

6.3 ML-MUD for CDMA: Problem Setup

- In this section, we address the full-blown CDMA problem, formulating it for ML.
 - no numerical results (yet)
 - demonstrates structure of detector with channel estimation error
 - demonstrates a causal form of the ML detector
 - sets it up for performance analysis by quadratic forms
 - suggests a more realistic solution method.

The work is incomplete, unpublished.

6.3.1 The MLSD Receiver for Perfect CSI

- As a warm-up, we'll derive the MLSD for perfect CSI. Recall that

$$\underline{y} = \underline{R} \underline{C} \underline{A} \underline{b} + \underline{v}$$

for multiple antennas, paths, users and symbol times.

- Conditional likelihood:

$$P_{y|b,c}(y | \underline{b}, \underline{c}) = \frac{1}{(2\pi)^{KLMN} |\underline{R}_v|} \exp \left[-\frac{1}{2} (\underline{y} - \underline{R} \underline{C} \underline{A} \underline{b})^+ \underline{R}_v^{-1} (\underline{y} - \underline{R} \underline{C} \underline{A} \underline{b}) \right]$$

Note $\underline{R}_v = N_0 \underline{R}$, independent of \underline{b}

- Log, discard hypothesis independent terms, factors

$$\hat{\underline{b}} = \underset{\underline{b}}{\operatorname{argmin}} (\underline{y} - \underline{RCA}\underline{b})^{\dagger} \underline{R}^{-1} (\underline{y} - \underline{RCA}\underline{b})$$

- Expand exponent, drop hypothesis independent stuff and we have the MLSD as a metric maximization:

$$\hat{\underline{b}} = \underset{\underline{b}}{\operatorname{argmax}} \underbrace{\operatorname{Re}[\underline{b}^{\dagger} \underline{A} \underline{C}^{\dagger} \underline{y}]}_{\text{correlation term}} - \underbrace{\frac{1}{2N_0} \underline{b}^{\dagger} \underline{A} \underline{C}^{\dagger} \underline{R} \underline{C} \underline{A} \underline{b}}_{\text{energy term}}$$

• Note incorporation of Rake/MRC through $\underline{C}^{\dagger} \underline{y}$ in the correlation term, which can be written $\operatorname{Re}[\underline{b}^{\dagger} \underline{A} \underline{\zeta}]$

• The energy term is important. If we drop it, then

$$\underset{\underline{b}}{\operatorname{argmax}} \operatorname{Re}[\underline{b}^{\dagger} \underline{A} \underline{\zeta}]$$

could be achieved by $\hat{b}_k(n) = \operatorname{sgn}[\zeta_k(n)]$, that is,

the conventional detector. It's the energy term

that makes it MLSD. Unfortunately, it is

quadratic in \underline{b} , so it seems that we must receive all of the packet to process any of it — no

causal form here ...

... or is there?

- A trick, originally formulated by Ungerboeck [Unge 74] gives an optimal causal form, even without whitening the noise. Here's how:

- Denote $\underline{C}^* \underline{R} \underline{C} = H$, so energy term is $(A\underline{b})^T H A\underline{b}$

- Decompose $H = \begin{matrix} \Delta^0 & & \\ H_e & H_d & \\ & & \nabla \end{matrix}$ $H_u = H_e^T$

- For convenience, denote $A\underline{b} = \underline{b}'$. Then energy term is

$$\begin{aligned} \underline{b}'^T H \underline{b}' &= \underline{b}'^T H_d \underline{b}' + \sum_{i=1}^{NK} b_i' \left(\sum_{j=1}^{i-1} [H_e]_{ij} b_j' \right) + \sum_{j=1}^{NK} b_j' \left(\sum_{i=1}^{j-1} [H_u]_{ij} b_i'^* \right) \\ &= \sum_{i=1}^{NK} \left(H_{ii} |b_i'|^2 + 2 \operatorname{Re} \left[b_i' \sum_{j=1}^{i-1} H_{ij}^* b_j'^* \right] \right) \end{aligned}$$

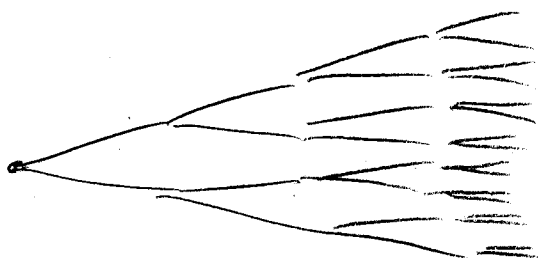
and this is causal.

