

## Solution to Midterm Examination No. 2

1. (a) Consider the  $3 \times 2$  matrix

$$\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 0 & 1 \end{bmatrix}.$$

Then  $\mathcal{C}(\mathbf{A})$  contains  $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ ,  $\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$  and  $\mathcal{C}(\mathbf{A}^T)$  contains  $(1, 1)$ ,  $(1, 2)$ .

- (b) Since the column space and nullspace both have three components, the desired matrix is 3 by 3, say  $\mathbf{B}$ . We can find  $\dim(\mathcal{N}(\mathbf{B})) = 1 \neq 2 = 3 - 1 = 3 - \text{rank}(\mathbf{B})$ , which is not possible. Therefore, no such matrix exists.
- (c) Suppose the desired matrix exists. By the required property, column rank =  $\dim(\mathcal{R}^4) = 4 \neq 3 = \dim(\mathcal{R}^3) = \text{row rank}$ , which is not possible. Therefore, no such matrix exists.

2. (a) Transform  $\mathbf{A}$  into the RRE form:

$$\mathbf{A} = \begin{bmatrix} 1 & 3 & 1 & 2 \\ 2 & 6 & 3 & 5 \\ -1 & -3 & 1 & 0 \end{bmatrix} \implies \mathbf{R} = \begin{bmatrix} 1 & 3 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$$

Therefore, a basis for the row space of  $\mathbf{A}$  can be given by

$$(1, 3, 0, 1), (0, 0, 1, 1).$$

- (b) The orthogonal complement of the column space of  $\mathbf{A}$  is the left nullspace of  $\mathbf{A}$ . We can have  $\mathbf{R} = \mathbf{E}\mathbf{A}$  where

$$\mathbf{E} = \begin{bmatrix} 3 & -1 & 0 \\ -2 & 1 & 0 \\ 5 & -2 & 1 \end{bmatrix}.$$

Since the last row of  $\mathbf{R}$  is a zero row, a basis for the left nullspace can be given by the last row of  $\mathbf{E}$ :

$$(5, -2, 1).$$

- (c) From  $\mathbf{R}$ , we know that  $(1, 2, -1)^T$ ,  $(1, 3, 1)^T$  form a basis for the column space of  $\mathbf{A}$ . Therefore, we can obtain

$$\begin{aligned} \mathbf{P}_c &= \begin{bmatrix} 1 & 1 \\ 2 & 3 \\ -1 & 1 \end{bmatrix} \left( \begin{bmatrix} 1 & 1 \\ 2 & 3 \\ -1 & 1 \end{bmatrix}^T \begin{bmatrix} 1 & 1 \\ 2 & 3 \\ -1 & 1 \end{bmatrix} \right)^{-1} \begin{bmatrix} 1 & 1 \\ 2 & 3 \\ -1 & 1 \end{bmatrix}^T \\ &= \begin{bmatrix} 1/6 & 1/3 & -1/6 \\ 1/3 & 13/15 & 1/15 \\ -1/6 & 1/15 & 29/30 \end{bmatrix}. \end{aligned}$$

(d) We can project  $\mathbf{x}$  onto the column space of  $\mathbf{A}$  and obtain

$$\mathbf{x}_c = \mathbf{P}_c \mathbf{x} = \begin{bmatrix} 1/6 & 1/3 & -1/6 \\ 1/3 & 13/15 & 1/15 \\ -1/6 & 1/15 & 29/30 \end{bmatrix} \begin{bmatrix} 5 \\ -1 \\ 3 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}.$$

Then we can have

$$\mathbf{x}_{ln} = \mathbf{x} - \mathbf{x}_c = \begin{bmatrix} 5 \\ -1 \\ 3 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 5 \\ -2 \\ 1 \end{bmatrix}.$$

3. (a) Let

$$\mathbf{A}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \hat{\mathbf{x}}_1 = [C_1], \text{ and } \mathbf{b}_1 = \begin{bmatrix} 2 \\ 0 \\ -3 \\ -5 \end{bmatrix}.$$

We know the choice of  $\hat{\mathbf{x}}_1$  which minimizes the squared error can be obtained by solving

$$\mathbf{A}_1^T \mathbf{A}_1 \hat{\mathbf{x}}_1 = \mathbf{A}_1^T \mathbf{b}_1$$

which gives

$$4C_1 = -6.$$

Hence the best least squares horizontal line fit is given by  $b = C_1 = -3/2$ .

