MATH-GA.2012.001 Selected Topics in Numerical Analysis: Convex and Nonsmooth Optimization, Spring 2020 Homework Assignment 2
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1. Prove that a function is convex if and only if its epigraph is a convex set. Suppose f is a convex function,  $f: \mathbf{R}^n \to \mathbf{R}$  then  $\forall (x, t_1), (y, t_2) \in \mathbf{epi}f$ , and  $\forall \theta \in [0, 1]$ , we want to show that  $\theta(x, t_1) + (1 - \theta)(y, t_2)$  is in  $\mathbf{epi}f$ . we have:

$$f(\theta x + (1 - \theta)y) \le \theta f(x) + (1 - \theta)f(y)$$
  
$$< \theta t_1 + (1 - \theta)t_2$$

thus **epi** f is convex. The other direction is similar  $\forall (x,t_1), (y,t_2) \in \mathbf{epi} f$ , **epi** f is a convex set, and  $\forall \theta \in [0,1]$ : Let  $t_1 = f(x), t_2 = f(y)$  thus  $\theta(x,t_1) + (1-\theta)(y,t_2) = (\theta x + (1-\theta)y, \theta t_1 + (1-\theta)t_2)$  is in **epi** f which implies:  $f(\theta x + (1-\theta)y) \leq \theta t_1 + (1-\theta)t_2 \Rightarrow f(\theta x + (1-\theta)y) \leq \theta f(x) + (1-\theta)f(y) \Rightarrow f$  is convex.

- 2. BV Ex. 2.31 Properties of dual cones. Let  $K^*$  be the dual cone of a convex cone K. Prove the following.
  - (a)  $K^*$  is indeed a convex cone.  $\forall y_1, y_2 \in K^*, \theta_1, \theta_2 \geq 0$ , and  $\forall x \in K$ ,  $x^T(\theta_1y_1 + \theta_2y_2) = \theta_1x^Ty_1 + \theta_2x^Ty_2 \geq 0$  thus  $K^*$  is a convex cone.
  - (b)  $K_1 \subseteq K_2$  implies  $K_2^* \subseteq K_1^*$ . Suppose  $y \in K_2^*$ ,  $\forall x \in K_1$ ,  $x^Ty \ge 0$ , and since  $x \in K_2$  also, then  $y \in K_1^*$  and  $K_2^* \subseteq K_1^*$ .
- 3. Show that if a convex cone K is closed, then  $(K^*)^*$ , the dual cone of the dual cone of K, is equal to K. source: class notes on the web pointing that  $(K^*)^*$  can be seen as the intersection of halfspaces.

Let K a convex closed cone, then  $(K^*)^* = \{x^*|y^Tx^* \geq 0, \forall y \in K^*\}$ . We can consider  $(K^*)^*$  as the intersection of halfspaces  $H_{x^* \in K} = \{y^Tx^* \geq 0, \forall y \in K^*\}$ . If  $x \in K$  then  $\forall y \in K^*, x^Ty = y^Tx \geq 0 \Rightarrow x \in (K^*)^*$ . K being convex and closed by the corollary of the separating hyperplanes,  $K = (K^*)^*$ .

- 4. BV Ex. 233 Find the dual cone of  $\{A \ x | x \geq 0\}$ , where  $A \in \mathbf{R}^{m \times n}$ . The dual of  $K = \{A \ x | x \geq 0\}$  is  $K^* = \{y | (Ax)^T y \geq 0, \forall x \geq 0\}$  or  $K^* = \{y | x^T (A^T y) \geq 0, x \geq 0\} = \{y | (A^T y)^T x \geq 0, x \geq 0\}$ . Given  $u = A^T y$ , we are looking for vectors u such that the inner product is non-negative for any  $x \geq 0$ . Let  $\{e_1, \cdots, e_n\}$  the canonical basis for  $\mathbf{R}^n$ , for any vector  $u = A^T y, y \in K^*$ , we have  $u^T e_i \geq 0 \Rightarrow u_i \geq 0, i \in [1, n]$ . Thus  $K^* = \{y | A^T y \geq 0, x \geq 0\}$ , this is sufficient as if  $x \geq 0$  then  $x^T A^T y \geq 0$ .
- 5. Show that the second-order cone defined on p.31 of BV is self-dual, that is, it satisfies  $K^* = K$ . Let C the second-order cone,  $C = \{(x,t) \in \mathbf{R}^n | \|x\|_2 \le t\}$ .  $C^* = \{(y,s) | \begin{bmatrix} x \\ t \end{bmatrix}^T \begin{bmatrix} y \\ s \end{bmatrix} \ge 0, \forall (x,t) \in C\}. \text{ if } (y,s) \in C \text{ then } x^Ty \le \|x\|_2 \|y\|_2$

