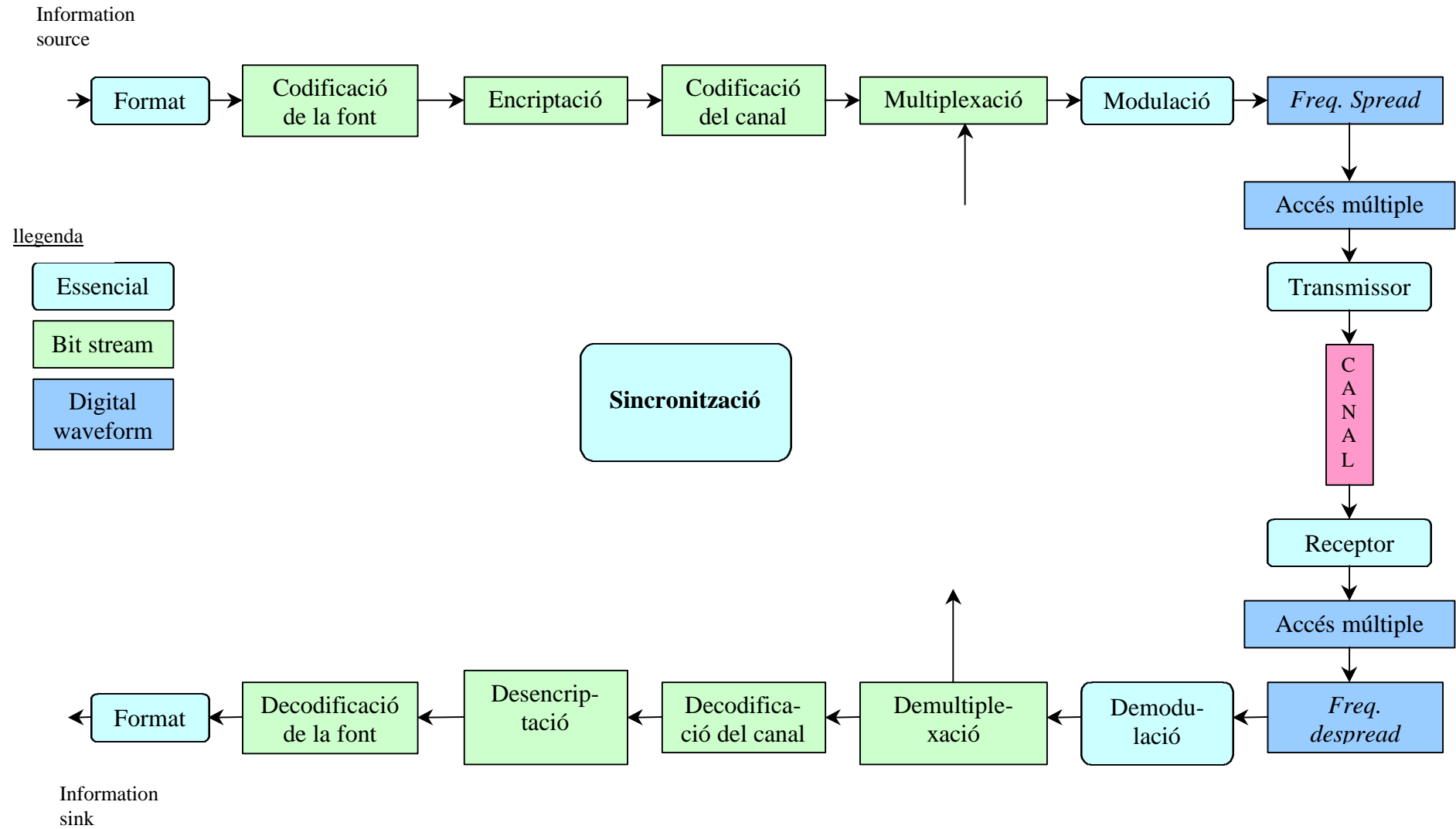


Esquema general d'un sistema de transmissió digital



El sistema de transmisión de datos

Pulse Amplitude Modulation (PAM)

