You are **NOT** allowed to use any type of calculators.

#### 1 (7+8=15 pts)

Linear equations

Three people play a game in which there are always two winners and one looser. They have the understanding that the loser gives each winner an amount equal to what the winner already has. After three games, each has lost just once and each has  $\in 24$ .

- (a) Let  $x_k$  be the amount of money Player k begins with. Assume that Player 1 has lost the first game, Player 2 the second game, and Player 3 the third game. Write down three equations in terms of  $x_1$ ,  $x_2$ , and  $x_3$  for the amount of money each has after three games.
- (b) Solve the linear equations obtained above to determine  $x_1$ ,  $x_2$ , and  $x_3$ .

# REQUIRED KNOWLEDGE: linear equations, Gauss-elimination, row reduced echelon form.

#### SOLUTION:

1a: After the first round of game, the players will have

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Player 1: 
$$x_1 - (x_2 + x_3) = x_1 - x_2 - x_3$$
  
Player 2:  $x_2 + x_2 = 2x_2$   
Player 3:  $x_3 + x_3 = 2x_3$ 

as Player 1 has lost the round. After the second round, they will have

Player 1: 
$$(x_1 - x_2 - x_3) + (x_1 - x_2 - x_3) = 2x_1 - 2x_2 - 2x_3$$
  
Player 2:  $2x_2 - (x_1 - x_2 - x_3) - 2x_3 = -x_1 + 3x_2 - x_3$   
Player 3:  $2x_3 + 2x_3 = 4x_3$ 

and after the third

Player 1: 
$$(2x_1 - 2x_2 - 2x_3) + (2x_1 - 2x_2 - 2x_3) = 4x_1 - 4x_2 - 4x_3$$
  
Player 2:  $(-x_1 + 3x_2 - x_3) + (-x_1 + 3x_2 - x_3) = -2x_1 + 6x_2 - 2x_3$   
Player 3:  $4x_3 - (2x_1 - 2x_2 - 2x_3) - (-x_1 + 3x_2 - x_3) = -x_1 - x_2 + 7x_3$ 

as the losers are respectively, Player 2 and 3. As they all have  $\in 24$  after three rounds, we arrive at the linear equation:

$$4x_1 - 4x_2 - 4x_3 = 24$$
$$-2x_1 + 6x_2 - 2x_3 = 24$$
$$-x_1 - x_2 + 7x_3 = 24$$

1b: We first form the augmented matrix:

$$\begin{pmatrix} -1 & -1 & 7 & \vdots & 24 \\ -2 & 6 & -2 & \vdots & 24 \\ 4 & -4 & -4 & \vdots & 24 \end{pmatrix}$$

Then, we perform row operations to put the augmented matrix into the *reduced row echelon form*:

$$\begin{pmatrix} -1 & -1 & 7 & \vdots & 24 \\ -2 & 6 & -2 & \vdots & 24 \\ 4 & -4 & -4 & \vdots & 24 \end{pmatrix}^{1st=-1\times 1st} \begin{pmatrix} 1 & 1 & -7 & \vdots & -24 \\ -2 & 6 & -2 & \vdots & 24 \\ 4 & -4 & -4 & \vdots & 24 \end{pmatrix}^{2nd=2\times 1st+2nd} \begin{pmatrix} 1 & 1 & -7 & \vdots & -24 \\ 0 & 8 & -16 & \vdots & -24 \\ 4 & -4 & -4 & \vdots & 24 \end{pmatrix}^{2nd=2\times 1st+2nd} \begin{pmatrix} 1 & 1 & -7 & \vdots & -24 \\ 0 & 8 & -16 & \vdots & -24 \\ 4 & -4 & -4 & \vdots & 24 \end{pmatrix}^{3rd=-4\times 1st+3rd} \begin{pmatrix} 1 & 1 & -7 & \vdots & -24 \\ 0 & 8 & -16 & \vdots & -24 \\ 0 & 8 & -16 & \vdots & -24 \\ 0 & 8 & -16 & \vdots & -24 \\ 0 & -8 & 24 & \vdots & 120 \end{pmatrix}^{2nd=\frac{1}{6}\times 2nd} \begin{pmatrix} 1 & 1 & -7 & \vdots & -24 \\ 0 & 8 & -16 & \vdots & -24 \\ 0 & -8 & 24 & \vdots & 120 \end{pmatrix}^{2nd=\frac{1}{6}\times 2nd} \begin{pmatrix} 1 & 1 & -7 & \vdots & -24 \\ 0 & 1 & -2 & \vdots & -3 \\ 0 & -8 & 24 & \vdots & 120 \end{pmatrix}^{3rd=\frac{1}{6}\times 2nd} \begin{pmatrix} 1 & 1 & -7 & \vdots & -24 \\ 0 & 1 & -2 & \vdots & -3 \\ 0 & -8 & 24 & \vdots & 120 \end{pmatrix}^{3rd=\frac{1}{6}\times 3rd} \begin{pmatrix} 1 & 1 & -7 & \vdots & -24 \\ 0 & 1 & -2 & \vdots & -3 \\ 0 & -1 & 3 & \vdots & 15 \end{pmatrix}^{3rd=2nd+3rd} \begin{pmatrix} 1 & 1 & -7 & \vdots & -24 \\ 0 & 1 & -2 & \vdots & -3 \\ 0 & 0 & 1 & \vdots & 12 \end{pmatrix}^{3rd=2nd+3rd} \begin{pmatrix} 1 & 1 & -7 & \vdots & -24 \\ 0 & 1 & -2 & \vdots & -3 \\ 0 & 0 & 1 & \vdots & 12 \end{pmatrix}^{2nd=2\times 3rd+2nd} \begin{pmatrix} 1 & 1 & -7 & \vdots & -24 \\ 0 & 1 & 0 & \vdots & 21 \\ 0 & 0 & 1 & \vdots & 12 \end{pmatrix}^{3rd=2\times 3rd+2nd} \begin{pmatrix} 1 & 1 & 0 & \vdots & 60 \\ 0 & 1 & 0 & \vdots & 21 \\ 0 & 0 & 1 & \vdots & 12 \end{pmatrix}^{3rd=2\times 3rd+1st} \begin{pmatrix} 1 & 0 & 0 & \vdots & 39 \\ 0 & 1 & 0 & \vdots & 21 \\ 0 & 0 & 1 & \vdots & 12 \end{pmatrix}^{3rd=-1\times 2nd+1st} \begin{pmatrix} 1 & 0 & 0 & \vdots & 39 \\ 0 & 1 & 0 & \vdots & 21 \\ 0 & 0 & 1 & \vdots & 12 \end{pmatrix}^{3rd=-1\times 2nd+1st} \begin{pmatrix} 1 & 0 & 0 & \vdots & 39 \\ 0 & 1 & 0 & \vdots & 21 \\ 0 & 0 & 1 & \vdots & 12 \end{pmatrix}^{3rd=-1\times 2nd+1st} \begin{pmatrix} 1 & 0 & 0 & \vdots & 21 \\ 0 & 0 & 1 & \vdots & 12 \end{pmatrix}^{3rd=-1\times 2nd+1st} \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & \vdots & 21 \\ 0 & 0 & 1 & \vdots & 12 \end{pmatrix}^{3rd=-1\times 2nd+1st} \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & \vdots & 21 \\ 0 & 0 & 1 & \vdots & 12 \end{pmatrix}^{3rd=-1\times 2nd+1st} \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & \vdots & 21 \\ 0 & 0 & 1 & \vdots & 12 \end{pmatrix}^{3rd=-1\times 2nd+1st} \end{pmatrix}$$

There are no free variables and hence the solution is unique:  $x_1 = 39$ ,  $x_2 = 21$ , and  $x_3 = 12$ .

Let

$$M = \begin{bmatrix} A & C \\ 0 & B \end{bmatrix}$$

where all four blocks are  $n \times n$  matrices.

