

Math Pset #2: Inner Product Spaces

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Problem 1.

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

PF:

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

As we are in \mathbb{R} ,

$$= \langle x, y \rangle$$

Problem 2.

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

Problem 3. $\langle x, x^5 \rangle = \int_0^1 x^6 dx = x^7/7|_0^1 = 1/7$, $\|x\| = \int_0^1 x^2 dx = x^3/3|_0^1 = 1/3$
and $\|x^5\| = \int_0^1 x^{10} dx = x^{11}/11|_0^1 = 1/11$. Therefore $\cos \theta = \sqrt{33}/7$ implies $\theta = 34.5$.

*Alberto did most of the Latexing

Problem 4. (i)

$$\|\cos(t)\| = \frac{1}{\pi} \int_{-\pi}^{\pi} \cos^2(t) dt = \frac{1}{\pi} \left. \frac{\cos(x) \sin(x) - x}{2} \right|_{-\pi}^{\pi} i = \frac{\pi}{\pi} = 1,$$

and similarly $\|\sin(t)\| = 1$. Also

$$\|\cos(2t)\| = \frac{1}{\pi} \int_{-\pi}^{\pi} \cos^2(2t) dt = \frac{1}{\pi} \left. \frac{\sin(4t) + 4t}{8} \right|_{-\pi}^{\pi} i = \frac{\pi}{\pi} = 1,$$

and similarly $\|\sin(2t)\| = 1$. Therefore the basis is normalized.

The following integrals:

$$\langle \cos(t), \sin(t) \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(t) \sin(t) dt = \frac{1}{\pi} \left. \frac{\sin^2(x)}{x} \right|_{-\pi}^{\pi} i = 0,$$

$$\langle \cos(t), \cos(2t) \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(t) \cos(2t) dt = \frac{1}{\pi} \left. \frac{3 \sin(t) - 2 \sin^3(t)}{3} \right|_{-\pi}^{\pi} i = 0,$$

$$\langle \cos(t), \sin(2t) \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(t) \sin(2t) dt = \frac{1}{\pi} \left. \frac{-2 \cos^3(t)}{3} \right|_{-\pi}^{\pi} i = 0,$$

$$\langle \cos(2t), \sin(2t) \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} \cos(2t) \sin(2t) dt = \frac{1}{\pi} \left. \frac{-\cos^2(2t)}{4} \right|_{-\pi}^{\pi} i = 0,$$

and so on, shows that S is an orthonormal basis.

(ii)

$$\|t\| = \frac{1}{\pi} \int_{-\pi}^{\pi} t dt = \frac{1}{\pi} \left. \frac{t^2}{2} \right|_{-\pi}^{\pi} = 0.$$

(iii) Since $\langle x, \cos(3x) \rangle = 0$ for any $x \in S$, $\text{proj}_X(\cos(3x)) = 0$.

(i)

$$\begin{aligned} \langle \sin(t), t \rangle &= \sin(t) - t \cos(t) \Big|_{-\pi}^{\pi} = 2\pi, \\ \langle \cos(t), t \rangle &= t \sin(t) - \cos(t) \Big|_{-\pi}^{\pi} = 0, \\ \langle \cos(2t), t \rangle &= (2t \sin(2t) + \cos(2t)) / 4 \Big|_{-\pi}^{\pi} = 0, \text{ and finally} \\ \langle \sin(2t), t \rangle &= \sin(2x) - 2x \cos(2x) / 4 \Big|_{-\pi}^{\pi} = -\pi. \end{aligned}$$

Therefore, $\text{proj}_X(t) = 2\pi \sin(t) - \pi \sin(2t)$

Problem 9. A rotation of angle θ in \mathbb{R}^2 represented as a matrix R in the standard basis is an orthonormal transformation since

$$\begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}^T \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix} =$$

$$\begin{bmatrix} \cos^2(\theta) + \sin^2(\theta) & 0 \\ 0 & \cos^2(\theta) + \sin^2(\theta) \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}.$$

Similarly, one shows that $RR^T = I$. Therefore, a rotation in \mathbb{R}^2 is an orthonormal transformation.

