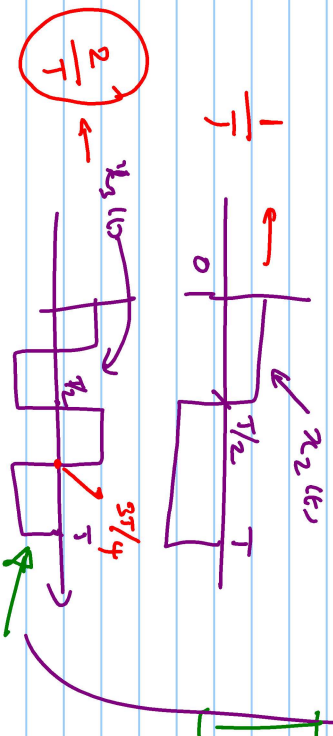
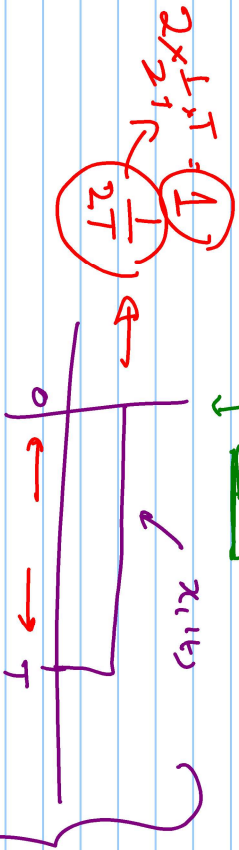
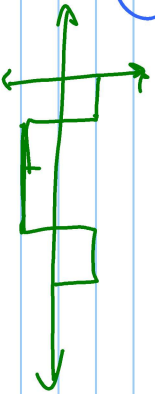


Lesson 1a -- Finding the basis signal set using Gram-Schmidt orthogonalisation

Note title

8/8/2017

Given M signals $\{s_1(t), s_2(t), \dots, s_M(t)\}$ find N basis functions $\{\phi_1(t), \phi_2(t), \dots, \phi_N(t)\}$



$$N \times 1 = \text{l.c.} \{i, j, k\}$$

3 vectors

$$N = 2WT$$

$N \rightarrow$ Bandwidth

$$2 \times \frac{4}{T} \rightarrow 8 \text{ fns.}$$

$$\int \phi_j^2 dt = 1 \quad \forall j=1, \dots, N$$

$$\int \phi_i(t) \phi_k(t) dt = \int_0^1 \delta_{ij} dt \quad [0, j \neq k]$$

Walsh- Hadamard sequences.

OVSF

Lehrbuchen-Lösung

Gram-Schmidt Procedure

$$\{s_1(t), \dots, s_n(t)\}$$

1. $\phi_1(t) = \frac{s_1(t)}{\sqrt{E_1}}$, $E_1 = \int_0^T s_1^2(t) dt$

← projection

2. $\tilde{\phi}_2(t) = s_2(t) - s_{21} \phi_1(t)$

where $s_{21} = \int_0^T s_2(t) \phi_1(t) dt$

3. $\phi_2(t) = \frac{\tilde{\phi}_2(t)}{\sqrt{E_2}}$, $E_2 = \int_0^T \tilde{\phi}_2^2(t) dt$

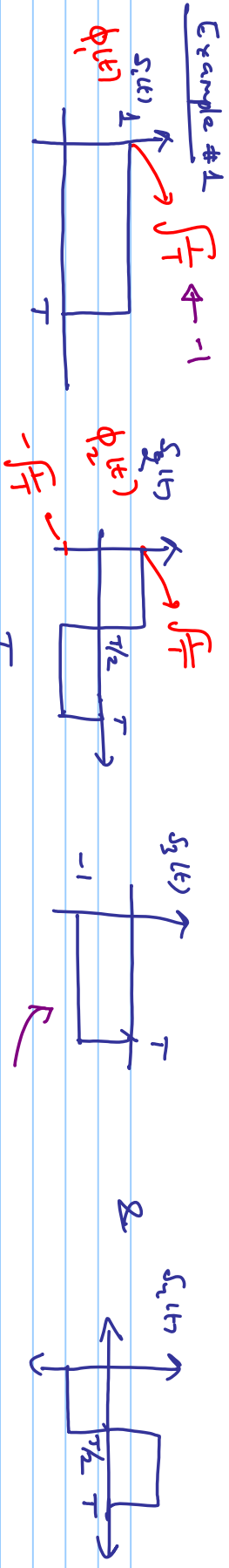
4. $\tilde{\phi}_3(t) = s_3(t) - s_{31} \phi_1(t) - s_{32} \phi_2(t)$