(a) If A and B are nonsingular matrices, show that M must also be nonsingular and that  $M^{-1}$  must be of the form

$$M^{-1} = \begin{bmatrix} A^{-1} & D\\ 0 & B^{-1} \end{bmatrix}$$

for some  $n \times n$  matrix D.

(b) Determine D.

# $REQUIRED\ KNOWLEDGE:$ nonsingular matrices, inverse of a matrix, partitioned matrices.

### SOLUTION:

**2a:** Let  $z \in \mathbb{R}^{2n}$  be a vector such that Mz = 0. Partition z as follows:

$$z = \begin{bmatrix} x \\ y \end{bmatrix}$$

where  $x, y \in \mathbb{R}^n$ . Then, we have

$$0 = Mz = \begin{bmatrix} A & C \\ 0 & B \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} Ax + Cy \\ By \end{bmatrix}.$$

Hence, we get Ax + Cy = By = 0. It follows from nonsingularity of B that y = 0. Then, we get Ax = 0. Since A is nonsingular, this can happen only if x = 0. Therefore, we have z = 0. Consequently, M is nonsingular.

Let the inverse of M be given by

$$M^{-1} = \begin{bmatrix} U & W \\ X & Y \end{bmatrix}$$

where all four blocks are  $n \times n$  matrices. Note that

$$I = MM^{-1} = \begin{bmatrix} A & C \\ 0 & B \end{bmatrix} \begin{bmatrix} U & W \\ X & Y \end{bmatrix} = \begin{bmatrix} AU + CX & AW + CY \\ BX & BY \end{bmatrix}.$$

This yields the following equations:

$$AU + CX = I$$
$$BX = 0$$
$$BY = I.$$

Since B is nonsingular, we get X = 0,  $Y = B^{-1}$ , and AU = I. Since A is nonsingular, the latter results in  $U = A^{-1}$ . Therefore,  $M^{-1}$  is of the form

$$M^{-1} = \begin{bmatrix} A^{-1} & D\\ 0 & B^{-1} \end{bmatrix}$$

for some matrix D.

**2b:** Note that

$$I = MM^{-1} = \begin{bmatrix} A & C \\ 0 & B \end{bmatrix} \begin{bmatrix} A^{-1} & D \\ 0 & B^{-1} \end{bmatrix} = \begin{bmatrix} I & AD + CB^{-1} \\ 0 & I \end{bmatrix}.$$

This yields  $AD + CB^{-1} = 0$ . By multiplying from left by  $A^{-1}$ , we obtain  $D = -A^{-1}CB^{-1}$ .

Let M be an  $4 \times 4$  matrix with the characteristic polynomial  $p_M(\lambda) = \lambda(\lambda - 1)(\lambda^2 + 1)$ .

- (a) Is M nonsingular? Explain.
- (b) Determine the eigenvalues of M.
- (c) Determine the determinant of M.
- (d) Determine the rank of M.
- (e) Is M diagonalizable? Explain.

## REQUIRED KNOWLEDGE: nonsingularity, eigenvalues, determinant, rank, diagonalizability.

### SOLUTION:

**3a:** A square matrix is singular if and only if one of its eigenvalues is zero. Since M has a zero eigenvalue, it has to be singular.

**3b:** Eigenvalues of a square matrix are the roots of its characteristic polynomial. By solving  $p_M(\lambda) = 0$ , we get  $\lambda_1 = 0$ ,  $\lambda_2 = 1$ , and  $\lambda_{3,4} = \pm i$  as eigenvalues.

**3c:** The determinant is equal to the product of the eigenvalues. Then, we get det(M) = 0.

**3d:** Every eigenvector corresponding a nonzero eigenvalue belong to the column space of the matrix. Since eigenvectors corresponding to distinct eigenvalues are linearly independent and M has three distinct nonzero eigenvalue, there must at least three linearly independent vectors in the column space of M. Then, we have

$$\operatorname{rank}(M) = \dim(\mathcal{R}(M)) \ge 3.$$

Since every eigenvector corresponding to the zero eigenvalue must belong to the null space of M, we have

$$\operatorname{null}(M) = \dim(\mathcal{N}(M)) \ge 1.$$

From the rank-nullity theorem, we know that

$$\operatorname{rank}(M) + \operatorname{null}(M) = 4.$$

Together with the above inequalities, this implies that

 $\operatorname{rank}(M) = 3$  and  $\operatorname{null}(M) = 1$ .

