#### 第二章: 随机信号与系统:

1. 随机信号(序列)通过线性时不变系统:时域频域的分析方法

$$y(t) = x(t) * h(t) = \int_{-\infty}^{\infty} h(\tau)x(t-\tau)d\tau , \quad Y(\omega) = H(j\omega)X(\omega)$$
$$S_{y}(\omega) = \int_{-\infty}^{\infty} R_{y}(\tau)e^{-j\omega\tau}d\tau = |H(j\omega)|^{2} S_{x}(\omega)$$

2. 平稳随机序列的参数模型: ARMA 模型, AR 模型, MA 模型: 自相关函数, 功率谱密度函数, 三种模型之间的联系, 模型的建立

$$y(n) = \sum_{k=0}^{q} b_k w(n-k) - \sum_{k=1}^{p} a_k y(n-k)$$
 滑动平均分量+自回归分量

3. 随机信号的复表示: 正弦型信号的复表示, 任意信号的复表示

第三章:信号检测:判断是否存在信号或者存在哪个信号问题:假设检验问题处理

1. 几种准则下的判决规则都具有如下似然比检验形式: 
$$\lambda(x) = \frac{f(x|H_1)}{f(x|H_0)} \stackrel{H_1}{\underset{H_0}{\gtrless}} th$$

最大后验概率准则: 
$$P(H_1|x) \underset{H_0}{\overset{H_1}{\geqslant}} P(H_0|x)$$
,  $th = \frac{P(H_0)}{P(H_1)}$ 

最小错误概率准则: 
$$\overline{P}_e = P(H_0)P(D_1|H_0) + P(H_1)P(D_0|H_1) \stackrel{R0,R1}{\rightarrow} \min$$
,  $th = \frac{P(H_0)}{P(H_1)}$ 

$$P(D_1|H_0) = \int_{R_1} f(x|H_0) dx$$

贝叶斯准则:

$$\begin{split} \overline{C} &= P(H_0) \Big[ C_{00} P(D_0 | H_0) + C_{10} P(D_1 | H_0) \Big] + P(H_1) \Big[ C_{01} P(D_0 | H_1) + C_{11} P(D_1 | H_1) \Big] \rightarrow \min \\ th &= \frac{P(H_0) (C_{10} - C_{00})}{P(H_1) (C_{01} - C_{11})} \end{split}$$

极小极大准则:

$$\overline{C}(p, p_{1}) = C_{00} p + C_{11}(1-p) + (C_{10} - C_{00})\alpha(p_{1})p + (C_{01} - C_{11})\beta(p_{1})(1-p) \ge \overline{C}(p, p) = \overline{C}_{\min}(p)$$

$$p_{1} = \underset{p}{\operatorname{arg max}} \overline{C}_{\min}(p), \quad \frac{\partial \overline{C}(p, p_{1})}{\partial p} = 0, \quad th = \frac{p_{1}(C_{10} - C_{00})}{(1-p_{1})(C_{01} - C_{11})}$$

纽曼-皮尔逊准则:  $\min P(D_0 \mid H_1)$  s.t.  $P(D_1 \mid H_0) = \alpha$ , th 由给定的虚警概率确定。

2. 对于 M 种假设的假设检验问题:

$$\overline{C} = \sum_{i=0}^{M-1} \sum_{i=0}^{M-1} C_{ij} P(D_i | H_j) P(H_j) \xrightarrow{R_0, R_1, \dots, R_{M-1}} \min$$

当 
$$C_{ii} = 0, C_{ij} = 1, i \neq j$$
:似然比检验形式  $\lambda(x) = \frac{f(x|H_i)}{f(x|H_j)} \ge \frac{P(H_j)}{P(H_i)}, j = 0, \dots, M-1, \neq i$ 

3. 多样本的假设检验问题:

$$\lambda(\mathbf{x}) = \frac{f(\mathbf{x}|H_1)}{f(\mathbf{x}|H_0)} = \frac{f(x_1, x_2, \dots, x_N|H_1)}{f(x_1, x_2, \dots, x_N|H_0)} \underset{H_0}{\overset{H_1}{\geq}} \frac{P(H_0)(C_{10} - C_{00})}{P(H_1)(C_{01} - C_{11})} = th$$

