

3.1

(i)

$$\begin{aligned}
(\|\mathbf{x} + \mathbf{y}\|^2 - \|\mathbf{x} - \mathbf{y}\|^2) &= \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle \\
&= 2\langle \mathbf{x}, \mathbf{y} \rangle + 2\langle \mathbf{y}, \mathbf{x} \rangle \\
&= 4\langle \mathbf{x}, \mathbf{y} \rangle
\end{aligned}$$

$$\frac{1}{4}(\|\mathbf{x} + \mathbf{y}\|^2 - \|\mathbf{x} - \mathbf{y}\|^2) = \langle \mathbf{x}, \mathbf{y} \rangle$$

as desired.

(ii)

$$\begin{aligned}
(\|\mathbf{x} + \mathbf{y}\|^2 + \|\mathbf{x} - \mathbf{y}\|^2) &= \|\mathbf{x}\|^2 + \|\mathbf{x}\|^2 + \|\mathbf{y}\|^2 + \|\mathbf{y}\|^2 \\
&= 2\|\mathbf{x}\|^2 + 2\|\mathbf{y}\|^2 \\
&= 2(\|\mathbf{x}\|^2 + \|\mathbf{y}\|^2)
\end{aligned}$$

$$\frac{1}{2}(\|\mathbf{x} + \mathbf{y}\|^2 + \|\mathbf{x} - \mathbf{y}\|^2) = \|\mathbf{x}\|^2 + \|\mathbf{y}\|^2$$

as desired.

3.2

$$\begin{aligned}
\frac{1}{4}(\|\mathbf{x} + \mathbf{y}\|^2 - \|\mathbf{x} - \mathbf{y}\|^2 + i\|\mathbf{x} - i\mathbf{y}\|^2 - i\|\mathbf{x} + i\mathbf{y}\|^2) \\
= \frac{1}{4}(4\langle \mathbf{x}, \mathbf{y} \rangle)
\end{aligned}$$

3.3

$$(i) \quad \cos \theta = \frac{\int_0^1 (x)(x^5) dx}{\sqrt{\int_0^1 x^2 dx} \sqrt{\int_0^1 x^{10} dx}} = \frac{1/7}{\sqrt{1/3} \sqrt{1/11}} = \frac{1/7}{\sqrt{1/33}} = \frac{\sqrt{33}}{7}; \quad \boxed{\theta = \cos^{-1} \left(\frac{\sqrt{33}}{7} \right)}$$

$$(ii) \quad \cos \theta = \frac{\int_0^1 x^6 dx}{\sqrt{\int_0^1 x^4 dx} \sqrt{\int_0^1 x^8 dx}} = \frac{1/7}{\sqrt{1/5} \sqrt{1/9}} = \frac{3\sqrt{5}}{7}; \quad \boxed{\theta = \cos^{-1} \left(\frac{3\sqrt{5}}{7} \right)}$$

3.8

- (i) *Proof.* Let V be the inner product space $C([- \pi, \pi]; \mathbb{R})$ with inner product

$$\langle f, g \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} f(t)g(t)dt$$

Let $X = \text{span}(S) \subset V$, where $S = \{\cos(t), \sin(t), \cos(2t), \sin(2t)\}$. Observe that

$$\begin{aligned}\langle \cos(t), \sin(t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(t) \sin(t) dt = 0 \\ \langle \cos(t), \cos(2t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(t) \cos(2t) dt = 0 \\ \langle \cos(t), \sin(2t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(t) \sin(2t) dt = 0 \\ \langle \sin(t), \cos(2t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(t) \cos(2t) dt = 0 \\ \langle \sin(t), \sin(2t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(t) \sin(2t) dt = 0 \\ \langle \cos(2t), \sin(2t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(2t) \sin(2t) dt = 0\end{aligned}$$

