

# Convex and Nonsmooth Optimization Homework 2

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1. Suppose a function  $f$  is convex. For two points  $(x_1, t_1), (x_2, t_2)$  in its epigraph, we have, for any  $\alpha \in [0, 1]$ ,

$$f(\alpha x_1 + (1 - \alpha)x_2) \leq \alpha f(x_1) + (1 - \alpha)f(x_2) \leq \alpha t_1 + (1 - \alpha)t_2.$$

Thus,  $(\alpha x_1 + (1 - \alpha)x_2, (\alpha t_1 + (1 - \alpha)t_2))$  is in the epigraph of  $f$ , which shows that the epigraph is a convex set.

Suppose the epigraph of a function  $f$  is a convex set. We take the conventional definition of a convex function here. Then, for any  $x_1, x_2 \in \mathbf{dom} f$ , we know that  $(x_1, f(x_1)), (x_2, f(x_2)) \in \mathbf{epi} f$ , and by convexity of the epigraph, for any  $\alpha \in [0, 1]$

$$f(\alpha x_1 + (1 - \alpha)x_2) \leq \alpha f(x_1) + (1 - \alpha)f(x_2).$$

We also need to show that  $\mathbf{dom} f$  is a convex set, which is true because  $\mathbf{epi} f = \{(x, t) | x \in \mathbf{dom} f, f(x) \leq t\}$  is a convex set, and  $\mathbf{dom} f$  is  $\mathbf{epi} f$  projected to the first coordinate. The projection is an affine transformation (and it is actually linear), and the image of a convex set under an affine function is convex, so  $\mathbf{dom} f$  is a convex set.

2. (a) Suppose  $y_1, y_2 \in K^*$ . For any  $t_1, t_2 \geq 0$ , we have, for any  $x \in K$ ,

$$x^T(t_1 y_1 + t_2 y_2) = t_1 x^T y_1 + t_2 x^T y_2 \geq 0,$$

so  $t_1 y_1 + t_2 y_2 \in K^*$  and  $K^*$  is a convex cone.

- (b) For any  $y \in K_2^*$ , we have

$$x^T y \geq 0, \forall x \in K_2.$$

Since  $K_1 \subseteq K_2$ ,  $x^T y \geq 0, \forall x \in K_1$ , which shows that  $y \in K_1^*$ . Since  $y$  is arbitrary in  $K_2^*$ , we have shown that  $K_2^* \subseteq K_1^*$ .

3. By the definition of the dual cone,  $y \in \mathbb{R}^m$  is in the dual cone if and only if

$$x^T A^T y \geq 0, \forall x \succeq 0. \tag{1}$$

We know that the cone  $\mathbb{R}_+^n$  is self-dual. Thus, (1) is satisfied if and only if  $A^T y \succeq 0$ . Therefore, the dual cone is

$$\{y \in \mathbb{R}^m | A^T y \succeq 0\}.$$

4. For any  $(x, t), (y, s) \in K$ , by Cauchy-Schwarz inequality,

$$|x^T y| \leq \|x\|_2 \|y\|_2 \leq ts \implies x^T y \geq -ts.$$

Thus,  $x^T y + ts \geq 0$ , and since  $(y, s)$  is an arbitrary element in  $K$ , we have  $(x, t) \in K^*$ , which implies that  $K \subseteq K^*$ .

On the other hand, suppose  $(x, t) \in K^*$  but  $(x, t) \notin K$ , then  $\|x\|_2 > t$ . Since (by Cauchy-Schwarz inequality)

$$\|x\|_2 = \sup_{z: \|z\|_2 \leq 1} x^T z,$$

there exists  $z$  such that

$$\|z\|_2 \leq 1, \quad x^T z > t.$$

Let  $y = -sz$  for some  $s > 0$ .  $\|y\|_2 \leq s$ , so  $(y, s) \in K$ , but

$$x^T y + ts = -sx^T z + ts < 0,$$

which contradicts with the assumption that  $(x, t) \in K^*$ . This shows that  $K^* \subseteq K$ , and thus,  $K^* = K$ .