$$x_T(t) = \sum_{n=-\infty}^{+\infty} a_n p(t - nT)$$

M posibles símbolos $b \text{ bits} = \log_2 M$
 $M = 2^b$

$$r = \frac{1}{T} \text{ [símbolos/seg]}$$

$$r_b = rb \text{ [bits/seg]}$$

$$R_{x_T x_T}(t + \mathbf{t}, t) = E\{x_T(t + \mathbf{t})x_T^*(t)\} : \text{cicloestacionario}$$

$$\bar{R}_{x_T x_T}(\mathbf{t}) = \frac{1}{T} \sum_{n=-\infty}^{+\infty} R_{aa}(n) R_{pp}(\mathbf{t} - nT)$$

$$S_{x_T x_T}(f) = F\{\bar{R}_{x_T x_T}(\mathbf{t})\} = \sum_{n=-\infty}^{+\infty} R_{aa}(n) e^{-2jfnT} \frac{|P(f)|^2}{T}$$

$$m_a = E\{a_n\}$$

$$a_n \equiv A_n + m_a$$

$$S_{x_T x_T}(f) = \left[\sum_{n=-\infty}^{+\infty} R_{AA}(n) e^{-2jfnT} \right] \frac{|P(f)|^2}{T} + \frac{|m_a|^2}{T^2} \sum_{n=-\infty}^{+\infty} |P\left(\frac{k}{T}\right)|^2 \mathbf{d}\left(f - \frac{k}{T}\right)$$

(a) $a[n]$ estadísticamente independientes

$$S_{x_T x_T}(f) = r \mathbf{S}_a^2 |P(f)|^2 + r^2 |m_a|^2 \sum_{n=-\infty}^{+\infty} |P(kr)|^2 \mathbf{d}(f - kr)$$

(b) $a[n]$ estadísticamente independientes y media nula

$$S_{x_T x_T}(f) = r \mathbf{S}_a^2 |P(f)|^2$$

$$x_R(t) \Big|_{t=t_k=kT} = \sum_{n=-\infty}^{+\infty} p_r(kT - nT) + w(t) = \underbrace{a_k p_r(0)}_{\text{señal}} + \underbrace{\sum_{n=-\infty, n \neq k}^{+\infty} p_r((k-n)T)}_{\text{ISI}} + \underbrace{w(t)}_{\text{ruido}}$$

$$p_r(t) \equiv p_T(t) * h_T(t) * h_C(t) * h_R(t)$$

$$w(t) \equiv n(t) * h_R(t)$$

Condición no ISI. Teorema de Nyquist

$$p(kT) = \begin{cases} 1, & k = 0 \\ 0, & k \neq 0 \end{cases} \Leftrightarrow r \sum_{k=-\infty}^{+\infty} P(f - kr) = cte$$

Teoría de la decisión

Suponemos que no hay ISI, y que trabajamos, pues, con pulsos ideales.

$$y_k = y(kT) = a_k + w(kT)$$

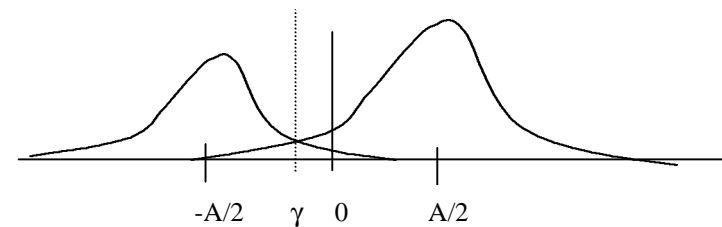
$$P_{error} = P_e = P(e|'1')P('1') + P(e|'0')P('0')$$

$$P_{error} = BER \Leftarrow \text{los símbolos son binarios}$$

$$H_0 = '0', H_1 = '1' \text{ hipótesis}$$

$$P_e = P(e|H_0)P(H_0) + P(e|H_1)P(H_1)$$

$$P(e|H_1) = \int_{-\infty}^{+\infty} f_{y|H_1}(y|H_1) dy = \int_{-\infty}^{\frac{g}{2}} f_w(y + \frac{A}{2}) dy$$



