

# MH5200 Advanced Investigations in Linear Algebra I

## Problem Sheet 1– Problems & Solutions

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### Overview of This Problem Sheet

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- **Problem 1:** Tridiagonal Toeplitz system; discrete 1D Laplacian with Dirichlet boundary; linear system solve (LU/Cholesky, spectral view).
- **Problem 2:** Parametrised linear system; singularity and infinite-solution conditions via rank, dependence of rows.
- **Problem 3:** Dimensions and indexed components of compositions; index-free vs. indexed formulations.
- **Problem 4:** Symmetry of a product of symmetric matrices; commuting condition; transpose of triangular matrices.
- **Problem 5:** Gram matrix as sum of rank-one outer products of rows; weighted Gram with diagonal scaling.
- **Problem 6:** Decomposition into symmetric and skew-symmetric parts; projections onto  $\pm 1$  eigenspaces of transpose.
- **Problem 7:** Orthogonal matrices; orthonormality of columns via  $Q^T Q = I$ ; geometric preservation of inner product.

# Problem 1

## Problem.

Consider the equations

$$-x_{i+1} + 2x_i - x_{i-1} = i$$

for  $i = 1, 2, 3, 4$  with  $x_0 = x_5 = 0$ . Write these equations in matrix form  $Ax = b$  where  $x = (x_1, x_2, x_3, x_4)^T$  and solve it to determine  $x_1, x_2, x_3, x_4$ .

(Notation:  $M^T$  denotes the transpose of the matrix  $M$ .)

## Solution

### Method 1: Direct / Elementary Approach

The system is

$$\begin{aligned} 2x_1 - x_2 &= 1, \\ -x_1 + 2x_2 - x_3 &= 2, \\ -x_2 + 2x_3 - x_4 &= 3, \\ -x_3 + 2x_4 &= 4, \end{aligned}$$

i.e.  $Ax = b$  with

$$A = \begin{bmatrix} 2 & -1 & 0 & 0 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ 0 & 0 & -1 & 2 \end{bmatrix}, \quad b = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}.$$

Perform Gaussian elimination (equivalently, Thomas algorithm for tridiagonal systems).

From the first equation,

$$2x_1 - x_2 = 1.$$

Eliminate  $x_1$  from the second equation:

$$-x_1 + 2x_2 - x_3 = 2 \Rightarrow \text{Row}_2 \leftarrow \text{Row}_2 + \frac{1}{2}\text{Row}_1 \Rightarrow \frac{3}{2}x_2 - x_3 = \frac{5}{2}.$$

Eliminate  $x_2$  from the third equation using the new second equation:

$$-x_2 + 2x_3 - x_4 = 3 \Rightarrow \text{Row}_3 \leftarrow \text{Row}_3 + \frac{2}{3}\text{Row}_2 \Rightarrow \frac{4}{3}x_3 - x_4 = \frac{14}{3}.$$

Eliminate  $x_3$  from the fourth equation using the new third equation:

$$-x_3 + 2x_4 = 4 \Rightarrow \text{Row}_4 \leftarrow \text{Row}_4 + \frac{3}{4}\text{Row}_3 \Rightarrow \frac{5}{4}x_4 = \frac{15}{2}.$$

Thus  $x_4 = 6$ . Back-substitute:

$$\begin{aligned} \frac{4}{3}x_3 - 6 &= \frac{14}{3} \Rightarrow x_3 = 8, \\ \frac{3}{2}x_2 - 8 &= \frac{5}{2} \Rightarrow x_2 = 7, \\ 2x_1 - 7 &= 1 \Rightarrow x_1 = 4. \end{aligned}$$

