

# Math 254: Introduction to Linear Algebra

## Notes #3.1 — Image & Kernel of a Linear Transform

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## Outline

- 1 Student Learning Objectives
  - SLOs: Image & Kernel of a Linear Transform
- 2 Subspaces of  $\mathbb{R}^n$  and Their Dimensions
  - Image & Kernel of a Linear Transformation
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## SLOs 3.1

## Image &amp; Kernel of a Linear Transform

After this lecture you should:

- Be able to **identify the *image*** of a linear transformation (and its associated matrix) —  $\text{im}(A)$
- Be able to **identify the *kernel*** of a linear transformation (and its associated matrix) —  $\text{ker}(A)$
- Know what the ***span*** of a set of vectors is.
- Know when  $\text{ker}(A) = \{\vec{0}\}$ ? — and the implications [THE CHARACTERISTICS OF INVERTIBLE MATRICES]



Fair Warning



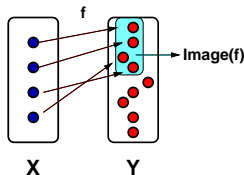
Things get quite “math-y” starting now.

## Image of a Linear Transformation

## Definition (Image of a Function (Linear Transformation))

The **image** of a function consists of all the values the function takes in its target space. If  $f : X \mapsto Y$ , then

$$\begin{aligned}\text{image}(f) &= \{ f(x) : x \in X \} \\ &= \{ b \in Y : b = f(x), \text{ for some } x \in X \}.\end{aligned}$$



**Figure:**  $X$  is the *domain* of  $f$ ;  $Y$  the *target space* of  $f$ ; and the shaded subset of  $Y$  is the *image* of  $f$ .

Fig: © 2019 Peter Blomgren

## Notational Hazard!



## Notational Warning: “Range”

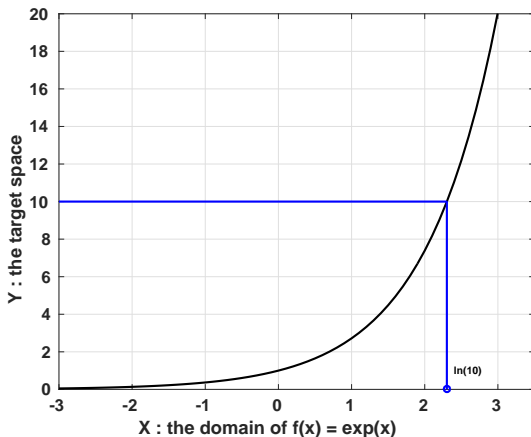
Sometimes you see the term **range** in the literature; and depending on who is speaking (writing), it may refer to what we call the *image*, or (occasionally) the entire *target space*.

**(In most literature range and image are the same.)**



Example  $e^x : \mathbb{R} \mapsto \mathbb{R}$ 

[NOT A LINEAR TRANSFORMATION]



**Figure:** The image of  $f(x) = e^x$  from  $\mathbb{R}$  to  $\mathbb{R}$  consists of  $\mathbb{R}^+$  (all positive real numbers). Every positive number  $b \in \mathbb{R}^+$  can be written as  $b = e^{\ln(b)} = f(\ln(b))$ .

[Figure: Copyright © 2019 Peter Blomgren]

Example:  $f(t) = [\cos(t) \quad \sin(t)]^T$

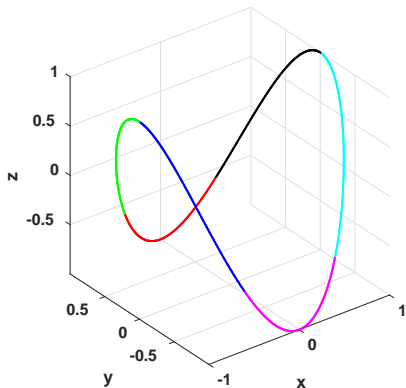
[NOT A LINEAR TRANSFORMATION]



**Figure:** The image of  $f(t) = \begin{bmatrix} \cos(t) \\ \sin(t) \end{bmatrix}$  from  $\mathbb{R}$  to  $\mathbb{R}^2$  consists of the unit circle centered at the origin;  $f$  is called the *parametrization* of the unit circle.

