

Exercise Problem Sets 5

Oct. 23, 2020

Problem 1. Complete the following.

1. Show that the p -norm on Euclidean space \mathbb{R}^n given by

$$\|x\|_p \equiv \begin{cases} \left(\sum_{i=1}^n |x_i|^p \right)^{\frac{1}{p}} & \text{if } 1 \leq p < \infty, \\ \max \{|x_1|, \dots, |x_n|\} & \text{if } p = \infty, \end{cases} \quad x = (x_1, \dots, x_n)$$

is indeed a norm on \mathbb{R}^n .

2. Show that for each $1 \leq p, q \leq \infty$ and $p \neq q$, $\|\cdot\|_p$ and $\|\cdot\|_q$ are equivalent norms.

Proof. 1. It suffices to show that $\|\cdot\|_p$ satisfies the triangle inequality, and the case of $p = 1$ and $p = \infty$ is left to the readers. First we prove Hölder's inequality: for $1 < p < \infty$,

$$\sum_{i=1}^n a_i b_i \leq \left(\sum_{i=1}^n |a_i|^p \right)^{\frac{1}{p}} \left(\sum_{i=1}^n |b_i|^{\frac{p}{p-1}} \right)^{\frac{p-1}{p}}.$$

Let $A = \left(\sum_{i=1}^n |a_i|^p \right)^{\frac{1}{p}}$ and $B = \left(\sum_{i=1}^n |b_i|^{\frac{p}{p-1}} \right)^{\frac{p-1}{p}}$. It suffices to show that

$$\sum_{i=1}^n \frac{a_i b_i}{A B} \leq 1.$$

By Young's inequality $ab \leq \frac{1}{p}a^p + \frac{p-1}{p}b^{\frac{p}{p-1}}$ for all $a, b \geq 0$, we find that

$$\frac{a_i b_i}{A B} \leq \frac{1}{p} \left(\frac{|a_i|}{A} \right)^p + \frac{p-1}{p} \left(\frac{|b_i|}{B} \right)^{\frac{p}{p-1}};$$

thus

$$\sum_{i=1}^n \frac{a_i b_i}{A B} \leq \frac{1}{p} \frac{1}{A^p} \sum_{i=1}^n |a_i|^p + \frac{p-1}{p} \frac{1}{B^{\frac{p}{p-1}}} \sum_{i=1}^n |b_i|^{\frac{p}{p-1}} = \frac{1}{p} + \frac{p-1}{p} = 1.$$

Having established Hölder's inequality, we find that

$$\begin{aligned} \|\mathbf{x} + \mathbf{y}\|_p^p &= \sum_{i=1}^n |x_i + y_i|^p \leq \sum_{i=1}^n |x_i + y_i|^{p-1} |x_i| + \sum_{i=1}^n |x_i + y_i|^{p-1} |y_i| \\ &\leq \left[\sum_{i=1}^n (|x_i + y_i|^{p-1})^{\frac{p}{p-1}} \right]^{\frac{p-1}{p}} \left(\sum_{i=1}^n |x_i|^p \right)^{\frac{1}{p}} + \left[\sum_{i=1}^n (|x_i + y_i|^{p-1})^{\frac{p}{p-1}} \right]^{\frac{p-1}{p}} \left(\sum_{i=1}^n |y_i|^p \right)^{\frac{1}{p}} \\ &= \left(\sum_{i=1}^n |x_i + y_i|^p \right)^{\frac{p-1}{p}} (\|\mathbf{x}\|_p + \|\mathbf{y}\|_p) = \|\mathbf{x} + \mathbf{y}\|_p^{p-1} (\|\mathbf{x}\|_p + \|\mathbf{y}\|_p). \end{aligned}$$

Therefore, $\|\mathbf{x} + \mathbf{y}\|_p \leq \|\mathbf{x}\|_p + \|\mathbf{y}\|_p$.

2. It suffices to show that every p -norm is equivalent to the ∞ -norm since if so, then for all $1 \leq p, q < \infty$ there exist C_1, C_2, C_3, C_4 such that

$$C_1 \|\mathbf{x}\|_p \leq \|\mathbf{x}\|_\infty \leq C_2 \|\mathbf{x}\|_p \quad \text{and} \quad C_3 \|\mathbf{x}\|_q \leq \|\mathbf{x}\|_\infty \leq C_4 \|\mathbf{x}\|_q \quad \forall \mathbf{x} \in \mathbb{R}^n.$$

Therefore,

$$\frac{C_1}{C_4} \|\mathbf{x}\|_p \leq \|\mathbf{x}\|_q \leq \frac{C_2}{C_3} \|\mathbf{x}\|_p \quad \forall \mathbf{x} \in \mathbb{R}^n.$$

Now we show that each p -norm is equivalent to the ∞ -norm. Note that

$$\|\mathbf{x}\|_\infty \leq \|\mathbf{x}\|_p \quad \forall 1 \leq p \leq \infty.$$

On the other hand,

$$\|\mathbf{x}\|_p = \left(\sum_{i=1}^n |x_i|^p \right)^{\frac{1}{p}} \leq \left(\sum_{i=1}^n \|\mathbf{x}\|_\infty^p \right)^{\frac{1}{p}} \leq n^{\frac{1}{p}} \|\mathbf{x}\|_\infty.$$

Therefore,

$$n^{-\frac{1}{p}} \|\mathbf{x}\|_p \leq \|\mathbf{x}\|_\infty \leq \|\mathbf{x}\|_p \quad \forall \mathbf{x} \in \mathbb{R}^n \text{ and } 1 \leq p \leq \infty. \quad \square$$

Problem 2. Complete the following.