Hence

$$x = \begin{bmatrix} 4 \\ 7 \\ 8 \\ 6 \end{bmatrix}.$$

**Method 2: Factorisation / Spectral Approach**

The coefficient matrix is symmetric positive definite and admits an  $LU$  (indeed, Cholesky) factorisation. One convenient factorisation is

$$A = LU, \quad L = \begin{bmatrix} 1 & 0 & 0 & 0 \\ -1/2 & 1 & 0 & 0 \\ 0 & -2/3 & 1 & 0 \\ 0 & 0 & -3/4 & 1 \end{bmatrix}, \quad U = \begin{bmatrix} 2 & -1 & 0 & 0 \\ 0 & 3/2 & -1 & 0 \\ 0 & 0 & 4/3 & -1 \\ 0 & 0 & 0 & 5/4 \end{bmatrix}.$$

To solve  $Ax = b$ , first solve  $Ly = b$  by forward substitution:

$$\begin{aligned} y_1 &= 1, \\ -\frac{1}{2}y_1 + y_2 &= 2 \Rightarrow y_2 = \frac{5}{2}, \\ -\frac{2}{3}y_2 + y_3 &= 3 \Rightarrow y_3 = \frac{14}{3}, \\ -\frac{3}{4}y_3 + y_4 &= 4 \Rightarrow y_4 = \frac{15}{2}. \end{aligned}$$

Then solve  $Ux = y$  by backward substitution:

$$\begin{aligned} \frac{5}{4}x_4 &= \frac{15}{2} \Rightarrow x_4 = 6, \\ \frac{4}{3}x_3 - x_4 &= \frac{14}{3} \Rightarrow x_3 = 8, \\ \frac{3}{2}x_2 - x_3 &= \frac{5}{2} \Rightarrow x_2 = 7, \\ 2x_1 - x_2 &= 1 \Rightarrow x_1 = 4. \end{aligned}$$

Thus again  $x = (4, 7, 8, 6)^T$ .

## Problem 2

### Problem.

Consider the system of equations

$$\begin{aligned}x + 4y - 2z &= 1 \\x + 7y - 6z &= 6 \\3y + qz &= t\end{aligned}$$

for  $x, y, z$ . For what value of  $q$  is the system singular and which value of  $t$  results in infinitely many solutions? Characterize the solutions in this case.

### Solution

#### Method 1: Row-Reduction Approach

Subtract the first equation from the second:

$$(x + 7y - 6z) - (x + 4y - 2z) = 6 - 1 \Rightarrow 3y - 4z = 5.$$

For the system to be singular with infinitely many solutions, the third equation must be compatible with this relation and not impose a new independent constraint. Thus its left-hand side must match  $3y - 4z$ , so  $q = -4$ , and its right-hand side must equal 5, so  $t = 5$ .

Hence the system is singular when  $q = -4$ ; for  $q = -4, t = 5$  the three equations reduce to two independent ones.

To parameterise the solution set, set  $z = s \in \mathbb{R}$ . Then

$$3y - 4s = 5 \Rightarrow y = \frac{5 + 4s}{3},$$

and from the first equation,

$$x + 4y - 2s = 1 \Rightarrow x = 1 - 4y + 2s = 1 - 4 \cdot \frac{5 + 4s}{3} + 2s = \frac{-17 - 10s}{3}.$$

Thus

$$(x, y, z) = \left( \frac{-17 - 10s}{3}, \frac{5 + 4s}{3}, s \right), \quad s \in \mathbb{R}.$$

For example, choosing  $s = 1$  gives the particular solution  $(-9, 3, 1)$ .

#### Method 2: Rank / Linear Dependence Approach

Write the coefficient rows as

$$r_1 = (1, 4, -2), \quad r_2 = (1, 7, -6), \quad r_3 = (0, 3, q).$$

We have

$$r_2 - r_1 = (0, 3, -4).$$

The system is singular precisely when  $r_3$  lies in the span of  $\{r_1, r_2\}$  in such a way that the three equations become linearly dependent. We seek  $\lambda$  with

$$r_3 = \lambda(r_2 - r_1) = \lambda(0, 3, -4).$$

Comparing components, this forces  $3 = \lambda \cdot 3$  and  $q = \lambda \cdot (-4)$ , so  $\lambda = 1$  and  $q = -4$ .

