

Ordered Statistics Decoding for Linear Block Codes over Intersymbol Interference Channels

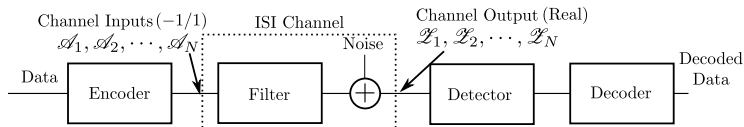
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University of Hawaii, Manoa

Joint work with :
Fabian Lim
Marc Fossorier



Problem Statement



Intersymbol Interference (ISI) channel with memory $I \geq 0$

Input-output relationship:

$$\mathcal{Z}_t = \sum_{i=0}^I \mathcal{A}_{t-i} h_i + \mathcal{W}_t \text{ for } 1 \leq t \leq N,$$

where $\mathcal{W}_t$ is i.i.d Gaussian noise.

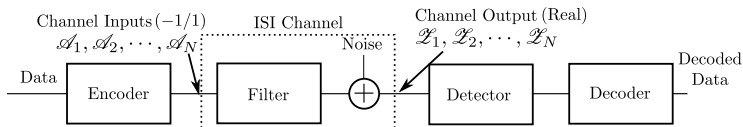
- Important model for e.g. *recording* channels.

- Consider *linear block codes* - important class of codes.

- Includes *BCH, low-density parity check, turbo*, etc.
- Includes codes used in standards, e.g. RS [255, 239, 1]



Problem Statement



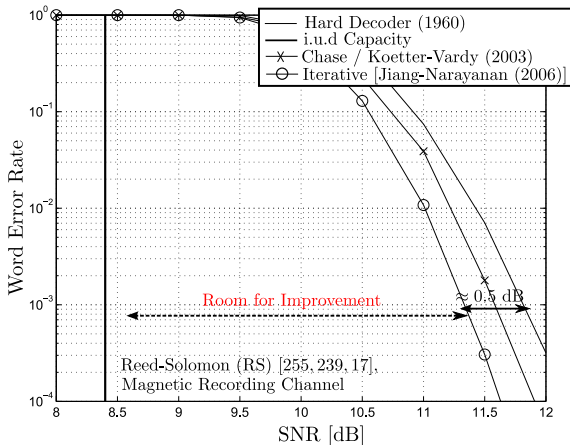
Problem:

Decode *linear block codes* over *ISI* channels.



Is This Problem Hard?

- Yes, *very*.
- Even *state-of-the-art* algorithms do not approach capacity.

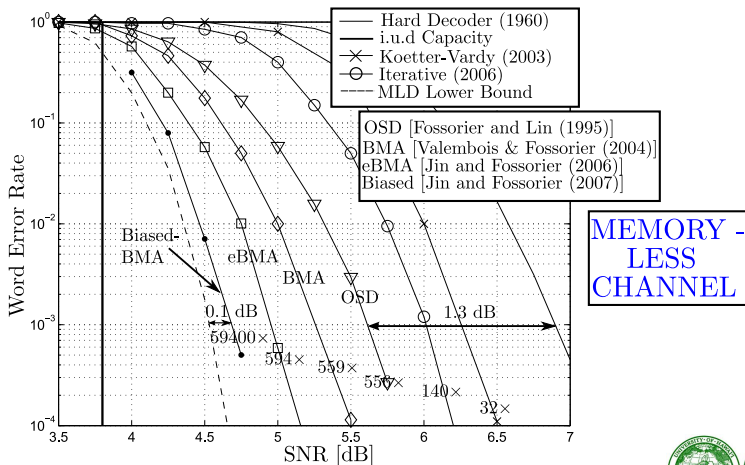


ISI
CHANNEL



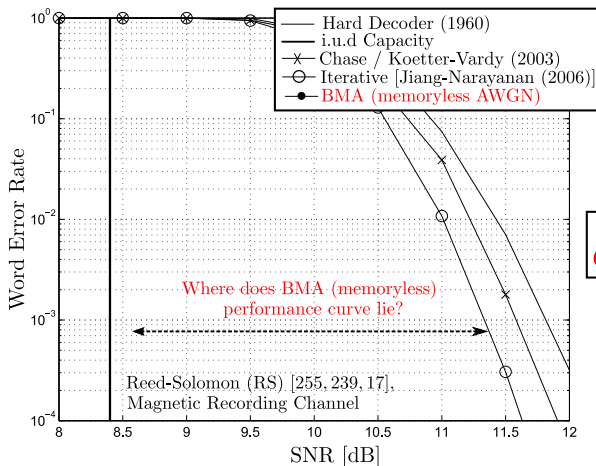
Ordered Statistics Decoding : Powerful Technique

RS [255, 239, 17] binary image, *memoryless* AWGN.



How Does OSD Perform on ISI Channels?

- Unimpressive gains observed (compared to memoryless).

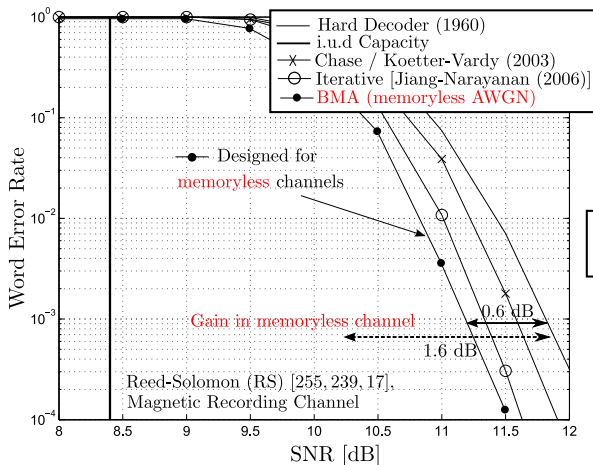


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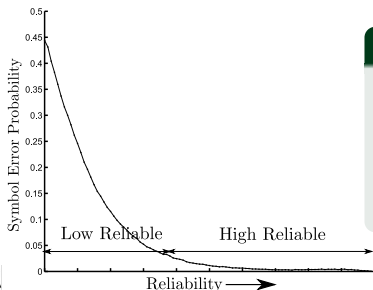
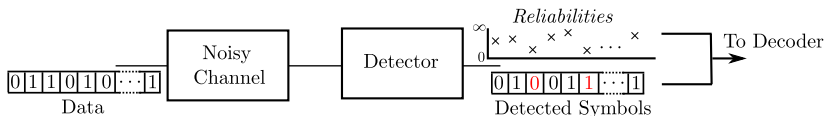


ISI
CHANNEL



What Went Wrong?

- OSD: *reliability-based* decoding algorithm.



Intuition behind OSD

High reliable symbols are:

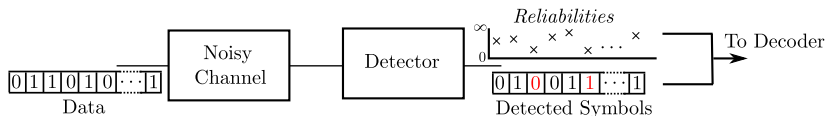
- Mostly received correctly.
- Able *decode/determine* low reliable ones.

- Simply *hypothesize high* reliable symbol values.



What Went Wrong?

- OSD: *reliability-based* decoding algorithm.



Issue: Prev. Method not suited for Channel with Memory

- OSD originally designed for a *memoryless* channel.
 - Symbol errors considered *individually*.
 - *Inefficient* in ISI channels - does not exploit memory.

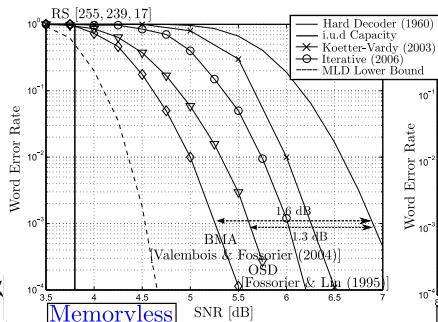
Past Extensions for other channels

- Rayleigh fading [Fossorier & Lin (1997)]
- Wireless channel [Chen & Hwang (2000)]

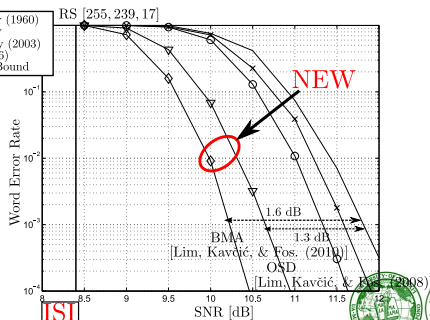


Main Challenges / Overview

- Decoder designs:
 - Ordered statistics decoding (OSD).
 - Box-and-Match algorithm (BMA).
- Complexity-saving measures.
- Analysis: Reliability distributions / Symbol Error Prob.



