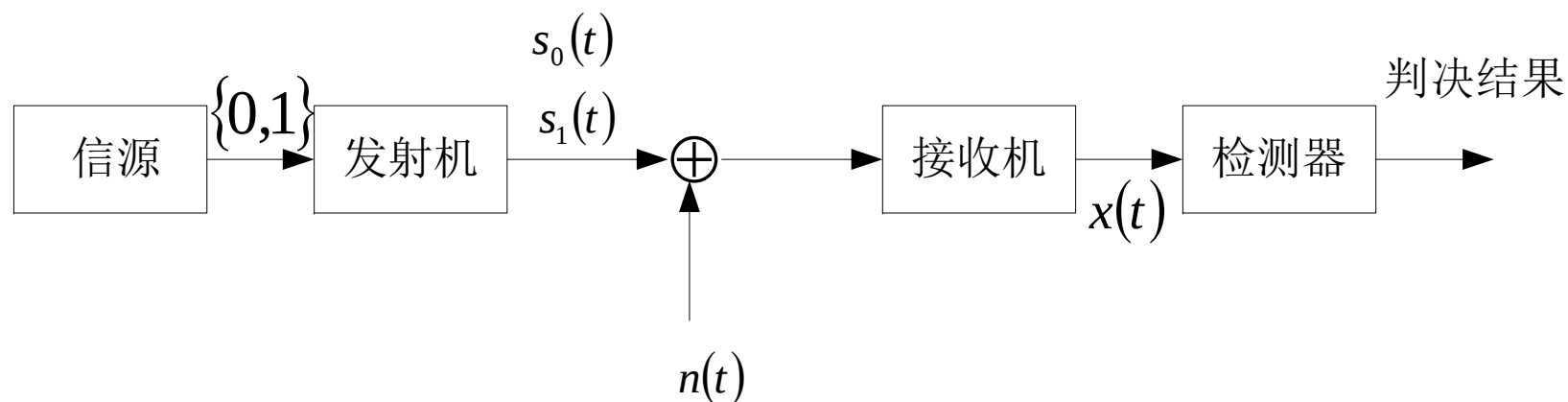




第四章 波形信号检测

- 二元信号波形检测模型
- 匹配滤波器
- 正交级数展开
- 波形信号检测

二元信号波形检测模型



信源输出

0

发射信号

$$s_0(t), \quad nT \leq t \leq (n+1)T$$

1

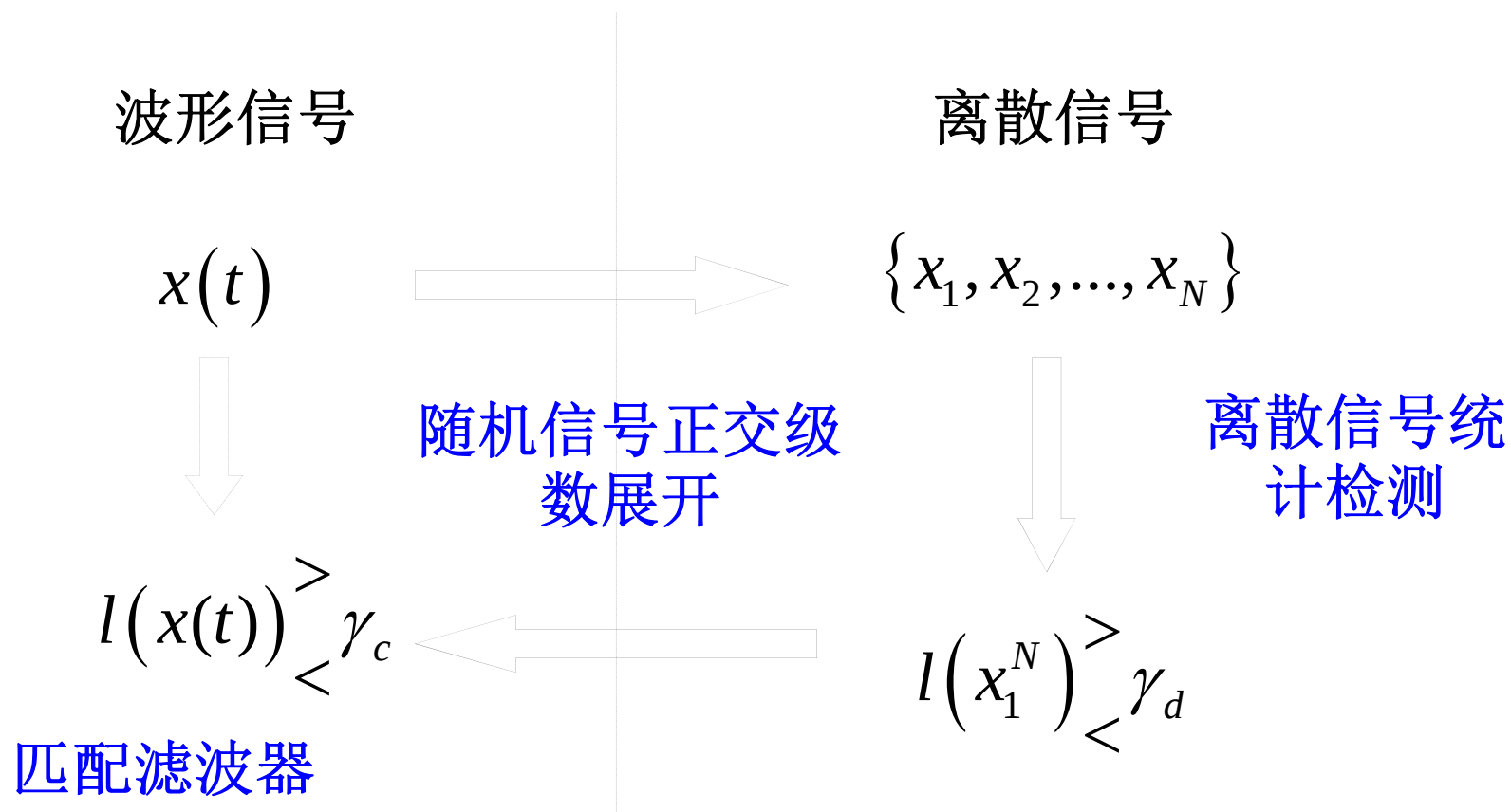
$$s_1(t), \quad nT \leq t \leq (n+1)T$$

信号在信道中传输，受到加性噪声的干扰，可描述为：

$$H_0: \quad x(t) = s_0(t) + n(t), \quad nT + t_0 \leq t \leq (n+1)T + t_0$$

$$H_1: \quad x(t) = s_1(t) + n(t), \quad nT + t_0 \leq t \leq (n+1)T + t_0$$

波形信号检测方法

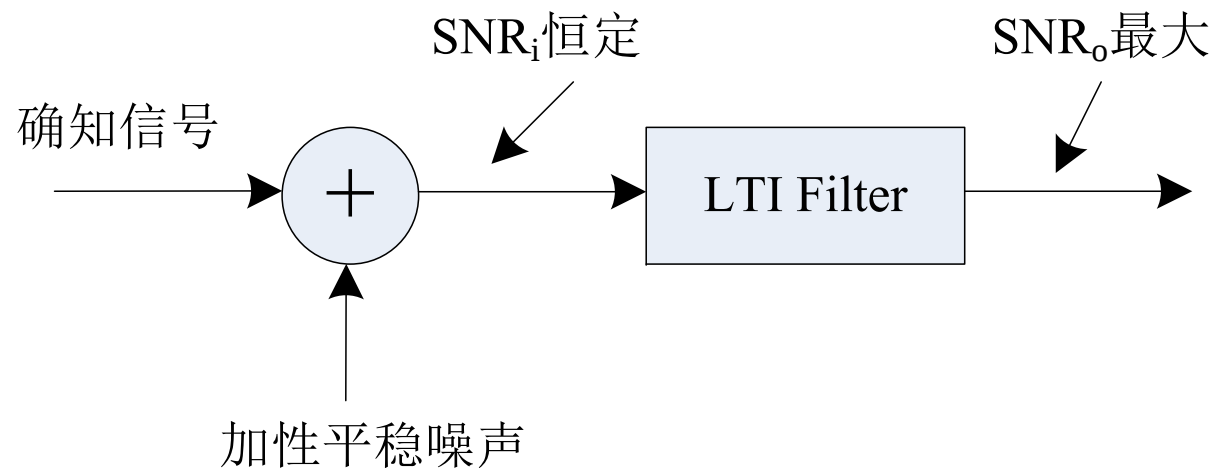




匹配滤波器 (MF)

● 定义

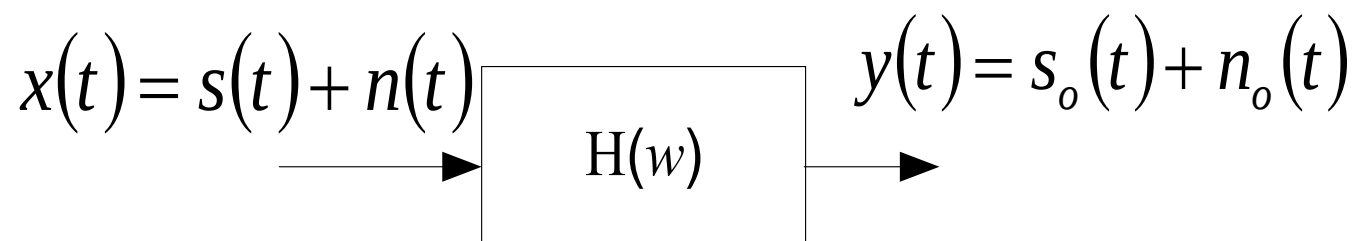
----若线性时不变 (LTI) 滤波器输入的信号是确知信号，噪声是加性平稳噪声，则在输入功率信噪比一定的条件下，使输出功率信噪比最大的滤波器，即为与输入信号匹配的最佳滤波器，称为匹配滤波器。



