

Low Rate Scalar Quantization for Gaussian Sources and Absolute Error

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Abstract—This paper considers low resolution scalar quantization for a memoryless Gaussian source with respect to absolute error distortion. It shows that slope of the the operational rate-distortion function of scalar quantization is infinite at the point D_{\max} where the rate becomes zero. Thus, unlike the situation for squared error distortion, or for Laplacian and exponential sources with squared or absolute error distortion, for a Gaussian source and absolute error, scalar quantization at low rates is far from the Shannon rate-distortion function, i.e. far from the performance of the best lossy coding technique.

I. INTRODUCTION

This paper considers the asymptotic low resolution, i.e. low rate, performance of scalar quantization for a Gaussian source and absolute error distortion measure. It follows a path somewhat similar to that taken in [1], where a Gaussian source and squared error distortion measure is considered. Specifically, we find the slope of the operational rate-distortion function of scalar quantization, $R(D)$, at $D = D_{\max}$, where D_{\max} is the minimum distortion attainable with zero rate. This slope determines the speed with which $R(D) \rightarrow 0$ as $D \rightarrow D_{\max}$.

Following [1], we write

$$R(D) = s \left(1 - \frac{D}{D_{\max}} \right) [1 + o_{D \rightarrow D_{\max}}],$$

where $o_{D \rightarrow D_{\max}}$ is a quantity that tends to zero as D goes to D_{\max} , and s is the magnitude of the slope with respect to normalized distortion.

The values of s in the case of exponential and Laplacian sources with both absolute and squared error distortion measures have been given in [2]; the values in the case of a uniform source and both distortion measures can be deduced from [3]; the value of s for Gaussian source and squared error was provided in [1]; finally, the value of s for a Gaussian source and absolute error is given in this paper. Table I below summarizes these values.

We observe from Table I that since the slopes of $R(D)$ at $D = D_{\max}$ equal 0 for exponential and Laplacian sources with squared error, they must equal the slopes of the corresponding

	exponential	Laplacian	uniform	Gaussian
squared error	0	0	∞	$\frac{\log_2 e}{2}$
absolute error	1	$\log_2 e$	∞	∞

TABLE I

MAGNITUDE OF THE SLOPE OF THE OPERATIONAL RATE-DISTORTION FUNCTION $R(D)$ AT $D = D_{\max}$.

Shannon rate-distortion functions (because the latter's magnitudes could be no larger). Furthermore, for a Gaussian source with squared error, and for Laplacian and exponential sources with absolute error, the Shannon rate-distortion functions are known [4], [5]³, and their slopes match the corresponding slopes of $R(D)$. Thus, in low resolution, scalar quantization for these sources and distortion measures is asymptotically optimal, i.e. as good as any quantization technique — vector or otherwise. For a uniform source with both distortion measures, and for a Gaussian source with absolute error, the slopes of $R(D)$ at $D = D_{\max}$ are negatively infinite, whereas the slopes of the corresponding Shannon rate-distortion functions cannot be negatively infinite, i.e., they must be finite (because these functions are convex). Thus, for these sources and distortion measures, low resolution scalar quantization is far from optimal.

The remainder of this paper is organized as follows. Section II provides background and introduces notation. The main result is given in Section III. Section IV offers concluding remarks. Finally, one lemma proof is left to the appendix.

II. BACKGROUND

The assumption throughout the paper is that the source to be quantized is stationary, memoryless and Gaussian with zero mean and variance σ^2 . We denote this source by $\mathcal{N}(0, \sigma^2)$.

A scalar quantizer q is a partition of the real line into cells S_k , each of which contains a reconstruction level r_k such that when the input lies in S_k , the output of the quantizer is r_k . The

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³Reference [5] makes a small error in applying its Theorem 2 to compute the rate-distortion function, with respect to absolute error, of an exponential source. Specifically, for $f(x) = \alpha e^{-\alpha x}$, $\alpha > 0$, a correct application of this theorem yields $\mathcal{R}(D) = -\ln(2(1 - e^{-\alpha D}))$, rather than the formula given in (24) of [5].

