

LECTURE 10

The Rank of a Matrix

Let \mathbf{A} be an $m \times n$ matrix.

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

Recall that the **column space** of \mathbf{A} is the subspace of \mathbb{R}^m spanned by the columns of \mathbf{A} :

$$ColSp(\mathbf{A}) = span(\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_n) \subset \mathbb{R}^m$$

where i^{th} column vector \mathbf{c}_i is defined by

$$(\mathbf{c}_i)_j \equiv a_{ji} \quad , \quad j = 1, \dots, m$$

Recall also that the **row space** of \mathbf{A} is the subspace of \mathbb{R}^n spanned by the rows of \mathbf{A} :

$$RowSp(\mathbf{A}) = span(\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_m) \subset \mathbb{R}^n$$

where the i^{th} row vector is defined by

$$(\mathbf{r}_i)_j = a_{ij} \quad , \quad j = 1, \dots, n$$

A priori there is no particular relationship between the column space of \mathbf{A} and the row space of \mathbf{A} ; indeed, they are not even subspaces of the same space.

LEMMA 10.1. *If a matrix \mathbf{A}' is row equivalent to a matrix \mathbf{A} then the row space of \mathbf{A}' is equal to the row space of \mathbf{A} .*

Proof. First we note that row operations can be built up from row operations of the following form

$$(10.1) \quad R_{ij}(\lambda_1, \lambda_2) : \begin{cases} \mathbf{r}_i \rightarrow \mathbf{r}'_i = \lambda_1 \mathbf{r}_i + \lambda_2 \mathbf{r}_j & i = j \\ \mathbf{r}_i \rightarrow \mathbf{r}'_i = \mathbf{r}_i & i \neq j \end{cases} \quad , \quad \lambda_1 \neq 0$$

For example, the interchange of i^{th} and j^{th} rows can be carried out as

$$\left\{ \begin{matrix} \mathbf{r}_i \\ \mathbf{r}_j \end{matrix} \right\} \xrightarrow{R_{ij}(1,1)} \left\{ \begin{matrix} \mathbf{r}'_i = \mathbf{r}_i + \mathbf{r}_j \\ \mathbf{r}'_j = \mathbf{r}_j \end{matrix} \right\} \xrightarrow{R_{ji}(-1,1)} \left\{ \begin{matrix} \mathbf{r}''_i = \mathbf{r}'_i - \mathbf{r}'_j = \mathbf{r}_i \\ \mathbf{r}''_j = -\mathbf{r}'_j + \mathbf{r}'_i = \mathbf{r}_i \end{matrix} \right\} \xrightarrow{R_{ij}(1,-1)} \left\{ \begin{matrix} \mathbf{r}'''_i = \mathbf{r}''_i + \mathbf{r}''_j = \mathbf{r}_j \\ \mathbf{r}'''_j = \mathbf{r}''_j = \mathbf{r}_i \end{matrix} \right\}$$

while the other two elementary row operations can be viewed as simply special cases of the row operation (??).

Now suppose \mathbf{v} is a vector lying in the span of row vectors of \mathbf{A} . I will show that it also lies in the span of the row vectors of the matrix \mathbf{A}' obtained by applying the row operation (??) to \mathbf{A} .

$$\begin{aligned} \mathbf{v} &\in span(\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_m) \Rightarrow \mathbf{v} = c_1 \mathbf{r}_1 + \dots + c_i \mathbf{r}_i + \dots + c_j \mathbf{r}_j + \dots + c_m \mathbf{r}_m \\ &= c_1 \mathbf{r}'_1 + \dots + c_i \left(\frac{1}{\lambda_1} (\mathbf{r}'_i - \lambda_2 \mathbf{r}'_j) \right) + \dots + c_j \mathbf{r}'_j + \dots + c_m \mathbf{r}'_m \\ &= c_1 \mathbf{r}'_1 + \dots + \left(\frac{c_i}{\lambda_1} \right) \mathbf{r}'_i + \dots + \left(c_j - \frac{c_i \lambda_2}{\lambda_1} \right) \mathbf{r}'_j + \dots + c_m \mathbf{r}'_m \\ &\in span(\mathbf{r}'_1, \dots, \mathbf{r}'_m) \end{aligned}$$