For infinite solutions the augmented rows must also be dependent with the same linear combination. The right-hand sides are 1, 6, and  $t$ ; thus

$$t = 1 \cdot (6 - 1) = 5.$$

When  $q = -4, t = 5$ , the rank of the coefficient matrix is 2 while the rank of the augmented matrix is also 2. With three variables,  $\dim \ker = 3 - 2 = 1$ , so the solution set is a line in  $\mathbb{R}^3$ , which can be parameterised as above.

## Problem 3

### Problem.

Let  $A = (a_{ij})$  be an  $m \times n$  matrix and  $B = (b_{ij})$  be a  $p \times m$  matrix. Let  $x = (x_i)$  be an  $n$ -dimensional column vector.

- (a) What is the dimension of  $BAx$ ? Write down an expression for the  $i$ -th element of  $BAx$ .
- (b) Let  $y = (y_i)$  be a  $p$ -dimensional column vector. What is the dimension of  $y^T BA$ ? Write down an expression for its  $i$ -th element.

## Solution

### Method 1: Indexed Computation

- (a) First  $Ax$  is an  $m \times 1$  vector, then  $BAx$  is a  $p \times 1$  vector. Its  $i$ -th component is

$$(BAx)_i = \sum_{j=1}^m b_{ij}(Ax)_j = \sum_{j=1}^m b_{ij} \left( \sum_{k=1}^n a_{jk} x_k \right) = \sum_{j=1}^m \sum_{k=1}^n b_{ij} a_{jk} x_k.$$

- (b)  $y^T$  is  $1 \times p$ , so  $y^T BA$  is  $1 \times n$ . Its  $i$ -th component is

$$(y^T BA)_i = \sum_{j=1}^p y_j (BA)_{ji} = \sum_{j=1}^p \sum_{k=1}^m y_j b_{jk} a_{ki}.$$

### Method 2: Operator / Index-Free View

Interpret

$$A : \mathbb{R}^n \rightarrow \mathbb{R}^m, \quad B : \mathbb{R}^m \rightarrow \mathbb{R}^p.$$

Then  $BA : \mathbb{R}^n \rightarrow \mathbb{R}^p$ , so for  $x \in \mathbb{R}^n$ ,  $BAx \in \mathbb{R}^p$ : a  $p$ -dimensional column vector, consistent with the indexed computation.

For  $y \in \mathbb{R}^p$ ,  $y^T BA$  is the covector in  $(\mathbb{R}^n)^T$  represented by

$$y^T BA = (A^T B^T y)^T,$$

so as a row vector it has dimension  $1 \times n$ . If  $e_i$  is the  $i$ -th standard basis vector in  $\mathbb{R}^n$ , then

$$(y^T BA)_i = y^T B A e_i = \langle B^T y, A e_i \rangle,$$

and expanding in the standard basis recovers the indexed formula

$$(y^T BA)_i = \sum_{j=1}^p \sum_{k=1}^m y_j b_{jk} a_{ki}.$$

## Problem 4

**Problem.**

- (a) Let  $A$  and  $B$  be two  $n \times n$  symmetric matrices. Is the product  $AB$  a symmetric matrix? If not, characterize the conditions under which it is a symmetric matrix.
- (b) Prove that the transpose of an upper triangular matrix is lower triangular.

## Solution

### Method 1: Direct Transpose and Index Argument

- (a) In general  $AB$  is not symmetric. Since  $A^T = A$  and  $B^T = B$ , we have

$$(AB)^T = B^T A^T = BA.$$

Thus  $AB$  is symmetric if and only if

$$AB = (AB)^T \iff AB = BA,$$

i.e.  $A$  and  $B$  commute.

