

# [C2-001] 기초수학

## Lecture 04: Matrix Inverse

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# Topics

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- Matrix Inverse

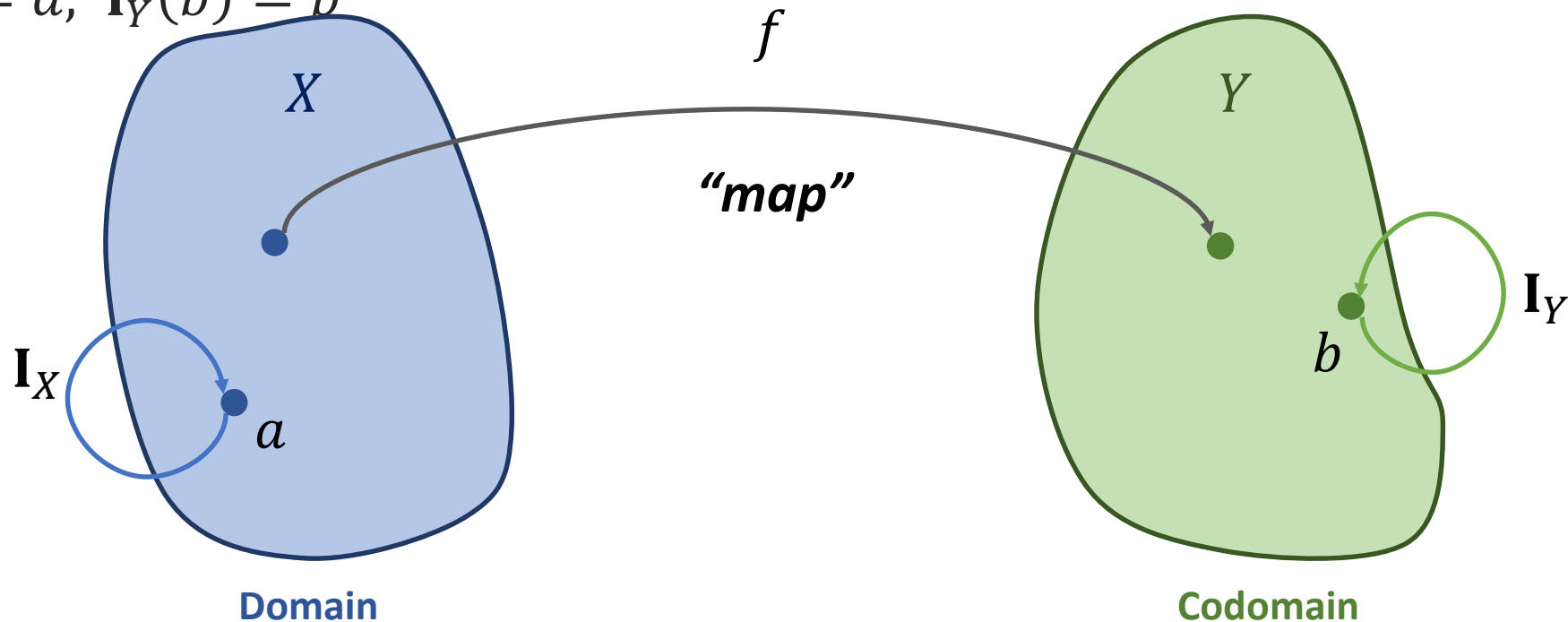
# Introduction: The Inverse of a Function

- **Identity Function**

- An identity function (identity relation; map; identity transformation) is a function that always returns the same value that was used as its argument

- $\mathbf{I}_X: X \rightarrow Y, a \in X, b \in Y,$

- $\mathbf{I}_X(a) = a, \mathbf{I}_Y(b) = b$



# Introduction: The Inverse of a Function

- **Inverse of a Function**

- An inverse function is a function that reverse another function.
- If the function  $f$  applied to an input  $x$  gives a result of  $y$ , then applying its inverse function  $g$  to  $y$  gives the result  $x$ .
  - $f(x) = y \iff g(y) = f^{-1}(y) = x$
- $f: X \rightarrow Y$  is invertible  $\iff$  there exists a function,  $f^{-1}: Y \rightarrow X$  such that  $f^{-1} \circ f = \mathbf{I}_X$  and  $f \circ f^{-1} = \mathbf{I}_Y$

# Introduction: The Inverse of a Function

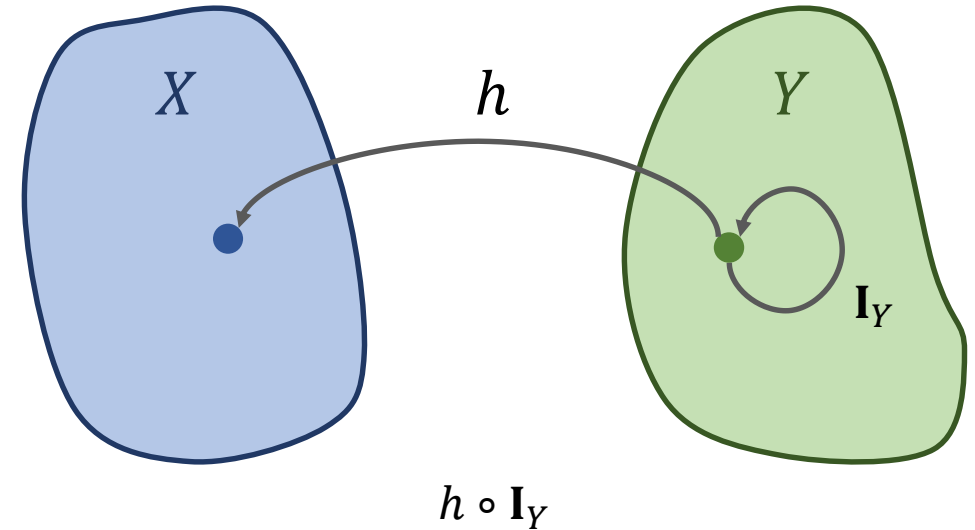
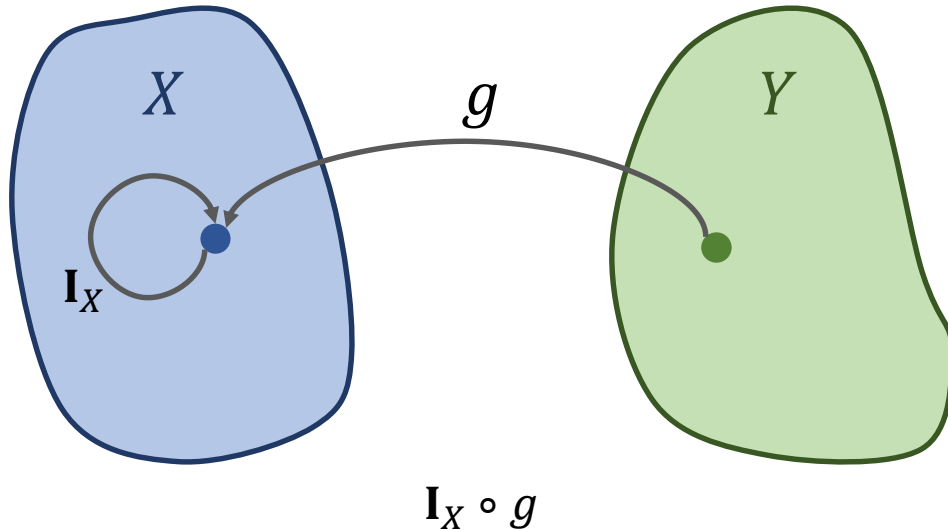
- Inverse of a Function

- Q)  $f$  is invertible, then is  $f^{-1}$  unique?