4. 判决性能的计算

根据判决规则,确定检验统计量,计算检验统计量在两种假设下的概率密度函数。

$$P(D_1|H_1) = \int_{th}^{+\infty} f(\lambda|H_1) d\lambda$$

5. 复合的假设检验问题

$$\lambda\left(\mathbf{x}\right) = \frac{f\left(\mathbf{x}\middle|H_{1}\right)}{f\left(\mathbf{x}\middle|H_{0}\right)} = \frac{\int_{\left(\mathbf{\Theta}\right)} f\left(\mathbf{x}\middle|\mathbf{\Theta}, H_{1}\right) f_{1}\left(\mathbf{\Theta}\right) d\mathbf{\Theta}}{\int_{\left(\mathbf{\Phi}\right)} f\left(\mathbf{x}\middle|\mathbf{\Phi}, H_{0}\right) f_{0}\left(\mathbf{\Phi}\right) d\mathbf{\Phi}} \underset{H_{0}}{\overset{H_{1}}{\geq}} \frac{\left(C_{10} - C_{00}\right) P\left(H_{0}\right)}{\left(C_{01} - C_{11}\right) P\left(H_{1}\right)} = th$$

6. 高斯白噪声中已知信号的检测: 根据似然比判决设计最佳接收机,计算系统性能。

$$f\left(x(t)\middle|H_{i}\right) = \lim_{N \to \infty} f\left(x_{1}, x_{2}, \dots, x_{N}\middle|H_{i}\right) = F \exp\left\{-\frac{1}{N_{0}} \int_{0}^{T} \left[x(t) - s_{i}(t)\right]^{2} dt\right\}$$

$$\lambda(x(t)) = \underset{H_0}{\overset{H_1}{\geq}} th \Rightarrow \int_0^T \left[ s_1(t) - s_0(t) \right] x(t) dt \underset{H_0}{\overset{H_1}{\geq}} \frac{N_0}{2} \ln th + \frac{1}{2} \int_0^T \left[ s_1^2(t) - s_0^2(t) \right] dt$$

通信接收机(最小错误概率准则):

$$P_{e} = 1 - \Phi\left(\sqrt{(1-\rho)E/N_{0}}\right), \quad \rho = \frac{1}{E} \int_{0}^{T} s_{0}(t) s_{1}(t) dt, \quad E = \frac{1}{2} (E_{0} + E_{1})$$

雷达接收机(NP 准则): 
$$P_D = 1 - \Phi \left( \Phi^{-1} \left( 1 - \alpha \right) - \sqrt{2E_1/N_0} \right)$$

7. 匹配滤波器

$$t = t_0$$
 时刻的瞬时输出信噪比  $SNR_o = \frac{\left|s_o\left(t_0\right)\right|^2}{E\left\{n_o^2\left(t\right)\right\}} \le \frac{E}{N_0/2}$ 。  $H\left(j\omega\right) = S^*\left(\omega\right)e^{-j\omega t_o}$ 。

$$y(T) = \int_0^T x(t)s(t)dt$$

$$\xrightarrow{x(t)} h(t) = s(T-t)$$

$$0 \le t \le T$$

$$0 \le t \le T$$

$$y(T)$$

$$t = T$$

8. 信号的分集接收:设计最佳接收机,计算系统性能

$$\lambda\left(\mathbf{x}(t)\right) = \frac{f\left(x_{1}(t), \dots, x_{M}(t)\middle|H_{1}\right)}{f\left(x_{1}(t), \dots, x_{M}(t)\middle|H_{0}\right)} = \prod_{i=1}^{M} \frac{f\left(x_{i}(t)\middle|H_{1}\right)}{f\left(x_{i}(t)\middle|H_{0}\right)} = \prod_{i=1}^{M} \lambda\left(x_{i}(t)\right) \underset{H_{0}}{\gtrless} th$$