- (b) Let  $U = (u_{ij})$  be an upper triangular matrix, so  $u_{ij} = 0$  whenever  $i > j$ . Its transpose  $U^T = (v_{ij})$  satisfies

$$v_{ij} = (U^T)_{ij} = u_{ji}.$$

If  $i < j$ , then  $v_{ij} = u_{ji}$  with  $j > i$ , so  $u_{ji} = 0$  because  $U$  is upper triangular. Hence  $v_{ij} = 0$  for all  $i < j$ , which means  $U^T$  is lower triangular.

### Method 2: Spectral / Structural View

- (a) If  $A$  and  $B$  are real symmetric and commute, the spectral theorem implies there exists an orthogonal matrix  $Q$  such that

$$Q^T A Q = \Lambda, \quad Q^T B Q =$$

are diagonal. Then

$$Q^T (AB) Q = (Q^T A Q)(Q^T B Q) = \Lambda,$$

which is diagonal, hence symmetric. Therefore  $AB$  is symmetric. Conversely, if  $AB$  is symmetric, then

$$AB = (AB)^T = B^T A^T = BA,$$

so commutativity is necessary and sufficient.

- (b) The set of upper triangular matrices forms a subspace of all  $n \times n$  matrices. The transpose map  $T(X) = X^T$  is a linear isomorphism that reverses the order of indices. The image of the subspace of upper triangular matrices under  $T$  is precisely the subspace of lower triangular matrices; therefore the transpose of any upper triangular matrix must be lower triangular.

## Problem 5

### Problem.

Suppose the matrix  $A$  has rows  $a_1^T, \dots, a_m^T$ . Show that

$$A^T A = a_1 a_1^T + \dots + a_m a_m^T.$$

If  $C$  is a diagonal matrix with diagonal elements  $c_1, \dots, c_m$ , find a similar expression for  $A^T C A$ .

### Solution

#### Method 1: Entrywise Verification

Write

$$A = \begin{bmatrix} a_1^T \\ \vdots \\ a_m^T \end{bmatrix},$$

so the  $(\ell, i)$ -entry of  $A$  is  $(a_\ell)_i$ . Then

$$(A^T A)_{ij} = \sum_{\ell=1}^m A_{\ell i} A_{\ell j} = \sum_{\ell=1}^m (a_\ell)_i (a_\ell)_j.$$

On the other hand,

$$\left( \sum_{\ell=1}^m a_\ell a_\ell^T \right)_{ij} = \sum_{\ell=1}^m (a_\ell a_\ell^T)_{ij} = \sum_{\ell=1}^m (a_\ell)_i (a_\ell)_j.$$

Thus every entry matches and

$$A^T A = \sum_{\ell=1}^m a_\ell a_\ell^T.$$

If  $C = \text{diag}(c_1, \dots, c_m)$ , then

$$(A^T C A)_{ij} = \sum_{\ell=1}^m C_{\ell\ell} A_{\ell i} A_{\ell j} = \sum_{\ell=1}^m c_\ell (a_\ell)_i (a_\ell)_j,$$

which is exactly the  $(i, j)$ -entry of

$$\sum_{\ell=1}^m c_\ell a_\ell a_\ell^T.$$

Hence

$$A^T C A = \sum_{\ell=1}^m c_\ell a_\ell a_\ell^T.$$

**Method 2: Gram Matrix / Inner-Product View**

View each row  $a_\ell^T$  as a vector  $a_\ell \in \mathbb{R}^n$ . Then  $A^T A$  is the Gram matrix of the rows:

$$(A^T A)_{ij} = \langle e_i, A^T A e_j \rangle = \sum_{\ell=1}^m \langle a_\ell, e_i \rangle \langle a_\ell, e_j \rangle = \sum_{\ell=1}^m (a_\ell)_i (a_\ell)_j.$$

This shows that

$$A^T A = \sum_{\ell=1}^m a_\ell a_\ell^T$$

is a sum of rank-one positive semidefinite operators. If  $C$  is diagonal with entries  $c_\ell \geq 0$ , then

$$A^T C A = (\sqrt{C} A)^T (\sqrt{C} A)$$

is the Gram matrix of the scaled rows  $\sqrt{c_\ell} a_\ell$ , giving

$$A^T C A = \sum_{\ell=1}^m c_\ell a_\ell a_\ell^T.$$

## Problem 6

### Problem.

$S$  is said to be a symmetric matrix if  $S^T = S$  and  $A$  is anti-symmetric (or skew-symmetric) if  $A^T = -A$ . Write down an expression for a general matrix  $M$  as a sum of a symmetric part and an anti-symmetric part:  $M = S + A$ . You should specify  $S$  and  $A$  in terms of  $M$ .