- A)  $g: Y \rightarrow X$ ,  $g \circ f = \mathbf{I}_X$  and  $f \circ g = \mathbf{I}_Y$

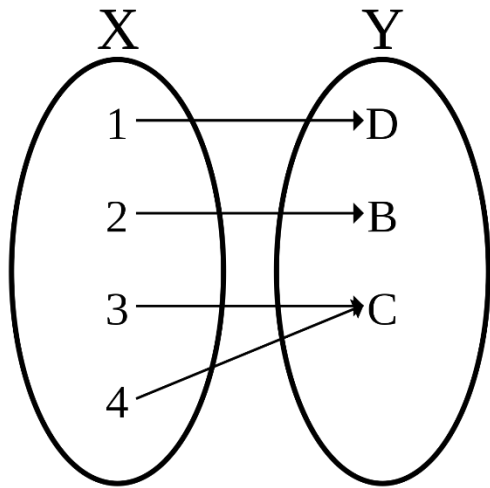
$$h: Y \rightarrow X, \quad h \circ f = \mathbf{I}_X \text{ and } f \circ h = \mathbf{I}_Y$$

$$\rightarrow g = \mathbf{I}_X \circ g = (h \circ f) \circ g = h \circ (f \circ g) = h \circ \mathbf{I}_Y = h$$



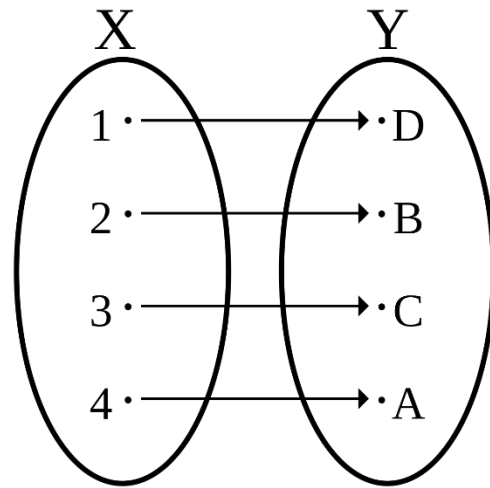
# Surjective Function (Onto)

- A surjective function (onto function) is a function  $f$  that maps an element  $x$  to every element  $y$ .
- If  $f: X \rightarrow Y$ , then  $f$  is said to be *surjective* if  $\forall y$  (every  $y$ )  $\in Y$ ,  
 $\exists$  *at least one*  $x \in X$  (there exists at least one  $x$ ) such that  $f(x) = y$



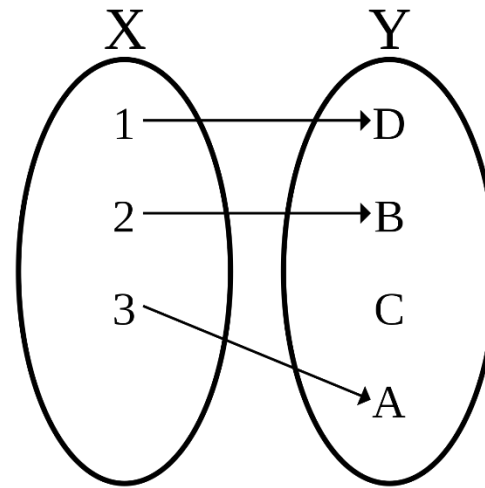
(a)

Surjective



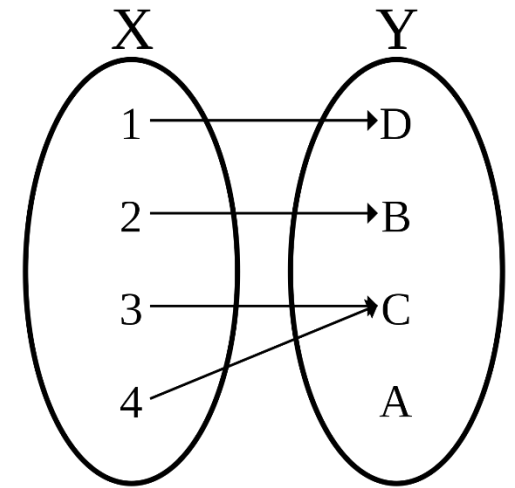
(b)

Surjective



(c)

Non-surjective

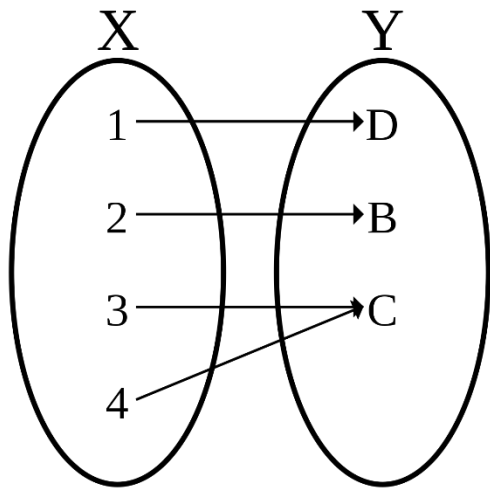


(d)

Non-surjective

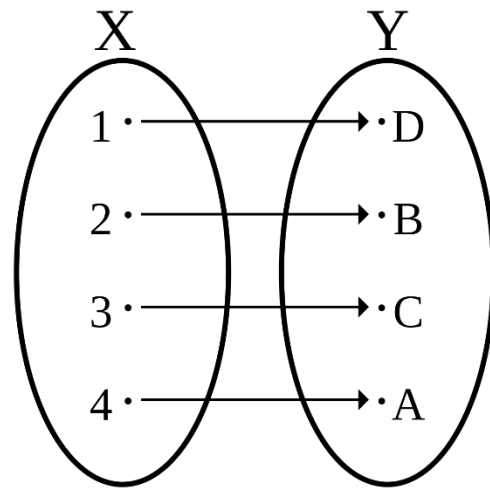
# Injective Function (One-to-one)

- A injective function (one-to-one function) is a function  $f$  that maps an distinct element  $x$  to distinct element  $y$ .
- If  $f: X \rightarrow Y$ , then  $f$  is said to be *injective* if  $\forall y$  (every  $y$ )  $\in Y$ ,  
 $\exists$  *at most one*  $x \in X$  such that  $f(x) = y$  (i.e.,  $a \neq b \Leftrightarrow f(a) \neq f(b)$ )



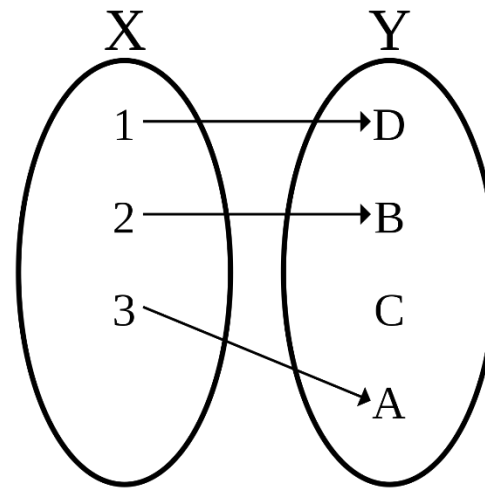
(a)

**Non-injective**



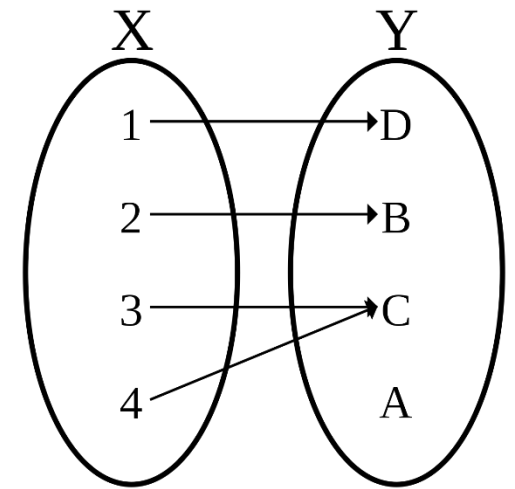
(b)

**Injective**



(c)