5. (a) The conjugate function is

$$f^*(y) = \begin{cases} 0, & \text{if } y \geq 0, \mathbf{1}^T y = 1, \\ \infty, & \text{otherwise.} \end{cases}$$

Here is the analysis. If there exists a  $y_i < 0$ , then we take  $x$  such that  $x_j = 0$  for all  $j \neq i$ , and let  $x_i \rightarrow -\infty$ . Then,

$$y^T x - f(x) = y_i x_i \rightarrow +\infty \implies f^*(y) = +\infty.$$

If there exists a  $y_i > 1$ , then we take  $x$  such that  $x_j = 0$  for all  $j \neq i$ , and let  $x_i \rightarrow +\infty$ . Then,

$$y^T x - f(x) = y_i x_i - x_i = (y_i - 1)x_i \rightarrow +\infty \implies f^*(y) = +\infty.$$

Now we consider  $y$  such that  $0 \leq y_i \leq 1$  for all  $i$ . Given any  $x$ , suppose  $x_i = \max_k x_k$ , then

$$y^T x - f(x) = y^T x - x_i \leq y^T (\mathbf{1} x_i) - x_i = x_i (y^T \mathbf{1} - 1),$$

since  $y_i \geq 0$  and the equality is attained if  $x_j = x_i$  for all  $1 \leq j \leq n$ .

If  $y^T \mathbf{1} \neq 1$ , then we can take  $x_i \rightarrow +\infty$  or  $-\infty$  so  $y^T x - f(x) \rightarrow +\infty$ . In other words,

$$f^*(y) = +\infty \quad \text{if } 0 \leq y_i \leq 1, y^T \mathbf{1} \neq 1.$$

If  $y^T \mathbf{1} = 1$ , then following the preceding analysis, we have

$$y^T x - f(x) \leq y^T (\mathbf{1} x_i) - x_i = 0,$$

which is attained if all  $x_i$ 's are equal.

Thus,

$$f^*(y) = 0 \quad \text{if } 0 \leq y_i \leq 1, y^T \mathbf{1} = 1.$$

(b) The conjugate function is

$$f^*(y) = \begin{cases} 0, & \text{if } y \in C, \\ +\infty, & \text{otherwise.} \end{cases}$$

with

$$C = \left\{ y \in \mathbb{R}^n \mid 0 \leq y \leq 1, y_{[j]} + \dots y_{[n]} \geq r + 1 - j, \forall 2 \leq j \leq r, \sum_{k=1}^n y_k = r \right\}.$$

In fact<sup>1</sup>, since  $0 \leq y_i \leq 1$ ,  $C$  can be simplified to

$$C = \left\{ y \in \mathbb{R}^n \mid 0 \leq y \leq 1, \sum_{k=1}^n y_k = r \right\}.$$

Here is the analysis.

If there is a  $y_i > 1$ , then let  $x$  such that  $x_j = 0$  for  $j \neq i$  and  $x_i \rightarrow +\infty$ , and we know that  $f^*(y) = +\infty$ .

Similarly, if there is a  $y_i < 0$ , then let  $x$  such that  $x_j = 0$  for  $j \neq i$  and  $x_i \rightarrow -\infty$ , and we know that  $f^*(y) = +\infty$ .

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<sup>1</sup>Upon looking at [this SE post](#) afterwards, I realized that my expression can be simplified

Now suppose  $0 \leq y \leq 1$ . If  $y \notin C$ , then there exists some  $y_{i_1} + \dots + y_{i_{n-j+1}} < r + 1 - j$  for some  $2 \leq j \leq r$ , or it might be that  $\sum_{k=1}^n y_k \neq r$ .

Consider the first case, where we have  $y_{i_1} + \dots + y_{i_{n-j+1}} < r + 1 - j$  for some  $2 \leq j \leq r$ . Then we can take  $x_{i_1} = x_{i_2} = \dots = x_{i_{n-j+1}} \rightarrow -\infty$  and let all other  $x_k = 0$ . Thus,

$$y^T x - f(x) = (y_{i_1} + \dots + y_{i_{n-j+1}})x_{i_1} - (r - j + 1)x_{i_1} = (y_{i_1} + \dots + y_{i_{n-j+1}} - (r - j + 1))x_{i_1} \rightarrow +\infty.$$

