

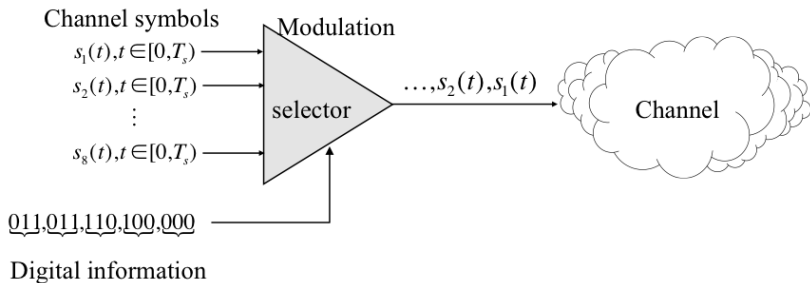
Digital Communications

Chapter 3: Digital Modulation Schemes

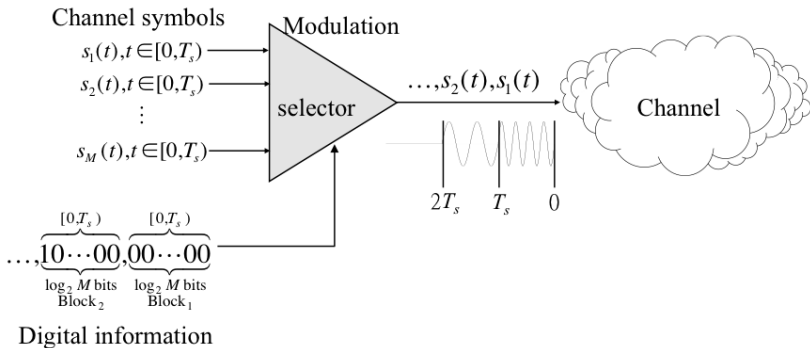
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3.1 Representation of digitally modulated signals



Note that the channel symbols are bandpass signals.



- Memoryless modulation: $s_{m_\ell}(t)$, $m_\ell \in \{1, 2, \dots, M\}$,
 $m_\ell = \text{function of Block}_\ell$

- Modulation with memory: $s_{m_\ell}(t)$,
 $m_\ell = \text{function of } (\text{Block}_\ell, \text{Block}_{\ell-1}, \dots, \text{Block}_{\ell-(L-1)})$

$L = \text{Constraint length of modulation (with memory)}$

Terminology

Signal $s_m(t)$, $1 \leq m \leq M$, $t \in [0, T_s)$

- Signaling interval: T_s (For convenience, we will sometimes use T instead.)
- Signaling rate (or symbol rate): $R_s = \frac{1}{T_s}$
- (Equivalent) Bit interval: $T_b = \frac{T_s}{\log_2 M}$
- (Equivalent) Bit rate: $R_b = \frac{1}{T_b} = R_s \log_2 M$
- Average signal energy (assume equal-probable in message m)

$$\mathcal{E}_{\text{avg}} = \frac{1}{M} \sum_{m=1}^M \int_0^{T_s} |s_m(t)|^2 dt$$

- (Equivalent) Average bit energy: $\mathcal{E}_{\text{bavg}} = \frac{\mathcal{E}_{\text{avg}}}{\log_2 M}$
- Average power: $P_{\text{avg}} = \frac{\mathcal{E}_{\text{avg}}}{T_s} = R_s \mathcal{E}_{\text{avg}} = \frac{\mathcal{E}_{\text{bavg}}}{T_b} = R_b \mathcal{E}_{\text{bavg}}$

3.2 Memoryless modulation methods

Example studies of memoryless modulation

- Digital pulse amplitude modulated (PAM) signals (Amplitude-shift keying or ASK)
- Digital phase-modulated (PM) signals (Phase shift keying or PSK)
- Quadrature amplitude modulated (QAM) signals
- Multidimensional modulated signals
 - Orthogonal
 - Bi-orthogonal
- Simplex signals

M-ary pulse amplitude modulation (M-PAM)

PAM bandpass waveform

$$s_m(t) = \mathbf{Re} \left\{ A_m g(t) e^{i2\pi f_c t} \right\} = A_m g(t) \cos(2\pi f_c t), \quad t \in [0, T_s),$$

where $A_m = (2m - 1 - M)d$, and $m = 1, 2, \dots, M$

Example 1 (M=4)

$$\begin{cases} s_1(t) &= -3 \cdot d \cdot g(t) \cdot \cos(2\pi f_c t) \\ s_2(t) &= -1 \cdot d \cdot g(t) \cdot \cos(2\pi f_c t) \\ s_3(t) &= +1 \cdot d \cdot g(t) \cdot \cos(2\pi f_c t) \\ s_4(t) &= +3 \cdot d \cdot g(t) \cdot \cos(2\pi f_c t) \end{cases}$$

The amplitude difference between two adjacent signals = 2d.

$$s_m(t) = \mathbf{Re} \{ A_m g(t) e^{i2\pi f_c t} \} = A_m g(t) \cos(2\pi f_c t), \quad t \in [0, T_s)$$

- $g(t)$ is the **pulse shaping function**.
- T_s is usually assumed to be a multiple of $\frac{1}{f_c}$ in principle.

Vectorization of M-PAM signals (Gram-Schmidt)

$$\phi_1(t) = \frac{g(t)}{\|g(t)\|} \sqrt{2} \cos(2\pi f_c t) = \frac{g(t)}{\sqrt{\mathcal{E}_g}} \sqrt{2} \cos(2\pi f_c t)$$

$$\mathbf{s}_m = \left[\frac{A_m}{\sqrt{2}} \cdot \|g(t)\| \right], \text{ a one-dimensional vector}$$

By the **correct** Gram-Schmidt procedure,

$$\begin{aligned} \phi_1(t) &= \frac{g(t) \cos(2\pi f_c t)}{\|g(t) \cos(2\pi f_c t)\|} \\ &\neq \frac{g(t) \cos(2\pi f_c t)}{\|g(t)\| \cdot \frac{1}{\sqrt{T_s}} \|\cos(2\pi f_c t)\|} = \frac{g(t) \cos(2\pi f_c t)}{\|g(t)\| \sqrt{1/2}} \end{aligned}$$

The idea behind the above derivation is to **single out** “ $\|g(t)\|$ ” in the expression! This justifies the necessity of introducing the lowpass equivalent signal where the influence of f_c has been relaxed.

For a time-limited signal, we can only claim $\mathcal{E}_{x_e} \approx 2\mathcal{E}_x$!

$$\begin{aligned}
\|\phi_1(t)\|^2 &= \frac{2}{\|g(t)\|^2} \int_0^{T_s} g^2(t) \cos^2(2\pi f_c t) dt \\
&= \frac{2}{\|g(t)\|^2} \int_0^{T_s} g^2(t) \left[\frac{1 + \cos(4\pi f_c t)}{2} \right] dt \\
&= \frac{1}{\|g(t)\|^2} \int_0^{T_s} g^2(t) dt \\
&\quad + \frac{1}{\|g(t)\|^2} \int_0^{T_s} g^2(t) \cos(4\pi f_c t) dt \\
&\approx \frac{1}{\|g(t)\|^2} \int_0^{T_s} g^2(t) dt = 1
\end{aligned}$$

If $g(t)$ is constant for $t \in [0, T_s)$ and T_s is a multiple of $\frac{1}{f_c}$, then the above “ \approx ” becomes “ $=$.”

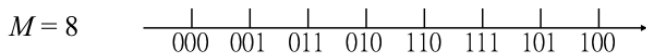
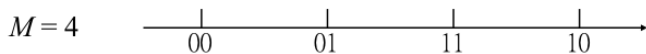
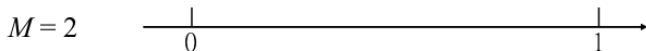
Based on the “pseudo”-vectorization,

- Transmission energy of M -PAM signals

$$\mathcal{E}_m = \int_0^{T_s} |s_m(t)|^2 dt \approx \frac{A_m^2 \|g(t)\|^2}{2} = \frac{1}{2} A_m^2 \mathcal{E}_g$$

- Error consideration
 - The most possible error is the erroneous selection of an adjacent amplitude to the transmitted signal amplitude.
 - Therefore, the mapping (from bit pattern to channel symbol) is assigned such that the adjacent signal amplitudes differ by exactly one bit. (Gray encoding)
 - In such way, the most possible bit error pattern (caused by the noise) is a single bit error.

Gray code (Signal space diagram : one dimension)



Euclidean distance

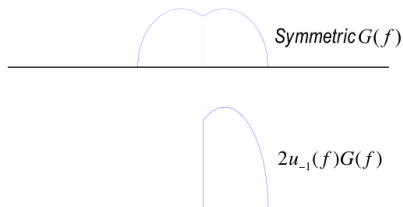
$$\begin{aligned}\|s_m(t) - s_n(t)\| &\approx \left| \frac{A_m \|g(t)\|}{\sqrt{2}} - \frac{A_n \|g(t)\|}{\sqrt{2}} \right| \\ &= \frac{\|g(t)\|}{\sqrt{2}} |(2m - 1 - M)d - (2n - 1 - M)d| \\ &= d\sqrt{2} \|g(t)\| |m - n|\end{aligned}$$

Single side band (SSB) PAM

- 1 $g(t)$ is real $\Leftrightarrow G(f)$ is Hermitian symmetric.
- 2 Consequently, the previous PAM is based on the double side band (DSB) transmission which requires twice the bandwidth.
- 3 Recall

$$\mathcal{F}^{-1}\{u_{-1}(f)G(f)\} = \frac{1}{2}[g(t) + j\hat{g}(t)] = g_+(t)$$

where $\hat{g}(t)$ is the Hilbert transform of $g(t)$.



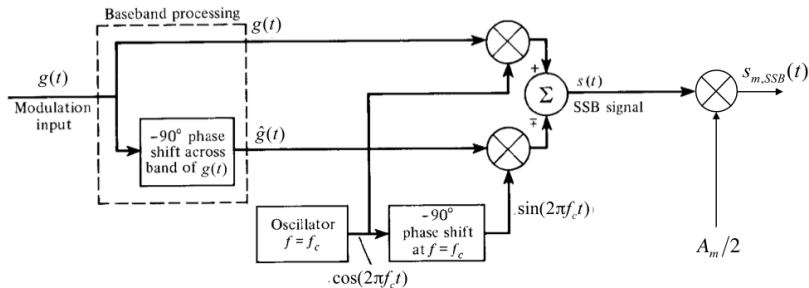
$$s_{m,SSB}(t) = \mathbf{Re} \{ A_m g_+(t) e^{i2\pi f_c t} \}$$

$$\phi_{1,SSB}(t) \approx \frac{\mathbf{Re} \{ A_m g_+(t) e^{i2\pi f_c t} \}}{\|g_+(t)\| \cdot \frac{1}{\sqrt{T_s}} \|\mathbf{Re} \{ A_m e^{i2\pi f_c t} \}\|} = \frac{\mathbf{Re} \{ \sqrt{2} g_+(t) e^{i2\pi f_c t} \}}{\|g_+(t)\|}$$

$$s_{m,SSB} = \left[\frac{A_m}{\sqrt{2}} \|g_+(t)\| \right]$$

$$\begin{aligned} & \|g_+(t)\|^2 \cdot \int_0^{T_s} \phi_{1,SSB}^2(t) dt \\ &= 2 \int_0^{T_s} \mathbf{Re} \{ g_+(t) e^{i2\pi f_c t} \}^2 dt \\ &= \frac{1}{2} \int_0^{T_s} [g_+(t) e^{i2\pi f_c t} + g_+^*(t) e^{-i2\pi f_c t}]^2 dt \\ &= \frac{1}{2} \int_0^{T_s} [|g_+(t)| e^{i2\pi f_c t + \angle g_+(t)} + |g_+(t)| e^{-i2\pi f_c t - \angle g_+(t)}]^2 dt \\ &= \int_0^{T_s} |g_+(t)|^2 dt + \int_0^{T_s} |g_+(t)|^2 \cos[4\pi f_c t + 2\angle g_+(t)] dt \\ &\approx \int_0^{T_s} |g_+(t)|^2 dt = \|g_+(t)\|^2 \end{aligned}$$

$$\begin{aligned}
 s_{m,SSB}(t) &= \text{Re} \left\{ \frac{A_m}{2} [g(t) \pm j \hat{g}(t)] e^{j2\pi f_c t} \right\} \\
 &= \frac{A_m}{2} g(t) \cos(2\pi f_c t) \mp \frac{A_m}{2} \hat{g}(t) \sin(2\pi f_c t)
 \end{aligned}$$



$$\|g_+(t)\|^2 = \left\| \frac{1}{2}g(t) + \imath \frac{1}{2}\hat{g}(t) \right\|^2 = \frac{1}{2} \|g(t)\|^2$$

Recall from Slide 2-22, $x_+(t) = \frac{1}{2}(x(t) + \imath \hat{x}(t))$ and $\mathcal{E}_x = 2\mathcal{E}_{x_+}$.

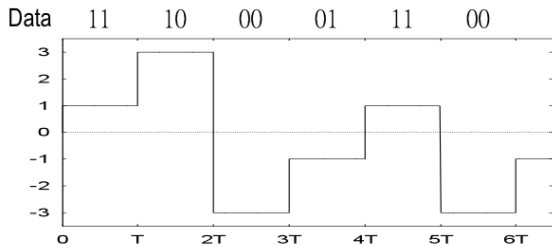
To summarize

$$\begin{cases} \phi_{1,(DSB)}(t) &= \frac{g(t)}{\|g(t)\|} \sqrt{2} \cos(2\pi f_c t) \\ s_{m,(DSB)} &= \frac{A_m}{\sqrt{2}} \|g(t)\| \end{cases}$$

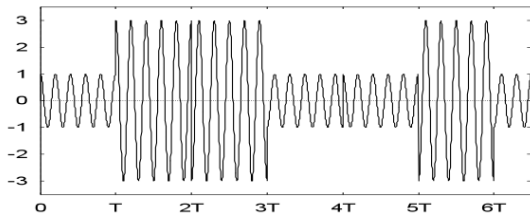
$$\begin{cases} \phi_{1,SSB}(t) &= \mathbf{Re} \left\{ \frac{g_+(t)}{\|g_+(t)\|} \sqrt{2} e^{\imath 2\pi f_c t} \right\} \\ s_{m,SSB} &= \frac{A_m}{\sqrt{2}} \|g_+(t)\| \end{cases}$$

2-level PAM signals are particularly named *antipodal* signals.
(± 1 signals)

Applications of PAM



Baseband PAM signal



Bandpass PAM signal

$$T_s = T$$

Phase-modulation (PM)

Bandpass PM

$$\begin{aligned} s_m(t) &= \mathbf{Re} \left[g(t) e^{i2\pi(m-1)/M} e^{i2\pi f_c t} \right] \\ &= g(t) \cos(2\pi f_c t + \theta_m) \\ &= \underbrace{\cos(\theta_m) g(t) \cos(2\pi f_c t)}_{\phi_1} - \underbrace{\sin(\theta_m) g(t) \sin(2\pi f_c t)}_{\phi_2} \end{aligned}$$

where $\theta_m = 2\pi(m-1)/M$, $m = 1, 2, \dots, M$

Example 2 (M=4)

$$\begin{cases} s_1(t) = g(t) \cos(2\pi f_c t) \\ s_2(t) = g(t) \cos(2\pi f_c t + \pi/2) \\ s_3(t) = g(t) \cos(2\pi f_c t + \pi) \\ s_4(t) = g(t) \cos(2\pi f_c t + 3\pi/2) \end{cases}$$

Signal space of PM signals

$$\begin{cases} \phi_1(t) \approx \frac{g(t)}{\|g(t)\|} \sqrt{2} \cos(2\pi f_c t) \\ \phi_2(t) \approx -\frac{g(t)}{\|g(t)\|} \sqrt{2} \sin(2\pi f_c t) \end{cases}$$

\implies

$$\mathbf{s}_m = \left[\frac{\|g(t)\|}{\sqrt{2}} \cos(\theta_m), \frac{\|g(t)\|}{\sqrt{2}} \sin(\theta_m) \right]$$

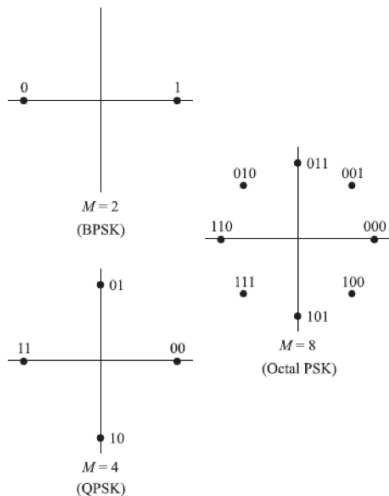
- Transmission energy of PM Signals

$$\mathcal{E}_m = \int_0^T s_m^2(t) dt \approx \frac{\|g(t)\|^2}{2} [\cos^2(\theta_m) + \sin^2(\theta_m)] = \frac{\mathcal{E}_g}{2}$$

Advantages of PM signals : Equal energy for every channel symbol

- Error consideration
 - The most possible error is the erroneous selection of an adjacent phase of the transmitted signal phase.
 - Therefore, we assign the mapping from bit pattern to channel symbol as the adjacent signal phases differ only by one bit. (Gray encoding)
 - The most possible bit error pattern caused by the noise is a single-bit error.

Signal space diagram of PM with Gray code



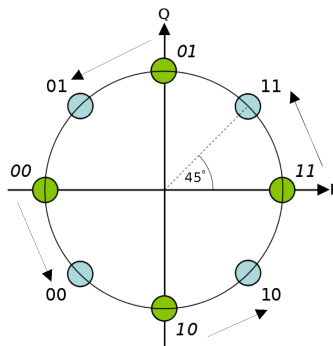
$$s_m = \left[\frac{\|g(t)\|}{\sqrt{2}} \cos(\theta_m), \frac{\|g(t)\|}{\sqrt{2}} \sin(\theta_m) \right]$$

- Euclidean distance

$$\begin{aligned} & \|s_m(t) - s_n(t)\| \\ &= \frac{\|g(t)\|}{\sqrt{2}} \sqrt{|\cos(\theta_m) - \cos(\theta_n)|^2 + |\sin(\theta_m) - \sin(\theta_n)|^2} \\ &= \|g(t)\| \sqrt{1 - \cos(\theta_m - \theta_n)} \end{aligned}$$

$\pi/4$ -QPSK

A variant of 4-phase PSK (QPSK), named $\pi/4$ -QPSK, is obtained by introducing an additional $\pi/4$ phase shift in the carrier phase in each symbol interval.



