

## Homework 6

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### Question #1

- a) If  $\lambda$  is an eigenvalue of  $A$ , then by definition,  $A\vec{v} = \lambda\vec{v}$  where the corresponding non-zero vector  $\vec{v} \neq 0$  is called the eigenvector of  $A$ .

Now,  $B = cA + dI$  then:

$$B\vec{v} = (cA + dI)\vec{v} = \underbrace{cA\vec{v}}_{A\vec{v} = \lambda\vec{v}} + d\vec{v} = c\lambda\vec{v} + d\vec{v} = (c\lambda + d)\vec{v}$$

$\therefore c\lambda + d$  is an eigenvalue of  $B = cA + dI$ .

- b) By mathematical induction:

Consider the base case where  $k=1$ :

$$A^1\vec{v} = \lambda^1\vec{v} \Leftrightarrow A\vec{v} = \lambda\vec{v}$$

which holds true per definition.

Now, assume that for some positive integer  $k=n$  where  $A^n\vec{v} = \lambda^n\vec{v}$ .

We want to prove that for  $k=n+1$  then  $A^{n+1}\vec{v} = \lambda^{n+1}\vec{v}$ .

$$A^{n+1}\vec{v} = A^n(A\vec{v}) \Leftrightarrow A^{n+1}\vec{v} = A(A^n\vec{v})$$

Using the assumption  $A^n\vec{v} = \lambda^n\vec{v}$ :

$$A^{n+1}\vec{v} = A(\lambda^n\vec{v})$$

$$A^{n+1}\vec{v} = \lambda^n(A\vec{v}) \quad \text{substitute base case}$$

$$A^{n+1}\vec{v} = \lambda^n(\lambda\vec{v})$$

$$A^{n+1}\vec{v} = \lambda^{n+1}\vec{v}$$

Since we have proven the base case & shown that if the statement holds for  $k=n$ , it also holds for  $k=n+1$ .

$\therefore$  By induction:  $A^k\vec{v} = \lambda^k\vec{v}$  for all positive  $k$ .  $\square$

c) By definition, we know that if  $\lambda$  is an eigenvalue of  $A$ , then  $A\vec{v} = \lambda\vec{v}$  where the corresponding non-zero vector  $\vec{v}$  is the eigenvector of  $A$ .

This means that for  $\lambda=0$ , we have  $A\vec{v} = 0\vec{v} = \vec{0}$ .

If  $A$  is singular (non-invertible), then  $A$  must have a  $\det(A)=0$  & a non-trivial kernel,  $\ker(A) \neq \{\vec{0}\}$ . As such, there exists a non-zero vector  $\vec{v} \in \ker(A)$ . Subsequently, the subspace consisting of eigenvector (eigenspace) with  $\lambda=0$  is precisely  $\ker(A)$  with the same dimension since kernel is the number of linearly independent solutions to  $A\vec{v} = \vec{0}$  that spans the subspace.

Conversely,  $A$  must be singular for  $\lambda=0$  to be its eigenvalue. If  $A$  is nonsingular (invertible), then  $A$  has a trivial kernel where there exists no non-zero vector  $\vec{v}$  such that  $A\vec{v} = \vec{0}$ .

d) Let  $A$  be an  $n \times n$  matrix where each entry is equal to 1:

$$A = \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix}$$

Observation:  $\text{rank}(A) = 1$  b/c all columns are multiple of each other &  $A$  has a non-trivial kernel. As per part c,  $A$  is clearly singular, thus  $\lambda=0$  is an eigenvalue of  $A$ .

Using the rank-nullity theorem, we get nullity of  $A$  as  $n - \text{rank}(A) = n-1$  which is equivalent to saying  $\dim(\ker(A)) = n-1$ . From part c, we found that the dimension of eigenspace corresponding to the eigenvalue  $\lambda=0$  is the nullity of  $A$  with multiplicity  $n-1$ . This means there exists one other non-zero eigenvalue.

continue

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Now, to find the non-zero eigenvalue & its vector:

Since the sum of each row of  $A$  is  $n$  as shown:

$$A\vec{v} = \begin{bmatrix} \sum_{i=1}^n v_i \\ \sum_{i=1}^n v_i \\ \vdots \\ \sum_{i=1}^n v_i \end{bmatrix}$$

we can take  $\vec{v} = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$  a column vector of all 1's then we have:

$$A\vec{v} = \begin{bmatrix} 1 & 1 & \dots & 1 \\ 1 & 1 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & \dots & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} = \begin{bmatrix} n \\ n \\ \vdots \\ n \end{bmatrix} = n \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} = \lambda \vec{v}$$

$\therefore \lambda = n$  is an eigenvalue of  $A$ , with its corresponding eigenvector  $\vec{v} = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$

Lastly, as already stated,  $\lambda = 0$  is also an eigenvalue of  $A$ , and its corresponding eigenvector is any non-zero  $\vec{v} \in \mathbb{R}^n$  such that  $A\vec{v} = \vec{0}$ .

e) By definition, we know that  $A\vec{v} = \lambda\vec{v}$ .

