

# Linear Systems

## Matrix Inverses

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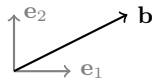
## References

- ▶ S Boyd, Applied Linear Algebra: Chapters 11.
- ▶ G Strang, Linear Algebra: Chapters 1.

## Representation of vectors in a basis

- Consider the vector space  $\mathbb{R}^n$  with basis  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ . Any vector in  $\mathbf{b} \in \mathbb{R}^n$  can be represented as a linear combination of  $\mathbf{v}_i$ s,

$$\mathbf{b} = \sum_{i=1}^n \mathbf{v}_i \mathbf{a}_i = \mathbf{V} \mathbf{a}; \quad \mathbf{a} \in \mathbb{R}^n, \quad \mathbf{V} = [\mathbf{v}_1 \quad \mathbf{v}_2 \quad \dots \quad \mathbf{v}_n] \in \mathbb{R}^{n \times n}$$



$\{\mathbf{v}_1, \mathbf{v}_2\}$ ,  $\{\mathbf{u}_1, \mathbf{u}_2\}$  and  $\{\mathbf{e}_1, \mathbf{e}_2\}$  are valid basis for  $\mathbb{R}^2$ , and the presentation for  $\mathbf{b}$  in each one of them is different.

- Finding out  $\mathbf{a}$  is easiest when we are dealing with an orthonormal basis  $\mathbf{U}$ , in which case  $\mathbf{a}$  is given by,

$$\mathbf{a} = \begin{bmatrix} \mathbf{u}_1^T \mathbf{b} \\ \mathbf{u}_2^T \mathbf{b} \\ \vdots \\ \mathbf{u}_n^T \mathbf{b} \end{bmatrix} = \mathbf{U}^T \mathbf{b} = \mathbf{b}_U$$

## Representation of vectors in a basis

Consider a vector  $\mathbf{b}$  whose representation in the standard basis is  $\mathbf{b} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ .

- Consider a basis  $V = \left\{ \begin{bmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix}, \begin{bmatrix} -2/\sqrt{5} \\ 1/\sqrt{5} \end{bmatrix} \right\}$ . Find out  $\mathbf{b}_V$ .

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- $U = \left\{ \begin{bmatrix} 1 \\ \frac{1}{2} \end{bmatrix}, \begin{bmatrix} -\frac{1}{2} \\ 1 \end{bmatrix} \right\}$ . Find out  $\mathbf{b}_U$ .

- $W = \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ \frac{1}{2} \end{bmatrix} \right\}$ . Find out  $\mathbf{b}_W$ .

## Matrix Inverse

- ▶ Consider the equation  $\mathbf{Ax} = \mathbf{y}$ , where  $\mathbf{A} \in \mathbb{R}^{n \times n}$  and  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ .
- ▶ Let us assume  $\mathbf{A}$  is non-singular  $\implies$  columns of  $\mathbf{A}$  represent a basis for  $\mathbb{R}^n$ .
- ▶ What does  $\mathbf{x}$  represent? It is the representation of  $\mathbf{y}$  in the basis consisting of the columns of  $\mathbf{A}$ .

$$\mathbf{y} = \mathbf{Ax} = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \dots & \mathbf{a}_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \sum_{i=1}^n \mathbf{a}_i x_i \implies \mathbf{x} = \mathbf{A}^{-1} \mathbf{y} = \begin{bmatrix} \tilde{\mathbf{b}}_1^T \\ \tilde{\mathbf{b}}_2^T \\ \vdots \\ \tilde{\mathbf{b}}_n^T \end{bmatrix} \mathbf{y} = \begin{bmatrix} \tilde{\mathbf{b}}_1^T \mathbf{y} \\ \tilde{\mathbf{b}}_2^T \mathbf{y} \\ \vdots \\ \tilde{\mathbf{b}}_n^T \mathbf{y} \end{bmatrix}$$

- ▶  $\mathbf{A}^{-1}$  is a matrix that allows change of basis to the columns of  $\mathbf{A}$  from the standard basis!

