

Math 33A
Linear Algebra and Applications

Discussion for July 11-14, 2022

Problem 1.

Here is an infinite-dimensional version of Euclidean space: In the space of all infinite sequences, consider the subspace ℓ_2 of square-summable sequences (namely, those sequences (x_1, x_2, \dots) for which the infinite series $x_1^2 + x_2^2 + \dots$ converges). For x and y in ℓ_2 , we define

$$\|\vec{x}\| = \sqrt{x_1^2 + x_2^2 + \dots} \quad \text{and} \quad \vec{x} \cdot \vec{y} = x_1y_1 + x_2y_2 + \dots.$$

A preliminary question is, why do $\|\vec{x}\|$ and $\vec{x} \cdot \vec{y}$ make sense, that is, why are they finite real numbers?

- Check that $\vec{x} = (1, 1/2, 1/4, 1/8, 1/16, \dots)$ is in ℓ_2 , and find $\|\vec{x}\|$. Recall the formula for the geometric series: $1 + a + a^2 + a^3 + \dots = 1/(1 - a)$ if $-1 < a < 1$.
- Find the angle between $(1, 0, 0, 0, \dots)$ and $(1, 1/2, 1/4, 1/8, \dots)$.
- Give an example of a sequence (x_1, x_2, \dots) that converges to 0 ($\lim_{n \rightarrow \infty} x_n = 0$) but does not belong to ℓ_2 .
- Let L be the subspace of ℓ_2 spanned by $(1, 1/2, 1/4, 1/8, \dots)$. Find the orthogonal projection of $(1, 0, 0, 0, \dots)$ onto L .

The Hilbert space ℓ_2 was initially used mostly in physics: Werner Heisenberg's formulation of quantum mechanics is in terms of ℓ_2 . Today, this space is used in many other applications, including economics. See, for example, the work of the economist Andreu Mas-Colell of the University of Barcelona.

Solution:

(a) Using the formula for the geometric series $\|\vec{x}\|^2 = 4/3$ so $\|\vec{x}\| = 2/\sqrt{3}$.

(b) Set $\vec{x} = (1, 0, 0, 0, \dots)$ and $\vec{y} = (1, 1/2, 1/4, 1/8, \dots)$, then

$$\theta = \arccos \left(\frac{\vec{x} \cdot \vec{y}}{\|\vec{x}\| \cdot \|\vec{y}\|} \right) = \arccos \left(\frac{1}{2/\sqrt{3}} \right) = \arccos \left(\frac{\sqrt{3}}{2} \right) = \frac{\pi}{6}.$$

(c) Consider $\vec{x} = (1, 1/\sqrt{2}, 1/\sqrt{3}, 1/\sqrt{4}, \dots)$, then

$$\|\vec{x}\|^2 = \sqrt{1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} + \dots} = \sqrt{\sum_{n=1}^{\infty} \frac{1}{n}}$$

which diverges since the harmonic series $\sum_{n=1}^{\infty} \frac{1}{n}$ diverges.

- (d) Let $\vec{x} = (1, 0, 0, 0, \dots)$ and $\vec{y} = (1, 1/2, 1/4, 1/8, \dots)$, we want the orthogonal projection of \vec{x} onto $L = \text{span}(\vec{y})$. For this, we first find a vector of length one in the direction of \vec{y} , namely

$$\vec{u} = \frac{\vec{y}}{\|\vec{y}\|} = \frac{\sqrt{3}}{2} \left(1, \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \dots \right)$$

and now we compute

$$\text{proj}_L(\vec{x}) = (\vec{x} \cdot \vec{u})\vec{u} = \left(\frac{\sqrt{3}}{2} \right) \frac{\sqrt{3}}{2} \left(1, \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \dots \right) = \left(\frac{3}{4}, \frac{3}{8}, \frac{3}{16}, \frac{3}{32}, \dots \right).$$

Problem 2.

Give an algebraic proof for the triangle inequality

$$\|\vec{v} + \vec{w}\| \leq \|\vec{v}\| + \|\vec{w}\|.$$

Draw a sketch.

