

Linear Operator Theory

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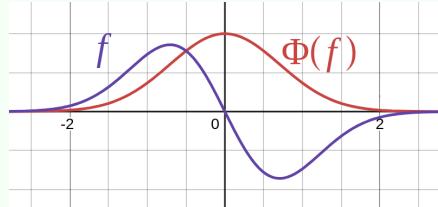
0. Contents

1 Topological Spaces	5
1.1 Random Notes	5
1.2 Solved Problems	8
1.2.1 Metric Spaces	8
1.2.2 Examples of Metric Spaces	12
1.2.3 Continuity	15
1.2.4 Convergence of Sequences	18
2 Algebraic Structure	23
2.1 Problems	26
2.1.1 Linear Spaces and Linear Subspaces	26
2.1.2 Linear Transformation	29
2.1.3 Inverse Transformation	30
2.1.4 Isomorphisms	32
2.1.5 Linear Independence and Dependence	34
3 Combined Topological and Algebraic Structures	39
3.1 Problems	39
3.1.1 Normed Linear Spaces	39

1. Topological Spaces

1.1 Random Notes

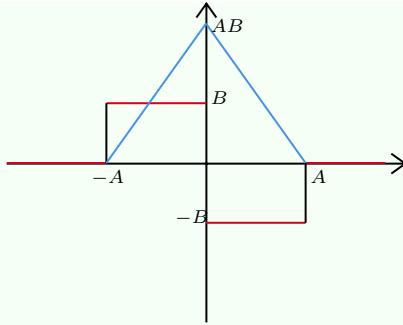
Observation 1.1.1 — Some notes on Example 2 page 63. In this example we study $X \subset L^2(-\infty, +\infty)$ the functions in $L^2(-\infty, +\infty)$ that their primitive functions also belongs to $L^2(-\infty, +\infty)$. I.e. they roughly look like the following function



Then we define the map $\Phi : X \rightarrow X$ given by

$$\Phi(x)(t) = \int_{-\infty}^t x(\tau)d\tau.$$

We want to show that this map is not continuous at any point of its domain. Since $L^2(-\infty, +\infty)$ is a vector space, it is enough to show this for the origin. So we want to show that this map is not continuous at the origin. Let x be the function whose graph is given as below (the red curve). It is easy to see that the blue curve is its primitive function.



The L^2 norm is roughly the area under the curve, so the L^2 norm of x is AB while the L^2 norm of $\Phi(x)$ is AB^2 . Now to prove that the mapping Φ is not continuous at the origin, we need to change A, B such that the L^2 norm x goes to zero but the norm of $\Phi(x)$ gets larger and larger. Suppose we want to make the norm of x to be half and double the norm of $\Phi(x)$. So we need to solve

$$\frac{A'B'}{AB} = \frac{1}{2}, \quad \frac{A'B'^2}{AB^2} = 1.$$

It is easy to see that we need to have

$$\frac{A'}{A} = 4, \quad \frac{B'}{B} = \frac{1}{2}.$$

Observation 1.1.2 — Some notes on Example 4, page 64. Putting this example in contrast to the example above, teaches us a lot! In the example above we observed that the “integration” of $L(-\infty, +\infty)$ functions is nowhere continuous. However, in this example, we will see that the integration operator in $L^2[0, T]$ is a uniformly continuous operator! This is a huge contrast and the surprising fact is that in both cases we are dealing with the “same” operator and the only difference is their domain. To see this let $X = L^2[0, T]$ and let $\Phi : L^2[0, T] \rightarrow L^2[0, T]$ defined as

$$[\Phi(x)](t) = \int_0^t x(\tau) d\tau.$$

■ **Remark 1.1** We need to check to see if $\Phi(x)$ really belongs to $L^2[0, T]$. However the argument will be very similar to what we will see below, so we have skipped this part.

Since $L^2[0, T]$ is a vector space and Φ is a linear map, it is enough to show the continuity

at the origin. Let $x \in L^2[0, T]$. Then

$$\begin{aligned} d(\Phi(x), 0)^2 &= \int_0^T \left| \int_0^t [\Phi(x)](\tau) d\tau \right|^2 dt \leq \int_0^T \left(\int_0^t |[\Phi(x)](\tau)| d\tau \right)^2 dt \\ &\leq \int_0^T \left(\int_0^T |[\Phi(x)](\tau)| d\tau \right)^2 dt \\ &\leq \int_0^T \left(\int_0^T 1 d\tau \right) \left(\int_0^T |[\Phi(x)](t)|^2 d\tau \right) dt \\ &= T^2 \int_0^T |[\Phi(x)](t)|^2 d\tau \\ &= T^2 d(x, 0). \end{aligned}$$

So we have

$$d(\Phi(x), 0) \leq T d(x, 0).$$

Since the expansion factor T does not depend on x , we can conclude that Φ is actually uniformly continuous.

Observation 1.1.3 — Discrete version of the example above. We can construct a discrete version of the example above, where the underlying space is \mathbb{R}^n , and the integration operator Φ is a linear map $\Phi : \mathbb{R}^n \rightarrow \mathbb{R}^n$ given by

$$\Phi : \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix} \mapsto \begin{bmatrix} x_1 \\ x_1 + x_2 \\ x_1 + x_2 + x_3 \\ \vdots \\ x_1 + \cdots + x_n \end{bmatrix}.$$

This can be represented by the matrix

$$M = \begin{bmatrix} 1 & 0 & 0 & 0 & \cdots & 0 \\ 1 & 1 & 0 & 0 & \cdots & 0 \\ 1 & 1 & 1 & 0 & \cdots & 0 \\ 1 & 1 & 1 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & 1 & 1 & \cdots & 1 \end{bmatrix}$$

However, from the Example 1 from page 62 we know that a linear map between vector spaces is continuous with the operator norm at most $\sum_{ij} |m_{i,j}|$. So the operator norm of this matrix will be

$$\|\Phi\| \leq \left(\frac{n(n+1)}{2} \right)^{1/2} \approx n.$$

Summary ↪ 1.1 — Some notes on Example 5 on page 65. The example 5 demonstrates the observation above in a very useful way. It explicitly shows that the example above is just

a very special matrix and we can have more general linear maps with arbitrary elements in their matrix representation. So the matrix operators between finite dimensional vector spaces correspond to the integration kernels. consider $L_2[0, T]$ and let $k \in L^2([0, T] \times [0, T])$, i.e.

$$\int_0^T \int_0^T |k(t, \tau)|^2 d\tau dt = M^2 < \infty.$$

And we define the linear map $\Phi_k : L_2[0, T] \mapsto L_2[0, T]$

$$[\Phi_k(x)](t) = \int_0^T k(t, \tau)x(\tau)d\tau.$$

The Example 5 shows that this is a continuous map (bounded linear map) with operator norm

$$\|\Phi_k\| \leq M.$$

Compare this with Example 1.

■ **Remark 1.2** One special kind of kernel integrals are convolutions where the kernel is of the form $k(t, \tau) = h(t - \tau)$ for some function h of appropriate type (I think it is enough to have $L_2[0, T]$, with the definition that $h(t - \tau) = 0$ if $t - \tau \notin [0, T]$). We often write convolution as

$$(f * h)(t) = \int_0^t h(t - \tau)f(\tau)d\tau.$$

1.2 Solved Problems

1.2.1 Metric Spaces

Summary ↗ 1.2 — A metric for bounded functions defined on any set S . Let S be a nonempty set and let $X = B(S)$ denote the collection of all bounded real-valued functions defined on S . Then

$$d(f, g) = \sup\{|f(s) - g(s)| : s \in S\},$$

is a metric on X .

Summary ↗ 1.3 — A metric for differential operators. Let $X(n)$ denote all the differential operators of the form

$$P(D) = p_0 D_0 + p_1 D_1 + \cdots + p_n D_n.$$

Then d is a metric on $X(n)$: $d(P(D), Q(D)) = \sum_{i=0}^n |p_i - q_i|$.

Summary ↗ 1.4 Let X denote the collection of all bounded closed intervals $[a, b]$ from the real line. Let $[a, b]$ and $[c, d]$ be two sets. Then the symmetric difference $[a, b] \delta [c, d]$ is the union of at most two bounded intervals. $\rho([a, b], [c, d])$ to be the sum of the lengths of these intervals. Then ρ is a pseudometric. In particular, the distance between any two single points is zero.

■ **Remark 1.3 TODO:** I am still thinking why the triangle inequality of the metric defined above holds.

■ **Problem 1.2.1.1** Let d be a metric on X and let $d_\alpha = \alpha d$, where $0 < \alpha < 1$. Show that $d_\alpha \not\equiv d$ if X has more than two points.

Solution Let X have more than one point. Then $\exists x, y \in X$ such that $d(x, y) \neq 0$. Then $d(x, y) \neq \alpha d(x, y) = d_\alpha(x, y)$. So $d \not\equiv d_\alpha$.

■ **Problem 1.2.1.2** Show that $d(x, y) = |x - y|$ is a metric on the real number \mathbb{R} . Show that it is also a metric on the set of complex numbers \mathbb{C} .

Solution First, we show (\mathbb{R}, d) is a metric space. We need to show:

- (i) First, we show that d is always positive. Since $|x| = \max\{x, -x\}$, we have $\pm x \leq |x|$. Adding to two inequalities we will get $0 \leq |x|$. Now we want to show that $|x - y| = 0$ then $x = y$. $|x - y| = 0$ implies $x - y = 0$, and then we will have $x = y$.
- (ii) Since $|x - y| = \max|x - y, y - x|$, we will have $|y - x| = \max|y - x, x - y| = |x - y|$.
- (iii) Triangle inequality: From the definition of absolute value we have $|x| = \max\{x, -x\}$. So $x \leq |x|$ and $-|x| \leq x$. Let $a, b \in \mathbb{R}$. Then

$$a \leq |a|, b \leq |b| \implies a + b \leq |a| + |b|.$$

Furthermore,

$$-|a| \leq a, -|b| \leq b \implies -|a| - |b| \leq a + b.$$

Combining these two inequalities, we will get

$$-|a| - |b| \leq a + b \leq |a| + |b|.$$

So

$$|a + b| \leq |a| + |b|.$$

Now we show that (\mathbb{C}, d) is a metric space. For this purpose we can utilize the properties of the complex conjugation and work with the definition $|z_1 - z_2| = \sqrt{(z_1 - z_2)(\overline{z_1 - z_2})}$. But instead, we want to show that there is a bijection from (\mathbb{C}, d) (we don't know yet if it is a metric space, otherwise we could call this map an isometric bijection) to $(\mathbb{R}^2, d_{\mathbb{R}^2})$. Recall

$$\mathbb{C} = \{a + ib : a, b \in \mathbb{R}\}.$$

Let $\phi : \mathbb{C} \rightarrow \mathbb{R}^2$ given by $a + ib \mapsto (a, b)$. This is a bijection, since for any $a + ib \in \mathbb{C}$ we have a unique $(a, b) \in \mathbb{R}^2$ (uniquely determined by a, b), and for any $(a, b) \in \mathbb{R}^2$ we have a uniquely determined $a + ib \in \mathbb{C}$. Also we claim that

$$d(z, w) = d_{\mathbb{R}^2}(\phi(z), \phi(w)).$$

This is true because

$$d(z, w) = d(z_1 + iz_2, w_1 + iw_2) = \sqrt{(z_1 - w_1)^2 + (z_2 - w_2)^2} = d_{\mathbb{R}^2}((z_1, z_2), (w_1, w_2)) = d_{\mathbb{R}^2}(\phi(z), \phi(w)).$$

We can use this map to transfer all of the properties of $d_{\mathbb{R}^2}$ in \mathbb{R}^2 to d in \mathbb{C} . So (\mathbb{C}, d) is indeed a metric space. So now we can call ϕ an isometric bijection.

■ **Remark 1.4** In the question above, for (\mathbb{R}, d) we used the definition $|x| = \max\{x, -x\}$, and for (\mathbb{C}, d) we used the definition $|x| = \sqrt{x\bar{x}}$.

■ **Problem 1.2.1.3** Let $d(x, y)$ be a metric on X . Show that

$$d_1(x, y) = \frac{d(x, y)}{1 + d(x, y)}, \quad d_2(x, y) = \min\{1, d(x, y)\},$$

are also metrics on X . Show that every set in the metric space (X, d_1) and (X, d_2) is bounded.

Solution (i) Showing that $d_1(x, y)$ is a metric: Positive definiteness: since the numerator and denominator are both positive, then $d_1(x, y)$ is also positive for all $x, y \in X$. Let $x, y \in X$ such that $d_1(x, y) = 0$. This implies $d(x, y) = 0$, hence $x = y$. Symmetry follows from d being symmetric. For the triangle inequality we use the fact that $f(x) = \frac{x}{1+x}$ is strictly increasing for $x \geq 0$ (because it has positive derivative), and also it is subadditive. To see the subadditivity, observe that

$$f(x+y) \leq \frac{x+y}{1+x+y} = \frac{x}{1+x+y} + \frac{y}{1+x+y} \leq \frac{x}{1+x} + \frac{y}{1+y} = f(x) + f(y).$$

Using the fact that for all $x, y, z \in X$ we have

$$d(x, z) \leq d(x, y) + d(y, z),$$

using the monotonicity of f and then its sub additivity, we will get

$$d_2(x, z) \leq d_2(x, y) + d_2(y, z).$$

(See the summary box below for a detailed argument of a more general setting).

