

1. Prove that a function is convex if and only if its epigraph is a convex set. Suppose  $f$  is a convex function,  $f : \mathbf{R}^n \rightarrow \mathbf{R}$  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:

$$\begin{aligned} f(\theta x + (1 - \theta)y) &\leq \theta f(x) + (1 - \theta)f(y) \\ &\leq \theta t_1 + (1 - \theta)t_2 \end{aligned}$$

thus  $\mathbf{epi} f$  is convex. The other direction is similar  $\forall (x, t_1), (y, t_2) \in \mathbf{epi} f$ ,  $\mathbf{epi} f$  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  $\mathbf{epi} f$  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  $\forall x \in K$ ,  $x^T(\theta_1 y_1 + \theta_2 y_2) = \theta_1 x^T y_1 + \theta_2 x^T y_2 \geq 0$  thus  $K^*$  is a convex cone.
- (b)  $K_1 \subseteq K_2$  implies  $K_2^* \subseteq K_1^*$ . Suppose  $y \in K_2^*, \forall x \in K_1, x^T y \geq 0$ , 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  $(K^*)^* = \{x^* | y^T x^* \geq 0, \forall y \in K^*\}$ . We can consider  $(K^*)^*$  as the intersection of halfspaces  $H_{x^* \in K} = \{y^T x^* \geq 0, \forall y \in K^*\}$ . If  $x \in K$  then  $\forall y \in K^*, x^T y = y^T x \geq 0 \Rightarrow x \in (K^*)^*$ .  $K$  being convex and closed by the corollary of the separating hyperplanes,  $K = (K^*)^*$ .

4. BV Ex. 2.33 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  $K^* = \{y | (A x)^T y \geq 0, \forall x \geq 0\}$  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  $\{e_1, \dots, e_n\}$  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,  $C = \{(x, t) \in \mathbf{R}^n | \|x\|_2 \leq t\}$ .

$$C^* = \{(y, s) | \begin{bmatrix} x \\ t \end{bmatrix}^T \begin{bmatrix} y \\ s \end{bmatrix} \geq 0, \forall (x, t) \in C\}. \text{ if } (y, s) \in C \text{ then } x^T y \leq \|x\|_2 \|y\|_2$$

using Cauchy-Schwarz or  $x^T y \leq t s$ .  $\begin{bmatrix} x \\ t \end{bmatrix}^T \begin{bmatrix} y \\ s \end{bmatrix} = x^T y + ts$ , and by the triangle inequality,  $\|x^T y + ts\| \geq t s - |x^T y| \geq 0 \Rightarrow y \in C^*$ . 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  $x^T y = -\frac{n}{s^2} y_m \leq -1$  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 inequalities in t.
% fashion: P = {x : a_i' * x <= b_i, i=1,...,m}

