MA2007B: LINEAR ALGEBRA I Midterm2/December 05, 2019

Please show all your work clearly for full credit! each problem 10 points, total 100 points)

(1) Find the symmetric factorization $A = LDL^{\top}$ of the 3 × 3 real symmetric matrix

$$A = \left[\begin{array}{rrrr} 1 & 4 & 5 \\ 4 & 2 & 6 \\ 5 & 6 & 3 \end{array} \right]$$

Solution: By the elimination, we have

$$A = \begin{bmatrix} 1 & 4 & 5 \\ 4 & 2 & 6 \\ 5 & 6 & 3 \end{bmatrix} \xrightarrow{\ell_{21}=4, \ \ell_{31}=5} \begin{bmatrix} 1 & 4 & 5 \\ 0 & -14 & -14 \\ 0 & -14 & -22 \end{bmatrix} \xrightarrow{\ell_{32}=1} \begin{bmatrix} 1 & 4 & 5 \\ 0 & -14 & -14 \\ 0 & 0 & -8 \end{bmatrix} = \boldsymbol{U}.$$

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Therefore, we obtain the symmetric factorization

$$A = LU = \begin{bmatrix} 1 & 0 & 0 \\ 4 & 1 & 0 \\ 5 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 4 & 5 \\ 0 & -14 & -14 \\ 0 & 0 & -8 \end{bmatrix}$$
$$= \begin{bmatrix} 1 & 0 & 0 \\ 4 & 1 & 0 \\ 5 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & -14 & 0 \\ 0 & 0 & -8 \end{bmatrix} \begin{bmatrix} 1 & 4 & 5 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} = LDL^{\top}.$$

(2) Show that if an $n \times n$ real matrix A is invertible, then A^{\top} is also invertible and $(A^{\top})^{-1} = (A^{-1})^{\top}$. **Proof:**

- \therefore *A* is invertible
- $\therefore \exists A^{-1}$ such that $A^{-1}A = I$ and $AA^{-1} = I$
- $\therefore (A^{-1}A)^{\top} = I^{\top} = I$ and $(AA^{-1})^{\top} = I^{\top} = I$
- $\therefore A^{\top}(A^{-1})^{\top} = I \text{ and } (A^{-1})^{\top}A^{\top} = I$
- $\therefore A^{\top}$ is invertible and $(A^{\top})^{-1} = (A^{-1})^{\top}$
- (3) Let $A \in \mathbb{R}^{m \times n}$ and $B \in \mathbb{R}^{n \times p}$. Show that C(A) is a subspace of \mathbb{R}^m and $C(AB) \subseteq C(A)$, where C(M) denotes the column space of matrix M.

Proof:

(i) Let $A = [A_1, A_2, \dots, A_n]$, where $A_i \in \mathbb{R}^m$, $1 \le i \le n$, are the column vectors of A. Then the column space of A is given by

$$C(A) = \{c_1A_1 + c_2A_2 + \cdots + c_nA_n | c_1, c_2, \cdots, c_n \in \mathbb{R}\} \subseteq \mathbb{R}^m.$$

Let $v := c_1 A_1 + c_2 A_2 + \dots + c_n A_n$, $w := d_1 A_1 + d_2 A_2 + \dots + d_n A_n \in C(A)$ and $c \in \mathbb{R}$. Then

$$v + w = (c_1A_1 + c_2A_2 + \dots + c_nA_n) + (d_1A_1 + d_2A_2 + \dots + d_nA_n)$$

= $(c_1 + d_1)A_1 + (c_2 + d_2)A_2 + \dots + (c_n + d_n)A_n \in C(A),$
 $cv = c(c_1A_1 + c_2A_2 + \dots + c_nA_n)$
= $cc_1A_1 + cc_2A_2 + \dots + cc_nA_n \in C(A).$

Therefore, C(A) is a subspace of \mathbb{R}^m .

(ii) Let $B = [B_1, B_2, \dots, B_p]$, where $B_i \in \mathbb{R}^n$, $i = 1, 2, \dots, p$, are the column vectors of B. Then $AB = A[B_1, B_2, \dots, B_p] = [AB_1, AB_2, \dots, AB_p]$, and

C(AB) = the set of all linear combinations of AB_1, AB_2, \cdots, AB_p .

Let $v \in C(AB)$. Then v is a linear combination of AB_1, AB_2, \cdots, AB_p .

 \therefore For each *i*, *AB*^{*i*} is a linear combination of columns of *A*

 \therefore *v* is a linear combination of columns of *A*

$$\therefore v \in C(A)$$

$$\therefore C(AB) \subseteq C(A)$$

(4) Let *V* be a vector space over the field \mathbb{F} and let *K* be a nonempty subset of *V*. Define

$$S := \{c_1v_1 + c_2v_2 + \cdots + c_Nv_N | \forall v_1, v_2, \cdots, v_N \in K, c_1, c_2, \cdots, c_N \in \mathbb{F}, \text{and } N \in \mathbb{N}\}.$$

Show that *S* is a subspace of *V* and *S* is the smallest subspace of *V* that contains *K*.

