Lecture 3->5

Detection in Gaussian noise:

Distinguish original transmitted signal yeter it was corrupted by a) sading b) Noise (additive, white beag)

Recall that we deal with discrete time complet model.

Scalar detection. (for now only noise).

WN NO. NO. 9+WER.

- Optimal Letter (ML)

take y, & compare $\Rightarrow P(x=x_A|y) \geq P(x=x_B|y) \Rightarrow \text{choose } x_A \text{ (else } t_B).$ $\Rightarrow P(x=x_A|y). P(y) \geq P(x=t_B|y) P(y)$

 $P(x_{A}, y) \ge P(x_{B}, y)$ $P(y|x_{A}) \cdot P(x_{A}) = P(y|x_{B}) \cdot P(x_{B})$

 $P(9/x_A).P(x_A) = P(9/x_B)$ $P(9/x_A) \stackrel{\times}{=} P(1/x_B).$

= 1 - (9- + E)/NO 14-xal = 14-xg) & detector chooses
nearest neighbor. 7 Optimal probability of error. (scalar depection). error when $(y-x_A)^2 > (y-x_B)^2$ & + > + A $(y=x_A+w)$. 29 (xB-+4) > xg -x4 y, xB+XA error event (is x>+A). (passes the 3 XATW > XBTX4 3) W7 XB-XA of error when w > (AKA - AKE)/ Per | x=x4 = P (y > +B+X4 | x4)= P (w > | WA - +B) = Q (| +4 - +B |)

P(3=xA | y) 3 P(x=xB | y) =

$$P(3=x_{A}|y) \stackrel{x_{A}}{=} P(x=x_{B}|y) \Rightarrow P(y|x_{A}) \geq P(y|x_{B})$$

$$\frac{1}{\sqrt{\pi N_{0}}} \cdot \frac{1}{e^{\|y-x_{A}\|^{2}}} N_{0}$$

$$\frac{1}{\sqrt{\pi N_{0}}} \cdot \frac{1}{e^{\|y-x_{A}\|^{2}}} N_{0}$$

$$\frac{1}{\sqrt{\pi N_{0}}} \cdot \frac{1}{e^{\|y-x_{A}\|^{2}}} N_{0}$$

$$\frac{1}{\sqrt{\pi N_{0}}} \cdot \frac{1}{e^{\|y-x_{B}\|^{2}}} N_{0}$$

P(9/4) =	N(+B
	711

When = +A & ever if | y-xa)2 > /y-+8/2 7 | xA+w-xA|2 |xA+w-+8|2 7 ||w|7 > ||w+xA-+8|| > www > (w+x) (wtx) + wtw > wtw + wty + (of) T + y 7 LRe {wtg} < -yyt. but all real. P(w4.y < - 1/21/2) = P((+A-+B) - w <-1/24-KB/2). Perr: P((xg-+)).w > ||xx-xx||). This just means that error when the. Note that $(\pm g - \pm a)^T w \sim N(0, \|\pm a - \pm g\|^2, N_0)$. In direction of sience etcecds the help-dist of the signal vector. ~ N(0, Ecitai) = N(0, Ecit No) $\frac{1}{2} \left\| \left(\frac{\left\| \left(\frac{1}{2} - \frac{1}{2} \right) \right\|^{2}}{\left\| \left(\frac{1}{2} - \frac{1}{2} \right) \right\|^{2}} \right\| = \left\| \frac{1}{2} \left\| \frac{1}{2} \left\| \frac{1}{2} - \frac{1}{2} \right\|^{2} \right\|^{2}$ - Only sunction of Euclidean distance.

$$\begin{array}{lll}
\underline{y} = \underline{x} + \underline{\omega} & \underline{\omega} \wedge N(0, \underline{\omega} \underline{I}) & \underline{z} = \underline{z} \wedge \alpha + \underline{B}. \\
\underline{het} & \underline{z} = \underline{z} \wedge \underline{l} & \underline{i} \underline{r} & \underline{z} = \underline{x} \wedge \alpha \\
\underline{het} & \underline{z} = \underline{z} \wedge \underline{l} & \underline{i} \underline{r} & \underline{z} = \underline{z} \wedge \alpha \\
\underline{-1} & \underline{i} \underline{r} & \underline{z} = \underline{z} \wedge \underline{B}
\end{array}$$

$$\begin{array}{ll}
\underline{z} = \underline{z} \wedge \alpha + \underline{B} \wedge \underline{z} \wedge \underline{z}$$

but
$$y' = y - \frac{x_A + x_B}{z} = \frac{x(x_A = x_B)}{z} + \frac{x_A + x_B}{z} + \frac{x_A + x_B}$$







7) can see that "transmitted vector" only in direction
$$\frac{V}{\|A - \lambda B\|}$$

