

Solutions to Homework



December 2, 2023

Contents

1	Solutions to Homework-2	1
2	Solutions to Homework-4	8
3	Solutions to Homework-6	12
4	Solutions to Homework-8	16





Chapter 1

Solutions to Homework-2

Exercise. Show that two different reduced row echelon system of linear equations have different solutions (if the solutions exist). Derive that the reduced row echelon matrix associated to a given matrix is unique.

Proof. For the first part, suppose reduced row echelon systems $A_1X = 0$, $A_2X = 0$ have the same solutions and show the row echelon system. Then the number of pivots and free unknowns are same, and the solutions of $A_1X = 0$, $A_2X = 0$ are given by the combinations of these unknowns and entries of A_1 and A_2 respectively. As a result $A_1 = A_2$ since $A_1X = 0$ and $A_2X = 0$ have the same solutions.

For the second part, if A is reduced to A_1 and A_2 , then $A_1X = 0$ has the same solutions as $A_2X = 0$, since both of them have the same solutions as $AX = 0$, and thus $A_1 = A_2$. \square

Exercise. Find all solutions of the equation $x_1 + x_2 + 2x_3 - x_4 = 3$.

Proof. All possible solutions in \mathbb{R} are given by

$$x = \begin{pmatrix} a \\ b \\ c \\ a + b + 2c - 3 \end{pmatrix},$$

where $a, b, c \in \mathbb{R}$. \square

Exercise. Find all solutions of the system of equations $AX = B$ when

$$A = \begin{pmatrix} 1 & 2 & 1 & 1 \\ 3 & 0 & 0 & 4 \\ 1 & -4 & -2 & 2 \end{pmatrix}$$

and

(a)

$$B = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

(b)

$$B = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$$



(c)

$$B = \begin{pmatrix} 0 \\ 2 \\ 2 \end{pmatrix}$$

Proof. For (a). By Gaussian elimination one has

$$[A \mid B] = \begin{pmatrix} 1 & 2 & 1 & 1 & 0 \\ 3 & 0 & 0 & 4 & 0 \\ 1 & 4 & -2 & 2 & 0 \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & 2 & 1 & 1 & 0 \\ 0 & -6 & -3 & 1 & 0 \\ 0 & 2 & -3 & 1 & 0 \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & 2 & 1 & 1 & 0 \\ 0 & -6 & -3 & 1 & 0 \\ 0 & 0 & -4 & \frac{4}{3} & 0 \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & 2 & 1 & 1 & 0 \\ 0 & -6 & -3 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

This shows all possible solutions over \mathbb{R} is given by

$$x = \begin{pmatrix} -\frac{3}{4}b \\ \frac{1}{6}(-3a + b) \\ a \\ b \end{pmatrix},$$

where $a, b \in \mathbb{R}$. Similarly one can show the solutions of (b) is empty set and (c) are given by

$$x = \begin{pmatrix} \frac{3}{2} - \frac{4}{3}b \\ -\frac{1}{3} - \frac{1}{2}a + \frac{1}{6}b \\ a \\ b \end{pmatrix},$$

where $a, b \in \mathbb{R}$. □

Exercise. Find the inverse of the following matrix by elementary row reduction:

$$\begin{pmatrix} 1 & & & & \\ 1 & 1 & & & \\ 1 & 2 & 1 & & \\ 1 & 3 & 3 & 1 & \\ 1 & 4 & 6 & 4 & 1 \end{pmatrix}$$

Proof. The inverse of above matrix is given by

$$\begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 & 0 \\ 1 & -2 & 1 & 0 & 0 \\ -1 & 3 & -3 & 1 & 0 \\ 1 & -4 & 6 & -4 & 1 \end{pmatrix}$$

□

Exercise. Write the following permutations as products of transpositions, and determine their sign.

(a) 1, 3, 5, 2, 4, 8, 6, 7,

(b) 9, 5, 3, 8, 4, 6, 2, 1, 7,

(c) 7, 1, 6, 2, 5, 3, 4.



Proof. For (a). Firstly note that the permutation

$$\begin{pmatrix} 1, 2, 3, 4, 5, 6, 7, 8 \\ 1, 3, 5, 2, 4, 8, 6, 7 \end{pmatrix}$$

can be written as $(1)(2, 3, 5, 4)(6, 8, 7)$, and thus it can be written as products of transpositions as follows

$$(1)(23)(25)(24)(68)(67),$$

which implies it's an odd permutation. By the same method one can write the other permutations as products of transpositions. \square

Exercise. Evaluate the following determinants:

(a) $\begin{pmatrix} 1 & i \\ 2-i & 3 \end{pmatrix},$

(b) $\begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix},$

(c) $\begin{pmatrix} 2 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 2 \end{pmatrix},$

(d) $\begin{pmatrix} 1 & 0 & 0 & 0 \\ 5 & 2 & 0 & 0 \\ 8 & 6 & 3 & 0 \\ 0 & 9 & 7 & 4 \end{pmatrix}.$

Proof. The determinants are $2 - 2i, -2, 3, 24$ respectively. \square

Exercise. Compute the determinant of the following $n \times n$ matrix using induction on n :

$$\begin{pmatrix} 2 & -1 & & & \\ -1 & 2 & -1 & & \\ & -1 & 2 & -1 & \\ & & -1 & \ddots & \\ & & & \ddots & 2 & -1 \\ & & & & -1 & 2 \end{pmatrix}.$$

Proof. Let T_n denote above $n \times n$ matrix. It's clear $\det T_1 = 2$ and $\det T_2 = 3$. Now let's prove $\det T_n = n + 1$ by induction. Suppose it holds for $n < k$. Then for $n = k$, by expansion along the first row one has

$$\det T_k = 2 \det T_{k-1} - (-1) \times (-1) \det T_{k-2} = 2k - (k - 1) = k + 1.$$

\square

Exercise. Suppose $R = K[t]$ is the polynomial ring and $A = (a_{ij}) \in M_n(R)$. Show that

(a)

$$\frac{\partial \det A}{\partial t} = \det A_1 + \cdots + \det A_n,$$

where A_i is obtained from A by taking the derivative of the i -th row and keep the other rows.



(b)

$$\frac{\partial \det A}{\partial t} = \sum_{i,j} (-1)^{i+j} \frac{\partial a_{ij}}{\partial t} \det a_{ij},$$

where $a_{ij} = A^c \begin{pmatrix} i \\ j \end{pmatrix}$.

