

4F5: Advanced Communications and Coding

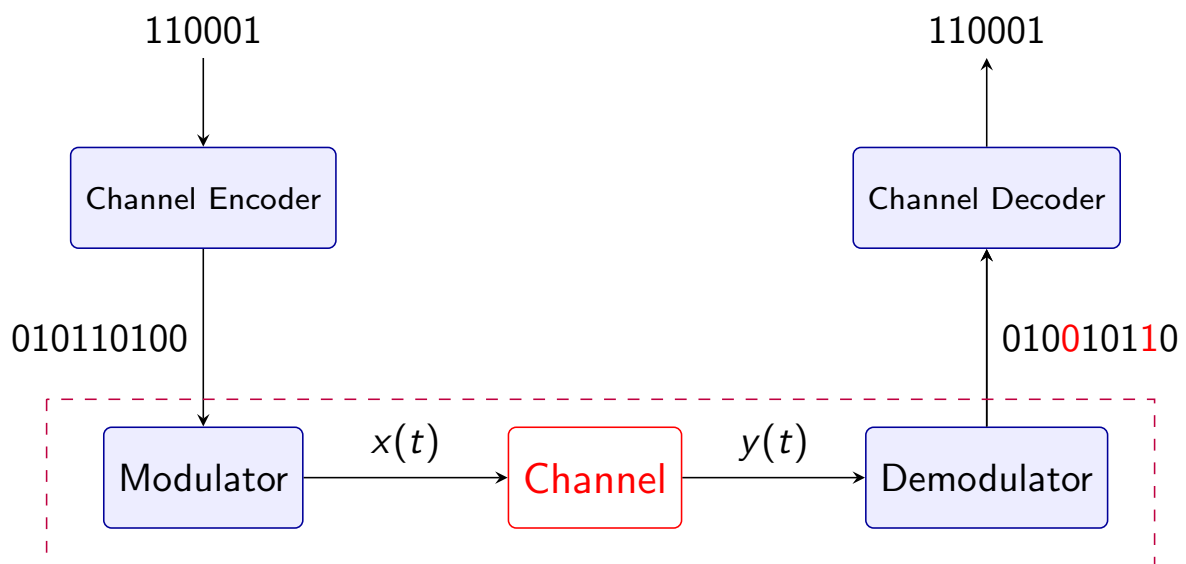
Handout 6: Signal Space and Modulation

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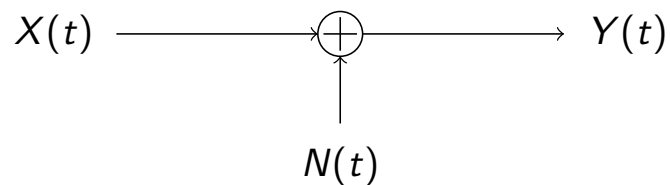


In the last five lectures of this module:

- We will look at some real-world channel models and learn to design modulators & demodulators for these channels
- We begin with the classic additive white Gaussian noise (AWGN) channel, and then study wireless channels, which have “fading” in addition to additive noise

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The Additive White Gaussian Noise (AWGN) Channel



The *continuous-time* AWGN channel:

$$Y(t) = X(t) + N(t)$$

- ① Input $X(t)$ is *power-limited* to P :
 \Rightarrow Average energy over time T must be $\leq P$ for large T
- ② $X(t)$ is *band-limited* \Rightarrow
Fourier Transform of $X(t)$ must be zero outside the interval $[f_c - W, f_c + W]$, where f_c is the carrier frequency and $f_c \gg W$
- ③ Noise $N(t)$ is a random process assumed to be *white* Gaussian
This means that for each t , $N(t)$ is a Gaussian rv with