**3e:** Since M has distinct eigenvalues, it is diagonalizable.

Consider the vector space of  $k \times k$  matrices, i.e.  $\mathbb{R}^{k \times k}$ . Let  $N_k$  denote the set of all nonsingular  $k \times k$  matrices,  $S_k$  denote the set of all  $k \times k$  symmetric matrices, and  $T_k$  denote the set of all  $k \times k$  skew-symmetric matrices.

- (a) Show that  $N_k$  is not a subspace of  $\mathbb{R}^{k \times k}$ . Show that both  $S_k$  and  $T_k$  are subspaces of  $\mathbb{R}^{k \times k}$ .
- (b) Determine a basis for  $S_3$  as well as its dimension. What would be the dimension of  $S_k$ ?
- (c) Determine a basis for  $T_3$  as well as its dimension. What would be the dimension of  $T_k$ ?
- (d) Show that  $\mathbb{R}^{k \times k}$  is the direct sum of  $S_k$  and  $T_k$ , that is  $\mathbb{R}^{k \times k} = S_k \oplus T_k$ .
- (e) Let  $L : \mathbb{R}^{k \times k} \to \mathbb{R}^{k \times k}$  be the transformation given by  $L(A) = \frac{A + A^T}{2}$ . Show that L is a linear transformation.

**Hint:** A square matrix M is called *skew-symmetric* if  $M = -M^T$ .

REQUIRED KNOWLEDGE: vector spaces, subspaces, basis, dimension, direct sum, linear transformations.

#### SOLUTION:

**4a:** Both  $I_k$  and  $-I_k$  are nonsingular and hence in  $N_k$ , but  $I_k + (-I_k) = 0$  is not nonsingular. This means that the set of nonsingular matrices  $N_k$  is not closed under matrix addition. As such,  $N_k$  is not a subspace.

Clearly, the zero matrix is symmetric and skew-symmetric at the same time. This means that both  $S_k$  and  $T_k$  are not empty sets. Since  $(A + B)^T = A^T + B^T$  and  $(\alpha A) = \alpha A^T$  for all matrices A, B and scalars  $\alpha$ . As such, we can conclude that both  $S_k$  and  $T_k$  are closed under matrix addition and scalar multiplication. Consequently, they are both subspaces.

**4b:** A matrix M belongs to  $S_3$  if and only if it is of the form

$$M = \begin{bmatrix} a & b & c \\ b & d & e \\ c & e & f \end{bmatrix}$$

for some real numbers a, b, c, d, e, and f. Then, we have

$$M = a \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} + b \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} + c \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{bmatrix} + d \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} + e \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} + f \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

Therefore, we can conclude that the six matrices above span  $S_3$ . Note that M = 0 if and only if a = b = c = d = e = f = 0. As such, these six matrices are linearly independent. Consequently, they form a basis for  $S_3$  and dim $(S_3) = 6$ .

The argument above suggests that the dimension of the subspace  $S_k$  would be the number of diagonal elements plus half of the number of the off-diagonal elements, that is

$$\dim(S_k) = k + \frac{k^2 - k}{2} = \frac{k^2 + k}{2} = \frac{k(k+1)}{2}.$$

**4c:** A matrix M belongs to  $T_3$  if and only if it is of the form

$$M = \begin{bmatrix} 0 & a & b \\ -a & 0 & c \\ -b & -c & 0 \end{bmatrix}$$

for some real numbers a, b, c, d, e, and f. Then, we have

$$M = a \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} + b \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ -1 & 0 & 0 \end{bmatrix} + c \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -1 & 0 \end{bmatrix}$$

In a similar fashion to the problem 4b, we can conclude that these three matrices form a basis and hence dim $(T_3) = 3$ . In addition, the dimension of the subspace  $T_k$  would be half of the number of the off-diagonal elements, that is

$$\dim(T_k) = \frac{k^2 - k}{2} = \frac{k(k-1)}{2}$$

**4d:** In order to prove that  $\mathbb{R}^{k \times k} = S_k \oplus T_k$ , one needs to show that every  $k \times k$  matrix M can be written as a unique sum of two matrices  $P \in S_k$  and  $Q \in T_k$ . For a given matrix M, let

$$P = \frac{M + M^T}{2}$$
 and  $Q = \frac{M - M^T}{2}$ .

