

Today: • When does a basis of eigenvectors exist?  
(a sufficient condition) 9.2

• How to go between different bases  
(equivalence of matrices) → 11.1, 11.2, some 11.4

Recall: Last time we finished at the Fundam. Thm of Algebra:

every polynomial over  $\mathbb{C}$  factors into a product of linear factors:

$$p(x) = a_n (x - \alpha_1) \cdots (x - \alpha_n) \quad (n = \deg(p)) \quad (\text{see 9.4})$$

↑ roots.      ↗

Comment: for  $n \geq 5$ , there is no formula for finding the roots! (Galois)

In practice, this means we cannot expect to find eigenvalues exactly. Numerically, can find eigenvalues approximately.

Def: If  $p(x) = a(x - \alpha)^k \cdot h(x)$   
the greatest such  $k$  is called the multiplicity of the root  $\alpha$ .

• If start with an  $n \times n$  matrix  $A$ ,  
write its characteristic polynomial  $p_A(\lambda) = \det(A - \lambda I)$

If  $p_A(\lambda)$  has  $n$  distinct roots in  $\mathbb{C}$   
(meaning every root has multiplicity = 1)  
then  $A$  will have a basis of eigenvectors

If there are multiple roots, it is more complicated.  
- there might still be a basis of eigenvectors, or not. Not easy to distinguish these cases.

Recall why eigenvalues are roots of  $P_A(\lambda)$ :

because  $\ker(A - \lambda I) \neq \{0\}$  when  $\lambda$  is an eigenvalue

Def:  $\dim \ker(A - \lambda I)$  is called the geometric multiplicity of the eigenvalue  $\lambda$ .

It is not 1 only if  $\lambda$  is also a multiple root of  $P_A(\lambda)$  and in this case, the multiplicity of  $\lambda$  as a root of the char. poly.  $P_A(\lambda)$  is called algebraic multiplicity.

Caution: geom. multiplicity  $\leq$  alg. multiplicity but they don't have to be equal!

To find geom. mult., you have to: do row reductions for  $A - \lambda I$ , find its kernel.

There is a method for finding a particularly good

basis for this kernel, but it's not part of Math 223  
And then need more basis vectors for  $V$

Thm If  $\underbrace{v_1^{(1)}, \dots, v_{k_1}^{(1)}}_{\text{---}}$  are lin. indep. eigenvectors for eigenvalue  $\lambda_1$ ,

$\underbrace{v_1^{(m)}, \dots, v_{k_m}^{(m)}}_{\text{---}}$  are lin. indep. eigenvectors for the eigenvalue  $\lambda_m$ .

then  $\{v_i^{(1)}, \dots, v_{k_m}^{(m)}\}$  are lin. indep.  
 ↑  
 all of them

(eigenvectors corr. to different eigenvalues are lin. indep.)  
 based on:

Lemma: If  $\lambda_1, \dots, \lambda_k$  are eigenvalues of  $A: V \rightarrow V$ ,  $\lambda_i \neq \lambda_j$  for  $i \neq j$   
 let  $v_i$  be an eigenvector for  $\lambda_i$   
 then  $v_1, \dots, v_k$  are lin. indep.

Pf: Induction on  $k$ .

Base:  $k=1$ :  $v_1$  - eigenvector for  $\lambda_1$   
 $v_1 \neq 0$  by def'n.  
 Nothing to prove.

Induction step: Suppose we know this statement for  $k=n$ . Want to prove it for  $k=n+1$ .

Let  $v_1, \dots, v_{n+1}$  be eigenvectors for  $\lambda_1, \dots, \lambda_{n+1}$ .

Suppose we have

$$(1) \quad \alpha_1 v_1 + \dots + \alpha_n v_n + \alpha_{n+1} v_{n+1} = 0 \quad \text{for some } \alpha_1, \dots, \alpha_{n+1} \in F.$$

↑  
our field.

Want to prove:  
 all  $\alpha_i = 0$

Apply  $A$ : Get.

$$\begin{aligned} & A(\alpha_1 v_1 + \dots + \alpha_{n+1} v_{n+1}) = 0 \\ &= \alpha_1 A v_1 + \dots + \alpha_{n+1} A v_{n+1} \\ &= \alpha_1 \lambda_1 v_1 + \dots + \alpha_{n+1} \lambda_{n+1} v_{n+1} = 0. \end{aligned} \quad (2)$$

↑  
 b/c eigenvectors.

Subtract  $\lambda_{n+1} \cdot (1)$  from  $(2)$ :

Get: (the last term cancels!)

$$(\alpha_1 \lambda_1 v_1 - \alpha_1 \lambda_{n+1} v_1) + \dots + (\alpha_n \lambda_n v_n - \lambda_{n+1} \alpha_n v_n) = 0$$

$$\Downarrow \alpha_1 (\lambda_1 - \lambda_{n+1}) v_1 + \dots + \alpha_n (\lambda_n - \lambda_{n+1}) v_n = 0.$$

By induction assumption,  $\{v_i\}$  are lin. indep.