Proof. For (a). Note that

$$\begin{aligned} \frac{\partial \det A}{\partial t} &= \frac{\partial}{\partial t} \left(\sum_{\sigma \in S_n} \text{sgn}(\sigma) a_{1\sigma(1)} \cdots a_{n\sigma(n)} \right) \\ &= \sum_k \sum_{\sigma \in S_n} \text{sgn}(\sigma) \left(a_{1\sigma(1)} \cdots \frac{\partial a_{k\sigma(k)}}{\partial t} \cdots a_{n\sigma(n)} \right) \\ &= \det A_1 + \cdots + \det A_n. \end{aligned}$$

For (b). It suffices to note that

$$A_k = \sum_i (-1)^{k+j} \frac{\partial a_{kj}}{\partial t} A_{k,j}.$$

Then by (a), one has

$$\det A = \det A_1 + \cdots + \det A_n = \sum_{i,j} (-1)^{i+j} \frac{\partial a_{ij}}{\partial t} A_{i,j}$$

□

Exercise. Suppose $\det \begin{pmatrix} x & y & z \\ 3 & 0 & 2 \\ 1 & 1 & 1 \end{pmatrix} = 1$. Compute the following determinant.

$$(1) \begin{pmatrix} 2x & 2y & 2z \\ \frac{3}{2} & 0 & 1 \\ 1 & 1 & 1 \end{pmatrix}.$$

$$(2) \begin{pmatrix} x & y & z \\ 3x+3 & 3y & 3z+2 \\ x+1 & y+1 & z+1 \end{pmatrix}.$$

$$(3) \begin{pmatrix} x-1 & y-1 & z-1 \\ 4 & 1 & 3 \\ 1 & 1 & 1 \end{pmatrix}.$$

Proof. By elementary operations, one can see all of above three determinants equal to the determinant of

$$\begin{pmatrix} x & y & z \\ 3 & 0 & 2 \\ 1 & 1 & 1 \end{pmatrix}$$

□

Exercise. Calculate the Vandermonde determinant

$$\det \begin{pmatrix} 1 & 1 & \cdots & 1 \\ \theta_1 & \theta_2 & \cdots & \theta_n \\ \vdots & \vdots & \ddots & \vdots \\ \theta_1^{n-1} & \theta_2^{n-1} & \cdots & \theta_n^{n-1} \end{pmatrix}.$$



Proof. Now let's prove the Vandermonde determinant equals $\prod_{1 \leq i < j \leq n} (\theta_j - \theta_i)$ by induction. It holds for $n = 2$, and suppose it holds for $n < k$. Let V denote the Vandermonde matrix. Then

$$\begin{aligned} \det V &= \det \begin{pmatrix} 1 & 1 & \cdots & 1 \\ 0 & \theta_2 - \theta_1 & \cdots & \theta_k - \theta_1 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & \theta_2^{k-1} - \theta_1^{k-1} & \cdots & \theta_k^{k-1} - \theta_1^{k-1} \end{pmatrix} \\ &= \det \begin{pmatrix} \theta_2 - \theta_1 & \cdots & \theta_k - \theta_1 \\ \theta_2^2 - \theta_1^2 & \cdots & \theta_k^2 - \theta_1^2 \\ \vdots & \ddots & \vdots \\ \theta_2^{k-1} - \theta_1^{k-1} & \cdots & \theta_k^{k-1} - \theta_1^{k-1} \end{pmatrix} \\ &= \det \begin{pmatrix} 1 & 1 & \cdots & 1 \\ \theta_2 + \theta_1 & \theta_3 + \theta_1 & \cdots & \theta_k + \theta_1 \\ \vdots & \vdots & \ddots & \vdots \\ \sum_{i=0}^{k-2} \theta_2^{k-2-i} \theta_1^i & \sum_{i=0}^{k-2} \theta_3^{k-2-i} \theta_1^i & \cdots & \sum_{i=0}^{k-2} \theta_k^{k-2-i} \theta_1^i \end{pmatrix} \begin{pmatrix} \theta_2 - \theta_1 & 0 & \cdots & 0 \\ 0 & \theta_3 - \theta_1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \theta_k - \theta_1 \end{pmatrix} \end{aligned}$$

Note that

$$\begin{aligned} &\begin{pmatrix} 1 & 1 & \cdots & 1 \\ \theta_2 + \theta_1 & \theta_3 + \theta_1 & \cdots & \theta_k + \theta_1 \\ \vdots & \vdots & \ddots & \vdots \\ \sum_{i=0}^{k-2} \theta_2^{k-2-i} \theta_1^i & \sum_{i=0}^{k-2} \theta_3^{k-2-i} \theta_1^i & \cdots & \sum_{i=0}^{k-2} \theta_k^{k-2-i} \theta_1^i \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 & \cdots & 0 \\ \theta_1 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & 0 \\ \theta_1^{n-2} & \theta_1^{n-3} & \cdots & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 & \cdots & 1 \\ \theta_2 & \theta_3 & \cdots & \theta_k \\ \vdots & \vdots & \ddots & \vdots \\ \theta_2^{k-2} & \theta_3^{k-2} & \cdots & \theta_k^{k-2} \end{pmatrix}. \end{aligned}$$

Then by induction hypothesis one has

$$\det V = \prod_{j=2}^k (x_j - x_1) \prod_{2 \leq i < j \leq k} (x_j - x_i) = \prod_{1 \leq i < j \leq k} (x_j - x_i)$$

as desired. \square

Exercise. Let a $2n \times 2n$ matrix be given in the form $M = \begin{pmatrix} A & B \\ C & D \end{pmatrix}$, where each block is an $n \times n$ matrix. Suppose that A is invertible and that $AC = CA$. Use block multiplication to prove that $\det M = \det(AD - CB)$. Give an example to show that this formula need not hold if $AC \neq CA$.

Proof. Note that

$$\begin{aligned} \det M &= \det \begin{pmatrix} A & B \\ C & D \end{pmatrix} \\ &= \det \begin{pmatrix} A & B \\ O & D - CA^{-1}B \end{pmatrix} \\ &= \det A \cdot \det(D - CA^{-1}B) \\ &= \det(AD - ACA^{-1}B) \\ &= \det(AD - CB). \end{aligned}$$

\square



Exercise. Suppose $a_{ii} > 0$ and $a_{ij} < 0$ for $i \neq j$. Suppose in addition that $\sum_{i=1}^n a_{ij} > 0$ for all j . Show that $\det(a_{ij}) > 0$.

Proof. Let's prove this by induction. For $n = 1$,

$$\det(a_{ij}) = a_{11} > 0.$$

Now suppose it holds for $n < k$. Then for $n = k$, note that

$$\begin{aligned} \det \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1k} \\ a_{21} & a_{22} & \cdots & a_{2k} \\ \vdots & \vdots & & \vdots \\ a_{k1} & a_{k2} & \cdots & a_{kk} \end{pmatrix} &= \det \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1k} \\ 0 & a_{22} - \frac{a_{12}}{a_{11}}a_{21} & \cdots & a_{2k} \\ \vdots & \vdots & & \vdots \\ 0 & a_{k2} - \frac{a_{12}}{a_{11}}a_{k1} & \cdots & a_{kk} \end{pmatrix} \\ &= a_{11} \det \begin{pmatrix} a_{22} - \frac{a_{12}}{a_{11}}a_{21} & \cdots & a_{2k} \\ \vdots & & \vdots \\ a_{k2} - \frac{a_{12}}{a_{11}}a_{k1} & \cdots & a_{kk} \end{pmatrix} \end{aligned}$$

