

# Like likelihood & Log of LR

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youva

Sign  $\left( \log \left( \frac{P(y_{l+1})}{P(y_{l-1})} \right) \right)$   
 ↗ LLR

$$y[0] = A + w[0]$$

$$y[k-1] = A + w[k-1]$$

$$\vec{h} = a \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \Rightarrow \text{Cov} = \sigma^2 I_{K \times K}$$

## Revision Lecture

### Link Budget Analysis

ex QAM - 256, 1 Mb/s  
 (8 bits)

$$\Rightarrow \text{Symbol Rate} = \frac{1}{8} \text{ Msym/s}$$

$$\Rightarrow \text{Minimum / ideal b/w} = \frac{1}{8} \text{ MHz}$$

$\Rightarrow a\%$  excess b/w  $\Rightarrow \times (1+a)$  factor

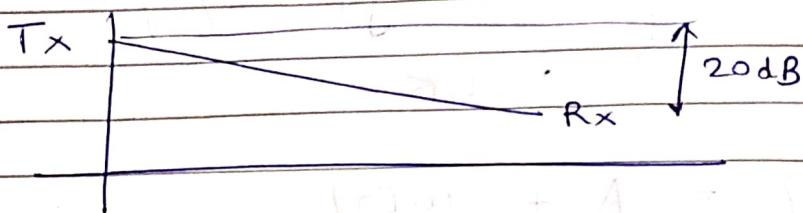
$$\frac{1}{8} \times (1+a) \rightarrow \text{Actual}$$

$\rightarrow$  If we say  $\text{BER} < 10^{-6} = Q(\dots)$

$E_s$  at receiver  
 $N_0$

Energy per symbol at receiver

→ Also take into account loss .....



$$L_{\text{ay}} = 20 \text{ dB} \quad R_x = 1 \text{ mW} \quad T_x = 100 \text{ mW}$$

• BER Gives us SNR measure .....

$\left( \frac{E_s}{N_0} \right)$  At Rx is a measure  
of SNR

How? Since it doesn't take into account

$$\frac{E_s}{N_0} = \frac{P \cdot T}{N_0} \approx \frac{P}{W N_0} \rightarrow \text{SNR}$$

→ Find  $E_s$ , Power req. at Rx =  $E_s$

⇒ Get Tx Power

~~Back to  
estimation  
problem~~

$$y_1 = A b_1 + w_1$$

$$y_k = A b_k + w_k$$

$$P(y | A=a) = f(y | A=a, b=+1) \times \frac{1}{2} + f(y | A=a, b=-1) \times \frac{1}{2}$$

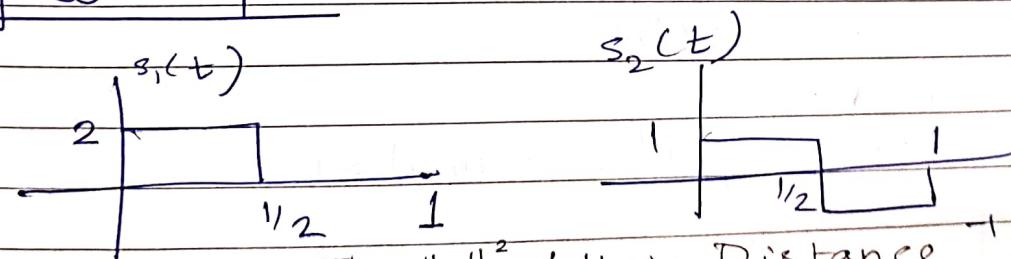
ML Gives us the formula

$$a = \frac{1}{k} \sum_{k=1}^K y^{(k)} \tanh\left(\frac{y_k a}{\sigma^2}\right)$$

→ If initial guess of sign of  $a$  is right we will converge to  $A$  (true value) else we will converge to  $-A$  (Reversed)

→ Reversed is still fine, can be detected using training (Known training Sequence)

