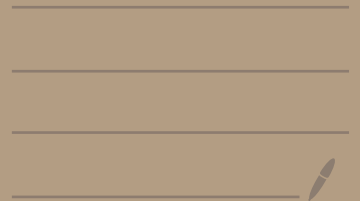


Principles of Digital Communication

March 26, 2021



Reading Assignment: 3. [5, 6, 7, 9].

where are we in the course

• Hypothesis testing

$$H \in \{0, \dots, m-1\} \quad P_H(i)$$

$$Y \in \mathcal{Y} \quad P_{Y|H}(y|i)$$

• MAP rule $\hat{H}(y)$, $\hat{H}(y) = \underset{i}{\operatorname{argmax}} \underbrace{P_{H|Y}(i|y)}_{= \operatorname{argmax}_i \underbrace{P_{Y|H}(y|i) P_H(i)}}.$

• Suff. stats & irrelevance.

$$Y = (t, u) \quad \text{if} \quad P_{H|Y}(i|(t, u)) \\ = f_{\text{ndt}}(i|t)$$

$\Rightarrow u$ is irrelevant.
 t is a suff stat.

• Special Case: $Y \in \mathbb{R}^n$ ($\mathbb{C}^n$).

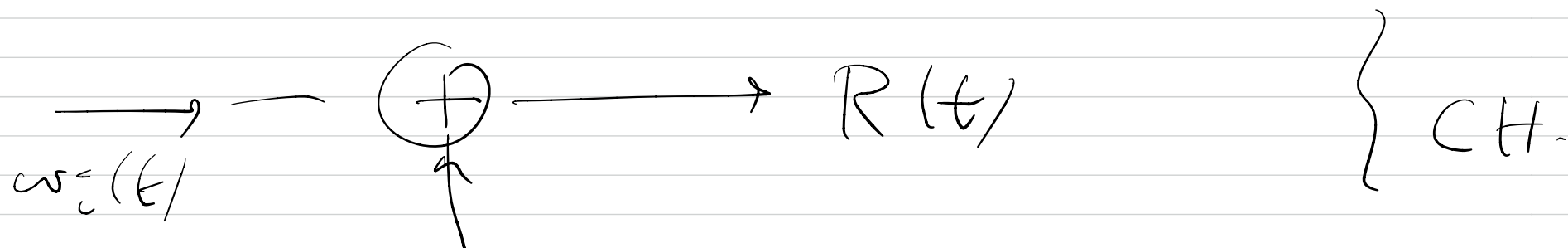
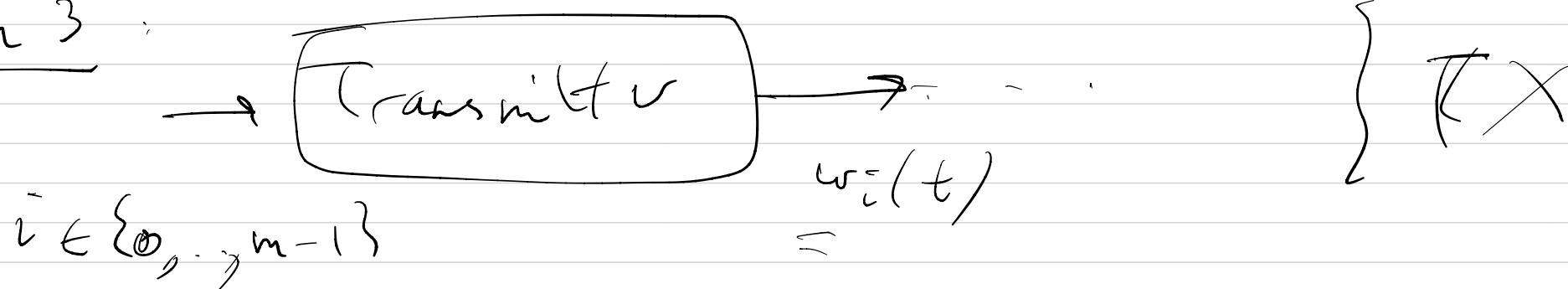
$$\text{when } H = i \quad Y = c_i + Z$$

