

On optimal weighted-sum rates for the modulo sum problem



Chandra Nair¹

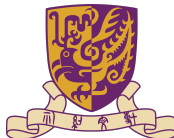


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Distributed Source Coding

Let (X^n, Y^n) be a sequence of random variables that are generated i.i.d. according to $p(x, y)$, and denote $Z^n = (f(X_1, Y_1), f(X_2, Y_2), \dots, f(X_n, Y_n))$.

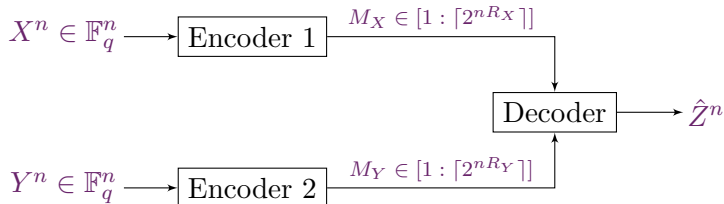


Figure 1: Distributed Source Coding



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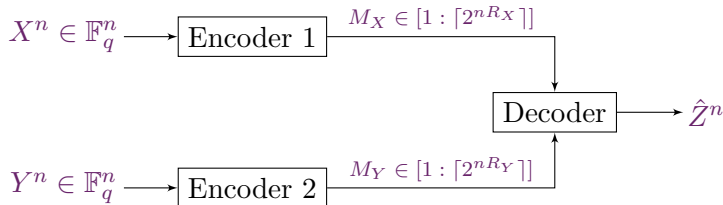


Figure 1: Distributed Source Coding

A rate pair (R_X, R_Y) is said to be achievable if there exists a sequence of (n, R_X, R_Y) -codes such that the $Pr(\hat{Z}^n \neq Z^n) \rightarrow 0$ as $n \rightarrow \infty$.

Problem

Consider $f(X_i, Y_i) = X_i \oplus Y_i$, what rate pairs (R_X, R_Y) are achievable?



Known results

Slepian-Wolf region: [Slepian-Wolf 73']

When $Z = (X, Y)$, the capacity region is given by

$$R_X \geq H(X|Y)$$

$$R_Y \geq H(Y|X)$$

$$R_X + R_Y \geq H(XY)$$



Known results

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Recovering (X, Y) with high probability is sufficient to decode any function $f(X, Y)$.



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Körner-Marton region: [Körner-Marton 79']

When $Z = X \oplus Y$, a rate pair (R_X, R_Y) is achievable by linear codes if

$$R_X \geq H(Z)$$

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This matches capacity region when (X, Y) follows Doubly Symmetric Binary Source distribution.

Remark:

First example showing that structured codes outperforms random coding for multiuser information theory problems.



Known results

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First example showing that structured codes outperforms random coding for multiuser information theory problems.

More distributions are discovered for optimality of this region on weighted sum rate. (this work)



Known results

Exercise 16.23 in [Csiszár-Körner]¹

When $H(Z) \geq \min\{H(X), H(Y)\}$, Slepian-Wolf rate region characterizes the capacity region \mathcal{C} for $Z = X \oplus Y$ in $GF(2)$.

¹I. Csiszár and J. Körner, Information theory: Coding theorems for discrete memoryless systems



Known results

Exercise 16.23 in [Csiszár-Körner]¹

More distributions are discovered for optimality of Slepian-Wolf region on weighted sum rate. (this work)

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Known results

Exercise 16.23 in [Csiszár-Körner]¹

More distributions are discovered for optimality of Slepian-Wolf region on weighted sum rate. (this work)

Ahlsvede-Han region: [Ahlsvede-Han 83']

When $Z = X \oplus Y$, a rate pair (R_X, R_Y) is achievable via a combination of linear codes and random binning if

$$\begin{aligned}R_X &\geq I(U; X|V) + H(Z|UV) \\R_Y &\geq I(V; Y|U) + H(Z|UV) \\R_X + R_Y &\geq I(UV; XY) + 2H(Z|UV)\end{aligned}$$

for some U and V that satisfy the Markov chain $U \rightarrow X \rightarrow Y \rightarrow V$.

