

Assignment 3, Linear Algebra

Spring 2021

Answers

1. The given system of equations is

$$a + b + 2c + d = 1,$$

$$a - b - c + d = 0,$$

$$b + c = -1,$$

$$a + b + d = 2.$$

The corresponding augmented matrix is

$$\begin{bmatrix} 1 & 1 & 2 & 1 & 1 \\ 1 & -1 & -1 & 1 & 0 \\ 0 & 1 & 1 & 0 & -1 \\ 1 & 1 & 0 & 1 & 2 \end{bmatrix}.$$

We can apply the following linear transformations to reduce this matrix to its REF (Gaussian elimination):

$$R_2 \leftrightarrow R_3, R_3 \rightarrow R_3 - R_1, R_4 \rightarrow R_4 - R_1$$

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

$$R_3 \rightarrow R_3 + 3R_2$$

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

$$R_3 \rightarrow R_3 - R_4$$

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

$$R_4 \rightarrow R_4 - 2R_3$$

$$\begin{bmatrix} 1 & 1 & 2 & 1 & 1 \\ 0 & 1 & 1 & 0 & -1 \\ 0 & 0 & -1 & 0 & -3 \\ 0 & 0 & 0 & 0 & 5 \end{bmatrix}$$

This is the REF of the augmented matrix. When we convert it back to linear equations, we get $0 = 5$, which is clearly absurd; hence the given system of equations has no solution.

2. The given system of equations is

$$2a + b = 3,$$

$$4a + b = 7,$$

$$2a + 5b = -1.$$

The corresponding augmented matrix is

$$\begin{bmatrix} 2 & 1 & 3 \\ 4 & 1 & 7 \\ 2 & 5 & -1 \end{bmatrix}.$$

We can apply the following linear transformations to reduce this matrix to its REF (Gaussian elimination):

$$R_2 \rightarrow R_2 - 2R_1, R_3 \rightarrow R_3 - R_1$$

$$\begin{bmatrix} 2 & 1 & 3 \\ 0 & -1 & 1 \\ 0 & 4 & -4 \end{bmatrix}$$

$$R_4 \rightarrow R_4 + 4R_2$$

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

This is the REF of the augmented matrix. When we convert the nonzero rows back to linear equations, we get

$$2a + b = 3,$$

$$-b = 1,$$

which we can solve by back-substitution to obtain the solution

$$b = -1,$$

$$a = 2.$$

3. The given pair of planes is

$$P_1 \equiv 3x + 2y + z = -1,$$

$$P_2 \equiv 2x - y + 4z = 5.$$

In order to find the line of intersection of these planes, we will use Gaussian elimination to solve the system of equations.

The augmented matrix is

$$\begin{bmatrix} 3 & 2 & 1 & -1 \\ 2 & -1 & 4 & 5 \end{bmatrix}.$$

We can apply the following linear transformations to reduce this matrix to its REF:

$$R_2 \rightarrow 3R_2$$

$$\begin{bmatrix} 3 & 2 & 1 & -1 \\ 6 & -3 & 12 & 15 \end{bmatrix}$$

$$R_2 \rightarrow R_2 - 3R_1$$

$$\begin{bmatrix} 3 & 2 & 1 & -1 \\ 0 & -7 & 10 & 17 \end{bmatrix}$$

This is the REF of the augmented matrix. We note that since there are only two equations, there must be one free variable in the system, which will act as a parameter. Let this variable be $z = t$. Converting the matrix in REF back to linear equations, we get

$$3x + 2y + z = -1,$$

$$-7y + 10z = 17,$$

and solving by back-substitution, we see that

$$y = \frac{10t - 17}{7},$$

$$x = \frac{27}{7}(1 - t),$$

$$z = t.$$

4. The coefficient matrix is

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

Its determinant is $|A| = 4$.

The matrices obtained by substituting the constant terms are

$$A_1 = \begin{bmatrix} 1 & 1 & -1 \\ 2 & 1 & 1 \\ 3 & -1 & 0 \end{bmatrix},$$

$$A_2 = \begin{bmatrix} 1 & 1 & -1 \\ 1 & 2 & 1 \\ 1 & 3 & 0 \end{bmatrix},$$

$$A_3 = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 2 \\ 1 & -1 & 3 \end{bmatrix}.$$

Their determinants are $|A_1| = 9$, $|A_2| = -3$, $|A_3| = 2$.

