

OSM Boot Camp: Math ProbSet2

Harrison Beard

2 July 2018

Exercise 3.1.

Question. Verify the polarization and parallelogram identities on a real inner product space, with the usual norm $\|\mathbf{x}\| = \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle}$ arising from the inner product:

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

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

It can be shown that in any normed linear space over \mathbb{R} for which (ii) holds, one can define an inner product by using (i).

Answer.

(i)

$$\begin{aligned} \frac{1}{4} (\|\mathbf{x} + \mathbf{y}\|^2 - \|\mathbf{x} - \mathbf{y}\|^2) &= \frac{1}{4} \left(\left(\sqrt{\langle \mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle} \right)^2 - \left(\sqrt{\langle \mathbf{x} - \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle} \right)^2 \right) \\ &= \frac{1}{4} (\langle \mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle - \langle \mathbf{x} - \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle) \\ &= \frac{1}{4} (\langle \mathbf{x} + \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{x} + \mathbf{y}, \mathbf{y} \rangle - \langle \mathbf{x} - \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{x} - \mathbf{y}, \mathbf{y} \rangle) \\ &= \frac{1}{4} (\overline{\langle \mathbf{x}, \mathbf{x} + \mathbf{y} \rangle} + \overline{\langle \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle} - \overline{\langle \mathbf{x}, \mathbf{x} - \mathbf{y} \rangle} + \overline{\langle \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle}) \\ &= \frac{1}{4} (\langle \mathbf{x}, \mathbf{x} + \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle - \langle \mathbf{x}, \mathbf{x} - \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle) \\ &= \frac{1}{4} (\cancel{\langle \mathbf{x}, \mathbf{x} \rangle} + \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle + \cancel{\langle \mathbf{y}, \mathbf{y} \rangle} - \cancel{\langle \mathbf{x}, \mathbf{x} \rangle} + \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle - \cancel{\langle \mathbf{y}, \mathbf{y} \rangle}) \\ &= \frac{1}{4} (\langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle) \\ &= \frac{1}{4} (\langle \mathbf{x}, \mathbf{y} \rangle + \overline{\langle \mathbf{x}, \mathbf{y} \rangle} + \langle \mathbf{x}, \mathbf{y} \rangle + \overline{\langle \mathbf{x}, \mathbf{y} \rangle}) \\ &= \frac{1}{4} (\langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{y} \rangle) \\ &= \frac{1}{4} (4 \cdot \langle \mathbf{x}, \mathbf{y} \rangle) \\ &= \langle \mathbf{x}, \mathbf{y} \rangle. \end{aligned}$$

■

(ii)

$$\begin{aligned}
\frac{1}{2} (\|\mathbf{x} + \mathbf{y}\|^2 + \|\mathbf{x} - \mathbf{y}\|^2) &= \frac{1}{2} \left(\left(\sqrt{\langle \mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle} \right)^2 + \left(\sqrt{\langle \mathbf{x} - \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle} \right)^2 \right) \\
&= \frac{1}{2} (\langle \mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle + \langle \mathbf{x} - \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle) \\
&= \frac{1}{2} (\langle \mathbf{x} + \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{x} + \mathbf{y}, \mathbf{y} \rangle + \langle \mathbf{x} - \mathbf{y}, \mathbf{x} \rangle - \langle \mathbf{x} - \mathbf{y}, \mathbf{y} \rangle) \\
&= \frac{1}{2} (\overline{\langle \mathbf{x}, \mathbf{x} + \mathbf{y} \rangle} + \overline{\langle \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle} + \overline{\langle \mathbf{x}, \mathbf{x} - \mathbf{y} \rangle} - \overline{\langle \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle}) \\
&= \frac{1}{2} (\langle \mathbf{x}, \mathbf{x} + \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{x} - \mathbf{y} \rangle - \langle \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle) \\
&= \frac{1}{2} (\langle \mathbf{x}, \mathbf{x} \rangle + \cancel{\langle \mathbf{x}, \mathbf{y} \rangle} + \cancel{\langle \mathbf{y}, \mathbf{x} \rangle} + \langle \mathbf{y}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{x} \rangle - \cancel{\langle \mathbf{x}, \mathbf{y} \rangle} - \cancel{\langle \mathbf{y}, \mathbf{x} \rangle} + \langle \mathbf{y}, \mathbf{y} \rangle) \\
&= \frac{1}{2} (\langle \mathbf{x}, \mathbf{x} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{x} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle) \\
&= \frac{1}{2} (2 \cdot \langle \mathbf{x}, \mathbf{x} \rangle + 2 \cdot \langle \mathbf{y}, \mathbf{y} \rangle) \\
&= \langle \mathbf{x}, \mathbf{x} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle \\
&= \|\mathbf{x}\|^2 + \|\mathbf{y}\|^2.
\end{aligned}$$

■

Exercise 3.2.

Question. Verify the polarization identity on a complex inner product space, with the usual norm $\|\mathbf{x}\| = \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle}$ arising from the inner product:

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

A nice consequence of the polarization identity on a real or complex inner product space is that if two inner products induce the same norm, then the inner products are equal.

Answer.

$$\begin{aligned}
& \frac{1}{4} \left(\|\mathbf{x} + \mathbf{y}\|^2 - \|\mathbf{x} - \mathbf{y}\|^2 + i \|\mathbf{x} - i\mathbf{y}\|^2 - i \|\mathbf{x} + i\mathbf{y}\|^2 \right) \\
= & \frac{1}{4} \left(\left(\sqrt{\langle \mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle} \right)^2 - \left(\sqrt{\langle \mathbf{x} - \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle} \right)^2 \right. \\
& \left. + i \left(\sqrt{\langle \mathbf{x} - i\mathbf{y}, \mathbf{x} - i\mathbf{y} \rangle} \right)^2 - i \left(\sqrt{\langle \mathbf{x} + i\mathbf{y}, \mathbf{x} + i\mathbf{y} \rangle} \right)^2 \right) \\
= & \frac{1}{4} (\langle \mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle - \langle \mathbf{x} - \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle \\
& + i \cdot \langle \mathbf{x} - i\mathbf{y}, \mathbf{x} - i\mathbf{y} \rangle - i \cdot \langle \mathbf{x} + i\mathbf{y}, \mathbf{x} + i\mathbf{y} \rangle) \\
= & \frac{1}{4} (\langle \mathbf{x} + \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{x} + \mathbf{y}, \mathbf{y} \rangle - \langle \mathbf{x} - \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{x} - \mathbf{y}, \mathbf{y} \rangle \\
& + i \cdot \langle \mathbf{x} - i\mathbf{y}, \mathbf{x} \rangle - i^2 \cdot \langle \mathbf{x} - i\mathbf{y}, \mathbf{y} \rangle - i \cdot \langle \mathbf{x} + i\mathbf{y}, \mathbf{x} \rangle - i^2 \cdot \langle \mathbf{x} + i\mathbf{y}, \mathbf{y} \rangle) \\
= & \frac{1}{4} \left(\overline{\langle \mathbf{x}, \mathbf{x} + \mathbf{y} \rangle} + \overline{\langle \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle} - \overline{\langle \mathbf{x}, \mathbf{x} - \mathbf{y} \rangle} + \overline{\langle \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle} \right. \\
& \left. + i \cdot \overline{\langle \mathbf{x}, \mathbf{x} - i\mathbf{y} \rangle} - i^2 \cdot \overline{\langle \mathbf{y}, \mathbf{x} - i\mathbf{y} \rangle} - i \cdot \overline{\langle \mathbf{x}, \mathbf{x} + i\mathbf{y} \rangle} - i^2 \cdot \overline{\langle \mathbf{y}, \mathbf{x} + i\mathbf{y} \rangle} \right) \\
= & \frac{1}{4} (\langle \mathbf{x}, \mathbf{x} + \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} + \mathbf{y} \rangle - \langle \mathbf{x}, \mathbf{x} - \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} - \mathbf{y} \rangle \\
& + i \cdot \langle \mathbf{x}, \mathbf{x} + i\mathbf{y} \rangle - i^2 \cdot \langle \mathbf{y}, \mathbf{x} + i\mathbf{y} \rangle - i \cdot \langle \mathbf{x}, \mathbf{x} - i\mathbf{y} \rangle - i^2 \cdot \langle \mathbf{y}, \mathbf{x} - i\mathbf{y} \rangle) \\
= & \frac{1}{4} (\langle \mathbf{x}, \mathbf{x} \rangle + \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{y}, \mathbf{y} \rangle - \langle \mathbf{x}, \mathbf{x} \rangle + \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle - \langle \mathbf{y}, \mathbf{y} \rangle \\
& + i \cdot \langle \mathbf{x}, \mathbf{x} \rangle + i^2 \cdot \langle \mathbf{x}, \mathbf{y} \rangle - i^2 \cdot \langle \mathbf{y}, \mathbf{x} \rangle - i^3 \cdot \langle \mathbf{y}, \mathbf{y} \rangle - i \cdot \langle \mathbf{x}, \mathbf{x} \rangle + i^2 \cdot \langle \mathbf{x}, \mathbf{y} \rangle \\
& - i^2 \cdot \langle \mathbf{y}, \mathbf{x} \rangle + i^3 \cdot \langle \mathbf{y}, \mathbf{y} \rangle) \\
= & \frac{1}{4} \left(\cancel{\langle \mathbf{x}, \mathbf{x} \rangle} + \cancel{\langle \mathbf{x}, \mathbf{y} \rangle} + \langle \mathbf{y}, \mathbf{x} \rangle + \cancel{\langle \mathbf{y}, \mathbf{y} \rangle} - \cancel{\langle \mathbf{x}, \mathbf{x} \rangle} + \cancel{\langle \mathbf{x}, \mathbf{y} \rangle} + \langle \mathbf{y}, \mathbf{x} \rangle - \cancel{\langle \mathbf{y}, \mathbf{y} \rangle} \right. \\
& \left. + i \cdot \cancel{\langle \mathbf{x}, \mathbf{x} \rangle} - \cancel{\langle \mathbf{x}, \mathbf{y} \rangle} + \langle \mathbf{y}, \mathbf{x} \rangle + i \cdot \cancel{\langle \mathbf{y}, \mathbf{y} \rangle} - i \cdot \cancel{\langle \mathbf{x}, \mathbf{x} \rangle} - \cancel{\langle \mathbf{x}, \mathbf{y} \rangle} + \langle \mathbf{y}, \mathbf{x} \rangle - i \cdot \cancel{\langle \mathbf{y}, \mathbf{y} \rangle} \right) \\
= & \frac{1}{4} (\langle \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle + \langle \mathbf{y}, \mathbf{x} \rangle) \\
= & \frac{1}{4} (\overline{\langle \mathbf{x}, \mathbf{y} \rangle} + \overline{\langle \mathbf{x}, \mathbf{y} \rangle} + \overline{\langle \mathbf{x}, \mathbf{y} \rangle} + \overline{\langle \mathbf{x}, \mathbf{y} \rangle}) \\
= & \frac{1}{4} (\langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{y} \rangle) \\
= & \frac{1}{4} (4 \cdot \langle \mathbf{x}, \mathbf{y} \rangle) \\
= & \langle \mathbf{x}, \mathbf{y} \rangle.
\end{aligned}$$