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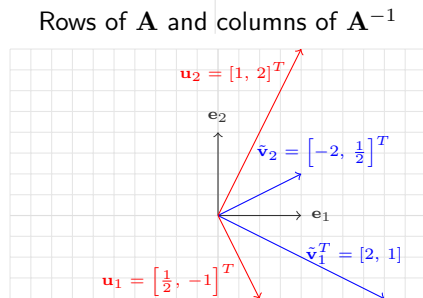
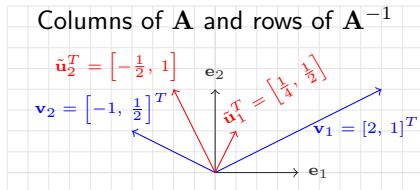
- ▶  $\mathbf{A}^{-1}$  is a matrix that allows change of basis to the columns of  $\mathbf{A}$  from the standard basis!

•  $W = \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ \frac{1}{2} \end{bmatrix} \right\}$ . Find  $\mathbf{b}_W$  by calculating the inverse of the matrix  $\mathbf{W} = \begin{bmatrix} 1 & -1 \\ 1 & \frac{1}{2} \end{bmatrix}$ . Does your answer match that of the previous approach?

• What about  $V = \left\{ \begin{bmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix}, \begin{bmatrix} -2/\sqrt{5} \\ 1/\sqrt{5} \end{bmatrix} \right\}$ . What is  $\mathbf{b}_V$ ?



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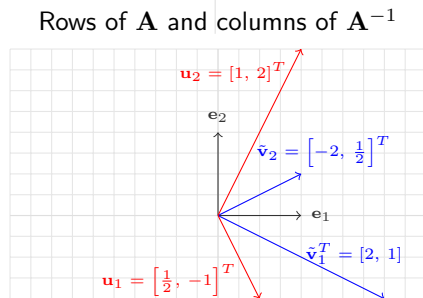
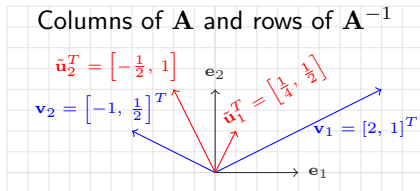
$$\mathbf{V} = [\mathbf{v}_1 \quad \mathbf{v}_2] = \begin{bmatrix} 2 & -1 \\ 1 & \frac{1}{2} \end{bmatrix}$$

$$\mathbf{V}^{-1} = \begin{bmatrix} \tilde{\mathbf{u}}_1^T \\ \tilde{\mathbf{u}}_2^T \end{bmatrix} = \frac{1}{2} \begin{bmatrix} \frac{1}{2} & 1 \\ -1 & 2 \end{bmatrix}$$

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Verify these for  $\mathbf{W} = \begin{bmatrix} 1 & -1 \\ 1 & \frac{1}{2} \end{bmatrix}$  and

$$\mathbf{V} = \begin{bmatrix} 1/\sqrt{5} & -2/\sqrt{5} \\ 2/\sqrt{5} & 1/\sqrt{5} \end{bmatrix}.$$

## Left Inverse

- ▶ Consider a rectangular matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$ . There exists no inverse  $\mathbf{A}^{-1}$  for this matrix.
- ▶ But, does there exist two matrices  $\mathbf{B}, \mathbf{C} \in \mathbb{R}^{n \times m}$ , such that,

$$\mathbf{CA} = \mathbf{I}_n \quad \text{and} \quad \mathbf{AB} = \mathbf{I}_m$$

- ▶ Both cannot be true for a rectangular matrix, only one can be true when the matrix is full rank.
- ▶ A rectangular matrix can only have either a left or a right inverse.

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Consider a matrix  $\mathbf{A} = \begin{bmatrix} 1 & -2 \\ 2 & 1 \\ 1 & 1 \end{bmatrix}$ . Let  $\mathbf{B}, \mathbf{C} \in \mathbb{R}^{2 \times 3}$ . Can you explain why only  $\mathbf{CA} = \mathbf{I}_2$  can be true and not  $\mathbf{AB} = \mathbf{I}_3$ ? Can you also explain why  $\mathbf{C}$  is not unique?

