Math Problem Set #3

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4.2

$$D = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{bmatrix}$$

D is upper triangular, so the eigenvalues are its diagonal entries $\implies D$ has no eigenvalues. Algebraic and geometric multiplicities are both zero.

4.4

(i) For non-zero $x \in V, \lambda \in \mathbb{F}$,

$$Ax = \lambda x$$
$$(Ax)^{H} = (\lambda x)^{H}$$
$$x^{H} A^{H} = x^{H} \lambda^{H}$$

Post-multiplying by x,

$$x^H A^H x = x^H \lambda^H x$$

Since A is Hermitian, $A^H = A$,

$$\Rightarrow x^{H}Ax = x^{H}\lambda^{H}x$$

$$x^{H}\lambda x = \lambda^{H}x^{H}x$$

$$\lambda x^{H}x = \lambda^{H}x^{H}x$$

$$\Rightarrow \lambda = \lambda^{H}$$

Thus, $\lambda \in \mathbb{R}$.

(ii)
$$\langle Ax, x \rangle = \langle \lambda x, x \rangle = \bar{\lambda} ||x||^2 = \langle x, A^H x \rangle = -\langle x, Ax \rangle = -\lambda ||x||^2 \implies \bar{\lambda} = -\lambda \implies \lambda$$
 is imaginary.

4.6

Take an upper triangular matrix $A \in M_n(\mathbb{F})$,

$$A = \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{bmatrix}$$

With arbitrary entries above the diagonal, and zeros below. Take any $\lambda \in \mathbb{F}$. Then,

$$\lambda I - A = \begin{bmatrix} \lambda - \lambda_1 & & & \\ & \lambda - \lambda_2 & & \\ & & \ddots & \\ & & & \lambda - \lambda_n \end{bmatrix}$$

 λ is an eigenvalue of A if and only if $\lambda I - A$ is not invertible. Equivalently, for upper triangular matrices, A is invertible if and only if all the diagonal entries are nonzero. Thus, λ is an eigenvalue if and only if it equals one of the numbers $\lambda_1, \lambda_2 \cdots \lambda_n$. Thus, $\lambda_1, \lambda_2 \cdots \lambda_n$ are the eigenvalues of A.

4.8

(i) $S = \{\sin(x), \cos(x), \sin(2x), \cos(2x)\}$ forms a basis if it's linearly independent, and if every vector $v \in V$ can be expressed as a linear combination of the elements of S.

(ii) $D = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 \\ 0 & 0 & 2 & 0 \end{bmatrix}$

(iii) The kernel of D is one of them.

4.13

$$A = \begin{bmatrix} 0.8 & 0.4 \\ 0.2 & 0.6 \end{bmatrix}$$

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

$$= \begin{vmatrix} \lambda - 0.8 & -0.4 \\ -0.2 & \lambda - 0.6 \end{vmatrix}$$

$$\implies \sigma(A) = \{1, 0.4\}$$

The eigenvectors corresponding to eigenvalues 1 and 0.4 are $\begin{bmatrix} 2 & 1 \end{bmatrix}^T$ and $\begin{bmatrix} -1 & 1 \end{bmatrix}^T$, respectively. Thus, the transition matrix is:

$$P = \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix}$$

4.15

Let set $S = \{x_1, x_2, \dots, x_n\}$ form an eigenbasis of $A \in M_n(\mathbb{F})$. Because A is semisimple, it is diagonalizable. Let $D = P^{-1}AP$ be the diagonal representation of A. Thus,

$$f(A) = a_0 I + a_1 A + \dots + a_n A^n$$

= $a_0 I + a_1 P^{-1} D P + \dots + a_n P^{-1} D^n P$

The eigenvalues of D are the entries along its diagonal. For $i = 1, \dots, n$,

$$\lambda_i = a_0 + a_1 \lambda_i + \dots + a_n \lambda_i^n$$

= $f(\lambda_i)$

as desired.

