

Exercise 1. Show that

$$d(x, y) = \left| \frac{1}{x} - \frac{1}{y} \right|$$

defines a metric on $(0, \infty)$.

Proof. Clearly, $d(x, y) \geq 0$ for all $x, y \in (0, \infty)$. If $d(x, y) = 0$, then $|\frac{1}{x} - \frac{1}{y}| = 0 \Rightarrow x = y$. Moreover, we have

$$d(x, y) = \left| \frac{1}{x} - \frac{1}{y} \right| = \left| \frac{1}{y} - \frac{1}{x} \right|$$

and

$$d(x, y) + d(y, z) = \left| \frac{1}{x} - \frac{1}{y} \right| + \left| \frac{1}{y} - \frac{1}{z} \right| \geq \left| \frac{1}{x} - \frac{1}{z} \right| = d(x, z).$$

Thus d is a metric on $(0, \infty)$. □

Exercise 2. If d is a metric on M , show that $|d(x, z) - d(y, z)| \leq d(x, y)$ for any $x, y, z \in M$.

Proof. Since d is a metric on M , we have $d(x, z) \leq d(x, y) + d(y, z)$, hence $d(x, z) - d(y, z) \leq d(x, y)$. Moreover, we have $d(y, z) \leq d(y, x) + d(x, z)$, hence $-d(x, y) \leq d(x, z) - d(y, z)$. Thus $|d(x, z) - d(y, z)| \leq d(x, y)$. □

Exercise 3. As it happens, some of our requirements for a metric are redundant. To see why this is so, let M be a set and suppose that $d : M \times M \rightarrow \mathbb{R}$ satisfies $d(x, y) = 0$ if and only if $x = y$, and $d(x, y) \leq d(x, z) + d(y, z)$ (*) for all $x, y, z \in M$. Prove that d is a metric.

Proof. First, we will prove that $d(x, y) = d(y, x)$. Indeed, let $z = x$ in (*), we get

$$d(x, y) \leq d(x, x) + d(y, x) = d(y, x).$$

Similarly, we will get $d(y, x) \leq d(x, y)$. Thus $d(x, y) = d(y, x)$. Notice that

$$0 = d(x, x) \leq d(x, y) + d(x, y) = 2d(x, y),$$

thus $0 \leq d(x, y)$ for all $x, y \in M$. Thus d is a metric. □

Exercise 6. If d is a metric on M , show that $\rho(x, y) = \sqrt{d(x, y)}$, $\sigma(x, y) = \frac{d(x, y)}{1+d(x, y)}$, and $\tau(x, y) = \min\{d(x, y), 1\}$ are also metrics on M .

Proof.

- $\rho(x, y)$.

Clearly, we have $\rho(x, y) \geq 0$. If $\rho(x, y) = 0$, then $\sqrt{d(x, y)} = 0$, which leads to $d(x, y) = 0$. Thus $x = y$. Otherwise, since $d(x, x) = 0$, it's not hard to check that $\rho(x, x) = 0$. Moreover, we have $\rho(x, y) = \sqrt{d(x, y)} = \sqrt{d(y, x)} = \rho(y, x)$. Also

notice that for $a, b \geq 0$, we have $\sqrt{x+y} \leq \sqrt{x} + \sqrt{y}$. Indeed, since $2\sqrt{xy} \geq 0$, we have

$$\begin{aligned} x+y &\leq x+y+2\sqrt{xy} \\ \Leftrightarrow (\sqrt{x+y})^2 &\leq (\sqrt{x} + \sqrt{y})^2 \\ \Leftrightarrow |\sqrt{x+y}| &\leq |\sqrt{x} + \sqrt{y}| \\ \Leftrightarrow \sqrt{x+y} &\leq \sqrt{x} + \sqrt{y}. \end{aligned}$$

Thus

$$\rho(x, z) = \sqrt{d(x, z)} \leq \sqrt{d(x, y) + d(y, z)} \leq \sqrt{d(x, y)} + \sqrt{d(y, z)} = \rho(x, y) + \rho(y, z).$$

Thus ρ defines a metrics on M .

- $\sigma(x, y)$.

Because $d(x, y) \geq 0$, we have $\sigma(x, y) \geq 0$ for all $x, y \in M$. If $\sigma(x, y) = 0$, then $\frac{d(x, y)}{1+d(x, y)} = 0$, thus $d(x, y) = 0$, which means $x = y$. Otherwise, since $d(x, x) = 0$, we can easily check that $\sigma(x, x) = 0$ too. Moreover,

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

Now, we have a small lemma as follow.

Lemma . Let $a, b, c \geq 0$, if $c \leq a + b$, then

$$\frac{c}{1+c} \leq \frac{a}{1+a} + \frac{b}{1+b}.$$

Proof. Indeed, we have

$$\begin{aligned} &\frac{c}{1+c} \leq \frac{a}{1+a} + \frac{b}{1+b} \\ \Leftrightarrow &\frac{c}{1+c} \leq \frac{a+b+2ab}{1+a+b+ab} \\ \Leftrightarrow &1 - \frac{1}{1+c} \leq 1 - \frac{1-ab}{1+a+b+ab} \\ \Leftrightarrow &\frac{1-ab}{1+a+b+ab} \leq \frac{1}{1+c}. \end{aligned} \quad (*)$$

If $1 - ab \leq 0$, then the left side is smaller or equal then 0, when the right side is larger or equal then 0. Thus the inequality is proved.

If $1 - ab \geq 0$, then

$$\begin{aligned} (*) \Leftrightarrow &1+c \leq \frac{1+a+b+ab}{1-ab} \\ \Leftrightarrow &1+c \leq 1 + \frac{a+b+2ab}{1-ab} \\ \Leftrightarrow &c \leq \frac{a+b+2ab}{1-ab} \\ \Leftrightarrow &c - abc \leq a+b+2ab. \end{aligned}$$

However, we have $c \leq a + b$ and $-abc \leq 0 \leq 2ab$, thus the lemma is proved. \square

Now let $a = d(x, z), b = d(x, y), c = d(y, z)$ in the lemma, we have $\sigma(x, z) \leq \sigma(x, y) + \sigma(y, z)$. Thus σ defines a metric on M .

- $\tau(x, y)$.

Since $d(x, y) \geq 0$, we have $\tau(x, y) = \min\{d(x, y), 1\} \geq 0$. If $\tau(x, y) = 0$, then $\min\{d(x, y), 1\} = 0$, hence $d(x, y) = 0$, which leads to $x = y$. Moreover, $\tau(x, x) = \min\{d(x, x), 1\} = \min\{0, 1\} = 0$. We also have $\tau(x, y) = \min\{d(x, y), 1\} = \min\{d(y, x), 1\} = \tau(y, x)$.

Lemma I. *f* $a, b \geq 0$, we have $\min\{a + b, 1\} \leq \min\{a, 1\} + \min\{b, 1\}$.

