

Mismatched Decoding and Bit-Interleaved Coded Modulation

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Joint work with

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Outline

Setup

Rates

Error Exponents

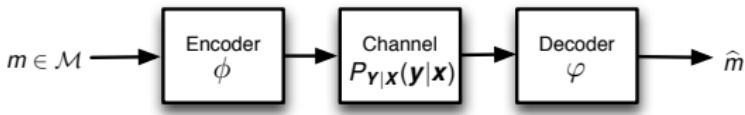
BICM



Setup Rates Error Exponents BICM



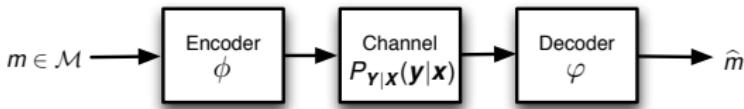
Setup



- ▶ Code $\mathcal{C}(n, |\mathcal{M}|) = \{\mathbf{x}^{(1)}, \dots, \mathbf{x}^{(|\mathcal{M}|)}\}$
 - ▶ codewords $\mathbf{x}^{(i)} \in \mathcal{X}^n$
 - ▶ rate $R = \frac{1}{n} \log |\mathcal{M}|$
- ▶ Discrete memoryless channel

$$P_{Y|X}(y|\mathbf{x}^{(i)}) = \prod_{k=1}^n P_{Y|X}(y_k|x_k^{(i)})$$

Setup



- ▶ Error probability

$$P_e(n, |\mathcal{M}|) = \mathbb{P}\{\hat{M} \neq M\}$$

- ▶ Maximum-Likelihood decoding

$$\hat{m} = \arg \max_{i=1, \dots, |\mathcal{M}|} P_{Y|X}(y|x^{(i)}) = \arg \max_{i=1, \dots, |\mathcal{M}|} \prod_{k=1}^n P_{Y|X}(y_k|x_k^{(i)})$$

Setup

- ▶ Capacity

$$C = \max_{P_X} I(X; Y)$$

- ▶ Random-coding error exponent for distribution Q

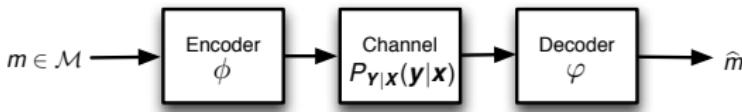
$$E_r(R, Q) = \max_{\rho \in [0, 1]} E_0(\rho, Q) - \rho R$$

$$E_0(\rho, Q) = -\log \sum_{x,y} Q(x) P_{Y|X}(y|x) \left(\frac{\sum_{\bar{x}} Q(\bar{x}) P_{Y|X}(y|\bar{x})^{\frac{1}{1+\rho}}}{P_{Y|X}(y|x)^{\frac{1}{1+\rho}}} \right)^\rho$$

C. Shannon, "A mathematical theory of communication," Bell Syst. Tech. Journal, vol. 27, pp. 379–423, July and Oct. 1948.

R. G. Gallager, *Information Theory and Reliable Communication*, John Wiley & Sons, Inc. New York, NY, USA, 1968.

Setup



- ▶ Maximum-Likelihood decoding

$$\hat{m} = \arg \max_{i=1, \dots, |\mathcal{M}|} P_{Y|X}(\mathbf{y}|\mathbf{x}^{(i)}) = \arg \max_{i=1, \dots, |\mathcal{M}|} \prod_{k=1}^n P_{Y|X}(y_k|x_k^{(i)})$$

- ▶ Maximum-Metric decoding

$$\hat{m} = \arg \max_{i=1, \dots, |\mathcal{M}|} q(\mathbf{x}^{(i)}, \mathbf{y}) = \arg \max_{i=1, \dots, |\mathcal{M}|} \prod_{k=1}^n q(x_k^{(i)}, y_k)$$