Quadrature amplitude modulation (QAM)

Bandpass QAM

$$s_m(t) = x_i(t) \cos(2\pi f_c t) - x_q(t) \sin(2\pi f_c t)$$

where $x_i(t)$ and $x_q(t)$ are quadrature components. Let

$x_i(t) = A_{mi}g(t)$ and $x_q(t) = A_{mq}g(t)$; then bandpass QAM is

$$s_m(t) = A_{mi}g(t) \cos(2\pi f_c t) - A_{mq}g(t) \sin(2\pi f_c t)$$

Advantage: Transmit more digital information by using both quadrature components as information carriers. As a result, the transfer rate of digital data is doubled.

Vectorization of QAM signals

$$s_m(t) = A_{mi} \underbrace{g(t) \cos(2\pi f_c t)}_{\phi_1} - A_{mq} \underbrace{g(t) \sin(2\pi f_c t)}_{\phi_2}$$

$$\begin{cases} \phi_1(t) \approx \frac{g(t)}{\|g(t)\|} \sqrt{2} \cos(2\pi f_c t) \\ \phi_2(t) \approx -\frac{g(t)}{\|g(t)\|} \sqrt{2} \sin(2\pi f_c t) \end{cases}$$

$$\implies \mathbf{s}_m = \left[\frac{A_{mi}}{\sqrt{2}} \|g(t)\|, \frac{A_{mq}}{\sqrt{2}} \|g(t)\| \right]$$

$$s_m = \left[\frac{A_{mi}}{\sqrt{2}} \|g(t)\|, \frac{A_{mq}}{\sqrt{2}} \|g(t)\| \right]$$

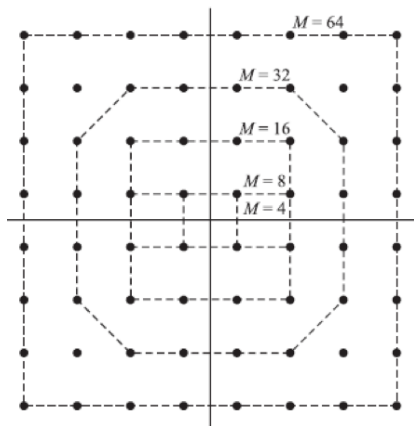
- Transmission energy of QAM signals

$$\begin{aligned} \mathcal{E}_m &= \int_0^T s_m^2(t) dt \\ &= \frac{1}{2} \|g(t)\|^2 A_{mi}^2 + \frac{1}{2} \|g(t)\|^2 A_{mq}^2 \\ &= \frac{1}{2} \|g(t)\|^2 (A_{mi}^2 + A_{mq}^2) \\ &= \frac{1}{2} \mathcal{E}_g (A_{mi}^2 + A_{mq}^2) \end{aligned}$$

- Euclidean Distance

$$\|s_m(t) - s_n(t)\| = \frac{\sqrt{\mathcal{E}_g}}{\sqrt{2}} \sqrt{|A_{mi} - A_{ni}|^2 + |A_{mq} - A_{nq}|^2}$$

Signal space diagram for rectangular QAM



$$s_m = \left[\frac{A_{mi}}{\sqrt{2}} \|g(t)\|, \frac{A_{mq}}{\sqrt{2}} \|g(t)\| \right],$$

where $A_{mi}, A_{mq} \in \{(2m-1-\sqrt{M}) : m=1, 2, \dots, \sqrt{M}\}$

- Minimum Euclidean distance (of square QAM)

$$\min_{m \neq n} \sqrt{\frac{\mathcal{E}_g}{2}} \sqrt{\underbrace{|A_{mi} - A_{ni}|^2}_{=4} + \underbrace{|A_{mq} - A_{nq}|^2}_{=0}} = \sqrt{2\mathcal{E}_g}$$

- Average symbol energy (of square QAM)

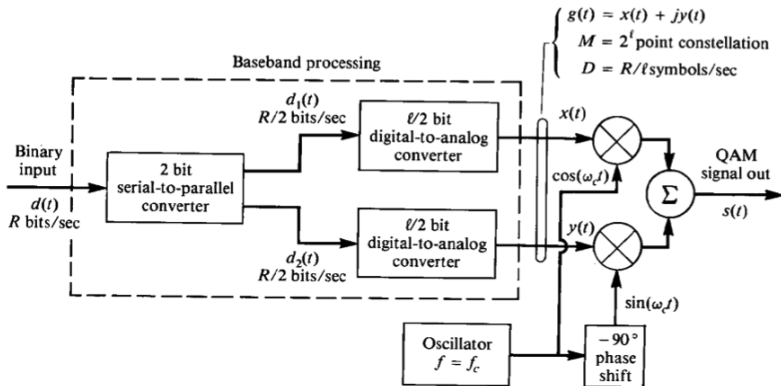
$$\mathcal{E}_{avg} = \frac{1}{M} \frac{\mathcal{E}_g}{2} \sum_{m=1}^{\sqrt{M}} \sum_{n=1}^{\sqrt{M}} (A_{mi}^2 + A_{nq}^2) = \frac{\mathcal{E}_g}{2M} \frac{2M(M-1)}{3} = \frac{M-1}{3} \mathcal{E}_g$$

- Average bit energy (of square QAM)

$$\mathcal{E}_{bavg} = \frac{M-1}{3 \log_2 M} \mathcal{E}_g$$

Example of applications of square QAM

- CCITT V.22 modem
 - Serial binary, asynchronous or synchronous, full duplex, dial-up
 - 2400 bps or 600 baud (symbols/sec)
 - QAM, 16-point rectangular-type signal constellation



Alternative viewpoint of QAM

QAM = PM (PSK) + PAM (ASK)

- Use both amplitude and phase as digital information bearers.

$$s_m(t) = \mathbf{Re} \left[V_{m1} e^{i\theta_{m2}} g(t) e^{i2\pi f_c t} \right] = V_{m1} g(t) \cos(2\pi f_c t + \theta_{m2})$$

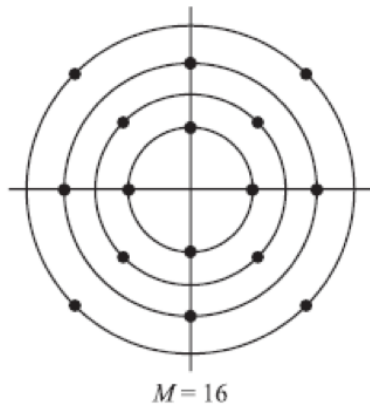
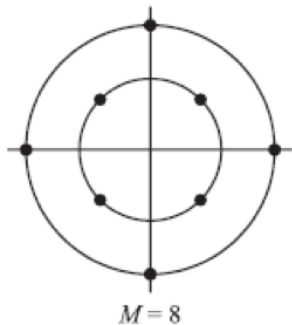
- Compare with the previous viewpoint

$$\begin{aligned} s_m(t) &= A_{mi} g(t) \cos(2\pi f_c t) - A_{mq} g(t) \sin(2\pi f_c t) \\ &= V_{m1} g(t) \cos(2\pi f_c t + \theta_{m2}) \end{aligned}$$

where $V_{m1} = \sqrt{A_{mi}^2 + A_{mq}^2}$ and $\theta_{m2} = \tan^{-1}(A_{mq}/A_{mi})$

- There is a one-to-one correspondence mapping from (A_{mi}, A_{mq}) domain to (V_{m1}, θ_{m2}) domain.

Signal space for non-rectangular QAM (AM-PSK)



Multi-dimensional signals

- PAM : one-dimensional
- PM : two-dimensional
- QAM : two-dimensional

- How to create three or higher dimensional signal?
 - 1 Subdivision of time
Example. N time slots can be used to form $2N$ vector basis elements (each has two quadrature bearers.)

 - 2 Subdivision of frequency
Example. N frequency subbands can be used to form $2N$ vector basis elements (each has two quadrature bearers.)

 - 3 Subdivision of both time and frequency

Frequency shift keying or FSK

- Subdivision of frequency
- Bandpass orthogonal multidimensional signals (Frequency shift keying or FSK)

$$\begin{aligned} s_m(t) &= \operatorname{Re} \left[\sqrt{\frac{2\mathcal{E}}{T}} e^{i2\pi(m\Delta f)t} e^{i2\pi f_c t} \right] \\ &= \sqrt{\frac{2\mathcal{E}}{T}} \cos(2\pi f_c t + 2\pi(m\Delta f)t) \end{aligned}$$

- Vectorization of FSK signals under orthogonality conditions (introduced in next few slides)

$$\phi_m(t) = \frac{1}{\sqrt{\mathcal{E}}} s_m(t) \text{ and } \mathbf{s}_m = [0, \dots, 0, \underbrace{\sqrt{\mathcal{E}}}_{\substack{\text{mth} \\ \text{position}}}, 0, \dots, 0]^T$$

Crosscorrelations of FSK signals

$$s_{m,\ell}(t) = \sqrt{\frac{2\mathcal{E}}{T}} e^{i2\pi(m\Delta f)t} \quad \text{and} \quad \|s_{m,\ell}(t)\| = \sqrt{2\mathcal{E}}$$

$$\begin{aligned} \rho_{mn,\ell} &= \frac{\langle s_{m,\ell}(t), s_{n,\ell}(t) \rangle}{\|s_{m,\ell}(t)\| \cdot \|s_{n,\ell}(t)\|} = \frac{1}{T} \int_0^T e^{i2\pi(m-n)\Delta f \cdot t} dt \\ &= \text{sinc}[T(m-n)\Delta f] e^{i\pi T(m-n)\Delta f} \end{aligned}$$

$$\begin{aligned} \frac{\langle s_m(t), s_n(t) \rangle}{\|s_m(t)\| \|s_n(t)\|} &= \text{Re}\{\rho_{mn,\ell}\} = \frac{\sin(\pi T(m-n)\Delta f)}{\pi T(m-n)\Delta f} \cos(\pi T(m-n)\Delta f) \\ &= \text{sinc}(2T(m-n)\Delta f) \end{aligned}$$

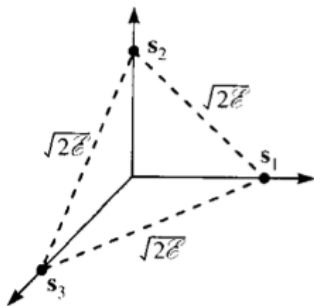
When $\Delta f = \frac{k}{2T}$, $\text{Re}\{\rho_{mn,\ell}\} = 0$ for $m \neq n$. In other words, the minimum frequency separation between adjacent (bandpass) signals for orthogonality is $\Delta f = \frac{1}{2T}$.

- Transmission energy of FSK signals

$$\mathcal{E}_m = \int_0^T |s_m(t)|^2 dt = \mathcal{E}$$

⇒ Equal transmission power for each channel symbol

- Signal space diagram for FSK



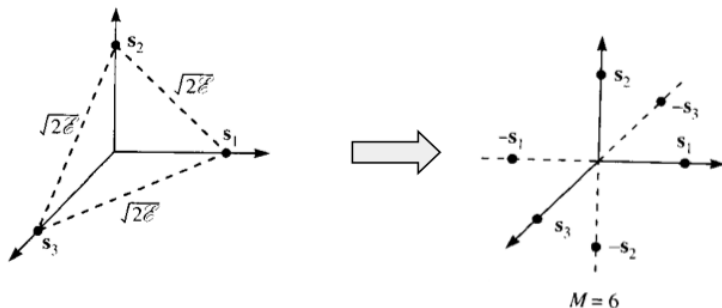
Euclidean distance between FSK signals

Equal distance between signals

$$[\mathbf{s}_1 \quad \mathbf{s}_2 \quad \cdots \quad \mathbf{s}_M] = \begin{bmatrix} \sqrt{\mathcal{E}} & 0 & \cdots & 0 \\ 0 & \sqrt{\mathcal{E}} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \sqrt{\mathcal{E}} \end{bmatrix}$$

$$\|\mathbf{s}_m - \mathbf{s}_n\| = \sqrt{2\mathcal{E}}$$

Biorthogonal multidimensional FSK signals



- Transmission energy for biorthogonal FSK signals

$$\mathcal{E}_m = \int_0^T |s_m(t)|^2 dt = \mathcal{E}$$

Still, equal transmission power for each channel symbol.

- Cross-correlation of baseband biorthogonal FSK signals

$$s_{m,\ell}(t) = \text{sgn}(m) \sqrt{\frac{2\mathcal{E}}{T}} e^{i2\pi|m|(\Delta f)t}, \quad m = \pm 1, \pm 2, \dots, \pm M/2$$

$$\rho_{mn,\ell} = \begin{cases} 1, & m = n \\ -1, & m = -n \\ 0, & \text{otherwise} \end{cases}$$

Euclidean distance between signals

$$\begin{aligned} & [\mathbf{s}_{-1} \quad \cdots \quad \mathbf{s}_{-M/2} \quad \mathbf{s}_1 \quad \cdots \quad \mathbf{s}_{M/2}] \\ &= \begin{bmatrix} -\sqrt{\mathcal{E}} & 0 & \cdots & 0 & \sqrt{\mathcal{E}} & 0 & \cdots & 0 \\ 0 & -\sqrt{\mathcal{E}} & \cdots & 0 & 0 & \sqrt{\mathcal{E}} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & -\sqrt{\mathcal{E}} & 0 & 0 & \cdots & \sqrt{\mathcal{E}} \end{bmatrix} \end{aligned}$$

Hence

$$\|\mathbf{s}_m - \mathbf{s}_n\| = \begin{cases} \sqrt{2\mathcal{E}} & \text{if } m \neq -n \\ 2\sqrt{\mathcal{E}} & \text{if } m = -n \end{cases}$$

Simplex signals

Given the vector representations of **orthogonal** and **equal-power** channel symbols (such as FSK)

$$\mathbf{s}_m = [a_{m1}, a_{m2}, \dots, a_{mk}]$$

for $m = 1, 2, \dots, M$, its **center** (of gravity under equal prior probability assumption) is

$$\mathbf{c} = \left[\frac{1}{M} \sum_{m=1}^M a_{m1}, \frac{1}{M} \sum_{m=1}^M a_{m2}, \dots, \frac{1}{M} \sum_{m=1}^M a_{mk} \right]$$

Define new channel symbol as

$$\mathbf{s}'_m = \mathbf{s}_m - \mathbf{c}$$

Then $\{\mathbf{s}'_1, \mathbf{s}'_2, \dots, \mathbf{s}'_M\}$ is called the **simplex signal**.

Transmission energy of simplex signals

$$\begin{aligned}\mathcal{E}'_m &= \int_0^T |s'_m(t)|^2 dt \\ &= \|\mathbf{s}_m - \mathbf{c}\|^2 \\ &= \|\mathbf{s}_m\|^2 + \|\mathbf{c}\|^2 - \langle \mathbf{s}_m, \mathbf{c} \rangle - \langle \mathbf{c}, \mathbf{s}_m \rangle \quad (\mathbf{c} = \frac{1}{M} \sum_{i=1}^M \mathbf{s}_i) \\ &= \|\mathbf{s}_m\|^2 + \|\mathbf{c}\|^2 - \frac{1}{M} \sum_{i=1}^M \langle \mathbf{s}_m, \mathbf{s}_i \rangle - \frac{1}{M} \sum_{i=1}^M \langle \mathbf{s}_i, \mathbf{s}_m \rangle \\ &= \|\mathbf{s}_m\|^2 + \frac{1}{M} \|\mathbf{s}_m\|^2 - \frac{2}{M} \|\mathbf{s}_m\|^2 \quad (\text{by orthogonality} \\ &\quad \text{and equal-power}) \\ &= \left(1 - \frac{1}{M}\right) \|\mathbf{s}_m\|^2\end{aligned}$$

- The transmission energy of a signal is reduced by “simplexing” it.

Crosscorrelation of simplex signals

$$\begin{aligned}\rho_{mn} &= \frac{\langle \mathbf{s}'_m, \mathbf{s}'_n \rangle}{\|\mathbf{s}'_m\| \|\mathbf{s}'_n\|} = \frac{\langle \mathbf{s}_m - \mathbf{c}, \mathbf{s}_n - \mathbf{c} \rangle}{\left(1 - \frac{1}{M}\right) \|\mathbf{s}_m\|^2} \\ &= \frac{\langle \mathbf{s}_m, \mathbf{s}_n \rangle - \langle \mathbf{s}_m, \mathbf{c} \rangle - \langle \mathbf{c}, \mathbf{s}_n \rangle + \langle \mathbf{c}, \mathbf{c} \rangle}{\left(1 - \frac{1}{M}\right) \|\mathbf{s}_m\|^2} \\ &= \begin{cases} \frac{\|\mathbf{s}_m\|^2 - \frac{2}{M} \|\mathbf{s}_m\|^2 + \frac{1}{M} \|\mathbf{s}_m\|^2}{\left(1 - \frac{1}{M}\right) \|\mathbf{s}_m\|^2} & m = n \\ \frac{0 - \frac{2}{M} \|\mathbf{s}_m\|^2 + \frac{1}{M} \|\mathbf{s}_m\|^2}{\left(1 - \frac{1}{M}\right) \|\mathbf{s}_m\|^2} & m \neq n \end{cases} \\ &= \begin{cases} 1 & m = n \\ -\frac{1}{M-1} & m \neq n \end{cases}\end{aligned}$$

Simplex signals are equally correlated !

