

Math Problem Set #2

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3.1

(i)

$$\langle x, y \rangle = \|x\| \|y\| \cos \theta$$

Law of cosines:

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

Subtracting and substituting gives:

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

(ii)

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

3.2

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

3.3

(i)

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

(ii)

$$\begin{aligned}\langle x^2, x^4 \rangle &= \int_0^1 x^2 x^4 dx = \frac{1}{7} \\ \|x^2\|^2 &= \int_0^1 x^4 dx = \frac{1}{5} \\ \|x^4\|^2 &= \int_0^1 x^8 dx = \frac{1}{9} \\ \implies \cos \theta &= \frac{\frac{1}{7}}{\frac{1}{5} \cdot \frac{1}{3}}\end{aligned}$$

3.8

(i) For a set to be orthonormal, the inner product of any distinct pair of elements in the set must be equal to zero (orthogonality), and the norm of any element must be 1 (normality):

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

For norms,

$$\begin{aligned}
 \langle \cos(t), \cos(t) \rangle &= \|\cos(t)\|^2 \\
 &= \|\sin(t)\|^2 \\
 &= \|\cos(2t)\|^2 \\
 &= \|\sin(2t)\|^2 \\
 &= 1
 \end{aligned}$$

(ii)

(iii)

(iv)

3.9

A matrix R is an orthonormal if and only if it satisfies $R^H R = I$:

$$\begin{aligned}
 R(\theta) &= \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \\
 R^H(\theta) &= \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} \\
 R^H R(\theta) &= \begin{bmatrix} \cos^2 \theta + \sin^2 \theta & \cos \theta \sin \theta - \cos \theta \sin \theta \\ \cos \theta \sin \theta - \cos \theta \sin \theta & \cos^2 \theta + \sin^2 \theta \end{bmatrix} \\
 &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I_2
 \end{aligned}$$

\implies the rotation matrix in \mathbb{R}^2 is an orthonormal transformation.

3.10

(i) Orthonormal transformations preserve inner products, i.e. $\langle x, y \rangle = \langle Qx, Qy \rangle$, where $x, y \in \mathbb{F}^n$ and $Q \in M_n(\mathbb{F})$. Thus, $\langle x, y \rangle = \langle Qx, Qy \rangle = \langle Q^H Qx, Qy \rangle \implies x = Q^H Qx \implies Q^H Q = I$.

(ii) $\|Qx\|^2 = \langle Qx, Qx \rangle = \langle Q^H Qx, Qx \rangle = \langle x, x \rangle = \|x\|^2$. Since $\|\cdot\|$ is strictly positive $\forall x \neq 0$, we take square roots to complete the desired result.

(iii) Q orthonormal $\implies Q^H Q = I \implies Q^H = Q^{-1}$. It is thus equivalent to show that Q^H is orthonormal: $Q^H Q = I = Q^{HH} Q^H = Q^H Q^{HH} \implies Q^H = Q^{-1}$ is orthonormal.

(iv) For orthonormal matrix $Q \in M_n(\mathbb{F})$, $QQ^H = I$, the matrix with ones along the diagonal and zeros everywhere else. Thus, from the rules of matrix multiplication, $q_i q_j^T = 1, \forall i = j$ columns in Q . If $i \neq j$, the product is equal to zero. In terms of inner products, we have $\langle q_i, q_i \rangle = \|q_i\|^2 = 1$, and $\langle q_i, q_j \rangle = 0, i \neq j \implies$ every column in Q

is orthonormal.

(v) $1 = \det(I_n) = \det(Q^H Q) = \det(Q^H) \det(Q) = \det(Q) \det(Q) = \det(Q)^2 \implies |\det(Q)| = 1$. The converse is not always true.

3.11

Suppose v_1, \dots, v_n is a linearly independent list of vectors in V . Applying the Gram-Schmidt procedure gives an orthonormal list of vectors e_1, \dots, e_n such that $\text{span}(e_1, \dots, e_n) = \text{span}(v_1, \dots, v_n)$. Take some vector $v_{n+1} \in \text{span}(v_1, \dots, v_n)$ and apply the Gram-Schmidt procedure:

$$e_{n+1} = \frac{v_{n+1} - \langle v_{n+1}, e_1 \rangle e_1 - \dots - \langle v_{n+1}, e_n \rangle e_n}{\|v_{n+1} - \langle v_{n+1}, e_1 \rangle e_1 - \dots - \langle v_{n+1}, e_n \rangle e_n\|}$$

Because any $v \in V$ can be written as $v = \langle v, e_1 \rangle e_1 + \dots + \langle v, e_n \rangle e_n$, the above expression is undefined (numerator and denominator both reduce to 0).