Memoryless



ISI



Main Results (ISI Channels)

- Ordered statistics decoding (OSD).
 - Works for any *linear block code*.
 - Develop hypothesis method - *error events*.
 - Efficient implementations.
- Box-and-Match algorithm (BMA).
 - OSD enhancement
 - Provide (*increased*) error correction guarantee.
 - Marginal complexity increment.
- Codeword optimality tests
 - Complexity saving techniques.
- Closed-form expressions for reliability distributions
 - Existing results not useful for our purposes.
- Approx. for ordered symbol error probabilities.
 - Quantifying the intuition behind OSD decoders.



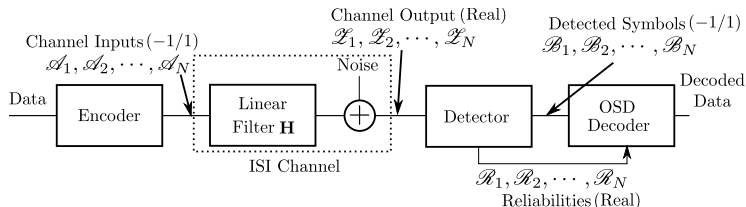
Outline

- 1 Introduction
- 2 Main Results
 - Detection
 - Ordered Statistics Decoding (OSD)
 - Box-and-Match (BMA)
 - Optimality Testing
 - Analysis
- 3 Conclusion

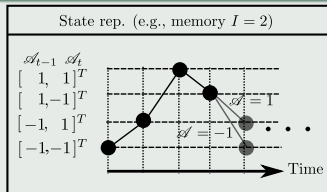


Detection in ISI Channels

• System overview.



ISI Channel Memory $I \geq 0$



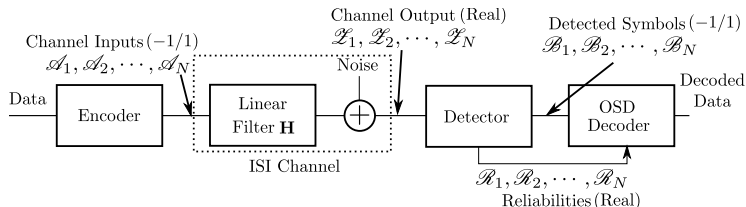
• Recall the following dependence

$$\mathcal{Z}_t = \mathcal{Z}_t(\mathcal{A}_{t-I}, \dots, \mathcal{A}_t).$$



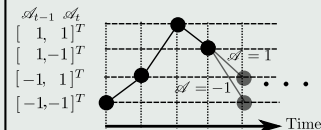
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ISI Channel Memory $I \geq 0$

State rep. (e.g., memory $I = 2$)



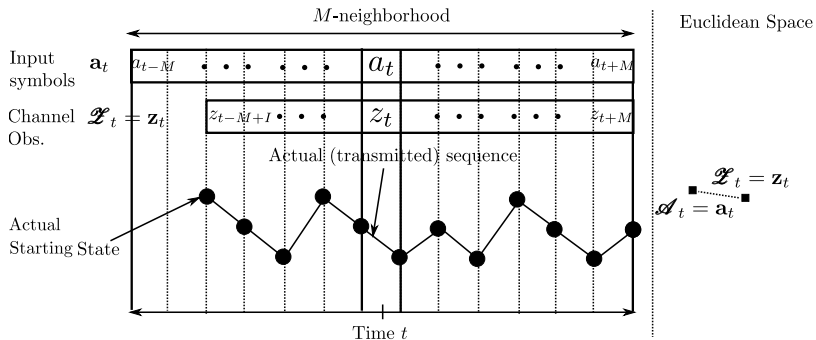
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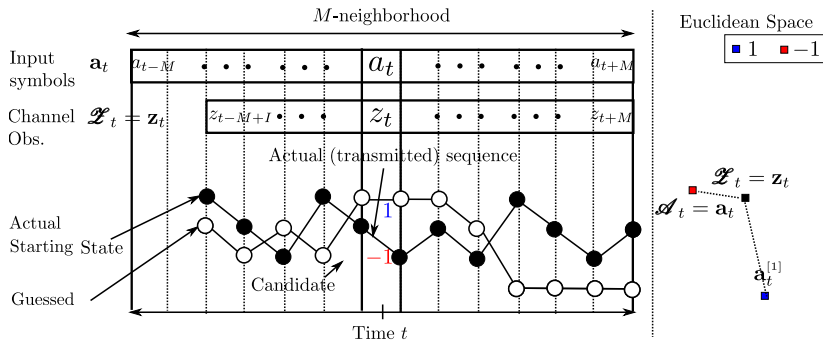
M -truncated max-log-map (MLM)

- Computes *symbol decision* b_t and *reliability* r_t .



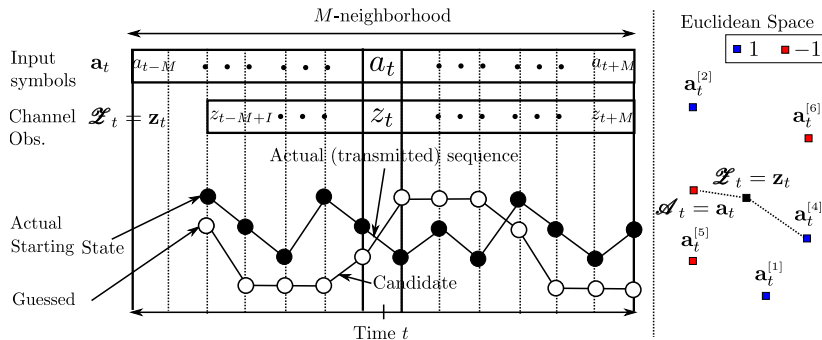
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Symbol Decision $b_1, b_2, \dots, b_N$

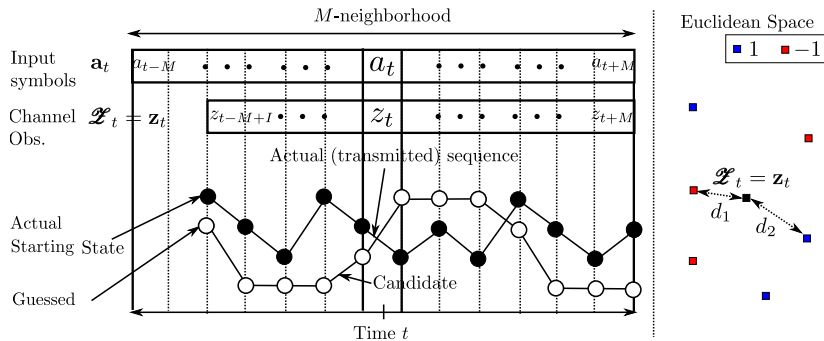
From *candidates* $\mathbf{a}_t = [a_{t-M}, \dots, a_{t+M}]^T$, choose

$$b_t = b_t(\mathbf{z}_t) \triangleq \arg \min_{a \in \{1, -1\}} \left\{ \min_{\mathbf{a}_t: a_t = a} |\mathbf{z}_t - \mathbf{H}\mathbf{a}_t|^2 \right\}$$



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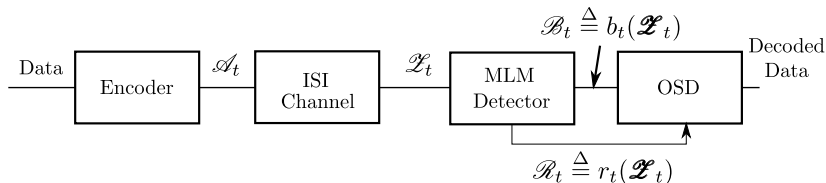
Reliability Sequence $r_1, r_2, \dots, r_N$

Measures a detection “margin”

$$r_t = r_t(\mathbf{z}_t) \triangleq \min_{\mathbf{a}: a_t \neq b_t(\mathbf{z}_t)} |\mathbf{z}_t - \mathbf{H}\mathbf{a}_t|^2 - \min_{\mathbf{a}: a_t = b_t(\mathbf{z}_t)} |\mathbf{z}_t - \mathbf{H}\mathbf{a}_t|^2 \geq 0$$

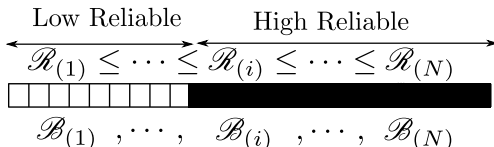


Order Statistics of Reliabilities



Order Statistics

- Ordered reliabilities: $\mathcal{R}_{(1)} \leq \mathcal{R}_{(2)} \leq \dots \leq \mathcal{R}_{(N)}$
- Ordered symbol decisions: $\mathcal{B}_{(1)}, \mathcal{B}_{(2)}, \dots, \mathcal{B}_{(N)}$
- OSD relies on high reliable decisions.
- High reliable decisions $\mathcal{B}_{(i)}$ recv. correctly most of the time.