■

Exercise 3.3.

Question. Let $\mathbb{R}[x]$ have the inner product

$$\langle f, g \rangle = \int_0^1 f(x)g(x) \, dx.$$

Using (3.8), find the angle θ between the following sets of vectors:

- (i) x and x^5 .
- (ii) x^2 and x^4 .

Answer.

- (i) In (3.8), we define $\cos \theta$ for the angle θ between vectors x and y as

$$\frac{\langle x, y \rangle}{\|x\| \|y\|}.$$

Solving in our case, we have

$$\begin{aligned} \cos \theta &= \frac{\langle x, x^5 \rangle}{\|x\| \|x^5\|} \\ &= \frac{\langle x, x^5 \rangle}{\sqrt{\langle x, x \rangle} \sqrt{\langle x^5, x^5 \rangle}} \\ &= \frac{\int_0^1 x^6 \, dx}{\sqrt{\int_0^1 x^2 \, dx} \sqrt{\int_0^1 x^{10} \, dx}} \\ &= \frac{\frac{1}{7} x^7 \Big|_0^1}{\frac{1}{3} x^3 \Big|_0^1 \cdot \frac{1}{11} x^{11} \Big|_0^1} \\ &= \frac{\frac{1}{7}}{\frac{1}{3} \cdot \frac{1}{11}} \\ &= \boxed{\frac{33}{7}}. \end{aligned}$$

- (ii)

$$\begin{aligned} \cos \theta &= \frac{\langle x^2, x^4 \rangle}{\|x^2\| \|x^4\|} \\ &= \frac{\langle x^2, x^4 \rangle}{\sqrt{\langle x^2, x^2 \rangle} \sqrt{\langle x^4, x^4 \rangle}} \\ &= \frac{\int_0^1 x^6 \, dx}{\sqrt{\int_0^1 x^4 \, dx} \sqrt{\int_0^1 x^8 \, dx}} \\ &= \frac{\frac{1}{7} x^7 \Big|_0^1}{\frac{1}{5} x^5 \Big|_0^1 \cdot \frac{1}{9} x^9 \Big|_0^1} \\ &= \frac{\frac{1}{7}}{\frac{1}{5} \cdot \frac{1}{9}} \\ &= \boxed{\frac{45}{7}}. \end{aligned}$$

Exercise 3.8.

Question. Let V be the inner product space $\mathcal{C}([-\pi, \pi]; \mathbb{R})$ 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 $S = \{\cos(t), \sin(t), \cos(2t), \sin(2t)\}$.

- (i) Prove that S is an orthonormal set.
- (ii) Compute $\|t\|$.
- (iii) Compute the projection $\text{proj}_X(\cos(3t))$.
- (iv) Compute the projection $\text{proj}_X(t)$.

Answer.

- (i) Normality:

$$\begin{aligned}
 \|\cos(t)\| &= \langle \cos(t), \cos(t) \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(t) \cos(t) dt \\
 &= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos^2(t) dt \\
 &= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\cos(2t) + \cos(0)) dt \\
 &= \underbrace{\frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} \cos(2t) dt}_{2t=:u \Rightarrow \begin{cases} \frac{du}{dt} = 2 \implies dt = \frac{1}{2} du \\ u \in 2 \cdot (-\pi, \pi) = (-2\pi, 2\pi) \end{cases}} + \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} dt \\
 &= \frac{1}{4\pi} \int_{-2\pi}^{2\pi} \cos(u) du + \frac{1}{2\pi} \cdot t \Big|_{-\pi}^{\pi} \\
 &= \frac{1}{4\pi} \cdot \sin(u) \Big|_{-2\pi}^{2\pi} + \frac{1}{2\pi} (\pi + \pi) \\
 &= \frac{1}{4\pi} (\sin(2\pi) - \sin(-2\pi)) + 1 \\
 &= \frac{1}{4\pi} (0) + 1 \\
 &= 1.
 \end{aligned}$$

$$\begin{aligned}
\|\sin(t)\| = \langle \sin(t), \sin(t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(t) \sin(t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \sin^2(t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\cos(0) - \cos(2t)) \, dt \\
&= \frac{1}{2\pi} \int_{-\pi}^{\pi} dt - \frac{1}{4\pi} \int_{-2\pi}^{2\pi} \cos(u) \, du \\
&= \frac{1}{2\pi} (2\pi) - \frac{1}{4\pi} \cdot \sin(u) \Big|_{-2\pi}^{2\pi} \\
&= 1.
\end{aligned}$$

$$\begin{aligned}
\|\cos(2t)\| = \langle \cos(2t), \cos(2t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(2t) \cos(2t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos^2(2t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\cos(4t) + \cos(0)) \, dt \\
&= \frac{1}{8\pi} \int_{-4\pi}^{4\pi} \cos(u) \, du + \frac{1}{2\pi} \cdot t \Big|_{-\pi}^{\pi} \\
&= \frac{1}{8\pi} \cdot \sin(u) \Big|_{-4\pi}^{4\pi} + \frac{1}{2\pi} (2\pi) \\
&= 1.
\end{aligned}$$

$$\begin{aligned}
\|\sin(2t)\| = \langle \sin(2t), \sin(2t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(2t) \sin(2t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \sin^2(2t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\cos(0) - \cos(4t)) \, dt \\
&= \frac{1}{2\pi} \int_{-\pi}^{\pi} dt - \frac{1}{8\pi} \int_{-4\pi}^{4\pi} \cos(u) \, du \\
&= \frac{1}{2\pi} (2\pi) - \frac{1}{8\pi} \cdot \sin(u) \Big|_{-4\pi}^{4\pi} \\
&= 1.
\end{aligned}$$