4.16

(i)

$$A = \begin{bmatrix} 0.8 & 0.4 \\ 0.2 & 0.6 \end{bmatrix}$$

$$= \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix}^{-1} B \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 0 \\ 0 & 0.4 \end{bmatrix} \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix}$$

$$\implies A^n = \begin{bmatrix} 1.5 & -1 \\ -0.5 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 0.4^n \end{bmatrix} \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix}$$

As $n \to \infty, 0.4 \to 0$.

$$\implies \lim_{n \to \infty} A^n = \begin{bmatrix} 1.5 & -1 \\ -0.5 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix}$$
$$= \begin{bmatrix} 1.5 & 0 \\ -0.5 & 0 \end{bmatrix}$$
$$= \begin{bmatrix} 3 & -1.5 \\ -1 & 0.5 \end{bmatrix}$$
$$\implies \|\lim_{n \to \infty} A^n\|_1 = 4$$

(ii)

$$\|\lim_{n\to\infty} A^n\|_{\infty} = 4.5$$

$$\|\lim_{n\to\infty} A^n\|_F = \sqrt{tr(A^T A)}$$

$$= \sqrt{12.5}$$

(iii) Let matrix $B = 3I + 5A + A^3$, with A as defined above.

$$\sigma(A) = \{1, 0.4\}
\sigma(B) = \{3 + 5(1) + (1)^3, 3 + 5(0.4) + (0.4)^3\}
= \{9, 5.064\}$$

4.18

$$Ax = \lambda x$$

Left-multiplying,

$$x^{T}Ax = x^{T}\lambda x$$
$$x^{T}Ax = \lambda x^{T}x$$
$$\implies x^{T}A = \lambda x^{T}$$

4.20

$$B = U^{H}AU$$

$$B^{H} = (U^{H}AU)^{H}$$

$$= U^{H}AU^{HH}$$

$$= U^{H}AU$$

$$= B$$

 \implies B is Hermitian.

4.24

(i)

$$p(x) = \frac{\langle x, Ax \rangle}{\|x\|^2} = \frac{\langle Ax, x \rangle}{\|x\|^2}$$

$$\implies \frac{\lambda \langle x, x \rangle}{\|x\|^2} = \frac{\bar{\lambda} \langle x, x \rangle}{\|x\|^2}$$

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

$$\implies \lambda \in \mathbb{R}$$

(ii)

$$p(x) = \frac{\langle x, Ax \rangle}{\|x\|^2} = \frac{\langle -Ax, x \rangle}{\|x\|^2}$$

$$\implies \frac{\lambda \langle x, x \rangle}{\|x\|^2} = \frac{-\bar{\lambda} \langle x, x \rangle}{\|x\|^2}$$

$$\implies -\lambda = \bar{\lambda}$$

$$\implies \lambda \in \mathbb{C}$$

4.25

Any $x_j, j = 1, \dots, n$ can be written as $x_1 x_1^H x_j + \dots + x_n x_n^H x_j = (x_1 x_1^H + \dots + x_n x_n^H) x_j \implies x_1 x_1^H + \dots + x_n x_n^H = I_n$.

4.27

Positive definite implies that $x^T A x > 0$ for all $x \in \mathbb{R}^n$. Let e_1, \dots, e_n be the standard basis. For positive definite matrix $A, e_i^T A e_i > 0, i = 1, \dots, n$. For some basis vector $e_i, e_i^T A e_i = a_{ii}$, that is, the i^{th} entry along the diagonal of A. Thus, the diagonal entries of A are all strictly greater than zero.