**Injective**



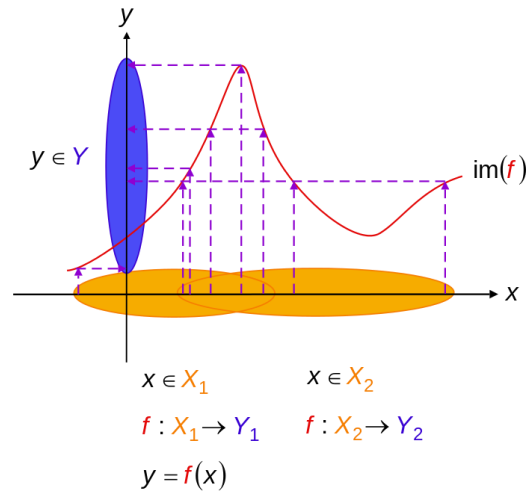
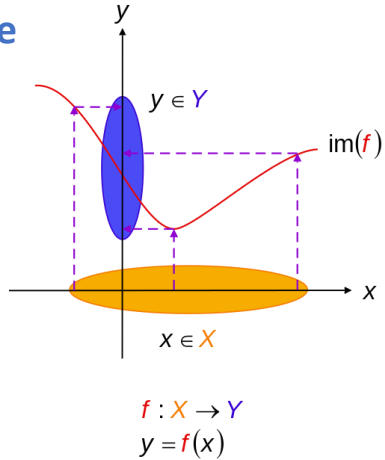
(d)

**Non-injective**

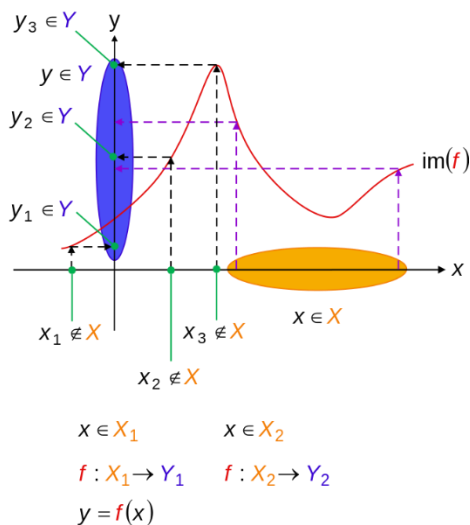
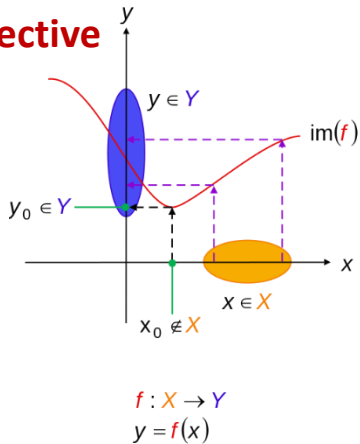
# Surjective & Injective Functions

- Surjective Function

Surjective

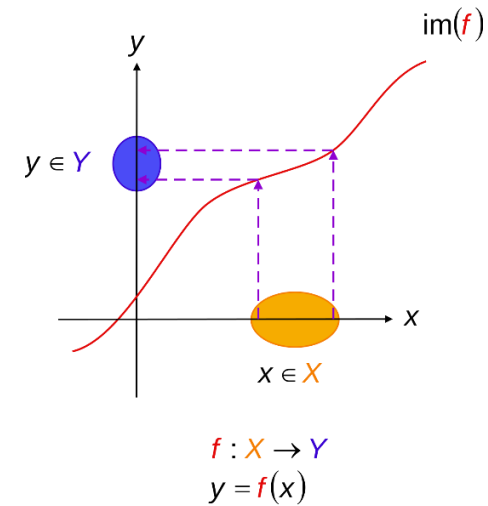


Non-surjective

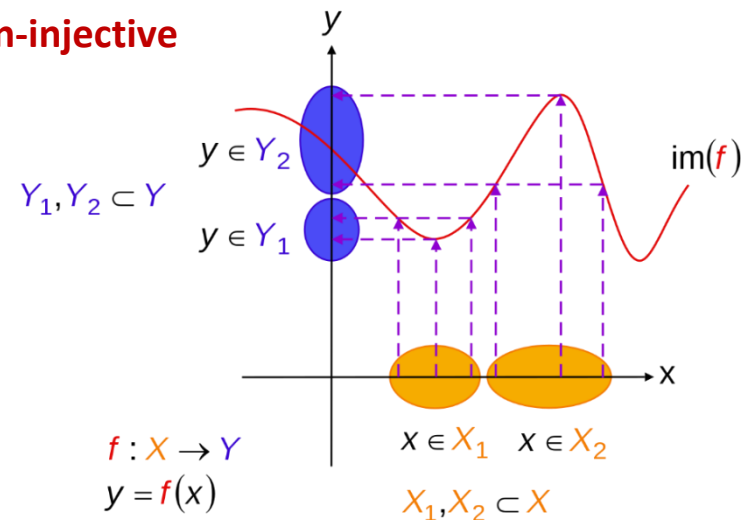


- Injective Function

Injective



Non-injective



[https://en.wikipedia.org/wiki/Surjective\\_function](https://en.wikipedia.org/wiki/Surjective_function)  
[https://en.wikipedia.org/wiki/Injective\\_function](https://en.wikipedia.org/wiki/Injective_function)

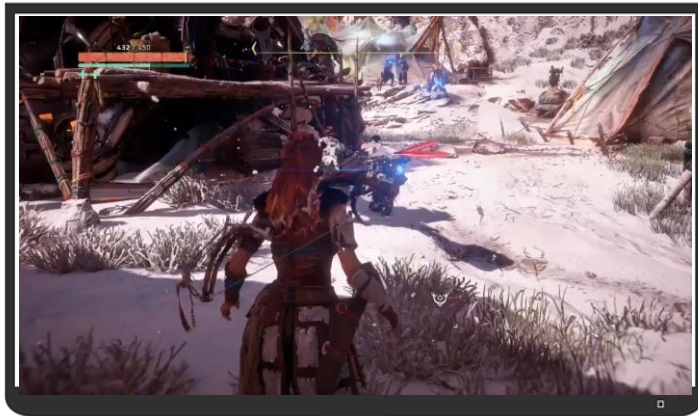


# Examples of Surjective & Injective Functions

- Example of Surjective Function



3D modeling



Projection  
onto 2D monitor

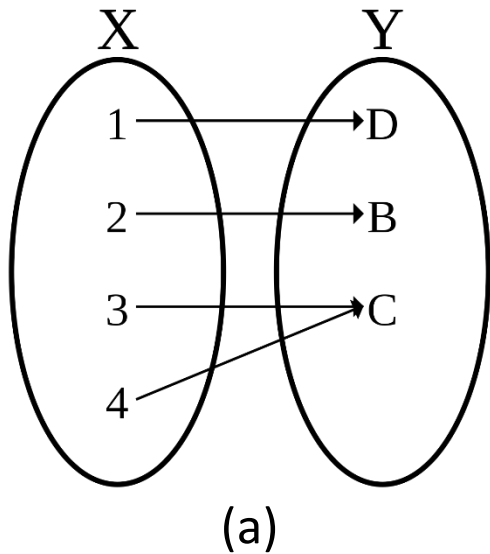
- Example of Injective Function



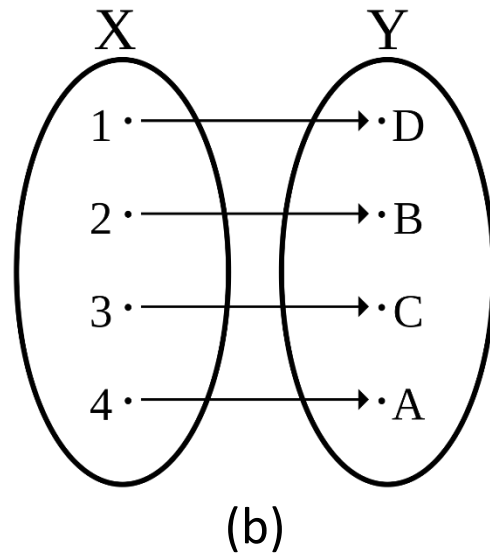
A character – Each item

# Relating Invertibility to being Onto and One-to-one

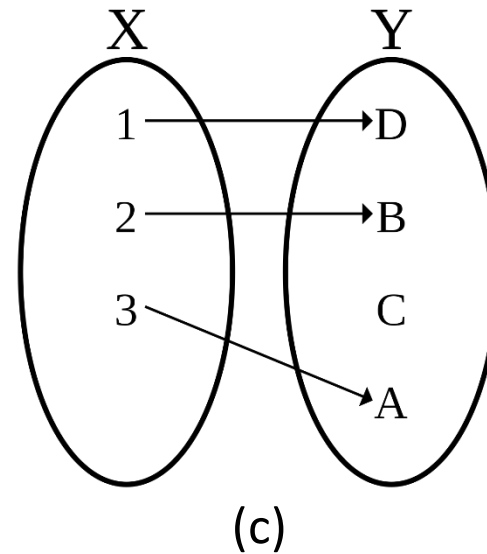
- $f$  is invertible  $\Leftrightarrow$  for **every**  $y \in Y$ , there exists a **unique**  $x \in X$  such that  $f(x) = y$ 
  - For **every**  $y \in Y$  : **Surjective (Onto)**
  - A **unique**  $x \in X$  : **Injective (One-to-one)**