Proof. Indeed, if $a, b > 1$, then the lemma becomes $a + b \leq a + b$. If $a > 1, b \leq 1$, the lemma becomes $a + b \leq a + 1$, which is true, same for the case $b > 1, a \leq 1$. And if $a, b \leq 1$, then obviously $a + b \leq 2$. And since $1 < 2$, we have $\min\{a + b, 1\} \leq 2 = \min\{a, 1\} + \min\{b, 1\}$. \square

Using the lemma, we have

$$\begin{aligned}\tau(x, z) &= \min\{d(x, z), 1\} \\ &\leq \min\{d(x, y) + d(y, z), 1\} \\ &\leq \min\{d(x, y), 1\} + \min\{d(y, z), 1\} \\ &= \tau(x, y) + \tau(y, z).\end{aligned}$$

Thus $\tau(x, y)$ defines a metric on M . \square

Exercise 9. Recall that 2^N denotes the set of all sequences (or "strings") of 0s and 1s. Show that $d(a, b) = \sum_{n=1}^{\infty} 2^{-n} |a_n - b_n|$, where $a = (a_n)$ and $b = (b_n)$ are sequences of 0s and 1s, defines a metric on 2^N .

Proof. Clearly, $d(a, b) \geq 0$. If $d(a, b) = 0$, then $|a_n - b_n| = 0$ for all n , thus $a = b$. Otherwise, $d(a, a) = \sum_{n=1}^{\infty} 2^{-n} |a_n - a_n| = 0$. Furthermore, we have

$$d(a, b) = \sum_{n=1}^{\infty} 2^{-n} |a_n - b_n| = \sum_{n=1}^{\infty} 2^{-n} |b_n - a_n| = d(b, a)$$

and

$$\begin{aligned}d(a, c) &= \sum_{n=1}^{\infty} 2^{-n} |a_n - c_n| \\ &\leq \sum_{n=1}^{\infty} 2^{-n} (|a_n - b_n| + |b_n - c_n|) \\ &= \sum_{n=1}^{\infty} 2^{-n} |a_n - b_n| + \sum_{n=1}^{\infty} 2^{-n} |b_n - c_n| \\ &= d(a, b) + d(b, c).\end{aligned}$$

Thus d defines a metric on 2^N . \square

Exercise 10. The Hilbert cube H^∞ is the collection of all real sequences $x = (x_n)$ with $|x_n| \leq 1$ for $n = 1, 2, \dots$.

(i) Show that $d(x, y) = \sum_{n=1}^{\infty} 2^{-n} |x_n - y_n|$ defines a metric on H^∞ .

Proof. Similar to exercise 9, d defines a metric on \mathbb{R}^∞ □

(ii) Given that $x, y \in H^\infty$ and $k \in \mathbb{N}$, let $M_k = \max\{|x_1 - y_1|, \dots, |x_k - y_k|\}$. Show that $2^{-k} M_k \leq d(x, y) \leq M_k + 2^{1-k}$.

Proof. For any $k \in \mathbb{N}$ and $i \leq k$, we have

$$2^{-k} |x_i - y_i| \leq 2^{-i} |x_i - y_i| \leq \sum_{i=1}^{\infty} 2^{-i} |x_i - y_i|.$$

Thus for $M_k = \max\{|x_1 - y_1|, \dots, |x_k - y_k|\}$, we have

$$2^{-k} M_k \leq d(x, y).$$

Furthermore, we have

$$\sum_{i=1}^k 2^{-i} |x_i - y_i| \leq \sum_{i=1}^k 2^{-i} M_k = M_k (1 - 2^{-k}) \leq M_k.$$

Also notice that $|x_i - y_i| \leq 2$, we also have

$$\sum_{i=k+1}^{\infty} 2^{-i} |x_i - y_i| \leq \sum_{i=k+1}^{\infty} 2^{1-i} = 2^{1-k}.$$

Thus

$$d(x, y) \leq M_k + 2^{1-k}.$$

□

Exercise 11. Let \mathbb{R}^∞ denote the collection of all real sequences $x = (x_n)$. Show that the expression

$$d(x, y) = \sum_{n=1}^{\infty} \frac{1}{n!} \frac{|x_n - y_n|}{1 + |x_n - y_n|}$$

defines a metric on \mathbb{R}^∞ .

Proof. Since every element of this summation is larger than 0, $d(x, y) \geq 0$. It's not hard to check that $d(x, x) = 0$, let $d(x, y) = 0$, then

$$\frac{1}{n!} \frac{|x_n - y_n|}{1 + |x_n - y_n|} = 0$$

for any n , thus $x_n = y_n$ for all n , which means $x = y$. Moreover, we have

$$d(x, y) = \sum_{n=1}^{\infty} \frac{1}{n!} \frac{|x_n - y_n|}{1 + |x_n - y_n|} = \sum_{n=1}^{\infty} \frac{1}{n!} \frac{|y_n - x_n|}{1 + |y_n - x_n|} = d(y, x)$$

and by exercise 6, we have

$$\begin{aligned} d(x, z) &= \sum_{n=1}^{\infty} \frac{1}{n!} \frac{|x_n - z_n|}{1 + |x_n - z_n|} \\ &\leq \sum_{n=1}^{\infty} \frac{1}{n!} \left(\frac{|x_n - y_n|}{1 + |x_n - y_n|} + \frac{|y_n - z_n|}{1 + |y_n - z_n|} \right) \\ &= d(x, y) + d(y, z). \end{aligned}$$

What is more, we have $\frac{|a-b|}{1+|a+b|} < 1$, thus $d(a, b) \leq \sum_{n=1}^{\infty} \frac{1}{n!}$ for all $a, b \in M$. Since $\sum_{n=1}^{\infty} \frac{1}{n!}$ converges, $\sum_{n=1}^{\infty} \frac{1}{n!} \frac{|x_n - z_n|}{1 + |x_n - z_n|}$ converges. Thus d defines a metric M . \square

Exercise 12. Check that $d(f, g) = \max_{a \leq t \leq b} |f(t) - g(t)|$ defines a metric on $C[a, b]$, the collection of all continuous, real-value functions defined on the closed interval $[a, b]$.

Proof. Clearly, $d(f, g) \geq 0$. We have $d(f, f) = \max_{a \leq t \leq b} |f(t) - f(t)| = 0$. If $d(f, g) = 0$, then $\max_{a \leq t \leq b} |f(t) - g(t)| = 0$. Therefor, $f(t) = g(t)$ for all $a \leq t \leq b$. Moreover, we have

$$d(f, g) = \max_{a \leq t \leq b} |f(t) - g(t)| = \max_{a \leq t \leq b} |g(t) - f(t)| = d(g, f)$$

and

$$\begin{aligned} d(f, h) &= \max_{a \leq t \leq b} |f(t) - h(t)| \\ &\leq \max_{a \leq t \leq b} (|f(t) - g(t)| + |g(t) - h(t)|) \\ &\leq \max_{a \leq t \leq b} |f(t) - g(t)| + \max_{a \leq t \leq b} |g(t) - h(t)| \\ &= d(f, g) + d(g, h). \end{aligned}$$

Thus d defines a metric on $C[a, b]$. \square

Exercise 14. We said that a subset A of a metric space M is bounded if there is some $x_0 \in M$ and some constant $C < \infty$ such that $d(a, x_0) \leq C$ for all $a \in A$. Show that a finite union of bounded sets is again bounded.

