12:00

Lelecoms

12/02/10

Summer 06:

Q5 4 brown modulation scheme is described bys-

$$S_i(t) = \int S_i(t) = A_i Cos(\omega_c t)$$
  $0 \le t \le T$   
 $LS_2(t) = A_2 Cos(\omega_c t)$   $0 \le t \le T$ 

Where T is an integer timeso /fc

Show:

(b) Average Signal energy per bit, Eb, is a fixed constant. I and I are also fixed, Prove that, to minimise to we require:

Solution

$$\Rightarrow \int_{\delta}^{T} [(A_{1} - A_{2}) C_{\infty}(\omega_{c} +)]^{2} dt \Rightarrow (A_{1} - A_{2})^{2} \int_{\delta}^{T} C_{0} s^{2}(\omega_{c} +) dt$$



$$= (A_1 - A_2)^2 \left( \frac{2\omega_c t}{4\omega_c t} + \frac{S_{in}(2\omega_c t)}{4\omega_c t} \right)^T$$

$$= (A_1 - A_2)^2 \frac{T}{2} \qquad S_{ince} \qquad S_{in}(2\omega_c t) = 0$$

$$S_{ince} \qquad T = 2 \frac{1}{5} \cdot n \in \mathbb{N}_o$$

$$P_e = Q \left( \frac{(A_1 - A_2)^2 T}{4 \cdot R} \right)^T$$

(b) Avg. signal energy per bit, 
$$E_b$$
, is:-
$$E_b = \frac{1}{2} \left[ \int_0^T \left( S_i^2(t) + S_2^2(t) \right) dt \right]$$

Hence:  

$$E_{0} = \frac{1}{2} \left[ (A_{1}^{2} + A_{2}^{2}) \int_{0}^{T} C_{2} \sigma^{2}(\omega_{e} t) dt \right] = (A_{1}^{2} + A_{2}^{2}) \left( \frac{1}{4} \right) = k$$

K is constant

We require i  $A_1^2 + A_2^2 = M$ , musa constant

To minimise le, we maximise its arguments.

$$\frac{ds}{dA_1} = \left(1 - \frac{dA_2}{dA_1}\right) 2 \left(A_1 - A_2\right) \frac{T}{412}$$



$$\left(1-\frac{dA}{dA}\right)=0$$
 Since  $A_1 \neq A_2$  because  $S_1(a)=B_2(a)$ 

Not Allowed

$$2A_1 + \frac{d}{dA_1}(A_2)^2 = 0$$

$$\frac{d}{dA_2}(A_2^2) \frac{dA_2}{dA_1} = 2A_2 \frac{dA_2}{dA_1}$$

$$\left(1+\frac{\Delta_{i}}{\Delta_{i}}\right)^{2}\left(A_{i}-A_{i}\right)\frac{T}{\Delta_{i}}=0 \qquad if d = 0$$



Example :

Suppose S. (t) and S. (t) must have the same energy. Prove that the optimum choice is:

$$S_{2}(t) = -S_{1}(t)$$

and that the resultant S.N.R is:

Where Eistle signal energy.

Proof?

$$(S/N)_{gp} = \frac{ZEd}{7}e$$

$$\Rightarrow \frac{2}{7} \int_{0}^{T} (S_{1}(t) - S_{2}(t))^{2} dt$$

$$= \frac{2}{7} \int_{0}^{7} (S_{1}^{2}(t) - Z(S_{1}(t) S_{2}(t)) + S_{2}^{2}(t)) dt$$

$$= \frac{L_1 E}{7} - \frac{4}{7} \int_0^T S_1(t) S_2(t) dt$$

ilong the Schools inequality ?-

$$\left|\int_{0}^{T} S_{s}(t)S_{2}(t)dt\right| \leq \left|\int_{0}^{T} S_{s}^{2}(t)dt\right|^{2} S_{2}^{2}(t)dt$$

Equality holds when Sa(t)= kS,(t)

		)

$$\Rightarrow k = \pm 1$$

Hence?

Thus by substituting we get

$$\left(\frac{S}{N}\right)_{0/p} = \frac{4E}{2} - \frac{4}{2} \int_{0}^{7} S_{1}(4)S_{2}(4)db$$

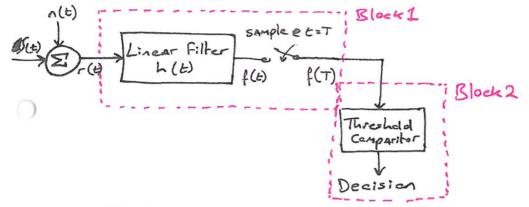
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Dr. C. Murphy

We now turn our attention to the more general guestran of minimising probability of error for any arbitrary modulation scheme.