## Left Inverse

- ▶ Any non-zero  $\mathbf{a} \in \mathbb{R}^{n \times 1}$  is left invertible:  $\mathbf{b}\mathbf{a} = 1$ ,  $\mathbf{b} \in \mathbb{R}^{1 \times n}$ ;  $\mathbf{b}^T = \frac{\mathbf{a}}{\|\mathbf{a}\|^2} + \alpha \mathbf{a}^\perp$
- ▶ This can be generalized to  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $m > n$ .

$$(\mathbf{C} + \hat{\mathbf{C}}) \mathbf{A} = \mathbf{I}_m \text{ where } \mathbf{C}, \hat{\mathbf{C}} \in \mathbb{R}^{n \times m}, \hat{\mathbf{C}}\mathbf{A} = \mathbf{0}$$

- ▶ Condition for left inverse of  $\mathbf{A}$  to exist: *Columns of  $\mathbf{A}$  must be independent.*  
 $\longrightarrow \text{rank}(\mathbf{A}) = n \longrightarrow \mathbf{A}\mathbf{x} = \mathbf{0} \implies \mathbf{x} = \mathbf{0}.$
- ▶  $\mathbf{A}\mathbf{x} = \mathbf{b}$  can be solved, if and only if  $\mathbf{A}(\mathbf{C}\mathbf{b}) = \mathbf{b}$ , where  $\mathbf{C}\mathbf{A} = \mathbf{I}_n$ .

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• Consider the system,  $\mathbf{A}\mathbf{x} = \mathbf{b}$ .  $\mathbf{A} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ ,  $\mathbf{x} = [x]$ , and  $\mathbf{b} = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$ . Find  $\mathbf{x}$ .

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• What happens when  $\mathbf{b} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$ . What is  $\mathbf{x}$ ?



## Right Inverse

- ▶ For  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $n > m$  with full rank,  $\mathbf{AB} = \mathbf{I}_m \longrightarrow \mathbf{B}$  is the right inverse.
- ▶ Right inverse of  $\mathbf{A}$  exists only if the rows of  $\mathbf{A}$  are independent, i.e.  $\text{rank}(\mathbf{A}) = m \longrightarrow \mathbf{A}^T \mathbf{x} = \mathbf{0} \implies \mathbf{x} = \mathbf{0}$
- ▶  $\mathbf{Ax} = \mathbf{b}$  can be solved for any  $\mathbf{b}$ .  $\mathbf{x} = \mathbf{Bb} \implies \mathbf{A}(\mathbf{Bb}) = \mathbf{b}$ .
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- Let  $\mathbf{A} = \begin{bmatrix} 1 & -2 & 1 \\ 2 & 1 & -1 \end{bmatrix}$ . Find a complete solution for the right inverse of  $\mathbf{A}$ .

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• Solve  $\mathbf{Ax} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ . Compare the solutions from Gauss-Jordan method and the ones obtained using right-inverses.

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- Solve  $\mathbf{Ax} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ . Compare the solutions from Gauss-Jordan method and the ones obtained using right-inverses.
- Let  $\mathbf{AB} = \mathbf{I}_m$ . What about the relationship between  $\mathbf{A}^T$  and  $\mathbf{B}^T$ ?

# Fundamental subspaces of left and right inverses

## Left Inverse

- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = n$
- ▶ Subspaces of  $\mathbf{A}$ :  
 $\mathcal{C}(\mathbf{A}) \in \mathbb{R}^m$      $\mathcal{N}(\mathbf{A}^T) \in \mathbb{R}^m$   
 $\mathcal{C}(\mathbf{A}^T) = \mathbb{R}^n$      $\mathcal{N}(\mathbf{A}) = \{\mathbf{0}\}$
- ▶ Let  $\mathbf{C} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{CA} = \mathbf{I}_n$ . What is  $\text{rank}(\mathbf{C})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{C})$
  - ▶  $\mathcal{N}(\mathbf{C}^T)$
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- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = n$
- ▶ Subspaces of  $\mathbf{A}$ :  
 $\mathcal{C}(\mathbf{A}) \in \mathbb{R}^m$      $\mathcal{N}(\mathbf{A}^T) \in \mathbb{R}^m$   
 $\mathcal{C}(\mathbf{A}^T) = \mathbb{R}^n$      $\mathcal{N}(\mathbf{A}) = \{\mathbf{0}\}$
- ▶ Let  $\mathbf{C} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{CA} = \mathbf{I}_n$ . What is  $\text{rank}(\mathbf{C})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{C}) = \mathcal{C}(\mathbf{A}^T) = \mathbb{R}^n$
  - ▶  $\mathcal{N}(\mathbf{C}^T) = \mathcal{N}(\mathbf{A}) = \{\mathbf{0}\}$
  - ▶  $\mathcal{C}(\mathbf{C}^T) = \mathcal{C}(\mathbf{A}) \in \mathbb{R}^m$
  - ▶  $\mathcal{N}(\mathbf{C}) = \mathcal{N}(\mathbf{A}^T) \in \mathbb{R}^m$

