2.5.1 Plansibility of MUD Claims 2.5.1 · Before we launch into Big Math, it is helful to establish some of the approaches with a very Simple example. Two signals in AWGN. 2 D 4, 42 orthonorm. basis A1/6 V2 A2 -V2 (4) V1, -V1 signal V2, -V2 vectors. A, A2 amplitudes o noise stelder. Receive $\Gamma = \pm \underline{v}_1 \pm \underline{v}_2 + \underline{n}$, get b, b, b, both ± 1 . · Single user bound. Correlate against the two pulse shapes (ie inner product with unit vectors u, uz). $y_1 = \pm A_1 \pm A_2 \cos \Theta + n_1$ E AL $y_2 = \pm A_1 \cos \Theta \pm A_2 + n_2$ If no interference Ai yi $P_{e_i} = O\left(\frac{A_i}{e_i}\right) = O\left(\sqrt{e_i}\right)$ $P_{e_2} = Q(\frac{A_2}{F}) = Q(\sqrt{\delta_2})$ $\int \sigma r convencence$ $X = \frac{A^2}{G^2} = 2 \frac{E_b}{N_0}$ Achievable only if $\theta = \pm \pi/2$.

Define $S/I = \delta_1/\delta_2 = A_1^2/A_2^2$

2.5.2
· Method 1. Ignore the Mith, pletchen it is the
white noise. Correlate ægamst pulse shapes.
A.>O Note cos O = p the pulse cross-correlation coefficient
A,-Azcoso A,+Azcoso random
$P_{e_1} = \pm Q\left(\frac{A_1 - A_2 \rho}{\sigma}\right) + \pm Q\left(\frac{A_1 + A_2 \rho}{\sigma}\right)$
$P_{e_2} = \frac{1}{2} \mathcal{Q} \left(\frac{A_2 - A_1 \rho}{\sigma} \right) + \frac{1}{2} \mathcal{Q} \left(\frac{A_2 + A_1 \rho}{\sigma} \right)$
If has and place 1, this is disastrous for both users
If has or deel, and place 1, it is disastrous for
the weak user and an annoyance to the strong one.
If 1p1<<1, for example by spreading code, then
close to single user - but additional users cause
the MAI to build
This is the conventional CDMA detector.

'n

2.53 Method 2. Zero forcing. Null out the other users by projecting I onto a hyperplane orthogonal to all the other users. θ ¥2 - hyperplane - Projection of I = ± Y ± Y2 + 1 onto the hyperplane does this: $\widetilde{\mathcal{U}}_{\mathbf{Z}} \cdot \underline{\Gamma} = \widetilde{\mathcal{U}}_{\mathbf{Z}} \cdot (\pm \underline{\mathcal{U}}_{\mathbf{Z}} \pm \underline{\mathcal{U}}_{\mathbf{Z}} + \underline{n}) = \pm A_{\mathbf{L}} \cos(\pi/2 - \theta) + n$ $=\pm A, \sqrt{1-\cos^2\theta} + n = \pm A, \sqrt{1-\rho^2} + n$ The unwanted user is eliminated, and $P_{e_1} = O\left(\frac{A_1 \sqrt{1-p^2}}{\sigma}\right) = O\left(\sqrt{X_1 (1-p^2)}\right), P_{e_2} = O\left(\sqrt{X_2 (1-p^2)}\right)$ irrespective of other user power. - By retaining only the component of 11 that is orthogonal to vz (& vice versa), we have weakened the effective signal - "noise enhancement". If pl≈1, it doesn't work, even for the strong user; a big sacrifice if $\lambda \ll 1$, Insensitive to near-far disperities

2.5.4 - In 2-space we can't add another signal, so dimensionality of measurements must at least equal the number of signals. Again, if spreading makes places, then not much cost - if only one user. Every additional user and orthogonality constraint chips away at desired user. . Method 3. MMSE. Why should the strong user sacrifice power to null a weak user. Why not simply reduce that user to the noise love ?? This allows some of white we way way the other user to be present without certhing own power as much, - The optimum w depends on the noise of so we must know SNR= 8, 82, unlike ZF. As &, &2 -> 00, the only disturbance is MAI, and MMSE solution -> ZF. At lower SNR, is much better than ZF, unless 101<<1. Insensitive to near-far disparity. - Also fails both users if 10121.