Example of simplex signals

$$[\mathbf{s}_1 \quad \cdots \quad \mathbf{s}_M] = \begin{bmatrix} \sqrt{\mathcal{E}} & 0 & \cdots & 0 \\ 0 & \sqrt{\mathcal{E}} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \sqrt{\mathcal{E}} \end{bmatrix}$$

⇓

$$[\mathbf{s}'_1 \quad \cdots \quad \mathbf{s}'_M] = \begin{bmatrix} (1 - \frac{1}{M})\sqrt{\mathcal{E}} & -\frac{1}{M}\sqrt{\mathcal{E}} & \cdots & -\frac{1}{M}\sqrt{\mathcal{E}} \\ -\frac{1}{M}\sqrt{\mathcal{E}} & (1 - \frac{1}{M})\sqrt{\mathcal{E}} & \cdots & -\frac{1}{M}\sqrt{\mathcal{E}} \\ \vdots & \vdots & \ddots & \vdots \\ -\frac{1}{M}\sqrt{\mathcal{E}} & -\frac{1}{M}\sqrt{\mathcal{E}} & \cdots & (1 - \frac{1}{M})\sqrt{\mathcal{E}} \end{bmatrix}$$

Subdivision of time: N time slots

For example: BPSK in each dimension

$$\mathbf{s}_m = [c_{m,0}, c_{m,1}, \dots, c_{m,N-1}], \quad 1 \leq m \leq M$$

where $NT_c = T$

- “ $c_{m,j} = 0$ ” \equiv “ $g_1(t)$ is transmitted at time slot j ”
- “ $c_{m,j} = 1$ ” \equiv “ $g_2(t)$ is transmitted at time slot j ”

$$g_1(t) = +\sqrt{\frac{2\mathcal{E}_c}{T_c}} \cos(2\pi f_c t), \quad g_2(t) = -\sqrt{\frac{2\mathcal{E}}{T}} \cos(2\pi f_c t),$$

with $t \in [0, T_c)$

$$s_m(t) = \sqrt{\frac{2\mathcal{E}_c}{T_c}} \sum_{j=0}^{N-1} (-1)^{c_{m,j}} \cos(2\pi f_c(t - jT_c)) \mathbf{1}\{jT_c \leq t < (j+1)T_c\}$$

- Crosscorrelation coefficient of adjacent signals (i.e., with only one distinct component)

- For those identical components

$$\int_0^{T_c} |g_1(t)|^2 dt = \int_0^{T_c} |g_2(t)|^2 dt = \mathcal{E}_c$$

- For the single distinct component

$$\int_0^{T_c} g_1(t)g_2^*(t) dt = \int_0^{T_c} -|g_1(t)|^2 dt = -\mathcal{E}_c$$

- Hence

$$\rho_{mn} = \frac{\langle \mathbf{s}_m, \mathbf{s}_n \rangle}{\|\mathbf{s}_m\| \|\mathbf{s}_n\|} = \frac{(N-1)\mathcal{E}_c - \mathcal{E}_c}{N\mathcal{E}_c} = 1 - \frac{2}{N}$$

- Minimum Euclidean distance between adjacent codewords

$$\begin{aligned} \min_{m \neq n} \|\mathbf{s}_m - \mathbf{s}_n\| &= \min_{m \neq n} \sqrt{\|\mathbf{s}_m\|^2 + \|\mathbf{s}_n\|^2 - \langle \mathbf{s}_m, \mathbf{s}_n \rangle - \langle \mathbf{s}_n, \mathbf{s}_m \rangle} \\ &= \sqrt{N\mathcal{E}_c + N\mathcal{E}_c - 2(N\mathcal{E}_c)\frac{N-2}{N}} = 2\sqrt{\mathcal{E}_c} \end{aligned}$$

- Transmission energy of multidimensional BPSK signals

$$\mathcal{E}_m = \int_0^T |s_m(t)|^2 dt = N \|g_1(t)\|^2 = N \int_0^{T_c} |g_1(t)|^2 dt = N\mathcal{E}_c$$

- Largest number of channel symbols

$$M \leq 2^N$$

- Vectorization of BPSK signals

$$\mathbf{s}_m = \begin{bmatrix} \pm\sqrt{\mathcal{E}_c} \\ \pm\sqrt{\mathcal{E}_c} \\ \vdots \\ \pm\sqrt{\mathcal{E}_c} \end{bmatrix}_{N \times 1}$$

Can we properly choose $\{\mathbf{s}_m\}_{m=1}^M$ such that they are orthogonal to each other ?

Orthogonal multidimensional signals: Hadamard signals

- **Definition:** The Hadamard signals of size $M = 2^n$ can be recursively defined as

$$H_n = \begin{bmatrix} H_{n-1} & H_{n-1} \\ H_{n-1} & -H_{n-1} \end{bmatrix}$$

with initial value $H_0 = [1]$.

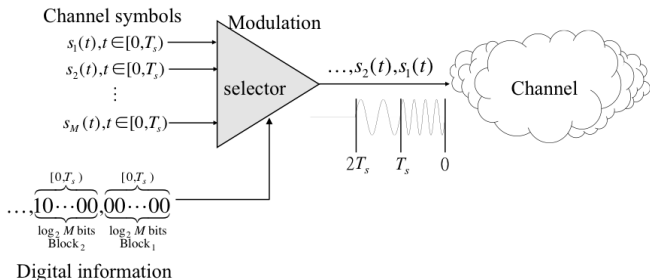
For example,

$$H_1 = \left[\begin{array}{c|c} 1 & 1 \\ \hline 1 & -1 \end{array} \right] \text{ and } H_2 = \left[\begin{array}{cc|cc} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ \hline 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{array} \right]$$

Hence, when $M = 4$, the Hadamard multidimensional orthogonal (BPSK) signals are

$$\left[\mathbf{s}_1 \quad \mathbf{s}_2 \quad \mathbf{s}_3 \quad \mathbf{s}_4 \right] = \begin{bmatrix} \sqrt{\mathcal{E}_c} & \sqrt{\mathcal{E}_c} & \sqrt{\mathcal{E}_c} & \sqrt{\mathcal{E}_c} \\ \sqrt{\mathcal{E}_c} & -\sqrt{\mathcal{E}_c} & \sqrt{\mathcal{E}_c} & -\sqrt{\mathcal{E}_c} \\ \sqrt{\mathcal{E}_c} & \sqrt{\mathcal{E}_c} & -\sqrt{\mathcal{E}_c} & -\sqrt{\mathcal{E}_c} \\ \sqrt{\mathcal{E}_c} & -\sqrt{\mathcal{E}_c} & -\sqrt{\mathcal{E}_c} & \sqrt{\mathcal{E}_c} \end{bmatrix}$$

3.3 Signaling schemes with memory



- Memoryless modulation: $s_{m_i}(t)$, $m_i \in \{1, 2, \dots, M\}$,
 m_i = function of Block $_i$
- Modulation with memory: $s_{m_i}(t)$,
 m_i = function of (Block $_i$, Block $_{i-1}$, \dots , Block $_{i-(L-1)}$)
- Linear modulation: The modulated part of $s_{m_i}(t)$ is a linear function of the digital waveform.

Linearity = Principle of superposition

If $a_1 \rightarrow b_1$ and $a_2 \rightarrow b_2$, then $a_1 + a_2 \rightarrow b_1 + b_2$.

- Non-linear modulation:

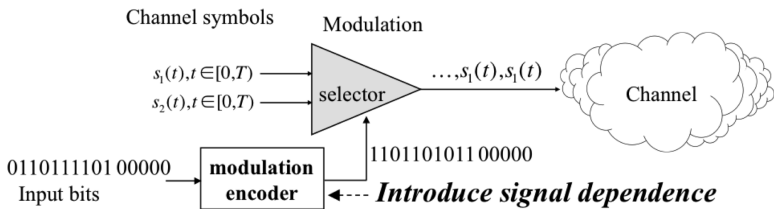
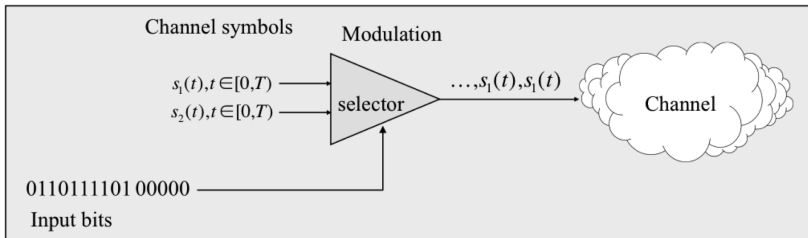
- Why introducing “memory” into signals?
 - The signal dependence is introduced for the purpose of shaping the spectrum of transmitted signal so that it matches the spectral characteristics of the channel.

- Linearity

- For example, $s_{m_i}(t) = \mathbf{Re} \{ A_{m_i} e^{2\pi f_c t} \}$.

$$\left\{ \begin{array}{l} -3 \longrightarrow \mathbf{Re} \{ -3e^{2\pi f_c t} \} \\ -1 \longrightarrow \mathbf{Re} \{ -1e^{2\pi f_c t} \} \\ +1 \longrightarrow \mathbf{Re} \{ +1e^{2\pi f_c t} \} \\ +3 \longrightarrow \mathbf{Re} \{ +3e^{2\pi f_c t} \} \end{array} \right.$$

- If the modulated part of $s_{m_i}(t)$ cannot be made as a linear function of the digital waveform, the modulation is classified as nonlinear.



Linear modulations with/without memory

- NRZ (Non-Return-to-Zero) = Binary PAM or binary PSK
: memoryless

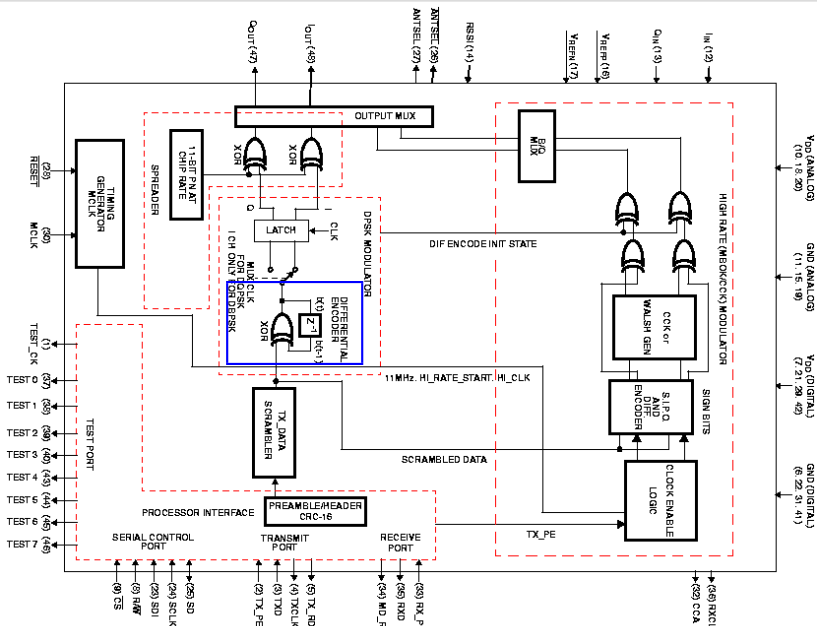
$$\text{channel code bit} = \text{input bit}$$

- NRZI (Non-Return-to-Zero, Inverted) = **Differential encoding** : with memory

$$(\text{channel code bit})_k = (\text{input bit})_k \oplus (\text{channel code bit})_{k-1}$$

$$\left\{ \begin{array}{l} (\text{channel code bit})_k = (\text{channel code bit})_{k-1}, \quad \text{when } (\text{input bit})_k = 0 \\ (\text{channel code bit})_k = \overline{(\text{channel code bit})_{k-1}}, \quad \text{when } (\text{input bit})_k = 1 \end{array} \right.$$

Application: DBPSK/DQPSK in Wireless LAN



Advantage of modulation with memory

Why adding differential encoding before BPSK ?

- For PSK modulations, digital information is carried by **absolute phase**.
 - Synchronization is often achieved by either adding a small pilot signal or using some self-synchronization scheme.
 - The demodulator needs to detect the phase, which may have a phase ambiguity due to noise and other constraints.

Example of phase ambiguity (frequency shift)

- Ideal (noiseless) case

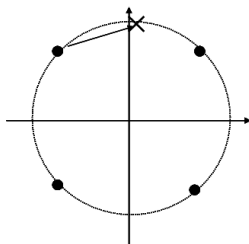
$$\begin{cases} f_{\text{transmitter}} = f_c : \text{receive } \cos(2\pi f_c t + \theta) \\ f_{\text{receiver}} = f_c : \text{estimate it based on } f_c \end{cases} \implies \text{estimate } \hat{\theta} = \theta$$

- Ambiguous case

$$\begin{cases} f_{\text{transmitter}} = f_c : \text{receive } \cos(2\pi f_c t + \theta) \\ f_{\text{receiver}} \neq f_c : \text{estimate it based on } f'_c \end{cases}$$

$$\implies \begin{cases} \text{receive } \cos(2\pi f'_c t + [2\pi(f_c - f'_c)t] + \theta) \\ \text{estimate it based on } f'_c \end{cases}$$

$$\implies \text{estimate } \hat{\theta} = [2\pi(f_c - f'_c)t] + \theta$$



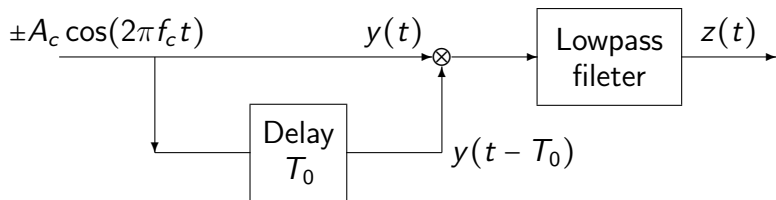
Advantage of differential encoding

$$(\text{channel code bit})_k = (\text{input bit})_k \oplus (\text{channel code bit})_{k-1}$$

- The phases or signs of the received waveforms are not important for detection.
- What is important is the change in the sign of successive pulses.
- The sign change can be detected even if the demodulating carrier has a phase ambiguity.

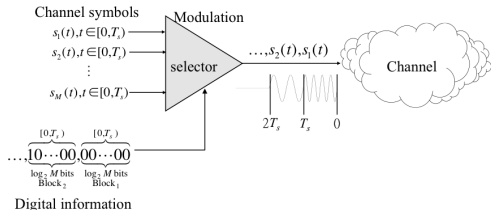
Advantage of diff encode (Noncoherent demod)

- No need to generate a local carrier at the receiver side.
- Use the received signal itself as a carrier.



$$z(t) = \begin{cases} A_c^2 \cos^2(2\pi f_c t) = \frac{A_c^2}{2} + \frac{1}{2} \cos(4\pi f_c t) \rightarrow \frac{1}{2} A_c^2, & \text{if } y(t) = y(t - T_0) \\ -A_c^2 \cos^2(2\pi f_c t) = -\frac{A_c^2}{2} - \frac{1}{2} \cos(4\pi f_c t) \rightarrow -\frac{1}{2} A_c^2, & \text{if } y(t) = -y(t - T_0) \end{cases}$$

Nonlinear modulation methods with memory



- Linear modulation: The modulated part of $s_{m_i}(t)$ is a linear function of the digital waveform.

Linearity = Principle of superposition

If $a_1 \rightarrow b_1$ and $a_2 \rightarrow b_2$, then $a_1 + a_2 \rightarrow b_1 + b_2$.

- Nonlinear modulation: The modulated part of $s_{m_i}(t)$ cannot be made as a linear function of the digital waveform.

(Linear (from the aspect of phase)) Frequency shift keying or FSK

$$s_m(t) = \text{Re} \left[\sqrt{\frac{2\mathcal{E}}{T}} e^{i 2\pi(m\Delta f)t} e^{i 2\pi f_c t} \right]$$

where $m = \pm 1, \pm 2, \dots, \pm(M-1)$

Motivation: Disadvantages of FSK

- Potential obstacles of multidimensional FSK with $(M - 1)$ oscillators for each desired frequency
 - Abrupt switching from one oscillator to another will result in relatively large spectral side lobes outside of the main spectral band of the signal.

Continuous-Phase FSK (CPFSK)

- Alternative implementation of multidimensional FSK
- A single carrier whose frequency is changed continuously.
- This is considered as a modulated signal with memory (we will explain this point in the next few slides).

Recall

$$s(t) = \mathbf{Re} \left\{ s_\ell(t) e^{i2\pi f_c t} \right\}, \quad s_\ell(t) = x_i(t) + i x_q(t)$$

$s_\ell(t)$ is the baseband version of the bandpass signal $s(t)$.

For ideal FSK signals

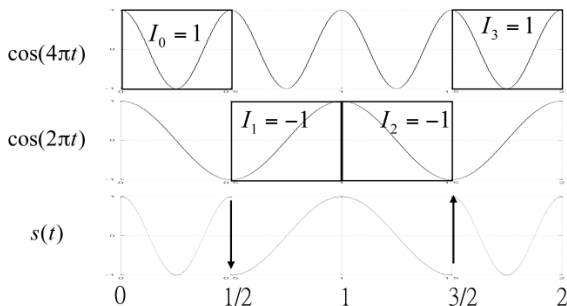
$$s_m(t) = \mathbf{Re} \left[\sqrt{\frac{2\mathcal{E}}{T}} e^{i2\pi(m\Delta f)t} e^{i2\pi f_c t} \right]$$
$$\implies s_{m,\ell}(t) = \sqrt{\frac{2\mathcal{E}}{T}} e^{i2\pi(m\Delta f)t}$$

where $\Delta f = f_d$ and $m = \pm 1, \pm 2, \dots, \pm(M-1)$.