### Solution

#### Method 1: Direct Construction

Define

$$S := \frac{1}{2}(M + M^T), \quad A := \frac{1}{2}(M - M^T).$$

Then

$$S^T = \frac{1}{2}(M^T + M) = S, \quad A^T = \frac{1}{2}(M^T - M) = -A,$$

so  $S$  is symmetric and  $A$  is skew-symmetric. Moreover,

$$S + A = \frac{1}{2}(M + M^T) + \frac{1}{2}(M - M^T) = M.$$

Thus  $M = S + A$  with the required properties.

#### Method 2: Projection via Transpose Involution

Consider the linear map  $T : \mathbb{R}^{n \times n} \rightarrow \mathbb{R}^{n \times n}$ ,  $T(X) = X^T$ . Then  $T^2 = \text{id}$ , so its eigenvalues are  $\pm 1$ . The eigenspaces are

$$\mathcal{S} = \{X : X^T = X\}, \quad \mathcal{A} = \{X : X^T = -X\},$$

the symmetric and skew-symmetric matrices. The operators

$$P_{\mathcal{S}}(M) = \frac{1}{2}(M + M^T), \quad P_{\mathcal{A}}(M) = \frac{1}{2}(M - M^T)$$

are the projections onto  $\mathcal{S}$  and  $\mathcal{A}$  along the other eigenspace, and  $M = P_{\mathcal{S}}(M) + P_{\mathcal{A}}(M)$  gives the desired decomposition.

## Problem 7

### Problem.

Suppose that  $Q^T$  equals  $Q^{-1}$ . Such a matrix is called an orthogonal matrix:  $Q^T Q = I$ .

- (a) Show that the columns  $q_i$  of  $Q$  are unit vectors:  $\|q_i\| = 1$ .
- (b) Show that every pair of columns of  $Q$  are orthogonal:  $q_i^T q_j = 0$ , for  $i \neq j$ .

### Solution

#### Method 1: Entrywise Argument from $Q^T Q = I$

Let  $q_1, \dots, q_n$  be the columns of  $Q$ . The  $(i, j)$ -entry of  $Q^T Q$  is the inner product of the  $i$ -th and  $j$ -th columns:

$$(Q^T Q)_{ij} = q_i^T q_j.$$

Since  $Q^T Q = I$ , we have

$$q_i^T q_j = \delta_{ij} = \begin{cases} 1, & i = j, \\ 0, & i \neq j. \end{cases}$$

Thus each column satisfies  $\|q_i\|^2 = q_i^T q_i = 1$ , so it is a unit vector, and distinct columns satisfy  $q_i^T q_j = 0$ , so they are mutually orthogonal.

#### Method 2: Inner-Product Preservation

For any  $v, w \in \mathbb{R}^n$ ,

$$\langle Qv, Qw \rangle = (Qv)^T (Qw) = v^T (Q^T Q) w = v^T w = \langle v, w \rangle,$$

so  $Q$  preserves the Euclidean inner product. Taking  $v = e_i$ ,  $w = e_j$  (standard basis vectors), we have

$$\langle Qe_i, Qe_j \rangle = \langle e_i, e_j \rangle = \delta_{ij}.$$

But  $Qe_i = q_i$  and  $Qe_j = q_j$ , so

$$q_i^T q_j = \delta_{ij}.$$

Hence each column has unit norm and different columns are orthogonal, as required.