% Generate the data
[~, 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), 2) <= 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 laying in a 2D polyhedron", txt});
    text(x_c(1), x_c(2), "\leftarrow center")
    axis([-1 1 -1 1])
    axis equal
    hold off
    txt = "chebyshev_center_" + num2str(n);
    saveas(gcf,txt, 'eps')

```

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  $b = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$ , we obtain the same circle which is tangent to the four

hyperplanes  $a_i^T x = b_i$  (see figure 1):

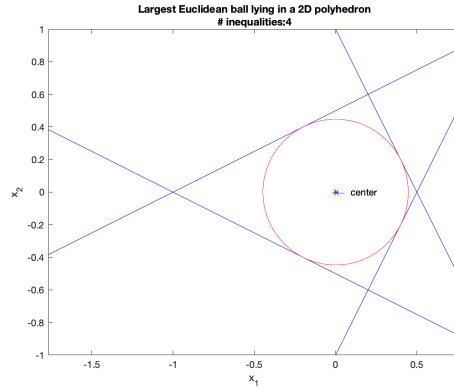


Figure 1: Sample example

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

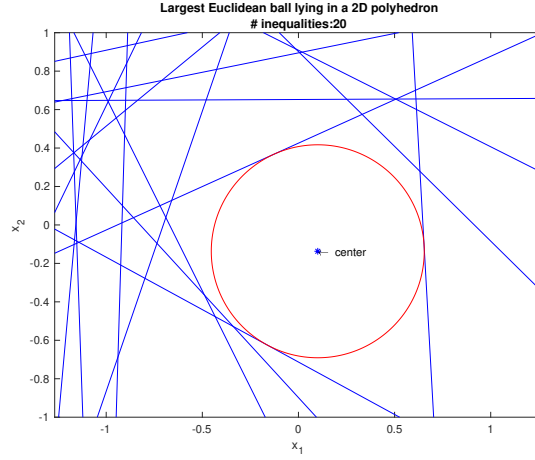


Figure 2: Example with 20 inequalities

- (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"  $\mathcal{B} = \{x_c + u \mid \|u\|_p \leq r\}$  that lies in the polyhedron described by a set of linear inequalities:  $\mathcal{P} = \{x \in \mathbf{R}^n \mid a_i^T x \leq b_i, i = 1, \dots, m\}$ . For any point of  $\mathcal{B}$  laying in one halfspace  $a_i^T x \leq b_i$ , similarly to the euclidean space, we have  $\|u\|_p \leq r \Rightarrow a_i^T x_c + r \|a_i\|_p \leq b_i$  since  $g_i = \sup \{a_i^T u \mid \|u\|_p \leq r\} = r \|a_i\|_p$ . And the Chebyshev center can be determined by solving the problem:

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

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)).

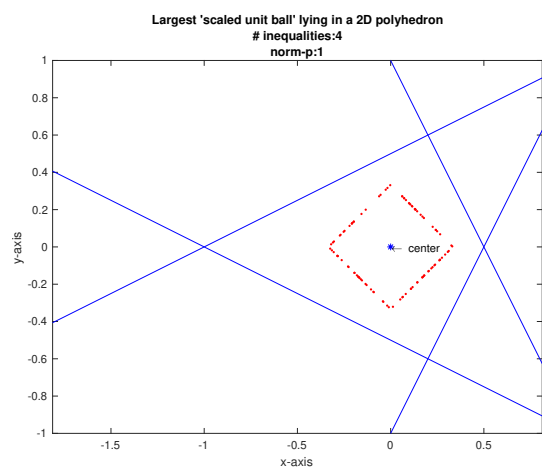


Figure 3: "Scaled" unit ball for  $p=1$

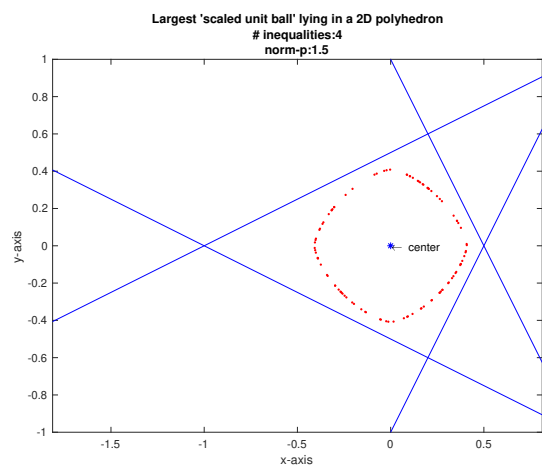


Figure 4: "Scaled" unit ball for  $p=1.5$

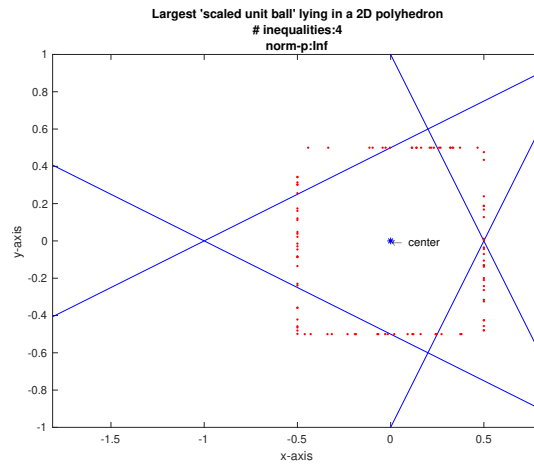


Figure 5: "Scaled" unit ball for  $p=\infty$

```
% 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 inequalities in t.
% fashion:  $P = \{x : a_i' * x \leq b_i, i=1, \dots, m\}$ 

rng('default')
format long g
[~, n] = size(A);

% Build and execute model
fprintf(1, 'Computing Chebyshev center...\n');
cvx_begin
    variable r(1)
    variable x_c(2)
    maximize ( r )
    for k=1:n
        A(:, k)' * x_c + r * norm(A(:, k), p) <= 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.* r + x_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' laying in a 2D polyhedron", txt1, tx2})
text(x_c(1), x_c(2), "\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);
end

```

- (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.

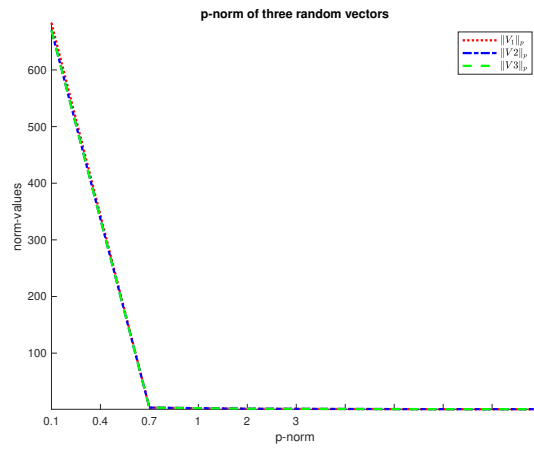


Figure 6: p-norms of three random vectors