(ii) Showing that $d_2(x, y) = d(x, y) \wedge 1$ is a metric: Positive definiteness: Since $d(x, y) \geq 0$ then $d(x, y) \wedge 1 \geq 0$. Let $x, y \in X$ such that $d(x, y) \wedge 1 = 0$. So $d(x, y) = 0$. This implies $x = y$. So d_2 is positive definite. Symmetric: Let $x, y \in X$. Then $d_2(x, y) = d(x, y) \wedge 1 = d(y, x) \wedge 1 = d_2(y, x)$, where we have used the fact that $d(x, y) = d(y, x)$. So $d_2(x, y) = d_2(y, x)$. For the triangle inequality we use the following lemma:

Lemma 1.1 Let $a, b \geq 0$. Then

$$a < b \implies a \wedge 1 \leq b \wedge 1.$$

Also

$$(a+b) \wedge 1 \leq a \wedge 1 + b \wedge 1.$$

Proof. For the first implication, when $a < b$, then there are three cases: $0 \leq a < b \leq 1$, $0 \leq a \leq 1 < b$, $1 < a < b$. In the first case $a \wedge 1 = a$, $b \wedge 1 = b$. So $a \wedge 1 \leq b \wedge 1$ holds. This holds for the other cases as well.

For the second implication, again we will show in cases: When $a+b < 1$, $a+b = 1$, $a+b > 1$. For the first case we can only have $a, b < 1$. So $(a+b) \wedge 1 = a+b$, $a \wedge 1 = a$, $b \wedge 1 = b$. So the desired inequality holds. For the second case, again $a = a \wedge 1$, $b = b \wedge 1$, and $(a+b) \wedge 1 = a+b = 1$. So the desired inequality holds. For the third case, WLOG we can assume that $a \leq b$. Since $a+b \geq 1$, then at least one of a or b should be larger than 1. WLOG we can assume $a \geq 1$. So $(a+b) \wedge 1 = 1$ and $a \wedge 1 = 1$. So the desired inequality holds. \square

So by the triangle inequality $d(x, z) \leq d(x, y) + d(y, z)$, and using the lemma above we can conclude that

$$d(x, z) \wedge 1 \leq d(x, y) \wedge 1 + d(y, z) \wedge 1.$$

Now to show that all of the sets in X , with the metrics above are bounded, it is enough to observe that for all $x, y \in X$ we have $d_1(x, y) \leq 1$ and $d_2(x, y) \leq 1$.

Summary ↗ 1.5 — Transformation of the metric. Using the proof ideas of the examples above, we can generalize it in the following proposition.

Proposition 1.1 Let (X, d) be a metric space and $\phi : [0, +\infty) \rightarrow \mathbb{R}$ be a real valued functions such that

- (i) $\phi(x) = 0$ if and only if $x = 0$.
- (ii) monotone increasing,
- (iii) subadditive: $\forall x, y \geq 0$ we have $\phi(x + y) \leq \phi(x) + \phi(y)$. Then $\tilde{d} = \phi \circ d$ is also a metric, and (X, \tilde{d}) is a metric space.

Proof. Since $\phi(x) = 0$ if and only if $x = 0$, then $\tilde{d} = 0$ if and only if $d = 0$, if and only if $x = y$. So \tilde{d} is positive definite. \tilde{d} is also symmetric, because $\forall x, y \in X$, we have $d(x, y) = d(y, x)$ that implies $\tilde{d}(x, y) = \phi(d(x, y)) = \phi(d(y, x)) = \tilde{d}(y, x)$. For the triangle inequality, we have

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

We apply (ii) above and we will get

$$\tilde{d}(x, z) \leq \phi(d(x, y) + d(y, z)),$$

and we now apply (iii) above to get

$$\tilde{d}(x, z) \leq \tilde{d}(x, y) + \tilde{d}(y, z).$$

□

Summary ↗ 1.6 — Metric transformation examples. Using the example above, and the remark below, the followings are example functions that can transform a metric d to a new metric $\tilde{d} = f(d)$.

- (i) $f_1(x) = x \wedge 1$.
- (ii) $f_2(x) = x/(1 + x)$.
- (iii) $f_3(x) = x^\alpha/(1 + x^\alpha)$ for $0 < \alpha \leq 1$.

■ **Remark 1.5** The reason that the example (iii) above works is that f_3 satisfies all the properties in the summary box above. To see the sub-additivity, it is enough to show the sub-additivity of x^α when $0 < \alpha \leq 1$. Let $\alpha = 1 - \beta$. Then

$$(x + y)^\alpha = \frac{x + y}{(x + y)^\beta} < \frac{x}{(x + y)^\beta} + \frac{y}{(x + y)^\beta} \leq \frac{x}{x^\beta} + \frac{y}{y^\beta} = x^\alpha + y^\alpha.z$$

■ **Problem 1.2.1.4** A real-valued function $\rho(x, y)$ is said to be a pseudometric on X if it satisfies conditions (M1), (M3), and (M4).

- (a) Show that $\rho(x, y) \equiv 0$ is a pseudometric on any set X .

- (b) Show that $\rho((x_1, x_2), (y_1, y_2)) = |x_1 - y_1|$ is a pseudometric in the plane \mathbb{R}^2 .

Solution (a) $\rho(x, y) \equiv 0$ satisfies (M1), (M3), and (M4) vacuously. However, if X has only one element, then ρ is indeed a metric. But if X has more than two elements, then $\exists x, y \in X$ such that $x \neq y$ but $\rho(x, y) = 0$.

- (b) Being symmetric (M3) and the triangle inequality (M4) follows from the properties of $|\cdot|$. However, for all $x, y \in \mathbb{R}^2$ that have the same first component (i.e. points on the vertical lines) have $d(x, y) = 0$. However, this will be a metric on the quotient space \mathbb{R}^2/W where $W = \text{Span}\{(0, 1)\}$.
-

■ **Problem 1.2.1.5** Show that if A is nonempty, in a metric space (X, d) , then $\text{diam } A = 0$ if and only if A consist of a single point. Is this true in a pseudometric space?

Solution For the forward direction assume for $A \subset X$ we have $\text{diam } A = 0$. So

$$\sup_{x, y \in A} \{d(x, y)\} = 0.$$

This implies that $\forall x, y \in A$ we have $d(x, y) = 0$. Since d is a metric, then $A = \{x\}$ is a singleton. For the converse, let $A = \{x\}$. Then it follows immediately that $\text{diam } A = 0$.

No this is not true in the case of the pseudometric spaces. In our proof above, in the forward direction, the logic breaks if d is a pseudometric. I.e. it is possible to have $\sup_{x, y \in A} \{d(x, y)\} = 0$ but A is not singleton.

1.2.2 Examples of Metric Spaces

■ **Problem 1.2.2.1** Let $X = \mathbb{R}^2$ and let

$$d(x, y) = (|x_1 - y_1|^{1/2} + |x_2 - y_2|^{1/2})^2.$$

Is (X, d) a metric space?

Solution No. Because the triangle inequality breaks. Let

$$x = (0, 0), \quad y = (0, 1/4), \quad z = (1/4, 1/4).$$

It is easy to check that the triangle inequality $d(x, z) \leq d(x, y) + d(y, z)$ does not hold. Because

$$1 = d(x, z) \leq d(x, y) + d(y, z) = 1/4 + 1/4 = 1/2,$$

is a contradiction.

■ **Remark 1.6** With a similar idea of proof, it is easy to show that d_p fails to satisfy the triangle inequality if $0 < p < 1$.

■ **Problem 1.2.2.2** In \mathbb{R}^2 let $A\{x = (x_1, x_2) : (|x_1|^2 + |x_2|^2)^{1/2} < 1\}$, and

$$d(x, y) = \{|x_1 - y_1|^2 + |x_2 - y_2|^2\}^{1/2}.$$

Compute $d(x, A)$. Show that $d(x, A) = 0$ if and only if

$$|x_1|^2 + |x_2|^2 \leq 1.$$

Solution Let $x = (x_1, x_2) \in \mathbb{R}^2$ (WLOG we assume that $x_1 > 0$) that does not belong to A . The ray that passes through x is given by $y = \frac{x_2}{x_1}x$, and this ray intersects the unit circle at

$$\tilde{x} = \left(\frac{1}{\sqrt{1 + (x_2/x_1)^2}}, \frac{x_2/x_1}{\sqrt{1 + (x_2/x_1)^2}} \right).$$

So the distance between x and the point above is

$$d(x, \tilde{x}) = \left(\left| \frac{1}{\sqrt{1 + (x_2/x_1)^2}} - x_1 \right|^2 + \left| \frac{x_2/x_1}{\sqrt{1 + (x_2/x_1)^2}} - x_2 \right|^2 \right)^{1/2}.$$

To show that $d(x, A) = 0$ if and only if $x_1^2 + x_2^2 \leq 1$, we first do the forward direction. Let $d(x, A) = 0$. From our explicit formula above, it is easy to see that we should have

$$x_1 = \frac{1}{\sqrt{1 + (x_2/x_1)^2}}, \quad x_2 = \frac{x_2/x_1}{\sqrt{1 + (x_2/x_1)^2}}.$$

It is easy to check that $x_1^2 + x_2^2 = 1$ which implies $x_1^2 + x_2^2 \leq 1$. For the converse direction, we assume $x_1^2 + x_2^2 \leq 1$. When $x_1^2 + x_2^2 < 1$, then $x \in A$, and by the definition of $d(x, A)$ we see $d(x, A) = 0$. When $x_1^2 + x_2^2 = 1$, then we can write $1 + (x_1/x_2)^2 = 1/x_1^2$. Using the fact that $x_1 > 0$ this implies that

$$d(x, A) = 0,$$

where we have used the explicit formula above.

■ **Problem 1.2.2.3** In example 5, the metric d_∞ was defined with a “sup” instead of “max”. In order to see the necessity of this let $x = (x_1, x_2, \dots), y = (y_1, y_2, \dots)$ be given by

$$x_n = \frac{1}{n+1}, \quad y_n = \frac{n}{n+1}.$$

And argue why “sup” should be used instead of “max”? What about Example 8 and Example 11?

Solution The reason is that if on ℓ_∞ we define d_∞ by “max”, then for the presented examples $d_\infty(x, y)$ does not exist. In the case of example 8, we **can** replace “max” with “sup”. The reason is that if $x, y \in C[0, T]$, then $f = |x - y| \in C[0, T]$. On the other hand, a continuous function defined on a compact set, attains its maximum and minimum. For Example 11, we can **not** use “max” in place of “sup”.

■ **Problem 1.2.2.4** Sketch the set of points $x = (x_1, x_2)$ in \mathbb{R}^2 for which

$$d_p(0, x) = 1,$$

where $0 = (0, 0)$ and d_p is defined in Example 1. With x, y fixed show that $d_p(x, y)$ is decreasing in p . Hence

$$d_p(x, y) \leq d_q(x, y).$$

Solution It is easy to find such sketches online. For the second part, fix $x, y \in \mathbb{R}^2$. Then

$$\frac{dd_p(x, y)}{dp} = -\frac{A}{p^2},$$

where A is a positive quantity (can be calculated explicitly by using the chain rule). So $d_p(x, y)$ is decreasing.

Summary ↗ 1.7 For fixed $x, y \in \mathbb{R}^n$, $d_p(x, y)$ is decreasing in p . So

$$p \leq q \implies d_p(x, y) \geq d_q(x, y).$$

■ **Problem 1.2.2.5** Let C be the unit circle in the complex plane, that is $C = \{z : |z| = 1\}$. Let X denote all complex-valued functions $f(z)$ defined on C for which

$$\int_C |f(z)| dz < \infty.$$

Show that

$$d(f, g) = \left(\int_C |f(z) - g(z)| dz \right)^{1/2} = \left(\int_0^{2\pi} |f(e^{2i\pi\theta}) - g(e^{2i\pi\theta})| d\theta \right)^{1/2}$$

is a metric on X .

Solution It is easy to check the other properties of being a metric. Thus we will show that d satisfies the triangle inequality. To see this let $f, g, h \in X$. Then

$$d(f, g) = \left(\int |\tilde{f} - \tilde{g}|^2 \right)^{1/2} = \left(\int |\tilde{f} - \tilde{g} \pm \tilde{h}|^2 \right)^{1/2} = \left(\int |u + v|^2 \right)^{1/2},$$

where $u = \tilde{f} - \tilde{h}$ and $v = \tilde{h} - \tilde{g}$, and $\tilde{f}(\theta) = f(e^{2i\pi\theta})$. Observe that

$$\int |u + v|^2 = \int (u + v)\overline{(u + v)} = \int (|u|^2 + |v|^2 + 2 \operatorname{Re}(u\bar{v})) = \int |u|^2 + \int |v|^2 + 2 \int \operatorname{Re}(u\bar{v}).$$

Also

$$\int \operatorname{Re}(u\bar{v}) \leq \operatorname{Re}(\int u\bar{v}) \leq \operatorname{Re}(\|u\|\|v\|) = \|u\|\|v\|,$$

where $\|\cdot\|$ is the norm that induces the metric. So

$$\int |u + v|^2 \leq \|u\|^2 + \|v\|^2 + 2\|u\|\|v\| = (\|u\| + \|v\|)^2.$$

so we can conclude

$$d(f, g) = \left(\int |u + v|^2 \right)^{1/2} \leq \|u\| + \|v\| = d(f, h) + d(h, g).$$

■ **Problem 1.2.2.6** Let X denote the class of all complex-valued functions $f(z)$ that are analytic for $|z| < 1$. Let $f(z) = \sum_{n=0}^{\infty} a_n z^n$ and $g(z) = \sum_{n=0}^{\infty} b_n z^n$ be the power series expansion for f, g in X . Show that the following functions are metrics, or pseudo-metrics.

- (a) $d(f, g) = \sup\{|f(z) - g(z)| : |z| \leq \rho\}$, where $0 < \rho < 1$.
- (b) $d(f, g) = |a_0 - b_0|$.
- (c) $d(f, g) = |a_0 - b_0| + |a_1 - b_1|$.
- (d) $d(f, g) = \sum_{n=0}^{\infty} |a_n - b_n| \rho^n$, where $0 < \rho < 1$.

(e)

$$d(f, g) = \sup \left\{ \left| \int_{|\xi|=0} \frac{f(\xi) - g(\xi)}{z - \xi} d\xi \right| : |z| < \rho \right\}, \quad \text{where } 0 < \rho < 1.$$

Solution Since we need to choose between a metric or a pseudometric, we don't check the triangle inequality.

