

1. Convexity

A function $f : V \rightarrow \mathbb{R}$ from a vector space V to the set of reals is *convex* if it satisfies

$$f(tx + (1 - t)y) \leq tf(x) + (1 - t)f(y)$$

for all $x, y \in V$, $t \in [0, 1]$.

Also, a set $K \subseteq V$ is *convex* if $x, y \in K \implies tx + (1 - t)y \in K$ for all $t \in [0, 1]$.

(a) If $f : V \rightarrow \mathbb{R}$ is convex, show that the *sub-level sets* defined by

$$E_\alpha = \{x \in V \mid f(x) \leq \alpha\}$$

are convex for all $\alpha \in \mathbb{R}$.

Solution. Let $x, y \in E_\alpha$, $\alpha \in \mathbb{R}$, and $t \in [0, 1]$. Therefore $f(x) \leq \alpha$ and $f(y) \leq \alpha$. Since f is convex, the following is true

$$\begin{aligned} f(tx + (1 - t)y) &\leq tf(x) + (1 - t)f(y) \\ &\leq t\alpha + (1 - t)\alpha = \alpha \\ \implies tx + (1 - t)y &\in E_\alpha. \end{aligned}$$

\therefore the sub-level sets E_α are convex for all $\alpha \in \mathbb{R}$.

(b) If V is a normed linear space, show that the unit ball $\{x \mid \|x\| \leq 1\}$.

Solution. The main piece of the proof is to use the triangle inequality of the norm on V . Let $x, y \in \text{u.b.}$ where u.b. denotes the unit ball and $t \in [0, 1]$. $\|x\|, \|y\| \leq 1$. Consider $\|tx + (1 - t)y\|$

$$\begin{aligned} \|tx + (1 - t)y\| &\leq \|tx\| + \|(1 - t)y\| \\ &= t\|x\| + (1 - t)\|y\| \\ &\leq t + (1 - t) = t \leq 1 \\ \implies tx + (1 - t)y &\in \text{u.b.} \end{aligned}$$

\therefore the unit ball $\{x \mid \|x\| \leq 1\}$ is convex.

2. Two norms $\|\cdot\|_a$ and $\|\cdot\|_b$ on a vector space are *equivalent* if they generate equivalent metrics.

(a) Show that the two norms are equivalent if and only if there exist constants c, C such that

$$c\|v\|_a \leq \|v\|_b \leq C\|v\|_a \quad \forall v \in V.$$

Solution. (\implies) Assume the norms are equivalent, then, by the definition of metric spaces

$$\begin{aligned} \forall \epsilon, \epsilon' > 0, \exists \delta, \delta' > 0 \text{ s.t. } \quad & \|v - v_0\|_a < \delta \implies \|v - v_0\|_b < \epsilon \\ \text{and} \quad & \|v - v_0\|_b < \delta' \implies \|v - v_0\|_a < \epsilon' \end{aligned}$$

we can construct constants

$$\begin{aligned} c &< \frac{\epsilon'}{\delta'} \\ C &> \frac{\epsilon}{\delta} \end{aligned}$$

such that

$$c\|v\|_a \leq \|v\|_b \leq C\|v\|_a.$$

(\Leftarrow) Assume we have constants c, C s.t.

$$c\|v\|_a \leq \|v\|_b \leq C\|v\|_a.$$

Construct a sequence $x_n \rightarrow x$ in l^b . Using the assumption, we get

$$c\|x_n - x\|_a \leq \|x_n - x\|_b \leq C\|x_n - x\|_a.$$

Pick $\delta' > 0$. Then

$$\|x_n - x\|_b < \delta' \implies \|x_n - x\|_a < c\delta' = \epsilon'.$$

Likewise, pick $\delta > 0$ and $x_n \rightarrow x$ in l^a , then

$$\|x_n - x\|_a < \delta \implies \|x_n - x\|_b < C\delta = \epsilon.$$

\therefore The norms are equivalent since we have

$$\begin{aligned} \forall \epsilon, \epsilon' > 0, \exists \delta, \delta' > 0 \text{ s.t. } \quad & \|v - v_0\|_a < \delta \implies \|v - v_0\|_b < \epsilon \\ \text{and} \quad & \|v - v_0\|_b < \delta' \implies \|v - v_0\|_a < \epsilon'. \end{aligned}$$

\therefore two norms are equivalent if and only if there exist constants c, C such that

$$c\|v\|_a \leq \|v\|_b \leq C\|v\|_a \quad \forall v \in V.$$

(b) Show that any two norms on a finite dimensional vector space are equivalent. (The one fact you can use without proof is that the unit l^1 (or any l^p) sphere in \mathbb{R}^n is compact, and in a compact set, every sequence has a convergent subsequence.)