Example of ideal (2-OSC) FSK signals

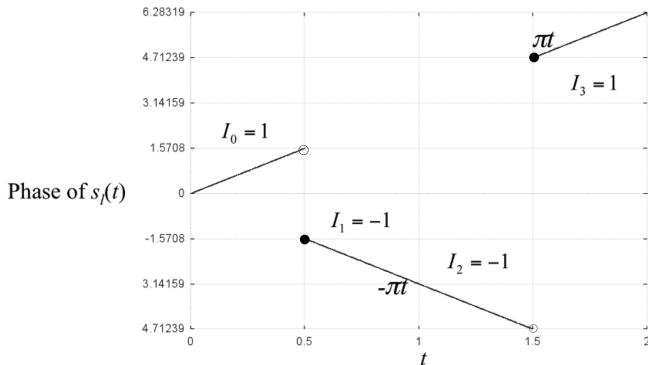
Let $T = 0.5$ sec, $\mathcal{E} = 0.25$, $f_d = 0.5$, $I_n (= m) \in \{1, -1\}$, and $f_c = 1.5$ Hz.

$$s(t) = \mathbf{Re} \left\{ s_\ell(t) e^{i2\pi f_c t} \right\} = \begin{cases} \cos(4\pi t) & I_n = 1 \\ \cos(2\pi t), & I_n = -1 \end{cases}$$



Discontinuous phase of (2-OSC) FSK

$$\text{Phase of } s_\ell(t) = \begin{cases} \pi t, & I_n = 1 \\ -\pi t, & I_n = -1 \end{cases} \quad \text{for } t \in [nT, (n+1)T)$$



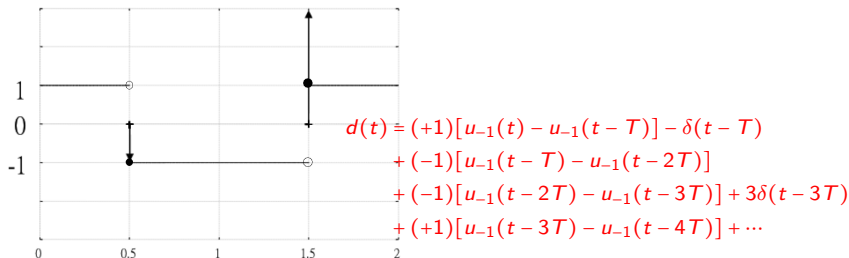
Phase change of (2-OSC) FSK

- (Normalized) phase change (for $t \in [nT, (n+1)T)$)

$$d(t) = \frac{\text{phase of } s_\ell(t)}{4\pi T f_d} = \frac{\frac{\partial}{\partial t}(2\pi I_n f_d t)}{4\pi T f_d} = \frac{I_n}{2T}$$

is the derivative of the phase!

Continue from the previous example with $T = 0.5$.



$$\begin{aligned}
 d(t) = & l_0 [u_{-1}(t) - u_{-1}(t - T)] + \mathbf{1}\{l_0 \neq l_1\} \cdot l_1 \cdot 1 \cdot \delta(t - T) \\
 & + l_1 [u_{-1}(t - T) - u_{-1}(t - 2T)] + \mathbf{1}\{l_1 \neq l_2\} \cdot l_2 \cdot 2 \cdot \delta(t - 2T) \\
 & + l_2 [u_{-1}(t - 2T) - u_{-1}(t - 3T)] + \mathbf{1}\{l_2 \neq l_3\} \cdot l_3 \cdot 3 \cdot \delta(t - 3T) \\
 & + l_3 [u_{-1}(t - 3T) - u_{-1}(t - 4T)] + \mathbf{1}\{l_3 \neq l_4\} \cdot l_4 \cdot 4 \cdot \delta(t - 4T) \\
 & + \dots
 \end{aligned}$$

- Phase change is the **derivative** of the phase!
- Phase is the **integration** of phase change!

$$s_\ell(t) = \sqrt{\frac{2\mathcal{E}}{T}} e^{i 4\pi T f_d \int_{-\infty}^t d(\tau) d\tau}$$

- Those $\delta(\cdot)$ functions result in “discontinuity” in integration! Hence, let us remove them to force “continuity” in phase.

Continuous phase FSK (CPFSK)

$$s_\ell(t) = \sqrt{\frac{2\mathcal{E}}{T}} e^{i4\pi T f_d \int_{-\infty}^t d(\tau) d\tau}$$

where

$$d(t) = \sum_{n=-\infty}^{\infty} I_n g(t - nT) \text{ and } g(t) = \frac{1}{2T} [u_{-1}(t) - u_{-1}(t - T)].$$

- $I_n \in \{\pm 1, \pm 3, \pm 5, \dots\}$ is the PAM information sequence.
- $g(t)$ is the “phase shaping function”.
 - It is now chosen as a rectangular pulse of height $1/(2T)$ and duration $[0, T)$ (hence, the area is $1/2$.)
- T is the symbol duration.

Re-express $s_\ell(t)$ as

$$s_\ell(t) = \sqrt{\frac{2\mathcal{E}}{T}} e^{i\phi(t; \mathbf{I})}$$

where

$$\phi(t; \mathbf{I})$$

$$= 4\pi T f_d \int_{-\infty}^t d(\tau) d\tau$$

$$= 4\pi T f_d \int_{-\infty}^t \left[\sum_{n=-\infty}^{\infty} I_n g(\tau - nT) \right] d\tau$$

$$= 4\pi f_d T \left[\sum_{k=-\infty}^{n-1} I_k \left(T \times \frac{1}{2T} \right) + I_n \frac{t - nT}{2T} \right] \quad \text{for } t \in [nT, (n+1)T)$$

$$= 2\pi f_d T \sum_{k=-\infty}^{n-1} I_k + 2\pi f_d (t - nT) I_n \quad \text{for } t \in [nT, (n+1)T)$$

(Cont.) For $t \in [nT, (n+1)T)$, $s_\ell(t) = \sqrt{\frac{2\mathcal{E}}{T}} e^{i\phi(t;I)}$ with

$$\begin{aligned}\phi(t; I) &= 2\pi f_d T \sum_{k=-\infty}^{n-1} I_k + 2\pi f_d (t - nT) I_n \\ &= \theta_n + 2\pi h \cdot I_n \cdot q(t - nT),\end{aligned}$$

where

$$\left\{ \begin{array}{l} h = 2f_d T \quad (\text{modulation index}) \\ \theta_n = \pi h \sum_{k=-\infty}^{n-1} I_k \quad (\text{accumulation of history/memory}) \\ q(t) = \begin{cases} 0 & t < 0 \\ \frac{t}{2T} & 0 \leq t < T \\ \frac{1}{2} & t \geq T \end{cases} \quad (\text{integration of } g(t)) \end{array} \right.$$

Generalization of CPFSK: CPM

We can further generalize $\phi(t; \mathbf{I})$ to

$$\phi(t; \mathbf{I}) = 2\pi \sum_{k=-\infty}^n h_k \cdot I_k \cdot q(t - kT)$$

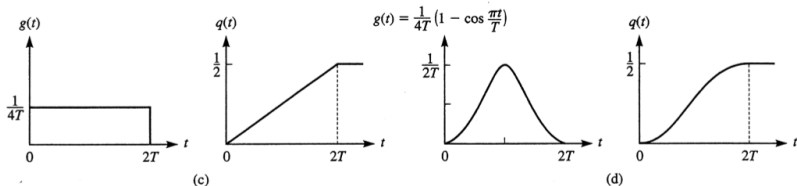
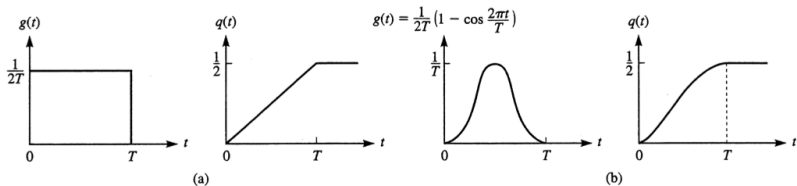
for $nT \leq t < (n+1)T$

where

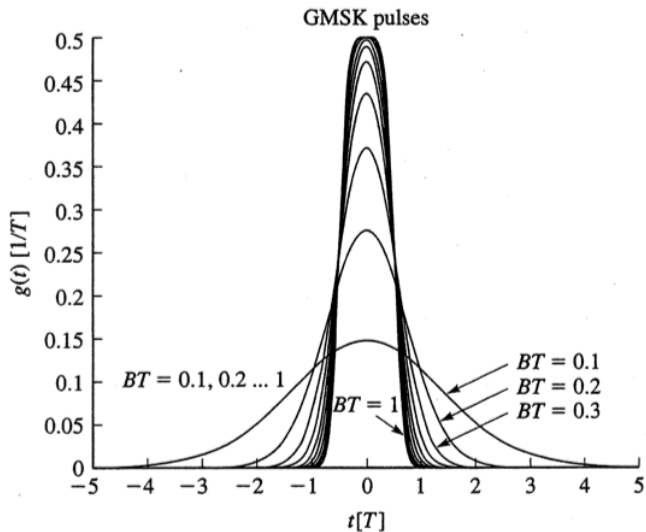
- 1 $\mathbf{I} = \{I_k\}_{k=-\infty}^{\infty}$ is the sequence of PAM symbols in $\{\pm 1, \pm 3, \dots, \pm(M-1)\}$.
- 2 h_k is the modulation index.
If h_k varies with k , it is called **multi- h CPM**.
- 3 $q(t) = \int_0^t g(\tau) d\tau$.

If $g(t) = 0$ for $t \geq T$ (and $t < 0$), $s_\ell(t)$ is called **full-response** CPM; otherwise it is called **partial-response** CPM.

Examples of CPMs



Examples of CPMs



Some commonly used CPM pulse shapes

- LREC (Rectangular): LREC with $L = 1$ is CPFSK

$$g(t) = \frac{1}{2LT} (u_{-1}(t) - u_{-1}(t - LT))$$

- LRC (Raised cosine)

$$g(t) = \frac{1}{2LT} (u_{-1}(t) - u_{-1}(t - LT)) \left(1 - \cos\left(\frac{2\pi t}{LT}\right) \right)$$

Some commonly used CPM pulse shapes

- GMSK (Gaussian minimum shift keying)

$$g(t) = Q\left(2\pi B\left(t - \frac{T}{2}\right) / \sqrt{\ln 2}\right) - Q\left(2\pi B\left(t + \frac{T}{2}\right) / \sqrt{\ln 2}\right)$$

where $Q(t) = \int_t^\infty \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$, and B is 3dB Bandwidth

- $g(t)$ is the response of filter $H(f) = 2^{-\frac{(f/B)^2}{2}}$ to a rectangular pulse of $u_{-1}(t + T/2) - u_{-1}(t - T/2)$.
- GMSK with $BT = 0.3$ is used in the European digital cellular communication system, called GSM (2G).
- At $BT = 0.3$, the GMSK pulse may be truncated at $|t| = 1.5T$ with a relatively small error incurred for $t > 1.5T$.

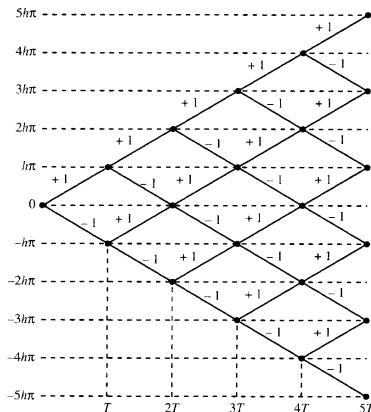
Representations of continuous-phase

- Phase trajectory or phase tree
- Phase trellis

Phase trajectory or phase tree

Binary CPFSK (i.e., $I_n = \pm 1$ and $g(t)$ is full response rectangular function)

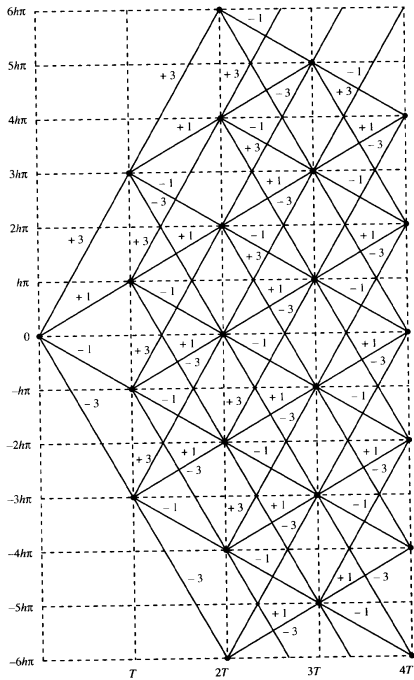
$$\phi(t; I) = \pi h \sum_{k=-\infty}^{n-1} I_k + 2\pi h I_n \cdot q(t - nT)$$



Example 3

Quaternary CPFSK (See the next page) with $I_n \in \{-3, -1, +1, +3\}$.

- *We observe that the phase trees for CPFSK are piecewise linear as a consequence of the fact that the pulse $g(t)$ is rectangular.*
- *Smoother phase trajectories and phase trees are obtained by using pulses that do not contain discontinuities.*

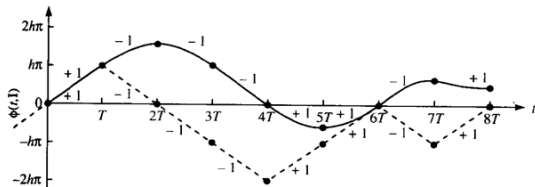


If $g(t)$ is continuous (especially at boundaries), phase trajectory becomes smooth.

Example 4

$$g(t) = \frac{1}{6T} \left(1 - \cos\left(\frac{2\pi t}{3T}\right) \right) = \text{raised cosine of length } 3T$$

with $(l_{-2}, l_{-1}, l_0, l_1, l_2, \dots) = (+1, +1, +1, -1, -1, -1, +1, +1, -1, +1, \dots)$



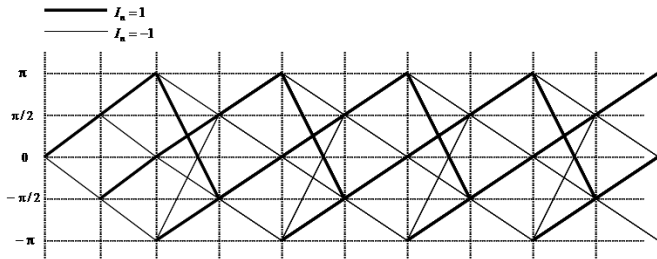
- * solid line = partial response CPM based on raised cosine pulse of length $3T$.
- * dashed line = binary CPFSK.

Phase trellis

Phase trellis = Phase trajectory is plotted with modulo 2π

Example 5

Binary CPFSK with $h = 1/2$ and $g(t)$ is a full response rectangular function.



Thus CPM can be decoded by **Viterbi trellis decoding**.

Minimum shift keying (MSK)

Recall for $nT \leq t < (n+1)T$, CPM has

$$\phi(t; \mathbf{I}) = 2\pi \sum_{k=-\infty}^n h_k \cdot I_k \cdot q(t - kT).$$

CPFSK is a special case of CPM with

$$g(t) = \frac{1}{2T} \text{ for } 0 \leq t < T$$

MSK is a special case of binary CPFSK with

$$h_k = \frac{1}{2}, g(t) = \frac{1}{2T} \text{ for } 0 \leq t < T \text{ and } I_n \in \{\pm 1\}$$

Thus for MSK, we have for $nT \leq t < (n+1)T$,

$$\phi(t; \mathbf{I}) = \frac{\pi}{2} \sum_{k=-\infty}^{n-1} I_k + \pi I_n q(t - nT) = \theta_n + \frac{1}{2} \pi I_n \left(\frac{t - nT}{T} \right)$$

$$\Phi(t; I) = \theta_n + \frac{1}{2}\pi I_n \left(\frac{t - nT}{T} \right) = 2\pi \left(\frac{I_n}{4T} \right) t - \frac{n\pi I_n}{2} + \theta_n$$

The corresponding modulated carrier wave is

$$\begin{aligned} s_{\text{MSK}}(t) &= A \cos(2\pi f_c t + \Phi(t; I)) \\ &= A \cos \left[2\pi \left(f_c + \frac{I_n}{4T} \right) t - \frac{n\pi I_n}{2} + \theta_n \right] \end{aligned}$$

Since $I_n \in \{\pm 1\}$, $s_{\text{MSK}}(t)$ has two frequency components:

$$\begin{aligned} f_1 &= f_c - \frac{1}{4T} \\ f_2 &= f_c + \frac{1}{4T} \end{aligned}$$

Minimum shift keying (MSK)

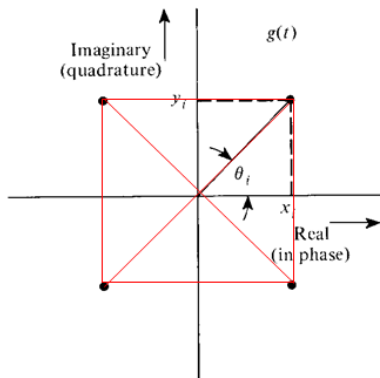
MSK is so named because $f_2 - f_1 = \frac{1}{2T}$ = the minimum (frequency) shift that makes the two frequency components **orthogonal**.

[See Slide 3-35] When $\Delta f = \frac{k}{2T}$, $\mathbf{Re}\{\rho_{mn,\ell}\} = 0$ for $m \neq n$. In other words, **the minimum frequency separation between adjacent (passband) signals for orthogonality is $\Delta f = \frac{1}{2T}$.**

MSK is sometimes regarded as a kind of OQPSK (Offset QPSK). Why?

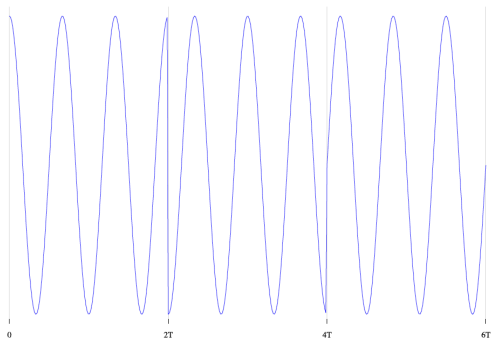
Offset QPSK

The original QPSK



There could be 180 degree of (sudden) phase change (so, not **continuous phase**), e.g., from $(+1, +1)$ to $(-1, -1)$.

$$s_{\text{QPSK}}(t) = \sum_{n=-\infty}^{\infty} I_{2n}g(t - 2nT) \cos(2\pi f_c t) - \sum_{n=-\infty}^{\infty} I_{2n+1}g(t - 2nT) \sin(2\pi f_c t)$$

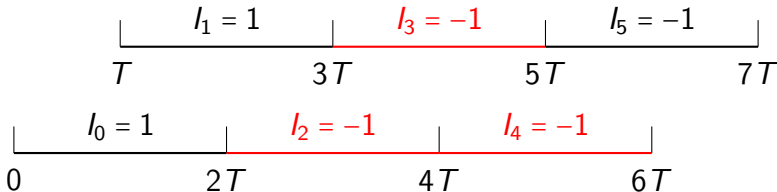


$(I_0, I_1) = (+1, +1)$, $(I_2, I_3) = (-1, -1)$ and $(I_4, I_5) = (-1, +1)$.
 $g(t)$ rectangular pulse of unit height and during $2T$.