$$\mathbf{g}_{\text{óptima}} \Rightarrow \frac{\partial}{\partial \mathbf{g}} P_e(\mathbf{g}) = 0$$

$$f_{y|H_0}(y | H_0)P(H_0) = f_{y|H_1}(y | H_1)P(H_1) \Big|_{y=g}$$

$$f_w(\mathbf{g} + \frac{A}{2})P(H_0) = f_w(\mathbf{g} - \frac{A}{2})P(H_1)$$

si $f_w(w) \approx \text{Gaussiana}(0, \mathbf{s}_w^2)$, $f_w(w) = \frac{1}{\sqrt{2\pi\mathbf{s}_w^2}} e^{-\frac{1}{2}\frac{w^2}{\mathbf{s}_w^2}}$

$$\Rightarrow Q(x) \equiv \int_x^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}w^2} dw \Rightarrow \mathbf{g}_{\text{óptima}} = -\frac{\mathbf{s}_w^2}{A} \ln \frac{P(H_1)}{P(H_2)}$$

Si $a[n]$ es una v.a, entonces

$$f_a(a) = P(H_0)\mathbf{d}(a + \frac{A}{2}) + P(H_1)\mathbf{d}(a - \frac{A}{2})$$

si $g = x + y$, con x e y v.as, $f_g(g) = f_x(x) * f_y(y)$ si x, y son indep.

$$f_y(y) = f_a(y) * f_w(y)$$

$$P(e | H_0) = Q\left(\frac{\mathbf{g} + A/2}{\mathbf{s}_w}\right) \quad P(e | H_1) = Q\left(\frac{\mathbf{g} - A/2}{\mathbf{s}_w}\right)$$

si $\mathbf{g} = 0 \Rightarrow P_e = Q\left(\frac{A}{2\mathbf{s}_w}\right)$

Teoría de la detección óptima

$$H_R^{\text{óptimo}}(f) = I \frac{P^*(f)}{S_{nn}(f)} e^{-2pf_0}$$

Si $n(t) \approx \text{AWGN}$, $S_{nn}(f) = \frac{N_0}{2} \Rightarrow H_R^{\text{óptimo}}(f) = I p^*(td - t)$

s'anomena Filtre Adaptat, i és el detector óptim amb soroll gaussià

$$h_r^{\text{óptima}}(t) = a p^*(td - t)$$

$$\mathbf{a} = \frac{1}{E_p} = \frac{1}{\int_{-\infty}^{+\infty} |p(t)|^2 dt} \Rightarrow h_r^{\text{óptima}}(t) = \frac{1}{E_p} p^*(td - t)$$

$$S_T(t) = E\{|x(t)|^2\} = E\{|\sum a_n p(t - nT)|\}$$

binari polar $S_T = \frac{1}{T} \int_T S_T(t) dt = r \mathbf{s}_a^2 E_p = r \left[\frac{1}{2} \left(\frac{A}{2}\right)^2 + \frac{1}{2} \left(-\frac{A}{2}\right)^2 \right] E_p = r \frac{A^2}{4} E_p = S_T$

$$E_b = S_T T = \frac{A^2}{2} = \sqrt{\frac{E_b}{2E_p}} \Rightarrow \text{BER} = P_e = Q\left(\frac{A}{2\mathbf{s}_w}\right) = Q\left(\sqrt{\frac{E_b}{2}}\right)$$

