MA1522 Linear Algebra for Computing Lecture 8: Row/Column Spaces, Rank and Nullity

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Outline

A Brief Revision of Section 4.1

A Brief Revision of Section 4.2

Selected Revision of Section 4.1

- Section 4.1 talks about Row and Column Spaces and Null Spaces.
- We take a brief revision, with emphasize on how to find basis in each of the spaces.
- ▶ Along the way, we will answer the questions and challenges posed by Dr. Teo.

Slide 3: Definition of Column and Row Space

Let **A** be an $m \times n$ matrix,

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

The *row space* of **A**, is the subspace of \mathbb{R}^n spanned by the rows of **A**,

$$\mathsf{Row}(\mathbf{A}) = \mathsf{span}\{\left(a_{11} \quad a_{12} \quad \cdots \quad a_{1n}\right), \left(a_{21} \quad a_{22} \quad \cdots \quad a_{2n}\right), ..., \left(a_{m1} \quad a_{m2} \quad \cdots \quad a_{mn}\right)\}.$$

The *column space* of **A**, is the subspace of \mathbb{R}^m spanned by the columns of **A**,

$$\mathsf{Col}(\mathbf{A}) = \mathsf{span} \left\{ \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix}, \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix}, ..., \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{pmatrix} \right\}.$$

Slide 4: Finding A Basis for Row Space

Algorithm of Finding a basis for $Row(\mathbf{A})$.

- 1. Reduce **A** into its reduced row echelon form **R**.
- 2. The nonzero rows of \mathbf{R} form a basis of $Row(\mathbf{A})$.

Remarks:

- ▶ It is based on the important **Fact** (see Slide 9): Row operations preserve row space.
- ▶ In fact, it suffices to reduce to the row echelon form and take the nonzero rows.
- Note: The basis are from the rows of **R**, not from **A**. We will come back to this point later.

Examples from Slide 11

1.
$$\mathbf{A} = \begin{pmatrix} 1 & 0 & 2 & 0 \\ 0 & 1 & 0 & 2 \\ 1 & 1 & 2 & 2 \end{pmatrix} \xrightarrow{RREF} \begin{pmatrix} 1 & 0 & 2 & 0 \\ 0 & 1 & 0 & 2 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$
So $\left\{ \begin{pmatrix} 1 & 0 & 2 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 & 0 & 2 \end{pmatrix} \right\}$ is a basis for Row(\mathbf{A}).

2.
$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 1 \\ 3 & 4 & 7 \\ -1 & 7 & -19 \\ 1 & 9 & -13 \end{pmatrix} \xrightarrow{RREF} \begin{pmatrix} 1 & 0 & 5 \\ 0 & 1 & -3 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}.$$

So $\{\begin{pmatrix}1 & 0 & 5\end{pmatrix}, \begin{pmatrix}0 & 1 & -3\end{pmatrix}\}$ is a basis for Row(**A**).

Challenge one from Section 4.1

Example continued:

3.
$$\mathbf{A} = \begin{pmatrix} 1 & 1 & 2 & -1 \\ 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1 \end{pmatrix} \xrightarrow{RREF} \begin{pmatrix} 1 & 0 & 0 & 2 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 1 \end{pmatrix}$$

So $\left\{\begin{pmatrix}1&0&0&2\end{pmatrix},\begin{pmatrix}0&1&0&-1\end{pmatrix},\begin{pmatrix}0&0&1&1\end{pmatrix}\right\}$ is a basis for Row(\mathbf{A}).

Challenge: However, in this case, we could have taken the original rows of ${\bf A}$

$$\left\{ \begin{pmatrix} 1 & 1 & 2 & -1 \end{pmatrix}, \begin{pmatrix} 0 & 1 & 1 & 0 \end{pmatrix}, \begin{pmatrix} 1 & 1 & 0 & 1 \end{pmatrix} \right\}$$

as a basis for Row(A) too. Why?

Answer to Challenge one from Section 4.1

$$\mathbf{A} = \begin{pmatrix} 1 & 1 & 2 & -1 \\ 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1 \end{pmatrix} \xrightarrow{RREF} \begin{pmatrix} 1 & 0 & 0 & 2 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 1 \end{pmatrix}$$

Q: We could have taken the original rows of A

$$\left\{\mathbf{u}_1 = \begin{pmatrix} 1 & 1 & 2 & -1 \end{pmatrix}, \mathbf{u}_2 = \begin{pmatrix} 0 & 1 & 1 & 0 \end{pmatrix}, \mathbf{u}_3 = \begin{pmatrix} 1 & 1 & 0 & 1 \end{pmatrix}\right\}$$

as a basis for $Row(\mathbf{A})$ too. Why?