(b) Let

$$\mathbf{A}_2 = \begin{bmatrix} 1 & -1 \\ 1 & 0 \\ 1 & 1 \\ 1 & 2 \end{bmatrix}, \hat{\mathbf{x}}_2 = \begin{bmatrix} C_2 \\ D_2 \end{bmatrix}, \text{ and } \mathbf{b}_2 = \begin{bmatrix} 2 \\ 0 \\ -3 \\ -5 \end{bmatrix}.$$

The least squares solution  $\hat{\mathbf{x}}_2$  can be obtained by solving

$$\mathbf{A}_2^T \mathbf{A}_2 \hat{\mathbf{x}}_2 = \mathbf{A}_2^T \mathbf{b}_2$$

or equivalently,

$$\begin{bmatrix} 4 & 2 \\ 2 & 6 \end{bmatrix} \begin{bmatrix} C_2 \\ D_2 \end{bmatrix} = \begin{bmatrix} -6 \\ -15 \end{bmatrix}.$$

Therefore, we have

$$\begin{bmatrix} C_2 \\ D_2 \end{bmatrix} = \begin{bmatrix} -3/10 \\ -12/5 \end{bmatrix}$$

and the best least squares straight line fit is given by  $b = C_2 + D_2 t = -3/10 - (12/5)t$ .

(c) Let

$$\mathbf{A}_3 = \begin{bmatrix} 1 & -1 & 1 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & 2 & 4 \end{bmatrix}, \hat{\mathbf{x}}_3 = \begin{bmatrix} C_3 \\ D_3 \\ E_3 \end{bmatrix}, \text{ and } \mathbf{b}_3 = \begin{bmatrix} 2 \\ 0 \\ -3 \\ -5 \end{bmatrix}.$$

The least squares solution  $\hat{\mathbf{x}}_3$  can again be obtained by solving

$$\mathbf{A}_3^T \mathbf{A}_3 \hat{\mathbf{x}}_3 = \mathbf{A}_3^T \mathbf{b}_3$$

which is given by

$$\begin{bmatrix} 4 & 2 & 6 \\ 2 & 6 & 8 \\ 6 & 8 & 18 \end{bmatrix} \begin{bmatrix} C_3 \\ D_3 \\ E_3 \end{bmatrix} = \begin{bmatrix} -6 \\ -15 \\ -21 \end{bmatrix}.$$

Finally, we obtain

$$\begin{bmatrix} C_3 \\ D_3 \\ E_3 \end{bmatrix} = \begin{bmatrix} -3/10 \\ -12/5 \\ 0 \end{bmatrix}$$

and the best least squares parabola fit is given by  $b = C_3 + D_3 t + E_3 t^2 = -3/10 - (12/5)t$ . In this case, the best parabola fit is identical to the best straight line fit.

4. (a) Let  $f_1(x) = 1$ ,  $f_2(x) = x$ , and  $f_3(x) = x^2$ . By the Gram-Schmidt process, we can obtain

$$\begin{aligned} F_1(x) &= f_1(x) = 1 \\ \implies q_1(x) &= \frac{F_1(x)}{\|F_1\|} = \frac{1}{\sqrt{\int_{-2}^2 1^2 dx}} = \frac{1}{2}. \end{aligned}$$

$$\begin{aligned} F_2(x) &= f_2(x) - \langle q_1, f_2 \rangle q_1(x) = x - \int_{-2}^2 \frac{1}{2} x dx \cdot \frac{1}{2} = x \\ \implies q_2(x) &= \frac{F_2(x)}{\|F_2\|} = \frac{x}{\sqrt{\int_{-2}^2 x^2 dx}} = \frac{\sqrt{3}}{4} x. \end{aligned}$$

$$\begin{aligned} F_3(x) &= f_3(x) - \langle q_1, f_3 \rangle q_1(x) - \langle q_2, f_3 \rangle q_2(x) \\ &= x^2 - \int_{-2}^2 \frac{1}{2} x^2 dx \cdot \frac{1}{2} - \int_{-2}^2 \frac{\sqrt{3}}{4} x^3 dx \cdot \frac{\sqrt{3}}{4} x = x^2 - \frac{4}{3} \\ \implies q_3(x) &= \frac{F_3(x)}{\|F_3\|} = \frac{x^2 - 4/3}{\sqrt{\int_{-2}^2 (x^2 - 4/3)^2 dx}} = \frac{3\sqrt{5}}{16} x^2 - \frac{\sqrt{5}}{4}. \end{aligned}$$

Therefore,

$$q_1(x) = \frac{1}{2}, \quad q_2(x) = \frac{\sqrt{3}}{4} x, \quad \text{and} \quad q_3(x) = \frac{3\sqrt{5}}{16} x^2 - \frac{\sqrt{5}}{4}$$

form an orthonormal basis for the subspace spanned by 1,  $x$ , and  $x^2$ .

