

# EE 419/519: Industrial Control Systems

## Lecture 4: State Space Systems

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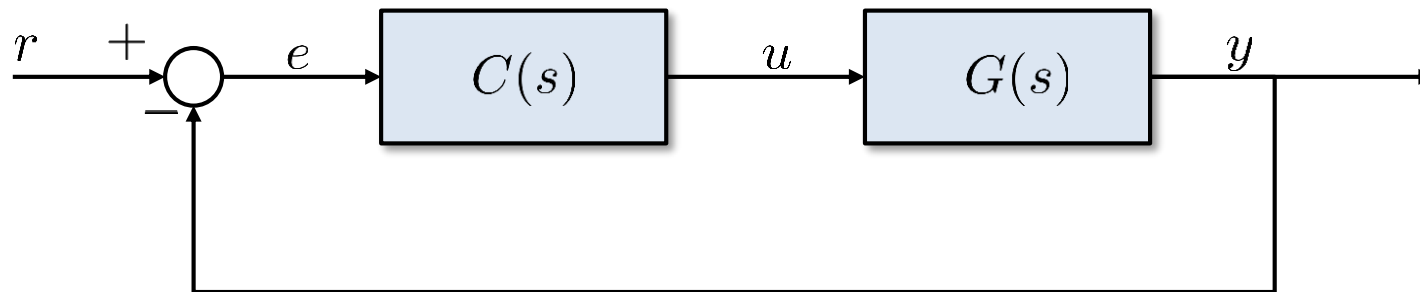
# Topic Outline

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- **Transfer Function vs State Space (Motivation)**
- **State Space Models and Examples**
- **From Nonlinear to Linear State Space Models**
- **Linear Algebra Review**
- **Jordan Canonical Form**

# Transfer Function (SISO) Drawbacks

- We have discussed the controller design for Single Input Single Output Systems (SISO)

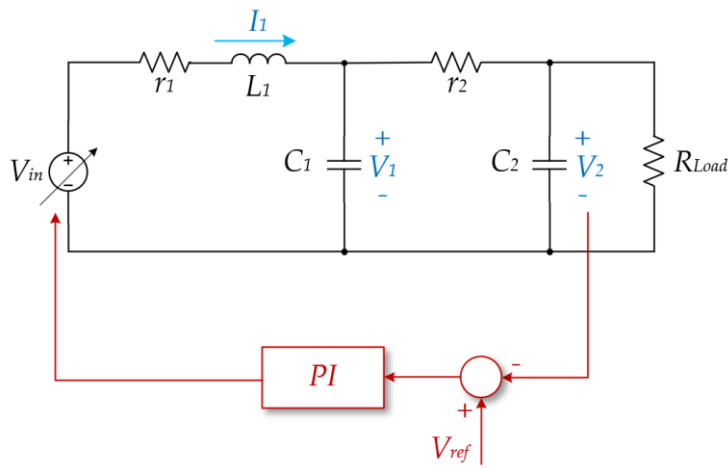


## Observations

- What if we had more sensors/observations available?
- What if we had more “inputs”?
- Could these be used to improve the closed loop system?

# Transfer Function (SISO) Drawbacks – Example 1

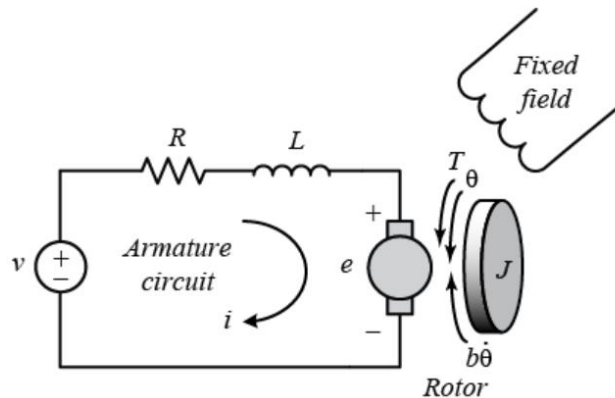
- In HW 2 problem 4, you will design a controller to regulate the  $C_2$  voltage
- **Assumptions:** Single input ( $V_{in}$ ) single output ( $V_2$ )



- What if we could also measure  $I_1$  and  $V_1$ ?

# Transfer Function (SISO) Drawbacks – Example 2

- We have also designed a controller to track the speed of a dc motor



- What if we could also measure the current  $i$ ?