Answer: From the RREF, we know that $\dim(\text{Row}(\mathbf{A})) = 3$. Clearly $\text{span}(\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3) = \text{Row}(\mathbf{A})$. By Slide 138 of Chap 3, we know that they are a basis of $\text{Row}(\mathbf{A})$.

Slide 16: Finding A Basis for Column Space

Algorithm of Finding a basis for $Col(\mathbf{A})$.

- 1. Reduce **A** into its reduced row echelon form **R**.
- The columns of A corresponding to the pivot columns in R form a basis for Col(A).

Remarks:

- ▶ It is based on the important **Fact** (see Slide 15): Row operations preserve linear relations between columns.
- ▶ In fact, it suffices to reduce to the row echelon form and do step 2.
- ▶ Back to row space of A, if we apply the above algorithm for A^T, we will get a basis for Row(A) consisting only row vectors of A.

Example (on Slide 17)

Q:

Which columns of **A** form a basis for Col(**A**)?

Answer: Let \mathbf{v}_i ($i=1,\ldots,5$) denote the column vectors of \mathbf{A} . Since the first and third columns of \mathbf{R} are the pivot columns, $\{\mathbf{v}_1,\mathbf{v}_3\}$ forms a basis for $\operatorname{Col}(\mathbf{A})$.

Illustrating Preservation of Linear Relations

We also use this example to illustrate "Row operations preserve linear relations between columns".

$$\begin{pmatrix} \frac{1}{2} \\ 0 \\ 0 \\ 0 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}, \quad \begin{pmatrix} \frac{5}{6} \\ -\frac{1}{6} \\ 0 \\ 0 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} 5 \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} - \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \end{pmatrix} \end{pmatrix}, \quad \begin{pmatrix} \frac{1}{3} \\ \frac{1}{3} \\ 0 \\ 0 \end{pmatrix} = \frac{1}{3} \begin{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \end{pmatrix} \end{pmatrix}$$

$$\begin{pmatrix} 1 \\ 2 \\ 1 \\ 3 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 2 \\ 4 \\ 2 \\ 6 \end{pmatrix}, \quad \begin{pmatrix} 1 \\ 3 \\ 2 \\ 4 \end{pmatrix} = \frac{1}{6} \left(5 \begin{pmatrix} 2 \\ 4 \\ 2 \\ 6 \end{pmatrix} - \begin{pmatrix} 4 \\ 2 \\ -2 \\ 6 \end{pmatrix} \right), \quad \begin{pmatrix} 2 \\ 2 \\ 0 \\ 4 \end{pmatrix} = \frac{1}{3} \left(\begin{pmatrix} 2 \\ 4 \\ 2 \\ -2 \\ 6 \end{pmatrix} + \begin{pmatrix} 4 \\ 2 \\ -2 \\ 6 \end{pmatrix} \right)$$

Challenge Two from Section 4.1

Q: We could take any 2 columns of **A** except columns 1 and 2, to be a basis for Col(A). Why?

Answer: We know that $\dim(\operatorname{Col}(\mathbf{A})) = 2$. It suffices to check that any such 2 columns are linearly independent. For example, take **v**₂ and \mathbf{v}_4 .

$$\begin{pmatrix} 1 & 1 \\ 2 & 3 \\ 1 & 2 \\ 3 & 4 \end{pmatrix} \xrightarrow{\text{by above}} \mathbf{R} = \begin{pmatrix} 1/2 & 5/6 \\ 0 & -1/6 \\ 0 & 0 \\ 0 & 0 \end{pmatrix} \xrightarrow{\text{rref}} \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \\ 0 & 0 \end{pmatrix},$$

which shows that \mathbf{v}_2 and \mathbf{v}_4 are linearly independent. Other pairs are similar.



Question One from Section 4.1

Which of the following statements is/are true?

- 1. Suppose **A** is a 3×3 matrix whose reduced row-echelon form is $\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$. Then the set $\left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \right\}$ is a basis for the column space of A.
- 2. Suppose ${\bf A}$ is a 4×3 matrix whose reduced row-echelon form is $\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$. Then we can conclude that the first 2 rows

of **A** are linearly independent.