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- Box-and-Match (BMA)
- Optimality Testing
- Analysis

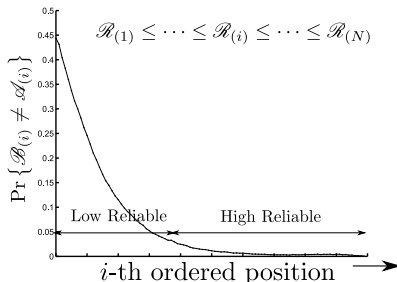
3 Conclusion



Ordered Statistics Decoding (OSD)

Result : OSD for ISI channels

F. Lim, A. Kavčić, and M. Fossorier, "OSD of linear block codes over ISI channels," *IEEE Trans. on Magn.*, vol. 44, no. 11, pp. 3765–3768, Nov. 2008.



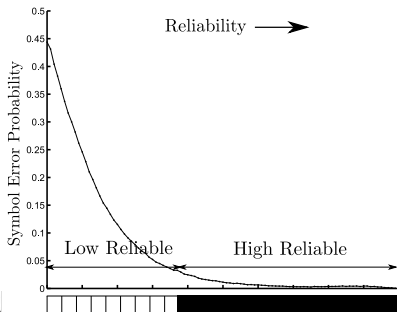
- Hypothesize errors over *high* reliable positions.
- Erasure decode *low* reliable positions.
- **Good** coding gains: **efficient** error hypothesis method.



Ordered Statistics Decoding (OSD)

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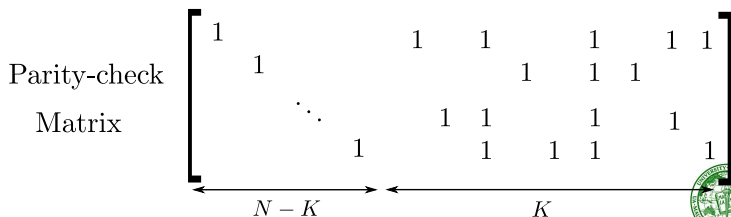
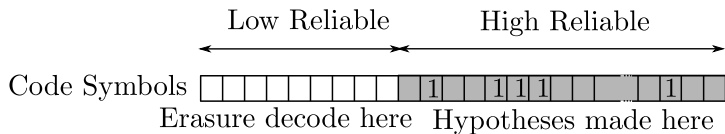


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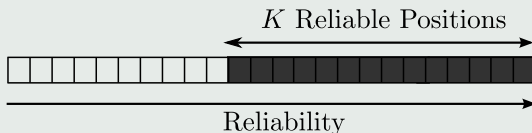
Erasure Decoding In a Nutshell

- (High) reliable symbols are *hypothesized* (guessed).
- Low reliable symbols decoded all at once (by erasing).
- $[N, K, d_{\min}]$ linear block codes $\mathcal{C}$.



OSD is a List Decoding Technique

Strategy of OSD for decoding $[N, K, d_{\min}]$ block codes $\mathcal{C}$



- 1) Select a set of K *reliable* positions.
- 2) Construct a codeword list $\mathcal{L}_{\text{OSD}} \subset \mathcal{C}$.
- 3) Each $\mathbf{c} \in \mathcal{L}_{\text{OSD}}$ obtained by
 - Hypothesize possible errors over reliable positions.
 - Erasing low reliability positions.
- 3) Finally choose best codeword in the list $\mathcal{L}_{\text{OSD}}$

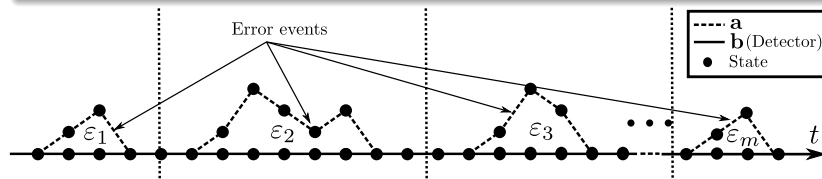
$$\mathbf{c}_{\text{OSD}} = \arg \min_{\mathbf{c} \in \mathcal{L}_{\text{OSD}}} |\mathbf{z} - \mathbf{H}\mathbf{c}|^2$$



Hypothesis Technique for ISI Channels

Hypotheses made according to *bursty* error mechanism

- Strategy: Consider set of “reliable” *error events*.
- Related to the memory state evolution in channel.



Definition: Err. pair $p = (\mathbf{a}, \mathbf{b})$, error event decomposition

- Err. pair $p = (\mathbf{a}, \mathbf{b})$ decomposed into error events

$$p = (\mathbf{a}, \mathbf{b}) = \varepsilon_1 \circ \varepsilon_2 \circ \cdots \circ \varepsilon_m$$

Definition: The set of reliable error events $\mathcal{E}$

- Reliable error event $\varepsilon \in \mathcal{E}$ flips *at least* one reliable position.

Summary of the OSD Algorithm

OSD Parameter $\mathcal{O}$

- Largest num. of events in $\mathcal{E}$, *collectively hypothesized*.
- Shorthand notation for list size $|\mathcal{L}_{\text{OSD}}| = \sum_{i=0}^{\mathcal{O}} \binom{|\mathcal{E}|}{i}$.



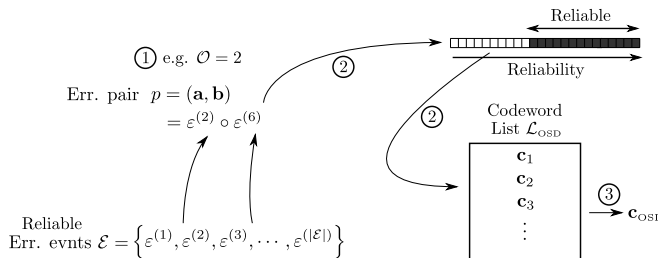
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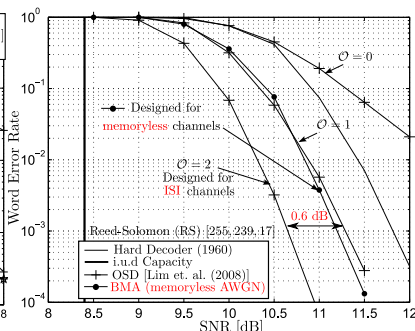
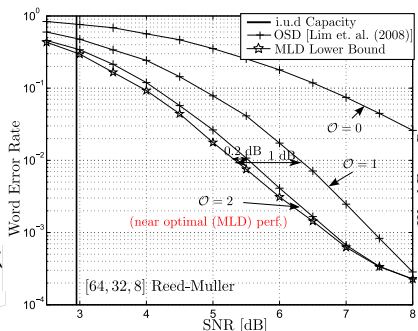
OSD decoding includes the following steps:

- 1) Pick a subset (of size *at most* $\mathcal{O}$) from $\mathcal{E} \rightarrow$ error pair p .
- 2) Form error hypothesis pair p and erasure decode.
- 3) Choose best codeword $\mathbf{c}_{\text{OSD}}$ in $\mathcal{L}_{\text{OSD}}$.



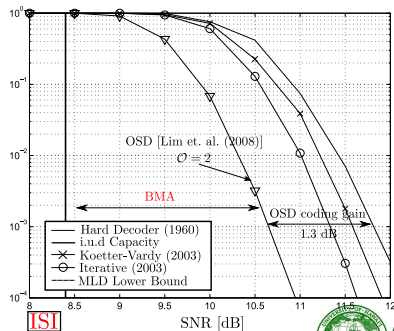
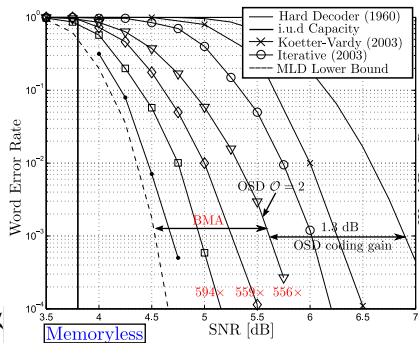
Closing Points : Ordered Statistics Decoding (OSD)

- OSD generalized for ISI channels.
- Consider reliable events $\varepsilon \in \mathcal{E}$ to exploit memory.
 - The set $\mathcal{E}$ is *very* complex (compared to memoryless case).
 - Handled using a generalized *Viterbi* alg. and *Battail's* alg.
- Our approach obtains much more impressive gains, than when *directly* applying past techniques.



Is There a Need for Further Improvement?

- For Reed-Muller [64, 32, 8], OSD is near-optimal.
- However, for RS [255, 239, 17], further improvement is desired.
- Consider *box-and-match (BMA)* enhancements.



