

ROB 501 HW5

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1

1.1

$$A_3 = \begin{bmatrix} 3 & 10 & 0 \\ 0 & 2 & 4 \\ 0 & 0 & 1 \end{bmatrix}$$
$$A_3 - \lambda I = \begin{bmatrix} 3-\lambda & 10 & 0 \\ 0 & 2-\lambda & 4 \\ 0 & 0 & 1-\lambda \end{bmatrix}$$

Characteristic polynomial of A_3 is $\det(A_3 - \lambda I) = (3 - \lambda)(2 - \lambda)(1 - \lambda)$.

$$\det(A_3 - \lambda I) = 0 \Rightarrow \lambda = 1, 2, 3$$

When $\lambda = 3$, $(A_3 - \lambda I)x_1 = \mathbf{0} \Rightarrow x_1 = [1, 0, 0]^T$

When $\lambda = 2$, $(A_3 - \lambda I)x_2 = \mathbf{0} \Rightarrow x_2 = [-10, 1, 0]^T$

When $\lambda = 1$, $(A_3 - \lambda I)x_3 = \mathbf{0} \Rightarrow x_3 = [20, -4, 1]^T$

Let $\alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3 = \mathbf{0}$,

$$\underbrace{\begin{bmatrix} 1 & -10 & 20 \\ 0 & 1 & -4 \\ 0 & 0 & 1 \end{bmatrix}}_A \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{bmatrix} = \mathbf{0}$$
$$\Rightarrow \det(A) = 1 > 0$$
$$\Rightarrow \alpha_1 = \alpha_2 = \alpha_3 = 0$$

The e-vectors of A are linear independent.

1.2

$$A_4 = \begin{bmatrix} 5 & 1 & 1 \\ 0 & 5 & 3 \\ 0 & 0 & 2 \end{bmatrix}$$

The e-values: $\lambda = 5, 5, 2$.

When $\lambda = 5$, $(A_4 - \lambda I)x_1 = \mathbf{0} \Rightarrow x_1 = [1, 0, 0]^T$

When $\lambda = 2$, $(A_4 - \lambda I)x_2 = \mathbf{0} \Rightarrow x_2 = [0, -1, 1]^T$

No, since $\text{span}\{x_1, x_2\} \neq \mathbb{R}^3$. E.g. $v = [0, 0, 1]$ can not be represented by any combination of x_1 and x_2 .

2

Given $B = P^{-1}AP$, show that $\det(A - \lambda I) = \det(B - \lambda I)$.

Proof.

$$\begin{aligned}
 \det(B - \lambda I) &= \det(P^{-1}AP - \lambda I) \\
 &= \det(P^{-1}AP - P^{-1}\lambda IP) \\
 &= \det(P^{-1}(A - \lambda I)P) \\
 &= \det(P^{-1})\det(A - \lambda I)\det(P) \\
 &= \det(A - \lambda I)
 \end{aligned}$$

□

3

Given A_3 , whose e-values are $\lambda = 3, 2, 1$ and cooresponding e-vectors are $[1, 0, 0]^T$, $[-10, 1, 0]^T$, $[20, -4, 1]^T$, show that A is similar to a diagonal matrix.

Proof.

$$\begin{aligned}
 A_3[x_1 \mid x_2 \mid x_3] &= [\lambda_1 x_1 \mid \lambda_2 x_2 \mid \lambda_3 x_3] \\
 &= \underbrace{[x_1 \mid x_2 \mid x_3]}_P \underbrace{\begin{bmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \lambda_3 \end{bmatrix}}_{\Lambda} \\
 \Rightarrow A_3 &= P\Lambda P^{-1}
 \end{aligned}$$