# From Transfer Function to State Space Models – Case 1

- Let's consider a strictly proper transfer function of the form:

$$\frac{Y(s)}{U(s)} = \frac{1}{s^n + a_1 s^{n-1} + \cdots + a_{n-1} s + a_n}$$

# From Transfer Function to State Space Models – Case 2

- Let's consider a **more general** strictly proper transfer function

$$\frac{Y(s)}{U(s)} = \frac{b_1 s^{n-1} + \dots + b_{n-1} s + b_n}{s^n + a_1 s^{n-1} + \dots + a_{n-1} s + a_n}$$

# From Transfer Function to State Space Models – Case 2

- Let's consider a **more general** strictly proper transfer function

$$\frac{Y(s)}{U(s)} = \frac{b_1 s^{n-1} + \dots + b_{n-1} s + b_n}{s^n + a_1 s^{n-1} + \dots + a_{n-1} s + a_n}$$

- We can also use the command **tf2ss(num, den)**

```
1 %% Write the transfer function first
2 num = [b1 b2 ... bn];
3 den = [1 a1 ... an];
4 [A, B, C, D] = tf2ss(num, den)
```

$$A = \begin{pmatrix} -a_1 & -a_2 & \dots & \dots & -a_n \\ 1 & 0 & \dots & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \dots & \dots & 1 & 0 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} b_1 & b_2 & \dots & \dots & b_n \end{pmatrix}$$
$$D = 0$$



# Example – Transfer Function to State Space

- Transform the following transfer function to state space:

$$\frac{Y(s)}{U(s)} = \frac{2s + 1}{s^2 + 7s + 10}$$

# Topic Outline

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- Transfer Function vs State Space (Motivation)
- **State Space Models and Examples**
- **Linear Approximation of Nonlinear Systems**
- Linear Algebra Review
- Jordan Canonical Form

# Review: Nonlinear State Space System Definition

- A (nonlinear) State Space model with outputs is of the form:

$$\dot{x} = f(x, u)$$

$$y = h(x)$$

- The vector  $x \in \mathbb{R}^n$  are the **states**
- The vector  $u \in \mathbb{R}^m$  are the **inputs**
- The vector  $y \in \mathbb{R}^p$  are the **outputs**

# Review: Linear State Space System Definition

- An LTI State Space model with outputs is of the form:

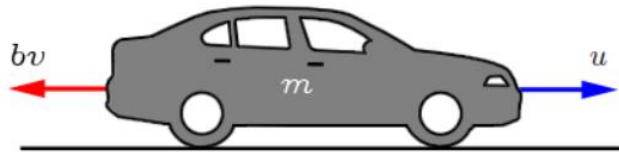
$$\dot{x} = Ax + Bu, \quad x(0) = x_0$$

$$y = Cx$$

- The vector  $x \in \mathbb{R}^n$  are the **states**
- The vector  $u \in \mathbb{R}^m$  are the **inputs**
- The vector  $y \in \mathbb{R}^p$  are the **outputs**

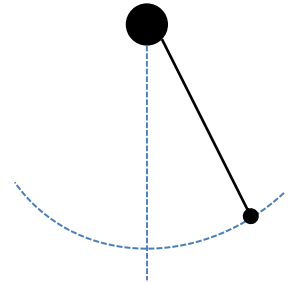
# Example 1: Cruise Control

- Derive a state space model of a car as shown in the figure



# Example 2: Pendulum

- Derive the state space differential equations for a pendulum



# Equilibrium Point Definition

- Many systems can be characterized by a nonlinear state space system

$$\dot{x} = f(x, u)$$

- An **equilibrium point**  $(x_e, u_e)$  is a point for which if the system starts there, it will remain there for all future time
- With respect to our state space model, this implies:  $0 = f(x_e, u_e)$

# Linear Approximation of Nonlinear Systems

- Many systems can be characterized by a nonlinear state space system

$$\dot{x} = f(x, u)$$

- However, close to an **equilibrium point**, many nonlinear systems can be **approximated** by a linear system



# Linear Approximation of Nonlinear Systems (2)

- Many systems can be characterized by a nonlinear state space system

$$\dot{x} = f(x, u)$$

- How can we obtain the  $A$  and  $B$  matrices which approximate  $f(x, u)$  close to  $(x_e, u_e)$ ?