These two questions are related to (or trying to extend) the facts:

- Row operations preserve row space.

Remarks (Slide 20)

1. Row operations do not preserve column space. Consider

$$\mathbf{A} = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \xrightarrow{R_1 \leftrightarrow R_2} \mathbf{B} = \begin{pmatrix} 0 & 0 \\ 1 & 1 \end{pmatrix}.$$

$$\mathsf{Col}(\mathbf{A}) = \mathsf{span} \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right\} \neq \mathsf{span} \left\{ \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right\}.$$

2. Row operations do no preserve linear relations between the rows. Consider

$$\mathbf{A} = \begin{pmatrix} 1 & 1 \\ 2 & 2 \end{pmatrix} \xrightarrow{R_2 - 2R_1} \mathbf{B} = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix}.$$

row 2 of $\mathbf{A} = 2 \times \text{ row } 1$ of \mathbf{A} , row 2 of $\mathbf{B} = 0 \times \text{ row } 1$ of \mathbf{B} .

Answer to Question One in Section 4.1, part 1

Q: Suppose $\bf A$ is a 3×3 matrix whose reduced row-echelon form is $\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$. Then the set $\left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \right\}$ is a basis for the column space of $\bf A$.

Note: Here, unlike the algorithm, the columns are picked from the RREF, instead of $\bf A$.

Answer: False. For example, let

$$\mathbf{A} = \begin{pmatrix} 0 & 0 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \xrightarrow{rref} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}.$$

But $(1,0,0)^T$ is not even in the column space of **A**.

Answer to Question One in Section 4.1, part 2

Q: Suppose \boldsymbol{A} is a 4×3 matrix whose reduced row-echelon form is

$$\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}.$$
 Then we can conclude that the first 2 rows of **A** are

linearly independent.

Note: Here, clearly the first two rows of the RREF being linearly independent, does it also hold for rows of **A**?

Answer: False. For example, let

$$\mathbf{A} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & -1 \end{pmatrix} \xrightarrow{rref} \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}.$$

The first two rows of **A** is clearly linearly dependent.



Slide 23: Nullspace

Definition

The <u>nullspace</u> of an $m \times n$ matrix **A** is the solution space to the homogeneous system $\mathbf{A}\mathbf{x} = \mathbf{0}$ with coefficient matrix **A**. It is denoted as

$$\mathsf{Null}(\mathbf{A}) = \{ \ \mathbf{v} \in \mathbb{R}^n \mid \mathbf{A}\mathbf{v} = \mathbf{0} \ \}.$$

The *nullity* of $\bf A$ is the dimension of the nullspace of $\bf A$, denoted as

$$\operatorname{nullity}(\mathbf{A}) = \dim(\operatorname{Null}(\mathbf{A})).$$

Algorithms for finding a basis of nullspace: Same as finding a basis for solution space.

Question Two in Section 4.1

Let

- 1. Find a basis for the nullspace of A.
- 2. What is the nullity of A?

From the RREF, we immediately get the general solution:

$$x_1 = -\frac{1}{2}r - \frac{5}{6}s - \frac{1}{3}t, x_2 = r, x_3 = \frac{1}{6}s - \frac{1}{3}t, x_4 = s, x_5 = t,$$

where $r, s, t \in \mathbb{R}$.

Answer to Question Two in Section 4.1

(continue from previous slide) Answer: The general solution of $\mathbf{A}\mathbf{x}=\mathbf{0}$ is

$$\mathbf{x} = r egin{pmatrix} -rac{1}{2} \ 1 \ 0 \ 0 \ 0 \ \end{pmatrix} + s egin{pmatrix} -rac{5}{6} \ 0 \ rac{1}{6} \ 1 \ 0 \end{pmatrix} + t egin{pmatrix} -rac{1}{3} \ 0 \ -rac{1}{3} \ 0 \ 1 \end{pmatrix}.$$

The vectors above form a basis and the Nullity of A is 3.

Slide 27: Rank

Let **A** be an $m \times n$ matrix.

Definition

Define the \underline{rank} of **A** to be the dimension of its column or row space

$$rank(\mathbf{A}) = dim(Col(\mathbf{A})) = dim(Row(\mathbf{A})).$$

Justification: Let **R** be the reduced row-echelon form of **A**.