Note that

$$P^{T} = (\frac{M + M^{T}}{2})^{T} = \frac{M^{T} + M}{2} = P \quad \text{and} \quad Q^{T} = (\frac{M - M^{T}}{2})^{T} = \frac{M^{T} - M}{2} = -Q.$$

In other words,  $P \in S_k$  and  $Q \in T_k$ . Observe that

$$P + Q = \frac{M + M^T}{2} + \frac{M - M^T}{2} = M.$$

Therefore, we proved that every matrix can be written as a sum of a symmetric and a skewsymmetric matrix. To prove that this sum is unique. Let  $P' \in S_k$  and  $Q' \in T_k$  be such that

$$M = P' + Q'.$$

Then, we have P + Q = P' + Q'. Equivalently, P - P' = Q' - Q. Since both  $S_k$  and  $T_k$  are subspaces, we can conclude that  $P - P' = Q' - Q \in S_k \cap T_k$ . In other words, the matrix P - P' is symmetric and skew-symmetric at the same time. This is possible only if P = P' and hence Q = Q'.

4e: Note that

$$L(A+B) = \frac{(A+B) + (A+B)^T}{2} = \frac{A+A^T}{2} + \frac{B+B^T}{2} = L(A) + L(B)$$

for any matrices  $A, B \in \mathbb{R}^{k \times k}$ .

Note also that

$$L(\alpha A) = \frac{\alpha A + (\alpha A)^T}{2} = \frac{\alpha A + \alpha A^T}{2} = \alpha \frac{A + A^T}{2} = \alpha L(A)$$

for any matrix  $A \in \mathbb{R}^{k \times k}$  and real number  $\alpha$ . Therefore, L is a linear transformation. A tridiagonal matrix is a square matrix that has nonzero elements only on the main diagonal, the first diagonal below the main diagonal, and the first diagonal above the main diagonal. Consider the sequence of tridiagonal matrices  $A_n \in \mathbb{R}^{n \times n}$  given by:

$$A_{1} = 3, \ A_{2} = \begin{bmatrix} 3 & 1 \\ 2 & 3 \end{bmatrix}, \ A_{3} = \begin{bmatrix} 3 & 1 & 0 \\ 2 & 3 & 1 \\ 0 & 2 & 3 \end{bmatrix}, \ A_{4} = \begin{bmatrix} 3 & 1 & 0 & 0 \\ 2 & 3 & 1 & 0 \\ 0 & 2 & 3 & 1 \\ 0 & 0 & 2 & 3 \end{bmatrix}, \ A_{5} = \begin{bmatrix} 3 & 1 & 0 & 0 & 0 \\ 2 & 3 & 1 & 0 & 0 \\ 0 & 2 & 3 & 1 & 0 \\ 0 & 0 & 2 & 3 & 1 \\ 0 & 0 & 0 & 2 & 3 \end{bmatrix}, \dots$$

Let  $d_n := \det(A_n)$ .

- (a) Verify that  $d_1 = 3$  and  $d_2 = 7$ .
- (b) By using cofactor expansion, determine the numbers p and q such that  $d_k = pd_{k-1} + qd_{k-2}$  for all  $k \ge 3$ .
- (c) Note that  $\begin{bmatrix} d_k \\ d_{k-1} \end{bmatrix} = \begin{bmatrix} p & q \\ 1 & 0 \end{bmatrix} \begin{bmatrix} d_{k-1} \\ d_{k-2} \end{bmatrix}$  for all  $k \ge 3$  where p and q are as obtained above. Define  $M = \begin{bmatrix} p & q \\ 1 & 0 \end{bmatrix}$ . Show that  $d_k = \begin{bmatrix} 1 & 0 \end{bmatrix} M^{k-2} \begin{bmatrix} d_2 \\ d_1 \end{bmatrix}$  for all  $k \ge 3$ .
- (d) Determine a nonsingular matrix X and a diagonal matrix D such that  $X^{-1}MX = D$ . Determine  $M^k$  for all  $k \ge 0$ . Determine  $d_k$  for all  $k \ge 1$ .