系统模型



- 假设线性时不变滤波器的冲激相应为 $\mathbf{h(t)}$ ，系统函数为 $H(w)$ ，滤波器的输入信号为 $\mathbf{x(t)=s(t)+n(t)}$ ，滤波器的输出信号为 $\mathbf{y(t)=s_o(t)+n_o(t)}$ 。



$$S(\omega) = \int_{-\infty}^{\infty} s(t) e^{-j\omega t} dt \quad E_s = \int_{-\infty}^{\infty} s^2(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |S^2(\omega)| d\omega$$

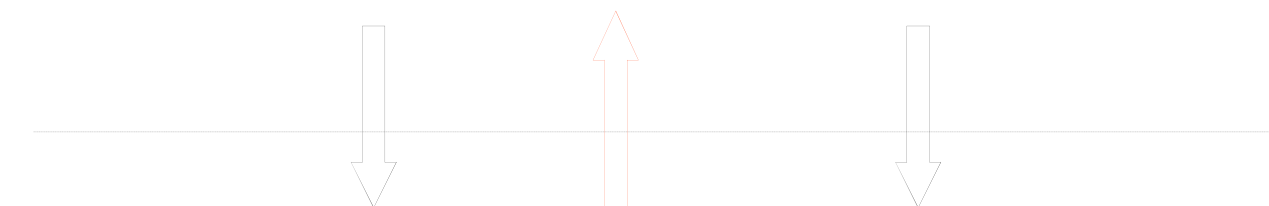
- 设计目标：输出功率信噪比最大。

设计思路



时域

$$s(t) \xrightarrow{h(t)} s_0(t)$$



频域

$$S(w) \xrightarrow{H(w)} S_o(w)$$

$$S(w) = FT \{s(t)\} = \int_{-\infty}^{\infty} s(t) e^{-j\omega t} dt$$

$$s(t) = IFT \{S(w)\} = \frac{1}{2\pi} \int_{-\infty}^{\infty} S(w) e^{j\omega t} dw$$



Formulation

- 目标：输出信号功率信噪比定义为输出信号 $s_o(t)$ 的峰值功率与噪声 $n_o(t)$ 的平均功率之比。

$$SNR_o \stackrel{\text{def}}{=} \frac{s_o(t) \text{ 的功率}}{n_o(t) \text{ 的平均功率}}$$

- 已知条件：

$$s_o(t) = s(t) * h(t) \quad S_o(\omega) = S(\omega)H(\omega)$$

$$\begin{aligned} E_s &= \int_{-\infty}^{\infty} s^2(t) dt < \infty & s_o(t) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S_o(\omega) e^{j\omega t} d\omega \\ & & &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S(\omega) H(\omega) e^{j\omega t} d\omega \\ P_{n_o}(\omega) &= |H(\omega)|^2 P_n(\omega) \end{aligned}$$

Solution



- 假设输出信号在 t_0 出现峰值，则输出信号的峰值功率为

$$|s_o(t_0)|^2 = \left| \frac{1}{2\pi} \int_{-\infty}^{\infty} S(\omega) H(\omega) e^{j\omega t_0} d\omega \right|^2$$

- 输出噪声的平均功率为

$$\begin{aligned} E[n_o^2(t)] &= \frac{1}{2\pi} \int_{-\infty}^{\infty} P_{n_o}(\omega) d\omega \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} |H(\omega)|^2 P_n(\omega) d\omega \end{aligned}$$

Solution



$$\begin{aligned} SNR_o &\stackrel{def}{=} \frac{s_o(t) \text{ 的功率}}{n_o(t) \text{ 的平均功率}} \\ &= \frac{\left| \frac{1}{2\pi} \int_{-\infty}^{\infty} H(\omega) S(\omega) e^{j\omega t_0} d\omega \right|^2}{\frac{1}{2\pi} \int_{-\infty}^{\infty} |H(\omega)|^2 P_n(\omega) d\omega} \end{aligned}$$

Cauchy-Schwarz不等式



$$\left| \frac{1}{2\pi} \int_{-\infty}^{\infty} F^*(t)Q(t)dt \right|^2 \leq \frac{1}{2\pi} \int_{-\infty}^{\infty} F^*(t)F(t)dt \cdot \frac{1}{2\pi} \int_{-\infty}^{\infty} Q^*(t)Q(t)dt$$

等号成立的条件为

$$Q(t) = \alpha F(t)$$

$$F^*(\omega) = \frac{S(\omega)e^{j\omega t_0}}{\sqrt{P_n(\omega)}} \quad Q(\omega) = H(\omega)\sqrt{P_n(\omega)}$$

Alternative form:

$$\|\langle a, b \rangle\| \leq \|a\| \cdot \|b\|$$

Solution



变量替换

$$F^*(\omega) = \frac{S(\omega)e^{j\omega t_0}}{\sqrt{P_n(\omega)}} \quad Q(\omega) = H(\omega)\sqrt{P_n(\omega)}$$

$$\left| \frac{1}{2\pi} \int_{-\infty}^{\infty} H(\omega)S(\omega)e^{j\omega t_0} d\omega \right|^2 = \left| \frac{1}{2\pi} \int_{-\infty}^{\infty} H(\omega)\sqrt{P_n(\omega)} \frac{S(\omega)e^{j\omega t_0}}{\sqrt{P_n(\omega)}} d\omega \right|^2$$

$$\leq \frac{1}{2\pi} \int_{-\infty}^{\infty} \left| H(\omega)\sqrt{P_n(\omega)} \right|^2 d\omega \cdot \frac{1}{2\pi} \int_{-\infty}^{\infty} \left| \frac{S(\omega)e^{j\omega t_0}}{\sqrt{P_n(\omega)}} \right|^2 d\omega$$

Solution



$$\begin{aligned} SNR_o &\stackrel{\text{def}}{=} \frac{s_o(t) \text{ 的功率}}{n_o(t) \text{ 的平均功率}} = \frac{\left| \frac{1}{2\pi} \int_{-\infty}^{\infty} H(\omega) S(\omega) e^{j\omega t_0} d\omega \right|^2}{\frac{1}{2\pi} \int_{-\infty}^{\infty} |H(\omega)|^2 P_n(\omega) d\omega} \\ &\leq \frac{\frac{1}{2\pi} \int_{-\infty}^{\infty} |H(\omega) \sqrt{P_n(\omega)}|^2 d\omega \cdot \frac{1}{2\pi} \int_{-\infty}^{\infty} \left| \frac{S(\omega) e^{j\omega t_0}}{\sqrt{P_n(\omega)}} \right|^2 d\omega}{\frac{1}{2\pi} \int_{-\infty}^{\infty} |H(\omega) \sqrt{P_n(\omega)}|^2 d\omega} \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{|S(\omega)|^2}{P_n(\omega)} d\omega \end{aligned}$$

Solution



由于

$$F^*(\omega) = \frac{S(\omega)e^{j\omega t_0}}{\sqrt{P_n(\omega)}} \quad Q(\omega) = H(\omega)\sqrt{P_n(\omega)}$$

所以, 当 $Q(\omega) = \alpha F(\omega)$ 时, 即

$$H(\omega)\sqrt{P_n(\omega)} = \alpha \frac{S^*(\omega)e^{-j\omega t_0}}{\sqrt{P_n(\omega)}}$$

$$H(\omega) = \alpha \frac{S^*(\omega)e^{-j\omega t_0}}{P_n(\omega)}$$

Solution



信道噪声为高斯白噪声的情况

$$P_n(\omega) = \frac{N_0}{2}$$

$$H(\omega) = 2\alpha \frac{S^*(\omega)e^{-j\omega t_0}}{N_0} = kS^*(\omega)e^{-j\omega t_0} \quad h(t) = ks(t_0 - t)$$

由Parseval定理

$$SNR_O = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{2|S(\omega)|^2}{N_0} d\omega = \frac{2E_s}{N_0}$$



Parseval定理

● Parseval定理

$$E_s = \int_{-\infty}^{\infty} |s(t)|^2 dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |S(w)|^2 dw$$

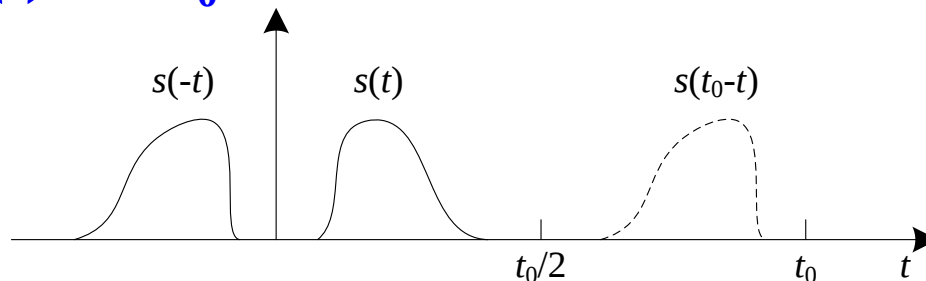
● Proof:

$$\begin{aligned} E_s &= \int_{-\infty}^{\infty} |s(t)|^2 dt = \int_{-\infty}^{\infty} s^*(t) \frac{1}{2\pi} \int_{-\infty}^{\infty} S(w) e^{j\omega t} d\omega dt \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} S(w) \int_{-\infty}^{\infty} s^*(t) e^{j\omega t} dt d\omega = \frac{1}{2\pi} \int_{-\infty}^{\infty} S(w) S^*(w) dw \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} |S(w)|^2 dw \end{aligned}$$



MF的性质

- $h(t)$ 与 $s(t)$ 对于 $t_0/2$ 呈对偶关系



- $h(t)$ 必须是物理可实现的，有

$$h(t) = \begin{cases} s(t_0 - t), & t \geq 0 \\ 0, & t < 0 \end{cases}$$

- 为了确保输入信号 $s(t)$ 的全部都能对输出信号有贡献， t_0 应满足 $s(t) = 0, t > t_0$
 t_0 至少选择在 $s(t)$ 的末尾