Therefore,

$$f^*(y) = +\infty.$$

Now suppose  $0 \leq y \leq 1$ ,  $y_{[j]} + \dots + y_{[n]} \geq r + 1 - j, \forall 2 \leq j \leq r$ , but  $\sum_{k=1}^n y_k \neq r$ . If  $y^T \mathbf{1} > r$ , then take  $x_j = x_1$  for all  $j$ , we have

$$y^T x - f(x) = y^T \mathbf{1} x_1 - r x_1 \rightarrow +\infty$$

as  $x_1 \rightarrow +\infty$ , so

$$f^*(y) = +\infty.$$

If  $y^T \mathbf{1} < r$ , then let  $x_j = x_1$  for all  $j$ , we have, as  $x_1 \rightarrow -\infty$ ,

$$y^T x - f(x) = y^T \mathbf{1} x_1 - r x_1 \rightarrow +\infty.$$

So

$$f^*(y) = +\infty.$$

Therefore, if  $y \notin C$ ,  $f^*(y) = +\infty$ .

Now we consider the case where  $y \in C$ . Given any  $x \in \mathbb{R}^n$ , we denote  $x_{i_j} = x_{[j]}$ . Since  $y \in C$ , we have

$$\begin{aligned} y^T x - f(x) &\leq y_{i_1} x_{i_1} + \dots + y_{i_{r-1}} x_{i_{r-1}} + (y_{i_r} + \dots + y_{i_n} - 1) x_{i_r} - (x_{i_1} + \dots + x_{i_{r-1}}) \\ &\leq y_{i_1} x_{i_1} + \dots + y_{i_{r-2}} x_{i_{r-2}} + (y_{i_{r-1}} + y_{i_r} + \dots + y_{i_n} - 2) x_{i_{r-1}} - (x_{i_1} + \dots + x_{i_{r-2}}) \\ &\leq \dots \\ &\leq (y_{i_1} + \dots + y_{i_n} - r) x_{i_1} \\ &= (y^T \mathbf{1} - r) x_{i_1} \\ &= 0. \end{aligned}$$

Thus,  $f^*(y) \leq 0$ . On the other hand, we can take  $x$  such that  $x_i = 0$  for all  $i$  to get  $y^T x - f(x) = 0$ .

So

$$f^*(y) = 0, \quad \forall y \in C.$$

6. The code is listed below. Here in Fig 1, we rerun the example on the webpage with our function with **A** and **b** and the result is on the left, which agrees with the true result and shows the reliability of our function. On the right, we add two other constraints, which are

$$\begin{aligned} -2x - y &\leq 1, \\ 4x - y &\leq 1. \end{aligned}$$

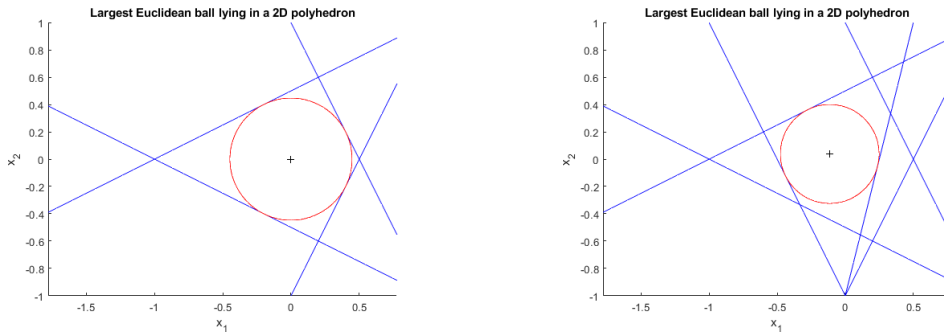


Figure 1: Left: reproduce the example on the webpage; Right: add two additional constraints

```

1 function [x_c, r] = ChebyCenter(A,b)
2 % each column of A corresponds to a line
3 % --- input ---
4 % A: a matrix; b: a vector -- the RHS
5 % A(:,k) ... b(k)
6 % --- output ---
7 % x_c: the location of the center
8 % r: the radius
9
10 n = size(A,2);
11 B = A';
12 len = sqrt(sum(B.^2 ,2));
13
14
15 % Create and solve the model
16 cvx_begin
17     variable r(1)
18     variable x_c(2)
19     maximize ( r )
20     B*x_c + r* len <= b;
21 cvx_end
22
23 % Generate the figure
24 x = linspace(-2,2);
25 theta = 0:pi/100:2*pi;
26
27 hold on
28
29 for i = 1:n
30     plot( x, -x*A(1,i)./A(2,i) + b(i)./A(2,i), 'b-');
31 end
32 plot( x_c(1) + r*cos(theta), x_c(2) + r*sin(theta), 'r');
33 plot(x_c(1),x_c(2), 'k+')
34 xlabel('x.1')
35 ylabel('x.2')
36 title('Largest Euclidean ball lying in a 2D polyhedron');
37 axis([-1 1 -1 1])
38 axis equal