Solution: Note that

$$\begin{aligned} \|\vec{v} + \vec{w}\|^2 &= (\vec{v} + \vec{w}) \cdot (\vec{v} + \vec{w}) = \vec{v} \cdot \vec{v} + \vec{v} \cdot \vec{w} + \vec{w} \cdot \vec{v} + \vec{w} \cdot \vec{w} = \\ &= \|\vec{v}\|^2 + 2(\vec{v} \cdot \vec{w}) + \|\vec{w}\|^2 \leq \|\vec{v}\|^2 + 2(\|\vec{v}\| \cdot \|\vec{w}\|) + \|\vec{w}\|^2 = (\|\vec{v}\| + \|\vec{w}\|)^2 \end{aligned}$$

where we have used the Cauchy-Schwarz inequality. Thus $\|\vec{v} + \vec{w}\| \leq \|\vec{v}\| + \|\vec{w}\|$.

Problem 3.

- (a) Consider a vector \vec{v} in \mathbb{R}^n , and a scalar k . Show that $\|k\vec{v}\| = |k|\|\vec{v}\|$.
 (b) Show that if \vec{v} is a nonzero vector in \mathbb{R}^n , then $\vec{u} = \frac{\vec{v}}{\|\vec{v}\|}$ is a unit vector.

Solution:

- (a) Note that

$$\|k\vec{v}\|^2 = (k\vec{v}) \cdot (k\vec{v}) = k^2(\vec{v} \cdot \vec{v}) = k^2\|\vec{v}\|^2$$

and thus taking square roots $\|k\vec{v}\| = |k|\|\vec{v}\|$ since $|k| = \sqrt{k^2}$.

- (b) We compute

$$\|\vec{u}\| = \left\| \frac{\vec{v}}{\|\vec{v}\|} \right\| = \left\| \frac{1}{\|\vec{v}\|} \vec{v} \right\| = \frac{1}{\|\vec{v}\|} \|\vec{v}\| = 1$$

using what we just proved.

Problem 4(★).

Can you find a line L in \mathbb{R}^n and a vector \vec{x} in \mathbb{R}^n such that $\vec{x} \cdot \text{proj}_L \vec{x}$ is negative? Explain, arguing algebraically.

Solution: No. Let $\vec{x} = \vec{x}^{\parallel} + \vec{x}^{\perp}$ be the decomposition of \vec{x} into the components parallel and perpendicular to L . In particular $\vec{x}^{\parallel} = \text{proj}_L \vec{x}$ and $\vec{x}^{\perp} \cdot \vec{x}^{\parallel} = 0$. Now

$$\vec{x} \cdot \text{proj}_L \vec{x} = (\vec{x}^{\parallel} + \vec{x}^{\perp}) \cdot \vec{x}^{\parallel} = \vec{x}^{\parallel} \cdot \vec{x}^{\parallel} + \vec{x}^{\perp} \cdot \vec{x}^{\parallel} = \|\vec{x}^{\parallel}\|^2 \geq 0.$$

Problem 5(★).

The following is one way to define the quaternions, discovered in 1843 by the Irish mathematician Sir W. R. Hamilton. Consider the set H of all 4×4 matrices M of the form

$$M = \begin{bmatrix} p & -q & -r & -s \\ q & p & s & -r \\ r & -s & p & q \\ s & r & -q & p \end{bmatrix}$$

where p, q, r, s are arbitrary real numbers. We can write M more succinctly in partitioned form as

$$M = \begin{bmatrix} A & -B^T \\ B & A^T \end{bmatrix}$$

where A and B are rotation–scaling matrices.

- Show that H is closed under addition: If M and N are in H , then so is $M + N$.
- Show that H is closed under scalar multiplication: If M is in H and k is an arbitrary scalar, then kM is in H .
- The above show that H is a subspace of the linear space $\mathbb{R}^{4 \times 4}$. Find a basis of H , and thus determine the dimension of H .
- Show that H is closed under multiplication: If M and N are in H , then so is MN .
- Show that if M is in H , then so is M^T .
- For a matrix M in H , compute $M^T M$.
- Which matrices M in H are invertible? If a matrix M in H is invertible, is M^{-1} necessarily in H as well?
- If M and N are in H , does the equation $MN = NM$ always hold?