Now, if  $A$  is a nonsingular matrix, then applying  $A^{-1}$  to both sides of  $A\vec{v} = \lambda\vec{v}$  gives:

$$A^{-1}A\vec{v} = A^{-1}\lambda\vec{v} \Rightarrow \vec{v} = A^{-1}\lambda\vec{v}$$

Dividing both sides by  $\lambda$ :

$$\frac{1}{\lambda}\vec{v} = A^{-1}\vec{v} \Leftrightarrow A^{-1}\vec{v} = \lambda^{-1}\vec{v}$$

$\therefore \lambda^{-1}$  is an eigenvalue of  $A^{-1}$  with  $\lambda \neq 0$  as  $A$  is invertible.



## Question #2

a) Since  $u$  is a unit vector, we know that  $\|u\|=1$  &  $u^T u = 1$ .

We also know that  $A = uu^T$  which is constructed in such a way that the columns of  $A$  are spanned by  $\vec{u}$   $\therefore \text{rank}(A) = 1$ .

This problem is similar to part 1c & part 1d, where we concluded that such a singular  $\text{rank}(A)=1$  matrix contains an eigenvalue of  $\lambda=0$  with multiplicity  $n-1$  & one other non-zero eigenvalue.

To find the non-zero eigenvalue, we can let  $\vec{v} = \vec{u}$ , then:

$$A\vec{v} = \lambda\vec{v} \Rightarrow (\underbrace{\vec{u}\vec{u}^T}_1)\vec{u} = \lambda\vec{u} \Rightarrow \vec{u} = \lambda\vec{u}$$

which is equivalent to saying  $A\vec{u} = (1)\vec{u}$  as  $A\vec{u} = \lambda\vec{u}$ .

$\therefore \lambda=1$  is the eigenvalue of  $A$  with the corresponding eigenvector  $\vec{u}$ .

Similarly, for the eigenvalue of  $\lambda=0$ , its corresponding eigenvector is any non-zero  $\vec{v} \in \mathbb{R}^n$  such that  $A\vec{v} = \vec{0}$ .

b) Using the eigenvalue equation of  $A\vec{v} = \lambda\vec{v}$ , we can substitute  $A = H = (I - 2\vec{u}\vec{u}^T)$ :

$$(I - 2\vec{u}\vec{u}^T)\vec{v} = \lambda\vec{v}$$

$$\Rightarrow \vec{v} - 2(\vec{u}\vec{u}^T\vec{v}) = \lambda\vec{v} \quad \text{then rearrange}$$

$$\Rightarrow \vec{v} - \lambda\vec{v} = 2(\vec{u}\vec{u}^T\vec{v})$$

$$\Rightarrow (1 - \lambda)\vec{v} = 2(\vec{u}^T\vec{v})\vec{u}$$

Here, we can consider 2 cases:  $\vec{v}$  is parallel to  $\vec{u}$  &  $\vec{v}$  is orthogonal to  $\vec{u}$ .

case 1 ( $\vec{v}$  is parallel to  $\vec{u}$ ):

$$\text{Let } \vec{v} = \vec{u}:$$

$$(1 - \lambda)\vec{u} = 2(\underbrace{\vec{u}^T\vec{u}}_1)\vec{u} \Rightarrow (1 - \lambda)\vec{u} = 2\vec{u} \Rightarrow 1 - \lambda = 2 \Rightarrow \lambda = -1$$

$\therefore A\vec{v} = H\vec{v} = -\vec{u}$  where  $\vec{u}$  is the corresponding eigenvector to  $\lambda = -1$ .

case 2 ( $\vec{v}$  is orthogonal to  $\vec{u}$ ):

If  $\vec{v}$  is orthogonal to  $\vec{u}$ , then  $\vec{u}^T\vec{v} = 0$ .

$$(1 - \lambda)\vec{v} = 2(\underbrace{\vec{u}^T\vec{v}}_0)\vec{u} \Rightarrow (1 - \lambda)\vec{v} = 2(0)\vec{u} \Rightarrow (1 - \lambda)\vec{v} = \vec{0} \Rightarrow 1 - \lambda = 0$$

$\therefore \lambda = 1$  where its corresponding eigenvectors are all  $\vec{v}$  orthogonal to  $\vec{u}$ .