Problem 1. 0 (i) Suppose Q represents an orthonormal operator on \mathbb{F}^n . Then $\langle x, y \rangle = \langle Q(x), Q(y) \rangle$ for each $x, y \in \mathbb{F}^n$. Since $\langle Q(x), Q(y) \rangle = (Qx)^H(Qy) = x^H Q^H Q y$, it equals $x^H y$ for all $x, y \in \mathbb{F}^n$ only if $Q^H Q = I$. On the other hand if $Q^H Q = Q Q^H = I$, then $\langle Q(x), Q(y) \rangle = (Qx)^H(Qy) = x^H Q^H Q y = x^H y = \langle x, y \rangle$.
(ii)

$$\|Qx\| = \sqrt{\langle Qx, Qx \rangle} = \sqrt{x^H Q^H Q x} = \sqrt{\langle x, x \rangle} = \|x\|.$$

(iii) If $Q^H Q = Q Q^H = I$, then $Q^{-1} = Q^H$. Since $(Q^H)^H = Q$, Q^H is also orthonormal:

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

(iv) Let q_i denote the i^{th} column of Q . Since Q is orthonormal, $(Q^H Q)_{ij} = q_i^H q_j = \langle q_i, q_j \rangle$ is 1 if $i = j$ and 0 if $i \neq j$. Thus, the columns of Q are orthonormal.

(v) The matrix

$$\begin{bmatrix} 2 & 0 \\ 0 & 1/2 \end{bmatrix}$$

shows that the converse is not true.

(vi)

$$(Q_1 Q_2)^H Q_1 Q_2 = Q_2^H Q_1^H Q_1 Q_2 = Q_2^H Q_2 = I$$

and

$$Q_1 Q_2 (Q_1 Q_2)^H = Q_1 Q_2 Q_2^H Q_1^H = Q_1 Q_1^H = I.$$

Therefore, $Q_1 Q_2$ is orthonormal.

Problem 1. 1 Fix $N \in \mathbb{N}$, $N > 0$, and suppose $\{x_i\}_{i=1}^N$ is a set of linearly dependent vectors in V . Also, suppose, without loss of generality, that for $2 < k < N$, $\{x_i\}_{i=1}^{k-1}$ is a linearly independent set and $\{x_i\}_{i=1}^k$ is a linearly dependent set. Then $\{q_i\}_{i=1}^{k-1}$ (as they are defined in the book) is also a linearly independent set. However, since $x_k \in \text{span}(\{x_i\}_{i=1}^{k-1})$, we have that $q_k = 0$. Therefore the Gram-Schmidt orthonormalization process brakes down.

Problem 1. 6 (i) Let $A \in \mathbb{M}_{m \times n}$ where $\text{rank}(A) = n \leq m$. Then there exist orthonormal $Q \in \mathbb{M}_{m \times m}$ and upper triangular $R \in \mathbb{M}_{m \times n}$ such that $A = QR$. Since $\tilde{Q} = -Q$ is still orthonormal ($-Q(-Q)^H = -Q(-Q^H) = QQ^H = I$ and similarly one shows $(-Q)^H(-Q) = I$) and $\tilde{R} = -R$ is still upper triangular, $A = QR = \tilde{Q}\tilde{R}$. Therefore QR-decomposition is not unique.

(ii) Now take a reduced QR-decomposition $A = \hat{Q}\hat{R}$, where $\hat{Q} \in \mathbb{M}_{m \times n}$ is orthonormal and $\hat{R} \in \mathbb{M}_{n \times n}$ is upper triangular. Since A has full column rank, \hat{R} has full rank and is therefore nonsingular. Then,

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

and premultiplying both LHS and RHS of the last equation by \hat{R}^{-1} gives $\hat{R}x = \hat{Q}^H b$.

Problem 2. 3 Let $x, y \in V$ and define $v := -y$. Since a norm is nonnegative and satisfies the triangular property, $\|x\| - \|v\| \leq \|x\| + \|v\| \leq \|x + v\|$. Then our definition of v implies $\|x\| - \|y\| = \|x\| - \|-y\| \leq \|x - y\|$. Interchanging the role of x and y and using the homogeneity property of norms we have $\|y\| - \|x\| \leq \|y - x\| = \|- (y - x)\| = \|x - y\|$, and the result follows.