[Figure: Copyright © 2019 Peter Blomgren]

Example:  $f(t) = [\cos(t) \quad \sin(t) \quad \cos(2t)]^T$  [NOT A LINEAR TRANSFORMATION]



**Figure:** The image of  $f(t) = \begin{bmatrix} \cos(t) \\ \sin(t) \\ \cos(2t) \end{bmatrix}$  from  $\mathbb{R}$  to  $\mathbb{R}^3$  consists of the figure above. (Here projected from 3D to 2D for your viewing pleasure!)

[Figure: Copyright © 2019 Peter Blomgren]



## Image of an Invertible Function

### Image of an Invertible Function

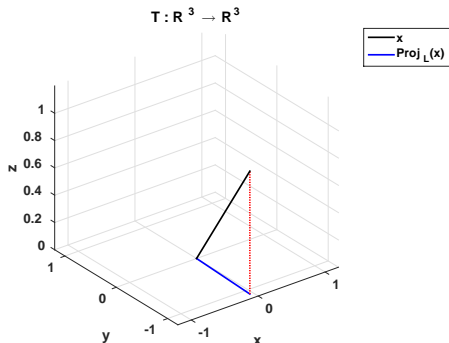
- If the function  $f : X \mapsto Y$  is *invertible*, then the image of  $f$  is (all of)  $Y$ . " $\forall b \in Y \exists x \in X : b = f(x)$ ."
- In this case  $x = f^{-1}(b)$ :

$$b = f(f^{-1}(b))$$

See also [NOTES#2.4].

Image of the Projection onto the  $x$ - $y$ -Plane

Consider  $T : \mathbb{R}^3 \mapsto \mathbb{R}^3$  that projects a vector  $\vec{v}$  orthogonally onto the  $x$ - $y$ -plane:



**Figure:** The image of  $T$  is the  $x$ - $y$ -plane in  $\mathbb{R}^3$ , consisting of all vectors of the form

$$\begin{bmatrix} x \\ y \\ 0 \end{bmatrix}.$$

[Figure: Copyright © 2019 Peter Blomgren]

$$T \left( \begin{bmatrix} x \\ y \\ z \end{bmatrix} \right) = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix} \Leftrightarrow T(\vec{x}) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix}$$

$$T(\vec{x}) = A\vec{x}$$

$$\mathbb{R}^2 \mapsto \mathbb{R}^2$$

Consider  $T(\vec{x}) = A\vec{x}$ , with  $\vec{x} \in \mathbb{R}^2$ , and  $A \in \mathbb{R}^{2 \times 2}$ , where

$$A = \begin{bmatrix} 1 & 3 \\ 2 & 6 \end{bmatrix}.$$

The image is described by

$$\begin{bmatrix} 1 & 3 \\ 2 & 6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + x_2 \begin{bmatrix} 3 \\ 6 \end{bmatrix} = (x_1 + 3x_2) \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

which is the line of all scalings of  $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$ .



**Figure:** The unit circle in domain space  $X = \mathbb{R}^2$ , and the image of the unit circle in target space  $Y = \mathbb{R}^2$ . Note: we can fill  $X = \mathbb{R}^2$  with circles of radii  $r \in [0, \infty)$ , so the image of  $T$  can be described by all scalings of the *image* of the unit circle; since  $T(k\vec{x}) = Ak\vec{x} = kA\vec{x} = kT(\vec{x})$ .