Proof:

(i) Obviously, $S \subseteq V$. Let $v, w \in S$ and $c \in \mathbb{F}$. Then

$$v = c_1v_1 + c_2v_2 + \cdots + c_Nv_N$$
 for some $v_1, v_2, \cdots, v_N \in K$ and $c_1, c_2, \cdots, c_N \in \mathbb{F}$,

$$w = d_1w_1 + d_2w_2 + \cdots + d_Mw_M$$
 for some $w_1, w_2, \cdots, w_M \in K$ and $d_1, d_2, \cdots, d_M \in \mathbb{F}$.

We obtain

- $v + w = c_1v_1 + c_2v_2 + \cdots + c_Nv_N + d_1w_1 + d_2w_2 + \cdots + d_Mw_M$ is a linear combination of vectors in $K \Longrightarrow v + w \in S$.
- $cv = c(c_1v_1 + c_2v_2 + \dots + c_Nv_N) = cc_1v_1 + cc_2v_2 + \dots + cc_Nv_N$ is a linear combination of vectors in $K \Longrightarrow cv \in S$.

Therefore, S is a subspace of V.

(ii) Assume that *T* is a subspace of *V* that contains *K*. We want to show that $S \subseteq T$. Let $v \in S$. Then $\exists v_1, v_2, \dots, v_N \in K$ and $c_1, c_2, \dots, c_N \in \mathbb{F}$ s.t. $v = c_1v_1 + c_2v_2 + \dots + c_Nv_N$. \therefore *T* is a subspace of *V* and $K \subseteq T$ and *v* is a linear combination of vectors in *K* \therefore $v = c_1v_1 + c_2v_2 + \dots + c_Nv_N \in T$ \therefore $S \subseteq T$

(5) Let $A \in \mathbb{R}^{m \times n}$ and $B \in \mathbb{R}^{n \times p}$. Show that $\operatorname{rank}(AB) \leq \operatorname{rank}(B)$.

Proof: Let $B = [B_1, B_2, \dots, B_p]$. Then $AB = A[B_1, B_2, \dots, B_p] = [AB_1, AB_2, \dots, AB_p]$. If the *j*-th column B_j of B is a linear combination of previous columns of B, i.e.,

$$B_j = c_1 B_1 + c_2 B_2 + \dots + c_{j-1} B_{j-1}$$
, for some $c_1, c_2, \dots, c_{j-1} \in \mathbb{R}$,

then we have

- B_i is a free column, i.e., B_i is not a pivot column.
- The *j*-th column of *AB* is

$$AB_{j} = A(c_{1}B_{1} + c_{2}B_{2} + \dots + c_{j-1}B_{j-1}) = c_{1}AB_{1} + c_{2}AB_{2} + \dots + c_{j-1}AB_{j-1}.$$

i.e., the *j*-th column AB_j of AB is a linear combination of previous columns of AB. Therefore, AB_j is not a pivot column of AB.

We can conclude that

rank(AB) = number of pivot columns of $AB \leq$ number of pivot columns of B = rank(B).

(6) Consider the linear system Ax = b,

$$A = \begin{bmatrix} 1 & 2 & 3 & 5 \\ 2 & 4 & 8 & 12 \\ 3 & 6 & 7 & 13 \end{bmatrix}, \quad b = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

Reduce the augmented matrix $[A \ b]$ to $[U \ c]$ so that Ax = b becomes a triangular system Ux = c and find the condition on b_1, b_2, b_3 for Ax = b solvable.

Solution:

$$\begin{bmatrix} A & b \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 & 5 & b_1 \\ 2 & 4 & 8 & 12 & b_2 \\ 3 & 6 & 7 & 13 & b_3 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & 3 & 5 & b_1 \\ 0 & 0 & 2 & 2 & b_2 - 2b_1 \\ 0 & 0 & -2 & -2 & b_3 - 3b_1 \end{bmatrix}$$
$$\rightarrow \begin{bmatrix} 1 & 2 & 3 & 5 & b_1 \\ 0 & 0 & 2 & 2 & b_2 - 2b_1 \\ 0 & 0 & 0 & 0 & b_3 + b_2 - 5b_1 \end{bmatrix} = \begin{bmatrix} \mathbf{U} & \mathbf{c} \end{bmatrix}$$

Therefore, the solvability condition of Ax = b is $b_3 + b_2 - 5b_1 = 0$.

(7) Let $b := [0, 6, -6]^{\top}$ in Problem (6). Find the reduced row echelon form $[\mathbf{R} \ d]$ and find the complete solution to Ax = b.