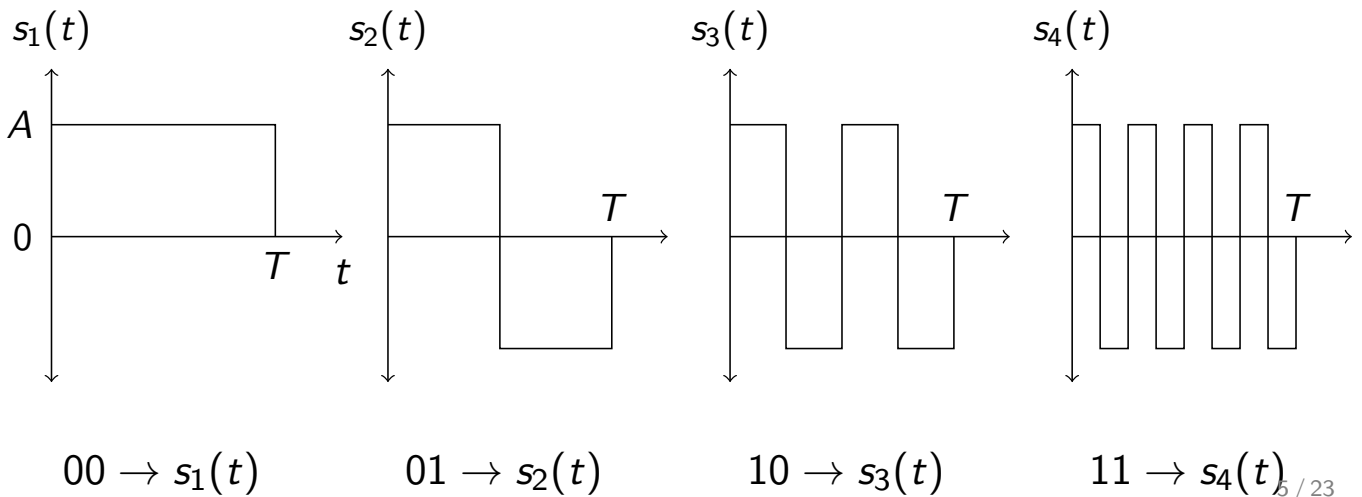
$$\mathbb{E}[N(t)] = 0, \quad \mathbb{E}[N(t)N(s)] = \frac{N_0}{2} \delta(t - s), \quad \text{for all } t, s$$

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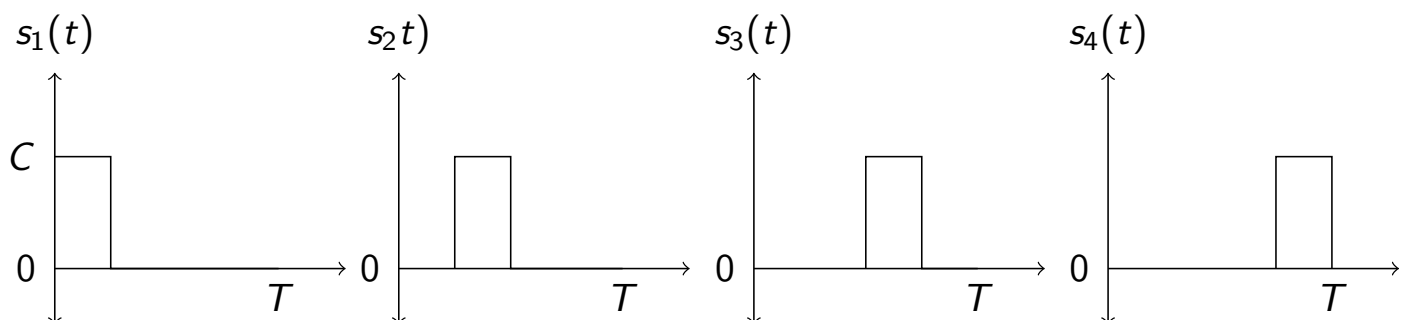
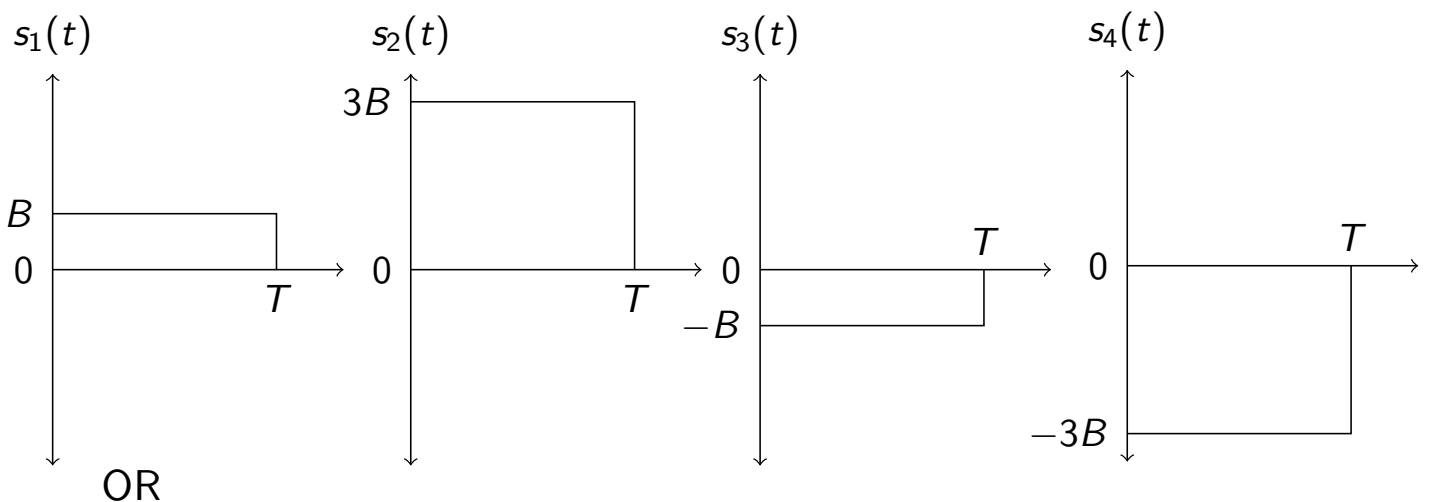
Mapping bits to waveforms