Offset QPSK (OQPSK)

How to reduce the 180° phase change to only 90° ?

Simple solution: Do not let the “two bits” l_{2n} and l_{2n+1} change at the same time!



$$s_{\text{OQPSK}}(t) = \sum_{n=-\infty}^{\infty} l_{2n} g(t - 2nT) \cos(2\pi f_c t) - \sum_{n=-\infty}^{\infty} l_{2n+1} g(t - (2n+1)T) \sin(2\pi f_c t)$$

To synchronize with the textbook, we reverse $\{I_{2n+1}\}$ to obtain

$$\begin{aligned} s_{\text{OQPSK}}(t) &= \sum_{n=-\infty}^{\infty} I_{2n} g(t - 2nT) \cos(2\pi f_c t) \\ &\quad + \sum_{n=-\infty}^{\infty} I_{2n+1} g(t - (2n+1)T) \sin(2\pi f_c t) \end{aligned}$$

MSK can be regarded as a kind of (memoryless) OQPSK. Why?

$$\text{MSK: } \phi(t; \mathbf{I}) = \theta_n + \frac{1}{2} \pi I_n \left(\frac{t-nT}{T} \right) = \theta_0 + \frac{\pi}{2} \sum_{k=0}^{n-1} I_k + \pi \left(\frac{I_n}{2T} \right) t - \frac{n\pi}{2} I_n \quad \text{for } nT \leq t < (n+1)T$$

Proof: Suppose without loss of generality,

$$\theta_0 = \frac{\pi}{2} \sum_{k=-\infty}^{-1} I_k = \frac{3\pi}{2}.$$

Then for $nT \leq t < (n+1)T$ (and $n \geq 1$),

$$s_{\text{MSK},\ell}(t) = e^{j\phi(t; \mathbf{I})}$$

$$= e^{j\pi \left(\frac{I_n}{2T} \right) t} \cdot e^{-j\pi \frac{n\pi}{2} I_n} \cdot e^{j\pi \frac{\pi}{2} \sum_{k=0}^{n-1} I_k} \cdot e^{j\theta_0}$$

Note $(-j)^n = 1$.

$$= \left[\cos \left(\pi \frac{t}{2T} \right) + j I_n \sin \left(\pi \frac{t}{2T} \right) \right] (-I_n j)^n \left(\prod_{k=0}^{n-1} (I_k j) \right) (-j)$$

$$= I_n^{n+1} \left(\prod_{k=0}^{n-1} I_k \right) \sin \left(\pi \frac{t}{2T} \right) + j I_n^n \left(\prod_{k=0}^{n-1} I_k \right) \sin \left(\pi \frac{(t-T)}{2T} \right)$$

n	$I_n^{n+1} \left(\prod_{k=0}^{n-1} I_k \right)$	$I_n^n \left(\prod_{k=0}^{n-1} I_k \right)$
0	$J_0 = I_0 = J_{2\lfloor 0/2 \rfloor}$	
1	$J_0 = I_0 = J_{2\lfloor 1/2 \rfloor}$	$J_1 = I_0 I_1 = J_{2\lfloor (1-1)/2 \rfloor + 1}$
2	$J_2 = I_0 I_1 I_2 = J_{2\lfloor 2/2 \rfloor}$	$J_1 = I_0 I_1 = J_{2\lfloor (2-1)/2 \rfloor + 1}$
3	$J_2 = I_0 I_1 I_2 = J_{2\lfloor 3/2 \rfloor}$	$J_3 = I_0 I_1 I_2 I_3 = J_{2\lfloor (3-1)/2 \rfloor + 1}$
4	$J_4 = I_0 I_1 I_2 I_3 I_4 = J_{2\lfloor 4/2 \rfloor}$	$J_3 = I_0 I_1 I_2 I_3 = J_{2\lfloor (4-1)/2 \rfloor + 1}$
5	$J_4 = I_0 I_1 I_2 I_3 I_4 = J_{2\lfloor 5/2 \rfloor}$	$J_5 = I_0 I_1 I_2 I_3 I_4 I_5 = J_{2\lfloor (5-1)/2 \rfloor + 1}$
6	$J_6 = I_0 I_1 I_2 I_3 I_4 I_5 I_6 = J_{2\lfloor 6/2 \rfloor}$	$J_5 = I_0 I_1 I_2 I_3 I_4 I_5 = J_{2\lfloor (6-1)/2 \rfloor + 1}$

For $nT \leq t < (n+1)T$,

$$\begin{aligned}
 \text{SMSK},\ell(t) &= J_{2\lfloor n/2 \rfloor} (-1)^{\lfloor n/2 \rfloor} \underbrace{\sin \left(\pi \frac{(t - 2\lfloor n/2 \rfloor T)}{2T} \right)}_{g(t - 2\lfloor n/2 \rfloor T)} \\
 &- j J_{2\lfloor (n-1)/2 \rfloor + 1} (-1)^{\lfloor (n-1)/2 \rfloor + 1} \underbrace{\sin \left(\pi \frac{(t - 2\lfloor (n-1)/2 \rfloor T - T)}{2T} \right)}_{g(t - 2\lfloor (n-1)/2 \rfloor T - T)}
 \end{aligned}$$

For $2mT \leq t < (2m+1)T$ (i.e., $n = 2m$),

$$s_{\text{MSK},\ell}(t) = J_{2m}(-1)^m g(t - 2mT) - j J_{2m-1} \underbrace{(-1)^m}_{=(-1)^{\lceil(2m-1)/2\rceil}} g(t - (2m-1)T)$$

For $(2m+1)T \leq t < (2m+2)T$ (i.e., $n = 2m+1$),

$$s_{\text{MSK},\ell}(t) = J_{2m}(-1)^m g(t - 2mT) - j J_{2m+1} \underbrace{(-1)^{m+1}}_{=(-1)^{\lceil(2m+1)/2\rceil}} g(t - (2m+1)T)$$

For $(2m+2)T \leq t < (2m+3)T$ (i.e., $n = 2m+2$),

$$s_{\text{MSK},\ell}(t) = J_{2(m+1)}(-1)^{m+1} g(t - 2(m+1)T) - j J_{2m+1} \underbrace{(-1)^{m+1}}_{=(-1)^{\lceil(2m+1)/2\rceil}} g(t - (2m+1)T)$$

with $g(t) = \sin\left(\pi \frac{t}{2T}\right) [u_{-1}(t) - u_{-1}(t - 2T)]$.

MSK can be regarded as a memoryless OQPSK by setting

$$s_{\text{MSK}}(t) = \left[\sum_{n=-\infty}^{\infty} \tilde{I}_{2n} g(t - 2nT) \right] \cos(2\pi f_c t) \\ + \left[\sum_{n=-\infty}^{\infty} \tilde{I}_{2n+1} g(t - (2n+1)T) \right] \sin(2\pi f_c t)$$

with

$$\tilde{I}_n = (-1)^{\lceil n/2 \rceil} J_n = (-1)^{\lceil n/2 \rceil} \prod_{k=0}^n I_k.$$

- MSK can be “composed” using “memoryless” circuits with “with-memory” information sequence \tilde{I} .
- Please be noted that the textbook **abuses** the notation by using $g(t)$ to denote both **amplitude** and **phase** pulse shaping functions for CPM signals!

A linear representation of CPM

The **key** of OQPSK representation of MSK is that **phase** can be “pulled down” as a multiplicative adjustment in **amplitude** when $I_n \in \{-1, +1\}$!

For example,
$$e^{j2\pi\left(\frac{I_n}{4T}\right)t} = \cos\left(\pi\frac{t}{2T}\right) + j I_n \sin\left(\pi\frac{t}{2T}\right).$$

(1986 Laurent)

- CPM can also be represented as a linear superposition of AM signal waveforms (if $I_n \in \{\pm 1\}$).
- Such a representation provides an alternative method for synthesizing CPM signal at the transmitter and for demodulating the signal at the receiver.

An important and useful fact

For $l \in \{-1, +1\}$,

$$e^{jA \cdot l} = \frac{\sin(B - A)}{\sin(B)} + e^{jB \cdot l} \frac{\sin(A)}{\sin(B)}.$$

Proof:

$$\begin{aligned} \sin(B)e^{jA \cdot l} &= \sin(B)[\cos(A) + j l \sin(A)] \\ &= \sin(B)\cos(A) + j \sin(B \cdot l)\sin(A) \\ &= \sin(B - A) + \cos(B)\sin(A) + j \sin(B \cdot l)\sin(A) \\ &= \sin(B - A) + \sin(A)[\cos(B \cdot l) + j \sin(B \cdot l)] \\ &= \sin(B - A) + \sin(A)e^{jB \cdot l} \end{aligned}$$

□

For general h and $g(\cdot)$ function of duration L and of integral $1/2$ (but each $I_n \in \{\pm 1\}$), we have for $nT \leq t < (n+1)T$ (for a binary CPM signal),

$$\begin{aligned}
 s_{\text{b-CPM},\ell}(t) &= e^{i\phi(t;I)} \\
 &= e^{i(\pi h \sum_{k=-\infty}^{n-L} I_k + 2\pi h \sum_{k=n-L+1}^n I_k q(t-kT))} \\
 &= e^{i\pi h \sum_{k=-\infty}^{n-L} I_k} \prod_{k'=0}^{L-1} e^{i2\pi h I_{n-k'} q(t-(n-k')T)} \quad (n-k' = k) \\
 &= e^{i\pi h \sum_{k=-\infty}^{n-L} I_k} \prod_{k'=0}^{L-1} \left(\frac{\sin(B - 2\pi h q(t - (n - k')T))}{\sin(B)} \right. \\
 &\quad \left. + e^{iB \cdot I_{n-k'}} \frac{\sin(2\pi h q(t - (n - k')T))}{\sin(B)} \right),
 \end{aligned}$$

where $B = \pi h$.

Define

$$s_0(t) = \begin{cases} \frac{\sin(2\pi h q(t))}{\sin(B)} & 0 \leq t < LT \\ \frac{\sin(B - 2\pi h q(t - LT))}{\sin(B)} & LT \leq t < 2LT \\ 0 & \text{otherwise} \end{cases}$$

Since $q(0) = 0$ and $q(LT) = 1/2$, $s_0(t)$ is continuous for $t \in \mathbb{R}$.

Continue the derivation:

$$\begin{aligned}
 s_{b\text{-CPM},\ell}(t) &= e^{j\pi h \sum_{k=-\infty}^{n-L} I_k} \prod_{k'=0}^{L-1} \left(\frac{\sin(B - 2\pi h q(t - (n - k')T + LT - LT))}{\sin(B)} \right. \\
 &\quad \left. + e^{jB \cdot I_{n-k'}} \frac{\sin(2\pi h q(t - (n - k')T))}{\sin(B)} \right) \\
 &= e^{j\pi h \sum_{k=-\infty}^{n-L} I_k} \prod_{k'=0}^{L-1} (s_0(t - (n - k')T + LT) \\
 &\quad + e^{jB \cdot I_{n-k'}} s_0(t - (n - k')T))
 \end{aligned}$$

$nT \leq t < (n+1)T$ and $0 \leq k' \leq L-1$ imply that

$$0 \leq t - (n - k')T < LT \text{ and } LT \leq t - (n - k')T + LT < 2LT.$$

$$\begin{aligned}
& \prod_{k'=0}^{L-1} (s_0(t - (n - k')T + LT) + e^{zB \cdot I_{n-k'}} s_0(t - (n - k')T)) \\
= & \left(\underbrace{s_0(t - nT + 0 \cdot T + LT)}_{a_{i,0}=1 \ (k'=0)} + e^{zB \cdot I_{n-0}} \underbrace{s_0(t - nT + 0 \cdot T)}_{a_{i,0}=0 \ (k'=0)} \right) \\
\times & \left(\underbrace{s_0(t - nT + 1 \cdot T + LT)}_{a_{i,1}=1 \ (k'=1)} + e^{zB \cdot I_{n-1}} \underbrace{s_0(t - nT + 1 \cdot T)}_{a_{i,1}=0 \ (k'=1)} \right) \\
& \vdots \\
\times & \left(\underbrace{s_0(t - nT + (L-1) \cdot T + LT)}_{a_{i,L-1}=1 \ (k'=L-1)} + e^{zB \cdot I_{n-(L-1)}} \underbrace{s_0(t - nT + (L-1) \cdot T)}_{a_{i,L-1}=0 \ (k'=L-1)} \right) \\
= & \sum_{i=0}^{2^L-1} e^{zB \sum_{k'=0}^{L-1} (1-a_{i,k'}) I_{n-k'}} \prod_{k'=0}^{L-1} s_0(t - nT + k' T + a_{i,k'} LT)
\end{aligned}$$

where $(a_{i,0}, a_{i,1}, \dots, a_{i,L-1})$ is the binary representation of i with $a_{i,0}$ being the most significant bit.

Continue the derivation:

$$\begin{aligned}
 s_{\text{b-CPM},\ell}(t) &= e^{iB \sum_{k=-\infty}^{n-L} I_k} \sum_{i=0}^{2^L-1} e^{iB \sum_{k'=0}^{L-1} (1-a_{i,k'}) I_{n-k'}} \prod_{k'=0}^{L-1} s_0(t - nT + k'T + a_{i,k'}LT) \\
 &= \sum_{i=0}^{2^L-1} \underbrace{e^{i\pi h A_{i,n}}}_{\text{complex amplitude}} \underbrace{c_i(t - nT)}_{\text{pulse shaping function}}
 \end{aligned}$$

where

$$A_{i,n} = \sum_{k=-\infty}^n I_k - \sum_{k'=0}^{L-1} a_{i,k'} I_{n-k'} \quad \text{and} \quad c_i(t) = \prod_{k'=0}^{L-1} s_0(t + k'T + a_{i,k'}LT).$$

Binary CPM can be expressed as a weighted sum of 2^L real-valued pulses $\{c_i(t)\}$ where the complex amplitudes depends on the information sequence. This is useful, especially when L is small!

Property of $c_i(t)$

- Duration:** $c_i(t) = 0$ if any of $s_0(t + k'T + a_{i,k'}LT) = 0$.
Hence, $c_i(t) \neq 0$ only possible in

$$\max_{0 \leq k' < L} (-k'T - a_{i,k'}LT) \leq t < \min_{0 \leq k' < L} [(-k'T - a_{i,k'}LT) + 2LT]$$
$$\Leftrightarrow - \underbrace{\left(\min_{\substack{0 \leq k' \leq L \\ \text{and } a_{i,k'}=0}} k' \right)}_{\substack{\text{"}\leq L\text{" for the case} \\ \text{of } a_{i,k'} = 1 \forall 0 \leq k' < L}} T \leq t < LT - \underbrace{\left(\max_{\substack{-1 \leq k' < L \\ \text{and } a_{i,k'}=1}} k' \right)}_{\substack{\text{"}\leq -1\text{" for the case} \\ \text{of } a_{i,k'} = 0 \forall 0 \leq k' < L}} T$$

where we define $a_{i,L} = 0$ and $a_{i,-1} = 1$. So, the duration is equal to:

$$\left(L - \underbrace{\left(\max_{\substack{-1 \leq k' < L \\ \text{and } a_{i,k'}=1}} k' \right)}_{k_{\max_1}} + \underbrace{\left(\min_{\substack{0 \leq k' \leq L \\ \text{and } a_{i,k'}=0}} k' \right)}_{k_{\min_0}} \right) T.$$

$$L = 3$$

i	$a_{i,0}a_{i,1}a_{i,2}$	$-k_{\min_0}$	$L - k_{\max_1}$	$(L - k_{\max_1}) - (-k_{\min_0})$
0	000	0	4	4
1	001	0	1	1
2	010	0	2	2
3	011	0	1	1
4	100	-1	3	4
5	101	-1	1	2
6	110	-2	2	4
7	111	-3	1	4

It can be shown that $L - k_{\max_1} + k_{\min_0} \leq L + 1$, and the upper bound can always be achieved by $i = 0$.