Note that $a_{22} - \frac{a_{12}}{a_{11}}a_{21} > 0$ and $-\frac{a_{12}}{a_{11}}(a_{21} + \cdots + a_{k1}) > -\frac{a_{12}}{a_{11}}(-a_{11}) = a_{12}$. Then by induction hypothesis

$$\det \begin{pmatrix} a_{22} - \frac{a_{12}}{a_{11}}a_{21} & \cdots & a_{2k} \\ \vdots & & \vdots \\ a_{k2} - \frac{a_{12}}{a_{11}}a_{k1} & \cdots & a_{kk} \end{pmatrix} > 0.$$

This completes the proof. □

Exercise. Suppose $A \in M_{n \times s}(R)$ and $B \in M_{s \times n}(R)$. Prove

$$\det(AB) = \begin{cases} 0, & n > s; \\ \det A \cdot \det B, & n = s; \\ \sum_{1 \leq k_1 < k_2 < \cdots < k_n \leq s} \det A \begin{pmatrix} 1 & 2 & \cdots & n \\ k_1 & k_2 & \cdots & k_n \end{pmatrix} \cdot \det B \begin{pmatrix} k_1 & k_2 & \cdots & k_n \\ 1 & 2 & \cdots & n \end{pmatrix}, & n < s. \end{cases}$$

Proof. If $n > s$, then there exists a non-zero x such that $Bx = 0$, and thus $ABx = 0$. This shows the system of linear equations $ABX = 0$ has a non-zero solution, and thus $\det AB = 0$. If $n = s$, then both A, B are square matrices, so

$$\det AB = \det A \cdot \det B$$

by properties of determinants. If $n < s$, consider the matrix

$$M = \begin{pmatrix} A & O \\ I_s & B \end{pmatrix},$$

which is a $(n + s) \times (n + s)$ matrix. On one hand, one has

$$\det M = \det \begin{pmatrix} A & O \\ I_s & B \end{pmatrix} = \det \begin{pmatrix} O & -AB \\ I_s & B \end{pmatrix} = (-1)^{ns+n} \det AB.$$

On the other hand, by Laplacian expansion one has

$$\det M = \sum_{1 \leq k_1 < \cdots < k_n \leq s} (-1)^{\frac{n(n+1)}{2} + k_1 + \cdots + k_n} \det A \begin{pmatrix} 1 & 2 & \cdots & n \\ k_1 & k_2 & \cdots & k_n \end{pmatrix} \det M \begin{pmatrix} n+1 & n+2 & \cdots & n+s \\ k_{n+1} & k_{n+2} & \cdots & k_{n+s} \end{pmatrix},$$



where $\{k_1, \dots, k_{n+s}\}$ is a permutation of $\{1, \dots, n+s\}$. Note that among the first n rows, the last n columns are zeros, so if

$$\det A \begin{pmatrix} 12 \dots n \\ k_1 k_2 \dots k_n \end{pmatrix} \neq 0,$$

we must have $1 \leq k_1, \dots, k_n \leq s$. In particular, one has

$$M \begin{pmatrix} n+1n+2 \dots n+s \\ k_{n+1}k_{n+2} \dots k_{n+s} \end{pmatrix} = \begin{pmatrix} I_s \begin{pmatrix} 12 \dots s \\ \mu_1, \dots, \mu_{s-n} \end{pmatrix} & B \end{pmatrix},$$

where $\{k_1, \dots, k_n\} \cup \{\mu_1, \dots, \mu_{s-n}\} = \{1, 2, \dots, s\}$. Again by Laplacian expansion one has

$$\det M \begin{pmatrix} n+1n+2 \dots n+s \\ k_{n+1}k_{n+2} \dots k_{n+s} \end{pmatrix} = (-1)^{\frac{(s-n)(s-n+1)}{2} + \mu_1 + \dots + \mu_{s-n}} \det B \begin{pmatrix} k_1 k_2 \dots k_n \\ 12 \dots n \end{pmatrix}.$$

Then

$$\det M = (-1)^{\frac{n(n+1)}{2} + \frac{(s-n)(s-n+1)}{2} + k_1 + \dots + k_n + \mu_1 + \dots + \mu_{s-n}} \sum_{1 \leq k_1 < \dots < k_n \leq s} \det A \begin{pmatrix} 12 \dots n \\ k_1 k_2 \dots k_n \end{pmatrix} \cdot \det B \begin{pmatrix} k_1 k_2 \dots k_n \\ 12 \dots n \end{pmatrix}.$$

Note that

$$\begin{aligned} \frac{n(n+1)}{2} + \frac{(s-n)(s-n+1)}{2} + k_1 + \dots + k_n + \mu_1 + \dots + \mu_{s-n} &= n^2 + s^2 + s - ns \\ &\equiv ns + n \pmod{2}. \end{aligned}$$

This completes the proof. □



Chapter 2

Solutions to Homework-4

Exercise. Show that $\text{rank}(AB) = \text{rank } B$ if and only if the solution space of $AB\mathbf{x} = \mathbf{0}$ is the same as the solution space of $B\mathbf{x} = \mathbf{0}$. Moreover, show that in this case, for any C , we have $\text{rank}(ABC) = \text{rank}(BC)$ whenever the product is well defined.

Proof. Firstly it's clear the solution space of $B\mathbf{x} = \mathbf{0}$ is included in the one of $AB\mathbf{x} = \mathbf{0}$. In other words, $\ker B \subseteq \ker AB$. Then $\ker A = \ker AB$ if and only if $\dim \ker B = \dim \ker AB$, which is equivalent to $\text{rank } B = \text{rank } AB$ since $\text{rank } B = n - \dim \ker B$ and $\text{rank } AB = n - \dim \ker AB$, where n is the number of columns of B .

In the case of $\ker B = \ker AB$, suppose $C = (v_1, \dots, v_m)$ with $v_i \in \mathbb{R}^n$. Then $Bv_{i_1}, \dots, Bv_{i_k}$ are linearly independent if and only if

$$c_1 v_{i_1} + \dots + c_k v_{i_k} \in \ker B \implies c_1 = \dots = c_k = 0.$$

But $\ker B = \ker AB$, this shows that $Bv_{i_1}, \dots, Bv_{i_k}$ are linearly independent if and only if $ABv_{i_1}, \dots, ABv_{i_k}$ are linearly independent. This shows $\text{rank } ABC = \text{rank } BC$. \square

Exercise. Suppose $A \in M_n(\mathbb{R})$. Show that $\text{rank}(A^t A) = \text{rank } A$.

Proof. It suffices to show that the equations $A\mathbf{x} = \mathbf{0}$ and $A^t A\mathbf{x} = \mathbf{0}$ have the same solution space. It's clear that $A\mathbf{x} = \mathbf{0}$ implies $A^t A\mathbf{x} = \mathbf{0}$. On the other hand, if $A^t A\mathbf{x} = \mathbf{0}$, then

$$\mathbf{x}^t A^t A \mathbf{x} = (A\mathbf{x})^t A \mathbf{x} = 0.$$

This shows $A\mathbf{x} = \mathbf{0}$. \square

Exercise. Find a basis of the space of symmetric and skew-symmetric matrices over a field K , and compute their dimensions.

