## EE4061 Communication Systems

Week 8

Coherent Signal Detection

## **Optimum Coherent Detection**

Suppose that we have M signals  $s_1(t), s_2(t), \ldots, s_M(t)$  that are defined over the time interval  $0 \le t \le T$ , where T is the baud period or baud interval. The baud rate is R = 1/T.

Note that  $T_b = T/\log_2 M$  and  $R = R_b/\log_2 M$ , where  $R_b = 1/T_b$  is the bit rate.

We select one of signals, say  $s_i(t)$ , for transmission over an AWGN channel.

The received signal is

$$r(t) = s_i(t) + n(t)$$

where n(t) is AWGN with power spectral density  $N_o/2$  watts/Hz.

**Problem:** By observing r(t) determine which one of the M signals in the set  $\{s_1(t), s_2(t), \ldots, s_M(t)\}$  was (most likely) transmitted.

Repeat this process once every T seconds for each modulated signal that is transmitted.

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### **Correlation Detector**

Any signal set  $\{s_1(t), s_2(t), \ldots, s_M(t)\}$  can be expressed in terms of a set of orthonormal basis functions  $\{f_1(t), f_2(t), \ldots, f_N(t)\}$  where N is the dimension of the signal space. Recall the Gram-Schmidt orthonormalization procedure.

However, the basis functions do not span the noise space, i.e., the noise waveform n(t) cannot be represented exactly in terms of the N basis functions. We have

$$r(t) = \sum_{k=1}^{N} r_i f_i(t) + \hat{n}(t)$$

where

$$r_k = \int_0^T r(t)f_k(t)dt$$
  
= 
$$\int_0^T s_m(t)f_k(t)dt + \int_0^T n(t)f_k(t)dt$$
  
= 
$$s_{mk} + n_k$$

Hence, the projection of the received signal r(t) onto the *signal space* yields the received vector  $\mathbf{r} = (r_1, r_2, \dots, r_N)$ .

## Remainder Process

Note that

$$r(t) = \sum_{k=1}^{N} r_k f_k(t) + \hat{n}(t)$$

where  $\hat{n}(t)$  is the **remainder process** 

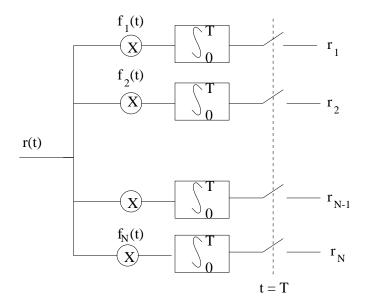
$$\hat{n}(t) = n(t) - \sum_{k=1}^{N} n_k f_k(t)$$

We will see later that the remainder process  $\hat{n}(t)$  is irrelevant when deciding as to which  $s_i(t)$  was sent.

The vector  $\mathbf{r} = (r_1, r_2, \dots, r_N)$  is said to provide *sufficient statistics*, meaning that nothing else is required other than the vector  $\mathbf{r}$  for the receiver to decide which  $s_i(t)$  was sent.

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# Correlation Detector



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## **Noise Statistics**

The noise components  $n_k$  have mean

$$E[n_k] = \int_0^T E[n(t)] f_k(t) dt = 0$$

and covariance

$$E[n_{j}n_{k}] = \int_{0}^{T} \int_{0}^{T} E[n(t)n(s)]f_{j}(t)f_{k}(s)dtds$$

$$= \frac{N_{o}}{2} \int_{0}^{T} \int_{0}^{T} \delta(t-s)f_{j}(t)f_{k}(s)dtds$$

$$= \frac{N_{o}}{2} \int_{0}^{T} f_{j}(t)f_{k}(t)dt$$

$$= \frac{N_{o}}{2} \delta_{jk}$$

Therefore, the  $r_k$  are independent Gaussian random variables with mean  $s_{mk}$  and variance  $N_o/2$ .