Solution:

(a) When we add two matrices in H we obtain another matrix in H

$$\begin{bmatrix} A & -B^T \\ B & A^T \end{bmatrix} + \begin{bmatrix} C & -D^T \\ D & C^T \end{bmatrix} = \begin{bmatrix} (A+C) & -(B+D)^T \\ (B+D) & (A+C)^T \end{bmatrix}.$$

(b) When we multiply a matrix in H by a real scalar we obtain a matrix in H

$$k \begin{bmatrix} A & -B^T \\ B & A^T \end{bmatrix} = \begin{bmatrix} (kA) & -(kB)^T \\ (kB) & (kA)^T \end{bmatrix}.$$

(c) The general element of H has four arbitrary constants, so H has dimension 4. A basis is

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & -1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}.$$

(d) When we multiply two matrices in H we obtain another matrix in H

$$\begin{bmatrix} A & -B^T \\ B & A^T \end{bmatrix} \begin{bmatrix} C & -D^T \\ D & C^T \end{bmatrix} = \begin{bmatrix} (AC - B^T D) & -(BC + A^T D)^T \\ (BC + A^T D) & (AC - B^T D)^T \end{bmatrix}$$

where it is useful to notice that since all A, B, C, D are rotation-scaling matrices, they commute with each other.

(e) When we transpose a matrix in H we obtain another matrix in H

$$\begin{bmatrix} A & -B^T \\ B & A^T \end{bmatrix}^T = \begin{bmatrix} (A^T) & -(-B)^T \\ (-B) & (A^T)^T \end{bmatrix}.$$

(f) We expand $M^T M$ as

$$\begin{bmatrix} p & q & r & s \\ -q & p & -s & r \\ -r & s & p & -q \\ -s & -r & q & p \end{bmatrix} \begin{bmatrix} p & -q & -r & -s \\ q & p & s & -r \\ r & -s & p & q \\ s & r & -q & p \end{bmatrix} = (p^2 + q^2 + r^2 + s^2) \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}.$$

(g) If $M \neq 0$ then $p^2 + q^2 + r^2 + s^2 \neq 0$ so by the above

$$M^T M = (p^2 + q^2 + r^2 + s^2) \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

and thus

$$\left(\frac{1}{(p^2 + q^2 + r^2 + s^2)} M^T \right) M = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

so

$$M^{-1} = \frac{1}{(p^2 + q^2 + r^2 + s^2)} M^T = \frac{1}{(p^2 + q^2 + r^2 + s^2)} \begin{bmatrix} p & q & r & s \\ -q & p & -s & r \\ -r & s & p & -q \\ -s & -r & q & p \end{bmatrix}.$$

Problem 6.

Consider a consistent system $A\vec{x} = \vec{b}$.

- Show that this system has a solution \vec{x}_0 in $(\ker A)^\perp$. Justify why an arbitrary solution \vec{x} of the system can be written as $\vec{x} = \vec{x}_h + \vec{x}_0$, where \vec{x}_h is in $\ker(A)$ and \vec{x}_0 is in $(\ker A)^\perp$.
- Show that the system $A\vec{x} = \vec{b}$ has only one solution in $(\ker A)^\perp$.
- If \vec{x}_0 is the solution in $(\ker A)^\perp$ and \vec{x}_1 is another solution of the system $A\vec{x} = \vec{b}$, show that $\|\vec{x}_0\| < \|\vec{x}_1\|$. The vector \vec{x}_0 is called the minimal solution of the linear system $A\vec{x} = \vec{b}$.