- (b) According to (a), we can write

$$x^2 + 2x = \frac{8}{3} q_1(x) + \frac{8}{\sqrt{3}} q_2(x) + \frac{16}{3\sqrt{5}} q_3(x).$$

5. (a) Yes, it is true. Since  $\mathbf{A}$  is not invertible, we have  $|\mathbf{A}| = 0$ . Then we can have  $|\mathbf{AB}| = |\mathbf{A}||\mathbf{B}| = 0 \cdot |\mathbf{B}| = 0$ . Hence  $\mathbf{AB}$  is not invertible.
- (b) No, it is false. For example, let  $\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$  and  $\mathbf{B} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$ . Then we have  $|\mathbf{A} - \mathbf{B}| = \begin{vmatrix} 1 & 0 \\ 0 & -1 \end{vmatrix} = -1$ , and  $|\mathbf{A}| - |\mathbf{B}| = 0 - 0 = 0$ . Hence  $|\mathbf{A} - \mathbf{B}| \neq |\mathbf{A}| - |\mathbf{B}|$ .
- (c) Yes, it is true. For a skew-symmetric matrix satisfies  $\mathbf{A}^T = -\mathbf{A}$ , we have  $|\mathbf{A}^T| = |-\mathbf{A}|$ . Since  $|\mathbf{A}^T| = |\mathbf{A}|$  and  $|-\mathbf{A}| = (-1)^n |\mathbf{A}|$ , we can obtain  $|\mathbf{A}| = (-1)^n |\mathbf{A}|$ . Therefore, if  $n$  is odd, we have  $|\mathbf{A}| = -|\mathbf{A}|$ , which implies  $|\mathbf{A}| = 0$ .
6. (a) Let  $S_n = |\mathbf{A}_n|$  where  $\mathbf{A}_n$  is an  $n$  by  $n$  matrix. For  $n \geq 3$ , we have

$$S_n = \begin{vmatrix} 3 & 1 & 0 & \cdots & 0 \\ 1 & & & & \\ 0 & & \mathbf{A}_{n-1} & & \\ \vdots & & & & \\ 0 & & & & \end{vmatrix} = \begin{vmatrix} 3 & 1 & 0 & 0 \cdots & 0 \\ 1 & 3 & 1 & 0 \cdots & 0 \\ 0 & 1 & & & \\ 0 & 0 & & \mathbf{A}_{n-2} & \\ \vdots & \vdots & & & \\ 0 & 0 & & & \end{vmatrix}.$$

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

$$\begin{aligned} S_n &= 3 \cdot (-1)^{1+1} |\mathbf{A}_{n-1}| + 1 \cdot (-1)^{1+2} \begin{vmatrix} 1 & 1 & 0 & \cdots & 0 \\ 0 & & & & \\ 0 & & \mathbf{A}_{n-2} & & \\ \vdots & & & & \\ 0 & & & & \end{vmatrix} \\ &= 3S_{n-1} - 1 \cdot (-1)^{1+1} |\mathbf{A}_{n-2}| \quad (\text{apply the cofactor formula to the first column}) \\ &= 3S_{n-1} - S_{n-2}. \end{aligned}$$

Then we can obtain  $a = 3$  and  $b = -1$ .

- (b) We have

$$\begin{aligned} S_1 &= 3 \\ S_2 &= 8 \\ S_3 &= 3S_2 - S_1 = 21 \\ S_4 &= 3S_3 - S_2 = 55 \\ S_5 &= 3S_4 - S_3 = 144. \end{aligned}$$

7. (a)

$$\begin{aligned}
 |\mathbf{A}_5| &= \begin{vmatrix} 0 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 0 & 1 & 1 \\ 1 & 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 & 0 \end{vmatrix} \\
 &= \begin{vmatrix} 0 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 0 & 1 & 1 \\ 1 & 1 & 1 & 0 & 1 \\ 4 & 4 & 4 & 4 & 4 \end{vmatrix} \quad [\text{add all rows (except the last) to the last row}] \\
 &= 4 \begin{vmatrix} 0 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 0 & 1 & 1 \\ 1 & 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 & 1 \end{vmatrix} \\
 &= 4 \begin{vmatrix} -1 & 0 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & -1 & 0 \\ 1 & 1 & 1 & 1 & 1 \end{vmatrix} \quad [\text{subtract the last row from each preceding row}] \\
 &= 4(-1)(-1)(-1)(-1)(1) \quad [\text{all other terms in the big formula are zero}] \\
 &= 4.
 \end{aligned}$$

(b) We have that the  $(1, 1)$  entry of  $\mathbf{A}_4^{-1}$  is equal to

$$(\mathbf{A}_4^{-1})_{11} = \frac{C_{11}}{\det(\mathbf{A}_4)} = \frac{\det(\mathbf{A}_3)}{\det(\mathbf{A}_4)} = \frac{\begin{vmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{vmatrix}}{\begin{vmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{vmatrix}} = -\frac{2}{3}.$$