Method 4, Max Likelihood, joint detection. Detect
 both user data at once by selecting the point
 closest to r.
 (+,+) * Note



* Note This method works even with p=1 (same spreading code) provided A, 7A2 (or a mutual phace shift. if complex).

Consider user 1 error rate, with +v, sent. Then Pe, = 12 Pr[user1error]++sent] + 12 Pr[user1error]+-sent]

$$\leq \frac{1}{2} \left(\operatorname{Pr} \left[+ + \rightarrow - + \right] + \operatorname{Pr} \left[+ + \rightarrow - - \right] \right)$$

$$+ \frac{1}{2} \left(\operatorname{Pr} \left[+ - \rightarrow - - \right] + \operatorname{Pr} \left[+ - \rightarrow - + \right] \right)$$

$$= \frac{1}{2} \left(\operatorname{O} \left(\frac{A_{1}}{\sigma} \right) + \operatorname{O} \left(\frac{\sqrt{A_{1}^{2} + A_{2}^{2} + 2A_{1}A_{2}\rho}}{\sigma} \right) \right)$$

$$+ \frac{1}{2} \left(\operatorname{O} \left(\frac{A_{1}}{\sigma} \right) + \operatorname{O} \left(\frac{\sqrt{A_{1}^{2} + A_{2}^{2} - 2A_{1}A_{2}\rho}}{\sigma} \right) \right)$$

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$$= \frac{1}{2} \left(\operatorname{O} \left(\frac{A_{1}}{\sigma} \right) + \operatorname{O} \left(\sqrt{A_{1}^{2} + A_{2}^{2} - 2A_{1}A_{2}\rho} \right) \right)$$

$$= \frac{1}{2} \left(\operatorname{O} \left(\frac{A_{1}}{\sigma} \right) \right)$$

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$$= \frac{1}{2} \left(\operatorname{O} \left(\frac{A_{1}}{\sigma} \right) + \operatorname{O} \left(\sqrt{A_{1}^{2} + A_{2}^{2} - 2A_{1}A_{2}\rho} \right) \right)$$

$$= \frac{1}{2} \left(\operatorname{O} \left(\frac{A_{$$

2.5,5



$$\rho = 0.6$$
 $\lambda = 2$

Notes:

* A large value of ρ .

* ZF worse than conv. at low SNR, because noise enhancement.

* ML union bound

diverges at low SNR. * ML bound is too loose for the strong user - a problem with standard union bound.

User 1 performance, various detectors



User 2 performance, various detectors

Notes:

* Conv. detector crippled by MAI.

* ML benefits this weaker user, effectively eliminating MAI.

. The multiser detectors are readily understood from simple sketches. They do offer significant improvement over conventional detection (for the cases examined).

25.7

Here are a few more questions I'd like you to think about if you ever have a moment.

1. We determined that Rake processing produces sufficient statistics? Does this mean that the conventional Rake receiver is optimum? If so, in what sense and for what environment?

2. In our toy system of Section 2.5, the signal space had dimensionality 2 for convenience in sketching. If it were 3 (e.g., because of 3 chips/symbol or because of 3 antennas), then:

(a) How many users can you support with zero forcing? Is this a hard limit?

(b) How many users can you support with MMSE? Is this a hart limit or a practical limit?

(c) Is there a limit, soft or hard, on the number of users you can support with ML detection?

(d) If the signals are linearly dependent, e.g., v3 is expressible as a linear combination of the other two, can ZF separate them? Can MMSE? Can ML?

(e) In a special case of linearly dependent signals, where two of them are colinear and the third is linearly independent of the other two, what can zero forcing salvage?