

Solution to Midterm Examination No. 2

1. Perform Gaussian elimination to obtain the reduced row echelon form \mathbf{R} of \mathbf{A} as follows:

$$\begin{aligned} \mathbf{A} = \begin{bmatrix} 0 & 1 & 2 & 3 & 4 \\ 0 & 1 & 2 & 4 & 6 \\ 0 & 0 & 0 & 1 & 2 \end{bmatrix} &\implies \begin{bmatrix} 0 & 1 & 2 & 3 & 4 \\ 0 & 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 1 & 2 \end{bmatrix} \\ &\implies \begin{bmatrix} 0 & 1 & 2 & 0 & -2 \\ 0 & 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} = \mathbf{R}. \end{aligned}$$

- Row space:

The row space of \mathbf{A} is the same as that of \mathbf{R} . We have a basis of the row space of \mathbf{A} as

$$(0 \ 1 \ 2 \ 0 \ -2), (0 \ 0 \ 0 \ 1 \ 2).$$

- Column space:

It is obvious that the second and fourth columns of \mathbf{R} form a basis for the column space of \mathbf{R} . Hence the second and fourth columns of \mathbf{A} form a basis for the column space of \mathbf{A} . Therefore, a basis for the column space of \mathbf{A} is

$$\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 4 \\ 1 \end{bmatrix}.$$

- Nullspace:

By \mathbf{R} , we have the special solutions

$$\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ -2 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 2 \\ 0 \\ -2 \\ 1 \end{bmatrix}$$

which form a basis for $\mathcal{N}(\mathbf{A})$.

- Left nullspace:

To find the left nullspace of \mathbf{A} , we perform Gaussian elimination to obtain the reduced row echelon form of \mathbf{A}^T :

$$\begin{aligned} \begin{bmatrix} 0 & 0 & 0 \\ 1 & 1 & 0 \\ 2 & 2 & 0 \\ 3 & 4 & 1 \\ 4 & 6 & 2 \end{bmatrix} &\implies \begin{bmatrix} 0 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 2 & 2 \end{bmatrix} \\ &\implies \begin{bmatrix} 0 & 0 & 0 \\ 1 & 0 & -1 \\ 0 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}. \end{aligned}$$

Then we have the special solution

$$\begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

which forms a basis for $\mathcal{N}(\mathbf{A}^T)$.

2. (a) Since the 5×5 matrix $[\mathbf{A} \ \mathbf{b}]$ is invertible, there are 5 nonzero pivots and all the columns are independent. Then \mathbf{b} cannot be a linear combination of the columns of \mathbf{A} . Therefore, $\mathbf{Ax} = \mathbf{b}$ has no solution.
- (b) We know that \mathbf{A} is a 5×4 matrix with rank 4. That is to say \mathbf{A} is with full column rank. Since $[\mathbf{A} \ \mathbf{b}]$ is singular and \mathbf{A} is with full column rank, there is only one free variable in the column of \mathbf{b} . Then \mathbf{b} can be a linear combination of the columns of \mathbf{A} . Therefore, $\mathbf{Ax} = \mathbf{b}$ is solvable.
3. Since \mathbf{x}_r is in the row space of \mathbf{A} , we have

$$\mathbf{x}_r = \mathbf{A}^T \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \text{ for some } x_1, x_2 \in \mathcal{R}.$$

Then

$$\mathbf{Ax}_r = \mathbf{AA}^T \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \mathbf{b}$$

where

$$\mathbf{AA}^T = \begin{bmatrix} 1 & 0 & 2 \\ 1 & 1 & 4 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 0 & 1 \\ 2 & 4 \end{bmatrix} = \begin{bmatrix} 5 & 9 \\ 9 & 18 \end{bmatrix} \text{ and } \mathbf{b} = \begin{bmatrix} 13 \\ 27 \end{bmatrix}.$$

We can solve x_1 and x_2 by

$$\begin{bmatrix} 5 & 9 \\ 9 & 18 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 13 \\ 27 \end{bmatrix} \implies \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}.$$

Hence,

$$\mathbf{x}_r = \mathbf{A}^T \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 6 \end{bmatrix}.$$

4. Let \mathbf{K}_1 be the subspace spanned by the first column of \mathbf{A} and \mathbf{K}_2 be the column space of \mathbf{A} . Obviously, \mathbf{K}_1 is contained in \mathbf{K}_2 . For any vector \mathbf{x} , the projection vector of \mathbf{x} onto \mathbf{K}_1 is

$$\mathbf{P}_1\mathbf{x} \in \mathbf{K}_1 \subset \mathbf{K}_2.$$

Since $\mathbf{P}_1\mathbf{x} \in \mathbf{K}_2$, the projection vector of $\mathbf{P}_1\mathbf{x}$ onto \mathbf{K}_2 is $\mathbf{P}_1\mathbf{x}$ itself. Thus,

$$\mathbf{P}_2\mathbf{P}_1\mathbf{x} = \mathbf{P}_2(\mathbf{P}_1\mathbf{x}) = \mathbf{P}_1\mathbf{x}, \text{ for any vector } \mathbf{x}.$$