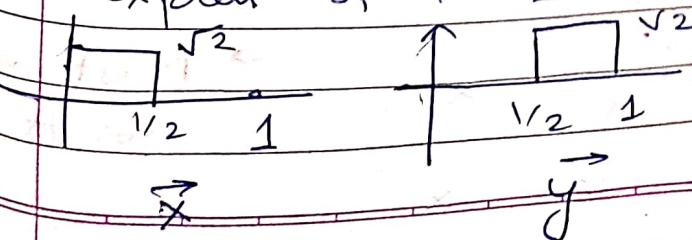
## Signal Space



The  $\| \cdot \|_2^2$  of their Distance

Find  $\int_0^1 (s_1(t) - s_2(t))^2 dt$  using vector

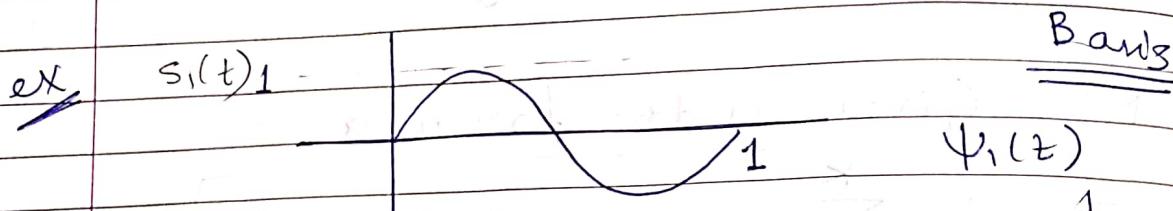
Express  $s_1$  &  $s_2$  in the basis



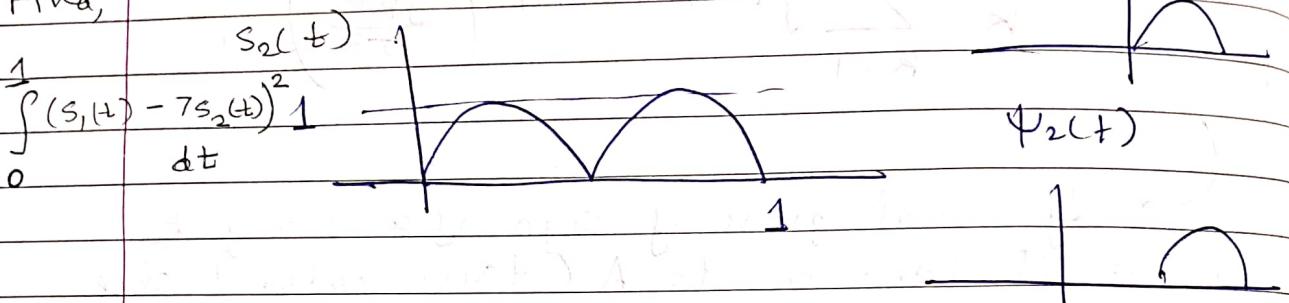
$$s_1 = \begin{bmatrix} \sqrt{2} \\ 0 \end{bmatrix}$$

$$s_2 = \begin{bmatrix} \sqrt{2} \\ 1 \end{bmatrix}$$

$$\left[ \begin{array}{c} \frac{\sqrt{2}-1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{array} \right] = \frac{\sqrt{2}+1}{2} - \frac{\sqrt{2}-1}{2} + \frac{1}{2} = 1$$



Find,



→ Whenever there are a finite number of signals we can do this.