# Linear Approximation of Nonlinear Systems (3)

- Many systems can be characterized by a nonlinear state space system

$$\dot{x} = f(x, u)$$

- A linear approximation of the system 'close' to an equilibrium point is given as follows

- Define  $\tilde{x} = x - x_e$  and  $\tilde{u} = u - u_e$

$$f(x, u) \approx \underbrace{\left. \frac{\partial f}{\partial x} \right|_{x_e, u_e}}_{\triangleq A} (x - x_e) + \underbrace{\left. \frac{\partial f}{\partial u} \right|_{x_e, u_e}}_{\triangleq B} (u - u_e)$$

- Then the behavior of the original nonlinear system close to the equilibrium point can be modeled by:

$$\dot{\tilde{x}} = A\tilde{x} + B\tilde{u}$$

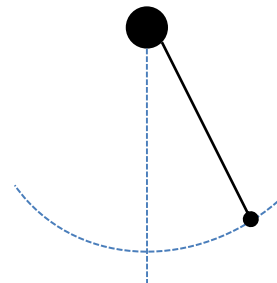
# Example: Pendulum Linearization

- The nonlinear state space model of a pendulum is

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = -\frac{g}{l} \sin(x_1) - \frac{k}{m} x_2$$

- Obtain a linear approximation at  $x_e = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$



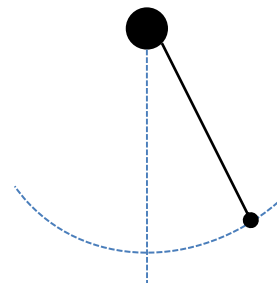
# Example: Pendulum Linearization (2)

- The nonlinear state space model of a pendulum is

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# Topic Outline

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- State Space Models and Examples
- From Nonlinear to Linear State Space Models
- **Linear Algebra Review**
- Jordan Canonical Form

# Motivation: Linear State Space Systems

- We will analyze linear time invariant (LTI) models in state space form:

$$\dot{x} = Ax + Bu, \quad x(0) = x_0$$

$$y = Cx$$

$$x \in \mathbb{R}^n, \quad u \in \mathbb{R}^m, \quad y \in \mathbb{R}^p$$

- Let's consider for example  $n = 2$ ,  $m = 1$ ,  $p = 1$
- When we say  $x \in \mathbb{R}^2$ , how can we visualize this? What **basis** are we referring to?  
What if we used a different basis?

# Linear Vector Spaces

- A *Linear Vector Space* is a set,  $\mathcal{V}$ , over a field,  $\mathbb{R}$ , in which having **two** operations:

1. Addition :

$$x, y \in \mathcal{V}, \quad z = x + y \in \mathcal{V}$$

Example:  $\mathcal{V} = \mathbb{R}^2$

2. Scalar multiplication :

$$x \in \mathcal{V}, \quad \alpha \in \mathbb{R}, \quad z = \alpha x \in \mathcal{V}$$

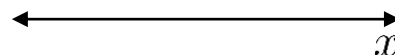
- (cont'd) it satisfies the following properties (Axioms):  $x, y, z \in \mathcal{V}$

1. (Commutativity)  $x + y = y + x$
2. (Associativity)  $(x + y) + z = x + (y + z)$
3. (Zero vector)  $\mathbf{0} + x = x, \quad x, \mathbf{0} \in \mathcal{V}$
4. (Additive inverse)  $\forall x \in \mathcal{V}$  there exists  $-x \in \mathcal{V}$  such that  $x + (-x) = \mathbf{0}$
5. (Identity scalar)  $1x = x$  for  $1 \in \mathbb{R}$
6. (Compatibility of scalar mult.)  $(ab)x = a(bx) \quad a, b \in \mathbb{R}$
7. (Distributivity w.r.t addition)  $a(x + y) = ax + ay \quad a \in \mathbb{R}$
8. (Distributivity w.r.t scalar mult.)  $(a + b)x = ax + bx \quad a, b \in \mathbb{R}$

# Linear Vector Spaces - Examples

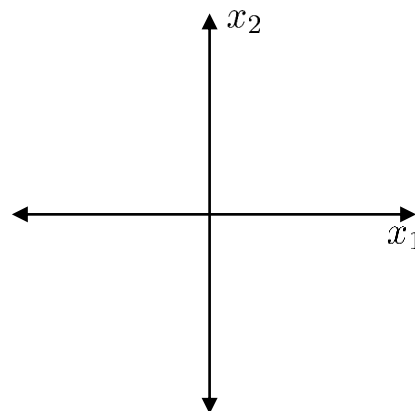
- What are some familiar examples of **Vector Spaces**?

