

Simple SDP Example using CVX

This script demonstrates formulating and solving SDP problems in CVX.

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Requirements

This script requires the CVX toolbox: <http://cvxr.com/cvx/>

Problem Data in Standard Form

```
c = [1 -1];
A1 = [1 0; 0 1];
A2 = [0 1; 1 -1];
B = [1 0; 0 0];
```

Primal Problem in CVX

```
% Initialize CVX environment
cvx_begin sdp % Specify that matrix inequalities are semidefinite constraints

    % Decision variables
    variable z(2,1) % Vector decision variable

    % Objective
    minimize c*z

    % Semidefinite Constraint
    z(1)*A1 + z(2)*A2 >= B;
cvx_end
```

Calling SDPT3 4.0: 3 variables, 1 equality constraints

```
num. of constraints = 1
dim. of sdp    var = 2,    num. of sdp blk = 1
```

SDPT3: Infeasible path-following algorithms

```
version predcorr gam expon scale_data
HKM      1      0.000  1      0
it pstep dstep pinfeas dinfeas gap      prim-obj      dual-obj      cputime
-----
0|0.000|0.000|5.0e-01|6.1e+00|2.0e+02| 1.000000e+01  0.000000e+00| 0:0:00| chol  1  1
1|1.000|1.000|2.5e-07|6.4e-02|1.0e+01| 8.074566e+00 -5.519999e-01| 0:0:00| chol  1  1
2|0.979|1.000|6.6e-08|6.4e-03|4.2e-01|-1.168985e-01 -5.240860e-01| 0:0:00| chol  1  1
3|1.000|0.926|2.4e-08|1.1e-03|5.5e-02|-1.592323e-01 -2.125979e-01| 0:0:00| chol  1  1
4|0.987|0.989|6.6e-09|7.4e-05|7.0e-04|-1.994838e-01 -2.000637e-01| 0:0:00| chol  1  1
5|0.989|0.989|1.6e-10|7.1e-06|7.7e-06|-1.999943e-01 -1.999908e-01| 0:0:00| chol  1  1
6|0.989|0.989|1.6e-11|7.8e-08|8.4e-08|-1.999999e-01 -1.999999e-01| 0:0:00| chol  1  1
7|0.996|1.000|1.6e-13|3.2e-12|1.2e-09|-2.000000e-01 -2.000000e-01| 0:0:00|
stop: max(relative gap, infeasibilities) < 1.49e-08
```

```
-----
number of iterations      = 7
primal objective value    = -1.99999999e-01
dual  objective value     = -2.00000000e-01
gap := trace(XZ)          = 1.16e-09
relative gap              = 8.30e-10
actual relative gap       = 8.26e-10
rel. primal infeas        = 1.55e-13
rel. dual  infeas         = 3.21e-12
norm(X), norm(y), norm(Z) = 1.0e+00, 2.0e-01, 1.0e+00
norm(A), norm(b), norm(C) = 2.6e+00, 2.0e+00, 2.2e+00
Total CPU time (secs)     = 0.10
CPU time per iteration    = 0.01
termination code          = 0
DIMACS: 1.6e-13  0.0e+00  3.6e-12  0.0e+00  8.3e-10  8.3e-10
-----
```

```
-----
Status: Solved
Optimal value (cvx_optval): +0.8
```

Primal optimal value returned by CVX

cvx_optval

z

cvx_optval =

8.0000e-01

z =

1.2000e+00
4.0000e-01

Dual Problem in CVX

```
% Initialize CVX environment
cvx_begin sdp % Specify that matrix inequalities are semidefinite constraints

% Decision variables
variable Y(2,2) % Matrix decision variable

% Objective
maximize trace(B*Y)

% Linear Equality Constraints
trace(A1*Y) == c(1);
trace(A2*Y) == c(2);

% Semidefinite Constraint
Y >= 0;
cvx_end
```

Warning: This linear matrix inequality appears to be unsymmetric. This is very likely an error that will produce unexpected results. Please check the LMI; and, if necessary, re-enter the model.