Proof. Let these bounded sets be A_1, A_2, \dots, A_n and $x_i \in A_i$ for $i = 1, 2, \dots, n$, let a_1, a_2, \dots, a_n be n constants such that for any $1 \leq i \leq n$, we have $d(t, x_i) \leq a_i$ for any $t \in A_i$. Finally, let $A = \cup_{i=1}^n A_i$.

For any set A_k , an $t \in A_k$ ($1 \leq k \leq n$), we have

$$d(x_1, t) \leq d(x_1, x_k) + d(x_k, t) \leq d(x_1, x_k) + a_k.$$

Notice that $d(x_1, x_k) + a_k$ is a constant, let $d_k = d(x_1, x_k) + a_k$, then $d = \max\{d_1, d_2, \dots, d_n\}$ is an upper bound for A . Thus a finite union of bounded sets is bounded. \square

Exercise 15. We define the diameter of a nonempty subset A of M by $\text{diam}(A) = \sup\{d(a, b) : a, b \in A\}$. Show that A is bounded if and only if $\text{diam}(A)$ is finite.

Proof. If A is bounded, then exist $x \in A$ and a number c such that $d(a, x) \leq c$ for all $a \in A$. Therefor, for any $a, b \in A$, we have

$$d(a, b) \leq d(a, x) + d(x, b) \leq 2c.$$

Thus the set $\{d(a, b) : a, b \in A\}$ has an upper bound, which means $\text{diam}(A) = \sup\{d(a, b) : a, b \in A\}$ is finite. Moreover, if $\text{diam}(A) = c$ a finite number, then $d(a, x) \leq c$ for all $a \in A$ and an $x \in A$. \square

Exercise 16. Let V be a vector space, and let d be a metric on V satisfying $d(x, y) = d(x - y, 0)$ and $d(ax, ay) = |a|d(x, y)$ for every $x, y \in V$ and every scalar a . Show that $\|x\| = d(x, 0)$ defines a norm on V . Give an example of a metric on the vector space \mathbb{R} that fails to be associated with a norm in this way.

Proof. For any $x \in V$, because d defines a matrix on V , thus $\|x\| = d(x, 0) \geq 0$. We have $\|0\| = d(0, 0) = 0$ and if $\|x\| = 0$, then $d(x, 0) = 0$. Therefor, $x = 0$ since d is a metric on V . We also have

$$\|ax\| = d(ax, 0) = |a|d(x, 0) = |a|\|x\|.$$

and

$$\begin{aligned} \|x + y\| &= d(x + y, 0) \\ &= d(x, -y) \\ &\leq d(x, 0) + d(0, -y) \\ &= d(x, 0) + d(y, 0) \\ &= \|x\| + \|y\|. \end{aligned}$$

One example of a metric on \mathbb{R} that fails to be associate with a norm this way is

$$d(x) = \begin{cases} 0 & \text{if } x = y \\ 1 & \text{if } x \neq y \end{cases}.$$

Because if letting $\|x\| = d(x, 0)$, then $\|ax\| = 1 \neq |a| = |a|\|x\|$ for any nonzero vector x and scalar a . \square

Exercise 17. Recall that for $x \in \mathbb{R}^n$ we have defined $\|x\|_1 = \sum_{i=1}^n |x_i|$ and $\|x\|_\infty = \max_{1 \leq i \leq n} |x_i|$. Check that each of these is indeed a norm on \mathbb{R}^n .

Proof. Obviously $\|x\|_1 \geq 0$. We have $\|0\|_1 = \sum_{i=1}^n 0 = 0$, whereas if $\|x\|_1 = 0$, then $|x_i| = 0$ for all $1 \leq i \leq n$. Thus $x = 0$. Moreover, $\|ax\|_1 = \sum_{i=1}^n |ax_i| = |a| \sum_{i=1}^n |x_i| = |a| \cdot \|x\|_1$. Last but not least, we have

$$\begin{aligned} \|x + y\|_1 &= \sum_{i=1}^n |x_i + y_i| \\ &\leq \sum_{i=1}^n |x_i| + \sum_{i=1}^n |y_i| \\ &= \|x\|_1 + \|y\|_1. \end{aligned}$$

Similarly, $\|x\|_\infty \geq 0$ is obvious. We have $\|0\|_\infty = \max_{1 \leq i \leq n} |x_i| = 0$, and if $\|x\|_\infty = 0$, then $\max_{1 \leq i \leq n} |x_i| = 0$. Therefore, $|x_i| = 0$ for all i , which means $x = 0$. Moreover, $\|ax\|_\infty = \max_{1 \leq i \leq n} |ax_i| = |a| \max_{1 \leq i \leq n} |x_i| = |a| \cdot \|x\|_\infty$. Lastly,

$$\begin{aligned} \|x + y\|_\infty &= \max_{1 \leq i \leq n} |x_i + y_i| \\ &\leq \max_{1 \leq i \leq n} (|x_i| + |y_i|) \\ &\leq \max_{1 \leq i \leq n} |x_i| + \max_{1 \leq i \leq n} |y_i| \\ &= \|x\|_\infty + \|y\|_\infty. \end{aligned}$$

Thus $\|x\|_1$ and $\|x\|_\infty$ defines two metrics on \mathbb{R}^n . □

Exercise 18. Show that $\|x\|_\infty \leq \|x\|_2 \leq \|x\|_1$ for any $x \in \mathbb{R}^n$. Also check that $\|x\|_1 \leq n\|x\|_\infty$ and $\|x\|_1 \leq \sqrt{n}\|x\|_2$.

Proof. We have

$$\begin{aligned} \|x\|_\infty &\leq \|x\|_2 \\ \Leftrightarrow \max_{1 \leq i \leq n} |x_i| &\leq \sqrt{\sum_{i=1}^n x_i^2} \\ \Leftrightarrow \max_{1 \leq i \leq n} x_i^2 &\leq \sum_{i=1}^n x_i^2 \text{ (square both sides),} \end{aligned}$$

which is obvious since $x_i^2 \geq 0$ for all x_i and one of the addends on the right side equals $\max_{1 \leq i \leq n} x_i^2$. Therefore, $\|x\|_\infty \leq \|x\|_2$. Very similarly, we have

$$\begin{aligned} \|x\|_2 &\leq \|x\|_1 \\ \Leftrightarrow \sqrt{\sum_{i=1}^n x_i^2} &\leq \sum_{i=1}^n |x_i| \\ \Leftrightarrow \sum_{i=1}^n x_i^2 &\leq \sum_{i=1}^n x_i^2 + \sum_{i \neq j} |x_i| |x_j|, \end{aligned}$$

which is obvious since $|x_i||x_j| \geq 0$. Hence, $\|x\|_\infty \leq \|x\|_2 \leq \|x\|_1$ for any $x \in \mathbb{R}^n$.

