

(103)

11/21/2016 → Lots of properties of det now follow.

COR (4.8.4  
4.8.5  
4.8.6  
4.8.16)

(i)  $\det(AB) = \det(A) \det(B)$

(ii)  $\det(P^{-1}) = \frac{1}{\det P}$

(iii)  $\det(P^{-1}AP) = \det A$

(iv)  $\chi_{P^{-1}AP}(t) = \chi_A(t)$

proof:

(i): If  $\det B = 0$  then  $\exists \vec{v} \neq \vec{0}$  in  $\ker B$  i.e.  $B\vec{v} = \vec{0}$   
so  $AB\vec{v} = \vec{0}$  also  $\Rightarrow \det(AB) = 0$   
("A0") (and (i) holds)Otherwise if  $\det B \neq 0$ , then use column operations  $\rightarrow$   $B \xrightarrow{\text{col ops}} I_n$   
and write  $B = E_1 E_2 \dots E_N$  for some elementary matrices  $E_i$ .We know  $\det B = c_1 c_2 \dots c_N$  where

$$c_i = \begin{cases} 1 & \text{if } E_i \text{ adds a multiple of a column to another} \\ -1 & \text{if } E_i \text{ swaps two columns} \\ c_i & \text{if } E_i \text{ scales some column by } c_i \end{cases}$$

But then we also know

$$\begin{aligned} \det(AB) &= \det(AE_1 E_2 \dots E_N) \\ &= c_1 c_2 \dots c_N \det A \\ &= \det B \cdot \det A, \end{aligned}$$

(ii):  $P^{-1} \cdot P = I_n$

$$\Rightarrow \det(P^{-1} \cdot P) = \det I_n$$

("by (i)")                      "1"

$$\det(P^{-1}) \cdot \det(P)$$

(iii):  $\det(P^{-1}AP) = \underbrace{\det(P^{-1})}_{(ii) = 1/\det P} \cdot \det A \cdot \det P = \det A$

(iv):  $\chi_{P^{-1}AP}(t) = \det(tI_n - P^{-1}AP) = \det(tP^{-1}I_n P - P^{-1}AP)$

$$= \det(P^{-1} \cdot (tI_n - A) P)$$

$$= \det(tI_n - A)$$

$$= \chi_A(t)$$

via (iii), although this is slightly tricky: (iii) applies whenever we choose  $t \in \mathbb{R}$ , not as a polynomial variable. But this shows the polynomial  $\det(tI_n - A)$  of degree  $n$  has more than  $n$  roots, so it is zero!  $\blacksquare$

(w4) There is a lot more to say about det, but let's get back to eigenvectors!

CoR: Every non <sup>(real)</sup> matrix  $A$  has at least one eigenvalue  $\lambda \in \mathbb{C}$  and an accompanying eigenvector  $\vec{v} \in \mathbb{C}^n$  with  $A\vec{v} = \lambda\vec{v}$

proof:  $\chi_A(t) = \det(tI_n - A) = \det \begin{bmatrix} t-a_{11} & -a_{12} & \dots \\ -a_{21} & \ddots & \\ \vdots & & t-a_{nn} \end{bmatrix} = t^n - (a_{11} + a_{22} + \dots + a_{nn})t^{n-1} + \dots \pm \det A$

is a ~~non-constant~~ polynomial in  $t$  of degree  $n$ , so it has at least one root  $t = \lambda \in \mathbb{C}$  by Fundamental Theorem of Algebra.

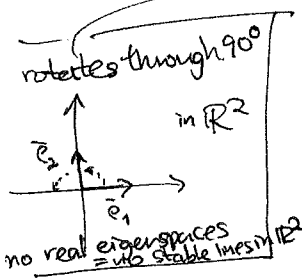
But then one can find  $\vec{v} \in \mathbb{C}^n$  solving  $(A - \lambda I_n)\vec{v} = \vec{0}$  using row reduction over  $\mathbb{C}$

(we only need the fact that every  $z \in \mathbb{C} - \{0\}$  has a multiplicative inverse  $\frac{1}{z} = \frac{1}{x+iy} \cdot \frac{x-iy}{x-iy} = \frac{x-iy}{x^2+y^2}$  for scaling pivot entries to 1)

EXAMPLE:

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

has  $\chi_A(t) = \det \begin{bmatrix} t & -1 \\ +1 & t \end{bmatrix} = t^2 - (-1) = t^2 + 1 = (t+i)(t-i)$  ← no real roots!



