

**EE5120 Linear Algebra: Tutorial 6, July-Dec 2017-18**  
Covers sec 4.2, 5.1, 5.2 of GS

1. State True or False with proper explanation:

- (a) All vectors are eigenvectors of the Identity matrix.
- (b) Any matrix can be diagonalized.
- (c) Eigenvalues must be nonzero scalars.
- (d)  $A$  and  $B$  are said to be *Similar* matrices if there exists an invertible matrix  $P$  such that  $P^{-1}AP = B$ .  $A$  and  $B$  always have the same eigenvalues.
- (e) The sum of two eigenvectors of an operator  $T$  is always an eigenvector of  $T$ .

**Solution:**

- (a) **True.** We know,  $S^{-1}AS = \Lambda$ . If  $A = I$ ,  $S^{-1}IS$  is always diagonal ( $\Lambda$  is just  $I$ ). The only requirement is that  $S$  should be invertible.
- (b) **False.** Any matrix with distinct eigenvalues can be diagonalized.
- (c) **False.** They can be zero as well. But, eigenvectors have to be nonzero. Having zero eigenvalue implies that the matrix is non-invertible.
- (d) **True.** If  $A$  and  $B$  are similar, there is some invertible matrix  $P$  such that  $P^{-1}AP = B$ . Thus,  $P^{-1}A = BP^{-1}$  or  $AP = PB$ .  
If  $Av = \lambda v$ , we have  $B(P^{-1}v) = \lambda P^{-1}v$ . Similarly, if  $Bv = \lambda v$ , then we have  $A(Pv) = \lambda Pv$ . Thus both have same eigenvalues  $\lambda$ .
- (e) **False.** For example, vectors  $(1, -1)^t$  and  $(0, 1)^t$  are eigenvectors of the matrix

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

But the sum of them  $(1, 0)^t$  is not an eigenvector of the same matrix.

2. Let  $T$  be the linear operator on  $n \times n$  real matrices defined by  $T(A) = A^t$ . Show that  $\pm 1$  are the only eigenvalues of  $T$ . Describe the eigenvectors corresponding to each eigenvalue of  $T$ .

*Hint: Write the Eigenvalue equation as  $T(A) = A^t = \lambda A$  and proceed.*

**Solution:** If  $T(A) = A^t = \lambda A$  for some  $\lambda$  and some nonzero matrix  $A$ , say  $A_{ij} \neq 0$ , we have

$$A_{ij} = \lambda A_{ji}$$

and

$$A_{ji} = \lambda A_{ij}$$

and so

$$A_{ij} = \lambda^2 A_{ij}$$

This means that  $\lambda$  can be only 1 or -1. And these two values are eigenvalues due to the existence of symmetric and skew-symmetric matrices.

The set of nonzero symmetric matrices are the eigenvectors corresponding to eigenvalue 1, while the set of nonzero skew-symmetric matrices are the eigenvectors corresponding to eigenvalue -1.

3. Prove that the geometric multiplicity of an eigenvalue,  $\gamma_A(\lambda_i)$ , can not exceed its algebraic multiplicity,  $\mu_A(\lambda_i)$ . Thus, from here conclude (and prove that)  $1 \leq \gamma_A(\lambda_i) \leq \mu_A(\lambda_i) \leq n$

**Solution:** See <http://www.ee.iitm.ac.in/uday/2017b-EE5120/multiplicity.pdf>

4. Consider the following  $N \times N$  matrix:

$$\mathbf{A} = \begin{bmatrix} x & -x & 0 & 0 & 0 & 0 & \dots & 0 & 0 & 0 \\ x & x & -x & 0 & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & x & x & -x & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & 0 & x & x & -x & 0 & \dots & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & \dots & \dots & \dots & x & x & -x \\ 0 & 0 & 0 & \dots & \dots & \dots & \dots & 0 & x & x \end{bmatrix}$$

This implies for  $N = 1, 2, 3$ , matrix  $\mathbf{A}$  looks like,

$$[x] \quad \begin{bmatrix} x & -x \\ x & x \end{bmatrix} \quad \begin{bmatrix} x & -x & 0 \\ x & x & -x \\ 0 & x & x \end{bmatrix}$$

Show that the determinant of  $\mathbf{A}$  is  $(F_{N-1} + F_{N-2})x^N$ , where  $F_1 = 1$ ,  $F_2 = 2$  and  $F_N = F_{N-1} + F_{N-2}$ .