¹I. Csiszár and J. Körner, Information theory: Coding theorems for discrete memoryless systems



Known results

Cut-set lower bound: [Körner-Martón 79']

Any achievable rate pair (R_X, R_Y) for the modulo sum problem must satisfy

$$\begin{aligned}R_X &\geq H(Z|Y) = H(X|Y) \\R_Y &\geq H(Z|X) = H(Y|X) \\R_X + R_Y &\geq H(Z).\end{aligned}$$



Main result: A lower bound

Theorem 1

Any achievable rate pair (R_X, R_Y) for the modulo sum problem must satisfy the following constraints for any $\lambda \geq 1$:

$$R_X + \lambda R_Y \geq H(XY) + \min_{U \rightarrow X \rightarrow Y} \lambda H(Z|U) - H(Y|U)$$
$$\lambda R_X + R_Y \geq H(XY) + \min_{V \rightarrow Y \rightarrow X} \lambda H(Z|V) - H(X|V)$$



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Remark: From [Nair 10' 13']

$$\begin{aligned} &\min_{U \rightarrow X \rightarrow Y} \lambda H(Z|U) - H(Y|U) \\ &= -\mathfrak{C}_{\mu(x)}[H(Y) - \lambda H(Z)]|_{p(x)}, \end{aligned}$$

where $\mathfrak{C}_x[f]|_{x_0}$ denotes the upper concave envelope of the function $f(x)$ with respect to x evaluated at $x = x_0$.



Proof sketch

The multi-letter extensions of Ahlswede-Han region tends to the capacity region.

$$\begin{aligned} & n(R_X + \lambda R_Y) + n\varepsilon \\ & \geq I(U; X^n) + \lambda I(V; Y^n|U) + (1 + \lambda)H(Z^n|UV) \end{aligned}$$



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The equalities (a) follows from Markov chain $V \rightarrow Y^n \rightarrow (U, Z^n)$



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The equalities (a) (b) follows from Markov chain $V \rightarrow Y^n \rightarrow (U, Z^n)$



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The equalities (a) (b) follows from Markov chain $V \rightarrow Y^n \rightarrow (U, Z^n)$ and (c) is due to $H(Z^n|UY^n) = H(X^n Z^n|UY^n) = H(X^n|UY^n)$.



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 & = H(X^n Y^n) + \lambda I(V; Y^n|UZ^n) + \lambda H(Z^n|U) - H(Y^n|U) + I(Z^n; Y^n|U, V) \\
 & \geq nH(XY) + \lambda H(Z^n|U) - H(Y^n|U)
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Single-letterize the lower bound

Lemma 1

Let $\lambda \geq 1$ and let (X^n, Y^n) be i.i.d distributed according to $p(x, y)$ where X, Y take values in a finite field. Let Z^n be obtained as $Z_i = X_i \oplus Y_i, i = 1, \dots, n$, i.e. the component-wise modulo sum on the field. Then for any $\lambda \geq 1$ the following holds:

$$\begin{aligned} & \min_{\hat{U}: \hat{U} \rightarrow X^n \rightarrow Y^n} \lambda H(Z^n | \hat{U}) - H(Y^n | \hat{U}) \\ &= n \left(\min_{U: U \rightarrow X \rightarrow Y} \lambda H(Z | U) - H(Y | U) \right). \end{aligned}$$



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- Taking i.i.d. copies of the minimizer of the right hand side, left hand side is at most the value of right hand side.



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Proof sketch

- Taking i.i.d. copies of the minimizer of the right hand side, left hand side is at most the value of right hand side.
- The other direction follows from Markov chain $(\hat{U}, Y_{i+1}^n, Z^{i-1}) \rightarrow X_i \rightarrow (Y_i, Z_i)$ and Körner-Marton identity.



Conditions for lower bound to be tight

Lemma 2

The lower bound for the weighted sum-rate $R_X + \lambda R_Y$, for $\lambda \geq 1$ given in Theorem 1 is optimal, i.e. matches the weighted sum-rate of the capacity region, if either of the following conditions hold:

- (i) $\mathfrak{C}_{\mu(x)}[H(Y) - \lambda H(Z)]|_{p(x)} = H(Y) - \lambda H(Z)$ and Y is independent of Z ,
- (ii) $\mathfrak{C}_{\mu(x)}[H(Y) - \lambda H(Z)]|_{p(x)} = H(Y|X) - \lambda H(Z|X)$.

Further if condition (i) holds for some $\lambda_1 > 1$, then it will also hold for $1 \leq \lambda \leq \lambda_1$; and if condition (ii) holds for some $\lambda_2 \geq 1$, then it will also hold for $\lambda \geq \lambda_2$.



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Remark:

A relatively easier condition to verify is the convexity of $H(Y) - \lambda H(Z)$ with respect to the distribution of X .



