

# Section 3.6

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## Dimensions

### Objective

- What is the **dimension** of a vector space?
- How to compute dimension for a vector space?
- What are some equivalent conditions for a set to be a basis for a vector space?

### Theorem 3.6.1

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Let  $V$  be a vector space which has a basis  
 $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  with  $k$  vectors.

1. Any subset of  $V$  with more than  $k$  vectors is always linearly dependent.
2. Any subset of  $V$  with less than  $k$  vectors cannot span  $V$ .

Recall Thm 3.4.7:

Any subset of  $\mathbf{R}^n$  with more than  $n$  vectors is linearly dep.

Recall Thm 3.2.7:

Any subset of  $\mathbf{R}^n$  with less than  $n$  vectors cannot span  $\mathbf{R}^n$ .

### Theorem 3.6.1 & Remark 3.6.2

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Let  $V$  be a vector space which has a basis  
 $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  with  $k$  vectors.

1. Any subset of  $V$  with more than  $k$  vectors is always linearly dependent.
2. Any subset of  $V$  with less than  $k$  vectors cannot span  $V$ .

$> k$  : too many vectors to be a basis

$< k$  : too few vectors to be a basis

All bases for a vector space  
have the same number of vectors

# What is dimension of a vector space?

## Definition 3.6.3

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The **dimension** of a vector space  $V$   
denoted by  $\dim(V)$   
is the **number of vectors in a basis** for  $V$ .

**Recall:**

The **basis for zero space** is defined to be the **empty set**.

The number of vector in this “basis” is 0.

$$\dim(\{\mathbf{0}\}) = 0$$

## Example 3.6.4.1-3

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1. The dimension of  $\mathbf{R}^n$  is  $n$ ,  
i.e.  $\dim(\mathbf{R}^n) = n$ .
2. Except  $\{\mathbf{0}\}$  and  $\mathbf{R}^2$ , all subspaces of  $\mathbf{R}^2$  are  
lines through the origin  $\text{span}\{\mathbf{u}\}$   
they are of dimension 1.
3. Except  $\{\mathbf{0}\}$  and  $\mathbf{R}^3$ , all subspaces of  $\mathbf{R}^3$  are  
either lines through the origin  $\text{span}\{\mathbf{u}\}$   
they are of dimension 1,  
or planes containing the origin,  $\text{span}\{\mathbf{u}, \mathbf{v}\}$   
they are of dimension 2.

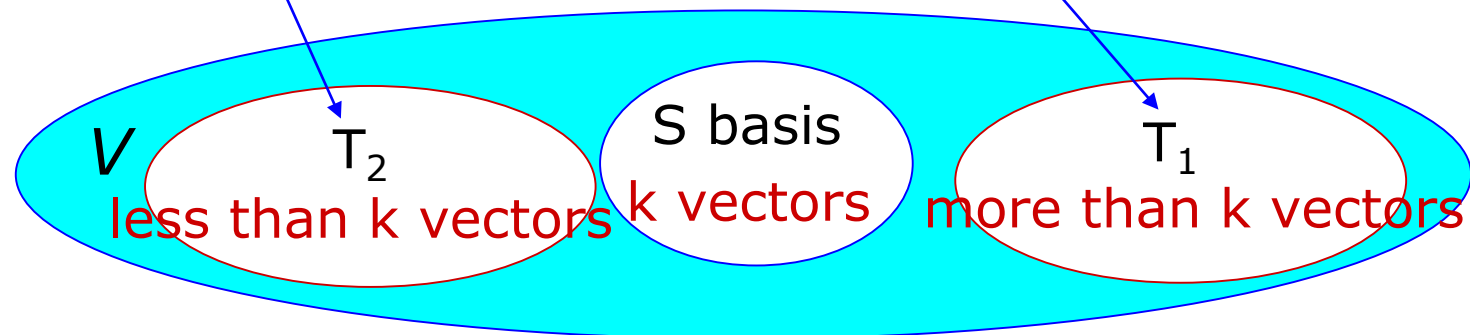
Number of vectors in a basis

→ dimension of the vector space

## Theorem 3.6.1

Let  $V$  be a vector space which has a basis  $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  with  $k$  vectors.  $\dim V = k$

1. Any subset of  $V$  with more than  $k$  vectors is always linearly dependent.
2. Any subset of  $V$  with less than  $k$  vectors cannot span  $V$ .



