

Solutions to Homework #4, Math 116

Keziah Cook and Michael McElroy

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Problem 3.1 (Luenberger)

Let x, y be vectors in a pre-Hilbert space. Show that $|\langle x|y\rangle| = \|x\| \cdot \|y\|$ iff $\exists \alpha, \beta \in \mathbb{R}$ s.t. $\alpha x + \beta y = 0$.

Solution to 3.1

This is really just Cauchy-Schwartz, which says that $|\langle x|y\rangle|^2 \leq \|x\|^2 \|y\|^2 = \langle x|x\rangle \langle y|y\rangle$, with equality iff x and y are linearly dependent. This is done in lemma 1 on page 47 of Luenberger. \square

Problem 3.3 (Luenberger)

Let H consist of all $m \times n$ real matrices with addition and scalar multiplication defined as the usual corresponding operations with matrices and when the inner product of two matrices is defined as

$$(A | B) = \text{Trace} [A'QB]$$

where A' denotes the transpose of the matrix A and Q is a symmetric, positive-definite $m \times m$ matrix. Prove that H is a Hilbert space.

Solution to 3.3

First we need to show that $(A | B) = \text{Trace} [A'QB]$ defines an inner product. Then we'll show that H with respect to the norm $\|X\| = \sqrt{(X | X)}$ is complete.

Inner products must satisfy four properties:

1. $(A | B) = \overline{(B | A)}$

As H is defined to be the space of real valued matrices, our base field is \mathbb{R} and thus $\overline{(B | A)} = (B | A)$.

$$\begin{aligned} (A | B) &= \text{Trace} [A'QB] \\ &= \text{Trace} [(B'QA)'] \\ &= \text{Trace} [B'QA] \\ &= (B | A) \end{aligned}$$

By the facts that $\text{Trace} [A] = \text{Trace} [A']$ and that $(AB)' = B'A'$.

2. $(A + B | C) = (A | C) + (B | C)$

$$\begin{aligned} (A + B | C) &= \text{Trace} [(A + B)'QC] \\ &= \text{Trace} [A'QC + B'QC] \\ &= \text{Trace} [A'QC] + \text{Trace} [B'QC] \\ &= (A | C) + (B | C) \end{aligned}$$

3. $(\lambda A | B) = \lambda(A | B)$

$$\begin{aligned} (\lambda A | B) &= \text{Trace} [(\lambda A)'QB] \\ &= \text{Trace} [\lambda A'QB] \\ &= \lambda \text{Trace} [A'QB] \\ &= \lambda(A | B) \end{aligned}$$

4. $(A | A) \geq 0$ and $(A | A) = 0 \iff A = \mathbf{0}$

Since Q is positive definite, we know that for every $x \in R^m$, $x \neq 0$, $x'Qx > 0$. If we write $A = [x_1, x_2, \dots, x_n]$, where x_i is the i th column of A , then $A'QA$ has entries of the form $(x_i)'Qx_j$. Since Q is positive definite, every entry of $A'QA$ is greater than or equal to zero. Thus, $\text{Trace}[A'QA] \geq 0$. Equality only happens when all the columns of A are equal to 0.

This shows H is a pre-Hilbert space. Now we need to show that every Cauchy sequence in H converges to an element of H . Let $\{A_n\}$ be a Cauchy sequence in H . This means that for $\epsilon > 0$, $\exists N$ such that $n, m > N \rightarrow \|A_n - A_m\| < \epsilon$. $\|A_n - A_m\| = \sqrt{\text{Trace} [(A_n - A_m)'Q(A_n - A_m)]}$ This is a linear combination of elements of $A_n - A_m$. Since each entry of the diagonal of $(A_n - A_m)'Q(A_n - A_m)$ is positive, $\{A_n\}$ Cauchy implies that the sequence $\{(a_{ij}^{(n)} - a_{ij}^{(m)})\}$ is Cauchy in R . Since R is complete, each of these sequences converges to a real number. We let A be the matrix $\{A_n\}$ converges to point-wise. Clearly, A is an $m \times n$ real-valued matrix.

Now we need to show $\{A_n\} \rightarrow A$ in the inner-product norm. $\text{Trace} [(A_n - A)'Q(A_n - A)]$ depends only on a linear combination of elements of the form $(a_{ij}^{(n)} - a_{ij})$. There are at most $n(n-1)/2$ such elements (as Q is symmetric). We need each element to be less than some fraction of ϵ . For each entry, there exists an N_{ij} such that this holds for $n > N_{ij}$. We let $N = \max_{i,j} N_{ij}$. Then for $n > N$, $\|A_n - A\| < \epsilon$. Thus $\{A_n\} \rightarrow A$, and H is complete. Therefore H is a Hilbert space. \square

Problem 3.4 (Luenberger)

Show that if $g(x_1, x_2, \dots, x_n) = 0$, the normal equations possess a solution but it is not unique.