Application to binary alphabets GF(2)

Notation: We will parameterize the space of distributions over pairs of binary alphabets, $p(x, y)$ as follows:

$$P(X = 0) = x, P(Y = 0|X = 0) = c, P(Y = 1|X = 1) = d.$$



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Proposition 1: Optimality of Slepian-Wolf region

The optimal weighted sum-rate of the capacity region is given by the Slepian Wolf region if any of the following conditions hold:

- (i) For any λ , if $(c - \frac{1}{2})(d - \frac{1}{2}) \leq 0$, or
- (ii) $\lambda \geq \left(\frac{c-\bar{d}}{c-d}\right)^2$, $c \neq d$, and $(c - \frac{1}{2})(d - \frac{1}{2}) > 0$.

where $\bar{d} = 1 - d$.



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Remarks:

- (i) The condition (i) above is already known and stated as exercise 16.23 page 390 of Csiszár and Körner's book. One can verify that that $H(Z) \geq H(Y)$ is equivalent to $(c - \frac{1}{2})(d - \frac{1}{2}) \leq 0$.
- (ii) Note that an equivalent proposition can also be stated for the alternate parameterization: $P(Y = 0) = y, P(X = 0|Y = 0) = \hat{c}, P(X = 1|Y = 1) = \hat{d}$.



Application to binary alphabets GF(2)

Proposition 2: Optimality of linear codes

Let $P(X = 0) = x$, $P(Y = 0|X = 0) = c$, $P(Y = 1|X = 1) = d$ where $x = \frac{\sqrt{d\bar{d}}}{\sqrt{d\bar{d}} + \sqrt{c\bar{c}}}$.

The optimal weighted sum-rate of the capacity region is given by the Körner-Marton region, i.e. using linear codes, if any of the following conditions hold:

- (i) For any λ , if $c = d$, or
- (ii) $1 \leq \lambda \leq \lambda_1$, $c \neq d$, and $(c - \frac{1}{2})(d - \frac{1}{2}) > 0$, where λ_1 is the larger root of the quadratic equation

$$\lambda^2(c - d)^2 + \lambda(2(c - d)(c - \bar{d}) - 4d\bar{d}(c - \bar{c})^2) + (c - \bar{d})^2 = 0.$$

where $\bar{d} = 1 - d$, $\bar{c} = 1 - c$.



Application to binary alphabets GF(2)

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where $\bar{d} = 1 - d, \bar{c} = 1 - c$.

Remarks:

- (i) As long as $(c - \frac{1}{2})(d - \frac{1}{2}) > 0$ and $x = \frac{\sqrt{d\bar{d}}}{\sqrt{d\bar{d}} + \sqrt{c\bar{c}}}$, the optimal sum-rate will be given by the Körner-Marton region, i.e. using linear codes.
- (ii) An equivalent Proposition can also be stated for the alternate parameterization: $P(Y = 0) = y, P(X = 0|Y = 0) = \hat{c}, P(X = 1|Y = 1) = \hat{d}$.



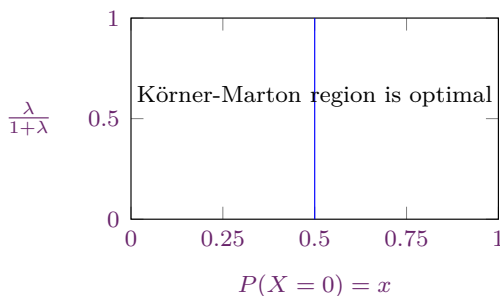
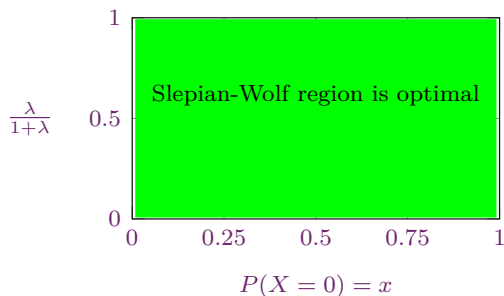
Application to binary alphabets GF(2)

Notation: $P(X = 0) = x, P(Y = 0|X = 0) = c, P(Y = 1|X = 1) = d.$

Known results: Optimality of weighted sum rates.