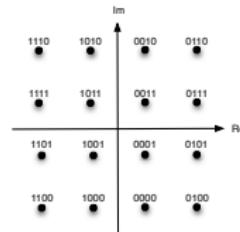
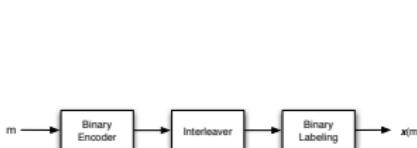
Examples

- ▶ Zero-error
- ▶ Zero undetected error
- ▶ Channel uncertainty $q(x, y) = \hat{P}_{Y|X}(y|x)$
- ▶ Practical constraints
 - ▶ nearest-neighbor (non AWGN)
 - ▶ metric quantization

Examples

Bit-Interleaved Coded Modulation

- ▶ Pragmatic approach
- ▶ Simple decoder
- ▶ Minimal capacity penalty
- ▶ Simple and flexible design
- ▶ Used in most standards (DVB, WiFi, WiMAX, DSL, 4G...)



E. Zehavi, "8-PSK trellis codes for a Rayleigh fading channel", *IEEE Trans. Commun.*, 1992.

G. Caire, G. Taricco and E. Biglieri, "Bit-Interleaved Coded Modulation", *IEEE Trans. Inf. Theory*, 1998.



Examples

Bit-Interleaved Coded Modulation



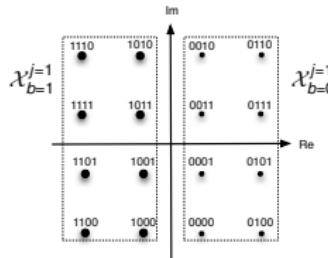
$$\begin{aligned}
 \hat{m} &= \arg \max_{i=1, \dots, |\mathcal{M}|} q(\mathbf{x}^{(i)}, \mathbf{y}) \\
 &= \arg \max_{i=1, \dots, |\mathcal{M}|} \prod_{k=1}^n q(x_k^{(i)}, y_k) \\
 &= \arg \max_{i=1, \dots, |\mathcal{M}|} \prod_{k=1}^n \prod_{j=1}^m q_j(b_j(x_k^{(i)}), y_k)
 \end{aligned}$$

Examples

Bit-Interleaved Coded Modulation

$$q_j(b_j(x) = b, y) = \sum_{x' \in \mathcal{X}_b^j} P_{Y|X}(y|x') P_X(x')$$

$$\hat{m} = \arg \max_{i=1, \dots, |\mathcal{M}|} \prod_{k=1}^n \prod_{j=1}^m \sum_{\substack{x' \in \mathcal{X}_b^j \\ b_j(x_k^{(i)})}} P_{Y|X}(y_k|x') P_X(x')$$



$$C_q = ?$$

Rates

i.i.d. Random Coding

$$\mathbb{P}[\mathbf{X} = \mathbf{x}] = \prod_{k=1}^n Q(x_k)$$

- Mismatched decoder

$$\hat{m} = \arg \max_{i=1, \dots, |\mathcal{M}|} q(\mathbf{x}^{(i)}, \mathbf{y})$$

- Can achieve rate

$$\sum_{x,y} Q(x) P_{Y|X}(y|x) \log \frac{q(x,y)}{\sum_{\bar{x}} Q(\bar{x}) q(\bar{x},y)}$$

$$I(X; Y) = \sum_{x,y} Q(x) P_{Y|X}(y|x) \log \frac{P_{Y|X}(y|x)}{\sum_{\bar{x}} Q(\bar{x}) P_{Y|X}(y|\bar{x})}$$

T. R. M. Fischer, "Some remarks on the role of inaccuracy in Shannon's theory of information transmission," in Trans. 8th Prague Conf. on Inf. Theory, 1971, pp. 211–226.



Rates

i.i.d.

