

Problem Set 2

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Exercise 3.1

Part i)

We can write

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

as $\langle \mathbf{y}, \mathbf{x} \rangle = \overline{\langle \mathbf{x}, \mathbf{y} \rangle} = \langle \mathbf{x}, \mathbf{y} \rangle$ in \mathbb{R}^n .

Using $\langle \mathbf{x}, -\mathbf{y} \rangle = (-1)\langle \mathbf{x}, \mathbf{y} \rangle$, we can write

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

Combining,

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

Part ii)

Combining equations 1 and 2, we find

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

Exercise 3.2

$$\begin{aligned}
\langle \mathbf{x}, \mathbf{y} \rangle &= \frac{1}{4}(\|\mathbf{x} + \mathbf{y}\|^2 - \|\mathbf{x} - \mathbf{y}\|^2 + i\|\mathbf{x} - i\mathbf{y}\|^2 - i\|\mathbf{x} + i\mathbf{y}\|^2) \\
&= \frac{1}{4}(\|\mathbf{x} + \mathbf{y}\|^2 - \|\mathbf{x} - \mathbf{y}\|^2) + \frac{1}{4}(i\|\mathbf{x} - i\mathbf{y}\|^2 - i\|\mathbf{x} + i\mathbf{y}\|^2) \\
&= \langle \mathbf{x}, \mathbf{y} \rangle + \frac{1}{4}(i\|\mathbf{x} - i\mathbf{y}\|^2 - i\|\mathbf{x} + i\mathbf{y}\|^2) \\
&= \langle \mathbf{x}, \mathbf{y} \rangle + \frac{1}{4}(i\langle \mathbf{x} - i\mathbf{y}, \mathbf{x} - i\mathbf{y} \rangle - i\langle \mathbf{x} + i\mathbf{y}, \mathbf{x} + i\mathbf{y} \rangle) \\
&= \langle \mathbf{x}, \mathbf{y} \rangle + \frac{1}{4}(i\langle \mathbf{x}, \mathbf{x} - i\mathbf{y} \rangle + i^2\langle \mathbf{y}, \mathbf{x} - i\mathbf{y} \rangle - i\langle \mathbf{x}, \mathbf{x} + i\mathbf{y} \rangle + (-i)^2\langle \mathbf{y}, \mathbf{x} + i\mathbf{y} \rangle) \\
&= \langle \mathbf{x}, \mathbf{y} \rangle + \frac{1}{4}(i\langle \mathbf{x}, \mathbf{x} \rangle + i^2\langle \mathbf{x}, \mathbf{y} \rangle + i^2\langle \mathbf{y}, \mathbf{x} \rangle + i^3\langle \mathbf{y}, \mathbf{y} \rangle \\
&\quad - i\langle \mathbf{x}, \mathbf{x} \rangle + (-i)^2\langle \mathbf{x}, \mathbf{y} \rangle + (-i)^2\langle \mathbf{y}, \mathbf{x} \rangle + (-i^3)\langle \mathbf{y}, \mathbf{y} \rangle) \\
&= \langle \mathbf{x}, \mathbf{y} \rangle
\end{aligned}$$

Exercise 3.3

Part i)

$$\begin{aligned}
\langle \mathbf{x}, \mathbf{x}^5 \rangle &= \int_0^1 x^6 dx = \left. \frac{x^7}{7} \right|_0^1 = \frac{1}{7} \\
\langle \mathbf{x}, \mathbf{x} \rangle &= \int_0^1 x^2 dx = \left. \frac{x^3}{3} \right|_0^1 = \frac{1}{3} \implies \|\mathbf{x}\| = \sqrt{\frac{1}{3}} \\
\langle \mathbf{x}^5, \mathbf{x}^5 \rangle &= \int_0^1 x^{10} dx = \left. \frac{x^{11}}{11} \right|_0^1 = \frac{1}{11} \implies \|\mathbf{x}^5\| = \sqrt{\frac{1}{11}}
\end{aligned}$$

$$\cos(\theta) = \frac{\sqrt{3}\sqrt{11}}{7} = \frac{\sqrt{33}}{7} \implies \theta \approx 35^\circ$$