Orthogonality:

$$\begin{aligned}
\langle \cos(t), \sin(t) \rangle = \langle \sin(t), \cos(t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(t) \cos(t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\sin(2t) - \sin(0)) \, dt \\
&= \frac{1}{4\pi} \int_{-2\pi}^{2\pi} \sin(u) \, du - 0 \\
&= \frac{1}{4\pi} (-\cos(u)) \Big|_{-2\pi}^{2\pi} \\
&= \frac{1}{4\pi} (-1 + 1) \\
&= 0.
\end{aligned}$$

$$\begin{aligned}
\langle \cos(t), \cos(2t) \rangle = \langle \cos(2t), \cos(t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(2t) \cos(t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\cos(3t) + \cos(-t)) \, dt \\
&= \frac{1}{6\pi} \int_{-3\pi}^{3\pi} \cos(u) \, du - \frac{1}{2\pi} \cdot \sin(v) \Big|_{\pi}^{-\pi} \\
&= \frac{1}{6\pi} \sin(u) \Big|_{-3\pi}^{3\pi} - \frac{1}{2\pi} (\sin(-\pi) - \sin(\pi)) \\
&= \frac{1}{6\pi} (\sin(\pi) - \sin(-\pi)) - \frac{1}{2\pi} (\sin(-\pi) - \sin(\pi)) \\
&= \frac{1}{6\pi} (\sin(\pi) - \sin(-\pi)) - \frac{1}{2\pi} (\sin(-\pi) - \sin(\pi)) \\
&= 0.
\end{aligned}$$

$$\begin{aligned}
\langle \cos(t), \sin(2t) \rangle = \langle \sin(2t), \cos(t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(2t) \cos(t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\sin(3t) - \sin(-t)) \, dt \\
&= \frac{1}{6\pi} \int_{-3\pi}^{3\pi} \sin(u) \, du - \frac{1}{2\pi} \cdot \cos(-t) \Big|_{-\pi}^{\pi} \\
&= \frac{1}{6\pi} \cdot -\cos(u) \Big|_{-3\pi}^{3\pi} - \frac{1}{2\pi} (\cos(-\pi) - \cos(\pi)) \\
&= \frac{1}{6\pi} (-\cos(\pi) + \cos(-\pi)) - \frac{1}{2\pi} (-1 + 1) \\
&= \frac{1}{6\pi} (1 - 1) - 0 \\
&= 0.
\end{aligned}$$

$$\begin{aligned}
\langle \sin(t), \cos(2t) \rangle = \langle \cos(2t), \sin(t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(2t) \sin(t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\sin(3t) - \sin(t)) \, dt \\
&= \frac{1}{6\pi} \int_{-3\pi}^{3\pi} \sin(u) \, du - \frac{1}{2\pi} \cdot \cos(t) \Big|_{-\pi}^{\pi} \\
&= \frac{1}{6\pi} \cdot -\cos(u) \Big|_{-3\pi}^{3\pi} - \frac{1}{2\pi} (\cos(\pi) - \cos(-\pi)) \\
&= \frac{1}{6\pi} (-\cos(\pi) + \cos(-\pi)) - \frac{1}{2\pi} (1 - 1) \\
&= \frac{1}{6\pi} (1 - 1) - 0 \\
&= 0.
\end{aligned}$$

$$\begin{aligned}
\langle \sin(t), \sin(2t) \rangle = \langle \sin(2t), \sin(t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(2t) \sin(t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\cos(-t) + \cos(3t)) \, dt \\
&= -\frac{1}{2\pi} \int_{3\pi}^{-3\pi} \cos(u) \, du - \frac{1}{6\pi} \cdot \int_{-3\pi}^{3\pi} \cos(v) \, dv \\
&= -\frac{1}{2\pi} \cdot \sin(u) \Big|_{3\pi}^{-3\pi} - \frac{1}{6\pi} \cdot \sin(v) \Big|_{-3\pi}^{3\pi} \\
&= -\frac{1}{2\pi} (\sin(-\pi) - \sin(\pi)) - \frac{1}{6\pi} (\sin(\pi) - \sin(-\pi)) \\
&= 0.
\end{aligned}$$

$$\begin{aligned}
\langle \cos(2t), \sin(2t) \rangle = \langle \sin(2t), \cos(2t) \rangle &= \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(2t) \cos(2t) \, dt \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\sin(3t) - \sin(0)) \, dt \\
&= \frac{1}{6\pi} \int_{-3\pi}^{3\pi} \sin(u) \, du - 0 \\
&= \frac{1}{6\pi} \cdot -\cos(u) \Big|_{-3\pi}^{3\pi} \\
&= \frac{1}{6\pi} (-\cos(\pi) + \cos(-\pi)) \\
&= \frac{1}{6\pi} (1 - 1) \\
&= 0.
\end{aligned}$$

■

(ii)

$$\begin{aligned}
\|t\| &= \sqrt{\langle t, t \rangle} \\
&= \sqrt{\frac{1}{\pi} \int_{-\pi}^{\pi} t^2 \, dt} \\
&= \sqrt{\frac{1}{\pi} \cdot \frac{1}{3} t^3 \Big|_{-\pi}^{\pi}} \\
&= \sqrt{\frac{1}{\pi} \cdot \left(\frac{1}{3} \pi^3 - \frac{1}{3} (-\pi)^3 \right)} \\
&= \sqrt{\frac{2}{3} \pi^2} \\
&= \boxed{\sqrt{\frac{2}{3}} \pi}.
\end{aligned}$$

(iii)

$$\begin{aligned}
\text{proj}_X(\cos(3t)) &\triangleq \sum_{i=1}^m \langle \mathbf{x}_i, \cos(3t) \rangle \mathbf{x}_i \text{ for } \mathbf{x} \in \text{basis}(X) \text{ because } X \subset V \text{ is orthonormal} \\
&= \langle \cos(t), \cos(3t) \rangle \cdot \cos(t) \\
&\quad + \langle \sin(t), \cos(3t) \rangle \cdot \sin(t) \\
&\quad + \langle \cos(2t), \cos(3t) \rangle \cdot \cos(2t) \\
&\quad + \langle \sin(2t), \cos(3t) \rangle \cdot \sin(2t) \text{ because } X := \text{span}(\{\cos(t), \sin(t), \cos(2t), \sin(2t)\}) \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(t) \cos(3t) dt \cdot \cos(t) \\
&\quad + \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(t) \cos(3t) dt \cdot \sin(t) \\
&\quad + \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(2t) \cos(3t) dt \cdot \cos(2t) \\
&\quad + \frac{1}{\pi} \int_{-\pi}^{\pi} \sin(2t) \cos(3t) dt \cdot \sin(2t) \\
&= \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\cos(4t) + \cos(2t)) dt \cdot \cos(t) \\
&\quad + \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\sin(4t) - \sin(2t)) dt \cdot \sin(t) \\
&\quad + \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\cos(5t) + \cos(t)) dt \cdot \cos(2t) \\
&\quad + \frac{1}{\pi} \int_{-\pi}^{\pi} \frac{1}{2} (\sin(5t) - \sin(t)) dt \cdot \sin(2t) \\
&= \frac{\cos(t)}{2\pi} \left(\underbrace{\int_{-\pi}^{\pi} \cos(4t) dt + \int_{-\pi}^{\pi} \cos(2t) dt}_{4t=:u_1 \Rightarrow \begin{cases} \frac{du_1}{dt} = 4 \Rightarrow dt = \frac{1}{4} du_1 \\ u_1 \in 4 \cdot (-\pi, \pi) = (-4\pi, 4\pi) \end{cases}} \right) \\
&\quad + \frac{\sin(t)}{2\pi} \left(\int_{-\pi}^{\pi} \sin(4t) dt - \int_{-\pi}^{\pi} \sin(2t) dt \right) \\
&\quad + \frac{\cos(2t)}{2\pi} \left(\int_{-\pi}^{\pi} \cos(5t) dt + \int_{-\pi}^{\pi} \cos(t) dt \right) \\
&\quad + \frac{\sin(2t)}{2\pi} \left(\int_{-\pi}^{\pi} \sin(5t) dt - \int_{-\pi}^{\pi} \sin(t) dt \right) \\
&= \frac{\cos(t)}{2\pi} \left(\frac{1}{4} \int_{-4\pi}^{4\pi} \cos(u_1) du_1 + \frac{1}{2} \int_{-2\pi}^{2\pi} \cos(u_2) du_2 \right) \\
&\quad + \frac{\sin(t)}{2\pi} \left(\frac{1}{4} \int_{-4\pi}^{4\pi} \sin(u_3) du_3 - \frac{1}{2} \int_{-2\pi}^{2\pi} \sin(u_4) du_4 \right) \\
&\quad + \frac{\cos(2t)}{2\pi} \left(\frac{1}{5} \int_{-5\pi}^{5\pi} \cos(u_5) du_5 + \int_{-\pi}^{\pi} \cos(t) dt \right) \\
&\quad + \frac{\sin(2t)}{2\pi} \left(\frac{1}{5} \int_{-5\pi}^{5\pi} \sin(u_6) du_6 - \int_{-\pi}^{\pi} \sin(t) dt \right)
\end{aligned}$$

