# Assignment 14

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Abstract—This document explains the proof such that if A is an  $m \times n$  matrix and B is an  $n \times m$  matrix and n < m, then AB is not invertible

Download all latex-tikz codes from

https://github.com/venkateshelangovan/IIT— Hyderabad—Assignments/tree/master/ Assignment14 Matrix Theory

#### 1 Problem

Prove that if **A** is an  $m \times n$  matrix, **B** is an  $n \times m$  matrix and n < m, then **AB** is not invertible

#### 2 RANK OF A MATRIX

## 2.1 Definition

The rank of a matrix is defined as

- 1. The maximum number of linearly independent column vectors in the matrix or
- 2. The maximum number of linearly independent row vectors in the matrix.

## 2.2 To prove Row Rank=Column Rank

Consider the matrix **A** of order  $m \times n$ ,

The row rank is the maximum number of linearly independent rows in the matrix A

$$RowRank(\mathbf{A}) \le m$$
 (2.2.1)

The column rank is the maximum number of linearly independent column in the matrix  $\bf A$ 

$$ColumnRank(\mathbf{A}) \le n$$
 (2.2.2)

A matrix  $P_n$  is a permutation matrix of order  $n \times n$  if and only if it is obtained from  $n \times n$  Identity matrix  $I_n$  by performing one or more interchanges of the rows and columns of  $I_n$ .

One of the  $3 \times 3$  permutation matrix is given by,

$$\mathbf{P_3} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \tag{2.2.3}$$

Let  $P_n$  be an  $n \times n$  permuation matrix,

$$\mathbf{AP_n} = \begin{pmatrix} \mathbf{X} & \mathbf{W} \end{pmatrix} \tag{2.2.4}$$

1

where columns of X are the d pivot columns of A.

Every column of W is a linear combination of the columns of X, so there is a matrix K such that,

$$\mathbf{W} = \mathbf{X}\mathbf{K} \tag{2.2.5}$$

where the columns of K contain the coefficients of each of those linear combinations.

Substituting the equation (2.2.5) in equation (2.2.4), we get

$$\mathbf{AP_n} = \begin{pmatrix} \mathbf{X} & \mathbf{XK} \end{pmatrix} \tag{2.2.6}$$

$$\implies \mathbf{AP_n} = \mathbf{X} \begin{pmatrix} \mathbf{I_d} & \mathbf{K} \end{pmatrix} \tag{2.2.7}$$

where  $I_d$  represents the  $d \times d$  identity matrix

Transforming the matrix A into reduced row echelon form, we get ,

$$\mathbf{B} = \mathbf{E}\mathbf{A} \tag{2.2.8}$$

where **E** is the product of elementary matrices and **B** is the reduced row echelon form of **A** 

Multiplying  $\mathbf{E}$  to the equation (2.2.7) on both sides,

$$EAP_{n} = EX(I_{d} K)$$
 (2.2.9)

$$\implies \mathbf{BP_n} = \mathbf{EX} \left( \mathbf{I_d} \quad \mathbf{K} \right) \tag{2.2.10}$$

Where,

$$\mathbf{EX} = \begin{pmatrix} \mathbf{I_d} \\ 0 \end{pmatrix} \tag{2.2.11}$$

Substituting the equation (2.2.11) in equation (2.2.10),

$$\mathbf{BP_n} = \begin{pmatrix} \mathbf{I_d} & \mathbf{K} \\ 0 & 0 \end{pmatrix} \tag{2.2.12}$$

Here ,we could see that the nonzero d rows of the reduced row echelon form with the same permutation on the columns as we did for **A**. Therefore we

could say that,

$$(\mathbf{I_d} \quad \mathbf{K}) = \mathbf{YP_n}$$
 (2.2.13)

Substituting the equation (2.2.13) in equation (2.2.7), we get,

$$\mathbf{AP_n} = \mathbf{XYP_n} \tag{2.2.14}$$

Here  $P_n$  is a permutation matrix and it is invertible. This implies,

$$\mathbf{A} = \mathbf{XY} \tag{2.2.15}$$

where matrix **X** is of order  $m \times d$  and matrix **Y** is of order  $d \times n$ 