MF的性质

- $t_0=T$

$$h(t) = s(t_0 - t), 0 \leq t \leq t_0$$

$$\begin{aligned} s_0(t) &= \int_0^{t_0} h(\tau) s(t - \tau) d\tau \\ &= \int_0^{t_0} s(t_0 - \tau) s(t - \tau) d\tau \\ &= \int_0^{t_0} s^2(u) du \end{aligned}$$

- 既然 $s(t)$ 的持续时间为 $(0, T)$ ，选择 $t_0=T$ 使得 $s_0(t)$ 最大，从而输出SNR最大。选取 $t_0>T$ 不会改善输出SNR，只会延迟做出判决的时间



MF的性质

- 匹配滤波器的输出信号功率信噪比

$$SNR_O = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{2|S(\omega)|^2}{N_0} d\omega = \frac{2E_s}{N_0}$$

MF的性质



- 对于振幅和时延参量不同的信号，匹配滤波器具有适应性

$$H(\omega) = kS^*(\omega)e^{-j\omega t_0}$$

$$s_1(t) = As(t - \tau) \quad S_1(\omega) = AS(\omega)e^{-j\omega\tau}$$

$$\begin{aligned} H_1(\omega) &= kS_1^*(\omega)e^{-j\omega t_1} = AkS^*(\omega)e^{j\omega\tau}e^{-j\omega t_1} \\ &= Ak \frac{H(\omega)}{e^{-j\omega t_0}k} e^{j\omega\tau}e^{-j\omega t_1} = AH(\omega)e^{-j\omega[t_1 - (\tau + t_0)]} \\ &= AH(\omega) \quad \text{if} \quad t_1 = \tau + t_0 \end{aligned}$$



MF的性质

- 对于频移信号，匹配滤波器不具有适应性。

$$H(\omega) = kS^*(\omega)e^{-j\omega t_0}$$

$$s_1(t) = s(t)e^{jvt} \quad S_1(\omega) = S(\omega + v)$$

$$H_1(\omega) = kS_1^*(\omega)e^{-j\omega t_1} = kS^*(\omega + v)e^{-j\omega t_1}$$

$v \neq 0$ 时， $H_1(\omega)$ 的频率特性与 $H(\omega)$ 的频率特性不同。

MF的性质



● 匹配滤波器与相关器的关系

对于平稳输入信号 $x(t) = s(t) + n(t)$ ，自相关器的输出为：

$$\begin{aligned} r_x(\tau) &= \int_{-\infty}^{\infty} x(t)x(t+\tau)dt \\ &= \int_{-\infty}^{\infty} (s(t) + n(t))(s(t+\tau) + n(t+\tau))dt \\ &= r_s(\tau) + r_n(\tau) + r_{sn}(\tau) + r_{ns}(\tau) \end{aligned}$$

MF的性质



对于平稳输入信号 $x_1(t) = s(t) + n(t)$ 和 $x_2(t) = s_0(t)$ ，互相关器的输出为：

$$\begin{aligned} r_{x_1 x_2}(\tau) &= \int_{-\infty}^{\infty} x_1(t) x_2(t + \tau) dt \\ &= \int_{-\infty}^{\infty} (s(t) + n(t)) s_0(t + \tau) dt \\ &= r_{ss_0}(\tau) + r_{ns_0}(\tau) \end{aligned}$$

MF的性质



假设本地信号为 $s(t)$, 相关器的输入信号 $x_1(t) = s(t) + n(t) \quad 0 \leq t \leq T$

相关器的输出信号为

$$y_c(t) = \int_0^t x(u)s(u)du$$

$$t \geq T \quad y_c(t \geq T) = \int_0^T x(u)s(u)du$$

匹配滤波器的输出信号为 $h(t) = s(T-t)$

$$y_f(t) = \int_0^t x(t-\tau)h(\tau)d\tau = \int_0^t x(t-\tau)s(T-\tau)d\tau$$

$$y_f(t=T) = \int_0^T x(T-\tau)s(T-\tau)d\tau = \int_0^T x(u)s(u)du$$

Relation to Optimal detection



● Minimal Distance Criterion

$$\mathbf{r} = \mathbf{s}_m + \mathbf{w}, m = 1, \dots, M$$

$$\begin{aligned}\hat{\mathbf{s}} &= \arg \min_{\mathbf{s}_m, m=1, \dots, M} \|\mathbf{r} - \mathbf{s}_m\|^2 \\ &= \arg \min_{\mathbf{s}_m, m=1, \dots, M} \left[\|\mathbf{r}\|^2 - 2\langle \mathbf{r}, \mathbf{s}_m \rangle + \|\mathbf{s}_m\|^2 \right] \\ &= \arg \max_{\mathbf{s}_m, m=1, \dots, M} \left[2\langle \mathbf{r}, \mathbf{s}_m \rangle - \|\mathbf{s}_m\|^2 \right] \\ &= \arg \max_{\mathbf{s}_m, m=1, \dots, M} \left[2\langle \mathbf{r}, \mathbf{s}_m \rangle - E_m \right]\end{aligned}$$