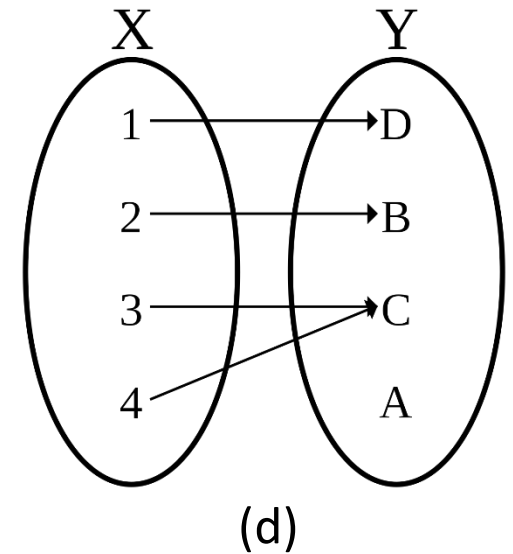
**Non-invertible**



**Invertible**



**Non-invertible**



**Non-invertible**

# Invertibility of Linear Transformation

- Matrix Condition for Onto Transformation
  - $\mathcal{T}(\mathbf{x}) = \mathbf{A}\mathbf{x} = x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + \cdots + x_n\mathbf{v}_n \rightarrow \text{Span}(\mathbf{v}_1, \cdots, \mathbf{v}_n) = C(\mathbf{A}) = \mathbb{R}^m$
  - $\mathcal{T}$  is onto  $\Leftrightarrow C(\mathbf{A}) = \mathbb{R}^m \rightarrow rref(\mathbf{A})$  has a pivot entry in every row
    - 1)  $m$  pivot entries
    - 2)  $rank(\mathbf{A}) = m$  ( $rank(\mathbf{A}) = \dim(C(\mathbf{A})) = \# \text{ of basis for } C(\mathbf{A})$ )
- $\mathcal{T}: \mathbb{R}^2 \rightarrow \mathbb{R}^3, \mathcal{T}(\mathbf{x}) = \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 \\ 0 & 2 \\ 0 & 4 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 0 & 4 \end{bmatrix} \rightarrow \begin{bmatrix} \mathbf{1} & 0 \\ 0 & \mathbf{1} \\ 0 & 0 \end{bmatrix}$
- $rank(\mathcal{T}) = 2 \neq 3 (\because \mathbb{R}^3) \rightarrow \mathcal{T}$  is not onto.  $\rightarrow \mathcal{T}$  is not invertible.

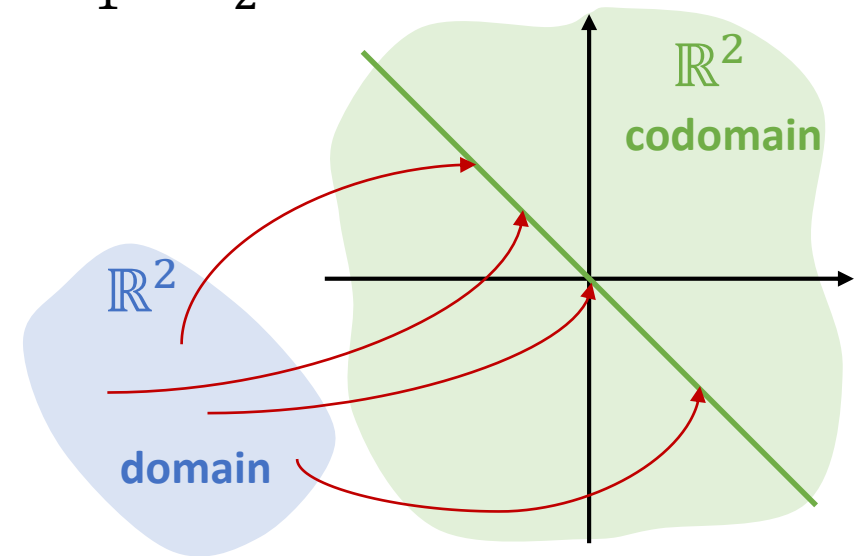
# Invertibility of Linear Transformation

- Exploring the Particular Solution Set of  $A\mathbf{x}=\mathbf{b}$

- $\mathcal{T}: \mathbb{R}^2 \rightarrow \mathbb{R}^2, \mathcal{T}(\mathbf{x}) = \begin{bmatrix} 1 & -3 \\ -1 & 3 \end{bmatrix} \mathbf{x} \rightarrow \mathbf{Ax} = \mathbf{b}$

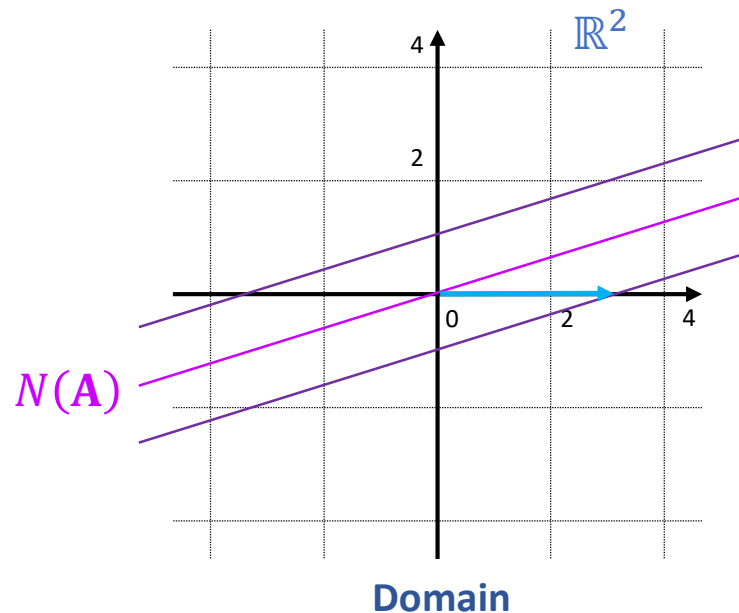
$$\begin{bmatrix} 1 & -3 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \rightarrow \left[ \begin{array}{cc|c} 1 & -3 & b_1 \\ -1 & 3 & b_2 \end{array} \right] \rightarrow \left[ \begin{array}{cc|c} 1 & -3 & b_1 \\ 0 & 0 & b_1 + b_2 \end{array} \right]: \text{Non-surjective}$$

- $\mathcal{T}$  is not surjective because  $\text{Im}(\mathcal{T})$  is not the entire codomain.
- Only members  $\mathbf{b} \in \mathbb{R}^2$  that have solutions are the ones  $b_1 + b_2 = 0$ .
- Constraint:  $x_1 - 3x_2 = b_1 \rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 3 \\ 1 \end{bmatrix}$

