriving robust/adaptive counterparts for mathematical programs, so it is then natural to think we may have stronger results in the univariate case than the high-dimensional case. And this is often true. For example, in [2], the authors are able to characterize exactly the optimal probability bounds in the univariate case (can be solved via semidefinite programming) while they show in high dimension such a characterization is NP-hard. The key in their univariate construction is the usage of the linear matrix inequalities (LMI) for nonnegative univariate polynomials.

I believe combining LMI and the univariate S-Lemma might bring about new insights into the robust univariate optimization problems, where similar ideas have been adopted by [3] to derive new counterparts for data-driven chance constrained problems. During the project, I would attempt to explore in depth on this issue. Not sure if it would be a serious research attempt in the future, but at least I think it will be much fun to put my efforts into it.

## References

- [1] Simai He, Man Hong Wong, and Shuzhong Zhang. “On the S-Lemma for Univariate Polynomials.”
- [2] Dimitris Bertsimas, and Ioana Popescu. “Optimal inequalities in probability theory: A convex optimization approach.” SIAM Journal on Optimization 15.3 (2005): 780-804.
- [3] Ruiwei Jiang, and Yongpei Guan. “Data-driven chance constrained stochastic program.” Mathematical Programming 158.1-2 (2016): 291-327.