# Math 221: LINEAR ALGEBRA

# Chapter 7. Linear Transformations §7-2. Kernel and Image

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Slides are adapted from those by Karen Seyffarth from University of Calgary.

What are the Kernel and the Image?

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# What are the Kernel and the Image?

#### Definition

Let V and W be vector spaces, and  $T:V\to W$  a linear transformation.

1. The kernel of T (sometimes called the null space of T) is defined to be the set

$$\ker(\mathbf{T}) = \{ \vec{\mathbf{v}} \in \mathbf{V} \mid \mathbf{T}(\vec{\mathbf{v}}) = \vec{\mathbf{0}} \}.$$

2. The image of T is defined to be the set

$$im(T) = \{T(\vec{v}) \mid \vec{v} \in V\}.$$

#### Remark

If A is an  $m\times n$  matrix and  $T_A:\mathbb{R}^n\to\mathbb{R}^m$  is the linear transformation induced by A, then

- $ightharpoonup \ker(T_A) = \text{null}(A);$
- $ightharpoonup \operatorname{im}(T_A) = \operatorname{im}(A).$

#### Example

Let  $T: \mathcal{P}_1 \to \mathbb{R}$  be the linear transformation defined by

$$T(p(x))=p(1) \text{ for all } p(x)\in \mathcal{P}_1.$$

$$\begin{aligned} \ker(T) &= & \{ p(x) \in \mathcal{P}_1 \mid p(1) = 0 \} \\ &= & \{ ax + b \mid a, b \in \mathbb{R} \quad \text{and} \quad a + b = 0 \} \\ &= & \{ ax - a \mid a \in \mathbb{R} \}. \end{aligned}$$

$$\begin{split} \operatorname{im}(T) &=& \{p(1) \mid p(x) \in \mathcal{P}_1\} \\ &=& \{a+b \mid ax+b \in \mathcal{P}_1\} \\ &=& \{a+b \mid a,b \in \mathbb{R}\} \\ &=& \mathbb{R}. \end{split}$$

#### Theorem

Let V and W be vector spaces and  $T:V\to W$  a linear transformation. Then  $\ker(T)$  is a subspace of V and  $\operatorname{im}(T)$  is a subspace of W.

# Proof. (that ker(T) is a subspace of V)

- 1. Let  $\vec{0}_V$  and  $\vec{0}_W$  denote the zero vectors of V and W, respectively. Since  $T(\vec{0}_V) = \vec{0}_W$ ,  $\vec{0}_V \in \ker(T)$ .
- 2. Let  $\vec{v}_1, \vec{v}_2 \in \ker(T)$ . Then  $T(\vec{v}_1) = \vec{0}$ ,  $T(\vec{v}_2) = \vec{0}$ , and

$$T(\vec{v}_1 + \vec{v}_2) = T(\vec{v}_1) + T(\vec{v}_2) = \vec{0} + \vec{0} = \vec{0}.$$

Thus  $\vec{v}_1 + \vec{v}_2 \in \ker(T)$ .

3. Let  $\vec{v}_1 \in \ker(T)$  and let  $k \in \mathbb{R}$ . Then  $T(\vec{v}_1) = \vec{0}$ , and

$$T(k\vec{v}_1) = kT(\vec{v}_1) = k(\vec{0}) = \vec{0}.$$

Thus  $k\vec{v}_1 \in \ker(T)$ .

By the Subspace Test, ker(T) is a subspace of V.

# Proof. (that im(T) is a subspace of W)

- 1. Let  $\vec{0}_V$  and  $\vec{0}_W$  denote the zero vectors of V and W, respectively. Since  $T(\vec{0}_V) = \vec{0}_W$ ,  $\vec{0}_W \in \operatorname{im}(T)$ .
- 2. Let  $\vec{w}_1, \vec{w}_2 \in \text{im}(T)$ . Then there exist  $\vec{v}_1, \vec{v}_2 \in V$  such that  $T(\vec{v}_1) = \vec{w}_1$ ,  $T(\vec{v}_2) = \vec{w}_2$ , and thus

$$\vec{w}_1 + \vec{w}_2 = T(\vec{v}_1) + T(\vec{v}_2) = T(\vec{v}_1 + \vec{v}_2).$$

Since  $\vec{v}_1 + \vec{v}_2 \in V$ ,  $\vec{w}_1 + \vec{w}_2 \in \text{im}(T)$ .