Therefore,

$$\mathbf{P}_2 \mathbf{P}_1 = \mathbf{P}_1 = \frac{\begin{bmatrix} 1 \\ 2 \\ 0 \\ 1 \end{bmatrix} [1 \ 2 \ 0 \ 1]}{[1 \ 2 \ 0 \ 1] \begin{bmatrix} 1 \\ 2 \\ 0 \\ 1 \end{bmatrix}} = \frac{1}{6} \begin{bmatrix} 1 & 2 & 0 & 1 \\ 2 & 4 & 0 & 2 \\ 0 & 0 & 0 & 0 \\ 1 & 2 & 0 & 1 \end{bmatrix}.$$

5. (a) Consider the least-square linear fit to $(0, b_1)$, $(1, b_2)$, and $(2, b_3)$, and let

$$\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix}, \quad \hat{\mathbf{x}} = \begin{bmatrix} C \\ D \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}, \quad \text{and } \mathbf{b} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}.$$

Since

$$\mathbf{A}^T \mathbf{A} \hat{\mathbf{x}} = \mathbf{A}^T \mathbf{b}$$

we have

$$\begin{aligned} \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ -2 \end{bmatrix} &= \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \\ \implies \begin{bmatrix} -3 \\ -7 \end{bmatrix} &= \begin{bmatrix} b_1 + b_2 + b_3 \\ b_2 + 2b_3 \end{bmatrix}. \end{aligned}$$

The equations that b_1 , b_2 , and b_3 must satisfy are

$$\begin{cases} b_1 + b_2 + b_3 = -3 \\ b_2 + 2b_3 = -7. \end{cases}$$

- (b) Since all the three points $(0, b_1)$, $(1, b_2)$, and $(2, b_3)$ fall on the line $1 - 2t$, we can have

$$\begin{aligned} b_1 &= 1 - 2 \times 0 = 1 \\ b_2 &= 1 - 2 \times 1 = -1 \\ b_3 &= 1 - 2 \times 2 = -3. \end{aligned}$$

That is to say

$$\mathbf{b} = (b_1 \ b_2 \ b_3)^T = \begin{bmatrix} 1 \\ -1 \\ -3 \end{bmatrix}.$$

Then we can have

$$\begin{cases} b_1 + b_2 + b_3 = 1 + (-1) + (-3) = -3 \\ b_2 + 2b_3 = -1 + 2 \times (-3) = -7 \end{cases}$$

which satisfy the equations in (a).

6. (a) Since $\text{rank}(\mathbf{A}) = \text{rank}(\mathbf{R}) = 2$, the maximum number of columns of \mathbf{A} that form an independent set of vectors is equal to 2.
- (b) Since the row space of \mathbf{A} is equal to that of \mathbf{R} , take

$$\mathbf{a}_1 = \begin{bmatrix} 1 \\ 2 \\ 0 \\ 2 \end{bmatrix} \quad \text{and} \quad \mathbf{a}_2 = \begin{bmatrix} 0 \\ 0 \\ 1 \\ -9 \end{bmatrix}$$

as a basis for the row space of \mathbf{A} . Then perform the Gram-Schmidt process as follows.

1. Take $\mathbf{A}_1 = \mathbf{a}_1$. Then $\mathbf{q}_1 = \frac{\mathbf{A}_1}{\|\mathbf{A}_1\|} = \frac{1}{3} \begin{bmatrix} 1 \\ 2 \\ 0 \\ 2 \end{bmatrix}$.

2. Take $\mathbf{A}_2 = \mathbf{a}_2 - (\mathbf{q}_1^T \mathbf{a}_2) \mathbf{q}_1 = \begin{bmatrix} 2 \\ 4 \\ 1 \\ -5 \end{bmatrix}$. Then $\mathbf{q}_2 = \frac{\mathbf{A}_2}{\|\mathbf{A}_2\|} = \frac{1}{\sqrt{46}} \begin{bmatrix} 2 \\ 4 \\ 1 \\ -5 \end{bmatrix}$.