1. For $f \in \mathcal{C}([a, b]; \mathbb{R})$, define

$$\|f\|_p = \begin{cases} \left(\int_a^b |f(x)|^p dx \right)^{\frac{1}{p}} & \text{if } 1 \leq p < \infty, \\ \max_{x \in [a, b]} |f(x)| & \text{if } p = \infty. \end{cases}$$

Show that $\|\cdot\|_p$ is a norm on $\mathcal{C}([a, b]; \mathbb{R})$.

2. Are $\|\cdot\|_p$ and $\|\cdot\|_q$ equivalent norms on $\mathcal{C}([a, b]; \mathbb{R})$ for any $1 \leq p, q \leq \infty$?

Proof. 1. For a continuous function $h : [a, b] \rightarrow \mathbb{R}$,

$$\int_a^b h(x) dx = \lim_{n \rightarrow \infty} \sum_{i=1}^n h\left(a + i \frac{b-a}{n}\right) \frac{b-a}{n}.$$

Therefore, with c_i and d_i denoting $f\left(a + i \frac{b-a}{n}\right)$ and $g\left(a + i \frac{b-a}{n}\right)$, respectively, we have

$$\|f + g\|_p = \lim_{n \rightarrow \infty} \left(\sum_{i=1}^n \left| (f + g)\left(a + i \frac{b-a}{n}\right) \right|^p \frac{b-a}{n} \right)^{\frac{1}{p}} = (b-a)^{\frac{1}{p}} \lim_{n \rightarrow \infty} \left[n^{-\frac{1}{p}} \left(\sum_{i=1}^n |c_i + d_i|^p \right)^{\frac{1}{p}} \right],$$

and similarly,

$$\|f\|_p = (b-a)^{\frac{1}{p}} \lim_{n \rightarrow \infty} \left[n^{-\frac{1}{p}} \left(\sum_{i=1}^n |c_i|^p \right)^{\frac{1}{p}} \right], \quad \|g\|_p = (b-a)^{\frac{1}{p}} \lim_{n \rightarrow \infty} \left[n^{-\frac{1}{p}} \left(\sum_{i=1}^n |d_i|^p \right)^{\frac{1}{p}} \right].$$

By Minkowski's inequality in Problem 1,

$$n^{-\frac{1}{p}} \left(\sum_{i=1}^n |c_i + d_i|^p \right)^{\frac{1}{p}} \leq n^{-\frac{1}{p}} \left(\sum_{i=1}^n |c_i|^p \right)^{\frac{1}{p}} + n^{-\frac{1}{p}} \left(\sum_{i=1}^n |d_i|^p \right)^{\frac{1}{p}};$$

thus the desired conclusion follows from passing to the limit as $n \rightarrow \infty$.

2. The 1-norm and the ∞ -norm are not equivalent. For each $n \in \mathbb{N}$, consider the function $f_n : [0, 1] \rightarrow \mathbb{R}$ defined by

$$f_n(x) = \begin{cases} -n^2x + n & \text{if } 0 \leq x \leq \frac{1}{n}, \\ 0 & \text{otherwise.} \end{cases}$$

Then $\|f_n\|_1 = \frac{1}{2}$ but $\|f_n\|_\infty = n$. Therefore,

$$\frac{\|f_n\|_\infty}{\|f_n\|_1} = 2n$$

which does not belong to any given bounded interval $[C_1, C_2]$ when n is large. In fact, any p -norm and q -norm cannot be equivalent since for every $n > 0$ one can also find a function $f : [0, 1] \rightarrow \mathbb{R}$ such that $\|f\|_p = 1$ and $\|f\|_q > n$ if $p < q$. \square

Problem 3. Let $\mathcal{M}_{n \times m}$ be the collection of all $n \times m$ real matrices. Define a function $\|\cdot\|_{p,q} : \mathcal{M}_{n \times m} \rightarrow \mathbb{R}$ by

$$\|A\|_{p,q} = \sup_{\|\mathbf{x}\|_p=1} \|A\mathbf{x}\|_q.$$

here we recall that $\|\cdot\|_p$ is the p -norm on Euclidean space. If $p = q$, we simply use $\|A\|_p$ to denote $\|A\|_{p,q}$. Complete the following.

1. Show that $\|A\|_{p,q} = \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_q}{\|\mathbf{x}\|_p}$ for all $p, q \geq 1$.
2. Show that $\|A\|_{p,q} = \inf \{M \in \mathbb{R} \mid \|A\mathbf{x}\|_q \leq M\|\mathbf{x}\|_p \ \forall \mathbf{x} \in \mathbb{R}^m\}$.
3. $\|A\mathbf{x}\|_q \leq \|A\|_{p,q}\|\mathbf{x}\|_p$ for all $\mathbf{x} \in \mathbb{R}^m$.
4. Let $\{A_k\}_{k=1}^\infty \subseteq \mathcal{M}_{n \times m}$, and $p, q \geq 1$ be given. Show that $\lim_{k \rightarrow \infty} \|A_k\|_{p,q} = 0$ if and only if each entry of A_k converges to 0. In other words, by writing $A_k = [a_{ij}^{(k)}]_{1 \leq i \leq n, 1 \leq j \leq m}$, show that $\lim_{k \rightarrow \infty} \|A_k\|_{p,q} = 0$ if and only if $\lim_{k \rightarrow \infty} a_{ij}^{(k)} = 0$ for all $1 \leq i \leq m, 1 \leq j \leq n$.

