

HW2 - 236201

313511602
931190987



HW2 - theory

Q1

a.

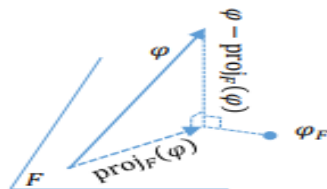
a)

First, we want to denote the n-term approximation of f in F using $Vec(\beta_{i_1}, \dots, \beta_{i_n})$.

Let us define this approximation as $\tilde{f}_n(t) = \sum_{i=i_1}^{i=i_n} f_i \beta_i(t)$ where $\{f_i\}_{i=i_1}^{i=i_n}$ are the representation coefficients.

The SE is:

$$\|f - \tilde{f}\|_2^2 = \left\| \underbrace{f - \text{proj}_F(f)}_{\perp F} + \underbrace{\text{proj}_F(f) - \tilde{f}_n}_{\in F} \right\|_2^2 \stackrel{\text{Pythagoras}}{=} \underbrace{\|f - \text{proj}_F(f)\|_2^2}_{\text{constant}} + \underbrace{\|\text{proj}_F(f) - \tilde{f}_n\|_2^2}_{\text{we want to minimize}}$$



At the draw $\varphi = f_n$.

As we can see, first element is a constant, so for minimizing the SE we must minimize the second element. We can see that choosing $\tilde{f} = \text{proj}_F(f)$ will give us zero, and because

$$\|\text{proj}_F(f) - \tilde{f}_n\|_2^2 \geq 0 \text{ it is the best choose for minimizing.}$$

We also know from linear algebra $\text{proj}_F(f) = \sum_{i=i_1}^{i=i_n} \underbrace{\langle f, \beta_i \rangle}_{\text{L2 is a Euclidean space}} \beta_i$.

With this knowledge we can say $\tilde{f}_n = \sum_{i=i_1}^{i=i_n} \langle f, \beta_i \rangle \beta_i$

$$\text{and SE is } \|f - \text{proj}_F(f)\|_2^2 \underset{\text{Pythagoras}}{=} \|f\|_2^2 - \|\text{proj}_F(f)\|_2^2 = \underbrace{\int_{\mathbb{R}} |f|^2}_{f \in L^2(\mathbb{R}, \mathbb{C})} - \left\| \left(\sum_{i=1}^{i_n} \langle f, \beta_i \rangle \beta_i \right) \right\|_2^2 \underset{(\beta_{i_1}, \dots, \beta_{i_n}) \text{ orthonormal functions}}{=} \int_{\mathbb{R}} |f|^2 - \sum_{i=1}^{i_n} \langle f, \beta_i \rangle^2$$

We can see in the same way that for k-term approximation,

$$\tilde{f}_k(t) = \sum_{i=1}^{i_k} \langle f(t), \beta_i(t) \rangle \beta_i(t) \quad \text{where } \tilde{f}_k(t) \text{ is the k-term approximation of } f \text{ in } F.$$

$$\text{So, we get } SE = \int_{\mathbb{R}} |f(t)|^2 - \sum_{i=1}^{i_k} \langle f(t), \beta_i(t) \rangle^2$$

Also, we can see from here that for minimizing the SE we have to take the k functions which

$$\langle f, \beta_i \rangle \text{ are the largest (so } \sum_{i=1}^{i_k} \langle f, \beta_i \rangle^2 \text{ will be maximum) .}$$

b)

The k functions for the best k-term approximation of f in F in the SE sense from $\binom{n}{k}$ possibilities are the k functions which $\langle f, \beta_i \rangle$ are the largest. The reason is as we explain above for minimizing the SE we need to maximize $\sum_{i=1}^{i_k} \langle f, \beta_i \rangle^2$.

We can get those by sorting all the $(\langle f, \beta_i \rangle)^2$ large to small, then we will take the corresponding first k β_i 's from this sorted list.

Therefore the associated $SE = \int_{\mathbb{R}} |f(t)|^2 - \sum_{i=1}^{i_k} \langle f(t), \beta_i(t) \rangle^2$ and choosing those β_i 's will minimize it the most.

It can happen that there will be $j_1 \neq j_2$ which $(\langle f, \beta_{j_1} \rangle)^2 = (\langle f, \beta_{j_2} \rangle)^2$ where we chose β_{j_1} for our k-term approximation of f in F in the SE sense, as we explained above, but β_{j_2} isn't in this choice (β_{j_2} should be the k + 1 choice). We can see that choosing the same group of function but replacing β_{j_1} with β_{j_2} will give us the same SE so the k-term approximation isn't unique.

b.

a)

As we saw at the last section we can say:

$$f_n = \sum_{i=1}^{i_n} \langle f, \beta_i \rangle \beta_i \quad \text{and} \quad \tilde{f}_n = \sum_{i=1}^{i_n} \langle f, \tilde{\beta}_i \rangle \tilde{\beta}_i$$

Where f_n will be n-approximation of f in F in the SE sense using β_i 's family

and \tilde{f}_n will be n-approximation of f in F in the SE sense using $\tilde{\beta}_i$'s family.

Both families are orthonormal bases of F (F is a subspace of E of finite dimension n).

So, we can say $f_n = \text{proj}_F(f) = \tilde{f}_n$, both n-approximation are the projections of f in F, which is a space of size n, means both n-approximation are the same.

b)

We can say now that because they are different finite families, the k-term approximation in this case will be apparently different because the projection will be different.

But, It's possible that β_i 's family $\neq \tilde{\beta}_i$'s family but for $i = i_1, \dots, i_k$ ($k < n$) $\beta_i = \tilde{\beta}_i$ (as we saw above we have to choose the k biggest $\langle f, \beta_i \rangle$ or $\langle f, \tilde{\beta}_i \rangle$).

In this case obviously the k-term approximation will be the same for both families, basically we can say that k-term approximation on each family probably won't be the same, but there is case that it will be.

Q2) Haar matrix and Walsh-Hadamard matrix

a.

i)

We need to show $H_4^* \cdot H_4 = I_{4 \times 4} = H_4 \cdot H_4^*$, and therefore H_4 is unitary.

$$\frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & -1 & -1 \\ \sqrt{2} & -\sqrt{2} & 0 & 0 \\ 0 & 0 & \sqrt{2} & -\sqrt{2} \end{bmatrix} \cdot \frac{1}{2} \begin{bmatrix} 1 & 1 & \sqrt{2} & 0 \\ 1 & 1 & -\sqrt{2} & 0 \\ 1 & -1 & 0 & \sqrt{2} \\ 1 & -1 & 0 & -\sqrt{2} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\frac{1}{2} \begin{bmatrix} 1 & 1 & \sqrt{2} & 0 \\ 1 & 1 & -\sqrt{2} & 0 \\ 1 & -1 & 0 & \sqrt{2} \\ 1 & -1 & 0 & -\sqrt{2} \end{bmatrix} \cdot \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & -1 & -1 \\ \sqrt{2} & -\sqrt{2} & 0 & 0 \\ 0 & 0 & \sqrt{2} & -\sqrt{2} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

ii)

We could construct the orthonormal Haar functions $\{\psi_i^H(t)\}_{i=1}^4$ with direct transform:

$$(\psi_1^H(t), \psi_2^H(t), \psi_3^H(t), \psi_4^H(t)) = \sqrt{4} \cdot (1\Delta_1(t), 1\Delta_2(t), 1\Delta_3(t), 1\Delta_4(t))H_4$$

Where $(1\Delta_1(t), 1\Delta_2(t), 1\Delta_3(t), 1\Delta_4(t))$ are the standard basis function and $\sqrt{4}$ is for normalization.

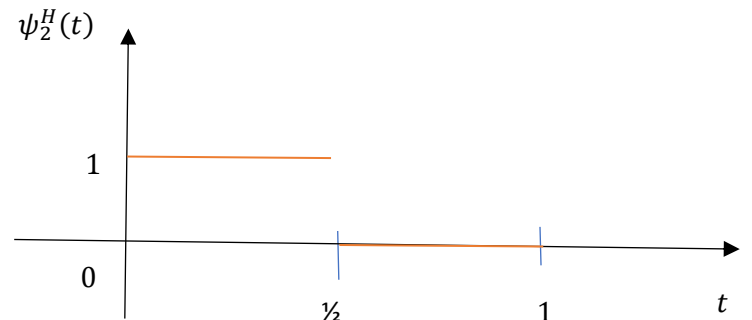
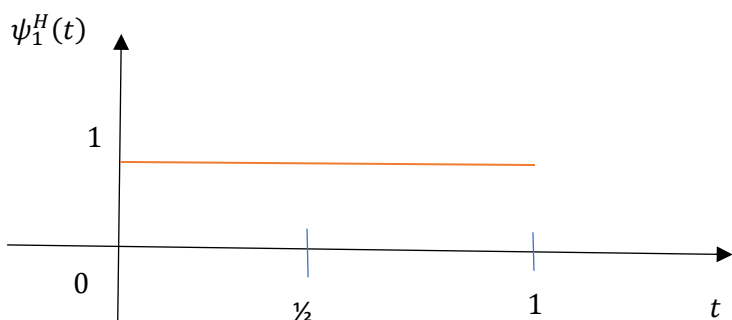
By that, we get:

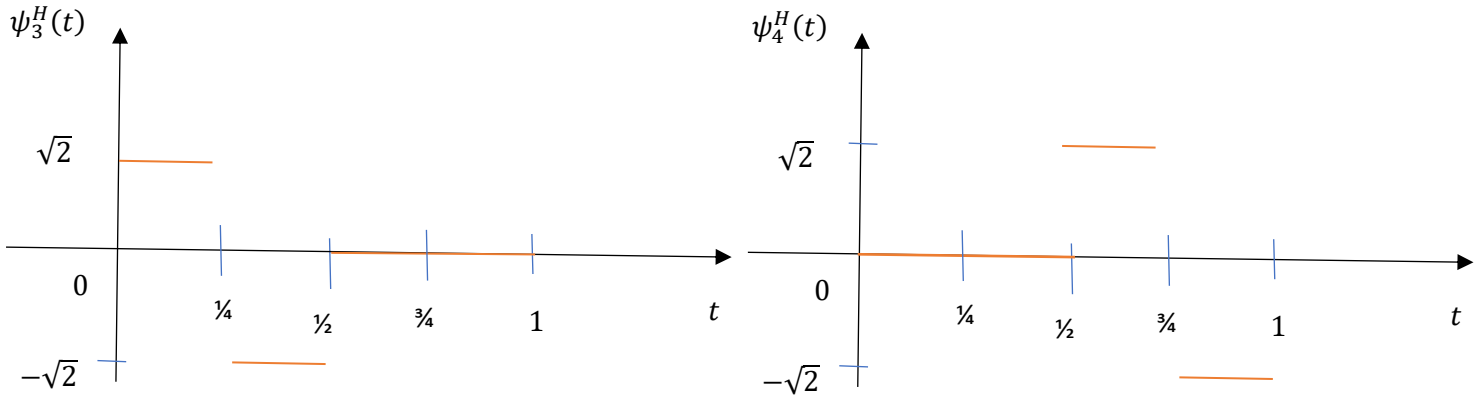
$$(\psi_1^H(t), \psi_2^H(t), \psi_3^H(t), \psi_4^H(t)) = 2 \cdot (1\Delta_1(t), 1\Delta_2(t), 1\Delta_3(t), 1\Delta_4(t)) \frac{1}{2} \begin{bmatrix} 1 & 1 & \sqrt{2} & 0 \\ 1 & 1 & -\sqrt{2} & 0 \\ 1 & -1 & 0 & \sqrt{2} \\ 1 & -1 & 0 & -\sqrt{2} \end{bmatrix} =$$

$$\left(1_{[0,1]}(t), 1_{[0,\frac{1}{2})}(t) + \left(-1_{[\frac{1}{2},1]}(t) \right), \sqrt{2}_{[0,\frac{1}{4})}(t) + (-\sqrt{2}_{[\frac{1}{4},\frac{1}{2})}(t)) + 0_{[\frac{1}{2},1]}(t), 0_{[0,\frac{1}{2})}(t) + \sqrt{2}_{[\frac{1}{2},\frac{3}{4})}(t) + \right.$$

$$\left. (-\sqrt{2}_{[\frac{3}{4},1]}(t)) \right)$$

$$\begin{pmatrix} \psi_1^H(t) \\ \psi_2^H(t) \\ \psi_3^H(t) \\ \psi_4^H(t) \end{pmatrix} = \begin{pmatrix} 1_{[0,1]}(t) \\ 1_{[0,\frac{1}{2})}(t) + \left(-1_{[\frac{1}{2},1]}(t) \right) \\ \sqrt{2}_{[0,\frac{1}{4})}(t) + (-\sqrt{2}_{[\frac{1}{4},\frac{1}{2})}(t)) + 0_{[\frac{1}{2},1]}(t) \\ 0_{[0,\frac{1}{2})}(t) + \sqrt{2}_{[\frac{1}{2},\frac{3}{4})}(t) + (-\sqrt{2}_{[\frac{3}{4},1]}(t)) \end{pmatrix}$$





Note: outside the boundary of $t \in [0, 1]$, $\psi_i^H(t)$ isn't define.

iii)

The approximation of ϕ using the Haar basis is $\tilde{\phi}(t) = \sum_{i=1}^4 \phi_i^{opt} \psi_i^H(t)$ and from the tutorial we know the best coefficients for each i in the MSE sense are $\phi_i^{opt} = \langle \phi(t), \psi_i^H(t) \rangle$ so,

$$\tilde{\phi}(t) = \sum_{i=1}^4 \langle \phi(t), \psi_i^H(t) \rangle \psi_i^H(t)$$

$$\begin{aligned} 1. \langle \phi(t), \psi_1^H(t) \rangle &= \langle a + b\cos(2\pi t) + c \cdot \cos^2(\pi t), 1_{[0,1]}(t) \rangle = \int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] \cdot 1_{[0,1]}(t) dt \\ &= \int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt = a \int_0^1 [1] dt + b \int_0^1 [\cos(2\pi t)] dt + c \int_0^1 [\cos^2(\pi t)] dt \\ &= a + 0 + c \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_0^1 = a + \frac{c}{2} \end{aligned}$$

$$\begin{aligned} 2. \langle \phi(t), \psi_2^H(t) \rangle &= \langle a + b\cos(2\pi t) + c \cdot \cos^2(\pi t), 1_{[0, \frac{1}{2}]}(t) + (-1)_{[\frac{1}{2}, 1]}(t) \rangle = \\ &= \int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] \cdot \left(1_{[0, \frac{1}{2}]}(t) + (-1)_{[\frac{1}{2}, 1]}(t) \right) dt \stackrel{\text{linearity}}{=} \\ &= \int_0^{\frac{1}{2}} [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt - \int_{\frac{1}{2}}^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt \end{aligned}$$

$$\begin{aligned} &= \left(\frac{a}{2} + 0 + c \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_0^{\frac{1}{2}} \right) - \left(\frac{a}{2} + 0 + c \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_{\frac{1}{2}}^1 \right) = \frac{a}{2} + \frac{c}{4} - \frac{a}{2} - \frac{c}{4} = 0 \end{aligned}$$

$$\begin{aligned} 3. \langle \phi(t), \psi_3^H(t) \rangle &= \langle a + b\cos(2\pi t) + c \cdot \cos^2(\pi t), \sqrt{2} 1_{[0, \frac{1}{4}]}(t) + (-\sqrt{2})_{[\frac{1}{4}, \frac{1}{2}]}(t) + 0_{[\frac{1}{2}, 1]}(t) \rangle = \\ &= \int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] \cdot \left(\sqrt{2} 1_{[0, \frac{1}{4}]}(t) + (-\sqrt{2})_{[\frac{1}{4}, \frac{1}{2}]}(t) + 0_{[\frac{1}{2}, 1]}(t) \right) dt \stackrel{\text{linearity}}{=} \\ &= \sqrt{2} \int_0^{\frac{1}{4}} [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt - \sqrt{2} \int_{\frac{1}{4}}^{\frac{1}{2}} [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt + 0 \cdot \dots = \end{aligned}$$

$$\begin{aligned} &= \sqrt{2} \left(\frac{a}{4} + \frac{b}{2\pi} [\sin(2\pi t)]_0^{\frac{1}{4}} + c \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_0^{\frac{1}{4}} \right) - \sqrt{2} \left(\frac{a}{4} + \frac{b}{2\pi} [\sin(2\pi t)]_{\frac{1}{4}}^{\frac{1}{2}} + c \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_{\frac{1}{4}}^{\frac{1}{2}} \right) = \\ &= \sqrt{2} \left(0 + \frac{b}{\pi} + \frac{c}{2\pi} \right) = \frac{\sqrt{2}}{\pi} \left(b + \frac{c}{2} \right) \end{aligned}$$

$$4. \langle \phi(t), \psi_4^H(t) \rangle = \langle a + b \cos(2\pi t) + c \cdot \cos^2(\pi t), 0_{[0, \frac{1}{2}]}(t) + \sqrt{2} \left[\frac{1}{2}, \frac{3}{4} \right](t) + (-\sqrt{2}) \left[\frac{3}{4}, 1 \right](t) \rangle =$$

$$\int_0^1 [a + b \cos(2\pi t) + c \cdot \cos^2(\pi t)]^* \left(0_{[0, \frac{1}{2}]}(t) + \sqrt{2} \left[\frac{1}{2}, \frac{3}{4} \right](t) + (-\sqrt{2}) \left[\frac{3}{4}, 1 \right](t) \right) dt \stackrel{\text{linearity}}{=} \int_0^1 [a + b \cos(2\pi t) + c \cdot \cos^2(\pi t)] dt - \sqrt{2} \int_{\frac{1}{2}}^{\frac{3}{4}} [a + b \cos(2\pi t) + c \cdot \cos^2(\pi t)] dt =$$

$$0 \cdot \dots + \sqrt{2} \int_{\frac{1}{2}}^{\frac{3}{4}} [a + b \cos(2\pi t) + c \cdot \cos^2(\pi t)] dt - \sqrt{2} \int_{\frac{3}{4}}^1 [a + b \cos(2\pi t) + c \cdot \cos^2(\pi t)] dt =$$

$$\sqrt{2} \left(\frac{a}{4} + \frac{b}{2\pi} \left[\sin(2\pi t) \right]_{\frac{1}{2}}^{\frac{3}{4}} + \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_{\frac{1}{2}}^{\frac{3}{4}} \right) - \sqrt{2} \left(\frac{a}{4} + \frac{b}{2\pi} \left[\sin(2\pi t) \right]_{\frac{3}{4}}^1 + \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_{\frac{3}{4}}^1 \right) =$$

$$\sqrt{2} \left(0 - \frac{b}{\pi} - \frac{c}{2\pi} \right) = -\frac{\sqrt{2}}{\pi} \left(b + \frac{c}{2} \right)$$