Binary Signal Detection & Hypothesis Testing
Once again, we confine our attention to ANGN and a distortantess
channel, i.e. no inter-signal interference (1.5.1), The following system diagram
summarises the receiver structure:-



The transmitted signal over a symbol interval (0,T) is represented by:  $S:(t): S:(t) = 0 \le t \le T$  for logic 1  $S:(t) = S:(t) = 0 \le t \le T$  for logic 0

at the receiver

The received signal r(t): represented by:r(t)= S:(t)+n(t) i=1,2 0=t=T
where n(t) is a zero-mean AMGN (Additive White Gaussian Noise).

As before, there are two steps involved in signal detection. The first step reduces the received signal r(t) to a (statement ) single number f(T). This operation can be performed by a linear filter followed by a sampler, as shown in Block I above.

The output of Black I, sampled a toT, yields:

z(T) = a: (T) + n. (T)

where a: (T) is the signal (ONLY) component of z(T) and Mo(T) is the noise (ONLY) component of z(T)

1=1,2

(after filtering)



For simplicity, we usually writes.

Z= a;+10 ::1,2

Note that the noise component to is also a zero-near Gaussian random variable (rv) ( We will confirm this later). Hence z is also a Gaussian rv with a near of either a, or az, depending on whether S, (E) or Sz (E) was sent. The semple z is often called the "test statistic".

The second step in the signal detection process consists of comparing z to a threshold level & in Black 2. The decision is then made according to :-

Z NA A

where H, & Hz are the 2 possible hypetheses. Choosing H, is equivalent to deciding that signal S.(c) was sent, and choosing Hz is equivalent to deciding that Sz(E) was sent. The inequality above indicates that H, is above if z is larger than 2, and Hz is chosen if z is less than 2. If z=2 an arbitrary decision can be made.

As before, the words probability of error, Pe, is given by:

Pe = P(S,)P(Hz|S,)+P(Sz)P(H,|Sz)

Probability of deciding etc.
He when S, was set

where  $P(S_i)_{i=1,2}$  is the probability of reading  $S_i(t)$ .

In the equi-probable case,  $P(S_i) = P(S_Z) = \frac{1}{2}$  and  $P(S_Z) = \frac{1}{2} \left(P(H_Z|S_i) + P(H_I|S_Z)\right)$ 

MAXIMUM Like hood Detector

A popular (without) criterion for choosing  $\lambda$  is based an animality R.

The calculation for this "minimum error" value f  $\lambda = \lambda_0$  starts with this likelihood rates test:  $\frac{f(z|s_i)}{f(z|s_0)} \stackrel{\text{He}}{=} \frac{P(s_0)}{P(s_0)}$ 

where f (ZISi) is the anditrocal poly known as the "likelihood of Si". Heree we can write:

 $\Lambda(z) = \frac{f(z|s_i)}{f(z|s_2)} + \frac{1}{4}$  in the guapropole case, since  $P(s_i) = P(s_i) = \frac{1}{2}$ 



If, in addition, the likelihood,  $f(Z|S_i)$  i. 1,2 are symmetric, then this further reduces to:- $Z \gtrsim \lambda_0$  where  $\lambda_0 = \frac{a_1 + a_2}{Z}$ .

We will show that he = 2 is the optimum threshold for minimistry Pe in the equipodade case (ha) Since this detector minimises Pe, it is known as a "Maximum likelihood detector"

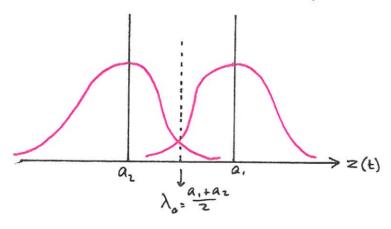
Probability of Error with Gaussian Noise

Assuming Gaussian noise it follows that here we have:-(z-a,) =/zma=

f(z|s,) = Jzma== e (z-az) =/zma==

f(z|s,) = Jzma== e (z-az) =/zma==

Where on = stendard deviation of the filtered noise.