[Figure: Copyright © 2019 Peter Blomgren]

$$T(\vec{x}) = A\vec{x}$$

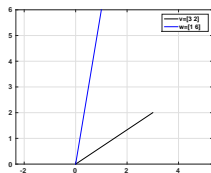
$$\mathbb{R}^2 \mapsto \mathbb{R}^2$$

Consider  $T(\vec{x}) = A\vec{x}$ , with  $\vec{x} \in \mathbb{R}^2$ , and  $A \in \mathbb{R}^{2 \times 2}$ , where

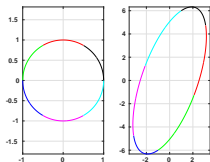
$$A = \begin{bmatrix} 1 & 3 \\ 6 & 2 \end{bmatrix}.$$

The image is described by

$$\begin{bmatrix} 1 & 3 \\ 6 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 6 \end{bmatrix} + x_2 \begin{bmatrix} 3 \\ 2 \end{bmatrix}$$



which fills out all of  $\mathbb{R}^2$  since  $\begin{bmatrix} 1 \\ 6 \end{bmatrix}$  and  $\begin{bmatrix} 3 \\ 2 \end{bmatrix}$  are not parallel.



**Figure:** The unit circle in domain space  $X = \mathbb{R}^2$ , and the image of the unit circle in target space  $Y = \mathbb{R}^2$ . Note: we can fill  $X = \mathbb{R}^2$  with circles of radii  $r \in [0, \infty)$ , so the image of  $T$  can be described by all scalings of the *image* of the unit circle; since  $T(k\vec{x}) = Ak\vec{x} = kA\vec{x} = kT(\vec{x})$ .

[Figure: Copyright © 2019 Peter Blomgren]

## The Span

## Describing the Linear Transformation

## Definition (The Span)

Consider the vectors  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m \in \mathbb{R}^n$ . The set of all linear combinations

$$c_1 \vec{v}_1 + c_2 \vec{v}_2 + \cdots + c_m \vec{v}_m, \quad c_1, \dots, c_m \in \mathbb{R}$$

of the vectors is called their **span**:

$$\text{span}(\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m) = \{c_1 \vec{v}_1 + c_2 \vec{v}_2 + \cdots + c_m \vec{v}_m : c_1, c_2, \dots, c_m \in \mathbb{R}\}.$$

Image of a Linear Transformation — Image of  $A$  / Column Space of  $A$ 

## Theorem (Image of a Linear Transformation)

*The image of a linear transformation  $T(\vec{x}) = A\vec{x}$  is the span of the column vectors of  $A$ . We denote the image of  $T$  by  $\text{im}(T)$  or  $\text{im}(A)$ .*



Notational Hazard (Language)



Since  $\text{im}(A)$  is the span of the columns of  $A$ , it is sometimes referred to as the **Column Space** of  $A$ , denoted  $C(A)$  [GS5–3.1].

## Describing the Linear Transformation

The theorem pretty much proves itself; it follows directly from how we multiply vectors and matrices:

$$T(\vec{x}) = A\vec{x} = \left[ \begin{array}{c|c|c} | & & | \\ \vec{v}_1 & \dots & \vec{v}_m \\ | & & | \end{array} \right] \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix} = x_1 \vec{v}_1 + \dots + x_m \vec{v}_m.$$

The vertical bars are there to illustrate that we are expressing the matrix column-wise, using the  $\vec{v}$ -vectors.

Properties of  $\text{im}(T)$ 

## Theorem (Properties of the Image)

The image of a linear transformation  $T : \mathbb{R}^m \mapsto \mathbb{R}^n$  has the following properties:

- The zero vector  $\vec{0}$  in  $\mathbb{R}^n$  is in the image of  $T$ .
- The image of  $T$  is **closed under addition**: if  $\vec{v}_1$  and  $\vec{v}_2$  are in the image of  $T$ , then so is  $\vec{v}_1 + \vec{v}_2$ .
- The image of  $T$  is **closed under scalar multiplication**: if  $\vec{v} \in \text{im}(T)$  and  $k \in \mathbb{R}$ , then  $k\vec{v} \in \text{im}(T)$ .



