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New Bounds for Kullback-Leibler Divergence based on a Lagrange Theorem

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Abstract

Entropy, conditional entropy and mutual information for discrete-valued random variables play important roles in the information theory. The purpose of this paper is to present new bounds for relative entropy $D(p||q)$ of two probability distributions and then to apply them to simple entropy and mutual information. The relative entropy upper bound obtained is a refinement of a bound previously presented into literature. For this purpose we develop a generalization for the classical Lagrange's Theorem and we apply it to refine the wanted result.

Keywords: Lagrange's theorem, entropy, bounds, refinements, generalizations

1. Introduction

The relative entropy $D(p||q)$ (see [1],[2]) is the measure of distance between two distributions. It can also be expressed like a measure of the inefficiency of assuming that the distribution is q when the true distribution is p .

Definition 1. (Relative Entropy) *The relative entropy, of the Kullback-Leibler distance, between two probability mass functions $p(x)$ and $q(x)$ is defined as*

$$D(p||q) := \sum_{x \in X} p(x) \log \left(\frac{p(x)}{q(x)} \right) = E_p \log \left(\frac{p}{q} \right),$$

where \log is the natural logarithm.

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A fundamental property of the relative entropy is the following

Theorem 2. *Let $p(x), q(x), x \in X$ be two probability mass functions. Then*

$$D(p||q) \geq 0,$$

with equality if and only if $p(x) = q(x), \forall x \in X$.

Obtaining a lower bound, the above fundamental inequality can be improved as follows (see [2])

Theorem 3. *Let $p(x), q(x), x \in X$ be two probability mass functions. Then*

$$D(p||q) \geq \frac{1}{2} \left(\sum_{x \in X} |p(x) - q(x)| \right)^2.$$

As an upper bound for relative entropy, we have taken into consideration the result presented in Theorem 1, [3] by S.S. Dragomir et al. Other bounds can also be found in [4, 5, 6, 7].

Theorem 4. (S.S. Dragomir et al.) *Let $p(x), q(x) > 0, x \in X$ be two probability mass functions. Then*

$$D(p||q) \leq \sum_{x \in X} \frac{p^2(x)}{q(x)} - 1 = \frac{1}{2} \sum_{x, y \in X} p(x)q(x) \left(\frac{p(x)}{q(x)} - \frac{p(y)}{q(y)} \right) \left(\frac{q(y)}{p(y)} - \frac{q(x)}{p(x)} \right),$$

with equality if and only if $p(x) = q(x), \forall x \in X$.

Furthermore, in order to refine the previous inequality, we continue presenting Lagrange's Theorem, for which we will express a generalization that will help us tighten the result.

Theorem 5. (Lagrange) *Let $f : [a, b] \rightarrow \mathbb{R}$ be a continuous function on $[a, b]$ and derivable on (a, b) , with $b > a > 0$. Then exists $c \in (a, b)$ such that*

$$\frac{f(b) - f(a)}{b - a} = f'(c).$$

2. Generalization of Lagrange's Theorem and a New Inequality for Strictly Convex Functions

We continue with a generalization of Lagrange's theorem. A similar previous result can be found in [8], as follows

Theorem 6. *Let n be a natural number, and $f : [a, b] \rightarrow \mathbb{R}$ a continuous function on $[a, b]$ and derivable on (a, b) , with $b > a > 0$. Then exists $c_1, c_2, \dots, c_n \in (a, b)$ such that*

$$\frac{f(b) - f(a)}{b - a} = \frac{f'(c_1) + f'(c_2) + \dots + f'(c_n)}{n}.$$

Proof. First we will split the $[a, b]$ interval into n equal intervals, as $[a, x_1], [x_1, x_2], \dots, [x_{n-1}, b]$, with

$$x_1 - a = x_2 - x_1 = \dots = b - x_{n-1} = \frac{b-a}{n}.$$

Now applying the classical Lagrange Theorem for each interval, we obtain that exists $c_1 \in [a, x_1], c_2 \in [x_1, x_2], \dots, c_n \in [x_{n-1}, b]$ such that

$$\frac{f(x_1) - f(a)}{\frac{b-a}{n}} = f'(c_1), \frac{f(x_2) - f(x_1)}{\frac{b-a}{n}} = f'(c_2), \dots, \frac{f(b) - f(x_{n-1})}{\frac{b-a}{n}} = f'(c_n).$$