- Mismatched decoder

$$\hat{m} = \arg \max_{i=1, \dots, |\mathcal{M}|} q(\mathbf{x}^{(i)}, \mathbf{y}) = \arg \max_{i=1, \dots, |\mathcal{M}|} q(\mathbf{x}^{(i)}, \mathbf{y})^s$$

- Generalized Mutual Information

$$I^{\text{GMI}}(Q) = \sup_{s \geq 0} \sum_{x,y} Q(x) P_{Y|X}(y|x) \log \frac{q(x,y)^s}{\sum_{\bar{x}} Q(\bar{x}) q(\bar{x},y)^s}$$

G. Kaplan and S. Shamai, "Information rates and error exponents of compound channels with application to antipodal signaling in a fading environment," Arch. Elek. Über., vol. 47, no. 4, pp. 228–239, 1993.



Rates

Constant Composition Random Coding

- ▶ Codewords have the same empirical distribution

$$p_{\mathbf{x}}(\mathbf{x}) = \frac{\text{Number of symbols } x \text{ in } \mathbf{x}}{n}$$

- ▶ If metric $q(x, y)$ is replaced by $q(x, y)^s e^{a(x)}$

$$\prod_{k=1}^n q(x_k, y_k)^s e^{a(x_k)} = \left(\prod_{k=1}^n q(x_k, y_k) \right)^s e^{\sum_{k=1}^n a(x_k)}$$

- ▶ LM rate

$$I_{LM}(Q) = \sup_{s \geq 0, a(\cdot)} \sum_{x,y} Q(x) P_{Y|X}(y|x) \log \frac{q(x, y)^s e^{a(x)}}{\sum_{\bar{x}} Q(\bar{x}) q(\bar{x}, y)^s e^{a(\bar{x})}}$$

J. Hui, "Fundamental issues of multiple accessing," Ph.D. dissertation, MIT, 1983.

I. Csiszár and J. Körner, "Graph decomposition: A new key to coding theorems," IEEE Trans. Inf. Theory, Jan. 1981.



Rates

Cost Constrained Random Coding

- ▶ Codewords meet a cost function

$$\mathbb{P}\{\mathbf{X} = \mathbf{x}\} = \frac{1}{\mu_n} \prod_{k=1}^n Q(x_k) \mathbb{1} \left\{ \left| \frac{1}{n} \sum_{k=1}^n a(x_k) - \mathbb{E}_Q[a(X)] \right| \leq \frac{\delta}{n} \right\}$$

- ▶ LM rate is also achieved

$$I_{LM}(Q) = \sup_{s \geq 0, a(\cdot)} \sum_{x,y} Q(x) P_{Y|X}(y|x) \log \frac{q(x, y)^s e^{a(x)}}{\sum_{\bar{x}} Q(\bar{x}) q(\bar{x}, y)^s e^{a(\bar{x})}}$$

A. Ganti, A. Lapidoth, and E. Telatar, "Mismatched decoding revisited: General alphabets, channels with memory, and the wide-band limit," IEEE Trans. Inf. Theory, Nov. 2000.

S. Shamai and I. Sason, "Variations on the Gallager bounds, connections, and applications," IEEE Trans. Inf. Theory, Dec. 2002

Rates

Properties

- ▶ Different ensembles achieve different rates
- ▶ Ensemble tightness: $\bar{P}_e \rightarrow 1$ when
 - ▶ $R > I_{\text{GMI}}(Q)$ (i.i.d. random coding)
 - ▶ $R > I_{\text{LM}}(Q)$ (constant-composition random coding)
 - ▶ $R > I_{\text{LM}}(Q)$ (cost-constrained random coding)
- ▶ Data Processing Inequality: $I_{\text{LM}}(Q) \leq I(X; Y)$ with equality iff

$$\log q(x, y) = \alpha(x) + \beta(y) + c \log P_{Y|X}(y|x)$$

for some $\alpha(x), \beta(y), c > 0$

- ▶ $I_{\text{LM}}(Q)$ can be *non-convex* in Q

N. Merhav, G. Kaplan, A. Lapidoth, and S. Shamai, "On information rates for mismatched decoders," IEEE Trans. Inf. Theory, Nov. 1994.