Solution. Let $1 \leq p \leq \infty$. First show that every norm is bounded above by the 1-norm, by the factor of some constant C . To do this, hopefully we get $\|\mathbf{x}\|_p \leq C\|\mathbf{x}\|_1$. Construct a basis for \mathbb{R}^n , e_i , s.t. $\mathbf{x} = \sum_{i=1}^n x_i e_i$. Then

$$\left\| \sum_{i=1}^n x_i e_i \right\|_p \leq \sum_{i=1}^n \|x_i e_i\|_p \leq \sum_{i=1}^n |x_i| \|e_i\|_p \leq \max_i |e_i| \sum_{i=1}^n |x_i| = \max_i |e_i| \|\mathbf{x}\|_1 < \infty.$$

Therefore, every p -norm is bounded by the 1-norm. Now, if $\|\mathbf{x}\|_1 \leq C\|\mathbf{x}\|_p$ we are done.

Use proof by contradiction to show that $\|\mathbf{x}\|_1 \leq C\|\mathbf{x}\|_p$. Suppose $\exists x_n$, a subsequence of \mathbf{x} for all n s.t. $\|x_n\|_1 > n\|x_n\|_p$. Define

$$y_n = \frac{x_n}{\|x_n\|_1}$$

which is on the l^1 unit ball: a compact set. Using the hint given; we can use the fact that every compact set contains a convergent subsequence, call it $w_n \rightarrow w$ in l^1 . Therefore, using earlier result of this exercise,

$$\|w_n - w\|_p \leq C\|w_n - w\|_1$$

which shows that $\|w_n - w\|_p \rightarrow 0$. But,

$$\begin{aligned} \|y_n\|_p &= \frac{\|x_n\|_p}{\|x_n\|_1} \leq \frac{1}{n\|x_n\|_1} \|x_n\|_1 = \frac{1}{n} \\ \text{from above } n\|y_n\|_p &< \|y_n\|_1 \\ \implies \|y_n\|_1 &> 1. \end{aligned}$$

However, since $\|y_n\|_p \rightarrow 0$ and $\|w_n - w\|_p \rightarrow 0 \implies w = 0$. However, by our assumption, $\|w_n\|_1 = 0 \not> 1$. Contradiction.

$\therefore \|\mathbf{x}\|_1 \leq C\|\mathbf{x}\|_p$.

Therefore, two norms are equivalent on a finite dimensional space.

3. **Seminorms** V is a vector space over \mathbb{R} . $p : V \rightarrow [0, \infty)$ is a seminorm, if it satisfies the triangle inequality and $p(\alpha v) = |\alpha|p(v)$, $\forall v \in V$ and $\alpha \in \mathbb{R}$.

(a) Show that p is a convex function.

Solution. Since p is a seminorm, the triangle inequality holds. We can construct $v, w \in V$ s.t. $w = tx$ and $v = (1-t)y$ for some $x, y \in V$, $t \in [0, 1]$. So, by the triangle inequality,

$$p(tx + (1-t)y) \leq p(tx) + p((1-t)y).$$

Use the second property of seminorms given to get

$$\begin{aligned} p(tx + (1-t)y) &\leq p(tx) + p((1-t)y) \\ &= |t|p(x) + |1-t|p(y) \\ &= tp(x) + (1-t)p(y). \end{aligned}$$

which is the definition of a convex function.

$\therefore p$ is convex

(b) If $L : V \rightarrow \mathbb{R}$ is a linear map, show that $p(v) = |L(v)|$ is a seminorm.

Solution. First show the triangle inequality. Assume $v, w \in V$, then

$$\begin{aligned} |L(v+w)| &= \left| \sum_i \alpha_i (v_i + w_i) \right| \quad \text{since } L \text{ is linear} \\ &= \left| \sum_i \alpha_i v_i + \sum_i \alpha_i w_i \right| \\ &\leq \left| \sum_i \alpha_i v_i \right| + \left| \sum_i \alpha_i w_i \right| \\ &= |L(v)| + |L(w)| \end{aligned}$$

which shows the triangle inequality holds.

