

## EEE424 Homework 2

(Clearly justify all answers.)

(Due 16 March 2023)

- 1- It is given that  $V$  is an  $N$ -dimensional inner product space (standard inner product), and  $S_2$  be a subspace in  $V$ .  $Q = \{\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_M\}$ ,  $M < N$ , is a basis for  $S_2$ . The problem is to find a solution to  $\underset{\hat{\mathbf{y}}}{\operatorname{argmin}} \|\mathbf{y} - \hat{\mathbf{y}}\|^2$ , where  $\|\cdot\|$  is the inner product induced norm,  $\mathbf{y} \in V$  and  $\hat{\mathbf{y}} \in S_2$ .

a) Prove that the solution to the above problem must satisfy,

i-  $(\mathbf{y} - \hat{\mathbf{y}}) \perp S_2$

ii-  $(\mathbf{y} - \hat{\mathbf{y}}) \perp \hat{\mathbf{y}}$

iii-  $(\mathbf{y} - \hat{\mathbf{y}}) \perp \mathbf{p}_j, \quad j = 1, \dots, M$

b) Derive the normal equations in terms of inner products  $\langle p_i, p_j \rangle$  and  $\langle y, p_j \rangle$  to solve for coefficients,  $c_i$ 's where  $\hat{\mathbf{y}} = \sum_{i=1}^M c_i \mathbf{p}_i$  using the facts given in (a).

c)  $\mathbf{x}$  is a vector in  $S_1$  which is an  $M$ -dimensional inner product space (standard inner product), and  $\mathbf{A}$  is an  $N \times M$  matrix,  $N > M$ , whose columns are linearly independent, representing a linear transform. Derive the left pseudoinverse  $\mathbf{A}_{ps}$  of  $\mathbf{A}$  which would yield  $\hat{\mathbf{x}} = \mathbf{A}_{ps} \mathbf{y}$  such that  $\hat{\mathbf{x}}$  is the solution to  $\hat{\mathbf{x}} = \underset{\mathbf{x}}{\operatorname{argmin}} \|\mathbf{y} - \mathbf{A}\mathbf{x}\|^2$

d) Let

$$A = \begin{bmatrix} 1 & -1 \\ 2 & 1 \\ -1 & 1 \end{bmatrix} \quad \mathbf{y} = \begin{bmatrix} 3 \\ -1 \\ 1 \end{bmatrix}$$

Find  $\hat{\mathbf{x}} = \underset{\mathbf{x}}{\operatorname{argmin}} \|\mathbf{y} - \mathbf{A}\mathbf{x}\|^2$ .

e) Find  $\hat{\mathbf{y}} = \mathbf{A}\hat{\mathbf{x}}$  and show that  $(\mathbf{y} - \hat{\mathbf{y}})$  is orthogonal to  $\hat{\mathbf{y}}$ .

- 2- Functions  $\phi_i(t)$ ,  $i = 1, 2, 3$  are given as,

$$\phi_1(t) = \begin{cases} t & \text{if } t \in [0, 1] \\ 0 & \text{else} \end{cases}$$

$$\phi_2(t) = \begin{cases} -1 & \text{if } t \in [0, 2] \\ 0 & \text{else} \end{cases}$$

$$\phi_3(t) = \begin{cases} -t+1 & \text{if } t \in [1, 2] \\ 0 & \text{else} \end{cases}$$

- a) Plot these functions. Are these functions orthogonal to each other?
- b) Find the range space of  $\{\phi_1(t), \phi_2(t), \phi_3(t)\}$ . Select an arbitrary  $(c_1, c_2, c_3)$  (indicate your selection). Plot a typical function in that range space using your selected coefficients, i.e., plot  $c_1\phi_1(t) + c_2\phi_2(t) + c_3\phi_3(t)$  for your  $c_i$ 's.