Solution:

- Since the system $A\vec{x} = \vec{b}$ is consistent, it has at least one solution \vec{x} . Let $\vec{x} = \vec{x}^\parallel + \vec{x}^\perp$ be the decomposition of \vec{x} into the components parallel and perpendicular to $V = \ker(A)$. In particular \vec{x}^\perp is in $(\ker(A))^\perp$ and $\vec{x}^\parallel = \text{proj}_V \vec{x}$ is in $\ker(A)$ so $A\vec{x}^\parallel = \vec{0}$. Now

$$\vec{b} = A\vec{x} = A(\vec{x}^\parallel + \vec{x}^\perp) = A\vec{x}^\parallel + A\vec{x}^\perp = A\vec{x}^\perp$$

so $\vec{x}_0 = \vec{x}^\perp$ is a solution of the system in $(\ker(A))^\perp$ and $\vec{x}_h = \vec{x}^\parallel$ is in $\ker(A)$.

- Suppose that $A\vec{x} = \vec{b}$ has two solutions \vec{x}_1 and \vec{x}_2 in $(\ker(A))^\perp$. Since $(\ker(A))^\perp$ is a linear subspace, then $\vec{x}_1 - \vec{x}_2$ is in $(\ker(A))^\perp$. Thus $A(\vec{x}_1 - \vec{x}_2) = A\vec{x}_1 - A\vec{x}_2 = \vec{b} - \vec{b} = \vec{0}$ so $\vec{x}_1 - \vec{x}_2$ is in $\ker(A)$. Now $\vec{x}_1 - \vec{x}_2$ is both in $\ker(A)$ and $(\ker(A))^\perp$, but $\vec{0}$ is the only element in both subspaces, so $\vec{x}_1 - \vec{x}_2 = \vec{0}$. Thus $\vec{x}_1 = \vec{x}_2$.
- Let $\vec{x}_1 = \vec{x}_1^\parallel + \vec{x}_1^\perp$ be the decomposition of \vec{x}_1 into the components parallel and perpendicular to $V = \ker(A)$. Now by the first part above we have that \vec{x}_1^\perp is a solution of the system in $(\ker(A))^\perp$. Since \vec{x}_0 is also a solution of the system

in $(\ker(A))^\perp$, by the second part above we have $\vec{x}_1^\perp = \vec{x}_0$. Since $\vec{x}_1 \neq \vec{x}_0$ we have $\vec{x}_1^\perp \neq \vec{0}$, so $\|\vec{x}_1^\perp\| > 0$ and by the Pythagoras theorem

$$\|\vec{x}_1\| = \|\vec{x}_1^\perp + \vec{x}_0\| \geq \|\vec{x}_1^\perp\| + \|\vec{x}_0\| > \|\vec{x}_0\|.$$

Problem 7.

Define the term minimal least-squares solution of a linear system. Explain why the minimal least-squares solution \vec{x}^* of a linear system $A\vec{x} = \vec{b}$ is in $(\ker(A))^\perp$.

Solution: We know that the least-squares solution of a linear system $A\vec{x} = \vec{b}$ are the exact solutions of the consistent linear system $A^T A\vec{x} = A^T \vec{b}$. In the previous problem we defined the term minimal solution of a consistent linear system. We then define the minimal least-squares solution of the linear system $A\vec{x} = \vec{b}$ to be the minimal solutions of the consistent linear system $A^T A\vec{x} = A^T \vec{b}$.

We first prove that $\ker(A) = \ker(A^T A)$, this will be useful. Let \vec{v} be in $\ker(A)$, then $A^T A\vec{v} = A^T \vec{0} = \vec{0}$ so \vec{v} is in $\ker(A^T A)$. Let \vec{v} be in $\ker(A^T A)$, then $\vec{0} = A^T A\vec{v} = A^T(A\vec{v})$ so $A\vec{v}$ is in $\ker(A^T)$. Now $A\vec{v}$ is in $\text{im}(A)$, and also in $\ker(A^T) = (\text{im}(A))^\perp$, but $\vec{0}$ is the only element in both subspaces, so $A\vec{v} = \vec{0}$, so \vec{v} is in $\ker(A)$.

Now, let \vec{x}^* be the minimal least-squares solution of the linear system $A\vec{x} = \vec{b}$. Then \vec{x}^* is the minimal solutions of the consistent linear system $A^T A\vec{x} = A^T \vec{b}$, so by the previous exercise \vec{x}^* is in $(\ker(A^T A))^\perp = (\ker(A))^\perp$.