

Memoryless Modulation

PAM

$$s_m(t) = \text{Re}[A_m g(t) e^{j2\pi f_c t}] = A_m g(t) \cos 2\pi f_c t \quad m = 1, 2, 3, \dots, M$$

$$A_m = (2m - 1 - M)d \quad \text{Energy } E_m = \frac{1}{2} A_m^2 E_g$$

$$P_{\text{error}} = Q\left(\sqrt{\frac{2E_b}{N_0}}\right) \quad P_b = Q\left(\sqrt{\frac{d^2}{2N_0}}\right)$$

M-ary PAM

$$\text{error} = \frac{2(M-1)}{M} Q\left(\sqrt{\frac{d^2 E_g}{N_0}}\right)$$

PSK

$$s_m(t) = \text{Re}\left[g(t) e^{j2\pi \frac{(m-1)}{M} t} e^{j2\pi f_c t}\right] = g(t) \cos\left[2\pi f_c t + \frac{2\pi}{M}(m-1)t\right]$$

$$= g(t) \cos\frac{2\pi}{M}(m-1) \cos 2\pi f_c t - g(t) \sin\frac{2\pi}{M}(m-1) \sin 2\pi f_c t$$

constant envelope **Energy** $E_m = \frac{1}{2} E_g$

Orthogonal Signaling FSK

$$s_m(t) = A \cos(2\pi [f_0 + m\Delta f] t) = \text{Re}\left[A e^{2\pi f_0 t} e^{2\pi m \Delta f t}\right]$$

$$\Delta f_{\text{bandpass min}} = \frac{1}{2T} \quad \Delta f_{\text{lowpass min}} = \frac{1}{T} \quad \text{distance} = \sqrt{2E} \text{ or } 2\sqrt{E} \text{ for biorthogonal}$$

Dimensionality theorem
 $N \approx 2WT$ where W = bandwidth and T = duration

Union Bound for SNR > 4 ln2 **Union Bound for SNR < 4 ln2**

$$P_M < 2^k e^{-kE_b/2N_0} \quad P_M < 2e^{-k(\sqrt{E_b N_0} - \sqrt{\ln 2})^2}$$

$$\text{Given } \frac{E_b}{N_0} > 2 \ln 2 = 1.39 = 1.42 \text{ dB} \quad \text{Given } \frac{E_b}{N_0} > \ln 2 = 0.693 = -1.6 \text{ dB}$$

$$P_e(M) \leq (M-1) Q\left(\sqrt{\frac{d_{\text{min}}^2}{2N_0}}\right)$$

Simplex Signaling

$$\text{if } s(t) = \frac{1}{M} \sum_{i=1}^M s_i(t) \quad s'_m(t) = s_m(t) - s(t) \quad \text{Energy } E' = \frac{M-1}{M} E$$

Memory Modulation

NRZI

$$b_k = a_k \text{ xor } b_{k-1} \quad 10110001 \rightarrow 0, 11011110$$

CPFSK

$$s(t) = \cos\left[2\pi f_0 t + 4\pi T f_d \int_{-\infty}^t v(\tau) d\tau\right] \quad v(t) = \sum_n a_n g(t - nT_s)$$

$$\text{at any instant } f_i(t) = f_0 + f_d a_n \quad h = 2T f_d \quad \theta_n = \pi h \sum_{k=-\infty}^{n-1} a_k$$

$$\theta(t; a) = 2\pi h \int_{-\infty}^t v(\tau) d\tau = \theta_n + 2\pi h a_n q(t - nT)$$

MSK

$$s(t) = \cos\left(2\pi f_0 t + \frac{2\pi t}{4T} a_n + \theta - \frac{n\pi}{2} a_n\right) \text{ this is CPFSK with } h = \frac{1}{2}$$

$$\theta(t; a) = \frac{\pi}{2} \sum_{k=-\infty}^{n-1} a_k + \pi a_n q(t - nT) \quad \Delta f = \frac{1}{2T}$$

Power spectra density

$$S_x(f) = \frac{1}{T} S_I(f) |G(f)|^2 \quad R_i(m) = \frac{1}{2} E[I_{n+m} I_n]$$

