Disciplined Convex Programming and CVX

- convex optimization solvers
- modeling systems
- disciplined convex programming
- CVX

Convex optimization solvers

LP solvers

lots available (GLPK, Excel, Matlab's linprog, . . .)

cone solvers

- typically handle (combinations of) LP, SOCP, SDP cones
- several available (SDPT3, SeDuMi, CSDP, . . .)

general convex solvers

- some available (CVXOPT, MOSEK, . . .)
- plus lots of special purpose or application specific solvers
- could write your own

(we'll study, and write, solvers later in the quarter)

Transforming problems to standard form

- you've seen lots of tricks for transforming a problem into an equivalent one that has a standard form (e.g., LP, SDP)
- these tricks greatly extend the applicability of standard solvers
- writing code to carry out this transformation is often painful
- modeling systems can partly automate this step

Modeling systems

a typical modeling system

- automates most of the transformation to standard form; supports
 - declaring optimization variables
 - describing the objective function
 - describing the constraints
 - choosing (and configuring) the solver
- when given a problem instance, calls the solver
- interprets and returns the solver's status (optimal, infeasible, . . .)
- (when solved) transforms the solution back to original form

Some current modeling systems

- AMPL & GAMS (proprietary)
 - developed in the 1980s, still widely used in traditional OR
 - no support for convex optimization
- YALMIP ('Yet Another LMI Parser')
 - first matlab-based object-oriented modeling system with special support for convex optimization
 - can use many different solvers; can handle some nonconvex problems
- CVXMOD/CVXOPT (in alpha)
 - python based, completely GPLed
 - cone and custom solvers
- CVX
 - matlab based, GPL, uses SDPT3/SeDuMi

Disciplined convex programming

- describe objective and constraints using expressions formed from
 - a set of basic atoms (convex, concave functions)
 - a restricted set of operations or rules (that preserve convexity)
- modeling system keeps track of affine, convex, concave expressions
- rules ensure that
 - expressions recognized as convex (concave) are convex (concave)
 - but, some convex (concave) expressions are not recognized as convex (concave)
- problems described using DCP are convex by construction

CVX

- uses DCP
- runs in Matlab, between the cvx_begin and cvx_end commands
- relies on SDPT3 or SeDuMi (LP/SOCP/SDP) solvers
- refer to user guide, online help for more info
- the CVX example library has more than a hundred examples

Example: Constrained norm minimization

```
A = randn(5, 3);
b = randn(5, 1);
cvx_begin
    variable x(3);
    minimize(norm(A*x - b, 1))
    subject to
        -0.5 <= x;
        x <= 0.3;
cvx_end</pre>
```

- between cvx_begin and cvx_end, x is a CVX variable
- statement subject to does nothing, but can be added for readability
- inequalities are intepreted elementwise

What CVX does

after cvx_end, CVX

- transforms problem into an LP
- calls solver SDPT3
- overwrites (object) x with (numeric) optimal value
- assigns problem optimal value to cvx_optval
- assigns problem status (which here is Solved) to cvx_status

(had problem been infeasible, cvx_status would be Infeasible and x would be NaN)

Variables and affine expressions

• declare variables with variable name [(dims)] [attributes]

```
- variable x(3);
- variable C(4,3);
- variable S(3,3) symmetric;
- variable D(3,3) diagonal;
- variables y z;
```

• form affine expressions

```
- A = randn(4, 3);
- variables x(3) y(4);
- 3*x + 4
- A*x - y
- x(2:3)
- sum(x)
```

Some functions

function	meaning	attributes
norm(x, p)	$ x _p$	CVX
square(x)	$\int x^2$	CVX
square_pos(x)	$(x_{+})^{2}$	cvx, nondecr
pos(x)	x_{+}	cvx, nondecr
<pre>sum_largest(x,k)</pre>	$x_{[1]} + \dots + x_{[k]}$	cvx, nondecr
sqrt(x)	$\sqrt{x} (x \ge 0)$	ccv, nondecr
<pre>inv_pos(x)</pre>	1/x (x > 0)	cvx, nonincr
max(x)	$\max\{x_1,\ldots,x_n\}$	cvx, nondecr
<pre>quad_over_lin(x,y)</pre>	$x^2/y (y > 0)$	cvx, nonincr in y
<pre>lambda_max(X)</pre>	$\lambda_{\max}(X)$ $(X = X^T)$	CVX
huber(x)	$\begin{cases} x^2, & x \le 1 \\ 2 x - 1, & x > 1 \end{cases}$	CVX

Composition rules

- \bullet can combine atoms using valid composition rules, e.g.:
 - a convex function of an affine function is convex
 - the negative of a convex function is concave
 - a convex, nondecreasing function of a convex function is convex
 - a concave, nondecreasing function of a concave function is concave
- for convex h, $h(g_1, \ldots, g_k)$ is recognized as convex if, for each i,
 - $-g_i$ is affine, or
 - $-g_i$ is convex and h is nondecreasing in its ith arg, or
 - g_i is concave and h is nonincreasing in its ith arg
- for concave h, $h(g_1, \ldots, g_k)$ is recognized as concave if, for each i,
 - $-g_i$ is affine, or
 - g_i is convex and h is nonincreasing in ith arg, or
 - g_i is concave and h is nondecreasing in ith arg

Valid (recognized) examples

u, v, x, y are scalar variables; X is a symmetric 3×3 variable

convex:

```
- norm(A*x - y) + 0.1*norm(x, 1)
- quad_over_lin(u - v, 1 - square(v))
- lambda_max(2*X - 4*eye(3))
- norm(2*X - 3, 'fro')
```

• concave:

```
- \min(1 + 2*u, 1 - \max(2, v))
- sqrt(v) - 4.55*inv_pos(u - v)
```