What is more, because $|x_i| \leq \max_{1 \leq i \leq n} |x_i|$ for all $1 \leq i \leq n$, we have

$$\sum_{i=1}^n |x_i| \leq n \max_{1 \leq i \leq n} |x_i|.$$

Hence $\|x\|_1 \leq n\|x\|_\infty$. Moreover, by Cauchy Schwarz inequality, we have

$$1|x_1| + 1|x_2| + \cdots + 1|x_n| \leq \sqrt{(1^2 + 1^2 + \cdots + 1^2)(x_1^2 + x_2^2 + \cdots + x_n^2)}.$$

Thus

$$\sum_{i=1}^n x_i \leq \sqrt{n \sum_{i=1}^n x_i^2},$$

which is synonymous to $\|x\|_1 \leq \sqrt{n}\|x\|_2$. □

Exercise 19. Show that we have $\sum_{i=1}^n x_i y_i = \|x\|_2 \|y\|_2$ if and only if x and y are proportional, that is, if and only if $x = \alpha y$ or $y = \alpha x$ for some $\alpha \leq 0$.

Proof. We will prove a stronger result, that is $\sum_{i=1}^n x_i y_i \leq \|x\|_2 \|y\|_2$ and the equality is if x and y are proportional by mathematical induction. Let x_k be a random vector in \mathbb{R}^k . If $n = 1$, the inequality becomes $x_1 y_1 \leq |x_1| |y_1|$, which is obvious. And since both x and y are in \mathbb{R} , they are obviously proportional. Now assume that the result holds for $n - 1$, notice that by AM-GM inequality, we have

$$\begin{aligned} & 2x_n y_n \sqrt{(x_1^2 + \cdots + x_{n-1}^2)(y_1^2 + \cdots + y_{n-1}^2)} \\ \leq & 2|x_n||y_n| \sqrt{(x_1^2 + \cdots + x_{n-1}^2)(y_1^2 + \cdots + y_{n-1}^2)} \\ \leq & x_n^2(y_1^2 + \cdots + y_{n-1}^2) + y_n^2(x_1^2 + \cdots + x_{n-1}^2). \end{aligned}$$

Thus

$$\begin{aligned} & (x_1^2 + \cdots + x_{n-1}^2)(y_1^2 + \cdots + y_{n-1}^2) + 2x_n y_n \sqrt{(x_1^2 + \cdots + x_{n-1}^2)(y_1^2 + \cdots + y_{n-1}^2)} + x_n^2 y_n^2 \\ \leq & (x_1^2 + \cdots + x_{n-1}^2)(y_1^2 + \cdots + y_{n-1}^2) + x_n^2(y_1^2 + \cdots + y_{n-1}^2) + y_n^2(x_1^2 + \cdots + x_{n-1}^2) + x_n^2 y_n^2. \end{aligned}$$

After cleaning this mess, we reach something that is not very messy.

$$\left(x_n y_n + \sqrt{(x_1^2 + \cdots + x_{n-1}^2)(y_1^2 + \cdots + y_{n-1}^2)} \right)^2 \leq (x_1^2 + \cdots + x_n^2)(y_1^2 + \cdots + y_n^2).$$

Take the square root both sides, we get

$$|x_n y_n| + \sqrt{(x_1^2 + \cdots + x_{n-1}^2)(y_1^2 + \cdots + y_{n-1}^2)} \leq \sqrt{(x_1^2 + \cdots + x_n^2)(y_1^2 + \cdots + y_n^2)},$$

or

$$x_n y_n + \|x_{n-1}\|_2 \cdot \|y_{n-1}\|_2 \leq |x_n y_n| + \|x_{n-1}\|_2 \cdot \|y_{n-1}\|_2 \leq \|x_n\|_2 \cdot \|y_n\|_2.$$

But thanks to the induction assumption, we have $\sum_{i=1}^{n-1} x_i y_i \leq \|x_{n-1}\|_2 \cdot \|y_{n-1}\|_2$. Thus

$$\sum_{i=1}^n x_i y_i \leq \|x_n\|_2 \|y_n\|_2.$$

The equality is when equality happens in every inequality that we used, which are $x_n y_n \leq |x_n y_n|$ and the AM-GM inequality. The first equality is equivalent to x_n and y_n have the same sign. The second one is equivalent to

$$x_n \sqrt{y_1^2 + \cdots y_{n-1}^2} = y_n \sqrt{x_1^2 + \cdots x_{n-1}^2},$$

Hence

$$\frac{x_n}{y_n} = \frac{\sqrt{x_1^2 + \cdots x_{n-1}^2}}{\sqrt{y_1^2 + \cdots y_{n-1}^2}} = \alpha$$

by the induction assumption. Thus $x = \alpha y$ for $\alpha \geq 0$. \square

Exercise 22. Show that $\|x\|_\infty \leq \|x\|_2$ for any $x \in \ell_2$, and that $\|x\|_2 \leq \|x\|_1$ for any $x \in \ell_1$.

Proof. Let $x \in \ell_1$, then $\sum_{i=1}^\infty |x_i|$ exists. Notice that $\sqrt{\sum_{i=1}^n x_i^2} \leq \sum_{i=1}^n |x_i|$ for all n , thus the sequence $\sqrt{\sum_{i=1}^n x_i^2}$ is both increasing and upper bounded. Thus $\sqrt{\sum_{i=1}^\infty x_i^2}$ also exists. Thus by the inequality above, we get $\|x\|_2 \leq \|x\|_1$.

Now let $x \in \ell_2$, we have $\sqrt{\sum_{i=1}^\infty x_i^2}$ exists. Similar to the upper case, we have $\max_{1 \leq i \leq n} x_i$ is both increasing and upper bounded, thus $\lim_{n \rightarrow \infty} (\max_{1 \leq i \leq n} x_i)$ exists. Since $\max_{1 \leq i \leq n} x_i \leq \sqrt{\sum_{i=1}^n x_i^2}$, we have $\|x\|_\infty \leq \|x\|_2$ for any $x \in \ell_2$. \square

Exercise 23. The subset of ℓ_∞ consisting of all sequences that converge to 0 is denoted by c_0 . Show that we have the following proper set inclusions: $\ell_1 \subset \ell_2 \subset c_0 \subset \ell_\infty$.

Proof. As we already checked in exercise 22, for any $x \in \ell_1$, then $\|x\|_2$ is finite. Therefore, $x \in \ell_2$ too, which means $\ell_1 \subset \ell_2$. Moreover $c_0 \subset \ell_\infty$ is by definition. Thus all we have to check left is $\ell_2 \subset c_0$. Notice that if $x = (x_1, x_2, \dots) \in \ell_2$, then $\sqrt{\sum_{i=1}^\infty x_i^2}$ exists. Therefore, $x_n \rightarrow 0$, which leads to $x \in c_0$. Thus $\ell_1 \subset \ell_2 \subset c_0 \subset \ell_\infty$. \square

Exercise 24. Prove that Holder's Inequality is also holds in the case $p = 1$ and $q = \infty$.