- The projection of y' outo direction I to v coutains only noise, & that noise is I (3) independent) of the noise in direction of signal. 7 g=v".(y-+(xA+xB)) sufficient steating 7 9 = x"g' Equivalents: $v^{\#}. y' = \frac{(x_A - x_B)}{\|x_A - x_B\|} \cdot x \cdot (x_A - x_B) + v^{\#}. \omega = x \|x_A - x_B\| + n.$ $0. y' = \left[x(x_A - x_B)\right] + 0 \omega$ I retaine la since (4/21

7 this approach is called "matched site". Project y in directión of signal space. 7 now scalar desection problem 7 essective half distance | XA-XB/ 3 As before | Per = Q (1/ + A - +B/1). - Argument beneatized { * ... * ... * project y an (un , ...) * more highed option, & sugg set extistic. - Is X --- Xm is colinear i.e Xi = +i. h 7 pojed y auto h.

VectorPetection: complet

9= tru t= Eta e Ch. wn qu(s, I.No)

Signal direction $V = \frac{\pm A - \pm B}{\| \pm A - \pm B \|}$

9= 9- = (+4++8)

Since $t \in \mathbb{R}$ π can set simple suffer suff state $\operatorname{Re}\left\{\widetilde{g}\right\} = \left| t \left| \frac{1}{2} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t - \frac{1}{2} \left| t \right| + \operatorname{Re}\left\{\widetilde{w}\right\} \left| t - \frac{1}{2} \left| t$

Per = Q ($\frac{||\Delta x_4 - x_B||}{2\sqrt{\frac{n_b}{7}}}$) ess. halfdigance $\frac{||A_4 - x_B||}{2}$

benowlitation: if tx-vectors of form h.ti, tiet > h.y suffer h.ti +ier > Re{h.y'suffer h.y's suffer h.y's refer h.y's suffer h.ti +ier > Re{h.y's}.

Detection over sading channels - Main diff over AWEN case: much higher of (err). lecal y [m] = E he [m] x[m-e] Tw[m] (ondider slet fading.) w [m) ~ (n/o, No) (h [m) (N/o)) (y [m] = h[m] x [m] + w[ac] Consider BASK: +[m] = I a (index. in time)

random sading

Consider no knowledge of h[m] at ex.

Fler 3 - 1 Fler 3 - 1 Need for coding.

But
$$y[0] \nmid A$$
, $y[1] \nmid AA$ are indep (note $P(ha_{TW_1} | w_2)$)

 $= g(h_{TW_1}) \cdot V$.

 $= g(h_{TW_1}) \cdot V$.

$$= \log \left[\frac{-\frac{1}{3} (3) |x_{1}|}{\frac{1}{2} (3) |x_{2}|} \cdot \frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{-\frac{1}{3} (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{1}{3} (3) |x_{2}|} \right] = \log \left[\frac{(3) - \frac{1}{3} (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) - \frac{1}{3} (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) - \frac{1}{3} (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) - \frac{1}{3} (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) - \frac{1}{3} (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) - \frac{1}{3} (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) - \frac{1}{3} (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) - \frac{1}{3} (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} \right] = \log \left[\frac{(3) (3) |x_{2}|}{\frac{1}{2} (3) |x_{2}|} - \frac{(3) (3) |$$

 $1(9) = \frac{(3(0) - 3(1))}{(a^2 + N_0) \frac{N_0}{a^2}} \stackrel{\text{A}}{>} 0$

(|9 60) | > |9 (1)

Pr (err/xa) =
$$\rho(|951)^{\frac{1}{2}} > |950|^{\frac{1}{2}} |x_A|$$

To ealculate this, note

 $|9(0)|^2 \sim \exp[a^2 + N_0]$, $|951|^2 \sim \exp\{N_0\}$

Then recall $M_0 \in \mathcal{O}_F = \exp[distribution] \sim X$
 $E\{e^{\frac{1}{2}X}\} = \frac{1}{1+5}$

To get $err|_{XA} = \rho(|951|^2 > |950|^2 |x_A|) = (2 + \frac{a^2}{N_0})^2 = err|_{XA}$

Perrior Signaling of optimal decog Received SNR of signal per complex symbol of to period of the source of the of Piz averege

There =
$$\frac{1}{2+\frac{q^2}{4N_0}} = \frac{1}{2(1+\frac{q^2}{2N_0})} = \frac{1}{2(1+\frac{$$

(ourgane with detection with Awan channel.