Problem 2. 4 (i) Since $|f(t)| \geq 0$ for every t , so is $\int_a^b |f(t)|dt$. In addition, if $f = 0$, then $\int_a^b |f(t)|dt = 0$. On the other hand, if $\int_a^b |f(t)|dt = 0$ and $|f(t)| \geq 0$, it must be that $|f(t)| = 0$ for all t , implying that $f = 0$. Now take a constant $c \in \mathbb{F}$, then $\int_a^b |cf(t)|dt = \int_a^b |c||f(t)|dt = |c| \int_a^b |f(t)|dt$, since c does not depend on t . Finally, take $g \in C([a, b]; \mathbb{F})$. Since $|f(t) + g(t)| \leq |f(t)| + |g(t)|$ for all t and the integral is a linear operator, we have that $\int_a^b |f(t) + g(t)|dt \leq \int_a^b |f(t)|dt + \int_a^b |g(t)|dt$.

(ii) Since $|f(t)|^2 \geq 0$ for every t , so is $\int_a^b |f(t)|^2 dt$ and its square root. In addition, if $f = 0$, then $|f(t)|^2 = 0$ for all t and $\sqrt{\int_a^b |f(t)|^2 dt} = 0$. On the other hand, if $\sqrt{\int_a^b |f(t)|^2 dt} = 0$, then $\int_a^b |f(t)|^2 dt = 0$ and since $|f(t)|^2 \geq 0$ for all t , it must be that $|f(t)|^2 = 0$ for all t , implying that $f = 0$. Now take a constant $c \in \mathbb{F}$, then $\sqrt{\int_a^b |cf(t)|^2 dt} = \sqrt{\int_a^b |c|^2 |f(t)|^2 dt} = |c| \sqrt{\int_a^b |f(t)|^2 dt}$, since c does not depend on t . Finally, take $g \in C([a, b]; \mathbb{F})$. Since $|f(t) + g(t)| \leq |f(t)| + |g(t)|$ for all t , $x \mapsto x^2$ and $x \mapsto \sqrt{x}$ are monotonically increasing for nonnegative x and the integral is a linear operator, we have that $\sqrt{\int_a^b |f(t) + g(t)|^2 dt} \leq \sqrt{\int_a^b |f(t)|^2 dt + \int_a^b |g(t)|^2 dt} \leq \|f\|_{L^2} + \|g\|_{L^2}$.

(iii) Since $|f(x)| \geq 0$ for all x , so is the $\sup_{x \in [a, b]} |f(x)|$. In addition, if $f = 0$, then $\sup_{x \in [a, b]} |f(x)|$ is also zero. On the other hand, since $|f(x)| \geq 0$ for all x , $0 \leq \sup_{x \in [a, b]} |f(x)| = 0$ implies that we must have $f = 0$. Now take a constant $c \in \mathbb{F}$, then $\sup_{x \in [a, b]} |cf(x)| = \sup_{x \in [a, b]} |c||f(x)| = |c| \sup_{x \in [a, b]} |f(x)|$. Finally, take $g \in C([a, b]; \mathbb{F})$. Since $|f(x) + g(x)| \leq |f(x)| + |g(x)|$ for all x , we have that $\sup_{x \in [a, b]} |f(x) + g(x)| \leq \sup_{x \in [a, b]} \{|f(x)| + |g(x)|\} \leq \sup_{x \in [a, b]} |f(x)| + \sup_{x \in [a, b]} |g(x)|$.

Problem 2. 6 We show that topological equivalence is an equivalence relation. Let $\|\cdot\|_r$ be a norm on X for $r \in \{a, b, c\}$. Clearly $\|\cdot\|_r$ is topologically equivalent with itself, just pick any $0 < m \leq 1$ and any $M \geq 1$ to show this. Also, 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 $0 < 1/M' \leq 1/m'$. Finally, if $\|\cdot\|_a$ is topologically equivalent to $\|\cdot\|_b$ with constants $0 < m \leq M$ and so is $\|\cdot\|_b$ with $\|\cdot\|_c$ with constants $0 < m' \leq M'$, then $\|\cdot\|_a$ is topologically equivalent to $\|\cdot\|_b$ with constants $0 < mm' \leq MM'$.

