PENLIB/SDP User's Guide

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In this guide we give a description of parameters of function sdp, solving linear semidefinite programming problems with linear constraints. This function is part of the library PENLIB.

We solve the dual linear SDP problem with linear constraints:

$$\min_{x \in \mathbb{R}^n} \sum_{i=1}^n f^i x^i$$
s.t.
$$\sum_{k=1}^n b_i^k x^k \le c_i, \qquad i = 1, \dots, m_\ell$$

$$A_i^0 + \sum_{k=1}^n A_i^k x^k \le 0 \qquad i = 1, \dots, m.$$

The matrix constraints can be written as one constraint with block diagonal matrices as follows:

$$\begin{pmatrix} A_0^1 & & & \\ & A_0^2 & & \\ & & \ddots & \\ & & A_0^m \end{pmatrix} + \begin{pmatrix} A_1^1 & & \\ & A_1^2 & & \\ & & \ddots & \\ & & & A_1^m \end{pmatrix} x_1 \\ + \begin{pmatrix} A_1^1 & & & \\ & A_2^2 & & \\ & & A_2^2 & & \\ & & & A_2^m \end{pmatrix} x_2 + \dots + \begin{pmatrix} A_n^1 & & & \\ & A_n^2 & & \\ & & & A_n^m \end{pmatrix} x_n$$

Here we use the abbreviations n := vars, $m_{\ell} := \text{constr}$, and m := mconstr.

The input parameters are explained below. We assume that the linear constraint vectors b_i can be sparse, so we give them in standard sparse format. Similarly, we assume that the matrix constraints data can be sparse. Here we distinguish two cases: Some (many) of the matrices A_k^i for the k-th constraint can be empty; so we only give those matrices (for each matrix constraint) that are nonempty. Further, each of the nonempty matrices A_k^i can still be sparse; hence we give the matrices A_k^i in sparse format (value, column index, row index). Further, as all the matrices are symmetric, we only give the upper triangle.

The function sdp is declared as

(1) vars	number of variables
integer number	
(2) constr	number of linear constraints
integer number	
(3) mconstr	number of matrix constraints (diagonal blocks in each A_k)
integer number	indiano of matrix constraints (diagonal stocks in each 11 _k)
(4) msizes	sizes of the diagonal blocks $A_k^1, A_k^2, \dots, A_k^{\texttt{mconstr}}$
integer array	length: mconstr
(5) fx	on exit: objective function value
double array	length: 1
(6) x0	on entry: initial guess for the solution
(0) ==0	on exit: solution vector x
	(Not referenced, if $x0 = 0$ on entry)
double array	length: vars
(7) uoutput	on exit: linear multipliers u_i $(i = 1,, constr)$ followed by upper
	triangular parts of matrix multipliers U^{j} $(j = 1,, mconstr)$ in
	rowise storage format
	(Not referenced, if uoutput = 0 on entry)
double array	length: constr + msizes(1)*(msizes(1)+1)/2 +
	msizes(2)*(msizes(2)+1)/2 ++
(0) 6 1 :	msizes(mconstr)*(msizes(mconstr)+1)/2
(8) fobj	objective vector f in full format
double array	length: vars
(9) ci	right-hand side vector of the linear constraint c in full format
double array	length: constr
(10) bi_dim	number of nonzeros in vector b_i for each linear constraint
integer array	length: constr
(11) bi_idx	indices of nonzeros in each vector b_i $length: bi_dim(1)+bi_dim(2)++bi_dim(constr)$
integer array	~
(12) bi_val double array	nonzero values in each vector b_i corresponding to indices in bi_idx length: bi_dim(1)+bi_dim(2)++bi_dim(constr)
(13) ai_dim	number of nonzero blocks $A_0^i, A_1^i, \ldots, A_{\mathtt{vars}}^i$ for each matrix constraint $i=1,2,\ldots,\mathtt{mconstr}$
integer array	length: mconstr
(14) ai_idx	indices of nonzero blocks for each matrix constraint
integer array	length: ai_dim(1)+ai_dim(2)++ai_dim(mconstr)
(15) ai_nzs	number of nonzero values in each nonzero block
(10) 413125	$A^i_{\mathtt{ai_idx}(1)}, A^i_{\mathtt{ai_idx}(2)}, \ldots, A^i_{\mathtt{ai_idx}(\mathtt{ai_dim}(\mathtt{i}))}$ for each matrix con-
	${ m straint}\ i=1,2,\ldots,{ m mconstr}$
integer array	<pre>length: ai_dim(1)+ai_dim(2)++ai_dim(mconstr)</pre>
(16) ai_val	nonzero values in the upper triangle of each nonzero block
	$A^{i}_{\mathtt{ai_idx}(1)}, A^{i}_{\mathtt{ai_idx}(2)}, \dots, A^{i}_{\mathtt{ai_idx}(\mathtt{ai_dim}(i))}$ for each matrix con-
	straint $i = 1, 2, \ldots, mconstr$
double array	<pre>length: ai_nzs(1)+ai_nzs(2)++ai_nzs(length(ai_nzs))</pre>
(17) ai_col	column indices of nonzero values in the upper triangle of each
	nonzero block $A^{i}_{\mathtt{ai_idx}(1)}, A^{i}_{\mathtt{ai_idx}(2)}, \dots, A^{i}_{\mathtt{ai_idx}(\mathtt{ai_dim}(\mathtt{i}))}$ for each
	matrix constraint $i = 1, 2, \dots, mconstr$
integer array	length: ai_nzs(1)+ai_nzs(2)++ai_nzs(length(ai_nzs))
(18) ai_row	row indices of nonzero values in the upper triangle of each
	nonzero block $A^{i}_{\mathtt{ai_idx}(1)}, A^{i}_{\mathtt{ai_idx}(2)}, \dots, A^{i}_{\mathtt{ai_idx}(\mathtt{ai_dim}(i))}$ for each
integer array	matrix constraint $i = 1, 2, \dots, mconstr$ $longth; oi mgg(1) + oi mgg(2) + oi mgg(1) + oi $
integer array	<pre>length: ai_nzs(1)+ai_nzs(2)++ai_nzs(length(ai_nzs))</pre>

