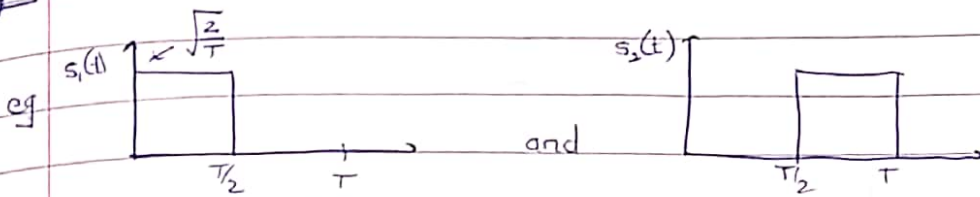


V NON-COHERENT DETECTION



Coherent :- We need f_c and ϕ to do template matching

Non-coherent :- No information encoded in f_c and ϕ

In above example, find energy in $(0, T/2)$ and $(T/2, T)$, conclude symbol based on which e is higher

- Can't use BPSK - no information in phase allowed $\therefore -1 = 1e^{j\pi}$ (energy will not be different.)
- Can't use QPSK at all.
- In PAM-4, can't distinguish b/w ± 3 or ± 1 .

• Since we don't care about phase,

$$y = s_p e^{j\theta} + w$$

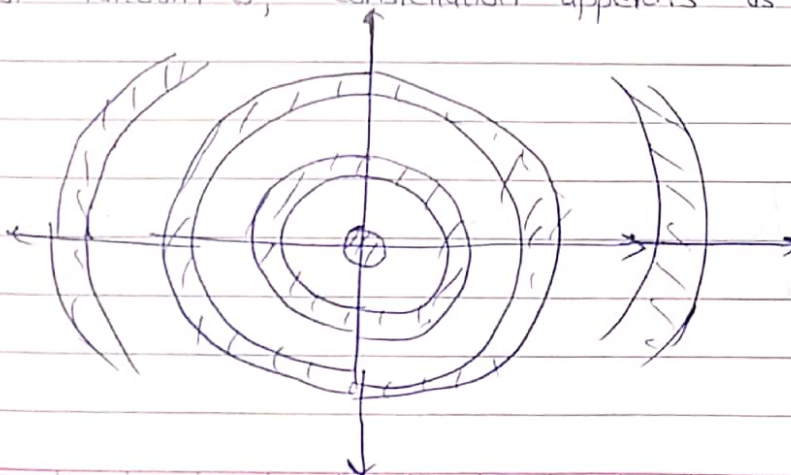
θ is random, because of f_c & ϕ mismatch.

eg $s_i(t) = i \text{ rect}\left(\frac{t-T}{T}\right)$ for $i = 0, 1, 2, 3$.

Because of random θ , constellation appears as concentric circles

(bands)

of radii $\approx 0, 1, 2, 3$



😊 No need to detect exact f_c and ϕ .

😊 When you find energy, we ~~are~~ get ω^2 term, which is no longer Gaussian. **it is chi squared distribution**
 - Minimum distance decoder may not be optimal.
 (still reasonable)

→ MPE

Coherent:- $\arg \max_i \langle y, s_i \rangle$
 Non-coherent:- $\arg \max_i |\langle y, s_i \rangle|$

non coherent estimation is same as coherent estimation but with a phase offset. But here we know that there is no information stored in the phase so we can take mod and $e^{j\theta}$ term will go away

A] Complex Gaussian.

Proper Complex Gaussian:-

U, V are complex jointly Gaussian vectors
 $E[(U - E(U))(V - E(V))] = 0$

$X_c + jX_s$

- N complex Gaussian vectors $\overset{\text{equivalent}}{\equiv} 2N$ real Gaussian vectors
- Elements of ~~cover~~ ^{covariance} correlation matrix can be complex for complex Gaussian.

$$C = \begin{bmatrix} C_{cc} & C_{sc} \\ C_{cs} & C_{ss} \end{bmatrix} = E \left(\begin{bmatrix} X_c \\ X_s \end{bmatrix} \begin{bmatrix} X_c & X_s \end{bmatrix} \right)$$

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- IF $X_{n \times 1} = \begin{bmatrix} x_{c_{n \times 1}} \\ x_{s_{n \times 1}} \end{bmatrix}$ is a $2n \times 1$ matrix
'Covariance matrix'

$$2n \times 2n \text{ covariance matrix} = \begin{bmatrix} C_{cc} & C_{cs} \\ C_{sc} & C_{ss} \end{bmatrix}$$

- C_{cs} and C_{sc} are transposes of each other, ~~as~~ but may not be equal.
- All four elements are real.
- $C_x \triangleq C_{cc} + C_{ss} + j(C_{sc} - C_{cs})$

Then $C_x^H = C_x$

- IF $\langle s_0, s_1 \rangle = 0$, then $\langle n, s_0 \rangle$ and $\langle n, s_1 \rangle$ are independent with variance $2\sigma^2 \|s_i\|^2$

$$\rightarrow \text{cov}(\langle n, s_0 \rangle, \langle n, s_1 \rangle) = \langle s_0, s_1 \rangle$$

- In C_x , relation between real and imaginary parts is unknown.

