

Homework-3

Navidha Jain

2020223

Problem 1:

$$\text{Let } A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ 0 & a_{22} & & & \\ 0 & 0 & a_{33} & & \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & a_{nn} \end{bmatrix}$$

$A \in \mathbb{R}^{n \times n}$, an upper triangular matrix.

For an eigenvalue λ
 $(A - \lambda I) \mathbf{x} = 0$

$$A - \lambda I = \begin{bmatrix} a_{11} - \lambda & a_{12} & a_{13} & \dots & a_{1n} \\ 0 & a_{22} - \lambda & & & \\ 0 & 0 & a_{33} - \lambda & & \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & a_{nn} - \lambda \end{bmatrix}$$

We know determinant of a triangular matrix is the product of its diagonal entries so
 $\det(A - \lambda I) = \prod_{i=1}^n (a_{ii} - \lambda)$

$$\det(A - \lambda I) = (a_{11} - \lambda)(a_{22} - \lambda) \dots (a_{nn} - \lambda)$$

Also $p_n(\lambda)$ (the characteristic equation),

$$p_n(\lambda) = \det(A - \lambda I)$$

$$\Rightarrow p_n(\lambda) = \prod_{i=1}^n (a_{ii} - \lambda)$$

Eigenvalues are the roots of the characteristic polynomial.

Thus eigenvalues for this matrix are
 $a_{11}, a_{22}, \dots, a_{nn}$

Same is for a lower triangular matrix.
Since proved.

Problem 2:

We know A is nonsingular.

~~Therefore~~

We know a nonsingular matrix is diagonalizable.
Thus A is diagonalizable.

Since it is diagonalizable,

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

where X is a non singular matrix

Λ is a diagonal matrix.

X is invertible so

$$XAX^{-1} = \Lambda$$

We know multiplication of a matrix with an invertible matrix does not change its rank.

$$\text{rank}(AX) = \text{rank}(A)$$

$$\text{rank}(X^{-1}AX) = \text{rank}(A)$$

$$\text{rank}(\Lambda) = \text{rank}(A)$$

Λ is a diagonal matrix and rank of a diagonal is equal to the non zero entries and the entries of a diagonal matrix are its eigenvalues (Problem 1)

$$\text{rank}(A) = \text{non zero entries in } \Lambda = \text{non zero eigenvalues}$$

Hence proved

Problem 3:

Claim: $A = uv^T$ is a rank 1 matrix.

Proof: $u \in \mathbb{R}^n, v \in \mathbb{R}^n$

The product uv^T will be a matrix $\in \mathbb{R}^{n \times n}$ containing multiples of u , with the multiples as the elements of v . Since uv^T has all columns of u , its rank = 1.

$$\text{Thus } \text{rank}(A) = \text{rank}(uv^T) = 1$$

Using the result of problem 2 since $\text{rank}(A) = 1$, the number of non-zero eigenvalues is also 1.

Since $u \in \text{col}(A)$ and $\text{col}(A)$ is 1 dimensional, u must be an eigenvector.

$$Au = (uv^T)u = u(v^T u) = (v^T u)u$$

$v^T u$ is a constant and also the non-zero eigenvalue of A .

Since $\dim(\text{col}(A)) = 1$, $\text{null}(A) = n - 1$

where $A \in \mathbb{R}^{n \times n}$

so 0 is an eigenvalue with multiplicity $n - 1$.
 \Rightarrow Eigenvalue of $A \Rightarrow \underline{v^T u}, 0$.

Column space of $A = uv^T$ is 1-dimensional and spanned by u . Thus the power method finds the eigenvector-eigenvalue pair in one step.

Problem 4:

Let A be the real symmetric ~~can~~ positive definite matrix.
 $A \in \mathbb{R}^{n \times n}$.

~~Also~~

\bar{A} = conjugate of A

$\bar{A} = A$ as $A \in \mathbb{R}^{n \times n}$

$A = A^T$ (A is symmetric)

$$A = A^T = \bar{A} \quad \text{--- (1)}$$

Let (λ, x) be an eigen pair for A , $x \neq 0$

$$\lambda \bar{x}^T x = \bar{x}^T (\lambda x)$$

$$= \bar{x}^T (Ax)$$

$$= (A^T \bar{x})^T x$$

$$= (\bar{A} \bar{x})^T x$$

$$= (\bar{\lambda} \bar{x})^T x$$

$$(\because \bar{A} \bar{x} = \bar{\lambda} \bar{x})$$

$$\lambda \bar{x}^T x = \bar{\lambda} \bar{x}^T x$$

Since $x \neq 0$, $\bar{x}^T x \neq 0$ so

$$\lambda = \bar{\lambda}$$

Hence eigenvalues are real.