The second property follows; for $a \in \mathbb{R}$

$$\begin{aligned} |L(av)| &= \left| \sum_i \alpha_i (av_i) \right| \\ &= |a| \left| \sum_i \alpha_i v_i \right| \\ &= |a| |L(v)|. \end{aligned}$$

Therefore the second property of seminorms holds.

$\therefore p(v) = |L(v)|$ is a seminorm.

(c) If $q : V \rightarrow \mathbb{R}$ is a convex function that satisfies $q(\alpha v) = |\alpha|q(v)$, $\forall v \in V$, show that q is a seminorm.

Solution. Let $x, y \in \mathbb{R}$. We only need to show the triangle inequality holds. Since q is convex satisfying $q(\alpha v) = |\alpha|q(v)$,

$$\begin{aligned} q(tx + (1-t)y) &\leq tq(x) + (1-t)q(y) \\ &\leq q(tx) + q((1-t)y). \end{aligned}$$

Let $v, w \in V$, $v = tx$, $w = (1-t)y$, we get the triangle inequality $q(v+w) \leq q(v) + q(w)$.

$\therefore q$ is a seminorm.

(d) If p_α is a family of seminorms on V , which are indexed by an arbitrary set $\alpha \in A$, show that p is a seminorm on a linear space V_p , where

$$p(v) = \sup_{\alpha \in A} p_\alpha(v), \quad V_p = \{v \in V | p(v) < \infty\}.$$

Solution. Start with the triangle inequality. We know that

$$p(v+w) = \sup_{\alpha \in A} p_{\alpha}(v+w) \leq \sup_{\alpha \in A} [p_{\alpha}(v) + p_{\alpha}(w)] = \sup_{\alpha \in A} p_{\alpha}(v) + \sup_{\alpha \in A} p_{\alpha}(w) = p(v) + p(w)$$

which shows the triangle inequality on (V_p, p) .

The second property follows almost immediately; let $a \in \mathbb{R}$,

$$p(av) = \sup_{\alpha \in A} p_{\alpha}(av) = \sup_{\alpha \in A} |a|p_{\alpha}(v) = |a| \sup_{\alpha \in A} p_{\alpha}(v) = |a|p(v).$$

$\therefore p$ is a seminorm on V_p .

(e) Find a countable family $L_k, k \in \mathbb{N}$ of linear maps from which we can obtain the l^p norm on \mathbb{R}^n as

$$\|\mathbf{x}\|_p = \sup_k |L_k(\mathbf{x})|.$$

Solution. Let $L : \mathbb{R}^n \rightarrow \mathbb{R}$ be a linear function. Then we know that $\dim(\text{null}(L)) = (n-1)$. We wish to associate every point on the unit ball with a linear function. To do so we will assign every point with a line tangent to the unit ball at that point. First we choose a coordinate on the unit ball or \mathbb{R}^n by choosing the first $(n-1)$ coordinates to be rational numbers, let them be $(a_1, a_2, a_3, \dots, a_{n-1})$. Next we have the restriction that this point in \mathbb{R}^n must lie on the unit ball, thus we have

$$a_1^p + a_2^p + \dots + a_{n-1}^p + a_n^p = 1,$$

or,

$$1 - (a_1^p + a_2^p + \dots + a_{n-1}^p) = a_n^p.$$

The dimension of the null space of a linear function is $(n-1)$, therefore the $(n-1)$ coordinates that we have chosen will uniquely determine the last coordinate $a_n \in \mathbb{R}$. To find the equation of the line, choose a hyperplane $\mathbf{z} \in \mathbb{R}^n$ that lies tangent to the unit ball. This tangent hyperplane can be expressed by $\mathbf{z} = (a - \mathbf{x})$ where \mathbf{x} is the hyperplane from the end of \mathbf{z} through the origin. Thus the equation for this tangent hyperplane is

$$\nabla f \cdot \mathbf{z} = 0.$$

with $f = a_1^p + a_2^p + \dots + a_{n-1}^p + a_n^p$. Now using calculus in higher dimensions

$$\begin{aligned} \nabla f \cdot \mathbf{z} &= (pa_1^{p-1} + \dots + pa_n^{p-1}) \cdot (a - \mathbf{x}) \\ &= a_1^p + \dots + a_n^p - (a_1^{p-1}x_1 + \dots + a_n^{p-1}x_n) = 0 \\ \implies 1 &= a_1^{p-1}x_1 + \dots + a_n^{p-1}x_n. \end{aligned}$$