But if noise is 'proper', $C_{sc} = C_{cs}$ makes life chill.

$$H_1: y = s_1 e^{j\theta} + n$$

$$z_0 = \langle y, s_0 \rangle$$

$$z_1 = \langle y, s_1 \rangle$$

$$P_{\text{elo}} = P[|z_1| > |z_0| | H_0]$$

$$\text{Define } \underline{z} = \begin{bmatrix} z_0 \\ z_1 \end{bmatrix}$$

→ Mean :-

$$E[\langle y, s_0 \rangle] = E[\langle s_0(l) e^{j\theta} + n(l), s_0(l) \rangle] \\ = e^{j\theta} \|s_0\|^2 + E[\langle n, s_0 \rangle]$$

* Over one symbol duration, θ is ~ constant
 θ is the same for both H_0 and H_1

$$\therefore \text{Mean of } \underline{z} = \underline{m_z} = \begin{bmatrix} e^{j\theta} \|s_0\|^2 \\ e^{j\theta} \langle s_0, s_1 \rangle \end{bmatrix}$$

$$\rightarrow \text{Correlation}_1 = \overset{\text{coefficient}}{\rho} = \frac{\langle s_0, s_1 \rangle}{\|s_0\| \|s_1\|} \quad \dots S \text{ is complex}$$

$$\rightarrow C_z = 2\sigma^2 e^{j\theta} \begin{bmatrix} 1 & \rho^* \\ \rho & 1 \end{bmatrix} \quad \star \text{Verify}$$

$$\rightarrow \text{Use } \|s_0\| = \|s_1\| = \sqrt{E_s}$$

$$\therefore \underline{m_z} = e^{j\theta} \begin{bmatrix} \|s_0\|^2 \\ \langle s_0, s_1 \rangle \end{bmatrix}$$

$$= e^{j\theta} E_s \begin{bmatrix} 1 \\ \frac{\langle s_0, s_1 \rangle}{\|s_0\| \|s_1\|} \end{bmatrix}$$

→ For C_z ,

$$E[|z_0 - E_s e^{j\theta}|^2]$$

$$= E[(z_0 - E_s e^{j\theta})(z_0^* - E_s^* e^{-j\theta})]$$

⋮

$$\text{Gives } C_z = 2\sigma^2 e^{j\theta} \begin{bmatrix} 1 & s^* \\ s & 1 \end{bmatrix}$$

Z0 and Z1

* We have 2 complex RV's \Rightarrow 4 real RV's.

Usually we will require 4 C_z correlation values.

But because both are 'white' noises, we require only 2 correlation values ($\text{Re}(s)$ and $\text{Im}(s)$)

$$* E[\langle y, s_1 \rangle \langle y, s_1 \rangle^*] = E\left[\int y(t) s_1^*(t) dt \int y^*(\tau) s_1(\tau) d\tau\right]$$

$$= E\left[\iint s_1^*(t) s_1(\tau) y(t) y^*(\tau) dt d\tau\right]$$

$$= \int s_1^*(t) s_1(\tau) E\left[\int y(t) y^*(\tau) dt d\tau\right]$$

→ Performance depends only on $|s|$

$$\langle s_0, s_1 \rangle = E_s \quad \Rightarrow \quad \langle s_0, s_1 \rangle / (|s_0| |s_1|)$$

has magnitude atmost 1 so can be represented as $\cos\theta + j\sin\theta$

$s_1 \rightarrow s_1 e^{j\arg(s)}$ We will redesign s_1 as $s_1 e^{j\arg(s)}$

$$\therefore \langle s_0, s_1 e^{j\arg(s)} \rangle = \langle s_0, s_1 \rangle e^{-j\arg(s)}$$

$$= E_s s e^{-j\arg(s)}$$

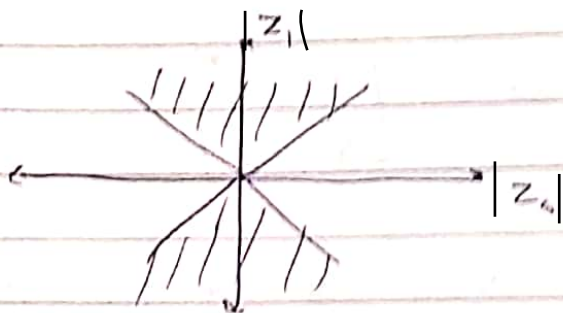
$$= E_s |s|$$

P.T.O.