# Fundamental subspaces of left and right inverses

## Left Inverse

- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = n$
- ▶ Subspaces of  $\mathbf{A}$ :
 
$$\begin{aligned} \mathcal{C}(\mathbf{A}) &\in \mathbb{R}^m & \mathcal{N}(\mathbf{A}^T) &\in \mathbb{R}^m \\ \mathcal{C}(\mathbf{A}^T) &= \mathbb{R}^n & \mathcal{N}(\mathbf{A}) &= \{\mathbf{0}\} \end{aligned}$$
- ▶ Let  $\mathbf{C} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{CA} = \mathbf{I}_n$ . What is  $\text{rank}(\mathbf{C})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{C}) = \mathcal{C}(\mathbf{A}^T) = \mathbb{R}^n$
  - ▶  $\mathcal{N}(\mathbf{C}^T) = \mathcal{N}(\mathbf{A}) = \{\mathbf{0}\}$
  - ▶  $\mathcal{C}(\mathbf{C}^T) = \mathcal{C}(\mathbf{A}) \in \mathbb{R}^m$
  - ▶  $\mathcal{N}(\mathbf{C}) = \mathcal{N}(\mathbf{A}^T) \in \mathbb{R}^m$

## Right Inverse

- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = m$
- ▶ Subspaces of  $\mathbf{A}$ :
 
$$\begin{aligned} \mathcal{C}(\mathbf{A}) &= \mathbb{R}^m & \mathcal{N}(\mathbf{A}^T) &= \{\mathbf{0}\} \\ \mathcal{C}(\mathbf{A}^T) &\in \mathbb{R}^n & \mathcal{N}(\mathbf{A}) &\in \mathbb{R}^n \end{aligned}$$
- ▶ Let  $\mathbf{B} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{AB} = \mathbf{I}_m$ . What is  $\text{rank}(\mathbf{B})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{B})$
  - ▶  $\mathcal{N}(\mathbf{B}^T)$
  - ▶  $\mathcal{C}(\mathbf{B}^T)$
  - ▶  $\mathcal{N}(\mathbf{B})$

# Fundamental subspaces of left and right inverses

## Left Inverse

- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = n$
- ▶ Subspaces of  $\mathbf{A}$ :
 
$$\begin{aligned} \mathcal{C}(\mathbf{A}) &\in \mathbb{R}^m & \mathcal{N}(\mathbf{A}^T) &\in \mathbb{R}^m \\ \mathcal{C}(\mathbf{A}^T) &= \mathbb{R}^n & \mathcal{N}(\mathbf{A}) &= \{\mathbf{0}\} \end{aligned}$$
- ▶ Let  $\mathbf{C} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{CA} = \mathbf{I}_n$ . What is  $\text{rank}(\mathbf{C})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{C}) = \mathcal{C}(\mathbf{A}^T) = \mathbb{R}^n$
  - ▶  $\mathcal{N}(\mathbf{C}^T) = \mathcal{N}(\mathbf{A}) = \{\mathbf{0}\}$
  - ▶  $\mathcal{C}(\mathbf{C}^T) = \mathcal{C}(\mathbf{A}) \in \mathbb{R}^m$
  - ▶  $\mathcal{N}(\mathbf{C}) = \mathcal{N}(\mathbf{A}^T) \in \mathbb{R}^m$