Proof. 1. If $\mathbf{x} \neq \mathbf{0}$, then $\mathbf{y} = \frac{\mathbf{x}}{\|\mathbf{x}\|_p}$ satisfies that $\|\mathbf{y}\|_p = 1$; thus if $\mathbf{x} \neq \mathbf{0}$,

$$\frac{\|A\mathbf{x}\|_q}{\|\mathbf{x}\|_p} = \|A\mathbf{y}\|_q \leq \sup_{\|\mathbf{x}\|_p=1} \|A\mathbf{x}\|_q = \|A\|_{p,q}.$$

Therefore, $\sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_q}{\|\mathbf{x}\|_p} \leq \|A\|_{p,q}$.

On the other hand, if $\|\mathbf{x}\|_p = 1$, then $\mathbf{x} \neq \mathbf{0}$; thus if $\|\mathbf{x}\|_p = 1$,

$$\|A\mathbf{x}\|_q = \frac{\|A\mathbf{x}\|_q}{\|\mathbf{x}\|_p} \leq \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_q}{\|\mathbf{x}\|_p}.$$

Therefore, $\|A\|_{p,q} = \sup_{\|\mathbf{x}\|_p=1} \|A\mathbf{x}\|_q \leq \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_q}{\|\mathbf{x}\|_p}$.

2. 2 follows from Problem 3 in Exercise 2.

3. By 1, $\frac{\|A\mathbf{x}\|_q}{\|\mathbf{x}\|_p} \leq \|A\|_{p,q}$ for all $\mathbf{x} \neq \mathbf{0}$ or equivalently,

$$\|A\mathbf{x}\|_q \leq \|A\|_{p,q} \|\mathbf{x}\|_p \quad \forall \mathbf{x} \neq \mathbf{0}.$$

Since the inequality above also holds for $\mathbf{x} = \mathbf{0}$, we conclude that

$$\|A\mathbf{x}\|_q \leq \|A\|_{p,q} \|\mathbf{x}\|_p \quad \forall \mathbf{x} \in \mathbb{R}^m.$$

4. Let $B = [b_{ij}] \in M_{n \times m}$, and $|b_{k\ell}| = \max_{1 \leq i \leq n, 1 \leq j \leq m} |b_{ij}|$; that is, the maximum of the absolute value of entries of B occurs at the (k, ℓ) -entry. Let \mathbf{e}_ℓ be the unit vector whose ℓ -th component is 1. Since $B\mathbf{e}_\ell$ is the ℓ -th column of B , for $1 \leq i \leq n$ and $1 \leq j \leq m$,

$$|b_{ij}| \leq |b_{k\ell}| \leq \|B\mathbf{e}_\ell\|_q \leq \|B\|_{p,q} \|\mathbf{e}_\ell\|_p = \|B\|_{p,q};$$

thus

$$|b_{ij}| \leq \|B\|_{p,q} \quad \forall 1 \leq i \leq n, 1 \leq j \leq m. \quad (\star)$$

On the other hand, there exists $\mathbf{x} \in \mathbb{R}^m$ such that $\|\mathbf{x}\|_p = 1$ and $\|B\mathbf{x}\|_q \geq \frac{\|B\|_{p,q}}{2}$. Therefore, if $1 \leq q < \infty$,

$$\begin{aligned} \frac{\|B\|_{p,q}}{2} &\leq \|B\mathbf{x}\|_q = \left(\sum_{i=1}^n \left| \sum_{j=1}^m b_{ij}x_j \right|^q \right)^{\frac{1}{q}} \leq \left[\sum_{i=1}^n \left(\sum_{j=1}^m |b_{ij}| \right)^q \right]^{\frac{1}{q}} \leq m \left[\sum_{i=1}^n \left(\frac{1}{m} \sum_{j=1}^m |b_{ij}| \right)^q \right]^{\frac{1}{q}} \\ &\leq m \left(\sum_{i=1}^n \frac{1}{m} \sum_{j=1}^m |b_{ij}|^q \right)^{\frac{1}{q}} \leq m^{1-\frac{1}{q}} \left(\sum_{i=1}^n \sum_{j=1}^m |b_{ij}|^q \right)^{\frac{1}{q}} \leq m \left(\sum_{i=1}^n \sum_{j=1}^m |b_{ij}|^q \right)^{\frac{1}{q}}, \end{aligned}$$

while if $q = \infty$,

$$\frac{\|B\|_{p,q}}{2} \leq \|B\mathbf{x}\|_\infty = \max_{1 \leq i \leq n} \left| \sum_{j=1}^m b_{ij}x_j \right| \leq \max_{1 \leq i \leq n} \sum_{j=1}^m |b_{ij}| \leq \sum_{i=1}^n \sum_{j=1}^m |b_{ij}|.$$

In either cases, we conclude that

$$\|B\|_{p,q} \leq f(|b_{11}|, |b_{12}|, \dots, |b_{nm}|) \quad (\diamond)$$

for some function f of nm variables satisfying that $f(\mathbf{y}) \rightarrow 0$ as $\mathbf{y} \rightarrow 0$.