[PROOF IN THE SUPPLEMENTAL SLIDES]



[Proof] Properties of  $\text{im}(T)$ 

[FOCUS :: MATH]

Properties of the Image.

a.  $\vec{0} = A\vec{0} = T(\vec{0})$ .

b.  $\exists \vec{w}_1, \vec{w}_2 \in \mathbb{R}^m: \vec{v}_1 = T(\vec{w}_1), \vec{v}_2 = T(\vec{w}_2)$ . Then  
$$\vec{v}_1 + \vec{v}_2 = T(\vec{w}_1) + T(\vec{w}_2) \stackrel{\text{L.T.}}{=} T(\vec{w}_1 + \vec{w}_2) \Rightarrow$$
$$(\vec{v}_1 + \vec{v}_2) \in \text{im}(T).$$

c. If  $\vec{v} = T(\vec{w})$ , then  $k\vec{v} = kT(\vec{w}) \stackrel{\text{L.T.}}{=} T(k\vec{w}) \Rightarrow k\vec{v} \in \text{im}(T)$ .

[b.] + [c.]  $\Rightarrow \text{im}(T)$  is closed under linear combinations.

[Parsing the Proof] Properties of  $\text{im}(T)$ 

[FOCUS :: MATH]

Properties of the Image.

- a.  $\vec{0} = A\vec{0} = T(\vec{0})$ .
- This follows straight from how we compute matrix-vector products; given  $A \in \mathbb{R}^{n \times m}$ , and  $T(\vec{x}) = A\vec{x}$ , we immediately get  $A\vec{0}_m = \vec{0}_n$ , where the subscript on the  $\vec{0}$ -vector indicates its number of components.
- b.  $\exists \vec{w}_1, \vec{w}_2 \in \mathbb{R}^m: \vec{v}_1 = T(\vec{w}_1), \vec{v}_2 = T(\vec{w}_2)$ .
- Since  $\vec{v}_1$  and  $\vec{v}_2$  are in the image; there must exist (“ $\exists$ ”) vectors  $\vec{w}_1$ , and  $\vec{w}_2$  so that  $\vec{v}_1 = T(\vec{w}_1), \vec{v}_2 = T(\vec{w}_2)$  {some input must generate the output!}

Then  $(\vec{v}_1 + \vec{v}_2) \stackrel{1}{=} T(\vec{w}_1) + T(\vec{w}_2) \stackrel{\text{L.T.}}{=} T(\vec{w}_1 + \vec{w}_2) \Rightarrow (\vec{v}_1 + \vec{v}_2) \in \text{im}(T)$ .

- First we write the vector we want to show is in the image ( $\vec{v}_1 + \vec{v}_2$ ); then
- (“ $\stackrel{1}{=}$ ”) we use the fact that each vector is in the image; followed by
- (“ $\stackrel{\text{L.T.}}{=}$ ”) the fact that  $T$  is a linear transformation; and we can conclude
- (“ $\Rightarrow$ ”) that we wrote  $(\vec{v}_1 + \vec{v}_2)$  as the linear transformation of some vector  $\vec{w}^* = (\vec{w}_1 + \vec{w}_2)$ , which makes  $\vec{v}^* = (\vec{v}_1 + \vec{v}_2)$  a member of the image.



[Parsing the Proof] Properties of  $\text{im}(T)$ 

[FOCUS :: MATH]

Properties of the Image.

c. If  $\vec{v} = T(\vec{w})$ , then  $k\vec{v} = kT(\vec{w}) \stackrel{\text{L.T.}}{=} T(k\vec{w}) \Rightarrow k\vec{v} \in \text{im}(T)$ .

- This is very similar to part ., given a vector  $\vec{v}$  in the image; there must be a vector  $\vec{w}$  in the domain, so that  $\vec{v} = T(\vec{w})$
- We want to show that  $k\vec{v}$  is in the image; so we use  $k\vec{v} = kT(\vec{w})$ ,
- then the fact that  $T$  is a linear transformation:  $kT(\vec{w}) = T(k\vec{w})$ ;
- and conclude as in part b.