```

The main function for this question:

```

1 close all;
2 clear variables;
3
4 %A = [2,2,-1,-1;1,-1,2,-2]; % original
5 A = [2,2,-1,-1,-2, 4; 1,-1,2,-2,-1,-1];
6 %b = ones(4,1); % original
7 b = ones(6,1);
8 [x_c, r] = ChebyCenter(A,b);

```

7. If there is no point inside the polyhedron, the output  $r$  is nonpositive. In the plot, we can get a circle, though. This is because the way we plot the result includes  $r * \cos(\theta)$  and  $r * \sin(\theta)$  which are well-defined with any  $r$ . But since the polyhedron is empty, we shouldn't expect to get any meaningful result. The function doesn't pop out an error because we don't require  $r \geq 0$  in our LP.

Here is an example where we add the following additional constraint to the original problem

$$2x + y \leq -3.$$

The result is as follows, which gives  $r = -0.149071$ .

```

1 close all;
2 clear variables;

```

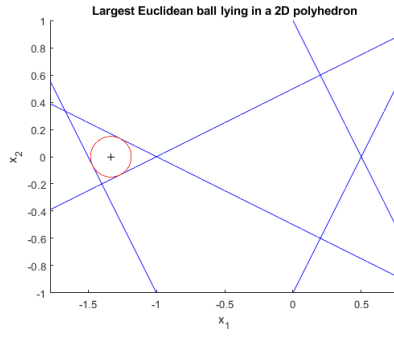


Figure 2: Polyhedron doesn't contain any point

```

3
4 A = [2, 2, -1, -1, 2; 1, -1, 2, -2, 1];
5 b = [1; 1; 1; 1; -3];
6
7 [x_c, r] = ChebyCenter(A, b);

```

8. Following the book, now we require that

$$\|u\|_p \leq r \implies a_i^T(x_c + u) \leq b_i.$$

Since

$$\sup \{a_i^T u \mid \|u\|_p \leq r\} = r \|a_i\|_q,$$

where  $q$  is conjugate to  $p$  in the sense that  $1/p + 1/q = 1$  ( $p$  or  $q$  can be  $\infty$ ),

we can write the condition as

$$a_i^T x_c + r \|a_i\|_q \leq b_i,$$

which is a linear inequality in  $x_c$  and  $r$ .

The optimization problem is an LP as

$$\begin{aligned} & \text{maximize} && r \\ & \text{subject to} && a_i^T x_c + r \|a_i\|_q \leq b_i, \quad i = 1, \dots, m. \end{aligned}$$