Solution: First, note that $b_3 + b_2 - 5b_1 = -6 + 6 - 0 = 0$.

$$\begin{bmatrix} \mathbf{U} & \mathbf{c} \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 & 5 & 0 \\ 0 & 0 & 2 & 2 & 6 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & 3 & 5 & 0 \\ 0 & 0 & 1 & 1 & 3 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & 0 & 2 & -9 \\ 0 & 0 & 1 & 1 & 3 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} \mathbf{R} & \mathbf{d} \end{bmatrix}$$

Free columns: 2, 4 free variables: x_2 , x_4

Pivot columns: 1, 3 pivot variables: x_1 , x_3

Consider $Ax = 0 \Leftrightarrow Rx = 0$.

Set $x_2 = 1$, $x_4 = 0$: $x_1 = -2$ and $x_3 = 0$.

Set $x_2 = 0$, $x_4 = 1$: $x_1 = -2$ and $x_3 = -1$.

We have two special solutions to Ax = 0:

$$s_1 = \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \end{bmatrix}$$
 and $s_2 = \begin{bmatrix} -2 \\ 0 \\ -1 \\ 1 \end{bmatrix}$

Consider $Ax_p = b \Leftrightarrow Rx_p = d$.

Set $x_2 = 0$, $x_4 = 0$: $x_1 = -9$ and $x_3 = 3$.

We obtain a particular solution to $Ax_p = b$:

$$\boldsymbol{x}_p = \begin{bmatrix} -9\\0\\3\\0 \end{bmatrix}$$

Therefore, the complete solution to Ax = b is

$$x = x_p + x_n = \begin{bmatrix} -9 \\ 0 \\ 3 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -2 \\ 0 \\ -1 \\ 1 \end{bmatrix}.$$

(8) Suppose you know that the 3×4 real matrix A has the vector $s = [2, 3, 1, 0]^{\top}$ as the only special solution to Ax = 0. (a) What is the rank of A and the complete solution to Ax = 0? (b) What is the exact reduced row echelon form R of A? (c) How do you know that Ax = b can be solved for all $b \in \mathbb{R}^3$?

Solution:

A has only one free variable x_3 and one free column, the 3rd column.

(a) rank(A) = number of pivot columns = 4 - 1 = 3. The complete solution to Ax = 0 is

$$\boldsymbol{x} = c\boldsymbol{s} = c \begin{bmatrix} 2\\ 3\\ 1\\ 0 \end{bmatrix}.$$

(b) Since $s = [2, 3, 1, 0]^{\top}$, the exact reduced row echelon form *R* of *A* is given by

$$\boldsymbol{R} = \begin{bmatrix} 1 & 0 & -2 & 0 \\ 0 & 1 & -3 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}.$$

(c) Ax = b can be solved for all $b \in \mathbb{R}^3$ because A and R have full row rank r = 3.

(9) Find the largest possible number of independent vectors among

$$v_{1} = \begin{bmatrix} 1\\ -1\\ 0\\ 0 \end{bmatrix}, v_{2} = \begin{bmatrix} 1\\ 0\\ -1\\ 0 \end{bmatrix}, v_{3} = \begin{bmatrix} 1\\ 0\\ 0\\ -1 \end{bmatrix}, v_{4} = \begin{bmatrix} 0\\ 1\\ -1\\ 0 \end{bmatrix}, v_{5} = \begin{bmatrix} 0\\ 1\\ 0\\ -1 \end{bmatrix}, v_{6} = \begin{bmatrix} 0\\ 0\\ 1\\ -1 \end{bmatrix}.$$

Solution:

(i) Claim: v_1 , v_2 , v_3 are linearly independent.

Let $x_1v_1 + x_2v_2 + x_3v_3 = 0$. Then

$$x_{1}\begin{bmatrix}1\\-1\\0\\0\end{bmatrix}+x_{2}\begin{bmatrix}1\\0\\-1\\0\end{bmatrix}+x_{3}\begin{bmatrix}1\\0\\0\\-1\end{bmatrix}=\begin{bmatrix}0\\0\\0\\0\end{bmatrix}\Longrightarrow x_{1}=0, \quad x_{2}=0, \quad x_{3}=0.$$

Therefore, v_1 , v_2 , v_3 are linearly independent.

(ii) $v_4 = v_2 - v_1 \implies -v_1 + v_2 - v_4 = \mathbf{0}$ $v_5 = v_3 - v_1 \implies -v_1 + v_3 - v_5 = \mathbf{0}$ $v_6 = v_3 - v_2 \implies -v_2 + v_3 - v_6 = \mathbf{0}$

By (i) and (ii), 3 is the largest number of independent vectors.

(10) Show that if an $n \times n$ real matrix A is invertible, then the set of all column vectors of A is linearly independent.

Proof:

Let $A = [A_1, A_2, \dots, A_n]$, where $A_i, 1 \le i \le n$, are the column vectors of A.

- Assume that $x_1A_1 + x_2A_2 + \cdots + x_nA_n = 0$, i.e., Ax = 0. Then
- \therefore *A* is invertible, so A^{-1} exists

$$\therefore x = A^{-1}Ax = A^{-1}\mathbf{0} = \mathbf{0}$$

 \therefore A_1, A_2, \cdots, A_n are linearly independent