1. The real numbers  $\mathbb{R}^1$



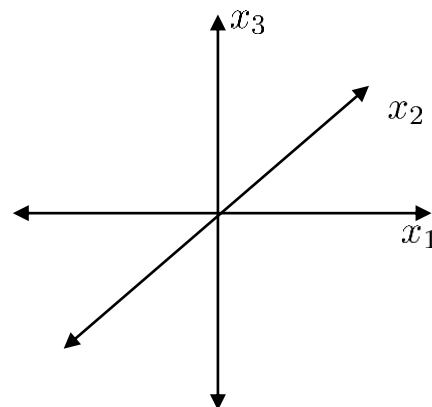
2. The 2 dimensional space,  $\mathbb{R}^2$

How to identify an element in  $\mathbb{R}^2$ ?



3. The 3 dimensional space,  $\mathbb{R}^3$

How to identify an element in  $\mathbb{R}^3$ ?





# Linear Vector Spaces – Examples (cont'd)

- Other not so common Linear Vector spaces:
  - The set of **polynomials** of degree less than or equal to  $n$ , e.g.  $n = 2$

$$\text{if } p, q \in \mathbb{P}_2 \Rightarrow p(x) = c_2x^2 + c_1x + c_0, \quad q(x) = d_2x^2 + d_1x + d_0$$

$$\text{e.g. } h(x) = 3x^2 + 1 \quad m(x) = 2x^2 + x - 3$$

- Define addition and scalar multiplication:

$$p(x) + q(x) = (c_2 + d_2)x^2 + (c_1 + d_1)x + (c_0 + d_0)$$

$$h(x) + m(x) =$$

## 2. Scalar multiplication:

$$\alpha \in \mathbb{R} \quad \alpha p(x) = \alpha c_2x^2 + \alpha c_1x + \alpha c_0$$

$$3h(x) =$$

- $\mathbb{P}_2$  and in general  $\mathbb{P}_n$  satisfy all the axioms for a vector space

# Subspace of a Vector Space

- Vector spaces provide us with a field to conduct extensive analysis
- **Definition:**  $\mathcal{S}$  is called a *subspace* of a vector space  $\mathcal{V}$  if  $\mathcal{S}$  is a subset of  $V$  and  $\mathcal{S}$  satisfies:
  - (i)  $\mathbf{0} \in \mathcal{S}$  ( $\mathcal{S}$  is non-empty)
  - (ii) For all  $x, y \in \mathcal{S}$ ,  $x + y \in \mathcal{S}$  ( $\mathcal{S}$  is closed under addition)
  - (iii) For all  $x \in \mathcal{S}$ ,  $\alpha x \in \mathcal{S}$ , where  $\alpha \in \mathbb{R}$  ( $\mathcal{S}$  is closed under scalar multiplication)
- To verify if a subset,  $\mathcal{S}$ , of a vector space,  $\mathcal{V}$ , is a subspace we need to check the three conditions above      **Remember them!**

# Subspace Example 1

- **Example:** Consider the following subset of  $\mathbb{R}^3$ :

$$\mathcal{S} = \left\{ \begin{pmatrix} x \\ y \\ z \end{pmatrix} \in \mathbb{R}^3 \mid x = 3y, z = -2y \right\}$$

- Is it a subspace of  $\mathbb{R}^3$ ?

## Subspace Example 2

- **Example:** Consider the following subset of  $\mathbb{R}^2$ :

$$\mathcal{S} = \left\{ \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \in \mathbb{R}^2 \mid x_2 = 3x_1 \right\}$$

- Is it a subspace of  $\mathbb{R}^2$ ?

# Subspace Example 3

- **Example:** Consider the following subset of  $\mathbb{R}^2$ :

$$\mathcal{S} = \left\{ \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \in \mathbb{R}^2 \mid x_1 \geq 0, x_2 \geq 0 \right\}$$

- Is it a subspace of  $\mathbb{R}^2$ ?