3. Let  $\vec{w}_1 \in \text{im}(V)$  and let  $k \in \mathbb{R}$ . Then there exists  $\vec{v}_1 \in V$  such that  $T(\vec{v}_1) = \vec{w}_1$ , and

$$k\vec{w}_1 = kT(\vec{v}_1) = T(k\vec{v}_1).$$

Since  $k\vec{v}_1 \in V$ ,  $k\vec{w}_1 \in im(T)$ .

By the Subspace Test, im(T) is a subspace of W.

#### Definition

Let V and W be vector spaces and  $T: V \to W$  a linear transformation.

 The dimension of ker(T), dim(ker(T)) is called the nullity of T and is denoted nullity(T), i.e.,

$$\operatorname{nullity}(T) = \dim(\ker(T)).$$

 The dimension of im(T), dim(im(T)) is called the rank of T and is denoted rank (T), i.e.,

$$rank(T) = dim(im(T)).$$

## Example

If A is an  $m \times n$  matrix, then

$$im(T_A) = im(A) = col(A).$$

It follows that

$$rank (T_A) = dim(im(T_A)) = dim(col(A)) = rank (A).$$

Also, 
$$ker(T_A) = null(A)$$
, so

$$\operatorname{nullity}(T_A) = \operatorname{dim}(\operatorname{null}(A)) = n - \operatorname{rank}(A).$$

# Finding bases of the kernel and the image

# Example (continued)

For the linear transformation T defined by  $T : \mathcal{P}_1 \to \mathbb{R}$ 

$$T(p(x))=p(1) \text{ for all } p(x)\in \mathcal{P}_1,$$

we found that

$$ker(T) = \{ax - a \mid a \in \mathbb{R}\} \text{ and }$$
  
 $im(T) = \mathbb{R}.$ 

From this, we see that  $\ker(T) = \operatorname{span}\{(x-1)\}$ ; since  $\{(x-1)\}$  is an independent subset of  $\mathcal{P}_1$ ,  $\{(x-1)\}$  is a basis of  $\ker(T)$ . Thus

$$\dim(\ker(T)) = 1 = \text{nullity}(T).$$

Since  $\operatorname{im}(T) = \mathbb{R}$ ,  $\operatorname{dim}(\operatorname{im}(T)) = 1 = \operatorname{rank}(T)$ .

#### Problem

Let  $T: \mathbf{M}_{22} \to \mathbf{M}_{22}$  be defined by

$$T\left[\begin{array}{cc} a & b \\ c & d \end{array}\right] = \left[\begin{array}{cc} a+b & b+c \\ c+d & d+a \end{array}\right] \text{ for all } \left[\begin{array}{cc} a & b \\ c & d \end{array}\right] \in \boldsymbol{M}_{22}.$$

Then T is a linear transformation (you should be able to prove this). Find a basis of ker(T) and a basis of im(T).

#### Solution

Suppose 
$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \in \ker(T)$$
. Then
$$T \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} a+b & b+c \\ c+d & d+a \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}.$$

$$\begin{bmatrix} c & d \end{bmatrix} = \begin{bmatrix} c+d & d+a \end{bmatrix} = \begin{bmatrix} 0 & 0 \end{bmatrix}.$$

This gives us a system of four equations in the four variables a, b, c, d:

$$a + b = 0; b + c = 0; c + d = 0; d + a = 0.$$

# Solution (continued)

This system has solution a=-t, b=t, c=-t, d=t for any  $t\in\mathbb{R},$  and thus

$$\ker(T) = \left\{ \left[ \begin{array}{cc} -t & t \\ -t & t \end{array} \right] \ \middle| \ t \in \mathbb{R} \right\} = \operatorname{span} \left\{ \left[ \begin{array}{cc} -1 & 1 \\ -1 & 1 \end{array} \right] \right\}.$$