(\Rightarrow) Using (\star), we find that for each $1 \leq i \leq n$ and $1 \leq j \leq m$,

$$0 \leq |a_{ij}^{(k)}| \leq \|A_k\|_{p,q}.$$

Since $\lim_{k \rightarrow \infty} \|A_k\|_{p,q} = 0$, by the Sandwich Lemma we conclude that

$$\lim_{k \rightarrow \infty} |a_{ij}^{(k)}| = 0 \quad \forall 1 \leq i \leq n, 1 \leq j \leq m.$$

(\Leftarrow) Suppose that $\lim_{k \rightarrow \infty} |a_{ij}^{(k)}| = 0$ for all $1 \leq i \leq n, 1 \leq j \leq m$. Then (\diamond) implies that

$$0 \leq \|A_k\|_{p,q} \leq f(|a_{11}^{(k)}|, |a_{12}^{(k)}|, \dots, |a_{nm}^{(k)}|) \quad (\diamond)$$

for some function f of nm variables satisfying that $f(\mathbf{y}) \rightarrow 0$ as $\mathbf{y} \rightarrow 0$. Therefore, the

Sandwich Lemma implies that $\lim_{k \rightarrow \infty} \|A_k\|_{p,q} = 0$. □

Problem 4. Show that

$$\|A\|_1 = \max \left\{ \sum_{i=1}^n |a_{i1}|, \sum_{i=1}^n |a_{i2}|, \dots, \sum_{i=1}^n |a_{im}| \right\}.$$

Hint: Use Problem 4 and 5 of Exercise 2.

Proof. By Problem 5 of Exercise 2,

$$\|\mathbf{x}\|_1 = \sup_{\|\mathbf{y}\|_\infty=1} \mathbf{x} \cdot \mathbf{y} \quad \text{and} \quad \|\mathbf{y}\|_\infty = \sup_{\|\mathbf{x}\|_1=1} \mathbf{x} \cdot \mathbf{y},$$

where $\mathbf{x} \cdot \mathbf{y}$ denotes the standard inner product of \mathbf{x} and \mathbf{y} in the Euclidean space. Therefore,

$$\|A\|_1 = \sup_{\|\mathbf{x}\|_1=1} \|A\mathbf{x}\|_1 = \sup_{\|\mathbf{x}\|_1=1} \sup_{\|\mathbf{y}\|_\infty=1} (A\mathbf{x}) \cdot \mathbf{y} = \sup_{\|\mathbf{x}\|_1=1} \sup_{\|\mathbf{y}\|_\infty=1} \mathbf{x} \cdot (A^T \mathbf{y}),$$

and Problem 4 of Exercise further implies that

$$\|A\|_1 = \sup_{\|\mathbf{y}\|_\infty=1} \sup_{\|\mathbf{x}\|_1=1} (A^T \mathbf{y}) \cdot \mathbf{x} = \sup_{\|\mathbf{y}\|_\infty=1} \|A^T \mathbf{y}\|_\infty = \|A^T\|_\infty.$$

By the fact that the ∞ -norm of an $n \times m$ real matrix is the maximum of the sum of the absolute value of entries of row vectors, we find that

$$\|A\|_1 = \|A^T\|_\infty = \max \left\{ \sum_{i=1}^n |a_{i1}|, \sum_{i=1}^n |a_{i2}|, \dots, \sum_{i=1}^n |a_{im}| \right\}. \quad \square$$

Alternative proof. Let $\mathbf{x} = (x_1, \dots, x_m) \in \mathbb{R}^m$ and $\|\mathbf{x}\|_1 = 1$. Then for $A = [a_{ij}] \in \mathcal{M}_{n \times m}$, we have

$$\begin{aligned} \|A\mathbf{x}\|_1 &= \sum_{i=1}^n \left| \sum_{j=1}^m a_{ij} x_j \right| \leq \sum_{i=1}^n \sum_{j=1}^m |a_{ij}| |x_j| = \sum_{j=1}^m \sum_{i=1}^n |a_{ij}| |x_j| = \sum_{j=1}^m |x_j| \left(\sum_{i=1}^n |a_{ij}| \right) \\ &\leq \sum_{j=1}^m |x_j| \left(\max_{1 \leq j \leq m} \sum_{i=1}^n |a_{ij}| \right) = \left(\max_{1 \leq j \leq m} \sum_{i=1}^n |a_{ij}| \right) \sum_{j=1}^m |x_j| = \left(\max_{1 \leq j \leq m} \sum_{i=1}^n |a_{ij}| \right) \|\mathbf{x}\|_1 \\ &= \max_{1 \leq j \leq m} \sum_{i=1}^n |a_{ij}|. \end{aligned}$$

Therefore, $\|A\|_1 = \sup_{\|\mathbf{x}\|_1=1} \|A\mathbf{x}\|_1 \leq \max_{1 \leq j \leq m} \sum_{i=1}^n |a_{ij}|$.