Continue with summing the above equalities, yields that

$$\frac{f(b) - f(a)}{b-a} = \frac{f'(c_1) + f'(c_2) + \dots + f'(c_n)}{n}$$

and we are done. \square

Applying the condition of strictly convex functions to the above function f , we obtain the following

Theorem 7. *Let n be a natural number, and $f : [a, b] \rightarrow \mathbb{R}$ a continuous function on $[a, b]$, derivable on (a, b) and strictly convex, with $b > a > 0$. Then*

$$f'(a) + \sum_{i=1}^{n-1} f' \left(a + i \frac{b-a}{n} \right) < n \frac{f(b) - f(a)}{b-a} < \sum_{i=1}^{n-1} f' \left(a + i \frac{b-a}{n} \right) + f'(b).$$

Proof. As f is a strictly convex function implies that f' is an increasing function, so because from the above theorem exists $c_1 \in [a, x_1], c_2 \in [x_1, x_2], \dots, c_n \in [x_{n-1}, b]$, with $x_1 - a = x_2 - x_1 = \dots = b - x_{n-1} = \frac{b-a}{n}$, implies that

$$f'(a) < f'(c_1) < f'(x_1), f'(x_1) < f'(c_2) < f'(x_2), \dots, f'(x_{n-1}) < f'(c_n) < f'(b)$$

and considering the result of the previous theorem and summing, we get the wanted result. \square

We continue with the following remark

Remark 8. *It is easy to see that for any natural positive number n*

$$nf'(a) \leq f'(a) + \sum_{i=1}^{n-1} f' \left(a + i \frac{b-a}{n} \right) \text{ and } \sum_{i=1}^{n-1} f' \left(a + i \frac{b-a}{n} \right) + f'(b) \leq nf'(b),$$

because $a < a + i \frac{b-a}{n}$ and $a + i \frac{b-a}{n} < b$, for all $i = 1, 2, \dots, n-1$.

3. New Bounds for Relative Entropy

In order to present a general inequality for $-\log x$ we start with a helpful result, which can be deduced by simple calculus, as follows

Lemma 1. *Let a, b, t, T be real numbers with $b \neq 0$ and $T > t > 0$, then the following two inequalities are equivalent*

$$t < \frac{a}{b} < T$$

and

$$b \frac{T+t - \frac{b}{|b|}(T-t)}{2} < a < b \frac{T+t + \frac{b}{|b|}(T-t)}{2}.$$

Now, applying Theorem 7 to the function $-\log x$ and taking into consideration the previous Lemma, yields

Corollary 9. *Let $a, b > 0$, with $m = \min\{a, b\}$, $M = \max\{a, b\}$ and $n \geq 1$ a natural number, then*

$$\begin{aligned} (a-b)\frac{1}{a} &\leq (a-b) \left(\sum_{i=1}^{n-1} \frac{1}{nm+i(M-m)} + \frac{m+M+b-a}{2nmM} \right) \leq \log a - \log b \\ &\leq (a-b) \left(\sum_{i=1}^{n-1} \frac{1}{nm+i(M-m)} + \frac{m+M+a-b}{2nmM} \right) \leq (a-b)\frac{1}{b}, \end{aligned}$$

with equality holding for $a = b$.