- c) Find the parameters,  $c_i$ 's to minimize  $\left\| y(t) - \sum_{i=1}^3 c_i \phi_i(t) \right\|^2$ , where

$$y(t) = \begin{cases} -1 & \text{if } t \in [-1, 1] \\ 2 & \text{if } t \in [2, 3] \\ 0 & \text{else} \end{cases}$$

Plot  $y(t)$  and  $\hat{y}(t) = c_1\phi_1(t) + c_2\phi_2(t) + c_3\phi_3(t)$  using the optimum  $c_i$ 's you found on the same graph.

- d) Find and plot the error function  $e(t) = y(t) - \hat{y}(t)$ , and show that  $e(t)$  is orthogonal to  $\hat{y}(t)$ .

## EEE424 Digital Signal Processing Homework 2

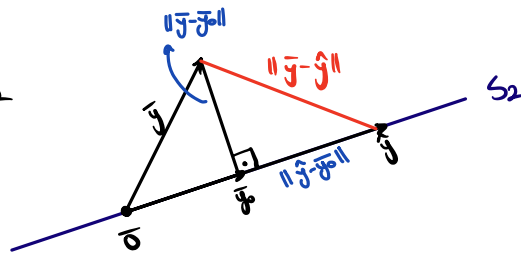
1.

It is given that  $V$  is an  $N$ -dimensional inner product space (standard inner product), and  $S_2$  be a subspace in  $V$ .  $Q = \{\mathbf{p}_1, \mathbf{p}_2, \dots, \mathbf{p}_M\}$ ,  $M < N$ , is a basis for  $S_2$ . The problem is to find a solution to  $\underset{\hat{\mathbf{y}}}{\operatorname{argmin}} \|\mathbf{y} - \hat{\mathbf{y}}\|^2$ , where  $\|\cdot\|$  is the inner product induced norm,  $\mathbf{y} \in V$  and  $\hat{\mathbf{y}} \in S_2$ .

a) Prove that the solution to the above problem must satisfy,

1.a.i)  $(\mathbf{y} - \hat{\mathbf{y}}) \perp S_2$

let  $\bar{\mathbf{y}}_0 \in S_2$ ,  $\bar{\mathbf{y}} - \bar{\mathbf{y}}_0 \perp S_2$



$$\underset{\hat{\mathbf{y}}}{\operatorname{argmin}} \|\bar{\mathbf{y}} - \hat{\mathbf{y}}\|^2 = \underset{\hat{\mathbf{y}}}{\operatorname{argmin}} [\|\hat{\mathbf{y}} - \bar{\mathbf{y}}_0\|^2 + \underbrace{\|\bar{\mathbf{y}} - \bar{\mathbf{y}}_0\|^2}_{\text{Constant}}] = \underset{\hat{\mathbf{y}}}{\operatorname{argmin}} \|\hat{\mathbf{y}} - \bar{\mathbf{y}}_0\| = \bar{\mathbf{y}}_0$$

$\Rightarrow$  Therefore  $\hat{\mathbf{y}} = \bar{\mathbf{y}}_0$  and since  $\bar{\mathbf{y}} - \bar{\mathbf{y}}_0 \perp S_2$   $(\bar{\mathbf{y}} - \hat{\mathbf{y}}) \perp S_2$

The proof is complete. ■

1.a.ii)  $(\mathbf{y} - \hat{\mathbf{y}}) \perp \hat{\mathbf{y}}$

Since  $\hat{\mathbf{y}} \in S_2$  and  $(\bar{\mathbf{y}} - \hat{\mathbf{y}}) \perp S_2$ ,  $(\bar{\mathbf{y}} - \hat{\mathbf{y}}) \perp \hat{\mathbf{y}}$  is also correct and proved. ■

1.a.iii)  $(\mathbf{y} - \hat{\mathbf{y}}) \perp \mathbf{p}_j, \quad j = 1, \dots, M$

Since  $\bar{\mathbf{p}}_j \in S_2$  for  $j = 0, 1, \dots, M$

and  $(\bar{\mathbf{y}} - \hat{\mathbf{y}}) \perp S_2$

then,  $(\bar{\mathbf{y}} - \hat{\mathbf{y}}) \perp \mathbf{p}_j$  is also correct and proved. ■