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- **Box-and-Match (BMA)**
- Optimality Testing
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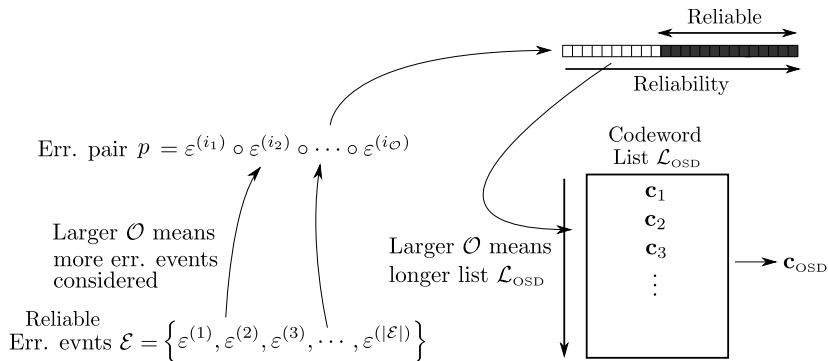
3 Conclusion



OSD: A naive strategy?

Is it effective to keep increasing $\mathcal{O}$?

- Larger $\mathcal{O}$ gives better error correction, however,
- $|\mathcal{E}|$ is a big number.



OSD: A naive strategy?

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- $|\mathcal{E}|$ is a big number.

The *box-and-match algorithm* (BMA) is an elegant idea to

- process error events more efficiently, using
- similar complexity as OSD.

Result : Box-and-Match (BMA) for ISI channels

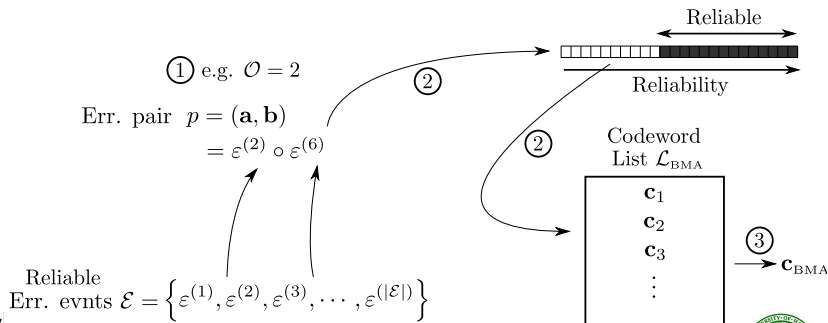
F. Lim, A. Kavčić, and M. Fossorier, “List decoding techniques for ISI channels using ordered statistics,” in IEEE JSAC, Vol. 28, No. 2, pp 241 - 251, Jan 2010.



Box-and-Match (BMA) : A Simple Idea

BMA Parameter $\mathcal{O}$

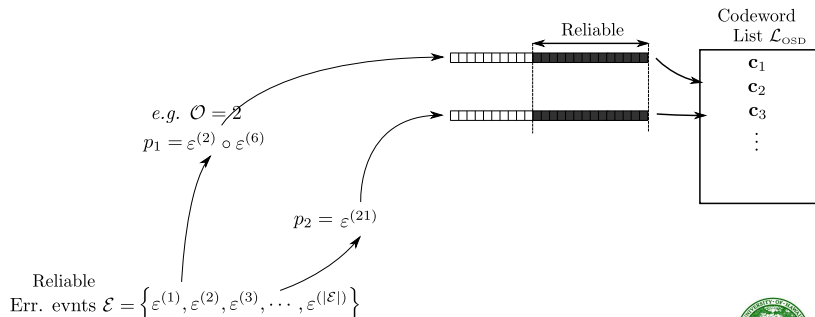
- Hypothesizes $\mathcal{O}$ subsets of error events in $\mathcal{E}$.
- List size $|\mathcal{L}_{\text{BMA}}| > \sum_{i=0}^{\mathcal{O}} \binom{|\mathcal{E}|}{i} = |\mathcal{L}_{\text{OSD}}|$.



Box-and-Match (BMA) : A Simple Idea

The Box-and-Match (BMA) decoder improves the OSD. How?

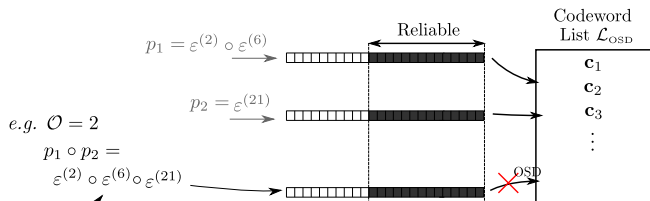
- Let's say the following 3 error events occur.
- Will not be corrected by OSD with order $\mathcal{O} = 2$.
- May be corrected by the BMA with order $\mathcal{O} = 2$.



Box-and-Match (BMA) : A Simple Idea

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e.g. $\mathcal{O} = 2$

$$p_1 \circ p_2 = \varepsilon^{(2)} \circ \varepsilon^{(6)} \circ \varepsilon^{(21)}$$

Reliable

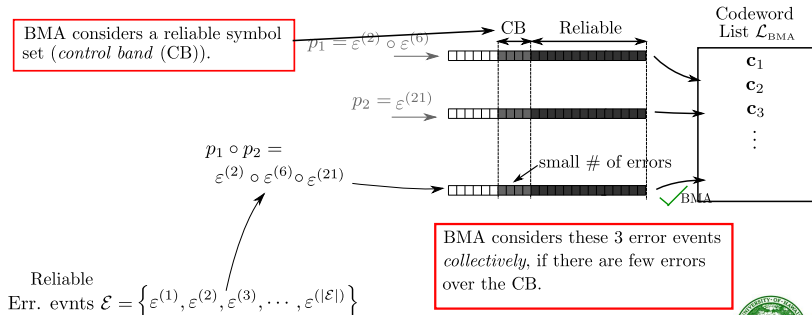
$$\text{Err. evnts } \mathcal{E} = \left\{ \varepsilon^{(1)}, \varepsilon^{(2)}, \varepsilon^{(3)}, \dots, \varepsilon^{(|\mathcal{E}|)} \right\}$$



Box-and-Match (BMA) : A Simple Idea

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Summary of Idea Behind BMA

- BMA extends error correction to $> \mathcal{O}$ error events.
- BMA considers $> \mathcal{O}$ error events, **IF**
 - **BEFORE** codeword is actually formed (computed),
 - the BMA **KNOWS** that codeword disagrees only with small # *control band* (CB) symbols.
- BMA *will not* test all $> \mathcal{O}$ combinations of error events.

Q: How does the BMA know ...

WHICH combinations of $> \mathcal{O}$ error events, map to codewords, with only a small number of errors over the CB?

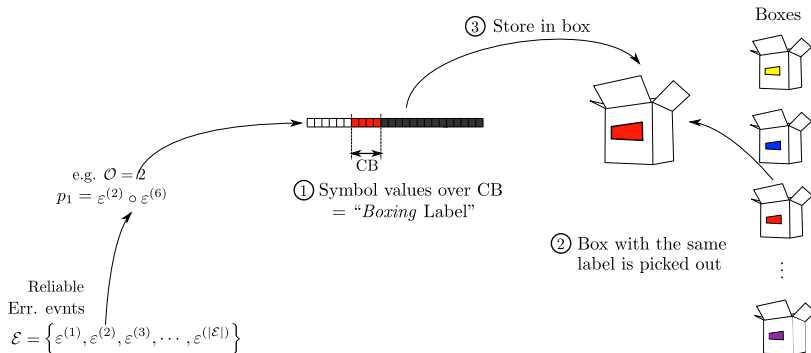
A:

Using the **box-and-match** technique.



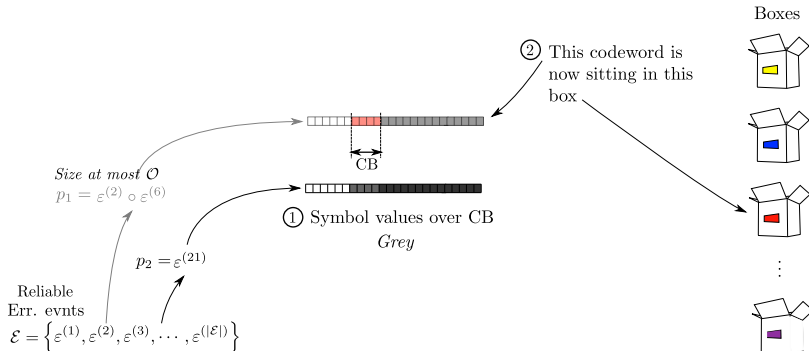
“Boxing” and “Matching” Procedure

- Each codeword is boxed according to its “label”.
- Each codeword is “matched” to other codewords.