*Hint:* Use Mathematical Induction.

**Solution:** (Use Mathematical Induction) It can be easily verified that the result is true for  $N = 1, 2, 3$ . Let us assume that it is true upto dimension  $N - 1 \times N - 1$ , where  $N > 4$ . If we prove that the result holds for when the matrix dimension is  $N \times N$ , then we are done. Carefully observe that the determinant of the  $N \times N$  matrix  $\mathbf{A}$  is given by,

$$\begin{aligned} & x(\det. \text{ of } N - 1 \times N - 1 \text{ matrix}) - (-x)(x(\det. \text{ of } N - 2 \times N - 2 \text{ matrix})) \\ &= x((F_{N-2} + F_{N-3})x^{N-1}) + x^2((F_{N-3} + F_{N-4})x^{N-2}) \\ &= F_{N-1}x^N + F_{N-2}x^N. \end{aligned}$$

Hence proved.

Alt proof: Partition  $A_n$  as  $A_n = \begin{bmatrix} x & -x & \dots \\ x & A_{n-1} & \\ \vdots & & \ddots \end{bmatrix}$ .

OR

(Using row transformations) Using the property that subtracting a multiple of one row from another row leaves the same determinant.

Subtracting  $Row2 = Row2 - Row1$ , followed by  $Row3 = Row3 - \frac{Row2}{2}$ , and soon to matrix  $A$  to get a upper triangular matrix as shown below,

$$\det(\mathbf{A}) = \begin{vmatrix} x & -x & 0 & 0 & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & \frac{2}{1}x & -x & 0 & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & 0 & \frac{3}{2}x & -x & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & 0 & 0 & \frac{5}{3}x & -x & 0 & \dots & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & \frac{8}{5}x & -x & \dots & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \end{vmatrix}$$

It can be observed that the diagonal elements are of the following form  $a_{ii} = \frac{F_{i-1}}{F_{i-2}}x$ .

$$\det(\mathbf{A}) = \begin{vmatrix} x & -x & 0 & 0 & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & \frac{F_2}{F_1}x & -x & 0 & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & 0 & \frac{F_3}{F_2}x & -x & 0 & 0 & \dots & 0 & 0 & 0 \\ 0 & 0 & 0 & \frac{F_4}{F_3}x & -x & 0 & \dots & 0 & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & \dots & \dots & \dots & 0 & \frac{F_{N-1}}{F_{N-2}}x & -x \\ 0 & 0 & 0 & \dots & \dots & \dots & \dots & 0 & 0 & \frac{F_N}{F_{N-1}}x \end{vmatrix}$$

As the matrix is upper triangular, determinant is product of diagonal elements,

i.e.  $\det(\mathbf{A}) = \frac{F_1}{F_0}x \times \frac{F_1}{F_2}x \times \dots \times \frac{F_{N-1}}{F_{N-2}}x \times \frac{F_N}{F_{N-1}}x = F_N x^N = (F_{N-1} + F_{N-2})x^N$

5. Let  $p(\lambda) = \prod_{i=1}^n (\lambda_i - \lambda)$  be the characteristic polynomial of the  $n \times n$  matrix  $\mathbf{A}$ . Derive the characteristic polynomial of  $\mathbf{A}^2 - \mathbf{I}$ , where  $\mathbf{I}$  is an identity matrix of appropriate dimension.  
*Hint: Use properties of eigen values and definition of a characteristic polynomial.*

**Solution:**

Let  $\Lambda$  be such that,

$$\Lambda = \begin{bmatrix} \lambda_1 & 0 & 0 & \dots & 0 \\ 0 & \lambda_2 & 0 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & \lambda_n \end{bmatrix}$$