- 1.b) b) Derive the normal equations in terms of inner products  $\langle p_i, p_j \rangle$  and  $\langle y, p_j \rangle$  to solve for coefficients,  $c_i$ 's where  $\hat{y} = \sum_{i=1}^M c_i p_i$  using the facts given in (a).

$$\bar{y} - \hat{y} = \bar{y} - \sum_{i=1}^M c_i \bar{p}_i \quad \text{By using the equation from 1.a.iii)}$$

$$\Rightarrow (\bar{y} - \sum_{i=1}^M c_i \bar{p}_i) \perp \bar{p}_j \quad \text{for } j=1, \dots, M$$

$$\Rightarrow \langle (\bar{y} - \sum_{i=1}^M c_i \bar{p}_i), \bar{p}_j \rangle = 0 \quad \text{for } j=1, \dots, M$$

$$\langle \hat{y}, \bar{p}_j \rangle = \sum_{i=1}^M c_i \langle \bar{p}_i, \bar{p}_j \rangle \quad \text{for } j=1, \dots, M$$

$$\langle \hat{y}, \bar{p}_j \rangle = \sum_{i=1}^M c_i \langle \bar{p}_i, \bar{p}_j \rangle \Rightarrow \bar{p} = \bar{R} \cdot \bar{c} \quad , \quad \boxed{\bar{c} = \bar{R}^{-1} \cdot \bar{p}} \quad \begin{array}{l} \hookrightarrow \text{Normal eq's for} \\ \text{the given set.} \end{array}$$

- 1.c)  $\mathbf{x}$  is a vector in  $S_1$  which is an  $M$ -dimensional inner product space (standard inner product), and  $\mathbf{A}$  is an  $N \times M$  matrix,  $N > M$ , whose columns are linearly independent, representing a linear transform. Derive the left pseudoinverse  $\mathbf{A}_{ps}$  of  $\mathbf{A}$  which would yield  $\hat{\mathbf{x}} = \mathbf{A}_{ps} \mathbf{y}$  such that  $\hat{\mathbf{x}}$  is the solution to  $\hat{\mathbf{x}} = \underset{\mathbf{x}}{\operatorname{argmin}} \|\mathbf{y} - \mathbf{A}\mathbf{x}\|^2$

$$\text{Let } \bar{\mathbf{A}} = [\bar{a}_1 \bar{a}_2 \dots \bar{a}_M] \quad ,$$

$$\Rightarrow \hat{\mathbf{x}} = \underset{\mathbf{x}}{\operatorname{argmin}} \|\bar{\mathbf{y}} - \bar{\mathbf{A}}\mathbf{x}\|^2 \quad , \quad \hat{\mathbf{y}} = \bar{\mathbf{A}}\hat{\mathbf{x}} \quad \text{Then the minimization}$$

becomes  $\underset{\hat{\mathbf{y}}}{\operatorname{argmin}} \|\bar{\mathbf{y}} - \hat{\mathbf{y}}\|^2$  where  $\hat{\mathbf{y}} \in \mathcal{L}(\bar{\mathbf{A}})$ . By using the relations proved in previous parts:

$$\Rightarrow (\bar{\mathbf{y}} - \hat{\mathbf{y}}) \perp \mathcal{L}(\bar{\mathbf{A}}) \quad \hat{\mathbf{y}} = \sum_{i=1}^M \hat{x}_i \cdot \bar{a}_i \Rightarrow \langle \bar{\mathbf{y}} - \sum_{i=1}^M \hat{x}_i \cdot \bar{a}_i, \bar{a}_j \rangle = 0 \quad \text{for } j=1, 2, \dots, M$$

$$\Rightarrow \langle \bar{\mathbf{y}}, \bar{a}_j \rangle = \sum_{i=1}^M \hat{x}_i \cdot \langle \bar{a}_i, \bar{a}_j \rangle \quad \text{for } j=1, 2, \dots, M \quad \Rightarrow \text{The Normal Eqn}$$