By that we got:

$$\tilde{\phi}_t = \left(a + \frac{c}{2} \right) \psi_1^H(t) + 0 \cdot \psi_2^H(t) + \frac{\sqrt{2}}{\pi} \left(b + \frac{c}{2} \right) \psi_3^H(t) - \frac{\sqrt{2}}{\pi} \left(b + \frac{c}{2} \right) \psi_4^H(t) = \left(a + \frac{c}{2} + \frac{2}{\pi} \left(b + \frac{c}{2} \right) \right)_{[0, \frac{1}{4}]} - \left(a + \frac{c}{2} + \frac{2}{\pi} \left(b + \frac{c}{2} \right) \right)_{[\frac{1}{4}, \frac{1}{2}]} \left(a + \frac{c}{2} - \frac{2}{\pi} \left(b + \frac{c}{2} \right) \right)_{[\frac{1}{2}, \frac{3}{4}]} + \left(a + \frac{c}{2} + \frac{2}{\pi} \left(b + \frac{c}{2} \right) \right)_{[\frac{3}{4}, 1]},$$

Associated MSE is:

$$\int_0^1 \left(\phi(t) - \tilde{\phi}(t) \right)^2 dt = \int_0^1 \phi^2(t) dt - 2 \int_0^1 \phi(t) \tilde{\phi}(t) dt + \int_0^1 \tilde{\phi}^2(t) dt \stackrel{\text{as we saw in class}}{=} \int_0^1 \phi^2(t) dt - \sum_{i=1}^4 \overbrace{(\phi_i^{opt})^2}^{**} dt =$$

$$\int_0^1 [a + b \cos(2\pi t) + c \cdot \cos^2(\pi t)]^2 dt - \sum_{i=1}^4 \left(\langle \phi(t), \psi_i^H(t) \rangle \right)^2 dt$$

$$* = \int_0^1 [a^2 + 2ab \cos(2\pi t) + 2ac \cdot \cos^2(\pi t) + b^2 \cos^2(2\pi t) + 2bc \cdot \cos(2\pi t) \cos^2(\pi t) + c^2 \cos^4(\pi t)] dt \stackrel{\text{linearity}}{=} a^2 \int_0^1 dt + 2ab \int_0^1 \cos(2\pi t) dt + 2ac \int_0^1 \cos^2(\pi t) dt +$$

$$b^2 \int_0^1 \cos^2(2\pi t) dt + 2bc \int_0^1 \cos(2\pi t) \cos^2(\pi t) dt + c^2 \int_0^1 \cos^4(\pi t) dt = a^2 + 0 + ac + \frac{b^2}{2} + \frac{bc}{2} + \frac{3c^2}{8}$$

$$** = \left(a + \frac{c}{2} \right)^2 + 0^2 + \left(\frac{\sqrt{2}}{\pi} \left(b + \frac{c}{2} \right) \right)^2 + \left(-\frac{\sqrt{2}}{\pi} \left(b + \frac{c}{2} \right) \right)^2 = a^2 + ac + \frac{c^2}{4} + \frac{4}{\pi^2} \left(b^2 + bc + \frac{c^2}{4} \right) =$$

$$a^2 + ac + c^2 \left(\frac{1}{4} + \frac{1}{\pi^2} \right) + \frac{4b^2}{\pi^2} + \frac{4bc}{\pi^2}$$

, By that

$$MSE = a^2 + ac + \frac{b^2}{2} + \frac{bc}{2} + \frac{3c^2}{8} \left(a^2 + ac + c^2 \left(\frac{1}{4} + \frac{1}{\pi^2} \right) + \frac{4b^2}{\pi^2} + \frac{4bc}{\pi^2} \right) = b^2 \left(\frac{1}{2} - \frac{4}{\pi^2} + \right)$$

$$bc \left(\frac{1}{2} - \frac{4}{\pi^2} \right) + c^2 \left(\frac{1}{8} - \frac{1}{\pi^2} \right)$$

iv)

First, let us sort the coefficients by the squared coefficients $(\phi_i^{opt})^2 = (\langle \phi(t), \psi_i^H(t) \rangle)^2$ in decreasing order, we can simply sort the absolute values of each:

$$|\phi_1^{opt}| = a + \frac{c}{2} > |\phi_4^{opt}| = \frac{\sqrt{2}}{\pi} \left(b + \frac{c}{2}\right) = |\phi_3^{opt}| > |\phi_2^{opt}| = 0$$

Because $a + \frac{c}{2} \geq b + \frac{c}{2}$ and $\frac{\sqrt{2}}{\pi} < 1$ and for each $x \in \mathbb{R}, |x| \geq 0$

As we saw in class, for k -term approximation $MSE = \int_0^1 \phi^2(t) dt - \sum_{i=1}^k (\phi_i^{opt})^2$ and for minimizing it we have to choose the first k functions with the biggest squared projections (maximize $\sum_{i=1}^k (\phi_i^{opt})^2$ will minimize the MSE).

1. The best 1-term approximation($k=1$)

$$\tilde{\phi}_1(t) = \phi_1^{opt} \cdot \psi_1^H(t) = \left(a + \frac{c}{2}\right) \cdot 1_{[0,1]}(t)$$

2. The best 2-term approximation($k=2$)

$$\begin{aligned} \tilde{\phi}_2(t) &= \phi_1^{opt} \cdot \psi_1^H(t) + \phi_4^{opt} \psi_4^H(t) = \left(a + \frac{c}{2}\right) \cdot 1_{[0,1]}(t) - \frac{\sqrt{2}}{\pi} \left(b + \frac{c}{2}\right) (0_{[0, \frac{1}{2}]}(t) + \\ &\sqrt{2}_{[\frac{1}{2}, \frac{3}{4}]}(t) + \left(-\sqrt{2}_{[\frac{3}{4}, 1]}(t)\right)) = \left(a + \frac{c}{2}\right)_{[0, \frac{1}{2}]}(t) + \left(a + c \left(\frac{1}{2} + \frac{1}{\pi}\right) + \frac{2b}{\pi}\right)_{[\frac{1}{2}, \frac{3}{4}]} + \\ &\left(a + c \left(\frac{1}{2} - \frac{1}{\pi}\right) - \frac{2b}{\pi}\right)_{[\frac{3}{4}, 1]} \end{aligned}$$

3. The best 3-term approximation($k=3$)

We can see $\phi_2^{opt} = 0$, so 3-term approximation will be the same as 4-term approximation, and they both will be the same as we saw at section (iii) because this Haar basis is 4-dimension. So, we get

$$\begin{aligned} \tilde{\phi}_3(t) &= \tilde{\phi}_4(t) = \left(a + \frac{c}{2} + \frac{2}{\pi} \left(b + \frac{c}{2}\right)\right)_{[0, \frac{1}{4}]} + \left(a + \frac{c}{2} - \frac{2}{\pi} \left(b + \frac{c}{2}\right)\right)_{[\frac{1}{4}, \frac{1}{2}]} + \left(a + \frac{c}{2} - \right. \\ &\left. \frac{2}{\pi} \left(b + \frac{c}{2}\right)\right)_{[\frac{1}{2}, \frac{3}{4}]} + \left(a + \frac{c}{2} + \frac{2}{\pi} \left(b + \frac{c}{2}\right)\right)_{[\frac{3}{4}, 1]} \end{aligned}$$

v)

$$a = \frac{1}{\pi}, b = 1, c = \frac{3}{2}$$

First, let us sort the coefficients by the squared coefficients $(\phi_i^{opt})^2 = (\langle \phi(t), \psi_i^H(t) \rangle)^2$ in decreasing order, we can simply sort by the absolute values of each:

$$\phi_1^{opt} = a + \frac{c}{2} = \frac{1}{\pi} + \frac{3}{4} > |\phi_4^{opt}| = \frac{\sqrt{2}}{\pi} \left(b + \frac{c}{2}\right) = \frac{7\sqrt{2}}{4\pi} = |\phi_3^{opt}| > |\phi_2^{opt}| = 0$$

