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# Pointwise Redundancy in One-Shot Lossy Compression via Poisson Functional Representation

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**Abstract**—We present a construction of one-shot variable-length lossy source coding schemes using the Poisson functional representation, and give bounds on its pointwise redundancy. This allows us to describe the distribution of the encoding length in a precise manner.

## I. INTRODUCTION

Variable-length lossy source coding has been considered, for example, in  $D$ -semifaithful codes [1], [2] where the distortion must be bounded almost surely. The redundancy of  $D$ -semifaithful codes, i.e., the difference between the encoding length and the rate distortion function, has been studied in [3]–[6].

For one-shot variable-length lossy source coding with the expected distortion constraint  $\mathbb{E}[d(X, Y)] \leq D$ ,<sup>1</sup> it was proved in [7] that there is a prefix-free code with expected length  $\leq R(D) + \log(R(D) + 1) + 6$ , showing that the optimal one-shot expected length is always within a logarithmic gap from the rate-distortion function  $R(D)$ . The proof utilizes the Poisson functional representation [7], [8], where the codebook is constructed as a Poisson process. Also see [9]–[11] for related results.

In this work, we utilize the Poisson functional representation to construct one-shot variable-length lossy source coding schemes, and give bounds on their pointwise redundancy. This allows us to describe the distribution of the encoding length in a more precise manner, compared to only bounding its expectation. The proofs and details of the results mentioned in this abstract, and the generalization to the lossy Gray-Wyner system [12], can be found in the preprint [13].

## II. MAIN RESULTS

A one-shot variable-length lossy compression scheme for the source  $X \in \mathcal{X}$ ,  $X \sim P_X$  with reconstruction space  $\mathcal{Y}$  is a pair  $(P_{M|X}, g)$ , where  $P_{M|X}$  is a stochastic encoder (a conditional distribution from  $\mathcal{X}$  to  $\{0, 1\}^*$ , where  $\{0, 1\}^*$  is the set of bit sequences of any length), and  $g : \{0, 1\}^* \rightarrow \mathcal{Y}$  is a decoding function. The encoder observes  $X \sim P_X$  and outputs the description  $M|X \sim P_{M|X}$ . The decoder observes  $M$  and outputs the reconstruction  $\tilde{Y} = g(M)$ . We can choose

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<sup>1</sup>Note that the probability of excess distortion  $\mathbb{P}(d(X, Y) > D) = \mathbb{E}[\mathbf{1}\{d(X, Y) > D\}]$  can also be written as an expected distortion.

whether to impose the prefix-free condition on  $M$  or not. We may impose an expected distortion constraint  $\mathbb{E}[d(X, \tilde{Y})] \leq D$ , where  $d : \mathcal{X} \times \mathcal{Y} \rightarrow [0, \infty)$  is a distortion function.

We can also replace the variable-length description  $M$  by a positive integer  $K$ , and assume that the encoder produces a positive integer description. Note that we can convert  $K$  into a variable-length description with  $\lfloor \log K \rfloor$  bits without the prefix-free condition [14], or  $\leq \log K + 2 \log(\log K + 1) + 1$  bits with the prefix-free condition using the Elias delta code [15].

The following theorem can be proved using the Poisson functional representation construction similar to [7, Theorem 2], with an analysis using techniques in [8]. Refer to [13] for the proof.

**Theorem 1:** Fix any  $P_X$ ,  $P_{Y|X}$  and  $Q_Y$  satisfying  $P_{Y|X}(\cdot|x) \ll Q_Y$  for  $P_X$ -almost all  $x$ 's. Fix any collection of functions  $\psi_i : \mathcal{X} \times \mathcal{Y} \times \mathbb{Z}_{>0} \rightarrow \mathbb{R}$  that are nondecreasing in the third argument for  $i = 1, \dots, \ell$ . Then there exists a lossy compression scheme with description  $K \in \mathbb{Z}_{>0}$  and reconstruction  $\tilde{Y}$  such that

$$\mathbb{E}[\psi_i(X, \tilde{Y}, K)] \leq \mathbb{E}[\psi_i(X, Y, \ell J)]$$

for  $i = 1, \dots, \ell$ , where  $(X, Y) \sim P_X P_{Y|X}$ , and  $J \in \mathbb{Z}_{>0}$  is distributed as

$$J|(X, Y) \sim \text{Geom}\left(\left(\frac{dP_{Y|X}(\cdot|X)}{dQ_Y}(Y) + 1\right)^{-1}\right).$$

This theorem is quite general. For example, to bound the expected distortion, take  $\psi_i(x, y, k) = d(x, y)$ . To bound the excess distortion probability, take  $\psi_i(x, y, k) = \mathbf{1}\{d(x, y) > D\}$ . To bound the probability that  $K$  cannot be encoded into  $n$  bits (for a fixed-length code), take  $\psi_i(x, y, k) = \mathbf{1}\{k > 2^n\}$ . To bound the expected length with (resp. without) the prefix-free condition, we may take  $\psi_i(x, y, k) = \log k$  (resp.  $\psi_i(x, y, k) = \log k + 2 \log(\log k + 1) + 1$ ).

