

# SI231 - Matrix Computations, 2022 Fall

## Homework Set #4

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### Acknowledgements:

- 1) Deadline: **2022-12-02 23:59:59**
  - 2) **Late Policy details** can be found on piazza.
  - 3) Submit your homework in **Homework 4** on **Gradescope**. Entry Code: **4V2N55**. **Make sure that you have correctly select pages for each problem.** If not, you probably will get 0 point.
  - 4) No handwritten homework is accepted. You need to write  $\text{\LaTeX}$ . (If you have difficulties in using  $\text{\LaTeX}$ , you are allowed to use **MS Word or Pages** for the first and the second homework to accommodate yourself.)
  - 5) Use the given template and give your solution in English. Solution in Chinese is not allowed.
  - 6) Your homework should be uploaded in the PDF format, and the naming format of the file is not specified.
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## I. EIGENVALUE

**Problem 1.** (8 points + 2 points + 12 points + 8 points) For the matrix below

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

- 1) Calculate the characteristic polynomial of  $A$ ,
- 2) Find the eigenvalues of  $A$ ,
- 3) Find a basis for each eigenspace of  $A$ ,
- 4) Determine whether or not  $A$  is diagonalizable. If  $A$  is diagonalizable, then find an invertible matrix  $P$  and a diagonal matrix  $D$  such that  $P^{-1}AP = D$ .

**Solution:**

- 1) The characteristic polynomial of  $A$  is

$$\begin{aligned} \det(A - \lambda I) &= \begin{vmatrix} -1 - \lambda & -3 & -3 \\ 3 & 5 - \lambda & 3 \\ -1 & -1 & 1 - \lambda \end{vmatrix} = (-1) \begin{vmatrix} 1 + \lambda & 3 & 3 \\ 3 & 5 - \lambda & 3 \\ -1 & -1 & 1 - \lambda \end{vmatrix} \\ &= (-1) \begin{vmatrix} 0 & 2 - \lambda & 4 - \lambda^2 \\ 0 & 2 - \lambda & 6 - 3\lambda \\ -1 & -1 & 1 - \lambda \end{vmatrix} \\ &= (-1)(-1) \begin{vmatrix} 2 - \lambda & 4 - \lambda^2 \\ 2 - \lambda & 6 - 3\lambda \end{vmatrix} = (2 - \lambda) \begin{vmatrix} 1 & 4 - \lambda^2 \\ 1 & 6 - 3\lambda \end{vmatrix} \\ &= (2 - \lambda) (6 - 3\lambda - (4 - \lambda^2)) = (2 - \lambda) (\lambda^2 - 3\lambda + 2) \\ &= -(\lambda - 1)(\lambda - 2)^2. \end{aligned}$$

(8 points)

- 2)  $A$  has eigenvalues 1 and 2, with algebraic multiplicities 1 and 2 respectively. (2 points)
- 3) The eigenspace of  $A$  associated to the eigenvalue 1 is the null space of the matrix  $A - I$ . To find a basis for the eigenspace, row reduce this matrix.

$$A - I = \begin{bmatrix} -2 & -3 & -3 \\ 3 & 4 & 3 \\ -1 & -1 & 0 \end{bmatrix} \longrightarrow \cdots \begin{bmatrix} 1 & 0 & -3 \\ 0 & 1 & 3 \\ 0 & 0 & 0 \end{bmatrix}$$

Thus, the general solution to the equation  $(A - I)\mathbf{x} = \mathbf{0}$  is  $\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3 \\ -3 \\ 1 \end{bmatrix} x_3$  where  $x_3$  is arbitrary. Letting

$x_3 = 1$  gives  $\mathcal{B}_1 = \left\langle \begin{bmatrix} 3 \\ -3 \\ 1 \end{bmatrix} \right\rangle$  as a basis of the eigenspace associated to the eigenvalue 1. The eigenspace of  $A$  associated to the eigenvalue 2 is the null space of the matrix  $A - 2I$ . To find a basis for the eigenspace, row reduce this matrix.

$$A - 2I = \begin{bmatrix} -3 & -3 & -3 \\ 3 & 3 & 3 \\ -1 & -1 & -1 \end{bmatrix} \longrightarrow \cdots \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

Thus, the general solution to the equation  $(A - 2I)\mathbf{x} = \mathbf{0}$  is