Then, there exists an invertible matrix  $\mathbf{B}$  such that  $\mathbf{A} = \mathbf{B}\Lambda\mathbf{B}^{-1}$ . Further, we get  $\mathbf{A}^2 =$

$\mathbf{B}\Lambda\mathbf{B}^{-1}\mathbf{B}\Lambda\mathbf{B}^{-1} = \mathbf{B}\Lambda^2\mathbf{B}^{-1}$ . Now, the characteristic polynomial of  $\mathbf{A}^2 - \mathbf{I}$  is given by,

$$\begin{aligned} \det\left((\mathbf{A}^2 - \mathbf{I}) - \lambda\mathbf{I}\right) &= \det(\mathbf{B}\Lambda^2\mathbf{B}^{-1} - (\lambda + 1)\mathbf{I}) = \det(\mathbf{B}\Lambda^2\mathbf{B}^{-1} - (\lambda + 1)\mathbf{B}\mathbf{B}^{-1}) \\ &= \det\left(\mathbf{B}(\Lambda^2 - (\lambda + 1)\mathbf{I})\mathbf{B}^{-1}\right) = \det(\mathbf{B})\det(\Lambda^2 - (\lambda + 1)\mathbf{I})\det(\mathbf{B}^{-1}) \\ &= \det(\Lambda^2 - (\lambda + 1)\mathbf{I}). \end{aligned}$$

Note that  $\Lambda^2 - (\lambda + 1)\mathbf{I}$  is a diagonal matrix. Hence, the characteristic polynomial is  $\prod_{i=1}^n (\lambda_i^2 - (\lambda + 1))$ .

6. Prove that a linear transformation  $\mathbf{T}$  on a finite dimensional vector space is invertible iff zero is not an eigen value of  $\mathbf{T}$

*Hint: Use properties of eigen values.*

**Solution:**  $\mathbf{T}$  is invertible iff  $\det(\mathbf{T}) \neq 0$ .  $\det(\mathbf{T}) = \text{product of eigen values}$ .  $\det(\mathbf{T}) \neq 0 \Rightarrow$  eigen values are non zero

7. (a) What is wrong with this proof that projection matrices have  $\det P = 1$ ?

$$P = A(A^T A)^{-1} A^T \quad \text{so} \quad |P| = |A| \frac{1}{|A^T||A|} |A^T| = 1$$

*Hint: Invertibility.*

- (b) Suppose the 4by4 matrix  $M$  has four equal rows all containing  $a, b, c, d$ . We know that  $\det(M) = 0$ . Find the  $\det(I + M)$  by any method?

$$\det(I + M) = \begin{bmatrix} 1+a & b & c & d \\ a & 1+b & c & d \\ a & b & 1+c & d \\ a & b & c & 1+d \end{bmatrix}$$

*Hint: Use properties of determinants.*

**Solution:**

- (a) The proof is valid only if  $A$  is square and invertible, which is not the case every-time. If  $A$  is not invertible, then  $(A^T A)^{-1} \neq A^{-1} A^{T-1} \Rightarrow |(A^T A)^{-1}| \neq \frac{1}{|A||A^T|}$ .
- (b) Using the property that subtracting a multiple of one row from another row leaves the same determinant.

$$\det(I + M) = \begin{bmatrix} 1+a & b & c & d \\ -1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 1 \end{bmatrix}$$

Using the property that subtracting a multiple of one column from another column leaves the same determinant.

$$\det(I + M) = \begin{vmatrix} 1 + a + b + c + d & b & c & d \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{vmatrix} = 1 + a + b + c + d$$

8. Find the eigenvalues and eigenvectors for both of these Markov matrices  $A$  and  $A^\infty$ . Explain why  $A^{100}$  is close to  $A^\infty$ :

$$A = \begin{bmatrix} .6 & .2 \\ .4 & .8 \end{bmatrix} \quad A^\infty = \begin{bmatrix} 1/3 & 1/3 \\ 2/3 & 2/3 \end{bmatrix}$$

*Hint:* Use diagonalization.

**Solution:** Eigen values and eigenvectors of  $A$  are 0.4, 1 and  $\begin{bmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$ ,  $\begin{bmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix}$ , respectively.