## Right Inverse

- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = m$
- ▶ Subspaces of  $\mathbf{A}$ :
 
$$\begin{aligned} \mathcal{C}(\mathbf{A}) &= \mathbb{R}^m & \mathcal{N}(\mathbf{A}^T) &= \{\mathbf{0}\} \\ \mathcal{C}(\mathbf{A}^T) &\in \mathbb{R}^n & \mathcal{N}(\mathbf{A}) &\in \mathbb{R}^n \end{aligned}$$
- ▶ Let  $\mathbf{B} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{AB} = \mathbf{I}_m$ . What is  $\text{rank}(\mathbf{B})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{B}) = \mathcal{C}(\mathbf{A}^T) \in \mathbb{R}^n$
  - ▶  $\mathcal{N}(\mathbf{B}^T)$
  - ▶  $\mathcal{C}(\mathbf{B}^T)$
  - ▶  $\mathcal{N}(\mathbf{B})$

# Fundamental subspaces of left and right inverses

## Left Inverse

- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = n$
- ▶ Subspaces of  $\mathbf{A}$ :
 
$$\begin{aligned} \mathcal{C}(\mathbf{A}) &\in \mathbb{R}^m & \mathcal{N}(\mathbf{A}^T) &\in \mathbb{R}^m \\ \mathcal{C}(\mathbf{A}^T) &= \mathbb{R}^n & \mathcal{N}(\mathbf{A}) &= \{\mathbf{0}\} \end{aligned}$$
- ▶ Let  $\mathbf{C} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{CA} = \mathbf{I}_n$ . What is  $\text{rank}(\mathbf{C})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{C}) = \mathcal{C}(\mathbf{A}^T) = \mathbb{R}^n$
  - ▶  $\mathcal{N}(\mathbf{C}^T) = \mathcal{N}(\mathbf{A}) = \{\mathbf{0}\}$
  - ▶  $\mathcal{C}(\mathbf{C}^T) = \mathcal{C}(\mathbf{A}) \in \mathbb{R}^m$
  - ▶  $\mathcal{N}(\mathbf{C}) = \mathcal{N}(\mathbf{A}^T) \in \mathbb{R}^m$

## Right Inverse

- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = m$
- ▶ Subspaces of  $\mathbf{A}$ :
 
$$\begin{aligned} \mathcal{C}(\mathbf{A}) &= \mathbb{R}^m & \mathcal{N}(\mathbf{A}^T) &= \{\mathbf{0}\} \\ \mathcal{C}(\mathbf{A}^T) &\in \mathbb{R}^n & \mathcal{N}(\mathbf{A}) &\in \mathbb{R}^n \end{aligned}$$
- ▶ Let  $\mathbf{B} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{AB} = \mathbf{I}_m$ . What is  $\text{rank}(\mathbf{B})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{B}) = \mathcal{C}(\mathbf{A}^T) \in \mathbb{R}^n$
  - ▶  $\mathcal{N}(\mathbf{B}^T) = \mathcal{N}(\mathbf{A}) \in \mathbb{R}^n$
  - ▶  $\mathcal{C}(\mathbf{B}^T)$
  - ▶  $\mathcal{N}(\mathbf{B})$

# Fundamental subspaces of left and right inverses

## Left Inverse

- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = n$
- ▶ Subspaces of  $\mathbf{A}$ :
 
$$\mathcal{C}(\mathbf{A}) \in \mathbb{R}^m \quad \mathcal{N}(\mathbf{A}^T) \in \mathbb{R}^m$$

$$\mathcal{C}(\mathbf{A}^T) = \mathbb{R}^n \quad \mathcal{N}(\mathbf{A}) = \{\mathbf{0}\}$$
- ▶ Let  $\mathbf{C} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{CA} = \mathbf{I}_n$ . What is  $\text{rank}(\mathbf{C})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{C}) = \mathcal{C}(\mathbf{A}^T) = \mathbb{R}^n$
  - ▶  $\mathcal{N}(\mathbf{C}^T) = \mathcal{N}(\mathbf{A}) = \{\mathbf{0}\}$
  - ▶  $\mathcal{C}(\mathbf{C}^T) = \mathcal{C}(\mathbf{A}) \in \mathbb{R}^m$
  - ▶  $\mathcal{N}(\mathbf{C}) = \mathcal{N}(\mathbf{A}^T) \in \mathbb{R}^m$