Sistemas multinivel

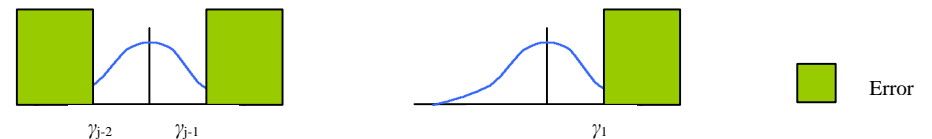
$$a_k = \left\{ \pm \frac{A}{2}, \pm 3\frac{A}{2}, \pm 5\frac{A}{2}, \dots, \pm(M-1)\frac{A}{2} \right\} \quad M \text{ símbolos equiprobables } p(H_i) = \frac{1}{M}$$

$$h_r(t) = \frac{1}{E_p} p^*(td - t)$$

$$f_{y|H_i}(y | H_i) = f_w(y - a_i)$$

$$\min\{P_e\} = \sum_{i=1}^M P(e | H_i)P(H_i) = \frac{1}{M} \sum_{i=1}^M P(e | H_i)$$

2 casos: centrales y extremos



$$P_e = \frac{2}{M} \overbrace{P(e | H_1)}^{\text{extrem}} + \frac{1}{M} \sum_{i=2}^{M-1} \overbrace{P(e | H_i)}^{\text{centrals}} = \frac{2}{M} \int_{g_1}^{+\infty} f_w(y - a_1) dy + \frac{1}{M} \sum_{i=2}^{M-1} \left(1 - \int_{g_{i-1}}^{g_i} f_w(y - a_i) dy\right)$$

$$P_e(g_1, g_2, \dots, g_{M-1}) \Rightarrow \frac{\partial}{\partial g_j} P_e = 0 \quad \forall j \Leftrightarrow f_w(g_j - a_j) = f_w(g_j - a_{j-1})$$

$$g_j^{\text{optima}} = \frac{a_{j+1} + a_j}{2} \Leftrightarrow \begin{cases} P\{H_i\} = \frac{1}{M} \\ f_w(w) = f_w(-w) \end{cases}$$

$$P_e = \frac{2M-2}{M} Q\left(\frac{A}{2s_w}\right) \quad \text{BER} \cong \frac{P_e}{\log_2 M}$$

$$E_b \rightarrow \frac{E_b}{N_o} \rightarrow \text{Governa la qualitat del sistema}$$

$$E_b \cong S_T \cdot T_b \cong S_T \frac{T}{b}$$

$$T_b \cong \frac{T}{b} = \frac{T}{\log_2 M}$$

$$S_T \begin{cases} R_{X_T X_T}(0) \\ \frac{1}{T} \int_T E\{|x(t)|^2\} dt \\ \text{estats independents} \end{cases}$$

$$s(t) = a_k p(t - kT)$$

$$E_s = \int_{-\infty}^{+\infty} E\{|S(t)|^2\} dt = E\{|a_k|^2\} \int_{-\infty}^{+\infty} \overbrace{E\{|p(t)|^2\}}^{E_p} dt = E\{|a_k|^2\} E_p$$

$$E_b = \frac{E_s}{b} = \frac{m^2 - 1}{12} A^2 \frac{E_p}{M \log_2 M}$$

$$P_e = \frac{2M-2}{M} Q\left(\sqrt{\frac{6 \log_2 M}{m^2 - 1}} \left(\frac{E_b}{N_o}\right)\right)$$

Polsos de Nyquist

$$p(t) = \text{sinc}(rt) = \frac{\sin(prt)}{prt} \Leftrightarrow P(f) = T \prod(Tf)$$

$$p_N(t) \equiv p_b \text{sinc}(rt) \Leftrightarrow P_N(f) = P_B(f) * \prod(Tf), P_B(f) = 0 \forall |f| \geq BW_B$$

$$BW_N = \frac{r}{2} + BW_B \equiv \frac{r}{2} (1 + b)$$

$$b \equiv \frac{BW_B}{r/2} : \text{factor de ROLL - OFF (\%)}$$

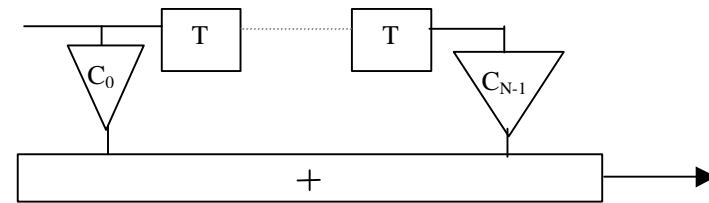
Ecuación

Igualación (compensación) del canal

Canal: forzadores de ceros, si el ruido afecta poco a la calidad del sistema

Canal + ruido: equalización, forzadores de ceros.

Forzador de ceros



$$\begin{matrix} P \\ P(1) & 0 & 0 \\ P(2) & P(1) & 0 \\ P(3) & P(2) & P(1) \end{matrix} \begin{matrix} C \\ C_0 \\ C_1 \\ C_2 \end{matrix} = \begin{matrix} 1 \\ 0 \\ 0 \end{matrix} \text{ donde haya más señal}$$

$$P C = 1$$

$$C = P^{-1} 1$$

$$C_{opt}^{ZF} = (P^T P)^{-1} P^T \cdot 1 \leftarrow \text{MECM}$$

El uso de forzadores de cero incrementa la pot

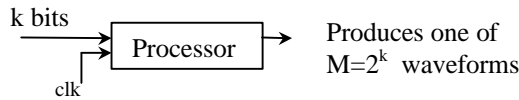
Waveform coding

Orthogonal signals

$$\Rightarrow \frac{1}{E} \int_0^T s_i(t)s_j(t)dt = \begin{cases} 1 & i = j \\ 0 & \text{otherwise} \end{cases}$$

$$E_i = \int_0^T s_i^2(t)dt$$

M-ary signaling



Orthogonal codes

Data set	Orthogonal codeword set
0	0 0
1	0 1

$$H_K = \begin{bmatrix} H_{k-1} & H_{k-1} \\ H_{k-1} & \bar{H}_{k-1} \end{bmatrix}$$

$$z_{ij} = \mathbf{d}[i - j]$$

$$P_E(k) \leq (2^k - 1)Q\left(\sqrt{\frac{kE_b}{N_0}}\right)$$

$$\frac{P_B(k)}{P_E(k)} = \frac{2^{k-1}}{2^{k-1} - 1}$$

Biorthogonal codes

Biorthogonal signal set of M signals or codewords can be obtained from an orthogonal set of M/2 signals. It consists of a combination of orthogonal and antipodal signals.

$$B_k = \begin{bmatrix} H_{k-1} \\ \bar{H}_{k-1} \end{bmatrix} \quad z_{ij} = \mathbf{d}[i - j] - \mathbf{d}\left[|i - j| - \frac{M}{2}\right] = \begin{cases} 1 & i = j \\ -1 & |i - j| = \frac{M}{2} \\ 0 & \text{otherwise} \end{cases}$$

$$P_E(k) \leq (2^k - 2)Q\left(\sqrt{\frac{kE_b}{N_0}}\right) + Q\left(\sqrt{\frac{2kE_b}{N_0}}\right) \quad P_B(k) \cong \frac{P_E(k)}{2}, k > 3$$

i.e.:

Data set	Orthogonal codeword set	Data set	Biorthogonal codeword set
0 0	$H_2 = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 \end{bmatrix}$	000	$H_2 = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 \end{bmatrix}$
0 1		001	
1 0		010	
1 1		011	
		100	
	101		
	110		
	111		

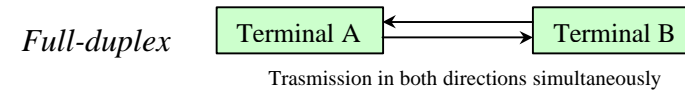
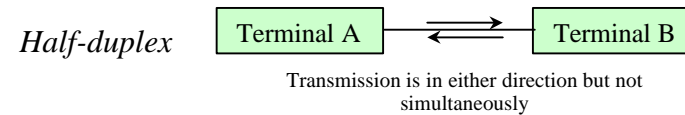
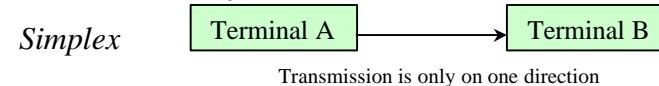
Transorthogonal (Simplex) Codes

Results from deleting the first digit of each orthogonal codeword.

$$z_{ij} = \begin{cases} 1 & i = j \\ -1 & i \neq j \end{cases} \quad \text{represents minimum energy}$$

Types of error control

Terminal connectivity



Automatic Repeat Request (ARQ methods)

- Stop-and-wait ARQ (half-duplex)* Transmitter waits for an ACK or a NAK.
- Continuous ARQ with pullback (full-duplex)* When the transmitter receives a NAK it also receives in which transmission block was the error and restarts transmission from that transmission block.
- Continuous ARQ with selective repeat (full-duplex)* Only the corrupted message is repeated.

Forward Error Correction (FEC)

- A reverse channel is not available or the delay with ARQ would be excessive.
- The retransmission strategy is not conveniently implemented.
- The expected number of errors, without corrections, would require excessive retransmissions.

Structured sequences

Orthogonal coding techniques have inefficient use of bandwidth.

Channel Modes

Discrete Memory Less Channel (DMC)

- Discrete input alphabet $P(j | i) \{1 \leq i \leq M, 1 \leq j \leq Q\}$
- Discrete output alphabet i : modulator M - ary symbol
- Set of conditional probabilities j : demodulator Q - ary output symbol
- $P(i | j)$: prob. of receiving j given that i was transmitted.

Input sequence $U = u_1, \dots, u_m$
 Output sequence $Z = z_1, \dots, z_m$

$$P(Z | U) = \prod_{n=1}^m P(z_n | u_n)$$

Binary Symmetric channel (BSC) "Hard Decisions"

$$P(0 | 1) = P(1 | 0) = p$$

$$P(1 | 1) = P(0 | 0) = 1 - p$$

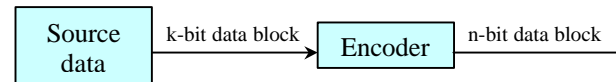
$$p = Q\left(\sqrt{\frac{2E_c}{N_0}}\right)$$

Gaussian Channel "Soft Decisions"

Discrete input

Continuous output $p(z | u_k) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(z-u_k)^2}{2\sigma^2}}$

Code Rate and Redundancy



n-k bits: redundant bits, parity bits or check bits.

Fundamentals of Statistical Decision Theory

Bayes' Theorem

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

$$P(s_i|z_j) = \frac{P(z_j|s_i)P(s_i)}{P(z_j)} = \frac{P(z_j|s_i)P(s_i)}{\sum_{p=1}^M P(z_j|s_p)P(s_p)}$$

$P(s_i)$: a priori probability
 $P(z_j|s_i)$: likelihood
 $P(s_i|z_j)$: a posteriori probability

Mixed form:

$$P(s_i|z) = \frac{P(z|s_i)P(s_i)}{P(z)} = \frac{P(z|s_i)P(s_i)}{\sum_{p=1}^M P(z|s_p)P(s_p)}$$

z : continuous sample

Decision theory

$$z(T) = a_i(T) + n_0(T)$$

$$r(n) = a(n) + w(n)$$

MAP: $\arg_i \max\{P(s_i|z)\} = \arg_i \max\{P(z|s_i)P(s_i)\}$

ML: $\arg_i \max\{P(z|s_i)\}$

Binary

MAP: $P\{z|s_1\}P\{s_1\} > P\{z|s_2\}P\{s_2\} \Rightarrow H_{s_1}$

ML: $P\{z|s_1\} > P\{z|s_2\} \Rightarrow H_{s_1}$

MLSE (Maximum Likelihood Sequence Estimation)



M_1 : samples before maximum

M_2 : samples after maximum

L symbols $\Rightarrow y(-M_1) \dots y(L-1-M_2)$

$$y_{received}(t)$$

Gaussian Noise: $\hat{s}_i = \arg_{s_i} \min\left\{ \sum_{k=-M_1}^{L-1-M_2} (y_r(k) - s_i(k))^2 \right\}$