## Finding dimension of a subspace

### Example 3.6.4.4

Not the same as the “dimension” of the vectors in the subspace

Find a **basis** for and determine the **dimension** of the subspace  $W = \{(x, y, z) \mid y = z\}$  of  $\mathbf{R}^3$ .

Note:  $\dim(W) \neq 3$

Explicit:  $(x, y, y) = x(1, 0, 0) + y(0, 1, 1)$

So  $W = \text{span}\{(1, 0, 0), (0, 1, 1)\}$   
linearly independent

**basis** for  $W$  :  $\{(1, 0, 0), (0, 1, 1)\}$

**dim**( $W$ ) = 2

## Dimension of solution space

### Example 3.6.6

$\mathbb{R}^5$

Solution space

$$su_1 + tu_2$$

Find a **basis** for and determine the **dimension** of the solution space of the homogeneous system

$$\begin{cases} 2v + 2w - x + z = 0 \\ -v - w + 2x - 3y + z = 0 \\ x + y + z = 0 \\ v + w - 2x - z = 0 \end{cases} \xrightarrow{\text{general solution}} \begin{pmatrix} v \\ w \\ x \\ y \\ z \end{pmatrix} = \begin{pmatrix} -s-t \\ s \\ -t \\ 0 \\ t \end{pmatrix} = s \begin{pmatrix} -1 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + t \begin{pmatrix} -1 \\ 0 \\ -1 \\ 0 \\ 1 \end{pmatrix}$$

$\mathbf{u}_1$                        $\mathbf{u}_2$

solution space =  $\text{span}\{\mathbf{u}_1, \mathbf{u}_2\}$

**basis** for the solution space =  $\{\mathbf{u}_1, \mathbf{u}_2\}$

**dim**(solution space) = 2

no. of parameters in the general solution

linearly indep.



## Dimension of solution space

### Discussion 3.6.5 (Example)

homogeneous system with 6 variables:  $u, v, w, x, y, z$

Gaussian Elimination

general solution with 4 parameters:  $s, t, r, q$

$$\begin{pmatrix} u \\ v \\ w \\ x \\ y \\ z \end{pmatrix} = \begin{pmatrix} s + 5t - 3r \\ s \\ t \\ 2r - 3q \\ r \\ q \end{pmatrix} = s \begin{pmatrix} 1 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} + t \begin{pmatrix} 5 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + r \begin{pmatrix} -3 \\ 0 \\ 0 \\ 2 \\ 1 \\ 0 \end{pmatrix} + q \begin{pmatrix} 0 \\ 0 \\ 0 \\ -3 \\ 0 \\ 1 \end{pmatrix}$$

$$s\mathbf{u}_1 + t\mathbf{u}_2 + r\mathbf{u}_3 + q\mathbf{u}_4$$

linearly independent

the solution space =  $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3, \mathbf{u}_4\}$

$\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3, \mathbf{u}_4\}$  is a basis for the solution space

$\dim(\text{solution space}) = 4$

### Discussion 3.6.5

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homogeneous system  $\longrightarrow$  row echelon form  $\mathbf{R}$

number of non-pivot columns in  $\mathbf{R}$

||

number of parameters in general solution

||

number of vectors in basis for solution space

||

the dimension of the solution space

## Showing a set form a basis (alternative ways)

### Theorem 3.6.7

Let  $V$  be a vector space of dimension  $k$  and  $S$  a subset of  $V$ .

The following are equivalent:

1.  $S$  is a **basis** for  $V$
2.  $S$  is **linearly independent** and  $|S| = k = \dim(V)$
3.  $S$  **spans**  $V$  and  $|S| = k = \dim(V)$

To show  $S$  is a basis for  $V$  :

$S$  lin. indep  
 $S$  spans  $V$

or

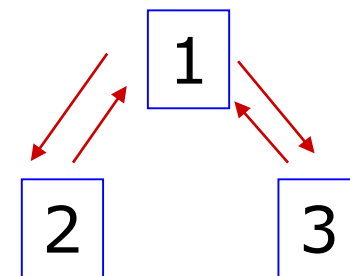
$S$  lin. indep  
 $|S| = \dim V$

or

$S$  spans  $V$   
 $|S| = \dim V$

## The proof

1.  $S$  is a basis for  $V$ .
2.  $S$  is lin indep and  $|S| = k$ .
3.  $S$  spans  $V$  and  $|S| = k$ .