On the other hand, suppose that $\max_{1 \leq j \leq m} \sum_{i=1}^n |a_{ij}| = \sum_{i=1}^n |a_{ik}|$; that is, the maximum of the sum of absolute value of column entries of A occurs at the k -th column. Let $\mathbf{x} = (x_1, \dots, x_m) \in \mathbb{R}^m$ be defined by

$$x_j = \begin{cases} 0 & \text{if } j \neq k, \\ 1 & \text{if } j = k. \end{cases}$$

Then

$$\|A\mathbf{x}\|_1 = \sum_{i=1}^n \left| \sum_{j=1}^m a_{ij} x_j \right| = \sum_{i=1}^n |a_{ik}| = \max_{1 \leq j \leq m} \sum_{i=1}^n |a_{ij}|;$$

thus $\|A\|_1 = \sup_{\|\mathbf{x}\|_1=1} \|A\mathbf{x}\|_1 \geq \max_{1 \leq j \leq m} \sum_{i=1}^n |a_{ij}|$. □

Problem 5. Let $(X, \|\cdot\|_X)$, $(Y, \|\cdot\|_Y)$, $(Z, \|\cdot\|_Z)$ be three normed vector spaces such that $X, Y \subseteq Z$ and

$$\|\mathbf{x}\|_Z \leq C\|\mathbf{x}\|_X \quad \forall \mathbf{x} \in X \quad \text{and} \quad \|\mathbf{y}\|_Z \leq C\|\mathbf{y}\|_Y \quad \forall \mathbf{y} \in Y.$$

Define

$$E = \{\mathbf{a} \in Z \mid \|\mathbf{a}\|_E \equiv \max\{\|\mathbf{a}\|_X, \|\mathbf{a}\|_Y\} < \infty\}$$

and

$$F = \{\mathbf{a} \in Z \mid \|\mathbf{a}\|_F \equiv \inf_{\substack{\mathbf{a}=\mathbf{x}+\mathbf{y} \\ \mathbf{x} \in X, \mathbf{y} \in Y}} (\|\mathbf{x}\|_X + \|\mathbf{y}\|_Y) < \infty\}.$$

1. Show that $(E, \|\cdot\|_E)$ is a normed vector space, and $E = X \cap Y$.
2. Show that $(F, \|\cdot\|_F)$ is a normed vector space. The space F is usually denoted by $X + Y$.

Proof. We note that $E, F \subseteq Z$, to show that E and F are vector spaces it suffices to show that E and F are vector subspaces of Z .

1. The case of E : Let $\mathbf{a}, \mathbf{b} \in E$, and $\lambda \in \mathbb{F}$. Then

$$\max\{\|\mathbf{a}\|_X, \|\mathbf{a}\|_Y\} < \infty;$$

thus

$$\max\{\|\lambda\mathbf{a}\|_X, \|\lambda\mathbf{a}\|_Y\} = |\lambda| \max\{\|\mathbf{a}\|_X, \|\mathbf{a}\|_Y\} < \infty \quad (\star)$$

which shows that

$$\lambda\mathbf{a} \in E \quad \forall \lambda \in \mathbb{F} \text{ and } \mathbf{a} \in E. \quad (\diamond)$$

Moreover,

$$\|\mathbf{a} + \mathbf{b}\|_X \leq \|\mathbf{a}\|_X + \|\mathbf{b}\|_X \quad \text{and} \quad \|\mathbf{a} + \mathbf{b}\|_Y \leq \|\mathbf{a}\|_Y + \|\mathbf{b}\|_Y$$

which implies that

$$\begin{aligned} \max\{\|\mathbf{a} + \mathbf{b}\|_X, \|\mathbf{a} + \mathbf{b}\|_Y\} &\leq \max\{\|\mathbf{a}\|_X + \|\mathbf{b}\|_X, \|\mathbf{a}\|_Y + \|\mathbf{b}\|_Y\} \\ &\leq \max\{\|\mathbf{a}\|_X, \|\mathbf{a}\|_Y\} + \max\{\|\mathbf{b}\|_X, \|\mathbf{b}\|_Y\} < \infty. \end{aligned} \quad (\star\star)$$

Therefore,

$$\mathbf{a} + \mathbf{b} \in E \quad \forall \mathbf{a}, \mathbf{b} \in E. \quad (\diamond\diamond)$$

Combining (\diamond) and $(\diamond\diamond)$, we conclude that

$$\lambda\mathbf{a} + \mu\mathbf{b} \in E \quad \forall \lambda, \mu \in \mathbb{F}, \mathbf{a}, \mathbf{b} \in E;$$

thus Lemma 2.9 shows that E is a subspace of Z .

2. The case of F : Let $\mathbf{a}, \mathbf{b} \in F$ and $\lambda \in \mathbb{F}$. Then there exists $\mathbf{x}_1, \mathbf{x}_2 \in X$, $\mathbf{y}_1, \mathbf{y}_2 \in Y$ such that

$$\begin{aligned} \mathbf{a} &= \mathbf{x}_1 + \mathbf{y}_1 & \text{and} & \quad \|\mathbf{x}_1\|_X + \|\mathbf{y}_1\|_Y \leq \|\mathbf{a}\|_F + 1, \\ \mathbf{b} &= \mathbf{x}_2 + \mathbf{y}_2 & \text{and} & \quad \|\mathbf{x}_2\|_X + \|\mathbf{y}_2\|_Y \leq \|\mathbf{b}\|_F + 1. \end{aligned}$$

Therefore,

$$\lambda \mathbf{a} = \lambda \mathbf{x}_1 + \lambda \mathbf{y}_1 \quad \text{and} \quad \|\lambda \mathbf{x}_1\|_X + \|\lambda \mathbf{y}_1\|_Y = |\lambda|(\|\mathbf{x}_1\|_X + \|\mathbf{y}_1\|_Y) < \lambda(\|\mathbf{a}\| + 1) < \infty$$

which implies that

$$\lambda \mathbf{a} \in F \quad \forall \lambda \in \mathbb{F} \text{ and } \mathbf{a} \in F.$$

Moreover, with \mathbf{x} and \mathbf{y} denoting $\mathbf{x}_1 + \mathbf{x}_2$ and $\mathbf{y}_1 + \mathbf{y}_2$, respectively, we find that $\mathbf{x} \in X$, $\mathbf{y} \in Y$, $\mathbf{a} + \mathbf{b} = \mathbf{x} + \mathbf{y}$ and

$$\|\mathbf{x}\|_X + \|\mathbf{y}\|_Y \leq \|\mathbf{x}_1\|_X + \|\mathbf{x}_2\|_X + \|\mathbf{y}_1\|_Y + \|\mathbf{y}_2\|_Y < \|\mathbf{a}\|_F + \|\mathbf{b}\|_F + 2 < \infty.$$

This implies that

$$\mathbf{a} + \mathbf{b} \in F \quad \forall \mathbf{a}, \mathbf{b} \in F.$$

Similar to the case of E , by Lemma 2.9 we conclude that F is a subspace of Z .

Next we show that $\|\cdot\|_E$ and $\|\cdot\|_F$ defined in the problem are indeed norms on E and F , respectively. It is clear that $\|\cdot\|_E$ and $\|\cdot\|_F$ satisfy Property (a) in the definition of the norm vector space, so we only prove Property (b)-(d).

1. The case of E :

(b) By the definition of $\|\cdot\|_E$,

$$\|\mathbf{a}\|_E = 0 \Leftrightarrow \max\{\|\mathbf{a}\|_X, \|\mathbf{a}\|_Y\} = 0 \Leftrightarrow \|\mathbf{a}\|_X = \|\mathbf{a}\|_Y = 0 \Leftrightarrow \mathbf{a} = \mathbf{0}.$$

(c) Let $\lambda \in \mathbb{F}$ and $\mathbf{a} \in E$ be given. Then (\star) implies that $\|\lambda \mathbf{a}\|_E = |\lambda| \|\mathbf{a}\|_E$.

(d) Let $\mathbf{a}, \mathbf{b} \in E$. Then $(\star\star)$ implies that $\|\mathbf{a} + \mathbf{b}\|_E \leq \|\mathbf{a}\|_E + \|\mathbf{b}\|_E$.

Finally, $\mathbf{a} \in E$ if and only if $\|\mathbf{a}\|_X < \infty$ and $\|\mathbf{a}\|_Y < \infty$; thus $\mathbf{a} \in E$ if and only if $\mathbf{a} \in X$ and $\mathbf{a} \in Y$. This shows that $E = X \cap Y$.

2. The case of F :

(b) Since $\mathbf{0} = \mathbf{0} + \mathbf{0}$,

$$\|\mathbf{0}\|_F = \inf_{\substack{\mathbf{a}=\mathbf{x}+\mathbf{y} \\ \mathbf{x} \in X, \mathbf{y} \in Y}} \|\mathbf{x}\|_X + \|\mathbf{y}\|_Y \leq \|\mathbf{0}\|_X + \|\mathbf{0}\|_Y = 0.$$

Suppose that $\|\mathbf{a}\|_F = 0$. For each $n \in \mathbb{N}$, there exists $\mathbf{x}_n \in X$ and $\mathbf{y}_n \in Y$ such that $\mathbf{a} = \mathbf{x}_n + \mathbf{y}_n$ and

$$\|\mathbf{x}_n\|_X + \|\mathbf{y}_n\|_Y < \frac{1}{n}.$$

The inequality above implies that $\mathbf{x}_n \rightarrow \mathbf{0}$ and $\mathbf{y}_n \rightarrow \mathbf{0}$ as $n \rightarrow \infty$; thus

$$\mathbf{a} = \lim_{n \rightarrow \infty} (\mathbf{x}_n + \mathbf{y}_n) = \mathbf{0}.$$

(c) Let $\lambda \in \mathbb{F}$, $\mathbf{a} \in E$, and $\varepsilon > 0$ be given. W.L.O.G. we can assume that $\lambda \neq 0$. Then there exists $\mathbf{x}_1 \in X$ and $\mathbf{y}_1 \in Y$ such that

$$\mathbf{a} = \mathbf{x}_1 + \mathbf{y}_1 \quad \text{and} \quad \|\mathbf{x}_1\|_X + \|\mathbf{y}_1\|_Y < \|\mathbf{a}\|_F + \frac{1}{|\lambda|}\varepsilon$$