$$\begin{array}{c} \downarrow \qquad \qquad \downarrow \\ \underbrace{\sum_{i=1}^M \begin{bmatrix} \bar{a}_i^{*T} \\ \vdots \\ \bar{a}_M^{*T} \end{bmatrix}}_{\bar{\mathbf{A}}_{adj}} \cdot \bar{\mathbf{y}} = \underbrace{\begin{bmatrix} \bar{a}_1^{*T} \\ \bar{a}_2^{*T} \\ \vdots \\ \bar{a}_M^{*T} \end{bmatrix}}_{\bar{\mathbf{A}}_{adj}} \cdot \underbrace{[\bar{a}_1 \bar{a}_2 \dots \bar{a}_M]}_{\bar{\mathbf{A}}} \cdot \hat{\mathbf{x}} \end{array} \quad \begin{array}{l} \nearrow \bar{\mathbf{A}}_{adj} \cdot \bar{\mathbf{y}} = \bar{\mathbf{A}}_{adj} \cdot \bar{\mathbf{A}} \cdot \hat{\mathbf{x}} \Rightarrow \hat{\mathbf{x}} = \underbrace{(\bar{\mathbf{A}}_{adj} \cdot \bar{\mathbf{A}})^{-1}}_{\bar{\mathbf{A}}_{ps}} \cdot \bar{\mathbf{A}}_{adj} \cdot \bar{\mathbf{y}} \\ \boxed{\bar{\mathbf{A}}_{ps} = (\bar{\mathbf{A}}_{adj} \cdot \bar{\mathbf{A}})^{-1} \cdot \bar{\mathbf{A}}_{adj}} \end{array}$$

Since  $\bar{\mathbf{A}}$  has  $L$  columns  $\bar{\mathbf{A}}_{ps}$  definitely exists

1.d)

d) Let

$$A = \begin{bmatrix} 1 & -1 \\ 2 & 1 \\ -1 & 1 \end{bmatrix} \quad y = \begin{bmatrix} 3 \\ -1 \\ 1 \end{bmatrix}$$

By using the equation derived in part 1.c,  $\arg\max_{\bar{x}} \|\bar{y} - \bar{A}\bar{x}\|^2 = (\bar{A}_{adj} \cdot \bar{A})^{-1} \cdot \bar{A}_{adj} \cdot \bar{y}$

$$\hat{x} = (\bar{A}_{adj} \cdot \bar{A})^{-1} \cdot \bar{A}_{adj} \cdot \bar{y}$$

$$\bar{A}_{adj} \cdot \bar{A} = \begin{bmatrix} 1 & 2 & -1 \\ -1 & 1 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 & -1 \\ 2 & 1 \\ -1 & 1 \end{bmatrix} = \begin{bmatrix} 6 & 0 \\ 0 & 3 \end{bmatrix}$$

$$(\bar{A}_{adj} \cdot \bar{A})^{-1} = \begin{bmatrix} 1/6 & 0 \\ 0 & 1/3 \end{bmatrix}$$

$$\hat{x} = \begin{bmatrix} 1/6 & 0 \\ 0 & 1/3 \end{bmatrix} \cdot \begin{bmatrix} 1 & 2 & -1 \\ -1 & 1 & 1 \end{bmatrix} \cdot \begin{bmatrix} 3 \\ -1 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} 1/6 & 1/3 & -1/6 \\ -1/3 & 1/3 & 1/3 \end{bmatrix} \cdot \begin{bmatrix} 3 \\ -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$$

$$\hat{x} = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$$

which is  $\arg\min_{\bar{x}} \|\bar{y} - \bar{A}\bar{x}\|^2$

1.e) e) Find  $\hat{y} = A\hat{x}$  and show that  $(y - \hat{y})$  is orthogonal to  $\hat{y}$ .