Consider the example,

$$\mathbf{A} = \begin{pmatrix} 4 & 8 & 9 \\ 2 & 4 & 7 \end{pmatrix} \tag{2.2.16}$$

$$\mathbf{AP_3} = \begin{pmatrix} 4 & 9 & 8 \\ 2 & 7 & 4 \end{pmatrix} \tag{2.2.17}$$

Let,

$$\mathbf{X} = \begin{pmatrix} 4 & 9 \\ 2 & 7 \end{pmatrix} \tag{2.2.18}$$

$$\mathbf{W} = \mathbf{X}\mathbf{K} = \begin{pmatrix} 8\\4 \end{pmatrix} \tag{2.2.19}$$

$$\mathbf{I_2} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \tag{2.2.20}$$

$$\mathbf{AP_3} = \begin{pmatrix} \mathbf{X} & \mathbf{W} \end{pmatrix} \tag{2.2.21}$$

$$\mathbf{AP_3} = \begin{pmatrix} \mathbf{X} & \mathbf{XK} \end{pmatrix} \tag{2.2.22}$$

$$\mathbf{AP_3} = \mathbf{X} \begin{pmatrix} \mathbf{I_2} & \mathbf{K} \end{pmatrix} \tag{2.2.23}$$

From equation (2.2.19),

$$\mathbf{K} = \mathbf{X}^{-1}\mathbf{W} \tag{2.2.24}$$

$$\mathbf{K} = \begin{pmatrix} 2 \\ 0 \end{pmatrix} \tag{2.2.25}$$

Let **B** be reduced row echelon form of matrix **A**,

$$\mathbf{B} = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{2.2.26}$$

$$\mathbf{BP_n} = \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 0 \end{pmatrix} \tag{2.2.27}$$

Let,

$$\begin{pmatrix} \mathbf{I_2} & \mathbf{K} \end{pmatrix} = \mathbf{Y} \mathbf{P_n} \tag{2.2.28}$$

$$\begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 0 \end{pmatrix} = \mathbf{Y} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \tag{2.2.29}$$

$$\implies \mathbf{Y} = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \end{pmatrix} \tag{2.2.30}$$

From the above equations (2.2.16),(2.2.18) and (2.2.30), it can be seen that,

$$A = XY$$

Now for the matrix **A** of order  $m \times n$ ,

Every row of A is the linear combination of the rows of Y

$$rowspace(\mathbf{A}) \subseteq span(\{\mathbf{Y}_1, \mathbf{Y}_2, \dots, \mathbf{Y}_d\})$$
 (2.2.31)

where  $Y_1, Y_2, \dots, Y_d$  were the rows of Y

The dimension of the row space is atmost d

Every column of A is the linear combination of the columns of X

$$columnspace(A) \subsetneq span(\{X_1, X_2, \dots, X_d\})$$
 (2.2.32)

where  $X_1, X_2, \dots, X_d$  were the columns of  $\boldsymbol{X}$ 

The dimension of the column space is atmost d.

Let **A** is of the order  $m \times n$ . If the dimension of the row space of **A** is r(Row Rank),then  $\mathbf{A} = \mathbf{X}\mathbf{Y}$  for some  $m \times r$  matrix  $\mathbf{X}$  and  $r \times m$  matrix  $\mathbf{Y}$ 

Let  $Y_1, Y_2, \ldots, Y_r$  be a basis for the row space of A and let  $A_1, A_2, \ldots, A_m$  be the rows of A. Then,

$$\mathbf{A_{1}} = x_{11}\mathbf{Y_{1}} + x_{12}\mathbf{Y_{2}} + \dots + x_{1r}\mathbf{Y_{r}}$$

$$\mathbf{A_{2}} = x_{21}\mathbf{Y_{1}} + x_{22}\mathbf{Y_{2}} + \dots + x_{2r}\mathbf{Y_{r}}$$

$$\vdots$$

$$\mathbf{A_{m}} = x_{m1}\mathbf{Y_{1}} + x_{m2}\mathbf{Y_{2}} + \dots + x_{mr}\mathbf{Y_{r}}$$

Thus A = XY where Y is the matrix with rows  $Y_1, ..., Y_r$  and X is the matrix of coefficients  $X(i, j) = x_{ij}$ .

$$ColumnRank \le RowRank$$
 (2.2.33)