# Span of a Set of Vectors

- Suppose  $v_1, v_2, \dots, v_n$  are vectors defined in a vector space  $\mathcal{V}$

- A *Linear Combination* of these is a vector:

$$\alpha_1 v_1 + \alpha_2 v_2 + \cdots + \alpha_n v_n \in \mathcal{V} \quad \alpha_1, \dots, \alpha_n \in \mathbb{R}$$

- The set of all linear combinations of  $v_1, v_2, \dots, v_n$  is called **span**  $\{v_1, v_2, \dots, v_n\}$

- Example:

# Span of a Set of Vectors - Example

- Determine if  $v = \begin{pmatrix} 2 \\ -2 \\ -1 \end{pmatrix} \in \text{span} \left\{ \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \begin{pmatrix} 2 \\ -3 \\ 1 \end{pmatrix}, \begin{pmatrix} 5 \\ -4 \\ 0 \end{pmatrix} \right\}$

# Span of a Set of Vectors - Subspace

- **Theorem:** if  $v_1, \dots, v_n$  are elements of a vector space  $\mathcal{V}$ , then the  $\mathcal{S} = \text{span}\{v_1, v_2, \dots, v_n\}$  is a subspace of  $\mathcal{V}$

**Proof:**



# Spanning Set for a Vector Space

- Let  $\mathcal{V}$  be a vector space
- For a certain set of vectors  $v_1, \dots, v_n \in \mathcal{V}$ , we know that  $\text{span}\{v_1, \dots, v_n\} \subset \mathcal{V}$
- Is it possible that instead  $\text{span}\{v_1, \dots, v_n\} = \mathcal{V}$

# Spanning Set for a Vector Space (2)

- Let  $\mathcal{V}$  be a vector space
  - For a certain set of vectors  $v_1, \dots, v_n \in \mathcal{V}$ , we know that  $\text{span}\{v_1, \dots, v_n\} \subset \mathcal{V}$
  - Is it possible that instead  $\text{span}\{v_1, \dots, v_n\} = \mathcal{V}$
- **Definition:** The set  $v_1, v_2, \dots, v_n \in \mathcal{V}$  is a *spanning set* for  $\mathcal{V}$  if and only if every vector in  $\mathcal{V}$  can be written as a linear combination of  $v_1, v_2, \dots, v_n$

# Spanning Set for a Vector Space - Example

- **Example:** Determine if  $\text{span} \left\{ \begin{pmatrix} 3 \\ -5 \\ 2 \end{pmatrix}, \begin{pmatrix} 8 \\ 4 \\ 1 \end{pmatrix}, \begin{pmatrix} 0 \\ 3 \\ 0 \end{pmatrix} \right\} = \mathbb{R}^3$

# Spanning Set and Linearly Dependent Set of Vectors

- Consider two spanning sets for  $\mathbb{R}^2$

- Let  $\{v_1, v_2, \dots, v_n\}$  be a **linearly dependent** set of vectors  
 $\Rightarrow$  There is at least one vector, say  $v_1$ , that can be written as a sum of the others
- Then, the span  $\{v_1, v_2, \dots, v_n\} = \text{span}\{v_2, v_3, \dots, v_n\}$

# Linearly Dependent Vectors

- In general, given  $\{v_1, v_2, \dots, v_n\}$ , it is possible to write one of the vectors as a linear combination of the others  $n - 1$  vectors **if and only if** there exists  $c_1, c_2, \dots, c_n \in \mathbb{R}$  *not all zero* such that  $c_1v_1 + c_2v_2 + \dots + c_nv_n = 0$

# Linear Independence

- The vectors  $v_1, v_2, \dots, v_n$  in a vector space  $\mathcal{V}$  are said to be *Linearly Independent* if

$$c_1v_1 + c_2v_2 + \dots + c_nv_n = 0 \Rightarrow c_1 = c_2 = \dots = c_n = 0$$

- If  $\{v_1, v_2, \dots, v_n\}$  is a *minimal spanning* set for  $\mathcal{V}$  then  $v_1, v_2, \dots, v_n$  are *linearly independent*!
- If  $\{v_1, v_2, \dots, v_n\}$  is a *linearly independent* set of vectors, then  $\text{span}\{v_1, v_2, \dots, v_n\} = \mathcal{V}$  is a *minimal spanning* set for  $\mathcal{V}$ !