## Theorem 3.6.7

"1  $\Rightarrow$  2" and "1  $\Rightarrow$  3" is immediate.

2  $\Rightarrow$  1 : (prove by contradiction)

Assume that  $S$  is **not** a basis for  $V$

Given  $S$  is linearly independent and  $|S| = k$ .

So  $\text{span}(S) \neq V$ .

There is a vector  $\mathbf{u}$  in  $V$  and  $\mathbf{u} \notin \text{span}(S)$ .

$\mathbf{u}$  is not redundant in  $\text{span}(S)$

Let  $S' = S \cup \{\mathbf{u}\}$

$k + 1$  vectors

**Contradiction**

$\Rightarrow S'$  is linearly indep.

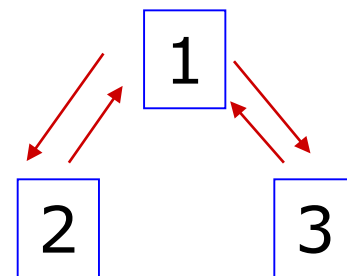
$\Rightarrow S'$  is linearly dep. see Theorem 3.6.1.1

see Theorem 3.4.10

So  $S$  is a basis for  $V$

## The proof

1.  $S$  is a basis for  $V$ .
2.  $S$  is lin indep and  $|S| = k$ .
3.  $S$  spans  $V$  and  $|S| = k$ .



## Theorem 3.6.7

$3 \Rightarrow 1$  : (prove by contradiction)

Assume  $S$  not a basis for  $V$

Given  $S$  spans  $V$  and  $|S| = k$ .

$S$  is linearly dependent.

There is a redundant vector  $\mathbf{v}$  in  $S$ .

Let  $S'' = S - \{\mathbf{v}\} \Rightarrow \text{span}(S'') = \text{span}(S) = V$   
see Theorem 3.2.12

$k - 1$  vectors

$\Rightarrow \text{span}(S'') \neq V$

Contradiction

see Theorem 3.6.1.2

So  $S$  is a basis for  $V$

## Showing a set form a basis (alternative ways)

### Example 3.6.8

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Show that

$\mathbf{u}_1 = (2, 0, -1)$ ,  $\mathbf{u}_2 = (4, 0, 7)$  and  $\mathbf{u}_3 = (-1, 1, 4)$   
form a basis for  $\mathbf{R}^3$ .

Since  $\dim \mathbf{R}^3 = 3$ ,  
we only need to show the set of 3 vectors is  
either linear independent or spans  $\mathbf{R}^3$ .

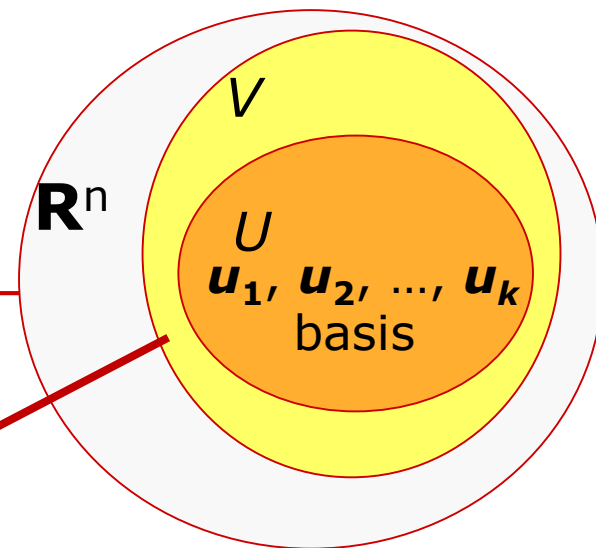
If we don't know the dimension of a vector space  $V$ ,  
to show a set is a basis for  $V$ ,  
we still need to check  
the set is both linear independent and spans  $V$ .