Therefore, by Cramer's Rule, we can obtain the solution

$$x = \frac{|A_1|}{|A|} = \frac{9}{4},$$

$$y = \frac{|A_2|}{|A|} = \frac{-3}{4},$$

$$z = \frac{|A_3|}{|A|} = \frac{2}{4} = \frac{1}{2}.$$

5. The coefficient matrix is

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

Its determinant is $|A| = 2$.

The matrices obtained by substituting the constant terms are

$$A_1 = \begin{bmatrix} 1 & 1 & -3 \\ 1 & 1 & 1 \\ 1 & 0 & 1 \end{bmatrix},$$

$$A_2 = \begin{bmatrix} 2 & 1 & -3 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \end{bmatrix},$$

$$A_3 = \begin{bmatrix} 2 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}.$$

Their determinants are $|A_1| = 4, |A_2| = 0, |A_3| = 2$.

Therefore, by Cramer's Rule, we can obtain the solution

$$x = \frac{|A_1|}{|A|} = \frac{4}{2} = 2,$$

$$y = \frac{|A_2|}{|A|} = \frac{0}{2} = 0,$$

$$z = \frac{|A_3|}{|A|} = \frac{2}{4} = \frac{1}{2}.$$

6. We know that $\det(XY) = \det(X)\det(Y)$ for any two matrices X and Y . Therefore,

$$\det(AB) = \det(A)\det(B),$$

which we know is equal to $\det(B)\det(A)$ [commutativity of multiplication on real numbers]. But this is the same as $\det(BA)$.

Hence, we can conclude that $\det(AB) = \det(BA)$, QED.

7. If A is idempotent, we know that $A^2 = A$. Taking the determinant on both sides, we get $\det(A^2) = \det(A)$. From the property proved above, we get $\det(A)\det(A) = \det(A) \implies \det(A)(\det(A) - 1) = 0$. Therefore the only possible values of $\det(A)$ are 0 and 1.

8. We need to show that every vector in $\text{null}(A)$ is orthogonal to every vector in $\text{row}(A)$, *i.e.*, $n \cdot r = 0$ for all $x \in \text{null}(A), r \in \text{row}(A)$.
Let the rows of A be the vectors r_1, r_2, \dots, r_m , *i.e.*, if

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix},$$

then $r_i = (a_{i1}, a_{i2}, \dots, a_{in})$.

We know that, for all $x \in \text{null}(A)$, $A \cdot n = 0$. Expanding this, we see that

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}.$$

By the definition of matrix multiplication, we see that $0 = \sum_{j=1}^n a_{ij} \cdot x_j = r_i \cdot x$, for all r_i .

Therefore every row vector of A is orthogonal to every $x \in \text{null}(A)$.

Consider an arbitrary vector $v \in \text{row}(A)$. We can write $v = c_1 r_1 + c_2 r_2 + \cdots + c_k r_k$. When we take $v \cdot x$ for an arbitrary $x \in \text{null}(A)$, we get $(c_1 r_1 + c_2 r_2 + \cdots + c_k r_k) \cdot x = c_1(r_1 \cdot x) + c_2(r_2 \cdot x) + \cdots + c_k(r_k \cdot x)$, by the distributive law.

Since $r_i \cdot x = 0$ for all r_i , this is equal to $c_1 \cdot 0 + c_2 \cdot 0 + \cdots + c_k \cdot 0 = 0$. Therefore $v \cdot x = 0$ for all $x \in \text{null}(A), v \in \text{row}(A)$, QED.

9. We need to prove that if U is invertible then $\text{rank}(UA) = \text{rank}(A)$. Consider the equivalent transformations U_T and A_T .

- (i) First, we show that the rank of a matrix is equal to the rank of the corresponding transformations, *i.e.*, $\text{rank}(X) = \text{rank}(X_T)$. The rank of a matrix is the number of linearly independent column vectors it contains. Consider the set $M = \{Xv | v \in \text{dom}(X_T)\}$. By the definition of matrix multiplication, M is the set of linear combinations of the columns of X , *i.e.*, the set spanned by the columns of X . Let C be the set of column vectors of X .

