

# Linear Algebra I

30/01/2014, Thursday, 9:00-12:00

---

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

---

1 (12+3=15 pts)

Linear equations

---

Consider the linear equation

$$\begin{bmatrix} 1 & 3 & 5 & -2 \\ 1 & 4 & 6 & -2 \\ -1 & -1 & -3 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} a \\ b \\ b - 2a \end{bmatrix}$$

where  $a$  and  $b$  are real numbers.

- (a) Find all values of  $a$  and  $b$  for which the equation is consistent. For these values find the general solution of the equation.
- (b) Find all values of  $a$  and  $b$  for which the equation has a unique solution.
- 

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

---

**SOLUTION:**

We begin with performing row operations in order to put the augmented matrix into the row echelon form:

$$\begin{aligned} & \begin{bmatrix} 1 & 3 & 5 & -2 & a \\ 1 & 4 & 6 & -2 & b \\ -1 & -1 & -3 & 2 & b - 2a \end{bmatrix} \xrightarrow{\text{2nd} = -1 \times \text{1st} + \text{2nd}} \begin{bmatrix} 1 & 3 & 5 & -2 & a \\ 0 & 1 & 1 & 0 & b - a \\ -1 & -1 & -3 & 2 & b - 2a \end{bmatrix} \\ & \begin{bmatrix} 1 & 3 & 5 & -2 & a \\ 0 & 1 & 1 & 0 & b - a \\ -1 & -1 & -3 & 2 & b - 2a \end{bmatrix} \xrightarrow{\text{3rd} = \text{1st} + \text{3rd}} \begin{bmatrix} 1 & 3 & 5 & -2 & a \\ 0 & 1 & 1 & 0 & b - a \\ 0 & 2 & 2 & 0 & b - a \end{bmatrix} \\ & \begin{bmatrix} 1 & 3 & 5 & -2 & a \\ 0 & 1 & 1 & 0 & b - a \\ 0 & 2 & 2 & 0 & b - a \end{bmatrix} \xrightarrow{\text{3rd} = -2 \times \text{2nd} + \text{3rd}} \begin{bmatrix} 1 & 3 & 5 & -2 & a \\ 0 & 1 & 1 & 0 & b - a \\ 0 & 0 & 0 & 0 & a - b \end{bmatrix}. \end{aligned}$$

**1a:** It is consistent if and only if  $a - b = 0$ , i.e.  $a = b$ . For these values, we can proceed with putting the augmented matrix into the row reduced echelon form in order to solve the equations:

$$\begin{bmatrix} 1 & 3 & 5 & -2 & a \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \xrightarrow{\text{1st} = -3 \times \text{2nd} + \text{1st}} \begin{bmatrix} 1 & 0 & 2 & -2 & a \\ 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}.$$

Therefore,  $x_1$  and  $x_2$  are the lead variables and  $x_3$  and  $x_4$  are the free. Then, the solution can be given as

$$x_1 = a - 2x_3 + 2x_4 \quad \text{and} \quad x_2 = -x_3.$$

**1b:** When  $a \neq b$ , the equations are inconsistent. When  $a = b$ , there are two free variables. Therefore, the system has never a unique solution.

---

Let

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

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

- (a) Show that  $M$  is nonsingular if and only if  $A - BC$  is nonsingular.  
 (b) Let  $D = A - BC$ . Suppose that  $D$  is nonsingular. Show that

$$M^{-1} = \begin{bmatrix} D^{-1} & -D^{-1}B \\ -CD^{-1} & I + CD^{-1}B \end{bmatrix}.$$

**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

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

Hence, we get  $Ax + By = 0$  and  $Cx + y = 0$ . From the latter, we get  $y = -Cx$  and hence  $(A - BC)x = 0$  from the former. Then, we can conclude that  $M$  is nonsingular if and only if so is  $A - BC$ .

**2b:** Note that

$$\begin{aligned} \begin{bmatrix} A & B \\ C & I \end{bmatrix} \begin{bmatrix} D^{-1} & -D^{-1}B \\ -CD^{-1} & I + CD^{-1}B \end{bmatrix} &= \begin{bmatrix} AD^{-1} - BCD^{-1} & -AD^{-1}B + B + BCD^{-1}B^{-1} \\ CD^{-1} - CD^{-1} & -CD^{-1}B + I + CD^{-1}B \end{bmatrix} \\ &= \begin{bmatrix} (A - BC)D^{-1} & -(A - BC)D^{-1}B + B \\ 0 & I \end{bmatrix} \\ &= \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix} \end{aligned}$$

since  $D = A - BC$ . Consequently, we have

$$M^{-1} = \begin{bmatrix} D^{-1} & -D^{-1}B \\ -CD^{-1} & I + CD^{-1}B \end{bmatrix}.$$

Let  $A \in \mathbb{R}^{m \times n}$ ,  $B \in \mathbb{R}^{n \times r}$ , and  $C = AB$ . Show that

- (a) the column space of  $C$  is a subspace of the column space of  $A$ .
  - (b) the row space of  $C$  is a subspace of the row space of  $B$ .
  - (c)  $\text{rank}(C) \leq \min\{\text{rank}(A), \text{rank}(B)\}$ .
- 

**REQUIRED KNOWLEDGE: column space and rank of a matrix**

---

**SOLUTION:**

**3a:** Let  $y \in R(C)$ . Then, there exists  $x \in \mathbb{R}^r$  such that  $y = Cx = ABx$ . Hence, we see that  $y \in R(A)$ . Therefore,  $R(C) \subseteq R(A)$ .