4.28

4.31

(i)

$$||A||_{2} = \sup_{x \neq 0} \frac{||Ax||_{2}}{||x||_{2}} = \sup_{x \neq 0} \frac{||U\Sigma V'x||_{2}}{||x||_{2}}$$

$$= \sup_{x \neq 0} \frac{||\Sigma V'x||_{2}}{||x||_{2}}$$

$$= \sup_{y \neq 0} \frac{||\Sigma y||_{2}}{||Vy||_{2}}$$

$$= \sup_{y \neq 0} \frac{(\sum_{i=1}^{r} \sigma_{i}^{2} |y_{i}|^{2})^{\frac{1}{2}}}{(\sum_{i=1}^{r} |y_{i}|^{2})^{\frac{1}{2}}}$$

$$\leq \sigma_{1}$$

For $y = \begin{bmatrix} 1 & 0 & \cdots & 0 \end{bmatrix}^T$, $\|\Sigma y\|_2 = \sigma_1$, and the supremum is attained.

(ii)

$$||A||_{2} = \sup_{x \neq 0} \frac{||A^{-1}x||_{2}}{||x||_{2}} = \sup_{x \neq 0} \frac{||(U\Sigma V')^{-1}x||_{2}}{||x||_{2}}$$

$$= \sup_{x \neq 0} \frac{||V\Sigma U^{T}x||_{2}}{||x||_{2}}$$

$$= \sup_{y \neq 0} \frac{||\Sigma^{-1}y||_{2}}{||Vy||_{2}}$$

$$= \sup_{y \neq 0} \frac{(\sum_{i=1}^{r} \frac{1}{\sigma_{i}^{2}} |y_{i}|^{2})^{\frac{1}{2}}}{(\sum_{i=1}^{r} |y_{i}|^{2})^{\frac{1}{2}}}$$

$$\implies ||A^{-1}||_{2} = \sigma_{n}^{-1}$$

(iii)

(iv)

$$||UAV||_{2} = \sup \frac{||UAVx||_{2}}{||x||_{2}}$$

$$= \sup \frac{\langle UAVx, x \rangle^{\frac{1}{2}}}{||x||_{2}}$$

$$= \sup \frac{\langle Ax, x \rangle^{\frac{1}{2}}}{||x||_{2}}$$

$$= \sup \frac{||Ax||_{2}}{||x||_{2}}$$

$$= ||A||_{2}$$

4.32

$$||UAV||_F = \sqrt{Tr[(UAV)(UAV)^H]}$$

$$= \sqrt{Tr(UAVV^HA^HU^H)}$$

$$= \sqrt{Tr(UAA^HU^H)}$$

$$= \sqrt{Tr(AA^HUU^H)}$$

$$= \sqrt{Tr(AA^H)}$$

$$= \sqrt{Tr(A^2)}$$

$$= ||A||_F$$

4.33

4.36

The matrix

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

Has $\det(A) = -1$, eigenvalues $\{i, -i\}$, and singular values $\{1, 1\}$.

4.38

(i)

$$AA^{\dagger} = U_1 \Sigma_1 V_1^H V_1 \Sigma_1^{-1} U_1^H$$
$$= I$$
$$\implies AA^{\dagger} A = A$$

(ii)
$$AA^{\dagger} = I \implies A^{\dagger}AA^{\dagger} = A^{\dagger}.$$

(iii)
$$(AA^{\dagger})^{H} = A^{\dagger H}A^{H}$$

$$= (V_{1}\Sigma_{1}^{-1}U_{1}^{H})^{H}(U_{1}\Sigma_{1}V_{1}^{H})^{H}$$

$$= U_{1}E_{1}^{-1}V_{1}^{H}V_{1}\Sigma_{1}U_{1}^{H}$$

$$= AA^{\dagger}$$

(iv)
$$(A^{\dagger}A)^{H} = A^{H}A^{\dagger H}$$

$$= (U_{1}\Sigma_{1}V_{1}^{H})^{H}(V_{1}\Sigma_{1}^{-}1U_{1}^{H})^{H}$$

$$= V_{1}\Sigma_{1}U_{1}^{H}U_{1}\Sigma_{1}^{-1}V^{H}$$

$$= A^{\dagger}A$$