$$c_i \in \mathbb{R}^n, \quad Z \sim \mathcal{N}(0, I_n \sigma^2).$$

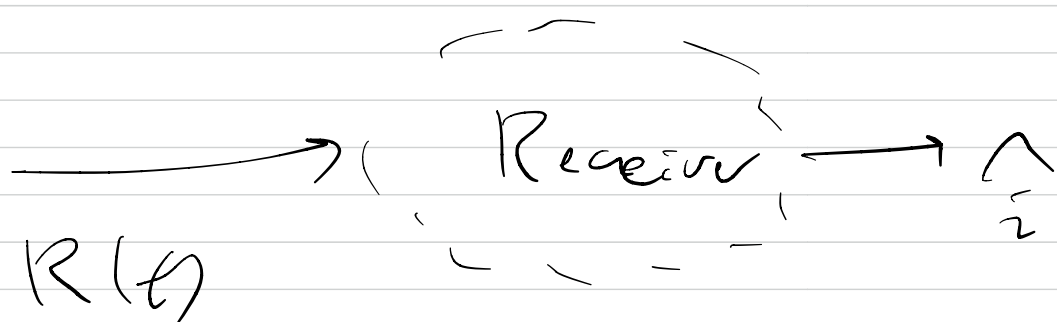
$$\hat{H}(y) = \underset{i}{\operatorname{argmax}} p_H(i) \frac{1}{\sigma^2} \exp\left(-\frac{1}{2\sigma^2} \|y - c_i\|^2\right)$$

$$\equiv \underset{i}{\operatorname{argmin}} \|y - c_i\|^2 - 2\sigma^2 \ln p_H(i).$$

Ch 3



$z(t)$: additive white Gaussian noise



To understand how to design the Receiver:

Suppose $\psi_1(t), \dots, \psi_n(t)$ are L -normal signals

such that $\operatorname{span}\{\psi_1, \dots, \psi_n\} = \operatorname{span}\{\underbrace{w_0, \dots, w_{m-1}}\}$

At the receiver form

$$Y_j = \langle R, \psi_j \rangle = \int R(t) \psi_j(t) dt$$

Claim $(Y_1 \dots Y_n)$ is a sufficient statistic

for any measurements we could have performed that include the Y 's.

$\equiv$ given $(\langle R, \psi_j \rangle = Y_j : j=1 \dots n)$ additional

measurements $(U_j = \langle R, g_j \rangle : j=1 \dots l)$

are irrelevant.

Recall $\text{span}(\psi) = \text{span}(w)$ so

$$w_0(t) = c_{01} \psi_1(t) + \dots + c_{0n} \psi_n(t)$$

$$w_i(t) = c_{i1} \psi_1(t) + \dots + c_{in} \psi_n(t)$$

So we have m vectors $c_0, \dots, c_{m-1}$

$$R(t) = \sum_{j=1}^n c_{ij} \psi_j(t) + \underbrace{Z(t)}_{H=0}$$

The receive now forms

$$y_j = \langle R, \psi_j \rangle \quad j=1 \dots n$$

$$= \left\langle \sum_{l=1}^n c_{il} \psi_l + z, \psi_j \right\rangle$$

when
 $H=i$

$$= \sum_{l=1}^n c_{il} \underbrace{\langle \psi_l, \psi_j \rangle}_{\delta_{lj}} + \underbrace{\langle z, \psi_j \rangle}_{z_j}$$

$$= c_{ij} + z_j$$

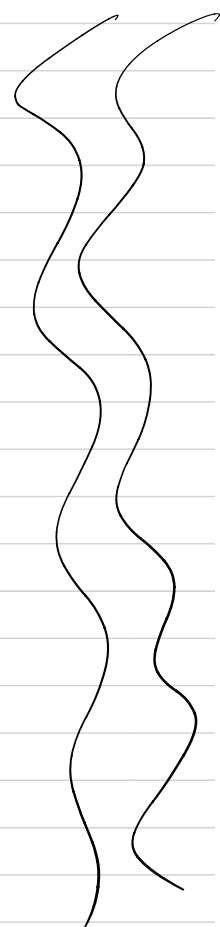
So when $H=i$

$$\begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} = \begin{pmatrix} c_{i1} \\ \vdots \\ c_{in} \end{pmatrix} + \begin{pmatrix} z_1 \\ \vdots \\ z_n \end{pmatrix}$$

y

c_i

$$z \sim \mathcal{N}(0, \frac{N_0}{2} I_n)$$



MAP decision of $\hat{i}$ from $(Y_1, \dots, Y_n)$.
 need to compute

$$\|y - c_i\|^2 - N_0 \ln p_H(i) \quad \text{for } i = 0, \dots, m-1$$

choose smallest.