Solution to 3.4

First, we note that we can write the normal equations as:

$$[\alpha_1 \quad \dots \quad \alpha_n] \cdot \begin{bmatrix} (y_1 | y_1) & (y_1 | y_2) & \dots & (y_1 | y_n) \\ (y_2 | y_1) & \cdot & \dots & \cdot \\ \cdot & \cdot & \dots & \cdot \\ \cdot & \cdot & \dots & \cdot \\ (y_n | y_1) & \cdot & \dots & (y_n | y_n) \end{bmatrix} = \begin{bmatrix} (x | y_1) \\ (x | y_2) \\ \cdot \\ \cdot \\ (x | y_n) \end{bmatrix}$$

$g(x_1, x_2, \dots, x_n)$ is the determinant of the matrix G , above. From linear algebra we know that $g = 0$ means the columns of the matrix G are linearly dependent. This tells us that we have more unknowns than (independent) equations, or equivalently that either many vectors (all elements of a dimension $k \geq 1$ subspace) satisfy the normal equations, or (if the equations are inconsistent) that no vectors satisfy the normal equations. The Projection Theorem guarantees that a solution exists (i.e. the equations must be consistent), and thus an entire subspace of solutions exist. \square

Problem 3.5 (Luenberger)

Find the linear function $x(t) = a + bt$ minimizing $\int_{-1}^1 [t^2 - x(t)]^2 dt$.

Solution to 3.5

If we consider ourselves working in $L_2[-1, 1]$, and considering the subspace $P = \{a + bt \mid a, b \in \mathbb{R}\}$, we see that this is a minimum norm problem from $t^2 \in L_2[-1, 1]$ to the subspace P . The projection theorem tells us that there is a unique solution, and we may find it using orthogonality. Our inner product is $\langle x|y \rangle = \int_{-1}^1 x(t)y(t) dt$. We could alternatively use calculus, but this does not give us the existence statement.

Method 1 Let $p = a + bt$ be the optimal affine function. The projection theorem tells us that $t^2 - p \perp P$, i.e. $\langle t^2 - p|\alpha + \beta t \rangle = 0$ for all $\alpha, \beta \in \mathbb{R}$.

$$\begin{aligned} 0 &= \langle t^2 - p|\alpha + \beta t \rangle = \int_{-1}^1 (t^2 - a - bt)(\alpha + \beta t) dt \\ &= \alpha \left[\frac{1}{3}t^3 - \frac{b}{2}t^2 - at \right]_{-1}^1 + \beta \left[\frac{1}{4}t^4 - \frac{b}{3}t^3 - \frac{a}{2}t^2 \right]_{-1}^1 \\ &= \alpha \left[\frac{2}{3} - 2a \right] - \frac{2}{3}\beta b. \end{aligned}$$

Therefore $a = \frac{1}{3}$ and $b = 0$.

Method 2 Using simple calculus, we have $f(a, b) = \int_{-1}^1 [t^2 - a - bt]^2 dt$, so an extremum point must obey the twin equations $df/da = 0, df/db = 0$. We have

$$f(a, b) = \int_{-1}^1 t^4 - 2bt^3 + (b^2 - 2a)t^2 + 2abt + a^2 dt = \frac{2}{5} + \frac{2}{3}(b^2 - a) + 2a^2.$$

$$\begin{aligned} 0 &= \frac{df}{da} = -\frac{4}{3} + 4a \implies a = \frac{1}{3}. \\ 0 &= \frac{df}{db} = \frac{4}{3}b \implies b = 0. \end{aligned}$$

This is the only extremal point, and it is easy to show that it's a minimum: $\frac{d^2f}{da^2} = 4, \frac{d^2f}{db^2} = 0$.

So the (unique) solution is $x(t) = \frac{1}{3}$. □

Problem 3.6 (Luenberger)

Given $x \in L^2[0, 1]$, we seek a polynomial p of degree less than or equal to n which minimizes $\int_0^1 |x(t) - p(t)|^2 dt$ while obeying the restriction that $\int_0^1 p(t) dt = 0$.