Challenge One of Section 4.2

Prove the following theorem.

Theorem

The linear system $\mathbf{A}\mathbf{x} = \mathbf{b}$ is consistent if and only if the rank of \mathbf{A} is equal to the rank of the augmented matrix $(\mathbf{A} \mid \mathbf{b})$,

$$rank(\mathbf{A}) = rank((\mathbf{A} \mid \mathbf{b})).$$

Preparation: Suppose that **A** is an $m \times n$ matrix and $\mathbf{v}_1, \dots, \mathbf{v}_n$ be the column vectors of **A**. Let $\mathbf{x} = (c_1, \dots, c_n)^T$. Then the linear system $\mathbf{A}\mathbf{x} = \mathbf{b}$ can be expressed as:

$$c_1\mathbf{v}_1+\cdots+c_n\mathbf{v}_n=\begin{pmatrix}\mathbf{v}_1&\dots&\mathbf{v}_n\end{pmatrix}\begin{pmatrix}c_1\\\vdots\\c_n\end{pmatrix}=\mathbf{b}.$$

Answer to Challenge One Section 4.2, the "only if" direction

Q: The linear system $\mathbf{A}\mathbf{x} = \mathbf{b}$ is consistent if and only if

$$rank(\mathbf{A}) = rank((\mathbf{A} \mid \mathbf{b})).$$

Answer: (\Rightarrow) Suppose that the linear system $\mathbf{A}\mathbf{x} = \mathbf{b}$ is consistent. By previous slide, there are c_1, \ldots, c_n such that

$$c_1\mathbf{v}_1+\cdots+c_n\mathbf{v}_n=\mathbf{b},$$

that is, $\mathbf{b} \in Col(\mathbf{A})$. Thus,

$$rank(\mathbf{A}) = dim(Col(\mathbf{A})) = dim(Col(\mathbf{A} \mid \mathbf{b})) = rank((\mathbf{A} \mid \mathbf{b})).$$

Answer to Challenge One Section 4.2, the "if" direction

Q: The linear system $\mathbf{A}\mathbf{x} = \mathbf{b}$ is consistent if and only if

$$rank(\mathbf{A}) = rank((\mathbf{A} \mid \mathbf{b})).$$

Answer: (\Leftarrow) Suppose that $rank(\mathbf{A}) = rank((\mathbf{A} \mid \mathbf{b}))$, that is,

$$\dim(\mathsf{Col}(\mathbf{A})) = \dim(\mathsf{Col}(\mathbf{A} \mid \mathbf{b})).$$

Let W be a set of basis of $Col(\mathbf{A})$, then W is linearly independent and also $W \subseteq Col(\mathbf{A} \mid \mathbf{b})$. Since $|W| = \dim(Col(\mathbf{A} \mid \mathbf{b}))$, we have $span(W) = Col(\mathbf{A} \mid \mathbf{b})$. Therefore, $\mathbf{b} \in span(W)$, hence also in the span of column vectors $\mathbf{v}_1, \ldots, \mathbf{v}_n$ of \mathbf{A} , i.e., there are c_1, \ldots, c_n , such that,

$$c_1\mathbf{v}_1+\cdots+c_n\mathbf{v}_n=\mathbf{b},$$

that is, $(c_1, \ldots, c_n)^T$ is a solution of $\mathbf{A}\mathbf{x} = \mathbf{b}$.

Question in Section 4.2

Show that if **A** and **B** are row equivalent matrices, then $rank(\mathbf{A}) = rank(\mathbf{B})$.

Answer: Since elementary row operations does not change the row space, we have $Row(\mathbf{A}) = Row(\mathbf{B})$. Hence,

$$rank(\mathbf{A}) = dim(Row(\mathbf{A})) = dim(Row(\mathbf{B})) = rank(\mathbf{B}).$$

Challenge Two in Section 4.2

Let **A** and **B** be matrices of the same size. Prove that

$$rank(\mathbf{A} + \mathbf{B}) \le rank(\mathbf{A}) + rank(\mathbf{B}).$$

Answer: Let $\mathbf{A} = (\mathbf{u}_1, \dots, \mathbf{u}_n)$ where \mathbf{u}_i are the column vectors of \mathbf{A} , and $\mathbf{B} = (\mathbf{v}_1, \dots, \mathbf{v}_n)$ where \mathbf{v}_j are the column vectors of \mathbf{B} . Form the matrix $\mathbf{C} = (\mathbf{u}_1, \dots, \mathbf{u}_n, \mathbf{v}_1, \dots, \mathbf{v}_n)$. Then we have

