

On Source Coding with Side Information for General Distortion Measures

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Abstract — High resolution bounds in lossy coding of a real memoryless source are derived when side information is present. We consider locally quadratic distortion measures which may be functions of the side information and give asymptotically tight bounds on the conditional rate-distortion function $R_{X|Y}(D)$ (the side information is available at both the encoder and the decoder) and on the Wyner-Ziv rate-distortion function $R^{\text{WZ}}(D)$ (the side information is only available at the decoder). The rate loss $R^{\text{WZ}}(D) - R_{X|Y}(D)$ is also investigated.

I. INTRODUCTION

Consider the source coding scenario depicted in Figure 1. The sequence $\{(X_k, Y_k)\}$ consists of independent and identically distributed copies of a pair of real random variables (X, Y) , where X is called the *source* and Y is called the *side information*. Fidelity of the reconstruction $\{\hat{X}_k\}$ is measured by means of a nonnegative single-letter distortion measure $d(x, y, \hat{x})$ which depends on the current value y of the side information (context).

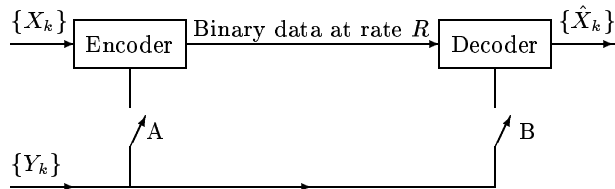


Figure 1: Source coding with side information

When switches A and B are closed, both the encoder and the decoder have access to the side information. Let $R_{X|Y}(D)$ denote the minimum achievable rate at distortion D . Then

$$R_{X|Y}(D) = \inf_{\hat{X}} I(X; \hat{X}|Y),$$

where the infimum of the conditional mutual information $I(X; \hat{X}|Y)$ is taken over all joint distributions of (Y, X, \hat{X}) such that $\mathbf{E}[d(X, Y, \hat{X})] \leq D$.

If switch A is open and switch B is closed, only the decoder knows the side information. Let $R^{\text{WZ}}(D)$ denote the minimum rate achievable at distortion D . Wyner and Ziv showed that

$$R^{\text{WZ}}(D) = \inf_Z I(X; Z|Y),$$

where Z is a real random variable, and where the infimum is taken over all joint distributions of (X, Y, Z) such that Y and Z are conditionally independent given X , and there exists a measurable function $f(Y, Z)$ with $\mathbf{E}[d(X, Y, f(Y, Z))] \leq D$.

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II. RESULTS

We will assume that $d(x, y, \hat{x})$ is a sufficiently smooth function such that $d(x, y, \hat{x}) = 0$ if and only if $x = \hat{x}$, and if $|x - \hat{x}|$ is small then the behavior of $d(x, y, \hat{x})$ is determined by the second order term in its Taylor expansion with respect to \hat{x} around (x, y, x) . That is,

$$d(x, y, \hat{x}) = m(x, y)(x - \hat{x})^2 + o(|x - \hat{x}|^2)$$

as $|x - \hat{x}| \rightarrow 0$, where $m(x, y) = \frac{1}{2}(\partial^2/\partial\hat{x}^2)d(x, y, \hat{x})|_{\hat{x}=x}$. Assume X has a density, finite variance and finite differential entropy $h(X)$, and suppose $I(X; Y) < \infty$.

Theorem 1 As $D \rightarrow 0$, the asymptotic behavior of $R_{X|Y}(D)$ is given by

$$R_{X|Y}(D) = h(X|Y) - \frac{1}{2} \log(2\pi eD) + \frac{1}{2} \mathbf{E}[\log m(X, Y)] + o(1),$$

where $o(1) \rightarrow 0$ as $D \rightarrow 0$.

Theorem 2 As $D \rightarrow 0$, $R^{\text{WZ}}(D)$ is upper bounded as

$$R^{\text{WZ}}(D) \leq h(X|Y) - \frac{1}{2} \log(2\pi eD) + \frac{1}{2} \mathbf{E}[\log \bar{m}(X)] + o(1),$$

where $\bar{m}(X) = \mathbf{E}[m(X, Y)|X]$. If additionally we assume that X and Y are independent, then the upper bound is asymptotically tight, i.e.,

$$R^{\text{WZ}}(D) = h(X) - \frac{1}{2} \log(2\pi eD) + \frac{1}{2} \mathbf{E}[\log \bar{m}(X)] + o(1).$$

We conjecture that the asymptotic upper bound in Theorem 2 is tight even for dependent X and Y .

The nonnegative rate loss $R^{\text{WZ}}(D) - R_{X|Y}(D)$ was investigated in [1] and it was established there that for difference distortion measures which do not depend on the side information (i.e., when $d(x, \hat{x}) = \rho(x - \hat{x})$) the loss becomes asymptotically negligible as $D \rightarrow 0$. Our conclusion for side information dependent distortion measures is different. At least for the case of independent X and Y , if $m(X, Y)$ is not a function of X alone, then Jensen's inequality implies

$$\begin{aligned} \lim_{D \rightarrow 0} (R^{\text{WZ}}(D) - R_{X|Y}(D)) \\ = \frac{1}{2} \mathbf{E}(\log \mathbf{E}[m(X, Y)|X]) - \frac{1}{2} \mathbf{E}[\log m(X, Y)] > 0. \end{aligned}$$

This fact would not be surprising if the value of \hat{x} minimizing $d(x, y, \hat{x})$ for given x and y depended on y , the side-information that is not available at the encoder in the Wyner-Ziv problem. Note, however, that the minimizing \hat{x} is equal to x for all y and still the rate loss is positive.

REFERENCES

- [1] R. Zamir, "The rate loss in the Wyner-Ziv problem," *IEEE Trans. Inform. Theory*, vol. IT-42, pp. 2073–2084, Nov. 1996.