Part ii)

$$\begin{aligned}
\langle \mathbf{x}^2, \mathbf{x}^4 \rangle &= \int_0^1 x^6 dx = \left. \frac{x^7}{7} \right|_0^1 = \frac{1}{7} \\
\langle \mathbf{x}^2, \mathbf{x}^2 \rangle &= \int_0^1 x^4 dx = \left. \frac{x^5}{5} \right|_0^1 = \frac{1}{5} \implies \|\mathbf{x}^2\| = \sqrt{\frac{1}{5}} \\
\langle \mathbf{x}^4, \mathbf{x}^4 \rangle &= \int_0^1 x^8 dx = \left. \frac{x^9}{9} \right|_0^1 = \frac{1}{9} \implies \|\mathbf{x}^4\| = \sqrt{\frac{1}{9}}
\end{aligned}$$

$$\cos(\theta) = \frac{\sqrt{9}\sqrt{5}}{7} = \frac{\sqrt{45}}{7} \implies \theta \approx 17^\circ$$

Exercise 3.8

i)

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

Therefore, S is an orthonormal set.

$$\text{ii) } \|t\| = \sqrt{\langle t, t \rangle} = \int_{-\pi}^{\pi} t^2 dt = \left. \frac{t^3}{3} \right|_{-\pi}^{\pi} = \frac{2}{3} \pi^2$$

$$\text{iii) } \text{proj}_X(\cos(3t)) = \sum_{i=1}^m \langle \mathbf{x}_i, \cos(3t) \rangle \mathbf{x}_i = 0$$

$$\text{iv) } \text{proj}_X(t) = \sum_{i=1}^m \langle \mathbf{x}_i, t \rangle \mathbf{x}_i = 1$$

Exercise 3.9

By Theorem 3.2.15, a matrix Q is orthonormal if and only if $Q^H Q = Q Q^H = 1$.

The rotation matrix is given by

$$R_{\theta} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

Calculating $R_{\theta} R$, we find

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

So R_{θ} is an orthonormal transformation.

Exercise 3.10

Part i)

Assume Q is orthonormal, which implies $\langle \mathbf{e}_i, \mathbf{e}_j \rangle = \langle Q\mathbf{e}_i, Q\mathbf{e}_j \rangle$.

$$\begin{aligned}\mathbf{e}_i^H \mathbf{e}_j &= \langle \mathbf{e}_i, \mathbf{e}_j \rangle \\ &= \langle Q\mathbf{e}_i, Q\mathbf{e}_j \rangle \\ &= (Q\mathbf{e}_i)^H (Q\mathbf{e}_j) \\ &= \mathbf{e}_i^H Q^H Q \mathbf{e}_j\end{aligned}$$

$\mathbf{e}_i^H \mathbf{e}_j = \mathbf{e}_i^H Q^H Q \mathbf{e}_j$ only if $Q^H Q = I$.

Part ii)

$$\begin{aligned}\|Q\mathbf{x}\| &= \sqrt{\langle Q\mathbf{x}, Q\mathbf{x} \rangle} \\ &= \sqrt{\mathbf{x}^H Q^H Q \mathbf{x}} \\ &= \sqrt{\mathbf{x}^H \mathbf{x}} \quad \text{as } Q^H Q = I \\ &= \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle} \\ &= \|\mathbf{x}\|\end{aligned}$$

Part iii)

$Q^H Q = I$ so $Q^H = Q^{-1}$.

$$\begin{aligned}\langle Q^{-1}\mathbf{x}, Q^{-1}\mathbf{x} \rangle &= \mathbf{x}^H (Q^{-1})^H Q^{-1} \mathbf{x} \\ &= \mathbf{x}^H (Q^H)^H Q^H \mathbf{x} \quad \text{as } Q^H = Q^{-1} \\ &= \mathbf{x}^H Q Q^H \mathbf{x} \\ &= \mathbf{x}^H \mathbf{x} \quad \text{as } Q Q^H = I \\ &= \langle \mathbf{x}, \mathbf{x} \rangle\end{aligned}$$

So Q^{-1} is orthonormal.