1. The best 1-term approximation($k=1$)

$$\tilde{\phi}_1(t) = \phi_1^{opt} \cdot \psi_1^H(t) = \left(\frac{1}{\pi} + \frac{3}{4}\right) \cdot 1_{[0,1]}(t) = 1.068 \cdot 1_{[0,1]}(t)$$

2. The best 2-term approximation($k=2$)

$$\begin{aligned} \tilde{\phi}_2(t) &= \phi_1^{opt} \cdot \psi_1^H(t) + \phi_4^{opt} \psi_4^H(t) = \left(a + \frac{c}{2}\right)_{[0, \frac{1}{2}]}(t) + \left(a + c \left(\frac{1}{2} + \frac{1}{\pi}\right) + \frac{2b}{\pi}\right)_{[\frac{1}{2}, \frac{3}{4}]} + \\ &\left(a + c \left(\frac{1}{2} - \frac{1}{\pi}\right) - \frac{2b}{\pi}\right)_{[\frac{3}{4}, 1]} = (1.068)_{[0, \frac{1}{2}]}(t) + (2.182)_{[\frac{1}{2}, \frac{3}{4}]} - (0.0458)_{[\frac{3}{4}, 1]} \end{aligned}$$

3. The best 3-term approximation($k=3$), as we said 3-term approximation = 4-term approximation

$$\begin{aligned}\tilde{\phi}_3(t) = \tilde{\phi}_4(t) &= \left(a + \frac{c}{2} + \frac{2}{\pi}\left(b + \frac{c}{2}\right)\right)_{\left[0, \frac{1}{4}\right)} + \left(a + \frac{c}{2} - \frac{2}{\pi}\left(b + \frac{c}{2}\right)\right)_{\left[\frac{1}{4}, \frac{1}{2}\right)} + \left(a + \frac{c}{2} - \frac{2}{\pi}\left(b + \frac{c}{2}\right)\right)_{\left[\frac{1}{2}, \frac{3}{4}\right)} \\ &+ \left(a + \frac{c}{2} + \frac{2}{\pi}\left(b + \frac{c}{2}\right)\right)_{\left[\frac{3}{4}, 1\right]} = (2.182)_{\left[0, \frac{1}{4}\right)} + (-0.0458)_{\left[\frac{1}{4}, \frac{1}{2}\right)} + (-0.0458)_{\left[\frac{1}{2}, \frac{3}{4}\right)} + (2.182)_{\left[\frac{3}{4}, 1\right]} \\ &= (2.182)_{\left[0, \frac{1}{4}\right)} + (-0.0458)_{\left(\frac{1}{2}, \frac{3}{4}\right)} + (2.182)_{\left[\frac{3}{4}, 1\right]}\end{aligned}$$

b.

i)

We need to show $W_4^* \cdot W_4 = I_{4 \times 4} = W_4 \cdot W_4^*$, and therefore H_4 is unitary.

$$(W_4^* = W_4)$$

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

ii)

We could construct the orthonormal WH functions $\{\psi_i^W(t)\}_{i=1}^4$ with direct transform:

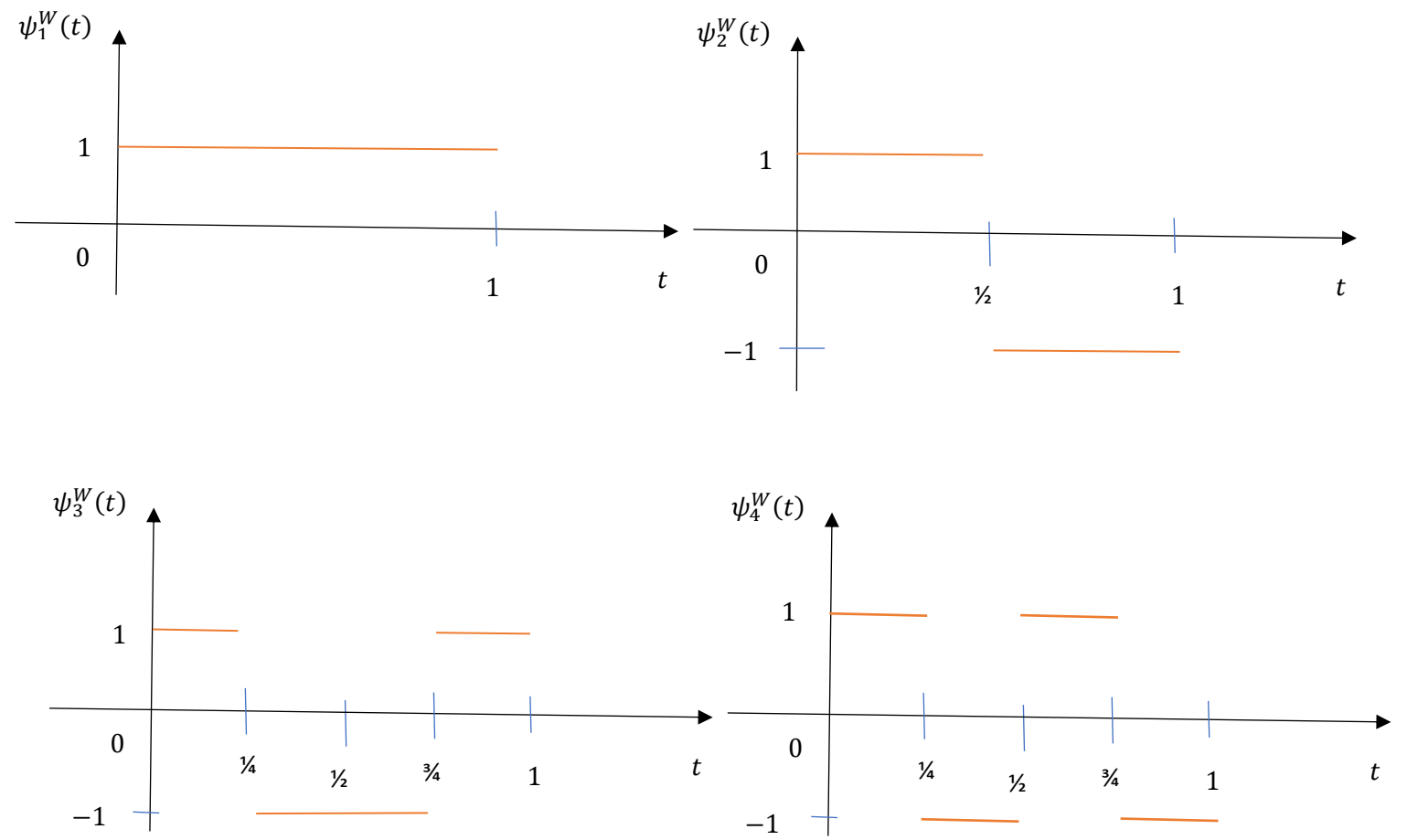
$$(\psi_1^W(t), \psi_2^W(t), \psi_3^W(t), \psi_4^W(t)) = \sqrt{4} \cdot (1\Delta_1(t), 1\Delta_2(t), 1\Delta_3(t), 1\Delta_4(t))W_4$$

Where $(1\Delta_1(t), 1\Delta_2(t), 1\Delta_3(t), 1\Delta_4(t))$ are the standard basis function and $\sqrt{4}$ is for normalization.

By that, we get:

$$\begin{aligned}(\psi_1^W(t), \psi_2^W(t), \psi_3^W(t), \psi_4^W(t)) &= \sqrt{4} \cdot (1\Delta_1(t), 1\Delta_2(t), 1\Delta_3(t), 1\Delta_4(t)) \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \\ 1 & -1 & 1 & -1 \end{bmatrix} = \\ &\left(1_{[0,1]}(t), 1_{\left[0, \frac{1}{2}\right)}(t) + \left(-1_{\left(\frac{1}{2}, 1\right]}(t)\right), 1_{\left[0, \frac{1}{4}\right)}(t) + \left(-1_{\left(\frac{1}{4}, \frac{1}{2}\right)}(t)\right) + 1_{\left(\frac{3}{4}, 1\right]}(t), 1_{\left[0, \frac{1}{4}\right)}(t) + \left(-1_{\left(\frac{1}{4}, \frac{1}{2}\right)}(t)\right) + \right. \\ &\left. 1_{\left(\frac{1}{2}, \frac{3}{4}\right)}(t) + \left(-1_{\left(\frac{3}{4}, 1\right]}(t)\right)\right)\end{aligned}$$

$$\begin{pmatrix} \psi_1^W(t) \\ \psi_2^W(t) \\ \psi_3^W(t) \\ \psi_4^W(t) \end{pmatrix} = \begin{pmatrix} 1_{[0,1]}(t) \\ 1_{\left[0, \frac{1}{2}\right)}(t) + \left(-1_{\left(\frac{1}{2}, 1\right]}(t)\right) \\ 1_{\left[0, \frac{1}{4}\right)}(t) + \left(-1_{\left(\frac{1}{4}, \frac{1}{2}\right)}(t)\right) + 1_{\left(\frac{3}{4}, 1\right]}(t) \\ 1_{\left[0, \frac{1}{4}\right)}(t) + \left(-1_{\left(\frac{1}{4}, \frac{1}{2}\right)}(t)\right) + 1_{\left(\frac{1}{2}, \frac{3}{4}\right)}(t) + \left(-1_{\left(\frac{3}{4}, 1\right]}(t)\right) \end{pmatrix}$$