$$\hat{x} = \begin{bmatrix} 0 \\ -1 \end{bmatrix}, \quad \bar{A} = \begin{bmatrix} 1 & -1 \\ 2 & 1 \\ -1 & 1 \end{bmatrix}, \quad \hat{y} = \begin{bmatrix} 1 & -1 \\ 2 & 1 \\ -1 & 1 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ -1 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix}, \quad \bar{y} = \begin{bmatrix} 3 \\ -1 \\ 1 \end{bmatrix}$$

$$\bar{y} - \hat{y} = \begin{bmatrix} 2 \\ 0 \\ 2 \end{bmatrix}, \quad \hat{y} = \begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix}, \quad \langle \bar{y} - \hat{y}, \hat{y} \rangle = 2 \cdot 1 + 0 \cdot (-1) + 2 \cdot (-1) = \underline{\underline{0}}$$

$(\bar{y} - \hat{y}) \perp \hat{y}$  as expected.

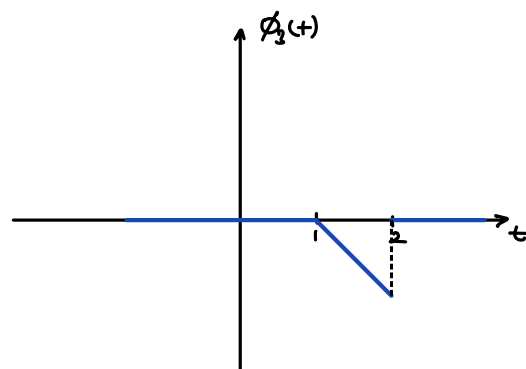
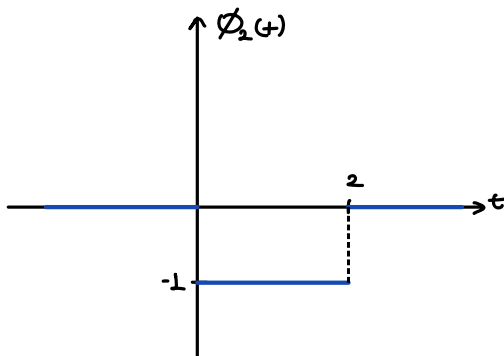
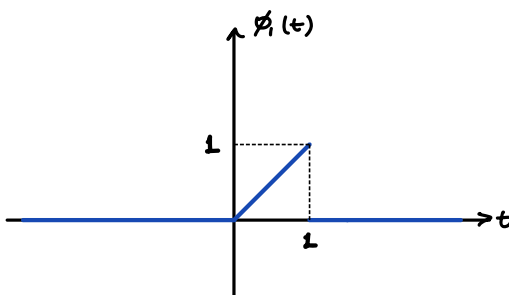
2.

2.a) Plot these functions. Are these functions orthogonal to each other?

$$\phi_1(t) = \begin{cases} t & \text{if } t \in [0, 1] \\ 0 & \text{else} \end{cases}$$

$$\phi_2(t) = \begin{cases} -1 & \text{if } t \in [0, 2] \\ 0 & \text{else} \end{cases}$$

$$\phi_3(t) = \begin{cases} -t + 1 & \text{if } t \in [1, 2] \\ 0 & \text{else} \end{cases}$$



$$\bullet \langle \phi_1(t), \phi_2(t) \rangle = \int_0^1 -t \, dt = \underline{\underline{-\frac{1}{2}}}$$

$$\bullet \langle \phi_2(t), \phi_3(t) \rangle = \int_1^2 t-1 \, dt = \left. \frac{t^2}{2} - t \right|_1^2 = \underline{\underline{\frac{1}{2}}}$$

$$\bullet \langle \phi_1(t), \phi_3(t) \rangle = \underline{\underline{0}}$$

$\phi_1(t) \perp \phi_3(t)$

$\phi_2(t)$  is orthogonal neither  $\phi_1(t)$  nor  $\phi_3(t)$ .