number of cells may be finite or infinite. P_k is the probability of the input lying in S_k . The (output) entropy of quantizer q is given by $H(q) = -\sum_k P_k \log P_k$, where all logarithms in this paper have base 2. The mean absolute error induced by the quantizer is

$$d(q) = \int_{-\infty}^{\infty} |x - q(x)| f(x) dx = \sum_k \int_{S_k} |x - r_k| f(x) dx,$$

where f is the Gaussian density of the source. It is well known that reconstruction levels at cell medians minimize mean absolute error, for a given partition. Specifically, for a contiguous cell S_k , i.e. $S_k = [a_k, b_k)$, where it matters not if the interval is open or closed on either side, the median r_k satisfies

$$\int_{a_k}^{r_k} f(x) dx = \int_{r_k}^{b_k} f(x) dx. \quad (1)$$

The operational rate-distortion function of scalar quantization for a Gaussian source with variance σ^2 and absolute error distortion measure is defined as follows:

$$R_{\sigma^2}(D) = \inf_{d(q) \leq D} H(q),$$

which specifies the least entropy of any scalar quantizer with distortion D or less.

Let D_{\max} denote the minimum distortion attainable when the rate is zero. Specifically, for a Gaussian source with variance σ^2 we have $D_{\max} = \sqrt{\frac{2}{\pi}} \sigma$.

Following the notation in [1], let the entropy function be defined as $\mathcal{H}(\dots, z_{-1}, z_0, z_1, \dots) = -\sum_{k=-\infty}^{\infty} z_k \log z_k$, where $0 < z_k \leq 1$ for all k , are a finite or countably infinite set of numbers that need not sum to one. Let $o_{x,y}$ denote a quantity that converges to zero when both $x \rightarrow \infty$ and $y \rightarrow \infty$. If this quantity depends on parameters other than x and y , its convergence to zero is uniform in such parameters. Finally, $G(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$ denotes the Gaussian density with zero mean and unit variance.

III. MAIN RESULT

The following lemma is used to show Theorem 2 below.

Lemma 1: Consider a scalar quantizer applied to a $\mathcal{N}(0, \sigma^2)$ source. If the cell containing the origin has boundaries $-a$ and b , has reconstruction level at the median, and contributes D_o to the mean absolute error of the quantizer, then

$$D_{\max} - D_o = \sigma \left(G\left(\frac{a}{\sigma}\right) + G\left(\frac{b}{\sigma}\right) \right) [1 + o_{a,b}].$$

Proof: Let r_o denote the median of the cell $(-a, b)$, and let f denote the Gaussian density with zero mean and variance σ^2 .

We evaluate D_o as follows:

$$\begin{aligned} D_o &= \int_{-a}^{r_o} (r_o - x) f(x) dx + \int_{r_o}^b (x - r_o) f(x) dx \\ &= r_o \left[\int_{-a}^{r_o} f(x) dx - \int_{r_o}^b f(x) dx \right] \\ &\quad - \int_{-a}^{r_o} x f(x) dx + \int_{r_o}^b x f(x) dx \\ &\stackrel{(a)}{=} \int_{r_o}^b x f(x) dx - \int_{-a}^{r_o} x f(x) dx \\ &\stackrel{(b)}{=} \sigma \left(2G\left(\frac{r_o}{\sigma}\right) - G\left(\frac{a}{\sigma}\right) - G\left(\frac{b}{\sigma}\right) \right), \end{aligned}$$

where (a) is due to the fact that r_o is the median of $(-a, b)$, and (b) follows from having $\int_x^\infty t G(t) dt = G(x)$. Next, we observe that $D_{\max} = \int_{-\infty}^{\infty} |x| f(x) dx = 2\sigma G(0)$. Therefore,

$$D_{\max} - D_o = \sigma \left(G\left(\frac{a}{\sigma}\right) + G\left(\frac{b}{\sigma}\right) + 2[G(0) - G\left(\frac{r_o}{\sigma}\right)] \right). \quad (2)$$

Finally, we have

$$\begin{aligned} 2\sigma [G(0) - G\left(\frac{r_o}{\sigma}\right)] &= 2\sigma \int_0^{\frac{r_o}{\sigma}} x G(x) dx \\ &\leq 2\sigma G(0) \left(\frac{r_o}{\sigma}\right)^2 \\ &= \sigma \left(G\left(\frac{a}{\sigma}\right) + G\left(\frac{b}{\sigma}\right) \right) o_{a,b}, \quad (3) \end{aligned}$$

where the last equality is due to having $\left(\frac{r_o}{\sigma}\right)^2 = \left(G\left(\frac{a}{\sigma}\right) + G\left(\frac{b}{\sigma}\right)\right) o_{a,b}$, as shown by Lemma A1 of the appendix. The lemma now follows from (2) and (3) \square

The following theorem is the principle result of this paper.