We know that if a set of linearly dependent vectors C spans M , then there is some linearly independent subset of C that spans M and therefore forms a basis. Hence, if C is linearly independent, it forms a basis for M ; otherwise some (linearly independent) subset of C does. Either way, the number of linearly independent vectors in C is the dimension of M . Therefore, the rank of X is the dimension of M .

However, we can see from the definition of M that it is nothing but $\text{range}(X_T)$, since $X_T(v) = Xv$. Therefore the rank of X is equal to

the dimension of the range of X_T ; but this is nothing but the rank of X_T . Hence we can conclude that the rank of X is equal to the rank of X_T .

- (ii) Now, let $\text{rank}(A) = n$ be the rank of the matrix A , which is also the rank of the linear transformation A_T . Let $B = \{b_1, b_2, \dots, b_n\}$ be a basis of $\text{range}(A_T)$. Then, consider the set $B_U = \{U_T(b_1), U_T(b_2), \dots, U_T(b_n)\}$. We will show that this set forms a basis of the transformation $U_T \circ A_T$, which corresponds to the matrix UA . First, we will show that B_U is linearly independent, which we can do by contradiction. Suppose B_U is linearly dependent, *i.e.* $c_1 U_T(b_1) + c_2 U_T(b_2) + \dots + c_n U_T(b_n) = U_T(c_1 b_1 + c_2 b_2 + \dots + c_n b_n) = 0$ for some set of c_i not all zero. Since U_T is invertible, only the null vector 0 is such that $U_T(0) = 0$. Therefore $c_1 b_1 + c_2 b_2 + \dots + c_n b_n = 0$, *i.e.*, B is linearly dependent, which is a contradiction. Therefore B_U is linearly independent.

Next, we will show that B_U spans $\text{range}(U_T \circ A_T)$. Given an arbitrary vector $U_T(A_T(x)) \in \text{range}(U_T \circ A_T)$, we can express it as $U_T(c_1 b_1 + c_2 b_2 + \dots + c_n b_n) = c_1 U_T(b_1) + c_2 U_T(b_2) + \dots + c_n U_T(b_n) \in \text{span}(B_U)$. Therefore $\text{range}(U_T \circ A_T) \subseteq \text{span}(B_U)$. Similarly, given a vector $x = c_1 U_T(b_1) + c_2 U_T(b_2) + \dots + c_n U_T(b_n) \in \text{span}(B_U)$, we can apply linearity and write $x = U_T(c_1 b_1 + c_2 b_2 + \dots + c_n b_n)$. But $c_1 b_1 + c_2 b_2 + \dots + c_n b_n \in \text{range}(A_T)$. Therefore $x \in \text{range}(U_T \circ A_T)$, which means that $\text{span}(B_U) \subseteq \text{range}(U_T \circ A_T)$.

From the above, we can conclude that $\text{span}(B_U) = \text{range}(U_T \circ A_T)$. Taken together with the fact that B_U is linearly independent, we see that B_U is a basis of $\text{range}(U_T \circ A_T)$. This means that the dimension of $\text{range}(U_T \circ A_T)$ is $|B_U| = |B| = n$.

Now, we have shown that $\text{range}(U_T \circ A_T) = \text{rank}(UA)$, which we know is n . Further, by assumption, $\text{rank}(A) = n$. Therefore $\text{rank}(UA) = \text{rank}(A)$, QED.

10. We need to prove that $\text{rank}(A + B) \leq \text{rank}(A) + \text{rank}(B)$. We have seen above that the rank of X is equal to the dimension of $\text{range}(X_T)$. Further, we know that the matrix $A + B$ corresponds to the transformation $A_T + B_T$. We will prove the given statement by contradiction.

Let $\text{rank}(A) = r$, $\text{rank}(B) = s$, $\text{rank}(A + B) = m$, where $m > r + s$. Further, suppose that $B_A = \{a_1, a_2, \dots, a_r\}$, $B_B = \{b_1, b_2, \dots, b_s\}$, $B_C = \{c_1, c_2, \dots, c_m\}$ are bases of $\text{range}(A)$, $\text{range}(B)$ and $\text{range}(A + B)$ respectively.

Consider the set $X = B_A \cup B_B$; we know that $|X| \leq r + s$. Now, any vector v in $\text{range}(A_T + B_T)$ can be expressed as $v = A_T(x) + B_T(x)$. But $A_T(x) \in \text{range}(A_T)$, and it can therefore be expressed as a linear combination of vectors in $\text{range}(B_A)$. Similarly, $B_T(x)$ can be expressed as a linear combination of vectors in $\text{range}(B_B)$.

This means that v is also expressible as a linear combinations of the vectors in $B_A \cup B_B$, *i.e.* of the vectors in X . But v is an arbitrary vector in $\text{range}(A_T + B_T)$, so $\text{range}(A_T + B_T) \subseteq \text{span}(X)$. Therefore X spans $\text{range}(A_T + B_T)$.

But the dimension of $\text{range}(A_T + B_T)$ is $|B_C| = m > |X|$. Therefore $\text{range}(A_T + B_T)$ is spanned by a set smaller than its basis, which is a contradiction. Therefore $m > r + s$ cannot be true.

Therefore, $m \leq r + s$, i.e., $\text{rank}(A_T + B_T) \leq \text{rank}(A_T) + \text{rank}(B_T)$. From the above property, however, $\text{rank}(A + B) \leq \text{rank}(A) + \text{rank}(B)$, QED.

11. [Assuming that the vector space is P^1 , the space of polynomials of degree ≤ 1] We need to find that change of basis matrix from $B = [1, x]$ to $C = [x, x + 1]$. Let this matrix be A .

The columns of A are simply the basis vectors in B , expressed in terms of the basis vectors in C .

Therefore, column 1 of A is

$$[b_1]_C = [1]_C = [(-1) \cdot x + (1) \cdot (x + 1)]_C = \begin{bmatrix} -1 \\ 1 \end{bmatrix}.$$

Similarly, column 2 of A is

$$[b_2]_C = [x]_C = [(1) \cdot x + (0) \cdot (x + 1)]_C = \begin{bmatrix} 1 \\ 0 \end{bmatrix}.$$

Hence, the change of basis matrix is

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

We can apply this to $p(x) = 2 - x$, which is $[p(x)]_B = \begin{bmatrix} 2 \\ -1 \end{bmatrix}$. To express $p(x)$ in the basis C , we premultiply $[p(x)]_B$ with A . Therefore,

$$[p(x)]_C = A[p(x)]_B = \begin{bmatrix} -1 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 2 \\ -1 \end{bmatrix} = \begin{bmatrix} -3 \\ 2 \end{bmatrix}.$$

We can verify this; $(-3) \cdot (x) + (2) \cdot (x + 1) = -3x + 2x + 2 = 2 - x = p(x)$.

12. The Taylor expansion of a function $f(x)$ about $x = a$ is

$$f(a) + \frac{f'(a)}{1}(x - a) + \frac{f''(a)}{2}(x - a)^2 + \frac{f'''(a)}{6}(x - a)^3 + \dots$$

When we let $f(x) = p(x)$, we see that since $p(x)$ is a quadratic polynomial, all its derivatives of the third degree and above vanish. Therefore only the first three terms written above contribute to the Taylor expansion.

The values of the derivatives are

$$p(-2) = [1 + 2x - 5x^2]_{x=-2} = 1 + 2(-2) - 5(-2)^2 = 1 - 4 - 20 = -23,$$

$$p'(-2) = [2 - 10x]_{x=-2} = 2 - 10(-2) = 2 + 20 = -22,$$

$$p''(-2) = [2]_{x=-2} = 2,$$

and

$$p'''(-2) = 0.$$

Therefore, the Taylor expansion of $p(x)$ is

$$\begin{aligned}p(-2) + \frac{p'(-2)}{1}(x+2) + \frac{p''(-2)}{2}(x+2)^2 \\&= (-23) + \frac{22}{2}(x+2) + \frac{2}{6}(x+2)^2 \\&= -23 - 11(x+2) + \frac{1}{3}(x+2)^2.\end{aligned}$$