Laplacian Noise: $\hat{s}_i = \arg_{s_i} \min\left\{ \sum_{k=-M_1}^{L-1-M_2} |y_r(k) - s_i(k)| \right\}$

Codificación de la fuente

$$I(X_j) : \text{autoinformación} \quad X_j : \text{símbolo} \quad H(X) : \text{entropía de la fuente}$$

$$I(X_j) = -\log_2 p_j \quad P(X_j) = p_j \quad H(X) = E\{I(X_j)\} = -\sum_{j=1}^N p_j \log_2 p_j$$

$$M_{\text{upla}} \Rightarrow P(X_1, \dots, X_M) = \prod_{m=1}^M P(X_m)$$

$$H_M(X) = H(X)$$

$$H_M(X)_{\text{con memoria}} < H_M(X)_{\text{sin memoria}}$$

Límites de la transmisión de datos

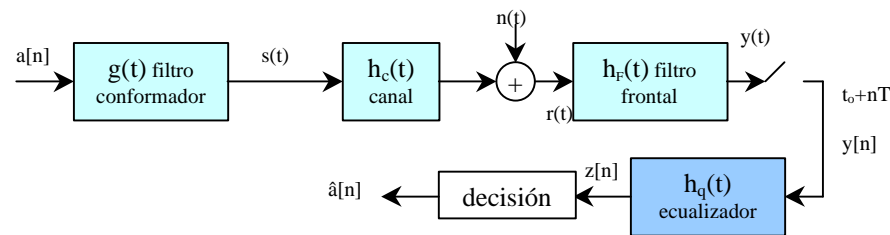
- Ancho de banda
- Potencia de señal
- Complejidad

1948, Shannon $C = W \log_2(1 + SNR)$ bps : capacidad

$$E = \frac{r_i}{W} \text{ (bits/s/Hz) : eficiencia espectral}$$

$$r_b = r_s \log_2 \underbrace{M}_{\text{n}^\circ \text{ de símbolos}} \text{ (bps)}$$

PAM



$$s(t) = \sum_{n=-\infty}^{+\infty} a(n)g(t - nT)$$

$$r(t) = s(t) * h_c(t) + n(t) = \sum_{n=-\infty}^{+\infty} a(n)g(t - nT) * h_c(t) + n(t)$$

$$y(t) = s(t) * h_c(t) * h_f(t) + n(t) * h_f(t)$$

$$\mathbf{h}(t) = n(t) * h_f(t)$$

$$x(t) \equiv g(t) * h_c(t) * h_f(t) : \text{respuesta impulsional}$$

$$x_d(t) \equiv g(t) * h_f(t) : \text{respuesta impulsional de diseño (canal ideal)}$$

$$y(t) = \sum_{n=-\infty}^{+\infty} a(n)x(t - nT) + \mathbf{h}(t)$$

$$y(t_0 + nT) = \sum_{m=-\infty}^{+\infty} a(m)x(t_0 + nT - mT) + \mathbf{h}(t_0 + nT)$$

$$y(n) = \sum_{m=-\infty}^{+\infty} a(m)x(n - m) + \mathbf{h}(n) = a(n) * x(n) + \mathbf{h}(n)$$

$$z(n) = a(n) * h(n)$$

$$h(n) \equiv h_q(n) * x(n)$$

Filtro frontal y filtro conformador

$$\mathbf{h}(t) = n(t) * h_f(t)$$

$$E\{\mathbf{h}^2(t)\} = \frac{N_0}{2} \int_{-W}^{+W} |H_f(f)|^2 df = \frac{N_0}{2} \int_0^t |h_f(t)|^2 dt$$

$$h_f(t) = g(t_0 - t) : \text{filtro adaptado}$$

$$\Rightarrow SNR_{\text{max}} = \frac{E\{a^2(n)\}}{\frac{N_0}{2}}$$