Subspaces associated to a matrix

There are three important subspaces associated to an  $m \times n$  matrix **A**. Let  $\mathbf{c}_1, \ldots, \mathbf{c}_n$  represent the columns of **A**. That is,

$$\mathbf{A} = \left[ \begin{array}{c|c} \mathbf{c}_1 & \mathbf{c}_2 & \dots & \mathbf{c}_n \end{array} 
ight].$$

These column vectors are vectors in  $\mathbb{R}^m$ .

Let  $\mathbf{r}_1, \ldots, \mathbf{r}_m$  represent the rows of **A**. That is,

$$\mathbf{A} = egin{bmatrix} \mathbf{r}_1 & & \ \hline & \mathbf{r}_2 & \ \hline & & \ \hline & & \ \hline & & \mathbf{r}_m \end{bmatrix}.$$

These row vectors are vectors in  $\mathbb{R}^n$ .

The column space of A. The column space of A is the span of the columns of A. We write

Col 
$$\mathbf{A} = \operatorname{Span} \{ \mathbf{c}_1, \ldots, \mathbf{c}_n \}.$$

The row space of A. The row space of A is the span of the rows of A. We write

Row 
$$\mathbf{A} = \operatorname{Span}\{\mathbf{r}_1, \ldots, \mathbf{r}_m\}.$$

The null space of A. The null space of A is the set of all vectors  $\mathbf{x}$  in  $\mathbb{R}^n$  such that

Ax = 0.

The null space of **A** is denoted by Nul **A**.

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**Theorem.** Let  $\mathbf{A}$  be an  $m \times n$  matrix. The column space of  $\mathbf{A}$  is a subspace of  $\mathbb{R}^m$ , and the null space and the row space of  $\mathbf{A}$  are subspaces of  $\mathbb{R}^n$ .

The consistency of a system of linear equations can be viewed as a statement about the column space of the coefficient matrix.

Fact. The linear system Ax = b is consistent if and only if b is an element of the column space of A.

Here is how Lay (p. 204) contrasts Nul A and Col A for an  $m \times n$  matrix A:

## Nul $\mathbf{A}$

- 1. Nul **A** is a subspace of  $\mathbb{R}^n$ .
- 2. Nul A is implicitly defined; that is, you are given only a condition (Ax = 0) that vectors in Nul A must satisfy.
- It takes time to find vectors in Nul A. Row operations on [A 0] are required.
- 4. There is no obvious relation between Nul **A** and the entries in **A**.
- 5. A typical vector  $\mathbf{v}$  in Nul  $\mathbf{A}$  has the property that  $\mathbf{A}\mathbf{v} = \mathbf{0}$ .
- 6. Given a specific vector **v**, it is easy to tell if **v** is in Nul **A**. Just compute **Av**.
- 7. Nul  $\mathbf{A} = \{\mathbf{0}\}$  if and only if the equation  $\mathbf{A}\mathbf{x} = \mathbf{0}$  has only the trivial solution.
- 8. Nul  $\mathbf{A} = \{\mathbf{0}\}$  if and only if the linear transformation  $\mathbf{x} \mapsto \mathbf{A}\mathbf{x}$  is one-to-one.

## $\operatorname{Col}\,\mathbf{A}$

- 1. Col **A** is a subspace of  $\mathbb{R}^m$ .
- 2. Col **A** is explicitly defined; that is, you are told how to build vectors in Col **A**.
- 3. It is easy to find vectors in Col **A**. The columns of **A** are displayed; others are formed from them.
- There is an obvious relation between Col A and the entries in A, since each column of A is in Col A.
- 5. A typical vector  $\mathbf{v}$  in Col  $\mathbf{A}$  has the property that the equation  $\mathbf{A}\mathbf{x} = \mathbf{v}$  is consistent.
- 6. Given a specific vector  $\mathbf{v}$ , it may take time to tell if  $\mathbf{v}$  is in Col  $\mathbf{A}$ . Row operations on  $[\mathbf{A} \quad \mathbf{v}]$  are required.
- 7. Col  $\mathbf{A} = \mathbb{R}^m$  if and only if the equation  $\mathbf{A}\mathbf{x} = \mathbf{b}$  has a solution for every  $\mathbf{b}$  in  $\mathbb{R}^m$ .
- 8. Col  $\mathbf{A} = \mathbb{R}^m$  if and only if the linear transformation  $\mathbf{x} \mapsto \mathbf{A}\mathbf{x}$  maps  $\mathbb{R}^n$  onto  $\mathbb{R}^m$ .

Item 8 in both lists suggest two subspaces that are intimately connected with any linear transformation from one vector space to another.

**Definition.** A transformation  $L: V_1 \to V_2$  from a vector space  $V_1$  to a vector space  $V_2$  is linear if

- 1.  $L(\mathbf{v}_1 + \mathbf{v}_2) = L(\mathbf{v}_1) + L(\mathbf{v}_2)$  for all vectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$  in  $V_1$ , and
- 2.  $L(r\mathbf{v}) = rL(\mathbf{v})$  for all  $\mathbf{v}$  in  $V_1$  and all r in  $\mathbb{R}$ .

**Example.** Let  $V_1$  be the vector space of all continuously differentiable functions  $f : \mathbb{R} \to \mathbb{R}$ and let  $V_2$  be the vector space of all continuous functions  $f : \mathbb{R} \to \mathbb{R}$ . The operation of differentiation is a linear transformation from  $V_1$  to  $V_2$ . That is, the transformation  $D: V_1 \to V_2$  given by

$$D(f) = f'$$

is a linear transformation.

Associated to any linear transformation are two important subspaces.

**Definition.** The kernel of  $L: V_1 \to V_2$  is the subset of  $\mathbf{V}_1$  given by

$$\{\mathbf{v}_1 \mid L(\mathbf{v}_1) = \mathbf{0}\}.$$

The range of L is the subset of  $\mathbf{V}_2$  given by

 $\{\mathbf{v}_2 \,|\, L(\mathbf{v}_1) = \mathbf{v}_2 \text{ for some } \mathbf{v}_1 \text{ in } V_1 \}.$ 

**Fact.** Both the kernel and the range of a linear transformation are subspaces. The kernel is a subspace of  $V_1$ , and the range is a subspace of  $V_2$ .

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For a matrix transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^m$  determined by the matrix **A**, its range is Col **A**, and its kernel is Nul **A**.

**Example.** What are the kernel and range of the transformation  $p : \mathbb{R}^3 \to \mathbb{R}^3$  determined by the matrix

$$\frac{1}{3} \begin{bmatrix} 2 & -1 & 1 \\ -1 & 2 & 1 \\ 1 & 1 & 2 \end{bmatrix}?$$

(This example was first introduced on September 20.)

**Example.** What are the kernel and the range of the differentiation transformation D mentioned above?

If we are careful, we can also use integration to define a linear transformation.

**Example.** Given a polynomial  $p(x) = a_n x^n + a_{n-1} x^{n-1} + \ldots + a_0$  in  $\mathbb{P}_n$  (the vector space of all polynomial functions of degree at most n), we can define

$$I(p) = \int_0^x p(t) dt = a_n \frac{x^{n+1}}{n+1} + a_{n-1} \frac{x^n}{n} + \ldots + a_0 x.$$

The map  $I: \mathbb{P}_n \to \mathbb{P}_{n+1}$  is a linear transformation. What are its kernel and range?