$$\begin{aligned}
&= \frac{\cos(t)}{2\pi} \left(\frac{1}{4} \cdot \sin(u_1)|_{-4\pi}^{4\pi} + \frac{1}{2} \cdot \sin(u_2)|_{-2\pi}^{2\pi} \right) \\
&\quad + \frac{\sin(t)}{2\pi} \left(\frac{1}{4} \cdot (-\cos(u_3))|_{-4\pi}^{4\pi} - \frac{1}{2} \cdot (-\cos(u_4))|_{-2\pi}^{2\pi} \right) \\
&\quad + \frac{\cos(2t)}{2\pi} \left(\frac{1}{5} \cdot \sin(u_5)|_{-5\pi}^{5\pi} + (\sin(t)|_{-\pi}^{\pi}) \right) \\
&\quad + \frac{\sin(2t)}{2\pi} \left(\frac{1}{5} \cdot (-\cos(u_6))|_{-5\pi}^{5\pi} - ((-\cos(t))|_{-\pi}^{\pi}) \right) \\
&= \frac{\cos(t)}{2\pi} \left(\frac{1}{4}(\sin(0) - \sin(0)) + \frac{1}{2}(\sin(0) - \sin(0)) \right) \\
&\quad + \frac{\sin(t)}{2\pi} \left(\frac{1}{4}(-\cos(0) + \cos(0)) - \frac{1}{2}(-\cos(0) + \cos(0)) \right) \\
&\quad + \frac{\cos(2t)}{2\pi} \left(\frac{1}{5}(\sin(\pi) - \sin(-\pi)) + (\sin(\pi) - \sin(-\pi)) \right) \\
&\quad + \frac{\sin(2t)}{2\pi} \left(\frac{1}{5}(-\cos(\pi) + \cos(-\pi)) - (-\cos(\pi) + \cos(-\pi)) \right) \\
&= \frac{\cos(2t)}{2\pi} \left(\frac{1}{5}(0-0) + (0-0) \right) + \frac{\sin(2t)}{2\pi} \left(\frac{1}{5}(1-1) - (1-1) \right) \\
&= \boxed{0}.
\end{aligned}$$

(iv)

$$\begin{aligned}
\text{proj}_X(t) &\triangleq \sum_{i=1}^m \langle \mathbf{x}_i, t \rangle \mathbf{x}_i \text{ for } \mathbf{x} \in \text{basis}(X) \text{ because } X \subset V \text{ is orthonormal} \\
&= \langle \cos(t), t \rangle \cdot \cos(t) \\
&\quad + \langle \sin(t), t \rangle \cdot \sin(t) \\
&\quad + \langle \cos(2t), t \rangle \cdot \cos(2t) \\
&\quad + \langle \sin(2t), t \rangle \cdot \sin(2t) \text{ because } X := \text{span}(\{\cos(t), \sin(t), \cos(2t), \sin(2t)\}) \\
&= \underbrace{\frac{1}{\pi} \int_{-\pi}^{\pi} \cos(t) \cdot t \, dt \cdot \cos(t)}_{\substack{\left\{ \begin{array}{l} u_1 = t \\ dv_1 = \cos(t) \end{array} \right\} \Rightarrow \left\{ \begin{array}{l} du_1 = dt \\ v_1 = \sin(t) \end{array} \right\}}} + \underbrace{\frac{1}{\pi} \int_{-\pi}^{\pi} \sin(t) \cdot t \, dt \cdot \sin(t)}_{\substack{\left\{ \begin{array}{l} u_2 = t \\ dv_2 = \sin(t) \end{array} \right\} \Rightarrow \left\{ \begin{array}{l} du_2 = dt \\ v_2 = -\cos(t) \end{array} \right\}}} \\
&\quad + \underbrace{\frac{1}{\pi} \int_{-\pi}^{\pi} \cos(2t) \cdot t \, dt \cdot \cos(2t)}_{\substack{\left\{ \begin{array}{l} u_3 = t \\ dv_3 = \cos(2t) \end{array} \right\} \Rightarrow \left\{ \begin{array}{l} du_3 = dt \\ v_3 = \frac{1}{2} \sin(2t) \end{array} \right\}}} + \underbrace{\frac{1}{\pi} \int_{-\pi}^{\pi} \sin(2t) \cdot t \, dt \cdot \sin(2t)}_{\substack{\left\{ \begin{array}{l} u_4 = t \\ dv_4 = \sin(2t) \end{array} \right\} \Rightarrow \left\{ \begin{array}{l} du_4 = dt \\ v_4 = -\frac{1}{2} \cos(2t) \end{array} \right\}}} \\
&= \frac{\cos(t)}{\pi} \left(t \sin(t) - \int_{-\pi}^{\pi} \sin(t) \, dt \right) + \frac{\sin(t)}{\pi} \left(-t \cos(t) + \int_{-\pi}^{\pi} \cos(t) \, dt \right) \\
&\quad + \frac{\cos(2t)}{\pi} \left(\frac{1}{2} t \sin(2t) - \frac{1}{2} \int_{-\pi}^{\pi} \sin(2t) \, dt \right) + \frac{\sin(2t)}{\pi} \left(-\frac{1}{2} t \cos(2t) + \frac{1}{2} \int_{-\pi}^{\pi} \cos(2t) \, dt \right)
\end{aligned}$$

$$\begin{aligned}
&= \frac{\cos(t)}{\pi} (t \sin(t) + \cos(t)) \Big|_{t=-\pi}^{\pi} + \frac{\sin(t)}{\pi} (-t \cos(t) + \sin(t)) \Big|_{t=-\pi}^{\pi} \\
&\quad + \frac{\cos(2t)}{\pi} \left(\frac{1}{2} t \sin(2t) + \frac{1}{4} \cos(w_1) \right) \Big|_{(t,w_1)=(-\pi,-2\pi)}^{(\pi,2\pi)} + \frac{\sin(2t)}{\pi} \left(-\frac{1}{2} t \cos(2t) + \frac{1}{4} \sin(w_2) \right) \Big|_{(t,w_2)=(-\pi,-2\pi)}^{(\pi,2\pi)} \\
&= \frac{\cos(\pi)}{\pi} (\pi \sin(\pi) + \cos(\pi)) - \frac{\cos(-\pi)}{\pi} (-\pi \sin(-\pi) + \cos(-\pi)) \\
&\quad + \frac{\sin(\pi)}{\pi} (-\pi \cos(\pi) + \sin(\pi)) - \frac{\sin(-\pi)}{\pi} (\pi \cos(-\pi) + \sin(-\pi)) \\
&\quad + \frac{\cos(2\pi)}{\pi} \left(\frac{1}{2} \pi \sin(2\pi) + \frac{1}{4} \cos(2\pi) \right) - \frac{\cos(-2\pi)}{\pi} \left(-\frac{1}{2} \pi \sin(-2\pi) + \frac{1}{4} \cos(-2\pi) \right) \\
&\quad + \frac{\sin(2\pi)}{\pi} \left(-\frac{1}{2} \pi \cos(2\pi) + \frac{1}{4} \sin(2\pi) \right) - \frac{\sin(-2\pi)}{\pi} \left(\frac{1}{2} \pi \cos(-2\pi) + \frac{1}{4} \sin(-2\pi) \right) \\
&= -\frac{1}{\pi} (-1) + \frac{1}{\pi} (-1) + \frac{1}{\pi} \left(\frac{1}{4} \right) - \frac{1}{\pi} \left(\frac{1}{4} \right) \\
&= \cancel{\frac{1}{\pi}} - \cancel{\frac{1}{\pi}} + \cancel{\frac{1}{4\pi}} - \cancel{\frac{1}{4\pi}} \\
&= \boxed{0}.
\end{aligned}$$

Exercise 3.9.