When $(c - \frac{1}{2})(d - \frac{1}{2}) \leq 0,$

When $c = d,$

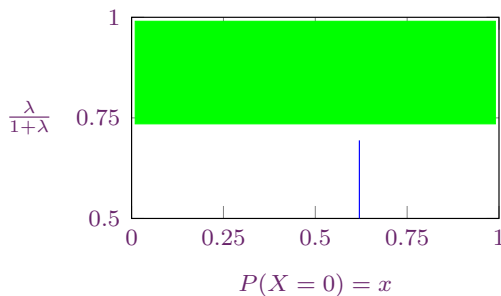


Application to binary alphabets GF(2)

Notation: $P(X = 0) = x, P(Y = 0|X = 0) = c, P(Y = 1|X = 1) = d.$

Our work: When $(c - \frac{1}{2})(d - \frac{1}{2}) > 0,$

Example 1: $c = 0.9, d = 0.6$

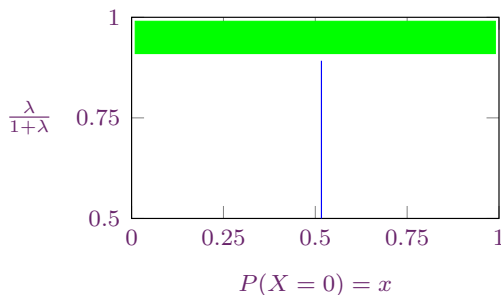


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Notation: $P(X = 0) = x, P(Y = 0|X = 0) = c, P(Y = 1|X = 1) = d.$

Our work: When $(c - \frac{1}{2})(d - \frac{1}{2}) > 0,$

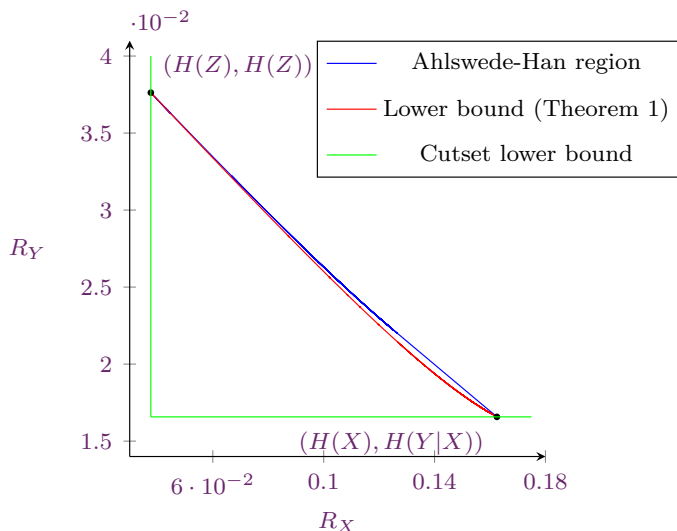
Example 2: $c = 0.7, d = 0.6$



Comparison of the bounds

In [Ahlsvede-Han 83'], Ahlsvede and Han chose the following $p(x, y)$ given by

$$p(x, y) = \begin{bmatrix} 0.003920 & 0.019920 \\ 0.976080 & 0.000080 \end{bmatrix}$$



Application to higher alphabet fields

Example 1

For $GF(3)$, one instance of $p(x, y)$ satisfying that Z is independent of Y and $\mathfrak{C}_{\mu(x)}[H(Y) - H(Z)]|_{p(x)} = H(Y) - H(Z)$ is given by the following distribution:

$$p(x, y) = \begin{bmatrix} 0.08 & 0.06 & 0.18 \\ 0.08 & 0.18 & 0.06 \\ 0.24 & 0.06 & 0.06 \end{bmatrix}$$



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Example 2

One instance of $p(x, y)$ satisfying $\mathfrak{C}_{\mu(x)}[H(Y) - H(Z)]|_{p(x)} = H(Y|X) - H(Z|X)$ is given by the following distribution:

$$p(x, y) = \begin{bmatrix} 0.02 & 0.02 & 0.48 \\ 0.02 & 0.06 & 0.16 \\ 0.06 & 0.02 & 0.16 \end{bmatrix}$$

Related open problems

Here is a conjecture verified by numerical simulations by different groups of researchers.

Conjecture [Sefidgaran-Gohari-Reza 15']

For binary random variables X, Y, U , and V that satisfy the Markov chain $U - X - Y - V$, and for $Z = X \oplus Y$, we have

$$I(X, Y; U, V) + 2H(X|U, V) \geq \min\{H(X, Y), 2H(Z)\}$$

If the conjecture is true, the smallest sum-rate yielded by Ahlswede-Han region is indeed the minimum of $\{H(XY), 2H(Z)\}$, i.e. the minimum of the Slepian-Wolf region and the Körner-Martón region.



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Another open problem is whether this lower bound can be applied to Gaussian Distributed Source Coding with distortion criterion?



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