9. 高斯色噪声中已知信号的检测

预白化方法:

$$x(t) = s(t) + n(t)$$
  $h_1(t)$   $s_1(t) + n_1(t)$   $h_2(t)$  输出  $h_2(t)$   $h_2(t)$ 

$$H(j\omega) = H_1(j\omega)H_2(j\omega) = \frac{1}{S_n^+(\omega)}\frac{S_n^+(\omega)}{S_n^-(\omega)}e^{-j\omega T} = \frac{S_n^*(\omega)}{S_n(\omega)}e^{-j\omega T}$$

卡亨南-洛维展开: 
$$\int_0^T R_n(t_1-t_2)f_j(t_2)dt_2 = \lambda_j f_j(t_1)$$
,  $0 \le t_1 \le T$ 

实信号下,
$$x(t) = \sum_{k} x_k f_k(t)$$
, $x_k = \int_0^T x(t) f_k(t) dt$ , $x_k$  间互不相关。

$$\lambda(x(t)) = \lim_{N \to \infty} \lambda(x_1, x_2, \dots, x_N) \underset{H_0}{\gtrless} th$$

$$\int_{0}^{T} x(t) \eta_{1}(t) dt - \int_{0}^{T} x(t) \eta_{0}(t) dt \underset{H_{0}}{\overset{H_{1}}{\geq}} th', \quad th' = \ln th + \frac{1}{2} \int_{0}^{T} s_{1}(t) \eta_{1}(t) dt - \frac{1}{2} \int_{0}^{T} s_{0}(t) \eta_{0}(t) dt$$

10. 高斯白噪声中随机相位信号的检测:设计最佳接收机,计算系统性能

$$\lambda(x(t)) = \frac{f(x(t)|H_1)}{f(x(t)|H_0)} = \frac{\int_0^{2\pi} f(x(t)|\theta, H_1) f(\theta) d\theta}{f(x(t)|H_0)} = \exp\left(-\frac{A^2T}{2N_0}\right) I_0\left(\frac{2Aq}{N_0}\right)_{H_0}^{H_1} th$$

判决规则:  $q = \sqrt{a^2 + b^2} \mathop{\gtrless}_{H_0}^{H_1} th'$  -------门限th' 由 $\alpha$ 确定(NP 准则)

$$a = q \sin \theta_0 = \int_0^T x(t) \cos \omega_c t dt$$
$$b = q \cos \theta_0 = \int_0^T x(t) \sin \omega_c t dt$$

计算性能:  $f(a,b|\theta,H_i) \rightarrow f(q,\theta_0|\theta,H_i) \rightarrow f(q|\theta,H_i) \rightarrow f(q|H_i)$ 

非相干频移键控信号的检测:设计最佳接收机,计算系统性能

$$\lambda(x(t)) = \frac{f(x(t)|H_1)}{f(x(t)|H_0)} = \frac{\int_0^{2\pi} f(x(t)|\theta, H_1) f(\theta) d\theta}{\int_0^{2\pi} f(x(t)|\phi, H_0) f(\phi) d\phi} = \frac{I_0\left(\frac{2Aq_1}{N_0}\right)}{I_0\left(\frac{2Aq_0}{N_0}\right)^{H_1}} \underset{H_0}{\overset{H_1}{\geqslant}} 1$$

$$q_i = \left[\left(\int_0^T x(t) \sin \omega_i t dt\right)^2 + \left(\int_0^T x(t) \cos \omega_i t dt\right)^2\right]^{\frac{1}{2}}, \quad i = 1, 2, \quad q_i > 0$$