This last line defines our linear functional, call it L_1 . Keep assigning $(n-1)$ coordinates with these linear functionals and indexing L_i every time you do it to build a countable

set of L_i 's. Note that this is countable since this is a countable union of $(n-1)$ rational numbers. If we take the p -norm to be the supremum of all of these functions we shall attain the irrational coordinates as well since \mathbb{N} is a dense set in \mathbb{R} . Therefore,

$$\|\mathbf{x}\|_p = \sup_k |L_k(\mathbf{x})|$$

4. Matrix Norms

(a) Problem 2.3.35 from Flaschka:

(i) For a matrix $\mathbf{A} = (a_{ij})$ on \mathbb{R}^n , show that

$$\begin{aligned} \|\mathbf{A}\|_\infty &\leq \max_i \sum_{j=1}^n |a_{ij}|, \\ \|\mathbf{A}\|_1 &\leq \max_j \sum_{i=1}^n |a_{ij}|. \end{aligned}$$

Solution. Use the definition of the ∞ -norm for matrices

$$\|\mathbf{A}\|_\infty = \max_{i,j} |a_{ij}| \leq \max_i \sum_{j=1}^n |a_{ij}|.$$

For the next inequality, use 1-norm definition and the fact that $\max_j |a_{ij}| \leq \sum_{j=1}^n |a_{ij}|$

$$\|\mathbf{A}\|_1 = \sum_{j=1}^n \sum_{i=1}^n |a_{ij}| \leq \max_j \sum_{i=1}^n |a_{ij}|.$$

\therefore the inequalities hold.

(ii) Show that there is always an $\mathbf{x} \in \mathbb{R}^n$ with $\|\mathbf{x}\|_\infty = 1$, for which

$$\|\mathbf{Ax}\|_\infty = \max_i \sum_{j=1}^n |a_{ij}|.$$

Hence deduce that $\|\mathbf{A}\|_\infty = \max_i \sum_{j=1}^n |a_{ij}|$.

Solution. Such a vector \mathbf{x} is easy to construct, take \mathbf{x} to be such that we know the k th row is the max row of \mathbf{A}

$$x_j = \begin{cases} -1 & \text{if } a_{kj} < 0 \\ 1 & \text{if } a_{kj} \geq 0 \end{cases}$$

Therefore, $\|\mathbf{Ax}\|_\infty = \max |\mathbf{Ax}|$. From our definition of \mathbf{x} and that k is the max row, the definition of maximum amounts to the maximum of a row sum of \mathbf{A} , that is

$$(\mathbf{Ax})_k = \|\mathbf{Ax}\|_\infty = \max_i \sum_{j=1}^n |a_{ij}|.$$

We can define the ∞ -norm of a matrix then by

$$\|\mathbf{A}\|_\infty = \max_{\|\mathbf{x}\|_\infty=1} \|\mathbf{Ax}\|_\infty = \max_i \sum_{j=1}^n |a_{ij}|.$$

(iii) Prove in a similar way that $\|\mathbf{A}\|_1 = \max_j \sum_{i=1}^n |a_{ij}|$.

Solution. Construct \mathbf{x} as a unit vector in some direction, call it k . That is, the k^{th} entry is 1, all others are 0. Pick k such that the matrix-vector product is maximal, that is where a_{ik} is maximal. From the definition of the 1-norm for vectors, we get

$$\|\mathbf{A}\|_1 = \max_{\|\mathbf{x}\|_1=1} \|\mathbf{Ax}\|_1 = \max_j \sum_{i=1}^n |a_{ij}|.$$

(b) Problem 2.3.39 from Flaschka:

Let \mathbf{A} be an $n \times n$ matrix. Show that

$$\|\mathbf{A}\|_2 \leq \sqrt{\|\mathbf{A}\|_1 \|\mathbf{A}\|_\infty}.$$

(Hint: Exercise 1.3.33.)