- (a) Is a metric.
- (b) Is a pseudometric. Because the distance between functions in which in their expansion only the first term is the same will be zero.
- (c) Is a pseudometric. Because of a similar argument as above.
- (d) Is a metric. Because if $d(f, g) = 0$, then every term in the positive series should be zero, hence $|a_n - b_n| = 0$ for all $n \in \mathbb{N}$.
- (e) Still thinking!

1.2.3 Continuity

■ **Problem 1.2.3.1** Let $C^\infty[0, T]$ denote the set of all real-valued smooth functions defined on $[0, T]$. Let D be the mapping of $C^\infty[0, T]$ to itself defined by

$$Df = \frac{df}{dt}.$$

- (a) Let d_1 be the sup-metric on $C^\infty[0, T]$. Is D a continuous mapping of $(C^\infty[0, T], d_1)$ into itself?
- (b) Define a metric d_2 on $C^\infty[0, T]$ by

$$d_2(x, y) = d_1(x, y) + d_1(Dx, Dy).$$

Is D a continuous mapping of $(C^\infty[0, T], d_2)$ into $(C^\infty[0, T], d_1)$?

Solution (a) No. Consider the sequence of functions $(f_n)_n$ where

$$f_n(t) = \frac{\sin(2^n x)}{n}.$$

Then $d_1(f_n, 0) \rightarrow 0$ as $n \rightarrow \infty$ but $d_1(Df_n, 0)$ becomes arbitrary large.

- (b) Yes. Let $f, g \in C^\infty[0, T]$ then

$$d_1(Df, Dg) \leq d_1(Df, Dg) + d_1(f, g) = d_2(f, g).$$

So for any $\epsilon > 0$ we let $\delta = \epsilon$. Then if $d_2(f, g) < \delta$ it implies that $d_1(f, g) < \epsilon$.

■ **Problem 1.2.3.2** A delay line is a device whose output is ideally a delayed version of its input, i.e.

$$y(t) = x(t - \tau).$$

Suppose that $x, y \in L_2(-\infty, +\infty)$, where $L_2(-\infty, +\infty)$ has the usual metric. Is the mathematical model of the delay line a continuous mapping of $L_2(-\infty, +\infty)$ into itself?

Solution We know that integration, in the sense of Lebesgue, is not sensitive to shift. I.e.

$$\int_{-\infty}^{+\infty} f(t)dt = \int_{-\infty}^{+\infty} f(t - \tau)dt,$$

for all $\tau \in \mathbb{R}$. So for any $f, g \in L_2(-\infty, +\infty)$ we have

$$d(f, g) = d(Sf, Sg).$$

■ **Problem 1.2.3.3** Let $Y = C[0, T]$ be given with the sup-metric $d(x, y)$. Let

$$X = (C[0, T], d) \times (C[0, T], d),$$

be the product space. Consider the mapping F of X into Y defined by $F(x) = x_1x_2$, where $x = (x_1, x_2)$. (That is F is a multiplier). Is F continuous? Is F uniformly continuous?

Solution Assume that the norm on X is $d_2(f, g) = d(f_1, g_1) + d(f_2, g_2)$ (we could chose other metrics on the product space as well, but we chose this one for the convenience). Then we can write

$$\begin{aligned} d(F(f), F(g)) &= \sup_{t \in [0, T]} |[F(f)](t) - [F(g)](t)| \\ &= \sup_{t \in [0, T]} |f_1(t)g_1(t) - f_2(t)g_2(t)| \\ &= \sup_{t \in [0, T]} |f_1(t)g_1(t) - f_2(t)g_2(t) \pm f_1(t)g_2(t)| \\ &= \sup_{t \in [0, T]} |f_1(t)(g_1(t) - g_2(t)) + g_2(t)(f_1(t) - f_2(t))| \\ &\leq \|f_1\|_\infty d(g_1, g_2) + \|g_2\|_\infty d(f_1, f_2) \\ &\leq \max\{\|f_1\|_\infty, \|g_2\|_\infty\}(d(g_1, g_2) + d(f_1, f_2)) \\ &= \max\{\|f_1\|_\infty, \|g_2\|_\infty\}d_2(f, g). \end{aligned}$$

So for a given $\epsilon > 0$ we can let $\delta = \epsilon / (\max\{\|f_1\|_\infty, \|g_2\|_\infty\})$. However, since the choice of δ depends on the point f, g , then the uniform continuity fails.

■ **Problem 1.2.3.4** Let (X, d) be a metric space, where X is nonempty. Let $Y = BC(X, \mathbb{R})$ denote the collection of all bounded, continuous real-valued functions defined on X .

(a) Show that the functions

$$\begin{aligned} f_1 : x &\mapsto \frac{d(x, x_0)}{1 + d(x, x_0)}, \\ f_2 : x &\mapsto d(x, x_1) - d(x, x_0), \\ f_3 : x &\mapsto 3, \end{aligned}$$

are in Y .

(b) Show that

$$\sigma(f, g) = \sup\{|f(x) - g(x)| : x \in X\}$$

is a metric on Y .

Solution (a) Fix $x_0 \in X$. We want to show the function $f : X \mapsto \mathbb{R}$ given as $f(x) = d(x, x_0)$ is a continuous function. Because

$$d(f(x), f(y)) = |f(x) - f(y)| = |d(x, x_0) + d(y, y_0)| \leq |d(x, y) + d(y, x_0) - d(y, x_0)| = d(x, y).$$

On the other hand since the function $g : \mathbb{R} \rightarrow \mathbb{R}$ given by $x \mapsto x/(1+x)$ is a bounded continuous function, then $f_1 = d/1+d$ is also a bounded continuous function. Continuity of f_2 follows from a similar reasoning. For the boundedness, observe that

$$d(x, x_1) - d(x, x_0) \leq d(x, x_0) + d(x_0, x_1) - d(x, x_0) = d(x_0, x_1).$$

(b) Follows from the result of Exercise 13 page 56.

■ **Problem 1.2.3.5** Let $f : X \rightarrow X$ be a continuous mapping, where X has a metric d . Let $G(f)$ denote the graph of f in $X \times X$, and Δ the diagonal set

$$\Delta = \{(x, y) : x = y\} = \{(x, x) : x \in X\}.$$

Assume that $X \times X$ has the metric

$$d((x_1, y_1), (x_2, y_2)) = d(x_1, x_2) + d(y_1, y_2).$$

Define $g : \Delta \rightarrow G(f)$ by

$$g(x, x) = (x, f(x)).$$

Show that g is continuous. Show that g is invertible. Is g^{-1} continuous?

Solution g is continuous. To see this let $(x_1, x_1), (x_2, x_2) \in \Delta$. So

$$\begin{aligned} d(g(x_1, x_1), g(x_2, x_2)) &= d((x_1, f(x_1)), (x_2, f(x_2))) \\ &= d(x_1, x_2) + d(f(x_1), f(x_2)) \\ &\leq (1 + A(x_1, x_2))d(x_1, x_2) \\ &= \frac{1}{2}(1 + A(x_1, x_2))d((x_1, x_1), (x_2, x_2)). \end{aligned}$$

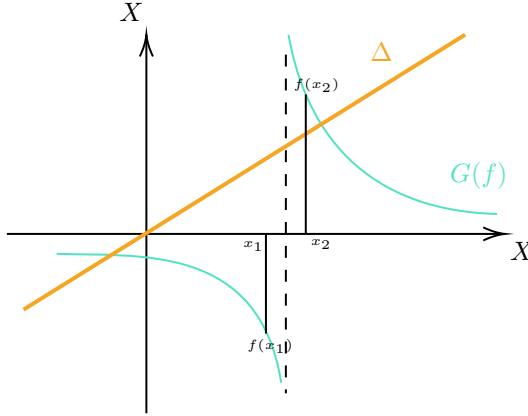
where we have used the continuity of f , i.e. $d(f(x_1), f(x_2)) \leq A(x_1, x_2)d(x_1, x_2)$. So g is a continuous mapping. Define the inverse of g as

$$h = g^{-1} : G(f) \rightarrow \Delta, \quad g^{-1}(x, f(x)) = (x, x).$$

g^{-1} is indeed inverse of g as it is straightforward to check $g^{-1} \circ g$ and $g \circ g^{-1}$ are both identity maps. However, $h = g^{-1}$ is not necessarily continuous. To see this let $(x_1, f(x_1)), (x_2, f(x_2)) \in G(f)$. Then

$$d(h(x_1, f(x_1)), h(x_2, f(x_2))) = d((x_1, x_1), (x_2, x_2)) = d(x_1, x_2) + d(x_1, x_2) = 2d(x_1, x_2).$$

So we can come up with examples where (x_1, x_1) and (x_2, x_2) are close to each other, but $(x_1, g(x_1))$ and $(x_2, g(x_2))$ are not. For instance consider the following graph: Note that the diagram above is just a schematic representation of the idea behind the provided example to demonstrate some cases where g^{-1} is not continuous, and there is no particular meaning behind the fact that $X \times X$ looks like the Euclidean 2 plane. The only case that this can fail is when all sets in X are bounded (see Exercise 3 page 46).



Observation 1.2.1 — Learning from a mistake. For the question above, after some discussions we found out that our answer is wrong. That is because, assuming the continuity for f , then $d(f(x_1), f(x_2))$ can not get very large, because $d(f(x_1), f(x_2)) \leq M(x_1, x_2)d(x_1, x_2)$.

1.2.4 Convergence of Sequences

■ **Problem 1.2.4.1** Let X denote the set of all bounded piecewise continuous functions defined on $0 \leq t \leq T$, with the sup-metric d_∞ .

- (a) Obviously $C[0, T]$ is a subspace of X . Let x_0 be an arbitrary point in $C[0, T]$. Suppose that x_0 is to be approximated by piecewise constant functions as shown in Figure 3.6.1 in text book. That is

$$x_n(t) = x_0(j \frac{T}{n}) \text{ for } j \frac{T}{n} \leq t < (j+1) \frac{T}{n} \text{ and } j = 0, 1, \dots, (n-1).$$

and

$$x_n(T) = x_0(\frac{n-1}{n}T).$$

Is it true that the sequence $\{x_n\}$ converges in (X, d) ? If so, is it true that $x_0 = \lim_{n \rightarrow \infty} x_n$? [Hint: Use the fact that a real-valued continuous function, defined on a bounded closed interval is uniformly continuous, compare with Exercise 13, Section 17.]

- (b) Suppose x_0 is not restricted to the subspace $C[0, T]$, and suppose that x_0 is approximated by function x_n as in (a). Is it true that $x_0 \lim_{n \rightarrow \infty} x_n$?
- (c) Consider a different metric on X , namely the d_2 metric. Does the sequence (x_n) converges to x_0 in (X, d_2) ?

Solution (a) Since the points in $C[0, T]$ are uniformly continuous, then for any choice of $\epsilon > 0$ we can choose $\delta > 0$ such that $|x - y| < \delta$ implies $|f(x) - f(y)| < \epsilon$. Let $n = T/(\delta/2)$. So on each interval $jT/n \leq t < (j+1)T/n$ we have

$$\sup_{jT/n \leq t < (j+1)T/n} |x_n(t) - x_0(jT/n)| < \epsilon.$$

This also implies

$$\sup_{0 \leq t \leq T} |x_n(t) - x_0(jT/n)| < \epsilon.$$

Thus $x_n \rightarrow x_0$ in d_∞ metric.

- (b) No. The uniform continuity is crucial as observed above. For instance, consider the function $x_0(t) = 0$ when $t \in [0, T)$ and $x_0(t) = 1$ when $t = 1$. With the approximation scheme as above, $x_n \equiv 0$ for all values of $n \in \mathbb{N}$, however, $d_\infty(x_n, x_0) = 1$ for all $n \in \mathbb{N}$.
- (c) Yes. On any closed interval that misses one of the points of discontinuity, the function x_0 is uniformly continuous, hence on this interval x_n converges to x_0 uniformly (in sup norm), thus in d_2 norm. On the intervals that contains a point of discontinuity, we can bound the L_2 norm as follows. Choose n large enough that each point of discontinuity sits inside one interval $jT/n \leq t < (j+1)T/n$, and the length of each interval is less than η . Let $\|x_0\|_\infty = M$ and let $K \in \mathbb{N}$ denote the number of discontinuities. So on the intervals that we have a discontinuity point we have

$$\int_{\text{on inter. w. dis. cont.}} |x(t) - x_0(t)|^2 \leq M^2 K \eta,$$

that goes zero as $\eta \rightarrow 0$.

■ **Problem 1.2.4.2** Suppose that in a metric space (X, d) a sequence $\{x_n\}$ converges to a point x_0 . Does it follow that

$$d(x_1, x_0) \geq d(x_2, x_0) \geq d(x_3, x_0) \geq \cdots \geq d(x_n, x_0) \geq \cdots ?$$

Either prove that it does, or give a counterexample.

Solution No it does not. For instance consider the sequence $\{\sin(n)\}$.

■ **Remark 1.7** However in the question above, we can say if $x_n \rightarrow x_0$ as $n \rightarrow \infty$, then there exists a subsequence $\{x_{n_k}\}$ such that

$$d(x_{k_1}, x_0) \geq d(x_{k_2}, x_0) \geq d(x_{k_3}, x_0) \geq \cdots$$

■ **Problem 1.2.4.3** Consider the metric space (X, d) given in Example 16, Section 3. Characterize the collection of all convergent sequences in (X, d)

Solution In this space a sequence is convergent if and only if it is eventually constant. Then it converges to that constant value.

■ **Problem 1.2.4.4** Consider the sequence (x_n) , where

$$x_n(t) = (\cos(n! \pi t))^{2n}, \quad n = 1, 2, \dots$$

In the metric space $C[0, 1]$ with the sup-metric d_∞ . Is (x_n) a convergent sequence?