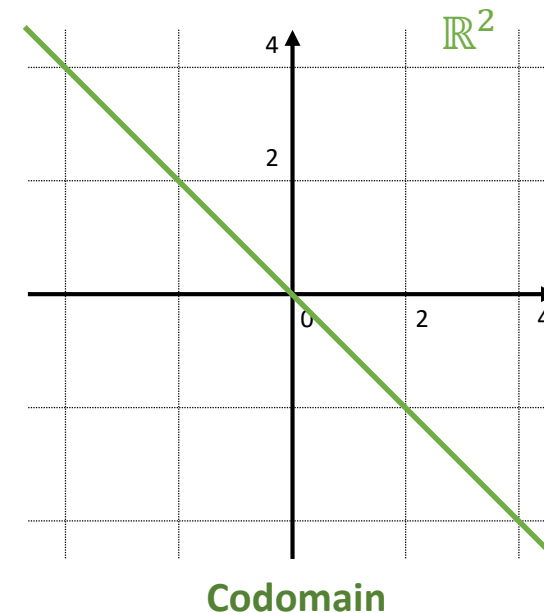


# Invertibility of Linear Transformation

- Exploring the Particular Solution Set of  $\mathbf{Ax} = \mathbf{b}$ 
  - For a particular  $\mathbf{b}$  that has a solution  $\mathbf{Ax} = \mathbf{b}$ , the solution set is  $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 3 \\ 1 \end{bmatrix}$
  - Assuming that  $\mathbf{Ax} = \mathbf{b}$  has a solution: the solution set  $\{\mathbf{x}_p + \mathbf{n} \mid \mathbf{n} = N(\mathbf{A})\}$ .
  - One-to-one  $\rightarrow$  At most one solution:  $N(\mathbf{A})$  has to just have the zero vector,  $\mathbf{0}$



$$\begin{aligned} &\mathcal{T} \\ &: \mathbb{R}^2 \rightarrow \mathbb{R}^2 \\ &\mathbf{x} \rightarrow \mathbf{Ax} = \mathbf{b} \end{aligned}$$



# Invertibility of Linear Transformation

- Matrix Condition for One-to-one Transformation
  - Any solution to the inhomogeneous system,  $\mathbf{Ax} = \mathbf{b}$ , will take the form,  $\mathbf{x} = \mathbf{x}_p + \mathbf{x}_h$
  - $\mathbf{Ax} = \mathbf{0} \rightarrow [\mathbf{A} \mid \mathbf{0}] \rightarrow [\text{rref}(\mathbf{A}) \mid \mathbf{0}]$   
 $\rightarrow \mathbf{x}_h = x_1 \mathbf{v}_1 + x_2 \mathbf{v}_2 + \cdots + x_n \mathbf{v}_n = N(\mathbf{A}) = \text{Span}(\mathbf{v}_1, \mathbf{v}_2, \cdots, \mathbf{v}_n)$
  - $\mathbf{Ax} = \mathbf{b} \rightarrow [\mathbf{A} \mid \mathbf{b}] \rightarrow [\text{rref}(\mathbf{A}) \mid \mathbf{b}']$   
 $\rightarrow \mathbf{x} = \mathbf{b}' + \mathbf{0} = \mathbf{b}' + x_1 \mathbf{v}_1 + x_2 \mathbf{v}_2 + \cdots + x_n \mathbf{v}_n = \mathbf{x}_p + \mathbf{x}_h$
- $\mathcal{T}: X \rightarrow Y$ , if  $\mathcal{T}$  is one-to-one transformation,  $\forall \mathbf{b} \in Y, \exists$  at most one solution to  $\mathbf{Ax} = \mathbf{b}$   
 $\rightarrow \mathbf{x} = \mathbf{x}_p + \mathbf{x}_h$  can be only one solution  $\rightarrow \mathbf{x}_h = N(\mathbf{A}) = x_1 \mathbf{v}_1 + \cdots + x_n \mathbf{v}_n = \{\mathbf{0}\}$   
 $\rightarrow x_1 = x_2 = \cdots = x_n = 0 : \text{Linearly independent}$   
 $\rightarrow \mathbf{v}_1, \mathbf{v}_2, \cdots, \mathbf{v}_n : \text{Basis for } C(\mathbf{A}) \rightarrow \therefore \dim(C(\mathbf{A})) = \text{rank}(\mathbf{A}) = n$

# Invertibility of Linear Transformation

- Invertibility of Linear Transformation
  - Onto (Surjective)  $\Leftarrow \text{rank}(\mathbf{A}) = m$
  - One-to-one (Injective)  $\Leftarrow \text{rank}(\mathbf{A}) = n$
  - Onto & One-to-one (invertible)  $\Rightarrow \text{rank}(\mathbf{A}) = m = n \Rightarrow n \times n$  square mtx

- $\mathbf{A}_{n \times n} = [\mathbf{v}_1 \ \mathbf{v}_2 \ \cdots \ \mathbf{v}_n]_{n \times n} \rightarrow \text{rref}(\mathbf{A}) = \begin{bmatrix} 1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & 1 \end{bmatrix}_{n \times n} = \mathbf{I}_n$

- $\therefore \mathcal{T}: \mathbb{R}^n \rightarrow \mathbb{R}^n, \mathcal{T}(\mathbf{x}) = \mathbf{A}\mathbf{x}, \mathcal{T}$  is only invertible if  $\text{rref}(\mathbf{A}) = \mathbf{I}_n$

# Finding Inverse

- Example of Finding Inverse Matrix,  $\mathbf{A}^{-1}$

$$\bullet \mathbf{A} = \begin{bmatrix} 1 & -1 & -1 \\ -1 & 2 & 3 \\ 1 & 1 & 4 \end{bmatrix} \xrightarrow{\mathcal{T}_1} \begin{bmatrix} 1 & -1 & -1 \\ 0 & 1 & 2 \\ 0 & 2 & 5 \end{bmatrix} \xrightarrow{\mathcal{T}_2} \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{bmatrix} \xrightarrow{\mathcal{T}_3} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\bullet \mathcal{T}_1 \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} x_1 \\ x_1 + x_2 \\ x_3 - x_1 \end{bmatrix}, \quad \mathcal{T}_2 \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} x_1 + x_2 \\ x_2 \\ x_3 - 2x_2 \end{bmatrix}, \quad \mathcal{T}_3 \left( \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \begin{bmatrix} x_1 - x_3 \\ x_2 - 2x_3 \\ x_3 \end{bmatrix}$$

$$\bullet \mathcal{T}_1(\mathbf{x}) = \mathbf{S}_1\mathbf{x}, \quad \mathcal{T}_2(\mathbf{x}) = \mathbf{S}_2\mathbf{x}, \quad \mathcal{T}_3(\mathbf{x}) = \mathbf{S}_3\mathbf{x}$$

$$\bullet \mathbf{S}_1 = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}, \quad \mathbf{S}_2 = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & -2 & 1 \end{bmatrix}, \quad \mathbf{S}_3 = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & -2 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\bullet [\mathbf{A} \mid \mathbf{I}] \rightarrow [\mathbf{S}_1\mathbf{A} \mid \mathbf{S}_1\mathbf{I}] \rightarrow [\mathbf{S}_2\mathbf{S}_1\mathbf{A} \mid \mathbf{S}_2\mathbf{S}_1\mathbf{I}] \\ \rightarrow [\mathbf{S}_3\mathbf{S}_2\mathbf{S}_1\mathbf{A} \mid \mathbf{S}_3\mathbf{S}_2\mathbf{S}_1\mathbf{I}] = [\mathbf{I} \mid \mathbf{A}^{-1}]$$



# Finding Inverse

- Example of Finding Inverse Matrix,  $\mathbf{A}^{-1}$

- $\mathbf{A} = \begin{bmatrix} 1 & -1 & -1 \\ -1 & 2 & 3 \\ 1 & 1 & 4 \end{bmatrix} \rightarrow \mathbf{A}^{-1} = ?$

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

$$\left[ \begin{array}{ccc|ccc} 1 & 0 & 1 & 2 & 1 & 0 \\ 0 & 1 & 2 & 1 & 1 & 0 \\ 0 & 0 & 1 & -3 & -2 & 1 \end{array} \right] \rightarrow \left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & 5 & 3 & -1 \\ 0 & 1 & 0 & 7 & 5 & 2 \\ 0 & 0 & 1 & -3 & -2 & 1 \end{array} \right] = [\text{rref}(\mathbf{A}) \mid \mathbf{A}^{-1}]$$

# Determinant

- 2 x 2 Determinant

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

- $\left[ \begin{array}{cc|cc} a & b & 1 & 0 \\ c & d & 0 & 1 \end{array} \right] \xrightarrow{\mathcal{T}_1} \left[ \begin{array}{cc|cc} a & b & 1 & 0 \\ 0 & ad-bc & -c & a \end{array} \right] \xrightarrow{\mathcal{T}_2} \left[ \begin{array}{cc|cc} (ad-bc)a & 0 & ad & -ab \\ 0 & ad-bc & -c & a \end{array} \right]$