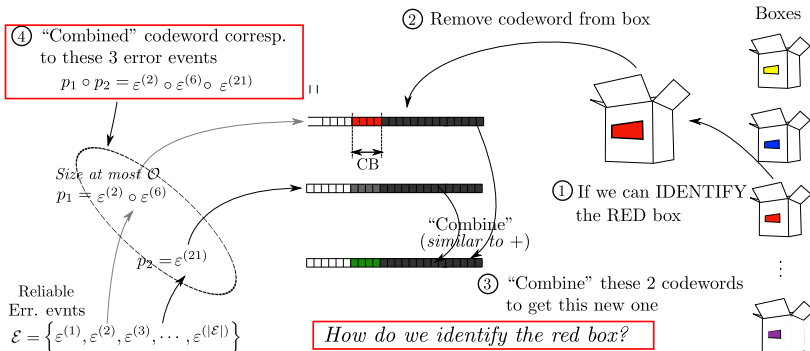
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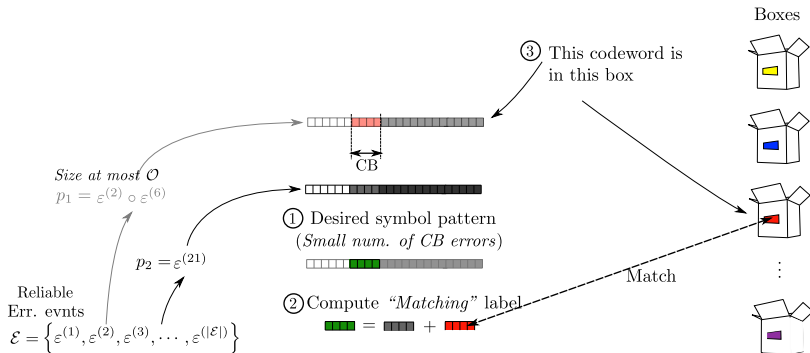
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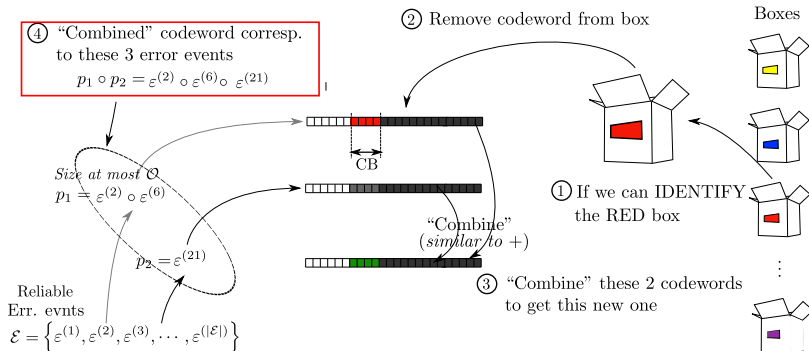
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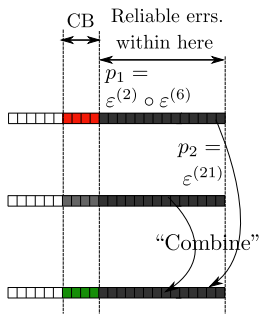
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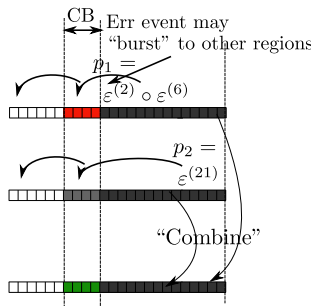


“Boxing” and “Matching” Procedure

- Each codeword is boxed according to its “label”.
- Each codeword is “matched” to other codewords.
- Main difficulty faced.



[Valembois & Fosson (2004)]
(memoryless)



[Lim, Kavčić, & Fos. (2010)]
(ISI)



BMA : Error Correction Guarantee

BMA extends error correction to $> \mathcal{O}$ error events.

If the transmitted codeword $\mathbf{c}$ has the decomposition

$$(\mathbf{c}, \mathbf{b}) = p_1 \circ p_2 \circ p_{\text{CB}} \circ (\mathbf{a}, \mathbf{b})$$

satisfying

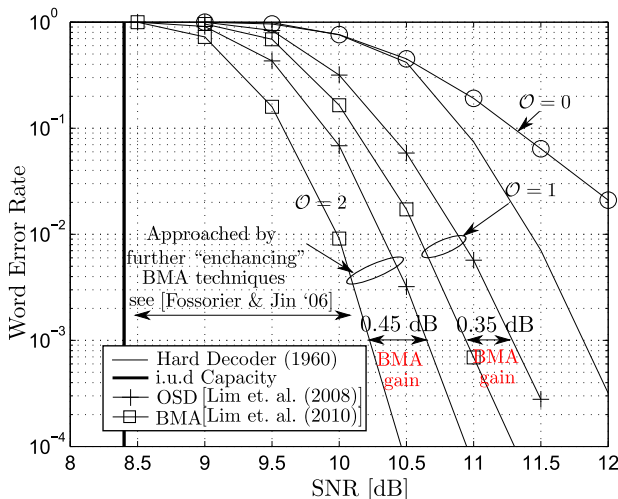
- Both p_1, p_2 each consists of $\leq \mathcal{O}$ reliable EEs
- $p_1 \circ p_2$ consists of $> \mathcal{O}$ reliable EEs
- p_{CB} is a hypothesized control band (CB) error pair.
- $(\mathbf{a}, \mathbf{b})$ consists of all other error events.

then $\mathbf{c}$ is *guaranteed* to be placed into the BMA list.



Performance of BMA

- [255, 239, 17] Reed-Solomon binary image.



Complexity of the BMA

Under some mild assumptions, the complexity of the BMA is computed as follows.

Proposition

For a CB of size κ , the fractional increase in list size of the BMA is

$$\begin{aligned} |\mathcal{L}_{\text{BMA}}|/|\mathcal{L}_{\text{OSD}}| &= 1 + \frac{(|\mathcal{L}_{\text{OSD}}| + 1)|\mathcal{L}_{\text{OSD}}|}{2^{\kappa+1}} \\ &\approx 1 \text{ for large } \kappa \end{aligned}$$

Note

For large CB size κ we have

$$|\mathcal{L}_{\text{BMA}}| \approx |\mathcal{L}_{\text{OSD}}|$$



Outline

1 Introduction

2 Main Results

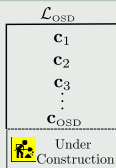
- Detection
- Ordered Statistics Decoding (OSD)
- Box-and-Match (BMA)
- Optimality Testing
- Analysis

3 Conclusion



Complexity Reduction by Optimality Testing

Optimality tests may help reduce OSD complexity



- Can we do better than $\mathbf{c}_{\text{OSD}}$?
- If *no*, stop and save unnecessary computations.

Q1 Is $\mathbf{c}_{\text{OSD}}$ better than *all other codewords*?

Q2 Is $\mathbf{c}_{\text{OSD}}$ better than *all vectors dist. $d_{\min}$ away*?

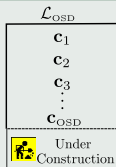
Result : Efficient optimality tests for ISI channels

F. Lim, A. Kavčić, and M. Fossorier, “On sufficient conditions for testing optimality of codewords in ISI channels,” submitted to *ISIT*, 2010.



Complexity Reduction by Optimality Testing

Optimality tests may help reduce OSD complexity

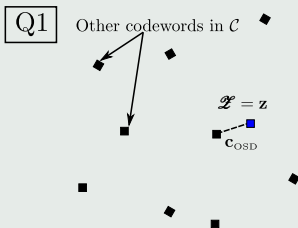


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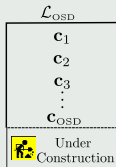
Q2 Is $\mathbf{c}_{\text{OSD}}$ better than **all vectors dist. $d_{\min}$ away**?

E.g. Memoryless ($\mathbf{H}$ is identity)



Complexity Reduction by Optimality Testing

Optimality tests may help reduce OSD complexity

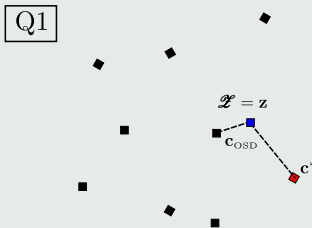


- Can we do better than c_{OSD} ?
- If *no*, stop and save unnecessary computations.

Q1 Is c_{OSD} better than **all other codewords**?