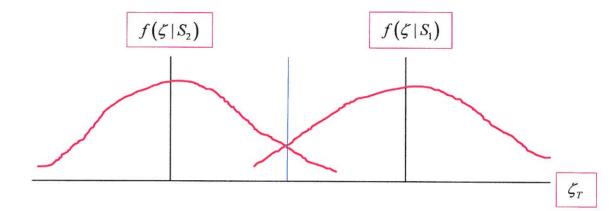


## **Telecommunications - Section 2 (Lecture 11?)**

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Assuming Gaussian noise it follows that here we have:-

$$\frac{-\left(\zeta-a_{1}\right)^{2}}{2V_{n0}^{2}}$$



Now:-

$$P(H_2 | S_1) = \int_{-\infty} f(\zeta | S_1) d\zeta$$

$$P(H_1 | S_2) = \int_{\lambda_0}^{\infty} f(\zeta | S_2) d\zeta$$

Because of the symmetry of  $f(\zeta | S_i)$ ,  $P_e$  reduces to:-

$$P_e = \frac{1}{2} (P(H_2 | S_1) + P(H_1 | S_2))$$

$$P_e = Q \left( \frac{a_1 - a_2}{2\delta n_0} \right)$$

where

$$Q(x) = \frac{1}{\sqrt{2\pi}} \int_{x} e \, dy$$

Example:

Prove that if  $P(S_1) = P(S_2)$  then  $\lambda_0 = \frac{(a_1 + a_2)}{2}$  minimises  $P_e$ .

Solution:



Assume the threshold is set at  $\lambda$ , and so:-

### **Optimum Detection**

### **Background Definitions**

### Wide-Sense Stationary Process

A random process x(t) is "wide-sense stationary" (wss) if its mean is constant, i.e.  $E\left[x(t)\right] = \mu_x = \int_{-\infty}^{\infty} x f_x\left(x;t\right) dx$  and its autocorrelation depends only on the time difference  $\tau$ . i.e.  $E\left[x(t)x(t+\tau)\right] = R_{xx}\left(\tau\right)$  (see below).

### Autocorrelation $R_{xx}(\tau)$

The autocorrelation of x(t) is defined by:-

$$R_{xx}(t) = E\left[x(t)x(t+\tau)\right]$$
$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x_1 x_2 f_x(x_1, x_2; t_1, t_2) dx_1 dx_2$$

And has the following properties:-

- 1. Even function:-  $R_{xx}(-\tau) = R_{xx}(\tau)$
- $2. |R_{xx}(\tau)| \le E[x^2(t)]$
- 3.  $R_{xx}(0) = E[x^2(t)] = \text{total average normalised power to a } 1\Omega \text{ load.}$

## Power Spectrum Density (PSD)

The PSD of x(t) is defined by the Fourier transform of  $R_{xx}(\tau)$ :-

$$S_{xx}(\omega) = \int_{-\infty}^{\infty} R_{xx}(t)e^{-j\omega t}dt$$
  
 $\Rightarrow R_{xx}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xx}(\omega)e^{j\omega t}d\omega$  Wiener-Khiachin Relations  
 $PSD S_{xx}(\omega)$  has the following properties:-

- 1  $S_{xx}(\omega)$  is real and  $S_{xx}(\omega) \ge 0$ .
- 2  $S_{xx}(-\omega) = S_{xx}(\omega)$  (Even also)

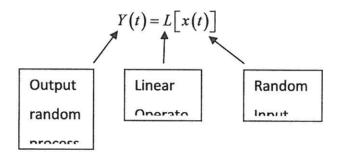
3 
$$R_{xx}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{xx}(\omega) d\omega = R_{xx}(0) = E[x^2(t)]$$

			$\bigcirc$

# Telecommunications – Section 2 (Lecture?) 4/2/10

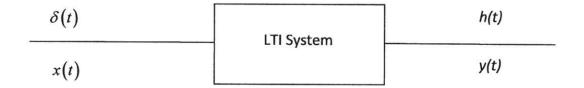
# Transmission of Random Processes Through Linear Systems

A linear LTI system can be represented by:-



Let h(t) be the impulse response of the LTI. Then:

$$Y(t) = h(t) * x(t)$$
$$= \int_{-\infty}^{\infty} h(\alpha) x(t - \alpha) d\alpha$$