## Right Inverse

- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = m$
- ▶ Subspaces of  $\mathbf{A}$ :
 
$$\mathcal{C}(\mathbf{A}) = \mathbb{R}^m \quad \mathcal{N}(\mathbf{A}^T) = \{\mathbf{0}\}$$

$$\mathcal{C}(\mathbf{A}^T) \in \mathbb{R}^n \quad \mathcal{N}(\mathbf{A}) \in \mathbb{R}^n$$
- ▶ Let  $\mathbf{B} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{AB} = \mathbf{I}_m$ . What is  $\text{rank}(\mathbf{B})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{B}) = \mathcal{C}(\mathbf{A}^T) \in \mathbb{R}^n$
  - ▶  $\mathcal{N}(\mathbf{B}^T) = \mathcal{N}(\mathbf{A}) \in \mathbb{R}^n$
  - ▶  $\mathcal{C}(\mathbf{B}^T) = \mathcal{C}(\mathbf{A}) = \mathbb{R}^m$
  - ▶  $\mathcal{N}(\mathbf{B})$

# Fundamental subspaces of left and right inverses

## Left Inverse

- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = n$
- ▶ Subspaces of  $\mathbf{A}$ :
 
$$\begin{aligned} \mathcal{C}(\mathbf{A}) &\in \mathbb{R}^m & \mathcal{N}(\mathbf{A}^T) &\in \mathbb{R}^m \\ \mathcal{C}(\mathbf{A}^T) &= \mathbb{R}^n & \mathcal{N}(\mathbf{A}) &= \{\mathbf{0}\} \end{aligned}$$
- ▶ Let  $\mathbf{C} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{CA} = \mathbf{I}_n$ . What is  $\text{rank}(\mathbf{C})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{C}) = \mathcal{C}(\mathbf{A}^T) = \mathbb{R}^n$
  - ▶  $\mathcal{N}(\mathbf{C}^T) = \mathcal{N}(\mathbf{A}) = \{\mathbf{0}\}$
  - ▶  $\mathcal{C}(\mathbf{C}^T) = \mathcal{C}(\mathbf{A}) \in \mathbb{R}^m$
  - ▶  $\mathcal{N}(\mathbf{C}) = \mathcal{N}(\mathbf{A}^T) \in \mathbb{R}^m$

## Right Inverse

- ▶  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = m$
- ▶ Subspaces of  $\mathbf{A}$ :
 
$$\begin{aligned} \mathcal{C}(\mathbf{A}) &= \mathbb{R}^m & \mathcal{N}(\mathbf{A}^T) &= \{\mathbf{0}\} \\ \mathcal{C}(\mathbf{A}^T) &\in \mathbb{R}^n & \mathcal{N}(\mathbf{A}) &\in \mathbb{R}^n \end{aligned}$$
- ▶ Let  $\mathbf{B} \in \mathbb{R}^{n \times m}$  be the left inverse of  $\mathbf{A}$ , such that  $\mathbf{AB} = \mathbf{I}_m$ . What is  $\text{rank}(\mathbf{B})$ ?
- ▶ What about the subspaces of the left inverse?
  - ▶  $\mathcal{C}(\mathbf{B}) = \mathcal{C}(\mathbf{A}^T) \in \mathbb{R}^n$
  - ▶  $\mathcal{N}(\mathbf{B}^T) = \mathcal{N}(\mathbf{A}) \in \mathbb{R}^n$
  - ▶  $\mathcal{C}(\mathbf{B}^T) = \mathcal{C}(\mathbf{A}) = \mathbb{R}^m$
  - ▶  $\mathcal{N}(\mathbf{B}) = \mathcal{N}(\mathbf{A}^T) = \{\mathbf{0}\}$