Let  $B_k = \left\{ \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix} \right\}$ . Since  $B_k$  is an independent subset of  $\mathbf{M}_{22}$  and  $\operatorname{span}(B) = \ker(T)$ ,  $B_k$  is a basis of  $\ker(T)$ .

$$\begin{split} \operatorname{im}(T) &= \left. \left\{ \left[ \begin{array}{cc} a+b & b+c \\ c+d & d+a \end{array} \right] \; \middle| \; a,b,c,d \in \mathbb{R} \right\} \\ &= & \operatorname{span} \left\{ \left[ \begin{array}{cc} 1 & 0 \\ 0 & 1 \end{array} \right], \left[ \begin{array}{cc} 1 & 1 \\ 0 & 0 \end{array} \right], \left[ \begin{array}{cc} 0 & 1 \\ 1 & 0 \end{array} \right], \left[ \begin{array}{cc} 0 & 0 \\ 1 & 1 \end{array} \right] \right\}. \end{split}$$

Let

$$S = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 1 \end{bmatrix} \right\}.$$

## Solution (continued)

S is a dependent subset of  $M_{22}$ , but (check this yourselves)

$$B_i = \left\{ \left[\begin{array}{cc} 1 & 0 \\ 0 & 1 \end{array}\right], \left[\begin{array}{cc} 0 & 1 \\ 1 & 0 \end{array}\right], \left[\begin{array}{cc} 0 & 0 \\ 1 & 1 \end{array}\right] \right\}$$

is an independent subset of S. Since  $span(B_i) = span(S) = im(T)$  and  $B_i$  is independent,  $B_i$  is a basis of im(T).

# Surjections and Injections

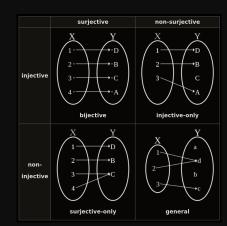
#### Definition

Let V and W be vector spaces and  $T: V \to W$  a linear transformation.

- 1. T is onto (or surjective) if im(T) = W.
- 2. T is one-to-one (or injective) if, for  $\vec{v}, \vec{w} \in V$ ,  $T(\vec{v}) = T(\vec{w})$  implies that  $\vec{v} = \vec{w}$ .

#### Example

Let V be a vector space. Then the identity operator on V,  $1_V:V\to V,$  is one-to-one and onto.



#### Theorem

Let V and W be vector spaces and T : V  $\rightarrow$  W a linear transformation. Then T is one-to-one if and only if  $\ker(T) = \{\vec{0}\}.$ 

#### Proof.

 $(\Rightarrow)$  Let  $\vec{v} \in \ker(T)$ . Then

$$T(\vec{v}) = \vec{0} = T(\vec{0}).$$

Since is one-to-one,  $\vec{v} = \vec{0}$ . But  $\vec{v}$  is an arbitrary element of ker(T), and thus ker T =  $\{\vec{0}\}$ .

( $\Leftarrow$ ) Conversely, suppose that  $\ker(T) = \{\vec{0}\}$ , and let  $\vec{v}, \vec{w} \in V$  be such that

$$T(\vec{v}) = T(\vec{w}).$$

Then  $T(\vec{v}) - T(\vec{w}) = \vec{0}$ , and since T is a linear transformation

$$T(\vec{v} - \vec{w}) = \vec{0}.$$

By definition,  $\vec{v} - \vec{w} \in \ker(T)$ , implying that  $\vec{v} - \vec{w} = \vec{0}$ . Therefore  $\vec{v} = \vec{w}$ , and hence T is one-to-one.

#### Problem

Let  $T: \mathbf{M}_{22} \to \mathbb{R}^2$  be a linear transformation defined by

$$T\left[\begin{array}{cc} a & b \\ c & d \end{array}\right] = \left[\begin{array}{cc} a+d \\ b+c \end{array}\right] \text{ for all } \left[\begin{array}{cc} a & b \\ c & d \end{array}\right] \in \textbf{M}_{22}.$$

Prove that T is onto but not one-to-one.

#### Proof.

Let 
$$\begin{bmatrix} x \\ y \end{bmatrix} \in \mathbb{R}^2$$
. Since  $T \begin{bmatrix} x & y \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix}$ , T is onto.

Observe that  $\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \in \ker(T)$ , so  $\ker(T) \neq \vec{0}_{22}$ . By the previous Theorem, T is not one-to-one.

#### Problem

Suppose U is an invertible  $m \times m$  matrix and let  $T: \mathbf{M}_{mn} \to \mathbf{M}_{mn}$  be defined by

$$T(A) = UA \text{ for all } A \in \mathbf{M}_{mn}.$$

Then T is a linear transformation (this is left to you to verify). Prove that T is one-to-one and onto.

#### Proof.

Suppose  $A, B \in \mathbf{M}_{mn}$  and that T(A) = T(B). Then UA = UB; since U is invertible

$$\begin{array}{rcl} U^{-1}(UA) & = & U^{-1}(UB) \\ (U^{-1}U)A & = & (U^{-1}U)B \\ & I_{mm}A & = & I_{mm}B \\ & A & = & B. \end{array}$$

Therefore, T is one-to-one.

## Proof. (continued)

To prove that T is onto, let  $B \in \mathbf{M}_{mn}$  and let  $A = U^{-1}B$ . Then

$$T(A) = UA = U(U^{-1}B) = (UU^{-1})B = I_{mm}B = B,$$

and therefore T is onto.

#### Problem

Let  $S: \mathcal{P}_2 \to \mathbf{M}_{22}$  be a linear transformation defined by

$$S(ax^{2} + bx + c) = \begin{bmatrix} a+b & a+c \\ b-c & b+c \end{bmatrix} \text{ for all } ax^{2} + bx + c \in \mathcal{P}_{2}$$

Prove that S is one-to-one but not onto.

## Proof.

By definition,

$$ker(S) = \{ax^2 + bx + c \in \mathcal{P}_2 \mid a+b = 0, a+c = 0, b-c = 0, b+c = 0\}.$$

Suppose  $p(x) = ax^2 + bx + c \in ker(S)$ . This leads to a homogeneous system of four equations in three variables. Putting the augmented matrix in reduced row-echelon form:

$$\begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 1 & 1 & 0 \end{bmatrix} \rightarrow \cdots \rightarrow \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$$

Since the unique solution is a = b = c = 0,  $ker(S) = {\vec{0}}$ , and thus S is one-to-one.

## Proof. (continued)

To show that S is **not** onto, show that  $\operatorname{im}(S) \neq \mathcal{P}_2$ ; i.e., find a matrix  $A \in \mathbf{M}_{22}$  such that for every  $p(x) \in \mathcal{P}_2$ ,  $S(p(x)) \neq A$ . Let

$$A = \left[ \begin{array}{cc} 0 & 1 \\ 0 & 2 \end{array} \right],$$

and suppose  $p(x) = ax^2 + bx + c \in \mathcal{P}_2$  is such that S(p(x)) = A. Then

$$a + b = 0$$
  $a + c = 1$   
 $b - c = 0$   $b + c = 2$ 

Solving this system

$$\begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 \\ 0 & 1 & -1 & 0 \\ 0 & 1 & 1 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & -1 & 1 & 1 \\ 0 & 1 & -1 & 0 \\ 0 & 1 & 1 & 2 \end{bmatrix}.$$

Since the system is inconsistent, there is no  $p(x) \in \mathcal{P}_2$  so that S(p(x)) = A, and therefore S is not onto.

Problem ( One-to-one linear transformations preserve independent sets )  $\,$ 

Let V and W be vector spaces and  $T:V\to W$  a linear transformation. Prove that if T is one-to-one and  $\{\vec{v}_1,\vec{v}_2,\ldots,\vec{v}_k\}$  is an independent subset of V, then  $\{T(\vec{v}_1),T(\vec{v}_2),\ldots,T(\vec{v}_k)\}$  is an independent subset of W.

