

Polynomial optimization using MOSEK and Julia

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- Julia package for polynomial optimization (requires Julia 0.4).
- Implements the Lasserre hierarchy of moment relaxations.
- Uses the MOSEK conic optimizer to solve the relaxations.

Installation

Pkg.clone("https://github.com/MOSEK/Polyopt.jl.git")

We consider polynomials optimization problems

minimize
$$f(x)$$

subject to $g_i(x) \ge 0$, $i = 1, ..., m$
 $h_i(x) = 0$, $i = 1, ..., l$
 $x \in \mathbb{R}^n$

for real polynomials $f, g_i, h_j : \mathbb{R}^n \mapsto \mathbb{R}$.

- Solved by a sequence of relaxations.
- An important recent application of semidefinite optimization.
- The relaxations can be difficult to solve numerically.

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- GloptiPoly, standard moment relaxations.
- **SparsePoP**, sparse moment relaxations.
- **SOSTools**, general sum-of-squares problems.
- Yalmip, general sums-of-squares and polynomial optimization.

- Test and improve the MOSEK semidefinite solver.
- Have full control of the generated semidefinite problems.
- Investigate other approaches for exploiting sparsity.
- Implement it in Julia to remove dependency on Matlab.

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• Standard moment relaxation:

minimize
$$p^T y$$

subject to $y_0 = 1$
 $M_k(y) \succeq 0$
 $M_{k-d_{g_j}}(g_j y) \succeq 0, j = 1, \dots, m$
 $M_{k-d_{h_i}}(h_i y) = 0, i = 1, \dots, l.$

• Dual problem (which we feed into MOSEK):

maximize
$$t$$

subject to $\sum_{j=1}^{m} A_0^j \bullet X^j + \sum_{k=1}^{l} B_0^k \bullet Z^k = p_0 - t$
 $\sum_{j=1}^{m} A_j^j \bullet X^j + \sum_{k=1}^{l} B_i^k \bullet Z^k = p_i, \quad i = 1, \dots, r$
 $X^j \succeq 0, Z^k$ are free symmetric matrices.

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How to solve a simple example in Julia

$$\begin{array}{ll} \text{minimize} & -x_1 - x_2 \\ \text{subject to} & 2x_1^4 - 8x_1^3 + 8x_1^2 - x_2 + 2 \ge 0 \\ & 4x_1^4 - 32x_1^3 + 88x_1^2 - 96x_1 - x_2 + 36 \ge 0 \\ & 0 \le x_1 \le 3, \quad 0 \le x_2 \le 4. \end{array}$$

More examples on Github...

Concluding remarks

Package overview:

- Lasserre's hierarchy of moment relaxations in Julia.
- Correlative sparsity and chordal relaxations by Waki et al.
- No solution extracting method by Henrion and Lasserre; perturb problem to extract a single global optimizer.

Modern features of Julia facilitate lean implementation:

- polynomial.jl 258 lines of code.
- cliques.jl 66 lines of code.
- Polyopt.jl 268 lines of code.
- solver_mosek.jl 134 lines of code.

Important for improving conic solver in upcoming MOSEK 8.0.

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