## The Kernel of a Linear Transformation

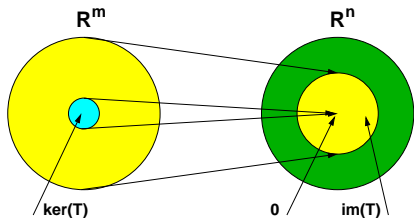
### Definition (Kernel / Null Space)

The **kernel** (aka “*null space*”) of a linear transformation  $T(\vec{x}) = A\vec{x}$  from  $\mathbb{R}^m$  to  $\mathbb{R}^n$  consists of all zeros of the transformation; that is, the solutions of the equation  $T(\vec{x}) = A\vec{x} = \vec{0}$ .

In other words, the kernel of  $T$  is the solution of the set of linear equations

$$A\vec{x} = \vec{0}$$

We denote the kernel of  $T$  by  $\ker(T)$  or  $\ker(A)$ .



**Figure:**  $\ker(T)$  are the elements in the domain that are transformed to  $0$  in the output space; the rest of the domain “paints”  $\text{im}(T)$ . Notice that there may be element of the output space that are NOT part of  $\text{im}(T)$ .

[Figure: Copyright © 2019 Peter Blomgren]

$$T : \mathbb{R}^m \mapsto \mathbb{R}^n$$

$$\text{im}(T) \subset \mathbb{R}^n$$

$$\ker(T) \subset \mathbb{R}^m$$

For the linear transformation  $T : \mathbb{R}^m \mapsto \mathbb{R}^n$ ,

- $\text{im}(T) = \{T(\vec{x}) : \vec{x} \in \mathbb{R}^m\}$  is a subset of the *target space*  $\mathbb{R}^n$  of  $T$ ;
- $\ker(T) = \{\vec{x} \in \mathbb{R}^m : T(\vec{x}) = \vec{0}\}$  is a subset of the *domain*.



Notational Hazard (Language)



[GS5–3.2] uses the notation  $N(A)$  for the null space (and [GS5–3.1]  $C(A)$  for the image / column space). We will use  $\ker(A)$  and  $\text{im}(A)$  exclusively.

A more common notational variant for the kernel is  $\text{null}(A)$ .

Example:  $\mathbb{R}^3 \mapsto \mathbb{R}^3$ Projection onto the  $x$ - $y$  plane in  $\mathbb{R}^3$ 

Consider, again, the linear transformation:

$$T\left(\begin{bmatrix} x \\ y \\ z \end{bmatrix}\right) = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix} \Leftrightarrow T(\vec{x}) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} x \\ y \\ 0 \end{bmatrix}$$

Clearly,

$$T\left(\begin{bmatrix} 0 \\ 0 \\ z \end{bmatrix}\right) = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \quad \forall z \in \mathbb{R}.$$

Therefore,

$$\ker(T) = \left\{ \begin{bmatrix} 0 \\ 0 \\ z \end{bmatrix} : \forall z \in \mathbb{R} \right\}, \quad \text{also } \text{im}(T) = \left\{ \begin{bmatrix} x \\ y \\ 0 \end{bmatrix} : \forall x, y \in \mathbb{R} \right\}.$$

Example:  $\mathbb{R}^5 \mapsto \mathbb{R}^4$ Find  $\ker(A)$ Consider  $T(\vec{x}) = A\vec{x}$ , where

$$A = \begin{bmatrix} 1 & 2 & 2 & -5 & 6 \\ -1 & -2 & -1 & 1 & -1 \\ 4 & 8 & 5 & -8 & 9 \\ 3 & 6 & 1 & 5 & -7 \end{bmatrix}$$

Let's find the *kernel* (solve  $A\vec{x} = \vec{0}$ )

$$\left[ \begin{array}{ccccc|c} 1 & 2 & 2 & -5 & 6 & 0 \\ -1 & -2 & -1 & 1 & -1 & 0 \\ 4 & 8 & 5 & -8 & 9 & 0 \\ 3 & 6 & 1 & 5 & -7 & 0 \end{array} \right]$$