#### Proof.

Let  $\vec{0}_{V}$  and  $\vec{0}_{W}$  denote the zero vectors of V and W, respectively. Suppose that

$$a_1T(\vec{v}_1)+a_2T(\vec{v}_2)+\cdots+a_kT(\vec{v}_k)=\vec{0}_W$$

for some  $a_1, a_2, \ldots, a_k \in \mathbb{R}$ . Since linear transformations preserve linear combinations (addition and scalar multiplication),

$$T(a_1\vec{v}_1 + a_2\vec{v}_2 + \dots + a_k\vec{v}_k) = \vec{0}_W.$$

Now, since T is one-to-one,  $ker(T) = {\vec{0}_V}$ , and thus

$$a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_k \vec{v}_k = \vec{0}_V.$$

However,  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k\}$  is independent, and hence  $a_1 = a_2 = \dots = a_k = 0$ . Therefore,  $\{T(\vec{v}_1), T(\vec{v}_2), \dots, T(\vec{v}_k)\}$  is independent.

Problem (Onto linear transformations preserve spanning sets)

Let V and W be vector spaces and  $T: V \to W$  a linear transformation. Prove that if T is onto and  $V = \text{span}\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k\}$ , then

$$W = \operatorname{span}\{T(\vec{v}_1), T(\vec{v}_2), \dots, T(\vec{v}_k)\}.$$

#### Proof.

Suppose that T is onto and let  $\mathbf{w} \in W$ . Then there exists  $\mathbf{v} \in V$  such that  $T(\mathbf{v}) = \mathbf{w}$ . Since  $V = \mathrm{span}\{\vec{v}_1, \vec{v}_2, \ldots, \vec{v}_k\}$ , there exist  $a_1, a_2, \ldots a_k \in \mathbb{R}$  such that  $\mathbf{v} = a_1\mathbf{v}_1 + a_2\mathbf{v}_2 + \cdots + a_k\mathbf{v}_k$ . Using the fact that T is a linear transformation,

$$\begin{split} \boldsymbol{w} &= T(\boldsymbol{v}) &= T(a_1\boldsymbol{v}_1 + a_2\boldsymbol{v}_2 + \dots + a_k\boldsymbol{v}_k) \\ &= a_1T(\boldsymbol{v}_1) + a_2T(\boldsymbol{v}_2) + \dots + a_kT(\boldsymbol{v}_k), \end{split}$$

i.e.,  $\mathbf{w} \in \mathrm{span}\{T(\vec{v}_1), T(\vec{v}_2), \dots, T(\vec{v}_k)\}$ , and thus

$$W \subseteq \operatorname{span}\{T(\vec{v}_1), T(\vec{v}_2), \ldots, T(\vec{v}_k)\}.$$

Since  $T(\vec{v}_1), T(\vec{v}_2), \ldots, T(\vec{v}_k) \in W$ , it follows that  $\operatorname{span}\{T(\vec{v}_1), T(\vec{v}_2), \ldots, T(\vec{v}_k)\} \subseteq W$ , and therefore  $W = \operatorname{span}\{T(\vec{v}_1), T(\vec{v}_2), \ldots, T(\vec{v}_k)\}$ .

Suppose A is an  $m \times n$  matrix. How do we determine if  $T_A : \mathbb{R}^n \to \mathbb{R}^m$  is onto? How do we determine if  $T_A : \mathbb{R}^n \to \mathbb{R}^m$  is one-to-one?

#### Theorem

Let A be an  $m \times n$  matrix, and  $T_A : \mathbb{R}^n \to \mathbb{R}^m$  the linear transformation induced by A.

- 1.  $T_A$  is onto if and only if rank (A) = m.
- 2.  $T_A$  is one-to-one if and only if rank (A) = n.

# Proof. (sketch)

- 1.  $T_A$  is onto if and only if  $\operatorname{im}(T_A) = \mathbb{R}^m$ . This is equivalent to  $\operatorname{col}(A) = \mathbb{R}^m$ , which occurs if and only if  $\operatorname{dim}(\operatorname{col}(A)) = m$ , i.e.,  $\operatorname{rank}(A) = m$ .
- 2.  $\ker(T_A) = \operatorname{null}(A)$ , and  $\operatorname{null}(A) = \{\vec{0}\}$  if and only if  $A\vec{x} = \vec{0}$  has the **unique** solution  $\vec{x} = \vec{0}$ . Thus and row echelon form of A has a leading one in every column, which occurs if and only if rank (A) = n.