- Combine with the correlation term, and we have the metric up to bit I

$$\Lambda(I) = \sum_{i=1}^I \left(\underbrace{\operatorname{Re} [b_i'^* \xi_i] + H_{ii} |b_i'|^2 + 2 \operatorname{Re} \left[b_i' \sum_{j=1}^{i-1} H_{ij}^* b_j'^* \right]}_{\lambda_i} \right)$$

a branch metric
only the third term has history
and it's limited because H is tri-diagonal

- Now it's all set up for MLSD by
 - Viterbi. Hmmm - the state set is finite (because $H = C^*RC$ is block tridiagonal) but it's large.
 - Viterbi with per survivor - well, it's better than DF, but by how much? No results.
 - tree search? We'll see shortly.



eventually folds back on itself to make a messy trellis.

- or even simple DF (not really ML, but if you're desperate...?)

Exercise:

- Cast the whitened receiver of Section 5.3 into the format of sequence metric and branch metric - show how a tree search algorithm will produce MLSD.

6.3.2 The MLSD Receiver for Imperfect CSI

- Model: the true channel gains and the estimates are related by

$$C = \hat{C} + E$$

Where the $KL MN \times KN$ matrix E consists of mutually uncorrelated Gaussian estimation errors that are also uncorrelated with \hat{C} .

- Then we write the measurements as

$$\begin{aligned} \underline{y} &= \underline{R} C \underline{A} \underline{b} + \underline{v} \\ &= \underline{R} \hat{C} \underline{A} \underline{b} + \underline{R} E \underline{A} \underline{b} + \underline{v}. \end{aligned}$$

The conditional statistics are

$$\mu_{y|b} = \underline{R} \hat{C} \underline{A} \underline{b}$$

$$R_{y|b} = \frac{1}{2} \mathbb{E}_{\underline{v}} \left[(\underline{R} E \underline{A} \underline{b} + \underline{v})(\underline{R} E \underline{A} \underline{b} + \underline{v})^T \right]$$

$$= \frac{1}{2} \mathbb{E} \left[\underbrace{\underline{R} E \underline{A} \underline{b} \underline{b}^T \underline{A}^T E^T \underline{R}^T}_H + N_0 \underline{R} \right]$$

$$= \sigma_e^2 \underline{R} \text{tr}[H] \underline{I} \underline{R}^T + N_0 \underline{R}$$

$$= \sigma_e^2 LMN \sum_{k=1}^K |A_k|^2 \underline{R}^2 + N_0 \underline{R} \quad \text{independent of } \underline{b} !$$

$$= R_y$$

• The MLSD and its metric:

$$P_{y|b, \hat{c}} = \frac{1}{2\pi} \frac{\exp\left(-\frac{1}{2N_0} (y - \mu_{y|b})^T R_y^{-1} (y - \mu_{y|b})\right)}{|R_y|}$$

Take log:

$$\arg \min_b (y - \mu_{y|b})^T R_y^{-1} (y - \mu_{y|b})$$

or

$$\arg \max_b \underbrace{2 \operatorname{Re}[\mu_{y|b}^T R_y^{-1} y] - \mu_{y|b}^T R_y^{-1} \mu_{y|b}}_{\Lambda(b, \hat{c})}$$

• Substitute for conditional mean:

$$\Lambda(b, \hat{c}) = \operatorname{Re}[\underline{b}^T \underline{A}^T \hat{\underline{c}}^T \underline{R} R_y^{-1} y] - \frac{1}{2} \underline{b}^T \underline{A}^T \hat{\underline{c}}^T \underline{R} R_y^{-1} \underline{R} \hat{\underline{c}} \underline{A} \underline{b}$$

Same form as before — another recursive formulation. An interesting receiver.

We can take it further through application of matrix inversion lemma to R_y — but not here!