3.16

(i) Any matrix $A \in M_{m \times n}$ with $m \geq n$ can be reduced to the product of an $m \times m$ orthonormal matrix Q and an $m \times n$ upper triangular matrix R . Since the bottom $(m - n)$ rows of any $m \times n$ matrix are entirely zeros, R and Q can be partitioned:

$$A = QR = Q \begin{bmatrix} R_1 \\ 0 \end{bmatrix} = [Q_1, Q_2] \begin{bmatrix} R_1 \\ 0 \end{bmatrix} = Q_1 R_1$$

where R_1 is an $n \times n$ upper triangular matrix, 0 is an $(m - n) \times n$ zero matrix, Q_1 is $m \times n$, Q_2 is $m \times (m - n)$ and Q_1 and Q_2 are both orthonormal. For some $Q'_2 \neq Q_2$,

$$[Q_1, Q_2] \begin{bmatrix} R_1 \\ 0 \end{bmatrix} = Q_1 R_1 = [Q_1, Q'_2] \begin{bmatrix} R_1 \\ 0 \end{bmatrix} = A$$

(ii) Suppose that $A = Q_1 R_1 = Q_2 R_2$ where Q_1, Q_2 are orthonormal and R_1, R_2 are upper triangular with positive diagonal entries. Then,

$$M := R_1 R_2^{-1} = Q_1^H Q_2.$$

Since M is orthonormal and upper triangular, it must be diagonal. Further, the diagonal entries of M are positive, because the upper triangular matrices R_1 and R_2^{-1} have positive diagonal entries, and of modulus one, because M is a diagonal orthonormal matrix. Thus, $M = I$, implying $R_1 = R_2$ and $Q_1 = Q_2$.

3.16

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

3.17

$$\begin{aligned}
A^H Ax &= A^H b \\
(\hat{Q}\hat{R})^H \hat{Q}\hat{R}x &= \hat{R}^H \hat{Q}^H b \\
\hat{R}^H \hat{Q}^H \hat{Q}\hat{R}x &= \hat{R}^H \hat{Q}^H b \\
\hat{R}^H \hat{R}x &= \hat{R}^H \hat{Q}^H b \\
\hat{R}x &= (\hat{R}^H)^{-1} \hat{R}^H \hat{Q}^H b \\
\hat{R}x &= \hat{Q}^H b
\end{aligned}$$

As desired.

3.23

This result follows directly from the triangle inequality because $\|x\| = \|x - y + y\| \leq \|x - y\| + \|y\| \implies \|x\| - \|y\| \leq \|x - y\|$ for all $x, y \in V$.

3.24

(i) Positivity is trivial. Scale preservation: for some scalar $x \in \mathbb{R}$,

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

Triangle inequality:

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

(ii) Positivity is once again trivial. For some scalar $x \in \mathbb{R}$,

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

Triangle inequality:

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

(iii) Positivity is trivial. Scale preservation:

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

Triangle inequality:

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

3.26

(i) $\|x\|_2 \leq \|x\|_1$ follows from the triangle inequality. To show $\|x\|_1 \leq \sqrt{n}$, let y be a vector of all ones, where the sign is equal to the sign of x . Then $\|x\|_1 = |\langle x, y \rangle| \leq \|y\|_2 \|x\|_2 = \sqrt{n} \|x\|_2$.

(ii) $\|x\|_2^2 = \sum_{i=1}^m x_i^2 \leq m \cdot \sup_{1 \leq m} x_i^2 = m(\sup_{1 \leq m} x_i)^2 = m\|x\|_\infty^2 \implies \|x\|_2 \leq \sqrt{m} \|x\|_\infty$

3.28

(i) $\|A\|_2 \leq \|A\|_F = \sqrt{\text{tr}(A^H A)}$, which can be interpreted as the square root of the sum of all squared entries in the A . Thus, $\frac{\|A\|_2^2}{n}$ is the average squared column sum,

and $\frac{\|A\|_2}{\sqrt{n}}$ is the average column sum. Since $\|A\|_1$ is the maximum column sum in A , $\frac{\|A\|_2}{\sqrt{n}}$ is clearly less than or equal to $\|A\|_1$.

3.29

Orthonormal transformations preserve inner products and thus preserve norms, i.e. $\|Qx\| = \|x\| \implies \|Q\| = 1$. For $R_x : M_n(\mathbb{F}) \rightarrow \mathbb{F}^n$,

$$\begin{aligned}\|R_x\| &= \sup \frac{\|R_x A\|}{\|A\|} \\ &= \sup \frac{\|Ax\|}{\|A\|} \\ &= \sup \frac{\|Ax\|}{\sup_{\|x\|_2} \frac{\|Ax\|}{\|x\|_2}} \\ &\leq \|x\|_2\end{aligned}$$

By Gram-Schmidt, any vector x with norm $\|x\|_2 = 1$ is part of an orthonormal basis, and hence is the first column of an orthonormal matrix.

3.37

$$\begin{aligned}q &= \sum_{i=1}^n L(x_i)e_i \\ &= L(1) \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + L(x) \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + L(x^2) \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \\ &= 0 \cdot \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + 1 \cdot \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + 2 \cdot \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix} \\ &= \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}\end{aligned}$$

3.38

The matrix of the derivative operator with respect to the power basis is:

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

The adjoint of D is its Hermitian conjugate:

$$D^* = \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 2 & 0 \end{bmatrix}$$

3.44

If $Ax = b$ has a solution, then for any vector y we have $\langle y, Ax \rangle = \langle y, b \rangle$. Equivalently, $\langle y, Ax \rangle = \langle A^H y, x \rangle$. It follows that for any vector y we must have $\langle A^* y, x \rangle = \langle y, b \rangle$. If y satisfies $A^* y = 0$, we must also have $\langle y, b \rangle = 0$.