Thus, each function is orthogonal to each other. Now, observe that

$$\begin{aligned}\frac{1}{\pi} \int_{-\pi}^{\pi} \cos(t) \cos(t) dt &= 1 \\ \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(t) \sin(t) dt &= 1 \\ \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(2t) \cos(2t) dt &= 1 \\ \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(2t) \sin(2t) dt &= 1\end{aligned}$$

Therefore, S is an orthonormal set. □

(ii) $\|t\| = \frac{1}{\pi} \int_{-\pi}^{\pi} t^2 dt = \frac{1}{\pi} \left(\frac{2\pi^3}{3} \right) = \frac{2\pi^2}{3}$

(iii) Recall from part (i) that each component of S is orthogonal to one another. Using this fact, $\text{proj}_X(\cos(3t)) = 0$

(iv) Note that $\langle \cos(t), t \rangle = 0$ and $\langle \cos(2t), t \rangle = 0$.

Now, $\langle \sin(t), t \rangle \sin(t) = 2 \sin(t)$ and $\langle \sin(2t), t \rangle \sin(2t) = -\sin(2t)$

Take the sum of each of these yields $\boxed{\text{proj}_X(t) = 2 \sin(t) - \sin(2t)}$

3.9

Proof. $R_\theta = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$ Now observe that

$$\begin{aligned} R_\theta^H R_\theta &= \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} \\ &= \begin{bmatrix} \cos^2 \theta + \sin^2 \theta & \cos \theta \sin \theta - \sin \theta \cos \theta \\ \sin \theta \cos \theta - \cos \theta \sin \theta & \sin^2 \theta + \cos^2 \theta \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \\ &= I \end{aligned}$$

By Theorem 3.2.15, the rotation is an orthonormal transformation. \square

3.10

(i) Let $Q \in M_n(\mathbb{F})$ be orthonormal. So, $\langle m, n \rangle = \langle Qm, Qn \rangle$. Observe that

$$(Qm)^H Qn = m^H n \Leftrightarrow m^H Q^H Qn = m^H n$$

Hence, $Q^H Q = I$. Note also that $\langle Qm, Qn \rangle = \langle m, n \rangle$. So,

$$m^H n = (Qm)^H Qn = m^H Q^H Qn$$

(ii) *Proof.* Let $Q \in M_n(\mathbb{F})$ be an orthonormal matrix. Observe that

$$\|x\|^2 = \langle x, x \rangle = \langle Qx, Qx \rangle = \|Qx\|^2$$

□

(iii) *Proof.* Let $Q \in M_n(\mathbb{F})$ be an orthonormal matrix. Observe that

$$\begin{aligned} QQ^H &= Q^H Q = I \\ \Rightarrow Q^H &= Q^{-1} \end{aligned}$$

Now observe that

$$\begin{aligned} (Q^H)^H &= Q \\ (Q^H)(Q^H)^H &= (Q)^H Q = I \end{aligned}$$

So $Q^H = Q^{-1}$ is an orthonormal matrix. □

(iv)

(v) No. Let $Q = \begin{bmatrix} 2 & 0 \\ 0 & \frac{1}{2} \end{bmatrix}$ so $Q^T = \begin{bmatrix} 2 & 0 \\ 0 & \frac{1}{2} \end{bmatrix}$. Note that $\det(Q) = 1$. Observe that

$$\begin{bmatrix} 2 & 0 \\ 0 & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & \frac{1}{2} \end{bmatrix} = \begin{bmatrix} 4 & 0 \\ 0 & \frac{1}{4} \end{bmatrix} \neq I$$

and $Q \in M_n(\mathbb{F})$

(vi) Let $Q_1, Q_2 \in M_n(\mathbb{F})$ be orthonormal matrices. Hence, $Q_1^H Q_1 = I$ and $Q_2^H Q_2 = I$. Let $Q = Q_1 Q_2$. Observe that

$$\begin{aligned} Q^H Q &= (Q_1 Q_2)^H (Q_1 Q_2) \\ &= Q_2^H Q_1^H Q_1 Q_2 \\ &= Q_2^H Q_2 \\ &= I \end{aligned}$$

Therefore, $Q_1 Q_2$ is an orthonormal matrix.