- $\xrightarrow{\mathcal{T}_3} \left[ \begin{array}{cc|cc} 1 & 0 & \frac{ad}{(ad-bc)a} & \frac{-ab}{(ad-bc)a} \\ 0 & 1 & \frac{-c}{(ad-bc)} & \frac{a}{(ad-bc)} \end{array} \right] \rightarrow \therefore \mathbf{A}^{-1} = \begin{bmatrix} \frac{d}{ad-bc} & \frac{-b}{ad-bc} \\ \frac{-c}{ad-bc} & \frac{a}{ad-bc} \end{bmatrix} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$

- Definition of Determinant

- $ad - bc \neq 0 \iff \mathbf{A}$  is invertible

- $\det(\mathbf{A}) = |\mathbf{A}| = \begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$

- $\therefore \mathbf{A}^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} = \frac{1}{\det(\mathbf{A})} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$

# Determinant

- 3 x 3 Determinant

- $\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \rightarrow \det(\mathbf{A}) = a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$

- Example of Finding 3 x 3 Determinant

- $\mathbf{B} = \begin{bmatrix} 1 & 2 & 4 \\ 2 & -1 & 3 \\ 4 & 0 & 1 \end{bmatrix} \rightarrow \det(\mathbf{B}) = 1 \begin{vmatrix} -1 & 3 \\ 0 & 1 \end{vmatrix} - 2 \begin{vmatrix} 2 & 3 \\ 4 & 1 \end{vmatrix} + 4 \begin{vmatrix} 2 & -1 \\ 4 & 0 \end{vmatrix}$   
 $= (-1 - 0) - 2(2 - 12) + 4(0 + 4) = 35: \text{Invertible}$

# Determinant

- $n \times n$  Determinant

- $\mathbf{A}_{n \times n} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix}$ ,  $\tilde{\mathbf{A}}_{ij} = (n-1) \times (n-1)$  matrix

- $\det(\mathbf{A}) = a_{11}|\tilde{\mathbf{A}}_{11}| - a_{12}|\tilde{\mathbf{A}}_{12}| + a_{13}|\tilde{\mathbf{A}}_{13}| - \cdots \pm a_{1n}|\tilde{\mathbf{A}}_{1n}|$   
 $= a_{11}\det(\tilde{\mathbf{A}}_{11}) - a_{12}\det(\tilde{\mathbf{A}}_{12}) + a_{13}\det(\tilde{\mathbf{A}}_{13}) - \cdots \pm a_{1n}\det(\tilde{\mathbf{A}}_{1n})$

- Example

- $B = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 0 & 2 & 0 \\ 0 & 1 & 2 & 3 \\ 2 & 3 & 0 & 0 \end{bmatrix} = 1 \begin{vmatrix} 0 & 2 & 0 \\ 1 & 2 & 3 \\ 3 & 0 & 0 \end{vmatrix} - 2 \begin{vmatrix} 1 & 2 & 0 \\ 0 & 2 & 3 \\ 2 & 0 & 0 \end{vmatrix} + 3 \begin{vmatrix} 1 & 0 & 0 \\ 0 & 1 & 3 \\ 2 & 3 & 0 \end{vmatrix} - 4 \begin{vmatrix} 1 & 0 & 2 \\ 0 & 1 & 2 \\ 2 & 3 & 0 \end{vmatrix}$

- $= 1 \cdot (-2 \cdot (-9)) - 2 \cdot (-2 \cdot (-6)) + 3 \cdot (1 \cdot (-9)) - 4 \cdot (1 \cdot (-6) + 2 \cdot (-2)) = 7$

# Determinant

- Determinant along Other Rows/Columns

$$\bullet B = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 0 & 2 & 0 \\ 0 & 1 & 2 & 3 \\ \color{blue}{2} & \color{blue}{3} & \color{blue}{0} & \color{blue}{0} \end{bmatrix} = \color{red}{-}\color{blue}{2} \begin{vmatrix} 2 & 3 & 4 \\ 0 & 2 & 0 \\ 1 & 2 & 3 \end{vmatrix} + \color{blue}{3} \begin{vmatrix} 1 & 3 & 4 \\ 1 & 2 & 0 \\ 0 & 2 & 3 \end{vmatrix} - \color{red}{0} \begin{vmatrix} 1 & 2 & 4 \\ 1 & 0 & 0 \\ 0 & 1 & 3 \end{vmatrix} + \color{blue}{0} \begin{vmatrix} 1 & 2 & 3 \\ 1 & 0 & 2 \\ 0 & 1 & 2 \end{vmatrix}$$

$$\bullet = -2 \cdot (2 \cdot 2) + 3 \cdot (-2 \cdot (-4) + 3 \cdot (-1)) = -8 + 3 \cdot (8 - 3) = 7$$

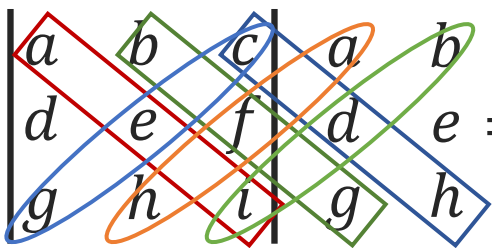
- Sign of Determinant

$$\bullet \text{sign}(i, j) = (-1)^{(i+j)}, \text{ (e.g.,) } 4 \times 4: \begin{bmatrix} \color{blue}{+} & \color{red}{-} & \color{blue}{+} & \color{red}{-} \\ \color{red}{-} & \color{blue}{+} & \color{red}{-} & \color{blue}{+} \\ \color{blue}{+} & \color{red}{-} & \color{blue}{+} & \color{red}{-} \\ \color{red}{-} & \color{blue}{+} & \color{red}{-} & \color{blue}{+} \end{bmatrix}$$

# Determinant

- Rule of Sarrus for Determinant

$$\begin{aligned}
 \bullet \begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} &= a \begin{vmatrix} e & f \\ h & i \end{vmatrix} - b \begin{vmatrix} d & f \\ g & i \end{vmatrix} + c \begin{vmatrix} d & e \\ g & h \end{vmatrix} \\
 &= a(ei - fh) - b(di - fg) + c(dh - eg) \\
 &= \textcolor{red}{a} \textcolor{green}{e} \textcolor{blue}{i} + \textcolor{green}{b} \textcolor{blue}{f} \textcolor{red}{g} + \textcolor{blue}{c} \textcolor{red}{d} \textcolor{green}{h} - \textcolor{red}{a} \textcolor{blue}{f} \textcolor{green}{h} - \textcolor{green}{b} \textcolor{red}{d} \textcolor{blue}{i} - \textcolor{blue}{c} \textcolor{green}{e} \textcolor{red}{g}
 \end{aligned}$$

$$\bullet \begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} \begin{matrix} a & b \\ d & e \\ g & h \end{matrix} \rightarrow \begin{vmatrix} \textcolor{red}{a} & \textcolor{green}{b} & \textcolor{blue}{c} \\ \textcolor{blue}{d} & \textcolor{red}{e} & \textcolor{green}{f} \\ \textcolor{green}{g} & \textcolor{blue}{h} & \textcolor{red}{i} \end{vmatrix} = \textcolor{red}{a} \textcolor{green}{e} \textcolor{blue}{i} + \textcolor{green}{b} \textcolor{blue}{f} \textcolor{red}{g} + \textcolor{blue}{c} \textcolor{red}{d} \textcolor{green}{h} - \textcolor{red}{a} \textcolor{blue}{f} \textcolor{green}{h} - \textcolor{green}{b} \textcolor{red}{d} \textcolor{blue}{i} - \textcolor{blue}{c} \textcolor{green}{e} \textcolor{red}{g}$$


# More Determinant

- Determinant When Row Multiplied by Scalar

- $\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc = \det(\mathbf{A}), \begin{vmatrix} a & b \\ kc & kd \end{vmatrix} = k(ad - bc) = k \cdot \det(\mathbf{A}) = \det(\mathbf{A}')$