**3b:** Note that  $C^T = B^T A^T$ . By applying the previous result, we get  $R(C^T) \subseteq R(B^T)$ . In other words, the row space of  $C$  is a subspace of the row space of  $B$ .

**3c:** Note that  $\text{rank}(C) = \dim(R(C)) = \dim(R(C^T))$ . From 3a and 3b, we have

$$\begin{aligned}\text{rank}(C) &= \dim(R(C)) \leq \dim(R(A)) = \text{rank}(A) \\ \text{rank}(C) &= \dim(R(C^T)) \leq \dim(R(B^T)) = \text{rank}(B).\end{aligned}$$

By combining these inequalities, we obtain  $\text{rank}(C) \leq \min\{\text{rank}(A), \text{rank}(B)\}$ .

---

---

Consider the vector space of  $k \times k$  matrices, i.e.  $\mathbb{R}^{k \times k}$ .

(a) Let  $\lambda \in \mathbb{R}$  and

$$V_\lambda = \{A \in \mathbb{R}^{k \times k} \mid \lambda \text{ is an eigenvalue of } A\}.$$

Show that  $V_\lambda$  is *not* a subspace of  $\mathbb{R}^{k \times k}$ .

(b) Let  $x \in \mathbb{R}^k$  be a nonzero vector and

$$V_x = \{A \in \mathbb{R}^{k \times k} \mid x \text{ is an eigenvector of } A\}.$$

Show that  $V_x$  is a subspace of  $\mathbb{R}^{k \times k}$ .

---

**REQUIRED KNOWLEDGE: vector spaces, and subspaces.**

---

**SOLUTION:**

**4a:** Suppose that  $\lambda \neq 0$ . Then,  $V_\lambda$  is closed neither under scalar multiplication nor vector addition. To verify the latter, observe that  $0_{k \times k} \notin V_\lambda$ . To verify the former, observe that  $\lambda$  is not an eigenvalue of  $A + A = 2A$  even if it is an eigenvalue of  $A$ . Suppose, now, that  $\lambda = 0$ . Then,  $V_0$  is closed under scalar multiplication. However, it is not so under vector addition. For instance, one can take  $k = 2$ ,  $A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ , and  $B = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$  and observe that  $A, B \in V_0$  but  $A + B \notin V_0$ .

**4b:** Note first that the zero (or the identity) matrix belongs to  $V_x$ . As such,  $V_x$  is a nonempty set. To show that it is closed under scalar multiplication. Let  $A \in V_x$  and  $\alpha \in \mathbb{R}$ . Note that  $x$  is also an eigenvector of  $\alpha A$ . For the closedness under vector addition, let  $A, B \in V_x$  and note that  $x$  is an eigenvector of  $A + B$ .

---

Consider the matrix

$$M = \begin{bmatrix} 0 & -2 & 1 \\ 1 & 3 & -1 \\ 0 & 0 & \alpha \end{bmatrix}$$

where  $\alpha$  is a real number.

- Find the determinant of  $M$ .
- Find all values of  $\alpha$  for which  $M$  is nonsingular.
- Find the eigenvalues of  $M$ . [**Hint:**  $\alpha$  is an eigenvalue.]
- Find all values of  $\alpha$  for which  $M$  is diagonalizable.
- Let  $\alpha = 1$ . Find a nonsingular matrix  $T$  and a diagonal matrix  $D$  such that  $M = TDT^{-1}$ .

---

REQUIRED KNOWLEDGE: determinants, eigenvalues, and diagonalization.

---

SOLUTION:

**5a:** By using cofactor expansion with respect to the last row, we get

$$\det(M) = \alpha \det\left(\begin{bmatrix} 0 & -2 \\ 1 & 3 \end{bmatrix}\right) = 2\alpha.$$

**5b:** A square matrix is nonsingular if and only if its determinant is not zero. Therefore,  $M$  is nonsingular if and only if  $\alpha \neq 0$ .

**5c:** Note that

$$\det(M - \lambda I) = \det\left(\begin{bmatrix} -\lambda & -2 & 1 \\ 1 & 3 - \lambda & -1 \\ 0 & 0 & \alpha - \lambda \end{bmatrix}\right) = (\alpha - \lambda) \det\left(\begin{bmatrix} -\lambda & -2 \\ 1 & 3 - \lambda \end{bmatrix}\right) = (\alpha - \lambda)(\lambda^2 - 3\lambda + 2).$$

Therefore, the eigenvalues can be found as  $\lambda_1 = \alpha$ ,  $\lambda_2 = 1$ , and  $\lambda_3 = 2$ .