We can also use Theorem 1 to bound the pointwise redundancy. We consider three different notions of pointwise redundancy: **Pointwise rate redundancy (PRR)**, studied in [5], [16], is given by

$$|M| - R(D),$$

i.e., the difference between the length  $|M|$  of the description  $M$  and the rate-distortion function  $R(D)$  where  $D = \mathbb{E}[d(X, \tilde{Y})]$ .

**Pointwise source-wise redundancy (PSR)**, studied in [5], is given by

$$|M| - j(X, D),$$

where  $j(x, D)$  is the  $d$ -tilted information [5], [17], [18]  $j(x, D) := -\log \mathbb{E}[2^{-\lambda^*(d(x, Y^*) - D)}]$ , where  $Y^* \sim P_Y$  follows the  $Y$ -marginal of  $P_X P_{Y|X}$  where  $P_{Y|X}$  is the conditional distribution that attains the minimum in  $R(D)$  (assume unique minimizer), and  $\lambda^* := -R'(D)$ . **Pointwise source-distortion-wise redundancy (PSDR)** is defined as

$$|M| - j(X, D, d(X, \tilde{Y})),$$

where we write  $j(x, D, \delta) := -\log \mathbb{E}[2^{-\lambda^*(d(x, Y^*) - \delta)}] = j(x, D) - \lambda^*(\delta - D)$ , which can be interpreted as the amount of information needed to convey  $x$  within a distortion  $\delta$  when the overall expected distortion is  $D$ . The expectations of these three redundancies must be nonnegative for prefix-free codes, but might be negative if we do not impose the prefix-free condition. We first state a corollary of Theorem 1 that can bound any of the three pointwise redundancies for the case without the prefix-free condition.

*Corollary 2:* Fix any  $P_X, P_{Y|X}$ , distortion function  $d : \mathcal{X} \times \mathcal{Y} \rightarrow [0, \infty)$ , function  $\eta : \mathcal{X} \times \mathcal{Y} \rightarrow \mathbb{R}$  and  $\gamma \in \mathbb{R}$ . Then there exists a lossy compression scheme without prefix-free condition such that  $\mathbb{E}[d(X, \tilde{Y})] \leq \mathbb{E}[d(X, Y)]$ , and

$$\begin{aligned} \mathbb{P}(|M| - \eta(X, \tilde{Y}) \geq \gamma) \\ \leq \mathbb{E} \left[ \min \{ 2^{-\eta(X, Y) - \gamma + 1} (2^{\ell_{X;Y}(X;Y)} + 1), 1 \} \right], \end{aligned}$$

where  $(X, Y) \sim P_X P_{Y|X}$ .

The result for PSDR is especially simple.

*Corollary 3:* For  $D > 0$ , under the regularity conditions in [18],<sup>2</sup> there exists a lossy compression scheme without prefix-free condition, with  $\mathbb{E}[d(X, \tilde{Y})] \leq D$ , and with PSDR satisfying

$$\mathbb{P}(|M| - j(X, D, d(X, \tilde{Y})) \geq \gamma) \leq 2^{-\gamma + 2}$$

for every  $\gamma \in \mathbb{R}$ .

The results for prefix-free codes are slightly more complicated.

*Corollary 4:* Fix any  $P_X, P_{Y|X}$ , distortion function  $d : \mathcal{X} \times \mathcal{Y} \rightarrow [0, \infty)$ , function  $\eta : \mathcal{X} \times \mathcal{Y} \rightarrow \mathbb{R}$ , and  $\gamma \in \mathbb{R}$ . Then there exists a prefix-free lossy compression scheme such that  $\mathbb{E}[d(X, \tilde{Y})] \leq \mathbb{E}[d(X, Y)]$ , and

$$\begin{aligned} \mathbb{P}(|M| - \eta(X, \tilde{Y}) \geq \gamma) \\ \leq \mathbb{E} \left[ \min \{ 2^{-\eta(X, Y) - \gamma + 2} ([\eta(X, Y) + \gamma]_+ + 1)^2 \right. \\ \left. \cdot (2^{\ell_{X;Y}(X;Y)} + 1), 1 \} \right], \end{aligned}$$

where  $(X, Y) \sim P_X P_{Y|X}$ .

<sup>2</sup>The regularity conditions in [18] are:  $R(\delta)$  is finite for some  $\delta$ , there exists a finite set  $\mathcal{E} \subseteq \mathcal{Y}$  such that  $\mathbb{E}[\min_{y \in \mathcal{E}} d(X, y)] < \infty$ , and the minimum in  $R(D)$  is achieved by a unique  $P_{Y|X}$ .

*Corollary 5:* For  $D > 0$ ,  $\gamma \in \mathbb{R}$ , under the regularity conditions in [18] (see Corollary 3), there exists a prefix-free lossy compression scheme with  $\mathbb{E}[d(X, \tilde{Y})] \leq D$ , and with PSDR satisfying

$$\begin{aligned} \mathbb{P}(|M| - j(X, D, d(X, \tilde{Y})) \geq \gamma) \\ \leq 2^{-\gamma + 3} \mathbb{E}[(\ell_{X;Y}(X;Y) + \gamma)_+ + 1]^2. \end{aligned}$$

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