Thus, the row spaces of \mathbf{A} and \mathbf{A}' are the same. If \mathbf{A}' is row equivalent to \mathbf{A} , then by definition there must be a sequence of row operations that converts \mathbf{A} into \mathbf{A}' .

$$\mathbf{A} \rightarrow \mathbf{A}^{(1)} \rightarrow \mathbf{A}^{(2)} \rightarrow \dots \rightarrow \mathbf{A}^{(k)} = \mathbf{A}'$$

From the preceding paragraph, we know at each intermediate stage we have $\text{RowSp}(\mathbf{A}^{(i)}) = \text{RowSp}(\mathbf{A}^{(i+1)})$ so we conclude

$$\text{RowSp}(\mathbf{A}') = \text{RowSp}(\mathbf{A})$$

LEMMA 10.2. *Let \mathbf{A} be an $m \times n$ matrix and let \mathbf{A}' be its reduction to row echelon form. Then the non-zero rows of \mathbf{A}' form a basis for the row space of \mathbf{A} .*

The basis idea underlying the proof of this lemma is best illustrated by an example. Suppose \mathbf{A} is a 4×5 matrix that is row equivalent to the following matrix in reduced row-echelon form

$$\mathbf{A}'' = \begin{bmatrix} 1 & 1 & 0 & 0 & 3 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Clearly the span of the row vectors of \mathbf{A}' is just the span of the first three row vectors (that is to say, the contribution of the last row to the row space of \mathbf{A} is just $\mathbf{0}$). On the other hand, it's clear the only way we can satisfy

$$\mathbf{0} = c_1 \mathbf{r}_1 + c_2 \mathbf{r}_2 + c_3 \mathbf{r}_3$$

is by taking $c_1 = c_2 = c_3 = 0$; because that's the only way to kill off the components of the total sum that come from the pivots of \mathbf{r}_1 , \mathbf{r}_2 and \mathbf{r}_3 (that is, we can't force a cancellation of terms coming from two different rows because only the pivot row will have a non-zero entry in the component corresponding a column with a pivot). Thus,

$$\begin{aligned} \mathbf{0} &= c_1 \mathbf{r}_1 + c_2 \mathbf{r}_2 + c_3 \mathbf{r}_3 \Rightarrow c_1 = c_2 = c_3 = 0 \\ &\Rightarrow \{\mathbf{r}_1, \mathbf{r}_2, \mathbf{r}_3\} \text{ is a basis for } \text{span}(\mathbf{r}_1, \mathbf{r}_2, \mathbf{r}_3) = \text{RowSp}(\mathbf{A}'') = \text{RowSp}(\mathbf{A}) \end{aligned}$$

However, this isn't quite the statement of the lemma. For the lemma says the row vectors of a matrix in (un-reduced) echelon form should be a basis for the row space of \mathbf{A} . However, we can conclude this simply by noting that

$$\begin{aligned} \dim(\text{RowSp}(\mathbf{A})) &= \text{number of vectors in a basis for } \text{RowSp}(\mathbf{A}) \\ &= \text{number of non-zero rows in reduced echelon-form } \mathbf{A}'' \text{ of } \mathbf{A} \\ &= \text{number of non-zero rows in an echelon-form } \mathbf{A}' \text{ of } \mathbf{A} \end{aligned}$$

But because the row vectors of the matrix in echelon-form span $\text{RowSp}(\mathbf{A})$, and because the number of these row vectors is the same as the dimension of $\text{RowSp}(\mathbf{A})$, we can use Statement 3(b) of Theorem 9.6 (at the end of Lecture 9) to conclude that the row vectors of \mathbf{A}' form a basis for $\text{RowSp}(\mathbf{A})$.