5. $\phi_3(t) = \tilde{\phi}_3(t) / \sqrt{E_3}$, $E_3 = \int_0^T \tilde{\phi}_3^2(t) dt$

$$s_{31} = \int_0^T s_3(t) \phi_1(t) dt$$

$$s_{32} = \int_0^T s_3(t) \phi_2(t) dt$$

Example #1



$$1. \phi_1(t) = \frac{s_1(t)}{\sqrt{E_1}} \quad E_1 = \int_0^T s_1^2(t) dt = T$$

$$2. \tilde{\phi}_2(t) = s_2(t) - s_{21} \phi_1(t), \quad s_{21} = \int_0^T s_2(t) \phi_1(t) dt = 0$$

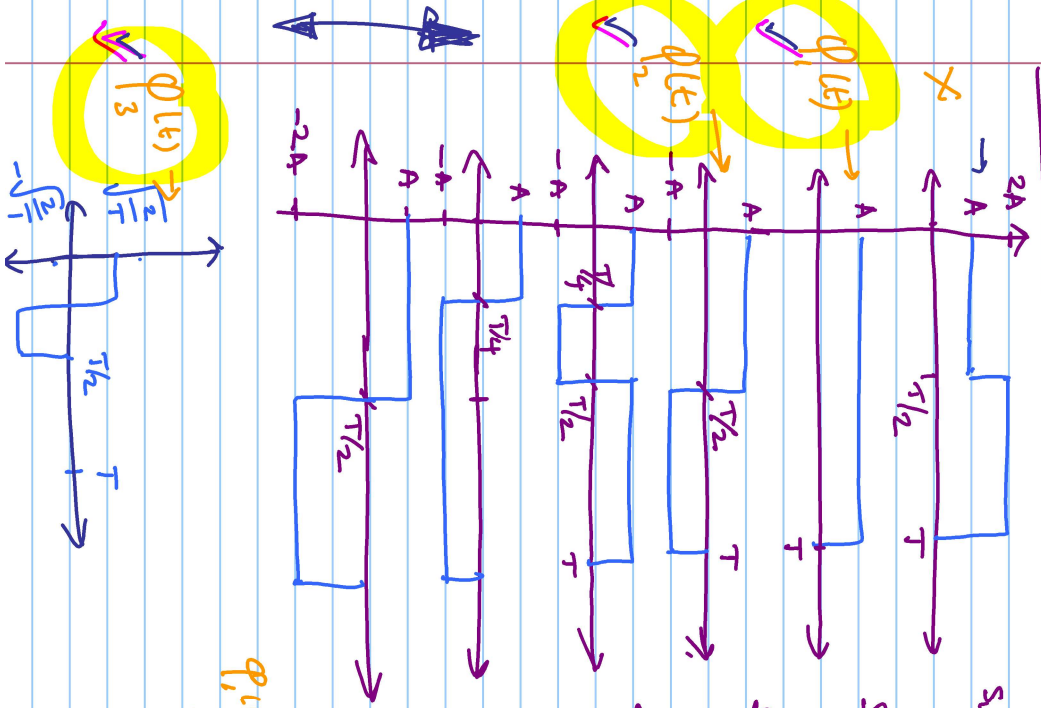
$$\Rightarrow \tilde{\phi}_2(t) = s_2(t)$$

$$3. \phi_2(t) = \frac{\tilde{\phi}_2(t)}{\sqrt{E_2}} = \frac{s_2(t)}{\sqrt{T}} \quad s_{31} = \int_0^T s_3(t) \phi_1(t) dt = -\sqrt{T} \quad ;$$

$$4. \tilde{\phi}_3(t) = s_3(t) - (s_{31} \phi_1(t)) - s_{32} \phi_2(t) \quad s_{32} = 0 \quad ;$$