Calling SDPT3 4.0: 3 variables, 1 equality constraints
For improved efficiency, SDPT3 is solving the dual problem.

```
-----
num. of constraints = 1
dim. of sdp      var = 2,   num. of sdp blk = 1
*****
SDPT3: Infeasible path-following algorithms
*****
version predcorr gam expon scale_data
HKM      1      0.000 1      0
it pstep dstep pinfeas dinfeas gap      prim-obj      dual-obj      cputime
-----
0|0.000|0.000|5.0e-01|6.1e+00|2.0e+02| 1.000000e+01  0.000000e+00| 0:0:00| chol 1 1
1|1.000|1.000|2.5e-07|6.4e-02|1.0e+01| 8.074566e+00 -5.519999e-01| 0:0:00| chol 1 1
2|0.979|1.000|6.6e-08|6.4e-03|4.2e-01|-1.168985e-01 -5.240860e-01| 0:0:00| chol 1 1
3|1.000|0.926|2.4e-08|1.1e-03|5.5e-02|-1.592323e-01 -2.125979e-01| 0:0:00| chol 1 1
4|0.987|0.989|6.6e-09|7.4e-05|7.0e-04|-1.994838e-01 -2.000637e-01| 0:0:00| chol 1 1
5|0.989|0.989|1.6e-10|7.1e-06|7.7e-06|-1.999943e-01 -1.999908e-01| 0:0:00| chol 1 1
6|0.989|0.989|1.6e-11|7.8e-08|8.4e-08|-1.999999e-01 -1.999999e-01| 0:0:00| chol 1 1
7|0.996|1.000|1.6e-13|3.2e-12|1.2e-09|-2.000000e-01 -2.000000e-01| 0:0:00|
stop: max(relative gap, infeasibilities) < 1.49e-08
-----
number of iterations = 7
primal objective value = -1.99999999e-01
dual objective value = -2.00000000e-01
gap := trace(XZ) = 1.16e-09
relative gap = 8.30e-10
actual relative gap = 8.26e-10
rel. primal infeas = 1.55e-13
rel. dual infeas = 3.21e-12
norm(X), norm(y), norm(Z) = 1.0e+00, 2.0e-01, 1.0e+00
```

```

norm(A), norm(b), norm(C) = 2.6e+00, 2.0e+00, 2.2e+00
Total CPU time (secs) = 0.08
CPU time per iteration = 0.01
termination code      = 0
DIMACS: 1.6e-13  0.0e+00  3.6e-12  0.0e+00  8.3e-10  8.3e-10
-----
-----

```

```

Status: Solved
Optimal value (cvx_optval): +0.8

```

Dual optimal value returned by CVX

Note that the dual optimal value is equal to the primal optimal value

```

cvx_optval
Y

```

```

cvx_optval =

    8.0000e-01

Y =

    8.0000e-01   -4.0000e-01
   -4.0000e-01    2.0000e-01

```

CVX can also return the primal and dual optimal variables

```

% Initialize CVX environment
% The quiet key word can be used to suppress the solvers output
cvx_begin sdp quiet

    % Decision variable
    variable z(2,1)

    % Dual variable
    % The dimension of the dual variable will be inferred from the
    % corresponding constraint so it is not necessary to specify the
    % dimension
    dual variable Y

    % Objective
    minimize c*z

    % Semidefinite Constraint with corresponding dual variable Y
    Y : z(1)*A1 + z(2)*A2 >= B;
cvx_end

```

```
cvx_status
```

```
cvx_status =
```

```
Solved
```

Primal and Dual Optimal Variables

Note the primal and dual optimal variables are the same as those found previously.

```
z  
Y
```

```
z =
```

```
1.2000e+00  
4.0000e-01
```

```
Y =
```

```
8.0000e-01 -4.0000e-01  
-4.0000e-01 2.0000e-01
```

Attribution

This example supplements the book "Networks of Dissipative Systems: Compositional Certification of Stability, Performance, and Safety" by Murat Arcak, Chris Meissen, and Andrew Packard.