→ Operations are Easier in Signal Space

~~4.3.10~~

~~$\hat{\theta}$~~

Find  $\hat{\theta}$

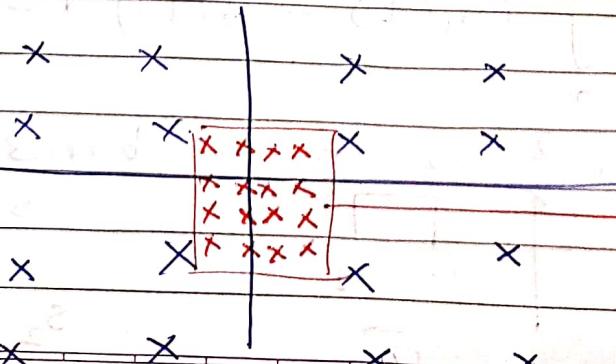
→ Recap of ML estimator

Consider QAM-16

$$y = As + w$$

$$A = 0.1$$

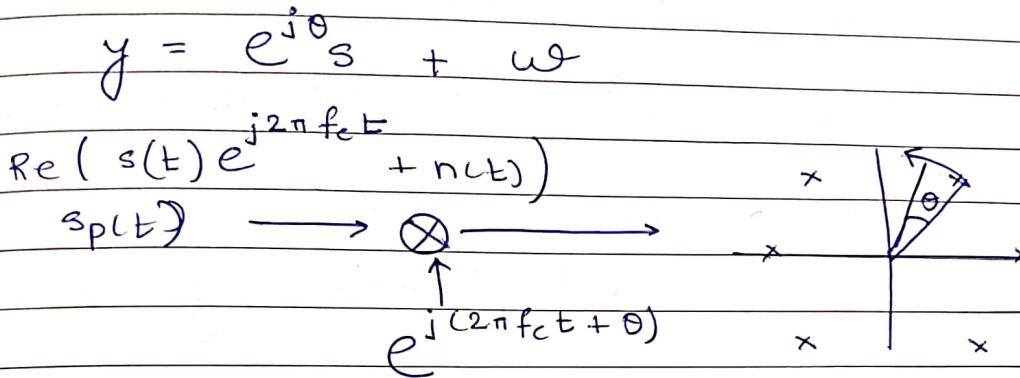
Must correct  
for A



... due  $(z^{-1})^{1/2}$  imaginary

ML estimate,  $y(\theta)$ , find  $\hat{\theta}$  that Maximizes  
 $\underset{\theta}{\operatorname{argmax}} \quad P(y(\theta) | \theta)$

### Synchronization Issue



→ How to find  $\theta$ ? Use ML estimator

$$\operatorname{Re}(\langle y, e^{j\theta} s \rangle) - \frac{\|s e^{j\theta}\|^2}{2}$$

$$\|s e^{j\theta}\| = \|s\| \quad (\text{like a Rotation})$$

$$= \operatorname{Re}(\langle e^{j\theta} s + \omega, e^{j\theta} s \rangle) - \frac{\|s\|^2}{2}$$

$$\begin{aligned} &= \cancel{\frac{\|s\|^2}{2}} + \underbrace{\operatorname{Re}(\langle \omega, s \rangle)}_{\theta \text{ dependent}} + \underbrace{\operatorname{Re}(\langle \omega, e^{j\theta} s \rangle)}_{\theta \text{ dependent}} \\ &\text{Ignore } \cancel{\frac{\|s\|^2}{2}} \end{aligned}$$

$$\underset{\theta}{\operatorname{argmax}} \operatorname{Re} \langle y, s e^{j\theta} \rangle = \underset{\theta}{\operatorname{argmax}} \operatorname{Re} \langle \omega, e^{j\theta} s \rangle$$

$$\langle y, s e^{j\theta} \rangle = z_c + j z_s$$

$$\min \tan^{-1} \left( \frac{z_s}{z_c} \right) = \underset{\theta}{\operatorname{argmax}} ?$$

ex

$$\theta = \frac{\pi}{2}, \text{ QPSK } 1, -1, i, -i$$

$$y[0] = e^{j\theta} + w[0]$$

$$y[1] = e^{j\theta} + w[1]$$

$$\operatorname{Re} \langle s e^{j\theta} + w, s e^{j\theta} \rangle$$

$$\zeta = \|s\|^2 + \operatorname{Re} \langle w, s e^{j\theta} \rangle$$

$\rightarrow$  Not very useful since ~~w~~ w is not known, y is

S.B.

$\rightarrow$  Sufficient Statistic

$$b^*[0] y[0] + b^*[1] y[1] + \dots + b^*[N-1] y[N-1]$$

### Delay estimation

$$y(t) = s(t - \tau) + w(t)$$

$\rightarrow$  Find  $\tau$  that maximizes ?