Proof. For $p = 1$ and $q = \infty$, the inequality becomes $\sum_{i=1}^\infty |x_i y_i| \leq \|x\|_1 \cdot \|y\|_\infty$, which means $\sum_{i=1}^\infty |x_i y_i| \leq \sum_{i=1}^\infty |x_i| \cdot \sup_i |y_i|$. However, this result is obvious because $|y_n| \leq \sup_i |y_i|$ for any n . Thus the Holder's inequality is also holds in the case $p = 1$ and $q = \infty$. \square

Exercise 25. The same techniques can be used to show that $\|f\|_p = (\int_0^1 |f(t)|^p dt)^{\frac{1}{p}}$ defines a norm on $C[0, 1]$ for any $1 < p < \infty$. State and prove the analogues of lemma 3.7 (Holder's Inequality) and Theorem 3.8 (Minkowski's Inequality) in this case.

Proof. First, notice that

$$\begin{aligned} \|f^{p-1}\|_q &= \left(\int_0^1 |f(x)^{p-1}|^q dx \right)^{\frac{1}{q}} \\ &= \left(\int_0^1 |f(x)|^p dx \right)^{\frac{1}{q}} \\ &= \left(\int_0^1 |f(x)|^p dx \right)^{\frac{p-1}{p}} \\ &= \|f\|_p^{p-1}. \end{aligned}$$

Now it's time to restate Lemma 3.7.

Lemma 3.7. (Holder's Inequality) Let $1 < p < \infty$ and let q be defined by $\frac{1}{p} + \frac{1}{q} = 1$. Given f and g in $C[0, 1]$, we have

$$\int_0^1 |f(t)g(t)| dt \leq \|f\|_p \|g\|_q.$$

Proof. By Young's inequality, we have

$$\frac{|f(t)| \cdot |g(t)|}{\|f\|_p \|g\|_q} \leq \frac{1}{p} \left| \frac{|f(t)|}{\|f\|_p} \right|^p + \frac{1}{q} \left| \frac{|g(t)|}{\|g\|_q} \right|^q.$$

Therefore

$$\int_0^1 \frac{|f(t)| \cdot |g(t)|}{\|f\|_p \|g\|_q} dt \leq \frac{1}{p} \int_0^1 \frac{|f(t)|^p}{\|f\|_p^p} dt + \frac{1}{q} \int_0^1 \frac{|g(t)|^q}{\|g\|_q^q} dt = \frac{1}{p} + \frac{1}{q} = 1.$$

Thus we have the Holder's inequality. □

With lemma 3.7, we can restate and prove theorem 3.8 as follow.

Lemma 3.8. Let $1 < p < \infty$, then we have $\|f + g\|_p \leq \|f\|_p + \|g\|_p$ for $f, g \in C[0, 1]$.

Proof. Indeed, we have

$$\begin{aligned} \|f + g\|_p^p &= \int_0^1 |f + g|^p \\ &= \int_0^1 |f + g|^{p-1} |f + g| \\ &\leq \int_0^1 |f + g|^{p-1} |f| + \int_0^1 |f + g|^{p-1} |g| \\ &\leq \|(|f + g|)^{p-1}\|_q \|f\|_p + \|(|f + g|)^{p-1}\|_q \|g\|_p \\ &= \|f + g\|_p^{p-1} (\|f\|_p + \|g\|_p). \end{aligned}$$

Thus $\|f + g\|_p \leq \|f\|_p + \|g\|_p$. □

So $\|f\|_p$ satisfies the triangular inequality. Moreover, it's easy to check that $\|f\|_p \geq 0$. If $\|f\|_p = 0$, then $\int_0^1 f(t)^p dt = 0$. And since f is continuous on $[0, 1]$, we have $f(t) = 0$ for all $t \in [0, 1]$. If $f(t) = 0$, then it's also easy to check that $\|f\|_p = 0$. Moreover, for a scalar a , then

$$\|af\|_p = \left(\int_0^1 |af(t)|^p dt \right)^{\frac{1}{p}} = |a| \left(\int_0^1 |f(t)|^p dt \right)^{\frac{1}{p}} = |a| \cdot \|f\|_p.$$

Thus $\|f\|_p$ is a norm on $C[0, 1]$. □

Exercise 26. Given $a, b > 0$, show that $\lim_{p \rightarrow \infty} (a^p + b^p)^{\frac{1}{p}} = \max\{a, b\}$. What happens as $p \rightarrow 0$? as $p \rightarrow -1$? as $p \rightarrow \infty$?

Proof. Assume that $a < b$, let $r = \frac{a}{b}$, we have

$$\begin{aligned} (a^p + b^p)^{\frac{1}{p}} &= b \cdot \left(1 + \left(\frac{a}{b} \right)^p \right)^{\frac{1}{p}} \\ &= b \cdot (1 + r^p)^{\frac{1}{p}} \\ &= b \cdot e^{\log(1+r^p) \frac{1}{p}}. \end{aligned}$$

Notice that as $p \rightarrow \infty$, we have $\log(1+r^p)^{\frac{1}{p}} \rightarrow 0 \cdot \log(1) \cdot 0 = 0$. Thus $(a^p + b^p)^{\frac{1}{p}} \rightarrow b \cdot 1 = b = \max\{a, b\}$. □

Exercise 27. Show that $\text{diam}(B_r(x)) \leq 2r$, and give an example where strict inequality occurs.

Proof. For any $a, b \in B_r(x)$, we have $d(a, b) \leq d(a, x) + d(x, b) < 2r$. Thus $\text{diam}(B_r(x)) = \sup\{d(a, b) : a, b \in B_r(x)\} \leq 2r$. An example for strict inequality is the discrete space because $\text{diam}(B_1(0)) = 0 < 2$. □

Exercise 28. If $\text{diam}(A) < r$, show that $A \subset B_r(a)$ for some $a \in A$.

Proof. For any $x \in A$, because $\text{diam}(A) < r$, we have $d(a, x) < r$. Therefore, $x \in B_r(a)$. Thus $A \subset B_r(a)$. □

Exercise 30. If $A \subset B$, show that $\text{diam}(A) \leq \text{diam}(B)$.

Proof. Since $A \subset B$, we have $\{d(a, b) : a, b \in A\} \subset \{d(a, b) : a, b \in B\}$. Thus $\text{diam}(A) = \sup\{d(a, b) : a, b \in A\} \leq \sup\{d(a, b) : a, b \in B\} = \text{diam}(B)$. □

Exercise 32. In a normed vector space $(V, \|\cdot\|)$ show that $B_r(x) = x + B_r(0) = \{x + y : \|y\| < r\}$ and that $B_r(0) = rB_1(0) = \{rx : \|x\| < 1\}$.

Proof. For any $a \in B_r(x)$, we have $\|a - x\| < r$. Therefore $a = x - (a - x) \in x + B_r(0)$. Thus $B_r(x) \subset x + B_r(0)$. And for any $a \in x + B_r(0)$, then there exists $y \in B_r(0)$ such that $a = x + y$. Thus $\|a - x\| = \|y\| < r$. Thus $a \in B_r(x)$, which means $x + B_r(0) \subset B_r(x)$. Hence $B_r(x) = x + B_r(0)$.