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} x_2 + \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} x_3$$

where  $x_2$  and  $x_3$  are arbitrary. Thus  $\mathcal{B}_2 = \left\langle \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\rangle$  as a basis of the eigenspace associated to the eigenvalue 2. (12 points)

- 4)  $A$  is diagonalizable since there is a basis of  $\mathbb{R}^3$  consisting of eigenvectors of  $A$ . Specifically, concatenate  $\mathcal{B}_1$  and  $\mathcal{B}_2$  to get such a basis

$$\mathcal{B} = \left\langle \begin{bmatrix} 3 \\ -3 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\rangle$$

If we set

$$P = \begin{bmatrix} \frac{3}{\sqrt{19}} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ -\frac{3}{\sqrt{19}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{19}} & 0 & \frac{1}{\sqrt{2}} \end{bmatrix}$$

then  $P$  is invertible and

$$P^{-1}AP = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}.$$

(8 points)

## II. EIGENSPACE

### Problem 2. (20 points)

Let  $T : \mathbb{R}^{2 \times 2} \rightarrow \mathbb{R}^{2 \times 2}$  be the linear transformation given by  $T(A) = A^T$  where  $A^T$  is the transpose of  $A$ .

If  $\exists \lambda \in \mathbb{R}, X \in \mathbb{R}^{2 \times 2} \implies T(X) = \lambda X$ , then  $\lambda$  is the eigenvalue of  $T$  and  $X$  is the eigenvector associated to the eigenvalue  $\lambda$ . Please find the eigenvalues of  $T$  and the dimensions of the eigenspaces.

### Solution:

This can be done by writing a matrix of  $A$ , but it can actually be done directly. Suppose we have

$$\begin{pmatrix} a & c \\ b & d \end{pmatrix} = T \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \lambda \begin{pmatrix} a & b \\ c & d \end{pmatrix}.$$

Then,  $a = \lambda a, c = \lambda b, b = \lambda c$ , and  $d = \lambda d$ . If  $a$  or  $d$  is nonzero, these imply immediately that  $\lambda = 1$ . Otherwise, either  $c$  or  $b$  is not zero, then either  $c = \lambda b = \lambda^2 c$  or  $b = \lambda^2 b$  implies that  $\lambda = \pm 1$ . Thus, the eigenvalues of  $T$  are 1 and -1.

For  $\lambda = 1$ , we must have  $c = b$  and no other conditions. Thus, the eigenspace for  $\lambda = 1$  is

$$\left\{ \begin{pmatrix} a & b \\ b & d \end{pmatrix} : a, b, d \in \mathbb{R} \right\} = \text{Span} \left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \right\}$$

and this eigenspace has dimension equal to 3.

For  $\lambda = -1$ , we must have  $a = 0$  since  $a = -a$  and similarly  $d = 0$ . We also have  $b = -c$ . Thus, the eigenspace is

$$\left\{ \begin{pmatrix} 0 & b \\ -b & 0 \end{pmatrix} : b \in \mathbb{R} \right\} = \text{Span} \left\{ \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} \right\}$$

and this eigenspace has dimension equal to 1. (20 points)

### III. EIGENVECTORS OF SYMMETRIC MATRIX

**Problem 3. (8 points + 12 points + 15 points)** Let  $\alpha, \beta \in \mathbb{R}^n$  be two linearly independent vectors, with unit norm ( $\|\alpha\|_2 = \|\beta\|_2 = 1$ ). Define the symmetric matrix  $\mathbf{A} = \alpha\beta^T + \beta\alpha^T$ .

- 1) Prove that  $\alpha + \beta$  and  $\alpha - \beta$  are eigenvectors of  $\mathbf{A}$ , and determine the corresponding eigenvalues.

**Hint:** The notation  $c = \alpha^T \beta$  may be useful.

- 2) Find the nullspace and rank of  $\mathbf{A}$ .
- 3) Find an eigenvalue decomposition of  $\mathbf{A}$ , in terms of  $\alpha, \beta$ .

**Solution:**

- 1)  $\mathbf{A}\alpha = c\alpha + \beta$ ,  $\mathbf{A}\beta = \alpha + c\beta$ , then  $\mathbf{A}(\alpha + \beta) = (c + 1)(\alpha + \beta)$ ,  $\mathbf{A}(\alpha - \beta) = (c - 1)(\alpha - \beta)$ , where  $c = \alpha^T \beta$ .