□

4

4.1

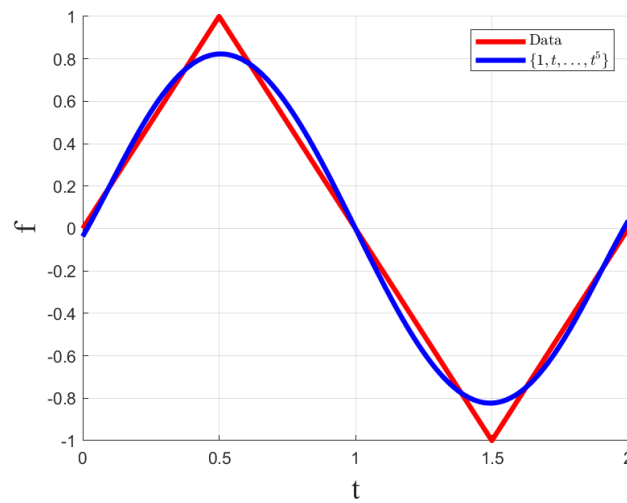


Figure 1: Least square fit of the data using $\{1, t, \dots, t^5\}$

Coefficients: $\alpha = [-0.0370, 2.2158, 2.8901, -11.2271, 7.6978, -1.5396]^T$.

4.2

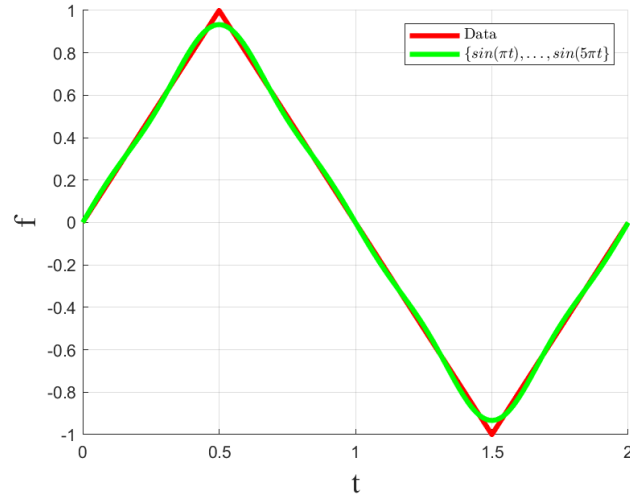


Figure 2: Least square fit of the data using $\{\sin(\pi t), \dots, \sin(5\pi t)\}$

Coefficients: $\alpha = [0.8106, 0.0000, -0.0901, 0.0000, 0.0325]^T$.

5

Function: $f(t) = \alpha_0 + \alpha_1 t + \alpha_2 t^2 + \alpha_3 t^3$

$$f'(t) = \frac{df(t)}{dt} = 3\alpha_3 t^2 + 2\alpha_2 t + \alpha_1$$

$$f'(0.3) = -0.4016$$

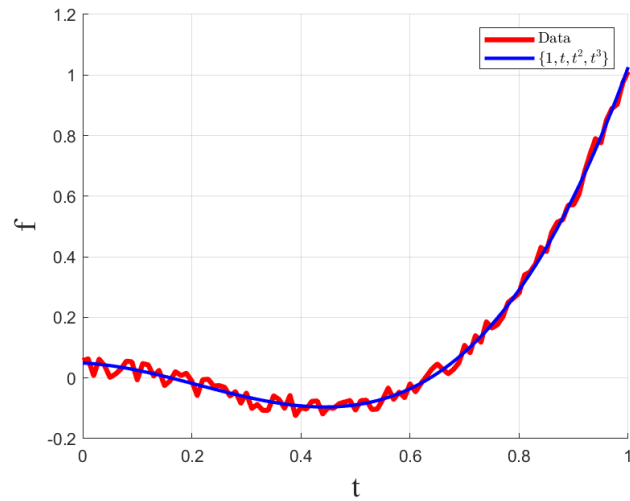


Figure 3: Least square fit of the data using $\{1, t, t^2, t^3\}$

Coefficients: $\alpha = [0.0495, -0.0847, -1.8287, 2.8900]^T$.

6

Show that on $(\mathbb{C}^n, \mathbb{C})$, $\langle x, y \rangle = x^T \bar{y}$ satisfies the definition of inner product used in lecture, while $\langle x, y \rangle = \bar{x}^T y$ satisfies the definition of inner product in Nagy's book.