Solution Let $p \in \mathbb{Q} \cap [0, 1]$ be a rational number. Then we can write $p = \frac{a}{b}$ for $a, b \in \mathbb{N}$, were $(a, b) = 1$. We claim $\forall n > b$ we have $x_n(p) = 1$. That is because

$$x_n(p) = (\cos\left(n! \pi \frac{a}{b}\right))^{2n} = (\cos\left(n(n-1)(n-1) \cdots (b!) \pi \frac{a}{b}\right))^{2n} = (\cos(2k\pi))^{2n} = 1,$$

where $k \in \mathbb{Z}$. On the other hand, for $q \in [0, 1] \setminus \mathbb{Q}$, since the cos function is equal to one if and only if its argument is a multiple of π , then $\cos(n! \pi q) \neq 1$ for all $n \in \mathbb{N}$. So

$$x_n(q) = (\cos(n! \pi q))^{2n} \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

So the limiting function x can be written as

$$x(t) = \begin{cases} 1 & t \in \mathbb{Q} \cap [0, 1] \\ 0 & t \in [0, 1] \setminus \mathbb{Q} \end{cases}.$$

However, $x \notin C[0, 1]$ so the sequence does not converge.

■ **Problem 1.2.4.5** If $\{x_n\}$ and $\{y_n\}$ are convergent sequences in metric space (X, d) , show that the sequence of real number $\{d(x_n, y_n)\}$ converges to $d(x_0, y_0)$ where $x_0 = \lim_{n \rightarrow \infty} x_n$ and $y_0 = \lim_{n \rightarrow \infty} y_n$. (This exercise should be reconsidered after studying Section 7).

Solution By the definition of converges $x_n \rightarrow x_0$ as $n \rightarrow \infty$ is equivalent to $d(x_n, x_0) \rightarrow 0$ as $n \rightarrow \infty$. Similarly for $\{y_n\}$ we have $d(y_n, y_0) \rightarrow 0$ as $n \rightarrow \infty$. So using the triangle inequality for the metric function we can write

$$d(x_n, y_n) \leq d(x_n, x_0) + d(y_n, x_0) \leq d(x_n, x_0) + d(y_n, y_0) + d(x_0, y_0),$$

and also

$$d(x_0, y_0) \leq d(x_n, x_0) + d(x_n, y_0) \leq d(x_n, x_0) + d(x_n, y_n) + d(y_n, y_0).$$

For any given $\epsilon > 0$ we can choose n large enough such that

$$d(x_n, x_0) < \epsilon/2, \quad \text{and} \quad d(y_n, y_0) < \epsilon/2.$$

So combining the inequalities above we will get

$$d(x_0, y_0) - \epsilon \leq d(x_n, y_n) \leq d(x_0, y_0) + \epsilon.$$

Or equivalently

$$|d(x_n, y_n) - d(x_0, y_0)| < \epsilon.$$

So $d(x_n, y_n) \rightarrow d(x_0, y_0)$ as $n \rightarrow \infty$.

■ **Problem 1.2.4.6** Let a be a real number satisfying $0 \leq a \leq 1$ and set $b = 1 - a$. Let $y_0 = 0$ and

$$y_{n+1} = \frac{1}{2}(b + y_n^2).$$

(a) Show that the sequence $\{y_n\}$ is bounded and monotone, and therefore converges. Let $y = \lim y_n$, and $x = 1 - y$. Show that $x^2 = a$.

(b) Modify part (a) for the case $a > 1$.

Solution (a) First we show the sequence being bounded. We do it by proof by induction. First, observe that $y_0 = 0 \leq 1$. Then assume $y_k \leq 1$ and we show that $y_{k+1} \leq 1$. This is true because

$$y_{k+1} = \frac{1}{2}(b + y_k^2) \leq \frac{1}{2}(1 + 1) = 1.$$

So by induction $y_n \leq 1$ for all $n \in \mathbb{N}$.

To show that the sequence is monotonically increasing, we also use the proof by induction. Since $y_0 = 0$, we have $y_1 = b/2$. So $y_1 - y_0 = b/2 \geq 0$. Now assume $y_k - y_{k+1} \geq 0$. We want to show $y_{k+1} - y_k \geq 0$. Observe that

$$y_{k+1} - y_k = \frac{1}{2}(b + y_k^2) - \frac{1}{2}(b + y_{k-1}^2) = \frac{1}{2}(y_k - y_{k-1})(y_k + y_{k+1}) \geq 0.$$

So by induction $y_{n+1} - y_n \geq 0$ for all $n \in \mathbb{N}$.

Since $\{y_n\}$ is increasing and bounded, then it converges. So taking the limit from both sides we will have

$$y = \frac{1}{2}(b + y^2).$$

By completing the square, we will have $(y - 1) = \sqrt{1 - b} = \sqrt{a}$. So $x = y - 1 = \sqrt{a}$.

(b) If $a > 1$ then we can use the following identity.

$$\sqrt{a} = \frac{1}{\sqrt{\frac{1}{a}}}$$

In other words, we calculate the square root for $a' = 1/a \leq 1$ and then we have $\sqrt{a} = 1/\sqrt{a'}$.

■ **Problem 1.2.4.7** Let $\{x_n\}$ be a sequence in a metric space (X, d) with the property that for some $\epsilon > 0$ one has $d(x_n, x_m) \geq \epsilon$ for all n, m . Show that $\{x_n\}$ is not convergent.

Solution Assume otherwise. So $\exists x_0 \in X$ such that for all $\epsilon > 0$, exists $N \in \mathbb{N}$ such that $\forall n > N$ we have $d(x_n, x_0) < \epsilon$. This contradicts the assumption.

■ **Problem 1.2.4.8** Let $X = \mathbb{R}^n$ be given with the metric $d(x, y) = \sum_{i=1}^n |x_i - y_i|$. Show that a sequence $\{x_n\}$ in X converges to x_0 if and only if $x_{n,i} \rightarrow x_{0,i}$ for all $i = 1, \dots, n$.

Solution \Rightarrow Assume $x_n \rightarrow x_0$. Then $d(x_n, x_0) \rightarrow 0$. So

$$\sum_{i=1}^N |x_{n,i} - x_{0,i}| \rightarrow 0.$$

So each term in the sum should go to zero, i.e. $|x_{n,i} - x_{0,i}| \rightarrow 0$ as $n \rightarrow \infty$. So $x_{n,i} \rightarrow x_{0,i}$ as $n \rightarrow \infty$.

\Leftarrow Assume $x_{n,i} \rightarrow x_{0,i}$ for all $i = 1, \dots, n$. So by definition $|x_{n,i} - x_{0,i}| \rightarrow 0$ as $n \rightarrow \infty$. So $\sum_{i=1}^N |x_{n,i} - x_{0,i}| \rightarrow 0$ as $n \rightarrow \infty$. So $x \rightarrow x_0$ in (\mathbb{R}^N, d) .

■ **Problem 1.2.4.9** Let $x_n(t)$ be a sequence of continuous real-valued functions where $x_n(t)$ is periodic with period $\tau_n > 0$. Assume that $x_n(t) \rightarrow x(t)$ uniformly for $t \in \mathbb{R}$ and $\tau_n \rightarrow \tau$. Show that $x(t)$ is periodic in t with period τ .

Solution First, observe that $x_n(t + \tau_n) \rightarrow x(t + \tau)$ as $n \rightarrow \infty$. Indeed, since $t + \tau_n \rightarrow t + \tau$ as $n \rightarrow \infty$, for any $k \in \mathbb{N}$ we have $x_k(t + \tau_n) \rightarrow x_k(t + \tau)$ that follows from continuity of x_n . Because of the uniform convergence of x_n , we can write $x_n(t + \tau_n) \rightarrow x(t + \tau)$ as $n \rightarrow \infty$. On the other hand, we want to show $x_n(t + \tau_n) \rightarrow x(t)$. To see this

$$|x_n(t + \tau_n) - x(t)| = |x_n(t) - x(t)| \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

So $x_n(t + \tau_n) \rightarrow x(t)$ and $x_n(t + \tau_n) \rightarrow x(t + \tau)$ both uniformly. Then it follows that $x(t) = x(t + \tau)$.

■ **Problem 1.2.4.10** Let f and g be functions in $C[0, T]$. Define $x_0 = f$ and $x_1 = g$. Let

$$x_{n+1} = \frac{1}{2}(x_n + x_{n-1}), \quad n = 1, 2, \dots.$$

Show that $\lim x_n = \frac{1}{3}(f + 2g)$ when $C[0, T]$ has the sup-metric.

Solution Let $t \in [0, T]$. Define $a_0 = f(t)$ and $a_1 = g(t)$, and define $a_n = x_n(t)$. Then $\{a_n\}$ is a sequence of real numbers where

$$a_{n+1} = \frac{1}{2}(a_n + a_{n-1}).$$

We can write this recursive equation as

$$\begin{bmatrix} a_n \\ a_{n+1} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1/2 & 1/2 \end{bmatrix} \begin{bmatrix} a_{n-1} \\ a_n \end{bmatrix}.$$

So we will have

$$\begin{bmatrix} a_n \\ a_{n+1} \end{bmatrix} = A^n \begin{bmatrix} a_0 \\ a_1 \end{bmatrix}.$$

Writing A in its Jordan block form we will have

$$A = P \begin{bmatrix} -1 & 0 \\ 0 & 2 \end{bmatrix} P^{-1},$$

where

$$P = \begin{bmatrix} -2 & 1 \\ 1 & 1 \end{bmatrix}, \quad P^{-1} = \frac{1}{3} \begin{bmatrix} -1 & 1 \\ 1 & 2 \end{bmatrix}.$$

So

$$\lim A^n = \frac{1}{3} \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix}.$$

Then it follows that $\lim a_n = 1/3(f + 2g)$. So the sequence $\{x_n\}$ converges to the function $h = \frac{1}{3}(f + 2g)$ point-wise. To show the uniform convergence, i.e. the converges with the sup norm, we use the fact that the point-wise convergence of the uniform /.

2. Algebraic Structure

Summary ↗ 2.1 Let $A : V \rightarrow W$ be a linear map between vector spaces. Then the pre-image of any linearly independent set contains a linearly independent set. Let $\{w_n\}$ be a linearly independent set of vectors in W . Let $v_n = A^{-1}(w_n)$, i.e. **one** of the pre-images of w_n (if A is not injective, then w_n can have multiple pre-images. But we choose just one pre-image. Any of them will work). Then let $\sum_n \beta_n v_n = 0$. Using the fact that $v_n = A^{-1}(w_n)$, one can write

$$0 = A\left(\sum_n \beta_n v_n\right) = \sum_n \beta_n A(A^{-1}(v_n)) = \sum_n \beta_n w_n,$$

which implies $\beta_n = 0$ for all n . So $\{v_n\}$ is linearly independent.

But note that the image of a linearly independent set is not necessarily a linearly independent set. Because a linear map can collapse some of the vectors to the origin, and the resulting collection will not be a linearly independent set of vectors.

Summary ↗ 2.2 Even in infinite dimensional vector spaces, an infinite sum of the form $\sum_{n=1}^{\infty} \alpha_n v_n$ for $\alpha_n \in F$ and $v_n \in V$ for all n , is meaningless, and one needs a topological structure on the space to make sense of such infinite sums.

Summary ↗ 2.3 — Causality and subspaces. Consider the space of functions defined on the real line (no regularity condition is necessary). Denote this space with X . Let $A : X \rightarrow X$ be a linear map between these two space, and let L_T denote the subspace that $x \in L_T$ iff $x(t) = 0$ for all $t \leq T$. L is a causal map iff L_T is invariant under A , that is $A(L_T) \subset L_T$.

Proof. Assume A is causal map and we want to show that L_T is invariant. Let $x \in L_T$. So $x(t) = 0$ for all $t \leq T$. The origin $x_0 \in L_T$ also satisfies $x_0(t) = 0$ for all $t \leq T$. From causality of A we need to have $[Ax](t) = [Ax_0](t)$ for all $t \leq T$. Since A is linear it sends the origin to the origin, i.e. $A(x_0) = 0$. This implies $[Ax](t) = 0$ for all $t \leq T$. So $Ax \in L_T$. So L_T is invariant under L .

Now assume L_T is invariant under A and we want to show that A is causal. Let $x, y \in X$ with $x(t) = y(t)$ for all $t \leq T$. Then $(x - y)(t) = 0$ for all $t \leq T$, this $x - y \in L_T$. So is $A(x - y) = Ax - Ay$. So $(Ax - Ay)(t) = 0$ for all $t \leq T$. So $[Ax](t) = [Ay](t)$ for all $t \leq T$. So A is causal. \square

Summary 2.4 Note that $\cup_{\alpha} B_{\alpha}$ can be a linear subspace and in the [Problem 2.1.1.5](#) we did not rule out the possibility of $\cup_{\alpha} B_{\alpha}$ to be a linear subspace. For instance, consider the space of all functions defined on the real line X , and let X_T denote the subspace of functions that vanish to the left of $T \in \mathbb{R}$, i.e. $x(t) = 0$ for all $t \leq T$. In this case $\cup_T X_T$ is the space of all functions that vanishes to the left of some finite time. This is a subspace of X .

Summary 2.5 It is well known that the set of all linear transformations from a linear space to another linear space is itself a linear space. However, there is even more! Let X be an arbitrary nonempty set and let Y be a linear space. Then \mathcal{F} , the set of all mappings of X into Y , is a linear space (with the addition of mappings and scalar multiplies of mappings are defined as in Example 2 Section 2).

Summary 2.6 A linear transformation is one-to-one (injective) if its kernel only contains the origin.

Proof. Let L be injective, and let $a \in \ker L$. Then $La = 0$, and from linearity of L we have $L0 = 0$. From injectivity of L we must have $a = 0$.

Let $\ker L = \{0\}$. Let $x, y \in V$ such that $f(x) = f(y)$. Then from linearity of f we can write $f(x - y) = 0$. Since the kernel only contains the origin we will have $x - y = 0$, so $x = y$. \square

Summary 2.7 The inverse of a linear transformation, if exists, is linear.