c) The eigenvalue equation respective to  $P$  is:  $P\vec{v} = \lambda\vec{v}$

Now, applying  $P^2 = P$  to the equation:

$$P^2\vec{v} = P\vec{v}$$

Substituting  $P\vec{v} = \lambda\vec{v}$ :

$$P(\lambda\vec{v}) = \lambda\vec{v}$$

$$\lambda P\vec{v} = \lambda\vec{v}$$

$$\lambda(\lambda\vec{v}) = \lambda\vec{v}$$

$$\lambda^2\vec{v} = \lambda\vec{v}$$

$$\lambda^2\vec{v} - \lambda\vec{v} = 0$$

$$\lambda^2 - \lambda = 0 \Leftrightarrow \lambda(\lambda - 1) = 0$$

$\therefore \lambda = 0$  &  $\lambda = 1$  where  $\lambda = 0$  corresponding eigenvectors is orthogonal of the subspace  
&  $\lambda = 1$  is the corresponding to eigenvectors in the subspace where  $P$  projects.

### Question #3

Finding the eigenvalues of the matrix  $A = \begin{bmatrix} 0 & c & -b \\ -c & 0 & a \\ b & -a & 0 \end{bmatrix}$

$$\begin{aligned} \det(A - \lambda I) &= \det \begin{bmatrix} -\lambda & c & -b \\ -c & -\lambda & a \\ b & -a & -\lambda \end{bmatrix} = -\lambda \det \begin{bmatrix} -\lambda & a \\ -a & -\lambda \end{bmatrix} - c \begin{bmatrix} -c & a \\ b & -\lambda \end{bmatrix} - b \begin{bmatrix} -c & -\lambda \\ b & -a \end{bmatrix} \\ &= -\lambda(\lambda^2 + a^2) - c(c\lambda - ab) - b(ca + b\lambda) \\ &= -\lambda^3 - a^2\lambda - c^2\lambda + abc - abc - b^2\lambda \\ &= -\lambda^3 - a^2\lambda - c^2\lambda - b^2\lambda \\ &= -\lambda(\lambda^2 + a^2 + c^2 + b^2) = 0 \end{aligned}$$

Eigenvalues:  $\lambda_1 = 0$  &  $\lambda^2 + a^2 + c^2 + b^2 = 0$

$$\lambda^2 = -a^2 - b^2 - c^2$$

$$\lambda_{2,3} = \pm i\sqrt{a^2 + b^2 + c^2}$$

$A$  is diagonalizable as indicated by the eigenvalues. Since there are 3 distinct eigenvalues, there is 3 linearly independent eigenvector forming a basis of  $\mathbb{R}^3$   
 $\therefore A$  is diagonalizable.

# Question #4

$\lambda_1 = 0, \lambda_2 = 2, \lambda_3 = -2$  with corresponding eigenvectors  $\begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 3 \end{bmatrix}$

$$S = \begin{bmatrix} -1 & 2 & 0 \\ 1 & -1 & 1 \\ 0 & 1 & 3 \end{bmatrix} \leftarrow \text{augmented eigenvectors}$$

$$\Lambda = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & -2 \end{bmatrix} \leftarrow \begin{array}{l} \text{diagonal matrix where} \\ \text{diagonal entries} \\ \text{are the eigenvalues} \end{array}$$

computing  $S^{-1}$ :

$$S^{-1} = \left[ \begin{array}{ccc|ccc} -1 & 2 & 0 & 1 & 0 & 0 \\ 1 & -1 & 1 & 0 & 1 & 0 \\ 0 & 1 & 3 & 0 & 0 & 1 \end{array} \right] \xrightarrow{\text{swap}} \left[ \begin{array}{ccc|ccc} 1 & -1 & 1 & 0 & 1 & 0 \\ -1 & 2 & 0 & 1 & 0 & 0 \\ 0 & 1 & 3 & 0 & 0 & 1 \end{array} \right] \xrightarrow{R_2 + R_1} \left[ \begin{array}{ccc|ccc} 1 & -1 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 1 & 0 \\ 0 & 1 & 3 & 0 & 0 & 1 \end{array} \right] \xrightarrow{R_3 - R_2}$$

$$\Rightarrow \left[ \begin{array}{ccc|ccc} 1 & -1 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 1 & 0 \\ 0 & 0 & 2 & -1 & -1 & 1 \end{array} \right] \xrightarrow{\frac{R_3}{2}} \left[ \begin{array}{ccc|ccc} 1 & -1 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \end{array} \right] \xrightarrow{R_2 - R_3} \left[ \begin{array}{ccc|ccc} 1 & -1 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & \frac{3}{2} & \frac{3}{2} & -\frac{1}{2} \\ 0 & 0 & 1 & -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \end{array} \right] \xrightarrow{R_1 + R_2 - R_3}$$

$$\Rightarrow \begin{bmatrix} 2 & 3 & -1 \\ \frac{3}{2} & \frac{3}{2} & -\frac{1}{2} \\ -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