Eigen values and eigenvectors of  $A^\infty$  are 0, 1 and  $\begin{bmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$ ,  $\begin{bmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix}$ , respectively.

We could see that the eigenvectors are linearly independent for  $A$ . So, matrix  $A$  can be diagonalizable as shown below.

$$A = S\Lambda S^{-1}$$

where  $S = \begin{bmatrix} -1/\sqrt{2} & 1/\sqrt{5} \\ 1/\sqrt{2} & 2/\sqrt{5} \end{bmatrix}$  and  $\Lambda = \begin{bmatrix} 0.4 & 0 \\ 0 & 1 \end{bmatrix}$ .

$$\Rightarrow A^2 = S\Lambda S^{-1}S\Lambda S^{-1} = S\Lambda^2 S^{-1}$$

Similarly,

$$\Rightarrow A^n = S\Lambda^n S^{-1} \quad \text{and} \quad A^\infty = S\Lambda^\infty S^{-1} = S\Lambda^\infty S^{-1}$$

where  $\Lambda^\infty = \begin{bmatrix} 0.4^\infty & 0 \\ 0 & 1^\infty \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$ . So, the eigen values with magnitude less than the one, will have less significance at higher powers.  $0.4^{100} = 1.6069 \times 10^{-40} \approx 0$ .

9. When  $a + b = c + d$ , show that  $(1,1)$  is an eigenvector and find both eigenvalues:

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

*Hint:* Use definition of eigen vector,  $Ax = \lambda x$  and substitute given vector for  $x$ .

**Solution:** Let  $a + b = c + d = f$ . Eigen vector is the solution to the equation  $Ax = \lambda x$ . If  $(1,1)$  is an eigen vector, then

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} a + b \\ c + d \end{bmatrix} = \begin{bmatrix} f \\ f \end{bmatrix} = f \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

which satisfies the equation  $Ax = \lambda x$  with eigen value  $\lambda = f = a + b = c + d$ .

10. EXTRA: Find  $u(t)$  that satisfies the differential equation  $du/dt = Pu$ , when  $P$  is a projection:

$$\frac{du}{dt} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} u \quad \text{with} \quad u(0) = \begin{bmatrix} 5 \\ 3 \end{bmatrix}$$

Here  $u(t)$  is a vector of time-varying functions, i.e., we can write  $u(t) = \begin{bmatrix} v(t) \\ w(t) \end{bmatrix}$ . You will find that a part of  $u$  increases exponentially while another part stays constant.

*Hint:* Find eigen values and eigen vectors of  $P$  and substitute given initial condition.

**Solution:** Solving  $\det(P - \lambda I) = 0$  gives the eigen values of  $P$  as  $\lambda_1 = 1$  and  $\lambda_2 = 0$  (This is true for all projection matrices).

Solving  $Px = \lambda x$  for each of the eigen values gives the corresponding eigen vectors as

$$x_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \text{ and } x_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}.$$

In the differential equation, this produces the special solutions  $u = e^{\lambda t} x$ . They are the pure exponential solutions to  $du/dt = Pu$ . Let these be  $u_1 = e^{\lambda_1 t} x_1$  and  $u_2 = e^{\lambda_2 t} x_2$ . Any linear combinations of  $u_1$  and  $u_2$  will also be solutions to the differential equation. The complete solution is given by  $u(t) = c_1 e^{\lambda_1 t} x_1 + c_2 e^{\lambda_2 t} x_2$ . We now find  $c_1$  and  $c_2$  using the initial condition  $u(0) = \begin{bmatrix} 5 \\ 3 \end{bmatrix}$ , i.e.,

$$c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 5 \\ 3 \end{bmatrix}$$

or

$$\begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 5 \\ 3 \end{bmatrix}$$

This gives  $c_1 = 4$  and  $c_2 = 1$ . Therefore the complete solution is  $u(t) = 4e^t \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 1 \\ -1 \end{bmatrix}$ .

Here, the first part of  $u$  increases exponentially while the nullspace part (corresponding to eigen value 0) remains fixed.