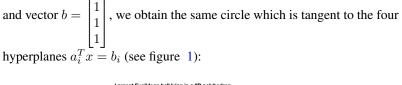
using Cauchy-Schwarz or  $x^Ty \leq t$  s.  $\begin{bmatrix} x \\ t \end{bmatrix}^T \begin{bmatrix} y \\ s \end{bmatrix} = x^Ty + ts$ , and by the triangle inequality,  $\|x^Ty + ts\| \geq t$  s  $-|x^Ty| \geq 0 \Rightarrow y \in C^*$ . Suppose  $(y,s) \notin C$ , then  $\|y\|_2 > s$  and let m the index of the largest component of y, thus  $\|y\|_2 = (\sum_{i=1,n} y_i^2)^{\frac{1}{2}} \leq (n^2|y_m|^2)^{\frac{1}{2}} = n|y_m| \Rightarrow$ . WLOG  $|y_m| = y_m$ , then  $y_m > \frac{n}{s^2}$  and let x the vector with the only component non-zero  $x_m = -\frac{n}{s^2}$  then  $x^Ty = -\frac{n}{s^2}$   $y_m \leq -1$  so  $y \notin C^*$ . In conclusion,  $C = C^*$ , C is self-dual.

## 6. "Chebyshev center" problem

(a) function chebyshev\_center(A, b) takes a matrix A of dimension (2, n) and a vector b(n) to find the largest Euclidean ball that lies in a polyhedron described by n linear inequalities.

```
% Compute the Chebyshev center of a polyhedron
% Boyd & Vandenberghe "Convex Optimization"
function [x_sol, r_sol] = chebyshev_center(A, b)
% The goal is to find the largest Euclidean ball (i.e. its center and
% radius) that lies in a polyhedron described by linear inequalites in t
% fashion: P = \{x : a_i' * x \le b_i, i=1,...,m\}
% Generate the data
[\tilde{n}, n] = size(A);
% Build and execute model
fprintf(1, 'Computing Chebyshev center...');
cvx begin
    variable r(1)
    variable x_c(2)
    maximize ( r )
    for k=1:n
        A(:, k) \times x + r \times norm(A(:, k), 2) \le b(k);
    end
cvx end
fprintf(1, 'Done! \n');
x_sol = x_c;
r_sol = r;
% Display results
fprintf(1,'The Chebyshev center coordinates are: \n');
disp(x_c);
fprintf(1,'The radius of the largest Euclidean ball is: \n');
disp(r);
% Generate the figure
x = linspace(-2, 2);
for k=1:n
```

```
plot (x, -x * A(1,k)./A(2,k) + b(k)./A(2,k),"b-");
     hold on
end
theta = 0:pi/100:2*pi;
plot(x_c(1) + r*cos(theta), x_c(2) + r*sin(theta), "r");
plot (x_c(1), x_c(2), b*');
xlabel("x_1")
ylabel("x_2")
txt = "# inequalities:" + num2str(n);
title({"Largest Euclidean ball lying in a 2D polyhedron", txt});
text(x_c(1), x_c(2), " \setminus leftarrow center")
axis([-1 \ 1 \ -1 \ 1])
axis equal
hold off
txt = "chebyshev_center_" + num2str(n);
saveas(gcf,txt,'epsc')
For the same example on the web page where matrix A = \begin{bmatrix} 2 & 2 & -1 & -1 \\ 1 & -1 & 2 & -2 \end{bmatrix}
and vector b = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, we obtain the same circle which is tangent to the four
```



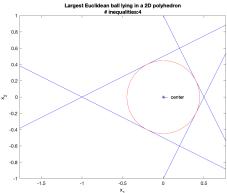


Figure 1: Sample example

We solve the same optimization problem with more inequalities and an interior center inside the polyhedron (see figure 2):

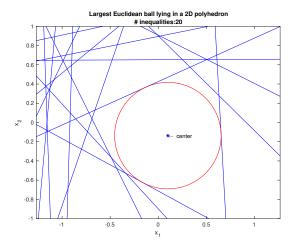


Figure 2: Example with 20 inequalities

- (b) If we choose A and b such that there is no interior point: if the hyperplanes intersect at the same point then CVX finds a center with radius zero, if they do not intersect then CVX will not find a solution.
- (c) The problem to solve now is to find the largest "scaled unit ball"  $\mathcal{B}=\{x_c+u|\|u\|_p\leq r\}$  that lies in the polyhedron described by a set of linear inequalities:  $\mathcal{P}=\{x\in\mathbf{R}^n|a_i^Tx\leq b_i,i=1,\cdots,m\}$ . For any point of  $\mathcal{B}$  lying in one halfspace  $a_i^Tx\leq b_i$ , similarly to the euclidean space, we have  $\|u\|_p\leq r\Rightarrow a_i^Tx_c+r\|a_i\|_p\leq b_i$  since  $g_i=\sup\{a_i^Tu\|u\|_p\leq r\}=r\|a_i\|_p$ . And the Chebyshev center can be determined by solving the problem:

maximize 
$$r$$
  
subject to  $a_i^T x_c + r ||a_i||_p \le b_i, i = 1, \dots, m$ 

This is still an LP problem since the inequalities are linear. We obtain different solutions for the same matrix A and vector b, corresponding to different p-norm, p=1 is the diamond shape ball (fig. 3), for p=1.5 the ball has a shape between the diamond and the circle p=2 (fig. 4), and as we increase p, for  $p=\infty$  the ball becomes a square (where either  $\|x\|=1$  and  $\|y\|\leq 1$  or  $\|x\|\leq 1$  and  $\|y\|=1$ , (fig. 5)).

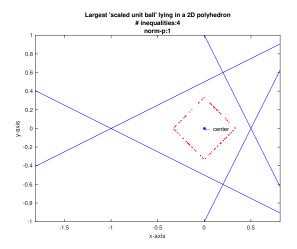


Figure 3: "Scaled" unit ball for p=1

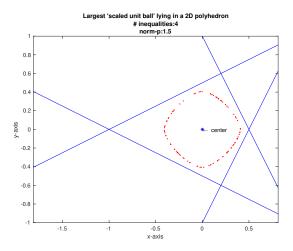


Figure 4: "Scaled" unit ball for p=1.5

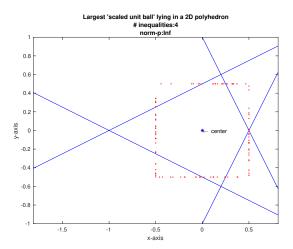


Figure 5: "Scaled" unit ball for  $p=\infty$ 

```
% Compute the Chebyshev center of a polyhedron
function [x_sol, r_sol] = chebyshev_center_with_norm(A, b, p)
% The goal is to find the largest Euclidean ball (i.e. its center and
% radius) that lies in a polyhedron described by linear inequalites in t
% fashion: P = \{x : a_i' * x <= b_i, i=1,...,m\}
rng('default')
format long g
[\tilde{n}, n] = size(A);
% Build and execute model
fprintf(1,'Computing Chebyshev center...');
cvx_begin
    variable r(1)
    variable x_c(2)
    maximize ( r )
    for k=1:n
        A(:, k)' * x_c + r * norm(A(:, k), p) \le b(k);
    end
cvx_end
fprintf(1, 'Done! \n');
x_sol = x_c;
r_sol = r;
% Display results
fprintf(1, "The Chebyshev center coordinates are: \n");
disp(x_c);
```

```
txt = "Radius of largest 'scaled unit ball' using norm p:" + num2str(p)
fprintf(1,txt);
disp(r);
% Generate the figure
x = linspace(-2,2);
for k=1:n
    plot (x, -x * A(1,k)./A(2,k) + b(k)./A(2,k),"b-");
    hold on
end
n_{vecs} = 100;
[x, y] = gen_random_vectors(n_vecs, p);
x = x \cdot * r + x \cdot c(1);
y = y.* r + x_c(2);
for i=1:n_vecs
    plot(x(i), y(i), "r.");
    hold on
end
plot(x_c(1), x_c(2), b*');
xlabel("x-axis")
ylabel("y-axis")
txt1 = "# inequalities:" + num2str(n);
tx2 = "norm-p:" + num2str(p);
title({"Largest 'scaled unit ball' lying in a 2D polyhedron", txt1, tx2}
text(x_c(1), x_c(2), "	ext{leftarrow center"})
axis([-1 \ 1 \ -1 \ 1])
axis equal
hold off
txt = "chebyshev_center_norm_" + num2str(p);
saveas(gcf,txt,'epsc')
function [x, y] = gen_random_vectors(n, p)
    r = randn(n, 2); % Use a large n
    for i=1:n
        norm_r = norm(r(i,:), p);
        r(i, :) = r(i, :) ./ norm_r;
    end
    x = r(:, 1);
    y = r(:, 2);
```

(d) As we decrease p, the p-norm of  $a_i$  grows exponentially (as shown in figure 6), thus CVX, to solve the problem, finds that the solution for the center is the intersection of the hyperplanes if it exists and r to be zero.

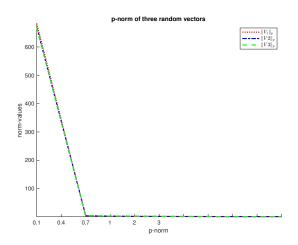


Figure 6: p-norms of three random vectors