A is positive definite so

$$\forall x \in \mathbb{R}^n \quad x^T A x > 0$$

(definition)

$$x^T A x = x^T (\lambda x) > 0$$

((λ, x) is an eigen pair)

$$\Rightarrow \lambda x^T x > 0$$

$$\lambda \|x\|^2 > 0$$

Length of x is $\|x\|^2$ is positive ~~and~~

$$\text{so } \underline{\lambda > 0} \text{ for } \|x\|^2 > 0$$

Hence proved.

Problem 5:

$$e^A = \sum_{k=0}^{\infty} \frac{1}{k!} A^k$$

$$e^A = I + A + \frac{1}{2!} A^2 + \dots \quad \text{--- (1)}$$

$$e^{A-\lambda I} = e^{A-\lambda I + \lambda I} = e^{\lambda I} e^{(A-\lambda I)} \quad \text{--- (2)}$$

$$e^{A-\lambda I} = \sum_{k=0}^{\infty} \frac{1}{k!} (A-\lambda I)^k$$

Let (λ, v) be an eigen pair, so:

$$e^{A-\lambda I} v = \left(I + (A-\lambda I) + \frac{1}{2!} (A-\lambda I)^2 + \dots \right) v$$

$$= v + (A-\lambda I)v + \frac{1}{2} (A-\lambda I)v(A-\lambda I) + \dots \quad \text{--- (3)}$$

Since $Av = \lambda v$

$$(A-\lambda I)v = 0 \quad \text{--- (4)}$$

(3) and (4):

$$e^{A-\lambda I} v = v I$$

$$\Rightarrow e^{A-\lambda I} = I \quad \text{--- (5)}$$

Using (5) and (2):

$$e^A = e^{\lambda I} e^{A-\lambda I} = e^{\lambda I} I = e^{\lambda I}$$

$$\text{So } e^A = e^{\lambda I} = e^{\lambda} = \sum_{k=0}^{\infty} \frac{1}{k!} (\lambda)^k$$

Problem 6:

Power Method:

Largest Eigenvalue (magnitude) is 11.0

eigenvector is

[[0.5]

[1.]

[0.75]]

Inverse Power Method:

Smallest Eigenvalue (magnitude) is 1.9999999999999996

eigenvector is

[[-0.2]

[-0.4]

[1.]]

According to the library:

Eigenvalues are [11. -2. -3.]

eigenvectors are

[[3.71390676e-01 1.82574186e-01 2.17732649e-17]

[7.42781353e-01 3.65148372e-01 -5.54700196e-01]

[5.57086015e-01 -9.12870929e-01 8.32050294e-01]]

Problem 7:

Shifted inverse method:

Eigenvalue closest to 2: 2.133074475348525

eigenvector is

[[-0.60692002]

[1.]

[0.34691451]]

According to the library

Eigenvalues are [7.28799214 2.13307448 0.57893339]

Eigenvectors are

[[0.86643225 0.49742503 -0.0431682]

[0.45305757 -0.8195891 -0.35073145]

[0.20984279 -0.28432735 0.9354806]]

Problem 8:

convergence rate for 1 th iteration is -2.0105113838524735

convergence rate for 2 th iteration is 0.23830609100914138

convergence rate for 3 th iteration is 0.34811990082400446

convergence rate for 4 th iteration is 0.4336109012595084

convergence rate for 5 th iteration is 0.5883847596701156

convergence rate for 6 th iteration is 1.0

convergence rate for 7 th iteration is 1.0

convergence rate for 8 th iteration is 1.0

convergence rate for 9 th iteration is 1.0

rayleigh quotient iteration:

eigenvalue: 11.0

eigenvector:

[[0.5]

[1.]

[0.75]]

According to the library, 11. was the largest eigenvalue and we have obtained 11.0 eigenvalue out of the iteration.

Problem 9:

qr iteration matrix(A_k) for matrix in problem 6 is

[[11. 0.72199487 -6.18642278]

[0. -3. -3.8996021]

[0. 0. -2.]]

qr iteration matrix(A_k) for matrix in problem 7 is

[[7.28799214e+00 5.53782393e-16 -1.20642700e-16]

[0.00000000e+00 2.13307448e+00 1.34800587e-16]

[0.00000000e+00 0.00000000e+00 5.78933386e-01]]