

# MATH 40 LECTURE 10: BASES, DIMENSION AND RANK

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In this lecture we return to our discussion of subspaces. We learn what a basis is, and use it to define the dimension of a subspace. We also revisit the notion of rank, and obtain a second part of the Fundamental Theorem of Line Integrals.

**Definition 1.** Let  $S$  be a subspace of  $\mathbb{R}^n$ . A subset  $B \subset S$  of vectors in  $S$  is called a basis of  $S$  if and only if

- (1)  $B$  spans  $S$  and
- (2)  $B$  is linearly independent.

**Example 2.** Consider the following.

- (1) The vectors  $\vec{e}_1, \dots, \vec{e}_n$  form a basis of  $\mathbb{R}^n$ .
- (2) The vectors  $(1, 0)$  and  $(1, 1)$  are also a basis of  $\mathbb{R}^2$ .

**Definition 3.** The vectors  $\{e_1, \dots, e_n\}$  are called the standard basis of  $\mathbb{R}^n$ .

**Example 4.** Find a basis for  $\text{span}\{(1, 2), (2, 7), (-3, -6)\}$ . Note that  $\{(1, 2), (2, 7)\}$  is a linearly independent set, and  $(-3, -6) \in \text{span}\{(1, 2), (2, 7)\}$ . Thus  $\{(1, 2), (2, 7)\}$  is a basis of

$$\text{span}\{(1, 2), (2, 7), (-3, -6)\}.$$

**Remark 5.** We will study a matrix  $A$  by searching for bases of the row space, column space and null space of  $A$ . How can we do so in practice? Let  $R$  be the reduced row echelon form of  $A$ . Then the nonzero row vectors of  $R$  form a basis of  $\text{row}(A)$ . Also, the leading columns form a basis of  $\text{col}(A)$ . We use the free variables of  $R\vec{x} = \vec{0}$  to determine a basis of  $\text{null}(A)$ .

**Theorem 6 (Basis Theorem).** Let  $S$  be a subspace of  $\mathbb{R}^n$ . Then any two bases of  $S$  have the same number of vectors.

**Definition 7.** Let  $S$  be a subspace of  $\mathbb{R}^n$ . The dimension of  $S$  is the number of elements in any basis of  $S$ .

**Example 8.**

$$\boxed{\dim(\mathbb{R}^n) = n.}$$

**Theorem 9.** For any matrix  $A$ ,

$$\dim \text{row}(A) = \dim \text{col}(A).$$

**Corollary 10.** The rank of  $A$  is equal to the dimension of its row space, which is the same as the dimension of the column space of  $A$ .

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These are lecture notes for HMC Math 40: Introduction to Linear Algebra and roughly follow our course text *Linear Algebra* by David Poole.

**Example 11.** Consider

$$A = \begin{pmatrix} 1 & 2 & -3 \\ 2 & 7 & -6 \end{pmatrix}.$$

We saw above that  $\dim \text{col}(A) = 2$ , thus  $\text{rank}(A) = 2$ . Note that we can also determine this from the rows of  $A$ . The reduced echelon form of  $A$  is

$$R = \begin{pmatrix} 1 & 0 & -3 \\ 0 & 1 & 0 \end{pmatrix}.$$

Thus  $R$  has two nonzero rows. Hence, again,  $\text{rank}(A) = 2$ .

**Definition 12.** The nullity of  $A$  is the dimension of the null space of  $A$ ,

$$\text{nullity}(A) = \dim \text{null}(A).$$

**Theorem 13** (Rank-Nullity Theorem). If  $A$  is an  $m \times n$  matrix, then

$$\text{rank}(A) + \text{nullity}(A) = n.$$

PROOF. Let  $R$  be the reduced row echelon form of  $A$ . Suppose  $\text{rank}(R) = r$ . Then  $R$  has  $r$  leading 1's. Thus, by the Rank Theorem, there are  $n - r$  free variables in the homogeneous system  $(A|\vec{0})$ . Each free variable has a corresponding basis element in the null space of  $A$ . Thus  $\dim \text{null}(A) = n - r$ . Therefore

$$\text{rank}(A) + \text{nullity}(A) = r + (n - r) = n. \quad \square$$

**Theorem 14** (Fundamental Theorem of Invertible Matrices Part II). Let  $A$  be an  $n \times n$  matrix. The following are equivalent.

- i  $A$  is invertible.
- ii  $\text{rank}(A) = n$ .
- iii  $\text{nullity}(A) = 0$ .
- iv The columns of  $A$  are linearly independent.
- v The columns of  $A$  span  $\mathbb{R}^n$ .
- vi The columns of  $A$  are a basis of  $\mathbb{R}^n$ .
- vii The rows of  $A$  are linearly independent.
- viii The rows of  $A$  span  $\mathbb{R}^n$ .
- ix The rows of  $A$  form a basis of  $\mathbb{R}^n$ .

PROOF.

- (i)  $\Rightarrow$  (ii) By FTIM I,  $A$  is invertible if and only if its reduced row echelon form is  $I_n$ . But  $I_n$  has  $n$  nonzero rows. Thus  $\text{rank}(A) = n$ .
- (ii)  $\Leftrightarrow$  (iii) This holds by the Rank-Nullity Theorem.
- (ii)  $\Rightarrow$  (iv) If  $\text{rank}(A) = n$ , then the linear system  $A\vec{x} = \vec{0}$  has no free variables, and hence has only the trivial solution. Therefore the columns of  $A$  are linearly independent. Note that, by FTIM I, this also shows that  $A$  is invertible.
- (iv)  $\Rightarrow$  (v) If the columns of  $A$  are linearly independent, then  $A\vec{x} = \vec{0}$  has only the trivial solution. Thus, by FTIM I,  $A\vec{x} = \vec{b}$  has a unique solution for every  $\vec{b} \in \mathbb{R}^n$ . Therefore every vector  $\vec{b}$  can be written as a linear combination of the columns of  $A$ .

(v)  $\Rightarrow$  (vi) Suppose the columns of  $A$  span  $\mathbb{R}^n$ . Then  $\dim \text{col}(A) = n$ . But  $\text{rank}(A) = \dim \text{col}(A)$ , so  $\text{rank}(A) = n$ . This is (ii). But we already showed (ii)  $\Rightarrow$  (iv). Thus, the columns of  $A$  are linearly independent. They also span  $\mathbb{R}^n$  by assumption. Therefore they are a basis of  $\mathbb{R}^n$ .

(vi)  $\Rightarrow$  (ii)

We have now proved (i)  $\Rightarrow$  (ii)  $\Leftrightarrow$  (iii)  $\Rightarrow$  (iv)  $\Rightarrow$  (v)  $\Rightarrow$  (vi)  $\Rightarrow$  (ii)  $\Rightarrow$  (i). Hence these statements are equivalent. Now, we claim that  $\text{rank}(A) = \text{rank}(A^T)$  for any matrix  $A$ . Indeed,

$$\begin{aligned}\text{rank}(A^T) &= \dim \text{col}(A^T) \\ &= \dim \text{row}(A) \\ &= \text{rank}(A).\end{aligned}$$

Also, we know that  $A$  is invertible if and only if  $A^T$  is invertible.

Therefore, we may apply our proven results to  $A^T$ . But the columns of  $A^T$  are the rows of  $A$ . This completes the proof.  $\square$