**5d:** An  $n \times n$  matrix is diagonalizable if and only if it has  $n$  linearly independent eigenvalues. In particular, this holds whenever the matrix has distinct eigenvalues. Therefore, we can conclude that  $M$  is diagonalizable if  $\alpha \notin \{1, 2\}$ . Then, we need to check two cases:  $\alpha = 1$  and  $\alpha = 2$ . For the case  $\alpha = 1$ , we can find the eigenvectors corresponding the eigenvalue  $\lambda = 1$  by solving the following linear equation:

$$0 = (M - I)x = \begin{bmatrix} -1 & -2 & 1 \\ 1 & 2 & -1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}.$$

This yields two linearly independent eigenvalues, for instance,  $x = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$  and  $x = \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}$ . As such, one can conclude that  $M$  is diagonalizable if  $\alpha = 1$ . For the case  $\alpha = 2$ , we can find the eigenvectors corresponding the eigenvalue  $\lambda = 2$  by solving the following linear equation:

$$0 = (M - 2I)x = \begin{bmatrix} -2 & -2 & 1 \\ 1 & 1 & -1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}.$$

This would yield at most one linearly independent eigenvector, say  $x = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$ . Hence, we can conclude that  $M$  is not diagonalizable whenever  $\alpha = 2$ . Consequently,  $M$  is diagonalizable if and

only if  $\alpha \neq 2$ .

**5e:** For  $\alpha = 1$ , we have already found two linearly independent eigenvalues corresponding to the eigenvalue  $\lambda = 1$  in the previous problem. For the eigenvalue  $\lambda = 2$ , we can find an eigenvector by solving:

$$0 = (M - 2I)x = \begin{bmatrix} -2 & -2 & 1 \\ 1 & 1 & -1 \\ 0 & 0 & -1 \end{bmatrix} x.$$

This would yield, for instance,  $x = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$ . Then, we can take

$$T = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & -1 \\ 1 & 2 & 0 \end{bmatrix} \quad \text{and} \quad D = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{bmatrix}.$$

---

Let  $x_i$  and  $y_i$  be real numbers with  $i = 1, 2, \dots, n$ . Suppose that

$$x_1 + x_2 + \cdots + x_n = y_1 + y_2 + \cdots + y_n = 0.$$

Show that the least squares solution of

$$\begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ \vdots & \vdots \\ x_n & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

is given by

$$a = \frac{x_1 y_1 + x_2 y_2 + \cdots + x_n y_n}{x_1^2 + x_2^2 + \cdots + x_n^2} \quad \text{and} \quad b = 0.$$

---

**REQUIRED KNOWLEDGE: least-squares problem.**

---

**SOLUTION:**

Note that the normal equations to solve are given by

$$\begin{aligned} \begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ \vdots & \vdots \\ x_n & 1 \end{bmatrix}^T \begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ \vdots & \vdots \\ x_n & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} &= \begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ \vdots & \vdots \\ x_n & 1 \end{bmatrix}^T \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} \\ \begin{bmatrix} x_1 & x_2 & \cdots & x_n \\ 1 & 1 & \cdots & 1 \end{bmatrix} \begin{bmatrix} x_1 & 1 \\ x_2 & 1 \\ \vdots & \vdots \\ x_n & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} &= \begin{bmatrix} x_1 & x_2 & \cdots & x_n \\ 1 & 1 & \cdots & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} \\ \begin{bmatrix} x_1^2 + x_2^2 + \cdots + x_n^2 & x_1 + x_2 + \cdots + x_n \\ x_1 + x_2 + \cdots + x_n & n \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} &= \begin{bmatrix} x_1 y_1 + x_2 y_2 + \cdots + x_n y_n \\ y_1 + y_2 + \cdots + y_n \end{bmatrix} \\ \begin{bmatrix} x_1^2 + x_2^2 + \cdots + x_n^2 & 0 \\ 0 & n \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} &= \begin{bmatrix} x_1 y_1 + x_2 y_2 + \cdots + x_n y_n \\ 0 \end{bmatrix}. \end{aligned}$$

Then, we have

$$\begin{aligned} \begin{bmatrix} a \\ b \end{bmatrix} &= \begin{bmatrix} x_1^2 + x_2^2 + \cdots + x_n^2 & 0 \\ 0 & n \end{bmatrix}^{-1} \begin{bmatrix} x_1 y_1 + x_2 y_2 + \cdots + x_n y_n \\ 0 \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{x_1^2 + x_2^2 + \cdots + x_n^2} & 0 \\ 0 & \frac{1}{n} \end{bmatrix} \begin{bmatrix} x_1 y_1 + x_2 y_2 + \cdots + x_n y_n \\ 0 \end{bmatrix} \\ &= \begin{bmatrix} \frac{x_1 y_1 + x_2 y_2 + \cdots + x_n y_n}{x_1^2 + x_2^2 + \cdots + x_n^2} \\ 0 \end{bmatrix}. \end{aligned}$$


---