- New $m_z = E_s e^{j\theta} \begin{bmatrix} 1 \\ |s| \end{bmatrix}$

$$C_z = 2\sigma^2 E_s \begin{bmatrix} 1 & |s| \\ |s| & 1 \end{bmatrix}$$

- Hypothesis : $P(|z_1| > |z_0| \mid H_0)$



This BER comes out to be $e^{-\frac{E_s}{2N_0} (1 - |s|)}$

$$\sim e^{-\text{SNR}}$$

* The Q function decays faster than $e^{-\text{SNR}}$

\therefore BER of coherent < BER of non-coherent

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* Mean has power terms, Covariance has power² terms
 \Rightarrow We are detecting power instead of phase, amplitude, etc.

• Transformation

$$z \rightarrow v$$

(Divide by $\sigma\sqrt{E_s}$)

$$\text{New } m_v = \frac{\sqrt{E_s}}{\sigma} \begin{bmatrix} 1 \\ |s| \end{bmatrix}$$

$$C_v = 2 \begin{bmatrix} 1 & |s| \\ |s| & 1 \end{bmatrix}$$

$$z_0 = \langle y, s_0 \rangle = \langle (s_0^* e^{j\theta} + n), s_0 \rangle$$

$$|z_0| = |e^{j\theta} E_s + \langle n, s_0 \rangle|$$

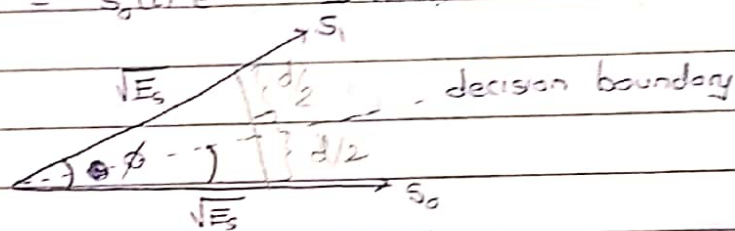
$$|z_0 e^{-j\theta}| = |E_s + \langle n e^{-j\theta}, s_0 \rangle|$$

$n e^{-j\theta}$ and n are identically distributed

\therefore The above two expressions are equal

Put $z \rightarrow \frac{z}{\sigma \sqrt{E_s}}$ to obtain above V , m_0 and C_0

eg $H_0: y(t) = s_0(t) e^{j\theta} + n(t)$



* If $u(t) = \alpha s_0(t)$ and $v(t) = \beta s_1(t)$,

angle between u and v

$$= \phi = \cos^{-1} \left(\frac{\text{Re} \langle u, v \rangle}{\|u\| \|v\|} \right) = \cos^{-1}(|S|)$$

$$s_0^* s_1$$

Decision boundary is at $\phi/2$

$$|s_0| * |s_1| \cos \phi$$

$$\sin^2 \phi/2 = \frac{1 - \cos \phi}{2} = \frac{1 - |S|}{2}$$

$$\sin \phi/2 = \sqrt{\frac{1 - |S|}{2}} = \frac{d}{2\sqrt{E_s}}$$

$$\text{Error rate} = Q\left(\frac{d}{2\sigma}\right) = Q\left(\frac{\sqrt{E_s (1 - |S|)}}{\sqrt{2N_0}}\right)$$

only dependent on $|S|$

→ Differential PSK



$$y = h s + w$$

↳ stays constant between consecutive symbols

- We can detect h from first symbol and use it for subsequent symbols

- But h will change with time \therefore

$$\begin{aligned} y[0] &= h s[0] + w[0] & \dots s[0] \text{ is known} \\ y[1] &= h s[1] + w[1] \end{aligned}$$

$$y[1] y[0]^* \propto s[1] s[0]^* \dots \text{Phase of } h \text{ does not matter.}$$

$$P_e = \frac{1}{2} e^{-E_s/N_0} |h|^2$$

$$y[i] = h s[i] + w[i] \quad \dots s[0] \text{ is known}$$

$i \in \{0, 1, \dots, N\}$

Try all 2^N possibilities of $s[i]$'s and make an ML estimate on them

→ ~~FF~~ Guess the $s[i]$'s which were most probably sent

- Using this, we estimate h .

Estimation of h is better for higher N

- Performance approaches coherent detection performance.