Proof. If $a = b$ then the inequality is obvious, so if $a \neq b$, by applying Theorem 7 to the function $-\log x$ defined on the interval $I = [m, M]$ and taking cont that $\frac{f(M)-f(m)}{M-m} = \frac{f(b)-f(a)}{b-a}$ we get

$$-\frac{1}{nm} - \sum_{i=1}^{n-1} \frac{1}{nm+i(M-m)} < \frac{\log a - \log b}{b-a} < -\sum_{i=1}^{n-1} \frac{1}{nm+i(M-m)} - \frac{1}{nM}$$

and now taking into consideration the equivalence from Lemma 1, yields

$$\begin{aligned} (a-b) \frac{2 \sum_{i=1}^{n-1} \frac{1}{nm+i(M-m)} + \frac{m+M}{nmM} - \frac{b-a}{M-m} \frac{m-M}{nmM}}{2} &< \log a - \log b \\ &< (a-b) \frac{2 \sum_{i=1}^{n-1} \frac{1}{nm+i(M-m)} + \frac{m+M}{nmM} + \frac{b-a}{M-m} \frac{m-M}{nmM}}{2}, \end{aligned}$$

which is equivalent with

$$\begin{aligned} (a-b) \left(\sum_{i=1}^{n-1} \frac{1}{nm+i(M-m)} + \frac{m+M+b-a}{2nmM} \right) &< \log a - \log b \\ &< (a-b) \left(\sum_{i=1}^{n-1} \frac{1}{nm+i(M-m)} + \frac{m+M+a-b}{2nmM} \right) \end{aligned}$$

and by comparing with the previous remark we get the wanted result. \square

We continue with the main result of the paper namely, new bounds for the relative entropy, where we have considered two probability mass function $p(x)$ and $q(x)$, $x \in X$.

Theorem 10. *Let $p(x), q(x) > 0$, $x \in X$ be two probability mass functions, with $m(x) = \min\{p(x), q(x)\}$ and $M(x) = \max\{p(x), q(x)\}$, $x \in X$. If $r(x) = p(x) - q(x)$, then*

$$\begin{aligned} \sum_{x \in X} p(x)r(x) \left(\sum_{i=1}^{n-1} \frac{1}{nm(x)+i(M(x)-m(x))} + \frac{m(x)+M(x)+r(x)}{2nm(x)M(x)} \right) \\ \geq D(p||q) \\ \geq \sum_{x \in X} p(x)r(x) \left(\sum_{i=1}^{n-1} \frac{1}{nm(x)+i(M(x)-m(x))} + \frac{m(x)+M(x)-r(x)}{2nm(x)M(x)} \right), \end{aligned}$$

with equality if and only if $p(x) = q(x)$, $\forall x \in X$.

Proof. Setting $a = q(x)$ and $b = p(x)$ in Corollary 9 we obtain

$$\begin{aligned} & -r(x) \left(\sum_{i=1}^{n-1} \frac{1}{nm(x) + i(M(x) - m(x))} + \frac{m(x) + M(x) + r(x)}{2nm(x)M(x)} \right) \\ & \leq \log q(x) - \log p(x) \\ & \leq -r(x) \left(\sum_{i=1}^{n-1} \frac{1}{nm(x) + i(M(x) - m(x))} + \frac{m(x) + M(x) - r(x)}{2nm(x)M(x)} \right), \end{aligned}$$

and multiplying by $-p(x)$ yields

$$\begin{aligned} & p(x)r(x) \left(\sum_{i=1}^{n-1} \frac{1}{nm(x) + i(M(x) - m(x))} + \frac{m(x) + M(x) + r(x)}{2nm(x)M(x)} \right) \\ & \geq p(x) \log \frac{p(x)}{q(x)} \\ & \geq p(x)r(x) \left(\sum_{i=1}^{n-1} \frac{1}{nm(x) + i(M(x) - m(x))} + \frac{m(x) + M(x) - r(x)}{2nm(x)M(x)} \right), \end{aligned}$$

from which summing over $x \in X$ we get the wanted result. \square

We continue with the following remark

Remark 11. From Corollary 9 and Theorem 10 we can deduce that

$$\begin{aligned} & \sum_{x \in X} \frac{p^2(x)}{q(x)} - 1 \geq \\ & \sum_{x \in X} p(x)r(x) \left(\sum_{i=1}^{n-1} \frac{1}{nm(x) + i(M(x) - m(x))} + \frac{m(x) + M(x) + r(x)}{2nm(x)M(x)} \right) \\ & \geq D(p||q), \end{aligned}$$

which leads us to the conclusion that the upper bound of relative entropy provided by Theorem 10 is stronger than the one from [3].