$\lambda_1 = +i: \begin{bmatrix} +i & -1 \\ +1 & +i \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow \vec{v}_1 \in \mathbb{C} \begin{bmatrix} -i \\ 1 \end{bmatrix}$

$\lambda_2 = -i: \begin{bmatrix} -i & -1 \\ +1 & -i \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow \vec{v}_2 \in \mathbb{C} \begin{bmatrix} i \\ 1 \end{bmatrix}$

We may not have an eigenbasis for  $A$ , e.g.  $A = \begin{bmatrix} 4 & 0 & 0 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$ ,  $\chi_A(t) = (t-4)(t-1)^2$

was discussed in section -  $\vec{e}_1$  is an eigenvector for  $\lambda_1 = 4$

but  $\vec{e}_2$  and its multiples are the only eigenvectors for  $\lambda_2 = 1$ ; can't find a third lin. indep. eigenvector!

(105)

The double root at  $t=1$  was part of the problem.

Thm. 2.7.4: If  $A$  has eigenvectors  $\vec{v}_1, \dots, \vec{v}_k$  with distinct eigenvalues  $\lambda_1, \dots, \lambda_k$  (i.e.  $\lambda_i \neq \lambda_j \nRightarrow i \neq j$ ) then  $\vec{v}_1, \dots, \vec{v}_k$  are lin. independent.

In particular, if  $\chi_A(t) = (t-\lambda_1)\dots(t-\lambda_n)$  has no repeated roots (i.e.  $\lambda_i \neq \lambda_j \forall 1 \leq i < j \leq n$ )

then  $A$  is diagonalizable, since the  $\lambda_i$ -eigenvectors  $\vec{v}_i$  give an eigenbasis for  $A$ .

Proof: If  $\vec{v}_1, \dots, \vec{v}_k$  are lin. dependent,

write down a nontrivial lin. dependence  $c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k \stackrel{(*)}{=} \vec{0}$

that has the fewest nonzero coefficients, and assume  $c_1 \neq 0$

(by re-indexing, if needed). We'll get a contradiction by manufacturing one with fewer nonzero coefficients:

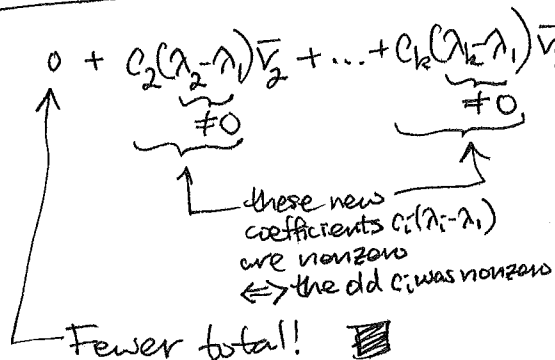
Apply  $A$  to  $\vec{v}_i$  to get  $c_1 A\vec{v}_1 + c_2 A\vec{v}_2 + \dots + c_k A\vec{v}_k = \vec{0}$

i.e.  $c_1\lambda_1\vec{v}_1 + c_2\lambda_2\vec{v}_2 + \dots + c_k\lambda_k\vec{v}_k = \vec{0}$

and subtract

$\lambda_1(x)$  i.e.  $c_1\lambda_1\vec{v}_1 + c_2\lambda_1\vec{v}_2 + \dots + c_k\lambda_1\vec{v}_k = \vec{0}$

giving  $0 + c_2(\lambda_2 - \lambda_1)\vec{v}_2 + \dots + c_k(\lambda_k - \lambda_1)\vec{v}_k = \vec{0}$



11/23/2016 >

REMARK: This already shows most matrices  $A$  are diagonalizable, since

$\chi_A(t)$  usually has distinct roots. In fact, with a little more work, one can write down a polynomial in the entries  $(a_{ij})$  of  $A$  that must vanish to get repeated roots in  $\chi_A(t)$ .

e.g. for  $n=2$   $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$  has repeated roots for  $\chi_A(t) = \det \begin{bmatrix} t-a & -b \\ -c & t-d \end{bmatrix} = t^2 - (a+d)t + ad-bc$

$\Leftrightarrow 0 = B^2 - 4C = (a+d)^2 - 4(ad-bc)$

$= a^2 + 2ad + d^2 - 4ad + 4bc$

$= (a-d)^2 + 4bc$