~~$$\|y\|^2 - 2\langle y, c_i \rangle + \|c_i\|^2 - N_0 \ln p_H(i)$$~~

$$\|w_i\|^2$$

observe:

$$w_i(t) = c_{i1} \psi_1(t) + \dots + c_{in} \psi_n(t)$$

$$\begin{aligned} w_i(t)^2 &= c_{i1}^2 \psi_1(t)^2 + \dots + c_{in}^2 \psi_n(t)^2 \\ &\quad + \dots + \psi_1(t) \psi_2(t) + \dots + \psi_{n-1}(t) \psi_n(t) \end{aligned}$$

$$\int dt = c_{i1}^2 + c_{i2}^2 + \dots + c_{in}^2 = \|c_i\|^2$$

what about

$$\langle y, c_i \rangle = \sum_j c_{ij} \int R(t) \psi_j(t) dt$$

$$= \int R(t) \left(\sum_j c_{ij} \psi_j(t) \right) dt$$

$$S_0 \langle \gamma, c_i \rangle = \langle R, w_i \rangle,$$

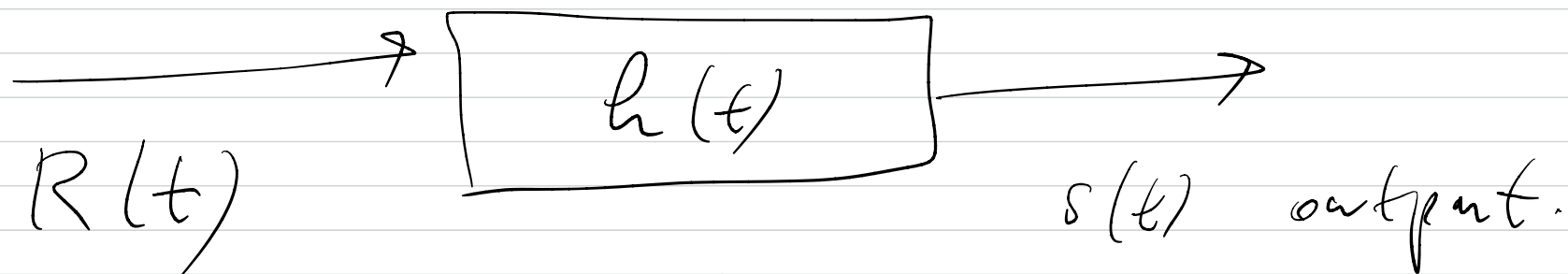
So the max rule: compute for
 $i = 0, \dots, m-1$

$$- 2 \int R(t) w_i(t) dt + \int w_i(t)^2 dt - N_0 \ln p_{1+}(i)$$

and find which is smallest

$$\equiv \arg \max_i \underbrace{\int R(t) w_i(t) dt}_{q_i} - \frac{\|w_i\|^2}{2} + \frac{N_0}{2} \ln p_{1+}(i)$$

A note on computing inner products of functions. $\int R(t) w(t) dt$



$$s(t) = \int R(\tau) h(t - \tau) d\tau$$

$$s(0) = \int R(\tau) h(-\tau) d\tau$$

$$= \langle \underline{R}, \underline{g} \rangle$$

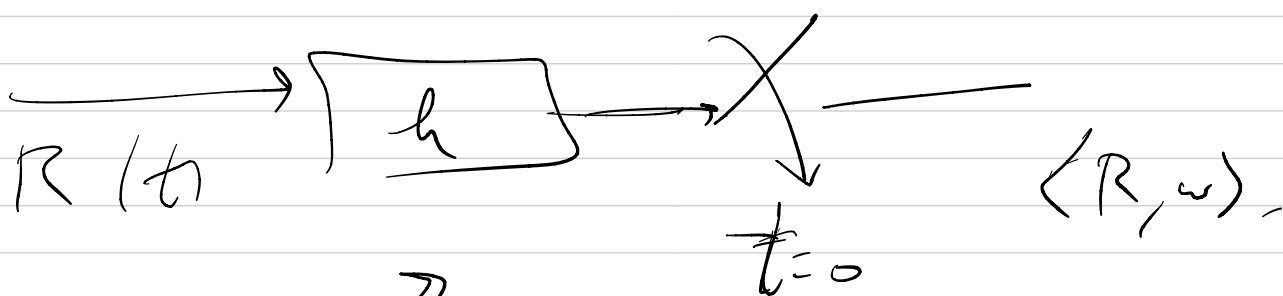
$$\underline{g(t)} = \underline{h(-t)}$$

$$s(t_0) = \langle R, \tilde{g} \rangle$$

$$\tilde{g}(\tau) = h(t_0 - \tau)$$

So to compute $\langle R, \underline{w} \rangle$ we could design a

filter with impulse response $h(t) = w(-t)$ and



matched filter implementation of $\langle R, w \rangle$.