Theorem 2: For a $\mathcal{N}(0, \sigma^2)$ source and absolute error distortion, the operational rate-distortion function of scalar quantization satisfies

$$\lim_{D \rightarrow D_{\max}} \frac{R_{\sigma^2}(D)}{D_{\max} - D} = \infty.$$

Proof: It suffices to consider only scalar quantizers with contiguous cells, as follows from [6]. By definition of $R_{\sigma^2}(D)$, for any $D \in (0, D_{\max})$ there exists a quantizer q_D such that

$$H(q_D) \leq R_{\sigma^2}(D) + \varepsilon(D) \quad \text{and} \quad d(q_D) \leq D, \quad (4)$$

where $\varepsilon(D)$ is some function of D such that $\varepsilon(D) > 0$ and $\lim_{D \rightarrow D_{\max}} \frac{\varepsilon(D)}{D_{\max} - D} = 0$. (The choices of q_D and $\varepsilon(D)$ are not unique, but any fixed choices will do.) Let $S_{o,D} = (-A_D, B_D)$, denote the cell of q_D containing the origin (it is immaterial if the cell is open or closed on either side). As $D \rightarrow D_{\max}$, $A_D, B_D \rightarrow \infty$. Note that either A_D or B_D (but not both simultaneously) might be infinite. Let $D_{o,D}$ be the contribution to distortion of cell $S_{o,D}$, where the reconstruction level of $S_{o,D}$ lies at the median of the cell. It follows from Lemma 1 that

$$D_{\max} - D_{o,D} = \sigma \left(G\left(\frac{A_D}{\sigma}\right) + G\left(\frac{B_D}{\sigma}\right) \right) [1 + o_{A_D, B_D}]. \quad (5)$$

Next, applying Lemma 5 from [1], which shows that $\mathcal{H}(Q(x)) = \frac{\log e}{2} x G(x) [1 + o_x]$, where $Q(x) = \int_x^\infty G(t) dt$ is the usual “ Q function”, we obtain

$$\begin{aligned} & \mathcal{H}\left(Q\left(\frac{A_D}{\sigma}\right)\right) + \mathcal{H}\left(Q\left(\frac{B_D}{\sigma}\right)\right) \\ &= \frac{\log e}{2} \left(\frac{A_D}{\sigma} G\left(\frac{A_D}{\sigma}\right) + \frac{B_D}{\sigma} G\left(\frac{B_D}{\sigma}\right) \right) [1 + o_{A_D, B_D}]. \end{aligned} \quad (6)$$

Finally, we have that

$$\begin{aligned} \liminf_{D \rightarrow D_{\max}} \frac{R_{\sigma^2}(D)}{D_{\max} - D} &\stackrel{(a)}{\geq} \liminf_{D \rightarrow D_{\max}} \frac{H(q_D) - \varepsilon(D)}{D_{\max} - d(q_D)} \\ &\stackrel{(b)}{\geq} \liminf_{D \rightarrow D_{\max}} \frac{\mathcal{H}\left(Q\left(\frac{A_D}{\sigma}\right)\right) + \mathcal{H}\left(Q\left(\frac{B_D}{\sigma}\right)\right)}{D_{\max} - D_{o,D}} \\ &\stackrel{(c)}{=} \infty, \end{aligned}$$

where (a) follows from (4), (b) is due to an elementary property of entropy and from having $D_{o,D} \leq d(q_D)$, and (c) derives from (5) and (6). Thus, $\lim_{D \rightarrow D_{\max}} \frac{R_{\sigma^2}(D)}{D_{\max} - D} = \liminf_{D \rightarrow D_{\max}} \frac{R_{\sigma^2}(D)}{D_{\max} - D} = \infty$, as needed to show. \square

IV. CONCLUSIONS

This paper considered the asymptotic low resolution performance of scalar quantizers for a Gaussian source with absolute error distortion measure. This performance is determined by the slope of the operational rate-distortion function of such quantizers at $D = D_{\max}$. It has been shown that the slope of the operational rate-distortion function of scalar quantization is infinite, and hence does not match the slope of the Shannon rate-distortion function, which is finite. Consequently, scalar quantization is not an optimal coding technique, in asymptotically low rate, for the given source and distortion measure. This is somewhat surprising since, as noted earlier, scalar quantization is optimal for a Gaussian source and squared error distortion measure, and for Laplacian source and both squared and absolute error distortion measures.