Therefore,  $\alpha + \beta$  and  $\alpha - \beta$  are eigenvectors of  $\mathbf{A}$ , with eigenvalues  $c + 1, c - 1$ . (8 points)

- 2) For any  $\mathbf{x} \in \mathcal{N}(\mathbf{A})$ ,  $\mathbf{A}\mathbf{x} = \mathbf{0}$ .  $\mathbf{A}\mathbf{x} = \alpha(\beta^T \mathbf{x}) + \beta(\alpha^T \mathbf{x}) = \mathbf{0}$ .

Since  $\beta^T \mathbf{x}$  and  $\alpha^T \mathbf{x}$  are scalars, we can rewrite above equation as:  $\mathbf{A}\mathbf{x} = (\beta^T \mathbf{x})\alpha + (\alpha^T \mathbf{x})\beta = \mathbf{0}$ .

We have  $\alpha, \beta$  are linearly independent, then  $\beta^T \mathbf{x} = \alpha^T \mathbf{x} = 0$ . So  $\mathcal{N}(\mathbf{A})$  is the set of vectors orthogonal to  $\alpha$  and  $\beta$ ,  $\mathcal{N}(\mathbf{A}) = \text{span}\{\alpha, \beta\}^\perp$ . (6 points)

According to the fact of  $\mathbf{A}$  is symmetric,  $\mathcal{R}(\mathbf{A}) = \mathcal{R}(\mathbf{A}^T) = \mathcal{N}(\mathbf{A})^\perp = \text{span}\{\alpha, \beta\}$ .

Therefore,  $\text{rank}(\mathbf{A}) = 2$ . (6 points)

- 3) First, we need to prove that  $\lambda_{1,2} = c \pm 1 \neq 0$ . Since  $\alpha, \beta$  are two linearly independent vectors and  $c = \alpha^T \beta = \langle \alpha, \beta \rangle$ ,  $|c| \leq \|\alpha\|_2 \|\beta\|_2 = 1$ , while equality holds only when  $\alpha = \pm \beta$ . Therefore,  $|c| < 1$ ,  $\lambda_{1,2} \neq 0$ . Thus, we have found two linearly independent eigenvectors  $\xi_1 = \alpha + \beta$  and  $\xi_2 = \alpha - \beta$  that do not belong to the nullspace of  $\mathbf{A}$ . (6 points)

Then, the eigenvalue decomposition of  $\mathbf{A}$  is  $\mathbf{A} = (c + 1)\nu_1\nu_1^T + (c - 1)\nu_2\nu_2^T$ ,

where  $\nu_1 = \xi_1 / \|\xi_1\|_2 = (\alpha + \beta) / \sqrt{2 + 2c}$ ,  $\nu_2 = \xi_2 / \|\xi_2\|_2 = (\alpha - \beta) / \sqrt{2 - 2c}$ . (6 points)

Hence the eigenvalue decomposition of  $\mathbf{A}$  becomes  $\mathbf{A} = \frac{1}{2}((\alpha + \beta)(\alpha + \beta)^T - (\alpha - \beta)(\alpha - \beta)^T)$ . (3 points)

## IV. DIAGONALIZATION

**Problem 4. (13 points + 12 points)** Let  $A$  be a real symmetric  $n \times n$  matrix with 0 as a simple eigenvalue (that is, the algebraic multiplicity of the eigenvalue 0 is 1), and given a vector  $\mathbf{v} \in \mathbb{R}^n$ .

- 1) Prove that for sufficiently small positive real  $\epsilon$  (Equivalently,  $\epsilon$  is smaller than all the absolute value of nonzero eigenvalues), the equation  $A\mathbf{x} + \epsilon\mathbf{x} = \mathbf{v}$  has a unique solution  $\mathbf{x} = \mathbf{x}(\epsilon) \in \mathbb{R}^n$
- 2) Evaluate  $\lim_{\epsilon \rightarrow 0^+} \epsilon\mathbf{x}(\epsilon)$  in terms of  $\mathbf{v}$ , the eigenvectors of  $A$ , and the inner product  $\langle \cdot, \cdot \rangle$  on  $\mathbb{R}^n$ .