For part I,

(a) **Hermitian Symmetry:** $\forall x, y \in \mathbb{C}^n, \overline{\langle y, x \rangle} = \overline{y^T \bar{x}} = \overline{\sum_{i=1}^n y_i \bar{x}_i} = \sum_{i=1}^n \bar{y}_i x_i = x^T \bar{y} = \langle x, y \rangle$

(b) **Non-negativity:** $\forall x \in \mathbb{C}^n, \langle x, x \rangle = x^T \bar{x} = \sum_{i=1}^n x_i \bar{x}_i \geq 0$; when $\langle x, x \rangle = 0, |x_i|^2 = 0 \Rightarrow x = \mathbf{0}$

(c) **Linearity in the first argument:**

$$\forall \alpha_1, \alpha_2 \in \mathbb{C}, x_1, x_2, y \in \mathbb{C}^n,$$

$$\langle \alpha_1 x_1 + \alpha_2 x_2, y \rangle = (\alpha_1 x_1 + \alpha_2 x_2)^T \bar{y} = \alpha_1 x_1^T \bar{y} + \alpha_2 x_2^T \bar{y} = \alpha_1 \langle x_1, y \rangle + \alpha_2 \langle x_2, y \rangle$$

For part II,

(a) **Hermitian Symmetry:** $\forall x, y \in \mathbb{C}^n, \overline{\langle y, x \rangle} = \overline{\bar{y}^T x} = \overline{\sum_{i=1}^n \bar{y}_i x_i} = \sum_{i=1}^n y_i \bar{x}_i = \bar{x}^T y = \langle x, y \rangle$

(b) **Non-negativity:** $\forall x \in \mathbb{C}^n, \langle x, x \rangle = \bar{x}^T x = \sum_{i=1}^n \bar{x}_i x_i \geq 0$; when $\langle x, x \rangle = 0, |x_i|^2 = 0 \Rightarrow x = \mathbf{0}$

(c) **Linearity in the second argument:**

$$\forall \alpha_1, \alpha_2 \in \mathbb{C}, x, y_1, y_2 \in \mathbb{C}^n,$$

$$\langle x, \alpha_1 y_1 + \alpha_2 y_2 \rangle = \bar{x}^T (\alpha_1 y_1 + \alpha_2 y_2) = \alpha_1 \bar{x}^T y_1 + \alpha_2 \bar{x}^T y_2 = \alpha_1 \langle x, y_1 \rangle + \alpha_2 \langle x, y_2 \rangle$$

7

Define the inner product $\langle p, q \rangle = \int_{-1}^1 p(x)q(x)dx$ in $\mathbb{P}_3([-1, 1])$.

Show that the set $P = \{p_0, p_1, p_2, p_3\}$ is an orthogonal basis in \mathbb{P}_3 .

Proof. (a) P is linear independent Let $\alpha_0 p_0 + \alpha_1 p_1 + \alpha_2 p_2 + \alpha_3 p_3 = 0$,

$$\begin{aligned} \frac{5}{2}\alpha_3 x^3 + \frac{3}{2}\alpha_2 x^2 + (\alpha_1 - \frac{3}{2}\alpha_3)x + (\alpha_0 - \frac{1}{2}\alpha_2) &= 0 \\ \begin{cases} \alpha_3 = 0 \\ \alpha_2 = 0 \\ \alpha_1 - \frac{3}{2}\alpha_3 = 0 \\ \alpha_0 - \frac{1}{2}\alpha_2 = 0 \end{cases} &\Rightarrow \alpha_0 = \alpha_1 = \alpha_2 = \alpha_3 = 0 \end{aligned}$$

(b) $\text{span}\{p_0, \dots, p_3\} = \mathbb{P}_3 \forall x \in \mathbb{P}_3, x = \beta_3 x^3 + \beta_2 x^2 + \beta_1 x + \beta_0$, we show that x can be represented by P .

$$\begin{cases} \frac{5}{2}\alpha_3 = \beta_3 \\ \frac{3}{2}\alpha_2 = \beta_2 \\ \alpha_1 - \frac{3}{2}\alpha_3 = \beta_1 \\ \alpha_0 - \frac{1}{2}\alpha_2 = \beta_0 \end{cases} \Rightarrow \begin{cases} \alpha_0 = \beta_0 + \frac{1}{3}\beta_2 \\ \alpha_1 = \beta_1 + \frac{3}{5}\beta_3 \\ \alpha_2 = \frac{2}{3}\beta_2 \\ \alpha_3 = \frac{2}{5}\beta_3 \end{cases}$$