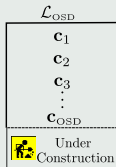
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Complexity Reduction by Optimality Testing

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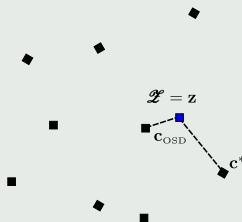
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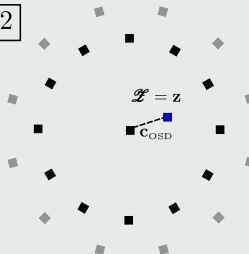
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Q1

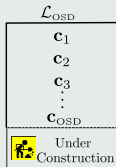


Q2



Complexity Reduction by Optimality Testing

Optimality tests may help reduce OSD complexity



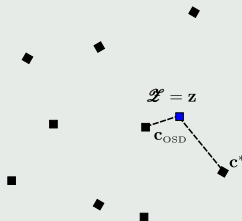
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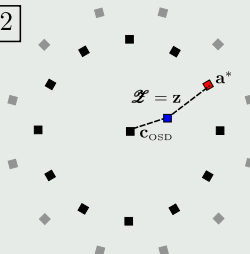
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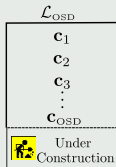


Q2



Complexity Reduction by Optimality Testing

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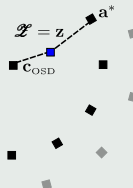
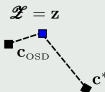
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Q1

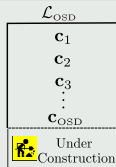
Q2

If the ans
to Q2 is yes
then the ans.
to Q1 is yes



Complexity Reduction by Optimality Testing

Optimality tests may help reduce OSD complexity



- Can we do better than c_{OSD} ?

- If *no*, stop and save unnecessary computations.

Q1 Is c_{OSD} better than **all other codewords**?

Q2 Is c_{OSD} better than **all vectors dist. $d_{\min}$ away**?

Exist *efficient* methods to solve Q2 for *memoryless* channels ...
 [Taipale (1991), Kaneko (1994)]



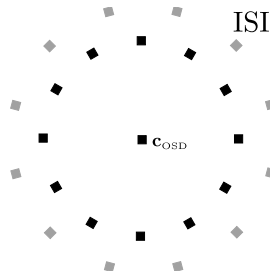
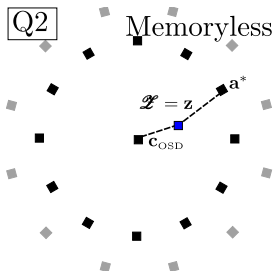
Efficiency is Key

Q2A Integer programming problem

$$\begin{aligned} \mathbf{a}^* &= \arg \max \mathbf{z}^T \mathbf{a} \\ \text{s. t. } d_H(\mathbf{a}, \mathbf{c}_{\text{OSD}}) &\geq d_{\min}, a_i \in \{-1, 1\} \end{aligned}$$

Q2A has simple solution [Taipale (1991)]

- If $\mathbf{H} = \mathbf{I}$, then Q2A $\equiv$ Q2.



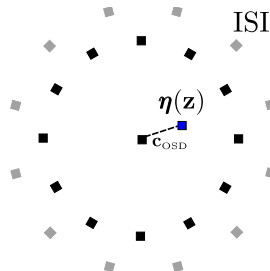
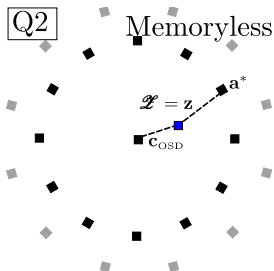
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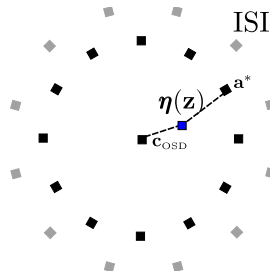
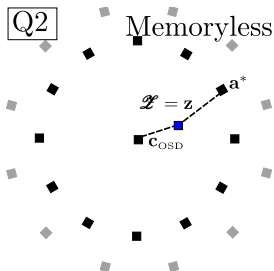
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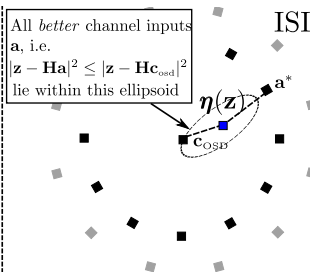
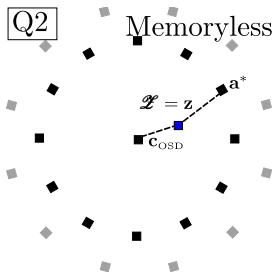
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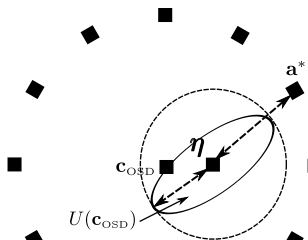
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Optimality Tests for ISI Channels

Defn: ($\mathbf{C}_{\text{OSD}}$ is diagonal matrix with diagonal $\mathbf{c}_{\text{OSD}}$)

- $\boldsymbol{\eta}(\mathbf{z}) \triangleq (1 - \mathbf{C}_{\text{OSD}}(\mathbf{H}^T \mathbf{H})^{-1} \mathbf{H} \mathbf{z})/2$.
- Dist. $U(\mathbf{c}_{\text{OSD}}) \triangleq \frac{|\mathbf{H} \mathbf{C}_{\text{OSD}} \boldsymbol{\eta}|^2}{\lambda_{\min}}$, where $\lambda_{\min}$ is min. eig. of $\mathbf{H}^T \mathbf{H}$.



All *better* channel inputs $\mathbf{a}$, i.e.
 $|\mathbf{z} - \mathbf{H}\mathbf{a}|^2 \leq |\mathbf{z} - \mathbf{H}\mathbf{c}_{\text{osd}}|^2$
 lie within this ellipsoid

Proposition

If we have

$$U(\mathbf{c}_{\text{OSD}}) < |\boldsymbol{\eta} - \mathbf{a}^*|^2, \quad (*)$$

then $\mathbf{a}^*$ is *better* than all $\mathcal{C} \setminus \{\mathbf{c}_{\text{OSD}}\}$.

In other words, if $(*)$ is satisfied...
 ans to Q2 will be YES.

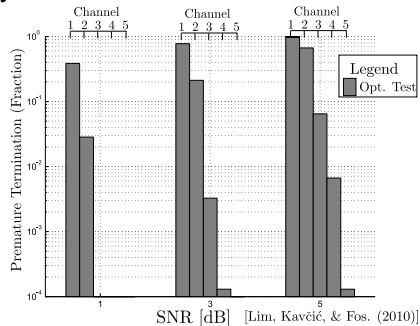


Performance (Computer Simulations)

Performance for $[24, 12, 8]$ Golay Code

- 5 different channels.

No.	$\lambda_{\min}$
1	0.7873
2	0.4309
3	0.2052
4	0.1408
5	0.0783



Outline

1 Introduction

2 Main Results

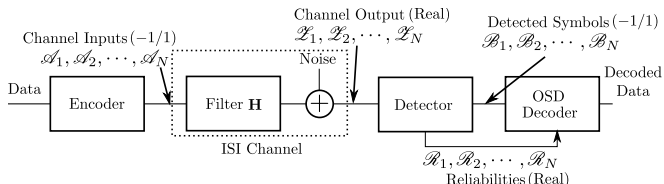
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- Optimality Testing
- **Analysis**

3 Conclusion



Overview of Analytic Results

Recall: System Overview



We provide *expressions* for:

- 1 *reliability distributions* $\Pr \{ \mathcal{R}_{t_1} \leq r_1, \dots, \mathcal{R}_{t_m} \leq r_m \}$, and *error probabilities* $\Pr \{ \mathcal{B}_{t_1} \neq \mathcal{A}_{t_1}, \dots, \mathcal{B}_{t_m} \neq \mathcal{A}_{t_m} \}$.
- 2 *ordered error probabilities*

$$\Pr \{ \mathcal{B}_{(i)} \neq \mathcal{A}_{(i)} \},$$

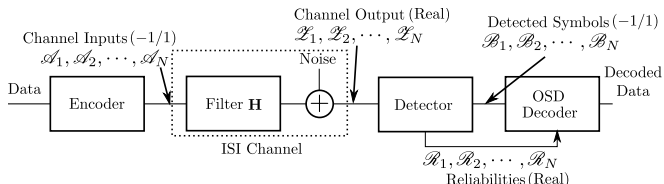
where the i -th ordered position obtained from

$$\mathcal{R}_{(1)} \leq \dots \leq \mathcal{R}_{(i)} \leq \dots \leq \mathcal{R}_{(N)}.$$



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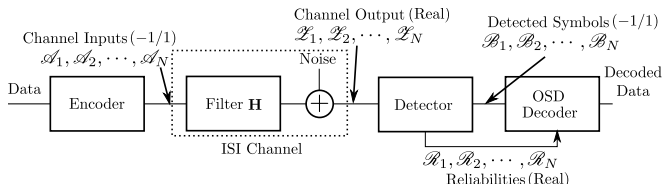
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Analysis : Past Work / Our Contribution

Result : M -truncated Max-Log-Map (MLM) Detector

F. Lim and A. Kavčić, "On the distribution the reliabilities of truncated max-log-map decoders," *submitted*, 2010.

Problem I: Detector Analysis - Solved (Practical Scheme)

- Detection schemes invented in 1970's, however
- their analysis is difficult.
- Recent results include,
 - High signal-to-noise (SNR) approx [Reggiani et. al. (2002)]
 - Memory-2 conv. codes [Lentmaier et. al. (2004)]
- We have a *closed-form* for the decoder presented earlier.



