

## Math 218D Problem Session

Week 9

### 1. Some simple examples

- a) The matrix  $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$  has characteristic polynomial  $(\lambda-1)^2$ , the only eigenvalue is  $\lambda_1 = 1$ , the  $\lambda_1$ -eigenspace is  $\mathbf{R}^2$  with basis  $\{(1,0), (0,1)\}$ , the matrix is diagonal and diagonalizable.
- b) The matrix  $\begin{pmatrix} 2 & 0 \\ 0 & -2 \end{pmatrix}$  has characteristic polynomial  $(\lambda-2)(\lambda+2)$ , the eigenvalues are  $\lambda_1 = 2$  and  $\lambda_2 = -2$ , the  $\lambda_1$ -eigenspace is  $\text{Span}\{(1,0)\}$  and the  $\lambda_2$ -eigenspace is  $\text{Span}\{(0,1)\}$ , the matrix is diagonal and diagonalizable.
- c) The matrix  $\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$  has characteristic polynomial  $\lambda^2$ , the only eigenvalue is  $\lambda_1 = 0$ , the  $\lambda_1$ -eigenspace is  $\mathbf{R}^2$ , the matrix is diagonal and diagonalizable.
- d) The matrix  $\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$  has characteristic polynomial  $(\lambda-1)(\lambda+1)$ , the eigenvalues are  $\lambda_1 = 1$  and  $\lambda_2 = -1$ , the  $\lambda_1$ -eigenspace is  $\text{Span}\{(1,1)\}$  and the  $\lambda_2$ -eigenspace is  $\text{Span}\{(1,-1)\}$ , the matrix is not diagonal but is diagonalizable.
- e) The matrix  $\begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$  has characteristic polynomial  $\lambda(\lambda-2)$ , the eigenvalues are  $\lambda_1 = 0$  and  $\lambda_2 = 2$ , the  $\lambda_1$ -eigenspace is  $\text{Span}\{(1,-1)\}$  and the  $\lambda_2$ -eigenspace is  $\text{Span}\{(1,1)\}$ , the matrix is not diagonal but is diagonalizable.
- f) The matrix  $\begin{pmatrix} 2 & 1 \\ 0 & 2 \end{pmatrix}$  has characteristic polynomial  $(\lambda-2)^2$ , the only eigenvalue is  $\lambda_1 = 2$ , the  $\lambda_2$ -eigenspace is  $\text{Span}\{(1,0)\}$ , the matrix is neither diagonal nor diagonalizable.
- g) The matrix  $\begin{pmatrix} 2 & 1 \\ -1 & 2 \end{pmatrix}$  has characteristic polynomial  $\lambda^2-4\lambda+5$ . Since  $4^2-4\cdot 5 < 0$ , this polynomial has no real root. This means it has no real eigenvalues, and cannot be diagonalized via real matrices. It is not diagonal.

## 2. A $2 \times 2$ diagonalization

Consider the matrix  $A = \begin{pmatrix} -1 & 2 \\ -3 & 4 \end{pmatrix}$ .

a) The characteristic polynomial is  $\lambda^2 - 3\lambda + 2 = (\lambda - 1)(\lambda - 2)$

b) The two eigenvalues are  $\lambda_1 = 1$  and  $\lambda_2 = 2$ .

c) A  $\lambda_1 = 1$  eigenvector is  $(1, 1)$ .

d) A  $\lambda_2 = 2$  eigenvector is  $(2, 3)$ .

e)  $D = \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix}$ ,  $C = \begin{pmatrix} 1 & 2 \\ 1 & 3 \end{pmatrix}$ ,  $A = \begin{pmatrix} 1 & 2 \\ 1 & 3 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 1 & 3 \end{pmatrix}^{-1}$ .

f) It is not hard to "guess" that  $(1, 2) = -(1, 1) + (2, 3)$ , i.e.  $c_1 = -1$ ,  $c_2 = 1$ . If you already computed the inverse  $C^{-1} = \begin{pmatrix} 3 & -2 \\ -1 & 1 \end{pmatrix}$ , you could also do  $(c_1, c_2) = C^{-1}(1, 2) = (3, -1) + 2(-2, 1) = (-1, 1)$ .

This means that  $A^n(1, 2) = A^n(-(1, 1) + (2, 3)) = -\lambda_1^n(1, 1) + \lambda_2^n(2, 3) = -(1, 1) + 2^n(2, 3)$ .

g) When  $n$  is very large, the ratio  $\frac{\|A^{n+1}(1,2)\|^2}{\|A^n(1,2)\|^2} = \frac{(2 \cdot 2^{n+1} + 1)^2 + (3 \cdot 2^{n+1} + 1)^2}{(2 \cdot 2^n + 1)^2 + (3 \cdot 2^n + 1)^2}$  is approximately 4 (the +1's are negligible compared to the large  $2^n$  terms). This means that the ratio  $\frac{\|A^{n+1}(1,2)\|}{\|A^n(1,2)\|}$  is approximately 2.

h) For any  $n$ ,  $\|A^{n+1}(1, 1)\|/\|A^n(1, 1)\|$  is not just approximately, but exactly, equal to 1.

i) If you were given a random vector  $w$ , you would expect  $\|A^{n+1}w\|/\|A^n w\|$  to be approximately 2 when  $n$  is very large - most vectors are not in the  $\lambda_1 = 1$  eigenspace, and for any vector not in that eigenspace, the same logic as in g) would apply.