(19) ioptions	integer valued options (see below)
integer array	length: 8
(20) foptions	real valued options (see below)
double array	length: 7
(21) iresults	on exit: integer valued output information (see below)
	(Not referenced, if iresults = 0 on entry)
integer array	length: 4
(22) fresults	on exit: real valued output information (see below)
	(Not referenced, if fresults = 0 on entry)
double array	length: 3
(23) info	on exit: error flag (see below)
integer array	length: 1

IOPTIONS	name/value	meaning	default
ioption(0)	DEF		
	0	use default values for all options	
	1	use user defined values	
ioption(1)	PBM_MAX_ITER	maximum number of iterations of the overall	50
		algorithm	
ioption(2)	UM_MAX_ITER	maximum number of iterations for the uncon-	100
		strained minimization	
ioption(3)	OUTPUT	output level	1
	0	no output	
	1	summary output	
	2	brief output	
	3	full output	
ioption(4)	DENSE	check density of the Hessian	0
	0	automatic check. For very large problems with	
		dense Hessian, this may lead to memory diffi-	
		culties.	
	1	dense Hessian assumed	
ioption(5)	LS	linesearch in unconstrained minimization	0
	0	do not use linesearch	
	1	use linesearch	
ioption(6)	XOUT	write solution vector x in the output file	0
	0	no	
	1	yes	
ioption(7)	UOUT	write computed multipliers in the output file	0
	0	no	
	1	yes	

FOPTIONS	name/value	meaning	default
foption(0)	U0	scaling factor for linear constraints; must be positive	1.0
foption(1)	MU	restriction for multiplier update for linear constraints	0.7
foption(2)	MU2	restriction for multiplier update for matrix con-	0.1
		straints	
foption(3)	PBM_EPS	stopping criterium for the overall algorithm	1.0e-7
foption(4)	P_EPS	lower bound for the penalty parameters	1.0e-6
foption(5)	UMIN	lower bound for the multipliers	1.0e-14
foption(6)	ALPHA	stopping criterium for unconstrained minimization	1.0e-2

IRESULTS	meaning
iresults(0)	number of outer iterations
<pre>iresults(1)</pre>	number of Newton steps
iresults(2)	number of linesearch steps
iresults(3)	elapsed time in seconds

FRESULTS	meaning
fresults(0)	relative precision at x_{opt}
fresults(1)	feasibility of linear inequalities at x_{opt}
fresults(2)	feasibility of matrix inequalities at x_{opt}

INFO	meaning
info(0)	no errors occured
info(1)	no progress in objective value, problem probably infeasible.
info(2)	Cholesky factorization of Hessian failed. The result still may be useful.
info(3)	Maximum iteration limit exceeded. The result still may be useful.
info(4)	Linesearch failed. The result still may be useful.
info(5)	Wrong input parameters (ioptions, foptions).
info(6)	Memory error.
info(7)	Unknown error, please contact PENOPT GbR (contact@penopt.com).