Error Exponents

- ▶ Upper bound

$$\bar{P}_e \leq \mathbb{E} [\min \{1, (|\mathcal{M}| - 1)\mathbb{P}\{q(\mathbf{X}', \mathbf{Y}) \geq q(\mathbf{X}, \mathbf{Y}) | \mathbf{X}, \mathbf{Y}\}\}]$$
- ▶ Lower bound

$$\bar{P}_e \geq \frac{1}{4} \mathbb{E} [\min \{1, (|\mathcal{M}| - 1)\mathbb{P}\{q(\mathbf{X}', \mathbf{Y}) \geq q(\mathbf{X}, \mathbf{Y}) | \mathbf{X}, \mathbf{Y}\}\}]$$
- ▶ Ensemble tight exponent given by

$$\liminf_{n \rightarrow \infty} -\frac{1}{n} \log \mathbb{E} [\min \{1, (|\mathcal{M}| - 1)\mathbb{P}\{q(\mathbf{X}', \mathbf{Y}) \geq q(\mathbf{X}, \mathbf{Y}) | \mathbf{X}, \mathbf{Y}\}\}]$$
- ▶ Method of types can be used to find exponent (primal form)

J. Scarlett, A. Martínez, and A. Guillén i Fàbregas, "Mismatched Decoding: Error Exponents, Second-Order Rates and Saddlepoint Approximations", IEEE Trans. Inf. Theory, vol. 60, no. 5, pp. 2647-2666, May 2014.

I. Csiszár and J. Körner, Information Theory: Coding Theorems for Discrete Memoryless Systems, 2nd ed. Cambridge University Press, 2011.

R. Gallager, "Fixed composition arguments and lower bounds to error probability," <http://web.mit.edu/gallager/www/notes/notes5.pdf>.



Error Exponents

Code Ensembles

- i.i.d.

$$\mathbb{P}\{\mathbf{X} = \mathbf{x}\} = \prod_{k=1}^n Q(x_k)$$

- Constant composition

$$\mathbb{P}\{\mathbf{X} = \mathbf{x}\} = \begin{cases} \frac{1}{|\mathcal{T}(Q)|} & \mathbf{x} \in \mathcal{T}(Q) \\ 0 & \text{otherwise} \end{cases}$$

- Multiple cost constraints, $\phi_I = \mathbb{E}_Q[a_I(X)]$

$$\mathbb{P}\{\mathbf{X} = \mathbf{x}\} = \frac{1}{\mu_n} \prod_{k=1}^n Q(x_k) \mathbb{1} \left\{ \left| \frac{1}{n} \sum_{k=1}^n a_I(x_k) - \phi_I \right| \leq \frac{\delta}{n}, I = 1, \dots, L \right\}$$

J. Scarlett, A. Martínez, and A. Guillén i Fàbregas, "Mismatched Decoding: Error Exponents, Second-Order Rates and Saddlepoint Approximations", IEEE Trans. Inf. Theory, vol. 60, no. 5, pp. 2647-2666, May 2014.



Error Exponents

Ensemble Tightness

$$E_r(R, Q) = \max_{0 \leq \rho \leq 1} E_0(\rho, Q) - \rho R$$

$$E_0^{\text{iid}}(\rho, Q) = \sup_{s \geq 0} -\log \sum_{x,y} Q(x) P_{Y|X}(y|x) \left(\frac{\sum_{\bar{x}} Q(\bar{x}) q(\bar{x}, y)^s}{q(x, y)^s} \right)^\rho$$

$$E_0^{\text{cc}}(\rho, Q) = \sup_{s \geq 0, a(\cdot)} -\sum_x Q(x) \log \sum_y P_{Y|X}(y|x) \left(\frac{\sum_{\bar{x}} Q(\bar{x}) q(\bar{x}, y)^s e^{a(\bar{x})}}{q(x, y)^s e^{a(x)}} \right)^\rho$$