3.11 When we apply the Gram-Schmidt orthonormalization process to a collection of linearly *dependent* vectors then the k th step will produce **0**

since \mathbf{x}_k is a linear combination of $\mathbf{x}_1, \dots, \mathbf{x}_{k-1}$. Note also that by producing the zero vector or any multiple of it, the length of it will not be equal to one, which is a necessary condition to be orthonormal.

3.16

(i)

(ii)

3.17

Proof. Let $A \in M_{m \times n}$ have rank $n \leq m$, and let $A = \hat{Q}\hat{R}$ be a reduced QR decomposition. Note that \hat{Q} is an $m \times n$ orthonormal matrix and \hat{R} is an $n \times n$ upper-triangular matrix (see Remark 3.3.10). Observe that

$$\begin{aligned} A^H A \mathbf{x} &= A^H \mathbf{b} \\ (\hat{Q}\hat{R})^H \hat{Q}\hat{R} \mathbf{x} &= (\hat{Q}\hat{R})^H \mathbf{b} \\ \hat{R}^H \hat{Q}^H \hat{Q}\hat{R} \mathbf{x} &= \hat{R}^H \hat{Q}^H \mathbf{b} \\ \hat{R}^H \hat{R} \mathbf{x} &= \hat{R}^H \hat{Q}^H \mathbf{b} \\ \hat{R} \mathbf{x} &= \hat{Q}^H \mathbf{b} \end{aligned}$$

□

3.23 Let $(V, \|\cdot\|)$ be a normed linear space. Observe that

$$\begin{aligned} \|x - y + y\| &\leq \|x - y\| + \|y\| \\ \|x\| &\leq \|x - y\| + \|y\| \\ \|x\| - \|y\| &\leq \|x - y\| \end{aligned}$$

Also observe that

$$\begin{aligned} \|y - x + x\| &\leq \|y - x\| + \|x\| \\ \|y\| &\leq \|x - y\| + \|x\| \\ \|y\| - \|x\| &\leq \|x - y\| \end{aligned}$$

Therefore, $|||x| - |y||| \leq ||x - y||$

3.24 Let $C([a, b]; \mathbb{F})$ be the vector space of all continuous functions from $[a, b] \subset \mathbb{R}$ to \mathbb{F} .

(i) $||f||_{L^1} = \int_a^b |f(t)| dt$

- positivity: $|f(t)| \geq 0$ by definition. If $f(t) = 0$, then $|f(t)| = 0$ and if $f(t) \neq 0$ then $|f(t)| \neq 0$. Thus $\int_a^b |f(t)| dt$ is positive.
- scale preservation: Let $\alpha \in \mathbb{R}$ be a scalar. Note that $||\alpha f(t)||_{L^1} = \int_a^b |\alpha f(t)| dt = \int_a^b |\alpha| |f(t)| dt = |\alpha| \int_a^b |f(t)| dt$
- triangle inequality:

$$\begin{aligned} ||f + g||_{L^1} &= \int_a^b |f(t) + g(t)| dt \\ &\leq \int_a^b |f(t)| + |g(t)| dt \\ &= \int_a^b |f(t)| dt + \int_a^b |g(t)| dt \end{aligned}$$

Therefore, $||f||_{L^1}$ is a norm on $C([a, b]; \mathbb{F})$.

(ii) $||f||_{L^2} = \left(\int_a^b |f(t)|^2 dt \right)^{\frac{1}{2}}$

- positivity: $|f(t)|$ is positive (see exercise 3.24 (i)). It follows that $|f(t)|^2$ is also positive. Hence $\int_a^b |f(t)|^2 dt$ is positive. Now, if $\int_a^b |f(t)|^2 dt = 0$ then $\left(\int_a^b |f(t)|^2 dt \right)^{\frac{1}{2}} = 0$. But if $\int_a^b |f(t)|^2 dt \neq 0$ then $\left(\int_a^b |f(t)|^2 dt \right)^{\frac{1}{2}} \neq 0$ and so $\left(\int_a^b |f(t)|^2 dt \right)^{\frac{1}{2}}$ is positive.