Proof. It depends on the characteristic of the field K . If $\text{char } K \neq 2$, then we have already shown in the Homework3 that

$$\{E_{ij} + E_{ji}\}_{i \neq j} \cup \{E_{ii}\}$$

gives a basis of the space of symmetric matrices, and thus the dimension of the space of symmetric matrices is $n(n+1)/2$. On the other hand,

$$\{E_{ij} - E_{ji}\}_{i \neq j}$$

gives a basis of the space of skew-symmetric matrices, and thus the dimension of the space of skewsymmetric matrices is $n(n-1)/2$. However, if $\text{char } K = 2$, then the space of symmetric matrices and skew-symmetric matrices coincide, both have dimension $n(n+1)/2$. \square

Exercise. Let $K = \mathbb{Z}_p$ be a finite field with p elements, where p is a prime. For positive integer n , compute the number of different basis of K^n .



Proof. Note that it suffices to compute $\# \text{GL}(n, \mathbb{Z}_p)$, since any two basis of K^n differs a unique element in $\text{GL}(n, \mathbb{Z}_p)$. (In other words if you like, $\text{GL}(n, \mathbb{Z}_p)$ acts on the set of basis of K^n transitively with trivial stabilizer.)

For $A \in \text{GL}(n, \mathbb{Z}_p)$, there are $p^n - 1$ choices for the first row, and if we have fixed the first columns, there are $p^n - p$ choices for the second row since the second row has to be linearly independent with the first column. Repeat above arguments one can see there are

$$(p^n - 1)(p^n - p) \dots (p^n - p^{n-1}).$$

□

Exercise.

- (a) Prove that the set $\mathbf{B} = ((1, 2, 0)^t, (2, 1, 2)^t, (3, 1, 1)^t)$ is a basis of \mathbb{R}^3 .
- (b) Find the coordinate vector of the vector $v = (1, 2, 3)^t$ with respect to this basis.
- (c) Let $\mathbf{B}' = ((0, 1, 0)^t, (1, 0, 1)^t, (2, 1, 0)^t)$. Determine the basechange matrix P from \mathbf{B} to \mathbf{B}' .

Proof. For (a). It suffices to compute the determinant of the matrix given by this basis.

For (b). It suffices to solve a system of linear equations.

For (c). It suffices to solve three systems of linear equations.

□

Exercise. Let U, V, W be three subspaces of a vector space. Is the following formula correct? Find a proof or a counterexample.

$$\begin{aligned} \dim(U + V + W) &= \dim(U) + \dim(V) + \dim(W) \\ &\quad - \dim(U \cap V) - \dim(U \cap W) - \dim(V \cap W) \\ &\quad + \dim(U \cap V \cap W) \end{aligned}$$

Proof. It's wrong. Just consider the

$$U = \{(x, 0) \mid x \in \mathbb{R}\}, \quad V = \{(x, 0) \mid x \in \mathbb{R}\}, \quad W = \{(x, x) \mid x \in \mathbb{R}\}.$$

Then

$$U \cap V = U \cap W = V \cap W = U \cap V \cap W = \{0\}$$

But

$$\dim(U + V + W) = 2 \neq 3 = \dim U + \dim V + \dim W.$$

□

Exercise. Consider the linear transform: $T: \mathbb{R}^3 \rightarrow \mathbb{R}^2$ such that $(x_1, x_2, x_3)^t \mapsto (x_1 + 2x_2, x_1 - x_2)^t$. Compute the matrix of T with respect to the basis $\alpha_1, \alpha_2, \alpha_3$ of \mathbb{R}^3 and β_1, β_2 of \mathbb{R}^2 :

- (a) $\alpha_1 = (1, 0, 0)^T, \alpha_2 = (0, 1, 0)^T, \alpha_3 = (0, 0, 1)^T; \beta_1 = (1, 0)^T, \beta_2 = (0, 1)^T;$
- (b) $\alpha_1 = (1, 1, 1)^T, \alpha_2 = (0, 1, 1)^T, \alpha_3 = (0, 0, 1)^T; \beta_1 = (1, 1)^T, \beta_2 = (1, 0)^T;$
- (c) $\alpha_1 = (1, 2, 3)^T, \alpha_2 = (0, 1, -1)^T, \alpha_3 = (-1, -2, 3)^T; \beta_1 = (1, 2)^T, \beta_2 = (2, 1)^T.$

Proof. A routine computation.

□

Exercise. Let θ be a real number. Consider the complex matrices

$$A = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}, \quad B = \begin{pmatrix} e^{\sqrt{-1}\theta} & 0 \\ 0 & e^{-\sqrt{-1}\theta} \end{pmatrix}.$$

Find a complex matrix P such that $P^{-1}AP = B$.



Proof. Note that

$$e^{\sqrt{-1}\theta} = \cos \theta + \sqrt{-1} \sin \theta.$$

Then

$$\begin{pmatrix} \sqrt{-1} & 1 \\ 1 & \sqrt{-1} \end{pmatrix}^{-1} \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} \sqrt{-1} & 1 \\ 1 & \sqrt{-1} \end{pmatrix} = \begin{pmatrix} e^{\sqrt{-1}\theta} & 0 \\ 0 & e^{-\sqrt{-1}\theta} \end{pmatrix}.$$

□

Exercise. Let W be a subspace of V , let $\pi: V \rightarrow V/W$ be the projection map. Let $g: V/W \rightarrow V/W$ be a linear transformation. Is there always a linear transformation $f: V \rightarrow V$ such that $g \circ \pi = \pi \circ f$?

Proof. Suppose $\dim W = k$ with basis $\{w_1, \dots, w_k\}$. Firstly we extend the basis of W to a basis of V as $\{e_1, \dots, e_{n-k}, w_1, \dots, w_k\}$, and then $\{e_1 + W, \dots, e_{n-k} + W\}$ is a basis of V/W .

Given a linear transformation $g: V/W \rightarrow V/W$, one has

$$\begin{aligned} g \circ \pi(e_i) &= g(e_i + W) = g(e_i) + W \\ g \circ \pi(w_j) &= g(W) = W. \end{aligned}$$

Then we define a linear transformation $f: V \rightarrow V$ by evaluating on basis $\{e_1, \dots, e_{n-k}, w_1, \dots, w_k\}$ as

$$\begin{aligned} f(e_i) &= g(e_i) \\ f(w_i) &= w_i. \end{aligned}$$

Then it's a linear transformation which extends g .

□

Exercise. Let $f(x) \neq 0 \in K[x]$, where K is a field. Let $f(x) \cdot K[x]$ be the subspace of $K[x]$ consisting of polynomials divisible by $f(x)$.

- Find a basis of $V = K[x]/(f(x) \cdot K[x])$ and compute its dimension.
- Consider the linear transformation $T: V \rightarrow V$ such that $\bar{g}(x) \mapsto \bar{x} \cdot \bar{g}(x)$. Find the matrix representing T with respect to your basis.