Note: outside the boundary of $t \in [0, 1]$, $\psi_i^W(t)$ isn't define.

iii)

The approximation of ϕ using the WH basis is $\tilde{\phi}(t) = \sum_{i=1}^4 \phi_i^{opt} \psi_i^W(t)$ and from the tutorial we know the best coefficients for each i in the MSE sense are $\phi_i^{opt} = \langle \phi(t), \psi_i^W(t) \rangle$ so,

$$\tilde{\phi}(t) = \sum_{i=1}^4 \langle \phi(t), \psi_i^W(t) \rangle \psi_i^W(t)$$

$$\begin{aligned} 1. \langle \phi(t), \psi_1^W(t) \rangle &= \langle a + b\cos(2\pi t) + c \cdot \cos^2(\pi t), 1_{[0,1]}(t) \rangle = \int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)]^* 1_{[0,1]}(t) dt \\ &= \int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt = a \int_0^1 [1] dt + b \int_0^1 [\cos(2\pi t)] dt + c \int_0^1 [\cos^2(\pi t)] dt \\ &= a + 0 + c \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_0^1 = a + \frac{c}{2} \end{aligned}$$

$$\begin{aligned} 2. \langle \phi(t), \psi_2^W(t) \rangle &= \langle a + b\cos(2\pi t) + c \cdot \cos^2(\pi t), 1_{[0, \frac{1}{2}]}(t) + \left(-1_{[\frac{1}{2}, 1]}(t) \right) \rangle = \\ &= \int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)]^* \left(1_{[0, \frac{1}{2}]}(t) + \left(-1_{[\frac{1}{2}, 1]}(t) \right) \right) dt \stackrel{\text{linearity}}{=} \\ &= \int_0^{\frac{1}{2}} [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt - \int_{\frac{1}{2}}^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt \end{aligned}$$

$$= \int_0^{\frac{1}{2}} [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt - \int_{\frac{1}{2}}^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt = \left(\frac{a}{2} + 0 + \right.$$

$$\left. c \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_0^{\frac{1}{2}} \right) - \left(\frac{a}{2} + 0 + c \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_{\frac{1}{2}}^1 \right) = \frac{a}{2} + \frac{c}{4} - \frac{a}{2} - \frac{c}{4} = 0$$

$$3. \langle \phi(t), \psi_3^W(t) \rangle = \langle a + b\cos(2\pi t) + c \cdot \cos^2(\pi t), 1_{[1, \frac{1}{4}]}(t) + (-1_{(\frac{1}{4}, \frac{3}{4}]}(t)) + 1_{(\frac{3}{4}, 1]}(t) \rangle =$$

$$\int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)]^* (1_{[1, \frac{1}{4}]}(t) + (-1_{(\frac{1}{4}, \frac{3}{4}]}(t)) + 1_{(\frac{3}{4}, 1]}(t)) dt \stackrel{\text{linearity}}{=} \int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt - \int_{\frac{1}{4}}^{\frac{3}{4}} [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt + \int_{\frac{3}{4}}^1 [a +$$

$$b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt = \left(\frac{a}{4} + \frac{b}{2\pi} [\sin(2\pi t)]_0^{\frac{1}{4}} + c \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_0^{\frac{1}{4}} - \frac{a}{2} - \frac{b}{2\pi} [\sin(2\pi t)]_{\frac{1}{4}}^{\frac{3}{4}} - c \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_{\frac{1}{4}}^{\frac{3}{4}} + \frac{a}{4} + \frac{b}{2\pi} [\sin(2\pi t)]_{\frac{3}{4}}^1 + c \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_{\frac{3}{4}}^1 \right) = \frac{2}{\pi} \left(b + \frac{c}{2} \right)$$

$$4. \langle \phi(t), \psi_4^W(t) \rangle = \langle a + b\cos(2\pi t) + c \cdot \cos^2(\pi t), 1_{[1, \frac{1}{4}]}(t) + (-1_{(\frac{1}{4}, \frac{2}{4}]}(t)) + 1_{(\frac{2}{4}, \frac{3}{4}]}(t) + (-1_{(\frac{3}{4}, 1]}(t)) \rangle = \int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)]^* \left(1_{[1, \frac{1}{4}]}(t) + (-1_{(\frac{1}{4}, \frac{2}{4}]}(t)) + 1_{(\frac{2}{4}, \frac{3}{4}]}(t) + (-1_{(\frac{3}{4}, 1]}(t)) \right) dt \stackrel{\text{linearity}}{=} \int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt - \int_{\frac{1}{4}}^{\frac{2}{4}} [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt + \int_{\frac{2}{4}}^{\frac{3}{4}} [a +$$

$$b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt - \int_{\frac{3}{4}}^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)] dt = \left(\frac{a}{4} + \frac{b}{2\pi} [\sin(2\pi t)]_0^{\frac{1}{4}} + \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_0^{\frac{1}{4}} - \frac{a}{4} - \frac{b}{2\pi} [\sin(2\pi t)]_{\frac{1}{4}}^{\frac{2}{4}} - \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_{\frac{1}{4}}^{\frac{2}{4}} + \frac{a}{4} + \frac{b}{2\pi} [\sin(2\pi t)]_{\frac{2}{4}}^{\frac{3}{4}} + \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_{\frac{2}{4}}^{\frac{3}{4}} - \frac{a}{4} - \frac{b}{2\pi} [\sin(2\pi t)]_{\frac{3}{4}}^1 - \left[\frac{\sin(2\pi t)}{4\pi} + \frac{t}{2} \right]_{\frac{3}{4}}^1 \right) = 0$$

By that we got:

$$\tilde{\phi}(t) = \left(a + \frac{c}{2} \right) \psi_1^H(t) + 0 \cdot \psi_2^H(t) + \frac{2}{\pi} \left(b + \frac{c}{2} \right) \psi_3^H(t) - 0 \cdot \psi_4^H(t) = \left(a + \frac{c}{2} + \frac{2}{\pi} \left(b + \frac{c}{2} \right) \right)_{0, \frac{1}{4}} + \left(a + \frac{c}{2} - \frac{2}{\pi} \left(b + \frac{c}{2} \right) \right)_{\frac{1}{4}, \frac{2}{4}} + \left(a + \frac{c}{2} - \frac{2}{\pi} \left(b + \frac{c}{2} \right) \right)_{\frac{2}{4}, \frac{3}{4}} + \left(a + \frac{c}{2} + \frac{2}{\pi} \left(b + \frac{c}{2} \right) \right)_{\frac{3}{4}, 1}$$

Associated MSE is:

$$\int_0^1 \left(\phi(t) - \tilde{\phi}(t) \right)^2 dt = \int_0^1 \phi^2(t) dt - 2 \int_0^1 \phi(t) \tilde{\phi}(t) dt + \int_0^1 \tilde{\phi}^2(t) dt \stackrel{\text{as we saw in class}}{=} \int_0^1 \phi^2(t) dt - \sum_{i=1}^4 \overbrace{(\phi_i^{opt})^2}^{**} dt = \int_0^1 [a + b\cos(2\pi t) + c \cdot \cos^2(\pi t)]^2 dt - \sum_{i=1}^4 \left(\langle \phi(t), \psi_i^W(t) \rangle \right)^2 dt$$

$$* = \int_0^1 [a^2 + 2ab\cos(2\pi t) + 2ac \cdot \cos^2(\pi t) + b^2 \cos^2(2\pi t) + 2bc \cdot \cos(2\pi t)\cos^2(\pi t) + c^2 \cos^4(\pi t)] dt \stackrel{\text{linearity}}{=} a^2 \int_0^1 dt + 2ab \int_0^1 \cos(2\pi t) dt + 2ac \int_0^1 \cos^2(\pi t) dt + b^2 \int_0^1 \cos^2(2\pi t) dt + 2bc \int_0^1 \cos(2\pi t)\cos^2(\pi t) dt + c^2 \int_0^1 \cos^4(\pi t) dt = a^2 + 0 + ac + \frac{b^2}{2} + \frac{bc}{2} + \frac{3c^2}{8}$$

$$** = \left(a + \frac{c}{2}\right)^2 + 0^2 + \left(\frac{2}{\pi}\left(b + \frac{c}{2}\right)\right)^2 + (0)^2 = a^2 + ac + \frac{c^2}{4} + \frac{4}{\pi^2}\left(b^2 + bc + \frac{c^2}{4}\right) = a^2 + ac + c^2\left(\frac{1}{4} + \frac{1}{\pi^2}\right) + \frac{4b^2}{\pi^2} + \frac{4bc}{\pi^2}$$

, By that

$$MSE = a^2 + ac + \frac{b^2}{2} + \frac{bc}{2} + \frac{3c^2}{8} \left(a^2 + ac + c^2\left(\frac{1}{4} + \frac{1}{\pi^2}\right) + \frac{4b^2}{\pi^2} + \frac{4bc}{\pi^2}\right) + \frac{1}{2} \left(a^2 + ac + c^2\left(\frac{1}{4} + \frac{1}{\pi^2}\right) + \frac{4b^2}{\pi^2} + \frac{4bc}{\pi^2}\right)$$

First, let us sort the coefficients by the squared coefficients $(\phi_i^{opt})^2 = (\langle \phi(t), \psi_i^W(t) \rangle)^2$ in decreasing order, we can simply sort the absolute values of each:

$$|\phi_1^{opt}| = a + \frac{c}{2} > |\phi_3^{opt}| = \frac{2}{\pi}\left(b + \frac{c}{2}\right) > |\phi_4^{opt}| = 0 = |\phi_2^{opt}|$$

Because $a + \frac{c}{2} \geq b + \frac{c}{2}$ and $\frac{2}{\pi} < 1$ and for each $x \in \mathbb{R}, |x| \geq 0$

As we saw in class, for k -term approximation $MSE = \int_0^1 \phi^2(t) dt - \sum_{i=1}^k (\phi_i^{opt})^2$ and for minimizing it we have to choose the first k functions with the biggest squared projections (maximize $\sum_{i=1}^k (\phi_i^{opt})^2$ will minimize the MSE).

1. The best 1-term approximation ($k=1$)

$$\tilde{\phi}_1(t) = \phi_1^{opt} \cdot \psi_1^W(t) = \left(a + \frac{c}{2}\right) \cdot 1_{[0,1]}(t)$$

2. The best 2-term approximation ($k=2$)

We can see $\phi_4^{opt} = 0 = \phi_2^{opt}$, so 2-term approximation will be the same as 3/4-term approximation, and all of them will be the same as we saw at section (iii) because this WH basis is 4-dimension. So, we get

$$\begin{aligned} \tilde{\phi}_2(t) &= \phi_1^{opt} \cdot \psi_1^W(t) + \phi_3^{opt} \cdot \psi_3^W(t) = \left(a + \frac{c}{2}\right) \cdot 1_{[0,1]}(t) + \frac{2}{\pi}\left(b + \frac{c}{2}\right) \left(1_{\left[0, \frac{1}{4}\right]}(t) + \right. \\ &\quad \left. (-1_{\left(\frac{1}{4}, \frac{3}{4}\right)}(t)) + 1_{\left(\frac{3}{4}, 1\right]}(t)\right) = \left(a + \frac{c}{2} + \frac{2}{\pi}\left(b + \frac{c}{2}\right)\right) 1_{\left[0, \frac{1}{4}\right]}(t) + \left(a + \frac{c}{2} - \frac{2}{\pi}\left(b + \frac{c}{2}\right)\right) 1_{\left(\frac{1}{4}, \frac{3}{4}\right)}(t) \\ &\quad + \left(a + \frac{c}{2} + \frac{2}{\pi}\left(b + \frac{c}{2}\right)\right) 1_{\left(\frac{3}{4}, 1\right]}(t) \end{aligned}$$

v)

$$a = \frac{1}{\pi}, b = 1, c = \frac{3}{2}$$

First, let us sort the coefficients by the squared coefficients $(\phi_i^{opt})^2 = (\langle \phi(t), \psi_i^H(t) \rangle)^2$ in decreasing order, we can simply sort by the absolute values of each:

$$|\phi_3^{opt}| = \frac{2}{\pi}\left(b + \frac{c}{2}\right) = \frac{7}{2\pi} > |\phi_1^{opt}| = a + \frac{c}{2} = \frac{1}{\pi} + \frac{3}{4} > |\phi_4^{opt}| = 0 = |\phi_2^{opt}|$$

1. The best 1-term approximation(k=1)

$$\tilde{\phi}_1(t) = \phi_3^{opt} \cdot \psi_3^W(t) = \frac{2}{\pi} \left(b + \frac{c}{2} \right) \cdot (1_{[0, \frac{1}{4})}(t) + (-1_{(\frac{1}{4}, \frac{3}{4})}(t)) + 1_{(\frac{3}{4}, 1]}(t)) = 1.114 \cdot 1_{[0, \frac{1}{4})}(t) - 1.114 \cdot (-1)_{(\frac{1}{4}, \frac{3}{4})}(t) + 1.114 \cdot 1_{[\frac{3}{4}, 1]}(t)$$

2. The best 2-term approximation(k=2), as we said 2-term approximation = 3/4-term approximation

$$\tilde{\phi}_2(t) = \phi_3^{opt} \cdot \psi_3^W(t) + \phi_1^{opt} \psi_3^W(t) = \left(a + \frac{c}{2} + \frac{2}{\pi} \left(b + \frac{c}{2} \right) \right)_{[0, \frac{1}{4})}(t) + \left(a + \frac{c}{2} - \frac{2}{\pi} \left(b + \frac{c}{2} \right) \right)_{(\frac{1}{4}, \frac{3}{4})}(t) + \left(a + \frac{c}{2} + \frac{2}{\pi} \left(b + \frac{c}{2} \right) \right)_{[\frac{3}{4}, 1]}(t) = (2.182)_{[0, \frac{1}{4})}(t) - (0.04578)_{(\frac{1}{4}, \frac{3}{4})}(t) + (2.182)_{[\frac{3}{4}, 1]}(t)$$

$$\begin{aligned}
 H_1 &= [1], \\
 H_2 &= \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}, \\
 H_4 &= \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix},
 \end{aligned}$$

matrix hadamard

Q3

a.

We will use induction :

Base H_1 can be written as $H_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$ where $\lambda_{2^1} = \frac{1}{\sqrt{2}}$ $A = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$

H_1 is symmetric and real

$$H_1 H_1^T = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} = I_{2^1 \times 2^1} \rightarrow$$

therefore, H_1 is also unitary

step : we assume the condition applies for $H_{2^{n-1}}$

By definition: $H_{2^n} = \frac{1}{\sqrt{2}} \begin{bmatrix} H_{2^{n-1}} & H_{2^{n-1}} \\ H_{2^{n-1}} & -H_{2^{n-1}} \end{bmatrix}$

$$\begin{aligned} H_{2^n}^T &= \frac{1}{\sqrt{2}} \begin{bmatrix} H_{2^{n-1}} & H_{2^{n-1}} \\ H_{2^{n-1}} & -H_{2^{n-1}} \end{bmatrix}^T = \\ &= \frac{1}{\sqrt{2}} \begin{bmatrix} H_{2^{n-1}}^T & H_{2^{n-1}}^T \\ H_{2^{n-1}}^T & -H_{2^{n-1}}^T \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} H_{2^{n-1}} & H_{2^{n-1}} \\ H_{2^{n-1}} & -H_{2^{n-1}} \end{bmatrix} = H_{2^n} \end{aligned}$$

↑
by induction $H_{2^{n-1}}^T$ is symmetric $\rightarrow H_{2^{n-1}}^T = H_{2^{n-1}}$

Therefore, H_{2^n} is symmetric

Because $H_{2^{n-1}}$ is real, it can be seen that H_{2^n} is also real

$$\begin{aligned} H_{2^n} H_{2^n}^T &= H_{2^n} H_{2^n} = \frac{1}{2} \begin{bmatrix} H_{2^{n-1}} & H_{2^{n-1}} \\ H_{2^{n-1}} & -H_{2^{n-1}} \end{bmatrix} \begin{bmatrix} H_{2^{n-1}} & H_{2^{n-1}} \\ H_{2^{n-1}} & -H_{2^{n-1}} \end{bmatrix} = \\ &= \frac{1}{2} \begin{bmatrix} 2H_{2^{n-1}}^2 & 0 \\ 0 & 2H_{2^{n-1}}^2 \end{bmatrix} = \begin{bmatrix} H_{2^{n-1}}^2 & 0 \\ 0 & H_{2^{n-1}}^2 \end{bmatrix} = I_{2^n \times 2^n} \end{aligned}$$

↑
by induction $H_{2^{n-1}}^2 = I_{2^{n-1} \times 2^{n-1}}$

Therefore, H_{2^n} is unitary

$$\begin{aligned} H_{2^n} &= \frac{1}{\sqrt{2}} \begin{bmatrix} H_{2^{n-1}} & H_{2^{n-1}} \\ H_{2^{n-1}} & -H_{2^{n-1}} \end{bmatrix} = \\ &= \frac{1}{\sqrt{2}} \begin{bmatrix} \lambda_{2^{n-1}} A_{n-1} & \lambda_{2^{n-1}} A_{n-1} \\ \lambda_{2^{n-1}} A_{n-1} & -\lambda_{2^{n-1}} A_{n-1} \end{bmatrix} = \\ &= \frac{\lambda_{2^{n-1}}}{\sqrt{2}} \begin{bmatrix} A_{n-1} & A_{n-1} \\ A_{n-1} & -A_{n-1} \end{bmatrix} \end{aligned}$$

and we will denote $\lambda_{2^n} = \frac{\lambda_{2^{n-1}}}{\sqrt{2}}$, $A_n = \begin{bmatrix} A_{n-1} & A_{n-1} \\ A_{n-1} & -A_{n-1} \end{bmatrix}$

and they hold the conditions.