- scalar preservation: For some scalar $\alpha \in \mathbb{R}$ observe that

$$\begin{aligned}
\|\alpha f\|_{L^2} &= \left(\int_a^b |\alpha f(t)|^2 dt \right)^{\frac{1}{2}} \\
&= \left(\int_a^b |\alpha|^2 |f(t)|^2 dt \right)^{\frac{1}{2}} \\
&= \left(|\alpha|^2 \int_a^b |f(t)|^2 dt \right)^{\frac{1}{2}} \\
&= |\alpha| \left(\int_a^b |f(t)|^2 dt \right)^{\frac{1}{2}}
\end{aligned}$$

- triangle inequality: $\|f+g\|_{L^2} = \left(\int_a^b |f(t)+g(t)|^2 dt \right)^{\frac{1}{2}}$. Observe that

$$\begin{aligned}
(\|f+g\|_{L^2})^2 &= \int_a^b |f(t)+g(t)|^2 dt \\
&= \int_a^b |f(t)|^2 + 2|f(t)g(t)| + |g(t)|^2 dt \\
&= \|f\|_{L^2}^2 + 2|f(t)g(t)| + \|g\|_{L^2}^2 \\
&\leq \|f\|_{L^2}^2 + 2\|f\|_{L^2}\|g\|_{L^2} + \|g\|_{L^2}^2 \\
&= (\|f\|_{L^2} + \|g\|_{L^2})^2
\end{aligned}$$

Hence $\|f+g\|_{L^2} \leq \|f\|_{L^2} + \|g\|_{L^2}$

Therefore $\|f\|_{L^2}$ is a norm on $C([a, b]; \mathbb{F})$.

$$(iii) \quad \|f\|_{L^\infty} = \sup_{x \in [a, b]} |f(x)|$$

- positivity: $|f(x)| \geq 0$ by definition. If $f(x) = 0$ then $|f(x)| = 0$, and if $f(x) \neq 0$ then $|f(x)| \neq 0$. Thus $\sup_{x \in [a, b]} |f(x)|$ is positive.

- scalar preservation: For some scalar $\alpha \in \mathbb{R}$ observe that

$$\begin{aligned}
\|\alpha f\|_{L^\infty} &= \sup_{x \in [a,b]} |\alpha f(x)| \\
&= \sup_{x \in [a,b]} |\alpha| |f(x)| \\
&= |\alpha| \sup_{x \in [a,b]} |f(x)|
\end{aligned}$$

- triangle inequality: Observe that

$$\begin{aligned}
\|f + g\|_{L^\infty} &= \sup_{x \in [a,b]} |f(x) + g(x)| \\
&\leq \sup_{x \in [a,b]} (|f(x)| + |g(x)|) \\
&= \sup_{x \in [a,b]} |f(x)| + \sup_{x \in [a,b]} |g(x)|
\end{aligned}$$

Therefore, $\|f\|_{L^\infty}$ is a norm on $C([a, b]; \mathbb{F})$.

3.26 Reflexive: Note that $\|x\|_a = \|x\|_a$. Given $0 \leq m \leq M$ we have $m\|x\|_a \leq \|x\|_a \leq M\|x\|_a$ where $\|x\|_a \leq \|x\|_a \leq \|x\|_a$ for $m, M = 1$. Thus, $\|x\|_a \sim \|x\|_a$.

