

Introduction

What is SeDuMi?

- Optimization over symmetric cones
 - linear, second order, semidefinite
 - complex variables
- Interior point method
 - primal-dual predictor-corrector scheme
 - self dual embedding
- Open source: GPL, written in Matlab and C
- Modelling languages:
 - YALMIP (Johan Löfberg)
 - CVX (Michael C. Grant)
 - Gloptipoly (Didier Henrion)
 - SOSTools (Stephen Prajna et al.)
- Widely used: both industry and academics
- Advantages: very high numerical accuracy, robustness, efficient sparse system handling, mixed second-order/semidefinite problems, Matlab
- Weaknesses: large dense problems, memory requirements, Matlab

History

late 1997: Jos F. Sturm starts SeDuMi

summer 1998: SeDuMi 1.0

November 2002: SeDuMi 1.05R5 (last version by Jos)

November 2003: Jos dies

October 2004: AdvOL at McMaster takes over

June 2005: SeDuMi 1.1 (new version)

October 2006: SeDuMi 1.1R2 (latest version)

Theoretical background

- Primal-dual conic optimization

$$\begin{array}{ll} \min c^T x & \max b^T y \\ Ax = b & A^T y + s = c \\ x \in \mathcal{K} & s \in \mathcal{K} \end{array} \quad (\text{CO})$$

where $x, c, s \in \mathbb{R}^n$, $A \in \mathbb{R}^{m \times n}$, $b, y \in \mathbb{R}^m$ and $\mathcal{K} \subset \mathbb{R}^m$ is a closed, convex, pointed, solid, self-dual cone

- Primal-dual path-following method
- Predictor-corrector scheme
- Self-dual embedding
- Complexity: $\mathcal{O}(\sqrt{n} \log \frac{n}{\epsilon})$ to find an ϵ -optimal solution

Modelling languages

GAMS, AMPL, AIMMS, LINDO, etc.

- None of the major modelling languages supports SDP
- User input required

YALMIP

- Very advanced and developed
 - More general problems (SDP, NLP, MILP), 30+ solvers
 - Automatic dualization, robustification, convexity analysis
- Widely used for optimal control problems

Lyapunov stability:

$$\begin{array}{l} P \succeq 0 \\ A^T P + PA \preceq 0 \end{array}$$

```
A = [-1 2 0; -3 -4 1; 0 0 -2];
P = sdpvar(3,3);
F = set(P>=0)+set(A'*P+P*A<=0);
F = F + set(trace(P) == 1);
solvesdp(F);
```

CVX

- Disciplined Convex Programming
- Least squares, LP, QP, GP, NLP is planned
- Currently uses only SeDuMi, more solvers will be added

Norm optimization:

$$\begin{array}{l} \min \|Ax - b\|_2 \\ Cx = d \\ \|x\|_\infty \leq 0.4 \end{array}$$

```
m = 20; n = 10; p = 4;
A = randn(m,n); b = randn(m,1);
C = randn(p,n); d = randn(p,1);
cvx.begin
variable x(n)
minimize( norm(A*x-b, 2) )
subject to
C * x == d;
norm( x, Inf ) <= 0.4;
cvx.end
```

Gloptipoly

- Polynomial systems, nonconvex problems
- Relaxations, bounds, global optimality detection

$$\begin{array}{l} \max x_1^2 + x_2^2 \\ 2x_1^2 + 3x_2^2 + 2x_1x_2 \leq 1 \\ 3x_1^2 + 2x_2^2 - 4x_1x_2 \leq 1 \end{array}$$

```
P1 = [0 0 -1; 0 0 0; -1 0 0];
P2 = [1 0 -3; 0 -2 0; -2 0 0];
P3 = [1 0 -2; 0 4 0; -3 0 0];
gloptipoly(P);
```

SOSTOOLS

- Sum-of-squares optimization
- Univariate polynomial minimization
- SDP relaxation for semialgebraic problems

SOS representation for:

$$p(x) = 4x^4y^6 + x^2 - xy^2 + y^2$$

```
syms x y;
p = 4*x^4*y^6+x^2-x*y^2+y^2;
[Q,Z]=findsos(p,'rational');
```

Examples are taken from the official websites.

Current projects

Parallelization/Improved linear algebra

- BLAS/LAPACK/ScaLAPACK
 - ACML, Intel MKL, ATLAS
 - matrix products/factorizations
 - automatic parallelization
 - linear speedup
- OpenMP
 - compilers: Pathscale, IBM, Intel, GCC (from 4.2)
 - forming the normal equation
 - very good speedup
- Better linear algebra
 - more efficient sparse/dense handling

Preprocessing

- Even elementary techniques work (identical/opposite sign/fixed variables, redundant constraints)
- Finding block-diagonal structure
- Decomposing narrow-band matrices
- Conic decomposition (Kojima et. al, Plaza Martínez/Krishnan, Young/Anjos)

Adaptive techniques

- Optimal parameter selection
 - tuning on a small problem, testing on the large one
 - significant savings within a problem group
- Change parameters during the iterations (corrector type, update method, neighbourhood parameters, step differentiation)
 - Online learning methods, reinforcement learning
- Starting point selection

Advanced infeasibility detection

- Early identification of infeasibility to save time
 - modelling errors
 - unrealistic designs
- Change the algorithm if infeasibility is suspected
- Full infeasibility analysis
 - Detect contradicting constraints
 - Handle weak infeasibility

Future plans

Reimplementation

- Limitations of Matlab
 - Price
 - Difficult to embed in another application
 - Memory limitations
- Our goal is to make/keep SeDuMi
 - free
 - platform-independent
 - efficient
 - modular
 - embeddable
- Candidate languages:
 - C/C++
 - Python (with SciPy and NumPy)
 - Octave

Bibliography

- [1] A. Ben-Tal and A. Nemirovski. *Lectures on Modern Convex Optimization: Analysis, Algorithms, and Engineering Applications*. MPS-SIAM Series on Optimization. SIAM, Philadelphia, PA, 2001.
- [2] M. Grant, S. Boyd, and Y. Ye. Disciplined convex programming. In Leo Liberti and Nelson Maculan, editors, *Global Optimization: From Theory to Implementation*, Nonconvex Optimization and its Applications. Kluwer, Dordrecht, 2005.
- [3] M. Grant, S. Boyd, and Y. Ye. *CVX Users' Guide*, 2007.
- [4] D. Henrion and J. B. Lasserre. GloptiPoly: Global optimization over polynomials with Matlab and SeDuMi. *ACM Transactions on Mathematical Software*, 29(2):165–194, June 2003.
- [5] J. Löfberg. YALMIP: A toolbox for modeling and optimization in MATLAB. In *Proceedings of the CACSD Conference*, Taipei, Taiwan, 2004.
- [6] I. Pólik and T. Terlaky. Detecting infeasibility in conic optimization: Theory and practice. Technical report, McMaster University, Advanced Optimization Lab, 2007. In preparation.
- [7] S. Prajna, A. Papachristodoulou, P. Seiler, and P. A. Parrilo. *SOSTOOLS: Sum of squares optimization toolbox for MATLAB*, 2004.
- [8] J. F. Sturm. *Primal-Dual Interior Point Approach to Semidefinite Programming*. Phd thesis, Tinbergen Institute Research Series vol. 156, Tilburg University, 1997.
- [9] J.F. Sturm. Using SeDuMi 1.02, a MATLAB toolbox for optimization over symmetric cones. *Optimization Methods and Software*, 11–12:625–653, 1999. Special issue on Interior Point Methods (CD supplement with software).