Proof. Let $L : V \rightarrow V$ be a linear transformation that its inverse exists denoted by L^{-1} . Let $y_1, y_2 \in V$. Then exists $x_1, x_2 \in V$ such that $y_1 = Lx_1$ and $y_2 = Lx_2$. Consider

$$L^{-1}(\alpha y_1 + y_2) = L^{-1}(\alpha Lx_1 + Lx_2) = L^{-1}(L(\alpha x_1 + x_2)) = \alpha x_1 + x_2 = \alpha L^{-1}(y_1) + L^{-1}(y_2).$$

So L^{-1} is linear when exists. \square

Summary 2.8 Let $X = l_2(-\infty, \infty)$ and Z be the linear space made up of all complex-valued functions $f(z)$ defined on the unit circle of the complex plane such that

$$\frac{1}{2\pi i} \int_C |f(z)|^2 dz/z < \infty.$$

The usual assumption is made that if $f_1(z)$ and $f_2(z)$ differ only on a set of measure zero, then f_1 and f_2 are considered to be the same function. These two spaces are isomorphic as linear spaces with

$$(\dots, \xi_{-1}, \xi_0, \xi_1, \dots) \mapsto f(z)$$

with

$$f(z) = \sum_{n=-\infty}^{\infty} \xi_n z^{-n}.$$

This is the two-sided z transform.

■ **Remark 2.1** In the example above, using appropriate change of variable, and utilizing the fact that functions in Z are defined on the unit circle we can write

$$\frac{1}{2\pi i} \int_C |f(z)|^2 dz/z = \frac{1}{2\pi i} \int_0^{2\pi} |f(e^{i\theta})|^2 d\theta.$$

In the second form, the connection to the Fourier transform is more clear.

■ **Remark 2.2** In the example above, the subspace $L \subset X$ where $x \in L$ if $x_i = 0$ for all $i < 0$ maps to the Hardy space H^2 via the constructed isomorphism. Hardy space is the space of all that are restriction of harmonic functions to the unit sphere.

Summary 2.9 — Some definitions. There are some important nuances in some definitions in linear algebra. We will highlight those definitions here.

Definition 2.1 — Linear Combination. Let V be a vector space, and $A \subset V$. Then $x \in V$ is said to be a linear combination of elements in A if there is a **finite** sum such that

$$x = \sum_{i=1}^n \alpha_i v_i, \quad v_i \in A \text{ for all } i.$$

Accordingly, $\text{Span}\{A\}$ is defined to be the set of all **finite** linear combinations of A .

■ **Remark 2.3** Regardless of the nature of the linear space (finite or infinite dimensional), a linear combination is always a finite sum.

Proposition 2.1 For any collection of vectors $A \subset V$, $\text{Span}\{A\}$ is **the smallest subspace** of V that **contains** A .

Proposition 2.2 — Linear Independence. Let $A \subset V$ be a collection of vectors. Then the followings are equivalent.

- (a) A is a linearly independent set of vectors.
- (b) For every $x \in A$ we have $x \notin \text{Span}\{A \setminus \{x\}\}$. That is x is not a linear combination of the points in $A \setminus \{x\}$.
- (c) For every **finite** sub-collection $\{v_1, \dots, v_n\} \subset A$ we have

$$\sum_{i=1}^n \alpha_i v_i = 0 \implies v_i = 0 \quad \forall i = 1, \dots, n.$$

- (d) For $v \neq 0 \in \text{Span}\{A\}$ there exists one and only one finite subset of A , say $\{v_i\}_{i=1}^n$ and a unique n -tuple of non-zero elements such that

$$v = \sum_{i=1}^n \alpha_i v_i.$$

- (e) A a linearly independent set of vectors if and only if there exists no proper subset $A_0 \subset A$ such that $\text{Span}\{A_0\} = \text{Span}\{A\}$. I.e. for all $B \in 2^A$ one has $\text{Span}\{B\} \subseteq \text{Span}\{A\}$.

■ **Remark 2.4** Note that in item (d) above we have excluded the origin, and also forced the n -tuple to have non-zero elements. Because $0 = 0v_1 + 0v_2 = 0v_3 + 0v_4$, etc assuming $v_1, v_2, v_3, v_4 \in A$. Furthermore, let $v \in A$, that trivially $v \in \text{Span}\{A\}$. Then one can write $v = 1v$, $v = 1v + 0v_1$, etc. I.e. one can add more vectors from A with zero coefficients.

Summary  **2.10** In Example 5, page 179, we discuss a very important example about the subspace $A_x \subseteq l_2(-\infty, \infty)$ and its connection to Z , the set of all complex valued functions defined on \mathbb{T} such that

$$\int_C |f(z)|^2 dz/z = \int_0^{2\pi} |f(e^{i\theta})|^2 d\theta < \infty.$$

Z and $l_2(-\infty, \infty)$ are isomorphic via ϕ given as:

$$(\dots, \xi_{-1}, \xi_0, \xi_1, \dots) \mapsto \sum_{n=-\infty}^{+\infty} \xi_n z^{-n}.$$

It turns out

$$\phi[S_r^n x](t) = t^{-n} \phi[x](t),$$

i.e. shifting the sequence x to the right by n -positions, is the same as multiplying $[\phi x](t)$ by the monomial t^{-n} . So $A_x = \{\sum_{n=M}^N \alpha_n S_r^n x\}$ is the same as $L \subset Z$ given by

$$L = \{g(t) \cdot [\phi x](t) : g(t) \text{ is a Laurent polynomial}\}.$$

A Laurent polynomial is a polynomial in variables z and $1/z$.

■ **Remark 2.5** The importance of the monomial like $f(z) = z^n$ becomes more evident when one restricts f to the unit-circle in the complex domain, i.e. $\mathbb{T} = \{|z| = 1\}$. Because

$$f|_{\mathbb{T}}(z) = f(e^{i\theta}), \quad \text{where } z = e^{i\theta}.$$

Thus restricting $f(z) = z$ to \mathbb{T} we will get

$$f(e^{i\theta}) = e^{in\theta} = \cos(n\theta) + i \sin(n\theta).$$

2.1 Problems

2.1.1 Linear Spaces and Linear Subspaces

■ **Problem 2.1.1.1** Let X be the linear space \mathbb{R}^4 . For what values of r , if any, is the set

$$A_r = \{x \in \mathbb{R}^4 : x_1 + x_2 + x_3 + x_4 = r\},$$

where r is a real number, a linear subspace of X ? For what values of r , if any, is the set

$$B_r = \{x \in \mathbb{R}^4 : x_1^2 + x_2^2 + x_3^2 + x_4^2 = r^2\}$$

a linear space of X ?

Solution Let $x, y \in A_r$. Then we need to have

$$\alpha x + y \in A_r \quad \forall \alpha \in F$$

Thus we need to have $r\alpha + r = r$ for all α in the underlying field. Thus we need to have

$$r = 0.$$

Let $x \in B_r$. Since we need to have $\alpha x \in B_r$ for all $\alpha \in F$, it follows that $\alpha r^2 = r^2$, thus $r = 0$. So B_0 , i.e. the origin, is the only case that B_r can be a linear subspace of \mathbb{R}^4 .

■ **Problem 2.1.1.2** Let X be the linear space made up of all complex-valued functions $T(s)$ defined on the imaginary axis of the complex plane such that

$$\left| \int_{-i\infty}^{i\infty} |T(s)| ds \right| < \infty,$$

where the integral is along the imaginary axis. That is, $X = L_2(-i\infty, i\infty)$. Let A be the set made up of all rational functions, that is all functions of the form

$$T(s) = \frac{a_0 s^m + \cdots + a_m}{b_0 s^n + \cdots + b_n},$$

where $a_0 \neq 0$ and $b_0 \neq 0$, and m, n are integers. Is the set A a linear subspace of X ? Next consider the subset B of A made up of all rational functions with $n > m$. Is B a linear subspace of X ? What about the subset C of B made up of all functions with all their finite poles in the left hand plane?

Solution The first part of the question is quite vague. Because $T(s) = 1/s$ belongs to A but not to X as

$$\int_{-\infty}^{\infty} \frac{1}{|x|^2} dx = \infty$$

So A is not a subset of X . However, if we want to consider $A \cap X$, then it is a linear subspace of X . That is because $T, L \in A$, then $\alpha T + \beta L$ still has the rational function form, and since $T, L \in X$, and X is a linear space, then $\alpha T + \beta L$ still satisfies the integral inequality, hence belongs to $A \cap X$.

STILL THINKING ON THE OTHER PARTS OF THE QUESTION.

■ **Problem 2.1.1.3** Show that the set of all $n \times m$ matrices can be viewed as a linear space.

Solution We exhibit an homomorphism between $M_{n \times m}$ and \mathbb{R}^{mn} . Let $A \in M_{n \times m}$ and $v \in \mathbb{R}^{mn}$. Let $\phi : M_{n \times m} \rightarrow \mathbb{R}^{mn}$ be defined as $\phi(A) = v$, where

$$v_{ni+j} = a_{ij}.$$

It is straightforward to show that ϕ is bijection. Using this function we can transfer all of the properties of \mathbb{R}^{mn} to $M_{m \times n}$. So $M_{m \times n}$ is a linear space, and ϕ is in fact an homomorphism of vector spaces. So $M_{m \times n}$ has dimension mn .

■ **Problem 2.1.1.4** Let X be the linear space $C[0, T]$. Which, if any, of the following subsets of X are linear subspaces?

- (a) $B_1 = \{x \in C[0, T] : x(0) = x(T)\},$
- (b) $B_2 = \{x \in C[0, T] : x(0) = x(T) = 0\},$
- (c) $B_3 = \{x \in C[0, T] : x(t_1) = x(t_2) \text{ for all } t_1, t_2 \text{ such that } t_1 + t_2 = T\},$
- (d) $B_4 = \{x \in C[0, T] : x(0) = 1\},$
- (e) $B_5 = \{x \in C[0, T] : \int_0^T x(\tau) d\tau = 1\},$

$$(f) \quad B_6 = \{x \in C[0, T] : |x(t_1) - x(t_2)| \leq 10|t_1 - t_2| \text{ for all } t_1, t_2 \in [0, T]\}.$$

Solution

Yes. $x, y \in B_1$, then $x(0) = x(T)$ and $y(0) = y(T)$. Then it follows that $(\alpha x + y)(0) = \alpha x(0) + y(0) = \alpha x(1) + y(1) = (\alpha x + y)(1)$. So $\alpha x + y \in B_1$.

Yes. Special case of above.

Yes. Let $x, y \in B_3$. Then $x(t) = x(T-t)$ and $y(t) = y(T-t)$. Then one can write

$$(\alpha x + y)(t) = \alpha x(t) + y(t) = \alpha x(T-t) + y(T-t) = (\alpha x + y)(T-t).$$

So $\alpha x + y \in B_3$.

No. $x \equiv 1$ is in B_4 , but $2x$ is not.

No. Let $x \in B_5$. Then $\int_0^T (2x)(\tau) d\tau = 2 \neq 1$, so $2x \notin B_5$.

No. $x(t) = 10t$ is in B_6 but $2x$ is not.

■ **Problem 2.1.1.5** Show that if $\{B_\alpha\}$ is a family of linear subspaces of a linear space X , then $B = \cap_\alpha B_\alpha$ is a linear subspace of X . What about $\cup_\alpha B_\alpha$?

Solution Let $x, y \in B$. Then $\forall \alpha$ we have $x, y \in B_\alpha$. Since B_α is a linear subspace, it follows that $ax + by \in B_\alpha$, which is true for all α . Thus $ax + by \in B$. The same is not true in the case of $\cup_\alpha B_\alpha$. For instance in the case of \mathbb{R}^2 , consider the subspaces $B_1 = \{(\alpha, 0) : \alpha \in \mathbb{R}\}$ and $B_2 = \{(0, \beta) : \beta \in \mathbb{R}\}$. Then $(1, 0) \in B_1$ and $(0, 1) \in B_2$ but $(1, 1) \notin B_1 \cup B_2$ (in fact $(1, 1) \in B_1 + B_2$, where $B_1 + B_2$ is the smallest subspace that contains B_1 and B_2).

■ **Remark 2.6** Note that $\cup_\alpha B_\alpha$ can be a linear subspace and in the problem above we did not rule out the possibility of $\cup_\alpha B_\alpha$ to be a linear subspace. For instance, consider the space of all functions defined on the real line X , and let X_T denote the subspace of functions that vanish to the left of $T \in \mathbb{R}$, i.e. $x(t) = 0$ for all $t \leq T$. In this case $\cup_T X_T$ is the space of all functions that vanishes to the left of some finite time. This is a subspace of X .

■ **Problem 2.1.1.6** Let X be the linear space made up of all real-valued sequences. Show that A_1 , the set of all sequences that have a finite number of nonzero entries only, is a linear subspace of X . Show that A_2 , the set of all sequences that have an infinite number of nonzero entries, is not a linear subspace of X .

Solution Let $a, b \in A_1$ with their number of non-zero elements as n_1, n_2 respectively. Then $\alpha a + \beta b$ has at most $n_1 + n_2$ number of non-zero elements, hence in A_1 . A_2 is not a linear subspace of X because it does not contain the origin.

■ **Problem 2.1.1.7** Often in systems theory the linear space $L_2(-\infty, \infty)$ is a good mathematical model for the set X of all inputs to a system as well as the set Y contains the range. Let \mathcal{A} be the set of all mappings (linear and nonlinear) of X into Y . Show that \mathcal{A} can be viewed as a linear space. Show that the subset $\mathcal{L} \subset \mathcal{A}$ of all mappings representing causal (Section 2.8) systems is a linear subspace of \mathcal{A} .

Solution Let $\phi, \psi \in \mathcal{A}$, and define the addition and scalar multiplication in this space as

$$(\alpha\phi + \beta\psi)(f) = \alpha\phi(f) + \beta\psi(f).$$

It is straightforward to check that with these definitions, \mathcal{A} is a vector space.