Solution. From exercise 1.3.33, we get the inequality

$$\begin{aligned} \|\mathbf{Ax}\|_q &\leq M^{1/p} N^{1/q} \|\mathbf{x}\|_q \\ &= \|\mathbf{A}\|_\infty^{1/p} \|\mathbf{A}\|_1^{1/q} \|\mathbf{x}\|_q. \end{aligned}$$

Let $p = q = 2$. The following is true

$$\begin{aligned} \|\mathbf{A}\|_2 &= \max_{\|\mathbf{x}\|_2=1} \|\mathbf{Ax}\|_2 \\ &\leq \max_{\|\mathbf{x}\|_2=1} \|\mathbf{A}\|_\infty^{1/2} \|\mathbf{A}\|_1^{1/2} \|\mathbf{x}\|_2 \\ &= \|\mathbf{A}\|_\infty^{1/2} \|\mathbf{A}\|_1^{1/2} \\ &= \sqrt{\|\mathbf{A}\|_\infty \|\mathbf{A}\|_1}. \end{aligned}$$

\therefore the conjecture is true.

5. The space c is the spaces of all sequences $\{x_n\}$ that converge as $n \rightarrow \infty$ and the space $c_0 \subset c$ is the space of all sequences which converge to 0. These spaces are equipped with the l^∞ norm.

Determine the duals for the spaces $l^p(\mathbb{R}, \mathbb{N})$ for $1 \leq p < \infty$ and for the spaces c_0, c .

Solution. Claim: The dual of $l^p(\mathbb{R}, \mathbb{N})$ is $l^q(\mathbb{R}, \mathbb{N})$. I can show this by looking at the set of sequences $\{x | x_n = 0 \forall n > N\}$. Given x in this set and ϕ a linear map, consider $\phi(x) = |\alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_N x_N|$. Now determine if the linear map is bounded.

$$\begin{aligned}\phi(x) &= |\alpha_1 x_1 + \alpha_2 x_2 + \dots + \alpha_N x_N| \\ &\leq \left(\sum_{i=1}^N |\alpha_i|^q \right)^{\frac{1}{q}} \|x\|_p \\ \implies \|\phi\|_{\text{ind}} &= \sup_{\|x\|_p=1} |\phi(x)| \leq \|\alpha\|_q\end{aligned}$$

where Hölder's inequality is used. Since α is finite dimensional, clearly ϕ is a bounded map. Now, we need to show equality. To do this construct x such that $\|x\|_p = 1$ and $\|\phi(x)\|_{\text{ind}} \geq \|\alpha\|_q$. The following works

$$\begin{aligned}x^* &= \left\{ x_n^* = \frac{\text{sgn}(\alpha_n) |\alpha_n|^{q-1}}{\|\alpha\|_q^{q/p}} \mid n < N \right\} \\ \implies |\phi(x^*)| &= \frac{\|\alpha\|_q^q}{\|\alpha\|_q^{q/p}} = \|\alpha\|_q \\ \implies \|\phi\|_{\text{ind}} &\geq |\phi(x^*)| = \|\alpha\|_q.\end{aligned}$$

which implies equality of the induced norm; that is, ϕ induces norm $\|\cdot\|_q$. The argument can be extended in the same fashion to all N via the appropriate definition of ϕ . Therefore the claim is true: $l^q(\mathbb{R}, \mathbb{N})$ is the dual for the space $l^p(\mathbb{R}, \mathbb{N})$ where $q = \frac{p}{p-1}$.

Next, we need to determine the dual for the space c_0 . If $x \in c_0$, $\forall \epsilon \exists M$ s.t. $N > M \implies |x_N| < \epsilon$. Since we have l^∞ , $\|x\|_\infty < \infty$. Given some linear map $\phi(x) = \sum_{i=1}^\infty \alpha_i x_i$, we have the following

$$\begin{aligned}|\phi(x)| &= \left| \sum_{i=1}^\infty \alpha_i x_i \right| \leq \sum_{i=1}^\infty |\alpha_i x_i| \leq \|\alpha\|_1 \|x\|_\infty \\ \implies \|\phi\|_{\text{ind}} &= \sup_{\|x\|_\infty=1} |\phi(x)| \leq \|\alpha\|_1\end{aligned}$$

The first inequality is from Hölder's inequality taking limiting case of $p = 1$, $q = \infty$. We are justified in using Hölder's inequality since x is a convergent sequence. Hence ϕ is a bounded linear map. Now for equality follow the same steps as above:

$$\begin{aligned}x^* &= \text{sgn}(\alpha_i) \implies \|x^*\|_\infty = 1 \\ |\phi(x^*)| &= \|\alpha\|_1 \\ \implies \|\phi\|_{\text{ind}} &\geq |\phi(x^*)| = \|\alpha\|_1.\end{aligned}$$

which, with the boundedness piece shows equality. Therefore the dual of the space c_0 is $l^1(c_0)$.

