# Hybrid Coding

Young-Han Kim, UCSD

Living Information Theory Workshop Happy <70.42><sup>th</sup> birthday, Professor Berger!

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Sung Hoon Lim, Paolo Minero, and Abbas El Gamal







INFORMATION AND CONTROL 13, 254-273 (1968)

#### Rate Distortion Theory for Sources with Abstract Alphabets and Memory

TOBY BERGER<sup>†</sup>

Advanced Development Laboratory, Raytheon Company Wayland, Massachusetts 01778 INFORMATION AND CONTROL 13, 254-273 (1968)

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**DEFINITION 1.** A source  $[X, \mu]$  is *block ergodic* if it is  $\tau$ -ergodic for every positive  $\tau \in M$ .

We show in the appendix that block ergodicity lies between ergodicity and weak mixing in restrictiveness. Note that  $\tau$ -ergodicity of  $[X, \mu]$  and ergodicity of  $[X, \mu]_{\tau}$  are equivalent.



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#### THE INFORMATION THEORY APPROACH TO COMMUNICATIONS

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MULTITERMINAL SOURCE CODING

TOBY BERGER School of Electrical Engineering Cornell University, Ithaca, New York



- Strong typicality
- (Extended) Markov lemma

#### RATE DISTORTION THEORY

A Mathematical Basis For Data Compression

Cornell University

Prentice-Hall, Inc. Englewood Cliffs, New Jersey



TOBY BERGER

#### **Rate Distortion Theory**

A MATHEMATICAL BASIS FOR DATA COMPRESSION

#### INFORMATION AND SYSTEM SCIENCES SERIES

Thomas Kailath Editor



But life is short and information endless: nobody has time for everything. In practice we are generally forced to choose between an unduly brief exposition and no exposition at all. Abbreviation is a necessary evil and the abbreviator's business is to make the best of a job which, though intrinsically bad, is still better than nothing. He must learn to simplify, but not to the point of falsification. He must learn to concentrate upon the essentials of a situation, but without ignoring too many of reality's qualifying side issues.

> —Aldous Huxley Brave New World Revisited



#### Question

Find the sufficient and necessary condition to achieve

 $\limsup_{n\to\infty}\frac{1}{n}\sum_{i=1}^n d(S_i,\hat{S}_i) \leq D$ 



#### Answer: Source-channel separation (Shannon 1948, 1959)

Sufficient and necessary condition:

$$R(D) = \min_{p(\hat{s}|s): \mathsf{E}(d(S,\hat{S})) \le D} I(S;\hat{S}) < \max_{p(x)} I(X;Y) = C$$



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Digital source-channel interface







 Optimal for quadratic Gaussian source coding over the Gaussian channel with average power constraint (Goblick 1965)



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• "To code or not to code" (Gastpar, Rimoldi, and Vetterli 2003)



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- "To code or not to code" (Gastpar, Rimoldi, and Vetterli 2003)
- Analog source–channel interface

## Hybrid Coding



## Hybrid Coding





I(U;S) < I(U;Y)

for some p(u|s), x(u, s),  $\hat{s}(u, y)$  such that  $E(d(S, \hat{S})) \leq D$ 



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Analog/digital hybrid source-channel interface



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  - *U* = Ø: uncoded transmission



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  - $U = (\hat{S}, X) \sim p(\hat{s}|s)p(x)$ : separate source and channel coding



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- "Beaten?" (Shamai, Verdú, and Zamir 1998, Mittal and Phamdo 2002, Tinguely and Lapidoth 2006, Caire and Narayanan 2007, Kochman, Khina, Erez, and Zamir 2009, Tian, Diggavi, and Shamai 2010)



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Random codebook generation

 $2^{nR}$  sequences  $u^n(m) \sim \prod_{i=1}^n p_U(u_i)$  for  $m \in [1:2^{nR}]$ 



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#### Joint typicality encoding

Find an index *m* such that  $(u^n(m), s^n) \in \mathcal{T}_{\epsilon'}^{(n)}$  and transmit  $x_i = x(u_i(m), s_i)$ :



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Find the unique index  $\hat{m}$  such that  $(u^n(\hat{m}), y^n) \in \mathcal{T}_{\epsilon}^{(n)}$  and reconstruct  $\hat{s}_i = \hat{s}(u_i(\hat{m}), y_i)$ :



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## Correlated Sources over a MAC



#### Hybrid coding

Berger–Tung source coding + MAC coding

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Berger–Tung source coding + MAC coding

 $R_1 > I(U_1; S_1),$  $R_2 > I(U_2; S_2)$


### Hybrid coding

Berger–Tung source coding + MAC coding

 $R_1 > I(U_1; S_1),$  $R_2 > I(U_2; S_2)$   $\begin{aligned} R_1 &< I(U_1; Y, U_2), \\ R_2 &< I(U_2; Y, U_1), \\ R_1 &+ R_2 &< I(U_1, U_2; Y) + I(U_1; U_2) \end{aligned}$ 