### REQUIRED KNOWLEDGE: determinants, diagonalization.

#### SOLUTION:

5a: By definition, we have

$$d_1 = \det(A_1) = \det(3) = 3$$
 and  $d_2 = \det(A_2) = \det\left(\begin{bmatrix} 3 & 1\\ 2 & 3 \end{bmatrix}\right) = 3 \cdot 3 - 1 \cdot 2 = 7.$ 

**5b:** Note that

 $d_k = \det(A_k)$ 

$$= 3 \det(A_{k-1}) - \det\left(\left[\frac{2}{0_{k-2}} | 1 \cdot 0 \cdots \cdot 0 - A_{k-2}\right]\right) \quad [\text{Here } 0_{k-2} \text{ denotes the zero vector of } \mathbb{R}^{k-2}]$$
$$= 3d_{k-1} - 2 \det(A_{k-2})$$
$$= 3d_{k-1} - 2d_{k-2}.$$

for all  $k \ge 3$ . Therefore, p = 3 and q = -2.

5c: Note that

$$\begin{bmatrix} d_3 \\ d_2 \end{bmatrix} = M \begin{bmatrix} d_2 \\ d_1 \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} d_4 \\ d_3 \end{bmatrix} = M \begin{bmatrix} d_3 \\ d_2 \end{bmatrix} = M^2 \begin{bmatrix} d_2 \\ d_1 \end{bmatrix}.$$

By repeating the same argument above, we obtain

$$\begin{bmatrix} d_k \\ d_{k-1} \end{bmatrix} = M^{k-2} \begin{bmatrix} d_2 \\ d_1 \end{bmatrix}$$

for all  $k \ge 3$ . By pre-multiplying by  $\begin{bmatrix} 1 & 0 \end{bmatrix}$ , we obtain

$$d_k = \begin{bmatrix} 1 & 0 \end{bmatrix} M^{k-2} \begin{bmatrix} d_2 \\ d_1 \end{bmatrix}.$$

**5d:** Note that

$$\det(M - \lambda I) = \det\left(\begin{bmatrix} 3 - \lambda & -2\\ 1 & -\lambda \end{bmatrix}\right) = \lambda^2 - 3\lambda + 2 = (\lambda - 1)(\lambda - 2).$$

Therefore, the eigenvalues of the matrix M are  $\lambda_1 = 1$  and  $\lambda_2 = 2$ . To compute eigenvectors, we should solve

$$0 = (M - \lambda_1 I) x_1 = \begin{bmatrix} 2 & -2 \\ 1 & -1 \end{bmatrix} x_1$$
  
$$0 = (M - \lambda_2 I) x_2 = \begin{bmatrix} 1 & -2 \\ 1 & -2 \end{bmatrix} x_2.$$

These would yield, for instance,

$$x_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
 and  $x_2 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ .

Let X be defined by

$$X := \begin{bmatrix} x_1 & x_2 \end{bmatrix} = \begin{bmatrix} 1 & 2\\ 1 & 1 \end{bmatrix}.$$

Note that

$$X^{-1} = \begin{bmatrix} -1 & 2\\ 1 & -1 \end{bmatrix}$$

and also that

$$X^{-1}MX = \begin{bmatrix} -1 & 2\\ 1 & -1 \end{bmatrix} \begin{bmatrix} 3 & -2\\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2\\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0\\ 0 & 2 \end{bmatrix} = D.$$