- $k\mathbf{A} = k \begin{bmatrix} a & b \\ c & d \end{bmatrix} \rightarrow \det(k\mathbf{A}) = \begin{vmatrix} ka & kb \\ kc & kd \end{vmatrix} = k^2(ad - bc) = k^2 \det(\mathbf{A})$

- $\begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} = -d \begin{vmatrix} b & c \\ h & i \end{vmatrix} + e \begin{vmatrix} a & c \\ g & i \end{vmatrix} - f \begin{vmatrix} a & b \\ g & h \end{vmatrix} = \det(\mathbf{B})$

- $\begin{vmatrix} a & b & c \\ kd & ke & kf \\ g & h & i \end{vmatrix} = -kd \begin{vmatrix} b & c \\ h & i \end{vmatrix} + ke \begin{vmatrix} a & c \\ g & i \end{vmatrix} - kf \begin{vmatrix} a & b \\ g & h \end{vmatrix} = k \det(\mathbf{B}) = \det(\mathbf{B}')$

- $\therefore \det(\mathbf{A}') = k \det(\mathbf{A}), \det(k\mathbf{A}) = k^n \det(\mathbf{A})$

# More Determinant

- Determinant When Row is Added

$$\bullet \mathbf{X} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \vdots & \vdots & \cdots & \vdots \\ x_1 & x_2 & \cdots & x_n \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix}, \mathbf{Y} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \vdots & \vdots & \cdots & \vdots \\ y_1 & y_2 & \cdots & y_n \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix}, \mathbf{Z} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \vdots & \vdots & \cdots & \vdots \\ x_1 + y_1 & x_2 + y_2 & \cdots & x_n + y_n \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix}$$

$$\bullet \det(\mathbf{X}) = \sum_{j=1}^n (-1)^{(i+j)} x_j |\tilde{\mathbf{A}}_{ij}|$$

$$\bullet \det(\mathbf{Y}) = \sum_{j=1}^n (-1)^{(i+j)} y_j |\tilde{\mathbf{A}}_{ij}|$$

$$\bullet \det(\mathbf{Z}) = \sum_{j=1}^n (-1)^{(i+j)} (x_j + y_j) |\tilde{\mathbf{A}}_{ij}| = \det(\mathbf{X}) + \det(\mathbf{Y})$$



# More Determinant

- Example of Determinant When Row is Added

- $\mathbf{X} = \begin{bmatrix} a & b \\ x_1 & x_2 \end{bmatrix}, \mathbf{Y} = \begin{bmatrix} a & b \\ y_1 & y_2 \end{bmatrix}, \mathbf{Z} = \begin{bmatrix} a & b \\ x_1 + y_1 & x_2 + y_2 \end{bmatrix}$

- $\det(\mathbf{X}) = ax_2 - bx_1$

- $\det(\mathbf{Y}) = ay_2 - by_1$

- $\det(\mathbf{Z}) = a(x_2 + y_2) - b(x_1 + y_1) = ax_2 - bx_1 + ay_2 - by_1 = \det(\mathbf{X}) + \det(\mathbf{Y})$

# More Determinant

- Determinant with Duplicate Row

$$\bullet \mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \vdots & \vdots & \cdots & \vdots \\ a_{i1} & a_{i2} & \cdots & a_{in} \\ \vdots & \vdots & \vdots & \vdots \\ a_{j1} & a_{j2} & \cdots & a_{jn} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} = \begin{bmatrix} \mathbf{r}_1 \\ \vdots \\ \mathbf{r}_i \\ \vdots \\ \mathbf{r}_j \\ \vdots \\ \mathbf{r}_n \end{bmatrix} \rightarrow \mathbf{S}_{ij} = \begin{bmatrix} \mathbf{r}_1 \\ \vdots \\ \mathbf{r}_j \\ \vdots \\ \mathbf{r}_i \\ \vdots \\ \mathbf{r}_n \end{bmatrix} \rightarrow \det(\mathbf{S}_{ij}) = -\det(\mathbf{A})$$

- If  $\mathbf{r}_i = \mathbf{r}_j$ ,  $\mathbf{S}_{ij} = \mathbf{A} \rightarrow \det(\mathbf{S}_{ij}) = \det(\mathbf{A}) = -\det(\mathbf{A}) = 0$  : **Not invertible**
- Matrix  $\mathbf{A}$  is invertible  $\Leftrightarrow rref(\mathbf{A}) = \mathbf{I}_n$
- Duplicate row  $\rightarrow$  Never get  $rref(\mathbf{A}) = \mathbf{I}_n \rightarrow$  Not invertible  $\rightarrow \det(\mathbf{A}) = 0$

# More Determinant

- Determinant After Row Operations

$$\bullet \mathbf{A} = \begin{bmatrix} \mathbf{r}_1 \\ \vdots \\ \mathbf{r}_i \\ \vdots \\ \mathbf{r}_j \\ \vdots \\ \mathbf{r}_n \end{bmatrix}, \mathbf{B} = \begin{bmatrix} \mathbf{r}_1 \\ \vdots \\ \mathbf{r}_i \\ \vdots \\ \mathbf{r}_j - c\mathbf{r}_i \\ \vdots \\ \mathbf{r}_n \end{bmatrix} \rightarrow \det(\mathbf{B}) = \begin{vmatrix} \mathbf{r}_1 \\ \vdots \\ \mathbf{r}_i \\ \vdots \\ \mathbf{r}_j \\ \vdots \\ \mathbf{r}_n \end{vmatrix} + \begin{vmatrix} \mathbf{r}_1 \\ \vdots \\ \mathbf{r}_i \\ \vdots \\ -c\mathbf{r}_i \\ \vdots \\ \mathbf{r}_n \end{vmatrix} = \begin{vmatrix} \mathbf{r}_1 \\ \vdots \\ \mathbf{r}_i \\ \vdots \\ \mathbf{r}_j \\ \vdots \\ \mathbf{r}_n \end{vmatrix} - c \begin{vmatrix} \mathbf{r}_1 \\ \vdots \\ \mathbf{r}_i \\ \vdots \\ \mathbf{r}_i \\ \vdots \\ \mathbf{r}_n \end{vmatrix} = \det(\mathbf{A})$$

# More Determinant

- Upper/Lower Triangular Determinant

- $\mathbf{A} = \begin{bmatrix} a & b \\ 0 & d \end{bmatrix} \rightarrow \det(\mathbf{A}) = ad$

- $\mathbf{B} = \begin{bmatrix} a & b & c \\ 0 & e & f \\ 0 & 0 & i \end{bmatrix} \rightarrow \det(\mathbf{B}) = a \begin{vmatrix} e & f \\ 0 & i \end{vmatrix} - b \begin{vmatrix} 0 & f \\ 0 & i \end{vmatrix} + c \begin{vmatrix} 0 & e \\ 0 & 0 \end{vmatrix} = aei$

- $$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ 0 & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_{nn} \end{bmatrix} \rightarrow \det(\mathbf{A}) a_{11} \begin{vmatrix} a_{22} & a_{23} & \cdots & a_{2n} \\ 0 & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_{nn} \end{vmatrix} = a_{11} a_{22} \begin{vmatrix} a_{33} & a_{34} & \cdots & a_{3n} \\ 0 & a_{44} & \cdots & a_{4n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_{nn} \end{vmatrix}$$
$$= a_{11} a_{22} \cdots a_{nn}$$

# More Determinant

- Simpler 4 x 4 Determinant

- Our goal is to make a given matrix the upper triangular matrix form.