- Suppose we want to transmit m bits in T seconds using a waveform $x(t)$, $t \in [0, T)$.
- Note that each m -bit pattern indexes one of $M = 2^m$ messages
- We choose a collection of M waveforms $\{s_1(t), \dots, s_M(t)\}$. Map each m -bit sequence to one of these waveforms.

Example: $m = 2$ bits. One choice of $M = 4$ waveforms could be



We could also choose something like:



- We need a systematic way to represent finite-energy functions defined in an interval $[0, T]$.
- We will see that by fixing a set of basis functions, signal waveforms can be represented as “vectors”, which obey properties similar to vectors in Euclidean space \mathbb{R}^n .
- Let us first review what a vector space is . . .

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Vector Spaces

A vector space \mathcal{V} is a set of elements (called “vectors”) that is **closed under addition and scalar multiplication**. The scalars are often chosen to be real or complex-valued, but they can be chosen over some other field as well.

That is, if $\underline{v}_1, \underline{v}_2 \in \mathcal{V}$, then $a\underline{v}_1 + b\underline{v}_2 \in \mathcal{V}$, for any scalars a, b .

You are already familiar with the Euclidean vector space \mathbb{R}^k .

A set of *linearly independent* vectors, say $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_k\}$, is called a **basis** of the vector space \mathcal{V} , if $\underline{v} \in \mathcal{V}$ can be expressed as a linear combination of the form

$$\underline{v} = a_1 \underline{v}_1 + a_2 \underline{v}_2 + \dots + a_k \underline{v}_k$$

for some scalars a_1, \dots, a_k .

If the number of vectors in the basis is a finite number k , the vector space is said to have *dimension* k .

Q: Specify a basis for \mathbb{R}^3 . Is the basis unique?

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Inner product, Orthogonality, Norm

You are familiar with the *inner-product* (“dot-product”) in Euclidean space \mathbb{R}^n : if

$$\underline{v}_1 = x_1 \underline{e}_1 + x_2 \underline{e}_2 + \dots + x_n \underline{e}_n,$$

$$\underline{v}_2 = y_1 \underline{e}_1 + y_2 \underline{e}_2 + \dots + y_n \underline{e}_n,$$

then

$$\langle \underline{v}_1, \underline{v}_2 \rangle = x_1 y_1 + \dots + x_n y_n.$$

(Here $\underline{e}_1, \dots, \underline{e}_n$ are the unit vectors along the n axes of \mathbb{R}^n)

- Recall that vectors \underline{v}_1 and \underline{v}_2 are *orthogonal* if $\langle \underline{v}_1, \underline{v}_2 \rangle = 0$. E.g. in \mathbb{R}^n , $\underline{e}_i, \underline{e}_j$ are orthogonal iff $i \neq j$.
- The **norm** (“length”) of a vector \underline{v} is defined as

$$\|\underline{v}\| = \sqrt{\langle \underline{v}, \underline{v} \rangle}$$

We will now extend these concepts to a vector space of *signals*. This vector space is called the *signal space*.