Similarly, if $a \in B_r(0)$, then $\|a\| < r$. Notice that because $r > 0$, we have $\|\frac{a}{r}\| = \frac{1}{r}\|a\| < 1$. And since $a = r \cdot \frac{a}{r}$, we have $a \in rB_1(0)$. Thus $B_r(0) \subset rB_1(0)$. Moreover, if $a \in rB_1(0)$, then there exists $x \in B_1(0)$ such that $a = rx$. Then $\|a\| = \|rx\| = |r| \cdot \|x\| < r$. Thus $a \in B_r(0)$, which leads to $rB_1(0) \subset B_r(0)$. Thus $B_r(0) = rB_1(0)$. □

Exercise 33. *Limits are unique.*

Proof. Assume that x and y are two identical limits of (x_n) , then $d(x, y) > 0$. Let $\epsilon > 0$ such that $2\epsilon < d(x, y)$, there exists N such that if $n > N$, then $x_n \in B_\epsilon(x) \cap B_\epsilon(y)$. Therefore, $2\epsilon < d(x, y) \leq d(x, x_n) + d(x_n, y) < 2\epsilon$, contradiction. Thus the limit of (x_n) must be unique. \square

Exercise 34. *If $x_n \rightarrow x$ in (M, d) , show that $d(x_n, y) \rightarrow d(x, y)$ for any $y \in M$. More generally, if $x_n \rightarrow x$ and $y_n \rightarrow y$, show that $d(x_n, y_n) \rightarrow d(x, y)$.*

First, we will prove a small lemma.

Lemma . *For $a, b, c \in M$, we have $|d(a, b) - d(b, c)| \leq d(a, c)$.*

Proof. Indeed, since $d(a, b) \leq d(a, c) + d(b, c)$, we have $d(a, b) - d(b, c) \leq d(a, c)$. Moreover $d(b, c) \leq d(b, a) + d(a, c)$, hence $-d(a, c) \leq d(a, b) - d(b, c)$. Thus $|d(a, b) - d(b, c)| \leq d(a, c)$. \square

Now back to the problem.

Proof. By the lemma, we have $|d(x_n, y) - d(x, y)| \leq d(x, x_n)$. Since $x_n \rightarrow x$, $d(x, x_n)$ can be sufficiently small. Thus $d(x_n, y) \rightarrow d(x, y)$.

Also by the lemma, we have $|d(x_n, y_n) - d(x, y)| \leq |d(x_n, y_n) - d(x_n, y)| + |d(x_n, y) - d(x, y)| \leq d(y_n, y) + d(x, x_n)$. Similarly, since we can make $d(x_n, x)$ and $d(y_n, y)$ sufficiently small, we have $d(x_n, y_n) \rightarrow d(x, y)$. \square

Exercise 35. *If $x_n \rightarrow x$, then $x_{n_k} \rightarrow x$ for any subsequence (x_{n_k}) of (x_n) .*

Proof. For any $\epsilon > 0$, there exists $N \in \mathbb{N}$ such that for any $n > N$, we have $d(x_n, x) < \epsilon$. Since $\{1, 2, \dots, N\}$ is finite, there exists some $k_0 \in \mathbb{N}$ such that $\{1, 2, \dots, N\} \cap \{n_{k_0+1}, n_{k_0+2}, \dots\} = \emptyset$. Let $K = k_0$, then if $k > K$, because $n_k > N$, we have $d(x_{n_k}, x) < \epsilon$. Thus $x_{n_k} \rightarrow x$. \square

Exercise 36. *A convergent sequence in Cauchy, and a Cauchy sequence is bounded.*

Proof. Let $x_n \rightarrow x$, we will prove that (x_n) is Cauchy. For any $\epsilon > 0$, we can find N such that if $n > N$, then $x_n \in B_{\frac{\epsilon}{2}}(x)$. And because $\text{diam}(B_{\frac{\epsilon}{2}}(x)) < 2\frac{\epsilon}{2} = \epsilon$, for any $m, n > N$, $d(x_m, x_n) < \epsilon$. Thus (x_n) is Cauchy.

If (x_n) is Cauchy, then there exists $N \in \mathbb{N}$ such that for any $m, n > N$, $d(x_m, x_n) < 1$. Fix m , let $t = \max(\{d(x_k, x_m) : 1 \leq k \leq N\} \cup \{1\})$. Then clearly $x_n \in B_t(x_m)$ for any $n \in \mathbb{N}$. Thus (x_n) is bounded. \square

Exercise 37. *A Cauchy sequence with a convergent subsequence converges.*

Proof. Let (x_n) be Cauchy and $x_n \rightarrow x$, where (x_{n_k}) is a subsequence of (x_n) . For any $\epsilon > 0$, we can find N_1 and N_2 such that for all $m, n > N_1$ and $n_p > N_2$, then $d(x_m, x_n) < \frac{\epsilon}{2}$ and $d(x, x_{n_p}) < \frac{\epsilon}{2}$. Let $N = \max\{N_1, N_2\}$ and a natural number $n_{k_0} > N$. For any $n > N$, we have

$$d(x_n, x_{n_{k_0}}) < \frac{\epsilon}{2} \text{ and } d(x_{n_{k_0}}, x) < \frac{\epsilon}{2}.$$

Therefore,

$$d(x_n, x) \leq d(x_n, x_{n_{k_0}}) + d(x_{n_{k_0}}, x) < \frac{\epsilon}{2} + \frac{\epsilon}{2} = \epsilon.$$

Thus $x_n \rightarrow x$. \square

Exercise 40. Given any fixed element $x \in \ell_1$, show that the sequence $x^{(k)} = (x_1, \dots, x_k, 0, \dots) \in \ell_1$ converges to x in ℓ_1 -norm. Show that the same holds true in ℓ_2 , but give an example showing that it fails in ℓ_∞ .

Proof. Since $x \in \ell_1$, we have $\sum_{i=1}^{\infty} |x_i| < \infty$, thus $\lim_{k \rightarrow \infty} \sum_{i=k}^{\infty} |x_i| = 0$. Now, for any $\epsilon > 0$, there exists K such that if $k > K$, then

$$\|x - x^{(k)}\|_1 = \sum_{i=k+1}^{\infty} |x_i| < \epsilon.$$

Thus $x^{(k)} \rightarrow x$ in ℓ_1 .

Similarly, if $x \in \ell_2$, then $\lim_{k \rightarrow \infty} (\sum_{i=k}^{\infty} x_i^2)^{\frac{1}{2}} = 0$. Thus for any $\epsilon > 0$, there exists K such that if $k > K$, then

$$\|x - x^{(k)}\|_2 = \left(\sum_{i=k+1}^{\infty} x_i^2 \right)^{\frac{1}{2}} < \epsilon.$$

Thus $x^{(k)} \rightarrow x$ in ℓ_2 .

One example showing that ℓ_∞ false is $x = (1, 1, \dots)$. We have $\|x - x^{(k)}\|_\infty = \max\{0, 1\} = 1$. Thus no matter how large k is, $d(x, x^{(k)}) = 1$, which means $x^{(k)}$ does not converge to x . \square

Exercise 41. Given $x, y \in \ell_2$, show that if $x^{(k)} \rightarrow x$ and $y^{(k)} \rightarrow y$ in ℓ_2 , then $\langle x^{(k)}, y^{(k)} \rangle \rightarrow \langle x, y \rangle$.