Let  $X_1, X_2, \dots, X_r$  be a basis for the column space of

A and let  $A_1, A_2, \ldots, A_n$  be the columns of A.Then,

$$\mathbf{A_1} = \mathbf{X_1}y_{11} + \mathbf{X_2}y_{21} + \dots + \mathbf{X_r}y_{r1}$$
  
 $\mathbf{A_2} = \mathbf{X_1}y_{12} + \mathbf{X_2}y_{22} + \dots + \mathbf{X_r}y_{r2}$ 

:

$$\mathbf{A_n} = \mathbf{X_1} y_{1n} + \mathbf{X_2} y_{2n} + \dots + \mathbf{X_r} y_{rn}$$

Thus A = XY where X is the matrix with columns  $X_1, \ldots, X_r$  and Y is the matrix of coefficients  $Y(i, j) = y_{ij}$ .

$$RowRank \le ColumnRank$$
 (2.2.34)

From equations (2.2.33) and (2.2.34), we get,

$$RowRank = ColumnRank$$
 (2.2.35)

From the rank definition and the above equations (2.2.1),(2.2.2) and (2.2.35),

$$Rank(\mathbf{A}) \le min(m, n)$$
 (2.2.36)

# 2.3 Properties

For a matrix **A** of order  $m \times n$ ,

- (a) If m is less than n, then the rank of the matrix will be atmost m.
- (b) If m is greater than n, then the rank of the matrix will be atmost n.

#### 3 Proof

Given, Matrix **A** is of order  $m \times n$  and Matrix **B** is of order  $n \times m$ 

$$n < m \tag{3.0.1}$$

From equation (2.2.36), since given n < m,

$$Rank(\mathbf{A}) \le n$$
 (3.0.2)

$$Rank(\mathbf{B}) \le n \tag{3.0.3}$$

 $\mathbf{A}^T$  will be of order  $n \times m$ 

From equation (2.2.36),

$$Rank(\mathbf{A}^T) \le min(n, m)$$
 (3.0.4)

$$\implies Rank(\mathbf{A}^T) = Rank(\mathbf{A})$$
 (3.0.5)

The maximum possible rank of A and B is given by

$$Rank(\mathbf{A}) = n \tag{3.0.6}$$

$$Rank(\mathbf{B}) = n \tag{3.0.7}$$

**AB** will be of order  $m \times m$ 

Consider a vector v.

$$\mathbf{v} \in col((\mathbf{AB})) \tag{3.0.8}$$

$$\mathbf{v} = (\mathbf{A}\mathbf{B})\mathbf{x} \tag{3.0.9}$$

$$\mathbf{v} = \mathbf{A}(\mathbf{B}\mathbf{x}) \tag{3.0.10}$$

$$\mathbf{v} \in col((\mathbf{A})) \tag{3.0.11}$$

From equations (3.0.8) and (3.0.11), we could say that for every vector in the column space of **AB** the same vector will be there in column space of **A** aswell.

$$Rank(\mathbf{AB}) \le Rank(\mathbf{A})$$
 (3.0.12)

From equation (3.0.5),

$$Rank(\mathbf{AB}) = Rank((\mathbf{AB})^T)$$
 (3.0.13)

$$\implies Rank((\mathbf{AB})^T) = Rank(\mathbf{B}^T \mathbf{A}^T)$$
 (3.0.14)

From equation (3.0.12),

$$\implies Rank(\mathbf{B}^T \mathbf{A}^T) \le Rank(\mathbf{B}^T)$$
 (3.0.15)

$$\implies Rank(\mathbf{B}^T \mathbf{A}^T) \le Rank(\mathbf{B})$$
 (3.0.16)

$$\implies Rank(\mathbf{AB}) \le Rank(\mathbf{B})$$
 (3.0.17)

From the equations (3.0.12) and (3.0.17),

$$Rank(\mathbf{AB}) \le min(Rank(\mathbf{A}), Rank(\mathbf{B}))$$
 (3.0.18)

$$Rank(\mathbf{AB}) \le n \tag{3.0.19}$$

The maximum possible rank of **AB** of order  $m \times m$  is given by

$$Rank(\mathbf{AB}) = n < m \tag{3.0.20}$$

From (3.0.20) we could say that **AB** does not have a full rank and if the matrix does not have a full rank then it is not invertible. Hence, **AB** is not invertible.

Hence Proved