Example. $h = 1/2$ and $q(t) = \begin{cases} 0 & t < 0 \\ t/(6T) & 0 \leq t < 3T \\ 1/2 & \text{otherwise} \end{cases}$. Then

$$s_0(t) = \begin{cases} \sin\left(\frac{\pi}{6T}t\right) & 0 \leq t < 6T \\ 0 & \text{otherwise} \end{cases}$$

$$A_{i,n} = \sum_{k=-\infty}^n I_k - \sum_{k'=0}^2 a_{i,k'} I_{n-k'} \quad \text{and} \quad c_i(t) = \prod_{k'=0}^2 s_0(t+k'T+a_{i,k'}LT).$$

$a_{i,0}a_{i,1}a_{i,2}$	duration	$c_i(t)$	$e^{i\pi hA_{i,n}}$
0 \equiv 000	[0,4T)	$s_0(t)s_0(t+T)s_0(t+2T)$	$e^{i\theta_{n+1}}$
1 \equiv 001	[0,T)	$s_0(t)s_0(t+T)s_0(t+5T)$	$e^{i(\theta_{n-2}+\pi hl_n+\pi hl_{n-1})}$
2 \equiv 010	[0,2T)	$s_0(t)s_0(t+4T)s_0(t+2T)$	$e^{i(\theta_{n-1}+\pi hl_n)}$
3 \equiv 011	[0,T)	$s_0(t)s_0(t+4T)s_0(t+5T)$	$e^{i(\theta_{n-2}+\pi hl_n)}$
4 \equiv 100	[-T,3T)	$s_0(t+3T)s_0(t+T)s_0(t+2T)$	$e^{i\theta_n}$
5 \equiv 101	[-T,T)	$s_0(t+3T)s_0(t+T)s_0(t+5T)$	$e^{i(\theta_{n-2}+\pi hl_{n-1})}$
6 \equiv 110	[-2T,2T)	$s_0(t+3T)s_0(t+4T)s_0(t+2T)$	$e^{i\theta_{n-1}}$
7 \equiv 111	[-3T,T)	$s_0(t+3T)s_0(t+4T)s_0(t+5T)$	$e^{i\theta_{n-2}}$

Note that

$$\left\{ \begin{array}{l} c_4(t) = c_0(t+T) \\ e^{i\pi hA_{4,n}} = e^{i\pi hA_{0,n-1}} \end{array} \right. \quad \left\{ \begin{array}{l} c_6(t) = c_0(t+2T) \\ e^{i\pi hA_{6,n}} = e^{i\pi hA_{0,n-2}} \end{array} \right. \quad \left\{ \begin{array}{l} c_7(t) = c_0(t+3T) \\ e^{i\pi hA_{7,n}} = e^{i\pi hA_{0,n-3}} \end{array} \right.$$

and

$$\left\{ \begin{array}{l} c_5(t) = c_2(t+T) \\ e^{i\pi hA_{5,n}} = e^{i\pi hA_{2,n-1}} \end{array} \right.$$

For $nT \leq t < (n+1)T$,

$$\begin{aligned}
 s_{\text{b-CPM},\ell}(t) &= e^{j\phi(t;I)} = \sum_{i=0}^7 e^{j\pi h A_{i,n}} c_i(t - nT) \\
 &= e^{j\pi h A_{0,n}} c_0(t - nT) + e^{j\pi h A_{1,n}} c_1(t - nT) + e^{j\pi h A_{2,n}} c_2(t - nT) \\
 &+ e^{j\pi h A_{3,n}} c_3(t - nT) + e^{j\pi h A_{4,n}} c_4(t - nT) + e^{j\pi h A_{5,n}} c_5(t - nT) \\
 &+ e^{j\pi h A_{6,n}} c_6(t - nT) + e^{j\pi h A_{7,n}} c_7(t - nT) \\
 &= e^{j\pi h A_{0,n}} c_0(t - nT) + e^{j\pi h A_{1,n}} c_1(t - nT) + e^{j\pi h A_{2,n}} c_2(t - nT) \\
 &+ e^{j\pi h A_{3,n}} c_3(t - nT) + e^{j\pi h A_{0,n-1}} c_0(t - (n-1)T) \\
 &+ e^{j\pi h A_{2,n-1}} c_2(t - (n-1)T) + e^{j\pi h A_{0,n-2}} c_0(t - (n-2)T) \\
 &+ e^{j\pi h A_{0,n-3}} c_0(t - (n-3)T) \\
 &= \sum_{m=-\infty}^{\infty} \left[e^{j\pi h A_{0,m}} c_0(t - mT) + e^{j\pi h A_{1,m}} c_1(t - mT) \right. \\
 &\quad \left. + e^{j\pi h A_{2,m}} c_2(t - mT) + e^{j\pi h A_{3,m}} c_3(t - mT) \right] \\
 &= \sum_{m=-\infty}^{\infty} \left[\sum_{i=0}^{2^{3-1}-1} e^{j\pi h A_{i,m}} c_i(t - mT) \right]
 \end{aligned}$$

So, we notice that when $a_{i,0} = 1$, $c_i(t)$ is always a shift-version of some $c_j(t)$ with $0 \leq j \leq 2^{L-1} - 1$.

This concludes to that:

Theorem 1 (Laurent '86)

For $nT \leq t < (n+1)T$,

$$s_{b\text{-CPM},\ell}(t) = \sum_{m=-\infty}^{\infty} \left[\sum_{i=0}^{2^{L-1}-1} e^{i\pi h A_{i,m}} c_i(t - mT) \right]$$

where

$$A_{i,n} = \sum_{k=-\infty}^n I_k - \sum_{k'=1}^{L-1} a_{i,k'} I_{n-k'}$$

and

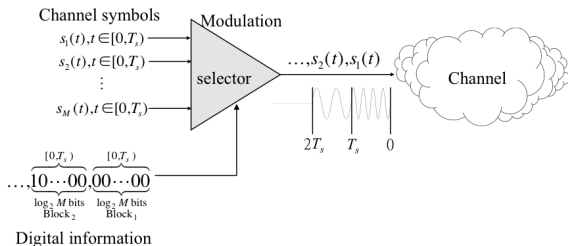
$$c_i(t) = s_0(t) \prod_{k'=1}^{L-1} s_0(t + k'T + a_{i,k'}LT)$$

with duration $0 \leq t < (L - k_{\max 1})T$.

3.4 Power spectrum of digital modulated signals

- Why studying spectral characteristics?
 - Bandwidth limitation in a real channel.
- Random process \implies Power spectral density
 - PAM
 - CPM

Power spectra of modulated signals



- The modulated waveform $s(t)$ is deterministic given the information sequence \mathbf{I} , so only the information sequence $\mathbf{I} = (\dots, I_{-2}, I_{-1}, I_0, I_1, I_2, \dots)$ is random!
- For convenience, we denote the waveform at $nT \leq t < (n+1)T$ as $s(t - nT; \mathbf{I}_n)$ if the modulation is memoryless, and as $s(t - nT; \mathbf{I}_n)$ if the modulation is with memory, where $\mathbf{I}_n = (\dots, I_{n-2}, I_{n-1}, I_n)$.

Hence, the modulated lowpass equivalent signal can be expressed as

$$\mathbf{v}_\ell(t) = \sum_{n=-\infty}^{\infty} s(t - nT; I_n).$$

Note that $\mathbf{v}_\ell(t)$ is usually not a (wide-sense) stationary process but a **cyclostationary** process.

Its spectral characteristics is then determined by the **time-averaged autocorrelation function** rather than the usual autocorrelation function for a WSS process.

2.7.2 Cyclostationary processes

- How to model a waveform source that carries digital information?
- For example,

$$X(t) = \sum_{n=-\infty}^{\infty} a_n \cdot g(t - nT)$$

where $\{a_n\}_{n=-\infty}^{\infty}$ is a discrete-time **random** sequence, and $g(t)$ is a **deterministic** pulse shaping function.

Cyclostationary processes

Given that $\{\mathbf{a}_n\}_{n=-\infty}^{\infty}$ is WSS, what is the statistical property of $\mathbf{X}(t)$?

- $\mathbf{X}(t)$ is not necessarily (strictly) stationary. Its mean becomes periodic with period T :

$$\mathbb{E}[\mathbf{X}(t)] = \mathbb{E}\left[\sum_{n=-\infty}^{\infty} \mathbf{a}_n g(t - nT)\right] = \mu_{\mathbf{a}} \sum_{n=-\infty}^{\infty} g(t - nT) = E[\mathbf{X}(t + KT)]$$

- Autocorrelation function becomes periodic with period T

$$\begin{aligned} R_{\mathbf{X}}(t_1, t_2) &= \mathbb{E}[\mathbf{X}(t_1)\mathbf{X}^*(t_2)] \\ &= \sum_{n=-\infty}^{\infty} \sum_{m=-\infty}^{\infty} \mathbb{E}[\mathbf{a}_n \mathbf{a}_m^*] g(t_1 - nT) g(t_2 - mT) \\ &= \sum_{n=-\infty}^{\infty} \sum_{m=-\infty}^{\infty} R_{\mathbf{a}}(n - m) g(t_1 - nT) g(t_2 - mT) \\ &= R_{\mathbf{X}}(t_1 + KT, t_2 + KT) \end{aligned}$$

Definition 1 (Cyclostationary process)

A random process is said to be *cyclostationary* or *periodically stationary in the wide sense* if its mean and autocorrelation function are both periodic.

- Time-average autocorrelation function

$$\bar{R}_{\mathbf{x}}(\tau) = \frac{1}{T} \int_0^T R_{\mathbf{x}}(t + \tau, t) dt$$

- Average power spectral density

$$\bar{S}_{\mathbf{x}}(f) = \mathcal{F} \{ \bar{R}_{\mathbf{x}}(\tau) \}$$

3.4-1 Power spectral density of a digitally modulated signal with memory

$$\mathbb{E}[\mathbf{v}_\ell(t)] = \sum_{n=-\infty}^{\infty} \mathbb{E}[I_n] g(t - nT) = \mu_I \sum_{n=-\infty}^{\infty} g(t - nT) = \mathbb{E}[\mathbf{v}_\ell(t + T)]$$

and

$$\begin{aligned} R_{\mathbf{v}_\ell}(t_1, t_2) &= \mathbb{E}[\mathbf{v}_\ell(t_1) \mathbf{v}_\ell^*(t_2)] \\ &= \sum_{n=-\infty}^{\infty} \sum_{m=-\infty}^{\infty} \mathbb{E}[I_n I_m^*] g(t_1 - nT) g^*(t_2 - mT) = R_{\mathbf{v}_\ell}(t_1 + T, t_2 + T) \end{aligned}$$

implies that $\mathbf{v}_\ell(t)$ is cyclostationary.

$$\begin{aligned} \bar{R}_{\mathbf{v}_\ell}(\tau) &= \frac{1}{T} \int_0^T \sum_{n=-\infty}^{\infty} \sum_{m=-\infty}^{\infty} R_I(n-m) g(t + \tau - nT) g^*(t - mT) dt \\ &= \frac{1}{T} \int_0^T \sum_{k=-\infty}^{\infty} \sum_{m=-\infty}^{\infty} R_I(k) g(t + \tau - kT - mT) g^*(t - mT) dt \\ &\quad (k = n - m) \end{aligned}$$

$$\begin{aligned}
&= \frac{1}{T} \sum_{k=-\infty}^{\infty} R_I(k) \sum_{m=-\infty}^{\infty} \int_0^T g(t+\tau-kT-mT)g^*(t-mT)dt \\
&\stackrel{u=t-mT}{=} \frac{1}{T} \sum_{k=-\infty}^{\infty} R_I(k) \sum_{m=-\infty}^{\infty} \int_{-mT}^{-(m-1)T} g(u+\tau-kT)g^*(u)du \\
&= \frac{1}{T} \sum_{k=-\infty}^{\infty} R_I(k) \int_{-\infty}^{\infty} g(u+\tau-kT)g^*(u)du \\
&= \frac{1}{T} \sum_{k=-\infty}^{\infty} g_k(\tau-kT)
\end{aligned}$$

where

$$g_m(\tau) = R_I(m) \int_{-\infty}^{\infty} g(u+\tau)g^*(u)du.$$

$$\begin{aligned}
G_m(f) &= \int_{-\infty}^{\infty} g_m(\tau) e^{-\imath 2\pi f \tau} d\tau \\
&= \int_{-\infty}^{\infty} \left(R_I(m) \int_{-\infty}^{\infty} g(u + \tau) g^*(u) du \right) e^{-\imath 2\pi f \tau} d\tau \\
&= R_I(m) \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} g(u + \tau) e^{-\imath 2\pi f \tau} d\tau \right) g^*(u) du \\
&\stackrel{v=u+\tau}{=} R_I(m) \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} g(v) e^{-\imath 2\pi f (v-u)} dv \right) g^*(u) du \\
&= R_I(m) \left(\int_{-\infty}^{\infty} g(v) e^{-\imath 2\pi f v} dv \right) \left(\int_{-\infty}^{\infty} g^*(u) e^{\imath 2\pi f u} du \right) \\
&= R_I(m) |G(f)|^2
\end{aligned}$$

$$\begin{aligned}
\Rightarrow \bar{S}_{v_\ell}(f) &= \mathcal{F}\{\bar{R}_{v_\ell}(\tau)\} = \frac{1}{T} \sum_{k=-\infty}^{\infty} \mathcal{F}\{g_k(\tau - kT)\} \\
&= \frac{1}{T} \sum_{k=-\infty}^{\infty} R_I(k) |G(f)|^2 e^{-\imath 2\pi k f T} \\
&= \frac{1}{T} S_I(f) |G(f)|^2 \quad \text{where } S_I(f) = \sum_{k=-\infty}^{\infty} R_I(k) e^{-\imath 2\pi k f T}.
\end{aligned}$$

Theorem 2

$$\bar{S}_{v_\ell}(f) = \frac{1}{T} S_I(f) |G(f)|^2$$

The average power spectrum density of PAM signals is determined by the pulse shape, as well as the input information.

Example

Input information is real and mutually uncorrelated

$$R_I(k) = \begin{cases} \sigma_I^2 + \mu_I^2, & k = 0 \\ \mu_I^2, & k \neq 0 \end{cases}$$

Hence

$$S_I(f) = \sigma_I^2 + \mu_I^2 \sum_{k=-\infty}^{\infty} e^{-j2\pi f k T} = \sigma_I^2 + \frac{\mu_I^2}{T} \sum_{k=-\infty}^{\infty} \delta\left(f - \frac{k}{T}\right)$$

and

$$\bar{S}_{v_e}(f) = \frac{\sigma_I^2}{T} |G(f)|^2 + \frac{\mu_I^2}{T^2} \sum_{k=-\infty}^{\infty} \delta\left(f - \frac{k}{T}\right) |G(f)|^2$$

$$\bar{S}_{v_\ell}(f) = \underbrace{\frac{\sigma_I^2}{T} |G(f)|^2}_{\text{continuous}} + \underbrace{\frac{\mu_I^2}{T^2} \sum_{k=-\infty}^{\infty} \delta\left(f - \frac{k}{T}\right) |G(f)|^2}_{\text{discrete}}$$

- Observation 1: Discrete spectrum vanishes when the input information has zero mean, which is often desirable for digital modulation techniques.
- Observation 2: With a zero-mean input information, the average power spectrum density is determined by $G(f)$.

Example 6

The average power spectrum density for rectangular pulses

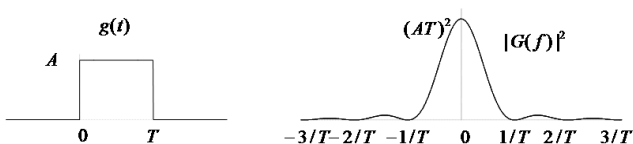
$$g(t) = A[u_{-1}(t) - u_{-1}(t - T)]$$

It shows

$$G(f) = AT \operatorname{sinc}(fT) e^{-j\pi fT} \Rightarrow |G(f)|^2 = A^2 T^2 \operatorname{sinc}^2(fT).$$

Hence

$$\begin{aligned} \bar{S}_{v_\ell}(f) &= \frac{\sigma_I^2}{T} |G(f)|^2 + \frac{\mu_I^2}{T^2} \sum_{k=-\infty}^{\infty} \delta\left(f - \frac{k}{T}\right) |G(f)|^2 \\ &= \sigma_I^2 A^2 T \operatorname{sinc}^2(fT) + \mu_I^2 A^2 \delta(f). \end{aligned}$$



Example 7

The average power spectrum density for raised cosine pulse

$$g(t) = \frac{A}{2} \left[1 + \cos \left(\frac{2\pi}{T} \left(t - \frac{T}{2} \right) \right) \right] (u_{-1}(t) - u_{-1}(t - T)).$$

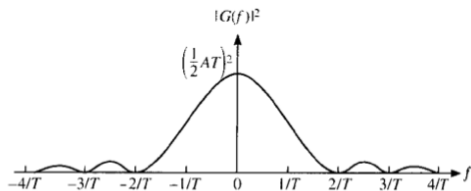
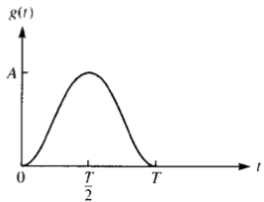
It gives

$$G(f) = \frac{AT}{2} \text{sinc}(fT) \frac{1}{1 - f^2 T^2} e^{-j\pi fT}.$$

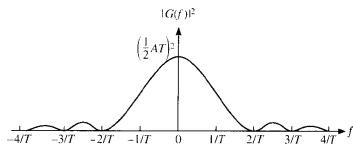
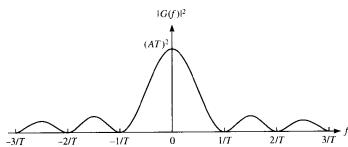
Hence

$$\begin{aligned} \bar{S}_{v_\ell}(f) &= \frac{\sigma_I^2}{T} |G(f)|^2 + \frac{\mu_I^2}{T^2} \sum_{k=-\infty}^{\infty} \delta \left(f - \frac{k}{T} \right) |G(f)|^2 \\ &= \frac{\sigma_I^2 A^2 T \text{sinc}^2(fT)}{4(1 - f^2 T^2)^2} + \frac{\mu_I^2 A^2}{4} \delta(f) + \frac{\mu_I^2 A^2}{16} \delta \left(f - \frac{1}{T} \right) + \frac{\mu_I^2 A^2}{16} \delta \left(f + \frac{1}{T} \right). \end{aligned}$$

$$\text{Note: } \lim_{x \rightarrow \pm 1} \frac{\text{sinc}^2(x)}{(1-x^2)^2} = \frac{1}{4}$$

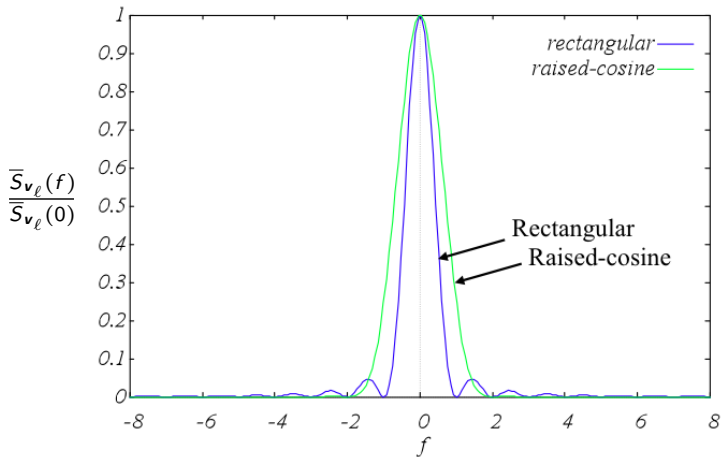


Comparison of the previous two examples

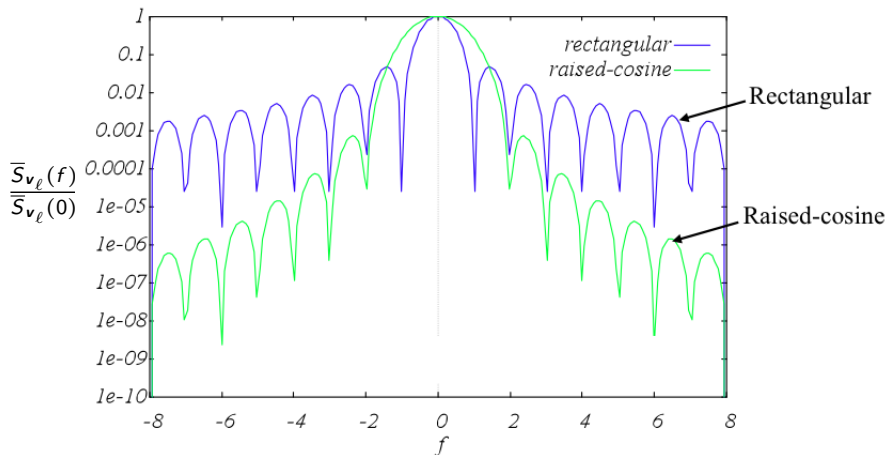


Broader side lobe
Faster decay in tail ($f^{-6} < f^{-2}$)

Assume $A = T = \sigma_I^2 = 1$ and $\mu_I = 0$



Assume $A = T = \sigma_I^2 = 1$ and $\mu_I = 0$



Assume $A = T = \sigma_I^2 = 1$ and $\mu_I = 0$

- The smoother (meaning, continuity of derivatives) the pulse shape, the greater the bandwidth efficiency (lower bandwidth occupancy).
- The raised cosine pulse shape will result in higher bandwidth efficiency than the rectangular pulse shape.