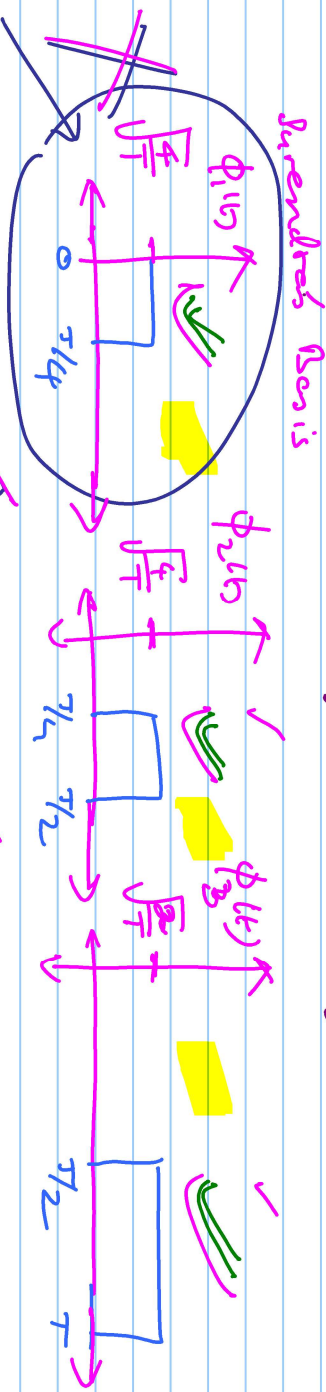
$$= 0 \quad ; \quad \text{|||} \quad \tilde{\phi}_4(t) = 0$$

Example #2 :

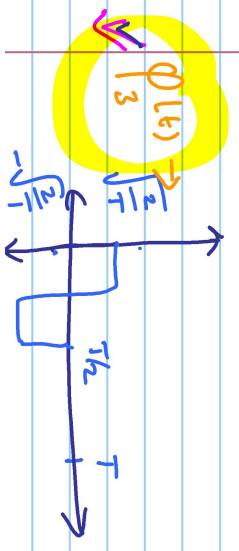


- (a) Find the basis set for this $\{s_i(t)\}$ $c = 1, \dots, 6$
- (b) Plot the signal constellation $N = 2WT$

(c) Develop the "most efficient" matched filter receiver for this signal set.



$s_1(t) = \sqrt{E_2}$
 $s_3(t) \rightarrow \phi_2(t)$
 $s_2(t) = \sqrt{E_2}$
 $s_4(t) = \sqrt{E_2}$
 $s_1(t) = \sqrt{E_2}$
 $s_2(t) = \sqrt{E_2}$
 $s_3(t) = \sqrt{E_2}$
 $s_4(t) = \sqrt{E_2}$

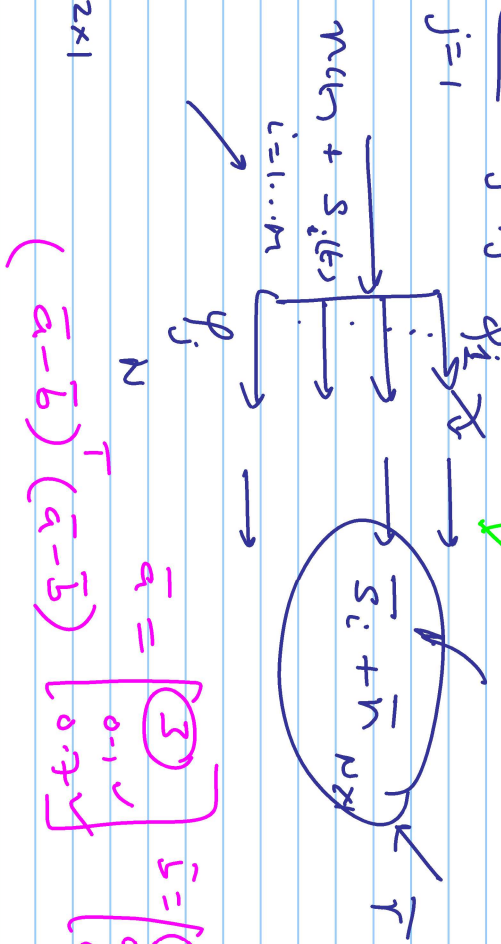
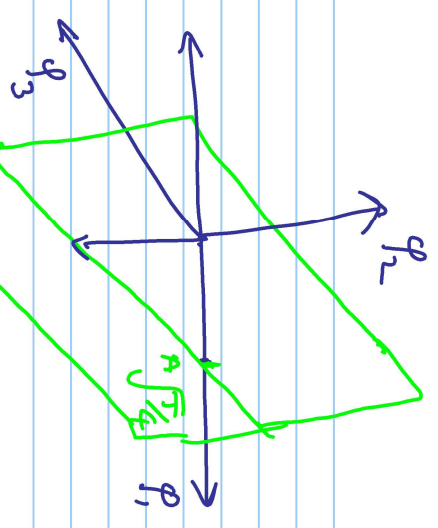
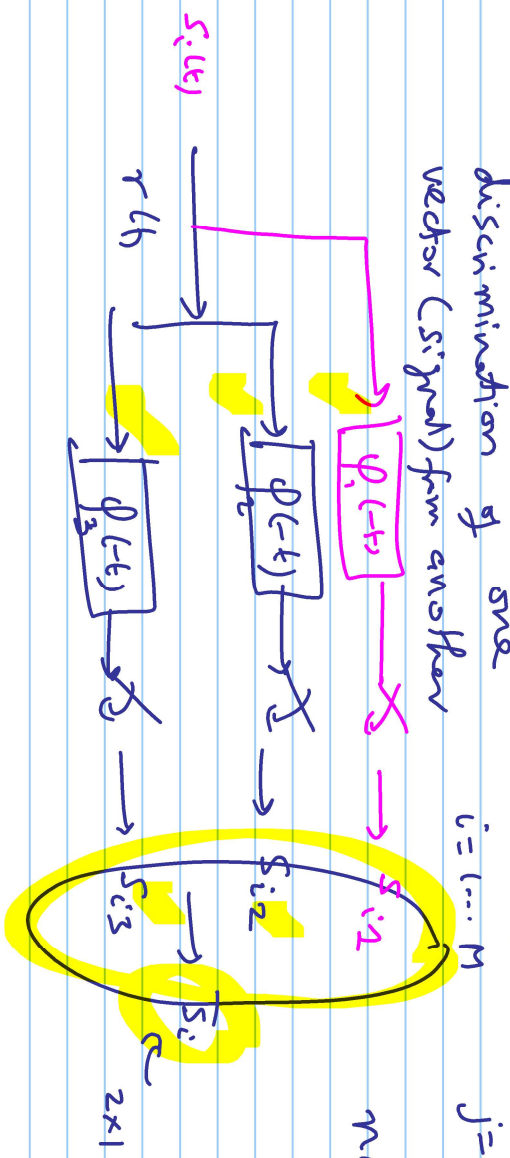