## 第四章: 参量估计

# 1. 估计准则

最大后验概率估计: 
$$\hat{\theta}_{MAP} = \underset{\theta}{\operatorname{arg max}} f\left(\theta \mid \mathbf{x}\right) = \underset{\theta}{\operatorname{arg max}} \ln f\left(\theta \mid \mathbf{x}\right)$$
 最大似然估计:  $\hat{\theta}_{ML} = \underset{\theta}{\operatorname{arg max}} f\left(\mathbf{x} \mid \theta\right) = \underset{\theta}{\operatorname{arg max}} \ln f\left(\mathbf{x} \mid \theta\right)$  最小均方误差估计:  $E\left\{e^{2}\left(\hat{\theta}\right)\right\} = \underset{(\theta)}{\int} \int_{(\theta)(\mathbf{x})} \left(\theta - \hat{\theta}\right)^{2} f\left(\theta, \mathbf{x}\right) d\theta d\mathbf{x} \rightarrow \min$ ,  $\hat{\theta}_{MS} = \int_{(\theta)} \theta f\left(\theta \mid \mathbf{x}\right) d\theta$  线性最小均方误差估计:  $E\left\{\left[\theta - \left(a + \sum_{k=1}^{N} b_{k} x_{k}\right)\right]^{2}\right\} \rightarrow \min$   $\hat{\theta}_{LMS} = a + \mathbf{b}^{T} \mathbf{x} = E\left\{\theta\right\} + Cov\left\{\theta, \mathbf{x}\right\} Cov^{-1}\left\{\mathbf{x}, \mathbf{x}\right\}\left[\mathbf{x} - E\left\{\mathbf{x}\right\}\right]$  正交条件 (充要):  $E\left\{\left(\theta - \hat{\theta}_{LMS}\right)\mathbf{x}^{T}\right\} = 0$  最小平均绝对误差估计:  $E\left\{\left|e\left(\hat{\theta}\right)\right|\right\} \rightarrow \min$ ,  $\int_{-\infty}^{\hat{\theta}_{ABS}} f\left(\theta \mid \mathbf{x}\right) d\theta = \int_{\hat{\theta}_{ABS}}^{\infty} f\left(\theta \mid \mathbf{x}\right) d\theta$  贝叶斯估计:  $E\left\{c\left(\hat{\theta}\right)\right\} \rightarrow \min$  最小二乘估计: 线性观测模型下 $\hat{\theta}_{LS} = \underset{\hat{\rho}}{\operatorname{arg min}} \left[\mathbf{x} - \mathbf{H}\hat{\mathbf{\theta}}\right]^{T} \left[\mathbf{x} - \mathbf{H}\hat{\mathbf{\theta}}\right]$ ,

### 2. 多参量估计

$$\begin{split} & \left[ \frac{\partial}{\partial \theta_{i}} \ln f \left( \mathbf{\theta} \, | \, \mathbf{x} \right) \right]_{\mathbf{\theta} = \hat{\mathbf{\theta}}_{MAP}} = 0 \,, \quad \left[ \frac{\partial}{\partial \theta_{i}} \ln f \left( \mathbf{x} \, | \, \mathbf{\theta} \right) \right]_{\mathbf{\theta} = \hat{\mathbf{\theta}}_{ML}} = 0 \,, \quad i = 1, 2, \dots, M \\ & E \left\{ \left[ \mathbf{\theta} - \mathbf{a} - \mathbf{B} \mathbf{x} \right]^{T} \left[ \mathbf{\theta} - \mathbf{a} - \mathbf{B} \mathbf{x} \right] \right\} \rightarrow \min \\ & \hat{\mathbf{\theta}}_{LMS} = E \left\{ \mathbf{\theta} \right\} + Cov \left\{ \mathbf{\theta}, \mathbf{x} \right\} Cov^{-1} \left\{ \mathbf{x}, \mathbf{x} \right\} \left[ \mathbf{x} - E \left\{ \mathbf{x} \right\} \right] \,, \qquad E \left\{ \left( \mathbf{\theta} - \hat{\mathbf{\theta}}_{LMS} \right) \mathbf{x}^{T} \right\} = 0 \end{split}$$

 $\hat{\boldsymbol{\theta}}_{LS} = \left(\mathbf{H}^T \mathbf{H}\right)^{-1} \mathbf{H}^T \mathbf{x} = \mathbf{H}^{\#} \mathbf{x}$ 

$$\hat{\boldsymbol{\theta}}_{LS} = \left(\mathbf{H}^T \mathbf{H}\right)^{-1} \mathbf{H}^T \mathbf{x} = \mathbf{H}^{\#} \mathbf{x}$$

3. 计算估计性能: 
$$E\{\hat{\theta}\}, E\{(\hat{\theta}-E\{\hat{\theta}\})^2\}, E\{(\theta-\hat{\theta})^2\}$$

无偏性: 
$$E\{\hat{\mathbf{\theta}}\} = \mathbf{\theta}$$
 或  $E\{\hat{\mathbf{\theta}}\} = E\{\mathbf{\theta}\}$ 

有效性:无偏估计量的均方误差达到最小。对于确定单参量无偏估计量 $\hat{ heta}$ ,有

$$E\left\{\left(\hat{\theta} - \theta\right)^{2}\right\} = \left\{E\left\{\left[\frac{\partial}{\partial \theta} \ln f\left(\mathbf{x} \mid \theta\right)\right]^{2}\right\}\right\}^{-1} = -\left\{E\left\{\frac{\partial^{2}}{\partial \theta^{2}} \ln f\left(\mathbf{x} \mid \theta\right)\right\}\right\}^{-1}$$

### 第五章:波形估计

1. 连续维纳滤波器

$$E\left\{e^{2}\left(t\right)\right\} = E\left\{\left[g\left(t\right) - y\left(t\right)\right]^{2}\right\} = E\left\{\left[g\left(t\right) - \int_{-\infty}^{\infty} h\left(\tau\right) \operatorname{Tx}\left(t - \tau\right) d\tau^{\mathsf{T}}\right]^{2}\right\} \xrightarrow{h(t)} \min$$

线性最小均方误差估计的正交条件:  $E\{e(t)x(\tau')\}=0$ ,  $\begin{cases} -\infty < \tau' < \infty & \text{非因果} \\ -\infty < \tau' \le t & \text{因果} \end{cases}$ 

维纳-霍夫方程: 
$$R_{gx}(\eta) = \int_{-\infty}^{\infty} h(\lambda) R_{x}(\eta - \lambda) d\lambda$$
, 
$$\begin{cases} -\infty < \eta < \infty & \text{非因果} \\ 0 \le \eta < \infty & \text{因果} \end{cases}$$

求解: 
$$H(j\omega) = \frac{S_{gx}(\omega)}{S_{x}(\omega)}$$
-----物理不可实现

$$H(s) = \frac{1}{S_x^+(s)} \left[ \frac{S_{gx}(s)}{S_x^-(s)} \right]^+$$
------ 物理可实现

$$E\left\{e^{2}\left(t\right)\right\}_{\min}=E\left\{e\left(t\right)g\left(t\right)\right\}=R_{g}\left(0\right)-\int_{-\infty}^{\infty}h(\lambda)R_{gx}(\lambda)d\lambda$$

2. 离散维纳滤波器

$$E\left\{e^{2}\left(k\right)\right\} = E\left\{\left[g\left(k\right) - y\left(k\right)\right]^{2}\right\} = E\left\{\left[g\left(k\right) - \sum_{i=-\infty}^{\infty} h(i)x(k-i)\right]^{2}\right\} \longrightarrow \min$$

线性最小均方误差估计的正交条件:  $E\{e(k)x(j)\}=0$ ,  $\begin{cases} -\infty < j < \infty & \text{非因果} \\ -\infty < j \le k & \text{因果} \end{cases}$ 

求解: 
$$H(z) = \frac{S_{gx}(z)}{S_x(z)}$$
-----物理不可实现

$$H(z) = \frac{1}{S_x^+(z)} \left[ \frac{S_{gx}(z)}{S_x^-(z)} \right]^+$$
------ 物理可实现

对应有限样本的维纳滤波器: 
$$\sum_{i=0}^{N-1} h(i) R_x(l-i) = R_{gx}(l), l = 0, 1, \dots, N-1$$

$$\mathbf{R}_{x}\mathbf{h}=\mathbf{r}_{gx}, \quad \mathbf{h}=\mathbf{R}_{x}^{-1}\mathbf{r}_{gx}$$