Our is a first, *exact* result for *joint distributions*, *irregardless* of memory length!



Analysis : Past Work / Our Contribution

Result : Conjectured approx. of ordered symbol error prob.

F. Lim, A. Kavčić, M. Fossorier, “Ordered statistics decoding for intersymbol interference channels,” *in preperation*, 2010.

Problem II: Ordered Error Probabilities - In Progress ...

- Solved for memoryless case [Lin & Fossorier (1995)], but
- unknown when sequence $\mathcal{R}_1, \dots, \mathcal{R}_N$ is *dependent*
- We conjecture *approximations*, based on
- known concentration results for dependent sequences.

We show empirical evidence to attest accuracy of our conjectures.



Analysis : Past Work / Our Contribution

Misconception that in approaches similar to [Reggiani et. al. (2002)], the reliability distribution $\mathcal{R}_t$ must be approximated.

[Lim & Kavčić (2010)]: because signaling is *binary*,
consideration of all cases is easy.

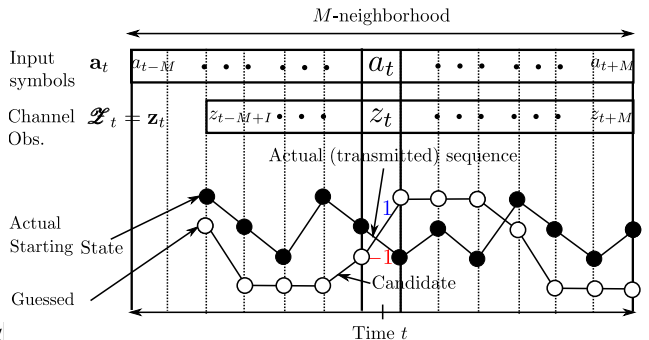
⇒ Discussion on Problem I (Detector Analysis) follows...



Detector Reliability Dist. and Error Probabilities

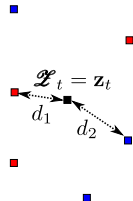
Recall: Reliability sequence $\mathcal{R}_1, \mathcal{R}_2, \dots, \mathcal{R}_N$

$$\mathcal{R}_t \triangleq \min_{\mathbf{a}_t: a_t \neq \mathcal{B}_t} |\mathcal{Z}_t - \mathbf{H}\mathbf{a}_t|^2 - \min_{\mathbf{a}_t: a_t = \mathcal{B}_t} |\mathcal{Z}_t - \mathbf{H}\mathbf{a}_t|^2 \geq 0$$



Euclidean Space

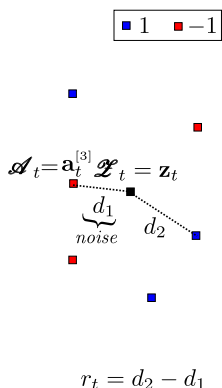
■ 1 ■ -1



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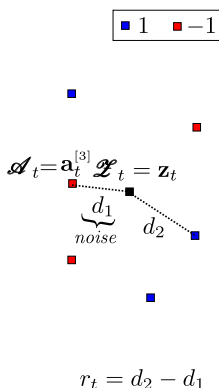
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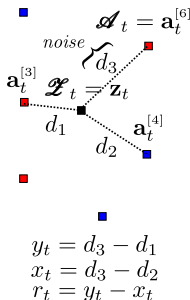
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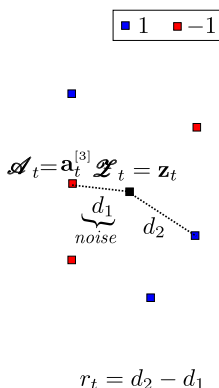
Euclidean Space



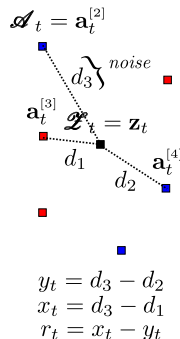
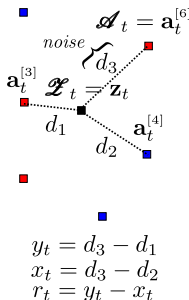
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Euclidean Space

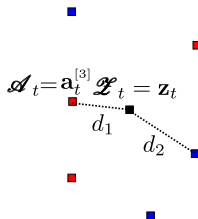


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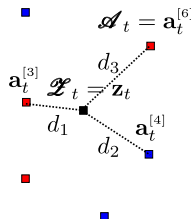
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■ 1 ■ -1

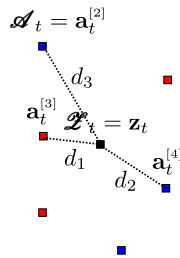


$$\begin{aligned} y_t &= d_1 - d_1 \\ x_t &= d_1 - d_2 \\ r_t &= y_t - x_t \end{aligned}$$

Euclidean Space



$$\begin{aligned} y_t &= d_3 - d_1 \\ x_t &= d_3 - d_2 \\ r_t &= y_t - x_t \end{aligned}$$



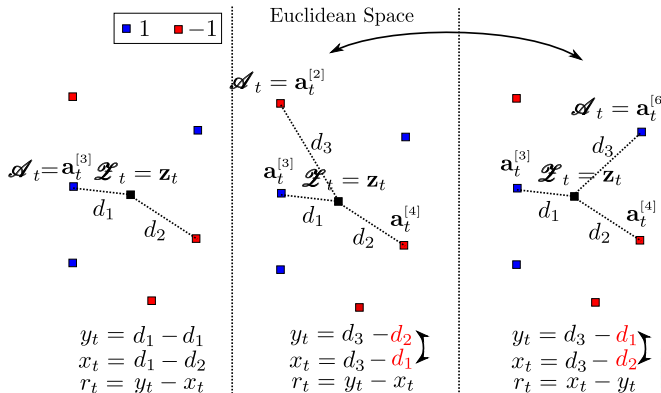
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Our Results

Recall: Reliability sequence $\mathcal{R}_1, \mathcal{R}_2, \dots, \mathcal{R}_N$

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$$\mathcal{Y}_t \triangleq \max_{\mathbf{a}_t: a_t = \mathcal{A}_t} |\mathcal{Z}_t - \mathbf{H}\mathbf{a}_t|^2 - |\mathcal{Z}_t - \mathbf{H}\mathbf{a}_t|^2,$$

$$\mathcal{X}_t \triangleq \max_{\mathbf{a}_t: a_t \neq \mathcal{A}_t} |\mathcal{Z}_t - \mathbf{H}\mathbf{a}_t|^2 - |\mathcal{Z}_t - \mathbf{H}\mathbf{a}_t|^2$$

Proposition [Lim '10]

- The reliability $\mathcal{R}_t$ satisfies

$$\mathcal{R}_t = |\mathcal{X}_t - \mathcal{Y}_t|.$$

- The symbol error event $\{\mathcal{B}_t \neq \mathcal{A}_t\}$ satisfies

$$\{\mathcal{B}_t \neq \mathcal{A}_t\} = \{\mathcal{X}_t \geq \mathcal{Y}_t\}.$$



Our Results

We show the distribution of $\mathcal{X}_t - \mathcal{Y}_t$ is of the form

$$\Pr \{ \mathcal{X}_t - \mathcal{Y}_t \leq r \} = \mathbb{E}_{\mathcal{A}_t, \mathcal{U}} \left\{ \int_0^\infty f_\Gamma(\gamma) \cdot \Phi(r + d(\gamma, \mathcal{A}_t, \mathcal{U})) d\gamma \right\}$$

where

- $\mathcal{U}$ is *uniformly distributed* on the *surface* of a $(2M)$ -dimensional hypersphere.
- $f_\Gamma(\gamma)$ is *standard chi density* with $2M$ degrees of freedom.
- $\Phi(\cdot)$ is a *Gaussian distribution* function.
- $d(\gamma, \mathbf{a}_t, \mathbf{u})$ is the *difference of affine maximums* in γ , i.e.

$$d(\gamma) = d(\gamma, \mathbf{a}_t, \mathbf{u}) = \max(\gamma \cdot \boldsymbol{\alpha} + \beta) - \max(\gamma \cdot \boldsymbol{\alpha}' + \beta')$$
 where $\boldsymbol{\alpha}, \boldsymbol{\alpha}', \beta$ and β' are length- (4^M) vectors depending on $\mathbf{a}_t$ and $\mathbf{u}$.