(c) The inner product of between elements in P equal to 0.

$$\langle p_0, p_3 \rangle = \int_{-1}^1 \frac{1}{2} (5x^3 - 3x) dx = \int_{-1}^1 \left(\frac{5}{2}x^3 - \frac{3}{2}x \right) dx = \left(\frac{5}{8}x^4 - \frac{3}{4}x^2 \right) \Big|_{-1}^1 = 0$$

$$\langle p_1, p_2 \rangle = \int_{-1}^1 \frac{1}{2} x (3x^2 - 1) dx = \int_{-1}^1 \left(\frac{3}{2}x^3 - \frac{1}{2}x \right) dx = \left(\frac{3}{8}x^4 - \frac{1}{4}x^2 \right) \Big|_{-1}^1 = 0$$

□

8

8.1

$$\begin{aligned}
& (A + BCD) \left[A^{-1} - A^{-1}B(C^{-1} + DA^{-1}B)^{-1}DA^{-1} \right] \\
&= \left[I - B(C^{-1} + DA^{-1}B)^{-1}DA^{-1} \right] + \left[BCDA^{-1} - BCDA^{-1}B(C^{-1} + DA^{-1}B)^{-1}DA^{-1} \right] \\
&= \left[I + BCDA^{-1} \right] - \left[B(C^{-1} + DA^{-1}B)^{-1}DA^{-1} + BCDA^{-1}B(C^{-1} + DA^{-1}B)^{-1}DA^{-1} \right] \\
&= I + BCDA^{-1} - (B + BCDA^{-1}B)(C^{-1} + DA^{-1}B)^{-1}DA^{-1} \\
&= I + BCDA^{-1} - BC(C^{-1} + DA^{-1}B)(C^{-1} + DA^{-1}B)^{-1}DA^{-1} \\
&= I + BCDA^{-1} - BCDA^{-1} \\
&= I
\end{aligned}$$

8.2

$$A = \text{diag}([0.5, 1, 1, 0.5, 1])$$

$$A^{-1} = \text{diag}([2, 1, 1, 2, 1])$$

$$\begin{aligned}
& A^{-1} - A^{-1}B(C^{-1} + DA^{-1}B)^{-1}DA^{-1} \\
&= A^{-1} - A^{-1} \begin{bmatrix} 3 \\ 0 \\ 2 \\ 0 \\ 1 \end{bmatrix} \left(4 + [3 \ 0 \ 2 \ 0 \ 1] A^{-1} \begin{bmatrix} 3 \\ 0 \\ 2 \\ 0 \\ 1 \end{bmatrix} \right)^{-1} [3 \ 0 \ 2 \ 0 \ 1] A^{-1} \\
&= A^{-1} - \begin{bmatrix} 6 \\ 0 \\ 2 \\ 0 \\ 1 \end{bmatrix} (27)^{-1} [6 \ 0 \ 2 \ 0 \ 1] \\
&= \begin{bmatrix} \frac{2}{3} & 0 & -\frac{4}{9} & 0 & -\frac{2}{9} \\ 0 & 1 & 0 & 0 & 0 \\ -\frac{4}{9} & 0 & \frac{23}{27} & 0 & -\frac{2}{27} \\ 0 & 0 & 0 & 2 & 0 \\ -\frac{2}{9} & 0 & -\frac{2}{27} & 0 & \frac{26}{27} \end{bmatrix}
\end{aligned}$$

9

9.1

Define $f(x) = (x^T A x)^{\frac{1}{2}}$ in $(\mathbb{R}^n, \mathbb{R})$

A is positive definite $\Rightarrow x^T A x > 0, \forall x \neq 0$ and there exists an invertible matrix B s.t. $A = BB^T$.

Show that $f(x)$ is a norm.