Let $F, G \in \mathcal{L}$ be causal maps. Then for all $T \in \mathbb{R}$, $x(t) = y(t)$ for $t < T$ implies $[Fx](t) = [Fy](t)$, and $[Gx](t) = [Gy](t)$ for all $t < T$. Then for $t \leq T$ one can write

$$[\alpha F + G](x)(t) = \alpha[Fx](t) + [Gx](t) = \alpha[Fy](t) + [Gy](t) = [\alpha F + G](y)(t).$$

Thus \mathcal{L} is a linear subspace.

■ **Remark 2.7 — Review of the causal maps.** $\Phi : L_2(-\infty, \infty) \rightarrow L_2(-\infty, \infty)$ is causal, if for all $T \in \mathbb{R}$, $x(t) = y(t)$ for all $t \leq T$, implies $(\Phi(x))(t) = (\Phi(y))(t)$ for all $t \leq T$.

■ **Problem 2.1.1.8** Let X be the linear space made up of all absolutely convergent sequences of real numbers. Show that B , the set of all absolutely convergent sequences of real numbers with limit zero, is a linear subspace of X .

Solution I can not understand what does absolutely convergent **sequence** means (in contrast to the absolutely convergent series). Also, not sure about the meaning of the abs. conv. sequences with limit zero.

■ **Problem 2.1.1.9** Let X be the set of all convergent sequences of real numbers. Is X a linear space?

Solution Yes. Viewing \mathbb{R} is a linear space, and using the continuity of the addition and scalar multiplication, if $x_n \rightarrow x$ and $y_n \rightarrow y$, then $\alpha x_n + \beta y_n \rightarrow \alpha x + \beta y$.

■ **Problem 2.1.1.10** Let X denote the collection of all real-valued Lipschitz-continuous functions $x(t)$ defined for $-\infty < t < \infty$. That is $x(t)$ satisfies $|x(t) - x(s)| \leq k|t - s|$ for some constant k which depends on x and for all t, s . Show that X is a real linear space.

Solution Let $x, y \in X$. Then we claim that $\alpha x + \beta y$ is also in X with Lipschitz constant $k \leq \alpha K_x + \beta K_y$. Because

$$|(\alpha x(t) + \beta y(t)) - (\alpha x(s) + \beta y(s))| \leq k_x |t - s| + k_y |t - s|.$$

2.1.2 Linear Transformation

■ **Problem 2.1.2.1** Let X and Y be linear spaces over the same scalar field. Show that $I : X \rightarrow X$ the identity transformation, and $0 : X \rightarrow Y$ the zero transformation are linear.

Solution Let $x, y \in X$. The $I(\alpha x + y) = \alpha x + y = \alpha I(x) + I(y)$, so I is linear. The zero transformation is trivially linear.

■ **Problem 2.1.2.2** Let X, Y , and Z be linear spaces over the same scalar field, and let $L_1 : X \rightarrow Y$ and $L_2 : Y \rightarrow Z$ be linear. Show that the composition $L_2 L_1 : X \rightarrow Z$ is linear.

Solution Let $x, y \in X$. Then

$$L_2(L_1(\alpha x + y)) = L_2(\alpha(L_1 x) + L_1 y) = \alpha L_2(L_1 x) + L_2(L_1 y) = \alpha(L_2 L_1)x + L_2 L_1 y.$$

So $L_2 L_1$ is linear.

■ **Problem 2.1.2.3** Suppose that we consider a system whose output is a delayed version of the input. That is, if $x(t)$ is the input, then the output $y(t) = x(t - \tau)$, where τ is a constant. Let X be the linear space $C(-\infty, \infty)$ of continuous real-valued functions defined on $(-\infty, \infty)$. Let D denote the system operation. Is D a linear transformation of X into itself? suppose that instead of being constant the delay τ is given by $\tau = e^{-t}$. Do we have a linear transformation? Then suppose that $\tau = \exp\left[-\int_{-\infty}^t |x(\xi)|d\xi\right]$, where of course the linear space X must be selected so that the integral exists. Do we have a linear transformation?

Solution In all of the cases D is a linear operator. Assume the delay is given as a function $\tau(t)$. Let $x, y \in C(-\infty, \infty)$. Then one can write

$$[D(\alpha x + y)](t) = (\alpha x + y)(t - \tau(t)) = \alpha x(t - \tau(t)) + y(t - \tau(t)) = \alpha [Dx](t) + [Dy](t).$$

■ **Problem 2.1.2.4** Let $Y = C([0, \infty], \mathbb{R}^n)$ be the linear space made up of all continuous mappings of $[0, \infty)$ into \mathbb{R}^n , that is, each component is continuous. Let $X = C^1([0, \infty], \mathbb{R}^n)$ be the linear subspace of Y made up of all elements of Y with continuous derivatives, that is, each component has a continuous derivative. Does the expression $y = Tx$, where

$$Tx = \frac{dx}{dt} - Ax$$

and A is a real $n \times n$ matrix, represent of linear transformation of X into Y ?

Solution Yes. Let $x, y \in X$. Then

$$T(\alpha x + y) = \alpha \frac{dx}{dt} + \frac{dy}{dt} - \alpha Ax - Ay = \alpha Tx + Ty,$$

where we have used the linearity of the differentiation operator and the linearity of multiplication by a matrix.

■ **Problem 2.1.2.5** Let $Y = BC(-\infty, \infty)$ denote the space of all bounded real-valued continuous functions $y(t)$ defined for $-\infty < t < \infty$ and let X denote the space of all Lipschitz continuous functions. Define $x = Ly$ be $x(t) = \int_0^t y(s)ds$. Show that L is a linear mapping of Y into X .

Solution Follows from the properties of integration: for $y_1, y_2 \in Y$

$$\int_0^t (\alpha y_1 + y_2)ds = \alpha \int_0^t y_1(s)ds + \int_0^t y_2(s)ds.$$

■ **Remark 2.8** Note that in the example above, integration improved the regularity of the functions. The input of the operator was $BC(-\infty, \infty)$ and its output is Lipschitz continuous functions.

2.1.3 Inverse Transformation

■ **Problem 2.1.3.1** Show that the linear transformation $y = Lx$ on $L_2(-\infty, \infty)$ given by

$$y(t) = \int_{-\infty}^t a^{-1} e^{a(t-\tau)} x(\tau) d\tau$$

is one-to-one. Hint: Show that $Lx = 0$ reduces to $\int_{-\infty}^t e^{a\tau} x(\tau) d\tau = 0$. Then differential and use theorem D.13.3.

Solution The transformation L is linear. So L is injective iff $\ker L = 0$. Let $x \in \ker L$. Then one can write

$$\int_{-\infty}^t a^{-1} e^{-a(t-\tau)} x(\tau) d\tau = a^{-1} e^{-at} \int_{-\infty}^t e^{a\tau} x(\tau) d\tau = 0 \quad \forall t.$$

This implies

$$\int_{-\infty}^t e^{at} x(\tau) d\tau = 0 \quad \forall t.$$

Differentiating with respect to t one gets

$$e^{at} x(\tau) \equiv 0.$$

This implies $x(\tau) \equiv 0$.

■ **Problem 2.1.3.2** Let $k(t, s)$ be continuous for $0 \leq s \leq t \leq T$ and consider

$$y(t) = x(t) + \int_0^t k(t, s)y(s)ds.$$

The following steps will lead to a proof that the relationship $y = Fx$ implicitly given above, does define a linear mapping F on $C[0, \alpha]$ provided α is a sufficiently small positive number.

(a) Define G by $x = Gy$, where

$$x(t) = y(t) - \int_0^t k(t, s)y(s)ds.$$

Show that G is linear.

- (b) Assume that for some $\alpha > 0$, G maps $C[0, \alpha]$ onto itself. Show that if G is one-to-one, then G^{-1} exists and is F , so F is linear.
- (c) Let M satisfy $|k(t, s)| \leq M$ for $0 \leq s \leq t \leq T$, then show that $Gy = 0$ reduces to

$$|y(t)| \leq M \int_0^t |y(s)| ds.$$

Now use the Gronwall inequality, see Cesari [1,p.35], to show that G is one-to-one.

Solution (a) Both terms in the RHS of

$$x(t) = y(t) - \int_0^t k(t, s)y(s)ds$$

with respect to the position of y (you know what I mean!!). So the linearity of G follows immediately.

- (b) By assumption G is onto and one-to-one. So G^{-1} exists. It remains to show that G^{-1} is F . So we need to show $GF = FG = I$. STILL THINKING ON THIS.
- (c) STILL THINKING ON THIS.

2.1.4 Isomorphisms

■ **Problem 2.1.4.1** Let χ denote the set of all linear spaces over a scalar field F . Show that the relation $X \sim Y \Leftrightarrow (X, Y \text{ are isomorphic})$ is an equivalent relation and, therefore, induces a partition on χ .

Solution Every linear space is isomorphic with itself with the identity operator. If U is isomorphic to Y via $\phi : U \rightarrow V$, then V is also isomorphic to U via $\phi^{-1} : V \rightarrow U$. Lastly, let $\phi : U \rightarrow V$ and $\psi : V \rightarrow W$ be isomorphism between spaces. Then U and W are also isomorphic via the isomorphism $\phi \circ \psi$.

■ **Problem 2.1.4.2** Referring to the Example 2 in Page 174 of the text book, under what conditions on real numbers $c_{11}, c_{12}, c_{21}, c_{22}$ does

$$T_2(x) = (c_{11}x_1 + c_{12}x_2) + (c_{21}x_1 + c_{22}x_2)t,$$

where $x = (x_1, x_2)$, define an isomorphism from \mathbb{R}^2 to Z ?

Solution Let $v = (v_1, v_2)$ denote the function $v_1 + v_2t \in Z$. Then we want the following matrix to be invertible

$$\begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}.$$

So the matrix should have a non-zero determinant. I.e.

$$c_{11}c_{22} - c_{12}c_{21} \neq 0.$$

■ **Problem 2.1.4.3** Show that the real linear space X made up of all functions of the form $x = a \cos(\omega t + \phi)$, where ω is fixed, is isomorphic to the complex numbers considered as a real linear space. This fact is the cornerstone of the so-called phasor method of analyzing alternating current electrical networks.

Solution Given $a \cos(\omega t + \phi)$ observe that

$$a \cos(\omega t + \phi) = \operatorname{Re}(a \cos(\omega t + \phi) + ia \sin(\omega t + \phi)) = \operatorname{Re}(ae^{i\phi} e^{i\omega t}).$$

So let the map between these two spaces be

$$a \cos(\omega t + \phi) \mapsto z = ae^{i\phi}.$$

Showing the linearity of the mapping above is very dense with different trigonometric identities. So we exhibit the mapping that is the other way around. Consider the following map

$$z \mapsto \operatorname{Re}(ze^{i\omega t}).$$

This map is linear and onto. Linearity follows from linearity of the $\operatorname{Re}(\cdot)$ map and it is onto because for any $a \cos(\omega t + \phi)$ we have a preimage given by $z = ae^{i\phi}$. It is also bijective because it is linear and its kernel contains only the origin. So The mapping above is isomorphism between vector spaces.

■ **Problem 2.1.4.4** Let $L_2^\sigma(-\infty, \infty)$ denote the linear space made up do all complex-valued functions x defined on $(-\infty, \infty)$ such that

$$\int_{-\infty}^{\infty} |x(t)|^2 e^{-2\sigma t} dt < \infty,$$

where σ is a real number. Show that $L_2^\sigma(-\infty, \infty)$ is isomorphic to $L_2^\tau(-\infty, \infty)$ for any σ and τ .

Solution Let $x \in L_2^\sigma$. Consider the mapping ϕ

$$x(t) \mapsto x(t)e^{(\sigma-\tau)t}.$$

This map is linear because

$$\phi(\alpha x + y) = (\alpha x(t) + y(t))e^{(\sigma-\tau)t} = \alpha\phi(x)(t) + \phi(y)(t).$$

The inverse map is ϕ^{-1} given by

$$\phi^{-1}(y)(t) = y(t)e^{-(\sigma-\tau)t}.$$

It is straightforward to check that $\phi\phi^{-1}$, and $\phi^{-1}\phi$ are the identities maps of the corresponding spaces. It only remains to show that for $x \in L_2^\sigma$, ϕx really belongs to L_2^τ . This is true because

$$\int_{-\infty}^{\infty} |[\phi x](t)|^2 e^{-2\tau t} dt = \int_{-\infty}^{\infty} |x(t)e^{(\sigma-\tau)t}|^2 e^{-2\tau t} dt = \int_{-\infty}^{\infty} |x(t)|^2 e^{-2\sigma t} dt < \infty.$$

■ **Problem 2.1.4.5** Show that $L_2(-\infty, \infty)$ is isomorphic to $L_2[0, \infty)$.

Solution Consider the following linear functions

$$\phi : L_2(-\infty, +\infty) \rightarrow L_2[0, +\infty), \quad [\phi x](t) = x(\log(t))/\sqrt{t},$$

and

$$\psi : L_2[0, +\infty) \rightarrow L_2(-\infty, +\infty), \quad [\psi y](t) = y(e^t)e^{t/2}.$$

It is easy to show that $\psi\phi$ and $\phi\psi$ are the identity maps of the corresponding spaces, so are inverses of each other. Now we need to show that for $x \in L_2(-\infty, +\infty)$, ϕx belongs to $L_2[0, +\infty)$. Consider the integral

$$\int_0^{\infty} |[\phi x](t)|^2 dt = \int_0^{\infty} |x(\log(t))/\sqrt{t}|^2 dt < \infty.$$

We do the change of variable $\tau = \log(|t|)$.

$$\int_0^{\infty} |[\phi x](t)|^2 dt = \int_{-\infty}^{\infty} |x(\tau)|^2 d\tau = \int_{-\infty}^{\infty} |x(\tau)|^2 d\tau < \infty.$$

A similar approach can be used to show that for $y \in L_2[0, \infty)$ we have $\phi y \in L_2(-\infty, \infty)$.