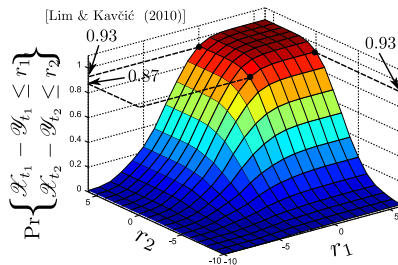
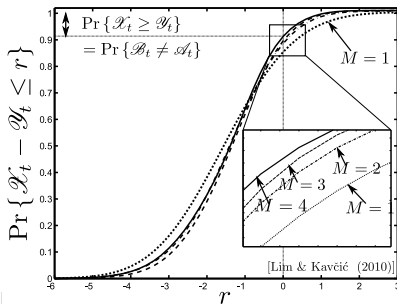


Our Results

We show the distribution of $\mathcal{X}_t - \mathcal{Y}_t$ is of the form

$$\Pr \{ \mathcal{X}_t - \mathcal{Y}_t \leq r \} =$$

$$\mathbb{E}_{\mathcal{A}_t, \mathcal{U}} \left\{ \int_0^\infty f_{\Gamma}(\gamma) \cdot \Phi(r + d(\gamma, \mathcal{A}_t, \mathcal{U})) d\gamma \right\}$$



Our Results

Discussion on Problem I (Detector Analysis) finished.

⇒ Discussion on Problem II (Ordered Symbol Error Prob.) follows...



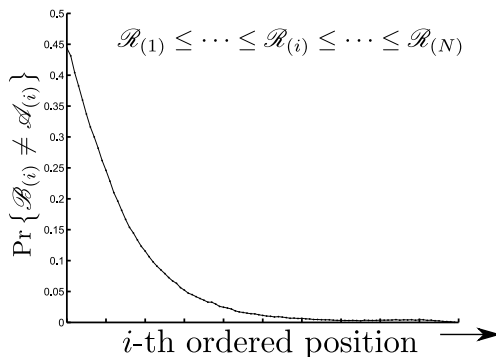
Conjectured Approx. for Ordered Symbol Error Prob.

- [Lim '10] Conjectured approx. for $\Pr \{ \mathcal{B}_{(i)} \neq \mathcal{A}_{(i)} \}$,

$$\int_0^\infty \Pr \{ \mathcal{B} \neq \mathcal{A} | \mathcal{R} = r \} f_{\mathcal{R}_{(i)}}(r) dr, \quad (1)$$

where $f_{\mathcal{R}_{(i)}}(r)$ is the density of $\mathcal{R}_{(i)}$.

Want to *analytically predict* $\Pr \{ \mathcal{B}_{(i)} \neq \mathcal{A}_{(i)} \}$



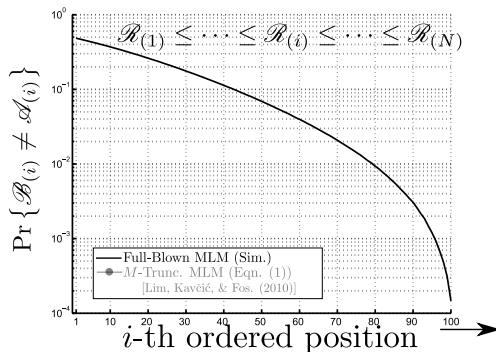
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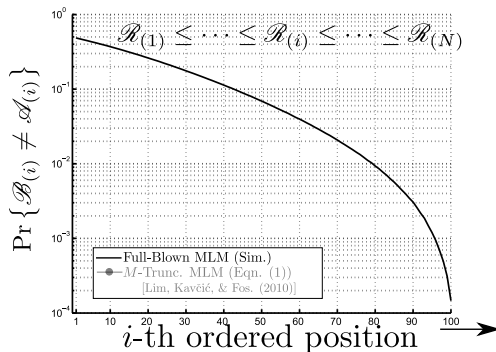
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where $f_{\mathcal{R}_{(i)}}(r)$ is the density of $\mathcal{R}_{(i)}$.

$\Pr \{ \mathcal{B} \neq \mathcal{A} | \mathcal{R} = r \} \Leftarrow \text{Problem I (for any given } M \text{)}.$



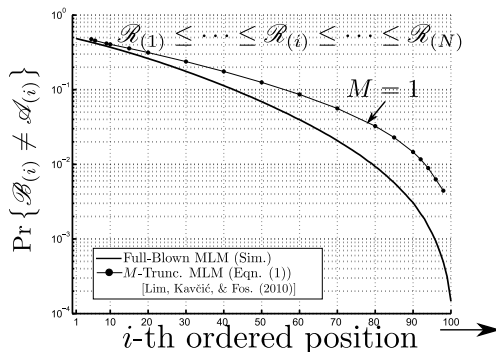
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where $f_{\mathcal{R}_{(i)}}(r)$ is the density of $\mathcal{R}_{(i)}$.

Eqn. (1) seems far off ... ?



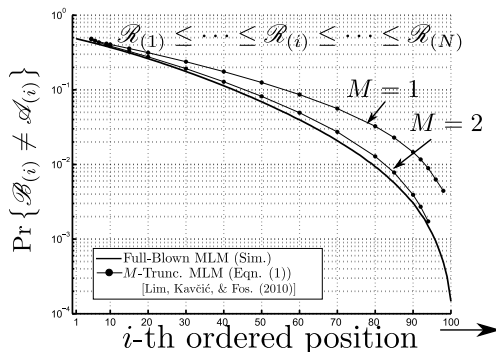
Conjectured Approx. for Ordered Symbol Error Prob.

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where $f_{\mathcal{R}_{(i)}}(r)$ is the density of $\mathcal{R}_{(i)}$.

Wait, we can increase truncation length M .



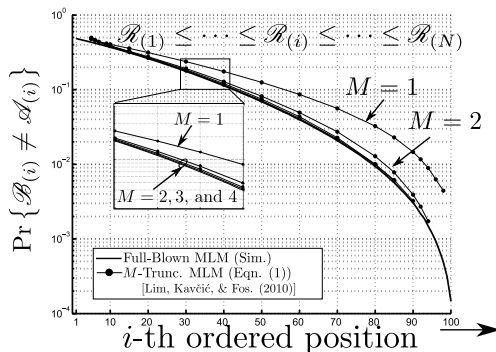
Conjectured Approx. for Ordered Symbol Error Prob.

- [Lim '10] Conjectured approx. for $\Pr \{ \mathcal{B}_{(i)} \neq \mathcal{A}_{(i)} \}$,

$$\int_0^\infty \Pr \{ \mathcal{B} \neq \mathcal{A} | \mathcal{R} = r \} f_{\mathcal{R}_{(i)}}(r) dr, \quad (1)$$

where $f_{\mathcal{R}_{(i)}}(r)$ is the density of $\mathcal{R}_{(i)}$.

Practically the same when $M = 3$ and 4.



Conjectured Approx. for Ordered Symbol Error Prob.

- [Lim '10] Conjectured approx. for $\Pr \{ \mathcal{B}_{(i)} \neq \mathcal{A}_{(i)} \}$,

$$\int_0^\infty \Pr \{ \mathcal{B} \neq \mathcal{A} | \mathcal{R} = r \} f_{\mathcal{R}_{(i)}}(r) dr, \quad (1)$$

where $f_{\mathcal{R}_{(i)}}(r)$ is the density of $\mathcal{R}_{(i)}$.

Unfortunately, we are unable to proof accuracy of (1),

remains a conjectured approx ...



Conjectured Approx. for Ordered Symbol Error Prob.

- [Lim '10] Conjectured approx for $\Pr \{ \mathcal{B}_{(i)} \neq \mathcal{A}_{(i)} \}$, is given as

$$\int_0^\infty \Pr \{ \mathcal{B} \neq \mathcal{A} | \mathcal{R} = r \} f_{\mathcal{R}_{(i)}}(r) dr, \quad (1)$$

where $f_{\mathcal{R}_{(i)}}(r)$ is the density of $\mathcal{R}_{(i)}$.

- [Sen (1968)] $\mathcal{R}_{(i)}$ is approx Gaussian for long N .

To compute (1), we will need

- | | | |
|--|---|---------------------------------|
| <ul style="list-style-type: none"> (i) Marginal $F_{\mathcal{R}}(r)$ (ii) Joint $F_{\mathcal{R}_t, \mathcal{R}_{t+j}}(r_1, r_2)$ (iii) Error probs. $\Pr \{ \mathcal{B} \neq \mathcal{A} \}$ | } | $\Leftarrow$ Problem I (Solved) |
| (iv) Concentration Result [Sen (1968)] | | |



Conclusion

Ordered statistics decoding (OSD) *generalized* for ISI channels.

- OSD/Box-and-Match (BMA)
 - powerful/efficient decoding algorithms
- Codeword optimality testing
 - complexity reduction.
- Reliability distributions / error probabilities obtained.
- Conjectured approx. for ordered error probabilities.