Now computing  $A = S\Lambda S^{-1}$ :

$$A = \begin{bmatrix} -1 & 2 & 0 \\ 1 & -1 & 1 \\ 0 & 1 & 3 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & -2 \end{bmatrix} \begin{bmatrix} 2 & 3 & -1 \\ \frac{3}{2} & \frac{3}{2} & -\frac{1}{2} \\ -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 4 & 0 \\ 0 & -2 & -2 \\ 0 & 2 & -6 \end{bmatrix} \begin{bmatrix} 2 & 3 & -1 \\ \frac{3}{2} & \frac{3}{2} & -\frac{1}{2} \\ -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \end{bmatrix} = \boxed{\begin{bmatrix} 6 & 6 & -2 \\ -2 & -2 & 0 \\ 6 & 6 & -4 \end{bmatrix}}$$



### Question #5

- a) Since  $A$  is a real & symmetric matrix, we can use  $\vec{v} = e_i$ , where  $e_i$  is the standard basis vector:

$$\vec{v} = e_i = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \leftarrow \text{i-th position}$$

Now, if we substitute  $\vec{v} = e_i$  in  $\vec{v}^T A \vec{v}$ , we will acquire  $e_i^T A e_i$ , which gives the  $(i, i)$  entry of  $A$ . To save on extensive notation,  $e_i^T$  summarize to taking the  $i$ -th row of  $A$ , then  $e_i$  takes the column of  $A$ , giving us the diagonal entries of  $A$ .

By the definition of positive semidefiniteness:

$$\vec{v}^T A \vec{v} \geq 0 \text{ for all } \vec{v}, \text{ then:}$$

$$e_i^T A e_i = A_{ii} \geq 0$$

$\therefore$  All diagonal entries of a positive semidefinite matrix are nonnegative.

- b) Since  $A$  is a real symmetric matrix, there exists an orthogonal matrix  $Q$  where its columns form an orthonormal basis of  $\mathbb{R}^n$ , which are eigenvectors of  $A$ . There also exists a diagonal matrix  $\Lambda$  such that  $A = Q \Lambda Q^T$ . We know that  $\vec{v}^T A \vec{v} \geq 0$ , then taking the spectral decomposition:

$$\vec{v}^T A \vec{v} = \vec{v}^T (Q \Lambda Q^T) \vec{v} = (\vec{v}^T Q) \Lambda (Q^T \vec{v}) \geq 0$$

We can let the change of basis be:

$$y = Q^T \vec{v} \text{ \& \> } y^T = \vec{v}^T Q$$

then substituting:

$$\vec{v}^T (Q \Lambda Q^T) \vec{v} \Leftrightarrow y^T \Lambda y \geq 0$$

Since  $\Lambda$  is the diagonal matrix with its diagonal entries being the eigenvalues &  $y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$  being the orthonormal basis, we can express the equation as the following summation:  $y^T \Lambda y = \sum_{i=1}^n \lambda_i y_i^2 \geq 0$

continue  
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Now, since  $\vec{v}^T A \vec{v} = \vec{y}^T \Lambda \vec{y} \geq 0$  when all  $\lambda_i \geq 0$ , then subsequently  $\vec{v}^T A \vec{v} \geq 0$  for all  $\vec{v} \in \mathbb{R}^n$  so  $A$  is a positive semidefinite matrix.

c) Given that  $A$  is a positive semidefinite matrix for all  $\vec{v} \in \mathbb{R}^n$ ;  $\vec{v}^T A \vec{v} \geq 0$ .

Proof by contradiction:

Assume that  $A$  has a negative eigenvalue  $\lambda < 0$  with a corresponding non-negative eigen vector.

Now, computing  $\vec{v}^T A \vec{v}$  with the eigenvalue equation  $A \vec{v} = \lambda \vec{v}$ :

$$\vec{v}^T A \vec{v} = \vec{v}^T (\lambda \vec{v}) = \lambda (\vec{v}^T \vec{v}) \geq 0$$

Here  $\vec{v}^T \vec{v} = \|\vec{v}\|^2 \geq 0$ , the magnitude, where  $\vec{v}$  is a non-zero eigenvector, then by the assumption that  $\lambda < 0$ ,

$$\vec{v}^T A \vec{v} = \lambda \|\vec{v}\|^2 < 0$$

This contradicts the definition of positive semidefiniteness, which requires  $\vec{v}^T A \vec{v} \geq 0$  for all  $\vec{v}$ , but we got  $\vec{v}^T A \vec{v} < 0$ .

This means that the eigenvalues of a positive semidefinite matrix  $A$  must be nonnegative.

Question #6

In MATLAB file.