And it was given that  $\lambda_i \neq \lambda_j$  for  $i \neq j$

so  $\lambda_i \neq \lambda_{n+1}$  when  $i = 1, \dots, n$ .

Then  $\alpha_1 = \alpha_2 = \dots = \alpha_n = 0$ .

Thus we have  $\alpha_{n+1} v_{n+1} = 0$ . So  $\alpha_{n+1} = 0$ .  
(b/c  $v_{n+1} \neq 0$  by def. of an eigenvector)

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Corollary: If the roots of the char. poly (over  $\mathbb{C}$ ) are distinct, the lin. op. has a basis of eigenvectors

Pf: You have  $n$  eigenvalues, their eigenvectors  $\overset{\dim(V)}{}$  are lin. independent, since there's  $n$  of them, they form a basis.

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The matrix of  $A$  in this basis is diagonal!

How to change bases:

## Changes of basis

- What happens to coordinates when you change the basis?
- What happens to matrices

Let  $v_1^{\text{old}}, \dots, v_n^{\text{old}}$  be a basis in  $V$

And we take another basis, call them  $v_1^{\text{new}}, \dots, v_n^{\text{new}}$

$$X \in V \quad X = x_1^{\text{old}} v_1^{\text{old}} + \dots + x_n^{\text{old}} v_n^{\text{old}} \\ = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}^{\text{old}} \quad \text{in the old basis.}$$

$$X = x_1^{\text{new}} v_1^{\text{new}} + \dots + x_n^{\text{new}} v_n^{\text{new}} \\ = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}^{\text{new}} \quad \text{in the new basis.}$$

Make the "transition matrix" or "change of basis" matrix

$$C = \left[ \begin{array}{c|c} c_{11} & \\ \vdots & \\ c_{in} & \end{array} \right]$$

$v_i^{\text{new}} = \begin{bmatrix} c_{1i} \\ \vdots \\ c_{ni} \end{bmatrix}$  ← coords in the old basis

columns are the coordinates of the new basis vectors w.r.t. the old basis.

We have:

$$X^{\text{old}} = C X^{\text{new}}$$

$$X^{\text{new}} = C^{-1} X^{\text{old}}$$

← easier to see

We proved it earlier in the course. It is easy to see because this relation holds for the new basis vectors, and then it holds for all vectors by linearity.

For linear operators

$A: V \rightarrow V$  - lin. op.

has matrix  $A^{old}$  w.r.t. the old basis  $\{v_1, \dots, v_n\}^{old}$   
 $A^{new}$  ————  $\{v_1, \dots, v_n\}^{new}$ .

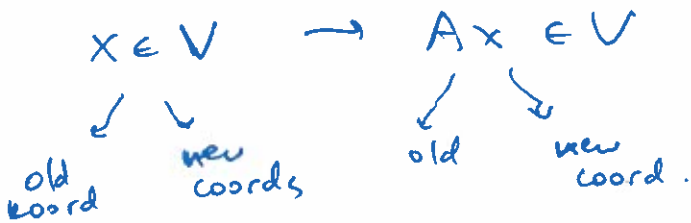
How are these matrices related?

$x^{new} = C^{-1} x^{old}$

$(A^{old} x^{old})^{new} = A^{new} x^{new}$

lin. operator  
 $Ax$

$(A x^{old})^{new}$  makes almost no sense



both are expressions for  $Ax$   
 the same vector, expressed in the new coordinates.

Then:  $C^{-1} A^{old} x^{old} = A^{new} x^{new}$

$C^{-1} A^{old} C x^{new} = A^{new} x^{new}$

$A^{new} = C^{-1} A^{old} C$

Example:  $C = \begin{bmatrix} 2 & 2 \\ 1 & -1 \end{bmatrix}$        $A = \begin{bmatrix} 1 & 4 \\ 1 & 1 \end{bmatrix}$

recall:  $\begin{bmatrix} 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ -1 \end{bmatrix}$  were eigenvectors for A

compute  $C^{-1} A C$ :  $C^{-1} = \begin{bmatrix} -1 & -2 \\ -1 & 2 \end{bmatrix} \cdot \frac{1}{-4}$

using  $\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{\det A} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} = \begin{bmatrix} 1/4 & 1/2 \\ 1/4 & -1/2 \end{bmatrix}$  "det A"