- When No delay,

$$\hat{\tau} = \operatorname{argmax}_i \int y(b) \phi_i(t) dt$$

Now take General Parameter  $\tau$

$$\operatorname{argmax}_{i, \tau} \int y(t) \phi_i(t - \tau) dt$$

Basis functions  
of

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$$\int y(t) \phi_i(t-\tau) dt$$

$$= \begin{bmatrix} g_1 \\ \vdots \\ g_k \end{bmatrix}$$

$$\int y(t) \phi_k(t-\tau) dt$$

find  $\tau$  which makes  
this closest to ~~s~~  $s_i$

~~S.B. 19~~

- Assume No frequency offset  $\rightarrow$  Get time offset  $\tau$

$$y(t) = A s(t-\tau) e^{j\theta} + n(t)$$

$\rightarrow$  We can't use signal space picture here.

$$y(t) = A s(t-\tau) e^{j\theta} + n(t)$$

unknowns :  $\underbrace{\tau, \theta, A}_{P(y|r)}$

$P(y|r)$

$\rightarrow$  Assume we know which symbol  $s(t)$  was sent.

Scaling + Phase Offset + Time offset

- To get time shift : Correlate with template

(OR) convolve with Matched filter

$A > 0$ , Real

$$L(y|r) = \exp \left\{ \frac{1}{\sigma^2} \left( Re \langle y, s_r \rangle - \frac{\|s_r\|^2}{2} \right) \right\}$$

$$\exp \left( \frac{1}{\sigma^2} \left( Re \langle y, s_r \rangle - \frac{\|s_r\|^2}{2} \right) \right)$$

(OR)

Simply Maximize

$$\max \left( Re \langle y, s_r \rangle - \frac{\|s_r\|^2}{2} \right)$$

$\downarrow$  is "Gamma"

$$\langle y, s_r \rangle$$

where,

$$s_r(t) = A' s'(t-\tau) e^{j\theta'}$$

$$\text{Ans} \quad \langle y, s \rangle ((A', \theta', \tau)) = (A, \theta, \tau)$$

~~Proof~~

$$\langle y, s_r \rangle = \int y(t) A' e^{-j\theta'} s^*(t-\tau) dt$$

$$A' e^{-j\theta'} \int y(t) s^*(t-\tau) dt$$

$$= A' e^{-j\theta'} (y * s_M)(\tau)$$

Optimize,

$$\operatorname{Re} \left( A' e^{-j\theta'} \int_{-\infty}^{\infty} y(t) s^*(t - \tau') dt \right) - A'^2 \|s\|^2$$

$$= J(\theta', \tau', A')$$

↳ negative cost function, Maximize this

Maximize  $J$  over  $\theta$

$$\operatorname{Re} \left( A' e^{-j\theta'} \int_{-\infty}^{\infty} y(t) s^*(t - \tau') dt \right)$$

→ Assume some  $\tau'$

$$\Rightarrow \theta' = \arg \left\{ \int y(t) s^*(t - \tau') dt \right\}$$

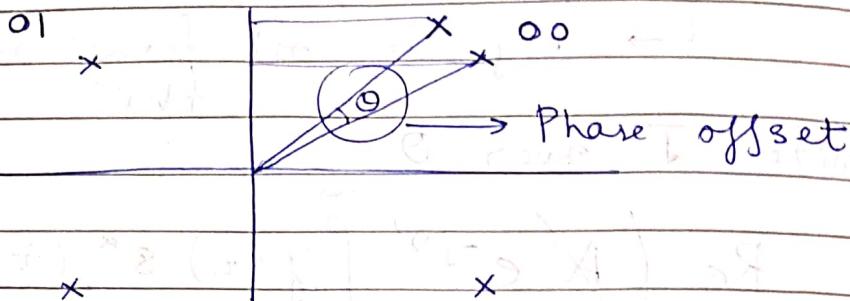
$$J(\tau, A) = A' \left| (y * s_{MF})(\tau') \right| - A'^2 \|s\|^2$$

$$\frac{\delta J}{\delta A'} = 0 = \left| (y * s_{MF})(\tau') \right| - A' \|s\|^2$$

$$A' = \frac{\left| (y * s_{MF})(\tau') \right|}{\|s\|^2}$$

→ Why does the likelihood function take the form

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



→ Effect of clock offsets ( $10 - 100 \text{ ppm}$ )

$$\frac{10}{10^6} \times 2.4 \times 10^9 \text{ Hz} = 24 \text{ kHz}$$

Must compensate for this amount of  $\Delta f$

→ How to determine frequency offset  
Unlike Amplitude/Phase 1 sample is not enough