*Minimal Spanning Set  $\equiv$  Basis*

# Linear Independence – Example 1

- Determine if  $\left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right\}$  are linearly independent

# Linear Independence – Example 2

- Determine if  $\left\{ \begin{pmatrix} 1 \\ 2 \\ 4 \end{pmatrix}, \begin{pmatrix} 2 \\ 1 \\ 3 \end{pmatrix}, \begin{pmatrix} 4 \\ -1 \\ 1 \end{pmatrix} \right\}$  are linearly independent



# Linear Independence – Example 3

- Determine if  $\left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \right\}$  are linearly independent

# Basis of a Vector Space

- **Theorem:** Suppose  $v_1, v_2, \dots, v_n$  are linearly independent vectors in a vector space  $\mathcal{V}$  and let  $v \in \text{span}\{v_1, v_2, \dots, v_n\}$ , **then**  $v$  can be written uniquely as a linear combination of  $v_1, v_2, \dots, v_n$

- **Definition:** The vectors  $v_1, v_2, \dots, v_n$  form a *basis* for a vector space  $\mathcal{V}$  iff:
  1.  $\{v_1, v_2, \dots, v_n\}$  are linearly independent
  2.  $\text{span}\{v_1, v_2, \dots, v_n\} = \mathcal{V}$

# Basis of a Vector Space – Example 1

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- In  $\mathbb{R}^3$  the *standard or natural* basis set is

## Basis of a Vector Space – Example 2

- Does the set  $v_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ ,  $v_2 = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$ ,  $v_3 = \begin{pmatrix} 2 \\ 0 \\ 1 \end{pmatrix}$  form a basis for  $\mathbb{R}^3$ ?

# Basis of a Vector Space – Example 3

- Does the set  $v_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ ,  $v_2 = \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix}$ ,  $v_3 = \begin{pmatrix} 0 \\ 2 \\ 4 \end{pmatrix}$  form a basis for  $\mathbb{R}^3$ ?

1. Are they linearly independent?

$$c_1 v_1 + c_2 v_2 + c_3 v_3 = 0 \Rightarrow \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 2 \\ 1 & 2 & 4 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \xrightarrow{\text{RREF}} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ c_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

2. Do they span  $\mathbb{R}^3$ ? can any vector  $x = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \in \mathbb{R}^3$  be written as a combination of  $v_1, v_2, v_3$ ?
- $$\alpha_1 v_1 + \alpha_2 v_2 + \alpha_3 v_3 = x$$

$$\Rightarrow \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 2 \\ 1 & 2 & 4 \end{pmatrix}}_{=A} \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \xrightarrow{\text{RREF}} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{pmatrix} \Rightarrow \det(A) = 0$$

# Dimension of a Vector Space and Basis Summary

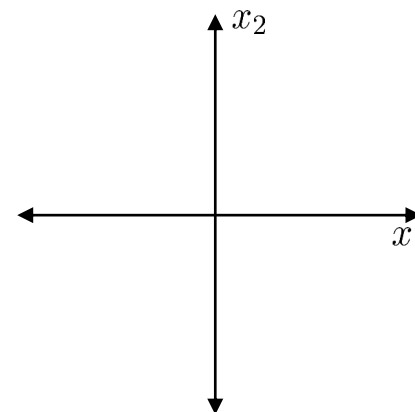
- **Theorem:** Let  $\mathcal{V}$  is a vector space of dimension  $n > 0$ . Then:
  1. Any set of  $n$  linearly independent vectors span  $\mathcal{V}$
  2. Any set of  $n$  vectors that span  $\mathcal{V}$  are linearly independent
  3. No set of less than  $n$  vectors can span  $\mathcal{V}$
  4. Any subset of less than  $n$  linearly indep. vectors can be extended to form a basis for  $\mathcal{V}$
  5. Any spanning set of  $> n$  vectors can be parted down to form a basis for  $\mathcal{V}$
- $\mathbb{R}^3$  is a vector space with dimension 3 since any basis must have 3 vectors, e.g.  $\{e_1, e_2, e_3\}$   
**No set of less than 3 vectors can span  $\mathbb{R}^3$ !**

# Change of Basis - Motivation

## Motivation:

- Many applied problems can be simplified by changing from one coordinate system to another, e.g. when integrating the volume of a solid is better to use spherical coordinates,  $\{r, \theta, \phi\}$ , rather than rectangular,  $\{x, y, z\}$
- When we specify a vector, we typically assume it is with respect to the standard basis
- For example in  $\mathbb{R}^2$ , a vector  $x = \begin{bmatrix} \alpha \\ \beta \end{bmatrix}$  is implied to be  $x = \alpha \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \beta \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