Example:  $\mathbb{R}^5 \mapsto \mathbb{R}^4$  $\text{rref}([A|\vec{b}])$ 

$$\left[ \begin{array}{ccccc|c} 1 & 2 & 2 & -5 & 6 & 0 \\ -1 & -2 & -1 & 1 & -1 & 0 \\ 4 & 8 & 5 & -8 & 9 & 0 \\ 3 & 6 & 1 & 5 & -7 & 0 \end{array} \right]$$

$$\left[ \begin{array}{ccccc|c} 1 & 2 & 2 & -5 & 6 & 0 \\ 0 & 0 & 1 & -4 & 5 & 0 \\ 0 & 0 & -3 & 12 & -15 & 0 \\ 0 & 0 & -5 & 20 & -25 & 0 \end{array} \right]$$

$$\left[ \begin{array}{ccccc|c} 1 & 2 & 0 & 3 & -4 & 0 \\ 0 & 0 & 1 & -4 & 5 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right]$$



Example:  $\mathbb{R}^5 \mapsto \mathbb{R}^4$ 

$$\left[ \begin{array}{ccccc|c} \textcircled{1} & 2 & 0 & 3 & -4 & 0 \\ 0 & 0 & \textcircled{1} & -4 & 5 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right]$$

$$\text{rank}(A) = 2$$

$$\text{number-of-leading-variables} = 2$$

$$\text{number-of-free-variables} = 3$$

Now, the equations

$$\begin{cases} x_1 = -2x_2 - 3x_4 + 4x_5 \\ x_3 = 4x_4 - 5x_5 \end{cases}$$

describe the kernel. As usual we let  $\{x_2 = s, x_4 = t, x_5 = u\}$ , and write:

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} -2s - 3t + 4u \\ s \\ 4t - 5u \\ t \\ u \end{bmatrix} = s \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} -3 \\ 0 \\ 4 \\ 1 \\ 0 \end{bmatrix} + u \begin{bmatrix} 4 \\ 0 \\ -5 \\ 0 \\ 1 \end{bmatrix}$$

Example:  $\mathbb{R}^5 \mapsto \mathbb{R}^4$

Given that the parameters,  $\{s, t, u\}$  are allowed to independently vary over  $(-\infty, \infty)$ , we are interested in all combinations of the 3 vectors...

Using the previously defined concept of *span*, we write

$$\ker(T) = \text{span} \left\{ \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 4 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 \\ 0 \\ -5 \\ 0 \\ 1 \end{bmatrix} \right\}$$

## Properties of the Kernel

Theorem (Some Properties of the Kernel)

Consider the linear transform  $T : \mathbb{R}^m \mapsto \mathbb{R}^n$ ,

- The zero vector  $\vec{0}$  in  $\mathbb{R}^m$  is in  $\ker(T)$ .
- The kernel is **closed under addition**.
- The kernel is **closed under scalar multiplication**.

The proofs for these properties are small modifications of the proofs of the analogous properties for the Image (SEE THE EXTENDED NOTES)... and are left as an exercise.

When is  $\ker(A) = \{\vec{0}\}$ ?

Theorem (When is  $\ker(A) = \{\vec{0}\}$ ?)

- Consider an  $(n \times m)$  matrix  $A$ . Then  $\ker(A) = \{\vec{0}\}$  if and only if  $\text{rank}(A) = m$ .
- Consider an  $(n \times m)$  matrix  $A$ . If  $\ker(A) = \{\vec{0}\}$ , then  $m \leq n$ . Equivalently, if  $m > n$ , then there are non-zero vectors in the kernel of  $A$ .
- For a square matrix  $A$ , we have  $\ker(A) = \{\vec{0}\}$  if and only if  $A$  is **invertible**.

## Characteristics of Invertible Matrices

IMPORTANT!