2.b) Find the range space of  $\{\phi_1(t), \phi_2(t), \phi_3(t)\}$ . Select an arbitrary  $(c_1, c_2, c_3)$  (indicate your selection). Plot a typical function in that range space using your selected coefficients, i.e., plot  $c_1\phi_1(t) + c_2\phi_2(t) + c_3\phi_3(t)$  for your  $c_i$ 's.

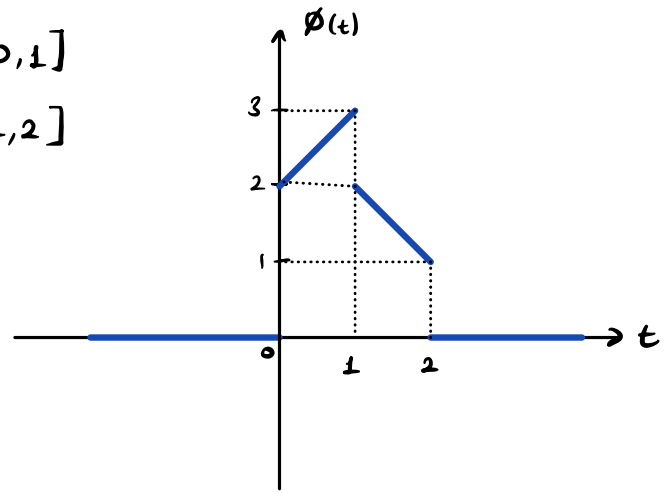
$$\phi(t) = c_1 \cdot \phi_1(t) + c_2 \cdot \phi_2(t) + c_3 \cdot \phi_3(t) \in \mathcal{R}(\{\phi_1(t), \phi_2(t), \phi_3(t)\}) \quad \forall c_1, c_2, c_3 \in \mathbb{R}$$

$$\phi(t) = \begin{cases} c_1 \cdot t - c_2 & t \in [0, 1] \\ -c_3 \cdot t - c_2 + c_3 & t \in [1, 2] \\ 0 & \text{else} \end{cases}$$

$$\forall c_1, c_2, c_3 \in \mathbb{R}$$

$$\phi(t) \in \mathcal{R}(\{\phi_1(t), \phi_2(t), \phi_3(t)\})$$

Let say  $\bar{c} = \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$ , then  $\phi(t) = \begin{cases} t+2 & t \in [0, 1] \\ 3-t & t \in [1, 2] \\ 0 & \text{else} \end{cases}$



2.c)

c) Find the parameters,  $c_i$ 's to minimize  $\left\| y(t) - \sum_{i=1}^3 c_i \phi_i(t) \right\|^2$ , where

$$y(t) = \begin{cases} -1 & \text{if } t \in [-1, 1] \\ 2 & \text{if } t \in [2, 3] \\ 0 & \text{else} \end{cases}$$

Plot  $(y(t))$  and  $\hat{y}(t) = c_1 \phi_1(t) + c_2 \phi_2(t) + c_3 \phi_3(t)$  using the optimum  $c_i$ 's you found on the same graph.

$c_i$ 's that minimize  $\left\| y(t) - \sum_{i=1}^3 c_i \phi_i(t) \right\|^2$  can be found by using the previous results.

$$\bar{\Phi} = [\phi_1(t) \quad \phi_2(t) \quad \phi_3(t)] \quad . \quad \text{Then,} \quad \bar{\Phi}_{\text{avg}} \cdot \bar{\Phi} \cdot \bar{c} = \bar{\Phi}_{\text{avg}} \cdot y(t)$$

$$\bar{c} = (\bar{\Phi}_{\text{avg}} \cdot \bar{\Phi})^{-1} \cdot \bar{\Phi} \cdot y(t)$$

$$\bullet \quad \bar{\Phi}_{\text{avg}} \cdot \bar{\Phi} = \begin{bmatrix} \phi_1(t) \\ \phi_2(t) \\ \phi_3(t) \end{bmatrix} \cdot [\phi_1(t) \quad \phi_2(t) \quad \phi_3(t)]$$