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The Signal Space

To understand communication over the continuous-time channel

$$Y(t) = X(t) + N(t),$$

it is useful to consider the vector-space of *finite-energy* signals.

Let \mathcal{L}_2 be the set of complex-valued signals (functions) $x(t)$ with finite energy, i.e.,

$$\int_{-\infty}^{\infty} |x(t)|^2 dt < \infty$$

It can be shown that \mathcal{L}_2 is a vector space (set of finite-energy signals is closed under addition and scalar multiplication.)

- The inner product in this space can be defined as follows. For $x(\cdot), y(\cdot) \in \mathcal{L}_2$,

$$\langle x, y \rangle = \int_{-\infty}^{\infty} x(t) y^*(t) dt$$

- The norm of a signal is the square-root of its energy:

$$\|x\| = \sqrt{\langle x, x \rangle} = \left[\int_{-\infty}^{\infty} |x(t)|^2 dt \right]^{1/2}$$

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Orthonormal Basis

For any vector space $\mathcal{L} \subset \mathcal{L}_2$, the set of functions $\{f_i(\cdot), i = 1, 2, \dots\}$ is called an **orthonormal** basis for \mathcal{L} if

- 1 Every $x(\cdot) \in \mathcal{L}$ can be expressed as

$$x(t) = \sum_i x_i f_i(t),$$

for some *scalars* x_1, x_2, \dots , and

- 2 The functions $\{f_i(\cdot), i = 1, 2, \dots\}$ are *orthonormal*, i.e.,

$$\langle f_\ell, f_m \rangle = \int f_\ell(t) f_m^*(t) dt = \begin{cases} 1 & \text{if } \ell = m \\ 0 & \text{if } \ell \neq m \end{cases}$$

If these two conditions are satisfied, the orthonormal basis $\{f_i\}$ is said to *span* the vector space \mathcal{L} .

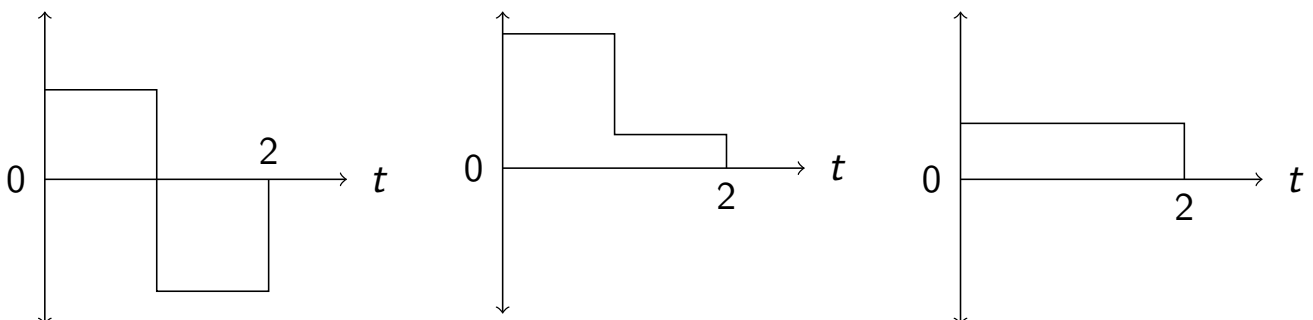
The number of elements (functions) in the basis is called the *dimension* of the vector space \mathcal{L}

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Example

Let \mathcal{L} be the set of all functions $f(t)$ defined for $t \in [0, 2]$ that are piece-wise constant in the intervals $[0, 1]$ and $(1, 2]$.

E.g., some such functions are:



- This set of functions is a vector space (why?)
- An orthonormal basis for this space is:



What is the dimension of this vector space?