Dimensions give the “size” of subspaces of  $\mathbf{R}^n$

## Theorem 3.6.9

Let  $U$  and  $V$  be subspaces of  $\mathbf{R}^n$

We say:  $U$  is a subspace of  $V$ .



(i) If  $U \subseteq V$ , then  $\dim(U) \leq \dim(V)$

(ii) If  $U \subseteq V$  and  $U \neq V$ , then  $\dim(U) < \dim(V)$

For (i),  $\dim(U) = k$

$u_1, u_2, \dots, u_k$  are  $k$  lin. indep. vectors in  $V$

So  $k \leq \dim(V)$

For (ii), suppose  $\dim(U) = \dim(V)$

Then  $\dim(V) = k$

contradiction

So  $V = \text{span}\{u_1, u_2, \dots, u_k\} = U$ .

Dimensions give the “size” of subspaces of  $\mathbf{R}^n$

### Example 3.6.10

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Given  $V$  a plane in  $\mathbf{R}^3$  containing the origin.

Suppose  $U$  is a subspace of  $V$  such that  $U \neq V$ .  
What can we say about  $U$ ?

$V$  is of dimension 2.

By Theorem 3.6.9,  $\dim(U) < 2$ .

So

either  $\dim(U) = 0 \iff U = \{\mathbf{0}\}$

or  $\dim(U) = 1 \iff U = \text{a line through the origin}$

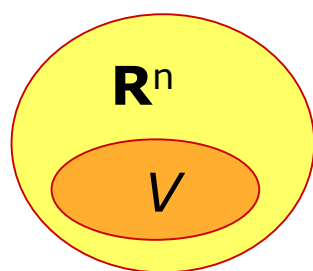


# True or False

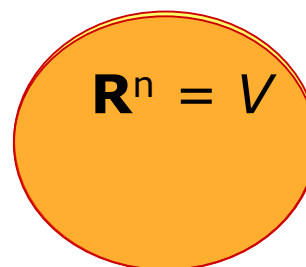
Let  $U$  and  $V$  be subspaces of  $\mathbf{R}^n$

No subspace of  $\mathbf{R}^n$  has dimension  $n$ , except  $\mathbf{R}^n$  itself.

A. If  $\dim(V) = n$ , then  $V = \mathbf{R}^n$  True

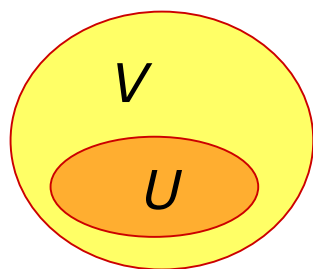


$V$  and  $\mathbf{R}^n$  have the same "size"

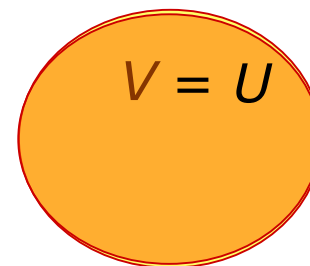


Theorem 3.6.9.2

B. If  $U \subseteq V$  and  $\dim(U) = \dim(V)$ , then  $U = V$  True



$U$  and  $V$  have the same "size"



## Theorem 3.6.11

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**A** is an  $n \times n$  matrix.

The following statements are **equivalent**:

1. **A** is invertible.
2. The linear system  $\mathbf{Ax} = \mathbf{0}$  has only the trivial solution.
3. The reduced row-echelon form of **A** is an identity matrix.
4. **A** can be expressed as a product of elementary matrices.
5.  $\det(\mathbf{A}) \neq 0$ .
6. The rows of **A** form a basis for  $\mathbf{R}^n$ .
7. The columns of **A** form a basis for  $\mathbf{R}^n$ .

1.  $\mathbf{A}$  is invertible
2.  $\mathbf{Ax} = \mathbf{0}$  has only trivial solution
7. The columns of  $\mathbf{A}$  form a basis for  $\mathbf{R}^n$

## Example

Suppose we know  $\begin{pmatrix} 1 & 2 & 1 \\ 3 & 1 & 0 \\ 2 & 0 & 1 \end{pmatrix}$  is invertible.