Yen = xent twent + En = ta $\ddot{y} = \text{Re}\{y[m]\}$ $\Rightarrow \text{Rerr} = Q\left(\frac{\alpha}{N_0}\right) = Q\left(\frac{z\alpha^2}{N_0}\right) = Q\left(\frac{z\alpha^2}{N_0$ -> Huge difference between Amon performance & Fading with non-coherent detection. (x: Pe=10 3 Pez 1/2(17e) 7 Cuin = 2 Pe -1 = 5.10 91 = 50 dB

There as Awar = C = 10 4 - C. lose = -5losio

7 C = 10 4 a = 10 4 a = 3 - 4.

First do suff statistics: project auto dir of signal has also take real. (ASK WHY)

$$\frac{h^{2}}{h^{2}} = Re \left\{ \frac{h^{4}}{\|h\|} \cdot y \right\} = Re \left\{ \frac{h^{4}}{\|h\|} \cdot (h + h + w) \right\} = Re \left\{ \frac{h^{4}}{\|h\|} \cdot y \right\}$$
Then rates

 $7 \tilde{r} = ||A|| \cdot x + 2 \qquad z = Re\left\{\frac{h''}{\|h\|} \omega\right\} \quad z \sim N\left(\frac{n}{2}\right).$ (help distince: $|h| \cdot m$)

to (00)

Per = $Eh\left\{le(er/h)\right\} = Eh\left\{q\left(\frac{lh|.q}{\sqrt{No/2}}\right)\right\}$, $e=\frac{a^2}{No}$ $= Eh\left\{q\left(\frac{lh|^2a^2}{No/2}\right)\right\} = Eh\left\{q\left(\frac{lh|^2a^2}{No/2}\right)\right\}.$ $=\int a\left(\sqrt{|h|^2}\cdot ze\right) = \frac{1}{z}\left[1-\left(\frac{e}{e+1}\right)^{\frac{1}{z}}\left[-\left(\frac{1}{e}\right)^{\frac{1}{z}}+eylor\right] + eylor} + eylor$ (a Perr % 40) Note: only 3db diff from I (non-coh). I major cause of error. is deep sade": |h/le cc/.

Deep sale: h: /h/° cc1. Strong essocietión to event of error. 1) lh/2e>>1 7 Per + a(1h/2e) = elh/2e>>0 7 sade 7 error 3) 14/2 ecc1 3 7(err) -> 1 7 sade prob error. to P (deep sede) zë z Perr.

- (Intilitions)) CSIR U.S. No CSIR not trep reaso of publications
2) Problem caused by deep feels most often.

Question: what is better: send . I'm or send . I'm or send . I'm the or send . I'm the send . I'

Let us try to use more dimensions than BPSK.

Use apsq (Quadrature plasse shift kegins). $X[m] \in \left\{ a(1+J), a(1-J), a(-1,+J), a(-1,-J) \right\}.$

- bits in I & Q dimensions are indep. detected. (due to noise midep).

- (ousider first AWGN case y[m] = x[m] + w[m]. y[m] = x[m] + w[m]. $y = y_R + i \cdot y_T$ $y = y_R + i \cdot y_T$ $y = y_R + i \cdot y_T$

I have (due to noise index) two index BPSt detections. $y_R = x_R + w_R \qquad , \quad y_{\pm} = x_{\pm} + w_{\pm} \qquad \forall_R \in \{x_1 - x_2\} \quad \text{the } \{x_1 - x_2\}.$

7 Peror | arx 2 ler | BESK but double the rate. - Pue to more efficient packing.

(* same Per, double rate) Now take similar approach; same Per, same late, less power when using map directed 7 Compare QPSK (seen), with 4-8AM (sever Simensions). 9 (-1ti) (ti) q -3h -b b 2b q(-i-i) (1-i)q. Per (4PAM, AWON) = Q (heledis) = Q (The)= Q (The) Perr (4PAir, Awar) = 2.9 (\frac{2b^2}{No)} + 2. \frac{1}{2} Q (\frac{2b^2}{No)} = \frac{1}{2} Q (\frac{2b^2}{No)}

DAWEN Per = 9 (Per = 9 (Tel Wo) We compare, setting Per (RASK, AWGN) = Per (4 PAM, AWGN) $Q\left(\begin{bmatrix} \frac{2e^2}{N_0} \end{pmatrix} = Q\left(\begin{bmatrix} \frac{2}{N_0} \\ \frac{N_0}{N_0} \end{pmatrix} \right) = Q\left(\begin{bmatrix} \frac{2}{N_0} \\ \frac{N_0}{N_0} \\ \frac{N_0}{N_0} \end{bmatrix} \right)$ I same prop of error, same rate (2 lits/c.u). arsk: E{|x|'}= 2a', 4-1AM: 6'+1796786 - 30=597. 7 E[1x12] = 222 < 522 - E[1x1] - Ine to packing essiciones, we have 2,5 times less power 16= 10 logo 2.5 210.(0.4) =4 LB - Common problem: we are stack Pers in f. 7 Neel Vivesity: