

Matrix Equations - Matrix Inversion - Invertible Matrix Theorem - Matrix Partitioning - Matrix Factorization

1. Given,

$$\mathbf{A}(\theta) = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}, \quad \theta \in (0, 2\pi]. \quad (1)$$

We now consider the *action* of  $\mathbf{A}$  on vectors from  $\mathbb{R}^2$ . That is, we wish to study the effect of the linear transformation  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  represented by the matrix  $\mathbf{A}(\theta)$  where  $\theta \in (0, 2\pi]$ .

- First show that the transformation is one-to-one.<sup>1</sup>
- Given this matrix representation of  $T$  find the matrix representation of the inverse transformation. That is find  $\mathbf{A}^{-1}$ .
- Let  $\mathbf{x} = [1 \ 0]^T$ . Describe or draw the action of the linear transformation  $\mathbf{A}\mathbf{x}$  for  $\theta \in S$  where  $S = \left\{0, \frac{\pi}{6}, \frac{\pi}{3}, \frac{\pi}{2}\right\}$ . What would the action of  $\mathbf{A}^{-1}$  be?
- Let  $\mathbf{A}(\theta)\mathbf{x} = \mathbf{b}$  for each  $\theta \in S$ . Calculate,  $\frac{\mathbf{x} \cdot \mathbf{b}}{|\mathbf{x}||\mathbf{b}|}$ .<sup>2</sup> How is this related to  $\theta$ ?<sup>3</sup>
- If we define the derivative of a matrix function as a matrix of derivatives then a typical product rule results. That is, if  $\mathbf{A}, \mathbf{B}$  have elements, which are functions of  $\theta$  then  $\frac{d[\mathbf{A}\mathbf{B}]}{d\theta} = \mathbf{A} \frac{d\mathbf{B}}{d\theta} + \frac{d\mathbf{A}}{d\theta} \mathbf{B}$ .<sup>4</sup> Using this and the identity  $\mathbf{A}\mathbf{A}^{-1} = \mathbf{I}$  to prove that  $\frac{d[\mathbf{A}^{-1}]}{d\theta} = -\mathbf{A}^{-1} \frac{d[\mathbf{A}]}{d\theta} \mathbf{A}^{-1}$ . Verify this formula using the matrix given above.

2. Given,

$$\mathbf{A} = \begin{bmatrix} 3 & 6 & 7 \\ 0 & 2 & 1 \\ 2 & 3 & 4 \end{bmatrix}.$$

- Calculate  $\mathbf{A}^{-1}$  and check your result with the appropriate matrix multiplication.
- Let  $\mathbf{A}_{\text{left}}^{-1} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T$ . Prove that  $\mathbf{A}_{\text{left}}^{-1}$  exists and show that  $\mathbf{A}_{\text{left}}^{-1} \mathbf{A} = \mathbf{I}$ .<sup>5</sup>
- Let  $\mathbf{A}_{\text{right}}^{-1} = \mathbf{A}^T (\mathbf{A} \mathbf{A}^T)^{-1}$ . Prove that  $\mathbf{A}_{\text{right}}^{-1}$  exists and show that  $\mathbf{A} \mathbf{A}_{\text{right}}^{-1} = \mathbf{I}$ .<sup>6</sup>
- Let  $\mathbf{A}_1 = [2 \ 2]^T$  and  $\mathbf{A}_2 = [2 \ 2]$ . Using the previous formula find the left-inverse of  $\mathbf{A}_1$  and the right-inverse of  $\mathbf{A}_2$ . Check your results with the appropriate multiplication.

<sup>1</sup>Recall that a transformation is one-to-one if and only if  $T(\mathbf{x}) = \mathbf{0}$  has only the trivial solution.

<sup>2</sup>Recall that  $\mathbf{x} \cdot \mathbf{y}$  and  $|\mathbf{x}|$  are the standard dot-product and magnitude, respectively, from vector-calculus. These operations hold for vectors in  $\mathbb{R}^n$  it now have the following definitions,  $\mathbf{x} \cdot \mathbf{y} = \mathbf{x}^T \mathbf{y}$  and  $|\mathbf{x}| = \sqrt{\mathbf{x}^T \mathbf{x}}$ .

<sup>3</sup>What we are trying to extract here is the standard result from calculus, which relates the dot-product or inner-product on vectors to the angle between them. This is clear when we have vectors in  $\mathbb{R}^2$  or  $\mathbb{R}^3$  since we have tools from trigonometry and geometry but when treating vectors in  $\mathbb{R}^n, n \geq 4$  these tools are no longer available. However, we would still like to have similar results to those of  $\mathbb{R}^n, n = 2, 3$ . To make a long story short, we will have these results for arbitrary vectors in  $\mathbb{R}^n$  but not immediately. The first thing we must do is show that  $|\mathbf{x} \cdot \mathbf{y}| \leq |\mathbf{x}||\mathbf{y}|$ , which is known as Schwarz's inequality. Without this we cannot be permitted to always relate  $\frac{\mathbf{x} \cdot \mathbf{b}}{|\mathbf{x}||\mathbf{b}|}$  to  $\theta$  via inverse trigonometric functions. These details will occur in chapter 6 where we find that by using the inner-product on vectors from  $\mathbb{R}^n$  we will define the notion of angle and from that distance. Using these definitions and Schwarz's inequality will then give us a triangle-inequality for arbitrary finite-dimensional vectors. This is to say that the algebra of vectors in  $\mathbb{R}^n$  carries its own definition of angle and length - very nice of it don't you think? Also, it should be noted that these results exist for certain so-called infinite-dimensional spaces, but are harder to prove - of course, and that the study of linear transformations of such spaces is the general setting for quantum mechanics - see MATH503:Functional Analysis for more details.

<sup>4</sup>To see why this is true differentiate an arbitrary element of  $\mathbf{A}\mathbf{B}$  to find  $\frac{d}{d\theta} [\mathbf{A}\mathbf{B}]_{ij} = \frac{d}{d\theta} \sum_{k=1}^n a_{ik} b_{kj} = \sum_{k=1}^n \frac{da_{ik}}{d\theta} b_{kj} + a_{ik} \frac{db_{kj}}{d\theta}$ .

<sup>5</sup>This matrix is called the left-inverse of  $\mathbf{A}$  and it can be shown that if  $\mathbf{A} \in \mathbb{R}^{m \times n}$  such that  $\mathbf{A}$  has a pivot in every column then the left inverse exists.

<sup>6</sup>This matrix is called the right-inverse of  $\mathbf{A}$  and it can be shown that if  $\mathbf{A} \in \mathbb{R}^{m \times n}$  such that  $\mathbf{A}$  has a pivot in every row then the right inverse exists.



3. Noting any theorems used from class or the text, prove the following statements:

- (a) If  $\mathbf{A}$  is an  $n \times n$  matrix and  $\mathbf{A}^{-1}$  exists, then the columns of  $\mathbf{A}$  span  $\mathbb{R}^n$ .
- (b) If  $\mathbf{A}$  is an  $n \times n$  matrix and  $\mathbf{Ax} = \mathbf{b}$  has a solution for each  $\mathbf{b} \in \mathbb{R}^n$ , then  $\mathbf{A}$  is invertible.
- (c) If the matrix  $\mathbf{A}$  is invertible, then the columns of  $\mathbf{A}^{-1}$  are linearly independent.
- (d) If the equation  $\mathbf{Ax} = \mathbf{b}$ , where  $\mathbf{A} \in \mathbb{R}^{n \times n}$ , has more than one solution for some  $\mathbf{b} \in \mathbb{R}^n$ , then the columns of  $\mathbf{A}$  do not span  $\mathbb{R}^n$ .
- (e) If the equation  $\mathbf{Ax} = \mathbf{b}$ , where  $\mathbf{A} \in \mathbb{R}^{n \times n}$ , is inconsistent for some  $\mathbf{b} \in \mathbb{R}^n$ , then the equation  $\mathbf{Ax} = \mathbf{0}$  has a non-trivial solution.
- (f) If  $\mathbf{A}$  is a square matrix with two identical columns then  $\mathbf{A}^{-1}$  does not exist.

4. Suppose that  $\mathbf{A} \in \mathbb{R}^{n \times n}$  is written in partitioned form as,

$$\mathbf{A} = \begin{bmatrix} \mathbf{P} & \mathbf{Q} \\ \mathbf{R} & \mathbf{S} \end{bmatrix}. \quad (2)$$

(a) Suppose that  $\mathbf{A}$  and  $\mathbf{P}$  are non-singular and prove that,

$$\mathbf{A}^{-1} = \begin{bmatrix} \mathbf{X} & -\mathbf{P}^{-1}\mathbf{Q}\mathbf{W} \\ -\mathbf{W}\mathbf{R}\mathbf{P}^{-1} & \mathbf{W} \end{bmatrix}, \quad (3)$$

where  $\mathbf{W} = (\mathbf{S} - \mathbf{R}\mathbf{P}^{-1}\mathbf{Q})^{-1}$  and  $\mathbf{X} = \mathbf{P}^{-1} + \mathbf{P}^{-1}\mathbf{Q}\mathbf{W}\mathbf{R}\mathbf{P}^{-1}$ .

(b) Suppose that  $\mathbf{A}$  and  $\mathbf{S}$  are non-singular and prove that,

$$\mathbf{A}^{-1} = \begin{bmatrix} \mathbf{X} & -\mathbf{X}\mathbf{Q}\mathbf{S}^{-1} \\ -\mathbf{S}^{-1}\mathbf{R}\mathbf{X} & \mathbf{W} \end{bmatrix}, \quad (4)$$

where  $\mathbf{X} = (\mathbf{P} - \mathbf{Q}\mathbf{S}^{-1}\mathbf{R})^{-1}$  and  $\mathbf{W} = \mathbf{S}^{-1} + \mathbf{S}^{-1}\mathbf{R}\mathbf{X}\mathbf{Q}\mathbf{S}^{-1}$ .

(c) Show that if  $\mathbf{P}, \mathbf{S}, \mathbf{A}$  are all non-singular matrices then the two previous forms are equivalent and that  $(\mathbf{S} - \mathbf{R}\mathbf{P}^{-1}\mathbf{Q})^{-1} = \mathbf{S}^{-1} + \mathbf{S}^{-1}\mathbf{R}\mathbf{X}\mathbf{Q}\mathbf{S}^{-1}$ .

(d) Finally, test these previous formula for  $\mathbf{P} = a$ ,  $\mathbf{Q} = b$ ,  $\mathbf{R} = c$ ,  $\mathbf{S} = d$  where  $a, b, c, d \in \mathbb{R}$  such that  $ad - bc \neq 0$ .

5. Determine the LU-Decomposition of the matrix  $\mathbf{A}$  and check your result for  $\mathbf{L}$  by multiplication of three elementary matrices.

$$\mathbf{A} = \begin{bmatrix} 1 & 4 & -1 & 5 \\ 3 & 7 & -2 & 9 \\ -2 & -3 & 1 & -4 \end{bmatrix}$$

**Hint:** The matrix  $\mathbf{U}$ , found by three steps of row reduction on  $\mathbf{A}$ , will have two pivot columns. These two pivot columns are used to determine the first two columns of  $\mathbf{L}_{3 \times 3}$ . The remaining column of  $\mathbf{L}$  is equal the last column of  $\mathbf{I}_3$ .



1) Given,

$$A(\theta) = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$$

$$a) A(\theta)\vec{x} = \vec{0} \Rightarrow \left[ \begin{array}{cc|c} \cos(\theta) & -\sin(\theta) & 0 \\ \sin(\theta) & \cos(\theta) & 0 \end{array} \right] \sim$$

$$\sim \left[ \begin{array}{cc|c} \cos(\theta)\sin(\theta) & -\sin^2(\theta) & 0 \\ \sin(\theta)\cos(\theta) & \cos^2(\theta) & 0 \end{array} \right] \sim$$

$$\sim \left[ \begin{array}{cc|c} \cos(\theta)\sin(\theta) & -\sin^2(\theta) & 0 \\ 0 & \cos^2(\theta) + \sin^2(\theta) & 0 \end{array} \right]$$

1

$$\Rightarrow x_2 = 0 \Rightarrow x_1 = 0 \Rightarrow \vec{x} = \vec{0} \text{ is the only Soln}$$

$\Rightarrow A$  is a 1 to 1 linear transformation.

b) By theorem 2.2.4 we have,

$$A^{-1} = \begin{bmatrix} \cos(\theta) & \sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{bmatrix} \cdot \frac{1}{\cos^2\theta + \sin^2\theta} \stackrel{\text{note}}{=} A^T$$



c) Case  $\theta = 0$ .

$$A\vec{x} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\theta = \frac{\pi}{6}$$

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

$$\theta = \frac{\pi}{3}$$

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

$$\theta = \frac{\pi}{2}$$

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

d) Case  $\theta = 0$ :

$$\frac{\vec{x} \cdot \vec{b}}{|\vec{x}| |\vec{b}|} = \frac{\begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix}}{1 \cdot 1} = 1 = \cos(\theta)$$

These transformations  
have rotated  $\vec{x}$

counter clockwise  
along the unit circle  
by an angle of  $\theta$ .



$$\text{Case } \theta = \frac{\pi}{6}$$

$$\frac{\vec{x} \cdot \vec{b}}{|\vec{x}| |\vec{b}|} = \frac{[1 \ 0] \begin{bmatrix} \sqrt{3}/2 \\ 1/2 \end{bmatrix}}{|\vec{x}| |\vec{b}|} = \frac{\sqrt{3}}{2} = \cos(\theta)$$

$$\text{Case } \theta = \frac{\pi}{3}$$

$$\frac{\vec{x} \cdot \vec{b}}{|\vec{x}| |\vec{b}|} = \frac{1}{2} = \cos(\theta)$$

$$\text{Case } \theta = \frac{\pi}{2}$$

$$\frac{\vec{x} \cdot \vec{b}}{|\vec{x}| |\vec{b}|} = 0 = \cos(\theta)$$

From this we notice,  
 $\frac{\vec{x}^T \vec{b}}{|\vec{x}| |\vec{b}|} = \cos(\theta)$ ,  $\theta$  is angle between  $\vec{x}, \vec{b}$ .

e) Consider,

$$\frac{d}{d\theta} [A A^{-1}] = A \frac{d[A^{-1}]}{d\theta} + \frac{d[A]}{d\theta} A^{-1} = \frac{d[I]}{d\theta} = 0$$

$$\Rightarrow \frac{d[A^{-1}]}{d\theta} = -A^{-1} \frac{dA}{d\theta} A^{-1}$$



$$\frac{dA^2}{d\theta} = - \begin{bmatrix} \cos(\theta) & \sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{bmatrix} \begin{bmatrix} -\sin(\theta) & -\cos(\theta) \\ +\cos(\theta) & -\sin(\theta) \end{bmatrix} \begin{bmatrix} \cos(\theta) & \sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{bmatrix} =$$

$$= - \begin{bmatrix} \cos(\theta) & \sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{bmatrix} \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} -\sin(\theta) & +\cos(\theta) \\ -\cos(\theta) & -\sin(\theta) \end{bmatrix} = dA^2$$



b) The Gauss-Jordan Method

$$\begin{aligned}
 & \left[ \begin{array}{ccc|ccc} 3 & 6 & 7 & 1 & 0 & 0 \\ 0 & 2 & 1 & 0 & 1 & 0 \\ 2 & 3 & 4 & 0 & 0 & 1 \end{array} \right] \begin{array}{l} R1 - 3R2 \\ 2R1 - 3R3 \end{array} \sim \left[ \begin{array}{ccc|ccc} 3 & 0 & 4 & 1 & -3 & 0 \\ 0 & 2 & 1 & 0 & 1 & 0 \\ 0 & 3 & 2 & 2 & 0 & -3 \end{array} \right] \begin{array}{l} \\ 2R3 - 3R2 \end{array} \\
 & \sim \left[ \begin{array}{ccc|ccc} 3 & 0 & 4 & 1 & -3 & 0 \\ 0 & 2 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 4 & -3 & -6 \end{array} \right] \begin{array}{l} R1 - 4R3 \\ R2 - R3 \end{array} \sim \left[ \begin{array}{ccc|ccc} 3 & 0 & 0 & -15 & -9 & 24 \\ 0 & 2 & 0 & -4 & 4 & 6 \\ 0 & 0 & 1 & 4 & -3 & -6 \end{array} \right] \begin{array}{l} \div 3 \\ \div 2 \end{array} \\
 & \sim \left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & -5 & 3 & 8 \\ 0 & 1 & 0 & -2 & 2 & 3 \\ 0 & 0 & 1 & 4 & -3 & -6 \end{array} \right] \quad A^{-1} = \begin{bmatrix} -5 & 3 & 8 \\ -2 & 2 & 3 \\ 4 & -3 & -6 \end{bmatrix}
 \end{aligned}$$

c) The Cofactor Representation

$$A^{-1} = \frac{1}{\det(A)} \begin{bmatrix} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \\ c_{31} & c_{32} & c_{33} \end{bmatrix} = \frac{1}{-1} \begin{bmatrix} 5 & -3 & 8 \\ 2 & -2 & -3 \\ -4 & 3 & 6 \end{bmatrix} = \begin{bmatrix} -5 & 3 & 8 \\ -2 & 2 & 3 \\ 4 & -3 & -6 \end{bmatrix}$$

d) Check your result by showing  $AA^{-1} = I$

$$AA^{-1} = \begin{bmatrix} 3 & 6 & 7 \\ 0 & 3 & 1 \\ 2 & 3 & 4 \end{bmatrix} \begin{bmatrix} -5 & 3 & 8 \\ -2 & 2 & 3 \\ 4 & -3 & -6 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = I$$

4.

- i.  $\det(A) = ad - bc$
- ii.  $\det(B) = cb - ad = -(ad - bc) = -\det(A)$
- iii.  $\det(D) = d(a + kc) - c(b + kd) = ad + kdc - cb - cdk = ad - bc = \det(A)$
- iv.  $\det(C) = adk - kcb = k(ad - bc) = k \cdot \det(A)$

$A \sim B$  by a row interchange and ii shows  $\det(A) = -\det(B)$

$A \sim C$  by a row scaling and iv shows  $\det(A) = k \cdot \det(C)$

$A \sim D$  by a row interchange where a multiple of one row is added to another.

iii shows that  $\det(A) = \det(D)$

5.

Forward Direction: Assume  $A_{3 \times 3}$  is such that  $\det(A) = 0$ . then the volume of the parallelepiped spanned by  $\vec{a}_1, \vec{a}_2, \vec{a}_3$  has zero volume. That is, the parallelepiped does not exist. This implies that all the vectors  $\vec{a}_1, \vec{a}_2, \vec{a}_3$  lie in the same plane, and form a linearly dependent set. Thus, by the invertible matrix theorem  $A^{-1}$  does not exist.

Backward Direction: Assume  $A$  is not invertible. Then the columns of  $A$  are linearly dependent and  $\text{span}\{\vec{a}_1, \vec{a}_2, \vec{a}_3\}$  is at most (in terms of dimension) a plane which has zero volume and cannot form a parallelepiped. Thus,  $\det(A) = 0$ .



Define:

$$b) A_{\text{left}}^{-1} = A_L^{-1} = (A^T A)^{-1} A^T$$

if  $A^{-1}$  exists then  $(A^T)^{-1}$  exists  $\Rightarrow (A^T A)^{-1}$  exists.

and

$$A_L^{-1} A = (A^T A)^{-1} A^T A = A^{-1} (A^T)^{-1} A^T A = A^{-1} I A = I$$

c)  $A_R^{-1} = A^T (A A^T)^{-1}$  exists by similar arguments.

$$A A_R^{-1} = A (A^T (A A^T)^{-1}) = A A^T (A A^T)^{-1} = I$$

d) Given  $A_1 = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$  then  $A_L^{-1} = \left( \begin{bmatrix} 2 & 2 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \end{bmatrix} \right)^{-1} \begin{bmatrix} 2 & 2 \end{bmatrix} = \frac{1}{8} \begin{bmatrix} 2 & 2 \end{bmatrix}$

and 
$$A_L^{-1} A_1 = \frac{1}{8} \begin{bmatrix} 2 & 2 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \end{bmatrix} = \frac{1}{8} \cdot 8 = 1$$

Also,  $A_2 = \begin{bmatrix} 2 & 2 \end{bmatrix} \Rightarrow A_R^{-1} = \begin{bmatrix} 2 \\ 2 \end{bmatrix} \left( \begin{bmatrix} 2 & 2 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \end{bmatrix} \right)^{-1} = \frac{1}{8} \begin{bmatrix} 2 \\ 2 \end{bmatrix}$

and

$$\begin{bmatrix} 2 & 2 \end{bmatrix} \cdot \frac{1}{8} \cdot \begin{bmatrix} 2 \\ 2 \end{bmatrix} = \frac{1}{8} \begin{bmatrix} 2 & 2 \end{bmatrix} \begin{bmatrix} 2 \\ 2 \end{bmatrix} = 1$$



5. a.

Proof

Since  $A$  is  $n \times n$  and  $A^{-1}$  exist we have that there exists a solution to

$$A\vec{x} = \vec{b}, \quad \vec{x} \in \mathbb{R}^n$$

for every  $\vec{b} \in \mathbb{R}^n$ , namely  $\vec{x} = A^{-1}\vec{b}$ . Thus by theorem 4 we have that the columns of  $A$  span  $\mathbb{R}^n$ .  $\square$

b.

We have that  $A$  is  $n \times n$  and

$$A\vec{x} = \vec{b}, \quad \vec{x} \in \mathbb{R}^n$$

has a solution for every  $\vec{b} \in \mathbb{R}^n$ . Thus, theorem 4 says that Every row of  $A$  must contain a pivot.

Since  $A$  is  $n \times n$  each column has a pivot. Hence the reduced row echelon form must be a diagonal matrix with 1's along the diagonal. This implies that  $A \sim I_n$  and by Theorem 7  $A$  must be invertible.  $\square$



3.

a. If the matrix  $A$  is invertible then  $A^{-1}$  is invertible by Theorem 2.26. If  $A^{-1}$  is invertible then by the invertible matrix theorem its columns are linearly independent.

b. If  $A\vec{x} = \vec{b}$  has more than 1 solution for some  $\vec{b} \in \mathbb{R}^n$ , then  $A$  does not have a pivot for each row (Presence of a free variable). Since  $A$  does not have a pivot for each row then by the IMT the columns of  $A$  do not span  $\mathbb{R}^n$ .

c. Note - this problem originally contained the typo,

" $A\vec{x} = \vec{b}$  is consistent for some  $\vec{b} \in \mathbb{R}^n$ ."

It should read inconsistent. We analyzed all cases here.

If the problem read "consistent", there would, naturally, be two cases.

Case 1:  $A\vec{x} = \vec{b}$  is consistent with nonunique solutions for some  $\vec{b} \in \mathbb{R}^n$ .

If  $A\vec{x} = \vec{b}$  is consistent with nonunique solns for some  $\vec{b} \in \mathbb{R}^n$  then there does not exist a pivot for each row of  $A$  and by the IMT the columns of  $A$  do not span  $\mathbb{R}^n$ .



Case 2:

If  $A\vec{x}=\vec{b}$  is consistent with a unique soln for some  $\vec{b} \in \mathbb{R}^n$  then  $A$  has a pivot for each row and by IMT the columns of  $A$  do span  $\mathbb{R}^n$ .

In this case the statement in 3c is false!

• If the problem read "inconsistent" there is only 1 case whose proof is identical to Case 1 above. It is worth saying that in this case the statement 3c<sub>o</sub> is always true.

d. If  $A$  is a square matrix with 2 identical columns then the columns of  $A$  are linearly dependent and by IMT  $A$  is not invertible. Hence,  $A^{-1}$  does not exist.



Let  $A = \begin{bmatrix} P & Q \\ R & S \end{bmatrix}$

a) Suppose  $A, P$  are nonsingular then  $A^{-1}$  exist s.t.

$$A^{-1} = \begin{bmatrix} X & -P^{-1}QW \\ -WRP^{-1} & W \end{bmatrix}$$

where  $W = (S - RP^{-1}Q)^{-1}$  and  $X = P^{-1} + P^{-1}QWRP^{-1}$ .

Proof:

$$[AA^{-1}]_{11} = PX - QWRP^{-1} =$$

$$= P(P^{-1} + P^{-1}QWRP^{-1}) - QWRP^{-1} = I + QWRP^{-1} - QWRP^{-1} = I$$

$$[AA^{-1}]_{12} = -PP^{-1}QW + QW = 0$$

$$[AA^{-1}]_{21} = RX - SWRP^{-1} = R(P^{-1} + P^{-1}QWRP^{-1}) - (W^{-1} + RP^{-1}Q)WRP^{-1} =$$

$$= RP^{-1} + RP^{-1}QWRP^{-1} - \underbrace{W^{-1}WRP^{-1}}_I - RP^{-1}QWRP^{-1} = 0$$

$$[AA^{-1}]_{22} = -RP^{-1}QW + SW = -RP^{-1}QW + (W^{-1} + RP^{-1}Q)W =$$

$$= -RP^{-1}QW + I + RP^{-1}QW = I$$

$$\Rightarrow AA^{-1} = \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix}, \text{ which is the } n \times n \text{ Identity matrix.}$$



b) Suppose  $A, S$  are nonsingular then  $A^{-1}$  exists s.t.

$$A^{-1} = \begin{bmatrix} x & -xQS^{-1} \\ -S^{-1}Rx & w \end{bmatrix}$$

s.t. when  $x = (P - QS^{-1}R)^{-1}$ ,  $w = S^{-1} + S^{-1}RxQS^{-1}$

$$[A^{-1}A]_{11} = xP - xQS^{-1}R =$$

$$= x(x^{-1} + QS^{-1}R) - xQS^{-1}R = I + xQS^{-1}R - xQS^{-1}R = I$$

$$[A^{-1}A]_{12} = xQ \cdot xQS^{-1} \overset{I}{\cancel{S}} = 0$$

$$[A^{-1}A]_{21} = -S^{-1}RxP + wR = -S^{-1}R(x^{-1} + QS^{-1}R) + (S^{-1} + S^{-1}RxQS^{-1})R =$$

$$= -S^{-1}R + -S^{-1}RQS^{-1}R + S^{-1}R + S^{-1}RxQS^{-1}R = 0$$

$$[A^{-1}A]_{22} = -S^{-1}RxQ + wS = -S^{-1}RxQ + (S^{-1} + S^{-1}RxQS^{-1})S =$$

$$= -S^{-1}RxQ + I + S^{-1}RxQ \overset{I}{\cancel{S}} = I$$

$$\Rightarrow [A^{-1}A] = \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix} \Rightarrow \text{which is the } n \times n \text{ Id. matrix.}$$

c) If  $P, S, A$  are nonsingular then all previous matrices are defined. Since  $A^{-1}$  is ~~exists~~ unique the two previous matrices are equal, which implies

$$-P^{-1}QW = -xQS^{-1} \quad \text{and} \quad -wRP^{-1} = -S^{-1}Rx$$

$$x = P^{-1} + P^{-1}QWRP^{-1} = (P - QS^{-1}R)^{-1}$$

$$w = (S - RP^{-1}Q)^{-1} = S^{-1} + S^{-1}RxQS^{-1}$$



d) To test the formula,

$$A = P^{-1}a, Q=b, R=c, S=d$$

$$\Rightarrow X = P^{-1} + P^{-1}QWRP^{-1} = \frac{1}{a} + \frac{1}{a} \cdot b \cdot W \cdot c \cdot \frac{1}{a}$$

$$\text{where } W = \left(d - \frac{c}{a} \cdot b\right)^{-1} = \frac{a}{ad-bc}$$

$$\Rightarrow X = \frac{1}{a} + \frac{1}{a} b \cdot \frac{a}{ad-bc} \cdot c \cdot \frac{1}{a} =$$

$$= \frac{1}{a} + \frac{bc}{a(ad-bc)} = \frac{a(ad-bc) + abc}{a(ad-bc)} = \frac{d}{ad-bc}$$

$$\text{also } = \frac{1}{a} \left(1 + \frac{bc}{ad-bc}\right) = \frac{1}{a} \left(\frac{ad-bc+bc}{ad-bc}\right)$$

$$-P^{-1}QW = -\frac{1}{a} \cdot b \cdot \frac{a}{ad-bc} = \frac{-b}{ad-bc}$$

and

$$-WRP^{-1} = \frac{-a}{ad-bc} \cdot c \cdot \frac{1}{a} = \frac{-c}{ad-bc}$$

$$\Rightarrow A^{-1} = \begin{bmatrix} X & -P^{-1}QW \\ -WRP^{-1} & W \end{bmatrix} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

which is the inverse of

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$



$$2. \quad A = \begin{bmatrix} 1 & 4 & -1 & 5 \\ 3 & 7 & -2 & 9 \\ -2 & -3 & 1 & -4 \end{bmatrix} \quad \begin{array}{l} R_2 = R_2 - 3R_1 \\ \sim \\ R_3 = R_3 + 2R_1 \end{array} \quad \begin{bmatrix} 1 & 4 & -1 & 5 \\ 0 & -5 & 1 & -6 \\ 0 & 5 & -1 & 6 \end{bmatrix} \quad \begin{array}{l} R_3 = R_3 + R_2 \\ \sim \end{array}$$

$$\sim \begin{bmatrix} 1 & 4 & -1 & 5 \\ 0 & -5 & 1 & -6 \\ 0 & 0 & 0 & 0 \end{bmatrix} = U, \quad L = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ -2 & 3 & 1 \end{bmatrix} \quad \begin{array}{l} \text{(By quick} \\ \text{Method)} \end{array}$$

$$E_1^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ +3 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad E_2^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix}, \quad E_3^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix}$$

$$E_1^{-1} E_2^{-1} E_3^{-1} \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ -2 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ -2 & -1 & 1 \end{bmatrix} = L$$

$$LU = \begin{bmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ -2 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 4 & -1 & 5 \\ 0 & -5 & 1 & -6 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 4 & -1 & 5 \\ 3 & 7 & -2 & 9 \\ -2 & -3 & 1 & -4 \end{bmatrix} = A.$$