Then we know that the linear system

$$\begin{pmatrix} 1 & 2 & 1 \\ 3 & 1 & 0 \\ 2 & 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \text{ has only trivial solution}$$

Write the linear system in vector equation form:

$$x \begin{pmatrix} 1 \\ 3 \\ 2 \end{pmatrix} + y \begin{pmatrix} 2 \\ 1 \\ 0 \end{pmatrix} + z \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \text{ has only zero coefficients}$$
$$x = y = z = 0$$

We conclude that  $\left\{ \begin{pmatrix} 1 \\ 3 \\ 2 \end{pmatrix}, \begin{pmatrix} 2 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \right\}$  is linearly independent  
hence form a basis for  $\mathbf{R}^3$

1. **A** is invertible
5.  $\det \mathbf{A} \neq 0$
7. The columns of **A** form a basis for  $\mathbf{R}^n$
6. The rows of **A** form a basis for  $\mathbf{R}^n$

## Example

Suppose we know  $\begin{pmatrix} 1 & 2 & 1 \\ 3 & 1 & 0 \\ 2 & 0 & 1 \end{pmatrix}$  is invertible.

Then we know that the determinant  $\begin{vmatrix} 1 & 2 & 1 \\ 3 & 1 & 0 \\ 2 & 0 & 1 \end{vmatrix} \neq 0$

Then the transpose determinant  $\begin{vmatrix} 1 & 3 & 2 \\ 2 & 1 & 0 \\ 1 & 0 & 1 \end{vmatrix} \neq 0$

So  $\begin{pmatrix} 1 & 3 & 2 \\ 2 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix}$  is invertible.

So the columns  $\left\{ \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}, \begin{pmatrix} 3 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 2 \\ 0 \\ 1 \end{pmatrix} \right\}$  form a basis for  $\mathbf{R}^3$

So the rows  $\{(1 \ 2 \ 1), (3 \ 1 \ 0), (2 \ 0 \ 1)\}$  form a basis for  $\mathbf{R}^3$

## Alternative method to check basis for $\mathbf{R}^n$

### Example 3.6.12 (Determinant method)

$$\mathbf{u}_1 = (1, 1, 1), \mathbf{u}_2 = (-1, 1, 2), \mathbf{u}_3 = (1, 0, 1)$$

Is  $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$  a basis for  $\mathbf{R}^3$ ? YES

$$\begin{vmatrix} 1 & 1 & 1 \\ -1 & 1 & 2 \\ 1 & 0 & 1 \end{vmatrix} = 3 \neq 0$$

$$\mathbf{u}_1 = (1, 1, 1, 1), \mathbf{u}_2 = (1, -1, 1, -1),$$

$$\mathbf{u}_3 = (0, 1, -1, 0), \mathbf{u}_4 = (2, 1, 1, 0)$$

Is  $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3, \mathbf{u}_4\}$  a basis for  $\mathbf{R}^4$ ? NO

$$\begin{vmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 0 & 1 & -1 & 0 \\ 2 & 1 & 1 & 0 \end{vmatrix} = 0$$

Cannot use this method to check basis for subspaces of  $\mathbf{R}^n$

# Section 3.7

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## Transition Matrices

### Objective

- What is a transition matrix?
- How to compute transition matrices?
- What is the relation between coordinate vectors w.r.t. different bases?

## From one basis to another

### Example 3.7.4.1

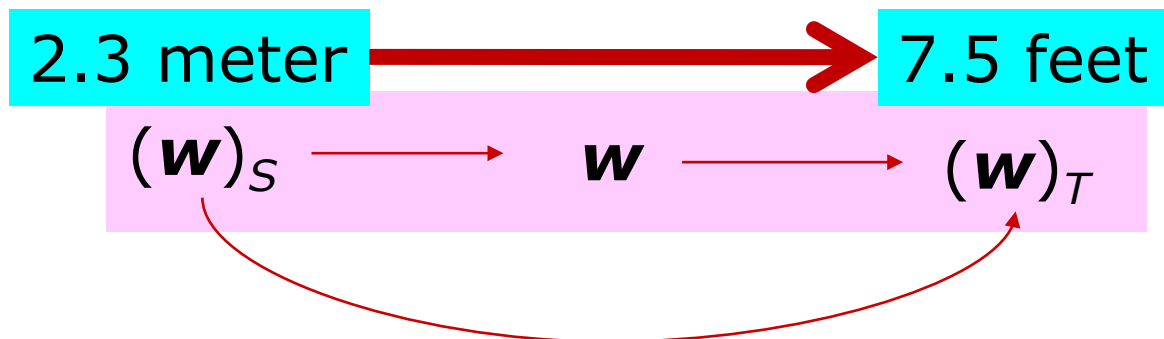
$S = \{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$  basis for  $\mathbf{R}^3$