What if I correlated?

Example 8

$$I_n = b_n + b_{n-1}$$

where $\{b_n\}$ mutually uncorrelated with zero mean and unit variance.

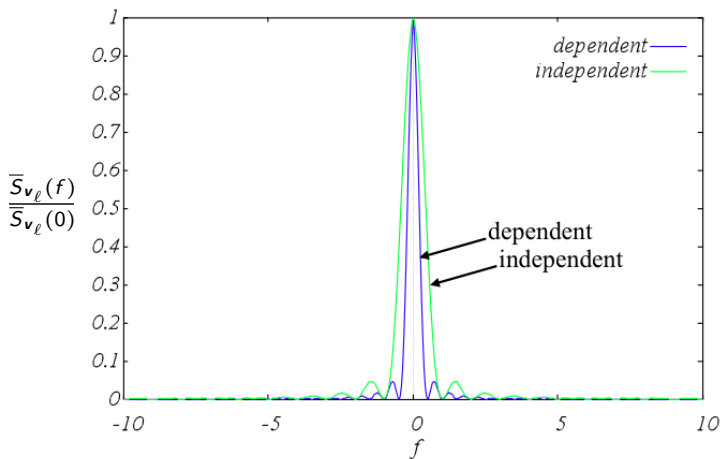
Then,

$$R_I(k) = \begin{cases} 2 & k = 0 \\ 1 & k = \pm 1 \\ 0 & \text{otherwise} \end{cases}$$

$$S_I(f) = 2 + e^{i2\pi fT} + e^{-i2\pi fT} = 2(1 + \cos(2\pi fT)) = 4 \cos^2(\pi fT)$$

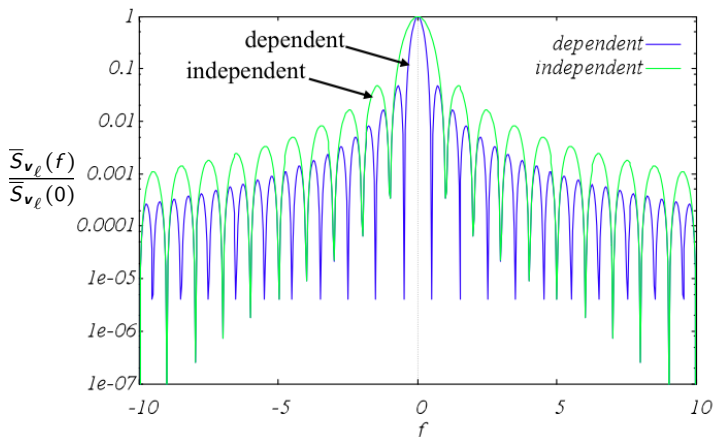
$$\bar{S}_{v_e}(f) = \frac{1}{T} |G(f)|^2 S_I(f) = \frac{4}{T} |G(f)|^2 \cos^2(\pi fT)$$

Rectangular pulse shape with $A = T = 1$



Rectangular pulse shape with $A = T = 1$

Dependence in **transmitted information** (not the original information) can improve the bandwidth efficiency.



Power spectra of CPFSK and CPM

CPM: Assume I i.i.d.

$$\mathbf{v}_\ell(t) = e^{i\phi(t;I)}$$

where

$$\phi(t;I) = 2\pi h \sum_{k=-\infty}^{\infty} I_k q(t - kT)$$

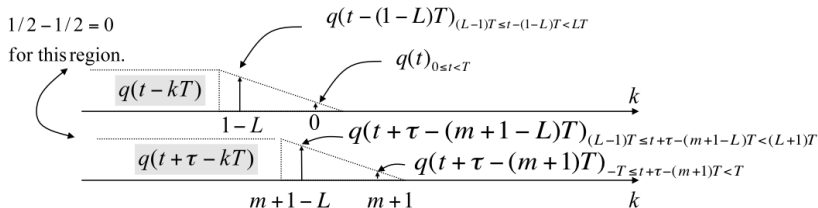
$$\begin{aligned} R_{\mathbf{v}_\ell}(t_1, t_2) &= \mathbb{E}[\mathbf{v}_\ell(t_1)\mathbf{v}_\ell^*(t_2)] \\ &= \mathbb{E}[e^{i\phi(t_1;I)}e^{-i\phi(t_2;I)}] \\ &= \mathbb{E}\left[\exp\left(i2\pi h \sum_{k=-\infty}^{\infty} I_k [q(t_1 - kT) - q(t_2 - kT)]\right)\right] \end{aligned}$$

$$\begin{aligned}
R_{\mathbf{v}_\ell}(t_1, t_2) &= \mathbb{E} \left[\prod_{k=-\infty}^{\infty} \exp(\imath 2\pi h l_k [q(t_1 - kT) - q(t_2 - kT)]) \right] \\
&= \prod_{k=-\infty}^{\infty} \mathbb{E} [\exp(\imath 2\pi h l_k [q(t_1 - kT) - q(t_2 - kT)])] \\
&= \prod_{k=-\infty}^{\infty} \left[\sum_{n \in \mathcal{S}} P_n \exp(\imath 2\pi h n [q(t_1 - kT) - q(t_2 - kT)]) \right],
\end{aligned}$$

where $l_k = n \in \mathcal{S}$ and $P_n \triangleq \Pr[l_k = n]$.

$$\begin{aligned}
\bar{R}_{\mathbf{v}_\ell}(\tau) &= \frac{1}{T} \int_0^T R_{\mathbf{v}_\ell}(t + \tau, t) dt \\
&= \frac{1}{T} \int_0^T \prod_{k=-\infty}^{\infty} \left[\sum_{n \in \mathcal{S}} P_n e^{\imath 2\pi h n [q(t+\tau-kT) - q(t-kT)]} \right] dt.
\end{aligned}$$

Assume $\tau \geq 0$. For $mT \leq \tau = \xi + mT < (m+1)T$ and $0 \leq t < T$ (i.e., the range of integration)



$$t + \tau - (m+1)T = t + \xi - T \text{ and } t + \tau - (m+1-L)T = t + \xi - (1-L)T.$$

$$\bar{R}_{v_\ell}(\tau)$$

$$= \frac{1}{T} \int_0^T \prod_{k=\min\{m+1-L, 1-L\}}^{\max\{m+1, 0\}} \left[\sum_{n \in \mathcal{S}} P_n e^{i2\pi hn [q(t+\tau-kT) - q(t-kT)]} \right] dt$$

$$\stackrel{m \geq 0}{=} \frac{1}{T} \int_0^T \prod_{k=1-L}^{m+1} \left[\sum_{n \in \mathcal{S}} P_n e^{i2\pi hn [q(t+\tau-kT) - q(t-kT)]} \right] dt.$$

Hermitian symmetry of $\bar{R}_{\mathbf{v}_\ell}(\tau)$

It suffices to derive $\bar{R}_{\mathbf{v}_\ell}(\tau)$ for $\tau \geq 0$ because $\bar{R}_{\mathbf{v}_\ell}(-\tau) = \bar{R}_{\mathbf{v}_\ell}^*(\tau)$.

Proof:

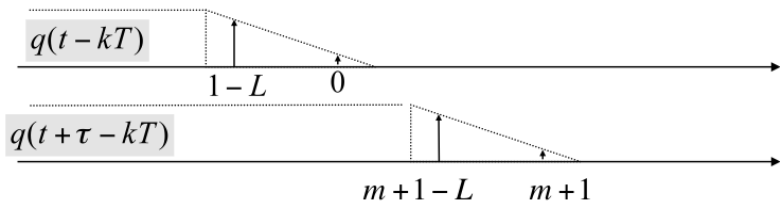
$$\begin{aligned}\bar{R}_{\mathbf{v}_\ell}^*(\tau) &= \frac{1}{T} \int_0^T \mathbb{E} \left[e^{-j2\pi h \sum_{k=-\infty}^{\infty} l_k [q(t+\tau-kT) - q(t-kT)]} \right] dt \\ &= \frac{1}{T} \int_0^T \mathbb{E} \left[e^{j2\pi h \sum_{k=-\infty}^{\infty} l_k [q(t-kT) - q(t+\tau-kT)]} \right] dt \\ &= \frac{1}{T} \int_\tau^{T+\tau} \mathbb{E} \left[e^{j2\pi h \sum_{k=-\infty}^{\infty} l_k [q(v-\tau-kT) - q(v-kT)]} \right] dv \\ &\quad (v = t + \tau) \\ &= \frac{1}{T} \int_0^T \mathbb{E} \left[e^{j2\pi h \sum_{k=-\infty}^{\infty} l_k [q(v-\tau-kT) - q(v-kT)]} \right] dv \\ &= \bar{R}_{\mathbf{v}_\ell}(-\tau).\end{aligned}$$

Average PSD of CPM

$$\begin{aligned}\bar{S}_{\mathbf{v}_\ell}(f) &= \int_{-\infty}^{\infty} \bar{R}_{\mathbf{v}_\ell}(\tau) e^{-i2\pi f\tau} d\tau \\ &= \int_{-\infty}^0 \bar{R}_{\mathbf{v}_\ell}(\tau) e^{-i2\pi f\tau} d\tau + \int_0^{\infty} \bar{R}_{\mathbf{v}_\ell}(\tau) e^{-i2\pi f\tau} d\tau \\ &= \int_0^{\infty} \bar{R}_{\mathbf{v}_\ell}(-\tau) e^{i2\pi f\tau} d\tau + \int_0^{\infty} \bar{R}_{\mathbf{v}_\ell}(\tau) e^{-i2\pi f\tau} d\tau \\ &= \int_0^{\infty} [\bar{R}_{\mathbf{v}_\ell}(\tau) e^{-i2\pi f\tau}]^* d\tau + \int_0^{\infty} \bar{R}_{\mathbf{v}_\ell}(\tau) e^{-i2\pi f\tau} d\tau \\ &= 2\mathbf{Re} \left[\int_0^{\infty} \bar{R}_{\mathbf{v}_\ell}(\tau) e^{-i2\pi f\tau} d\tau \right].\end{aligned}$$

$$\begin{aligned}
& \int_0^{\infty} \bar{R}_{v_\ell}(\tau) e^{-\iota 2\pi f \tau} d\tau \\
&= \int_0^{LT} \bar{R}_{v_\ell}(\tau) e^{-\iota 2\pi f \tau} d\tau + \int_{LT}^{\infty} \bar{R}_{v_\ell}(\tau) e^{-\iota 2\pi f \tau} d\tau \\
&= \int_0^{LT} \bar{R}_{v_\ell}(\tau) e^{-\iota 2\pi f \tau} d\tau + \sum_{m=L}^{\infty} \int_{mT}^{(m+1)T} \bar{R}_{v_\ell}(\tau) e^{-\iota 2\pi f \tau} d\tau.
\end{aligned}$$

For $m \geq L$, the two “regions” below are non-overlapping!



$$\begin{aligned}
\bar{R}_{\mathbf{v}_\ell}(\tau) &\stackrel{m \geq L}{=} \frac{1}{T} \int_0^T \prod_{k=1-L}^{m+1} \left[\sum_{n \in \mathcal{S}} P_n e^{i2\pi hn[q(t+\tau-kT)-q(t-kT)]} \right] dt \\
&= \frac{1}{T} \int_0^T \left(\prod_{k=1-L}^0 \left[\sum_{n \in \mathcal{S}} P_n e^{i2\pi hn[q(t+\tau-kT)-q(t-kT)]} \right] \right. \\
&\quad \left. \prod_{k=1}^{m-L} \left[\sum_{n \in \mathcal{S}} P_n e^{i2\pi hn[q(t+\tau-kT)-q(t-kT)]} \right] \right. \\
&\quad \left. \prod_{k=m+1-L}^{m+1} \left[\sum_{n \in \mathcal{S}} P_n e^{i2\pi hn[q(t+\tau-kT)-q(t-kT)]} \right] \right) dt \\
&= \frac{1}{T} \int_0^T \left(\prod_{k=1-L}^0 \left[\sum_{n \in \mathcal{S}} P_n e^{i2\pi hn[1/2-q(t-kT)]} \right] \right. \\
&\quad \left. \prod_{k=1}^{m-L} \left[\sum_{n \in \mathcal{S}} P_n e^{i2\pi hn[1/2-0]} \right] \right. \\
&\quad \left. \prod_{k=m+1-L}^{m+1} \left[\sum_{n \in \mathcal{S}} P_n e^{i2\pi hn[q(t+\tau-kT)-0]} \right] \right) dt
\end{aligned}$$

$$\begin{aligned}
\bar{R}_{\mathbf{v}_\ell}(\tau) &\stackrel{m \geq L}{=} \frac{1}{T} \int_0^T \left(\prod_{k=1-L}^0 \left[\sum_{n \in \mathcal{S}} P_n e^{i 2\pi h n [1/2 - q(t - kT)]} \right] [\Phi_I(h)]^{m-L} \right. \\
&\quad \left. \prod_{k'=1-L}^1 \left[\sum_{n \in \mathcal{S}} P_n e^{i 2\pi h n [q(t + \tau - k'T - mT)]} \right] \right) dt \quad (k' = k - m) \\
&= [\Phi_I(h)]^{m-L} \lambda(\tau - mT),
\end{aligned}$$

where $\Phi_I(h) = \sum_{n \in \mathcal{S}} P_n e^{i \pi h n}$ and

$$\begin{aligned}
\lambda(\xi) &= \frac{1}{T} \int_0^T \left(\prod_{k=1-L}^0 \left[\sum_{n \in \mathcal{S}} P_n e^{i 2\pi h n [1/2 - q(t - kT)]} \right] \right. \\
&\quad \left. \prod_{k'=1-L}^1 \left[\sum_{n \in \mathcal{S}} P_n e^{i 2\pi h n [q(t + \xi - k'T)]} \right] \right) dt.
\end{aligned}$$

$$\begin{aligned}
& \sum_{m=L}^{\infty} \int_{mT}^{(m+1)T} \bar{R}_{\mathbf{v}_\ell}(\tau) e^{-\imath 2\pi f \tau} d\tau \\
&= \sum_{m=L}^{\infty} \int_{mT}^{(m+1)T} [\Phi_I(h)]^{m-L} \lambda(\tau - mT) e^{-\imath 2\pi f \tau} d\tau \\
&= \sum_{m=L}^{\infty} \int_0^T [\Phi_I(h)]^{m-L} \lambda(\xi) e^{-\imath 2\pi f(\xi+mT)} d\xi \quad (\xi = \tau - mT) \\
&= \left(\sum_{m=L}^{\infty} [\Phi_I(h)]^{m-L} e^{-\imath 2\pi f m T} \right) \left(\int_0^T \lambda(\xi) e^{-\imath 2\pi f \xi} d\xi \right) \\
&= \begin{cases} \left(\frac{e^{-\imath 2\pi f L T}}{1 - \Phi_I(h) e^{-\imath 2\pi f T}} \right) \left(\int_0^T \lambda(\xi) e^{-\imath 2\pi f \xi} d\xi \right) & \text{if } |\Phi_I(h)| < 1 \\ \left(e^{-\imath 2\pi f L T} \sum_{m'=0}^{\infty} e^{-\imath 2\pi T(f - \nu/T)m'} \right) \left(\int_0^T \lambda(\xi) e^{-\imath 2\pi f \xi} d\xi \right) & \text{if } |\Phi_I(h)| = |e^{\imath 2\pi \nu}| = 1 \end{cases} \\
&= \begin{cases} \left(\frac{e^{-\imath 2\pi f L T}}{1 - \Phi_I(h) e^{-\imath 2\pi f T}} \right) \left(\int_0^T \lambda(\xi) e^{-\imath 2\pi f \xi} d\xi \right) & \text{if } |\Phi_I(h)| < 1 \\ e^{-\imath 2\pi f L T} \left(\frac{1}{2} + \frac{1}{2T} \sum_{m'=-\infty}^{\infty} \left(\delta\left(f - \frac{\nu+m'}{T}\right) - \imath \frac{1}{\pi(f - (\nu+m')/T)} \right) \right) \left(\int_0^T \lambda(\xi) e^{-\imath 2\pi f \xi} d\xi \right) & \text{if } |\Phi_I(h)| = |e^{\imath 2\pi \nu}| = 1 \end{cases}
\end{aligned}$$

$$g(t) \leftrightarrow G(f) \Rightarrow \begin{cases} g_{\delta}(t) = \sum_{n=-\infty}^{\infty} g(nT_s)\delta(t - nT_s) \\ G_{\delta}(f) = \sum_{n=-\infty}^{\infty} g(nT_s)e^{-i2\pi nT_s f} = \frac{1}{T_s} \sum_{n=-\infty}^{\infty} G\left(f - \frac{n}{T_s}\right) \end{cases}$$