**Solution:**

- 1) Recall that the eigenvalues of a real symmetric matrices are all real numbers and it is diagonalizable by an orthogonal matrix. Note that the equation  $A\mathbf{x} + \epsilon\mathbf{x} = \mathbf{v}$  can be written as

$$(A + \epsilon I)\mathbf{x} = \mathbf{v}, \quad (*)$$

where  $I$  is the  $n \times n$  identity matrix. Thus to show that the equation  $(*)$  has a unique solution, it suffices to show that the matrix  $A + \epsilon I$  is invertible. Since  $A$  is diagonalizable, there exists an invertible matrix  $S$  such that

$$S^{-1}AS = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix},$$

where  $\lambda_i$  are eigenvalues of  $A$ .

Since the algebraic multiplicity of 0 is 1, without loss of generality, we may assume that  $\lambda_1 = 0$  and  $\lambda_i, i > 1$  are nonzero.

Then we have

$$S^{-1}(A + \epsilon I)S = S^{-1}AS + \epsilon I = \begin{bmatrix} \epsilon & 0 & \cdots & 0 \\ 0 & \epsilon + \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \epsilon + \lambda_n \end{bmatrix}.$$

If  $\epsilon > 0$  is smaller than the lengths of  $|\lambda_i|, i > 1$ , then none of the diagonal entries  $\epsilon + \lambda_i$  are zero. Hence we have

$$\begin{aligned} \det(A + \epsilon I) &= \det(S)^{-1} \det(A + \epsilon I) \det(S) \\ &= \det(S^{-1}(A + \epsilon I)S) \\ &= \epsilon(\epsilon + \lambda_2) \cdots (\epsilon + \lambda_n) \neq 0. \end{aligned}$$

Since  $\det(A + \epsilon I) \neq 0$ , it yields that  $A + \epsilon I$  is invertible, hence the equation  $*$  has a unique solution

$$\mathbf{x}(\epsilon) = (A + \epsilon I)^{-1}\mathbf{v}.$$

**Remark:** This result is in general true for any square matrix. Instead of using the diagonalization, we can use the triangulation of a matrix.

2) Let  $A = V\Lambda V^T$  and  $v_i$  is the  $i$ -th column vector of  $V$ .

Then we compute

$$\begin{aligned} A\mathbf{x}(\epsilon) &= A(A + \epsilon I)^{-1}\mathbf{v} \\ &= V\Lambda V^T V(\Lambda + \epsilon I)^{-1}V^T v \\ &= V\Lambda(\Lambda + \epsilon I)^{-1}V^T v \\ &= \sum_{i=1}^n c_i \frac{\lambda_i}{\lambda_i + \epsilon} v_i \end{aligned}$$

with  $c_i = \langle v_i, v \rangle$ .

Therefore we have

$$\begin{aligned} \lim_{\epsilon \rightarrow 0^+} \epsilon \mathbf{x}(\epsilon) &= \lim_{\epsilon \rightarrow 0^+} (\mathbf{v} - A\mathbf{x}(\epsilon)) \\ &= \mathbf{v} - \lim_{\epsilon \rightarrow 0^+} (A\mathbf{x}(\epsilon)) \\ &= \sum_{i=1}^n c_i \mathbf{v}_i - \lim_{\epsilon \rightarrow 0^+} \left( \sum_{i=2}^n c_i \frac{\lambda_i}{\lambda_i + \epsilon} \mathbf{v}_i \right) \\ &= \sum_{i=1}^n c_i \mathbf{v}_i - \sum_{i=2}^n c_i \mathbf{v}_i \\ &= c_1 \mathbf{v}_1 \end{aligned}$$

Using the orthonormality of the basis  $\{\mathbf{v}_i\}$ , we have

$$\langle \mathbf{v}, \mathbf{v}_1 \rangle = \sum_{i=1}^n \langle c_i \mathbf{v}_i, \mathbf{v}_1 \rangle = c_1.$$

Hence the required expression is

$$\lim_{\epsilon \rightarrow 0^+} \epsilon \mathbf{X}(\epsilon) = \langle \mathbf{v}, \mathbf{v}_1 \rangle \mathbf{v}_1,$$

where  $\mathbf{v}_1$  is the unit eigenvector corresponding to the eigenvalue 0.