$\mathbf{u}_1 = (1, 0, -1)$ ,  $\mathbf{u}_2 = (0, -1, 0)$ ,  $\mathbf{u}_3 = (1, 0, 2)$ .

$T = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  basis for  $\mathbf{R}^3$

$\mathbf{v}_1 = (1, 1, 1)$ ,  $\mathbf{v}_2 = (1, 1, 0)$ ,  $\mathbf{v}_3 = (-1, 0, 0)$ .

Given  $(\mathbf{w})_S = (2, -1, 2)$ . Find  $(\mathbf{w})_T$ .



Is there a direct method?

## Notation 3.7.1

$S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$ : a basis for a vector space  $V$

$\mathbf{v}$ : a vector in  $V$

Write  $\mathbf{v} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \dots + c_k\mathbf{u}_k$

Then  $(\mathbf{v})_S = (c_1, c_2, \dots, c_k)$  row form of coordinate vector

$$[\mathbf{v}]_S = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_k \end{pmatrix} \quad \text{column form of coordinate vector}$$

We need to pre-multiply the coordinate-vector by a  $k \times k$  matrix



$$[\mathbf{w}]_S = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_k \end{pmatrix} \quad [\mathbf{w}]_T = \begin{pmatrix} d_1 \\ d_2 \\ \vdots \\ d_k \end{pmatrix}$$

## Discussion 3.7.2

$S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  and  $T = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$   
two bases for a vector space  $V$ .

Take a vector  $\mathbf{w}$  in  $V$

Relation between  $[\mathbf{w}]_S$  and  $[\mathbf{w}]_T$ ?

$\mathbf{w}$  in terms of  $\mathbf{u}_i$

$\mathbf{w}$  in terms of  $\mathbf{v}_i$

We will show that

does not depend on  $\mathbf{w}$

$[\mathbf{w}]_T = \mathbf{P} [\mathbf{w}]_S$  for some **fixed**  $k \times k$  matrix  $\mathbf{P}$   
transition matrix

## Finding transition matrix from S to T

### Definition 3.7.3

Read Discussion 3.7.2  
to see why it works

$S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  and  $T = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$   
two bases for a vector space  $V$ .

1. Express each  $\mathbf{u}_i$  as linear combination of  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$
2. Form the (column) coordinate vectors w.r.t.  $T$

$$[\mathbf{u}_1]_T = \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{k1} \end{pmatrix} \quad [\mathbf{u}_2]_T = \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{k2} \end{pmatrix} \quad \dots \quad [\mathbf{u}_k]_T = \begin{pmatrix} a_{1k} \\ a_{2k} \\ \vdots \\ a_{kk} \end{pmatrix}$$

3. Form the matrix  $\mathbf{P} = ( [\mathbf{u}_1]_T \ [\mathbf{u}_2]_T \ \dots \ [\mathbf{u}_k]_T )$

$$\mathbf{P} = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1k} \\ a_{21} & a_{22} & \cdots & a_{2k} \\ \vdots & \vdots & \ddots & \vdots \\ a_{k1} & a_{k2} & \cdots & a_{kk} \end{pmatrix}$$

transition matrix  
from  $S$  to  $T$

4.  $\mathbf{P} [\mathbf{w}]_S = [\mathbf{w}]_T$  for any vector  $\mathbf{w}$  in  $V$ .