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## Joint Conditional Density

Since elements of the vector  $\mathbf{r}$  are independent random variables the joint conditional density function of the vector  $\mathbf{r}$  has the product form

$$p(\mathbf{r}|\mathbf{s_m}) = \prod_{k=1}^{N} p(r_k|s_{mk})$$

$$= \prod_{k=1}^{N} \frac{1}{\sqrt{\pi N_o}} \exp\left\{-\frac{1}{N_o} (r_k - s_{mk})^2\right\}$$

$$= \frac{1}{(\pi N_o)^{N/2}} \exp\left\{-\frac{1}{N_o} \sum_{k=1}^{N} (r_k - s_{mk})^2\right\}$$

$$= \frac{1}{(\pi N_o)^{N/2}} \exp\left\{-\frac{1}{N_o} \|\mathbf{r} - \mathbf{s}_m\|^2\right\}$$

which is a multivariate Gaussian distribution.

Note that we have used the notation

$$\|\mathbf{r} - \mathbf{s}_m\|^2 = \sum_{k=1}^{N} (r_k - s_{mk})^2$$

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### Irrelevance

We have

$$E[\hat{n}(t)r_{k}] = E[\hat{n}(t)]s_{mk} + E[\hat{n}(t)n_{k}]$$

$$= E[\hat{n}(t)n_{k}]$$

$$= E\left[\left(n(t) - \sum_{j=1}^{N} n_{j}f_{j}(t)\right)n_{k}\right]$$

$$= \int_{0}^{T} E[n(t)n(\tau)]f_{k}(\tau)d\tau - \sum_{j=1}^{N} E[n_{k}n_{j}]f_{j}(t)$$

$$= \frac{1}{2}N_{o}f_{k}(t) - \frac{1}{2}N_{o}f_{k}(t) = 0$$

Hence, the vector  $\mathbf{r}$  is uncorrelated with  $\hat{n}(t)$  and, therefore,  $\hat{n}(t)$  is irrelevant since it does not contain any information about  $\mathbf{r}$ .

This is Wozencraft's irrelevance theorem which is certainly not irrelevant!

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## Matched Filter Receiver

Suppose that we filter the received signal r(t) with a bank of matched filters having the impulse responses

$$h_k(t) = f_k(T - t)$$
 ,  $0 \le t \le T$ 

and sample the filter outputs at time t = T.

The filter outputs are

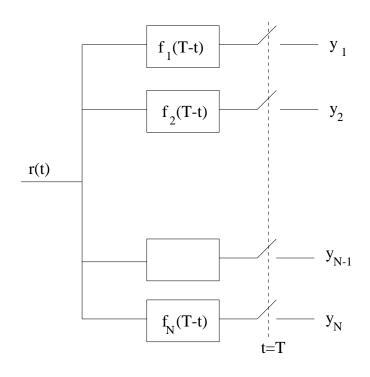
$$y_k(t) = \int_0^t r(\tau)h_k(t-\tau)d\tau$$
$$= \int_0^t r(\tau)f_k(T-t+\tau)d\tau$$

$$y_k \equiv y_k(T) = \int_0^T r(\tau) f_k(\tau) d\tau$$

Note that  $y_k = r_k$ , i.e., the matched filter outputs are identical to the correlator outputs.

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# Matched Filter Receiver



### Minimum Distance Decisions

With minimum distance decisions, the receiver first calculates the vector  $\mathbf{r}$ . The receiver then decides in favour of the signal point  $\mathbf{s}_i$  that is closest in *Euclidean distance* or squared *Euclidean distance* to the received vector  $\mathbf{r}$ .

The minimum distance decision rule is

$$\hat{\mathbf{s}} = \frac{\arg\min}{\mathbf{s}_i} \|\mathbf{r} - \mathbf{s}_i\|^2$$

Since, the vector  $\mathbf{r}$  has the joint conditional density function

$$p(\mathbf{r}|\mathbf{s}_m) = \frac{1}{(\pi N_o)^{N/2}} \exp\left\{-\frac{1}{N_o} ||\mathbf{r} - \mathbf{s}_m||^2\right\}$$

the choice of  $\mathbf{s}_i$  that minimizes  $\|\mathbf{r} - \mathbf{s}_i\|^2$  also maximizes the likelihood  $p(\mathbf{r}|\mathbf{s}_m)$ .

Hence minimum distance decisions are maximum likelihood decisions.

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