Example 1. Let $f = (1, 2, 3)^T$. Assume that we have no linear constraints. Assume further that we have two matrix inequality constraints, first of size (3x3), second of size (2x2). The first constraint contains full matrices, the second one diagonal matrices:

$$\left(\begin{array}{cc} A_0^1 & \\ & A_0^2 \end{array}\right) + \left(\begin{array}{cc} A_1^1 & \\ & A_1^2 \end{array}\right) x_1 + \left(\begin{array}{cc} A_2^1 & \\ & A_2^2 \end{array}\right) x_2 + \left(\begin{array}{cc} A_3^1 & \\ & A_3^2 \end{array}\right) x_3$$

The blocks A_k^1 have then 6 nonzero elements, block A_k^2 only two nonzero elements (recall that we only give elements of the upper triangular matrix). In this case

```
{\tt vars}\,=3
 constr = 0
mconstr = 2
 msizes = (3,2)
        x0 = (0.0,0.0,0.0) (for example)
    fobj = (1.0, 2.0, 3.0)
        ci = (0.0)
  bi_dim = (0)
  bi_idx = (0)
  bi\_val = (0.0)
  ai\_dim = (4,4)
  ai\_idx = (0,1,2,3,0,1,2,3)
  ai\_nzs = (6,6,6,6,2,2,2,2)
  \begin{array}{ll} \mathtt{ai\_val} \ = (A_0^1(1), \dots, A_0^1(6), \ A_1^1(1), \dots, A_1^1(6), \dots \dots, \\ A_0^2(1), A_0^2(2), A_1^2(1), A_1^2(2), A_2^2(1), A_2^2(2), A_3^2(1), A_3^2(2)) \end{array}
  ai\_col = (0,1,2,1,2,2,0,1,2,1,2,2,...,0,1,0,1,0,1,0,1)
  \mathtt{ai\_row} = (0,0,0,1,1,2,\,0,0,0,1,1,2,\ldots,\,0,1,0,1,0,1,0,1)
```

Example 2. Let again $f = (1, 2, 3)^T$. We have two linear constraints with

$$b_1 = (0,0,1)^T$$
, $b_2 = (5,6,0)^T$, $c = (3,-3)^T$

Assume further that we have two matrix inequality constraints, first of size (3x3), second of size (2x2). The first constraint contains sparse matrices, the second one diagonal matrices. Some of the matrices are empty, as shown below:

$$\begin{pmatrix}
0 & & & & \\
& 0 & & & \\
& & & 0 & & \\
& & & & 0
\end{pmatrix} + \begin{pmatrix}
2 & -1 & 0 & & & \\
2 & 0 & & & & \\
& & 2 & & & \\
& & & & 1
\end{pmatrix} x_1$$

$$+ \begin{pmatrix}
0 & & & & \\
& 0 & & & \\
& & & 0 & & \\
& & & & 3
\end{pmatrix} x_2 + \begin{pmatrix}
2 & 0 & -1 & & \\
& 2 & 0 & & \\
& & & 2 & & \\
& & & & 2
\end{pmatrix} x_3$$

In this case

$$vars = 3$$

$$constr = 2$$

$$mconstr = 2$$

$$msizes = (3,2)$$

$$x0 = (0.0, 0.0, 0.0)$$
 (for example)

$$fobj = (1.0, 2.0, 3.0)$$

$$ci = (3.0, -3.0)$$

$$\mathtt{bi_dim}\,=(1,\!2)$$

$$bi_idx = (2,0,1)$$

$$bi_val = (1.0, 5.0, 6.0)$$

$$ai_dim = (2,3)$$

$$ai_idx = (1,3,0,1,2)$$

$$ai_nzs = (4,4,1,2,2)$$

$$ai_val = (2.0,-1.0,2.0,2.0,2.0,-1.0,2.0,2.0,1.0,1.0,-1.0,3.0,-3.0)$$

$$ai_col = (0,1,1,2,0,2,1,2,1,0,1,0,1)$$

$$ai_row = (0,0,1,2,0,0,1,2,1,0,1,0,1)$$