## From one basis to another

### Example 3.7.4.1

$S = \{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$  basis for  $\mathbf{R}^3$

$$\mathbf{u}_1 = (1, 0, -1), \quad \mathbf{u}_2 = (0, -1, 0), \quad \mathbf{u}_3 = (1, 0, 2).$$

$T = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  basis for  $\mathbf{R}^3$

$$\mathbf{v}_1 = (1, 1, 1), \quad \mathbf{v}_2 = (1, 1, 0), \quad \mathbf{v}_3 = (-1, 0, 0).$$

(a) Find the transition matrix from  $S$  to  $T$ .

$$P = ([\mathbf{u}_1]_T \quad [\mathbf{u}_2]_T \quad [\mathbf{u}_3]_T)$$

(b)  $\mathbf{w}$  a vector in  $\mathbf{R}^3$  with  $(\mathbf{w})_S = (2, -1, 2)$ .

Find  $(\mathbf{w})_T$ .

$$[\mathbf{w}]_T = P [\mathbf{w}]_S$$

# Finding transition matrix

$$S = \{u_1, u_2, u_3\}$$

$$T = \{v_1, v_2, v_3\}$$

## Example 3.7.4.1(a)

$$u_1 = a_{11}v_1 + a_{21}v_2 + a_{31}v_3$$

$$u_2 = a_{12}v_1 + a_{22}v_2 + a_{32}v_3$$

$$u_3 = a_{13}v_1 + a_{23}v_2 + a_{33}v_3$$

find  $a_{11}, a_{21}, \dots, a_{33}$

Convert to three linear systems:

$$\begin{cases} a_{11} + a_{21} - a_{31} = 1 \\ a_{11} + a_{21} = 0 \\ a_{11} = -1 \end{cases}$$

$$\begin{cases} a_{12} + a_{22} - a_{32} = 0 \\ a_{12} + a_{22} = -1 \\ a_{12} = 0 \end{cases}$$

$$\begin{cases} a_{13} + a_{23} - a_{33} = 1 \\ a_{13} + a_{23} = 0 \\ a_{13} = 2 \end{cases}$$

$$\left( \begin{array}{ccc|ccc} 1 & 1 & -1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 0 & -1 & 0 \\ 1 & 0 & 0 & -1 & 0 & 2 \end{array} \right)$$

$v_1 \quad v_2 \quad v_3$

$u_1 \quad u_2 \quad u_3$

Gauss-Jordan  
Elimination

$$\left( \begin{array}{ccc|ccc} 1 & 0 & 0 & -1 & 0 & 2 \\ 0 & 1 & 0 & 1 & -1 & -2 \\ 0 & 0 & 1 & -1 & -1 & -1 \end{array} \right)$$

$[u_1]_T \quad [u_2]_T \quad [u_3]_T$

transition matrix from  $S$  to  $T$

## Example 3.7.4.1(b)

$$(\mathbf{w})_S = (2, -1, 2)$$

$$[\mathbf{w}]_T = (\text{Transition matrix from } S \text{ to } T)[\mathbf{w}]_S$$

$$\mathbf{P} = \begin{pmatrix} -1 & 0 & 2 \\ 1 & -1 & -2 \\ -1 & -1 & -1 \end{pmatrix}$$

$$[\mathbf{w}]_T = \mathbf{P}[\mathbf{w}]_S = \begin{pmatrix} -1 & 0 & 2 \\ 1 & -1 & -2 \\ -1 & -1 & -1 \end{pmatrix} \begin{pmatrix} 2 \\ -1 \\ 2 \end{pmatrix} = \begin{pmatrix} 2 \\ -1 \\ -3 \end{pmatrix}$$

$$\text{So } (\mathbf{w})_T = (2, -1, -3).$$

## From $S$ to $T$ and from $T$ to $S$

### Example 3.7.4.2

$$P [w]_S = [w]_T \text{ for any vector } w$$

$$Q [w]_T = [w]_S \text{ for any vector } w$$

$$S = \{u_1, u_2\} \quad u_1 = (1, 1), \quad u_2 = (1, -1).$$

$$T = \{v_1, v_2\} \quad v_1 = (1, 0), \quad v_2 = (1, 1).$$

two bases for  $\mathbf{R}^2$

transition matrix from  $S$  to  $T$       transition matrix from  $T$  to  $S$

$$\begin{cases} u_1 = 0v_1 + v_2 \\ u_2 = 2v_1 - v_2 \end{cases}$$

$$\begin{cases} v_1 = \frac{1}{2}u_1 + \frac{1}{2}u_2 \\ v_2 = u_1 + 0u_2 \end{cases}$$