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Exercise: Find an orthonormal basis for the vector space spanned by each of the three sets of functions in p.5–6. Also specify the dimension of the space in each case.

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Why do we care about an orthonormal basis for a given vector space \mathcal{L} ? Consider any function $x(\cdot) \in \mathcal{L}$, expressed in terms of an orthonormal basis $\{f_i(t)\}$ as

$$x(t) = \sum_i x_i f_i(t).$$

Each coefficient x_j can be calculated as

$$x_j = \langle x, f_j \rangle = \int \left(\sum_i x_i f_i(t) \right) f_j^*(t) dt = \sum_i x_i \int f_i(t) f_j^*(t) dt = x_j$$

The coefficients x_1, x_2, \dots are called the *projections* of the signal $x(t)$ along $f_1(t), f_2(t), \dots$, respectively.

The inner product between $x(t)$ and $y(t)$ is

$$\langle x(t), y(t) \rangle = \int \left(\sum_i x_i f_i(t) \right) \left(\sum_j y_j f_j(t) \right)^* dt = \sum_i x_i y_i^*$$

The energy of $x(t)$ can thus be written as

$$\int |x(t)|^2 dt = \langle x(t), x(t) \rangle = \sum_i |x_i|^2$$

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Thus, if we fix an orthonormal basis for \mathcal{L} , we can treat functions in \mathcal{L} just like vectors in Euclidean space, i.e.,

$$x(t) \leftrightarrow (x_1, x_2, \dots, x_k),$$

where k is the number of elements in the orthonormal basis, i.e., the *dimension* of the signal space.

- We can use this vector representation to easily compute inner products between signals, energies etc.
- We have effectively converted continuous-time operations (integrals over t) into discrete-time operations (summations)!

Signal space is a convenient framework to analyse modulation and demodulation.

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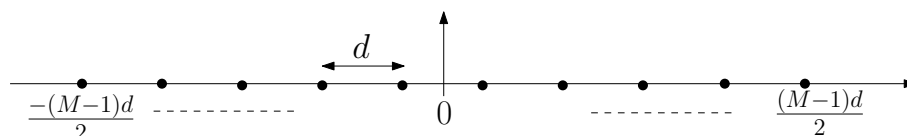
Modulation Techniques

We now describe some common modulation techniques in terms of their associated signal space over one symbol period $[0, T)$.

M-ary Pulse Amplitude Modulation (PAM):

$$x(t) = x \cdot p(t), \quad t \in [0, T)$$

- 1 $p(t)$ is a unit energy pulse, e.g., $p(t) = \frac{1}{\sqrt{T}}$ for $t \in [0, T)$
- 2 The symbol x can take one of M values, chosen from the set $\left\{ - (2i - 1 - M) \frac{d}{2} \right\}, i = 1, \dots, M$



$x(t)$ lies in a 1-dim. signal space with basis $f_1(t) = p(t)$. Hence

$$x(t) \longleftrightarrow x$$

where x belongs to the above constellation.

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Quadrature Amplitude Modulation (QAM)

In M -ary QAM, the symbol x is chosen from a *complex* constellation with M symbols. Over one symbol period:

$$x(t) = \text{Re}(x)p(t)\sqrt{2}\cos(2\pi f_c t) + \text{Im}(x)p(t)\sqrt{2}\sin(2\pi f_c t), \quad t \in [0, T)$$

- As in PAM, $p(t)$ is a baseband unit energy pulse, whose spectrum is (almost) zero for frequencies outside $[-W, W]$
- The carrier frequency $f_c \gg W \approx \frac{1}{2T}$. For example, $f_c = 1 \text{ GHz}$, $\frac{1}{T} = 1 \text{ MHz}$
- The real part of the symbol modulates the cosine, the imaginary part modulates the sine.