Rejected examples

u, v, x, y are scalar variables

- neither convex nor concave:
 - square(x) square(y)
 - norm(A*x y) 0.1*norm(x, 1)
- rejected due to limited DCP ruleset:
 - sqrt(sum(square(x))) (is convex; could use norm(x))
 - square(1 + x^2) (is convex; could use square_pos(1 + x^2), or 1 + $2*pow_pos(x, 2) + pow_pos(x, 4)$)

Sets

- some constraints are more naturally expressed with convex sets
- sets in CVX work by creating unnamed variables constrained to the set
- examples:
 - semidefinite(n)
 - nonnegative(n)
 - simplex(n)
 - lorentz(n)
- semidefinite(n), say, returns an unnamed (symmetric matrix)
 variable that is constrained to be positive semidefinite

Using the semidefinite cone

variables: X (symmetric matrix), z (vector), t (scalar) constants: A and B (matrices)

- X == semidefinite(n)
 - means $X \in \mathbf{S}^n_+$ (or $X \succeq 0$)
- A*X*A' X == B*semidefinite(n)*B'
 - means $\exists~Z\succeq 0$ so that $AXA^T-X=BZB^T$
- [X z; z' t] == semidefinite(n+1)
 - $\text{ means } \left[\begin{array}{cc} X & z \\ z^T & t \end{array} \right] \succeq 0$

Objectives and constraints

• objective can be

- minimize(convex expression)
- maximize(concave expression)
- omitted (feasibility problem)

• constraints can be

- convex expression <= concave expression</pre>
- concave expression >= convex expression
- affine expression == affine expression
- omitted (unconstrained problem)

More involved example

Defining new functions

- can make a new function using existing atoms
- example: the convex deadzone function

$$f(x) = \max\{|x| - 1, 0\} = \begin{cases} 0, & |x| \le 1\\ x - 1, & x > 1\\ 1 - x, & x < -1 \end{cases}$$

• create a file deadzone.m with the code

function
$$y = deadzone(x)$$

 $y = max(abs(x) - 1, 0)$

deadzone makes sense both within and outside of CVX

Defining functions via incompletely specified problems

- suppose f_0, \ldots, f_m are convex in (x, z)
- ullet let $\phi(x)$ be optimal value of convex problem, with variable z and parameter x

minimize
$$f_0(x,z)$$
 subject to $f_i(x,z) \leq 0, \quad i=1,\ldots,m$ $A_1x+A_2z=b$

- ullet ϕ is a convex function
- problem above sometimes called *incompletely specified* since x isn't (yet) given
- an incompletely specified concave maximization problem defines a concave function

CVX functions via incompletely specified problems

```
implement in cvx with

function cvx_optval = phi(x)
cvx_begin
    variable z;
    minimize(f0(x, z))
    subject to
        f1(x, z) <= 0; ...
        A1*x + A2*z == b;
cvx_end</pre>
```

- function phi will work for numeric x (by solving the problem)
- function phi can also be used inside a CVX specification, wherever a convex function can be used

Simple example: Two element max

• create file max2.m containing

```
function cvx_optval = max2(x, y)
cvx_begin
   variable t;
   minimize(t)
   subject to
       x <= t;
       y <= t;
cvx_end</pre>
```

- the constraints define the epigraph of the max function
- could add logic to return max(x,y) when x, y are numeric (otherwise, an LP is solved to evaluate the max of two numbers!)

A more complex example

- $f(x) = x + x^{1.5} + x^{2.5}$, with $\operatorname{dom} f = \mathbf{R}_+$, is a convex, monotone increasing function
- its inverse $g = f^{-1}$ is concave, monotone increasing, with $\operatorname{dom} g = \mathbf{R}_+$
- there is no closed form expression for g
- g(y) is optimal value of problem

maximize
$$t$$
 subject to $t_+ + t_+^{1.5} + t_+^{2.5} \le y$

(for y < 0, this problem is infeasible, so optimal value is $-\infty$)

• implement as

```
function cvx_optval = g(y)
cvx_begin
    variable t;
    maximize(t)
    subject to
        pos(t) + pow_pos(t, 1.5) + pow_pos(t, 2.5) <= y;
cvx_end</pre>
```

• use it as an ordinary function, as in g(14.3), or within CVX as a concave function:

```
cvx_begin
   variables x y;
   minimize(quad_over_lin(x, y) + 4*x + 5*y)
   subject to
      g(x) + 2*g(y) >= 2;
cvx_end
```

Example

- ullet optimal value of LP, $f(c) = \inf\{c^Tx \mid Ax \leq b\}$, is concave function of c
- by duality (assuming feasibility of $Ax \leq b$) we have

$$f(c) = \sup\{-\lambda^T b \mid A^T \lambda + c = 0, \ \lambda \succeq 0\}$$

• define f in CVX as

```
function cvx_optval = lp_opt_val(A,b,c)
cvx_begin
    variable lambda(length(b));
    maximize(-lambda'**b);
    subject to
        A'*lambda + c == 0; lambda >= 0;
cvx_end
```

• in lp_opt_val(A,b,c) A, b must be constant; c can be affine expression

CVX hints/warnings

- watch out for = (assignment) versus == (equality constraint)
- X >= 0, with matrix X, is an elementwise inequality
- X >= semidefinite(n) means: X is elementwise larger than some positive semidefinite matrix (which is likely not what you want)
- writing subject to is unnecessary (but can look nicer)
- make sure you include brackets around objective functions
 - yes: minimize(c'*x)
 - no: minimize c'*x
- double inequalities like 0 <= x <= 1 don't work;
 use 0 <= x; x <= 1 instead
- log, exp, entropy-type functions not yet implemented in CVX