### Mean and Autocorrelation of the Output

The mean of the output



$$\mu_{y}(t) = E[y(t)]$$

$$= E\left[\int_{-\infty}^{\infty} h(\alpha)x(t-\alpha)d\alpha\right]$$

$$= \int_{-\infty}^{\infty} h(\alpha)E[x(t-\alpha)]d\alpha$$

$$= \int_{-\infty}^{\infty} h(\alpha)\mu_{x}(t-\alpha)d\alpha$$

$$= h(t)*\mu_{x}(t)$$

If the input is wide-sense stationary we have:

$$E[y(t)] = \int_{-\infty}^{\infty} h(\alpha) \mu_x \, d\alpha$$
$$= \mu_x \int_{-\infty}^{\infty} h(\alpha) \, d\alpha$$
$$= \mu_x H(0)$$

Where H(0) is the frequency response of the linear system at  $\omega=0$ . Thus, the mean of the output is a constant. The autocorrelation of the output,  $R_{yy}(\tau)$ , is given by:

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(\alpha)h(\beta)R_{xx}(\tau+\alpha-\beta)d\alpha d\beta$$

And the output is also wide-sense-stationary.

#### **Power Spectral Density of Output**

Taking the Fourier transform of both sides of this last expression yields:

$$S_{\gamma\gamma}(\omega) = \int_{-\infty}^{\infty} R_{\gamma\gamma}(\tau) e^{-j\omega\tau} d\tau$$

$$which = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(\alpha)h(\beta) R_{xx}(\tau + \alpha - \beta) e^{-j\omega\tau} d\tau d\alpha d\beta$$

$$\Rightarrow s_{\gamma\gamma}(\omega) = |H(\omega)|^2 S_{xx}(\omega)$$

The autocorrelation of the output,  $R_{yy}(\tau)$ , is often best evaluated via the inverse Fourier transform of  $S_{yy}(\omega)$ :-

$$R_{yy}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} |H(\omega)|^2 S_{xx}(\omega) e^{j\omega t} d\omega$$

Hence, the average power in the output y(t) is



$$E[y^{2}(t)] = R_{yy}(0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} |H(\omega)|^{2} S_{xx}(\omega) d\omega$$

#### White Noise

A random process x(t) is called "white noise" if  $S_{xx}(\omega) = \frac{\eta}{2}$ 

It is usually assumed that the mean of white noise is zero. (Note  $S_{xx}(\tau)$  relates to  $|V|^2$  and the mean relates to V itself.)

### **Optimum Detection**

Recall earlier we had:

$$P_e = Q \left[ \frac{a_1 - a_2}{2\sigma_{n_0}} \right].$$

And so, to minimise Pe we search for the filter capable of maximising  $\frac{a_{\rm l}-a_{\rm 2}}{2\sigma_{n_{\rm 0}}}$ 

#### **Matched Filter**

A matched filter, as we will see, provides the maximum output signal-to-noise ratio for a given transmitted signal. Consider that a known signal S(t) plus AWGN n(t) is the input to a LTI filter, followed by a sampler. Let a(t) be the "signal only" response of the filter. Then, at t=T we have:-

$$\left(\frac{S}{N}\right)_{0} = \frac{a^{2}\left(T\right)}{E\left[n_{0}^{2}\left(T\right)\right]}$$

Assuming a  $1\Omega$  reference resistor

$$= \frac{a^2(T)}{\sigma_{n_0}^2} \to \text{since } \sigma_{n_0}^2 = E\left[\left(y - \overline{y}\right)^2\right]$$

But our input noise has zero mean  $\Rightarrow$  so too does our output noise  $\Rightarrow \overline{y} = 0$ 

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# Example:

Find the optimum filter Ho(w) that maximises the output SNR when the input noise is not white (charred noise).

Solution:

Let  $H(\omega)$  be the frequency response of the linear fitter Let  $S_{nn}(\omega)$  be the power spectrum of the input coloured noise. At t=T we have as (follows) before;

and the overage output noise power is;-

=> output SNR is; -

$$\left(\frac{5}{N}\right)_{0} = \frac{1}{2\pi} \left[ \int_{-\infty}^{\infty} H(\omega) S(\omega) e^{i\omega T} d\omega \right]^{2}$$

$$\int_{-\infty}^{\infty} S_{NN}(\omega) |H(\omega)|^{2} d\omega$$

To find the optimum filter set: -

$$f_1(\omega) = \sqrt{5nn(\omega)} H(\omega)$$

$$f_2(\omega) = \frac{5(\omega) e^{i\omega T}}{\sqrt{5nn(\omega)}}$$

we can now write using the Schwarz inequality:-