$C^{-1} \begin{bmatrix} 1 & 4 \\ 1 & 1 \end{bmatrix} C = \begin{bmatrix} 1/4 & 1/2 \\ 1/4 & -1/2 \end{bmatrix} \begin{bmatrix} 1 & 4 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 2 \\ 1 & -1 \end{bmatrix}$  ←  $A^{new}$

it does "it" →  $\begin{bmatrix} 3 & 0 \\ 0 & -1 \end{bmatrix}$  4

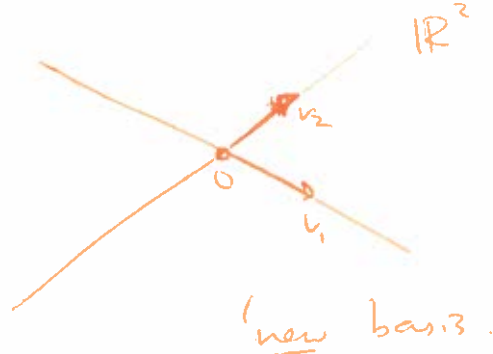
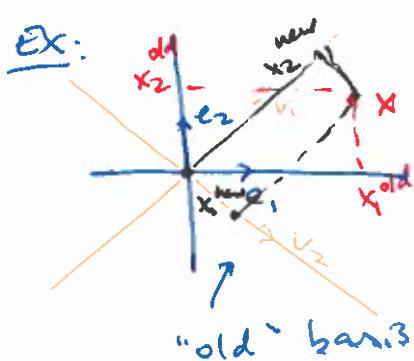
# An illustration for Changes of basis

$V$  - vector space over  $F$ .

$\{v_1^{old}, \dots, v_n^{old}\}$  - a basis of  $V$

$\{v_1^{new}, \dots, v_n^{new}\}$  - a new basis of  $V$ .

(Recall: they have the same number of elements,  $n = \dim(V)$ .)



(often: old basis is  $\{e_1, \dots, e_n\}$  - standard basis of  $F^n$ )

How do the old coords of  $x$  relate to the new coords of  $x$ ?

$$X = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}^{old} = x_1^{old} v_1^{old} + \dots + x_n^{old} v_n^{old}$$

Also

$$X = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}^{new} = x_1^{new} v_1^{new} + \dots + x_n^{new} v_n^{new}$$

Geometrically hard to relate  $\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}^{new}$  to  $\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}^{old}$ .

Above,

we solved this question algebraically:

we make a transition matrix  $C$ :

Columns of  $C$  are the coordinates of the new basis vectors in the old basis.

Example: Old basis:  $\{e_1, e_2\}$  in  $\mathbb{R}^2$

New basis:  $v_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$   $v_2 = \begin{bmatrix} 2 \\ -1 \end{bmatrix}$

$$X = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}^{\text{old}}$$

What are the new coordinates?

↑ usual standard coords.

We make  $C$ :  $\begin{bmatrix} 2 & 2 \\ 1 & -1 \end{bmatrix}$  - our transition matrix

Aside: you could make a mean example:  
 $\{v_1, \dots, v_n\}$  - not the standard basis; given in terms of  $e_1, \dots, e_n$   
 $\{w_1, \dots, w_n\}$  - another one - given in terms of  $e_1, \dots, e_n$ .

Transition from  $\{v_1, \dots, v_n\}$  to  $\{w_1, \dots, w_n\}$ ?

Answer:  $C_1$  = transition matrix from  $\{e_1, \dots, e_n\}$  to  $\{v_1, \dots, v_n\}$

$C_2$  - transition from  $\{e_1, \dots, e_n\}$  to  $\{w_1, \dots, w_n\}$

$$C_1 = \begin{bmatrix} \vdots \\ \vdots \\ \vdots \end{bmatrix}$$

↑ coords of  $v_i$  w.r.t.  $\{e_1, \dots, e_n\}$

$$C_2 = \begin{bmatrix} \vdots \\ \vdots \\ \vdots \end{bmatrix}$$

↑ coords of  $w_i$  w.r.t.  $\{e_1, \dots, e_n\}$

To go from the  $v$ 's to the  $w$ 's:

$$C^{-1} = C_2^{-1} C_1$$

$$\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}^{\text{new}, v} = C_1^{-1} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}^{\text{standard}}$$

$$\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}^w = C_2^{-1} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}^{\text{standard}}$$

We have:

$$X^{\text{old}} = C \cdot X^{\text{new}}$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = C \cdot \begin{bmatrix} x_1^{\text{new}} \\ x_2^{\text{new}} \end{bmatrix} = \begin{bmatrix} 2 & 2 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} x_1^{\text{new}} \\ x_2^{\text{new}} \end{bmatrix}$$



Then

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}^{\text{new}} = C^{-1} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}^{\text{old}}$$

↑ why does  $C^{-1}$  exist?

We know: columns of  $C$  form a basis, so  $\text{rk}(C) = n$   
then  $C^{-1}$  exists (b/c  $\dim(\ker(C)) = n - \text{rk}(C) = 0$ .)