Relation to Optimal detection

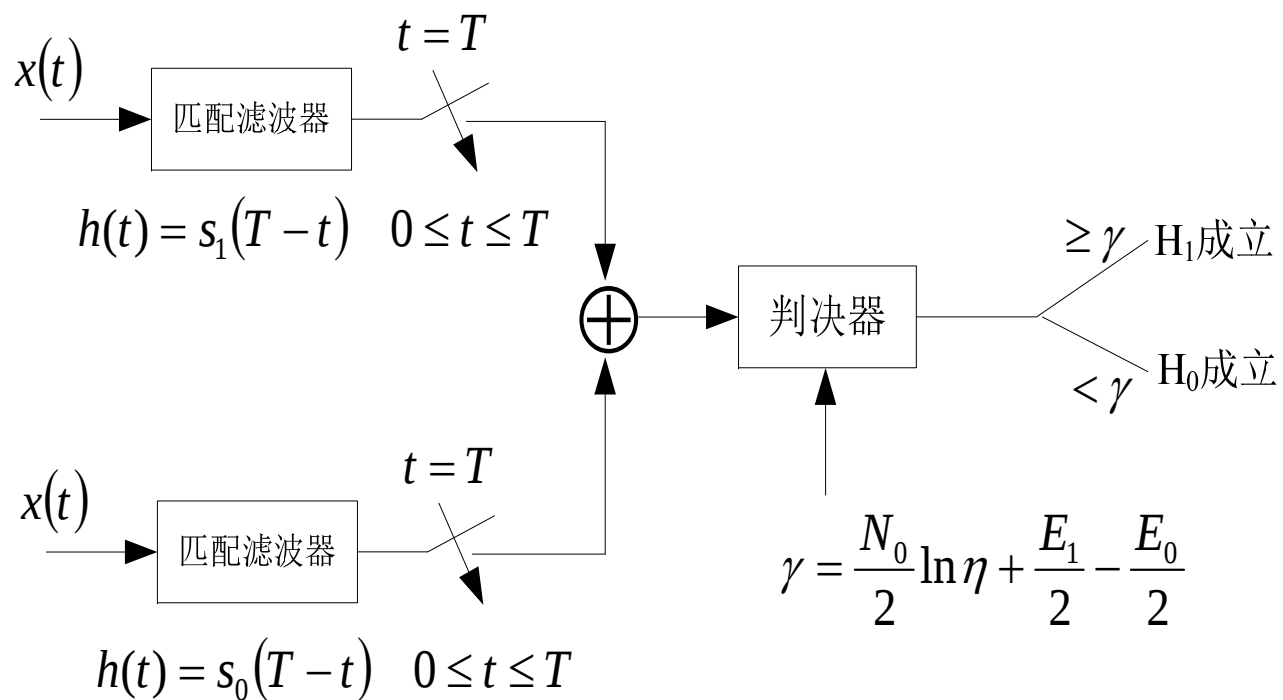


• Equal energy assumption

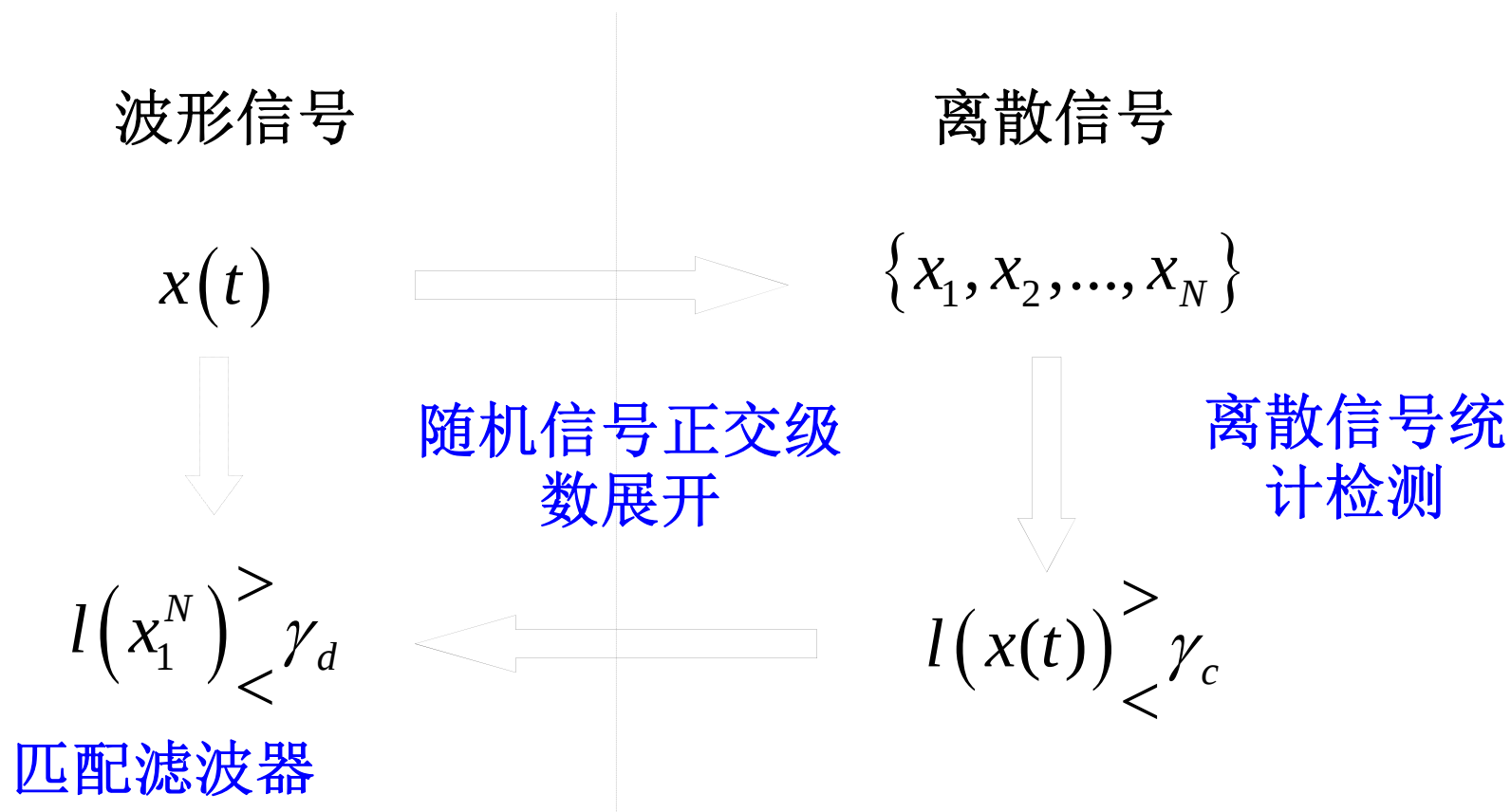
$$\begin{aligned}\hat{s} &= \arg \max_{s_m, m=1, \dots, M} \left[2 \langle \mathbf{r}, \mathbf{s}_m \rangle - E_m \right] \\ &= \arg \max_{s_m, m=1, \dots, M} \langle \mathbf{r}, \mathbf{s}_m \rangle \\ &= \arg \max_{s_m, m=1, \dots, M} \int_0^T r(t) s_m(t) dt\end{aligned}$$

• The minimal distance receiver can be implemented as a bank of matched filter

检测系统结构



波形信号检测方法



Problem Formulation



- 考虑联合PDF $P(x_1^N | H_j)$

如果 $x_k, k=1, \dots, N$ 相互独立, 则有

$$P(x_1^N | H_j) = \prod_{k=1}^N P(x_k | H_j)$$

- 目标: 正交级数展开得到的系数相互独立



完备的正交函数集

若实函数集 $\{f_k(t)\}, k=1,2,\dots$ 在 $(0, T)$ 时间内满足

$$\int_0^T f_k(t) f_j(t) dt = \begin{cases} 1, & j = k \\ 0, & j \neq k \end{cases}$$

且不存在归一化的非零函数 $g(t)$, 满足

$$\int_0^T f_k(t) g(t) dt = 0$$

则称函数集 $\{f_k(t)\}, k=1,2,\dots$ 是完备正交函数集。



确知信号的正交级数展开

$s(t)$ 是定义在 $(0, T)$ 时间内确知信号，且

$$E_s < \infty$$

该信号可用正交级数展开表示为：

$$s(t) = \lim_{N \rightarrow \infty} \sum_{k=1}^N s_k f_k(t)$$

$$s_k = \int_0^T f_k(t) s(t) dt$$

随机过程的正交级数展开



- 假设接收信号为

$$x(t) = s(t) + n(t)$$

其中 $s(t)$ 是确知信号， $n(t)$ 是零均值、自相关函数为 $R_n(\tau)$ 的平稳随机过程，则接收信号也是平稳随机过程。

- 由于随机过程是由很多样本函数构成的集合，而每个样本函数是时间的函数，所以对给定的样本函数，可以进行正交级数展开

$$x(t) = \lim_{N \rightarrow \infty} \sum_{k=1}^N x_k f_k(t) \quad x_k = \int_0^T f_k(t) x(t) dt$$

所有样本函数的展开系数，构成了一族随机变量。

随机过程的正交级数展开



- 目标：给出一种正交函数集的选择方法，以保证展开系数(高斯分布)之间是相互独立（互不相关）的随机变量。

假设随机过程为 $x(t) = s(t) + n(t)$ 正交函数集 $\{f_k(t)\}, k = 1, 2, \dots$

$x(t) = s(t) + n(t)$ 的展开系数是随机变量，且

$$x_k = \int_0^T f_k(t)x(t)dt$$

$$\begin{aligned} E[x_k] &= E\left[\int_0^T f_k(t)x(t)dt\right] = E\left[\int_0^T f_k(t)(s(t) + n(t))dt\right] \\ &= s_k \end{aligned}$$

随机过程的卡亨南-洛维展开



$$\begin{aligned} E[(x_k - s_k)(x_j - s_j)] &= E\left[\left(\int_0^T f_k(t)x(t)dt - s_k\right)\left(\int_0^T f_j(t)x(t)dt - s_j\right)\right] \\ &= E\left[\left(\int_0^T f_k(t)(s(t) + n(t))dt - s_k\right)\left(\int_0^T f_j(t)(s(t) + n(t))dt - s_j\right)\right] \\ &= E\left[\left(\int_0^T f_k(t)n(t)dt\right)\left(\int_0^T f_j(t)n(t)dt\right)\right] \\ &= E\left[\left(\int_0^T f_k(t)n(t)dt\right)\left(\int_0^T f_j(u)n(u)du\right)\right] \\ &= \int_0^T f_k(t)\left[\left(\int_0^T E(n(t)n(u))f_j(u)du\right)\right]dt \end{aligned}$$

随机过程的卡亨南-洛维展开



$$\begin{aligned} E\left[(x_k - s_k)(x_j - s_j)\right] &= \int_0^T f_k(t) \left[\left(\int_0^T E(n(t)n(u)) f_j(u) du \right) \right] dt \\ &= \int_0^T f_k(t) \left[\left(\int_0^T R_n(t-u) f_j(u) du \right) \right] dt \end{aligned}$$

为保证 $E\left[(x_k - s_k)(x_j - s_j)\right] = \lambda_j \delta_{kj}$

$$\int_0^T R_n(t-u) f_j(u) du = \lambda_j f_j(t)$$

白噪声情况下正交函数集的任意性



- 白噪声条件下

$$R_n(t-u) = \frac{N_0}{2} \delta(t-u)$$

$$\begin{aligned} E[(x_k - s_k)(x_j - s_j)] &= \int_0^T f_k(t) \left[\left(\int_0^T R_n(t-u) f_j(u) du \right) \right] dt \\ &= \int_0^T f_k(t) \left[\left(\int_0^T \frac{N_0}{2} \delta(t-u) f_j(u) du \right) \right] dt \\ &= \frac{N_0}{2} \int_0^T f_k(t) f_j(t) dt = \frac{N_0}{2} \delta_{kj} \end{aligned}$$

- 在白噪声条件下，可任意选取正交函数集，均可保证展开系数之间是不相关的。

一般二元信号的波形检测



● 信号模型

在一般二元信号的波形检测中，假设H0下和假设H1的接收信号分别为：

$$H_0: x(t) = s_0(t) + n(t), \quad 0 \leq t \leq T$$

$$H_1: x(t) = s_1(t) + n(t), \quad 0 \leq t \leq T$$

其中 $s_0(t)$ 是能量为 E_0 的确知信号， $s_1(t)$ 是能量为 E_1 的确知信号
 $n(t)$ 是均值为零，功率谱密度为 $N_0/2$ 的高斯白噪声。

检测方法概述



- 首先，利用随机过程的正交级数展开，将随机过程用一组随机变量来表示；
- 然后，针对展开得到的随机变量，利用第一章的统计检测方法，构建贝叶斯检测表达式；
- 最后，利用展开系数与随机过程之间的表示关系，构建波形信号的检测表达式。

判决表达式



$$H_0: x(t) = s_0(t) + n(t), \quad 0 \leq t \leq T$$

$$H_1: x(t) = s_1(t) + n(t), \quad 0 \leq t \leq T$$

步骤1, 选一组完备的正交函数集 $\{f_k(t), k = 1, 2, \dots\}$

对接收信号进行正交级数展开, 得到一组随机变量

$$x_k, k = 1, 2, \dots$$

$$x_k = \int_0^T f_k(t)x(t)dt \quad x(t) = \lim_{N \rightarrow \infty} \sum_{i=1}^N x_k f_k(t)$$

$$H_0: x_k = s_{0k} + n_k, \quad k = 1, 2, \dots \quad n_k = \int_0^T f_k(t)n(t)dt$$

$$H_1: x_k = s_{1k} + n_k, \quad k = 1, 2, \dots \quad s_{ik} = \int_0^T f_k(t)s_i(t)dt, \quad i = 0, 1$$

判决表达式



x_k 是高斯随机过程积分的结果，因而 x_k 服从高斯分布，且有

$$\begin{aligned} E[x_k | H_0] &= E\left[\int_0^T x(t) f_k(t) dt | H_0\right] = E\left[\int_0^T (s_0(t) + n(t)) f_k(t) dt\right] \\ &= \int_0^T E[n(t)] f_k(t) dt + \int_0^T s_0(t) f_k(t) dt = s_{0k} \end{aligned}$$

$$\text{Var}[x_k | H_0] = E[n_k^2] = E\left[\int_0^T n(t) f_k(t) dt \int_0^T n(u) f_k(u) du\right]$$

$$= \int_0^T f_k(t) \int_0^T E[n(t)n(u)] f_k(u) du dt = \frac{N_0}{2}$$

判决表达式



$$\begin{aligned} E[x_k|H_1] &= E\left[\int_0^T x(t)f_k(t)dt|H_1\right] = E\left[\int_0^T (s_1(t) + n(t))f_k(t)dt\right] \\ &= s_{1k} + \int_0^T E[n(t)]f_k(t)dt = s_{1k} \end{aligned}$$

$$\text{Var}[x_k|H_1] = \frac{N_0}{2}$$

$$p(x_k|H_0) = \frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{(x_k - s_{0k})^2}{N_0}\right) \quad k = 1, 2, \dots$$

$$p(x_k|H_1) = \frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{(x_k - s_{1k})^2}{N_0}\right) \quad k = 1, 2, \dots$$