Slide 2-9: $u_{-1}(t) = \begin{cases} 1, & t > 0 \\ \frac{1}{2}, & t = 0 \\ 0, & t < 0 \end{cases} \leftrightarrow U_{-1}(f) = \frac{1}{2}(\delta(f) - i \frac{1}{\pi f})$

$$\Rightarrow U_{-1,\delta}(f) = \sum_{n=-\infty}^{\infty} u_{-1}(nT_s)e^{-i2\pi nT_s f} = -\frac{1}{2} + \sum_{n=0}^{\infty} e^{-i2\pi nT_s f}$$

$$= \frac{1}{T_s} \sum_{n=-\infty}^{\infty} U_{-1}\left(f - \frac{n}{T_s}\right) = \frac{1}{2T_s} \sum_{n=-\infty}^{\infty} \left(\delta\left(f - \frac{n}{T_s}\right) - i \frac{1}{\pi\left(f - \frac{n}{T_s}\right)} \right)$$

$$\sum_{m'=0}^{\infty} e^{-i2\pi T(f-\nu/T)m'} = \frac{1}{2} + \frac{1}{2T} \sum_{m'=-\infty}^{\infty} \left(\delta\left(f - \frac{\nu + m'}{T}\right) - i \frac{1}{\pi\left(f - \frac{\nu + m'}{T}\right)} \right)$$

We finally obtain a **numerically computable/plotable** formula for the average PSD of CPM. For example, if $|\Phi_I(h)| < 1$,

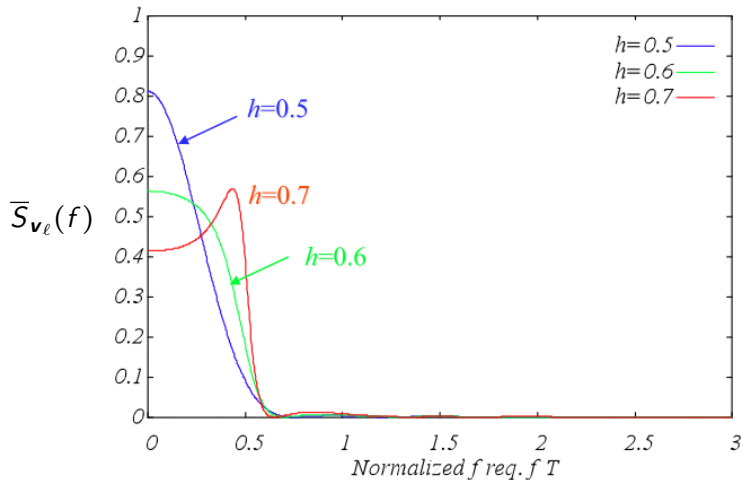
$$\bar{S}_{\mathbf{v}_\ell}(f) = 2 \operatorname{Re} \left[\int_0^{LT} \bar{R}_{\mathbf{v}_\ell}(\tau) e^{-j2\pi f\tau} d\tau + \left(\frac{1}{1 - \Phi_I(h) e^{-j2\pi fT}} \right) \left(\int_0^T \lambda(\xi) e^{-j2\pi f(\xi+LT)} d\xi \right) \right]$$

where for $0 \leq \tau = \xi + mT < LT$,

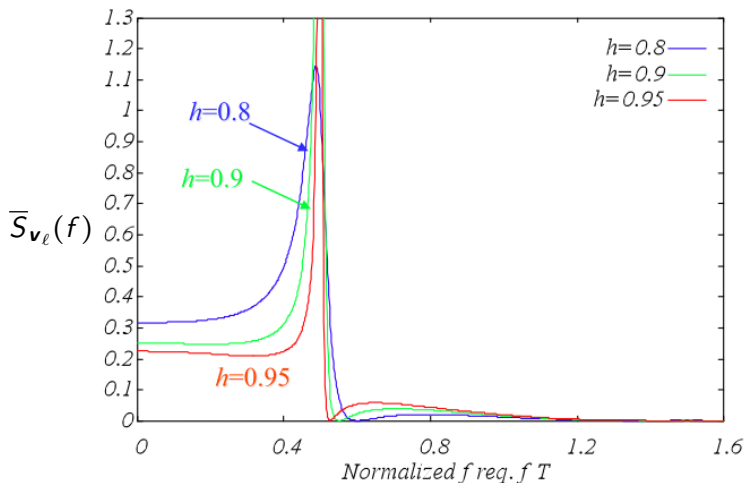
$$\bar{R}_{\mathbf{v}_\ell}(\tau) \stackrel{m \geq 0}{=} \frac{1}{T} \int_0^T \prod_{k=1-L}^{m+1} \left[\sum_{n \in \mathcal{S}} P_n e^{j2\pi hn[q(t+\tau-kT) - q(t-kT)]} \right] dt.$$

However, if $|\Phi_I(h)| = |e^{j2\pi\nu}| = 1$, where $0 \leq \nu < 1$, the average PSD of CPM signals has impulses at $f_{m'} = \frac{\nu+m'}{T}$ for integer m' .

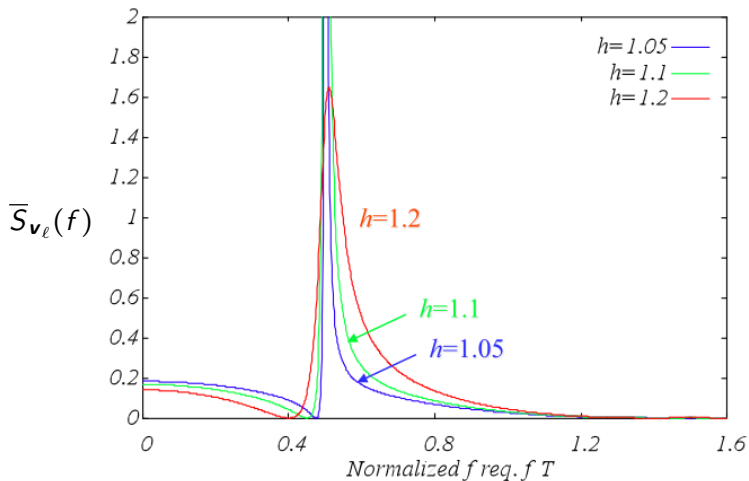
Numerically plotted average PSD of the equivalent lowpass CPFSK signal ($M = 2$, $T = 0.5$, P_n uniform over $\mathcal{S} = \{\pm 1\}$ and $\Phi_I(h) = \frac{1}{2}(e^{j2\pi h} + e^{-j2\pi h}) = \cos(2\pi h)$)



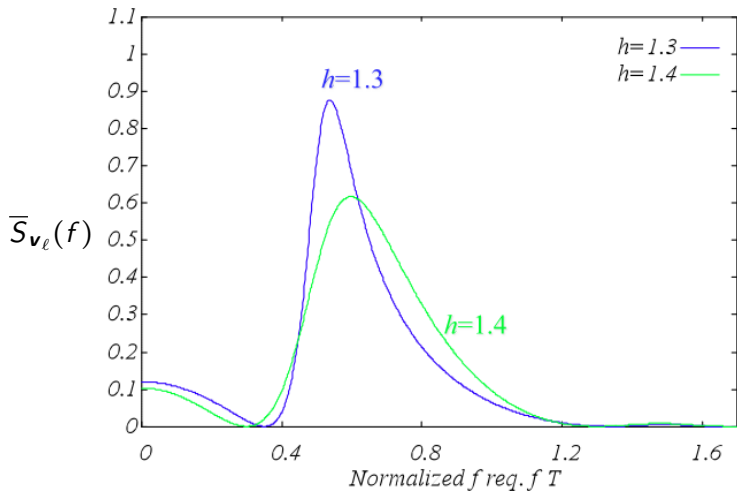
Numerically plotted average PSD of the equivalent lowpass CPFSK signal ($M = 2$, $T = 0.5$ and P_n uniform over $\mathcal{S} = \{\pm 1\}$)



Numerically plotted average PSD of the equivalent lowpass CPFSK signal ($M = 2$, $T = 0.5$ and P_n uniform over $\mathcal{S} = \{\pm 1\}$)



Numerically plotted average PSD of the equivalent lowpass CPFSK signal ($M = 2$, $T = 0.5$ and P_n uniform over $\mathcal{S} = \{\pm 1\}$)



Observation 1

For $h < 1$

- Its average PSD is relatively smooth and well confined.
- Almost all power is confined within

$$fT < 0.6 \text{ or } f < \frac{0.6}{T}$$

where T is the width of the channel symbols.

Observation 2

For $h > 1$

- Its average PSD becomes broader and hence the bandwidth is approximately

$$fT < 1.2 \text{ or } f < \frac{1.2}{T}$$

- This is the main reason why in communication systems, where CPFSK is used, the modulation index h is usually taken to be < 1 .

Example: Bluetooth RF specification (Version 1.0)

- GFSK (Gaussian FSK) with $BT = 0.5$
 - B = Bandwidth (for baseband symbol) = 0.5 MHz,
 $T = 1\mu$ sec
 - 1 = positive frequency deviation
 - 0 = negative frequency deviation
- Modulation index $0.28 \sim 0.35$
 - Modulation index = $2f_d T$, where f_d is the peak frequency deviation.
 - $0.28 < h = 2f_d T < 0.35 \implies 140\text{KHz} < f_d < 175\text{KHz}$

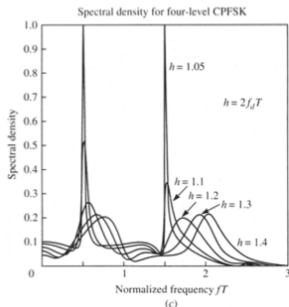
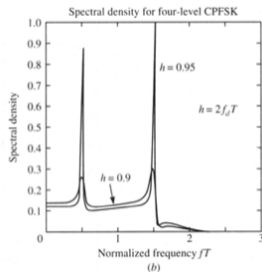
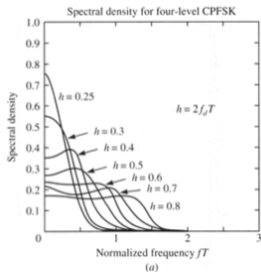
Observation3

By letting $h \rightarrow 1$

- we can observe M impulses in the average PSD of the equivalent lowpass CPFSK signal.

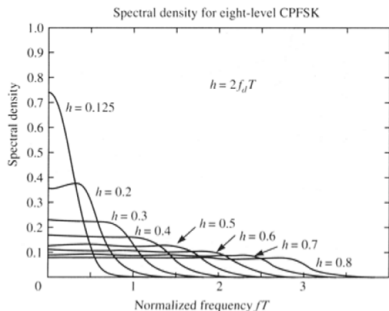
Numerically plotted average PSD of the equivalent low-pass CPFSK signal ($M = 4$, P_n uniform over $\mathcal{S} = \{\pm 1, \pm 3\}$ and $\Phi_I(h) = \frac{1}{2}(\cos(\pi h) + \cos(3\pi h))$)

- Approximately 4 impulses appear when $h \approx 1$.
- The bandwidth becomes broader than almost twice of that of $M = 2$.

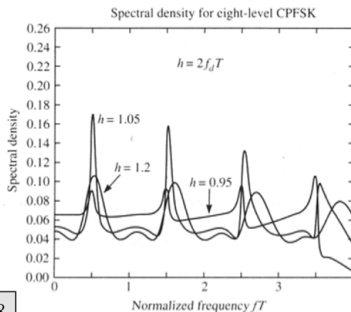


$M = 4$

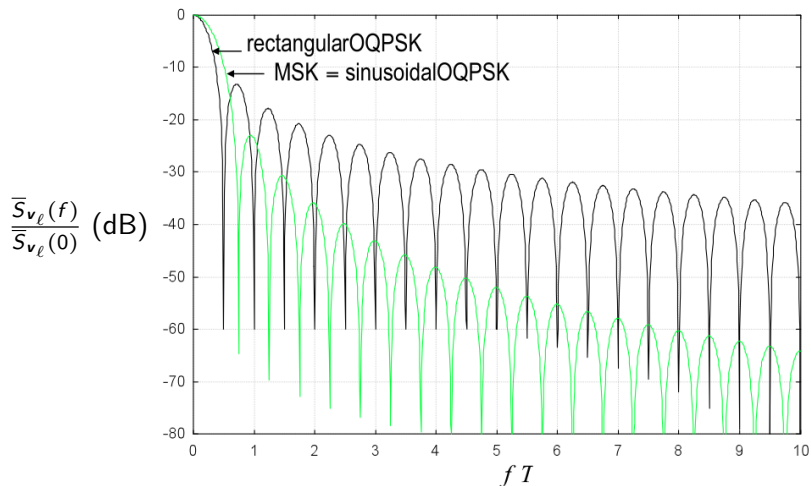
- Approximately 8 impulses are observed when $h \approx 1$.
- Bandwidth becomes broader than almost four times of that of $M = 2$.



$M = 8$



Re-visit MSK versus OQPSK



Observations

- Main Lobe: MSK is 50% wider than rectangular OQPSK, i.e., $\text{MSK} = 1.5 \times \text{rectangular OQPSK}$.
- Side Lobe:
 - Compare the bandwidth that contains 99% of the total power: $\text{MSK} = 1.2/T$ and $\text{rectangular OQPSK} = 8.0/T$.
 - MSK decreases much faster than OQPSK.
 - MSK is significantly more bandwidth efficient than rectangular OQPSK.
 - By further decreasing the modulation index h (i.e., making $h < 1/2$), the bandwidth efficiency of MSKs can be increased. However, in such case, MSK signals are no longer orthogonal. $f_d = 1/(4T) \Leftrightarrow h = 2f_d T = 1/2$

Appendix: Fractional out-of-band power

- Fractional in-band power

$$\Delta P_{\text{In-band}}(W) = \frac{1}{P_{\text{Total}}} \int_{-W}^W \bar{S}_{\mathbf{v}_\ell}(f) df,$$

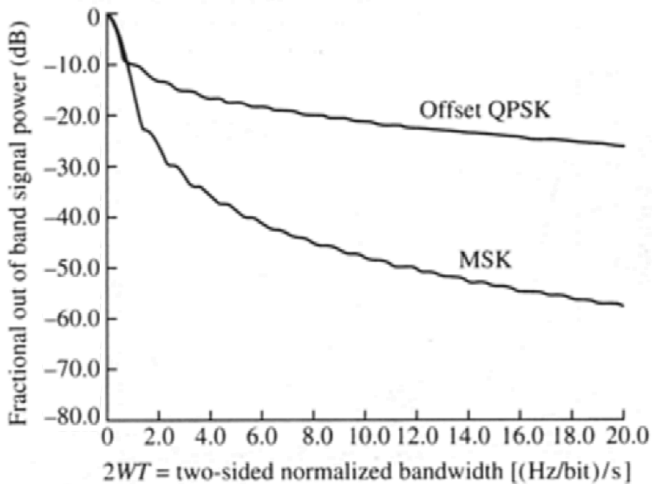
where

$$P_{\text{Total}} = \int_{-\infty}^{\infty} \bar{S}_{\mathbf{v}_\ell}(f) df.$$

- Fractional out-of-band power

$$\Delta P_{\text{Out-of-band}}(W) = 1 - \Delta P_{\text{In-band}}(W)$$

- This quantity is often used to measure the bandwidth efficiency of a modulation scheme. For example, finding the bandwidth W under some acceptable condition, say fractional-out-of-band power is no greater than 0.01.



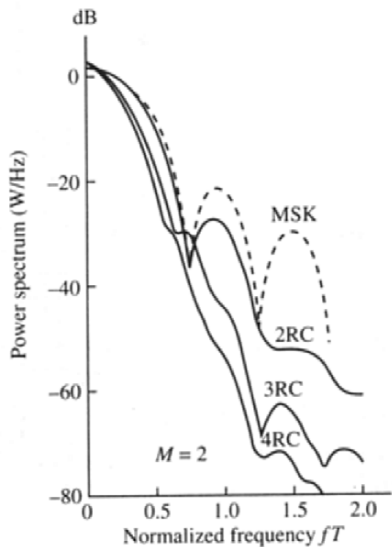
Summary of spectral characteristics of CPFSK signals

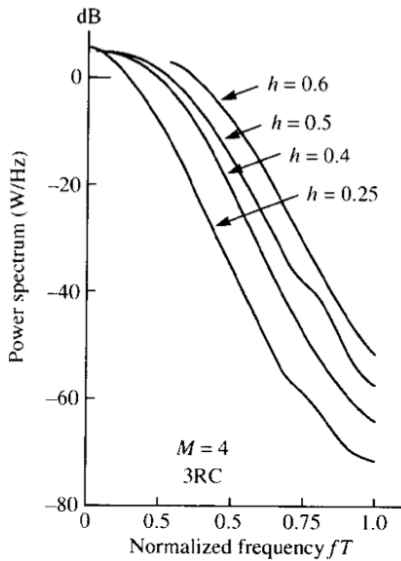
Modulation Index h

- In general, the lower the modulation index h , the higher the bandwidth efficiency.

Pulse shape $g(t)$

- The smoother (meaning, e.g., continuity of the derivatives) the $g(t)$, the greater the bandwidth efficiency.
- For example, the raised cosine $g(t)$ will result in higher bandwidth efficiency than the rectangular $g(t)$.
- For example, LRC (raised cosine $g(t)$ with duration LT) with larger L (i.e., smoother) will result in greater bandwidth efficiency.





What you learn from Chapter 3



- (Pseudo-)Vectorization of standard ASK, PSK and QAM signals
 - Computation of average energy based on signal space vector points
 - Euclidean distance based on signal space vector points
 - Gray code mapping from binary pattern to the signal space vector points (in terms of their Euclidean distances)

- (Good to know) QPSK versus $\pi/4$ -QPSK

- Vectorization of standard orthogonal (FSK or multi-dimensional) and bi-orthogonal signals
 - Computation of average energy based on signal space vector points
 - Euclidean distance based on signal space vector points
 - (Important) Cross-correlation of FSK bandpass and lowpass signals (Minimum shift keying)

- (Good to know) Simplex signals (from orthogonal signals)

- (Important) Why cyclo-stationarity for digitally modulated signals and its power spectrum
- Modulation with memory – CPM signals
 - Its basic formation

$$\phi(t; \mathbf{I}) = 4\pi T f_d \int_{-\infty}^t d(\tau) d\tau$$

based on phase change $d(t) = \sum_{n=-\infty}^{\infty} I_n g(t - nT)$

- (Good to know) Full response and partial response
- MSK versus OQPSK
- Linear representation of CPM
- (Important) Time-average autocorrelation and power spectrum (of cyclostationary PAM and MSK)