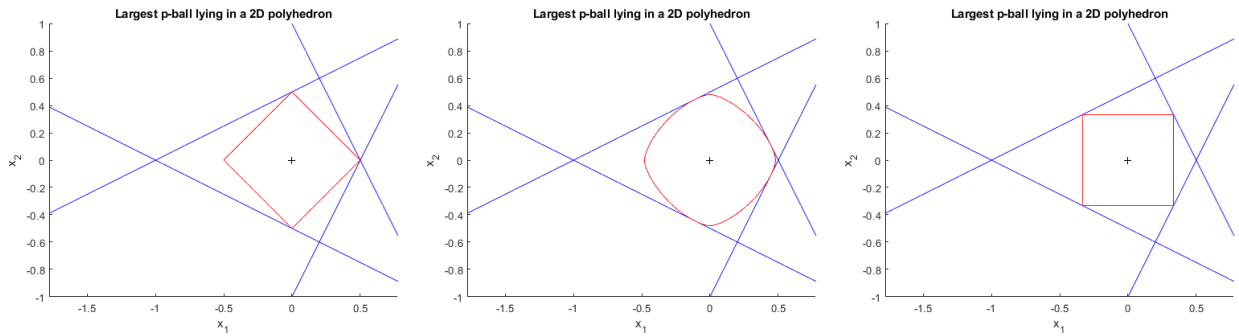


Figure 3: Left:  $p = 1$ ; Middle:  $p = 1.5$ ; Right:  $p = \infty$ .

```

1 function [x_c, r] = ChebyCenterQnorm(A, b, p)
2 % each column of A corresponds to a line
3 % --- input ---
4 % A: a matrix;
5 % b: a vector -- the RHS
6 % A(:,k) ... b(k)
7 % p: for p-norm
8 % --- output ---

```

```

9 % x_c: the location of the center
10 % r: the radius
11
12
13 n = size(A,2);
14 B = (A');
15 Babs = abs(B);
16 q = 1/(1-(1/p));
17
18 if p == 1
19     len = max(Babs,[],2);
20 else
21     len = (sum(Babs.^q ,2)).^(1/q);
22 end
23
24
25 % Create and solve the model
26 cvx_begin
27     variable r(1)
28     variable x_c(2)
29     maximize ( r )
30     B*x_c + r* len ≤ b;
31 cvx_end
32
33 % Generate the figure
34 x = linspace(-2,2);
35 %theta = 0:pi/100:2*pi;
36
37 hold on
38
39 for i = 1:n
40     plot( x, -x*A(1,i)./A(2,i) + b(i)./A(2,i), 'b-');
41 end
42 %plot( x_c(1) + r*cos(theta), x_c(2) + r*sin(theta), 'r');
43
44 up = r*(-1:0.01:1);
45 xp = up + x_c(1);
46 lxp = length(xp);
47 if p == inf
48     xp = [-r *ones(1,lxp)+x_c(1),xp, r*ones(1,lxp) + x_c(1)];
49     yup = [r*linspace(0,1,lxp), r*ones(1,lxp), r-r*linspace(0,1,lxp)];
50 else
51     yup = (r^p - abs(up).^p).^(1/p);
52 end
53 plot( xp, x_c(2) - yup, 'r');
54 plot( xp, x_c(2) + yup, 'r');
55
56 plot(x_c(1),x_c(2),'k+')
57 xlabel('x_1')
58 ylabel('x_2')
59 title('Largest p-ball lying in a 2D polyhedron');
60 axis([-1 1 -1 1])
61 axis equal

```

The main function is

```

1 close all;
2 clear variables;
3
4 A = [2,2,-1,-1; 1,-1,2,-2];
5 b = ones(4,1);
6
7 p= 1; %1.5 % Inf
8 [x_c, r] = ChebyCenterQnorm(A,b,p);

```

9. We cannot simply set  $p$  in the previous function when  $p < 1$ , because when formulating the LP, we use

the fact that

$$\sup \{a_i^T u \mid \|u\|_p \leq r\} = r \|a_i\|_q.$$

This duality relation between  $p$  and  $q$  norms is based on Holder's inequality for  $p \geq 1$ , and is no longer true when  $p < 1$ . So the preceding problem formulation doesn't hold for  $p < 1$ .

If we try it, say  $p = 0.5$ , we will get the following plot.

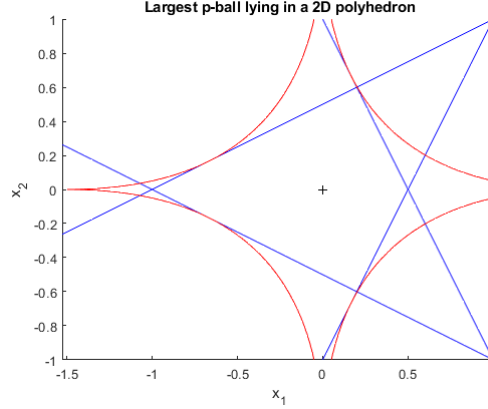


Figure 4:  $p = 0.5$

10. <sup>2</sup> The polyhedron is a convex set. If a  $p$ -unit ball is in the polyhedron, then its convex hull must also lie in the polyhedron.