Standard basis for  $\mathbb{R}^2 = \text{span}\{e_1, e_2\}$



# Change of Basis – General Basis

## Definition:

- Let  $\mathcal{V}$  be a finite dimensional vector space, and let  $\mathcal{B} = \{b_1, b_2, \dots, b_k\}$  be a basis for  $\mathcal{V}$ . Then any  $v \in \mathcal{V}$ , can be written uniquely as:  $v = \alpha_1 b_1 + \alpha_2 b_2 + \dots + \alpha_k b_k$
- The elements  $\alpha_1, \alpha_2, \dots, \alpha_k$  are called the **coordinates of  $v$  w.r.t. to  $\mathcal{B}$**

- $v_{\mathcal{B}} = \begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_k \end{bmatrix}$  is called the **coordinate vector of  $v$  w.r.t.  $\mathcal{B}$**

- $v = \begin{bmatrix} v_1 \\ \vdots \\ v_k \end{bmatrix}$  is called the **coordinate vector of  $v$  w.r.t. the standard basis**

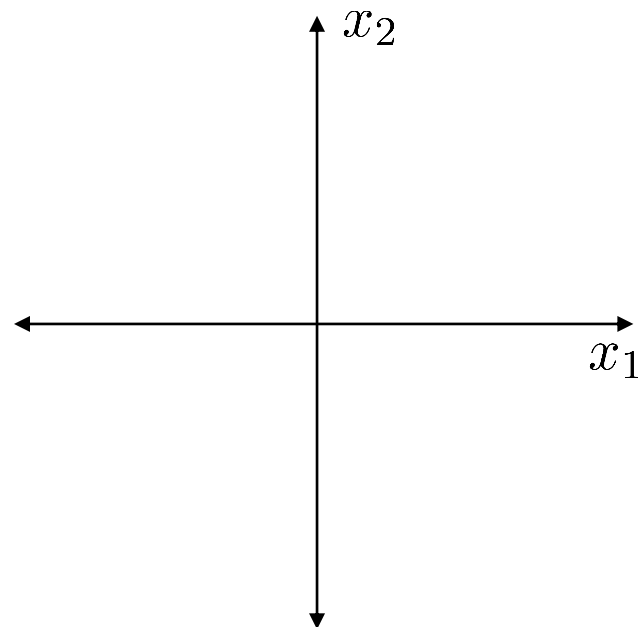


# Change of Basis – Coordinate Vector

- Given a vector  $x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  (w.r.t standard basis), we would like to find the coordinates w.r.t another basis  $\mathcal{B}$
- For example,  $\mathcal{B} = \left\{ \begin{bmatrix} 3 \\ 5 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\}$ , then  $x_{\mathcal{B}} = \begin{bmatrix} \alpha \\ \beta \end{bmatrix}$  implies that

# Change of Basis – Example (1)

- Example, given  $x = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$  find its coordinates w.r.t  $\mathcal{B} = \left\{ \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}$ , i.e.  $x_{\mathcal{B}}$



## Change of Basis – Example (2)

- Given two basis for  $\mathbb{R}^2$ :  $\mathcal{B} = \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\}$   $\mathcal{C} = \left\{ \begin{bmatrix} 3 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ -2 \end{bmatrix} \right\}$
- Let  $x_{\mathcal{B}} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$ , find  $x_{\mathcal{C}} = \begin{bmatrix} \alpha_1 \\ \alpha_2 \end{bmatrix}$

# Change of Basis – Transition Matrix

- For  $\mathbb{R}^n$ , let  $B = \begin{bmatrix} b_1 & b_2 & \dots & b_n \end{bmatrix}$  be a matrix whose columns form a basis for  $\mathbb{R}^n$ .  
Let  $v \in \mathbb{R}^n$ ,  $v = \alpha_1 b_1 + \dots + \alpha_n b_n$
- Similarly, let  $C = \begin{bmatrix} c_1 & \dots & c_n \end{bmatrix}$  be a matrix whose columns form a basis for  $\mathbb{R}^n$ ,  
then  $v = \gamma_1 c_1 + \dots + \gamma_n c_n$
- Therefore, we can find the following transition matrices:

$$\begin{aligned} C v_{\mathcal{C}} &= B v_{\mathcal{B}} \Rightarrow v_{\mathcal{C}} = \overbrace{C^{-1} B} v_{\mathcal{B}} \\ &\Rightarrow v_{\mathcal{B}} = \underbrace{B^{-1} C} v_{\mathcal{C}} \end{aligned}$$

