

Exercise 1

① The overall system is LTI, thus if we delay the input, we also delay the output.

Since $h(t) = \delta(t - T)$, $x(t)$ is delayed by T , hence after MF, if we sample at

$t = kT + T$, the samples y_k will be ISI-free.

$$\textcircled{2} R_p(t) = \int p(\alpha+t) p(\alpha)^* d\alpha = \langle p(\alpha+t), p(\alpha) \rangle = \langle p(\alpha), p(\alpha+t) \rangle^*$$

$$R_p^*(t) = \langle p(\alpha), p(\alpha+t) \rangle \Rightarrow R_p^*(-t) = \int p(\alpha) p(\alpha-t)^* d\alpha$$

$$\text{Let } \alpha' = \alpha' + t \Rightarrow R_p^*(t) = \int p(\alpha'+t) p(\alpha')^* d\alpha' = R_p(t) //$$

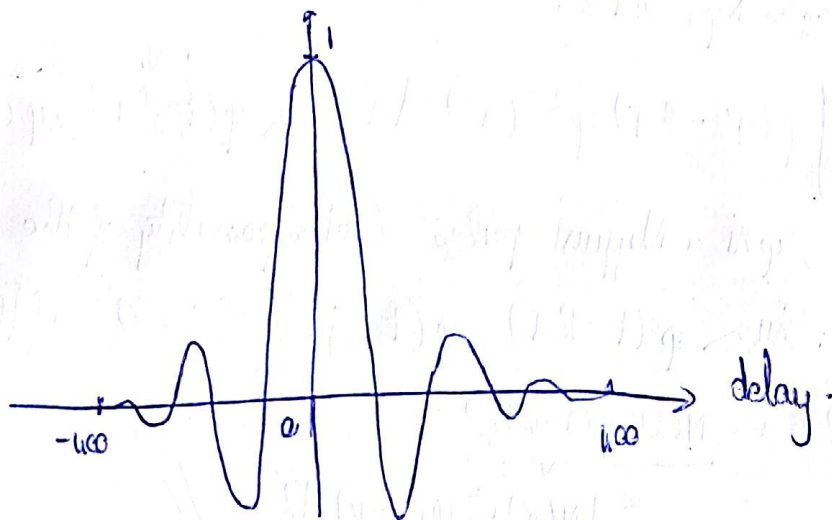
b) First note that $R_p(0) = \int p(\alpha) p^*(\alpha) d\alpha = \langle p(\alpha), p(\alpha) \rangle = \|p\|^2$.

By Cauchy-Schwarz ineq: $|R_p(t)| = |\langle p(\alpha+t), p(\alpha) \rangle| \leq \frac{\|p(\alpha+t)\|}{\|p\|} \cdot \frac{\|p(\alpha)\|}{\|p\|}$

$$\Leftrightarrow |R_p(t)| \leq \|p\|^2 //$$

Self similarity funct of RLC pulse.

c)



Exercise 2:

② From eq. [7]: $p(v) = \langle x(t), p(t-v) \rangle = R_p(v-\tau) + \langle s_{\text{data}}(t-\tau), p(t-v) \rangle + \langle n(t), p(t-v) \rangle$

a) we choose $p(t) = s_0 \varphi(t)$, thus:

$$R_p(v-\tau) = \int s_0 \varphi(\alpha + v - \tau) \varphi^*(\alpha) s_0^* d\alpha = |s_0|^2 \int \varphi(\alpha + v - \tau) \varphi^*(\alpha) d\alpha = |s_0|^2 R_p(v-\tau)$$

$$\langle s_{\text{data}}(t-\tau), p(t-v) \rangle = \int s_{\text{data}}(\alpha - \tau) p^*(\alpha - v) d\alpha = \sum_{k \geq 1} s_k \int \varphi(\alpha - \tau - kT) \varphi^*(\alpha - v) d\alpha$$

$$\langle s_{\text{data}}(t-\tau), p(t-v) \rangle = \sum_{k \geq 1} s_k s_0^* \int \varphi(\alpha - \tau - kT) \varphi^*(\alpha - v) d\alpha$$

$$\text{let } \alpha' = \alpha - v \Leftrightarrow \alpha = \alpha' + v \Leftrightarrow d\alpha = d\alpha'$$

$$\langle s_{\text{data}}(t-\tau), p(t-v) \rangle = \sum_{k \geq 1} s_k s_0^* \int \varphi(\alpha' + v - \tau - kT) \varphi^*(\alpha') d\alpha' = \sum_{k \geq 1} s_k s_0^* R_p(v - \tau - kT)$$

$$\text{since } k = 1 \Rightarrow k > 0 \quad (k \geq 1) \quad //$$

b) let's evaluate $\langle s_{\text{data}}(t-\tau), p(t-v) \rangle$ when $v = \tau$:

$$\langle s_{\text{data}}(t-\tau), p(t-\tau) \rangle = \sum_{k \geq 0} s_k s_0^* R_p(-kT) \quad \underline{k \geq 0!}$$

$$\text{Note that: } R_p(-kT) = \int \varphi(\alpha - kT) \varphi^*(\alpha) d\alpha = \langle \varphi(t - kT), \varphi(t) \rangle = 0$$

