

- Lecture 19 Recap:

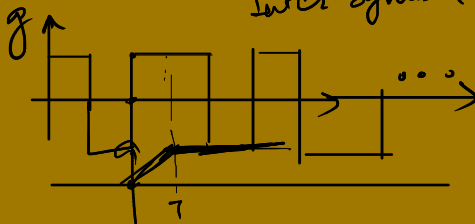
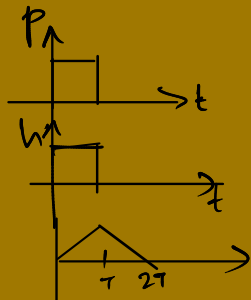
- ▶ Eye diagram.

- Channel distortion.

- ▶ Amplitude and Phase distortion may lead to ISI.

Delay distortion

Inter Symbol Interference

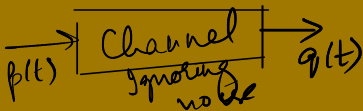
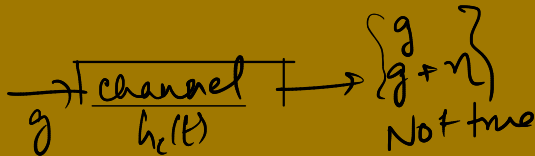


Modeling ISI

- Let us denote the transmitter pulse waveform by $p(t)$, the positive and negative impulse that encodes the m^{th} bit by a_m and the transmitted signal by

$$g(t) = \sum_m a_m p(t - mT)$$

T - duration of a symbol.
 $+V$ or $-V$
 bit 1 bit 0



$$r = h_c * g + n$$

$$\frac{r}{h_c * (g + n)}$$

Modeling ISI

- Let us denote the transmitter pulse waveform by p , the positive and negative impulse that encodes the m^{th} bit by a_m , and the transmitted signal by $g(t) = \sum_m a_m p(t - mT)$.
- Since the channel is assumed to be an LTI system, if for an input of $p(t)$ to the channel, the output is $q(t)$, the received signal can be written as

$$r(t) = \sum_m a_m q(t - mT) + \eta(t),$$

where η is AWGN.

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Lab 5:
$$r(t) = \sum_m a_m p(t - mT) + \eta(t)$$

- How should we arrive at a decision on what bits were transmitted?

Channel is deterministic, p is in your control
(h_c is known)
o. q is known!

MLE Estimation

- Assume that in some orthonormal basis $\{f_n, n \in \mathbb{Z}^+\}$, $\underbrace{r}_{\text{r(t)}} = \sum_n \underbrace{r_n}_{\text{scalar} \in \mathbb{R}} f_n$.

$$h(t) = \sum_n h_n f_n(t)$$

MLE Estimation

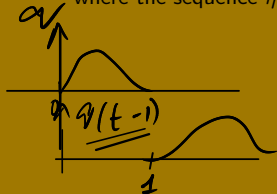
- Assume that in some orthonormal basis $\{f_n, n \in \mathbb{Z}^+\}$, $r = \sum_n r_n f_n$.
- Then, denoting $\langle q(t - mT), f_k \rangle$ by q_{km} and $\langle \eta(t), f_k \rangle$ by η_k , we get

$$q(t - mT) = \sum_k q_{km} f_k$$

$$r_n = \sum_k a_k q_{nk} + \eta_n$$

$$\eta_1, \eta_2, \dots, \eta_n, \dots$$

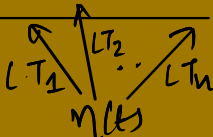
where the sequence η_n has a white Gaussian distribution.



$$E[\eta_k \eta_m]$$

$$= E[\langle \eta, f_k \rangle \langle \eta, f_m \rangle]$$

$$= 0 \text{ if } k \neq m$$



MLE Estimation

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- Then, denoting $\langle q(t - mT), f_k \rangle$ by q_{km} , and $\langle \eta(t), f_k \rangle$ by η_k , we get

$$r_n = \sum_k a_k q_{nk} + \eta_n,$$

independent / uncorrelated

where the sequence η_n has a ~~white~~ Gaussian distribution.

- Let us assume that L bits were transmitted. We need to find $\{a_0, \dots, a_{L-1}\}$, from N coefficients $\{r_0, \dots, r_{N-1}\}$ of the received signal.

$$\eta_n \triangleq r_n - \sum_k a_k q_{nk} \sim \mathcal{N}(0, N_0)$$

MLE Estimation

$$r = r_1 f_1 + r_2 f_2$$

$$\tilde{r} = r_1 f_1$$

- Assume that in some orthonormal basis $\{f_n, n \in \mathbb{Z}^+\}$, $r = \sum_n r_n f_n$
- Then, denoting $\langle q(t - mT), f_k \rangle$ by q_{km} , and $\langle \eta(t), f_k \rangle$ by η_k , we get

$$r_n = \sum_k a_k q_{nk} + \eta_n,$$

where the sequence η_n has a ~~white~~ Gaussian distribution.

η_1, \dots, η_N - are independent

- Let us assume that L bits were transmitted. We need to find $\{a_0, \dots, a_{L-1}\}$, from N coefficients $\{r_0, \dots, r_{N-1}\}$ of the received signal.

- Since $\eta_n = r_n - \sum_k a_k q_{nk} \sim \mathcal{N}(0, N_0)$, the likelihood is

$\eta_1, \eta_2, \eta_3, \dots$

$$p(\eta_1, \dots, \eta_N) = p(\underbrace{r_0, \dots, r_{N-1}}_{\text{received signal}}, \underbrace{\{a_0, \dots, a_{L-1}\}}_{\text{transmitted signal}}) = K \exp \left(-\frac{1}{2N_0} \sum_{n=0}^{N-1} \left| r_n - \sum_{k=0}^{L-1} a_k q_{nk} \right|^2 \right) \leftarrow$$

$$\equiv p(\eta_1) \cdot p(\eta_2) \cdot \dots \cdot p(\eta_N)$$

and the task boils down to

$$\min_{\{a_0, \dots, a_{L-1}\}} \sum_{n=0}^{N-1} \left| r_n - \sum_{k=0}^{L-1} a_k q_{nk} \right|^2$$

Let $N \rightarrow \infty$

- Also as the number of coefficients used $N \rightarrow \infty$, we get

$$\lim_{N \rightarrow \infty} \left| \sum_{n=0}^{N-1} r_n \sum_{k=0}^{L-1} a_k q_{nk} \right|^2 = J := \int_{-\infty}^{\infty} \left(r(t) - \sum_k a_k q(t - kT) \right)^2 dt$$

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► Then,

$$\begin{aligned} J = & \int_{-\infty}^{\infty} \underline{r^2(t)} dt - 2 \sum_k a_k \int_{-\infty}^{\infty} r(t) q(t - kT) dt \\ & + \sum_k \sum_l \underline{a_k a_l} \int_{-\infty}^{\infty} \underline{q(t - kT)} \underline{q(t - lT)} dt \end{aligned}$$

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$$z_k = z(kT)$$

► Let $\underline{z_k} = \int_{-\infty}^{\infty} r(t) q(t - kT) dt$ and $\underline{x_k} = \int_{-\infty}^{\infty} q(t) q(t + kT) dt$.

$$z_k = z(kT) = \int_{-\infty}^{\infty} h(t) q(t - kT) dt$$

Sampled MF response matched to $q(t)$

$$R_q(\tau) = \int_{-\infty}^{\infty} q(t) q(t + \tau) dt$$

$$x_k = R_q(\tau = kT)$$

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- Let $z_k = \int_{-\infty}^{\infty} \textcircled{r(t)} q(t - kT) dt$ and $x_k = \int_{-\infty}^{\infty} q(t) q(t + kT) dt$.
- Note that z_k can be obtained by sampling the output of the matched filter matched to q every T secs, and x is the sampled autocorrelation function of q .

$$\begin{aligned} h(t) &= \sum_m a_m q(t - mT) + n(t) \\ z_k &= ? \end{aligned}$$

- Also as the number of coefficients used $N \rightarrow \infty$, we get

$$\lim_{N \rightarrow \infty} \sum_{n=0}^{N-1} \left| r_n - \sum_{k=0}^{L-1} a_k q_{nk} \right|^2 = J := \int_{-\infty}^{\infty} \left(r(t) - \sum_k a_k q(t - kT) \right)^2 dt$$

- Then,

$$\begin{aligned} J &= \int_{-\infty}^{\infty} r^2(t) dt - 2 \sum_k a_k \int_{-\infty}^{\infty} r(t) q(t - kT) dt \\ &\quad + \sum_k \sum_l a_k a_l \int_{-\infty}^{\infty} q(t - kT) q(t - lT) dt \end{aligned}$$

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- Note that z_k can be obtained by sampling the output of the matched filter matched to q every T secs, and x is the sampled autocorrelation function of q .
- Since $r(t) = \sum_m a_m q(t - mT) + \eta(t)$,
 $z_k = \sum_m a_m \int_{-\infty}^{\infty} q(t - mT) q(t - kT) dt + \int_{-\infty}^{\infty} \eta(t) q(t - kT) dt$.

- $z_k = \sum_m a_m x_{k-m} + v_k$ ← Sampled output of Matched filter.

lab 5: $q = p \Rightarrow x(0) \neq 0, x(1) = x(-1) = x(2) = x(-2) \dots$
 $z_k = \boxed{a_k} + v_k$

If $x_k \neq \delta_k \Rightarrow ISI!!$

- If q is such that $x_{k-m} = \int_{-\infty}^{\infty} q(t-mT)q(t-kT) = 0$ for $|k-m| > L$,
$$z_k = \sum_{m=-L}^L a_m x_{k-m} + v_k.$$