LEMMA 10.3. *Let \mathbf{A} be an $m \times n$ matrix and let \mathbf{A}' be its reduction to row echelon form. Then the columns of \mathbf{A} corresponding to the columns of \mathbf{A}' containing the pivots of \mathbf{A}' form a basis for the column space of \mathbf{A} .*

This lemma is also best demonstrated by example. Suppose \mathbf{A} is a 4×5 matrix that is row equivalent to the following matrix in reduced row-echelon form

$$\mathbf{A}'' = \begin{bmatrix} 1 & 1 & 0 & 0 & 3 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Note that the pivots have been designated by bold-face type. Now the column space of \mathbf{A}'' will be identical to the row space of its transpose

$$(\mathbf{A}'')^T = \begin{bmatrix} \mathbf{1} & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & \mathbf{1} & 0 & 0 \\ 0 & 0 & \mathbf{1} & 0 \\ 3 & 1 & 1 & 0 \end{bmatrix} \xrightarrow{\substack{R_2 \rightarrow R_2 - R_1 \\ R_5 \rightarrow R_5 - 3R_1 - R_3 - R_4}} \begin{bmatrix} \mathbf{1} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & \mathbf{1} & 0 & 0 \\ 0 & 0 & \mathbf{1} & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

where we've used the rows containing the old pivots to clear out the rest of the matrix entries. Obviously, the remaining non-zero rows will be a basis for the $RowSp[(\mathbf{A}'')^T] = ColSp[\mathbf{A}''] = ColSp[\mathbf{A}]$. But the non-zero rows of $(\mathbf{A}'')^T$ are just (actually, linear combinations of) of the columns of \mathbf{A}' containing pivots. Therefore, the columns of \mathbf{A}' that contain pivots correspond to a basis for the column space of \mathbf{A} .

EXAMPLE 10.4. Find a basis for the column space of

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

- First we row reduce \mathbf{A} to row-echelon form

$$\begin{array}{l} R_1 \leftrightarrow R_2 \\ R_3 \rightarrow R_3 + R_2 \\ R_4 \rightarrow R_4 + R_3 \end{array} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 1 & 2 & 1 \\ 0 & 2 & 4 & 2 \end{bmatrix} \xrightarrow{\substack{R_3 \rightarrow R_3 - R_2 \\ R_4 \rightarrow R_4 - 2R_3}} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

The last matrix is a row-echelon form of \mathbf{A} . It has pivots in the 1st, 2nd, and 3rd columns. Therefore, the 1st, 2nd, and 3rd columns of the original matrix \mathbf{A} will form a basis for the column space of \mathbf{A} :

$$ColSp(\mathbf{A}) = span \left(\left(\begin{bmatrix} 0 \\ 1 \\ -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 2 \\ 2 \end{bmatrix} \right) \right)$$

THEOREM 10.5. Let \mathbf{A} be an $m \times n$ matrix. The dimension of the row space of \mathbf{A} is equal to the dimension of its column space.

This follows easily from the preceding two lemmas since the number of non-zero rows in a matrix in row-echelon form is exactly equal to the number of columns containing pivots. This theorem leads to the following definition.

DEFINITION 10.6. The **rank** of a matrix is the dimension of its row space (which equals the dimension of its column space).

Recall that the null space of an $m \times n$ matrix \mathbf{A} is the subspace of \mathbb{R}^n corresponding to the solution space of $\mathbf{A}\mathbf{x} = \mathbf{0}$.

THEOREM 10.7. Let \mathbf{A} be an $m \times n$ matrix. Then

$$n = [\text{number of columns of } \mathbf{A}] = \dim[\text{Null space of } \mathbf{A}] + \text{rank}(\mathbf{A})$$

To see why this theorem must be true, consider the following simple example.

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

This matrix is already in reduced row-echelon form. It has three pivots so

$$\text{rank}(\mathbf{A}) = \dim(\text{RowSp}(\mathbf{A})) = \dim(\text{ColSp}(\mathbf{A})) = 3$$

The dimension of its null space is evidently 1 since the solution of the corresponding homogeneous linear system $\mathbf{Ax} = \mathbf{0}$ implies

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow \begin{cases} x_1 = 0 \\ x_2 = 0 \\ x_3 = 0 \end{cases}$$

but leaves x_4 undetermined. Hence, the dimension of the null space of \mathbf{A} is 1. Thus,

$$4 = \text{number of columns of } \mathbf{A} = 3 + 1 = (\text{rank of } \mathbf{A}) + (\dim(\text{null space of } \mathbf{A}))$$

In the next lecture we shall develop a geometric interpretation of this fundamental fact.