Symmetric: Suppose $\|x\|_a \sim \|x\|_b$. There exists $m, M \in \mathbb{R}$ with $0 < m \leq M$ such that $m\|x\|_a \leq \|x\|_b \leq M\|x\|_a$. Hence $\frac{1}{M}\|x\|_b \leq \|x\|_a \leq \frac{1}{m}\|x\|_b$. Thus $\|x\|_b \sim \|x\|_a$.

Transitive: Suppose $\|x\|_a \sim \|x\|_b$ and $\|x\|_b \sim \|x\|_c$. There exists $m_1, M_1, m_2, M_2 \in \mathbb{R}$ with $0 < m_1 \leq M_1$ and $0 < m_2 \leq M_2$ such that $m_1\|x\|_a \leq \|x\|_b \leq M_1\|x\|_a$ and $m_2\|x\|_b \leq \|x\|_c \leq M_2\|x\|_b$. Now $m_2\|x\|_b \leq \|x\|_c$ so $m_1m_2\|x\|_a \leq \|x\|_c$ and $\|x\|_c \leq M_2\|x\|_b$. So $\|x\|_c \leq M_1M_2\|x\|_a$. Putting it all together we have $m_1m_2\|x\|_a \leq \|x\|_c \leq M_1M_2\|x\|_a$. Thus $\|x\|_a \sim \|x\|_c$.

Therefore, topological equivalence is an equivalence relation.

(i)

$$\begin{aligned} |x_1|^2 + |x_2|^2 + \dots + |x_n|^2 &\leq (|x_1| + |x_2| + \dots + |x_n|)^2 \\ (|x_1|^2 + |x_2|^2 + \dots + |x_n|^2)^{\frac{1}{2}} &\leq |x_1| + |x_2| + \dots + |x_n| \\ \|x\|_2 &\leq \|x\|_1 \end{aligned}$$

Now, $|\langle x, \mathbf{1} \rangle| \leq \|x\|_2 \|\mathbf{1}\|_2$ (Cauchy-Schwartz) where
 $\|\mathbf{1}\|_2 = (1^2 + 1^2 + \dots + 1^2)^{\frac{1}{2}} = \sqrt{n}$. So, $\|x\|_1 \leq \sqrt{n} \|x\|_2$.
Therefore, $\|x\|_2 \leq \|x\|_1 \leq \sqrt{n} \|x\|_2$

(ii)

3.28 Let A be an $n \times n$ matrix.

(i)

3.29

3.30

Proof. Let $S \in M_n(\mathbb{F})$ be an invertible matrix. Given any matrix norm $\|\cdot\|$ on M_n define $\|\cdot\|_S$ by $\|A\|_S = \|SAS^{-1}\|$. By definition 3.5.15, any norm $\|\cdot\|$ on the finite-dimensional vector space $M_n(\mathbb{F})$ that satisfies the submultiplicative property is called a *matrix norm*. \square

3.37

3.38

3.39

Proof. let V and W be finite-dimensional inner-product spaces.

(i) Let $S, T \in \mathcal{L}(V; W)$. Observe that

$$\langle (S + T)(\mathbf{v}), \mathbf{w} \rangle = \langle \mathbf{v}, (S + T)^*(\mathbf{w}) \rangle$$

which follows from the definition of adjoint. Now,

$$\begin{aligned}
\langle (S + T)(\mathbf{v}), \mathbf{w} \rangle &= \langle S(\mathbf{v}) + T(\mathbf{v}), \mathbf{w} \rangle \\
&= \langle S(\mathbf{v}), \mathbf{w} \rangle + \langle T(\mathbf{v}), \mathbf{w} \rangle \\
&= \langle \mathbf{v}, S^*(\mathbf{w}) \rangle + \langle \mathbf{v}, T^*(\mathbf{w}) \rangle \\
&= \langle \mathbf{v}, (S^* + T^*)(\mathbf{w}) \rangle
\end{aligned}$$

Hence, $(S + T)^* = S^* + T^*$ Now, observe that for $\alpha \in \mathbb{F}$

$$\langle (\alpha T)(\mathbf{v}), \mathbf{w} \rangle = \langle \mathbf{v}, (\alpha T)^*(\mathbf{w}) \rangle$$