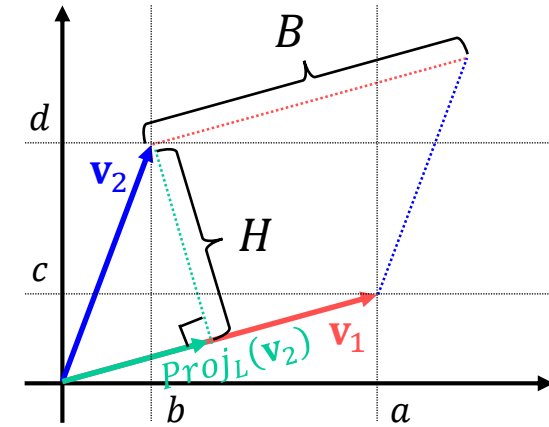
- $\mathbf{A} = \begin{bmatrix} 1 & 2 & 2 & 1 \\ 1 & 2 & 4 & 2 \\ 2 & 7 & 5 & 2 \\ -1 & 4 & -6 & 3 \end{bmatrix}$

- $\det(\mathbf{A}) = \begin{vmatrix} 1 & 2 & 2 & 1 \\ 1 & 2 & 4 & 2 \\ 2 & 7 & 5 & 2 \\ -1 & 4 & -6 & 3 \end{vmatrix} \rightarrow \begin{vmatrix} 1 & 2 & 2 & 1 \\ 0 & 0 & 2 & 1 \\ 0 & 3 & 1 & 0 \\ 0 & 6 & -4 & 4 \end{vmatrix} \rightarrow - \begin{vmatrix} 1 & 2 & 2 & 1 \\ 0 & 3 & 1 & 0 \\ 0 & 0 & 2 & 1 \\ 0 & 6 & -4 & 4 \end{vmatrix}$   
 $\rightarrow - \begin{vmatrix} 1 & 2 & 2 & 1 \\ 0 & 3 & 1 & 0 \\ 0 & 0 & 2 & 1 \\ 0 & 0 & -6 & 4 \end{vmatrix} \rightarrow - \begin{vmatrix} \mathbf{1} & 2 & 2 & 1 \\ 0 & \mathbf{3} & 1 & 0 \\ 0 & 0 & \mathbf{2} & 1 \\ 0 & 0 & 0 & \mathbf{7} \end{vmatrix} = -42$

# More Determinant: Area of a Parallelogram

- Determinant and Area of a Parallelogram

- $\mathbf{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} = [\mathbf{v}_1 \quad \mathbf{v}_2]$ ,  $\mathbf{v}_1 = \begin{bmatrix} a \\ c \end{bmatrix}$  and  $\mathbf{v}_2 = \begin{bmatrix} b \\ d \end{bmatrix}$



- Area of a parallelogram

- $B = \|\mathbf{v}_1\| \rightarrow B^2 = \|\mathbf{v}_1\|^2 = \mathbf{v}_1 \cdot \mathbf{v}_1$

- $H^2 = \|\mathbf{v}_2\|^2 - \|\text{Proj}_L(\mathbf{v}_2)\|^2 = \mathbf{v}_2 \cdot \mathbf{v}_2 - \left\| \frac{\mathbf{v}_2 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \mathbf{v}_1 \right\|^2 = \mathbf{v}_2 \cdot \mathbf{v}_2 - \left( \frac{(\mathbf{v}_2 \cdot \mathbf{v}_1)(\mathbf{v}_2 \cdot \mathbf{v}_1)}{(\mathbf{v}_1 \cdot \mathbf{v}_1)(\mathbf{v}_1 \cdot \mathbf{v}_1)} \mathbf{v}_1 \cdot \mathbf{v}_1 \right)$

- $S = BH \rightarrow S^2 = B^2 H^2 = (\mathbf{v}_1 \cdot \mathbf{v}_1) \left( \mathbf{v}_2 \cdot \mathbf{v}_2 - \left( \frac{(\mathbf{v}_2 \cdot \mathbf{v}_1)^2}{\mathbf{v}_1 \cdot \mathbf{v}_1} \right) \right) = (\mathbf{v}_1 \cdot \mathbf{v}_1)(\mathbf{v}_2 \cdot \mathbf{v}_2) - (\mathbf{v}_2 \cdot \mathbf{v}_1)^2$   
 $= (a^2 + c^2)(b^2 + d^2) - (ab + cd)^2 = a^2b^2 + a^2d^2 + c^2b^2 + c^2d^2 - (a^2b^2 + 2abcd + c^2d^2)$   
 $= a^2d^2 - 2abcd + c^2b^2 = (ad - bc)^2 = (\det(\mathbf{A}))^2$

- $\therefore S = |\det(\mathbf{A})|$

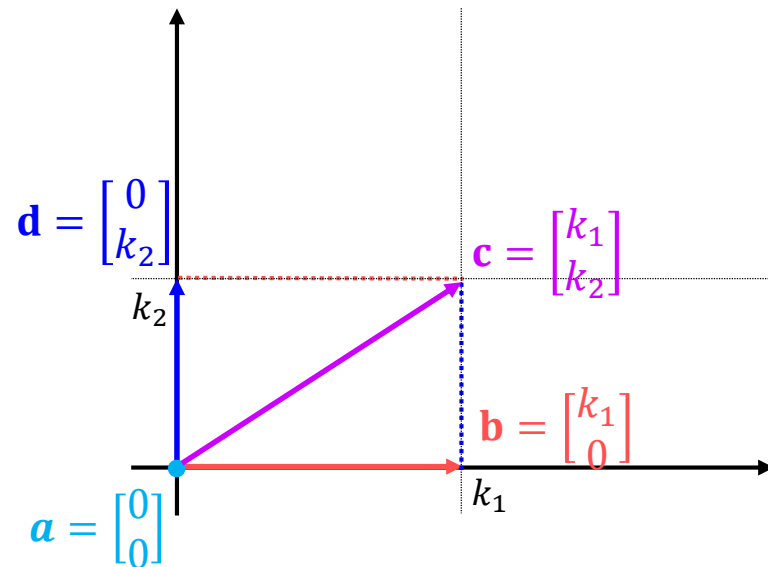
# More Determinant: Scaling

- Determinant as Scaling Factor

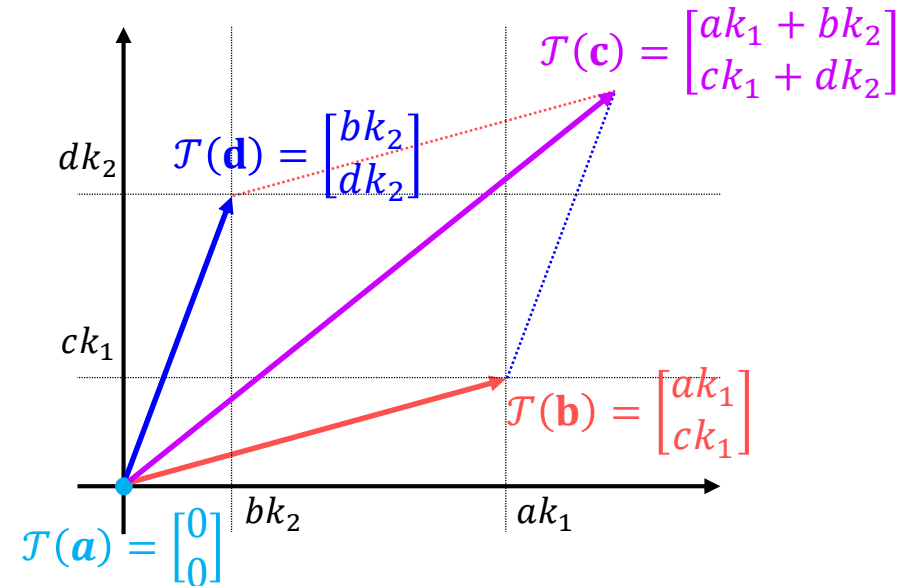
- $\mathcal{T}: \mathbb{R}^2 \rightarrow \mathbb{R}^2, \mathcal{T}(\mathbf{x}) = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \mathbf{x} = \mathbf{A}\mathbf{x}$

- $\mathbf{R} = \begin{bmatrix} k_1 & 0 \\ 0 & k_2 \end{bmatrix} \xrightarrow{\mathcal{T}} \mathcal{T}(\mathbf{R}) = \begin{bmatrix} ak_1 & bk_2 \\ ck_1 & dk_2 \end{bmatrix} = \mathbf{P}$

- $\det(\mathbf{P}) = |k_1k_2ad - k_1k_2bc| = |k_1k_2(ad - bc)| = |k_1k_2 \cdot \det(\mathbf{A})|$



$$\mathcal{T}: \mathbb{R}^2 \rightarrow \mathbb{R}^2$$



# Next Lecture

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- Affine Transformation