$$[u_1]_T = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad [u_2]_T = \begin{pmatrix} 2 \\ -1 \end{pmatrix}$$

$$[v_1]_S = \begin{pmatrix} \frac{1}{2} \\ \frac{1}{2} \end{pmatrix} \quad [v_2]_S = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$P = \begin{pmatrix} 0 & 2 \\ 1 & -1 \end{pmatrix} \quad \longleftrightarrow \quad Q = \begin{pmatrix} \frac{1}{2} & 1 \\ \frac{1}{2} & 0 \end{pmatrix}$$

inverse of each other

## The inverse of transition matrix

### Theorem 3.7.5

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$S$  and  $T$  : two bases of a vector space

$P$ : the transition matrix from  $S$  to  $T$ .

1.  $P$  is invertible.
2.  $P^{-1}$  is the transition matrix from  $T$  to  $S$ .

$S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$ ,  $T = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  bases

$P = ( [\mathbf{u}_1]_T \ [\mathbf{u}_2]_T \ \dots \ [\mathbf{u}_k]_T ) \Rightarrow P$  is invertible

$[\mathbf{u}_1]_T \ [\mathbf{u}_2]_T \ \dots \ [\mathbf{u}_k]_T$  are linearly independent

Let  $Q$  be the transition matrix from  $T$  to  $S$ .

$Q = ( [\mathbf{v}_1]_S \ [\mathbf{v}_2]_S \ \dots \ [\mathbf{v}_k]_S )$

To show  $QP = I$

# The proof: two observations

## Theorem 3.7.5

Let  $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  basis

**Obs. 1**

$$[\mathbf{u}_1]_S = \begin{pmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \quad [\mathbf{u}_2]_S = \begin{pmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \quad \dots \quad [\mathbf{u}_k]_S = \begin{pmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix}$$

standard basis vectors

$$\mathbf{u}_1 = 1\mathbf{u}_1 + 0\mathbf{u}_2 + \dots + 0\mathbf{u}_k$$

**Obs. 2**

any  $m \times n$  matrix  $\mathbf{A}$

$$\begin{pmatrix} a_{11} & \dots & a_{1,i-1} & a_{1i} & a_{1,i+1} & \dots & a_{1n} \\ a_{21} & \dots & a_{2,i-1} & a_{2i} & a_{2,i+1} & \dots & a_{2n} \\ \vdots & & \vdots & \vdots & \vdots & & \vdots \\ \vdots & & \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{m,i-1} & a_{mi} & a_{m,i+1} & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} a_{1j} \\ a_{2j} \\ 0 \\ \vdots \\ a_{mj} \end{pmatrix}$$

$j^{\text{th}}$  column of  $\mathbf{A}$

$j^{\text{th}}$  coordinate



## The proof

$$[\mathbf{u}_1]_S = \begin{pmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \quad [\mathbf{u}_2]_S = \begin{pmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \quad \dots \quad [\mathbf{u}_k]_S = \begin{pmatrix} 0 \\ 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix}$$

## Theorem 3.7.5

To show  $\mathbf{QP} = \mathbf{I}$

Examine the  $i^{\text{th}}$  column of  $\mathbf{QP}$  for  $i = 1, 2, \dots, k$

$i^{\text{th}}$  column of  $\mathbf{A} = \mathbf{A} [\mathbf{u}_i]_S$

$$i^{\text{th}} \text{ column of } \mathbf{QP} = \mathbf{QP} [\mathbf{u}_i]_S = \mathbf{Q} [\mathbf{u}_i]_T = [\mathbf{u}_i]_S = \begin{pmatrix} 0 \\ 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}$$

$\mathbf{P}$  : transition matrix from  $S$  to  $T$

$\mathbf{Q}$  : transition matrix from  $T$  to  $S$

$$\mathbf{QP} = \begin{pmatrix} 1 & 0 & & 0 \\ 0 & 1 & & 0 \\ 0 & 0 & \ddots & 0 \\ \vdots & \vdots & & \vdots \\ 0 & 0 & & 1 \end{pmatrix} = \mathbf{I}$$

So  $\mathbf{P}$  is invertible  
and  $\mathbf{P}^{-1} = \mathbf{Q}$