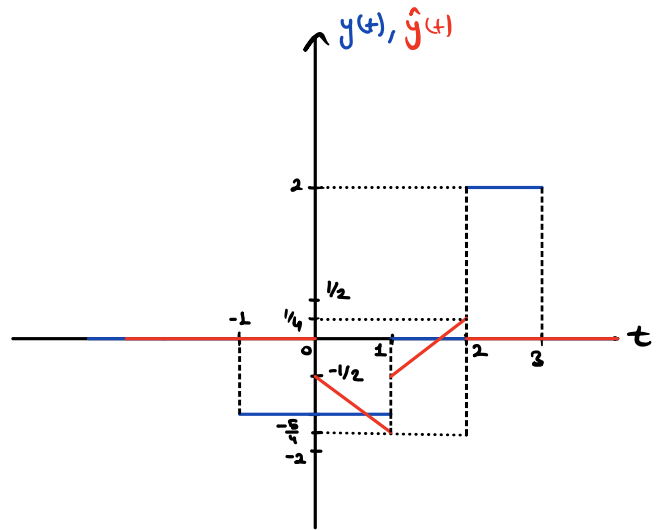
$$= \begin{bmatrix} \langle \phi_1(t), \phi_1(t) \rangle & \langle \phi_2(t), \phi_1(t) \rangle & \langle \phi_3(t), \phi_1(t) \rangle \\ \langle \phi_1(t), \phi_2(t) \rangle & \langle \phi_2(t), \phi_2(t) \rangle & \langle \phi_3(t), \phi_2(t) \rangle \\ \langle \phi_1(t), \phi_3(t) \rangle & \langle \phi_2(t), \phi_3(t) \rangle & \langle \phi_3(t), \phi_3(t) \rangle \end{bmatrix} = \begin{bmatrix} 1/3 & -1/2 & 0 \\ -1/2 & 2 & -1/2 \\ 0 & -1/2 & 1/3 \end{bmatrix}$$

$$(\bar{\Phi}_{\text{avg}} \cdot \bar{\Phi})^{-1} = \left( \begin{bmatrix} 1/3 & -1/2 & 0 \\ -1/2 & 2 & -1/2 \\ 0 & -1/2 & 1/3 \end{bmatrix} \right)^{-1} = \frac{\begin{bmatrix} 5/12 & 1/6 & -1/4 \\ 1/6 & 1/9 & -1/6 \\ -1/4 & -1/6 & 5/12 \end{bmatrix}^T}{\frac{1}{3} \cdot \left( \frac{2}{3} - \frac{1}{4} \right) + \frac{1}{2} \cdot \left( -\frac{1}{6} \right)} = \frac{\begin{bmatrix} 5/12 & 1/6 & -1/4 \\ 1/6 & 1/9 & -1/6 \\ -1/4 & -1/6 & 5/12 \end{bmatrix}}{\frac{5}{36} + \frac{1}{2} \cdot \left( -\frac{1}{6} \right)} = \frac{\begin{bmatrix} 5/12 & 1/6 & -1/4 \\ -1/6 & 1/9 & -1/6 \\ -1/4 & -1/6 & 5/12 \end{bmatrix}}{\frac{1}{18}} = \begin{bmatrix} 15/2 & 3 & -9/2 \\ 3 & 2 & -3 \\ 9/2 & -3 & 15/2 \end{bmatrix} = (\bar{\Phi}_{\text{avg}} \cdot \bar{\Phi})^{-1}$$