$R_p(-kT) = 0$ because φ is a Nyquist pulse (orthogonality of the shifted pulses):

$$\langle \varphi(t - kT), \varphi(t - jT) \rangle = 0 \quad \text{if } \begin{matrix} k, j \in \mathbb{Z} \\ k \neq j \end{matrix}$$

$$\text{Thus: } p(v) = |s_0|^2 R_p(v-\tau) + \underbrace{\langle n(t), p(t-v) \rangle}_{= \int N(\alpha) s_0^* \varphi(\alpha - v) d\alpha} //$$

c) It doesn't work, some sent symbols are lost. Indeed the estimated channel delay is way larger than the actual channel delay. This comes from the fact that $v = \tau$ is only a local maximum of $p(v)$, contribution of other data symbol for $v \gg \tau$ is non-negligible.

$$\textcircled{3} \quad p(t) = \sum_{i=0}^{l-1} B_i \varphi(t - iT).$$

$$1) \quad R_p(t) = \int p(\alpha+t) p(\alpha)^* d\alpha = \int \sum_{i=0}^{l-1} B_i \varphi(t - iT + \alpha) \cdot \sum_{j=0}^{l-1} B_j^* \varphi^*(\alpha - iT) d\alpha.$$

$$R_p(t) = \sum_{i=0}^{l-1} \sum_{j=0}^{l-1} B_i B_j^* \int \varphi(t - iT + \alpha) \varphi(\alpha - iT) d\alpha. \quad \text{Let } \alpha = \alpha' + iT \Rightarrow d\alpha = d\alpha'.$$

$$R_p(t) = \sum_{i=0}^{l-1} \sum_{j=0}^{l-1} B_i B_j^* \times R_\varphi(t).$$

$$\Rightarrow E[B_i B_j^*] = \begin{cases} 1 & i=j \text{ (since } E[X_{ij}] = \frac{1}{2} \times 1 + \frac{1}{2} \times (-1)^2 = 1) \\ 0 & i \neq j \text{ (since } i \neq j: E[X_{ij}] = i \times 1/4 + (i-1) \times 1/4 + (i+1) \times 1/4 + (i-1) \times 1/4 = 0) \end{cases}$$

$$\text{Thus: } E[R_p(t)] = \sum_{i=0}^{l-1} \sum_{j=0}^{l-1} E[B_i B_j^*] R_\varphi(t) = l \times R_\varphi(t).$$

φ has a sharp self-similarity function means that $R_\varphi(t)$ is large around 0 and negligible when we deviate from 0. Thus from [2], $R_p(t)$ is on average large around 0 and negligible for value of t not around 0.

$$2) \quad \langle q(t), p(t-v) \rangle = \int q(\alpha) p^*(\alpha - v) d\alpha = \int q(\alpha) \cdot \sum_{i=0}^{l-1} B_i^* \varphi^*(\alpha - v - iT) d\alpha.$$

$$\langle q(t), p(t-v) \rangle = \underbrace{\sum_{i=0}^{l-1} B_i^*}_X \cdot \underbrace{\int q(\alpha) \varphi^*(\alpha - v - iT) d\alpha}_Y.$$

since random processes $p(t)$ and $q(t)$ are independent, so are X and Y .

$$\text{Hence } E[\langle q(t), p(t-v) \rangle] = E[X \cdot Y] = \underbrace{E[X] \cdot E[Y]}_{X \text{ and } Y \text{ are indep.}}$$

$$E[X] = E\left[\sum_{i=0}^{l-1} B_i^*\right] = \sum_{i=0}^{l-1} E[B_i] = 0$$

linearity of expectation
and B_i are real valued

$$\text{Thus } E[\langle q(t), p(t-v) \rangle] = 0.$$

Exercise 3: (We assume b_i 's are real valued)

$$\textcircled{1} y(t) = x(t) * \varphi^*(-t) = \int x(\alpha) \varphi^*(\alpha-t) d\alpha = \int p(\alpha-t) \varphi^*(\alpha-t) d\alpha + \int q(\alpha) \varphi^*(\alpha-t) d\alpha$$