Question. Prove that a rotation (2.17) in \mathbb{R}^2 is an orthonormal transformation (with respect to the usual inner product).

Answer.

We can prove the orthonormality of this transformation by showing the following two identities:

$$\begin{aligned}
&\begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix}^T \begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix} \\
&= \begin{pmatrix} \cos^2(\theta) + \sin^2(\theta) & 0 \\ 0 & \cos^2(\theta) + \sin^2(\theta) \end{pmatrix} \\
&= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \\
&= I.
\end{aligned}$$

$$\begin{aligned}
&\begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix} \begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix}^T \\
&= \begin{pmatrix} \cos^2(\theta) + \sin^2(\theta) & 0 \\ 0 & \cos^2(\theta) + \sin^2(\theta) \end{pmatrix} \\
&= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \\
&= I.
\end{aligned}$$

■

Exercise 3.10.

Question. Recall the definition of an orthonormal matrix given in Definition 3.2.14. Assume the usual inner product on \mathbb{F}^n . Prove the following statements:

- (i) The matrix $Q \in M_n(\mathbb{F})$ is an orthonormal matrix if and only if $Q^H Q = Q Q^H = I$.
- (ii) If $Q \in M_n(\mathbb{F})$ is an orthonormal matrix, then $\|Q\mathbf{x}\| = \|\mathbf{x}\|$ for all $\mathbf{x} \in \mathbb{F}^n$.
- (iii) If $Q \in M_n(\mathbb{F})$ is an orthonormal matrix, then so is Q^{-1} .
- (iv) The columns of an orthonormal matrix $Q \in M_n(\mathbb{F})$ are orthonormal.
- (v) If $Q \in M_n(\mathbb{F})$ is an orthonormal matrix, then $|\det(Q)| = 1$. Is the converse true?
- (vi) If $Q_1, Q_2 \in M_n(\mathbb{F})$ are orthonormal matrices, then the product $Q_1 Q_2$ is also an orthonormal matrix.

Answer.

- (i) If Q is orthonormal, then $\langle \mathbf{x}, \mathbf{y} \rangle = \langle Q\mathbf{x}, Q\mathbf{y} \rangle$ for all $\mathbf{x}, \mathbf{y} \in \mathbb{F}^n$. Moreover, if $Q^H Q = Q Q^H = I$, then note that

$$\begin{aligned} \langle Q\mathbf{x}, Q\mathbf{y} \rangle &= (Q\mathbf{x})^H (Q\mathbf{y}) \\ &= \mathbf{x}^H Q^H Q \mathbf{y} \\ &\stackrel{Q^H Q = I}{=} \mathbf{x}^H \mathbf{y}, \end{aligned}$$

again for all $\mathbf{x}, \mathbf{y} \in \mathbb{F}^n$. It follows that (again if $Q^H Q = Q Q^H = I$)

$$\begin{aligned} \langle Q\mathbf{x}, Q\mathbf{y} \rangle &= (Q\mathbf{x})^H (Q\mathbf{y}) \\ &= \mathbf{x}^H Q^H Q \mathbf{y} \\ &= \mathbf{x}^H \mathbf{y} \\ &= \langle \mathbf{x}, \mathbf{y} \rangle. \end{aligned}$$

■

- (ii)

$$\begin{aligned} \|Q\mathbf{x}\| &= \sqrt{\langle Q\mathbf{x}, Q\mathbf{x} \rangle} \\ &= \sqrt{\mathbf{x}^H Q^H Q \mathbf{x}} \\ &= \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle} \\ &= \|\mathbf{x}\|. \end{aligned}$$

■

- (iii) We can observe that if $Q Q^H = Q^H Q = I$, then $Q^{-1} = Q^H$. Moreover, we know Q^H is orthonormal because of the fact that $(Q^H)^H = Q$. Then,

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

■

- (iv) Let \mathbf{v}_i be the i -th column (vector) of Q , which itself is orthonormal. Then, $(Q^H Q)_{ij} = \mathbf{v}_i^H \mathbf{v}_j = \langle \mathbf{v}_i, \mathbf{v}_j \rangle = \delta_{i,j}$ where $\delta_{i,j}$ is the Kronecker delta. Thus, the columns of Q are orthonormal. ■

(v) No. Counterexample:

$$\begin{pmatrix} 2 & 0 \\ 0 & \frac{1}{2} \end{pmatrix}.$$

■

(vi) Consider orthonormal matrices Q and P . Then,

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

and

$$\begin{aligned} QP(QP)^H &= QPP^H Q^H \\ &= QQ^H \\ &= I. \end{aligned}$$

■

Exercise 3.11.

Question. Describe what happens when we apply the Gram-Schmidt orthonormalization process to a collection of linearly *dependent* vectors.

Answer.

Consider a set $\{x_i\}_{i=1}^n$ of linearly dependent vectors in V , for $n \in \mathbb{N}$. Then let $\{x_i\}_{i=1}^{k-1}$ for $k \in (2, N)$ be a linearly *independent* set of vectors, also in V . Then, $\{q_i\}_{i=1}^{k-1}$ is a linearly independent set. But Gram Schmidt does not work in this situation; $q_k = 0$ because $x_k \in \text{span}\left(\{x_i\}_{i=1}^{k-1}\right)$.

Exercise 3.16.

Question. Prove the following results about the QR decomposition:

- (i) The QR decomposition is not unique. *Hint:* Consider matrices of the form QD and $D^{-1}R$, where D is a diagonal matrix.
- (ii) If A is invertible, then there is a unique QR decomposition of A such that R has only positive diagonal elements.

Answer.

- (i) Claim: $-Q$ for the Q in a resultant QR decomposition of $m \times n$ matrix A is orthonormal:

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

and

$$(-Q)^H(-Q) = I.$$

Next, note that $-R$ is clearly upper triangular. Then,

$$\begin{aligned} A &= QR \\ &= (-Q)(-R). \end{aligned}$$

Thus, there are A, Q, R that satisfy both $A = QR$ and $A = (-Q)(-R)$.

- (ii) If A is invertible and can be written as two different QR decompositions (QR and $\hat{Q}\hat{R}$) in which the diagonal entries of R and \hat{R} are strictly positive. This means that R and \hat{R} are both invertible. Thus, $\hat{R}^{-1}R = Q^H\hat{Q}$. Because R and \hat{R} are both upper triangular matrices, we know $\hat{R}^{-1}R$ is upper triangular. Moreover, since Q and \hat{Q} are both orthonormal, then $Q^H\hat{Q}$ is orthonormal. It follows that $\hat{R}^{-1}R = I$ and $R = \hat{R}$ and $Q = \hat{Q}$ (since the matrix inverse is always unique for a given matrix). ■

Exercise 3.17.

Question. Let $A \in M_{m \times n}$ have rank $n \leq m$, and let $A = \hat{Q}\hat{R}$ be a reduced QR decomposition. Prove that solving the system $A^H A \mathbf{x}$

Answer.

Let A have full column rank where $A = \hat{Q}\hat{R}$ is of reduced form; then, \hat{R} has full rank, and so \hat{R} is invertible. Then,

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

■

Exercise 3.23.

Question. Let $(V, \|\cdot\|)$ be a normed linear space. Prove that $|||\mathbf{x}| - \|\mathbf{y}||| \leq \|\mathbf{x} - \mathbf{y}\|$ for all $\mathbf{x}, \mathbf{y} \in V$. *Hint:* Prove $\|\mathbf{x}\| - \|\mathbf{y}\| \leq \|\mathbf{x} - \mathbf{y}\|$ and $\|\mathbf{y}\| - \|\mathbf{x}\| \leq \|\mathbf{x} - \mathbf{y}\|$.

Answer.

Let $\mathbf{x}, \mathbf{y} \in V$. By definition, norms satisfy nonnegativity and the triangle inequality, so note that

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

which implies by homogeneity that

$$\begin{aligned}\|\mathbf{y}\| - \|\mathbf{x}\| &\leq \|\mathbf{y} - \mathbf{x}\| \\ &= \|-(\mathbf{y} - \mathbf{x})\| \\ &= \|\mathbf{x} - \mathbf{y}\|.\end{aligned}$$

■

Exercise 3.24.