Part iv)

$Q_{n,m}$ is an orthonormal matrix. We can write Q as

$$Q = \begin{bmatrix} \mathbf{q}_1 & \mathbf{q}_2 & \cdots & \mathbf{q}_m \end{bmatrix}$$

Where \mathbf{q}_i is a $(n \times 1)$ column of Q , and $\{\mathbf{q}_i\}_{i=1}^m$ is a set of columns vectors.

As Q is orthonormal, we know $Q^H Q = I_m$. So we can write

$$\begin{aligned} Q^H Q &= \begin{bmatrix} \mathbf{q}_1 & \mathbf{q}_2 & \cdots & \mathbf{q}_m \end{bmatrix}^H \begin{bmatrix} \mathbf{q}_1 & \mathbf{q}_2 & \cdots & \mathbf{q}_m \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{q}_1^H \\ \mathbf{q}_2^H \\ \vdots \\ \mathbf{q}_m^H \end{bmatrix} \begin{bmatrix} \mathbf{q}_1 & \mathbf{q}_2 & \cdots & \mathbf{q}_m \end{bmatrix} \\ &= \begin{bmatrix} \mathbf{q}_1^H \mathbf{q}_1 & \mathbf{q}_2^H \mathbf{q}_1 & \cdots & \mathbf{q}_m^H \mathbf{q}_1 \\ \mathbf{q}_1^H \mathbf{q}_2 & \mathbf{q}_2^H \mathbf{q}_2 & \cdots & \vdots \\ \vdots & \vdots & \ddots & \\ \mathbf{q}_1^H \mathbf{q}_m & \cdots & & \mathbf{q}_m^H \mathbf{q}_m \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & \vdots \\ \vdots & \vdots & \ddots & \\ 0 & \cdots & & 1 \end{bmatrix} \end{aligned}$$

which satisfies the definition of an orthonormal set, ie. $\langle \mathbf{q}_i, \mathbf{q}_j \rangle = \delta_{i,j}$ where

$$\delta_{i,j} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases}$$

Part v)

It is not true that $|\det(Q)| = 1$ implies Q is orthonormal. For example

$$Q = \begin{bmatrix} 2 & \sqrt{3} \\ \sqrt{3} & 2 \end{bmatrix}$$

has $|\det(Q)| = 1$ but

$$Q^H Q = \begin{bmatrix} 2 & \sqrt{3} \\ \sqrt{3} & 2 \end{bmatrix} \begin{bmatrix} 2 & \sqrt{3} \\ \sqrt{3} & 2 \end{bmatrix} = \begin{bmatrix} 1 & 2\sqrt{3} - 3\sqrt{2} \\ 3\sqrt{2} - 2\sqrt{3} & -1 \end{bmatrix} \neq I$$

So Q is not orthonormal.

Part vi)

Q_1 and Q_2 are orthonormal, so $Q_1^H Q_1 = I$ and $Q_2^H Q_2 = I$ Given $Q = Q_1 Q_2$

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

Q is therefore orthonormal.

Exercise 3.11

If the Gram–Schmidt orthonormalisation process is applied to a collection of linearly dependent vectors, it outputs the $\mathbf{0}$ vector for some $\mathbf{x}_k - \mathbf{p}_{k-1}$, and therefore \mathbf{q}_k is undefined.

Exercise 3.16

Part i)

Consider a diagonal matrix D with elements on the diagonal equal to -1 . We can then write $Q^* = QD$ and $R^* = D^{-1}R$, and deduce that $A = Q^* R^* = QDD^{-1}R = QR$, showing that the QR decomposition is not unique.

Part ii)

Suppose $A = Q_1 R_1$ and $A = Q_2 R_2$, , where R_1, R_2 are upper triangular matrices with positive elements along the diagonal.

Then $Q_1 R_1 = Q_2 R_2 \implies Q_2^{-1} Q_1 = R_1^{-1} R_2$, which means that $R_1^{-1} R_2$ is orthonormal. As $R_1^{-1} R_2$ is an upper triangular matrix that is orthonormal, its diagonal elements must be equal to 1 (as the diagonal of R is positive). So $Q_2^{-1} Q_1 = I$ and $R_1^{-1} R_2 = I$, and thus $Q_1 = Q_2$ and $R_1 = R_2$, showing that the decomposition is unique in this case.

Exercise 3.17

$$\begin{aligned}
A^H A \mathbf{x} &= (\hat{Q} \hat{R})^H \hat{Q} \hat{R} \mathbf{x} \\
&= \hat{R}^H \hat{Q}^H \hat{Q} \hat{R} \mathbf{x} \\
&= \hat{R}^H \hat{R} \mathbf{x} \quad \text{as } \hat{Q}^H \hat{Q} = 1 \\
\hat{R}^H \hat{R} \mathbf{x} &= \hat{R}^H \hat{Q}^H \mathbf{b} \quad \text{as } A^H = \hat{R}^H \hat{Q}^H \\
\hat{R} \mathbf{x} &= \hat{Q}^H \mathbf{b} \quad \text{as required}
\end{aligned}$$

Exercise 3.23

By the triangle inequality,

$$\|\mathbf{x}\| = \|(\mathbf{x} - \mathbf{y}) + \mathbf{y}\| \leq \|\mathbf{x} - \mathbf{y}\| + \|\mathbf{y}\| \Leftrightarrow \|\mathbf{x}\| - \|\mathbf{y}\| \leq \|\mathbf{x} - \mathbf{y}\|$$

By scale preservation,

$$\|\mathbf{y}\| = \|-1\| \cdot \|\mathbf{y}\| = \|\mathbf{-y}\| = \|\mathbf{x} - \mathbf{y} + \mathbf{x}\| \leq \|\mathbf{x} - \mathbf{y}\| + \|\mathbf{y}\| \Leftrightarrow \|\mathbf{y}\| - \|\mathbf{x}\| \leq \|\mathbf{x} - \mathbf{y}\|$$

Exercise 3.24

Part i)

1. Positivity:

Consider $f, g \in C[a, b]$, with $f(t) \geq g(t)$ for all $t \in [a, b]$, then

$$\int_a^b f(t) dt \geq \int_a^b g(t) dt$$

With $g(t) = 0$, and $|f(t)| \geq 0$

$$\int_a^b f(t) dt \geq \int_a^b 0 dt = 0$$

2. Scale preservation:

Given $h \in \mathbb{F}$ and $f \in C[a, b]$,

$$\|hf\| = \int_a^b |hf(t)|dt = \int_a^b |h||f(t)|dt = |h| \int_a^b |f(t)|dt = |h|\|f\|$$

3. Triangle inequality:

Consider $f, g \in C[a, b]$.

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

Part ii)

Results follow from part i).

Part iii)

1. Positivity:

$|f(x)| \geq 0$ for all x , so it follows that $\sup_{x \in [a, b]} |f(x)| \geq 0$.

2. Scale preservation:

$$\|hf\| = \sup_{x \in [a, b]} |h||f(x)| = |h| \sup_{x \in [a, b]} |f(x)| = |h|\|f\|$$

3. Triangle inequality:

For any $x \in [a, b]$

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

As this holds for all $x \in [a, b]$,

$$\sup_{x \in [a, b]} |f(x) + g(x)| = \|f + g\| \leq \|f\| + \|g\|$$

Exercise 3.26

$$\underline{\|x\|_a \sim \|x\|_a}$$

$$\begin{aligned} m\|x\|_a &\leq \|x\|_a \leq M\|x\|_a \\ \implies m &\leq 1 \leq M \end{aligned}$$

with $m \in [0, 1]$, and $M \in [1, \infty]$.