Input Output signals

For a probabilistic signal let x = input and y = output

$$S_y(f) = S_x(f) |H(f)|^2 \quad S_{XY}(f) = S_X(f) H^*(f)$$

$$R_Y(\tau) = R_X(\tau) * h(\tau) * h^*(-\tau)$$

Match filter response

let $s =$ the signal, the match filter would be $h(t) = s(T - t)$ and the response would be $y(t) = \int_0^t s(\tau)s(T - t + \tau)d\tau$

Noise due to match filter

$$y(t) = \int_0^t r(\tau)h(t - \tau) = \int_0^t s(\tau)h(t - \tau) + \int_0^t n(\tau)h(t - \tau)d\tau = y_s(T) + y_n(T)$$

$$\text{variance of noise} = E[y_n^2(T)] = \frac{1}{2}N_2 \int_0^T h^2(T - t)dt \quad \text{SNR} = \frac{y_s^2(T)}{E[y_n^2(T)]} = \frac{2E}{N_0}$$

Frequency domain $H(f) = \int_0^T s(T - t)e^{-j2\pi ft} dt = S^*(f)e^{-j2\pi fT}$

MAP detection, MAX likelyhood detection

Let $r_1 = s + n_1$ and $r_2 = n_1 + n_2$ we want to find the best decision scheme such that $P(s = \sqrt{E}|r_1, r_2) > P(s = -\sqrt{E}|r_1, r_2)$ if we assume that both signals are equally likely we can use max likelyhood detection. $P(r_1, r_2|s = \sqrt{E}) > P(r_1, r_2|s = -\sqrt{E})$

$$P(r_1 = s + n_1, r_2 = n_1 + n_2|s = \sqrt{E}) > P(r_1 = s + n_1, r_2 = n_1 + n_2|s = -\sqrt{E})$$

$$P(r_1 = \sqrt{E} + n_1, r_2 = n_1 + n_2|s = \sqrt{E}) > P(r_1 = -\sqrt{E} + n_1, r_2 = n_1 + n_2|s = -\sqrt{E})$$

You want to rearrange the algebra to isolate the two noise signals since they are independent and can be split.

$$P(n_1 = r_1 - \sqrt{E})P(n_2 = r_2 - n_1) > P(n_1 = r_1 + \sqrt{E})P(n_2 = r_2 - n_1)$$

$$P(n_1 = r_1 - \sqrt{E})P(n_2 = r_2 - r_1 + \sqrt{E}) > P(n_1 = r_1 + \sqrt{E})P(n_2 = r_2 - r_1 - \sqrt{E})$$

From this you see that this becomes independent gaussian distributions which we can compare.

Information Theory

Individual information

$$\text{mutual information} = I(x_i; y_j) = \log \frac{P(x_i|y_j)}{P(x_i)}$$

$$\text{self info} = I(x_i) = -\log P(x_i)$$

$$\text{conditional info} = I(x_i) = -\log P(x_i|y_i)$$

$$I(x_i; y_j) = I(x_i) - I(x_i|y_j)$$

Average information

$$I(X; Y) = \sum_{i=1}^n \sum_{j=1}^m P(x_i, y_j)I(x_i, y_j) \quad H(X) = -\sum_{i=1}^n P(x_i)I(x_i)$$

$$H(X|Y) = -\sum_{i=1}^n \sum_{j=1}^m P(x_i, y_j)P(x_i|y_j) \quad I(X, Y) = H(X) - H(X|Y)$$

$$H(X_1, X_2, X_3 \dots) = H(X_1) + H(X_2|X_1) + \dots$$

Rate distortion function

$$R_q(D) = \frac{1}{2} \log_2 \frac{\sigma_x^2}{D} \quad (0 < D < \sigma_x^2) \quad R^*(D) \leq R(D) \leq \frac{1}{2} \log_2 \frac{\sigma_x^2}{D}$$

$$R^*(D) = H(X) - \frac{1}{2} \log_2 2\pi e D \text{ pg 109}$$

Channel Capacity

$$C = \max_{P(x_j)} \sum_{i=1}^{n-1} \sum_{j=1}^{m-1} I(X, Y) = \max_{P(x_j)} \sum_{i=1}^{n-1} \sum_{j=1}^{m-1} p(x_j, y_i) \log \frac{P(y_i|x_j)}{p(y_i)}$$

with power and bandwidth constrain, shannon's theorem

$$C = W \log \left(1 + \frac{P_{avg}}{WN_0} \right) \text{ since } P_{avg} = CE_b \text{ this could also be written as}$$

$$r = \frac{C}{W} = \log_2 \left(1 + \frac{C}{W} \frac{E_b}{N_0} \right) \text{ where } r \text{ is measured in bits/second/Hz. If we solve for SNR we have}$$

$$\frac{E_b}{N_0} = \frac{2^{C/W} - 1}{C/W} = \frac{2^r - 1}{r} \quad \frac{E_b}{N_0} = \lim_{C/W \rightarrow \infty} \frac{2^{C/W} - 1}{C/W} = \ln 2 = -1.6dB$$

This means that as SNR approach $\ln 2$, the bandwidth required approaches infinity.

R₀ Theorem , Error bounds

Average error for random coding $\leq 2^{n(R_0-R)}$ $P(E|X_m) \leq \sum_{m'=1}^M \int p^\lambda(y|x_{m'})p^{1-\lambda}(y|x_m)dy$

Special case when $\lambda = \frac{1}{2}$ we have

$$ED_\lambda(1-\lambda > 2) = \int (\sum_x p(x) \sqrt{P(y|x)})^2$$

Special case when $\lambda = \frac{1}{2}$ and $p(x) = \text{uniform}$, $\|x\| = Q$, and $p(x) = \frac{1}{Q}$

$$ED_\lambda(1-\lambda > 2) = \frac{1}{Q^2} \int \left(\sum_{i=1}^Q \sqrt{p(y|x_i)} \right)^2 dy$$

$$R_0 = 2\log_2 Q - \log_2 \int \left(\sum_{i=1}^Q \sqrt{p(y|x_i)} \right)^2 dy$$

$$R_0 \text{ for Binary Symmetric Channel} = \log_2 \frac{2}{1+2\sqrt{p(1-p)}}$$

$$R_0 \text{ for AWGN BSC} = \log_2 \frac{2}{1+e^{-E_c/N_0}}$$

Linear block code

You start off with a Generator matrix where $X_m G = \text{Code}$, where x are the bits, G is the transitional matrix. In the reduce form, $G = [I_k|P]$. We use the parity bit H to check for errors. $H = [-P|I_{n-k}]$

Hard code procedure

if $Y = C_{code} + e_{error}$ and $YH = (C_m + e)H' = eH' = S$ where S are the syndrom bits. There will be $2^{\#\text{syndrombits}}$ number of errors, we need to find all the error patterns that corresponds with the right syndrom. To find the error patterns, we first use up all the fewest bits, (1 bits) and move to different combinations of 2 and 3 .. bits. With this info, we can construct the standard array and the syndrom table. When we multiply H to Y, we get the syndroms and we look up the syndrom table for the possible error. We then subtract the error from the Code.

Rate relationship

coding gain = $R_c d_{min}$

d_{min} is the number of columns in the H matrix that are required to have an dependent columns.

if k = input bits and n = output bits $R_c = \frac{k}{n}$

$E = E_b k = E_c n$, $E_b R_c = E_c$

Probability error

where d_{min}^H is the Hamming distance

Soft decision $P_e \leq (m-1)Q \left(\sqrt{2R_c d_{min}^H \frac{E_b}{N_0}} \right)$

Hard decision $P_e \leq (M-1)[4p(1-p)]^{\frac{d_{min}^H}{2}}$

Hamming code(Type of systematic code)

$(n, k) = (2^m - 1, 2^m - 1 - m)$ where H (parity check) is the complete set of permutations

Convolutional Code

Transfer Function

-The transfer function is defined as $T(D) = \frac{X_e}{X_a}$ and when you solve for the equation you would end up with an expression.

-If you start from 000, a_d tells you the number of paths that has the distance d. So if you look below you have 1 branch with distance 6, 2 branches with distance 8 and etc.

$$\begin{aligned}
X_c &= D^3 X_a + D X_b & X_b &= D X_c + D X_d \\
X_d &= D^2 X_c + D^2 X_d & X_e &= D^2 X_b \\
T(D) &= \frac{D^6}{1-2D^2} = D^6 + 2D^8 + 4D^{10} + 8D^{12} + \dots = \sum_{d=6}^{\infty} a_d D^d
\end{aligned}$$