■ **Problem 2.1.4.6** Show that $l_2(-\infty, \infty)$ is isomorphic with $l_2(0, \infty)$.

Solution Map every negative indexed element of the sequence in $l_2(-\infty, \infty)$ to an odd index, and every positive indexed element of the sequence to an even index. Since the sequence is absolutely summable, so is the new sequence, hence in $l_2(0, \infty)$.

2.1.5 Linear Independence and Dependence

■ **Problem 2.1.5.1** Let T be an isomorphic mapping of a linear space X onto a linear space Y . Show that $A \subset X$ is linearly independent if and only if its image $T(A) \subset Y$ is linearly independent.

Solution Assume $\{T(A)\}$ is linearly independent. We want to prove that A is linearly independent. To see this consider a sub-collection $\{x_i\} \subset A$. Let the following be any linear combination of these vectors that is equal to zero

$$\sum_{i=1}^n \alpha_i x_i = 0$$

One can apply T and using its linearity

$$0 = T\left(\sum_{i=1}^n \alpha_i x_i\right) = \sum_{i=1}^n \alpha_i T(x_i).$$

Since $\{T(x_i)\}$ is linearly independent, it follows that $\alpha_i = 0$ for all i . So $\{x_i\}$.

For the converse let A be a linearly independent set of vectors. We want to show $\{T(A)\}$ is also linearly independent. Consider the following linear combination of $\{T(x_i)\}_{i=1}^n$ where

$$\sum_{i=1}^n \alpha_i T(x_i) = 0.$$

Applying T^{-1} to both sides and using its linearity we have

$$\sum_{i=1}^n \alpha_i x_i = 0.$$

Since $\{x_i\}$ is linearly independent, it follows that $\alpha_i = 0$ for all i . Since this is true for any sub-collection of vector in $\{T(A)\}$ it follows that $\{T(A)\}$ is linearly independent.

■ **Remark 2.9** Note that one $\{T(A)\}$ being linearly independent (L.I.) implies $\{A\}$ being L.I., however, the converse is not true. Because a map might not be injective and can send some of the vectors to the origin, thus the image will not be linearly independent set of vectors.

■ **Problem 2.1.5.2** Let $X = L_2(\Omega, \mathcal{F}, \mathbb{P})$, the linear space made up of all complex-valued random variables x defined on a probability space $(\Omega, \mathcal{F}, \mathbb{P})$ such that

$$\mathbb{E}[x\bar{x}] = \int_{\Omega} = \int_{\Omega} |x|^2 d\mathbb{P},$$

where \bar{x} denotes the complex conjugate of x . Let $A \subset X$ be the set containing two random variables x_1 and x_2 such that

$$\mathbb{E}[x_1\bar{x}_1] = 1, \quad \mathbb{E}[x_2\bar{x}_2] = 1, \quad \mathbb{E}[x_1\bar{x}_2] = 0.$$

Show that $y \in X$ is a linear combination of points in A if and only if

$$\mathbb{E}[y\bar{y}] = |\mathbb{E}[y\bar{x}]|^2 + |\mathbb{E}[y\bar{x}_2]|^2.$$

Moreover, show that A is linearly independent.

Solution We are dealing with a linear space that we have a notion of inner product, hence projection, etc. So I will use that notation here. For two random variables x, y define

$$\langle x, y \rangle = \mathbb{E}[xy].$$

So with this language the properties in the question statement can be written as

$$\langle x_1, x_1 \rangle = 1, \quad \langle x_2, x_2 \rangle = 1, \quad \langle x_1, x_2 \rangle = 0.$$

To show the forward direction let $y = \alpha x_1 + \beta x_2$. Then

$$\mathbb{E}[y\bar{y}] = \langle y, y \rangle = |\alpha|^2 + |\beta|^2 = \left| \mathbb{E}[y\bar{x}_1] \right|^2 + \left| \mathbb{E}[y\bar{x}_2] \right|^2.$$

For the converse, assume

$$\mathbb{E}[y\bar{y}] = \left| \mathbb{E}[y\bar{x}_1] \right|^2 + \left| \mathbb{E}[y\bar{x}_2] \right|^2$$

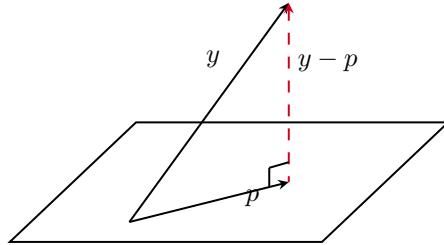
holds. We want to show $y = \alpha x_1 + \beta x_2$ for some α, β . Let $p = \alpha x_1 + \beta x_2$. Then

$$\langle y - p, x_1 \rangle = \langle y, x_1 \rangle - \langle p, x_1 \rangle = |\alpha|^2 - |\alpha|^2 = 0.$$

And for a similar reason one can write

$$\langle y - p, x_2 \rangle = 0.$$

So $y - p \perp \text{Span}\{x_1, x_2\}$. See the following figure



So we can write

$$\|y - p\|^2 = \|y\|^2 + \|p\|^2.$$

Using the fact that $\|y\|^2 = |\alpha|^2 + |\beta|^2$ and $\|p\|^2 = |\alpha|^2 + |\beta|^2$ one gets

$$\|y - p\| = 0.$$

Using the properties of the norm we have

$$y = p.$$

To show that A is linearly independent, consider the following linear combination

$$\alpha x_1 + \beta x_2 = 0.$$

By applying $\langle x_1, \cdot \rangle$ and $\langle x_2, \cdot \rangle$ to both sides one gets

$$\alpha = 0, \quad \beta = 0.$$

So A is a linearly independent set of vectors.

■ **Problem 2.1.5.3** Let X be the linear space $L_2[0, 2\pi]$ and A the set of all functions $x_n(t) = e^{int}$, $n = 0, 1, 2, \dots$. Show that A is linearly independent.

Solution In order not to deal with messy sums, we demonstrate the linear independence of the subset $\{1, e^{it}, e^{2it}\}$. The linear independence for arbitrary finite subset of A can be shown in a similar way. Consider

$$\alpha_0 + \alpha_1 e^{it} + \alpha_2 e^{2it} = 0.$$

It follows that $\alpha_0 = 0$. Furthermore, differentiating both sides and dividing by ie^{it} one gets

$$\alpha_1 + 2\alpha_2 e^{2it} = 0.$$

This time we get $\alpha_1 = 0$. Differentiating again and dividing by e^{2it} one gets $\alpha_2 = 0$. So $\{1, e^{it}, e^{2it}\}$ is linearly independent.

■ **Problem 2.1.5.4** Show that a finite set $A = \{x_1, \dots, x_n\}$ is a linear space X is linearly independent if and only if the only n -tuple of scalars satisfying the equation

$$a_1 x_1 + \dots + a_n x_n = 0$$

is $a_1 = \dots = a_n = 0$.

Solution Assume A is linearly independent. We want to show $a_1 x_1 + \dots + a_n x_n = 0$ implies $a_1 = \dots = a_n = 0$. We show this by contrapositive. Let $a_1 x_1 + \dots + a_n x_n = 0$ with not all of a_i equal to zero. Then one can write

$$x_i = \beta_1 x_1 + \dots + \beta_n x_n$$

, where x_i is the vector that its corresponding coefficient α_i was not zero. This implies that the set A is not linearly independent.

For the converse we want to show $a_1 x_1 + \dots + a_n x_n = 0$ implying $a_i = 0$ for all i implies A is linearly independent. We do it by proof by contrapositive. Assume A is not linearly independent. WLOG we can assume x_1 is linear combination of others:

$$x_1 = \sum_{i=2}^n \alpha_i x_i.$$

Rearranging this sum one gets

$$x_1 - \alpha_2 x_2 - \dots - \alpha_n x_n = 0$$

with not all of coefficients equal to zero. This finishes the proof.

■ **Problem 2.1.5.5** Let X be the linear space \mathbb{R}^n , and let A be the set containing the n vectors

$$x_i = (x_{1i}, x_{2i}, \dots, x_{ni}) \quad \text{for } i = 1, \dots, n.$$

Show that this set is linearly independent if and only if

$$\det \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1n} \\ x_{21} & x_{22} & \cdots & x_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ x_{n1} & x_{n2} & \cdots & x_{nn} \end{bmatrix} \neq 0.$$

Solution Assume the determinant is non zero and we want to show that A is linearly independent. We do this by contrapositive. Assume A is not linearly independent. So we can write one column of the matrix as the linear combination of the other column. Then by elementary column operations (that keeps the determinant intact) one can reduce that column to zero column and hence the determinant will be zero.

Now assume A is linearly independent and we want to show that the determinant is non-zero. Let $c = (c_1, \dots, c_n)$ be a vector of coefficients. Then columns of A being linearly independent is the same as saying the following system of equations has the only zero answer

$$Ac = 0 \quad \text{only when } c = (0, \dots, 0).$$

In other words, $\ker A = \{0\}$. This implies A is invertible (because it is automatically onto), and thus $\det(A) \neq 0$.

■ **Problem 2.1.5.6** Let the state of a dynamical system be a point in the linear space $X = \mathbb{R}^n$. Let the state at time $k = 0, 1, 2, \dots$ be denoted by x_k . Further, suppose that the evolution of the system is characterized by a linear transformation $T : X \rightarrow X$. In particular, $x_k = Tx_{k-1}$. Let x_0 be an arbitrary initial state, and consider the set

$$A = \{x_0, Tx_0, T^2x_0, \dots\}.$$

Show that there exists an integer p such that for

$$A_p = \{x_0, Tx_0, \dots, T^px_0\}$$

one has $V(A) = V(A_p)$, that is, A and A_p span exactly the same linear spaces.

Solution Let $A = \{x_1, x_2, \dots, x_N\}$ be the first linearly independent vectors that appear in A . Note that $N \leq n$, as the space X is n dimensional. Let $p = N$. Then $\text{Span}\{A\} = \text{Span}\{A_p\}$.

■ **Problem 2.1.5.7** Let X be the linear space made up of all real-valued random variables defined on some probability space. Let A be a set of random variables in X . Show that if a random variable $z \in X$ is stochastically independent of each random variable in A , then z is ot in $V(A)$.

Solution STILL THINKING ON THIS.

■ **Problem 2.1.5.8** Let X be a linear space. A set $K \subset X$ is said to be convex if

$$\lambda x + (1 - \lambda)y \in K \quad (0 \leq \lambda \leq 1),$$

whenever $x, y \in K$. Let K_1 and K_2 be two convex sets in X .

- (a) Show that $K_1 \cap K_2$ is convex.
- (b) Is $K_1 \cup K_2$ convex?

Solution (a) Let $x, y \in K_1 \cap K_2$. Then since K_1 and K_2 are convex we have

$$\lambda x + (1 - \lambda)y \in K_1 \quad \lambda x + (1 - \lambda)y \in K_2 \quad (0 \leq \lambda \leq 1).$$

Thus $\lambda x + (1 - \lambda)y \in K_1 \cap K_2$ for $\lambda \in [0, 1]$. Thus $K_1 \cap K_2$ is convex.

- (b) No. Let $x_1, x_2 \in X$ be two points in the vector space. Each of the sets $\{x_1\}$ and $\{x_2\}$ are convex, but $\{x_1, x_2\}$ is not.
-

■ **Problem 2.1.5.9** Let $y = f(x)$ be a C^2 function defined for $-\infty < x < \infty$. Find a condition on d^2f/dx^2 in order that

$$K = \{(x, y) \in \mathbb{R}^2 : y \geq f(x)\}$$

be convex.

Solution We use the fact that $f'' > 0$ implies f is convex, i.e.

$$f(\lambda x_1 + (1 - \lambda)x_2) \leq \lambda f(x_1) + (1 - \lambda)f(x_2), \quad \lambda \in [0, 1]$$

Choose $(x_1, y_1), (x_2, y_2) \in K$. So by definition $y_1 \geq f(x_1)$ and $y_2 \geq f(x_2)$. Thus

$$f(\lambda x_1 + (1 - \lambda)x_2) \leq \lambda y_1 + (1 - \lambda)y_2, \quad \lambda \in [0, 1].$$

So $\lambda y_1 + (1 - \lambda)y_2 \in K$ for all λ . Thus K (the epigraph of f) is convex if and only if $f'' > 0$.

3. Combined Topological and Algebraic Structures

Summary 3.1 — **A useful trick.** Let $x, y \in \mathbb{R}$. Then we want to show for $p \in (0, 1)$ we have

$$|x + y|^p \leq |x|^p + |y|^p.$$

To see this let $\beta = 1 - p$. Then we can write

$$|x + y|^p = |x + y|^{1-\beta} = \frac{|x + y|}{|x + y|^\beta} \leq \frac{|x|}{|x + y|^\beta} + \frac{|y|}{|x + y|^\beta} \leq \frac{|x|}{|x|^\beta} + \frac{|y|}{|y|^\beta} = |x|^p + |y|^p.$$

Summary 3.2

$$\|x\| = \int_0^1 |x(t)|^p dt < \infty$$

is not a norm but $d(x, y) = \|x - y\|$ is a metric.

Summary 3.3 — **Not norm but metric!** There are functions $\|\cdot\| : V \rightarrow \mathbb{R}$ defined on a vector space, that are not norm, but $d(x, y) = \|x - y\|$ is a metric. It is enough for a function $\|\cdot\|$ to be non-negative, and positive definite, and satisfy the triangle inequality and have $\|x\| = \|-x\|$. Then $d(x, y) = \|x - y\|$ will be a metric. There is no need for homogeneity condition.

Summary 3.4 Any metric d on a vector space V comes from a function $\|x\|$ that need to satisfy all norm properties except for the homogeneity condition. And it also need to satisfy $\|x\| = \|-x\|$. See [Problem 3.1.1.7](#).

3.1 Problems

3.1.1 Normed Linear Spaces

- **Problem 3.1.1.1** Show that the norm $\|\cdot\|$ considered as a mapping of a normed linear space X into the reals is continuous.