$$E_0^{\text{cost}}(\rho, Q, \{a_I(\cdot)\})$$

$$= \sup_{\substack{s \geq 0 \\ \{r_I\}, \{\tilde{r}_I\}}} -\log \sum_{x,y} Q(x) P_{Y|X}(y|x) \left(\frac{\sum_{\bar{x}} Q(\bar{x}) q(\bar{x}, y)^s e^{\sum_{l=1}^L \tilde{r}_l (a_l(\bar{x}) - \phi_l)}}{q(x, y)^s e^{\sum_{l=1}^L r_l (a_l(x) - \phi_l)}} \right)^\rho$$

J. Scarlett, A. Martínez, and A. Guillén i Fàbregas, "Mismatched Decoding: Error Exponents, Second-Order Rates and Saddlepoint Approximations", IEEE Trans. Inf. Theory, vol. 60, no. 5, pp. 2647-2666, May 2014.



Error Exponents

Proposition

$$E_r^{\text{iid}}(R, Q) \leq E_r^{\text{cost}}(R, Q, \{a_l\}) \leq E_r^{\text{cc}}(R, Q)$$

$$\sup_{a_1(\cdot), a_2(\cdot)} E_r^{\text{cost}}(R, Q, \{a_1, a_2\}) = E_r^{\text{cc}}(R, Q)$$

J. Scarlett, A. Martínez, and A. Guillén i Fàbregas, "Mismatched Decoding: Error Exponents, Second-Order Rates and Saddlepoint Approximations", *IEEE Trans. Inf. Theory*, vol. 60, no. 5, pp. 2647-2666, May 2014.

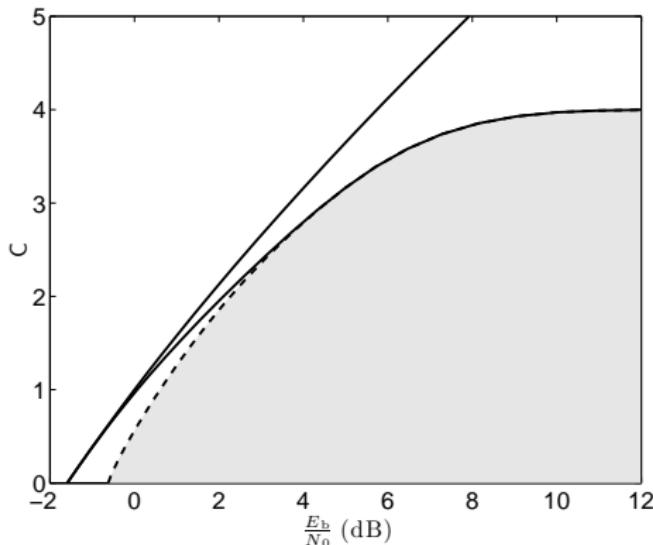
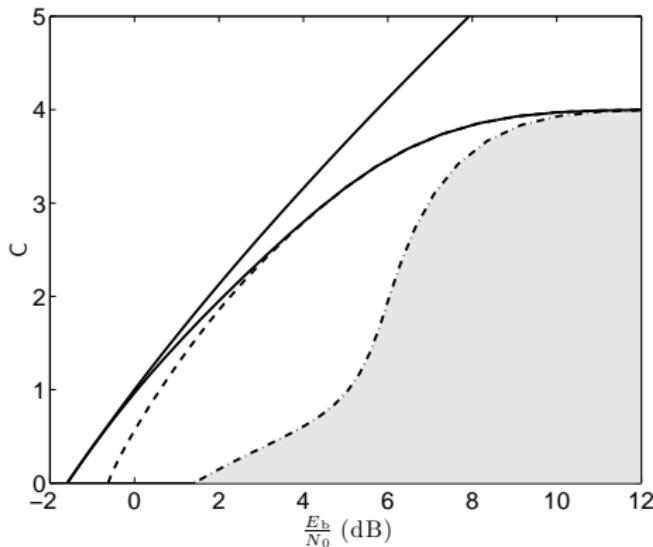