The conclusion and argument for the dual of the space c follows exactly from above where instead of considering sequences that converge to 0, we will consider sequences that converge to some $x_0 \in \mathbb{R}$. Consider instead the map

$$\phi(x - x_0) = \left| \sum_{i=1}^{\infty} \alpha_i(x_i - x_0) \right| \leq \sum_{i=1}^{\infty} |\alpha_i(x_i - x_0)| \leq \|\alpha\|_1 \|x - x_0\|_{\infty}.$$

For the same reasons as above ϕ is a bounded linear map with induced norm $\|\cdot\|_1$. Therefore the dual of the space c is $l^1(c)$. For the $\phi(x_0)$ part, check the notes on computing duals.

6. Define $p : \mathbb{R}^2 \rightarrow [0, \infty)$ by

$$p(x, y) = \begin{cases} |x| + |y| & xy \geq 0 \\ \sqrt{x^2 + y^2} & \text{otherwise} \end{cases}$$

(a) Show that p is a norm on \mathbb{R}^2 .

Solution. Need to show 3 properties: (i) positive-definiteness, (ii) positive homogeneity and (iii) triangle inequality. Check the notes on computing duals for additional info.

(i) Positive-definiteness follows directly from $|\cdot|$ on \mathbb{R} and $\sqrt{(\cdot)^2 + (\cdot)^2}$ on \mathbb{R}^2 . So, $p(x, y) = 0 \implies x = y = 0$.

(ii) Positive homogeneity follows since

$$\begin{aligned} p(ax, ay) &= \begin{cases} |ax| + |ay| & xy \geq 0 \\ \sqrt{(ax)^2 + (ay)^2} & \text{otherwise} \end{cases} = \begin{cases} |a|(|x| + |y|) & xy \geq 0 \\ |a|\sqrt{x^2 + y^2} & \text{otherwise} \end{cases} \\ &= |a|p(x, y). \end{aligned}$$

(iii) Triangle inequality

$$\begin{aligned} p(x_1 + x_2, y_1 + y_2) &= \begin{cases} |x_1 + x_2| + |y_1 + y_2| & xy \geq 0 \\ \sqrt{(x_1 + x_2)^2 + (y_1 + y_2)^2} & \text{otherwise} \end{cases} \\ &\leq \begin{cases} (|x_1| + |y_1|) + (|x_2| + |y_2|) & xy \geq 0 \\ \sqrt{x_1^2 + y_1^2} + \sqrt{x_2^2 + y_2^2} & \text{otherwise} \end{cases} \\ &= p(x_1, y_1) + p(x_2, y_2) \end{aligned}$$

from properties of $l^1(\mathbb{R}^2, \mathbb{N})$ and $l^2(\mathbb{R}^2, \mathbb{N})$.

$\therefore p$ is a norm.

(b) Determine the dual space (State clearly any identifications you make between spaces of functions in \mathbb{R}^2 .)

Solution. Consider a 2-d linear map $\phi : \mathbb{R}^2 \rightarrow \mathbb{R}$. Steps follow the same as in the previous problem. Let $\mathbf{x} = (x, y)$.

$$|\phi(\mathbf{x})| = \left| \sum_{i=1}^2 \alpha_i x_i \right| \leq \sum_{i=1}^2 |\alpha_i x_i| \leq \begin{cases} \|\alpha\|_\infty \|\mathbf{x}\|_1 & xy \geq 0 \\ \|\alpha\|_2 \|\mathbf{x}\|_2 & \text{otherwise} \end{cases}$$

To show equality, follow the same steps as in the previous exercise. Construct any \mathbf{x}^* s.t. the desired norm (1- or 2- in this case) on the unit ball to get the following

$$\begin{aligned} |\phi(\mathbf{x}^*)| &= \begin{cases} \|\alpha\|_\infty & xy \geq 0 \\ \|\alpha\|_2 & \text{otherwise} \end{cases} \\ \implies \|\phi\|_{\text{ind}} &\geq |\phi(\mathbf{x}^*)| = \begin{cases} \|\alpha\|_\infty & xy \geq 0 \\ \|\alpha\|_2 & \text{otherwise} \end{cases} \end{aligned}$$

Since the object $\alpha \in \mathbb{R}^2$, the dual space is

$$\begin{cases} l^\infty(\mathbb{R}^2) & xy \geq 0 \\ l^2(\mathbb{R}^2) & \text{otherwise} \end{cases}$$