(a) **Non-negative:** $\forall x \in \mathbb{R}^n, f(x) = (x^T A x)^{\frac{1}{2}} \geq 0$ and $f(x) = 0 \Leftrightarrow x = 0$

(b) **Scalability:** $\forall \alpha \in \mathbb{R}, \forall x \in \mathbb{R}^n, f(\alpha x) = [(\alpha x)^T A (\alpha x)]^{\frac{1}{2}} = |\alpha| (x^T A x)^{\frac{1}{2}}$

(c) **Triangular inequality:** $\forall x, y \in \mathbb{R}^n$

$$\begin{aligned}
f(x+y) &= [(x+y)^T A(x+y)]^{\frac{1}{2}} \\
&= [(x^T + y^T) A(x+y)]^{\frac{1}{2}} \\
&= [x^T A(x+y) + y^T A(x+y)]^{\frac{1}{2}} \\
&= [x^T Ax + x^T Ay + y^T Ax + y^T Ay]^{\frac{1}{2}} \\
&= [x^T Ax + y^T Ay + 2x^T Ay]^{\frac{1}{2}} \Leftarrow x^T Ay = y^T Ax, x^T Ay = (B^T x)^T (B^T y) \\
&\leq [x^T Ax + y^T Ay + 2\sqrt{x^T Ax \cdot y^T Ay}] \Leftarrow |<x, y>| \leq <x, x>^{1/2} <y, y>^{1/2} \\
&= [x^T Ax]^{\frac{1}{2}} + [y^T Ay]^{\frac{1}{2}} \\
&\text{Equality holds when } x = y \\
&= f(x) + f(y)
\end{aligned}$$

If A is replaced by $2A$, it is still a positive definite matrix, so proof holds.

9.2

Given $(\mathbb{R}^n, \mathbb{R}, \|\cdot\|_V)$ is a norm space.

Show that in $(\mathbb{R}^{n \times n}, \mathbb{R})$, $f_V(A) = \sup_{x \neq 0} \frac{\|Ax\|_V}{\|x\|_V}$ ($x \in \mathbb{R}^n$) is a norm.

(a) **Non-negative:** $\forall A$ in $\mathbb{R}^{n \times n}$, $f_V(A) \geq 0$, since $\|\cdot\|_V \geq 0$. $f_V(A) = 0$ holds when $A = 0$

(b) **Scalability:** $\forall \alpha \in \mathbb{R}$, $\forall A \in \mathbb{R}^{n \times n}$, $f_v(\alpha A) = \sup_{x \neq 0} \frac{\|\alpha Ax\|_V}{\|x\|_V} = \sup_{x \neq 0} \frac{|\alpha| \|Ax\|_V}{\|x\|_V} = |\alpha| f_V(A)$

(c) **Triangular inequality:** $\forall A, B \in \mathbb{R}^{n \times n}$

$$\begin{aligned}
f_v(A+B) &= \sup_{x \neq 0} \frac{\|(A+B)x\|_V}{\|x\|_V} \\
&= \sup_{x \neq 0} \frac{\|Ax + Bx\|_V}{\|x\|_V} \\
&\leq \sup_{x \neq 0} \frac{\|Ax\|_V + \|Bx\|_V}{\|x\|_V} \text{ [Equality form } (\mathbb{R}^n, \mathbb{R}, \|\cdot\|_V)] \\
&= \sup_{x \neq 0} \frac{\|Ax\|_V}{\|x\|_V} + \underbrace{\sup_{x \neq 0} \frac{\|Bx\|_V}{\|x\|_V}} \\
&= f_V(A) + f_V(B)
\end{aligned}$$

Hence,

$$\begin{aligned}
f_1(A) &= \sup_{x \neq 0} \frac{\|Ax\|_1}{\|x\|_1} = \sup_{x \neq 0} \frac{\sum_{i=1}^n \left| \sum_{j=1}^n a_{ij} x_j \right|}{\sum_{i=1}^n |x_i|} = \max_{1 \leq j \leq n} \sum_{i=1}^n |a_{ij}| \\
f_\infty(A) &= \sup_{x \neq 0} \frac{\|Ax\|_\infty}{\|x\|_\infty} = \sup_{x \neq 0} \frac{\max_{1 \leq i \leq n} \left| \sum_{j=1}^n a_{ij} x_j \right|}{\max_{1 \leq i \leq n} |x_i|} = \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}|
\end{aligned}$$