Then $\lambda\mathbf{a} = \lambda\mathbf{x}_1 + \lambda\mathbf{y}_1$, $\lambda\mathbf{x}_1 \in X$, $\lambda\mathbf{y}_1 \in Y$ and

$$\|\lambda\mathbf{a}\|_F \leq \|\lambda\mathbf{x}_1\|_X + \|\lambda\mathbf{y}_1\|_Y = |\lambda|(\|\mathbf{x}_1\|_X + \|\mathbf{y}_1\|_Y) < |\lambda|\|\mathbf{a}\|_F + \varepsilon.$$

Since $\varepsilon > 0$ is given arbitrarily, we find that

$$\|\lambda\mathbf{a}\|_F \leq |\lambda|\|\mathbf{a}\|_F \quad \forall \lambda \in \mathbb{F} \text{ and } \mathbf{a} \in F.$$

On the other hand, by the fact that $\lambda\mathbf{a} \in F$, there exists $\mathbf{x}_2 \in X$, $\mathbf{y}_2 \in Y$ such that

$$\lambda\mathbf{a} = \mathbf{x}_2 + \mathbf{y}_2 \quad \text{and} \quad \|\lambda\mathbf{a}\|_F \leq \|\mathbf{x}_2\|_X + \|\mathbf{y}_2\|_Y + |\lambda|\varepsilon.$$

Then $\mathbf{a} = \lambda^{-1}\mathbf{x}_2 + \lambda^{-1}\mathbf{y}_2$, $\lambda^{-1}\mathbf{x}_2 \in X$, $\lambda^{-1}\mathbf{y}_2 \in Y$, and

$$\|\mathbf{a}\|_F \leq \|\lambda^{-1}\mathbf{x}_2\|_X + \|\lambda^{-1}\mathbf{y}_2\|_Y = \frac{1}{|\lambda|}(\|\mathbf{x}_2\|_X + \|\mathbf{y}_2\|_Y) < \frac{1}{|\lambda|}\|\lambda\mathbf{a}\|_F + \varepsilon.$$

Since $\varepsilon > 0$ is given arbitrarily, we conclude that $\|\mathbf{a}\|_E \leq \frac{1}{|\lambda|}\|\lambda\mathbf{a}\|_F$. Therefore,

$$\|\lambda\mathbf{a}\|_F = |\lambda|\|\mathbf{a}\|_F \quad \forall \lambda \in \mathbb{F}, \mathbf{a} \in F.$$

(d) Let $\mathbf{a}, \mathbf{b} \in F$, and $\varepsilon > 0$ be given. There exists $\mathbf{x}_1, \mathbf{x}_2 \in X$ and $\mathbf{y}_1, \mathbf{y}_2 \in Y$ such that

$$\begin{aligned} \mathbf{a} &= \mathbf{x}_1 + \mathbf{y}_1, \|\mathbf{x}_1\|_X + \|\mathbf{y}_1\|_Y < \|\mathbf{a}\|_F + \frac{\varepsilon}{2}, \\ \mathbf{b} &= \mathbf{x}_2 + \mathbf{y}_2, \|\mathbf{x}_2\|_X + \|\mathbf{y}_2\|_Y < \|\mathbf{b}\|_F + \frac{\varepsilon}{2}. \end{aligned}$$

Let $\mathbf{x} = \mathbf{x}_1 + \mathbf{x}_2$ and $\mathbf{y} = \mathbf{y}_1 + \mathbf{y}_2$. Then $\mathbf{x} \in X$ and $\mathbf{y} \in Y$, $\mathbf{a} + \mathbf{b} = \mathbf{x} + \mathbf{y}$, and

$$\|\mathbf{a} + \mathbf{b}\|_F \leq \|\mathbf{x}\|_X + \|\mathbf{y}\|_Y \leq \|\mathbf{x}_1\|_X + \|\mathbf{x}_2\|_X + \|\mathbf{y}_1\|_Y + \|\mathbf{y}_2\|_Y < \|\mathbf{a}\|_F + \|\mathbf{b}\|_F + \varepsilon.$$

Since $\varepsilon > 0$ is given arbitrarily, we find that

$$\|\mathbf{a} + \mathbf{b}\|_F \leq \|\mathbf{a}\|_F + \|\mathbf{b}\|_F \quad \forall \mathbf{a}, \mathbf{b} \in F. \quad \square$$

Problem 6. Show that if $\langle \cdot, \cdot \rangle$ is an inner product on a vector space \mathcal{V} (over a scalar field \mathbb{F}). Then

1. $\langle \lambda\mathbf{v} + \mu\mathbf{w}, \mathbf{u} \rangle = \lambda\langle \mathbf{v}, \mathbf{u} \rangle + \mu\langle \mathbf{w}, \mathbf{u} \rangle$ for all $\mathbf{u}, \mathbf{v}, \mathbf{w} \in \mathcal{V}$.
2. $\langle \mathbf{u}, \lambda\mathbf{v} + \mu\mathbf{w} \rangle = \bar{\lambda}\langle \mathbf{u}, \mathbf{v} \rangle + \bar{\mu}\langle \mathbf{u}, \mathbf{w} \rangle$ for all $\mathbf{u}, \mathbf{v}, \mathbf{w} \in \mathcal{V}$.
3. $\langle \mathbf{v}, \lambda\mathbf{w} \rangle = \bar{\lambda}\langle \mathbf{v}, \mathbf{w} \rangle$ for all $\mathbf{v}, \mathbf{w} \in \mathcal{V}$.
4. $\langle \mathbf{0}, \mathbf{w} \rangle = \langle \mathbf{w}, \mathbf{0} \rangle = 0$ for all $\mathbf{w} \in \mathcal{V}$.