Solution We use the reverse triangle inequality, i.e. for $x, y \in X$ we have

$$|\|x\| - \|y\|| \leq \|x - y\|.$$

Let $\epsilon > 0$ given and let $d(x, y) = \|x - y\| < \epsilon$. Then it implies that

$$d(\|x\|, \|y\|) = |\|x\| - \|y\|| \leq \|x - y\| < \epsilon.$$

So $\|\cdot\|$ is continuous.

■ **Remark 3.1.1** In the problem above we implicitly assume that the real line (co-domain of the norm function) has the topology induced by $|\cdot|$.

■ **Problem 3.1.1.2** Show that the addition and scalar multiplications in a vector space are continuous.

Solution Let $\cdot + \cdot : X \times X \rightarrow X$ denote the addition operation in a vector space. We assume that X has the topology induced by $\|\cdot\|$ and $X \times X$ has the topology induced by $\|(x, y)\| = \|x\| + \|y\|$. Let $(a, u), (b, v) \in X \times X$. Let $\epsilon > 0$ such that $d((a, u), (b, v)) = \|(a - b, v - u)\| = \|a - b\| + \|v - u\| < \epsilon$. Then we can write

$$d(a + u, b + v) = \|a + u - b - v\| \leq \|a - b\| + \|u - v\| = d((a, u), (b, v)) < \epsilon.$$

For the scalar multiplication, let $\cdot : F \times X \rightarrow X$ denote the scalar multiplication. We assume F is \mathbb{R} with the usual topology. Then

$$\|\alpha u - \beta v\| = \|\alpha u \pm \alpha v - \beta v\| \leq \|\alpha(u - v)\| + \|v(\alpha - \beta)\| = |\alpha| \|u - v\| + |\alpha - \beta| \|v\|.$$

Let $\epsilon > 0$ given and choose $\|u - v\| < \epsilon/\alpha$ and $|\alpha - \beta| < \epsilon/\|v\|$. So when $d((\alpha, u), (\beta, v)) = \|u - v\| + |\beta - \alpha| < \delta$ where $\delta = \epsilon/\alpha + \epsilon/\|v\|$ we will have

$$\|\alpha u - \beta v\| < \epsilon.$$

So the scalar multiplication is continuous. Note that the choice of δ depends on v . So the scalar multiplication is **not** uniformly continuous.

■ **Problem 3.1.1.3** Characterize all possible norms on the real line \mathbb{R} , where \mathbb{R} is considered a real linear space. On the complex plane \mathbb{C} , where \mathbb{C} is considered as a complex linear space. Show that \mathbb{C} may have other norms when it is considered as a real linear space.

Solution We know that $|\cdot| : \mathbb{R} \rightarrow \mathbb{R}$ is a norm on \mathbb{R} . Let $f(x)$ be any other norm on this space. Then define $g(x) = f(x)/|x|$ for $x \neq 0$. For any $\alpha \in \mathbb{R}$ we will have

$$g(\alpha x) = g(x) \quad \forall \alpha, x \in \mathbb{R}.$$

This implies that g is a constant function. Because, let $x, y \in \mathbb{R} \setminus \{0\}$ then $g(x) - g(y) = g(x) - g(\beta x) = g(x) - g(x) = 0$ where $\beta = y/x$. So $f(x) = \alpha|x|$ where $g(x) \equiv \alpha$. So all of the norms on \mathbb{R} are of the form $\|x\|$.

On the complex plane \mathbb{C} as a vector space over \mathbb{C} we define the norm as

$$\|u\| = \sqrt{u\bar{u}}.$$

Similar to above, define $g(u) = f(u)/\|u\|$ where $f(u)$ is any other norm. Then we will have $g(\alpha u) = g(u)$ for all $u, \alpha \in \mathbb{C}$. Let $u, v \in \mathbb{C}$. Then $\exists \alpha \in \mathbb{C}$ such that $u = \alpha v$ (this is only true if \mathbb{C}

is a vector space over \mathbb{C}). So $g(u) - g(v) = g(u) - g(\alpha u) = 0$. So g is a constant function. So we conclude that every norm in (\mathbb{C}, \mathbb{C}) is in the form of $|\alpha| \|\cdot\|$ for $\alpha \in \mathbb{C}$.

However, if we consider \mathbb{C} as a vector space over \mathbb{R} , then there are other norms different from the form above. For instance $\|u\| = \sqrt{\operatorname{Im}(u)^2 + \operatorname{Re}(u)^2}$ is a norm.

■ **Problem 3.1.1.4** Let $(X, \|\cdot\|)$ be a normed linear space and let $S_r = \{x : \|x\| = r\}$ where $r > 0$. Assume that $X \neq \{0\}$. Show that $(X, \|\cdot\|)$ is a Banach space if and only if the metric space $(S_r, \|\cdot\|)$ is complete for some $r > 0$.

Solution First we show if S_r is complete then X is complete. Let $\{x_n\}$ be a Cauchy sequence in X . There are two cases: The first case is that x_n gets arbitrarily close to the origin, and the second case is that for large enough n , x_n is uniformly away from the origin. Since the sequence is Cauchy, these are the only two possible cases. For the first case we have $\forall \epsilon > 0, \exists N > 0$ such that $\forall n > N$ we have $\|x_n\| < \epsilon$, we have $x_n \rightarrow 0 \in X$, thus converges. The second case is that for n large enough, the sequence is uniformly away from the origin. For this case case, where $\exists \epsilon > 0$ and $\exists N > 0$ such that $\|x_n\| > 0$. We map the sequence $\{x_n\}$ to a sequence $\{y_n\} \subset S_r$ via

$$y_n = \frac{rx_n}{\|x_n\|}.$$

Since this map is continuous (using the fact that every Cauchy sequence is bounded, so is $\{x_n\}$), y_n is also Cauchy, thus it converges: $y_n \rightarrow y \in S_r$. Using the continuity of $\|\cdot\|$, the sequence $\{\|x_n\|\} \subset \mathbb{R}$ is Cauchy, thus it converges: i.e. $\|x_n\| \rightarrow a \in \mathbb{R}$. We claim $x_n \rightarrow x = ay/r$. Because if we assume other wise, then $\{y_n\}$ is not converging to y , which is against our hypothesis.

For the other direction, we assume X is Banach and we want to conclude S_r is complete. Let $\{x_n\} \subset S_r$ be a Cauchy sequence. Since $S_r \subset X$, then $\{x_n\}$ is also a Cauchy sequence in X , hence converges to $x \in X$. We want to show that $\|x\| = r$, thus belongs to S_r . To show this we use the fact that $\|\cdot\cdot\cdot\|$ is a continuous map. So since $\|x_n\| = r$ for all n , it follows that $\|x\| = 1$ as well.

■ **Problem 3.1.1.5** Let p satisfy $0 < p < 1$ and consider the space $L_p[0, 1]$ of all functions with

$$\|x\| = \int_0^1 |x(t)|^p dt < \infty.$$

Show that $\|x\|$ is not a norm on $L_p[0, 1]$. Show that $d(x, y) = \|x - y\|$ is a metric on $L_p[0, 1]$. Hint: Note that if $0 \leq \alpha \leq 1$ then $\alpha < \alpha^p \leq 1$.

Solution The fact that $\|\cdot\|$ as defined above is not a norm follows from the fact that it is not homogeneous. I.e. for $\alpha \in \mathbb{R}$ we have $\|\alpha x\| = |\alpha|^p \|x\|$. But $d(x, y) = \|x - y\|$ is a metric. That is because it satisfies the metric properties, in particular $d(x, y) \leq d(x, z) + d(z, y)$. See the following remark for the triangle inequality.

■ **Remark 3.2** Let $x, y \in \mathbb{R}$. Then we want to show for $p \in (0, 1)$ we have

$$|x + y|^p \leq |x|^p + |y|^p.$$

To see this let $\beta = 1 - p$. Then we can write

$$|x + y|^p = |x + y|^{1-\beta} = \frac{|x + y|}{|x + y|^\beta} \leq \frac{|x|}{|x + y|^\beta} + \frac{|y|}{|x + y|^\beta} \leq \frac{|x|}{|x|^\beta} + \frac{|y|}{|y|^\beta} = |x|^p + |y|^p.$$

■ **Problem 3.1.1.6** Define

$$\begin{aligned}\sigma_n(f) &= \sup\{|f(t)| : |t| \leq n\}, \\ \rho_n(f) &= \min(1, \sigma_n(f)), \\ \|f\| &= \sum_{n=1}^{\infty} 2^{-n} \rho_n(f),\end{aligned}$$

where $f \in C(-\infty, \infty)$.

- (a) Show that $\sigma_n(f)$ is a pseudo-norm on $C(-\infty, \infty)$.
- (b) Show that $\rho_n(f)$ and $\|f\|$ are not norms.
- (c) Show that $d(f, g) = \|f - g\|$ is a metric on $C(-\infty, \infty)$.
- (d) Show that $d(f_n, f) \rightarrow 0$ as $n \rightarrow \infty$ if and only if $f_n(t) \rightarrow f(t)$ uniformly on compact sets in $-\infty < t < \infty$.

Solution (a) For any choice of n , the function

$$f(t) = \begin{cases} 0 & |t| < n, \\ x - n & t \geq n, \\ -x - n & t \leq n \end{cases},$$

and the origin both has norm zero.

- (b) Let $f \equiv 5$. Then $\sigma_n(f) = 5$ for all $n \in \mathbb{N}$ and $\rho_n(f) = 1$ for all $n \in \mathbb{N}$. For all $\alpha \geq 1/5$ we also have $\rho_n(\alpha f) = \rho_n(f)$. Thus ρ is not homogeneous. So it is not a norm. For the same reason, $\|\cdot\|$ is not a norm.
- (c) $d(x, y) = \|x - y\|$ satisfies all of the metric properties. In particular, for the triangle inequality we have

$$\sigma_n(f + g) = \sup\{|f + g|\} \leq \sup\{|f| + |g|\} \leq \sup\{|f|\} + \sup\{|g|\} = \sigma_n(|f|) + \sigma_n(|g|),$$

where $\sup\{|f|\}$ is used as a short notation for $\sup\{|f(t)| : |t| < n\}$. Furthermore, we know that $\min(1, a + b)$ for $a, b > 0$ we have $\min(1, a + b) \leq \min(1, a) + \min(1, b)$. Because there are 4 cases: $a, b < 1$ with their sum $a + b < 1$, $a, b < 1$ with $a + b > 1$, $a, b > 1$, and $a < 1 < b$. In all of these cases the inequality holds. Furthermore, we have

$$\|f + g\| = \sum_{n=1}^{\infty} 2^{-n} \rho_n(f + g) \leq \sum_{n=1}^{\infty} 2^{-n} \rho_n(f) + \sum_{n=1}^{\infty} 2^{-n} \rho_n(g).$$

This is true because $\rho_n(f + g) \leq 1$ and $\rho_n(f + g) \leq \rho_n(f) + \rho_n(g) \leq 2$. So the $\sum 2^{-n} \rho_n(f + g)$ is absolutely convergent. This $\sum 2^{-n} \rho_n(f + g) \leq \sum 2^{-n} (\rho_n(f) + \rho_n(g)) \leq \sum 2^{-n} \rho_n(f) + \sum 2^{-n} \rho_n(g)$.

- (d) Still thinking!

■ **Problem 3.1.1.7** Let X be a linear space and let $\|x\|$ be a real-valued function defined on X . Show that $d(x, y) = \|x - y\|$ is a metric if and only if $\|x\|$ satisfies (N1), (N2), (N4), and $\|x\| = \|-x\|$ for all $x \in X$.

Solution Assume $\|x\|$ satisfies (N1), (N2), (N4), and $\|x\| = \|-x\|$. Then $d(x, y) = \|x - y\|$ is positive (from N1) and $d(x, y) = 0$ iff $x = y$ (from positive definiteness of $\|\cdot\|$), and $d(x, y) = d(y, x)$ from $\|x\| = \|-x\|$. $d(x, y) \leq d(x, z) + d(z, y)$ from the triangle inequality for $\|\cdot\|$.

Now assume $d(x, y) = \|x - y\|$ is a metric. $\|x\| = d(x, 0) \leq 0$, so $\|\cdot\|$ is non-negative. Also $d(x, y) = d(y, x)$ implies $\|x\| = \|-x\|$. Furthermore, in $d(x, y) \leq d(x, z) + d(z, y)$, let $z = 0$, and we will recover $\|x - y\| \leq \|x\| + \|-y\|$. Let $\tilde{y} = -y$. Then $\|x + \tilde{y}\| \leq \|x\| + \|\tilde{y}\|$, hence the triangle inequality for the norm. Lastly, for the positive definiteness, Let $\|x\| = 0$. Since $\|x\| = \|x - 0\| = d(x, 0)$, it follows that $d(x, 0) = 0$. Thus $x = 0$ from metric properties. Also if $x = 0$, then $\|0\| = \|0 - 0\| = d(0, 0) = 0$.

■ **Problem 3.1.1.8** Let $(X, \|\cdot\|)$ be a normed linear space and let $\{x_n\}$ be a sequence in X with $x = \lim_{n \rightarrow \infty} x_n$. Assume that $\|x_n - y\| \leq a$ for all n . Show that $\|x - y\| \leq a$.

Solution Since $x_n \rightarrow x$, from the continuity of the addition in vector space, it also follows that $x_n - y \rightarrow x - y$. From continuity of $\|\cdot\|$ it follows that the real sequence $\{\|x_n - y\|\}$ converges to $\|x - y\|$. Since all elements of the real sequence $\|x_n - y\|$ is less than or equal to a , then the limit will have $\|x - y\| \leq a$. This property for the real sequences can be shown with a simple proof by contradiction.

■ **Problem 3.1.1.9** Show that a Hamel basis for a Banach space is either finite or uncountably infinite. (Completeness is important for this. Use Exercise 17, section 3.13).