Question. Let $\mathcal{C}([a, b]; \mathbb{F})$ be the vector space of all continuous functions from $[a, b] \subset \mathbb{R}$ to \mathbb{F} . Prove that each of the following is a norm on $\mathcal{C}([a, b]; \mathbb{F})$:

- (i) $\|f\|_{L^1} = \int_a^b |f(t)| \, dt$.
- (ii) $\|f\|_{L^2} = \left(\int_a^b |f(t)|^2 \, dt \right)^{1/2}$.
- (iii) $\|f\|_{L^\infty} = \sup_{x \in [a, b]} |f(x)|$.

I will omit this answer.

■

Exercise 3.26.

Question. Two norms $\|\cdot\|_a$ and $\|\cdot\|_b$ on the vector space X are *topologically equivalent* if there exist constants $0 < m \leq M$ such that

$$m \|\mathbf{x}\|_a \leq \|\mathbf{x}\|_b \leq M \|\mathbf{x}\|_a \quad \forall \mathbf{x} \in X.$$

Prove that topological equivalence is an equivalence relation. Then prove that the p -norms for $p = 1, 2, \infty$ on \mathbb{F}^n are topologically equivalent by establishing the following inequalities:

- (i) $\|\mathbf{x}\|_2 \leq \|\mathbf{x}\|_1 \leq \sqrt{n} \|\mathbf{x}\|_2$.
- (ii) $\|\mathbf{x}\|_\infty \leq \|\mathbf{x}\|_2 \leq \sqrt{n} \|\mathbf{x}\|_\infty$.

Hint: Use the Cauchy-Schwarz inequality.

The idea of topological equivalence is especially important in Chapter 5.

Answer.

Let $\|\cdot\|_p$ be a norm on X , for $p \in \{a, b, c\}$. Clearly, $\|\cdot\|_p$ is topologically equivalent to itself for any $m \in (0, 1]$ and $M \geq 1$.

Suppose that $\|\cdot\|_a$ is topologically equivalent to $\|\cdot\|_b$ with constants $0 < m \leq M$. Then, $\|\cdot\|_b$ is topologically equivalent to $\|\cdot\|_a$ with constants $0 < \frac{1}{M'} \leq \frac{1}{m'}$. This leads us to: If $\|\cdot\|_a$ is topologically equivalent to $\|\cdot\|_b$ with constants $0 < m \leq M$, and $\|\cdot\|_b$ is topologically equivalent with $\|\cdot\|_c$ with constants $0 < m' \leq M'$, then $\|\cdot\|_a$ is topologically equivalent to $\|\cdot\|_c$ with constants $0 < mm' \leq MM'$.

Consider arbitrary $\mathbf{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n$. Then, we can show $\|\mathbf{x}\|_2 \leq \|\mathbf{x}\|_1 \leq \sqrt{n} \|\mathbf{x}\|_2$ and $\|\mathbf{x}\|_\infty \leq \|\mathbf{x}\|_2 \leq \sqrt{n} \|\mathbf{x}\|_\infty$ by the following:

$$\begin{aligned} \sum_{i=1}^n |x_i|^2 &\leq \left(\sum_{i=1}^n |x_i|^2 + 2 \sum_{i \neq j} |x_i| |x_j| \right) \\ &= \left(\sum_{i=1}^n |x_i| \right)^2 \end{aligned}$$

$$\begin{aligned} \sum_{i=1}^n |x_i| \cdot 1 &\leq \left(\sum_{i=1}^n |x_i|^2 \right)^{1/2} \left(\sum_{i=1}^n 1^2 \right)^{1/2} \\ &= \sqrt{n} \left(\sum_{i=1}^n |x_i|^2 \right)^{1/2}. \end{aligned}$$

$$\begin{aligned} \max_{i \in [1, n]} |x_i| &= \left(\max_{i \in [1, n]} |x_i|^2 \right)^{1/2} \\ &\leq \left(\sum_{i=1}^n |x_i|^2 \right)^{1/2} \end{aligned}$$

$$\sum_{i=1}^n |x_i|^2 \leq n \cdot \max_{i \in [1, n]} |x_i|^2.$$

■

Exercise 3.28.

Question. Let A be an $n \times n$ matrix. Prove that the operator p -norms are topologically equivalent for $p = 1, 2, \infty$ by establishing the following inequalities:

- (i) $\frac{1}{\sqrt{n}} \|A\|_2 \leq \|A\|_1 \leq \sqrt{n} \|A\|_2$.
- (ii) $\frac{1}{\sqrt{n}} \|A\|_\infty \leq \|A\|_2 \leq \sqrt{n} \|A\|_\infty$.

Answer.

- (i) From above, we know that

$$\begin{aligned} \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_1}{\|\mathbf{x}\|_1} &\leq \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_1}{\|\mathbf{x}\|_1} \\ &\leq \sqrt{n} \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_2}{\|\mathbf{x}\|_2}, \end{aligned}$$

and

$$\begin{aligned} \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_1}{\|\mathbf{x}\|_1} &\geq \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_2}{\|\mathbf{x}\|_1} \\ &\geq \frac{1}{\sqrt{n}} \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_2}{\|\mathbf{x}\|_2}. \end{aligned}$$

Thus, $\frac{1}{\sqrt{n}} \|A\|_2 \leq \|A\|_1 \leq \|A\|_2$. ■

(ii)

$$\begin{aligned} \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_2}{\|\mathbf{x}\|_2} &\leq \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\sqrt{n} \|A\mathbf{x}\|_\infty}{\|\mathbf{x}\|_\infty} \\ \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_2}{\|\mathbf{x}\|_2} &\geq \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|A\mathbf{x}\|_\infty}{\sqrt{n} \|\mathbf{x}\|_\infty}. \end{aligned}$$
■

Exercise 3.29.

Question. Take \mathbb{F}^n with the 2-norm, and let the norm on $M_n(\mathbb{F})$ be the corresponding induced norm. Prove that any orthonormal matrix $Q \in M_n(\mathbb{F})$ has $\|Q\| = 1$. For any $\mathbf{x} \in \mathbb{F}^n$, let $R_{\mathbf{x}} : M_n(\mathbb{F}) \rightarrow \mathbb{F}^n$ be the linear transformation $A \mapsto A\mathbf{x}$. Prove that the induced norm of the transformation $R_{\mathbf{x}}$ is equal to $\|\mathbf{x}\|_2$. *Hint:* First prove $\|R_{\mathbf{x}}\| \leq \|\mathbf{x}\|_2$. Then recall that by Gram-Schmidt, any vector \mathbf{x} with norm $\|\mathbf{x}\|_2 = 1$ is part of an orthonormal basis, and hence is the first column of an orthonormal matrix. Use this to prove equality.

Answer.

Consider arbitrary $\mathbf{x} \neq \mathbf{0}$ and let $\|\cdot\|$ be the norm induced by the inner product. Then

$$\begin{aligned} \|Q\mathbf{x}\| &= (\langle Q\mathbf{x}, Q\mathbf{x} \rangle)^{1/2} \\ &= (\langle Q^H Q \mathbf{x}, \mathbf{x} \rangle)^{1/2} \\ &= (\langle \mathbf{x}, \mathbf{x} \rangle)^{1/2} \\ &= \|\mathbf{x}\|, \end{aligned}$$

so

$$\begin{aligned} \|Q\| &= \sup_{\mathbf{x} \neq \mathbf{0}} \frac{\|Q\mathbf{x}\|}{\|\mathbf{x}\|} \\ &= 1. \end{aligned}$$

Then, for the $R_{\mathbf{x}}$ considered in the question, we have that

$$\begin{aligned} \|R_{\mathbf{x}}\| &= \sup_{A \neq \mathbf{0}} \frac{\|A\mathbf{x}\|}{\|A\|} \\ &= \sup_{A \neq \mathbf{0}} \frac{\|A\mathbf{x}\| \|\mathbf{x}\|}{\|A\| \|\mathbf{x}\|} \\ &\leq \sup_{A \neq \mathbf{0}} \left(\frac{\|A\mathbf{x}\| \|\mathbf{x}\|}{\|A\mathbf{x}\|} \right) \\ &= \|\mathbf{x}\|. \end{aligned}$$
■

Exercise 3.30.

Question. Let $S \in M_n(\mathbb{F})$ be an invertible matrix. Given any matrix norm $\|\cdot\|$ on M_n , define $\|\cdot\|_S$ by $\|A\|_S = \|SAS^{-1}\|$. Prove that $\|\cdot\|_S$ is a matrix norm on M_n .

Answer.