Proof. Assume that for some k_0 , $\|x^{(k_0)} - x\|_2 < \epsilon$ and $\|y^{(k_0)} - y\|_2 < \delta$, then applying the Cauchy-Schwarz, we have

$$\begin{aligned} |\langle x^{(k_0)}, y^{(k_0)} \rangle - \langle x, y \rangle| &= |\langle x^{(k_0)} - x, y^{(k_0)} - y \rangle + \langle x, y^{(k_0)} - y \rangle + \langle y, x^{(k_0)} - x \rangle| \\ &= |\langle x^{(k_0)} - x, y^{(k_0)} - y \rangle| + |\langle x, y^{(k_0)} - y \rangle| + |\langle y, x^{(k_0)} - x \rangle| \\ &\leq \|x^{(k_0)} - x\|_2 \|y^{(k_0)} - y\|_2 + \|x\|_2 \|y^{(k_0)} - y\|_2 + \|y\|_2 \|x^{(k_0)} - x\|_2 \\ &\leq \epsilon \delta + \delta \|x\|_2 + \epsilon \|y\|_2. \end{aligned}$$

Because $x, y \in \ell_2$, $\|x\|_2$ and $\|y\|_2$ are finite. Moreover, because $x^{(k)} \rightarrow x$ and $y^{(k)} \rightarrow y$, ϵ and δ can be as small as possible. Therefore $|\langle x^{(k)}, y^{(k_0)} \rangle - \langle x, y \rangle|$ can be as small as possible, thus $\langle x^{(k)}, y^{(k_0)} \rangle \rightarrow \langle x, y \rangle$. \square

Exercise 42. Two metrics d and ρ on a set M is said to be equivalent if they generate the same convergent sequences; that is, $d(x_n, x) \rightarrow 0$ if and only if $\rho(x_n, x) \rightarrow 0$. If d is any metric on M , show that the metrics ρ, σ, τ are all equivalent to d .

Proof. Assume that there exists x_n and $x \in M$ such that $d(x_n, x) \rightarrow 0$, then for any $0 < \epsilon < 1$, there exists N such that if $n > N$, then $d(x_n, x) < \epsilon$. Thus

$$\begin{aligned} \rho(x_n, x) &= \sqrt{d(x_n, x)} < \sqrt{\epsilon} < \epsilon, \\ \sigma(x_n, x) &= \frac{d(x_n, x)}{1 + d(x_n, x)} < \frac{d(x_n, x)}{1} < \epsilon, \end{aligned}$$

and

$$\tau(x_n, x) = \min\{d(x_n, x), 1\} = d(x_n, x) < \epsilon.$$

Therefore, we have $\rho(x_n, x) \rightarrow 0$, $\sigma(x_n, x) \rightarrow 0$, and $\tau(x_n, x) \rightarrow 0$.

Now assume that $\rho(x_n, x) \rightarrow 0$, we have

$$d(x_n, x) = \rho(x_n, x)^2 \rightarrow 0.$$

If $\sigma(x_n, x) \rightarrow 0$, first, we will calculate $d(x_n, x)$ respect to $\sigma(x_n, x)$. We have

$$\begin{aligned}\sigma(x_n, x) &= \frac{d(x_n, x)}{1 + d(x_n, x)} \\ \frac{1}{\sigma(x_n, x)} &= \frac{1 + d(x_n, x)}{d(x_n, x)} = 1 + \frac{1}{d(x_n, x)} \\ \frac{1}{\sigma(x_n, x)} - 1 &= \frac{1}{d(x_n, x)} \\ \frac{1}{\frac{1}{\sigma(x_n, x)} - 1} &= d(x_n, x).\end{aligned}$$

As $\sigma(x_n, x) \rightarrow 0$, we have $\frac{1}{\sigma(x_n, x)} \rightarrow \infty$, hence $d(x_n, x) = \frac{1}{\frac{1}{\sigma(x_n, x)} - 1} \rightarrow 0$. Now, if $\tau(x_n, x) \rightarrow 0$, then $\min\{d(x_n, x), 1\} \rightarrow 0$, thus $d(x_n, x) \rightarrow 0$. \square

Exercise 43. Show that the usual metric on \mathbb{N} is equivalent to the discrete metric. Show that any metric on a finite set is equivalent to the discrete metric.

Proof. For any $x_n, x \in \mathbb{N}$, if $|x_n - x| \rightarrow 0$, then let $\epsilon = \frac{1}{2}$, then there exists N such that if $n > N$, then $|x_n - x| < \frac{1}{2}$, therefore $x_n = x$. Let d be the discrete metric on N , then $d(x_n, x) = 0$ for $n > N$. Thus $d(x_n, x) \rightarrow 0$.

If $d(x_n, x) \rightarrow 0$, then for $\epsilon < 1$, there exist N such that if $n > N$, then $d(x_n, x) < \epsilon < 1$. Hence $d(x_n, x) = 0$, which means $x_n = x$. Thus $|x_n - x| \rightarrow 0$. Thus the usual metric on \mathbb{N} is equivalent to the discrete metric.

Let M be a finite set and d, m defines a discrete metric and a random metric on M . Since metric is positive, let $0 < \epsilon < \min\{m(x, y) : x, y \in M, x \neq y\}$. For $x_n, x \in M$, if $m(x_n, x) \rightarrow 0$, then there exists N such that for any $n > N$, then $m(x_n, x) < \epsilon$. But by the definition of ϵ , we have $x_n = x$. Therefore, $d(x_n, x) = 0$ for any $n > N$. So $d(x_n, x) \rightarrow 0$.

If $d(x_n, x) \rightarrow 0$, then similar to the \mathbb{N} case, there exists N such that for any $n > N$, we have $x_n = x$. Thus $m(x_n, x) = 0$ for $n > N$. Thus $m(x_n, x) \rightarrow 0$. \square

Exercise 44. Show that the metrics induced by $\|\cdot\|_1$, $\|\cdot\|_2$, and $\|\cdot\|_\infty$ on \mathbb{R}^n are all equivalent.

Proof. Assume that $\|x - x_k\| \rightarrow 0$ for some $x, x_k \in \mathbb{R}^n$. By exercise 18, we have $0 \leq \|x - x_k\|_\infty \leq \|x - x_k\|_2 \leq \|x - x_k\|_1$. Thus, we also have $\|x - x_k\|_\infty \rightarrow 0$ and $\|x - x_k\|_2 \rightarrow 0$.

Now, assume that $\|x - x_k\|_2 \rightarrow 0$, then $\sqrt{n}\|x - x_k\|_2 \rightarrow 0$. Also by exercise 18, we have $\|x - x_k\| \leq \sqrt{n}\|x - x_k\|_2$. Thus $\|x - x_k\|_1 \rightarrow 0$.

Similarly, if $\|x - x_k\|_\infty \rightarrow 0$, then $n\|x - x_k\|_\infty \rightarrow 0$. By exercise 18, we have $\|x - x_k\|_1 \leq n\|x - x_k\|_\infty$. Thus $\|x - x_k\|_1 \rightarrow 0$.