We conclude, H_{2^n} is real, symmetric, unitary and can be formulated as λ_{2^n} and A_n for all $n \in \mathbb{N}$.

b1.

We will consider the two possible scenarios, according to if the end of s_1 is the same as the start of s_2 or not.

$$S(s_1 s_2) = \begin{cases} S(s_1) + S(s_2) + 1 & \text{where the meeting point is different} \\ S(s_1) + S(s_2) & \text{where the meeting point is the same} \end{cases}$$

b2.

We will use induction to prove the theorem :

base : For $H_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$ it is clear that $S(r_1) = 0$, $S(r_2) = 1$, therefore satisfies the ensemble equality

step : we will assume that the equality holds for $H_{2^{n-1}}$

we will denote r_i to be the i - th row of $H_{2^{n-1}}$

Considering the structure $H_{2^n} = \begin{bmatrix} H_{2^{n-1}} & H_{2^{n-1}} \\ H_{2^{n-1}} & -H_{2^{n-1}} \end{bmatrix}$, we will denote the i - th row of H_{2^n} as R_i

We can see that $S(R_i) = S(r_i r_i)$, $S(R_{i+2^{n-1}}) = S(r_i(-r_i))$ for $1 \leq i \leq 2^{n-1}$

According to the previous subtraction :

$S(R_i) = S(r_i r_i) = 2S(r_i) + \delta_i$ where δ_i is 0 or 1 depending on the meeting point in between the two concatenated rows

Since putting a minus sign for each entry in a row does not change the number of signs

changes, we get that $S(r_i) = S(-r_i)$, and therefore :

$$S(R_{i+2^{n-1}}) = S(r_i(-r_i)) = 2S(r_i) + (1 - \delta_i)$$

Therefore, the rows of H_2 satisfy :

$$\{S(R_1), S(R_{1+2^{n-1}}), S(R_2), S(R_{2+2^{n-1}}), \dots, S(R_{2^{n-1}}), S(R_{2^n})\} = \\ \{2S(r_1), 2S(r_1) + 1, 2S(r_2), 2S(r_2) + 1, \dots, 2S(r_{2^{n-1}}), 2S(r_{2^{n-1}}) + 1\}$$

and by the induction assumption we get :

$$\{S(R_1), S(R_2), \dots, S(R_{2^n})\} = \{0, 1, \dots, 2^n - 1\}$$

Q 4

First we will prove a few properties of Kronecker product

$$A = \begin{bmatrix} a_{1,1} & \cdots & a_{1,n} \\ \cdots & \cdots & \cdots \\ a_{m,1} & \cdots & a_{m,n} \end{bmatrix} B = \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} & \cdots & b_{p,q} \end{bmatrix} C = \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ \cdots & \cdots & \cdots \\ c_{p,1} & \cdots & c_{p,q} \end{bmatrix}$$

First property – $A \otimes (B + C) \stackrel{?}{=} A \otimes B + A \otimes C$

$$A \otimes (B + C)$$

$$A \otimes \begin{bmatrix} b_{1,1} + c_{1,1} & \cdots & b_{1,q} + c_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} + c_{p,1} & \cdots & b_{p,q} + c_{p,q} \end{bmatrix} =$$

$$= \begin{bmatrix} a_{1,1} \cdot \begin{bmatrix} b_{1,1} + c_{1,1} & \cdots & b_{1,q} + c_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} + c_{p,1} & \cdots & b_{p,q} + c_{p,q} \end{bmatrix} & \cdots & a_{1,n} \cdot \begin{bmatrix} b_{1,1} + c_{1,1} & \cdots & b_{1,q} + c_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} + c_{p,1} & \cdots & b_{p,q} + c_{p,q} \end{bmatrix} \\ a_{m,1} \cdot \begin{bmatrix} b_{1,1} + c_{1,1} & \cdots & b_{1,q} + c_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} + c_{p,1} & \cdots & b_{p,q} + c_{p,q} \end{bmatrix} & \cdots & a_{m,n} \cdot \begin{bmatrix} b_{1,1} + c_{1,1} & \cdots & b_{1,q} + c_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} + c_{p,1} & \cdots & b_{p,q} + c_{p,q} \end{bmatrix} \end{bmatrix}$$

$$A \otimes B = \begin{bmatrix} a_{1,1} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} & \cdots & a_{1,n} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ B_{p,1} & & b_{p,q} \end{bmatrix} \\ a_{m,1} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} & & a_{m,n} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} \end{bmatrix}$$

$$A \otimes C = \begin{bmatrix} a_{1,1} \cdot \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} & \cdots & a_{1,n} \cdot \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \\ a_{m,1} \cdot \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} & & a_{m,n} \cdot \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \end{bmatrix}$$

$$A \otimes B + A \otimes C =$$

$$= \begin{bmatrix} a_{1,1} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} & \cdots & a_{1,n} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ B_{p,1} & & b_{p,q} \end{bmatrix} \\ a_{m,1} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} & & a_{m,n} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} \end{bmatrix} +$$

$$\begin{bmatrix} a_{1,1} \cdot \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} & \cdots & a_{1,n} \cdot \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \\ a_{m,1} \cdot \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} & & a_{m,n} \cdot \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \end{bmatrix} =$$

$$\begin{aligned}
& \left[\begin{array}{c} a_{1,1} \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ B_{p,1} & & b_{p,q} \end{bmatrix} + a_{1,1} \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \quad \cdots \quad a_{1,n} \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} + a_{1,n} \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \\ a_{m,1} \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} + a_{m,1} \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \quad \cdots \quad a_{m,n} \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} + a_{m,n} \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \end{array} \right] = \\
& = \left[\begin{array}{c} a_{1,1} \left(\begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} + \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \right) \quad \cdots \quad a_{1,n} \left(\begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} + \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \right) \\ a_{m,1} \left(\begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} + \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \right) \quad \cdots \quad a_{m,n} \left(\begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ b_{p,1} & & b_{p,q} \end{bmatrix} + \begin{bmatrix} c_{1,1} & \cdots & c_{1,q} \\ c_{p,1} & & c_{p,q} \end{bmatrix} \right) \end{array} \right] = \\
& = \left[\begin{array}{c} a_{1,1} \cdot \begin{bmatrix} b_{1,1} + c_{1,1} & \cdots & b_{1,q} + c_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} + c_{p,1} & \cdots & b_{p,q} + c_{p,q} \end{bmatrix} \quad \cdots \quad a_{1,n} \cdot \begin{bmatrix} b_{1,1} + c_{1,1} & \cdots & b_{1,q} + c_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} + c_{p,1} & \cdots & b_{p,q} + c_{p,q} \end{bmatrix} \\ a_{m,1} \cdot \begin{bmatrix} b_{1,1} + c_{1,1} & \cdots & b_{1,q} + c_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} + c_{p,1} & \cdots & b_{p,q} + c_{p,q} \end{bmatrix} \quad \cdots \quad a_{m,n} \cdot \begin{bmatrix} b_{1,1} + c_{1,1} & \cdots & b_{1,q} + c_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} + c_{p,1} & \cdots & b_{p,q} + c_{p,q} \end{bmatrix} \end{array} \right] = \\
& = A \otimes \begin{bmatrix} b_{1,1} + c_{1,1} & \cdots & b_{1,q} + c_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} + c_{p,1} & \cdots & b_{p,q} + c_{p,q} \end{bmatrix} = \\
& = A \otimes (B + C) \rightarrow \boxed{A \otimes (B + C) = A \otimes B + A \otimes C}
\end{aligned}$$