- $\|A\|_S = \|SAS^{-1}\| \geq 0$ for $A \in M_n(\mathbb{F})$ because $\|\cdot\|$ is a norm on $M_n(\mathbb{F})$ and $SAS^{-1} \in M_n(\mathbb{F})$.
- $\|\mathbf{0}\|_S = \|S\mathbf{0}S^{-1}\| = \|\mathbf{0}\| = 0$. If $0 = \|A\|_S = \|SAS^{-1}\|$, then $SAS^{-1} = \mathbf{0}$ implying $A = \mathbf{0}$.
- For arbitrary $a \in \mathbb{F}$,

$$\begin{aligned} \|aA\|_S &= \|SaAS^{-1}\| \\ &= \|aSAS^{-1}\| \\ &= |a| \|SAS^{-1}\| \\ &= |a| \|A\|_S. \end{aligned}$$

- For arbitrary $B \in M_n(\mathbb{F})$,

$$\begin{aligned} \|A + B\|_S &= \|S(A + B)S^{-1}\| \\ &= \|SAS^{-1} + SBS^{-1}\| \\ &\leq \|SAS^{-1}\| + \|SBS^{-1}\| \\ &= \|A\|_S + \|B\|_S. \end{aligned}$$

- So, $\|\cdot\|_S$ is a norm on $M_n(\mathbb{F})$. To show matrix norm:

$$\begin{aligned} \|AB\|_S &= \|SAB S^{-1}\| \\ &= \|SAS^{-1}ABS^{-1}\| \\ &\leq \|SAS^{-1}\| \|SBS^{-1}\|, \end{aligned}$$

$$\text{so } \|AB\|_S \leq \|A\|_S \|B\|_S. \quad \blacksquare$$

Exercise 3.37.

Question. Let $V = \mathbb{R}[x; 2]$ be the space of polynomials of degree at most two, which is a subspace of the inner product space $L^2([0, 1]; \mathbb{R})$. Let $L : V \rightarrow \mathbb{R}$ be the linear functional given by $L[p] = p'(1)$. Find the unique $q \in V$ such that $L[p] = \langle q, p \rangle$, as guaranteed by the Riesz representation theorem. *Hint:* Look at the discussion just before Theorem 3.7.1.

Answer.

Note that $V := \mathbb{R}[x; 2] \cong \mathbb{R}^3$, so $V \ni p = ax^2 + bx + c$ can be represented as a vector in \mathbb{R}^3 . Let $p = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$.

Take $q = \boxed{(2, 1, 0)}$ where $p'q = 2a + b = p'(1) = L[p]$.

Exercise 3.38.

Question. Let $V = \mathbb{F}[x; 2]$, which is a subspace of the inner product space $L^2([0, 1]; \mathbb{R})$. Let D be the derivative operator $D : V \rightarrow V$; that is, $D[p](x) = p'(x)$. Write the matrix representation of D with respect to the power basis $[1, x, x^2]$ of $\mathbb{F}[x; 2]$. Write the matrix representation of the adjoint of D with respect to this basis.

Answer.

Let $p = ax^2 + vx + c$ be an arbitrary element of $V = \mathbb{F}[x; 2]$. Because $p = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$ and $p' = D(p) = \begin{pmatrix} 0 \\ 2a \\ b \end{pmatrix}$, we know the matrix representation of D is

$$D = \begin{pmatrix} 0 & 0 & 0 \\ 2 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$$

and

$$D^H = D^T = \begin{pmatrix} 0 & 2 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}.$$

Exercise 3.39.

Question. Prove Proposition 3.7.12.

Proposition 3.7.12. Let V and W be fin-dim inner product spaces. The adjoint has the following properties:

- (i) If $S, T \in \mathcal{L}(V; W)$, then $(S + T)^* = S^* + T^*$ and $(\alpha T)^* = \bar{\alpha}T^*$, $\alpha \in \mathbb{F}$.
- (ii) If $S \in \mathcal{L}(V; W)$, then $(S^*)^* = S$.
- (iii) If $S, T \in \mathcal{L}(V)$, then $(ST)^* = T^*S^*$.
- (iv) If $T \in \mathcal{L}(V)$ and T is invertible, then $(T^*)^{-1} = (T^{-1})^*$.

Answer.

(i)

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

From above, we have that $(S + T)^* = S^* + T^*$. Also,

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

■

(ii)

$$\begin{aligned}
 \langle \mathbf{w}, S \mathbf{v} \rangle_W &= \langle S^* \mathbf{w}, \mathbf{v} \rangle_V \\
 &= \overline{\langle \mathbf{v}, S^* \mathbf{w} \rangle_V} \\
 &= \overline{\langle S^{**} \mathbf{v}, \mathbf{w} \rangle_W} \\
 &= \langle \mathbf{w}, S^{**} \mathbf{v} \rangle_W
 \end{aligned}$$

for $\mathbf{v} \in V$ and $\mathbf{w} \in W$, so it follows that $S = S^{**}$.

■

(iii)

$$\begin{aligned}
 \langle (ST)^* \mathbf{v}', \mathbf{v} \rangle_V &= \langle \mathbf{v}', (ST) \mathbf{v} \rangle_V \\
 &= \langle \mathbf{v}', S(T \mathbf{v}) \rangle_V \\
 &= \langle S^* \mathbf{v}', T \mathbf{v} \rangle_V \\
 &= \langle T^* S^* \mathbf{v}', \mathbf{v} \rangle_V,
 \end{aligned}$$

■

(iv) From (iii), we have that

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

■

Exercise 3.40.

Question. Let $M_n(\mathbb{F})$ be endowed with the Frobenius inner product (see Example 3.1.7). Any $A \in M_n(\mathbb{F})$ defines a linear operator on $M_n(\mathbb{F})$ by left multiplication: $B \mapsto AB$.

(i) Show that $A^* = A^H$.

(ii) Show that for any $A_1, A_2, A_3 \in M_n(\mathbb{F})$ we have $\langle A_2, A_3 A_1 \rangle = \langle A_2 A_1^*, A_3 \rangle$. *Hint:* Recall $\text{tr}(AB) = \text{tr}(BA)$.

(iii) Let $A \in M_n(\mathbb{F})$. Define the linear operator $T_A : M_n(\mathbb{F}) \rightarrow M_n(\mathbb{F})$ by $T_A(X) = AX - XA$, and show that $(T_A)^* = T_{A^*}$.

Answer.

(i) Consider arbitrary $B, C \in M_n(\mathbb{F})$. Then, by definition of Frobenius inner product, we have:

$$\begin{aligned}\langle B, AC \rangle_F &= \text{tr}(B^H AC) \\ &= \text{tr}((A^H B)^H C) \\ &= \langle A^H B, C \rangle_F.\end{aligned}$$

■

(ii)

$$\begin{aligned}\langle A_2, A_3 A_1 \rangle_F &= \text{tr}(A_2^H A_3 A_1) \\ &= \text{tr}(A_1 A_2^H A_3) \\ &= \text{tr}((A_2 A_1^H)^H A_3) \\ &= \langle A_2 A_1^H, A_3 \rangle_F \\ &= \langle A_2 A_1^*, A_3 \rangle.\end{aligned}$$

■

(iii) For arbitrary $B, C \in M_n(\mathbb{F})$, we have $\langle B, AC - CA \rangle = \langle B, AC \rangle - \langle B, CA \rangle$. Then, we showed above, we then have $\langle B, CA \rangle = \langle BA^*, C \rangle$. Moreover,

$$\begin{aligned}\langle B, AC \rangle &= \text{tr}(B^H AC) \\ &= \text{tr}((A^H B)^H C) \\ &= \langle A^H B, C \rangle \\ &= \langle A^* B, C \rangle.\end{aligned}$$

Thus, it follows that $T_A^* = T_A$. ■

Exercise 3.44.

Question. Given $A \in M_{m \times n}(\mathbb{F})$ and $\mathbf{b} \in \mathbb{F}^m$, prove the *Fredholm alternative*: Either $A\mathbf{x} = \mathbf{b}$ has a solution $\mathbf{x} \in \mathbb{F}^n$ or there exists $\mathbf{y} \in \mathcal{N}(A^H)$ such that $\langle \mathbf{y}, \mathbf{b} \rangle \neq 0$.

Answer.