Proof. For (a). Suppose the degree of $f(x)$ is n . Then

$$\{1, x, x^2, \dots, x^{n-1}\}$$

is a basis of $K[x]/(f(x)K[x])$. Indeed, it's clear above elements are linearly independent over K , and for any element $g(x)$ in $K[x]$ with degree higher than n , we can use division with remainders to write

$$g(x) = q(x)f(x) + r(x)$$

where $\deg r(x) < n$. This shows in $K[x]/(f(x)K[x])$ one has $g(x)$ is the same as $r(x)$, which implies $g(x)$ is a linear combination of $\{1, x, x^2, \dots, x^{n-1}\}$.

For (b). Suppose $f(x) = a_n x^n + \dots + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$. Then for x^i with $i < n-1$, one has $T(x^i) = x^{i+1}$, and

$$T(x^{n-1}) = x^n = -\frac{1}{a_n}(a_{n-1}x^{n-1} + \dots + a_1 x + a_0).$$

This shows T has the matrix representation as

$$\begin{pmatrix} 0 & 0 & 0 & \dots & 0 & -\frac{a_0}{a_n} \\ 1 & 0 & 0 & \dots & 0 & -\frac{a_1}{a_n} \\ 0 & 1 & 0 & \dots & 0 & -\frac{a_2}{a_n} \\ \vdots & \vdots & \vdots & & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 1 & -\frac{a_{n-1}}{a_n} \end{pmatrix}.$$

□



Exercise. Let V be a vector space.

- (a) Let V_1, V_2 and V'_1, V'_2 be subspaces of V such that $\dim V_i \cap V_j = \dim V'_i \cap V'_j$ for every possible i, j (in particular, $\dim V_i = \dim V'_i$). Show that there is an isomorphism $T: V \rightarrow V$ such that $T(V_i) = V'_i$.
- (b) **(open question, no need to submit)* Let V_1, V_2, V_3 and V'_1, V'_2, V'_3 be subspaces of V such that $\dim V_i \cap V_j \cap V_k = \dim V'_i \cap V'_j \cap V'_k$ and $\dim V_i \cap (V_j + V_k) = \dim V'_i \cap (V'_j + V'_k)$ for every possible i, j, k . Is there always an isomorphism $T: V \rightarrow V$ such that $T(V_i) = V'_i$?
- (c) **(open question, no need to submit)* What about subspaces V_1, V_2, V_3, V_4 , and more?

Proof. For (a). Let $\{e_1, \dots, e_k\}$ be a basis of $V_1 \cap V_2$ and $\{e'_1, \dots, e'_k\}$ be a basis of $V'_1 \cap V'_2$. Then we extend $\{e_1, \dots, e_k\}$ to a basis of V_1 by adding vectors u_1, \dots, u_m and extend $\{e_1, \dots, e_k\}$ to a basis of V_2 by adding vectors v_1, \dots, v_n . Finally we extend

$$\{e_1, \dots, e_k, u_1, \dots, u_m, v_1, \dots, v_n\}$$

to a basis of V by adding vectors $\varphi_1, \dots, \varphi_l$. Similarly, we can do the same thing to the basis $\{e'_1, \dots, e'_k\}$ and obtain a basis

$$\{e'_1, \dots, e'_k, u'_1, \dots, u'_m, v'_1, \dots, v'_n, \varphi'_1, \dots, \varphi'_l\}$$

of V . Then we define T as follows

$$\begin{aligned} T(e_\alpha) &= e'_\alpha \\ T(u_\beta) &= u'_\beta \\ T(v_\gamma) &= v'_\gamma \\ T(\varphi_\delta) &= \varphi'_\delta. \end{aligned}$$

□



Chapter 3

Solutions to Homework-6

Exercise. Suppose $A \in M_{m \times n}(K)$ and $B \in M_{n \times m}(K)$. Show that the nonzero eigenvalues of AB are the same as the nonzero eigenvalues of BA . If $m = n$, show that the eigenvalues of AB are the same as the eigenvalues of BA .

Proof. Note that

$$\begin{pmatrix} \lambda I_m & A \\ B & I_n \end{pmatrix} \rightarrow \begin{pmatrix} \lambda I_m - AB & A \\ 0 & I_n \end{pmatrix} \rightarrow \begin{pmatrix} \lambda I_m - AB & 0 \\ 0 & I_n \end{pmatrix}$$

and

$$\begin{pmatrix} \lambda I_m & A \\ B & I_n \end{pmatrix} \rightarrow \begin{pmatrix} \lambda I_m & A \\ 0 & I_n - \frac{1}{\lambda} BA \end{pmatrix} \rightarrow \begin{pmatrix} \lambda I_m & 0 \\ 0 & I_n - \frac{1}{\lambda} BA \end{pmatrix}$$

This shows that

$$\det |\lambda I_m - AB| = \det |\lambda I_m| \det \left| I_n - \frac{1}{\lambda} BA \right| = \lambda^{m-n} \det |\lambda I_n - BA|$$

As a consequence, they have the same nonzero eigenvalues. In particular, if $m = n$, then their characteristic polynomials are the same, and thus have the same eigenvalues. \square

Exercise. Find $\lim_{n \rightarrow \infty} A^n$, where $A = \begin{pmatrix} \frac{1}{7} & \frac{3}{7} & \frac{3}{7} \\ \frac{3}{7} & \frac{1}{7} & \frac{3}{7} \\ \frac{3}{7} & \frac{3}{7} & \frac{1}{7} \end{pmatrix}$.

Proof. Note that

$$A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & -2 & 1 \\ -2 & 1 & 1 \end{pmatrix} \begin{pmatrix} -\frac{2}{7} & & \\ & -\frac{2}{7} & \\ & & 1 \end{pmatrix} \frac{1}{3} \begin{pmatrix} 1 & 0 & -1 \\ 1 & -1 & 0 \\ 1 & 1 & 1 \end{pmatrix}$$

Then

$$\begin{aligned} \lim_{n \rightarrow \infty} A^n &= \begin{pmatrix} 1 & 1 & 1 \\ 1 & -2 & 1 \\ -2 & 1 & 1 \end{pmatrix} \begin{pmatrix} 0 & & \\ & 0 & \\ & & 1 \end{pmatrix} \frac{1}{3} \begin{pmatrix} 1 & 0 & -1 \\ 1 & -1 & 0 \\ 1 & 1 & 1 \end{pmatrix} \\ &= \frac{1}{3} \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} \end{aligned}$$

\square

Exercise. Find the inverse matrix of the matrix $A = \begin{pmatrix} 1 & 1 & 2 \\ 9 & 2 & 0 \\ 5 & 0 & 3 \end{pmatrix}$ using the Cayley-Hamilton theorem.



Proof. Note that the characteristic polynomials of A is $\lambda^3 - 6\lambda^2 - 8\lambda + 41$, and thus

$$A(A^2 - 6A - 8I_3) = -41I_3$$

As a consequence

$$A^{-1} = -\frac{1}{41}(A^2 - 6A - 8I_3) = \frac{1}{41} \begin{pmatrix} -6 & 3 & 4 \\ 27 & 7 & -18 \\ 10 & -5 & 7 \end{pmatrix}$$

□

Exercise. Let $A \in M_3(\mathbb{R})$ such that $\det A = 1$ and $(-1 + \sqrt{-3})/2$ is an eigenvalue of A .