判决表达式



步骤2，利用前N项展开系数，构建似然比检验

由于信道是加性高斯白噪声，由卡亨南-洛维展开可知，各展开系数是不相关的，因而也是相互独立的。

$$p(\mathbf{x}_N|H_0) = \prod_{k=1}^N \frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{(x_k - s_{0k})^2}{N_0}\right)$$

$$p(\mathbf{x}_N|H_1) = \prod_{k=1}^N \frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{(x_k - s_{1k})^2}{N_0}\right)$$

$$\mathbf{x}_N = (x_1, x_2, \dots, x_N)^T$$

判决表达式



由贝叶斯检测准则，得到

$$\Lambda(\mathbf{x}_N) = \frac{p(\mathbf{x}_N|H_1)}{p(\mathbf{x}_N|H_0)} \underset{H_0}{\overset{H_1}{\gtrless}} \eta \quad \frac{\prod_{k=1}^N \frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{(x_k - s_{1k})^2}{N_0}\right)}{\prod_{k=1}^N \frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{(x_k - s_{0k})^2}{N_0}\right)} \underset{H_0}{\overset{H_1}{\gtrless}} \eta$$

$$\exp\left(\sum_{k=1}^N \frac{(x_k - s_{0k})^2 - (x_k - s_{1k})^2}{N_0}\right) \underset{H_0}{\overset{H_1}{\gtrless}} \eta$$

$$\frac{2}{N_0} \sum_{k=1}^N x_k s_{1k} - \frac{2}{N_0} \sum_{k=1}^N x_k s_{0k} + \frac{1}{N_0} \sum_{k=1}^N s_{0k}^2 - \frac{1}{N_0} \sum_{k=1}^N s_{1k}^2 \underset{H_0}{\overset{H_1}{\gtrless}} \ln \eta$$

判决表达式



步骤3, 令 $N \rightarrow \infty$

$$\lim_{N \rightarrow \infty} \left(\frac{2}{N_0} \sum_{k=1}^N x_k s_{1k} - \frac{2}{N_0} \sum_{k=1}^N x_k s_{0k} + \frac{1}{N_0} \sum_{k=1}^N s_{0k}^2 - \frac{1}{N_0} \sum_{k=1}^N s_{1k}^2 \right) \stackrel{\text{def}}{=} \ln \Lambda(x(t))$$

$$\ln \Lambda(x(t)) \underset{H_0}{\overset{H_1}{\gtrless}} \ln \eta$$

$$\begin{aligned} \frac{2}{N_0} \lim_{N \rightarrow \infty} \sum_{k=1}^N x_k s_{ik} &= \frac{2}{N_0} \lim_{N \rightarrow \infty} \sum_{k=1}^N x_k \int_0^T s_i(t) f_k(t) dt \\ &= \frac{2}{N_0} \int_0^T s_i(t) \lim_{N \rightarrow \infty} \sum_{k=1}^N x_k f_k(t) dt = \frac{2}{N_0} \int_0^T s_i(t) x(t) dt \end{aligned}$$

判决表达式



$$\frac{1}{N_0} \lim_{N \rightarrow \infty} \sum_{k=1}^N s_{0k}^2 = \frac{1}{N_0} \int_0^T s_0^2(t) dt = \frac{E_0}{N_0}$$

$$\frac{1}{N_0} \lim_{N \rightarrow \infty} \sum_{k=1}^N s_{1k}^2 = \frac{1}{N_0} \int_0^T s_1^2(t) dt = \frac{E_1}{N_0}$$

$$\lim_{N \rightarrow \infty} \left(\frac{2}{N_0} \sum_{k=1}^N x_k s_{1k} - \frac{2}{N_0} \sum_{k=1}^N x_k s_{0k} + \frac{1}{N_0} \sum_{k=1}^N s_{0k}^2 - \frac{1}{N_0} \sum_{k=1}^N s_{1k}^2 \right) \stackrel{def}{=} \ln \Lambda(x(t))$$

$$= \frac{2}{N_0} \int_0^T s_1(t)x(t)dt - \frac{2}{N_0} \int_0^T s_0(t)x(t)dt + \frac{1}{N_0} E_0 - \frac{1}{N_0} E_1$$

判决表达式



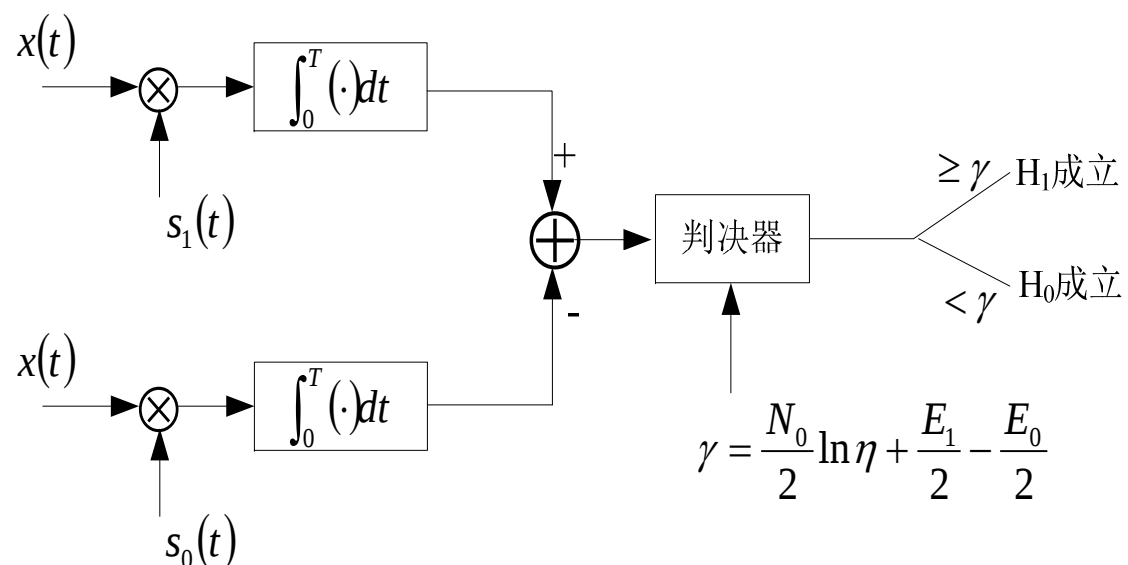
$$\frac{2}{N_0} \int_0^T s_1(t)x(t)dt - \frac{2}{N_0} \int_0^T s_0(t)x(t)dt + \frac{E_0}{N_0} - \frac{E_1}{N_0} \underset{H_0}{\overset{H_1}{\gtrless}} \ln \eta$$

$$\int_0^T s_1(t)x(t)dt - \int_0^T s_0(t)x(t)dt \underset{H_0}{\overset{H_1}{\gtrless}} \frac{N_0}{2} \ln \eta + \frac{E_1}{2} - \frac{E_0}{2}$$

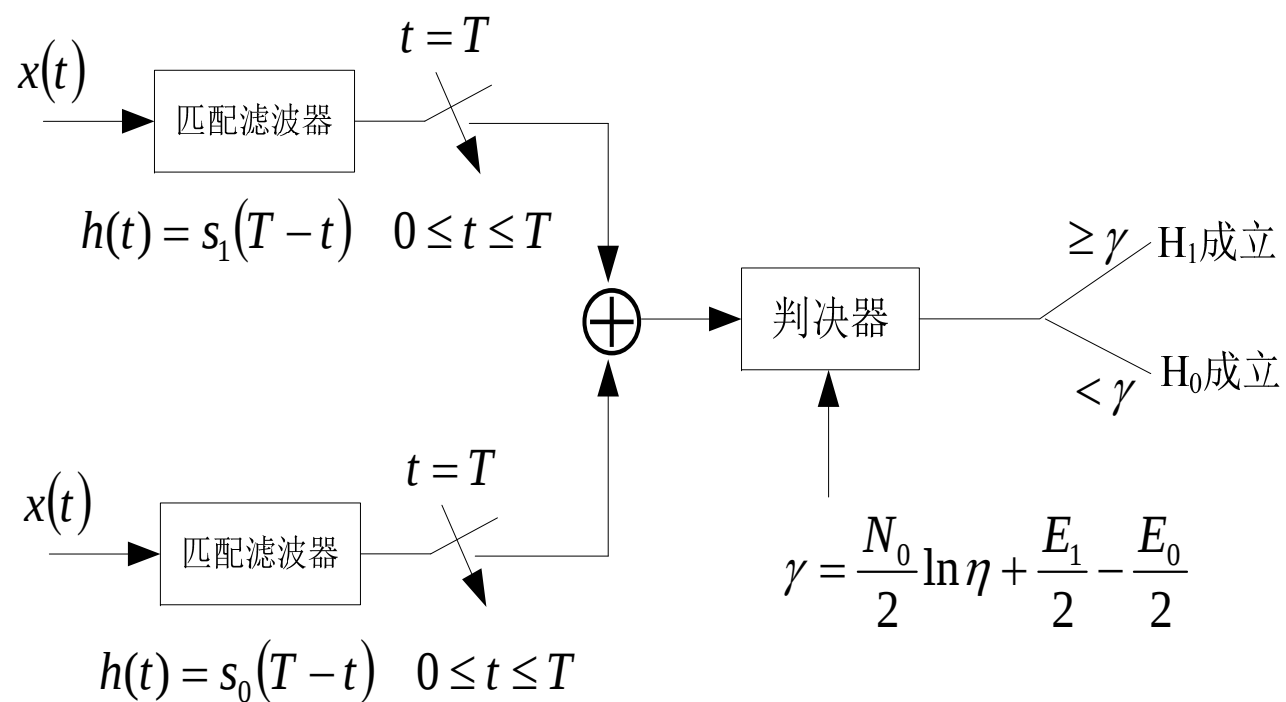
检测系统结构



$$\int_0^T s_1(t)x(t)dt - \int_0^T s_0(t)x(t)dt \underset{H_0}{\overset{H_1}{\gtrless}} \frac{N_0}{2} \ln \eta + \frac{E_1}{2} - \frac{E_0}{2}$$