- ▶ $Col(\mathbf{A} + \mathbf{B}) \subseteq Col(\mathbf{C})$, because every column vector in $\mathbf{A} + \mathbf{B}$ is $\mathbf{u}_i + \mathbf{v}_i$ for some i. thus $rank(\mathbf{A} + \mathbf{B}) \le rank(\mathbf{C})$.
- ▶ On the other hand, let $X \subseteq \{\mathbf{u}_1, \dots, \mathbf{u}_n\}$ and $Y \subseteq \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ be bases of $Col(\mathbf{A})$ and $Col(\mathbf{B})$ respectively. We have $rank(\mathbf{A}) = |X|$ and $rank(\mathbf{B}) = |Y|$. Since $Col(\mathbf{C}) \subseteq span(X \cup Y)$,

$$rank(\mathbf{C}) = dim(Col(\mathbf{C})) \le dim(span(X \cup Y))$$

 $\le |X| + |Y| = rank(\mathbf{A}) + rank(\mathbf{B}).$



Challenge Three in Section 4.2

Let **A** be a $m \times n$ matrix such that rank(**A**) = m. Suppose m > n. By the equivalent statements of full rank equals number of columns, (**A**^T**A**) invertible and (**A**^T**A**)⁻¹**A**^T is a left inverse of **A**.

Now consider the system $\mathbf{A}\mathbf{x} = \mathbf{b}$ for some vector \mathbf{b} in \mathbb{R}^m . Premultiplying the left inverse above on both sides of the equation, we get

$$\mathbf{x} = ((\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T) \mathbf{A} \mathbf{x} = ((\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T) \mathbf{b},$$

that is, $((\mathbf{A}^T\mathbf{A})^{-1}\mathbf{A}^T)\mathbf{b}$ is a solution to $\mathbf{A}\mathbf{x} = \mathbf{b}$. But this is true for every \mathbf{b} , which by the equivalent statements of full rank equals number of rows, means that the rank of \mathbf{A} is equal to m, the number of row. This is a contradiction to m > n.

What is the mistake in the argument above?

Side Remark: Matrices of the form $\mathbf{A}^T \mathbf{A}$ will appear in sections on Least Square Problems.



Slide 46: Full Rank Equals Number of Columns

Theorem

Suppose **A** is an $m \times n$ matrix. The following statements are equivalent.

- (i) **A** is full rank, where the rank is equal to the number of columns, $rank(\mathbf{A}) = n$.
- (ii) The rows of **A** spans \mathbb{R}^n , Row(**A**) = \mathbb{R}^n .
- (iii) The columns of **A** are linearly independent.
- (iv) The homogeneous system $\mathbf{A}\mathbf{x} = \mathbf{0}$ has only the trivial solution, that is, $\text{Null}(\mathbf{A}) = \{\mathbf{0}\}.$
- (v) $\mathbf{A}^T \mathbf{A}$ is an invertible matrix of order n.
- (vi) A has a left inverse.

Note: In this case, $m \ge n$.

Slide 49: Full Rank Equals Number of Rows

Theorem

Suppose **A** is an $m \times n$ matrix. The following statements are equivalent.

- (i) **A** is full rank, where the rank is equal to the number of rows, $rank(\mathbf{A}) = m$.
- (ii) The columns of **A** spans \mathbb{R}^m , $Col(\mathbf{A}) = \mathbb{R}^m$.
- (iii) The rows of **A** are linearly independent.
- (iv) The linear system $\mathbf{A}\mathbf{x} = \mathbf{b}$ is consistent for every $\mathbf{b} \in \mathbb{R}^m$.
- (v) $\mathbf{A}\mathbf{A}^T$ is an invertible matrix of order m.
- (vi) A has a right inverse.

Note: In this case, $m \leq n$.



Answer to the Challenge Three in Section 4.2

Q: After apply Full Rank Theorems (both the column form and the row form), some contradiction occurred. What is the mistake?

Answer: When m > n, one can only use the Theorem about full rank equals number of columns, the theorem for rows would not apply.