### 3. Traces and determinants

Recall that the trace  $\text{Tr}(A)$  is the sum of the diagonal entries of  $A$ .

- a) For example, for **a**),  $\lambda_1 = 1$  and  $\lambda_2 = 1$ . Therefore  $\text{Tr}\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = 1 + 1 = 2$ , while  $\det\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = 1 \cdot 1 = 1$ . For a non-diagonal example, look at **d**) - the eigenvalues are  $\lambda_1 = 1, \lambda_2 = -1$ ,  $\text{Tr}\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = 1 + (-1) = 0$  while  $\det\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = 1 \cdot (-1) = -1$ .
- b) For any  $n \times n$  matrix, the polynomial  $p(\lambda) = \det(A - \lambda I_n)$  can be factored as

$$p(\lambda) = (-1)^n(\lambda - \lambda_1) \cdots (\lambda - \lambda_n).$$

When you set  $\lambda = 0$  in  $\det(A - \lambda I_n)$ , you get  $\det(A)$ . When you set  $\lambda = 0$  in  $(-1)^n(\lambda - \lambda_1) \cdots (\lambda - \lambda_n)$ , you get  $(-1)^n(-\lambda_1) \cdots (-\lambda_n) = \lambda_1 \cdot \lambda_n$ . Therefore  $\det(A) = \lambda_1 \cdot \lambda_n$ .

- c) The determinant  $\det(A)$  has another product formula:

$$\det(A) = (-1)^k d_1 \cdots d_n,$$

when the  $A$  has REF with pivot entries  $d_1, \dots, d_n$ , found using Gaussian elimination w/o row scaling and with  $k$  row swaps. Even though this formula looks quite similar to the formula of **b**), eigenvalues and pivots are not at all the same.

An example of a  $2 \times 2$  matrix where the pivots  $d_1, d_2$  are not the same as the eigenvalues  $\lambda_1, \lambda_2$  is given by  $\begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}$ . This matrix has  $p(\lambda) = \lambda^2 - \lambda - 1$ , hence has eigenvalues  $\lambda_1, \lambda_2 = \frac{1+\sqrt{5}}{2}, \frac{1-\sqrt{5}}{2}$ . But the REF, with one row swap, is  $\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$ , with pivots 1, 1. This gives two different formula for the determinant  $\det\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} = -1 \cdot 1 = \frac{1+\sqrt{5}}{2} \cdot \frac{1-\sqrt{5}}{2}$ .

- d) For any  $n \times n$  matrix, we will show that  $\text{Tr}(A) = \lambda_1 + \cdots + \lambda_n$ . We'll do the same strategy as in **b**), but the details are much trickier.

**$p(\lambda)$ -side:** If you expand  $p(\lambda) = (-1)^n(\lambda - \lambda_1) \cdots (\lambda - \lambda_n)$  into  $p(\lambda) = (-1)^n \lambda^n + (?)\lambda^{n-1} + \cdots$ , the coefficient of  $\lambda^{n-1}$  is  $\lambda_1 + \cdots + \lambda_n$ .

For example,  $(-1)^3(\lambda - \lambda_1)(\lambda - \lambda_2)(\lambda - \lambda_3) = (-1)^3(\lambda^3 - (\lambda_1 + \lambda_2 + \lambda_3)\lambda^2 + (\lambda_1\lambda_2 + \lambda_2\lambda_3 + \lambda_3\lambda_1)\lambda - \lambda_1\lambda_2\lambda_3)$ .

**$\det(A - \lambda I)$ -side** What is the coefficient of  $\lambda^{n-1}$  for  $\det(A - \lambda I)$ ? Well, you have to think very carefully about the cofactor expansion, or really the formula you get when you do cofactor expansion  $n$  times, all the way to  $1 \times 1$  matrices. The only term in the cofactor expansion which has a possibility of having a  $\lambda^{n-1}$  term is the product  $(a_{11} - \lambda) \cdots (a_{nn} - \lambda)$ , coming from the  $(1, 1)$ -cofactor  $n$  times.

For example, when  $n = 3$ ,

$$\det \begin{pmatrix} (a_{11} - \lambda) & a_{12} & a_{13} \\ a_{21} & (a_{22} - \lambda) & a_{23} \\ a_{31} & a_{32} & (a_{33} - \lambda) \end{pmatrix} = (a_{11} - \lambda) \det \begin{pmatrix} (a_{22} - \lambda) & a_{23} \\ a_{32} & (a_{33} - \lambda) \end{pmatrix} \\ - a_{21} \det \begin{pmatrix} a_{12} & a_{13} \\ a_{32} & a_{33} - \lambda \end{pmatrix} \\ + a_{31} \det \begin{pmatrix} a_{12} & a_{13} \\ (a_{22} - \lambda) & a_{23} \end{pmatrix}.$$

Both  $\det \begin{pmatrix} a_{12} & a_{13} \\ a_{32} & (a_{33} - \lambda) \end{pmatrix}$  and  $\det \begin{pmatrix} a_{12} & a_{13} \\ (a_{22} - \lambda) & a_{23} \end{pmatrix}$  are degree one polynomials in  $\lambda$ , with no  $\lambda^{n-1} = \lambda^2$  term. The first term

$$(a_{11} - \lambda) \det \begin{pmatrix} (a_{22} - \lambda) & a_{23} \\ a_{32} & (a_{33} - \lambda) \end{pmatrix}$$

equals  $(a_{11} - \lambda)(a_{22} - \lambda)(a_{33} - \lambda) - a_{32}a_{23}$ , and only the first part of this,  $(a_{11} - \lambda)(a_{22} - \lambda)(a_{33} - \lambda)$ , can have  $\lambda^2$  terms.

Back to discussing general  $n$ . Since the  $\lambda^{n-1}$  term of  $\det(A - \lambda I_n)$  is the same as the  $\lambda^{n-1}$  term of  $(a_{11} - \lambda) \cdots (a_{nn} - \lambda)$ ,

$$\det(A - \lambda I_n) = (-1)^{n-1} \lambda^n + (a_{11} + \cdots + a_{nn}) \lambda^{n-1} + \cdots.$$

**Conclusion:** We then compare the  $\lambda^{n-1}$ -terms on both sides of  $\det(A - \lambda I_n) = (-1)^n (\lambda - \lambda_1) \cdots (\lambda - \lambda_n)$ , which gives

$$a_{11} + \cdots + a_{nn} = \lambda_1 + \cdots + \lambda_n,$$

i.e.

$$\text{Tr}(A) = \lambda_1 + \cdots + \lambda_n.$$

#### 4. Linear independence of eigenvectors

- a) Consider a matrix  $A$  with two distinct eigenvalues  $\lambda_1 \neq \lambda_2$ , with associated eigenvectors  $v_1$  and  $v_2$ . We will show that  $v_1$  is not a scalar multiple of  $v_2$ .

Suppose that they were scalar multiples  $v_1 = cv_2$ . Note that  $c \neq 0$ , since the eigenvector  $v_1$  can't be 0. Then  $Av_1 = A(cv_2) = cAv_2$ . Using the eigenvector equations  $Av_1 = \lambda_1 v_1$  and  $Av_2 = \lambda_2 v_2$ , this becomes  $\lambda_1 v_1 = c\lambda_2 v_2$ . Substituting  $v_1 = cv_2$ , this becomes  $\lambda_1(cv_2) = c\lambda_2 v_2$ . As  $v_2$  and  $c$  are not the zero vector/scalar, this implies  $\lambda_1 = \lambda_2$ , a contradiction.

Therefore  $v_1$  and  $v_2$  are not scalar multiples.

- b) Consider a matrix  $A$  with three distinct eigenvalues  $\lambda_1, \lambda_2, \lambda_3$ , with associated eigenvectors  $v_1, v_2$  and  $v_3$ . We will show that  $v_1, v_2$ , and  $v_3$  are linearly independent.

Suppose they were linearly dependent: we would have an equation

$$av_1 + bv_2 + cv_3 = 0,$$

where at least two of the scalars  $a, b, c$  are non-zero. If one of them is zero, we are actually in the situation of a) - we already checked that this was impossible.

Multiplying by  $A$ , we obtain another equation

$$\lambda_1 av_1 + \lambda_2 bv_2 + \lambda_3 cv_3 = 0.$$

Now, we may assume that  $\lambda_1 \neq 0$  (if it is zero, re-order the eigenvalues - the eigenvalues can't all be zero, since they are 3 distinct numbers). We can subtract  $\lambda_1$  times the equation  $av_1 + bv_2 + cv_3 = 0$  from  $\lambda_1 av_1 + \lambda_2 bv_2 + \lambda_3 cv_3 = 0$ , to get

$$(\lambda_2 - \lambda_1)bv_2 + (\lambda_3 - \lambda_1)cv_3 = 0.$$

Since all the eigenvalues were distinct, the coefficients  $(\lambda_2 - \lambda_1)b$  and  $(\lambda_3 - \lambda_1)c$  are both nonzero. Therefore the eigenvectors  $v_2$  and  $v_3$  are scalar multiples of each other. But this is impossible, due to a)!

Since all cases give rise to contradictions, we may conclude that the assumption that  $v_1, v_2$ , and  $v_3$  are linearly dependent is impossible. In other words, any three eigenvectors with distinct eigenvalues must be linearly independent.