检测系统结构



检测性能分析



判决表达式

$$\int_0^T s_1(t)x(t)dt - \int_0^T s_0(t)x(t)dt \underset{H_0}{\overset{H_1}{\gtrless}} \frac{N_0}{2} \ln \eta + \frac{E_1}{2} - \frac{E_0}{2}$$

充分统计量 $l = \int_0^T s_1(t)x(t)dt - \int_0^T s_0(t)x(t)dt$ 门限 $\gamma = \frac{N_0}{2} \ln \eta + \frac{E_1}{2} - \frac{E_0}{2}$

$$l \underset{H_0}{\overset{H_1}{\gtrless}} \gamma$$

$$P(H_1|H_0) = \int_{\gamma}^{\infty} p(l|H_0)dl \quad P(H_0|H_1) = \int_{-\infty}^{\gamma} p(l|H_1)dl$$

由于接收信号 $x(t)$ 是以高斯随机过程，所以统计量 l 为服从高斯分布的随机变量



检测性能分析

定义统计量

$$l = \int_0^T s_1(t)x(t)dt - \int_0^T s_0(t)x(t)dt$$

$$\begin{aligned} E[l|H_0] &= E\left[\int_0^T x(t)s_1(t)dt - \int_0^T x(t)s_0(t)dt|H_0\right] \\ &= E\left[\int_0^T (s_0(t) + n(t))s_1(t)dt - \int_0^T (s_0(t) + n(t))s_0(t)dt|H_0\right] \\ &= \int_0^T s_0(t)s_1(t)dt - \int_0^T s_0^2(t)dt = \rho\sqrt{E_1E_0} - E_0 \end{aligned}$$

波形相关系数

$$\rho \stackrel{\text{def}}{=} \frac{|\langle s_0(t), s_1(t) \rangle|}{\|s_0(t)\| \|s_1(t)\|} = \frac{1}{\sqrt{E_1E_0}} \int_0^T s_0(t)s_1(t)dt; \quad |\rho| \leq 1$$

检测性能分析



$$\begin{aligned} \text{Var}[l|H_0] &= E\left\{\left[(l|H_0) - E[(l|H_0)]\right]^2\right\} \\ &= E\left[\left(\int_0^T n(t)s_1(t)dt - \int_0^T n(t)s_0(t)dt\right)^2\right] \\ &= E\left[\left(\int_0^T n(t)s_1(t)dt\right)^2\right] + E\left[\left(\int_0^T n(t)s_0(t)dt\right)^2\right] \\ &\quad - 2E\left[\int_0^T n(t)s_1(t)dt \int_0^T n(u)s_0(u)du\right] \end{aligned}$$

检测性能分析



$$\begin{aligned} & E \left[\left(\int_0^T n(t) s_1(t) dt \right)^2 \right] \\ &= E \left[\int_0^T n(t) s_1(t) dt \int_0^T n(u) s_1(u) du \right] \\ &= \int_0^T s_1(t) \int_0^T E[n(t)n(u)] s_1(u) du dt \\ &= \int_0^T s_1(t) \int_0^T \frac{N_0}{2} \delta(t-u) s_1(u) du dt \\ &= \frac{N_0}{2} \int_0^T s_1(t) s_1(t) dt = \frac{N_0 E_1}{2} \end{aligned}$$

检测性能分析



$$\begin{aligned} & E \left[\int_0^T n(t) s_1(t) dt \int_0^T n(u) s_0(u) du \right] \\ &= \int_0^T s_1(t) \int_0^T E[n(t)n(u)] s_0(u) du dt \\ &= \int_0^T s_1(t) \int_0^T \frac{N_0}{2} \delta(t-u) s_0(u) du dt \\ &= \frac{N_0}{2} \int_0^T s_1(t) s_0(t) dt = \frac{N_0}{2} \rho \sqrt{E_1 E_0} \end{aligned}$$

检测性能分析



$$\begin{aligned} \text{Var}[l|H_0] &= E\left[\left(\int_0^T n(t)s_1(t)dt\right)^2\right] + E\left[\left(\int_0^T n(t)s_0(t)dt\right)^2\right] \\ &\quad - 2E\left[\int_0^T n(t)s_1(t)dt \int_0^T n(u)s_0(u)du\right] \\ &= \frac{N_0 E_1}{2} + \frac{N_0 E_0}{2} - 2\frac{N_0}{2}\rho\sqrt{E_1 E_0} \\ &= \frac{N_0}{2}(E_1 + E_0 - 2\rho\sqrt{E_1 E_0}) \end{aligned}$$

检测性能分析



$$E[l|H_1] = E_1 - \rho\sqrt{E_1E_0}$$

$$\text{Var}[l|H_1] = \frac{N_0}{2} (E_1 + E_0 - 2\rho\sqrt{E_1E_0})$$

$$\begin{aligned} d^2 &\stackrel{\text{def}}{=} \frac{(E(l|H_1) - E(l|H_0))^2}{\text{Var}(l|H_0)} = \frac{2}{N_0} \frac{(E_1 - \rho\sqrt{E_1E_0} - \rho\sqrt{E_1E_0} + E_0)^2}{E_1 + E_0 - 2\rho\sqrt{E_1E_0}} \\ &= \frac{2}{N_0} (E_1 + E_0 - 2\rho\sqrt{E_1E_0}) \end{aligned}$$



检测性能分析

$$P(H_1|H_0) = \int_{\gamma}^{\infty} p(l|H_0)dl = Q\left(\frac{\ln \eta}{d} + \frac{d}{2}\right)$$

$$P(H_0|H_0) = 1 - Q\left(\frac{\ln \eta}{d} + \frac{d}{2}\right)$$

$$P(H_1|H_1) = \int_{\gamma}^{\infty} p(l|H_1)dl = Q\left(\frac{\ln \eta}{d} - \frac{d}{2}\right)$$

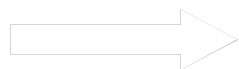
$$P(H_0|H_1) = 1 - Q\left(\frac{\ln \eta}{d} - \frac{d}{2}\right)$$

最佳信号波形设计



Problem

$$\begin{cases} \max d^2 = \frac{\left(E(l|H_1) - E(l|H_0)\right)^2}{\text{Var}(l|H_0)} = \frac{2}{N_0} \left(E_1 + E_0 - 2\rho\sqrt{E_1E_0}\right) \\ \text{s.t. } E_0 + E_1 \leq 2E_S \end{cases}$$



Solution

$$d^2 = \frac{8E_S}{N_0} \quad E_0 = E_1 = E_S \quad \rho = -1$$

$$s_0(t) = -s_1(t)$$

充分统计量法



正交级数展开法：信道噪声是白噪声，正交函数集可任意选取。

充分统计量法：选取特定的正交函数集，使得有关发送信号的信息只包含在有限的展开系数中。



充分统计量法

$$H_0: x(t) = s_0(t) + n(t), \quad 0 \leq t \leq T$$

$$H_1: x(t) = s_1(t) + n(t), \quad 0 \leq t \leq T$$

$$\rho = \frac{1}{\sqrt{E_0 E_1}} \int_0^T s_0(t) s_1(t) dt$$



充分统计量法

步骤1, 选择一组完备正交函数集 $\{f_k(t), k = 1, 2, \dots\}$

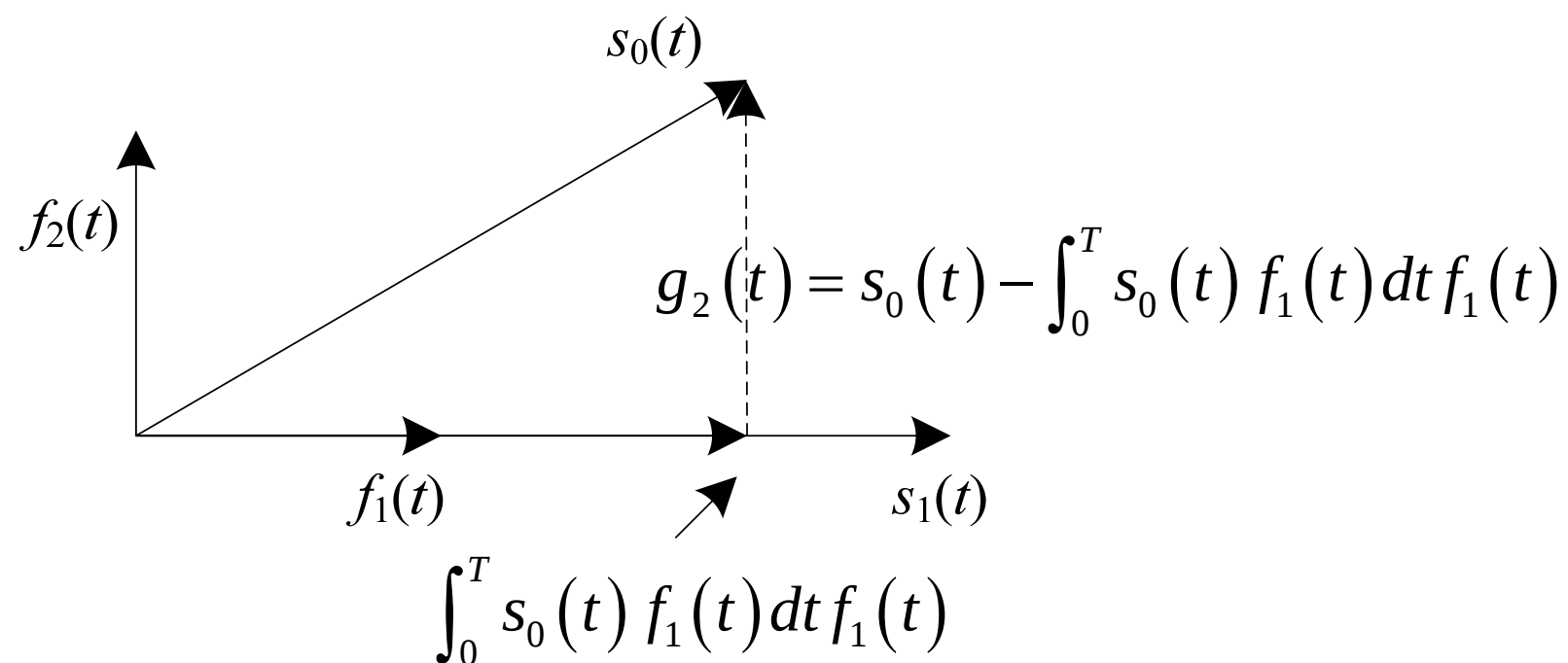
满足以下条件:

$$f_1(t) = \frac{1}{\sqrt{E_1}} s_1(t) \quad g_2(t) = s_0(t) - \int_0^T s_0(t) f_1(t) dt f_1(t) \\ = s_0(t) - \rho \sqrt{E_0} f_1(t)$$

$$f_2(t) = \frac{g_2(t)}{\sqrt{\int_0^T g_2^2(t) dt}} = \frac{1}{\sqrt{(1-\rho^2)E_0}} \left(s_0(t) - \rho \sqrt{\frac{E_0}{E_1}} s_1(t) \right)$$

$$\int_0^T g_2^2(t) dt = \int_0^T \left(s_0(t) - \rho \sqrt{E_0} f_1(t) \right)^2 dt \\ = \int_0^T \left(s_0(t) \right)^2 dt + \rho^2 E_0 \int_0^T \left(f_1(t) \right)^2 dt - 2\rho \sqrt{E_0} \int_0^T s_0(t) f_1(t) dt \\ = E_0 + \rho^2 E_0 - 2\rho \sqrt{E_0} \times \rho \sqrt{E_0} \\ = E_0 - \rho^2 E_0$$

充分统计量法



充分统计量法



$f_k(t)$ 是分别与 $f_1(t)$ 和 $f_2(t)$ 正交的单位能量函数, $k \geq 3$
任意两个函数 $f_k(t)$ 和 $f_j(t)$ 是正交的, $k \geq 1, j \geq 1, k \neq j$

步骤2, 利用选择的正交函数集 $\{f_k(t), k = 1, 2, \dots\}$
对接收信号进行正交级数展开。

$$\begin{aligned} H_0 : \quad x_1 &= \int_0^T x(t) f_1(t) dt \\ &= \int_0^T (s_0(t) + n(t)) \frac{1}{\sqrt{E_1}} s_1(t) dt \\ &= \rho \sqrt{E_0} + n_1 \end{aligned}$$



充分统计量法

$$\begin{aligned}x_2 &= \int_0^T x(t) f_2(t) dt = \int_0^T (s_0(t) + n(t)) f_2(t) dt \\&= \int_0^T (s_0(t) + n(t)) \frac{1}{\sqrt{(1-\rho^2)E_0}} \left(s_0(t) - \rho \sqrt{\frac{E_0}{E_1}} s_1(t) \right) dt \\&= \frac{1}{\sqrt{(1-\rho^2)E_0}} \int_0^T (s_0(t) + n(t)) \left(s_0(t) - \rho \sqrt{\frac{E_0}{E_1}} s_1(t) \right) dt \\&= \frac{1}{\sqrt{(1-\rho^2)E_0}} \left[E_0 - \rho \sqrt{\frac{E_0}{E_1}} \rho \sqrt{E_1 E_0} \right] + n_2 \\&= \sqrt{(1-\rho^2)E_0} + n_2\end{aligned}$$

$$x_k = \int_0^T x(t) f_k(t) dt = n_k \quad k \geq 3$$

充分统计量法



$$\begin{aligned} H_1: \quad x_1 &= \int_0^T x(t) f_1(t) dt = \int_0^T (s_1(t) + n(t)) \frac{1}{\sqrt{E_1}} s_1(t) dt \\ &= \sqrt{E_1} + n_1 \end{aligned}$$

$$x_2 = \int_0^T x(t) f_2(t) dt = \int_0^T (s_1(t) + n(t)) f_2(t) dt = n_2$$

$$x_k = \int_0^T x(t) f_k(t) dt = n_k \quad k \geq 3$$

充分统计量法



步骤3, 利用得到的展开系数, 构建似然比表达式

$$E[x_1|H_0] = E[\rho\sqrt{E_0} + n_1] = \rho\sqrt{E_0}$$

$$E[x_2|H_0] = E[\sqrt{(1-\rho^2)E_0} + n_2] = \sqrt{(1-\rho^2)E_0}$$

$$E[x_k|H_0] = E[n_k] = 0 \quad k \geq 3 \quad \text{Var}[x_k|H_0] = E[n_k^2] = \frac{N_0}{2}$$

$$E[x_1|H_1] = E[\sqrt{E_1} + n_1] = \sqrt{E_1} \quad E[x_k|H_1] = E[n_k] = 0 \quad k \geq 2$$

$$\text{Var}[x_k|H_1] = E[n_k^2] = \frac{N_0}{2}$$



充分统计量法

$$\mathbf{x} = (x_1, x_2, \dots, x_N, \dots)^T$$

当k大于2时，展开系数 x_k 是仅与信道噪声有关的随机变量，

与发送信号检测无关，因此可以利用前两个展开系数构建贝叶斯检测，即

$$\Lambda(\mathbf{x}) = \frac{p(\mathbf{x}|H_1)}{p(\mathbf{x}|H_0)} \underset{H_0}{\overset{H_1}{\gtrless}} \eta$$



$$\Lambda(\mathbf{x}) = \frac{p(x_1, x_2|H_1)}{p(x_1, x_2|H_0)} \underset{H_0}{\overset{H_1}{\gtrless}} \eta$$

$$\frac{\frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{(x_1 - \sqrt{E_1})^2}{N_0}\right) \frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{x_2^2}{N_0}\right)}{\frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{(x_1 - \rho\sqrt{E_0})^2}{N_0}\right) \frac{1}{\sqrt{\pi N_0}} \exp\left(-\frac{(x_2 - \sqrt{(1-\rho^2)E_0})^2}{N_0}\right)} \underset{H_0}{\overset{H_1}{\gtrless}} \eta$$

充分统计量法



两边取对数，并化简，得

$$\exp\left(\frac{1}{N_0}\left[\left(x_1 - \rho\sqrt{E_0}\right)^2 + \left(x_2 - \sqrt{(1-\rho^2)E_0}\right)^2\right] - \left[\left(x_1 - \sqrt{E_1}\right)^2 + x_2^2\right]\right) \underset{H_0}{\overset{H_1}{>}} \eta$$

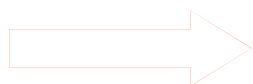
$$\frac{1}{N_0}\left[2\sqrt{E_1}x_1 - 2\rho\sqrt{E_0}x_1 - 2\sqrt{(1-\rho^2)E_0}x_2 - E_1 + E_0\right] \underset{H_0}{\overset{H_1}{>}} \ln \eta$$

$$\left(\sqrt{E_1} - \rho\sqrt{E_0}\right)x_1 - \sqrt{(1-\rho^2)E_0}x_2 \underset{H_0}{\overset{H_1}{>}} \frac{N_0}{2}\ln \eta + \frac{1}{2}(E_1 - E_0)$$

充分统计量法



$$\begin{aligned} l[x(t)] &= \left(\sqrt{E_1} - \rho \sqrt{E_0} \right) x_1 - \sqrt{(1 - \rho^2) E_0} x_2 \\ &= \left(\sqrt{E_1} - \rho \sqrt{E_0} \right) \int_0^T x(t) \frac{1}{\sqrt{E_1}} s_1(t) dt \\ &\quad - \sqrt{(1 - \rho^2) E_0} \int_0^T x(t) \frac{1}{\sqrt{(1 - \rho^2) E_0}} \left(s_0(t) - \rho \sqrt{\frac{E_0}{E_1}} s_1(t) \right) dt \\ &= \left(\sqrt{E_1} - \rho \sqrt{E_0} \right) \int_0^T x(t) \frac{1}{\sqrt{E_1}} s_1(t) dt - \int_0^T x(t) \left(s_0(t) - \rho \sqrt{\frac{E_0}{E_1}} s_1(t) \right) dt \\ &= \int_0^T x(t) s_1(t) dt - \int_0^T x(t) s_0(t) dt \end{aligned}$$



$$\int_0^T s_1(t) x(t) dt - \int_0^T s_0(t) x(t) dt \underset{H_0}{\overset{H_1}{\gtrless}} \frac{N_0}{2} \ln \eta + \frac{E_1}{2} - \frac{E_0}{2}$$

例1



考虑发送信号周期为 $T = 2\pi/\omega_0$ 的二元移频键控系统,在假设 H_1 和 H_0 下的发送信号分别为:

$$H_0: x(t) = a \sin \omega_0 t + n(t), \quad 0 \leq t \leq T$$

$$H_1: x(t) = a \sin 2\omega_0 t + n(t), \quad 0 \leq t \leq T$$

其中, 信号的振幅和频率已知,并假定两个假设先验等概.信号在传输中叠加了均值为零,功率谱密度为 $N_0/2$ 的高斯白噪声.现采用最小平均错误概率准则, 设计信号检测系统,并计算平均错误概率.