$$\bar{s}_3 = \begin{bmatrix} A\sqrt{\frac{T}{2}} \\ A\sqrt{\frac{T}{4}} \\ -A\sqrt{\frac{T}{2}} \end{bmatrix}; \quad \bar{s}_5 = \begin{bmatrix} A\sqrt{\frac{T}{2}} \\ -A\sqrt{\frac{T}{4}} \\ -A\sqrt{\frac{T}{2}} \end{bmatrix}; \quad \bar{s}_6 = \begin{bmatrix} A\sqrt{\frac{T}{4}} \\ A\sqrt{\frac{T}{4}} \\ -2A\sqrt{\frac{T}{2}} \end{bmatrix};$$

⇒ Since s_{i1} is common in all \bar{s}_i

⇒ It does not add to the discrimination of one vector (signal) from another

$$s_i(t) = \sum_{j=1}^N s_{ij} \phi_j(t)$$



$\bar{a} = \begin{bmatrix} 3 \\ 0.1 \\ 0.2 \end{bmatrix}$
 $\bar{b} = \begin{bmatrix} 3 \\ 0.1 \\ 0.1 \end{bmatrix}$
 $(\bar{a}-\bar{b})^T (\bar{a}-\bar{b})$

$$\bar{x} = \frac{1}{N} \sum_{i=1}^N x_i$$

Minimum Distance Rule

Pick $\hat{s} = s_i$

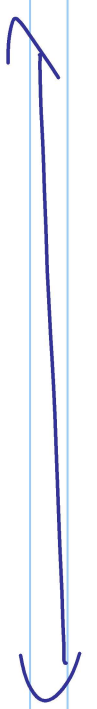
if $(\bar{x} - \bar{s}_i)^T (\bar{x} - \bar{s}_i) \leq (\bar{x} - \bar{s}_l)^T (\bar{x} - \bar{s}_l)$

$\forall l \neq i$
 $l = 1, \dots, N$

$$\|s_{cell}\| = \int_0^T |s_i(t)| dt$$

$$s_i(t)$$

$$i=1, \dots, N$$



$$\bar{s}_i$$

Triangular Inequality

$$\|s_1(t) + s_2(t)\| \leq \|s_1(t)\| + \|s_2(t)\|$$

$$\|\bar{s}_1 + \bar{s}_2\| \leq \|\bar{s}_1\| + \|\bar{s}_2\|$$

$$\int_a^b |x_1(t) + x_2(t)| dt \leq \int_a^b |x_1(t)| dt + \int_a^b |x_2(t)| dt$$

Cauchy-Schwarz Inequality

$$\|v_1 + v_2\| \leq \|v_1\| + \|v_2\|$$