(1) Find all eigenvalues of A .

(2) Suppose $A^{100} = aA^2 + bA + cI$, determine a, b, c .

Proof. For (1). The characteristic polynomial of A is of real coefficient, and $(-1 + \sqrt{-3})/2$ is a root. Then $(-1 - \sqrt{-3})/2$ is also a root which is also an eigenvalue. Since the product of all eigenvalues are $\det A = 1$, one has all eigenvalues of A are

$$\frac{-1 + \sqrt{-3}}{2}, \frac{-1 - \sqrt{-3}}{2}, 1.$$

For (2). Note that the characteristic polynomial of A is $\lambda^3 - 1$. By Cayley-Hamilton theorem one has $A^3 = I_3$. Therefore

$$A^{100} = A = aA^2 + bA + cI_3 \Rightarrow aA^2 + (b-1)A + cI_3 = 0$$

However, the eigenvalues of A are distinct, and thus the characteristic polynomial of A is the minimal polynomial of A as well. Then

$$a = 0, \quad b = 1, \quad c = 0.$$

□

Exercise. Let V be a vector space over K and $f, g \in V^*$ such that $f(v) = 0$ if and only if $g(v) = 0$. Show that $f = cg$ for some $0 \neq c \in K$.

Proof. If $f = cg$ for some $0 \neq c \in K$, it's clear $f(v) = 0$ if and only if $g(v) = 0$. On the other hand, assume that $f(v) = 0$ if and only if $g(v) = 0$.

(1) If $f = 0$, then for any $v \in V$ one has $g(v) = 0$ and thus $g = f = 0$.

(2) If $f \neq 0$, then $f(v) = 0$ if and only if $g(v) = 0$ is equivalent to say $\ker f = \ker g = W$, where W is a linear subspace with codimension one. Suppose $V = W \oplus S$ with $0 \neq v_0 \in S$, and take $c = f(v_0)/g(v_0)$. For any $x \in V$, it can be written uniquely as

$$x = tv_0 + w, \quad w \in W.$$

Thus

$$f(x) = f(tv_0 + w) = tf(v_0) = \frac{f(v_0)}{g(v_0)}tg(v_0) = cg(tv_0 + w) = cg(x).$$

Therefore $f = cg$.

□



Exercise. Let $\alpha_1 = (1, 0, -1), \alpha_2 = (1, 1, 1), \alpha_3 = (2, 2, 0)$ be a basis of \mathbb{C}^3 . Find the coordinates of the dual basis of α_i with respect to the dual basis of the standard basis of \mathbb{C}^3 .

Proof. Suppose $\{f^i\}$ is a dual basis of $\{\alpha_i\}$. For convenience we write $f^i = \sum_{j=1}^3 a_{ij}\epsilon^j$, where $\epsilon^1, \epsilon^2, \epsilon^3$ are dual basis for standard basis of \mathbb{C}^3 . Note that $f^i(\alpha_j) = \delta_{ij}$. Then it suffices to solve several systems of linear equations to find out

$$f^1 = \epsilon^1 - \epsilon^2, \quad f^2 = \epsilon^1 - \epsilon^2 + \epsilon^3, \quad f^3 = -\frac{1}{2}\epsilon^1 + \epsilon^2 - \frac{1}{2}\epsilon^3.$$

□

Exercise. Let V be the vector space of all polynomial functions p from \mathbb{R} to \mathbb{R} that have degree 2 or less:

$$p(x) = c_0 + c_1x + c_2x^2.$$

Define three linear functionals on V by

$$f_1(p) = \int_0^1 p(x)dx, \quad f_2(p) = \int_0^2 p(x)dx, \quad f_3(p) = \int_0^{-1} p(x)dx$$

Show that $\{f_1, f_2, f_3\}$ is a basis for V^* by exhibiting the basis for V of which it is dual.

Proof. For $p(x) = c_0 + c_1x + c_2x^2$, a direct computation shows

$$f_1(p) = c_0 + \frac{c_1}{2} + \frac{c_2}{3}, \quad f_2(p) = 2c_0 + 2c_1 + \frac{8}{3}c_2, \quad f_3(p) = -c_0 + \frac{c_1}{2} - \frac{c_2}{3}.$$

Then a dual basis can be taken as

$$p_1(x) = 1 + x - \frac{3}{2}x^2, \quad p_2(x) = -\frac{1}{6} + \frac{1}{2}x^2, \quad p_3(x) = -\frac{1}{3} + 1x - \frac{1}{2}x^2.$$

This is a basis since

$$\det \begin{pmatrix} 1 & 1 & -\frac{3}{2} \\ -\frac{1}{6} & 0 & \frac{1}{2} \\ -\frac{1}{3} & 1 & -\frac{1}{2} \end{pmatrix} = -\frac{1}{2} \neq 0$$

□

Exercise. Let W be the supspace of \mathbb{R}^5 spanned by the vectors $\alpha_1 = e_1 + 2e_2 + e_3, \alpha_2 = e_2 + 3e_3 + 3e_4 + e_5, \alpha_3 = e_1 + 4e_2 + 6e_3 + 4e_4 + e_5$, where e_i are the standard basis of \mathbb{R}^5 . Find a basis for W^\perp in terms of the dual basis of e_i .

Proof. Suppose $f \in W^\perp$ is given by $\sum_{i=1}^5 \alpha_i \epsilon^i$. Then one has the following system of linear equations

$$\begin{pmatrix} 1 & 0 & 0 & 4 & 3 \\ 0 & 1 & 0 & -3 & -2 \\ 0 & 0 & 1 & 2 & 1 \end{pmatrix} \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \\ \alpha_4 \\ \alpha_5 \end{pmatrix} = 0$$

Thus W^\perp can be viewed as a solution space, which is generated by

$$\{-4\epsilon^1 + 3\epsilon^2 - 2\epsilon^3 + \epsilon^4, -3\epsilon^1 + 2\epsilon^2 - \epsilon^3 + \epsilon^5\}.$$

□

Exercise. Let n be a positive integer and K a field. Let W be the set of vectors (x_1, \dots, x_n) in K^n such that $\sum x_i = 0$.



- (1) Prove that W^\perp consists of all linear functionals f of the form $f(x_1, \dots, x_n) = c \sum x_i$.
- (2) Suppose $K = \mathbb{R}$. Show that W^* can be naturally identified with the set of linear functionals $f = \sum c_i x_i$ on K^n such that $\sum c_i = 0$.

Proof. For (1). Suppose $f \in W^\perp$ and we denote $f(e_i) = c_i$ for convenience. Since $e_i - e_j \in W$, one has

$$c_i - c_j = f(e_i) - f(e_j) = f(e_i - e_j) = 0.$$

This shows $c_i = c$ for all i . Therefore

$$f(x_1, \dots, x_n) = f(x_1 e_1 + \dots + x_n e_n) = x_1 f(e_1) + \dots + x_n f(e_n) = c \sum x_i.$$

On the other hand, it's obvious that $f(x_1, \dots, x_n) = 0$ if f is of this form.