$x(t)$ lies in 2-dim. signal space with orthonormal basis functions

$$f_1(t) = \sqrt{2}p(t)\cos(2\pi f_c t), \quad f_2(t) = \sqrt{2}p(t)\sin(2\pi f_c t),$$

Therefore $x(t) = \sum_{k=1}^2 x_k f_k(t)$ where $[x_1, x_2] = [\text{Re}(x), \text{Im}(x)]$, where x belongs to the constellation

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The Orthonormal Basis for QAM

We want to show that

$$\int_0^T f_1^2(t)dt = 1, \quad \int_0^T f_2^2(t)dt = 1, \quad \int_0^T f_1(t)f_2(t)dt = 0$$

We have

$$\begin{aligned} \int_0^T f_1^2(t)dt &= \int_0^T 2p^2(t)\cos^2(2\pi f_c t)dt = \int_0^T p^2(t)(1 + \cos(4\pi f_c t))dt \\ &= 1 + \int_0^T p^2(t)\cos(4\pi f_c t)dt \end{aligned}$$

Note $\cos(4\pi f_c t)$ has period $\frac{1}{2f_c} \ll T$. Assume $T = \frac{K}{2f_c}$ for some integer K . So we can split the above integral as

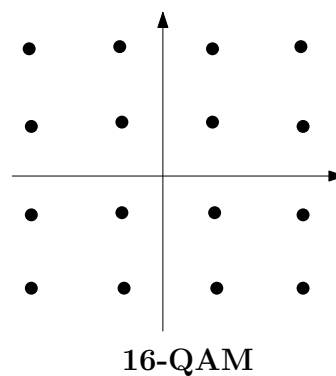
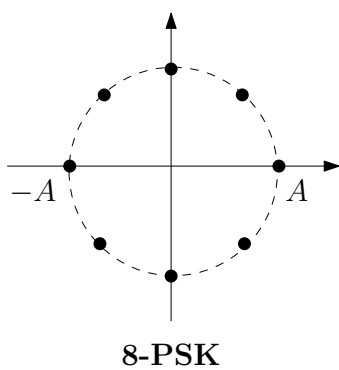
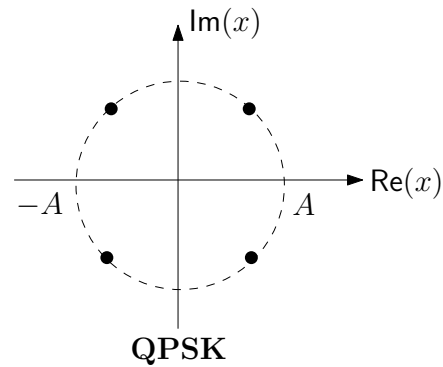
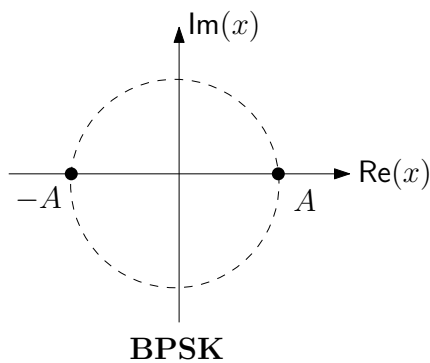
$$\int_0^{\frac{1}{2f_c}} p^2(t)\cos(4\pi f_c t)dt + \int_{\frac{1}{2f_c}}^{\frac{2}{2f_c}} p^2(t)\cos(4\pi f_c t)dt \dots + \int_{\frac{(K-1)}{2f_c}}^{\frac{K}{2f_c}} p^2(t)\cos(4\pi f_c t)dt \quad (1)$$

As $p(t)$ is a baseband signal, it can be assumed to be nearly constant over a period of $\frac{1}{2f_c}$. Therefore each of the K integrals ≈ 0 , and hence $\int f_1^2(t) \approx 1$. Similar calculations show that $\int f_2^2(t) \approx 1$ and $\int f_1(t)f_2(t)dt \approx 0$.

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Some QAM Constellations

$$x_{QAM}(t) = \text{Re}(x)p(t)\sqrt{2} \cos(2\pi f_c t) + \text{Im}(x)p(t)\sqrt{2} \sin(2\pi f_c t), \quad t \in [0, T)$$



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PAM and QAM are digital versions of amplitude modulation as the information symbols modulate the amplitude of the pulse/carrier.

We now study at a modulation scheme where the information symbols modulate the *frequency* of the carrier.