Since  $M = XDX^{-1}$ , we know that  $M^k = XD^kX^{-1}$ . Hence, we get

$$M^{k} = \begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1^{k} & 0 \\ 0 & 2^{k} \end{bmatrix} \begin{bmatrix} -1 & 2 \\ 1 & -1 \end{bmatrix} = \begin{bmatrix} 2^{k+1} - 1 & 2 - 2^{k+1} \\ 2^{k} - 1 & 2 - 2^{k} \end{bmatrix}$$

for all  $k \ge 0$ . Note that

$$d_{k} = \begin{bmatrix} 1 & 0 \end{bmatrix} M^{k-2} \begin{bmatrix} d_{2} \\ d_{1} \end{bmatrix}$$
$$= \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} 2^{k-1} - 1 & 2 - 2^{k-1} \\ 2^{k-2} - 1 & 2 - 2^{k-2} \end{bmatrix} \begin{bmatrix} 7 \\ 3 \end{bmatrix}$$
$$= \begin{bmatrix} 2^{k-1} - 1 & 2 - 2^{k-1} \end{bmatrix} \begin{bmatrix} 7 \\ 3 \end{bmatrix}$$
$$= 7(2^{k-1} - 1) + 3(2 - 2^{k-1})$$
$$= 4 \cdot 2^{k-1} - 1$$
$$= 2^{k+1} - 1.$$

for all  $k \ge 3$ . Since  $d_1 = 3$  and  $d_2 = 7$ , we have  $d_k = 2^{k+1} - 1$  for all  $k \ge 1$ .

Given the data

determine a line y = ax + b that best fits the data in the least squares sense.

# REQUIRED KNOWLEDGE: least-squares problem.

# SOLUTION:

From the table, we obtain the linear equations

$$\begin{bmatrix} -1 & 1\\ 0 & 1\\ 1 & 1\\ 2 & 1 \end{bmatrix} \begin{bmatrix} a\\ b \end{bmatrix} = \begin{bmatrix} 1.5\\ 4.5\\ 6.5\\ 7.5 \end{bmatrix}.$$

The solution of the least-squares problem can be found by solving the following equation:

$$\begin{bmatrix} -1 & 0 & 1 & 2\\ 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} -1 & 1\\ 0 & 1\\ 1 & 1 & 2 & 1 \end{bmatrix} \begin{bmatrix} a\\ b \end{bmatrix} = \begin{bmatrix} -1 & 0 & 1 & 2\\ 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1.5\\ 4.5\\ 6.5\\ 7.5 \end{bmatrix}$$
$$\begin{bmatrix} 6 & 2\\ 2 & 4 \end{bmatrix} \begin{bmatrix} a\\ b \end{bmatrix} = \begin{bmatrix} -1.5 + 6.5 + 15\\ 1.5 + 4.5 + 6.5 + 7.5 \end{bmatrix} = \begin{bmatrix} 20\\ 20 \end{bmatrix}$$
$$\begin{bmatrix} 3 & 1\\ 1 & 2 \end{bmatrix} \begin{bmatrix} a\\ b \end{bmatrix} = \begin{bmatrix} 10\\ 10 \end{bmatrix}$$
$$\begin{bmatrix} a\\ b \end{bmatrix} = \begin{bmatrix} 3 & 1\\ 1 & 2 \end{bmatrix}^{-1} \begin{bmatrix} 10\\ 10 \end{bmatrix}$$
$$\begin{bmatrix} a\\ b \end{bmatrix} = \begin{bmatrix} 3 & 1\\ 1 & 2 \end{bmatrix}^{-1} \begin{bmatrix} 10\\ 10 \end{bmatrix}$$
$$\begin{bmatrix} a\\ b \end{bmatrix} = \frac{1}{5} \begin{bmatrix} 2 & -1\\ -1 & 3 \end{bmatrix} \begin{bmatrix} 10\\ 10 \end{bmatrix}$$
$$\begin{bmatrix} a\\ b \end{bmatrix} = \frac{1}{5} \begin{bmatrix} 10\\ 20 \end{bmatrix} = \begin{bmatrix} 2\\ 4 \end{bmatrix}.$$