The key observation is that the convex hull of a  $p$ -unit ball is the 1-norm unit ball, at least in this  $\mathbb{R}^2$  case. Since the 1-norm unit ball is convex, we only need to prove that it is a subset of any convex set containing the  $p$ -unit ball  $B$ . By symmetry, we only need to prove that the region where  $x \geq 0, y \geq 0, x + y \leq 1$  is in the convex hull of  $B$ .

For any  $(x, y)$  such that  $x \geq 0, y \geq 0, x + y \leq 1$ , if  $y < 1$ , then the line going across  $(x, y)$  and  $(0, 1)$  will hit the  $x$ -axis at  $D = (x/(1 - y), 0)$ . Since  $0 \leq x/(1 - y) \leq 1$ ,  $D \in B$ . Since  $(0, 1) \in B$ , we have shown that such  $(x, y)$  is a convex combination of two points in  $B$  and thus,  $(x, y) \in \text{conv} B$ .

Thus, the 1-norm unit ball is the convex hull of the  $p$ -unit ball when  $p < 1$ . So the polyhedron contains a radius  $r$   $p$ -ball if and only if it contains a radius  $r$  1-norm ball with the same center.

Therefore, when  $p < 1$ , we simply consider the case for  $p = 1$  but draw the  $p$ -ball instead of 1-norm ball for the plot.

Here is an example with  $p = 0.5$ .

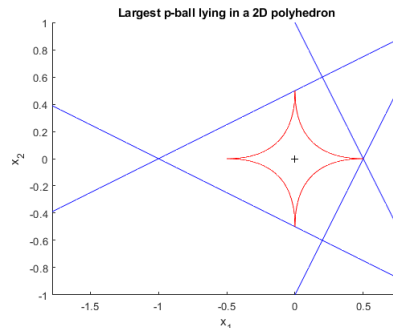


Figure 5:  $p = 0.5$

```
1 function [x_c, r] = ChebyCenterQnormGeneral(A, b, p)
2 % p can be smaller than 1
3 % each column of A corresponds to a line
```

<sup>2</sup>Thanks for the discussion between Sonia and Prof. Overton on campuswire.

```

4 % --- input ---
5 % A: a matrix;
6 % b: a vector -- the RHS
7 % A(:,k) ... b(k)
8 % p: for p-norm
9 % --- output ---
10 % x_c: the location of the center
11 % r: the radius
12
13 n = size(A,2);
14 B = (A');
15 Babs = abs(B);
16 q = 1/(1-(1/p));
17
18 if p ≤ 1
19     len = max(Babs,[],2);
20 else
21     len = (sum(Babs.^q ,2)).^(1/q);
22 end
23
24
25 % Create and solve the model
26 cvx_begin
27     variable r(1)
28     variable x_c(2)
29     maximize ( r )
30     B*x_c + r* len ≤ b;
31 cvx_end
32
33 % Generate the figure
34 x = linspace(-2,2);
35 %theta = 0:pi/100:2*pi;
36
37 hold on
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39 for i = 1:n
40     plot( x, -x*A(1,i)./A(2,i) + b(i)./A(2,i), 'b-');
41 end
42 %plot( x_c(1) + r*cos(theta), x_c(2) + r*sin(theta), 'r');
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44 up = r*(-1:0.01:1);
45 xp = up + x_c(1);
46 lxp = length(xp);
47 if p == inf
48     xp = [-r *ones(1,lxp)+x_c(1),xp, r*ones(1,lxp) + x_c(1)];
49     yup = [r*linspace(0,1), r*ones(1,lxp), -r*r*linspace(0,1)];
50 else
51     yup = (r^p - abs(up).^p).^(1/p);
52 end
53 plot( xp, x_c(2) - yup, 'r');
54 plot( xp, x_c(2) + yup, 'r');
55
56 plot(x_c(1),x_c(2),'k+')
57 xlabel('x_1')
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59 title('Largest p-ball lying in a 2D polyhedron');
60 axis([-1 1 -1 1])
61 axis equal

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The main function is

```

1 close all;
2 clear variables;
3
4
5 A = [2,2,-1,-1; 1,-1,2,-2];

```



```
6 b = ones(4,1);  
7 p= 0.5;  
8 [x_c, r] = ChebyCenterQnormGeneral(A,b,p);
```