# Linear Maps

- Let  $\mathcal{V}$  be a vector space of dimension  $n$
- and  $\mathcal{W}$  be a vector space of dimension  $m$  over the field of  $\mathbb{R}$

- **Definition:** A function  $T : \mathcal{V} \rightarrow \mathcal{W}$  is called *linear* if

$$T(u + v) = T(u) + T(v) \quad \forall u, v \in \mathcal{V}$$

$$T(\alpha v) = \alpha T(v) \quad \forall \alpha \in \mathbb{R}, v \in \mathcal{V}$$

# Linear Mapping Examples

- Show whether the transformation,  $L : \mathbb{R}^2 \rightarrow \mathbb{R}^3$ , is linear or not

$$L \left( \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \right) = \begin{bmatrix} 2x \\ x - 2y \\ 3x + 4y \end{bmatrix}$$

# Linear Mapping Examples

- Show whether the transformation,  $L : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ , is linear or not

$$L \left( \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \right) = \begin{bmatrix} 2x \\ xy \end{bmatrix}$$

# Matrix as a Linear Map

- **Theorem:** Let  $A \in \mathbb{R}^{m \times n}$ . The mapping  $L_A(x) : \mathbb{R}^n \rightarrow \mathbb{R}^m$  defined by:

$$L_A(x) = Ax, \quad \forall x \in \mathbb{R}^n$$

is a linear transformation.



# Matrix of a Linear Map

- **Theorem:** If  $L$  is a *linear* transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^m$ , then there is an  $m \times n$  matrix  $A$  such that :

$$L(x) = Ax$$

for all  $x \in \mathbb{R}^n$ . The  $j^{\text{th}}$  column of  $A$  is given by:

$$a_j = L(e_j), \text{ for } j = 1, \dots, n$$

That is  $L = L_A$ :

$$A = \begin{bmatrix} L(e_1) & L(e_2) & \cdots & L(e_n) \end{bmatrix}$$

The matrix  $A$  is called the **standard** matrix of a linear transformation  $L$ .

# Linear Mapping – Matrix Examples

- Find the standard matrix of the following linear transformation:

$$L\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = \begin{bmatrix} 2x \\ x - 2y \\ 3x + 4y \end{bmatrix}$$

# Linear Mapping – Different Basis

- We now know a linear transformation from two finite dimensional vector spaces, i.e.  $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$ , has a **standard** matrix representation
- What if we were to use a different (not standard) basis for  $\mathbb{R}^n$  and  $\mathbb{R}^m$ ? What is the matrix representation of  $L$  w.r.t. these different basis?

# Null Space of a Linear Map

- **Definition:** Let  $A \in \mathbb{R}^{m \times n}$ , i.e.  $A : \mathbb{R}^n \rightarrow \mathbb{R}^m$ , then the set

$$\mathcal{N}(A) = \{x \in \mathbb{R}^n \mid Ax = 0\}$$

is a subspace of  $\mathbb{R}^n$  and is called the **null space** of  $A$

# Range/Column Space of a Linear Map

- **Definition:** Let  $A \in \mathbb{R}^{m \times n}$ , i.e.  $A : \mathbb{R}^n \rightarrow \mathbb{R}^m$ , then the set

$$\mathcal{R}(A) = \{y \in \mathbb{R}^m \mid \exists x \in \mathbb{R}^n, Ax = y\} = \{y = Ax \in \mathbb{R}^m \mid x \in \mathbb{R}^n\}$$

is a subspace of  $\mathbb{R}^m$  and is called the **range/column space** of  $A$

- The range space for this matrix is also called the **column** space of  $A$  since:

$$\mathcal{R}(A) = \text{span} \{a_1, a_2, \dots, a_n\}$$

where  $a_i$  are the columns of  $A$

# Null/Range Space Example

- Find a **basis for** the null and column space of the following matrix:

$$A = \begin{bmatrix} 1 & 1 & 1 & 0 \\ 2 & 1 & 0 & 1 \end{bmatrix}$$

# Null/Range Space Example

- Find a **basis for** the null and column space of the following matrix:

$$A = \begin{bmatrix} 1 & 1 & 1 & 0 \\ 2 & 1 & 0 & 1 \end{bmatrix}$$

# Topic Outline

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- Transfer Function vs State Space (Motivation)
- State Space Models and Examples
- From Nonlinear to Linear State Space Models
- Linear Algebra Review
- **Jordan Canonical Form (LN5 Slides)**