Observe over the window  $[0, T_0]$

$$s(t) = \text{rect}\left(\frac{t - T_0/2}{T_0}\right)$$

$$\|s^2\| = \frac{\frac{1}{\sqrt{T_0}} \times \frac{1}{\sqrt{T_0}}}{T_0} = T_0$$

$$L(y|0) = \exp\left(\frac{1}{\sigma^2} (\operatorname{Re} < y, e^{j\theta} > - \frac{T_0}{2})\right)$$

$$\theta = \theta(t) : \theta(t)$$

$s(t)$  [Received waveform] =

$$\operatorname{rect}\left(\frac{t - T_0/2}{T_0}\right) e^{j\theta(t)}$$

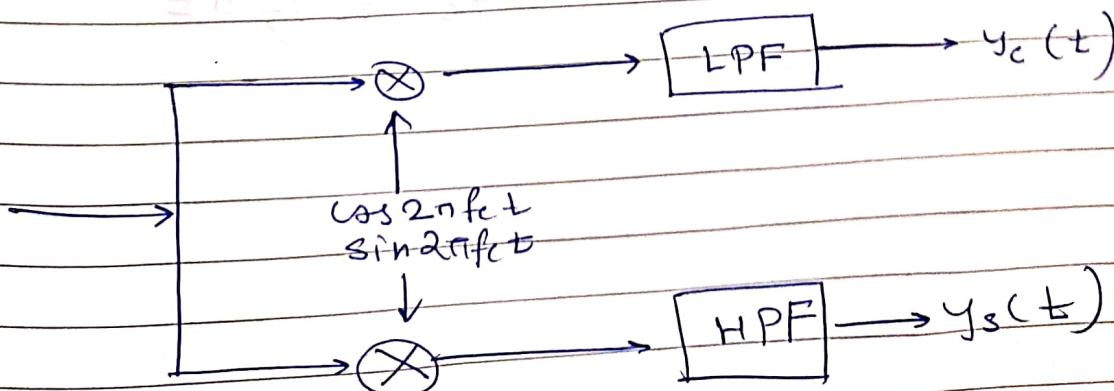
$$y(t) = s(t) + w(t) \quad 0 \leq t \leq T_0$$

$$\operatorname{Re} < y, s > - \frac{\|s\|^2}{2}$$

$$\operatorname{Re} < y, \operatorname{rect}\left(\frac{t - T_0/2}{T_0}\right) e^{j\theta(t)} > - \frac{T_0}{2}$$

$$J(\theta) = \operatorname{Re} < y, e^{j\theta} > = \int_0^{T_0} (y_c(t) \cos(\theta(t)) + y_s(t) \sin(\theta(t))) dt$$

$$= \int_0^{T_0} [y_c(t) \cos(\theta(t)) + y_s(t) \sin(\theta(t))] dt$$



$$\theta(t) = 2\pi(f_c - f_{c'})t$$

could be something else (crazy narration)

$$\operatorname{Re} \langle y, e^{j\theta} \rangle = \int_0^{T_0} (y_c(t) \cos(\theta(t)) + y_s(t) \sin(\theta(t))) dt$$

$$a \int_0^{T_0} -y_c(t) \sin(\theta(t)) + y_s(t) \cos(\theta(t)) dt$$

$\downarrow \frac{d}{d\theta}$

Use a VCO to perform up-downs in search for  $(\theta(t))$

ex Say,  $(f_c - f_{c'}) = 1 \text{ Hz}$

$$\int_0^{T_0} \cos(2\pi(f_c - f_{c'})t) dt$$

$$\int_0^{T_0} -\sin(2\pi(f_c - f_{c'})t) dt$$