$$\underline{\|x\|_a \sim \|x\|_b \implies \|x\|_b \sim \|x\|_a}$$

$$\begin{aligned} m\|x\|_a &\leq \|x\|_b \leq M\|x\|_a \\ \implies \frac{1}{m} &\leq \frac{\|x\|_b}{\|x\|_a} \leq \frac{1}{M} \\ \implies \frac{1}{M} &\geq \frac{\|x\|_a}{\|x\|_b} \geq \frac{1}{m} \\ \implies Y &\geq \frac{\|x\|_a}{\|x\|_b} \geq y \\ \implies y\|x\|_b &\leq \|x\|_a \leq Y\|x\|_b \end{aligned}$$

where $0 \leq y \leq Y$.

$$\underline{\|x\|_a \sim \|x\|_b \text{ and } \|x\|_b \sim \|x\|_c \implies \|x\|_a \sim \|x\|_c}$$

$$m\|x\|_a \leq \|x\|_b \leq M\|x\|_a \tag{3}$$

$$w\|x\|_b \leq \|x\|_c \leq W\|x\|_b \tag{4}$$

Multiply (3) by w , to get

$$mw\|x\|_a \leq w\|x\|_b \leq wM\|x\|_a \tag{5}$$

Combining (5) with (4), we see

$$mw\|\mathbf{x}\|_a \leq \|\mathbf{x}\|_c \leq W\|\mathbf{x}\|_b$$

Similarly,

$$mw\|\mathbf{x}\|_a \leq \|\mathbf{x}\|_c \leq MW\|\mathbf{x}\|_a$$

where $0 \leq mw \leq MW$, as required.

Part i)

$$\begin{aligned} \|\mathbf{x}\|_2 &= (|\mathbf{x}_1|^2 + |\mathbf{x}_2|^2 + \cdots + |\mathbf{x}_n|^2)^{1/2} \\ &\leq |\mathbf{x}_1| + |\mathbf{x}_2| + \cdots + |\mathbf{x}_n| \quad \text{by Cauchy-Schwarz} \\ &= \|\mathbf{x}\|_1 \end{aligned}$$

Exercise 3.29

The induced norm on $M_n(\mathbb{F})$ is given by

$$\|A\|_2 = \sup \frac{\|A\mathbf{x}\|_2}{\|\mathbf{x}\|_2}$$

Given an orthonormal matrix Q ,

$$\begin{aligned} \|Q\|_2 &= \sup \frac{\|Q\mathbf{x}\|_2}{\|\mathbf{x}\|_2} \\ &= \sup \frac{\|\mathbf{x}\|_2}{\|\mathbf{x}\|_2} \\ &= 1 \end{aligned}$$

Given a matrix norm, we know $\|A\mathbf{x}\| \leq \|A\|\|\mathbf{x}\|$. So

$$\begin{aligned}\|R_{\mathbf{x}}\| &= \sup \frac{\|A\mathbf{x}\|}{\|\mathbf{x}\|} \\ &= \sup \frac{\|A\mathbf{x}\|\|\mathbf{x}\|}{\|\mathbf{x}\|\|\mathbf{x}\|} \\ &\leq \sup \frac{\|A\mathbf{x}\|\|\mathbf{x}\|}{\|A\mathbf{x}\|} \\ &= \|\mathbf{x}\|\end{aligned}$$

Consider \mathbf{x} such that $\|\mathbf{x}\| = 1$. If $\|R_{\mathbf{x}}\| < \|\mathbf{x}\| = 1$, it must be that $\|A\|\|\mathbf{x}\| < \|A\mathbf{x}\|$. This can be written as

$$\|A\| < \frac{\|A\mathbf{x}\|}{\|\mathbf{x}\|}$$

Exercise 3.47

$$P = A(A^H A)^{-1} A^H.$$

- i. $P^2 = A(A^H A)^{-1} A^H A(A^H A)^{-1} A^H$
- ii. $P^H = (A(A^H A)^{-1} A^H)^H = A((A^H A)^{-1})^H A^H = A(A^H A)^H)^{-1} A^H = A(A^H A)^{-1} A^H = P$
- iii. P is an idempotent matrix by i), which means $\text{rank}(P) = \text{tr}(P)$.

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