### Hybrid coding

Berger–Tung source coding + MAC coding

 $I(U_1; S_1|Q) < I(U_1; Y, U_2|Q),$   $I(U_2; S_2|Q) < I(U_2; Y, U_1|Q),$  $I(U_1; S_1|Q) + I(U_2; S_2|Q) < I(U_1, U_2; Y|Q) + I(U_1; U_2|Q)$ 

for some  $p(q)p(u_1|s_1, q)p(u_2|s_2, q)$ ,  $x_j(u_j, s_j, q)$ ,  $\hat{s}_j(u_1, u_2, y, q)$  such that  $E(d_j(S_j, \hat{S}_j)) \le D_j$ , j = 1, 2



### Hybrid coding

Berger–Tung source coding + MAC coding



- Berger–Tung source coding + MAC coding
- Special cases
  - $(\hat{S}_1, \hat{S}_2) = (S_1, S_2)$ : lossless coding (Cover, El Gamal, and Salehi 1980)



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  - $(\hat{S}_1, \hat{S}_2) = (S_1, S_2)$ : lossless coding (Cover, El Gamal, and Salehi 1980)
  - $Y = (X_1, X_2)$ : noiseless channel (Berger 1978, Tung 1978)



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  - $(\hat{S}_1, \hat{S}_2) = (S_1, S_2)$ : lossless coding (Cover, El Gamal, and Salehi 1980)
  - $Y = (X_1, X_2)$ : noiseless channel (Berger 1978, Tung 1978)
  - Gaussian  $(S_1, S_2)$  and Gaussian MAC (Tinguely and Lapidoth 2006)



- Berger–Tung source coding + MAC coding
- Special cases
  - $(\hat{S}_1, \hat{S}_2) = (S_1, S_2)$ : lossless coding (Cover, El Gamal, and Salehi 1980)
  - $Y = (X_1, X_2)$ : noiseless channel (Berger 1978, Tung 1978)
  - Gaussian  $(S_1, S_2)$  and Gaussian MAC (Tinguely and Lapidoth 2006)
- Common part: Kaspi–Berger coding + Slepian–Wolf MAC coding







Hybrid coding

Conceptually simple yet performs very well



- Conceptually simple yet performs very well
- Paolo Minero's talk on Monday 10:05 am (right before David Tse)



#### Existing coding schemes

Decode–forward (digital-to-digital interface)



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- Decode–forward (digital-to-digital interface)
- Amplify–forward (analog-to-analog interface)



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- Noisy network coding (analog-to-digital interface)



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#### Question

Can we do better than these coding schemes?



#### Answer

Hybrid coding (analog-to-[analog/digital] interface)



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#### Answer

Hybrid coding (analog-to-[analog/digital] interface)



Naturally combines noisy network coding and amplify–forward





#### Cutset bound

 $C \leq \max_{p(x_1, x_2, x_3)} \min\{H(Y_2, Y_3), H(Y_2) + H(Y_4 | X_3), H(Y_3) + H(Y_4 | X_2), H(Y_4)\}$ 



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### General lower bound (Avestimehr, Diggavi, and Tse 2007)

 $C \ge \max_{p(x_1)p(x_2)p(x_3)} \min\{H(Y_2, Y_3), H(Y_2) + H(Y_4|X_3), H(Y_3) + H(Y_4|X_2), H(Y_4)\}$ 



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#### Hybrid coding lower bound

 $C \ge \max_{p(x_1)p(x_2|y_2)p(x_3|y_3)} \min\{H(Y_2, Y_3), H(Y_2) + H(Y_4|X_3), H(Y_3) + H(Y_4|X_2), H(Y_4)\}$ 





Binary erasure MAC  $y_4(x_2, x_3)$   $Y_4 = X_2 + X_3$   $X_2, X_3 \in \{0, 1\},$  $Y_4 \in \{0, 1, 2\}$ 





Binary erasure MAC  $y_4(x_2, x_3)$   $Y_4 = X_2 + X_3$   $X_2, X_3 \in \{0, 1\},$  $Y_4 \in \{0, 1, 2\}$ 

General lower bound =  $1.5 < \log 3$  = Hybrid coding lower bound

### Example 2: Gaussian Diamond Network



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## AF vs. NNC vs. Hybrid Coding



# AF vs. NNC vs. Hybrid Coding



## Example 3: Gaussian Two-Way Relay Channel



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## Hybrid Coding vs. Other Coding Schemes



## Hybrid Coding vs. Other Coding Schemes



### Hybrid coding

Versatile interface for source-channel coding and relaying

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- Joint source-channel coding can be simple (and fun too!)

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Back to reality

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### Back to reality

Barron and Oppenheim (2002): US Patent 6,441,764

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- · Barron and Oppenheim (2002): US Patent 6,441,764
- Meir Feder: Amimon high-definition wireless audio-video modem

## Hybrid coding

- Versatile interface for source-channel coding and relaying
- Joint source-channel coding can be simple (and fun too!)
- Hybrid coding > amplify–forward + noisy network coding
- Hybrid coding + structured coding > decode-forward

### Back to reality

- · Barron and Oppenheim (2002): US Patent 6,441,764
- Meir Feder: Amimon high-definition wireless audio-video modem
- Any good code for both (source) encoding and (channel) decoding?