解： 根据题设，得到两个信号的能量分别为

$$E_0 = \int_0^T (a \sin \omega_0 t)^2 dt = \frac{a^2 T}{2}$$

$$E_1 = \int_0^T (a \sin 2\omega_0 t)^2 dt = \frac{a^2 T}{2}$$

由于两个假设先验等概，因此在最小平均错误概率准则下，判决门限 $\eta = 1$
利用一般二元信号检测波形判决表达式，得

$$\int_0^T s_1(t)x(t)dt - \int_0^T s_0(t)x(t)dt \underset{H_0}{\overset{H_1}{\gtrless}} \frac{N_0}{2} \ln \eta + \frac{E_1}{2} \underset{H_1}{\overset{H_0}{-}} \frac{E_0}{2} = 0$$
$$\int_0^T x(t)a \sin 2\omega_0 t dt - \int_0^T x(t)a \sin \omega_0 t dt \underset{H_0}{\gtrless} 0$$



为求平均错误概率，首先需要计算偏移系数

$$d^2 \stackrel{\text{def}}{=} \frac{(E(l|H_1) - E(l|H_0))^2}{\text{Var}(l|H_0)}$$

$$l = \int_0^T x(t) a \sin 2\omega_0 t dt - \int_0^T x(t) a \sin \omega_0 t dt$$

$$E[l|H_0] = E\left[\int_0^T x(t) a \sin 2\omega_0 t dt - \int_0^T x(t) a \sin \omega_0 t dt | H_0\right]$$

$$\begin{aligned} &= E\left[\int_0^T (a \sin \omega_0 t + n(t)) a \sin 2\omega_0 t dt - \int_0^T (a \sin \omega_0 t + n(t)) a \sin \omega_0 t dt\right] \\ &= -\frac{a^2 T}{2} = -E_0 = -E_s \end{aligned}$$

$$\sin \alpha \sin \beta = -\frac{1}{2}(\cos(\alpha + \beta) - \cos(\alpha - \beta))$$



$$\begin{aligned} \text{Var}[l|H_0] &= E \left\{ \left[(l|H_0) - E[(l|H_0)] \right]^2 \right\} \\ &= E \left[\left(\int_0^T n(t) a \sin 2\omega_0 t dt - \int_0^T n(t) a \sin \omega_0 t dt \right)^2 \right] \\ &= E \left[\left(\int_0^T n(t) a \sin 2\omega_0 t dt \right)^2 \right] + E \left[\left(\int_0^T n(t) a \sin \omega_0 t dt \right)^2 \right] \\ &\quad - 2E \left[\int_0^T n(t) a \sin 2\omega_0 t dt \int_0^T n(u) a \sin \omega_0 u du \right] \\ &= \frac{N_0 E_1}{2} + \frac{N_0 E_0}{2} = N_0 E_s \end{aligned}$$



$$\begin{aligned} E[l|H_1] &= E\left[\int_0^T x(t)a \sin 2\omega_0 t dt - \int_0^T x(t)a \sin \omega_0 t dt | H_1\right] \\ &= E\left[\int_0^T (a \sin 2\omega_0 t + n(t))a \sin 2\omega_0 t dt - \int_0^T (a \sin 2\omega_0 t + n(t))a \sin \omega_0 t dt\right] \\ &= E_1 = E_s \end{aligned}$$

$$\begin{aligned} d^2 &\stackrel{\text{def}}{=} \frac{(E(l|H_1) - E(l|H_0))^2}{\text{Var}(l|H_0)} = \frac{(E_s - (-E_s))^2}{N_0 E_s} \\ &= \frac{4E_s}{N_0} \end{aligned}$$



$$P(H_1|H_0) = \int_{\gamma}^{\infty} p(l|H_0)dl = Q\left(\frac{\ln \eta}{d} + \frac{d}{2}\right) = Q\left(\frac{d}{2}\right) = Q\left(\sqrt{\frac{E_s}{N_0}}\right)$$

$$P(H_1|H_1) = \int_{\gamma}^{\infty} p(l|H_1)dl = Q\left(\frac{\ln \eta}{d} - \frac{d}{2}\right)$$

$$\begin{aligned} P(H_0|H_1) &= 1 - Q\left(\frac{\ln \eta}{d} - \frac{d}{2}\right) = 1 - Q\left(-\frac{d}{2}\right) = 1 - Q\left(-\sqrt{\frac{E_s}{N_0}}\right) \\ &= Q\left(\sqrt{\frac{E_s}{N_0}}\right) \end{aligned}$$

$$P_e = P(H_0)P(H_1|H_0) + P(H_1)P(H_0|H_1) = Q\left(\sqrt{\frac{E_s}{N_0}}\right)$$

例2

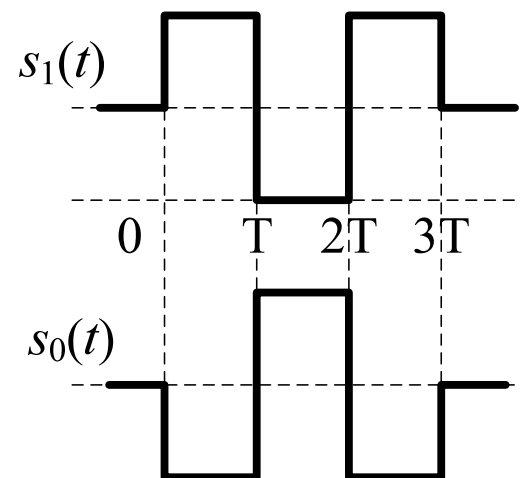


- 在二元数字通信系统中，两个假设下的接收信号分别为

$$\begin{cases} H_0 : x(t) = s_0(t) + n(t), 0 \leq t \leq 3T \\ H_1 : x(t) = s_1(t) + n(t), 0 \leq t \leq 3T \end{cases}$$

其中，信号 $s_1(t)$ 和 $s_0(t)$ 波形如右图所示，加性噪声是均值为0，功率谱密度为 $N_0/2$ 的AWGN；设信号先验等概，采用MAEP求 $E_s/N_0=2$ 时的平均错误概率 P_e ，其中 E_s 是平均能量，即

$$E_s = \frac{1}{2} \left[\int_0^{3T} s_0^2(t) dt + \int_0^{3T} s_1^2(t) dt \right]$$





解： 根据题设，得到两个信号的能量分别为

$$E_0 = \int_0^{3T} (s_0(t))^2 dt \quad E_1 = \int_0^{3T} (s_1(t))^2 dt$$

由于两个假设先验等概，因此在最小平均错误概率准则下，判决门限 $\eta = 1$

利用一般二元信号检测波形判决表达式，得

$$\int_0^{3T} s_1(t)x(t)dt - \int_0^{3T} s_0(t)x(t)dt \underset{H_0}{\overset{H_1}{\gtrless}} \frac{N_0}{2} \ln \eta + \frac{E_1}{2} - \frac{E_0}{2}$$

由于 $s_1(t) = -s_0(t)$

所以 $E_0 = E_1 = E_s$

$$\int_0^{3T} s_1(t)x(t)dt \underset{H_0}{\overset{H_1}{\gtrless}} 0$$



为求平均错误概率，首先需要计算偏移系数

$$d^2 \stackrel{\text{def}}{=} \frac{\left(E(l|H_1) - E(l|H_0)\right)^2}{\text{Var}(l|H_0)}$$

在两种假设下，统计量 $l = \int_0^{3T} x(t)s_1(t)dt$ 均是高斯随机变量，因此有

$$\begin{aligned} E[l|H_0] &= E\left[\int_0^{3T} x(t)s_1(t)dt|H_0\right] \\ &= E\left[\int_0^{3T} (s_0(t) + n(t))s_1(t)dt\right] \\ &= -E_s \end{aligned}$$



$$\begin{aligned} \text{Var}[l|H_0] &= E \left\{ \left[(l|H_0) - E[l|H_0] \right]^2 \right\} \\ &= E \left[\left(\int_0^{3T} n(t)s_1(t)dt \right)^2 \right] = \frac{N_0 E_s}{2} \end{aligned}$$

$$\begin{aligned} E[l|H_1] &= E \left[\int_0^{3T} x(t)s_1(t)dt | H_1 \right] \\ &= E \left[\int_0^{3T} (s_1(t) + n(t))s_1(t)dt \right] \\ &= E_s \end{aligned}$$



$$\begin{aligned} \text{Var}[l|H_1] &= E \left\{ \left[(l|H_1) - E[l|H_1] \right]^2 \right\} \\ &= E \left[\left(\int_0^{3T} n(t) s_1(t) dt \right)^2 \right] = \frac{N_0 E_s}{2} \end{aligned}$$

$$d^2 \stackrel{\text{def}}{=} \frac{\left(E[l|H_1] - E[l|H_0] \right)^2}{\text{Var}(l|H_0)} = \frac{\left(E_s - (-E_s) \right)^2}{N_0 E_s / 2}$$

$$= \frac{8E_s}{N_0}$$



$$P(H_1|H_0) = \int_{\gamma}^{\infty} p(l|H_0)dl = Q\left(\frac{\ln \eta}{d} + \frac{d}{2}\right) = Q\left(\frac{d}{2}\right) = Q\left(\sqrt{\frac{2E_s}{N_0}}\right)$$

$$P(H_1|H_1) = \int_{\gamma}^{\infty} p(l|H_1)dl = Q\left(\frac{\ln \eta}{d} - \frac{d}{2}\right)$$

$$\begin{aligned} P(H_0|H_1) &= 1 - Q\left(\frac{\ln \eta}{d} - \frac{d}{2}\right) = 1 - Q\left(-\frac{d}{2}\right) = 1 - Q\left(-\sqrt{\frac{2E_s}{N_0}}\right) \\ &= Q\left(\sqrt{\frac{2E_s}{N_0}}\right) \end{aligned}$$

$$P_e = P(H_0)P(H_1|H_0) + P(H_1)P(H_0|H_1) = Q\left(\sqrt{\frac{2E_s}{N_0}}\right)$$