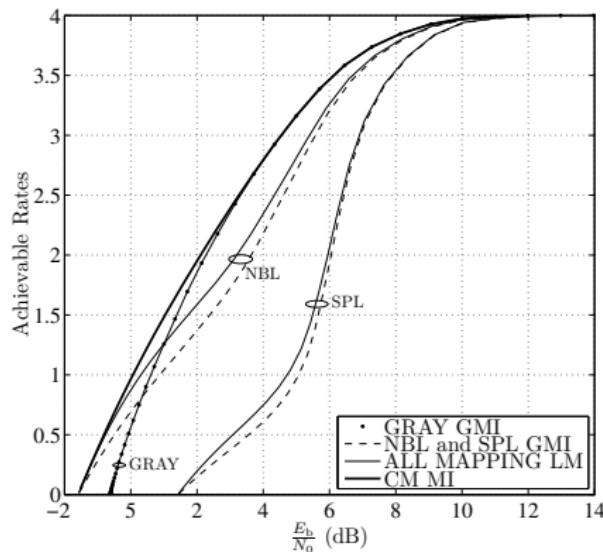
BICM

Theorem

$$I_{GMI}(Q) = \sum_{j=1}^m I(B_j; Y)$$

A. Martínez, A. Guillén i Fàbregas, G. Caire and F. Willems, "Bit-Interleaved Coded Modulation Revisited: A Mismatched Decoding Perspective", *IEEE Trans. Inf. Theory*, June 2009.

BICM**BICM**



Wideband Regime

- ▶ Information rates can be expanded as

$$C(\text{SNR}) = c_1 \text{SNR} + c_2 \text{SNR}^2 + o(\text{SNR}^2)$$

- ▶ c_1, c_2 depend on the transmission scheme
- ▶ c_1 and c_2 trade off power and bandwidth (Verdú 2002)
 - ▶ $c_1 = 1$ for many modulation formats
 - ▶ $c_2 = -\frac{1}{2}$ for proper complex modulations (Prelov, Verdú 2004)

BICM

Wideband Regime

Theorem (Wideband Regime BICM)

The low-SNR expansions of BICM achievable rates satisfy

$$C_{1,GMI} = C_{1,LM} = \sum_{j=1}^m \frac{1}{2} \sum_{b=0}^1 \left| \mathbb{E}[X_b^j] \right|^2 \quad (1)$$

$$C_{2,GMI} = -\frac{1}{2} \left(m \kappa(\mathcal{X}) - \sum_{i=1}^m \sum_b \frac{1}{2} \kappa(\mathcal{X}_b^i) \right) \quad (2)$$

$$C_{2,LM} = C_{2,GMI} + \frac{1}{2} \sum_x \frac{1}{2^m} \left(\sum_{i,j=1; i \neq j}^m r(\bar{\mathcal{X}}_{b_i(x)}^i, \bar{\mathcal{X}}_{b_j(x)}^j) \right)^2 \quad (3)$$

A. Martínez, L. Peng, A. Alvarado, A. Guillén i Fàbregas, "Improved Information Rates for Bit-Interleaved Coded Modulation," 2015 IEEE Int. Symp. Inf. Theory, Hong Kong.



BICM

Wideband Regime: Square 2^m -QAM

	C_1		C_2	
	GMI	LM	GMI	LM
BRGL	$\frac{3 \cdot 2^{2m}}{4 \cdot (2^{2m} - 1)} < 1$	$\frac{3 \cdot 2^{2m}}{4 \cdot (2^{2m} - 1)} < 1$	(2)	(3)
NBL	1	1	(2) ($< -\frac{1}{2}$)	$-\frac{1}{2}$