- (a) Show that the problem has a unique solution.
- (b) Show that we may solve the problem via first finding q of degree $\leq n$ which minimizes $\|x - q\|^2$ and then finding p of degree $\leq n$ which minimizes $\|q - p\|^2$ while maintaining the constraint $\int_0^1 p(t) dt = 0$.

Solution to 3.6

- (a) Let Q be the space of polynomials of degree n or less. Let $P \subset Q$ be the set of polynomials which integrate to zero, i.e. $p_0 \in P \implies \int_0^1 p_0(t) dt = 0$. N is a closed subspace of $L_2[0, 1]$, and P is a closed subspace of Q (and hence of $L_2[0, 1]$ also). Obviously $L_2[0, 1]$ is a Hilbert space, so the projection theorem tells us that there is a unique $p \in P$ such that $\forall p_0 \in P, \|x - p\| \leq \|x - p_0\|$. Note that $\int_0^1 |x(t) - q(t)|^2 dt = \|x - q\|^2$, and minimizing the norm is equivalent to minimizing its square.
- (b) Choose q and p as stated in the problem. By the projection theorem, we have that $x - q \perp Q$ and $q - p \perp P$. We want to show that $x - p \perp P$, so that the projection theorem tells us that $p \in P$ minimizes $\|x - p\|^2$. Since $0 \in P, Q$, we have the twin statements that

$$\forall q_0 \in Q, \quad \langle x - q | q_0 \rangle = 0 \quad \text{and} \quad \forall p_0 \in P, \quad \langle q - p | p_0 \rangle = 0.$$

Since for any $p_0 \in P$, we have $p - p_0 \in P \subset Q$, we have $\langle x - q | p - p_0 \rangle = 0$ and $\langle q - p | p - p_0 \rangle = 0$. Taking the difference of these two inner products gives our result, $\langle x - p | p - p_0 \rangle = 0$, or $x - p \perp P$. □

Problem 3.9 (Luenberger)

Prove $S^{\perp\perp} = \overline{[S]}$.

Solution to 3.9

As usual, we prove equality of the two sets by first showing that $S^{\perp\perp} \supseteq \overline{[S]}$, and then that $S^{\perp\perp} \subseteq \overline{[S]}$.

- $S^{\perp\perp} \supseteq \overline{[S]}$

By Prop. 1, part 1 (Luenberger, page 52), we know $S^{\perp\perp}$ is a closed subspace. Part 2 of the same proposition tells us that $S \subset S^{\perp\perp}$. By definition, $\overline{[S]}$ is the smallest closed subspace containing S . This means every closed subspace containing S also contains $\overline{[S]}$. Thus, $S^{\perp\perp} \supseteq \overline{[S]}$ as desired.

- $S^{\perp\perp} \subseteq \overline{[S]}$

It suffices to show that every closed subspace containing S must also contain $S^{\perp\perp}$. Let $B \supset S$ be a closed subspace. Then part 3 of Prop. 1 tells us $B^\perp \subset S^\perp$. Applying part 3 again yields $S^{\perp\perp} \subset B^{\perp\perp}$. But B is a closed subspace of a Hilbert space, and thus according to Theorem 1 (page 53) $B = B^{\perp\perp}$. So $S^{\perp\perp} \subset B$ for every closed B containing S . Thus $S^{\perp\perp}$ lies in the intersections of such B and is contained in $\overline{[S]}$, as desired. □

Problem 3.24 (Luenberger)

Let K be a closed, convex set in a Hilbert space H , and let $x \in H, x \notin K$ be given. Then there is a unique $k_0 \in K$ such that $\|x - k_0\| \leq \|x - k\|$ for all $k \in K$. Show that this is not true in an arbitrary Banach space.

Solution to 3.24

We will prove this via a counter-example. We choose $X = \ell_\infty$, the Banach space of bounded sequences. The sup-norm $\|\cdot\|_\infty$ is not derived from an inner product. We can show this from the parallelogram law: define $x = (1, 0, 0, \dots)$ and $y = (0, 1, 0, \dots)$. Then $\|x + y\|_\infty^2 + \|x - y\|_\infty^2 = 1 + 1 = 2$, which differs from $2\|x\|_\infty^2 + 2\|y\|_\infty^2 = 4$.

Let $K = [x] = (\alpha, 0, \dots)$, $\alpha \in \mathbb{R}$, that is, the set of elements with zeros everywhere after the first element. Let $y = (0, 1, 0, \dots)$. Then $\|y - k\|_\infty \geq 1$ for all $k \in K$, but $\|y - k\|_\infty = 1$ for all $k = (\alpha, 0, \dots)$, $\alpha \in [-1, 1]$. So the minimizer to K is not unique. □