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Orthogonal Signalling (M-ary FSK)

To send message $i \in \{1, \dots, M\}$ in the symbol period $[0, T)$:

$$x(t) = \sqrt{\frac{2E_s}{T}} \cos \left(2\pi \left(f_c + (2i - (M + 1)) \frac{\Delta_f}{2} \right) t \right),$$

where $\Delta_f = \frac{1}{2T} = \frac{f_c}{K}$ for some large integer K

- The M symbols are represented by M cosines, whose frequencies are separated by multiples of Δ_f :

$$f_c - \frac{(M-1)}{2} \Delta_f, \quad f_c - \frac{(M-3)}{2} \Delta_f, \dots, \quad f_c + \frac{(M-1)}{2} \Delta_f$$

- We have an M -**dimensional** signal space with ortho-basis:

$$f_i(t) = \left\{ \sqrt{\frac{2}{T}} \cos \left(2\pi \left(f_c + (2i - (M + 1)) \frac{\Delta_f}{2} \right) t \right) \right\},$$

$$t \in [0, T), \quad i = 1, \dots, M.$$

- Hence $x(t) = \sum_{k=1}^M x_k f_k(t)$. If the i th message is transmitted, the only non-zero term in the sum is $x_i = \sqrt{E_s}$, i.e.,

$$[x_1, \dots, x_i, \dots, x_M] = [0, \dots, \sqrt{E_s}, \dots, 0]$$

- E_s is the energy of $x(t)$ in one period $[0, T)$

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Orthonormal Basis for M-ary FSK

We want to show that $\int_0^T f_i^2(t) dt = 1$ for all $i \in \{1, \dots, M\}$, and $\int_0^T f_i(t) f_j(t) dt = 0$ for $i \neq j$.

$$\begin{aligned} \int_0^T f_i^2(t) dt &= \frac{1}{T} \int_0^T \left[1 + \cos \left(4\pi \left(f_c + (2i - (M + 1)) \frac{\Delta_f}{2} \right) t \right) \right] dt \\ &= 1 - \frac{1}{cT} \underbrace{\sin \left(4\pi \left(f_c + (2i - (M + 1)) \frac{\Delta_f}{2} \right) T \right)}_{= 0 \text{ as } f_c T = \frac{K}{2} \text{ and } \Delta_f T = \frac{1}{2}} \end{aligned}$$

Similarly, for $i \neq j$

$$\begin{aligned} \int_0^T f_i(t) f_j(t) dt &= \frac{1}{T} \int_0^T \left[\cos(2\pi (2f_c + ((i+j) - (M+1)) \Delta_f) t) \right. \\ &\quad \left. - \cos(2\pi (i-j) \Delta_f t) \right] dt \\ &= \frac{1}{cT} \underbrace{\sin(2\pi (2f_c + ((i+j) - (M+1)) \Delta_f) T)}_0 - \frac{1}{c'T} \underbrace{\sin(2\pi (i-j) \Delta_f T)}_0 \end{aligned}$$

(c, c' are constants whose exact values don't matter)

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Demodulation (without noise)

To summarize, we considered modulation schemes of the form:

$$\text{Message} \in \{1, \dots, M\} \longleftrightarrow [x_1, \dots, x_K] \longrightarrow \sum_{i=1}^K x_i f_i(t)$$

where

- $\{f_i(t)\}_{i=1}^K$ is a set of functions which form an orthonormal basis for the signal space of $x(t)$.
- $K = 1$ for PAM, $K = 2$ for QAM, $K = M$ for M -ary FSK.

Demodulation is the process of recovering the message — or, equivalently $[x_1, \dots, x_K]$ — from the received waveform $y(t)$.

- Assuming there is no noise, $y(t) = x(t)$
- To recover x_j , compute inner product of (x) with $f_j(t)$:

$$\langle x, f_j \rangle = \int \left(\sum_i x_i f_i(t) \right) f_j^*(t) dt = \sum_i x_i \int f_i(t) f_j^*(t) dt = x_j$$

- We can recover $[x_1, \dots, x_K]$ (and hence the message) by just computing inner products of $x(t)$ with $f_1(t), \dots, f_K(t)$.
- In the next handout, we'll study how to perform demodulation in the presence of noise.