

# Lecture 1

## More On Rank

We will prove some more theorems regarding the rank of a matrix today. The ultimate goal is to prove that  $\text{rank } A = \text{rank } A^T$ . Before we get there, here is a theorem

### Theorem 1.1

Let  $A$  be an  $m \times n$  matrix with rank  $r$ . Then  $r \leq m$  and  $r \leq n$  and, by means of a finite number of elementary row and column operations,  $A$  can be transformed into a matrix of the form below, denoted as  $D$ .

$$D = \begin{bmatrix} I_r & O_1 \\ O_2 & O_3 \end{bmatrix}$$

Where  $I_r$  is the  $r \times r$  identity matrix,  $O_1$  is the  $r \times (n - r)$  zero matrix,  $O_2$  is the  $(m - r) \times r$  zero matrix, and  $O_3$  is the  $(m - r) \times (n - r)$  zero matrix.

*Proof.*

□

## 1.1 Elementary Row & Column Operations

There are three kinds of elementary row operations. They are

1. Switch two rows, denoted as  $R_i \longleftrightarrow R_j$
2. Multiply a row by a non-zero scalar, denoted as  $R_i \rightarrow dR_i$ , for  $d \neq 0$
3. Add a scalar multiple of a row to another row, denoted as  $dR_i + R_j \rightarrow R_j$

Additionally, each of the elementary row operations above has a corresponding *elementary column operation*, which is performed on the columns instead of the rows.

### 1.1.1 Elementary Matrices

Each elementary row operation can be defined using an elementary matrix. The elementary matrix is obtained by performing that row operation on the identity. Namely, define the following

$E_{i \longleftrightarrow j}$  = matrix obtained after applying type 1 row operation to the identity

$E_{di}$  = matrix obtained after applying type 2 row operation to the identity

$E_{i+dj}$  = matrix obtained after applying type 3 row operation to the identity

Elementary matrices for column operations are defined similarly. The next fact shows how the elementary matrices can be used to apply row/column operations.

Let  $A'$  be the matrix obtained from  $A$  by a row (column) operation. Then  $A' = EA$  ( $A' = AG$ ) where  $E$  ( $G$ ) is the matrix corresponding to the row (column) operation as seen above.

Essentially the fact says that multiplying an elementary matrix to a matrix on the correct side has the same effect as applying that elementary row/column operation to that matrix. Using this fact, we can prove this corollary

### Corollary 1.1

Let  $A$  be an  $m \times n$  matrix such that  $\text{rank } A = r$ . Then there exist invertible matrices  $B$  and  $C$  of sizes  $m \times m$  and  $n \times n$  respectively such that  $D = BAC$  (where  $D$  is the matrix defined in theorem 13.1)

*Proof.* Theorem 13.1 says that there is a finite set of elementary row/column operations to transform any matrix into that form. Combining this with the last fact, let  $E_1, E_2, \dots, E_k$  be the elementary matrices for the row operations required and let  $G_1, G_2, \dots, G_\ell$  be the elementary matrices for the column operations. Define

$$E = E_k \cdots E_3 E_2 E_1$$

$$G = G_1 G_2 \cdots G_\ell$$

Then these matrices are the ones that will satisfy the theorem. These matrices are invertible since the product of invertible matrices is also invertible.  $\square$

The next corollary will give us an important result. Before, we will define some important terms related to the columns and rows of a matrix.

### Definition 1.1: Row & Column Rank

Let  $A$  be an  $m \times n$  matrix. Then we have

1. The **row rank** of  $A$  is the maximum number of linearly independent rows in  $A$
2. The **column rank** of  $A$  is the maximum number of linearly independent columns in  $A$ .

With the terms defined, we can present the corollary.

### Corollary 1.2

Let  $A$  be  $m \times n$  matrix. Then we have

1.  $\text{rank } A = \text{rank } A^T$
2.  $\text{rank } A = \text{row rank of } A$
3.  $\text{rank } A = \text{column rank of } A$

*Proof.* 1. PSS 1.

2. Follows from (1) and (3)

3. This is by definition of rank  $A$ .

$\square$

The next theorem gives us some extra ways to determine whether a matrix is invertible.

### Corollary 1.3

A matrix  $A$  is invertible if and only if  $A$  is a product of elementary matrices.

*Proof.* Since  $A$  is invertible, it follows that  $\text{rank } A = n$  (from rank-nullity theorem). So  $D = I_n$ . Then  $I_n = D = BAC$  and so we have

$$A = B^{-1}C^{-1}$$

$$= E_1^{-1}E_2^{-1} \cdots E_k^{-1}G_\ell^{-1} \cdots G_1^{-1}$$

And the inverse of an elementary matrix is also elementary. Thus, we have  $A$  is a product of elementary matrices as desired.

The other direction is easy since all elementary matrices are invertible, so their product is also invertible.  $\square$

The next theorem relates the rank of a matrix to the rank of a product of two matrices, which can be viewed as a composition of linear transformations.

**Theorem 1.2**

Let  $A$  and  $B$  be matrices such that the product  $AB$  is defined. Then, we have

1.  $\text{rank } AB \leq \text{rank } A$
2.  $\text{rank } AB \leq \text{rank } B$

*Proof.* Observe that by the definition of rank of a matrix, we have

$$\begin{aligned}\text{rank } AB &= \text{rank}(L_{AB}) \\ &= \dim(\text{Range } L_{AB})\end{aligned}$$

Now observe that  $\text{Range } L_{AB} \subseteq \text{Range } L_A$ . Now, we have that  $\text{Range } L_{AB} = L_A(\text{Range } L_B)$ , which is a subspace of the domain of  $A$ . Then, this implies that the result.

For the second part, observe that

$$\begin{aligned}\text{rank } AB &= \text{rank}(AB)^T \\ &= \text{rank}(B^T A^T) \\ &\leq \text{rank } B^T \\ &= \text{rank } B\end{aligned}$$

$\square$