For (2). Note that there is a natural identification between W and W^* by

$$v^* \longleftrightarrow \langle v, - \rangle.$$

Under this identification, the linear functional $f = \sum_i c_i x_i$ on K^n corresponds to the vector (c_1, \dots, c_n) . Thus it gives a linear functional on W if and only if $(c_1, \dots, c_n) \in W$, that is, $\sum_i c_i = 0$. \square

Exercise. Suppose $f \in M_n(\mathbb{R})^*$ such that $f(AB) = f(BA)$ for all $A, B \in M_n(K)$ and $f(I) = n$. Show that f is the trace function.

Proof. Let $E_{ij} \in M_n(\mathbb{R})$ denote the matrix with (i, j) entry 1 and the others. Then

$$f(E_{ij}) = f(E_{ik}E_{kj}) = f(E_{kj}E_{ik}) = \delta_{ij}f(E_{kk}).$$

On the other hand, note that

$$f(E_{ii}) = f(E_{ij}E_{ji}) = f(E_{ji}E_{ij}) = f(E_{jj}).$$

for any i, j . Thus

$$f(I) = n f(E_{ii}) = n \Rightarrow f(E_{ii}) = 1.$$

Therefore

$$f(E_{ij}) = \delta_{ij}.$$

Then for any $A \in M_n(\mathbb{R})$, one has

$$f(A) = f\left(\sum_{i,j} a_{ij} E_{ij}\right) = \sum_{i,j} a_{ij} f(E_{ij}) = \sum_{i,j} a_{ij} \delta_{ij} = \sum_{i=1}^n a_{ii} = \text{tr}(A).$$

Therefore f is the trace function. \square



Chapter 4

Solutions to Homework-8

Exercise. Let $A = (a_{ij})$ be a symmetric real matrix. Suppose $a_{ii} > \sum_{j \neq i} |a_{ij}|$. Show that A is positive definite.

Proof. As shown in Exercise 12 of Homework-2, one has all principle minors of A is positive, and thus A is positive definite. \square

Exercise. A real symmetric matrix $A = (a_{ij})$ of order n is semi-positive definite if $\sum_{i,j} a_{ij} x_i x_j \geq 0$ for all (x_1, \dots, x_n) in \mathbb{R}^n . Prove that the following are equivalent:

- (a) A is semi-positive definite
- (b) $A = P^t \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} P$ for some real invertible matrix P
- (c) $A = Q^t Q$ for some real matrix Q
- (d) all principal minors of A are non-negative.

State the corresponding conclusion for semi-positive definite Hermitian matrices. Is it equivalent to that all leading principal minors of A are non-negative?

Proof. From (a) to (b). Since A is a real symmetric matrix, there exists some invertible matrix P such that

$$A = P^t \begin{pmatrix} I_r & & \\ & -I_s & \\ & & O \end{pmatrix} P$$

for some $r, s \in \mathbb{Z}_{\geq 0}$. On the other hand, since A is semi-positive definite, s must be zero as desired.

From (b) to (c). It suffices to set $Q = \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} P$.

From (c) to (a). Note that

$$x^t A x = x^t Q^t Q x = (Qx)^t Qx \geq 0.$$

From (a) to (d). For the principal minors $A \begin{pmatrix} k_1 & k_2 & \dots & k_s \\ k_1 & k_2 & \dots & k_s \end{pmatrix}$, now we're going to show that it's semi-definite, and thus $\det A \begin{pmatrix} k_1 & k_2 & \dots & k_s \\ k_1 & k_2 & \dots & k_s \end{pmatrix} \geq 0$. Suppose $x = (x_{k_1}, \dots, x_{k_s})^t$ such that

$$x^t A \begin{pmatrix} k_1 & k_2 & \dots & k_s \\ k_1 & k_2 & \dots & k_s \end{pmatrix} x < 0.$$

Then consider $\tilde{x} = (0, \dots, x_{k_1}, \dots, x_{k_s}, \dots, 0)$, one has $\tilde{x} A \tilde{x} < 0$, a contradiction.



From (d) to (a). If all principal minors of A are non-negative, then the characteristic polynomials $f(\lambda) \geq \lambda^n$ for all $\lambda > 0$ since the coefficients of $f(\lambda)$ are positive combinations of principal minors. This shows all eigenvalues of A are non-negative, and thus A is semi-positive definite.

However, the corresponding conclusion for semi-positive definite Hermitian matrices fails. □

Exercise. Use Gram-Schmidt procedure to construct an orthonormal basis of \mathbb{R}^4 from the following:

(a) $(0, 0, 2, 1)^t, (0, 3, 7, 2)^t, (1, 1, 6, 2)^t, (-1, 4, -1, -1)^t$;

(b) $(1, 1, 1, 1)^t, (1, 0, 1, 1)^t, (1, 1, 0, 1)^t, (1, 1, 1, 0)^t$.

Proof. It's a routine computation, and here we only show the results.

For (a).

$$\frac{1}{\sqrt{5}}(0, 0, 2, 1)^t, \frac{1}{\sqrt{30}}(0, 5, 1, -2)^t, \frac{1}{\sqrt{42}}(6, -1, 1, -2)^t, \frac{1}{\sqrt{7}}(1, 1, -1, 2)^t.$$

For (b)

$$\frac{1}{2}(1, 1, 1, 1)^t, \frac{1}{\sqrt{12}}(1, -3, 1, 1)^t, \frac{1}{\sqrt{6}}(1, 0, -2, 1)^t, \frac{1}{\sqrt{2}}(1, 0, 0, -1)^t.$$

□

Exercise. Prove that the maximal entries of a positive definite, symmetric, real matrix A are on the diagonal.

Proof. Suppose $a_{i_0 j_0} = \max_{i,j} a_{ij}$. If $i_0 \neq j_0$, then $a_{i_0 i_0} a_{j_0 j_0} - a_{i_0 j_0}^2 > 0$, since the determinant of principal minors $A(i_0, j_0)$ is > 0 . Thus $a_{i_0 j_0} < \max\{a_{i_0 i_0}, a_{j_0 j_0}\}$ since both $a_{i_0 i_0}$ and $a_{j_0 j_0}$ are positive, a contradiction. □

Exercise. Let $\langle -, - \rangle$ be a positive definite Hermitian form on a complex vector space V , and let $\{ -, - \}$, and $[-, -]$ be its real and imaginary parts, the real-valued forms defined by

$$\langle v, w \rangle = \{v, w\} + [v, w]i.$$

Prove that when V is made into a real vector space by restricting scalars to \mathbb{R} , $\{ -, - \}$ is a positive definite symmetric form, and $[-, -]$ is a skew-symmetric form.