Second property – $(A \otimes B)^T \stackrel{?}{=} A^T \otimes B^T$

$$(A \otimes B)^T =$$

$$= \left[\begin{array}{ccc} a_{1,1} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} & \cdots & b_{p,q} \end{bmatrix} & \cdots & a_{1,n} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} & \cdots & b_{p,q} \end{bmatrix} \\ a_{m,1} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} & \cdots & b_{p,q} \end{bmatrix} & \cdots & a_{m,n} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{1,q} \\ \cdots & \cdots & \cdots \\ b_{p,1} & \cdots & b_{p,q} \end{bmatrix} \end{array} \right]^T =$$

$$= \left(\begin{bmatrix} a_{1,1}B & \cdots & a_{1,n}B \\ \cdots & \cdots & \cdots \\ a_{m,1}B & \cdots & a_{m,n}B \end{bmatrix} \in \mathfrak{R}^{mp \times nq} \right)^T =$$

$$* \\ = \begin{bmatrix} a_{1,1}B^T & \cdots & a_{m,1}B^T \\ \cdots & \cdots & \cdots \\ a_{1,n}B^T & \cdots & a_{m,n}B^T \end{bmatrix} \in \mathfrak{R}^{nq \times mp}$$

$$A^T \otimes B^T = \begin{bmatrix} a_{1,1} & \cdots & a_{m,1} \\ \cdots & \cdots & \cdots \\ a_{1,n} & \cdots & a_{m,n} \end{bmatrix} \otimes \begin{bmatrix} b_{1,1} & \cdots & b_{p,1} \\ \cdots & \cdots & \cdots \\ b_{1,q} & \cdots & b_{p,q} \end{bmatrix} =$$

$$= \left[\begin{array}{ccc} a_{1,1} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{p,1} \\ \cdots & \cdots & \cdots \\ b_{1,q} & \cdots & b_{p,q} \end{bmatrix} & \cdots & a_{m,1} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{p,1} \\ \cdots & \cdots & \cdots \\ b_{1,q} & \cdots & b_{p,q} \end{bmatrix} \\ a_{1,n} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{p,1} \\ \cdots & \cdots & \cdots \\ b_{1,q} & \cdots & b_{p,q} \end{bmatrix} & \cdots & a_{m,n} \cdot \begin{bmatrix} b_{1,1} & \cdots & b_{p,1} \\ \cdots & \cdots & \cdots \\ b_{1,q} & \cdots & b_{p,q} \end{bmatrix} \end{array} \right] =$$

$$= \begin{bmatrix} a_{1,1}B^T & \cdots & a_{m,1}B^T \\ \cdots & \cdots & \cdots \\ a_{1,n}B^T & \cdots & a_{m,n}B^T \end{bmatrix} \in \mathfrak{R}^{nq \times mp} = (A \otimes B)^T \rightarrow \boxed{(A \otimes B)^T = A^T \otimes B^T}$$

$$* (a_{1,1}B)^T = B^T a_{1,1}^T \text{ and since } a_{1,1} \text{ is a scalar} \rightarrow \text{it is equal to its transpose} \rightarrow (a_{1,1}B)^T = a_{1,1}B^T$$

Third property –

Let $A \in \mathbb{R}^{m \times n}$, $B \in \mathbb{R}^{r \times s}$, $C \in \mathbb{R}^{n \times p}$ and $D \in \mathbb{R}^{s \times t}$

$$(A \otimes B)(C \otimes D) \stackrel{?}{=} AC \otimes BD$$

$$(A \otimes B)(C \otimes D) =$$

$$= \begin{bmatrix} a_{1,1}B & \cdots & a_{1,n}B \\ \cdots & \cdots & \cdots \\ a_{m,1}B & \cdots & a_{m,n}B \end{bmatrix} \begin{bmatrix} c_{1,1}D & \cdots & c_{1,p}D \\ \cdots & \cdots & \cdots \\ c_{n,1}D & \cdots & c_{n,p}D \end{bmatrix} =$$

$$= \begin{bmatrix} \sum_{k=1}^n a_{1,k}c_{k,1}BD & \cdots & \sum_{k=1}^n a_{1,k}c_{k,p}BD \\ \cdots & \cdots & \cdots \\ \sum_{k=1}^n a_{m,k}c_{k,1}BD & \cdots & \sum_{k=1}^n a_{m,k}c_{k,p}BD \end{bmatrix} = AC \otimes BD \rightarrow \boxed{(A \otimes B)(C \otimes D) = AC \otimes BD}$$

a.

$$H_{2^{n+1}} = \begin{bmatrix} H_{2^n} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix} \\ I_{2^n} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix} \end{bmatrix}$$

$$H_2 = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$H_{2^n}^T = \left[\begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix} \end{bmatrix} \right]^T =$$

$$= \left[\left(H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix} \right)^T \quad \left(I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix} \right)^T \right] = \left[\left(H_{2^{n-1}}^T \otimes \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right) \quad \left(I_{2^{n-1}}^T \otimes \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right) \right] =$$

by property 2 $\boxed{(A \otimes B)^T = A^T \otimes B^T}$

by induction assumption $H_{2^{n-1}}^T = H_{2^{n-1}}$

$$= \left[\left(H_{2^{n-1}} \otimes \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right) \left(I_{2^{n-1}} \otimes \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right) \right] \neq \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix} \end{bmatrix}$$

$\rightarrow H_{2^n}$ is NOT symmetric

Of note, in the following we will prove that H_{2^n} is orthogonal. Another option to prove that H_{2^n} is not symmetric orthogonal matrix is by showing that $H_{2^n} \cdot H_{2^n} \neq I_{2^n \times 2^n}$

b.

Let us use induction :

$$\text{Base } H_{2^1} \text{ can be written as } H_{2^1} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

step : we shall assume $H_{2^{n-1}}$ is orthogonal

$$\text{By definition : } H_{2^n} = \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix} \end{bmatrix}$$

$$H_{2^n}^T H_{2^n} = \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix} \end{bmatrix}^T \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix} \end{bmatrix} =$$

$$= \begin{bmatrix} (H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix})^T & (I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix})^T \end{bmatrix} \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix} \end{bmatrix} =$$

$$\text{by property 2 } \boxed{(A \otimes B)^T = A^T \otimes B^T}$$

$$= \begin{bmatrix} \left(H_{2^{n-1}}^T \otimes \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right) & \left(I_{2^{n-1}}^T \otimes \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right) \end{bmatrix} \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix} \end{bmatrix} =$$

$$= \begin{bmatrix} \left(H_{2^{n-1}}^T \otimes \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right) (H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \end{bmatrix}) & \left(I_{2^{n-1}}^T \otimes \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right) (I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix}) \end{bmatrix} =$$

$$\text{by property 3 } \boxed{(A \otimes B)(C \otimes D) = AC \otimes BD}$$

$$= \left[\left((H_{2^{n-1}}^T H_{2^{n-1}}) \otimes \left(\begin{bmatrix} 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \end{bmatrix} \right) \right) \left((I_{2^{n-1}}^T I_{2^{n-1}}) \otimes \left(\begin{bmatrix} 1 \\ -1 \end{bmatrix} \otimes \begin{bmatrix} 1 & -1 \end{bmatrix} \right) \right) \right] =$$

by induction assumption $H_{2^{n-1}}^T H_{2^{n-1}} = I_{2^{n-1} \times 2^{n-1}}$

$$= \left[\left(I_{2^{n-1} \times 2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \right) \left(I_{2^{n-1} \times 2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix} \right) \right] =$$

by first property $A \otimes B + A \otimes C = A \otimes (B + C)$

$$= \left[I_{2^{n-1} \times 2^{n-1}} \otimes \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \right] = 2 I_{2^n \times 2^n} \rightarrow H_{2^n} \text{ is orthogonal}$$

$$\text{since } H_{2^n}^T H_{2^n} = \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix} \end{bmatrix}^T \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix} \end{bmatrix} = 2 \cdot I_{2^n \times 2^n} \rightarrow H_{2^n} \text{ is not unitary}$$

c.

we will prove by induction that $H_{2^n} = \frac{1}{\sqrt{2}} \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix} \end{bmatrix}$ is a unitary matrix

$$\text{Base } \widetilde{H}_{2^1} \text{ can be written as } H_{2^1} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

\widetilde{H}_{2^1} is a unitary matrix

step: we shall assume $H_{2^{n-1}}$ is unitary

$$H_{2^n}^T H_{2^n} = \frac{1}{2} \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix} \end{bmatrix}^T \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix} \end{bmatrix} =$$

by the proof in "e"

$$= \frac{1}{2} \left[I_{2^{n-1} \times 2^{n-1}} \otimes \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \right] = I_{2^n \times 2^n} \rightarrow H_{2^n} \text{ is unitary and the recursion}$$

$$\text{formula is } H_{2^n} = \frac{1}{\sqrt{2}} \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \end{bmatrix} \text{ with base } H_{2^1} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

d. See approval of property 2 above

e.

$$H_{2^n}^T = \frac{1}{\sqrt{2}} \begin{bmatrix} H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \\ I_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \end{bmatrix}^T = \frac{1}{\sqrt{2}} \left[\left(H_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \right)^T \quad \left(I_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \right)^T \right] =$$

$$\text{by property 2 } \boxed{(A \otimes B)^T = A^T \otimes B^T}$$

$$= \frac{1}{\sqrt{2}} \left[\left(H_{2^{n-1}}^T \otimes \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \right) \left| \left(I_{2^{n-1}} \otimes \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \right) \right] \right]$$