Probability error

$$\text{if we have } T(D) = \sum_{d=d_{free}} a_d D^d$$

$$P_e \leq \sum_{d=d_{free}}^{\infty} a_d \frac{1}{2} e^{-R_c d \frac{E_b}{N_0}} = \frac{1}{2} T(D) \Big|_{D=e^{-R_c d \frac{E_b}{N_0}}} \approx a_d \frac{1}{2} e^{-R_c d \frac{E_b}{N_0}}$$

Trellis Coded Modulation

1. Draw constellation 2. label constellation with numbers sequentially 3. set partition n_1 times. 4. Use the size v convolutional code to decide $\#$ v states needed $\# = 2^{k(l-1)}$ where k is num v input and l is num of boxes in convolutional code. 5. Use set partition to randomly assign the trellis. (the branches from the trellis must have the same parents) 6. Convert the branch num into binary bits 7. notice that branches from the same state have equal bits, keep the equal bits and make the branches single branches. 8 Now we have the state transition trellis for the convolutional code.

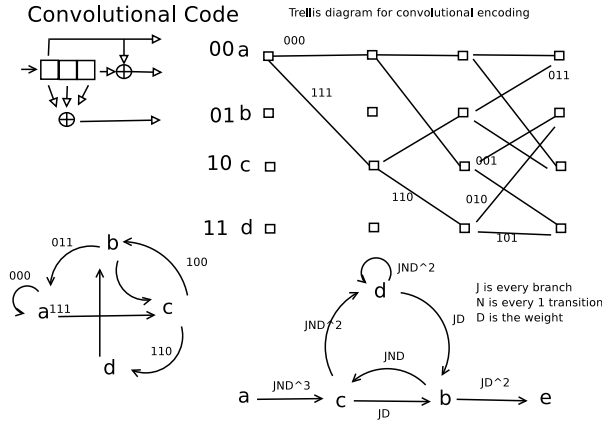


Figure 1: Convolutional code

Bandlimited Channels

$$S = \sum_{n=0}^{\infty} I_n g(t - nT) \rightarrow C(t) + n(t) = r(t) = I_n h(t - NT) + n(t)$$

$$h(t) = g(t) * c(t) \text{ and } y_k = I_k + \sum_{n=0}^{\infty} I_n X_{k-n} + V_k$$

to avoid ISI we need to satisfy $\sum_{m=-\infty}^{\infty} X(f + \frac{m}{T}) = T$

SNR

$$\frac{E_b}{N_0} = \frac{P_b T}{N_0} = \frac{P_b(1-\beta)}{N_0 W}$$

Raise Cosine function

$$X_{rc}(f) = T$$

$$X_{rc}(f) = \frac{T}{2} \left\{ 1 + \cos \left[\frac{\pi T}{\beta} \left(|f| - \frac{1-\beta}{2T} \right) \right] \right\}$$

$$X_{rc}(f) = 0$$

$$0 \leq f \leq \frac{1-\beta}{2T}$$

$$\frac{1-\beta}{2T} \leq |f| \leq \frac{1+\beta}{2T}$$

$$|f| \geq \frac{1+\beta}{2T}$$

Probability of error

$$P_M = 1 - (1 - P_{\sqrt{M}})^2 \quad P_{\sqrt{M}} = 2 \left(1 - \frac{1}{\sqrt{M}} \right) Q \left(\sqrt{\frac{3E_{avg}}{(M-1)N_0}} \right)$$

The key equation to note is $\frac{1+\beta}{2T} = \frac{BW}{2}$ Where $\frac{1}{T}$ is equal to the symbol Rate

(R_S) measured in symbols/second. If you need to transmit 9600 bits/s with 8PSK,(3 bits) You would have $\frac{9600bits/s}{3bits/symbol}$. In general we want a β value that's greater than .5.

Example:

1. If we have 4000 Hz voice-bandpass channel, what's the bit rate if we use BPSK? (let beta = 1/2)

$$\frac{1}{2T}(1 + 1/2) = 2000 \text{ from this we get } \frac{1}{T} = 2666.666symbol/sec$$

Since this is BPSK with 1 bit per symbol this means we also have 2666 bits/sec

2. If we use 8QAM?

We know that the symbol rate is 2666 symbol/sec, with 3 bits we have $2666*3 = 8001$ bits/sec

3. If we use 4FSK?

We divide the total bandwidth by 4 $\frac{4000Hz}{4} = \frac{1}{T} = 1000symbol/sec$ with this value we can multiply by 2 bits to get the bit rate.