$$\int q(\alpha) \varphi^*(\alpha-t) d\alpha = \int q(\alpha+t) \varphi^*(\alpha') d\alpha' = R_{q,p}(t) \quad (\text{take } \alpha = \alpha' + t \Rightarrow d\alpha = d\alpha')$$

$$\int p(\alpha-t) \varphi^*(\alpha-t) d\alpha = \int \sum_{i=0}^{L-1} b_i \varphi(\alpha-t-iT) \varphi^*(\alpha-t) d\alpha = \sum_{i=0}^{L-1} b_i \int \varphi(\alpha'+t-iT) \varphi^*(\alpha') d\alpha'$$

let $\alpha = \alpha' + t$

$$\int p(\alpha-t) \varphi^*(\alpha-t) d\alpha = \sum_{i=0}^{L-1} b_i R_{q,p}(t-iT)$$

$$\Rightarrow y(t) = \sum_{i=0}^{L-1} b_i R_{q,p}(t-iT) + R_{q,p}(t)$$

$$\textcircled{2} p(v) = \langle x(t), p(t-v) \rangle = \langle p(t-iT), p(t-v) \rangle + \langle q(t), p(t-v) \rangle$$

$$\begin{aligned} \langle q(t), p(t-v) \rangle &= \int q(\alpha) \tilde{p}(\alpha-v) d\alpha = \int q(\alpha) \sum_{i=0}^{L-1} b_i \tilde{p}(\alpha-v-iT) d\alpha \\ &= \sum_{i=0}^{L-1} b_i \int q(\alpha) \varphi^*(\alpha-v-iT) d\alpha = \sum_{i=0}^{L-1} b_i \int q(\alpha'+v+iT) \varphi^*(\alpha') d\alpha' \end{aligned}$$

$$\langle q(t), p(t-v) \rangle = \sum_{i=0}^{L-1} b_i R_{q,p}(v+iT)$$

$$\langle p(t-iT), p(t-v) \rangle = \int p(\alpha-iT) p^*(\alpha-v) d\alpha = \int p(\alpha'+v-iT) p^*(\alpha') d\alpha'$$

$$= R_p(v-iT) = \sum_{i=0}^{L-1} \sum_{j=0}^{L-1} b_i b_j^* R_{q,p}(v-iT)$$

$$\Rightarrow p(v) = \sum_{i=0}^{L-1} \sum_{j=0}^{L-1} b_i b_j^* R_{q,p}(v-iT) + \sum_{i=0}^{L-1} b_i R_{q,p}(v+iT)$$

$$p(mT_s) = \sum_{i=0}^{L-1} \sum_{j=0}^{L-1} b_i b_j^* R_{q,p}(mT_s-iT) + \sum_{i=0}^{L-1} b_i R_{q,p}(mT_s+iT)$$

Let's calculate $\langle y[m+k], \hat{p}_k \rangle = \sum_k y[m+k] \cdot \hat{p}_k = \sum_k \sqrt{T_s} y((m+k)T_s) \cdot \hat{p}_k$

$$\begin{aligned} \langle y[m+k], \hat{p}_k \rangle &= \sum_k \sum_{i=0}^{L-1} b_i \hat{p}_k \sqrt{T_s} \cdot R_{q,p}((m+k)T_s-iT) + \sum_k \hat{p}_k \sqrt{T_s} \cdot R_{q,p}((m+k)T_s) \\ &= \sum_k \sum_{i=0}^{L-1} b_i \hat{p}_k \sqrt{T_s} \cdot R_{q,p}((m+k)\frac{T}{J}-iT-iT) + \sum_k \hat{p}_k \sqrt{T_s} \cdot R_{q,p}((m+k)\frac{T}{J}) \\ &= \sum_{i=0}^{L-1} \sum_{j=0}^{L-1} b_i b_j^* \sqrt{T_s} R_{q,p}(mT_s-iT+iT-iT) + \sum_{i=0}^{L-1} b_i \sqrt{T_s} R_{q,p}(mT_s+iT) \end{aligned}$$

$$\langle y[m+k], \hat{p}_k \rangle = \left(\sum_{i=0}^{k-1} \sum_{j=0}^{k-1} p_{ij} R_{ij}(mT_s - \tau) + \sum_{i=0}^{k-1} p_i R_{q,i}(mT_s + iT) \right) \times \sqrt{T_s}.$$

$$\Rightarrow \langle y[m+k], \hat{p}_k \rangle = \sqrt{T_s} \times \rho(mT_s).$$