Additionally,

$$\begin{aligned}
\langle (\alpha T)(\mathbf{v}), \mathbf{w} \rangle &= \langle \alpha T(\mathbf{v}), \mathbf{w} \rangle \\
&= \alpha \langle \mathbf{v}, T^*(\mathbf{w}) \rangle \\
&= \langle \mathbf{v}, \bar{\alpha} T^*(\mathbf{w}) \rangle
\end{aligned}$$

Hence, $(\alpha T)^* = \bar{\alpha} T^*$

(ii) Let $S \in \mathcal{L}(V; W)$. Observe that

$$\langle (S^*)^*(\mathbf{v}), \mathbf{w} \rangle = \langle \mathbf{v}, (S^*)^*(\mathbf{w}) \rangle$$

Additionally,

$$\begin{aligned}
\langle S^*(\mathbf{w}), \mathbf{v} \rangle &= \overline{\langle \mathbf{v}, S^*(\mathbf{w}) \rangle} \\
&= \overline{\langle S(\mathbf{v}), \mathbf{w} \rangle} \\
&= \langle \mathbf{w}, S(\mathbf{v}) \rangle
\end{aligned}$$

Hence, $(T^*)^* = T$

(iii) Let $S, T \in \mathcal{L}(V)$. Observe that

$$\langle (ST)(\mathbf{v}), \mathbf{w} \rangle = \langle \mathbf{v}, (ST)^*(\mathbf{w}) \rangle$$

Additionally,

$$\begin{aligned}
\langle (ST)(\mathbf{v}), \mathbf{w} \rangle &= \langle S(T(\mathbf{v})), \mathbf{w} \rangle \\
&= \langle T(\mathbf{v}), S^*(\mathbf{w}) \rangle \\
&= \langle \mathbf{v}, T^*(S^*(\mathbf{w})) \rangle \\
&= \langle \mathbf{v}, (T^*S^*)(\mathbf{w}) \rangle
\end{aligned}$$

Hence, $(ST)^* = T^*S^*$

(iv) Let $T \in \mathcal{L}(V)$ and suppose T is invertible. Observe that

$$\begin{aligned}
(T^*)^{-1}T^* &= I \\
((T^*)^{-1}T^*)^* &= I^*
\end{aligned}$$

But $I^* = I$. From property (iii) we have that $T^{**}((T^*)^{-1})^* = I$. Note that $T^{**} = T$, so we have that

$$\begin{aligned}
T((T^*)^{-1})^* &= I \\
((T^*)^{-1})^* &= T^{-1} \\
(T^*)^{-1} &= (T^{-1})^*
\end{aligned}$$

□

3.40

3.44

3.45

3.46

3.47 Assume A is an $m \times n$ matrix of rank n . Let $P = A(A^H A)^{-1} A^H$

(i)

$$P^2 = A(A^H A)^{-1} A^H A(A^H A)^{-1} A^H \quad (0.1)$$

$$= A(A^H A)^{-1} A^H \quad (0.2)$$

$$(0.3)$$

(ii)

$$\begin{aligned} P^H &= (A(A^H A)^{-1} A^H)^H \\ &= A((A^H A)^{-1})^H A^H \\ &= A((A^H A)^H)^{-1} A^H \\ &= A(A^H A)^{-1} A^H \\ &= P \end{aligned}$$

(iii)

$$\text{rank}(P) = \text{rank}(A(A^H A)^{-1} A^H)$$

Now because P is idempotent by (i) we have that

$$\begin{aligned} \text{tr}(A(A^H A)^{-1} A^H) &= \text{tr}(A^H A(A^H A)^{-1}) \\ &= \text{tr}(I) \end{aligned}$$

where I is $n \times n$, so it has rank n . Therefore, $\text{rank}(P)=n$.

3.48

3.50