# The Dimension Theorem (Rank-Nullity Theorem)

Suppose A is an  $m \times n$  matrix with rank r. Since  $im(T_A) = col(A)$ ,

$$\dim(\operatorname{im}(T_A)) = \operatorname{rank}\ (A) = r.$$

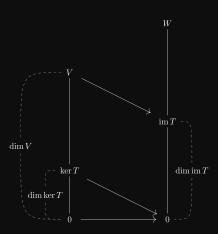
We also know that  $\ker(T_A)=\operatorname{null}(A)$ , and that  $\dim(\operatorname{null}(A))=n-r$ . Thus,  $\dim(\operatorname{im}(T_A))+\dim(\ker(T_A))=n=\dim\ \mathbb{R}^n.$ 

# Theorem (Dimension Theorem (Rank-Nullity Theorem))

Let V and W be vector spaces and  $T: V \to W$  a linear transformation. If  $\ker(T)$  and  $\operatorname{im}(T)$  are both finite dimensional, then V is finite dimensional, and

$$\dim(V) = \dim(\ker(T)) + \dim(\operatorname{im}(T)).$$

Equivalently,  $\dim(V) = \text{nullity}(T) + \text{rank}(T)$ .



# Proof. (Outline)

Let  $\vec{w} \in \text{im}(T)$ ; then  $\vec{w} = T(\vec{v})$  for some  $\vec{v} \in V$ . Suppose

$$\left\{T(\vec{b}_1), T(\vec{b}_2), \dots, T(\vec{b}_r)\right\}$$

is a basis of im(T), and that

$$\left\{\vec{f}_1, \vec{f}_2, \dots, \vec{f}_k\right\}$$

is a basis of ker(T). We define

$$B = \left\{ \vec{b}_1, \vec{b}_2, \dots, \vec{b}_r, \vec{f}_1, \vec{f}_2, \dots, \vec{f}_k \right\}.$$

To prove that B is a basis of V, it remains to prove that B spans V and that B is linearly independent.

Since B is independent and spans V, B is a basis of V, implying V is finite dimensional (V is spanned by a finite set of vectors). Furthermore, |B| = r + k, so

$$\dim(V) = \dim(\operatorname{im}(T)) + \dim(\ker(T)).$$

#### Remark

- It is not an assumption of the theorem that V is finite dimensional. Rather, it is a consequence of the assumption that both im(T) and ker(T) are finite dimensional.
- 2. As a consequence of the Dimension Theorem, if V is a finite dimensional vector space and either  $\dim(\ker(T))$  or  $\dim(\operatorname{im}(T))$  is known, then the other can be easily found.

#### Example

Let V and W be vector spaces and  $T:V\to W$  a linear transformation. If V is finite dimensional, then it follows that

$$\dim(\ker(T)) \le \dim(V)$$
 and  $\dim(\operatorname{im}(T)) \le \dim(V)$ .

## Problem

For  $a \in \mathbb{R}$ , recall that the linear transformation  $E_a : \mathcal{P}_n \to \mathbb{R}$ , the evaluation map at a, is defined as

$$E_a(p(x)) = p(a)$$
 for all  $p(x) \in \mathcal{P}_n$ .

Prove that E<sub>a</sub> is onto, and that

$$B = \{(x - a), (x - a)^2, (x - a)^3, \dots, (x - a)^n\}$$

is a basis of  $ker(E_a)$ .

#### Proof.

Let  $t \in \mathbb{R}$ , and choose  $p(x) = t \in \mathcal{P}_n$ . Then p(a) = t, so  $E_a(p(x)) = t$ , i.e.,  $E_a$  is onto.

By the Dimension Theorem,

$$n+1=\dim(\mathcal{P}_n)=\dim(\ker(E_a))+\dim(\operatorname{im}(E_a)).$$

Since  $E_a$  is onto,  $\dim(\operatorname{im}(E_a)) = \dim(\mathbb{R}) = 1$ , and thus  $\dim(\ker(E_a)) = n$ .

It now suffices to find n independent polynomials in ker(E<sub>a</sub>).

Note that  $(x-a)^j \in \ker(E_a)$  for  $j=1,2,\ldots,n,$  so  $B \subseteq \ker(E_a).$ 

Furthermore, B is independent because the polynomials in B have distinct degrees.