Proof. For $v, w \in V$, one has

$$\langle v, w \rangle + \langle w, v \rangle = \langle v, w \rangle + \overline{\langle v, w \rangle} = 2\{v, w\}.$$

This shows $\{v, w\} + \{w, v\} = 0$, and thus $[-, -]$ is a skew-symmetric form. On the other hand, one has

$$\{v, w\} = \frac{1}{2}(\langle x+y, x+y \rangle - \langle x, x \rangle - \langle y, y \rangle) = \{y, x\}.$$

This shows $\{ -, - \}$ is a symmetric form, and it's positive definite since $\{x, x\} = \langle x, x \rangle$. □

Exercise. Let $V = \mathbb{R}^{2 \times 2}$ be the vector space of real 2×2 matrices.

(a) Determine the matrix of the bilinear form $\langle A, B \rangle = \text{tr}(AB)$ on V with respect to the standard basis $\{e_{ij}\}$.

(b) Determine the signature of this form.

(c) Find an orthogonal basis for this form.



(d) Determine the signature of the form $\text{trace } AB$ on the space $\mathbb{R}^{n \times n}$ of real $n \times n$ matrices.

Proof. For (a). Note that

$$\text{tr}(AB) = a_{11}b_{11} + a_{12}b_{21} + a_{21}b_{12} + a_{22}b_{22}.$$

A direct computation shows that the matrix is

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}.$$

For (b) and (c). The orthogonal basis is given by

$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix},$$

and thus the signature is $3 - 1 = 2$.

For (d). One can construct an orthogonal basis of $\mathbb{R}^{n \times n}$ by define

$$\alpha_{ij} = \begin{cases} E_{ij} & i = j \\ E_{ij} + E_{ji} & i < j \\ E_{ij} - E_{ji} & i > j. \end{cases}$$

A direct computation shows

$$\langle \alpha_{ij}, \alpha_{kl} \rangle = \begin{cases} 0, & (i, j) \neq (k, l) \\ 1, & i = j = k = l \\ 2, & i = k, j = l, i < j \\ -2, & i = k, j = l, i > j. \end{cases}$$

This shows the signature is

$$\frac{n^2 + n}{2} - \frac{n^2 - n}{2} = n.$$

□

Exercise. Let W be a subspace of a Euclidean space/Hermitian space V . Show that $W = W^{\perp\perp}$.

Proof. On one hand, it's clear $W \subseteq W^{\perp\perp}$. On the other hand, one has

$$\dim W^{\perp\perp} + \dim W^\perp = \dim V = \dim W + \dim W^\perp.$$

This shows $W = W^{\perp\perp}$.

□

Exercise. Show that the Gram determinant $\det((\alpha_i, \alpha_j))$ of n real vectors $\alpha_1, \dots, \alpha_n$ in \mathbb{R}^n is non-zero if and only if the vectors are linearly independent.

Proof. Note that $\det((\alpha_i, \alpha_j)) = \det(A^t A) = \det^2 A \neq 0$ if and only if $\det A \neq 0$, which is equivalent to say the α_i 's are linearly independent.

□

Exercise. Let V be a Euclidean space.

(a) Prove the parallelogram law $|v + w|^2 + |v - w|^2 = 2|v|^2 + 2|w|^2$.

(b) Prove that if $|v| = |w|$, then $(v + w) \perp (v - w)$.



Proof. For (a).

$$|u + v|^2 + |u - v|^2 = |u|^2 + 2(u, v) + |v|^2 + |u|^2 - 2|u||v| + |v|^2 = 2|u|^2 + 2|v|^2.$$

For (b).

$$(v + w, v - w) = |v|^2 + (w, v) - (v, w) - |w|^2 = 0.$$

□

Exercise. Let T be a linear operator on $V = \mathbb{R}^n$ whose matrix A is a real symmetric matrix.

(a) Prove that V is the orthogonal sum $V = (\ker T) \oplus (\operatorname{im} T)$.

(b) Prove that T is an orthogonal projection onto $\operatorname{im} T$ if and only if, in addition to being symmetric, $A^2 = A$.

Proof. For (a), For $v \in \ker T$ and $u = T(w) \in \operatorname{im} T$, one has

$$(v, u) = (v, T(w)) = (T(v), w) = (0, w) = 0$$

Therefore $\ker T \perp \operatorname{im} T$. On the other hand, one has

$$\dim \ker T + \dim \operatorname{im} T = \dim V.$$

Thus V is their orthogonal sum.

For (b). Suppose T is an orthogonal projection onto $\operatorname{im} T$. Then for every $v \in V$ with $v = v_1 + v_2$, where $v_1 \in \ker T$, $v_2 \in \operatorname{im} T$, one has $Tv = v_2$ and $Tv_2 = v_2$. This shows $T^2v = Tv$ for every $v \in V$, and thus $A^2 = A$. Conversely, if $A^2 = A$, then for every $v \in V$, one has $A^2v = Av$, and thus $Av - v \in \ker T$. This shows T is an orthogonal projection onto $\operatorname{im} T$ since $v = v - Av + Av$. □

Exercise. Let W be the subspace of \mathbb{R}^3 spanned by the vectors $(1, 1, 0)^t$ and $(0, 1, 1)^t$. Determine the orthogonal projection of the vector $(1, 0, 0)^t$ to W .

Proof. Note that W^\perp can be spanned by $(1, -1, 1)^t$. Suppose

$$(1, 0, 0)^t = a(1, 1, 0)^t + b(0, 1, 1)^t + c(1, -1, 1)^t$$

Then $c = 1/3$, and thus the orthogonal projection is

$$(1, 0, 0)^t - \frac{1}{3}(1, -1, 1)^t = \frac{1}{3}(2, 1, -1)^t.$$

□

Exercise. Let V be the real vector space of 3×3 matrices with the bilinear form $\langle A, B \rangle = \operatorname{tr}(A^t B)$, and let W be the subspace of skew-symmetric matrices. Compute the orthogonal projection to W with respect to this form, of the matrix

$$\begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 1 & 3 & 0 \end{pmatrix}.$$

Proof. If $A \in W^\perp$, then $\operatorname{tr}(A^t B) = 0$ for all $B \in W$. Since W is spanned by an orthonormal basis can be taken as

$$E_{12} - E_{21}, E_{13} - E_{31}, E_{23} - E_{32},$$



it reduces to

$$\begin{aligned}\operatorname{tr}(A^t(E_{12} - E_{21})) &= 0 \\ \operatorname{tr}(A^t(E_{13} - E_{31})) &= 0 \\ \operatorname{tr}(A^t(E_{23} - E_{32})) &= 0.\end{aligned}$$

This is equivalent to A is a symmetric matrix. Thus W^\perp consists of the symmetric matrices. Note that

$$\begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \\ 1 & 3 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 1 & \frac{1}{2} \\ 1 & 0 & 2 \\ \frac{1}{2} & 2 & 0 \end{pmatrix} + \begin{pmatrix} 0 & 1 & -\frac{1}{2} \\ -1 & 0 & -1 \\ \frac{1}{2} & 1 & 0 \end{pmatrix}.$$

Thus the orthogonal projection is exactly

$$\begin{pmatrix} 0 & 1 & -\frac{1}{2} \\ -1 & 0 & -1 \\ \frac{1}{2} & 1 & 0 \end{pmatrix}.$$

□