APPENDIX

Lemma A1: Let $-a$ and b be the boundaries of the cell containing the origin for a quantizer applied to a $\mathcal{N}(0, \sigma^2)$ source. Let r_o be the median of $(-a, b)$. Then,

$$\left(\frac{r_o}{\sigma}\right)^2 = \left(G\left(\frac{a}{\sigma}\right) + G\left(\frac{b}{\sigma}\right)\right) o_{a,b}.$$

Proof: From (1) we obtain that

$$Q\left(\frac{r_o}{\sigma}\right) = \frac{Q\left(\frac{-a}{\sigma}\right) + Q\left(\frac{b}{\sigma}\right)}{2} = \frac{1 - Q\left(\frac{a}{\sigma}\right) + Q\left(\frac{b}{\sigma}\right)}{2}. \quad (\text{A1})$$

Next, let $v \in \mathbb{R}$ be arbitrary. If $v \geq 0$, then

$$Q(v) = \int_v^\infty G(x) dx = \frac{1}{2} - \int_0^v G(x) dx \leq \frac{1}{2} - vG(v),$$

from which it follows that $0 \leq v \leq \frac{\frac{1}{2} - Q(v)}{G(v)}$. Similarly, if $v < 0$, then

$$\begin{aligned} Q(v) &= 1 - Q(|v|) = 1 - \left(\frac{1}{2} - \int_0^{|v|} G(x) dx\right) \\ &= \frac{1}{2} + \int_0^{|v|} G(x) dx \geq \frac{1}{2} + vG(v) = \frac{1}{2} - vG(v), \end{aligned}$$

from which it follows that $\frac{\frac{1}{2} - Q(v)}{G(v)} \leq v < 0$. These two bounds to v imply $v^2 \leq \left(\frac{\frac{1}{2} - Q(v)}{G(v)}\right)^2$. This is now used as follows:

$$\begin{aligned} \left(\frac{r_o}{\sigma}\right)^2 &\leq \left(\frac{\frac{1}{2} - Q\left(\frac{r_o}{\sigma}\right)}{G\left(\frac{r_o}{\sigma}\right)}\right)^2 \stackrel{(a)}{=} \left(\frac{\frac{1}{2} - \frac{1}{2} + \frac{Q\left(\frac{a}{\sigma}\right)}{2} - \frac{Q\left(\frac{b}{\sigma}\right)}{2}}{G\left(\frac{r_o}{\sigma}\right)}\right)^2 \\ &\stackrel{(b)}{=} \left(\frac{Q\left(\frac{a}{\sigma}\right) - Q\left(\frac{b}{\sigma}\right)}{\sqrt{2\pi}[1 + o_{a,b}]}\right)^2 \stackrel{(c)}{=} \left(Q\left(\frac{a}{\sigma}\right) - Q\left(\frac{b}{\sigma}\right)\right) o_{a,b} \\ &\stackrel{(d)}{=} \left(\frac{G\left(\frac{a}{\sigma}\right)}{a/\sigma}[1 + o_a] - \frac{G\left(\frac{b}{\sigma}\right)}{b/\sigma}[1 + o_b]\right) o_{a,b} \\ &= \left(G\left(\frac{a}{\sigma}\right)[1 + o_a] o_a - G\left(\frac{b}{\sigma}\right)[1 + o_b] o_b\right) o_{a,b} \\ &= \left(G\left(\frac{a}{\sigma}\right) + G\left(\frac{b}{\sigma}\right)\right) o_{a,b}, \end{aligned}$$

where (a) follows from (A1), (b) is obtained from the fact that $\frac{r_o}{\sigma} \rightarrow 0$ as both a and b tend to infinity, (c) follows from having $Q\left(\frac{a}{\sigma}\right) - Q\left(\frac{b}{\sigma}\right) \rightarrow 0$ as both a and b tend to infinity, and (d) derives from having $Q(x) = \frac{1}{x} G(x) [1 + o_x]$ for any $x > 0$, which is obtained from the fact that for any $x > 0$, $\frac{1}{x}(1 - \frac{1}{x^2}) G(x) < Q(x) < \frac{1}{x} G(x)$, as shown in [7, pp. 82-83]. \square

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