Since  $|B| = n = \dim(\ker(E_a))$ , B spans  $\ker(E_a)$ .

Therefore, B is a basis of  $ker(E_a)$ .

#### Theorem

Let V and W be vector spaces,  $T:V\to W$  a linear transformation, and

$$B = \left\{ \vec{b}_{1}, \vec{b}_{2}, \dots, \vec{b}_{r}, \vec{b}_{r+1}, \vec{b}_{r+2}, \dots, \vec{b}_{n} \right\}$$

a basis of V with the property that  $\left\{\vec{b}_{r+1},\vec{b}_{r+2},\ldots,\vec{b}_n\right\}$  is a basis of ker(T). Then

$$\left\{T(\vec{b}_1), T(\vec{b}_2), \dots, T(\vec{b}_r)\right\}$$

is a basis of im(T), and therefore r = rank(T).

# Remark (How is this useful?)

Suppose V and W are vector spaces and  $T: V \to W$  is a linear transformation. If you find a basis of  $\ker(T)$ , then this may be used to find a basis of  $\operatorname{im}(T)$ : extend the basis of  $\ker(T)$  to a basis of V; applying the transformation T to each of the vectors that was added to the basis of  $\ker(T)$  produces a set of vectors that is a basis of  $\operatorname{im}(T)$ .

Problem

Let 
$$A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$
, and let  $T : \mathbf{M}_{22} \to \mathbf{M}_{22}$  be a linear transformation defined by

$$T(X) = XA - AX$$
 for all  $X \in \mathbf{M}_{22}$ .

Find a basis of ker(T) and a basis of im(T).

#### Solution

First note that by the Dimension Theorem,

$$\dim(\ker(T))+\dim(\operatorname{im}(T))=\dim(\boldsymbol{M}_{22})=4.$$

Let 
$$X = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$
. Then

$$T(X) = AX - XA$$

$$= \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} - \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} c & d \\ a & b \end{bmatrix} - \begin{bmatrix} b & a \\ d & c \end{bmatrix} = \begin{bmatrix} c - b & d - a \\ a - d & b - c \end{bmatrix}$$

# Solution (continued)

If  $X \in \ker(T)$ , then  $T(X) = \vec{0}_{22}$  so

$$c - b = 0, d - a = 0, a - d = 0, b - c = 0.$$

This system of four equations in four variables has general solution a=s, b=t, c=t, d=s for  $s,t\in\mathbb{R}$ . Therefore,

$$\ker(T) = \left\{ \left[ \begin{array}{cc} s & t \\ t & s \end{array} \right] \ \middle| \ s,t \in \mathbb{R} \right\} = \operatorname{span} \left\{ \left[ \begin{array}{cc} 1 & 0 \\ 0 & 1 \end{array} \right], \left[ \begin{array}{cc} 0 & 1 \\ 1 & 0 \end{array} \right] \right\}.$$

Let  $B_k = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \right\}$ . Since  $B_k$  is independent and spans Rer(T),  $B_k$  is a basis of ker(T).

To find a basis of im(T), extend the basis of ker(T) to a basis of  $M_{22}$ : here is one such basis

$$\left\{\left[\begin{array}{cc} 1 & 0 \\ 0 & 1 \end{array}\right], \left[\begin{array}{cc} 0 & 1 \\ 1 & 0 \end{array}\right], \left[\begin{array}{cc} 1 & 0 \\ 0 & 0 \end{array}\right], \left[\begin{array}{cc} 0 & 1 \\ 0 & 0 \end{array}\right]\right\}.$$

# Solution (continued)

Thus

$$B_i = \left\{ T \left[ \begin{array}{cc} 1 & 0 \\ 0 & 0 \end{array} \right], T \left[ \begin{array}{cc} 0 & 1 \\ 1 & 0 \end{array} \right] \right\} = \left\{ \left[ \begin{array}{cc} 0 & -1 \\ 1 & 0 \end{array} \right], \left[ \begin{array}{cc} 1 & 0 \\ 0 & -1 \end{array} \right] \right\}$$

is a basis of im(T).