Take $x \in \mathbb{R}^n$ Notice that

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

and that

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

prove that $\|x\|_2 \leq \|x\|_1 \leq \sqrt{n} \|x\|_2$.

Also notice that

$$\max_i |x_i| = \left(\max_i |x_i|^2 \right)^{1/2} \leq \left(\sum_{i=1}^n |x_i|^2 \right)^{1/2} =$$

and

$$\sum_{i=1}^n |x_i|^2 \leq n \cdot \max_i |x_i|^2$$

prove that $\|x\|_\infty \leq \|x\|_2 \leq \sqrt{n} \|x\|_\infty$.

Problem 2. 8 (i) Notice that (applying the results of the previous exercise)

$$\sup_{x \neq 0} \frac{\|Ax\|_1}{\|x\|_1} \leq \sup_{x \neq 0} \frac{\|Ax\|_1}{\|x\|_1} \leq \sqrt{n} \sup_{x \neq 0} \frac{\|Ax\|_2}{\|x\|_2},$$

and

$$\sup_{x \neq 0} \frac{\|Ax\|_1}{\|x\|_1} \geq \sup_{x \neq 0} \frac{\|Ax\|_2}{\|x\|_1} \geq \frac{1}{\sqrt{n}} \sup_{x \neq 0} \frac{\|Ax\|_2}{\|x\|_2}$$

imply that $\frac{1}{\sqrt{n}} \|A\|_2 \leq \|A\|_1 \leq \|A\|_2$.

(ii) Notice that

$$\sup_{x \neq 0} \frac{\|Ax\|_2}{\|x\|_2} \leq \sup_{x \neq 0} \frac{\sqrt{n} \|Ax\|_\infty}{\|x\|_\infty},$$

and

$$\sup_{x \neq 0} \frac{\|Ax\|_2}{\|x\|_2} \geq \sup_{x \neq 0} \frac{\|Ax\|_\infty}{\sqrt{n} \|x\|_\infty}.$$

Problem 2. 9 Take an arbitrary $x \neq 0$ and suppose $\|\cdot\|$ is an inner product induced norm. Since

$$\|Qx\| = (\langle Qx, Qx \rangle)^{1/2} = (\langle Q^H Qx, x \rangle)^{1/2} = (\langle x, x \rangle)^{1/2} = \|x\|,$$

then

$$\|Q\| = \sup_{x \neq 0} \frac{\|Qx\|}{\|x\|} = 1.$$

Now let $R_x : \mathbb{M}_n(\mathbb{F}) \rightarrow \mathbb{F}^n, A \mapsto Ax$ for every $x \in \mathbb{F}^n$. Notice that

$$\|R_x\| = \sup_{A \neq 0} \frac{\|Ax\|}{\|A\|} = \sup_{A \neq 0} \frac{\|Ax\| \|x\|}{\|A\| \|x\|} \leq \sup_{A \neq 0} \left(\frac{\|Ax\| \|x\|}{\|Ax\|} \right) = \|x\|.$$

Problem 3. 0 Take an arbitrary $A \in \mathbb{M}_n(\mathbb{F})$. First, $\|A\|_S = \|SAS^{-1}\| \geq 0$ for any A because $\|\cdot\|$ is a norm on $\mathbb{M}_n(\mathbb{F})$ and $SAS^{-1} \in \mathbb{M}_n(\mathbb{F})$. In addition, $\|0\|_S = \|S0S^{-1}\| = \|0\| = 0$ and if $0 = \|A\|_S = \|SAS^{-1}\|$, then $SAS^{-1} = 0$ which implies $A = 0$. Second, take $a \in \mathbb{F}$, then

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

Finally, let $B \in \mathbb{M}_n(\mathbb{F})$ and notice that

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

Therefore $\|\cdot\|_S$ is a norm on $\mathbb{M}_n(\mathbb{F})$.