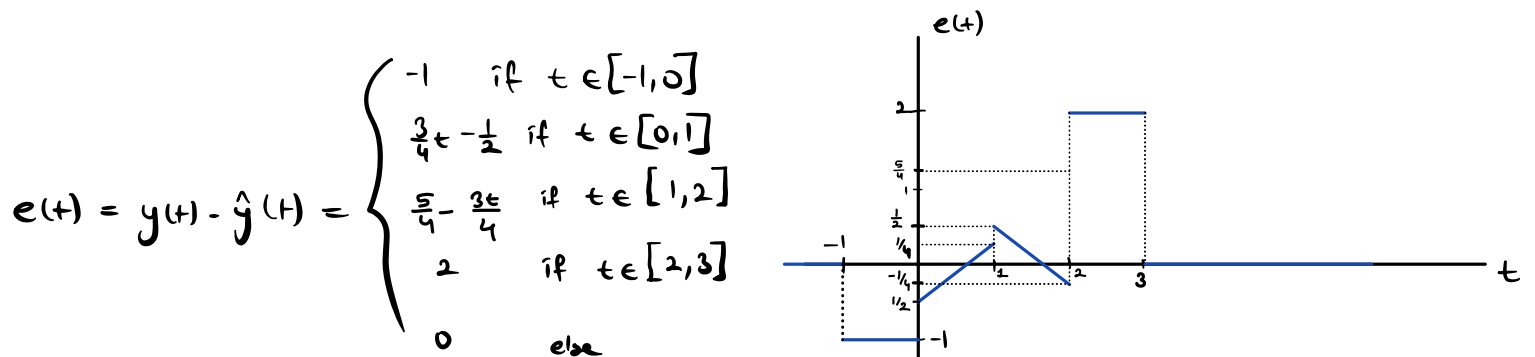
$$\bullet \bar{\bar{\Phi}}_{\mathbf{y}}(+) \cdot \mathbf{y}(+) = \begin{bmatrix} \langle \phi_1(+), \mathbf{y}(+) \rangle \\ \langle \phi_2(+), \mathbf{y}(+) \rangle \\ \langle \phi_3(+), \mathbf{y}(+) \rangle \end{bmatrix} = \begin{bmatrix} \int_0^1 (-t) dt \\ \int_0^1 dt \\ 0 \end{bmatrix} = \begin{bmatrix} -1/2 \\ 1 \\ 0 \end{bmatrix}$$

$$\bullet \bar{\mathbf{c}} = (\bar{\bar{\Phi}}_{\mathbf{y}} \cdot \bar{\bar{\Phi}})^{-1} \cdot (\bar{\bar{\Phi}}_{\mathbf{y}} \cdot \bar{\mathbf{y}}) = \begin{bmatrix} 15/2 & 3-9/2 \\ 3 & 2-3 \\ 9/2-3 & 15/2 \end{bmatrix} \cdot \begin{bmatrix} -1/2 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} -3/4 \\ 1/2 \\ -3/4 \end{bmatrix} = \bar{\mathbf{c}} = \begin{bmatrix} c_1 \\ c_2 \\ c_3 \end{bmatrix}$$

$$\bullet \hat{\mathbf{y}}(+) = \begin{cases} -\frac{3}{4}t - \frac{1}{2} & t \in [0, 1] \\ \frac{3}{4}t - \frac{5}{4} & t \in [1, 2] \\ 0 & \text{else} \end{cases}$$



2.d) Find and plot the error function  $e(t) = y(t) - \hat{y}(t)$ , and show that  $e(t)$  is orthogonal to  $\hat{y}(t)$ .



$$\bullet \langle e(t), \hat{y}(t) \rangle = -\int_0^1 \frac{9}{16}t^2 - \frac{1}{4} dt + \int_1^2 \frac{-9t^2}{16} + \frac{15t}{8} - \frac{25}{16} dt$$

$$= \left( -\frac{3t^3}{16} + \frac{t}{4} \right) \Big|_0^1 + \left( \frac{-3t^3}{16} + \frac{15t^2}{16} - \frac{25t}{16} \right) \Big|_1^2 = \frac{1}{16} + \left( -\frac{24}{16} + \frac{60}{16} - \frac{50}{16} + \frac{3}{16} - \frac{15}{16} + \frac{25}{16} \right) = \frac{1}{16} + \left( -\frac{1}{16} \right) = 0, \quad \boxed{e(t) \perp \hat{y}(t)}$$

as proved.