## Equivalent Statements: Invertible Matrices

For an  $(n \times n)$  matrix  $A$ , the following statements are equivalent; that is for a given  $A$ , they are either all true or all false:

- i.  $A$  is invertible ( $\exists A^{-1}$ )
- ii. The linear system  $A\vec{x} = \vec{b}$  has a unique solution  $\vec{x}$ ,  $\forall \vec{b} \in \mathbb{R}^n$
- ii.  $\text{rref}(A) = I_n$
- iv.  $\text{rank}(A) = n$
- v.  $\text{im}(A) = \mathbb{R}^n$
- vi.  $\text{ker}(A) = \{\vec{0}\}$

We will add to this list throughout the semester: [NOTES#2.4]<sup>✓</sup>, [NOTES#3.3], and [NOTES#7.1].

## Suggested Problems 3.1

**Available on Learning Glass videos:**

3.1 — 1, 7, 11, 14, 15, 17, 23, 24, 29, 39

## Lecture – Book Roadmap

Lecture	Book, [GS5–]
3.1	§3.1, §3.2, §3.3
3.2	§3.1, §3.2, §3.3, §3.4
3.3	§3.1, §3.2, §3.3, §3.4, §3.5
3.4	

## Metacognitive Exercise — Thinking About Thinking &amp; Learning

I know / learned	Almost there	Huh?!?
Right After Lecture		
After Thinking / Office Hours / SI-session		
After Reviewing for Quiz/Midterm/Final		



(3.1.1), (3.1.7), (3.1.11)

**(3.1.1)** Find vectors that span the *kernel* of

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

**(3.1.7)** Find vectors that span the *kernel* of

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 3 & 2 \\ 3 & 2 & 1 \end{bmatrix}$$

**(3.1.11)** Find vectors that span the *kernel* of

$$A = \begin{bmatrix} 1 & 0 & 2 & 4 \\ 0 & 1 & -3 & -1 \\ 3 & 4 & -6 & 8 \\ 0 & -1 & 3 & 4 \end{bmatrix}$$

(3.1.14), (3.1.15), (3.1.17)

**(3.1.14)** Find vectors that span the *image* of

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 2 & 3 \\ 1 & 2 & 3 \end{bmatrix}$$

**(3.1.15)** Find vectors that span the *image* of

$$A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 \end{bmatrix}$$

**(3.1.17)** Describe the *image* of the transformation  $T(\vec{x}) = A\vec{x}$  geometrically (e.g. as a line, a plane, etc. in  $\mathbb{R}^2$  or  $\mathbb{R}^3$ .)

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

(3.1.23), (3.1.24)

**(3.1.23)** Describe the *image* and *kernel* of the transformation  $T(\vec{x}) = A\vec{x}$  geometrically, where

$$T(\vec{x}) = \left\{ \begin{array}{c} \text{REFLECTION ABOUT THE LINE} \\ \{y = x/3\} \text{ IN } \mathbb{R}^2 \end{array} \right\}.$$

**(3.1.24)** Describe the *image* and *kernel* of the transformation  $T(\vec{x}) = A\vec{x}$  geometrically, where

$$T(\vec{x}) = \left\{ \begin{array}{c} \text{ORTHOGONAL PROJECTION ONTO} \\ \text{THE PLANE } \{x + 2y + 3z = 0\} \text{ IN } \mathbb{R}^3 \end{array} \right\}.$$

(3.1.29), (3.1.39)

**(3.1.29)** Give an example of a function whose image is the unit sphere

$$\mathbb{S}^2 = \{x^2 + y^2 + z^2 = 1\} \text{ in } \mathbb{R}^3.$$

**(3.1.39)** Consider a square matrix  $A$ :

- What is the relationship among  $\ker(A)$  and  $\ker(A^2)$ ? Are they necessarily equal?? Is one of them necessarily contained in the other? More generally what can you say about  $\ker(A)$ ,  $\ker(A^2)$ ,  $\ker(A^3)$ , ...?
- What can you say about  $\text{im}(A)$ ,  $\text{im}(A^2)$ ,  $\text{im}(A^3)$ , ...?