Problem 7. Let (M, d) be a metric space. Show that $\rho : M \times M \rightarrow \mathbb{R}$ defined by

$$\rho(x, y) = \frac{d(x, y)}{1 + d(x, y)}$$

is a metric on M .

Proof. Let $x, y, z \in M$.

1. Since $d(x, y) \geq 0$, we find that $\rho(x, y) \geq 0$.
2. $\rho(x, y) = 0 \Leftrightarrow d(x, y) = 0 \Leftrightarrow x = y$.
3. Since $d(x, y) = d(y, x)$, $\rho(x, y) = \frac{d(x, y)}{1 + d(x, y)} = \frac{d(y, x)}{1 + d(y, x)} = \rho(y, x)$.
4. Let $a = d(x, y)$, $b = d(x, z)$ and $c = d(z, y)$. Since $a \leq b + c$, we find that

$$\begin{aligned} \rho(x, z) + \rho(z, y) - \rho(x, y) &= \frac{b}{1 + b} + \frac{c}{1 + c} - \frac{a}{1 + a} = \frac{(b + c + 2bc)(1 + a) - a(1 + b + c + bc)}{(1 + a)(1 + b)(1 + c)} \\ &= \frac{b + c + 2bc + ab + ac + 2abc - a - ab - ac - abc}{(1 + a)(1 + b)(1 + c)} \\ &= \frac{b + c + 2bc + abc - a}{(1 + a)(1 + b)(1 + c)} \geq 0; \end{aligned}$$

thus ρ satisfies the triangle inequality. □

Problem 8. Let $d : \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$ be defined by

$$d(\mathbf{x}, \mathbf{y}) = \begin{cases} |x_1 - y_1| & \text{if } x_2 = y_2, \\ |x_1 - y_1| + |x_2 - y_2| + 1 & \text{if } x_2 \neq y_2, \end{cases} \text{ where } \mathbf{x} = (x_1, x_2) \text{ and } \mathbf{y} = (y_1, y_2).$$

Show that d is a metric on \mathbb{R}^2 .

Proof. Let $\mathbf{x} = (x_1, x_2)$, $\mathbf{y} = (y_1, y_2)$ and $\mathbf{z} = (z_1, z_2)$ in \mathbb{R}^2 .

1. Clearly $d(\mathbf{x}, \mathbf{y}) \geq 0$.
2. $d(\mathbf{x}, \mathbf{y}) = 0 \Leftrightarrow (x_2 = y_2) \wedge |x_1 - y_1| = 0 \Leftrightarrow (x_2 = y_2) \wedge (x_1 = y_1) \Leftrightarrow \mathbf{x} = \mathbf{y}$.
3. (a) The case $x_2 = y_2$: In this case $d(\mathbf{x}, \mathbf{y}) = |x_1 - y_1|$ and $d(\mathbf{y}, \mathbf{x}) = |y_1 - x_1|$; thus if $x_2 = y_2$ then $d(\mathbf{x}, \mathbf{y}) = d(\mathbf{y}, \mathbf{x})$.
- (b) The case $x_2 \neq y_2$: In this case

$$d(\mathbf{x}, \mathbf{y}) = |x_1 - y_1| + |x_2 - y_2| + 1 \quad \text{and} \quad d(\mathbf{y}, \mathbf{x}) = |y_1 - x_1| + |y_2 - x_2| + 1;$$

thus if $x_2 \neq y_2$ then $d(\mathbf{x}, \mathbf{y}) = d(\mathbf{y}, \mathbf{x})$.

In either cases, we have $d(\mathbf{x}, \mathbf{y}) = d(\mathbf{y}, \mathbf{x})$.

4. (a) The case $x_2 = y_2$: In this case

$$d(\mathbf{x}, \mathbf{y}) = |x_1 - y_1| \leq |x_1 - z_1| + |z_1 - y_1| \leq d(\mathbf{x}, \mathbf{z}) + d(\mathbf{z}, \mathbf{y}).$$

(b) The case $x_2 \neq y_2$: In this case z_2 is different from at least one of the second component x_2, y_2 . W.L.O.G. we assume that $z_2 \neq x_2$. Then

$$\begin{aligned}d(\mathbf{x}, \mathbf{y}) &= |x_1 - y_1| + |x_2 - y_2| + 1 \leq |x_1 - z_1| + |z_1 - y_1| + |x_2 - z_2| + |z_2 - y_2| + 1 \\ &= d(\mathbf{x}, \mathbf{z}) + |z_1 - y_1| + |z_2 - y_2| \leq d(\mathbf{x}, \mathbf{z}) + d(\mathbf{z}, \mathbf{y}).\end{aligned}$$

In either cases, d satisfies the triangle inequality. □