Furthermore we present new bounds for entropy and mutual information.

Corollary 12. Let X be a random variable whose range has $|X|$ elements and has the probability mass function $p(x) > 0$, with $m(x) = \min\{p(x), 1/|X|\}$ and $M(x) = \max\{p(x), 1/|X|\}$, $x \in X$. If $r(x) = p(x) - 1/|X|$, then

$$\begin{aligned} & \sum_{x \in X} p(x)r(x) \left(\sum_{i=1}^{n-1} \frac{1}{nm(x) + i(M(x) - m(x))} + \frac{m(x) + M(x) + r(x)}{2nm(x)M(x)} \right) \\ & \geq \log |X| - H(X) \\ & \geq \sum_{x \in X} p(x)r(x) \left(\sum_{i=1}^{n-1} \frac{1}{nm(x) + i(M(x) - m(x))} + \frac{m(x) + M(x) - r(x)}{2nm(x)M(x)} \right). \end{aligned}$$

The equality holds if and only if $p(x) = 1/|X|$.

Proof. It follows from Theorem 10 applied for $D(p||q)$, where now $p(x) = p(x)$ and $q(x) = 1/|X|$, i.e. $D(p(x)||1/|X|)$. \square

Corollary 13. *Let X, Y be two random variables with a joint probability mass function $p(x, y)$ and marginal probability mass function $p(x)$ and $p(y)$, with $p(x, y), p(x), p(y) > 0, x \in X, y \in Y$ and $m_{x,y} = \min\{p(x, y), p(x)p(y)\}$ and $M_{x,y} = \max\{p(x, y), p(x)p(y)\}, x \in X, y \in Y$. If $r_{x,y} = p(x, y) - p(x)p(y)$, then*

$$\begin{aligned} & \sum_{(x,y) \in X \times Y} p(x, y) r_{x,y} \left(\sum_{i=1}^{n-1} \frac{1}{nm_{x,y} + i(M_{x,y} - m_{x,y})} + \frac{m_{x,y} + M_{x,y} + r_{x,y}}{2nm_{x,y}M_{x,y}} \right) \\ & \geq I(X; Y) \\ & \geq \sum_{(x,y) \in X \times Y} p(x, y) r_{x,y} \left(\sum_{i=1}^{n-1} \frac{1}{nm_{x,y} + i(M_{x,y} - m_{x,y})} + \frac{m_{x,y} + M_{x,y} - r_{x,y}}{2nm_{x,y}M_{x,y}} \right), \end{aligned}$$

The equality holds if and only if X and Y are independent.

Proof. It follows from Theorem 10 applied for $D(p||q)$, where now $p(x) = p(x, y)$ and $q(x) = p(x)p(y)$, i.e. $D(p(x, y)||p(x)p(y))$ and where, this time $m(x) = m_{x,y}, M(x) = M_{x,y}, r(x) = r_{x,y}$. \square

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- [1] R. Ash, Information Theory, Interscience, New York, 1965.
- [2] T. M. Cover and J. A. Thomas, Elements of Information Theory, Jhon Wiley and Sons, Inc., 2006.
- [3] S. S. Dragomir, M. L. Scholz and J. Sunde, Some Upper Bounds for Relative Entropy and Applications, Computers and Mathematics with Applications, 39(2000), 91-100.
- [4] S. S. Dragomir and C. J. Goh, A counterpart of Jensen's discrete inequality for differentiable convex mappings and applications in information theory, Math. Comput. Modelling, 24(1996), 1-11.
- [5] M. Matic, C. E. M. Pearce and J. Pecaric, Refinements of some bounds in information theory, ANZIAM J., 42(2001), 387-398.
- [6] S. Simic, Jensen's inequality and new entropy bounds, Applied Mathematics Letters, 22(2009), 1262-1265.
- [7] N. Tapus and P.G. Popescu, New Entropy Upper Bound, Applied Mathematics Letters, 25 no. 11(2014), 1887-1890.
- [8] P. G. Popescu, O extindere a teoremei lui Lagrange, Revista Arhimede, 1-2(2004), 5-8.