Therefore, $\|\cdot\|_1$, $\|\cdot\|_2$, and $\|\cdot\|_\infty$ are equivalent. \square

Exercise 45. We say that two norms on the same vector space X are equivalent if the metrics they induce are equivalent. Show that $\|\cdot\|$ and $\|\cdot\|$ are equivalent of X if and only if they generate the same sequences tending to 0; that is, $\|x_n\| \rightarrow 0$ if and only if $\|x_n\| \rightarrow 0$.

Proof. Assume that $\|x_n\| \rightarrow 0$ if and only if $\|x_n\| \rightarrow 0$, we will prove that $\|\cdot\|$ and $\|\cdot\|$ are equivalent. Indeed, if $x_n, x \in X$ such that $\|x_n - x\| \rightarrow 0$, then notice that $y_n = x_n - x \in X$, we have $\|y_n\| \rightarrow 0$. However, by the assumption, we have $\|y_n\| \rightarrow 0$, which is synonymous with $\|x_n - x\| \rightarrow 0$. Similarly, if $\|x_n - x\| \rightarrow 0$, we also have $\|x_n - x\| \rightarrow 0$. Thus $\|\cdot\|$ and $\|\cdot\|$ are equivalent.

If $\|\cdot\|$ and $\|\cdot\|$ are equivalent, then obviously $\|x_n\| \rightarrow 0$ if and only if $\|x_n\| \rightarrow 0$. \square

Exercise 46. Given two metric spaces (M, d) and (N, p) , we can define a metric on the product $M \times N$ in a variety of ways. Our only requirement is that a sequence of pairs (a_n, x_n) in $M \times N$ should converge precisely when both coordinate sequences (a_n) and (x_n) converge. Show that each of the following define metrics on $M \times N$ that enjoy this property and that all three are equivalent:

$$d_1((a, x), (b, y)) = d(a, b) + p(x, y),$$

$$d_2((a, x), (b, y)) = (d(a, b)^2 + p(x, y)^2)^{1/2},$$

$$d_\infty((a, x), (b, y)) = \max\{d(a, b), p(x, y)\}.$$

Proof. First we will prove that d_1, d_2, d_∞ define metrics on $M \times N$. For any $(a, x), (b, y) \in M \times N$, we have

$$d_1((a, x), (b, y)) = d(a, b) + p(x, y) \geq 0,$$

$$d_2((a, x), (b, y)) = (d(a, b)^2 + p(x, y)^2)^{1/2} \geq 0,$$

$$d_\infty((a, x), (b, y)) = \max\{d(a, b), p(x, y)\} \geq 0.$$

The equality is when $d(a, b) = d(x, y) = 0$. Thus $d_1((a, x), (b, y)) = d_2((a, x), (b, y)) = d_\infty((a, x), (b, y)) = 0$ if and only if $d(a, b) = p(x, y) = 0$. We also have

$$d_1((a, x), (b, y)) = d(a, b) + p(x, y) = d(b, a) + p(y, x) = d_1((b, y), (a, x)),$$

$$d_2((a, x), (b, y)) = (d(a, b)^2 + p(x, y)^2)^{1/2} = (d(b, a)^2 + p(y, x)^2)^{1/2} = d_2((b, y), (a, x)),$$

$$d_\infty((a, x), (b, y)) = \max\{d(a, b), p(x, y)\} = \max\{d(b, a), p(y, x)\} = d_\infty((b, y), (a, x)).$$

Now, for any $i \geq 1$, applying the triangular inequality for d and p , and norm inequality, we have

$$\|(d(a, b), p(x, y))\|_i \leq \|(d(a, c) + d(c, b), p(x, z) + p(z, y))\|_i \leq \|(d(a, c), p(x, z))\|_i + \|(d(c, b), p(z, y))\|_i$$

Thus for $i = 1, 2, \infty$, we have three triangular inequalities in d_1, d_2 , and d_∞ . Thus d_1, d_2 , and d_∞ defines metrics on $M \times N$. Let $(a_n, x_n) \in M \times N$ such that $a_n \rightarrow a$ and $x_n \rightarrow x$ in (M, d) and (N, p) respectively. Then $d(a_n, a) \rightarrow 0$ and $p(x_n, x) \rightarrow 0$. Thus we easily have

$$d_1((a_n, x_n), (a, x)) = d(a_n, a) + p(x_n, x) \rightarrow 0,$$

$$d_2((a_n, x_n), (a, x)) = (d(a_n, a)^2 + p(x_n, x)^2)^{1/2} \rightarrow 0,$$

and

$$d_\infty((a_n, x_n), (a, x)) = \max\{d(a_n, a), p(x_n, x)\} \rightarrow 0.$$

Thus, d_1, d_2, d_∞ enjoy the property of this exercise. Now if $(a_n, x_n) \rightarrow (a, x)$ in $(M \times N, d_1)$, that is $d_1((a_n, x_n), (a, x)) \rightarrow 0$, then we have $d(a_n, a) + p(x_n, x) \rightarrow 0$, thus $d(a_n, a)$ and $p(x_n, x)$ converge to 0. Because d_2 defines a metric on $M \times N$, we also have $d_2((a_n, a), (x_n, x)) \rightarrow 0$, thus $(a_n, x_n) \rightarrow (a, x)$ in $(M \times N, d_2)$. Similarly, if $(a_n, x_n) \rightarrow (a, x)$ in $(M \times N, d_2)$, then we have $(a_n, x_n) \rightarrow (a, x)$ in $(M \times N, d_\infty)$, and if $(a_n, x_n) \rightarrow (a, x)$ in $(M \times N, d_\infty)$, then we have $(a_n, x_n) \rightarrow (a, x)$ in $(M \times N, d_1)$. Thus d_1, d_2 , and d_3 are equivalent. □

Lemma 1. Let $a, b, c, d \in \mathbb{R}_+$, if $a \leq c$ and $b \leq d$, then for any $i \geq 1$, we have $\|(a, b)\|_i \leq \|(c, d)\|_i$.

Proof. Indeed, if $i \geq 1$, then $a^i \leq c^i$ and $b^i \leq d^i$. And since $\frac{1}{i} > 0$, we have

$$\|(a, b)\|_i = (|a|^i + |b|^i)^{1/i} \leq (|c|^i + |d|^i)^{1/i} = \|(c, d)\|_i.$$

Thus $\|(a, b)\|_i \leq \|(c, d)\|_i$. □

Exercise Hint. Let $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ defines by $f(x, y) = xy$, prove that f is continuous.

Proof. For $(x_0, y_0) \in \mathbb{R}^2$, assume that $|x - x_0| < \epsilon$ and $|y - y_0| < \delta$, then we have

$$\begin{aligned} |xy - x_0y_0| &= |(x - x_0)(y - y_0) + x_0y + y_0x - 2y_0x_0| \\ &= |(x - x_0)(y - y_0) + x_0(y - y_0) + y_0(x - x_0)| \\ &\leq |x - x_0||y - y_0| + |x_0||y - y_0| + |y_0||x - x_0| \\ &\leq \epsilon\delta + |x_0|\delta + |y_0|\epsilon. \end{aligned}$$

Because $|xy - x_0y_0|$ can be as small as possible by letting ϵ and δ super small, f is continuous at (x_0, y_0) . Therefore, f is continuous everywhere. □