- If $\exists \mathbf{x} \in \mathbb{F}^n : A\mathbf{x} = \mathbf{b}$ then $\forall \mathbf{y} \in \mathcal{N}(A^H)$, so we have:

$$\begin{aligned}\langle \mathbf{y}, \mathbf{b} \rangle &= \langle \mathbf{y}, A\mathbf{x} \rangle \\ &= \langle A^H \mathbf{y}, \mathbf{x} \rangle \\ &= \langle \mathbf{0}, \mathbf{x} \rangle \\ &= 0.\end{aligned}$$

- If $\exists \mathbf{y} \in \mathcal{N}(A^H) : \langle \mathbf{y}, \mathbf{b} \rangle \neq 0$ then $\mathbf{b} \notin \mathcal{N}(A^H)^\perp = \mathcal{R}(A)$.
- Therefore, $\exists \mathbf{x} \in \mathbb{F}^n : A\mathbf{x} = \mathbf{b}$.

■

Exercise 3.45.

Question. Consider the vector space $M_n(\mathbb{R})$ with the Frobenius inner product (3.5). Show that $\text{Sym}_n(\mathbb{R})^\perp = \text{Skew}_n(\mathbb{R})$. (See Exercise 1.18 for the definition of Sym and Skew.)

Answer.

- Let $A \in \text{Sym}_n(\mathbb{R})$ and $B \in \text{Skew}_n(\mathbb{R})$. Then,

$$\begin{aligned}\langle B, A \rangle &= \text{tr}(B^\top A) \\ &= \text{tr}(AB^\top) \\ &= \text{tr}(A^\top(-B)) \\ &= -\langle A, B \rangle.\end{aligned}$$

(It follows that $\langle A, B \rangle = 0$ and $\text{Skew}_n(\mathbb{R}) \subset \text{Sym}_n(\mathbb{R})^\perp$.)

- Let $B \in \text{Sym}_n(\mathbb{R})^\perp$, so $B + B^\top \in \text{Sym}_n(\mathbb{R})$. Thus,

$$\begin{aligned}0 &= \langle B + B^\top, B \rangle \\ &= \text{tr}((B + B^\top)B) \\ &= \text{tr}(BB + B^\top B) \\ &= \text{tr}(BB) + \text{tr}(B^\top B),\end{aligned}$$

so $\langle B^\top, B \rangle = \langle -B, B \rangle$ so $B^\top = -B$. Thus $\text{Sym}_n(\mathbb{R})^\perp = \text{Skew}_n(\mathbb{R})$. ■

Exercise 3.46.

Question. Prove the following for an $m \times n$ matrix A :

- (i) If $\mathbf{x} \in \mathcal{N}(A^\mathsf{H}A)$, then $A\mathbf{x}$ is in both $\mathcal{R}(A)$ and $\mathcal{N}(A^\mathsf{H})$.
- (ii) $\mathcal{N}(A^\mathsf{H}A) = \mathcal{N}(A)$.
- (iii) A and $A^\mathsf{H}A$ have the same rank.
- (iv) If A has linearly independent columns, then $A^\mathsf{H}A$ is nonsingular.

Answer.

- (i) If $\mathbf{x} \in \mathcal{N}(A^\mathsf{H}A)$, then $\mathbf{0} = (A^\mathsf{H}A)\mathbf{x} = A^\mathsf{H}(A\mathbf{x})$ and $A\mathbf{x} \in \mathcal{N}(A^\mathsf{H})$. $A\mathbf{x} \in \text{im}(A)$ by definition. ■
- (ii) If $\mathbf{x} \in \mathcal{N}(A)$, then $A\mathbf{x} = \mathbf{0}$. Then, $A^\mathsf{H}A\mathbf{x} = A^\mathsf{H}\mathbf{0} = \mathbf{0}$, so $\mathbf{x} \in \mathcal{N}(A^\mathsf{H}A)$. If $\mathbf{x} \in \mathcal{N}(A^\mathsf{H}A)$, then $\|A\mathbf{x}\|^2 = \mathbf{x}^\mathsf{H}A^\mathsf{H}A\mathbf{x} = \mathbf{x}^\mathsf{H}\mathbf{0} = 0$, so $A\mathbf{x} = \mathbf{0}$ and $\mathbf{x} \in \mathcal{N}(A)$. ■
- (iii) $n = \text{rank}(A) + \dim(\mathcal{N}(A))$ by rank-nullity, and $n = \text{rank}(A^\mathsf{H}A) + \dim(\mathcal{N}(A^\mathsf{H}A))$. Then, by (ii), it follows that $\text{rank}(A) = \text{rank}(A^\mathsf{H}A)$. ■
- (iv) By (iii), we have that $n = \text{rank}(A) = \text{rank}(A^\mathsf{H}A)$. We know it is invertible by $A^\mathsf{H}A \in M_n$. ■

Exercise 3.47.

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

- (i) $P^2 = P$.
- (ii) $P^H = P$.
- (iii) $\text{rank}(P) = n$.

Whenever a linear operator satisfies $P^2 = P$, it is called a *projection*. Projections are treated in detail in Section 12.1.

Answer.

(i)

$$\begin{aligned}
 P^2 &= (A(A^H A)^{-1} A^H)(A(A^H A)^{-1} A^H) \\
 &= A(A^H A)^{-1} A^H A(A^H A)^{-1} A^H \\
 &= A(A^H A)^{-1} A^H \\
 &= P.
 \end{aligned}$$

■

(ii)

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

■

(iii) $\text{rank}(A) = n \implies \text{rank}(P) \leq n$. Thus, $\forall \mathbf{y} \in \text{im}(A) : \exists \mathbf{x} \in \mathbb{F}^n : \mathbf{y} = A\mathbf{x}$. So,

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

so $\mathbf{y} \in \text{im}(P)$. Thus, $\text{rank}(P) \geq \text{rank}(A)$, so $\text{rank}(P) = n$.

■

Exercise 3.48.

Question. Consider the vector space $M_n(\mathbb{R})$ with the Frobenius inner product (3.5). Let $P(A) = \frac{A+A^T}{2}$ be the map $P : M_n(\mathbb{R}) \rightarrow M_n(\mathbb{R})$. Prove the following:

- (i) P is linear.
- (ii) $P^2 = P$.
- (iii) $P^* = P$ (note that $*$ here means the adjoint with respect to the Frobenius inner product).
- (iv) $\mathcal{N}(P) = \text{Skew}_n(\mathbb{R})$.
- (v) $\mathcal{R}(P) = \text{Sym}_n(\mathbb{R})$.
- (vi) $\|A - P(A)\|_F = \sqrt{\frac{\text{tr}(A^T A) - \text{tr}(A^2)}{2}}$. Here $\|\cdot\|_F$ is the norm with respect to the Frobenius inner product.

Hint: Recall that $\text{tr}(AB) = \text{tr}(BA)$ and $\text{tr}(A) = \text{tr}(A^T)$.

I will omit this answer. ■

Exercise 3.50.

Question. Let $(x_i, y_i)_{i=1}^n$ be a collection of data points that we have reason to believe should lie (roughly) on an ellipse of the form $rx^2 + sy^2 = 1$. We wish to find the least squares approximation for r and s . Write A , \mathbf{x} , and \mathbf{b} for the corresponding normal equation in terms of the data x_i and y_i and the unknowns r and s .

Answer.

Consider the regression $\mathbf{y}^2 = \frac{1}{s} + \frac{r\mathbf{x}^2}{s}$ in the form $A\mathbf{x} = \mathbf{b}$. Let $b_i := y_i^2$ for each component $b_i, y_i \in \mathbf{b}, \mathbf{y}$, and let $\mathbf{x} = \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix}$ where $\beta_1 = \frac{1}{s}$ and $\beta_2 = \frac{r}{s}$. Then the corresponding normal equation is $A^H A \hat{\mathbf{x}} = A^H \mathbf{b}$, where

$$\begin{aligned} A^H A \hat{\mathbf{x}} &= \begin{pmatrix} \sum_i 1 & \sum_i x_i^2 \\ \sum_i x_i^2 & \sum_i x_i^4 \end{pmatrix} \begin{pmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{pmatrix} \\ &= \begin{pmatrix} n\hat{\beta}_1 - \hat{\beta}_2 \sum_i x_i^2 \\ \hat{\beta}_1 \sum_i x_i^2 - \hat{\beta}_2 \sum_i x_i^4 \end{pmatrix} \end{aligned}$$

and

$$A^H \mathbf{b} = \begin{pmatrix} \sum_i y_i^2 \\ \sum_i x_i^2 y_i^2 \end{pmatrix}.$$