## Pseudo Inverse

- ▶ Consider a tall, skinny matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$  with independent columns. It turns out the Gram matrix  $\mathbf{A}^T \mathbf{A} \in \mathbb{R}^{n \times n}$  is invertible. If that is the case then,

$$(\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{A} = \mathbf{I}_n; \quad (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \text{ is a left inverse.}$$

- ▶  $\mathbf{A}^\dagger = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T$  is called the *pseudo inverse* or the *Moore-Penrose inverse*.
- ▶ For the case of a fat, wide matrix, we have  $\mathbf{A}^\dagger = \mathbf{A}^T (\mathbf{A} \mathbf{A}^T)^{-1}$ .
- ▶ When  $\mathbf{A}$  is square and invertible,  $\mathbf{A}^\dagger = \mathbf{A}^{-1}$ .

## Pseudo Inverse

- ▶ Consider a tall, skinny matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$  with independent columns. It turns out the Gram matrix  $\mathbf{A}^T \mathbf{A} \in \mathbb{R}^{n \times n}$  is invertible. If that is the case then,

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- Solve  $\mathbf{A} \mathbf{x} = \mathbf{b}$  using the  $\mathbf{A}^\dagger$ .  $\mathbf{A} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ , and  $\mathbf{b} = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$ . Find  $\mathbf{x}$ .



## Pseudo Inverse

- ▶ Consider a tall, skinny matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$  with independent columns. It turns out the Gram matrix  $\mathbf{A}^T \mathbf{A} \in \mathbb{R}^{n \times n}$  is invertible. If that is the case then,

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- Solve  $\mathbf{A} \mathbf{x} = \mathbf{b}$  using the  $\mathbf{A}^\dagger$ .  $\mathbf{A} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ , and  $\mathbf{b} = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$ . Find  $\mathbf{x}$ .
- Compare  $\mathbf{A}^\dagger$  with that of the general left inverse  $\mathbf{C}$ . Calculate  $\|\mathbf{C}\|^2$  and find out the  $\min \|\mathbf{C}\|^2$ . What is  $\|\mathbf{A}^\dagger\|^2$ ?

## Matrix Inverse and Pseudo Inverse through QR factorization

- Consider an invertible, square matrix  $\mathbf{A} \in \mathbb{R}^{n \times n}$ .

$$\mathbf{A} = \mathbf{QR} \implies \mathbf{A}^{-1} = (\mathbf{QR})^{-1} = \mathbf{R}^{-1}\mathbf{Q}^{-1} = \mathbf{R}^{-1}\mathbf{Q}^T$$

where,  $\mathbf{R}, \mathbf{Q} \in \mathbb{R}^{n \times n}$ .  $\mathbf{R}$  is upper triangular, and  $\mathbf{Q}$  is an orthogonal matrix.

- In the case of a left invertible rectangular matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$ , we can factorize  $\mathbf{A} = \mathbf{QR}$ , with  $\mathbf{Q} \in \mathbb{R}^{m \times n}$  and  $\mathbf{R} \in \mathbb{R}^{m \times m}$ .

$$\mathbf{A}^\dagger = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T = (\mathbf{R}^T \mathbf{Q}^T \mathbf{Q} \mathbf{R})^{-1} \mathbf{R}^T \mathbf{Q}^T = (\mathbf{R}^T \mathbf{R})^{-1} \mathbf{R}^T \mathbf{Q}^T = \mathbf{R}^{-1} \mathbf{Q}^T$$

- For a right invertible wide, fat matrix, we can find out the pseudo-inverse of  $\mathbf{A}^T$ , and then take the transpose of the pseudo-inverse.

$$\mathbf{A} \mathbf{A}^\dagger = \mathbf{I} \implies \left(\mathbf{A}^\dagger\right)^T \mathbf{A}^T = \left(\mathbf{A}^T\right)^\dagger \mathbf{A}^T = \mathbf{I}$$

$$\mathbf{A}^T = \mathbf{QR} \implies \left(\mathbf{A}^T\right)^\dagger = \mathbf{R}^{-1} \mathbf{Q}^T = \left(\mathbf{A}^\dagger\right)^T \implies \mathbf{A}^\dagger = \mathbf{QR}^{-T}$$