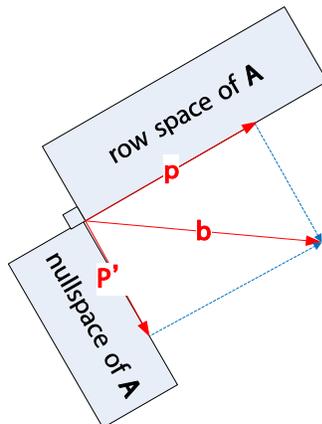
Therefore, we obtain an orthonormal basis for the row space of \mathbf{A} as

$$\frac{1}{3} \begin{bmatrix} 1 \\ 2 \\ 0 \\ 2 \end{bmatrix} \quad \text{and} \quad \frac{1}{\sqrt{46}} \begin{bmatrix} 2 \\ 4 \\ 1 \\ -5 \end{bmatrix}.$$

- (c) This is equivalent to finding the projection of \mathbf{b} onto the row space of \mathbf{A} . We can obtain

$$\mathbf{p} = (\mathbf{q}_1^T \mathbf{b}) \mathbf{q}_1 + (\mathbf{q}_2^T \mathbf{b}) \mathbf{q}_2 = \begin{bmatrix} 2 \\ 4 \\ 0 \\ 4 \end{bmatrix}.$$

- (d) By the following figure,



we can have

$$\mathbf{p}' = \mathbf{b} - \mathbf{p} = \begin{bmatrix} 0 \\ 1 \\ -9 \\ -1 \end{bmatrix}.$$

7. (a) Consider the first three columns:

$$\begin{bmatrix} x & x & x \\ x & x & x \\ 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$

The rank of this submatrix is at most 2. Therefore, the columns are dependent.

- (b) The big formula states that the determinant of \mathbf{A} is the sum of $5!$ simple determinants, times 1 or -1 , and every simple determinant chooses one entry from each row and column. If some simple determinant of \mathbf{A} avoids all the zero entries in \mathbf{A} , then it cannot choose one entry from each column. Thus every simple determinant of \mathbf{A} must choose at least one zero entry, and hence all the terms are zero in the big formula for $\det \mathbf{A}$.

8. (a) Let $|\mathbf{A}_n| = \det \mathbf{A}_n$. First observe that, for $n \geq 3$,

$$|\mathbf{A}_n| = \begin{vmatrix} & & 0 \\ & \mathbf{A}_{n-1} & \vdots \\ & & 0 \\ 0 & \cdots & 0 & 1 & a_n \end{vmatrix} = \begin{vmatrix} & & 0 & 0 \\ & \mathbf{A}_{n-2} & \vdots & \vdots \\ & & -1 & 0 \\ 0 & \cdots & 0 & 1 & a_{n-1} & -1 \\ 0 & \cdots & 0 & 0 & 1 & a_n \end{vmatrix}.$$

Applying the cofactor formula to the last row, we can have

$$\begin{aligned} |\mathbf{A}_n| &= a_n \cdot (-1)^{n+n} |\mathbf{A}_{n-1}| + 1 \cdot (-1)^{n+(n-1)} \begin{vmatrix} & & 0 \\ & \mathbf{A}_{n-2} & \vdots \\ & & 0 \\ 0 & \cdots & 0 & 1 & -1 \end{vmatrix} \\ &= a_n |\mathbf{A}_{n-1}| - (-1) \cdot (-1)^{(n-1)+(n-1)} |\mathbf{A}_{n-2}| \quad (\text{expansion along the last column}) \\ &= a_n |\mathbf{A}_{n-1}| + |\mathbf{A}_{n-2}|. \end{aligned}$$

- (b) (i) We can have

$$\begin{aligned} |\mathbf{A}_1| &= 1 \\ |\mathbf{A}_2| &= \begin{vmatrix} 1 & -1 \\ 1 & 2 \end{vmatrix} = 3 \\ |\mathbf{A}_3| &= 3 \cdot |\mathbf{A}_2| + |\mathbf{A}_1| = 3 \cdot 3 + 1 = 10 \\ |\mathbf{A}_4| &= 4 \cdot |\mathbf{A}_3| + |\mathbf{A}_2| = 4 \cdot 10 + 3 = 43 \\ |\mathbf{A}_5| &= 5 \cdot |\mathbf{A}_4| + |\mathbf{A}_3| = 5 \cdot 43 + 10 = 225 \\ |\mathbf{A}_6| &= 6 \cdot |\mathbf{A}_5| + |\mathbf{A}_4| = 6 \cdot 225 + 43 = 1393. \end{aligned}$$

(ii) We can have

$$|\mathbf{A}_1| = 5$$

$$|\mathbf{A}_2| = \begin{vmatrix} 5 & -1 \\ 1 & 4 \end{vmatrix} = 21$$

$$|\mathbf{A}_3| = 3 \cdot |\mathbf{A}_2| + |\mathbf{A}_1| = 3 \cdot 21 + 5 = 68$$

$$|\mathbf{A}_4| = 2 \cdot |\mathbf{A}_3| + |\mathbf{A}_2| = 2 \cdot 68 + 21 = 157$$

$$|\mathbf{A}_5| = 1 \cdot |\mathbf{A}_4| + |\mathbf{A}_3| = 1 \cdot 157 + 68 = 225$$

$$|\mathbf{A}_6| = 0 \cdot |\mathbf{A}_5| + |\mathbf{A}_4| = 0 \cdot 225 + 157 = 157.$$